

A parameter sensitivity study of the NCAR-NEON CLM simulation tool in the Southeastern United States

28th Annual CESM Workshop

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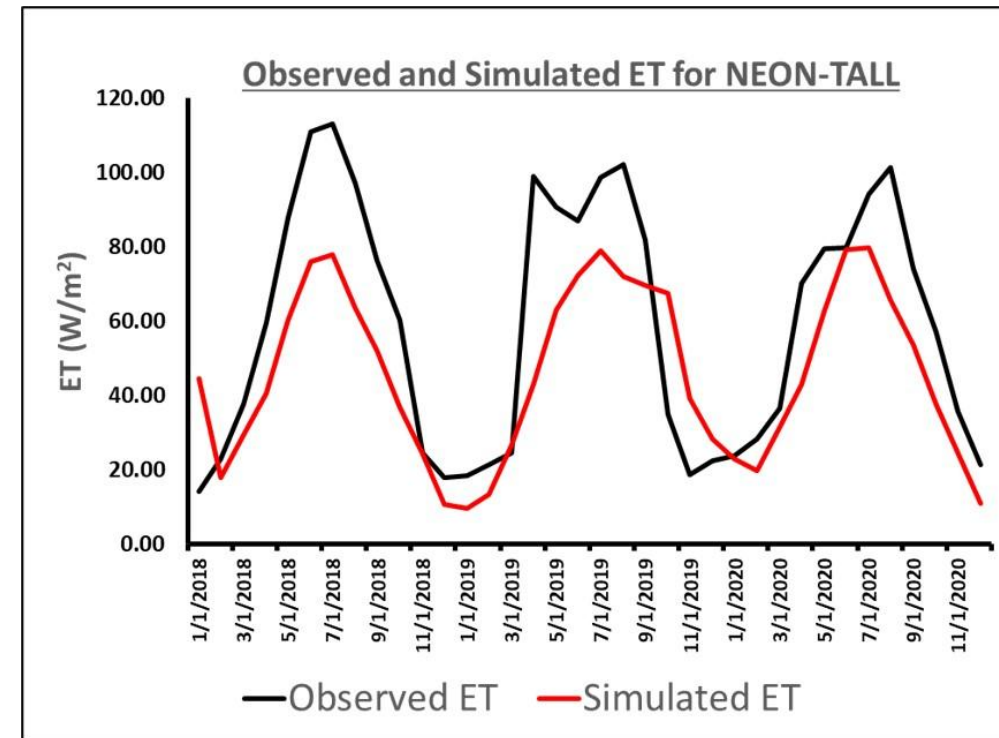
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College of
Forestry, Wildlife
& Environment



INTRODUCTION

- Estimates of change (e.g., ET) are key for understanding the terrestrial hydrological cycle under changing environments.
- Increasing complexity and comprehensiveness of LSMs - more uncertainty
- Significant biases when estimating water-carbon cycle processes



Motivation

❖ Growing Complexity and Uncertainty:

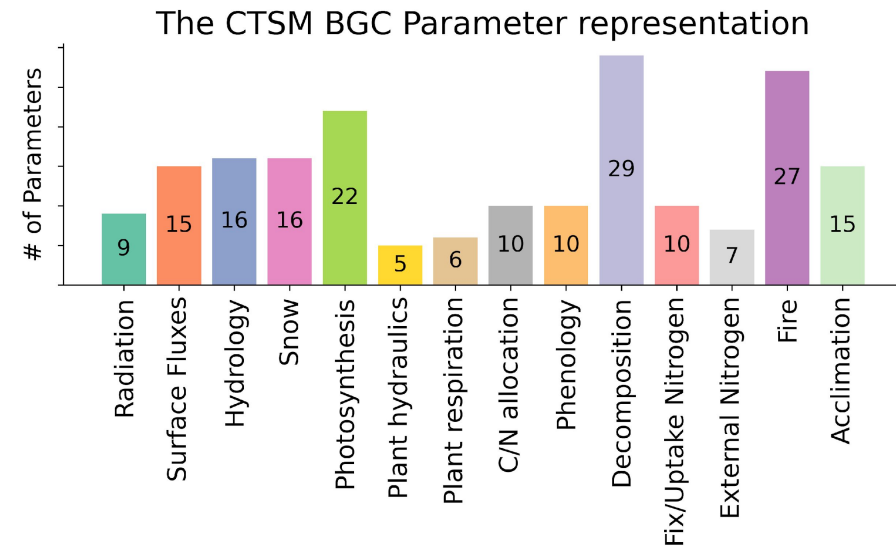
- Land models increasingly complex and comprehensive (CTSM5, includes over 200 parameters)
- Increasing number of uncertain parameters poses challenges in accurate simulations.

❖ Unquantified Parameter Uncertainty:

- The contribution of parameter uncertainty to overall uncertainty is expected to be substantial.
- However, the extent of this uncertainty remains largely unquantified.
- This hinders a comprehensive understanding of the model's reliability.

