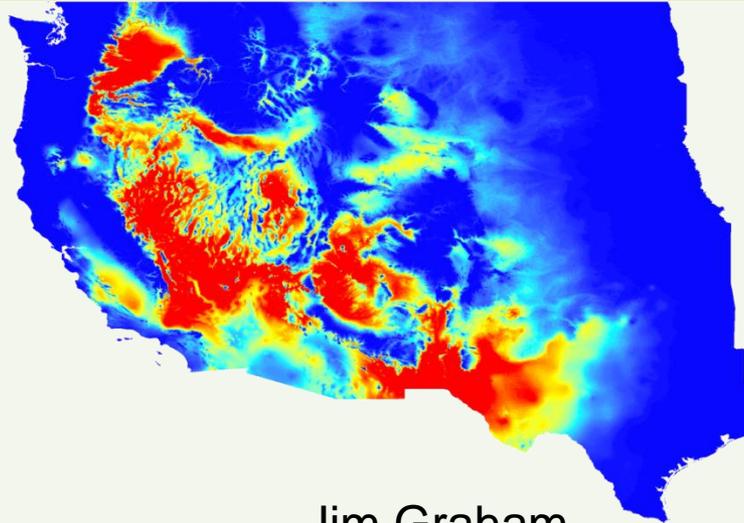


The Hyper-Envelope Modeling Interface (HEMI): A novel approach to habitat suitability modeling



Jim Graham

Catherine Jarnevich, Greg Newman, Nick Young,
Natural Resource Ecology Laboratory

Oregon State University / Colorado State University

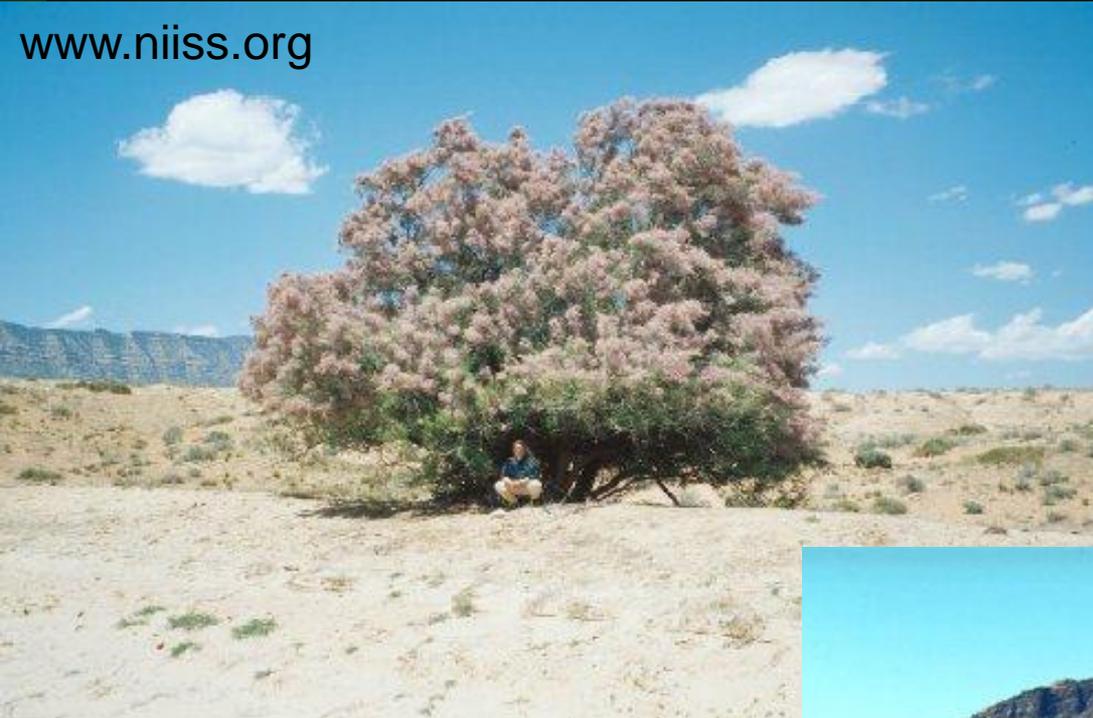


- New approach to Habitat Suitability Modeling, Species Distribution Modeling (SDM)
- Challenge:
 - Predicting potential habitat for a species
 - Over large spatial scales
 - Large datasets with large or undefined uncertainties



Invasive Tamarisk / *Tamarix*

www.niiss.org



Species:

- *T. chinensis*
- *T. ramosissima*
- And hybrids



Bugwood.org



NISS.org: Tamarisk Data

welcome guest | Login | My Profile | October 30, 2011

niiss The National Institute of Invasive Species Science

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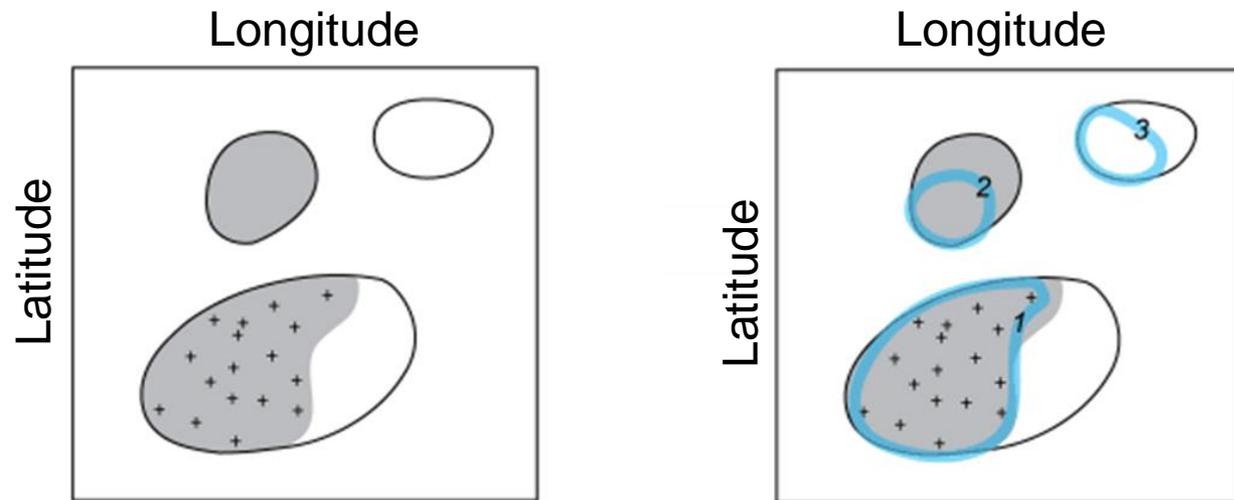
- Plants
 - Tamarisk**
- Backgrounds
 - Google: Terrain
 - Google: Map
 - Google: Satellite
 - Google: Hybrid

0 200 400 600 800 1000 1200 1400 1600 km

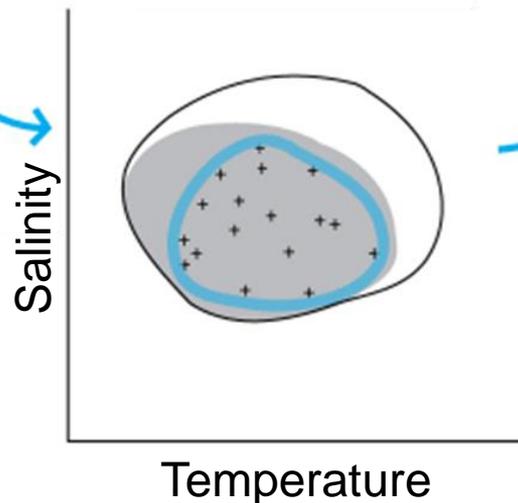
Projection: Google Mercator [Sources](#)

