### **New Project**

### A Neural Network Approach to Generating Leaf Area Index Estimates Using the Sentinel-2 Satellite Record

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Andrew Trlica presenting





### LAI Can Inform Forestry Decisions

- Need for an Accurate, Flexible, Accessible tool for LAI estimation
  - Current approaches pose barriers to forest practitioners
- Canopy LAI is critical for stand and sub-stand level management
- Understory LAI is critical for vegetation control decisions





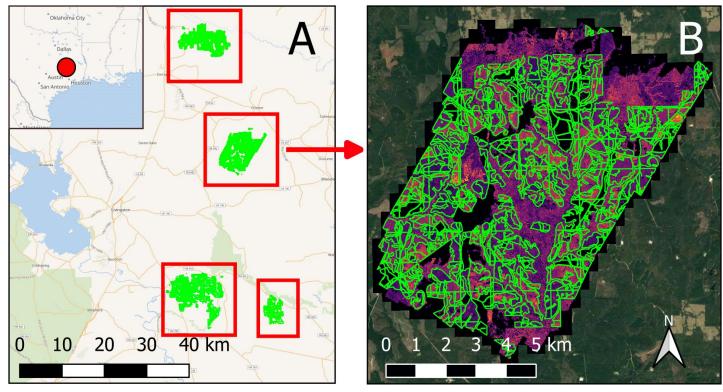
# Objectives

- Operationalize web interface with loblolly pine canopy "machine-learning" LAI model (MLAI)
- Expand to other regions and species for a national level LAI model for production forests
- Develop understory model to run in parallel with overstory model
- Use LAI model to develop potential productivity and response maps in conjunction with soils and climate data





### Predicting LiDAR LAI from Sentinel-2



A: Example Loblolly pine plantation boundaries provided by coop member

B: Corresponding LIDAR Canopy LAI, January 2018





# Current Standard LAI Estimation

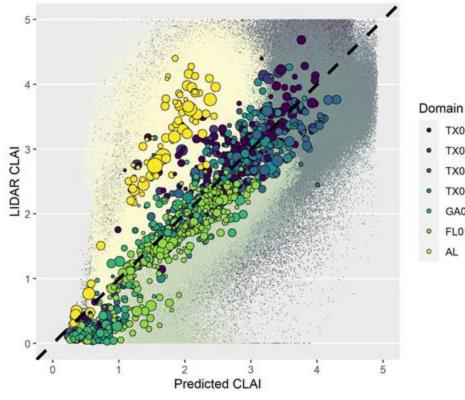
TX0102

TX03 TX04

TX05 GA010203

FL01

#### SR LAI model



Sentinel-2 Simple Ratio, Loblolly pine, 10 m

- Sentinel-2 (10-20 m) or Landsat (30 m)
- Linear model of Simple Ratio (SR)
- Accuracy (RMSE):
  - 0.78 (pixel)
  - 0.58 (whole stand)
- Requires manual data fetching+processing





### Machine Learning for Forest Remote Sensing

 Cloud data and distributed computing allows for new approaches

Current work:

- 1. Custom built Artificial Neural Net in TensorFlow
- 2. LiDAR from members & S2 data via Earth Engine for training
- 3. Google AI Platform hosts trained model to make predictions for any place/time with S2 coverage







# Deep Learning Approach

- Introducing MLAI: Artificial Neural Network for LAI estimation in plantation forests
- Uses information from all 9 bands
  + pixel neighbors → greater match to LiDAR
- S2 needs no upfront processing, just specify date(s) and place(s)
- 10 predictions + confidence estimate for each pixel

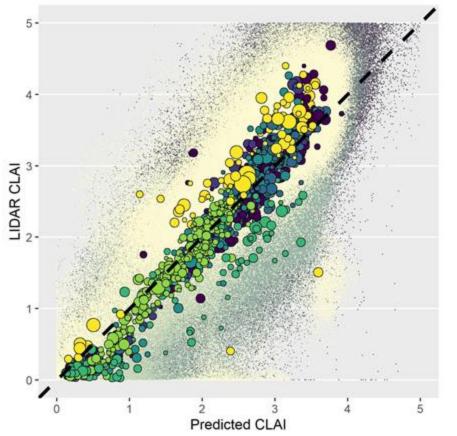






# The "MLAI" Model

#### MLAI ANN model



Sentinel-2 MLAI, Loblolly pine, 10 m

- Pixel RMSE: 0.56
- Stand RMSE: 0.38
- Applicable for Loblolly LAI anywhere on the globe, mid-2017 to present