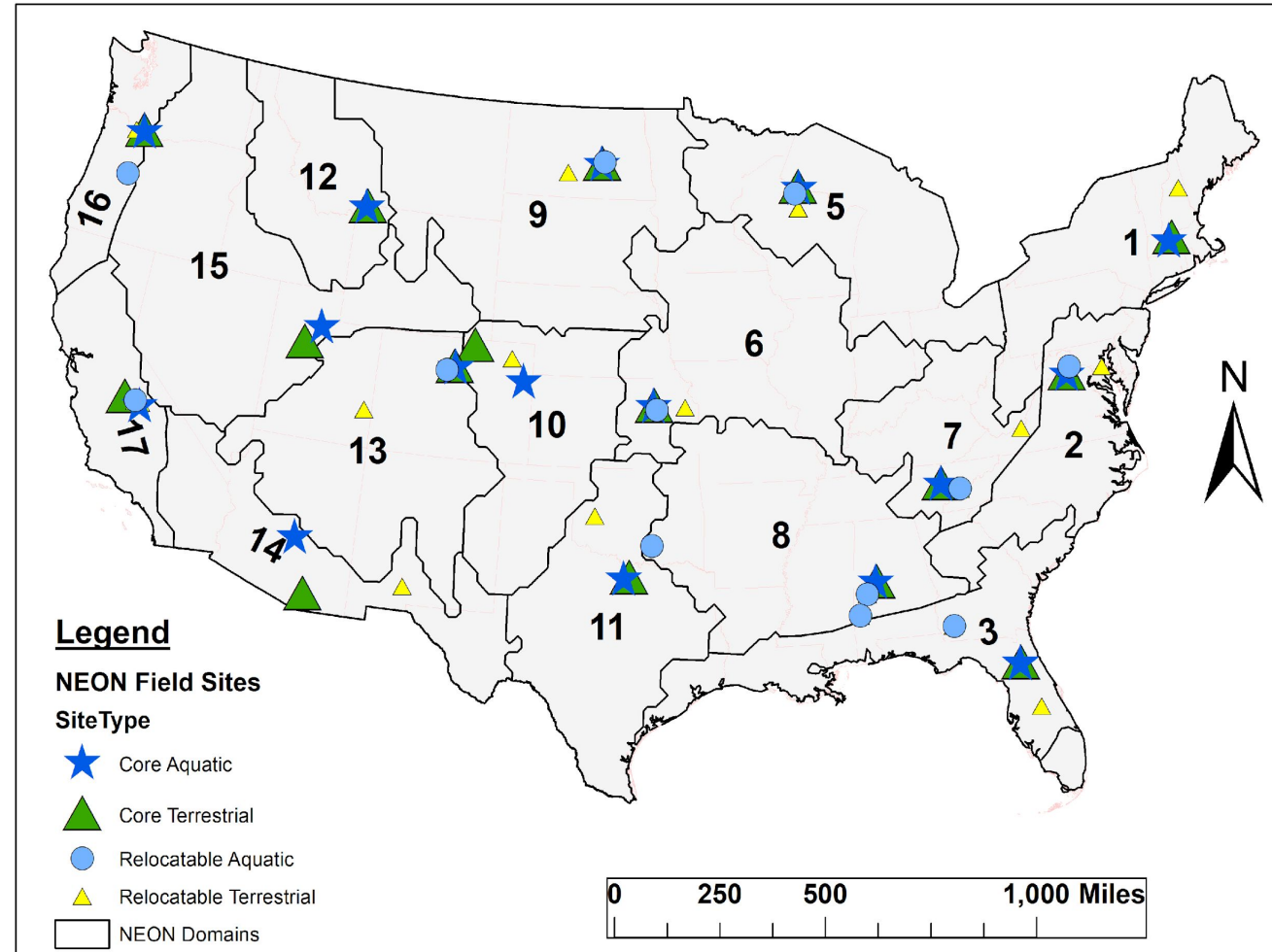
❖ Systematic Parameter Calibration for Actionable Science:

- Implementing a systematic parameter calibration approach is crucial
- It enhances the accuracy of simulations within land models like CTSM.
- The improved accuracy increases the suitability and accessibility of CTSM for actionable science applications.



