

Using museum data for species distribution modeling: The case of plants in Florida

Charlotte Germain-Aubrey, Julie Allen, Robert Guralnick, Shawn Laffan, Brent Mishler, Kurt Neubig, Douglas Soltis, Lucas Majure, Pamela Soltis



iDigBio is funded by a grant from the National Science Foundation's Advancing Digitization of Biodiversity Collections Program (Cooperative Agreement EF-1115210). Any opinions, findings, and conclusions or recommendations expressed in this material are those of the author(s) and do not necessarily reflect the views of the National Science Foundation. All images used with permission or are free from copyright.

Niche Modeling: A widespread tool

- PRACTICAL
 - Invasive species
 - Disease
 - Data deficient species
 - Conservation/Land Management
- THEORETICAL
 - Diversity through time
 - Evolutionary and ecological patterns

Modeling the Distribution of Species

- Location information and environmental data
- Software to model the range of each species
- Project onto future climate conditions
- For Florida plants:
 - 1548 plant species (of 4100 species)
 - >511,000 georeferenced points (GPS)
 - Environmental features: temperature, precipitation, soil, etc.

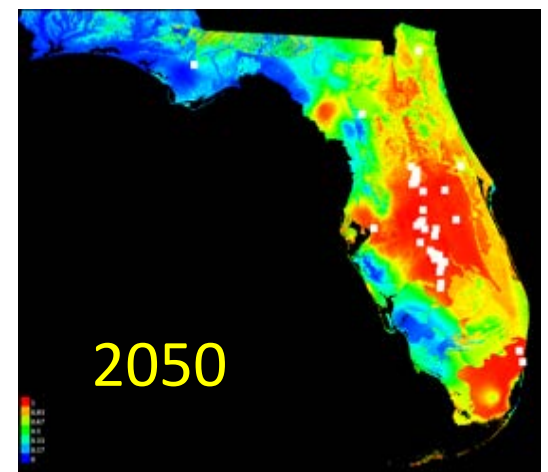
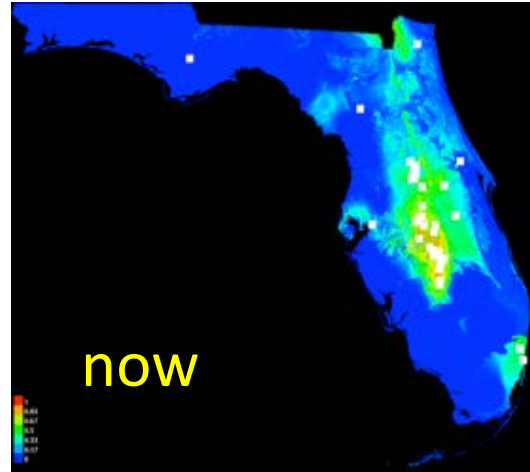


Niche Modeling: A widespread tool

- PRACTICAL
 - Invasive species
 - Disease
 - Data-deficient species
 - Conservation/Land Management
- THEORETICAL
 - Diversity through time
 - Evolutionary and ecological patterns

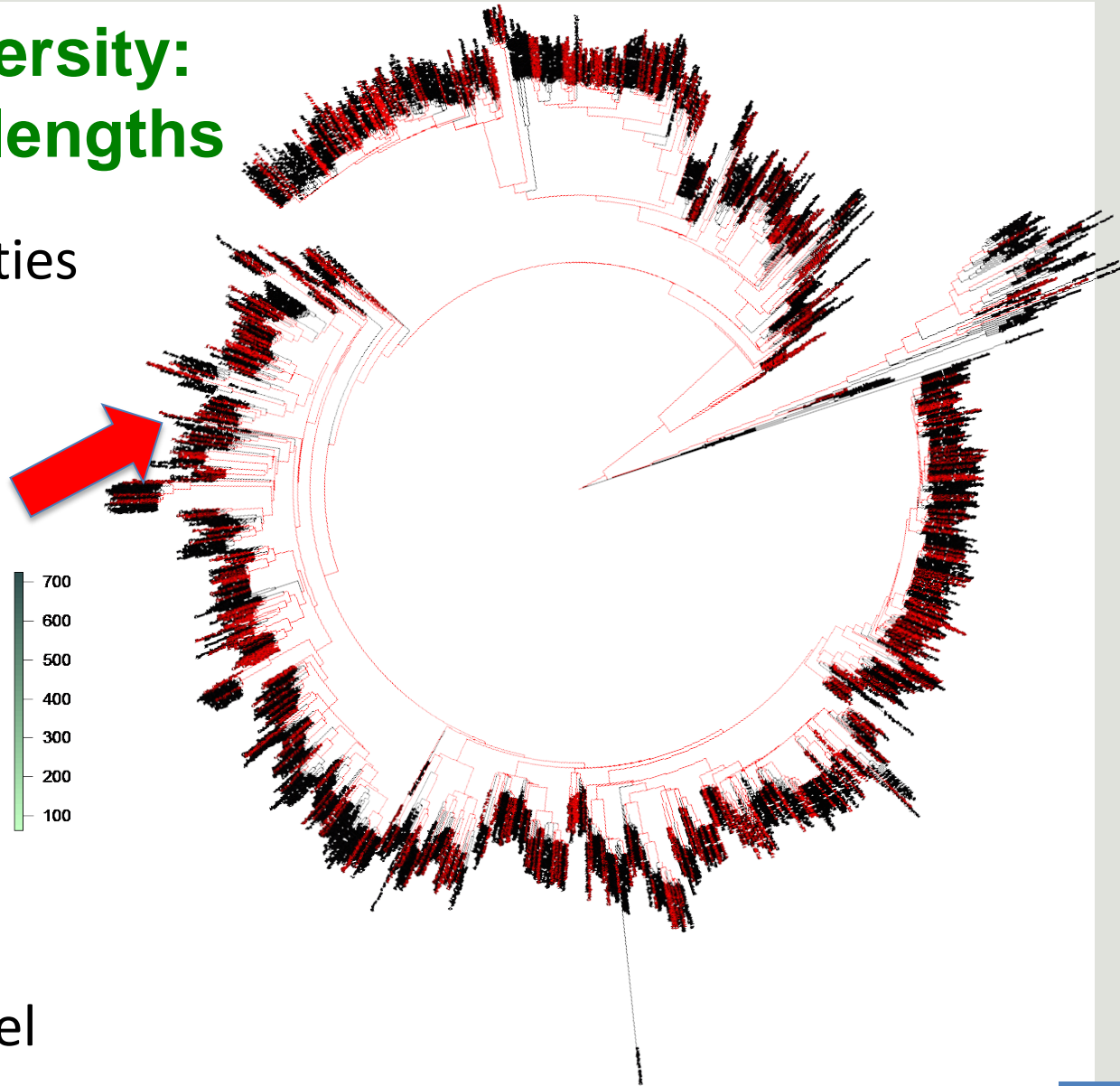
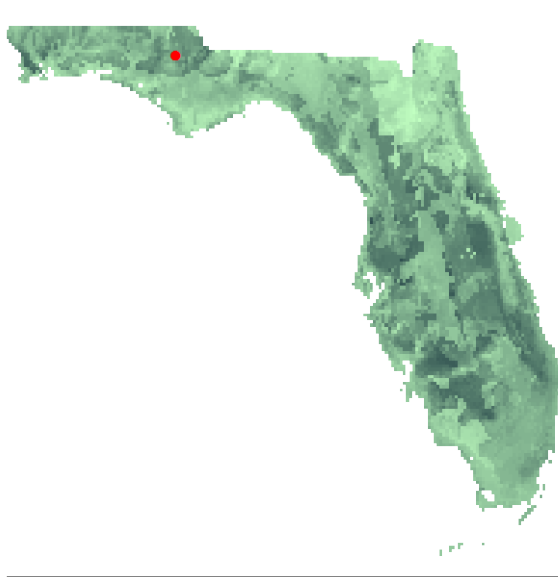
Response to Climate Change

