

Using Simple to Intermediate Climate Models to Understand Climate Change and other stuff

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Math & ClimateChange Nexus

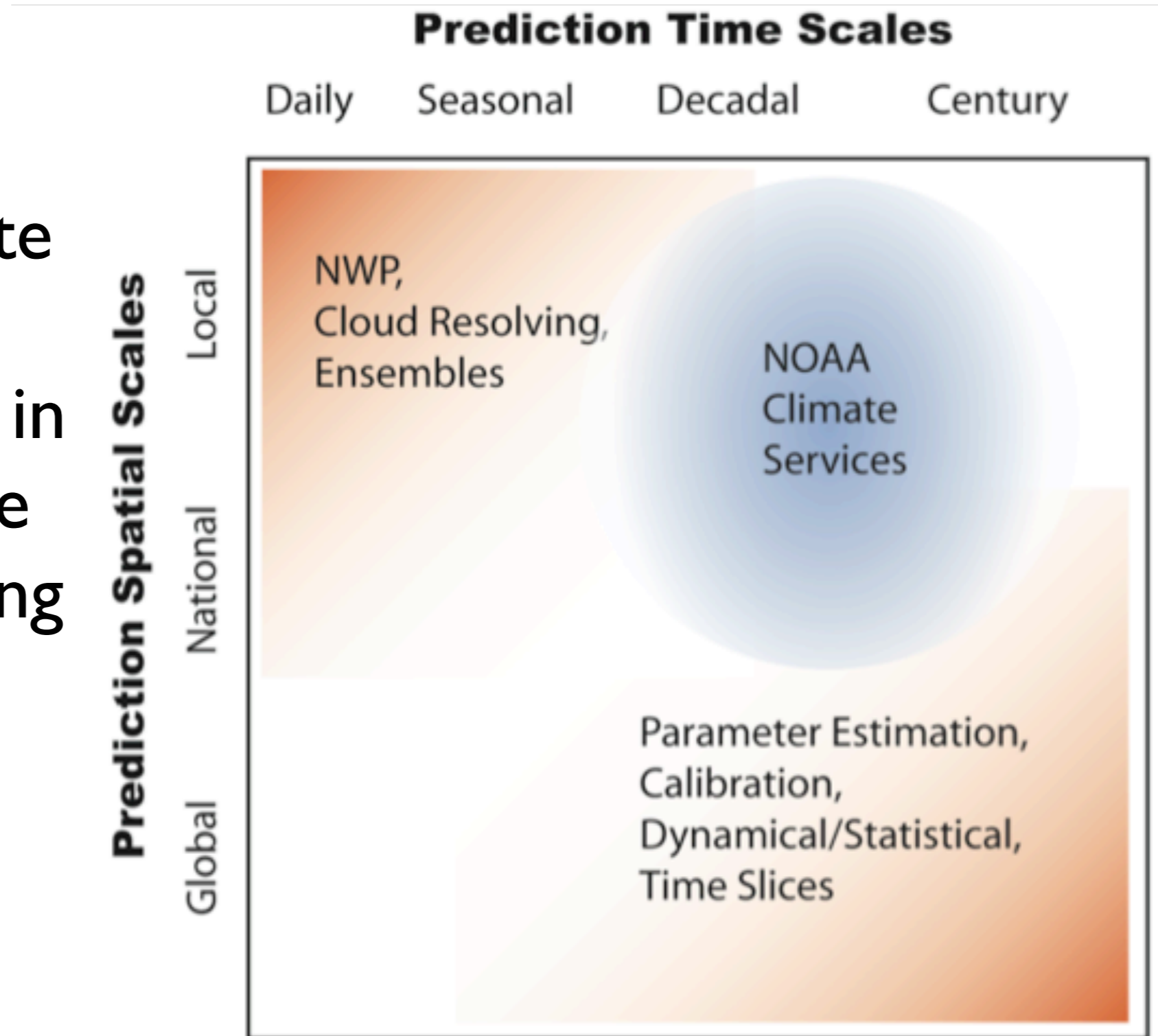
- **Model parameter estimation and calibration for predictions**
 - Prediction: Time-scales? Spatial scales? Variables? Model errors?
- **D&A Issues:** Variability on long time-scales, Attribution of climate change
- **Dynamics v. Response - What are we learning from data analysis? (properties v. parameters?)**
- **Tools for Uncertainty Estimation**
 - Sampling, Parameters, Structure, ...

