

**Powering the Continued Development Journey**



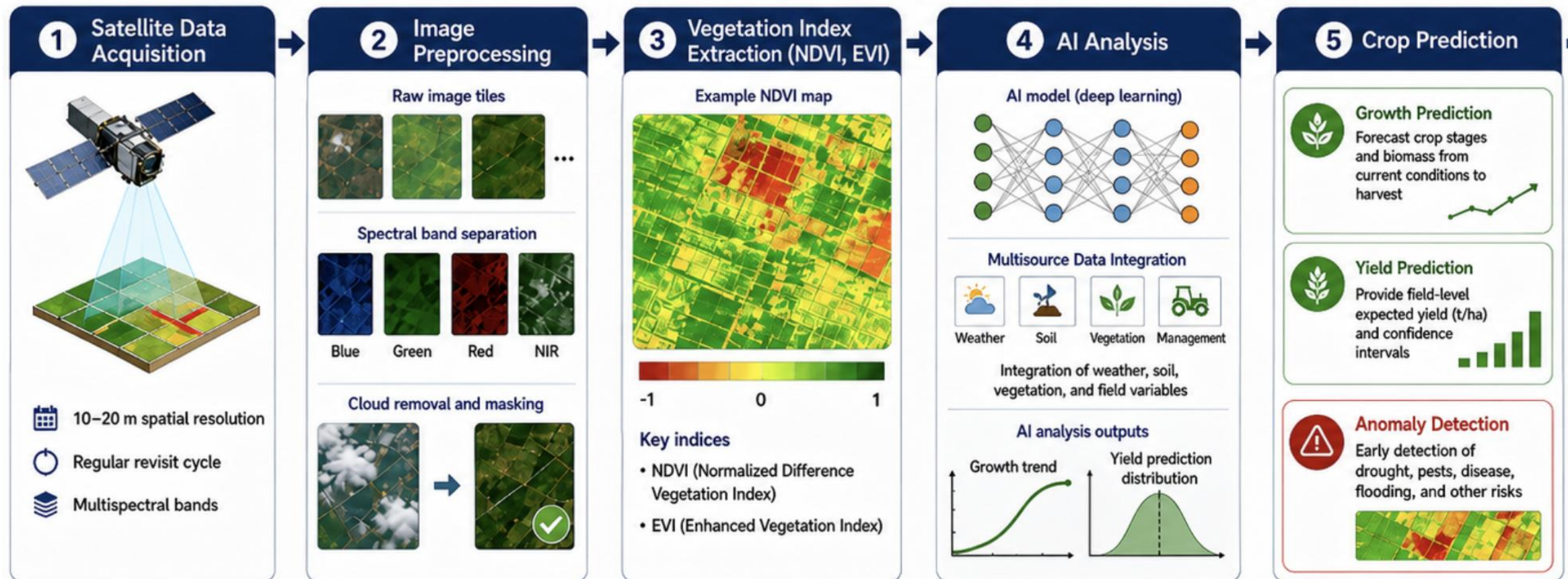
**Building the Methodological Foundation:  
A Six-Step Framework for Integrating Remote Sensing, AI, and  
Crop Models for Crop Area and Production Monitoring  
In Honduras**

Kwang Soo Kim

Professor, Seoul National University

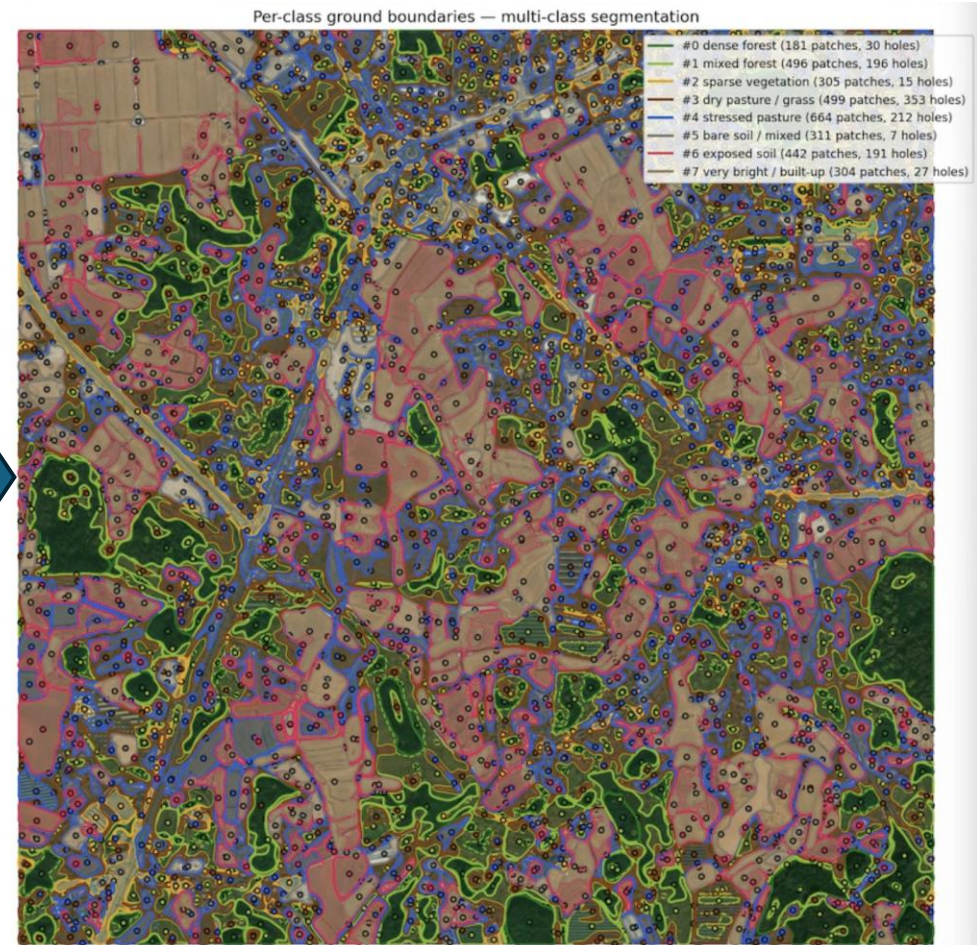
Director, National Center for AgroMeteorology