Updated 9/29/2011

Geographical Space



Environmental Space



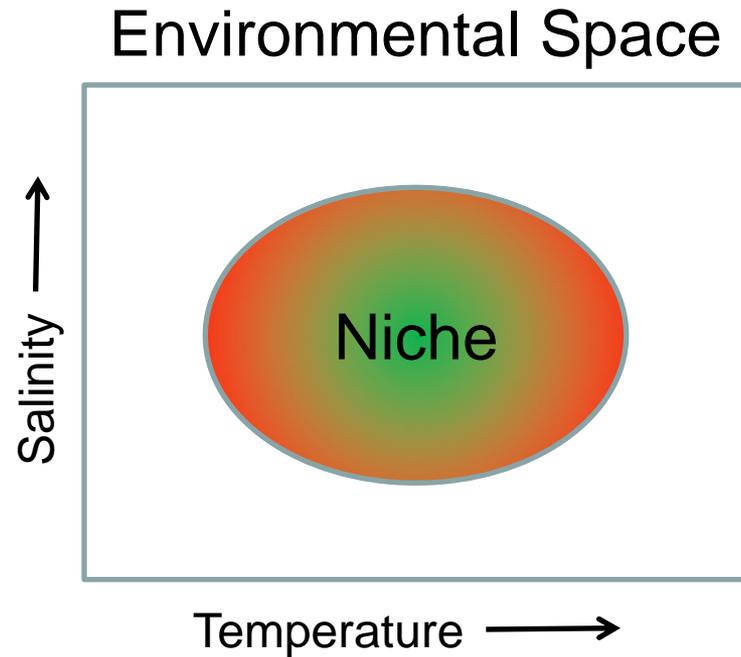
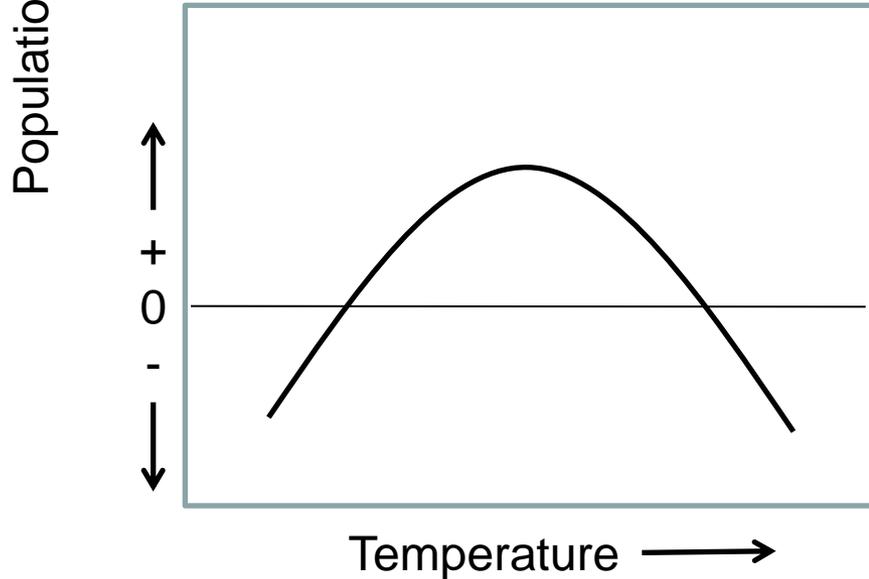
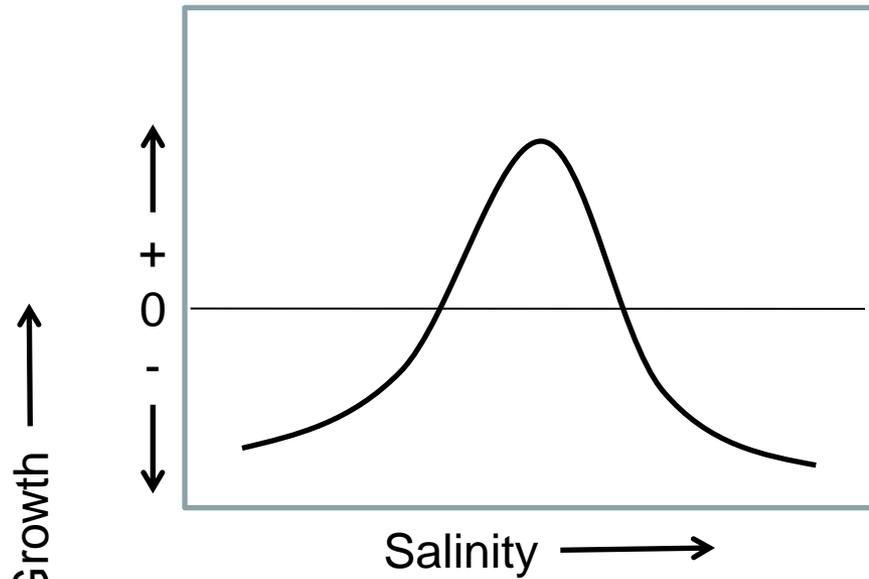
+ Observed Occurrences

