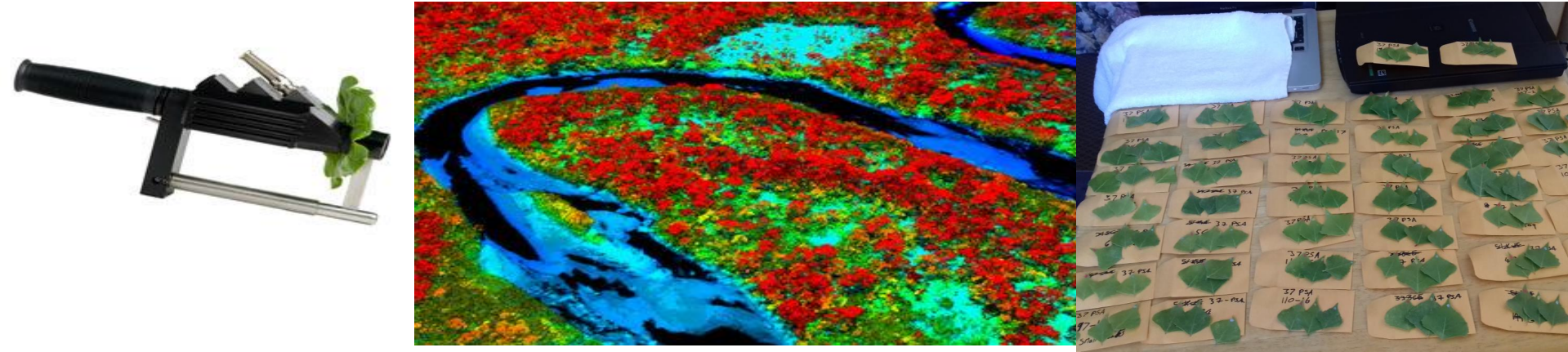


# High throughput phenotyping at the cottonwood common gardens



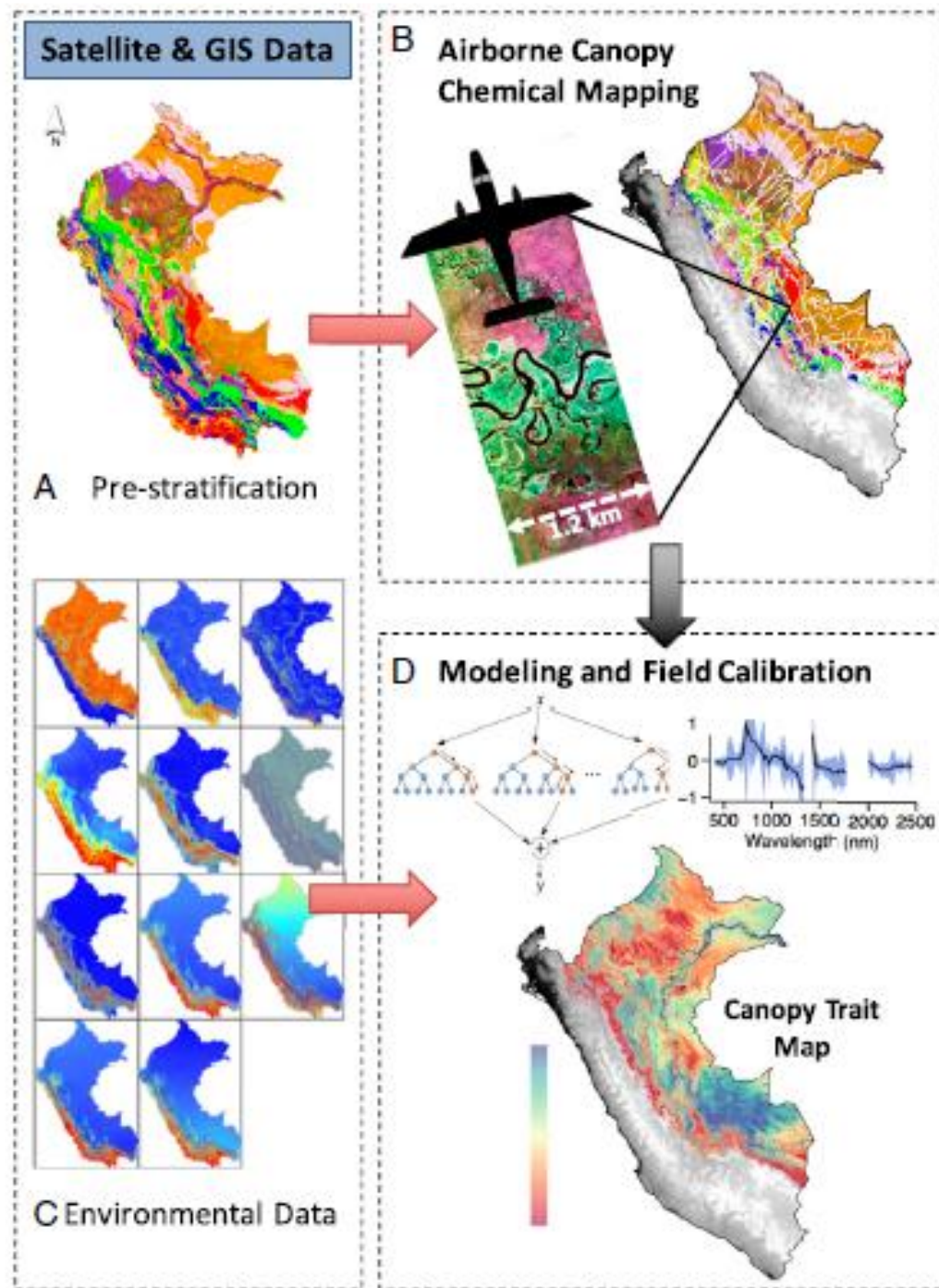
Chris Doughty and Eleanor Thomson



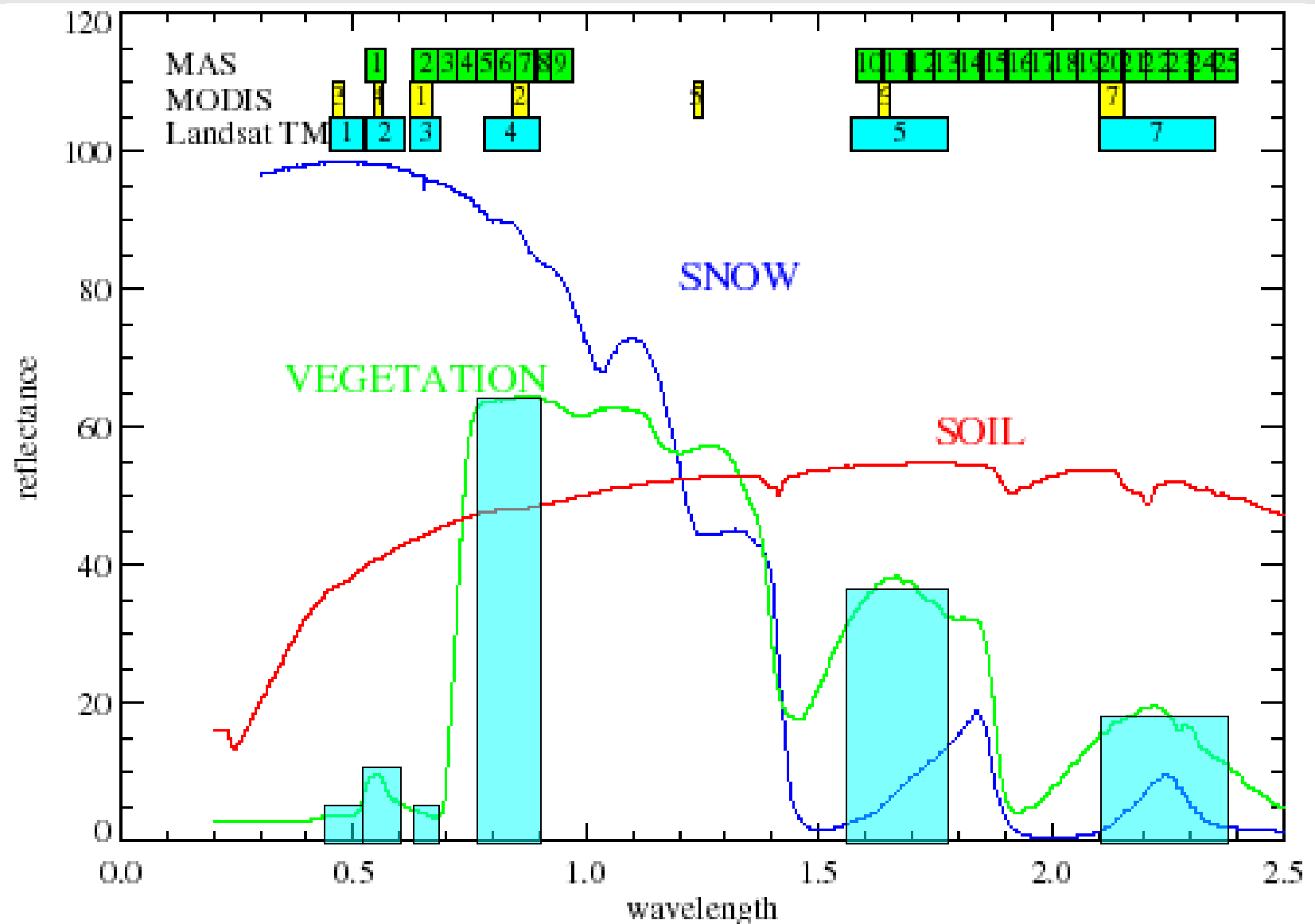
School of Informatics,  
Computing, and  
Cyber Systems



School of Informatics,  
Computing, and  
