

# National and regional BMSB models and factors affecting presence/absence and population abundance

Dave W. Crowder & Javier G. Illan  
Washington State University



*Funding*

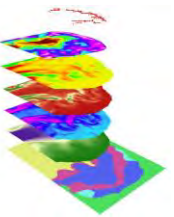
**USDA** United States Department of Agriculture National Institute of Food and Agriculture  
Specialty Crop Research Initiative

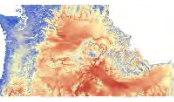


*Collaborating Institutions*

**OSU** Oregon State University  
**NC STATE UNIVERSITY**  
**PennState**  
**UNIVERSITY OF MARYLAND**  
**UNIVERSITY OF GEORGIA**  
**WASHINGTON STATE UNIVERSITY**  
**Cornell University**  
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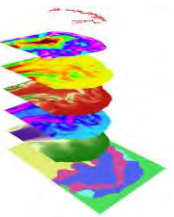
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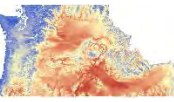


# Questions/Main aims

-  Can we effectively predict the risk of invasion (establishment) and abundance (impact) of BMSB using ecological modelling?

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 What are the most important environmental factors driving BMSB occurrence and abundance?

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# Team

- 40+ scientists from 25 research groups across 17 states
- The most responsive team we've ever worked with! Thank you all

# Strengths

- Our nation-wide BMSB monitoring scheme represent a very rare and ideal dataset to tackle these questions
- Novel study: SDMs are rarely applied to non-natural systems or agricultural insect pests

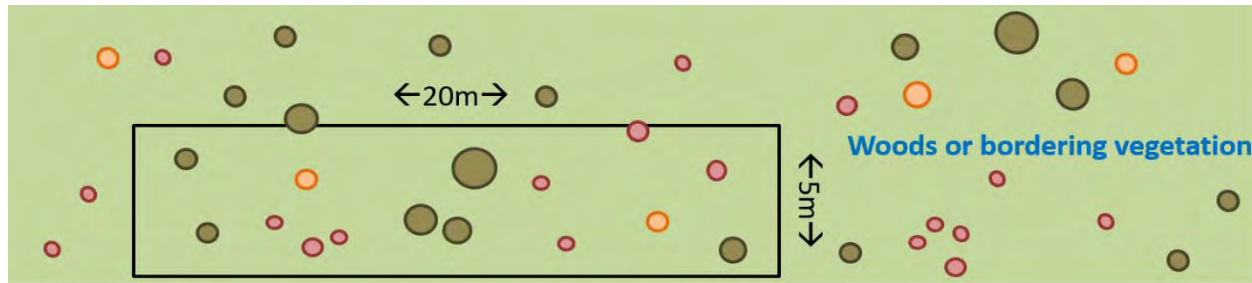
# Sampling protocol

## **BMSB sampling:**

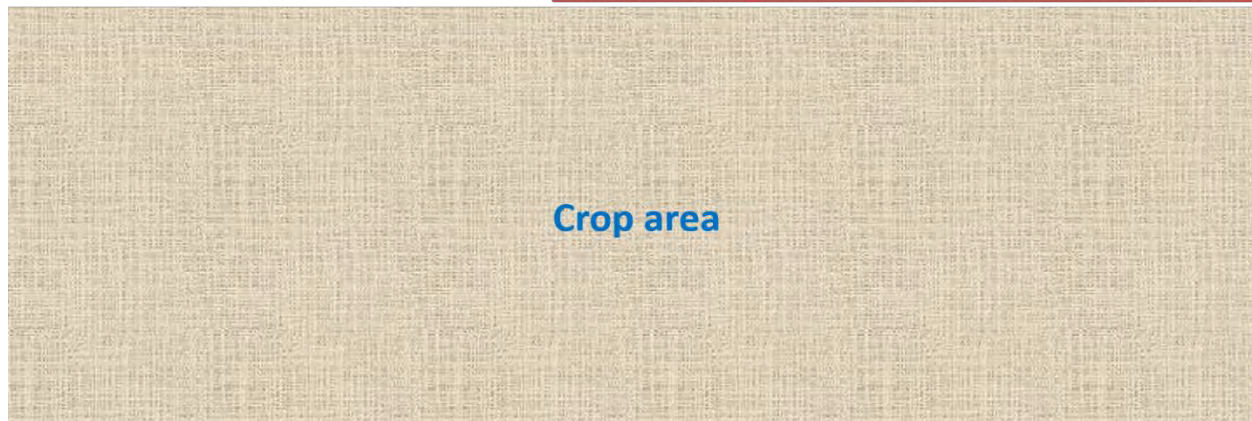
- Unique trapping location
- 3 traps per site (<50m apart)
- Interface of natural habitats and host crops
- Sampled from early spring into the fall



# Sampling protocol



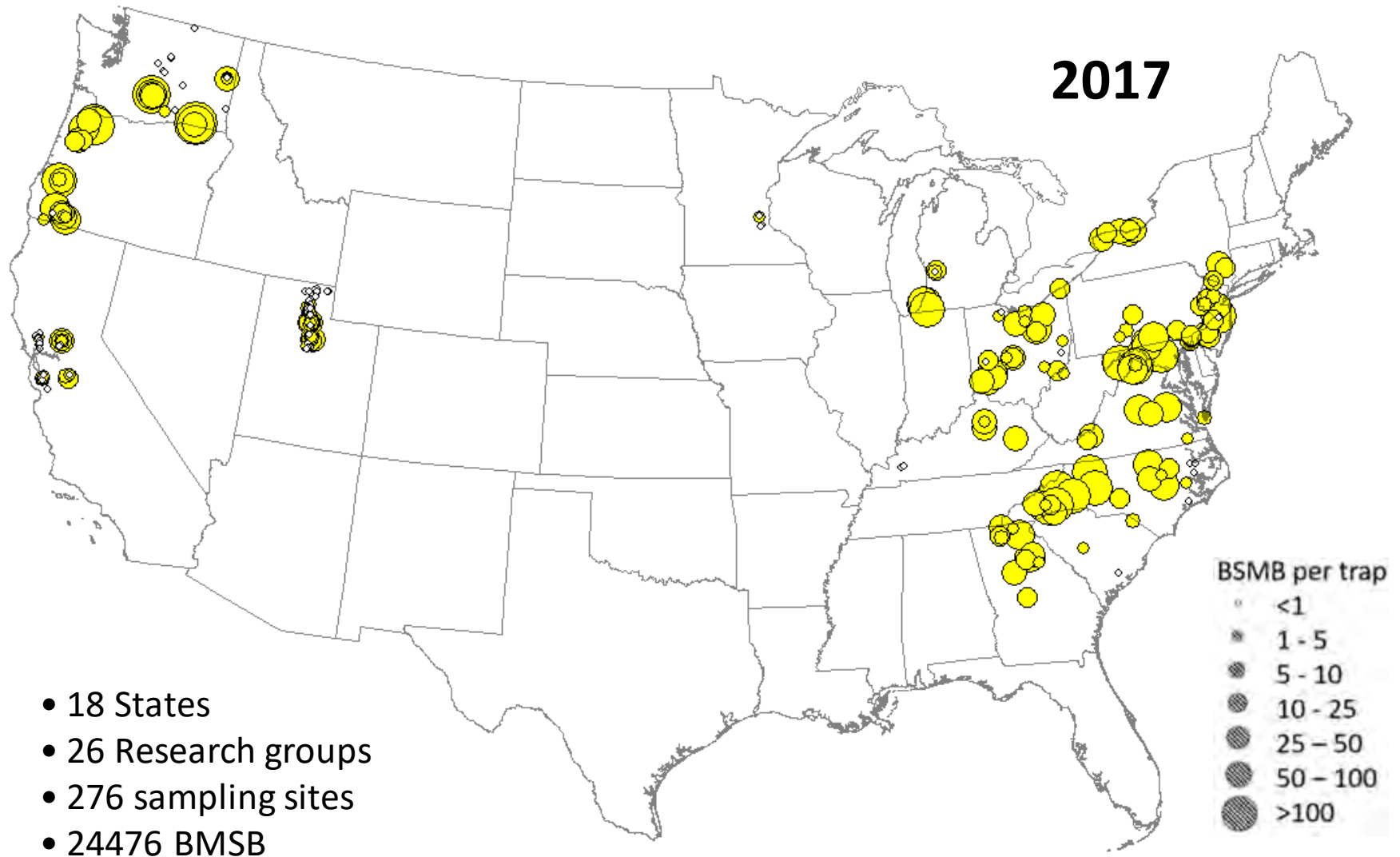
Example of vegetation assessment for BSMB trapping



## Vegetation sampling:

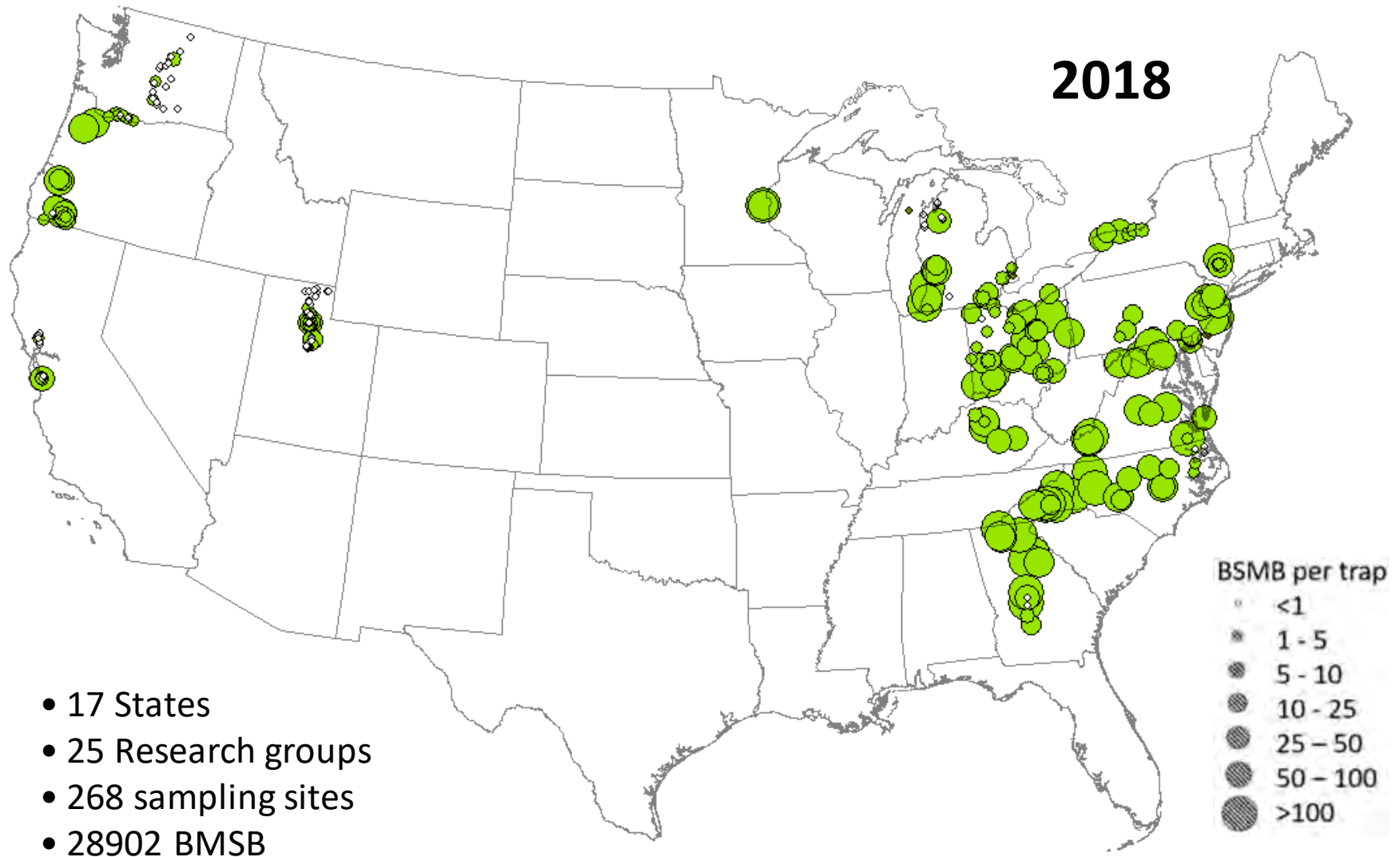
- Presence and abundance of host plants adjacent to traps
- 20 meter (65 ft) length of woods adjacent to each trap
- Once or twice during summer months

