

# Earth Observations to Combat Invasive Aquatic Vegetation



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Delta  
Stewardship  
Council





# The Sacramento – San Joaquin Delta

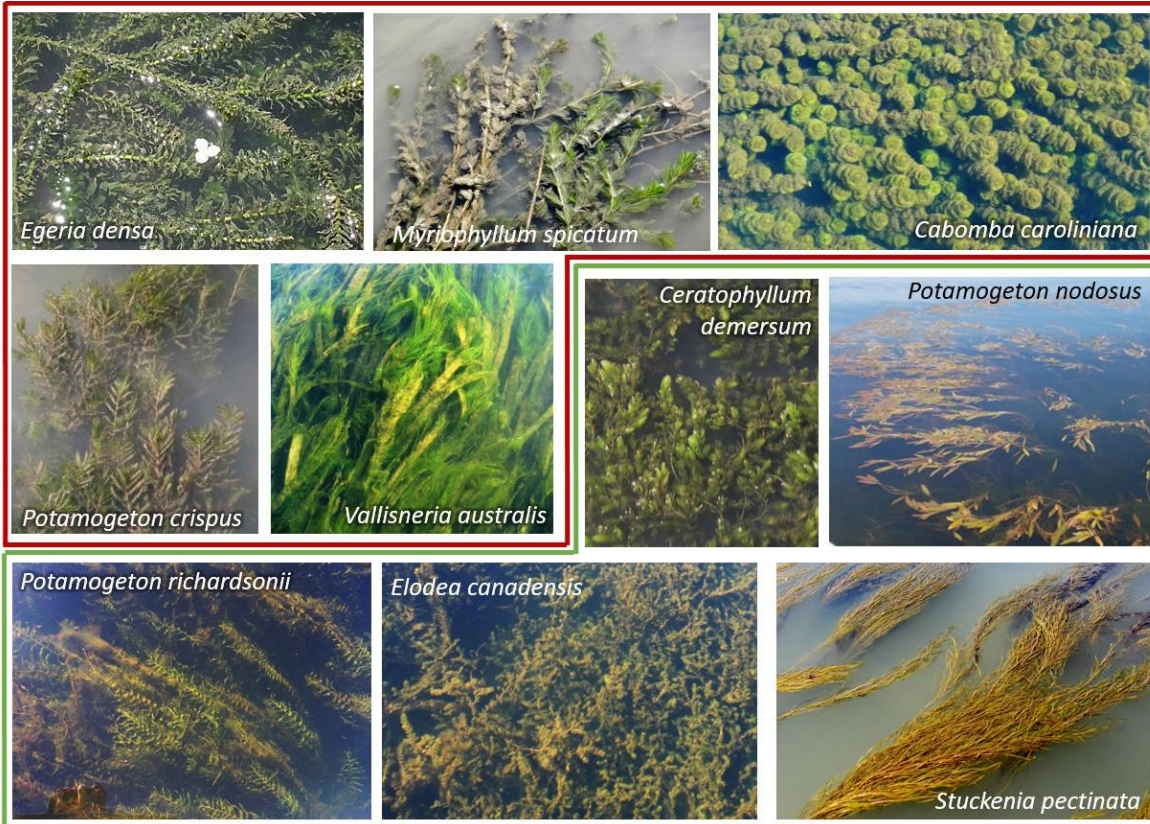
- Water source for 27 million people.
- \$5.3 billion economic output
- One of 25 global hotspots for biodiversity.
- One of the most invaded ecosystems in the world.
- Highly vulnerable to climate variability & change.
- Biodiversity conservation and restoration are required by law.





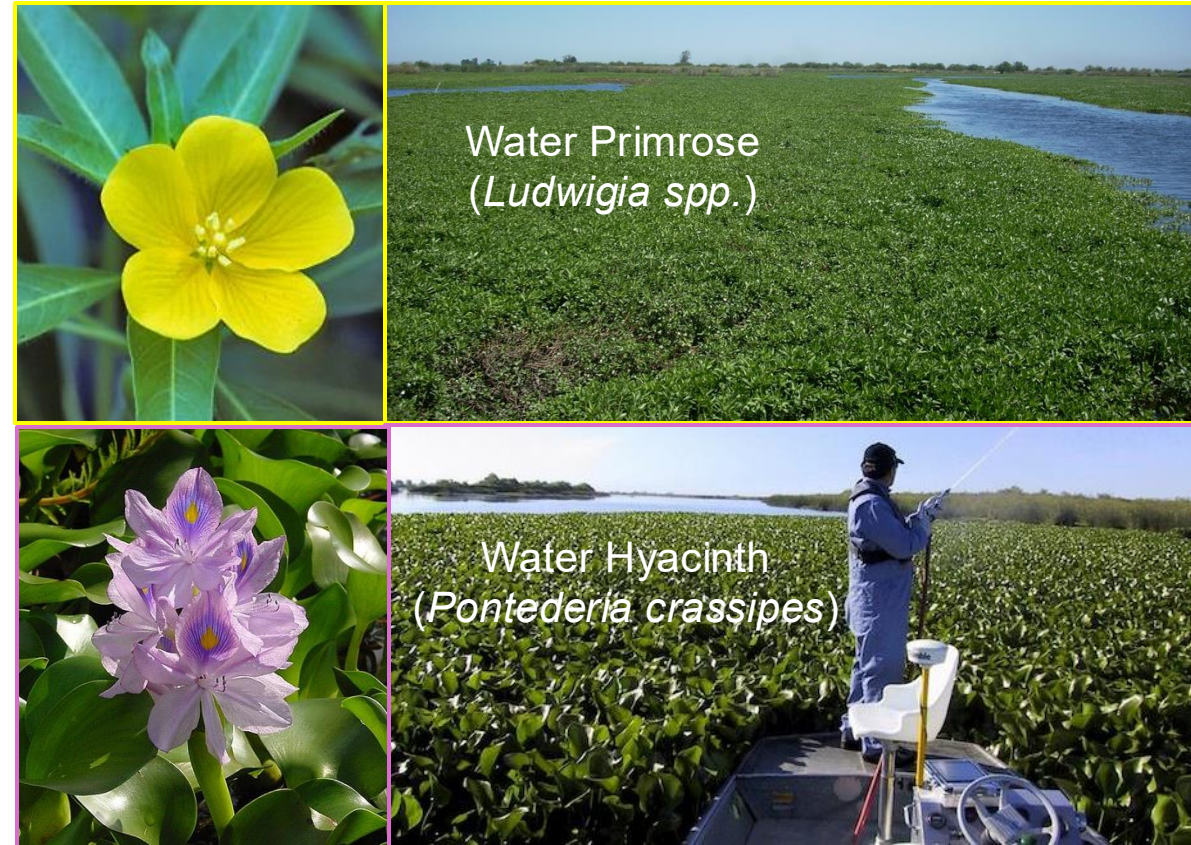
# Invasive aquatic vegetation (IAV) represent a major threat to global biodiversity

