

# ANALYSIS OF MODEL BIASES

ERT 474/574

Open-Source Hydro Data Analytics

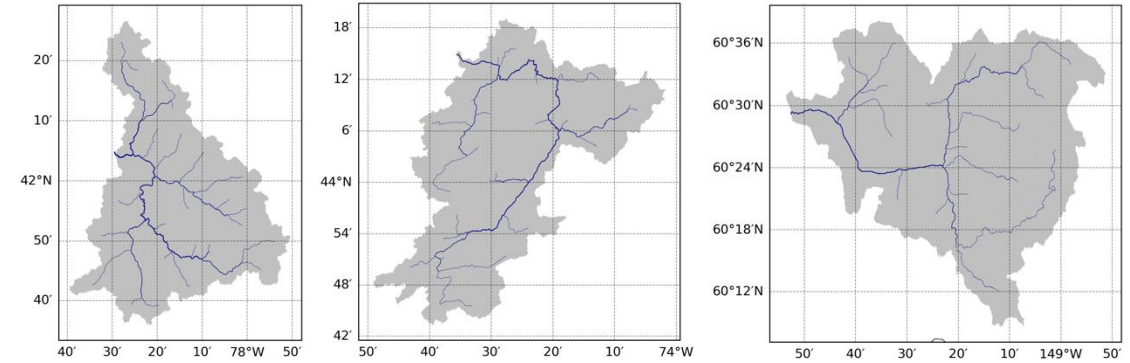
Nov 17<sup>th</sup> 2025

 **University at Buffalo** The State University of New York



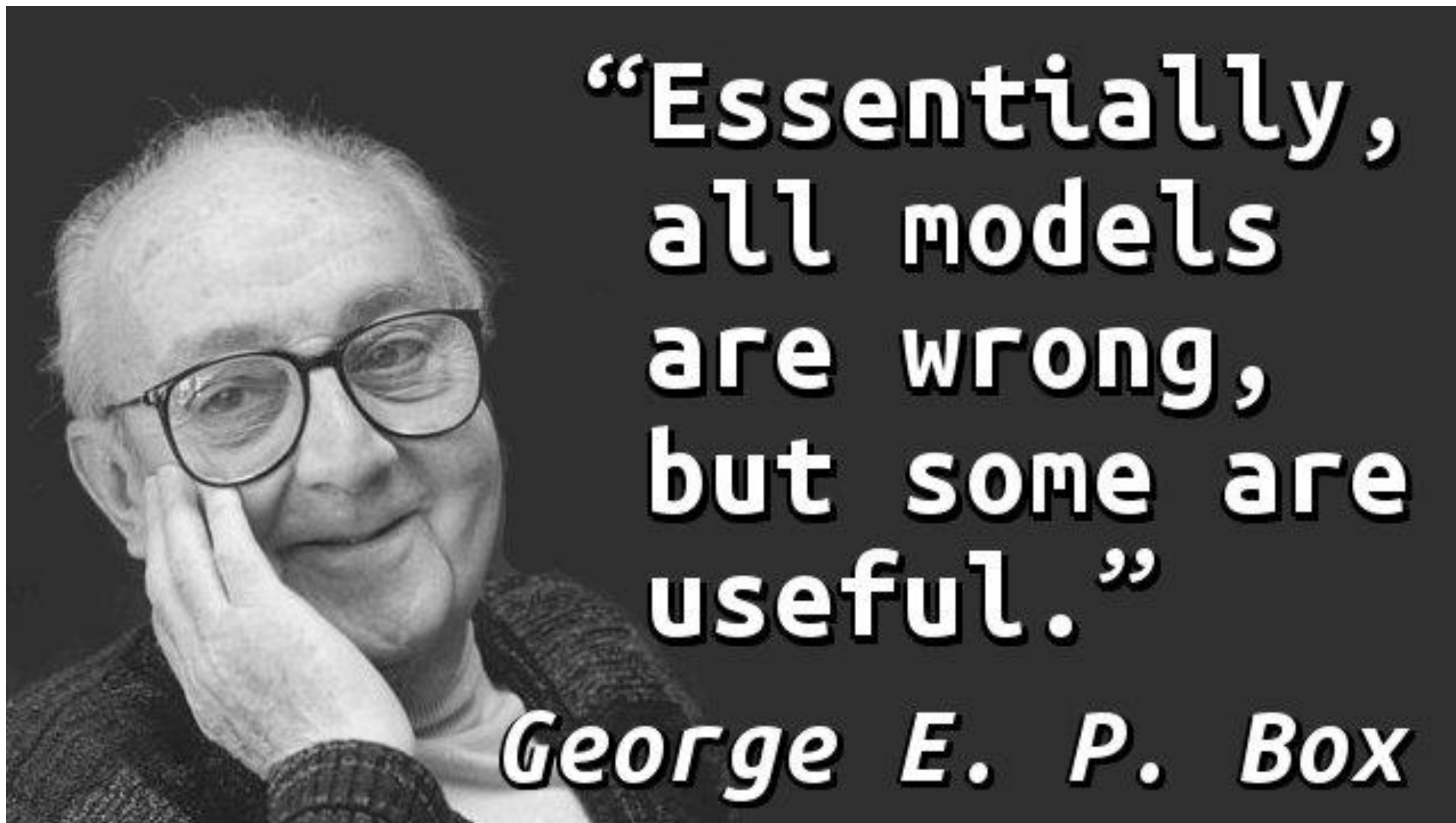
# Final Project

- Set up the VIC and mizuRoute models for the targeted river basin
  - Evaluate the model performance by comparing it against the USGS observations
- Analyze the model biases and identify the potential source of errors
- Based on the analysis of potential errors, conduct a parameter sensitivity analysis for ONE parameter
  - Choose one updated parameter value and check whether modifying this parameter will improve the model performance for your model



## Timeline:

- Nov 24: progress update (each group is expected to finish setting up VIC and mizuRoute models, and conducting initial evaluation against the USGS time series)
  - 5 min for each group
- Dec 8: Final presentation (30%)
  - 15 min for each group
- Dec 15: Final report (60%)
  - 10-page report for each group (including reference)



