

# Uncertainty Quantification and Calibration of E3SM Land Model



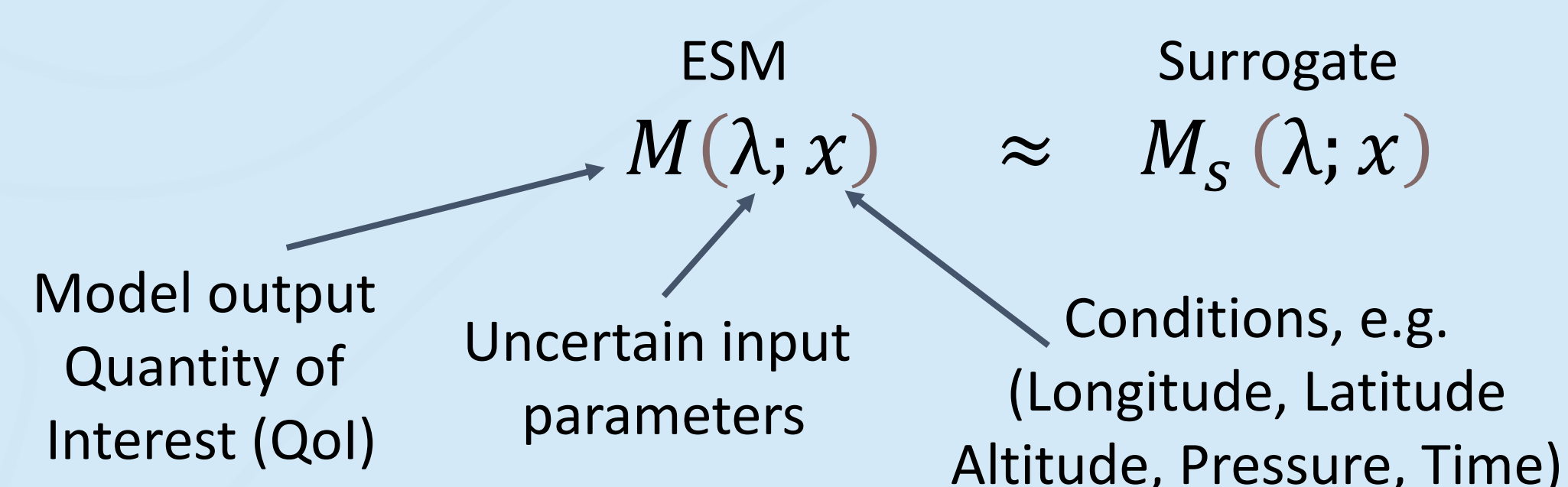
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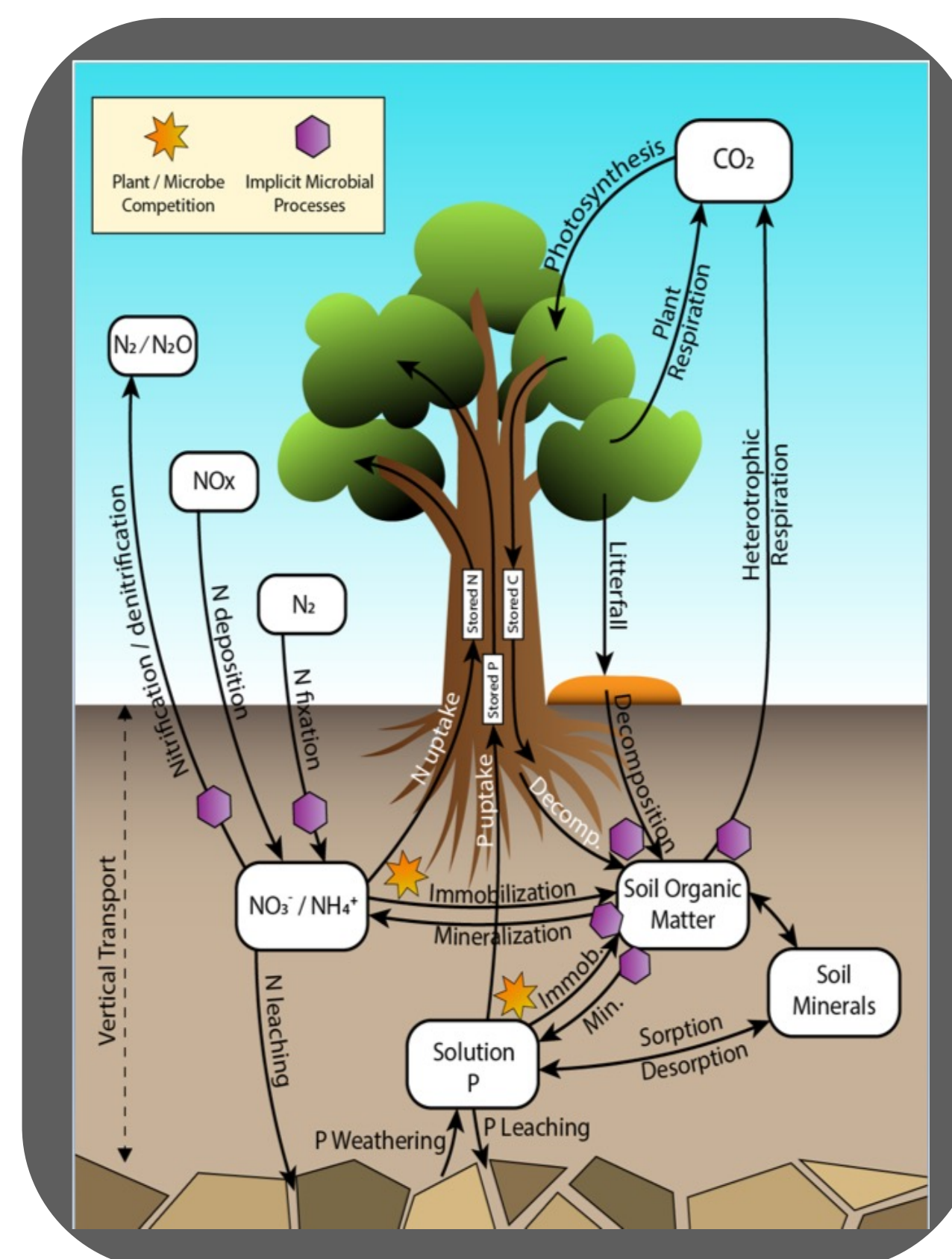
## Objective

