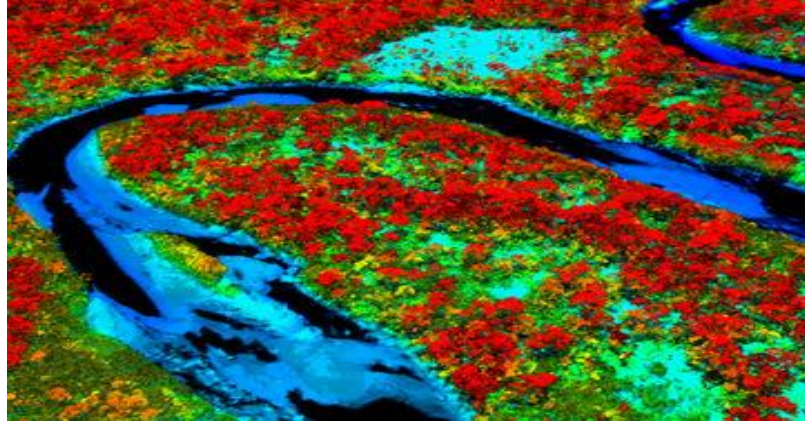


High throughput phenotyping at the cottonwood common gardens



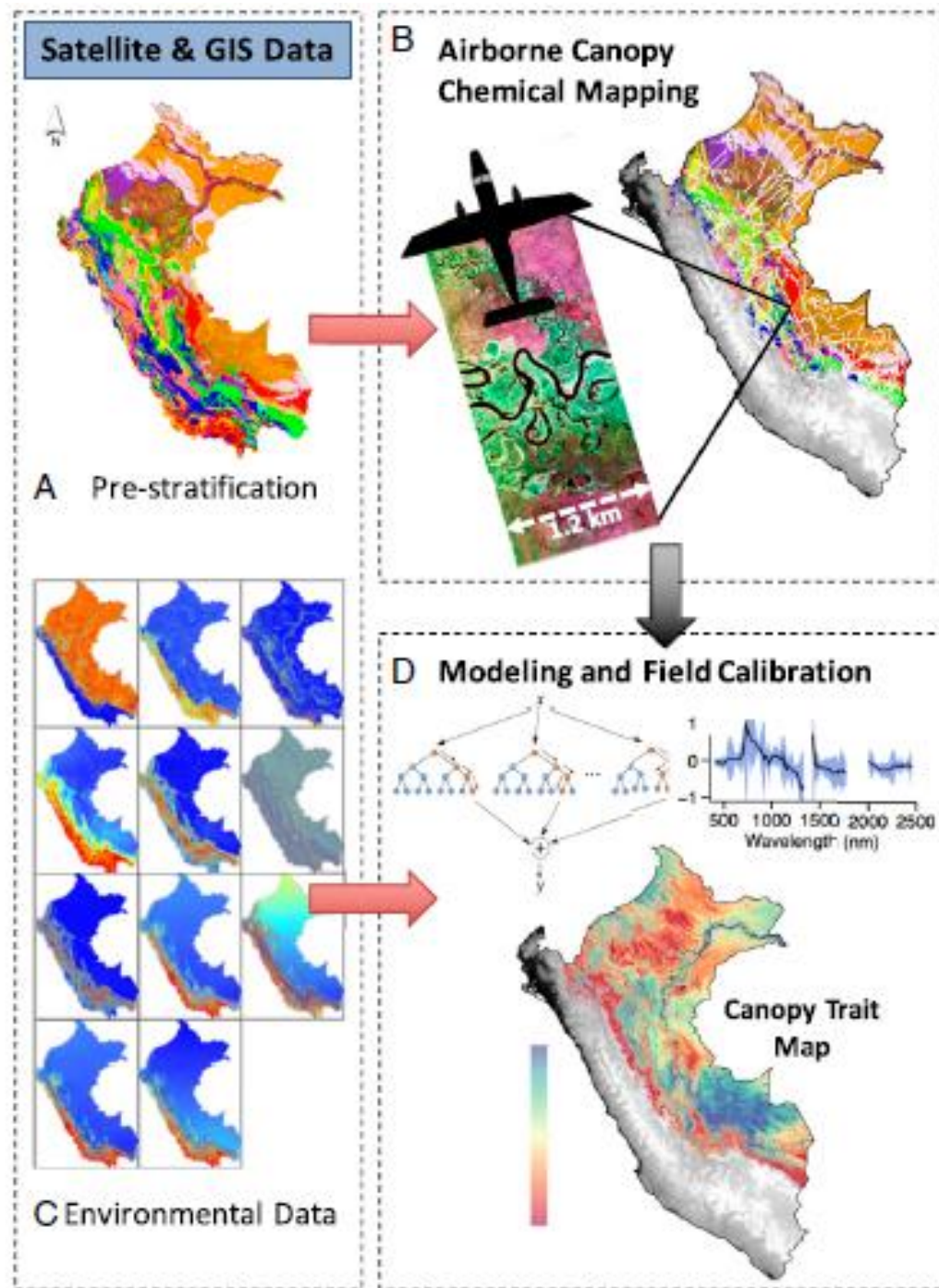
Chris Doughty and Eleanor Thomson



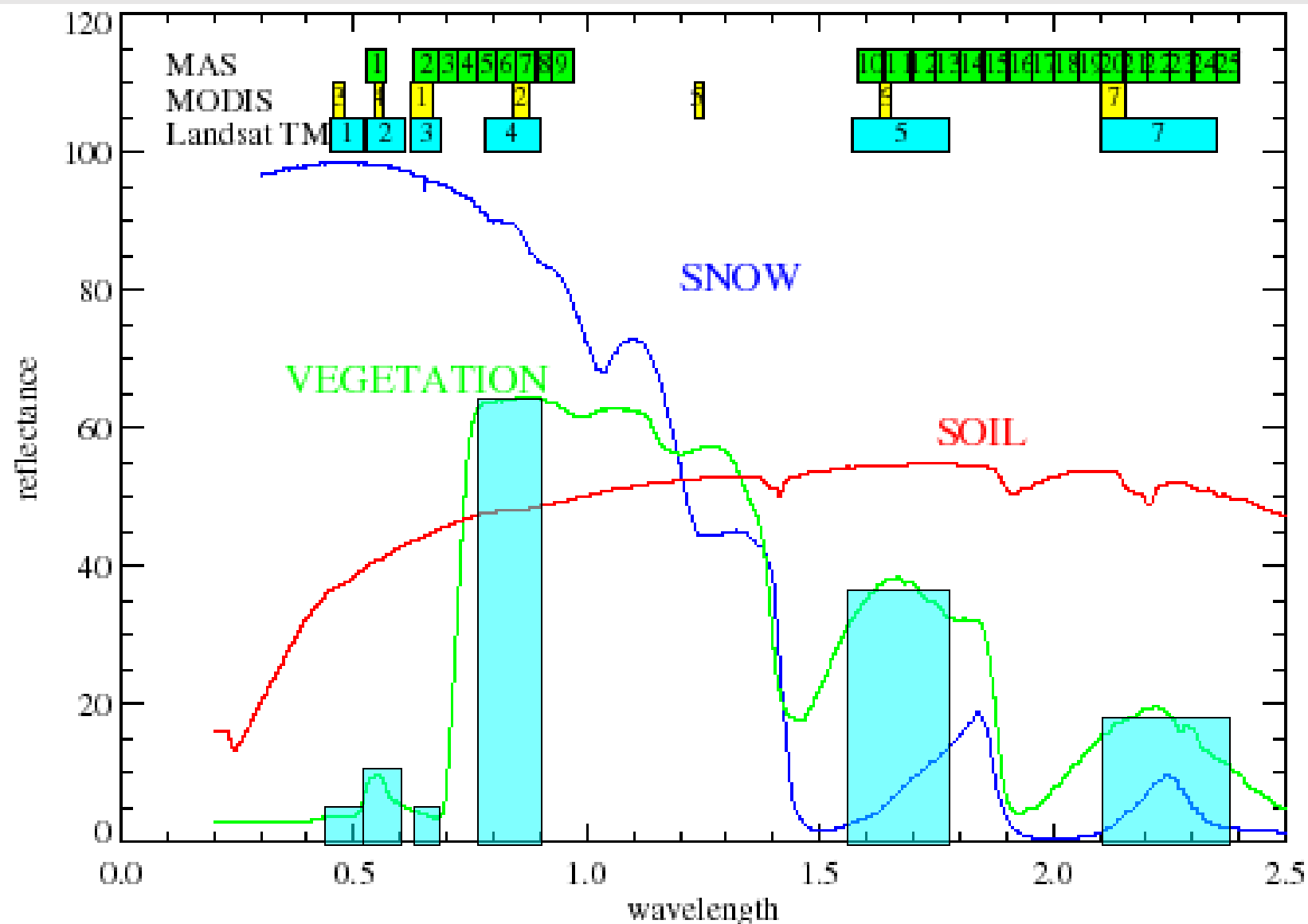
School of Informatics,
Computing, and
Cyber Systems