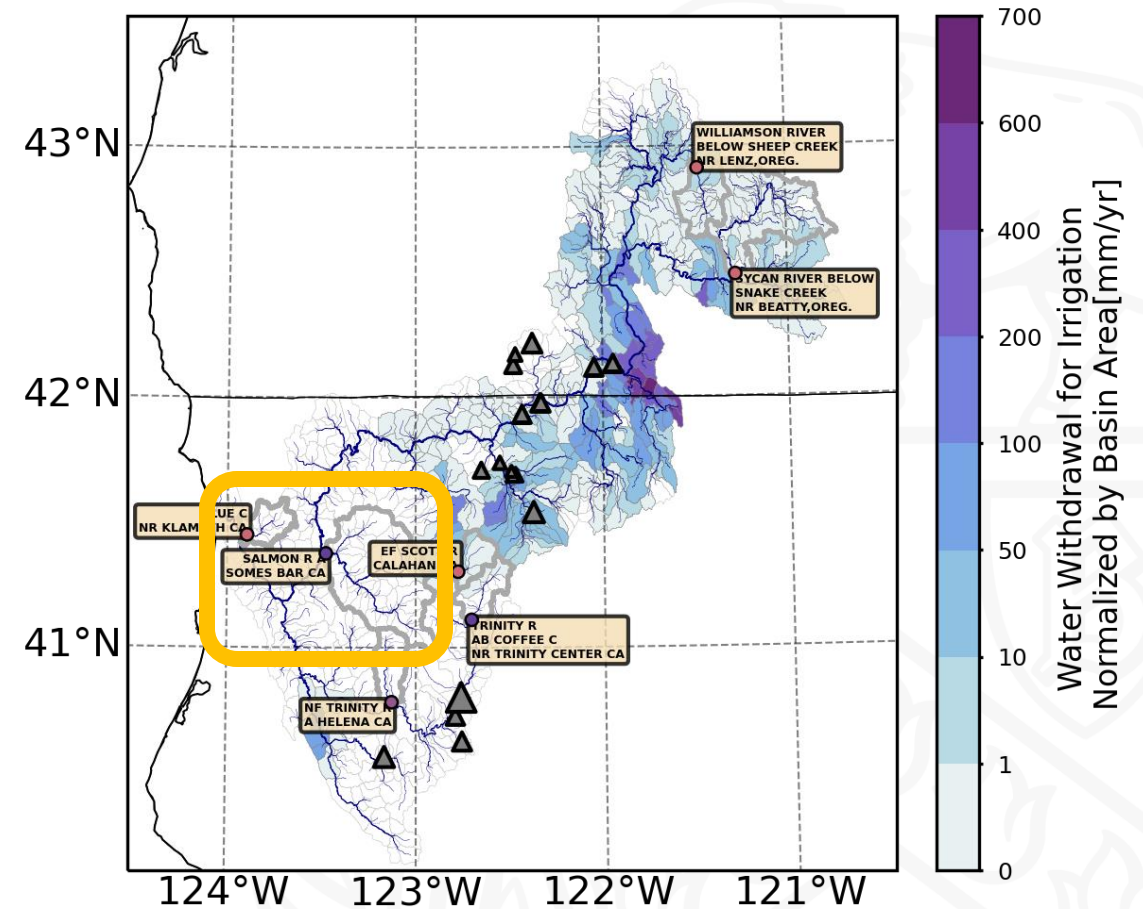
# Outline

- Type of biases
- Source of biases
- Metrics for evaluating biases
- Case studies – Klamath River Basin
- Procedures to analyze the source of biases



# Type of biases

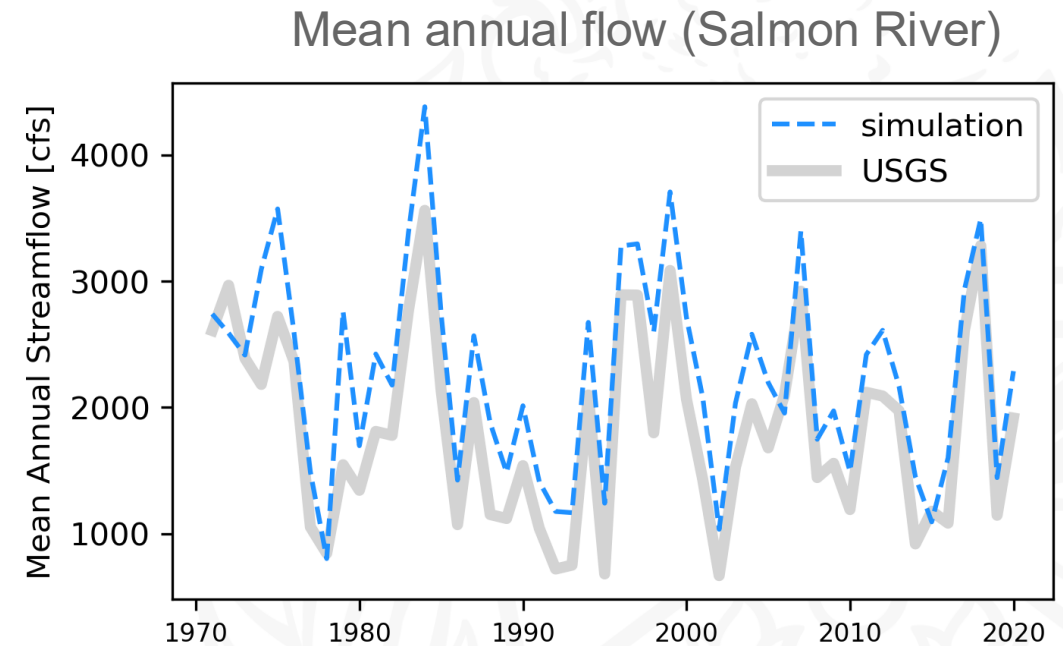
- Systematic Bias: Persistent over- or underestimation (e.g., streamflow).
- Seasonal Bias: Errors that vary by season (e.g., snowmelt timing).
- Event-Based Bias: Errors during extremes (e.g., floods, droughts).
- Spatial Bias: Differences across regions or catchments.



