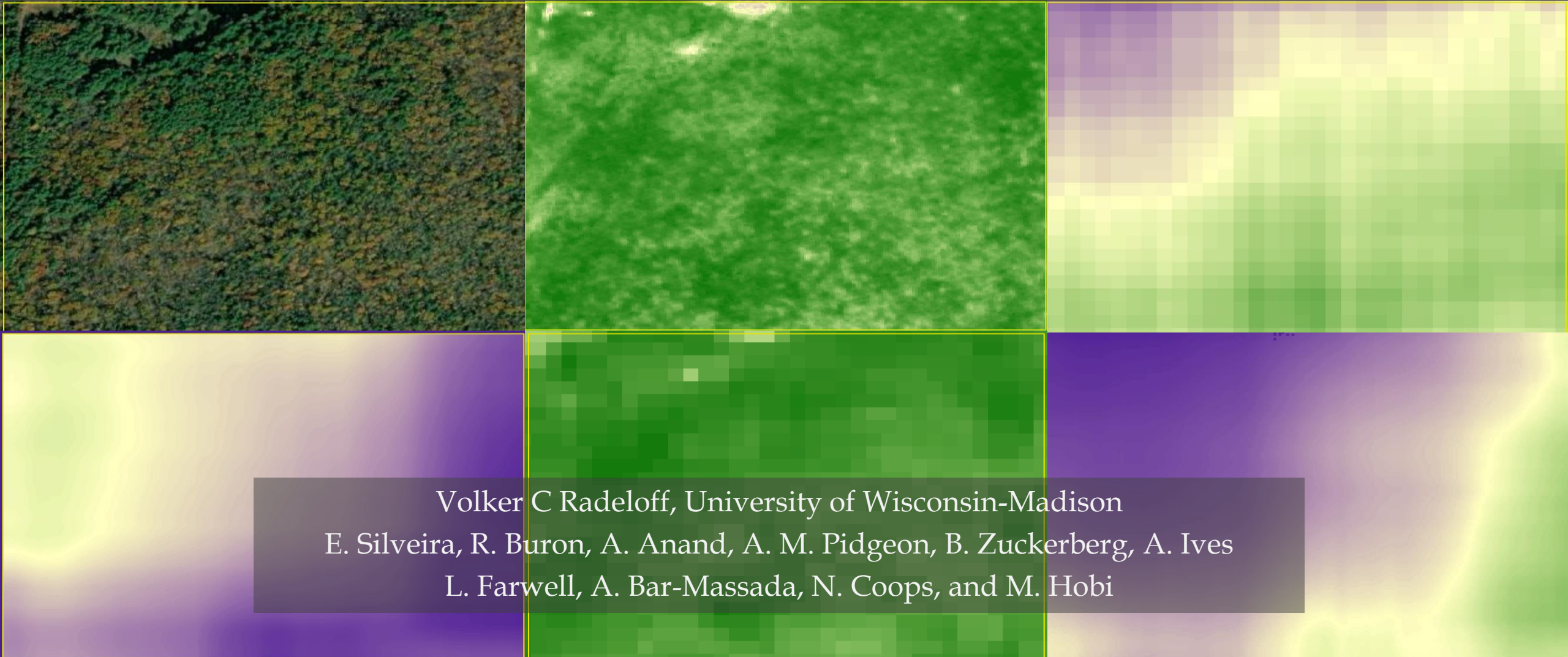
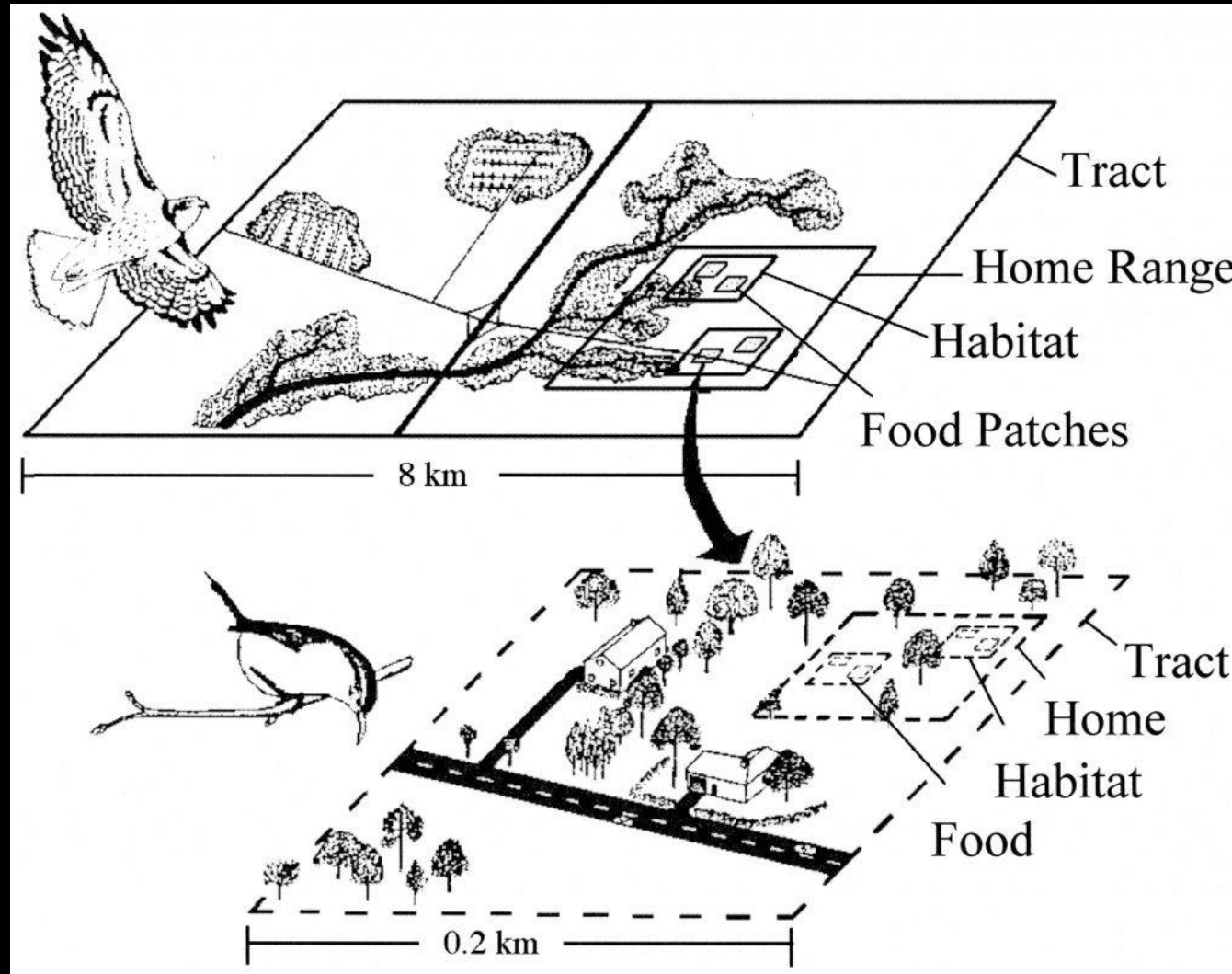


Multiple spatial scales, long-term trends, and synchrony of the dynamic habitat indices and bird populations

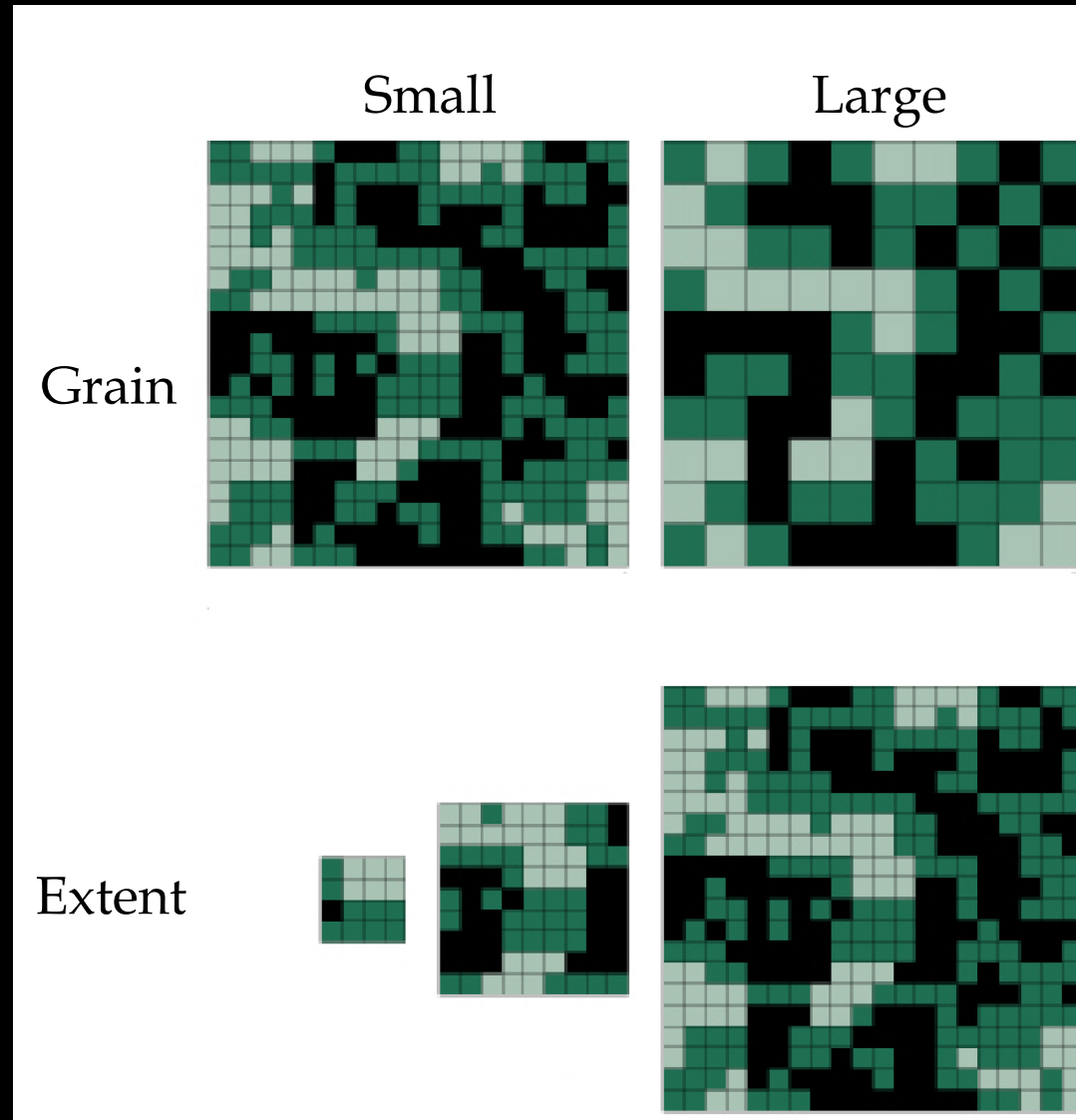


Volker C Radeloff, University of Wisconsin-Madison
E. Silveira, R. Buron, A. Anand, A. M. Pidgeon, B. Zuckerberg, A. Ives
L. Farwell, A. Bar-Massada, N. Coops, and M. Hobi