School of Informatics,
Computing, and
Cyber Systems

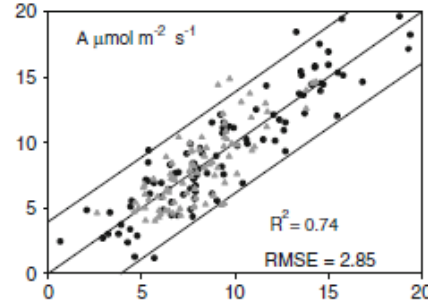
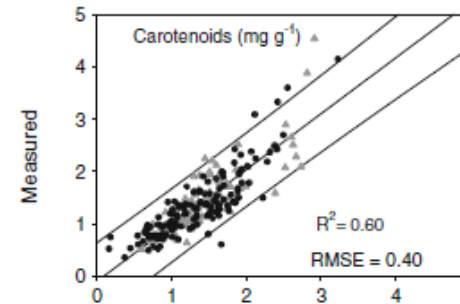
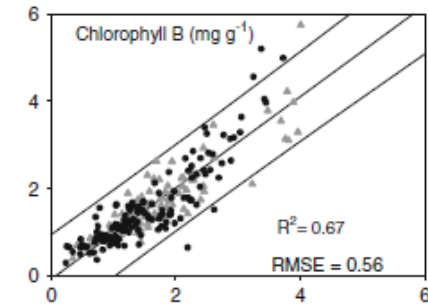
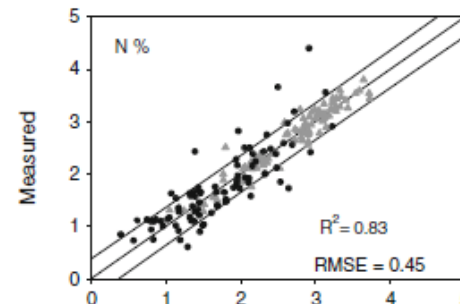
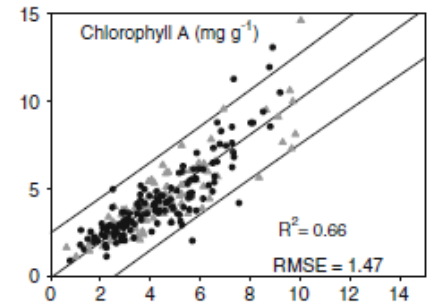
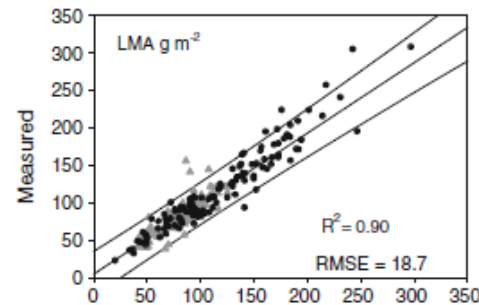
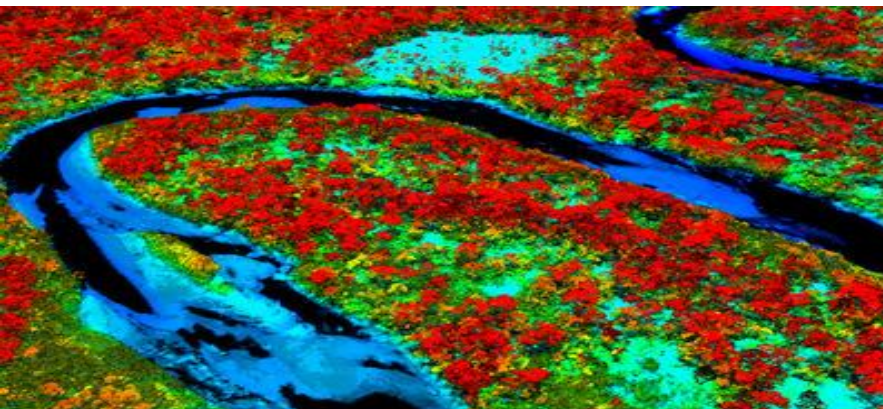