● Realized Niche/Current Distribution

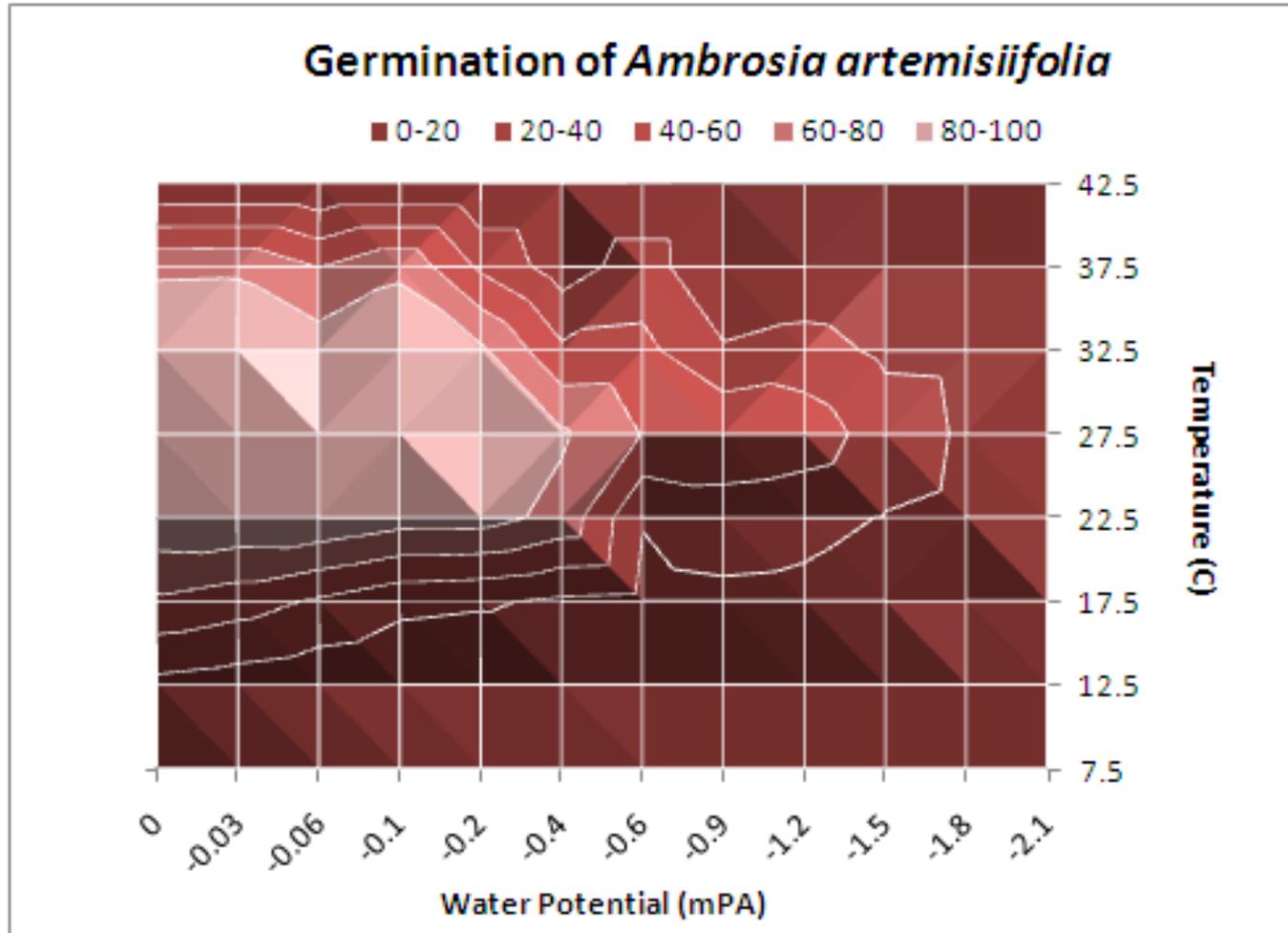
○ Fundamental Niche/Potential Habitat

○ Model Fitted to Occurrences

From the Theory of Biogeography

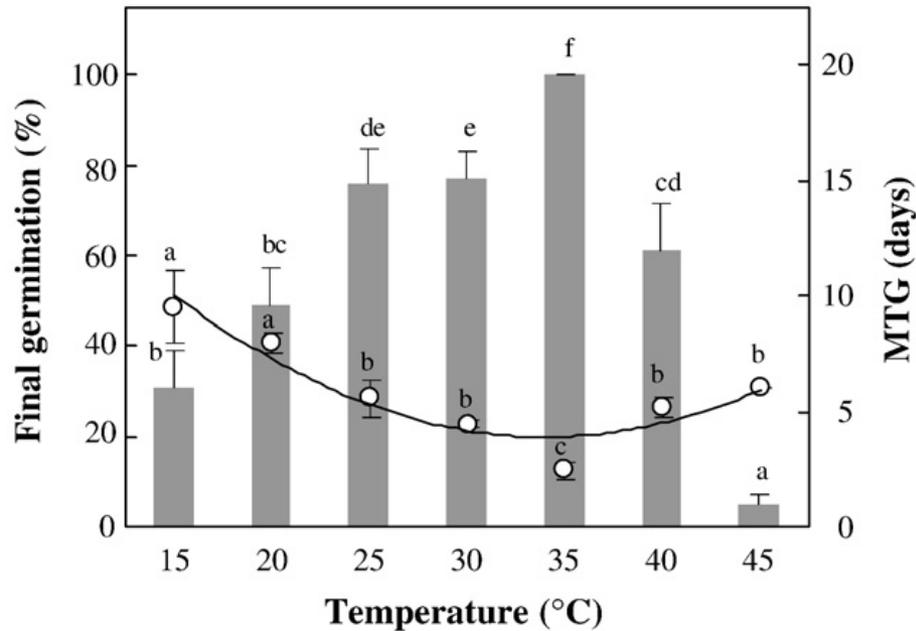


Germination Percentage



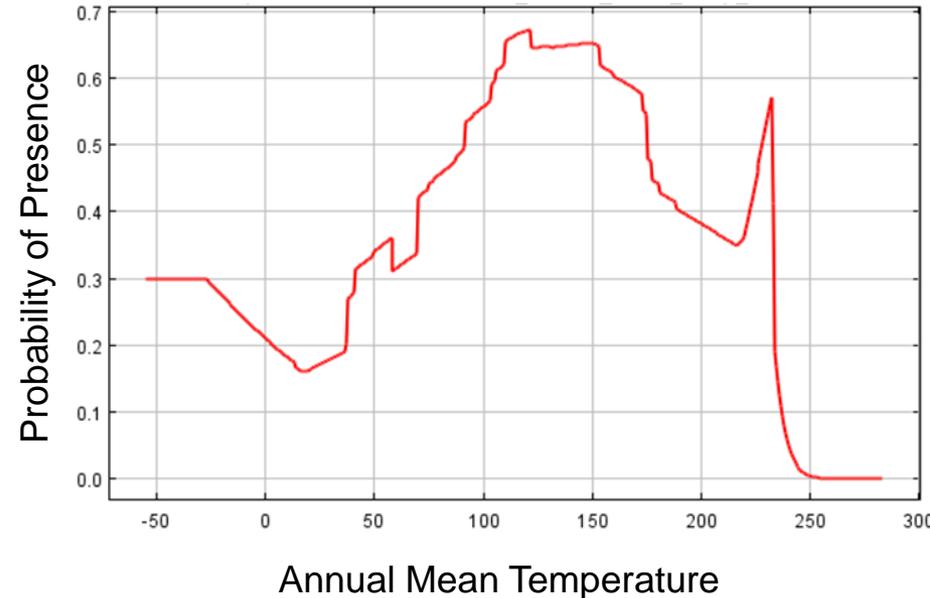
Shrestha, A., E. S. Roman, A. G. Thomas, and C. J. Swanton. 1999. Modeling germination and shoot-radicle elongation of *Ambrosia artemisiifolia*. *Weed Science* 47:557-562.

Measured germination of *Ziziphus lotus* (a buckthorn)



What should the model look like?

Maxent model for *Tamarix* in the US with temperature and precipitation



Are we over-fitting the data?



Maxent Model Parameters

- bio12_annual_percip_CONUS, -4.946359908378759, 52.0, 3269.0
- bio1_annual_mean_temp_CONUS, 0.0, -27.0, 255.0
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- 'bio1_annual_mean_temp_CONUS, 0.12827234634968193, -27.0, 19.5
- linearPredictorNormalizer, 2.2050375426546283
- densityNormalizer, 1311.2581836276431
- numBackgroundPoints, 10000
- entropy, 8.358957722359722

162 Parameters

Maxent model from Tamarix model of western US using precipitation and temperature.

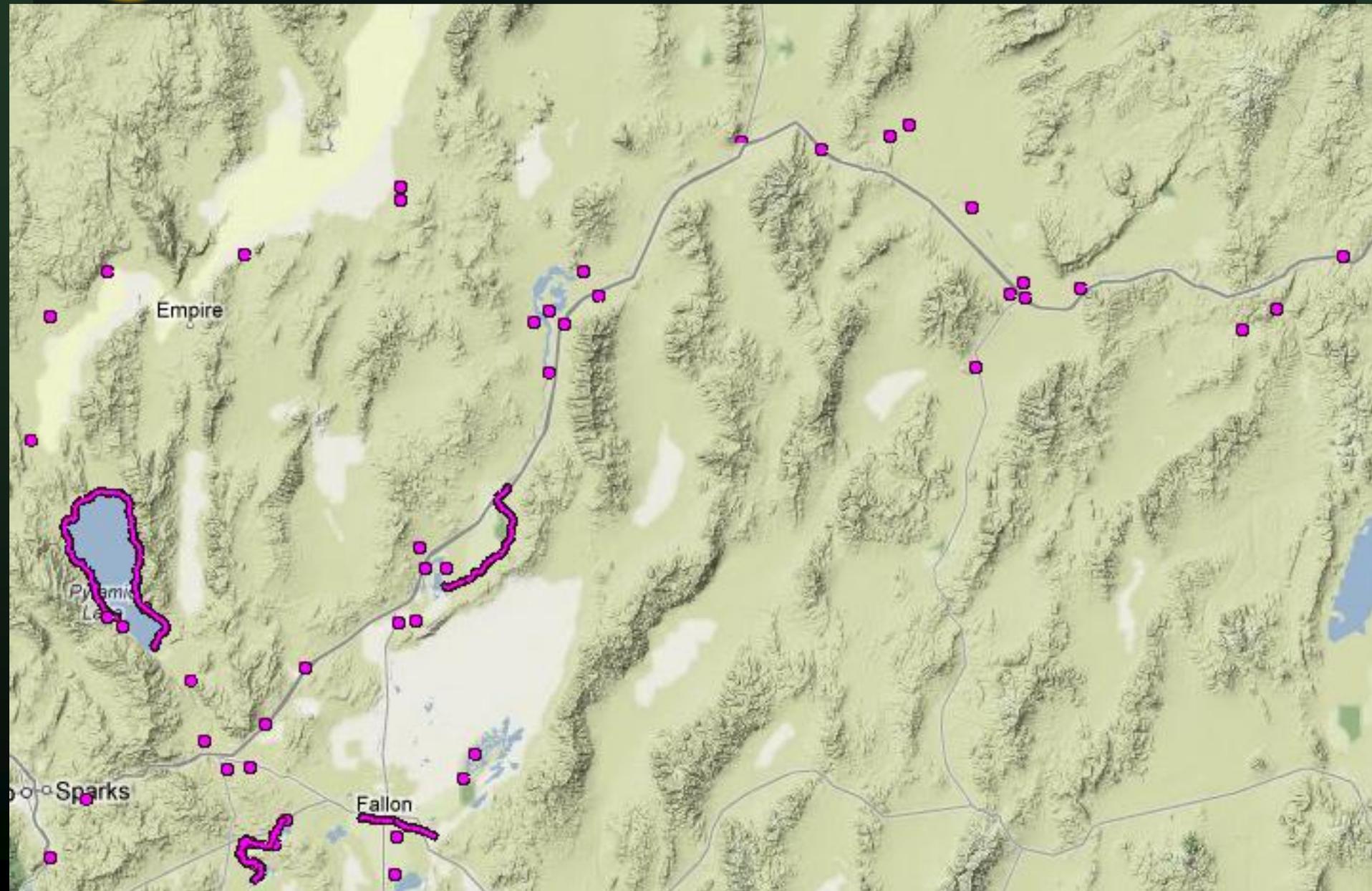


Some Caveats

- We are using “observations”
 - A. Modeling occurrences with some uncertainty
 - B. Modeling the realized niche if the data is a complete sample for the environmental space the species currently occupies
 - C. Modeling the fundamental niche if B is true and the species is covering it's full possible range of habitats
- To model suitable habitat
 - With some uncertainty