Cyber Systems



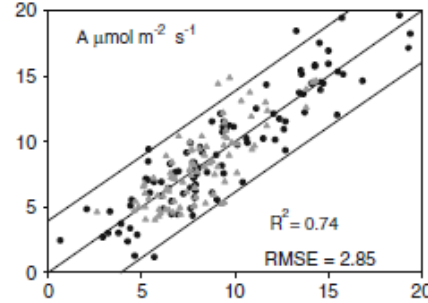
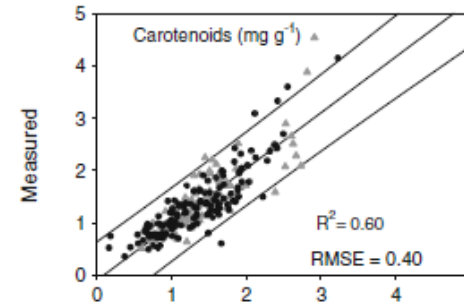
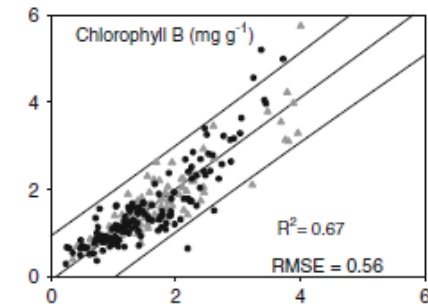
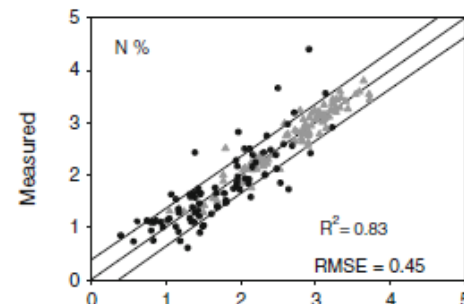
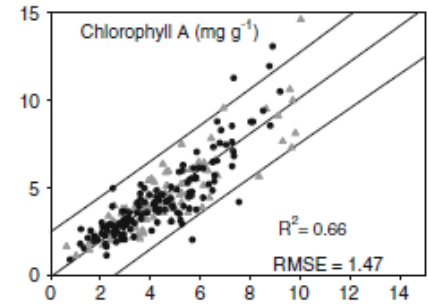
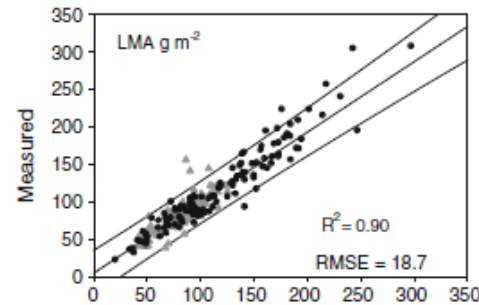
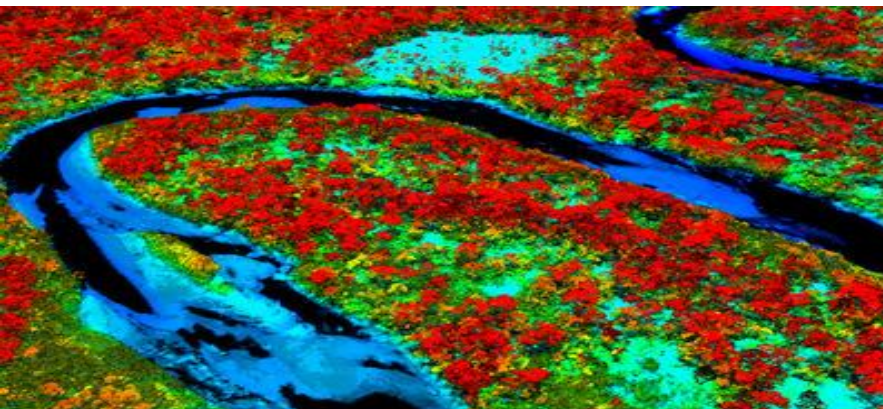
# REFLECTANCE SIGNATURES