REFLECTANCE SIGNATURES



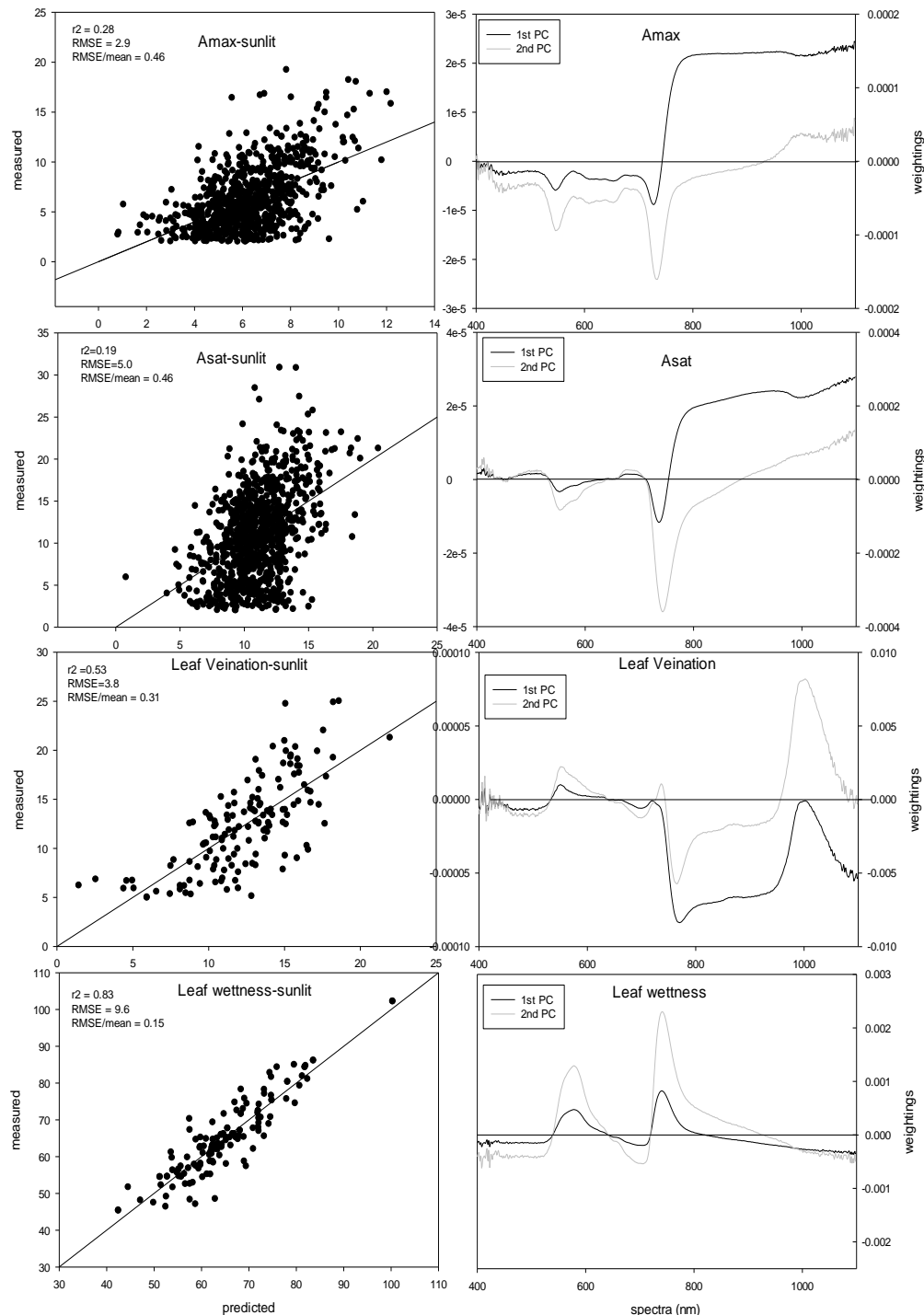
Using leaf spectroscopy to predict leaf traits

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



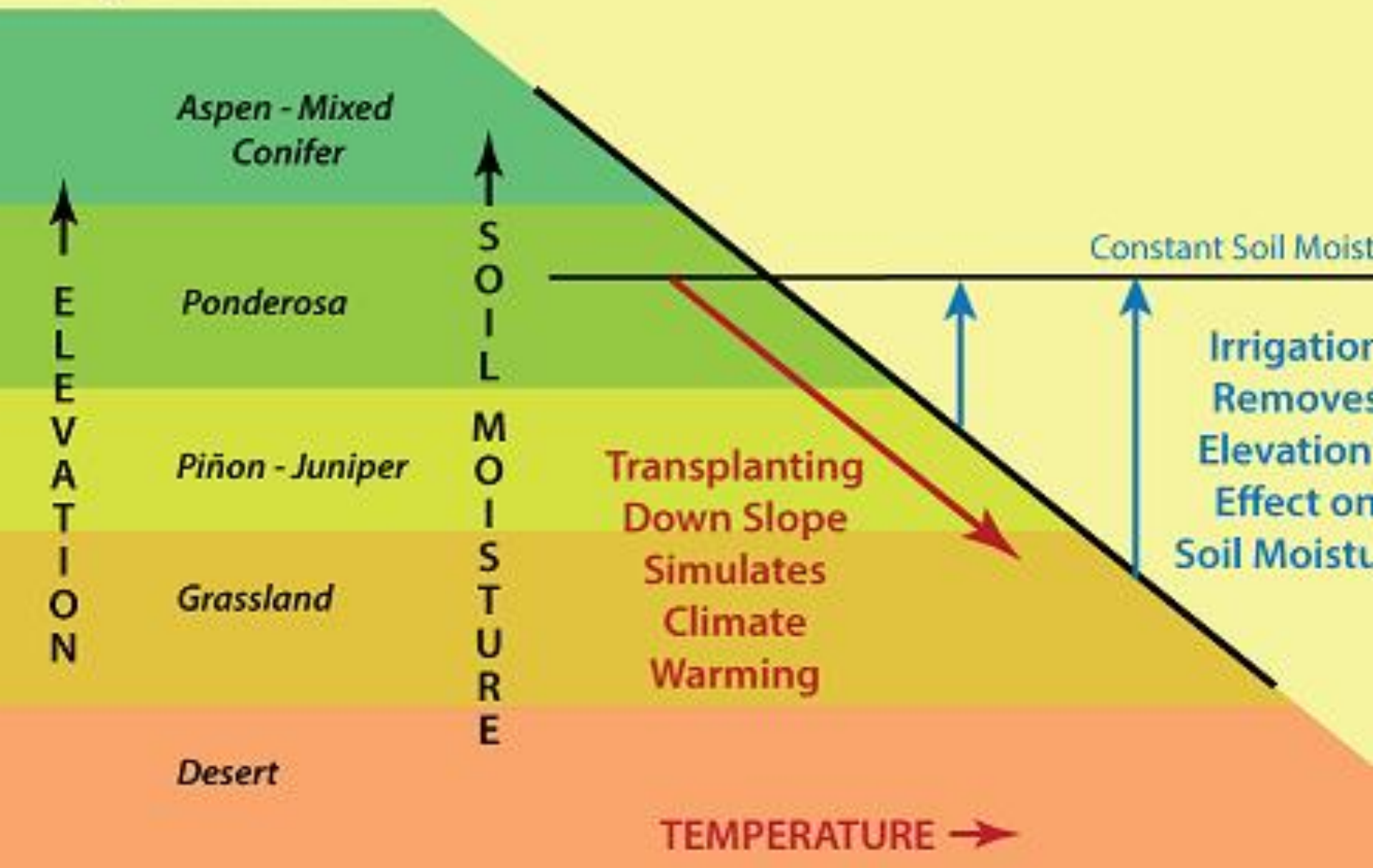
Can we predict other non-
leaf traits?

