

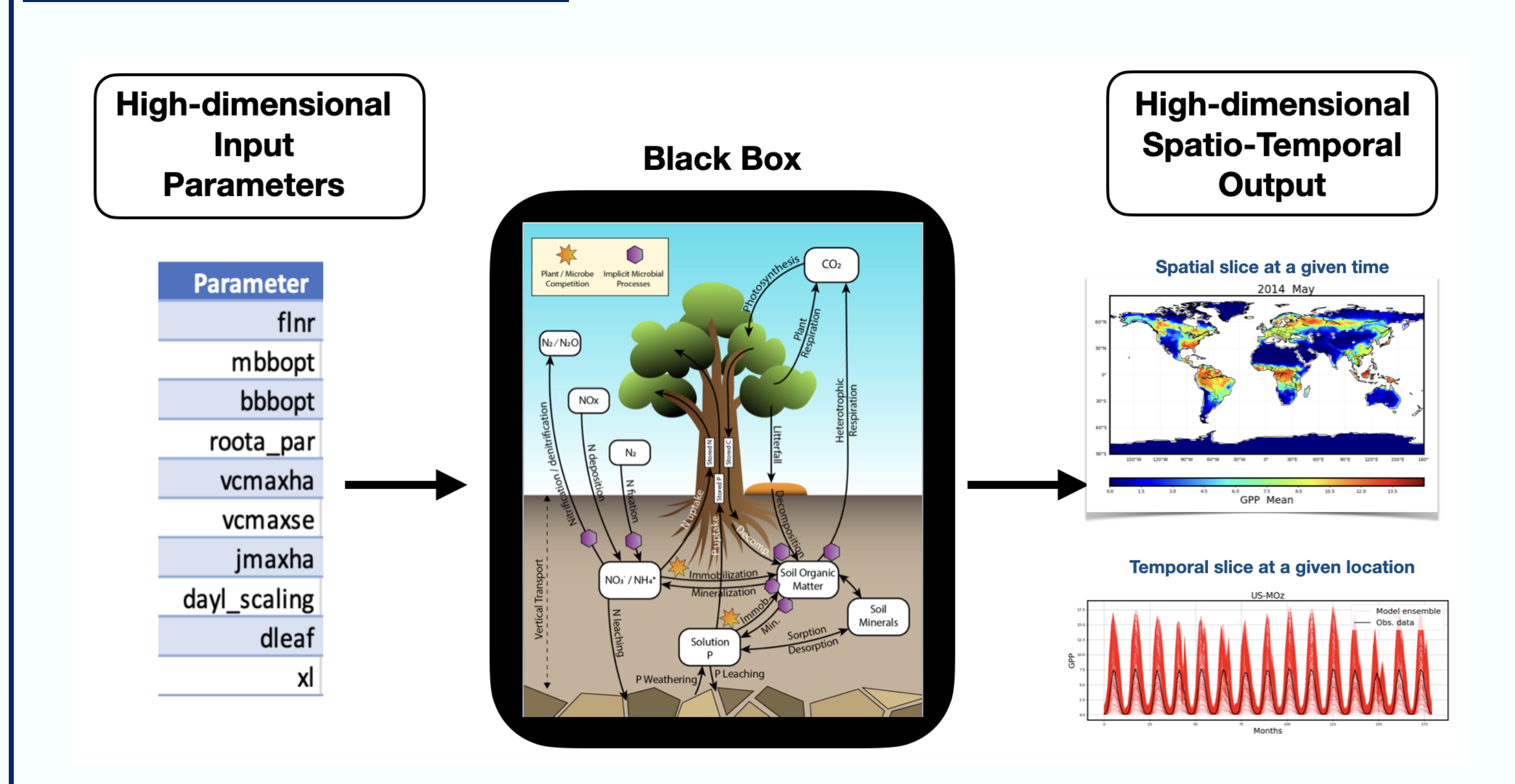
Hit twice by the curse of dimensionality: spatio-temporal land model calibration using Karhunen-Loève expansions

