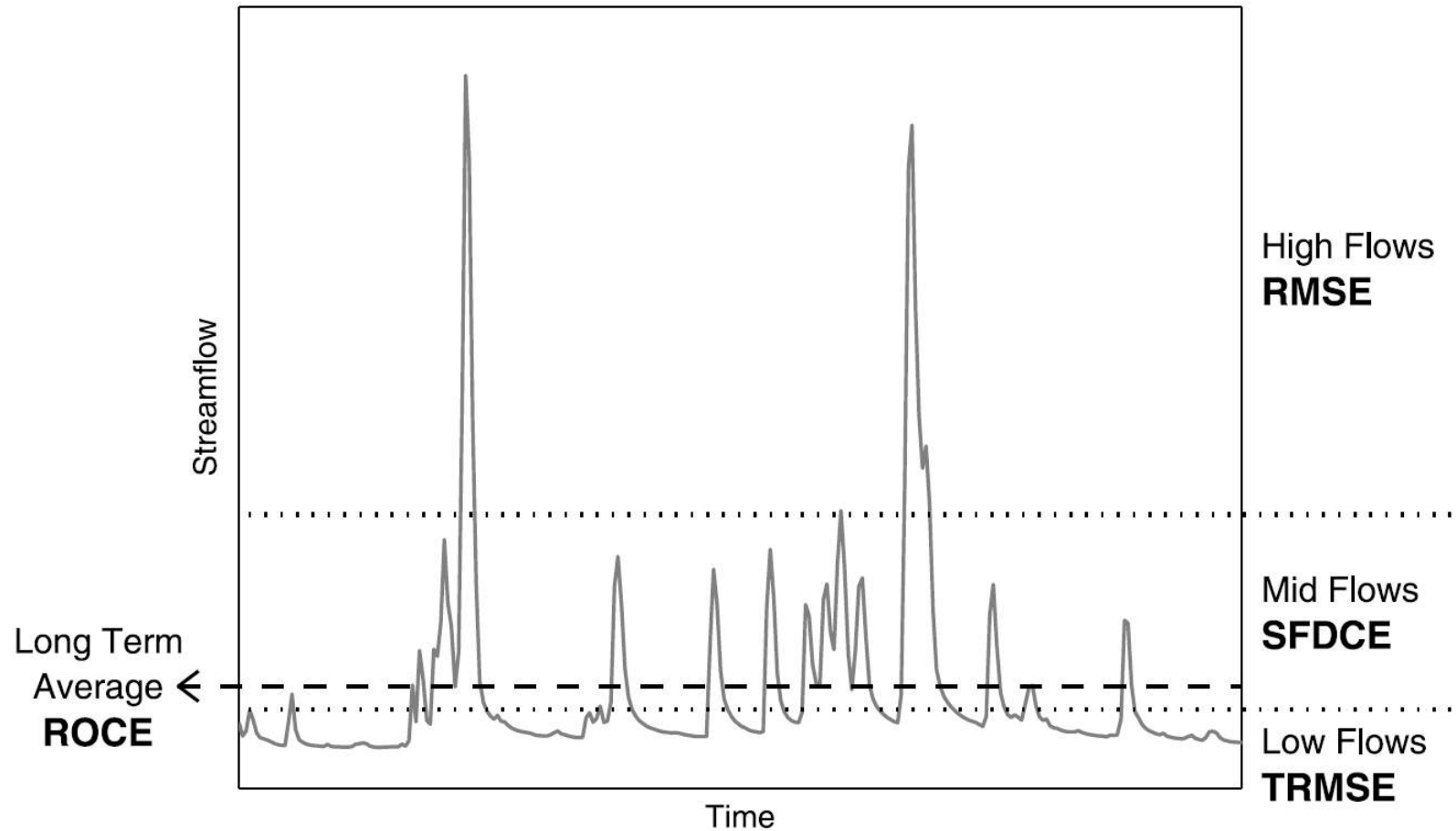
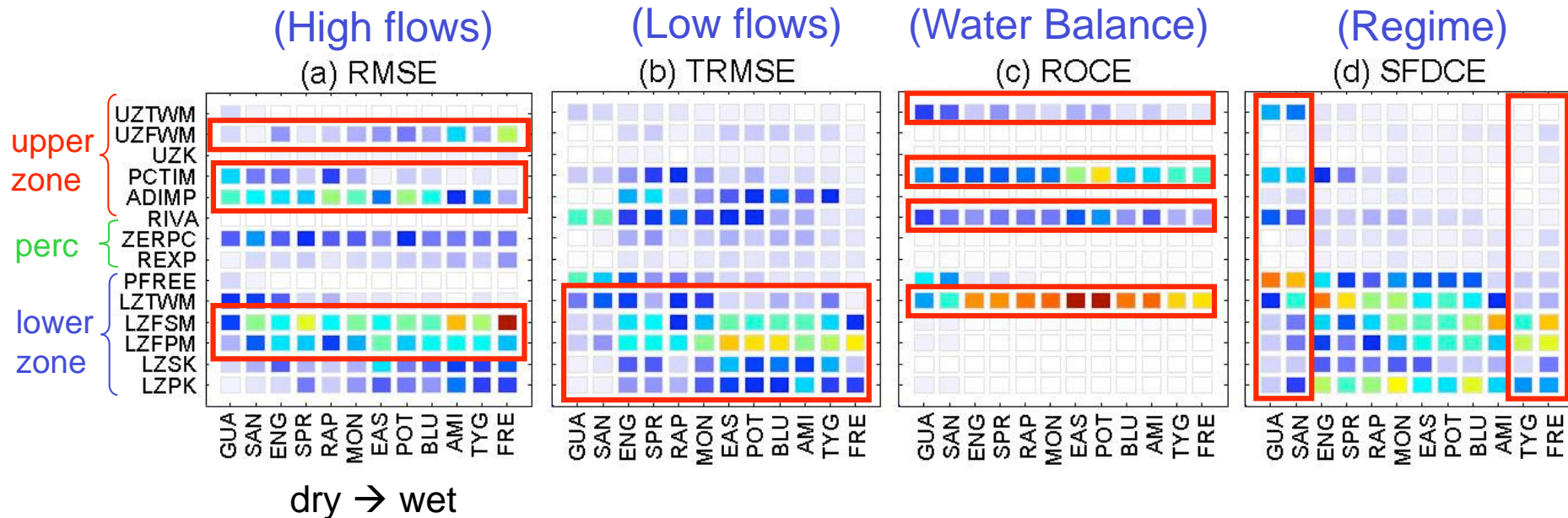