Assessing Uncertainty in Regional Climate Change Projections

- **Factors affecting regional climate**
 - Large-scale response of atmospheric circulation
 - Topography, tropical SSTs, local and global environment, etc.
- **Motivation**
 - Need to characterize differences in climate models by building *simplified models* to capture uncertainty

Getting from global to regional scales

The regional climate prediction capabilities are not in sync with the state of the art forecasting methods

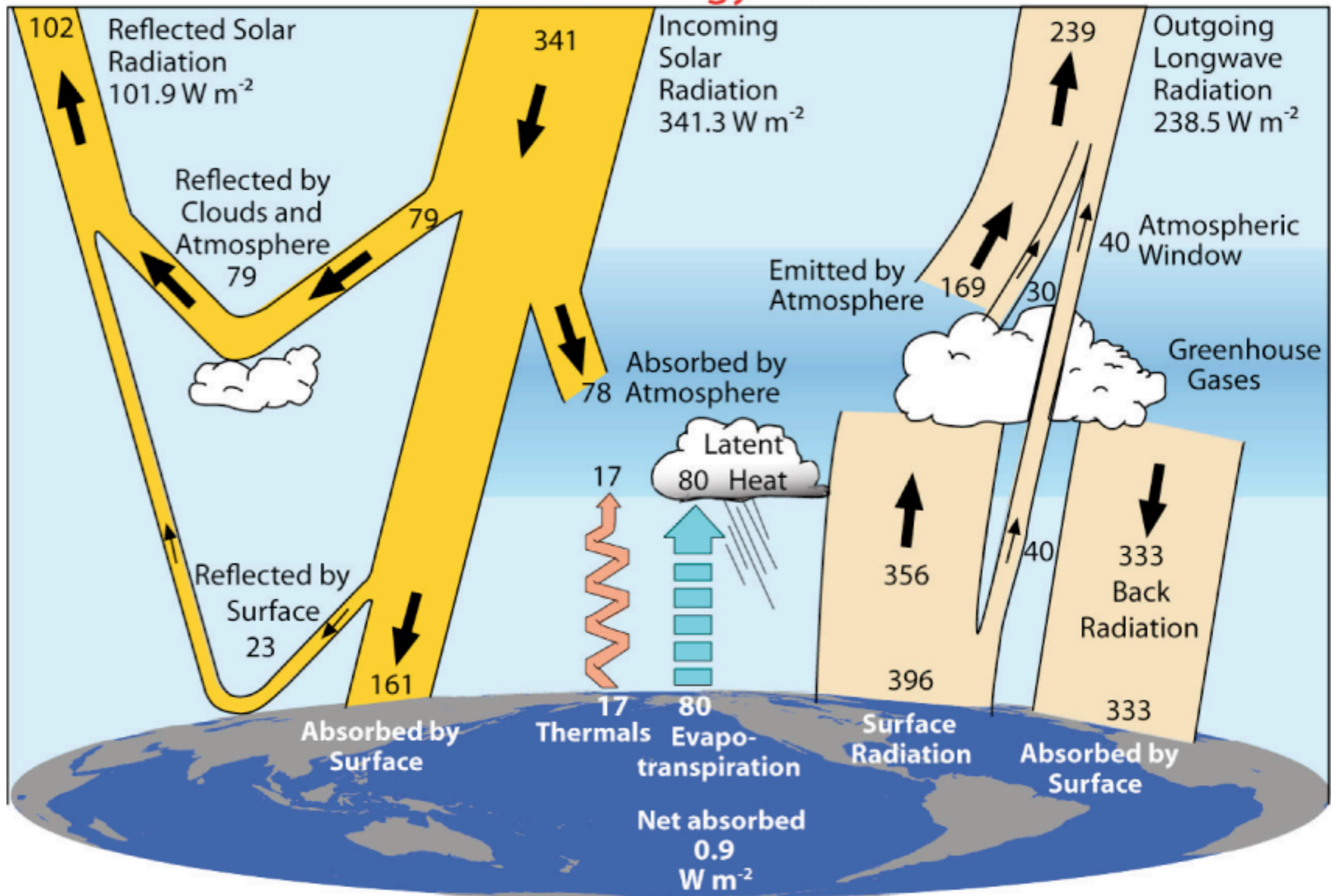


Forest et al. - White Paper on NOAA Climate Services

Overview

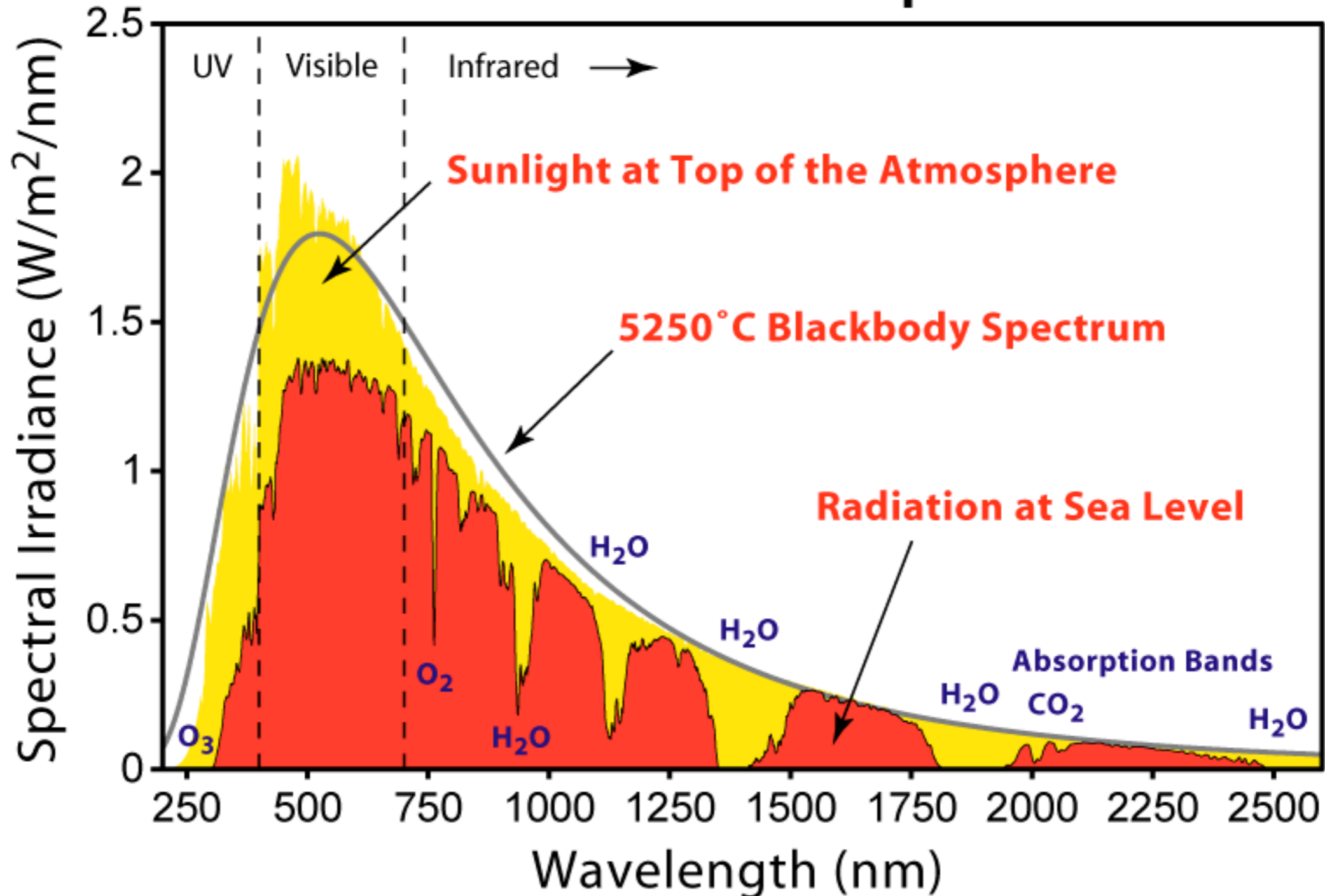
- **Other Simple Models**
 - **0D climate model**
 - **Stommel Box model (not discussed)**
- **Introduction to Earth/Climate Models of Intermediate Complexity (EMICs)**
- **Intro to MIT IGSM**
- **Uses of the MIT IGSM**
 - **Reproducing 20th century record**
 - **Future Climate Change**
 - **Understanding AOGCM behavior**