## Submerged Aquatic Vegetation (SAV)



NATIVE

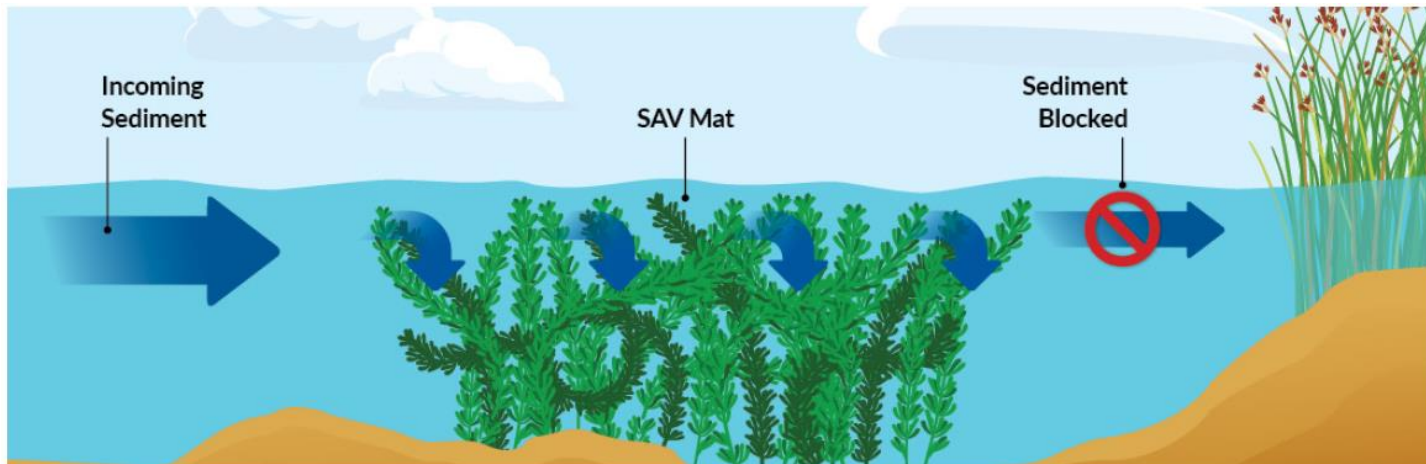
## Floating Aquatic Vegetation (FAV)



INVASIVE

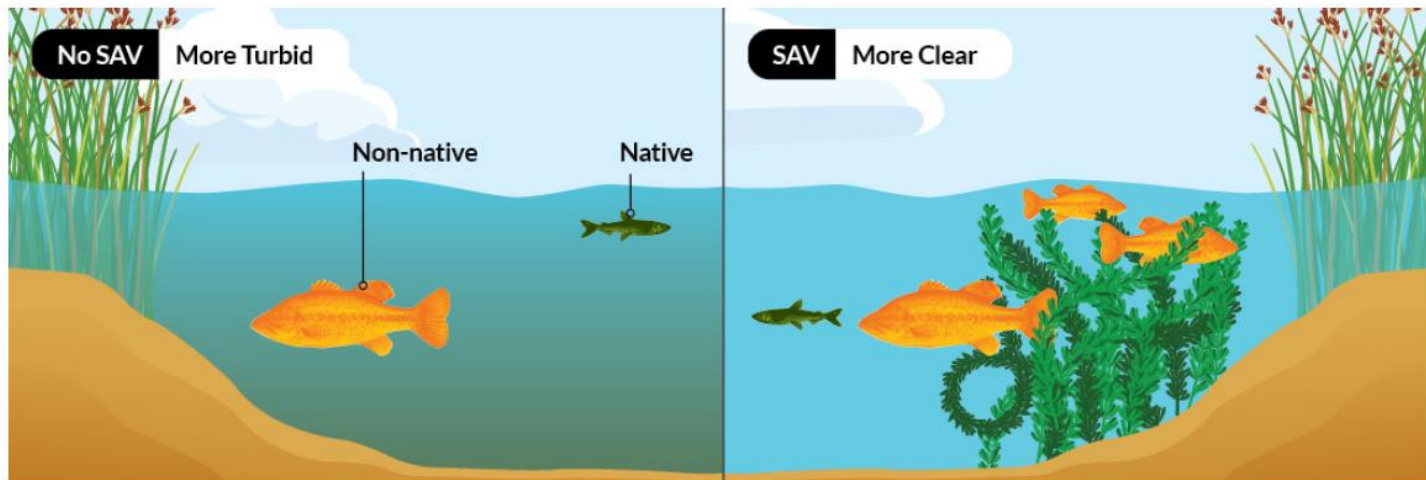


# Invasive aquatic vegetation (IAV) represent a major threat to global biodiversity



## Ecological Impacts:

- Alter environmental conditions
- Reduce native biodiversity
- Replace key habitats
- Increase invader establishment



## Human Impacts:

- Increase agricultural costs
- Cost CA \$14.5 million to manage/year
- Increase flood risk
- Increase disease risk