We utilize both statistical and 'hydrological' (signature) objective functions



We find significant variability in parameter sensitivity across the study region



- ✓ Patterns are correlated to hydroclimate, R up to 0.96
- ✓ Impervious area parameters important for peaks
- ✓ Lower zone impacts peaks through percolation
- ✓ Similar lower zone behavior for RMSE and TRMSE
- ✓ Importance of parameters that control ET losses
- ✓ Large differences between driest and wettest

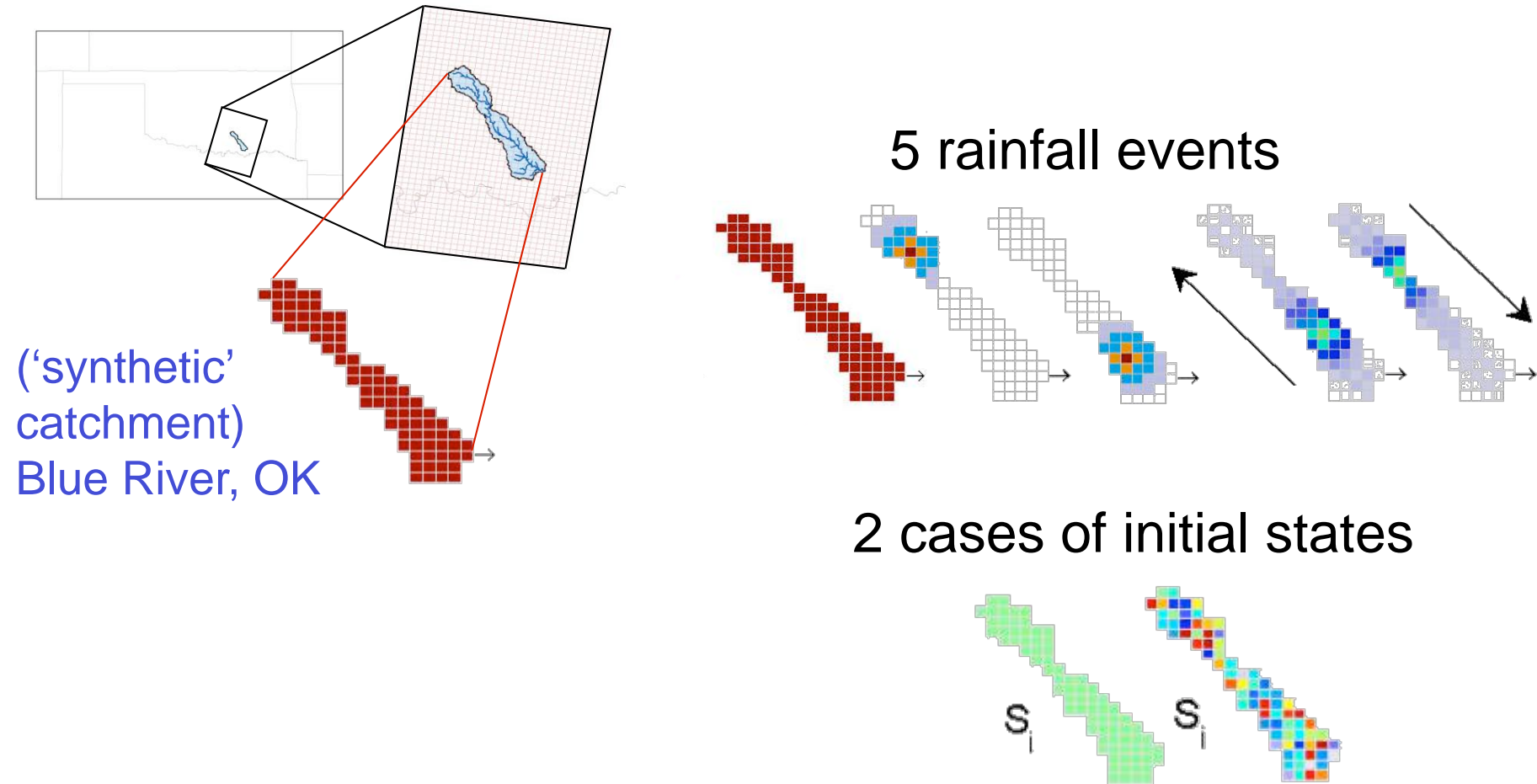
These results have important consequences for model calibration and evaluation

- ⇒ Parametric control **varies significantly**, though is traditionally assumed constant across watersheds **and** time periods
- ⇒ Greater **model complexity might be justified** for flexibility across watersheds, contrary to past assertions
- ⇒ **Aggregation is evil!**