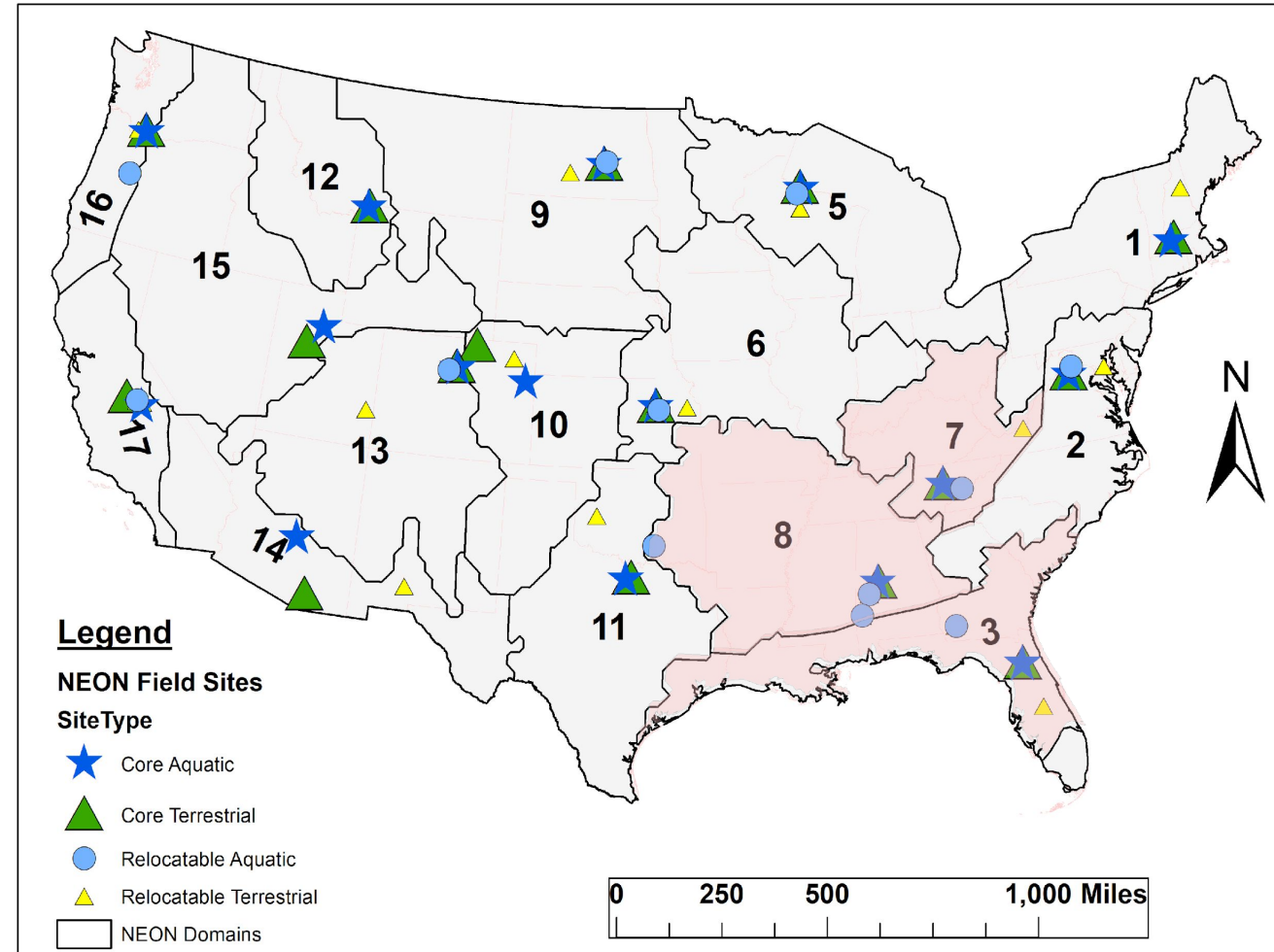
Methods and Data

- NEON Data Integration in CLM-5:
 - Climate data to drive CLM-5 simulations
 - Land use data to initialize and constrain model parameters