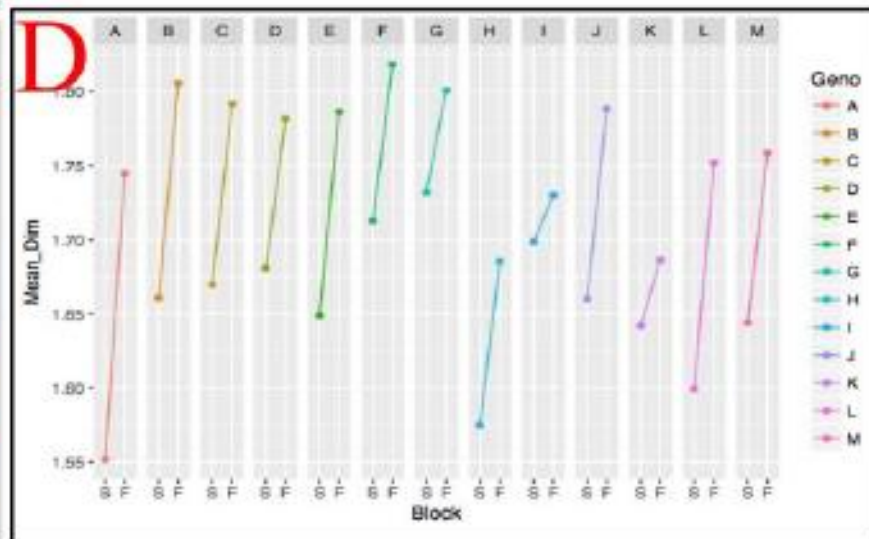
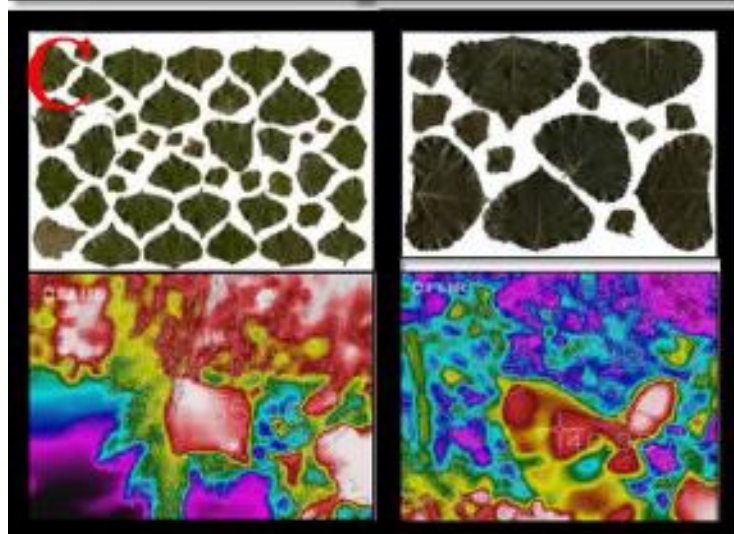
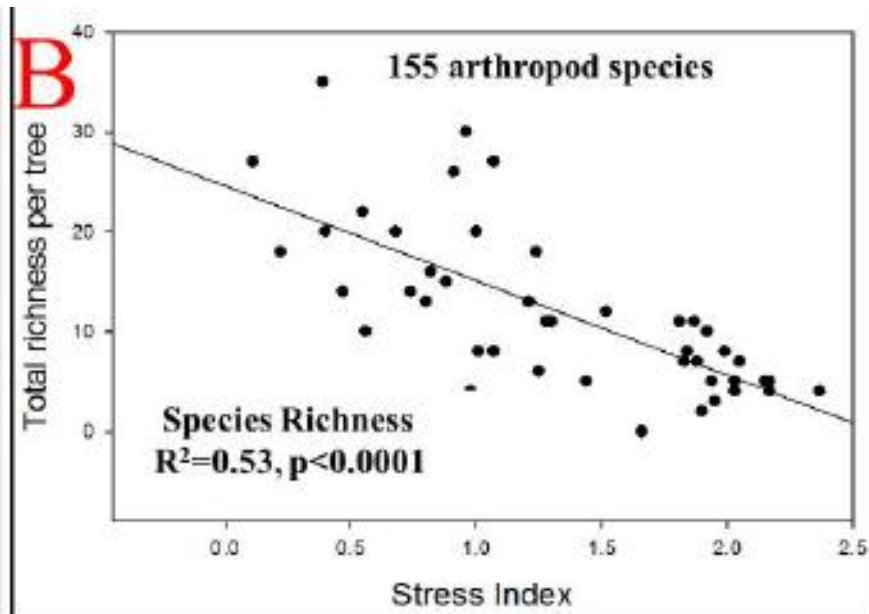
- We can predict traits like photosynthesis, wood density, veination and leaf wetness with spectroscopy