Prunus geniculata (scrub plum) – Lake Wales Ridge



Phylogenetic Diversity: \approx sum of branch lengths

8,045 pixels/communities
16 km² per pixel



Species list at each pixel
generated from ENMs

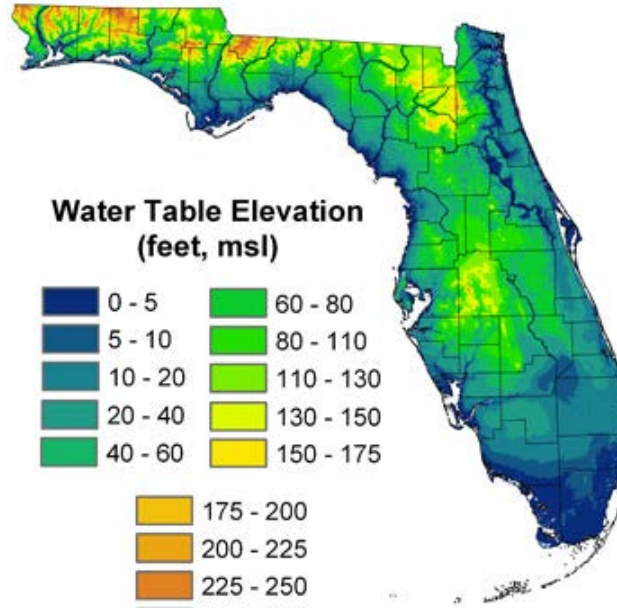
How does MaxEnt work?

- Uses presence only
- Using Maps:
 - Extract values from environmental layers
 - Creates background points as pseudo-absences
- Using Values:
 - You can feed MaxEnt the environmental values for each point
 - Allows one to extract values from different maps/inputs

Parameters to explore:

1. Using **yearly climate data** from time of collection to improve niche models using museum specimens
2. Using **reduced area** to train the models
3. **Smoothing response curve** to mimic more realistic response of organism
4. Using **fewer background points** to increase computing efficiency

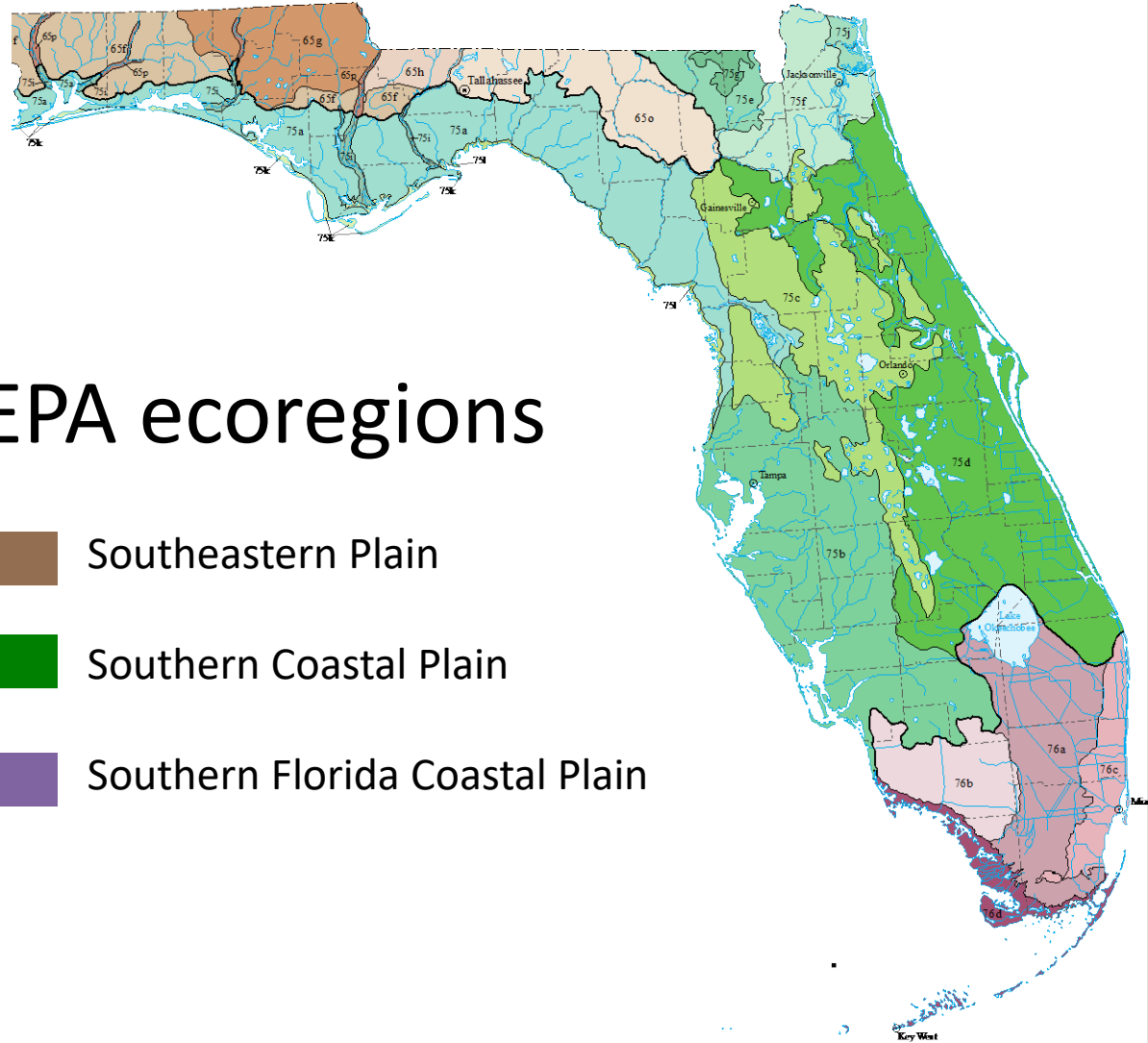
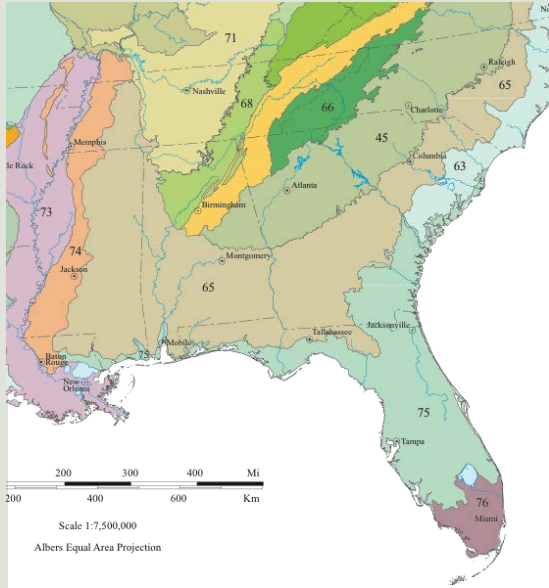
Florida



**Water Table Elevation
(feet, msl)**

0 - 5	60 - 80
5 - 10	80 - 110
10 - 20	110 - 130
20 - 40	130 - 150
40 - 60	150 - 175
175 - 200	
200 - 225	
225 - 250	
250 - 280	
280 - 328	