Bias from Roads





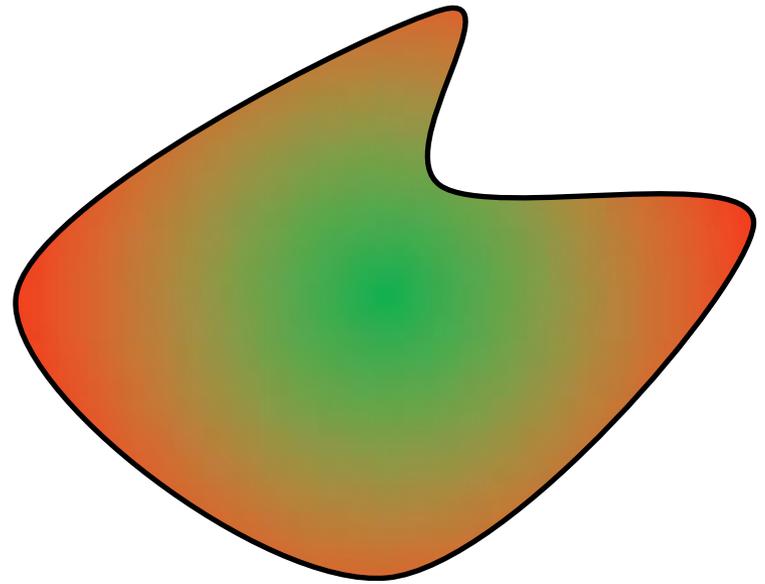
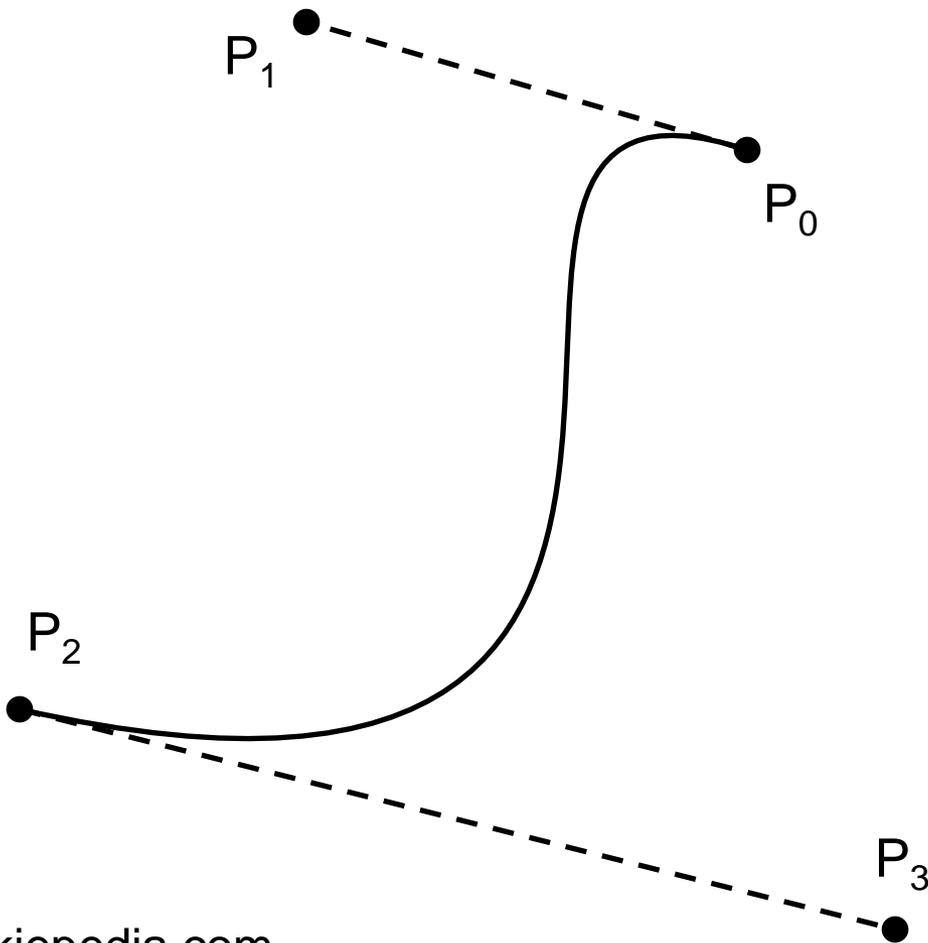
Can we?

- Create a habitat suitability modeling method that:
 - Uses a smaller number of parameters?
 - Minimize over-fitting the data?
 - Allows visualization of the model in environmental space?
 - Does not use absence points?
 - Allows editing the model to try scenarios?



Bezier Curves

$$\mathbf{B}(t) = (1 - t)^3 \mathbf{P}_0 + 3(1 - t)^2 t \mathbf{P}_1 + 3(1 - t) t^2 \mathbf{P}_2 + t^3 \mathbf{P}_3, \quad t \in [0, 1].$$



