Dynamic process connectivity for model diagnostics, evaluation, and intercomparison

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Background

Single-objective model evaluation does not tell the whole story of model performance. It tells us how wrong we are, but not why

Understanding interaction of processes when modeling hydrologic systems can be used to explore the "why" of model performance

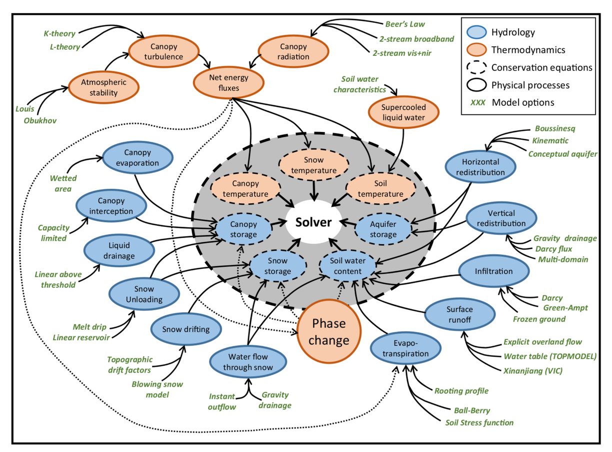
Information theory gives a systematic way to compare various processes across scales and units

SUMMA

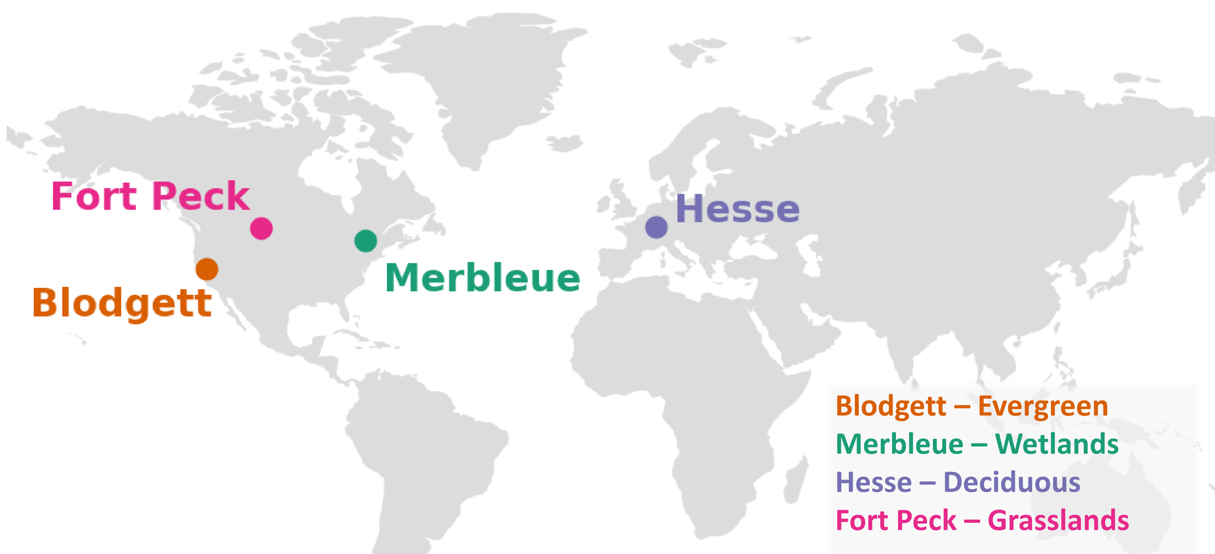
Framework for implementing hydrologic models

