

# CLOSING THE WATER BUDGET GAP: LEVERAGING SATELLITE-BASED INSIGHTS FOR ACCURATE IRRIGATION

## WATER-USE ESTIMATION ACROSS WESTERN U.S

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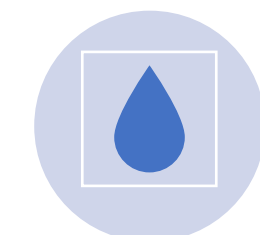
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### INTRODUCTION



One-third of cultivated lands across the United States rely on freshwater to produce approximately 40% of the nation's food



Groundwater reserves providing freshwater for irrigation across western US regions are declining at a rate faster than recharge. Projected to decline by 70% by the year 2050.



Accurately tracking water used to grow crops is useful for managing and sustaining water resources



Traditional approaches (e.g. National surveys, Metering devices, etc.) for monitoring seasonal water-use are cost and labor-intensive



Remotely sensed satellite data combined with machine learning approaches presents a unique opportunity to continuously estimate seasonal water use from irrigated areas.

### STUDY AREA

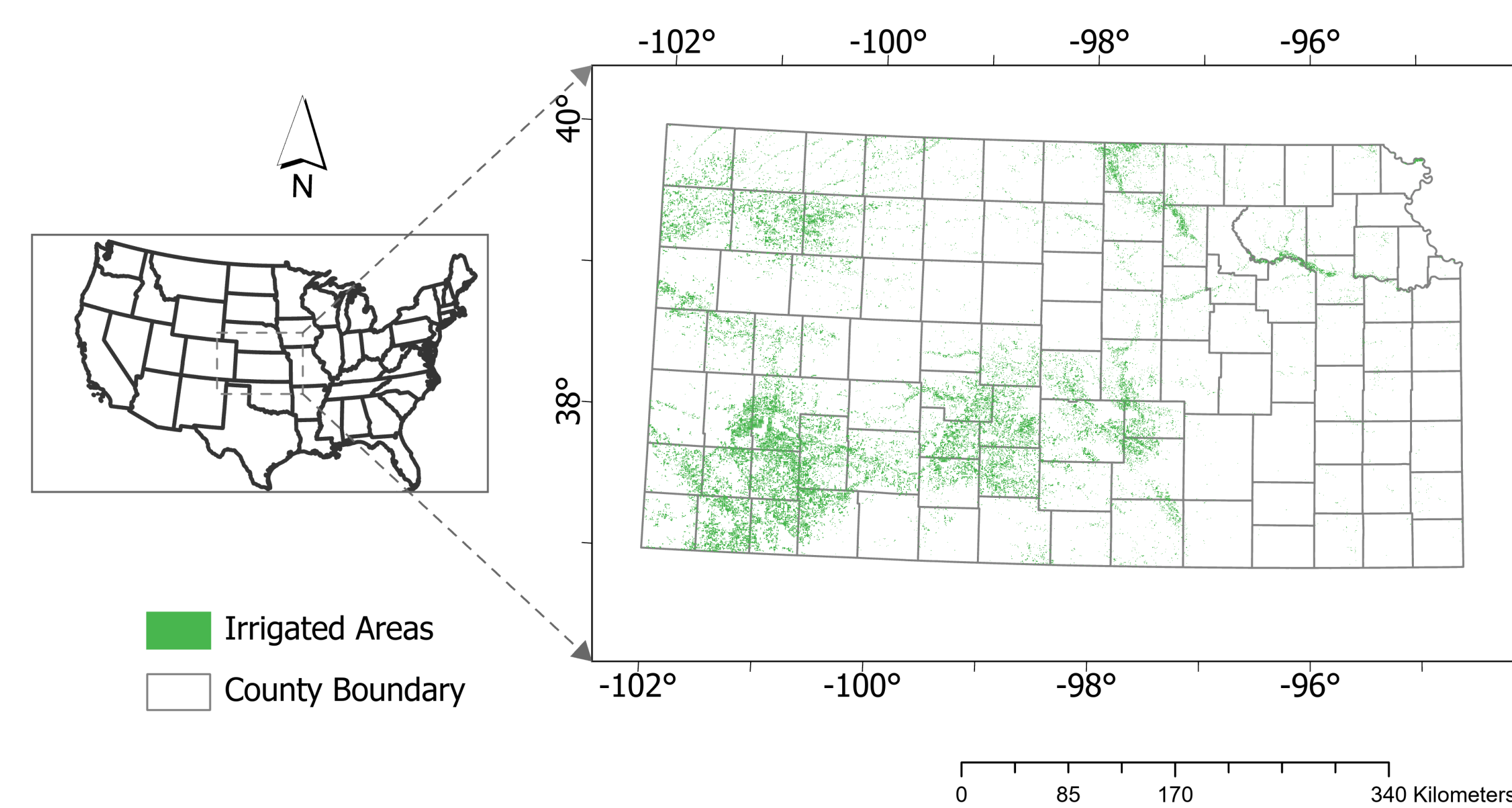


Figure 1: Map of Kansas showing irrigated areas extent across counties over 2012 – 2017 period.

