

# Understanding Limitations in Environmental Models through Diagnostic Analysis in the Context of Global Change

**Thorsten Wagener**

Civil & Environmental Engineering, Penn State

Institute of Hydrology, University of Freiburg

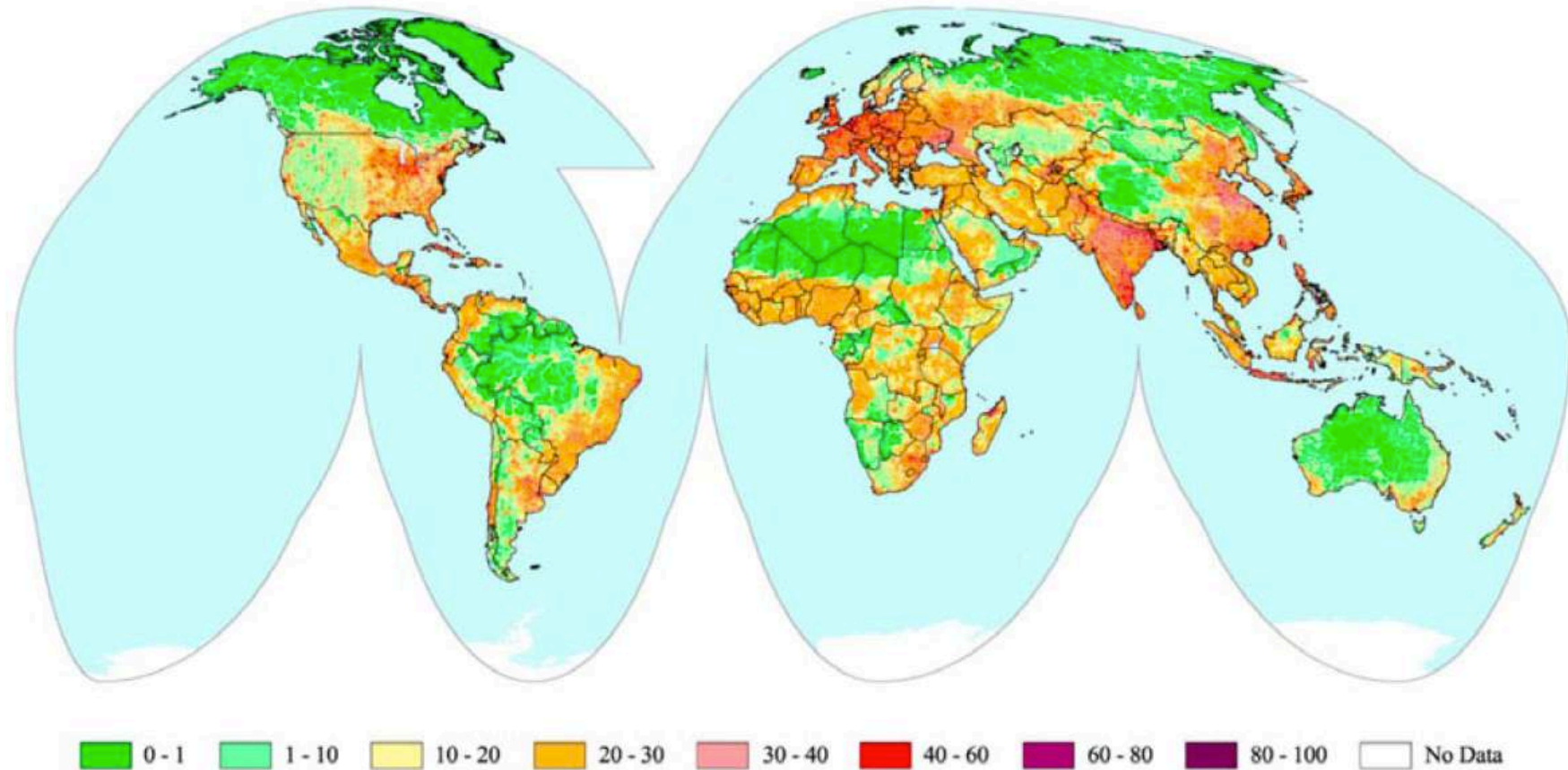
(Soon: Civil Engr., University of Bristol)

There is an increasing need for hydrologic predictions in support of a wide range of water resources services. These predictions need to be available everywhere and represent past, current and potential future conditions.

To achieve this, we need to ensure our models work for the right reasons, and use available data optimally.