# Remote Sensing Data and Artificial Intelligence



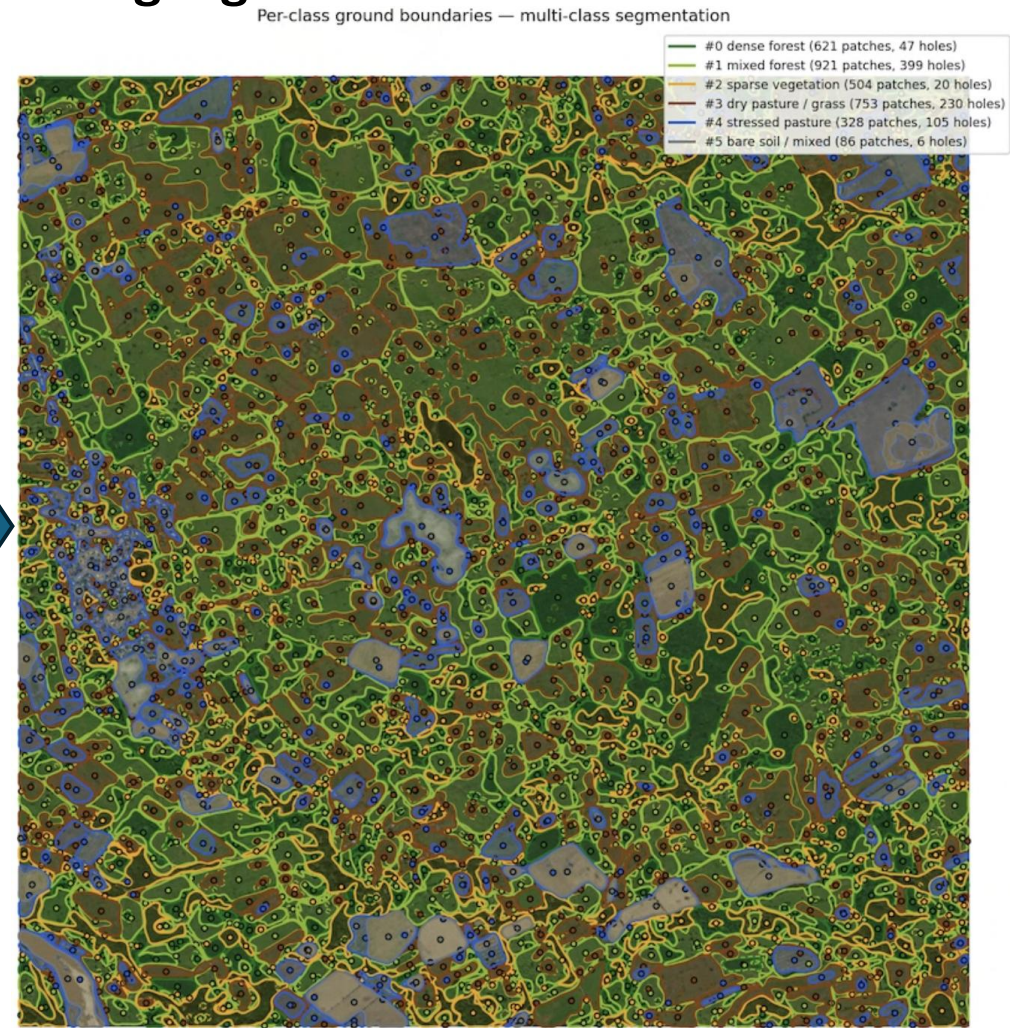
# Remote Sensing Based Crop Area

## Landcover Classification using