# Monitoring Network System

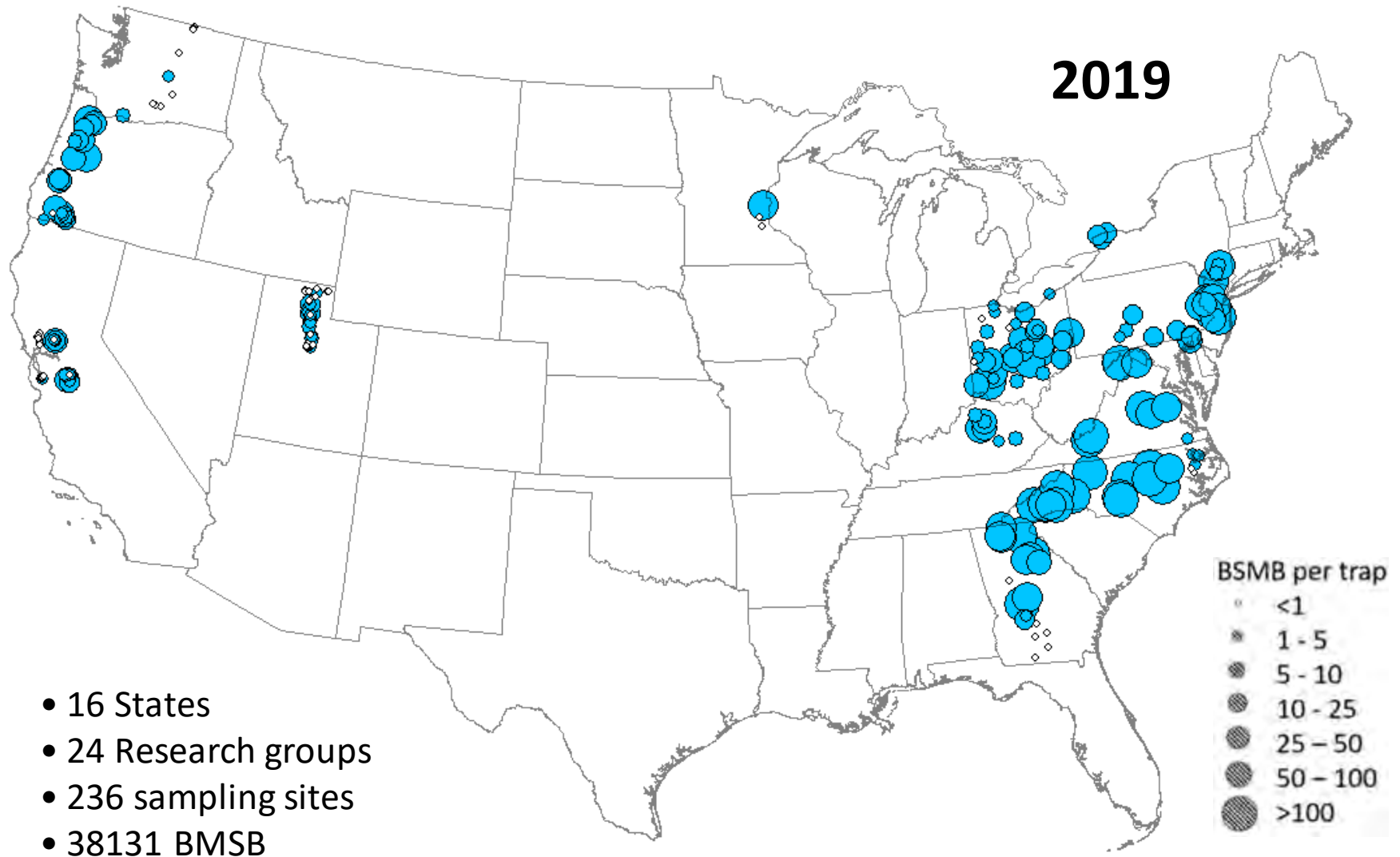


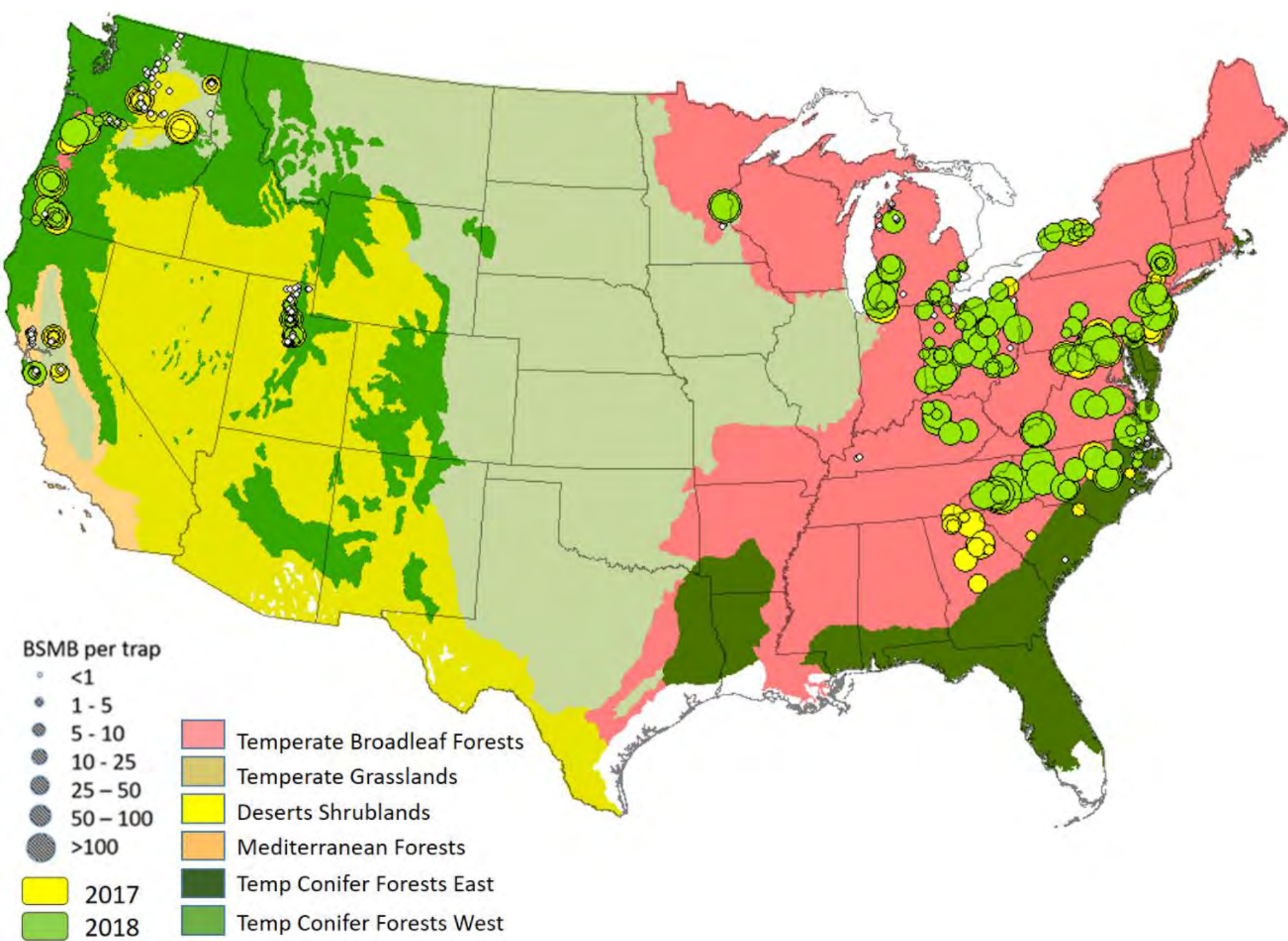


# Monitoring Network System



# Monitoring Network System





# North-East

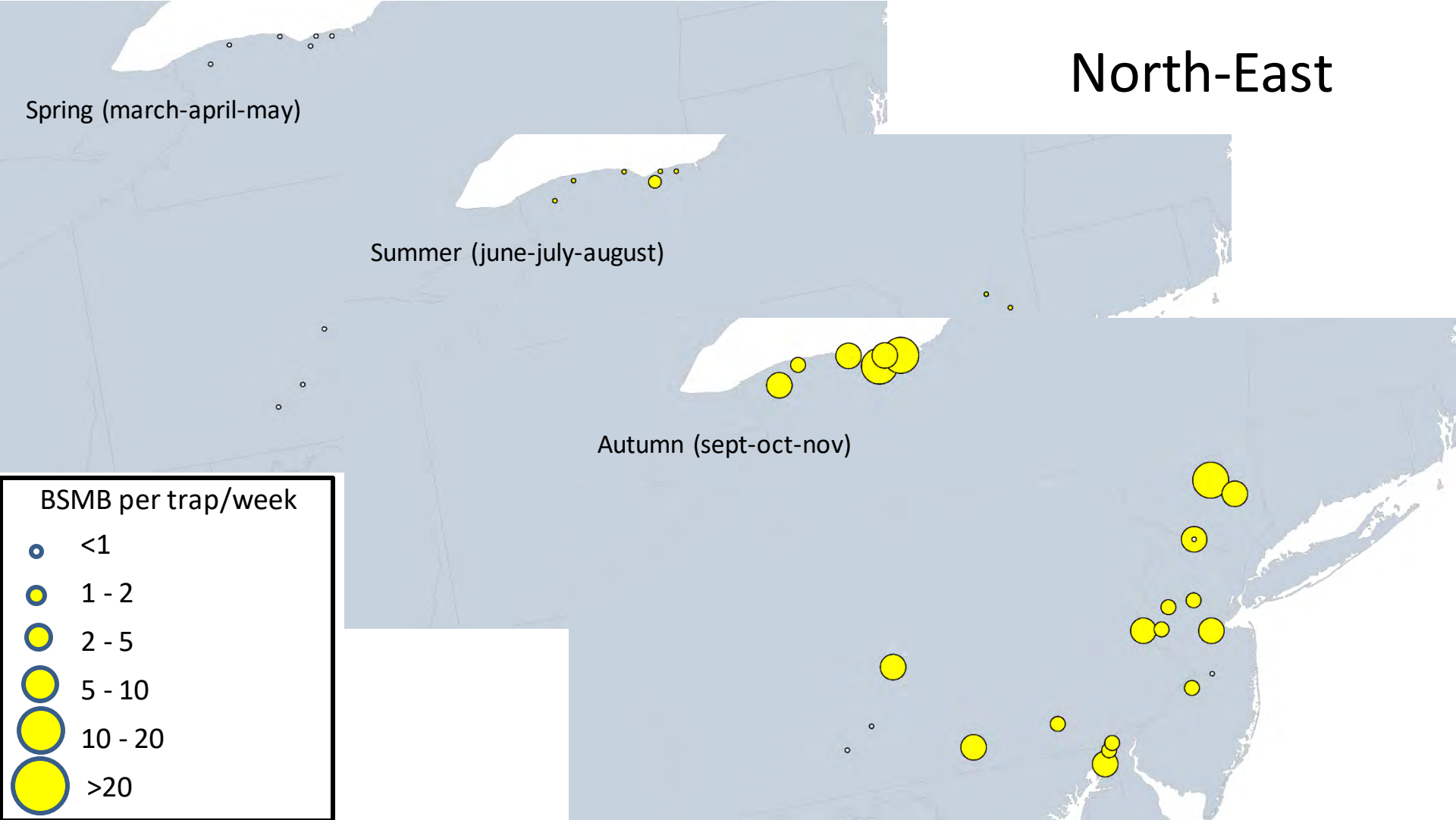
Spring (march-april-may)

Summer (june-july-august)

Autumn (sept-oct-nov)