Wagner, T., M. Sivapalan, P. A. Troch, B. L. McGlynn, C. J. Harman, H. V. Gupta, P. Kumar, P. S. C. Rao, N. B. Basu, and J. S. Wilson (2010), The future of hydrology: An evolving science for a changing world, *Water Resour. Res.*, 46, W05301, doi:10.1029/2009WR008906.

Such predictions are required in the context of a changing world (for which models have to account)



Human footprint index analysis showing that over 80% of the land surface is impacted by human activity (Sanderson et al. 2002, *BioScience*)

## Current

## Future

Humans are external to the hydrologic system

Humans are intrinsic to the hydrologic system, both as agents of change and as beneficiaries of ecosystem services

Assumption of stationarity: past is a guide to the future

Nonstationary world: past is no longer a sufficient guide to the future, expected variability could be outside the range of observed variability

Predicting response, assuming fixed system characteristics: boundary value problem with prescribed fixed topography, soils, vegetation, climate

Both system and response evolve: no longer a boundary value problem, boundary conditions and interfaces themselves evolve and are coupled. Becomes a complex adaptive system

Learning from studying individual places (often pristine experimental catchments) to extrapolate or upscale to other places

Comparative hydrology: learning from individual places embedded along gradients (e.g. changing climate, human imprint) and across spatial scales

Hydrologists as analysts of individual processes or features at small scales (akin to a microscope) or as synthesists of whole system behavior at large scales (akin to a telescope)

Hydrologists as both analysts *and* synthesists (akin to the *macroscope*) studying the coupled system across a range of time and space scales

Observations to characterize input-output behavior in individual (mostly) pristine places

Observations to track the evolution of both structure and response in coupled systems and subsystems

Observe and analyze pristine places and extrapolate to make predictions of human impacts

Observe and analyze real places where humans live and interact with the hydrologic system at range of scales

Model predictions derive credibility by reproducing historical observations

Model predictions derive credibility via more in-depth diagnostic evaluation of model consistency with underlying system and testing of behavior outside of observed range

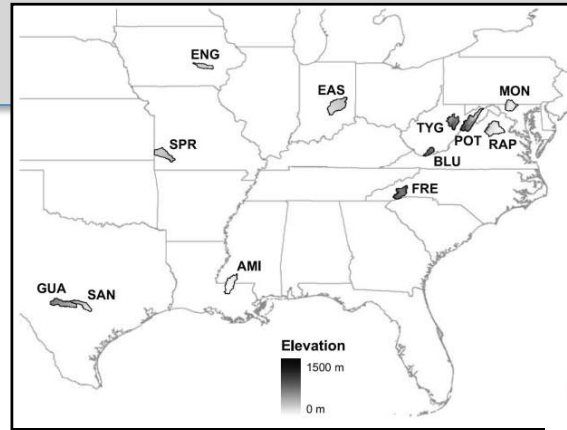
Observation, prediction (modeling) and management are separate exercises (without feedbacks!)

Real-time learning: observations (sensing, including participatory human sensing), modeling and management are interactive exercises with feedbacks and updating