Introduction



Introduction



A. Hansen, MSU

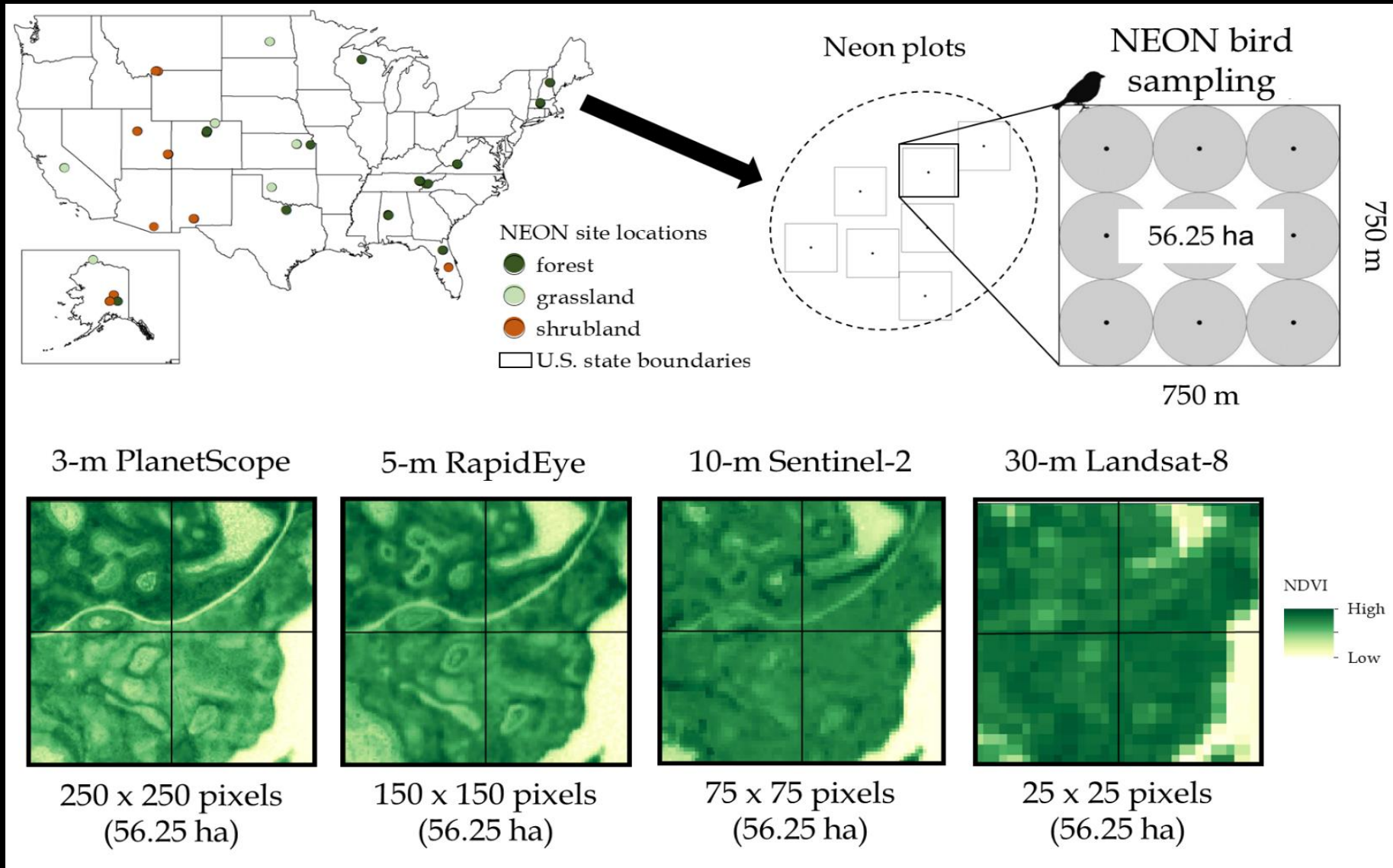


Outline

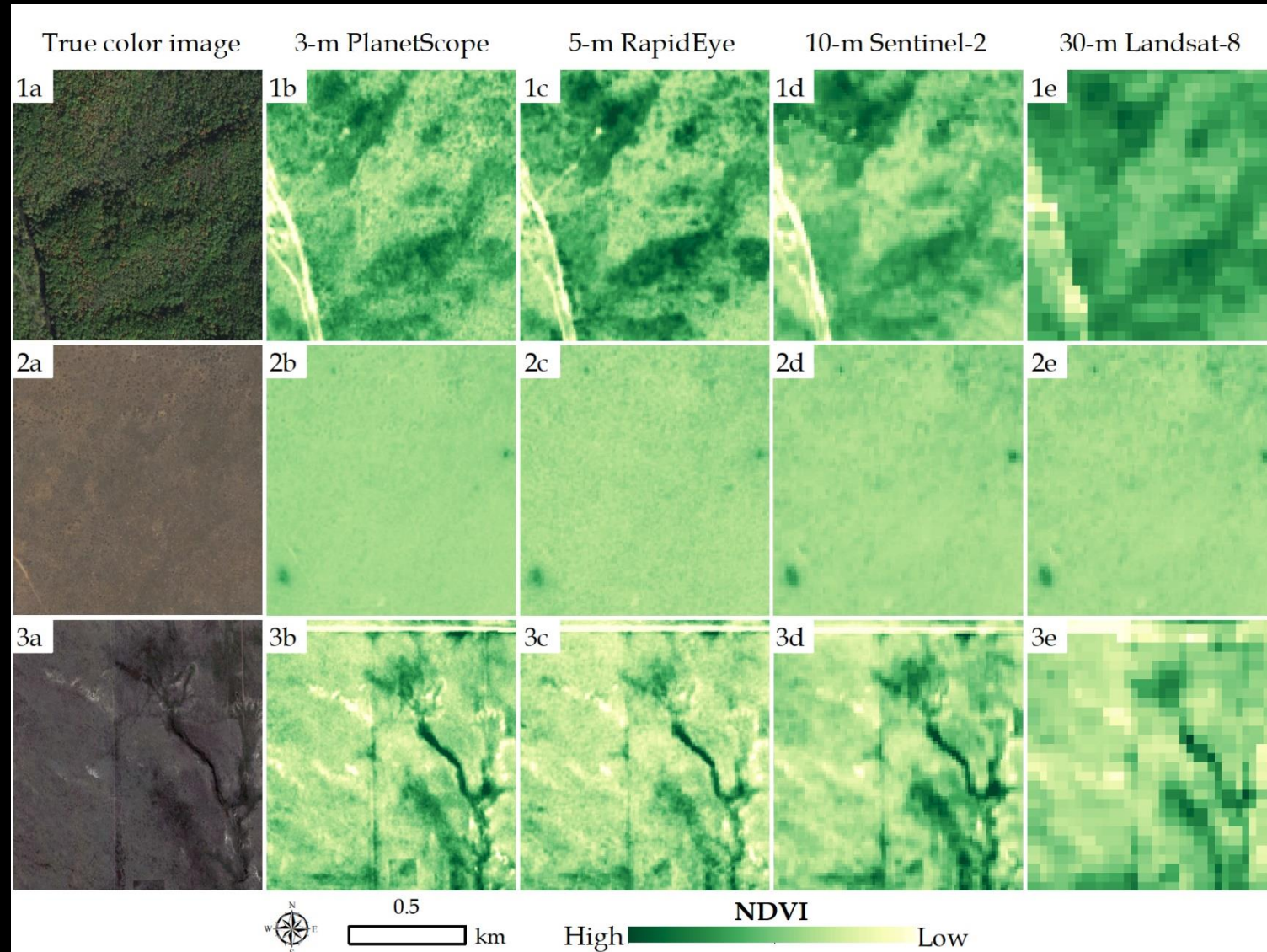
- Part I: Multi-grain habitat models
 - Can high-res imagery predict bird richness better than S2/L8?
 - Is there one optimal grain, or do multiple grains perform best?
 - Is the same grain best for different bird guilds?
- Part II: Scalograms
 - Can we synthesize habitat patterns from a range of extents?
- Part III: Conclusions