Develop workflows of surrogate construction for outputs of earth system models (ESM) to enable uncertainty quantification (UQ)

Sample-intensive studies, such as UQ and parameter calibration, for earth system models require a construction of a **surrogate model** that approximates the ESM behavior across a range of conditions and input parameters.

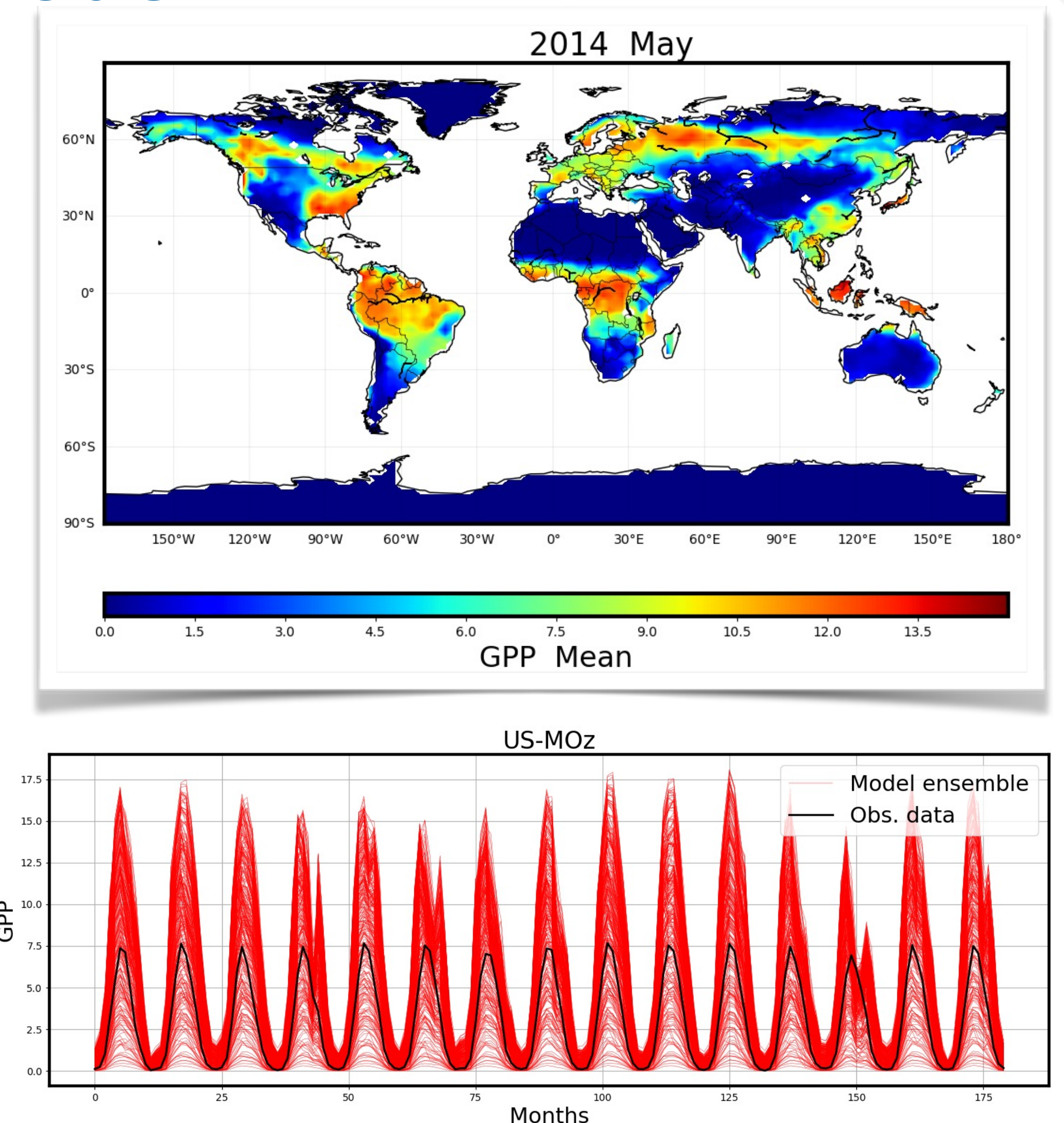


## E3SM Land Model

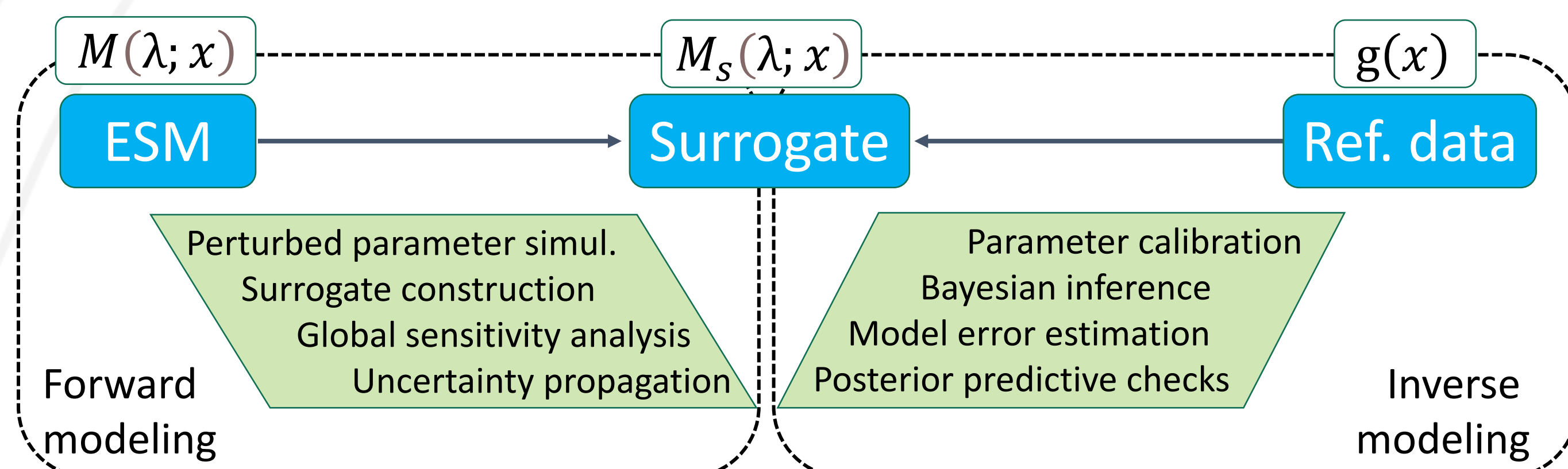


Satellite Phenology version used for this study

Quantity of Interest: Gross primary productivity (GPP)... resolved in space ... and time.



