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29 **expertise to improve ARID and its ability to make dramatic advances in our understanding of and capacity to**  
30 **support decisions for Earth's drylands. A full Acknowledgement of Contributors and their contributions can**  
31 **be found in Appendix F.8.**

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## Executive Summary

The NASA Terrestrial Ecology Program's **Adaptation and Response in Drylands (ARID)** scoping study provides a research agenda for the next generation of NASA's Terrestrial Ecology field campaigns. ARID will be an international research initiative bringing a global focus to dryland functioning and its critical role in climate mitigation and adaptation. Dryland ecosystems represent the planet's largest terrestrial biome, making up over 40% of the land surface and sustaining a third of the world's population, meaning one in three people in the world today lives in drylands. Defined by their aridity, dryland ecosystems provide vital services that include supplying 60% of global food production, supporting critical hotspots of biodiversity, representing 40% of terrestrial net primary production, and dominating the trend and the interannual variability in the terrestrial carbon sink. Dryland systems are also increasingly vulnerable to global change given compounded impacts from climate change and land use. Plant communities are shifting substantially in drylands around the world, with widespread observations of woody encroachment into grasslands and savannas, exotic grass invasion, large scale mortality events, lowered productivity, and losses of biodiversity in the face of exceptional drought, temperature, and wildfire. The fact that drylands have been identified as one of the most vulnerable global ecosystems to climate change has significant implications for food security, water availability, biodiversity, land degradation, geopolitical stability, as well as large potential feedbacks to future climate. ***Despite their global importance and high vulnerability to change, our scalable understanding and capacity to forecast dryland dynamics, as well as impacts to ecosystems and society, is notably poor.***

The ARID campaign would augment NASA's extensive current (e.g., ECOSTRESS, EMIT, SMAP, Landsat) and future (e.g., NISAR, SBG, and Landsat Next) satellites with a dedicated multi-scale field campaign using NASA airborne assets (e.g., AVIRIS, UAVSAR, and MASTER) and ground measurements. With these spaceborne, aircraft, and field sampling capabilities across scales, NASA is best positioned to bring the science and end-user communities together to address the ARID campaign's four interrelated science themes and overarching questions:

### ***Climate Variability and Drought***

*How are climate extremes like droughts, heatwaves, and large rain pulses impacting dryland systems and how do they interact with changing fire regimes, land cover change, and land-atmosphere interactions?*

### ***Ecosystem Structure, Function and Biodiversity***

*What are the main mechanisms driving the spatiotemporal distributions of dryland structure, function, and biodiversity and what is their vulnerability to change?*

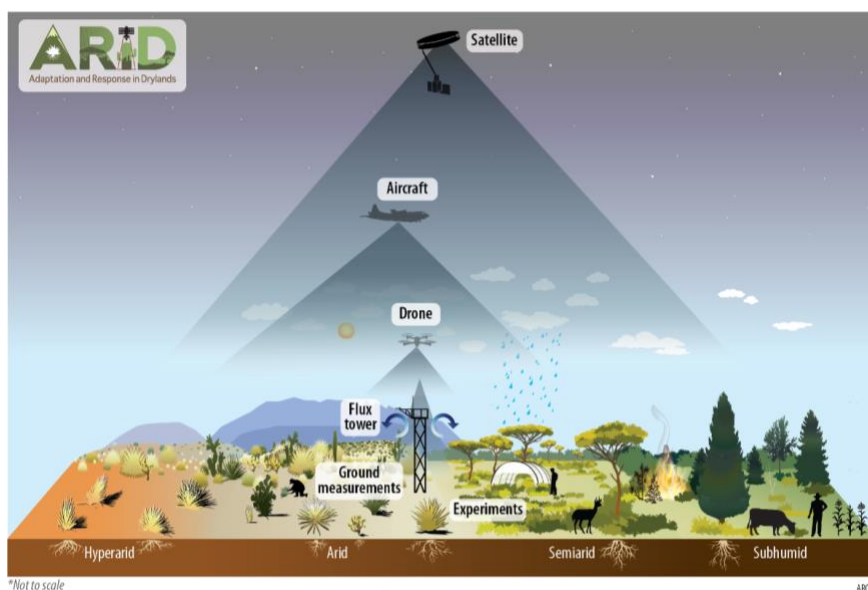
### ***Carbon Cycle Interannual Variability and Long-Term Trends***

*What is the contribution of drylands to the mean, trend, and interannual variability of terrestrial carbon uptake and what drives these patterns?*

### ***Social-Ecological Systems***

*What are the consequences of changes in drylands for social-ecological systems and what management (e.g., mitigation and adaptation) solutions can maintain the critical services provided by drylands even in the face of change?*

1 The interacting pressures of climate and land-use change are dramatically altering the  
2 structure and function of Earth's drylands. Although global efforts recognize the crisis these  
3 changes portend for social-ecological systems, recent research exposes a significant lack of  
4 knowledge regarding the mechanisms behind and consequences of dryland transformation. To  
5 date, there has been no sustained research and monitoring campaign targeting dryland systems  
6 to fill this gap. Research in these heterogeneous ecosystems requires an understanding  
7 contextualized within larger planetary scale biogeophysical dynamics and variability. As such,  
8 NASA Terrestrial Ecology field campaigns are the ideal mechanism for gathering the cross-scale,  
9 transdisciplinary knowledge needed to understand, quantify, and predict change in these critical  
10 ecosystems. The ARID scoping study provides an intensive research plan consistent with the  
11 mandate of NASA Earth Science Division's Earth System Observatory to "create a 3D, holistic  
12 view of Earth, from bedrock to atmosphere." Furthermore, ARID is well aligned with NASA's Earth  
13 Science to Action (ES2A) strategy, which has the goal of innovating and collaborating across  
14 multiple research and end-user sectors "to explore and understand the Earth system, make new  
15 discoveries, and enable solutions for the benefit of all".



*ARID will vastly improve the quantitative and predictive understanding of Earth's drylands using a nested approach that coordinates in-situ measurements, existing and new networks, and with drone and aircraft data that harmonize with and inform satellite observations. The datasets and new knowledge will cross disciplines and scales.*

33 The ARID framework was co-created by a large and diverse community of scientists, data users,  
34 and practitioners who together informed an actionable study design to dramatically advance our  
35 understanding and capacity to monitor, forecast, and inform decisions for Earth's rapidly changing  
36 drylands. Specifically, the scoping study provides a rationale for the importance of better  
37 understanding Earth's dryland ecosystems, poses key scientific questions around the  
38 foundational interconnected research themes, offers a conceptual campaign framework to  
39 address them, and underscores the necessary role of NASA in achieving these important goals.  
40 ARID integrates multiple observations and analytical methods (e.g., proximal, drone, airborne,  
41 and satellite remote sensing observations, ground-based measurements, tower-based fluxes,  
42 and multi-scale modeling), leveraging the wealth of archived, current, and future multi-mission  
43 NASA data. During scoping, the ARID team secured strategic, multi-sectoral partnerships for  
44 facilitating the research, for co-developing science-based information products with decision

1 makers, and for supporting and training the next generations of interdisciplinary Earth system  
2 scientists. ARID is co-designed with diverse end-users and knowledge providers (e.g., United  
3 States [US] Bureau of Land Management, Tribal authorities, Department of Defense, Department  
4 of Energy, US Geological Survey, Department of Agriculture, National Science Foundation) - as  
5 well as academic institutions and non-governmental organizations around the world - through  
6 high-level involvement, workshops, Town Halls, Round Tables, individual discussions, Working  
7 Groups, and webinars. ARID is responsive to and was developed with Native American  
8 perspectives during the co-design process, integrating Tribal Nations' expertise and viewpoints  
9 as part of the scoping study.

10 ARID is an international campaign, outlining strong research needs, partnerships, and synergistic  
11 field campaign initiatives for Australia, northern Mexico, southern Africa, and South America. At  
12 the same time there is a significant focus on the United States (US) due the strong national need  
13 for actionable dryland science. Drylands are of particular importance to US national resources  
14 and security, covering ~83% of the western US and under stress from the ongoing North American  
15 megadrought. Within the US, the Bureau of Land Management, US Forest Service, and Tribal  
16 authorities together manage >2 million km<sup>2</sup> of drylands for multisectoral land uses (rangeland,  
17 cultivation, forestry, conservation, energy), most of which are experiencing changes in vegetation  
18 and productivity relating to climate and land use change, including shrub encroachment and loss  
19 of grazing resources, invasive grasses, and changing fire regimes. Moreover, alternative energy  
20 siting and ambitious nature-based climate solutions (NbCS) programs frequently target drylands  
21 for development, restoration, and/or afforestation to enhance carbon sequestration and the  
22 human environment, but critical tradeoffs for biodiversity, water consumption, surface albedo, and  
23 social-ecological outcomes must be fully considered prior to acting on such recommendations.  
24 Yet the tools and scalable information to evaluate these actions, and for quantifying the intended  
25 and unintended consequences, remain remarkably limited.

26 By investigating how ecosystems are responding to climate change, weather extremes,  
27 management practices, and land-use change, ARID aims to better understand the shifting role of  
28 drylands in the Earth System, including global carbon and water cycles, thereby advancing our  
29 predictive capacity to support end-user decision making and improve forecasts of current and  
30 future feedbacks to climate change at the global scale. This white paper provides a broad  
31 research agenda based on engagement with the scientific community and end-users. The first  
32 three science themes - Climate Variability and Drought, Ecosystem Structure, Function, and  
33 Biodiversity, and Carbon Cycle Dynamics - focus mainly on the fundamental interactions between  
34 the *drivers of change* and *ecosystem responses*. The fourth theme, Social-Ecological Systems:  
35 Natural Resource Management and Adaptation-Mitigation, is related to *human systems*. Taken  
36 together, the white paper embraces drylands as multi-faceted, complex systems, meriting broad  
37 interdisciplinary approaches including terrestrial ecology, hydrology, biogeochemistry, remote  
38 sensing, and modeling. ***We stress that drylands represent a critical set of systems that – due  
39 to their global importance, responsiveness to perturbation, and poor representation in  
40 models – induce large uncertainties in our global understanding of anthropogenic change  
41 and feedbacks on Earth system processes.***

1           The white paper also provides an implementation plan to address central science  
2 questions using a set of multi-scaled remote sensing approaches, ground networks, and process  
3 models, with NASA assets at the forefront of these approaches. A core portion of ARID is  
4 proposed to be carried out in the western US, taking advantage of natural gradients of aridity,  
5 practitioner interest, and the opportunity to leverage existing NASA data, partner airborne  
6 campaigns, and ground assets (e.g., NASA-USGS GEMx campaigns, the AmeriFlux network, the  
7 NSF NEON network and LTER networks, the USDA LTAR networks, etc.). In this way, **ARID**  
8 **provides a nexus of collaboration across existing institutions, researchers, and end-users,**  
9 **that require the diverse satellite observations, Airborne Science Program, and leadership**  
10 **of a coordinated leaf-to-orbit research project only NASA and a Terrestrial Ecology Field**  
11 **Campaign can provide.** The domestic focal areas include the southwestern US, Great Plains,  
12 Mountain West, and Great Basin. We provide detailed motivations of the science questions and  
13 end-user management challenges in each of these areas. While the US focal domain represents  
14 a large portion of the aridity range of global drylands, international sites are necessary to address  
15 science questions at a global scale and to represent the diversity of dryland ecosystems across  
16 which models and hypotheses are tested. Designated candidate international sites were selected  
17 based on queried community needs, scientific track record, and assets of local partners (long-  
18 term field data and instrumentation), as well as logistical feasibility, dependability of plans, and  
19 safety and security. Candidate sites are located in southern Africa, northern Mexico, Australia,  
20 and the Cerrado and Caatinga regions of South America. Leading international partner institutions  
21 (e.g., TERN - Australia; SAEON - South Africa) are members of Global Ecosystem Research  
22 Infrastructure ([GERI](#)), along with the US's NEON, where they coordinate research on global  
23 issues such as ecological drought.

24           Guiding principles of the ARID campaign strategy are to utilize NASA's ground, aircraft, and  
25 spaceborne instruments and missions for: (i) ensuring observations and modeling at scales  
26 appropriate to the spatio-temporal variability and cross-scale feedbacks of drylands, (ii) leveraging  
27 existing assets (data and infrastructure) and investments by partners, (iii) capitalizing on existing  
28 international networks, including the use of supersites for coordinated, synergistic field data  
29 collection, (iv) multi-temporal airborne and UAS acquisitions, (v) supporting ongoing and  
30 upcoming NASA missions, and (vi) balancing operational risk with scientific reward. The  
31 implementation plan is modular and can follow a phased approach where the logistically  
32 straightforward study regions within the US can be targeted first, while more complex international  
33 campaigns are planned.

34           ARID aims to use its research to support end-user goals and is thus fully aligned with NASA's  
35 ES2A strategy to "*advance and integrate Earth science knowledge to empower humanity to create*  
36 *a more resilient world*". The ARID field campaign will support algorithm development and  
37 uncertainty estimation of fundamental remote sensing products that feed directly into existing  
38 workflows of agencies. The ARID field campaign is the embodiment of ES2A as it directly  
39 addresses science questions and informs both existing applications and future adaptation  
40 strategies for federal agencies and Tribal authorities. **The NASA ARID campaign will deliver**  
41 **actionable, high-impact information and insights to many US land managers and a suite of**  
42 **national and international end-users.**

1 ARID’s scoping efforts and white paper formulation build on co-development and  
2 knowledge sharing frameworks that are specific to science and practitioner needs as informed  
3 through over 120 community engagement events, which included workshops, webinars, technical  
4 Working Group meetings, individual discussions, roundtables and in-person visits, Tribal  
5 consultations, and conference town halls with a diverse range of data end-user groups. ARID  
6 events included larger workshops in Arizona (October 2023), New Mexico (May 2024), and  
7 Washington D.C. (July 2024) that brought together research partners, end-users, Tribal Nations,  
8 and agency partners (BLM, USGS, DoD, DOE, NSF, NPS, USFWS, USFS). In addition, the ARID  
9 scoping team has directly worked with end-users representing a suite of managers and decision-  
10 makers responsible for stewardship of dryland ecosystems and natural resources. We have  
11 interacted closely with Tribal groups in recognition of indigenous connections to drylands, and the  
12 powerful opportunities to co-create and share knowledge, as well as to support diverse capacity  
13 building and the meeting of critical Tribal information needs. Finally, the white paper provides  
14 details of international engagements, partners, and their support and commitment to the ARID  
15 campaign, and ARID has strong international commitments and connections across nations and  
16 continents.

17  
18 The ARID mission arrives at a critical juncture, as climate change impacts intensify and the need  
19 for sustainable management of dryland regions becomes more urgent than ever. While fine-scale  
20 spatial variability and rapid temporal dynamics of drylands have largely eluded prior Earth  
21 Observation (EO) technologies, ARID’s tiered observing plan, using multi-sensor airborne,  
22 uncrewed aircraft system (UAS), and in-situ data, will fill critical observation gaps, improving  
23 remote sensing and modeling capabilities. ARID will transform applications of remote sensing and  
24 Earth System Models to assess the current and future dynamics and sustainability of dryland  
25 ecosystems and communities in the face of changing climate and land use. By leveraging existing  
26 data and strong research-management partnerships, ARID can build critically needed actionable  
27 understanding through the use of transdisciplinary science in locations that are logistically  
28 dependable, safe for researchers, and with exceptional relevance for US and international  
29 priorities, making ARID a low risk and high return proposition.

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# Table of Contents

1		
2	<b>A. Dryland Motivation</b>	<b>10</b>
3	<b>A.1 Introduction and Motivation</b>	<b>10</b>
4	A.1.1 Dryland Importance and Change	10
5	A.1.2 Uncertainties in Dryland Processes and their Role in the Earth System	15
6	A.1.3 Dryland Social-Ecological Systems and Connection to End-Users	17
7	A.1.4 The need for a NASA Campaign in Drylands	19
8	A.2 Potential for a major scientific advancement	21
9	A.3 ARID Objectives	23
10	A.4 Relevance to Solicitation and NASA	25
11	A.4.1 NASA Terrestrial Ecology Program and A.4 Solicitation Relevance	25
12	A.4.2 NASA Earth Science Relevance	26
13	A.4.3 Relevance to Earth Science to Action Strategy	27
14	<b>B. Technical Approach and Methodology</b>	<b>29</b>
15	B.1 Science Themes and Questions	29
16	B.1.1 Science Theme 1: Climate Variability and Drought	34
17	B.1.1.1 Sub-Theme: Water Availability	36
18	B.1.1.2 Sub-Theme: Dryland Climate Variability: Pulses and Droughts	37
19	B.1.1.3 Sub-Theme: Fire	39
20	B.1.1.4 Sub-Theme: Land-Atmosphere Interactions	41
21	B1.1.5 Approaches to Studying Drought and Climate Variability	41
22	B.1.2 Science Theme 2: Ecosystem Structure, Function, and Biodiversity	44
23	B.1.2.1 Sub-Theme: Vegetation Structure and Heterogeneity	45
24	B.1.2.2 Sub-Theme: Dryland Biodiversity	45
25	B.1.2.3 Sub-Theme: Ecosystem Function	46
26	B.1.2.4 Sub-Theme: Dryland Geology and Soils	48
27	B.1.2.5 Approaches for Ecosystem Structure, Function, and Biodiversity	50
28	B.1.3 Science Theme 3: Carbon Cycle Interannual Variability and Long-Term Trends	52
29	B.1.3.1 Sub-theme: Carbon Stocks and Fluxes	53
30	B.1.3.2 Sub-theme: Integrated Approach to address Carbon Cycle Theme	55
31	B.1.4 Science Theme 4: Social Ecological Systems	56
32	B.1.4.1 Sub-Theme: Land Management	57
33	B.1.4.2 Sub-Theme: Adaptation and Mitigation Strategies	59
34	B.1.4.3 Research Approaches in Support of Social Ecological Systems	61
35	B.2 Overall Study Design	62
36	B.2.1 Remote Sensing of Drylands and the critical role of NASA sensors	63
37	B.2.2 Dryland Ecosystem Modeling and Model-Data Synthesis	82
38	B.2.3 Coordinated Ground Measurements	91
39	B.2.4 Study Domain and Selection Criteria	95
40	B.2.5 Leveraging Existing Field Campaigns and Networks	100
41	B.2.6 Proposed Field Campaign Implementation Strategy	104



1	B.2.7 Earth Science to Action Strategy and Application	117
2	B.2.8 Technical and Logistical Feasibility	120
3	B.2.9 Modularity and Sequence	121
4	<b>C. Community Engagement and Co-Development</b>	<b>122</b>
5	C.1 ARID Community Engagement Vision	122
6	C.2 ARID Research and Technology Partners and Collaborators	125
7	C.2.1 Domestic Research Partners and Collaborators	125
8	C.2.2 International Research Partners and Collaborators	127
9	C.2.3. Research and Data User Engagement	129
10	C.3 End-user Engagement and Decision Support	131
11	C.4 Tribal Engagement	135
12	C.5 Preparing the Next Generation of Earth Scientists and Practitioners	140
13	C.6 Diversity, Equity, Inclusion, and Accessibility	142
14	C.7 Capacity Building: Education and Training	142
15	<b>D. Management and Plan of Work</b>	<b>144</b>
16	D.1 Management Structure and Personnel Responsibilities	144
17	D.2 Code of Conduct	145
18	D.3 Data, Software, and Information Management	146
19	D.4 Computing Resources	148
20	<b>Acronym List</b>	<b>151</b>
21	<b>E. References</b>	<b>157</b>
22	<b>F. Appendix</b>	<b>179</b>
23	F.1 Science and Application Traceability Matrix	179
24	F.2 ARID Workshop 1: Tucson, AZ	183
25	F.3 ARID Workshop 2: Albuquerque, NM	183
26	F.4 ARID Science Theme Working Groups Participants	183
27	F.5 ARID Technical Working Group Participants	183
28	F.6 ARID Engagement Archive	183
29	F.7 Tribal Engagement Archive	183
30	F.8 Acknowledgement of ARID Scoping Participants	183
31	F.9 Community Feedback: Response to Forms	183
32	F.10 Airborne Remote Sensing Feedback	183
33	F.11 End-User Input	184
34	F.12 Archive of Dryland Efforts that ARID can Leverage	184
35		
36		
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38		

# 1 A. Dryland Motivation

## 2 A.1 Introduction and Motivation

### 3 **A.1.1 Dryland Importance and Change**

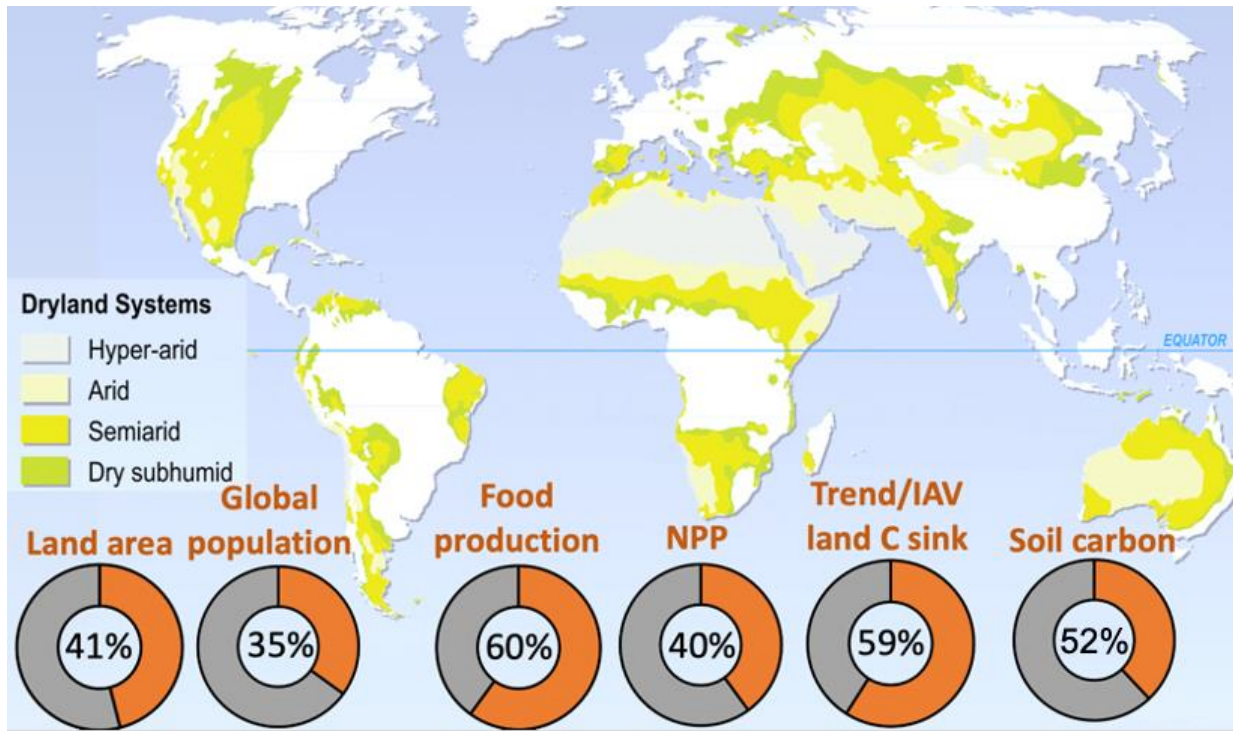
4 *“Traditionally neglected at the policy level, drylands are gaining importance...Globally, the most*  
5 *important emerging issues are: climate change, food security, biodiversity and human security*  
6 *including water scarcity. Such forces are highlighting the value of healthy drylands to the world,*  
7 *and their role in a secure global future.” - United Nations Environmental Management Group*  
8 *(United Nations, 2011).*  
9

10 In October 2011, the United Nations (UN) Environmental Management Group published “*Global*  
11 *Drylands: A UN System-Wide Response*” (United Nations, 2011), which recognized that  
12 drylands have been historically neglected, and which acknowledged drylands’ high value in the  
13 Earth System, as well as the large climatic and land use change threats faced by these  
14 ecosystems worldwide. The extent to which drylands have been subjected to desertification/land  
15 degradation – defined as the loss of biodiversity ecosystem function and ecosystem services  
16 caused by anthropogenic and natural drivers (D’Odorico et al., 2013) – has long been of  
17 significant global concern. Land degradation in drylands inspired the UN Convention to Combat  
18 Desertification (UNCCD) in 1994, to which the United States (US) is a signatory, along with 197  
19 other countries. More recently, UN Sustainable Development Goal (SDG) 15.3 was created,  
20 which states and aspires to accomplish: “*By 2030, combat desertification, restore degraded land*  
21 *and soil, including land affected by desertification, drought and floods, and strive to achieve a*  
22 *land degradation-neutral world*” (UN, 2017).” The science themes of Adaptation and Response  
23 in Drylands (ARID) are closely tied to numerous US and international goals, and ARID could  
24 directly inform and assess the indicators countries are asked to evaluate (e.g., trends in carbon  
25 stocks, resilience to drought, impacts to ecosystem function (Johnson et al., 2017). In this way,  
26 **the new knowledge and datasets ARID would deliver are extremely well-aligned with the**  
27 **targets and information needs outlined by multiple US and international policy efforts,**  
28 **which recognize the importance and high levels of change occurring in drylands, as well**  
29 **as the consequences for social-ecological systems.**  
30

31 Drylands, defined as systems where annual potential evaporative demand is at least 1.5 times  
32 greater than precipitation (Barrow, 1992), encompass a diverse array of ecosystem types, all  
33 unified by a scarcity of water (Figs. A.1, A.2). They span hyperarid to subhumid biomes and  
34 contain vegetation types from grasses, to shrubs, to forests amongst bare soil and biological soil  
35 crusts (Figs. A.1, A.2). In fact, the landscapes of these ecosystems vary greatly in vegetation  
36 and soil composition, including green forested areas to sparsely vegetated, with the only  
37 commonality that they are water-limited (Fig. A.2). Dryland ecosystems represent the planet’s  
38 largest terrestrial biome (~60 million km<sup>2</sup>), making up over 40% of the global land surface  
39 (Safriel et al., 2005; Wang et al., 2022). This area is projected to increase 11%–23% by the end  
40 of the century due to climate change (Huang et al., 2016), partly in response to drying  
41 conditions advecting into downwind wetter transitional regions (Koppa et al., 2024). Drylands  
42 also directly support more than two billion people, or about 35% of the global human population,  
43 meaning drylands are the home to one in three people in the world today (Maestre et al., 2012;  
44 Reynolds et al., 2007; Wang et al., 2022). These water-limited ecosystems have the highest

1 population growth rate of any ecological zone (Gaur & Squires, 2018; L. Wang et al., 2012),  
 2 supporting livelihoods that are commonly tightly coupled to local ecosystem services. Overall,  
 3 due to the location of vast croplands and rangelands, drylands produce 60% of the world's food  
 4 supply (van der Esch et al., 2022; Wang et al., 2022) and account for up to 44% of cultivated  
 5 lands (Burrell et al., 2020a; Cherlet et al., 2018). They also contain substantial mineral and  
 6 energy resources, are areas of rapidly growing alternative energy installations, and support  
 7 unique biological and cultural diversity across flora, fauna, and human communities (Alikhanova  
 8 & Bull, 2023; Cartereau et al., 2023). Importantly, the distinct mechanisms regulating drylands  
 9 are expected to control other ecosystem types in a warmer, drier world (Grünzweig et al., 2022).  
 10 Drylands are, therefore, of great economic, cultural, and ecological global importance.

11



12

13 **Figure A.1.** Map of global drylands and their role in Earth's structure, function, and  
 14 provision of services. Orange fractions of the circles show dryland contributions to global  
 15 land area, population, food production, net primary production, the trend and interannual  
 16 variability in the terrestrial carbon sink, and soil carbon (drylands store 32% of soil  
 17 organic carbon and 79% of soil inorganic carbon to a 2m depth, combining to represent  
 18 52% of Earth's soil carbon to 2m).

19

20 As mitigation options for climate change become an increasing focus, there is more interest  
 21 than ever before in understanding, quantifying, and managing ecosystem carbon cycles (Díaz-  
 22 Martínez et al., 2024). Although long considered an insignificant part of the global carbon cycle,  
 23 in part due to misconceptions of these ecosystems as unproductive "wastelands" (Hoover et al.,  
 24 2020; Mortimore et al., 2009), drylands represent 40% of terrestrial net primary productivity  
 25 (NPP), largely dominate the interannual variability (IAV) in atmospheric CO<sub>2</sub> concentrations,  
 26 and store a third of the planet's soil organic carbon and 79% of its soil inorganic carbon (Plaza

1 et al., 2018; together these pools make up 52% of Earth's total soil carbon to 2 m depth; Fig. 1).  
2 Drylands are estimated to provide an annual CO<sub>2</sub> drawdown on the order of 35 PgC through  
3 plant photosynthesis (Yao et al., 2020)– roughly on par with the tropics (including tropical  
4 forests, wetlands, and agricultural lands) – as well as additional significant drawdown through  
5 drylands' expansive photosynthetic soil communities (i.e., biological soil crusts; Elbert et al.,  
6 2012; Maestre et al., 2013; Wilske et al., 2008). Drylands also exert the greatest influence of  
7 any biome on the IAV of the global terrestrial carbon sink (Ahlström et al., 2015; Fan et al.,  
8 2019; Poulter et al., 2014; Sitch et al., 2024; X. Zhang et al., 2018). Because drylands are so  
9 responsive to changes in water availability, large year-to-year moisture changes (from ENSO,  
10 for example) can greatly affect surface carbon fluxes and substantially influence atmospheric  
11 carbon concentrations (Metz et al., 2023; Poulter et al., 2014). The lack of understanding of this  
12 dryland control on the carbon cycle IAV reduces our capability to quantify country-level emission  
13 reductions. Drylands also strongly contribute to the mean uptake trend (Ahlström et al., 2015; L.  
14 Wang et al., 2022), and have much greater tree carbon storage than formerly assumed (Brandt  
15 et al., 2020; Tucker et al., 2023a). Thus, although drylands have previously incorrectly been  
16 considered relatively unimportant in regulating Earth's carbon cycle and climate, we now know  
17 these ecosystems dominate core aspects of the planet's terrestrial carbon storage and  
18 flux. Nevertheless, our scalable and mechanistic understandings, and our model representation  
19 of the drivers and feedbacks for dryland contributions to the global carbon cycle remain notably  
20 poor.

21  
22 Drylands also have remarkably high biodiversity, including maintaining 35% and 20% of global  
23 biodiversity and plant diversity, respectively (Gross et al., 2024; Maestre et al., 2021). The long  
24 evolutionary history of dryland ecosystems and their role as the origin of many unique plant  
25 lineages has given rise to multiple dryland biodiversity hotspots around the world (Maestre et  
26 al., 2021). Moreover, new work indicates that drylands act as a global reservoir of plant  
27 phenotypic diversity, challenging the pervasive view that harsh environmental conditions reduce  
28 plant trait diversity (Gross et al., 2024). Thus efforts to describe and monitor the unique diversity  
29 of species, form, and function found in drylands are critical (Reynolds et al., 2011). However,  
30 anthropogenic impacts, for example, livestock grazing pressure and changing wildfire regimes,  
31 can cause rapid species decline and loss (Cartereau et al., 2023). We may lose key aspects of  
32 dryland biodiversity before they've been discovered (Cowie et al., 2011; Lewin et al., 2024).

33  
34 Rapidly expanding our knowledge of dryland structure and function is critical not only because  
35 of the biome's large and increasing global importance, but also because drylands are  
36 responding dramatically and rapidly to anthropogenic change. Many dryland ecosystems have  
37 passed or are believed to be close to passing tipping points, which, when crossed, lead to  
38 abrupt system transitions and persistent degraded states. Numerous dryland ecosystem  
39 services are under immediate threat, which, in turn, threatens global carbon cycles, climate  
40 feedbacks, and the well-being of myriad societies (Fraser et al., 2011; Gaur & Squires, 2018; M.  
41 W. Jones et al., 2022). Specifically, drylands are experiencing more extreme heatwaves,  
42 droughts, and fires; for example, in the western US, where the most severe drought in more  
43 than a thousand years is currently taking place, and record-high temperatures are putting  
44 pressure on the region's inhabitants, ecosystems, and their services (Dannenberg et al., 2022;  
45 Williams et al., 2022). Many drylands are drying in terms of annual mean water availability,  
46 including in the western US (F. Zhang et al., 2021). While increased evaporative demand  
47 through warming is causing higher humidity in most global ecosystems, drylands are not  
48 experiencing this higher atmospheric humidity due to already low surface soil moisture (F.

1 Zhang et al., 2021). These climate changes are resulting in rapid ecosystem vegetation shifts,  
2 such as mass dryland forest mortality in the face of “hot drought” (Allen et al., 2010; Anderegg  
3 et al., 2013) and large increases in the extent and severity of wildfire (M. W. Jones et al., 2022).

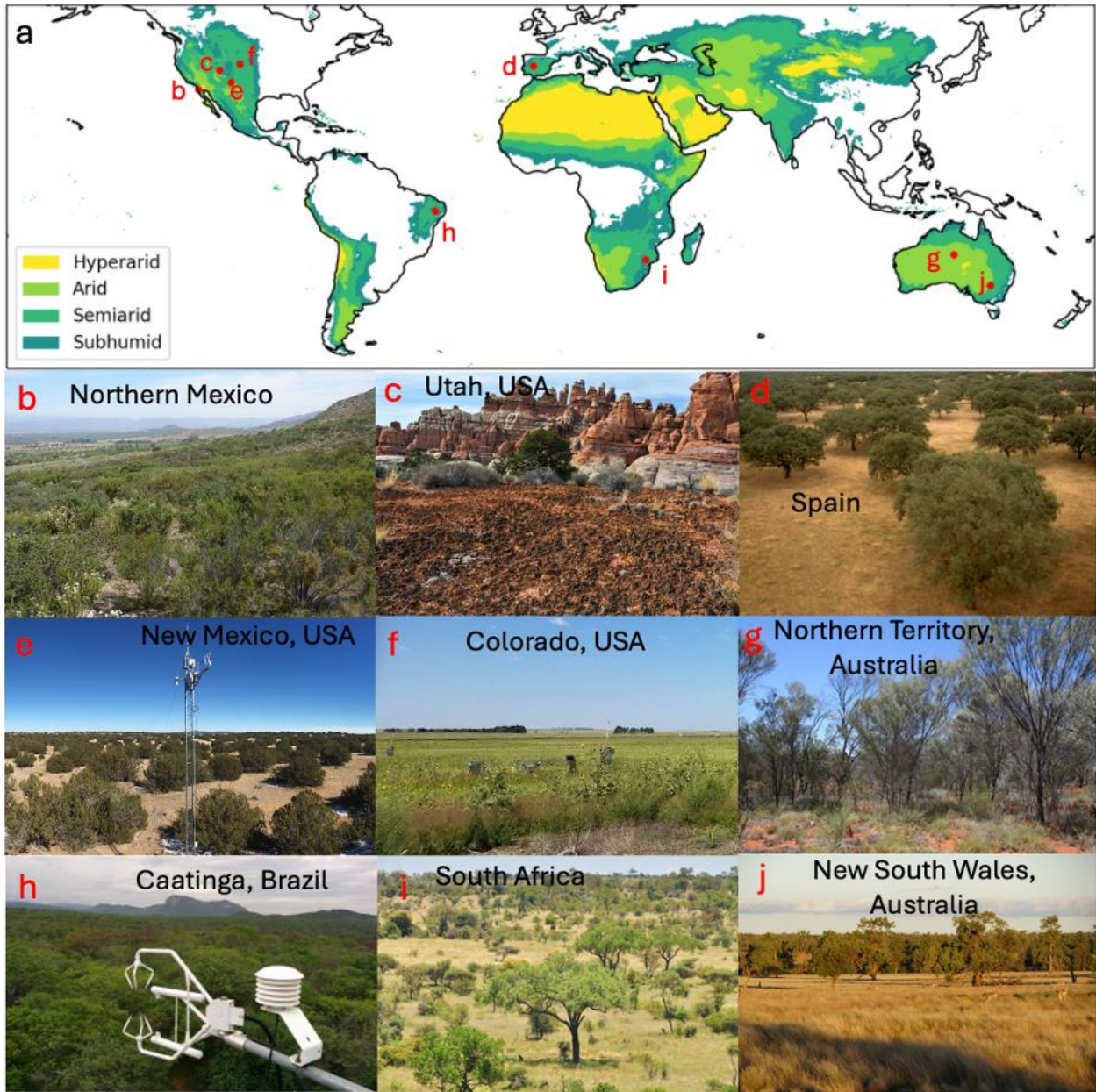
4  
5 Effects of extreme events on drylands have been extensively highlighted by the media of late,  
6 especially in the US. The southwestern US was struck with an extreme drought in 2020 which  
7 depleted the Colorado River flow to unprecedented lows (Fountain, 2020). This drought is a part  
8 of a multi-decadal megadrought that started in the early 2000s and is the most extreme in this  
9 region in over one thousand years (Williams et al., 2022). Furthermore, one of the largest heat  
10 waves struck the western US in 2021 with temperatures reported over 120 degrees Fahrenheit  
11 and with over one thousand heat-related deaths (Bartusek et al., 2022). Additionally, with  
12 unprecedented heat in summer 2024, many western US cities are breaking records for the  
13 amount of time spent above 110 degrees Fahrenheit, especially in Phoenix, AZ (Golembo & El-  
14 Baweb, 2024). These hot conditions are causing hundreds of fatalities (Ramirez, 2024), and are  
15 a part of the drying conditions that have interacted with large scale exotic plant invasion to result  
16 in enormous increases in wildfire and changes in fire seasonality. Numerous articles document  
17 how fires in the western US, in turn, are destroying homes, causing landslides, imperiling  
18 wildlife, and resulting in wholesale changes to plant communities. While only providing a brief  
19 sampling of these recent events, it is clear that the changes and consequences observed in  
20 drylands are extreme, and that they will continue in western US drylands and across the globe.  
21 Because of drylands’ vulnerability, these changes will put pressure on water resources,  
22 ecosystem function, and human well-being for decades to come. Our ability to understand,  
23 monitor, forecast, and make decisions for these rapidly changing ecosystems must evolve now  
24 to keep pace with exceptional rates of change.

25  
26 In recent decades, major land-use change has occurred in global drylands, with agriculture  
27 expanding to meet local and global demands for food, feed, and energy. For example, ~225,000  
28 km<sup>2</sup> of tree-covered drylands were converted to other land-cover types between 1992 and  
29 2015; 56% of that area transitioned to shrubland, while 40% was converted to cropland (L.  
30 Wang et al., 2022). These are expansive shifts in land cover type, with critical implications for  
31 carbon storage, land health, and energy exchange with the atmosphere. Drylands are also the  
32 source of a large proportion of the food and fiber used around the world (e.g., grasslands alone  
33 produce 27% and 23% of the world’s milk and meat, respectively), meaning that what happens  
34 in drylands does not only affect those who live in drylands. For both climate- and land use-  
35 induced change, vegetation transitions will have far-reaching consequences for biodiversity,  
36 carbon, water, and energy budgets, as well as the provision of vital ecosystem services.

37 **Improved quantification of these changes and effects are critically needed to refine**  
38 **model forecasts, inform management decisions, and evaluate mitigation and adaptation**  
39 **options.** Yet, drylands’ heterogeneous landscapes with sparse canopy cover and pulse-driven  
40 activity have historically presented significant remote sensing and modeling challenges for  
41 estimating water and carbon flux, land cover classification, aboveground biomass, and many  
42 other crucial applications.

43  
44 This white paper outlines a targeted plan to deploy field and NASA airborne instruments to  
45 vastly augment data derived from satellite observations that, when joined, will substantially  
46 advance quantification of drylands’ large and changing role in the Earth system. There are  
47 numerous management and conservation options for addressing the challenges of land use and  
48 climate change in drylands, but their utility at relevant scales requires an advanced generation  
49 of remote sensing tools, sensors, and dryland-specific algorithm development. This need is  
50 underscored by the fact that dryland ecology and remote sensing are understudied (Bond-  
51 Lamberty & Thomson, 2010; Ciais et al., 2011; Maestre, Quero, et al., 2012; W. K. Smith et al.,

1 2019a). We urgently need a NASA-driven field campaign to advance the science and  
2 technology that can provide a data-informed understanding of dryland processes and  
3 feedbacks to inform global resilience efforts.  
4



5  
6 **Figure A.2.** (a) Global dryland distribution based on aridity index computed from  
7 TerraClimate precipitation and radiation data (Abatzoglou et al., 2018). Aridity  
8 index (AI) is defined as in Wang et al. (2022) as precipitation/potential  
9 evapotranspiration with hyperarid as  $AI < 0.05$ , arid as  $AI = 0.05-0.2$ , semiarid as  
10  $AI = 0.2-0.5$ , and subhumid as  $AI = 0.5-0.65$ . (b-f) Sample pictures of different  
11 dryland landscapes (mainly arid and semi-arid, as defined by TerraClimate),  
12 which can vary substantially in structure and composition within the same regions  
13 and across the globe. (b) Arid shrubland site in Northern Mexico (MX-EMg  
14 MexFlux site; photo obtained from <https://ameriflux.lbl.gov/sites/siteinfo/MX->

1 *EMg#image-gallery* with permission to reuse from the site investigators (Cueva et  
2 al., 2020)). (c) Arid desert ecosystem with dominant biological soil crust coverage  
3 in Utah, United States of America (U.S.; photo used with permission from Bill  
4 Bowman). (d) Semi-arid grassland and evergreen broadleaf forest site in central  
5 Spain (eslm1 PhenoCam site, photo obtained from  
6 <https://phenocam.nau.edu/webcam/sites/eslm1/> with permission to reuse from  
7 the site investigators (Seyednasrollah et al., 2019)). (e) Semi-arid juniper  
8 savanna site in central New Mexico, U.S. (US-Wjs AmeriFlux site; photo owned  
9 by the authors). (f) Semi-arid cropland site in central Colorado, U.S. (US-xSL  
10 AmeriFlux and NEON site; photo obtained from  
11 <https://ameriflux.lbl.gov/sites/siteinfo/US-xSL> with permission to reuse from the  
12 site investigators (Metzger et al., 2019)). (g) Arid woodland site in Northern  
13 Territory, Australia (AU-ASM OzFlux site; photo obtained from  
14 [https://www.ozflux.org.au/monitoringsites/alicesprings/alicesprings\\_pictures.html](https://www.ozflux.org.au/monitoringsites/alicesprings/alicesprings_pictures.html)  
15 with permission to reuse from the site investigators (Cleverly et al., 2013). (h)  
16 Semi-arid needleleaf forest site in the Caatinga region of Brazil (BR-CST  
17 AmeriFlux; photo obtained from <https://ameriflux.lbl.gov/sites/siteinfo/BR-CST>  
18 with permission to reuse from the site investigators (Antonino, 2019)). (i) Semi-  
19 arid savanna ecosystem in Kruger National Park, South Africa (photo owned by  
20 the authors).  
21 (j) Arid savanna site in Yathong Nature Reserve in New South Wales, Australia  
22 (photo owned by the authors).

### 23 **A.1.2 Uncertainties in Dryland Processes and their Role in the Earth System**

24 While new measurements and modeling show the large importance of drylands in the global  
25 system, substantial uncertainties remain regarding the drivers, magnitude, and vulnerability of  
26 these contributions. Dryland systems are characterized by large interannual variability in  
27 weather conditions and heterogeneous vegetation coverage (Fig. A.2), which have hindered  
28 past estimates of their role in the Earth system. Current trends in climate and land use changes  
29 translate into dryland ecosystem dynamics becoming more variable and trajectories more  
30 uncertain. For instance, long-term drought conditions are resulting in vegetation mortality that  
31 further impacts numerous ecosystem services (Jiao et al., 2021). Such changes from forests to  
32 grasslands or from grasslands to shrublands (Eldridge et al., 2015) have large consequences,  
33 for example, on carbon and energy feedbacks to climate change. Yet the magnitude and even  
34 the sign of these feedbacks remains highly uncertain. Weather variability is being experienced  
35 through pulsed dynamics of periods of extreme dryness and heat stress (Bradford et al., 2020)  
36 interspersed with large rainfall events that result in flooding, gully erosion, and the movement of  
37 soils to create landslides, fill reservoirs, induce dust storms (Luo et al., 2020), and reduce soil  
38 health of these dryland systems.

39 Water resources are further stressed due to the rapid emergence of high levels of evaporative  
40 demand pulling moisture from the vegetation and soils (Ojima, 2021; Hobbins et al., 2016). In  
41 fact, many dryland regions, such as the western US, are experiencing longer dry spells and less  
42 annual rainfall (F. Zhang et al., 2021). At the same time, there is increased water use and  
43 consumption to meet food, potable water, and energy needs to support rapidly growing dryland  
44 urban centers. In addition, water resources in dryland regions that sit in the rain-shadow of  
45 mountain regions are experiencing warming that has accelerated snow melt and glacier loss  
46 that feed rivers and aquifers. The future availability of water for dryland inhabitants, ecosystems,  
47 and agriculture remains highly uncertain, with severe implications for society.

1 These water resource uncertainties translate to concerns for global food security. ~90% of the  
2 world's dryland population lives in developing countries, which are facing developmental  
3 pressure to increase food production under water-limited conditions (L. Wang et al., 2012) and  
4 which do not have the resources to withstand reductions in food. Rangelands constitute the  
5 world's largest land use, and the majority of these fall in drylands, which have a large  
6 contribution to beef production (Maestre et al., 2022a). Global rangelands are experiencing  
7 significant changes to community composition (e.g., from shrub encroachment) and to forage  
8 productivity due to current and future climate change (Boone et al., 2018; Godde et al., 2020),  
9 with significant impacts on the livelihoods of more than a billion people (Maestre et al., 2022a).  
10 Relatedly, there is significant likelihood that climatic change will result in agricultural  
11 abandonment of crops in drylands as irrigation water becomes more limiting (ref). Yet the  
12 effects of these changes on food security, dust production, and feedbacks to climate change  
13 remain unclear.

14 Dryland carbon fluxes and stocks are some of the most uncertain of any biome (Fawcett et al.,  
15 2022; Rodriguez-Veiga et al., 2019). Many global scale land surface models are unable to  
16 replicate the mean net or gross carbon uptake or its year-to-year variability observed at in-situ  
17 flux tower sites (MacBean et al., 2021; Teckentrup et al., 2021; L. Wang et al., 2022). It is critical  
18 to reduce the uncertainty of dryland contribution to the global carbon cycle with measurements  
19 given their large demonstrated role in carbon cycling at the global scale. *In-situ* water stores  
20 (soil moisture) and fluxes (evapotranspiration) are more accurately represented in models, but  
21 relationships between these variables are poorly represented (Short Gianotti et al., 2019).  
22 Additionally, water flux partitioning between plant transpiration and bare soil evaporation is  
23 poorly represented, suggesting that compensating errors are present (MacBean et al., 2020).  
24 Challenges for modeling drylands include their daily-timescale pulse-lag dynamics with often  
25 lagged surface responses to rainfall events and multiple day-to-week responses that models do  
26 not yet capture (Feldman, 2024a; Ogle et al., 2004; Post & Knapp, 2021). The complexity of  
27 varied and asynchronous dryland plant responses to limited and changing water availability is  
28 not represented in global scale models. This is partly due to limited data used in the models'  
29 development, evaluation, and calibration, and partly because of limited cross-scale model  
30 activities that could be leveraged to facilitate global scale model development. Additionally, most  
31 modeling activities were developed using process understanding from mesic ecosystems, which  
32 often does not translate to sufficiently represent drylands.

33 Surprisingly, despite increasing atmospheric aridity and often declining soil moisture trends,  
34 many drylands are greening (Dardel et al., 2014; Gonsamo et al., 2021; Lian et al., 2021; Song  
35 et al., 2018), highlighting the complicated uncertainties for this biome. While CO<sub>2</sub> fertilization is  
36 likely playing a large role in these trends (Zhu et al., 2016), it is uncertain how much and for how  
37 long the greening will continue (Burrell et al., 2020, Lian et al., 2021). Such greening might  
38 ultimately mitigate some of the dryland expansion effects (Berg & McColl, 2021) and the state  
39 changes observed in non-greening dryland ecosystems. However, current modeling approaches  
40 often do not explicitly account for rapid ecosystem change in their proportioning of the relative  
41 contributions of different drivers of greening (e.g., elevated CO<sub>2</sub> versus land-use change), which  
42 means they can miss or underestimate the rapid changes driven by processes like extreme fires  
43 and changes in land use (Burrell et al., 2020b; K. J. Wessels et al., 2012). Disentangling the  
44 roles of temperature, CO<sub>2</sub>, and land use in dryland productivity has been identified as a key  
45 knowledge gap by the UNCCD, IPCC, and the Intergovernmental Science–Policy Platform on  
46 Biodiversity and Ecosystem Services (ref). At the same time, drylands and, in particular,  
47 rangelands are being discussed as a focal location to store carbon for climate mitigation, and  
48 ranchers are already being paid for carbon sequestration in the US and internationally (Briske et



1 al. 2023). Yet improvements to remote sensing and modeling will certainly be required to  
2 quantify the carbon stored or lost and to inform any carbon payment or credit system.

3  
4 Carbon sequestration is only one area where drylands are explicitly targeted as climate  
5 mitigation or nature based solutions, and for which uncertainties in dryland contributions create  
6 challenges for decision making. Given their large land area, drylands are viewed for their  
7 afforestation and renewable energy potential as a natural climate solution. However, poor  
8 understanding of the dryland carbon and energy balance greatly limits the ability to quantify  
9 potential benefits (or unintended consequences) and prevents prediction of benefits under  
10 future climate and land cover conditions (Barron-Gafford et al., 2016; Liu et al., 2022; Novick et  
11 al., 2024). Indeed, in many instances in the wet tropics and temperate ecosystems, tree planting  
12 can increase carbon uptake and cool the surface (Duveiller et al., 2018). As such, the African  
13 Great Green Wall - a colossal initiative to restore 100 million hectares of degraded ecosystems  
14 across 11 countries in the Sahel region - is proposed as a major natural climate solution for  
15 increased terrestrial carbon uptake and surface cooling benefits (Mirzabaev et al., 2021).  
16 However, given water limitation in drylands, these benefits may only be conferred in some  
17 instances (Rohatyn et al., 2022; L. Wang & D'Odorico, 2019), and increasing vegetation may  
18 even *warm* the surface through reduced albedo without a sufficient compensating evaporative  
19 cooling (Rotenberg and Yakir 2010, Williams et al. 2021, *Sci. Adv.*, Feldman et al. 2023). Thus,  
20 uncertainties in how drylands will respond to management, including management intended as  
21 a climate solution - results in significant challenges for decision making in these regions.

22  
23 Historically remote sensing efforts to monitor drylands and land degradation were largely limited  
24 to time-series analysis of vegetation index data from low spatial resolution sensors (e.g.,  
25 MODIS), using trends in greenness as a proxy for ecosystem function and overall vegetation  
26 cover (Fensholt 2007, Wessels 2013, Wessels 2007). However, monitoring greenness alone  
27 could not differentiate changes in vegetation structure and composition of plant functional types  
28 the trees, shrubs, and grasses to understand the nature of the ecological changes (Smith 2018).  
29 These past remote sensing challenges have meant that, despite their global importance and  
30 high vulnerability to change, our scalable understanding and capacity to forecast dryland  
31 dynamics, as well as impacts to ecosystems and society, is notably poor.

32  
33 These limitations are now greatly reduced with the enhanced resolution and diversity of current  
34 and planned ground based, airborne, and spaceborne sensors. ARID airborne and UAS data in  
35 combination with advanced process models and new satellite sensors enable us to characterize  
36 vegetation composition, structure, and function at the appropriate scales. Accordingly, drylands  
37 represent a large but now addressable uncertainty in our understanding of terrestrial  
38 ecosystems. The ARID field campaign will investigate process level changes in ecosystem  
39 functions and services to further understand the impacts of these changes and potentially  
40 reduce uncertainties of dryland ecosystems.

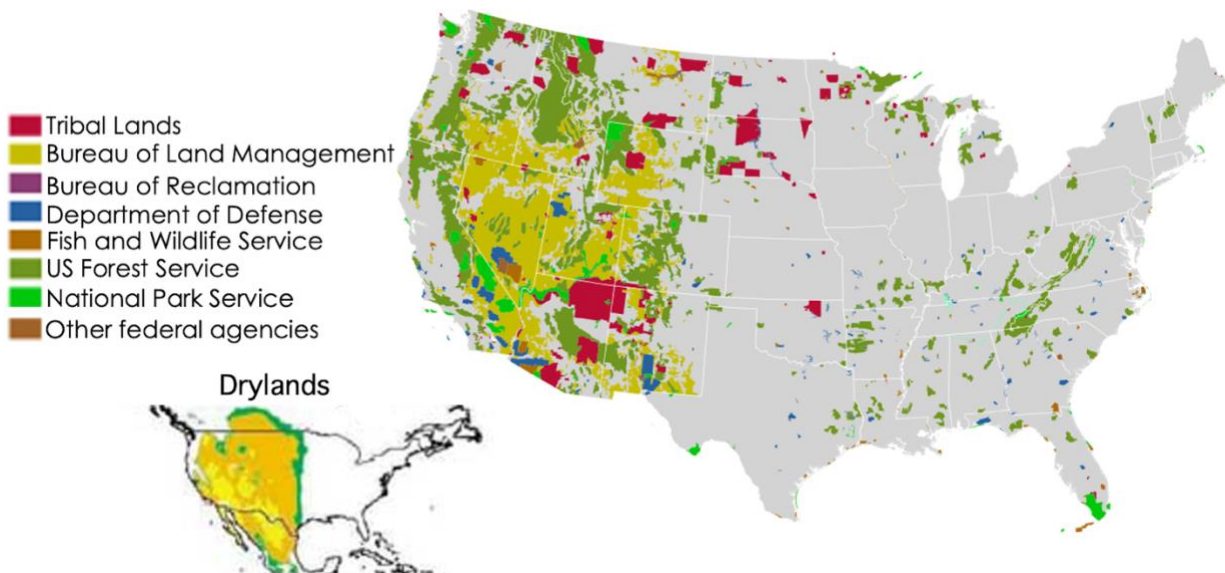
### 41 ***A.1.3 Dryland Social-Ecological Systems and Connection to End-Users***

42 Those who manage dryland ecosystems have the increasingly challenging task of making  
43 decisions for vulnerable social-ecological systems that are experiencing substantial change. In  
44 the US, drylands are managed by federal and state agencies, Tribal nations, major corporations  
45 such as mining and cement industries, non-governmental organizations, and private  
46 landowners. Lands managed by federal and tribal entities are particularly expansive, and there  
47 is high overlap between the location of these lands and the occurrence of drylands (Fig. A.3).  
48 For example, the US Bureau of Land Management (BLM) alone manages nearly 10% of all land  
49 within the US, with dominant lands in western drylands. This represents a sizable challenge -

1 drylands are notoriously difficult to manage due to their dry climates, low soil fertility, and  
2 responsiveness to perturbation - but also a substantial opportunity. The ARID Terrestrial  
3 Ecology field campaign would provide urgently needed information on the status, trends, and  
4 management options of drylands. An improved understanding of the drivers of change,  
5 ecosystem responses, and adaptation and mitigation potential for drylands can inform decisions  
6 and improve outcomes for vast areas of the US and globe. Land managers and decision makers  
7 are asking for this information and a coordinated campaign can meet these needs.

8  
9 A powerful component of ARID is the exceptionally strong partnerships with land managers in  
10 federal agencies and tribal authorities who were integral in producing ARID's research  
11 framework and themes. The NASA ARID campaign will bring sharp focus to the natural  
12 resources that decision-makers are struggling to manage across diverse sectors in the face of  
13 change. In addition to addressing key science questions, ARID has collaborated with diverse  
14 end-user communities, including numerous U.S. agencies motivated to improve science-based  
15 decision making, who are enthusiastic about the new knowledge and predictive understanding  
16 ARID will provide (Fig. A.3). Our approach is informed by and is extremely well aligned with  
17 NASA's Earth Science to Action initiative, and we have convened a global community of  
18 science, policy, and management practitioners to design a targeted field campaign addressing  
19 the most pressing questions facing drylands.

20  
21 **A project such as ARID that improves the understanding of current and future dryland**  
22 **conditions, quantifies the drivers and consequences of global change, and informs the**  
23 **options for improved and adaptive decision making will support science-based land**  
24 **management at an astounding scale. We stress that ARID's framework and approach are**  
25 **designed based on substantial input from both the scientific and data end-user**  
26 **communities.**



28  
29 **Figure A.3.** Comparisons of U.S. federal and tribal lands with the location of  
30 drylands (lower left inset) highlights the vast area of drylands managed by public  
31 and tribal land managers in western North America. Due to low precipitation and  
32 soil fertility, drylands are notoriously difficult to manage, and they are  
33 transforming rapidly in the face of global change. These conditions result in

1 *exceptional challenges for land managers, who have co-produced ARID to meet*  
2 *their information needs. ARID will support a diverse range of federal and tribal*  
3 *land managers in providing actionable science to inform decision making in*  
4 *drylands.*  
5

#### 6 **A.1.4 The need for a NASA Campaign in Drylands**

7 Drylands present key uncertainties in estimates of ecosystem processes and state variables  
8 that can only be quantified and disentangled through coordinated field, airborne, satellite-based,  
9 and modeling analysis. Drylands pose unique remote sensing and modeling challenges, as they  
10 are characterized by high spatial heterogeneity and high temporal variability caused by low and  
11 irregular precipitation. Only NASA can provide the satellite capability to evaluate ecosystem  
12 structure and function at the scale needed for Earth's heterogeneous drylands. A long-term  
13 NASA field and airborne campaign will fill multiple observation gaps (physical parameters) at the  
14 necessary spatio-temporal resolutions within a multi-scale framework to address critical science  
15 questions in drylands (see modified SATM E.1 for details). Previous NASA large-scale  
16 Terrestrial Ecology field campaigns have demonstrated the value of long-term field campaigns  
17 in addressing fundamental ecological questions. Like these previous campaigns, ARID will  
18 implement a multi-scaled approach to integrate remote sensing observations, ground-based  
19 observations, and modeling approaches to quantify and assess ecosystem drivers, responses,  
20 vulnerabilities, and consequences in the context of global change. We explicitly consider how  
21 dryland measurements must link across scales and address the considerable spatial and  
22 temporal heterogeneity found in these systems (see section B.2.1), as well as plan for the  
23 substantial leveraging of existing data, experimental field sites, global networks, cross-institution  
24 partnerships, and a wide range of research infrastructure (e.g., eddy covariance networks,  
25 global change experiments, natural aridity gradients).  
26

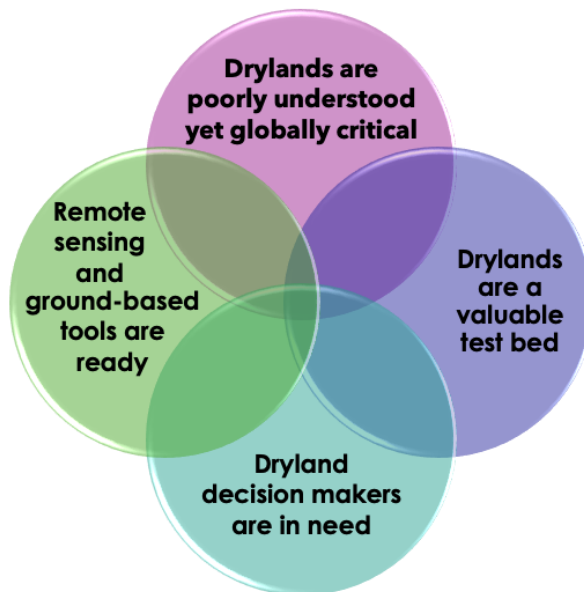
27 Due to drylands' high spatial and temporal heterogeneity, vast spatial expanse, and dramatic  
28 rates of change, satellite data are key to an accurate understanding of dryland trajectories under  
29 global change. However, **remote sensing has been challenged by drylands' low vegetation**  
30 **signals, soil contribution to observations, and very high spatial (often meter scale soil-**  
31 **plant variability) and temporal variability (hourly to daily ecosystem responses), which**  
32 **make it hard to track or predict short- and long-term key variable changes and attribute**  
33 **them to specific drivers** (W. K. Smith et al., 2019b). For example, estimating the fractional  
34 cover of different vegetation types in spatially heterogeneous and sparsely vegetated tree-  
35 grass-shrub ecosystems is challenging (Pervin et al., 2022), resulting in considerable global  
36 land surface model uncertainty in predicted carbon, water, and energy budgets (Hartley et al.,  
37 2017a). There is a need to join satellite data with high-resolution data from airborne and UAS  
38 data acquisition to support plot to regional scale analyses (W. K. Smith et al., 2019b). Moreover,  
39 photosynthetic soils (biological soil crusts) characteristic of drylands are thought to make up  
40 12% of Earth's land surface, but these organisms are typically not included in fractional cover  
41 classification algorithms or models. Their impacts on ecosystem carbon, water, and energy  
42 budgets, therefore, cannot be predicted, nor can their contribution to satellite observation data,  
43 representing a potentially important source of error.  
44

45 Furthermore, drylands are driven by periodic precipitation pulses interspersed with extended dry  
46 periods (Collins et al., 2014; Feldman, Feng, et al., 2024a; F. Zhang et al., 2021) and  
47 experience irregular growing seasons (Cawse-Nicholson et al., 2021). Global scale process  
48 base land surface models are not capturing such responses (MacBean et al., 2021; Renwick et  
49 al., 2019; Teckentrup et al., 2021; Whitley et al., 2016). There is, thus, a need for continuous or

1 higher-frequency measurements to capture ecological responses and better parameterize  
2 models (W. K. Smith et al., 2019b). Capturing the data at the appropriate spatiotemporal scales  
3 will require a coordinated multi-sensor field campaign that allows the research community to  
4 address observation gaps (linked to poor process understanding and predictive capability) and  
5 linkages across scales. **NASA is the clear lead for such an endeavor to scale observations  
6 and processes from leaf to orbit in drylands**

7  
8 In addition to improving our understanding of these important water-limited ecosystems,  
9 drylands provide a rigorous test of the ability of modern remote sensing to capture highly  
10 dynamic, spatially complex, and globally significant ecosystems. Drylands have consistently  
11 played impactful roles in remote sensing development and evaluation. Due to lower cloud cover  
12 and higher likelihoods of favorable atmospheric conditions, many remote sensing techniques  
13 were developed in drylands, including early retrievals of surface reflectance and estimates of  
14 vegetation condition using red and near-infrared bands, which would later become the  
15 Normalized Difference Vegetation Index (NDVI). Increased spatial resolution and the ever-  
16 expanding array of current and future space-based NASA missions provide ideal conditions for  
17 testing and innovating drylands.