# How can we understand how and why models work?



Diagnostic model evaluation

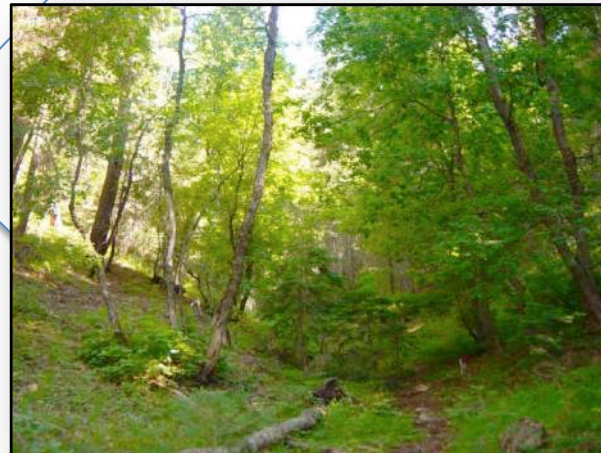


Understanding model behavior across environmental gradients



Identifying spatially distributed models

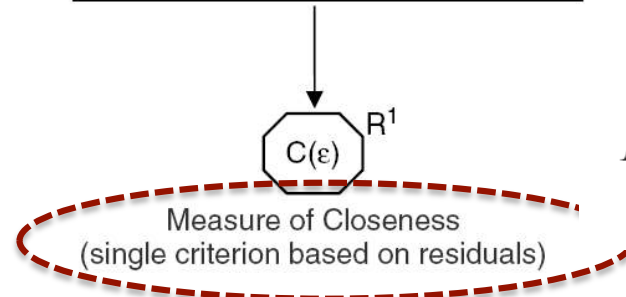
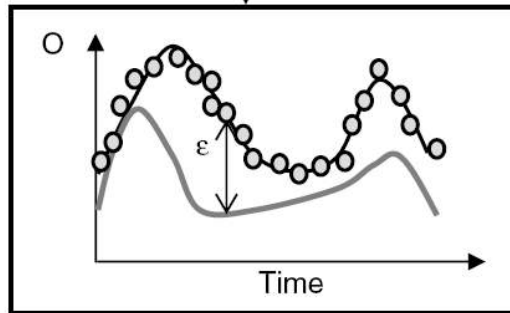
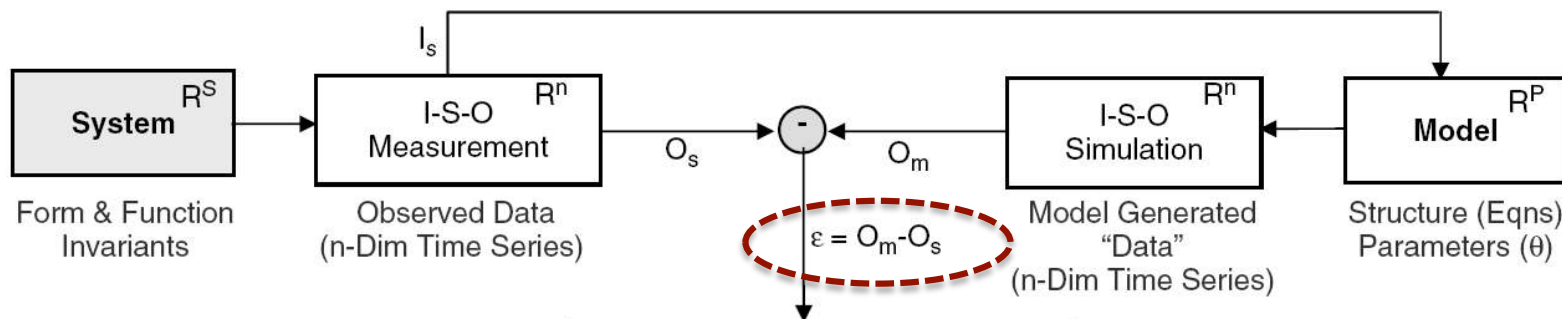
Predictions / Projections



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# DIAGNOSTIC MODEL EVALUATION

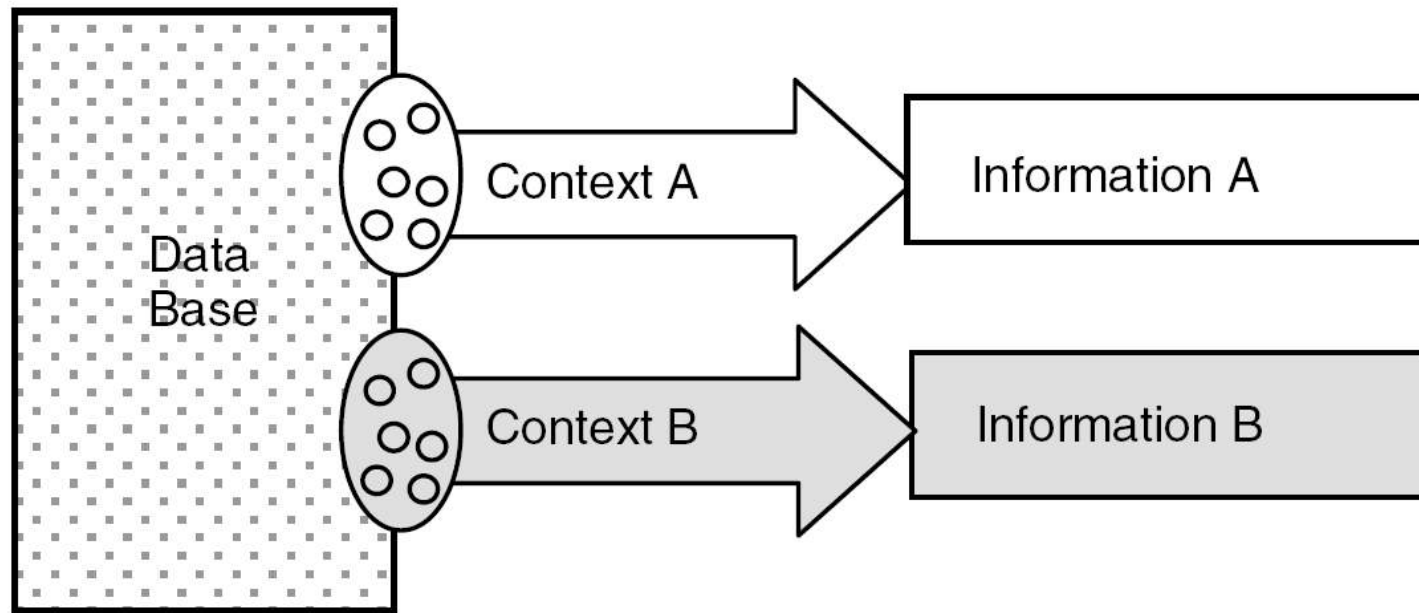
# Most (all?) models require some degree of calibration to observed data