BSMB per trap/week

- <1
- 1 - 2
- 2 - 5
- 5 - 10
- 10 - 20
- >20

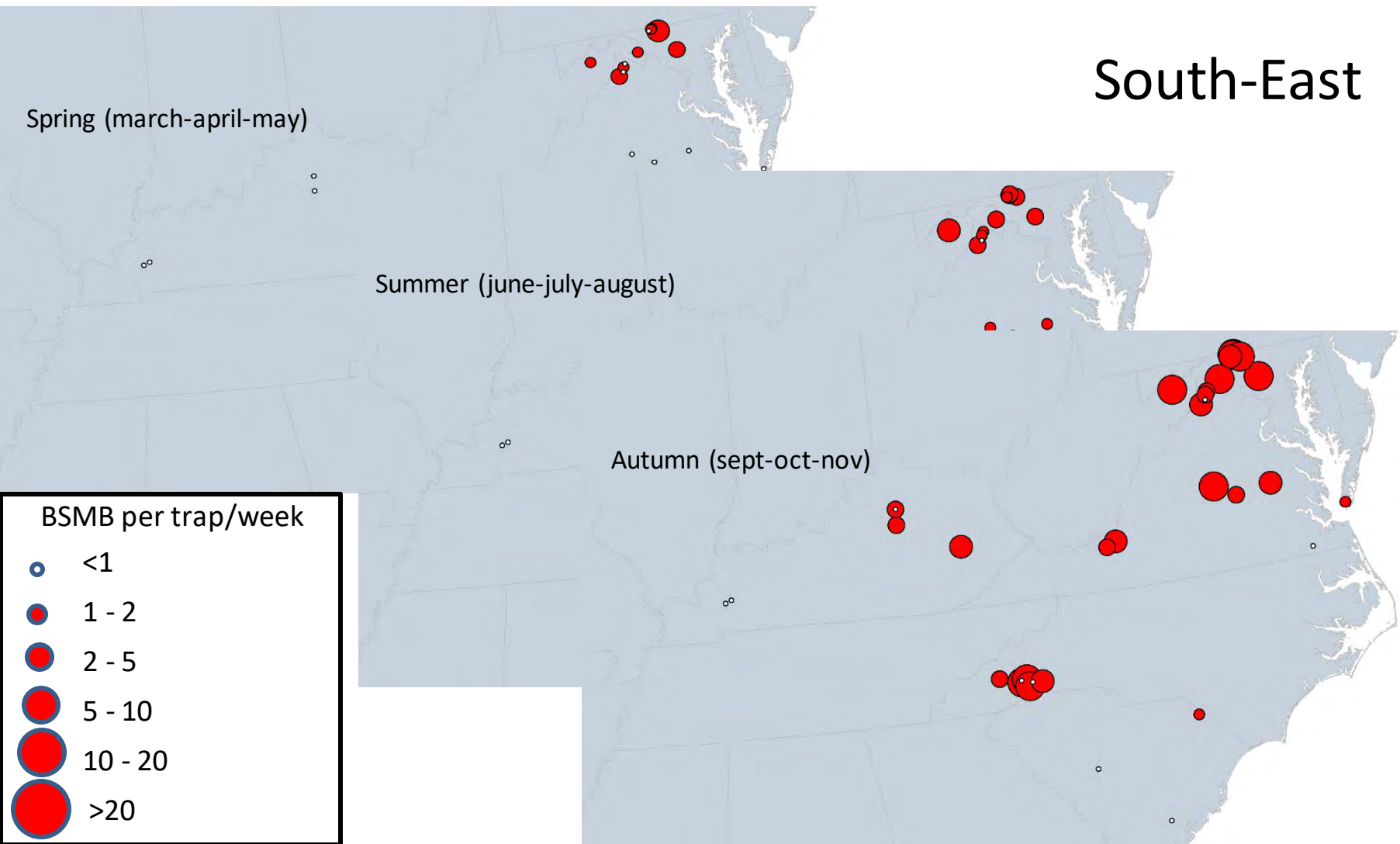
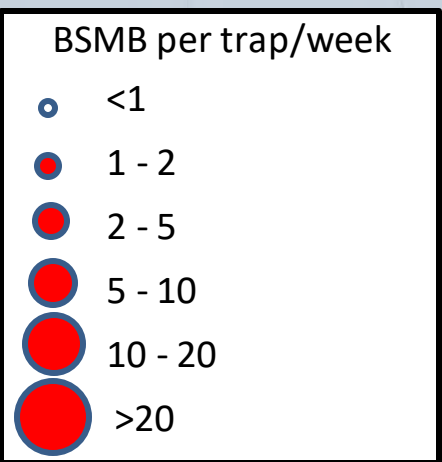


# South-East

Spring (march-april-may)

Summer (june-july-august)

Autumn (sept-oct-nov)

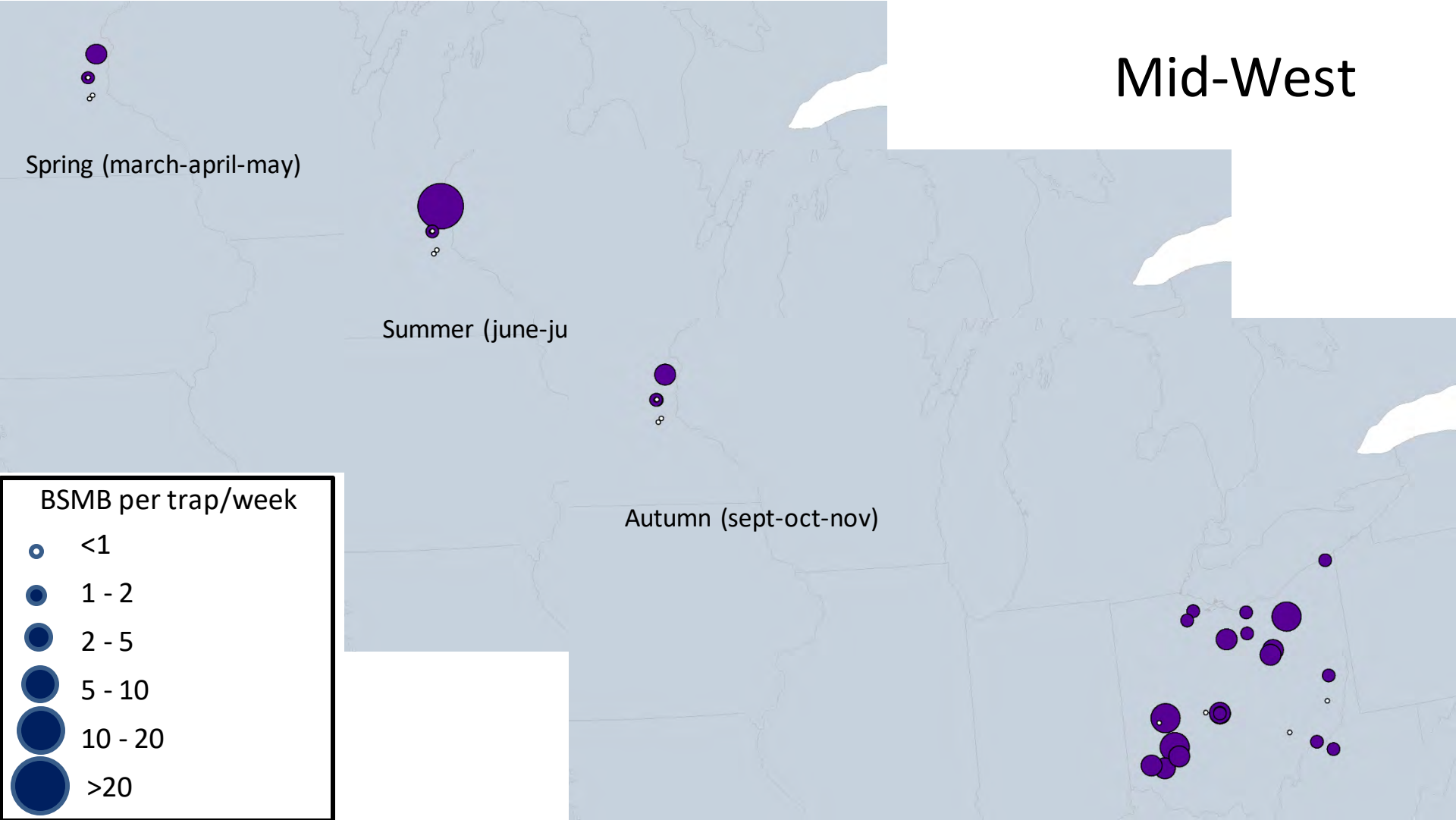
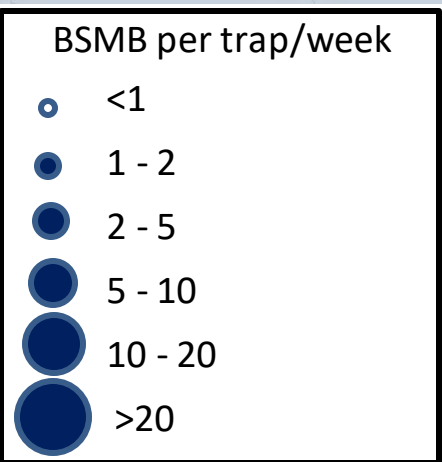


