

# Autotuning E3SMv3

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- **Motivation:** Expedite and formalize E3SMv3 tuning.
  - Identify an optimal tuning parameter set for the atmosphere to assist expert model tuning.
  - Identify plausible alternative parameter sets.
  - Timeline: 6-12 months after code freeze in early 2023.
- **Setup:**
  1. Sample E3SM by running 100's of E3SM forward simulations with ~5 perturbed tuning parameters.
    - Prescribed present-day SST, ~5-10 yrs.
  2. Create a surrogate model.
    - Input: uncertain E3SM model parameters.
    - Output: climatologies of E3SM spatial fields on a 2-D grid.
  3. Optimize and calibrate parameters by sampling the surrogate model
    - Identify parameters that minimize the cost function between surrogate-predicted spatial fields and observational fields
  4. Confirm parameter selection by running surrogate-predicted optimal and alternative parameter sets in E3SM.

Expert manual tuning	Automated tuning
Time-consuming (6-12 months)	Less than 6-12 months
Deterministic: one set of “tuned” parameters per model release	Probabilistic: a distribution of parameters per model release OR observational target
Non-reproducible	Reproducible
Computationally expensive	Computationally expensive (upfront only)

# Prototype problem using E3SMv2

## Prototype problem

- N=250 10-year E3SMv2 simulations, 5 uncertain atm. parameters
- Surrogate construction using PCA and polynomial chaos

## Capabilities

- E3SMv2 parameter sampling (LHS) and ensemble generation using Dakota<sup>1</sup>
- Surrogate construction for E3SMv2 spatial fields using machine learning and reduced order modeling
- Deterministic and Bayesian calibration for uncertain E3SMv2 parameters against an unlimited number of combined spatial fields
- Visualization as a function of atmosphere parameters

## Challenges

- Optimization
  - Surrogate errors are relatively large; multiple optimizations are needed to derive a parameter set with skill equal to the expertly tuned E3SMv2
  - Currently working to improve surrogate accuracy by experimenting with the surrogate construction loss function
- Calibration
  - Simple MSE-based objective function assumes fields and grid points are independent and model has no structural bias

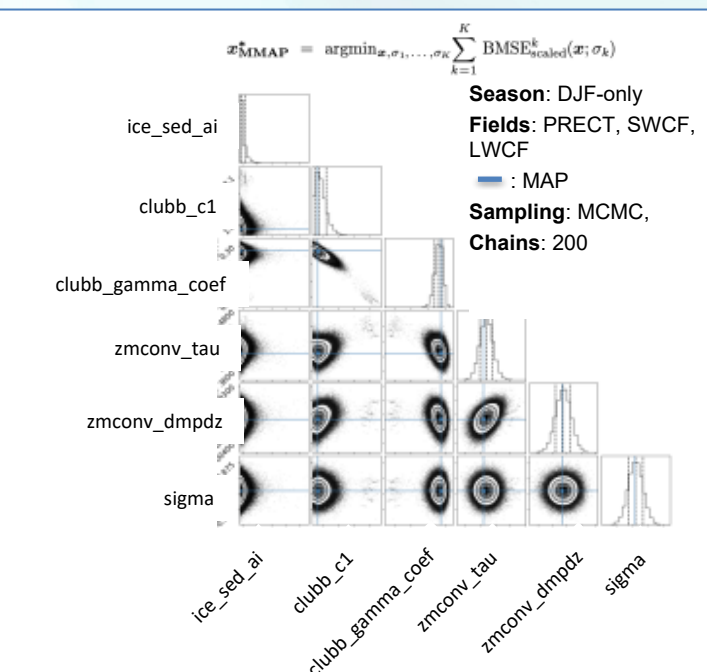


Figure: Corner plot of E3SMv2 parameter calibration against observations.

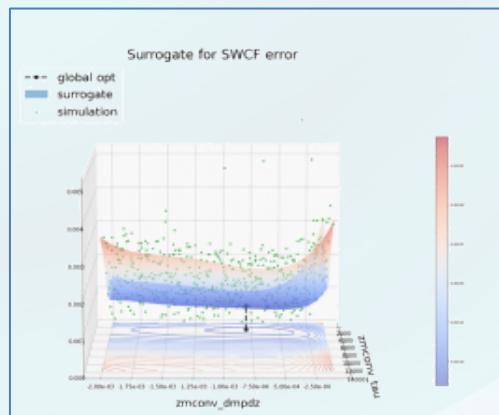


Figure: E3SMv2 SWCF error as a function of two E3SMv2 parameters as predicted by the surrogate (surface) and a ultra-low resolution configuration of E3SMv2 (green dots).

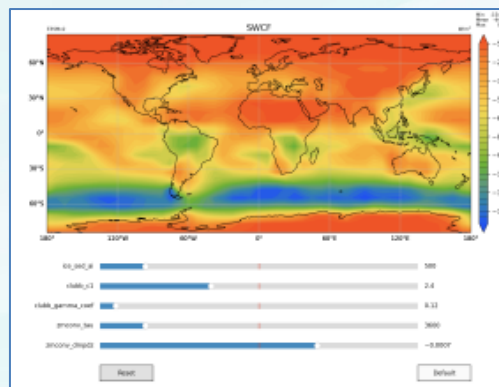


Figure: Screen shot from interactive visualization of fields as a function of E3SMv2 parameters.

## More information

- B. Wagman, K. Chowdhary, A. Salinger. *E3SM Atmosphere surrogate construction and calibration using machine learning and reduced order modeling (talk)*. SIAM UQ, April 2022
- Ensemble and surrogate software: <https://github.com/E3SM-Project/Autotuning-NGD>