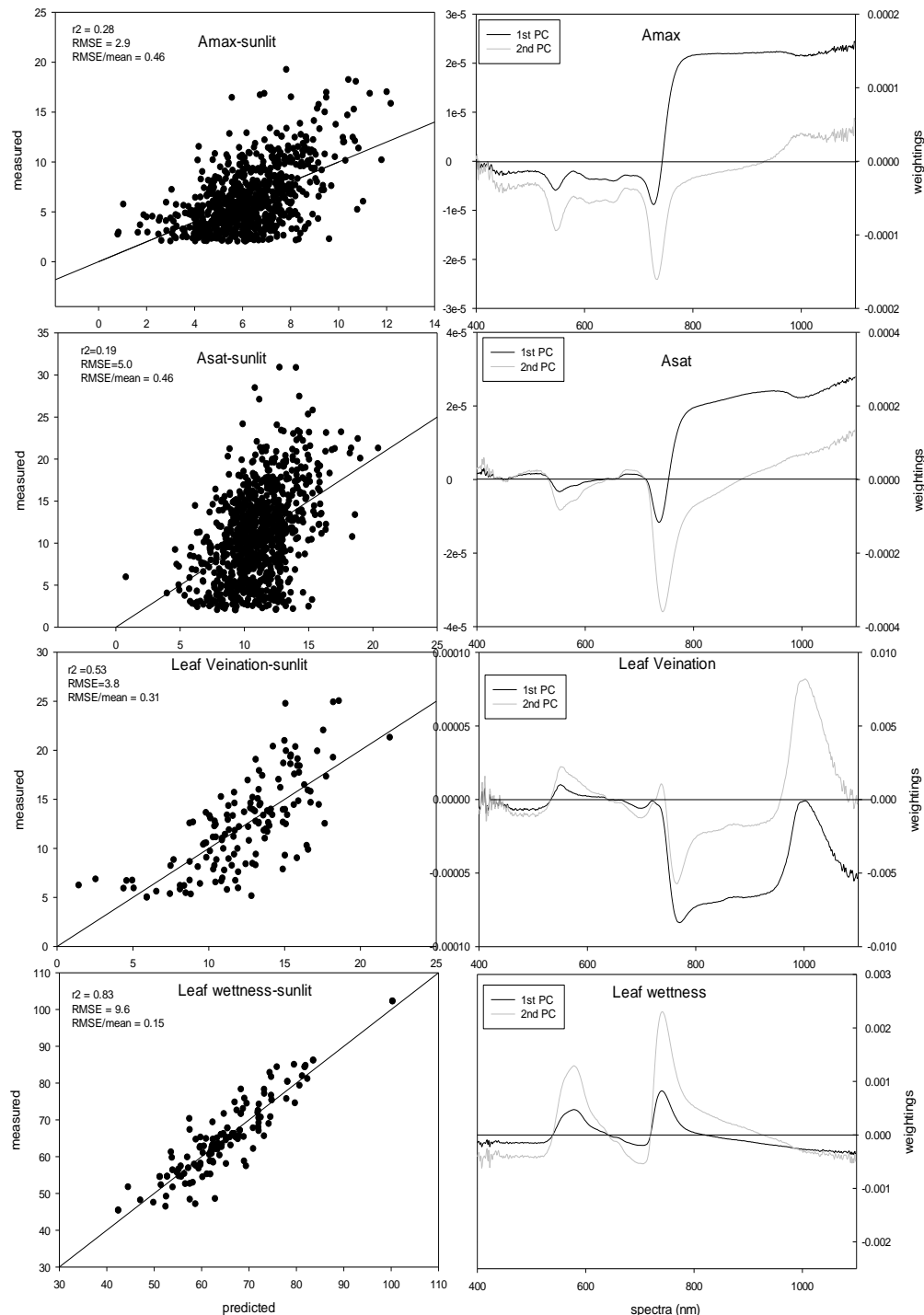
# Using leaf spectroscopy to predict leaf traits

- Many leaf traits, including photosynthesis ( $A_{max}$ ), can be predicted via leaf spectral properties.



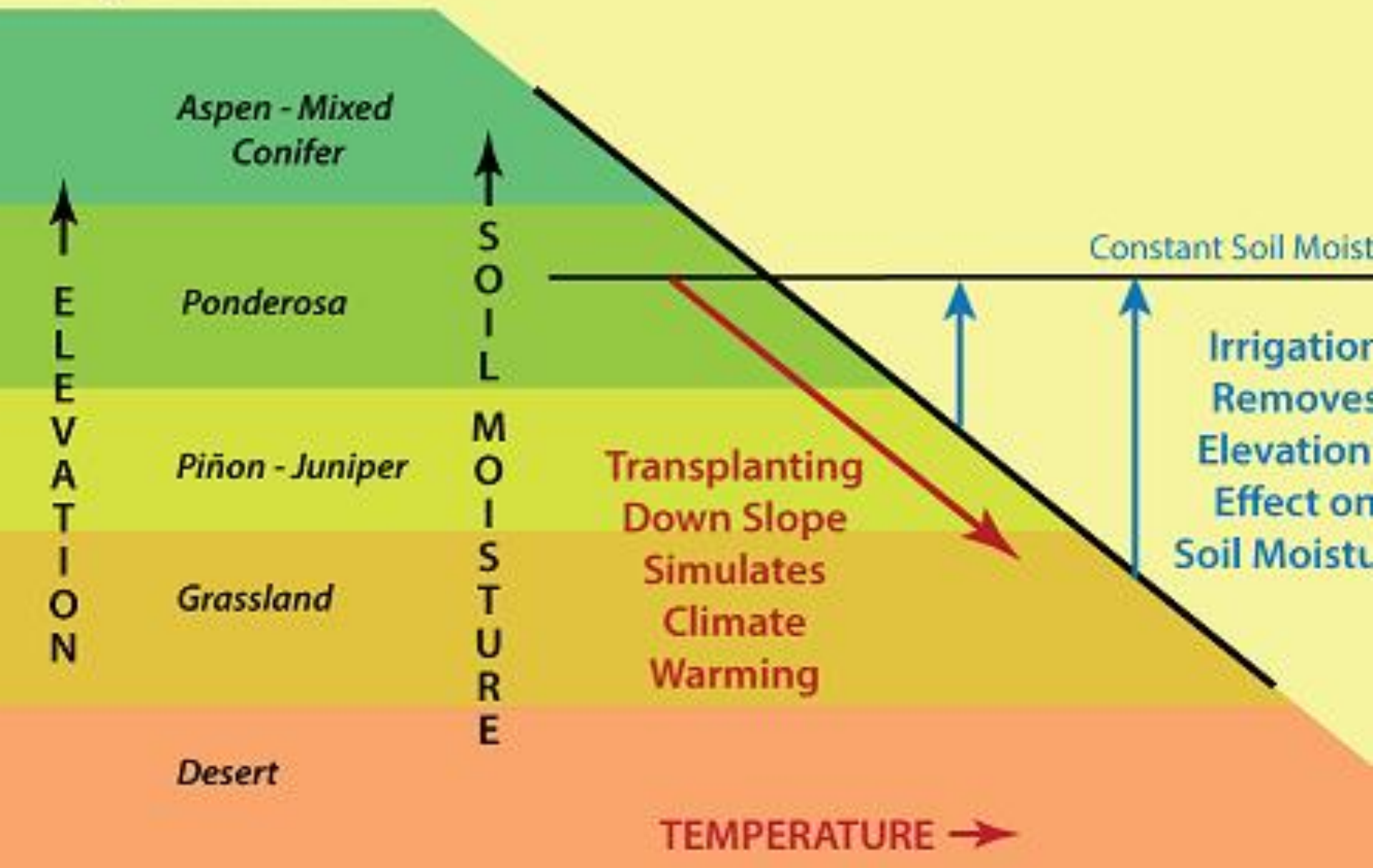
Can we predict other non-  
leaf traits?

