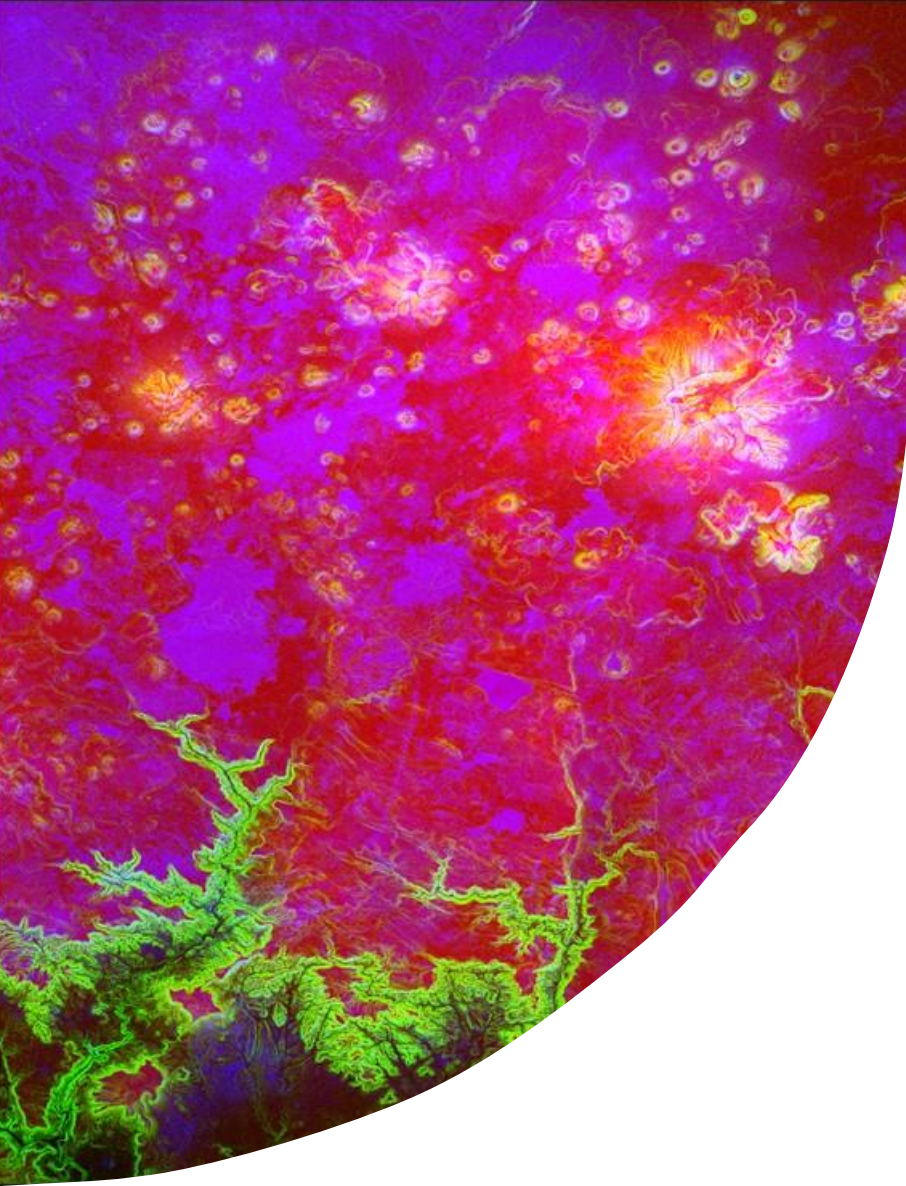




National Aeronautics and
Space Administration



Northern Arizona Ecological Conservation

Assessing Quaking Aspen Health in
Northern Arizona Using Earth Observations

Luke Collins, Margaret Cox, Ikram Morso, &
Melissa Schwan (Analytical Mechanics
Associates)

Pop-Up Project – Northern Arizona University | Summer 2025



Quaking Aspen



Image Credit: Sergey Guk

➤ What?

Deciduous species,
iconic fall color

➤ Where?

North America northern
latitudes & Western
U.S. mountainous
regions

➤ Why?

Culturally and
economically important,
ecologically vital



Aspen Ecosystem Benefits



Image Credits: Eleanor Bricetti, Ingrid Arlton

- Provide critical habitat for nesting Williamson's and Red-naped Sapsuckers
- Provide food source for deer and elk
- Act as a natural firebreak
- Protect water resources
- Fix nitrogen and sequester carbon



Threats to Aspen



Wildfires



Climate Factors



Disease & Insects



Ungulate Browsing



Conifer Encroachment

Image Credits: USFS, Jordi, Webb, Thiele, Smith, Biernacki

Partners

Image Credit: USFS



**USFS Northern
Research
Station**

Image Credit: NPS



**USFS South-
western
Region,
Arizona Zone
Forest Health**

**NPS Wupatki,
Sunset Crater
Volcano,
Walnut
Canyon**

**Northern
Arizona
University,
School of
Forestry**

**USFS
Coconino
National
Forest,
Flagstaff RD**

**USFS
Coconino
National
Forest,
Mogollon Rim
RD**

**Arizona Dept.
Of Forestry
and Fire
Managment**



Objectives



Describe Vegetation and Seasonality

Assess vegetation seasonality trends and peaks across the study area

Detect Aspen Trees

Train and validate a Random Forest model using Earth Observations and field-based measurements

Determine Change in Aspen Health

Calculate difference in vegetation index to assess aspen health change

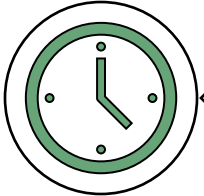
Support Management Using EO

Provide feasibility results that support aspen management using Earth Observations

Area of Interest

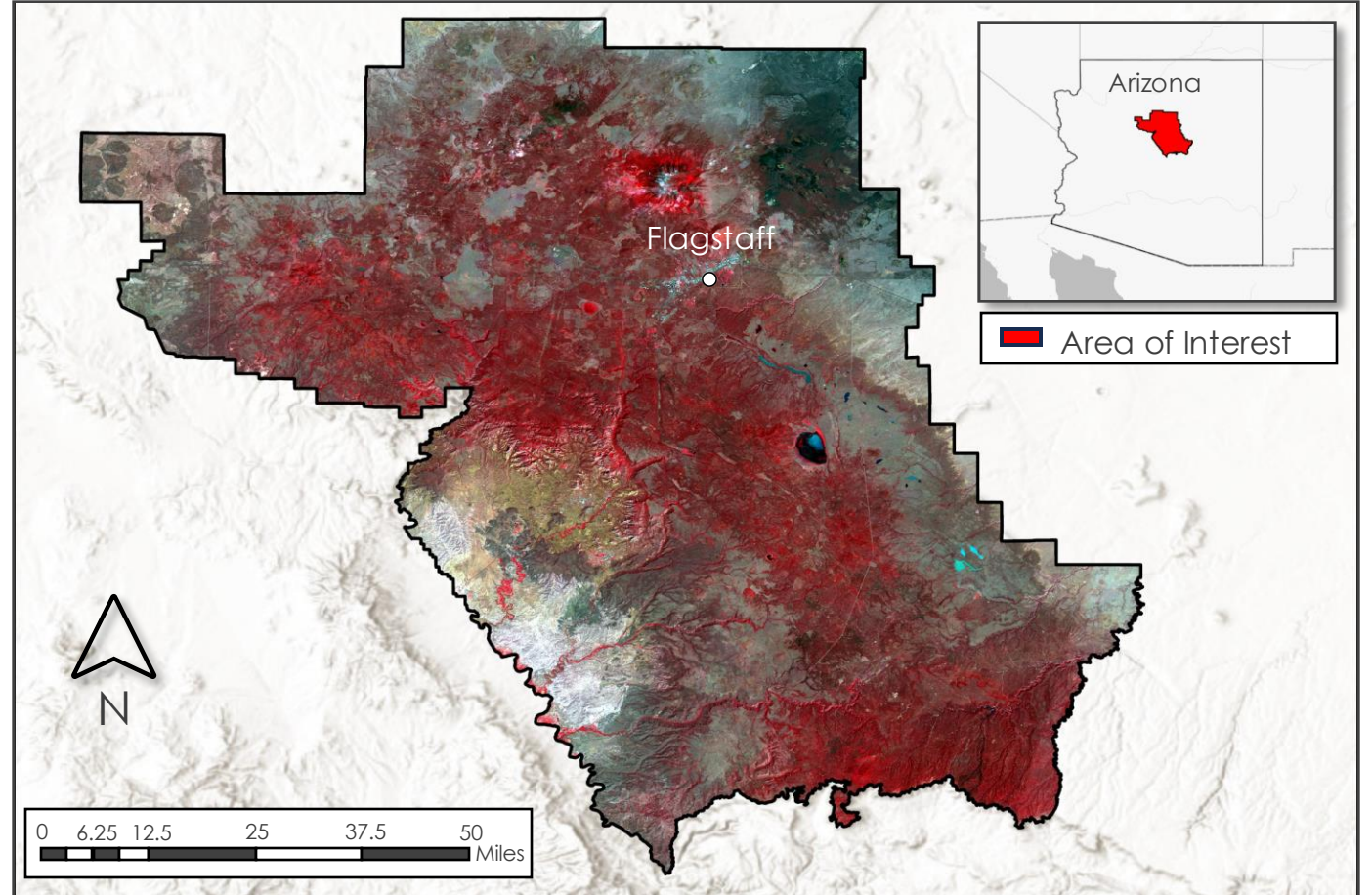


Coconino and
Southern Kaibab
National Forests



Jan. 2014 – Dec. 2024

- Area of interest spans approximately **4111.2 mi²**
- Elevational gradient from **2,600 to 12,633 ft.**