User can choose spatial discretization and flux parameterizations

Ensembles can be built in a controlled fashion



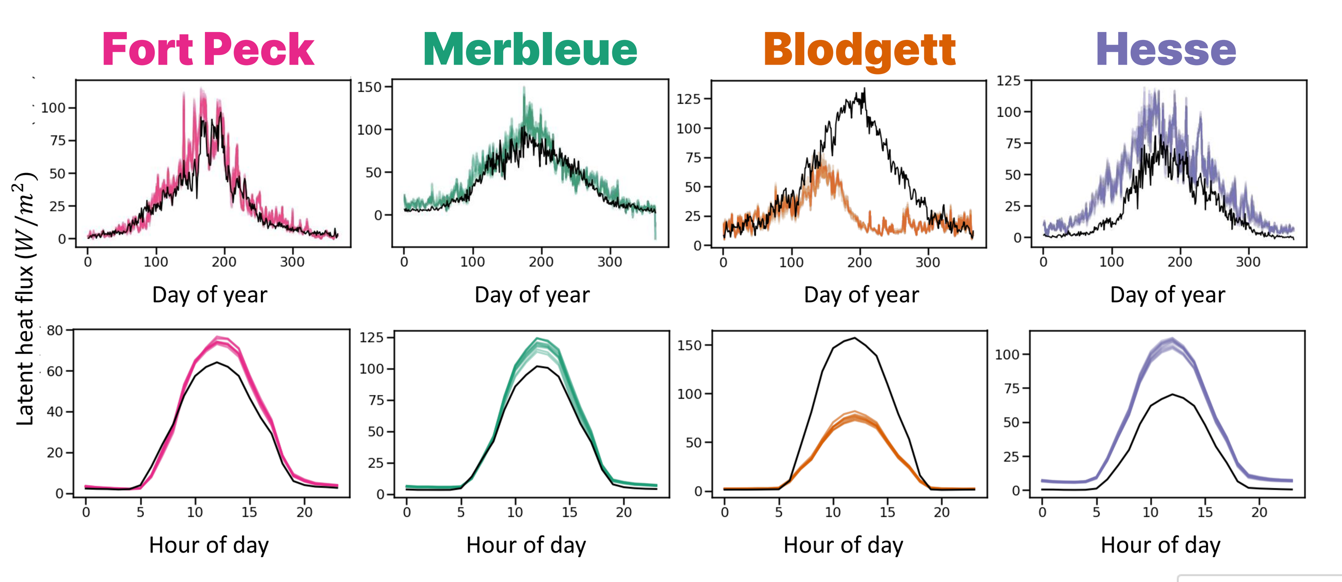
Simulations



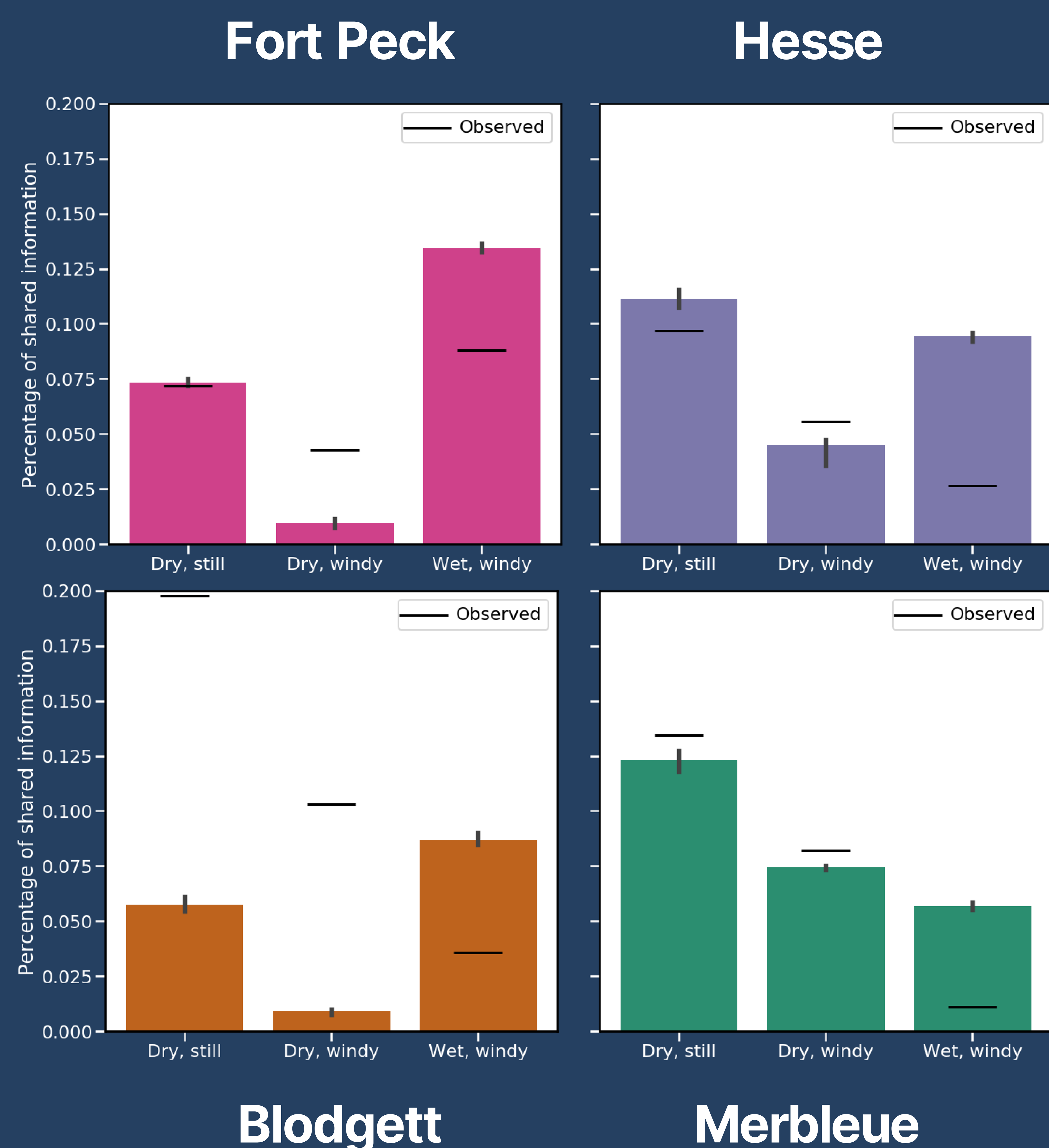
108 Model runs per site:

- 2 parameterizations of vegetation dynamics
- 2 parameterizations of canopy emissivity
- 3 parameterizations of canopy shortwave
- 3 values of soil thermal conductivity
- 3 values of specific heat of vegetation

Latent heat (Q) results



Mismatch in **shared information** between **latent heat** fluxes and **shortwave radiation** indicates poor understanding in **wet, windy conditions**



Computing normalized mutual information between latent heat and shortwave radiation ($NMI(Q;SW)$) for all model instances and the observations we can see whether the modeled interactions are realistic

Evaluation measures

Normilized mutual information (NMI): How much does measurement of one variable reduce prediction error in another?

