

NORTHERN BRAZIL AGRICULTURE

Measuring Soybean Yields in Northern Brazil
During El Niño-Southern Oscillation
Conditions to Evaluate Trends in
Agricultural Production and Support Crop
Forecasting, 1984 – 2023

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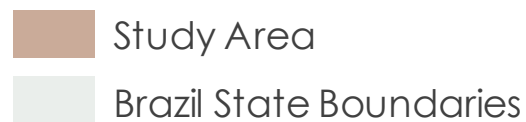
Grayson Shanley Barr

Sofya Goncharenko



PRESENTATION ROAD MAP

- Study Area and Period
- Study Objectives
- Partners
- Community Concerns
- Earth Observations
- Methods
- Results
- Errors and Uncertainty
- Feasibility
- Future Recommendations
- Acknowledgements



PROJECT TEAM



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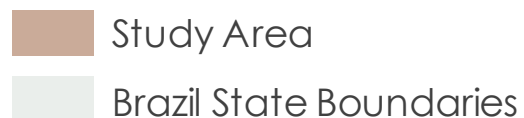
Study Area & Period

Study Area

- Northern & central Brazilian states of Pará, Tocantins, Bahia, and Mato Grosso

Study Period

- 1984–2023
- El Niño years: 1987–88, 1991–92, 1994–95, 1997–98, 2002–03, 2009–10, 2015–16, 2023–24



Background

- Reductions in crop yields due to climatic variability in Brazil affect food supply globally
- El Niño-Southern Oscillation (ENSO) causes variation in the Brazilian climate
- As El Niño returns, it is critical to assess the impact of ENSO conditions on crop production to inform estimates of agricultural production in the country



Image Credit: Photo by [PROJETO CAFÉ GATO-MOURISCO](#) on [Unsplash](#)

Study Objectives

- Examine the relationship between soybean production yield, normalized difference vegetation index (NDVI), and ENSO conditions
- Produce a time series analysis from 1984 to 2023 comparing crop yield and environmental conditions during ENSO phases
- Conduct statistical analysis to examine relationships at the state level



Image Credit: Photo by [PROJETO CAFÉ GATO-MOURISCO](#) on [Unsplash](#)

Community Concerns

- Brazil has emerged as one of the top exporters of agricultural commodities, supplying over 200 countries and territories around the world
- Brazil engages in 50% of world soybean trade and manages a top global corn production
- With Brazil's changing climate and the oncoming ENSO effects, global food security concerns are rising

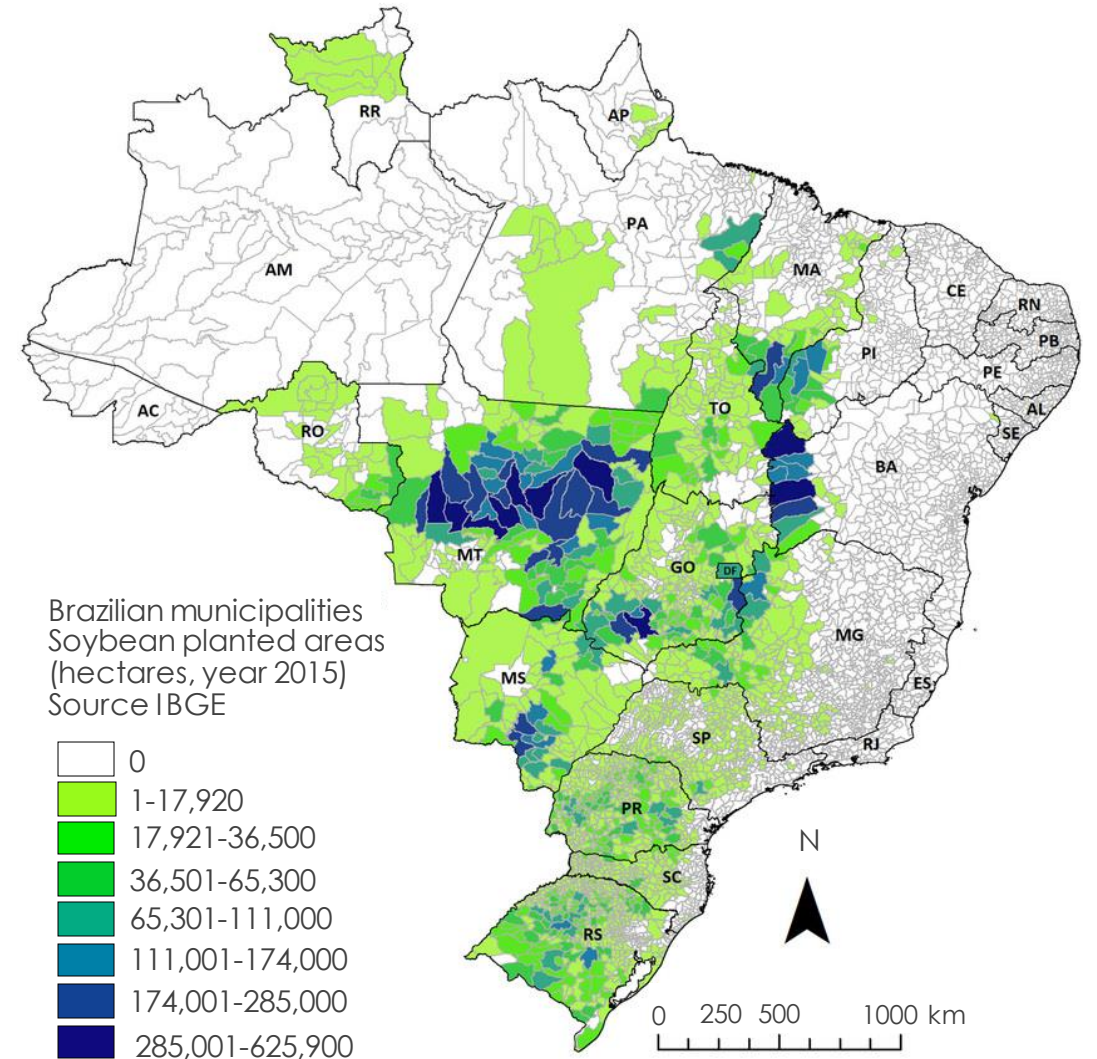
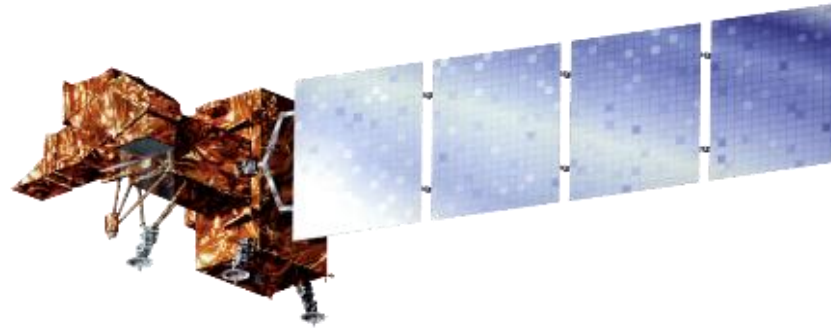


Image Credit: Martinelli et al., 2017

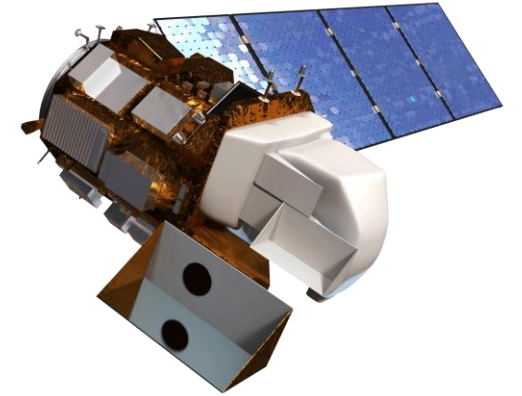
Earth Observations



Landsat 5 TM

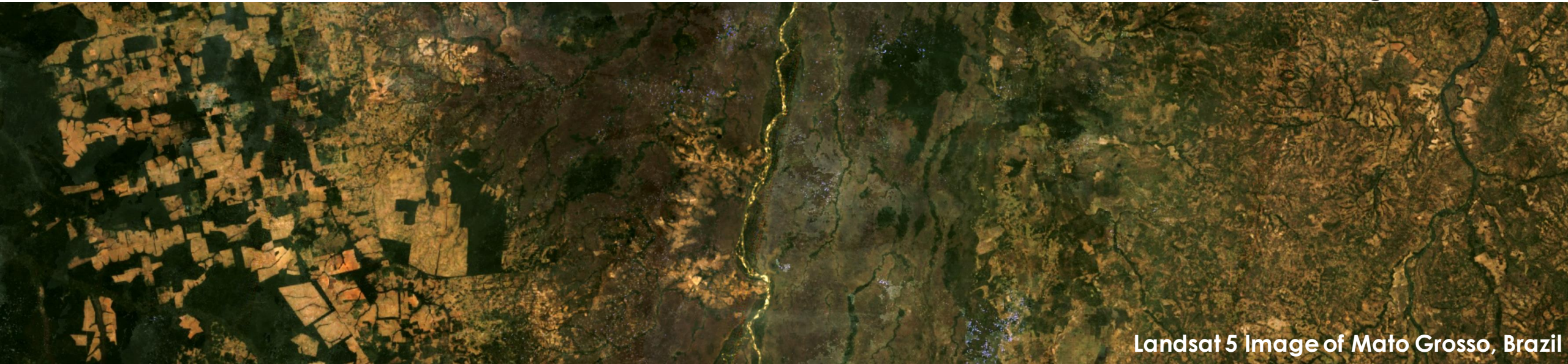


Landsat 7 ETM+



Landsat 8 OLI

Image Credit: NASA



Landsat 5 Image of Mato Grosso, Brazil

Project Partners

We are working with partners at the United States Department of Agriculture (USDA):