### DATA AND METHODS

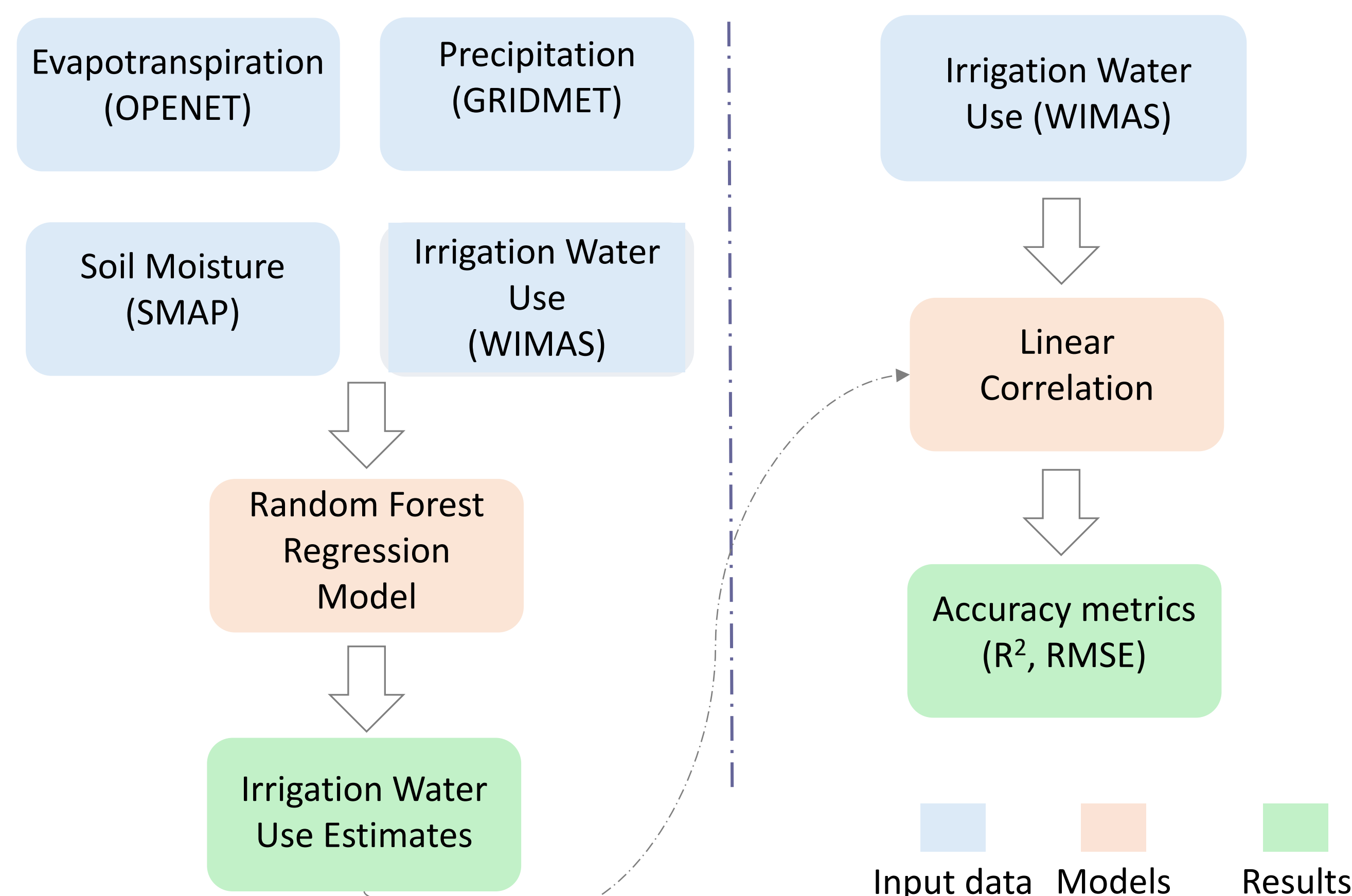


Figure 2: Workflow of input data, sources and machine learning modeling and evaluation process.

### OBJECTIVES

- Predict annual irrigation water withdrawals across Kansas using a machine learning model integrated with multi-source satellite-based soil moisture data.