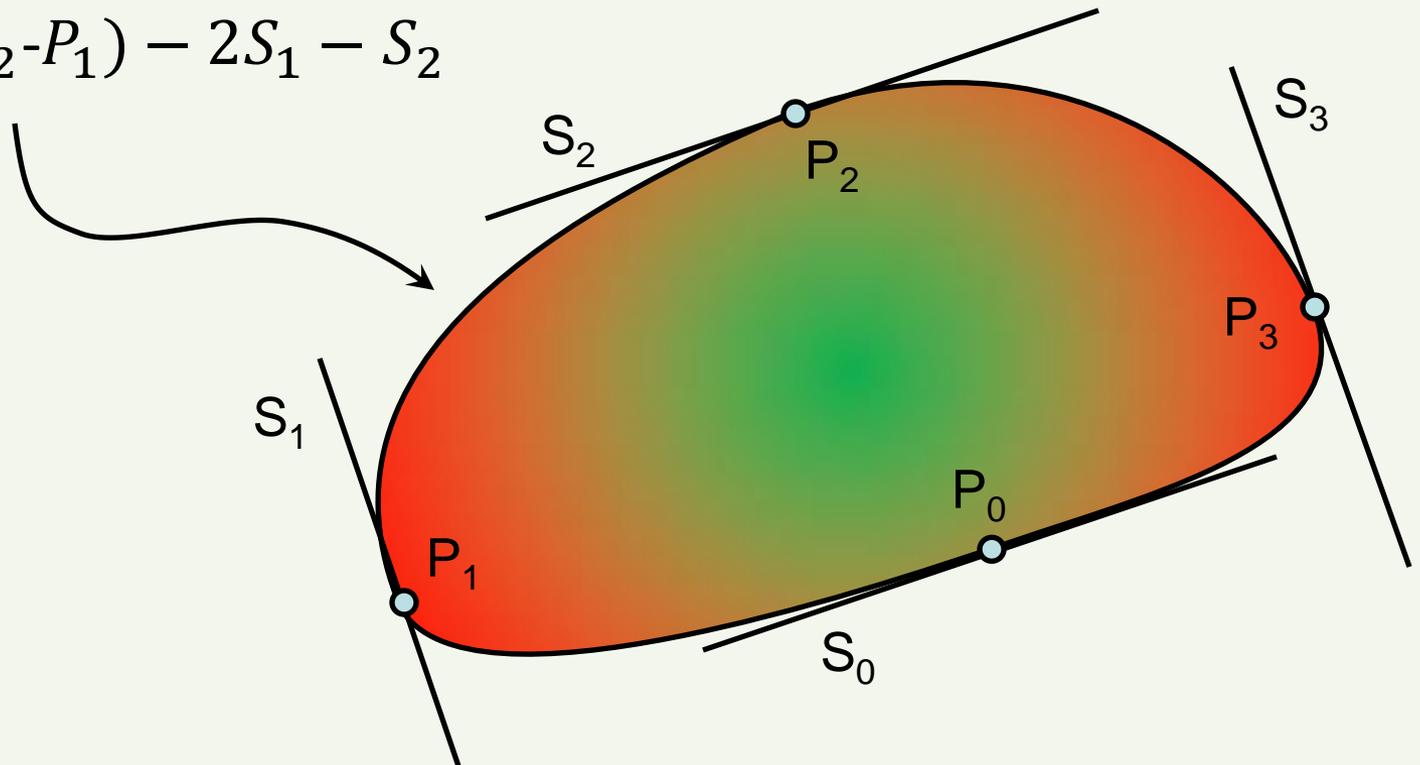


Modified Bezier Curves

$$B(t) = at^3 + at^2 + S_1t + P_1$$

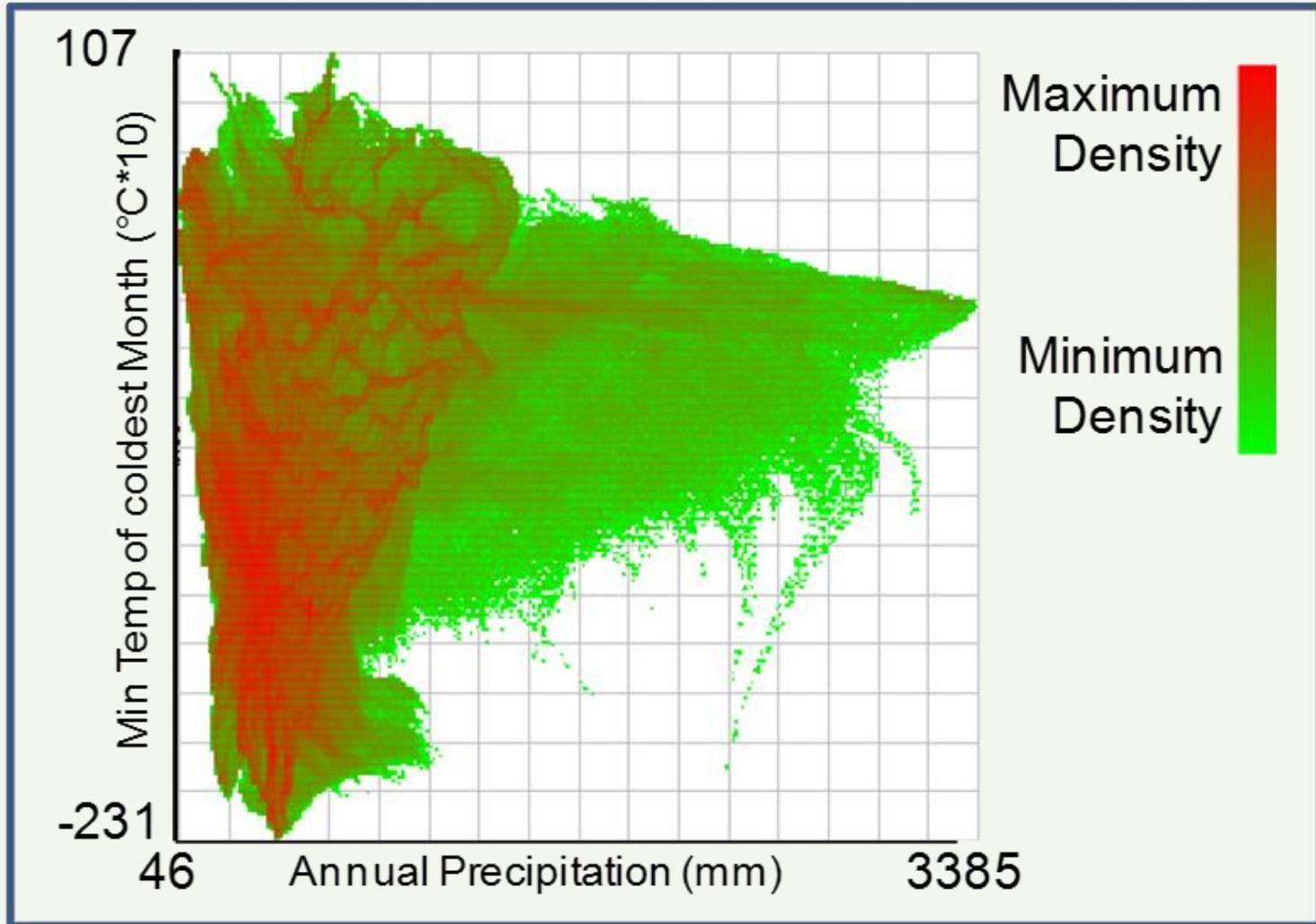
$$a = 2(P_1 - P_2) + S_1 + S_2$$

$$b = 3(P_2 - P_1) - 2S_1 - S_2$$

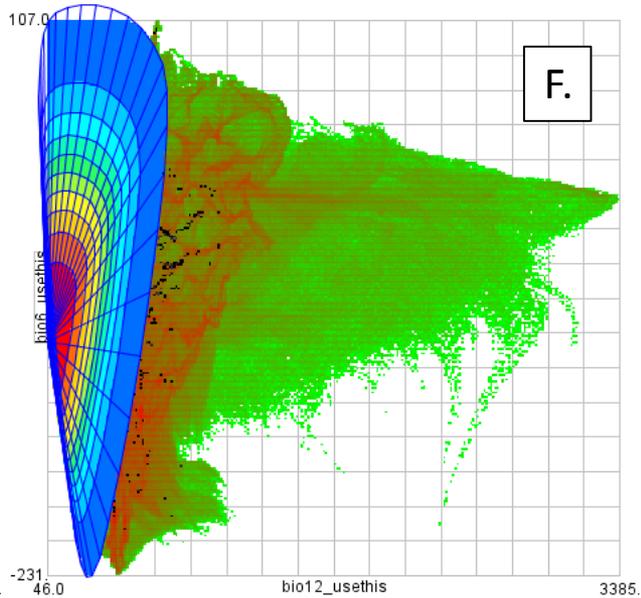
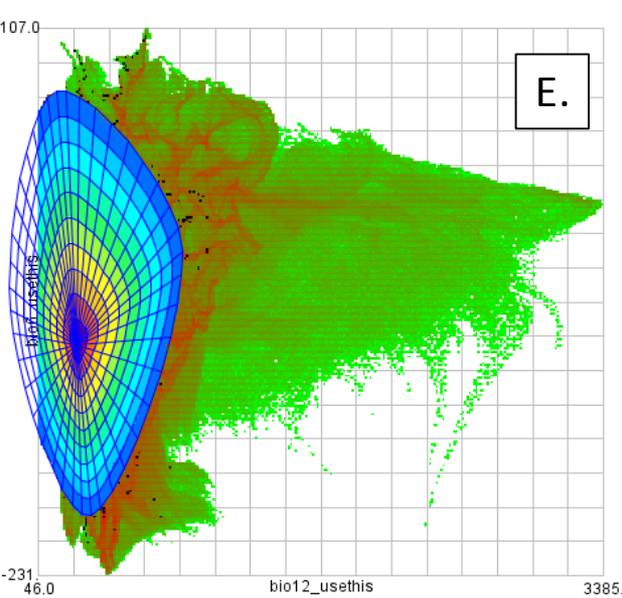
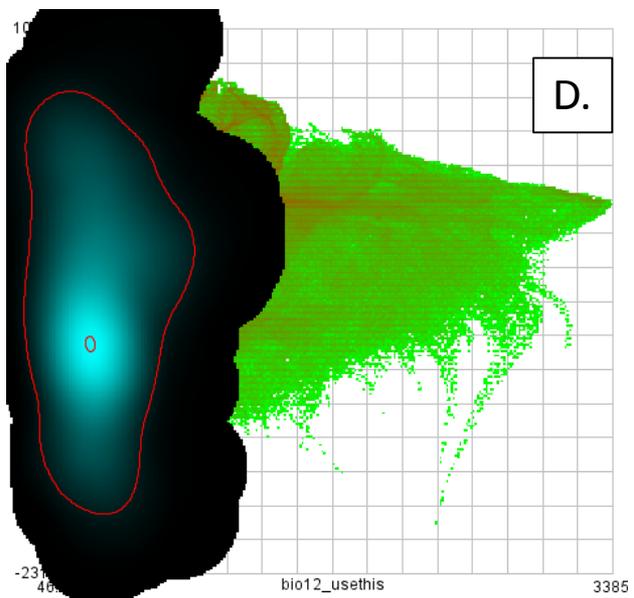
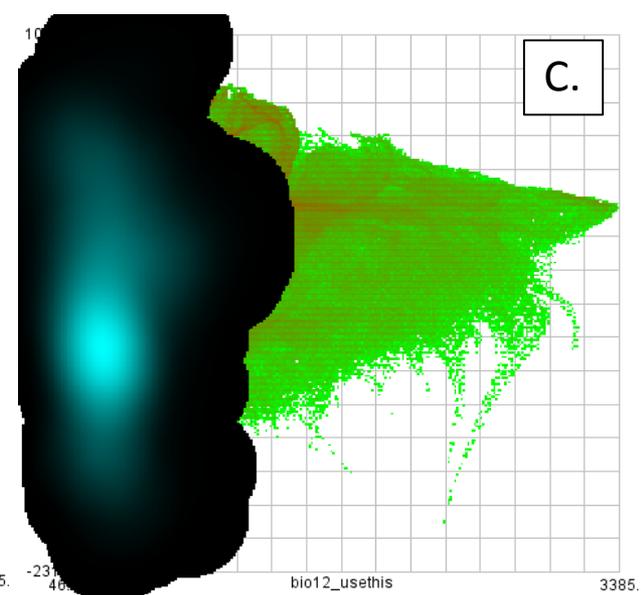
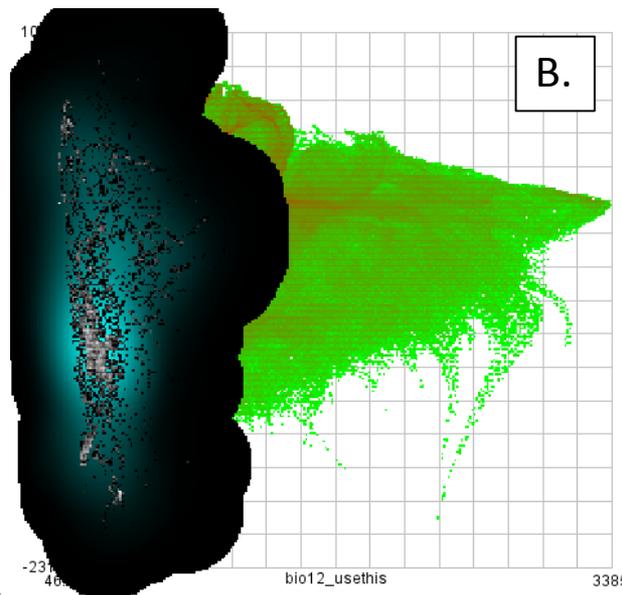
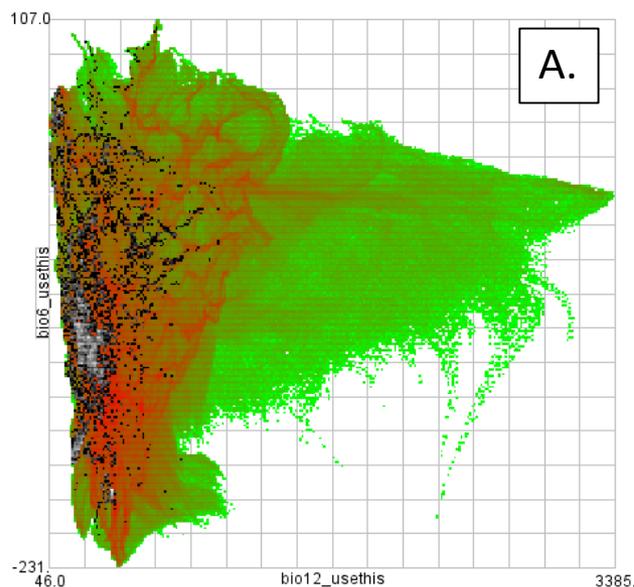