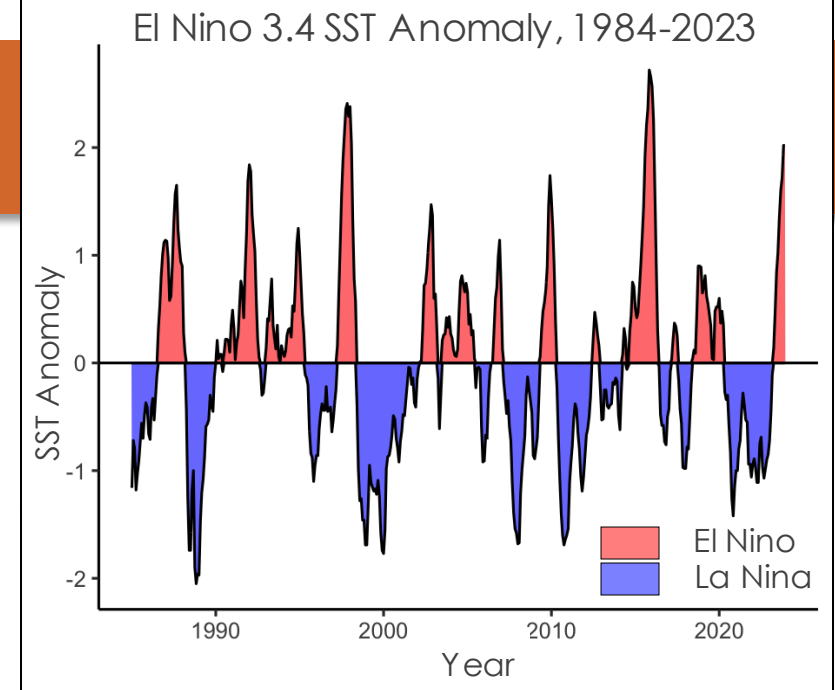
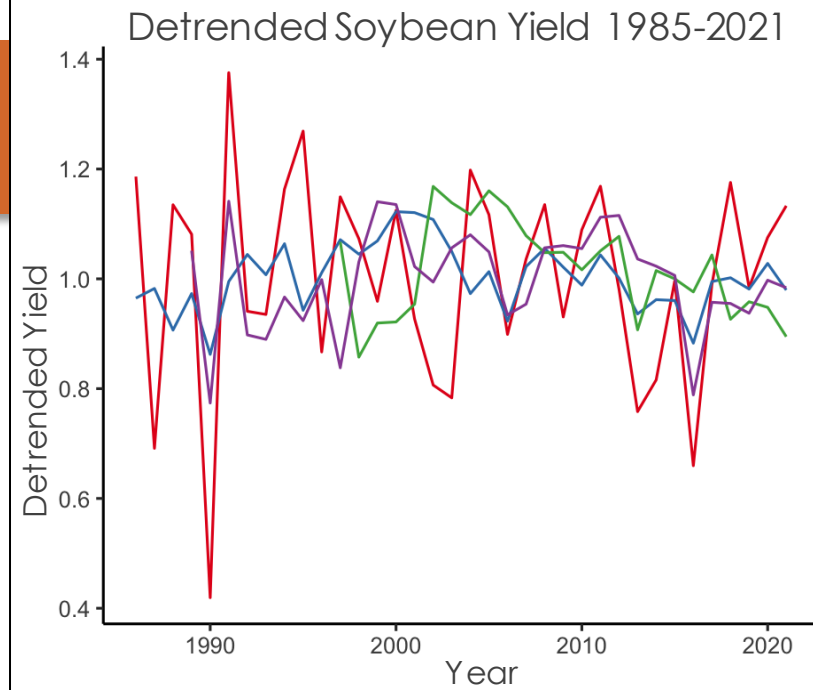
- USDA Foreign Agricultural Service International Production Assessment Division
- USDA Office of the Chief Economist and World Agriculture Outlook Board



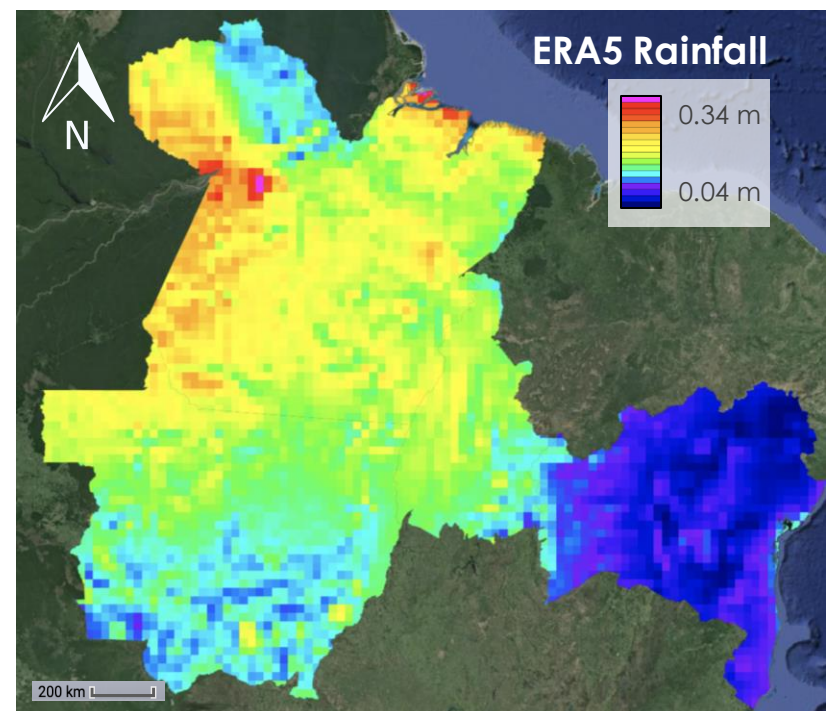
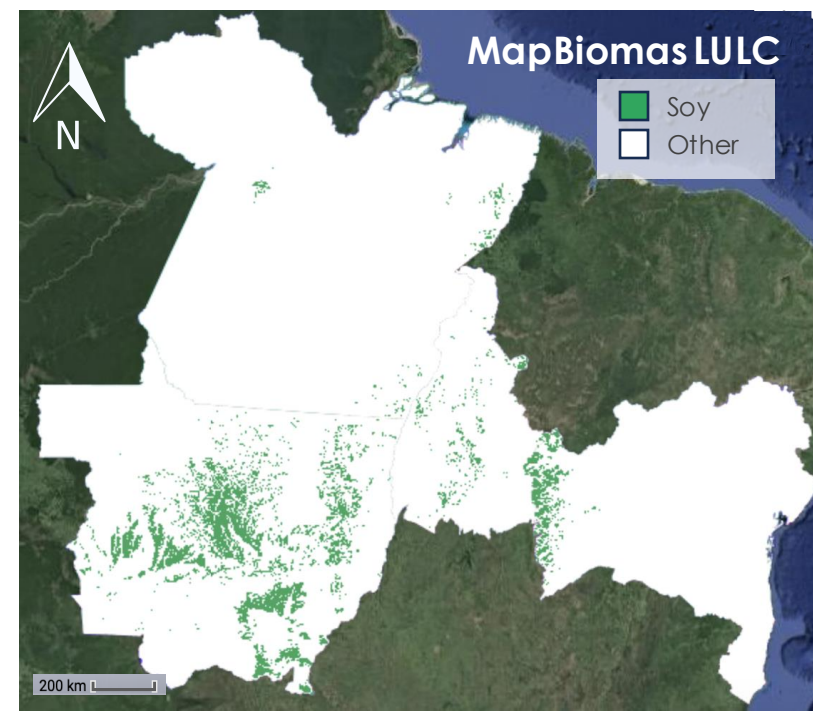
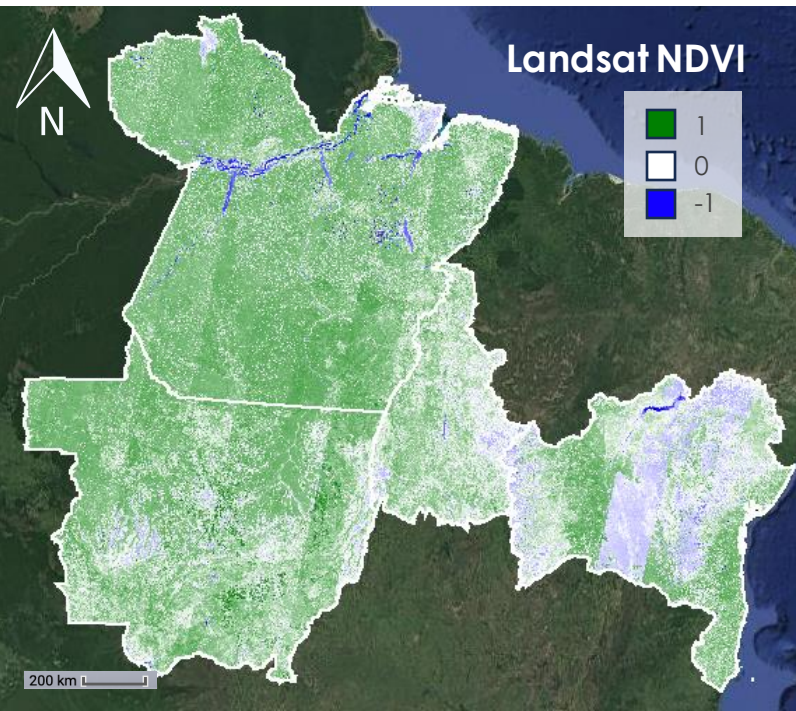
Datasets