# Mid-West

Spring (march-april-may)

Summer (june-ju

Autumn (sept-oct-nov)



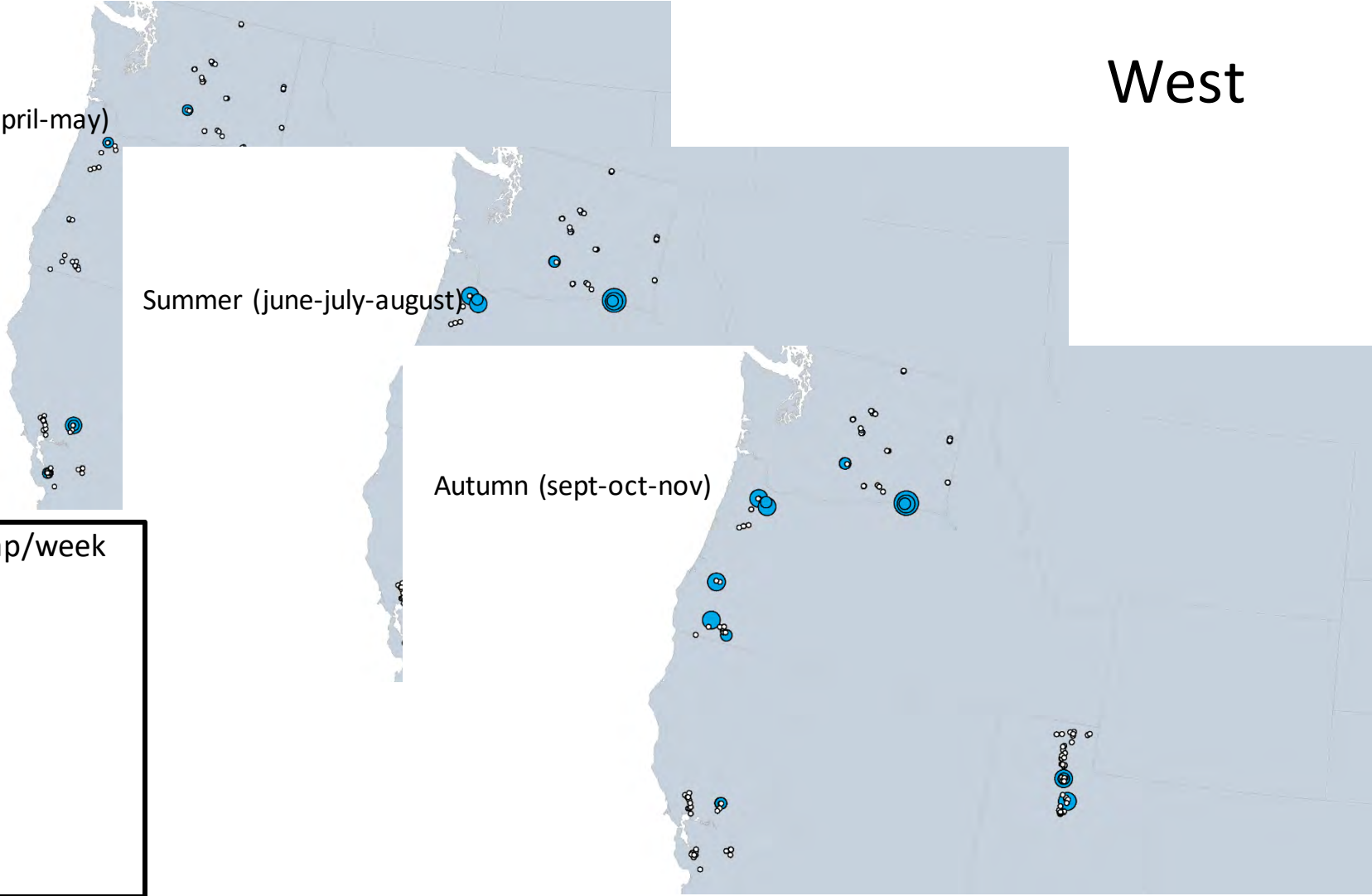
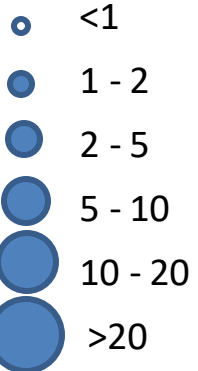
# West

Spring (march-april-may)

Summer (june-july-august)

Autumn (sept-oct-nov)

BSMB per trap/week

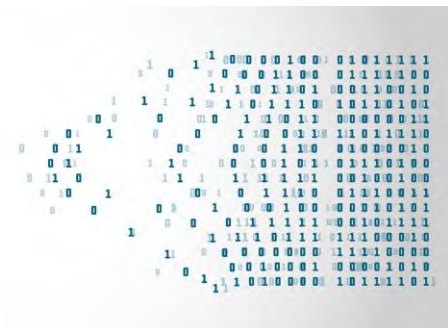


# Modeling approach:

- Occurrence models (establishment): **MAXENT**



- Abundance models (impact): **Boosted regression trees** implemented in “gbm” R package.





# Predictors

## a) Climatic (PRISM)

- Max temp summer
- Min temp winter
- Precipitation
- Vapor pressure deficit
- Evapotranspiration
- Growing degree days



# Predictors

## a) **Landscape**

- Land-cover
- Distance to water
- Soil ph
- photoperiod

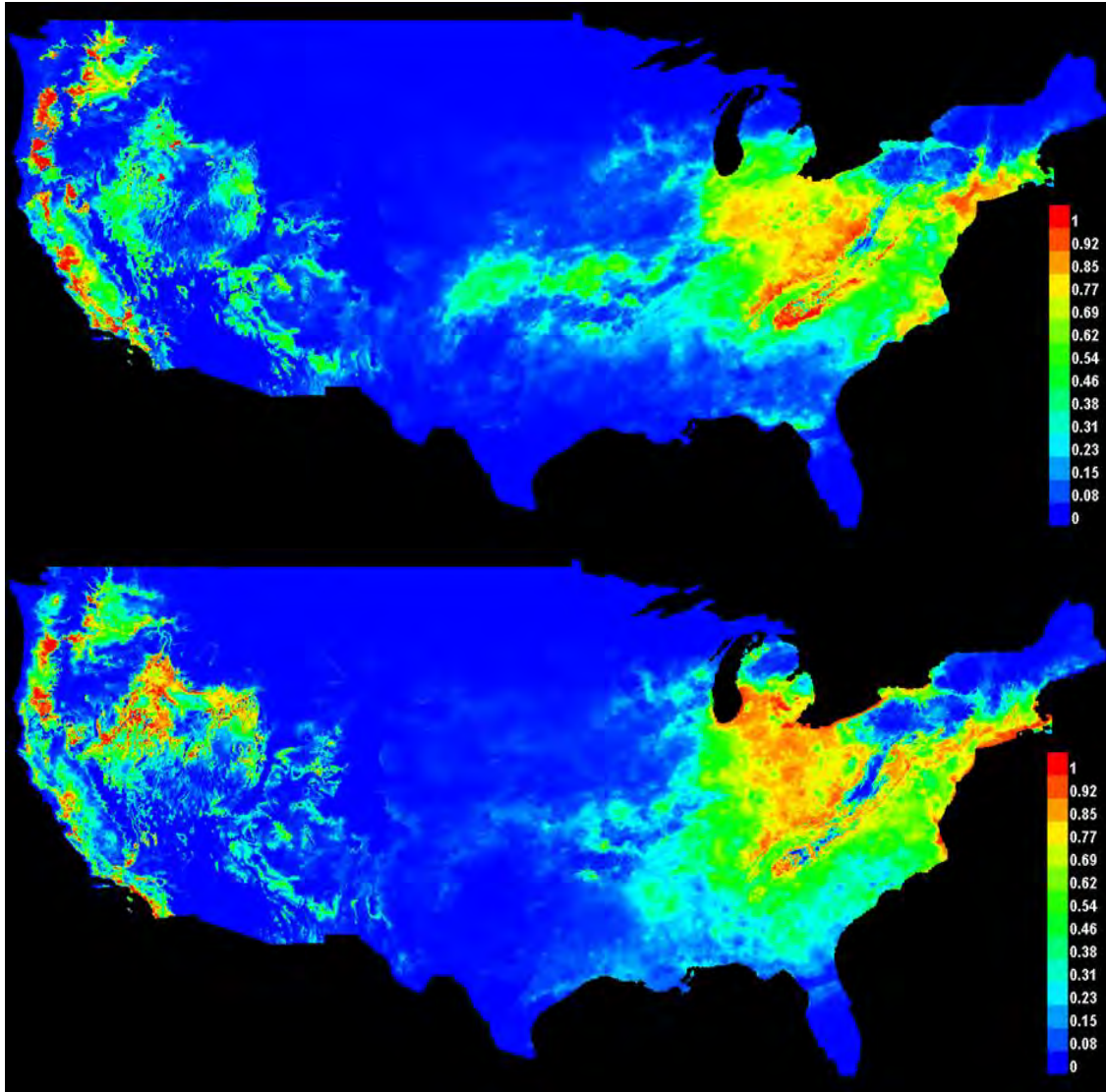


## Climatic and landscape variability (among sampling sites)

	2017	2018	2019
Minimum tp winter (°C)	-2.5 (-14.6 - 6.1)	-4.5 (-16.1 - 7.4)	-2.8 (-16.3 - 6.8)
Maximum tp summer (°C)	31.2 (25.9 - 36.7)	30.7 (26.2 - 35.2)	30.7 (23.3 - 34.3)
Precipitation (mm)	134.5 (23.7 - 544.7)	67.2 (15.6 - 267.9)	103.3 (11.4 - 326.0)
Elevational range (m)	476.7 (2.1 -1845)	499.5 (2.1 -1845.2)	429.6 (0.5 - 1934)
Land-use classes	27	28	26
Solar_photoperiod	1345.4 (1.2 - 2764.4)	1316.7 (2.8 - 2882.2)	1316.3 (0.2 - 2764.4)
Evapotranspiration	43.1 (16.3 - 73.5)	44.4 (14.5 - 71.9)	45.0 (14.5 - 80.0)
Soil ph	6.2 (5.0 - 8.1)	6.2 (5.0 - 8.1)	6.1 (5.0 - 8.1)
Distance to water	112 (1 - 504)	109 (1 - 498)	112 (1 - 438)