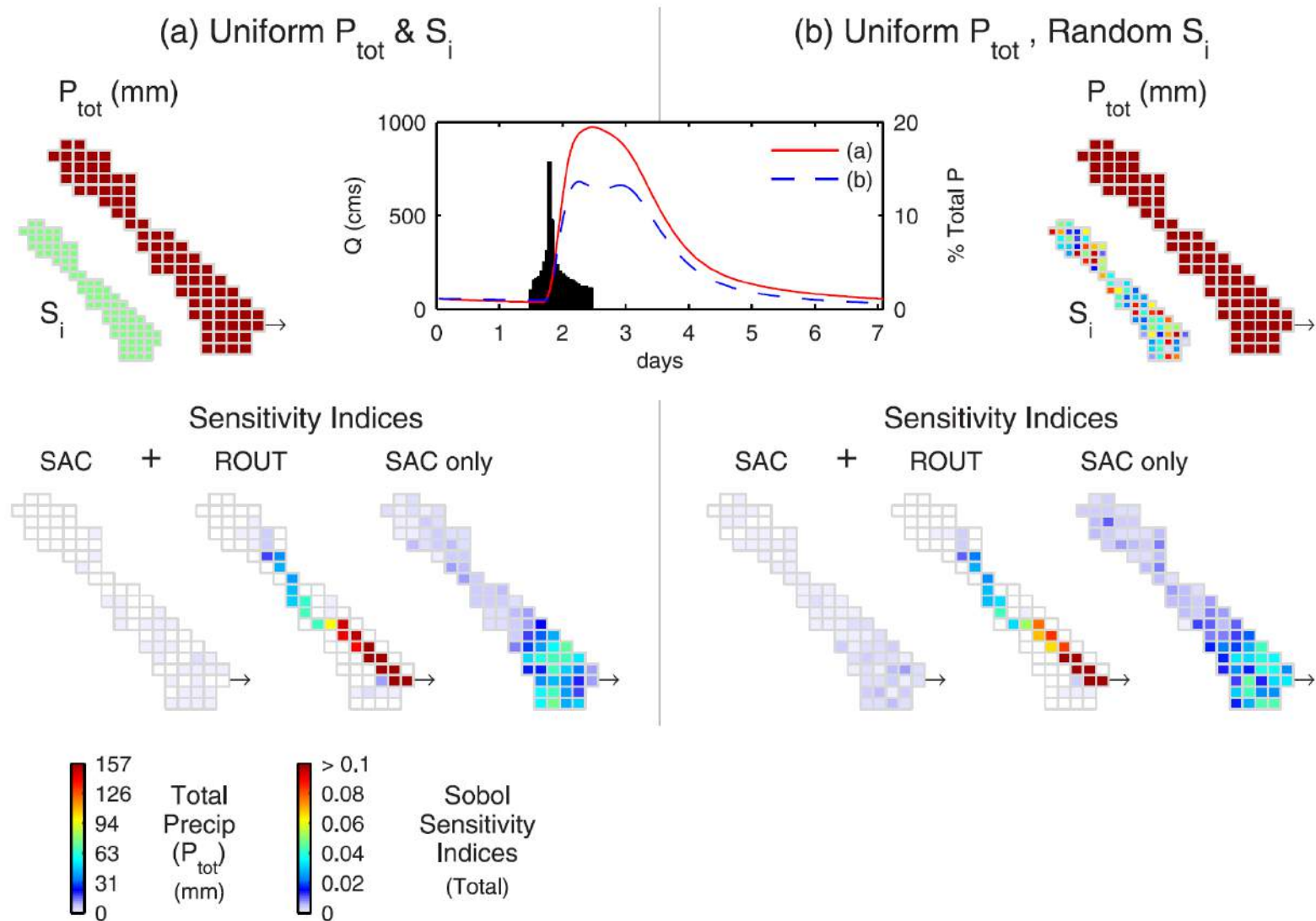
Methods for evaluation and identification that ignore model behavior or assume static behavior are ill-formulated and might bias results!

IDENTIFYING SPATIALLY DISTRIBUTED MODELS

We created a series of experiments to test the relationship between model forcing and its behavior



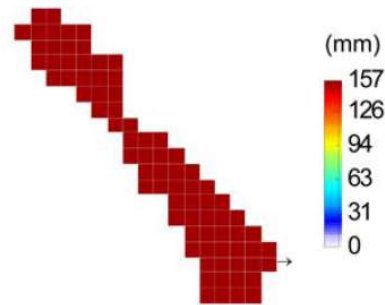
Uniform rainfall does not provide information about the upper part of the catchment



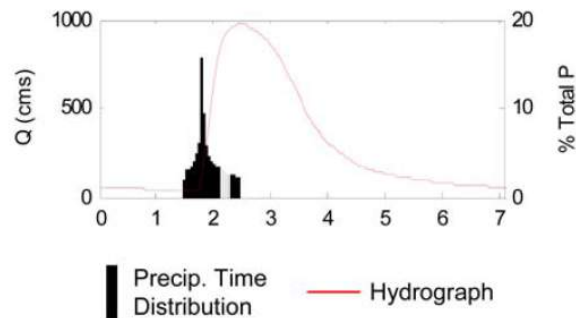
Vertical and horizontal sensitivity changes with objective function chosen

Experimental Setup:

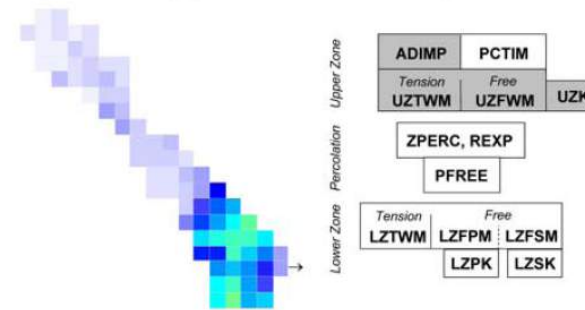
(a) Uniform Precipitation Distribution



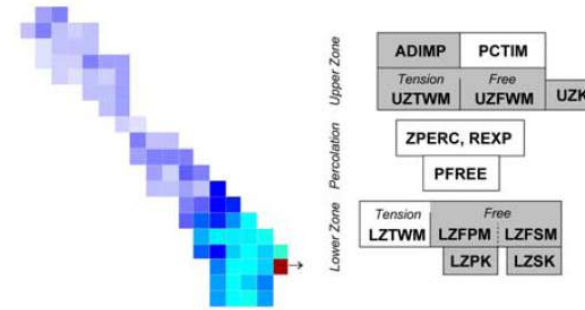
(b) Hyetograph and Hydrograph



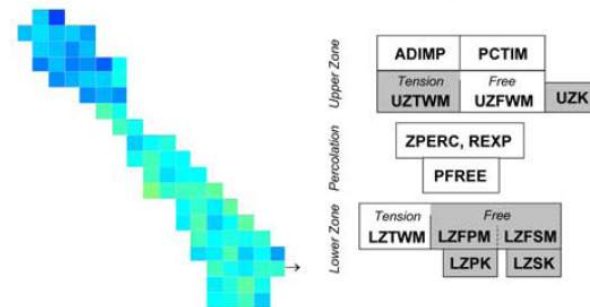
(c) RMSE Sensitivity



(d) TRMSE Sensitivity



(e) ROCE Sensitivity



Spatially distributed model identification strategies need to be dynamic to use information well!

⇒ Information content of streamflow data is **dynamic** and mainly **controlled by precipitation** (near surface)

Existing calibration approaches (e.g. multipliers) do not account for dynamically varying information in streamflow data → thus add bias to parameters!

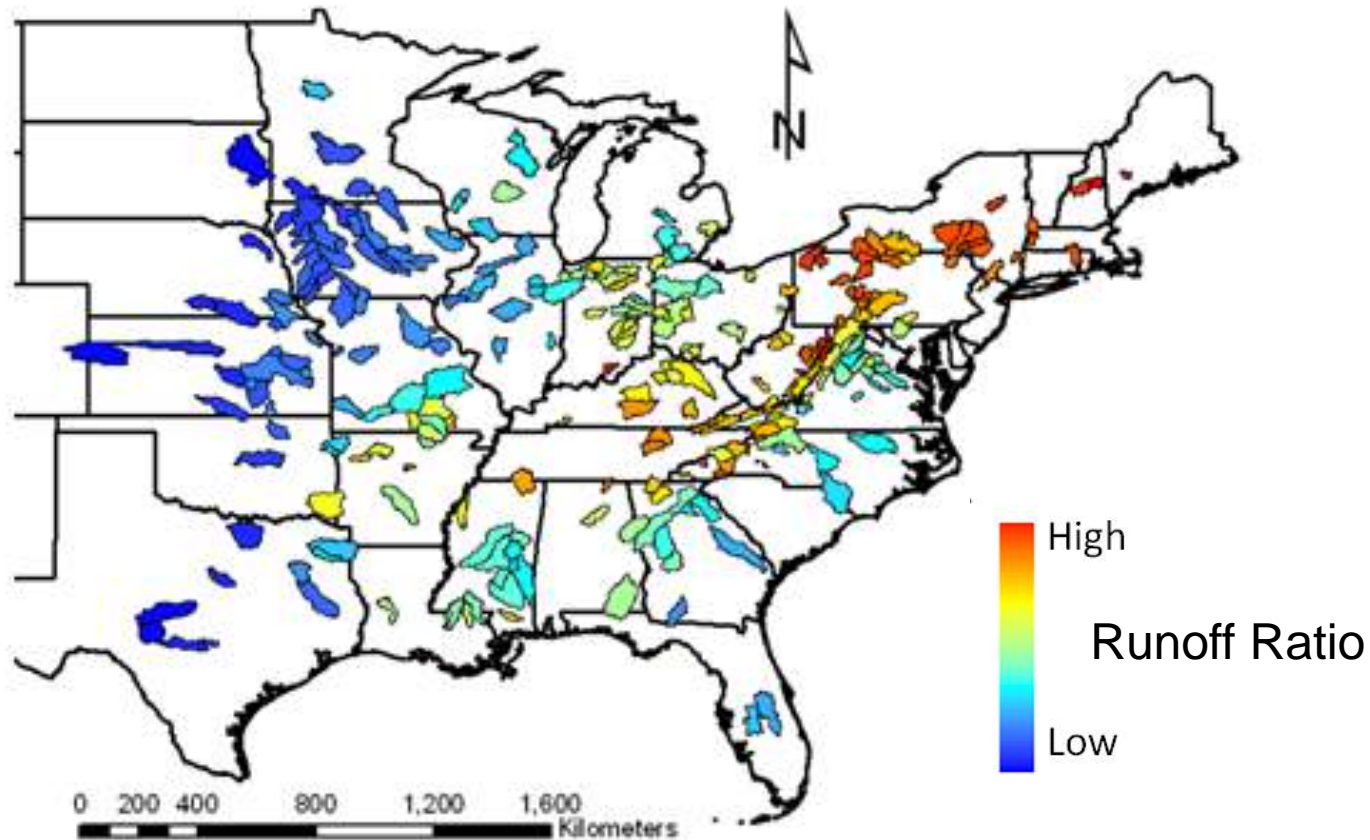
⇒ The value of streamflow data extends only into **portion of the watershed** upstream from the gauge

This needs to be incorporated into observational network design to maximize the value of streamflow observations and to provide information everywhere!