18  
19 The recent advances in the spectral, spatial, and temporal capabilities of remote sensing and  
20 ground-based technologies mean a NASA Terrestrial Ecology field campaign focused on  
21 drylands can now provide the same exceptional advances in our understanding and support of  
22 our planet’s dry ecosystems. **Now is the ideal time for a NASA Terrestrial Ecology field  
23 campaign focused on drylands (Fig. A.4), and only NASA can deliver the “leaf to orbit”  
24 perspective required to address the complexity, dynamics, and diversity of these critical  
25 ecosystems.** ARID was developed based on the ideas and information needs of scientific and  
26 data end user communities, and represents a strategic way forward for facilitating NASA’s ability  
27 to revolutionize the quantitative understanding and management options for drylands in the  
28 Earth System.



**Figure A.4.** *Now is the ideal time for a dryland-focused NASA terrestrial Ecology field campaign. Emerging data show the critical contributions of drylands to the Earth system, as well as their responsiveness to change. Remote sensing and ground-based tools finally have the spatial and temporal resolution needed to assess heterogeneous drylands, and these systems represent a valuable testbed for evaluating and improving the accuracy and utility of satellite data. Finally, land managers and other decision makers in drylands are asking for this research. They need the cross-scale, actionable science that can inform management, adaptation, and mitigation solutions for Earth’s drylands.*

48 We propose the Adaptation and Response in Drylands (ARID) field experiment, based on our  
49 scoping study between 2023-2024. We describe activities for an interdisciplinary project that  
50 addresses critical scientific questions and observation gaps by integrating novel airborne and  
51 space-based remote sensing with ground-based measurements and modeling. The proposed

1 four focal study areas within the U.S., as well as an additional four international sites that  
2 include northern Mexico, drylands across southern Africa and Australia, and the Cerrado,  
3 Caatinga, and Gran Chaco in South America.  
4

## 5 A.2 Potential for a major scientific advancement

6 Effective Earth Observation for the entire planet is within reach. Advances in remote sensing  
7 technologies have dramatically increased our ability to observe rapid changes at fine scales.  
8 Drylands provide a rigorous test of the ability of modern remote sensing to capture highly  
9 dynamic, spatially complex, and globally significant ecosystems.

10  
11 Over the past four decades, NASA's Terrestrial Ecology Field Campaigns have focused on  
12 temperate ecosystems (FIFE), boreal ecosystems (BOREAS), tropical forests (LBA), and Arctic-  
13 Boreal systems (ABOVE), and these efforts have demonstrated the power of coordinated field  
14 campaigns to vastly advance our understanding of complex systems, to build long lasting  
15 collaborations, and to train the next generation of Earth system scientists. While other  
16 ecosystems, such as boreal and tropical forests, have received long duration coordinated  
17 research efforts, drylands have not yet received this level of scientific attention. NASA has  
18 invested in dryland campaigns like NEESPI, SHIFT, EOS, HAPEX-Sahel and SAFARI-2000.  
19 However, they were shorter in duration, more limited in spatial domain, and/or did not involve  
20 personnel at the scale a NASA Terrestrial Ecology Field Campaign can support in ARID.  
21 Additionally, other agencies like the U.S. Department of Energy have invested in Arctic and  
22 tropical forest ecosystems through the Next Generation Ecosystem Experiments (NGEE), which  
23 have brought together diverse members of the science community. **Such a scale of effort and  
24 degree of community building has not yet occurred for dryland ecosystems and  
25 represents a substantial and timely research opportunity.** The remote sensing, modeling,  
26 and ground-based tools are also now much more capable of capturing the large spatial and  
27 temporal heterogeneity in dryland systems. Thus, **for the first time, ARID can provide a large-  
28 scale research effort to make substantial progress toward addressing critical dryland  
29 research and land management knowledge gaps.**

30  
31 A NASA dryland terrestrial field campaign would result in major scientific advancement in five  
32 core areas. First, ARID will provide a substantially improved understanding of dryland  
33 contributions to global processes and how they respond to change. Drylands are understudied  
34 despite being globally extensive, with growing evidence showing they play a disproportionate  
35 role in the climate system. Joining remote sensing approaches with ground-based  
36 measurements and monitoring, as well as with statistical and process-based modeling and data  
37 analysis, will dramatically clarify the role of drylands in the Earth system. Second, the current  
38 generation of space-based, aircraft, and UAS remote sensing sensors offers unprecedented  
39 potential to scale processes and services across areas that cannot be measured on the ground  
40 in heterogeneous drylands. ARID would advance our spatio-temporal understanding of how  
41 drylands have already changed - including their growth into more mesic areas, and identify the  
42 main drivers that determine the size and direction of change. Third, ARID would directly support  
43 model development and evaluation. Ecosystem and Land/Earth System Models are currently  
44 woefully poor at capturing dryland drivers and responses to change, which drive large  
45 uncertainties in our understanding of global biogeochemical cycles and forecasts of future  
46 climate (Humphrey et al., 2018; Simpson et al., 2024a; Teckentrup et al., 2021). ARID would  
47 directly address this critical knowledge gap by supplying the datasets needed to parameterize  
48 and evaluate dryland mechanisms, processes, and patterns across scales and into the future.

1 Fourth, ARID will offer a powerful opportunity for calibration and validation of current and  
 2 planned NASA satellite missions. Given a relative lack of focus on drylands throughout satellite  
 3 missions, there are large uncertainties in satellite retrievals of dryland geophysical properties.  
 4 For example, retrievals across measurement frequencies are hindered by mixed bare soil,  
 5 biocrust, and plant fractions, which creates errors in retrievals of dryland infrared VIs and LST.  
 6 Fifth, remote sensing tools in drylands offer impactful opportunities to support science-informed  
 7 decision making and for monitoring, evaluating, and prioritizing mitigation and adaptation  
 8 solutions .  
 9

10 Water scarcity is a key concern in drylands, but the exact mechanisms that govern water  
 11 distribution—the partitioning of precipitation into evaporation, transpiration, runoff, and  
 12 groundwater recharge—remain complex and not entirely predictable. The spatial and temporal  
 13 variability of soil moisture, influenced by factors like vegetation patchiness and proximity to  
 14 water sources, adds further layers of uncertainty. By employing NASA’s cutting-edge remote  
 15 sensing technologies and refining models, there is significant potential to uncover new insights  
 16 into these processes and controls, improving our ability to predict and manage water resources  
 17 in these fragile environments.  
 18

19 Climate change is expected to exacerbate the challenges facing drylands, with increasing  
 20 frequency and intensity of droughts, heatwaves, and wildfires. Yet, many questions remain  
 21 about how these changes will alter the structure and function of dryland ecosystems. For  
 22 instance, the feedback loops between vegetation changes, fire regimes, and climate are not  
 23 fully understood, and current models often struggle to capture the rapid, non-linear responses  
 24 typical of these systems. Exploring these dynamics through innovative research approaches  
 25 could lead to breakthroughs in our understanding of dryland resilience and the development of  
 26 more effective management strategies.  
 27

28 Moreover, drylands represent a complex interplay between ecological processes and human  
 29 activities, with land use practices such as grazing, agriculture, and energy production deeply  
 30 intertwined with ecosystem health. The impacts of these practices under changing climatic  
 31 conditions are still not fully known, particularly in terms of how they might affect the long-term  
 32 sustainability of both the environment and the communities that depend on it. This uncertainty  
 33 underscores the need for integrated research that not only addresses ecological questions but  
 34 also explores the socio-economic dimensions of dryland management, opening up possibilities  
 35 for new strategies that enhance resilience and sustainability in these vital regions.  
 36

37 Our proposed field campaign aims to address scientific questions spanning themes including  
 38 water availability, dryland climate variability, fire, land atmosphere interactions, carbon stocks  
 39 and fluxes, vegetation structure, heterogeneity and biodiversity, geology and soil processes,  
 40 and land management to inform adaptation and mitigation strategies (Table A.1). The scientific  
 41 advancements that would be enabled by the ARID campaign are summarized in Table A.1.  
 42

43 **Table A.1** *Expected advances in dryland science themes within the ARID campaign.*

Topic	Scientific discovery enabled by ARID
1.1.1. Water Availability	Key controls of water availability in drylands
1.1.2. Dryland Climate Variability: Pulses, Deluges, and Droughts	The impact of heterogeneous rainfall pulses on carbon, water and energy exchange and feedback to synoptic weather.

1.1.3. Fire	The impacts of fire on the composition, structure, and function of drylands at various time scales and feedbacks to the fire regime.
1.1.4. Land Atmosphere Interactions	Connecting the drivers of carbon, water, and energy exchange from local to global scales
1.2.1. Vegetation Structure	The controls and impacts of climate on three-dimensional vegetation structure and its spatial heterogeneity
1.2.2. Dryland Biodiversity	Quantification of vegetation biodiversity and identification of hotspots in drylands
1.2.3. Ecosystem Function	The magnitude and duration of plant hydraulic and photosynthetic response to climate
1.2.4. Dryland Geology and Soil Processes	Novel spectral descriptors of soil type and function, enabling new discovery
1.3.1. Carbon Stocks and Fluxes	Direct, observational quantification of the role of drylands in the variability of global carbon sink.
1.4.1. Land Management	Ability to monitor and forecast impacts of land management choices on biogeochemical and hydrological cycles
1.4.2. Adaptation and Mitigation Strategies: a Socio-Ecological System Perspective	Holistic information needed to understand social-ecological interactions and feedbacks.

1

## 2 A.3 ARID Objectives

3 ARID's core mission is to reduce human and ecosystem vulnerability to environmental change  
4 through groundbreaking research using multi-scale remote sensing, in-situ measurements, and  
5 modeling. While we know drylands are critical in the Earth system and that the interacting  
6 pressures of climate and land use change are dramatically altering these ecosystems, there are  
7 significant knowledge and observation gaps that limit the ability to predict the manner,  
8 magnitude, and options for mitigating/adapting to current and future changes in drylands. These  
9 gaps are the focus of ARID's research framework. The four themes and overarching science  
10 questions to be addressed by ARID are based on the need to develop a quantitative  
11 understanding of drylands and their responses to change, to predict future changes and  
12 feedback, and to inform decision making strategies to sustain drylands and the vital ecosystem  
13 services they provide. With this approach, ARID represents both cutting edge and actionable  
14 science.

15  
16 To address shared foundational and applied science goals, ARID worked extensively with the  
17 research and end-user communities to develop four main science themes and associated  
18 overarching questions (see Section B.1 for more details):

19  
20 ***Climate Variability and Drought***

1 How are climate extremes like droughts, heatwaves, and large rain pulses impacting  
2 dryland systems and how do they interact with changing fire regimes, land cover  
3 change, and land-atmosphere interactions?

4 **Ecosystem Structure, Function, and Biodiversity**

5 What are the main mechanisms driving the spatiotemporal distributions of dryland  
6 structure, function, and biodiversity and what is their vulnerability to change?

7 **Carbon Cycle Interannual Variability and Long-Term Trends**

8 What is the contribution of drylands to the mean, trend, and interannual variability of  
9 terrestrial carbon uptake and what drives these patterns?

10 **Social-Ecological Systems**

11 What are the consequences of changes in drylands for social-ecological systems and  
12 what management (e.g., mitigation and adaptation) solutions can maintain the critical  
13 services provided by drylands even in the face of change?

14 Using these questions and themes, ARID aims to elucidate the **drivers** of change within  
15 drylands, the **responses** of ecosystems and social-ecological **consequences** of those  
16 responses, as well as the options **mitigation and adaptation** could provide for maintaining  
17 dryland ecosystem structure and function (Fig. A.5). The interconnections between ARID's  
18 questions, themes, and approach allow for an iterative strategy that synergistically informs our  
19 larger understanding of drylands in the Earth system.

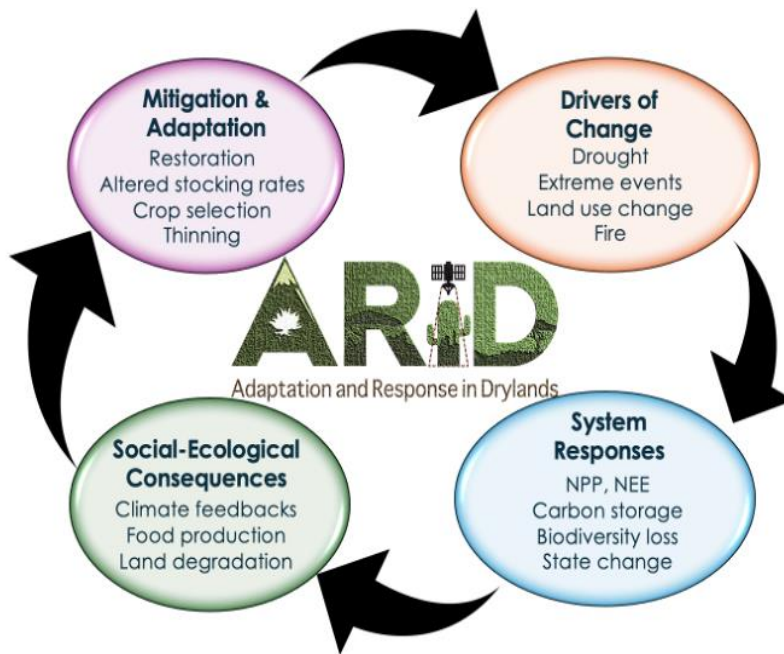


Figure A.5. ARID is structured to the cycles of change in Earth's drylands, elucidating the causes, consequences, and solutions. This includes determining the **Drivers** of patterns and change for dryland structure and function; the types and magnitude of **Systems' Responses**; the **Social-Ecological Consequences** of system responses; and the options science-informed **Mitigation and Adaptation** options could play in maintaining drylands.

41  
42  
43 **ARID will blend its science goals with critical NASA Terrestrial Ecology mandates for**  
44 **community and capacity building using the following joined objectives:**

45  
46 **Advance** the quantitative and scalable understanding of drylands' contributions to the Earth  
47 system



1  
2 **Transform** knowledge and model representation of the mechanisms behind and the  
3 consequences of the changes now observed in drylands and those yet to come  
4

5 **Build** a diverse and collaborative international community focused on multidisciplinary dryland  
6 science  
7

8 **Collaborate** with decision makers to determine and evaluate mitigation and adaptation  
9 solutions that sustain and build resilience in drylands in the face of change  
10

11 **Train and support** the next generation of diverse dryland researchers through multidisciplinary  
12 education, training, and network building  
13

14 **Revolutionize** multi-scale, multi-sensor remote sensing of dryland processes and utilize the  
15 unique attributes of drylands to improve and inform current and future NASA missions and  
16 effective interpretation and analysis of space-based data  
17

18 **Develop** data products and tools to inform science-based decision making and create a fully  
19 harmonized, accessible database for the global community

## 20 A.4 Relevance to Solicitation and NASA

21 The timing of ARID is particularly beneficial for NASA fueled scientific discovery. The advent of  
22 higher-resolution satellites and emerging technologies, such as UAVs, coupled with existing  
23 satellites and aircraft, offers a high chance of gaining new knowledge about drylands through a  
24 NASA field campaign. ARID addresses all components of the A.4 scoping study solicitation, as  
25 well as NASA's Earth System Directorate science and applications mission, including those  
26 within the NASA Terrestrial Ecology Program. With respect to the 2017 NASA Earth Science  
27 Decadal Survey, ARID addresses numerous "important", "very important", and "most important"  
28 questions within the hydrology, climate, solid-earth, weather, and ecosystems science focal  
29 areas. The ARID framework integrates observations from NASA's current program of record  
30 (PoR), including decommissioned missions, (e.g., HYSPIRI, Aqua and Terra) and planned  
31 Decadal missions. Observations from ground instruments, NASA airborne payloads, and  
32 spaceborne missions are evaluated through SATM processes, and explicitly link the science  
33 questions with geophysical parameter requirements. We also align completely with NASA  
34 guidelines for recent initiatives such as the Transform to Open Science (TOPS) initiative and the  
35 Science Mission Directorate's SPD-41a directive.  
36

### 37 ***A.4.1 NASA Terrestrial Ecology Program and A.4 Solicitation Relevance***

38 The NASA Terrestrial Ecology Program's Goal is to: "*Improve understanding of the structure*  
39 *and function of global terrestrial ecosystems, their interactions with the atmosphere and*  
40 *hydrosphere, and their role in the cycling of the major biogeochemical elements and water.*"  
41 With this goal in mind, ARID is addressing major science themes of "vegetation structure and  
42 biodiversity" and "carbon cycle interannual variability" (see Section B.1). Since drylands  
43 intimately link water and carbon cycles through water-limitation, the major hydrosphere theme of  
44 "climate variability and drought" is also evaluated within ARID's campaign. These science

1 themes and research questions are ultimately interdisciplinary and cover topics in other NASA  
2 programs including those in Carbon Cycle and Ecosystems, Terrestrial Hydrology, as well as  
3 many programs across Applied Sciences. ARID is also highly relevant to the NASA SERVIR  
4 and LCLUC programs which aim to address climate change, food security, water and related  
5 disasters, land use, and air quality issues in predominantly dryland regions.  
6

7 The A.4 field campaign solicitation also strives to “*identify new ecosystems, biomes or regions*  
8 *that merit intensive investigation*”. While the global importance and vulnerability of drylands are  
9 evident (section A), limited research resources have been dedicated to the study of these  
10 ecosystems. *We therefore assert that drylands are indeed a “new ecosystem” when it comes to*  
11 *coordinated long-term ecological investigations by a NASA field campaign*. While there have  
12 been aircraft campaigns conducted in the drylands of western US (GEMx, SHIFT, WDTS), their  
13 scope was limited to generating a specific dataset or technologically motivated to develop or  
14 validate a NASA campaign. Due to the goals of those campaigns, there was not significant  
15 investment in understanding dryland ecosystem processes or collecting contemporaneous  
16 ecological field data. As a result, very few scientists or end-users within the ARID working  
17 groups have used these datasets (see Section B.2.5 and working group queries in Appendix  
18 F.10). Accordingly, ARID can be considered a new, large-scale effort to spearhead intensive  
19 investigations in direct response to the solicitation, and can powerfully leverage the data  
20 collected in past distributed campaigns. Given the importance and vulnerability of drylands, the  
21 dominance of drylands in the western US, and the overwhelming engagement response by the  
22 research and end-user communities, we argue ARID is in fact, long overdue.  
23

24 As encouraged by the A.4 solicitation, the ARID field campaign will bring together diverse  
25 expertise across fields of study, spatial scales of focus, geographical locations, contributor  
26 demographics, and career stages. A broad range of expertise was engaged throughout the  
27 scoping, as exemplified by the diversity of participants making up the ARID community. ARID’s  
28 engagement reached early, middle, and later career scientists and we were particularly  
29 enthused to see significant engagement with early career researchers. One of the powerful  
30 aspects of NASA’s Terrestrial Ecology field campaigns is their relatively long time span, which  
31 not only allows for deep and iterative research, but which means early career scientists may  
32 participate in the same campaign as a student, post-doc, and PI. With this in mind it is  
33 imperative that early career researcher ideas and perspectives be included in scoping and  
34 project design. More than half of individual inputs and working group participants (>80, >175  
35 people respectively) were at a career stage less than ten years beyond receiving their PhD  
36 (Appendix F.5, F.9).  
37

#### 38 **A.4.2 NASA Earth Science Relevance**

39 The ARID scoping study provides a research plan consistent with the mandate of NASA Earth  
40 Science Division’s Earth System Observatory to “create a 3D, holistic view of Earth, from  
41 bedrock to atmosphere.” ARID addresses key components of NASA’s Earth System Directorate  
42 science and applications mission. ARID will integrate observations from NASA’s current  
43 program of record (PoR), including decommissioned missions (e.g., HYSPIRI, Aqua and Terra),

1 and planned Decadal missions (e.g., Surface Biology Geology). ARID will support validation and  
2 science application using ongoing NASA satellite missions like SMAP, OCO-2, EMIT,  
3 ECOSTRESS, and GEDI, especially in some of the most complex ecosystems for these  
4 platforms to observe. It will also support the development and validation of upcoming NASA  
5 missions like SBG and NISAR and incubator missions like PBL and STV. Observations from  
6 ground instruments, NASA airborne payloads, and spaceborne missions are evaluated through  
7 Science and Application Traceability Matrix (SATM) processes, explicitly linking the ARID  
8 science questions with geophysical parameter requirements. ARID also follows guidelines of  
9 NASA's recent open data initiatives, including the Transform to Open Science (TOPS) initiative  
10 and the Science Mission Directorate's SPD-41a directive.

11  
12 ARID addresses all four of NASA's Earth Science Directorate questions. ARID specifically  
13 addresses understanding the role of drylands in how the Earth system is changing now and, in  
14 the future, and what causes the changes (see section A.2). ARID also addresses the Earth  
15 Action-related ESD question of how Earth system science can provide societal benefit,  
16 specifically, within the social ecological systems focus in section B.1 in understanding dryland  
17 ecosystem services and land management.

18  
19 With respect to the 2017 NASEM Earth Science Decadal Survey, "Thriving on our Changing  
20 Planet: A Decadal Strategy for Earth Observation from Space", which drives NASA ESD  
21 questions and goals, ARID addresses numerous "important", "very important", and "most  
22 important" questions within the hydrology, climate, solid-earth, weather, and ecosystems  
23 science focal areas. ARID is also being developed following key messages from the most recent  
24 United States National Climate Assessment (NCA5) and in coordination with the United States  
25 Global Change Strategic Plan (USGCRP) four pillars that guide how we respond to change and  
26 manage critical risks including: (1) Advancing Science, (2) Engaging the Nation, (3) Informing  
27 Decisions, and (4) Collaborating Internationally.

28

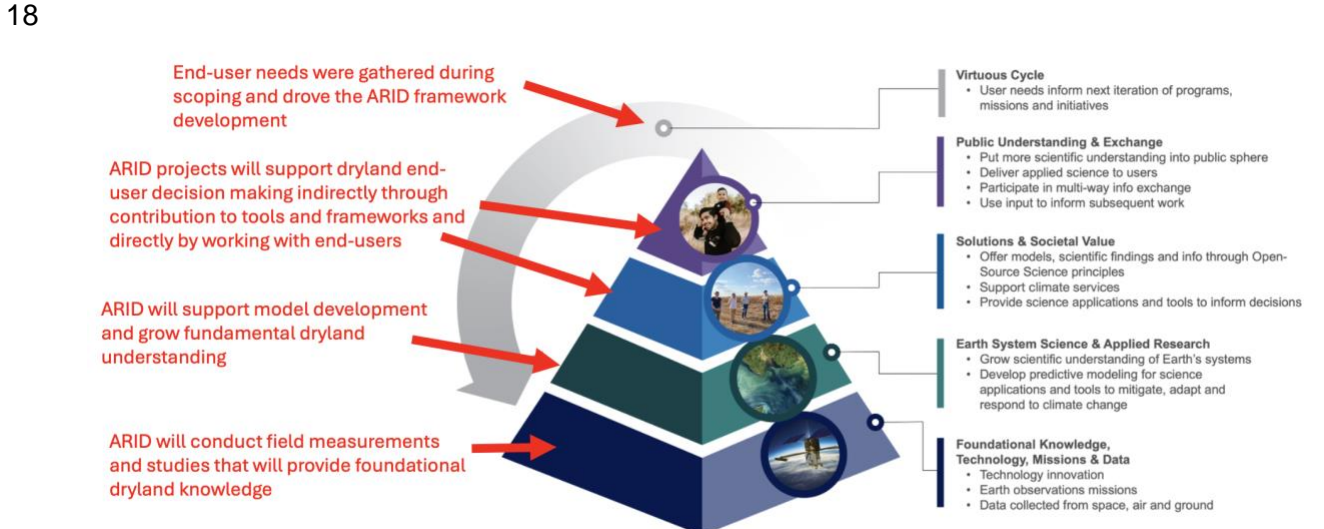
#### 29 ***A.4.3 Relevance to Earth Science to Action Strategy***

30 ARID is directly aligned with the 2024 launch of the NASA Earth Science to Action strategy  
31 (ES2A) by bringing together scientists with end-user communities including public and private  
32 end-users, tribal nations, industries with large land-bases, and representatives from government  
33 agencies in the implementation of a dedicated NASA dryland field campaign. In accordance with  
34 the field campaign plan, the ARID scoping phase demonstrated this scientist and end-user  
35 community engagement in several overarching ways:

- 36 1) Engagement with end-user communities guided our research questions and plan  
37 development with land management and conservation of ecosystem services in mind  
38 such that the data collected can be implemented at higher application readiness levels.
- 39 2) Both ARID workshops in Arizona and New Mexico included intensive discussions with  
40 end-user communities that manage and own land across the western US. The Arizona  
41 meeting (October 2023) included discussing with end-users across public and private  
42 domains in the southwest US. The New Mexico meeting (May 2024) focused on  
43 engagement with tribal communities that own and manage land in the southwest US and

- 1 Great Plains as well as hold traditional ecological knowledge (TEK) about managing  
 2 their land.
- 3 3) ARID’s inter-agency partnerships with USGS, BLM, and USDA are built upon using the  
 4 ARID field campaign data to support land management, especially throughout the  
 5 western US (see section B.2.7 for further details).
- 6 4) The ARID field campaign will improve the algorithm development and uncertainty  
 7 estimation of fundamental remote sensing products that feed directly into existing  
 8 workflows of agencies with established user-basis and decision-making frameworks.  
 9 These applications will facilitate adaptation and mitigation to changes in drylands.
- 10 5) The ARID team leadership included researchers that regularly work with end-users  
 11 including USDA ARS members (Joel Biederman, Russ Scott)

13 ARID also drives the ES2A strategy virtuous cycle by foremost providing foundational data that  
 14 can be used across a range of technical and societally beneficial uses (Fig. A.6). Additionally,  
 15 with the ES2A rollout in 2024, it is expected that the Terrestrial Ecology program can fund work  
 16 directly with end-users within ARID that can inform and directly impact aspects higher on the  
 17 pyramid (see Fig. A.6).



19 **Figure A.6.** The Earth Science to Action (ES2A) Virtuous Cycle obtained from  
 20 <https://science.nasa.gov/earth-science/earth-science-to-action/>. ARID informs  
 21 each level of the pyramid to varying degrees.

24 ARID has the potential to be a flagship program leading the ES2A ten-year strategy by being a  
 25 means to directly support projects that “discover and demonstrate innovative and practical uses  
 26 of NASA Earth Science observations and research through applied research and applications  
 27 projects carried out in partnership with end user organizations.” The ARID initiative will provide  
 28 ES2A with the opportunity to directly engage operational agencies as well as rural and tribal  
 29 communities, united around fundamental dryland science questions to address science-based  
 30 decision making. This will clearly demonstrate the societal value of NASA’s science to a large  
 31 population of diverse users and garner inter-agency support for ES2A.

# 1      **B. Technical Approach and Methodology**

## 2      **B.1 Science Themes and Questions**

3      The ARID mission is designed around four scientific themes (Fig. B.1) drawn from an extensive  
4      consultation process with scientists and end-users, a synthesis of the scientific literature, and  
5      assessment of technological challenges and opportunities.

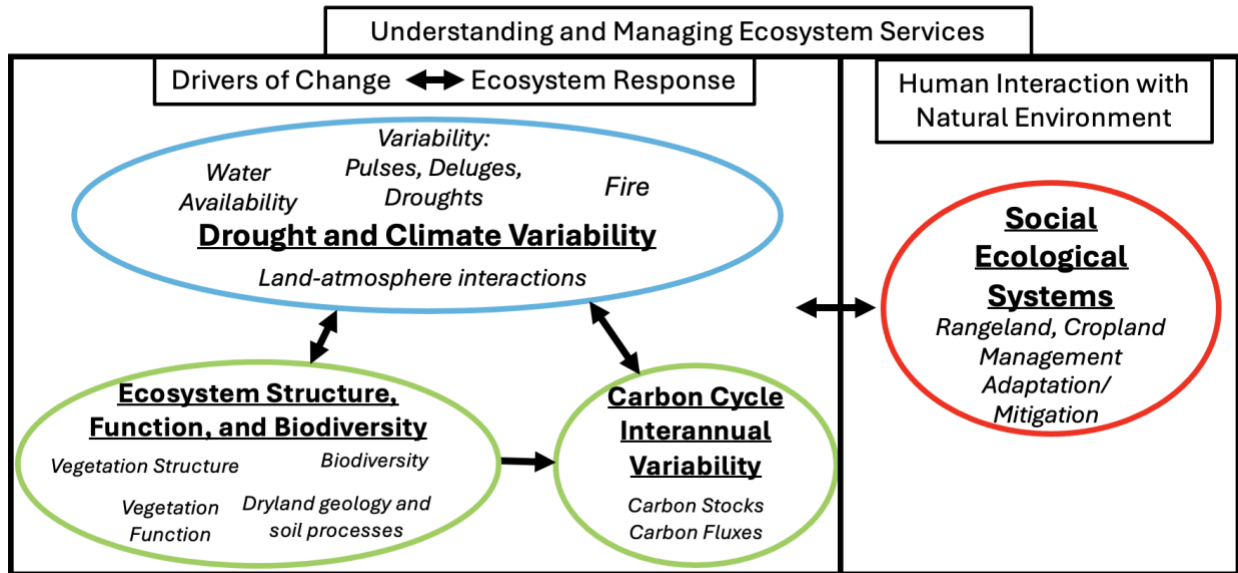
- 6          1. Science Theme 1: Climate Variability and Drought
- 7          2. Science Theme 2: Ecosystem Structure, Function, and Biodiversity
- 8          3. Science Theme 3: Carbon Cycle Interannual Variability and Long-Term Trends
- 9          4. Science Theme 4: Social-Ecological Systems

10  
11      Addressing these science themes contributes to understanding and managing the vast  
12      ecosystem services that drylands provide. These include, but are not limited to, water supply,  
13      carbon storage, crop/rangeland/livestock/forestry management, and biodiversity. Enabling  
14      research into dryland social-ecological systems connects the fundamental scientific discoveries  
15      from the first three science themes to the broader priorities in Earth Science to Action (Fig. B.1).  
16      Specifically, with a high population of humans living in dryland environments, social-ecological  
17      systems research evaluates the human interactions with the first three themes, including  
18      humans managing these ecosystems and adapting to them as they change. We envision  
19      research that not only represents cutting-edge discovery about drylands and their responses to  
20      change, but that creates usable information to support decisions and adaptations, and to  
21      promote resilience of drylands.

22      In this section we briefly outline the context of each science theme and sub-themes, along with  
23      overarching and specific science questions. We highlight that these science themes are  
24      interlinked, and each theme contributes to understanding the larger integrated system dynamics  
25      of dryland ecosystems (Fig. B.1). We briefly identify the approaches the ARID campaign would  
26      use to address these science themes and questions, noting that the associated ARID technical  
27      approaches are described in Section B.2.

28

# ARID Science Themes



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**Figure B.1.** Interactive components of ARID science themes. Linkage between the different ARID science themes as well as several main objectives of ARID. The first three science themes (green and blue ellipses) aim to understand the interaction between drivers of change and ecosystem responses. Dryland drivers are impacting biodiversity and ecosystem responses that feedback and further reinforce or suppress the climatic drivers, under land-atmosphere interactions. The fourth theme (red ellipse) aims to understand the interaction of humans with these three fundamental science themes. Altogether, they contribute to understanding dryland ecosystem services. The dryland biodiversity and ecological structure and function itself provides ecosystem services including water purification, forage and browse, grains, habitat, fuel wood, carbon storage, cultural settings, and many others. Research findings from a field campaign can quantify how ecosystem services are being impacted under global environmental changes, and work in partnership with land managers and other communities to co-develop usable information products to support decisions and adaptations in drylands.

19  
20  
21

**Table B.1.** Summary table of ARID research questions discussed throughout Section B.1. These include the four main theme questions and select questions from each of the sub-themes.

Research Theme	Theme Questions	
Theme 1: Climate Variability and Drought	<i>How are extremes like droughts, heatwaves, and large rain pulses across drylands impacting water availability, and are these extremes amplified by changing fire regimes, land cover change, and land-atmosphere interactions?</i>	
Sub-Theme	Central Sub-Theme Question	Select Process-level Questions

<p><b>1.1 Water Availability</b></p>	<p>How do changes in the amount, timing, intensity and phase of water inputs affect surface water partitioning among evaporation, transpiration, runoff, and groundwater recharge, thereby regulating the amount of water that is available to humans and ecosystems?</p>	<p>How do changes in atmospheric conditions, such as CO<sub>2</sub>, VPD, and energy balance, alter the fate and temporal variability of <b>soil moisture</b>?</p> <p>How can remote sensing detect, and Earth System Models predict, key physical parameters (e.g., precipitation, evapotranspiration, soil moisture, plant water status and runoff) to address these questions?</p>
<p><b>1.2 Dryland Climate Variability: Pulses and Droughts</b></p>	<p>How do dryland ecosystems process heterogeneous, highly dynamic moisture pulses?</p> <p>How does the timing and duration of lack of rainfall impact dryland function, structure, composition, and water availability and how is drought intensity, severity, and duration changing in drylands?</p>	<p>How are moisture pulses changing?</p> <p>How can Earth observation adequately capture pulse dynamics at the appropriate diurnal and daily-scale frequency?</p> <p>Do process models capture rain/non-rainfall pulses and ecosystem responses?</p> <p>How do higher temperatures and heatwaves influence drought onset and duration in drylands?</p> <p>Which dryland ecosystem components respond most to drought?</p>
<p><b>1.3 Fire</b></p>	<p>How do fire regimes on both rangeland, open woodland, and forested landscapes change the composition, structure, and function of drylands at various time scales?</p>	<p>How do changes in dryland composition, structure, and function feedback to impact fire regimes?</p> <p>What role does fire play in the expansion of invasive species, shrub encroachment, and the loss of ecosystem services as observed across multiple continents?</p> <p>How does fire impact carbon stocks and fluxes across diverse dryland ecosystem types and across spatio-temporal scales?</p>
<p><b>1.4 Land-atmosphere Interactions</b></p>	<p>How do land-atmosphere interactions influence climate extremes, water availability, and dryland ecosystem responses, including changes in air temperature, changes in the frequency and intensity of extreme events, as well as land use change?</p>	<p>To what degree are land-atmosphere interactions driving drylands' contribution to water, carbon, and energy fluxes?</p> <p>How much are climate extremes playing a role in these interactions?</p> <p>How does dryland heterogeneity of soil conditions (e.g., soil moisture, soil texture) and vegetation (structure and types) influence landscape-scale land-atmosphere interactions (like convection)?</p>
<p><b>Theme 2: Ecosystem Structure, Function, and Biodiversity</b></p>	<p><b><i>What are the main mechanisms driving the spatiotemporal distributions of dryland structure, function, and biodiversity?</i></b></p>	
<p><b>Sub-Theme</b></p>	<p><b>Central Sub-Theme Question</b></p>	<p><b>Select Process-level Questions</b></p>

<p><b>2.1 Vegetation Structure and Heterogeneity</b></p>	<p>How are ongoing changes in climate (rising CO<sub>2</sub>, increases in drought frequency and intensity, long-term changes in mean rainfall and changing fire regimes) and land use (changing fire regimes grazing and other land uses) impacting vegetation structure, function and habitat in drylands?</p>	<p>What are the rates and underlying causes of woody plant encroachment (WPE) in global drylands?</p> <p>How do invasions by exotic shrubs and grasses observed in drylands around the world impact ecosystem function and disturbance regimes?</p> <p>How are changes in structure and function impacting essential ecosystem services, including changes in water, carbon and nutrient cycling in drylands and provision of forage for livestock?</p>
<p><b>2.2 Dryland Biodiversity</b></p>	<p>What are the drivers of biodiversity (functional, phylogenetic, taxonomic) in drylands and how will these be changing?</p>	<p>What is the relationship between biodiversity, ecosystem function, carbon stocks and resilience to disturbances (drought and fire)?</p> <p>What are the impacts of climate change (increased aridity) and continued grazing pressure on biodiversity (functional, phylogenetic, taxonomic, etc.)?</p> <p>What is the impact of land cover transformation, land degradation, and cultivation on dryland biodiversity and species of special interest (e.g. endangered species)?</p> <p>What are the impacts of invasive plant species on biodiversity, rangeland condition and water-availability?</p>
<p><b>2.3 Ecosystem Function</b></p>	<p>Across different timescales, what are the dominant mechanisms driving dryland function such as plant hydraulics, leaf level photosynthesis, respiration, and nutrient cycling?</p>	<p>What role has CO<sub>2</sub> fertilization played in driving changes in the SPAC and dryland functions GPP, NEE, ET, and WUE?</p> <p>What role has the timing and intensity of precipitation and temperature-driven increases in VPD played in driving changes in the SPAC and dryland functions GPP, NEE, ET, and WUE?</p> <p>Can we utilize remote sensing data to improve the estimation of dryland vegetation physiological status (e.g. photosynthetic quantum yield, nutrient, pigment and enzyme concentrations, canopy stomatal and hydraulic properties, vegetation and soil water status, drought stress) for distinct PFT and biocrust communities?</p> <p>Can we improve retrievals of vegetation function (e.g. carbon, water, and energy fluxes) in drylands, either directly (e.g. thermal estimation of energy balance), or via use of improved structural, physiological, and phenological retrievals in process-based models?</p> <p>Can improvements of decomposition of soil organic matter be improved by night-time monitoring of carbon exchanges and soil moisture levels?</p>



<p><b>2.4 Dryland Geology and Soils</b></p>	<p>How do soil communities and physicochemical characteristics drive and respond to climate variability and ecosystem change?</p>	<p>What is the extent, composition, and function of dryland biological soil crusts (biocrusts)? What roles do soils play in dictating aboveground structure, function, and response to change?</p> <p>How much inorganic carbon is stored in dryland soils across different soil types, aridity index gradients, and natural vs. managed systems, and how vulnerable is the carbon to environmental shifts?</p> <p>What contribution do soil signals play in space-based observations of all terrestrial ecosystems?</p> <p>What are the sources, sinks, causes, and consequences of accelerated topsoil loss and dust in Earth's drylands?</p>
<p><b>Theme 3: Carbon Cycle Interannual Variability and Long-Term Trends</b></p> <p><i>What is the contribution of drylands to the mean, trend, and particularly the interannual variability of terrestrial carbon dynamics?</i></p>		
<p><b>Sub-Theme</b></p>	<p><b>Central Sub-Theme Question</b></p>	<p><b>Select Process-level Questions</b></p>
<p><b>3.1 Carbon Stocks and Fluxes</b></p>	<p>How large are the carbon stocks and fluxes in drylands, how do they vary at sub-annual to decadal timescales, and what is their response through space and time, to drivers of global change?</p>	<p>How vulnerable are dryland carbon stocks and fluxes to global and regional changes in water availability and atmospheric demand, combined with asynchronous plant responses to periods of water stress (B.1.2.1)?</p> <p>How do woody encroachment and desertification change dryland carbon cycling and feedbacks?</p> <p>What is the relative contribution of different plant functional types to carbon stocks and how are these changing along environmental gradients?</p> <p>What influence do rooting strategies and belowground carbon allocation play in ecosystem carbon storage?</p> <p>What is the potential of carbon capture in dryland vegetation and soils to work as a nature-based climate solution (NCS)?</p> <p>How can remote and in-situ observations be best used to model and quantify carbon stocks and fluxes?</p>
<p><b>Theme 4: Social Ecological Systems</b></p> <p><i>What are the consequences of changes in drylands for social-ecological and what management (e.g., mitigation and adaptation) solutions can maintain the critical services provided by drylands even in the face of change?</i></p>		
<p><b>Sub-Theme</b></p>	<p><b>Central Sub-Theme Question</b></p>	<p><b>Select Process-level Questions</b></p>
<p><b>4.1 Land Management</b></p>	<p>How are land and water resources and resource management being affected by drought and aridity changes?</p>	<p>How is the composition and productivity of rangelands changing under various management regimes and how are they predicted to change due to climate?</p> <p>How will remote sensing and modeling enhance assessment and forecasting of land resources (e.g.,</p>

		<p>soil moisture, productivity, water use, carbon storage) for various land uses?</p> <p>How will land use further impair wildlife corridors and migratory pathways?</p> <p>How will further fragmentation of conservation and rangeland areas affect biodiversity and conservation goals?</p> <p>How will reduced water availability impact forage production, cropping yields, livestock conditions, and habitat conditions?</p> <p>How will increased thermal stress affect agricultural production of rangelands, cropland, forests, livestock, wildlife, and wildlands systems?</p> <p>How will renewable energy deployment affect ecosystem services?</p> <p>Can multi-sensor observations and modeling improve our ability to assess land management effectiveness?</p>
<p><b>4.2 Adaptation and Mitigation Strategies</b></p>	<p><b><i>Adaptation and resilience of dryland social ecological systems</i></b>  How can dryland ecosystems enhance their resilience in the face of environmental stressors, and what adaptive strategies can local communities implement to build resilience?</p> <p><b><i>Mitigation options for different dryland livelihood strategies</i></b>  How will climate variability and water limitations hinder carbon sequestration efforts?</p>	<p><b><i>Adaptation and resilience of dryland social ecological systems</i></b>  What factors and information can enhance adaptive capacity of dryland social-ecological systems to meet the challenges to increased aridity?</p> <p>How can ecological forecasting of drought conditions enable decision making of dryland operators and natural resource managers?  Have land managers needed to change their strategies in drylands in the past few decades to accommodate change? What drives these changes?</p> <p><b><i>Mitigation options for different dryland livelihood strategies</i></b>  What impact will renewable energy development and carbon sequestration practices have on dryland livelihood strategies and ecosystem services in different dryland regions?</p> <p>How can improved observations enhance our assessment of renewable energy systems and carbon sequestration practices?  What increased information can be provided related to changes in ecosystem processes and ecosystem services related to mitigation practices?</p>

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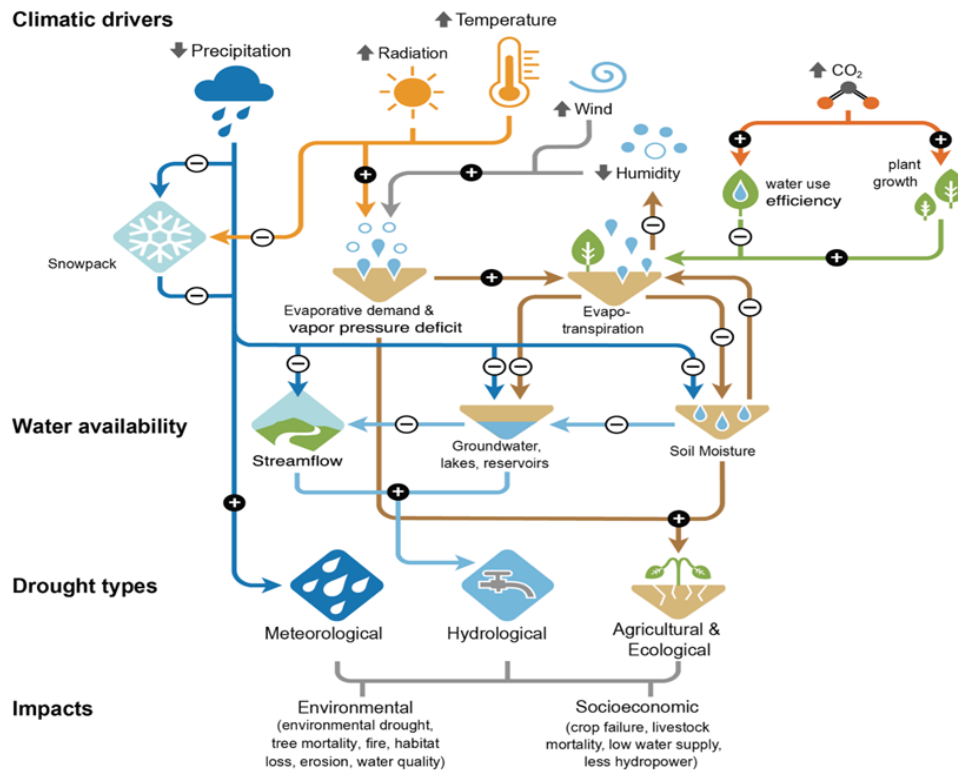
1 **B.1.1 Science Theme 1: Climate Variability and Drought**

2

3 A dominant climate feature across dryland regions of the world is the low precipitation levels  
4 and high evaporative demand. These climate drivers are leading to reduced water availability  
5 and increased drought impacts resulting in economic losses, altered levels of ecosystem  
6 services, damage to natural resources, and reduced capacity to maintain livelihoods. Dryland  
7 regions are facing increases in climate variability that further exacerbate conditions, including  
8 more frequent droughts and heatwaves, pulses of extreme rainfall events, increased frequency  
9 and intensity of fire, and sustained, stronger extremes through land-atmosphere interactions.

10 Droughts in dryland systems are not uncommon, however, recent increases in climate variability  
11 and exposure to extreme events are affecting local hydrological systems to further exacerbate  
12 drought conditions and severity of droughts. Droughts are characterized by meteorological  
13 droughts (e.g., when dry weather conditions persist) and hydrological droughts (e.g., when  
14 water availability is low) (Fig. B.2). In dryland systems these two conditions often coexist due to  
15 climate and human related causes. Future scenarios across the region suggest that elevated  
16 seasonal temperatures and increases in the atmospheric evaporative demand will exacerbate  
17 drought effects, through increased evapotranspiration, for example, despite the level of changes  
18 in precipitation taking place.

19



20

21 **Figure B.2.** Conceptual diagram describing climate, land use, and atmospheric  
22 CO<sub>2</sub> effects on water availability and resulting drought conditions, including water  
23 consumption under various land uses such as irrigation for cropping systems.  
24 Considerations of diurnal pattern of dew formation also needs consideration in  
25 these dryland landscapes (from IPCC WG1 Chapter 8, 2021)

1  
2 The consequences of these climate driven changes, human-caused changes in water allocation  
3 and usage for cropping and other consumptive activities, such as mining, has reduced the  
4 natural flows of water in many dryland systems. Extraction of groundwater supplies to meet  
5 demands of local communities for drinking water supply and irrigation as well as reservoir usage  
6 for power generation and irrigation have impacted water availability across drylands. In addition,  
7 other extreme events are associated with increasing climate variability such as dust storms,  
8 flooding following drought conditions contributing to landscape deformation through erosion or  
9 gully formation, and enhanced fire danger as fuel loads are persisting in flammable stages for  
10 longer periods of time.

11 Our overall question we ask in this theme is: *How are climate extremes like droughts,*  
12 *heatwaves, and large rain pulses impacting dryland systems and how do they interact with*  
13 *changing fire regimes, land cover change, and land-atmosphere interactions??* This major  
14 science theme, drought and climate variability, is further defined in four sub-themes: 1) water  
15 availability, 2) pulses and extreme events, 3) fires, and 4) land-atmosphere interactions in  
16 dryland systems.

17

#### 18 *B.1.1.1 Sub-Theme: Water Availability*

##### 19 *Motivation*

20 Water is the key limiting factor for ecosystems and societies in drylands. Quantifying how water  
21 availability is changing due to climate change and through land use activities is critical to  
22 understanding and predicting how dryland ecosystem structure and function will be altered (e.g.,  
23 structure, composition, vegetation productivity, water partitioning, carbon cycling) in the future  
24 (Fig. B.2). This understanding is essential to inform decision makers of how to adapt  
25 management practices to conserve and enhance dryland resources.

26 Dryland water availability is controlled by the amount, timing, and fate of water inputs (frozen or  
27 liquid and mainly through precipitation) at the land surface (Fig. B.2). Precipitation may  
28 evaporate from wet leaves and surfaces, run off laterally, or infiltrate into soil locally, recharging  
29 soil moisture. Soil moisture is subsequently partitioned among abiotic evaporation, root uptake,  
30 and subsequent transpiration by plants. Typically, recharge to groundwater is small in drylands  
31 except in and along mountain fronts, where precipitation is higher and runoff is concentrated  
32 (Scanlon et al. 2006).

33 Snowmelt is an important moisture source in many drylands at higher latitudes and/or higher  
34 elevations. In some drylands, especially those near large water bodies, fog and dew can  
35 contribute important fractions of the water budget (L. Wang et al., 2017). Soil water availability is  
36 often spatially heterogeneous in drylands, due to soil/geologic variability, topography, and  
37 variable, infrequent, patchy precipitation. Water availability at the land surface is further modified  
38 by ecosystem structure, biodiversity, and functioning that controls the land surface energy  
39 balance and regulates evapotranspiration (ET). Thus, changes in dryland land cover (e.g.,  
40 woody plant encroachment) and land use (e.g., urbanization, grazing) could have dramatic  
41 effects on precipitation partitioning and change the amount of water that is available for  
42 ecosystems and human needs.

1 Climate change is affecting water availability, with general predictions that dry locations are  
2 becoming drier from warming temperatures and higher evaporative demand (Held and Soden  
3 2006). The western US is also drying with trends toward reducing mean rainfall amounts (F.  
4 Zhang et al., 2021). While many ecosystems have higher atmospheric humidity due to  
5 increased ET from higher evaporative demand, the atmosphere above drylands is not becoming  
6 more humid (Simpson et al., 2024a).

#### 7 *Questions*

8 In seeking to quantify water availability as a key driver of change in drylands, we identify several  
9 key questions to be addressed by ARID. First, how do changes in the amount, timing, intensity  
10 and phase of water inputs affect surface water partitioning among evaporation, transpiration,  
11 runoff, and groundwater recharge, thereby regulating the amount of water that is available to  
12 humans and ecosystems? Second, how do changes in atmospheric conditions, such as CO<sub>2</sub>,  
13 VPD, and energy balance, alter the fate and temporal variability of **soil moisture**? Finally, how  
14 can remote sensing detect, and Earth System Models predict, key physical parameters (e.g.,  
15 precipitation, evapotranspiration, soil moisture, plant water status and runoff) to address these  
16 questions?

17

#### 18 *B.1.1.2 Sub-Theme: Dryland Climate Variability: Pulses and Droughts*

19 Two controls particularly powerful in regulating dryland structure and function are rainfall pulse  
20 dynamics and dry spells or drought between pulses (large changes in dryland resource  
21 availability and process rates across short timescales). Many dryland regions are currently  
22 experiencing more rainfall variability and extreme heat waves. Nominally, this includes longer  
23 dry spells and more intense rainfall (Pendergrass et al. 2017), as well as “hot droughts”, where  
24 drought conditions co-occur with warmer temperatures. The precipitation pulse dynamics of  
25 drylands are poorly understood, particularly how the rain pulse inputs influence the soil-  
26 vegetation system at hourly-to-daily timescales and consequently the water, carbon, energy  
27 cycles and erosion rates. With most terrestrial biosphere models developed to understand  
28 forests of more humid regions, they may not include dynamics at these daily timescales and  
29 how their impacts accumulate to influence annual flux estimates (Feldman, Feng, et al., 2024a).  
30 This increase in rainfall variability also extends to extreme events, with more frequent and  
31 intense droughts, heatwaves, dust storms, and severe weather events. We outline pulse  
32 dynamics and drought dynamics separately within this subtheme.

33

#### 34 *Pulses of moisture and ecosystem responses*

##### 35 *Motivation*

36 Pulsed events, both nominal and extreme, have major consequences on dryland water  
37 availability, landscape structure, and ecosystem dynamics. Pulse dynamics refers to the  
38 tendency of dryland biological activity to be controlled by relatively infrequent moisture pulses,  
39 and while it has long been known drylands can respond quickly to precipitation, the mechanisms  
40 behind pulsed controls and the expectations for their change with climate change, remain highly  
41 uncertain. Dryland moisture inputs may include rainfall, snowmelt, and in some cases,  
42 meaningful amounts of fog, dew, or water vapor adsorption. Pulse dynamics are a key feature of

1 dryland ecosystems. Because of the strong moisture limitation in these systems, periodic  
2 moisture inputs result in rapid changes in ecological activity which are dynamic over time scales  
3 of hours to days. Discrete moisture pulses temporarily alleviate the effects of water stress on  
4 plant and microbial activities, allowing higher rates of photosynthesis, evapotranspiration, and  
5 ecosystem respiration (Huxman et al., 2004; Noy-Meir, 1973). High-intensity rainfall events can  
6 also trigger flooding and/or erosion. The frequency and intensity of precipitation pulses is  
7 changing (Demaria et al., 2019; F. Zhang et al., 2021) and the impact of this change is highly  
8 uncertain.

### 9 *Questions*

10 The main question we seek to understand is: How do dryland ecosystems process  
11 heterogeneous, highly dynamic moisture pulses? Some more specific questions to consider  
12 are: How are moisture pulses changing? How can Earth observation adequately capture pulse  
13 dynamics at the appropriate diurnal and daily-scale frequency? What are the key determinants  
14 of the functional form of the pulse response (diurnal and daily-scale)? Do process models  
15 capture rain/non-rainfall pulses and ecosystem responses?

16

### 17 *Drought*

#### 18 *Motivation*

19 In contrast to pulses, droughts may occur as discrete short-term events or persist for years to  
20 decades or longer. Droughts may be characterized as meteorological (e.g., when dry weather  
21 conditions persist), hydrological (e.g., when surface water systems are relatively dry) or  
22 ecological (when plant and soil water availability are low) (Figure B.2). In dryland systems, these  
23 three conditions often coexist due to climate and human related causes. In addition to types of  
24 droughts, drought can manifest over a variety of temporal scales including flash droughts at  
25 weekly scales to decadal scale deficits. Future scenarios across the western US drylands  
26 suggest that increasing mean aridity conditions, like increasing temperature (and associated  
27 evaporative demand) and declining winter rainfall totals (e.g.(F. Zhang et al., 2021)), will lead to  
28 greater drought effects. Droughts are common in dryland systems. However, with recent climate  
29 change, the frequency and intensity of droughts are predicted to increase in most drylands (Dai,  
30 2011). Climate and land use change are also affecting local hydrological systems to further  
31 exacerbate drought severity and duration. The recent severe drought in the western US is  
32 unprecedented for thousand-year timescales and has exposed the vulnerability of agricultural  
33 livelihoods with significant socio-economic impacts. These drought effects substantially impact  
34 dryland productivity and biogeochemical processes. However, the ecosystem responses to  
35 drought in drylands are poorly understood and poorly modeled in terrestrial biosphere models  
36 (De Kauwe et al. 2015).

37

### 38 *Questions*

39 The main question we seek to understand is: how does the timing and duration of lack of rainfall  
40 impact dryland function, structure, composition, and water availability and how is drought  
41 intensity, severity, and duration changing in drylands? More specific questions to consider  
42 include: How do higher temperatures and heat waves influence drought onset and duration in

1 drylands? Which dryland ecosystem components respond most to drought? How can modeling  
2 of ecosystem responses to drought be improved? How does remote sensing help to monitor the  
3 drought onset, development, and drought impacts? How can we project changes in ecosystem  
4 structure with increased persistence of drought conditions?

5

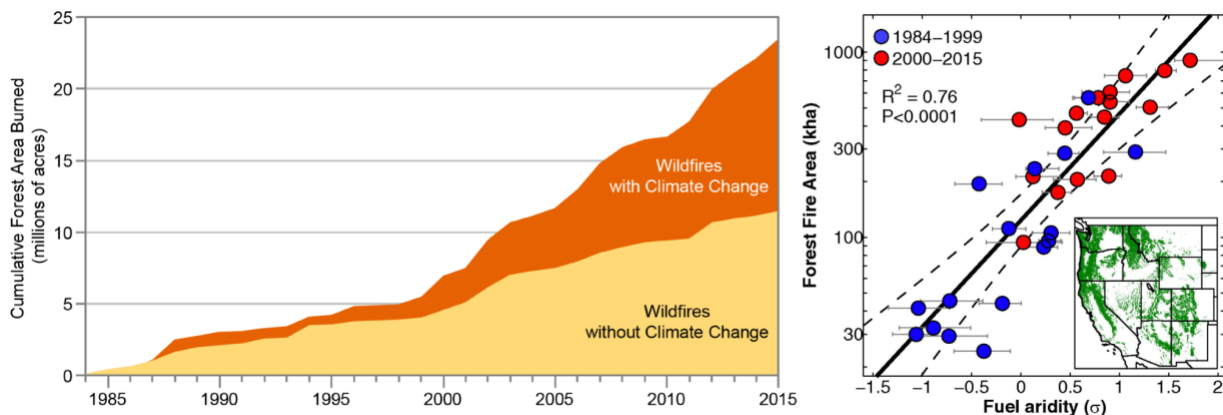
### 6 *B.1.1.3 Sub-Theme: Fire*

#### 7 *Motivation*

8 Climate change - specifically seasonal and interannual warming temperatures, increased  
9 evaporative demand, and declining water availability - is driving unprecedented increases in  
10 wildfire across dryland ecosystems globally, including increases in fire intensity and frequency  
11 (Senande-Rivera et al., 2022). While fire is a natural part of dryland ecosystems, persistent and  
12 repeated events and/or novel vegetation-fire feedbacks can cause threshold responses and  
13 ecosystem state changes as tipping points are crossed (Hoover et al., 2020). For instance, in  
14 rangeland and woodland ecosystems, invasive annual grasses are transforming natural fire  
15 regimes and displacing native grass and shrub species leading to large reductions in forage  
16 production and quality (Balch et al., 2013; Fusco et al., 2019; J. T. Smith et al., 2022; J. T.  
17 Smith, Allred, Boyd, Davies, Kleinhesselink, et al., 2023). Conversely, fire management that  
18 traditionally aimed to prevent or extinguish natural fire regimes has had unintended  
19 consequences, such as rapid shrub encroachment of large areas of natural grasslands with  
20 associated reductions in forage yields (Briggs et al., 2002; Keeley, 2006), as well as the  
21 unnatural accumulation of surface fuels (Barbero et al., 2015; Littell et al., 2009; Mueller et al.,  
22 2020).

23 Rangelands including grassland, shrubland, and open woodland ecosystems, experience the  
24 largest extent of burned area in the western US and globally (Crist, 2023). Rangeland wildfire is  
25 increasing with climate change due to the aridification of these ecosystems and changes in  
26 community composition (Fig. B.3; Balch et al., 2013). The most at-risk areas in the western US,  
27 such as low resistance and resilience sagebrush ecosystems, are quickly being converted to  
28 invasive annual grass-dominated ecosystems that further elevate fire risk (J. T. Smith et al.,  
29 2022).

30



31

32

33

**Figure B.3.** (a) The cumulative forest area burned by wildfires has greatly increased between 1984 and 2015, with analyses estimating that the area

1 *burned by wildfire across the western continental US over that period was twice*  
2 *what would have burned had climate change not occurred (Garfin et al., 2018,*  
3 *4th National Climate Assessment, Southwest Chapter adapted from Abatzoglou*  
4 *& Williams, 2016). (b) A strong correlation has been observed to fuel aridity since*  
5 *2000 and seems to be a key driver related to the area burned (Abatzoglou &*  
6 *Williams, 2016a).*

7  
8 Fire management on woodland and forested ecosystems has heavily targeted vulnerable dry  
9 mixed conifer forest that are fully within the ARID domain (Fig. B.3). Fire exclusion management  
10 on forest ecosystems was largely successful for much of the 20th century, but the confluence of  
11 accumulated surface fuels, unnaturally dense forests, and climate change has rapidly increased  
12 wildfire frequency and severity (Fig. B.3) (Abatzoglou & Williams, 2016b; Barbero et al., 2015;  
13 Garfin et al., 2018; Mueller et al., 2020). Landscape-scale, stand-replacing fires have the  
14 potential to push western US woodland and forest ecosystems beyond tipping points and  
15 thereby cause lasting changes in vegetation structure, composition, and succession (Batllori et  
16 al., 2020). Recognizing this threat, the forest management paradigm rapidly shifted to  
17 widespread deployment of forest restoration treatments, including mechanical thinning and  
18 managed or prescribed fire, aimed at reducing canopy cover and fuel loads to reduce wildfire  
19 frequency and severity (Fulé et al., 2012). Through ARID, we can further integrate multi-scale  
20 remote sensing observations, and coordinate with land managers to inform land management  
21 strategies towards increased forest resilience to wildfire without unintended negative  
22 consequences.

23 We have the opportunity to explore cutting-edge rangeland and wildfire management strategies  
24 that leverage novel multi-scale remote sensing technologies. For instance, with very high  
25 spatiotemporal resolution optical data (e.g., daily, 3-meter data from PlanetScope) and virtually  
26 fencing (Wätzold et al., 2024), we can determine where targeted grazing opportunities exist as a  
27 potential tool for managing fine fuels on rangeland ecosystems. Novel science-informed  
28 management approaches can be pioneered on western US rangelands and then knowledge can  
29 be tested on transferred to global rangeland socio-ecological systems through ARID  
30 international partnerships.

### 31 32 *Questions*

33 In seeking to understand how dryland fire regimes respond to climate change and in-turn drives  
34 change in dryland community composition, structure, and function, we have identified key  
35 overarching questions to be addressed by the ARID terrestrial campaign: How do fire regimes  
36 on both rangeland, open woodland, and forested landscapes change the composition, structure,  
37 and function of drylands at various time scales?; and how do these changes in dryland  
38 composition, structure, and function feedback to impact fire regimes?