- Evaluate model estimates with observed irrigation water withdrawals reports from the Water information management system (WIMAS)
- Assess spatial variability of irrigation water-use across Kansas counties from 2015 to 2021.

### RESULTS

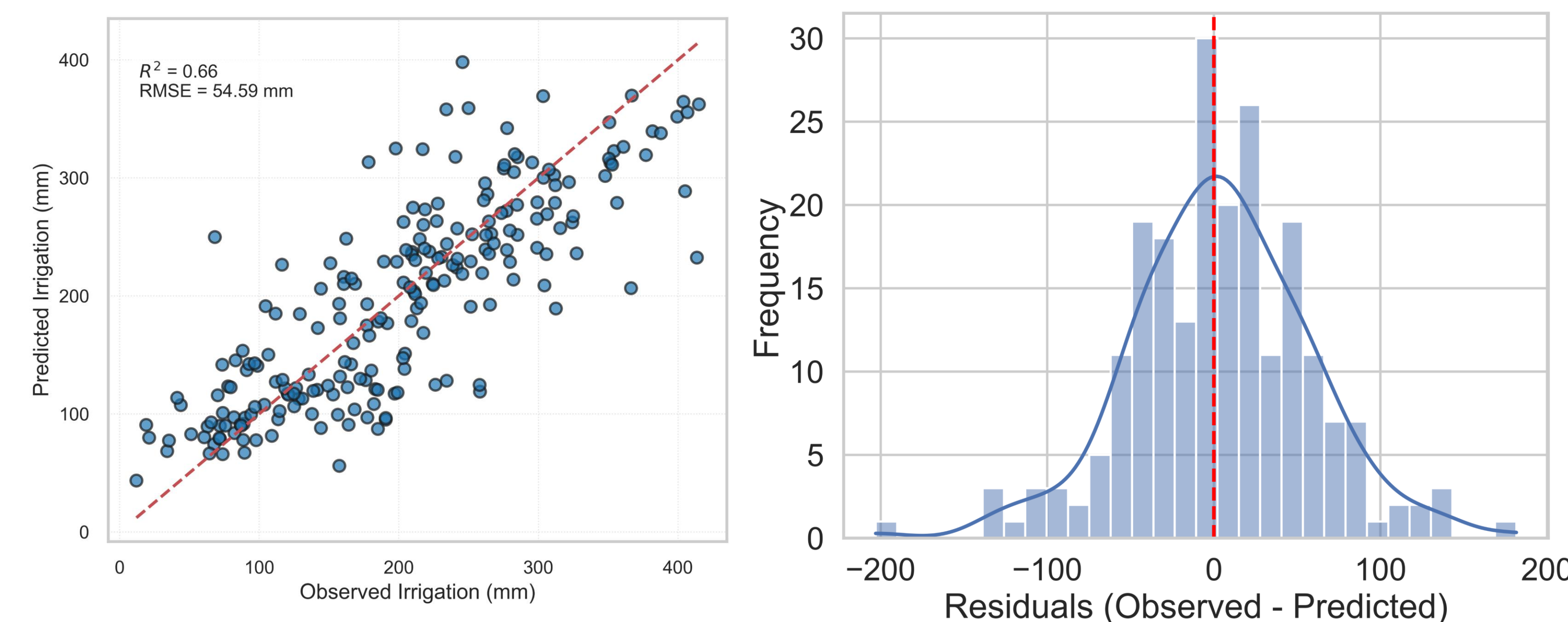


Figure 3: Model prediction accuracy defined by goodness of fit ( $R^2 = 0.66$ ) and error distribution