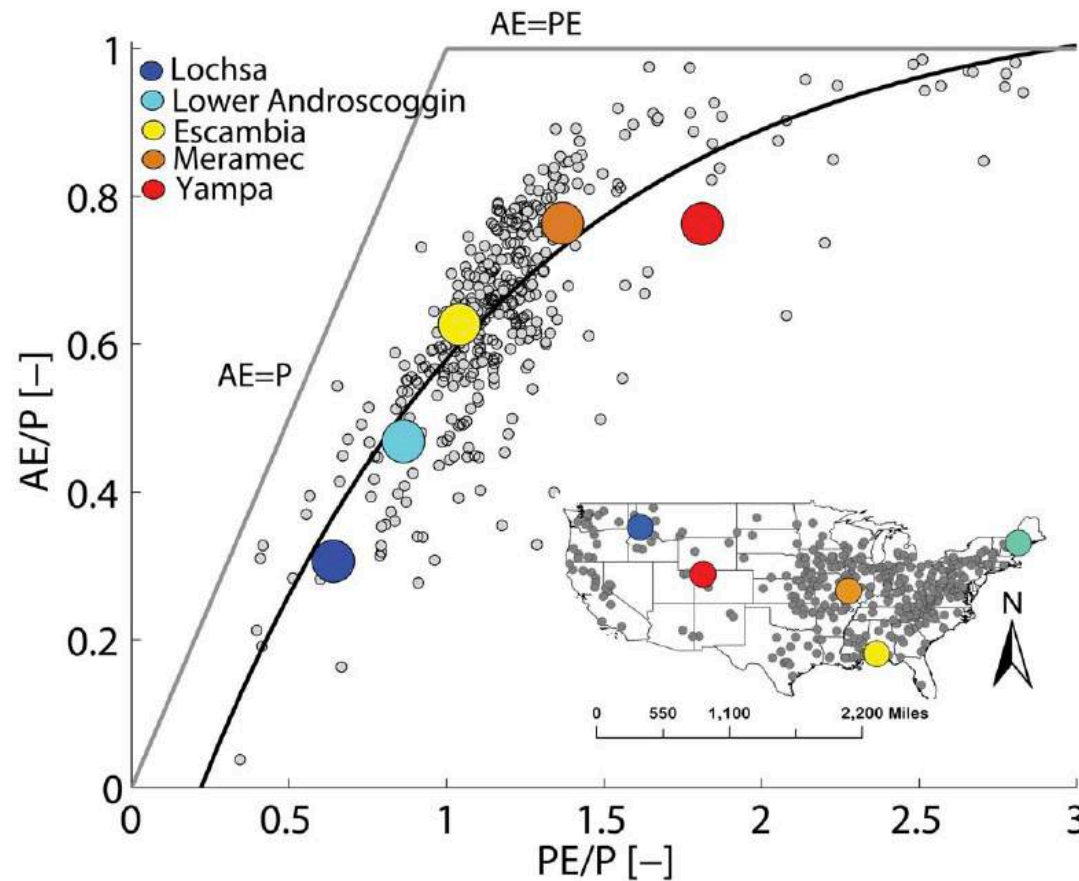
How can we assess models without local historical observations of streamflow?

PREDICTION OF UNGAUGED LOCATIONS AND OF CHANGE IMPACTS

We can assess signatures for a large number of catchments, e.g. regarding how catchments partition rainfall into runoff and evapotranspiration