Methods

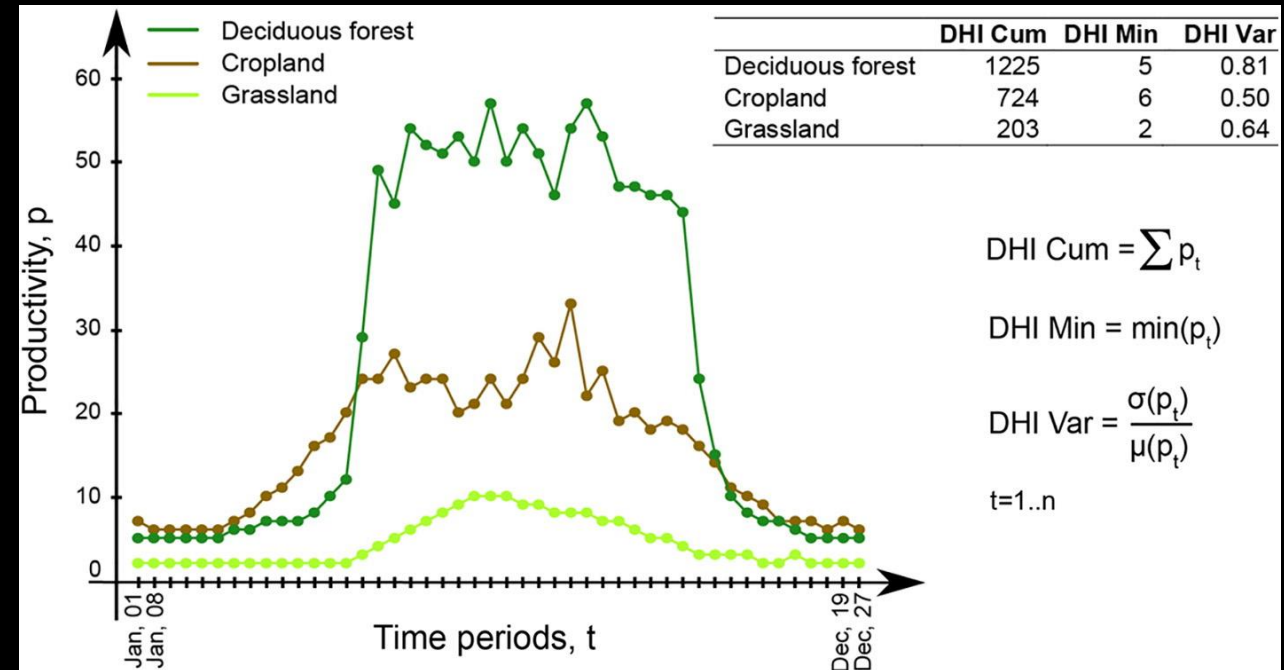


Results

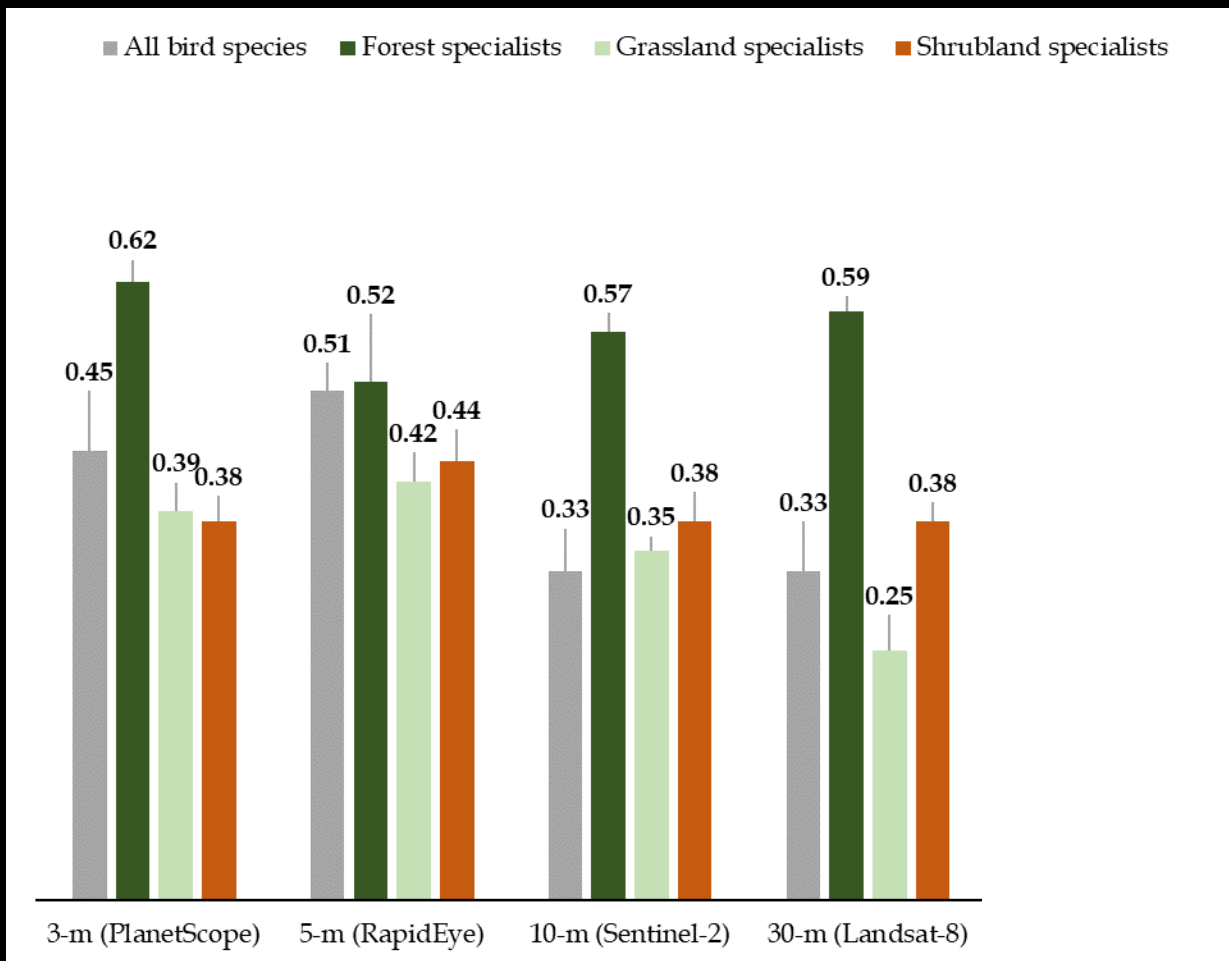


Methods

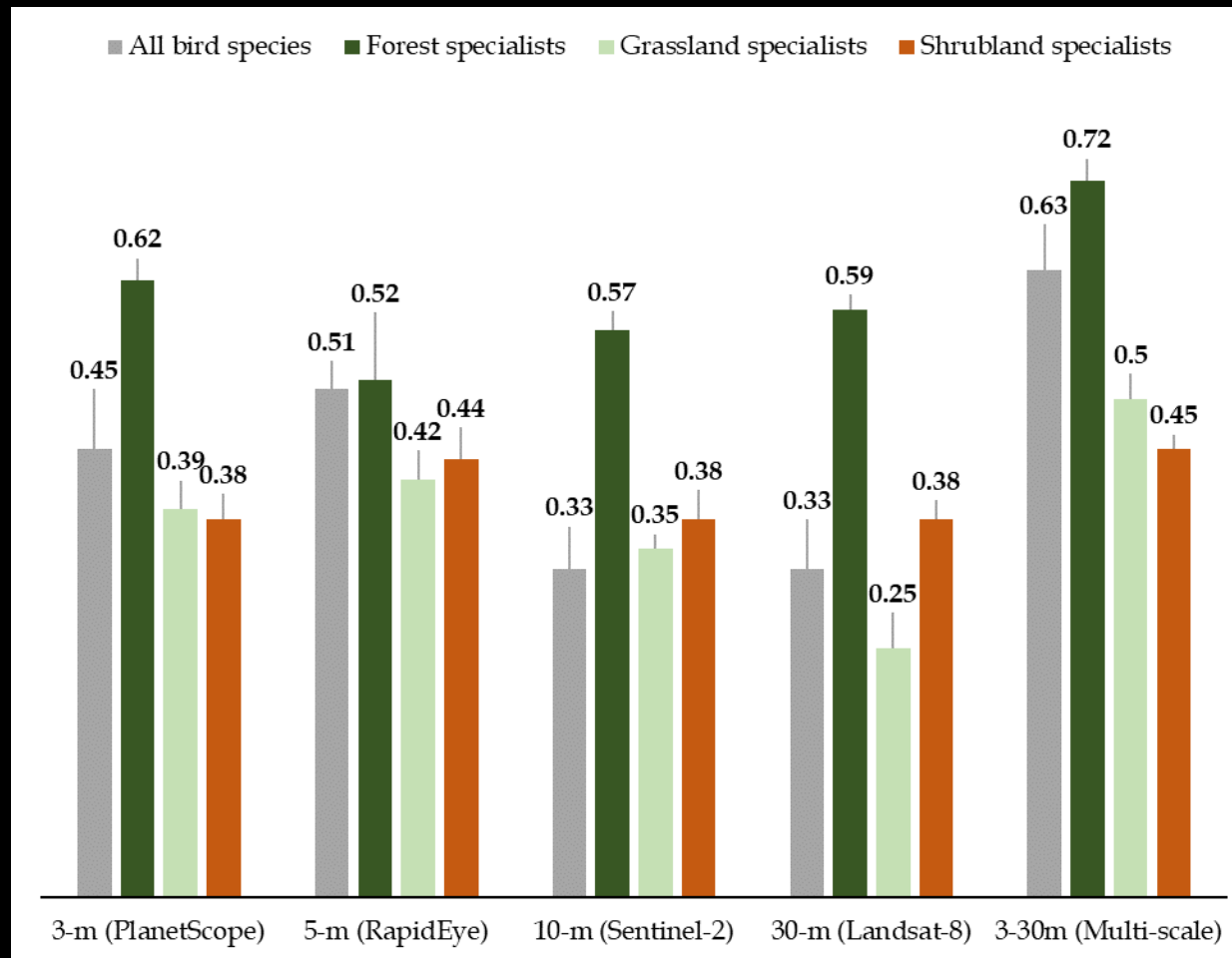
- Median monthly NDVI values from 2016-'20
- Dynamic Habitat Indices (DHIs)
 - Cumulative (sum)
 - Minimum (min)
 - Variation (CV)
- NDVI 90th percentile
- NDVI median