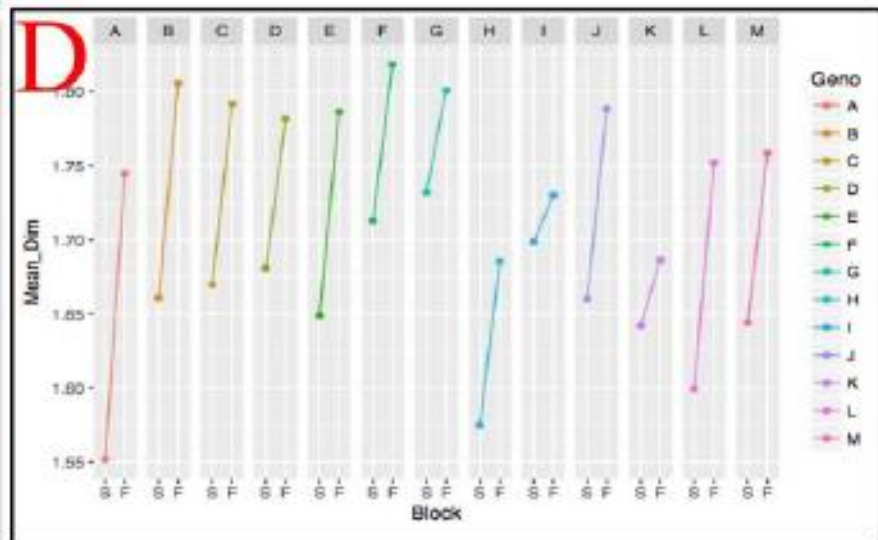
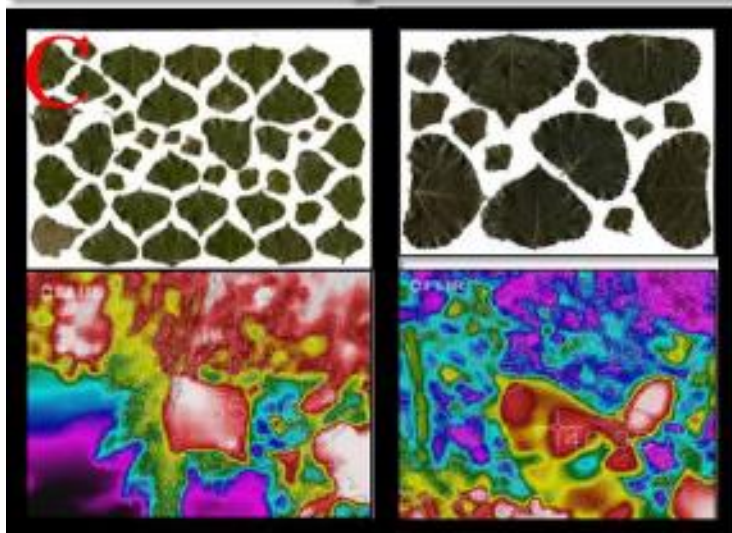
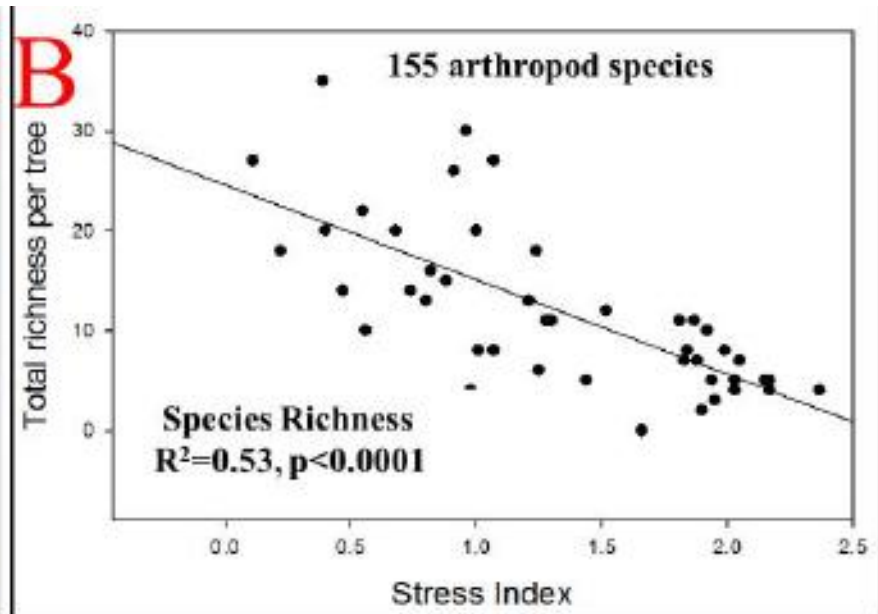
- We can predict traits like photosynthesis, wood density, veination and leaf wetness with spectroscopy