Examples will be shown for the Salmon River in the Klamath River Basin

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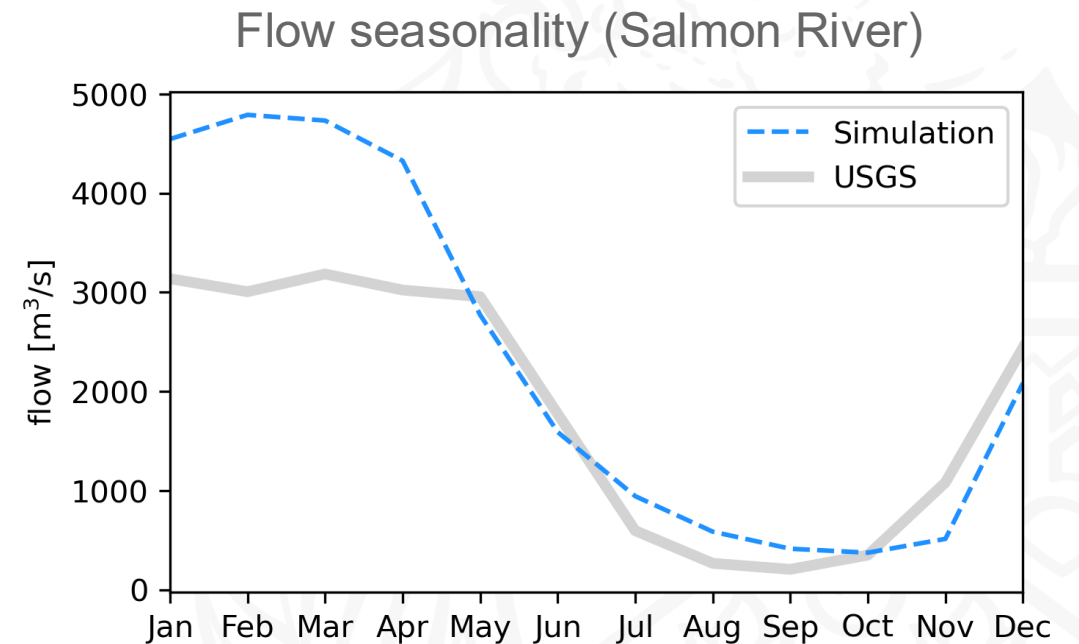
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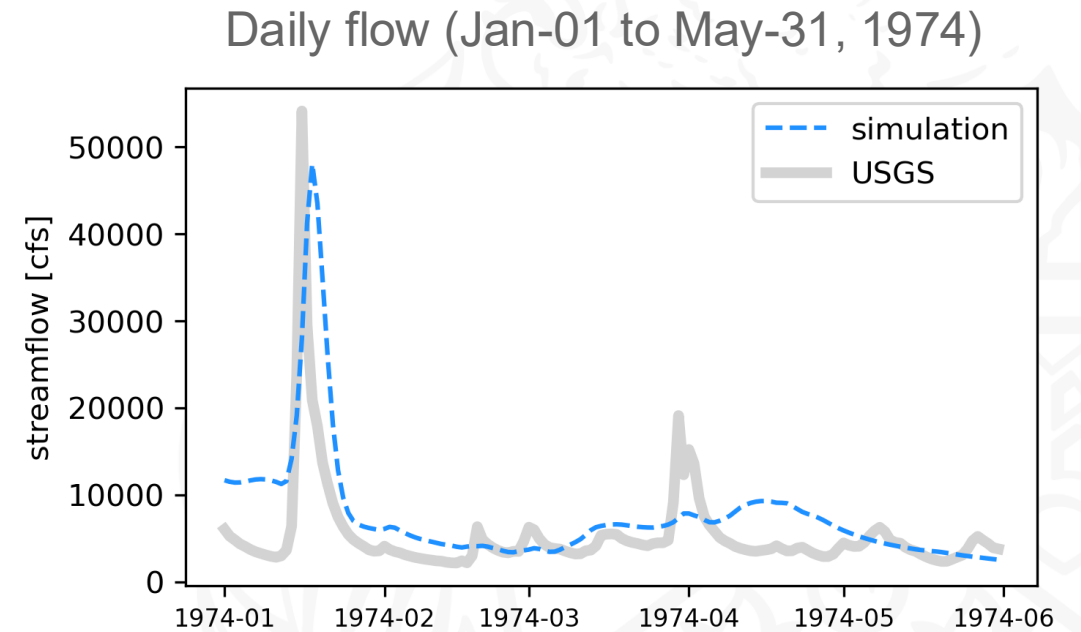
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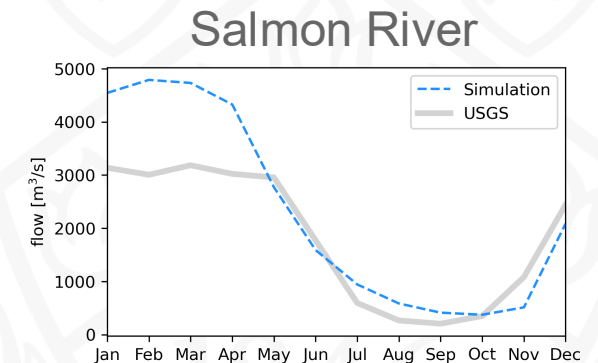
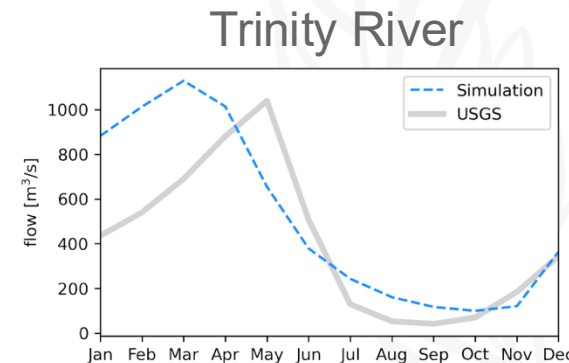
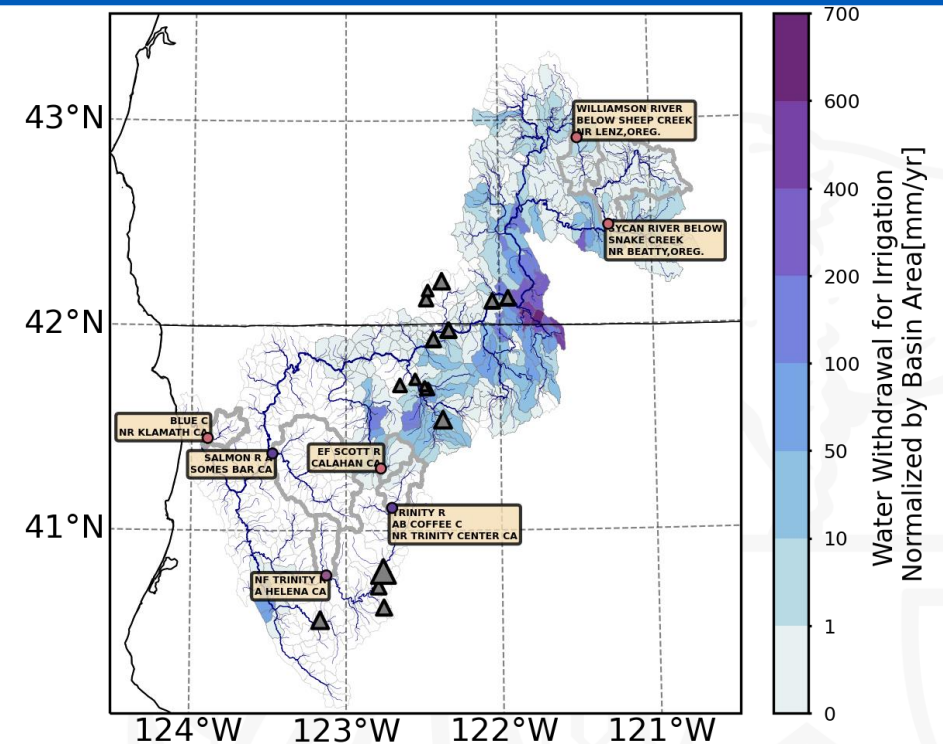
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## Source of biases