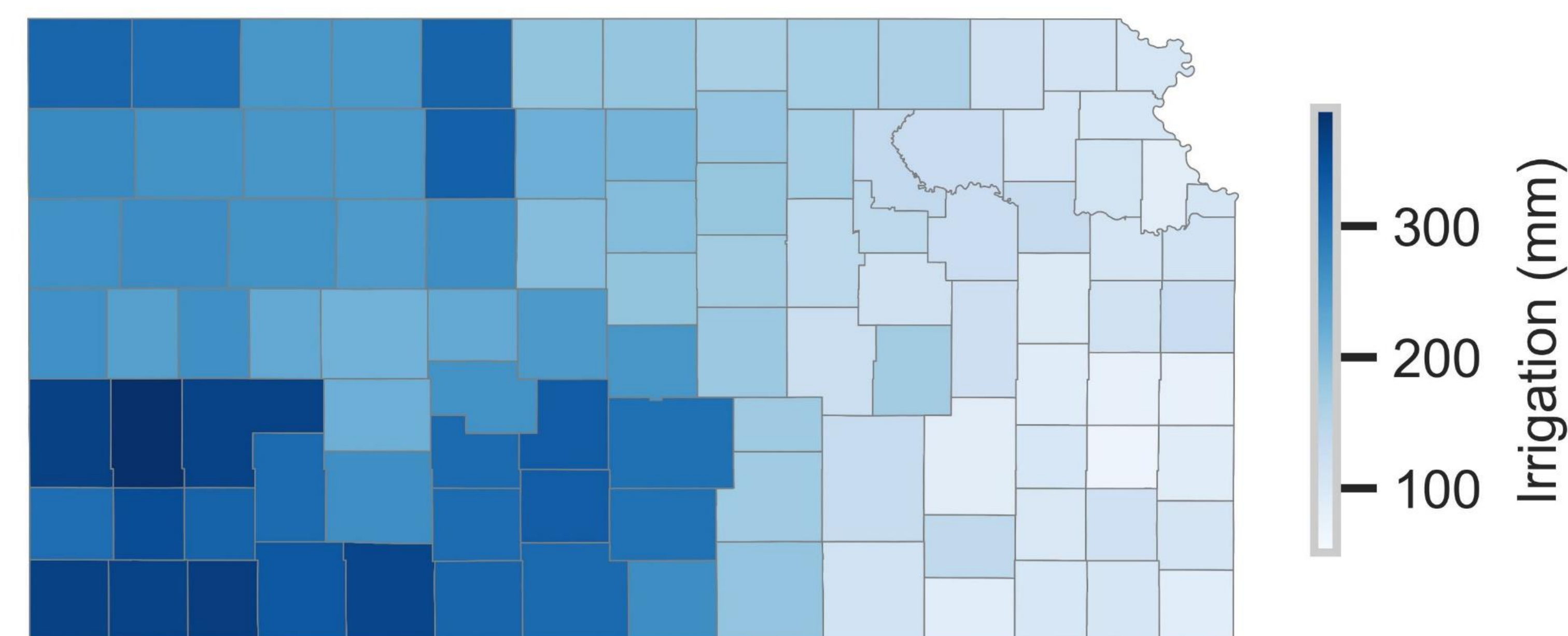


Figure 4: Spatial distribution of predicted Irrigation water use across Kansas Counties

### Result Highlights:

- A Random Forest machine learning model fitted to all variables explained 66% variability in seasonal irrigation water use across Kansas counties over the 2015 to 2021 period. The normally distributed errors also improved confidence in the model accuracy (Figure 3)
- Model estimates accurately captures the east to west increasing pattern of irrigation water use in Kansas (Figure 4)
- Surface soil moisture emerged as most influential variable in irrigation water use prediction (Figure 5)
- Soil moisture variables showed stronger relationship with irrigation than precipitation and actual evapotranspiration (Figure 6)

