

Continuing Project

Characterizing abiotic and biotic tree stress using hyperspectral information

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Reflectance spectroscopy can capture tree stress responses

- **Rapid and accurate** observation of forest traits is necessary for effective monitoring of forest health
- **Leaf spectral reflectance** has been used to investigate functional traits and to detect **diverse forest issues** (e.g., drought, air pollution, fire, diseases, invasive species, etc.) **at different spatial scales**

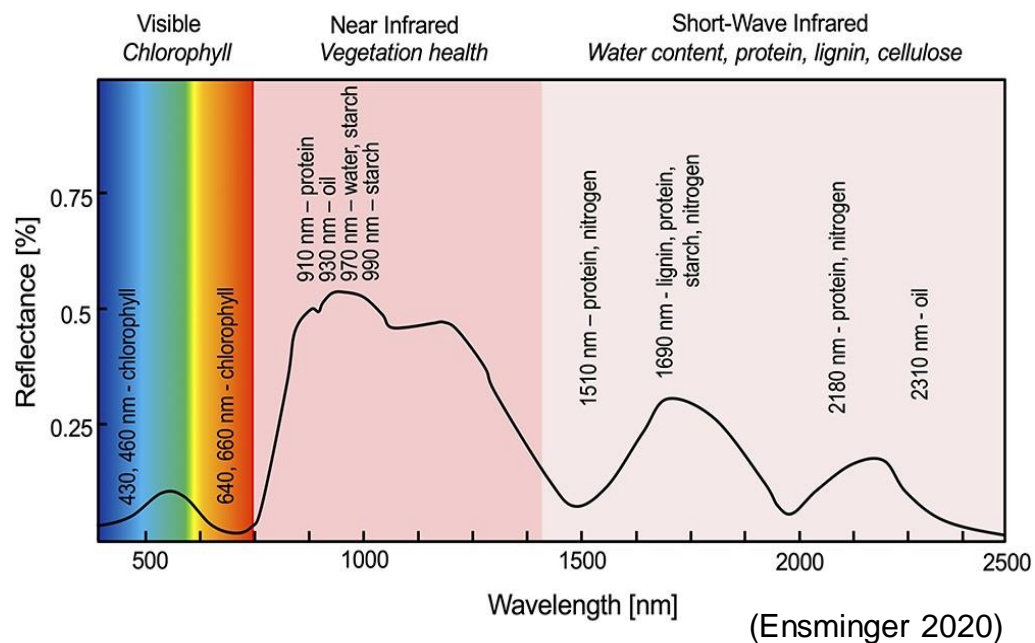


Rapid 'Ohi'a Death, Hawaii (Greg Asner)



Reflectance spectroscopy can capture tree stress responses

- **Retrieving foliar traits** relevant to specific stress response mechanisms from hyperspectral data helps identify different stressors
- Understanding **the link between physiochemical responses and spectral changes** is crucial to increase the potential for extracting a wider range of functional traits from leaf spectral profiles



To test different spectral regions to investigate their influences on model performance estimating key leaf functional traits

- Functional traits can provide information on tree health status
 - Growth & Survival*
(e.g., the photosynthetic rate, water content, specific leaf area, foliar N, sugars)
 - Defense & Stress response*
(e.g., phenolic compounds)
- Various wavelength ranges of a leaf spectral profile contain absorption features related to specific substances in leaves (e.g., chlorophylls, water, cellulose, proteins, nitrogen)



Methods

Abiotic and biotic stress

Fungal infection +
Soil quality

Nitrogen deficiency +
Drought

Fungal infection +
Drought

Nitrogen deficiency +
Salt deposition



Black walnut (*Juglans nigra* L.)
Red oak (*Quercus rubra* L.)