39 We have further identified key sub-questions with direct management implications that could  
40 also be targeted as part of the ARID campaign that include: What role does fire play in the  
41 expansion of invasive species, shrub encroachment, and the loss of ecosystem services as  
42 observed across multiple continents? How does fire impact carbon stocks and fluxes across  
43 diverse dryland ecosystem types and across spatio-temporal scales? To what extent can  
44 grazing be used as a management tool to reduce fire frequency and intensity in drylands? How



1 is a changing snowpack and a shift from snow to rain in winter impacting fire dynamics across  
2 diverse dryland ecosystem types? How can management (e.g., forest thinning, prescribed fire,  
3 invasive removals, grazing) be deployed to reduce fuels and increase dryland ecosystem  
4 resilience to fire? What management options (e.g., aerially seeding for restoration) could help  
5 ecosystems recover from fire when it does occur?

#### 6 *B.1.1.4 Sub-Theme: Land-Atmosphere Interactions*

##### 7 *Motivation*

8 Land-atmosphere interactions strongly influence dryland carbon, water, energy, and other  
9 biogeochemical fluxes, and are shifting with changes in large-scale atmospheric circulation  
10 under climate change (Zhou et al., 2021). These interactions also contribute to rainfall initiation  
11 in drylands as well as intensification of droughts and heatwaves. For example, soil moisture  
12 spatial heterogeneity, the amount of soil moisture, and the ET rates from different vegetation  
13 types can contribute to rainfall initiation in drylands (De Kauwe et al., 2013; Green et al., 2017;  
14 Koster et al., 2004; Taylor et al., 2011). Similarly, low soil moisture availability can reduce  
15 terrestrial carbon uptake and prolong drought conditions (Dannenberg et al., 2022; Williams et  
16 al., 2022). It is essential to understand this feedback between the land surface and atmosphere  
17 to accurately predict climate and weather patterns as well as the associated impacts on  
18 precipitation, soil moisture, ET, runoff, dust storms, and vegetation productivity and dynamics  
19 (e.g., stress response, phenology, vegetation mortality).

20

##### 21 *Questions*

22 We seek to address: How do land-atmosphere interactions influence climate extremes, water  
23 availability, and dryland ecosystem responses, including changes in air temperature, changes in  
24 the frequency and intensity of extreme events, as well as land use change? More specific  
25 questions to consider include: To what degree are land-atmosphere interactions driving  
26 drylands' contribution to water, carbon, and energy fluxes? How much are climate extremes  
27 playing a role in these interactions? How does dryland heterogeneity of soil conditions (e.g., soil  
28 moisture, soil texture) and vegetation (structure and types) influence landscape-scale land-  
29 atmosphere interactions (like convection)? Do land-atmosphere interactions amplify the effects  
30 of extreme weather in drylands? Do changed dryland land uses and rangeland management  
31 influence land-atmosphere interactions, and if so, at which scales? How does shrub  
32 encroachment, vegetation composition change, or reduced vegetation cover influence land-  
33 atmosphere interactions?

#### 34 *B1.1.5 Approaches to Studying Drought and Climate Variability*

35 To address these scientific questions, challenges quantifying and modeling the key physical  
36 parameters of the water cycle must be overcome. Key parameters include precipitation, surface  
37 temperature, topography, relative humidity, evaporative demand, evapotranspiration, vegetation  
38 structure and status, vegetation water status, fractional cover, wind speed, water use efficiency,  
39 surface roughness, groundwater, and soil moisture (W. K. Smith et al., 2019b). Additionally,  
40 non-rainfall inputs such as fog and dew are challenging to quantify. For soil moisture, only  
41 shallow (5 cm and potentially deeper), coarse resolution (~ 10's km) measurements are  
42 detectable by current satellites. In addition, geostationary data can provide greater temporal

1 resolution to estimate soil moisture, surface temperature, and water availability in dryland  
2 systems.

3 Meanwhile, high spatial resolution (<0.1 km) shallow soil moisture from satellite synthetic  
4 aperture radar (SAR) is available, but it is typically measured too infrequently (~weekly) to  
5 detect important features of rainfall pulses and drying dynamics, which often dominate the  
6 function of dryland ecosystems. Recent advances in estimating evaporation and transpiration  
7 with remote sensing observations and various modeling techniques provide useful starting  
8 points to evaluate dryland ET rates (Awada et al., 2022; Pan et al., 2020; Senay et al., 2017; K.  
9 Zhang et al., 2016). In addition, partitioning evapotranspiration into its evaporation and  
10 transpiration components can be difficult due to sparse vegetation, and thus mixed pixels of  
11 bare soil and vegetation need to be partitioned with the aid of very high resolution UAS and  
12 airborne data from the ARID campaign.

13 Groundwater, which greatly improves the productivity and diversity of both ecosystems and  
14 human activities, monitoring gravity storage changes from orbiting satellites has led to  
15 innovative global and large regional-scale understanding (Scanlon et al. 2021), though these  
16 analyses are often too large in scale to be useful to water resource managers. Estimates of  
17 groundwater can be improved with data fusion techniques to produce finer spatial resolution  
18 products as well as provide measurement continuity, which are critical to determine future  
19 groundwater availability in drylands. For example, the next generation GRACE satellite  
20 combined with well data via models and data assimilation shows great promise in improving the  
21 spatial and temporal resolutions of these measurements, enhancing their utility for decision-  
22 making. Additionally, supersites and some soil moisture networks (USCRN, SCAN) include soil  
23 moisture measurements at intervals between 5 cm and 100 cm and sometimes deeper, which  
24 can provide an understanding of the role of some of the deeper moisture sources have on  
25 ecosystem function.

26 New approaches are needed to better understand how changing water availability alters water  
27 movement and use at 1) throughout the soil-plant-atmosphere and how this translates to 2)  
28 structural change of the individuals and ecosystems, 3) ecosystem composition across the  
29 landscape, and 4) landscape to regional scale carbon, water, energy exchange. To address  
30 precipitation-related challenges, ARID will need to develop and test new high-temporal and  
31 spatial precipitation estimates by combining ground radar and gauges, aircraft measurements,  
32 existing satellites, algorithms, and modeling techniques.

33 To better measure plant available water related to soil moisture and soil water potential, new  
34 techniques based on remote sensing of water along the atmosphere-plant-soil continuum can  
35 be developed using microwave observations, which can be validated using GNSS sensors  
36 measuring VOD frequently at the site level (Feldman, 2024a). Such plant water status  
37 information is especially useful for integration with plant hydraulic schemes in terrestrial  
38 biosphere models. Information from models and remote sensing products, such as soil moisture  
39 and rainfall distribution, can be merged to provide finer spatial resolution measurements deeper  
40 into the soil of soil moisture, especially when combining high temporal frequency passive and  
41 high spatial resolution active microwave sensors. When combining this information with soil  
42 characteristics, it can be used to estimate soil water potential. For improved estimates of ET,  
43 new soil moisture products and higher resolution (<100m) thermal data from aircraft and  
44 spaceborne instruments can be integrated into ET models as a constraint, with improved ET

1 partitioning schemes. These can be further validated using the network of flux tower sites  
2 dispersed across dryland ecosystems.

3 To capture the pulse dynamics of extreme precipitation events or infrequent flooding events,  
4 opportunistic observational strategy is needed to follow these events and to monitor impacts of  
5 these events on landscape deformation and vegetation structures. Being able to monitor long  
6 term ecosystem responses to these pulse events will be a key feature of ARID research  
7 approach. In addition, experimentally imposing moisture pulses and then monitoring responses,  
8 and pulse chasing such as deploying aircraft, drone, mobile flux towers, and personnel for field  
9 measurements during and immediately following naturally occurring moisture pulses. It also  
10 includes utilizing high temporal frequency satellite measurements such as passive microwave  
11 remote sensing, which measure at near daily timescales in all-weather conditions, and  
12 geostationary satellites that measure across infrared spectra at sub-hourly scales. Finally, using  
13 these data, it is essential to develop characteristic responses of states and fluxes to pulse  
14 events for model development.

15 As increasing aridity across dryland regions affect vegetation condition and fuel moisture levels,  
16 new techniques are primed to rapidly accelerate our understanding of these vegetation-fire  
17 interactions with novel ecohydrologic applications. In particular, advances in active remote  
18 sensing techniques include light detection and ranging (LIDAR) and synthetic aperture radar  
19 (SAR) are enabling more accurate monitoring of forest structure and water content; while  
20 advances in visible to shortwave infrared (VSWIR; ~400 to 2500 nm) imaging spectroscopy  
21 (also known as hyperspectral imaging and hyperspectral remote sensing), thermal infrared, and  
22 solar-induced fluorescence remote sensing are enabling more accurate monitoring of  
23 ecosystem composition and function (W. K. Smith et al., 2019b).

24 In addition, LiDAR data from field to satellite (e.g., NASA GEDI) scale are increasingly being  
25 used to monitor key predictors of fire occurrence and severity including ecosystem surface fuels  
26 and forest density in forest and rangeland ecosystems. Wintertime LiDAR data can also be  
27 applied to monitor changes in winter snowpack, a key early indicator of ecosystem drought and  
28 critical fuel moisture conditions (Duncanson et al., 2022; Dwivedi et al., 2024; Painter et al.,  
29 2010). Microwave data from field to satellite (e.g., NASA SMAP) and the deployment of Landsat  
30 Next in 2030 are very complimentary to LiDAR data and can be used to explicitly track changes  
31 in ecosystem fuel moisture dynamics over seasonal timescales (Humphrey & Frankenberg,  
32 2023; Rao et al., 2020; Yao et al., 2020). Active microwave techniques, e.g., synthetic aperture  
33 radar (SAR), can be applied to achieve very high spatial resolution measurements for enhanced  
34 fuel moisture monitoring along fuel treatment gradients (e.g., NASA NISAR) (Rao et al., 2020;  
35 Santoro et al., 2021). Thermal infrared (e.g., NASA ECOSTRESS) and solar-induced  
36 fluorescence (e.g., NASA OCO-3) observations can also be applied as an early indicator of  
37 ecosystem drought stress and for active fire monitoring for improved fire response times and  
38 management (Farella et al., 2022). Imaging spectrometer observations from field to satellite  
39 scales (e.g., NASA EMIT) can be used to determine species composition dynamics both pre-  
40 and post-fire to better understand complex vegetation-fire feedbacks and to monitor ecosystem  
41 state changes following fire events (W. K. Smith et al., 2019b).

42 In addition, additional field, aircraft/drone, and remote sensing observations will enhance  
43 characterization of land-atmosphere interactions related to drought and climate variability  
44 dynamics. Several key approaches to address these science questions and challenges include:  
45 (1) Isolate the driving factors of land surface behavior using techniques such as (a)

1 experimental setups to isolate variables and interactions (i.e., using rainout shelters to keep soil  
2 moisture dry/constant compared to ambient conditions and measure ET or GPP differences), (b)  
3 model experiment isolations (like GLACE-CMIP5) where driving processes are isolated within  
4 the model scheme (Koster et al. 2004), and (c) statistical analyses using observational data  
5 (Granger Causality/convergent cross mapping/Sugihara causality) (Green et al. 2017, Sugihara  
6 et al. 2012). (2) Examine the role of land management and patchiness of vegetation and/or soil  
7 moisture on ET and GPP fluxes. (3) Simulate drought experiments or monitor natural ones to  
8 determine role of extremes on dryland ET and GPP fluxes and their recovery. (4) Quantify the  
9 sensitivity of dryland GPP and ET fluxes to soil water availability variations, which is currently an  
10 uncertain model parameter. These approaches can all be addressed using simultaneous soil  
11 moisture and flux measurements (or flux proxies) at any scale and using observed or modeled  
12 data. Importantly, many of these approaches require measuring land and atmosphere properties  
13 at hourly or daily scales where weekly timescales will miss these rapid dryland dynamics.  
14 Similarly, high spatial resolution measurements will be needed to distinguish between signals  
15 from soil and vegetation in remote sensing and flux tower footprints.

### 16 ***B.1.2 Science Theme 2: Ecosystem Structure, Function, and Biodiversity***

17 Drylands in the US and globally have experienced significant change in structure, composition,  
18 and productivity and more significant change is predicted under climate change and land  
19 management practices (Boone et al. 2017; Godde et al., 2020; Joyce et al., 2013;  
20 Kleinhesselink et al., 2023). Conventional remote sensing and modeling approaches have  
21 poorly represented vegetation structure in drylands characterized by low cover and density of  
22 woody plants across spatially heterogeneous landscapes. These landscapes are characterized  
23 by their horizontal, vertical, and temporal heterogeneity in the cover and density of bare soil,  
24 biocrust, and distinct plant functional types (PFTs) (W. K. Smith et al., 2018). The fine spatial  
25 and temporal scale of dryland variability has meant that historical remote sensing and modeling  
26 of drylands has represented vegetation structure, composition and function at aggregate scales  
27 not well-suited to actual dryland systems. Indeed, many remote sensing and modeling  
28 approaches assume full cover and dominance by a single PFT. Even in models representing  
29 distinct C3 and C4 photosynthetic pathways, the uncertainty in PFT fractional cover leads to  
30 higher uncertainty in modeled carbon, water, and energy fluxes than in more homogeneous  
31 ecosystems (Hartley et al., 2017b).

32 Drylands also have remarkably high biodiversity, including 35% and 20% of global diversity and  
33 plant diversity hotspots, respectively (Maestre et al., 2021). The long evolutionary history of  
34 dryland ecosystems and their role as the origin of many unique plant lineages have given rise to  
35 multiple biodiversity hotspots around the world (e.g. Succulent Karoo in South Africa, and  
36 Sonoran Desert in North America, Gran Chaco and Caatinga in South America, dry forests of  
37 Meso-America and Australia) (Maestre et al., 2021). The high diversity of drylands adds to the  
38 complexity of measuring and modeling vegetation structure and function, while highlighting the  
39 critical need for improved EO in global drylands.

40 The biodiversity and range of PFT's that characterizes the dryland systems also contributes to  
41 the variability in ecosystem functions and processes rates that respond to alternating periods of  
42 dryness and wetness, responses to extreme climate conditions, and diurnal cycles of water and  
43 carbon dynamics. Understanding the ecosystem responses to climate variability as changes to  
44 ecosystem structures change will be a key focus of ARID.

1 Our overall question we ask in this theme is: *What are the main mechanisms driving the*  
2 *spatiotemporal distributions of dryland structure, function, and biodiversity and what is their*  
3 *vulnerability to change?*

#### 4 *B.1.2.1 Sub-Theme: Vegetation Structure and Heterogeneity*

##### 5 *Motivation*

6 While recent NASA Terrestrial Ecology field campaigns have greatly advanced Earth  
7 observation and modeling of vegetation processes in mesic ecosystems (tropical, boreal and  
8 arctic), a new focus on dryland remote sensing, including improved detection and quantification  
9 of horizontal and vertical vegetation structure as well as PFT diversity, will pave the way  
10 towards greater structural and functional realism in remote sensing retrievals. Improvements in  
11 our ability to measure dryland vegetation structure and composition will enhance our  
12 understanding of dryland biodiversity and representation of dryland vegetation type and cover  
13 (and associated carbon, water and ecosystem services) in Earth system models.

14

##### 15 *Questions*

16 We seek to address: How are ongoing changes in climate (rising CO<sub>2</sub>, increases in drought  
17 frequency and intensity, long-term changes in mean rainfall and changing fire regimes) and land  
18 use (changing fire regimes grazing and other land uses) impacting vegetation structure, function  
19 and habitat in drylands? More specifically, what are the rates and underlying causes of woody  
20 plant encroachment (WPE) in global drylands? How do invasions by exotic shrubs and grasses  
21 observed in drylands around the world impact ecosystem function and disturbance regimes?  
22 How are changes in structure and function impacting essential ecosystem services, including  
23 changes in water, carbon and nutrient cycling in drylands and provision of forage for livestock?

24 These primary questions give rise to a series of sub-questions that will enhance our earth  
25 system observation capabilities and ecosystem understanding of drylands, including: (i) Can we  
26 optimize new sensors and sensor-synergies to retrieve separate biophysical/structural data (e.g.  
27 cover, height, biomass, leaf area, light interception, etc.) for distinct woody and herbaceous,  
28 grass, forb and succulents, C<sub>3</sub> and C<sub>4</sub> PFTs, biocrust and bare soil in drylands? (ii) Can we  
29 utilize remote sensing data to improve the estimation of dryland vegetation physiological status  
30 (e.g. photosynthetic quantum yield, nutrient, pigment and enzyme concentrations, canopy  
31 stomatal and hydraulic properties, vegetation and soil water status, drought stress) for distinct  
32 PFT and biocrust communities? (iii) Can we improve retrievals of vegetation function (e.g.  
33 carbon, water, and energy fluxes) in drylands, either directly (e.g. thermal estimation of energy  
34 balance), or via use of improved structural, physiological, and phenological retrievals in process-  
35 based models? (iv) Can we use enhanced remote sensing to detect changes in vegetation  
36 structure and function associated with management and disturbance events in drylands,  
37 including fire, wood harvest, agricultural clearance, grazing, and invasive species? (v) Can we  
38 improve predictions of future changes in vegetation composition, structure and function under  
39 climate change using novel datasets and data fusion for model development, initialization,  
40 evaluation, and data assimilation to optimize parameters related to vegetation processes?

41

1 *B.1.2.2 Sub-Theme: Dryland Biodiversity*

2 *Motivation*

3 While harsh environmental constraints have been hypothesized to limit functional diversity in  
4 drylands, these dry ecosystems host a higher-than-expected functional diversity. This  
5 “functional paradox” could be attributed to unpredictable environmental conditions allowing  
6 alternative plant strategies to coexist, leading to high functional, trait, and species diversity  
7 (Gross et al., 2024). Horizontal, vertical, and temporal heterogeneity in plant functional types  
8 (woody perennials, annual and perennial C3 and C4 grasses, annual and perennial forbs), and  
9 diverse biocrust communities make for diverse autotrophic communities in drylands. High  
10 spatiotemporal variability in water availability has driven unprecedented species diversity across  
11 global drylands, including dynamic mixtures of C3 and C4 grasses, shallow-rooted CAM cacti  
12 and agaves, and deep-rooted C3 shrub and tree species, each with specialized water  
13 acquisition strategies (Gross et al., 2024). Diverse dryland plant communities in turn support  
14 significant diversity in invertebrate and vertebrate herbivore and predator populations. However,  
15 the scale-dependent processes through which structural heterogeneity (MacArthur and  
16 MacArthur 1961; Wu 2004) and temporal variability due to disturbances (fire and rainfall pulses)  
17 drive dryland biodiversity is poorly understood.

18 Drylands are amongst the most vulnerable ecosystems to climate change (D. Li et al., 2018)  
19 with serious impacts on dryland biodiversity. For example, 44-88% of global dryland tree  
20 species face significant decline under predicted increases in aridity (Cartereau et al., 2023)  
21 along with cascading effects on other taxa. It is furthermore widely established that land  
22 degradation in drylands results in substantial loss in biodiversity and ecosystem services  
23 (Davies et al., 2012; Lewin et al., 2024; Montanarella et al., 2018.). Grazing, the dominant land  
24 use in drylands, can have major impacts on biodiversity that become more pronounced in  
25 rangelands with higher aridity, and a drier future (Gross et al., 2024; Maestre et al., 2022b). In  
26 addition, US rangelands have experienced significant changes in vegetation species  
27 composition with the cover and production of annual plants now exceed that of perennials on >  
28 21 million ha of rangeland managed by BLM, marking a fundamental shift in the ecology of  
29 these lands (Kleinhesselink et al., 2023). BLM furthermore needs species-specific products on  
30 the extent of Sagebrush (primarily for conservation of endangered Greater sage-grouse,  
31 *Centrocercus urophasianus*) and pinyon juniper cover, as well as invasive annual grass cover  
32 specifically.

33 *Questions*

34 We ask: What are the drivers of biodiversity (functional, phylogenetic, taxonomic) in drylands  
35 and how will these be changing? More specifically, we ask: what is the relationship between  
36 biodiversity, ecosystem function, carbon stocks and resilience to disturbances (drought and  
37 fire)? What are the impacts of climate change (increased aridity) and continued grazing  
38 pressure on biodiversity (functional, phylogenetic, taxonomic, etc.)? What is the impact of land  
39 cover transformation, land degradation, and cultivation on dryland biodiversity and species of  
40 special interest (e.g. endangered species)? What are the impacts of invasive plant species on  
41 biodiversity, rangeland condition and water-availability?

42

43 *B.1.2.3 Sub-Theme: Ecosystem Function*

44 *Motivation*

1 As a result of the complexity in community composition and ecosystem structure, dryland  
2 ecosystem function is also a complex mosaic of multiple functional groups moving between  
3 dormancy and activity over unique diurnal, daily, seasonal, and interannual time scales. The  
4 changing ecosystem functional dynamics of dryland ecosystems are affecting large-scale  
5 energy, water, and biogeochemical cycles (e.g., carbon, nitrogen, and phosphorus). The rates  
6 of change in these ecosystem processes are regulated by biodiversity and ecosystem  
7 conditions operating in response to climate anomalies, land and water use practices, and  
8 altered intensity and duration of disturbance events.

9 Drylands can be in a state of relative dormancy following dry events that endure for weeks to  
10 several months, until a rain event recharges surface soil moisture driving an initial pulse of  
11 heterotrophic respiration and loss of carbon from the ecosystem (Roby et al., 2022; Throop et  
12 al., 2020). Depending on the timing and intensity of the rain event and the depth and texture of  
13 the soil, root zone soil moisture (RZSM) will next recharged, driving a pulse of ecosystem  
14 carbon uptake through photosynthesis or gross primary productivity (GPP) that can continue  
15 until critical soil moisture thresholds are reached (Feldman, Feng, et al., 2024b; Z. Fu et al.,  
16 2024). Dryland GPP and ecosystem respiration can be co-limited by nutrients such as the  
17 macronutrient nitrogen, which can be regulated by nitrogen-fixing biocrusts and C3 tree species  
18 (Reed et al., 2011; Vitousek et al., 2013). The balance between GPP, total ecosystem  
19 respiration (heterotrophic and autotrophic), and carbon losses through disturbances such as  
20 fire, grazing, and erosion ultimately determine the net ecosystem exchange of carbon (NEE) - a  
21 critical measure of the contribution of drylands to the global carbon cycle (Ahlström et al., 2015;  
22 Poulter et al., 2014).

23 Dryland vegetation is fundamentally controlled by plant hydraulic considerations through the  
24 soil-plant-atmosphere water continuum (SPAC), such as root water uptake and leaf and xylem  
25 conductances. Ultimately, leaf water potential, influenced by both soil water availability and  
26 vapor pressure deficit, is a strong control of photosynthesis and evapotranspiration rates. For  
27 instance, hydraulic and photosynthetic strategies differ greatly between isohydric species that  
28 close their stomata relatively quickly under drought, while anisohydric species can keep their  
29 stomata open under drought and endure much more negative water potential (Martínez-Vilalta &  
30 Garcia-Forner, 2017). Additionally, CAM plants such as dryland succulents avoid stressful  
31 atmospheric conditions during the day by photosynthesizing at night and by storing more water  
32 to maintain function under stressful conditions. Understanding the complex SPAC in different  
33 dryland plants as it controls dryland carbon (C) dynamics is of particularly high priority since we  
34 know drylands play a disproportionately large regulator of global trends and variability in  
35 atmospheric CO<sub>2</sub> concentrations (Ahlström et al., 2015; Humphrey et al., 2018; Poulter et al.,  
36 2014) (see section B.1.3). As determined by the SPAC, dryland GPP is predominantly water  
37 limited and has been found to be tightly coupled with anomalies in RZSM and ET (Biederman et  
38 al., 2017a; Scott et al., 2015). Slight changes in the balance between GPP, NEE, and ET, i.e.,  
39 ecosystem water use efficiency (WUE), can have large consequences for dryland carbon and  
40 water cycling (F. Li et al., 2023).

41

## 42 *Questions*

43 In seeking to understand how dryland GPP, NEE, and ET dynamics respond to climate change  
44 and in-turn contribute to global carbon and water cycling, we have identified key overarching  
45 research questions to be addressed by the ARID terrestrial campaign: 1) Across different

1 timescales, what are the dominant mechanisms driving dryland function such as plant  
2 hydraulics, leaf level photosynthesis, respiration, and nutrient cycling? 2) To what extent can we  
3 use novel remote sensing techniques to track changes in the SPAC and resulting changes in  
4 ecosystem function including GPP, NEE, ET, and WUE?; and 3) How has dryland functioning  
5 related to GPP, NEE, ET, and WUE changed over time and by functional type and what are the  
6 major driving factors (e.g., atmospheric CO<sub>2</sub>, RZSM, VPD, nutrients)?

7 We have further identified key sub-questions with direct management implications that could  
8 also be targeted as part of the ARID campaign that include: (i) What role has CO<sub>2</sub> fertilization  
9 played in driving changes in the SPAC and dryland functions GPP, NEE, ET, and WUE? (ii)  
10 What role has the timing and intensity of precipitation and temperature-driven increases in VPD  
11 played in driving changes in the SPAC and dryland functions GPP, NEE, ET, and WUE? (iii)  
12 Can we utilize remote sensing data to improve the estimation of dryland vegetation  
13 physiological status (e.g. photosynthetic quantum yield, nutrient, pigment and enzyme  
14 concentrations, canopy stomatal and hydraulic properties, vegetation and soil water status,  
15 drought stress) for distinct PFT and biocrust communities? (iv) Can we improve retrievals of  
16 vegetation function (e.g. carbon, water, and energy fluxes) in drylands, either directly (e.g.,  
17 thermal estimation of energy balance), or via use of improved structural, physiological, and  
18 phenological retrievals in process-based models? (v) Can improvements of decomposition of  
19 soil organic matter be improved by night-time monitoring of carbon exchanges and soil moisture  
20 levels?

21

#### 22 *B.1.2.4 Sub-Theme: Dryland Geology and Soils*

##### 23 *Motivation*

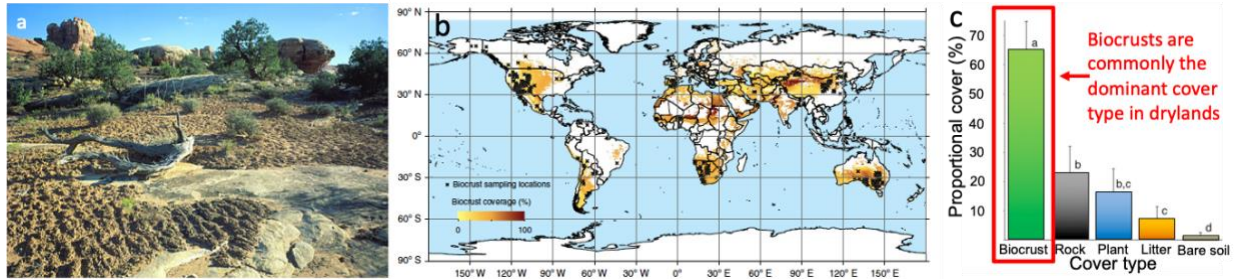
24 In drylands, bare rock, sediments, and soils strongly affect remotely sensed signals, facilitating  
25 characterization of geologic processes along with measurement of biologic functions. Exposed  
26 geology in drylands strongly connects to the wellbeing of humans, wildlife, natural and  
27 agricultural ecosystems.

28 Soils play critical roles in all terrestrial ecosystems, but the capacity to observe large amounts of  
29 soil from space-based sensors is rare outside of drylands. Because dryland soils are visible  
30 from space and the air, they provide exceptional opportunities to use coupled sensor space-,  
31 airborne, and ground-based approaches to 1) improve our quantitative and predictive  
32 understanding of dryland soil structure, function, and responsiveness to change and 2) improve  
33 our interpretation of soils' contribution to all space- and airborne based terrestrial data.  
34 Currently, 10-20% of drylands worldwide are classified as degraded or marginal. When  
35 considering that drylands also have the highest growth rate of any ecological zone (18.5%; Gaur  
36 & Squires, 2018), these statistics are concerning, as productive soils are a nonrenewable  
37 resource on human time scales, and recovery rates are notoriously slow in drylands.

38 Another way drylands are different from many wetter ecosystems is that they maintain extensive  
39 and diverse communities of photosynthetic soils, called biological soil crusts (biocrusts; Fig.  
40 B.4). Biocrusts are soil surface communities of mosses, lichens, and cyanobacteria that cover  
41 vast expanses of the terrestrial surface and play critical roles in soil stabilization, fertility, water  
42 cycling, and carbon exchange with the atmosphere (Elbert et al., 2012; Weber et al., 2022).  
43 Biocrusts are found on all of Earth's continents and global estimates suggest biocrusts make up



1 12% of Earth's land surface and represent up to 10% of desert NPP and > 25% of nitrogen  
 2 fixation (Elbert et al., 2012; Weber et al., 2022). However, the estimates of biocrust coverage  
 3 and quantitative understanding of biocrust contributions to global biogeochemical cycles remain  
 4 highly uncertain, in large part due to a relative lack of options for remotely sensing biocrusts.  
 5 Increased spatial resolution and advances in sensor diversity can build on past biocrust remote  
 6 sensing (Havrilla et al., 2020; Karnieli et al., 1999; Potter, 2016; Rodríguez-Caballero et al.,  
 7 2017; Rozenstein & Adamowski, 2017; W. K. Smith et al., 2019b; Yan et al., 2024), to  
 8 substantially improve quantitative assessments that monitor, track change, and scale  
 9 composition and function in biocrusts.



10  
 11 **Figure B.4.** (a) Landscape photograph showing biocrusts amidst shrubs, trees,  
 12 coarse woody debris, and bedrock. Biocrusts are the bumpy, brown and black  
 13 soil in the interspaces. (b) Estimated biocrust coverage (in %) within each  $0.5^\circ \times$   
 14  $0.5^\circ$  - black X's are sampling locations. Biocrusts cover 17.9 million km<sup>2</sup>: 12.2%  
 15 of the global terrestrial surface (Rodríguez-Caballero et al., 2018). Spatial  
 16 coverage was estimated with environmental modeling based on 911-point  
 17 measurements. (c) Surface cover of the main components of many drylands at a  
 18 site on the Colorado Plateau USA, showing biocrusts represent the dominant  
 19 cover type (Torres-Cruz et al., 2018).

20 The major global dust sources are found in arid regions. The role of dust in radiative forcing  
 21 (heating or cooling) is a major uncertainty in Earth System Models (Miller et al, 2006; Mahowald  
 22 et al, 2011). To address this gap in knowledge, the Earth Surface Mineral Dust Investigation  
 23 (EMIT) is collecting VSWIR imaging spectrometer data from the International Space Station  
 24 across the arid regions of the world (Thompson et al., 2023). Field studies of mineral  
 25 composition and abundance are needed on a more widespread basis for the validation of EMIT  
 26 results. Dust also contributes to the productivity (Swap et al., 1992) and water balance (Painter  
 27 et al., 2010) of non-dryland systems and represents an unquantified component of the global  
 28 carbon cycle (Webb et al., 2012). Climate shifts and land use intensification affect the  
 29 production of dust, yet the ecological importance of dust remains understudied (Field et al.,  
 30 2010; Mahowald et al., 2005).

31 In drylands, heavy rainfall over bare geologic parent material, rock outcrops and sediments, can  
 32 impart pulses of metallic element concentrations to hydrologic and biologic systems and result  
 33 in gully erosion events (Nordstrom 2009). Natural sources of acid rock drainage and heavy  
 34 metal contamination may be amplified in areas of legacy and active mining. Over 500,000  
 35 legacy mined areas are estimated in the U.S. with a significant fraction in the arid west;  
 36 however, no comprehensive or large regional assessments of the impact of natural acidic rock  
 37 and mined areas to stream water quality have yet been made.

1 Human activities affecting land, water, and air quality are starkly evident in arid regions.  
2 Population growth and shifts in national economies to sustainable energy systems are driving  
3 demands for mineral resources, such as the projected needs for lithium to enable a global  
4 movement from fossil fuels to electric vehicles and grid-level energy storage (Graham et  
5 al.,2021). The largest producers of lithium are in arid regions of Australia and Chile. Competition  
6 for such resources to make a “Green Energy Transition” is one example of recent trends in  
7 mineral extraction from drylands. Critical minerals, those important to national economies and  
8 security and subject to supply disruption, are being identified by national level studies (USGS,  
9 2022) and driving exploration from new sources and studies of extraction from mine waste  
10 (Sarker et al, 2022).

11

## 12 *Questions*

13 We ask: how do soil communities and physicochemical characteristics drive and respond to  
14 climate variability and ecosystem change? Further sub-questions capture the specific aspects of  
15 soil and geological interactions in dryland systems. These include: What is the extent,  
16 composition, and function of dryland biological soil crusts (biocrusts)? What roles do soils play  
17 in dictating aboveground structure, function, and response to change? How much inorganic  
18 carbon is stored in dryland soils across different soil types, aridity index gradients, and natural  
19 versus managed systems, and how vulnerable is the carbon to environmental shifts? What  
20 contribution do soil signals play in space-based observations of all terrestrial ecosystems? What  
21 are the sources, sinks, causes, and consequences of accelerated topsoil loss and dust in  
22 Earth’s drylands? What is the contribution of erosional (aeolian and hydrological) processes to  
23 dryland carbon cycling? How do changes in soil reflectance feedback to global climate (e.g.,  
24 albedo)? How does soil health contribute to dryland ecosystem resilience to change? How do  
25 soil environment changes affect the diversity and functions of soil microbial communities?

26

## 27 *B.1.2.5 Approaches for Ecosystem Structure, Function, and Biodiversity*

28 Historical remote sensing technologies generally did not provide data with sufficient spatial,  
29 spectral and temporal resolution to reliably separate different PFT in measurement of dryland  
30 vegetation structure, function, or species composition. However, new sensors and sensor-  
31 synergies on airborne and AUS platforms have significant new potential to retrieve separate  
32 structural metrics (e.g., density, height and size of individual woody plants, fractional cover, leaf  
33 area index, height and biomass of herbaceous PFT), enabling the separation of  
34 annual/perennial herbaceous grasses and forbs, woody plants, and biocrust. More work is  
35 needed using very high-resolution imagery and diverse passive and active sensors (LiDAR,  
36 Radar, VSWIR imaging spectroscopy) to retrieve these structural metrics. Hyperspatial  
37 approaches from very high-resolution remote sensing (cm scale) allows quantification of  
38 individual tree/shrub biomass, which can be calibrated with ground observations (e.g., TLS, tree  
39 rings, forest inventory) (Tucker et al., 2023a).

40 As a key implication of vegetation structure, wildfire characteristics in drylands (e.g., surface or  
41 crown fire) depend on fuel type (herbaceous, woody or litter), fuel amount, and flammability  
42 (Kahiu & Hanan, 2018). The challenge for remote sensing of fuels for fires in drylands is that  
43 changes in vegetation structure and productivity impacts the probability, propagation and type of

1 fire and therefore needs to be estimated at fine spatial and temporal scales. However, improved  
2 remote sensing of productivity, biomass and water content of different PFT provides new  
3 opportunities to track fuel loads and thus potential for fire occurrence and type of fire. The ARID  
4 campaign will improve high spatial and temporal resolution vegetation structure and productivity  
5 retrievals, with green and dry biomass assessments, to provide improved inputs on fuel loads  
6 for fire monitoring and modeling.

7 The ARID field campaign will also focus on improved quantification of dryland vegetation  
8 physiological state and function in systems frequently limited by drought and heat stress.  
9 Orbiting and geostationary remote sensing assets can contribute to this effort, with  
10 geostationary satellites now providing opportunities for near-continuous monitoring of land  
11 surface temperatures and greenness and orbiting platforms providing both high resolution and  
12 high return-frequency VSWIR imaging. In particular, infrared and microwave retrievals of  
13 vegetation water status offer significant potential as input for modeling dryland vegetation  
14 dynamics and land surface models, while VSWIR imaging spectrometers can provide enhanced  
15 measurement of physiological capacities relating to photosynthesis, evapotranspiration, growth,  
16 and phenology.

17 Enhanced observational capabilities are providing improved characterization of biodiversity  
18 across various landscapes (BioSCape campaign 2023) with AVIRIS and LiDAR data. In  
19 addition, airborne and UAS image spectroscopy and LiDAR, accompanied with intensive field  
20 work during the ARID field campaign can capture plant diversity and structure of PFTs at fine  
21 scales. The long-term field data on plant species composition recorded at LTER, NEON, and  
22 new supersites will be used to train models to map functional and taxonomic diversity, as well  
23 as specific grass and shrub species with the airborne and UAS data. These models will be  
24 scaled-up to space-based image spectroscopy (EMIT, SBG, EnMap, Landsat Next) and  
25 LiDAR+SAR (GEDI, NISAR, EDGE) to address science questions across vast drylands.

26 Satellite remote sensing has been used to estimate GPP, NEE, ET, and WUE dynamics across  
27 spatial scales back to the early 1980's (Fisher et al., 2017; Ryu et al., 2019). These estimates  
28 have helped shape our current understanding of dryland function within the context of the  
29 broader Earth system (Ahlström et al., 2015; Humphrey et al., 2018; Poulter et al., 2014).  
30 However, these key ecosystem functions cannot be directly observed and while the spatial  
31 variability of these products is generally consistent with field measurements, their temporal  
32 variability suffers from substantial uncertainties and bias, particularly across dryland ecosystems  
33 (Biederman et al., 2017a; Verma et al., 2014). New remote sensing techniques including solar-  
34 induced fluorescence (SIF) – a factor closely related to GPP and NEE – and land surface  
35 temperature (LST) – a factor closely related to GPP, NEE, and ET offer exciting new  
36 opportunities to overcome past limitations. Space-borne SIF estimates are currently being  
37 provided by the ISS OCO-3, the polar-orbiting OCO-2, and the geostationary TEMPO mission.  
38 LST estimates have been provided historically by the polar-orbiting Landsat and MODIS  
39 missions and new estimates include the ISS ECOSTRESS and the geostationary GOES  
40 mission.

41 Furthermore, microwave-based measurements of soil moisture and vegetation optical depth  
42 (from SMAP) and radar backscatter (from NISAR) can provide insights into plant hydraulics and  
43 the various dryland plant water use strategies under water limitation. Ground-, UAS-, and  
44 aircraft based GNSS microwave instruments can be further used to investigate plant hydraulic  
45 and stomatal conductance strategies across C3, C4, and CAM plants (Feldman, 2024b).

1 Dryland systems often have heterogeneous vegetation coverage and resulting bare ground and  
2 rock surfaces interspersed. Biological soil crusts are also part of the landscape matrix of dryland  
3 systems. These surface properties such bare ground fraction, surface mineralogy, digital  
4 elevation data, as well as surface biophysical data mentioned in B1.1.5 will provide information  
5 to more accurately interpret the surface and soil characteristics from space borne sensors.  
6 Remote sensing will be used to identify indicators of soil properties (e.g., soil moisture, organic  
7 matter content, nutrients, texture, salt accumulation) along the replicated climatic and ecological  
8 gradients (Viscarra Rossel, 2016).

9 Advances in remote sensing capabilities now provide opportunities to integrate remote sensing  
10 data with high-resolution soil biogeochemistry datasets derived from field experiments (multiple  
11 temporal and vertical measurements) to identify biological, physical, and chemical indicators of  
12 soil health/degradation. Satellite remote sensing observations and advanced analysis  
13 techniques such as microwave polar orbit (1-3 days), thermal polar orbit (VIIRS, Landsat, etc.),  
14 thermal geostationary (15 minute) (GOES, Himawari, MSG), ISS instruments (ECOSTRESS,  
15 OCO3) (1-5 day), rainfall, satellite and ground data fusion (e.g., GPM, PERSIANN-CDR) all  
16 provide greater capacity to analyze surface characteristics that influence soil and geological  
17 processes.

18 Recent campaigns have provided information on enhanced mineral observations useful in  
19 dryland landscapes. Remote sensing data from NASA's Earth Surface Mineral Dust Source  
20 Investigation (EMIT) mission & NASA-USGS Geological Earth Mapping Experiment (GEMx)  
21 field campaign are imaging spectroscopy missions that provide spectral information at 8-10  
22 nanometers from the visible to shortwave infrared wavelengths (VSWIR) at ~60 m and ≤15 m  
23 spatial resolution, respectively. The prime mission for EMIT and for GEMx (using AVIRIS  
24 instruments for the VSWIR and MASTER for the thermal infrared) is to map and understand  
25 mineral composition of dryland systems and the archive is available on the NASA Land  
26 Processes Distributed Active Archive Center (LPDAAC).

27 Given the role of drylands in providing forage for livestock and wildlife, quantifying the temporal  
28 phenology and spatial variability in green and senescent herbaceous and woody forage is  
29 challenging, but is a critical goal for ARID. New sensors and sensor-model fusion approaches  
30 have potential to estimate cover and biomass of green and senescent vegetation and thus  
31 graze and browse resources for domestic and wild herbivores. ARID will provide opportunities  
32 for development of growing season productivity and dry season decline using diverse remote  
33 sensing approaches, including high spatial resolution imagery for tree canopy characterization,  
34 high temporal resolution HLS or commercial data (e.g., Maxar, Planet) for sub-pixel unmixing.

35

### 36 ***B.1.3 Science Theme 3: Carbon Cycle Interannual Variability and Long-Term Trends***

37 The ARID field campaign has the potential to substantially increase our understanding of the  
38 role drylands play in the global carbon cycle. We ask, what is the contribution of drylands to the  
39 mean, trend, and particularly the interannual variability of terrestrial carbon dynamics? What are  
40 the underlying drivers of this dryland net carbon exchange? Can we improve quantification of  
41 dryland carbon stocks, fluxes, and controls? Addressing these issues is vital for near-term  
42 management and policy decisions aimed at climate change mitigation. Namely, it is challenging  
43 to distinguish the effects of rapid increases or decreases in anthropogenic carbon emissions,  
44 and thus any nation-level or global mitigation efforts, from the rapid increases or decreases in

1 terrestrial net carbon exchange we see in drylands. Disentangling drylands' contributions from  
2 interannual climate variability is needed to quantify the effects of climate action.

3 Currently, about half of anthropogenic carbon emissions are removed from the atmosphere  
4 annually by natural carbon sinks, with roughly equal amounts taken up by the land and the  
5 ocean. Drylands are characterized as having low organic carbon stocks on a per-area basis -  
6 particularly above ground - but high, variable, and poorly predicted carbon fluxes (Biederman et  
7 al., 2017a). Many global scale land surface models are presently unable to replicate the mean  
8 net or gross carbon uptake or its year-to-year variability at in-situ flux towers (MacBean et al.  
9 2021; Teckentrup et al., 2021; Wang et al., 2022). This is problematic because drylands are  
10 vast, and whether they act as a sink or a source of carbon in any given year can strongly  
11 influence the temporal variability in the global carbon cycle (Poulter et al. 2014, Ahlstrom et al.  
12 2015; Zhang et al., 2018). Untangling the dryland carbon interannual variability issue will  
13 strongly depend on the integrated studies outlined above in the Ecosystem Structure and  
14 Function sections, as well as the Climate Variability and Drought section. Whether drylands act  
15 as a sink or a source of carbon in any given year strongly influences the interannual variability in  
16 the CO<sub>2</sub> growth rate.

17 In addition, given their large land area, drylands are viewed for their potential for afforestation  
18 and renewable energy as natural climate solutions. However, poor understanding of the dryland  
19 carbon and energy balance prevents prediction of the benefits. Indeed, tree planting can  
20 increase carbon uptake and cool the surface (Duveller et al. 2018), however in drylands, any  
21 additional evapotranspiration associated with tree planting efforts may result in unintended  
22 consequences such as salinization of soils or depletion of soil moisture. Further, creating dense  
23 tree canopies in drylands will increase the risk of fire and other disturbances through the  
24 creation of above-average fuel loads - and thus the rapid loss of gained carbon stocks in  
25 catastrophic events. Also, adding trees or more vegetation may even warm the surface through  
26 reduced albedo compared to the underlying soil (Rotenberg and Yakir 2010, Williams et al.  
27 2021 Sci. Adv., Feldman et al. 2023). Therefore, given water-limitation in drylands, such carbon  
28 sequestration benefits of tree planting may only be conferred in some instances (Rohatyn et al.  
29 2022) and the new knowledge created by ARID could help quantify, scale, and forecast the  
30 utility or potential problems with these national and global efforts.

31

32 The overall question we ask in this theme is: *What is the contribution of drylands to the mean,*  
33 *trend, and interannual variability of terrestrial carbon uptake and what drives these patterns?*

34

### 35 *B.1.3.1 Sub-theme: Carbon Stocks and Fluxes*

#### 36 *Motivation*

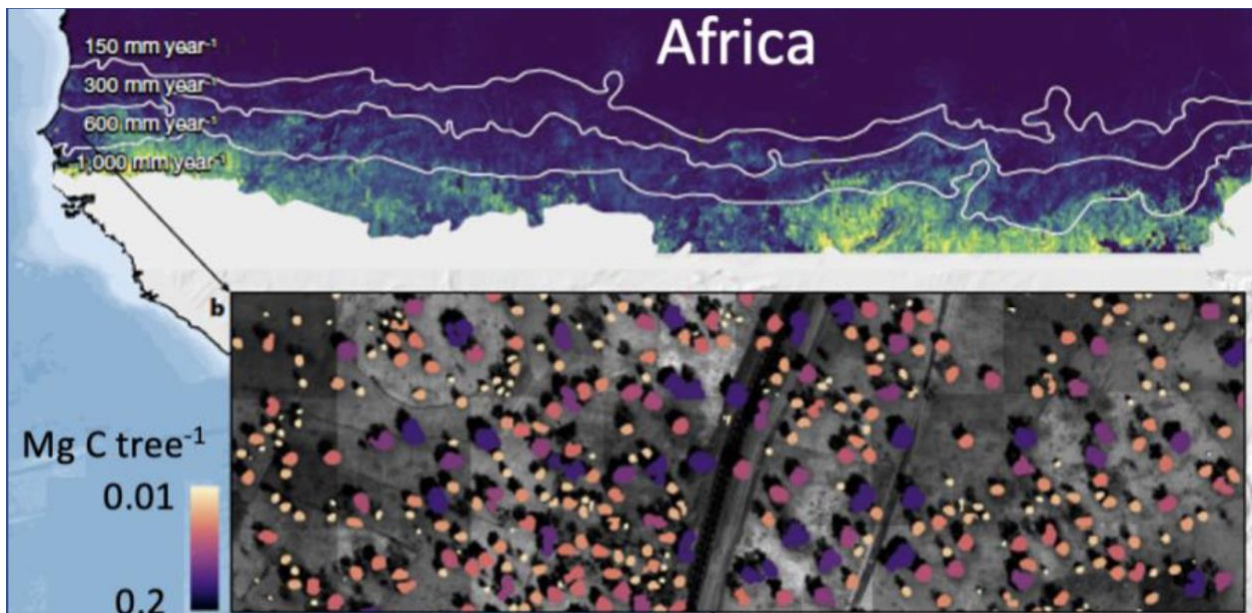
37 Persistent challenges remain in understanding the large dryland contribution to the global  
38 carbon cycle variability and its uncertainties, as well as in predicting the impacts of carbon cycle  
39 responses to climate and land use changes on vegetation biomass, structure, biodiversity and  
40 function. Global-scale, process-based land surface and dynamic vegetation models, which form  
41 the land component of earth system models, poorly capture dryland gross and net carbon  
42 fluxes, with considerable spread across model estimates (MacBean et al. 2021; Teckentrup et  
43 al., 2021; Fawcett et al., 2022). The combined effects of short-term pulse events, seasonal to  
44 interannual alterations of precipitation patterns, and changes in ecosystem structure, add to the  
45 complexity of dryland carbon cycle and fluxes. Furthermore, remote sensing-based production

1 efficiency models that predict GPP and NEE perform notoriously poorly in drylands (Biederman  
2 et al., 2017a, Smith et al. 2019).

3

4 Although historically challenged by high heterogeneity, new remote sensing tools are  
5 revolutionizing the ability to quantify carbon above and belowground in drylands. For example, a  
6 recent joining of high-resolution satellite imagery, deep learning, and ground-based allometric  
7 relationships showed the power to map tall vegetation across vast areas and estimate carbon  
8 stocks in drylands at the individual tree scale (Fig. B.5) (Tucker et al., 2023b). In Africa, 9.9  
9 billion trees were assessed for carbon, including wood, foliage, and root carbon. Comparisons  
10 with numerical simulation models for the area found that carbon stocks of trees were  
11 underestimated by some models and overestimated by others. Such high-resolution remote  
12 sensing of drylands, linked with modeling, has large potential for model benchmarking and for a  
13 much improved quantitative understanding of carbon storage and change in drylands. The  
14 increase in remote sensing observations that do not rely on greenness to assess carbon  
15 cycling, such as SIF, also bring new opportunities to improve estimates of carbon flux for  
16 dryland plants that often remain green even while their stomates are closed and carbon uptake  
17 is minimal (W. K. Smith et al., 2018; X. Wang et al., 2022; Y. Zhang et al., 2023). Dryland soils  
18 too store vast amounts of carbon in organic and inorganic forms and improving quantification of  
19 those stocks and their vulnerability to change will be critical in advancing estimates of carbon  
20 storage and flux in global drylands (see Section B.1.2.4). New capabilities for high temporal  
21 resolution assessment of soil dryland CO<sub>2</sub> exchange provide unprecedented opportunities to  
22 assess biocrust and soil heterotroph contributions to overall exchange (Darrouzet-Nardi et al.,  
23 2015; e.g., as assessed by eddy covariance tower), allowing for the attribution of carbon  
24 sources and sinks at the within-ecosystem scale.

25



26

27 **Figure B.5.** Carbon estimates of 9,947,310,221 trees across 9.7 million km<sup>2</sup>.  
28 The study covered much of Africa and estimated woody carbon stock of single  
29 trees. Such remote sensing tools would be incredibly valuable not only for

1 *improving estimates and forecasts of carbon cycling in global drylands, but*  
2 *evaluating and prioritizing mitigation and adaptation efforts (Tucker et al., 2023b).*

3

#### 4 *Questions*

5 We seek to answer the following questions: How large are the carbon stocks and fluxes in  
6 drylands, how do they vary at sub-annual to decadal timescales, and what is their response  
7 through space and time, to drivers of global change? Several more specific questions should be  
8 addressed: How vulnerable are dryland carbon stocks and fluxes to global and regional  
9 changes in water availability and atmospheric demand, combined with asynchronous plant  
10 responses to periods of water stress (B.1.2.1)? How do woody encroachment and  
11 desertification change dryland carbon cycling and feedback? What is the relative contribution of  
12 different plant functional types and biocrusts to carbon stocks and how are these changing  
13 along environmental gradients? What influence do rooting strategies and belowground carbon  
14 allocation play in ecosystem carbon storage? What is the potential of carbon capture in dryland  
15 vegetation and soils to work as a nature-based climate solution (NCS)? How can remote and in-  
16 situ observations be best used to model and quantify carbon stocks and fluxes?

17

#### 18 *B.1.3.2 Sub-theme: Integrated Approach to address Carbon Cycle Theme*

19 Estimating aboveground biomass in sparse, short stature, heterogenous drylands (e.g.  
20 savannas and shrublands) is challenging with space-based sensors and algorithms that are  
21 mainly designed for forests. This limits our ability to accurately quantify aboveground carbon  
22 stocks. However, new sensors and sensor-synergies offer unprecedented potential to retrieve  
23 separate structural metrics (e.g., density, height, and size of individual woody plants, fractional  
24 cover, leaf area index, height and biomass of herbaceous PFT), enabling the separation of  
25 annual/perennial herbaceous grasses and forbs, woody plants, and biocrust. More work is  
26 needed using very high-resolution imagery (Tucker et al., 2023) and diverse passive and active  
27 sensors (LiDAR, Radar, VSWIR imaging spectroscopy) to retrieve these structural metrics.  
28 Hyperspatial approaches from very high-resolution remote sensing (cm scale) allows  
29 quantification of individual tree/shrub biomass and thus high-resolution aboveground carbon  
30 storage, which can be calibrated with ground observations (e.g., terrestrial laser scanning, tree  
31 rings, forest inventory) (Babst et al. 2018, Tucker et al., 2023).

32 Production efficiency models that rely on satellite remote sensing inputs cannot partition GPP  
33 according to the fine scale plant functional type (PFT) heterogeneity. Satellite remote sensing  
34 has been used to estimate GPP, NEE, ET, and WUE dynamics across spatial scales since the  
35 early 1980's (Fisher et al., 2017; Ryu et al., 2019). However, these key ecosystem functions  
36 need to be better linked to appropriate PFT's and information on vegetation state (e.g.,  
37 photosynthetic capacity), as well as physiological function (e.g., photosynthetic activity). There  
38 is co-location of NASA instruments on the ISS, including the OCO-3 and ECOSTRESS, as well  
39 as Landsat Next missions and enhanced capabilities of geostationary platforms. These  
40 instruments together offer an exciting new opportunity to overcome past limitations, by providing  
41 first-time insights into the diurnal to seasonal dynamics of SIF and LST, factors closely related  
42 to GPP, NEE, and ET.

43 In addition, coupling these satellite-based observations with more detailed ground based tower  
44 flux data deployed over the diverse dryland landscapes will greatly improve ecosystem process  
45 understanding of these observations. Spaceborne instruments such as OCO-2 and 3 can

1 retrieve atmospheric columnar CO<sub>2</sub> (XCO<sub>2</sub>), which can be directly linked to net CO<sub>2</sub> exchange  
2 via atmospheric inversions. OCO-2 MIP is a useful top-down inversion that uses these satellite  
3 retrievals and can estimate the carbon cycle and its variability (Byrne et al., 2023).

4 Tower flux studies in combination with machine learning approaches can also be used to scale  
5 up carbon flux measurements. The NASA ARID campaign provides a unique opportunity to  
6 coordinate high resolution airborne and UAV campaigns with in-situ networks of towers and field  
7 sites to (i) validate spaceborne biomass and CO<sub>2</sub> flux data from current and future satellite  
8 missions (e.g. SBG, NISAR, OCO, EDGE candidate explorer mission), and (ii) to better test,  
9 develop, and optimize models so that they can more accurately represent and estimate dryland  
10 carbon stocks and fluxes. Modeling of carbon stocks and fluxes will be further improved by more  
11 accurate quantification of above- vs. belowground biomass ratios and of PFT fraction using  
12 satellite and airborne observations (Section B.2.1), calibrated by extensive ground observations  
13 (Section B.1.2.1). Finally, focused regional evaluation of in-situ measurements, new aircraft  
14 measurements, and/or focused calibration of carbon flux-based remote sensing indices can all  
15 contribute to informing both machine learning approaches and top-down inversions (Byrne et  
16 al., 2023) that scale dryland carbon fluxes to larger scales. All these fields, airborne, and  
17 hyperspatial approaches allow model development, testing, and optimization to improve  
18 representation and accuracy of dryland carbon cycling in process-based models.

19

20 The ARID field campaign will also focus on improved quantification of dryland vegetation  
21 physiological state and function in systems frequently limited by drought and heat stress and  
22 affected by fire and other disturbances affecting carbon stocks. In addition, the ARID campaign  
23 will enhance our ability to study ecosystem recovery from disturbances that influence the level  
24 of carbon uptake across these high stress and frequently disturbed landscapes. Orbiting and  
25 geostationary remote sensing assets can contribute to this effort, with satellite constellations  
26 and geostationary satellites now providing opportunities for near-continuous monitoring of land  
27 surface reflectance and temperatures. Together VSWIR surface reflectance, thermal infrared  
28 and microwave retrievals of vegetation water status offer significant potential: (i) to enhance  
29 measurement of physiological capacities relating to photosynthesis, evapotranspiration, growth,  
30 and phenology; and (ii) as observational constraints to improve the representation of dryland  
31 vegetation dynamics in land surface models (W. K. Smith et al., 2019b).

32

### 33 ***B.1.4 Science Theme 4: Social Ecological Systems***

34 The ARID field campaign will support efforts to study dryland social ecological system dynamics  
35 and livelihoods. Many of the products and insights gained from ecosystem process studies,  
36 ecosystem structural observations, and climate variability analyses will greatly improve  
37 understanding of drivers of change to dryland social-ecological systems and potential impacts to  
38 dryland livelihood practices. Research in these social-ecological systems will be conducted  
39 jointly with research activities described in the previous sections.

40 Dryland systems serve as critical water and land resources and habitat for wildlife and livestock,  
41 dryland crops, energy production, and other critical ecosystem services supporting rural  
42 livelihoods (Briske et al., 2023; McNeeley et al., 2017; Ojima et al., 2015). Dryland systems  
43 support livelihoods associated with ranching, farming, forestry, conservation, recreation, cultural  
44 amenities, and renewable energy production; and are reliant on key ecosystem processes  
45 underlying habitat integrity, biological productivity, water resource quality and quantity,



1 biodiversity, and soil health (White et al. 2000; Lund 2007; Fernández-Giménez et al. 2019;  
2 Briske et al. 2023). Dryland livelihoods and agricultural practices operate as tightly integrated  
3 social ecology systems (SES) (Havstad et al., 2007; Hruska et al., 2017; Mccollum et al., 2017).

4 Changes in climatic drivers, such as rainfall and growing season temperatures, influence  
5 biodiversity and ecosystem processes that include evapotranspiration, soil moisture retention,  
6 biogeochemical cycling (i.e., carbon, nitrogen, and phosphorus), and productivity (Ojima et al.,  
7 2015; Polley et al., 2013). Climate regimes within most dryland systems typically have low mean  
8 and high variability precipitation patterns associated with high evaporative demand (Asner et al.  
9 2004; Zomer et al. 2006; Havstad et al. 2007; Reeves et al. 2014). Shorter-term climate  
10 variability and longer-term climate change in dryland systems are environmental factors that  
11 affect temperature ranges and extremes and the frequency, intensity, form, and duration of  
12 precipitation events (Holmgren et al. 2006; Polley et al. 2013). These changes in critical  
13 ecosystem services act to further constrain the ability of people to meet livelihood needs in  
14 these dryland regions. Concerns about these changes are shared across U.S. public entities  
15 (BLM, USFS, etc.), tribal land managers, and private land managers.

16 Land use practices in drylands have also altered ecosystems and have affected water  
17 availability, further exacerbating stresses to various livelihoods, such as rangeland  
18 management, cropping systems, and tourism (Briske et al. 2023). Extensive systems include  
19 pastoral management of various livestock types ranging from sheep and goats to horses,  
20 camels, and cattle. Agropastoral and dryland cropping practices are common in semiarid  
21 ecosystems supporting small grain cropping such as sorghum and wheat production. Extensive  
22 land use also includes wildlife reserves and conservation areas that provide critical habitats and  
23 other natural resources such as wood products, browse, and forage. In areas where adequate  
24 water resources are available, irrigated cropping systems can be highly productive, though  
25 these systems are vulnerable to rapidly decreasing water resources due to ground water  
26 depletion or stream flow reductions. Recent deployment of renewable energy projects,  
27 especially solar and wind power installations, have increased land competition for rangeland  
28 activities and compromised conservation efforts.

29 Here, we ask: *What are the consequences of changes in drylands for social-ecological systems*  
30 *and what management (e.g., mitigation and adaptation) solutions can maintain the critical*  
31 *services provided by drylands even in the face of change?* This question is addressed through  
32 two sub-themes of land management and adaptation and mitigation.

33

#### 34 *B.1.4.1 Sub-Theme: Land Management*

##### 35 *Motivation*

36 Land-use management in drylands sustain ecosystems and their services, and can safeguard  
37 the future of rural livelihoods, agricultural economics, and communities who depend on  
38 drylands. Managed lands in drylands include rangelands, dryland cropping, as well as rainfed  
39 and irrigated croplands. Drylands also maintain both traditional and alternative energy  
40 production and use, mining, recreation, and forest management, including old growth forests.  
41 Land management thus encompasses a wide diversity of ecosystems, goals, and challenges.  
42 This includes natural resource management of wildlife, endangered species, exotic species,  
43 wildfire, and protection of our most cherished cultural resources.

1

2 Rangelands constitute the largest land use in the world and most of these fall in drylands  
3 (Maestre et al., 2022b). Global rangelands are experiencing significant changes in composition -  
4 notably shrub encroachment and invasive annual grasses - with impacts on forage production  
5 due to current and future climate changes and interactions with wildfire (Boone et al., 2018;  
6 Godde et al., 2020), with significant impacts on livelihoods of more than a billion people  
7 (Maestre et al., 2022b). Vast areas of rangelands in US drylands are managed by BLM (1  
8 million km<sup>2</sup>), USFS (781 000 km<sup>2</sup>), Tribal authorities (283,000 km<sup>2</sup>), and private landowners who  
9 require spatial data on their natural resources for informed decision making. In particular, US  
10 rangelands under BLM management have already recorded significant changes in vegetation  
11 composition and structure during the past 30 years with perennial grasses decreasing and  
12 shrubs increasing (Kleinhesselink et al., 2023). Wildfire is also a major issue for dryland land  
13 managers in the US and around the world and improved options for managing fuels, exotic plant  
14 invasion, and stabilization/restoration after fire is of considerable interest. Indeed, US  
15 rangelands are predicted to change significantly as the climate becomes warmer, drier, and  
16 more variable (Joyce et al., 2013). The ARID field campaign will help us understand the  
17 fundamental science behind the changes, the interactions with diverse land uses and  
18 management strategies, and the ways these could result in improved remote sensing monitoring  
19 capabilities.

20 Extensive land use also includes wildlife reserves, climate refugia, and conservation areas that  
21 provide critical habitat to maintain biodiversity and other natural resources such as wood  
22 products, browse, and forage. For example, numerous Tribal Nations are interested in buffalo  
23 population restoration and management throughout western US drylands. Many dryland  
24 systems also support agropastoral systems where small grain production is interspersed in  
25 rangeland systems.

