

Continuing Project

Assessing and mapping regional variation in potential site carrying capacity

CAFS 19.76

Dr. Mark Kimsey, University of Idaho

Dr. Aaron Weiskittel, University of Maine

Dr. Rachel Cook, North Carolina State University

Dr. Cristian Montes, University of Georgia

Dr. Douglas Mainwaring (OSU)

Dr. Eric Turnblom, University of Washington

Presenter: Jaslam Poolakkal, Dr. Mark Kimsey



Justification

- ❑ Understocked stands underutilize site resources and will not reach maximum potential productivity
- ❑ Overstocked stands are slow to develop and susceptible to wildfire, drought and insect outbreaks due to competition for limited resources.
- ❑ To date, forest carrying capacity research is regionalized, utilizes multiple modeling approaches, and not universally available spatially across the US

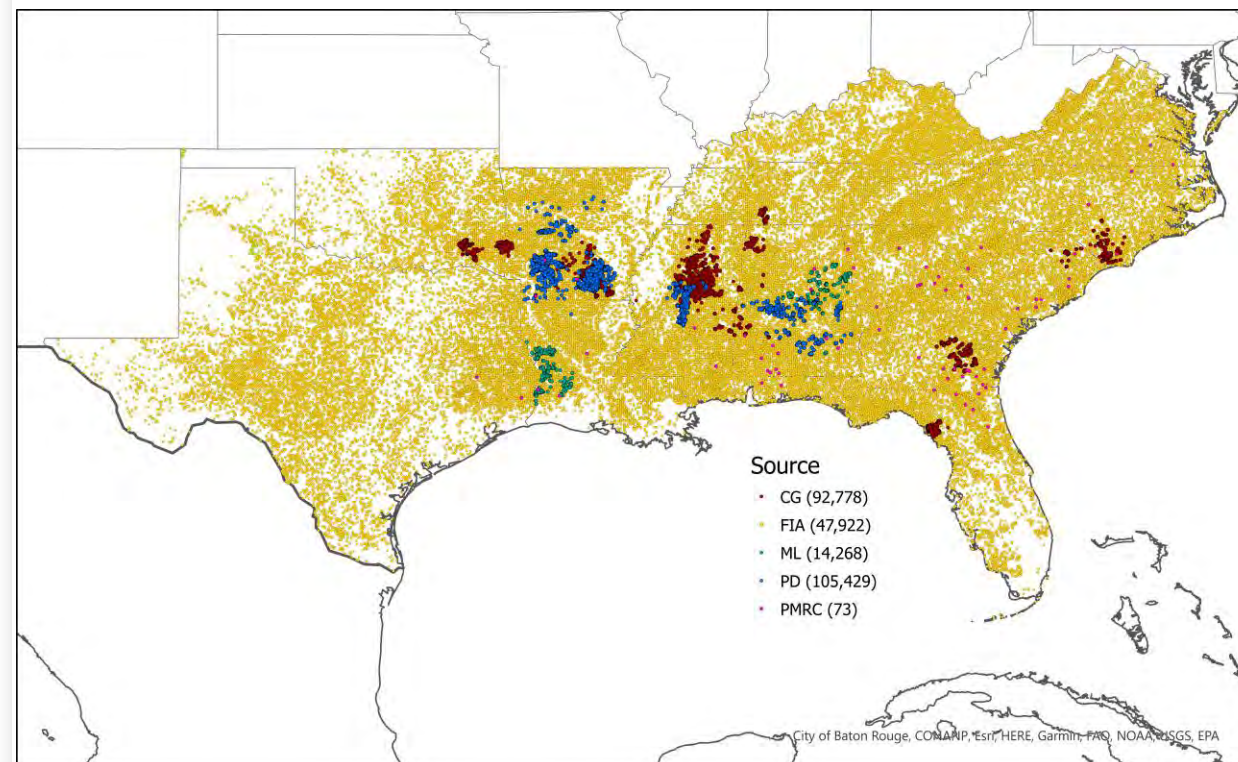
Objectives

- 1) Synthesize a nationwide forest inventory database from publicly available data and from CAFS members,
- 2) Standardize maximum carrying capacity modeling, and
- 3) Create efficiencies for multi-regional forest management organizations by providing consistent, species-site-silviculturally sensitive, wall-to-wall spatial models of SDI_{max} for commercial species of the United States.



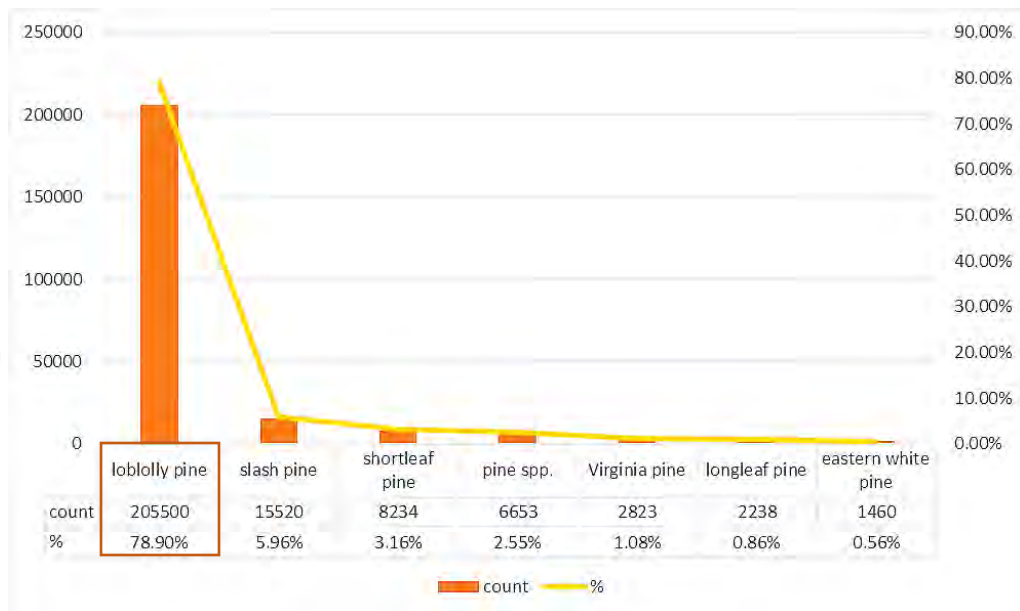
Data

- QMD , TPA , Species BA proportions and Silvicultural info of each record
- Topography extraction from 30m DEM – 10 Variables (e.g., Slope, Aspect, Topographic wetness index, Solar radiation)
- ClimateNA – 128 Variables (Annual, Month, Season)
- Geology and Soil layer - 56 Variables (SGMC & gSSURGO geodatabase) –
- Enhanced soil parent material database by FPC
(e.g., Depth to increase in clay content, Drainage class, Nature of surface soil modifier, Nature of subsoil (mineralogy) modifier)