# Previous Mapping Efforts

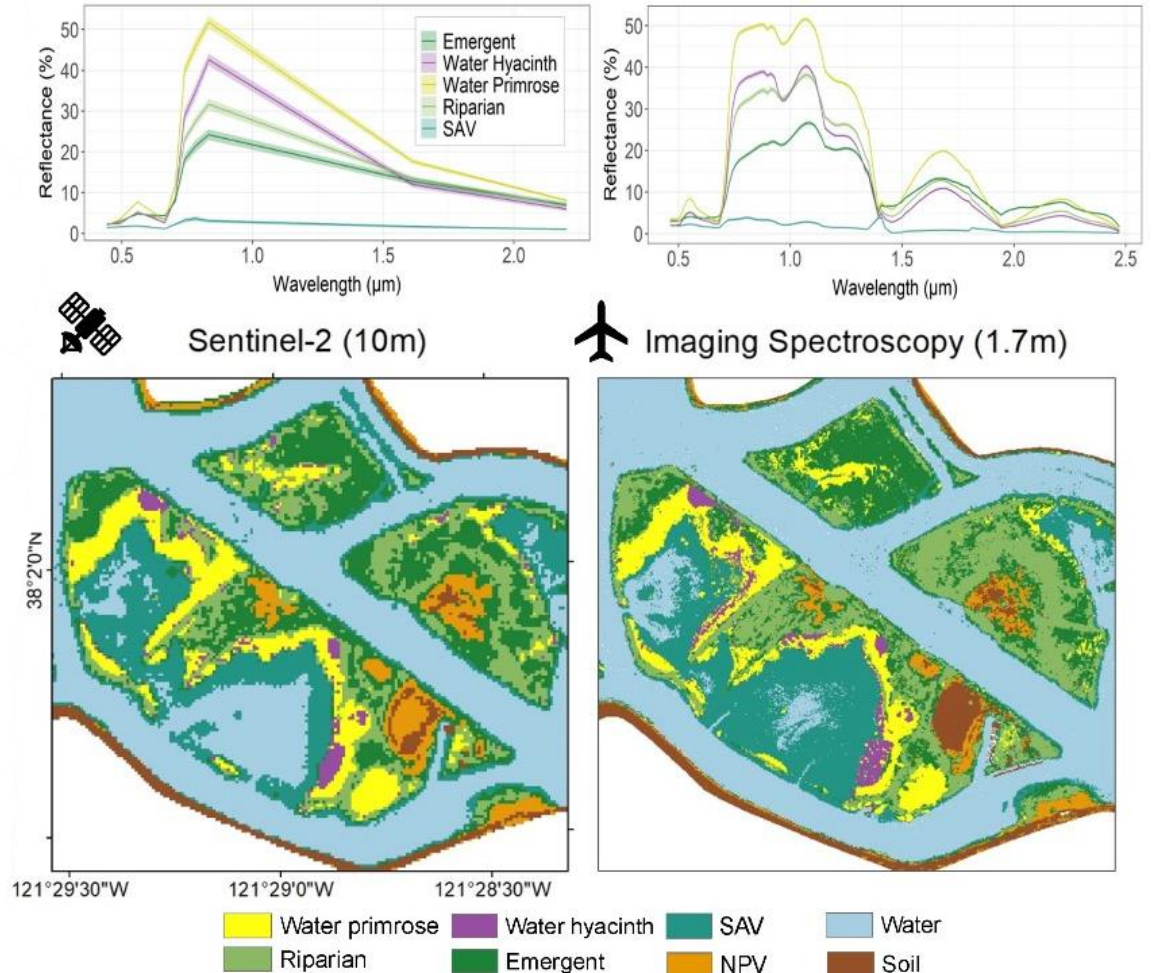
## Long Time-Series Airborne Imaging Spectroscopy (AIS)

- Annual vegetation classification maps of the Delta have been produced from AIS annually since 2004.
- CSTARS (UC Davis)
- <https://doi.org/10.5063/F1HH6HJX>

## IAV Mapping with Satellite Remote Sensing

- Advancement and successful development of IAV classification maps using Sentinel-2 data.
- Ade et al. (2022)

**\*\* Not intended to replace AIS dataset, just fill data gap.**



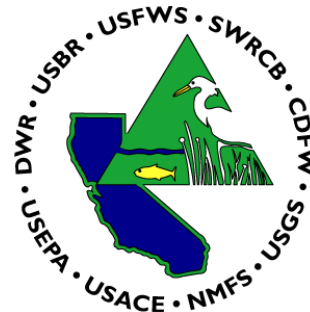
# Earth Observations to Combat IAV

## Project Goals

1. Operationalize the first sustainable mapping effort for vegetation in the Delta.
2. Advance modeling tools to assess IAV responses to past and future management actions.
3. Co-develop IAV mapping tool and integrate into the decision-making frameworks of stakeholders.



EARTH SCIENCE  
APPLIED SCIENCES



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# Goal 1: Operationalize S2 Class Mapper

## Updates:

- Updated/improved cal-val datasets
- Feature Reduction
- Created riparian mask

## Future work:

- Increase the number of in-situ cal-val observations
- Incorporate winter and spring in-situ observations
- Assess model performance against AIS class maps.

Class	Acc. Stat	AIS	S2 2022	S2 2025
	OA	0.92	0.87	0.90
	Kappa	0.90	0.85	0.88
Emergent	PA	0.93	0.85	0.92
	UA	0.88	0.77	0.92
Hyacinth	PA	0.90	0.79	0.87
	UA	0.88	0.84	0.90
Primrose	PA	0.94	0.91	0.94
	UA	0.95	0.83	0.88
SAV	PA	0.83	0.86	0.87
	UA	0.91	0.84	0.90
Water	PA	0.92	0.88	0.90
	UA	0.91	0.90	0.90