AI and Ray-Tracing Algorithm - Gimje, Korea



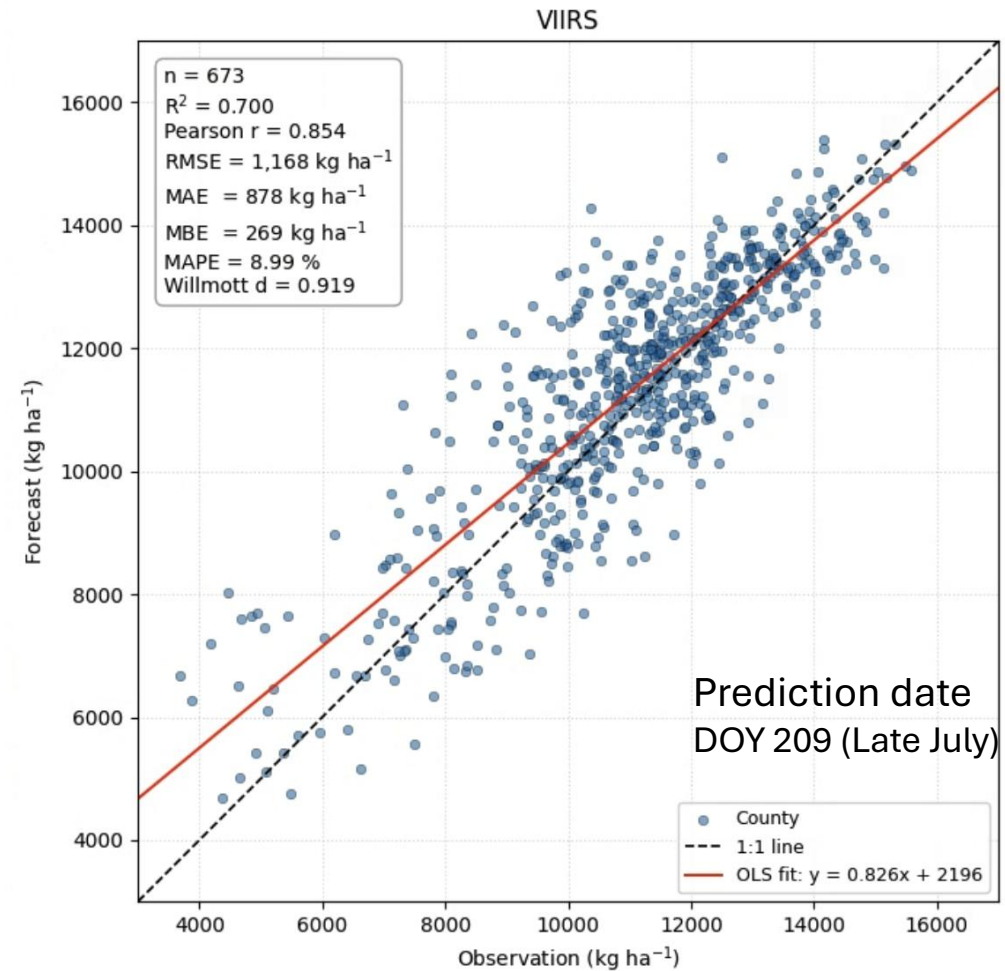
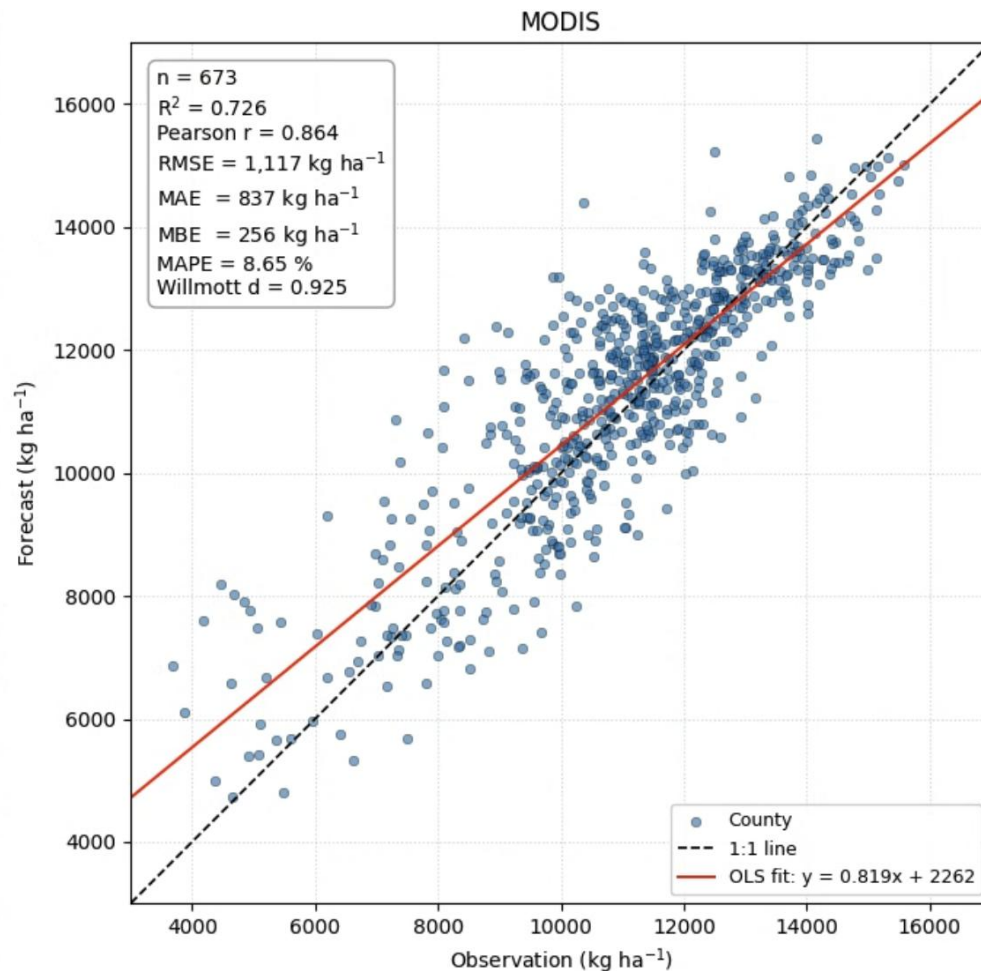
# Remote Sensing Based Crop Area