26 In areas where adequate water resources are available, irrigated and rainfed cropping systems  
27 can be highly productive, though these systems are vulnerable to rapidly decreasing water  
28 resources due to ground water depletion or stream flow reductions. Irrigated cropland systems  
29 constitute the largest consumption of water in drylands worldwide and sustainable water use is  
30 imperative. Recent deployment of renewable energy projects, especially solar and wind power  
31 installations, have increased land competition for rangeland activities and compromised  
32 conservation efforts (Briske et al., 2023) and can also represent high water demand.

33 Land use management of dryland systems is faced with balancing multiple objectives to  
34 maintain biodiversity, land productivity, access to water, and soil health. As management  
35 challenges are compounded by climate change and variability of water availability, land  
36 managers are faced with limited options and information to support land management decision-  
37 making options. Improved ecological forecasting and now-casting of dryland conditions can  
38 greatly help practitioners and managers to deal with emerging stresses to ecosystem services.

39

40 *Questions:*

41 How are land and water resources and resource management being affected by drought and  
42 aridity changes? How is the composition and productivity of rangelands changing under various  
43 management regimes and how are they predicted to change due to climate? How will remote

1 sensing and modeling enhance assessment and forecasting of land resources (e.g., soil  
2 moisture, productivity, water use, carbon storage) for various land uses?  
3 Sub-questions include: How will land use further impair wildlife corridors and migratory  
4 pathways? How will further fragmentation of conservation and rangeland areas affect  
5 biodiversity and conservation goals? How will reduced water availability impact forage  
6 production, cropping yields, livestock conditions, and habitat conditions? How will increased  
7 thermal stress affect agricultural production of rangelands, cropland, forests, livestock, wildlife,  
8 and wildlands systems? How will changes in ecosystem structure affect various land uses  
9 across dryland ecosystems? How will cascading effects related to droughts, fires, and extreme  
10 rainfall events affect land system conditions? Can invasive species be more effectively  
11 managed and how is their extend and effects on ecosystems affected by global change? Can  
12 we improve wildfire management success with fuels treatments, managed fire use, and post-fire  
13 management activities? How will renewable energy deployment affect ecosystem services? Can  
14 multi-sensor observations and modeling improve our ability to assess land management  
15 effectiveness? How can land management support resilience and resistance in dryland  
16 ecosystems?

17

#### 18 *B.1.4.2 Sub-Theme: Adaptation and Mitigation Strategies*

19 As dryland systems continue to be affected by climate and land-use changes the need to  
20 consider adaptation and mitigation strategies for these regions is crucial and growing.  
21 Development of climate resilient management strategies that assess multiple stresses, such as  
22 droughts, extreme heat, erosional flooding, dust storms and fires, are needed to enhance  
23 coping responses, to develop place-based adaptation and mitigation strategies, and to enable  
24 transformative actions when needed. These climate resilient strategies need to incorporate a  
25 SES perspective that links ecosystem services to livelihood strategies and to utilize co-design  
26 and co-production practices to develop useful options and actions. For example, development of  
27 climate resilient management strategies in response to climate change focusing on the onset of  
28 rapid drought events that impact ecosystem services and adaptation of various livelihood  
29 enterprises to climate changes will provide a conceptual example of the needed  
30 transdisciplinary research needed. Providing new insights on how dryland ecosystems are  
31 changing will enable development of enhanced management activities to build resilience and  
32 sustain livelihoods that would contribute to long-term mitigation to climate change practices and  
33 adaptation practices to meet livelihood needs. Within this subtheme, we outline adaptation and  
34 mitigation strategies separately.

#### 35 **Adaptation and resilience of dryland social ecological systems**

##### 36 *Motivation:*

37 As changing conditions in dryland regions affect the delivery and availability of key ecosystem  
38 services, such as forage, browse, fuel wood, grains, habitat, water resources, the need for  
39 adaptation strategies is becoming more urgent. ARID research, in partnership, with various  
40 practitioner and decision maker communities can enhance strategies for adapting to these  
41 changing conditions. For example, social-ecological systems research studies have  
42 incorporated improved drought assessments to enhance drought response strategies co-  
43 developed with natural resource managers (Fernández-Giménez et al., 2019; McNeeley et al.,  
44 2016). These analyses provide greater insight into the range of coping and adaptation choices

1 to make under various levels of adaptive capacity and the importance for co-development  
2 practices to be employed for greater end-user usage. Institutional responses to climate change  
3 are often best suited for dealing with emergency situations and isolated events, rather than for  
4 slower onset, cumulative, or systemic climate-related problems leading to disruption of  
5 ecosystem services. Efforts to enhance resilience of social-ecological systems in these dryland  
6 situations are supporting communities to adapt to climatic changes, such as flash-drought,  
7 increased fire events, loss of land productivity by engaging with operators at the local scale to  
8 enhance delivery of more nuanced information needed to deal with these climate stresses given  
9 local adaptive capacity. ARID research observations provided if this field campaign is deployed  
10 will provide additional climate information and analytical tools to better inform and enable natural  
11 resource managers to cope with water shortages, droughts, and extreme events.

12 *Questions:*

13 Key questions related to adaptation and resilience of dryland social-ecological systems: How  
14 can dryland ecosystems enhance their resilience in the face of environmental stressors, and  
15 what adaptive strategies can local communities implement to build resilience? Several sub-  
16 questions include: What factors and information can enhance adaptive capacity of dryland  
17 social-ecological systems to meet the challenges to increased aridity? How can ecological  
18 forecasting of drought conditions enable decision making of dryland operators and natural  
19 resource managers? Have land managers needed to change their strategies in drylands in the  
20 past few decades to accommodate change? What drives these changes?

21

## 22 **Mitigation options for different dryland livelihood strategies**

23 *Motivation:*

24 Dryland systems are being evaluated for enhancing carbon sequestration, and establishing  
25 renewable energy sites, such as solar and wind. The land use practices when integrated into  
26 livelihood goals and strategies can enhance the socio-economic resilience of many  
27 communities. Considerations of impacts on wildlife habitat, water availability, and further  
28 fragmentation of landscapes should be addressed in permitting and siting of large-scale  
29 renewable energy facilities. Various natural resource management approaches need further  
30 development to take advantage of these potentials and to contribute to various livelihood  
31 systems, while not undermining cultural norms and needs.

32 Likewise, mitigation practices that enhance carbon stock stability and reduce carbon losses due  
33 to reduced primary productivity or losses due to disturbances such as fires and erosion of soils  
34 should benefit from the research conducted above. Including renewable energy siting with  
35 greater understanding of extreme events and pulse events will improve long-term suitability of  
36 renewable energy facilities.

37 *Questions:*

38 Related to human dimension issues in drylands: How will climate variability and water limitations  
39 hinder carbon sequestration efforts? What impact will renewable energy development and  
40 carbon sequestration practices have on dryland livelihood strategies and ecosystem services in  
41 different dryland regions? Sub questions include: How can improved observations enhance our  
42 assessment of renewable energy systems and carbon sequestration practices? What increased

1 information can be provided related to changes in ecosystem processes and ecosystem  
2 services related to mitigation practices?

### 3 *B.1.4.3 Research Approaches in Support of Social Ecological Systems*

4 Developing research efforts to address climate change impacts on dryland management  
5 practices, the research community will need to assess various land management activities from  
6 extensive practices associated with ranching and conservation efforts to more intensive levels  
7 associated with irrigated agriculture and other cropping systems. Land use practices in these  
8 dryland regions are determined by water availability. Sources of surface water and groundwater  
9 sources will often dictate the type of land use practices feasible. Surface water sources are  
10 critical for livestock and wildlife management, whereas sources of groundwater may allow for  
11 more intensive agricultural systems including irrigated farming. Connections to water sources,  
12 such as mountain snowmelt and runoff, that allow for recharge of the aquifers and surface flows  
13 are also important resources that are undergoing changes. These factors determine how  
14 effective land management practices can be, especially as emerging practices are needed to  
15 develop adaptation and mitigation options.

16 ARID research efforts related to land management practices will rely heavily on the research  
17 findings of the science foci described above, especially as enhanced understanding emerges  
18 from how climate variability will affect critical ecosystem services, such as water availability,  
19 carbon fluxes, ecosystem structure and biodiversity. Through these research efforts connections  
20 to land use management practices will be aided by integrative modeling and observations  
21 efforts to support assessment of current trends and projections of changes affecting ecosystem  
22 services critical to land management decision making.

23 Coordinated research efforts with natural resource agencies and other organizations will be  
24 needed to co-develop usable products for dryland resource managers, across public, tribal, and  
25 private entities. ARID research described above has the potential to support land management  
26 decision making by improved forecasts of drought events, soil moisture levels, forage and  
27 browse availability, land cover and land use patterns, plant functional types, fire monitoring, and  
28 ecosystem forecasts. These improvements can enhance the ability to develop integrated model-  
29 observation system to support now-casting to forecasting of cropland and forage production  
30 (see Grass-Cast model - USDA/ARS); assessment of landscape features that differentiate  
31 between active vs non-photosynthetically active vegetation using SiF technologies; integrating  
32 soil moisture and actual ET assessments to enable better forage and crop yield estimates; and  
33 characterizing the vulnerability of different agricultural practices and other land uses to changes  
34 in aridity. This information can be used to guide observational protocol to provide better  
35 information on the timing, duration, and intensity of drought conditions in different dryland  
36 regions of the world.

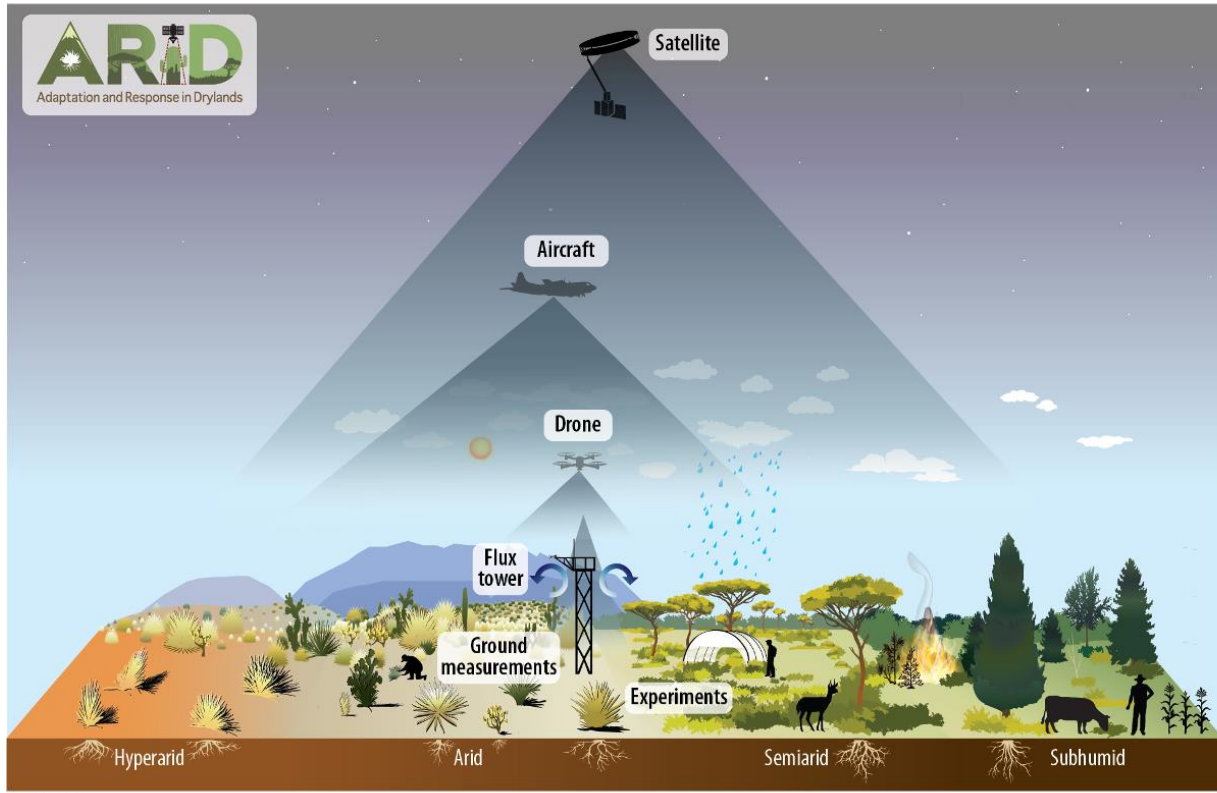
37 These advancements from better understanding of climate variability and water availability  
38 dynamics on ecosystem services will aid development of adaptation strategies and assess the  
39 effectiveness of mitigation strategies associated with renewable energy deployment and carbon  
40 sequestration activities.

41

1 **B.2 Overall Study Design**

2 In order to address the critical fundamental and applied sciences questions and observation  
3 gaps in Section B.1, we have designed an interdisciplinary, multi-scale ARID field campaign  
4 based on our scoping study. The field campaign will leverage existing assets and provide new  
5 field and aircraft measurements to scale from point to pixel to plant across field to aircraft to  
6 satellite instruments (Fig. B.6). The study domain will span diverse drylands from hyperarid to  
7 subhumid, in the US and internationally. The approach includes measurements across spatial,  
8 temporal, and spectral axes to observe drylands' highly spatiotemporally heterogeneous  
9 landscapes. It also includes methods to inform and be informed by process modeling (Section  
10 B.2.2).

11  
12 We first discuss approaches to address challenges with remote sensing, modeling, and ground  
13 observations in drylands in Sections B.2.1 to B.2.3. We emphasize the role of the proposed  
14 ARID campaign to address observation gaps. These approaches are general and can be  
15 applied across domains and focal areas. Next, in Section B.2.4., we motivate our main dryland  
16 domain across four focal study areas within the US, as well as an additional four international  
17 sites that include northern Mexico, drylands across southern Africa and Australia, and the  
18 Cerrado, Caatinga, and Gran Chaco in Brazil. We discuss specific existing field and aircraft  
19 assets we will leverage within these domains in Section B.2.5. We develop our field sampling  
20 strategy within these domains in Section B.2.6 using both new measurements and the existing  
21 assets. Finally, we discuss our proposed applied science activities in alignment with Earth  
22 Science to Action strategy in working with end-users to accomplish land management and  
23 mitigation goals.  
24



\*Not to scale

AR003

1  
2 **Figure B.6.** ARID will include a nested approach of coordinating measurements  
3 in the field, existing and new networks, drones, and aircraft to inform satellite  
4 measurements and improve understanding of drylands alongside NASA satellite  
5 measurements. Figure B.10 shows how these measurements integrate into and  
6 are informed by dryland modeling efforts.  
7

8 **B.2.1 Remote Sensing of Drylands and the critical role of NASA sensors**

9 Drylands are characterized by high spatial heterogeneity and high temporal variability which  
10 pose unique remote sensing challenges (outlined below). This demands fine resolution and  
11 frequent observations brought about by acquisitions at multiple scales, from multiple sensors  
12 and modalities (SAR, Hyperspectral, thermal) (W. K. Smith et al., 2019b). *NASA's airborne*  
13 *sensors along with the UAS and proximal sensing proposed by ARID will fill observation gaps of*  
14 *physical parameters at the scales required to address the science questions and applications.*  
15 For example, the recent Surface Biology and Geology (SBG) High-Frequency Time Series  
16 (SHIFT), an airborne and field campaign which took place in a California dryland, is  
17 investigating how the frequent acquisition of imaging spectrometer data within a growing season  
18 reveals the progression of ecosystem function (Chadwick et al., in revision, 2020a) by  
19 characterizing leaf density, leaf chemistry, plant health, vegetation type and other plant  
20 attributes, which are often conflated by multispectral proxies for greenness (Myneni et al.,  
21 1995).

1  
2 The historical remote sensing literature on dryland ecosystems, especially on the topic of land  
3 degradation, is dominated by the time-series analysis of trends in vegetation greenness  
4 (Normalized Difference Vegetation Index - NDVI) (K. J. Wessels et al., 2012), with no  
5 differentiation of trees, shrubs, grasses, or biocrust. Such analyses cannot therefore explain the  
6 ecological causes of observed trends in greenness leading to speculative conclusions.  
7 Contradictory interpretations of “browning and greening” trends have been controversial in the  
8 context of desertification/land degradation monitoring (Fensholt et al., 2012; Prince et al., 2007).  
9 In contrast, the ARID campaign will use current and emerging airborne, UAS, proximal, and  
10 space-based sensors that can fully characterize dryland ecosystems to understand changes in  
11 their structure, composition and function, as well as soil moisture and evapotranspiration,  
12 extending well beyond only the focus of greenness and its trends. Moreover, accurate  
13 estimation of fractional cover of plant functional types (PFTs) at the appropriate scale are  
14 fundamental for achieving accurate carbon, water, and energy fluxes from process-based land  
15 surface and dynamic global vegetation models (Hartley et al., 2017a). *This will require novel*  
16 *combinations of UAS and airborne VSWIR imaging spectrometers, multispectral infrared*  
17 *sensors, LiDAR, and SAR sensing within a robust scaling framework.*  
18

19 Addressing the ARID science questions will therefore require a fully integrated and coordinated  
20 approach that scales-up observations and process models “from leaf to orbit”. This section of  
21 the white paper focuses on remote sensing and will (i) demonstrate the critical role of NASA  
22 remote sensing data in estimating parameters required to address the ARID science questions,  
23 (ii) demonstrate the need for airborne, UAS and in-situ data to sense the required physical  
24 parameters and processes at the appropriate higher spatial and temporal scales, (iii) emphasize  
25 the need for a multi-scale, multi-sensor, multi-modality approach, (iv) highlight opportunities for  
26 novel approaches using proximal sensors on towers and UAS, and (v) highlight the role of ARID  
27 in the calibration and validation of remote sensing algorithms to utilize new and future NASA  
28 missions.  
29

### 30 *B.2.1.1 The challenges of remote sensing in Drylands*

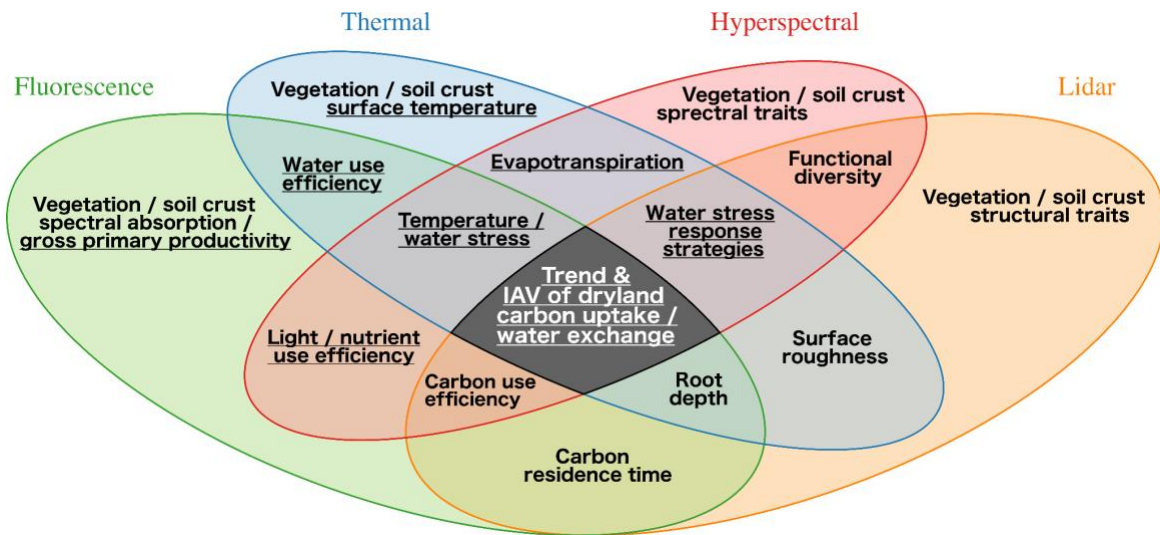
31 Drylands have consistently played impactful roles in remote sensing development. Due to lower  
32 cloud cover and higher likelihoods of favorable atmospheric conditions, many remote sensing  
33 techniques were developed in drylands, including early retrievals of surface reflectance and  
34 estimates of vegetation condition using red and near-infrared bands, which would later become  
35 the Normalized Difference Vegetation Index (NDVI) (Tucker, 1979). Despite this important role,  
36 our remote sensing capacity for core measurements of ecosystem structure, composition, and  
37 function (e.g., productivity, water exchange, etc.) in drylands show severe deficiencies  
38 (Biederman et al., 2017b; W. K. Smith et al., 2019b). We highlight the fundamental challenges  
39 of dryland remote sensing below:

- 40 1. The characterization of dryland composition and function is difficult without very high-  
41 resolution imaging spectrometer data due to dynamic mixtures of mineral soil, non-  
42 vascular biocrusts, C3 and C4 annual grass and forbs, C3 and C4 perennial grass and  
43 forbs, C3 trees and shrubs, and CAM succulents and cacti (Fig. A.2).



- 1 2. Traditional remote sensing-based ET and GPP products rely on parameters derived  
2 from optical reflectance such as FAPAR, which captures changes in vegetation  
3 greenness and photosynthetic capacity and not photosynthesis efficiency. Low  
4 confidence in the accuracy of dryland GPP and ET products derived from remotely  
5 sensed data limits their use in understanding broad spatial patterns in dryland  
6 ecosystem function, as well as their use for benchmarking process-based models (see  
7 section B.2.2).
- 8 3. The characterization of dryland structure is difficult without very high resolution optical or  
9 LiDAR data due to low density and low-stature vegetation (shrubs and grass) and  
10 horizontally and vertically complex tree structure and cover.
- 11 4. Episodic vegetation activity due to irregular and patchy rainfall is not captured frequently  
12 enough by many remote sensing platforms, making it difficult to characterize temporal  
13 variability in key structural and functional traits.
- 14 5. High exposure of soils and varied geologic substrates often confounds our ability to  
15 retrieve biological contributions to spectral reflectance.
- 16 6. Photosynthetic soils (biocrusts) can significantly contribute to carbon, water, and nutrient  
17 flux in drylands, but these contributions are poorly known since biocrusts are intermixed  
18 with both the mineral soil and vascular plants at fine spatial scales.
- 19 7. Soil moisture estimates do not capture high spatial-temporal variability required to model  
20 water fluxes.

21  
22 Drylands demand a fine and frequent observation approach using a multi-scale, multi-platform,  
23 and multi-sensor framework (Fig. B.7 Parameters and modalities, (W. K. Smith et al., 2019b).  
24 *The challenges of dryland remote sensing can be addressed by leveraging the ever-expanding*  
25 *array of current and future space-based NASA missions given reliable calibration and validation*  
26 *with in-situ, UAS, and airborne measurements planned for the ARID campaign (Fig. B.7). The*  
27 *multi-scale ARID campaign will also enable the development of improved algorithms and*  
28 *models for estimating essential physical parameters and reliable higher-level products (e.g.*  
29 *GPP, ET) that can be applied globally to more accurately account for drought impacts on*  
30 *dryland ecosystem processes and structure, biodiversity, and the global water and carbon cycle.*  
31



1 **Figure B.7.** Parameters and modalities. Note this figure does not include  
2 Microwave sensors which provide additional information related to soil moisture,  
3 ecosystem structure, and vegetation water content. This figure is used with  
4 permission from (W. K. Smith et al., 2019b).  
5

#### 6 B.2.1.2 Physical parameters for science questions and applications

7 The ARID science theme working groups listed physical parameters that are required to  
8 address science questions within each science theme and sub-theme. These are summarized  
9 in a modified Science and Application Traceability Matrix (SATM) along with the potential  
10 sensors on proximal (e.g. towers), UAS, airborne, and space-based satellite platforms  
11 (Appendix F.1). Table B.2 is a summary of the ARID SATM, listing the physical parameters  
12 most frequently required by the sub-theme science questions (indicated by subtheme numbers,  
13 Table B.1). Successfully observing these parameters during the ARID campaign at the  
14 appropriate resolutions would contribute to substantial advancement of dryland science across  
15 many focal areas. Some parameters have established remote sensing sensors and techniques  
16 (e.g. vegetation fractional cover), while others, such as, soil carbon and belowground biomass  
17 remain a significant challenge. Parameters such as GPP and ET are remote sensing-based  
18 model outputs requiring remotely sensed input parameters including chlorophyll fluorescence,  
19 surface temperature, precipitation, soil moisture, and PFT fractional cover (also included  
20 elsewhere in the SATM).  
21

22 The estimation of most physical parameters requires multiple sensors on multiple platforms (Fig.  
23 B.6). Key to the scaling-up process are (i) proximal remote sensing instruments installed at field  
24 sites capable of provide near-continuous, high spatial resolution measurements; (ii) UASs that  
25 augment the field measurements and provide reference data beyond the extent of field plots; (iii)  
26 airborne NASA sensors covering larger areas for higher level product generation with fine  
27 spatial resolution; and (iv) satellite platforms that provide measurements that can be validated,  
28 trained, and integrated into models to provide global coverage of key physical parameters (Fig.  
29 B.7). The NASA drylands field campaign will enable an effective scaling-up process and  
30 uncertainty estimation to unlock the full potential of remote sensing data to address critical  
31 science questions. **Below we highlight some of the critical physical parameters required to  
32 address ARID science questions and how the ARID field campaign will improve their  
33 estimation.**  
34

35 **Table B.2.** Select rows of the ARID Science and Application Traceability Matrix  
36 (for full SATM, see Appendix F.1). Selected physical parameters required to  
37 address science sub-themes (numbered in footnotes) and the remote sensing  
38 sensors available on proximal (e.g. towers), UAS, airborne and spaceborne  
39 platforms. Names of sensors are given in the footnotes of this table. The table  
40 lists the current and future sensors that can make the most significant advances  
41 and does not list all available options. Science Sub-themes: 1.1 Water  
42 availability, 1.2 Dryland climate variability: Pulses and Drought, 1.3 Fire, 1.4  
43 Land-Atmosphere interactions, 2.1 Vegetation structure and Heterogeneity, 2.2  
44 Biodiversity, 2.3. Ecosystem Function, 2.4 Dryland Geology and Soil Processes,

1  
2

3.1 Carbon stocks and fluxes, 4.1 Land Management, 4.2 Adaptation and Mitigation.

Physical Parameter or process / Observable	Science theme number	Comment: need and current status	Near surface and Proximal	UAS	Airborne	Spaceborne/ Satellite
Soil Moisture (SM)	1.1, 1.2, 2.1, 2.2, 2.3, 2.4, 3.1	Poorly estimated for the subsurface/root zone; identified need for higher resolution (10 m to <1 km) estimates Driver for local models Validation/diagnostic for water cycle in LSMs/ESMs	Soil moisture network, GNSS receiver above and below canopy	Thermal	UAVSAR / AirMoss, SMAPEX; NOAA airborne gamma NOHRSC, AirSWOT,	SMAP, ECOSTRESS, Hydrosat, SBG, SMOS, NISAR, Sentinel1,
Evapotranspiration (ET) Level 4 model output	1.1, 1.2, 1.3, 2.1, 2.2, 3.1	Calibration/Validation for model estimates of ET, E and T. Open ET models can use multiple RS inputs	Infrared Thermometer (IRT): Apogee SI-111-SS Thermal Camera: FLIR A700f; ICI P-Series	Thermal	HYTES, MASTER	ECOSTRESS, Landsat, Hydrosat, SBG
Surface temperature (ST) of soil vs. plants	1.2, 1.4, 2.3, 3.1	Input to ET, GPP, NEE, plant stress	Infrared Thermometer (IRT): Apogee SI-111-SS Thermal Camera: FLIR A700f; ICI P-Series	Thermal	HYTES, MASTER	ECOSTRESS, Landsat, Hydrosat, SBG
Vegetation fractional cover of Plant functional types (PFT)	1.3, 2.1, 2.2, 2.3, 3.1	PFT: Trees, shrubs, grasses, forbs, biocrust. Grass biomass poorly estimated. Local calibration/ initialization for models DGVM constraint Trait information	PhenoCam; Terrestrial Laser Scanning (TLS)	Multi-spectral Lidar	AVIRIS-NG, NEON-AOP, G-LiHT, LVIS	Landsat, SBG, EMIT
Vegetation height: Grass, shrubs, trees	1.3, 2.1, 2.2, 3.1	Shrub (<3m) biomass poorly estimated with optical and space-based LiDAR algorithms. Individual tree heights can be extracted using SfM from airborne imagery. Potential constraint for data assimilation	Terrestrial Laser Scanning (TLS)	Structure from motion (SfM); Lidar	Discrete Lidar, G-LiHT, LVIS, UAVSAR, AVIRIS-NG, NEON-AOP	NISAR, S1, GEDI, IceSat2, WorldView,
Above ground biomass: grass	1.3, 2.1, 2.3, 2.2, 3.1	Grass biomass poorly estimated with optical	Terrestrial Laser Scanning (TLS)	Multi-spectral; Lidar	AVIRIS-NG, G-LiHT, NEON-AOP	EMIT, Landsat NEXT, HLS, SBG,
Above ground biomass: shrubs and trees	1.3, 2.1, 2.2, 3.1	Shrub biomass is poorly estimated with optical and space-based LiDAR algorithms. Calibration/Validation for model estimates of NPP Potential constraint for data assimilation	Terrestrial Laser Scanning (TLS)	Lidar	Discrete Lidar, G-LiHT, LVIS, UAVSAR, AVIRIS-NG, NEON-AOP	NISAR, S1, GEDI, IceSat2, EDGE, STV

Photosynthetically active vegetation fractions of PFT's	1.3,2.3, 3.1	Essential to make a distinction between active and non-active components	Phenocams, spectrometer (e.g., Ocean Insight FLAME VNIR spectroradiometer)	Hyperspectral, Multi-spectral	AVIRIS-NG, LVIS	EMIT, EnMAP, SBG, Landsat
Vegetation moisture content	1.3, 2.3		TLS; dual GNSS receivers installed above and below the canopy	Hyperspectral, Multi-spectral, SAR	AVIRIS-NG, UAVSAR	S1, NISAR, EMIT, EnMAP, SBG, Landsat 8,9,10, NEXT, HLS, PACE
Invasive species: tree, grass, shrub, forbs	1.3, 2.1, 2.2	Individual Invasive plants species could be spectrally distinguished	Hyperspectral: ASD FieldSpec; Spectra Evolution RS-3500, 8800, PSR+	Hyperspectral	AVIRIS-NG	EMIT, EnMAP, SBG,
Solar-induced Fluorescence (SIF)	1.2, 2.3, 3.1		Ultra-hyperspectral VNIR: PhotoSpec; FluoroSpec; FloX Box	Ultra-hyperspectral VNIR	FIREFLY, CFIS	OCO-2,3, TEMPO
Fraction of photosynthetically active radiation absorbed (FAPAR) per PFT	2.3, 3.1	Input to carbon-flux models	Hyperspectral, Multi-spectral	Hyperspectral, Multi-spectral	AVIRIS-NG	EMIT, EnMAP, SBG, GOES-R
Gross Primary Production: Trees, grass, shrubs. Level 4 model output	1.2, 1.4, 2.3, 3.1	Modeled variable, require FAPAR of PFT and meteorological variables as input	Hyperspectral; Thermal; SIF	FAPAR of PFTs derived above	FAPAR of PFTs derived above	FAPAR of PFTs derived above, GOES-R

1 Abbreviations: S1: Sentinel1, SBG: Surface Biology Geology, AirMOSS: Airborne Microwave Observatory of Subcanopy and  
2 Subsurface, UAS-LS: Uncrewed Airborne System - Laser Scanner, FIREFLY: fluorescence imaging of red and far-red light yield,  
3 CFIS: Chlorophyll Fluorescence Imaging Spectrometer, FireSense is multi-band microwave, visible to thermal infrared observation  
4 program,, UAVSAR: Uninhabited aerial vehicle synthetic aperture radar, SMAP: Soil Moisture Active Passive, SMAPEX: SMAP  
5 Experiments, NOHRSC: National Operational Hydrologic Remote Sensing Center, AirSWOT: Air Surface Water and Ocean  
6 Topography, HyTES: Hyperspectral Thermal Emission Spectrometer, MASTER: MODIS/ASTER airborne simulator, AVIRIS-NG:  
7 Airborne Visible-Infrared Imaging Spectrometer - Next Generation, NEON-AOP: National Ecological Observatory Network-Airborne  
8 Observation Platform, G-LiHT: Goddard's LiDAR, Hyperspectral & Thermal Imager, LVIS: Land, Vegetation, and Ice Sensor,  
9 ECOSTRESS: Ecosystem Spaceborne Thermal Radiometer Experiment on the International Space Station, SMOS: Soil Moisture  
10 and Ocean Salinity, NISAR: NASA-ISRO SAR, EMIT: Earth Surface Mineral Dust Source Investigation, GEDI: Global Ecosystem  
11 Dynamic Investigation, STV: Surface Topography and Vegetation EnMAP: Environmental Mapping and Analysis Program, GOES:  
12 Geostationary Operational Environmental Satellite

13

#### 14 *Carbon stocks and fluxes partitioned among plant functional types (PFT)*

15 The reliable estimation of carbon stocks and fluxes, require their partitioning among plant  
16 functional types (PFT) including annual grasses and forbs, perennial grasses and forbs, shrubs,  
17 trees, biocrust and litter, as well as their photosynthetically active vs. inactive (e.g. litter)  
18 fractions. Accurate estimation of fractional cover of PFT at the appropriate scale is fundamental  
19 to parameterizing and reducing the uncertainty of carbon (GPP) and water (ET) flux models  
20 (Hartley et al., 2017a). In particular, assessing changes in the fractional cover and biomass of  
21 woody shrubs versus grass is essential for monitoring shrub encroachment in the US  
22 (Kleinhesselink et al., 2023) and across global rangelands, driven by a combination of CO<sub>2</sub>  
23 fertilization, changing rainfall frequency and intensity, and overgrazing (Stevens et al., 2017).

1 Aboveground woody biomass and carbon stocks have been poorly estimated for drylands by  
2 traditional remote sensing products originally developed for forests (Bastin et al., 2017).  
3 However, recent advances in AI, specifically deep learning, to extract individual trees from very  
4 high-resolution drone and airborne imagery, as well as continent-wide commercial satellite  
5 imagery (e.g. Maxar's WorldView and PlanetScope/LANET) are enabling the estimation of  
6 carbon in dryland trees (Tucker et al., 2023) (Fig. B.5). ARID's integration of airborne and UAS  
7 measurements can serve as training data to develop area-specific convolutional neural network  
8 models that will automatically extract individual trees (and show promise for shrubs) to estimate  
9 carbon density and its changes using commercial satellite imagery. Aircraft- and UAS-based  
10 assessment of ecosystem dynamics allow a more detailed spatial view than traditional satellites  
11 (Pervin et al., 2022; Villarreal et al., in review) and has shown utility for algorithmic development,  
12 particularly when paired with coordinated field campaigns (Chadwick et al., 2020). In a  
13 multiscale, multi-step process, PFT fraction can first be estimated with UAS-based LiDAR (ULS)  
14 combined with very-high resolution (5cm GSD) multi-spectral or imaging spectrometer data,  
15 assisted by AI and advanced image processing (Sankey et al., 2018). The UAS-based biomass  
16 and fractional cover estimates of the PFT's will be used as reference data to develop models  
17 that predict these variables using airborne imaging spectrometer (AVIRIS) and LiDAR data  
18 (LVIS and/or discrete LiDAR from NEON AOP) acquired contemporaneously over the UAS data  
19 and across the larger study sites. The airborne-derived estimates of biomass and fractional  
20 cover of the PFT's in turn serve as indispensable and reliable calibration and validation for  
21 model development to predict these variables with space-based imaging spectrometer data  
22 (e.g. EMIT, SBG) and SAR data (Sentinel 1, ALS02/4 or NISAR). *The ARID field campaign will*  
23 *significantly improve the estimation of fine-scale fractional cover and biomass for different*  
24 *dryland PFTs which will transform the modeling of carbon fluxes (GPP and NPP) and stocks in*  
25 *drylands* (Hartley et al., 2017b; M. O. Jones et al., 2021; Kleinhesselink et al., 2023; Robinson  
26 et al., 2019; Shi et al., 2022).

27  
28 Existing rangeland monitoring applications by partners BLM (e.g. [Rangeland Analysis Platform](#),  
29 [Climate Engine](#)) and USGS ([Rangeland Condition Monitoring Assessment and Projection](#),  
30 [RCMAP](#)) currently use estimates of PFT from Landsat to track rangeland condition (M. O. Jones  
31 et al., 2021; Kleinhesselink et al., 2023; Robinson et al., 2019; Shi et al., 2022). More accurate  
32 estimations of 16-day PFT and grass biomass with Landsat or SBG can feed directly into  
33 existing applications. In this manner, the ARID field campaign can revolutionize the PFT  
34 fractional cover products that are already embedded in decision making workflows of agencies  
35 and tribal authorities.

36  
37 *Soil moisture (SM) and vegetation water content (VWC)*  
38 Quantifying the amount of water stored within dryland soils, at plot-to-regional scales, is one of  
39 the most important aspects of ecosystem analysis and prediction, yet remains a considerable  
40 challenge for remote sensing and modeling communities. The NASA SMAP mission currently  
41 offers near daily retrievals of soil moisture, which is appropriate for evaluating pulse dynamics. It  
42 has two advantages in drylands, including lower retrieval error because of lack of an overlying  
43 vegetation layer and scattering (Chan et al. 2016, Colliander et al. 2019), as well as a deeper  
44 sensing depth due to drier soils and deeper emission (Moghaddam, Saatchi et al ; Feldman et

1 al. 2023). However, its spatial resolution is particularly limited at 30km. It is therefore essential  
2 to leverage aircraft measurements, such as from NASA Scanning L-band Active/Passive  
3 (SLAP), flights to retrieve soil moisture over 100m scales to quantify its spatial distribution in  
4 relation to drylands' heterogeneous vegetation distributions, elevation, and proximity to other  
5 water sources. Such aircraft measurements can be co-located with soil moisture networks, such  
6 as measurements at Mesonet, NEON, and/or AmeriFlux stations, which allows validation and  
7 further understanding of soil moisture spatial dynamics. Furthermore, these measurements tend  
8 to be limited to the upper soil layers and new measurements of deeper soil layers should be  
9 explored in the field. Terrestrial water storage (TWS) provided by the NASA Gravity Recovery  
10 and Climate Experiment (GRACE) can provide deeper soil moisture information. GRACE  
11 indicates especially strong TWS losses in drylands resulting from climate change coupled with  
12 direct human activity including land cover changes and unsustainable irrigation practices (L. An  
13 et al., 2021). However, the vertical representation of TWS is uncertain and low spatial resolution  
14 can be prohibitive, thus requiring use of soil moisture stations with measurements at a range of  
15 depths (Rodell et al. 2001). NISAR will also provide a soil moisture product that will be  
16 calibrated and used in the ARID campaign.

17  
18 Vegetation water content information can additionally be retrieved through microwave active  
19 and passive measurements such as from NISAR and SMAP, especially through the retrieved  
20 parameter vegetation optical depth (VOD) (Konings et al., 2019). These VWC retrievals  
21 experience some of the same challenges as soil moisture, with the addition that it is unclear  
22 how much sparse vegetation plays a role in the VOD retrieval. Such challenges can be tested  
23 with GNSS sensors and airborne measurements from SLAP that have finer resolution  
24 (Feldman, 2024a).

25  
26 *Gross Primary Productivity (GPP), Evapotranspiration (ET) and Water Use Efficiency (WUE)*  
27 GPP, ET, and WUE are key processes that link the water, carbon, and energy cycles, especially  
28 in drylands where water is the dominant constraint on vegetation productivity (Fisher et al.,  
29 2017). These parameters are key for addressing numerous ARID science questions related to  
30 the trends and interannual variability in carbon uptake, water availability along the soil-plant-  
31 atmosphere continuum (SPAC), rangeland resource availability, crop productivity and water  
32 stress, and dryland responses to drought and heatwave events (Table B.1). Satellite remote  
33 sensing is unmatched in spatiotemporal coverage and has been used to estimate GPP, ET, and  
34 WUE dynamics across spatial scales back to the early 1980's (Fisher et al., 2017; Ryu et al.,  
35 2019). These estimates have helped shape our current understanding of dryland function within  
36 the context of the broader Earth system (Ahlström et al., 2015; Humphrey et al., 2018; Poulter et  
37 al., 2014). However, these key ecosystem functions cannot be directly observed and thus rely  
38 heavily on physical parameters derived from the optical domain of the electromagnetic  
39 spectrum, such as the fraction of absorbed photosynthetically active radiation (FAPAR), which  
40 provide information on vegetation state (e.g., photosynthetic capacity), but not physiological  
41 function (e.g., photosynthetic activity). Operational GPP and ET products based on these data  
42 are widely used and include the NASA MODIS GPP and ET products, and more recently the  
43 NASA SMAP Level 4 C products. Although the spatial variability of these products is generally  
44 consistent with field measurements, their temporal variability suffers from substantial

1 uncertainties and bias, particularly across dryland ecosystems (Biederman et al., 2017a; Verma  
2 et al., 2014). For example, across the drylands of southwestern North America the MODIS ET  
3 and GPP products were only able to account for 20–30% of the interannual variation in  
4 measured GPP (Biederman et al., 2017a).

5  
6 Co-location of NASA instruments, including the NASA OCO-3 and ECOSTRESS missions, offers  
7 an exciting new opportunity to overcome past limitations, by providing first-time insights into the  
8 highly dynamic nature of solar-induced fluorescence (SIF) and land surface temperature (LST) –  
9 factors closely related to GPP and ET. By providing information directly related to vegetation  
10 physiological function, SIF and LST offer a significant advance relative to traditional reflectance-  
11 based parameters such as FAPAR (Farella et al., 2022; Y. Sun, Gu, et al., 2023; Y. Sun, Wen, et  
12 al., 2023). SIF is a direct measure of the light re-emitted from chlorophyll during photosynthesis  
13 and thus represents a direct proxy for photosynthetic activity (Y. Sun, Gu, et al., 2023; Y. Sun,  
14 Wen, et al., 2023). LST can be integrated with measurements of air temperature to infer changes  
15 in stomatal conductance and rates of canopy transpiration, processes fundamental to both GPP  
16 and ET (Fisher et al., 2017; Volk et al., 2024). Thermal sensors have been included on satellite  
17 platforms for decades, and LST has been used as a critical input to a variety of models including  
18 semi-empirical / physical or surface energy balance models that estimate ET, e.g., in the Open  
19 ET application managed by USDA (Volk et al., 2024). While these past missions continue to  
20 provide invaluable data, we need complementary sub-daily observations since SIF and LST vary  
21 at high temporal frequency – especially across drylands – in response to high temporal frequency  
22 changes in biophysical factors.

23  
24 SIF and LST measurements from towers, airborne, and space-based platforms have tremendous  
25 potential for more direct measurements of dryland physiological function. Drylands are an ideal  
26 testbed to further explore SIF-GPP and LST-ET relationships across spatial and temporal scales.  
27 SIF can be measured near-continuously from towers, regularly with VSWIR imaging  
28 spectrometers on UAS and with FIREFLY and CFIS sensors on aircrafts, and at multiple times  
29 per day with OCO-2, OCO-3, and the geostationary TEMPO platform (Table B.2). LST can also  
30 be measured near-continuously from towers, regularly with thermal imagers on UAS and with  
31 MASTER and HyTES sensor on aircrafts, and multiple times per day from space with the  
32 ECOSTRESS, SBG, and the HydroSat platforms (Table B.2). The integration of these cutting-  
33 edge data can help to revolutionize the accuracy of space-based GPP, LST, and WUE estimates  
34 across various spatiotemporal scales required to address urgent dryland science questions.

### 35 *B.2.1.3 Proximal Sensing*

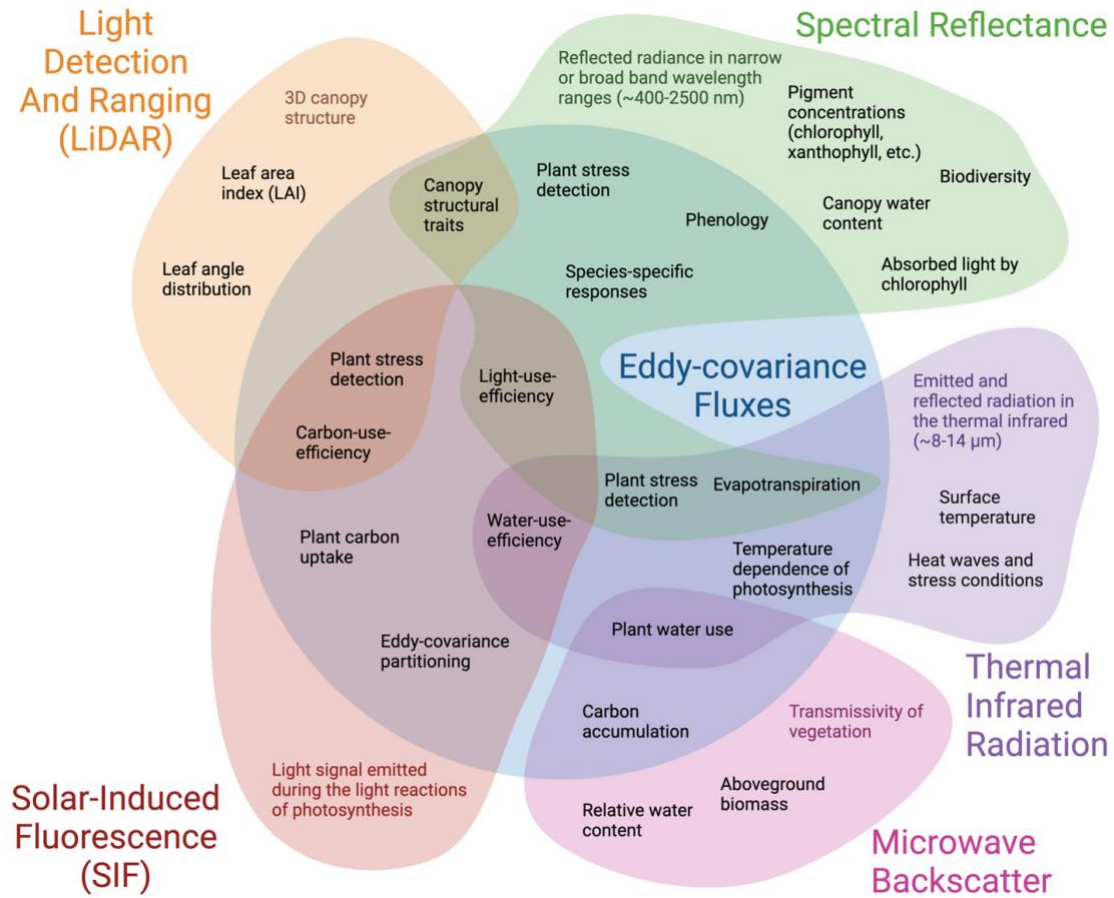
37 Proximal sensing, or fixed plot to site scale sensors, provides high spatial, temporal, and  
38 spectral resolution measurements enabling a more direct link to dryland field measurements  
39 including flux measurements from eddy-covariance towers (Fig. B.8) (Gamon, 2015; Schimel et  
40 al., 2019). Proximal sensing can be used to validate satellite measurements and products (Y.  
41 Zhang et al., 2023), drive and test the representation of ecosystem processes in models  
42 (Magney et al., 2019; Raczka et al., 2019), reveal spatial heterogeneity in fluxes at the field level  
43 (Javadian et al., 2022, 2024; Magney et al., 2019) and scale field-level measurements to the  
44 landscape as observed by spaceborne instruments (Farella et al., 2022) thereby providing

1 unprecedented insights into the processes that drive ecosystem fluxes (Pierrat, Z. et al., in  
2 review). Proximal sensing will be a critical component of the ARID campaign since we plan to  
3 deploy a suite of sensors across existing flux tower networks to fully link remote sensing and  
4 flux measurements in space and time and enable mechanistic insights into the link between  
5 remote sensing signals and biologic processes (Gamon, 2015; Schimel et al., 2019).

6  
7 Specifically, as a part of the ARID field campaign, we propose five different types of proximal  
8 sensing that cover ecosystem structure, composition, and function: visible-to-shortwave infrared  
9 spectral reflectance (VSWIR), solar-induced chlorophyll fluorescence (SIF), thermal infrared  
10 emittance (TIR), microwave backscatter (WVB), and light detection and ranging (LiDAR) (Table  
11 parameters). Early efforts have effectively deployed: 1) proximal SIF and LiDAR measurements  
12 to show the direct connection between SIF and GPP at the AmeriFlux Niwot Ridge dry mixed  
13 conifer flux site (Magney et al., 2019); and 2) proximal thermal and structure from motion (SfM)  
14 measurements to show the direct connection between canopy temperature and water stress at  
15 the AmeriFlux Mount Bigelow dry mixed conifer flux site (Javadian et al., 2022, 2024). ARID  
16 also intends to deploy mobile, low-cost towers that combine the aforementioned spectral  
17 sensors and full suite of meteorological and eddy covariance (Wong et al., 2023a)  
18 measurements to capture pulse dynamics or experimental treatments. Such mobile towers will  
19 provide significant opportunities for innovation. These data have the potential to resolve three  
20 key questions: What is the scale dependence (spectral, spatial, temporal) of dryland ecological  
21 processes and fluxes?; What are the underlying physical and biological drivers of observed  
22 remote sensing signals?; How can new technologies, synergies, algorithms and models  
23 developed at the site advance our understanding of global dryland ecology at scale?

24  
25 The ARID campaign should invest in proximal sensing to address many of the remaining  
26 barriers to entry, the standardization of products, and the coordination of efforts (Gamon, 2015;  
27 Schimel et al., 2019), following the example of the FLUXNET community (Pastorello et al.,  
28 2020). ARID can be the driving force in support of a first of its kind coordinated dryland proximal  
29 sensing network.  
30





1  
2 **Figure B.8.** Overview of synergies between proximal remote sensing and eddy-  
3 covariance flux data. Major proximal remote sensing techniques shown as  
4 different color overlapping polygons. A description of each technique is shown in  
5 colored text and key physical parameters associated with each technique and  
6 across multiple co-located techniques are listed in black text. Image from  
7 (Pierrat, Z. et al., in review).

8  
9 **B.2.1.4 Uncrewed Aircraft Systems (UAS) Remote Sensing**

10 UAV-based assessments of ecosystems allow finer spatial scale and more frequent data  
11 acquisition than satellites (Z. Sun et al., 2021; Villarreal et al., In Review) and have shown utility  
12 for remote sensing algorithmic development, particularly when paired with coordinated field  
13 campaigns (Chadwick et al., 2020b). UAS platforms and sensors (multi-spectra, LiDAR,  
14 thermal, and image spectroscopy) are evolving very rapidly and providing scientifically robust  
15 data that fill an observation scaling gap between in-situ and airborne data, at ever-reducing  
16 costs.

17 With their extremely high spatial resolution (GSD 5 cm to 1m), UAS data will play a key role in  
18 scaling up field measurements to airborne (1-15m GSD) and space-based measurements  
19 (>10m GSD)(Fig. B.9). For example, the UAS-derived very high resolution fractional cover of  
20 various PFTs can be generated from the fusion of image spectroscopy and LiDAR data (Sankey

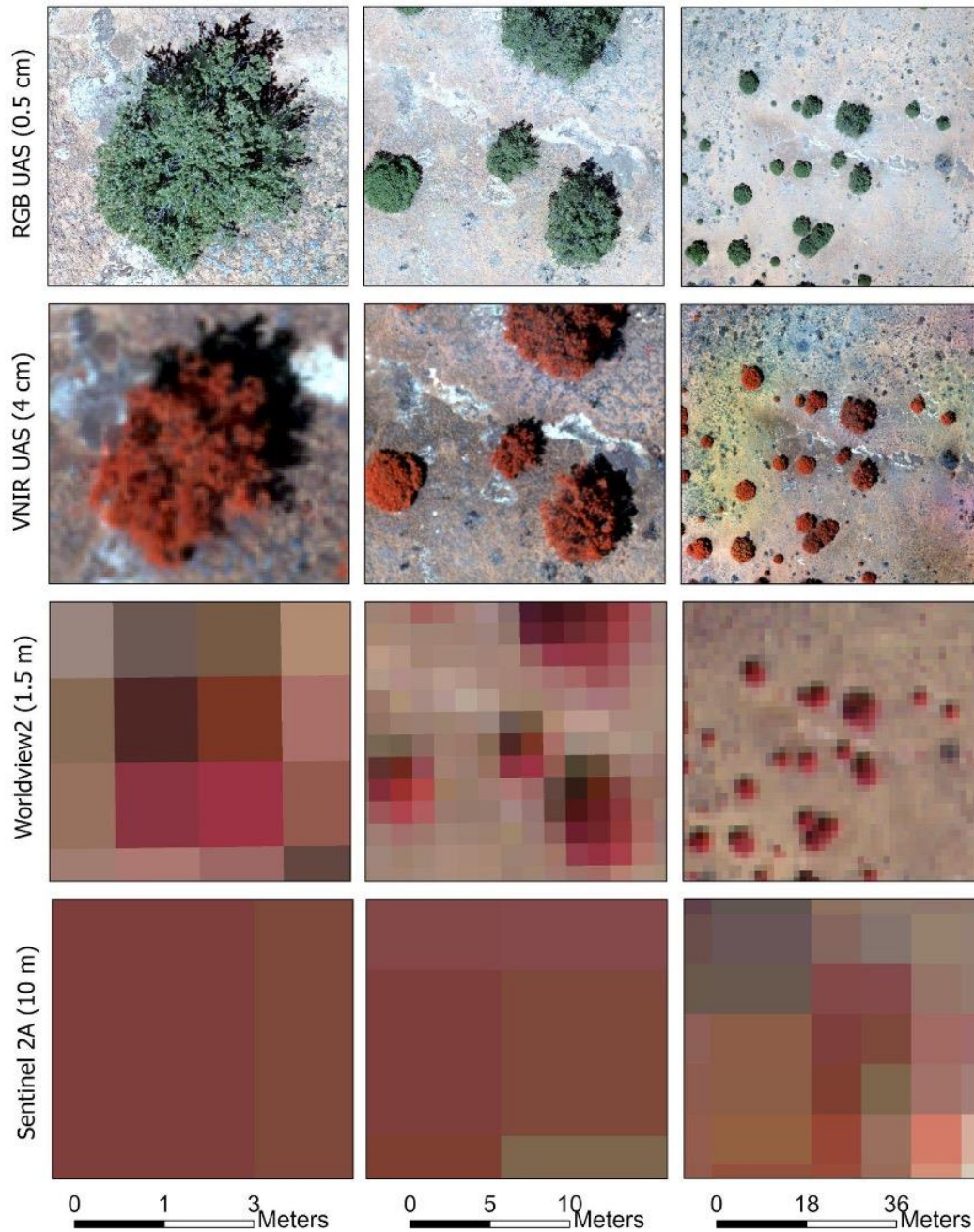
1 et al., 2018), which will be used to develop mixture models that accurately estimate these  
2 fractional covers with AVIRIS data over vast areas (see details above) (Pervin et al., 2022;  
3 Villarreal et al., In Review) (Fig. B.9). Characterizing the distribution of PFTs in drylands is  
4 critical given these ecosystem's high spatial heterogeneity. During the ABoVE field campaign,  
5 researchers demonstrated the indispensable role of UAS in measuring fine-scale, low-stature  
6 vegetation structure (similar to dryland vegetation) using RGB cameras and "structure from  
7 motion" (SFM) photogrammetry, that could not be captured with airborne and satellite data  
8 (Assmann et al., 2020; Cunliffe et al., 2020; Yang et al., 2022). The other significant benefit of  
9 UAS is the flexibility of on-demand data acquisition which can be as frequently as required to  
10 capture rapidly-changing variables during diurnal cycles or pulse dynamics following rainfall  
11 events. *UAS will play a key role in the ARID field campaign to capture the high spatial*  
12 *heterogeneity at flexible on-demand intervals to investigate temporally variable PFT phenology*  
13 *and function during rapid green-up, senescence or pulse events.*

14  
15 Other useful applications of note are using a combination of sensors mounted on UAS  
16 instruments to derive products like GPP and ET (Morgan & Caylor, 2023), which at the higher  
17 spatiotemporal resolutions of UAS, can advance GPP and ET retrievals from satellite platforms.

18  
19 University PI's and commercial service providers have been using UAS with less restrictions to  
20 collect data on government sponsored projects, e.g. in ABOVE (Assmann et al., 2020; Cunliffe  
21 et al., 2020). However, UAS use within US government agencies has been heavily restricted,  
22 leading to the grounding of all BLM and USGS DJI drones in 2019. (However, indications are  
23 that these limitations may be resolved in the next few years as US-built UAS with sufficient  
24 capability become more affordable.) Other ARID partners, e.g. NEON managed by Battelle,  
25 have developed significant drone remote sensing capabilities (including sensor calibration) and  
26 are not impacted by the DJI security concerns and grounding.

27  
28 The ARID mission plans to make extensive use of drone-based data collected by PI's, private  
29 service providers, and government partners (e.g. BLM and USGS). *We thus anticipate PI's and*  
30 *non-government partners (e.g. NEON managed by Battelle) will collect the majority of the UAS*  
31 *data with its limited administrative burden compared to federal government agencies. We*  
32 *furthermore plan to use standardized drone data processing/calibration workflows and the ARID*  
33 *cloud environment to process all drone data, similar to recent airborne data processing*  
34 *developments (see Section D) (e.g. SHIFT, BioScape). ARID cloud environment will support*  
35 *hierarchical data storage, include file and archive version control, provide digital object*  
36 *identifiers (DOIs) for data and use an API for data search, discovery and retrieval, similarly to*  
37 *CyVerse Open Science Workspace (Yang et al., 2022).*

38



**Figure B.9.** Illustration of different pixel resolutions and scales of UAS imagery and satellite imagery. The top row shows an RGB UAS image at 0.5 cm resolution displayed at scales ranging from tree crown to landscape (left to right), followed by a 4 cm resolution visible near infrared (VNIR) UAS image displayed as a false color composite (vegetation is red), a 1.5 m resolution Worldview2 false color composite satellite image and a 10 m resolution Sentinel 2A false color composite. Image from Villarreal et al., (In Review).

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B.2.1.5 Airborne Remote Sensing

1 NASA's airborne Sensors will play a critical role in sensing essential biophysical parameters in  
2 order to address science questions, and scaling up to space-based sensors. The ABoVE  
3 campaign has illustrated how airborne observations have the potential to link field-based,  
4 process-level studies with geospatial data products derived from satellite remote sensing,  
5 spanning the critical intermediate space and time scales essential for a comprehensive scaling  
6 of ecosystems (Miller et al., 2019). Airborne sensors fill the crucial intermediate spatial scale  
7 between field or UAS data and space-based sensors. *While previous NASA airborne campaigns*  
8 *typically only involved a single overpass without a revisit, ARID's community clearly*  
9 *communicated that our science questions require multi-temporal, seasonal, and inter-annual*  
10 *repeat acquisitions to capture the high temporal variability of drylands (see aircraft remote*  
11 *sensing feedback in Appendix F.10).* This poses new challenges and opportunities, as  
12 addressed below.

13  
14 A guiding principle of the ARID airborne acquisition strategy will be to leverage (i) existing  
15 complementary NASA campaigns (e.g. GEMx, FireSense; see Section B2.5), (ii) investments by  
16 other US institutions, notably our partner NEON, (iii) compatible airborne assets and capability  
17 of international partners and commercial service providers to manage costs and simplify  
18 logistics. While extensive airborne data (especially AVIRIS) have been acquired in parts of US  
19 drylands, these efforts were seldom focused on long-term terrestrial ecology science questions.  
20 However, we plan to leverage existing and on-going missions (e.g. GEMx) through coordination  
21 of complementary flights (Section B2.5).

22  
23 The ARID science questions and ARID community feedback (Table B.1, Appendix F.9)  
24 highlights the need for multi-date airborne acquisitions to capture vegetation dynamics within  
25 one growing season, as well as pulse events. This could be achieved by (i) e.g. coordinating  
26 with NEON to complement their peak-greenness acquisitions of their dryland NEON sites with  
27 2-3 additional acquisition of their NEON-AOP instruments, (ii) complementing other campaigns  
28 (e.g. GEMx, FireSense, Western Diversity Time Series) acquisitions for priority sites with 2-3  
29 additional AVIRIS acquisitions and (iii) acquiring very frequent (e.g. weekly) drone-based  
30 imaging spectrometer data over field sites within footprints of AVIRIS overflights. While such  
31 frequent acquisitions pose financial and logistic challenges, the benefit to quantifying rapidly  
32 varying dryland dynamics cannot be overstated, motivating the community to develop innovative  
33 practical solutions rather than defaulting to traditional one-off acquisitions. The process of  
34 optimally selecting potential areas for airborne acquisitions in the US and overseas is outlined in  
35 the mission planning section B.2.4.

36  
37 ARID will utilize a combination of foundational NASA sensors (LVIS, AVIRIS, UAVSAR,  
38 HYTES), PI-led NASA sensors (G-LiHT), and partner sensors (e.g. NEON-AOP). Multiple  
39 sensor modalities will be acquired and combined including, imaging spectroscopy, LiDAR and  
40 SAR, as briefly outlined below.