# Goal 2: SDMs to Assess IAV Responses to Management Actions

## Species Distribution Model (SDM)

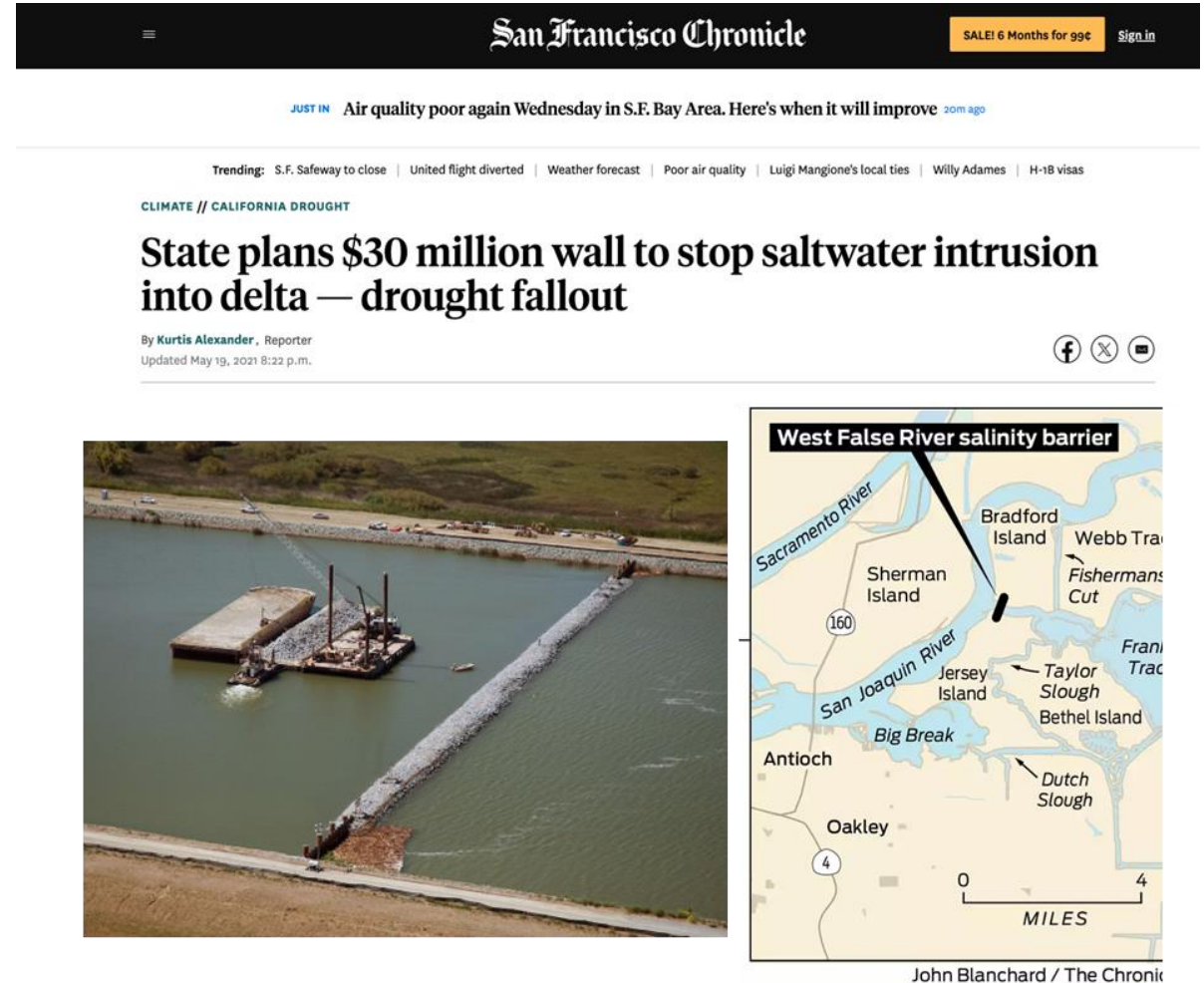
### Predictor Variables

- UnTrim 3D
- Salinity
  - Turbidity
  - Speed
- Bathymetry
- Depth

### Why did we choose these predictors?

- Management decisions affect these variables
- Need to understand their impact on IAV

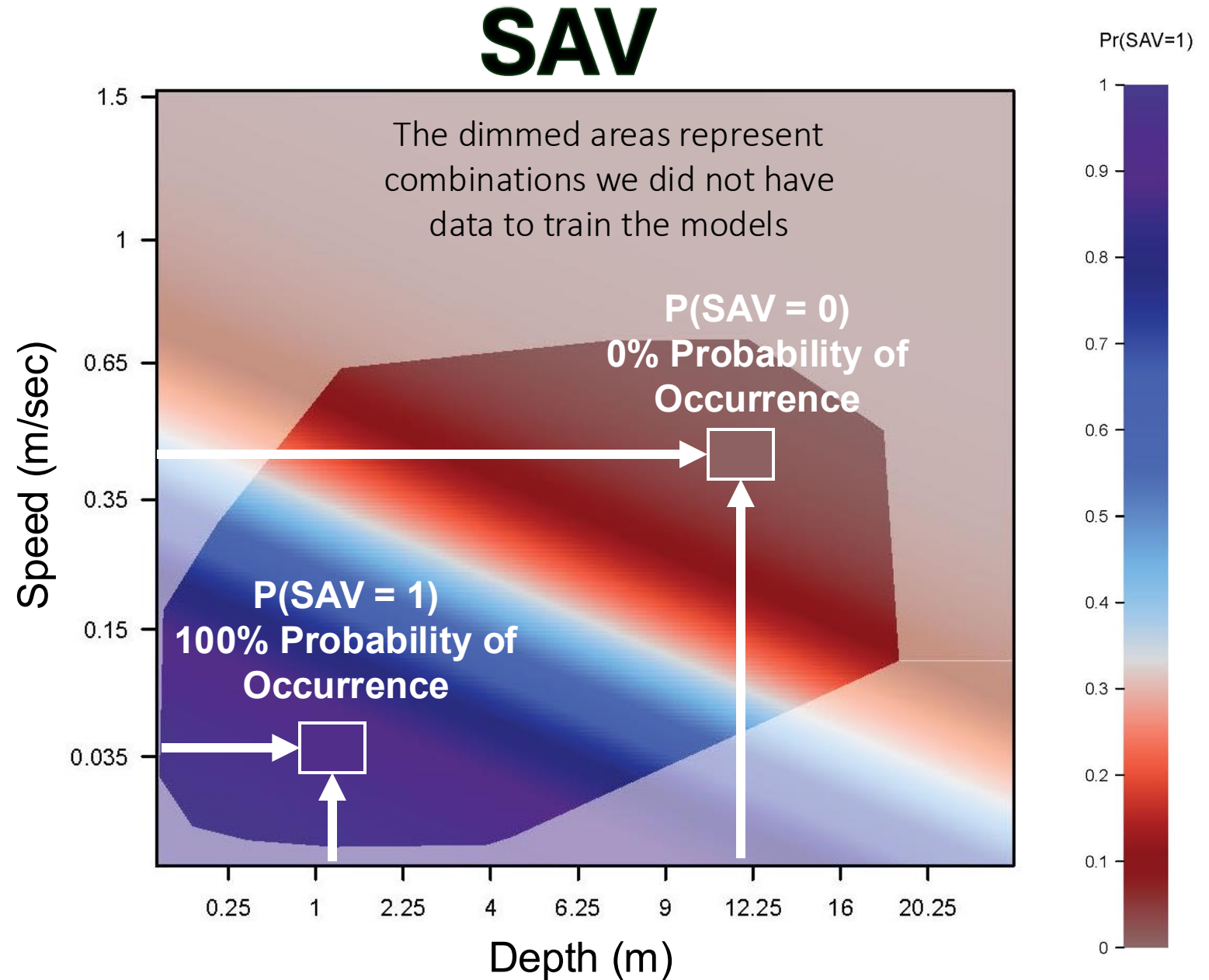
Response Variable = AIS derived IAV class maps