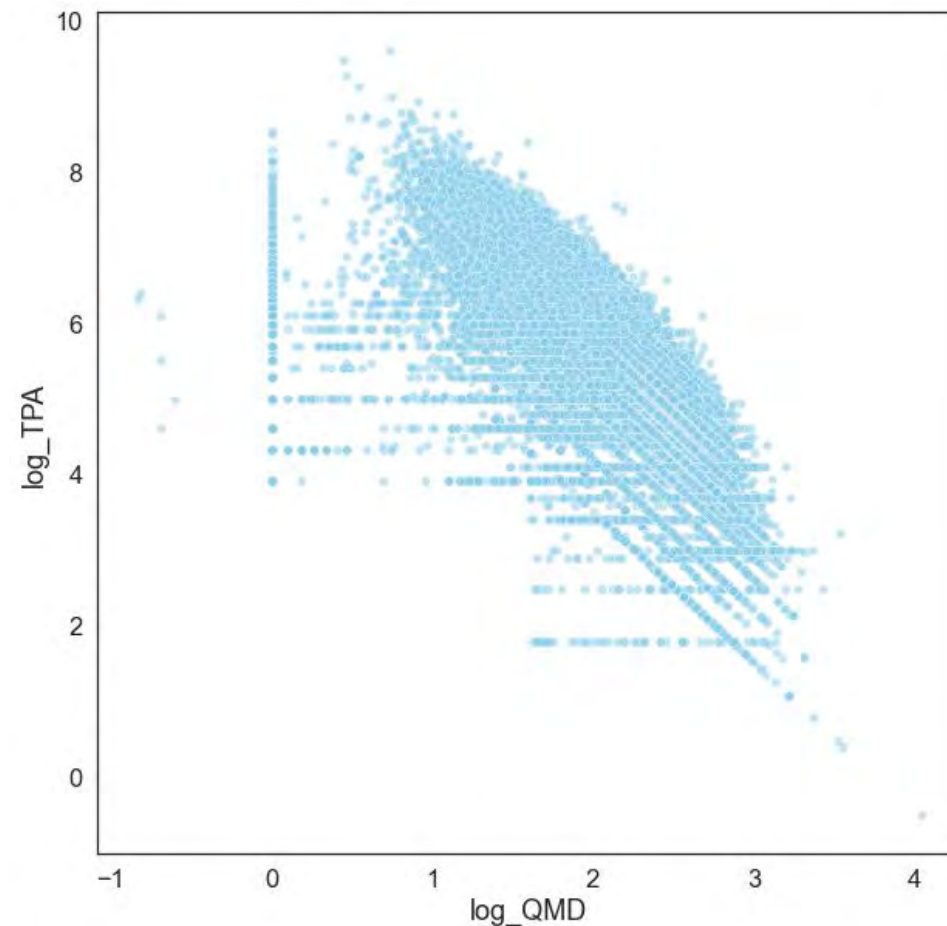
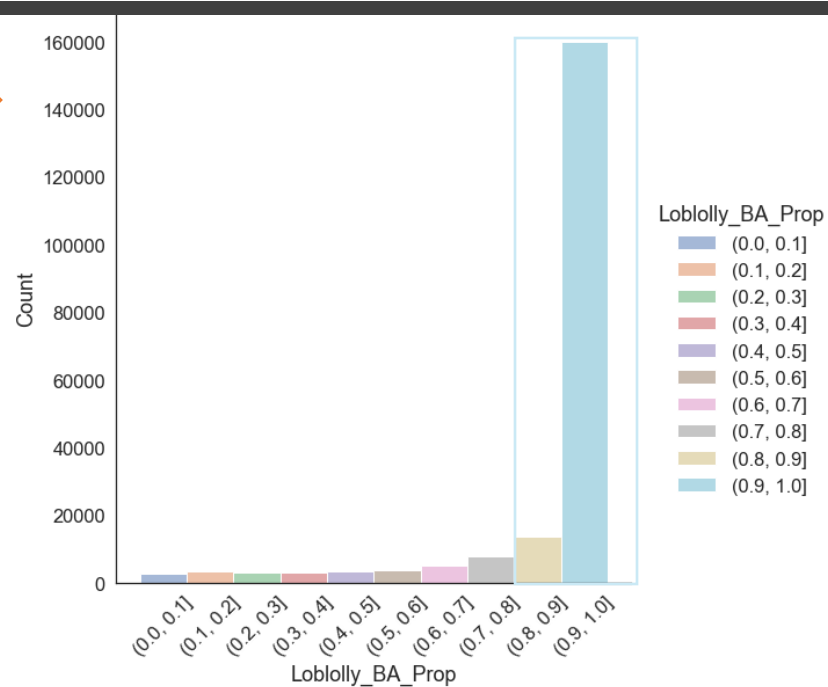


260,470 initial records from the forest land in the southern region.





174,370(84.8%)
records with
0.8-1.0 BA
proportion of
Loblolly Pine



Loblolly Pine Modeling Approach

☐ 165,313 plots with BA of 0.8 to 1.0

Data Cleaning:

Missing expansion factors, at least 10 TPA, QMD at least 2-inch, questionable & missing data

Quantile Regression:

Linear Quantile Mixed Models to determine the quantile line of $\log(\text{TPA}) \sim \log(\text{QMD})$

Variable Selection:

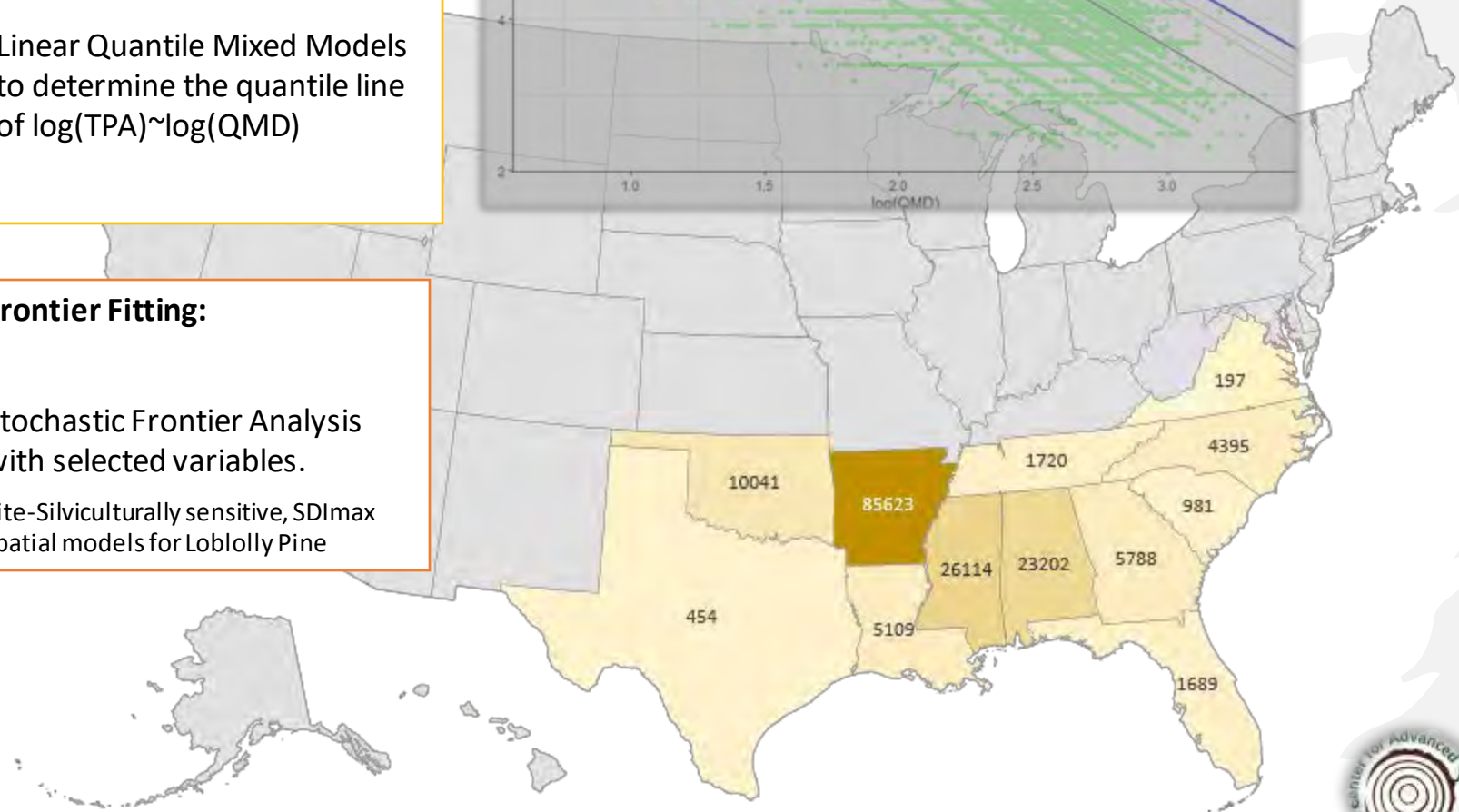
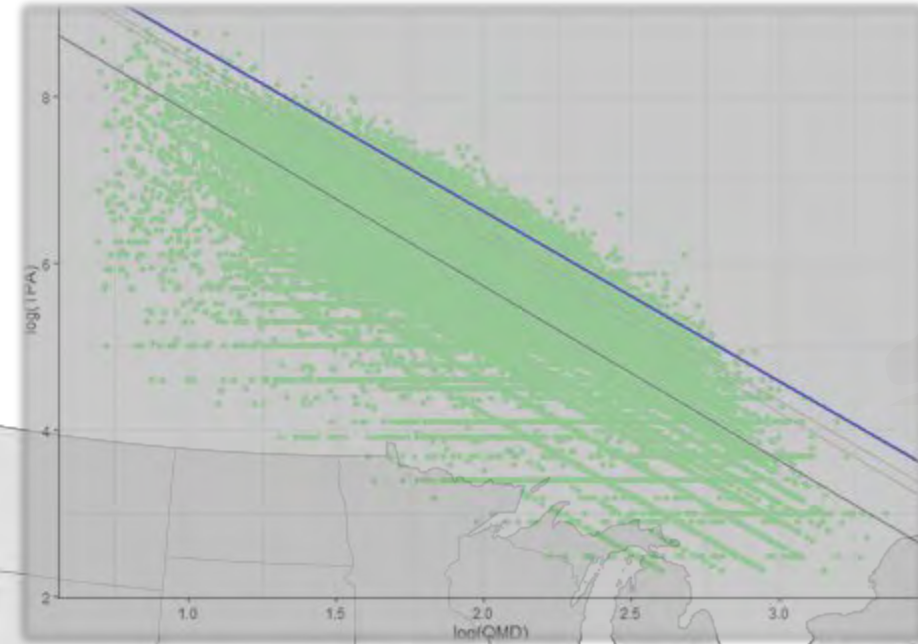
Featurewiz for feature selection and extraction

Based on Minimum Redundancy Maximum Relevance (MRMR) algorithms

Frontier Fitting:

Stochastic Frontier Analysis with selected variables.

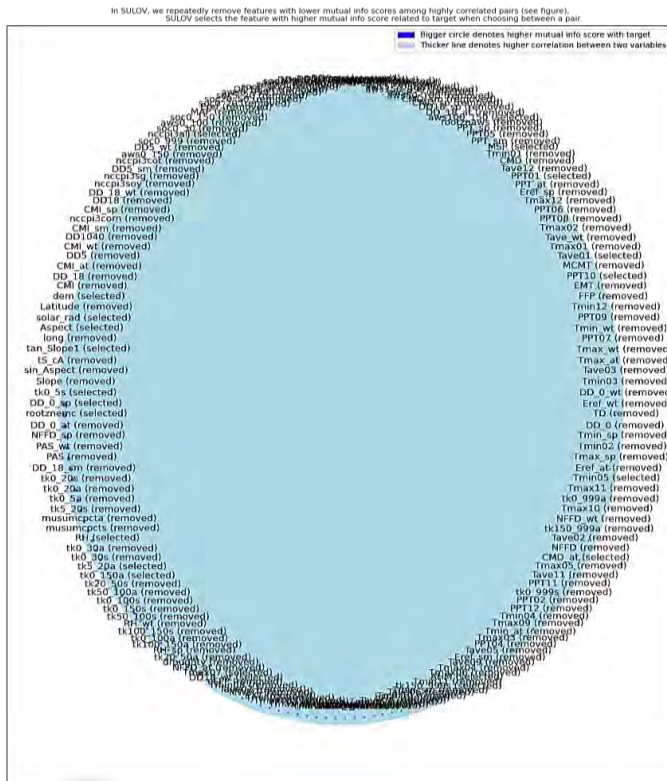
Site-Silviculturally sensitive, SDImax spatial models for Loblolly Pine



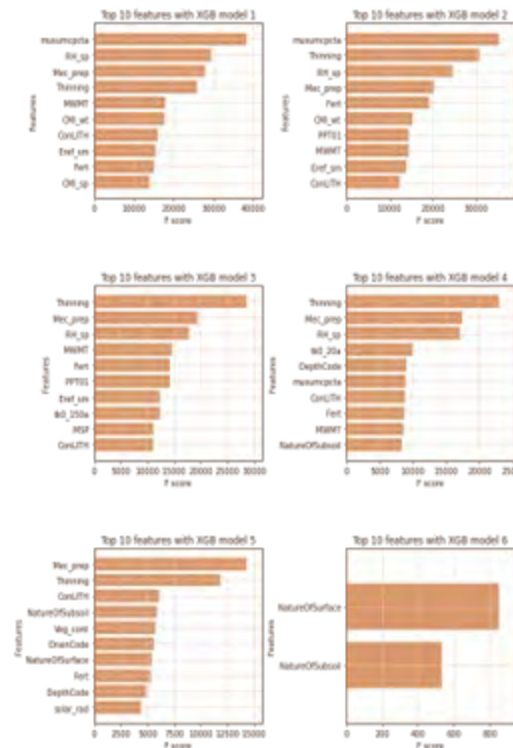
Variable Selection

featurewiz performs feature selection in 2 steps.