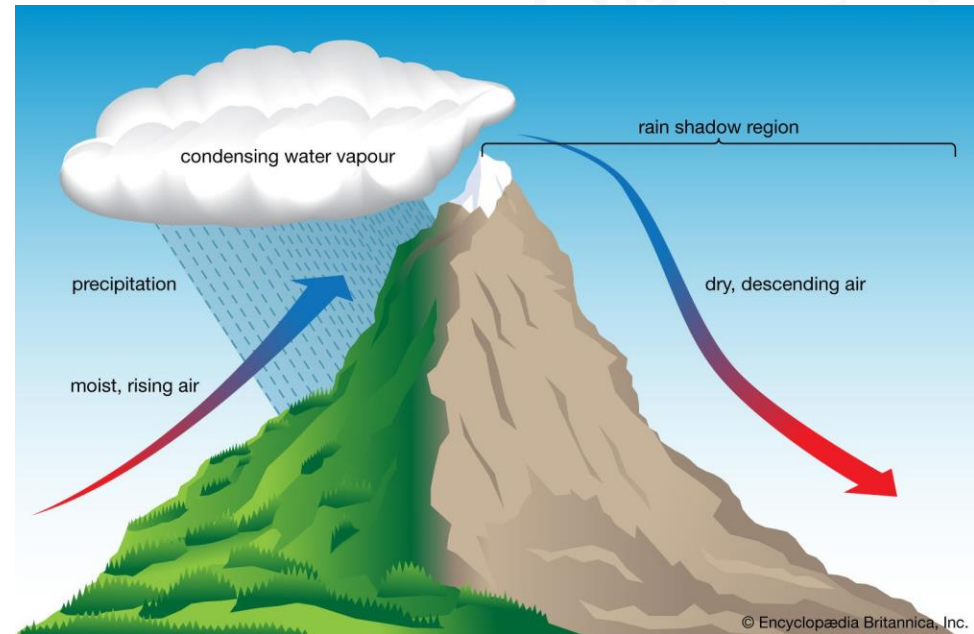
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- Parameter Uncertainty: Infiltration and baseflow parameters are often used in VIC model tuning.
- Model Structure: Simplifications in snowmelt or infiltration processes.
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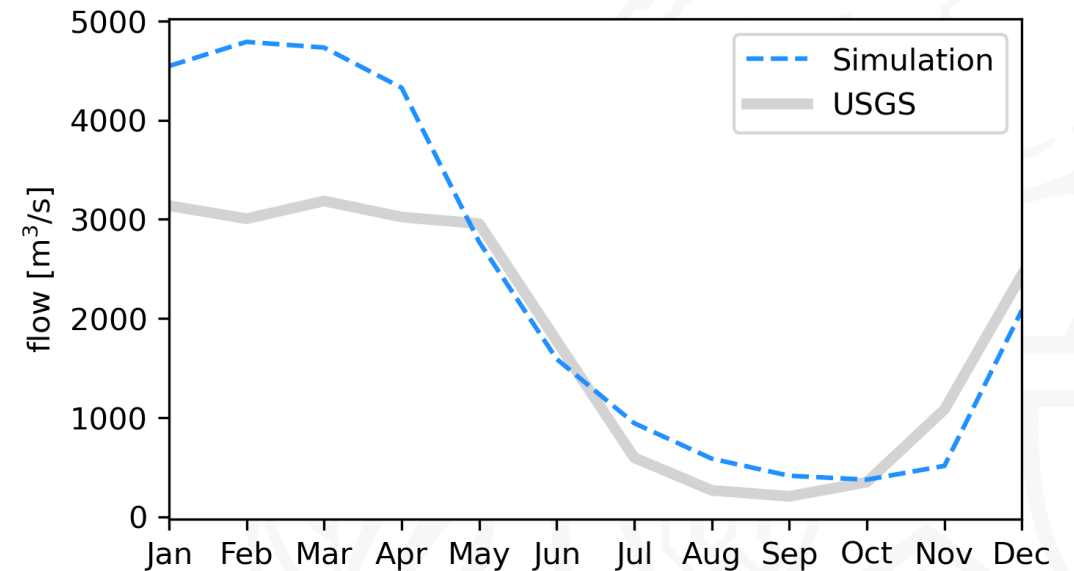
Why would coarse spatial resolution of met forcing be a problem?



Because of the topographic precipitation.