EPA ecoregions

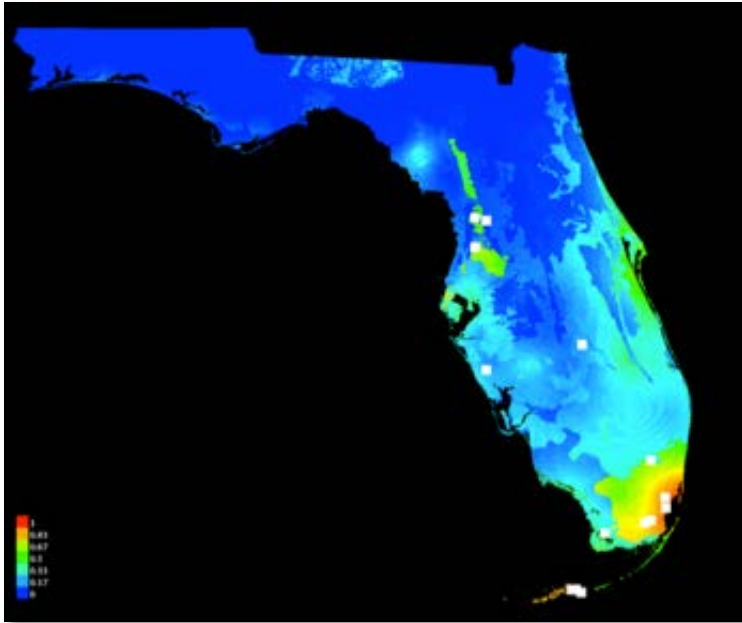
- Southeastern Plain
- Southern Coastal Plain
- Southern Florida Coastal Plain



~ 4,100 plant species

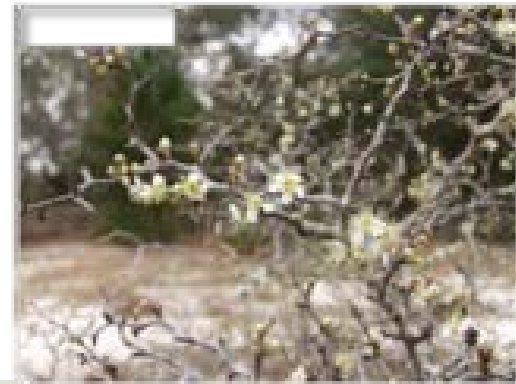
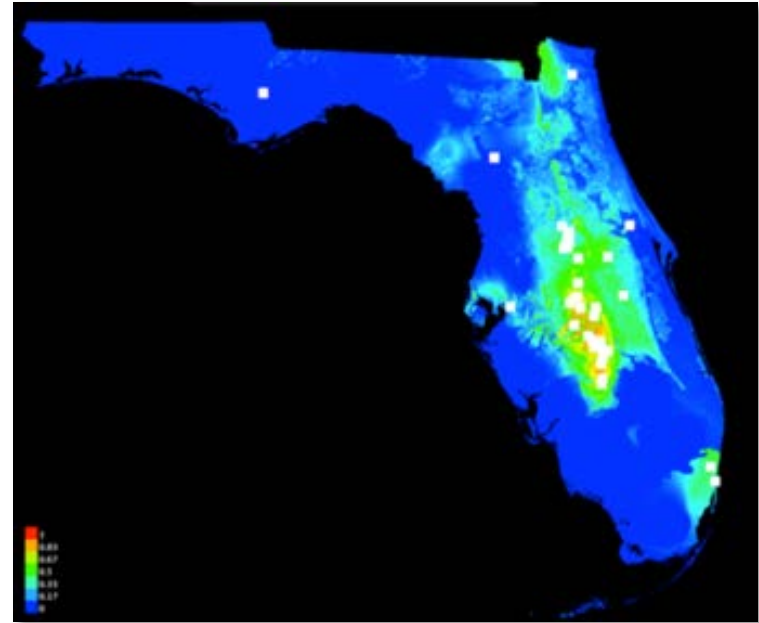
Flatspike Sedge

Abildgaardia obovata

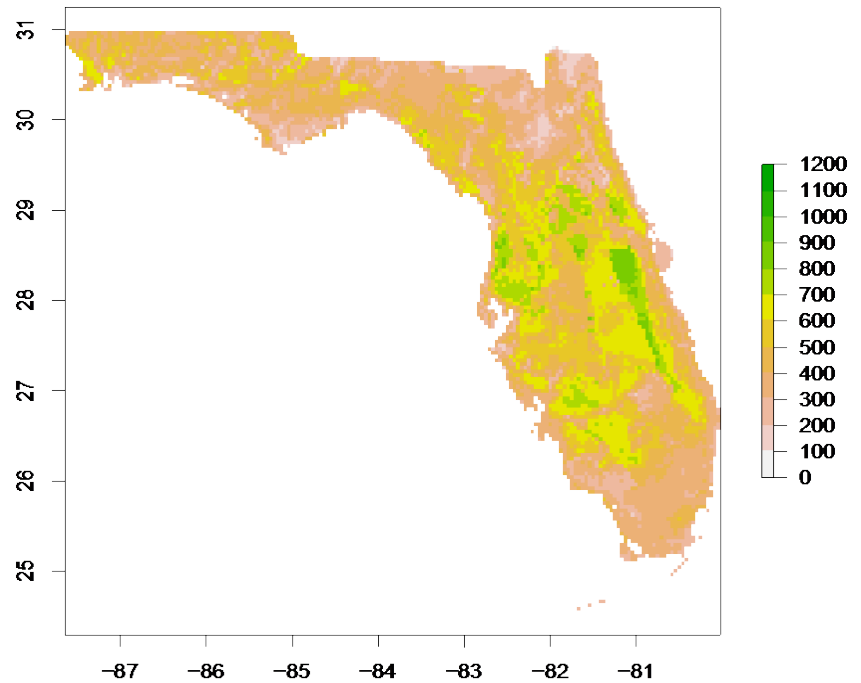
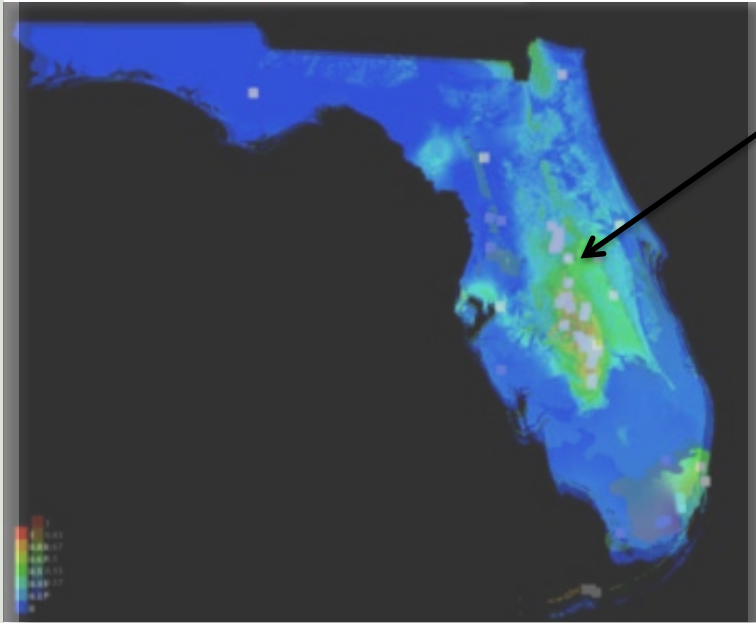


Scrub Plum

Prunus geniculata



What species are predicted to reside in this point?