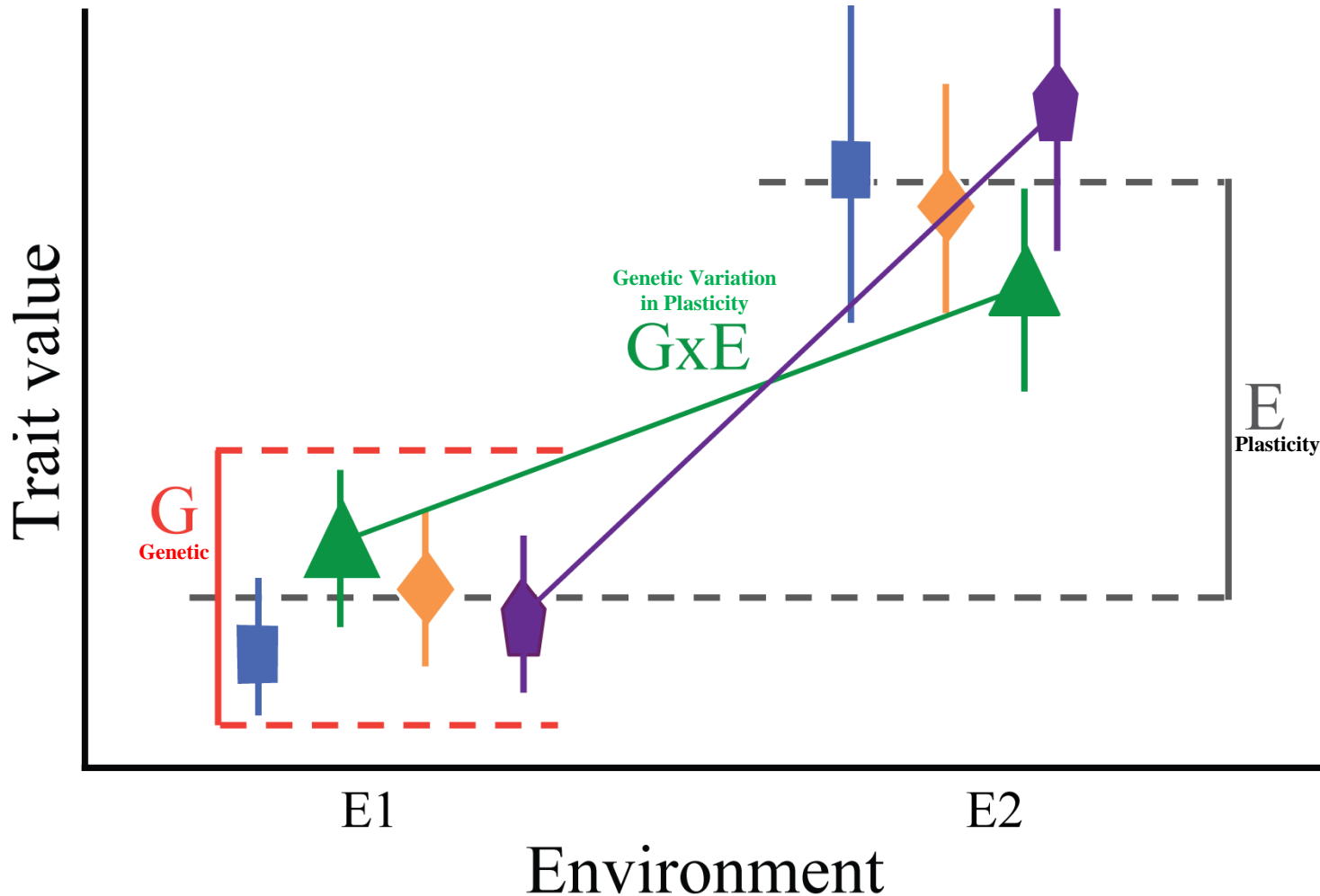
# SEGA

## Vegetation Zones







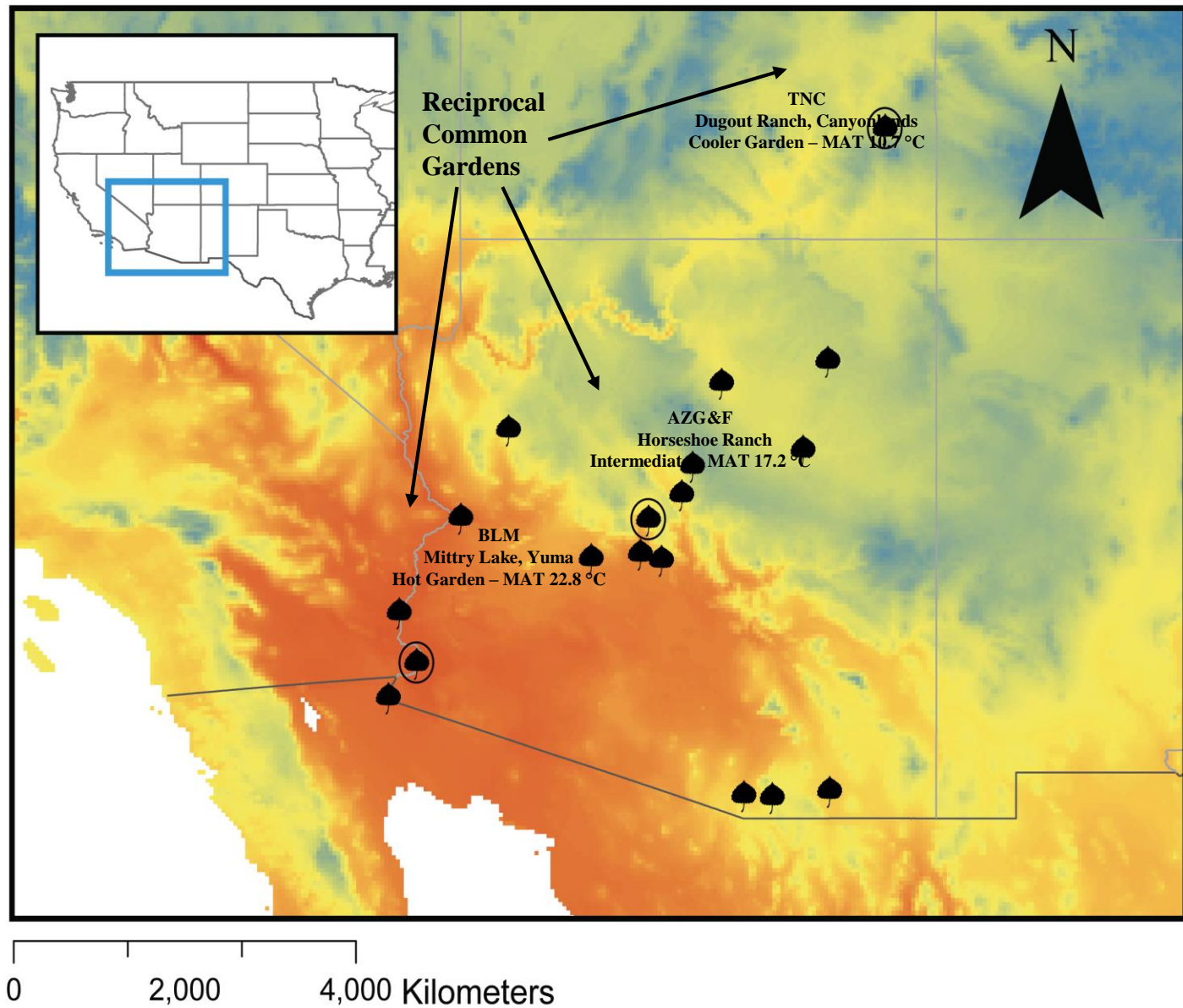


**A two-environment reaction norm showing the components of phenotypic variation of four genotypes:  $G$  = trait variation due to population genetics within a single environment,  $E$  = trait variation due to change in environment (plasticity),  $G \times E$  = the variation in plasticity among genotypes. Phenotypic variation ( $V_P$ ) =  $V_G + V_E + V_{G \times E}$ . From Cooper et al. 2018 Global Change Biology.**