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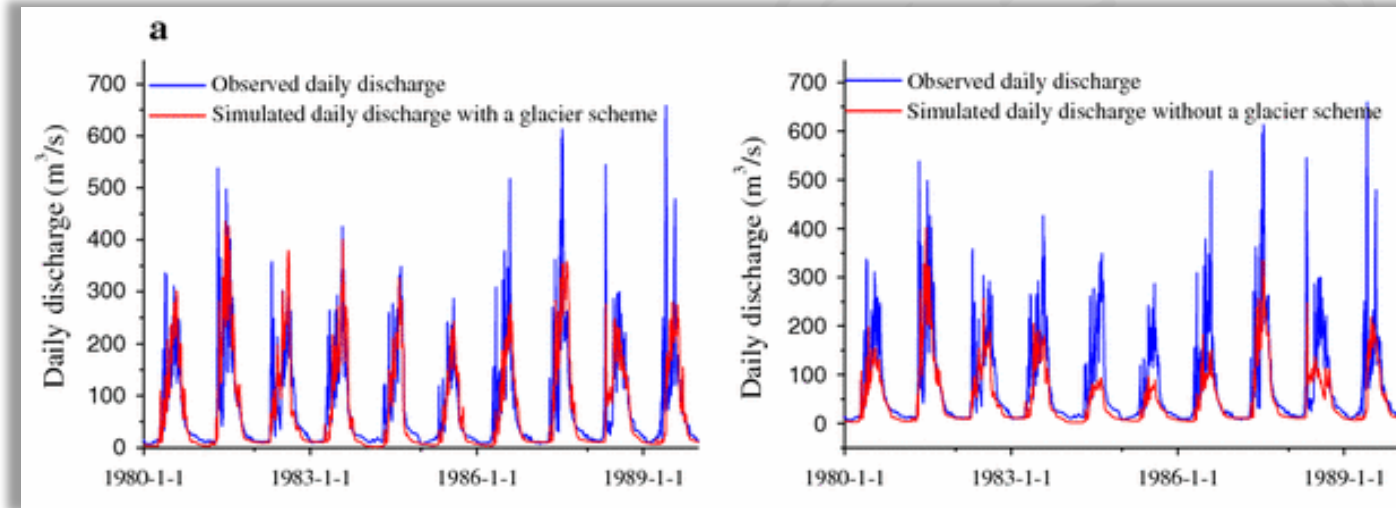
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During the summer, when the model simulation consistently overestimates the streamflow, it is highly possible that it results from the baseflow parameters

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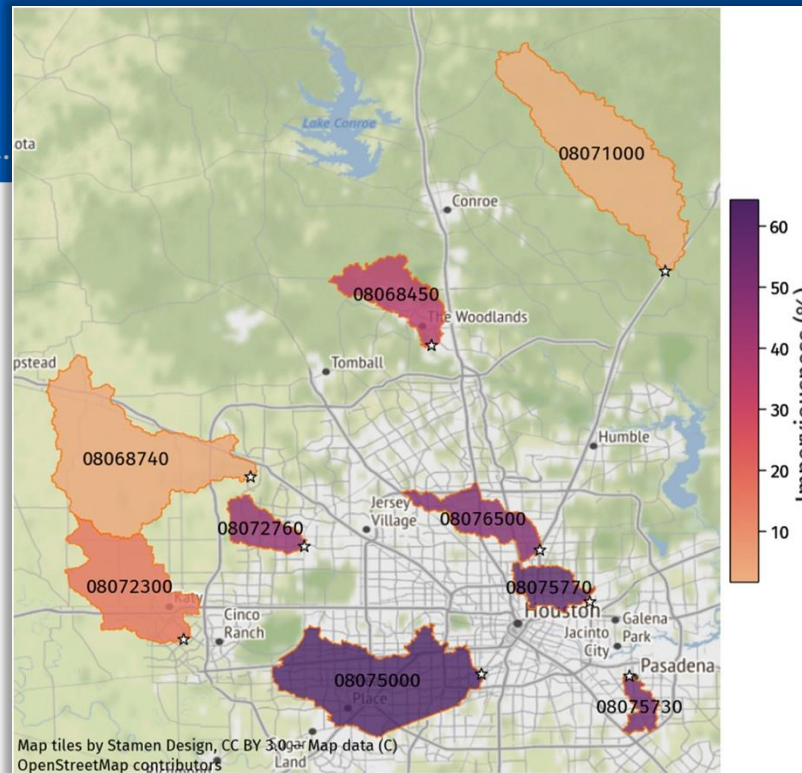


**Coupling a glacier melt model to the Variable Infiltration Capacity (VIC) model for hydrological modeling in north-western China**

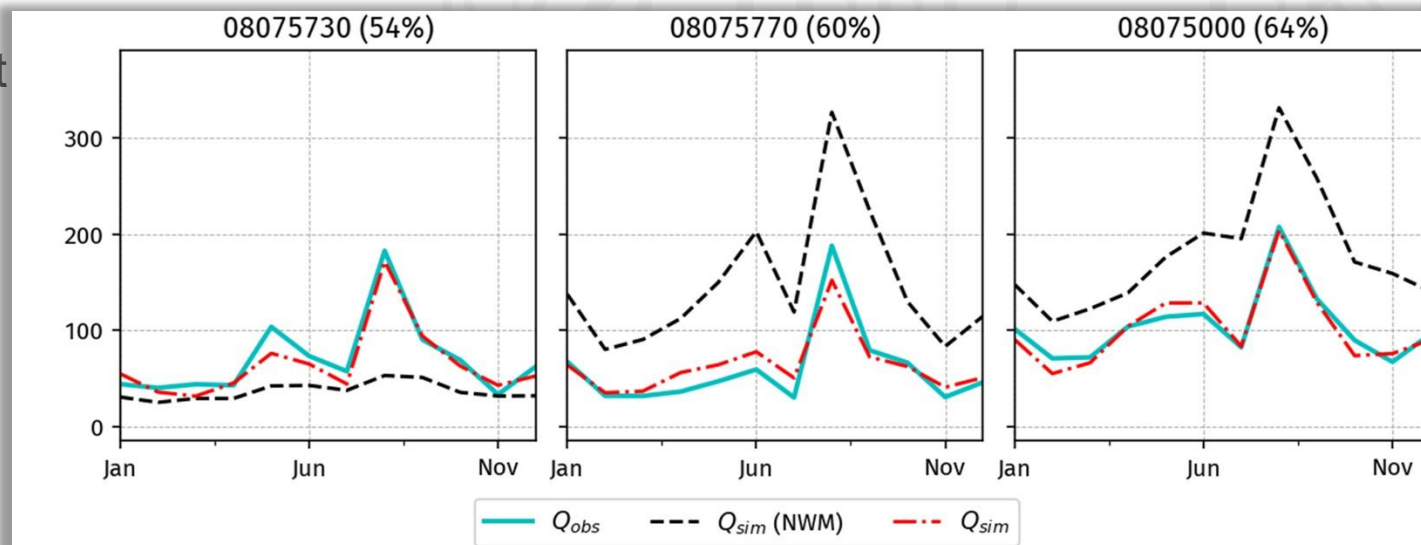


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A Scale-Adaptive  
Urban Hydrologic  
Framework:  
Incorporating  
Network-Level Storm  
Drainage Pipes  
Representation