# Results

## Presence (establishment) models



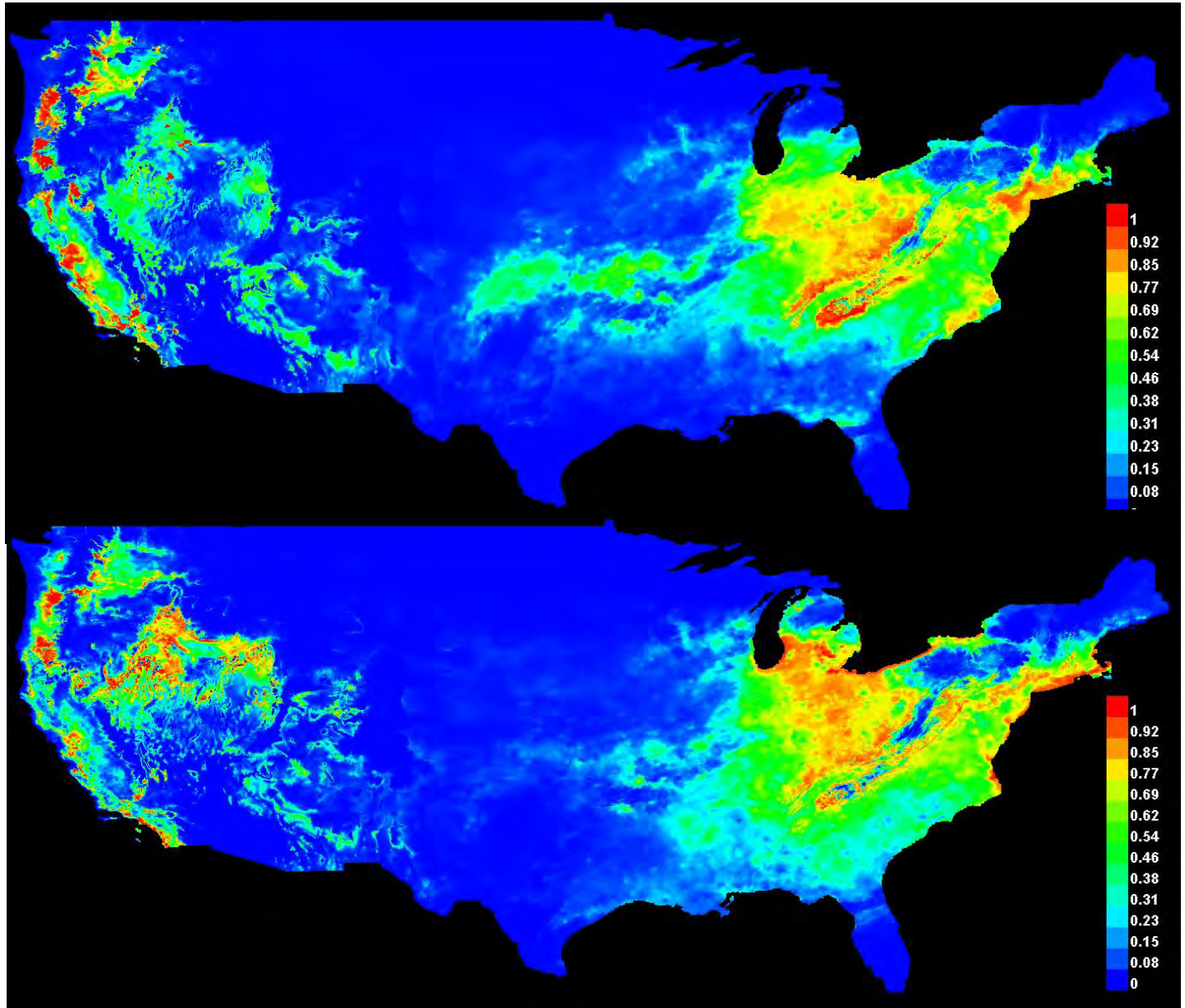
### Top climatic predictors:

- winter precipitation
- minimum tp winter
- growing degree days

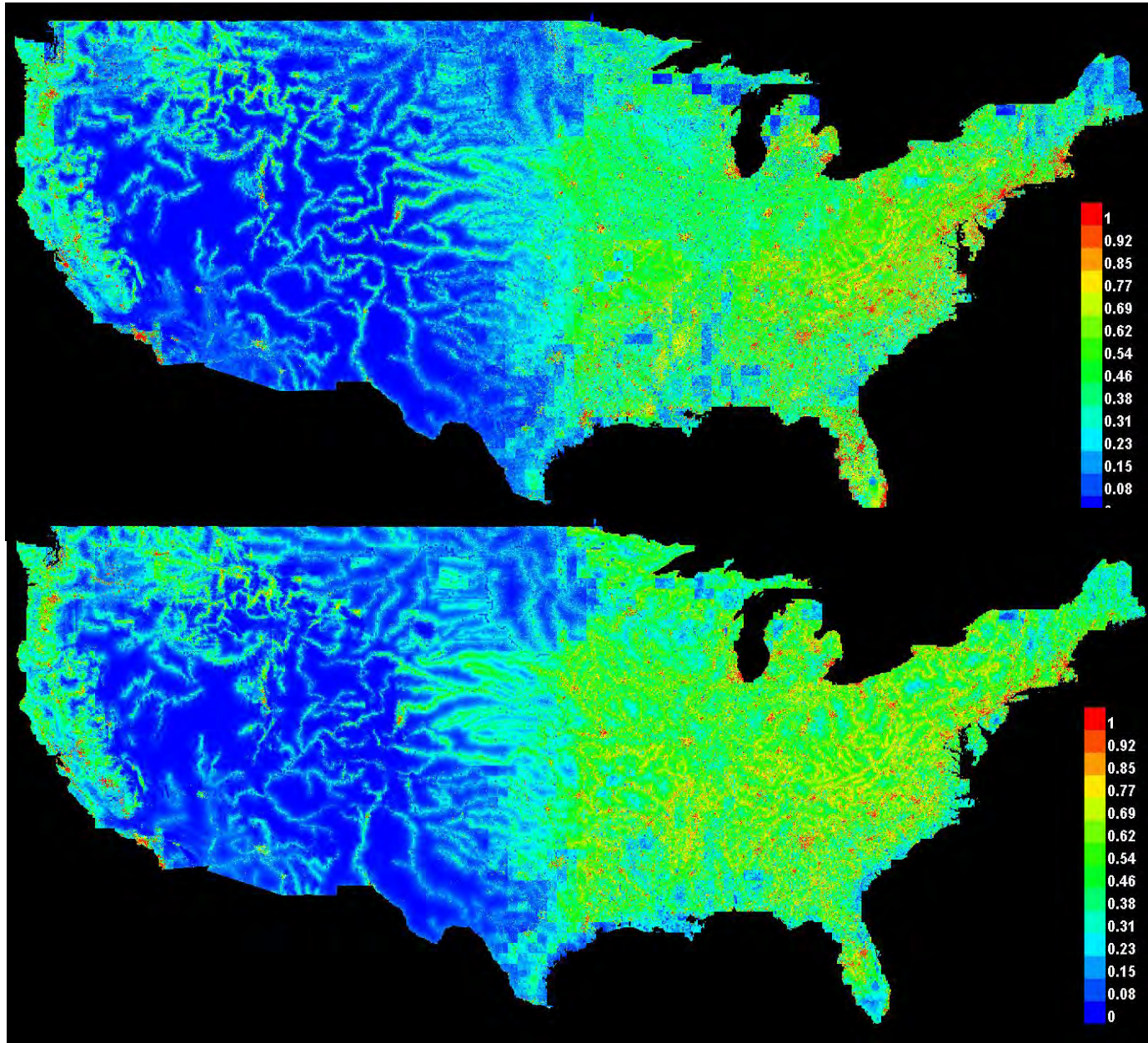
### Top landscape predictors:

- land use
- soil ph
- solar photoperiod

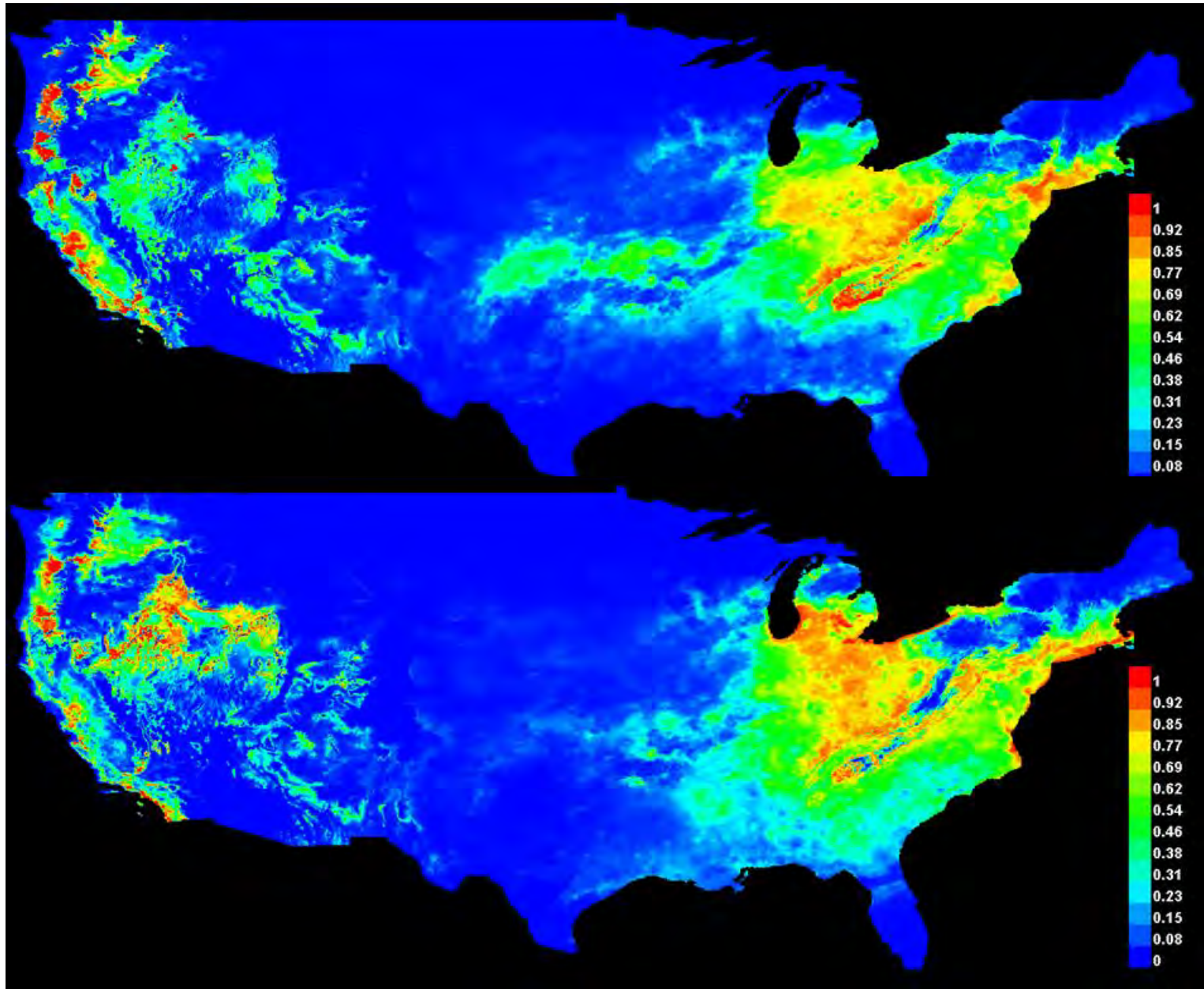
# Results – Climate only



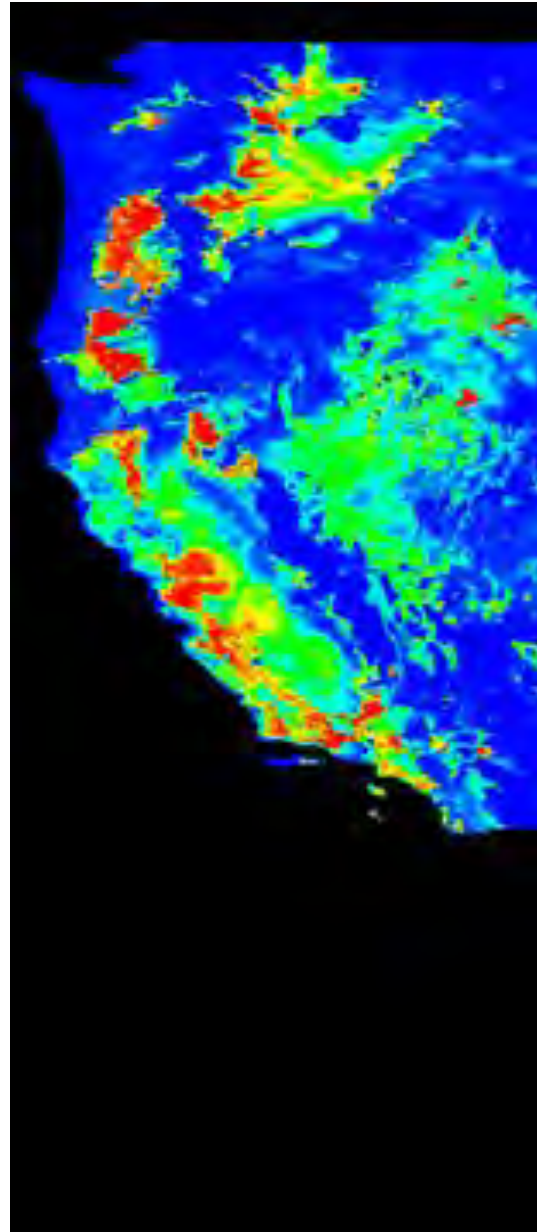
# Results – Landscape only



# Results – Combined

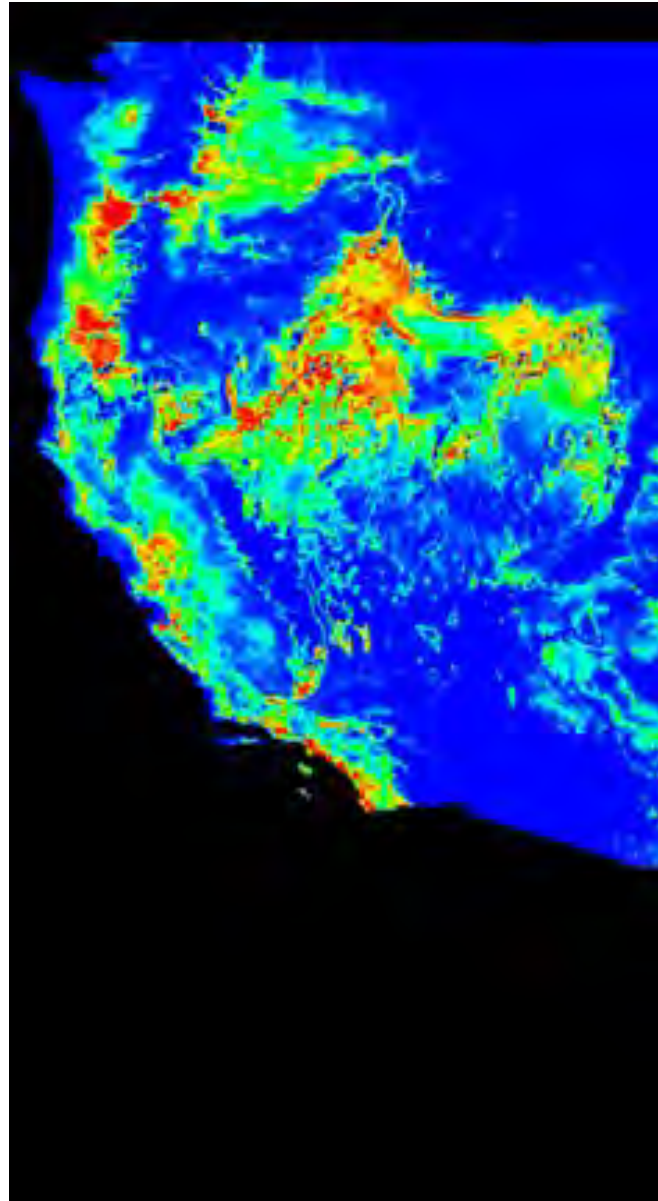


# Results – West Coast 2017





# Results – West Coast 2018



# Results

## Abundance (impact) models

Correlation observed vs predicted abundance:

- Climatic models: Spearman's  $\rho = 0.457$ ;  $p < 0.001$
- Climatic models: Spearman's  $\rho = 0.522$ ;  $p < 0.001$

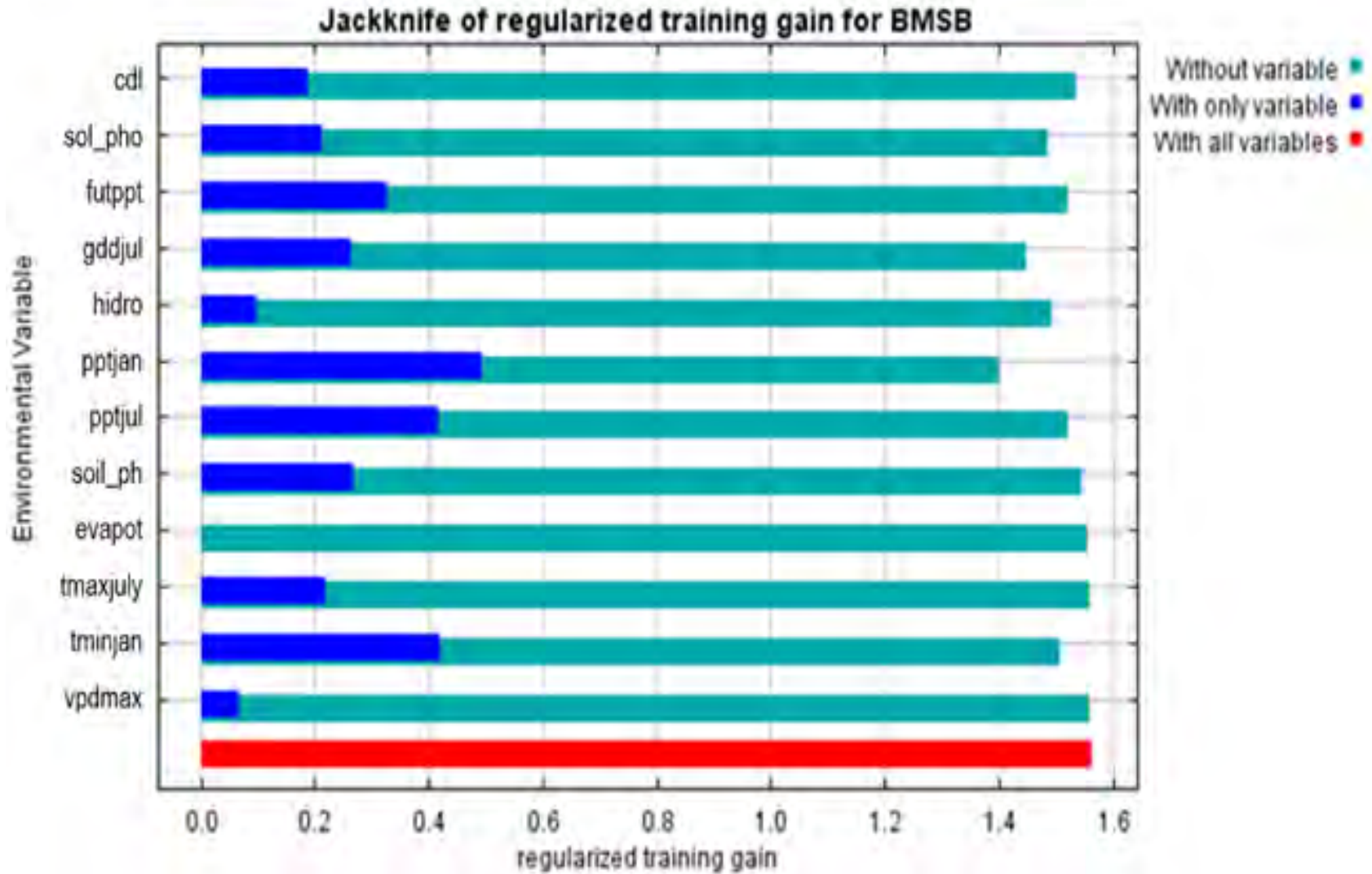
### Top climatic predictors:

- minimum tp winter
- evapotranspiration
- summer precipitation

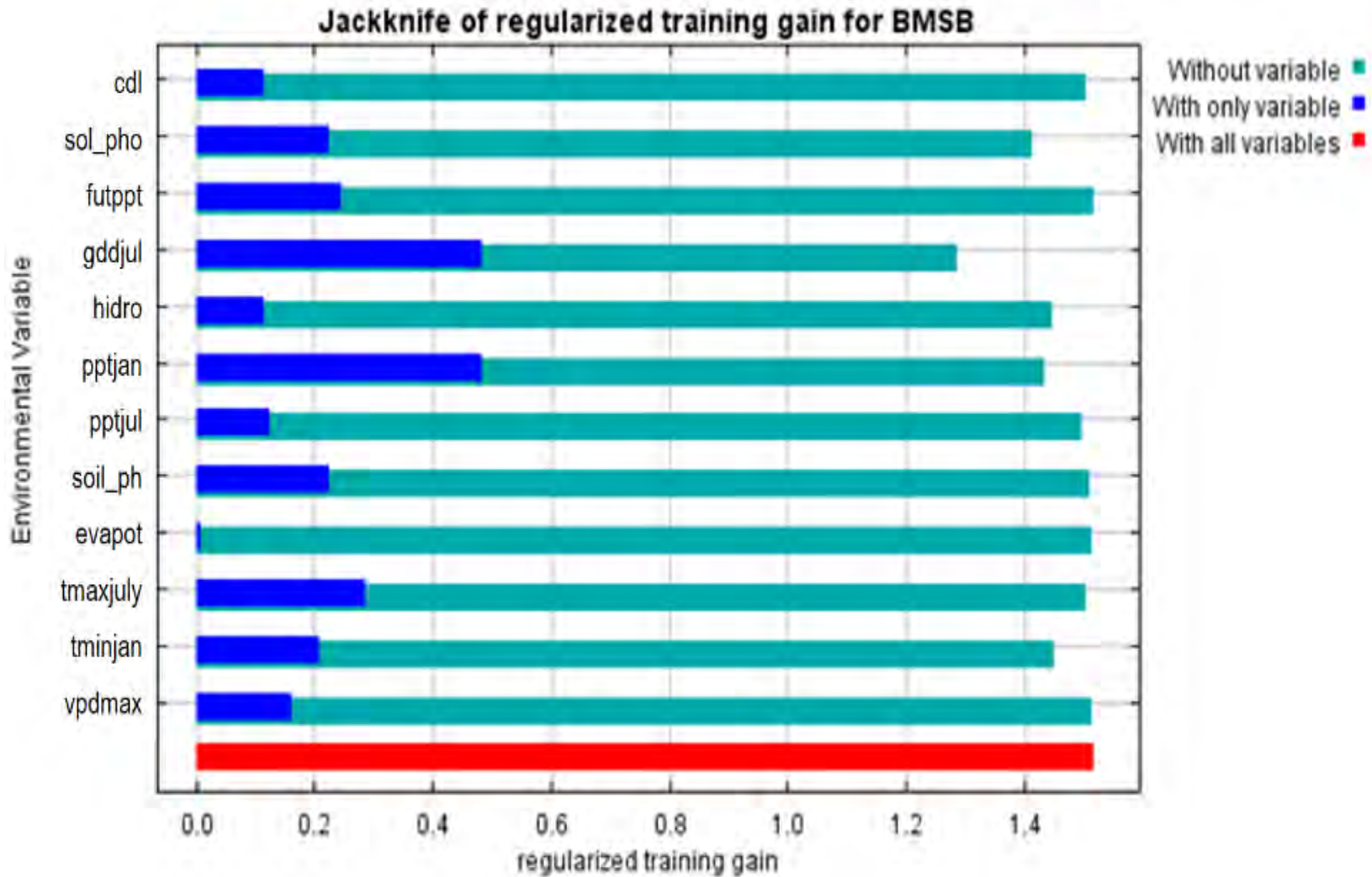
### Top landscape predictors:

- solar photoperiod
- soil ph
- distance to water

# Results - 2017



# Results - 2018



# Conclusions

- Our Ecological niche models can accurately describe both the occurrence and the abundance of BMSB.
- It appears that, while winter conditions are more important for BMSB presence, factors like photoperiod and evapotranspiration are driving BMSB abundance. In both cases, minimum temperature in winter seems to be a limiting factor.
- We have detected some areas environmentally suited for BMSB where the species has not yet been detected. Those are where new monitoring effort should concentrate.

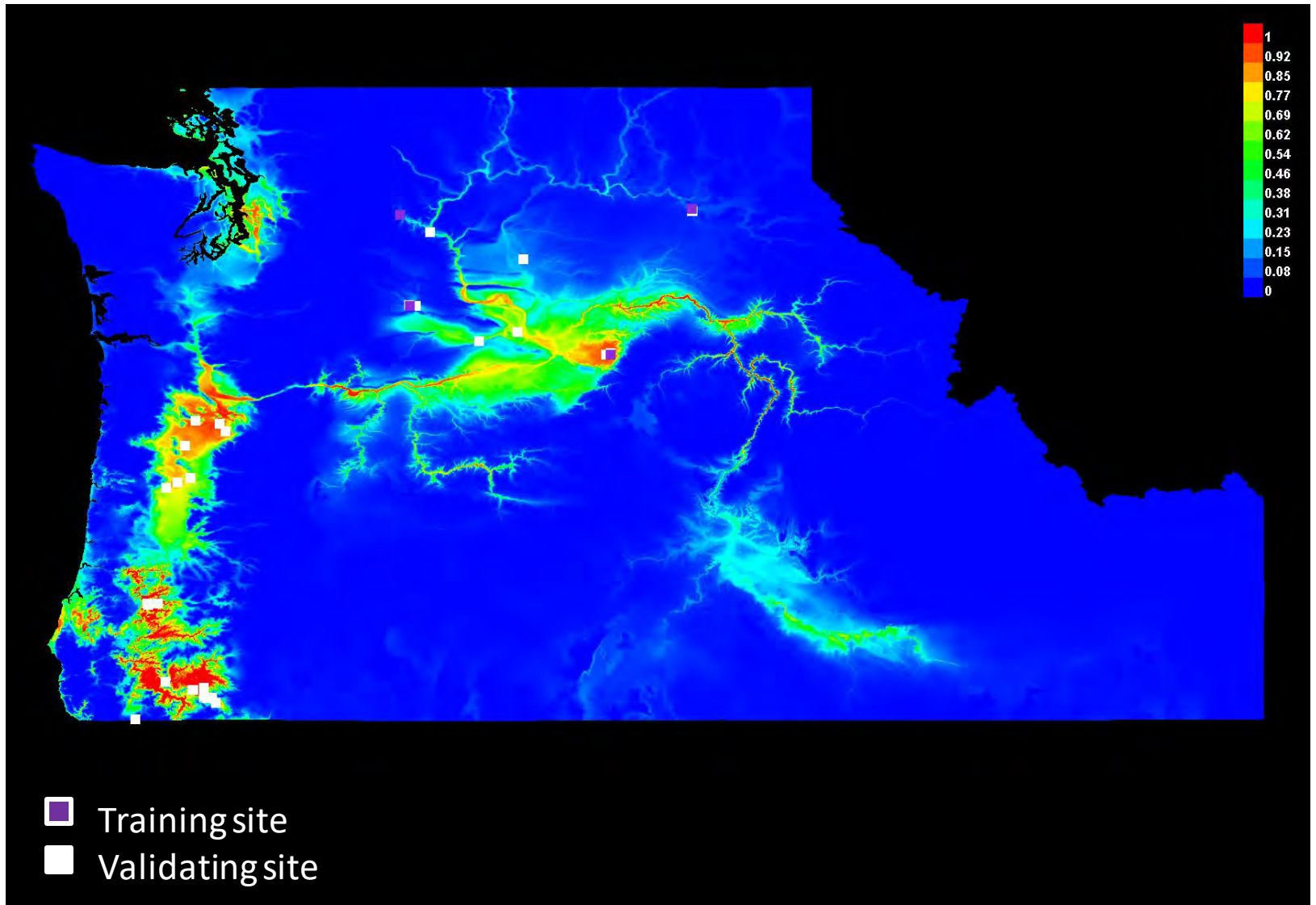
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## Example of smaller scale (regional) model



