

# An Efficient Method for Discerning Climate-Relevant Sensitivities in AGCMs

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- ▶ We're interested in time step sensitivities and convergence properties in CAM5
  - ▶ AMIP simulations need multiple years/decades to overcome natural variability
  - ▶ Model with small time step is expensive to integrate
- ⇒ Need an alternative experimentation strategy

## ▶ Our idea

Replace serial-in-time long-term climate simulations by **representative ensembles of shorter runs**

- ▶ Utility of the method goes far beyond time step sensitivity
- ▶ An uncertainty quantification (UQ) example is shown later
- ▶ Very useful in efficient model tuning and sensitivity analysis, especially for high-resolution models

## Similarities

- ▶ Both exploit the important role of fast processes in determining model sensitivities/uncertainties

## Differences

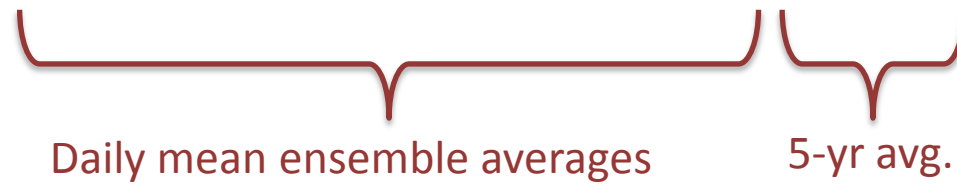
- ▶ Model biases v.s. sensitivities as focus
- ▶ In this study we are interested in parametric and structural sensitivities close to the model's equilibrium state

**We are trying to make use of the scientific basis of CAPT in more general ways**

# Evaluation Example (1): Time Step Sensitivity

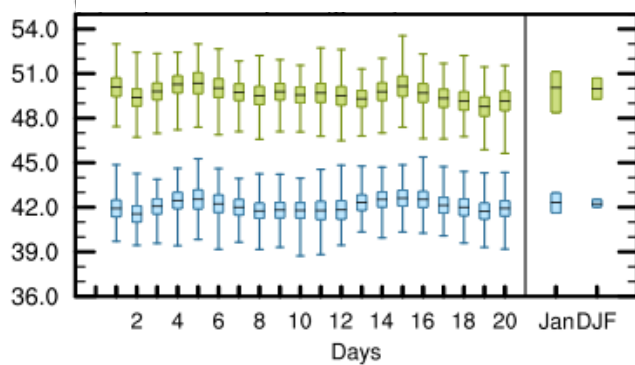
- ▶ Reference simulations
  - 1+5-yr simulations, 2 degree FV dycore
  - 30-minute and 4-minute time step
  - 5-yr mean DJF differences in clouds and precipitation
  
- ▶ Ensemble simulations
  - 50 members
  - Initial conditions sampled from DJF of a previously performed 20-yr simulation
  - 30-minute and 4-minute ensembles use the same set of initial conditions
  
- ▶ Compare 5-yr winter averages with 1-day 50-member ensemble averages

## Global Mean Total Cloud Cover (%)

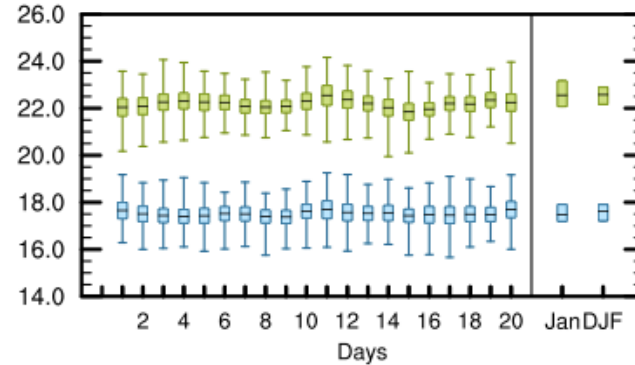


# Some Other Fields

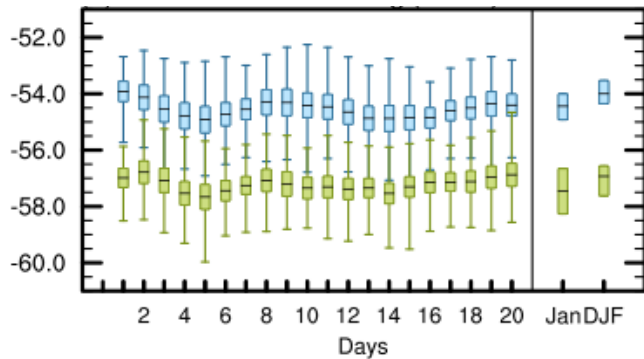
### Liquid Water Path ( $\text{g m}^{-2}$ )



