## Climate/Earth System Projections and their Uncertainties - An Overview

### Chris E Forest Pennsylvania State University

Uncertainty in Climate Change Research: An Integrated Approach

> NCAR TOY Workshop 2012 August 6-17







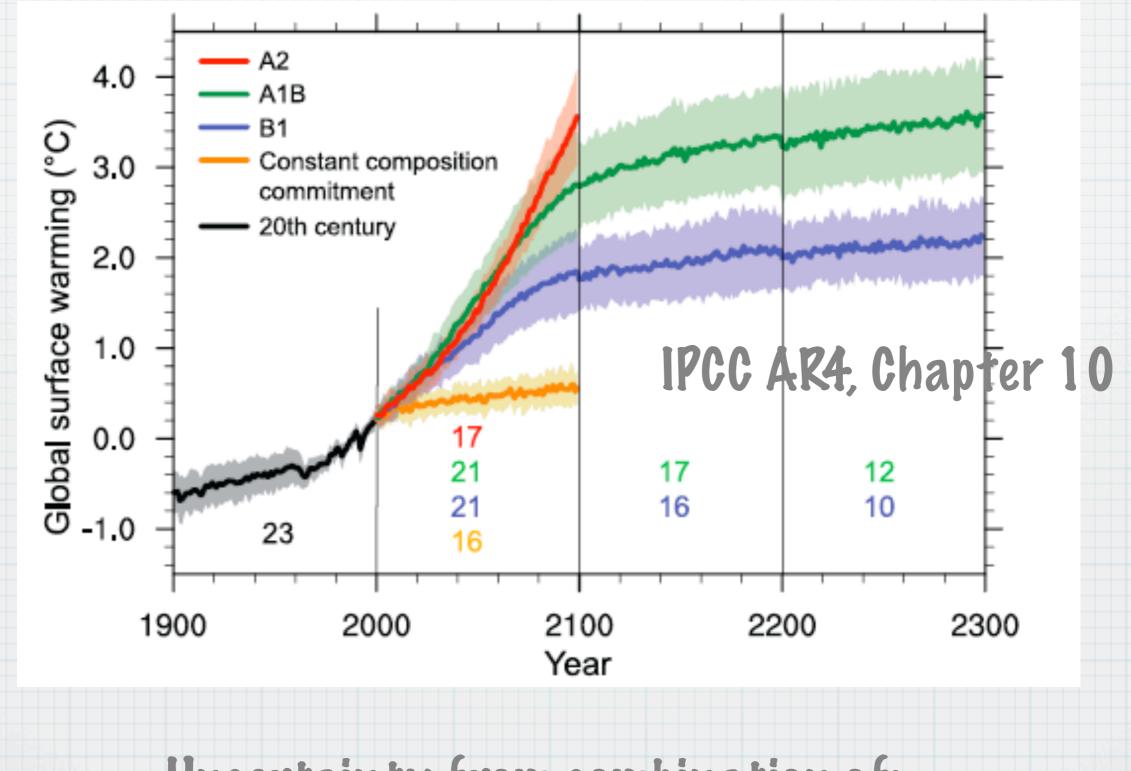
# Day 2 - Climate Talk outline

- \* Sources of uncertainty in climate predictions
- \* Intro to climate model hierarchy
- \* Climate System Response
- \* Characterizing Model uncertainty
- \* Segue to climate science talks



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# What drives uncertainty?

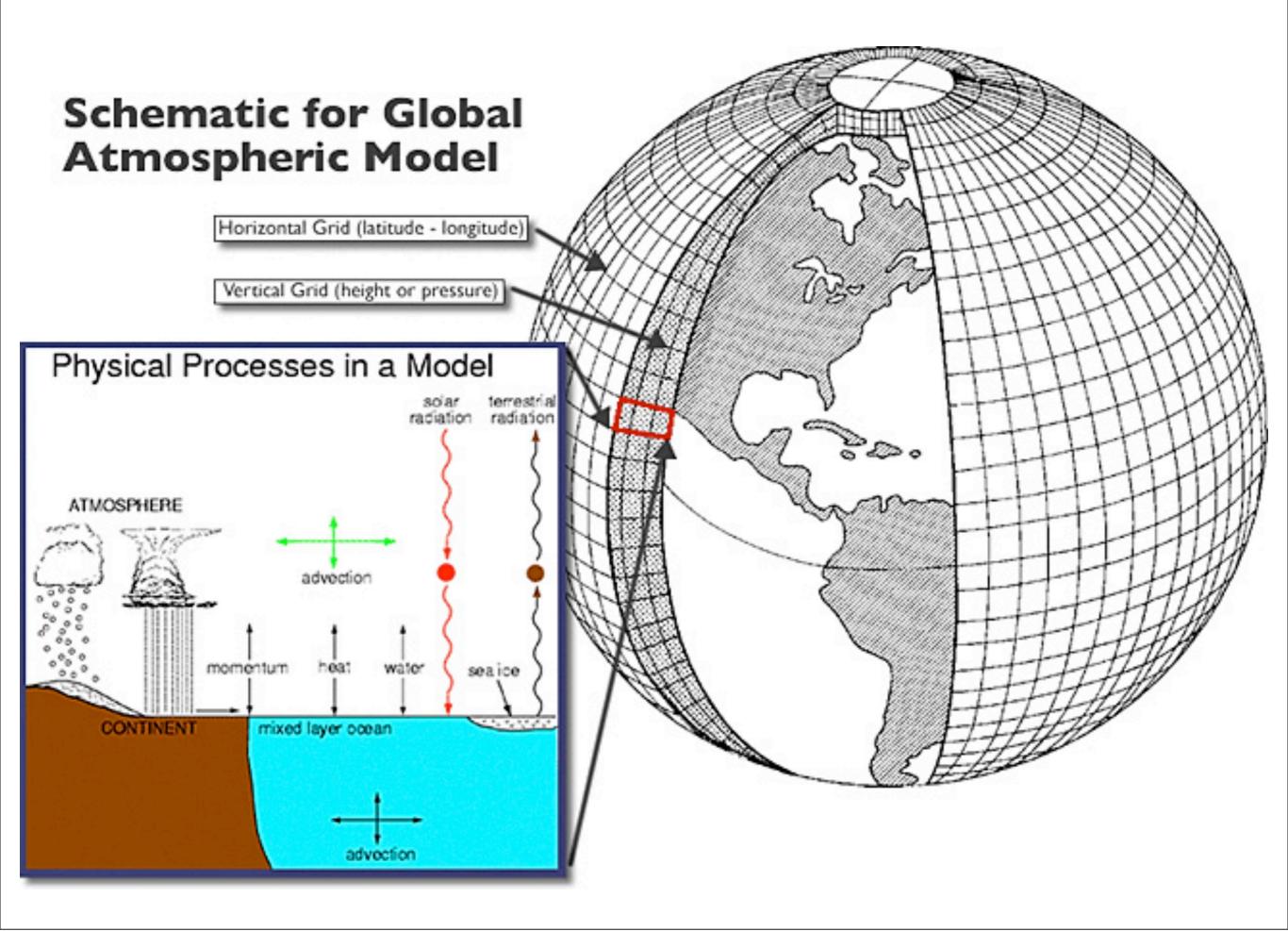




Uncertainty from combination of: Model response and forcings

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# Sources of Uncertainty in Climate Projections