State

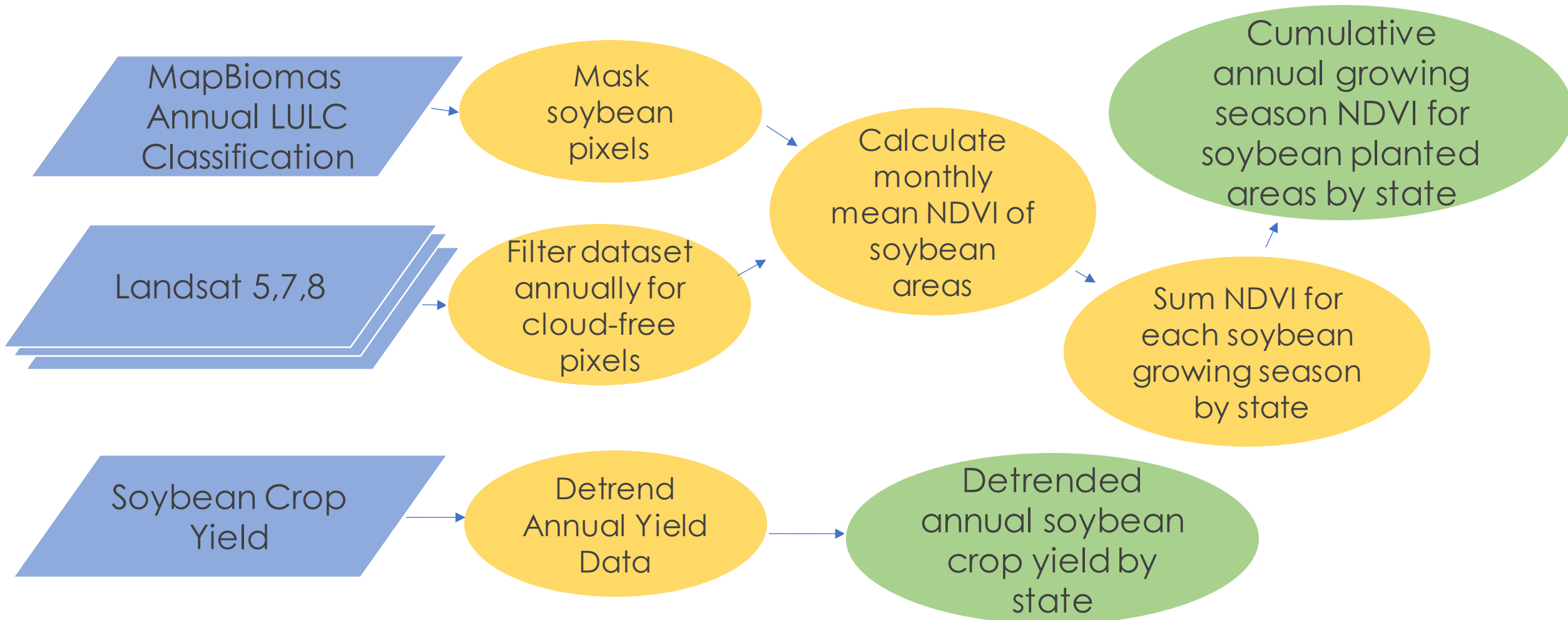
- Bahia
- Mato Grosso
- Pará
- Tocantins