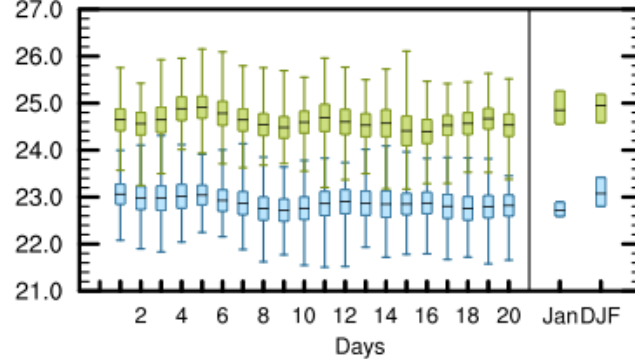
### Ice Water Path ( $\text{g m}^{-2}$ )



### Shortwave Cloud Forcing ( $\text{W m}^{-2}$ )

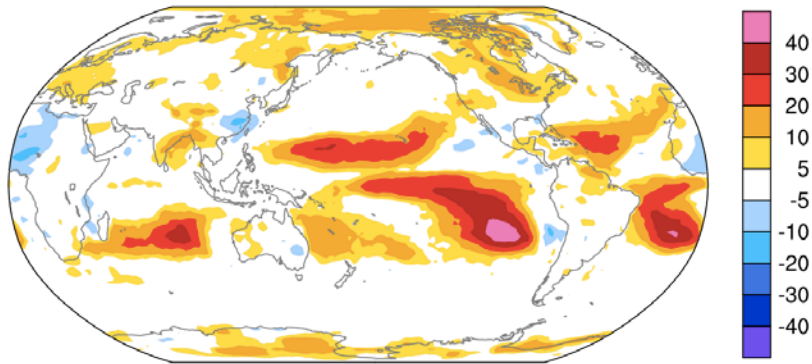


### Longwave Cloud Forcing ( $\text{W m}^{-2}$ )

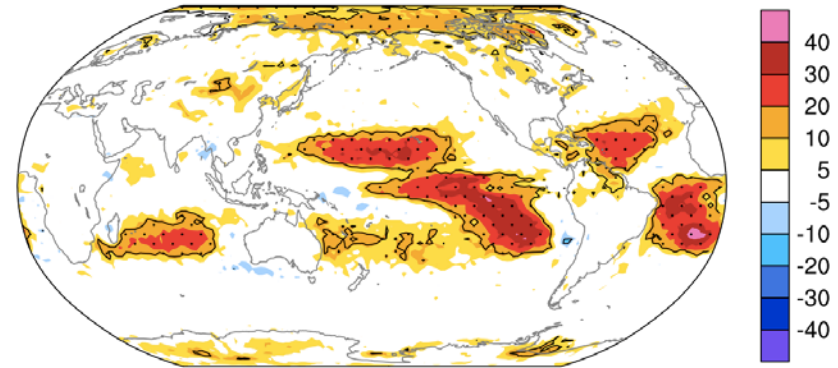


## Total Cloud Cover Difference (%) 4-minute minus 30-minute time step

5-yr DJF Average



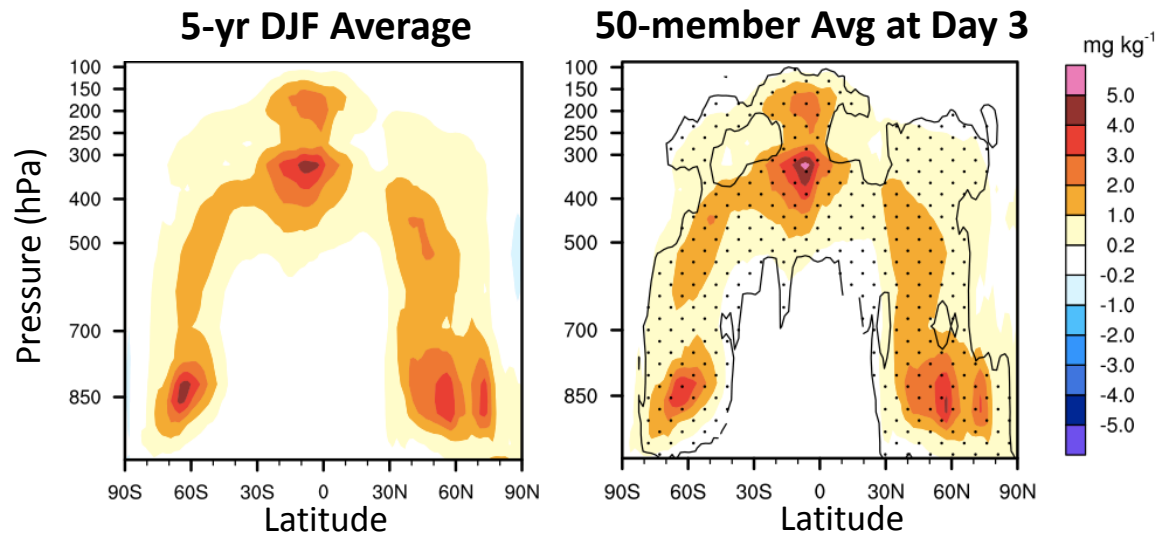
50-member Average at Day 3



Stippling in the right panel indicates differences significant at the 95% confidence level according to the local t-test.

## $\Delta$ Cloud Ice

4-minute minus 30-minute time step



Stippling in the right panel indicates differences significant at the 95% confidence level according to the local t-test.