Stress response measurement

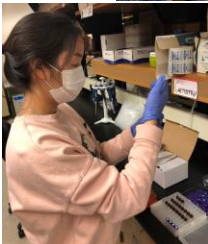
Trait data collection

Spectral data collection

Leaf functional traits
Gas-exchange
Water
Biochemical

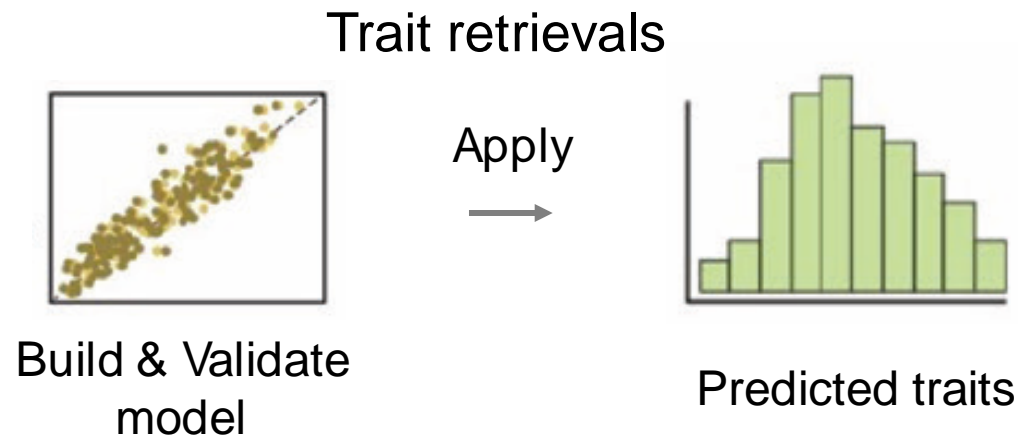
Leaf reflectance spectra
(400–2400 nm)

Spectroradiometer
(SVC HR-1024i)



Partial Least Squares Regression (PLSR)

- Multivariate statistical approach
- Train PLSR models on leaf reflectance spectra and six key leaf traits measured
 - The maximum photosynthetic rate (A_{max})
 - Leaf water content (LWC)
 - Specific leaf area (SLA)
 - Nitrogen (N)
 - Sugars
 - Gallic acid (Gal)
- Validate models and get leaf trait estimates



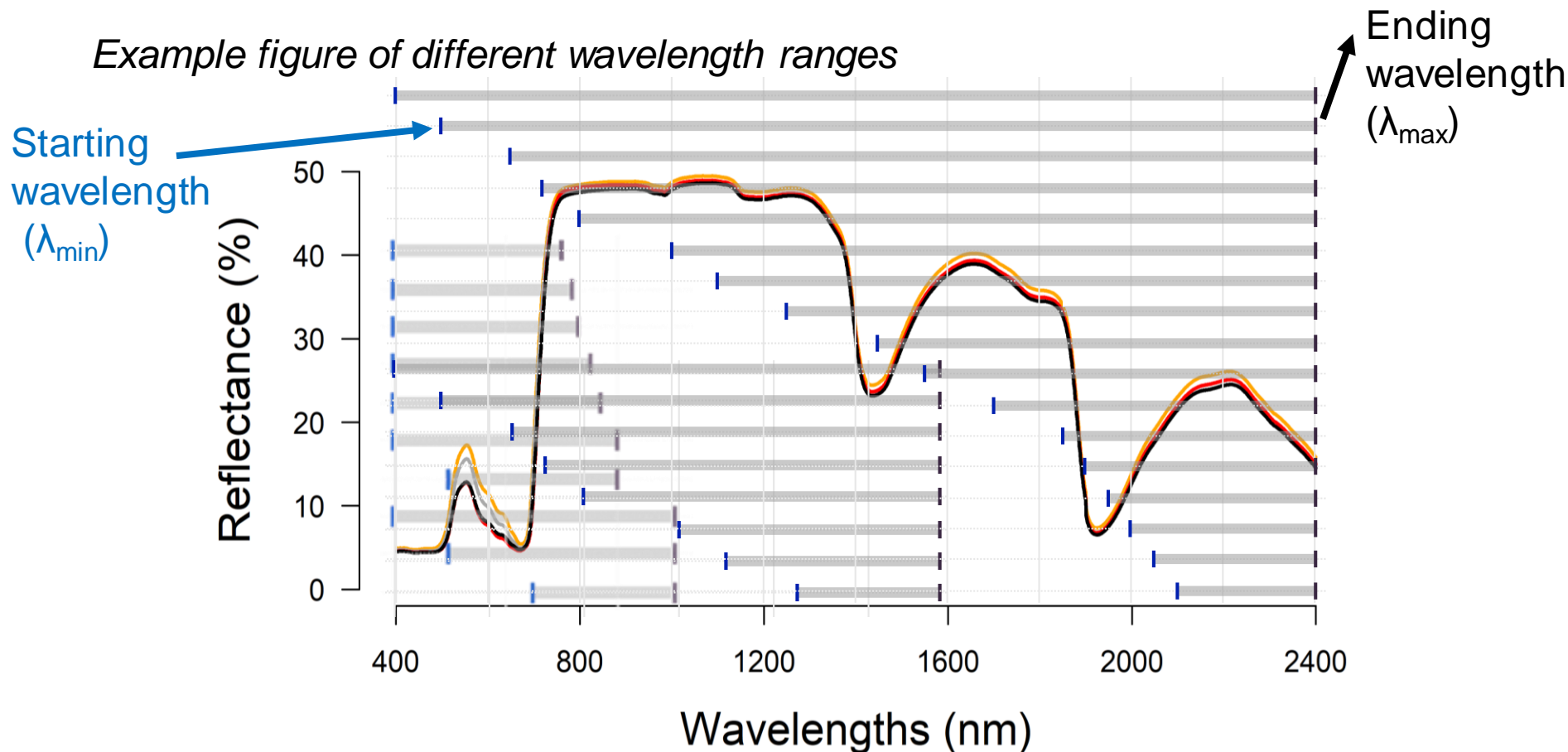
(Burnett et al., 2021)