Khachik Sargsyan¹, Daniel Ricciuto², Cosmin Safta¹

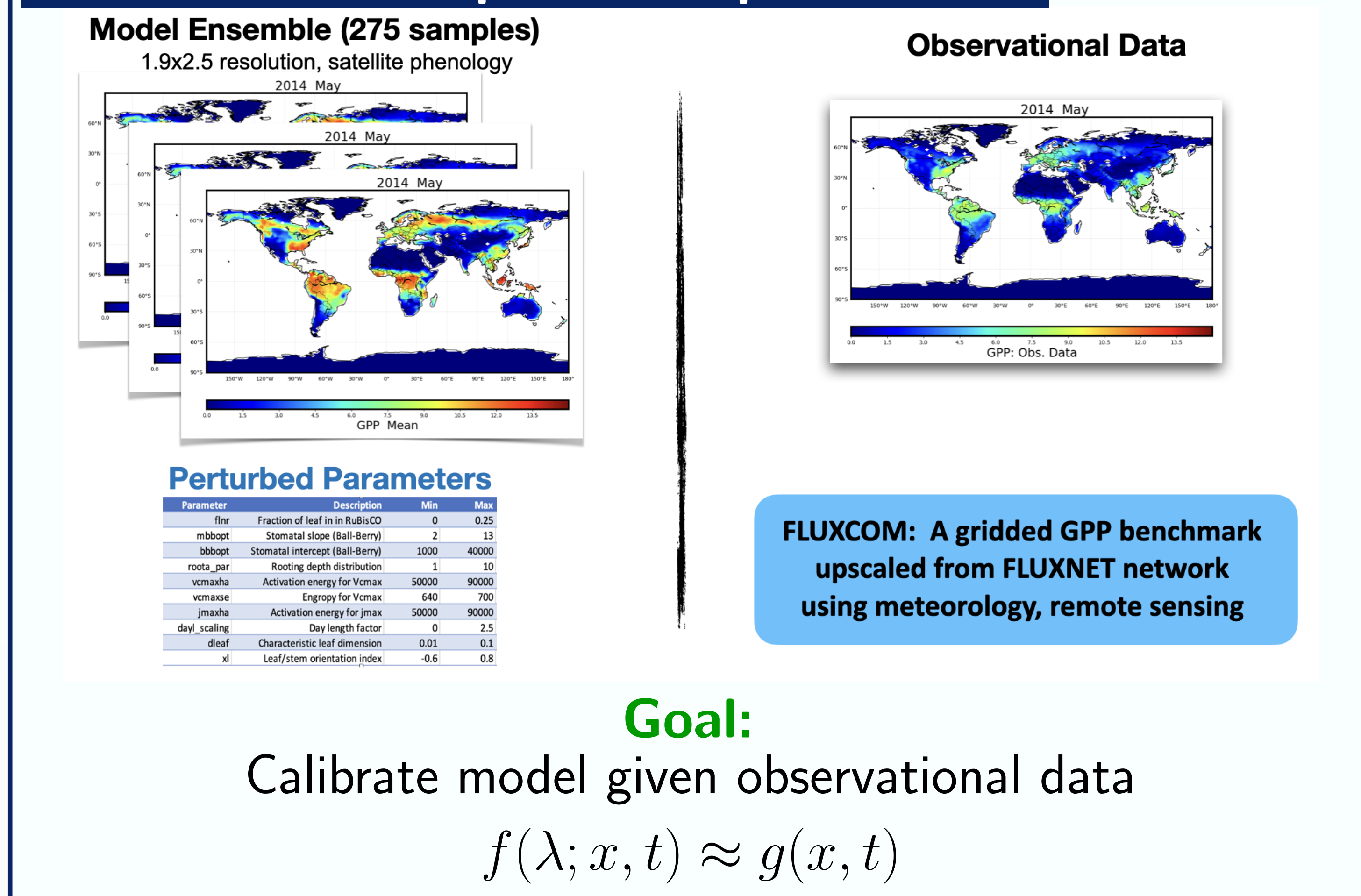
¹ Sandia National Laboratories, Livermore, CA, USA, ² Oak Ridge National Laboratory, Oak Ridge, TN, USA



E3SM Land Model



Calibration of a Spatio-Temporal Model



Need for a Surrogate Model

- **Forward modeling:** Global sensitivity analysis requires prohibitively many model evaluations
- **Inverse modeling:** Likelihood requires online evaluation of the model, e.g.

