

# Projecting future climate: methods, limitations, and challenges

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THIS TALK APPROVED



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# Why do we need climate models?

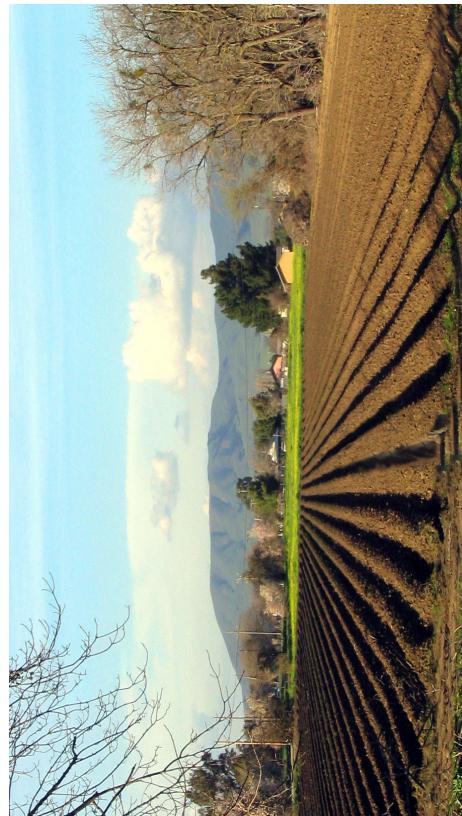
To understand effects of human influences on climate



Greenhouse gases



Particulate pollution



Surface properties



Surface properties

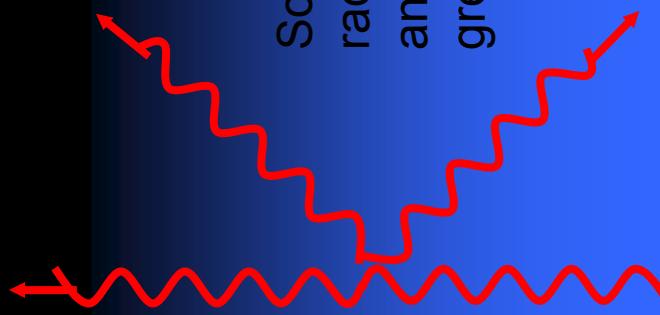
# Earth's radiation balance



Some solar radiation is reflected by the atmosphere

Atmosphere

Some solar radiation is reflected by the Earth



Some infrared radiation is absorbed and re-emitted by greenhouse gases

Most solar radiation is absorbed by the Earth

Infrared radiation is emitted by the Earth

Earth





What are climate models?

Climate models are large computer programs that simulate the atmosphere ocean, sea ice, etc.

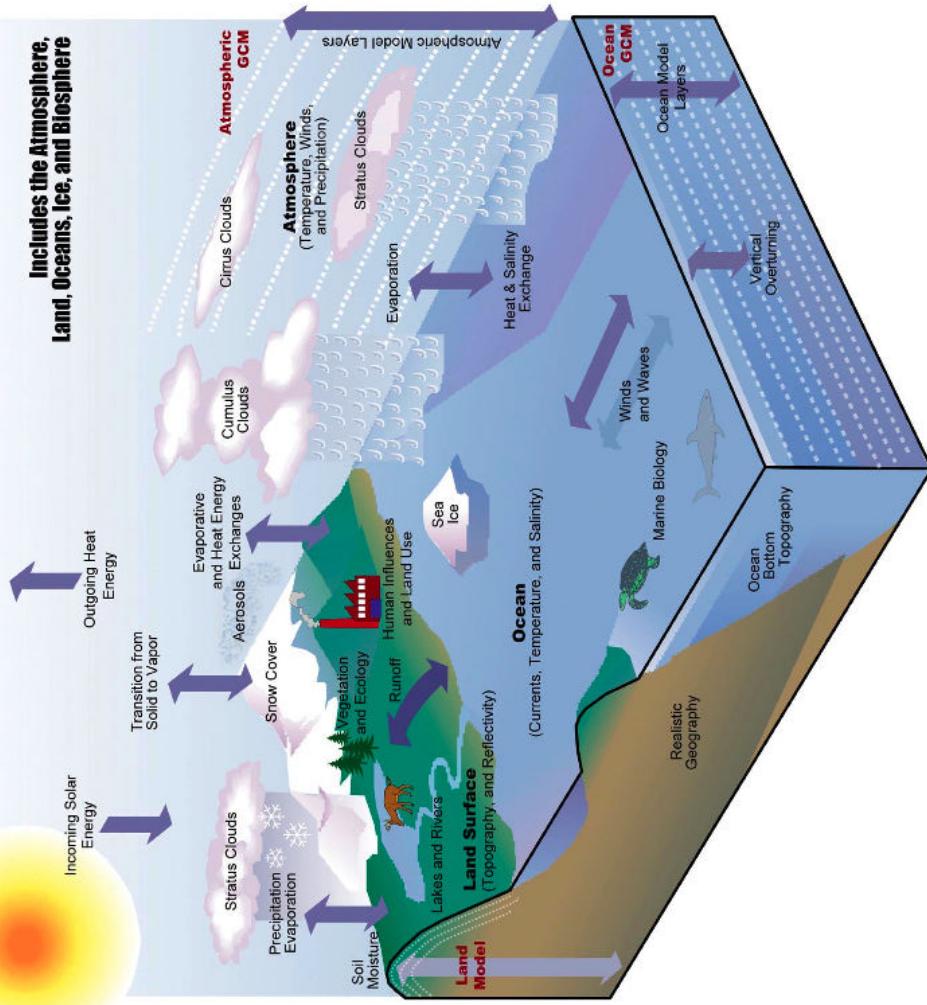
## Time scales:

Atmosphere - days

Sea ice – days to centuries  
Vegetation – days to centuries

Oceans – months to centuries  
Ice sheets - years

## Modeling the Climate System



# Atmospheric models solve differential equations

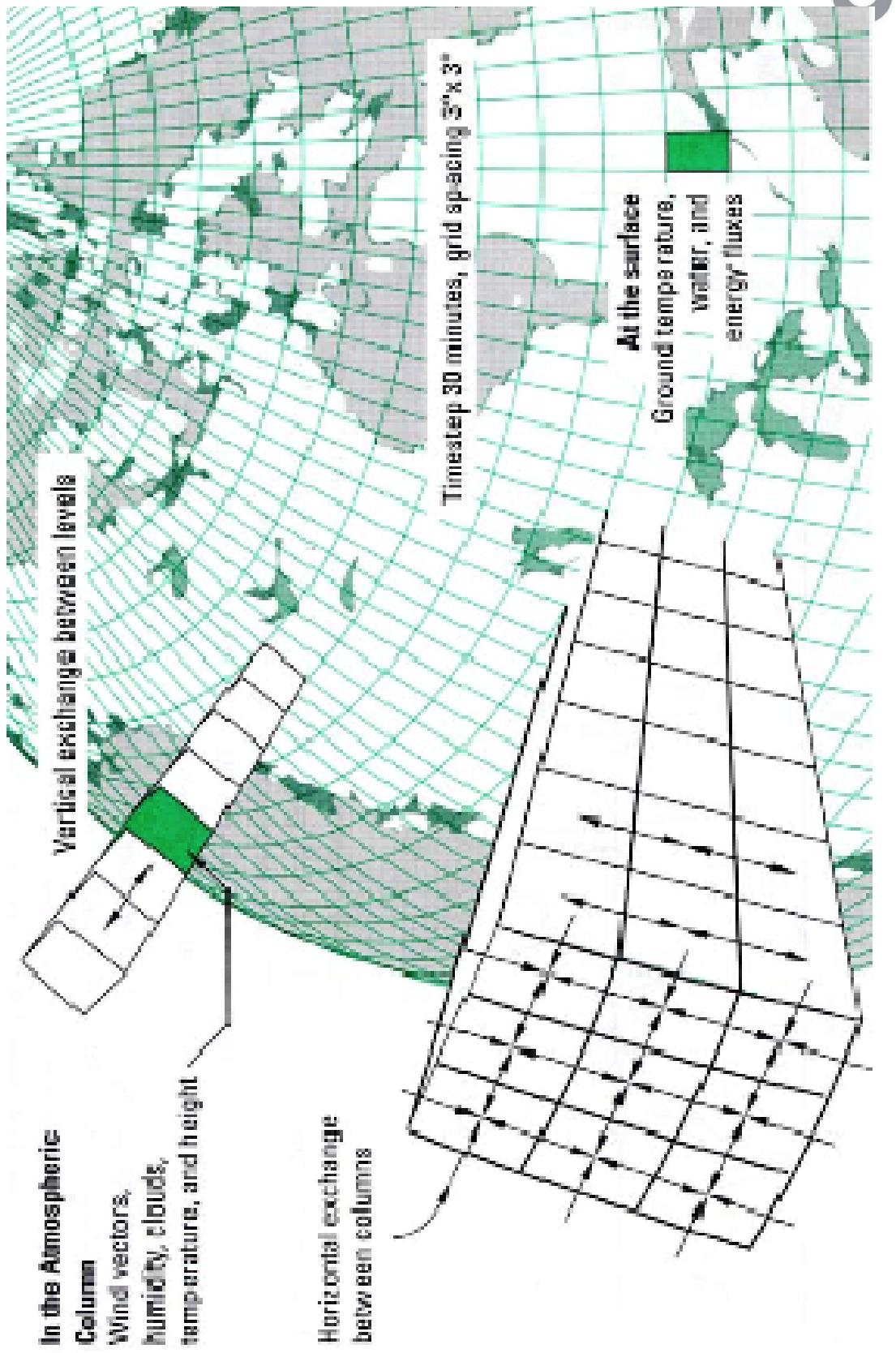
Conservation of momentum: $D\mathbf{v}/Dt = -2 \Omega \times \mathbf{v} - \mathbf{grad}(\rho) / \rho + \mathbf{g}$	Conservation of mass: $\partial_t \rho + \mathbf{div}(\rho \mathbf{v}) = 0$
Conservation of (thermal) energy: $c_v D T / Dt = -\rho (d\rho^{-1} / dt) + Q$	Equation of state: $\rho = \mu \rho / (R T)$

Unknowns:	Parameters:
$\rho$ = density $\rho$ = pressure $\mathbf{v}$ = velocity (3 components) $T$ = temperature	$\Omega$ = Coriolis parameter $\mathbf{g}$ = gravitational acceleration $Q$ = "heating rate" $c_v$ = volume heat capacity $R$ = gas constant $\mu$ = molecular weight

+ tracer-conservation law ( $q$  for atmosphere,  $S$  for ocean)  $\Rightarrow 7$  equations in 7 unknowns



# Climate models divide the world into little boxes (actually, not so little)





# Clouds: the Achilles heel of climate models





# Why are clouds hard to model?

## Clouds

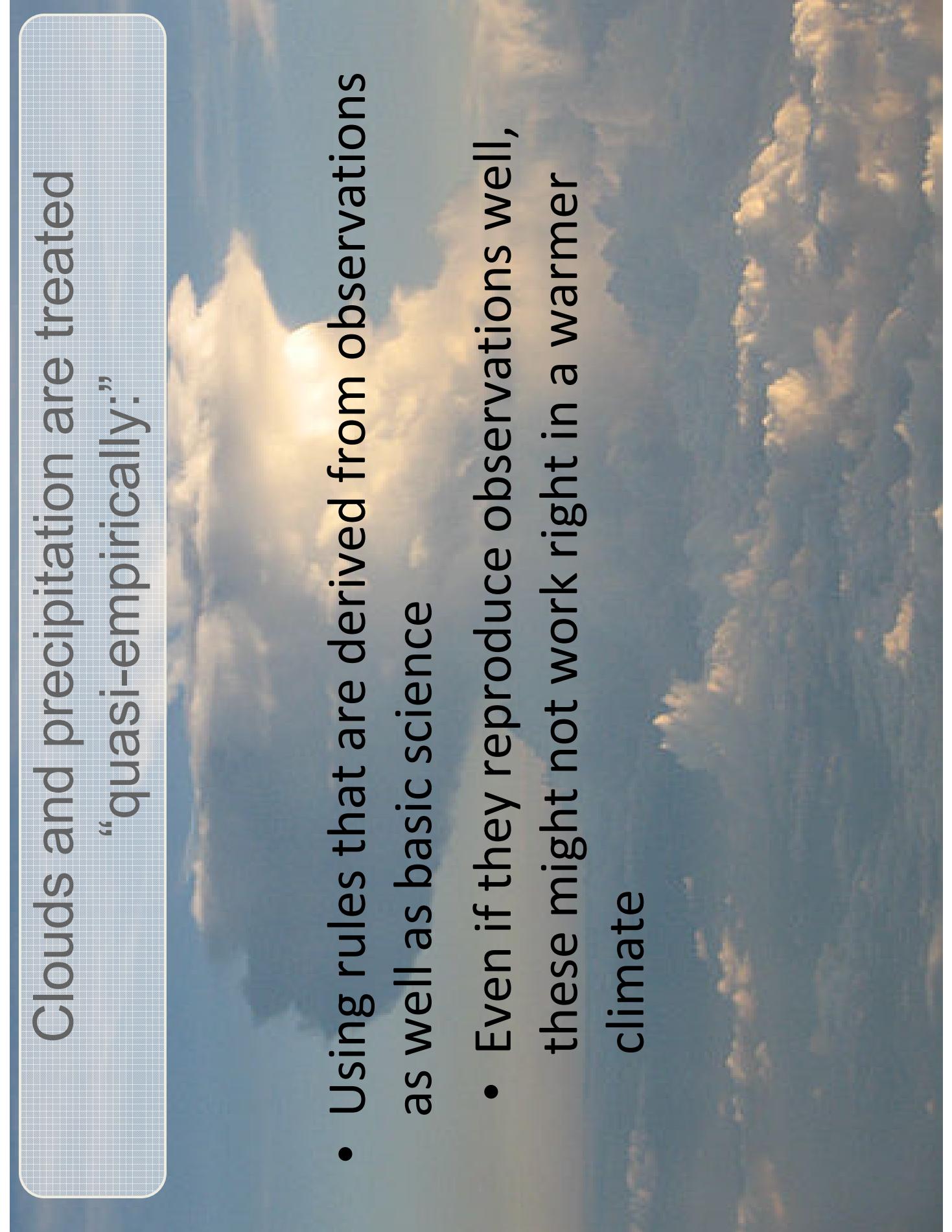
- Are smaller than climate model grid boxes
- Are not well-understood
- Respond in unknown ways to increasing greenhouse gases and other climate insults



“Computers only tell you what you  
already know.”