# Summary of Example (1)

## Effectiveness

- ▶ Ensembles of 20 to 50 three-day simulations are sufficient for clouds and precipitation
- ▶ The method can detect global mean differences AND identify climate regimes
- ▶ Ensembles can be combined with nudging to help understand the role of physics-dynamics interaction (not shown here)

## Computational efficiency

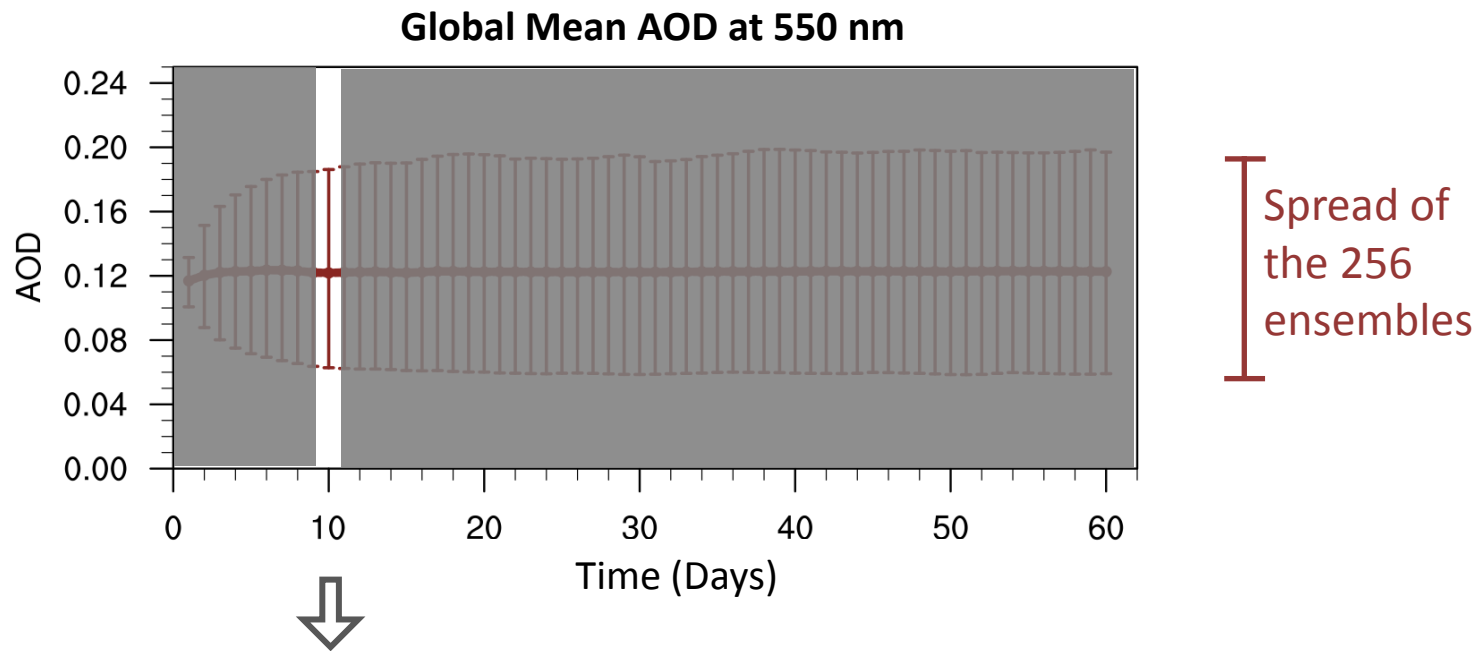
- ▶ 50 x 3-day simulations v.s. 1+5-yr climate run
- ▶ Total CPU time: 150 v.s. 2190 days, **a factor of 15**
- ▶ Throughput time: 20 minutes v.s. 4-7 days on Yellowstone, **a factor of several hundred**
- ▶ Contrast can be even stronger for certain variables and domain averages