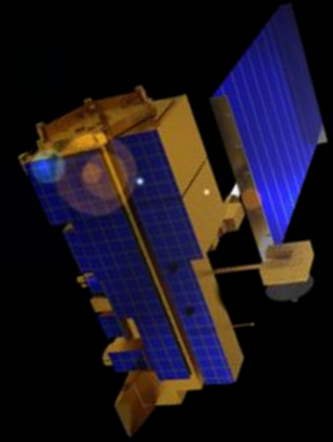
Basemap Credits: Esri, CGIAR, USGS, Sources: Esri, TomTom, Garmin, FAO, NOAA, USGS, (c) OpenStreetMap contributors, and the GIS User Community
Source: USGS Landsat 8 Collection 2, Level-2 Surface Reflectance. Sensor: OLI. © 2023 U.S. Geological Survey.



Earth Observations



Landsat 8 OLI

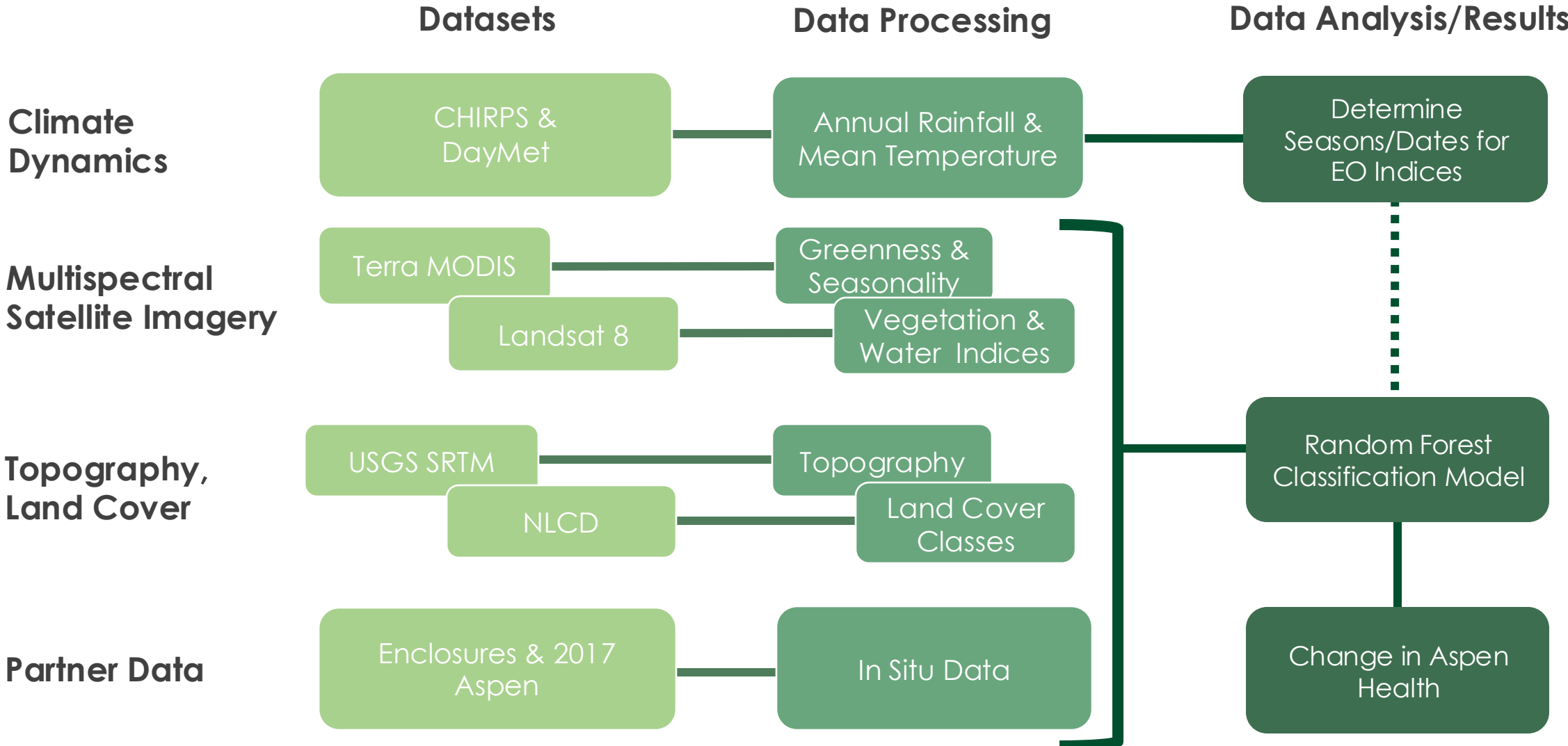