Source:  
<https://agupubs.onlinelibrary.wiley.com/doi/full/10.1029/2022WR037268>



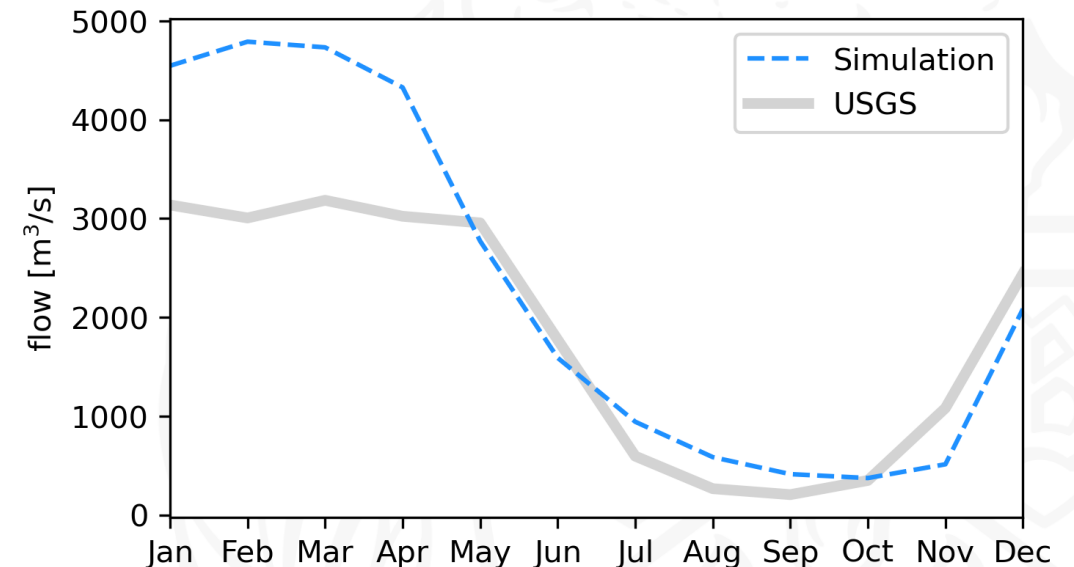
# Metrics for evaluating biases

- Mean Bias Error (MBE)
- Percent Bias (PBIAS)
- Root Mean Square Error (RMSE)
- Nash-Sutcliffe Efficiency (NSE)
- Kling-Gupta Efficiency (KGE)