e.g.

$$RMSE = \sqrt{\frac{1}{m} \sum_{t=1}^m (Q_{s,t} - Q_{o,t})^2}$$

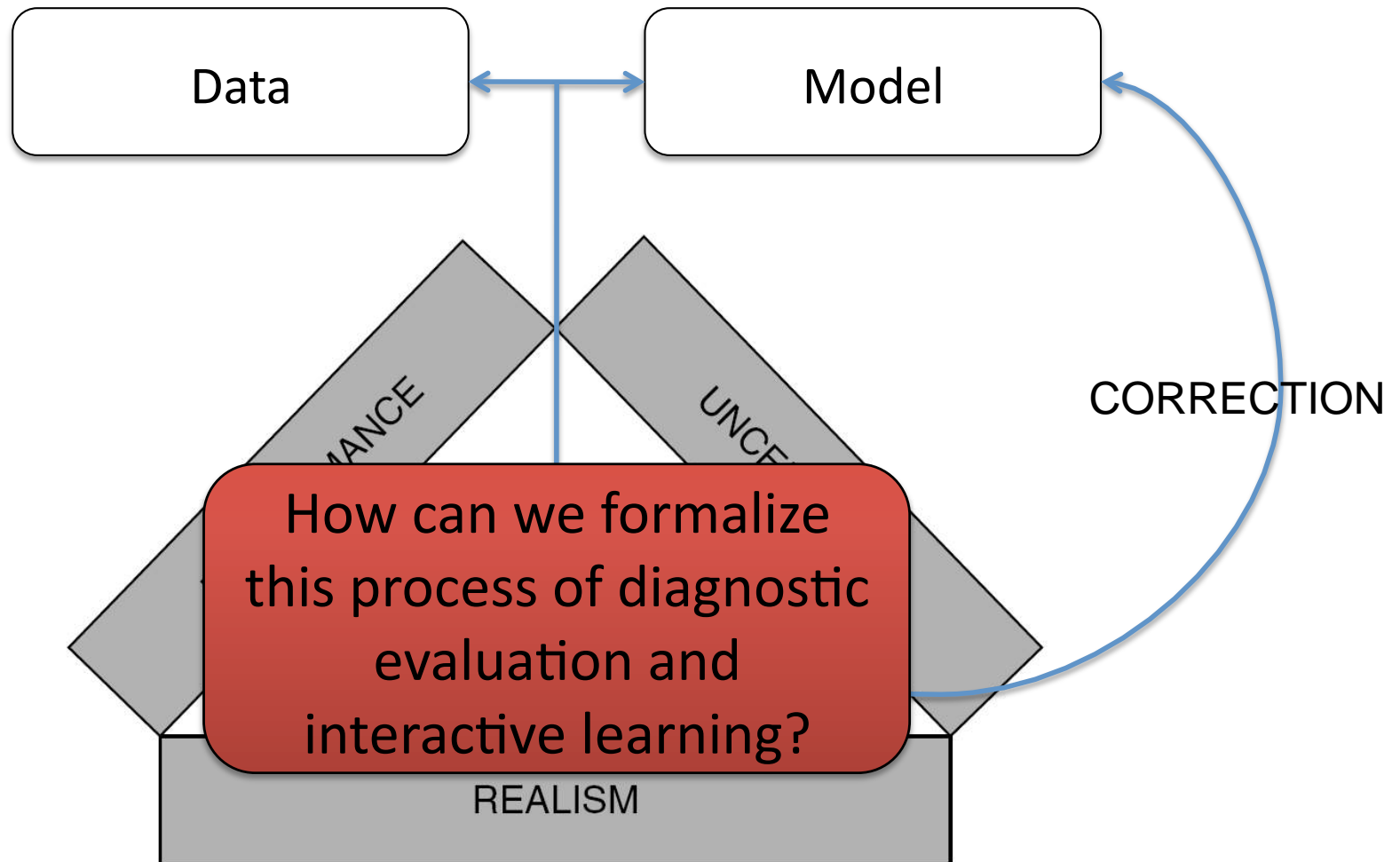
The measure of closeness (objective function) extracts information about model performance