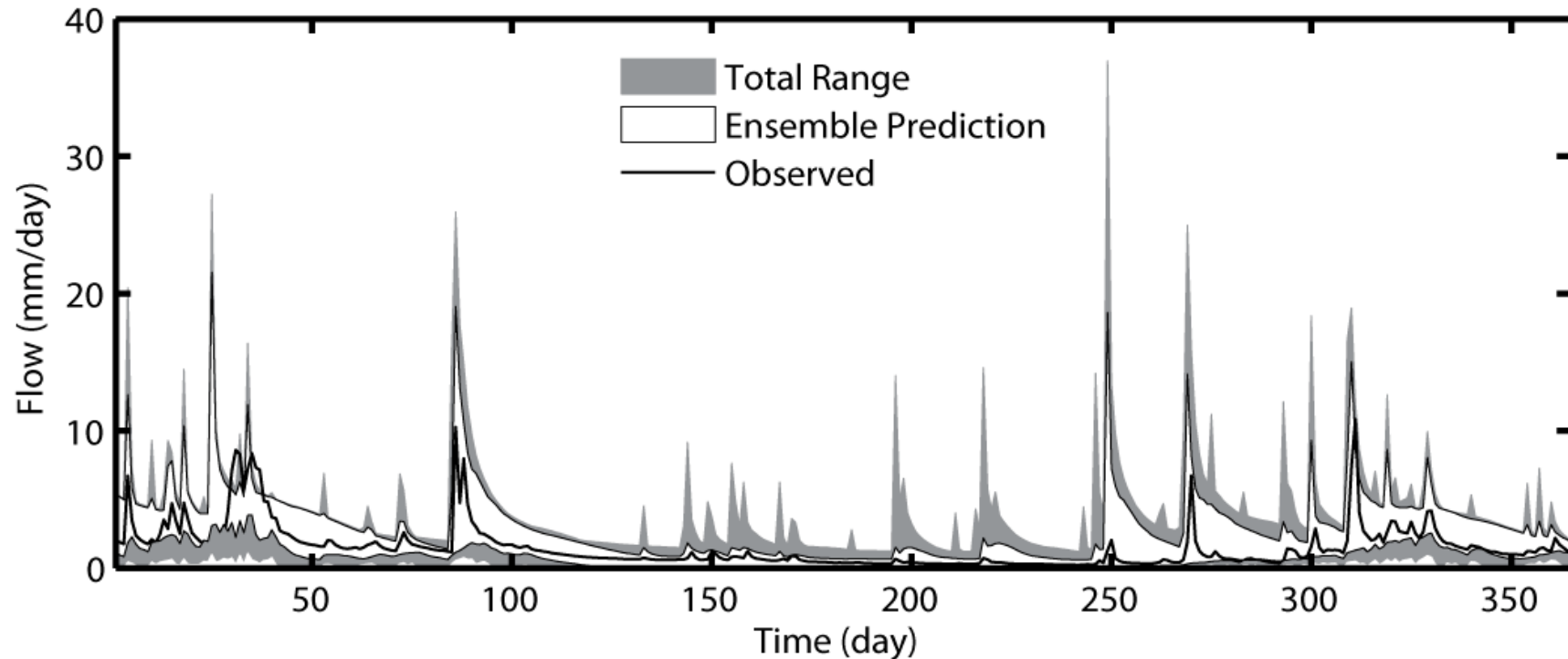


We can then build a model of this spatial variability



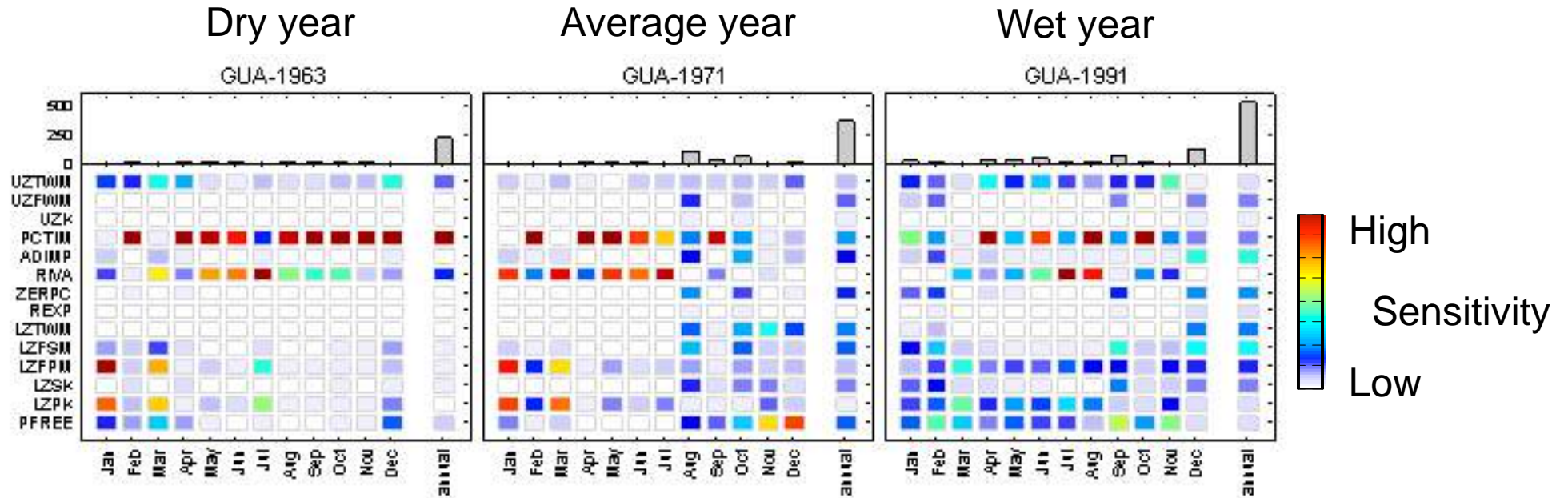
For example using the empirical model by Schreiber to estimate runoff ratio based on climate alone (PET and P).

We can then use this knowledge to reduce the uncertainty in PUB (and change projections) by constraining/conditioning ensemble predictions of watershed models!!!



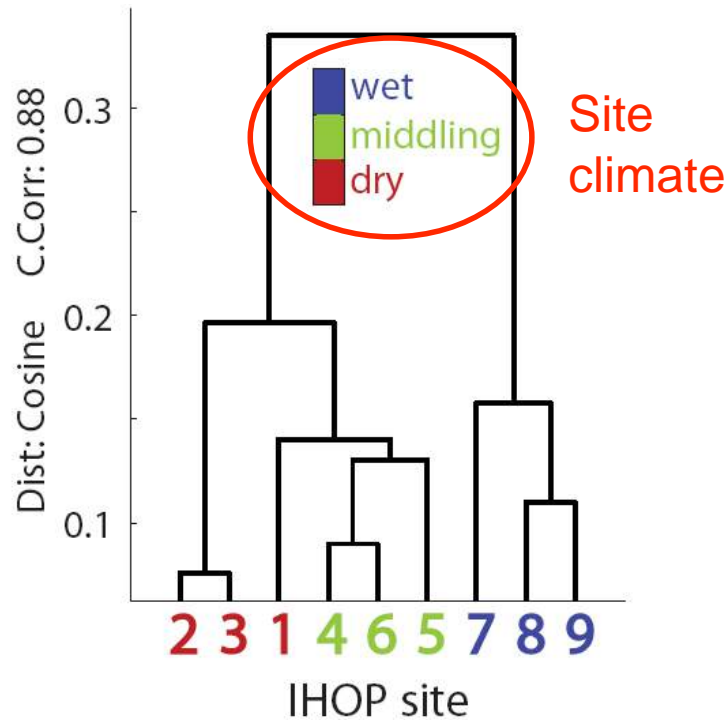
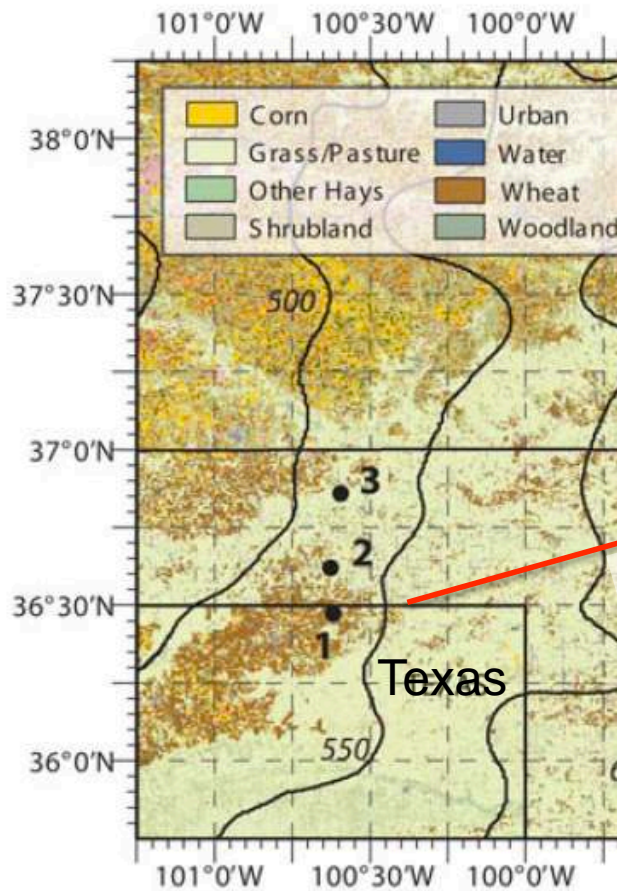
This approach is complementary to other strategies of deriving PUB!