Terra MODIS

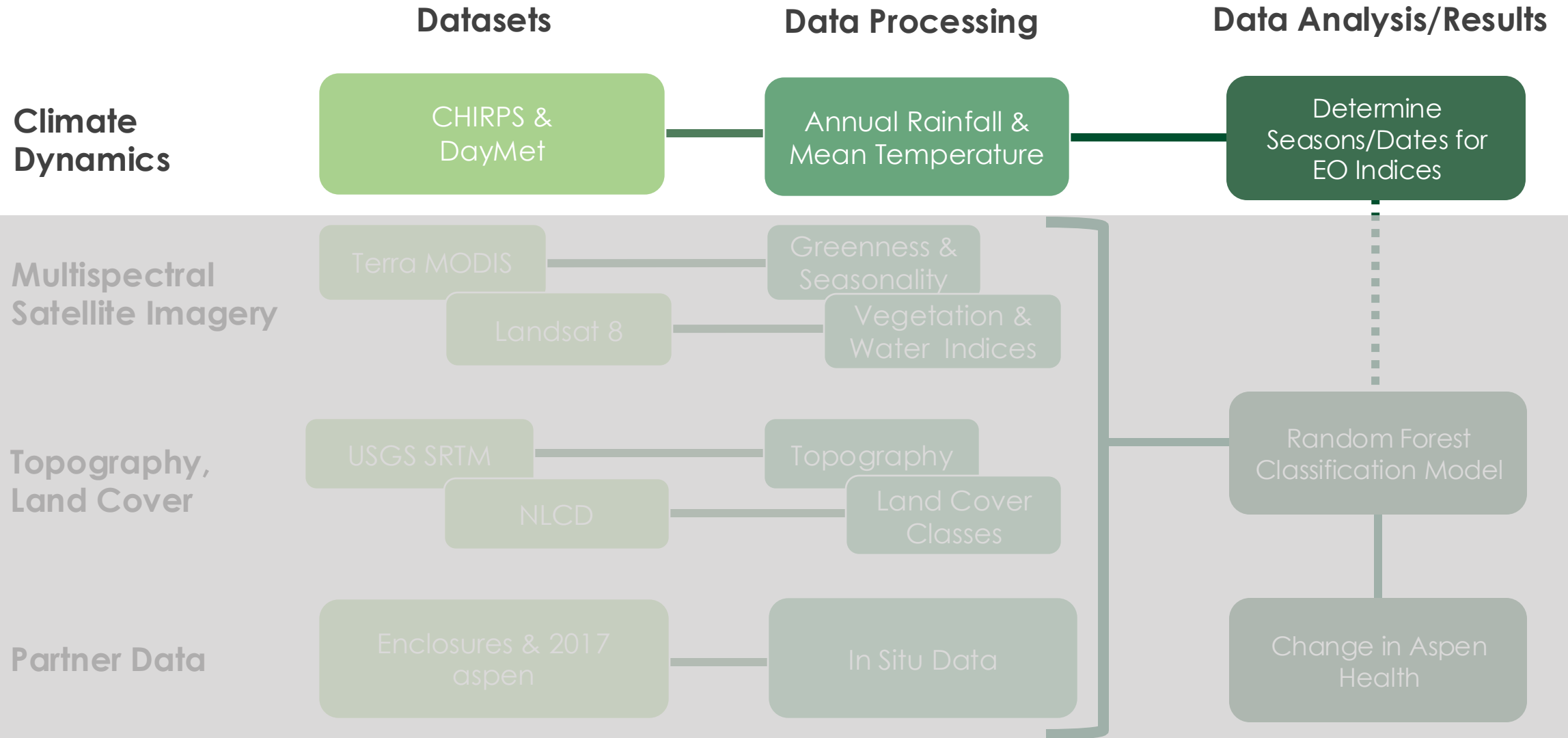
Image Credit: NASA Goddard



Methodology

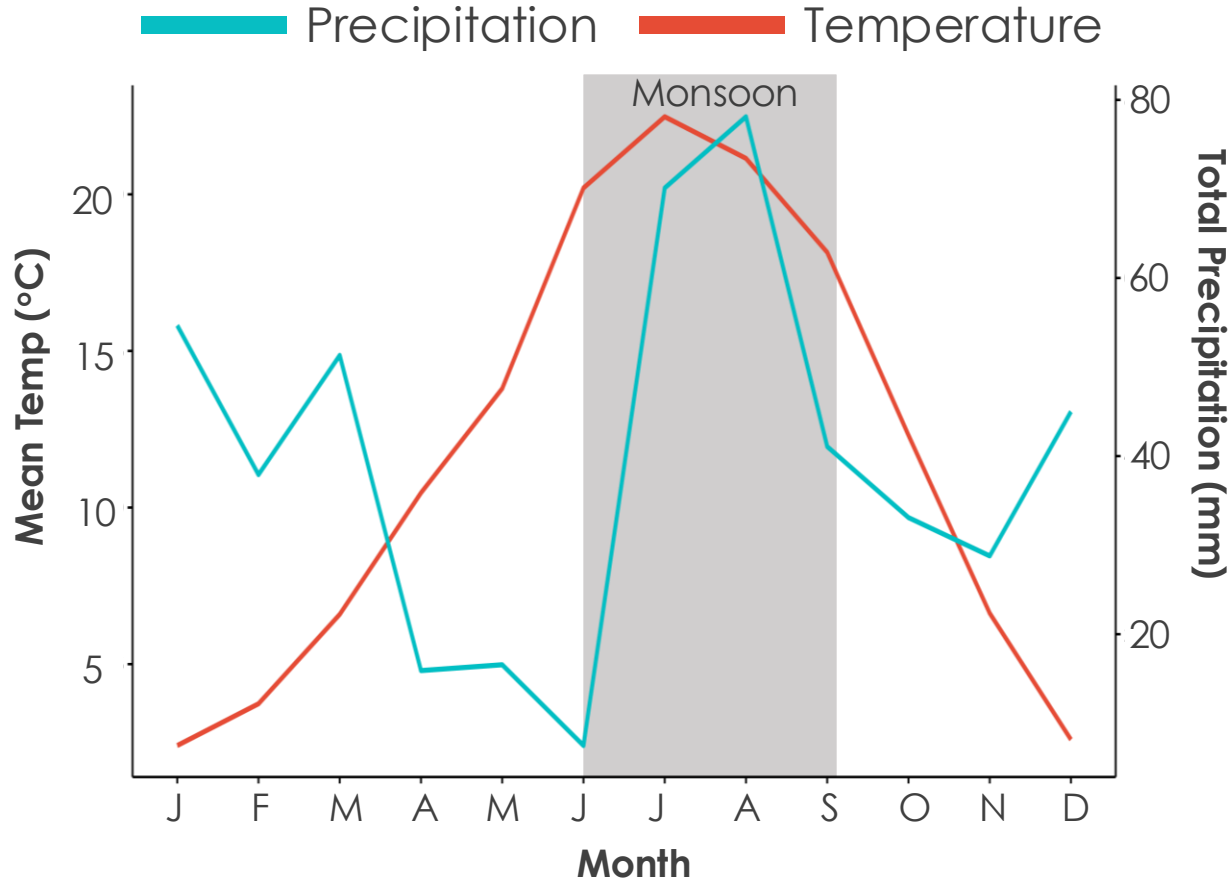


Methodology

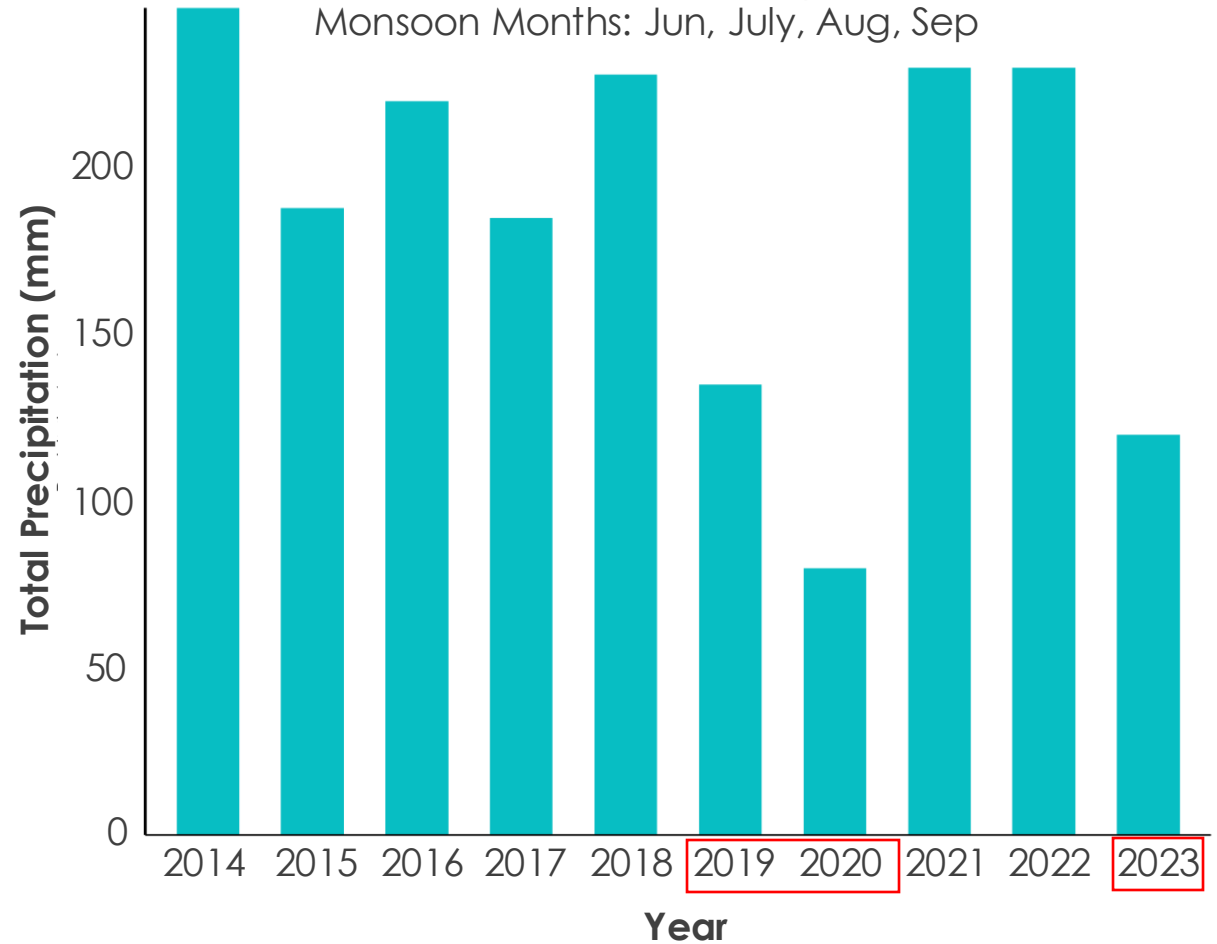


Climate Dynamics: Summer Monsoon

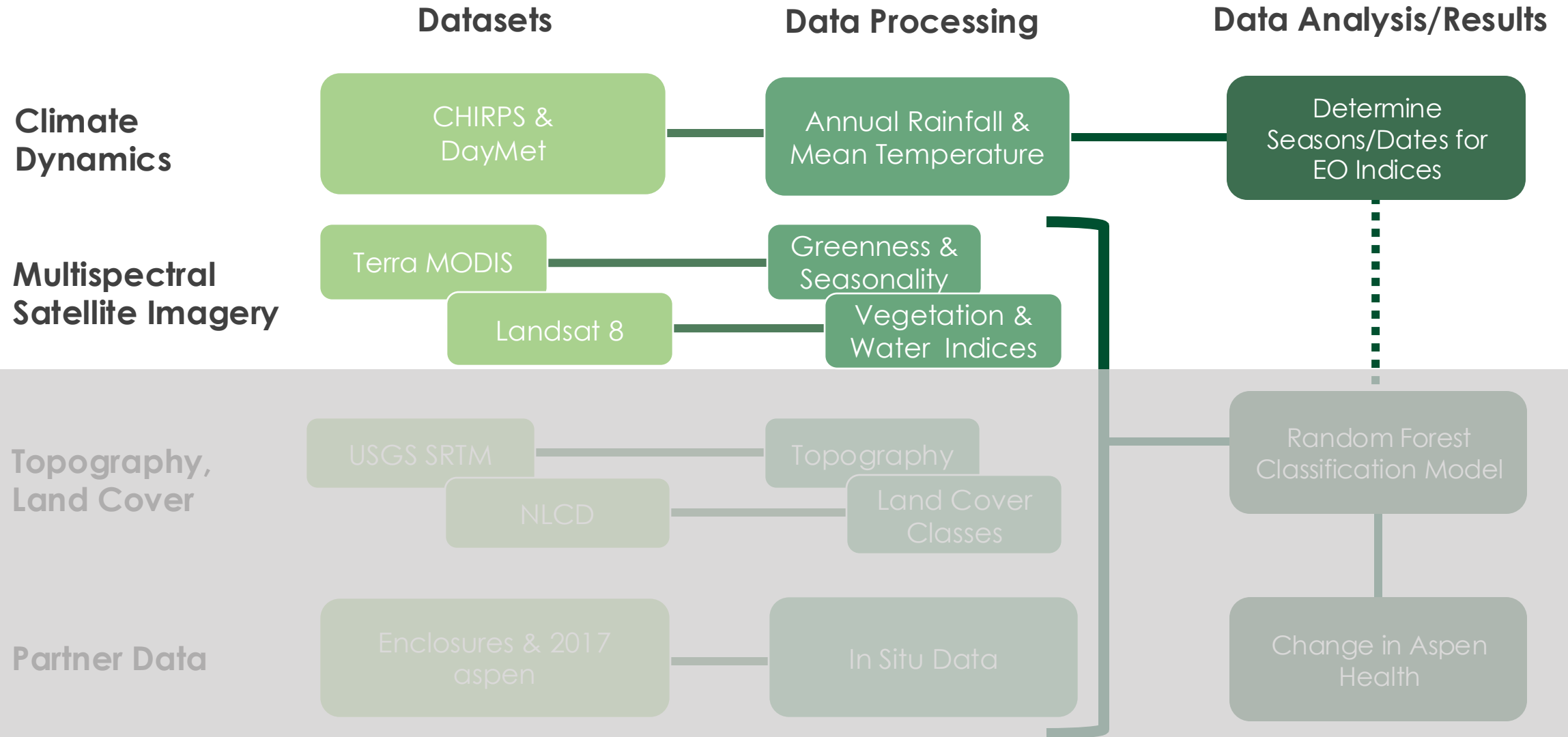
Monthly Precipitation and Temperature (averaged across 2014-2023)



Total Monsoon Precipitation 2014-2023