 - Enhancement of model performance and fidelity in simulating real-world ecosystems



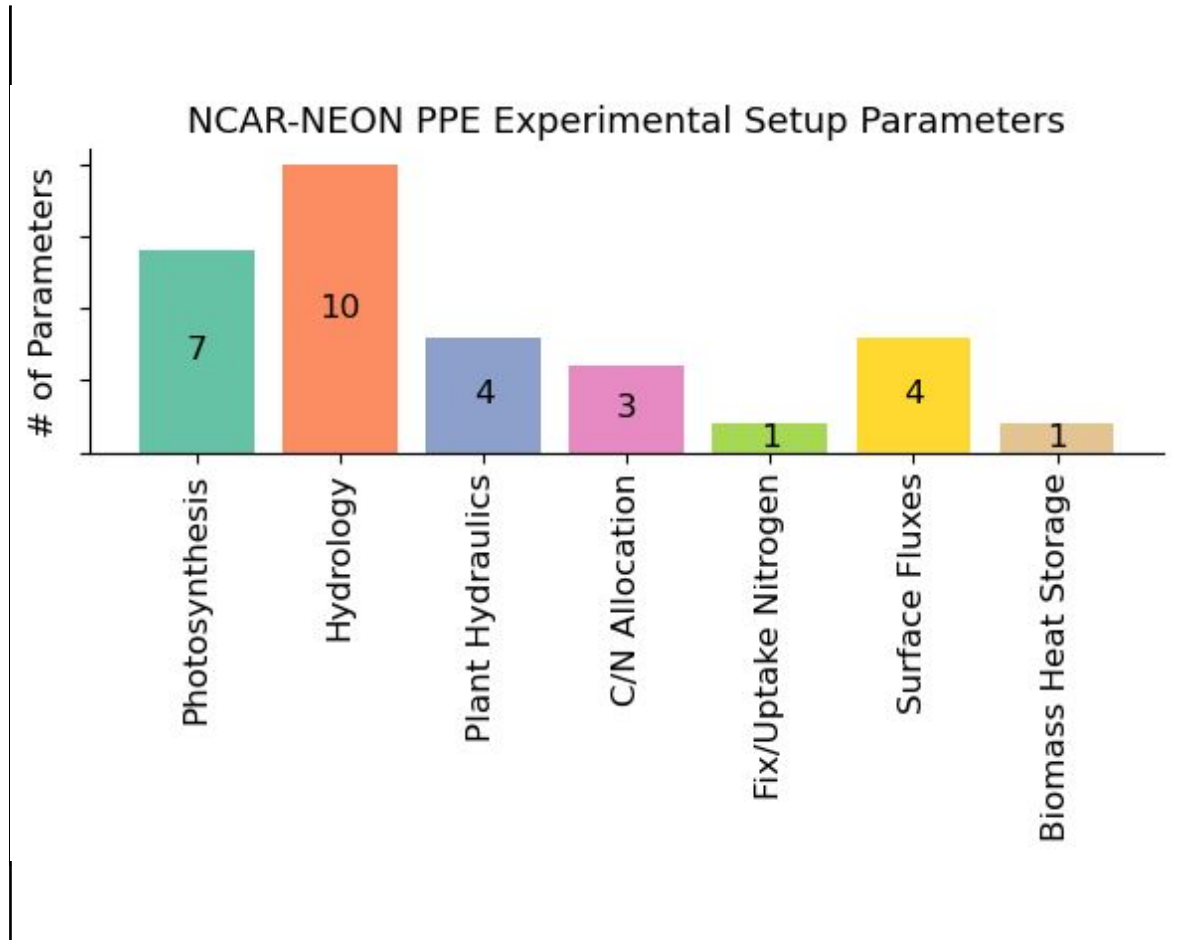
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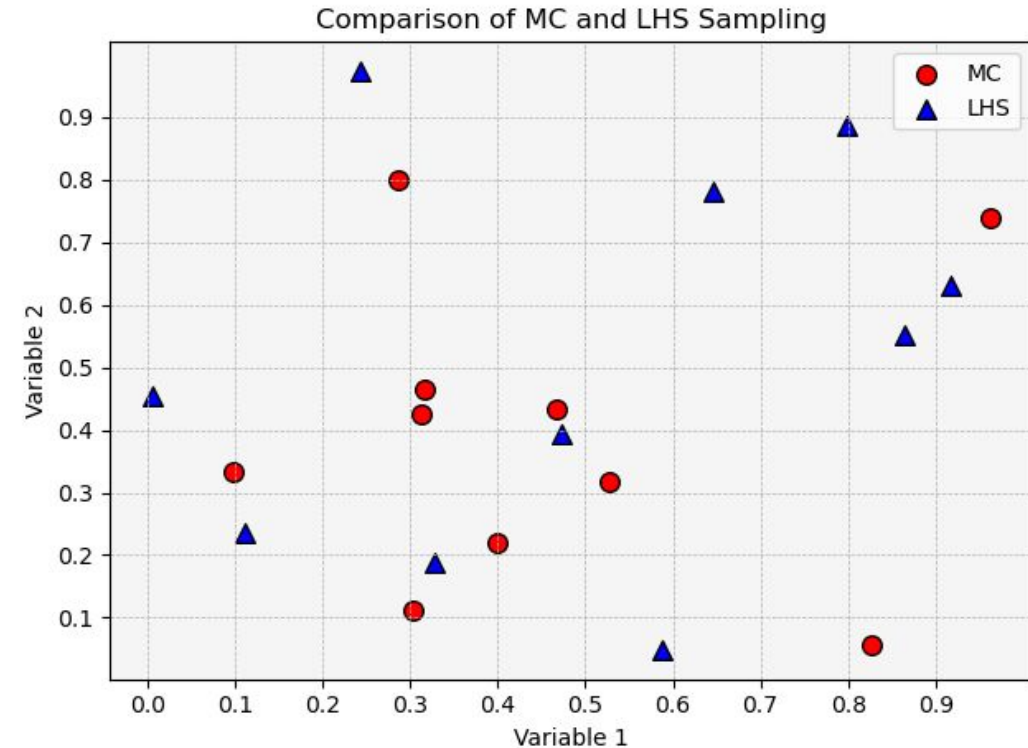
NEON-NCAR Sensitivity Experiment Setup