# Evaluation Example (2): Uncertainty Quantification

- ▶ Zhao et al. (2013, doi:10.5194/acp-13-10969-2013)
  - Parametric sensitivity of TOA radiative balance
  - Perturbed 16 empirical parameters in CAM5
  - Quasi-Monte Carlo sampling, 256 simulations, 1+4-yr AMIP
  
- ▶ Our ensemble experiments
  - Same 256 parameter combinations
  - 12 ensemble members representing 12 months of a year
  
- ▶ Compare 4-yr averages with 1-day 12-member averages

# Spin-up Time

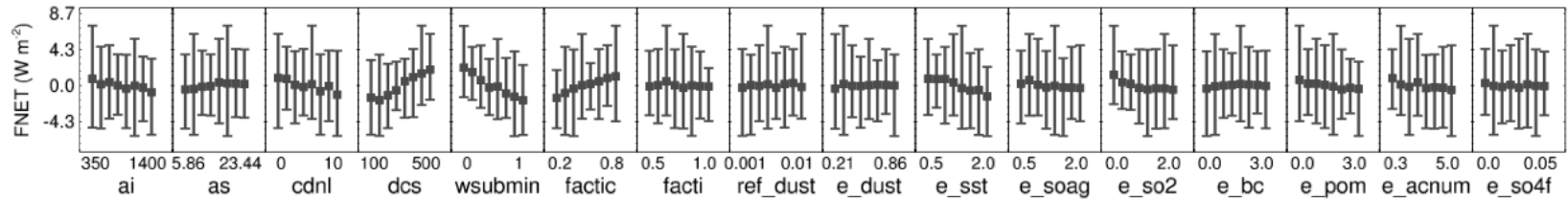
- ▶ 11 out of 16 parameters directly affect aerosols (e.g., tuning factors for emissions)
- ▶ Global mean aerosol life cycle is ~4 days in CAM5-MAM3 (Liu et al., 2012, GMD)
- ▶ Expect longer spin-up than in the 1<sup>st</sup> example



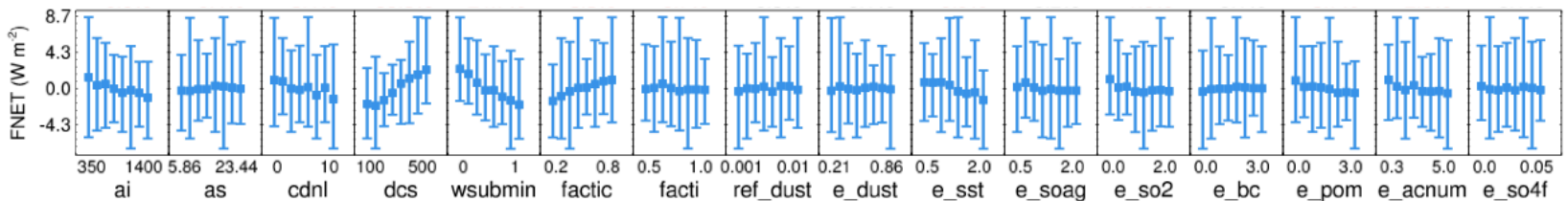
# Global Mean TOA Net Radiative Flux (FNET)