1. SULOV algorithm



2. Recursive XGBoost



Stochastic Frontier Regression

An initial model of predicting $\ln(\text{TPA})$ with the single variable $\ln(\text{QMD})$ served as the base model.

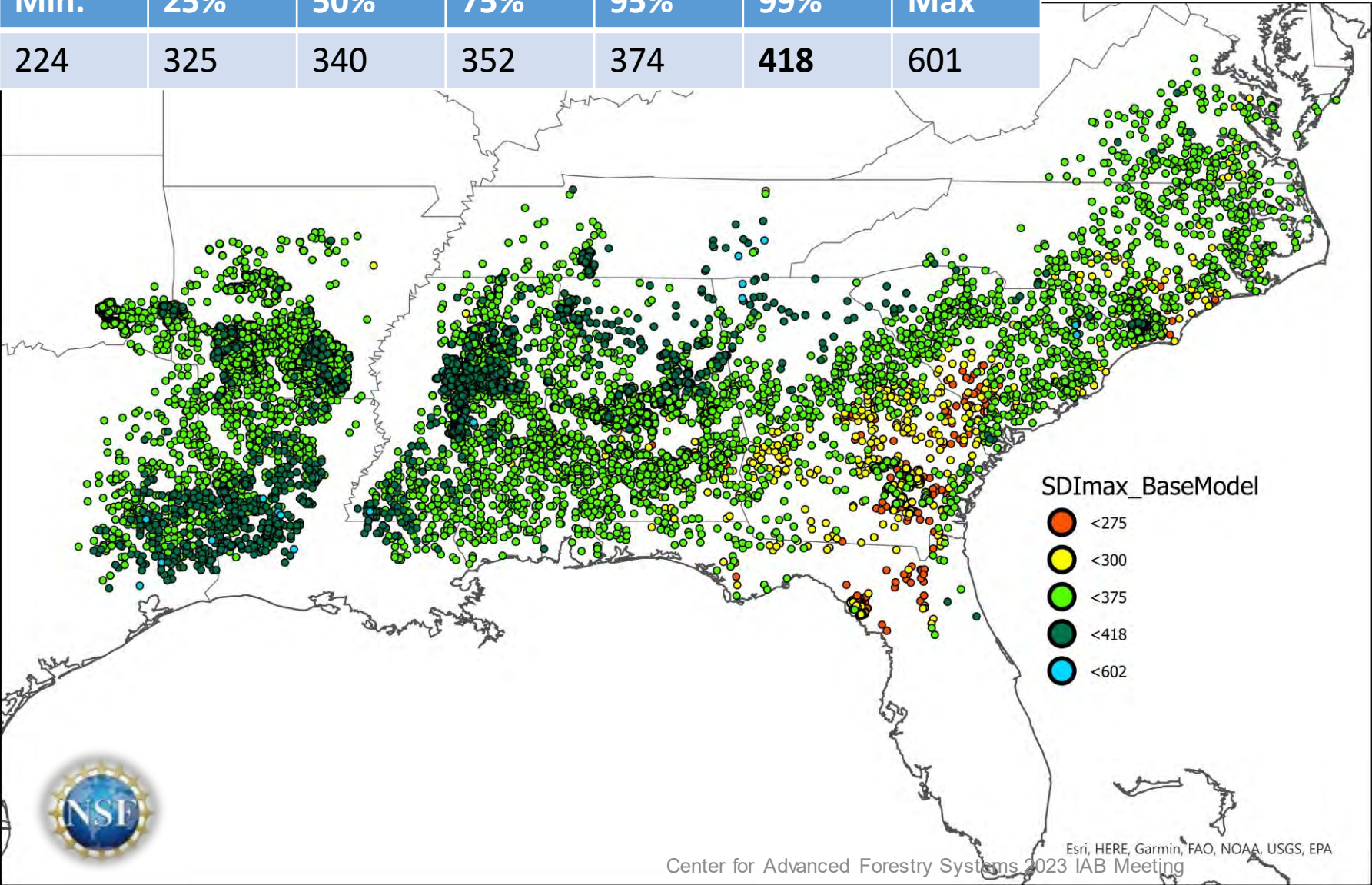
Then, through an iterative process, variables were introduced to the model and kept or removed based on variable significance and AIC score

- ☐ Topographic – Topographic wetness index, Cosine of Aspect, Tan of Slope
- ☐ Climatic - Autumn Hargreaves climatic moisture deficit, winter degree-days above 18°C, winter & april precipitation , May to September precipitation (mm),
- ☐ Geology and Soil – Consolidated lithology , Available water storage estimate in standard zone 1 & 6, FPC_DrainCode, FPC_DepthCode, FPC_Nature of Surface, FPC_Nature of Subsoil,
- ☐ Silvicultural info: Interaction variable of Fertilization and Thinning



SDImax for loblolly pine: No silviculture treatments implied or stated in given inventory records or shown in map – base model