Methodology



Vegetation Seasonality

2014-2023

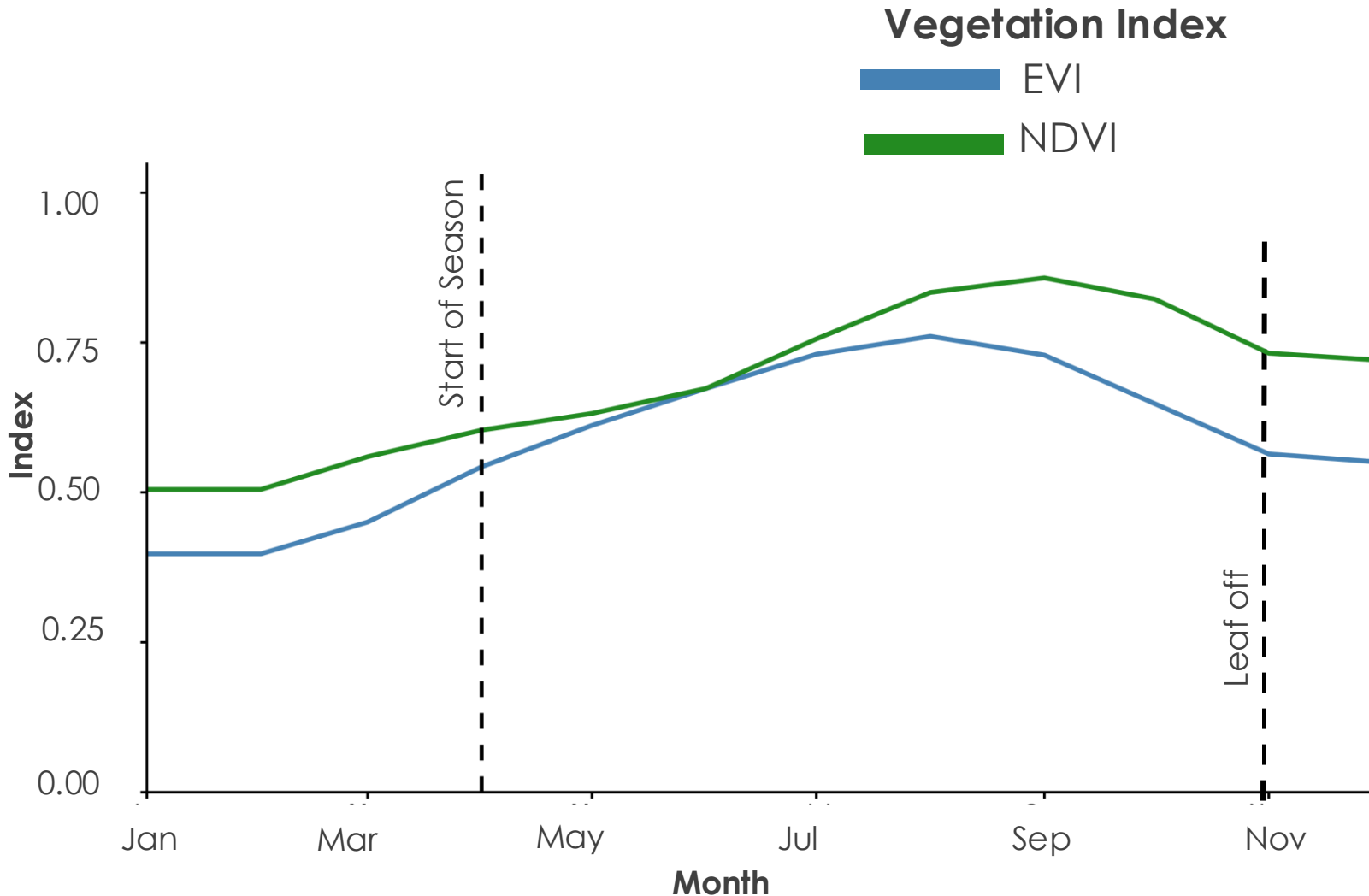
Start of season: March – April
(20% increase in NDVI from winter)

Peak growing season: August – September

End of season: End of October – early November

Dates support future analysis

Critical dates for calculating vegetation indices for the aspen classification model