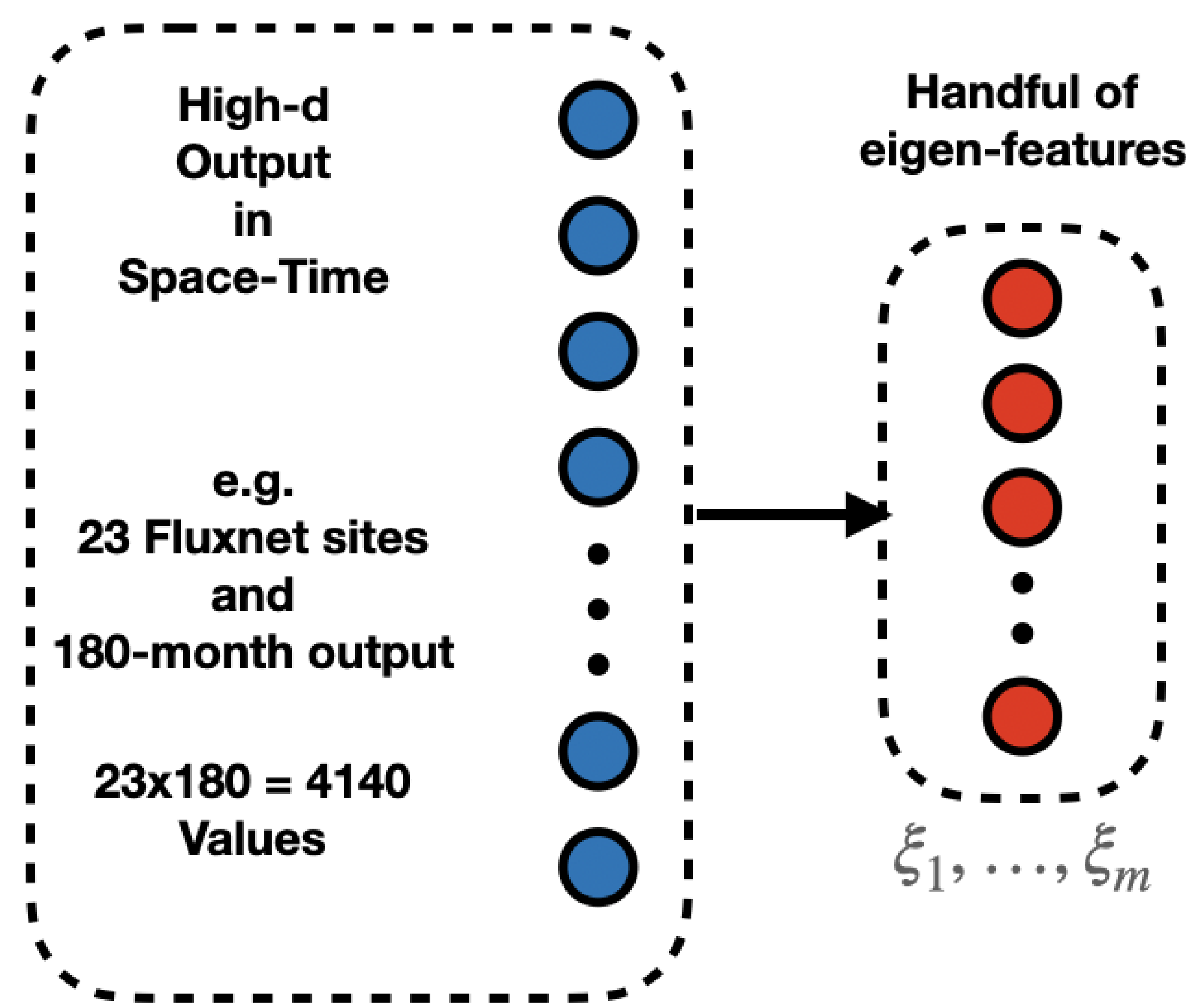
$$p(D|\lambda) \propto \exp\left(-\frac{1}{2}\|f(\lambda; x, t) - g(x, t)\|^2\right)$$
- Construct a surrogate, inexpensive approximation

$$f(\lambda; x, t) \approx f_s(\lambda; x, t)$$
 using the perturbed ensemble of model simulations.

High-Dimensional Output

- Karhunen-Loève (KL) Expansion:**
- Optimal linear expansion in terms of eigenvalues of the covariance kernel.