- Reconstructing species distributions
 - museum specimens
 - long-term monitoring data
 - *to generate ranges for use in biodiversity and PD analyses*
- Sampling in 3 EPA regions present in FL
- Taxonomic Name Resolution
- >30 points per species (10-20 for rare endemics)
372,241 unique points

1,145 species models (30%)

Parameters to explore:

1. Using **yearly climate data** from time of collection to improve niche models using museum specimens
2. Using **reduced area** to train the models
3. **Smoothing response curve** to mimic more realistic response of organism
4. Using **fewer background points** to increase computing efficiency

Using Yearly Climate Data to reconstruct a species niche

- Dated (year)
- Reflect distribution of plants through recent history
- Can use the climate conditions at the time of collection
- Use PRISM monthly data
 - Reconstruct Bioclim layers for each year of collection
 - Run MaxEnt model

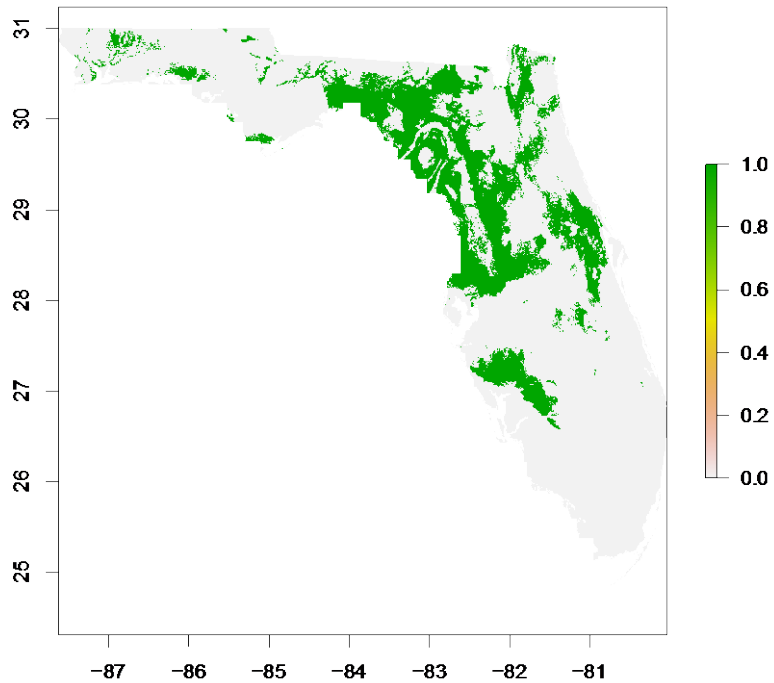


>1600 natural history
collections
in the US alone
1-2 billion specimens
in the US
3-4 billion specimens
worldwide

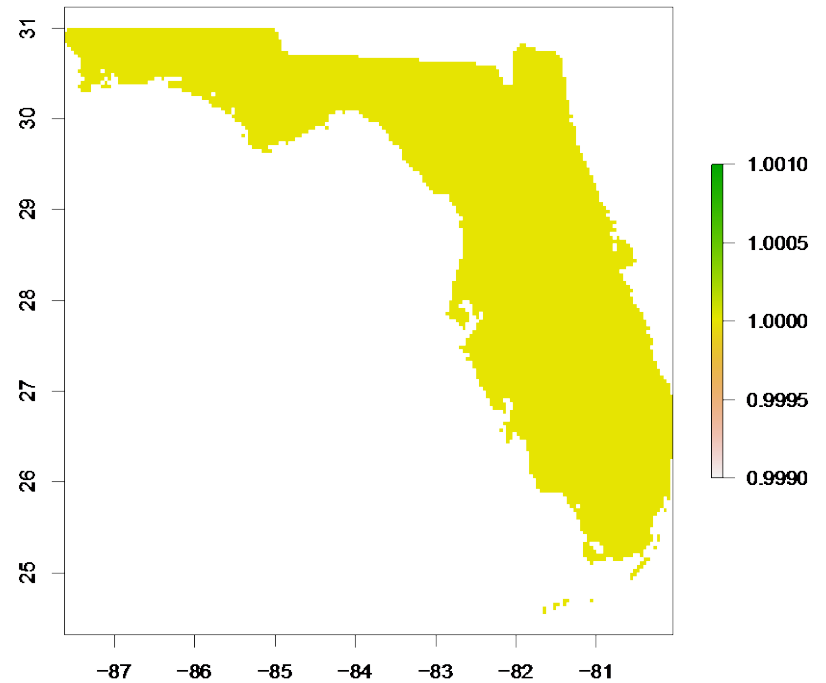


Long-lived species

Average Climate

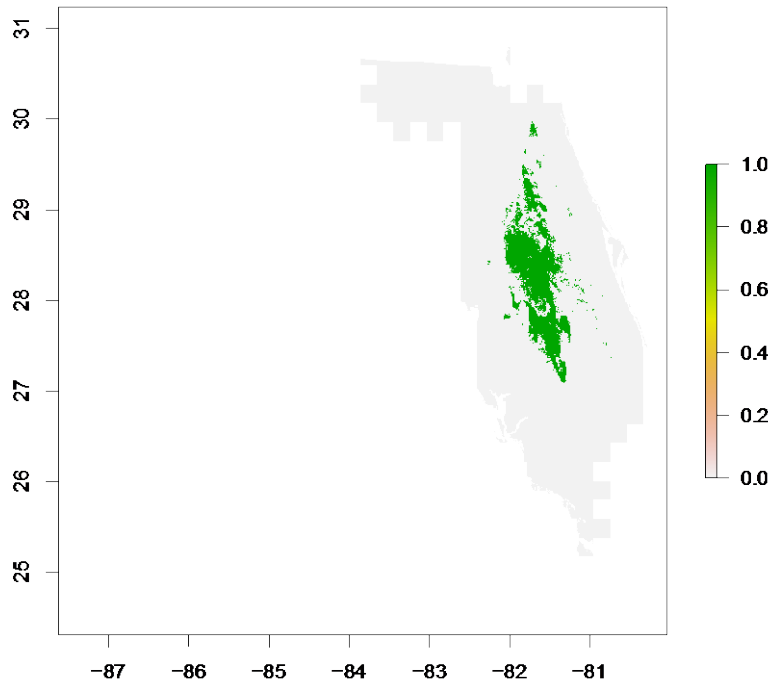


Yearly Climate

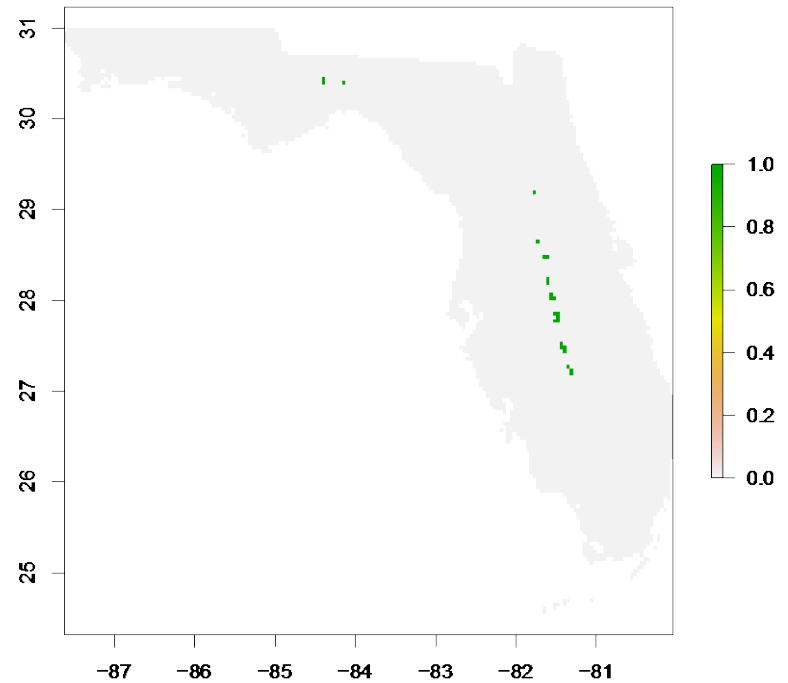


Short-lived narrow species

Average Climate



Yearly Climate



Parameters to explore:

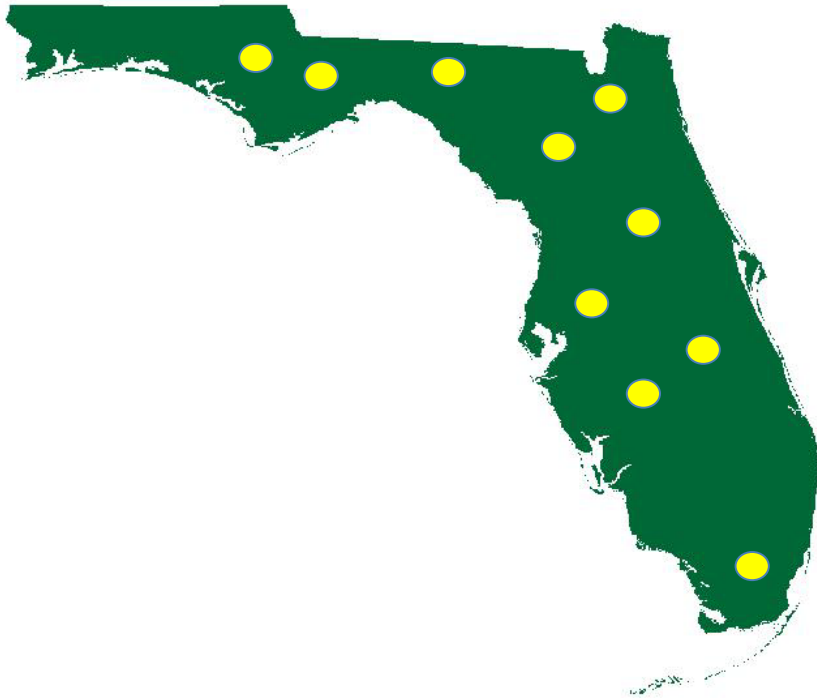
1. Using **yearly climate data** from time of collection to improve niche models using museum specimens
2. Using **reduced area** to train the models
3. **Smoothing response curve** to mimic more realistic response of organism
4. Using **fewer background points** to increase computing efficiency

Selecting an area to train the model

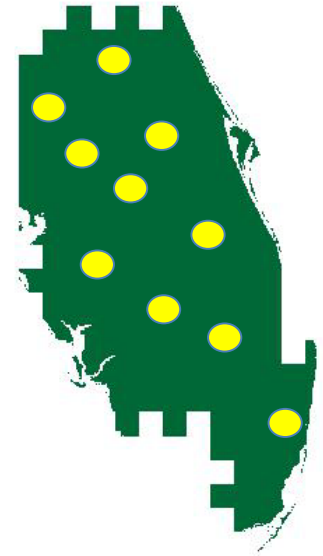
- Limiting the area in which the model trains
- If you want to recover suitable niche
 - Include an area where biotic interactions might be excluding the species
 - Large mask
- If you want to recover species distribution
 - Consider abiotic factors as part of the limiting factor
 - Delimit mask from presence points
 - Tighter mask