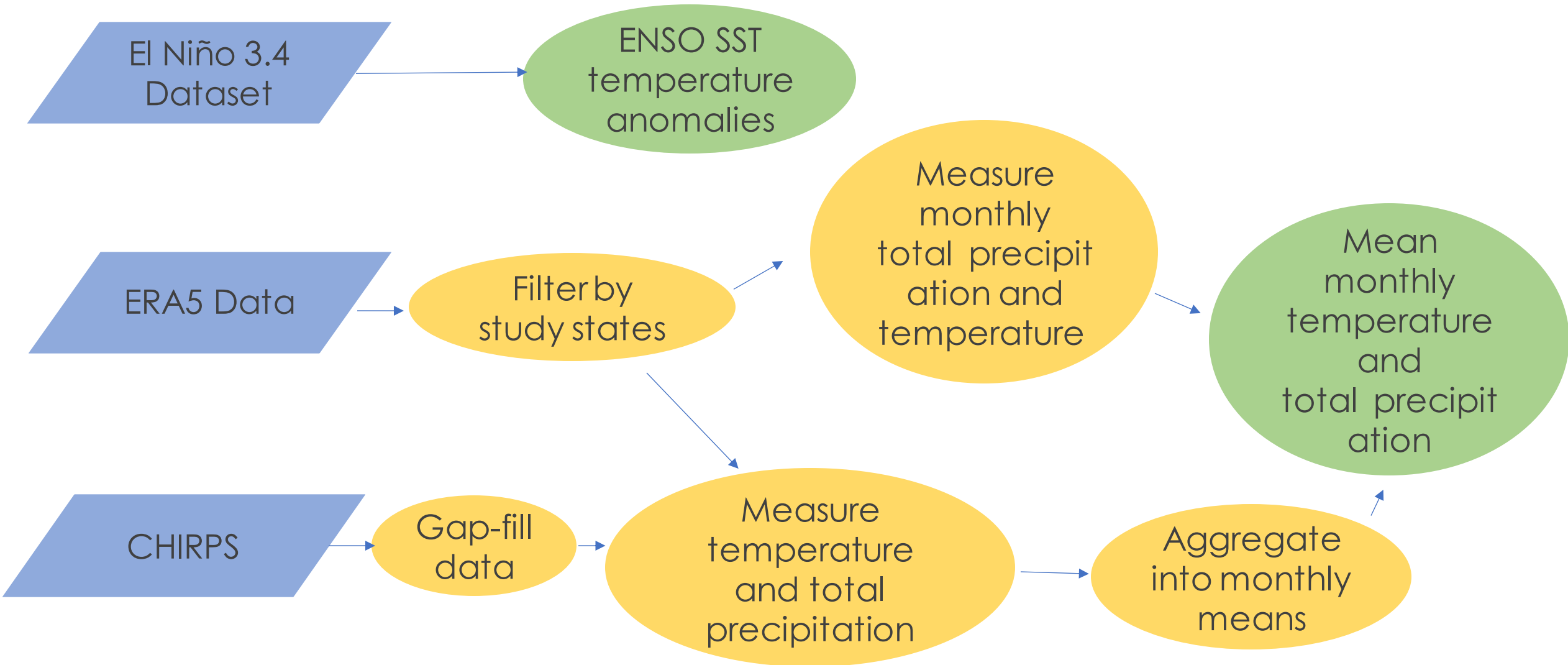
Base map credit: Google Satellite



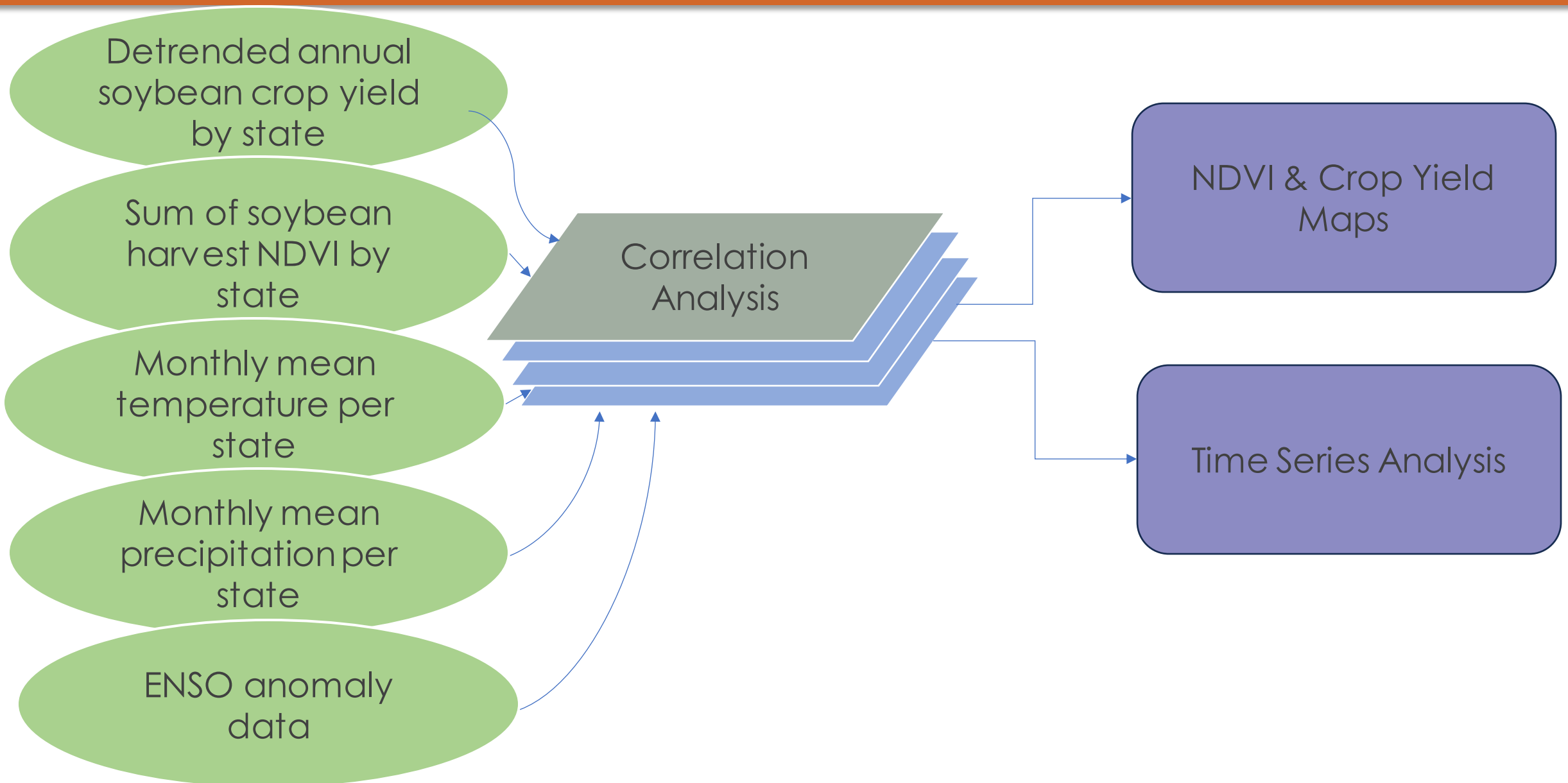
Methods: Calculating NDVI & Crop Yield



Methods: Temperature & Precipitation



Methods: Data Analysis

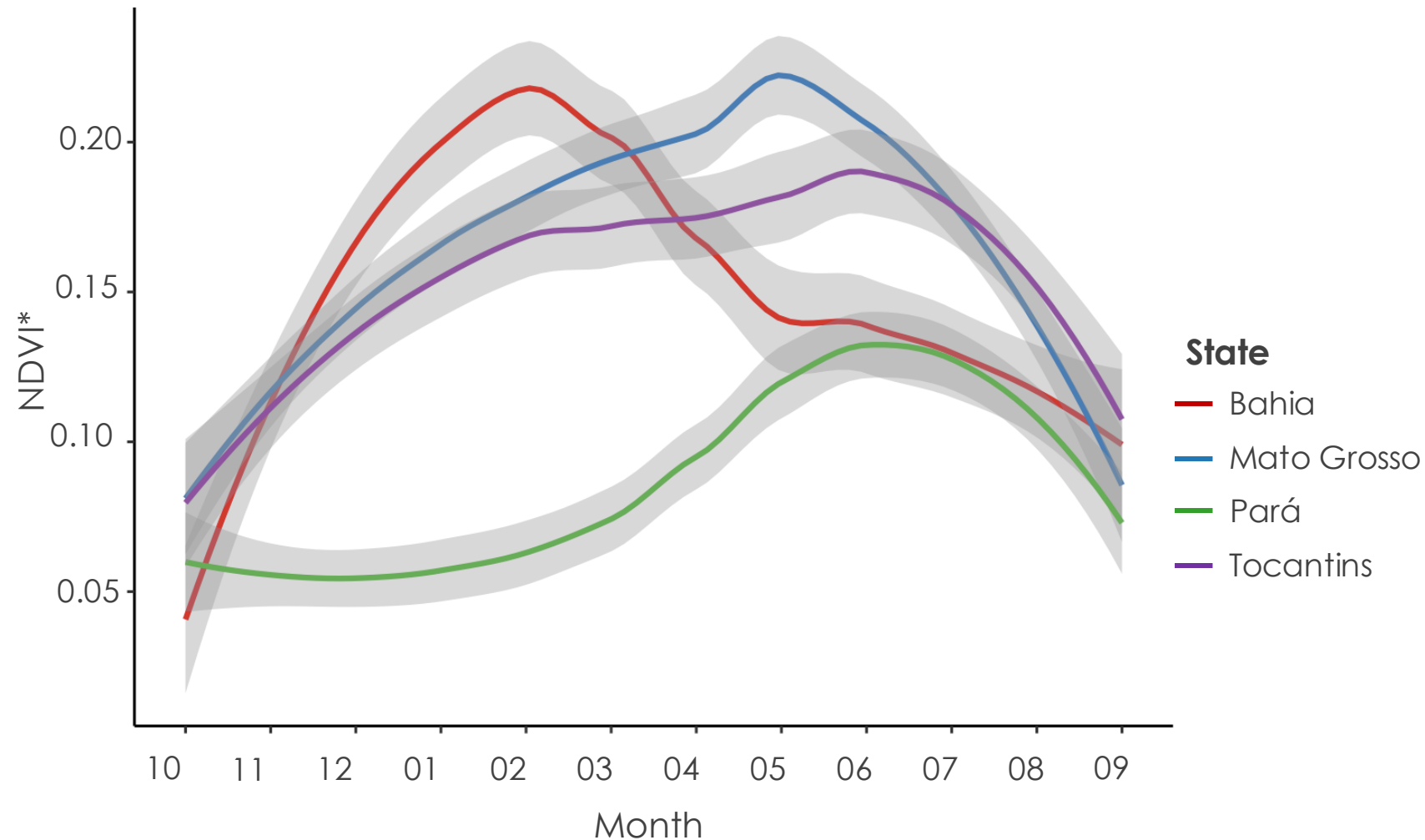


Results: NDVI Seasonality

State	Growing Season
Pará	Nov–July
Mato Grosso	Oct–June
Tocantins	Oct–June
Bahia	Oct–June

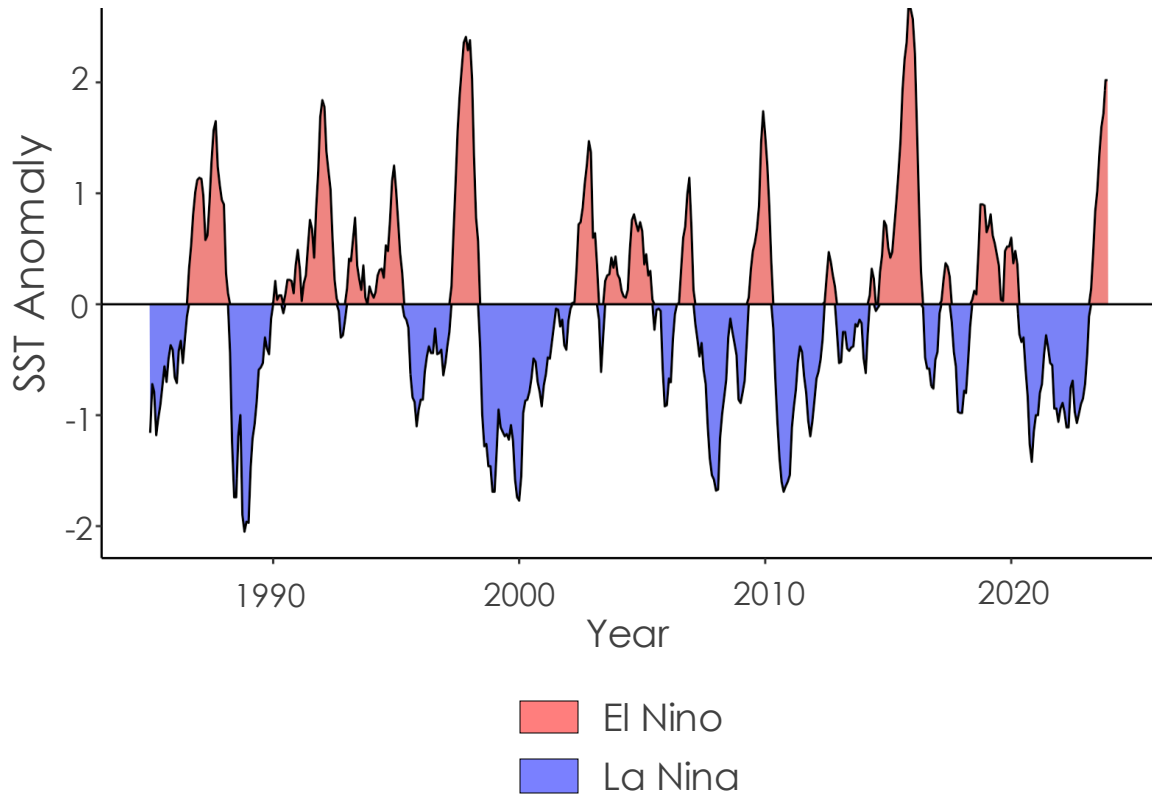
*Scale of -1 to 1, v values between 0 and 1 represent vegetation greenness, with higher v values corresponding with increased greenness