## Landcover Classification using AI and Ray-Tracing Algorithm - Catacamas Valley, Honduras



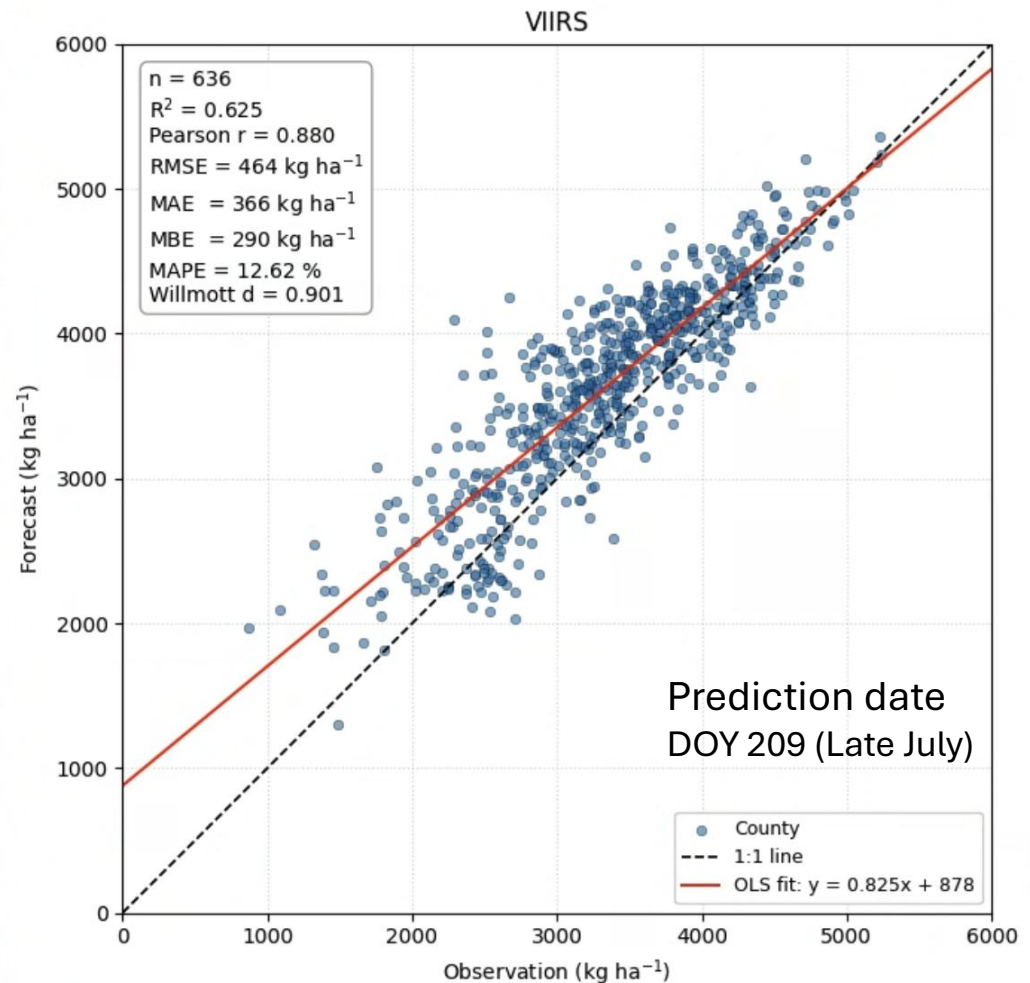
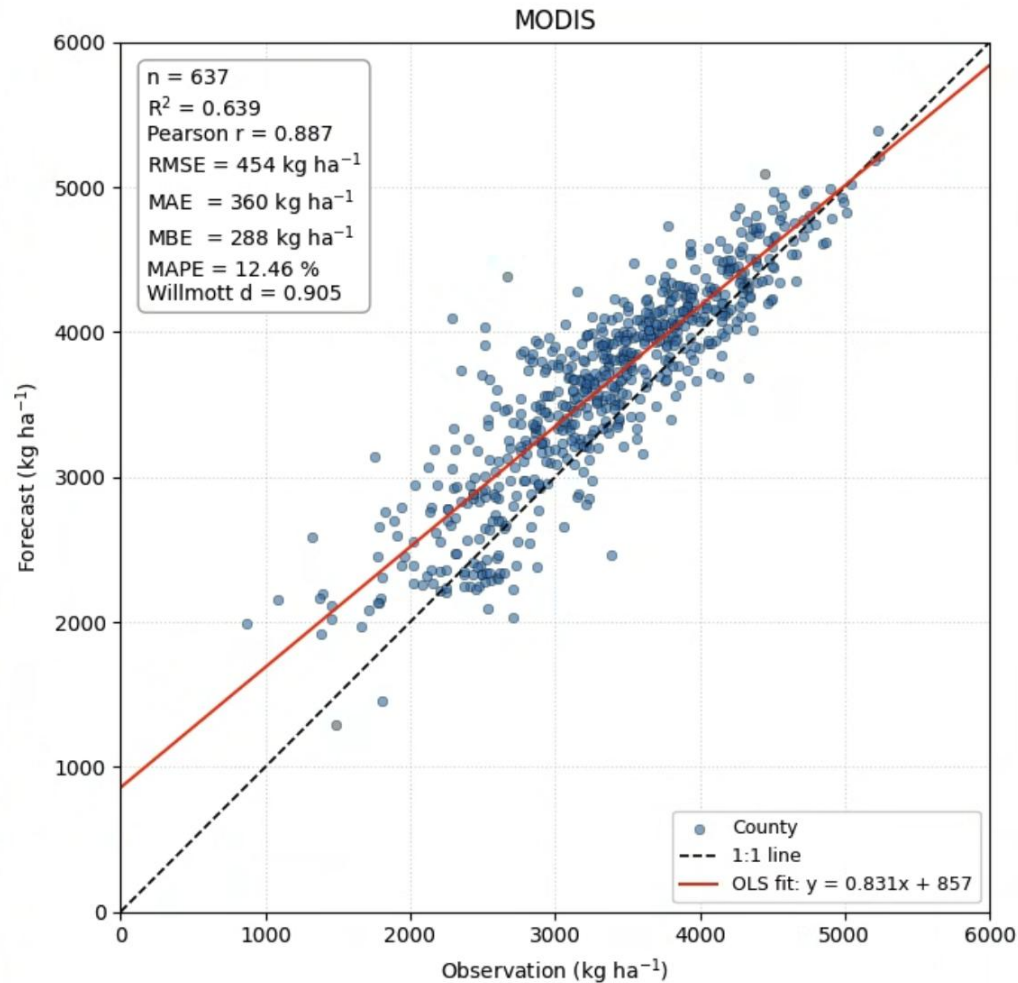
# Remote Sensing Based Yield Forecast in the US