Selecting a meaningful area to train the model

NO mask



WITH mask



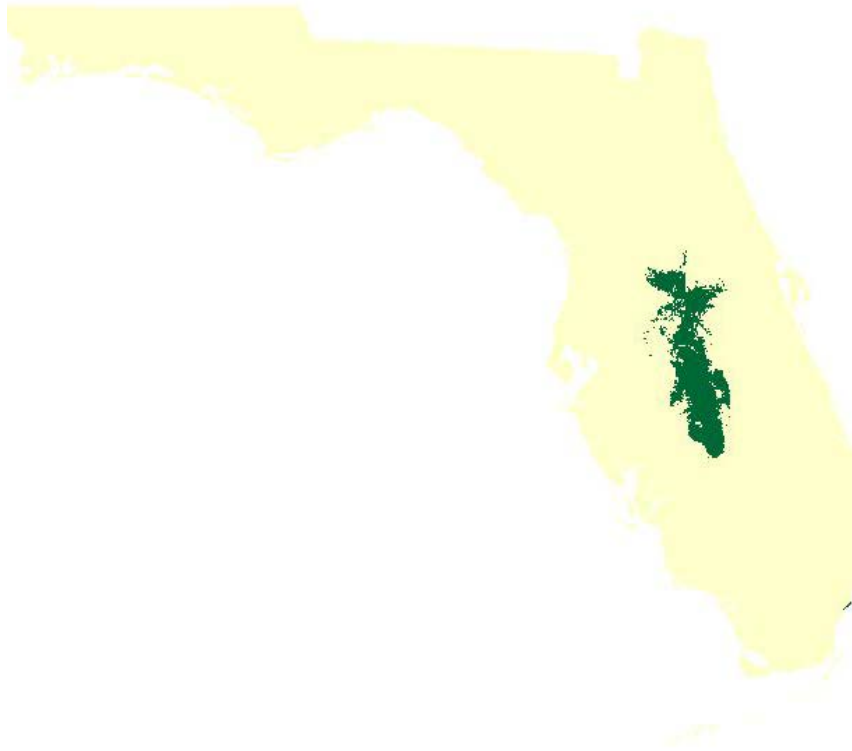
Prunus geniculata



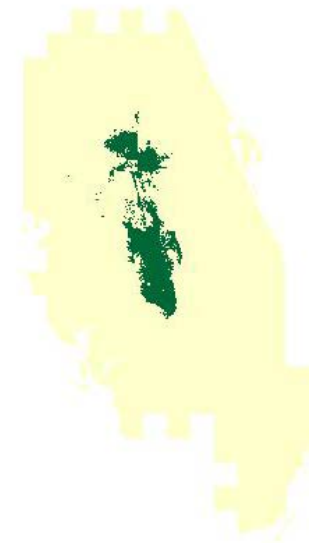
Credit: fws



NO mask



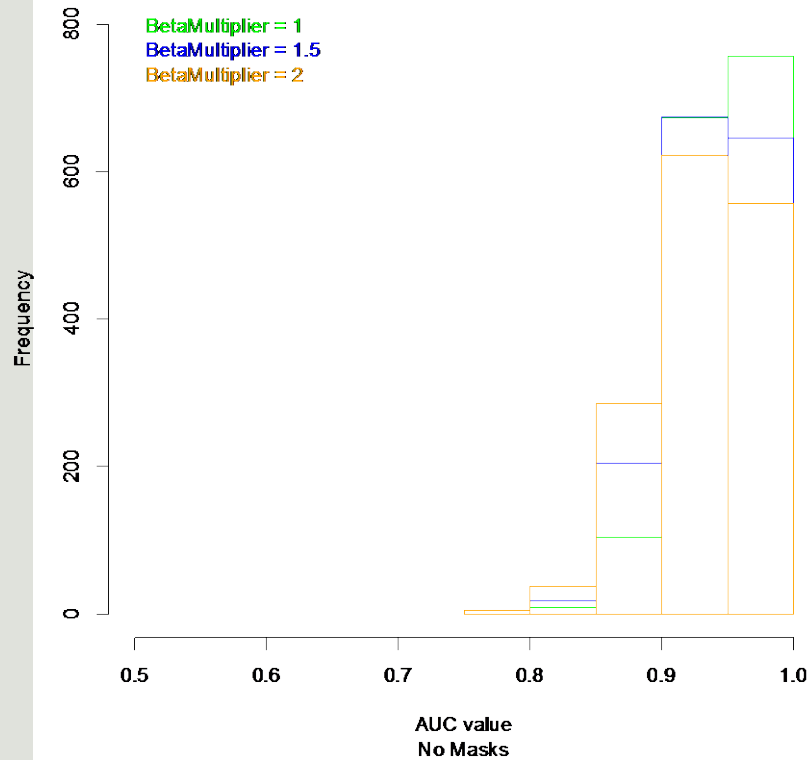
WITH mask



AUC scores

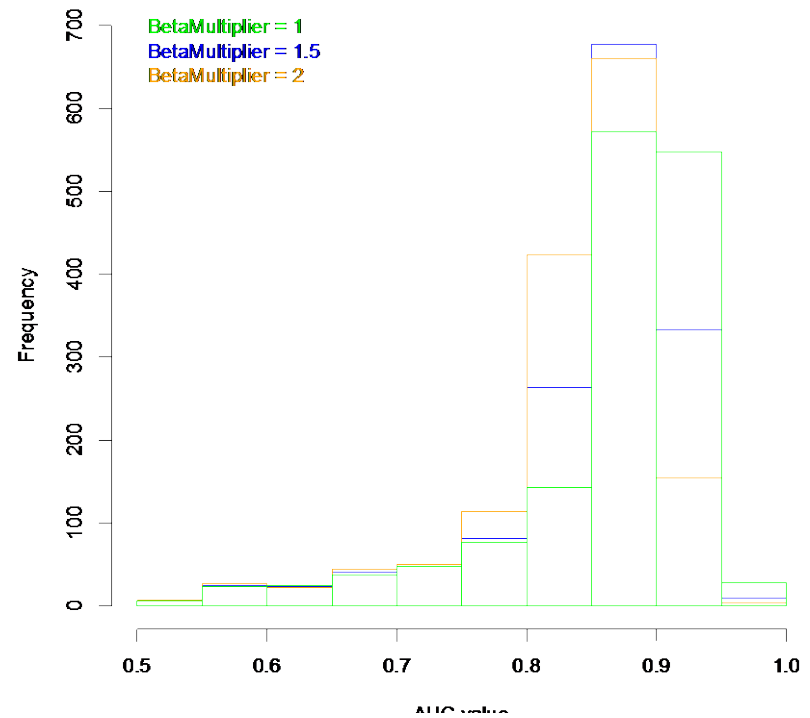
NO MASKS

Yearly Models – AUC values



MASKS

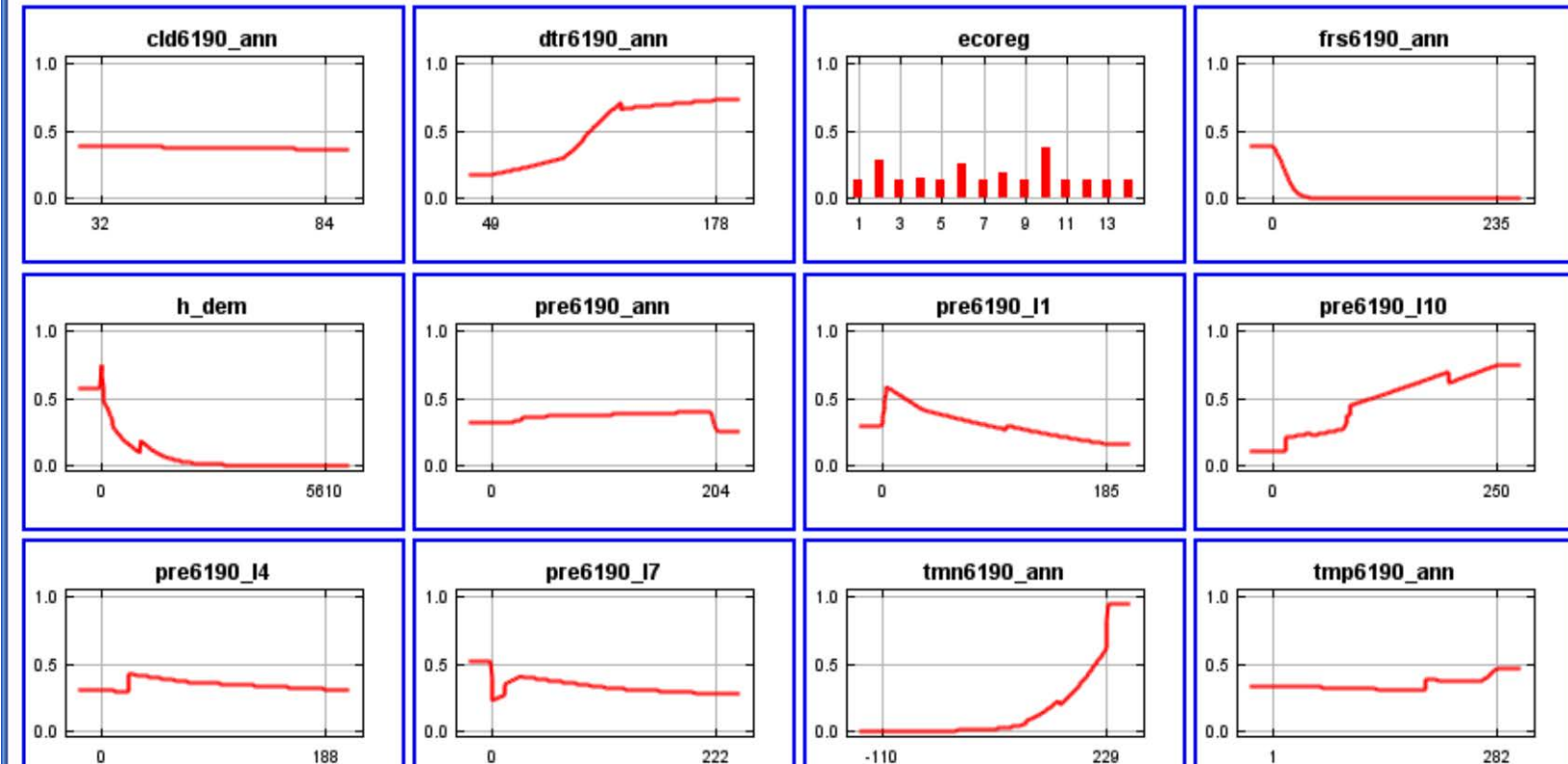
Yearly Models – AUC values



Parameters to explore:

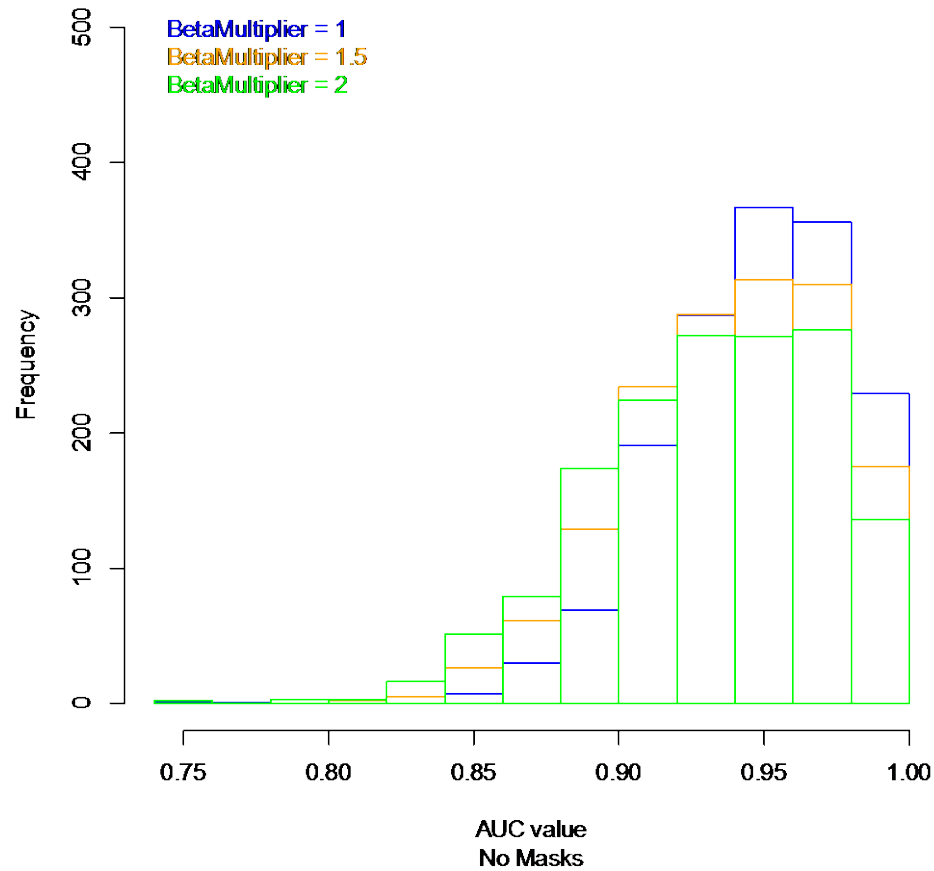
1. Using **yearly climate data** from time of collection to improve niche models using museum specimens
2. Using **reduced area** to train the models
3. **Smoothing response curve** to mimic more realistic response of organism
4. Using **fewer background points** to increase computing efficiency