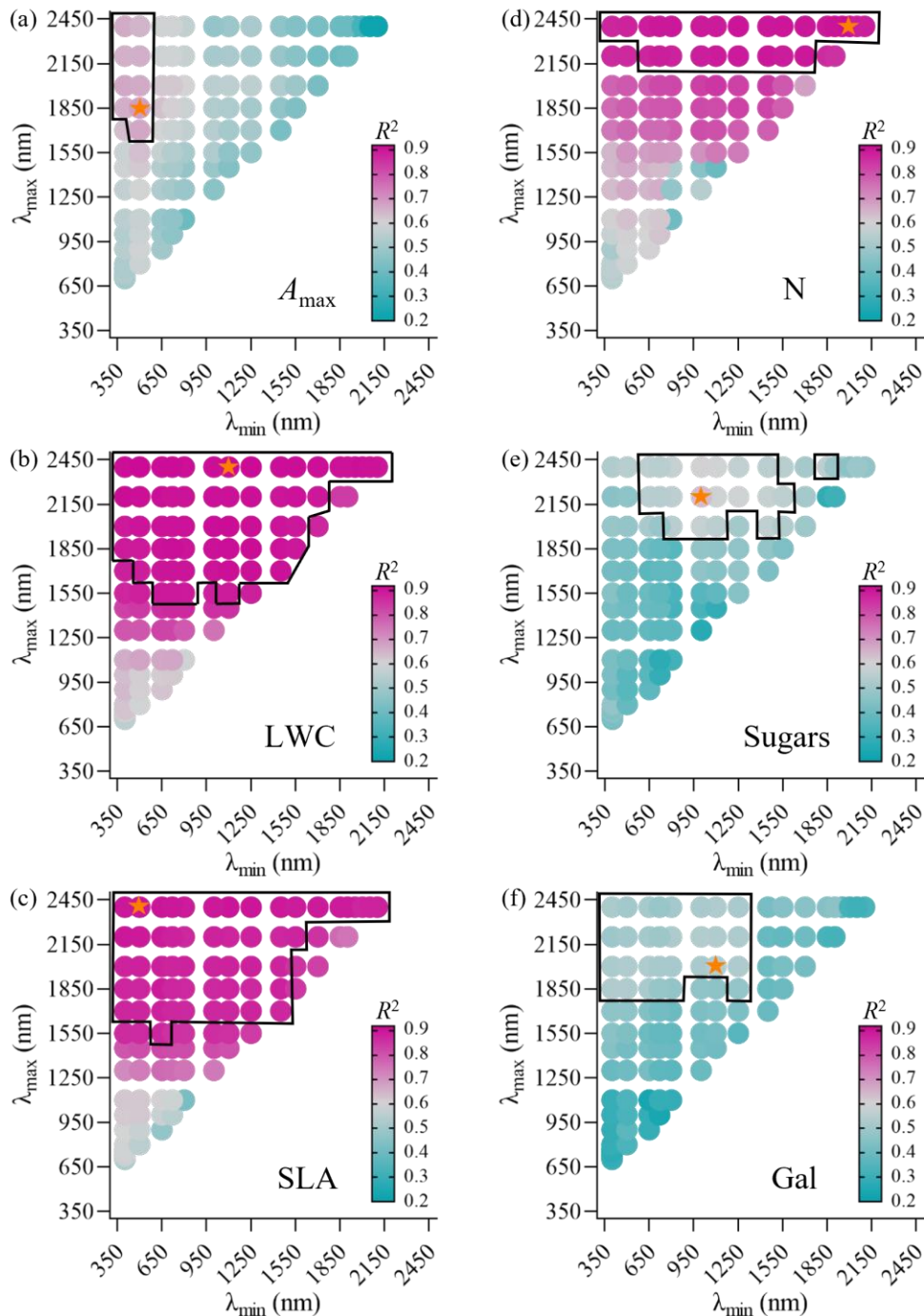
Partial Least Squares Regression (PLSR)