- 30 parameters affecting various processes
- Includes, plant functional types (pft), namelist and hard-coded parameters
- Obtained from Literature review and ongoing experiments
- Sampled using Latin Hypercube



Latin Hypercube Sampling Technique

- Better coverage of the parameter space, resulting in more representative samples and improved accuracy.
- Particularly beneficial when exploring complex systems with high-dimensional input spaces, as it efficiently covers the entire space while minimizing the number of samples required.

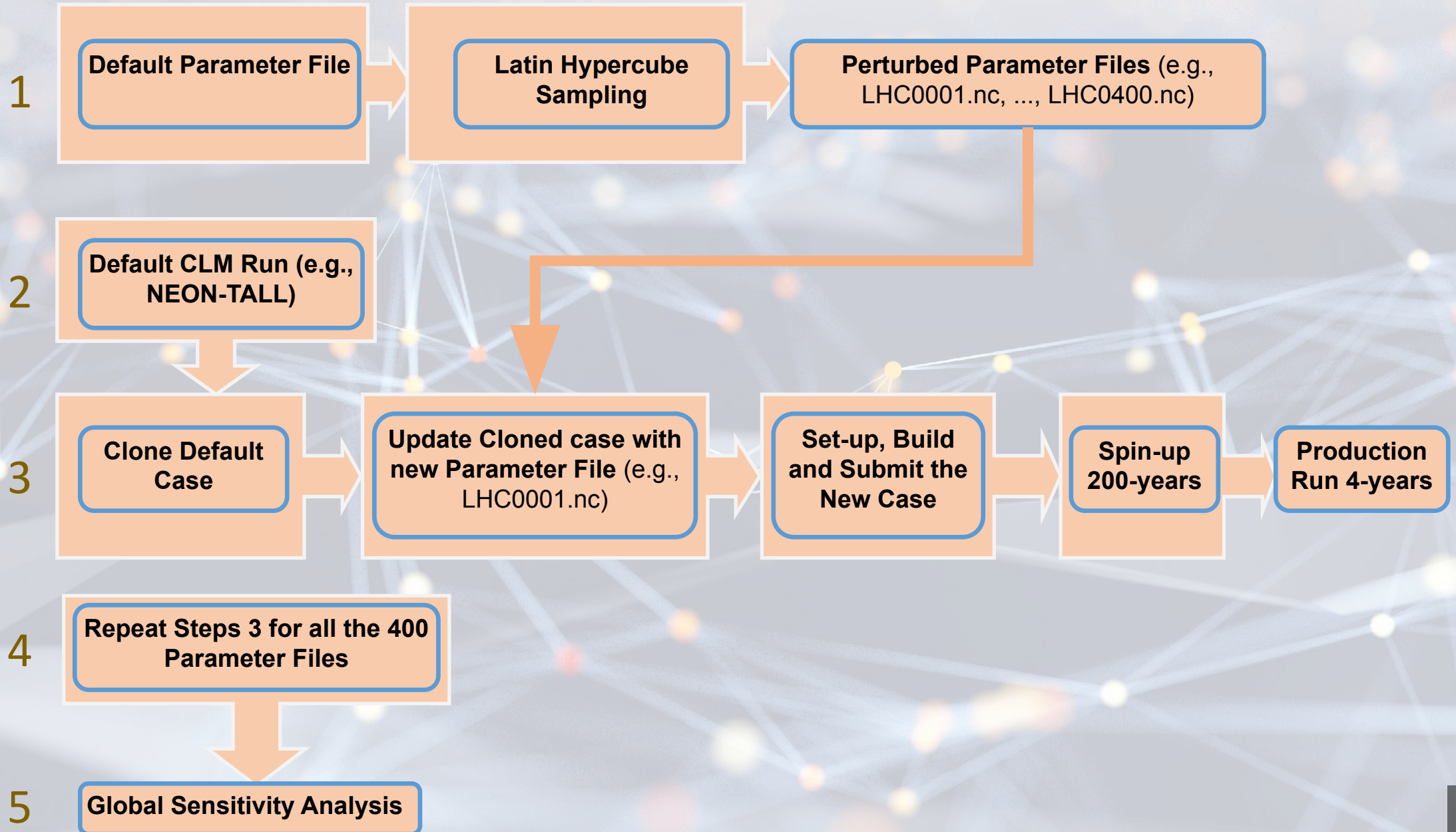


Parameter Sensitivity Analysis - VISCOUS

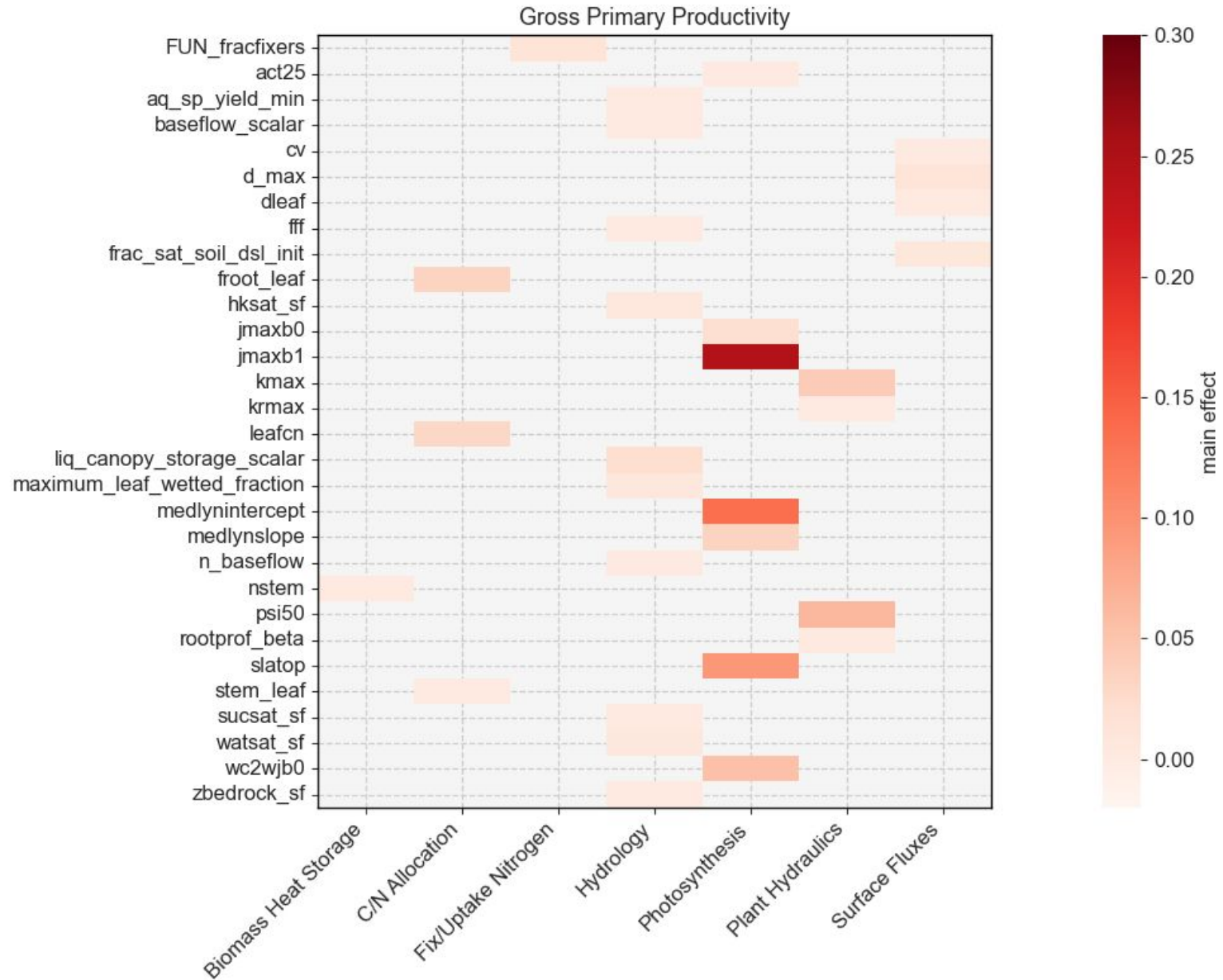
- **Data-driven approach:** Uses existing data to determine the importance of factors without assuming a specific distribution.
- **Joint probability density function:** Analyzes data using a joint probability density function to capture the relationship between input and output variables.
- **Efficient sensitivity analysis:** Calculates sensitivity indices without rerunning the model, providing insights into the relative importance of factors in a computationally efficient manner.

(Razi et al., 2021)