Min.	25%	50%	75%	95%	99%	Max
224	325	340	352	374	418	601



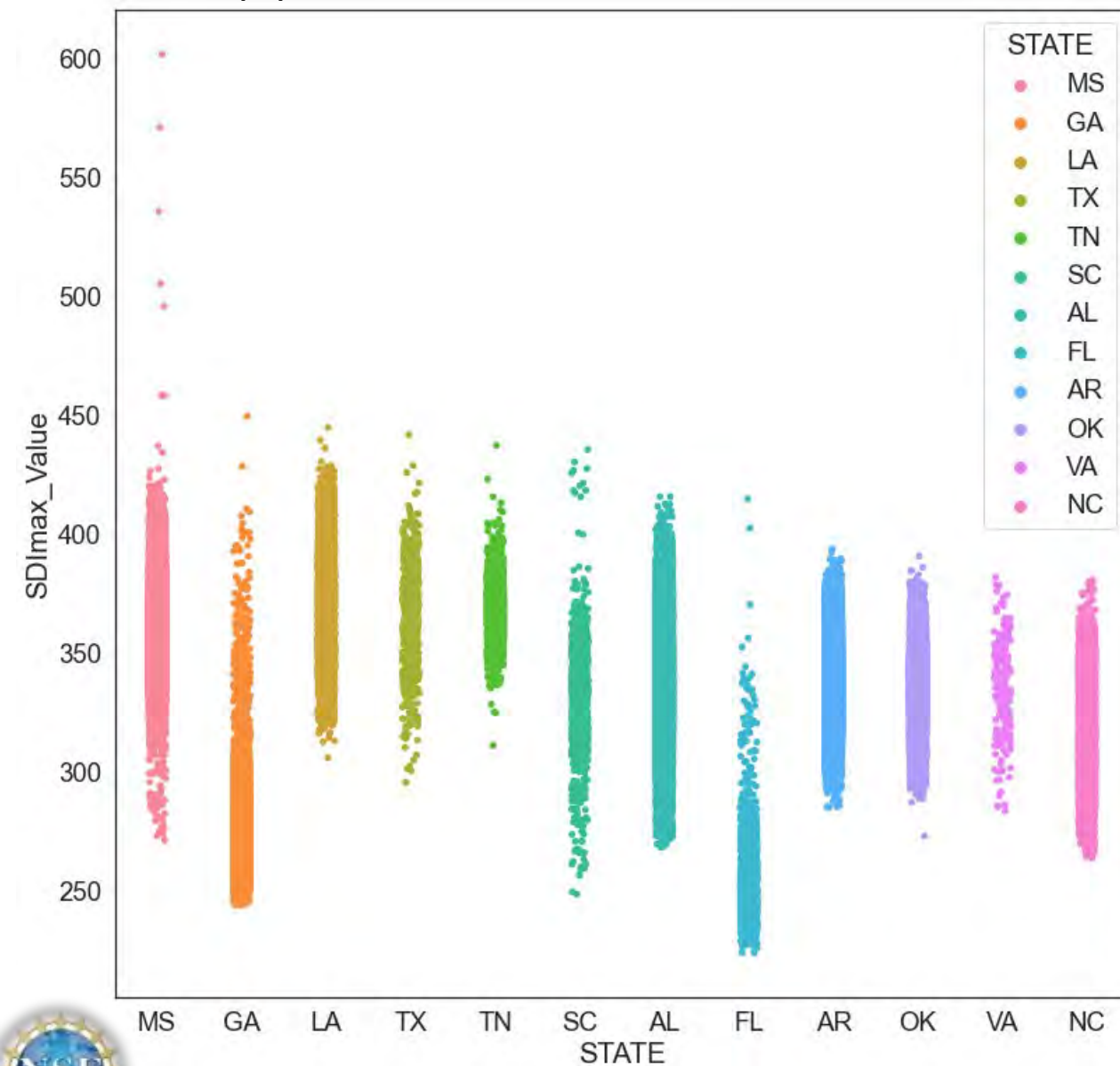
Major Findings

Comparative regional loblolly pine SDImax citations

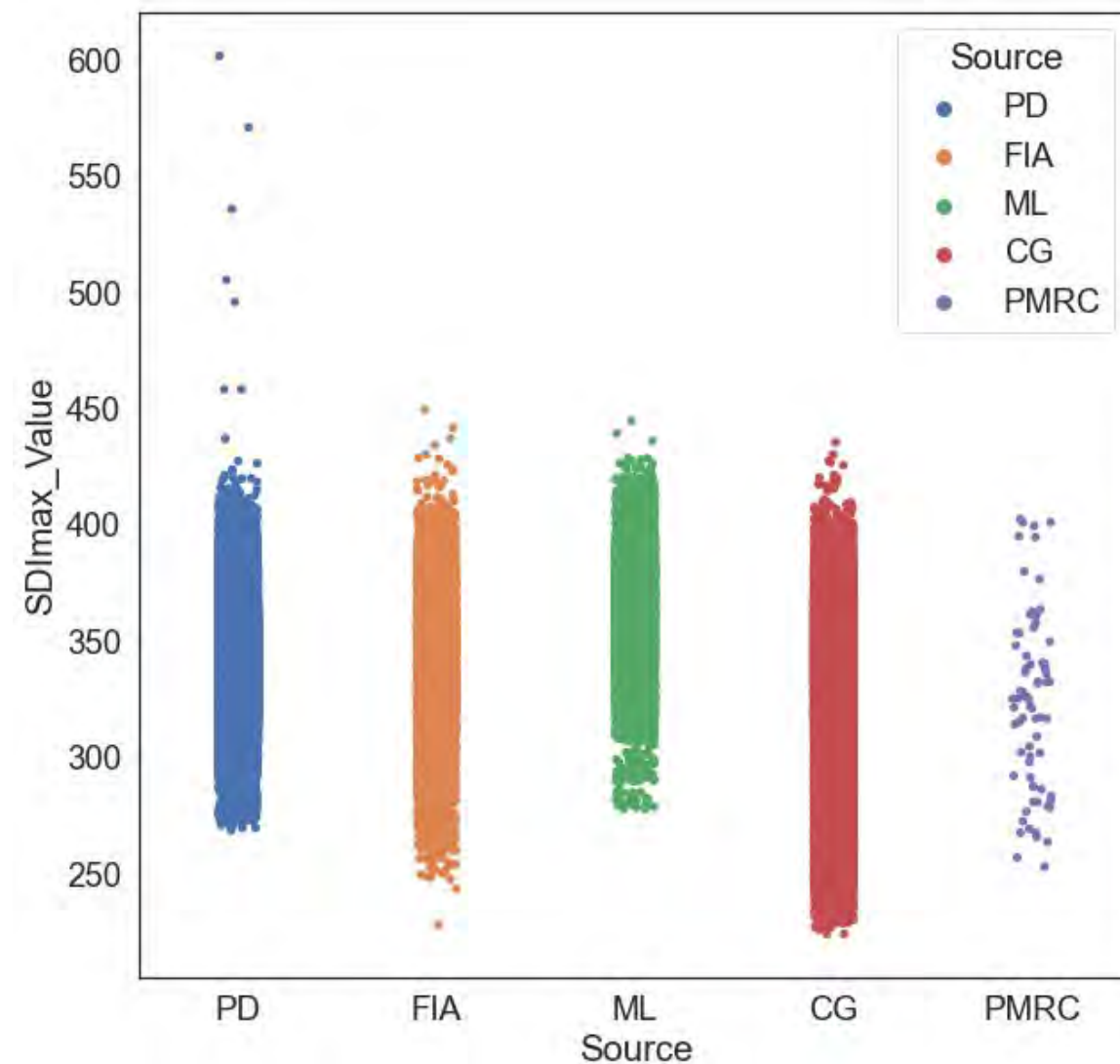
Max SDI	Source
243-571	Zhao et al,2020
400	Williams,1996
450	Dean et al,1993