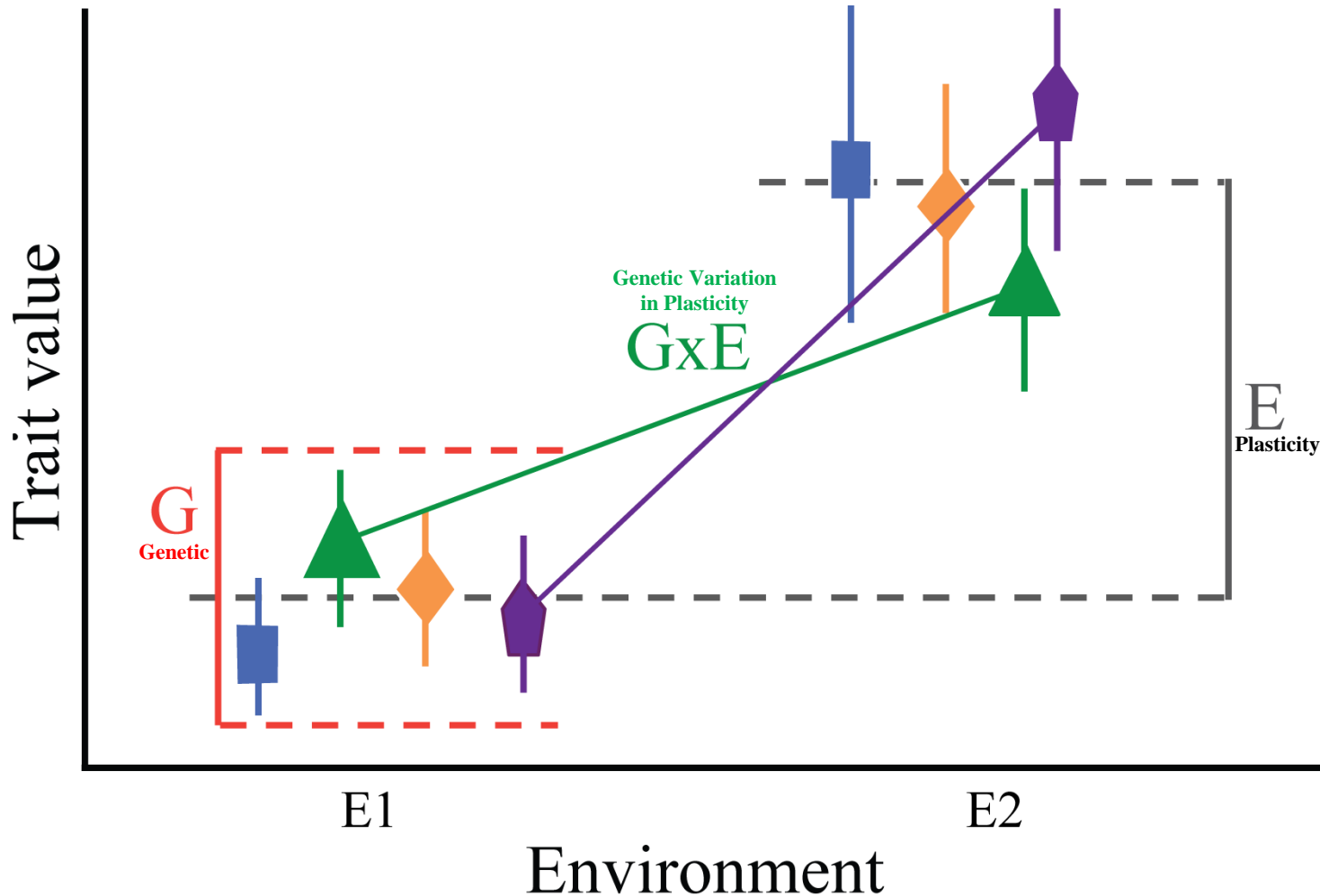


SEGA

Vegetation Zones



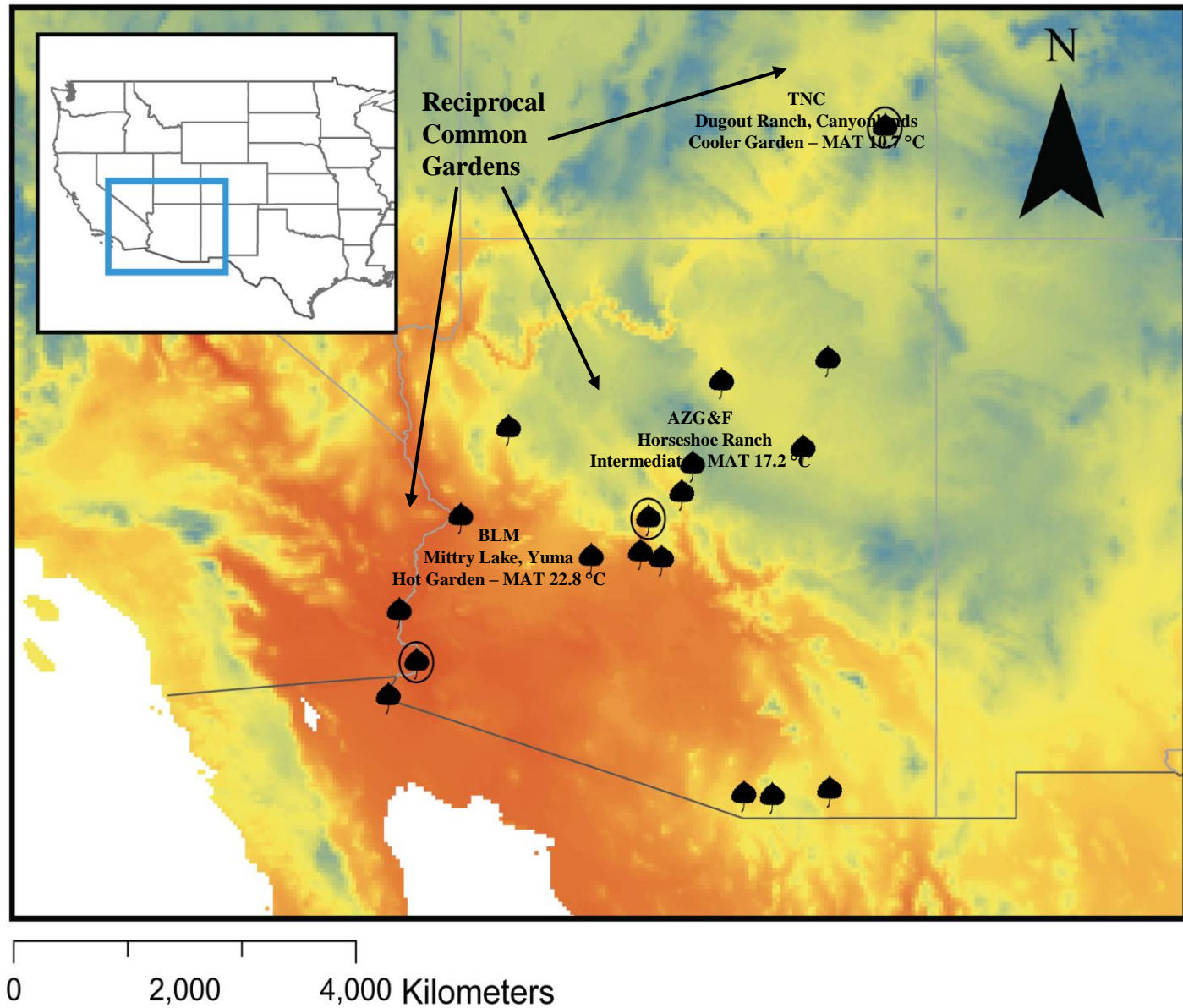




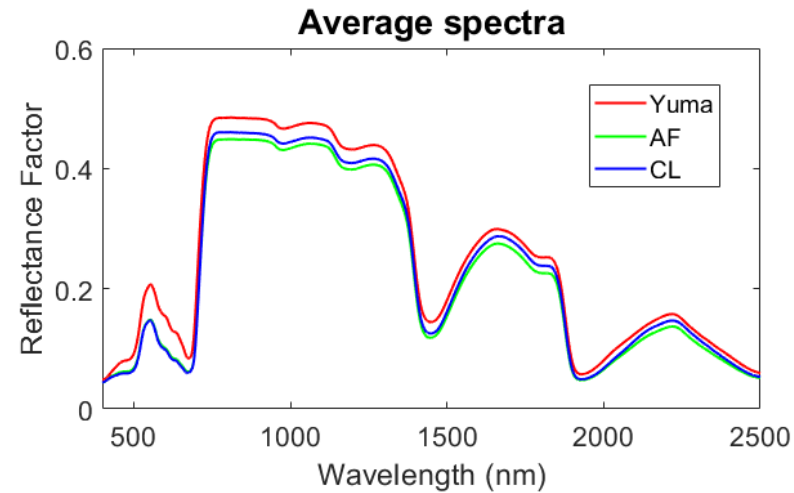
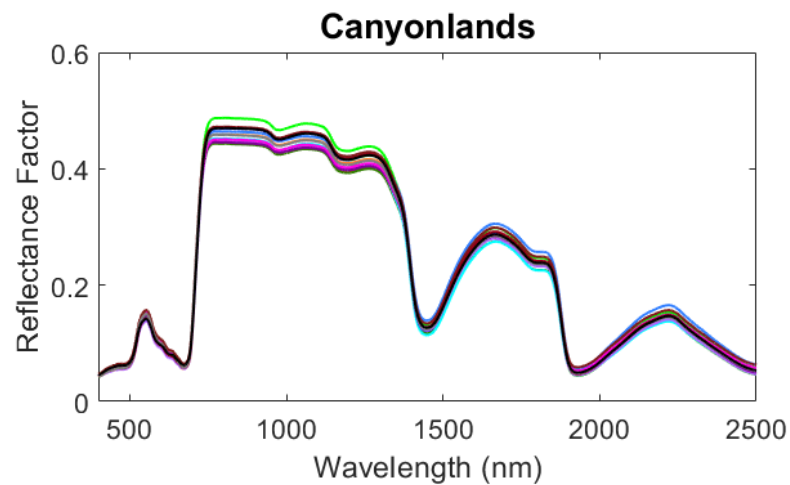
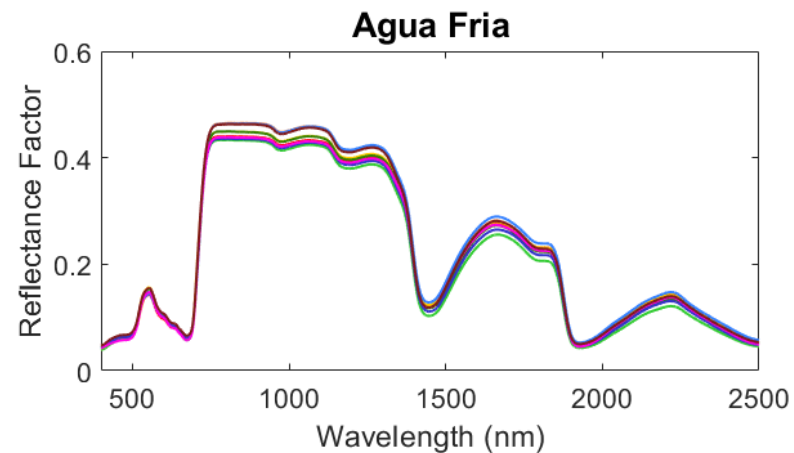
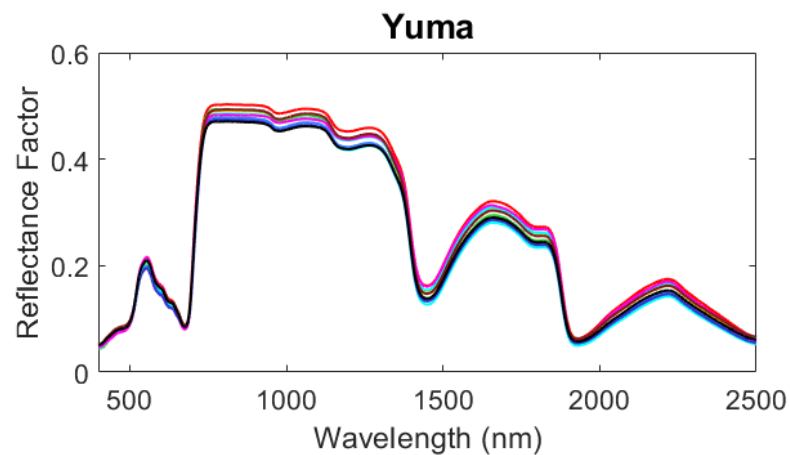
A two-environment reaction norm showing the components of phenotypic variation of four