## Maize Yield Forecast in 2024 using MODIS and VIIRS Data



# Remote Sensing Based Yield Forecast in the US

## Soybean Yield in 2024 using VIIRS and MODIS



# What is Missing?

Limited field-level data on planted area, growth status, yield, and management

Sparse historical yield records for corn and beans



- AI models trained on scarce data risk overfitting to specific years or regions
- Predictions become unstable beyond observed conditions

In data-limited environments,

- *Crop models can improve prediction stability beyond observed conditions by representing the biological processes that govern crop growth and yield formation*
- *Hybrid approaches can help reduce the risk of overfitting by integrating process-based crop responses with remote sensing data and AI-driven prediction methods*

# What Crop Models Can Offer

- **Estimate yield with sparse data**

Use weather and soil data to simulate growth and yield, even where field observations are limited

- **Interpret satellite signals**

Distinguish whether vegetation changes reflect biological stress or differences in cropping calendar

- **Augment AI training data**

Provide simulated yield outputs and process-based covariates when ground-truth labels are scarce.

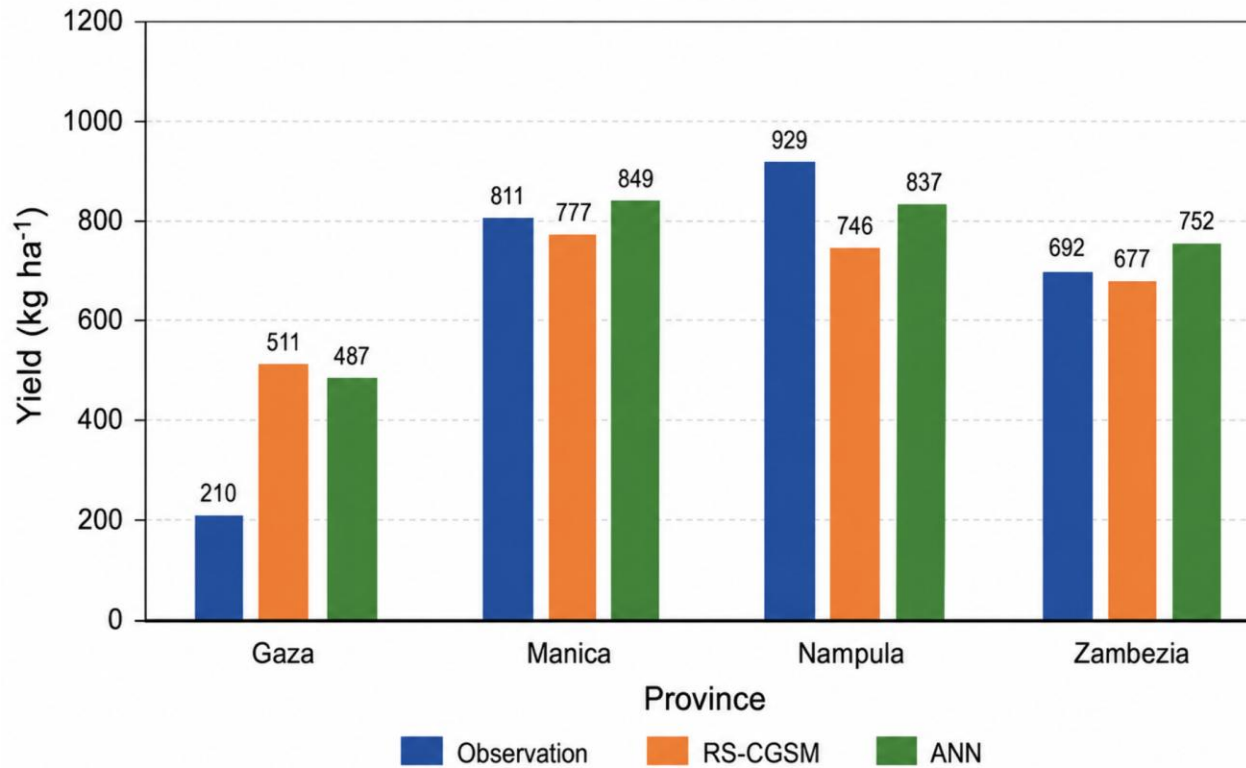
- **Project unobserved conditions**

Forecast outcomes under drought, heat, and delayed planting scenarios not seen in historical records.

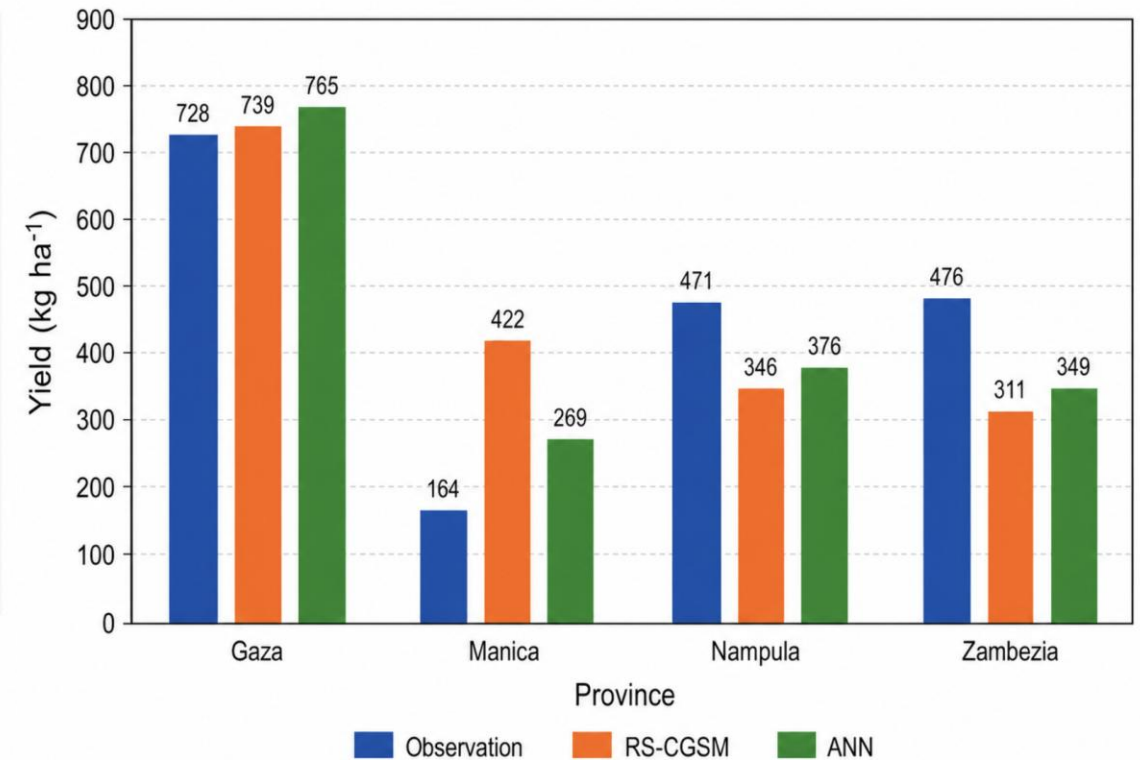
# Yield Forecast using RS and Crop Model in Mozambique