Sensitivity Analysis Experiment Design

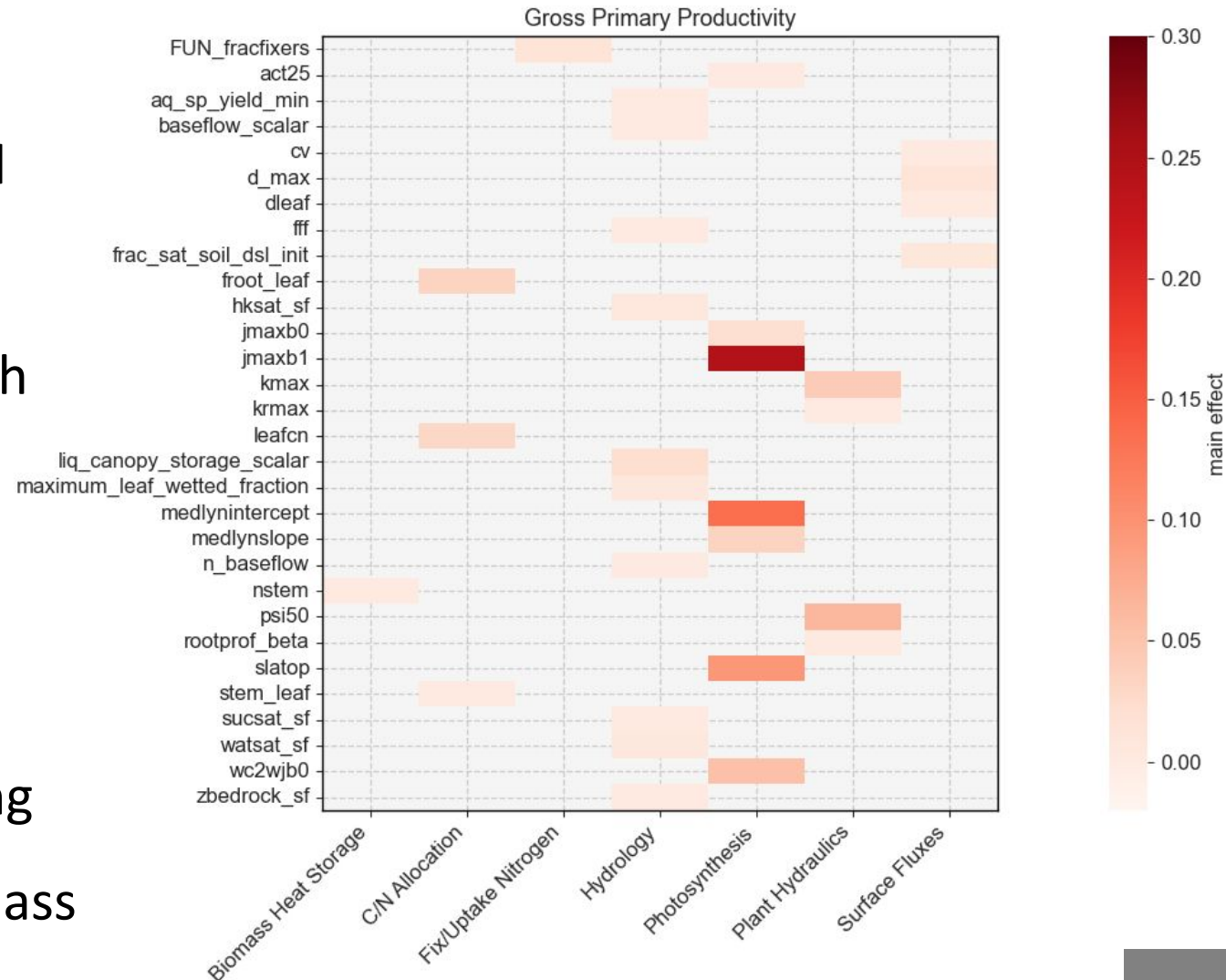


GPP Sensitivity Results

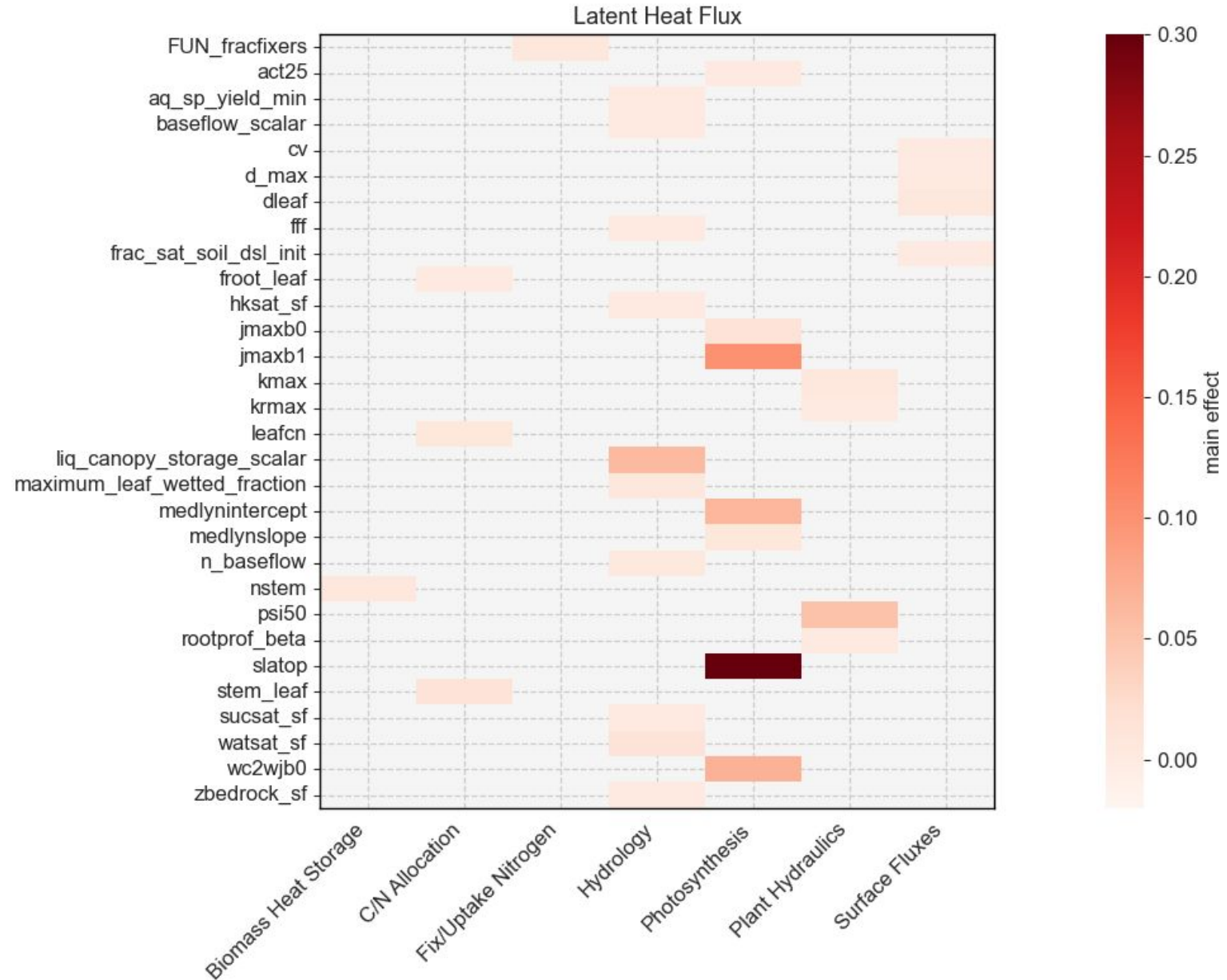


GPP Sensitivity Results

- ***Jmaxb1***: response of electron transport rate to light absorption
- ***Medlynintercept***: Sensitivity of stomatal conductance to atmospheric CO2
- ***Psi50***: Regulates water transport through plants
- ***Kmax***: Water flow rate through plant segments
- ***Froot-leaf***: Influences carbon partitioning between above and below-ground biomass