Results

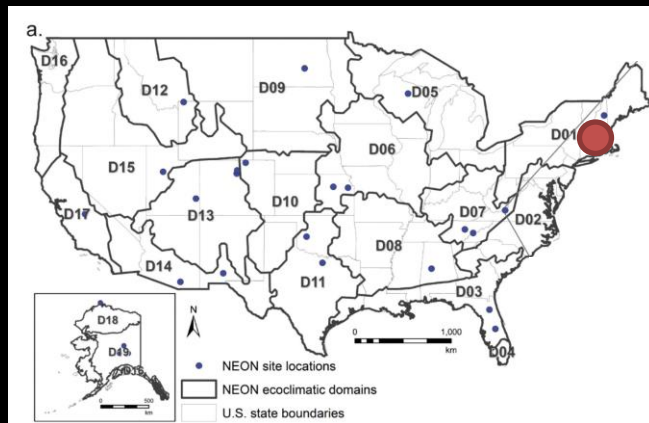


Results



Results

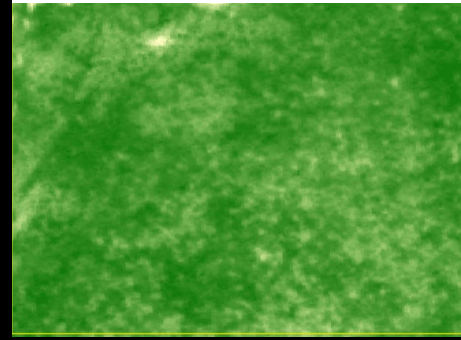
(Int)	L8 MinDHI HOM	L8 NDVIp90 MEAN	PS MinDHI SD	PS NDVIp90 MEAN	RE MinDHI UNIF	adjR ²
-8.325	-8.373	-17.42	108.2	40.5	73.89	0.72



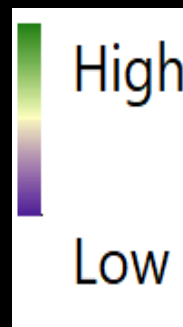
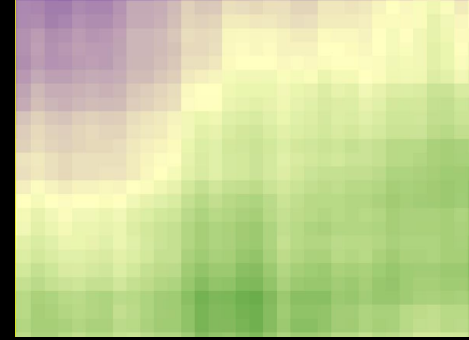
HARV 025: Deciduous forests



PS_NDVIp90_MEAN



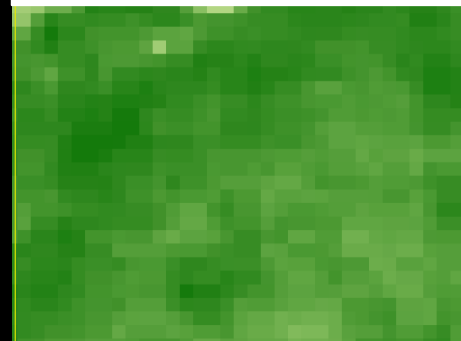
L8_MinDHI_HOM



PS_MinDHI_SD



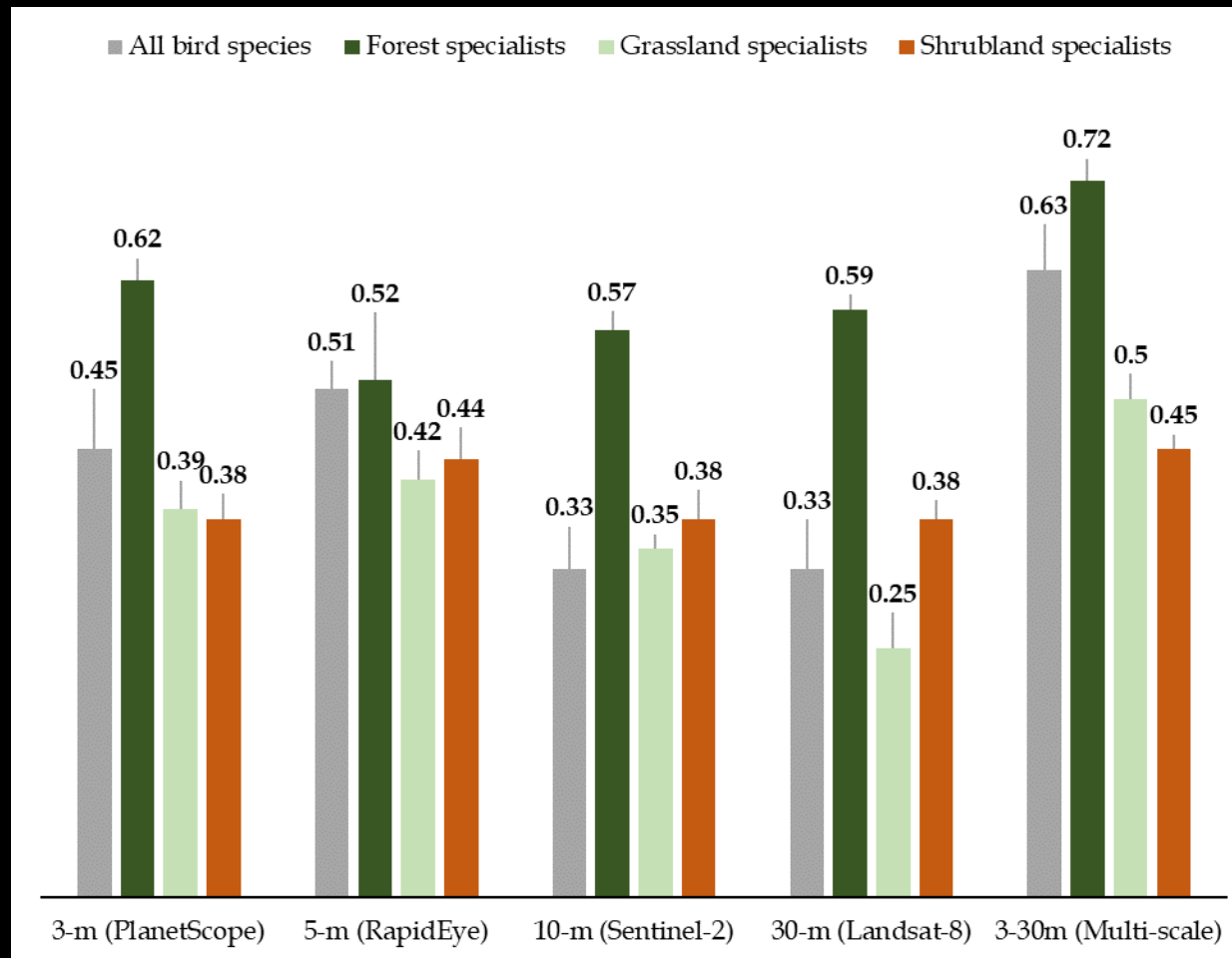
L8_NDVIp90_MEAN



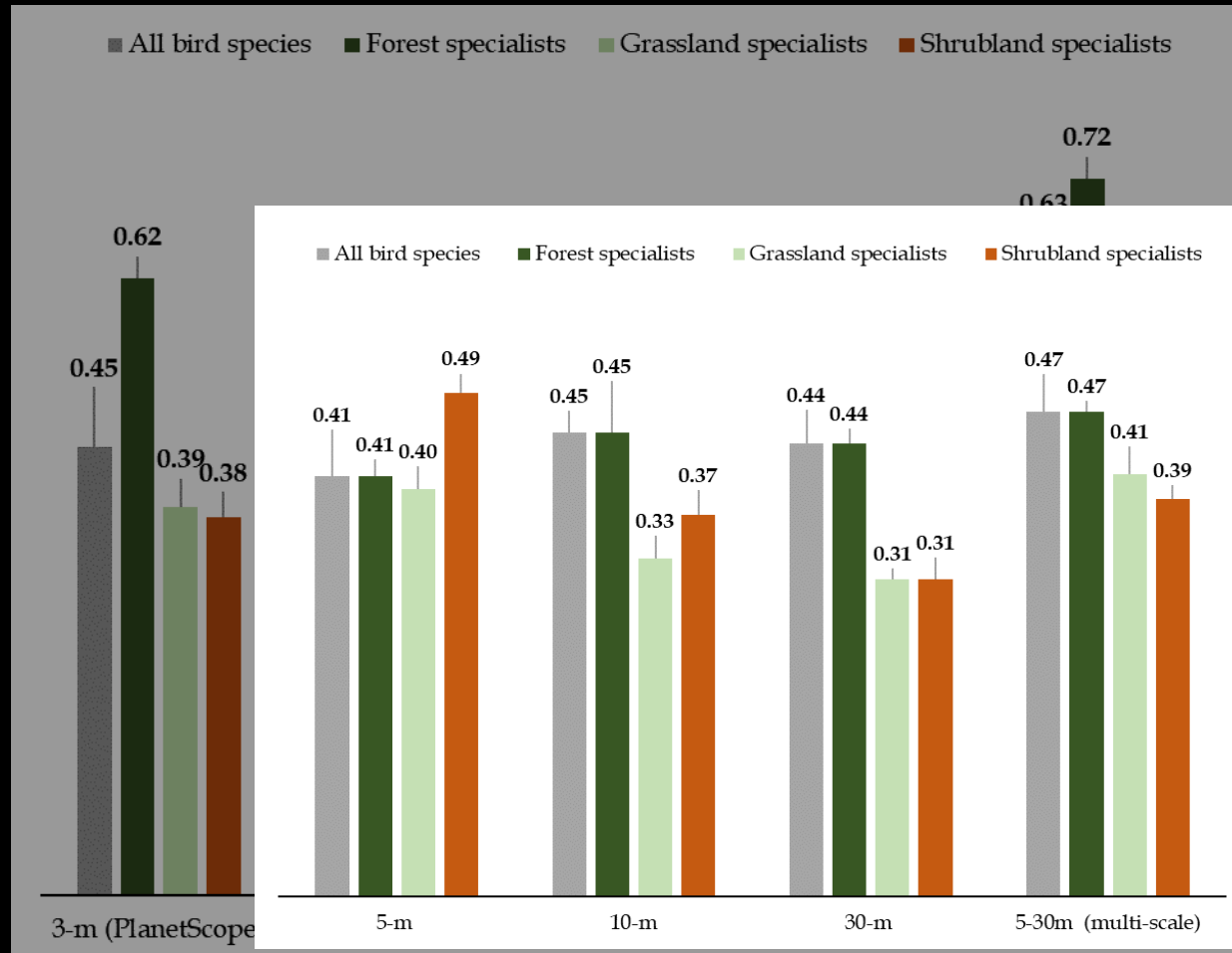
RE_MinDHI_UNIF



Results



Results



Results

1. Can high-resolution imagery predict bird richness better than S2/L8?
2. Is there one optimal grain, or do multi-grain models perform best?
3. Are the same grains best for different bird guilds?