- Build predictive models by relating wavelengths to leaf traits measured
- Test model performances with 100 different wavelength ranges
- Model performance parameters: R^2 , normalized RMSE

Example figure of different wavelength ranges



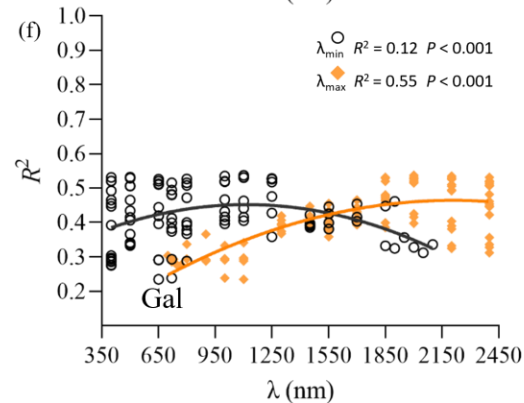
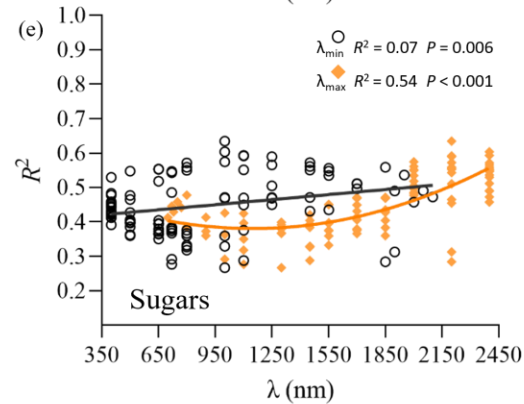
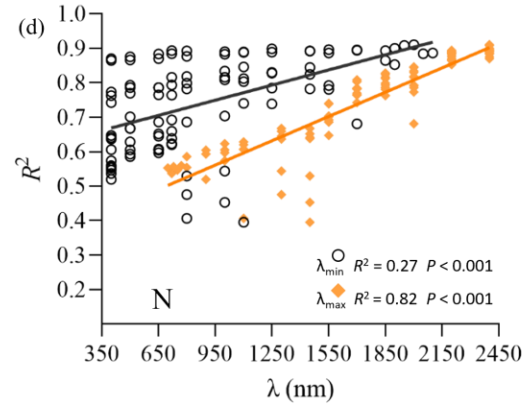
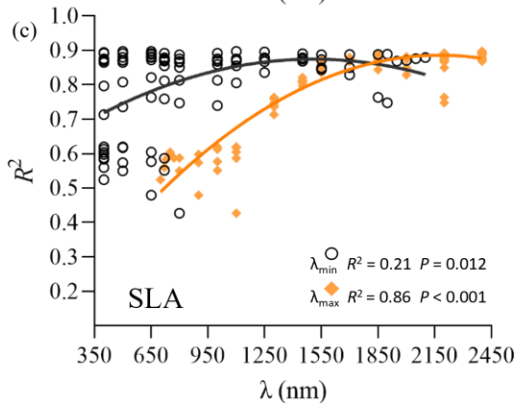
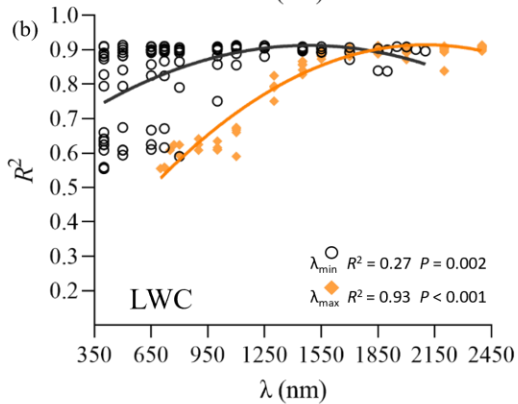
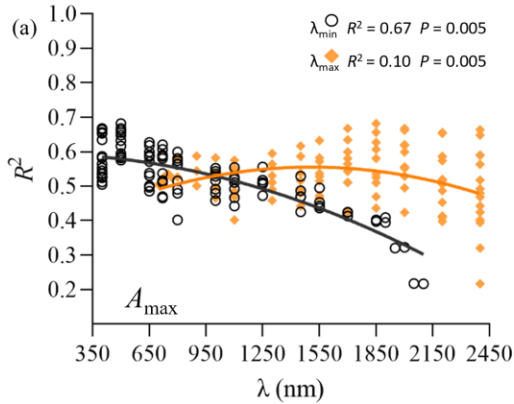
Major Findings