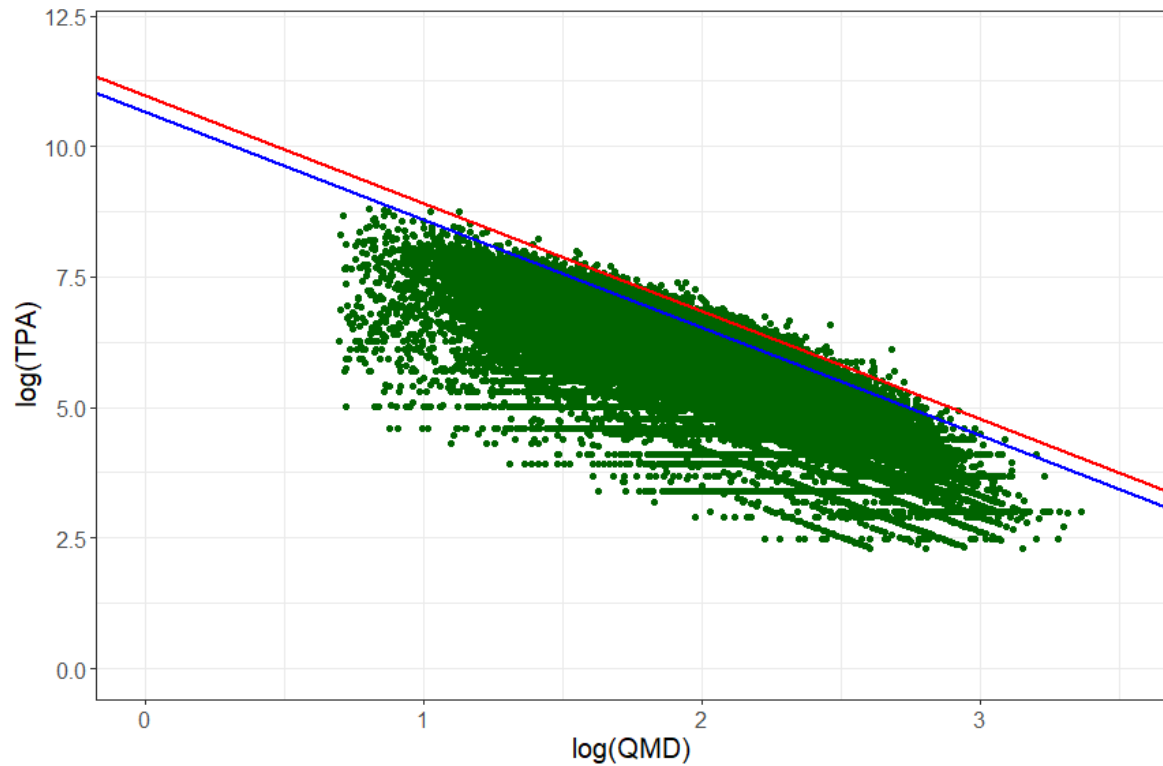
Strip plot: SDImax values of Southeast States



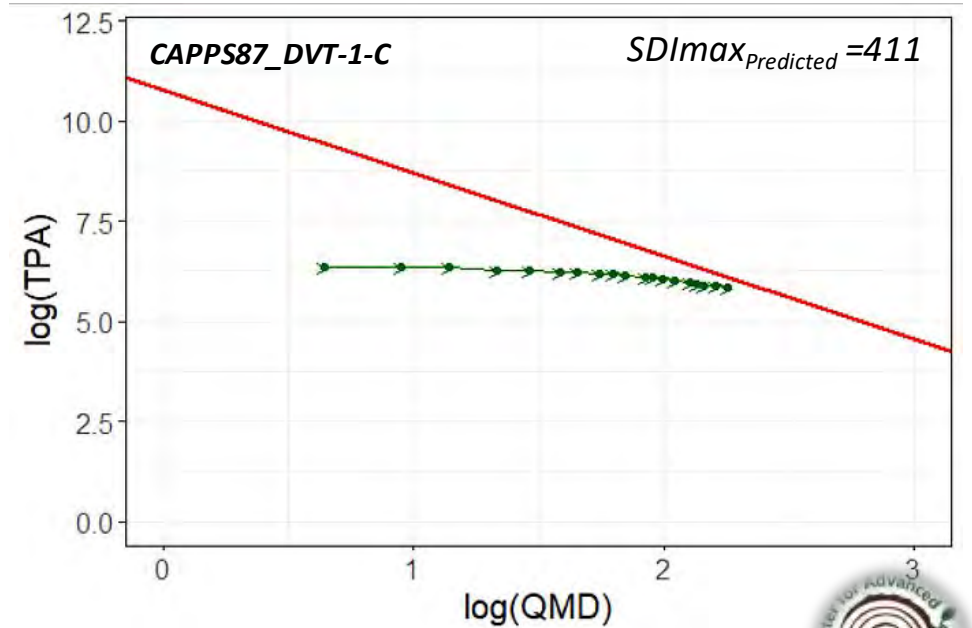
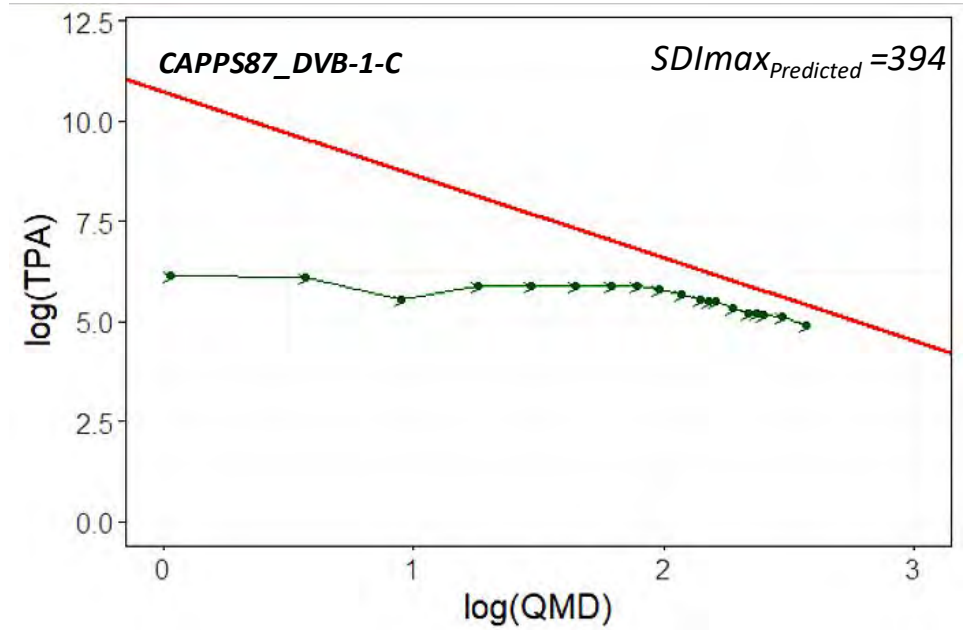
Strip plot: SDImax values of Data Sources



Evaluation of base model performance (i.e., no known silviculture influence on SDImax)