41  
42 *Imaging Spectroscopy (hyperspectral image spectroscopy - HIS):* The AVIRIS sensors (Classic,  
43 Next generation (NG), and recent versions 3 and 5) are capable of deriving the largest number  
44 of required physical parameters and functional traits (Table B.2, Appendix F.1). AVIRIS will be a

1 key workhorse for ARID given that the spectral signature will provide high resolution retrievals of  
2 carbon stock and fluxes (fractional cover of PFT), biocrust, vegetation composition, PFTs,  
3 vegetation, chlorophyll content, water content, and fuel loads (Pervin et al., 2022). All variables  
4 are otherwise challenging to retrieve with satellite sensors alone and will inform satellite-based  
5 algorithm development for EMIT and SBG. Adequately addressing dryland science questions  
6 will require multi-temporal acquisitions over selected “supersites” to capture changes vegetation  
7 during various phases of the growing season. The HIS data and field work will furthermore  
8 provide essential calibration/validation data for Landsat and SBG-based estimates of key  
9 variables in future updates to BLM’s and USGS’s rangeland monitoring applications (see  
10 Section B.2.7). NEON’s AOP sensors include an HSI that is very similar to AVIRIS-NG and can  
11 help meet measurement demands where necessary. NASA’s G-LiHT can also contribute HIS  
12 observations.

13  
14 *SAR - UAVSAR:* SAR, in particular polarimetric L-band SAR, has proven effective at estimating  
15 woody vegetation biomass and structure changes in arid savannas and shrublands (Naidoo et  
16 al., 2015, 2016; K. Wessels et al., 2019, 2023). The airborne *UAVSAR* (L-band) has been  
17 effective at estimating biomass across a wide range of environments, from tundra to tropical  
18 forests (Z. Zhang et al., 2017). During ARID, the *UAVSAR* sensor could thus be employed to  
19 estimate tree and shrub biomass and cover and to scale up models to regional coverage with  
20 ALOS2/4 and the imminent NISAR mission. *UAVSAR-L* has furthermore proven essential to  
21 estimating fuel loads and fuel moisture content for wildfire risk prediction (K. An et al., 2024).  
22 The multi-band (L & P) *UAVSAR* (L-band and AirMoss) can furthermore contribute to the  
23 accurate estimation of soil moisture and differentiating soil moisture and vegetation or crop  
24 water content (Tobin et al., 2022), especially in combination with AVIRIS to help partition PTF.  
25 The multi-band *UAVSAR* will also enable innovative tomographic SAR (TomoSAR) PolInSAR  
26 methods for measuring vertical vegetation structure in woodlands and savannas. It will  
27 ultimately inform algorithm development for NISAR and applications for spatially heterogeneous  
28 drylands with mixed woody and non-woody vegetation landscapes.

29  
30 *LiDAR:* LiDAR is essential for measuring woody vegetation height, 3D structure, biomass,  
31 carbon stocks, and their changes through time (Table B.2). Multiple discrete and waveform  
32 LiDAR sensors are available as part of various sensor payloads including LVIS (waveform 10 or  
33 20m footprint), G-LiHt (Discrete), NEON-AOP (both discrete and small footprint waveform). The  
34 airborne LiDAR data will be used to calibrate models to estimate biomass with space-based  
35 SAR (e.g., NISAR) and LiDAR (e.g. GEDI, EDGE and IceSAT2).

36  
37 *Multi-sensor Observations:* There is significant opportunity for innovation when combining  
38 various sensors and modalities within the ARID airborne acquisitions. These include  
39 LiDAR+UAVSAR or LiDAR+HSI for vegetation biomass estimation by PFTs.

40  
41 *SIF:* Airborne SIF measurements can be correlated with tower-based SIF and GPP estimates  
42 and can assist in calibrating space-based SIF estimates with OCO-2/3 and TEMPO. SIF is  
43 being measured by the FIREFLY (Fluorescence Imaging of REd and Far-red Light Yield) which

1 has been integrated with the multimodality NASA G-LiHT payload, as well as CFIS (Chlorophyll  
2 Fluorescence Imaging Spectrometer) built to validate OCO-2.

3  
4 Atmospheric Carbon: CARAFE: The Carbon Airborne Flux Experiment (CARAFE) is a unique  
5 NASA asset that combines in-situ atmospheric sensors and wavelet transform eddy covariance  
6 to directly quantify surface-atmosphere exchange of carbon dioxide, methane, water vapor, and  
7 sensible heat at spatial scales of ~1 km (Wolfe et al., 2018). Airborne fluxes can constrain NEE  
8 and energy balance across gradients in modeled and remotely-sensed surface characteristics,  
9 extending beyond the reach of stationary towers and providing opportunities for upscaling and  
10 process studies (Hannun et al., 2020; Poulter et al., 2023). CARAFE will provide the opportunity  
11 to quantify the spatial distribution of carbon fluxes across heterogeneous drylands, especially  
12 when flown over flux towers.

13  
14 MODIS/ASTER Airborne Simulator (MASTER): The MASTER instrument is maintained and  
15 operated by the Airborne Sensor Facility at NASA Ames Research Center. This sensor covers  
16 the midwave infrared with 15 bands, the thermal infrared with 10 bands, and the visible to  
17 shortwave infrared with 25 bands (Hook et al., 2001). This information across the optical and  
18 infrared regions facilitates the quantification of surface temperature, characterization of  
19 emissivity, derivation of ET, and mapping of surface cover elements and change (French et al.,  
20 2008). Use of MASTER data in conjunction with AVIRIS data has been shown to increase the  
21 accuracy of surface temperature quantification (Grigsby et al., 2015).

#### 22 23 *B.2.1.6 The Central Role of NASA Space-based/Satellite Remote Sensing*

24 The current and future NASA space-based sensors will be key in continuously sensing a host of  
25 parameters at a sufficient spectral, spatial, and temporal resolution across vast global drylands  
26 (Table B.2). The ARID field and airborne campaign will enable multi-scale retrieval model  
27 development for accurately estimating key parameters from space-based sensors. A host of  
28 new sensors and missions provide exciting opportunities for innovations that combine multiple  
29 sensing modalities (e.g., optical and SAR) to address key science questions and applications  
30 (e.g., rangeland monitoring) at the appropriate spatio-temporal scales. The combination of  
31 diverse NASA satellites offers critical measurements for improving our predictive understanding  
32 of dryland processes, controls, variability, and adaptation potential. **Given the high  
33 spatiotemporal complexity of dryland landscapes, a better understanding of drylands  
34 globally will require a full interpretation of current and future NASA spaceborne sensors  
35 and their increasingly high-resolution capabilities spectrally, spatially, and temporally.**

36  
37 Key parameters estimated by NASA space-based sensors include Level 2 and 3 products such  
38 as land surface temperature (LST; Landsat, ECOSTRESS, SBG), solar-induced fluorescence  
39 (SIF; OCO-2, OCO-3, TEMPO), vegetation structure (GEDI, ICESat-2, NISAR) and composition  
40 (EMIT, HISUI, SBG), and soil moisture (SMAP, NISAR); and Level 4 products such as gross  
41 primary productivity (GPP; VIIRS, Landsat, OCO-2, OCO-3), aboveground carbon stocks  
42 (SMAP, GEDI, ICESat-2), evapotranspiration (ET; Landsat, ECOSTRESS), water use efficiency  
43 (WUE; Landsat, ECOSTRESS), root zone soil moisture (RZSM; SMAP, NISAR) (Table B.2,  
44 Appendix F.1). NASA's soil moisture (SMAP, NISAR), vegetation water content (SMAP,

1 NISAR), terrestrial water storage (GRACE), water vapor (AIRS), and precipitation (GPM)  
2 estimates also provide an opportunity to capture a range of dryland mechanisms relevant to  
3 water availability and the soil-plant-atmosphere continuum (SPAC). Importantly, there is an  
4 urgent need to improve and expand upon these existing products through improved evaluation  
5 and calibration, integration of novel measurements and techniques, data fusion, algorithm  
6 advances, and data-model development (e.g., model benchmarking, parameterization, and data  
7 assimilation).

8  
9 *Capturing long-term trends and variability of global drylands:* Long-term satellite records from  
10 NASA polar orbit platforms provide invaluable baseline information across a diverse range of  
11 techniques. This includes 20-40 years of estimates of NDVI (MODIS, VIIRS, Landsat), LST  
12 (MODIS, VIIRS, Landsat), soil and vegetation moisture content (AMSR, SMAP), and, more  
13 recently, vegetation structure (ICESat) and SIF (OCO-2). These platforms provide critical  
14 observations at a fixed overpass time and are ideal for evaluating vegetation function,  
15 phenology, and long-term trends in vegetation dynamics (e.g., the global greening of semiarid  
16 areas) (Fensholt et al., 2012). The long-term NDVI record has played a central role in the  
17 controversial debate around human-induced desertification versus the impact of rainfall  
18 variability (Prince et al., 2007; K. J. Wessels et al., 2012). NASA's soil moisture (SMAP),  
19 terrestrial water storage (GRACE), water vapor (AIRS), and precipitation (GPM) estimates will  
20 play a crucial role in hydrological modeling at the fine temporal scales needed to capture pulse  
21 dynamics of dryland systems. More recent higher resolution remote sensing datasets and  
22 modeling will allow an ARID campaign to unpack the politicized, umbrella term of  
23 "desertification" into changes in composition, structure and function (e.g., water use efficiency)  
24 of dryland ecosystems (Brandt et al., 2020; Mugabowindekwe et al., 2023; Tucker et al., 2023).

25  
26 *Capturing high temporal variability of global drylands:* Satellite-based sensors can help address  
27 the challenge of capturing the high temporal variability of dryland ecosystems through novel  
28 data fusion techniques and better integration of measurements from geostationary satellite  
29 platforms. Advanced data fusion techniques are revolutionizing our ability to seamlessly  
30 integrate multiple satellite records towards monitoring drylands in unprecedented temporal  
31 detail. Leading the way, the NASA Harmonized Landsat and Sentinel-2 (HLS) initiative is  
32 integrating observation from Landsat 8, Landsat 9, Sentinel-2A, and Sentinel-2B satellites to  
33 generate a harmonized, analysis-ready surface reflectance data product at 30-meter spatial  
34 resolution and with observations every two to three days. There is also the emerging opportunity  
35 to better utilize high temporal frequency observations from geostationary satellites, which have  
36 served primarily as inputs for weather forecasting, and overcome many of the temporal  
37 limitations of orbital sensors. The geostationary constellation provides almost complete  
38 coverage of global drylands in North/South America (GOES-R, TEMPO), Central Asia/Australia  
39 (Himawari-8/9) and Africa/the Middle East/Southern Europe (SEVIRI). Of note, the TEMPO  
40 mission (Earth Ventures class platform) is providing first-time hourly SIF measurements over the  
41 entire US that can be used to track the diurnal cycle and pulse dynamics of dryland vegetation  
42 productivity. Emerging initiatives to fuse measurements across multiple low orbit and  
43 geostationary sensors is enabling the monitoring of key ecosystem variables such as GPP and  
44 ET at unprecedented spatial and temporal resolutions (Khan et al., 2022; Stisen et al., 2007).

1 For instance, the NASA GeoNEX project through data fusion is now providing harmonized,  
2 analysis-ready surface reflectance and LST products at hourly, daily, and sub-weekly temporal  
3 resolution at a near-global scale. Together these novel platforms and initiatives offer invaluable  
4 yet underexplored insight into pulse dynamics of dryland ecosystems and offer potential to  
5 revolutionize our understanding and informing critical societal decisions regarding our planet's  
6 drylands.

7  
8 *Capturing unique high spatial and temporal variability of global drylands from NASA's Venture*  
9 *class sensors on the ISS:* The NASA Ventures class sensors and missions operated by other  
10 national space agencies on the International Space Station (ISS) could transform our  
11 understanding of dryland structure (e.g., tree, shrub canopy height) and function (e.g., GPP and  
12 ET). While these sensors provide high temporal resolution in some cases, they have high  
13 capacity to evaluate dryland's heterogeneous surfaces with their mostly higher spatial  
14 resolutions (<100m). ECOSTRESS (70-m) measurements of LST (Level 2) and ET (Level 3)  
15 which can be combined with co-occurring OCO-3 (2-km) measurements of SIF (Level 2) and  
16 GPP (Level 3) to derive first-time integrated estimates of WUE (Level 4). These observationally-  
17 constrained estimates of ET, GPP, and WUE have the potential to fully overcome limitations of  
18 previous products and revolutionize our understanding of dryland function from the individual  
19 rainfall pulse to annual timescales. EMIT, DESIS and HISUI are providing first-time space-borne  
20 hyperspectral measurements that can be used to map dryland spectral traits across diverse  
21 functional groups (e.g., C3, C4, and CAM plants, diverse biocrust species). Additionally, these  
22 hyperspectral missions will revolutionize our knowledge of geological surface composition, as  
23 well as vascular and non-vascular plant fractional cover across global drylands. Finally, the  
24 GEDI mission is providing novel structural traits (e.g., vegetation height, plant area index, and  
25 total aboveground biomass) that will for the first time capture the true high structural biodiversity  
26 of drylands and enable first time assessments of how dryland structure relates to and potentially  
27 regulated dryland function. These instruments offer observations at variable overpass times and  
28 thus can be jointly used to gain insights into diurnal dynamics at high spatial resolution. We  
29 acknowledge that it is possible that several of the sensors listed here will no longer be available  
30 during ARID (e.g., DESIS). However, several like EMIT, ECOSTRESS, GEDI are likely to  
31 continue collecting data from the ISS for several more years during the ARID campaign,  
32 depending on various factors. Continued acquisitions from these three sensors during the ARID  
33 study is strongly desired and will allow synergistic use of these different data types and  
34 preparation for SBG (Stavros et al., 2017).

35  
36 *Capturing high spatial and temporal variability of global drylands with commercial small satellite*  
37 *constellations:* Emerging polar-orbiting small satellite constellations have the potential to  
38 provide first-time high spectral and spatial resolution observations in the optical and thermal  
39 regions of the electromagnetic spectrum. The PlanetScope constellation is providing surface  
40 reflectance across key optical bands at 3-m spatial resolution and a daily revisit frequency. The  
41 newly launched Hydrosat constellation will provide complementary observations in the thermal  
42 infrared region of the electromagnetic spectrum, which can be used to estimate LST, surface  
43 water availability, and the evolution of drought dynamics across drylands at 20-m spatial  
44 resolution and a daily revisit frequency. Through ARID, we will partner with these initiatives and



1 explore options to continue and expand the NASA NextView License to make these data fully  
2 available to the ARID research community (Neigh et al., 2013).

3  
4 *Key Future Sensors: NASA's Surface Biology Geology (SBG) Designated Observable mission*  
5 objectives are strongly aligned with ARID's objectives by focusing on terrestrial vegetation  
6 physiology, functional traits, and health (Cawse-Nicholson et al., 2021), as well as wildfire  
7 ecology and risk applications, at high spatial resolution. Specifically, the SATM in Appendix F.1  
8 indicates that SBG Thermal and VSWIR components will play a key role in estimation of  
9 multiple parameters related to fractional cover of PFTs, and other vegetation and fuel  
10 characteristics in drylands. SBG's hyperspectral component will be especially important for  
11 parsing vegetation and soil. The airborne AVIRIS data will enable the development of accurate  
12 spectral unmixing models with which these parameters can be estimated, and methods applied  
13 using Landsat and SBG (Cawse-Nicholson et al., 2021). These spaceborne instruments in  
14 combination with structural parameters and woody vegetation biomass estimated with NISAR  
15 and space-based LiDAR (GEDI, EDGE) will revolutionize the estimation of dryland vegetation  
16 composition, structure, and biomass (Duncanson et al., 2022; X. Li et al., 2024), especially in  
17 spatially heterogeneous drylands. This will include dryland applications related to rangeland  
18 condition, shrub encroachment, invasive grasses, fuel moisture and fuel loads. Additionally, the  
19 ARID campaign will be critical to the calibration and validation of SBG and NISAR observations  
20 and derived products, including soil moisture, vegetation water content, and fuel moisture  
21 products from NISAR. Landsat-NEXT will also provide 26-bands of multi-spectral bands  
22 including 5 bands in TIR wavelengths at 10- to 60-m resolution with a very high temporal 6-daily  
23 frequency provided by the triplet of satellites. With a planned launch year of 2030, ARID and its  
24 follow-on applications may still provide benefit to, and, in turn, benefit from Landsat NEXT.

25  
26 The Decadal Survey (2017) (Medicine et al., 2019), highlighted three Explorer mission  
27 observables, including Terrestrial Ecosystem Structure, and a mission called EDGE ([Earth](#)  
28 [Dynamics Geodetic Explorer](#)) has just been approved for Phase A development. The space-  
29 based waveform LiDAR on EDGE is specifically tailored to have the ability to characterize  
30 dryland ecosystems with discontinuous, short stature vegetation structure, in addition to  
31 forested ecosystems. The Surface Topography and Vegetation (STV) designated observable is  
32 currently in incubation phase, but could enter the next development phase during the course of  
33 the ARID campaign. If successful, both STV and EDGE have clear science objectives (see  
34 SATM) to accurately measure short stature vegetation structure and biomass in drylands. ARID  
35 will contribute to their calibration and validation efforts while the science community will benefit  
36 from the accurate measurements of dryland carbon stocks and PFTs.

### 37 *Sensors from other Agencies and commercial companies*

38 While this section mainly aimed to highlight the contribution of NASA sensors, it is clear that  
39 ARID will greatly benefit from the sensors made available by other agencies and commercial  
40 companies. The European Space Agency's (ESA) series of Copernicus satellites will contribute  
41 in several ways including (i) increasing the temporal frequency of wide area observations  
42 through the Harmonized LandSat Sentinel-2 (HLS) which will achieve a nominal 3 daily  
43 observation rate, (ii) providing hyperspectral data via EnMAP at 30 x 30m and Copernicus

1 Hyperspectral Imaging Mission for the Environment (CHIME) scheduled to launch before the  
2 end of the decade, (iii) providing SIF observations from TROPOMI and the upcoming FLEX  
3 mission, (iv) providing microwave data from ALOS4 L-band and SAR from JAXA, and (v) as  
4 discussed above, providing high frequency and fine-scale GOES observations which are being  
5 used to develop GPP and ET products at a sub-hourly temporal resolution (Khan et al., 2021).  
6 We will also continue to build collaborations with commercial platforms including WorldView,  
7 PlanetScope, and Hydrospec under the NASA NextView License framework and explore ways  
8 in which we can make these data freely available to the ARID research community (Neigh et al.,  
9 2013).

10

### 11 ***B.2.2 Dryland Ecosystem Modeling and Model-Data Synthesis***

12 We envisage two key roles for models as part of the ARID field campaign. First, the ARID field  
13 campaign will provide critical opportunities to address long-standing scientific questions  
14 regarding dryland functioning and improve representation of processes within models. Data on  
15 vegetation structure and composition, vegetation phenology, water availability and plant water  
16 use, and fire dynamics will allow us to improve model parameterizations, enabling more  
17 accurate predictions of dryland carbon and water flux variability, plant responses to pulsed  
18 rainfall and extreme climate events, and long-term vegetation shifts. These efforts will  
19 significantly enhance our ability to predict dryland ecosystem responses to global change, as  
20 well as dryland ecosystem feedbacks to the climate system, and ensure models can better  
21 support land management and policy decisions.

22 Second, models provide useful direction for data collection efforts and in the design of remote  
23 sensing instruments. In underrepresented dryland ecosystems, where our process  
24 understanding, mapping, and modeling all remain poor, models can play a crucial role in  
25 supporting hypothesis testing, planning for data collection, in evaluating the precision and  
26 resolution of emerging remote sensing technologies, and in aiding development of algorithms  
27 for retrieving land surface parameters through Observing System Simulation Experiments  
28 (OSSEs).

29 Overall, ARID provides an opportunity to systematically link ground observations of important  
30 ecosystem process responses to climate variability, water availability, and ecosystem structure,  
31 involving remote sensing observations across different spatial, spectral, and temporal scales, to  
32 process-based models via comprehensive and extensive data-model synthesis. The linkage  
33 between models and different scales of observation serves as a bridge between ecosystem  
34 theory and multiscale observational platforms to better understand, as well as predict, how  
35 dryland systems respond to emerging climate and land use changes across dryland regions of  
36 the world.

37

#### 38 ***B.2.2.1 Interfacing Dryland Modeling and Dryland Remote Sensing***

39 Process-based models at all scales generally do not fully represent the complex structure,  
40 functioning, and dynamics that are characteristic of dryland ecosystems. This includes models  
41 from smaller scale ecological (process-specific models), to landscape or habitat scales, to  
42 global-scale dynamic vegetation models and land surface models implemented within Earth

1 system models. Such lack of representation is in part because models were largely developed  
2 and tested for more mesic, homogeneous forested ecosystems where data and experiments are  
3 more numerous and understanding of relevant processes more advanced. We outline several  
4 dryland modeling limitations here.

5  
6 First, the representation of the spatial complexity of mixed woody-herbaceous ecosystems and  
7 associated processes (e.g., shrub-grass competition for water) is largely absent from all  
8 modeling frameworks. For example, the radiative and aerodynamic transfer schemes  
9 implemented in most global-scale land surface models (LSM) assume homogeneous vertical  
10 distribution of leaf area, causing overestimation of light interception (Bégué et al., 1996), and  
11 incorrect parameterization of surface layer aerodynamic conductance (Dolman & Wallace, 1991;  
12 Kabat et al., 1997; Young et al., 2021), with implications for carbon and water cycling and  
13 energy balance calculations. Many models also lack representation of dryland specific plant  
14 functional types (PFTs), including biocrusts, annual and perennial grasses, and forbs, and  
15 shrubs. In addition, while both C3 and C4 photosynthetic pathways are included in many  
16 models, fractional cover of these key PFTs are likely poorly estimated and not well validated  
17 and there is generally no representation of warm versus cool season types, both of which are  
18 pivotal for understanding drylands such as in the western US. The methods used to create  
19 global scale PFT fractional cover maps used in LSMs have the highest uncertainties in  
20 heterogeneous, sparsely vegetated dryland regions, which results in a large spread in modeled  
21 carbon, water, and energy fluxes due to uncertainties in PFT fractional cover (and notably is  
22 larger than the inter-model spread due to differences in model structures) (Hartley et al., 2017b).

23  
24 Second, key processes for modeling dryland ecosystem responses to rainfall pulses, carbon  
25 flux variability, woody plant encroachment, or vegetation recovery from disturbance (e.g., fires)  
26 and climate extremes (e.g., severe drought) are either only recently under development and  
27 have not yet been widely tested for drylands, or missing in models altogether. Such processes  
28 include dryland plant specific phenology schemes, fire dynamics and vegetation responses to  
29 fire, mechanistic descriptions of plant water uptake and use (e.g., plant hydraulics or stomatal  
30 optimization; dryland plant specific rooting profiles and competitive interactions), hydraulic  
31 redistribution, and groundwater recharge. Most phenology schemes have been developed for  
32 boreal and temperate mesic, forested ecosystems (MacBean et al., 2015) and therefore dryland  
33 specific PFT phenology schemes need to be developed (Dahlin et al., 2017; Hufkens et al.,  
34 2016; Renwick et al., 2019), which will require new data collection and analysis efforts (B.2.1  
35 and B.2.3) and harmonization of existing data. Most models still only include an empirical water  
36 stress function that performs poorly in capturing plant water stress (De Kauwe et al., 2015). New  
37 mechanistic approaches to modeling plant water uptake and use (e.g., plant hydraulics and  
38 stomatal optimization) are currently being developed for LSMs (Paschalis et al., 2024; Sabot et  
39 al., 2022), but have not yet been widely tested in dryland ecosystems (though see Hawkins et  
40 al., 2022). Other processes and model parameterizations relevant for capturing plant responses  
41 to water stress, such as dryland plant specific rooting profiles and groundwater have not widely  
42 been implemented in models. While recent studies have implemented hydraulic redistribution  
43 (HR) by roots in LSMs (C. Fu et al., 2018; Niu et al., 2020), further research is needed to assess  
44 the consequent impacts on simulations of carbon and water fluxes, vegetation dynamics, and

1 plant responses to drought across a range of dryland ecosystem types, and how well these new  
2 HR schemes perform when applied in conjunction to other developments in modeled  
3 ecohydrological functioning (e.g., plant hydraulics). Fire modules are often not activated, or  
4 even included, in LSMs and where they are used can perform poorly (Baudena et al., 2015;  
5 Hantson et al., 2020; Teckentrup et al., 2021). Even where fire processes are included in  
6 models, the parameterizations are often too simple to represent the complex interactions  
7 between fuel types and fuel flammability controlling the probability of ignition and spread of  
8 crown fires, common in dry forests, relative to surface fires, common in open herbaceous  
9 systems, the duration of fires in dryland regions, and the role of human activities in fire ignition  
10 (Hantson et al., 2020; Teckentrup et al., 2019; Thavhana et al., 2024).

11  
12 Third, model performance is poor where dryland-specific processes should in theory be  
13 represented in the models. These processes include asynchronous dryland plant responses  
14 (e.g., PFT-specific leaf phenology and physiological activation) to large seasonal changes in  
15 water availability, periods of water stress, or pulsed rainfall dynamics, (De Kauwe et al., 2017;  
16 MacBean et al., 2021; Whitley et al., 2016). Unsurprisingly given these three factors, dryland  
17 model evaluations have shown that vegetation dynamics (e.g. phenology, long-term trends),  
18 carbon, and water fluxes (including partitioning between transpiration and bare soil evaporation)  
19 and the response of these processes to warming and elevated CO<sub>2</sub> are not well represented in  
20 models (Dahlin et al., 2015; Dashti et al., 2021; Fawcett et al., 2022; MacBean et al., 2015,  
21 2020, 2021; Renwick et al., 2019; Teckentrup et al., 2021; Traore et al., 2014; Whitley et al.,  
22 2016) with implications for predictions of land-atmospheric feedbacks (Humphrey et al., 2018;  
23 (Simpson et al., 2024b).

24  
25 Models should play a vital role in improving our knowledge of the role of drylands in global  
26 carbon and water cycles and land-atmosphere feedbacks, and for forecasting the effects of  
27 climate change on ecosystems that underpin provisioning of vital ecosystem services for a  
28 range of livelihoods in both developed and developing countries. However, we cannot currently  
29 rely on process-based models either for attributing observed changes in dryland ecosystems  
30 (shrub encroachment as one example) to environmental change drivers, or for accurately  
31 predicting dryland ecosystem responses to changing climate variability, extremes, and land  
32 management. *There is an urgent need to assess where and why models are not performing well  
33 in drylands, and to use new data to improve our understanding of dryland ecosystems.*

34  
35 ARID is of vital importance for improving our ability within the ecological and Earth system  
36 science communities to model drylands. *Integrating field and satellite data from the ARID  
37 campaign will provide an unprecedented opportunity to develop models that can reliably  
38 address a number of ecosystem challenges in the dryland system.* The field campaign proposed  
39 in ARID will provide the requisite synthesis and harmonization of existing data together with new  
40 datasets and analyses that are sorely needed to improve modeling of dryland ecosystems.  
41 Additionally, modeling itself fosters collaboration among dryland scientists, remote sensing  
42 scientists, and land managers, all with a goal of driving innovative projects to fill significant  
43 knowledge gaps. In this section, we outline the scope of model improvements that could be  
44 made with the ARID field campaign (aligning with planned data collection efforts), how we

1 envisage such a model improvement effort could be organized, and the technical and scientific  
2 questions we plan to answer with ARID modeling activities.

### 3 4 *B.2.2.2 How ARID can co-improve dryland modeling and model-data synthesis for improved 5 process understanding*

6 Modeling and model-data fusion techniques enhance the utility of remote sensing data for  
7 improving process understanding by integrating observations with model predictions to estimate  
8 quantities that cannot be observed remotely. By fusing process models with remote sensing  
9 data, models can be constrained and adjusted, improving the accuracy of predictions related to  
10 ecosystem processes such as below ground processes, carbon fluxes, and water use. For  
11 example, the NASA ABoVE domain estimates of above ground biomass or leaf area derived  
12 from remote sensing were used to reduce bias in model biomass and improve estimates of  
13 regional carbon flux and belowground carbon pools (Huo *et al.* 2024). This approach not only  
14 improves model performance but also allows for better understanding of complex ecosystem  
15 dynamics, particularly in heterogeneous environments where direct measurements are limited.

16 A dryland focused remote sensing field and satellite campaign will also provide much needed  
17 datasets with which we can improve modeling of dryland ecosystems. The ARID field campaign  
18 will allow for coordinated data collection to address multiple different facets of poor dryland  
19 ecosystem process representation (i.e., simultaneously evaluate and constrain processes  
20 related to vegetation dynamics, carbon and water fluxes with multiple datasets). New data  
21 collection together with ARID supported synthesis and harmonization of existing datasets will  
22 support dryland model initialisation (e.g., PFT maps), process development, and testing.

23  
24 Four core processes that are key for modeling dryland ecosystem responses to climate  
25 variability and extremes (Section B.1.1 to B.1.3) will be targeted for improvement based on new  
26 data collection and harmonization in ARID. These include inaccurate representation of: 1)  
27 vegetation and biocrust composition and structure in sparsely vegetated, heterogeneous  
28 ecosystems; 2) dryland plant specific phenologies; 3) processes related to water availability,  
29 uptake and use (including groundwater, hydraulic redistribution, plant hydraulics or stomatal  
30 optimization schemes and dryland plant specific rooting structures); and 4) fire and recovery  
31 from fire. If we can improve model representation of these four core processes we should be  
32 able to better predict plant responses to pulses and droughts (from improvements to modeled  
33 plant water use and phenology) (B.1.1.2), carbon cycle inter-annual variability (related to  
34 interactions between pulsed ecosystem responses and seasonal dynamics (B.1.1.1 and  
35 B.1.3.1), and long-term changes in vegetation composition and carbon sequestration (related to  
36 responses to disturbance and competition for resources between different PFTs and biocrusts)  
37 (B.1.2 and B.1.3). All four of these categories are closely aligned with data collection efforts  
38 planned in the ARID field campaign, outlined in the remote sensing approach in B.2.1 and  
39 SATM Table B.2 as well as in the science theme approaches in Section B.1. For example, data  
40 collected during the ARID campaign on vegetation fractional cover and height will assist in  
41 initializing land surface models with static vegetation or in testing shifting vegetation composition  
42 and competition predicted by vegetation demographic models; LAI, FAPAR, SIF, NIRv, and PRI  
43 data for specific dryland PFTs will help develop new dryland PFT phenology schemes with  
44 impacts on our ability to simulate dryland energy balance, carbon and water fluxes; biomass

1 and burnt area data will help to test and develop fire models; and soil moisture, hydrogeodesic  
2 (groundwater), VOD, and leaf water potential measurements together with partitioned ET data  
3 will help to assess new plant hydraulics and hydraulic redistribution schemes that in turn will  
4 enable models to better represent drought stress (with feedbacks on simulations of phenology,  
5 pulsed responses, and drought induced mortality). All above mentioned datasets, together with  
6 Level 4 remote sensing and upscaled in-situ flux products (e.g., GPP and ET) will be utilized to  
7 quantify and constrain uncertainties in models across scales via data assimilation and emerging  
8 machine learning methods for parameter and state estimation.

#### 9 10 *B.2.2.3 Plan for modeling activities within ARID*

11 We envisage a four-phase modeling effort within ARID, with a continuous flow of information  
12 between modeling and data/experimental communities at each stage (Table B.3):

13  
14 **Phase 1: Data Model Fusion Framework:** Development of a framework that can support the  
15 linkage between various observations of dryland ecosystem structure and processes to the  
16 various ecosystem process representation available in different modeling schemes will enable  
17 better information exchange. This DMF framework will provide additional guidance on how  
18 ecosystem processes can be better inferred by remote sensing observations. Estimates of  
19 ecosystem dynamics based on modeling analysis can be cross referenced to remote sensing or  
20 ground-based observations.

21  
22 **Phase 2a: Model inter-comparison.** Modeling of dryland ecosystems is unique in that the  
23 processes needed to accurately represent dryland ecosystem structure, functioning, and  
24 dynamics are not necessarily included in the models, and knowing which processes are  
25 important is a considerable challenge. Therefore, the goal of the initial model-intercomparison  
26 should be to benchmarking multiple emergent dryland ecosystem variables at broad spatial  
27 scales (e.g., model-data comparison of GPP, ET, above- and belowground carbon stocks, and  
28 soil moisture using iLAMB or other available tools) to identify where/when models agree or  
29 disagree and better understand possible causes of poor dryland model performance.

30  
31 **Phase 2b: Detailed, assumptions-based process-evaluation.** Building on Phase 2a, in  
32 Phase 2b centrally funded personnel (see below) would coordinate a voluntary multi-model  
33 inter-comparison comprising a suite of site-based simulations (including experimental  
34 manipulations) in order to perform tests and assessments of model assumptions used to  
35 simulate various variables in an assumption-based evaluation (Medlyn et al., 2015) of core  
36 processes (for example, carbon assimilation, phenology, root water uptake, litter input quantity  
37 and quality) that contribute to emergent ecosystem level responses. The results of these  
38 analyses will be published in a journal article detailing the “roadmap” of dryland model  
39 developments that are needed and that will be supported by ARID.

40  
41 **Phase 3: Model developments and optimization:** Based on our existing knowledge of what  
42 processes are misrepresented or missing in models, we expect that new model developments  
43 will be needed. We anticipate that dryland model developments will be incorporated into many  
44 NASA Requests for Proposals (RFP). However, we advocate that the roadmap of model

1 developments resulting from Phase 2a,b efforts should be the target of dedicated core funding.  
2 Given the significant resources and continuity of personnel efforts that are required for rigorous  
3 model developments and model optimization, we suggest targeted global scale modeling  
4 groups (with different core structures – e.g., land surface model and dynamic  
5 vegetation/vegetation demographic models) be funded for a specific, competitive RFP. This  
6 funding should include a requirement that these modeling groups collaborate with individual or  
7 smaller groups of scientists with expertise on specific processes – i.e., those working finer  
8 scale, process-specific models, and/or datasets that can help test hypotheses and guide model  
9 developments. We also advocate that teams submitting proposals for this dedicated RFP  
10 include experts on data assimilation and model integration. We envisage this dedicated RFP will  
11 help to build teams involving strong collaborations between empirical scientists and modelers to  
12 develop new mathematical functions of missing processes (e.g., dryland specific plant  
13 phenology, biocrusts, etc.). The goal of funding only 2-3 modeling groups is that each new  
14 development can be carefully tested and evaluated so that other modeling groups can benefit  
15 directly from knowledge gained about which processes improve dryland ecosystem modeling,  
16 and how to best parameterize those processes in their models. We believe this framework is the  
17 most optimal for maximizing limiting resources for the whole modeling community.

18  
19 **Phase 4: Regional to global scale modeling to address ARID science themes.** The ultimate  
20 goal of the ARID modeling effort will be to use the new state-of-the-art dryland modeling  
21 schemes developed throughout the 10-year field campaign to make more reliable predictions of  
22 the impact that climate and environmental change will have on dryland ecosystems from near  
23 term to decadal and centennial time scales, and from ecosystem to global spatial scales (see  
24 more detailed science questions below). This fourth and final phase will include model  
25 implementations and output interpretation of the several main models designated for  
26 development within ARID. This phase, which will start in the final stages of ARID and last well  
27 beyond the planned field campaign, will also see the re-evaluation of models within a global  
28 benchmarking framework, using new airborne, upscaled, and satellite data products developed  
29 within the ARID field campaign.

30  
31 *Table B.3. Three phase modeling framework envisaged for ARID describing the*  
32 *3 phases, who the work would be led by, the funding required, and the overall*  
33 *goal for that Phase.*

Phase	1	2a	2b	3	4
<b>Modeling Activity</b>	Data-Model Fusion framework to develop links (observation operators) between ARID datasets and model processes	Broad-scale model intercomparison of emergent ecosystem scale processes (GPP, ET)	Assumptions-based evaluation of core processes (photosynthesis; stomatal conductance; etc)	Model development and optimization to improve core processes (e.g., plant hydraulics, phenology)	Model application to science questions; Re-evaluation with new ARID field and satellite derived products)
<b>Led by</b>	Modeling Task Force selected competitively (part of science team) to assess needs for observation operators. New observation operators developed by RS and modeling community via RFPs and supported by centrally funded postdoc	Modeling Task Force to develop simulation protocol. Postdoc for input data processing and synthesis and MIP. Voluntary simulations from modeling community	Postdoc carries out focused model evaluation to inform data collection and model developments needed (overseen by Model Task Force)	Two teams led by global scale modeling center or institution working with a global scale model as PI	Modeling community for model applications; Postdoc to perform re-evaluation from 1st stage
<b>Funding mechanism</b>	Core/directed funding for postdoc and RFPs	Core/directed funding for postdoc	Core/directed funding for postdoc	Competitive funding (either core or via dedicated model development RFP) but only for 2 modeling groups	RFPs for model application; Core/directed funding for postdoc
<b>Goal</b>	New observation operators where needed to ensure consistent linkages between ARID data and models	Informs data collection needs and core process evaluations	Informs needs for model development	Improves models representation of dryland dynamics along our science themes (pulse behavior, droughts, C cycle IAV, fires etc); quantify and reduce uncertainties in structure and parameters	Address science theme questions; Evaluate reduction in model spread due to model improvements; Assess utility of new ARID data for improving modeling

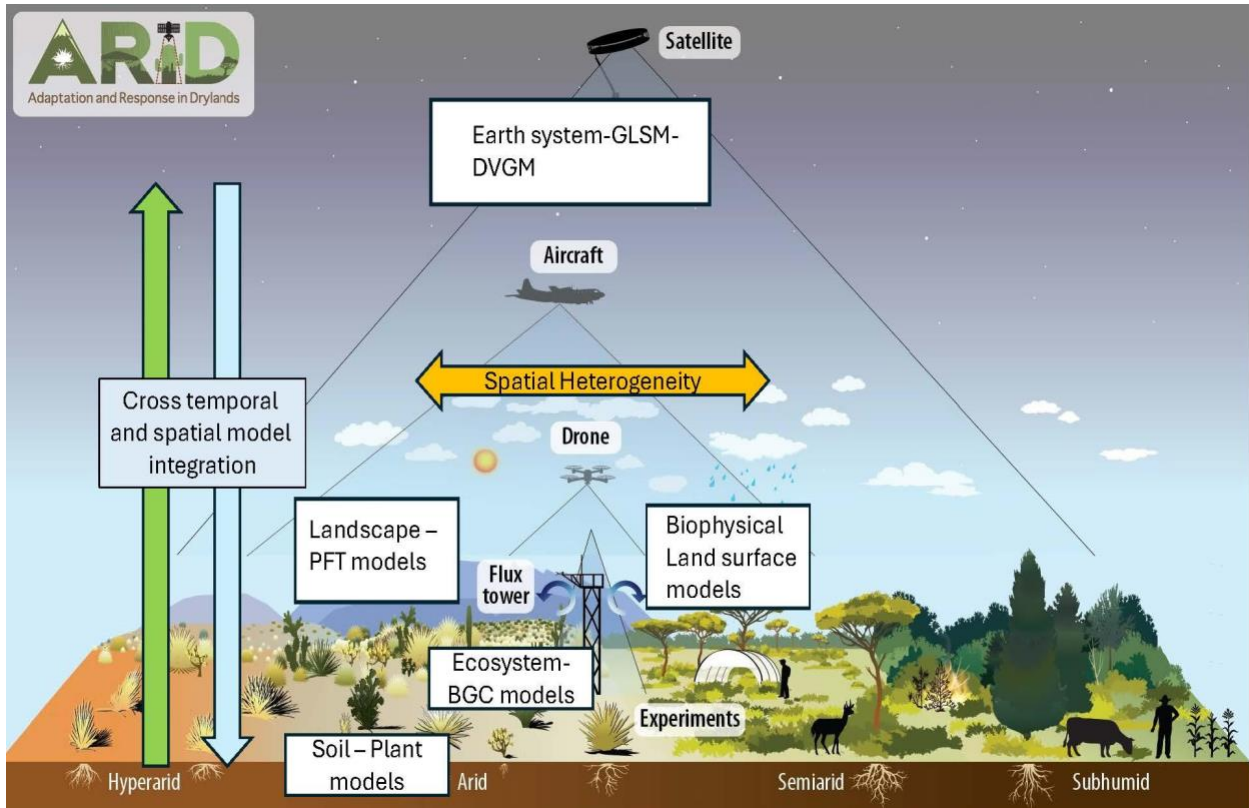
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In addition to model evaluation and developments, the wealth of new and existing dryland datasets collected and/or synthesized within ARID will be utilized by all modeling groups for model parameter optimization and state estimation using traditional data assimilation and emerging machine learning methods. Currently, most parameters in global scale models have been calibrated for more mesic ecosystems and likely do not represent dryland plant and soil traits. Literature and trait database reviews combined with parameter sensitivity analyses are needed to assess what new trait data collection efforts are needed and to inform prior bounds for parameter data assimilation experiments. This work will predominantly be funded via RFPs to individual modeling groups. A large part of this research will need to be devoted to developing the technical expertise to assimilate diverse, multi-variable dryland datasets into process-based models, and thus needs to occur in parallel to model developments (see B.2.2.4 Technical Questions). We anticipate that machine learning and artificial intelligence methods applied to ARID data should be used in some cases where process understanding is still lacking – for example, in understanding fire occurrence and spread (see B.2.2.4 Technical Questions).

As indicated in Phase 4, ARID data collection efforts across multiple scales enables our multi-scale modeling framework (Fig. B.10). It is crucial that we leverage data from field and airborne remote sensing through to space-borne observations including those from geo-stationary observations. Ecophysiological, ecohydrological, process-specific models at ecosystem scales, landscape and habitat models, and land surface and dynamic vegetation models working in concert will provide enhanced analytical capacity to integrate information of ecosystem dynamics gained from multiple sources of remote sensing observations in terms of spatial, temporal, and spectral scales. Process-specific field or high resolution airborne and satellite datasets will enable hypothesis testing with smaller scale, more process-oriented models, from which process understanding will be implemented into larger scaled models. Higher level,



1 coarser resolution data products will be used to evaluate and calibrate ecosystem scale fluxes  
 2 at broader spatial scales.



3  
4  
5  
6

**Figure B.10.** The role of multi-scale modeling within ARID.

7 While we have targeted four key processes for model development in ARID (vegetation and  
 8 biocrust composition and structure, phenology, plant water use, and fire), we note that  
 9 challenges exist in modeling belowground dynamics, such as soil CO<sub>2</sub> pulses following rainfall  
 10 (i.e., the “Birch Effect”), nutrient cycling, and inorganic carbon dynamics (not currently  
 11 represented in any global scale land surface or dynamic vegetation model). These are important  
 12 issues that need to be addressed for adequately modeling carbon cycle variability and long-term  
 13 carbon sequestration potentials of dryland ecosystems. Field data of relevance for testing and  
 14 improving modeled belowground processes will be collected as part of the ARID field campaign  
 15 (Section B.2.3). However, given the challenges measuring these quantities, these goals are  
 16 more aspirational, and we encourage individual research groups to pursue them.

17  
18  
19

*B.2.2.4 Technical modeling and science questions addressed by ARID modeling activities*

20 *Technical modeling related questions we hope to answer within ARID in collaboration with*  
 21 *remote sensing and field data scientists:*

22

- 1 ● Can models capture pulse dynamics (fast plant and soil responses to changing rainfall)?
- 2 If not, what is missing in the models to be able to better simulate this key characteristic
- 3 of dryland ecosystems?
- 4 ● How can we better characterize dryland vegetation and biocrust fractional cover and
- 5 dynamic vegetation change over time (e.g., shrub encroachment)?
- 6 ● How much (species) diversity do we need to account for in order to accurately predict
- 7 surface fluxes and ecosystem resilience in drylands?
- 8 ● Which data (and data characteristics) are needed to constrain dryland model parameters
- 9 and initial conditions via Bayesian data assimilation and novel machine learning-based
- 10 parameter estimation techniques?
- 11 ● Are PFT specific parameters useful in dryland ecosystems, or is spatialization of dryland
- 12 PFT parameters needed (Dahlin et al., 2017)?
- 13 ● In what ways do demography based dynamic vegetation models need to be adapted to
- 14 improve representation of vegetation and biocrust composition, structure and
- 15 competition in drylands?
- 16 ● Can machine/deep learning methods be used to identify patterns and relationships that
- 17 could help improve modeling of fire occurrence, spread and burn severity?
- 18 ● What dryland management practices are currently missing in models? How can we best
- 19 represent these management practices?
- 20 ● Could state data assimilation be used to update dryland carbon stocks and water stores
- 21 following disturbance or management in cases when adequate process-based modeling
- 22 of these processes cannot be achieved?
- 23 ● How can important facets of terrain complexity that exert important controls on dryland
- 24 ecosystem composition and functions be better represented or parameterized in LSMs?
- 25 ● How can groundwater and lateral water flow processes be better represented for
- 26 drylands with complex terrains?

27

28 *Science questions we hope to answer with data-constrained, newly improved models within*  
 29 *ARID:*

- 30 ● Will all dryland regions experience continued declines in water availability, and what will
- 31 be the impact on dryland vegetation-carbon-water interactions?
- 32 ● Will increased drought and fire frequency lead to widespread mortality in dryland
- 33 ecosystems? What would be the consequences for dryland species composition, carbon
- 34 sequestration, soil quality and water retention?
- 35 ● What are the mechanisms of dryland plant adaptation to the changing climate (warming,
- 36 water stress)?
- 37 ● Is CO<sub>2</sub> fertilization causing enhanced productivity and/or water use efficiency in dryland
- 38 ecosystems? If so, are increases in WUE enough to offset declines in water availability
- 39 in drought stressed periods?
- 40 ● Which processes contribute to high inter-annual variability in dryland carbon cycling?
- 41 And which dryland regions are “hotspots” of global carbon cycle inter-annual variability?

- 1 • What are the causes of shrub encroachment in drylands, and are they the same in all  
2 dryland regions? How does shrub encroachment alter dryland productivity and water  
3 use?
- 4 • How do shifting patterns of trees versus grasses impact land-atmosphere coupling?  
5 Conversely, how is changing atmospheric circulation affecting vegetation composition  
6 and ecosystem functioning in drylands?
- 7 • How does topographic complexity (elevation, aspect, slope) serve to buffer or  
8 exacerbate changes in the composition and functioning of dryland ecosystems?
- 9 • How do different dryland management practices impact above- and belowground carbon  
10 stocks, surface and groundwater storage and the surface energy balance?
- 11 • What nature-based solutions can be utilized in drylands to combat dryland degradation?  
12

### 13 ***B.2.3 Coordinated Ground Measurements***

14 In-situ observations of surface states and fluxes aligned with airborne campaigns and satellite  
15 observations are essential to addressing ARID scientific questions. One of ARID's strengths is  
16 the ability to leverage rich networks of available ground measurements of soil moisture, water  
17 fluxes, carbon fluxes, and vegetation states, all of which are targets of remote sensing  
18 platforms. Observational and manipulative experimental data offer direct means to address  
19 long-standing knowledge gaps for dryland ecosystems at representative sites, and for improving  
20 interpretation and use of remote sensing data. ARID will take the essential step of  
21 systematically combining ground measurements with proximal, aerial, and satellite remote  
22 sensing and modeling efforts to quantify dryland functioning and forecast future responses  
23 across dryland regions (MacBean et al., 2021). ARID's approach to incorporate ground  
24 networks will closely follow the blueprint that NASA Terrestrial Ecology Program campaigns  
25 have followed since FIFE and continuing through ABoVE by using distributed field observations  
26 in representative land cover types and across environmental gradients of key forcing variables  
27 (e.g., mean annual temperature and precipitation, time since disturbance). Here, we discuss our  
28 vision and approaches for how existing ground networks and new ground measurements will be  
29 used within ARID to address dryland-specific challenges and opportunities. Note that while  
30 several ground networks are mentioned in this section, a more comprehensive list of ground  
31 networks that will be leveraged in our domains is presented in Section 2.5.

32  
33 Unique challenges in drylands include the ability to monitor key parameters and processes to  
34 understand mechanisms across vast landscapes that have large spatial complexity and  
35 heterogeneity. ARID will leverage the richness of existing infrastructure, including weather  
36 stations, flux towers, manipulative experiments, and other in-situ networks, as well as a wealth  
37 of existing datasets including soil assessments and long-term vegetation monitoring. However,  
38 critical new components proposed here will facilitate effective coordination of ground  
39 measurements with modeling and remote sensing efforts. These include: 1) augmenting existing  
40 sites deployed across identified environmental gradient transects (see potential locations in Fig.  
41 B.11) with proximal remote sensing instruments, and enhanced sensor networks to facilitate  
42 upscaling efforts (Super Sites described in B.2.3.2 below); 2) making use of relocatable flux  
43 towers for short-term campaigns to increase measurements in ecosystem types and

1 disturbance regimes that are not well represented in the existing networks; 3) facilitating  
2 contemporaneous measurements with ground based LiDAR, UAS, and airborne flyover  
3 campaigns to understand spatial scale heterogeneity and link high resolution fractional cover,  
4 thermal drone imagery, and SIF to carbon, water, and energy fluxes with multitemporal and  
5 airborne data at flux towers and Super Sites; and 4) utilizing ground campaigns, experimental  
6 manipulations, and in-situ sensor networks to fill in the gaps between long-term sites along  
7 airborne transects and enhance measurement capability at those sites (e.g., sapflux, tree  
8 growth, soil moisture down to 1 m, belowground carbon, rooting depth, plant water status, and  
9 direct measures of net primary productivity).

#### 11 *B.2.3.1 Vision for Effective Upscaling Strategies and Improving Sensing Capabilities*

12 ARID's ground-based measurements are ultimately the highest resolution and thus most  
13 essential for upscaling our dryland processing understanding across the globe and testing our  
14 remote sensing capabilities. This is especially important for observing drylands' complex  
15 landscapes. Using the ground measurements in our domains, ARID will use several strategies  
16 for upscaling process understanding and sensing techniques. (1) As perhaps the primary  
17 technique, UAS and aircraft fly-overs of in-situ networks, experiments, and particularly ARID  
18 Super Sites, present unique opportunities to calibrate and validate retrieval methods of target  
19 physical parameters using the in-situ knowledge, as well as understand the spatial scales of  
20 representation of the target physical parameters. (2) Gaining an understanding of spatial  
21 variability and representation of target physical variables from these ground measurements  
22 (augmented with UAS and aircraft measurements) presents an opportunity to develop  
23 quantitative frameworks for optimal placement of new sensors in existing measurement  
24 networks. (3) The ground measurements leveraged in ARID present a unique opportunity to test  
25 sensing capabilities and innovative development of proxies for remotely sensing particularly  
26 challenging variables to observe. These variables include evapotranspiration, carbon uptake of  
27 both soil and vegetation, soil carbon, soil types, deeper soil moisture, and other belowground  
28 variables. (4) Experimental manipulation sites provide opportunities to test whether observation-  
29 based approaches with satellites (often using statistical models) can replicate results from  
30 manipulations, such as rainfall mean and variability manipulations (e.g., DroughtNet) (Beier et  
31 al., 2012).

#### 32 *B.2.3.2 Super Sites and Distributed Flux Towers*

33 We propose targeted augmentation of a coordinated subset of existing eddy covariance tower  
34 sites to be designated as ARID Super Sites that are strategically located to provide multi-year,  
35 data-intensive time series across environmental gradients. Most existing flux towers in dryland  
36 sites across the globe make ground-based observations, including local meteorology, land-  
37 surface water vapor, CO<sub>2</sub> and energy exchanges, and soil moisture and temperature states  
38 (Baldocchi et al., 2001; Novick et al., 2018). Sites will be selected from among long-term flux  
39 towers already operating in the following networks to be designated as ARID Super Sites, with  
40 an inclination towards sites with more feasibility of access and that contain a high amount of  
41 instrumentation. These will be chosen from existing networks within the ARID domain including  
42 AmeriFlux, MexFlux, OzFlux, ICOS, NEON, LTER, USDA LTAR, NPS, BLM (AIM), USFS,  
43 NEON, TERN. In addition to standard instrumentation to quantify land-atmosphere exchange of  
44 carbon, water and energy, augmentations at Super Sites will include RGB imagery, hyper/multi-

1 spectral data, thermal data, biocrust and vegetation composition, cover, and structure (Table  
2 B.2 SATM). Given the key role of soil moisture in drylands (Novick et al., 2024), Super Sites will  
3 be augmented with multi-depth soil sensor profiles of volumetric water content and soil water  
4 potential, spatially distributed to capture the dominant cover types of each site. Within the flux  
5 tower footprint, we suggest installing soil autochambers to make continuous measurements of  
6 soil CO<sub>2</sub> efflux to the atmosphere, capturing biocrust, root, soil heterotroph, and inorganic  
7 carbon cycling contributions. We will partition soil exchange of CO<sub>2</sub> among these dryland  
8 sources (i.e., heterotrophs, autotrophs, inorganic carbon) with isotopic assessment, including in-  
9 line measurements of <sup>13</sup>C:<sup>12</sup>C ratios for CO<sub>2</sub>, and targeted measurements of <sup>14</sup>CO<sub>2</sub> to  
10 determine the age and sources of respired carbon, as well as the turnover times of soil organic  
11 carbon. We will assess the biocrust community composition and spectral properties of the soil  
12 within the autochambers to link carbon cycling to the soil microclimate, weather, and remote  
13 sensing data concurrently collected at the Super Site. On soils adjacent to but outside of the  
14 tower footprint, we will manipulate the biocrust cover, for example mimicking climate and  
15 physical disturbance, to quantify the effect on overall soil CO<sub>2</sub> exchange. These in-situ  
16 measurements will be coupled with laboratory incubations that allow assessment of the  
17 mechanisms, vulnerability, and potential future contributions of soil microbial heterotrophs and  
18 soil inorganic carbon cycling to dryland carbon.

19

#### 20 *B.2.3.3 In-Situ Networks and New Relocatable Towers*

21 In addition to regular distribute sites and Super Sites, both modeling and remote sensing efforts  
22 will take advantage of a number of existing in-situ measurement networks in drylands. For  
23 example, these include networks SPECNET (optical measurements linked with surface fluxes),  
24 SNOTEL and SNOWPIX (for snowpack dynamics), the National Soil Moisture Network  
25 (standardized, distributed soil moisture), the Phenocam network (ecosystem greenness), etc.  
26 There are similar in-situ networks in ARID's international drylands domains; for example, Brazil  
27 has a national network of 360 soil moisture and rainfall stations in semi-arid sites.

28

29 To test specific hypotheses, field-based research can greatly benefit from the use of relocatable  
30 towers. Relocatable towers will be used for short-term deployments of diverse instrument suites,  
31 including eddy covariance, spectrometry, and RGB camera systems using scanning point  
32 measurements of user-specified targets (i.e., vegetation, ground, sky, etc.). Relocatable towers  
33 will thereby monitor hyperspectral reflectance in the visible and NIR regions and have capability  
34 to resolve far-red SIF, coupled with RGB imagery (Wong et al., 2023b). These deployments will:  
35 1) assist with mapping key parameters required to address process-based questions and  
36 knowledge gaps across environmental, land use, and disturbance gradients and 2) augment  
37 testing the ability of remote sensing platforms to accurately quantify and represent variability in  
38 these key parameters (e.g., soil water availability, belowground carbon profiles to depths of 1 m,  
39 and rooting depths. (For more detail see section B 2.1.3 Proximal Sensing)

40

#### 41 *B.2.3.4 Manipulative Field Experiments and Long-term Ecological Field Sites*

42 We will leverage existing field-scale manipulative experiments where key environmental  
43 variables such as precipitation amount, timing, temperature, and nutrients are manipulated to  
44 directly test and observe how and why these variables alter dryland function (M. D. Smith et al.,

1 2024). Existing precipitation manipulations in drylands across the US alone include RainMan  
2 (Javadian et al., 2023; Y. Zhang et al., 2023), DroughtNet (Knapp et al., 2017), and Mean  
3 Variance Experiment at the Sevilleta LTER). Augmenting existing manipulations with additional  
4 ground-based remote sensing measurements, (e.g. thermal imagery, spectral measurements,  
5 relationships among surface states and fluxes) will provide a direct link to the airborne and  
6 remote sensing campaigns. These leveraged manipulative field experiments will provide the  
7 means to test the ability of remote sensing platforms to detect and quantify change, and develop  
8 a more process-based understanding for model parameterization/characterization of processes  
9 that are challenging to observe from remote sensing. As learned from extensive scoping with  
10 practitioners, ARID can also coordinate with large land management treatments that can act as  
11 experiments for researchers. For example, land management agencies such as the BLM and  
12 US Forest Service will be applying large scale restoration treatments that can be assessed and  
13 tracked through time. These ‘experiments’ allow scientists and managers to work together to  
14 improve understanding of the drivers of treatment outcomes to advance success in future  
15 management efforts, including in the context of climate change.

16

#### 17 *B.2.3.5 Vegetation Surveys*

18 Within the US and internationally there is a wealth of long-term data assessing plant and  
19 biocrust community composition and cover across time and space in drylands. For example, the  
20 BLM’s Assessment, Inventory, and Monitoring (AIM) program and the National Park Service’s  
21 Inventory and Monitoring program have thousands of sites in the western US where plant and  
22 biocrust data are collected. These existing datasets can be blended with past remote sensing  
23 data and with the collection of new data, including ground-based, UAS, and airplane remote  
24 sensing, to improve the understanding of the controls over and changes to vegetation patterns  
25 and to build predictions for future change. This includes assessment of changes to biodiversity,  
26 increases in the abundance of exotic invasive plants, and losses or gains of functional types that  
27 lend insight into the kinds of vegetation most likely to be successful in a warmer, drier world.  
28 The sites also allow for improvement of remote sensing interpretation, as remotely sensed data  
29 can be ground truthed with vegetation survey data, as well as informing emerging new  
30 opportunities for remotely sensing biocrusts. More information can be found in B.2.5.2.

31

#### 32 *B.2.3.6 Soil Properties*

33 Advancing our understanding of dryland controls and resilience must include consideration of  
34 the roles soils play in regulating ecosystem function and response to change. Soil maps and  
35 field-based studies can join with remote sensing tools to provide a much improved assessment  
36 of soil controls over dryland structure and function. Remote sensing of soils is challenging, and  
37 since drylands have a large coverage of soils, ARID presents a unique opportunity to both  
38 understand their role in dryland processes from rich networks of in-situ measurements as well  
39 as develop and test techniques to remotely sense soil types and soil carbon.

40

41 Although deeper soils and integrated soil carbon stocks are not currently able to be assessed  
42 with remote sensing, we can link remotely sensed to field based soil collections using models,  
43 as well as inform emerging soil-focused remote sensing tools. ARID can strategically conduct  
44 field studies across common land cover types (i.e., grassland, shrubland, woodland) and aridity

1 classes to elucidate climate, plant community, and edaphic controls over soil carbon. ARID will  
2 collaborate with and contribute to ongoing US and international mapping efforts by the USGS,  
3 NRCS, and USDA, and can stratify the landscape according to vegetation cover and  
4 composition, mapped soil characteristics, parent material, topography, surface hydrology, and  
5 disturbance history. An improved understanding of the relationship between these potential  
6 controls and soil organic and inorganic carbon will provide insight into the determinants of soil  
7 carbon and its spatial distribution, over the vulnerability of organic and inorganic carbon to  
8 climate and land cover change, and to the potential management actions that could protect or  
9 even increase soil carbon in drylands. Large scale disturbances and land management  
10 treatments (e.g., wildfire, landscape restoration) could be targeted for assessment, as could  
11 research manipulations (e.g., DroughtNet) and time-since-disturbance/treatments, in order to  
12 reveal the longer-term consequences of change and/or management. To elucidate soil  
13 contributions to dryland ecosystem carbon exchange with the atmosphere, we will use a  
14 relocatable automated soil efflux system that can be deployed in areas that have experienced  
15 relevant change (e.g., after fire) or that are undergoing scientific manipulation (e.g., DroughtNet)  
16 to complement the same system being used in Super Site tower footprints (see Section  
17 B.2.3.1).

18  
19 There are also emerging capabilities for assessing soil organic and soil inorganic carbon using  
20 remote sensing tools. For example, although, remote sensing has rarely been used for  
21 estimating soil inorganic carbon and monitoring its changes - because optical remote sensing  
22 cannot penetrate the land surface and radar-based remote sensing is often too coarse in spatial  
23 resolution (Jarmer et al., 2005) - small-scale exploration in the use of satellite remote sensing  
24 (e.g., Landsat, Sentinel) for regional soil inorganic mapping has recently emerged and does  
25 provide useful new insight for new approaches designed to quantify and monitor soil inorganic  
26 dynamics (Jarmer et al., 2005; Kusumo 2018). ARID can build on these advances with an  
27 explicit assessment of soil carbon at Super Sites and associated linkages to remote sensing  
28 measurements.

#### 30 **B.2.4 Study Domain and Selection Criteria**

31 The ARID field campaign will adopt a strategy for detailed, geographically focused and  
32 comprehensive field, airborne and orbiting data collection in an *Intensive* (threshold) research  
33 region of the western US, with complementary programs at internationally *Distributed* (baseline)  
34 field sites representing global drylands. This strategy will promote the synergies and efficiencies  
35 inherent in the concentration of resources and research teams in the *Intensive* region, while  
36 *Distributed* sites will sample the large range of structural, physiological, bioclimatic, edaphic and  
37 land use diversity in global drylands. Thus, where the *Intensive* sites may be the primary locus  
38 of technological advances in dryland processes, remote sensing science, and modeling during  
39 ARID, the *Distributed* sites will provide opportunities for application, parameterization and  
40 validation of new EO technologies across globally diverse drylands.

41 **Intensive Study Region:** Given the global representativeness of drylands in the western US  
42 (Figure B.11; see below), the western U.S. is ARID's threshold (or required) domain. Thus,

1 ARID science questions and themes (Section B.1) can be largely addressed in this domain if  
2 descoping is necessary. While all shaded areas in Figure B.11 are a part of ARID's domain, we  
3 recommend focused studies in four locations: grasslands, shrublands and dry deciduous  
4 woodlands of the southwest US; rangelands and croplands of the western Great Plains; Mojave  
5 Desert in the Great Basin; and cold, higher elevation deserts of the Mountain West (see  
6 Implementation Plan in Section B.2.6). Given the high density of AmeriFlux, NEON, LTER, and  
7 LTAR sites in the region, we anticipate that airborne campaigns would be anchored by existing  
8 (and potentially new) field data, while sampling the larger heterogeneity characteristic of dryland  
9 regions.