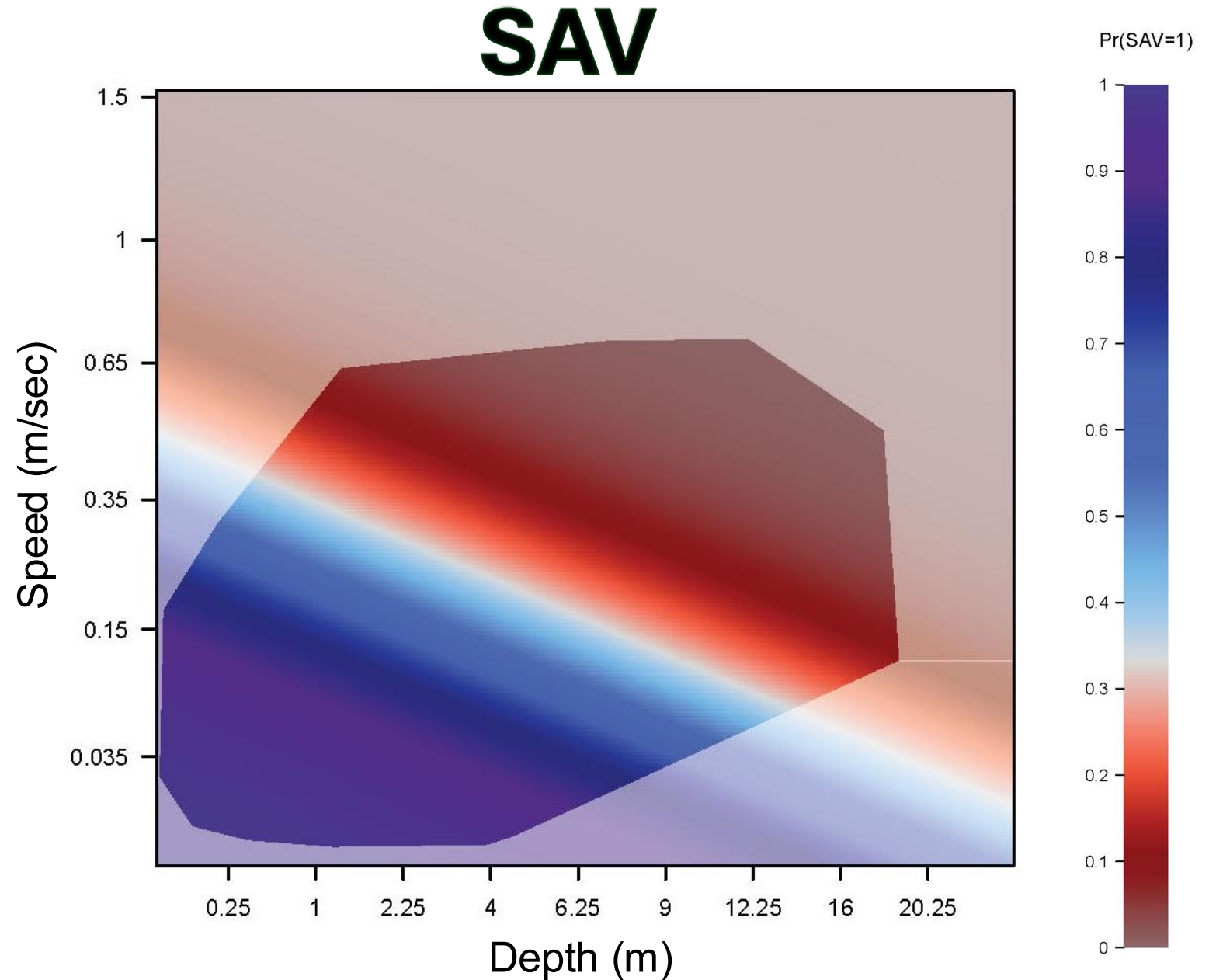
# How to interpret heat maps

Simulate **probability of occurrence** for all possible **environmental combinations**



# How to interpret heat maps

SAV can colonize at greater depths, especially when velocities (speed) are low.