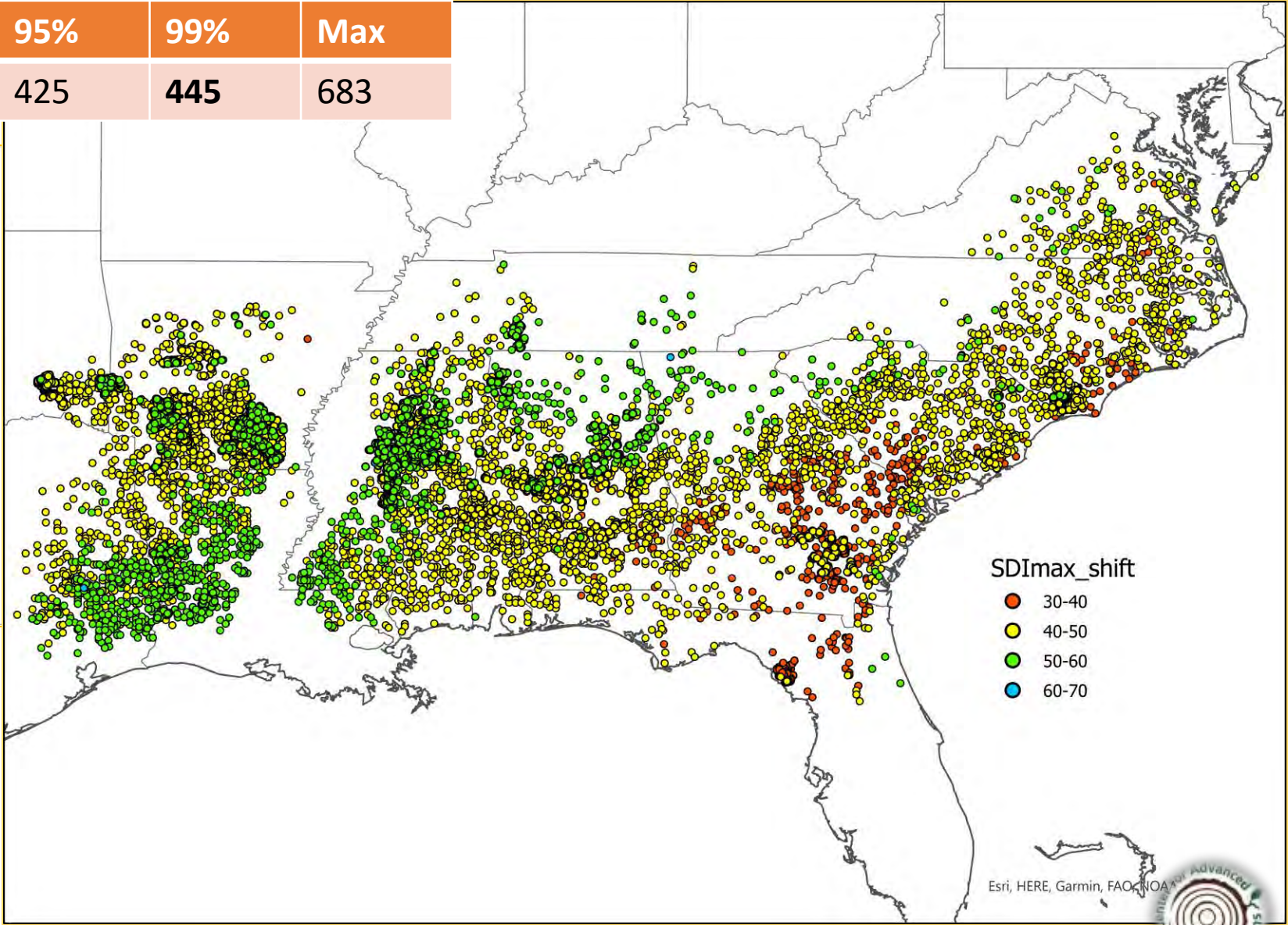
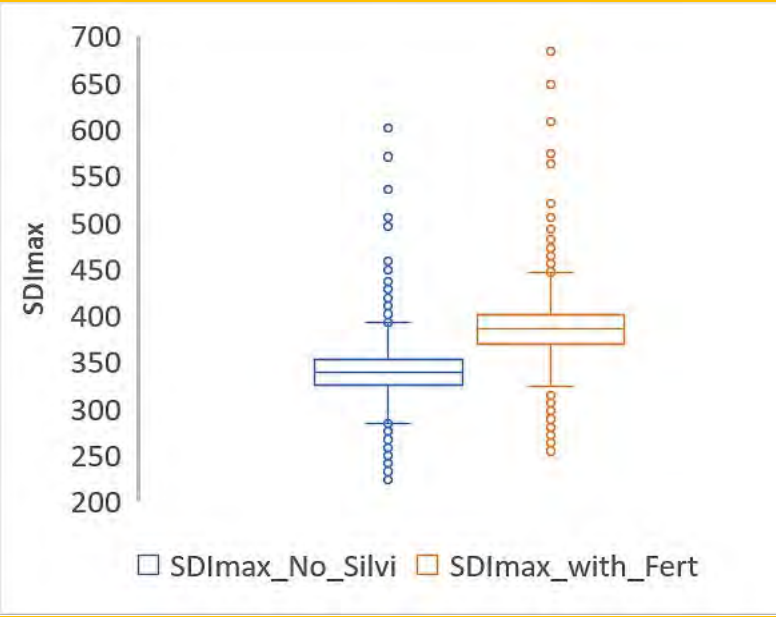


Models perform relative to the 90th (red line) and 50th percentile (blue line) of site growing conditions