genotypes: G = trait variation due to population genetics within a single environment, E = trait variation due to change in environment (plasticity), $G \times E$ = the variation in plasticity among genotypes. Phenotypic 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



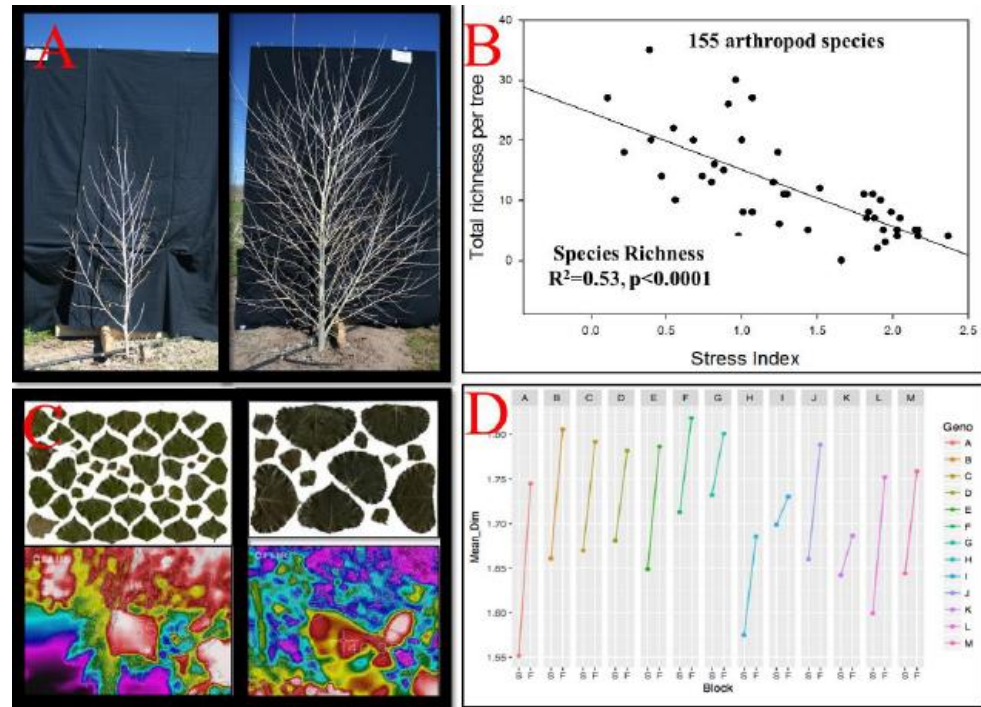
(Cooper et al. 2018 Global Change Biology)