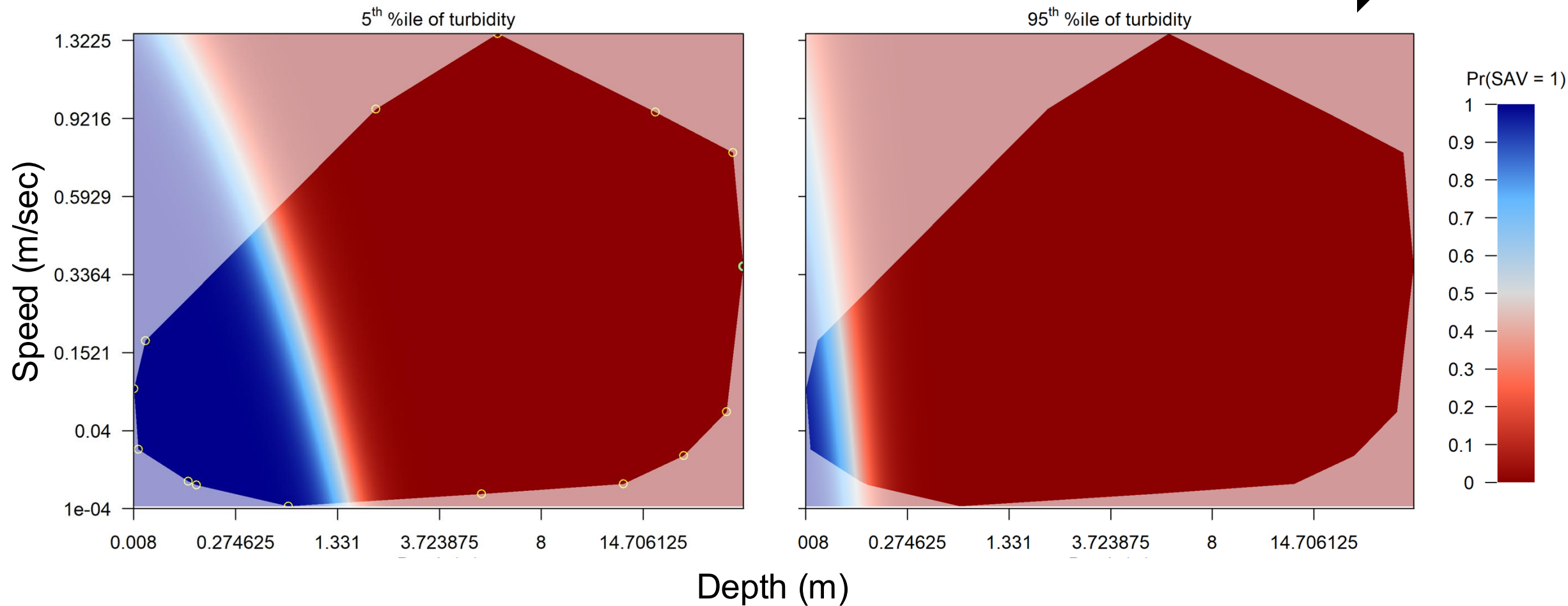


# SAV

LOW

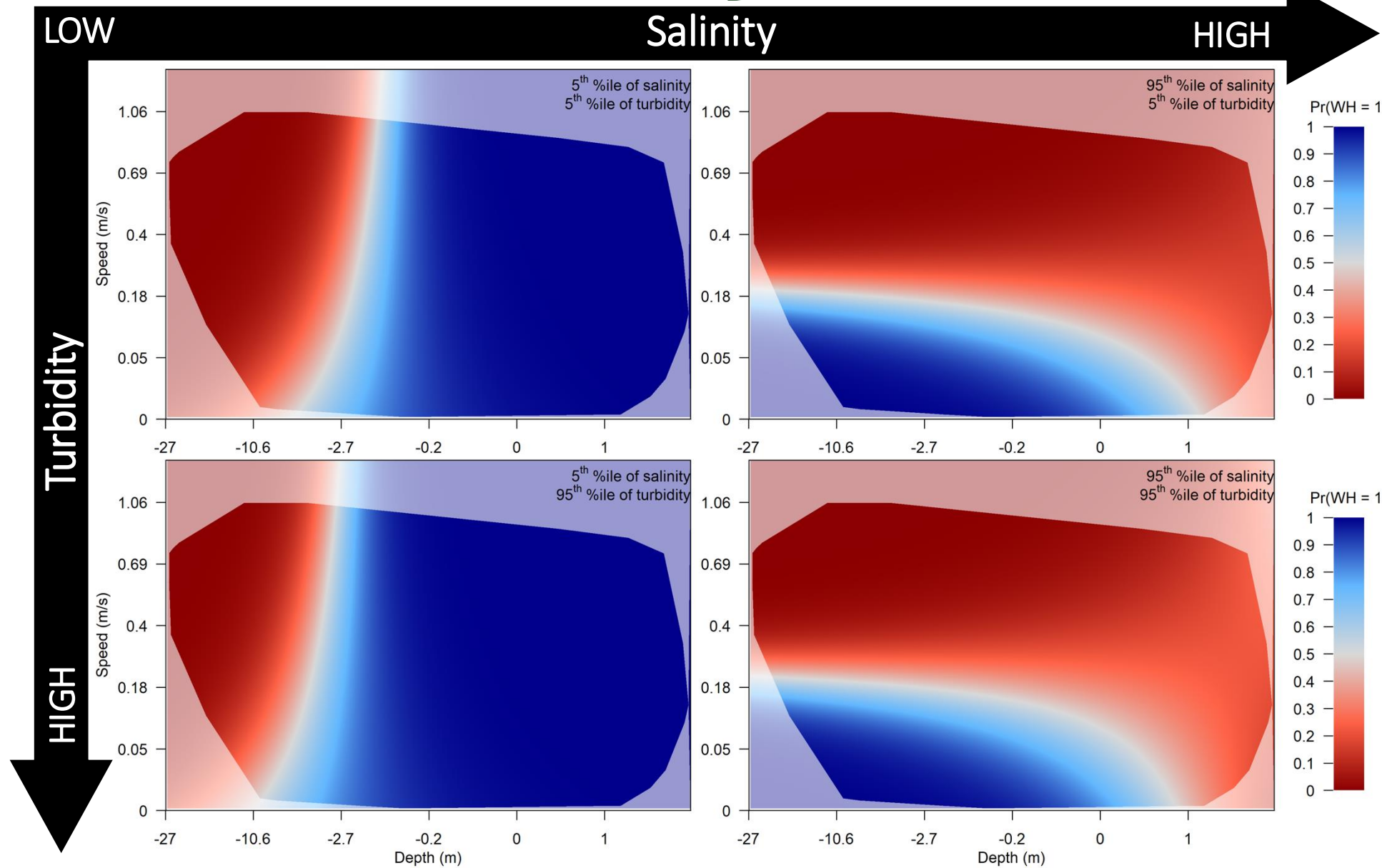
HIGH

TURBIDITY



# Water Hyacinth

CHANNEL HABITATS





# Goal 3: IAV Mapping Tool



## Testing CSIRO's EASI Platform for deployment on the cloud

- Partnered with the state through climate action MOU
- Cloud-based analysis
- Python and R supported through Jupyter Notebooks.
- Platform maintenance made easy?

```
[*]: require(raster, quietly = T, warn.conflicts = F)
require(shapefiles, quietly = T, warn.conflicts = F)
require(terra, quietly = T, warn.conflicts = F)
require(caret, lib.loc = "/home/jovyan/Rlib", quietly = T, warn.conflicts = F)
require(randomforest, lib.loc = "/home/jovyan/Rlib", quietly = T, warn.conflicts = F)
require(leaflet, lib.loc = "/home/jovyan/Rlib", quietly = T, warn.conflicts = F)

[3]: dat = as.data.frame(shapefile("calval_data/rf_20181007_bm_reduced_classes.shp"))
dat$CLASS5 = as.factor(dat$CLASS5)

[4]: train = dat[dat$rMod == "trn", ]
test = dat[dat$rMod == "test", ]

[5]: load("models/rf_20181007_bm_reduced_classes.RData")
rf5

Call:
randomforest(formula = CLASS5 ~ Blue + Green + SWIR1 + NDVI + SAVI + NDVIre2 + NDMI + PNDVI, data = train, ntree = 500)
Type of random forest: classification
Number of trees: 500
No. of variables tried at each split: 2

OOB estimate of error rate: 9.6%

Confusion matrix:
      EMR FLT-hyac FLT-prim SAV WATER class.error
EMR    44      2      2      1      1      0.12
FLT-hyac 3      44      3      0      0      0.12
FLT-prim 1      2      46      1      0      0.08
SAV      1      1      0      46      2      0.08
WATER    1      0      0      3      46      0.08

[6]: pred = predict(rf5, newdata = test)
test$predicted = pred

[7]: cm = caret::confusionMatrix(test$predicted, test$CLASS5)

[8]: cm_m = matrix(cm$table, nrow = 5, ncol = 5)
diag_vals = as.vector(diag(as.matrix(cm_m)))
PA = as.character(round(diag_vals/apply(cm_m, 1, FUN = sum), digits = 2))
UA = as.character(round(c(diag_vals/apply(cm_m, 2, FUN = sum), sum(diag_vals/row(test))), digits = 2))

cm_table = matrix(as.character(cm$table), nrow = 5, ncol = 5)
cm_table = cbind(cm_table, PA)
cm_table = rbind(cm_table, UA)