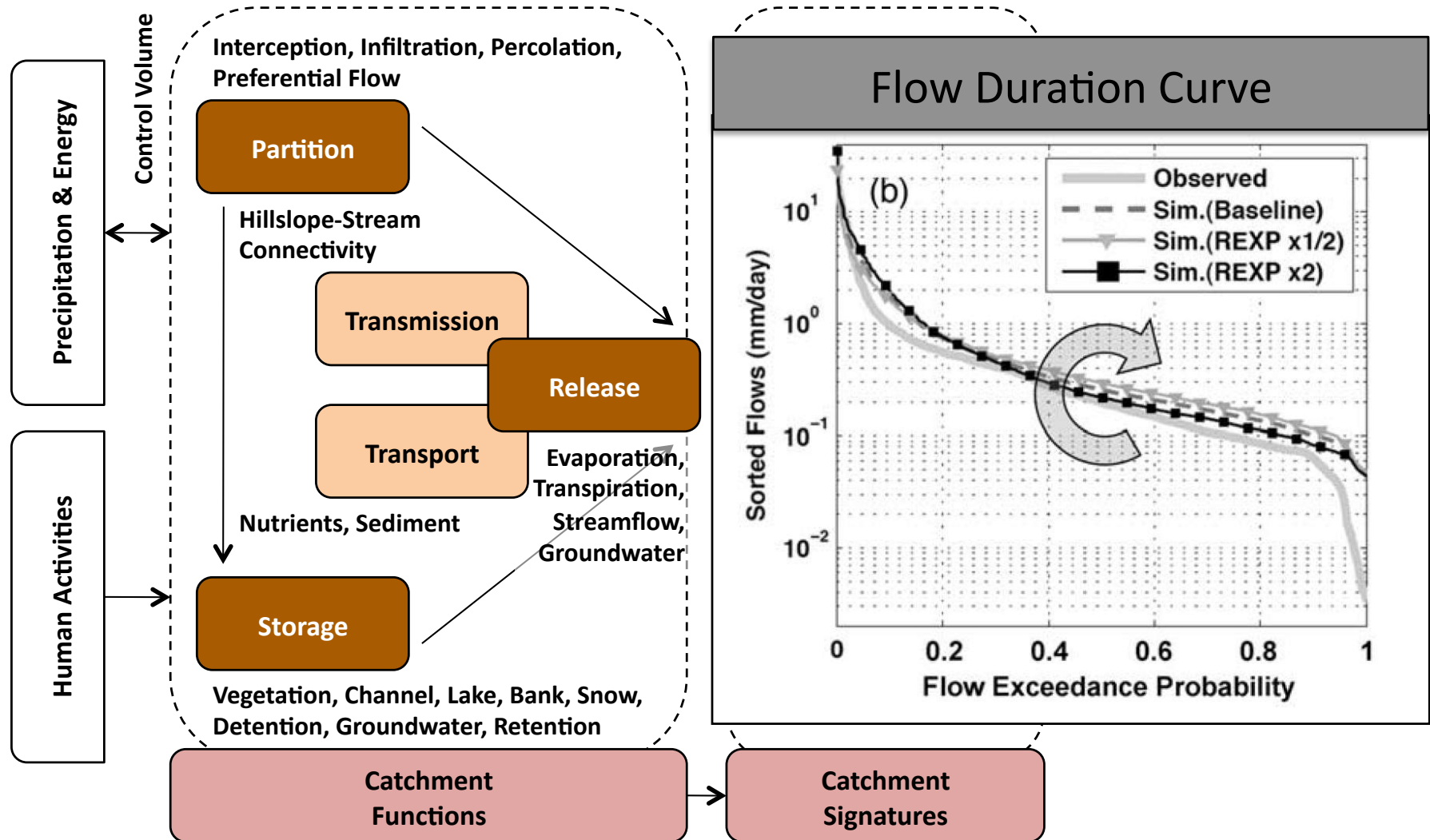
However, traditional statistical measures extract little information useful for diagnostic evaluation!