- Wavelength ranges including the SWIR regions (1300–2400 nm) produced enhanced model performances for six traits
- The full spectral range was not always the most optimal range

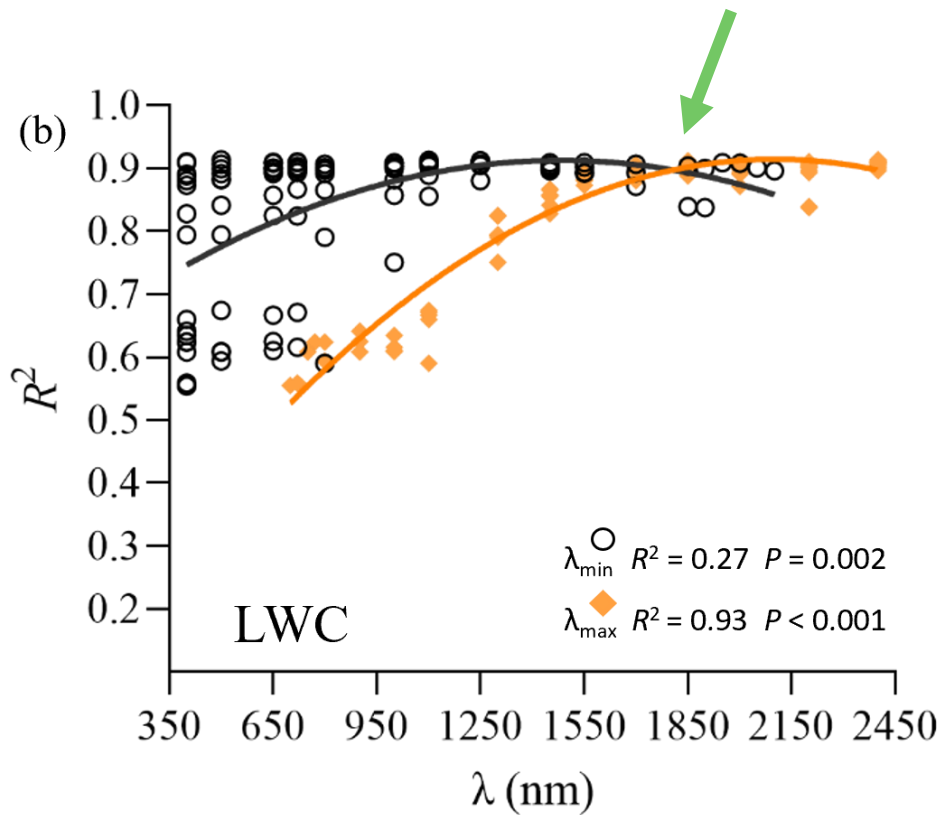
Wavelength ranges
[λ_{\min} , λ_{\max}]