BMSB risk of invasion (Climate) 2017/2018 in the PNW



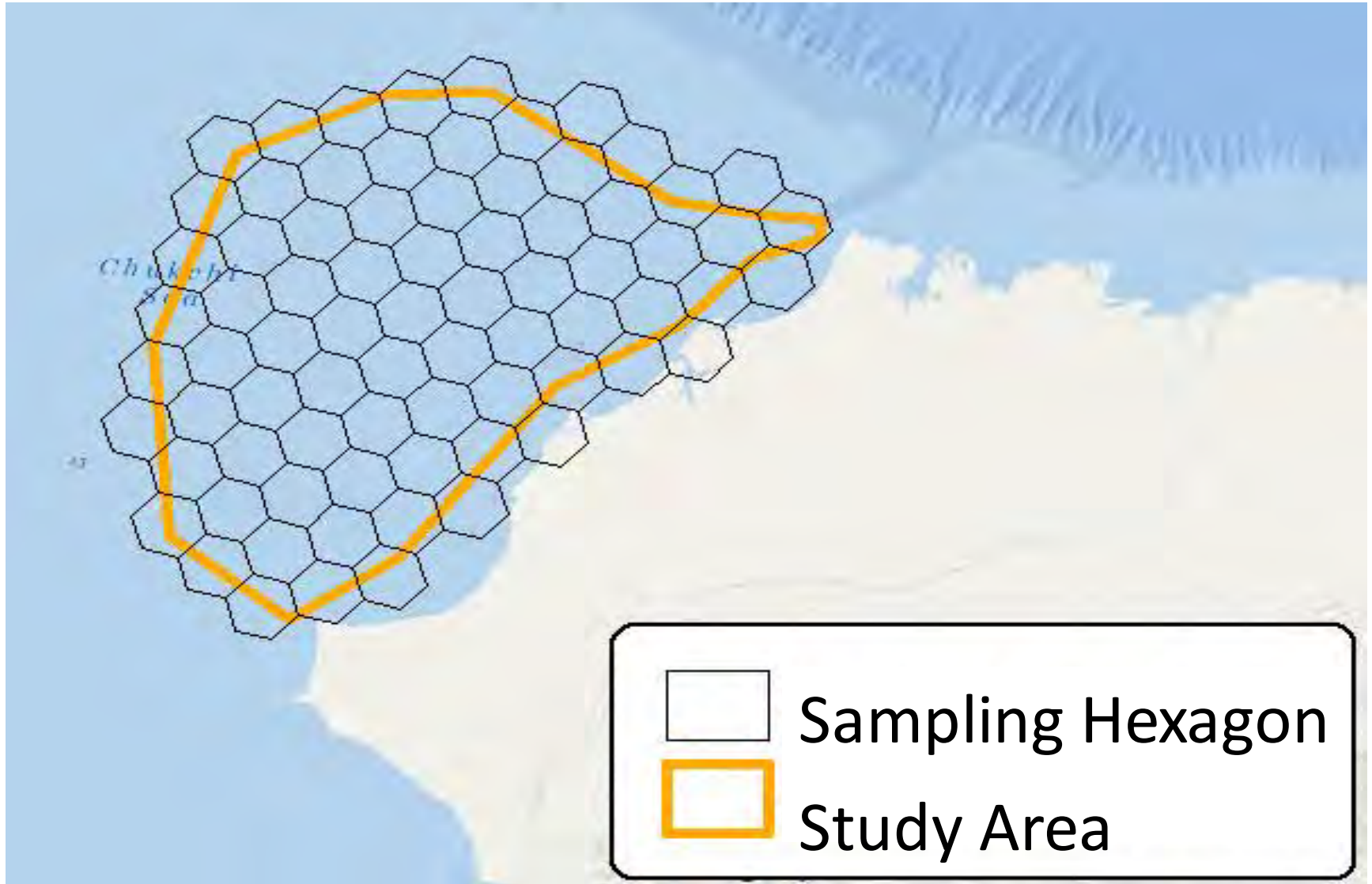
# Caveats and further research...

- Results presented here are solely based on adults. Data for larvae instars are available for modelling.
- After previous meetings with participants, new predictors will be considered (e.g. distance to urban areas).
- The temporal aspect has not yet been considered. Phenology models are the next critical step.
- We will calculate predicted change in suitable habitat based in IPCC future climatic scenarios (e.g. moderate climatic scenario: 1.5 °C and 10% ppt).

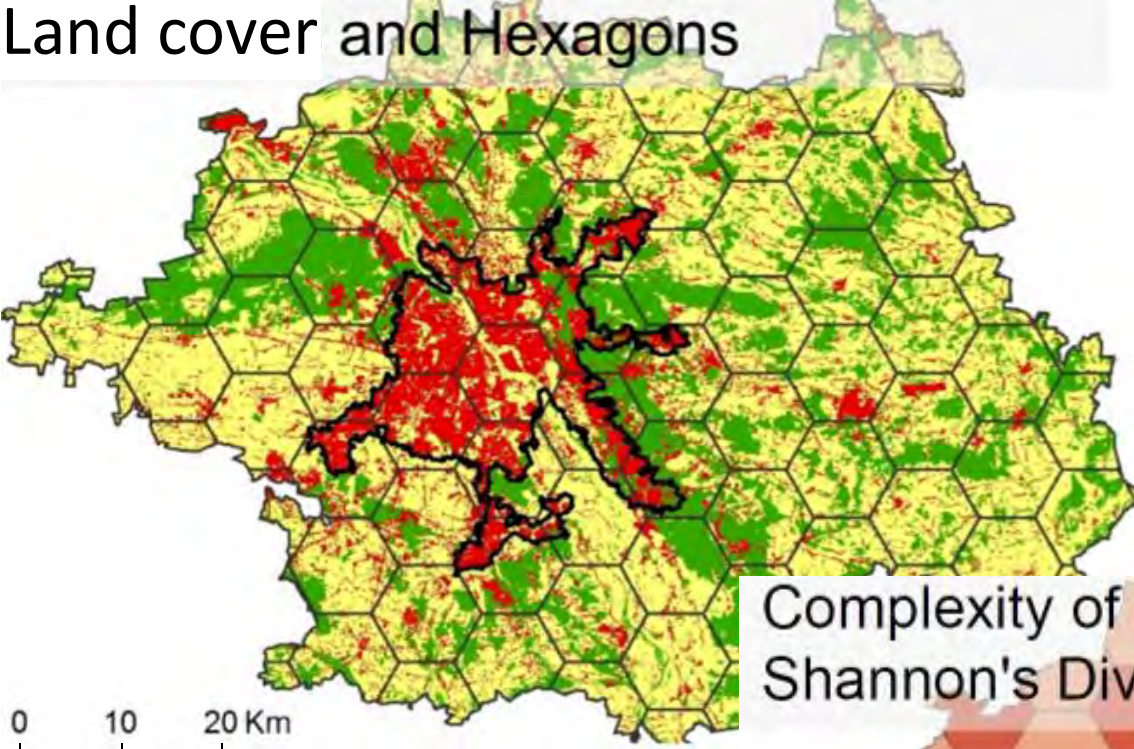
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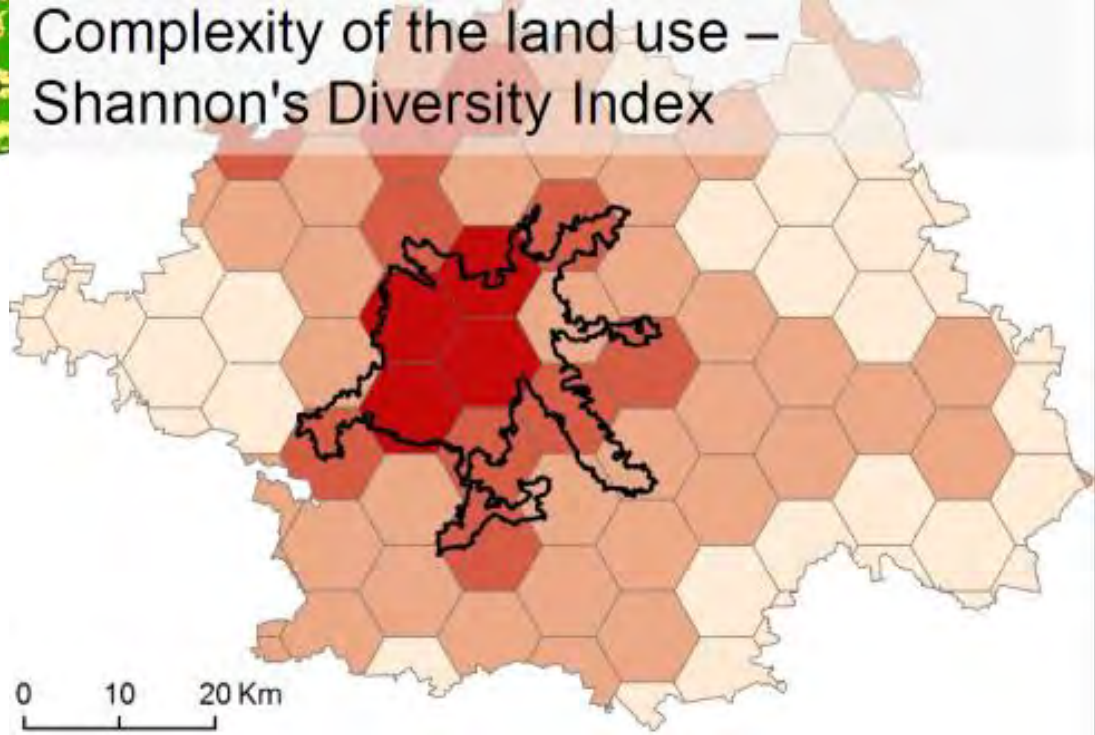
- Landscape structure models are being developed (diversity, fragmentation, connectivity).



# Land cover and Hexagons



# Complexity of the land use – Shannon's Diversity Index



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# Further research (*Nosema maddoxi*)...

State	% infected	# high infection	# low infection	# uninfected	# total	Degrees W	Degrees N
DE	0.00%	0	0	25	25	-75.742267	39.667969
GA	0.00%	0	0	16	16	-83.738785	32.658514
KY	20.00%	1	5	25	30	-84.536792	38.013029
MD	40.00%	5	3	12	20	-76.75111	38.91111
MD	13.33%	4	0	26	30	-77.43028	39.16556
MD	23.33%	2	5	23	30	-76.132361	39.658991
MI	0.00%	0	0	30	30	-84.788288	42.635395
MI	3.33%	0	1	29	30	-86.513024	42.041258
MI	6.67%	0	2	28	30	-86.381256	41.977244
MI	3.33%	0	1	29	30	-85.789065	42.930614
MN	0.00%	0	0	30	30	-92.995727	45.334428
NC	0.00%	0	0	26	26	-82.9764	35.45366
NC	28.00%	5	2	18	25	-81.047	36.052
NJ	0.00%	0	0	30	30	-75.43337102	39.56705
NJ	13.33%	3	1	26	30	-75.126754	39.615989
NJ	0.00%	0	0	30	30	-74.522188	40.113816
NY	3.33%	0	1	29	30	-73.965735	41.746148
NY	0.00%	0	0	15	15	-76.2074211	42.6041713
NY	0.00%	0	0	30	30	-76.515286	42.541403
OH	10.00%	3	0	27	30	-83.06333	39.98877
OR	0.00%	0	0	25	25	-123.274	44.553
OR	0.00%	0	0	30	30	-122.654448	45.279876
OR	0.00%	0	0	30	30	-122.934975	42.3354722
OR	0.00%	0	0	24	24	-121.5186083	45.6850611
PA	5.00%	0	1	19	20	-77.292468	39.953894
PA	23.33%	4	3	23	30	-77.308069	39.955059
PA	6.67%	0	1	14	15	-76.68038	40.011086
TN	0.00%	0	0	30	30	-88.846325	35.623219
UT	0.00%	0	0	21	21	-111.90633	41.03606
UT	0.00%	0	0	28	28	-112.00829	41.33215
VA	4.00%	0	1	24	25	-78.22492	39.1305
VA	3.33%	1	0	29	30	-80.41778	37.23
WV	19.23%	1	4	21	26	-77.88759	39.35523

# Further research...

- Validating model more completely (2020)

Sampling in novel areas

More intensive sampling in a few regions



# Thank you!



## *Funding*



United States  
Department of  
Agriculture

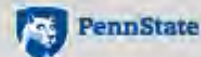
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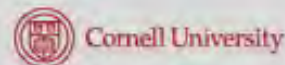


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