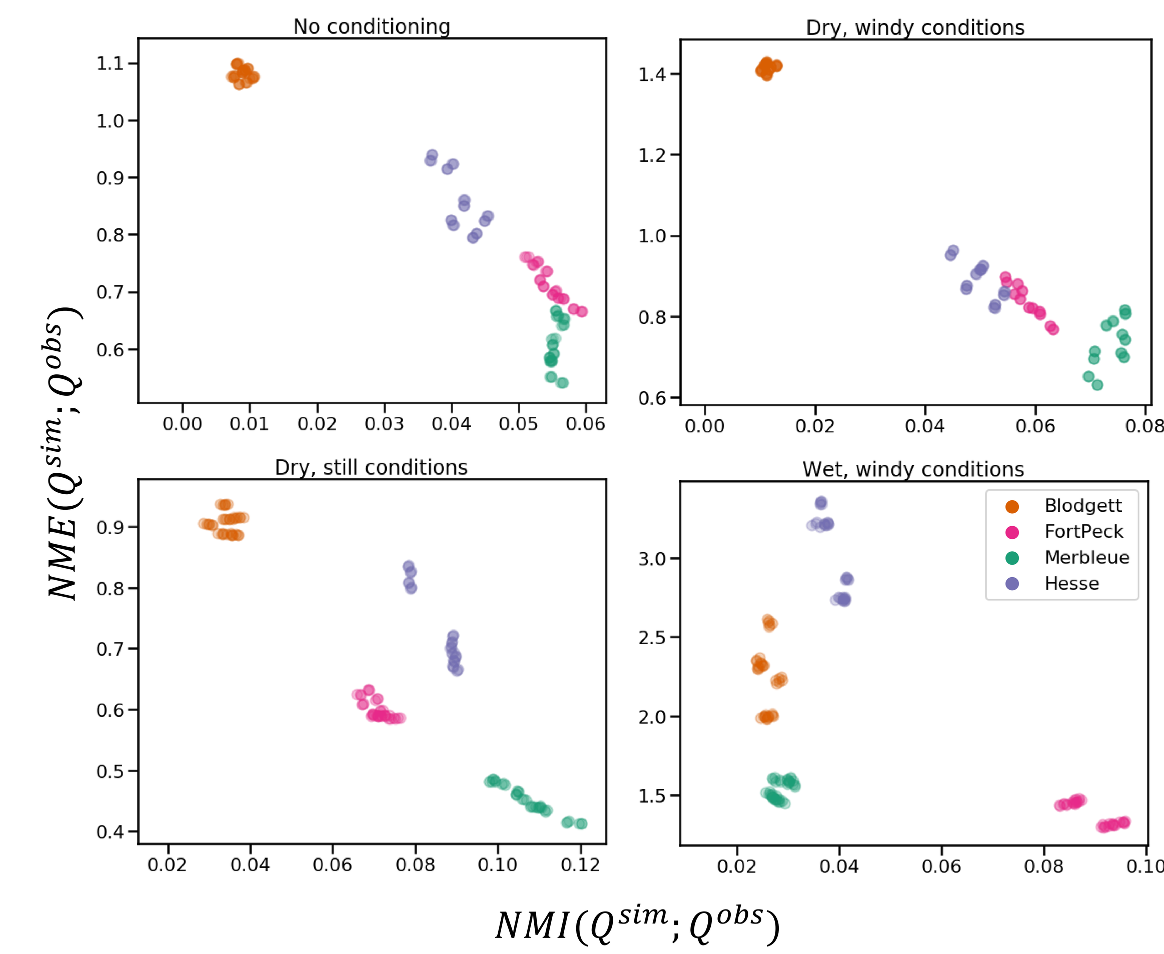
$$NMI = \frac{I(X;Y)}{\min[H(X), H(Y)]}$$

Normalized mean error (NME): How does error in my prediction relate to observation?

$$NME = \frac{\sum |x - y|}{\sum |\bar{y} - y|}$$

NME > 1 indicates that simulation variability is greater than average observed variability

Information versus error



Decreasing error is well correlated with increasing information in all conditions except wet, windy conditions

Departure of trend in wet, windy conditions indicates some process representation deficiency where error is uncorrelated with information

Not enough data to compute wet, still conditions

Discussion

Poor model performance at Blodgett stems from poor usage of both temperature and shortwave information across conditions – most likely bad vegetation parameters

Model performance in wet, windy conditions is poor across sites due to inflated connection with shortwave

Models tend to use shortwave information correctly under dry, still conditions

Further investigation into parameterizations and/or measurements of latent heat in wet, windy conditions should be conducted

Acknowledgements

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