Domain

TX0102

TX03

TX04 TX05

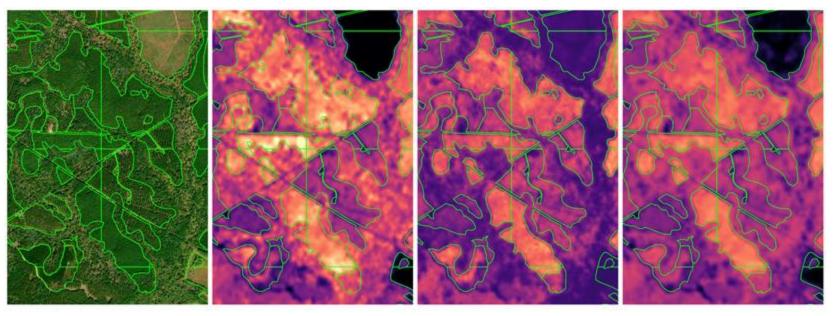
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Best-so-far accuracy

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### Map-to-map comparisons



True color



SR model, 10m

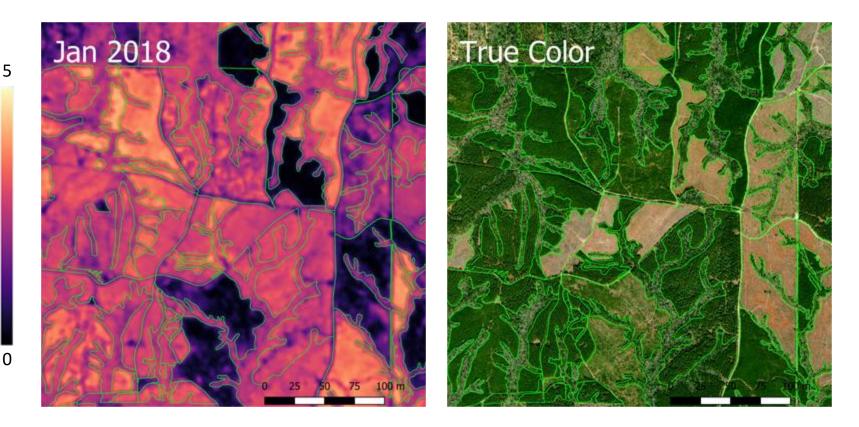
MLAI, 10m

#### Several ways to measure LAI — which one is "correct"?





### **Continuous monitoring**

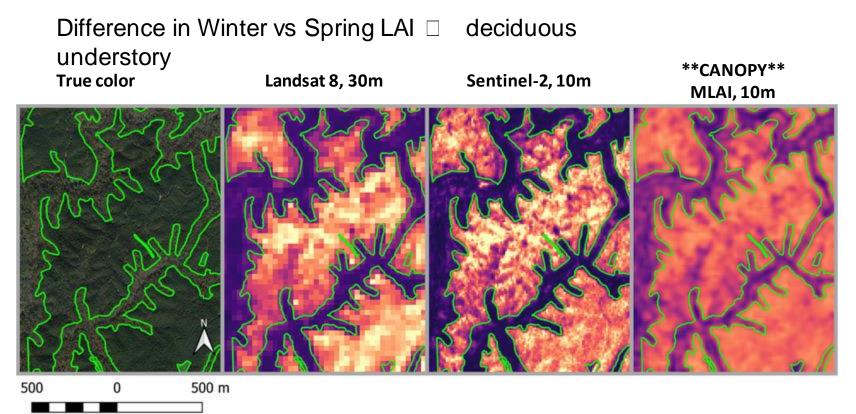


#### Loblolly plots in Jan 2018 LIDAR track, East Texas





## Understory identification



Loblolly LAI, Jan 2020, Alabama Coastal Plain

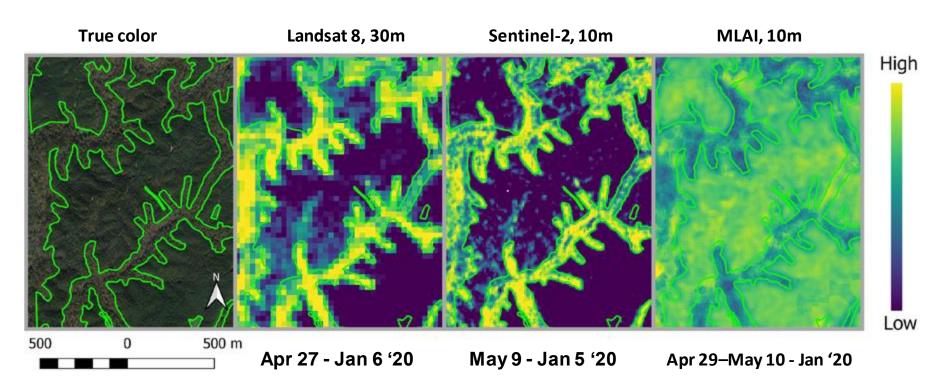




Center for Advanced Forestry Systems 2021 Meeting

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### **Understory identification**



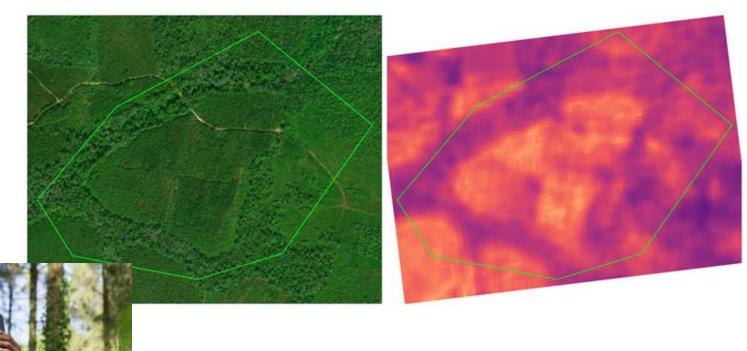
Little available springtime LiDAR at present to inform MLAI estimation





**Company Benefits** 

### On-demand forest information



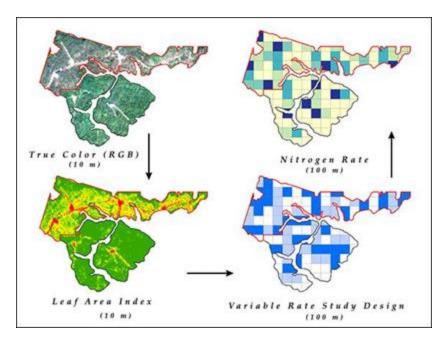
- In field, on site estimation (new windshield cruise)
- Identify stands + sub-stands to prioritize management





### Company Benefits Optimize management across land base

- Prioritize stands more efficiently for fertilization or vegetation control