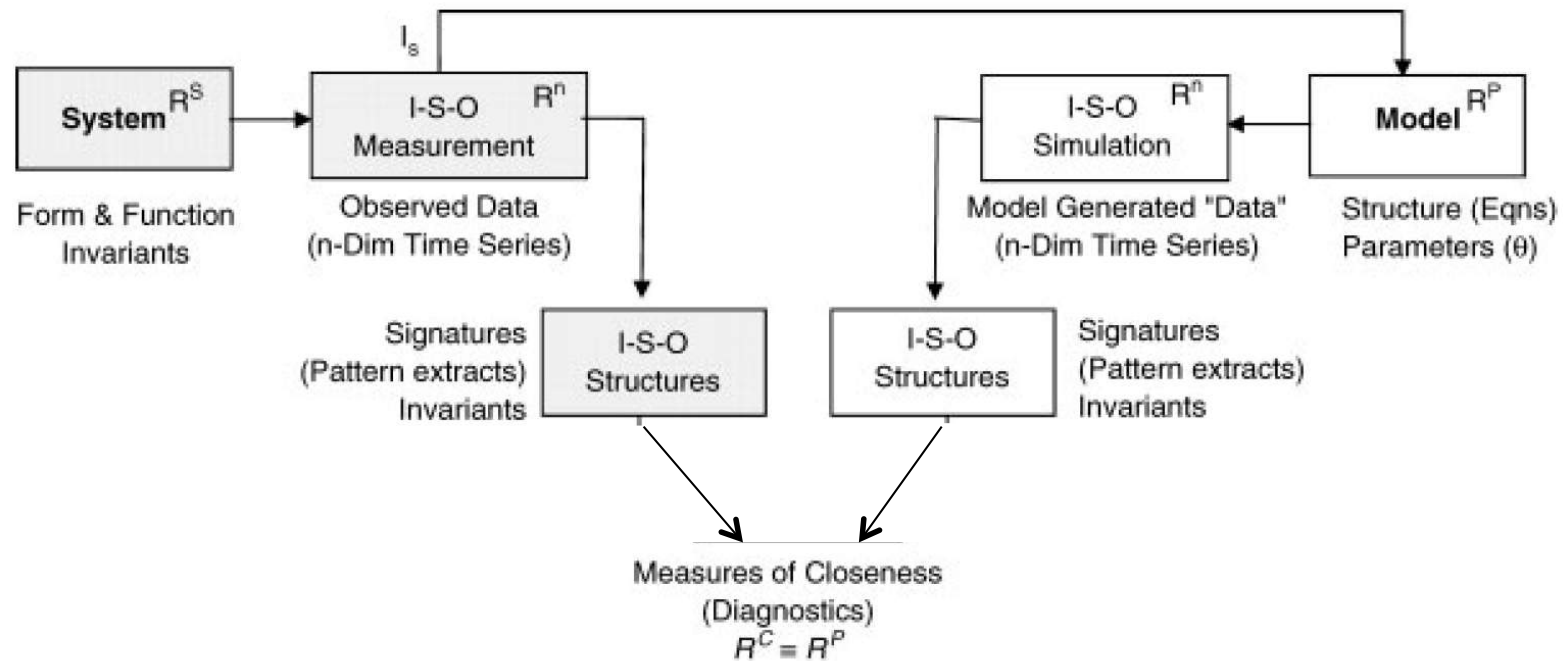
# Evaluation of environmental models has at least 3 dimensions