10 ***Distributed International Study Regions:*** To ensure ARID science represents global  
11 drylands, we propose several baseline (desired) study regions, that would provide additional  
12 insight into variability with different vegetation types, a wider range of hyperarid to subhumid  
13 climatic conditions, and possibly different forcing from extreme events (Fig. B.12). We anticipate  
14 that final *Distributed* site selections will occur during the initial planning stages of the ARID field  
15 experiment, based on cost, logistics, and local collaborator contributions. However, we  
16 recommend core baseline regions including four rotating regions in the Chihuahuan and  
17 Sonoran desert of northern Mexico, dry-wet transect of southeastern Australia, desert to  
18 subhumid woodland gradients in southern Africa, and dry regions of the Caatinga, Cerrado and  
19 Gran Chaco in South America (Fig. B.12) (We note that some of the Cerrado is not classified as  
20 drylands, but some parts are, and the ecosystem represents critical transition point for drylands  
21 and thus, based on community input, it is a focal ecosystem). This list is not exclusive, and  
22 international sites will be added based on local interest and PI-led proposals. See our  
23 implementation plan in Section 2.6 for more details.

24 ***Dryland Definition:*** ARID can ultimately support projects across all global drylands. Our ARID  
25 dryland definition is based on aridity index (AI; Wang et al., 2022) of regions where aridity index  
26 (precipitation/potential ET) is less than 0.65, including hyperarid (AI<0.05), arid (AI = 0.05-0.2),  
27 semi-arid (AI=0.2-0.5), and sub-humid (AI=0.5-0.65). Other dryland definitions were explored,  
28 including alternative rainfall and water availability metrics, and classifications based on  
29 dominant vegetation/biome. However, aridity index is the most widely accepted, given its  
30 inclusion of both precipitation and radiation.

31 ***Selection Criteria and Domain Motivation:*** Such an *Intensive* ARID domain that is focused on  
32 drylands of the western US and several distributed international sites is chosen for several  
33 reasons:

34 *Representativeness:* The western US drylands represent global drylands in terms of mean  
35 aridity level. Additionally, it has an extensive representative distribution of semi-arid sites with  
36 some arid and sub-humid regions. Vegetation includes grasses (C3 and C4), shrubs, trees, and  
37 CAM plants. It contains hot-cold gradients from north to south as well as along elevation  
38 gradients that include wet-dry and hot-cold seasonalities. A dry to wet gradient exists generally  
39 from west to east between hyperarid to subhumid ecosystems. Land cover gradients also exist  
40 between rangelands throughout western US eastward into croplands of the dry western half of  
41 the Great Plains. All of these gradients allow for climate changes assessments in evaluating  
42 spatial relationships that can indicate changes in time. The western US also includes the  
43 contribution of mountains to climate both with orographic effects and snowpack impacts on



1 dryland water availability. International sites add several elements. For example, southern Africa  
2 includes gradients between hyperarid and sub-humid tropical woodlands, which is wider climatic  
3 gradient than that observed in the western US. It is fog-driven and highly seasonally rainfall  
4 driven. Australia also provides a testbed for understanding strongly ENSO driven dryland  
5 systems and its contribution to the interannual variability of the carbon cycle.

6 *Previous Measurement/Existing Infrastructure:* The western US includes a vast amount of  
7 existing field network infrastructure and previous measurements and airborne campaigns to  
8 leverage within the ARID campaign. Namely, ARID airborne campaigns can leverage airborne  
9 campaigns conducted by USGS (GEMx), NEON, and NASA (Western Diversity Time Series,  
10 SHIFT) to create multiple overpasses of locations within a year. More detailed coverage of this  
11 topic can be found in Section 2.5. International baseline domains were also selected based on  
12 the existence of measurement infrastructure, interested partners, and ongoing research.

13 *Supporting Personnel:* ARID will support an extensive community in its domestic domain of  
14 U.S.-based researchers, rights holders, and end-users who live in, manage land in, and/or study  
15 drylands. International candidate domains were also selected based on proximity to research  
16 partners that are studying drylands or working with end-users on land management needs and  
17 that would continue to build on the ARID funded work after the campaign has ended.

18 *Community Input:* Community feedback across over 100 engagement events and over 350  
19 written inputs (see Section C) revealed a strong interest in ARID having both domestic and  
20 international domains, with suggestions showing almost equal interest in each. The community  
21 showed broad interest across the dryland western United States as a whole (Fig. B.11),  
22 including many suggestions to consider the Great Basin and Great Plains. The strongest  
23 international interest was in Africa, particularly Southern Africa. Mexico is also of  
24 disproportionately high interest to the community given the Chihuahuan and Sonoran deserts  
25 are shared by the U.S. and northern Mexico.

26 *Feasibility:* Conducting NASA research and field campaigns domestically is simplified by lower  
27 cost, less time, and logistical ease of instrument transportation (relative to international transport  
28 of instruments, personnel, etc.), ease of use of NASA funds domestically on researchers at U.S.  
29 institutions, safety, and high density of airports with feasible air traffic permissions. Several  
30 examples of successful domestic campaigns, and particularly those ARID will leverage, are  
31 discussed in Section 2.5. International sites are also selected partly based on research partner  
32 availability and their ability to augment deployment of NASA campaign components.

33

#### 34 **Further Motivation of ARID Distributed International Domains**

35 *Southern Africa:* In addition to strong interest from the community, southern Africa includes  
36 gradients between hyper-arid ecosystems in Namibia to dry tropical forests in Mozambique.  
37 There are existing efforts by USAID, NASA SERVIR, university partners in Mozambique, and  
38 Conservation International in Botswana to augment sustainable land management practices.  
39 These are opportunities for NASA to assist with co-development efforts. There are also  
40 opportunities to build on former NASA field campaigns (BioSCape) in South Africa. Additionally,  
41 the ARID team has connected with Gobabeb-Namib Research Institute and University of  
42 Namibia leadership in Namibia and Okavango Research Institute in Botswana, providing further

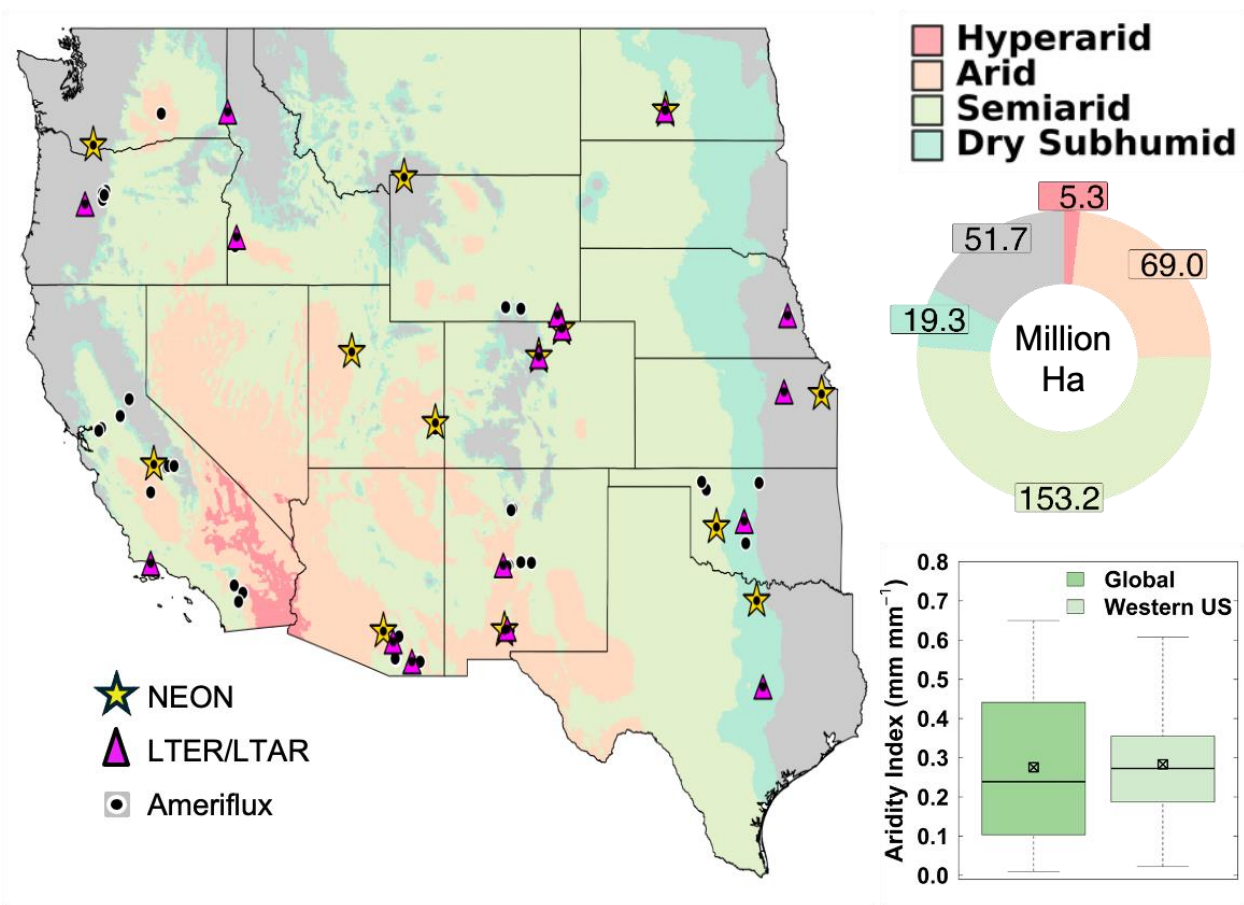
1 opportunities to co-conduct field campaigns with in-place knowledge holders and field sites as  
2 well as contribute to co-training students (see letters of support).

3  
4 *Australia:* Australia is likely a strong driver of carbon cycle interannual variability (Metz et al.,  
5 2023; Poulter et al., 2014) especially with its strong sensitivity to ENSO events. There is also  
6 strong technical ecosystem science experience among Australian researchers and high  
7 potential for co-development of work. For example, ARID team members are well connected  
8 with and have partnered with TERN and OzFlux site networks, as well as government  
9 institutions like CSIRO. Additionally, the ARID team has partnered with leadership of a proposed  
10 Centre of Excellence to be funded through Australia's Research Council focused on effects of  
11 heatwaves on terrestrial ecosystems that would have a similar timeframe and scale as ARID.

12  
13 *Northern Mexico:* Northern Mexico was commonly mentioned by the community to consider  
14 given it shares the Sonoran and Chihuahuan Deserts with the US and might provide an  
15 opportunity for US-Mexico transects. These include gradients between arid desert and semi-arid  
16 croplands. Agencies like Mexico's SAGARHPA and Institute for Forestry, Agriculture and  
17 Livestock Research (INIFAP) assist farming, ranching, and forest industries, and there are  
18 opportunities to assist end-users. Co-development of dryland ecosystem function is likely with  
19 ARID partners including site networks MexFlux and NEON and partners at Universidad  
20 Nacional Autonoma de Mexico (UNAM) and the Instituto Tecnologica de Sonora (ITSON).  
21 There are also opportunities to co-produce with the Mexican National Council of Humanities,  
22 Science and Technology (CONAHCYT), which is a funding ministry for Mexican researchers  
23 that funds networks like Network of Social-Ecological Participatory Observatories (SEPO) and  
24 International Network for Drylands Sustainability (RIZA).

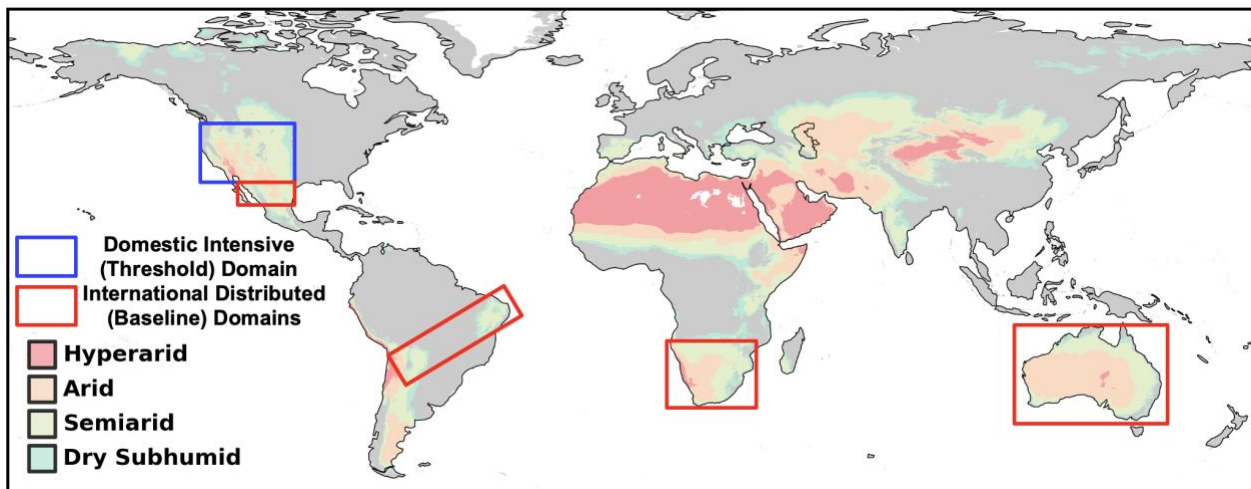
25  
26 *South America:* ARID also received a high proportion of community feedback with written inputs  
27 to include South America, and particularly to include existing efforts in the Gran Chaco,  
28 Cerrado, and Caatinga (i.e., the South American Dry Diagonal). NASA has conducted work  
29 particularly in Brazil during the LBA campaign, but not in drylands east and south of the Amazon  
30 tropical forest. These areas are experiencing dramatic change due to climate and land use  
31 changes. While an arid region was registered for the first time in northeast Brazil, the semi-arid  
32 region is expanding at a rate of 75,000 km<sup>2</sup> per decade (Tomasella et al., 2023). There are vast  
33 ongoing research efforts for ARID to complement including NeoTropTree, DryFlor, SECO,  
34 FAPESP e-phenology, among others, which are established communities and measurement  
35 networks. In fact, in September 2024, ARID leadership held a workshop with about 50  
36 researchers in South America or those who conduct work with these networks to understand  
37 research needs. There was interest in participation of all dryland sites across South America,  
38 which **more than 30% of South America is drylands**, and including research in seasonally  
39 semi-arid regions like the Cerrado which are experiencing increased atmosphere aridity and  
40 land transformation may be a proxy and testbed for understanding expansion of drylands, also  
41 in direction of tropical and subtropical moist forests. Furthermore, ARID has partnered with the  
42 Organization of American States, which will be pivotal for connecting the work ARID does with  
43 end-users and land managers in South America.

44



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**Figure B.11.** ARID's proposed intensive, threshold domain as shown in the colored shading. Most of the region is arid and semi-arid (pie-chart) and its mean aridity is approximately the same as the global dryland mean aridity (see boxplots).



7

1 **Figure B.12.** *Global dryland distribution based on aridity index. Potential ARID*  
2 *field domains are outlined by borders and include shaded dryland regions (based*  
3 *on aridity index) within those borders. A core intensive domain includes shaded*  
4 *drylands within the western US highlighted approximately by the blue borders.*  
5 *Candidate distributed international domains are shown as shaded regions*  
6 *outlined by the red borders. More specific proposed transects and sites are*  
7 *proposed in Section B.2.6. These are ultimately focal areas that ARID will focus,*  
8 *and are not meant to be exclusive. Research and measurements can be*  
9 *proposed and included within ARID outside of these focal areas during the*  
10 *campaign.*  
11

## 12 **B.2.5 Leveraging Existing Field Campaigns and Networks**

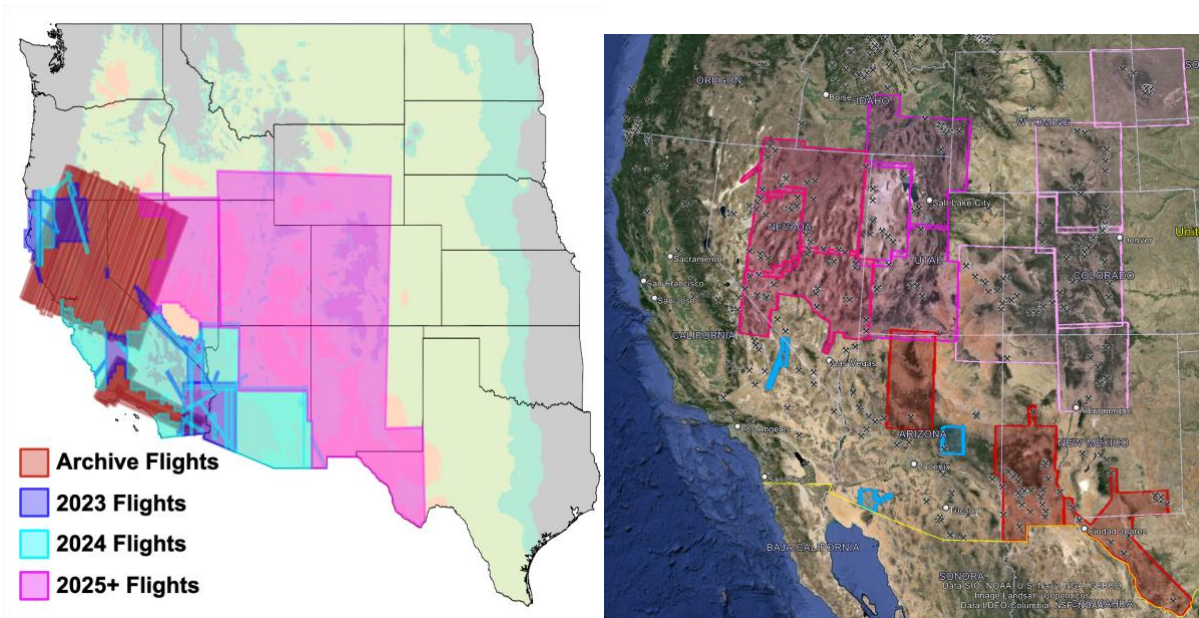
13 A defining feature of the ARID field campaign is a strong ability to leverage previous and  
14 ongoing field (including airborne) campaigns. While NASA and other agencies (i.e., USGS)  
15 have invested in airborne campaigns in drylands, and particularly in the western US, these  
16 datasets have not been used substantially to advance understanding of dryland ecosystem  
17 processes. In fact, the ARID team queried their working group members about their use and  
18 need for airborne measurements (see Appendix F.10 for more details). The 67 members  
19 queried included scientists at varying career stages that have worked to varying degrees in  
20 drylands. Of these researchers, very few had any knowledge of the range of aircraft campaigns  
21 that NASA was involved in (including GEMx, Western Diversity Time Series, SHIFT, etc.). This  
22 might be attributable to the campaigns addressing a specific set of research questions (often  
23 unrelated to dryland ecosystem science), limited community engagement, and/or being focused  
24 on technical readiness level demonstration for upcoming satellite sensors. This also may be due  
25 to the fact that previous NASA airborne acquisitions were not coordinated with field  
26 measurements and optimally timed within a cohesive dryland ecosystem field campaign as  
27 proposed by ARID. Despite the apparent lack of awareness of previous NASA campaigns, all  
28 respondents indicated that future aircraft measurements across VSWIR imaging spectroscopy,  
29 lidar, microwave, infrared, etc. would be useful to their work in dryland ecosystems.  
30 Furthermore, the respondents overwhelmingly indicated that temporal revisits, especially at  
31 seasonal timescales, would be most useful for addressing their research questions. This  
32 feedback creates a strong case for multi-temporal, multi-sensor, strategically located airborne  
33 acquisitions that complement existing and on-going to address ARID science questions. It is  
34 clear that the existing airborne campaigns by themselves cannot sufficiently advance our  
35 understanding of dryland processes.

36  
37 We describe several campaigns here that ARID should specifically leverage in its intensive  
38 domestic domain in the western US and in some cases internationally. These include both  
39 airborne and ground measurement networks. This list is not exhaustive, and we anticipate a  
40 wide array of data will be leveraged within ARID, though we anticipate that the campaigns and  
41 networks mentioned here are critical to leverage to address ARID's research questions. For  
42 example, we anticipate that several AmeriFlux and/or NEON sites will serve as ARID Super  
43 Sites (see Section 2.3).

### 44 **B.2.5.1 Ongoing Aircraft Campaigns ARID will leverage**

45 **Geological Earth Mapping Experiment (GEMx) (2023-2026):** GEMx is an interagency effort  
46 by USGS and NASA to collect VSWIR imaging spectrometer and multispectral infrared data for  
47

1 surface mineral identification and mapping across a large portion of the western US using  
 2 NASA's ER-2 high altitude aircraft. GEMx runs through 2026 with funding up to \$4M per year  
 3 provided by the USGS Earth Mapping Resources Initiative (Earth MRI). Flights occur between  
 4 April and September each year. GEMx goals include improving knowledge of the nation's  
 5 geology, in particular critical mineral resources. Sensor's flown have been AVIRIS-Classic and  
 6 MASTER (MODIS/Aster Airborne Simulator) in 2025 and 2026, the sensor suite is expected to  
 7 include AVIRIS-3. Lower altitude collections of AVIRIS-NG and HyTES (Hyperspectral Thermal  
 8 Emission Spectrometer) are also possible in the coming years. GEMx activities complement the  
 9 Earth Surface Mineral Dust Source Investigation (EMIT) mission. To date, 480,000 sq. km of the  
 10 western US have been covered, with plans for further coverage throughout 2025-2026 across  
 11 all of the southwestern US states (Fig. B.13). Given that GEMx is ongoing, and GEMx  
 12 leadership is also on the ARID leadership team (Raymond Kokaly), it is possible to plan timing  
 13 and location of GEMx flights to benefit both ARID's and GEMx goals.  
 14



15  
 16 **Figure B.13. (A)** left, GEMx flight coverage compared to ARID domain using  
 17 AVIRIS-Classic and MASTER instruments. **(B)** right, Prospective GEMx  
 18 coverage in 2025 to 2027 with red boxes indicating flights for April to June 2025,  
 19 dark magenta boxes (NV) for flights in June-Sep 2025, and pink boxes for flights  
 20 in 2026/2027. Blue boxes include a re-fly in 2025 for areas that were cloudy.  
 21

22 **Hyperspectral Infrared Imager (HyspIRI) and Western Diversity Time Series (WDTS)**  
 23 **(2004-present):** HyspIRI was a NASA satellite mission to study global vegetation state and  
 24 monitor natural disasters (drought, volcano, wildfire) with imaging spectrometer and  
 25 multispectral infrared data. It has included several airborne campaigns since 2004 with AVIRIS-  
 26 Classic and MASTER mainly in southern California and parts of Nevada (Lee et al., 2015). As a  
 27 continuation, WDTS is a NASA field campaign in California starting in 2013 and continuing to  
 28 provide critical information on natural disasters like volcanoes, wildfires, and droughts. The  
 29 initial HyspIRI-based airborne campaign was mainly used to demonstrate technical readiness  
 30 for upcoming satellite missions (Lee et al., 2015). WDTS is aiming to establish a time series of  
 31 repeated collection in areas of the western US. To date and to our knowledge, these data have  
 32 not yet been widely used to address dryland science questions as proposed by ARID. Flights  
 33 include MASTER, AVIRIS-Classic, HyTES, and PICARD sensors flown on ER-2 aircraft. The

1 ARID team connected with WDTS leadership at University of Wisconsin (Prof. Phil Townsend)  
2 and, in stating a need for more scientific use of the datasets and ongoing dataset development  
3 efforts, they encouraged the use of the datasets within ARID.  
4

5 **Goddard's LiDAR, Hyperspectral & Thermal Imager (G-LiHT) (2011-Present):** G-LiHT is a  
6 NASA airborne campaign ongoing since 2011 including many flight paths across the western  
7 US (Cook et al., 2013). It measures Lidar, thermal, and VNIR imaging spectrometer data. With  
8 ongoing funding through NASA Headquarters, it is expected to continue into the timeframe of  
9 the proposed ARID field campaign.

10  
11 **FireSense (2023-2028):** FireSense is an annual NASA airborne campaign starting in late 2023  
12 to improve US wildfire management. Previous and planned flights include AVIRIS-3, MASTER,  
13 UAVSAR, and SLAP throughout the western U.S mostly in Colorado, Arizona, Utah, and  
14 California.

15  
16 **NEON:** These are NSF funded sites operated by Batelle, with many sites throughout the US  
17 (Fig. B.11). These sites include associated aircraft flights during peak greenness at all sites  
18 annually or bi-annually including from hyperspectral, thermal, and LiDAR instruments. Flight  
19 schedules can be found at [https://www.neonscience.org/data-collection/flight-schedules-](https://www.neonscience.org/data-collection/flight-schedules-coverage)  
20 [coverage.](https://www.neonscience.org/data-collection/flight-schedules-coverage)

21  
22 **Other noteworthy airborne campaigns:** There are other campaigns ARID can leverage in  
23 specific cases including WHyMSIE airborne campaign includes flights across the western US  
24 and Great Plains in 2024, SMAPVEX flights in 2015 in Tucson, AZ, the SBG SHIFT campaign in  
25 California in 2022, FarmFlux across the Great Plains and western US beyond 2027, and  
26 upcoming Catalyst. More information about other campaigns is provided in Appendix F.12.

#### 27 *B.2.5.2 Field Networks ARID will directly leverage*

28  
29 **NEON:** These sites provide typically a range of field measurements including water and carbon  
30 fluxes, soil moisture, temperature, and radiation. There are approximately 13 of these sites  
31 across the western U.S. that ARID can leverage.

32  
33 **GERI:** Leading international partners institutions (e.g. TERN - Australia, SAEON - South Africa)  
34 are members of Global Ecosystem Research Infrastructure ([GERI](#)), along with NEON. Together,  
35 they are coordinating research on global ecological drought.

36  
37 **AmeriFlux:** These sites provide a standardized set of measurements including water and  
38 carbon fluxes, soil moisture, temperature, humidity, wind speed and radiation across the U.S.  
39 (Fig. B.11). With AmeriFlux as an ARID partner, obtaining datasets from site PI is more  
40 streamlined and there are opportunities to install new instruments at these sites within the ARID  
41 campaign. Of particular interest are use of the net ecosystem exchange and evapotranspiration  
42 measurements which can be used to develop and scale remote sensing products of these  
43 quantities.

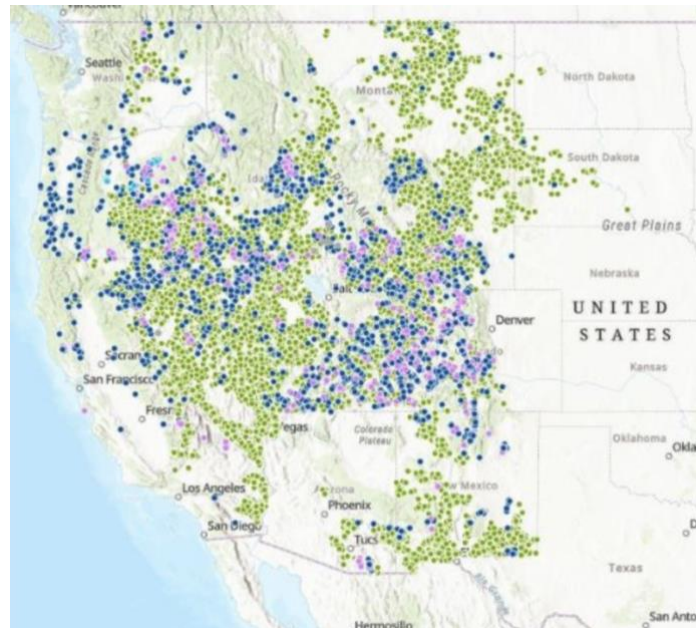
44  
45 **LTAR/LTER:** These sites are often coupled with AmeriFlux towers, but also integrate strategic  
46 research, coordinated experimentation, and common measurements to address ecological and  
47 agricultural challenges within and across major US ecoclimate zones. LTAR sites particularly  
48 focus on agricultural lands including croplands, grazing lands, and integrated agroecosystems.

49  
50 **NOAA Fire Prediction Sites:** NOAA received funding to set up four fixed field sites in Flagstaff,  
51 Arizona, Berkeley, California, Boise, Idaho, and Gunnison, Colorado. These sites will include

1 soil moisture and temperature, doppler and LiDAR, eddy covariance, wind speed, and others.  
2 They will also include some UAS flights potentially annually. ARID has partnered with this effort  
3 and there are opportunities to collaborate by adding additional measurements at these sites  
4 based on the leasing arrangements at each site.

5  
6 **Snowtopography:** These sites were designed to monitor a wide variety of ecohydrologic factors  
7 including daily snow depth, snow water equivalent, soil moisture, etc. along gradients of forest  
8 management including mechanical thinning, commercial harvest, and managed fire. This  
9 network is ideally suited for quantitative evaluation of the impacts of management on forest  
10 ecohydrologic processes and physical parameters.

11  
12 **BLM-AIM:** The Bureau of Land Management Assessment Inventory and Monitoring (BLM-AIM)  
13 program uses standardized ground-based measurements to assess the natural resource  
14 conditions and trends on BLM-managed public lands (Fig. B.14). The AIM Strategy provides  
15 quantitative data and tools to guide and justify policy actions, land uses, and adaptive  
16 management decisions and offers a wealth of data available to ARID, which could be joined with  
17 airborne and spaceborne data. AIM has currently assessed over 20,000 dryland sites that could  
18 be leveraged for AIM.



20  
21 **Figure B.14.** Map of BLM AIM sites. There are currently >50,000 BLM AIM sites  
22 in the western US.

23  
24 **USFS FIA:** Many Forest Inventory and Analysis (FIA) sites are in dryland forests, such as  
25 pinyon-juniper woodlands and dry mixed conifer forests, which represent the most extensive  
26 forest cover types in the U.S. The FIA annual inventory provides data on the extent, condition,  
27 volume, growth, depletion, and health of US forest resources. FIA Plots consist of one field  
28 sample site - designed to cover a 1-acre sample area - for every 6,000 acres of forest. Common  
29 variables measured regularly across all FIA plots include diameter at breast height (DBH), total  
30 height, age, mortality, and crown base height, which can be used for diameter growth increment  
31 analysis and biomass assessment through allometric relationships. Additional measurements  
32 including tree increment cores, downed woody material (DWM), soil chemistry, and understory

1 plant composition are collected for a subset of plot and can be used to assess ecosystem-level  
2 carbon pools and fluxes, forest health and sensitivity to climate, fuels and fire hazard, and  
3 trends in forest health, insects and pathogens, and invasive species (Tinkham et al., 2018).  
4

5 **NPS I&M:** The U.S. National Park Service maintains 32 Inventory and Monitoring networks  
6 across the country, many of which are in drylands. NPS I&M uses standardized inventory and  
7 monitoring to track natural resources through time - including plant cover, composition, diversity  
8 - and these ground-based data represent a strong potential link with remote sensing data,  
9 models of ecosystem change, and assessments of the accuracy of satellite sensors in capturing  
10 shifts in plant communities and associated resources through time.

11  
12 **NRCS-NRI:** Natural Resources Conservation Service National Resources Inventory (NRCS  
13 NRI). The NRI is a statistical survey of land use and natural resource conditions and trends on  
14 U.S. non-Federal lands, offering a powerful inventory complement to BLM-AIM, USFS-FIA, and  
15 NPS-I&M, which focus on federal lands. The NRI collects and produces information on the  
16 status, condition, and trends of land, soil, water, and related natural resources on the nation's  
17 non-federal lands.  
18

19 **Landscape Data Commons:** The Landscape Data Commons (LDC) is an inter-agency  
20 monitoring data repository and portal, led by the Jornada Experimental Range in collaboration  
21 with USDA-ARS and New Mexico State University, that connects standardized monitoring data  
22 to analysis tools to support land management and research. The LDC aggregates and  
23 harmonizes core methods and data collected across agencies and monitoring programs (e.g.,  
24 BLM-AIM, NPS I&M, NRCS grazing land on-site program, the National Wind Erosion Research  
25 Network, and smaller research and monitoring efforts). There are currently 85,000 locations  
26 housed within the LDC. With these aggregated data, the LDC supports natural-resource  
27 management, modeling, and research.  
28

29 **National Wind Erosion Research Network:** The National Wind Erosion Research Network  
30 (NWERN) is a multi-partner network established in 2014 that collects standardized  
31 measurements of aeolian sediment transport, dust emission fluxes, meteorological conditions,  
32 and soil and vegetation. NWERN vegetation inventory and monitoring methods follow those  
33 implemented by the BLM AIM and NRCS NRI programs, enabling aeolian sediment transport  
34 processes to be linked to dryland ecosystem processes, services and management. NWERN is  
35 currently the only network collecting long-term, standardized aeolian process data to support  
36 dust model parameterization across ecosystems.  
37

38 **CrustNet:** In 2023, the National Science Foundation funded a global network focused on  
39 biological soil crusts: CrustNet. Biocrust researchers from around the world will assess biocrust  
40 and plant community composition and function in drylands and will provide these data to a  
41 centralized data repository that can be linked with ARID. Although the collections of spectral  
42 data are not planned for CrustNet, ARID and CrustNet could each be benefitted and leveraged  
43 by using the same communities and samples to gather remote sensing data that could then be  
44 scaled using satellite and modeling approaches.  
45

## 46 ***B.2.6 Proposed Field Campaign Implementation Strategy***

### 47 ***B.2.6.1 Field Campaign Strategy Overview and Motivation***

48 Here, we develop an ARID implementation strategy within the domains defined in Section B.2.4  
49 and leveraging existing networks and campaigns in Section B.2.5 to address dryland science



1 themes and questions in Section B.1. We motivate and outline an overarching strategy in  
2 several focal areas within the ARID domains. However, we emphasize that the campaign plan is  
3 flexible and expect a variety of approaches will advance the understanding and sensing  
4 capabilities of dryland ecosystems. This section focuses on addressing foundational science  
5 questions, while B.2.7 expands the approach to applied sciences relevant to NASA Earth Action  
6 programs.

7  
8 The guiding principles of the field campaign strategy include but are not limited to:

9 (i) The overarching dryland science questions laid out in Section A.2 and expanded science  
10 themes in Section B.1

11 (ii) Multi-temporal airborne acquisitions needed to address the science questions

12 (iii) Scaling and development needs for remote sensing observations and modeling techniques

13 (iv) Leveraging existing assets (data and infrastructure) and investments by partners, and  
14 particularly supersites (see Section 2.5)

15 (v) The safety and security of team members and affiliated scientists

16 (vi) Optimal location of US and international focal areas that bring the most science return and  
17 potential for co-development with partners while being feasible and low risk.

18  
19 *High spatio-temporal sampling with multiple sensors:*

20 Given the high spatial heterogeneity of drylands and rapidly changing conditions, the ARID field  
21 campaign strategy must accommodate sampling across spatial, temporal, and spectral scales.

22 This requires integration of a broad range of sampling strategies that allow for not only capturing  
23 of heterogeneity, but also synthesis of approaches and datasets for high interoperability and

24 advances in our understanding of drylands. The optimal field implementation approach will  
25 depend on the science themes being addressed. For example, high spatial sampling is

26 necessary for vegetation structure and biodiversity themes. High temporal sampling is

27 necessary for pulse dynamics, drought, and fire themes. A range of spectral measurements are  
28 necessary for understanding interactions between themes (water-carbon coupling). High spatial

29 scaling must include use of in-situ, drone, aircraft, and satellite measurements, underscoring the  
30 relevance of this effort as a NASA Terrestrial Ecology Field Campaign. Upscaling is especially

31 important for integration with gridded models that predict global climate and contribute to ARID's  
32 central goal of estimating dryland contributions to water, carbon, and energy cycles. Specifically,

33 dryland ecosystems are highly spatially heterogeneous with high vegetation biodiversity, which  
34 requires measurements at high resolution (<1m) and linkages with larger scales. In addition to

35 capturing dryland structure at the necessary scale, assessments at these scales can provide  
36 insight into satellite remote sensing observation capability and algorithm development.

37 Temporal scaling requires capturing diurnal to interannual timescales to characterize the rapid  
38 response of drylands to weather and climate. Drylands are driven by individual moisture pulses,

39 and thus field campaign measurements can blend satellite remote sensing assessments with a  
40 focus on these pulse-drydown sequences using complementary ground-based and airborne

41 approaches. Finally, a range of remote sensing modalities and scales are needed, with

42 microwave sensing providing information on water availability and carbon stocks, hyperspectral  
43 on ecosystem composition and soils processes, and LiDAR on ecosystem structure. In this way,

44 ARID's aspiration is to create a new understanding that is multi-scale, multi-sensor, and

1 hypertemporal, joining new and emerging techniques and process models to assess  
2 heterogenous, dynamic drylands.

3  
4 *ARID Super Sites Strategy:*

5 Some of the challenges of high spatiotemporal sampling needs of drylands can be overcome  
6 with leveraging *Super Sites* (Fig. B.11). Previous NASA field campaigns have demonstrated the  
7 need for collocations of field plots, instrumentation (e.g., flux towers and proximal sensing), and  
8 UAS surveys across projects so that data may be properly integrated to effectively address  
9 complex science questions. Super Sites are high priority field sites, with a rich diversity of  
10 historic field data and infrastructure, field experiments that chronicle long-term management  
11 practices and treatments, as well as archived and ongoing airborne data (e.g. NEON and LTER  
12 sites). They also provide measurements of challenging parameters to estimate with remote  
13 sensing, including carbon and water fluxes. In many cases, these Super Sites will form the core  
14 locations of ARID airborne flights and effectively create a hotspot of scientific collaboration.  
15 Experience from the ABoVE team suggests that such Super Site locations should be very  
16 carefully planned (especially those to leverage) at the outset of the field campaign to encourage  
17 individual projects to focus field effort to these locations.

18  
19 *Multi-temporal Airborne Acquisition Strategy:*

20 ARID identified a community-motivated need for multi-temporal airborne sampling in order to  
21 understand the dynamics of drylands, and to address our science themes. Of particular interest  
22 are temporal revisits for evaluation of ecosystem responses to disturbance, seasonal changes,  
23 and rain pulses and drought. Several multi-temporal sampling strategies can be employed  
24 during ARID:

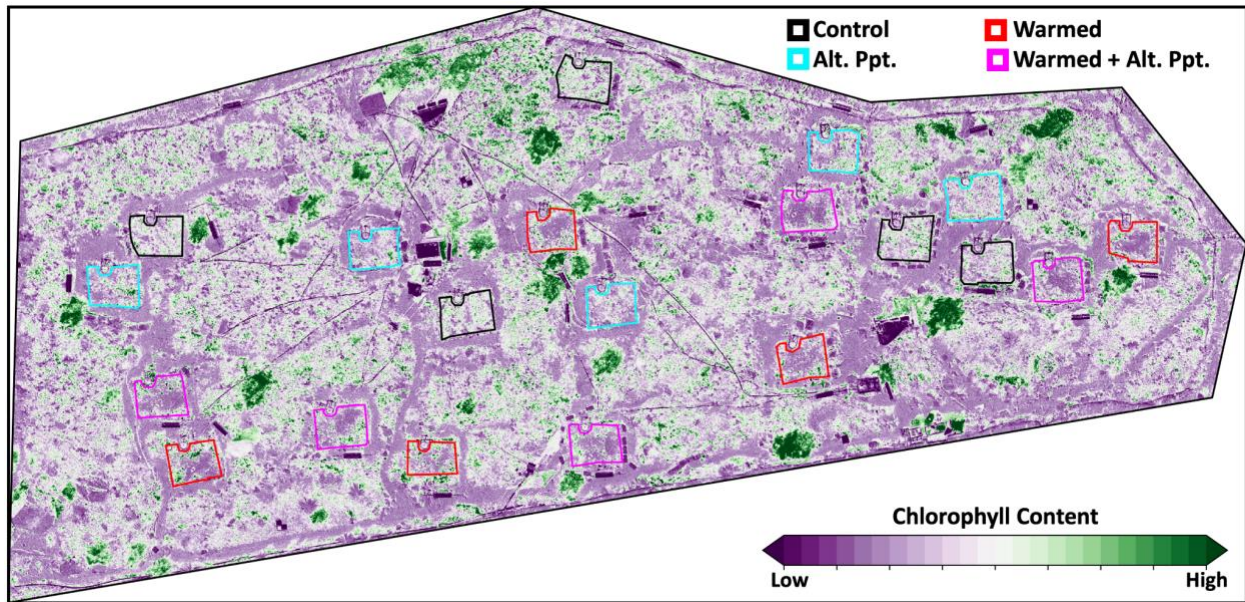
- 25 • Pulse chasing: make aircraft measurements immediately after rain event or at the peak  
26 of an extreme event such as drought/heatwave over ARID Super Sites
- 27 • Seasonal Sampling: remeasure the same location in different phases of growing season  
28 (e.g., repeated sampling of NEON sites that are traditionally sampled only at peak  
29 greenness)
- 30 • Fusion approach: Drones can be used to resample on demand at a specific temporal  
31 cadence across Super Sites within the bounds of a single aircraft overpass

32  
33 To create multi-temporal measurements, ARID will take advantage of existing assets (see  
34 Section 2.5) to augment the aforementioned acquisition strategies, especially for building time  
35 series and wide spatial extents of measurements. For example, ARID will conduct new aircraft  
36 measurements over previously-sampled locations to identify ecosystem changes, coordinate  
37 aircraft measurements with ongoing campaigns to capture pulses or seasonal changes (GEMx,  
38 NEON, FireSense). As such, a time series can be built by leveraging existing NASA, USGS,  
39 and other investments.

40  
41  
42 *Strategy to Leverage Manipulation Experiments:*

43 There is a wealth of in-situ climate and land-use experiments in drylands, as well as myriad  
44 large scale land management activities that offer exceptional opportunities for addressing

1 ARID's science themes and questions. For example, NutNet, DroughtNet, and Dragnet are  
2 globally distributed manipulation experiments in grasslands that use standardized methods to  
3 alter nutrient inputs and forage consumption, precipitation amount, and land disturbance,  
4 respectively (*DRAGNet*, n.d.; *Drought Network*, n.d.; *Nutrient Network*, n.d.). In the US alone  
5 there are also dryland warming experiments (Fig. B.15), grazing experiments, and experiments  
6 that simultaneously explore the interacting effects of multiple global change drivers. These  
7 experiments were designed to simulate global change to assess the mechanisms behind shifts  
8 in plant and soil community composition, carbon storage and flux, and factors related to the  
9 numerous ecosystem services provided by drylands. They now offer ARID unprecedented  
10 opportunity to use remote sensing and ground-based tools to improve the capacity to remotely  
11 sense, scale, and model the effects of global change across larger dryland areas. The unique  
12 power of an ability to "see" what this change looks like in controlled settings where background  
13 soil, climate, and atmospheric conditions can be kept constant. In addition to research  
14 experiments that can be leveraged to improve our understanding and remote sensing tools of  
15 dryland change, the large observable land management actions taken in drylands offer an  
16 opportunity to quantify and predict land treatment success and failure. For example, over the  
17 past two years, the Bipartisan Infrastructure Law (BIL) and Inflation Reduction Act (IRA) have  
18 provided management agencies billions of dollars, much of this going to drylands of the western  
19 US for water, climate resilience, and habitat restoration activities. These recent and planned  
20 activities represent the potential for an extremely strong synergy between ARID and US land  
21 management agencies, as managers are asking for partners and tools with which to assess the  
22 outcomes from this substantial investment in dryland systems, and as these treatments are  
23 large-scale experiments that can be learned from and leveraged. ARID research can help learn  
24 from the outcome of these treatments in support of future management, and the work these  
25 agencies do on the ground offer an unprecedented opportunity to improve our understanding  
26 and ability to measure ecosystems and change at the landscape scale. Our manager partners  
27 are willing to time and to design these actions in collaboration with ARID PIs to facilitate this co-  
28 production of actionable science.  
29



1  
2  
3 **Figure B.15.** An example integration of remote sensing measurements into a  
4 long-term climate change experimental manipulation. The base map of  
5 chlorophyll content at 10-cm spatial resolution was generated from UAS  
6 multispectral data. Experimental plot replicates are shown as colored polygons  
7 overlaying the map. Visual inspection reveals that warming greatly reduced  
8 chlorophyll content across these biocrust-dominated experimental plots (Phillips  
9 et al., 2022).

10 **B.2.6.2 Focal Area Selection Criteria**

11 The selection of focal areas within our domains considers multiple factors, including:

- 12  
13  
14  
15  
16  
17  
18  
19  
20  
21  
22  
23  
24
1. Ability to contribute to addressing ARID science themes
  2. Representation of diverse dryland environments and environmental gradients, or locations that fill a specific knowledge gap
  3. Experiencing the most pronounced climatic events (e.g. extended drought) or change
  4. Experiencing a historical disturbance or management regime so as to serve as macro experiments
  5. The existence of established field sites with long-term data and instrument infrastructure (e.g. NEON/AmeriFlux sites, LTER/LTAR, BLM-AIM sites)
  6. Co-location with on-going airborne campaigns and/or with long-term airborne data (e.g. NEON sites)
  7. Containing priority locations identified by end-users and Tribal Authorities

25 In addition to the above criteria, international focal sites also consider the following criteria:

- 26  
27  
28  
29  
30  
31  
32
1. Expanding the global dryland representativeness for data and models
  2. Addressing ARID science questions from a global perspective
  3. Having an existing track-record of dryland research by local partners (e.g. SAEON, TERN, MexFlux)
  4. Containing compatible airborne assets, and capability of international partners and commercial service providers to conduct airborne surveys if needed

- 5. Availability of logistic support and infrastructure for fieldwork
- 6. Having high levels of safety for researchers during field work

Within our western US and international domains, we propose several candidate focal areas: Four in the western US and four international areas in Mexico, southern Africa, Australia, and South America. We propose a rotating field sampling strategy, where one domestic and one international focal area is intensively sampled for an approximate two-year period, and rotating to a different site for the next ~two year period and so-on (Table B.4; see Section B.2.9 for information on potential coordination with other field campaigns). However, scientific activities that focus on each of these areas should be continued throughout the duration of the ARID campaign.

**Table B.4.** Rotating field campaign strategy with a new domestic and international focal region every approximately two-year period. The ordering of domains is not in order of priority and can be shifted based on feasibility.

	Year 1/Year 2	Year 3/Year 4	Year 5/Year 6	Year 7/Year 8
Domestic Domain	Southwest US	Great Plains	Great Basin	Mountain West
International Domain	Mexico	Southern Africa	Australia	South America

We note that this modularity allows for descoping as necessary (see Section B.2.9). At a minimum, a western US-only domain can accomplish many of our goals and science questions. The western US is nearly identical to the global mean of aridity, and we can scale the new collective understanding to other drylands. However, we emphasize the high scientific value our international sites provide in scaling to global domains, including more representation of drier hyperarid and wetter sub-humid biomes, different climate forcing and variability conditions, different vegetation types, and different natural resource management challenges. These international domains also allow deeper insights into several of our science themes by providing a wide range of data points and testing a wider variety of modeling approaches. More details about the value of these domestic and international focal areas follow.

### B.2.6.3 Domestic Focal Areas

Here, we motivate our four focal areas and provide example studies to be conducted within the campaign to address ARID’s research questions.

**Southwest:** The Southwest focal area includes Arizona, New Mexico, and southwestern Colorado (Fig. B.16). A key feature of this region is extensive field sampling and strong in-place knowledge by university, tribal, and government-based partners. The region includes existing in-situ networks with dense coverage between AmeriFlux, LTER/LTAR, NEON, and Snowtopography sites. Additionally, existing aircraft measurements include GEMx aircraft coverage in 2023 and 2024 (scheduled to continue into 2026) and several G-LiHT flights between 2011 and present. Therefore, there is high potential to scale understanding of dryland processes between in-situ, aircraft, and satellite scales. The region also offers a southwest-to-northeast gradient of hot to cold conditions (with elevation gain) and transition from summer-dominated to winter-dominated seasonal precipitation (Fig. B.17).

Example studies:

- 1) *Vegetation Structure Heterogeneity Tests*: The region's highly heterogeneous landscape with mixture of grassland, shrubland, woodland, and forest offers a testbed for evaluating remote sensing indices' ability to quantify vegetation structure at finer spatial scales and scale up to that of satellite observation resolutions. Single aircraft overpasses over instrumented supersites (NEON, AmeriFlux, LTER sites) are particularly useful for meeting this goal.
- 2) *Tribal Land Management*: Given strong tribal interest in water availability at site scales, UAVSAR and NISAR measurements can be used to both monitor soil water availability but also map spatial distributions of soil moisture at high resolution (<200m) across managed tribal lands.
- 3) *Water Availability*: High resolution soil moisture sampling from in-situ networks and microwave aircraft measurements (UAVSAR, SLAP) can provide insight into the spatial distribution of soil moisture, particularly with regard to proximity to higher vegetation density and riparian zones. It also can provide insight into the effect this soil moisture heterogeneity has at NISAR's 200m scale and SMAP's ~30km scale.
- 4) *Drought and Fire Disturbance*: With high potential for drought and fire disturbance in the region, ARID can include AVIRIS flights after such a disturbance, while using GEMx AVIRIS 2023-2024 measurements as a baseline for change detection.
- 5) *Pulse Chasing*: With strong summer monsoonal effects in the Southwest US, pulse chasing is particularly useful for capturing vegetation and soil pulse responses to large rainfall events. Both in-situ field sampling teams and airborne teams can begin sampling after a large rainfall event to monitor both vegetation (pulse-reserve) and soil (birch effect) responses. Such responses are particularly important for model integration.

**Great Plains**: This focal area extends from northern Texas to the northern Great Plains in South Dakota, as well as between eastern Colorado to western Nebraska and Kansas. The goal for this region is to address rangeland and cropland management questions, given a gradient between rangelands of eastern Colorado and croplands throughout the midwestern states in the domain. Airborne assets to leverage include the upcoming FarmFlux campaign (post 2027) with carbon and nitrogen flux measurements, NEON flights at Colorado and Oklahoma sites, and upcoming WHyMSIE planetary boundary layer flights. In-situ assets include Mesonet soil moisture sites, especially throughout Oklahoma and Nebraska.

Example studies:

- 1) *Rangeland and Cropland Management*: AVIRIS flights are particularly useful for fractional cover of plant functional types (PFTs) and biomass, which are variables that are useful for BLM's decision-making tools.
- 2) *Cropland Carbon Uptake Contribution*: Given potential for CARAFE flights in this region and as a part of ARID, there is an opportunity for quantification of carbon fluxes for specific vegetation types throughout croplands and rangelands. Such flights can be coupled to flux tower measurements such as from NEON for scaling of the site and aircraft data to drive remote sensing-based carbon flux estimates.
- 3) *Land-atmosphere Interactions*: The dry to wet gradient from west to east in this region is an opportunity to test the contribution of soil moisture to dryland ET and consequent convection development in this region. Possibly, there is less potential for convection initiation further west where the ET magnitude is lower. AmeriFlux and NEON towers, as well as potential lower atmosphere aircraft measurements from WHyMSIE, allow testing of these hypotheses from drier to wetter sites.

**Great Basin/Mojave**: This focal area is recommended mainly from southern Idaho through Nevada and extending into hyper-arid regions of central California and the Imperial Valley. The

1 Great Basin is a notoriously undersampled area and a goal for ARID is to provide more field and  
2 airborne sampling in this region. Under climate change, the Great Basin and the Mojave deserts  
3 are experiencing exceptional increases in fire frequency, invasive grasses, and woody plant  
4 encroachment, with severe implications for carbon storage, endangered species, dust  
5 production, and grazing (J. T. Smith, Allred, Boyd, Davies, Jones, et al., 2023; J. T. Smith et al.,  
6 2022). Therefore, understanding fire regimes is a focal point in this region. GEMx airborne  
7 measurements have been made in 2023 with plans for GEMx measurements in 2025 in this  
8 region and potential FireSense measurements through 2028. These serve as a baseline for  
9 change detection if VSWIR imaging or thermal measurements are made in future cases post  
10 2028. There is also potential for the NASA Catalyst biodiversity effort to take place in this  
11 region. We acknowledge that southern Nevada may be infeasible to sample because of military  
12 airspace restrictions, but ARID has built strong partnerships with the Department of Defense  
13 and there is a likelihood of permission, at least with ground-based sampling.  
14

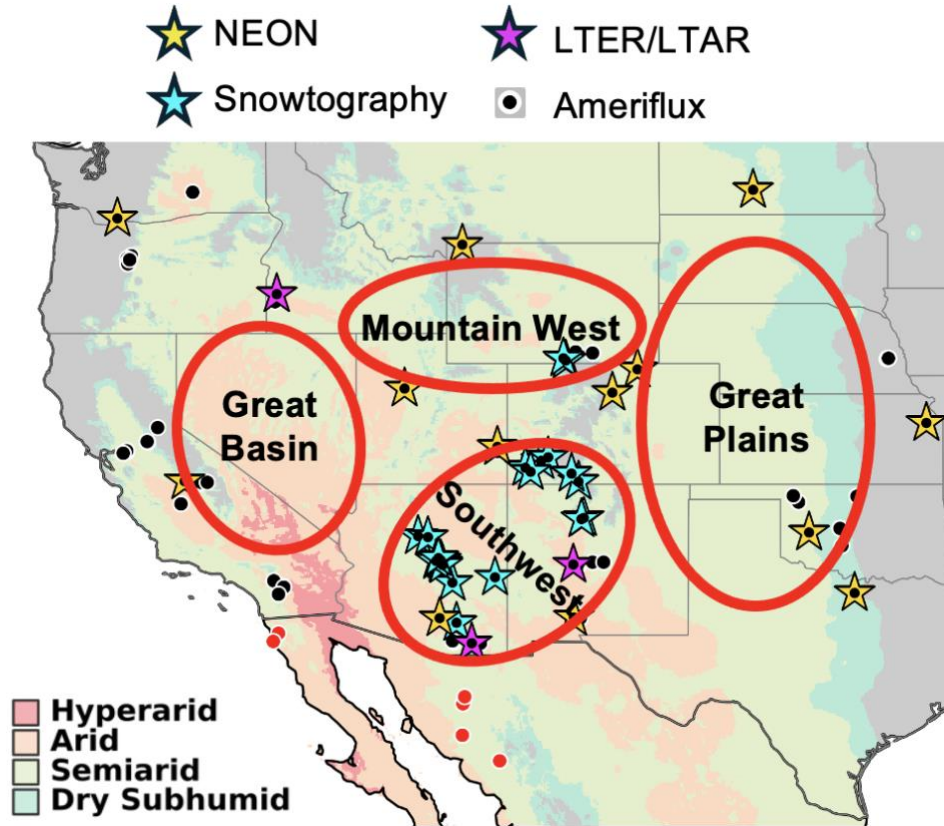
15 Example studies:

- 16 1) *Invasive Species and Fire Management*: AVIRIS flights are particularly useful for  
17 vegetation type identification and fraction of vegetation cover, which are variables that  
18 are exceptionally useful for land manager decision making tools (e.g., BLM). Multiple  
19 aircraft measurements in the same location provide an opportunity to evaluate post-fire  
20 impacts and/or change in species composition (native versus invasive vegetation), as  
21 well as the ability to assess the efficacy of post-fire management actions. Such  
22 measurements coupled with ground information (such as surveys from BLM and USFS)  
23 will allow testing of high-resolution satellite remote sensing (from Landsat, EMIT, and  
24 SBG) to characterize vegetation coverage and species types.
- 25 2) *Soil and Geology*: VSWIR imaging spectroscopy from AVIRIS flights (GEMx) and EMIT  
26 measurements allow characterization of soil types, minerals, and biocrust across these  
27 lower vegetation cover deserts. This can be linked with ground-based data collections,  
28 such as through ARID's partnerships with NRCS, ARS, and NSF's CrustNet.  
29

30 **Mountain West:** This focal area consists of higher elevation, cold drylands across Idaho,  
31 Wyoming, and Utah (Figure B.17) with a gradient of hot to cold semi-arid environments from  
32 west to east. It includes Snowtopography and AmeriFlux sites, as well as G-LiHT flights, though  
33 this region is generally more sparsely sampled. This region will provide insights into the effects  
34 of lower temperatures on dryland vegetation function, differences of dryland vegetation structure  
35 at higher elevations, and effects of water sources (soil water versus snowmelt) on ecosystem  
36 function.  
37

38 Example studies:

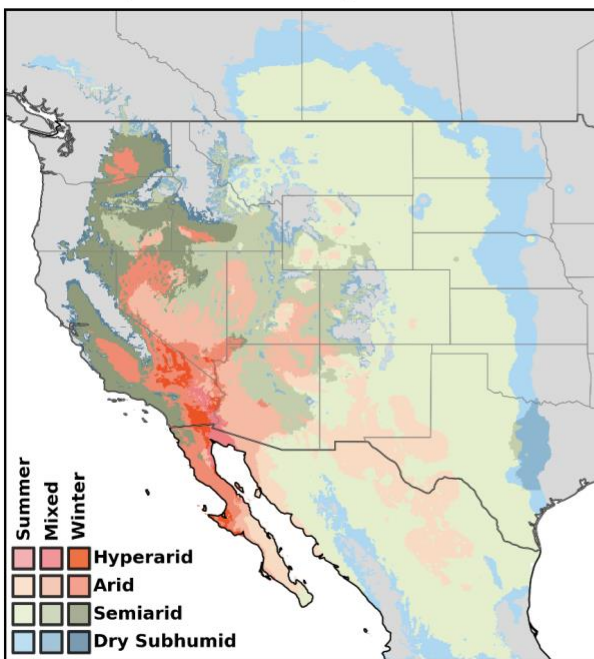
- 39 1) *Vegetation Structure and Composition*: VSWIR imaging spectrometer and LiDAR  
40 measurements allow characterization of how the vegetation structure and composition  
41 changes between lower to higher elevation regions.
- 42 2) *Rangeland Management/Water Availability*: Evaluating soil moisture dynamics of this  
43 region and its rangelands is pivotal for understanding rangeland response to water-  
44 limitation. Particularly, this can involve evaluating seasonal moisture availability and  
45 partitioning sources of moisture between from rain versus snowmelt, which likely have  
46 varying seasonal inputs. These measurements can be directly linked with models of  
47 western US soil moisture (e.g., SOILWAT), which can blend the remotely sensed and  
48 ground-based data into contemporary and future projects of soil moisture and edaphic  
49 drought (Lauenroth & Bradford, 2012).  
50



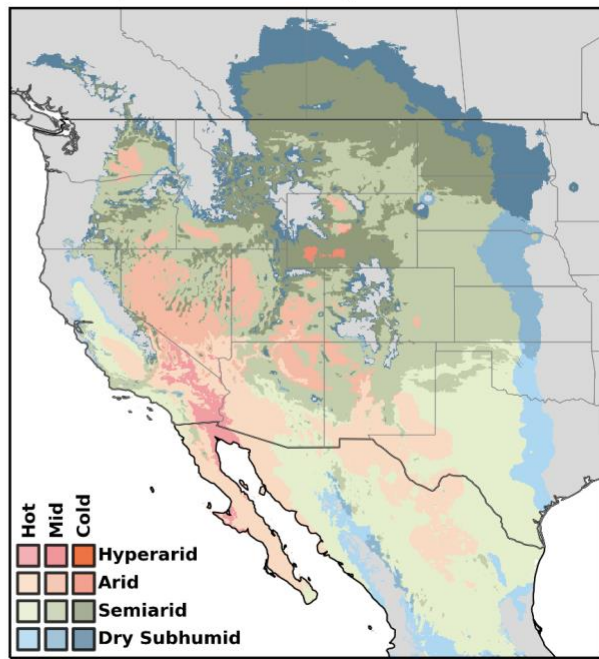
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*Figure B.16. Proposed focal areas within ARID's western US domain. Red borders roughly outline focal regions.*

**A. Precipitation – Aridity Gradient**



**B. Temperature – Aridity Gradient**



5



1 **Figure B.17.** The ARID focal areas take advantage of varying temperature and  
2 rainfall seasons across the western US.

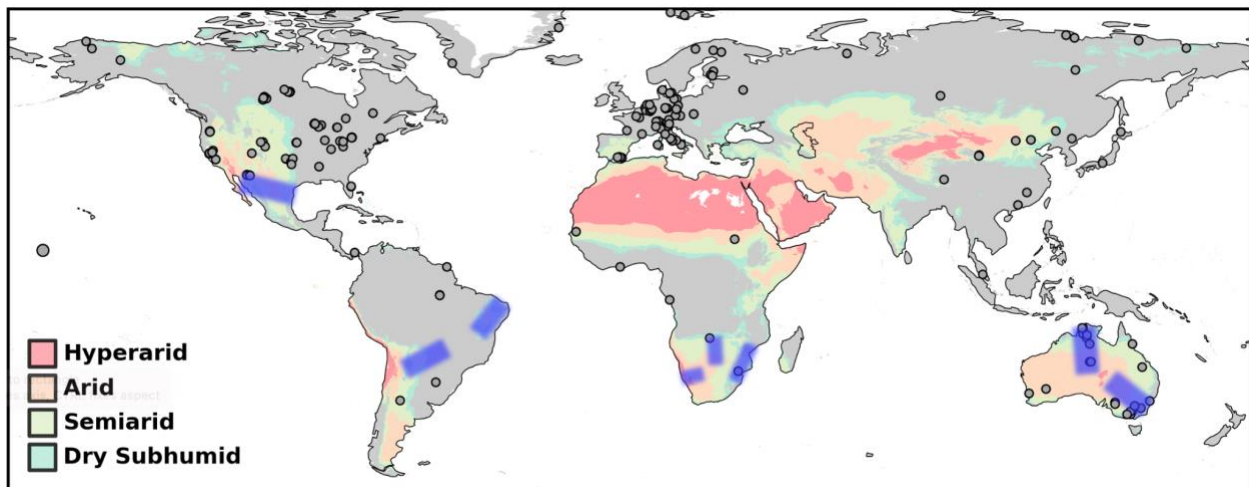
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4 **B.2.6.4 International Focal Areas**

5 Here, we provide motivation for our international focal areas. However, we note that our  
6 international implementation strategy differs from our domestic strategy by aligning activities  
7 with existing, in-place international research efforts in these focal areas that will address ARID's  
8 science themes (Fig. B.18). By aligning and co-developing with international research partners,  
9 we ensure continuity of the campaign activities (with the assets and measurements being  
10 leveraged beyond the bounds of the ARID campaign), allow for leadership of in-place  
11 researchers to champion the ARID campaign activities, and ultimately mitigate against ill-  
12 informed "parachute" science strategies.