Ernesto Colnago



Clouds and precipitation are treated  
“quasi-empirically”:

- Using rules that are derived from observations as well as basic science
- Even if they reproduce observations well, these might not work right in a warmer climate



**How well do climate  
models work?**

# Climate simulation by Warren Washington, circa 1969

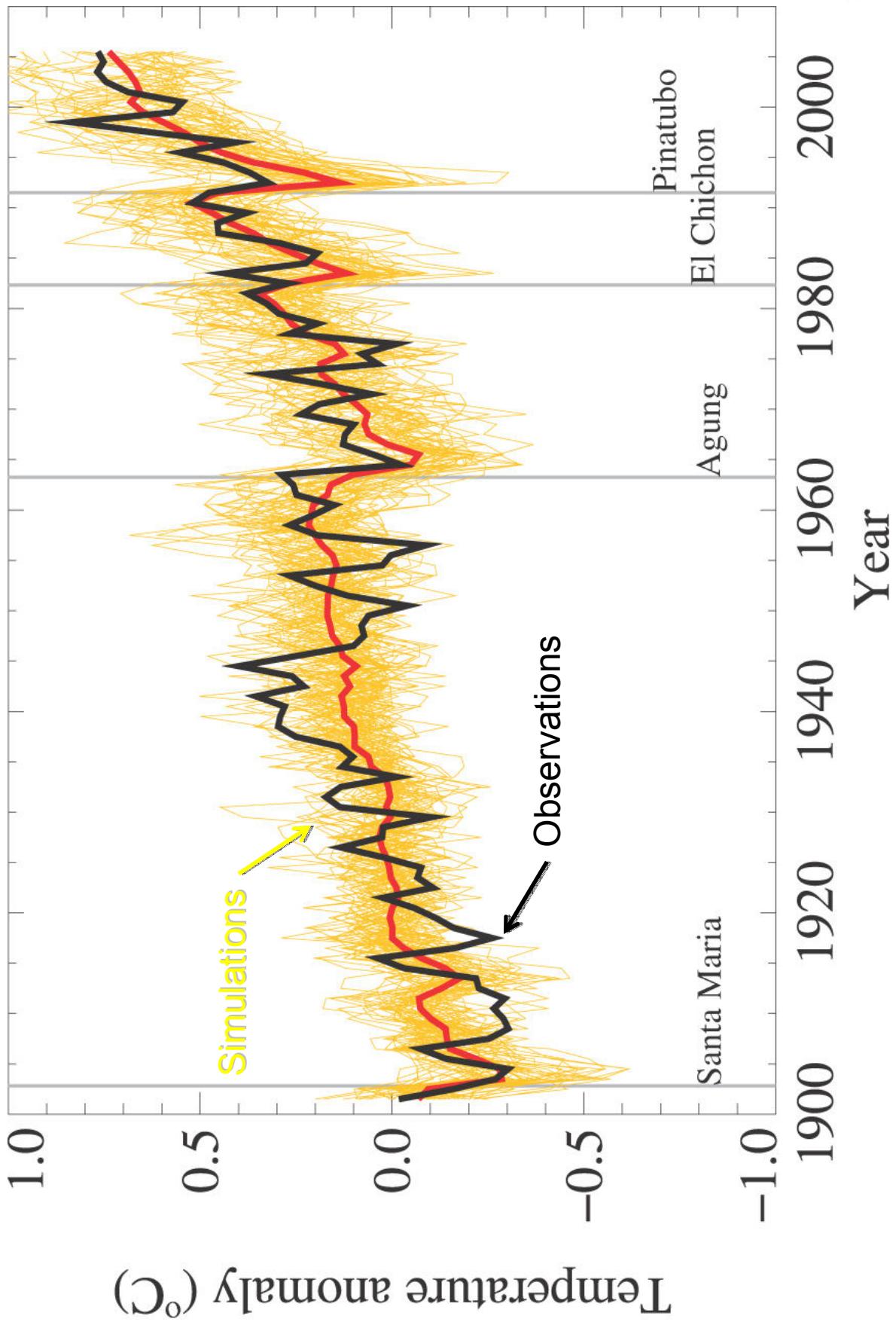


# Climate model evaluation

- Models are thoroughly evaluated
  - They are not perfect, but we know their flaws
  - The naysayers who claim that climate models are not evaluated are not telling the truth.
- We evaluate models by comparing to the past, and hope that this tells us how well they predict the future.
  - It is difficult to directly evaluate the predictions of climate models (unlike weather models).

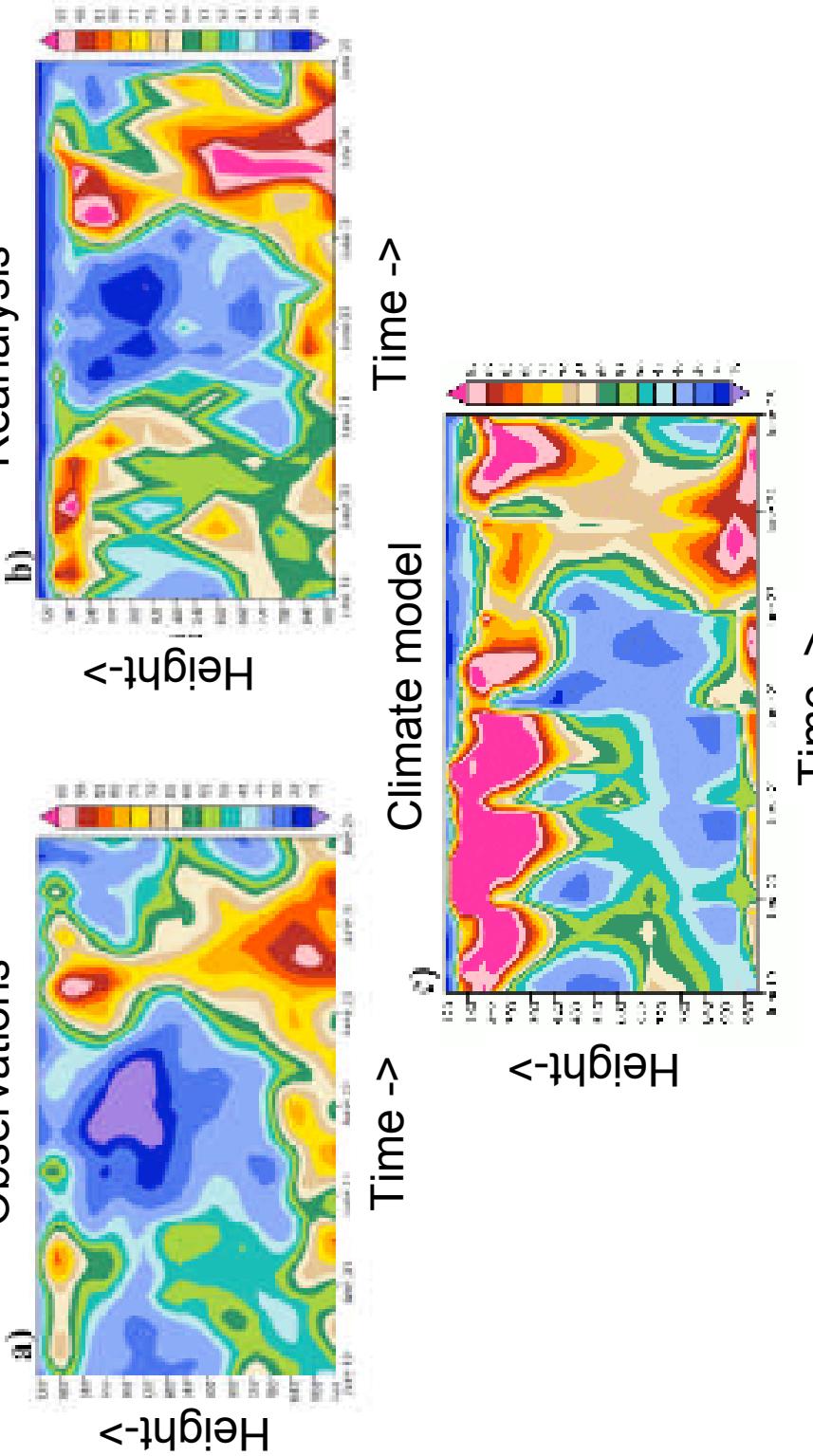


Models reproduce the 20<sup>th</sup> century  
pretty well



# We evaluate climate models by using them to forecast weather

“Reanalysis”

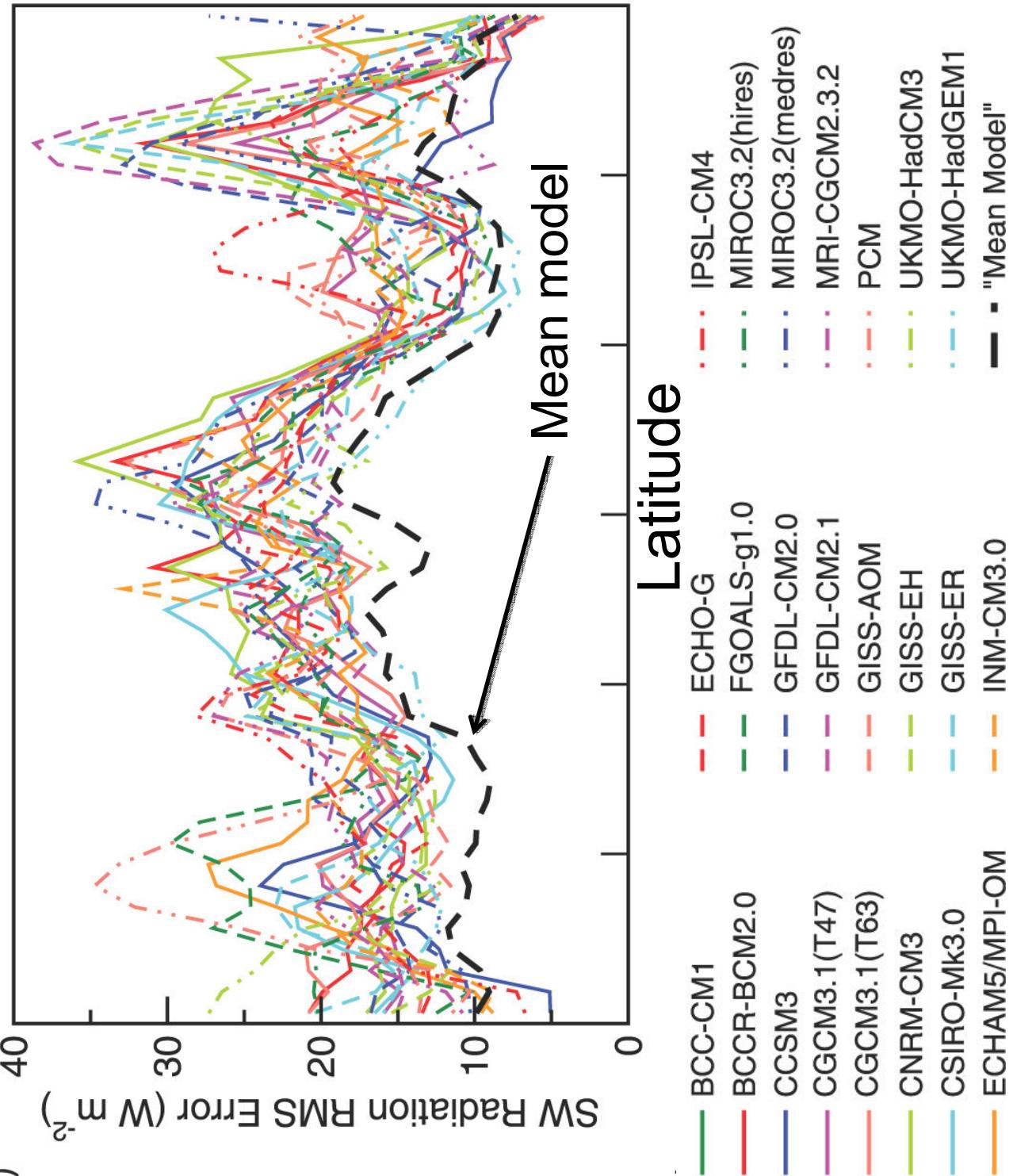


**FIGURE 4** Panels (a)–(d) vertical profile of atmospheric stability (CAPE) at the JRKV SGP site on June 18, 1993, for the period from 0000–0600 UTC, as obtained from (a) ARM observational retrievals, (b) the NCAR reanalysis, (c) a sequence of GFDL forecasts that are initialized at 00Z, each day, and valid for the 1200 UTC, (d) forecasts initialized at 00Z, each day, using the GFDL climate model initialized at 00Z, each day, with the relative humidity profile initialized at 00Z, each day. Note, the vertical dimension (CAPE) is identical to the relative humidity profile in (c), its stability evolution will be identical to the one shown in (a).



# RMS errors in simulated outgoing solar radiation

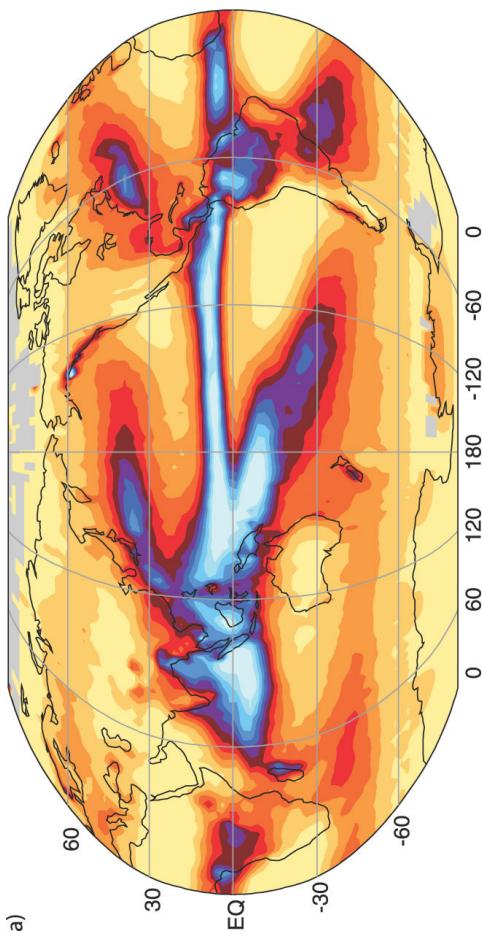
a)



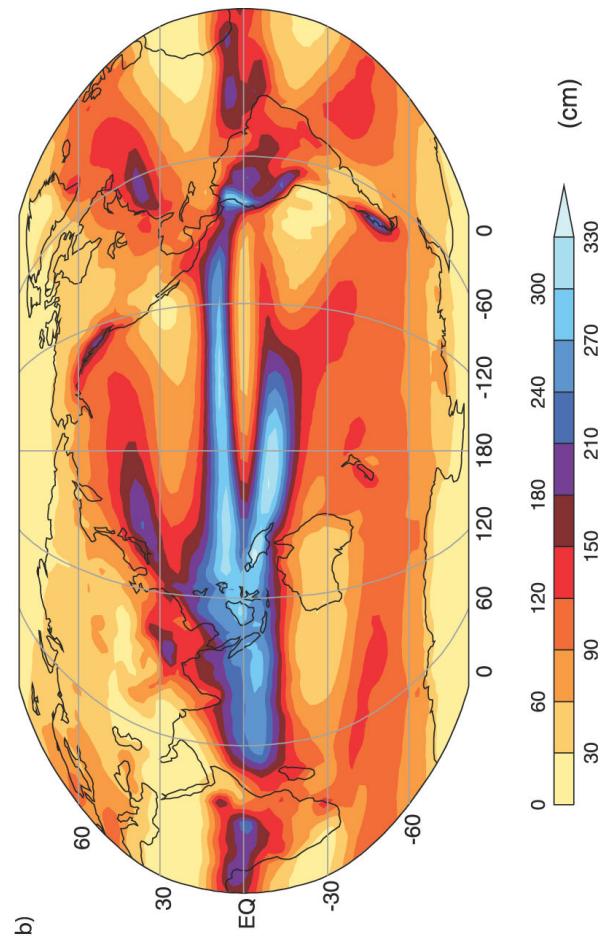
Global climate models do well on  
the global scale...