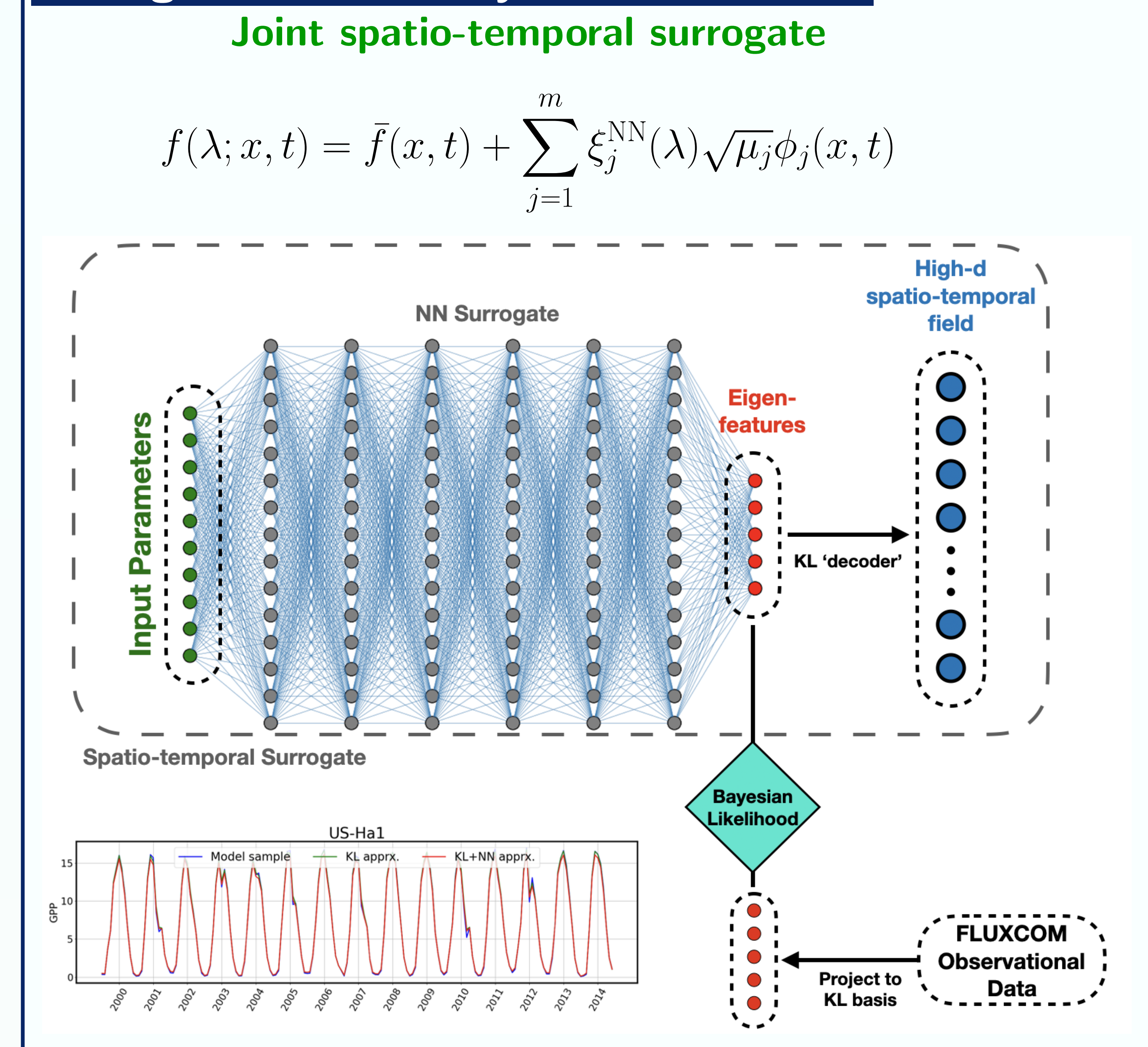
$$f(\lambda; x, t) = \bar{f}(x, t) + \sum_{j=1}^{\infty} \xi_j(\lambda) \sqrt{\mu_j} \phi_j(x, t)$$
 - Similar to principal components analysis, but in the continuous, (x, t) space.
 - The analysis of spatio-temporal field $f(\lambda; x, t)$ reduces to a handful of eigen-features $\xi_j(\lambda)$
 - KL expansion is an optimal linear auto-encoder



High-Dimensional Input

- Option 1: Sparse Polynomial Chaos (PC) Surrogates for each $\xi_j(\lambda)$**
- $$\xi_i(\lambda) \approx \xi_i^{PC}(\lambda) = \sum_{k=0}^K c_{ik} \Psi_k(\lambda)$$
- Option 2: Joint Neural Network (NN) Surrogate for all $\xi_j(\lambda)$'s**
- $$\xi(\lambda) \approx \xi^{NN}(\lambda)$$

Surrogate-enabled Bayesian Calibration



Results and Summary