Major Findings



- PLSR model performance was significantly associated with both the starting or ending wavelengths
- The importance of wavelength ranges differs depending on leaf traits

Wavelength ranges
 $[\lambda_{min}^{\circ}, \lambda_{max}^{\diamond}]$

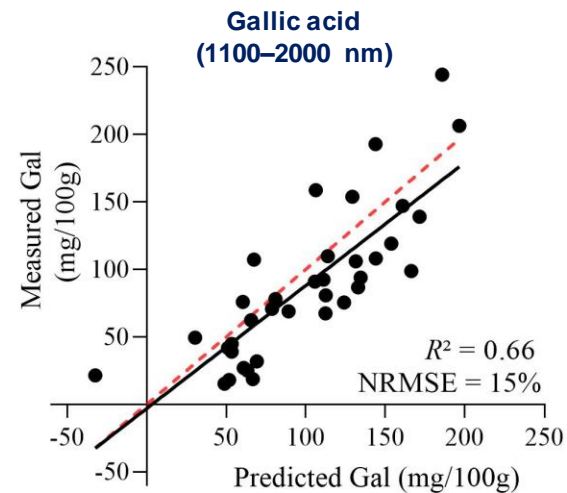
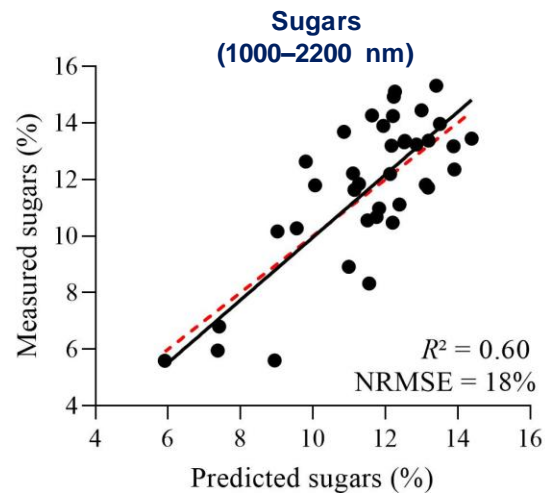
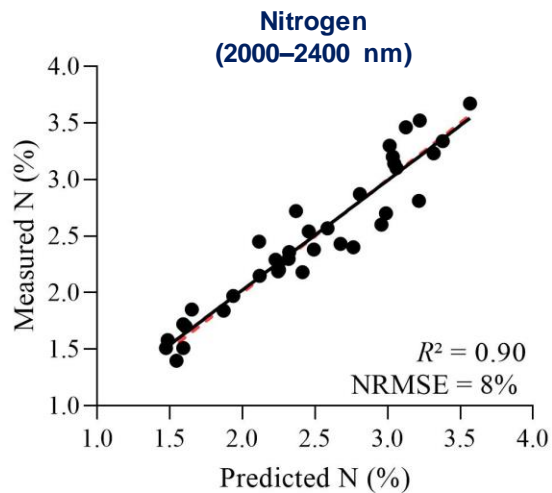
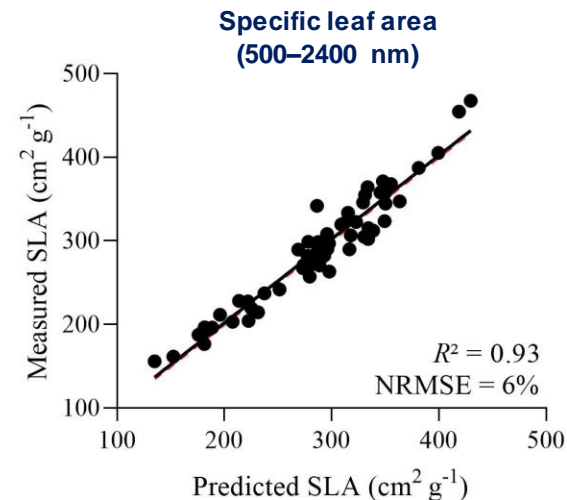
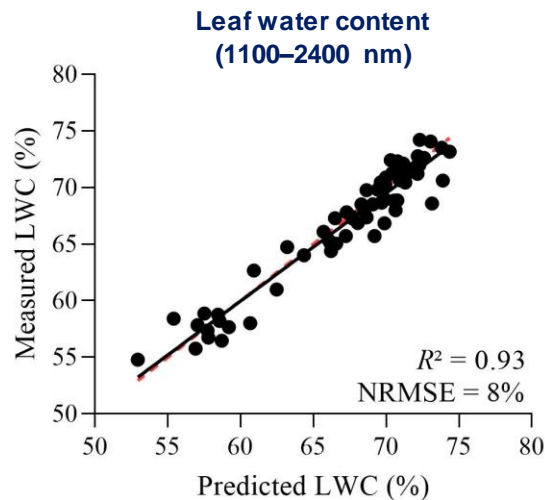
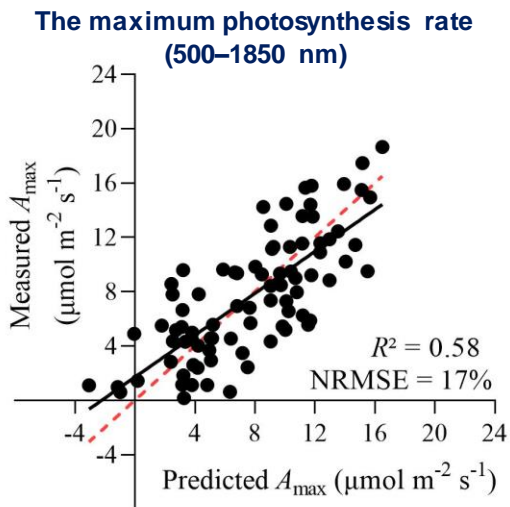