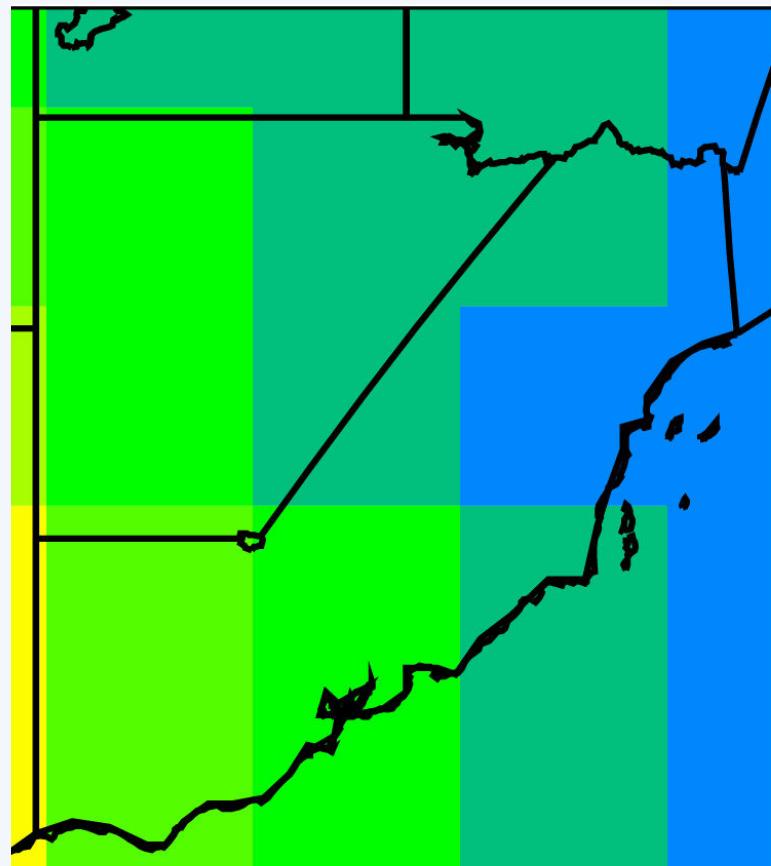
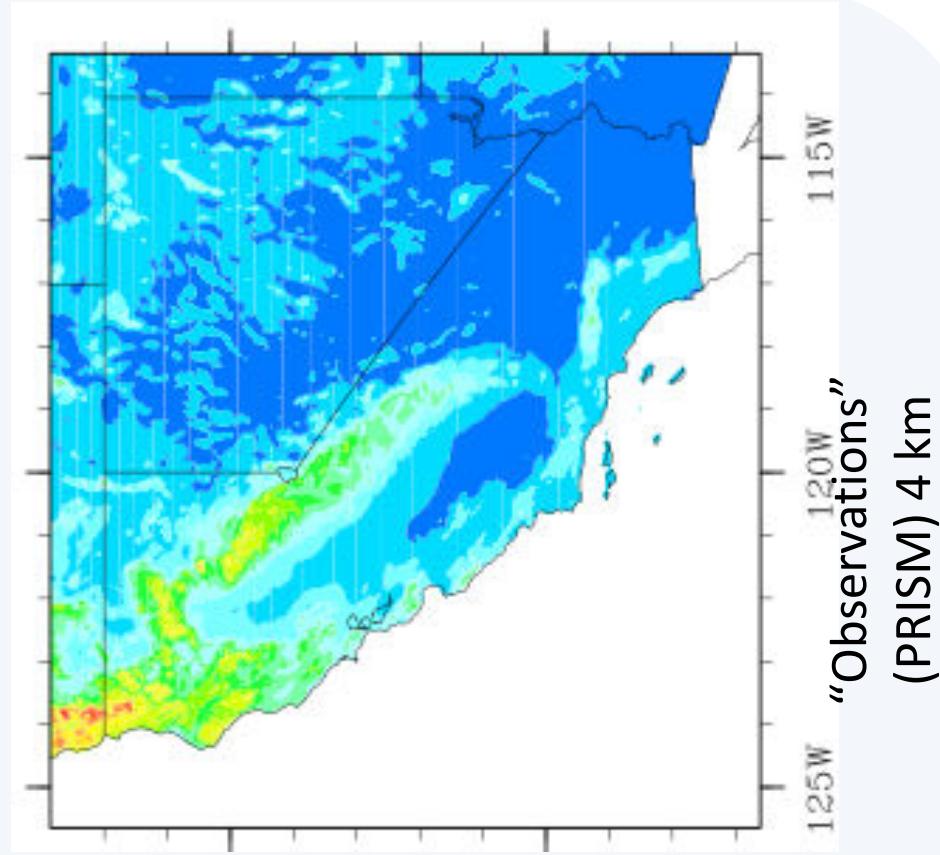
Observed precipitation



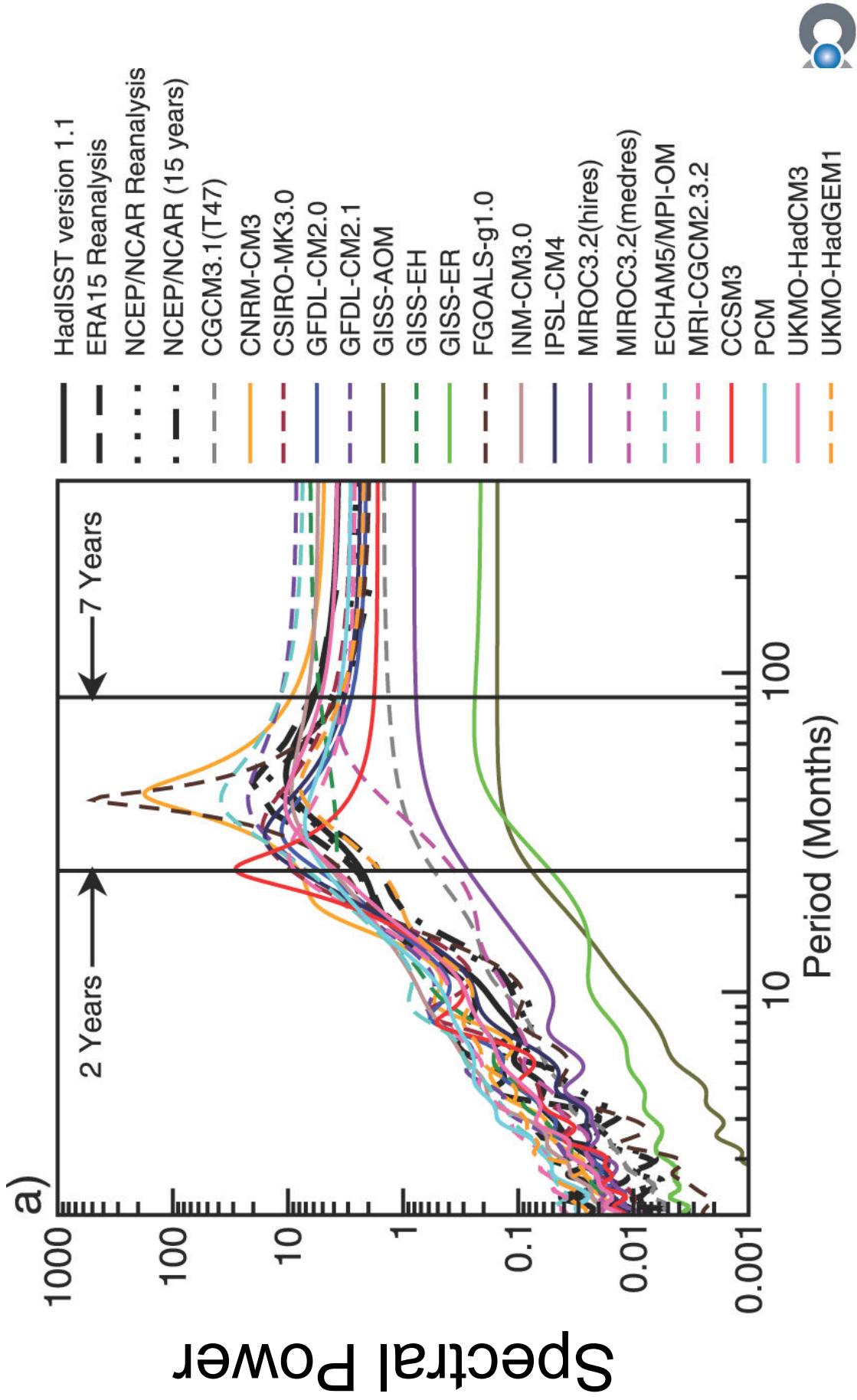
Simulated precipitation  
(multimodel mean)



...but less well on smaller scales



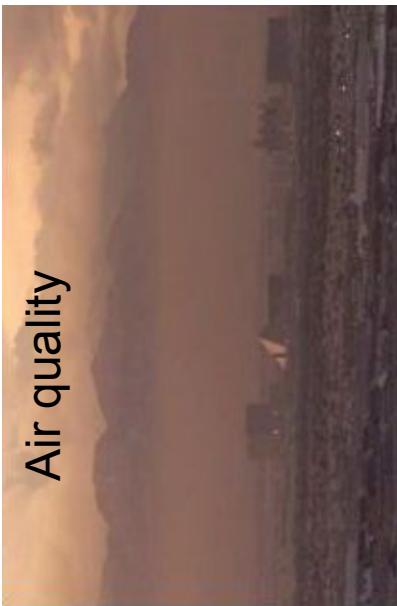
We evaluate simulated variability as  
well as means





# Societal Impacts of Climate Change

# Societal impacts of climate change: The basis of policy decisions



Water availability



# Mitigation

- Reducing GHG emissions to minimize climate change;
- Requires understanding of societal impacts because we need to know “*how much climate change is OK.*”



# Adaptation

- Significant climate change is inevitable;
- *We need to develop coping strategies.*
- This requires understanding of societal impacts.



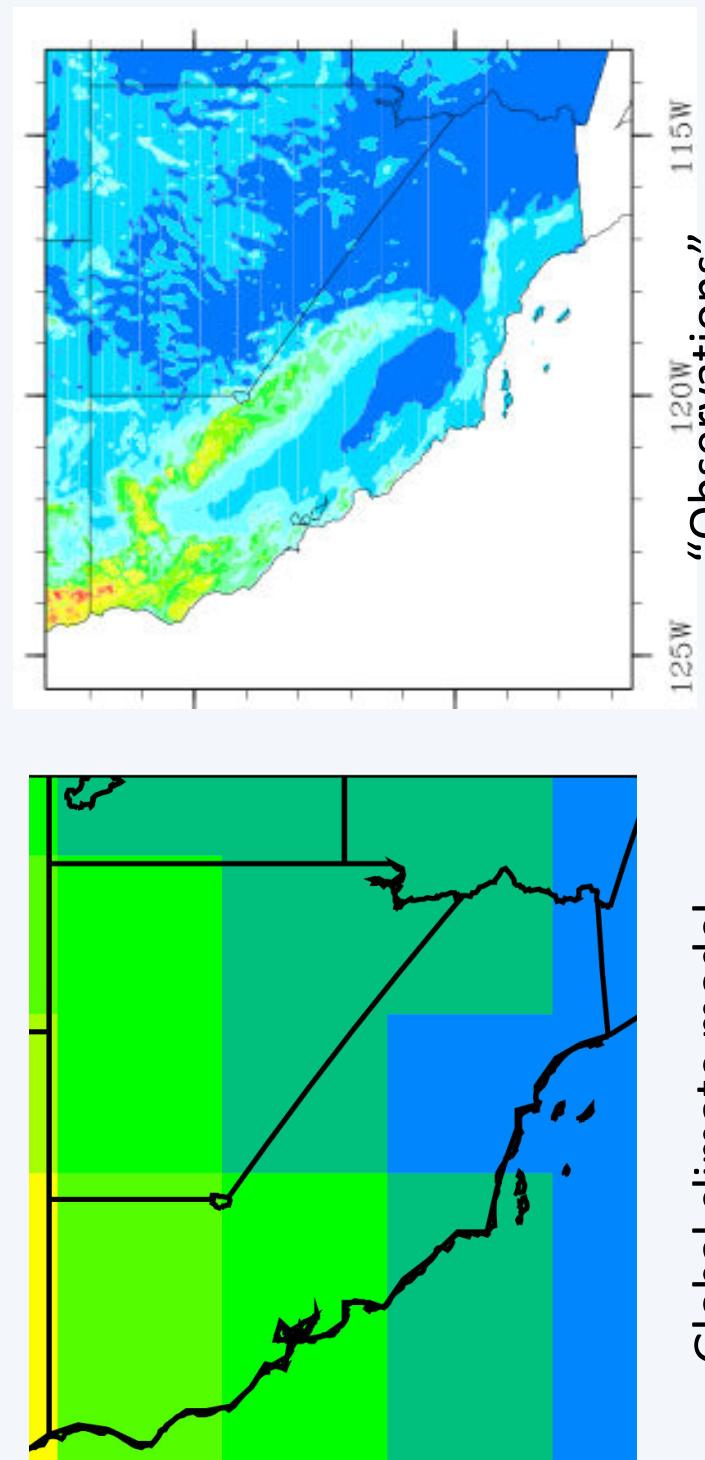


# Societal-impacts studies need climate projections having:

- **Fine resolution**
    - to provide regional-scale fidelity
  - Reliable information on **extremes**
    - because these have disproportionate societal impacts
  - Quantified **uncertainties**
    - usually by analyzing a large family of simulations
- ~~It's difficult impossible to make projections  
having all these properties!~~