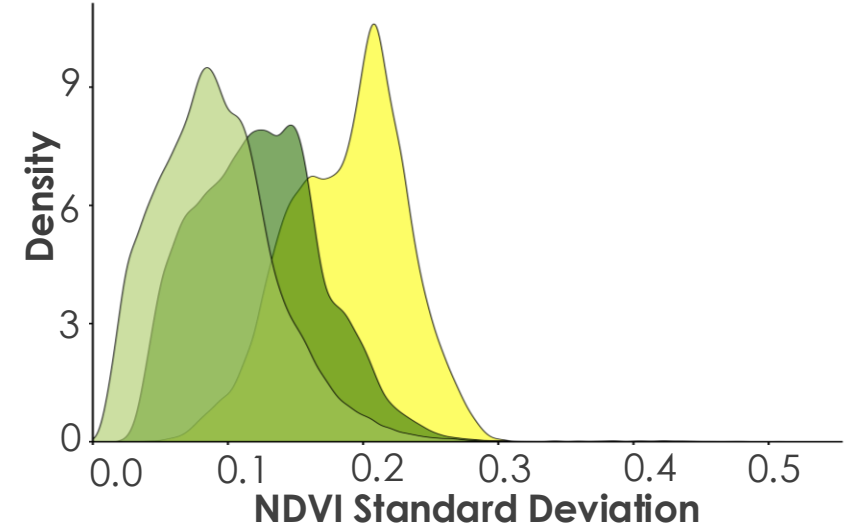


Phenological Differences in Vegetation

Normalized Difference Vegetation Index



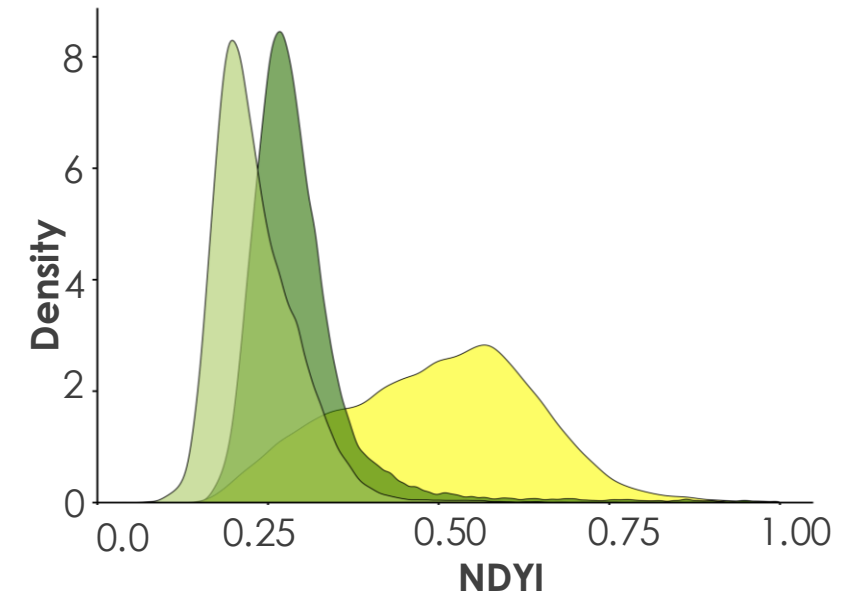
$$\text{NDVI} = \frac{\text{NIR} - \text{Red}}{\text{NIR} + \text{Red}}$$