Average NDVI of Soybean Planted Areas per Month 1985–2021

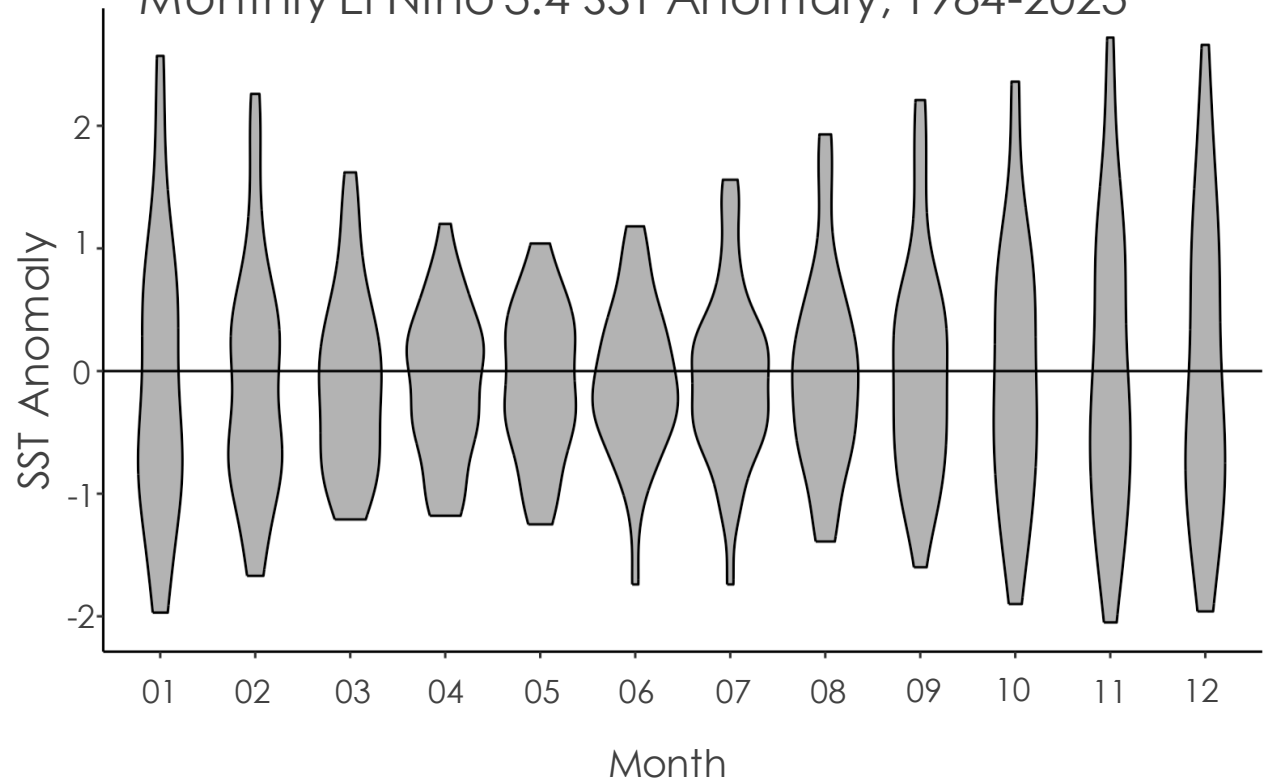


Results: ENSO Anomaly

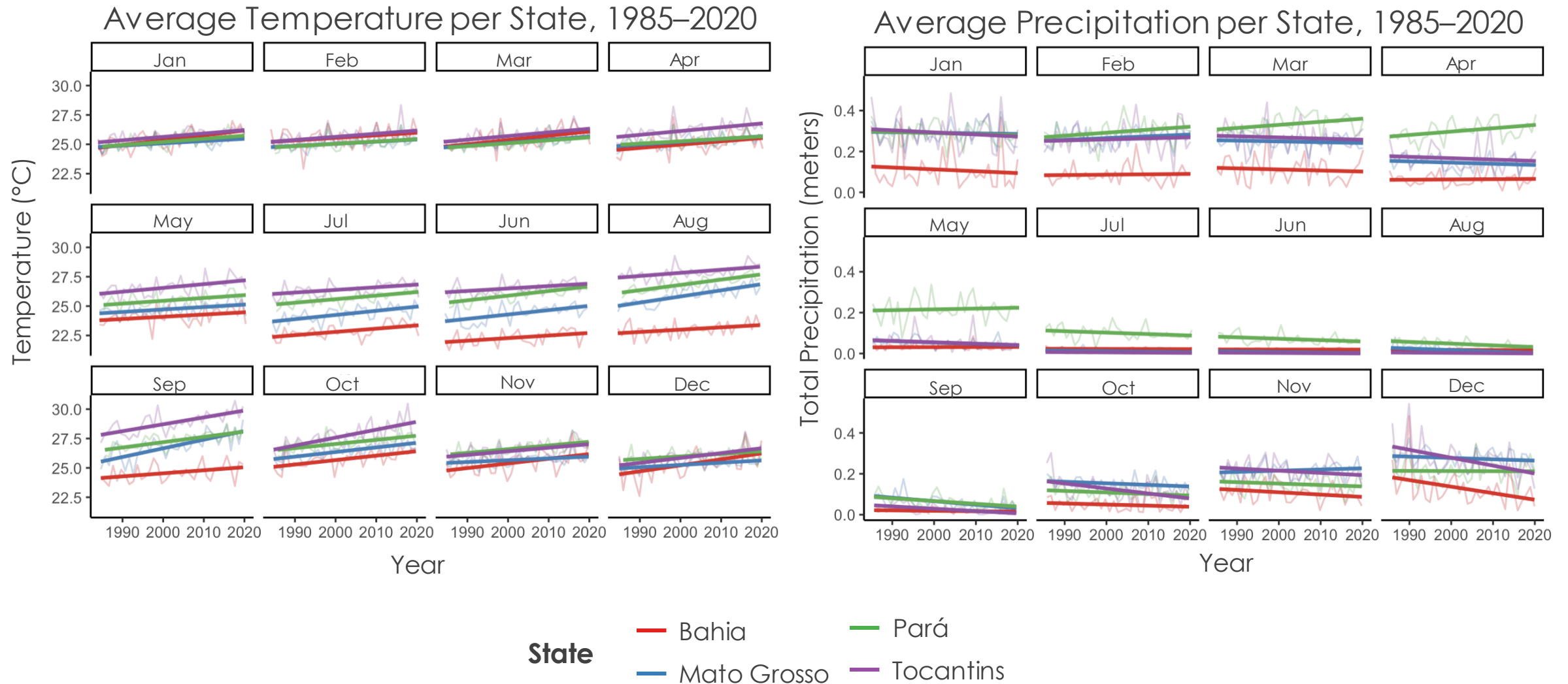
El Nino 3.4 Sea Surface Temperature (SST)
Anomaly, 1984-2023