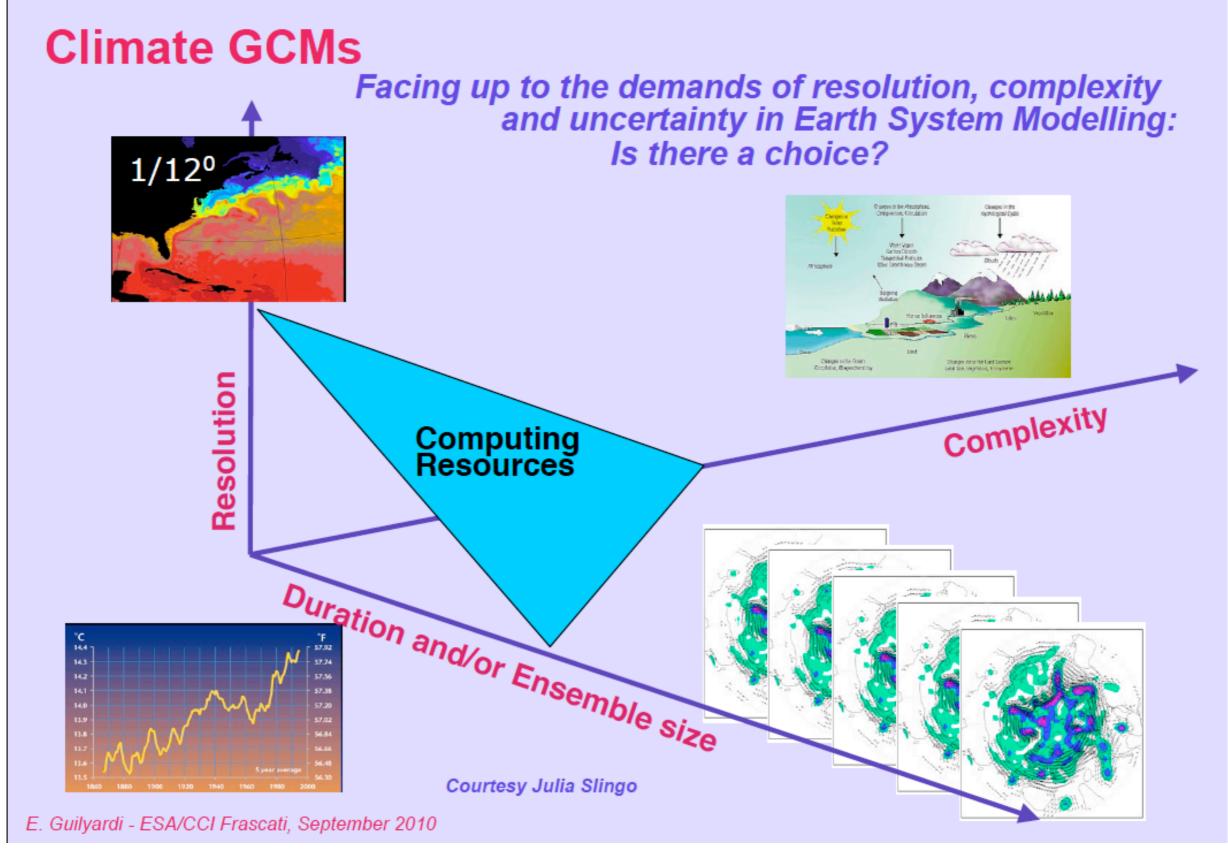


- \* Forcing, Response, Internal Variability
- \* Response time-scales, spatial-scales
- \* How to quantify Uncertainties
  - Forward sensitivity runs
     Inverse estimates of parameters



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### What limits our ability to understand uncertainty in models?



#### Courtesy of Julia Slingo (via Eric Guilyardi)

# Climate Model Hierarchy

- \* Simplest model = Energy Balance Model
- \* EMIC = Earth-system model of Intermediate Complexity
- \* Most complex = Earth System Model
- \* Climate Models are designed for specific purposes and uncertainty analysis is not often one of them.



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## Model Complexity: Components

- \* Atmosphere/Ocean/Land/Ice = Atmosphere-Ocean General Circulation Model := AOGCM
- \* Add: Atmospheric Chemistry, Carboncycle, Vegetation = Earth System Model = ESM
- \* Add Human/Societal dimension = Integrated Earth System Model = iESM



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## Model Complexity: Structure