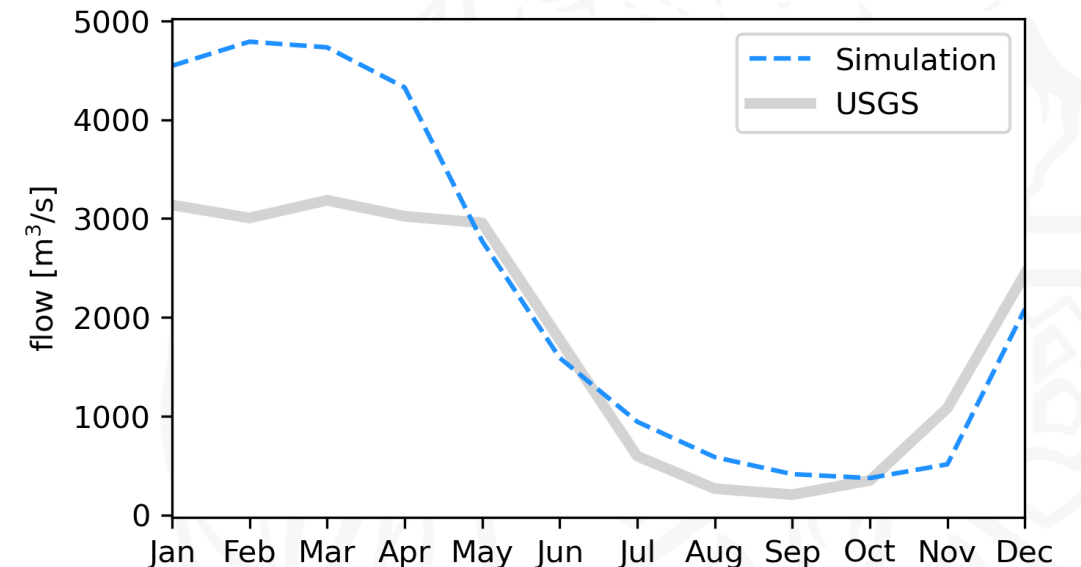
# Metrics for evaluating biases

- Mean Bias Error (MBE)
- Percent Bias (PBIAS)
  - For MBE and PBIAS, we can use it to calculate not only mean annual values, but also seasonal biases.
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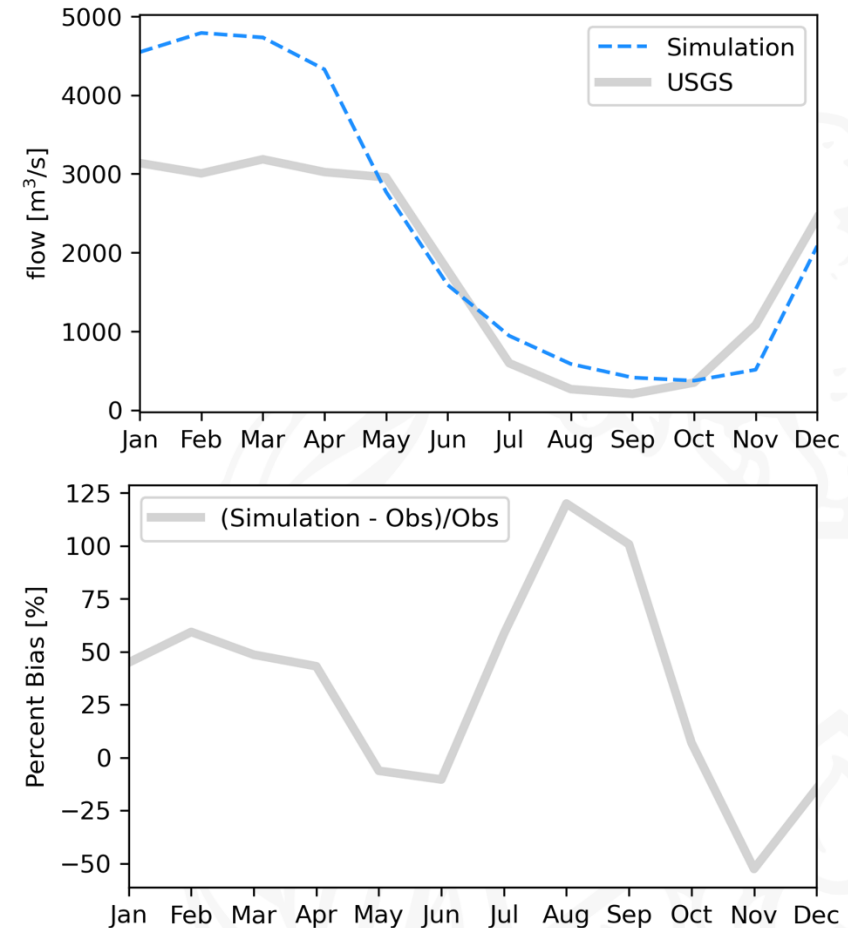
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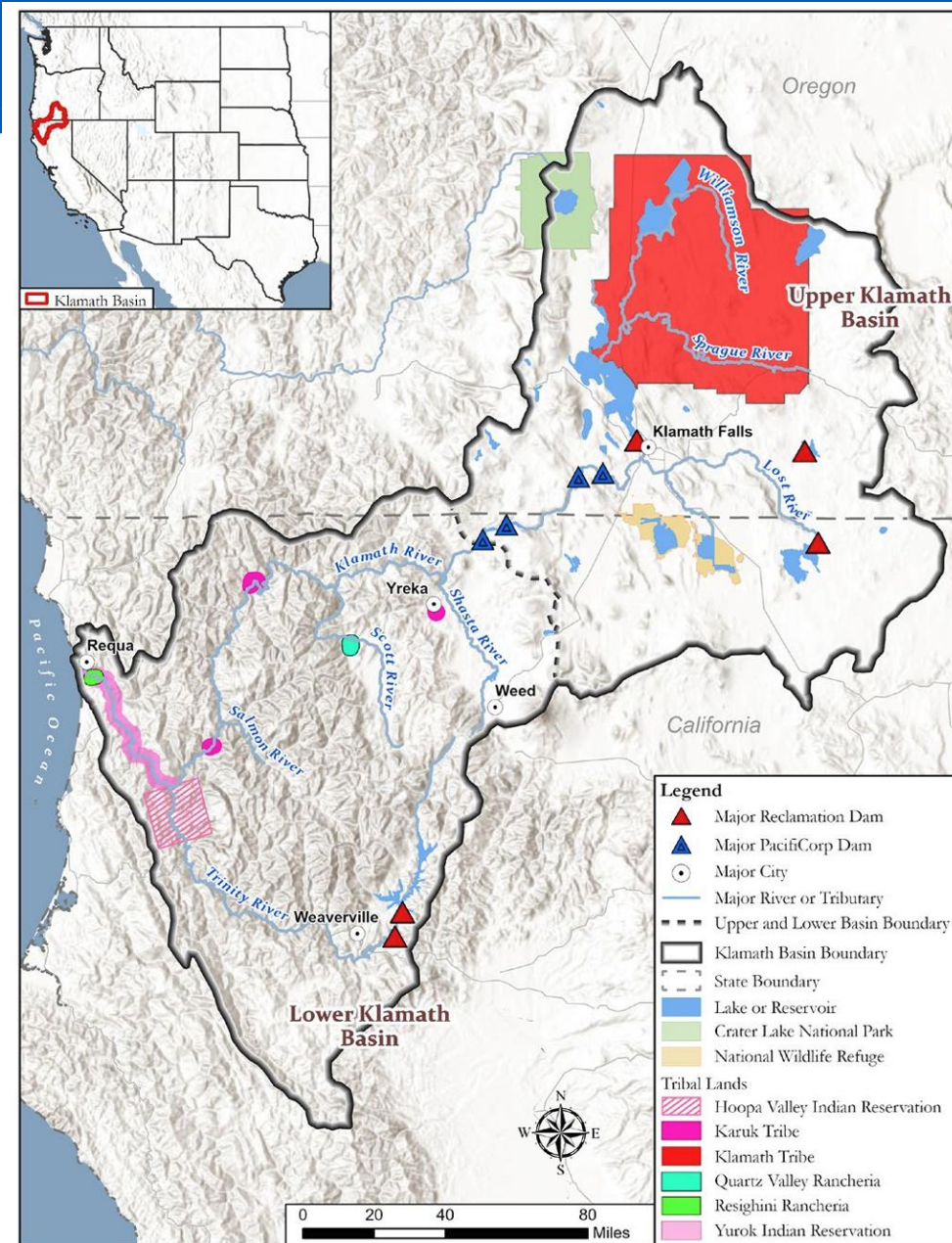


Even though the largest absolute bias occurred in winter, the largest relative bias occurred in the summer.