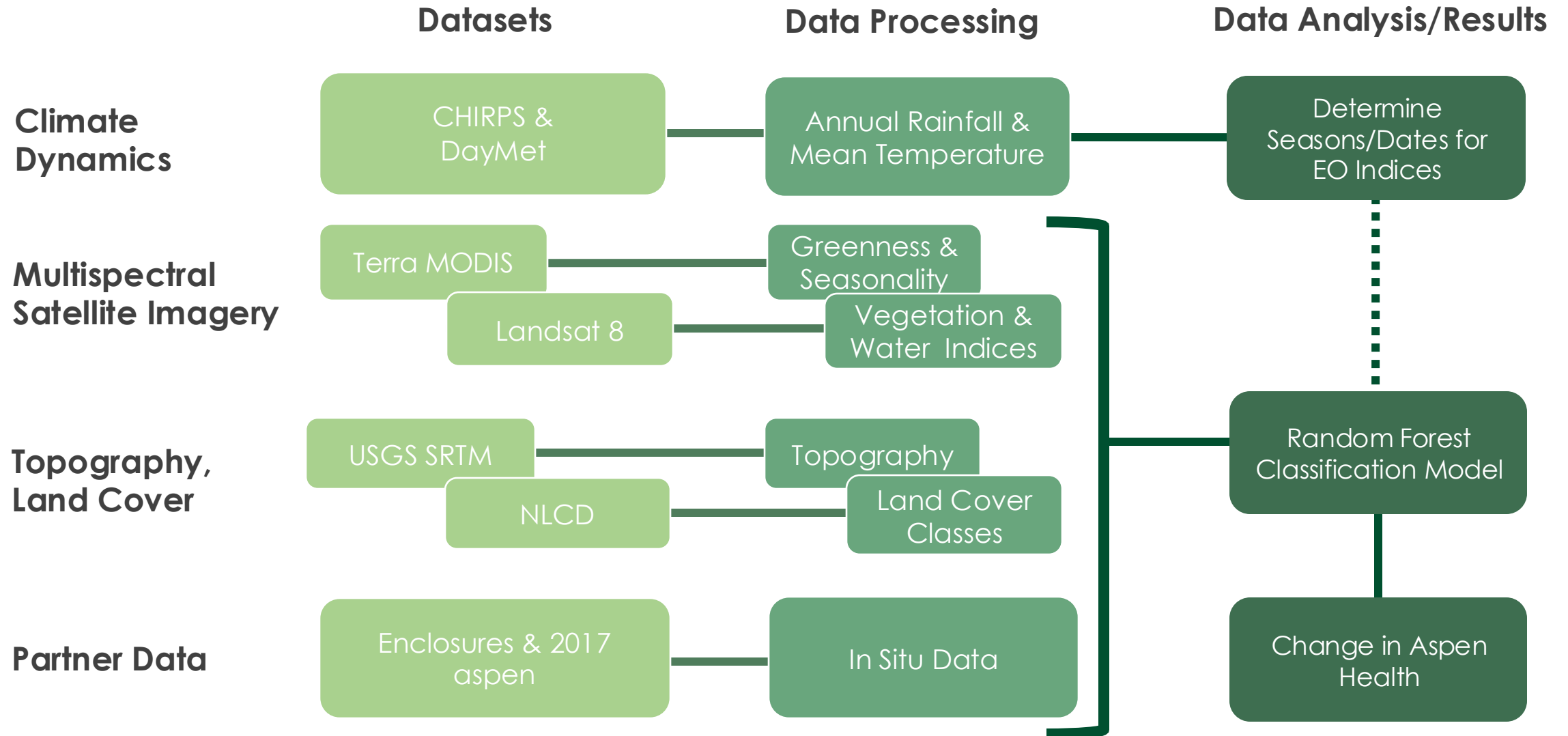
Normalized Difference Yellowness Index



$$\text{NDYI} = \frac{\text{Blue} - \text{Green}}{\text{Blue} + \text{Green}}$$



Methodology



Building the Random Forest Model

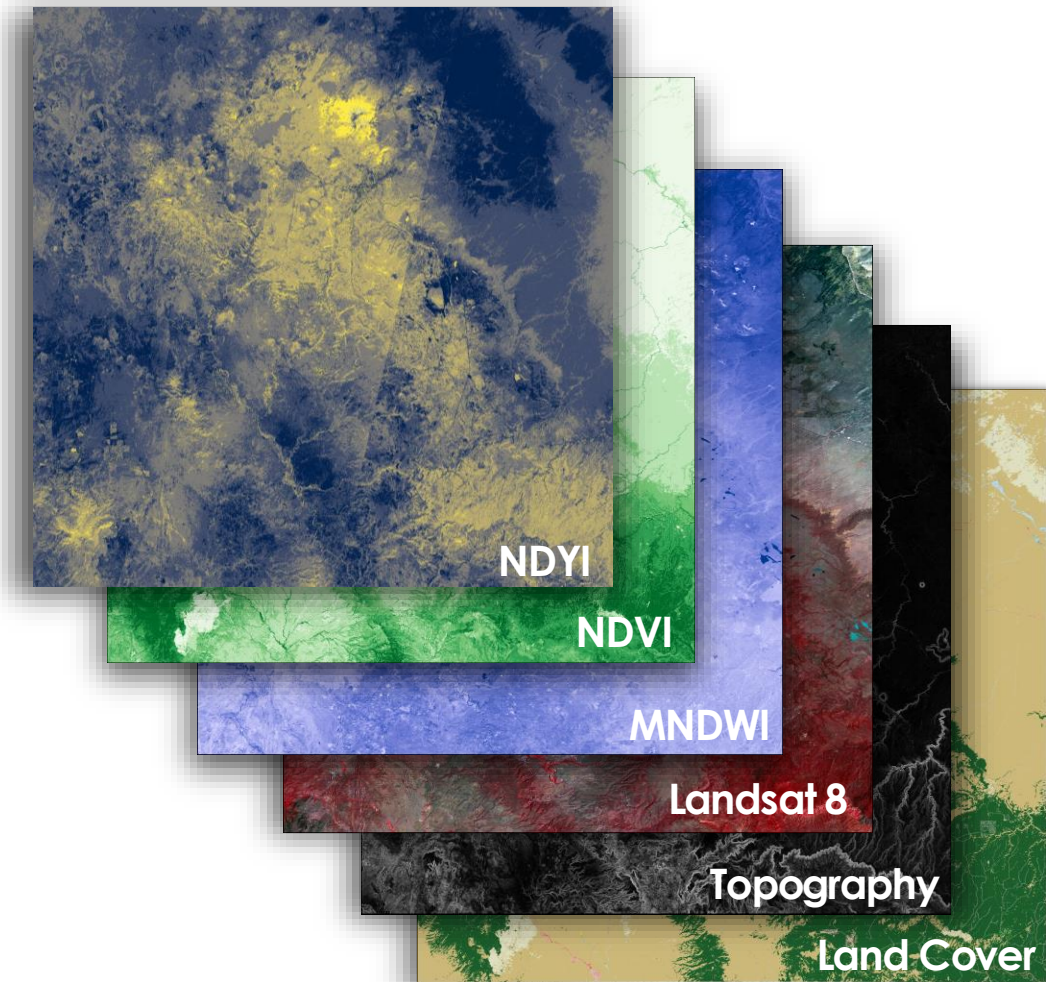
Remote Sensing data: Predictors



In situ data



Random Forest (RF) Model



Aspen Polygons

Mixed aspen polygons

Conifer polygons

Grassland/ shrub points

Polygons Sampling:
Polygon Leave-One-Out (POLO) cross-validation

Points Sampling:
Random stratified

Balanced classes:
Same number of points per class

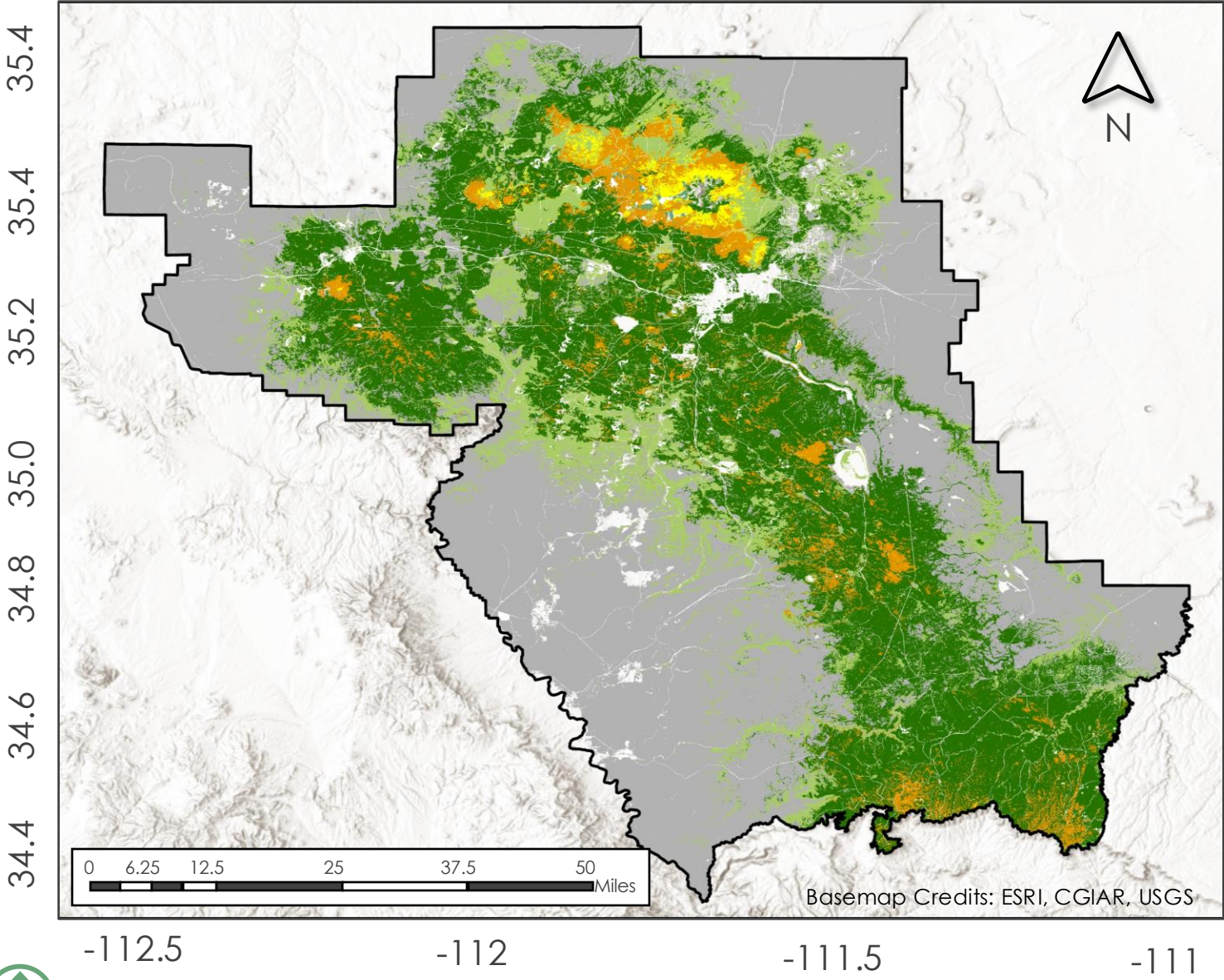
Training dataset: 75%

Testing dataset: 25%

nTrees in RF: 500

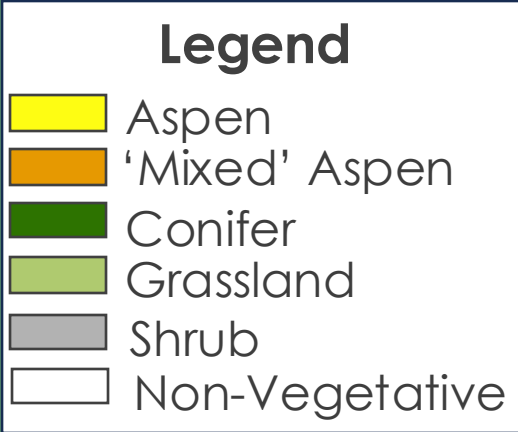


Results | Random Forest



Detected Aspen in 2017

- Model accuracy: **79.15%**
 - **41.41 sq. miles** of aspen
 - **195.45 sq. miles** of mixed aspen
- **15.5%** of forested area is aspen/mixed aspen



Results | Random Forest

Classification Statistics

Class	Producer's Accuracy (Recall)	User's Accuracy (Precision)	F1-score
Aspen	0.754	0.736	0.745
Mixed-Aspen	0.678	0.727	0.702
Overall Accuracy (All classes)			0.7915