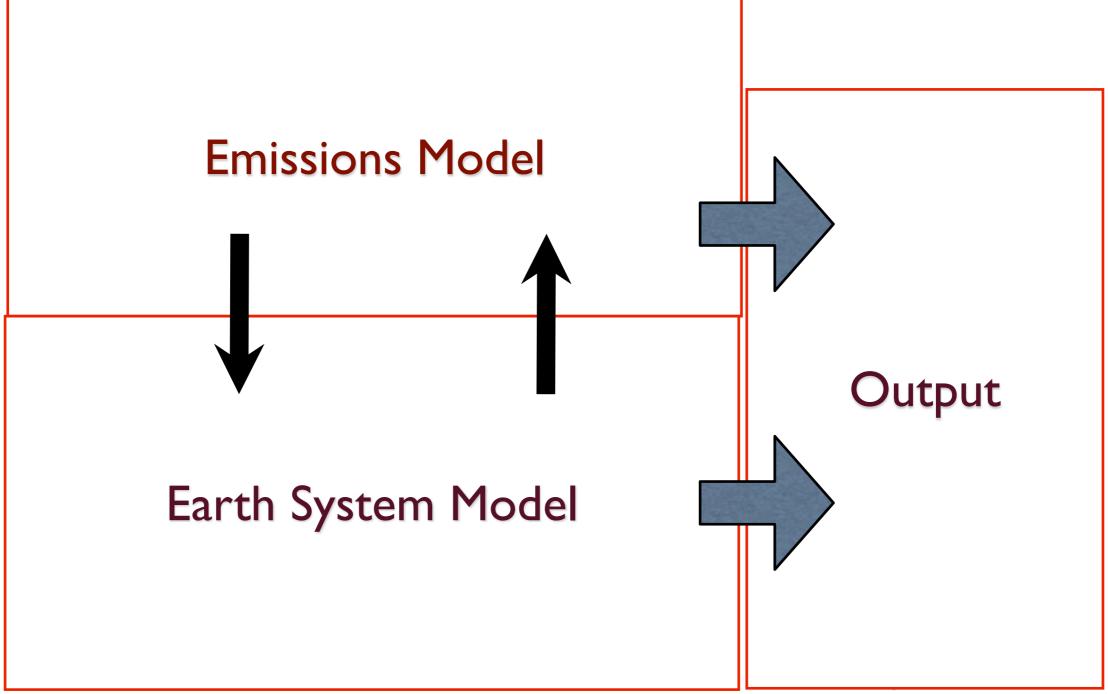


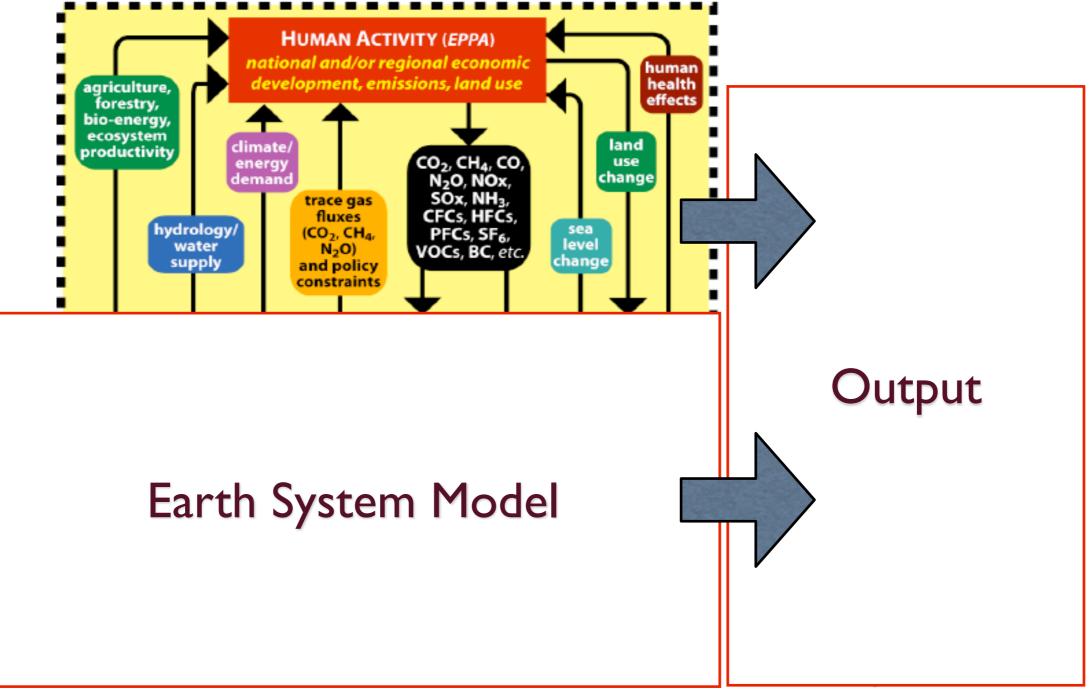
- \* Reduced dimensions (3D model to 2D)
- \* Reduce governing equations
  - \* Conservation of energy, mass, moisture, momentum, angular momentum