Results

1. Can high-resolution imagery predict bird richness better than S2/L8?
 - Yes! RapidEye was best in univariate models for all-species.
 - PlanetScope/RapidEye single-grain multivariate models best for all guilds.



Results

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Results

1. Can high-resolution imagery predict bird richness better than S2/L8?
 - Yes! RapidEye was best in univariate models for all-species.
 - PlanetScope/RapidEye multivariate models best for all guilds.
2. Is there one optimal grain, or do multi-grain models perform best?
 - No! There is not one optimal grain.
 - Yes! Multi-grain models performed best for all guilds
3. Is the same grain best for different bird guilds?
 - No! In univariate models, forest specialists best predicted by S2/L8, grassland specialists by PlanetScope/Rapideye, and shrubland specialist by any/all

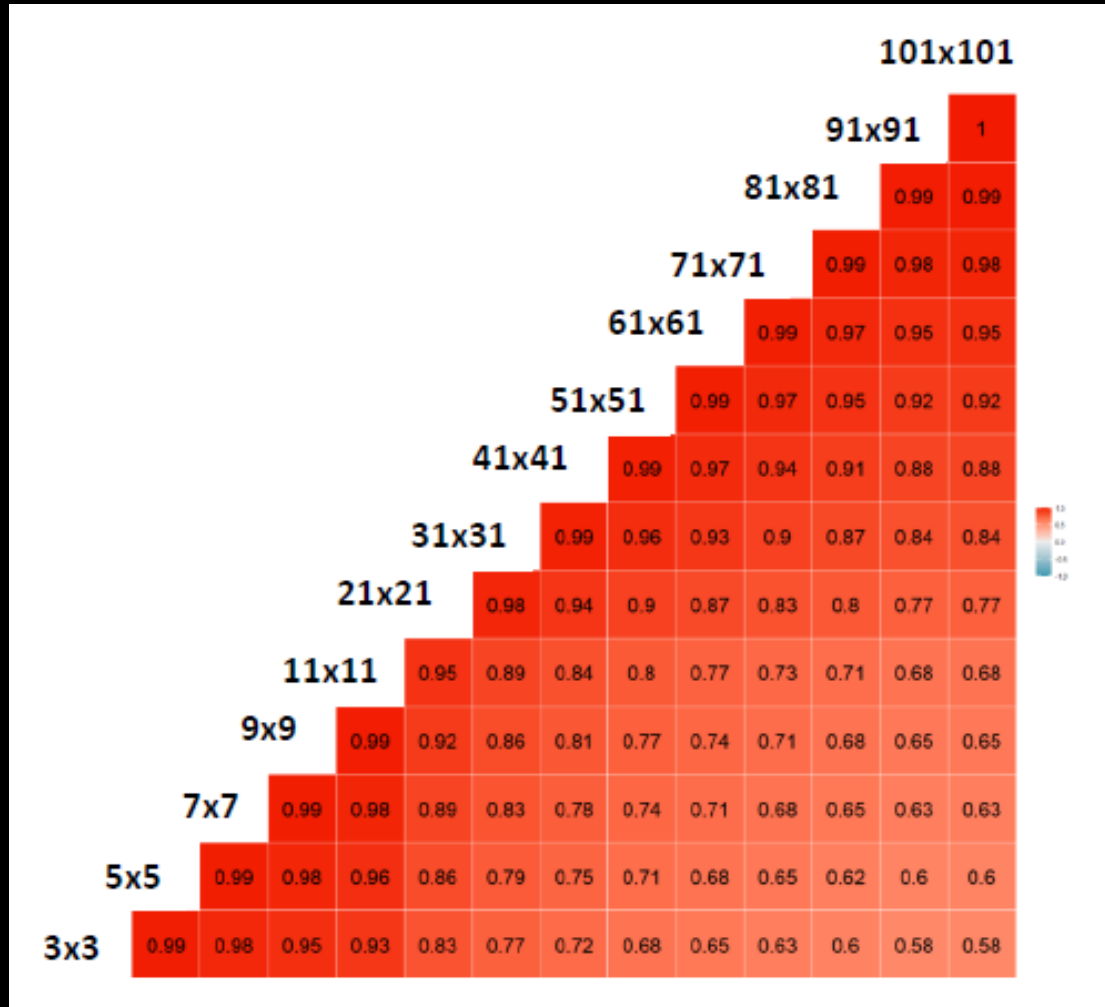


Outline

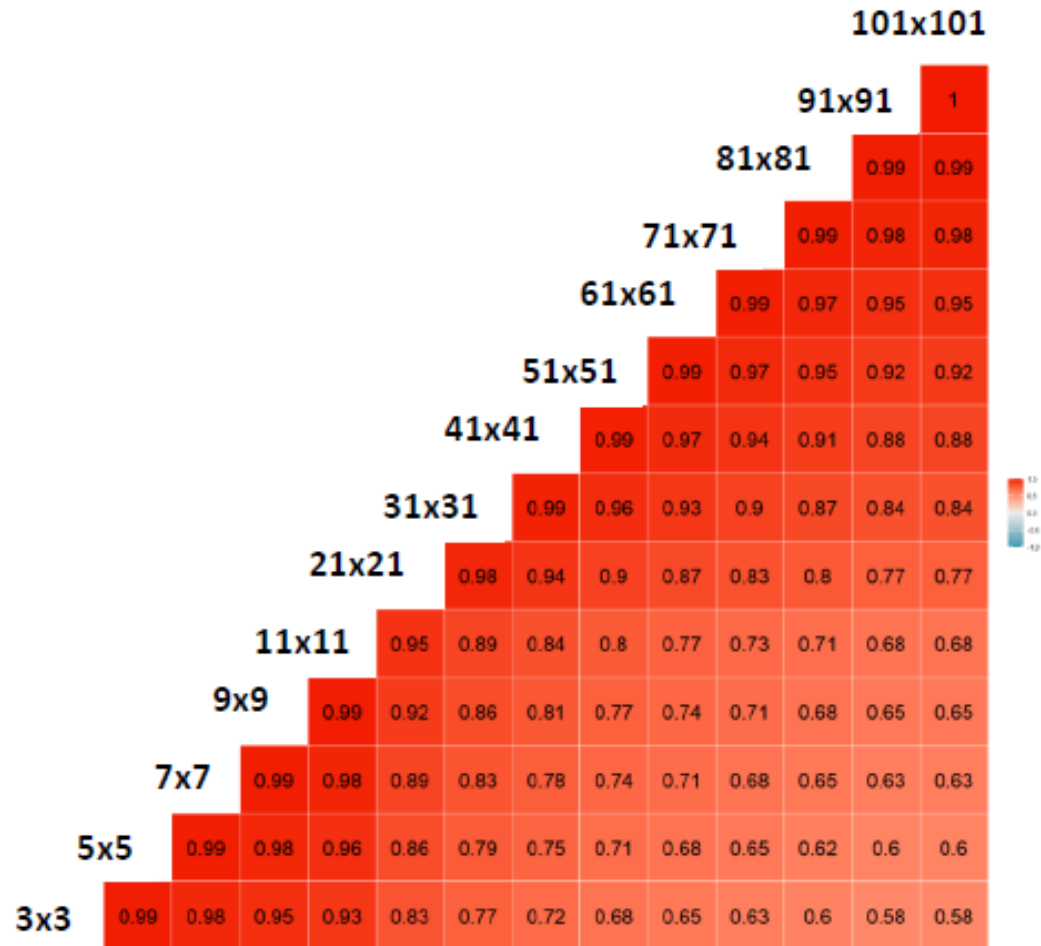
- Part I: Multi-grain habitat models
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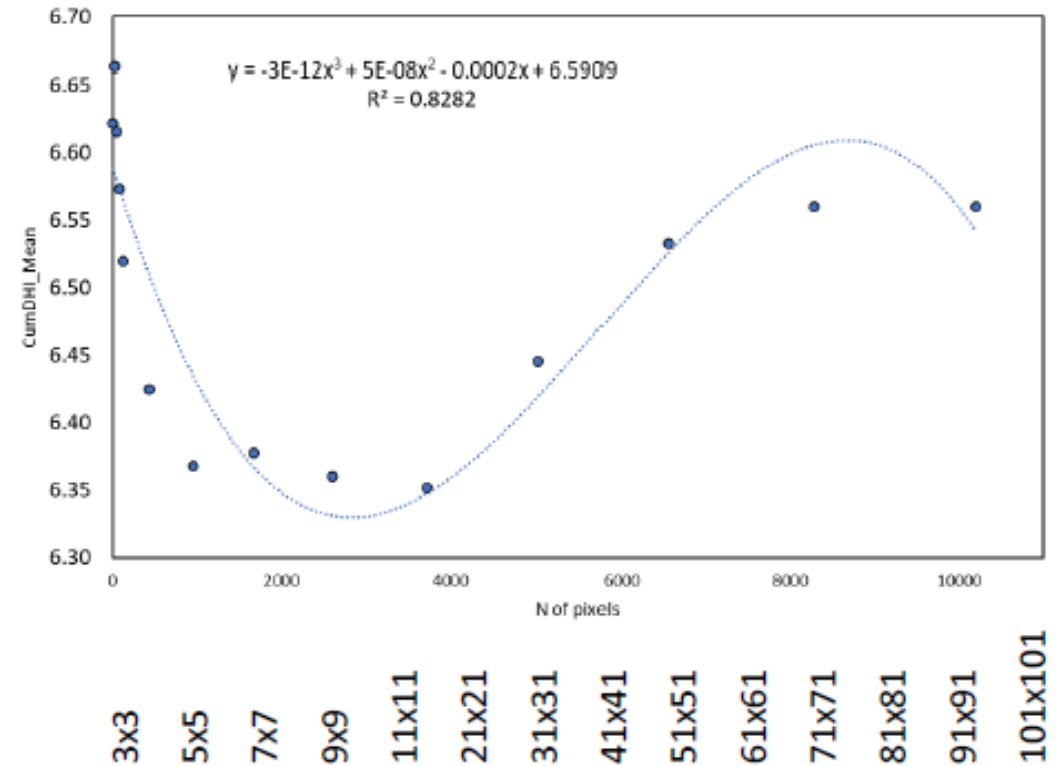
Results