### RESULTS

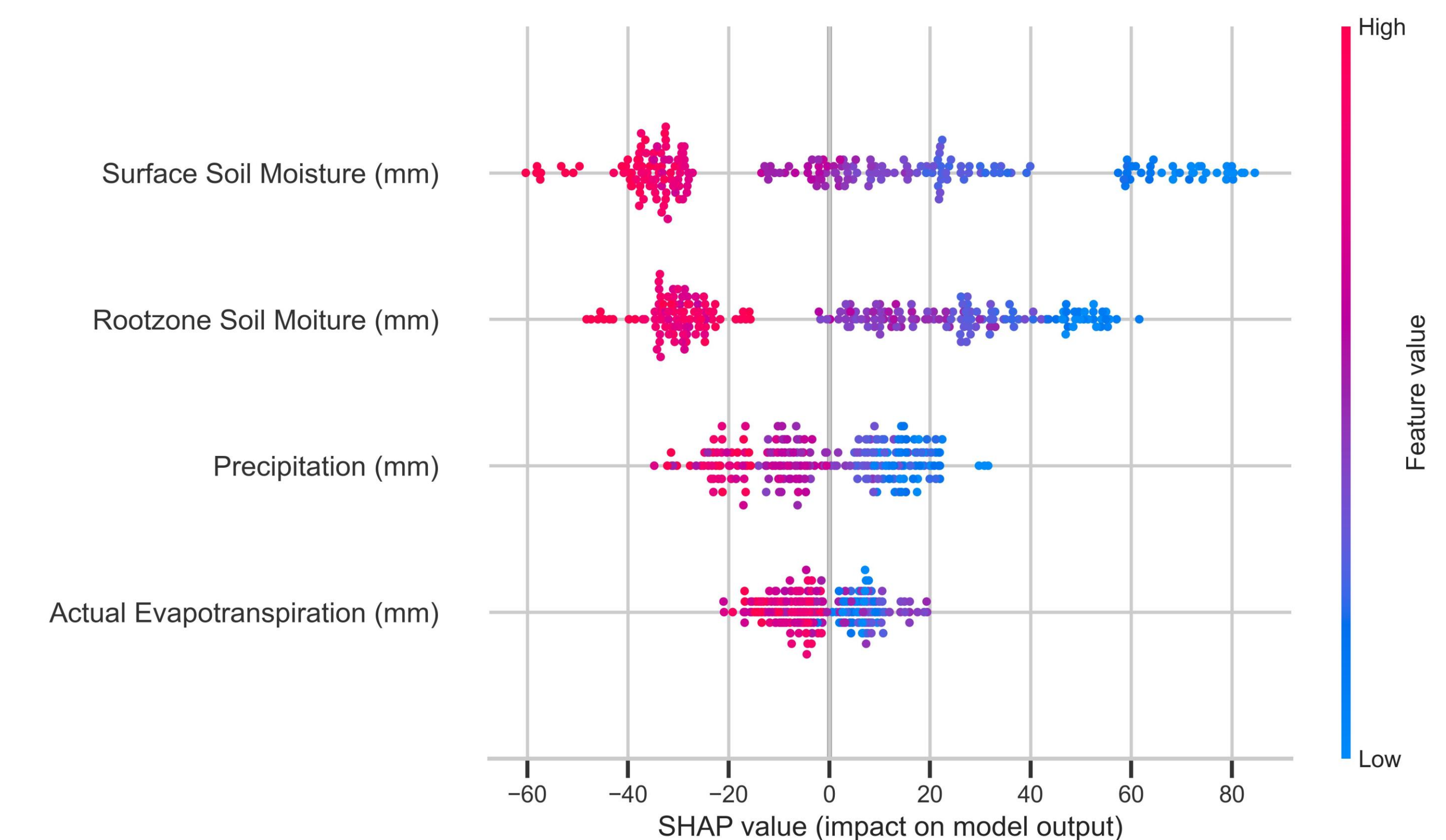


Figure 5: Importance of variables for predicting Irrigation ranked based on impact on model output