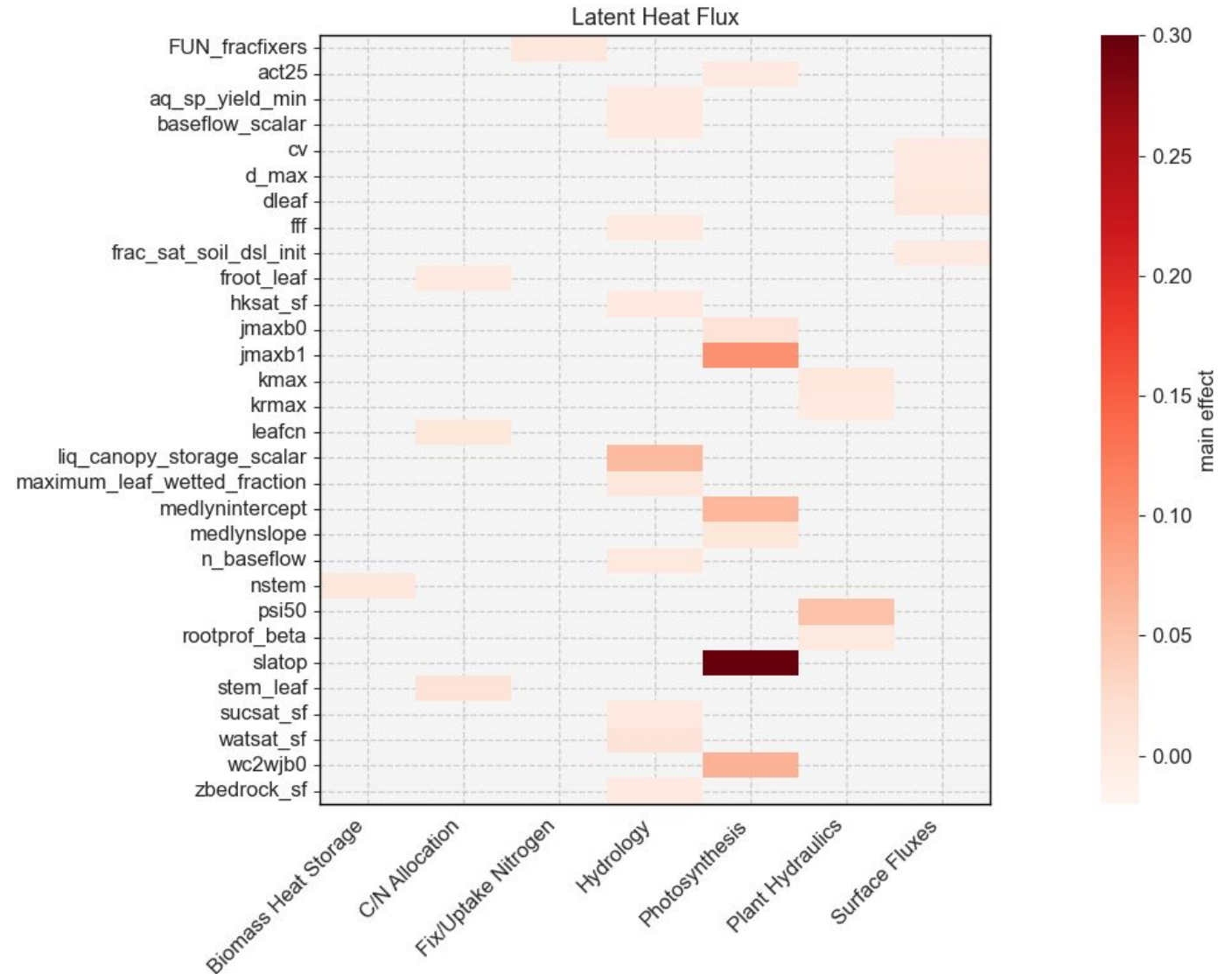


Latent Heat Flux Sensitivity Results

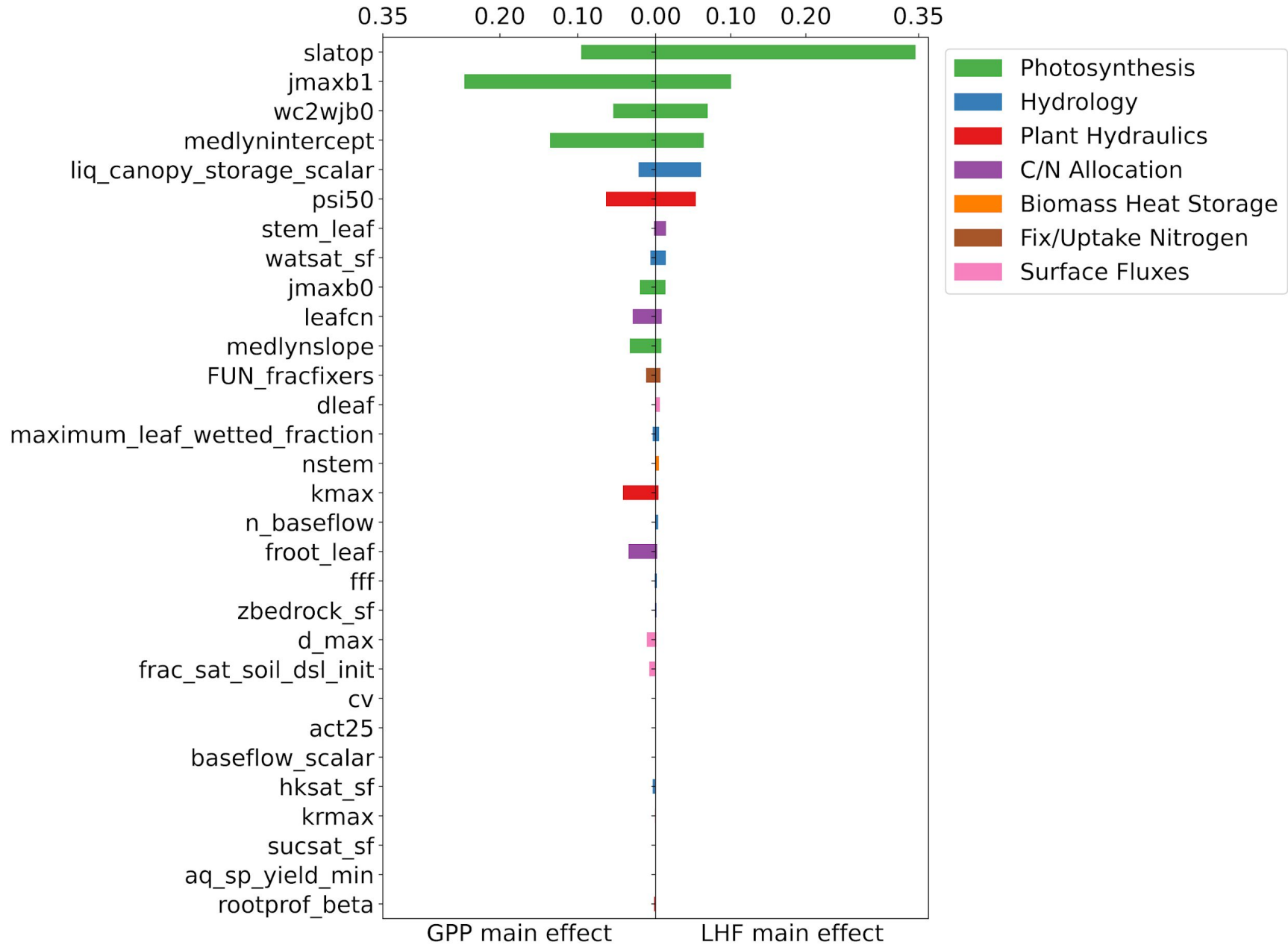


Latent Heat Flux Sensitivity Results

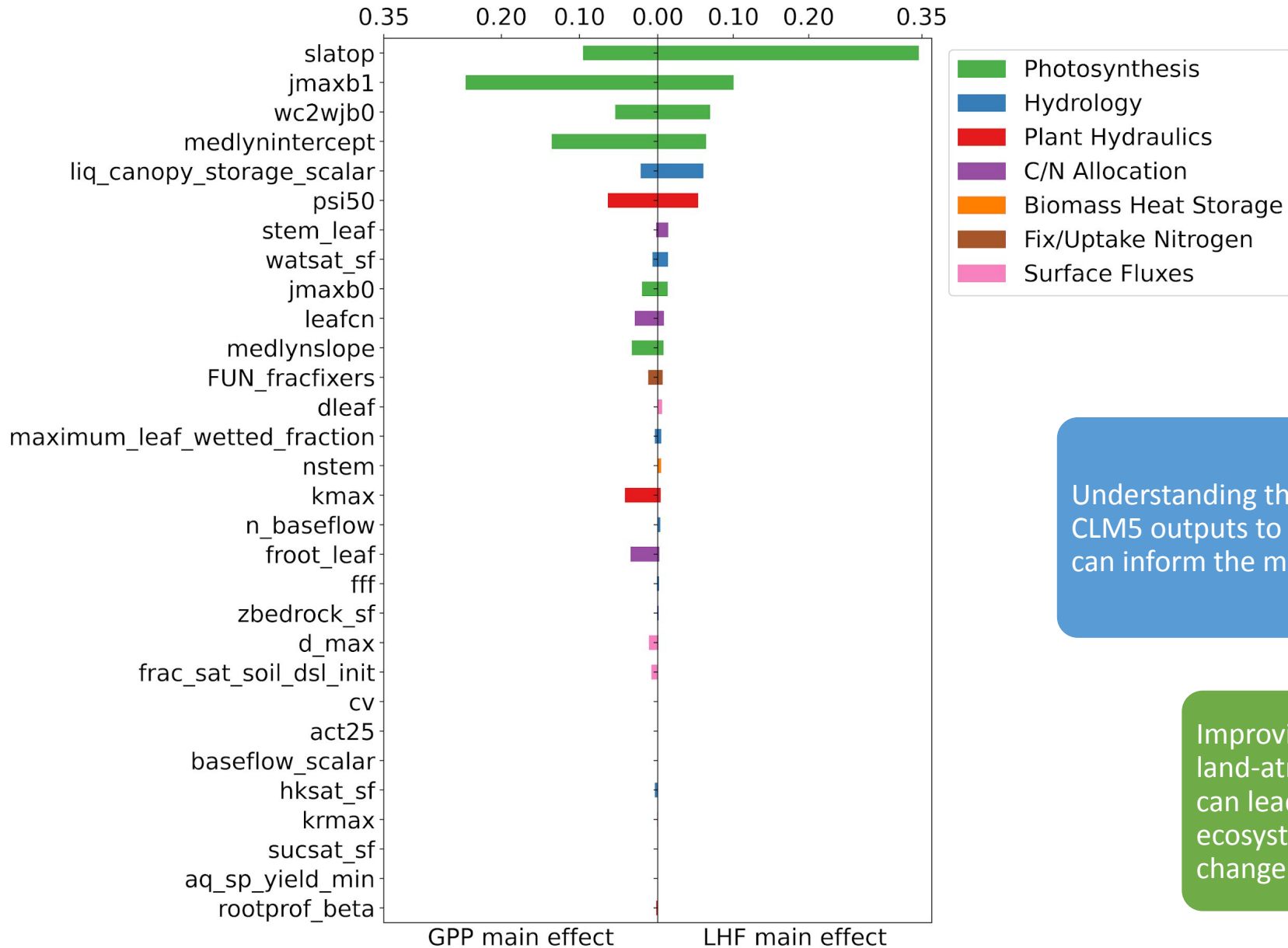
- **Slatop**: Relationship between transpiration and absorbed light
- **Liq_canopy_storage_scalar**: Canopy water storage
- **Wcwjb0**: Simulates the effects of temperature on the balance between Rubisco-limited and light-limited photosynthesis



Conclusion



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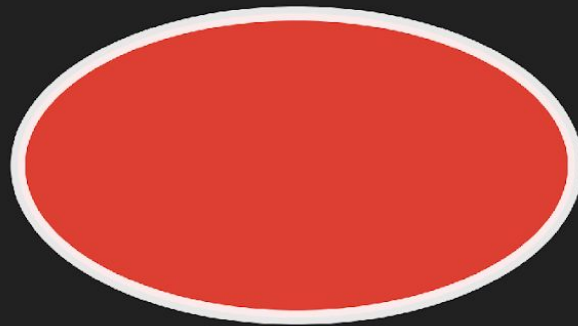


Understanding the sensitivity of CLM5 outputs to these parameters can inform the model calibration

Improving our understanding of land-atmosphere interactions can lead to better predictions of ecosystem function and climate change impact.



**Thanks for
listening!**



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