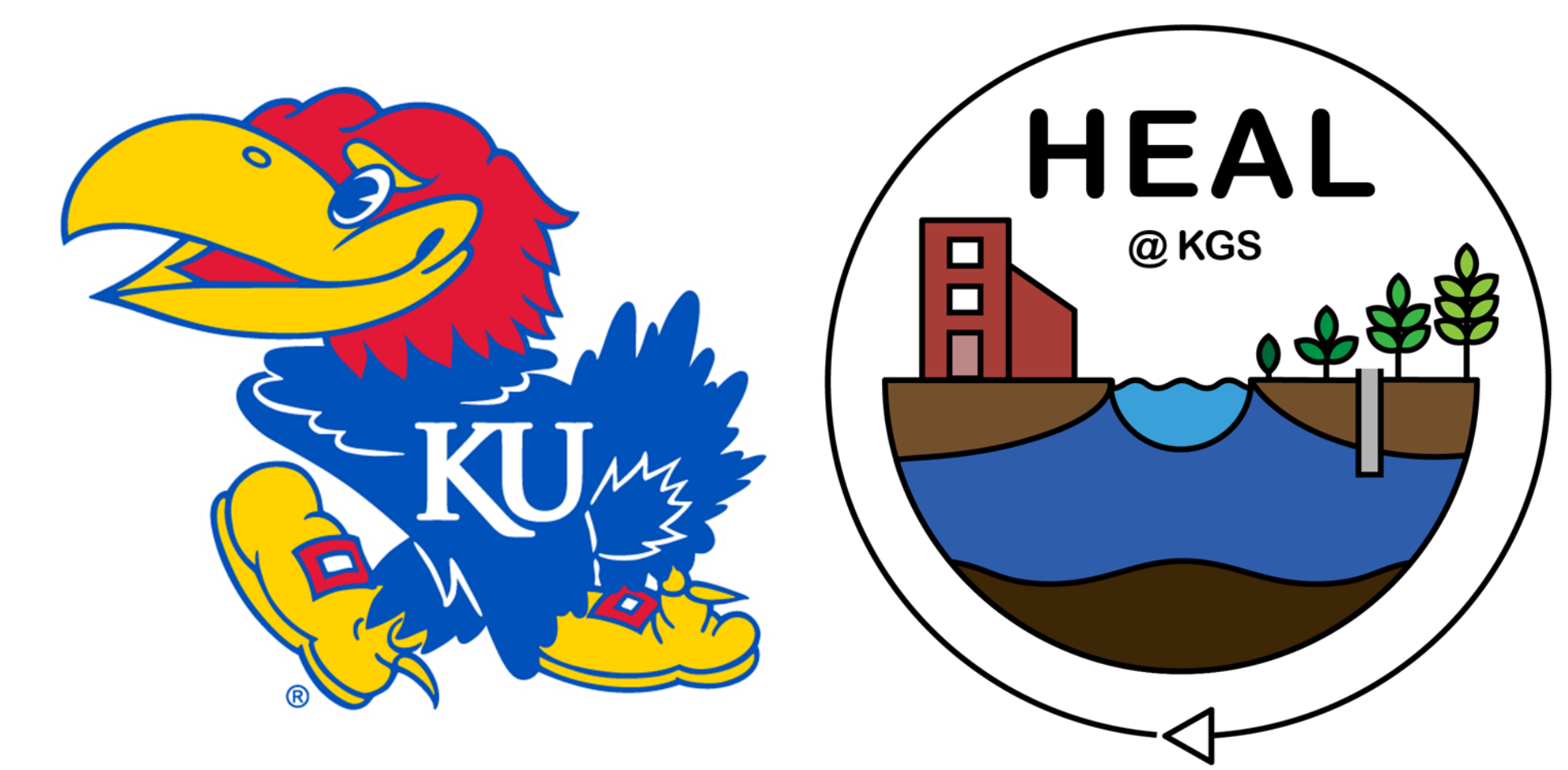


Predicting streamflow in intensively-managed agricultural basins across High Plains Aquifer