- \* Resolution

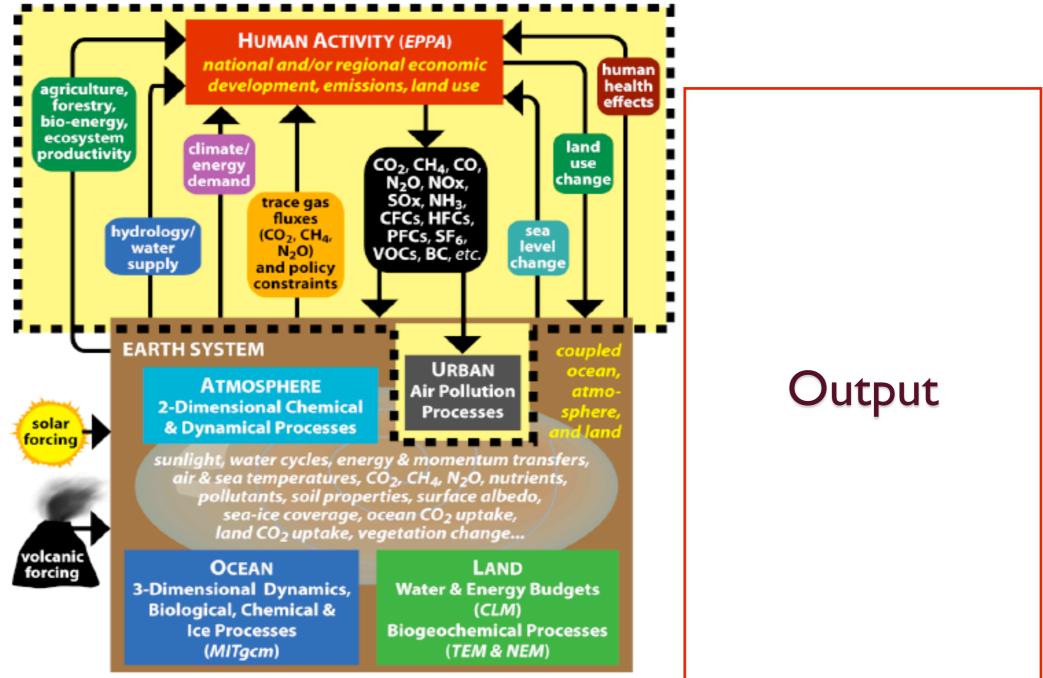


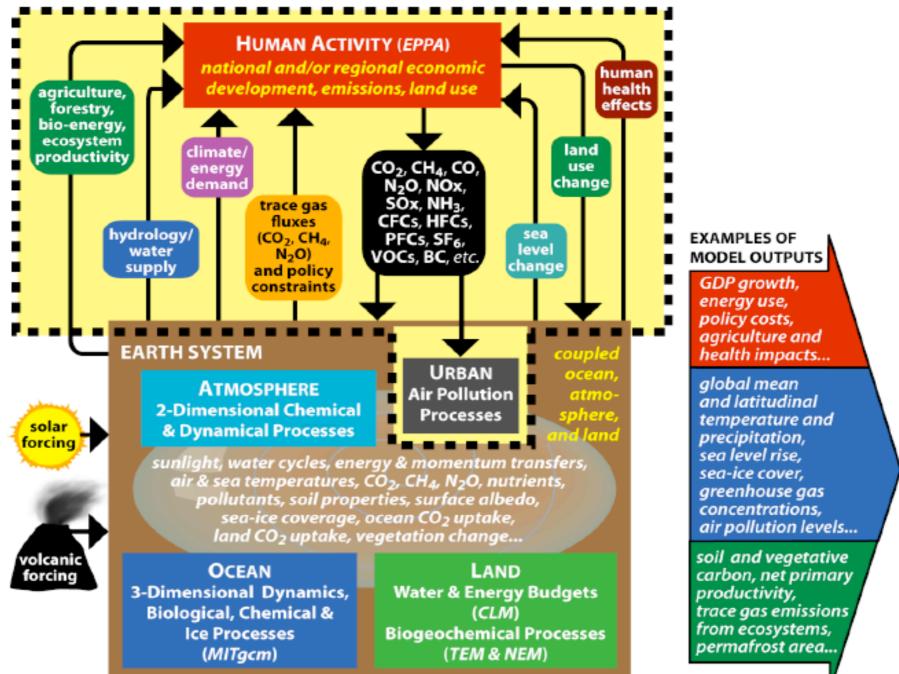
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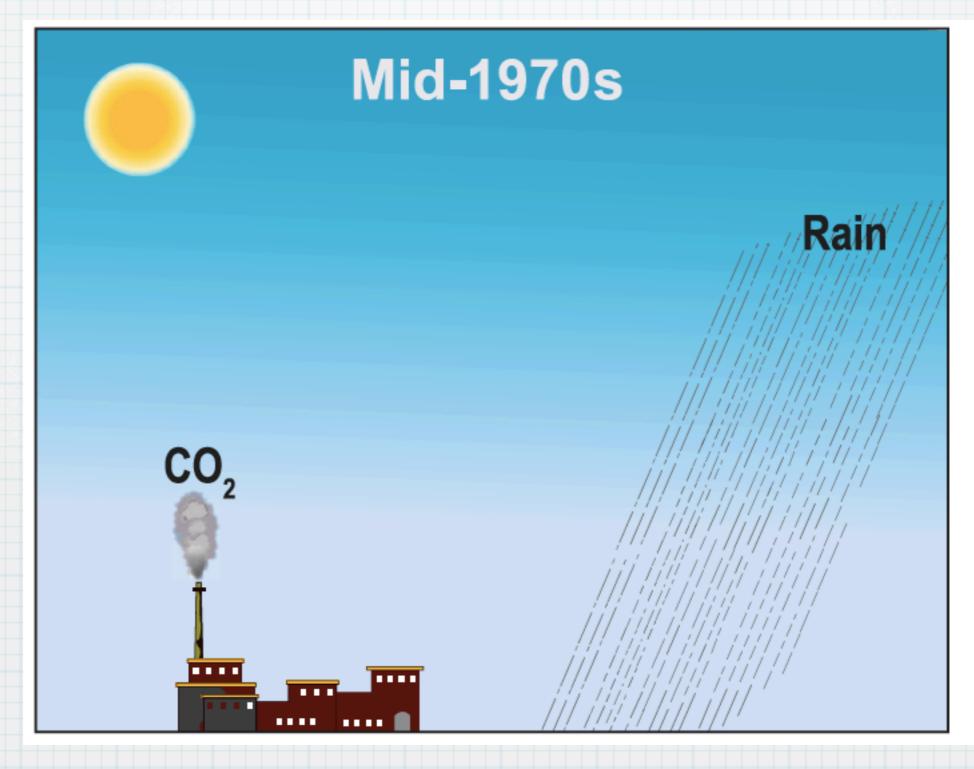