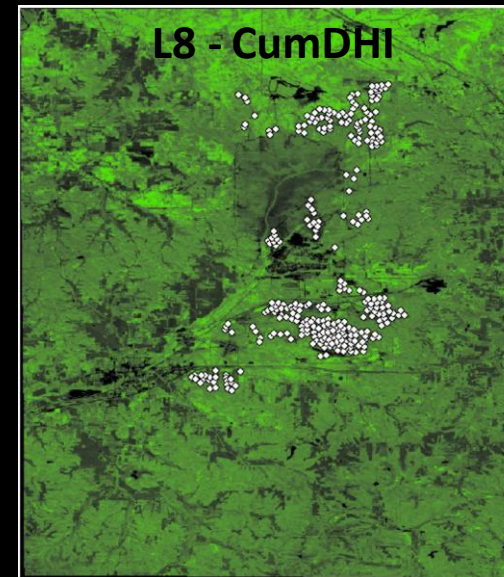
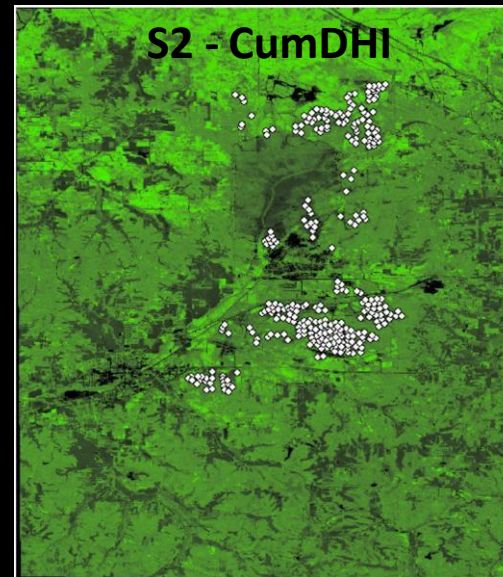
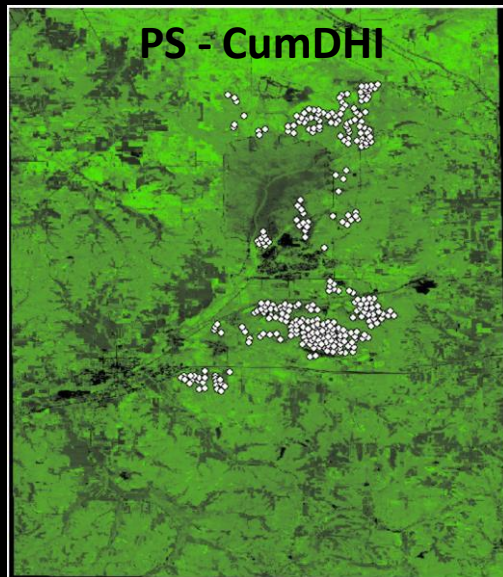
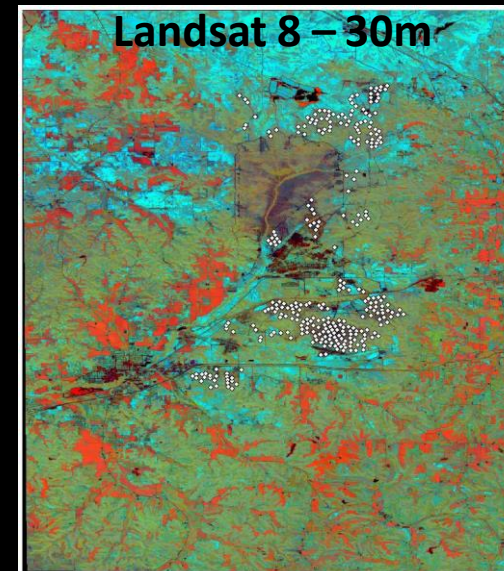
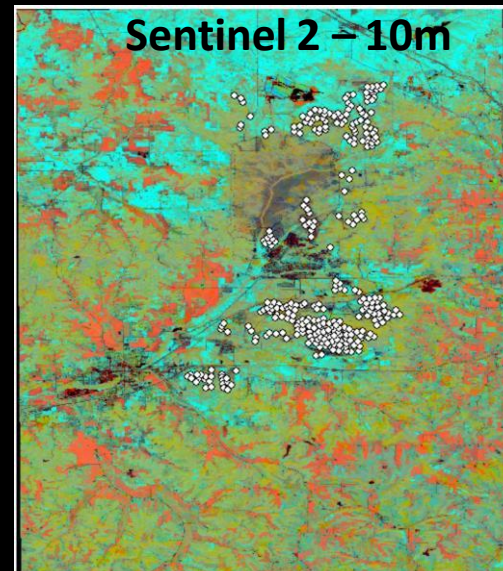
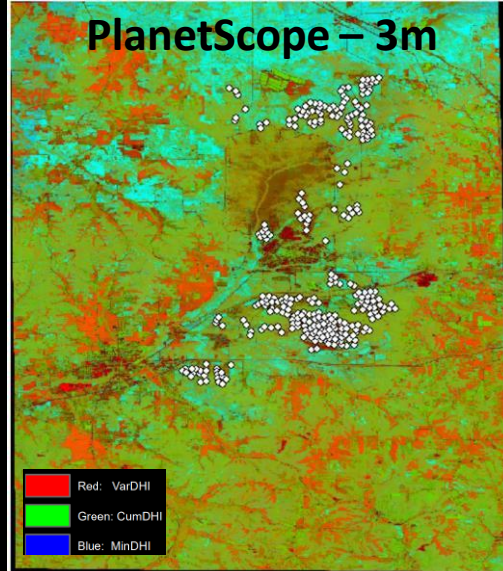
Results