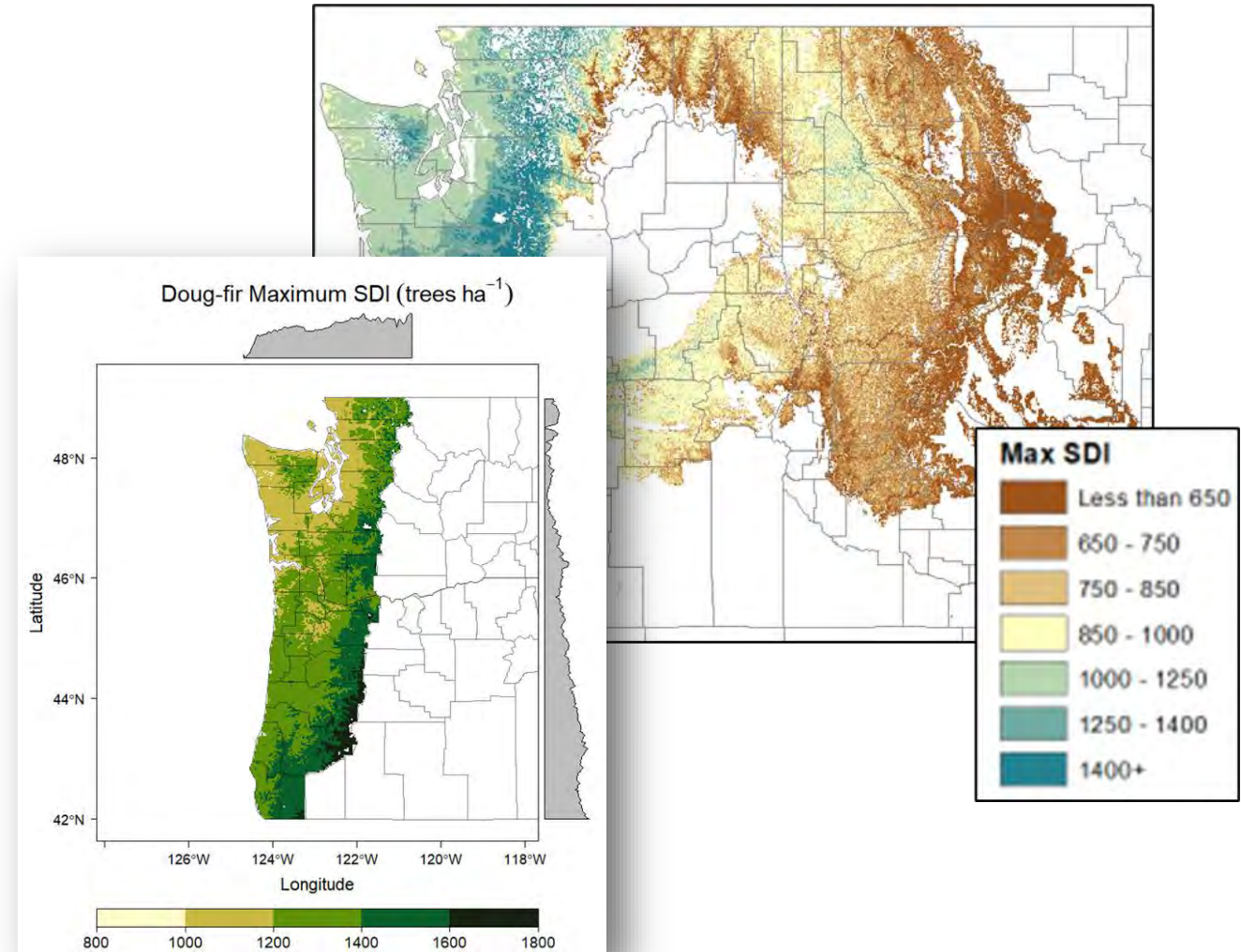
SDImax for loblolly pine using base model with fertilizer application

Min.	25%	50%	75%	95%	99%	Max
254	370	386	384	425	445	683



The primary deliverables will include:

- 1) Harmonized dataset for major commercial species throughout US forestlands;
- 2) Machine learning models relating SDImax to various species functional traits, stand attributes (e.g., structure, diameter distribution, site index), and environmental factors (e.g., soils, topography, climate);
- 3) High-resolution (10-30 m) raster maps of predicted SDImax for use in multiple digital platforms;
- 4) Annual progress reports and presentations and a final report and presentation + peer-review publications and conference presentations.

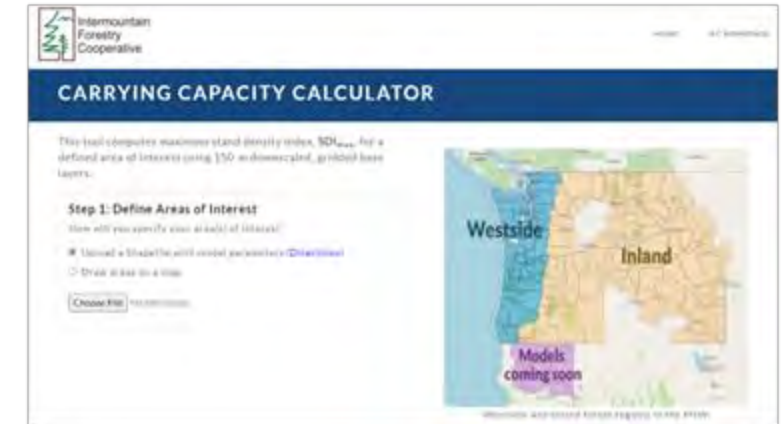
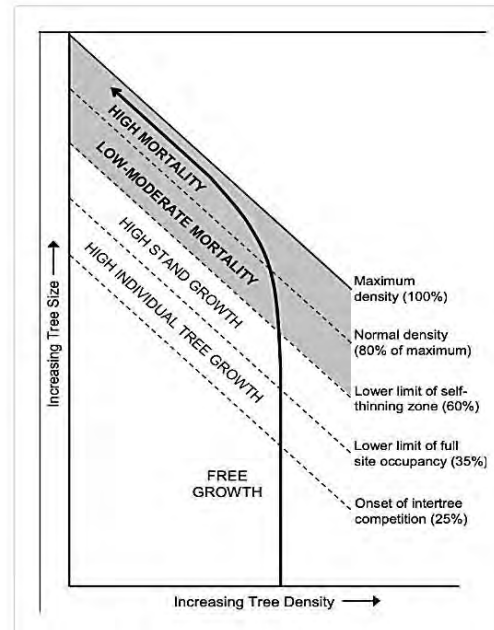
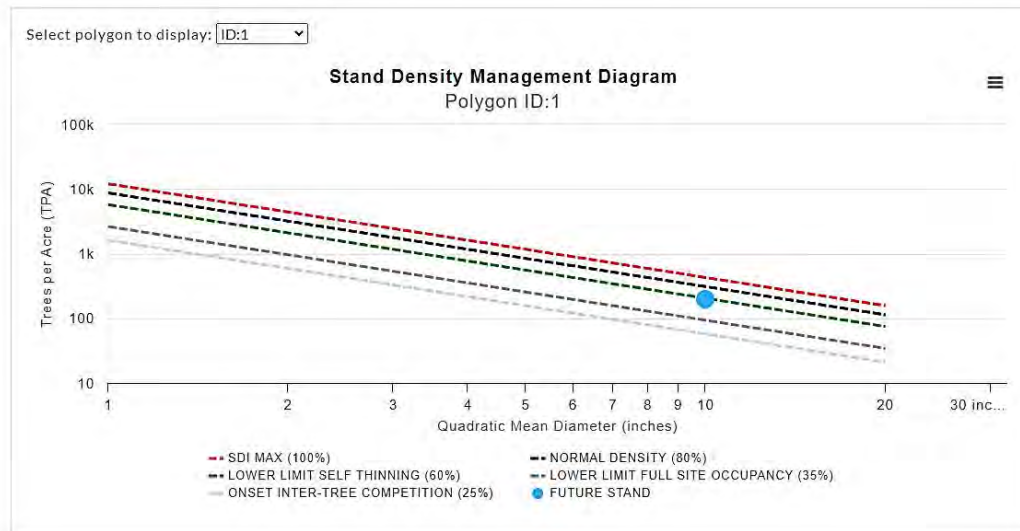


Company Benefits

❑ Identification of optimal planting or thinning residual densities as a function of:

- Species composition
- Site resources
- Silvicultural treatments

❑ Consistent methodology/platform for identifying and managing forest density across multi-regional land holdings



Summary