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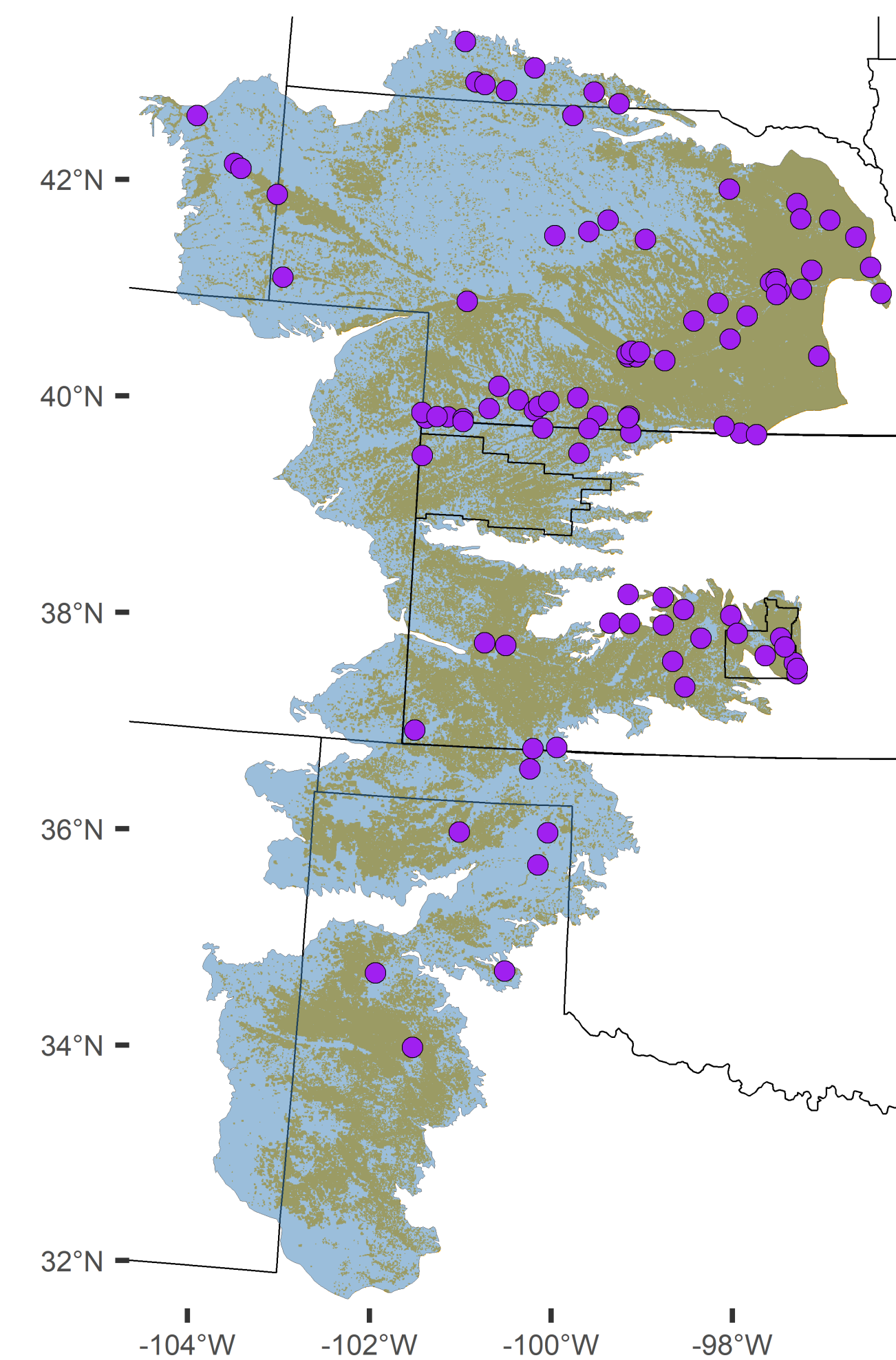
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Motivation

- Deep learning models are promising in terms of their predictive capabilities, but most studies to date have evaluated performance in small catchments with minimal human disturbance.
- The High Plains Aquifer in the central US is important for global food production, but irrigation has led to declining aquifer levels and altered stream-aquifer interactions.