Scalogram

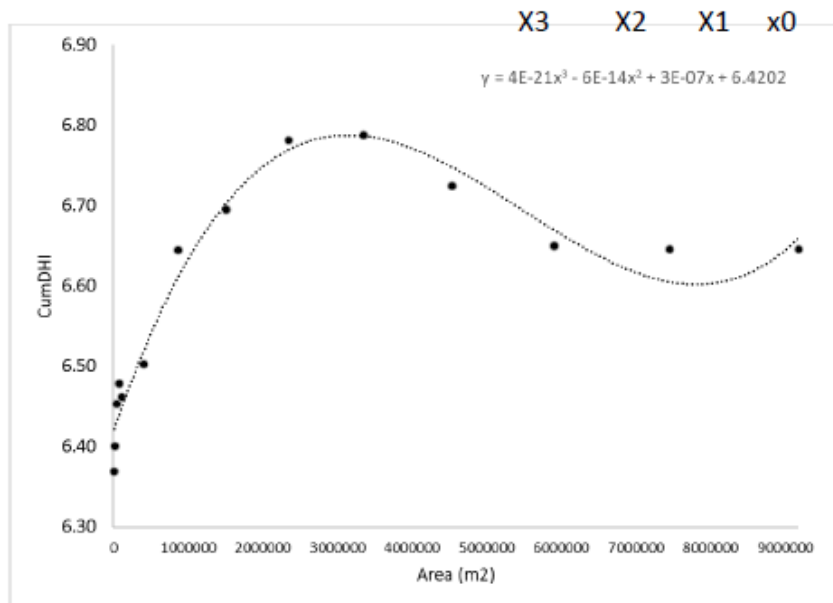


Results



Results

	BAOR	BHCO	BRTH	CCSP	CHSP	DICK	EABL	EAKI	EAME	EATO	EAWP	FISP	GCFL	GRCA	GRSP	HOLA	HOWR	INBU	LASP	MODO	OROR	OVEN	RBGR	REVI	RHWO	SCTA	UPSA	VESP
L8_CumDHI_MEAN_3	-0.05	0.16	-0.13	-0.15	-0.10	-0.19	-0.22	-0.22	-0.34	0.34	0.53	-0.08	0.23	0.06	-0.59	-0.32	-0.04	0.37	-0.23	-0.28	-0.12	0.68	0.47	0.56	-0.02	0.57	-0.30	-0.57
L8_CumDHI_MEAN_5	-0.05	0.17	-0.14	-0.16	-0.10	-0.21	-0.22	-0.23	-0.35	0.35	0.54	-0.08	0.25	0.04	-0.60	-0.32	-0.05	0.38	-0.22	-0.29	-0.12	0.69	0.48	0.57	-0.03	0.57	-0.31	-0.57
L8_CumDHI_MEAN_7	-0.05	0.17	-0.16	-0.17	-0.09	-0.23	-0.23	-0.25	-0.37	0.35	0.55	-0.09	0.27	0.01	-0.63	-0.34	-0.06	0.38	-0.22	-0.32	-0.14	0.69	0.49	0.58	-0.05	0.58	-0.32	-0.57
L8_CumDHI_MEAN_9	-0.06	0.16	-0.18	-0.17	-0.09	-0.24	-0.25	-0.25	-0.38	0.34	0.57	-0.10	0.29	-0.02	-0.64	-0.35	-0.08	0.37	-0.23	-0.33	-0.15	0.70	0.51	0.58	-0.07	0.59	-0.33	-0.56
L8_CumDHI_MEAN_11	-0.06	0.15	-0.19	-0.16	-0.08	-0.25	-0.26	-0.25	-0.39	0.33	0.57	-0.10	0.29	-0.04	-0.65	-0.36	-0.08	0.38	-0.23	-0.33	-0.15	0.70	0.51	0.58	-0.07	0.59	-0.34	-0.54
L8_CumDHI_MEAN_21	-0.04	0.17	-0.17	-0.16	0.00	-0.30	-0.25	-0.22	-0.41	0.35	0.56	-0.05	0.29	-0.01	-0.69	-0.40	-0.04	0.42	-0.21	-0.29	-0.14	0.65	0.50	0.53	-0.06	0.58	-0.37	-0.45
L8_CumDHI_MEAN_31	-0.03	0.16	-0.16	-0.18	0.02	-0.33	-0.25	-0.18	-0.42	0.33	0.54	-0.05	0.28	0.01	-0.69	-0.43	-0.04	0.41	-0.22	-0.30	-0.12	0.63	0.50	0.52	-0.07	0.57	-0.39	-0.39
L8_CumDHI_MEAN_41	-0.04	0.13	-0.15	-0.16	0.02	-0.32	-0.25	-0.16	-0.42	0.31	0.51	-0.05	0.27	0.02	-0.67	-0.44	-0.06	0.39	-0.22	-0.29	-0.10	0.60	0.48	0.49	-0.09	0.55	-0.40	-0.34
L8_CumDHI_MEAN_51	-0.04	0.13	-0.13	-0.13	0.03	-0.32	-0.24	-0.15	-0.42	0.30	0.49	-0.04	0.25	0.04	-0.65	-0.44	-0.05	0.38	-0.25	-0.27	-0.08	0.58	0.46	0.48	-0.09	0.53	-0.40	-0.31
L8_CumDHI_MEAN_61	-0.03	0.13	-0.12	-0.09	0.04	-0.33	-0.22	-0.13	-0.43	0.30	0.46	-0.02	0.24	0.07	-0.64	-0.43	-0.03	0.36	-0.26	-0.25	-0.06	0.56	0.44	0.47	-0.08	0.50	-0.39	-0.29
L8_CumDHI_MEAN_71	-0.03	0.13	-0.11	-0.06	0.06	-0.34	-0.19	-0.10	-0.43	0.28	0.42	-0.01	0.25	0.08	-0.61	-0.41	-0.01	0.34	-0.25	-0.24	-0.03	0.55	0.43	0.46	-0.08	0.47	-0.38	-0.26
L8_CumDHI_MEAN_81	-0.02	0.14	-0.10	-0.04	0.08	-0.35	-0.17	-0.08	-0.44	0.26	0.40	0.00	0.25	0.09	-0.60	-0.39	0.00	0.32	-0.24	-0.23	-0.01	0.53	0.42	0.46	-0.08	0.45	-0.36	-0.23
L8_CumDHI_MEAN_91	-0.01	0.15	-0.08	-0.03	0.09	-0.36	-0.15	-0.07	-0.45	0.25	0.38	0.00	0.24	0.10	-0.58	-0.37	0.02	0.31	-0.23	-0.22	0.01	0.51	0.41	0.45	-0.07	0.43	-0.34	-0.20
L8_CumDHI_MEAN_101	-0.01	0.15	-0.08	-0.03	0.09	-0.36	-0.15	-0.07	-0.45	0.25	0.38	0.00	0.24	0.10	-0.58	-0.37	0.02	0.31	-0.23	-0.22	0.01	0.51	0.41	0.45	-0.07	0.43	-0.34	-0.20



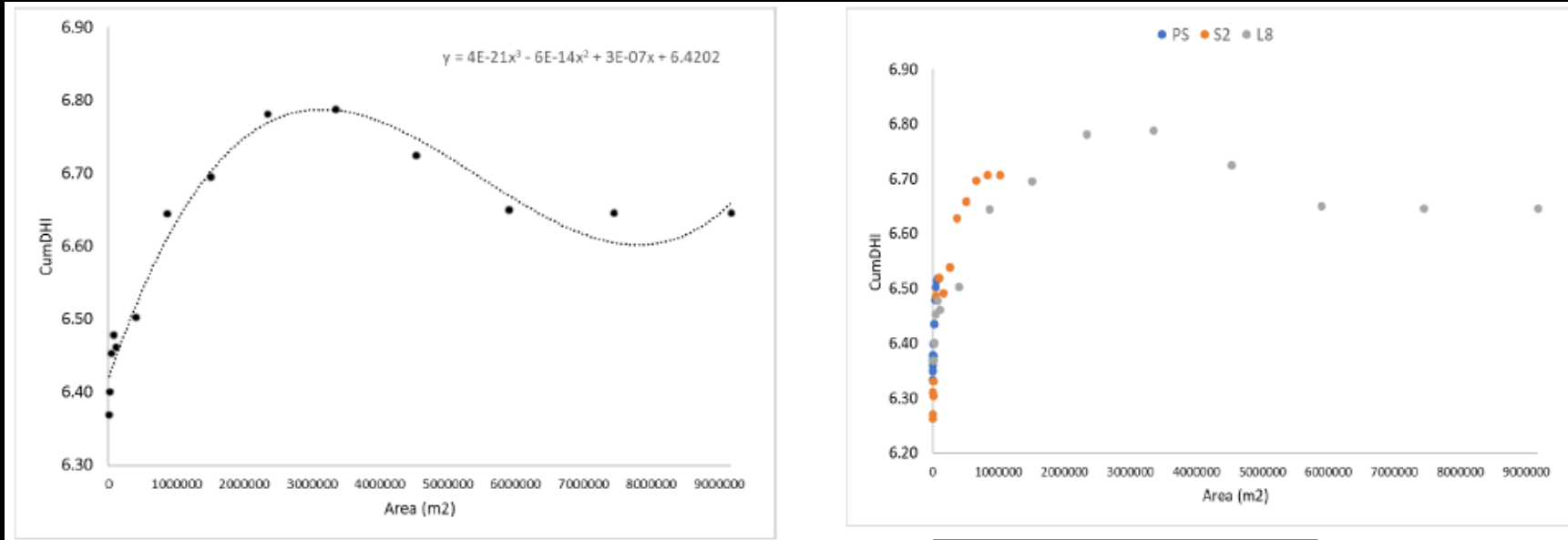
OVEN – Optimal scale (9x9 / 11x11) - $r = 0.7$

	OVEN
X3	L8_CumDHI_MEAN_3 -0.22
X2	L8_CumDHI_MEAN_2 0.26
X1	L8_CumDHI_MEAN_1 -0.34
x0	L8_CumDHI_MEAN_0 0.70

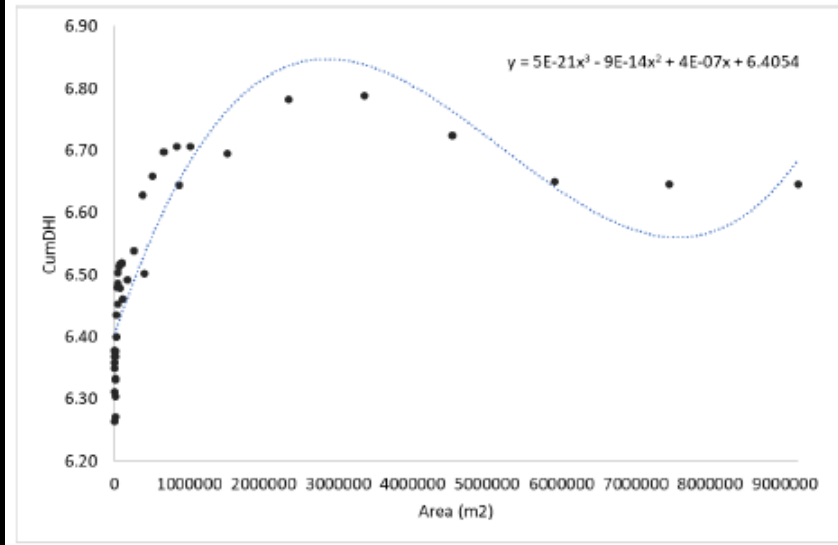
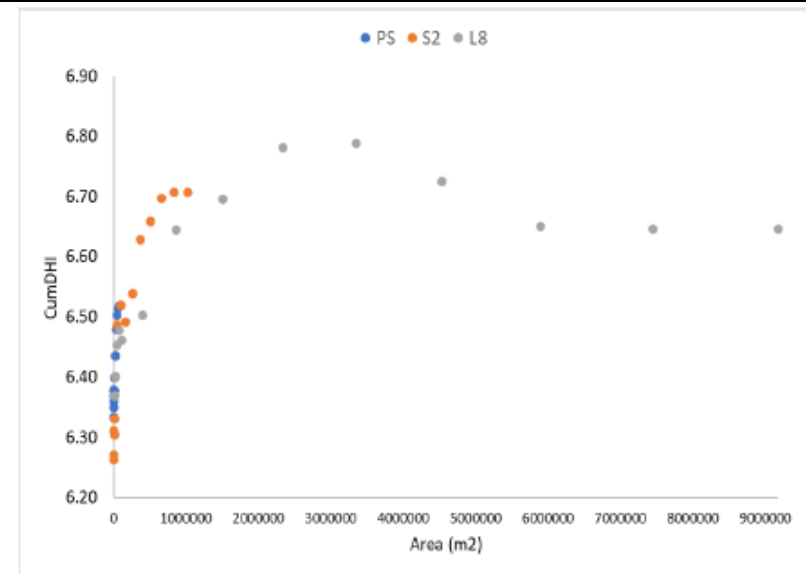
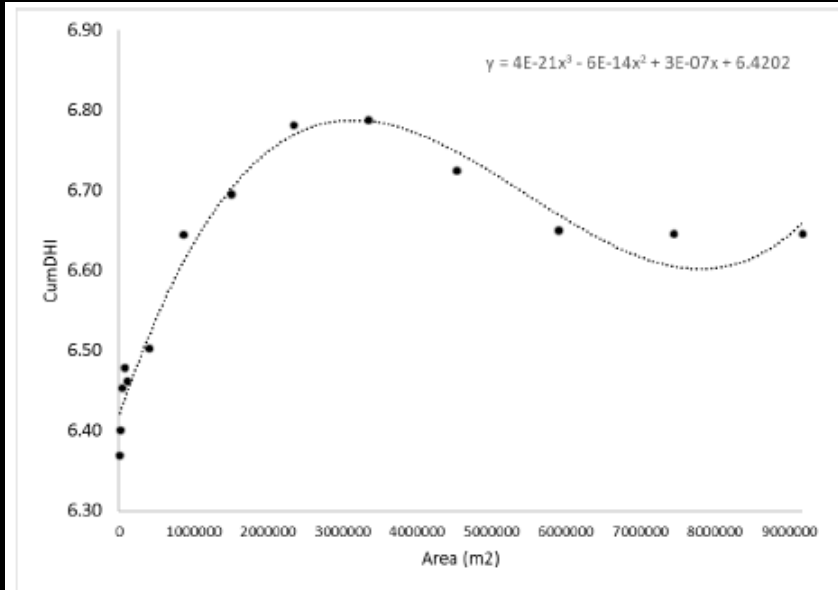
Multiple regression: $r = 0.71$