Components/Complexity

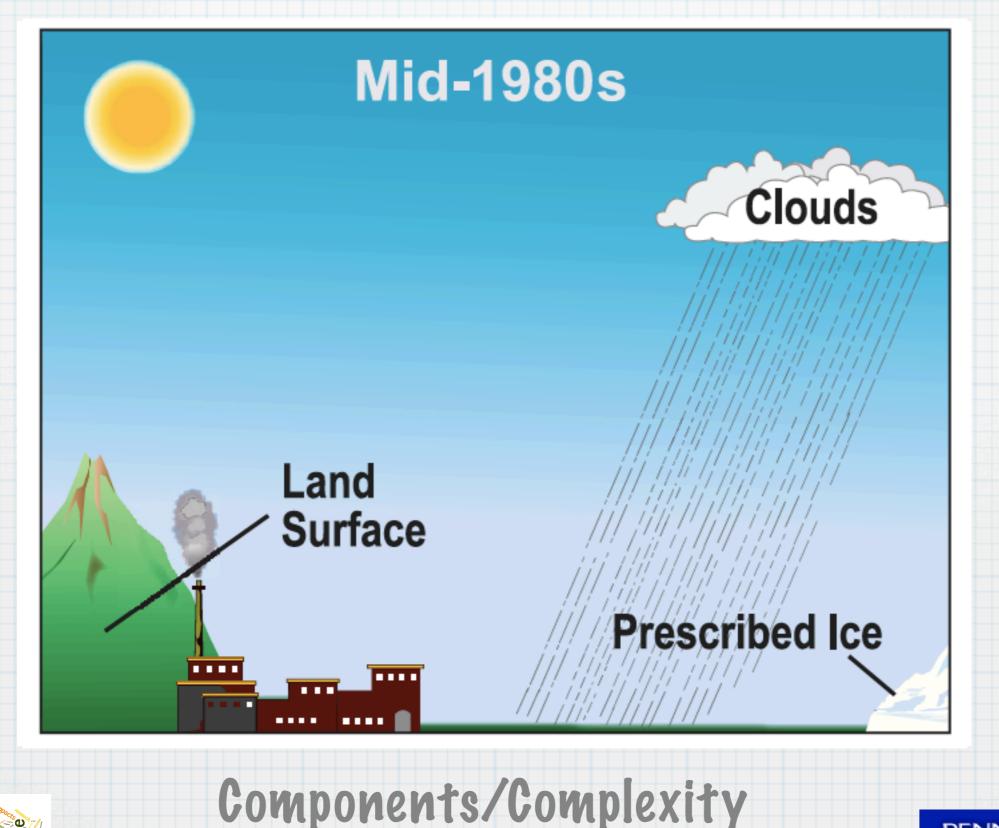




**Components/Complexity** 

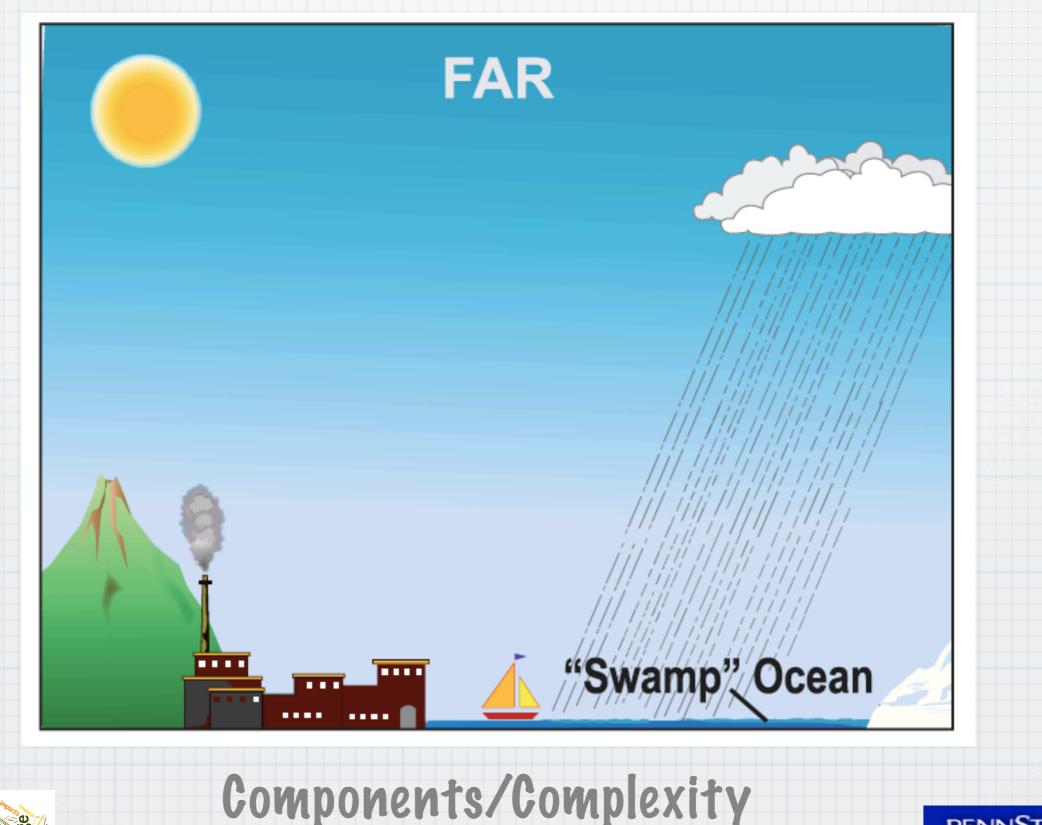


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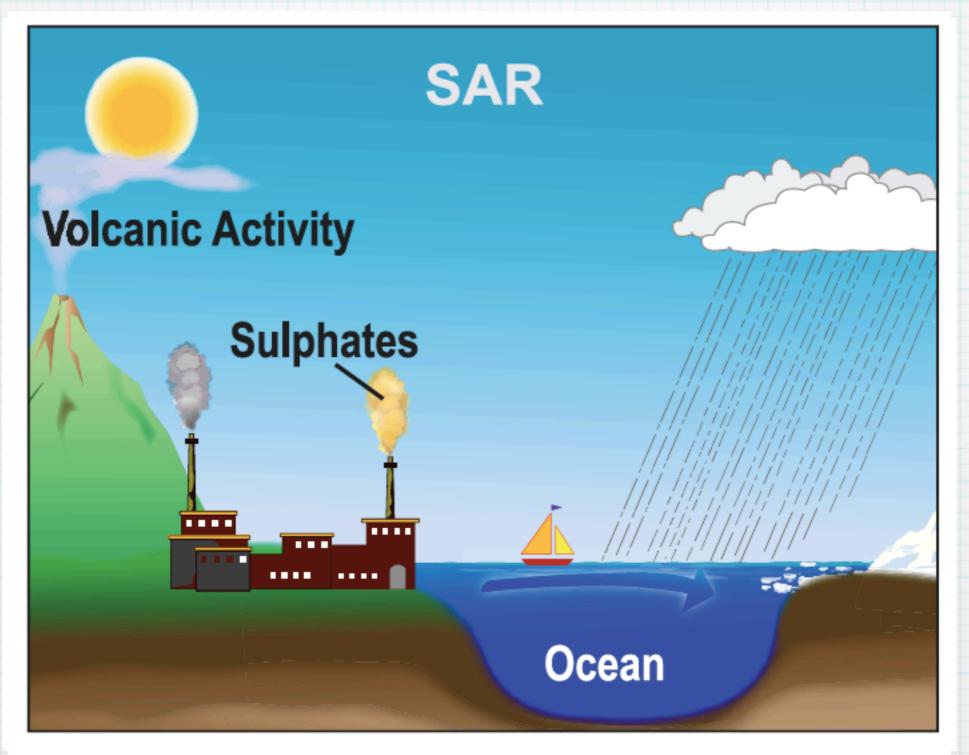


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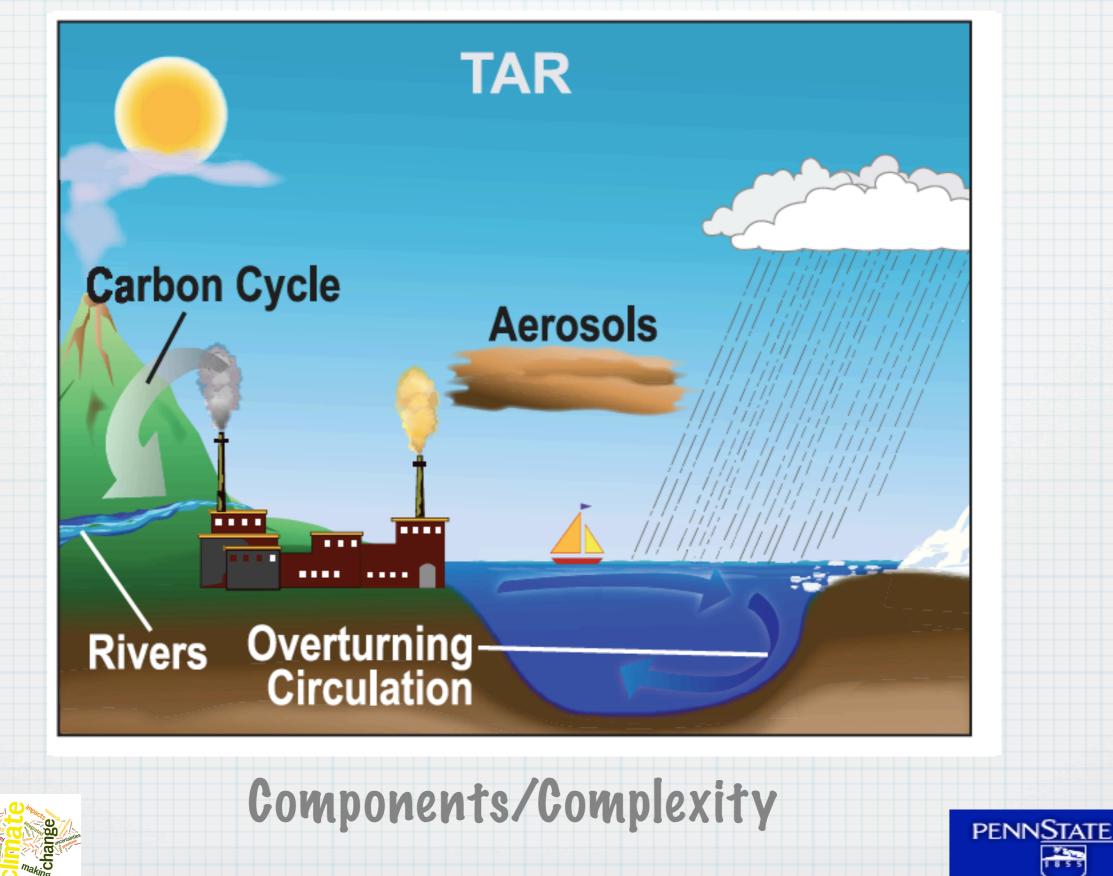
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**Components/Complexity** 