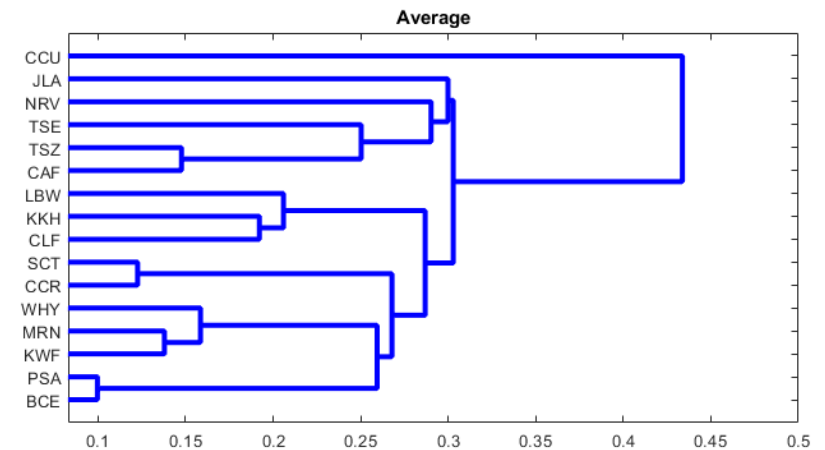
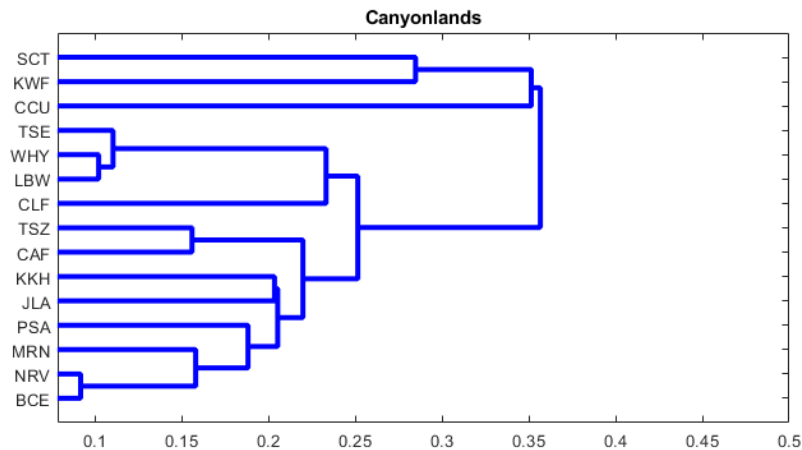
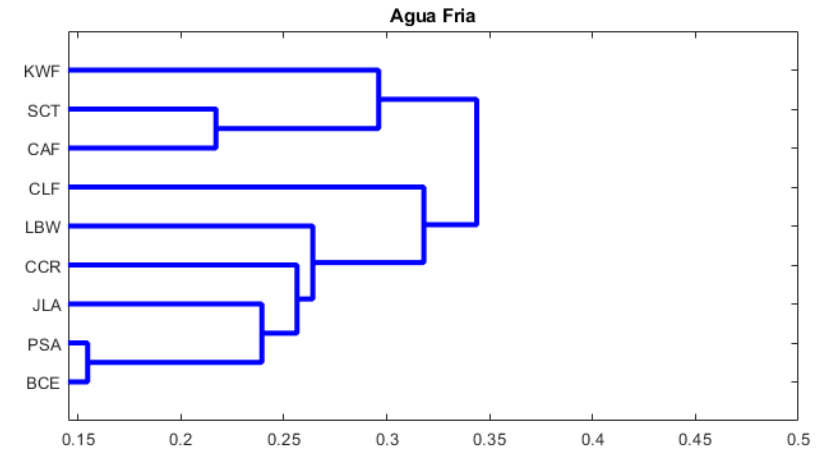
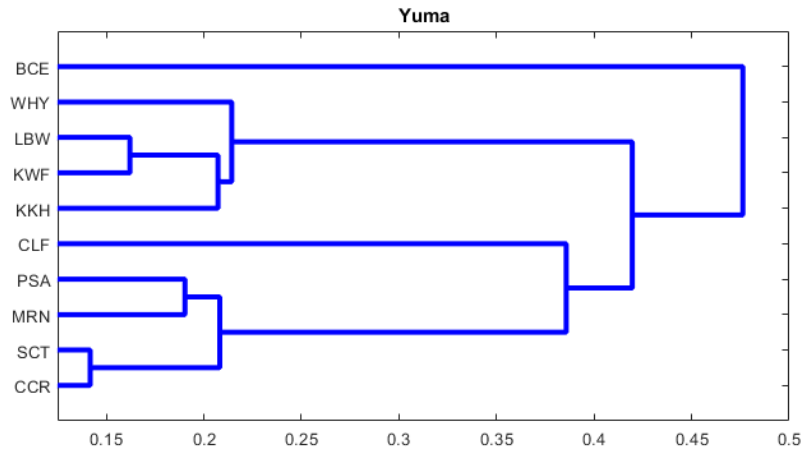
- BCE
- CAF
- CCR
- CCU
- CLF
- JLA
- KKH
- KWF
- LBW
- MRN
- NRV
- PSA
- SCT
- TSE
- TSZ
- WHY

Preliminary Methods

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

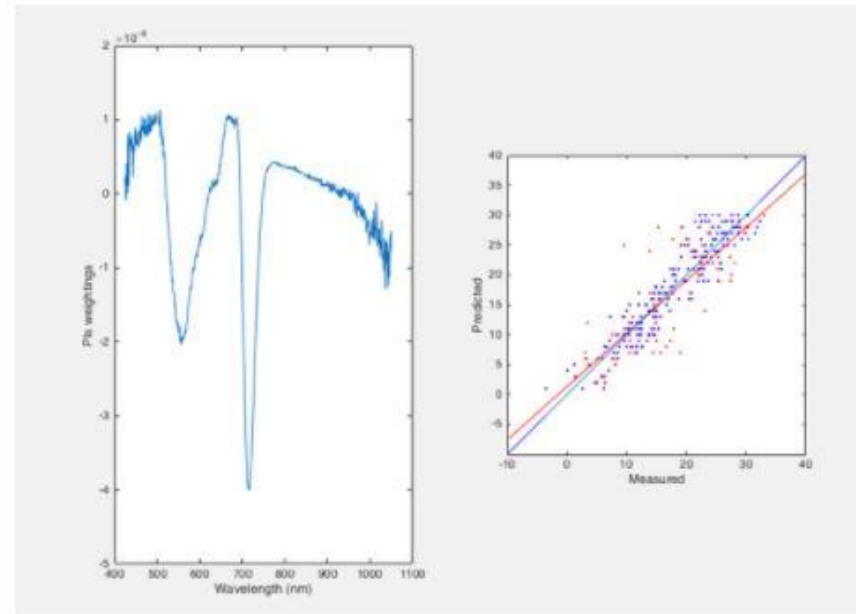


Hierarchical cluster analysis applied to the average spectra of each population sampled in each garden.