Monthly El Nino 3.4 SST Anomaly, 1984-2023

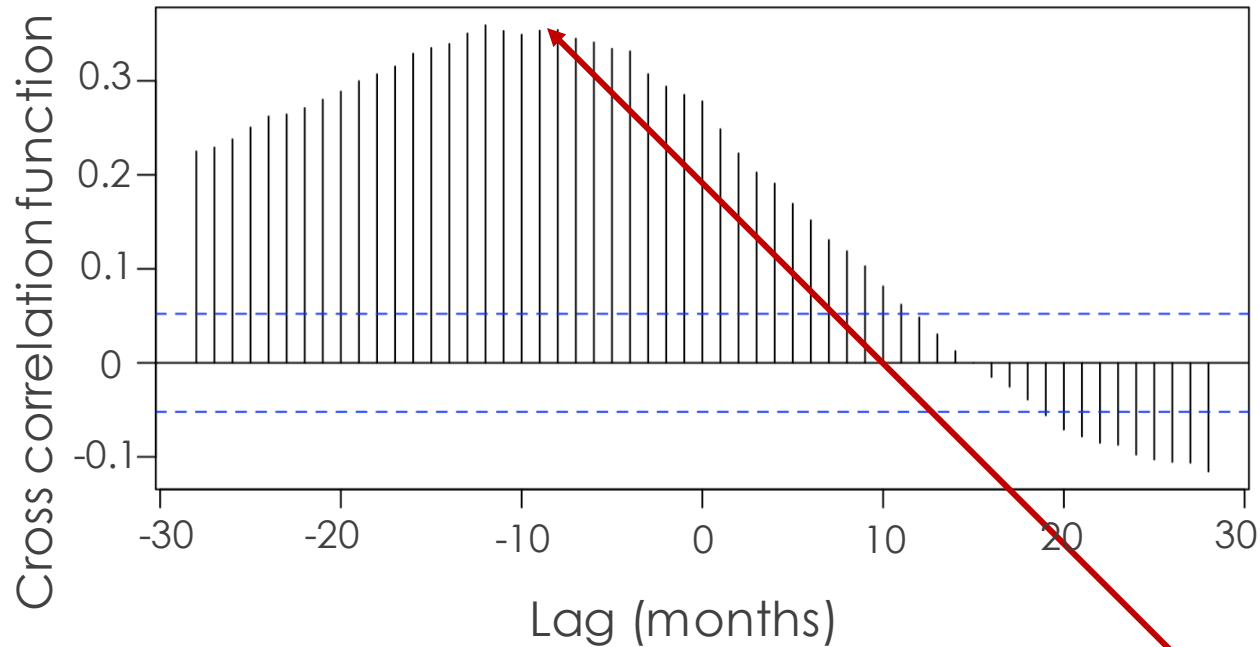


Results: Temperature and Precipitation Trends

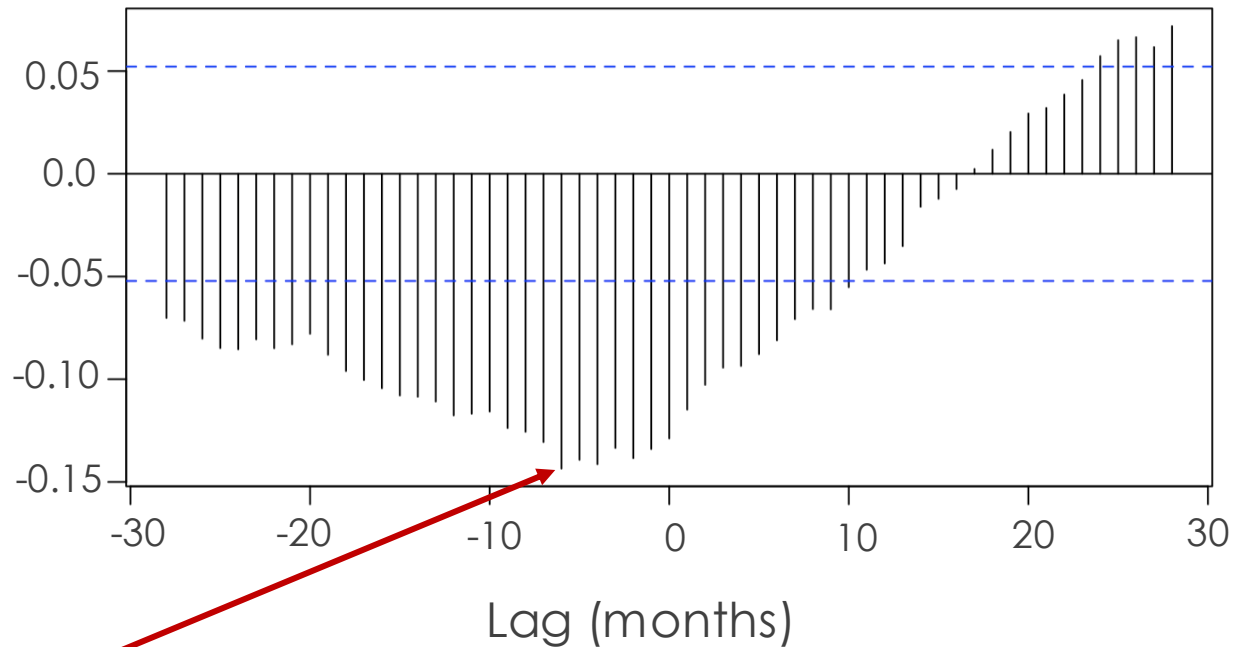


Results: ENSO and Local Climate Lag

ENSO and Temperature Anomaly

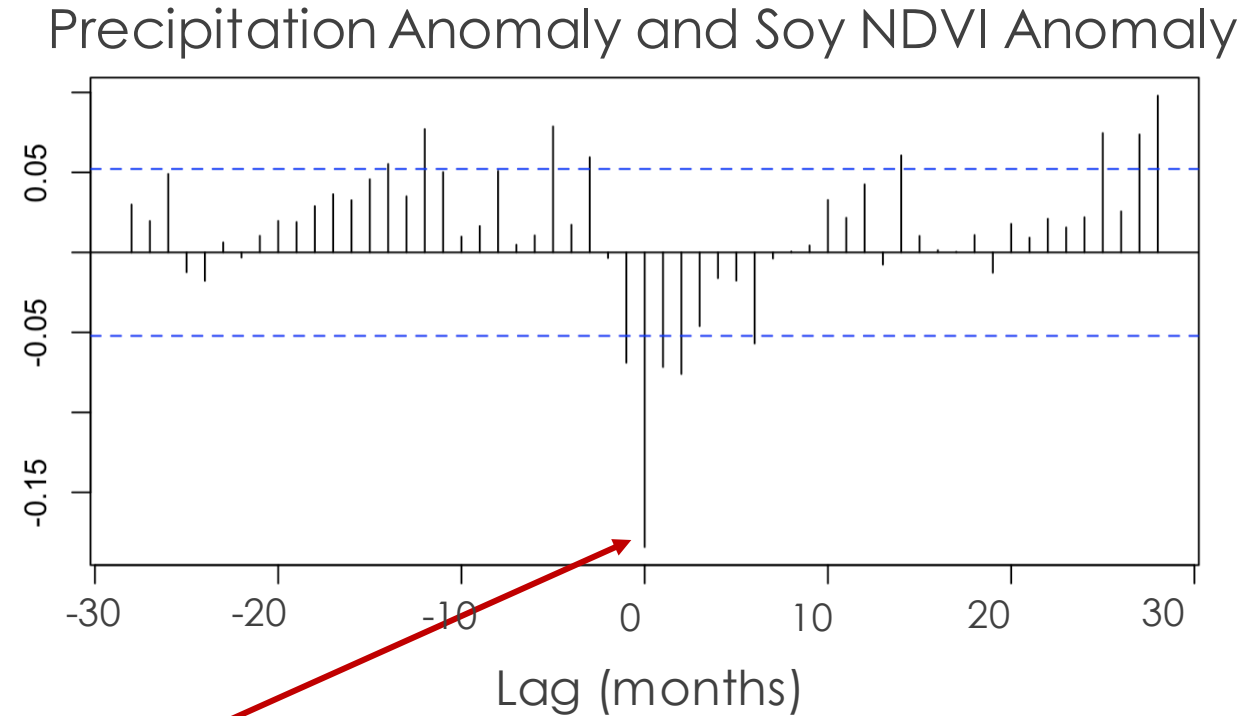
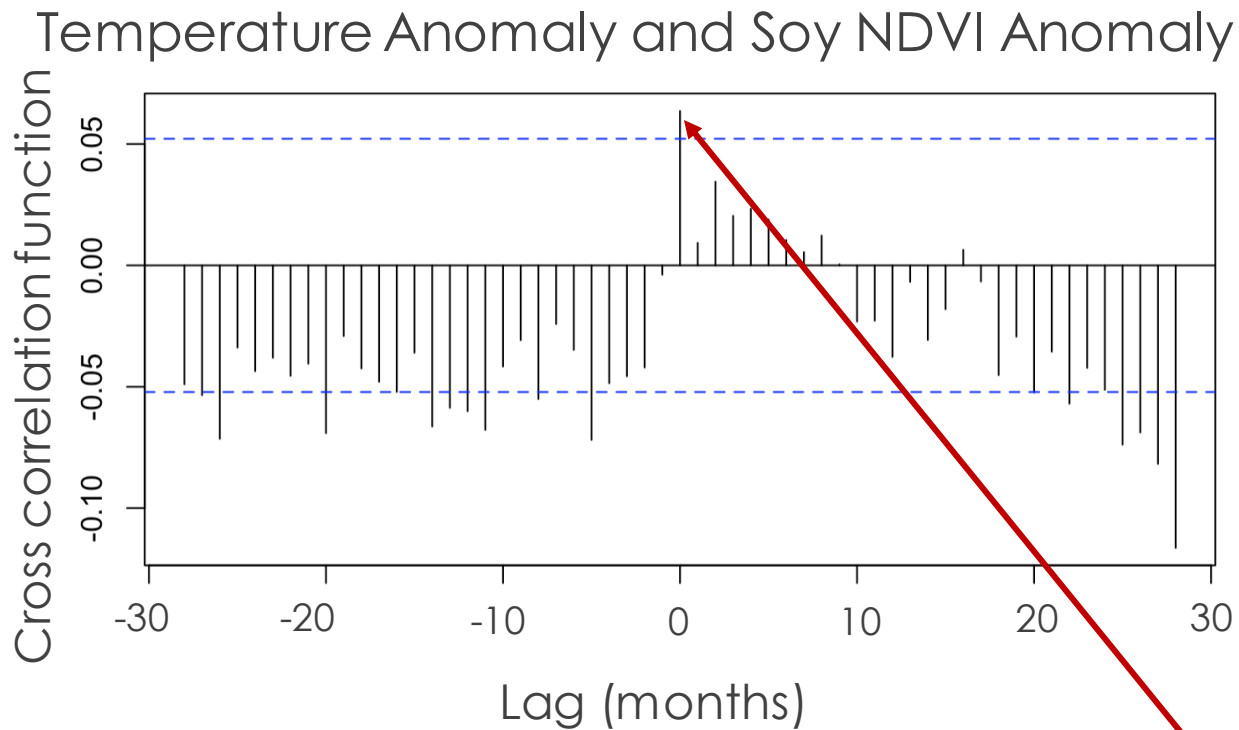


ENSO and Precipitation Anomaly



8-month lag between ENSO
anomalies and precipitation and temperature
anomalies in study region