# Why we need fine resolution:

Global climate model results are too coarse to be reliable on a regional scale



Global climate model  
~300 km

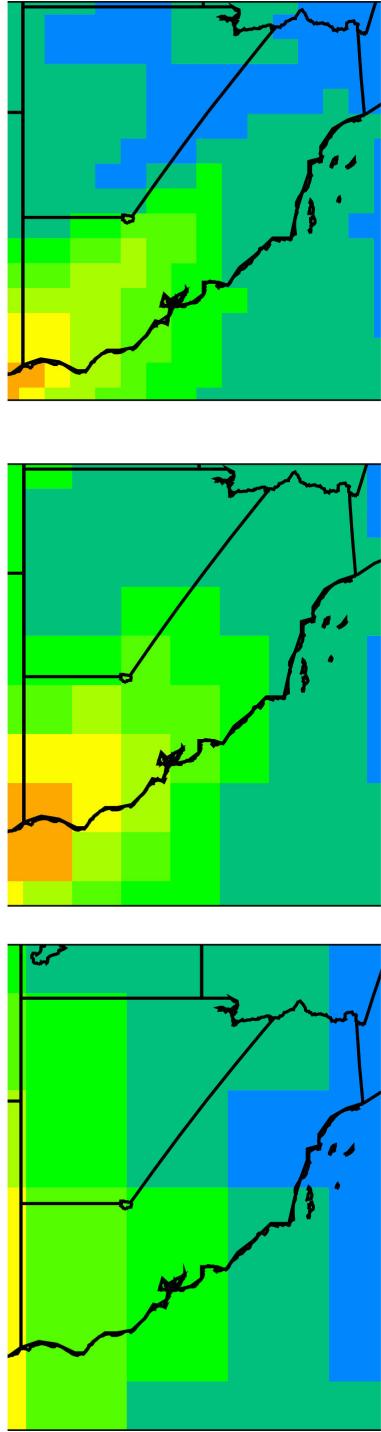
Annual mean precipitation

"Observations"  
(PRISM) 4 km

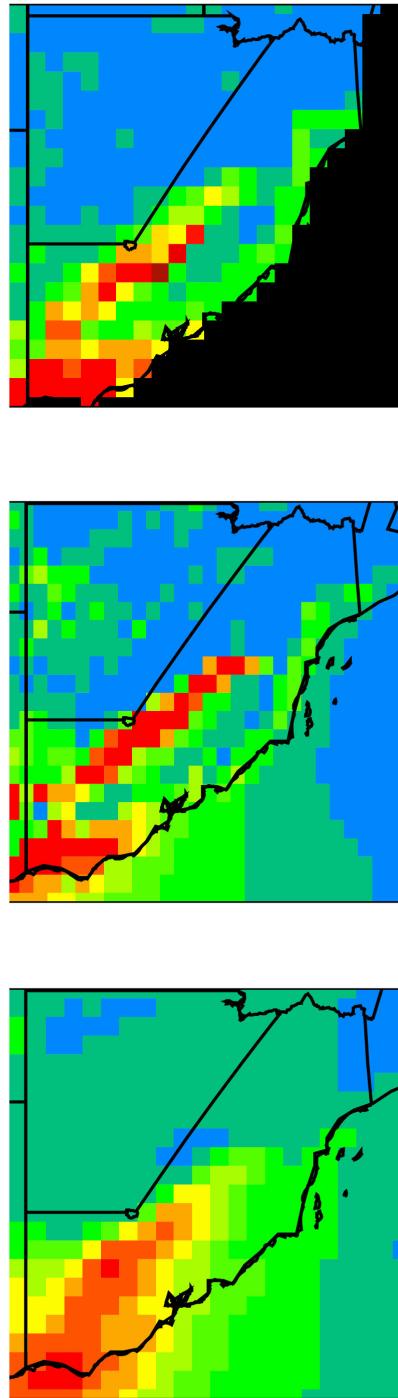
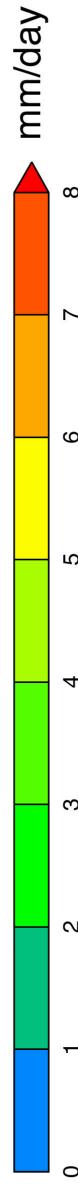


# Refining resolution improves fidelity...

Wintertime precipitation rate



T42 (300 km)      T85 (150 km)      T170 (75 km)



$0.4^\circ \times 0.5^\circ$  (40 x 50 km)      Observations (VEMAP)



... at a high computational price

- A 2x decrease in horizontal grid dimensions  
—> an 8x or 16x increase in CPU time
- Our simulations at 50 km resolution are 200x slower than simulations at the standard resolution of 300 km