rownames(cm_table) = c("EMR", "FLT-hyac", "FLT-prim", "SAV", "WATER", "UA")
colnames(cm_table) = c("EMR", "FLT-hyac", "FLT-prim", "SAV", "WATER", "PA")
cm_table

A matrix 6 x 6 of type chr
      EMR FLT-hyac FLT-prim SAV WATER PA
EMR    46      0      3      0      1 0.92
FLT-hyac 3      45      3      1      0 0.87
FLT-prim 0      3      44      0      0 0.94
SAV      1      2      0      45      4 0.87
WATER    0      0      0      4      45 0.92
```

# Next Steps and Plans



## **S2 Class Models:**

- Construct 2021-2025 cal-val datasets
- Test all-season vs. season S2 models

## **SDMs:**

- AVIRIS-3 flights continue for 2025
- Additional environmental variables:
  - Bulk Water Temperature
  - Surface Temperature
  - Fine resolution bathymetry
- Try more sophisticated models, address scale and hydro-cells

## **Mapping tool:**

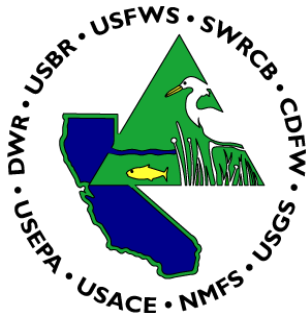
- Iterate with stakeholders
- Ensure suitability for decision-making frameworks





# Thank You

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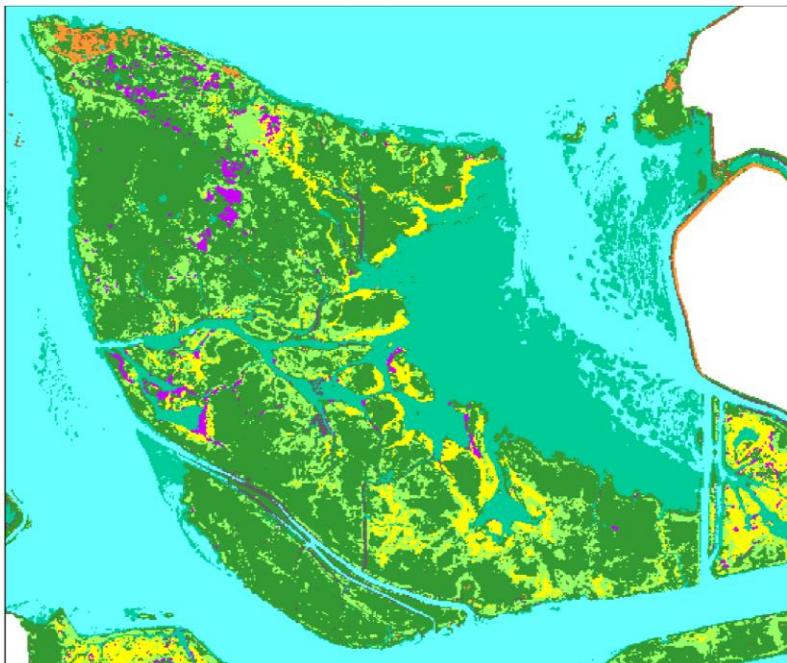


Agency	Primary Role in IAV Management
Delta Stewardship Council (DSC)	Oversees implementation of the Delta Plan; sets policy direction; tracks progress and promotes coordination across agencies. Provides science funding and policy support but does <b>not implement field control measures</b> .
California Department of Fish and Wildlife (CDFW)	Leads <b>on-the-ground species monitoring, enforcement, and habitat protection</b> . Its Invasive Species Program helps prevent, detect, and manage non-native species impacts.
Interagency Ecological Program (IEP)	A <b>science consortium</b> (including CDFW and DWR) that conducts long-term ecological monitoring and targeted research to inform adaptive management, including studies on IAV trends, impacts, and ecosystem effects.
Department of Water Resources (DWR)	Supports IAV management through <b>restoration projects, water operations</b> , and coordination with the Division of Boating and Waterways (DBW), which carries out much of the direct IAV treatment (e.g., herbicide spraying). Also contributes to long-term monitoring and modeling.