13  
14 International focal area suggestions were selected based on:

- 15 (i) Community interest in the region
- 16 (ii) Contribution of region to addressing ARID science questions at global scale
- 17 (iii) Scientific partners and their assets (data, and infrastructure)
- 18 (iv) Feasibility and operational risk

19  
20 Based on these criteria, the suggested international domains are: Australia, northern Mexico,  
21 southern Africa, and the dry diagonal in South America. We do not provide extensive details on  
22 study type examples, but offer an overall vision for the types of research and engagement in  
23 international domains.



30 **Figure B.18.** Potential focal areas within ARID's distributed international  
31 domains. Existing FluxNet sites are shown as gray circles with black outlines.  
32 Blue shading roughly outlines potential focal regions, though these will likely  
33 change with input from partners and mission feasibility.

34 Australia: The ARID team has partnered with CSIRO (Commonwealth Science and Industrial  
35 Research Organization, Earth Observation Center), TERN, and OzFlux community, as well as a  
36 multi-university effort that is initiating a Centre of Excellence focused on heatwave impacts on  
ecosystem function. Australia has been identified as a potential strong source of carbon cycle  
interannual variability. Therefore, scaling between OzFlux sites, ground sampling, and potential  
hyperspectral aircraft measurements will be critical to understand relationships between  
vegetation spectra and vegetation carbon flux magnitudes. Sampling should also occur in EI-



1 opportunities. ARID also has strong ability to address end-user needs in Mexico, especially  
2 related to cropland management and related water management. Such needs are motivated by  
3 food supply demands in northern Mexico amongst indigenous communities and dry conditions  
4 that limit food production, which partnerships with SAGHARPA can be leveraged to address  
5 these end-user gaps. **More than 50% of Mexico is drylands.**

6  
7 Southern Africa: Southern Africa hosts a wide variety of dryland ecosystems, including arid  
8 shrublands, savannas, and woodlands, which supply the vast majority of the rural population's  
9 energy needs (e.g. fuelwood and charcoal) and where livelihoods are tightly linked to dryland  
10 ecosystem services. This region is also a hub of scientific activity which can help address  
11 dryland science questions in a continent that otherwise has a dearth of data and capacity.  
12 Community inputs motivated the ARID team to focus on this region. NASA also successfully  
13 carried the BioScape airborne campaign in South Africa, with demonstrated logistic support,  
14 safety, and scientific collaboration. We have consulted with NGOs (Conservation International)  
15 in Botswana, NASA SERVIR, USAID, and universities in Mozambique and Namibia, as well as  
16 research institutes in Namibia (Gobabeb-Namib Research Institute) and Botswana (Okavango  
17 Research Institute) (see letters of support). The South African Environmental Observation  
18 Network ([SAEON](#)) (similar to NEON) conducts research and provides facilities, instruments (flux  
19 towers), infrastructure (airborne sensors), datasets, models and staff across a diversity of  
20 ecosystems, including two nodes in drylands (see letter of support). SAEON can provide  
21 significant support and research partnership, as demonstrated during BioScape in South Africa.

22  
23 Southern Africa, including at these partners' locations, offers a high potential to address key  
24 ARID research questions from a different perspective with in-place knowledge. Savannas have  
25 the potential to be a major carbon sink, but fire regimes, fuel wood removal, overgrazing and  
26 elephant impacts can lead to a net loss of carbon, resulting in controversial scientific and  
27 management debates. For example, both Conservation International and USAID are assisting  
28 end-users with sustainable rangeland management to prevent overgrazing and invasive species  
29 establishment. Partners at Eduardo Mondlane University in Mozambique are evaluating the fire  
30 regimes of sub-humid miombo woodlands and attempting to create tools for locals to conduct  
31 sustainable management practices. Okavango Research Institute in northern Botswana have  
32 collaborated with US Army Corps of Engineers studying the dry to wet transitions between dry  
33 savanna and dry tropical forest and particularly how carbon fluxes change across this gradient.  
34 Namibia's Gobabeb Research Institute studies the hyperarid desert ecosystems which are  
35 relatively sparse in the US, and which are excellent examples of ecosystems that receive inputs  
36 from non-rainfall dew and fog as a dominant water source. ARID also has the opportunity to  
37 strongly build upon BioSCape campaign, which, although focused on biodiversity in the Cape  
38 Floristic Region (Fynbos), also included additional dryland sites in South Africa. ARID would  
39 also benefit from the heritage of SAFARI2000 and the Kalahari transect. All partners offered to  
40 guide NASA and ARID for their respective country's governmental processes to fly aircraft for  
41 research purposes, and they noted experience and confidence for success with permissions  
42 processes.

43  
44 South America: South America has a range of dryland ecosystems and infrastructure that can  
45 be leveraged and contributed to by ARID. For example, Brazil's national natural disaster  
46 monitoring and alert center (Cemaden/MCTI) uses a remote-sensing based dryland drought  
47 monitoring to provide monthly updates of an integrated drought index based on soil moisture  
48 and vegetation data. There are strong opportunities for ARID to partner with this organization.  
49 Brazil also maintains a semi-arid ecosystem rainfall and soil moisture network with 495 sites  
50 that offer strong opportunities to collect complementary data and to ground truth remote  
51 sensing. Multiple eddy covariance towers in semi-arid South American ecosystems maintain a

1 focus on dryland responses to flash drought and could represent candidate international Super  
2 Sites. South America maintains an exciting phenological network (e-phenology) that crosses the  
3 continent and is focused on seasonally dry forests and woodlands. This network assesses  
4 productivity and functional traits, links with water cycles, woody vegetation and grassland  
5 dynamics, carbon sinks and sources, and provides a powerful opportunity to scale with drones,  
6 satellites, and models. This and other networks (e.g., DRYFLOR and SECO) focuses on  
7 biodiversity and other ecosystem services and could be used as test beds for exploring remote  
8 sensing capabilities for biodiversity in tropical drylands. The Deforestation Dryland Alert System  
9 (DDAS) is an innovative system designed to monitor and quantify deforestation in dryland  
10 regions, with a focus on the Caatinga. The DDAS uses advanced geospatial technology,  
11 satellite imagery, and cloud computing to detect and analyze changes in vegetation cover and  
12 can provide near real-time alerts on deforestation activities, enabling timely enforcement  
13 actions. Areas of potential focus include the Caatinga, which spans 884,453 km<sup>2</sup> and is the  
14 largest tropical dry forest in South America, and the Cerrado, a vast ecoregion of tropical  
15 savanna in eastern Brazil and second largest of Brazil's major habitat types (after the  
16 Amazonian rainforest), accounting for 21% of the country's land area. Not all of the Cerrado are  
17 drylands but many parts are and the region represents an area of transition where more of the  
18 Cerrado could be drylands in the future. Extensive research, expertise, and infrastructure in the  
19 region offer substantial scientific opportunities. Finally, the Gran Chaco semi-arid lowland area  
20 than brings together more than 50 different ecosystems united by the same pattern of  
21 vegetation and climate and one of the highest deforestation rates on the planet (every month  
22 >340 km<sup>2</sup> is lost). There is a particularly strong emergent need to understand and forecast the  
23 impact of fires in the South American Dry Diagonal to socio-ecological systems and global  
24 carbon budgets. The region is facing dramatic heat, drought, and wildfire combinations, which  
25 are consuming large areas of dry forests from the Chiquita no (Bolivia) to Caatinga (Brazil).  
26 ARID has built partnerships with the scientists of South America, with great mutual interest in  
27 collaborating on a NASA field campaign focused on drylands. **More than 30% of South**  
28 **America is drylands.**  
29  
30

### 31 ***B.2.7 Earth Science to Action Strategy and Application***

32 ARID is fully aligned with NASA's ES2A goal to "advance and integrate Earth science  
33 knowledge to empower humanity to create a more resilient world." ARID envisions three types  
34 of application strategies. (1) As with former Terrestrial Ecology field campaigns, ARID will  
35 generate datasets that drive foundational knowledge development that can indirectly inform  
36 societal applications with further development of the data and knowledge into the decision  
37 framework, or information development lower on the ES2A pyramid (Fig. A.6). Such a strategy  
38 is more indirect. However, with the rollout of ES2A, ARID is committed to conducting projects  
39 that will directly involve end-users. In fact, much of the ARID engagement activities during  
40 scoping were focused on understanding end-user needs before we started developing the  
41 science themes or field implementation strategy. This produced two other strategies which  
42 involve directly working with and co-developing with end-users, including (2) broader NASA-  
43 directed efforts between a group of scientists and an end-user agency (like BLM, see below)  
44 and (3) through individual efforts that can be conducted as a part of competed projects. We  
45 describe our strategy for these latter two more direct efforts and then highlight several specific  
46 use-cases we envision, developed with extensive conversations with our partners, where ARID  
47 will advance decision making and products used inside existing and future applications.

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*End-User Co-Development Strategy*

The ARID field campaign will directly engage operational agencies (e.g. BLM, USFS, USDA), rural and tribal communities, united around fundamental dryland science questions to address science-based decision making. These partnerships began during scoping with their needs discussed and used to develop our science themes and implementation strategy. The ARID field campaign aims to improve the algorithm development and uncertainty estimation of fundamental remote sensing products that feed directly into existing workflows of agencies with established user-basis and decision-making frameworks. These applications will facilitate adaptation and mitigation to changes in drylands, demonstrating the societal value of NASA’s science to a large population of diverse users and fostering inter-agency support for ES2A.

Our suggested strategy to improve and facilitate end-user decision making includes first understanding the end-user needs and constraints. We have gained many insights about end-user needs (see Section C), but the scientist(s) will need to converse with the end-users in each specific case. Boundary organizations like the NASA Satellite Needs Working Group can help guide this process. The next step is to develop the data product and/or monitoring tool and then iterate again by discussing with the developments with the end-user. In the final phase of implementing the new information into the decision-making framework, the ARID-funded scientist needs to continue facilitating and revising the data product as necessary and respond to any shortcomings identified by the end-user. Such early, iterative, and continuous engagement with end-users is critical to build the new knowledge and data products into the specific decision-making frameworks.

The impact of these end-user engagements can be measured by consulting the “*Defining and Measuring Impact in NASA Ecological Conservation Projects*” handout. After an applied science project is carried out, such a framework can be consulted to quantify the impact the work had on the end-user in terms of knowledge gain, extent of use, change in behavior, benefit, awareness and perception, and sustainability.

*BLM Rangeland Monitoring*

BLM has co-developed remote sensing-based rangeland monitoring applications and NASA’s Ecological Conservation program funded the development of one of these approaches (BLM, LandCart). These applications have a very large and diverse user base (e.g., [Rangeland Analysis Platform](#) has > 22,000 sites visits per year, referenced > 155 publications) and they have developed through multiple versions and stages of maturity. The approach uses basic algorithms to estimate PFT fractional cover and biomass from Landsat data to track rangeland condition (M. O. Jones et al., 2021; Kleinhesselink et al., 2023; Robinson et al., 2019). Similarly, the USGS has developed the Rangeland Condition, Monitoring, Assessment, and Projection (RCMAP) products of fractional cover components (shrub, grass, annual / perennial grass, sage-brush) across the western US, trained with field estimates of fractional cover types and limited high resolution commercial satellite imagery (1985–2020) (Rigge et al., 2019; Shi et al., 2022). *The ARID field campaign can significantly improve the estimation of fine-scale fractional cover and biomass of PFT through improved calibration and validation of algorithms trained with*

1 *fraction PFT data from multi-temporal hyperspectral airborne and UAS data, but applied to*  
2 *Landsat imagery (and SBG in future) to ensure wide area coverage (see section B.2.1.2 for*  
3 *details). These improved fundamental input products will feed into the established workflows to*  
4 *inform BLM and other decision makers. These applications have evolved in their operational*  
5 *capabilities and BLM has recently entered into multi-year contracts to support these applications*  
6 [\(Climate Engine\)](#).

7  
8 BLM furthermore need species-specific products on the extent of Sagebrush (primarily for  
9 conservation of endangered Greater sage-grouse, *Centrocercus urophasianus*) and pinyon  
10 juniper cover, as well as invasive annual grass cover specifically. BLM furthermore uses the  
11 fundamental and derived data products for decisions related to habitat conservation, restoration  
12 and reclamation effectiveness. *The BLM rangeland application will allow ES2A to demonstrate*  
13 *societal impact of ARID campaign and science.* The ARID-derived fundamental PFT data  
14 products can feed into multiple applications (e.g. fuel properties for wildfire risk predictions) for  
15 multiple agencies, including USFS and National Park Service. The BLM rangeland application  
16 will allow ES2A to demonstrate societal impact of ARID campaign and science. Such an effort is  
17 more aligned with the second strategy noted above and will require a team of scientists to  
18 consistently work with BLM, and should be encouraged to be a directed effort or competed effort  
19 with more continuous funding beyond three years.

#### 20 21 *Tribal Rangeland Monitoring*

22 Three specific tribal land management needs that arose during the scoping study were: (1)  
23 water availability, specific mainly to tribes in Arizona and New Mexico; (2) rangeland  
24 productivity, specific to tribes in the Great Plains, Arizona, and New Mexico; and (3) buffalo  
25 restoration, specific to tribes in the Great Plains. Such work can be funded as a part of the third  
26 strategy noted above with more individual-based collaborations between scientists and the  
27 tribes. In this case, it is encouraged that tribal members are leaders of the funded effort or co-  
28 funded between NASA funded scientists and tribal scientists or managers.

29  
30 For water availability, Natural Resources Departments, for example at the Pueblo of Santa Ana  
31 or Pueblo of Jemez, need to be consulted first with their elders first providing permissions and  
32 recommendations on any measurements or research going forward. Soil moisture sensors can  
33 be deployed, and a higher resolution microwave measurement like from active satellite radar  
34 (NISAR) or aircraft (SLAP or UAVSAR) can be used to monitor spatial distributions of soil  
35 moisture at 100s of meter scales. Additionally, OpenET can be used which is a proxy for  
36 moisture availability via Landsat at 30m resolutions. Such datasets can be developed into a  
37 simplified framework for the tribe to use continuously. Finally, NASA co-developed the Navajo  
38 Drought Severity Evaluation Tool with the Navajo Nation. Measurements generated from ARID  
39 can further constrain that tool's outputs.

40  
41 For rangeland productivity, it was recognized that increasing precipitation variability and  
42 extremes driven by climate change are already having significant impacts on rangelands of the  
43 western US, with critical consequences for livestock grazing and wildlife. There is an urgent  
44 need for monitoring and forecasting the seasonal productivity of these vulnerable

1 agroecosystems in order to support effective resource management and conservation efforts.  
2 The United States Department of Agriculture (USDA) Grass-Cast rangeland productivity  
3 forecast tool is available for the Great Plains and was recently expanded to the Southwest U.S.  
4 and provides short-term seasonal forecasts of rangeland productivity with the major objective of  
5 providing early decision support for rangeland managers. For Grass-Cast Southwest, Navajo  
6 Nations shared a wealth of field-based estimates of grassland aboveground net primary  
7 productivity that was key for parameterizing this model. There are opportunities to expand  
8 Grass-Cast to other tribal lands and western US ecoclimate zones that we hope to initiate with  
9 ARID.

10

11 For buffalo restoration, climate adaptation centers, such as the Rosebud Sioux Tribe, are  
12 focused on broader buffalo restoration efforts across the Great Plains. There are potential  
13 applications for Landsat and ECOSTRESS to generate maps of grazable land and vegetation  
14 biomass in conjunction with buffalo population distribution data via machine learning or other  
15 statistical approaches. Focus can be placed on sustainably managing these lands to enable  
16 growth of buffalo populations.

17

## 18 ***B.2.8 Technical and Logistical Feasibility***

19 ***The ARID field campaign is a low-risk, high-reward endeavor*** designed to significantly  
20 advance our understanding of Earth's dryland regions and their dynamic responses to  
21 environmental changes. ARID is low-risk in that it leverages existing networks, sites,  
22 infrastructure, and local knowledge and proposes new measurements across easily accessible  
23 dryland systems. This method reduces costs, increases field safety, and provides high value to  
24 additional field and aircraft measurements in their ability to advance dryland science. ARID is  
25 also high-reward in applying state-of-the-art yet readily available satellite remote sensing  
26 technologies along with field deployments to understand a poorly understood and generally  
27 understudied ecosystem: drylands. With research findings increasingly emphasizing drylands'  
28 large role in the Earth system (see Section A.1), ARID presents an opportunity to bring the  
29 science community together to gain substantial new understanding of the global relevance of  
30 these ecosystems, which encompass 41% of the global land surface. These findings are  
31 essential for informing sustainable land management and mitigating the impacts of rapidly  
32 evolving climate change in drylands.

33 Most ARID field activities can be conducted within the extensive range of dryland ecosystems in  
34 the U.S., or in collaboration with local international partners where successful field operations  
35 with NASA have already been carried out. The U.S. provides a highly tractable environment with  
36 established infrastructure and logistical advantages that ensure the campaign's success while  
37 maximizing its potential for groundbreaking scientific discoveries. Conducting operations  
38 domestically allows NASA to allocate funds efficiently, supporting U.S.-based personnel,  
39 technology, and research. The familiarity and extensive training of U.S. personnel with NASA  
40 assets enable a rapid start to the campaign and the production of high-impact results.  
41 Logistically, planning for aircraft operations is more feasible due to fewer regulatory constraints,  
42 allowing flexible scheduling of flights across the western United States. The dense distribution of  
43 airports further enhances flight safety, enabling "pulse-chasing" experiments that capitalize on  
44 the rapid, day-to-day evolution of dryland environments following rain events.

1 By focusing on these elements, the campaign aims to maximize the feasibility and success of its  
2 flights and fieldwork while ensuring the safety of all field and flight crews. This approach is  
3 crucial for achieving the campaign's objectives, allowing for efficient resource use and  
4 minimizing uncertainties associated with large-scale field operations.

5 Internationally, the ARID campaign stands to benefit from established collaborations with highly  
6 capable scientific teams in regions like South Africa and Australia. These partnerships,  
7 strengthened by previous NASA field campaigns, enhance the likelihood of obtaining necessary  
8 permissions and flight permits, making international operations more feasible. For instance,  
9 South Africa's experience with NASA's BioSCape and Australia's airborne and in-situ assets  
10 can provide valuable support to the ARID initiative. Additionally, using local aircraft and UAS  
11 technology in these regions reduces the need for long-distance transport of NASA equipment,  
12 lowering costs and simplifying logistical planning. In situations where NASA aircraft operations  
13 are not feasible due to logistical constraints, local partners or private companies could provide  
14 equivalent airborne data, such as the HyMap HSI in Australia, ensuring the continuity and  
15 success of the ARID campaign.

16 Technically, the ARID field campaign is positioned to succeed with minimal risk, employing  
17 mature remote sensing technologies that ensure robust and reliable data collection. While the  
18 campaign may include some new or experimental sensor technologies, such as multi-temporal,  
19 multi-spectral drone data collection, the risks are manageable and represent opportunities for  
20 significant innovation in sensor technology. For example, cross-calibration with NASA assets  
21 could fall within the scope of a proposed PI-led research project. Moreover, proximal sensing  
22 from towers, involving innovative yet low-risk sensors, could further enhance the campaign's  
23 capabilities.

24 The ARID initiative is well-prepared to deliver high-impact results by building on a foundation of  
25 strong existing expertise and collaborations. With ongoing partnerships with agencies like the  
26 Bureau of Land Management (BLM) and tribal nations, the campaign is positioned to secure  
27 necessary permits and permissions, ensuring high accessibility to field sites. By leveraging  
28 partners such as NEON and USGS for UAS data acquisitions, and employing a mix of U.S.-  
29 produced and international UAS technologies, ARID minimizes risks and ensures the collection  
30 of scientific-grade data. *This strategic approach to logistics, technical readiness, and*  
31 *international collaboration underscores ARID's potential for groundbreaking discoveries, all*  
32 *while maintaining a low-risk profile.*

### 33 **B.2.9 Modularity and Sequence**

34 The ARID study domain and research framework allow for a modular approach to the collection,  
35 analysis, harmonization, and communication of data. For example, the sequence and priority of  
36 study domains would be straightforward to plan and modify due to the lack of many of the  
37 logistical challenges that hinder operations in locations where research regulations,  
38 permissions, and safety are more challenging and uncertain. New data collection in the ARID's  
39 US domain could begin quickly due to the high feasibility and access to US sites for ground-  
40 based and airborne data collection, as well as the ability to access substantial existing  
41 infrastructure and partnerships. In addition to the US, there are many strong existing  
42 collaborators and networks in Mexico and Australia on which ARID could build, and there is a  
43 history of NASA field work in our focal regions in southern Africa. These allow for more rapid  
44 spin-up of field work. There is also the possibility for measurements in Mexico where takeoff and  
45 landing occur in the US, and the feasibility of such an approach can be investigated. Moreover,



1 there is a range of data that already exists for drylands of the US, which could be leveraged,  
2 added to, and synthesized in new ways. Although many of these datasets were collected to  
3 inform sensor and mission development and not to address ecological questions, ARID can  
4 draw on many datasets already collected by NASA (section B2.5), providing even more return  
5 on investment. There are aspects of the ARID campaign that would be more challenging to  
6 separate in space or time. For example, transect and Super Site data collections require the  
7 concurrent assessment of ecosystems with coordinated ground based and remote sensing  
8 tools, thus it would be preferred to collect the necessary data at focal sites in series (i.e., not all  
9 the sites at the same time) over not collecting the full suite of complementary data needed to  
10 answer core questions.

11  
12 This modularity and ability to perform data collections in sequence could be valuable for a  
13 number of reasons (e.g., unexpected changes in funding, an opportunity to leverage a large-  
14 scale US land management treatment), but there are also options unique to a framework in  
15 which both ARID and PANGEA were selected to advance as field campaigns. First, the ability  
16 for ARID to start quickly could allow for a system in which ARID began data collections first,  
17 while PANGEA coordinated their collections with international partners. ARID could then spend  
18 time analyzing collected data, parameterizing, and evaluating model developments, while  
19 PANGEA performed a period of new data collection. ARID could then return to a local or  
20 international data collection phase while PANGEA transitioned to its data analysis phase and  
21 then subsequently returned to data collection. This “ebb and flow” approach could continue for  
22 the lifetime of the campaign. There is a great deal of interest from other federal partners that  
23 could result in additional funding for ARID or for large scale coordinated treatments co-designed  
24 for science and management and conducted by agency partners, and this framework could  
25 allow for more time to plan and coordinate. Second, in addition to coordinating data collection  
26 and processing for ARID and PANGEA, there are geographically tractable and climatically  
27 strong gradients that exist spanning from drylands to the wet tropics that could be used to  
28 integrate questions and expand inference. For example, in South America, Africa, and Australia  
29 (see Sections B.2.4, B.2.6). This offers a chance to study key aspects of ecosystem structure  
30 and function (e.g., biodiversity, carbon cycling) along environmental gradients that could  
31 address both ARID and PANGEA goals, offer information on what drives the transition between  
32 drylands and tropical forests, and could lend a deeper insight into larger terrestrial function and  
33 how ecosystems may transition in a drier, hotter future.

34

## 35 **C. Community Engagement and Co-Development**

### 36 **C.1 ARID Community Engagement Vision**

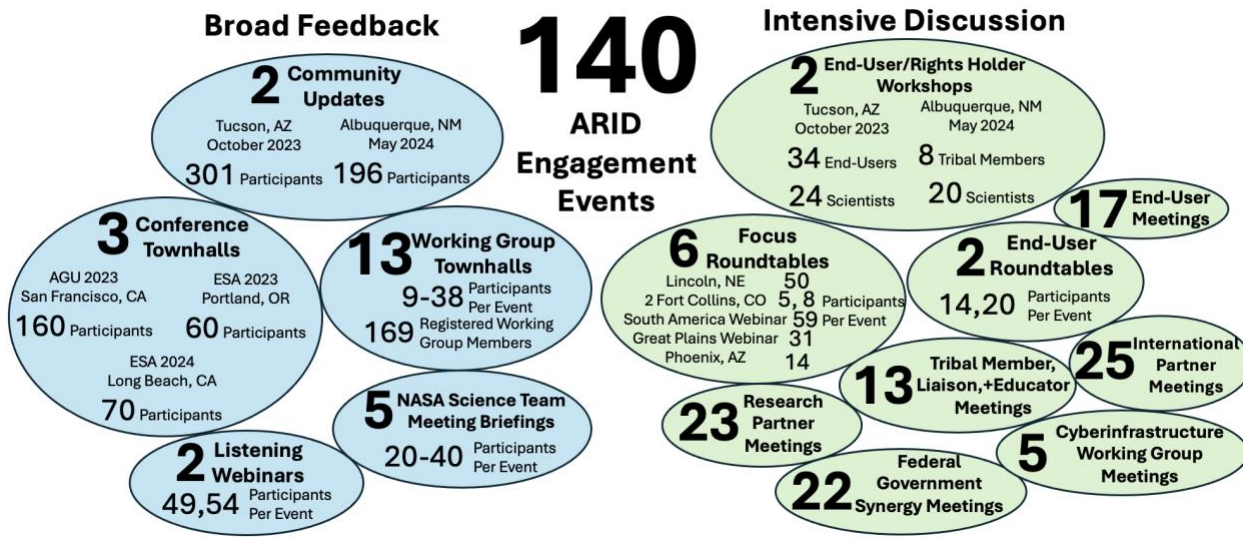
37 ARID has engaged and partnered with numerous communities that provided input throughout  
38 the scoping study and helped develop the perspectives, questions, and framework outlined  
39 here. If the ARID campaign is selected, this community is extremely enthusiastic to be involved  
40 in writing proposals, conducting research, and providing new knowledge and capability to

1 understand and make decisions for drylands. In particular, we have worked to cultivate  
2 relationships with three major types of communities - researchers, data end-users, and  
3 indigenous stewards - to contribute to improving the design, implementation, and usage of the  
4 knowledge generated by ARID. These communities are not mutually exclusive of each other,  
5 and some groups contribute to multiple aspects of research and end-user project design, such  
6 as tribal nations and natural resource managers.

7 ARID's networks were developed based on both broad feedback during wide-scoped knowledge  
8 gathering sessions, as well as intensive discussions in smaller groups on more focused topics  
9 (Fig. C.1). Overall, the ARID team conducted 140 engagement events, including roundtables  
10 with end-users and conference town halls (Fig. C.1, C.2). This is in addition to ARID's many  
11 working group and leadership team meetings. A full list of events and communities consulted  
12 are listed in Appendix F.6, F.7.

13 Consequently, the ARID team has established an advisory network of partners who would  
14 directly contribute to knowledge development, guide or advise the campaign, benefit from the  
15 campaign, and/or, in some cases, co-fund the campaign (Fig. C.3). If funded, this structure can  
16 persist, with end users helping to inform and/or be a part of ARID designated and competed  
17 science teams (see Section D.1 for more description of ARID's management structure). The  
18 interactions are anticipated to be primarily mutualistic. For example, ARID's data gathering  
19 would inform end-user issues and decisions and could be used to design or inform decision  
20 making tools. End-user data use will guide more optical data collection. Boundary organizations  
21 like NASA's Satellite Needs Working Group (SNWG) and Western Water Applications Office  
22 (WWAO) can help guide this communication with end-users. While making use of NASA  
23 satellite products throughout the campaign to advance dryland science, ARID can also inform  
24 the development of current and future NASA missions (SBG, NISAR).

25 In this section, we discuss our domestic and international research partners, end-user partners,  
26 and tribal partners, including discussing who our partners are, how they would co-develop the  
27 ARID campaign, and our philosophy for a continued engagement strategy. We also discuss  
28 capacity building of the next generation of scientists, including options for "place based"  
29 researchers and end-users. Finally, we discuss our overall vision for diversity, equity, and  
30 inclusion. Note that a summary of our ARID engagement and strategies appears in our *Earth's*  
31 *Future* publication (Feldman, Reed, et al., 2024).

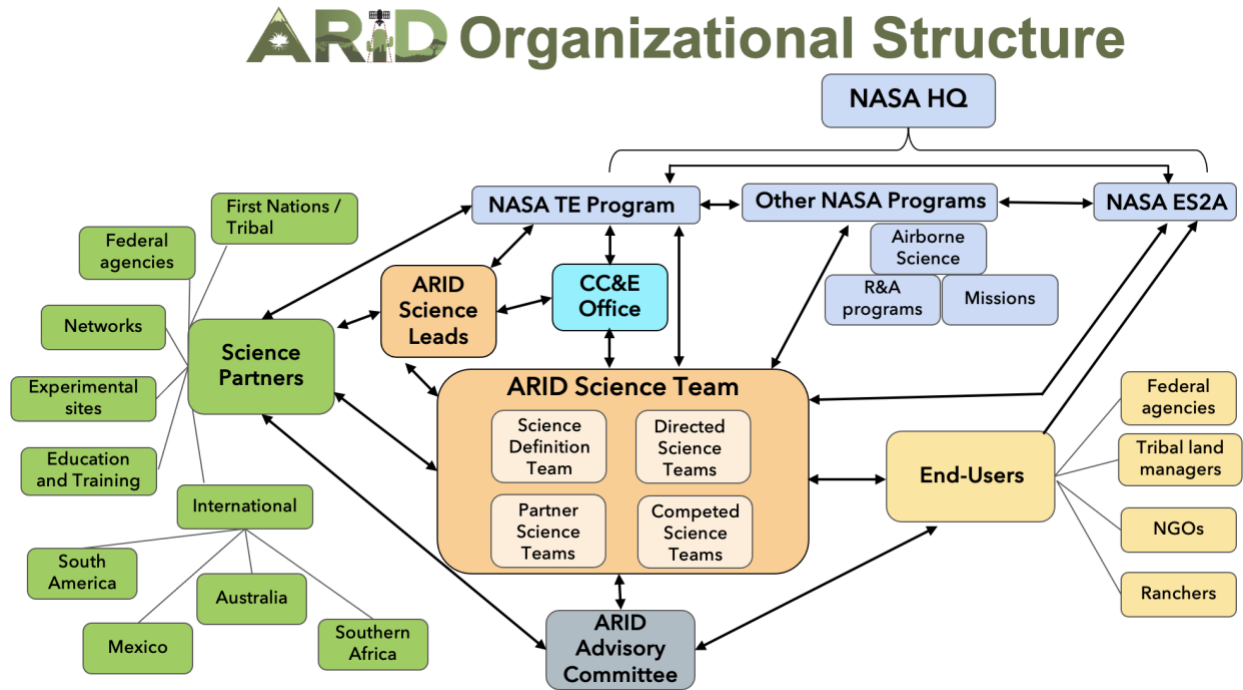


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**Figure C.1.** Summary of ARID engagement events, including both broad feedback (from listening sessions) to intensive discussions, primarily occurring between October 2023 and September 2024. Total participants are based on event documentation from ARID team members. Most engagements consisted of approximately 1-1.5-hour meetings. However, the workshops included 1–2-day discussions with end-users and rights-holders across several topics and two-day science-focused workshops. Note that “end-user meetings” included with end-users and those who liaise between researchers and end-users. Numbers do not include internal meetings between ARID team members.



1 **Figure C.2.** (a) Data end-user meeting and (b) science meeting at the ARID  
 2 Kickoff meeting at University of Arizona in October 2023. ARID town halls at the  
 3 (c) 2023 Ecological Society of America (ESA) and (d) 2023 American  
 4 Geophysical Union (AGU) conferences. The photos are owned by the authors.  
 5



6  
 7 **Figure C.3.** An overarching ARID network describing how the ARID Science  
 8 Team and Leads would be informed by and inform domestic and international  
 9 partners, end-users, and NASA-based groups.  
 10

## 11 C.2 ARID Research and Technology Partners and Collaborators

### 12 C.2.1 Domestic Research Partners and Collaborators

13 The ARID Field Campaign is envisioned to be multi-national in scope, but also to rely heavily on  
 14 a US focused implementation plan, building on domestic collaboration across multiple  
 15 institutions, tribal nations, and agencies in the western US. The domestic efforts will incorporate  
 16 partners at various levels of the field campaign, ranging from site level operators conducting on-  
 17 the-ground observations and experiments to flux tower operators, aircraft and drone operators,  
 18 modeling groups, and remote sensing specialists. These technical groups will work with  
 19 resource managers in both a research capacity, as well as in the translation of new knowledge  
 20 to be used in management decisions.

21 Our domestic partners are summarized in a list below, consisting of (1) ground networks that  
 22 have existing infrastructure that ARID can use in collaboration with the leadership of the  
 23 networks; (2) domestic institutions interested in co-funding research and/or campaigns and that

1 have networks of personnel interested in collaborating and leading projects of their own. Others  
2 include (3) private institutions and companies that have an interest in ARID data for their  
3 applications; (4) Federal agencies that ARID researchers can work with to meet their decision  
4 and management needs, as well as provide support in conducting the campaign on public land  
5 (e.g., BLM, DOD); and (5) Tribal Nations interested in co-developing research and land  
6 management with ARID.

7  
8 Of note, the ARID scoping study team established strong partnerships with a number of US  
9 federal agencies, and especially BLM and USFS, which would serve as direct end users of  
10 ARID data with clear use cases. For example, BLM National Science Advisor, Karen Prentice,  
11 and other BLM staff, such as BLM's new Climate Science Program Manager (Dr. Julian Reyes)  
12 emphasized the critical importance of the data ARID would provide for the lands that BLM  
13 manages in the western US. They described themselves as data 'super users' and made clear  
14 their excitement and willingness to engage in ARID. Dr. Julian Reyes accompanied the ARID  
15 team to several NASA HQ meetings to emphasize these points (Appendix F.6).

16 This adds to a strong partnership with the USGS, which is highly motivated to facilitate and  
17 collaborate on ARID as an opportunity for additional successful collaboration between USGS  
18 and NASA (e.g., building on GEMx). There are also numerous potential opportunities to co-fund  
19 ARID, which can be considered during the science definition team phase. For example, the DoD  
20 program manager mentioned a strong interest in DoD understanding dryland ecosystem  
21 disturbance, habitability, and response because of the large US military installations across the  
22 western US drylands (e.g., 5% of New Mexico and Nevada are DoD lands, an extremely large  
23 amounts of federal drylands) and because of global instability within many drylands. If budgets  
24 remained at their current states, there could be strong opportunity for DoD and the U.S. Navy to  
25 support aspects of ARID through the funding of PIs, thus leveraging the funding provided by  
26 NASA. The NSF program managers further supported ARID's plans to work directly with the  
27 NEON site PI's to co-conduct work. The DOE talked about the needs of improved dryland  
28 understanding for informing their Earth System Models. Finally, NASA terrestrial hydrology  
29 program managers mentioned the possibility their program could consider co-funding  
30 components that relate to GEWEX activities and water and energy cycle topics, and there are  
31 strong linkages to the LCLUC, Fire, and ES2A programs.

32

33 Here, we list several of our main domestic U.S. partnerships. This list is non-exclusive (see  
34 letters of support for more details):

35

- 36 • U.S. Federal Agencies: BLM, USFS, DoD, Navy, USDA ARS (Jornada, Fort Collins,  
37 Southwest Watershed Research Center, Beltsville, Biose) USGEO, BoR, NPS, NOAA
- 38 • Research Networks: NEON, AmeriFlux/FLUXNET, LTAR, LTER, Mesonet, BLM AIM,  
39 PSINet, NOAA Fire Sites
- 40 • Domestic Efforts: GEWEX, RUBISCO
- 41 • University Partners: ASU Global Drylands Center, UNM Arid Institute, UA Arizona  
42 Experiment Station, NMSU Agricultural Experiment Stations, CSU Natural Resource  
43 Ecology Laboratory

- 1 • Tribal Nations Partners, Liaisons, and Collaborators: SIPI, Greyhills Academy High  
2 School, Pueblo of Jemez, Pueblo of Santa Ana, Ute Mountain Ute, Rosebud Sioux,  
3 NASA IPI, USGS CASC North Central, USGS CASC Southwest, NASA MUREP  
4 MAIANSE
- 5 • Private Industry: BHP Copper, Rio Tinto, Ball Aerospace

### 6 7 **C.2.2 International Research Partners and Collaborators**

8 The science that ARID would produce will also provide new information that will be useful to the  
9 international research community, and international partners will play various roles in ARID's  
10 implementation phase. International partners and collaborators will enable a more robust test of  
11 process understanding of ecosystem changes in relation to different stresses and under control  
12 of different biodiversity controls. These partners can conduct simultaneous field campaigns to  
13 coincide with those conducted in the western US and in line with shared remote sensing tools  
14 and techniques. In addition, these collaborators will expand our observations of ecosystem  
15 processes determining the extent ecosystem services would be impacted under different climate  
16 and land use changes. Harmonizing flux emissions and other ecosystem variables for dryland  
17 systems will be feasible with the coordinated efforts among domestic and international partners.

18 ARID's international partners are summarized in lists below. Candidate international countries  
19 and partners were identified according to following criteria: (1) significance of drylands in that  
20 region/country (uniqueness of climate, vegetation, specific management techniques, etc.); (2)  
21 existing research networks and connections (university, institutes, government agencies); (3)  
22 community input (See Appendix F.9) (4) existing measurement infrastructure; (4) previous  
23 efforts in these regions (especially related to NASA); (5) feasibility and security; and (6) maturity  
24 of scientific relations with US.

25 There are various international efforts that strongly align with ARID research approach and  
26 goals. Countries, such as Australia and South Africa, have comparable research assets to  
27 conduct similar observational and research efforts coincident with ARID campaign timeframe.  
28 Other countries can provide ground-based observations to assist in verification studies and  
29 information to inform observational and modeling efforts to incorporate in the research analysis  
30 of ARID across various ecosystem structural types and processes. International partnerships  
31 also provide a broader context for dryland land use and stewardship approaches across a range  
32 of social-ecological systems.

33 While global drylands and associated livelihoods face similar challenges due to climate change,  
34 they vary greatly in their environmental conditions, drivers of change and management  
35 objectives. International expansion of the ARID approach allows for a broader range of  
36 environmental gradients and environmental responses as well as greater range of results and  
37 applicability of models. ARID created an international network of partners that will collectively  
38 address science questions by leveraging existing field data, instrument, and remote sensing  
39 data sets. International partners will contribute local knowledge and data. They can also benefit  
40 greatly from NASA airborne acquisitions where feasible. Several alternative scenarios are  
41 available for airborne data acquisition: (i) NASA aircraft and NASA sensors, (ii) local aircraft and

1 NASA sensors, and (iii) local aircraft and local sensors. There are also impactful opportunities  
2 for the training and education of early career scientists from other nations during the  
3 international campaign.

4 As an example, the South African Environmental Observation Network ([SAEON](#)) is a  
5 government-funded long-term environmental observation and research facility, similar to NEON  
6 and also part of Global Ecosystem Research Infrastructure ([GERI](#)). SAEON conducts research  
7 and provides facilities, instruments (flux towers), infrastructure (airborne sensors), datasets,  
8 models and staff across a diversity of ecosystems, including two Expanded Freshwater and  
9 Terrestrial Environmental Observation Network (EFTEON) in drylands. SAEON was a major  
10 partner in NASA's successful BioScape airborne campaign in the western Cape which  
11 demonstrated logistic support, safety and scientific collaboration of this focal area. SAEON has  
12 expressed their willingness to be a leading participant in ARID and their existing relations with  
13 TERN (Australia) and NEON will be of great benefit (see letter of support).

14

15 Here, we list several of our main international partnerships. This list is non-exclusive (see letters  
16 of support for more details):

17

- 18 • *Southern Africa*: South African Environmental Observation Network SAEON, BioScape  
19 (South Africa), University of Cape Town (South Africa), Okavango Research Institute  
20 (Botswana), University of Namibia, Southern African Science Service Centre for Climate  
21 Change and Adaptive Land Management (SASSCAL)
- 22 • *Australia*: ARC Centre of Excellence (which includes partners at University of  
23 Melbourne, Australian National University, Western Sydney University), OzFlux, CSIRO,  
24 and TERN
- 25 • *Mexico*: National Autonomous University of Mexico (UNAM), MexFlux, SAGARHPA
- 26 • *South America*: SECO, DryFlor, e-phenology/UNESP, CEMADEN/MCTI, SOS sertão  
27 (NGO), Biospheric and Climate Laboratory IAG/USP
- 28 • *Central Asia and Mongolia*: National University of Mongolia, Mongolian Ministry of  
29 Environment and Tourism, Samarkand State University, Uzbekistan
- 30 • *International Agency Efforts*: NASA SERVIR, USAID, FEWSNET
- 31 • *NGOs*: Conservation International, Organization of American States

32

### 33 **C.2.3. Research and Data User Engagement**

34 The science community was engaged in several ways throughout the scoping study, and we  
35 received valuable feedback on the ARID framework from a large number of diverse research  
36 communities and range of career stages throughout scoping effort.

37 Conference town halls and science webinars provided broad feedback from across the research  
38 community. These events served as an efficient means to describe the plans for the ARID  
39 campaign while allowing time for question-and-answer and general feedback. These events  
40 were well attended, with 200 and 300 virtual attendees at our two ARID science meeting  
41 briefings and over 60, 70, and 160 in-person attendees at our three conference town halls: the  
42 ESA Annual Meeting 2023, ESA Annual Meeting 2024, and AGU Fall Meeting 2023,

1 respectively. These events were most useful early in the ARID scoping effort, mainly between  
2 August 2023 to December 2023, for ARID to receive broad input and to look for commonalities  
3 in the feedback. The ARID leadership team members carefully evaluated the feedback to  
4 identify common threads of input, which guided discussions about the main science themes,  
5 research questions, and measurement techniques to be used throughout the ARID campaign.

6 More specific topics and thoughts were discussed during working group meetings and with  
7 write-in responses. ARID formed several types of working groups including science theme  
8 working groups and technological working groups. The science theme working groups primarily  
9 consisted of the main contributors of this ARID white paper and were organized around the  
10 ARID sub-themes discussed in Section B.1. They were tasked with outlining and drafting the  
11 ARID science sub-themes, and initiated their activities during the October 2023 Tucson, AZ  
12 kickoff workshop (Appendix F.2). The ARID technological working groups were organized into  
13 four groups including ground network and in-situ measurements, remote sensing  
14 measurements, model development, and cyberinfrastructure (Appendix F.5). These groups  
15 were tasked with having more focused group discussions about the respective approach.  
16 Across the three working group topics of ground network measurements, remote sensing  
17 measurements, and model development, there were 169 registered members of the ARID  
18 working groups and 9-38 attendees in any of these 13 total townhalls. Subject experts from the  
19 community were invited to give seminars about their approaches and how they can be used in  
20 or inform the ARID campaign. These inputs were most useful mid-way to later in the scoping  
21 effort, conducted mainly between April to August 2024. These engagement activities specifically  
22 allowed querying about our more updated framework. The cyberinfrastructure working group  
23 conducted more specific discussions across 5 meetings. Beside the working group activities,  
24 throughout the duration of ARID, we received over 350 written inputs. The ARID scoping  
25 steering committee (authors of this white paper) met weekly and sometimes twice-weekly (ARID  
26 writing committee) to plan these activities, discuss the received feedback, and revise the ARID  
27 framework.

28 As an example of how input was used to develop ARID's framework, the community drove the  
29 ARID scoping team to broaden our scope and revise some of our originally proposed science  
30 themes (Appendix F.9). For example, biodiversity and wildfire were only sub-components of our  
31 original framework, and cropland and rangeland management were not prominently considered.  
32 However, extensive community recommendations and excitement about these topics warranted  
33 them being explicitly elevated as central themes. The community also confirmed that dryland  
34 water availability and carbon stocks and fluxes were priority interest topics. In another example,  
35 there is strong interest in ARID having both domestic and international domains, which  
36 encouraged the domain to be both domestic, taking advantage of reliable logistics and exciting  
37 opportunities to meet US information needs - and international - allowing us to broaden the  
38 range of insight and build a transdisciplinary international dryland research community.

39  
40 While there is strong interest in sites in the southwestern US, we received overwhelming input  
41 to include the full dryland western US in the domestic ARID domain, including the Great Basin  
42 and western Great Plains (Appendix F.9). Community interest in the Great Plains encouraged  
43 us to conduct roundtable discussions with researchers and end-users in rangeland and cropland



1 landscapes across Colorado, Nebraska, and Texas. This included a meeting at University of  
2 Nebraska in February 2024 as a part of the “Harnessing the Heartland” effort, as well as  
3 roundtables in Fort Collins, CO with Colorado State University and USDA Agricultural Research  
4 Service (ARS) throughout Spring 2024. Furthermore, a virtual Great Plains roundtable was held  
5 with over 30 scientists and end-users (across universities and agencies in Texas, Oklahoma,  
6 Colorado, Wyoming, Kansas, and Nebraska) in June 2024 with the goal of understanding  
7 rangeland and cropland research and land management gaps, as well as understanding how to  
8 leverage monitoring sites like National Ecological Observatory Network (NEON) and Mesoscale  
9 Network (Mesonet).

10  
11 Furthermore, while we received strong input to include many global dryland locations across  
12 Australia, South America (such as the Cerrado and Caatinga), and Asia (such as Mongolian  
13 grasslands), the community showed substantial interest in Africa and, particularly, southern  
14 African countries. Community input also included disproportionately high interest in Mexico, both  
15 because of the relevant research that could be linked, and also because ARID’s proposed  
16 western US domain already included portions of the Chihuahuan and Sonoran deserts, which  
17 are shared by the U.S. and northern Mexico. There are also significant knowledge gaps for  
18 Mexico, and science in the region could help address requests by indigenous and rural  
19 communities in Mexico for improved understanding of their lands and the ways they are  
20 changing. Additionally, this feedback motivated engagement with, for example, the Okavango  
21 Research Institute (Botswana), the Gobabeb-Namib Research Institute (Namibia), and  
22 investigators of the MexFlux sites located in northern Mexico. We also held an ARID South  
23 America Workshop which included 59 researchers, many of which were from research  
24 institutions across South America and particularly Brazil.

25  
26 The ARID scoping study team also met with NASA science teams to determine their field and  
27 airborne data needs and to find synergies with their planned activities. This included meetings  
28 with the NASA Land-Cover and Land-Use Change (LCLUC) program, NASA SBG team, NASA  
29 Soil Moisture Active Passive (SMAP) team, NASA ECOSTRESS team, NASA Orbiting Carbon  
30 Observatory-2/3 (OCO-2/3) team, and NASA GEDI team. Our ARID team members are also  
31 finding synergies with a diverse array of efforts, including with ongoing domestic field network  
32 efforts (e.g., NEON, LTER, AmeriFlux), international institutes and initiatives, non-government  
33 organizations (NGOs), established university dryland centers and institutes, and others such  
34 that ARID can leverage resources and personnel across many partnerships.

35

## 36 C.3 End-user Engagement and Decision Support

### 37 ***C.3.1. ARID’s End-User Engagement Vision***

38 ARID envisions an active engagement effort throughout the duration of the field campaign, with  
39 end-users and practitioners dealing with environmental changes affecting the ecosystem  
40 services communities rely upon for sustaining livelihoods. Our end-users include managers and  
41 decision makers from a diverse set of groups, such as natural resource managers, tribal  
42 communities, rangeland managers, agronomists, and conservationists. The engagement is

1 envisioned to operate in a co-production paradigm, with close collaboration between ARID  
2 research community and the end-user groups. Particularly, it is key to start discussion with end-  
3 users to determine the data or process that best fits their needs. For achieving the highest  
4 application readiness levels, it is also critical to establish the datasets, knowledge, and/or tools  
5 needed that are directly useful and feasible for these end-user needs. However, not all  
6 proposed ARID field campaign measurements are well-positioned to address end-user needs  
7 and thus ARID science definition team and individual proposing PIs would need to champion  
8 applications in collaboration with end-users (see Section B.2.7).

9 This vision was developed from our end-user engagement during scoping. In speaking with  
10 these different groups of end-users, we've identified groups and individuals extremely interested  
11 in working with ARID-funded scientists, and these established relationships can continue to be  
12 developed throughout the ARID campaign. We also identified a range of end-user needs  
13 specific to drylands and the drylands in which end-users live and work.

### 14 **C.3.2. ARID's End-User Engagement Strategy**

15 To facilitate engagement activities during the implementation of ARID, engagement with diverse  
16 end-user and practitioner groups was undertaken to develop a better sense of the scope of  
17 information needs and data products that might be useful to these end-user groups. The  
18 strategies outlined here can be further developed throughout the ARID campaign. In order to  
19 reach the larger goal of actionable science, it is essential to continue fostering relationships with  
20 ARID's end-users.

21 *Our engagement strategy included in-person meetings in small group settings to speak with*  
22 *end-users, rights-holders (tribal), and scientists who work with end-users communities. Because*  
23 *of the specific needs of each end-user, smaller group discussions and one-on-one meetings*  
24 *with end-users were prioritized. We also allowed the end-user(s) to guide the conversations*  
25 *given the unique (and often less predictable) topics for each discussion and the fact that they*  
26 *were almost always more well-equipped to address their needs than ARID personnel. This is in*  
27 *contrast to meetings with the scientific research community, which tended to include more*  
28 *generalizable discussion topics.*

29 We identified organizations, agencies, and individual personnel we felt would potentially co-  
30 develop the campaign and also directly benefit from using ARID field campaign measurements  
31 and developed knowledge. We invited these end-users to our in-person meetings and, in some  
32 cases, for one-on-one virtual discussions.

33 For relationship and capacity building, we felt it was important to travel to where the end-users  
34 live as much and as early as possible in the scoping to conduct these discussions. This included  
35 in-person meetings in Arizona, New Mexico, Utah, Nevada, Colorado, and on tribal lands. ARID  
36 could be conducted in these locations and co-developed and led by in-place partners. For  
37 example, our ARID leadership team held a kickoff meeting on October 23rd-26th, 2023, at the  
38 University of Arizona in Tucson, Arizona (Fig. C.2). Given that applying research for action  
39 requires starting with understanding end-user needs, we wanted the first day of the first ARID  
40 event to begin with listening to the perspectives, expertise, and input of the data user  
41 community. The event began with keynote speakers from the U.S. Bureau of Land Management

1 (BLM). We then divided into four smaller groups to allow our visiting data end-users to offer their  
2 thoughts on the value of a NASA field campaign focused on drylands, and the utility of a  
3 ground-to-satellite approach to better characterize and understand the drylands they manage.  
4

5 We believe it is critical for ARID to leverage boundary personnel and organizations, or those  
6 that support end-user engagement of existing efforts to co-develop usable products from  
7 various NASA activities. These groups include personnel that are experts at gathering input  
8 from end-users and communicating with scientists about information needs. By streamlining the  
9 process of connecting with end-users in having formed relationships, they also mitigate end-  
10 user “burnout” from end-users being contacted to excess by scientists with similar questions.  
11 The ARID team identified and engaged with several of these groups during scoping. These  
12 groups agreed that they can, to varying degrees, informally advise the ARID science team with  
13 overall co-development with end-users. They include NASA SERVIR and USAID partners  
14 conducting work in southern Africa (primarily in Mozambique), NASA Western Water  
15 Applications Office (WWAO), NASA Carbon Monitoring System (CMS) team members who  
16 work with end-users to varying degrees especially in the western US, NASA’s Harvest, NASA’s  
17 Indigenous People’s Initiative (IPI), and the NASA Satellite Needs Working Group (SNWG) and  
18 USGS National Land Imaging Program that both extensively gather end-user needs to inform  
19 satellite mission development.  
20

### 21 **C.3.3. End-User Engagement Summary**

22 ARID’s end-user engagement included workshops, roundtable discussions, written input, and  
23 individual meetings with boundary personnel and other US agencies. The ARID scoping team  
24 held end-user focused workshops and roundtables in Arizona, New Mexico, and Colorado,  
25 engaging a vast array of end-user knowledge in the western US. They also held several  
26 individual meetings with personnel from several US agencies that would likely serve as end-  
27 users. A more detailed list of meetings and dates can be found in Appendix F.6.  
28

29 *Tucson Workshop:* An ARID end-user focused workshop was held at University of Arizona in  
30 Tucson, AZ, USA on October 23, 2023. Over 30 end-users attended in person with more  
31 attending virtually. It consisted of two keynote speakers from BLM - Karen Prentice, National  
32 Science Advisor for the BLM and Jon Norred, BLM Branch Chief of Resource Data Services -  
33 and breakout group discussions. The majority of these data end-users are land and resource  
34 managers in Arizona, New Mexico, and Utah with affiliations in the Bureau of Land  
35 Management, the National Park Service, the U.S. Department of Agriculture, the University of  
36 Arizona, U.S. Army Corp of Engineers, The Salt River Project, The Nature Conservancy,  
37 several non-profit and private companies, and private landowners, including cattle ranchers.  
38 Several end-users could not attend the meeting but provided written input.  
39

40 *Albuquerque Workshop:* An ARID sovereign nation-focused workshop was held at University of  
41 New Mexico in Albuquerque, NM, USA on May 7, 2024. This meeting included discussions with  
42 tribal scientists, natural resources departments, educators, and liaisons about tribal land  
43 management needs, education needs, and data sovereignty. Eight tribal members were in  
44 attendance, and we acknowledge their affiliations across Rosebud Sioux Tribe, Navajo Nation,

1 Hopi Tribe of Arizona, Cherokee Nation, Pueblo of Isleta, and Pueblo of Santa Ana. Our  
2 meeting also included a visit to a high school in Tuba City, Arizona on the Navajo Nation to  
3 demonstrate to over 100 students the use of field instruments and discuss how ARID can  
4 provide wider use of such instruments. Follow-up meetings were conducted with USGS-based  
5 and NASA-based tribal liaisons based in New Mexico as well as natural resources departments  
6 of several Pueblos in New Mexico.

7  
8 *Roundtables:* ARID team leadership held in-person roundtables with several USDA ARS land  
9 managers in Fort Collins, CO in February and May 2024 and discussed cropland and rangeland  
10 management in the Great Plains. In September 2024, they also spoke at and held listening  
11 sessions at virtual webinars at the Applied Earth Observations Innovation Partnership (AEOIP)  
12 webinar which included approximately 20 end-users from the US Forest Service, and held a  
13 round table with approximately 15 USFS Rocky Mountain Research Station scientists and  
14 managers.

15  
16 *Washington DC End-User Focused Roundtables:* Several ARID team leadership members held  
17 meetings in Washington DC on July 17-19, 2024, with several federal agencies to define  
18 synergies and partnerships that the ARID campaign could both leverage and use to extend the  
19 work's impact. In particular, end-user governmental personnel at DOI and BLM, USDA Office of  
20 Energy and Environmental Policy, and USGCRP are directly considered end-users, while others  
21 can connect the ARID field campaign data to end users.

22  
23 *Boundary Organization Meetings:* ARID team members held individual meetings with personnel  
24 at boundary organizations including NASA SERVIR, USAID, NASA WWAO, NASA CMS, NASA  
25 IPI, NASA's SNWG, and USGS National Land Imaging Program.

26  
27 *Other End-User Meetings:* ARID team members also conducted individual meetings with USFS,  
28 BLM, USDA ARS, NRCS, DoD, Navy, BoR, NPS, USFWS, and mining companies.

#### 29 30 **C.3.4. Summary of End-user Needs**

31 To meet end-user needs, it would be necessary to maintain continued engagement with the  
32 end-user communities and practitioners, including tribal communities. These engagement  
33 efforts would facilitate co-production of products useful in decision making for managing  
34 ecosystem services and utilizing research products from ARID research. We envision ARID  
35 working to sustain connections with end-user communities throughout the course of the  
36 campaign. The early portion of the campaign could support a working group of end users to  
37 initiate development of new usable products derived from currently available remote sensing  
38 products. This initial effort could establish a framework to launch new co-production efforts from  
39 ARID's measurements and research findings.

40  
41 While we received a diverse array of feedback from end-users throughout the engagement  
42 activities highlighted in C.2.3, several themes arose. We provide a few examples here.

1 *Data Access and Use Challenges:* A common theme was that most end-users have extensive  
2 field sampling capacity and, often, strong science backgrounds in understanding dryland  
3 ecosystem processes. However, despite NASA's and other space agencies' freely available  
4 data products that could support data end-users, land managers find it challenging to access  
5 products and tools directly applicable for their needs. For example, during our ARID kick-off  
6 meeting, we learned that BLM faces technical challenges including storage and retrieval  
7 capacity for very large remote sensing datasets and limited in-house knowledge for how to  
8 process and use the data and technologies. There is substantial value in external scientific  
9 support to help build technical frameworks to process data into products, tools, and services  
10 useful for land managers and agencies such as the BLM in their decision-making process.

11  
12 *High Resolution Data Needs:* End-users and boundary personnel consistently discussed the  
13 need for plot level data, or high spatial resolution information relevant to their site conditions for  
14 monitoring vegetation types, soil moisture, evapotranspiration, animals, fire, streamflow, and  
15 mineral mining. They often mentioned needing data-informed decision tools, such as dryland  
16 plant productivity and vegetation type maps, so that they can more closely monitor, for example,  
17 vegetation growth and/or invasive species. A NASA Terrestrial Ecology field campaign would  
18 provide more in-situ and airborne measurements as well as efforts to scale field data to satellite  
19 observations and generate end-user relevant variables in the service of decision-support tools.

20  
21 *Fostering the Application Process:* Incorporating research and data products into the decision-  
22 making process with end-users, especially at BLM, requires time. The data product is not the  
23 end of the application process; there is a need for scientists to directly work with decision  
24 makers early and iteratively throughout the process. With the longer multi-year-to-decadal  
25 timeline often permitted in a NASA Terrestrial Ecology field campaign, these relationships could  
26 truly be solidified with significant benefit for science being useful and used.

27  
28 *Early engagement:* It is critical to engage with end-users early. End-users requested that ARID  
29 scientists develop data products by starting from end-user needs and working backwards. If this  
30 step is avoided, there is often a large gap between the data need and the data product and,  
31 consequently, applications and implementation into end-user decision frameworks does not  
32 occur.

33

## 34 C.4 Tribal Engagement

35 Tribal Nations (First Nations) are sovereign nations, and are located throughout ARID's  
36 proposed domestic western US domain. These personnel have extensive experience and have  
37 passed knowledge along for generations (at times for millennia) about their land, and they are  
38 thus pivotal for enhancing the ARID campaign and its outcomes. ARID recognizes the value of  
39 traditional knowledge as a systematic way of knowing that informs and increases our  
40 understanding of drylands, creates a more comprehensive knowledge base, and provides  
41 innovative solutions to help mitigate and adapt to dryland transformation in the face of change.  
42 Tribal members often have first-hand experience and a keen interest in how climate change is

1 impacting their environment. ARID can work to include this expertise in the assessment of  
2 drylands ecosystems, their change, and opportunities for adaptation, and outreach activities  
3 specifically designed to inform Tribal communities of the results of ARID will be developed.  
4 Tribes have their own unique goals, needs, and resources that do not necessarily align with all  
5 aspects of end-users of NASA data and tools as discussed in Section C.2. Rather, engaging  
6 with Tribal Nations should be focused on co-developing work such that both parties (Tribal  
7 Nations and non-tribal NASA funded researchers) mutually benefit from any interactions.

8  
9 The ARID team received feedback early during the scoping to prominently include Tribal  
10 Nations within the ARID scoping effort and within the field campaign. Acknowledging deeply  
11 impaired relations between tribes and the US government, engaging with tribes requires time  
12 and care through trust building. As such, early engagement with tribes is critical for allowing the  
13 time to build relationships and define shared values. We, therefore, share our goals,  
14 engagement techniques, engagement summaries, understanding of tribal needs, and proposed  
15 co-development strategy. While we feel we have made substantial progress engaging with  
16 tribes during the ARID scoping study, we acknowledge that our tribal engagement effort could  
17 be greatly improved and, therefore, we provide these details here in such that the ARID  
18 campaign and any future efforts that co-develop with indigenous peoples can build on and refine  
19 our framework and strategies.

#### 20 21 **C.4.1. ARID Tribal Engagement Scoping Goals**

22 During scoping, we aimed to accomplish the following with regard to engagement with Tribal  
23 Nations:

- 24 1) Engage with Tribal personnel to a degree that the following goals can be sufficiently met  
25 while avoiding overburdening their time commitments.
- 26 2) Provide a general awareness of a potential ARID campaign and define opportunities to  
27 directly be involved
- 28 3) Define potential NASA ARID resources and Tribal Nations resources that, together, can  
29 co-develop new knowledge about drylands and/or new land management strategies
- 30 4) Gain an initial understanding of common goals and needs of different Tribal Nations,  
31 particularly related to land and resource management
- 32 5) Understand methods to co-develop new knowledge while applying and protecting  
33 traditional ecological knowledge
- 34 6) Define boundaries and required permissions that tribes have with regard to ability to  
35 conduct the ARID campaign within Tribal Nations
- 36 7) Include Tribal personnel directly within ARID leadership and especially discussions  
37 related to Tribal co-development
- 38 8) Ensure open discussions of data sovereignty and the way data can and cannot be used  
39 and shared
- 40 9) Build a Tribal engagement strategy to be used and refined by NASA and other agencies  
41 for years to come.