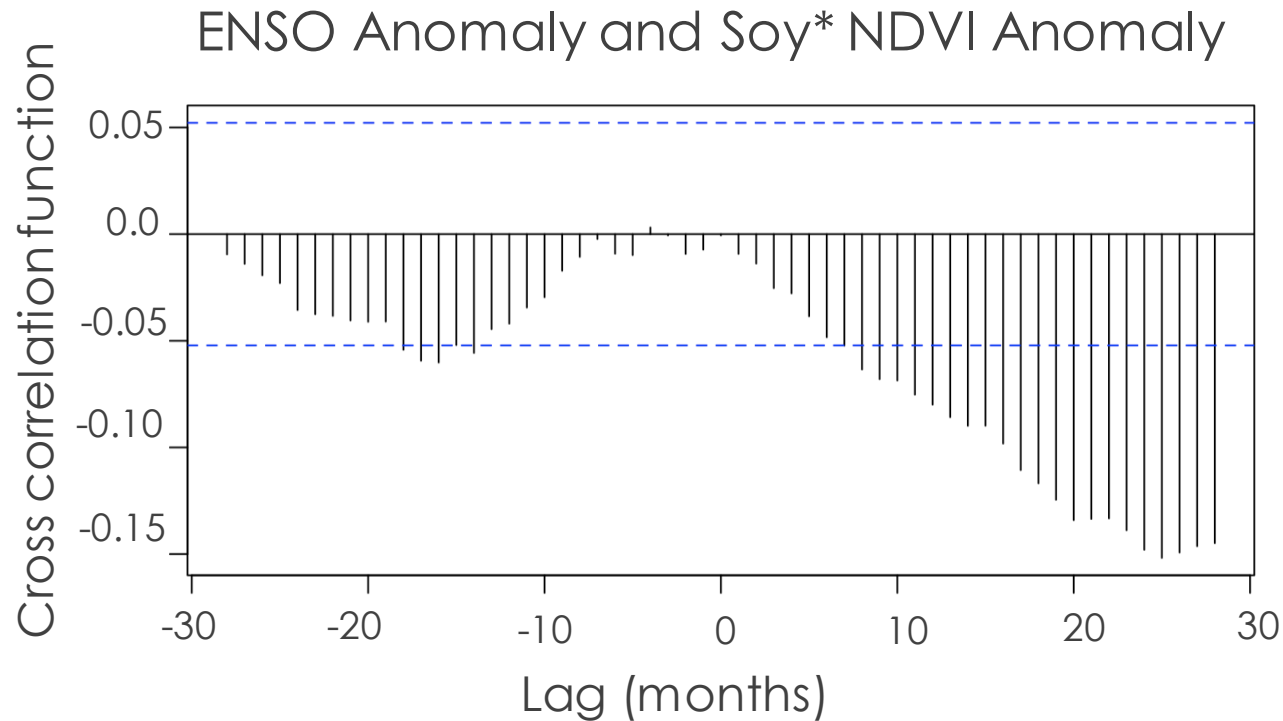
Results: Climate and Soy NDVI Lag



No lag (0-month) between precipitation and temperature anomalies and NDVI anomaly in study region

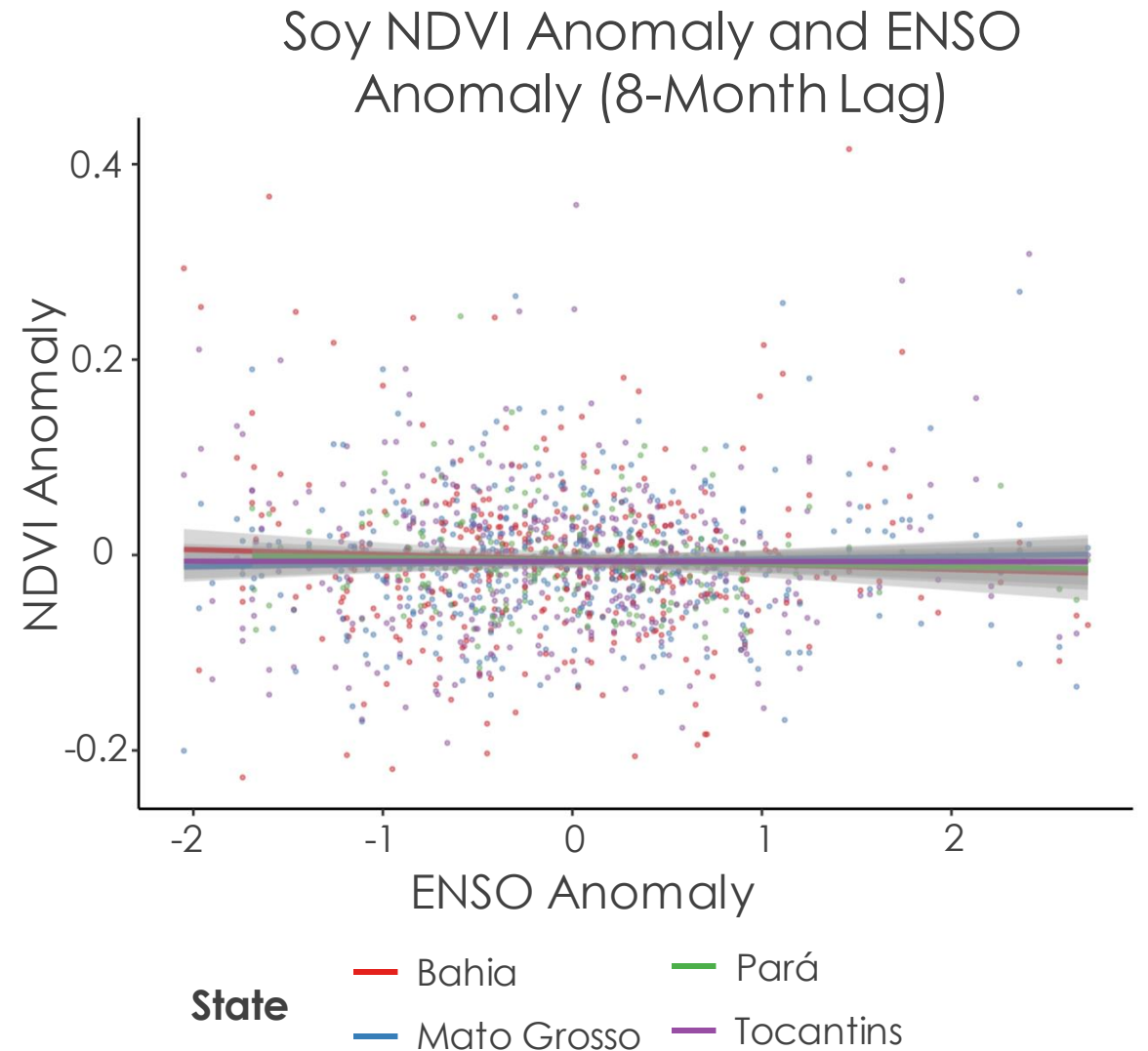
*Soy crop mask by MapBiomass

Results: ENSO and NDVI lag

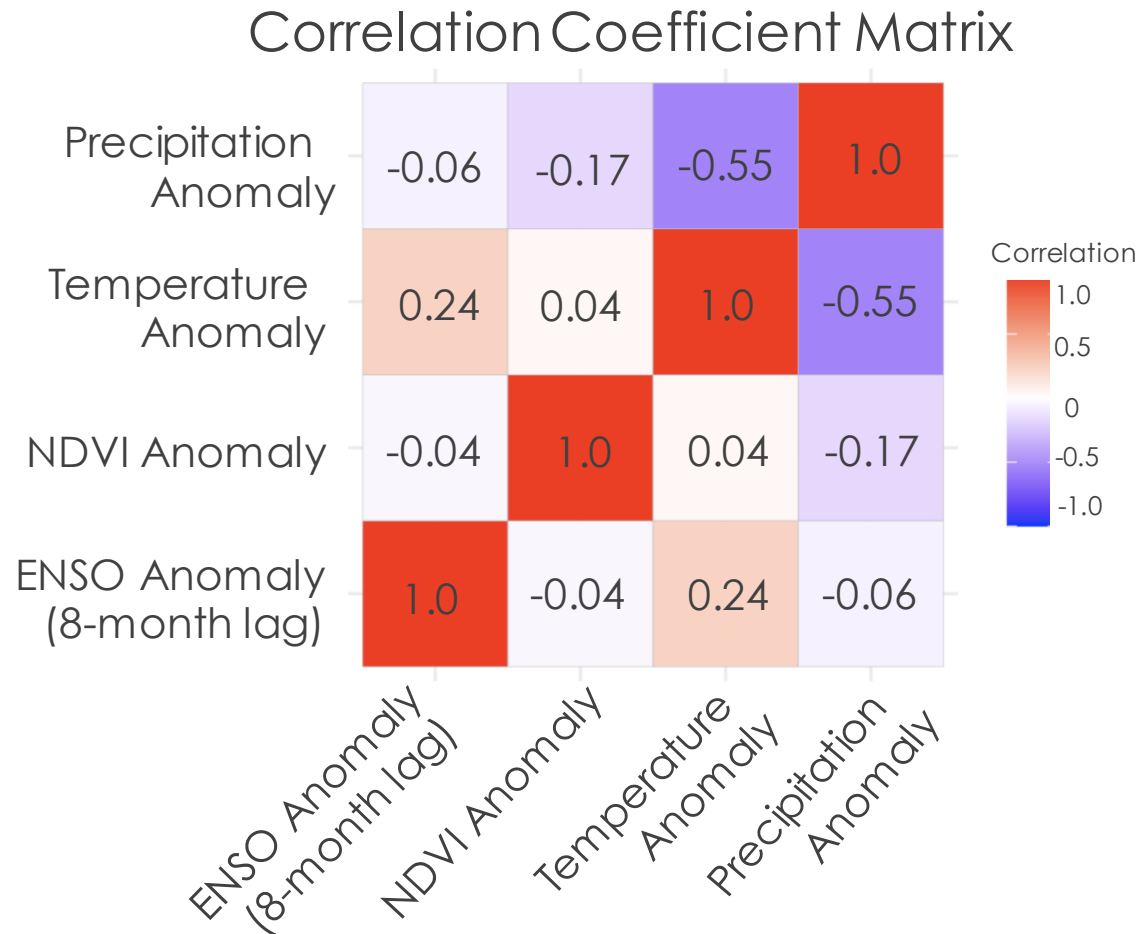


No clear lag or statistically significant relationship between ENSO Anomaly and NDVI