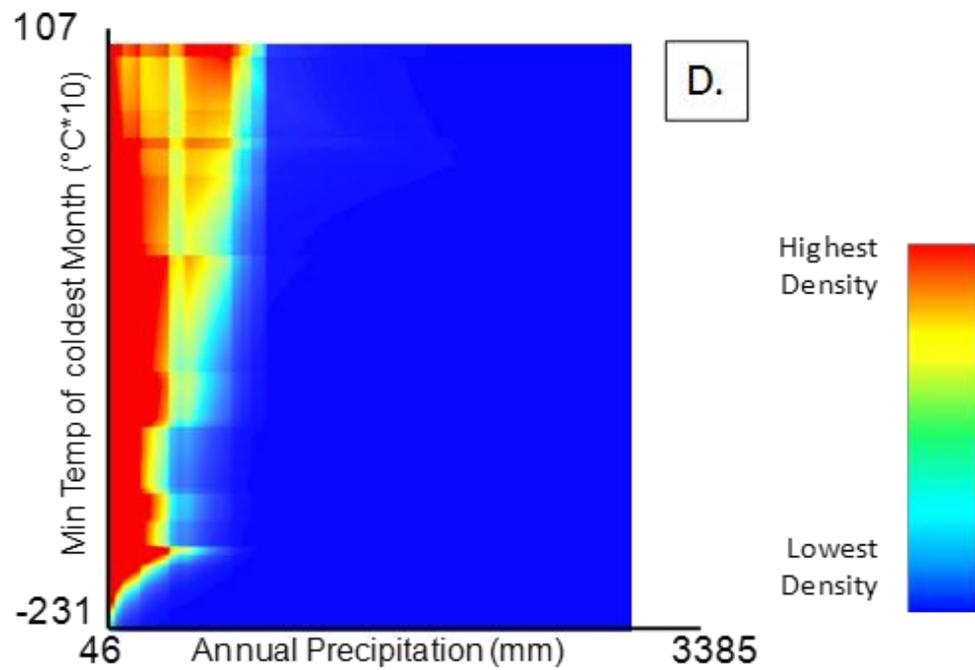
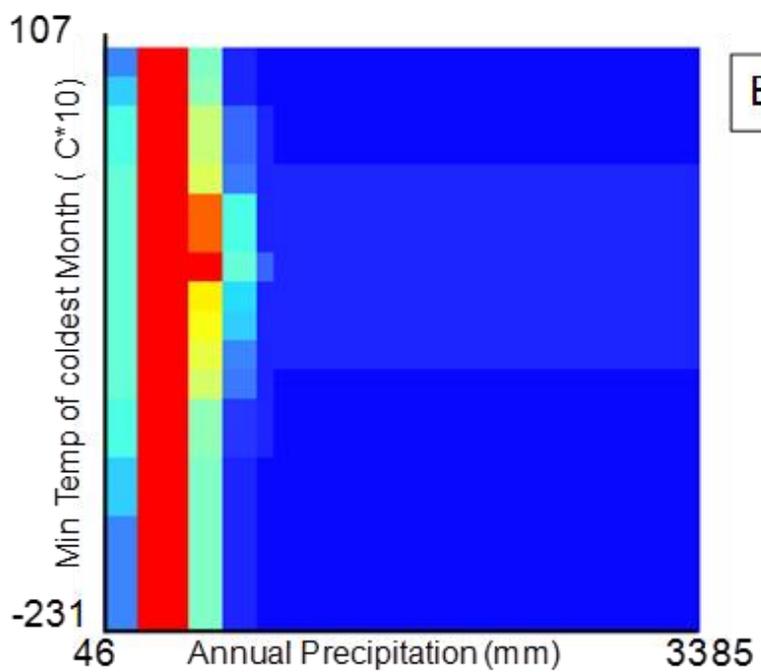
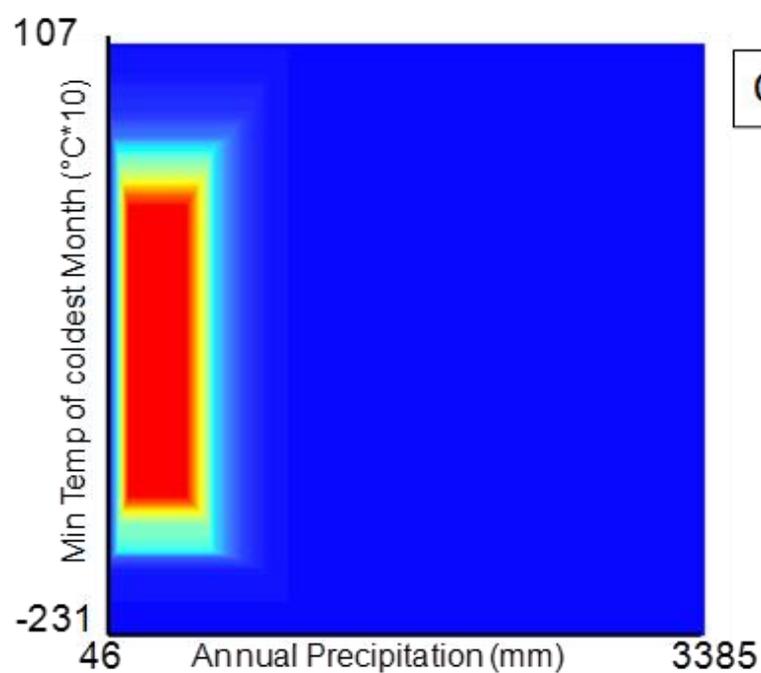
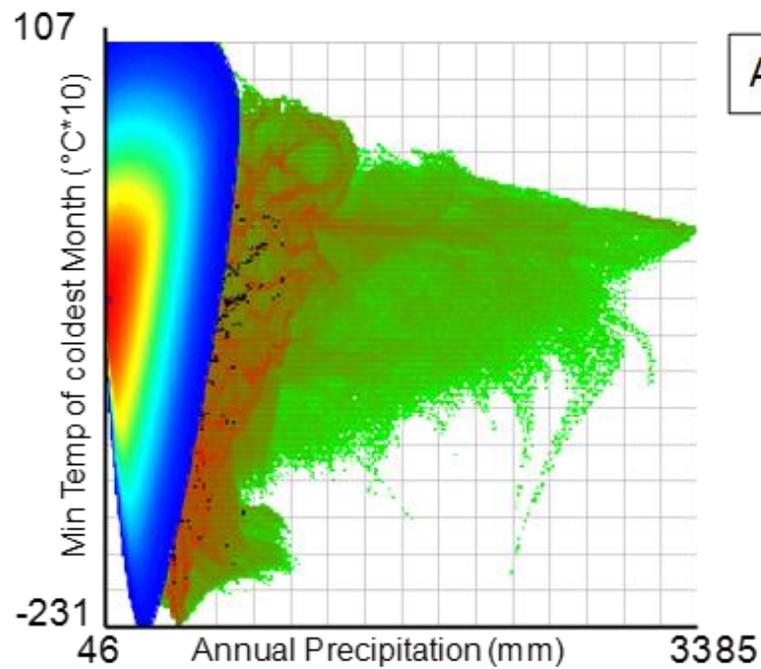