Beta parameter: smoothing environmental response

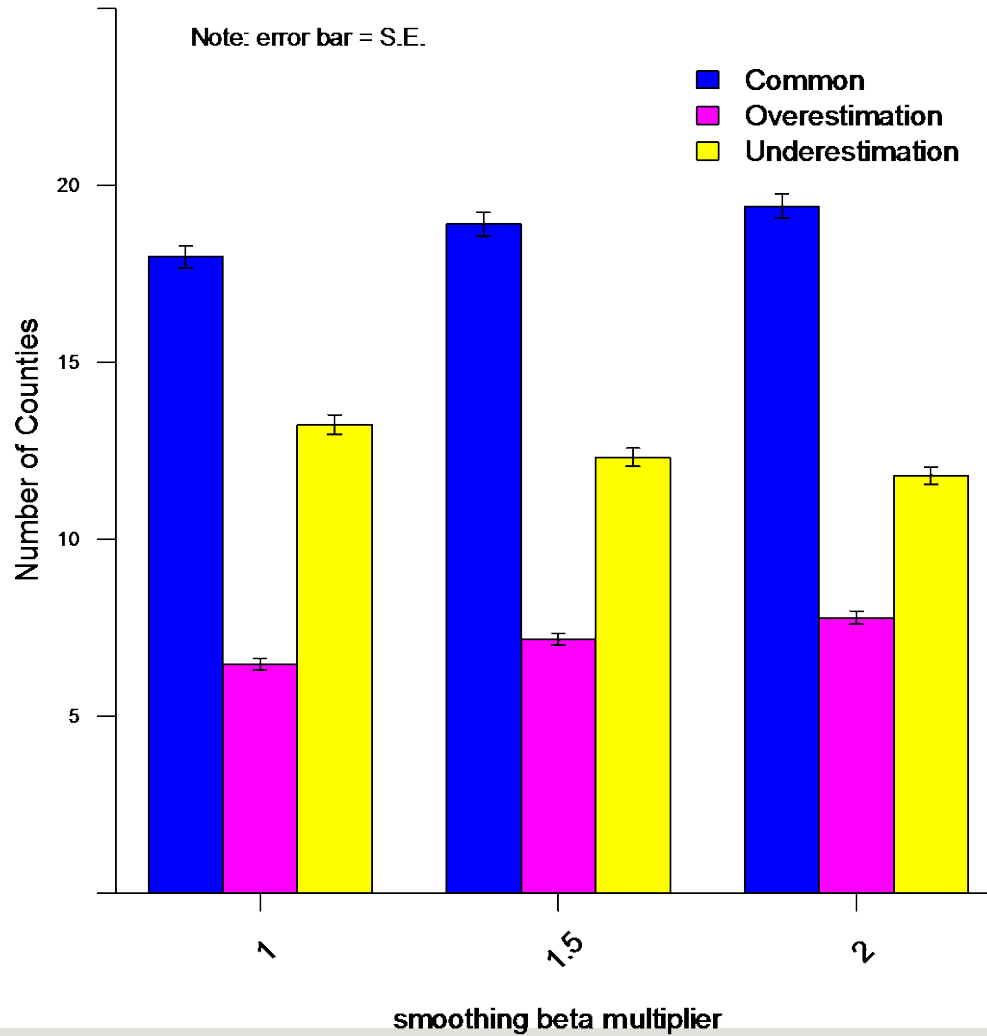


AUC values for different beta

Yearly Models – AUC values



Model performance



Parameters to explore:

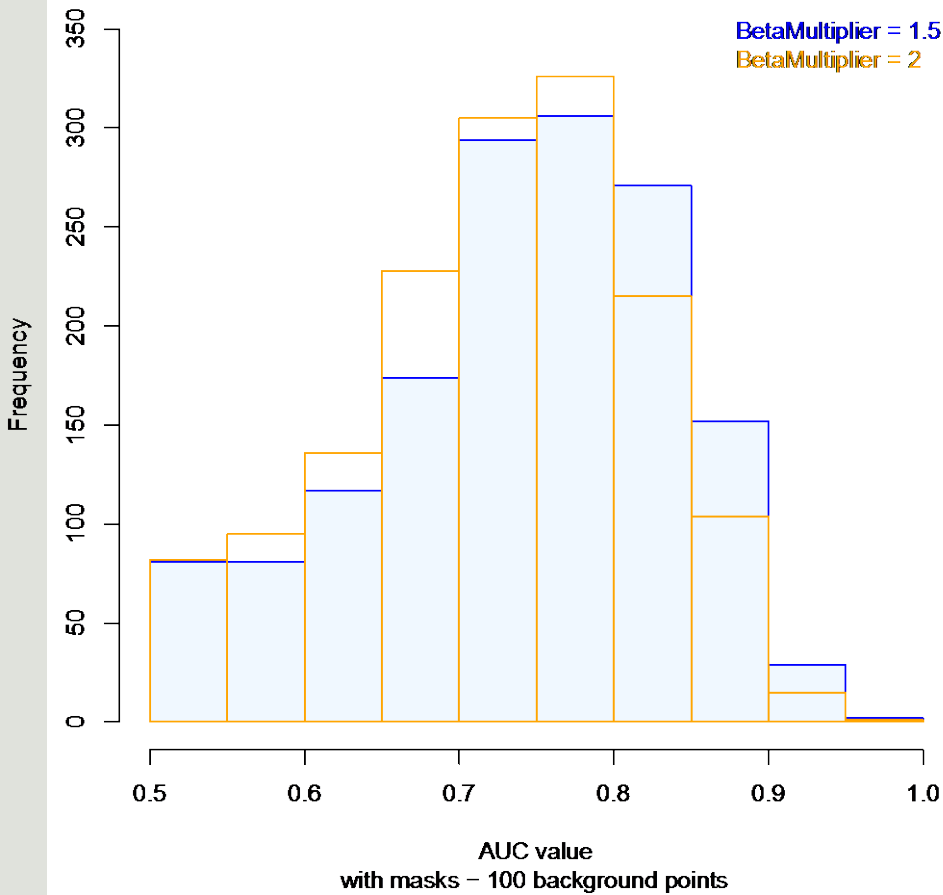
1. Using **yearly climate data** from time of collection to improve niche models using museum specimens
2. Using **reduced area** to train the models
3. **Smoothing response curve** to mimic more realistic response of organism
4. Using **fewer background points** to increase computing efficiency