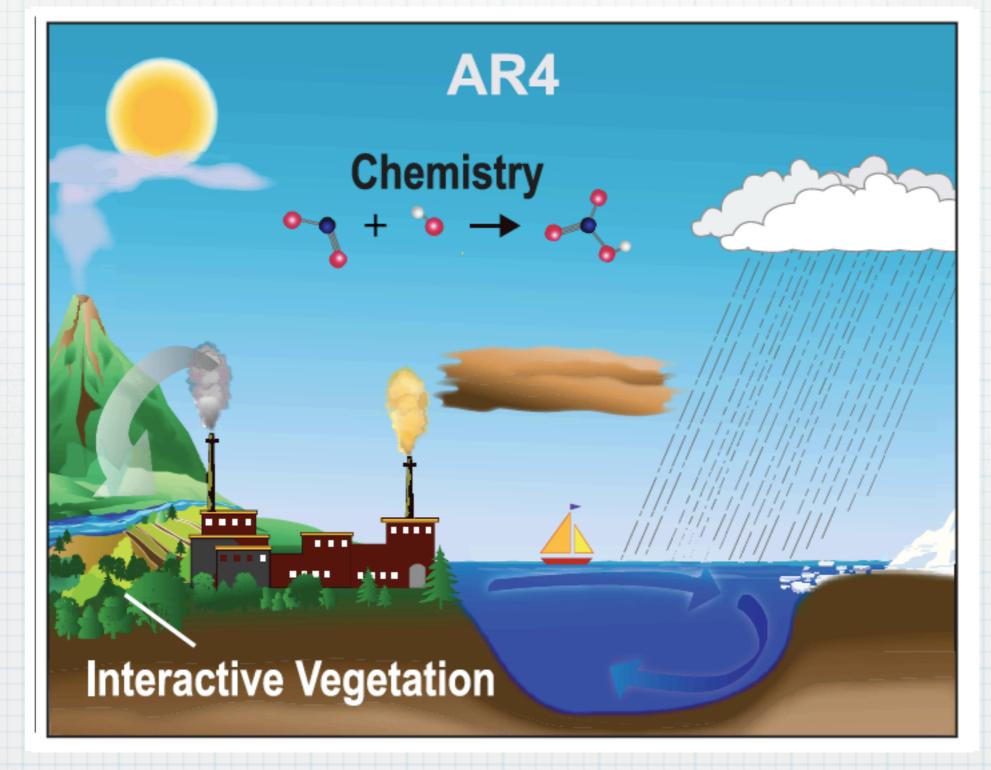
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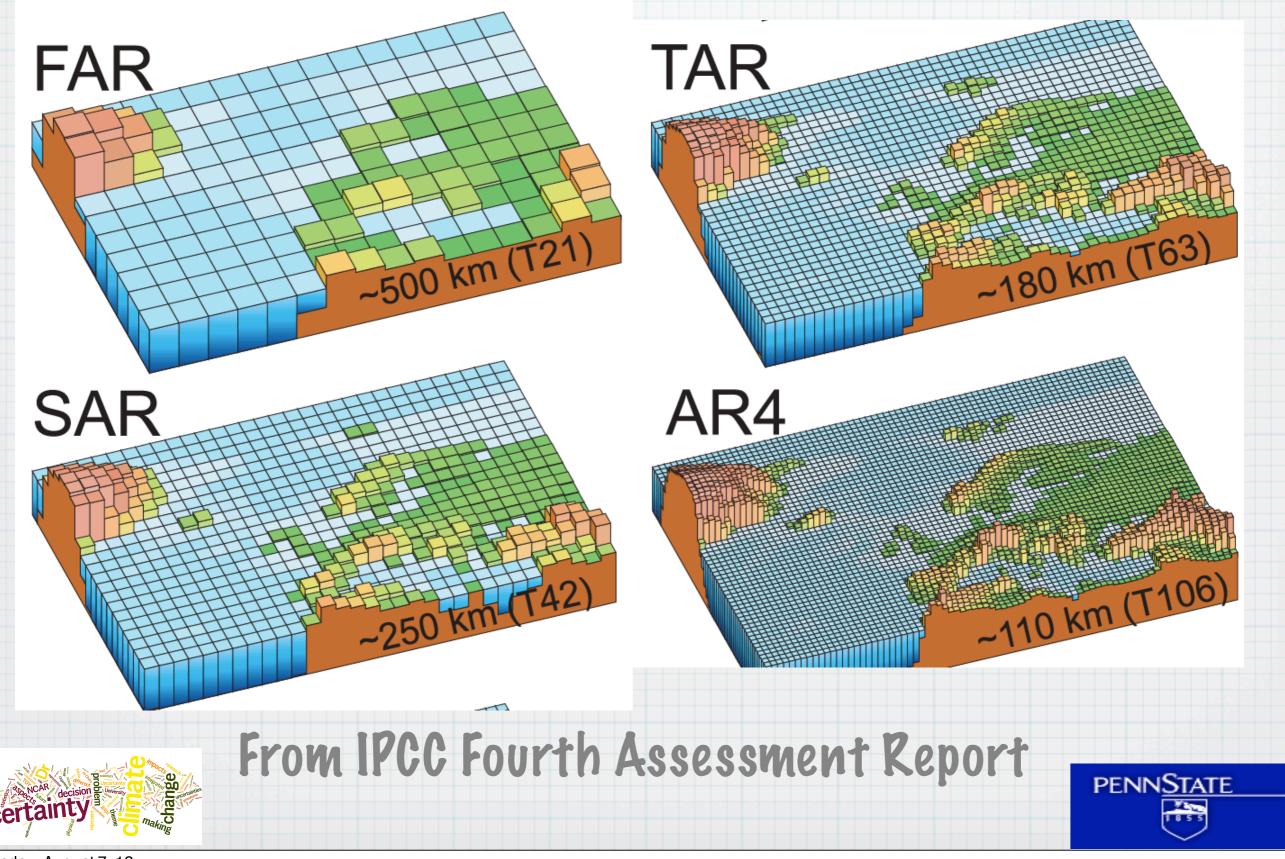












# What matters for long-term climate prediction?



- \* Long-term warming
- \* Delay by ocean
- \* Net forcing

# What matters for long-term climate prediction?

 Controls on:
 Long-term warming
 Pelay by ocean
 Net forcing
 Uncertainties in:
 Climate Sensitivity
 Rate of Ocean Heat Uptake
 Forcing by Aerosols, Carbon-cycle

Consider the energy balance equation for the global-mean surface temperature anomaly  $(\Delta T)$ :

$$c_p \frac{d\Delta T(t)}{dt} = F(t) - \lambda \Delta T(t) - \Phi_o(K_v)$$

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Change in global mean heat content

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Future Forcings

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Future Forcings

Net Feedbacks  $\lambda = 1/S$ 

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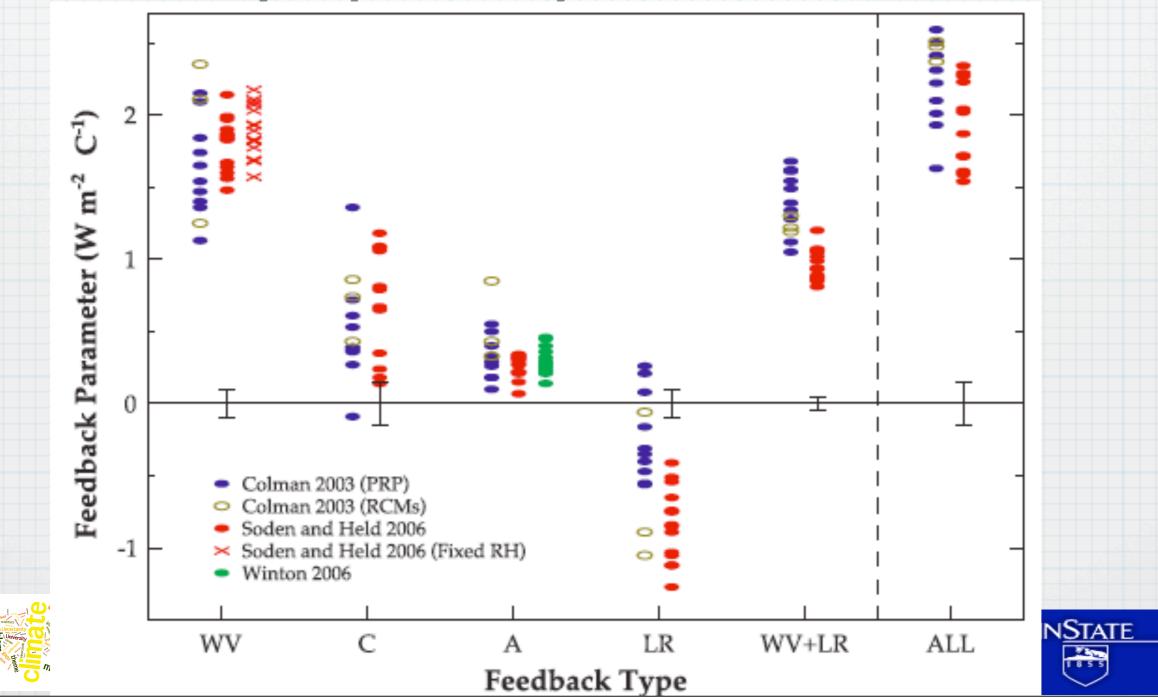
Net Feedbacks 
$$\lambda = 1/S$$

Flux of heat into deepocean

**Conceptually:** This is a good framework for organizing where the uncertainty exists.

In practice: For state-of-the-art models, each uncertainty is an aggregate quantity and cannot be identified with any one specific model component or process.