#### 42 43 **C.4.2 Tribal Engagement Techniques**

1 *Gaining an understanding of tribal needs:* we learned that Tribal needs are unique and diverse  
2 across different communities and across individuals. It is not possible to view one tribal member  
3 as a spokesperson for all Tribal Nations, and in some cases for their own Tribal affiliation. We  
4 therefore held multiple discussions within tribes with liaisons, educators, students, and land  
5 managers to broaden our understanding of goals and needs. We also held discussions with  
6 members with affiliations across over 10 tribes across the western US.

7  
8 *Building long-term relationships and trust:* A general strategy was to reach out to tribal members  
9 early during scoping and make clear that their input would guide our campaign. Early  
10 engagement is critical for building trust with typically a more positive reception and allowing  
11 more time to build relationships. However, response rates of Tribal Nations to “cold contacting”  
12 tended to be low, especially when there was not an existing relationship. A successful strategy  
13 was to first leverage existing relationships that ARID team members and NASA colleagues had.  
14 From these initial conversations stemming from established relationships, Tribal personnel  
15 would commonly recommend and facilitate building new relationships with other Tribal  
16 personnel that would provide value to the scoping. Tribal liaisons are also key figures for  
17 facilitating this process because they often have tribal affiliation(s) themselves and have built  
18 relationships. They are also aware of common needs and goals across many tribes. Finally, it is  
19 critical that the relationships are long term and thus collaborations should be set up to start early  
20 and last beyond the bounds of the field campaign.

21  
22 *Prioritizing in-person interactions:* To convey our sincerity for meaningful and continued  
23 engagement, most of our interactions with tribal members included ARID team members  
24 traveling to meet them where they live.

25  
26 *Inviting Tribal members to guide conversations:* It is common for tribal members in the western  
27 US to spend time listening, pausing to think, and waiting to be invited to contribute to a  
28 conversation. This often results in non-tribal members from western cultures dominating the  
29 conversation. Therefore, within our interactions, we prioritized asking questions, allowing  
30 pauses and silence, and invited tribal members to provide their thoughts.

### 31 **C.4.3 Tribal Engagement Examples**

32 We provide several examples of ARID’s engagement with Tribal Nations, which provided  
33 awareness of the ARID campaign in the form of both deeper-dive conversations and broad  
34 feedback.

35  
36 *Tribal Engagement Workshop:* The ARID team held a Tribal focused workshop at University of  
37 New Mexico in Albuquerque, NM on May 7th, 2024, with 8 tribal members in attendance that  
38 are liaisons from NASA and USGS, educators (high school and university), and land managers.  
39 They included affiliations across the Rosebud Sioux Tribe, Navajo Nation, Hopi Tribe of Arizona,  
40 Apache tribe, Cherokee Nation, Pueblo of Isleta, and Pueblo of Santa Ana. Tribal members  
41 discussed development of a tribal engagement framework, college-level educational needs,  
42 research experiences and pitfalls, and opportunities for liaison support. ARID Team leadership  
43 also traveled to Sante Fe, NM to meet with a NASA IPI liaison.  
44



2  
3 **Figure C.4. (Left)** ARID team discussing tribal engagement strategies at  
4 **University of New Mexico. (Right)** Faculty at SIPI, a tribal college, giving a talk  
5 about SIPI's tribal program and collaboration opportunities with NASA and ARID  
6

7 *Navajo Nation High School Student Engagement:* ARID Team leadership traveled to Tuba City,  
8 AZ to meet with Navajo Nation and Hopi Nation high school students. The ARID team taught  
9 two classes with 50 students in each about how to analyze multi-spectral remote sensing data  
10 using NASA STELLA instruments as well as discussed, the ARID campaign and how it would  
11 present research and educational opportunities. ARID team leadership also conducted a  
12 broadcast to KGR 91.3 FM about the potential impact of ARID on Navajo Nation and its next  
13 generations and NASA's engagement with the high school. This broadcast was released on  
14 September 2024 and has been running on a 30 minute segment every few hours on the KGR  
15 91.3 FM, which services much of the northwest Navajo Nation in northeastern Arizona.  
16

17 *Washington D.C. Tribal Focused Meetings:* James Rattling Leaf Sr. joined ARID team  
18 leadership in D.C. as a representative of the Rosebud Sioux Tribe and their new Climate  
19 Center. In all meetings, he described the importance of early engagement with Tribes, not only  
20 within NASA campaigns but also across agencies. He highlighted how the Rosebud Sioux  
21 Tribe's Climate Center is developing climate adaptation plans and determining how data can be  
22 combined with traditional ecological knowledge for Tribal Nations. James discussed excitement  
23 about ARID and its framework for engaging Tribes early and often and for considering how to  
24 incorporate Tribal data needs, expertise, and actionable science opportunities. He noted the  
25 importance of campaigns like ARID including Tribal leadership directly on NASA science teams.  
26 Such principles were encouraged to be used not only for ARID but across any future campaign.  
27

28 *Tribal Land Management Focused Discussions:* ARID Team leadership met with tribal land  
29 managers across three tribes in February, April, and May 2024 to discuss their land  
30 management challenges, discuss their climate adaptation plans, and/or find synergies with how  
31 they can co-develop new knowledge and methods with ARID. This included Ute Mountain Ute  
32 Tribal Council, Pueblo of Jemez, and Pueblo of Santa Ana.  
33

34 *Lakota meeting:* TBD October 2024



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**C.4.4 Tribal Resources, Goals, and Needs**

In having conversations with tribal members during our visits, we summarize several goals, resources, and needs here:

*Land Management:* While land management goals range widely across tribes, common themes arose. These include buffalo restoration throughout the Great Plains, including tracking and protecting their movements. Most of their needs in the southwestern US are related to mitigating issues of low water availability for crops and grazers. They also mentioned issues with invasive species and monitoring needs. High resolution satellite and drone measurements can be useful for helping to identify these species instead of counting and surveying individual plants manually.

*Remote sensing assistance:* Typically, Tribal Nations have natural resources departments and different instruments, such as in-situ monitoring devices and drones. These departments range widely in availability of resources and personnel on staff. However, like our US agency end-user partners, not all were well-versed in processing and applying remote sensing data, but agreed the data would be valuable for their decision-making processes. Therefore, ARID personnel can directly be involved through providing satellite measurement experience and expertise and guiding the downloading, storage, and use of the datasets.

*Climate Adaptation Plans:* Tribal Nations often have climate adaptation plans they are creating or revising for their specific land and needs, and how they plan to address these needs under varying and changing climate conditions. Measurements and findings made during ARID can inform and supplement these plans.

*Educational:* Tribal students are interested in learning science and remote sensing, and often specifically for how it would address their Tribe’s needs. Therefore, ARID can provide a range of opportunities to train Tribal students. However, ARID leadership learned that it is a sensitive topic of what Tribal students want to do after completing their education, with often a preference to live near family and/or on their tribal land. Therefore, it is highly important for ARID to provide a range of opportunities for Tribal students for them to choose.

**C.4.5. Tribal Co-Development Strategy**

We provide several strategies and identify resources ARID would bring that can facilitate co-development with Tribal Nations. For a successful completion of ARID in co-development with tribes, these strategies will need to be advanced and updated.

*ARID Team to include Tribal personnel:* Including Tribal personnel on the ARID leadership team is critical for facilitating the co-development of the ARID campaign with tribes, identifying resources in both parties, and utilizing and protecting traditional ecological knowledge. In these cases, tribes may need to appoint a leader based on decisions of tribal elders. Therefore, it is critical to begin the process early of requesting tribal personnel to join the ARID team.

1 *Tribal PI's and Co-I's*: It is encouraged for tribal members to lead ARID projects as PIs and for  
2 non-tribal PI's to include tribal Co-I's. Such collaborations can facilitate co-development  
3 practices directly.

4 *Mentorship and Educational Opportunities*: ARID can facilitate a range of opportunities for Tribal  
5 early career involvement. NASA's MUREP MAIANSE provides NASA funding for tribal early  
6 career scientists to conduct research alongside NASA researchers virtually or onsite. MAIANSE  
7 coordinators are aware of ARID and are excited to facilitate opportunities for Tribal students to  
8 be involved in the ARID campaign. Similarly, on-site NASA center fellowships, like the NASA  
9 Postdoctoral Program, is another avenue for more intensive exposure and training with remote  
10 sensing data and the ARID campaign. Other options include exposing tribal students to NASA  
11 field work and instruments in partnership with professors at Tribal colleges. ARID team  
12 members have various means to contact Tribal students about opportunities including via tribal  
13 college professors who have connected with ARID during the scoping study phase, and tribal  
14 conferences like AISES.

15 *Training and Informational Communication*: Given new datasets and new products that would  
16 be developed during ARID, it will be critical to collate and share these developments with tribes.  
17 A task of the ARID science definition team can be to develop a method to assist tribal members  
18 with accessing, downloading, and/or processing publicly available NASA data. Available online  
19 NASA ARSET trainings can be leveraged in this regard.

20 *Continued Engagement*: It will be critical for the ARID scoping team to continue to meet with  
21 existing tribal partners and form new tribal partnerships to facilitate the building of those  
22 relationships with, if funded, the ARID field campaign leadership team. Relationships should be  
23 developed into long-lasting collaborations that should continue long after a field campaign. If  
24 ARID is funded, outreach should be more widespread given the ability to more clearly  
25 communicate funding opportunities and available resources than during scoping.

26

#### 27 **C.4.6. ARID Proposed Data Sovereignty and Data Policies**

28 A more comprehensive data management plan is discussed in Section D, but here we state our  
29 commitment to good stewardship of data sovereignty. Data collected by Tribal Nations will only  
30 be used if permission is granted. Traditional ecological knowledge (TEK) should only be used in  
31 a project and published if permission is granted by Tribal members and leaders. If  
32 measurements on Tribal lands are desired, we will directly seek permissions for any airborne or  
33 ground-based measurements campaigns and assure that data will not be collected on tribal  
34 lands without tribal approval.

35

## 36 **C.5 Preparing the Next Generation of Earth Scientists and Practitioners**

### 37 *Early Career Scientist Engagement during Scoping*

38 The ARID team involved early career scientists throughout scoping through several means. (1)  
39 Over 50% of the 169 ARID working group participants were early career scientists,  
40 demonstrating that the proposed ARID campaign was co-designed with their direct technical

1 input. This number appears representative of similar proportions of early career scientists who  
2 participated in conference townhalls, roundtables, workshops, and other engagement events.  
3 (2) Over 50% of over 350 written inputs sent to the ARID team also came from early career  
4 scientists, demonstrating both their interest levels in the ARID campaign and their contributions  
5 that contributed to building the ARID framework (see Appendix F.9). (3) The ARID team  
6 leadership included early career scientists. A. Feldman was co-I and co-lead of the ARID  
7 scoping study and seven other early career scientists were members of the ARID scoping team,  
8 showing that ARID’s vision was directly developed by the scientists that would participate in the  
9 duration of the ARID campaign. Including early career researchers on the leadership team also  
10 ensured more early career involvement throughout the engagement process.

11  
12 There were many notable engagement events that directly engaged or focused on early career  
13 researchers during ARID scoping. Examples include: (1) ARID team leadership met with  
14 approximately 100 high school students at a high school in Navajo Nation (Fig. C.5; see  
15 Appendix F.7) and provided a general overview and proximal remote sensing instrument  
16 demonstration about how multi-spectral instruments can observe relevant features of the land  
17 surface. (2) The ARID team discussed educational needs with a SIPI professor (see Appendix  
18 F.2) about how to involve tribal early career researchers and provide training opportunities. (3)  
19 ARID team leadership attended the 6th Federal UxS Workshop at SIPI (see Appendix F.7) and  
20 had direct interactions with students about dryland science and remote sensing. (4) ARID team  
21 leadership mentored DEVELOP summer interns regarding a dryland project and noted how  
22 such internship opportunities provide training and career pathways into NASA and ARID-related  
23 research.



**Figure C.5.**  
*ARID team leadership discussing remote sensing and field campaigns like ARID with students at Greyhills Academy High School in Navajo Nation.*

42  
43 *Vision for Early Career Scientist Involvement in ARID*  
44 ARID is committed to fostering the development of early career researchers throughout the  
45 ARID campaign, and propose several diverse approaches: (1) Inclusion of early career

1 scientists on ARID science definition team throughout the campaign. (2) Encouragement and  
2 commitment to recruit underrepresented early career scientists to be participants on ARID-  
3 funded grants, such as through ESA SEEDS and from tribal conferences like AISES. (3)  
4 Encouragement for ARID-funded scientists to fund and mentor early career researchers on  
5 ARID grants as well as through internships like NASA’s summer internship program, NASA’s  
6 MUREP MAIANSE (tribal internship opportunities through NASA), and fellowships like the  
7 NASA Postdoctoral Program. (4) Encourage ARID scientists, end-users, and partners to  
8 participate in NASA’s ARSET virtual and in-person training, as well as establishing more  
9 specific in-person remote sensing training such as HYR-SENSE, as conducted by NASA SBG.

10  
11 Expanding more on the third point in the previous paragraph about early career funding,  
12 research projects supported by NASA typically fund graduate and undergraduate students, as  
13 well as post-docs, and we expect this to continue throughout ARID. With one of ARID’s focal  
14 regions being the western U.S., we have also worked with other U.S. agencies to outline  
15 educational support that could be independent of PI-funded projects in ARID, for example, the  
16 USGS Mendenhall Program and the Presidential Management Program. Our partnership with  
17 the Nature Conservancy also highlights the NatureNet Science Fellows, which is specifically  
18 designed to “embrace existing and emerging technologies”. There are also opportunities for  
19 graduate students to develop their own independent research, independent from their advisor’s  
20 projects, through NASA’s Earth and Space Science Fellowship program. ARID can facilitate  
21 further involvement of undergraduate students via collaborations with the numerous projects  
22 funded by NSF, such as, NSF’s Research Experience for Undergraduates (REU) program, as  
23 well as through NASA’s GLOBE program, and the collaborations between NSF and GLOBE.

## 25 C.6 Diversity, Equity, Inclusion, and Accessibility

26 ARID is committed to building a diverse and inclusive community with equitable and accessible  
27 opportunities and interactions throughout the scoping study. ARID adhered to these values  
28 throughout the scoping study and will continue to value these efforts throughout the ARID  
29 campaign. We describe our vision here, but refer the reader to our tribal and early career  
30 scientist engagement efforts (throughout Section C) for efforts that demonstrate the scoping  
31 team’s vision for diversity, equity, inclusion, and accessibility.

32  
33 As a part of commitment to diversity and inclusion, we made and will continue to make strong  
34 efforts to include tribal voices directly within ARID. ARID team leadership worked hard to build  
35 relationships with tribal members. Such partnerships and co-development bring invaluable  
36 ideas, perspectives, and in-place knowledge to ARID. This includes membership directly on the  
37 ARID leadership team, including Raymond Kokaly and James Rattling Leaf Sr. and extensive  
38 tribal engagement throughout scoping (see Section C.3). We will maintain commitment to  
39 gender diversity, with scoping including a PI and several members of ARID leadership who are  
40 women. Ultimately, we span a wide range of backgrounds, institutions, and career stages. Such  
41 commitment to diversity needs to remain active throughout the ARID campaign. It is especially  
42 critical to recruit underrepresented researchers to be funded participants of ARID, including at

1 early career stages to build a diverse NASA and Earth science community. This includes  
2 recruiting from networks that support inclusion of underrepresented scientists (i.e., Black, Latinx,  
3 Tribal, etc.) including at AISES and ESA SEEDS.

4  
5 To maintain equitable and accessible interactions between scientists, we developed a Code of  
6 Conduct (Section D.3) to set expectations for safe and meaningful interactions and to ultimately  
7 retain underrepresented ARID participants. These ideals were upheld and enforced throughout  
8 the scoping study. For all ARID participants, ARID should require field safety and bystander  
9 intervention trainings, such that all are aware of the Code of Conduct, understand  
10 consequences, and know how to intervene. Such trainings are often virtually available at NASA  
11 centers, which can be leveraged during the campaign. Finally, the ARID leadership team will  
12 develop a reporting system that has anonymous options and is independent of any human  
13 resources department such that victims feel comfortable discussing and reporting offenses. This  
14 will include several designated reporting officers which serve as points of contacts and will be of  
15 varying demographic backgrounds (career stage, gender, race, etc.).

## 17 C.7 Capacity Building: Education and Training

18 ARID's capacity building plan will develop and strengthen the skills, capabilities, available  
19 resources, and structures to meet NASA and community goals to educate, train, and uplift  
20 dryland communities. This will encompass a wide range of activities, including training,  
21 mentoring, and resource allocation, aimed at enhancing capabilities in areas including research,  
22 data accessibility, and technical proficiency. We seek to empower people and organizations to  
23 be well-trained and self-sufficient, supporting the dryland community in addressing challenges  
24 and seizing opportunities over the long term. ARID's process is not only targeted at providing  
25 technical knowledge and tools; it also emphasizes the creation of enabling environments, such  
26 as inclusive policies (see C6. Diversity, Equity, Inclusion, and Equitability), access to information  
27 (see D.3 Data Management and D.4 Computing Resources), and collaborative networks (see  
28 C.1 and D.2 Code of Conduct). Ultimately, ARID's capacity building aims to create long-lasting  
29 improvements that allow organizations or communities to grow, thrive, and be resilient in the  
30 face of change.

31 ARID was built strongly on NASA's long-standing commitment and resources in support of  
32 capacity building, as well as on the opportunities and knowledge provided by agencies such as  
33 USAID. NASA SERVIR, ARSET, DEVELOP, GLOBE, and the Indigenous Peoples Initiative  
34 have been integral in designing opportunities for ARID to build capacity, and, during the scoping  
35 study, as described in the other sections of Section C, we performed consultations and needs  
36 assessments, end-user/practitioner mapping, service design, as well as monitoring, evaluation,  
37 and learning and used SERVIR's capacity building model as framework for international  
38 capacity building.

39  
40 ARID has specific focus on building capacity and providing educational opportunities for tribal  
41 communities. Strong opportunities exist to directly collaborate with these communities to build

1 capacity in several ways. Tribal communities have powerful knowledge and understanding of  
2 dryland ecosystems and are actively involved in documenting and adapting to change.  
3 As we learned through our tribal partners during scoping, some of the training and skill building  
4 opportunities that work well for other scientists may not be appropriate for some tribal members  
5 who are otherwise ready to make great strides in advancing science. Additional options for tribal  
6 capacity building are needed to provide opportunities for tribal students and scientists to be  
7 exposed to NASA ARID, with both field measurements and satellite data. From NASA's  
8 foundational work and from our scoping, we know this can include seminars and training for  
9 non-tribal scientists who are engaging with tribal scientists and students, on-site trainings for  
10 tribal and other rural communities where the scientist comes to them, exposure to instruments  
11 that does not require travel, and support in working with data and analysis remotely. ARID could  
12 fund installment of instruments at tribal colleges and universities (TCUs), as well as scientist  
13 visits to TCUs to provide educational and training opportunities on field instruments. Professor  
14 Dennis Dye at SIPI Tribal College suggested a powerful approach in ARID, for example,  
15 supporting trainings such as Hyr-Sense through ESIL (ARID team member Cibebe Amaral is a  
16 staff scientist at ESIL and has extensive experience in working with tribal partners.) Also  
17 possible are NASA Internships and fellowships, such as MAIANSE, Summer Intern Program,  
18 FINESST, and NASA Postdoctoral Program.

## 19 **D. Management and Plan of Work**

### 20 **D.1 Management Structure and Personnel Responsibilities**

21 We describe our vision for the ARID science team structure, with the overall network illustrated  
22 in Fig. C.3. We envision a core "ARID Science Team" that consists of the science definition  
23 team, any directed science teams, and the competed science team funded through ROSES, as  
24 well as the potential for partner science teams. We recommend that the NASA CC&E office  
25 supports and advises ARID. It is critical that the ARID Science Team includes representation  
26 across the science community from early to late career, domestic and international researchers,  
27 and with Tribal Nation representation. Officers should be defined within the ARID team that will  
28 focus on continued community engagement, field campaign safety (handling enforcement of  
29 ARID's code of conduct in Section D.2), and coherence of ARID's efforts across satellite,  
30 airborne, ground, and modeling strategies.

31  
32 The ARID team will also have an independent advisory committee which includes science  
33 community, agency (BLM, USFS), Tribal, international, and end-user/practitioner representation.  
34 This independent advisory committee can consist of a wide range of communities that can  
35 assist with end-user engagement (boundary organizations and personnel experienced in  
36 working with end-users and practitioners) such as the NASA SNWG and NASA CMS team.  
37 NASA IPI can advise tribal engagement. NASA SERVIR can advise international engagement  
38 and field deployment. The committee can also provide independent annual reviews of ARID's  
39 activities in scientific merit of findings, campaign planning strategies, field safety, and  
40 inclusiveness of campaign efforts and engagement.

1  
2 Throughout the ARID implementation in Section B.2, there are two main recommended directed  
3 efforts to include within the ARID Science Team. First, ARID's modeling efforts should partially  
4 be directed, with two different modeling teams focusing on Earth system modeling and dynamic  
5 vegetation modeling (see Section B.2.2). Second, we recommend a larger, directed end-user  
6 engagement effort with BLM and USFS to support their modeling and decision-making  
7 frameworks (see Section B.2.7). Both of these efforts across modeling and applied sciences  
8 efforts are recommended to be directed given the continuity involved beyond a typical three-  
9 year ROSES effort.

10  
11 We additionally recommend different management and campaign planning strategies for  
12 domestic and international efforts. For domestic efforts, the science themes of ARID outlined in  
13 B.1 and the implementation strategy in B.2.6 should be prioritized. By contrast, international  
14 activities should strongly rely on in-place champions that ARID has already partnered with.  
15 While the ARID team can define research goals and field site locations, they should emphasize  
16 coherence with the in-place research activities. For each international location, a committee of  
17 international researchers should be defined that will be assisting or directly a part of the ARID  
18 campaign.

19

## 20 D.2 Code of Conduct

21 We envision ARID as a community founded upon our shared interest in dryland science, and we  
22 are open to everyone from anywhere in the world. ARID is fully committed to providing a safe  
23 and inclusive experience for all those involved. Together, we can make an ARID community of  
24 which we are all proud to be a part. The following Code of Conduct was developed during  
25 scoping and posted to our ARID scoping website upon its establishment early in scoping on  
26 November 2023 (<https://aridscoping.arizona.edu>). It will need to be adapted in real time as the  
27 ARID field research and collaborations develop.

28 Participants in our programs, field work, research, and events agree to the below Code Of  
29 Conduct:

- 30
- 31 ● Follow the rules. Please respect the rules and policies of meeting venues, hotels, online  
32 platforms, or any other venue. During online events, we will create structure to ensure  
33 respect for our facilitators, presenters, and attendees. For example, you may be asked  
34 to hold questions until the end or keep your microphones and phones on silent.
  - 35 ● Be respectful. Use good practices for intercultural collaborations. Disagree with ideas  
36 openly but respectfully, without demeaning or embarrassing others or calling individuals  
37 out (i.e., criticize ideas, not people). Be open to challenges to your own ideas, and don't  
38 dismiss the ideas of others. Acknowledge other's identities by using their correct  
39 pronouns.
  - 40 ● Be kind. Assume everyone is bringing their best self and treat everyone with dignity. Be  
41 cautious of using humor or sarcasm, especially in online communication. Avoid jokes  
that demean others or make fun of people, even if indirectly.

- 1 ● Be an ally. Speak up if you see or hear something that might be inconsistent with the  
2 ARID Code of Conduct. You are empowered to make others aware of their actions if you  
3 feel they are in conflict with expected, collaborative, respectful behavior.
- 4 ● Be accountable. When we as organizers or participants fail to meet these guidelines, we  
5 will work together to identify problems and adjust policy and practice together.

6 We do not tolerate harassment of program or event participants in any form, including:

- 7 ● Physical, verbal, written, or other forms of abuse of any attendee, speaker, volunteer,  
8 exhibitor, staff member, service provider, or other guest.
- 9 ● Examples of abuse include, but are not limited to, verbal comments related to gender,  
10 sexual orientation, disability, physical appearance, body size, race, religion, national  
11 origin, or socioeconomic class; inappropriate use of nudity and/or sexual images in  
12 public spaces or in presentations.
- 13 ● Sustained disruption of talks or other events. This includes interrupting speakers or other  
14 guests, dismissing others' ideas, or dominating the time in a manner that overpowers  
15 other voices.
- 16 ● Personal attacks, intimidation, stalking, or unwelcome following, whether in person or  
17 online.

18 We understand that some of these behaviors might be done unintentionally, so we ask that if  
19 you are made aware that your words or actions are offensive, stop immediately. If you do not  
20 stop, you will need to leave and could be removed from future participation.

21 Note, ARID adapted this Code of Conduct from the existing American Geophysical Union (AGU)  
22 and Ecological Society of America (ESA) Codes of Conduct. We will fall back to these more  
23 thorough documents for any issues or situations that are not explicitly covered here.

24 If you experience or witness any behavior that violates these guidelines or any other behavior  
25 that makes you or someone else uncomfortable, please reach out to the ARID leadership team  
26 or NASA Carbon Cycle & Ecosystems Office.

27

### 28 D.3 Data, Software, and Information Management

29 ARID holds at its core the best practices of Open Science to accelerate data, software, and  
30 knowledge sharing responsibly and ethically. This includes adherence to NASA's SPD-41a  
31 scientific information policy, NASA Earth science data & information policy, and community  
32 guiding principles and best practices. ARID will follow the science community-defined FAIR  
33 principles (Wilkinson et al., 2006) for data and metadata curation and management to improve  
34 the findability, accessibility, interoperability, and reusability of ARID products and CARE  
35 principles (Carroll et al., 2020) for data sovereignty to ensure that ARID products involving  
36 Indigenous Peoples are managed in a way that respects their rights, interests, and values.  
37 Below, we provide several guiding strategies and critical considerations that suggest the best  
38 practices and policies ARID data, software, and information should be built upon and adhered to  
39 during the ARID campaign.

40



- 1
- 2 ● ARID Products: “ARID products” represent any dataset, software, or information  
3 generated to meet ARID goals, such as primary observations, monitoring data, site  
4 characterization information, algorithms, codes, and output, remotely sensed products,  
5 documentation, and ancillary information, supported explicitly by NASA and its partner  
6 institutions. Domestic and international partners will define, share, and agree upon  
7 general policies for data standardization, uncertainty assessment, metadata, code  
8 documentation, data governance, and stewardship.
  - 9 ● Data and Software Sharing: ARID data and software will be publicly available as soon as  
10 possible after validation and quality control. For NASA airborne acquisitions, project  
11 teams will have no period of exclusive use. However, PI-led projects may request an  
12 embargo from the public release of collected data for an extended period to ensure  
13 master students and doctoral candidates can publish their papers and theses before  
14 data is publicly available. All ARID datasets will be archived through the assigned  
15 DAAC(s) or other public repositories (non-NASA data) and openly accessible through  
16 NASA Earthdata, other NASA-relevant data portals, and the ARID website. Data will be  
17 accessible through standard services and Application Program Interfaces (APIs) as  
18 applicable. In addition to working with the assigned DAAC(s) to make data publicly  
19 available, ARID cyberinfrastructure will support the integration of relevant NASA and  
20 non-NASA datasets. It will facilitate ARID multi-source data sharing and processing for  
21 project affiliates (see section D5). ARID will request that PIs share their codes, Jupyter  
22 Notebooks with “vignettes” and examples of using relevant APIs, and Docker containers  
23 on ARID’s GitHub and DockerHub to foster collaborations within the ARID network and  
24 ensure the reproducibility of workflows for the wider community. The ARID website will  
25 centralize the links to relevant data and infrastructure facilitating access to all ARID  
26 products.
  - 27 ● Data and Software Credit: Each published ARID dataset, and software will be assigned  
28 a Digital Object Identifier (DOI) and a formal citation. Following NASA Earth Science  
29 Data and Information System (ESDIS) Project Data Use Policy and community  
30 practices, ARID will ensure that data and software are credited appropriately, either by  
31 co-authorship, citation, or acknowledgment, depending on the participants' involvement  
32 in data collection and software creation in the publication.
  - 33 ● Protecting Students’ Data: ARID recommends that PI-led projects may request an  
34 embargo from the public release of collected data for a defined period to ensure master  
35 students and doctoral candidates can publish their scientific papers and theses before  
36 data is publicly available. Such an embargo is at odds with SPD-41a, and the ability to  
37 allow an embargo and define its length will be done in consultation with NASA program  
38 managers.
  - 39 ● Indigenous Data Sovereignty: ARID recommends the CARE principles and relevant  
40 policies to ensure data collected over tribal lands and Indigenous datasets are  
41 appropriately managed, shared, and used. ARID will include tribal representative  
42  
43  
44

1 collaboration to define data collection and governance. Provenance information,  
2 contextual metadata, protocols, access permissions, and data use and circulation will be  
3 added for these specific datasets. ARID PIs shall work with tribal partners to clearly  
4 define the expected benefits for the communities directly involved or affected by any  
5 ARID products related to tribal lands.  
6

- 7 ● ARID Products Format and Archive Recommendations: ARID will collaborate with the  
8 NASA ESDS program to streamline the DAAC assignment process and ensure that  
9 researchers understand the need to archive their NASA data with the assigned  
10 DAAC(s). ARID will advise that non-NASA data be archived in the PIs institutional  
11 repositories. As early as possible, All ARID project PIs will receive training on optimal  
12 data formatting and archiving practices to ensure the accessibility, reproducibility, and  
13 interoperability of ARID products, regardless of the public repository in which they are  
14 archived.  
15
- 16 ● Standardization and Training: As a critical step, ARID will work closely with the assigned  
17 DAACs and other data experts to apply standard formats and data curation/management  
18 best practices to improve the interoperability and usability of ARID data to better support  
19 meta-analysis and model integration. These standard formats and best practices will  
20 apply to all directed field and airborne datasets, as well as PI-led project-generated  
21 datasets, to improve data formatting and organization, reporting of uncertainties,  
22 provision of metadata, etc. Lessons learned from past campaigns suggest such efforts  
23 must be planned early instead of at the conclusion of the funded projects. Sufficient and  
24 timely training, in close collaboration with assigned DAACs and NASA Open Science  
25 programs, needs to be provided to the ARID-funded PIs and participants to ensure the  
26 established standards and practices are appropriately followed. As of scoping, we  
27 envision ARID datasets will be stored in formats such as Climate & Forecast (CF)  
28 compliant NetCDF and Cloud-optimized GeoTIFF (CoG). Archival at the assigned  
29 DAAC(s) will also enable public search of ARID data through multiple interfaces,  
30 including a SpatioTemporal Asset Catalog (STAC) API, the Earthdata Search, and the  
31 Common Metadata Repository (CMR) API. This will enable the use of ARID data across  
32 a wide range of programming applications and allow it to be processed in a range of  
33 environments from personal computers (within the limits of data volume) to cloud  
34 services. ARID will conform to SPD-41a and follow community practices to ensure  
35 research software developed by ARID is publicly available with a permissive software  
36 license (such as MIT or BSD 3-Clause License) and is citable no later than the research  
37 publication.  
38
- 39 ● Near and Long-term ARID Resources Preservation: ARID will follow NASA's Earth  
40 Science Data Preservation Content Specification (PCS) and the Preservation Content  
41 Implementation Guidance, which complement the above-described policies and  
42 practices, to preserve valuable ARID data, software, and information beyond the life of  
43 the ARID-related project. It will enable concurrent and future projects and users to  
44 understand and utilize ARID products and derive information, knowledge, and policy

1 recommendations. These ARID resources may include instrument and platform  
2 description, site characteristics, calibration data and methods, science data product  
3 validation method and results, product quality, processing and algorithm version history,  
4 science data access, and analysis tools.  
5

## 6 D.4 Computing Resources

7 The ARID Cyberinfrastructure (ARID CI) is committed to leveraging existing NASA  
8 cyberinfrastructure, nurturing its open science mission, and fostering collaboration among  
9 scientists and end-users working on dryland adaptation and response studies. It will seamlessly  
10 integrate NASA and non-NASA data repositories and cloud computing sources, provide a user-  
11 friendly interface, and establish open data governance, licensing, and indexing standards. ARID  
12 CI aims to make ARID data, softwares, and CI resources accessible to everyone, supporting  
13 crucial local and global dryland research. Its cloud environment is a vital component of ARID's  
14 mission. It enables collaborative knowledge production by offering a centralized platform for  
15 NASA and non-NASA data sharing and cloud computing resources for collaboration among  
16 diverse teams of scientists and end-users (Fig. D.1).  
17

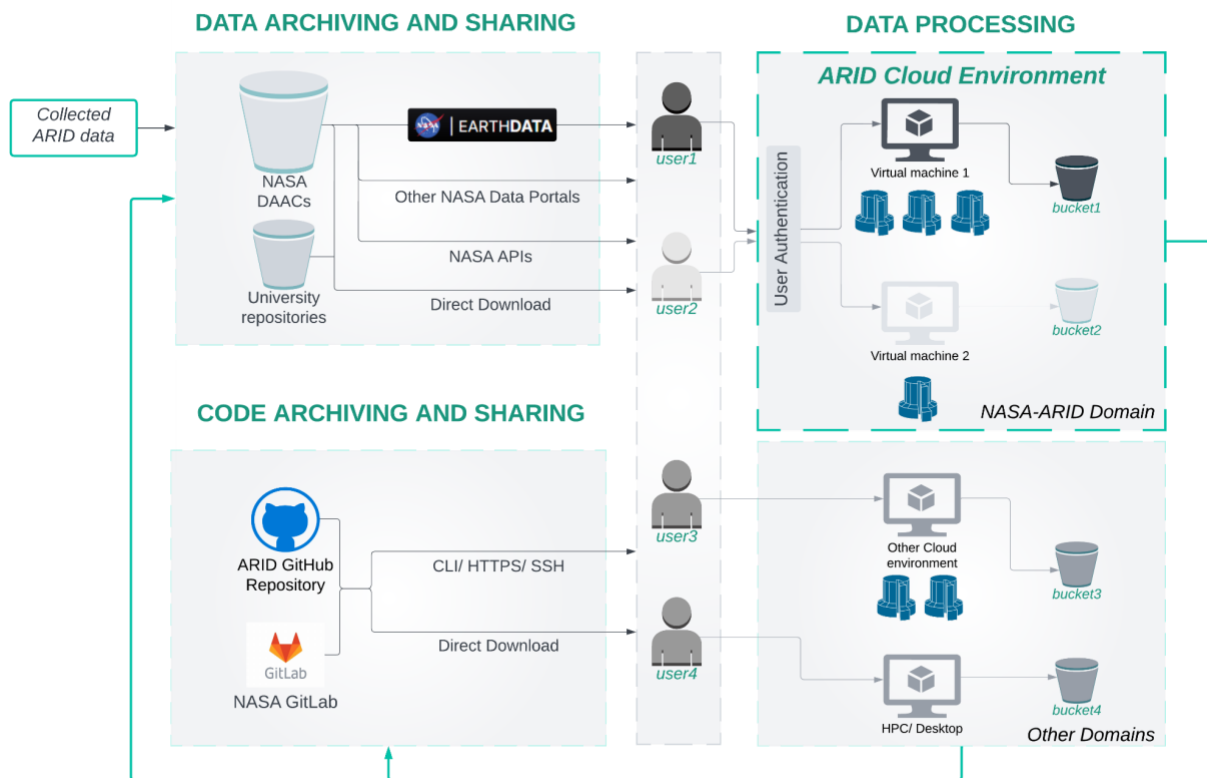
18 The ARID cloud environment interface such as the Science Managed Cloud Environment  
19 (SMCE) will empower scientists from a range of career levels, disciplines, and locations to  
20 efficiently store data and deploy virtual machines for running Jupyter notebooks and Docker  
21 images, thereby enabling the processing and integration of substantial multi-modal datasets. As  
22 for scoping, we envision one environment where participants can run Jupyter notebooks from  
23 ARID collection, which includes Earthdata cloud-based services, such as Harmony, OPeNDAP,  
24 and EGIS, that will facilitate large dataset transformation and harmonization directly from NASA  
25 Earthdata Cloud. PIs-led Jupyter Notebooks and Docker containers will also foster  
26 collaborations within the ARID network and ensure the reproducibility of ARID workflows for the  
27 wider community. This will facilitate scientists and end-users in processing their data within the  
28 ARID cloud environment and leverage other NASA and non-NASA CI tools such as VEDA,  
29 MAAP, Cyverse, and PIs institutional HPCs and local computers.  
30

31 To grant a diverse end-user community access to datasets, existing NASA tools, including CMR  
32 STAC APIs and the APPEARS tool for simplified data access, management, and download, as  
33 well as dashboard applications such as NASA EIS, VEDA, and MAAP, will be accessible to  
34 users on the ARID website. These applications will enable users without expertise in cloud  
35 computing to access, process, and visualize multiple datasets and information resulting from  
36 ARID research. Committed to the CARE principles and respecting data and knowledge  
37 sovereignty, the utilization of ARID CI resources is optional. End-users have the freedom to  
38 download their datasets of interest and process them on local machines, applying their  
39 knowledge without constraints and with individualized security.  
40

41 We further propose ARID invest in capacity building to promote open science and scalable  
42 computing best practices within the ARID community and beyond. We expect that ARID PIs

1 understand and are committed to nurturing NASA Open Science and DEI vision by following  
 2 NASA SPD41a, as well as the community-driven FAIR and CARE principles. ARID may also  
 3 encourage training on the best use of ARID datasets and cyberinfrastructure through  
 4 partnerships with programs such as NASA TOPS and OpenScapes. The ARID webpage will  
 5 provide easy access to data, existing NASA resources, and the ARID cloud environment,  
 6 offering links to all ARID-related resources. Recorded trainings on best practices for open data  
 7 and code management and publication will be accessible to the ARID and broader community.  
 8 Furthermore, ARID CI will be established to streamline data synthesis among ARID-funded  
 9 scientists and end-users and ensure a lasting ARID products legacy for the future generation of  
 10 dryland scientists, managers, and local communities.

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**Figure D.1.** Diagram delineating the ARID cyberinfrastructure data and code storage, sharing, processing, and archiving procedures. The ARID cloud environment will be developed to facilitate data synthesis for recipients of NASA's ARID awards. The workflow adheres to the FAIR principles, ensuring the findability, accessibility, interoperability, and reproducibility of ARID data and codes. At the same, it gives freedom to end-users to download and apply their knowledge to ARID datasets on local machines. ARID awardees and the wider community will be granted access to training sessions covering best practices for open data and code management and publication. The ARID webpage will provide easy access to data, existing NASA resources, and the ARID cloud environment, offering links to all ARID-related CI resources.

# 1 Acronym List

Acronym	Definition
ABoVE	NASA Arctic-Boreal Vulnerability Experiment
ACT-America	Atmospheric Carbon and Transport - America
AEOIP	Applied Earth Observations Innovation Partnership
AGB	Aboveground Biomass
AGU	American Geophysical Union
AI	Artificial intelligence
AIM	Bureau of Land Management Assessment, Inventory and Monitoring
AIRS	Atmospheric Infrared Sounder
AISES	American Indian Science and Engineering Society
API	Application Programming Interface
APPEARS	NASA Application for Extracting and Exploring Analysis Ready Samples
ARC	Australian Research Council
ARID	Adaptation and Response in Drylands (scoping study for NASA)
ARID CI	ARID Cyberinfrastructure
ARM	Atmospheric Radiation Measurement
ARSET	Applied Remote Sensing Training Program
ASD	Analytical Spectral Device
ASU	Arizona State University
AVIRIS	Airborne Visible/Infrared Imaging Spectrometer
AWS	NASA Amazon Web Service
AZ	Arizona
BHP	BHP Copper Co
BIL	Bipartisan Infrastructure Law
BLM	Bureau of Land Management
BOREAS	Boreal Ecosystem-Atmosphere Study (NASA-terrestrial field campaign)
CA	California
CAM	Crassulacean Acid Metabolism
CARAFE	Carbon Airborne Flux Experiment
CARE	Collective Benefit, Authority to Control, Responsibility, Ethics
CARVE	NASA Carbon in Arctic Reservoirs Vulnerability Experiment
CASC	USGS Climate Adaptation Science Centers
CCE	NASA Carbon Cycle and Ecosystems
CF	Climate and Forecast
CFIS	Chlorophyll Fluorescence Imaging Spectrometer
CHIME	Copernicus Hyperspectral Imaging Mission for the Environment
CMR	NASA EarthData common metadata repository
CMS	NASA Carbon Monitoring System
CO	Colorado
COG	Cloud Optimized GeoTIFF

ONAHCYT	Mexican National Council of Humanities, Science and Technology
CREW	Climate Response Early Warning
	Commonwealth Science and Industrial Research Organization, Earth
CSIRO	Observation Center
CSU	Colorado State University
DC	District of Columbia
DGVM	Dynamic Global Vegetation Model
DOD	Department of Defense
DOE	Department of Energy
DOI	Department of Interior
DryFlor	Latin American Seasonally Dry Tropical Forest Floristic Network
ECOSTRESS	ECOsysteM Spaceborne Thermal Radiometer Experiment on Space Station
EDGE	Earth Dynamics Geodetic Explorer
EFTEON	Expanded Freshwater and Terrestrial Environmental Observation Network
EIS	Earth Information System
EMIT	Earth Surface Mineral Dust Investigation
ENSO	El Niño-Southern Oscillation
EO	Earth Observation technologies
EOS	Earth Observing System
ESA	European Space Agency and Ecological Society of America
ESD	NASA's Earth Science Division
ESIL	Environmental Data Science Innovation & Inclusion Lab
ESM	Earth System Models
ESTO	NASA Earth Science Technology Office
ET	Evapotranspiration
FAIR	Findability, Accessibility, Interoperability, and Reusability
FAPAR	Fraction of photosynthetically active radiation absorbed
FEWSNET	USGS Famine Early Warning Systems Network
	First International Satellite Land Surface Climatology Project Field
FIFE	Experiment (NASA- terrestrial field campaign)
FIREFLY	Fluorescence Imaging of Red and Far-Red Light Yield
FLAME VNIR	Flame (Near-Infrared) spectrometer from Ocean Optics
FLEX	NASA Flame Extinguishment Experiment
GEDI	Global Ecosystem Dynamics Investigation
GEMX	NASA-USGS Geological Earth Mapping Experiment
GERI	Global Ecosystem Research Infrastructure
GEWEX	NASA's Global Energy and Water Cycle Experiment
GIS	Geographic Information System
	Global Land-Atmosphere Coupling Experiment–Coupled Model
GLACE-CMIP5	Intercomparison Project phase 5
GLIHT	NASA Goddard's LiDAR, Hyperspectral & Thermal Imager
GLOBE	NSF Global Learning and Observations to Benefit the Environment
GNSS	Global Navigation Satellite System

GOES	Geostationary Operational Environmental Satellites
GOSAT	Greenhouse gases observing satellite
GPM	Global Precipitation Measurement
GPP	Gross Primary Production
GPR	Ground Penetrating Radar
GRACE	NASA Gravity Recovery and Climate Experiment
GRACE-FO	Gravity Recovery and Climate Experiment Follow-On mission
GSD	Global Spectral Deconvolution
GSFC	NASA Goddard Space Flight Center
HIS	Hyperspectral Image Spectroscopy
HISUI	Hyperspectral Imager Suite
HLS	Harmonized Landsat and Sentinel-2
HQ	Head Quarters
HSI	Hyperspectral Imaging
HYR-SENSE	Hyperspectral and Thermal Remote Sensing for Environmental Justice
HYSPIRI	Hyperspectral Infrared Imager
HYTES	Hyperspectral Thermal Emission Spectrometer
IAV	Interannual Variability
ICOS	Integrated Carbon Observation System
IN	Indiana
INIFAP	Institute for Forestry, Agriculture and Livestock Research
IPBES	Intergovernmental Science-Policy Platform on Biodiversity and Ecosystem Services
IPCC	Intergovernmental Panel on Climate Change
IPI	NASA's Indigenous Peoples Initiative
IRA	Inflation Reduction Act
IRT	Infrared Thermometer
ISS	International Space Station
ITSON	Instituto Tecnologica de Sonora
JAXA	Japan Aerospace Exploration Agency
JPL	Jet Propulsion Laboratory
LAI	Leaf Area Index
LandSat	Land Satellite
LASI	Latin America Sustainability Institute
LBA	Large-Scale Biosphere-Atmosphere (NSAS-terrestrial field campaign)
LCLUC	NASA Land-Cover and Land-Use Change
LDC	Landscape Data Commons
LIDAR	Light Detection and Ranging
LPDAAC	NASA Land Processes Distributed Active Archive Center
LST	Land Surface Temperature
LSTM	NASA Land Surface Temperature Monitoring
LTAR	Long-Term Agroecosystem Research
LTER	Long Term Ecological Research

LVIS	Land, Vegetation, and Ice Sensor
MA	Massachusetts
MAAP	NASA Multi-Mission Algorithm and Analysis Platform
MAIANSE	NASA MUREP for American Indian and Alaskan Native STEM Engagement
MASTER	Spaceborne Thermal Emission and Reflection Radiometer (ASTER)
MC	NASA Mission Concepts
MD	Maryland
MERLIN	NASA The MISR Enhanced Research and Lookup Interface
MODIS	Moderate Resolution Imaging Spectroradiometer
MRI	USGS Earth Mapping Resources Initiative
MSG	Microgravity Science Glovebox
MUREP	NASA Minority University Research & Education Project
NASA	National Aeronautics and Space Administration
NASEM	National Academies of Sciences, Engineering, and Medicine
NC CASC	North Central Climate Adaptation Science Center
NCA5	United States National Climate Assessment
NDVI	Normalized Difference Vegetation Index
NEE	Net Ecosystem Exchange
NEESPI	Northern Eurasia Earth Science Partnership Initiative (NASA research program)
NEON	National Ecological Observatory Network
NEON-AOP	NEON Airborne Observation Platform
NG	Airborne Visible InfraRed Imaging Spectrometer - Next Generation
NGEE	Next-Generation Ecosystem Experiments
NIR	Near-Infrared
NISAR	NASA-ISRO Synthetic Aperture Radar
NM	New Mexico
NMSU	New Mexico State University
NOAA	National Oceanic and Atmospheric Administration
NOHRSC	National Operational Hydrologic Remote Sensing Center
NPP	Net Primary Production
NPS- I&M	National Park Service Inventory and Monitoring
NRCS	Natural Resources Conservation Service
NRCS-NR	Natural Resources Conservation Service National Resources Inventory
NSF	National Science Foundation
OCO	NASA Orbiting Carbon Observatory
OCO-2 v10	Orbiting Carbon Observatory-2 (OCO-2) Model Intercomparison Project
MIP	(MIP)
ORNL DAAC	Oak Ridge National Laboratory Distributed Active Archive Center
OSSE	Observing System Simulation Experiments
PACE	Plankton, Aerosol, Cloud, ocean Ecosystem
PALS	Passive Active L-band System
PBL	Planetary Boundary Layer



PERSIANN-CDR	Precipitation Estimation from Remotely Sensed Information using Artificial Neural Networks–Climate Data Record
PFT	Plant Functional Types
PI	Principal Investigator
PICARD	Pushbroom Imager for Cloud and Aerosol Research and Development
PRI	Photochemical Reflectance Index
RCMAP	Rangeland Condition Monitoring Assessment and Projection,
REU	Research Experiences for Undergraduates
RFP	NASA Requests for Proposals
RGB	Red Green Blue
RIZA	International Network for Drylands Sustainability
RS	Remote Sensing
RUBISCO	Ribulose-1,5-bisphosphate carboxylase/oxygenase
RZSM	Root Zone Soil Moisture
SAEON	South African Environmental Observation Network
SAFARI	Southern African Regional Science Initiative
	Southern African Science Service Centre for Climate Change and Adaptive
SASSCAL	Land Management
SBG	Surface Biology and Geology
SDC	NASA Surface Deformation and Change
SDG	Sustainable Development Goals
SECO	Bosque Seco Biosphere Reserve
SEEDS	Strategies for Ecology Education, Diversity and Sustainability
SEPO	Social-Ecological Participatory Observatories
SERDP	Strategic Environmental Research and Development Program
	Regional Visualization and Monitoring System (SERVIR's name is derived
SERVIR	from the Spanish word meaning “to serve”)
SES	Social Ecology Systems
SEVIRI	Spinning Enhanced Visible and Infrared Imager
SFM	Structure From Motion
SHIFT	High-Frequency Time Series (NASA research Program)
SIC	Soil Inorganic Carbon
SIF	Solar-Induced Fluorescence
SIPI	Southwestern Indian Polytechnic Institute
SLAP	Scanning L-band Active/Passive
SM	Soil Moisture
SMAP	Soil Moisture Active Passive
SMAPVEX15	Soil Moisture Active Passive Validation Experiment 15
SMCE	Science Managed Cloud Environment
SMOS	Soil Moisture and Ocean Salinity
SNOTEL	Natural Resources Conservation Service "Snow Telemetry"
SNWG	NASA’s Satellite Needs Working Group
SOILWAT	water balance simulation model

SPAC	Soil Plant Atmosphere Water Continuum
SPD-41a	Scientific Information Policy for the Science Mission Directorate
SPECNET	Spectral Network
SSAI	Science Systems and Applications, Inc
STAC	SpatioTemporal Asset Catalog
STELLA	NASA Science and Technology Education for Land/Life Assessment
STV	Surface Topography and Vegetation
SWE	Snow Water Equivalent
TEK	Traditional Ecological Knowledge
TEMPO	Tropospheric Emissions: Monitoring of Pollution
TERN	Terrestrial Ecosystem Research Network
TIR	Thermal Infrared Emittance
TLS	Terrestrial Laser Scanning
TOPS	Transform to Open Science
TRISHNA	Thermal Infra-Red Imaging Satellite for High-resolution Natural Resource Assessment
TRUST	Transparency, Responsibility, User focus, Sustainability, and Technology
TWS	Terrestrial Water Storage
UA	University of Arizona
UAS	Unmanned Aircraft System
UAS-LS	Uncrewed Airborne System - Laser Scanner
UAV	Unmanned Aerial Vehicle
UAVSAR	Uninhabited Aerial Vehicle Synthetic Aperture Radar
UDSA	United States Department of Agriculture
UK	United Kingdom
ULS	Unmanned Laser Scanning
UN	United Nations
UNAM	Universidad Nacional Autonoma de Mexico
UNCCD	United Nations Convention to Combat Desertification
UNM	University of New Mexico
US	United States
USA	United States of America
USAID	United States Agency for International Development
USDA	United States Department of Agriculture
USDA/ARS	USDA Agricultural Research Service
USFS	United States Forest Service
USFS FIA	Forest Inventory and Analysis
USFWS	United States Fish and Wildlife Service
USGCRP	United States Global Change Strategic Plan
USGEO	United States Group on Earth Observations
USGS	United States Geological Survey
VA	Virginia
VEDA	NASA Visualization, Exploration, and Data Analysis

VHR	Very High Resolution Satellite Imagery
VIIRS	Visible Infrared Imaging Radiometer Suite
VOD	Vegetation Optical Depth
VPD	Vapor Pressure-Deficit
VSWIR	HyspIRI Visible to Short Wavelength InfraRed
VWC	Vegetation Water Content
WDTS	Western Diversity Time Series
WGEW	Walnut Gulch Experimental Watershed
WPE	Woody Plant Encroachment
WTDS	Wisconsin TeraGrid Data System
WUE	Water Use Efficiency
WWAO	Western Water Applications Office

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## F. Appendix

We have redacted most Appendix sections given personally identifying information. This information is not publicly available and will only be available for internal review.

### F.1 Science and Application Traceability Matrix

**Table F.1.** Modified Science and Application Traceability Matrix (SATM, Appendix F.1) Physical parameters required to address science sub-themes (numbered in footnotes) and the various remote sensing sensors available on proximal (e.g. towers), UAS, airborne and spaceborne platforms. Names of sensors are given in the footnotes of this table. The table lists the current and future sensors that can make the most significant advances and do not attempt to list all available options. Science Sub-themes: 1.1 Water availability, 1.2 Dryland climate variability: Pulses and Drought, 1.3 Fire, 1.4 Land-Atmosphere interaction, 2.1 Vegetation Structure, 2.2 Dryland Biodiversity, 2.3 Ecosystem Function, 2.4 Dryland Geology and Soil Processes, 3.1 Carbon stocks and fluxes, 4.1 Land Management, 4.2 Adaptation and Mitigation.

Physical Parameter or process / Observable	Science theme number	Comment: need and current status	Near surface and Proximal	UAS	Airborne	Spaceborne/ Satellite
Precipitation (P)	1.1, 1.2					GPM
Soil Moisture (SM)	1.1, 1.2, 2.1, 2.2	Poorly estimated for the subsurface/root zone; identified need for higher resolution (10 m to <1 km) estimates Driver for local models Validation/diagnostic for water cycle in LSMs/ESMs	Soil moisture network, GNSS receiver above and below canopy	Thermal	UAVSAR / AirMoss, SMAPEX; NOAA airborne gamma NOHRSC, AirSWOT,	SMAP, ECOSTRESS, Hydrosat, SMOS, NISAR, Sentinel1,
Evapotranspiration (ET) Level 4 model output	1.1, 1.2, 1.3, 2.1	Calibration/Validation for model estimates of ET, E and T. Open ET models can use multiple RS inputs	Infrared Thermometer (IRT): Apogee SI-111-SS Thermal Camera: FLIR A700f; ICI P-Series	Thermal	HYTES, MASTER	ECOSTRESS, Landsat,
Surface temperature (ST) of soil vs. plants	1.4	Input to ET, GPP, NEE, plant stress	Infrared Thermometer (IRT): Apogee SI-111-SS Thermal Camera: FLIR A700f; ICI P-Series	Thermal	HYTES, MASTER	ECOSTRESS, Landsat,
Vegetation fractional cover of Plant	1.3, 2.1,	PFT: Trees, shrubs, grasses, forbs,	PhenoCam; Terrestrial Laser	Multi-spectral Lidar	AVIRIS-NG, NEON-AOP, G-	Landsat, SBG, EMIT

functional types (PFT)	3.1	biocrust. Grass biomass poorly estimated. Local calibration/ initialization for models DGVM constraint Trait information	Scanning (TLS)		LiHT, LVIS	
Vegetation height: Grass, shrubs, trees	1.3, 2.1, 3.1	Shrub (<3m) biomass poorly estimated with optical and space-based LiDAR algorithms. Individual tree heights can be extracted using SFM from airborne imagery. Potential constraint for data assimilation	TLS	RGB SfM LiDAR	Discrete Lidar, G-LiHT LVIS, UAVSAR, AVIRIS-NG, NEON-AOP	NISAR, S1, GEDI, IceSat2, WorldView,
Above ground biomass: grass	1.3, 2.1, 3.1	Grass biomass poorly estimated with optical	TLS	Multi-spectral Lidar	AVIRIS-NG, G-LiHT, NEON-AOP	EMIT, LandSat NEXT, HLS, SBG,
Above ground biomass: shrubs	1.3, 2.1, 3.1	Shrub biomass is poorly estimated with optical and space-based LiDAR algorithms. Calibration/Validation for model estimates of NPP Potential constraint for data assimilation	TLS	LiDAR	Discrete Lidar, G-LiHT LVIS, UAVSAR, AVIRIS-NG, NEON-AOP	NISAR, S1, GEDI, IceSat2, EDGE, STV
Above ground biomass: trees	1.3, 2.1, 3.1	Lidar is most accurate. SAR-based estimates of AGB with models trained with airborne Lidar data. Individual tree biomass can be estimated with VHR commercial imagery and AI. Calibration/Validation for model estimates of NPP. Potential constraint for data assimilation	TLS	RGB SfM LiDAR	Discrete Lidar, G-LiHT LVIS, UAVSAR, AVIRIS-NG, NEON-AOP	NISAR, S1, GEDI, IceSat2, EDGE, STV, WorldView,
Below ground biomass	3.1	Very poorly estimated, require ground penetrating radar (GPR)	Ground Penetrating Radar (GPR)			
Photosynthetically active vegetation fractions of PFT's	1.3, 3.1	Essential to make a distinction between active and non-active components	Phenocams, spectrometer (e.g., Ocean Insight FLAME VNIR spectroradiometer)	Hyperspectral, Multi-spectral	AVIRIS-NG, LVIS	EnMap, SBG, Landsat
Coarse and fine fuel loads	1.3	Derived in similar way to PFT biomass	TLS; GNSS receiver	Hyperspectral, Multi-spectral,	AVIRIS-NG, LVIS	EnMap, SBG,

				LiDAR		
Sub-canopy fuel loads	1.3	Currently poorly estimated. Requires LiDAR for 3D structure estimates	TLS; GNSS receiver	LiDAR	Discrete Lidar, G-LiHT LVIS, AVIRIS-NG, NEON-AOP	GEDI, IceSat2, EDGE
Vegetation moisture content	1.3		TLS; dual GNSS receivers installed above and below the canopy	Hyperspectral, Multi-spectral, SAR	AVIRIS-NG, UAVSAR	S1, NISAR, EnMap, SBG, Landsat 8,9,10, NEXT, HLS, PACE
Vegetation greenness / curing	1.3	Requires long time series of greenness	Phenocams; spectrometer (e.g., Ocean Insight FLAME VNIR spectroradiometer)	Hyperspectral, Multi-spectral	AVIRIS-NG, LVIS	EnMap, SBG, Landsat 8,9,10, NEXT, HLS
Vegetation Chlorophyll content	2.1		Phenocams; spectrometer (e.g., Ocean Insight FLAME VNIR spectroradiometer)	Hyperspectral, Multi-spectral	AVIRIS-NG	EnMap, SBG, Landsat 8,9,10, NEXT, HLS
Vegetation Nitrogen content	2.1	Poorly estimated in discontinuous canopies	Hyperspectral: ASD FieldSpec; Spectra Evolution RS-3500, 8800, PSR+	Hyperspectral	AVIRIS-NG, SHIFT	EnMap, SBG
Invasive species: tree, grass, shrub, forbs	2.1	Individual Invasive plants species could be spectrally distinguished	Hyperspectral: ASD FieldSpec; Spectra Evolution RS-3500, 8800, PSR+	Hyperspectral	AVIRIS-NG	EnMap, SBG,
Solar-induced Fluorescence (SIF)	1.2, 3.1		Ultra-hyperspectral VNIR: PhotoSpec; FluoroSpec; FloX Box	Ultra-hyperspectral VNIR	FIREFLY, CFIS	OCO-2,3, TEMPO
Fraction of photosynthetically active radiation absorbed (FAPAR) per PFT	3.1	Input to carbon-flux models	Hyperspectral, Multi-spectral	Hyperspectral, Multi-spectral	AVIRIS-NG	EnMap, SBG, GOES-R
Gross Primary Production: Trees, grass, shrubs. Level 4 model output	1.2, 1.4, 3.1	Modeled variable, require FAPAR of PFT and meteorological variables as input	Hyperspectral; Thermal; SIF	FAPAR of PFTs derived above	FAPAR of PFTs derived above	FAPAR of PFTs derived above, GOES-R
Net Primary Production (NPP). Level 4 model output	3.1	Modeled variable, require FAPAR, GPP and estimated of respiration	Hyperspectral; Thermal; SIF			
Plant Stress Level 4 model output		Requires inputs: ST, Leaf moisture content	Hyperspectral; Thermal; SIF			
Snow Water			TLS; GNSS receiver		NOAA airborne	

Equivalent (SWE)					gamma NOHRSC	
Soil Carbon Organic, Inorganic??)	2.1, 2.2, 3.1		Hyperspectral			
Biocrust fractional cover	2.2	Validation for C stocks in ESM model spin-up	Hyperspectral	Hyperspectral, Multi-spectral	AVIRIS-NG	EnMap, SBG,
Biocrust community composition	2.2		Hyperspectral	Hyperspectral, Multi-spectral	AVIRIS-NG	EnMap, SBG,
Atmospheric Carbon concentrations and flux (Level 4 model output)		Calibration/Validation for model estimates of GPP Potential constraint for data assimilation	Flux tower		CARVE, ATom, ACT-America	OCO-2/-3
Geology/ geochemistry		Geologic constraints on nutrient and water availability; geologic materials affecting environment and human health; acid rock drainage; dust source mineral composition and quantification	Hyperspectral	VSWIR hyperspectral; Infrared multispectral and hyperspectral (MWIR 3-5 microns and TIR 8-12 microns)	AVIRIS-NG, -3, -5; MASTER; HYTES, HYTES-2	Present: EMIT, ECOSTRESS, other national space agency imaging spectrometers Planned: SBG VSWIR and TIR; Landsat NEXT;

- 1 Abbreviations: SfM - RGB: Structure from motion, S1: Sentinel1, SBG: Surface Biology
- 2 Geology, AirMOSS: Airborne Microwave Observatory of Subcanopy and Subsurface, UAS-LS:
- 3 Uncrewed AirborneSystem - Laser Scanner, FIREFLY: fluorescence imaging of red and far-red
- 4 light yield
- 5
- 6

## 7 F.2 ARID Workshop 1: Tucson, AZ

- 8 [Redacted]
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## 10 F.3 ARID Workshop 2: Albuquerque, NM

- 11 [Redacted]
- 12

## 13 F.4 ARID Science Theme Working Groups Participants

- 14 [Redacted]
- 15

## 16 F.5 ARID Technical Working Group Participants

- 17 [Redacted]

1

2 F.6 ARID Engagement Archive

3 [Redacted]

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5 F.7 Tribal Engagement Archive

6 [Redacted]

7

8 F.8 Acknowledgement of ARID Scoping Participants

9 [Initially Redacted, May Appear in Final Version]

10

11 F.9 Community Feedback: Response to Forms

12 [Redacted]

13

14 F.10 Airborne Remote Sensing Feedback

15 [Redacted]

16

17 F.11 End-User Input

18 [Redacted]

19

20 F.12 Archive of Dryland Efforts that ARID can Leverage

21 [Initially Redacted, May Appear in Final Version]

22