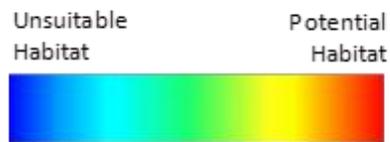
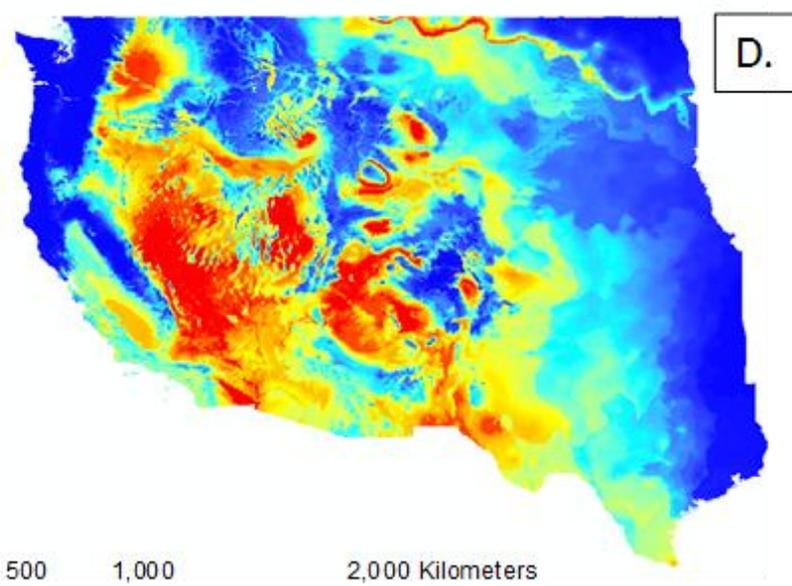
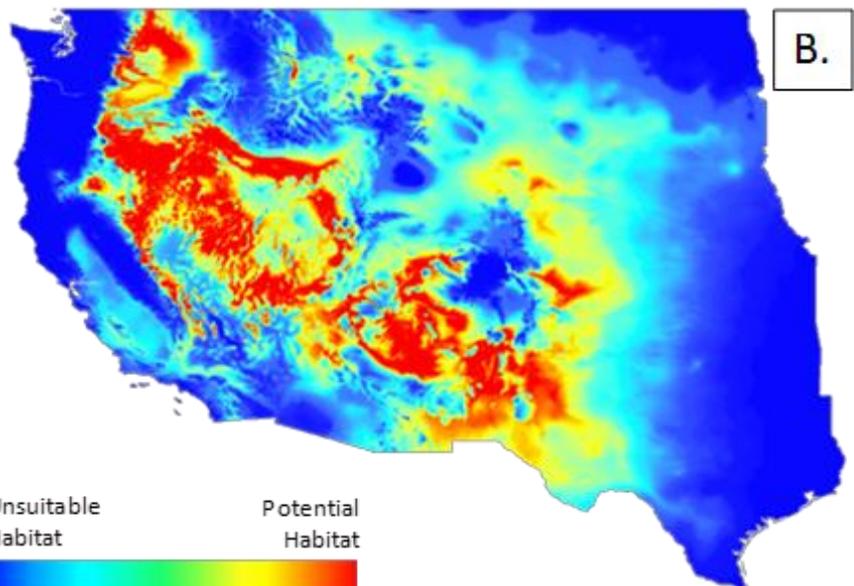
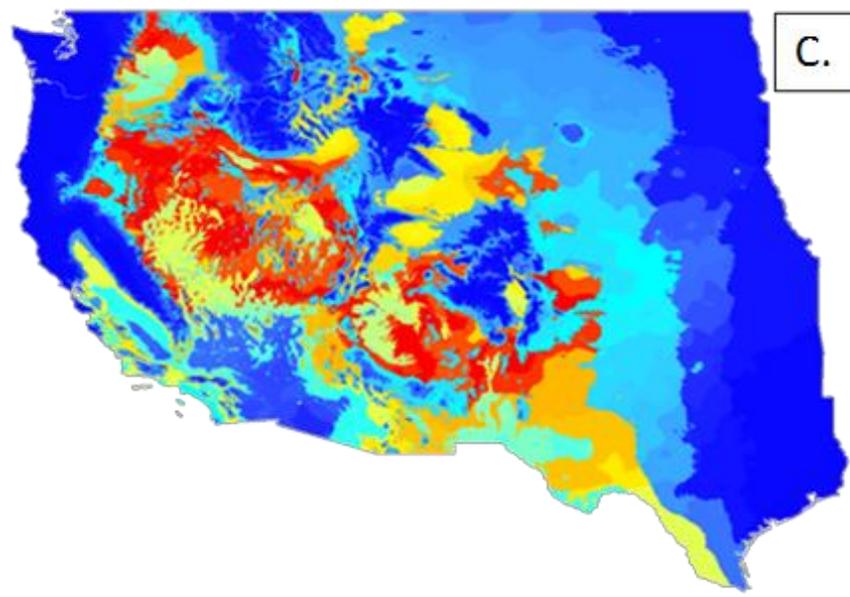
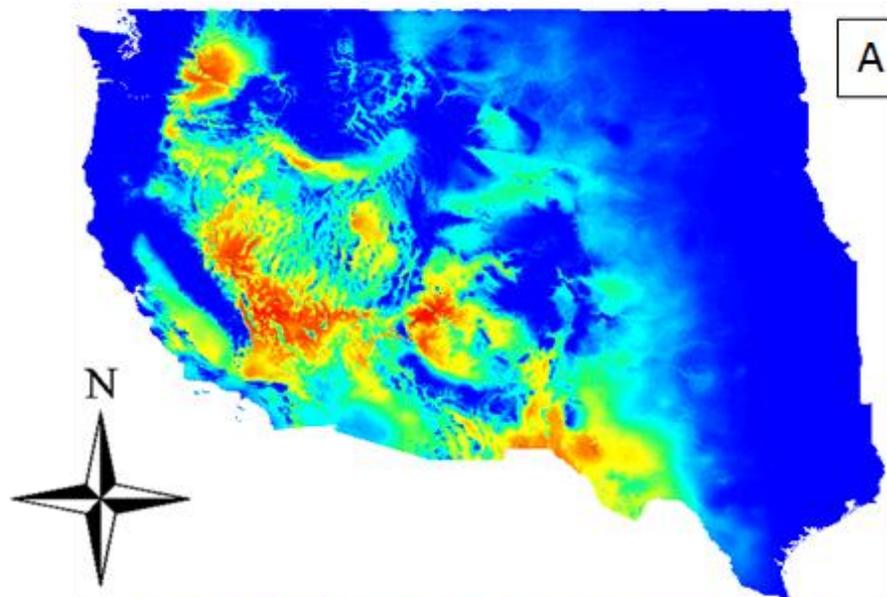
Available Environment