Sensitivity also varies in time within the same catchment

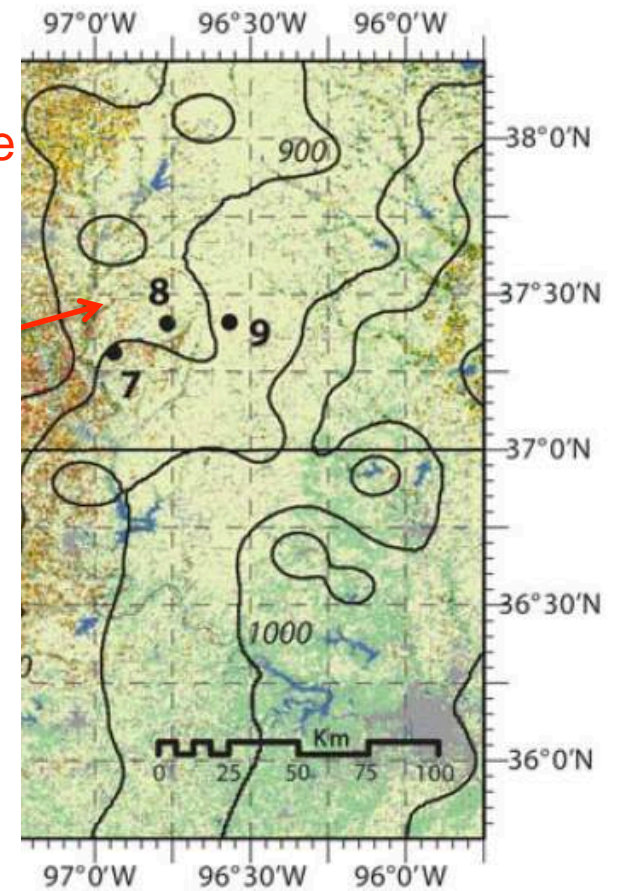


This suggests that model behavior and hence model parameters also vary in time!

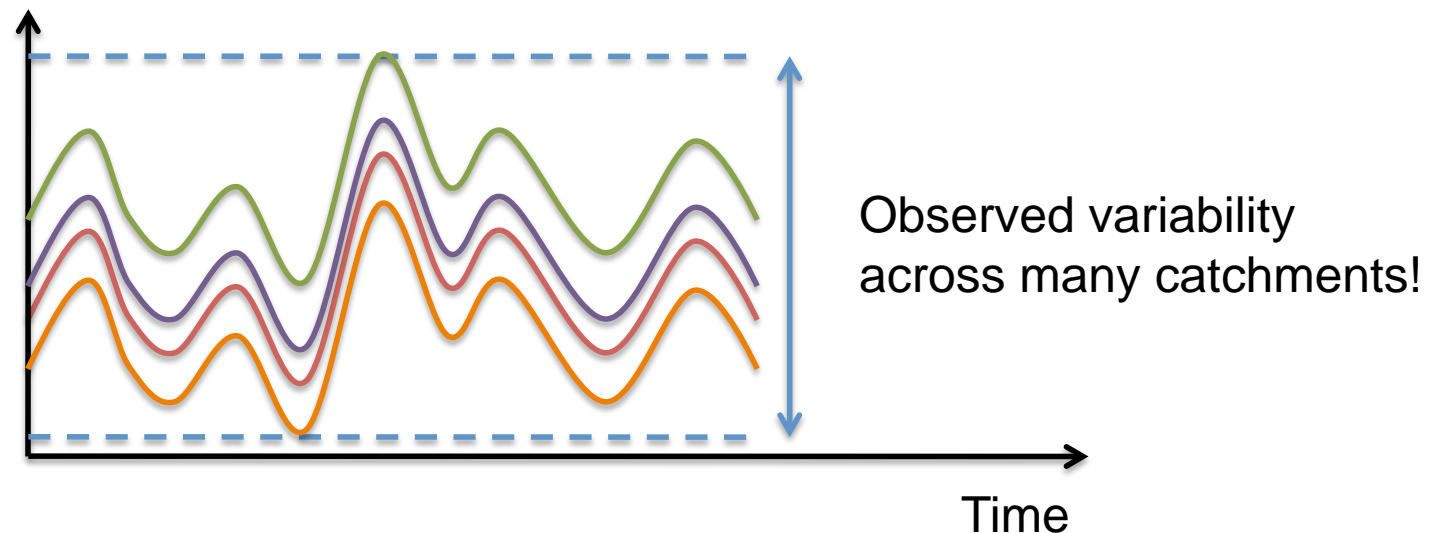
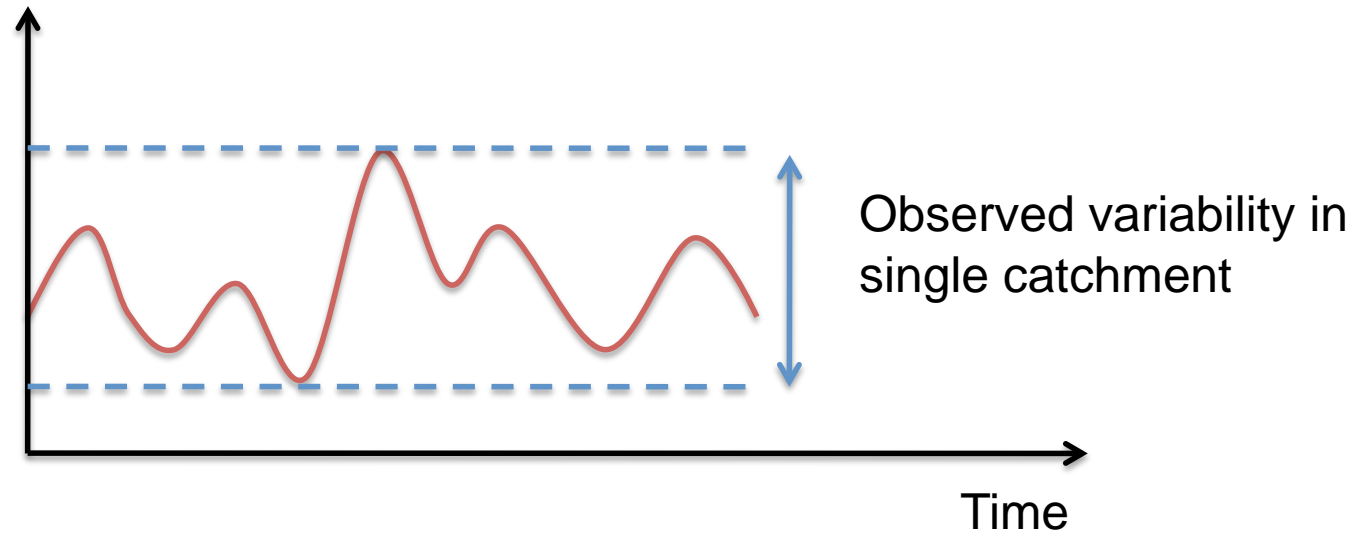
We found further evidence of this climatic control on parameters