Predictor Variable Importance

High Importance

1. Elevation
2. Oct. NDVI
3. Oct. NDYI
4. Nov. NDVI
5. NDVI StdDev

Mid Importance

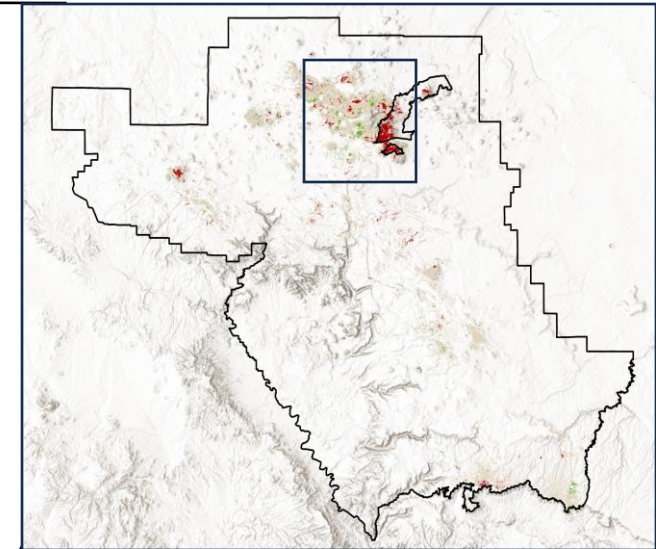
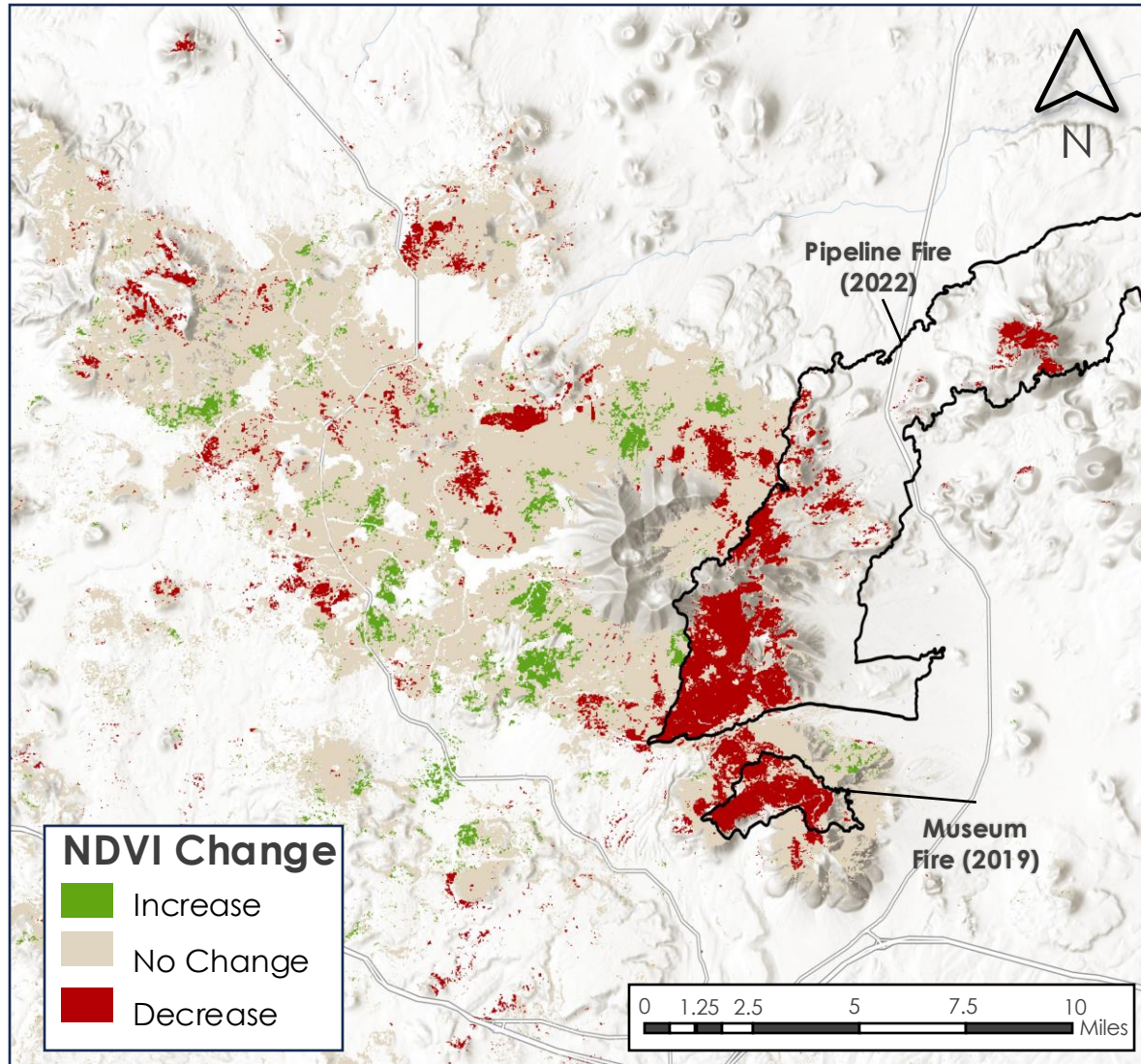
1. NDVI Mean
2. MNDWI StdDev
3. Sep. Red
4. Aug. NDVI
5. Aug. NIR

Least Importance

Slope
Green band StdDev
Aspect
Sep. NDYI



Results | Change in Aspen Health (2017-2024)



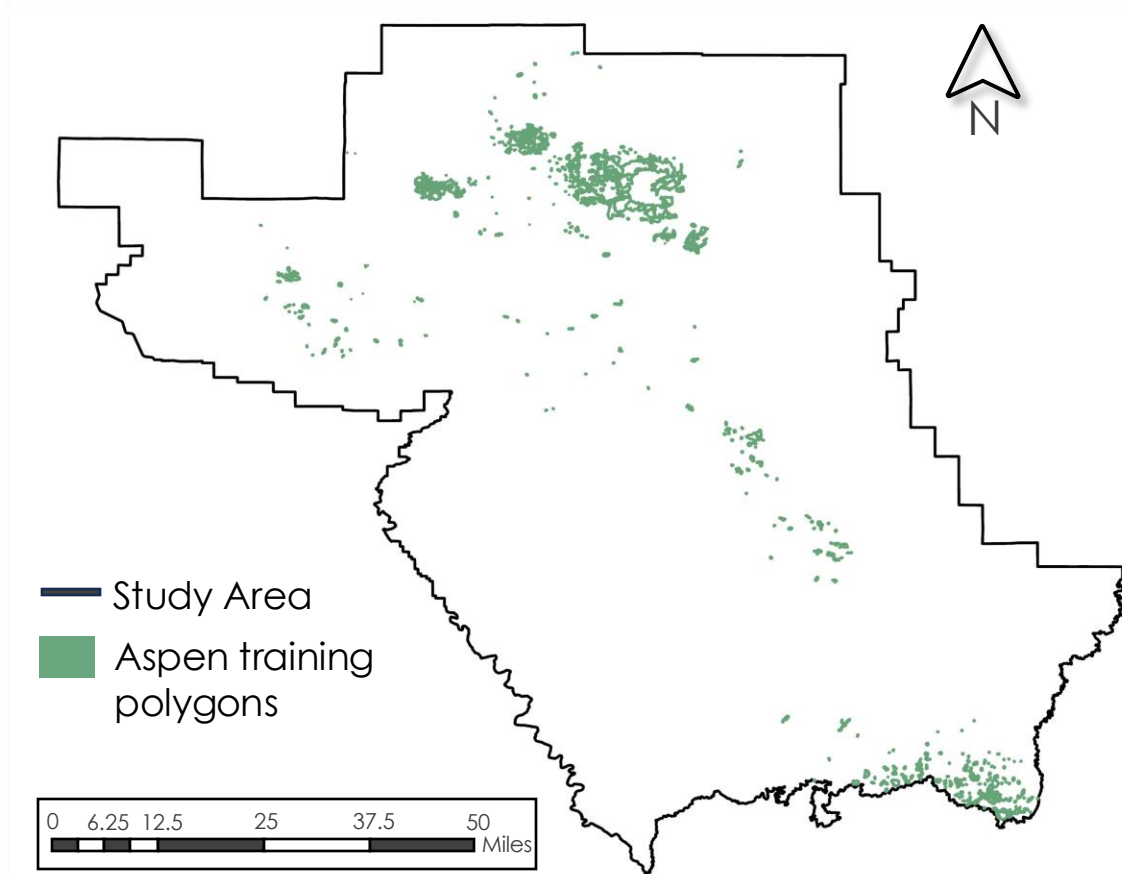
Basemap Credits: ESRI, CGIAR, USGS

- **NDVI Increase:** 6.4 mi²
- **No Change:** 324.7 mi²
- **NDVI Decrease:** 30.6 mi²
- **Net Decrease:** 24.8mi²
- **Potential Causes:** Fires, Drought, Diseases

Basemap Credits: ESRI, CGIAR, USGS



Limitations & Uncertainties



Resolution Limitation

- Coarse (30 m) spatial resolution in detecting aspen patches
- Hyperspectral imagery could improve aspen identification

Training and Testing Data

- Limited coverage of in situ dataset

Imagery Complications

- Cloud cover in fall imagery limited the ability to use NDVI to capture leaf change
- Temporal alignment & mosaics



Feasibility & Partner Implementation

01

It is feasible to use NASA Earth observations to identify aspen with **74.5%** accuracy

02

The assessment of vegetation health in areas of predicted aspen provides insight into priority areas for **management** of aspen & mixed-aspen areas

03

The **replicable** aspen detection **workflow** offers partners a tutorial that can be applied to recent years imagery



Conclusions

- Understanding **vegetation seasonality** prior to applying machine learning models enables the selection of relevant datasets, helping to **reduce overfitting** and improve **classification accuracy**.
- Elevation, October **NDVI** & **NDYI**, and November **NDVI** were the most **important predictors** of aspen in the RF classification.
- **Negative NDVI** change could indicate **vegetation loss or health decline** in aspen stands impacted by wildfire, for example.