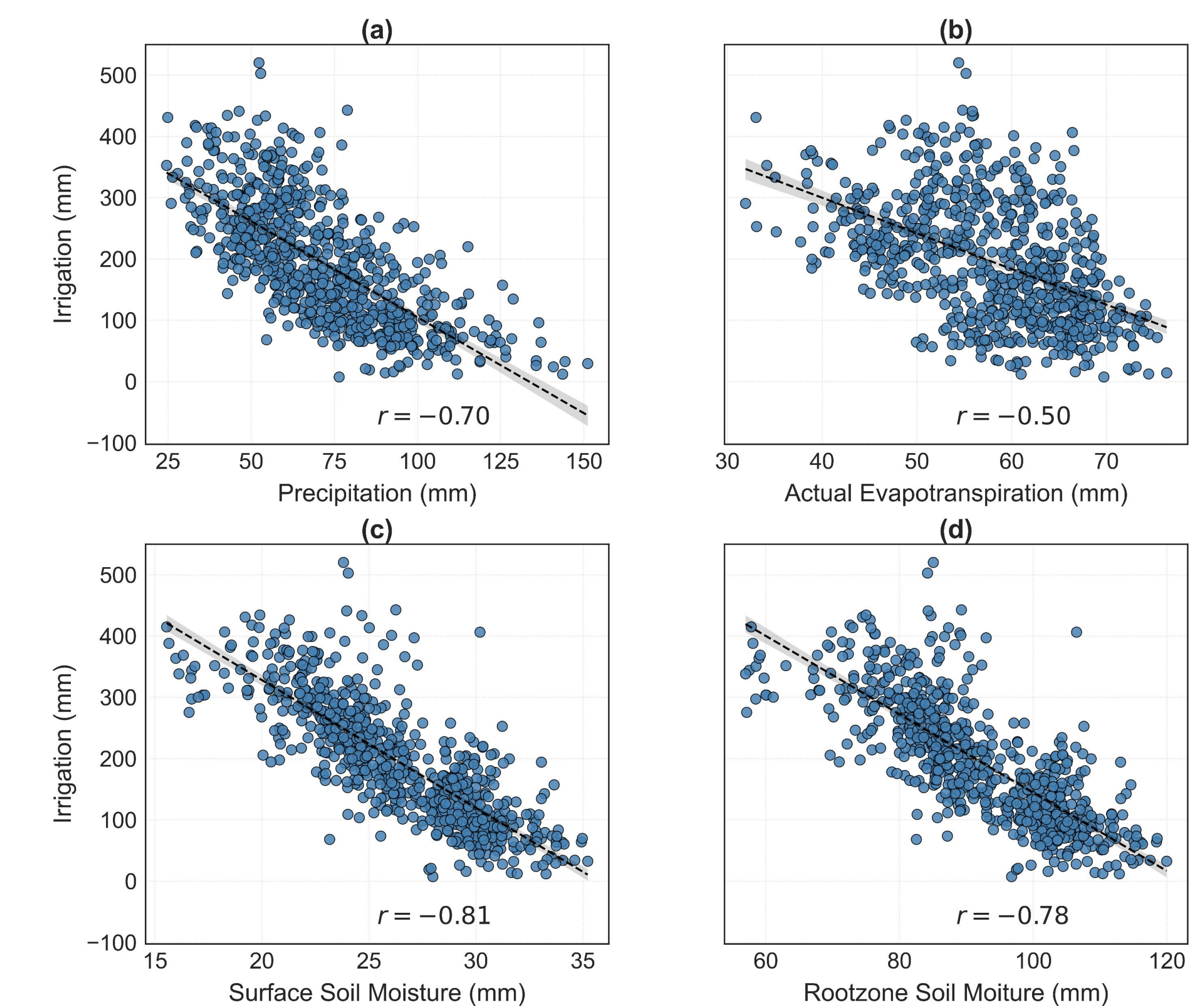


Figure 6: Pearson correlation irrigation with (a) Precipitation (b) evapotranspiration (c) Surface and (d) Rootzone soil moisture

### CONCLUSION & FUTURE WORK

- Remotely sensed data integrated into machine learning model shows potential for continuous water use monitoring
- Model explained nearly 70% variation in seasonal irrigation withdrawals across Kansas counties
- Surface soil moisture is crucial to enhancing the accuracy of seasonal irrigation water use estimation.
- Future study will evaluate the influence of spatial resolution, depth of extraction and time of acquisition of soil moisture datasets on model performance accuracy