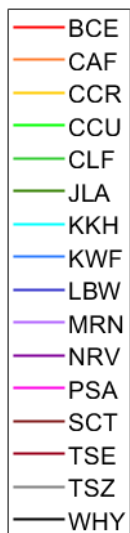
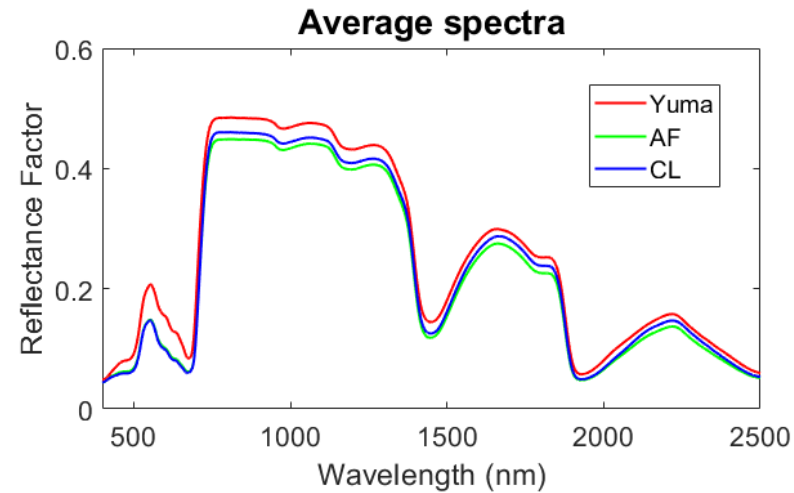
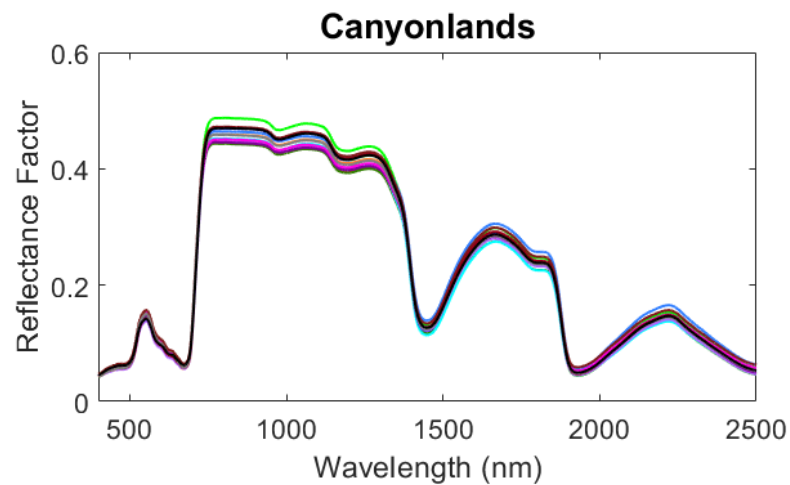
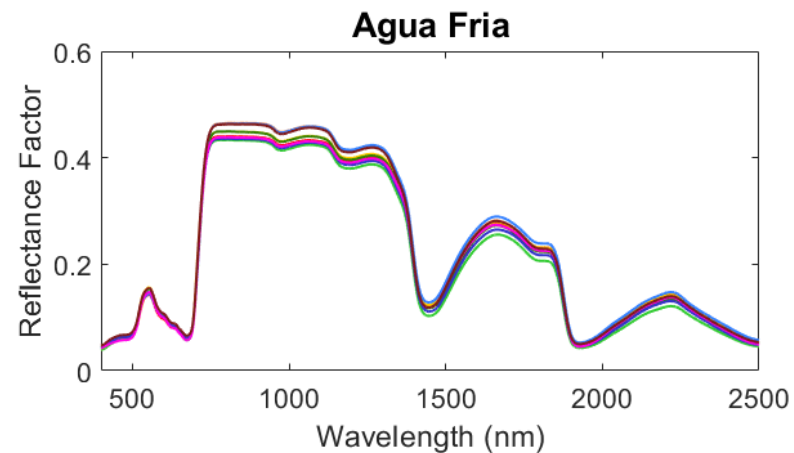
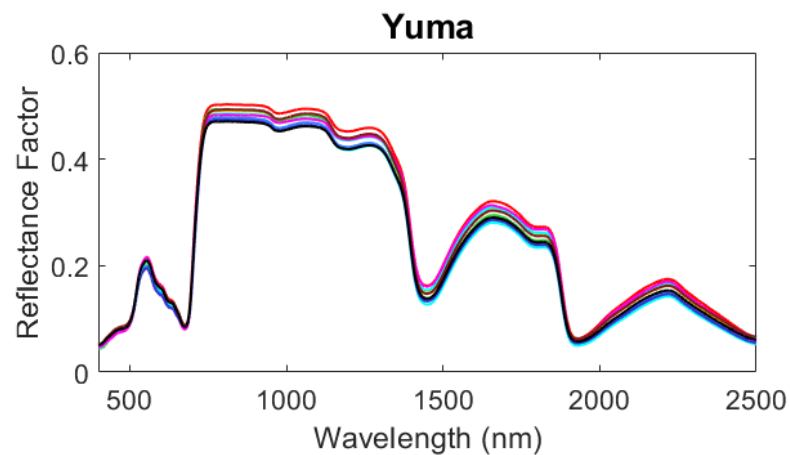




# Reciprocal common gardens show finer scale local adaptation within the Sonoran desert ecotype



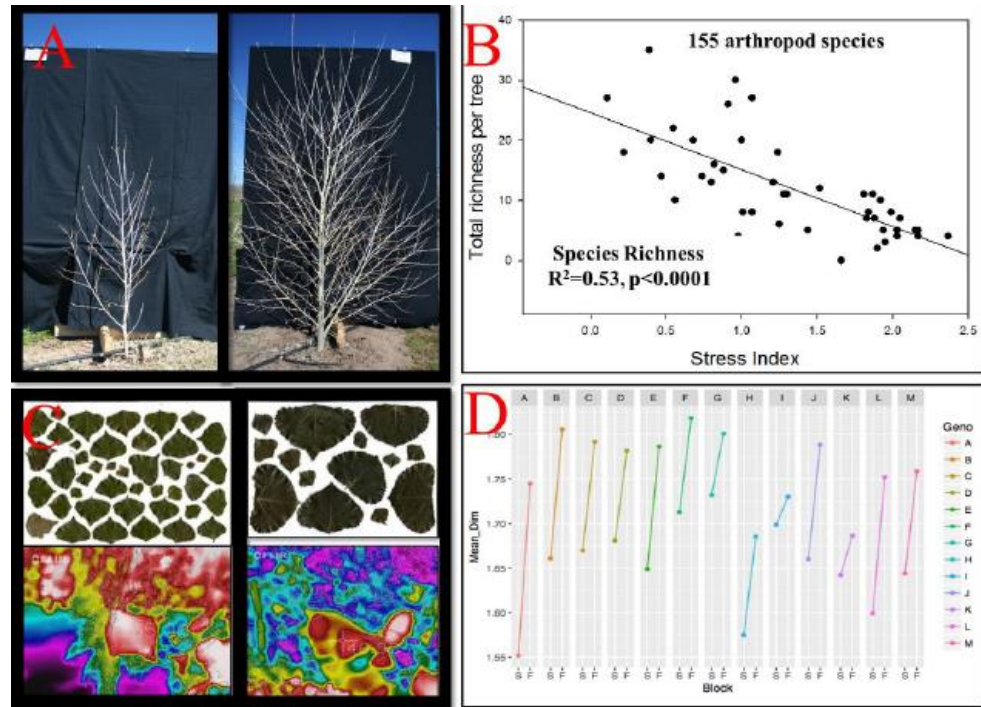
(Cooper et al. 2018 Global Change Biology)