Acknowledgements

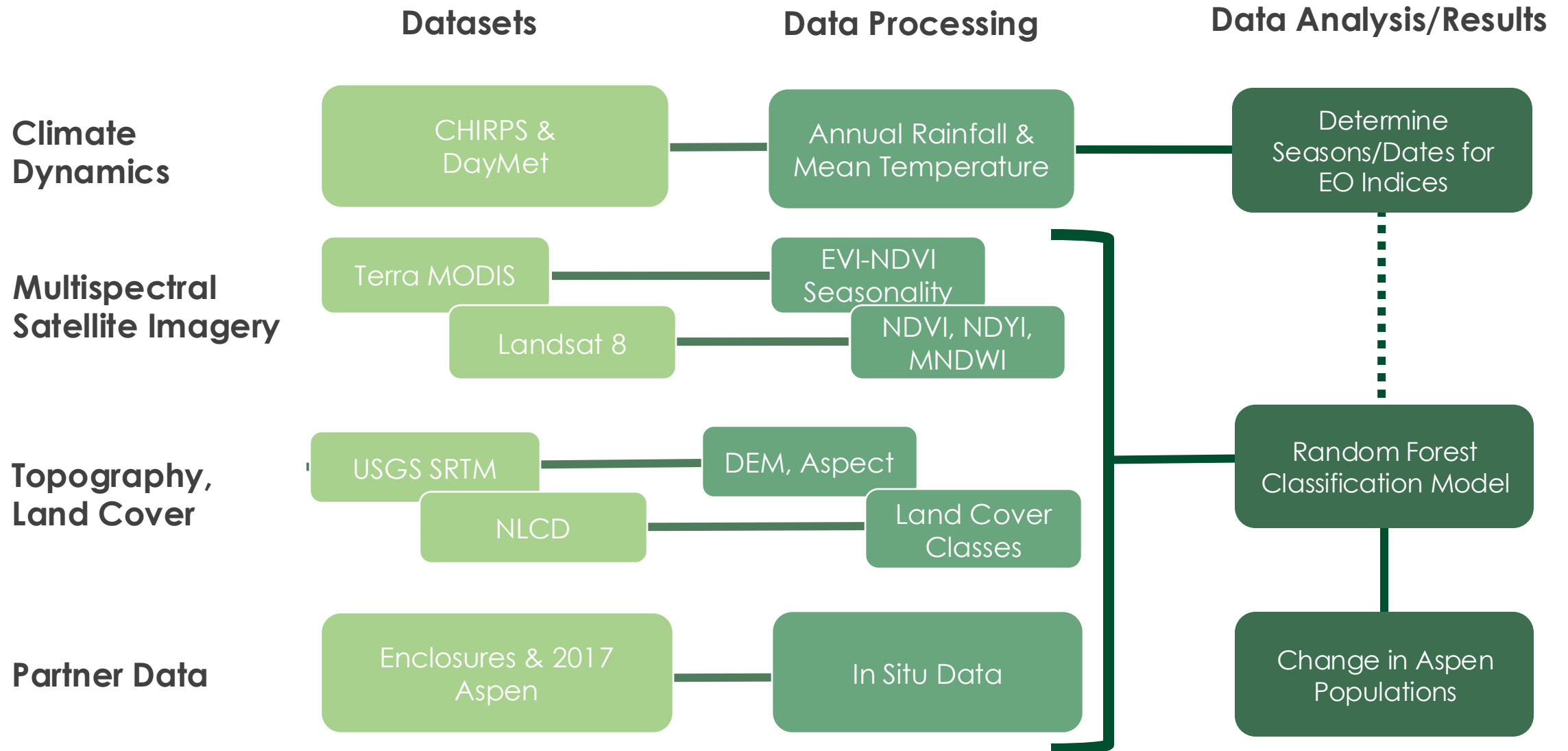
- DEVELOP Lead: Jane Zugarek
- Science Advisors: Dr. Paul Arellano, Dr. Alexander Shenkin, Dr. Xia Cai
- DEVELOP Support: Kait Lemon (Collaboration Coordinator), Maya Hall (Impact Analysis Fellow)
- Partners at USFS, NPS, NAU, AZDFFM: Mark Nabel, Andy Pigg, Christopher Curley, Jack Daugherty, Noah Bard, Mark Szydlo, Aly McAlexander, Kristen Waring, Connor Crouch, Amanda Grady, Nicholas Wilhelmi, Mary Price, Brett Miller
- *This work utilized data made available through the NASA Commercial Smallsat Data Acquisition (CSDA) Program.*



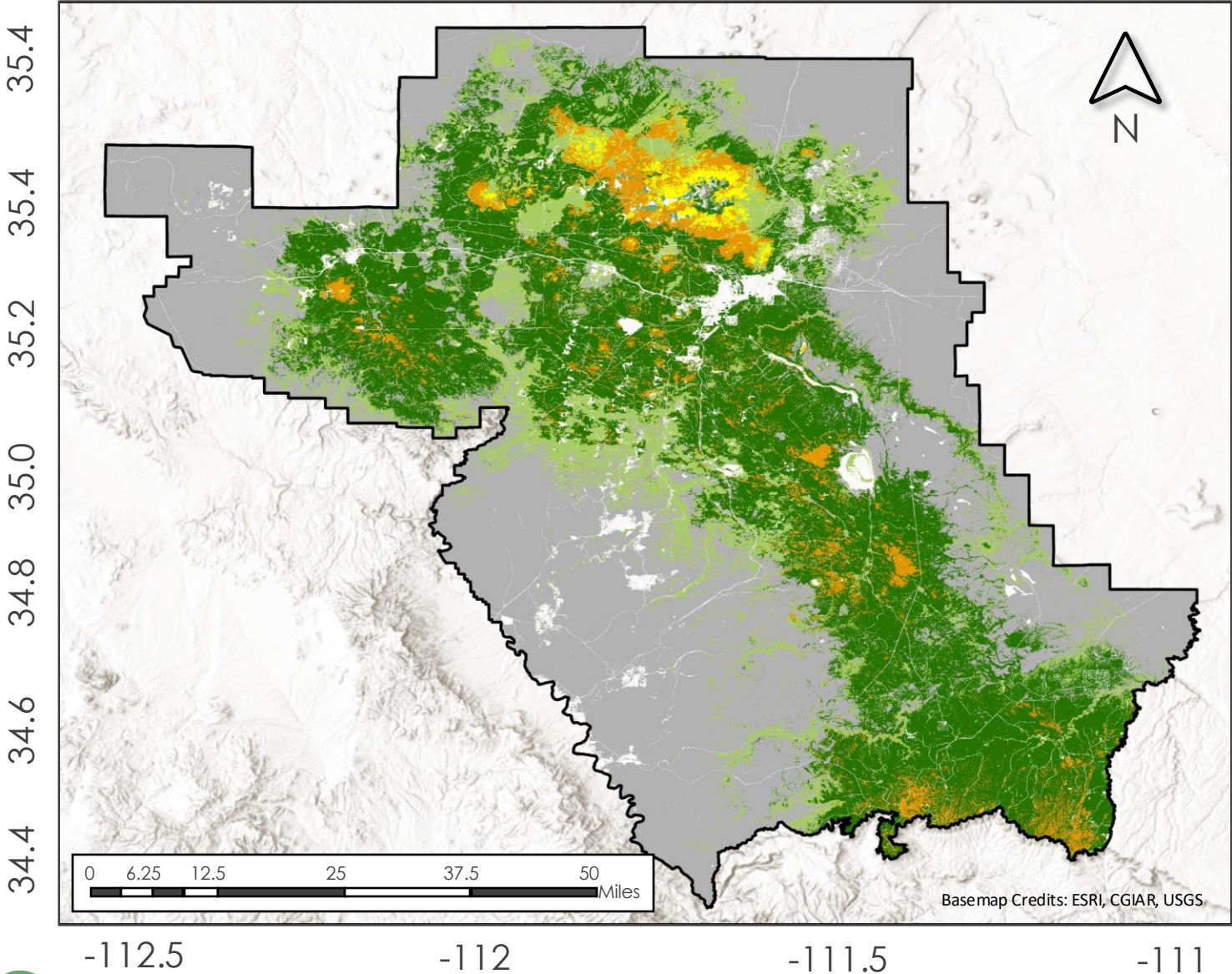
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Methodology



Results | Random Forest



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