Objective: Use a *Long Short-Term Memory* model to predict streamflow for historical conditions across the intensively-managed High Plains Aquifer and evaluate model performance.



• USGS Stream Gauges (n = 96)

Figure 1. USGS stream gauges (n = 96) overlying the High Plains Aquifer with at least 20 years of data used in our study.

Model Overview

- Long Short-Term Memory* (LSTM) models are a type of recurrent neural network capable of learning long-term relationships between inputs and outputs, and are well-suited for hydrological modeling.
- Calibrated a regional LSTM using meteorological forcing variables and static catchment attributes on all stream gauges in the High Plains Aquifer.

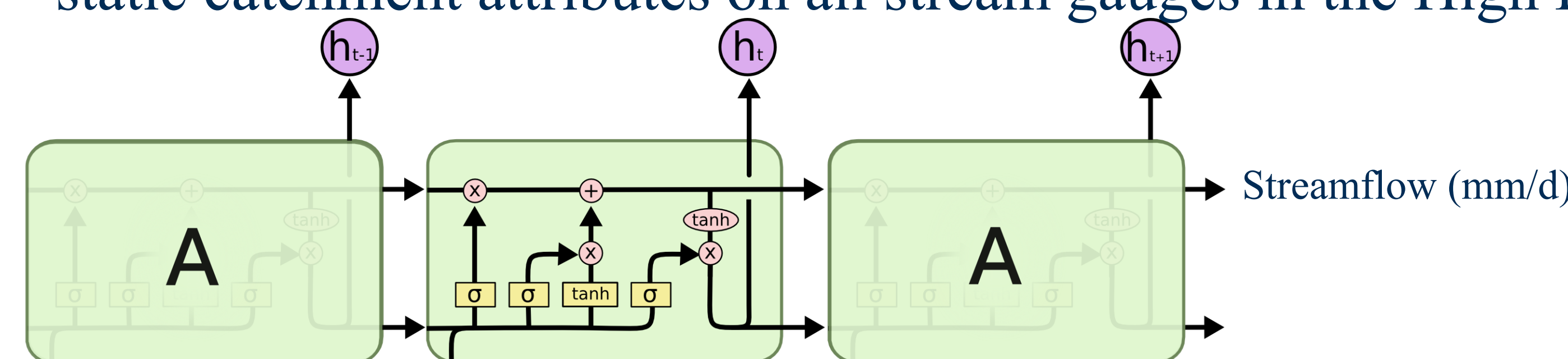


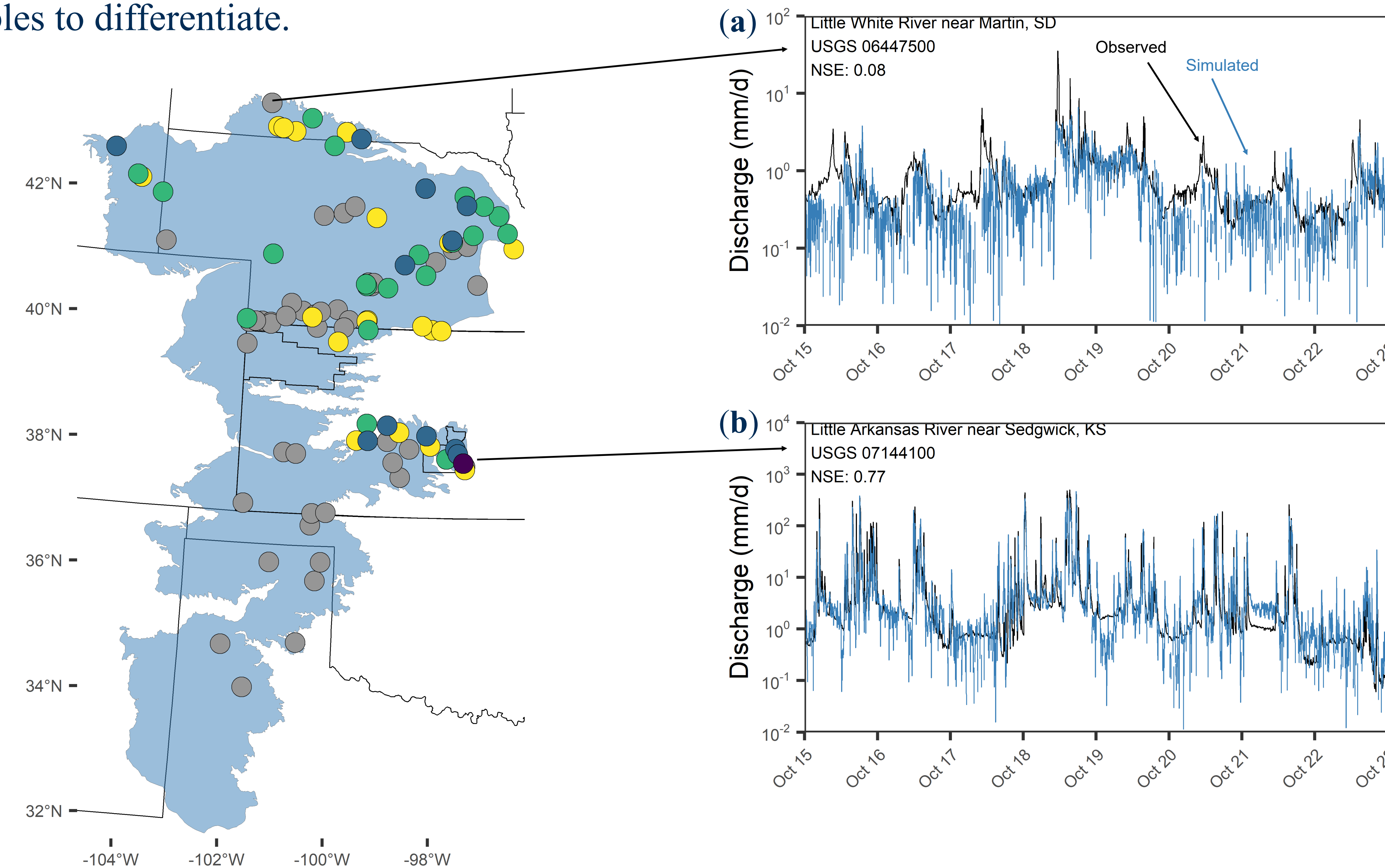
Figure 2. The *forget*, *input*, and *output* gates regulate the flow of information from short-term (*recurrent state*) to long-term (*cell state*) memory.

- Training: 1979 to 2005
- Validation: 2005 to 2015
- Testing: 2015 to 2023

Static climatic and physiographic covariates

Model Performance

- LSTM performance varied spatially with clusters of higher NSE values occurring along the Lower Arkansas River, Elkhorn, and Niobrara Rivers, while the more intensively-managed Republican River had worse performance.
- Hyperparameter tuning was not conducted, but could improve model performance, as well as including additional static catchment attributes to differentiate rainfall-runoff dynamics in the LSTM.
- We interpret poor performance at some stream gauges to be a result of non-correlated rainfall-runoff dynamics between basins that require additional predictor variables to differentiate.



NSE: ● -Inf to 0 ● 0 to 0.25 ● 0.25 to 0.5 ● 0.5 to 0.75 ● 0.75 to 1

Figure 3. Nash-Sutcliffe efficiency (NSE) for modeled sites. Nash-Sutcliffe efficiency: $(-\infty, 1]$; values closer to 1 are desirable.