## Uncertainty in Atmospheric Model Feedbacks \* Uncertainty in Water Vapor, Cloud, Albedo, Lapse Rate, and All Combined



# Characterizing Model Uncertainty



- \* Perturbed Physics Ensemble (PPE)
  - (Or Perturbed Parameter Ensemble)



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# Model Intercomparison Projects = MIPs



- \* Each group creates its "best" model
- \* Samples Structural Uncertainty due to model development choices





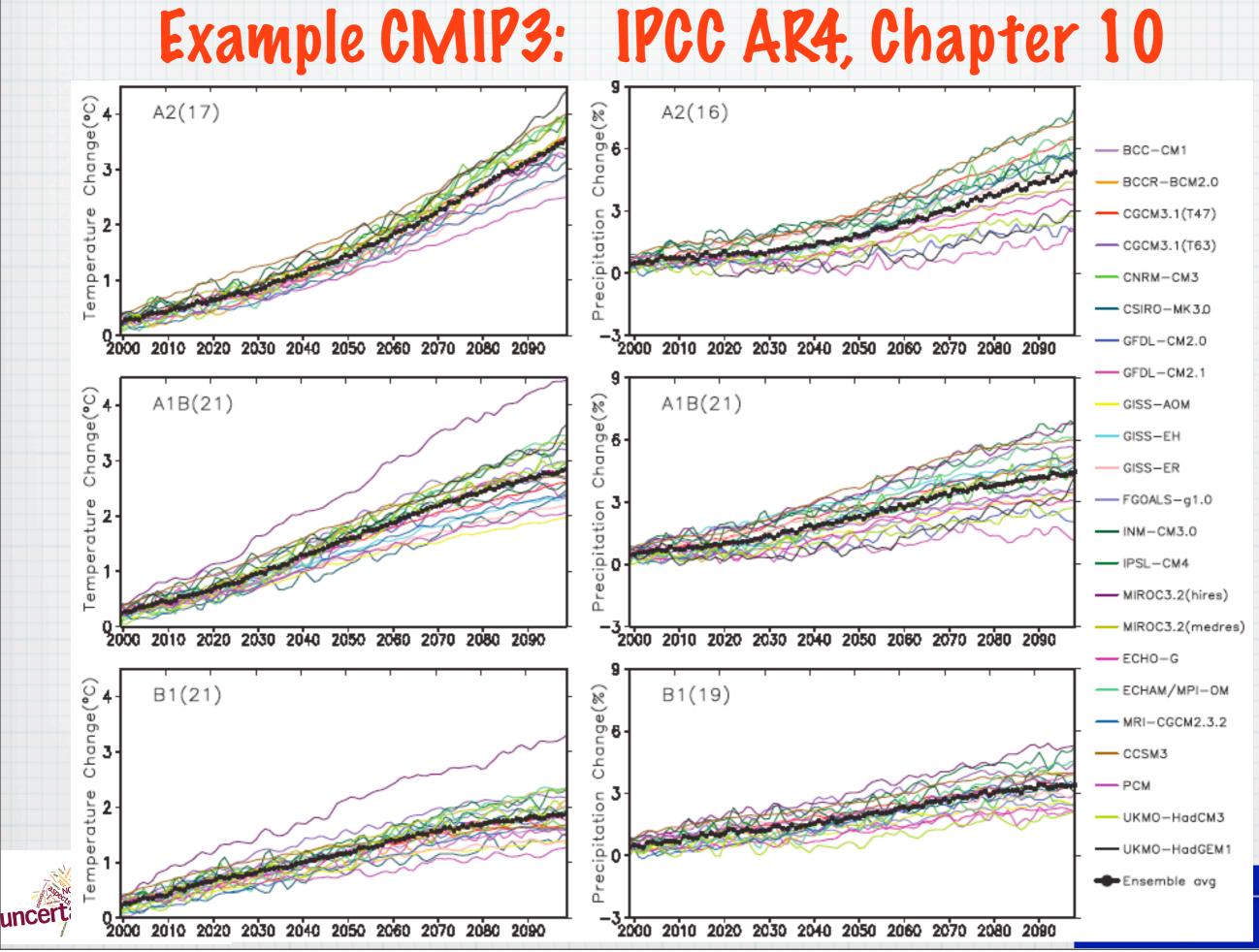
# Model Intercomparison Projects = MIPs

\* Examples:
\* AMIP = Atmospheric-GCM MIP
\* CMIP = Coupled-AOGCM MIP
\* PMIP = Paleoclimate MIP
\* CFMIP = Cloud Feedback MIP
\* CMIP1, CMIP2, CMIP3, CMIP5, ....
\* New models, new MIP.

\* Program for Climate Model Diagnostics and Intercomparison = PCMDI







# Perturbed Physics Ensembles (PPE)

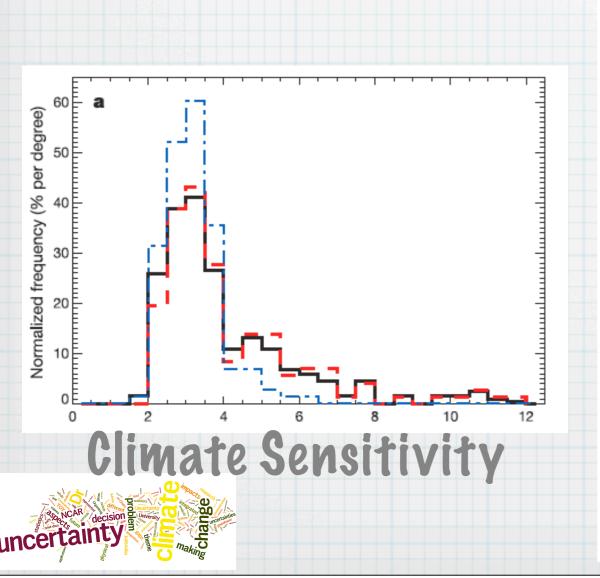
- \* Single-model framework
  - \* Individual model parameters varied to adjust physics components
- Samples parametric uncertainty leading to model response uncertainty
- \* Focus on perturbing feedbacks in the climate system

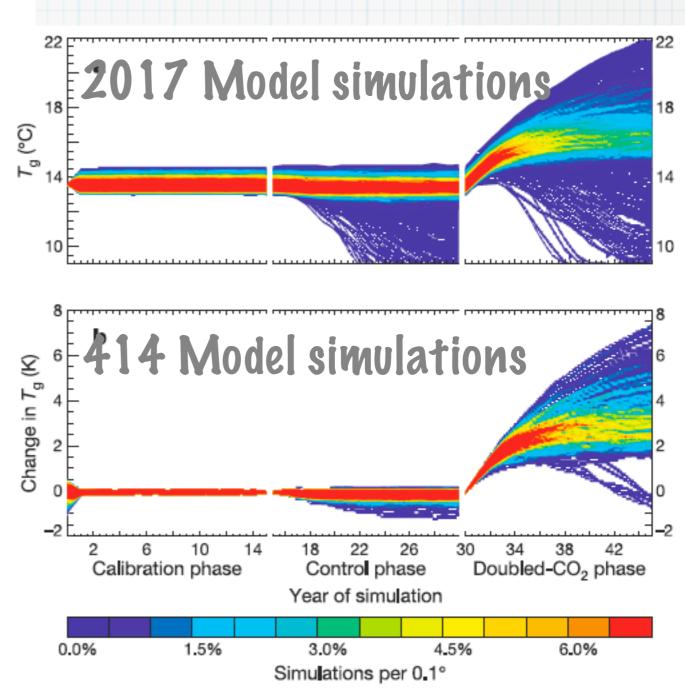


# Perturbed Physics Ensembles (PPE)

#### Uncertainty in predictions of the climate response to rising levels of greenhouse gases

D. A. Stainforth<sup>1</sup>, T. Aina<sup>1</sup>, C. Christensen<sup>2</sup>, M. Collins<sup>3</sup>, N. Faull<sup>1</sup>, D. J. Frame<sup>1</sup>, J. A. Kettleborough<sup>4</sup>, S. Knight<sup>1</sup>, A. Martin<sup>2</sup>, J. M. Murphy C. Piani<sup>1</sup>, D. Sexton<sup>3</sup>, L. A. Smith<sup>5</sup>, R. A. Spicer<sup>6</sup>, A. J. Thorpe<sup>7</sup> & M. R. Allen<sup>1</sup>