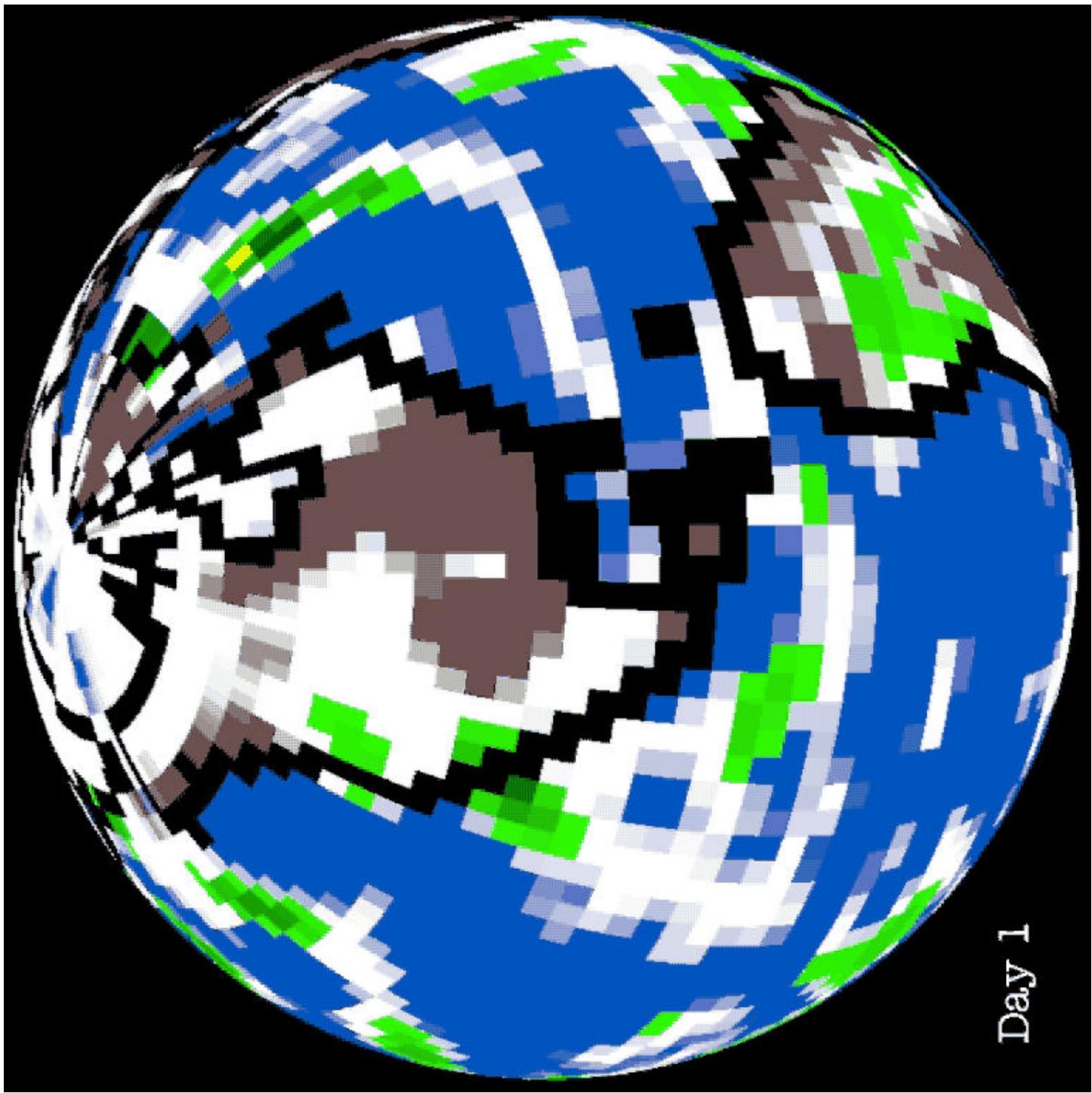


Lawrence Livermore National Laboratory  
Thunder - April 2004

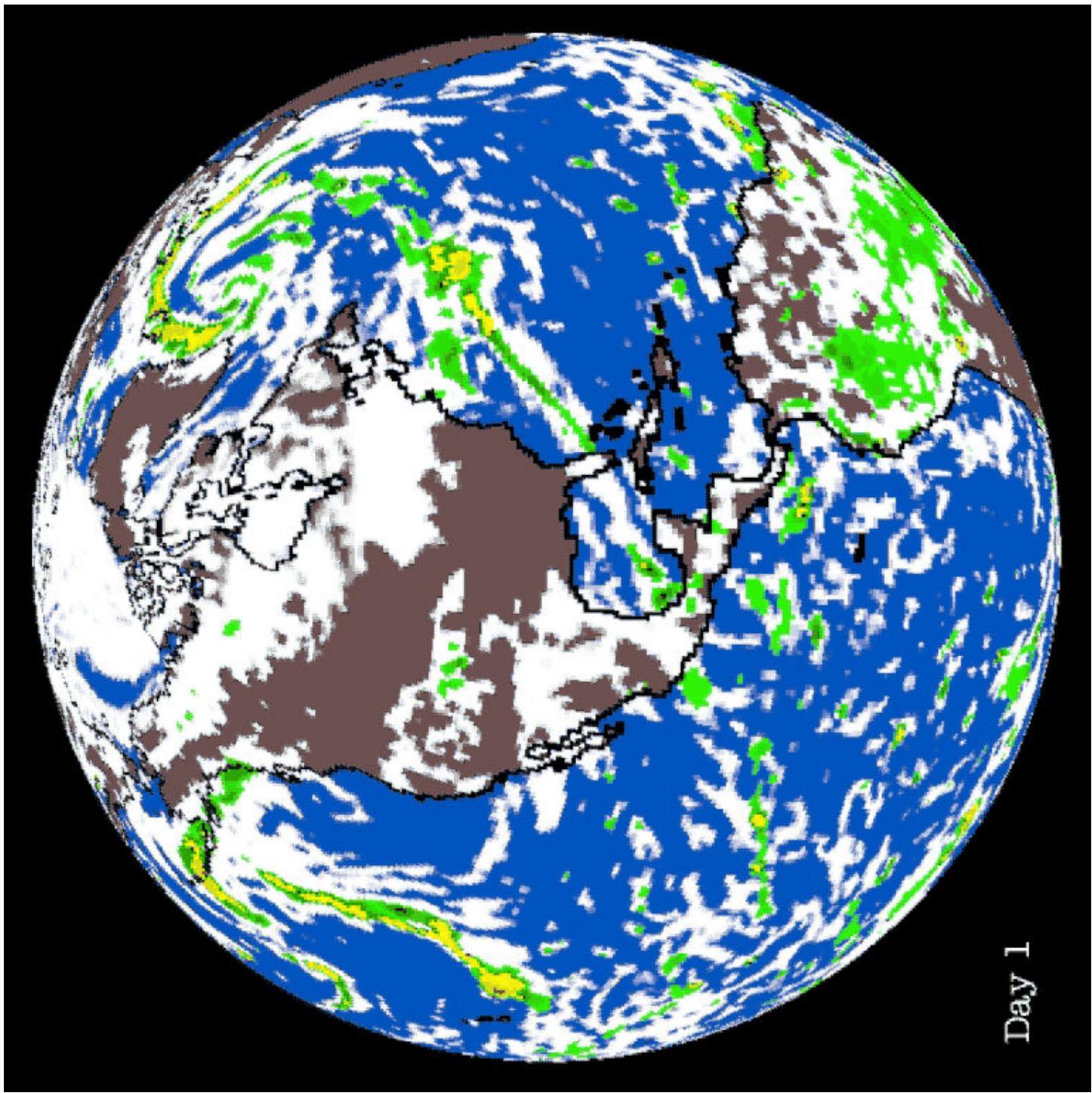


The Earth Simulator Center

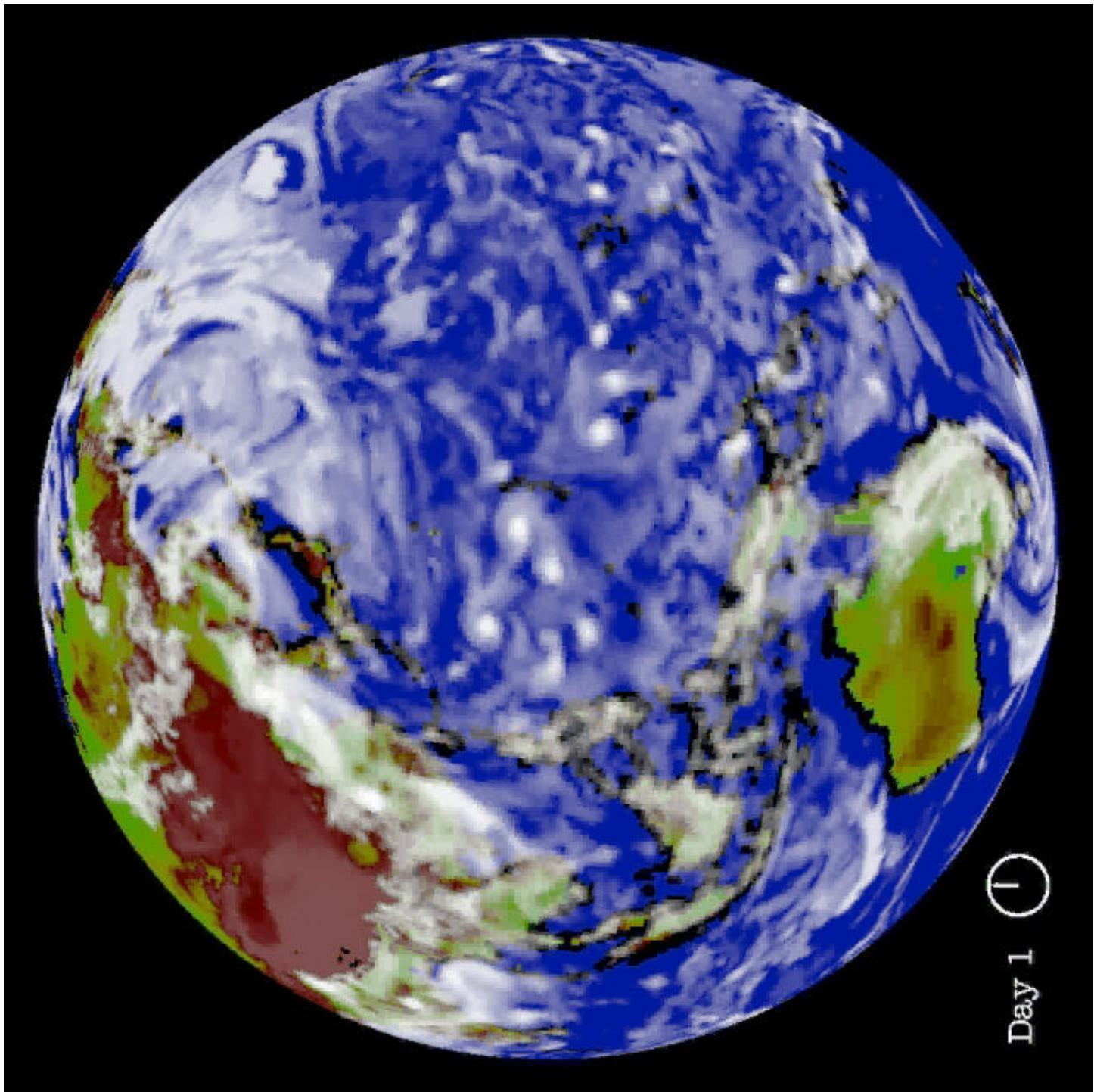




300 km  
grid spacing



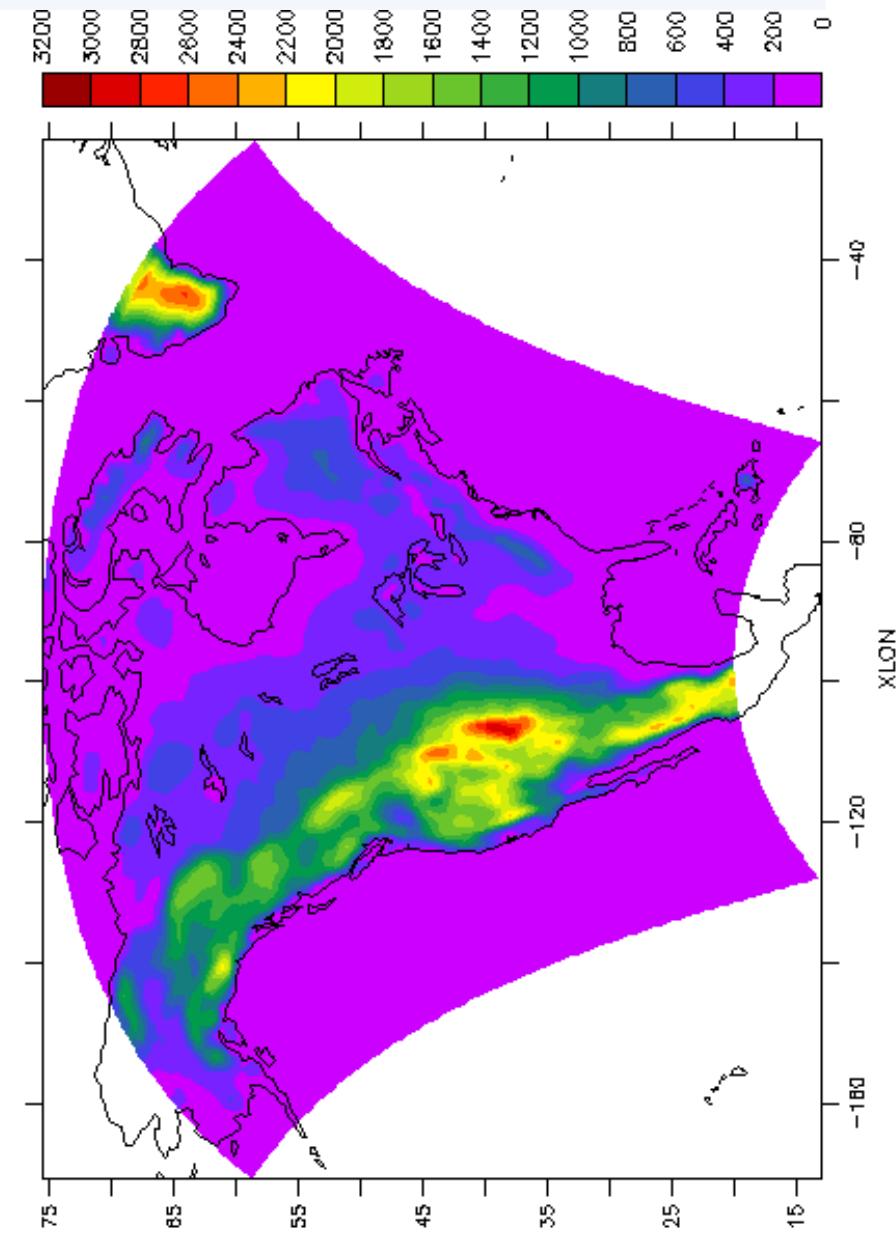
50 km  
grid spacing



Day 1

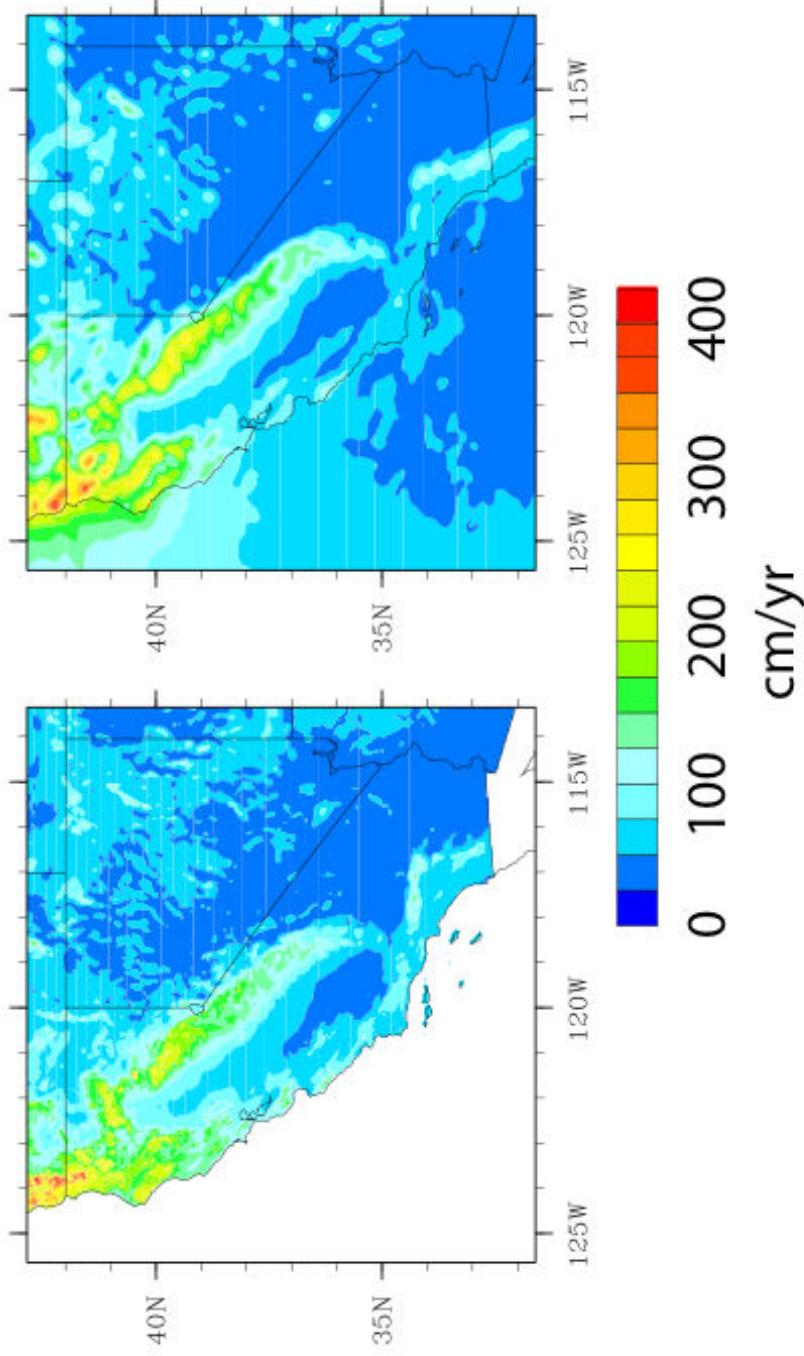
# Dynamical downscaling:

Uses a nested, limited-domain climate model that is based on physical laws



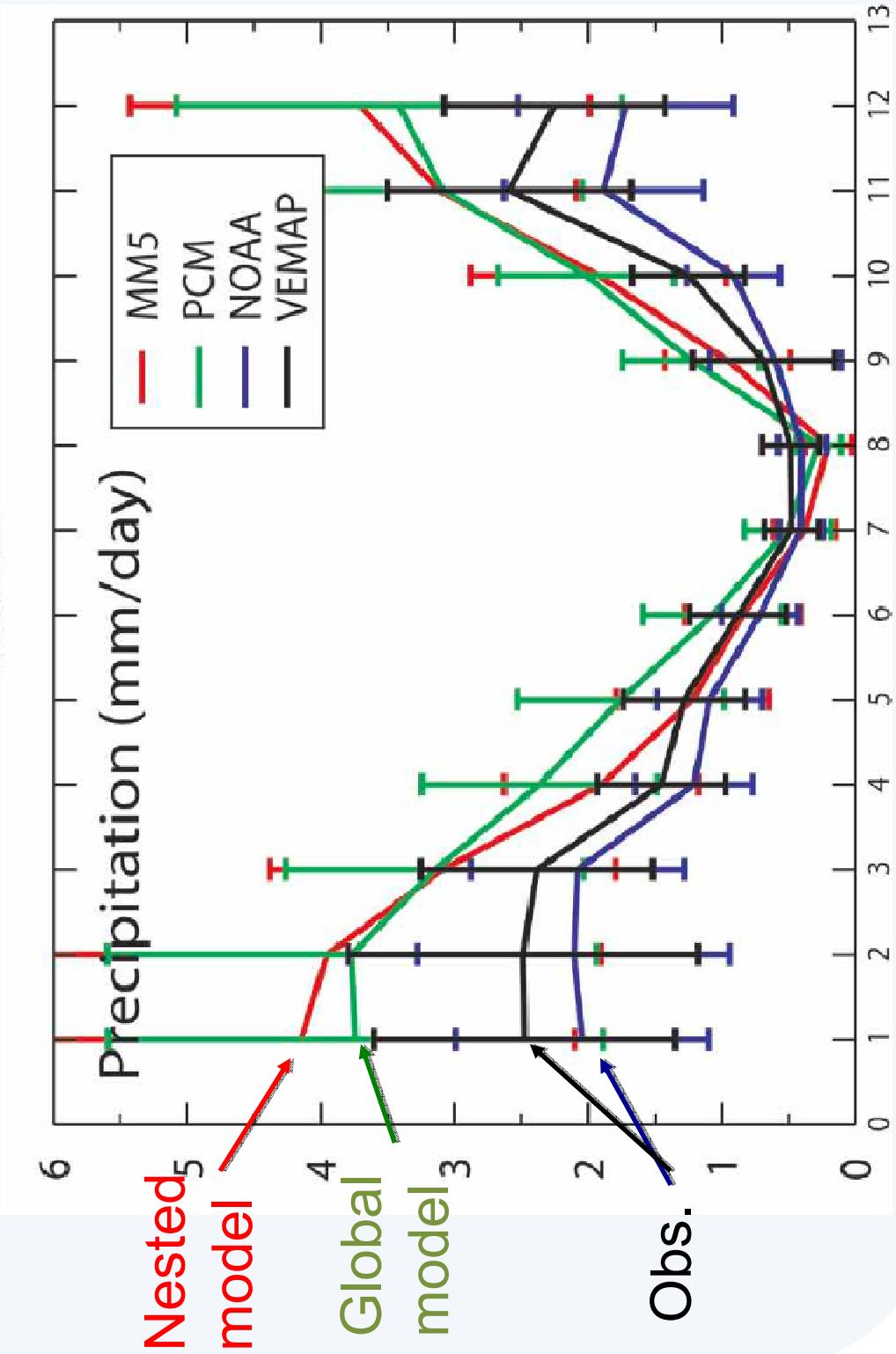
Nested models *can* work beautifully

Annual Mean Precipitation  
Observations  
“Nested” models

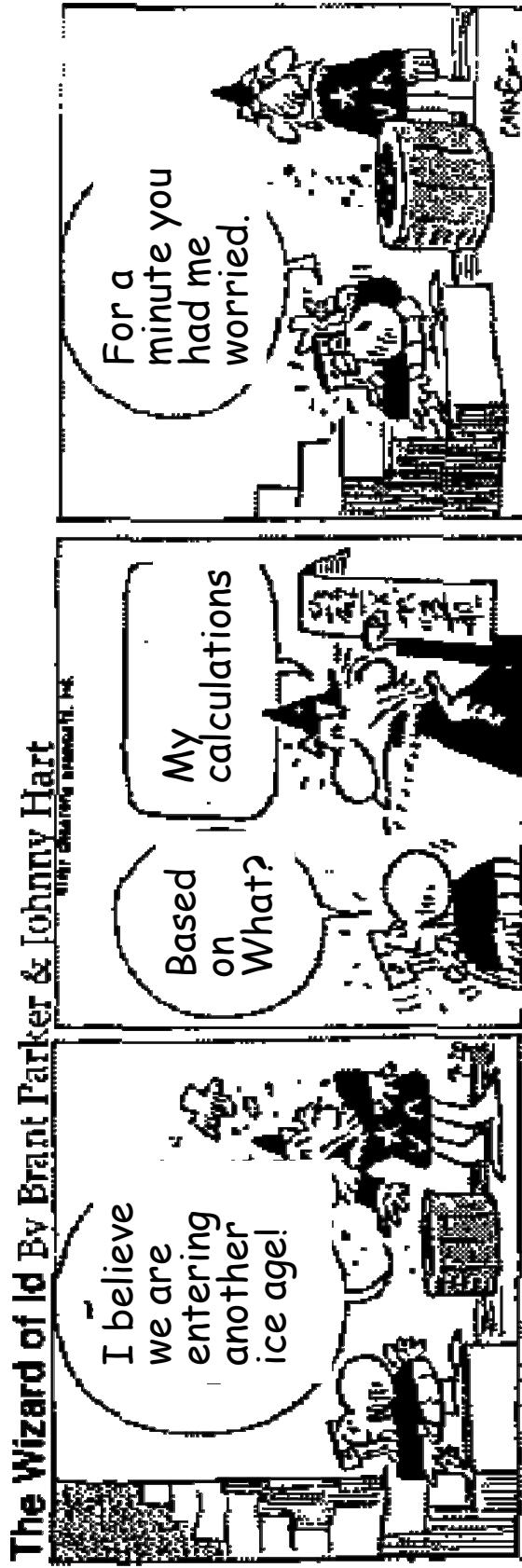


# Dynamical downscaling: GIGO

Month of year->



# Uncertainty: what are limits of climate prediction?



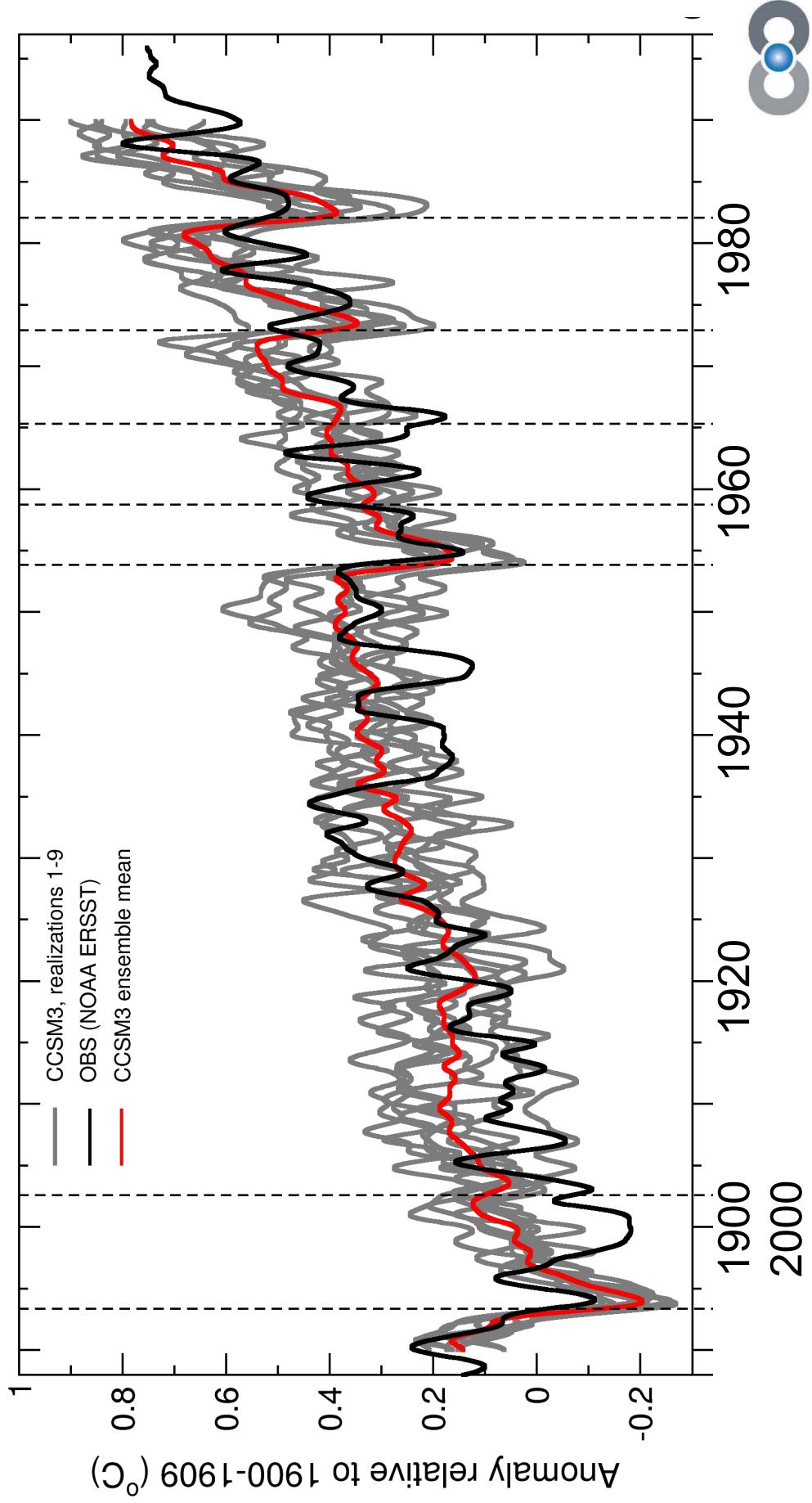


## Sources of uncertainty: imperfect knowledge of

- initial conditions in the atmosphere, etc.;

# Example of initial condition uncertainty

Simulated and observed regional sea-surface temperatures  
courtesy Ben Santer

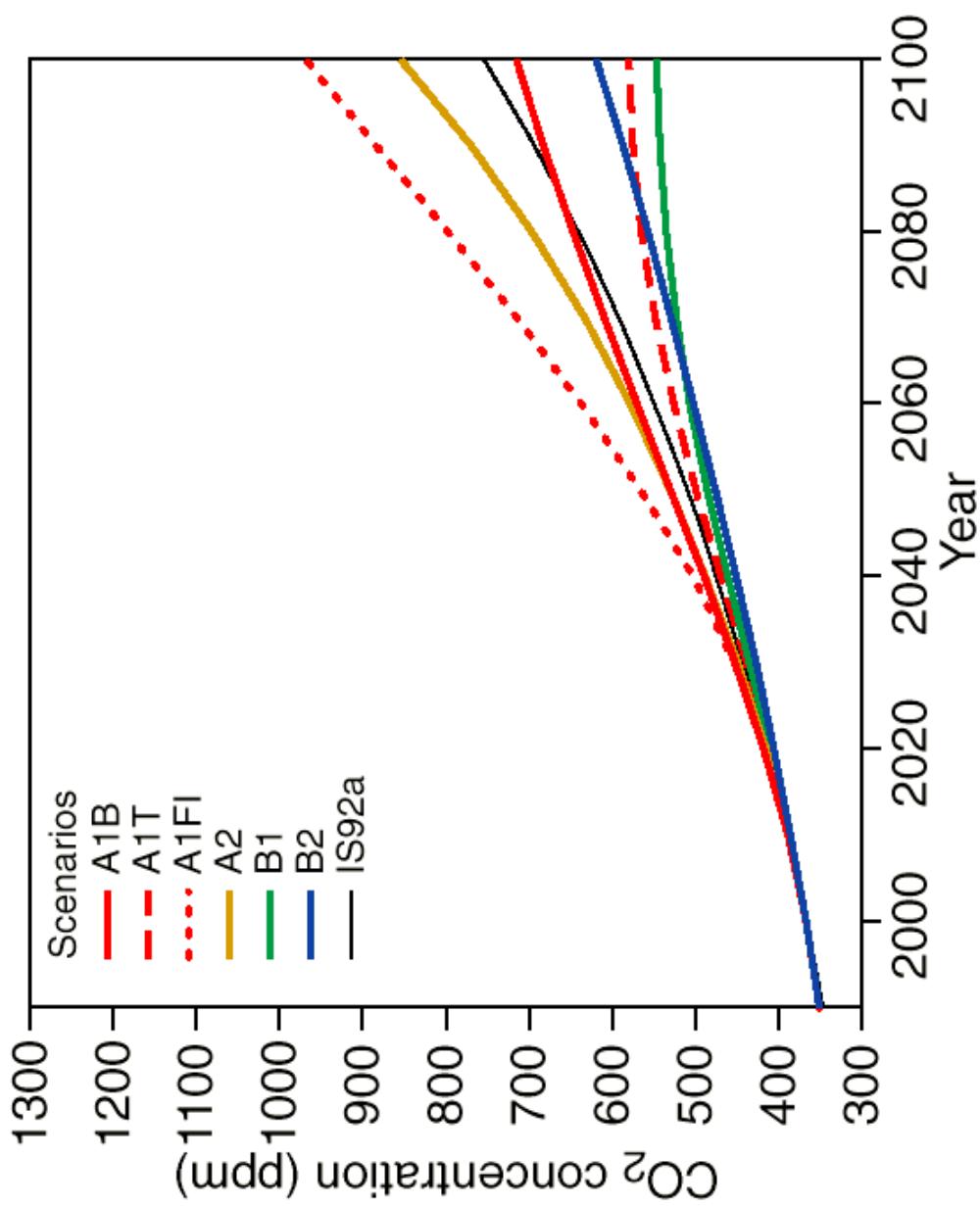




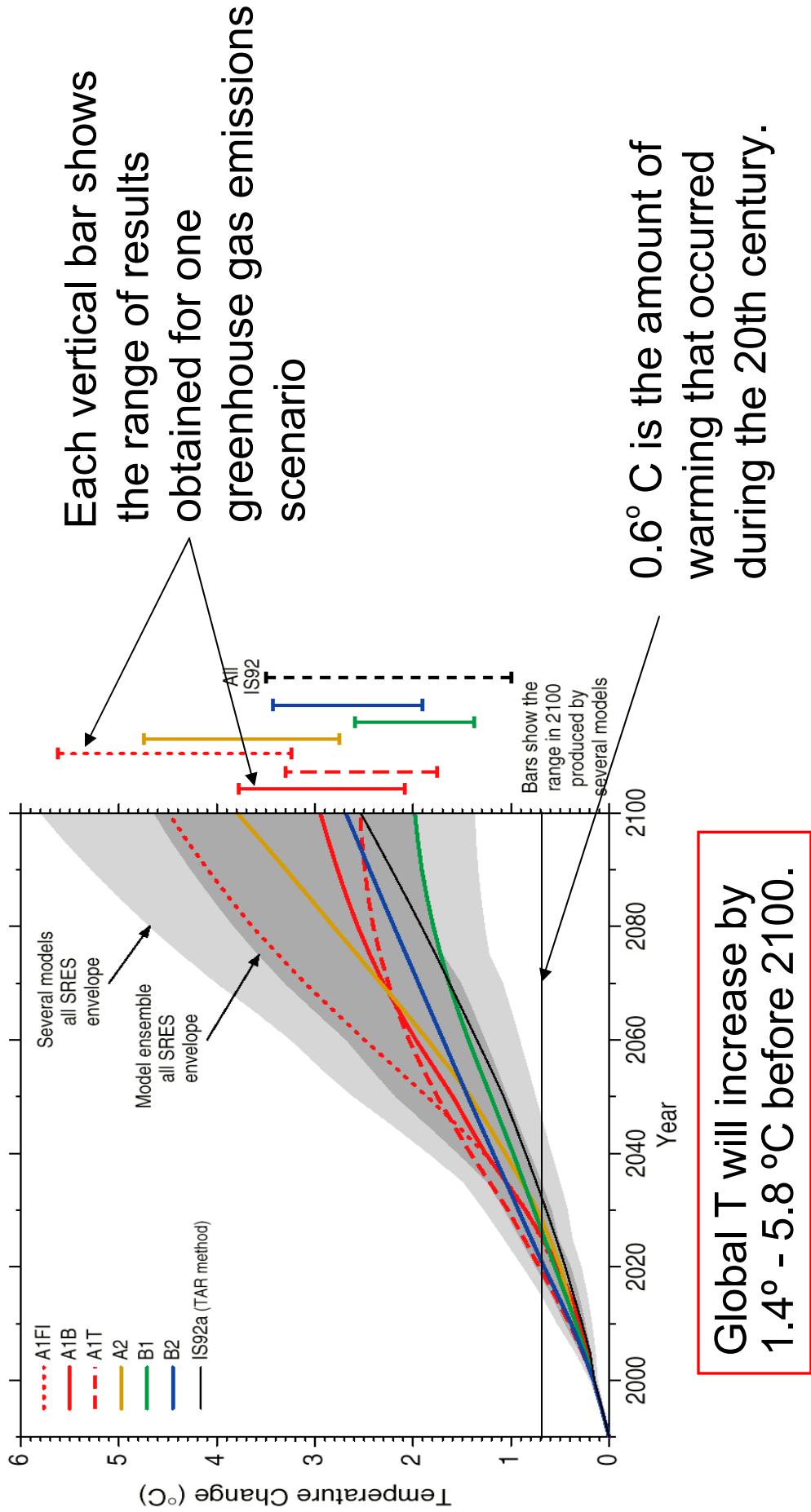
## Sources of uncertainty: imperfect knowledge of

- future behavior of climate “forcings,” e.g. greenhouse gas concentrations;

Future CO<sub>2</sub> concentrations are *unknowable*; this is true of other influences also



About half of future uncertainty in temperature comes from uncertainty in future CO<sub>2</sub> emissions.