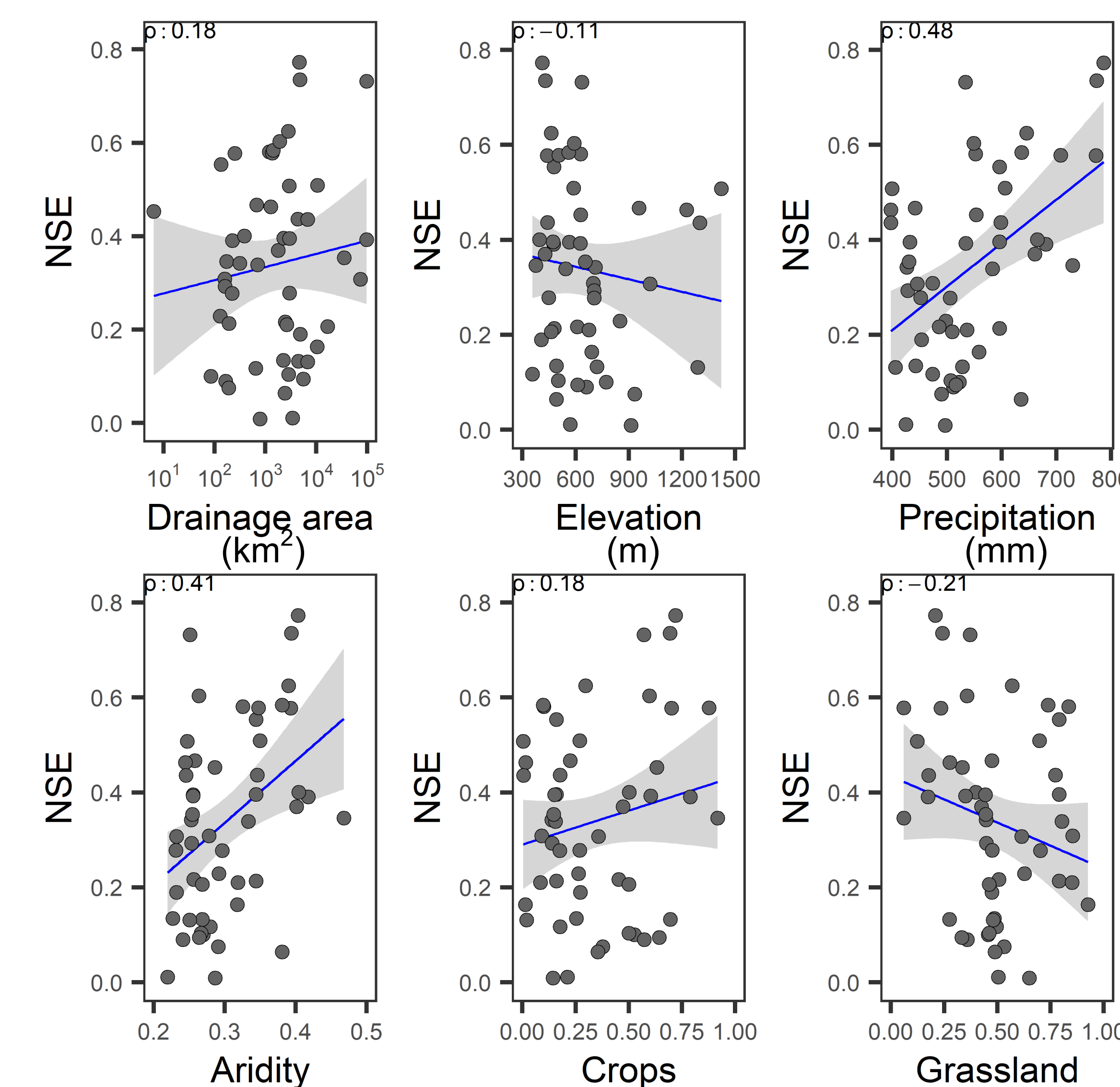


Figure 6. Pearson correlation (ρ) between NSE and catchment climatic and physiographic attributes.

Precipitation and aridity (P/refET) represent long-term annual averages since 1979.

Figure 4. Simulated streamflow at a site closest to the median NSE (a) L. White River near Martin, SD, and at the best performing site (b) L. Arkansas River near Sedgwick, KS.