- Optimize return on investment of management
- Learn from operational "experiments" on a landscape scale







Deliverables: Year 1

# Develop User Interface

- Web-based app to interface MLAI model
- Input place + time  $\rightarrow$  LAI map
- Develop underlying understory estimation process for parallel delivery







# Species-specific LAI models Year 1-2

- Douglas fir
- Slash pine
- Western larch
- NE conifers
- Flexible intake + training + prediction → Anywhere with enough LiDAR can form the basis of a new specialized model

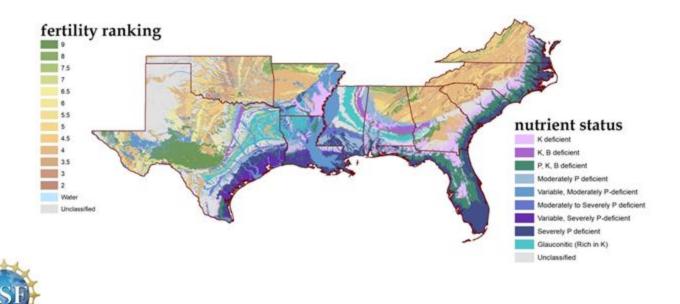






#### Deliverables Year 2-4 Create Site Specific Recommendations

- Overlay with soils map and operational responses over time to fine tune silvicultural prescriptions
- Model with soil and climate variables to make better predictions for future response and current conditions





#### Summary

### Moving towards data-driven forestry

- Accurate, Flexible, Accessible tools are in the wings to better use the growing body of forest information
- Need industry and university accumulated LiDAR+stand data from multiple regions to further develop models
- Will require continued IT support for updates, development, and maintenance