HEMI Modeling Process









Area Under the Curve Metric

- Area Under the Curve (AUC)
 - 1=Perfect
 - 0.5=Random

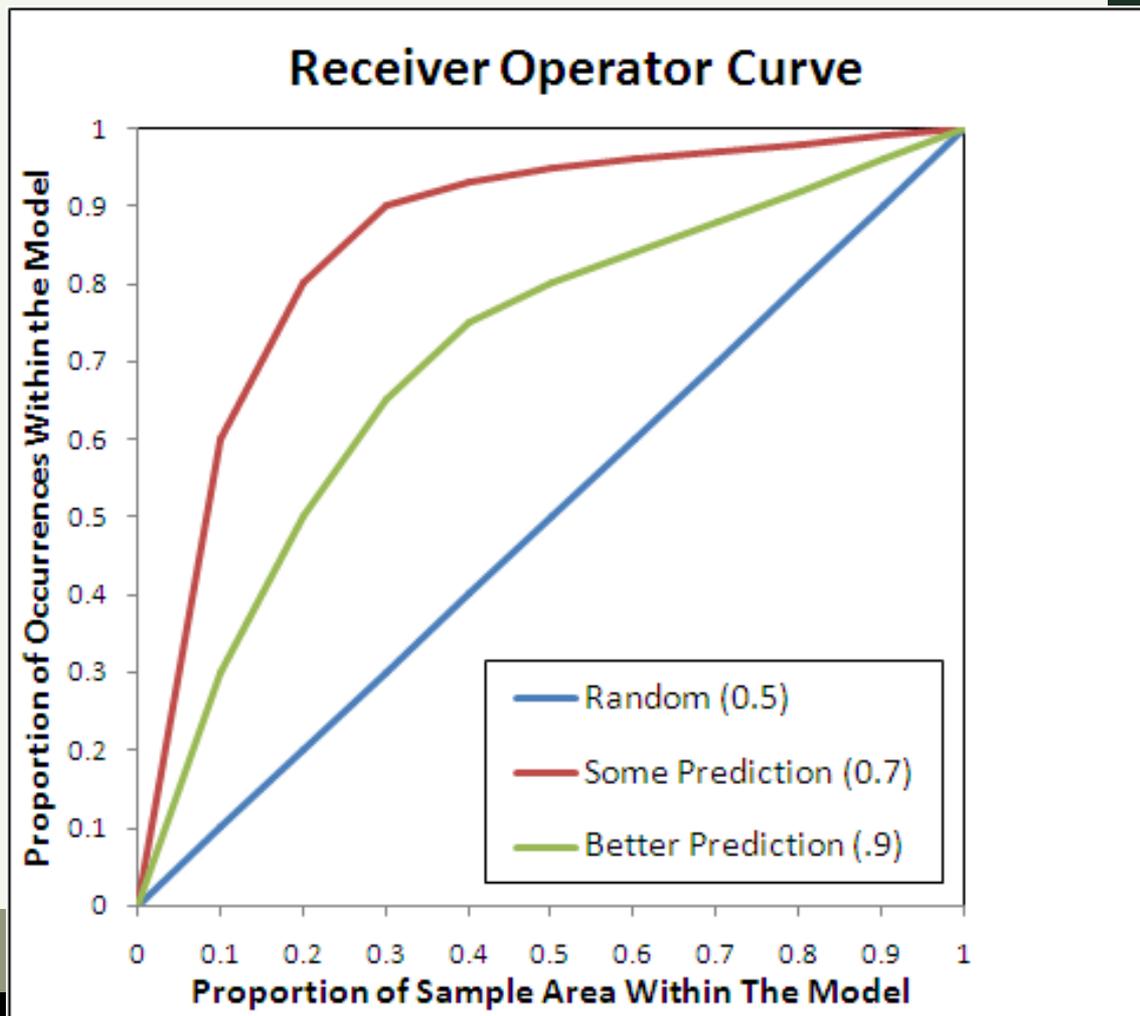




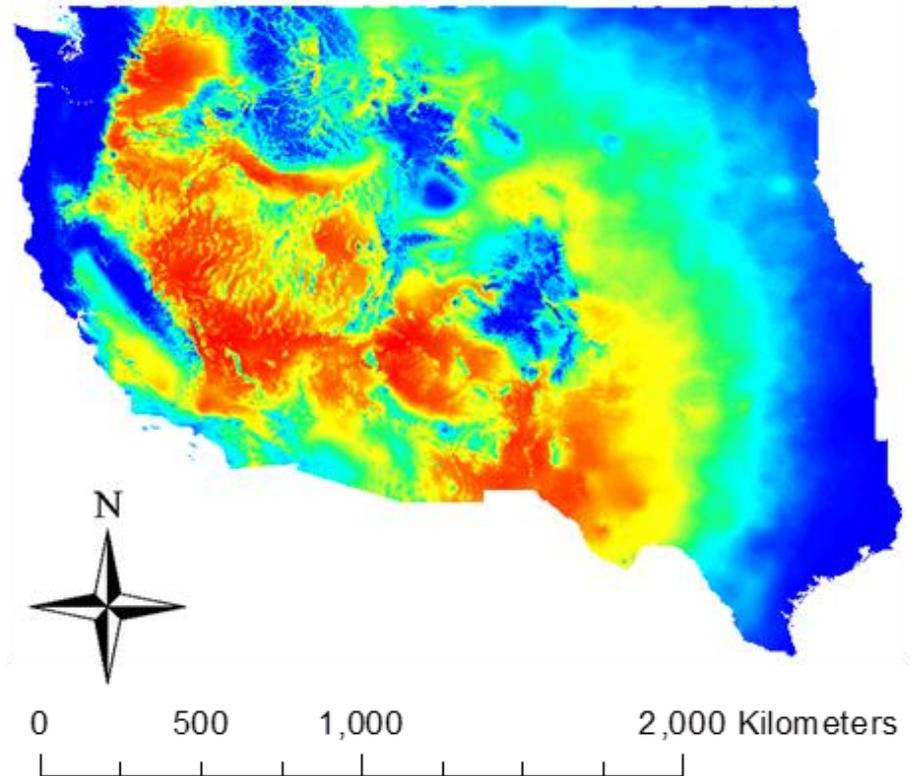
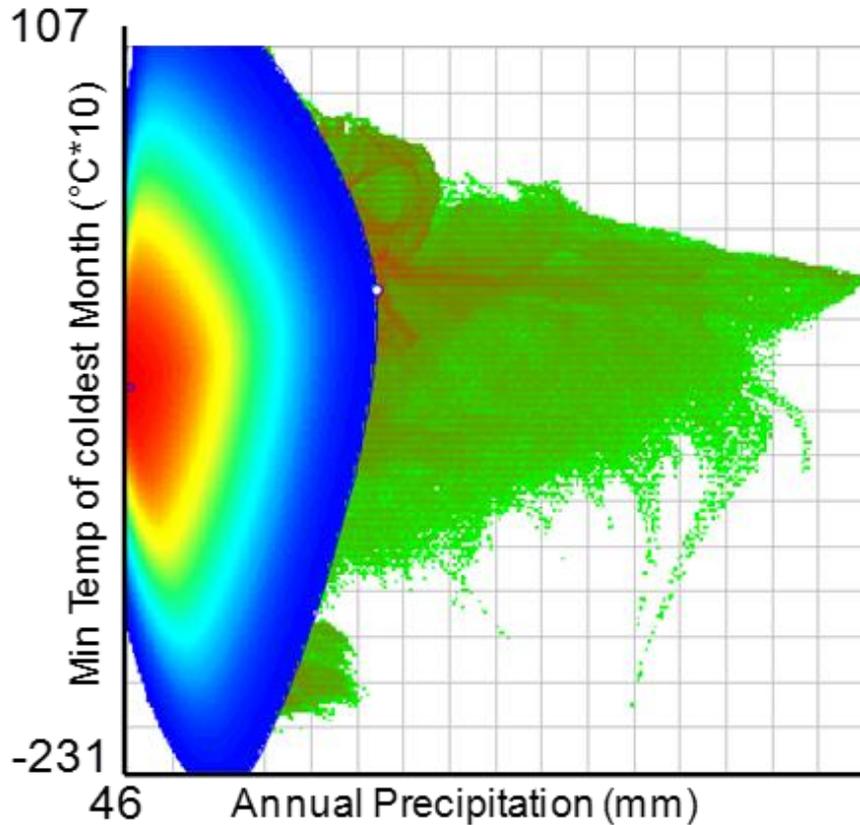
Table of AUC Values

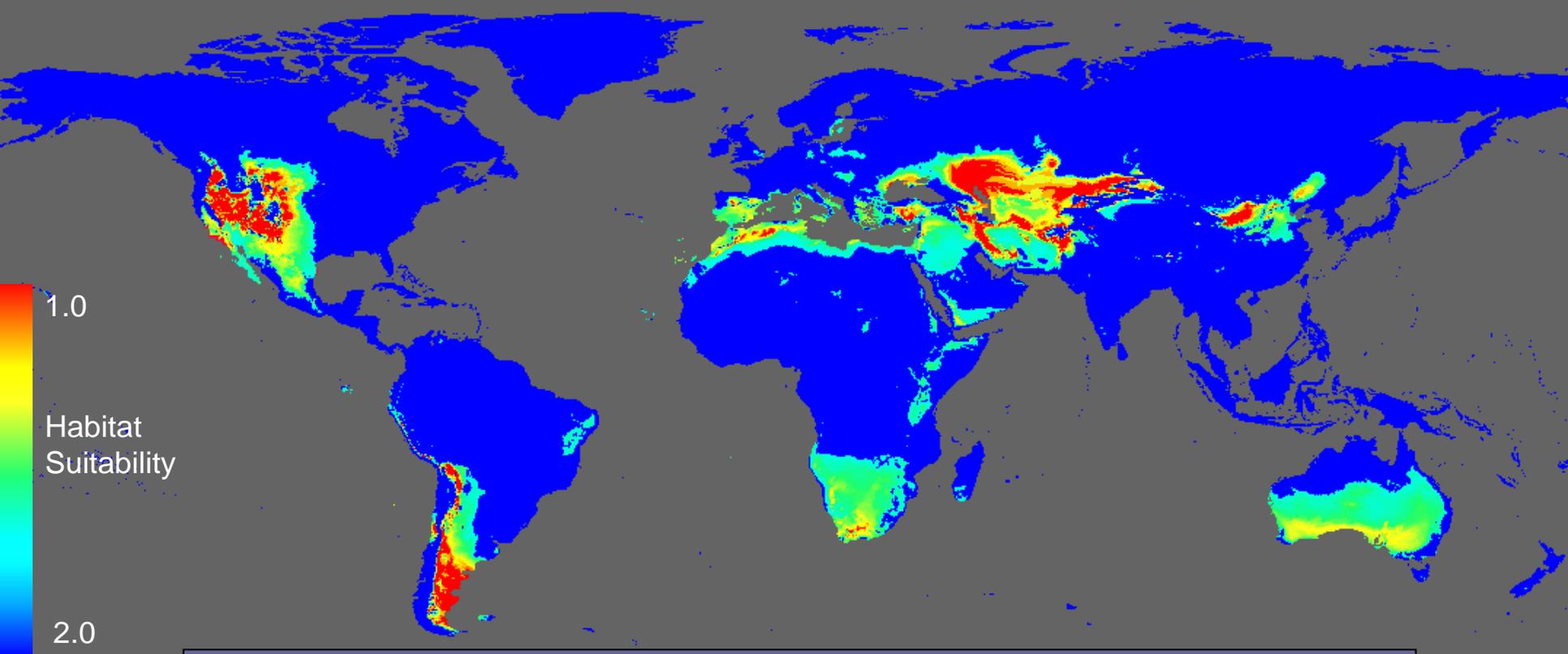
Method	Train AUC	Parameters	Test AUC
HEMI Automatic	0.75	10	0.76
HEMI Edited	0.73	10	0.73
BioClim	0.67	8	0.68
BioMapper	0.74	625	0.72
Median			
Maxent	0.77	162	0.78



Editable Control Points

- Expanded habitat







Conclusions

- Number of parameters can be reduced while minimizing over-fit with a small loss of model accuracy to the data
- We can visualize models in environmental space
 - At least in 2 dimensions
- Absence points are not required
- We can edit the models for scenarios



Next Steps

- Expand to N dimensions
- Faster parameter optimization
- Evaluate with additional species
- Include uncertainty analysis and error surfaces
- Improved data!



Develop Guidelines

- “Over-fitting”
 - Conservative estimate of suitable habitat
 - Appropriate for threatened and endangered species where we want to ensure we are preserving suitable habitat
- “Under-fitting”
 - Expanded estimate of suitable habitat
 - Appropriate for invasive species where we want to check anywhere the species may be becoming established



Acknowledgements

- Tom Stohlgren, Lane Carter, Abby Flory, Paul Evangelista, Sunil Kumar, Alycia Crall, Kirstin Holfelder, Sara Simonson, Lee Casuto, Annie Simpson, Jeff Morisette, and many more...
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 - NSF Grant #OCI-0636210
- BlueSpray by SchoonerTurtles, Inc.