- The modeling performance (R^2) shows a convergence in the effects of starting and ending wavelengths, particularly at the wavelengths corresponding to known absorption features



Final trait models using the most optimal spectral regions

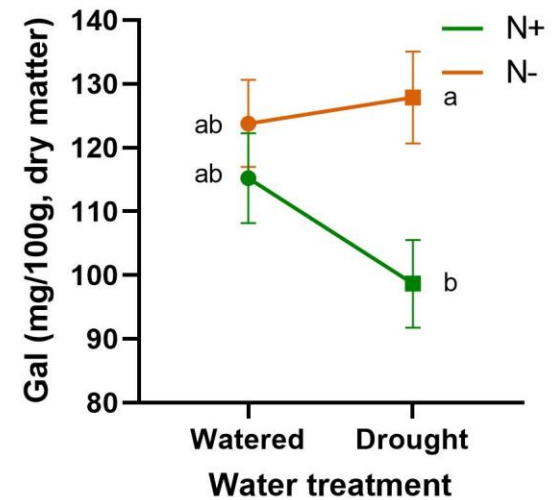
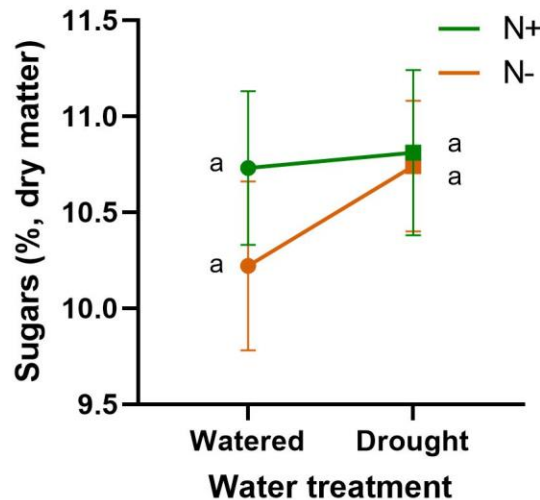
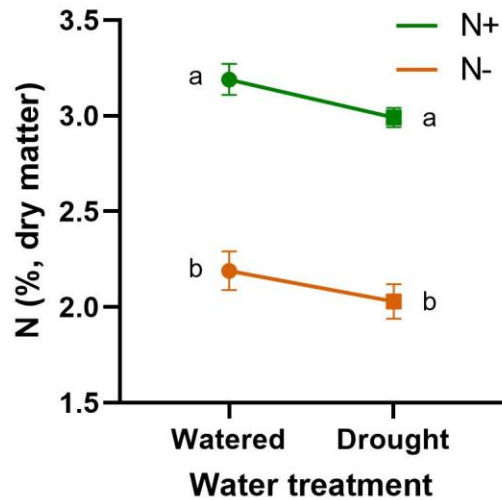
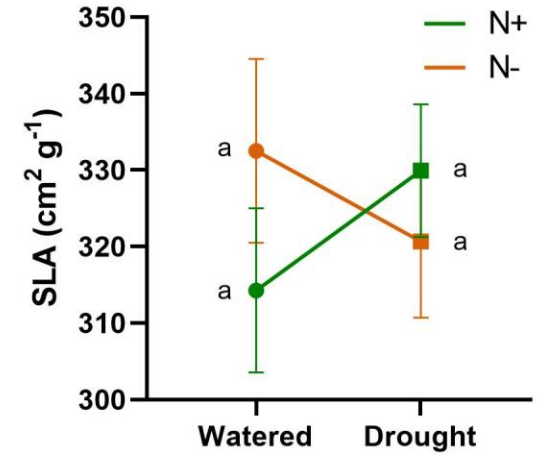
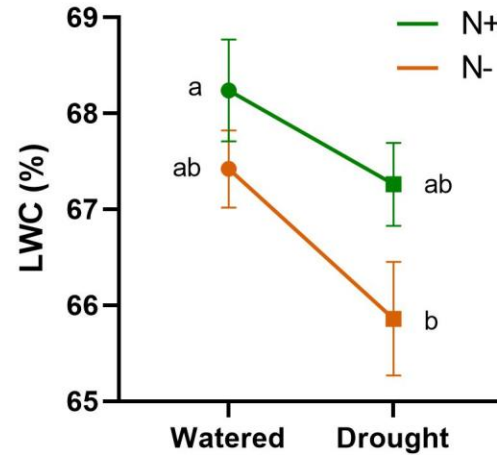
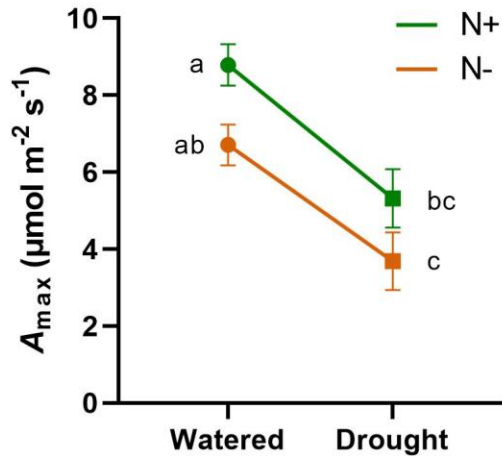
Major Findings



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Trait retrievals + Spectral phenotyping (Water & Nutrient stress) **Major Findings**



Leaf functional traits predicted from spectral data show responses to stress treatment provide detailed information on shifts in tree health status

- Standardized analyses, spectral measurements, and robust statistical modeling allow us to build trait models to retrieve relevant foliar traits for stress monitoring
- The inclusion of short infrared wavelength ranges (1300–2400 nm) was essential in enhancing the prediction of all six leaf traits using PLSR
- Future research finding the most optimal spectral regions to improve predictions of important traits in other species and conditions are needed



- Report with predictive models predicting leaf functional traits related to growth, nutritional status, defense
- Public presentation of findings at HTIRC and Digital Forestry meetings at Purdue, CAFS annual meeting, and international conferences (ESA and AGU)
- A manuscript for peer-reviewed journal *in preparation* (*Methods in Ecology and Evolution*)



- Generate outcomes that can directly inform potential management decisions involving forest plantation management through more efficient and specific characterization of tree health using RS data
- This project will be at a national-scale and relevant for all industry members



Acknowledgements



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Plant-Insect Chemical Ecology Lab

Regeneration and Restoration Silviculture Lab

Forestry Entomology Lab

Mickelbart lab