► Sensitivity of FNET to individual parameters

4-yr Annual Mean



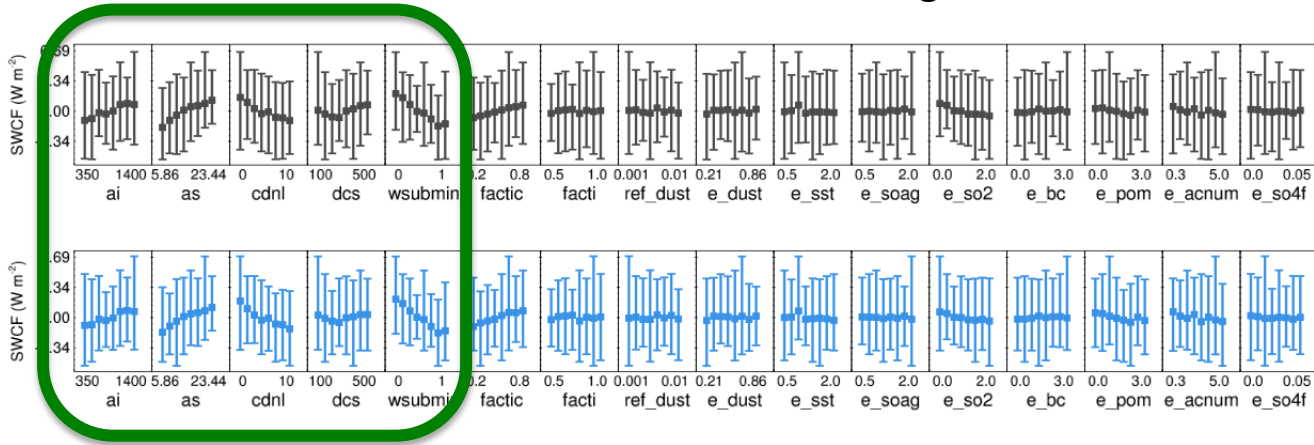
12-member Ensemble Average at Day 10



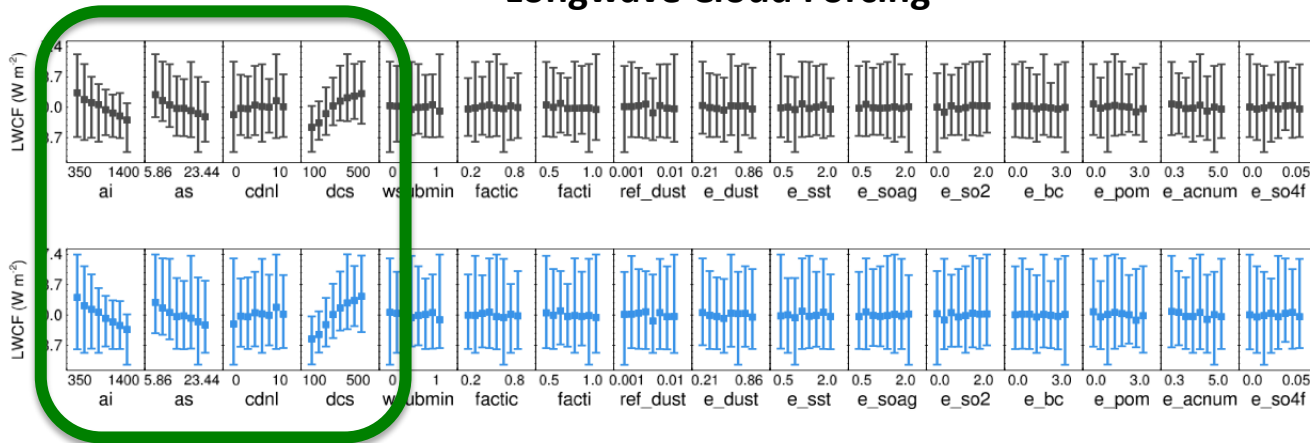
# Cloud Forcing

Black: 4-yr mean; Blue: 12-member ensemble average at day 10

## Shortwave Cloud Forcing



## Longwave Cloud Forcing



# Summary of Example (2)

## Effectiveness

- ▶ Short ensembles correctly reproduces parametric sensitivities of the TOA radiative budget

## Computational efficiency

- ▶ 12 x10-day simulations v.s. 1+4-yr AMIP run
- ▶ Total CPU time: 120 v.s. 1825 days, **a factor of 15**
- ▶ Throughput time: 12 x 256 simulations finished overnight on Yellowstone
- ▶ If more nodes had been available to allow 3000+ simulations to run simultaneously, the entire ensemble UQ experiment could have been completed within 15 minutes!

## The ensemble method

- ▶ Exploits the fact that **fast processes** are an important source of model sensitivities and uncertainties
- ▶ Is very **effective and efficient**
- ▶ Does not address slow modes or slow feedbacks, but
- ▶ Can provide a first-order assessment of model sensitivity at substantially reduced computational cost
- ▶ Can be very useful for speeding up investigations, especially for expensive models/studies

**We plan to test and use the ensemble strategy in other applications**  
(e.g., aerosol lifecycle and climate effects)