*Soy crop mask by MapBiomass

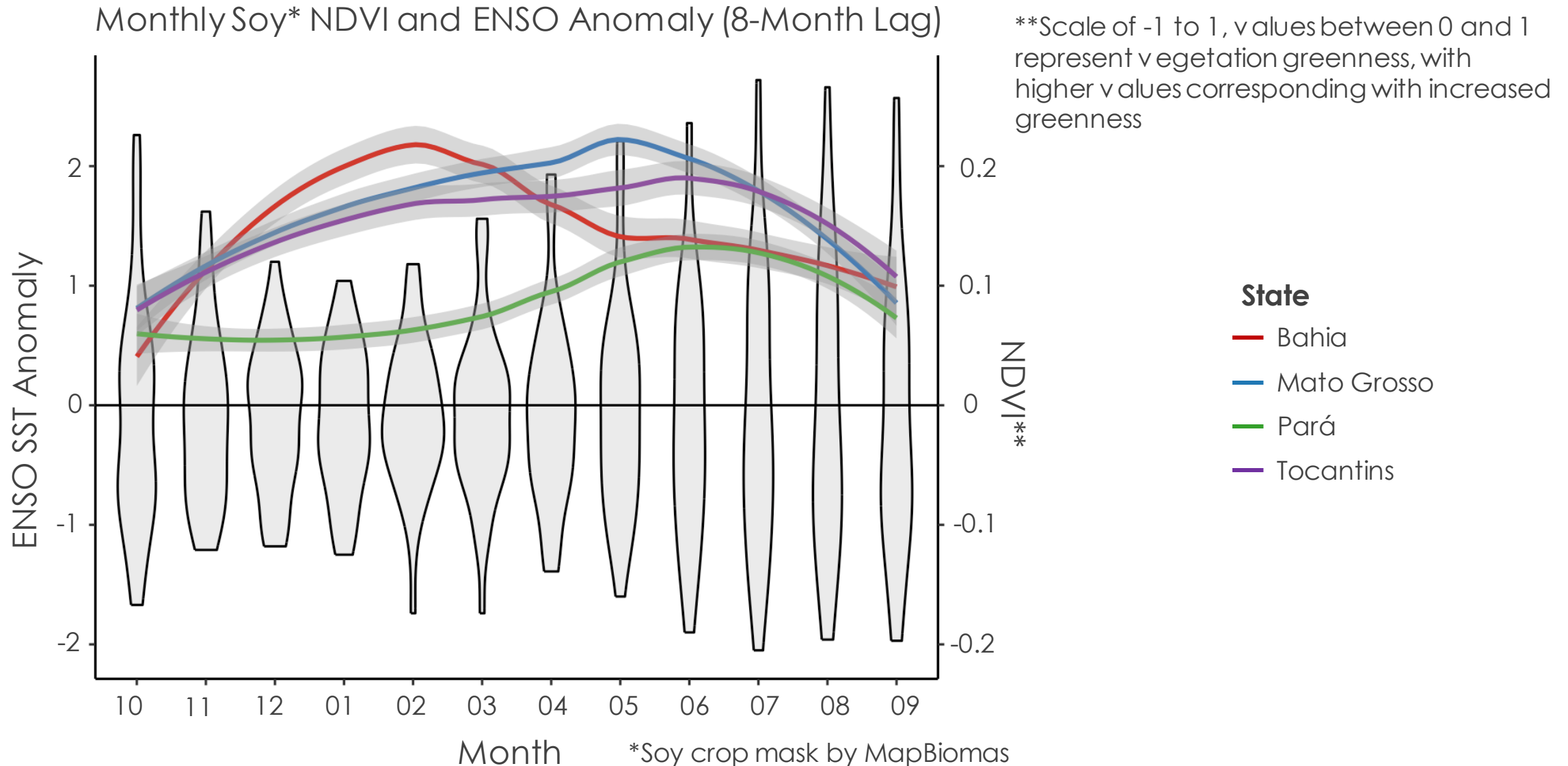


Results: Principal Component Analysis



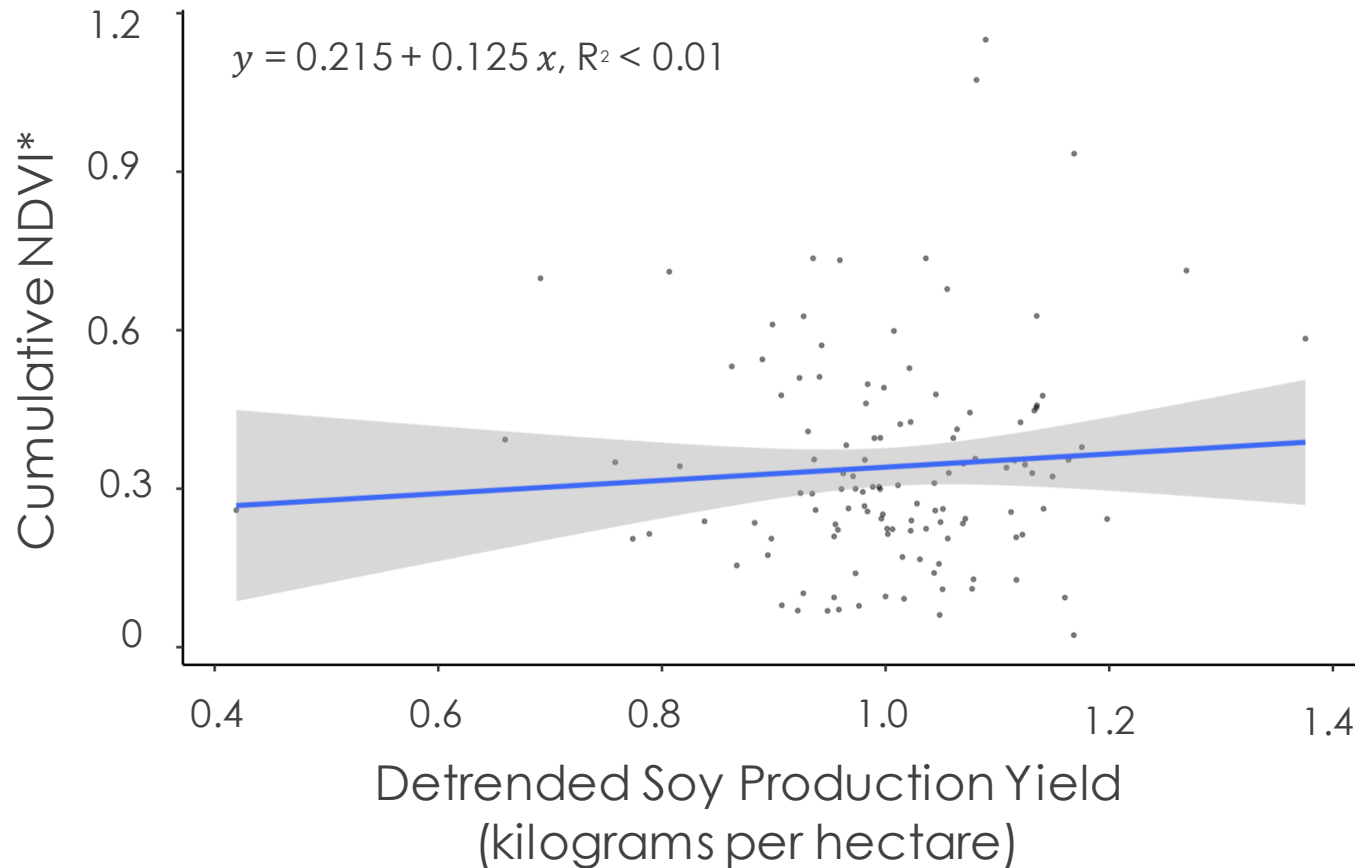
- Little to no correlation between NDVI and the 8-month lag ENSO
- This suggests that ENSO anomaly cannot be used to forecast NDVI anomaly

Results: Soy and Local ENSO Seasonality



Results: Production Yield

Detrended Yield by Cumulative Soy Growing Season NDVI



*Scale of -1 to 1, values between 0 and 1 represent vegetation greenness, with higher values corresponding with increased greenness

- No correlation between cumulative soy growing season NDVI and the detrended production yield statistics
- This suggests NDVI is not a feasible indicator for forecasting soy production yield

Errors and Uncertainties

Missing values:

- Some missing monthly NDVI values due to clouds, so we relied on gap - filling
- Due to missing temperature and precipitation years in ERA5, we supplemented with data from CHIRPS

Methodological limitations:

- MapBiomas LULC classification
- Changing crop seasonality

Feasibility



Challenges using ENSO conditions to forecast soybean crop productivity



Temperature and precipitation data is viable to examine relationships between NDVI and seasonality



Correlation analyses can assist with producing time series and crop yield maps to examine factors that affect production yield

Conclusions



Correlation and time-series analysis can identify trends and determine which factors are affecting soy NDVI



Temperature and precipitation data can provide insights into the relationships between NDVI and seasonality through a temporal analysis



Monthly soy NDVI anomaly had little to no correlation with monthly ENSO anomaly. ENSO have complex impact on soy in the study area and period warranting further study



Monthly ENSO anomaly cannot be used to forecast monthly soy NDVI anomaly nor soy yearly production yield

Future Recommendations



Conduct analysis with a multivariate ENSO index



Examine multi-year events



Assess methods on 2nd season corn production



Focus on a smaller geographic region – state, sub-state level

Acknowledgments

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