Behavioral parameters of a land surface model cluster with climate of site

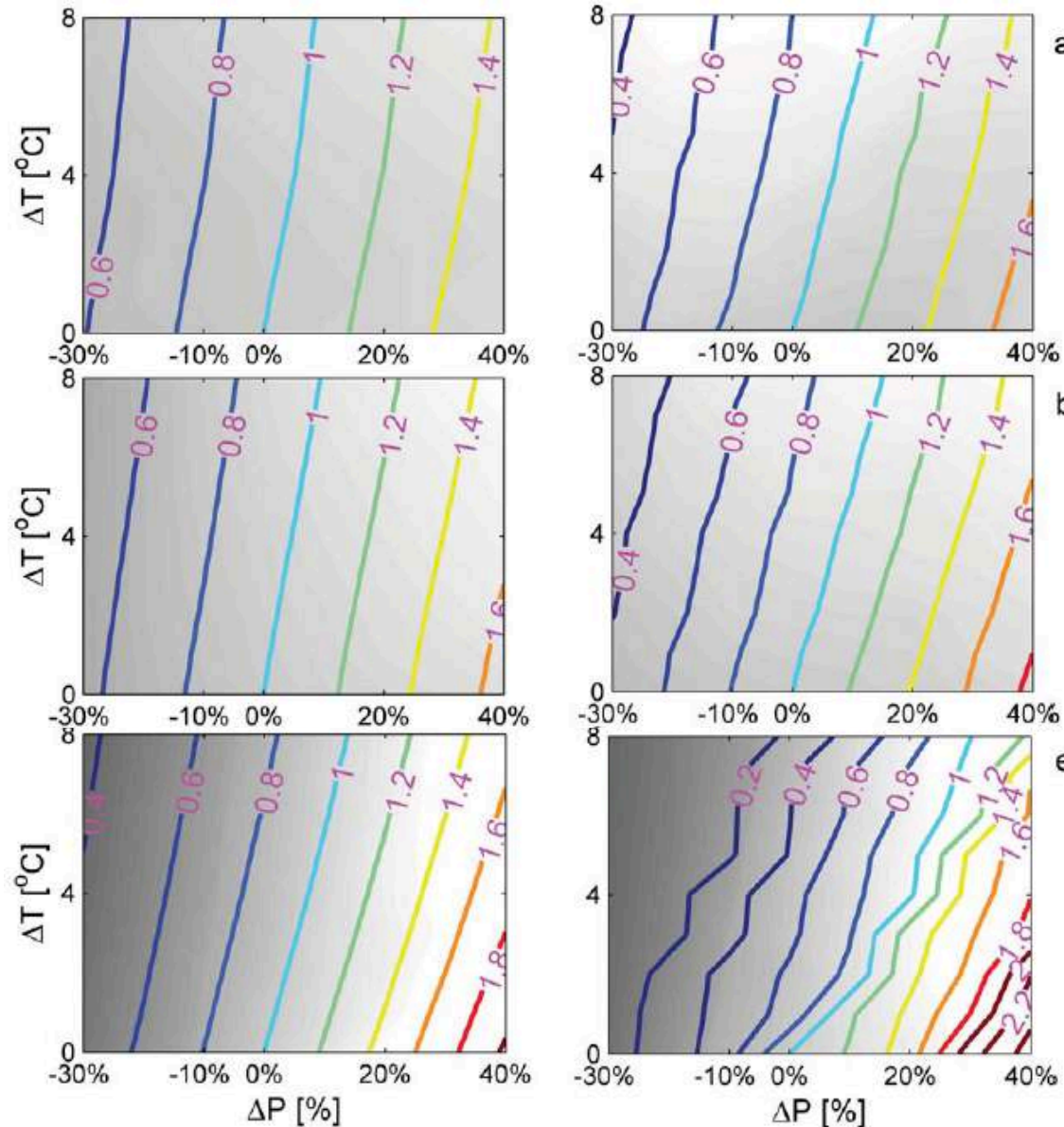


The observed historical variability in hydrologic variables at one place is often limited, and hence our ability to know a catchment's response



We can also do this assuming a temporal gradient at the same location, i.e. we can trade space for time

Increasingly Drier



In summary, there is a need to re-assess how we identify and evaluate models for change impact projections/predictions