# [1] We can use signatures to provide insight into how the system functions

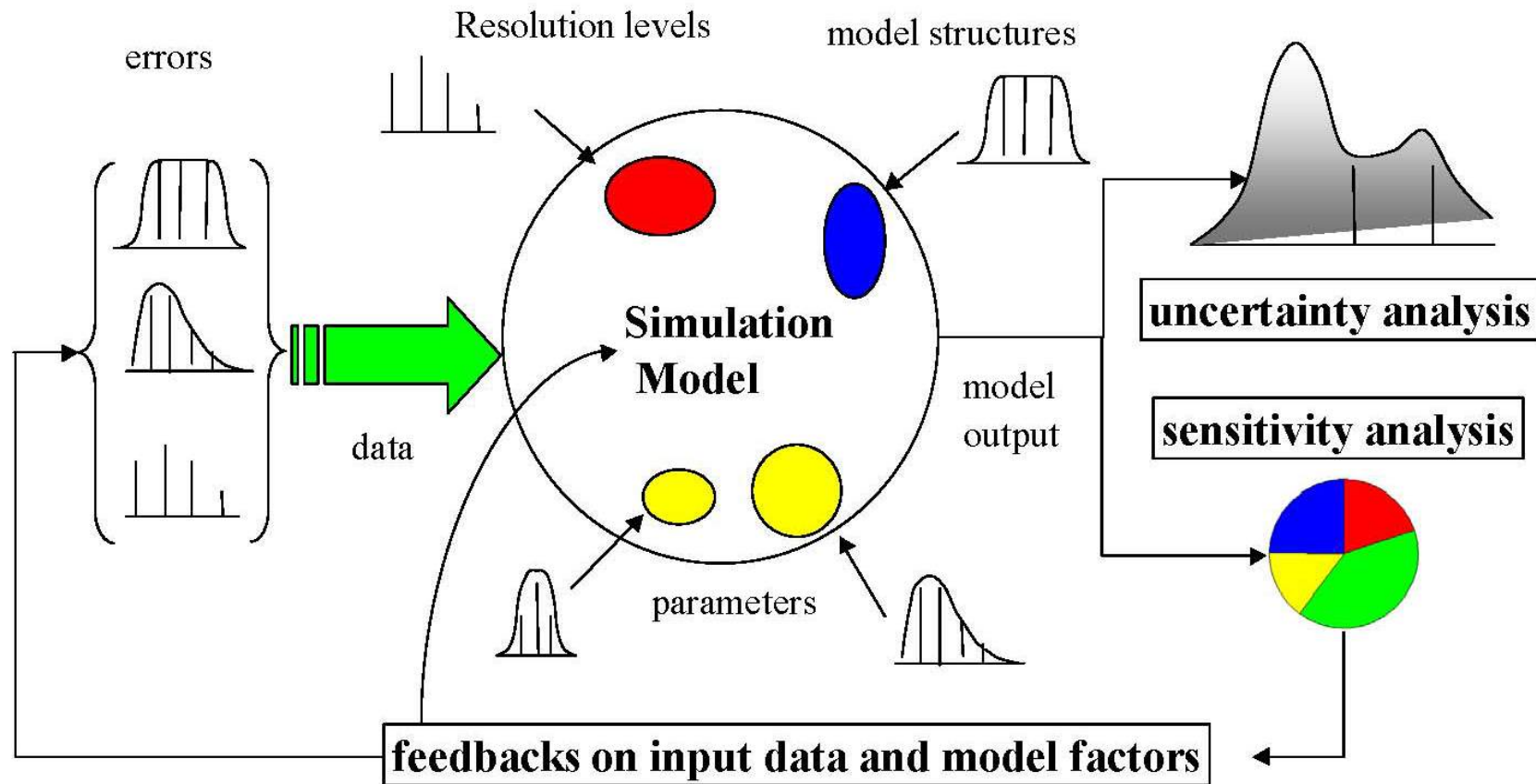


# Comparing signatures and understanding how the model produces them can be diagnostic...



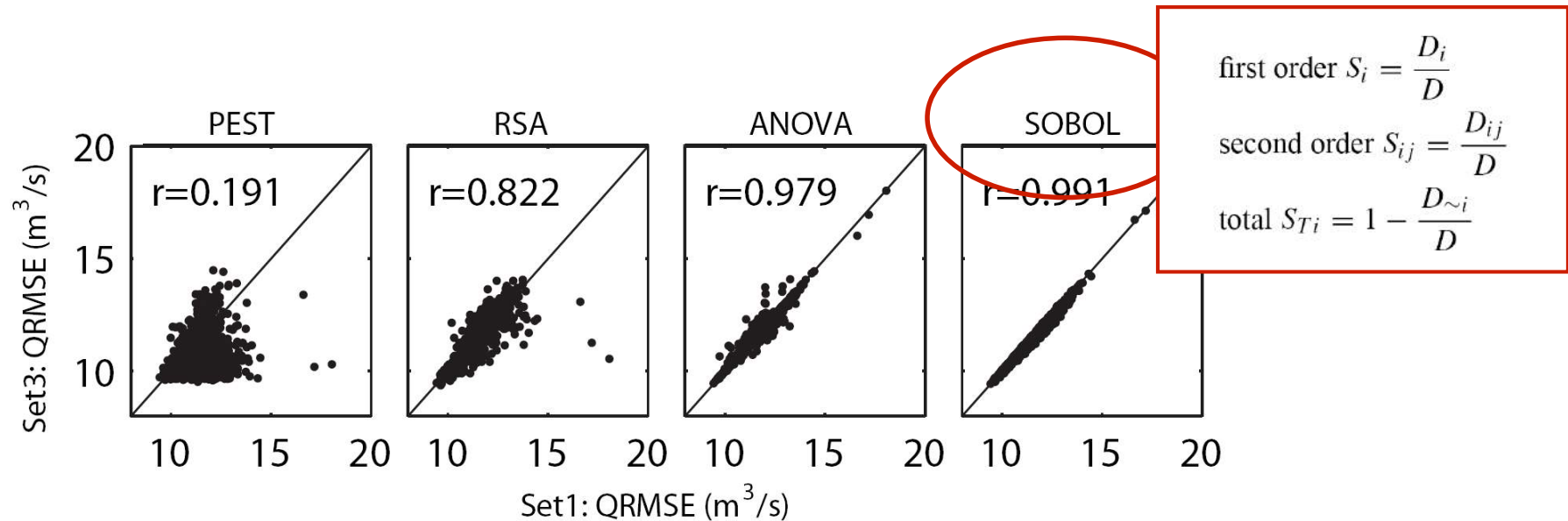
...if we can assess whether the model reproduces the functional behavior of the system in a way consistent with our theory