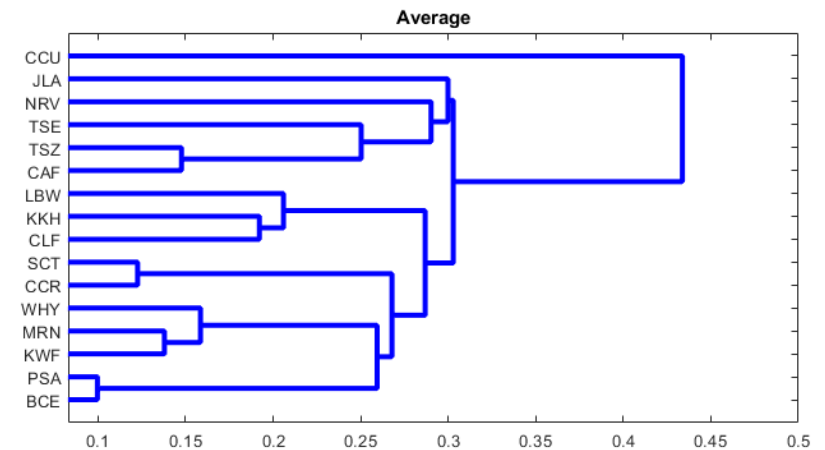
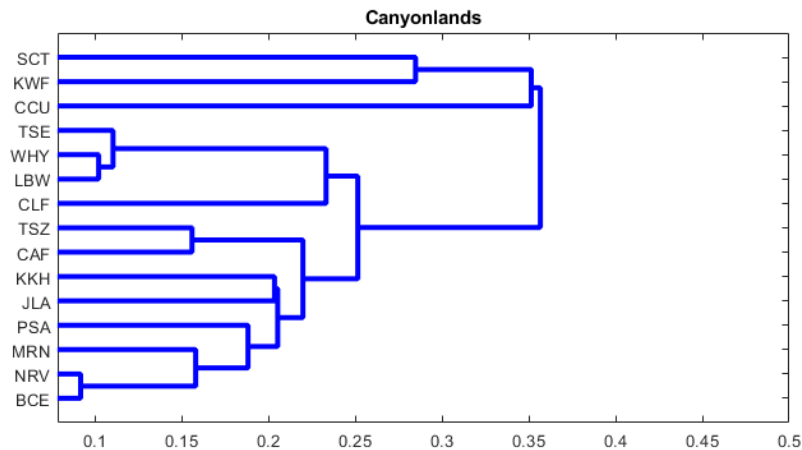
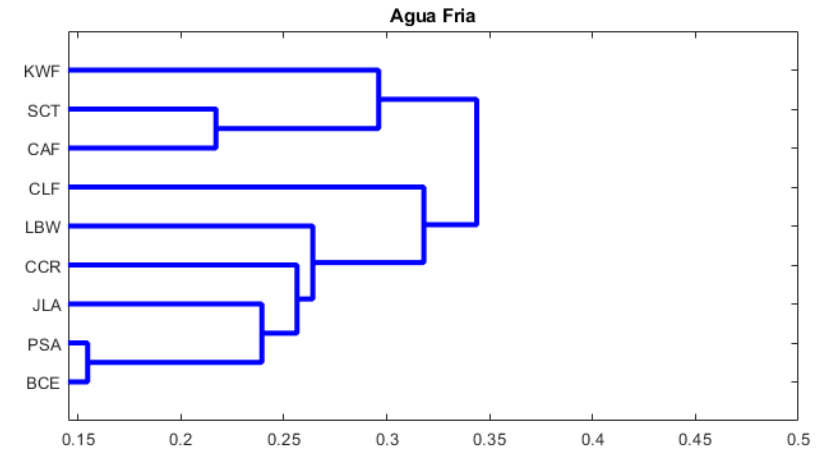
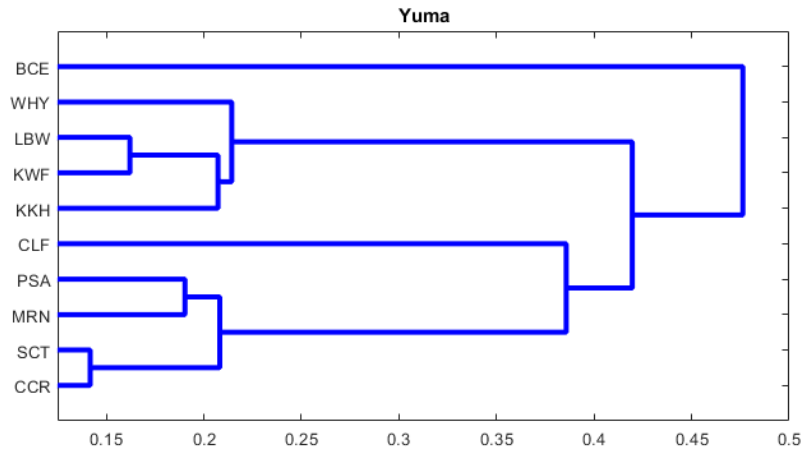


# Preliminary Methods

- Hierarchical cluster analysis
- Partial least squares regression
- Need more traits for comparison!