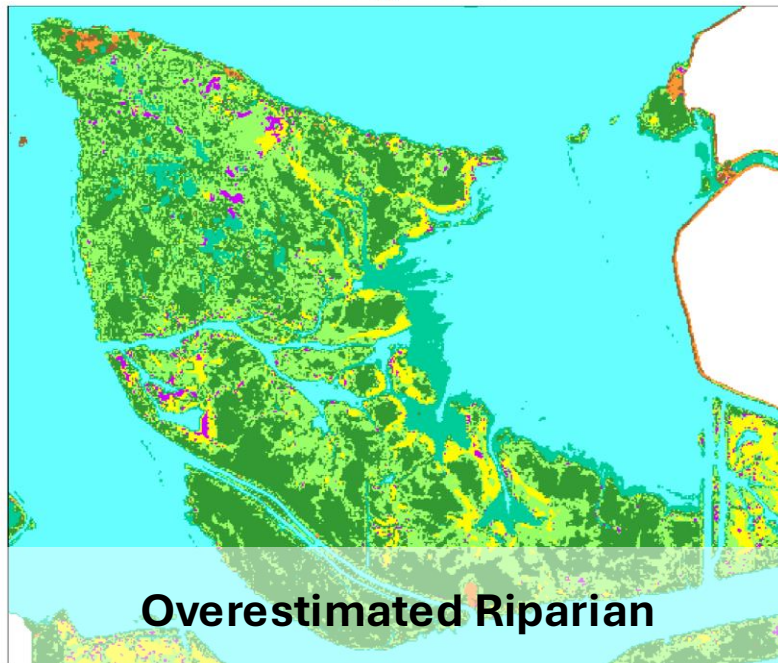


# Goal 1: Operationalize S2 Class Mapper

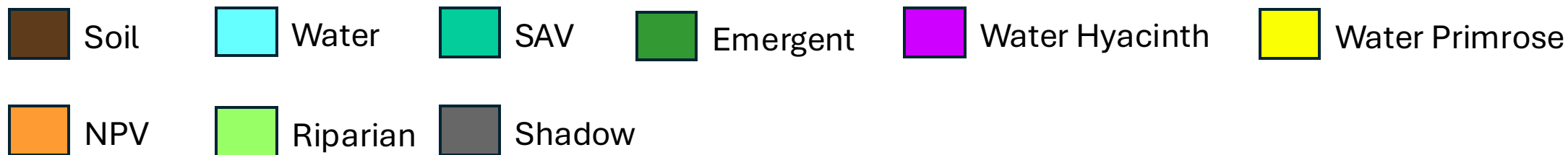
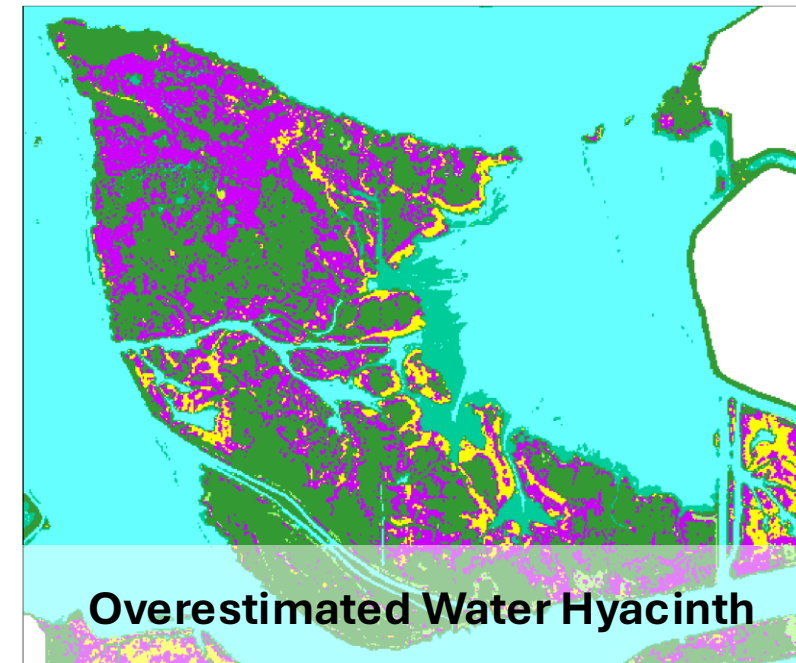
AIS 10/7/2018



S2 Prototype 10/7/2018

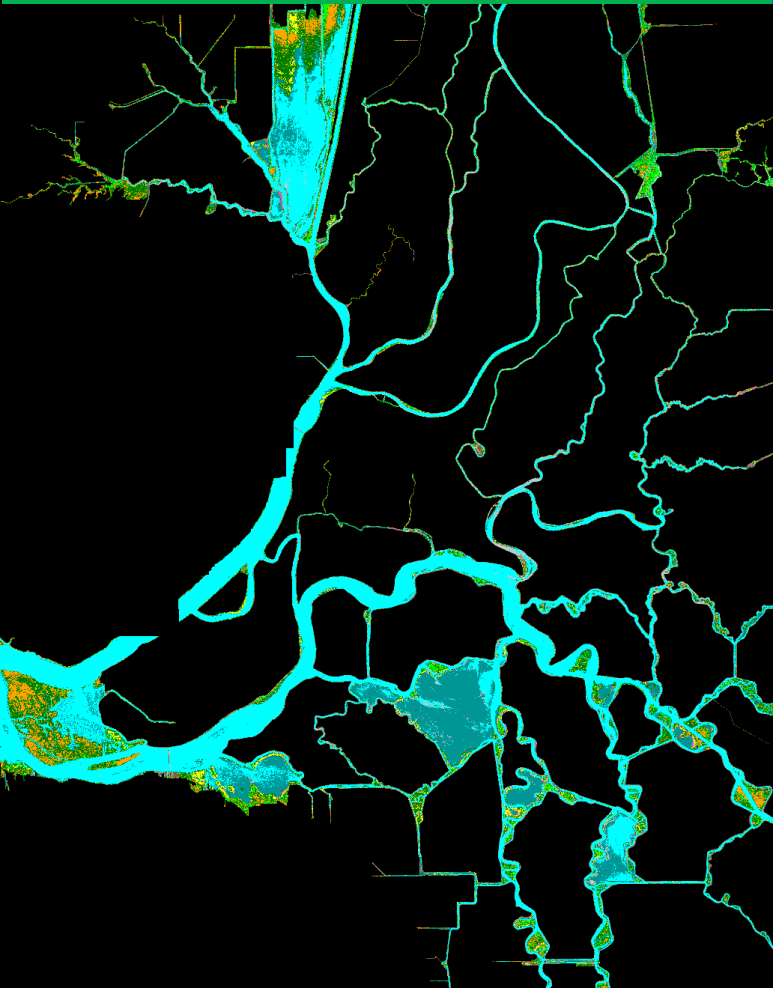


S2 Operationalize 10/7/2018



# Goal 2: SDMs to Assess IAV Responses to Management Actions

Response Variable



Environmental Predictors