[2] We can use sensitivity analysis to understand how the model produces these signatures (= realism)

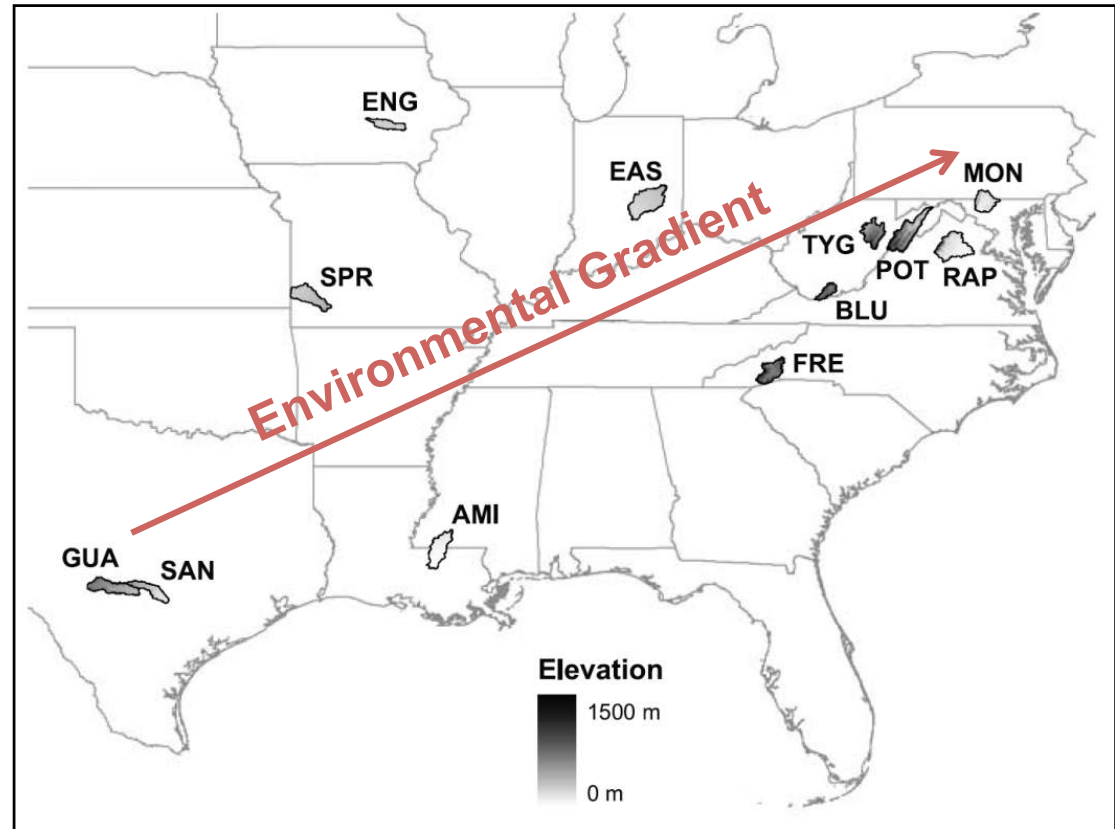


Both sensitivity analysis and signatures form elements of a diagnostic model evaluation strategy

We found the global variance-based approach by Sobol to be robust and effective



We compared four different sensitivity analysis approaches. The perfect approach would result in all points falling on a line.



# UNDERSTANDING MODEL BEHAVIOR ACROSS ENVIRONMENTAL GRADIENTS

# The model evaluated is a popular lumped watershed model, i.e. the Sacramento model