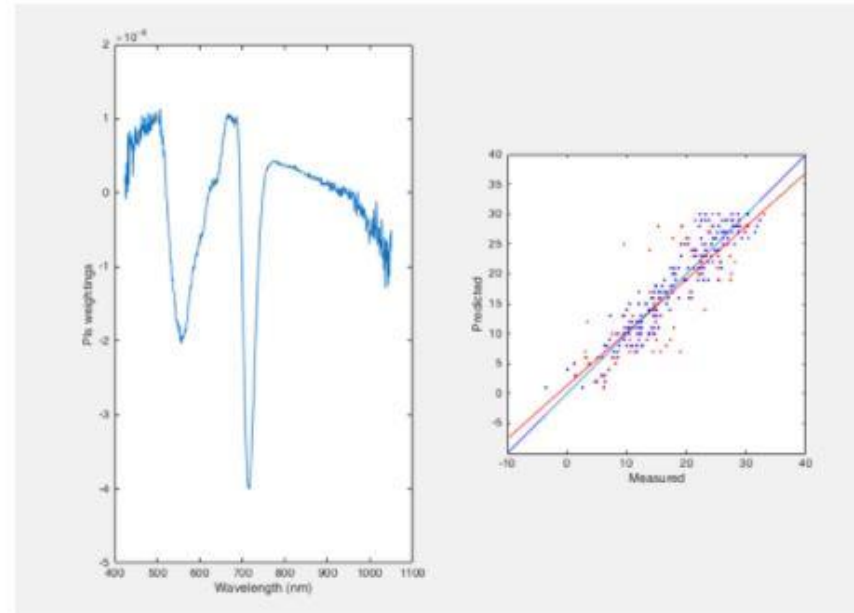


# Hierarchical cluster analysis applied to the average spectra of each population sampled in each garden.

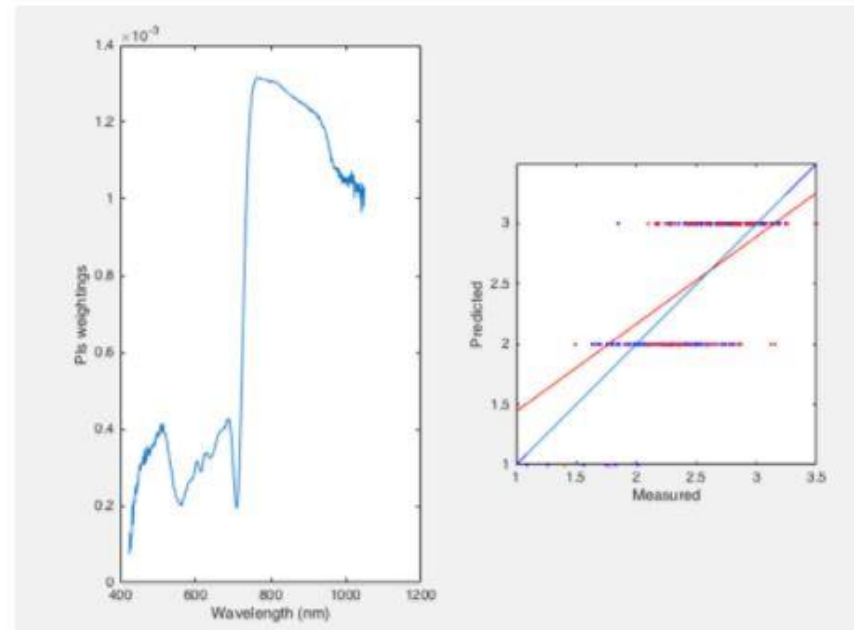


## Genotypes

Strong predictions of both genotypes and plant height

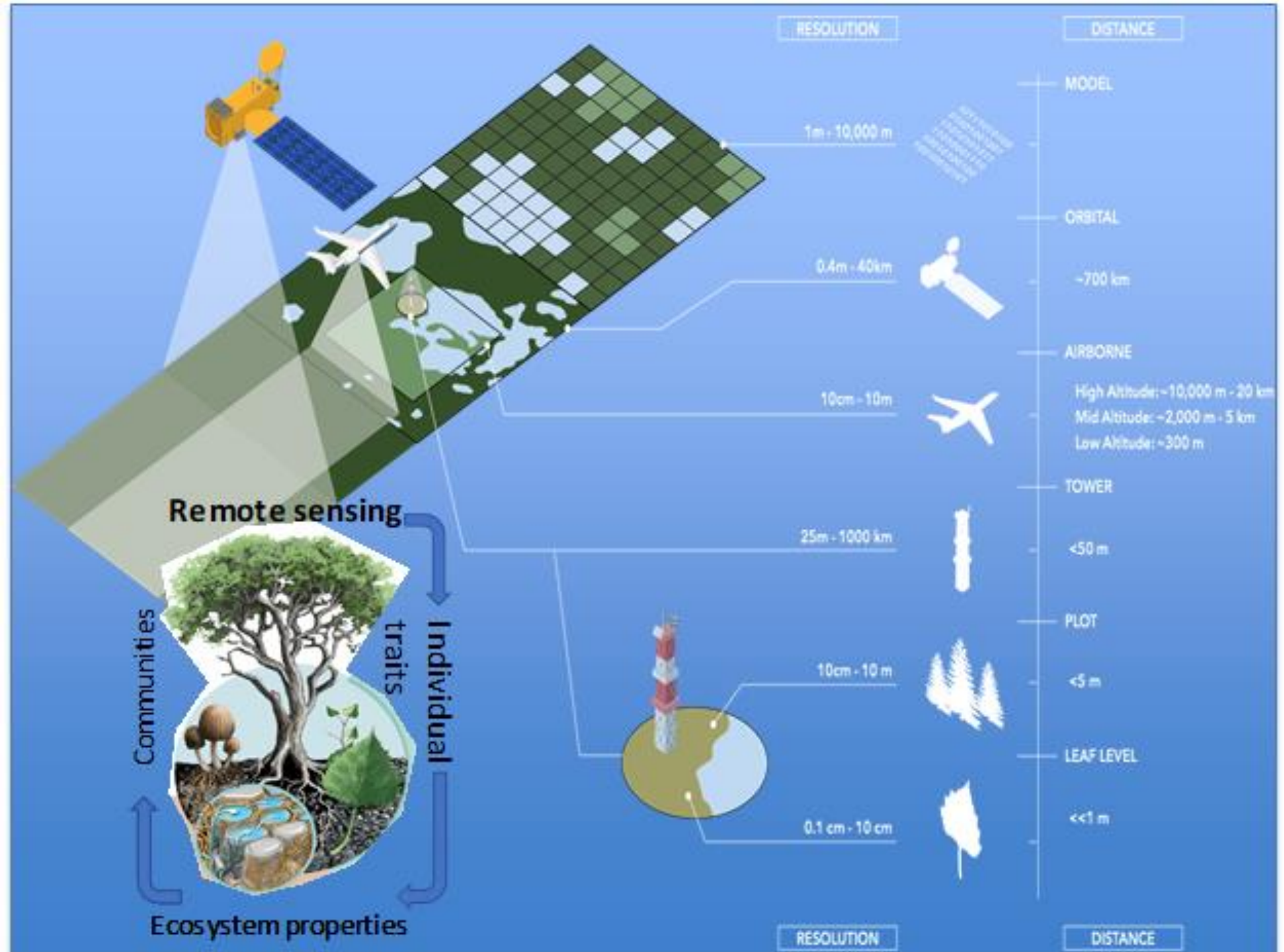


## Height



<b>Genotypes</b>	Individual leaves (3 per plant) n=307	One averaged spectrum for each plant n=103
RMSE	5.89	7.7
RMSEmean	0.33	0.43
R <sup>2</sup> - cal	0.87	0.72
R <sup>2</sup> - test	0.47	0.18
<b>S/M/T</b>	Individual leaves (3 per plant) n=307	One averaged spectrum for each plant n=103
RMSE	0.52	0.49
RMSEmean	0.21	0.19
R <sup>2</sup> - cal	0.43	0.27
R <sup>2</sup> - test	0.13	0.15

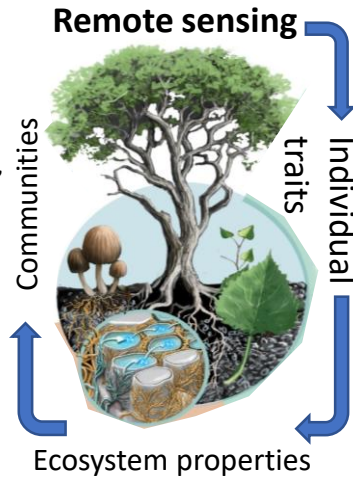




# Genes-to-Ecosystems: Scaling from Critical Plant Species to Global Ecosystems for Precision Conservation & Security of Natural Resources

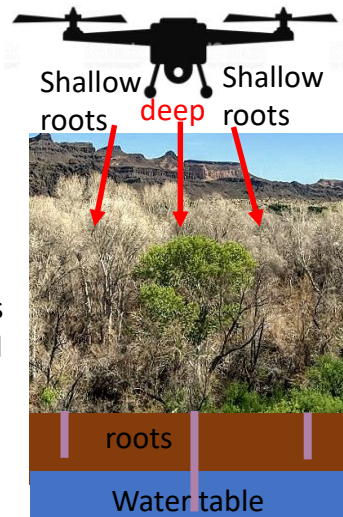
## MISSION

- Climate change and invasive species are altering ecosystems at an unprecedented rate, leading to massive mortality of plants that provide key ecosystem services including food, fiber, fuel, and clean water.
- By combining high performing plant genotypes with microbes that promote stress tolerance, we will restore ecosystem services to degraded lands, promoting global health and security.



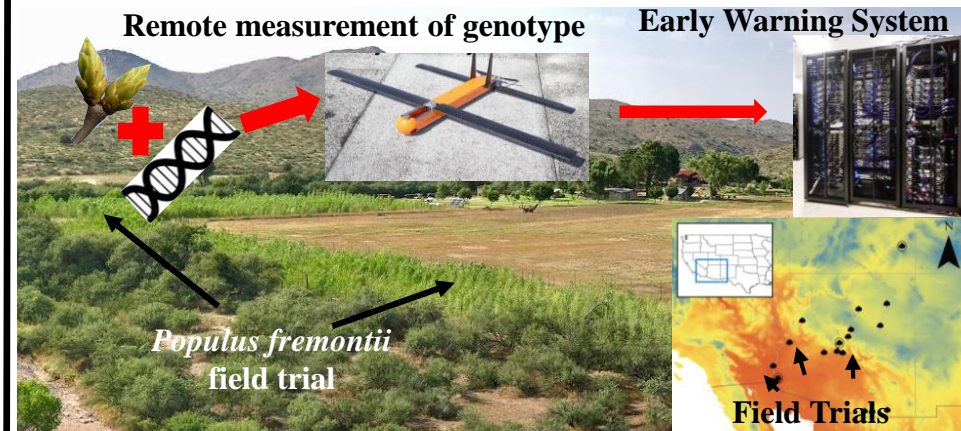
## WIN STRATEGY/KEY TECHNOLOGIES

- We can integrate genetics and environment to develop predictive models of landscape scale shifts as a function of the environment.
- Hyperspectral remote sensing and lidar predicts tree genetics and phenotype, enabling prediction of future plant traits that promote drought tolerance needed in a warmer world.



## NEEDS DESCRIPTION

- We have demonstrated that individual leaf spectroscopy predicts key plant traits.
- A suite of technologies is needed to scale trait detection to landscape and global levels.
- We can succeed by combining Raytheon's expertise and capacities in hyperspectral remote sensing with NAU's infrastructure and expertise in genetics, ecology and remote sensing.



## END USER/CUSTOMER

- Predict vulnerable environmental regions before an environmental catastrophe leads to famine, instability and migration.
- Customers: DoD, DHS, FEMA
- Infer gene by environment interactions to predict vulnerability to climate change.
- Customers: USFS, NPS, USDA, BLM, BOR