Importance of the number of background points

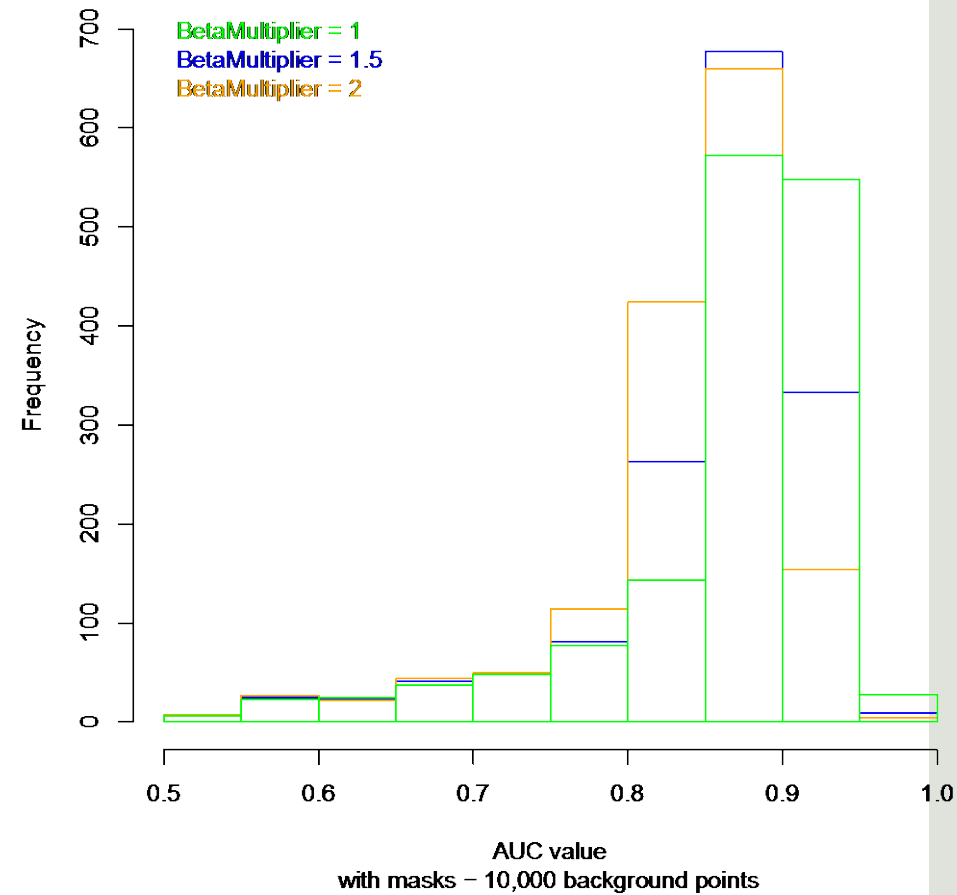
- Pseudo absences are ‘soft’ absences:
 - for the evaluation of the model
 - not the fitting
- Fewer background points
 - assign a higher probability to each cell
 - Hard to distinguish regions of high probability
- Large number of background points is computationally more demanding

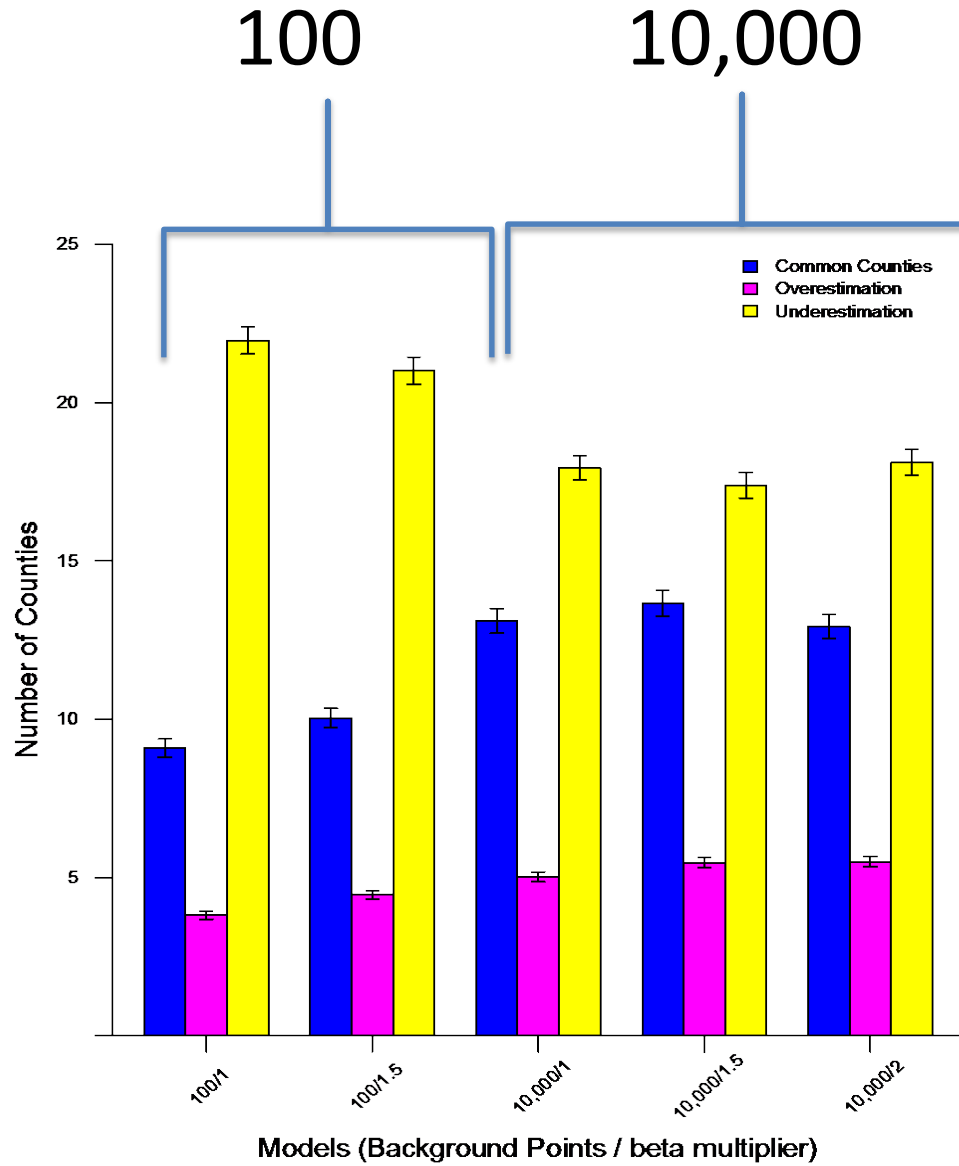
AUC

Yearly Models – AUC values



Yearly Models – AUC values



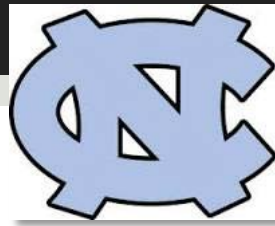


Parameters to explore:

1. Using **yearly climate data** from time of collection to improve niche models using museum specimens
2. Using **reduced area** to train the models
3. **Smoothing response curve** to mimic more realistic response of organism
4. Using **fewer background points** to increase computing efficiency

CONCLUSIONS

- Yearly Climate Data:
 - Annual, small distribution performs better
 - Perennials and/or widespread doesn't
 - Use of Masks:
 - Improves models, refines the local delimitation
 - Mask has to be carefully chosen
 - Smoothing parameter
 - Test with a subset of species to determine
 - Number of background points:
 - Worth the extra time !!!
- ✧ Think about why you are modeling the species niche (fundamental or realized?)
- ✧ Beware of statistical performance vs. biological accuracy



THANK YOU !!!!



THANK YOU !!!



www.idigbio.org



facebook.com/iDigBio



twitter.com/iDigBio



vimeo.com/idigbio



idigbio.org/rss-feed.xml



<webcal://www.idigbio.org/events-calendar/export.ics>