# Case Study #1 – Klamath River Basin

## Local hydrology summary

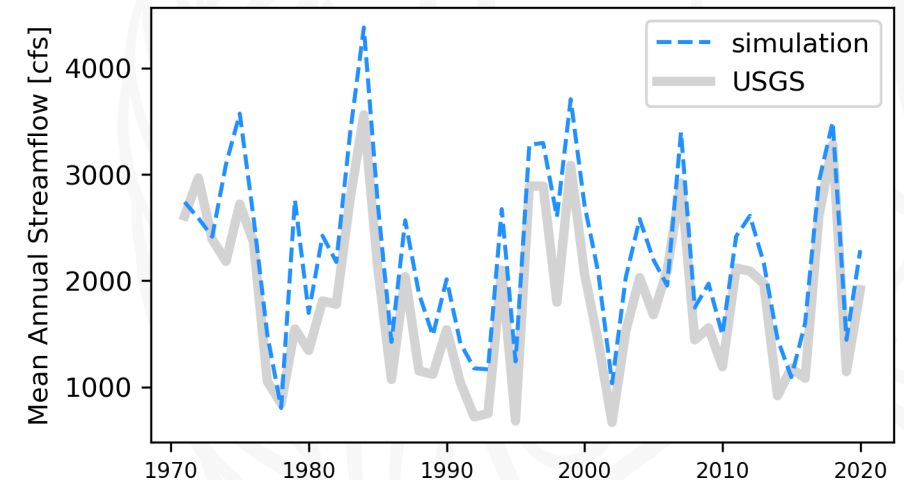
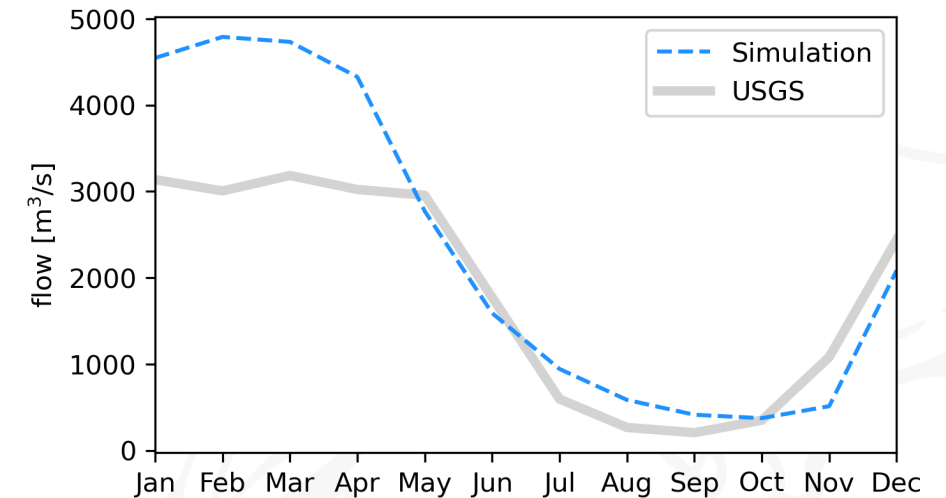
- Cold and wet winter, hot and dry summer
- Winter precipitation may accumulate temporarily as snowpack in the Klamath Mountains and the Cascade Range,



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Bias identification for Salmon River

- Consistent overestimation of mean annual streamflow
- Overestimation of winter streamflow
- Overestimation of summer streamflow





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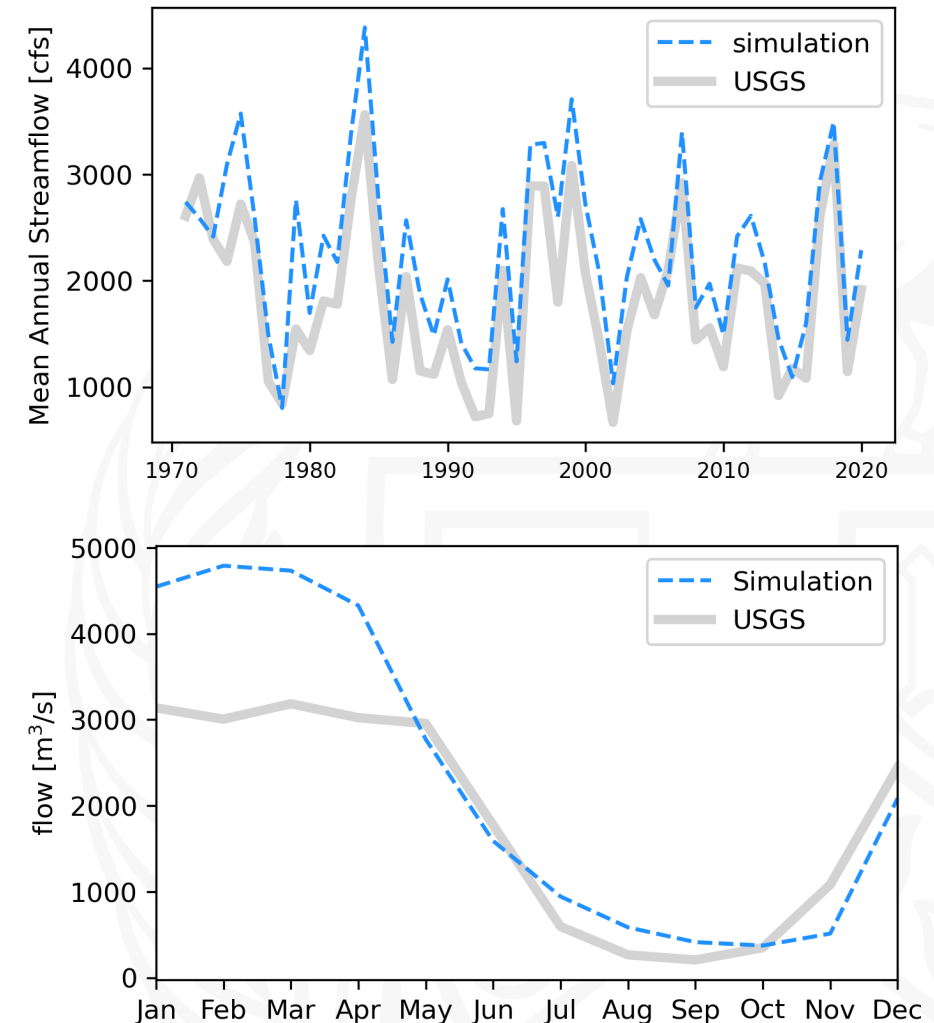
Precipitation bias

Parameters that partition runoff/ET

Precipitation did not stored in snow

Temperature bias (too hot)

Parameters related to ET and Baseflow



# Procedures to analyze the bias in hydrologic modeling

- First, learn the hydrology in this region.
  - What type of climate is in this region (dry/humid, precipitation/temperature seasonality)?
  - Does this area snow?
- Second, identify the type of biases.
  - Consistently overestimation or only certain seasons have large biases.
  - Postulate certain hypotheses
- Third, conduct follow-up analyses to test your hypotheses

For example:

If consistently overestimating/underestimating streamflow, check whether there are biases in 1) model forcing data, such as too much/too little precipitation, and 2) parameters that affect the partition of runoff/evapotranspiration.

If the seasonality of streamflow is off during the spring/summer seasons, check whether there is snow and snow-related parameters.

If the flow in the low-flow season is consistently underestimated, check the baseflow parameters.

# Questions?