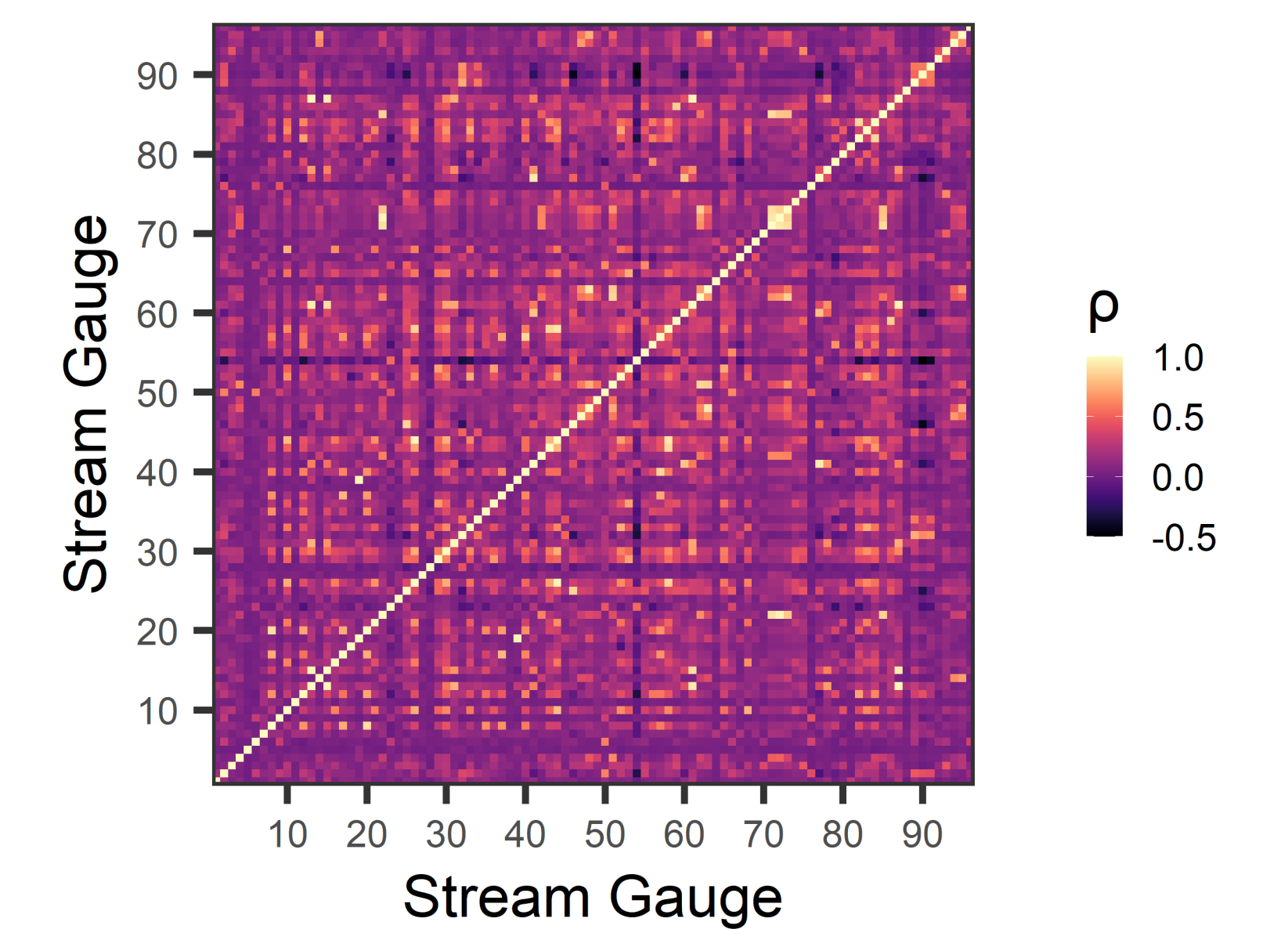


Figure 5. Pearson correlation (ρ) between observed daily flow at each gauge.

Key Findings

- LSTM performance varied spatially across the High Plains Aquifer with NSE ranging from $-\infty$ to 0.77.
- Median NSE during testing was 0.07.
- Better model performance in wetter (northern and eastern) portions of domain.
- Relatively low inter-basin streamflow correlation compared to other datasets (e.g., CAMELS) challenging for LSTM to learn.

Acknowledgements

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