**Figure 1** Frequency distributions of  $T_g$  (colours indicate density of trajectories per 0.1 K interval) through the three phases of the simulation. **a**, Frequency distribution of the 2,017 distinct independent simulations. **b**, Frequency distribution of the 414 model versions. In **b**,  $T_g$  is shown relative to the value at the end of the calibration phase and where initial-condition ensemble members exist, their mean has been taken for each time point.

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# Returning to Simpler Models

- \* Example of Uncertainty Propagation using the IGSM2 System
- Sample uncertainty across multiple components





## Included Uncertainties

## Emissions Uncertainty from MIT EPPA4

 Population: 6-13 billion, Energy Resources, Efficiency/ Technology

## Climate System Response

(Calibrated in Forest et al. 2008)

- Climate Sensitivity
- Rate of Heat uptake by Deep Ocean
- Radiative Forcing Strength of Aerosols
- Carbon Cycle Uncertainty:
  - CO<sub>2</sub> Fertilization Effect on Ecosystem
  - Rate of Carbon Uptake by Deep-Ocean

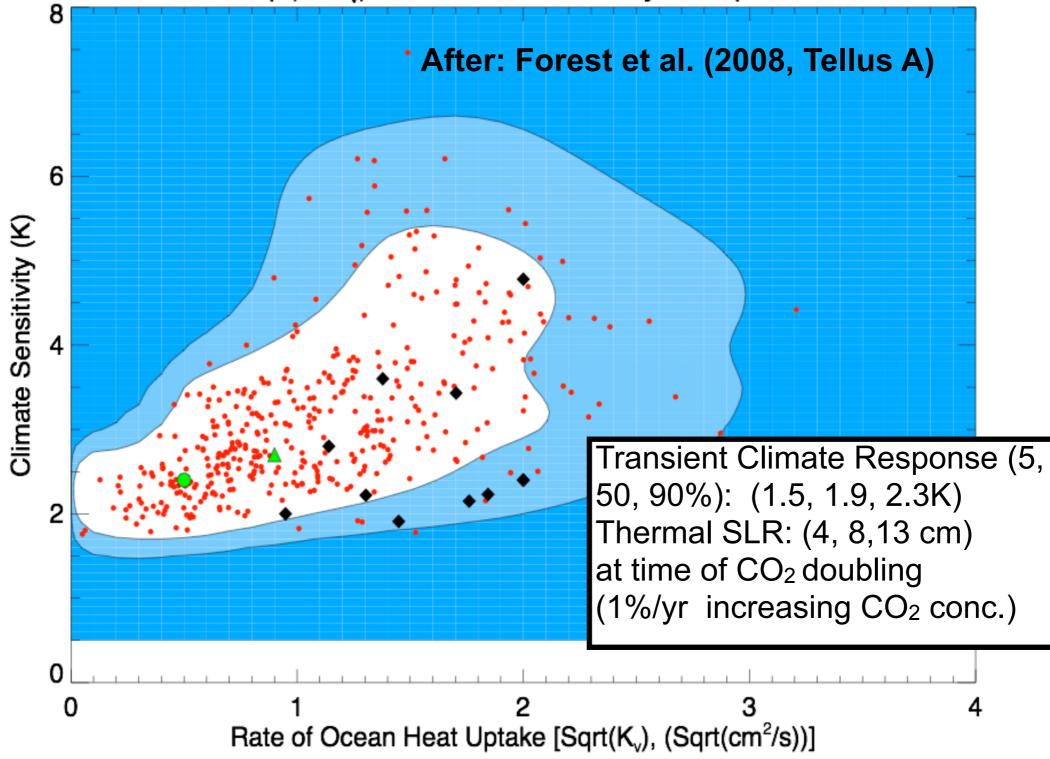
## • Trends in Precip. Freq. on $CH_4 + N_2O$

(Statistics scaled using by AR4 model trends)

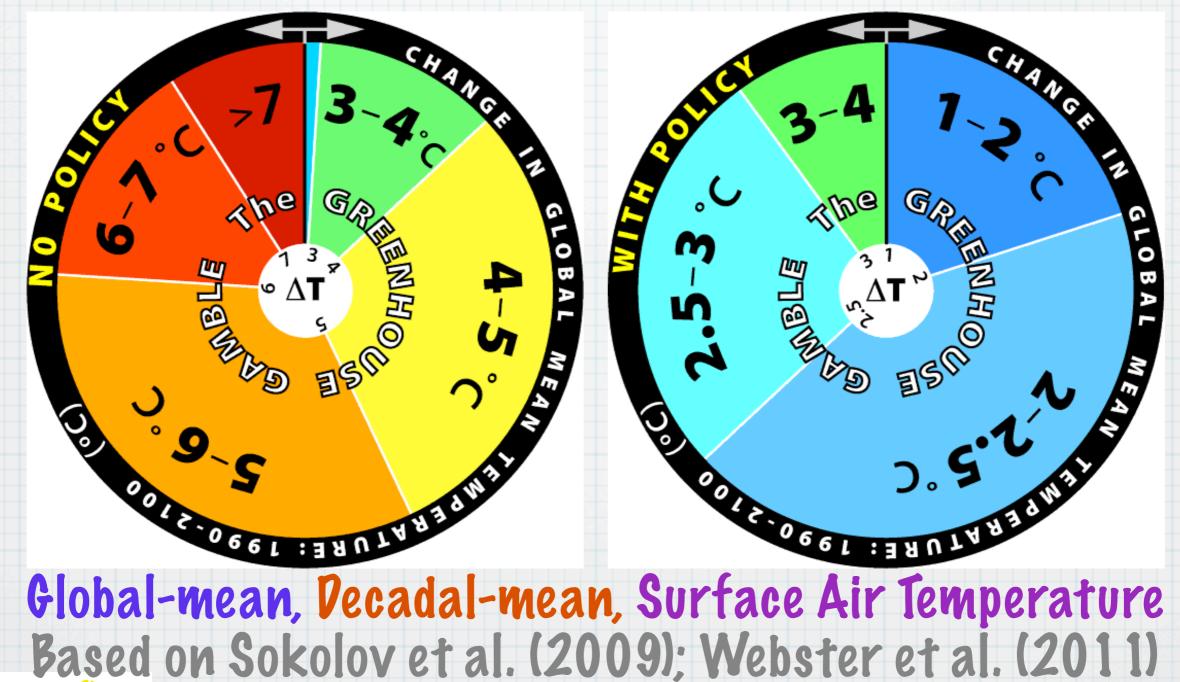
## Climate Sensitivity and Ocean Heat Update Consistent with

## Observations

p(S,K<sub>v</sub>): IGSM2 Uncertainty Sample



## Uncertainty in Global Climate Response in 2100





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# The Uncertainty Cake





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