## Crop Yield Forecast using Remote Sensing Data and Process-Based Crop Model - Prediction Date: Jan 15 2020

Maize Yield in 2019/20

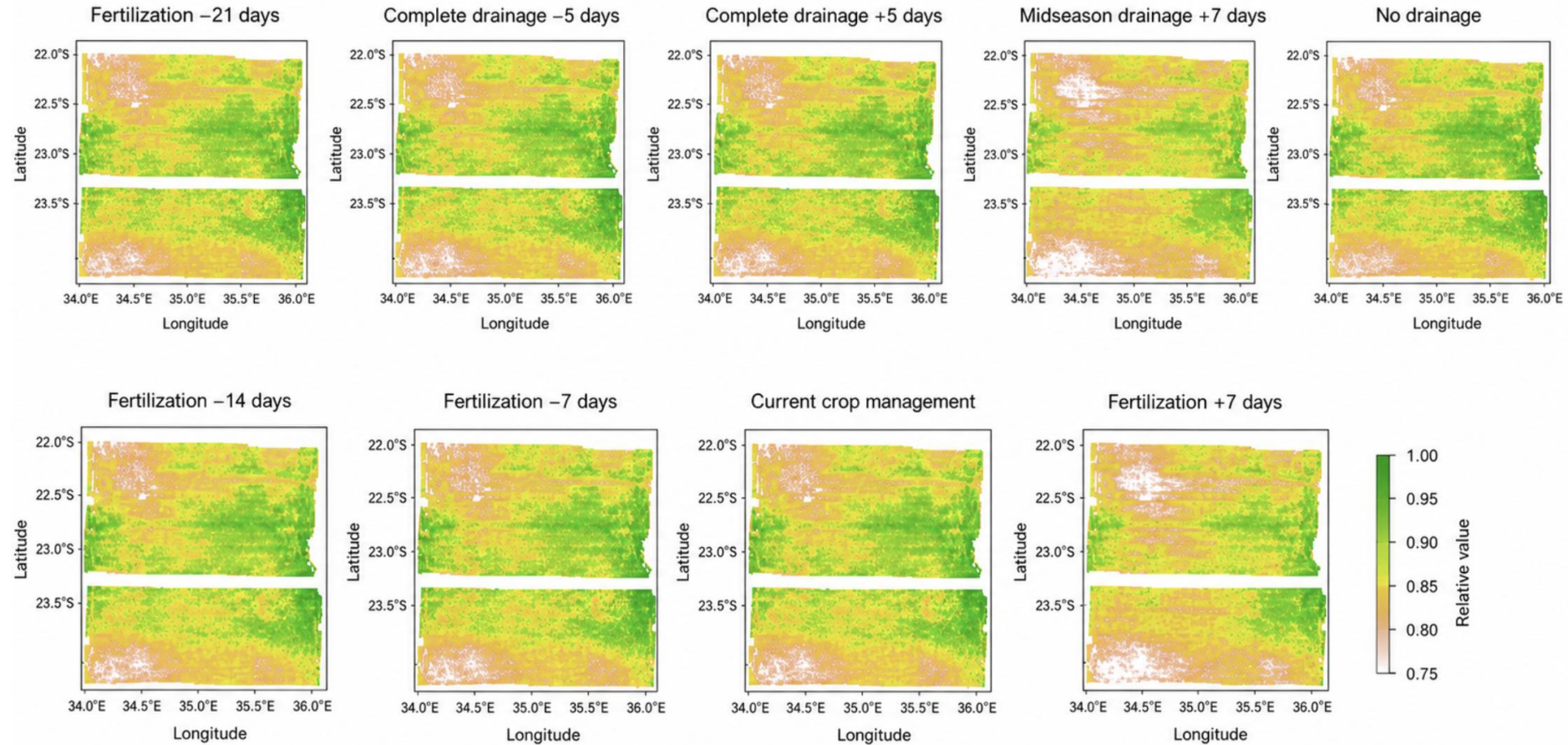


Rice Yield in 2019/20



# Assimilation of Drone Data into Crop Growth Simulations

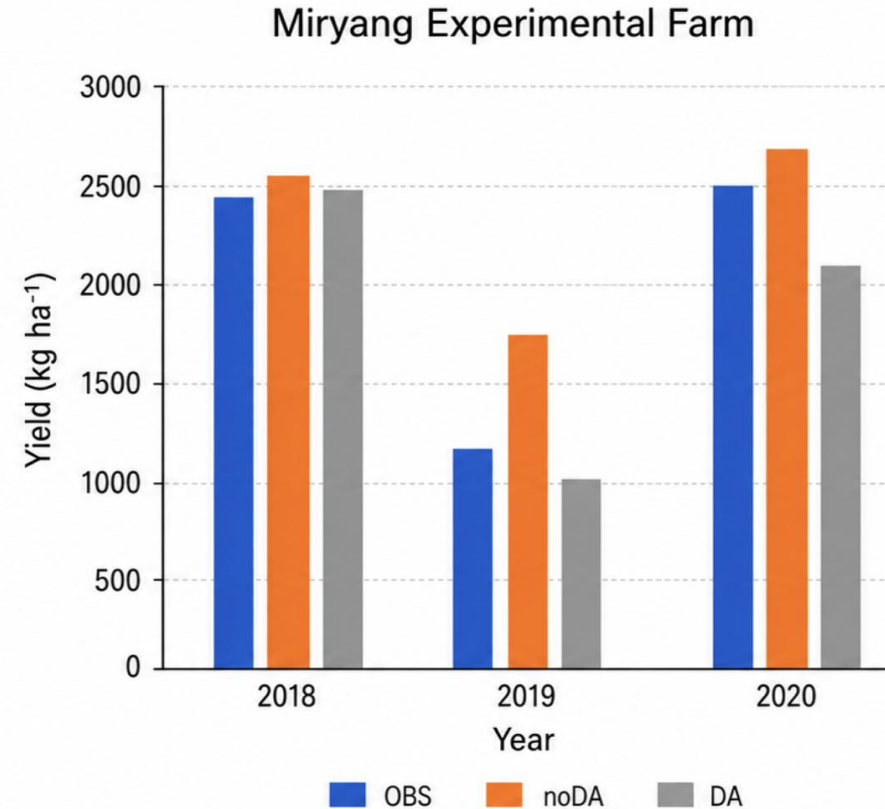
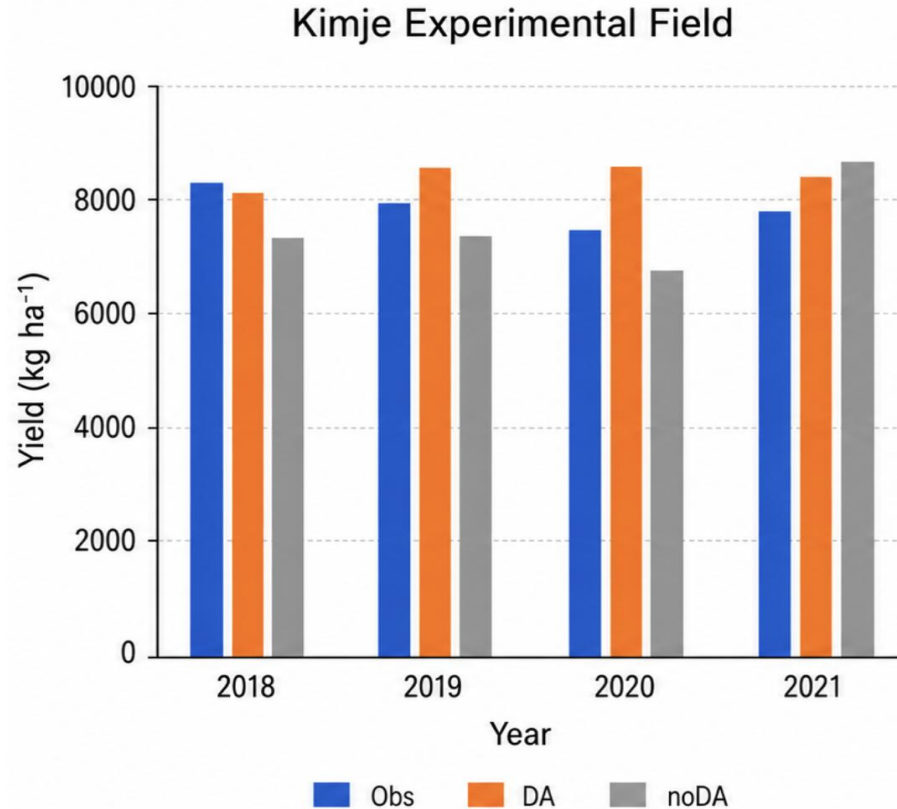
## Effects of Fertilization and Drainage Timing on Spatial Crop Conditions



# Assimilation of Drone Data into Crop Growth Simulations

## Incorporating drone imagery improved crop yield prediction accuracy

- The prediction error decreased from 11% to 8% for rice on average
- The prediction error decreased from 21% to 9% for soybean on average



# Integrated approaches are proposed

Component	Primary Role
Remote Sensing	Observe planted area, growth status, and vegetation indices over wide regions
AI	Learn spatial-temporal patterns and scale predictions across the country
Crop Model	Compensate for sparse observations; explain growth and yield formation
Drones	Sample at strategic points to anchor representative national estimates

# Application to Honduras (Olancho & El Paraíso)

1

## Data Acquisition

- Satellite imagery
- Drone imagery
- Weather data
- Field data

2

## Crop Area Mapping

- NDVI/EVI
- AI classification of crop types

3

## Crop Growth Monitoring

- Track seasonal vegetation
- Detect anomalies

4

## Crop Model Assimilation

- Simulate growth
- Assimilate LAI & biomass

5

## AI Yield Estimation

- ML/DL models on satellite + weather time-series
- Bias Correction

6

## Production Monitoring

- Area × Yield
- Regional production
- Decisions

# Conclusion

## ❖ From Observation to Operation

- Remote sensing, AI, and crop models transform vegetation signals into actionable production intelligence

## ❖ Integrated Monitoring Pipeline

- A consistent workflow provides cultivated area, in-season crop condition, and pre-harvest yield estimates

## ❖ Spatiotemporal Crop Monitoring

- Spatially and temporally explicit information supports food security assessment, market planning, and evidence-based policy decisions

# Thank You

- Acknowledgments

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