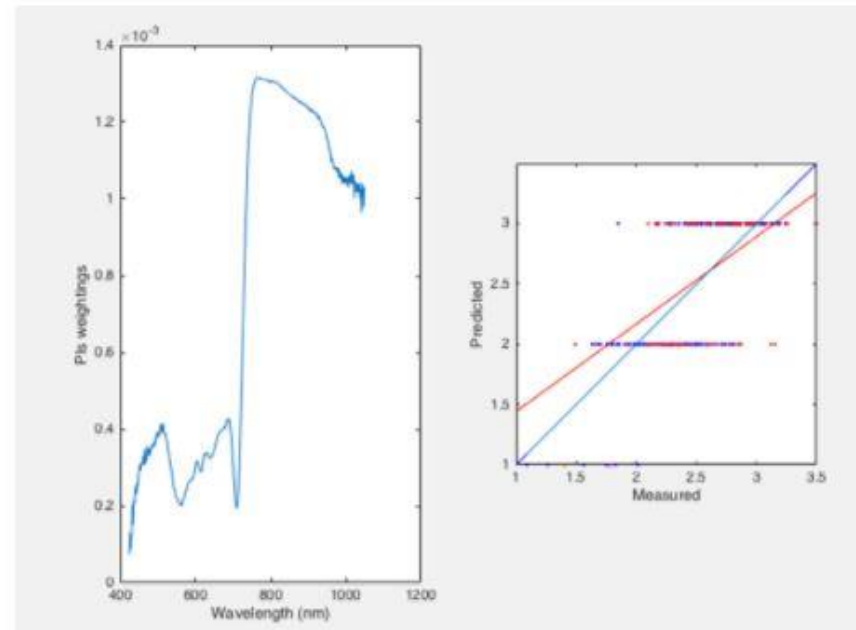


Genotypes

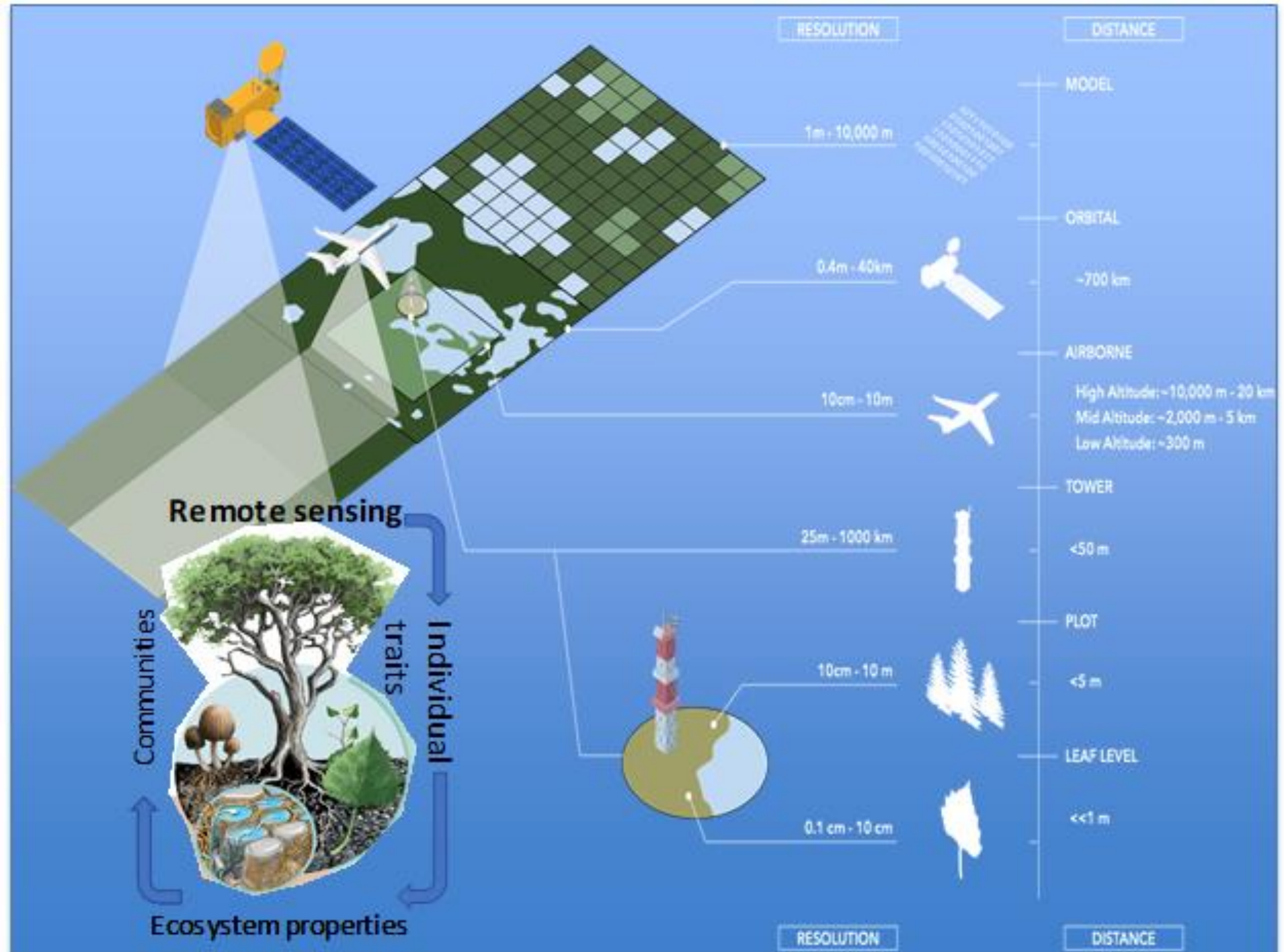
Strong predictions of both genotypes and plant height



Height



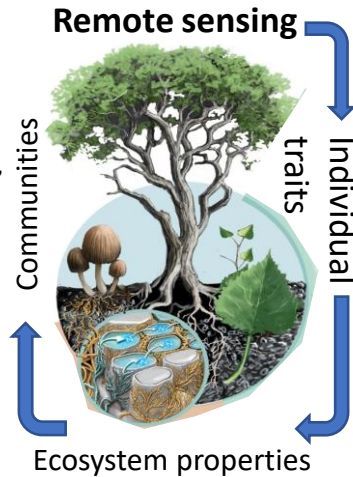
Genotypes	Individual leaves (3 per plant) n=307	One averaged spectrum for each plant n=103
RMSE	5.89	7.7
RMSEmean	0.33	0.43
R ² - cal	0.87	0.72
R ² - test	0.47	0.18
S/M/T	Individual leaves (3 per plant) n=307	One averaged spectrum for each plant n=103
RMSE	0.52	0.49
RMSEmean	0.21	0.19
R ² - cal	0.43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



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

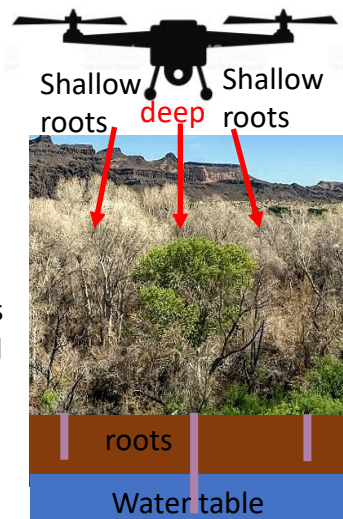
MISSION

- 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.



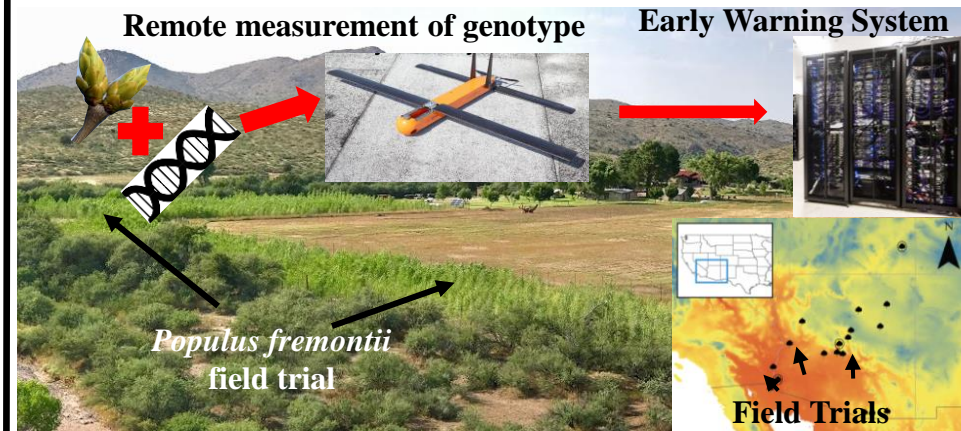
WIN STRATEGY/KEY TECHNOLOGIES

- 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*