Results



Results



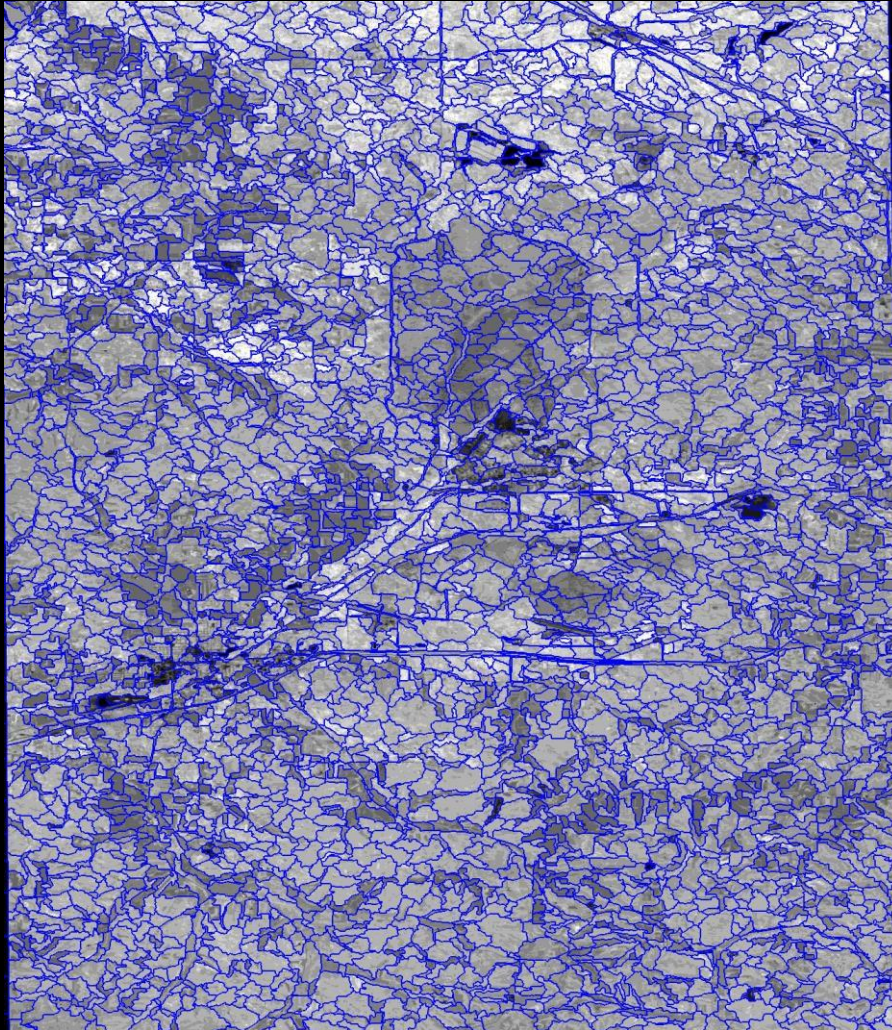
	OVEN
PS_S2_L8_cumDHI_MEAN_3	-0.14
PS_S2_L8_cumDHI_MEAN_2	0.19
PS_S2_L8_cumDHI_MEAN_1	-0.3
PS_S2_L8_cumDHI_MEAN_0	0.71

Multiple regression: $r = 0.74$

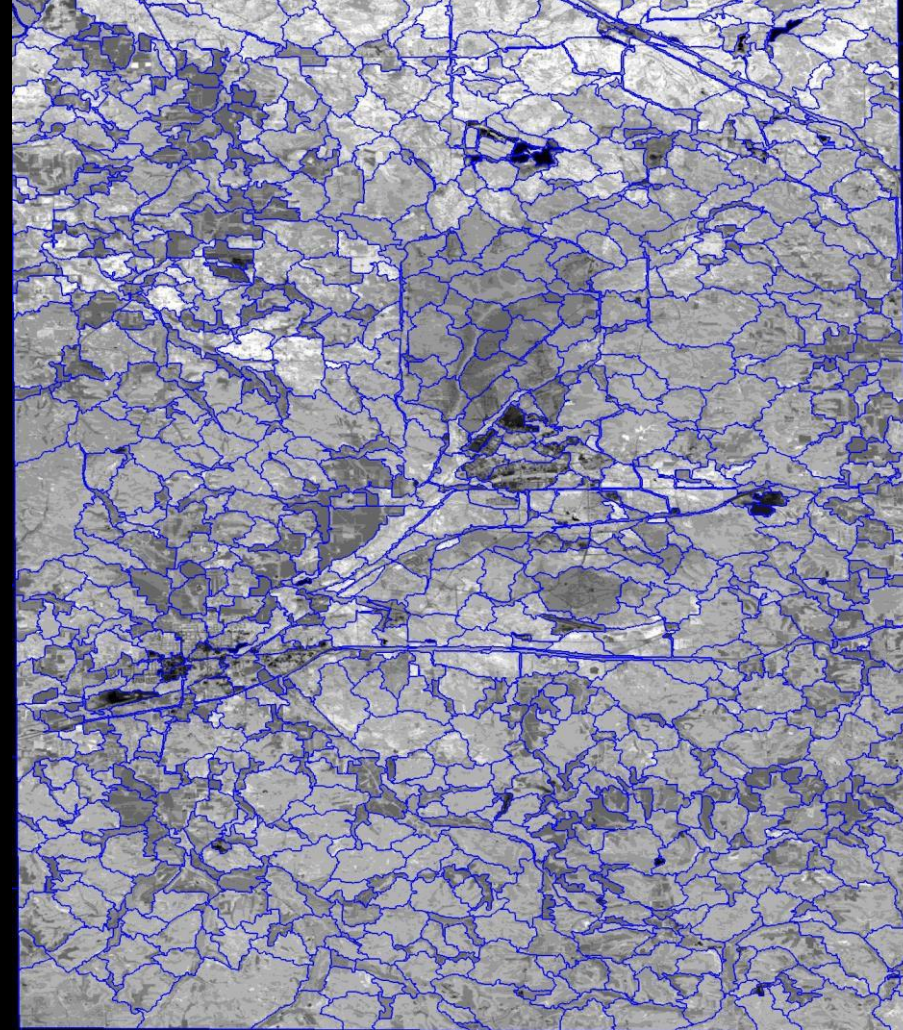


Results

Landsat 8 - Scale - 5



Landsat 8 - Scale - 10



Outline

- Part I: Multi-grain habitat models
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Conclusions

Habitat selection in birds is:

- Hierarchical
- Multi-scale
- Variable among species



Conclusions

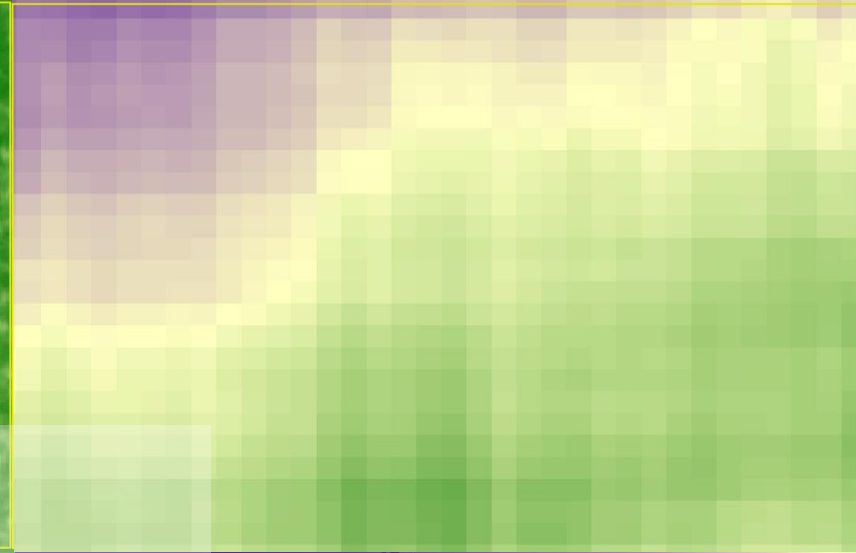
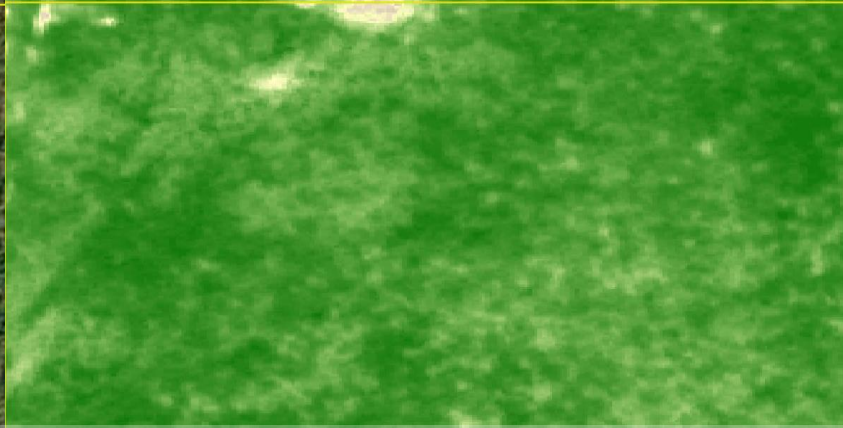
Habitat selection in birds is:

- Hierarchical
- Multi-scale
- Variable among species

Our habitat models should be too



THANK YOU!!!



radeloff@wisc.edu

<http://silvis.forest.wisc.edu>