# Sources of uncertainty: imperfect knowledge of

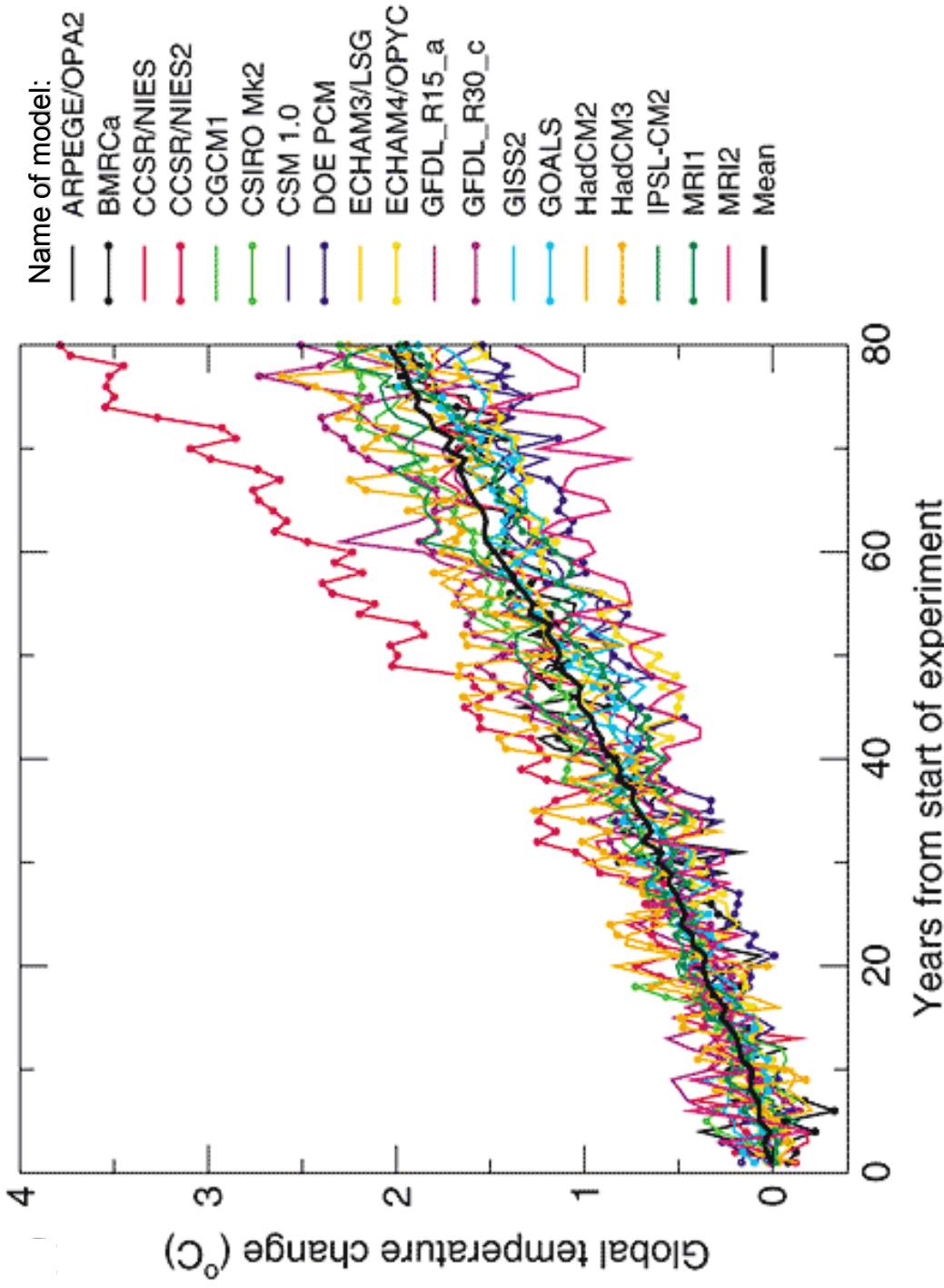
- how the climate system behaves.

These errors arise from:

- Imperfect representation of unresolved phenomena (notably clouds)
- numerical discretization
- “unknown unknowns”.



# Different models respond differently to same inputs



Simulated temperature responses to 1%/yr  $\text{CO}_2$  increase



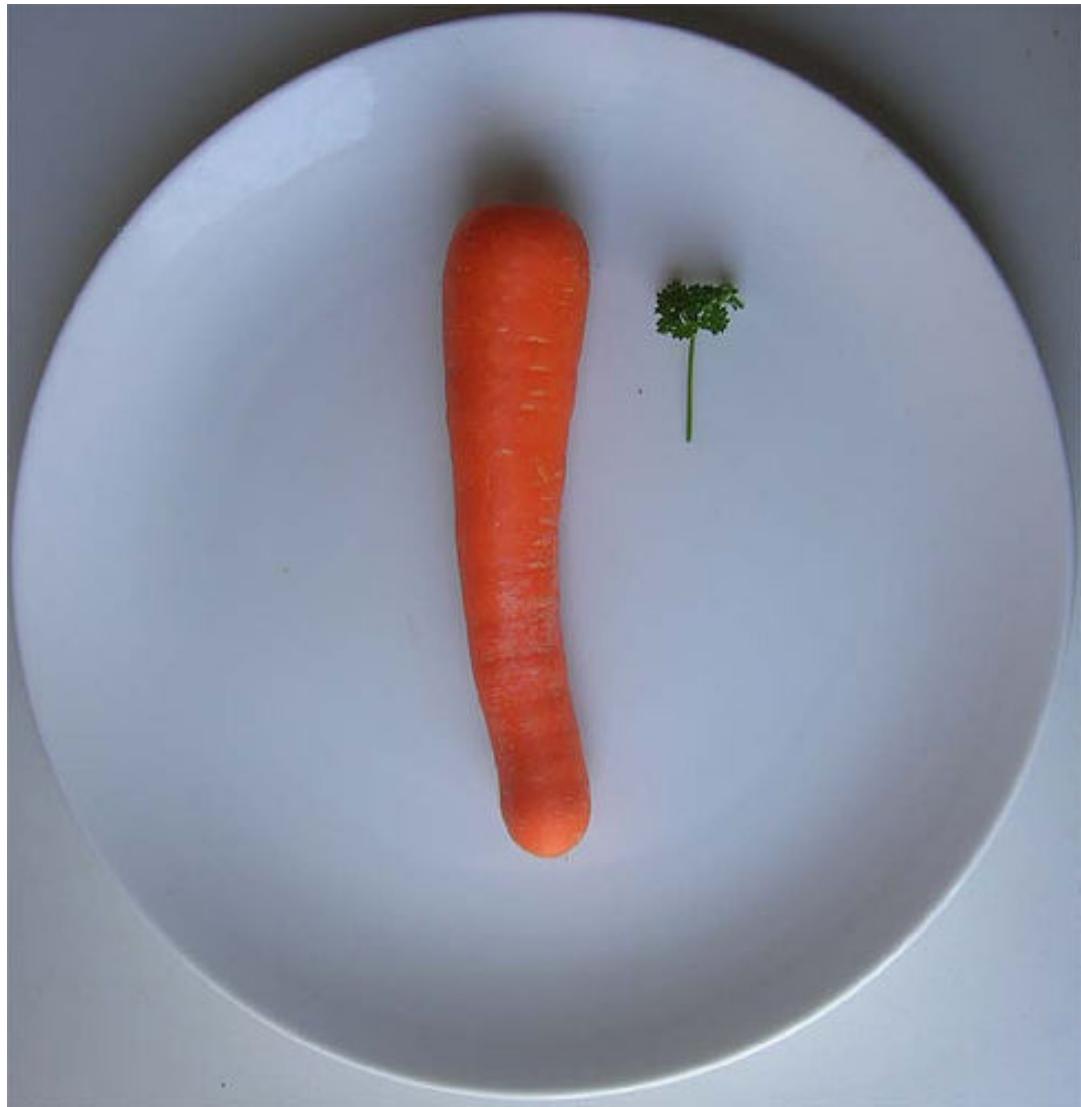
# Parting Thoughts

- Climate models work amazingly well.
- Climate models have serious errors.
- Some important sources of error in future climate predictions are irreducible.
- Climate prediction is no longer an academic exercise!
- The need to incorporate climate change into real-world decisions has “raised the bar” for climate modelers.
- Quantifying and reducing uncertainties are major challenges.



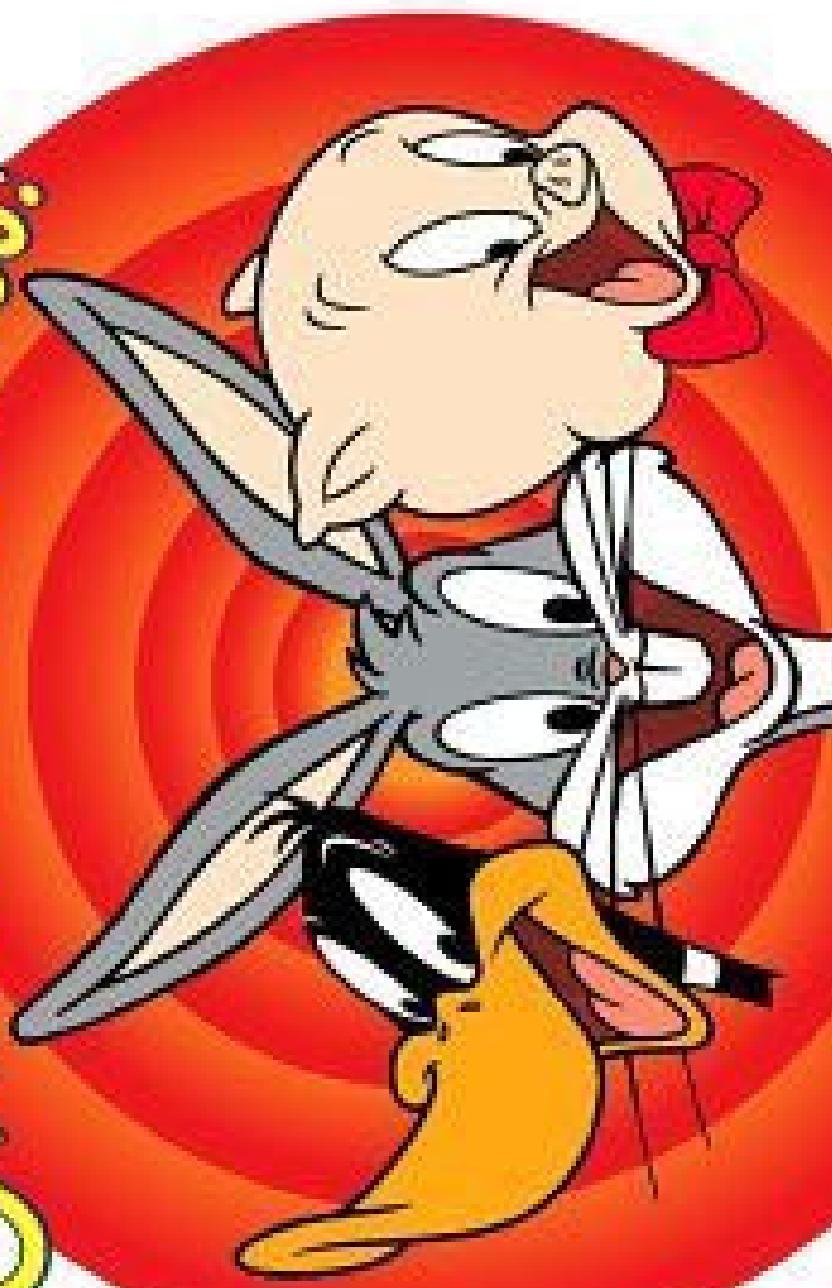


Let's have dinner!





*"Cartoon Network"*



Cartoon Songs From  
**MERRIE MELODIES & LOONEY TUNES**

# How do we estimate climate uncertainty?

- Expert elicitation
- Ensembles of opportunity
- Perturbed physics ensemble
- Why none of these is perfect



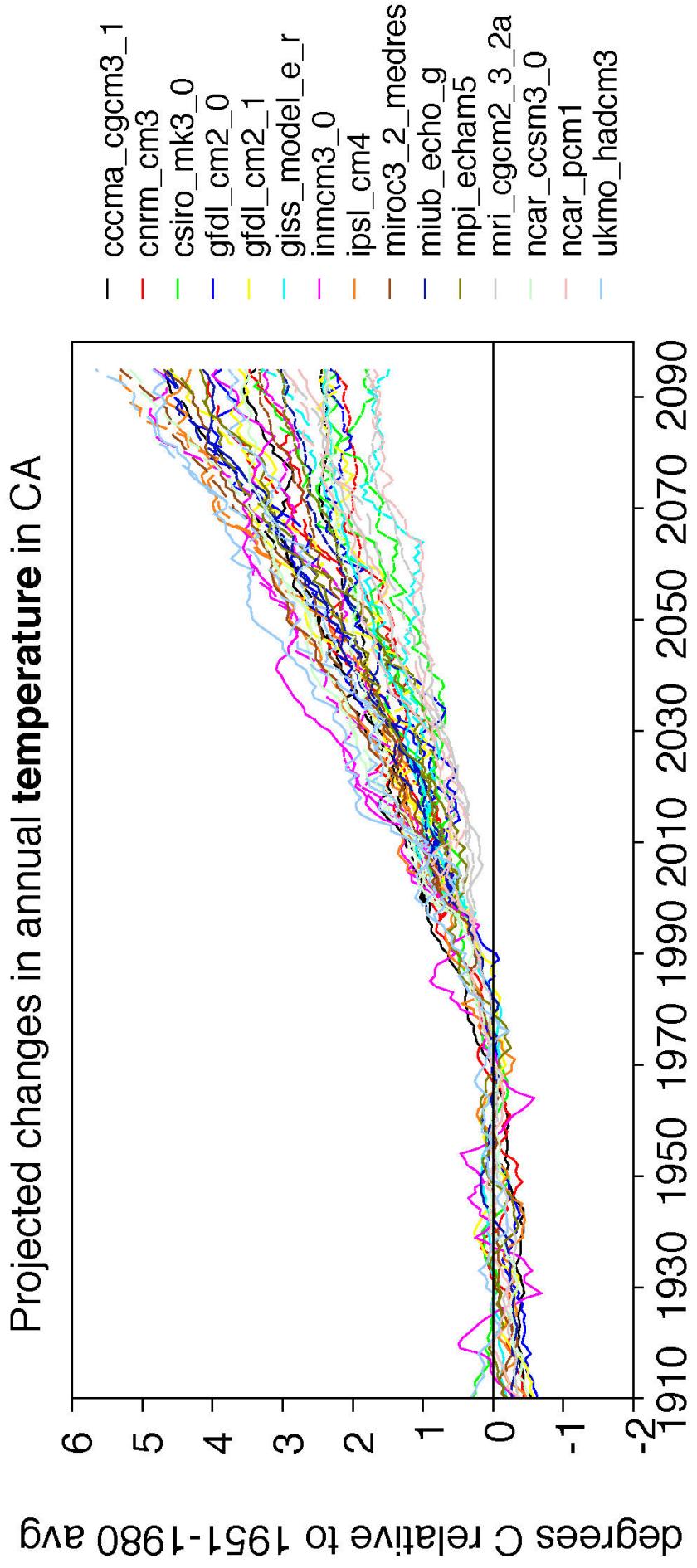


# “Expert Elicitation”

- Fancy term for asking a bunch of so-called experts.
- Why I don't like this approach:
  - It's completely subjective
  - (but often made to look quantitative)
  - Groupthink creates false consensus

# “Ensemble of opportunity:”

a collection of results from a number of available models



Results from 15 models, each simulating 3 CO<sub>2</sub> scenarios





# What's **good** about quantifying uncertainty in this way?

1. It's a start

# What's **good** about quantifying uncertainty in this way?

1. It's a start
2. The mean of a large number of models consistently performs better than any single model
  - This is true in climate simulation and in seasonal weather prediction
    - So having results from multiple models seems to give a better estimate of the most likely outcome.