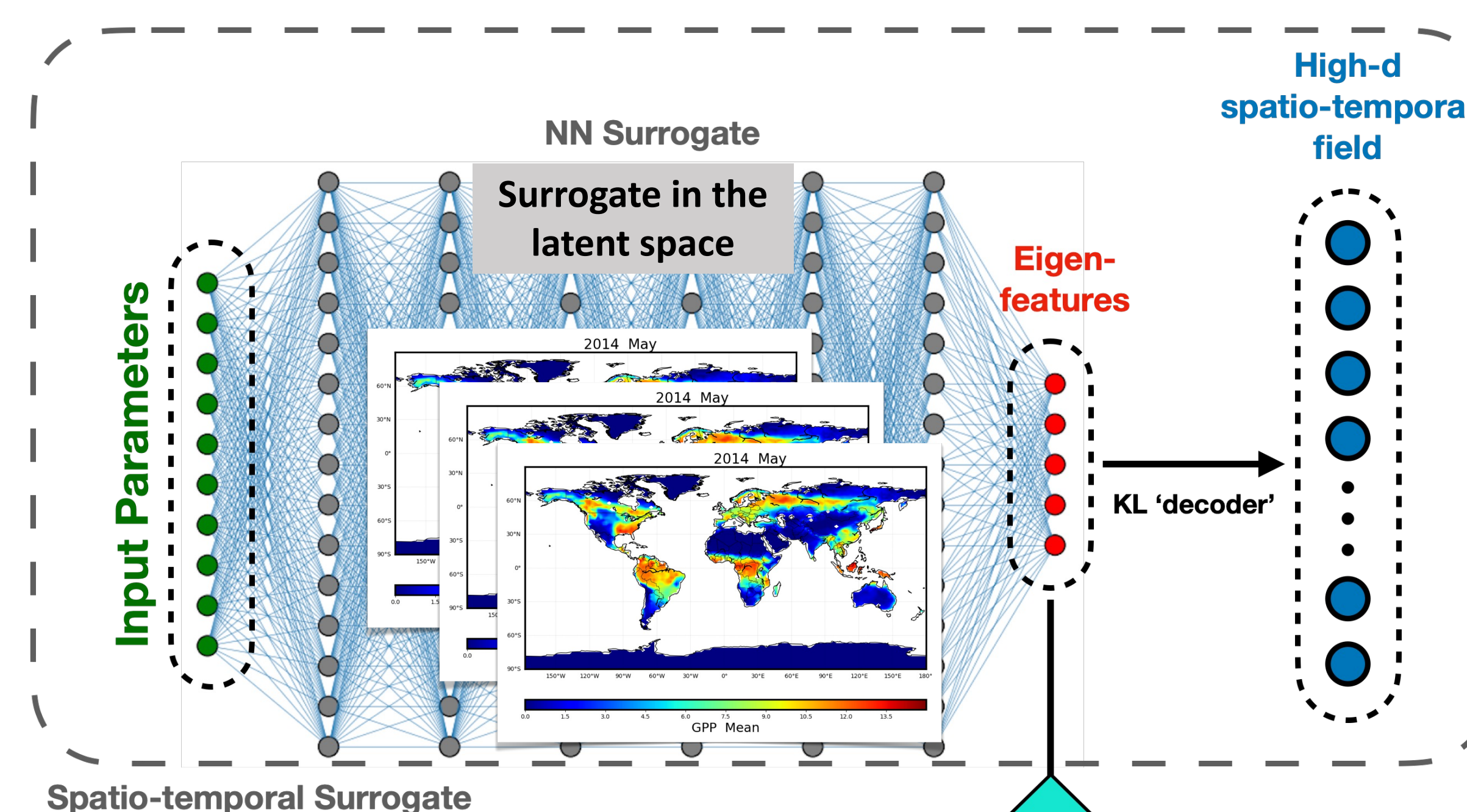
## UQ Workflow



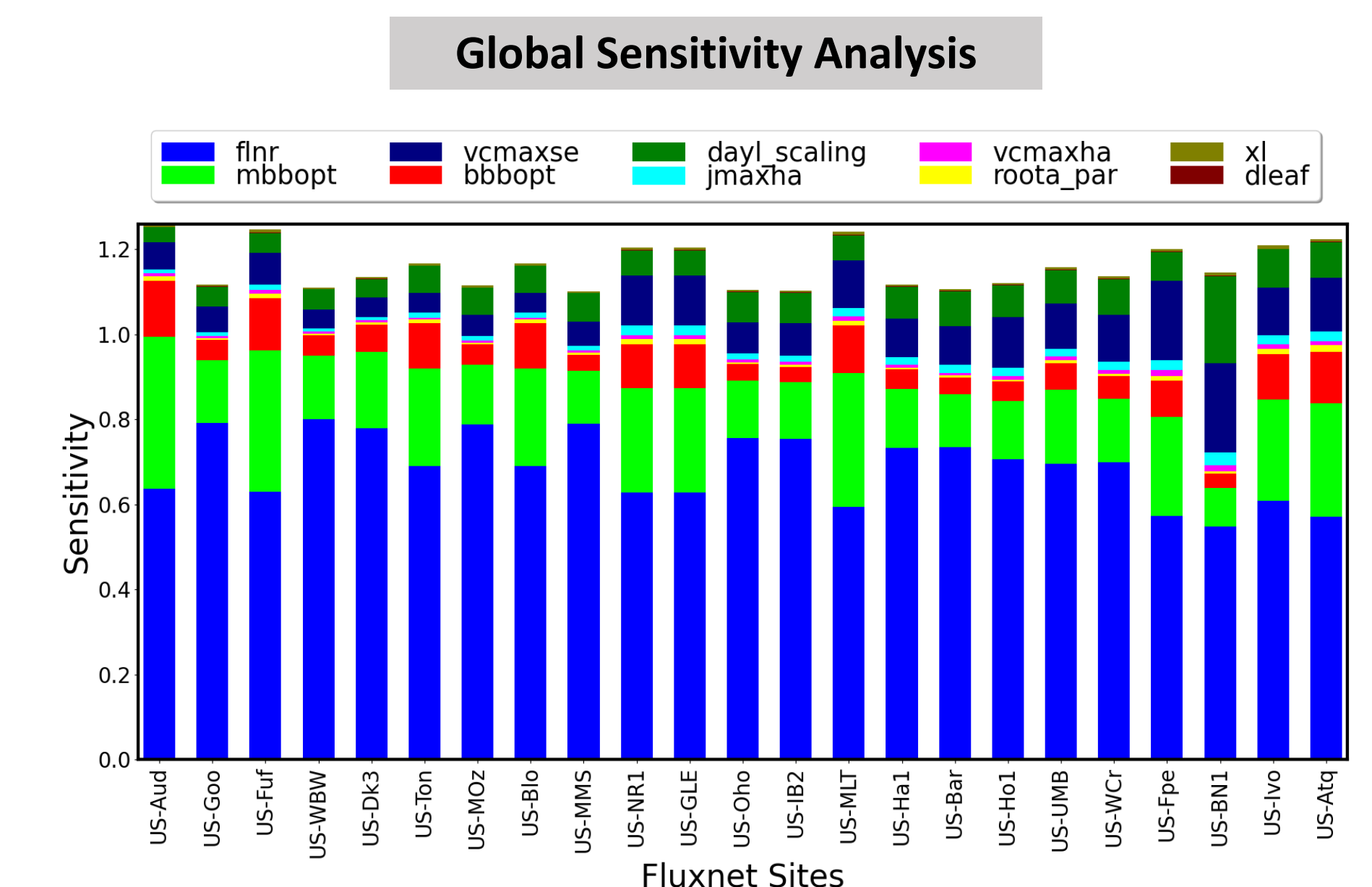
### Major challenges:

- **Large number of conditions / high-dimensional output fields**
  - ✓ Employ Karhunen-Loève decomposition to reduce dimensionality
$$M(\lambda; x) \approx \bar{M}(\lambda; x) + \sum_{j=1}^J \xi_j(\lambda) \sqrt{\mu_j} \phi_j(x)$$
  - ✓ Construct surrogate in the latent eigen-space  $\xi(\lambda) \approx \xi_s(\lambda)$
- **Large number of uncertain inputs / high-dimensional stochastic space**
  - ✓ Employ polynomial surrogates with compressed sensing to pick only relevant parameter combinations
$$\xi_s(\lambda) = \sum_{k=1}^K c_k \Psi_k(\lambda)$$
- **Expense of ESM / low number of training simulations**
  - ✓ No real remedy but cross-validation and hyperparameter optimization help.

## Land Model Calibration enabled by Spatio-Temporal Neural Network Surrogate



- ~5000 spatio-temporal cells down to ~10 eigenfeatures with 1% loss of accuracy
- The overall spatio-temporal surrogate has a fraction of ELM cost



### Model calibration via Bayesian inference