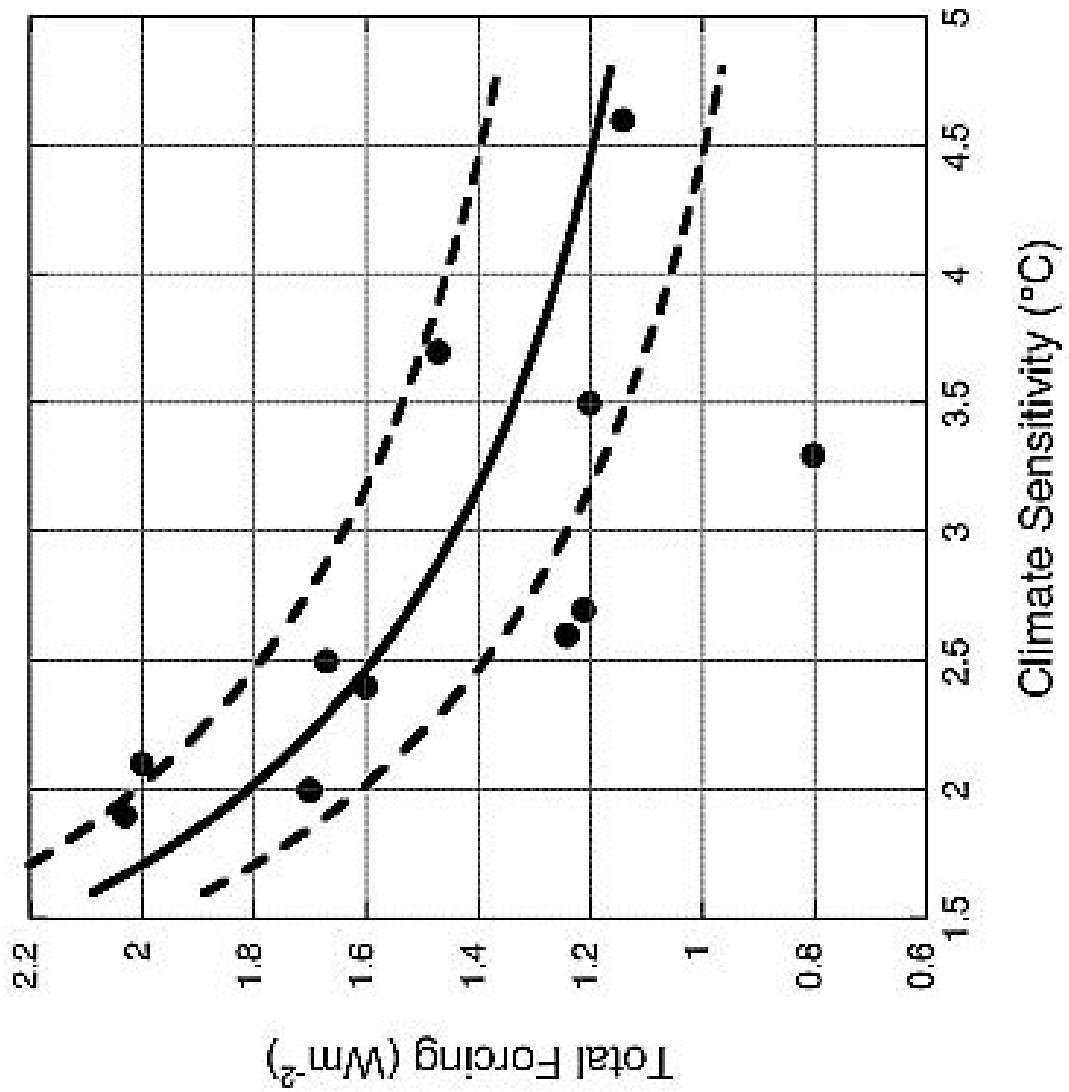
# What's **bad** about quantifying uncertainty in this way?

1. Results can be influenced by selection of models, which can be haphazard.
2. Can be misleading because errors common to many models may be important. I.e., even if models agree with each other, they could all be wrong.
  - Superiority of mean model *suggests* that this is not important
  - Hence this approach measure consensus more than uncertainty



# What's **bad** ...

3. Some evidence that GCMs have been unconsciously "tuned"

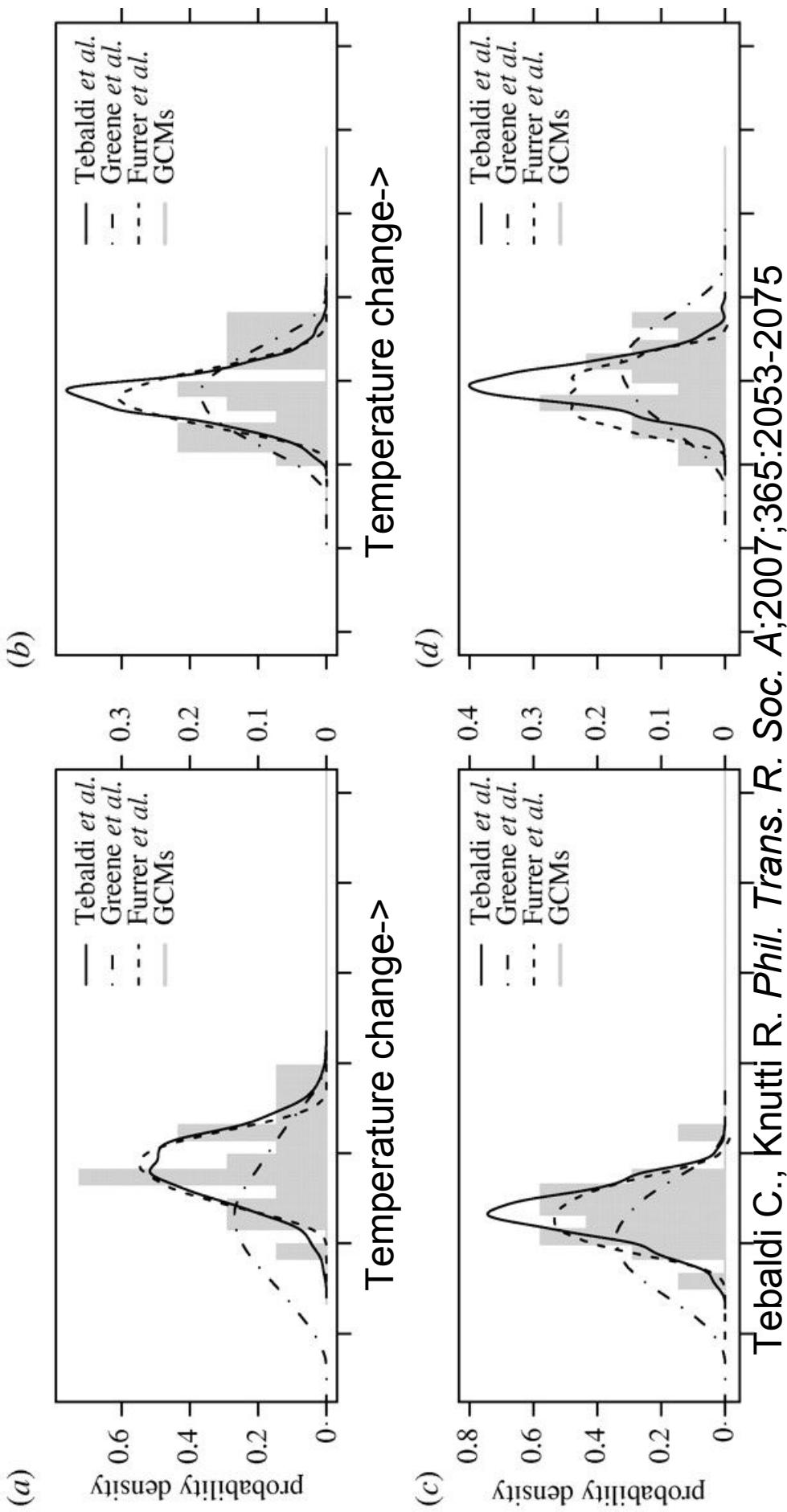


Source: Kiehl, GRL (2007)



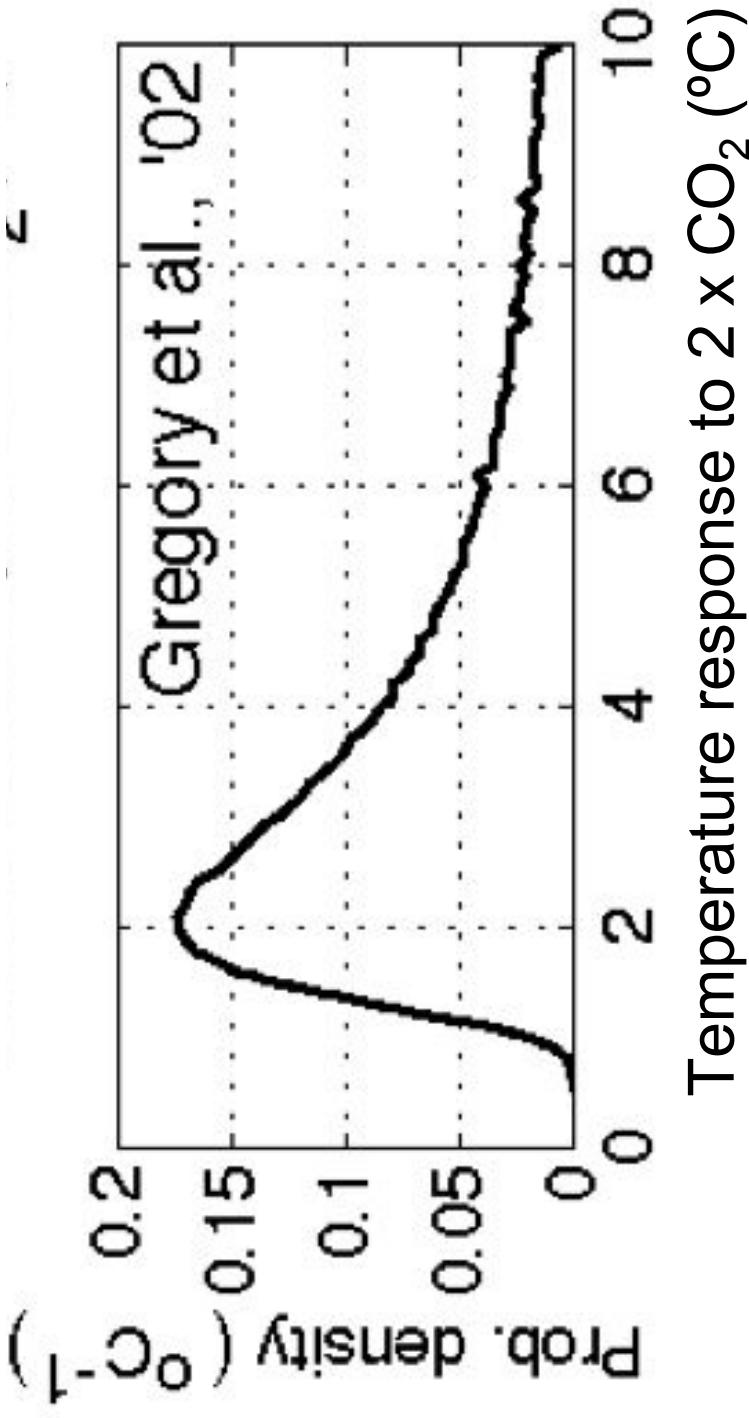
# What's **bad** ...

4. Often values all models equally, which *can't* be optimal
  - But we can't agree on best way to combine models



## What's **bad** ...

5. Does not include outcomes that all agree have low (but non-zero) likelihood.



A range of model results estimates the uncertainty in the most likely outcome, not the full range of possible values.

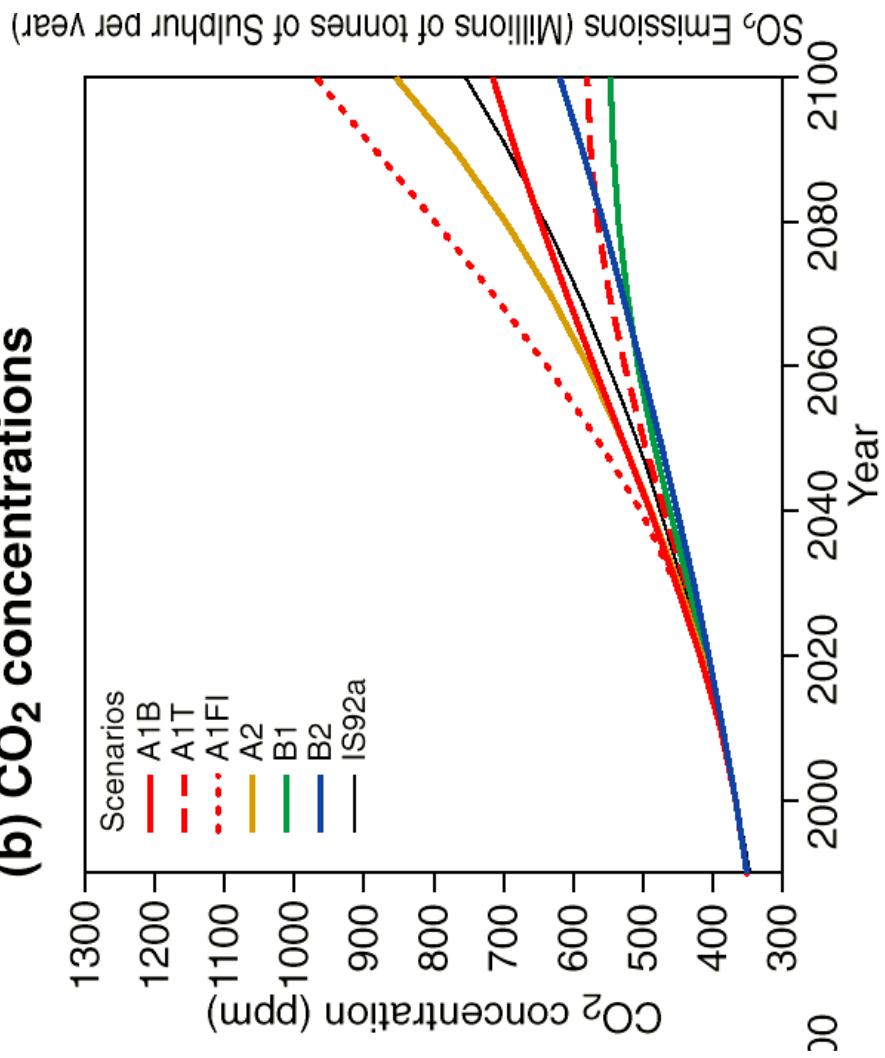
Source: Roe and Baker, UW



# What's **bad** ...

6. Uncertainty in future forcings (e.g. greenhouse gases) is difficult to quantify.

(b)  $\text{CO}_2$  concentrations



A better and cooler way to quantify uncertainty: climateprediction.net

- 48,000 participants are running a climate model “in background” on their computers.
- 43,672,873 simulated years had been run as of April 23.
- Each participant runs a slightly different model version, with a unique combination of parameter values.
- The result is a thorough exploration of parameter space.