Global Energy Flows $W m^{-2}$

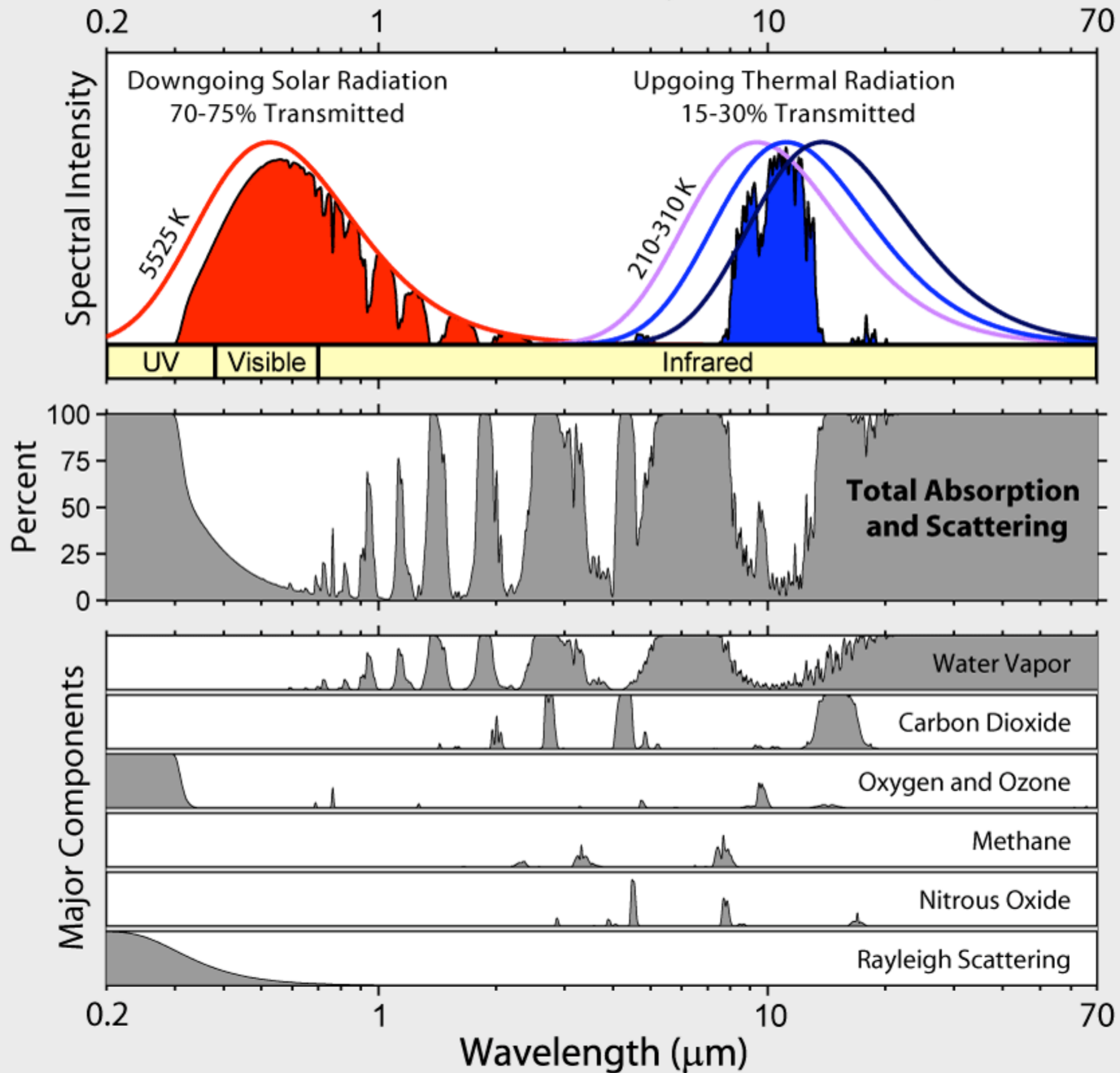


Is the atmosphere transparent to Solar wavelengths?

Solar Radiation Spectrum



Radiation Transmitted by the Atmosphere



0D Climate Model

- **Incoming absorbed solar = outgoing IR**

$$Q_o = \frac{S_o(1-\alpha)}{4} = \sigma T_e^4$$

- **where T_e = Effective blackbody temperature of Earth required to balance incoming radiation**

$$T_e = \left(\frac{S_o(1-\alpha)}{4\sigma} \right)^{\frac{1}{4}} = 255K$$

Simplest Greenhouse Model

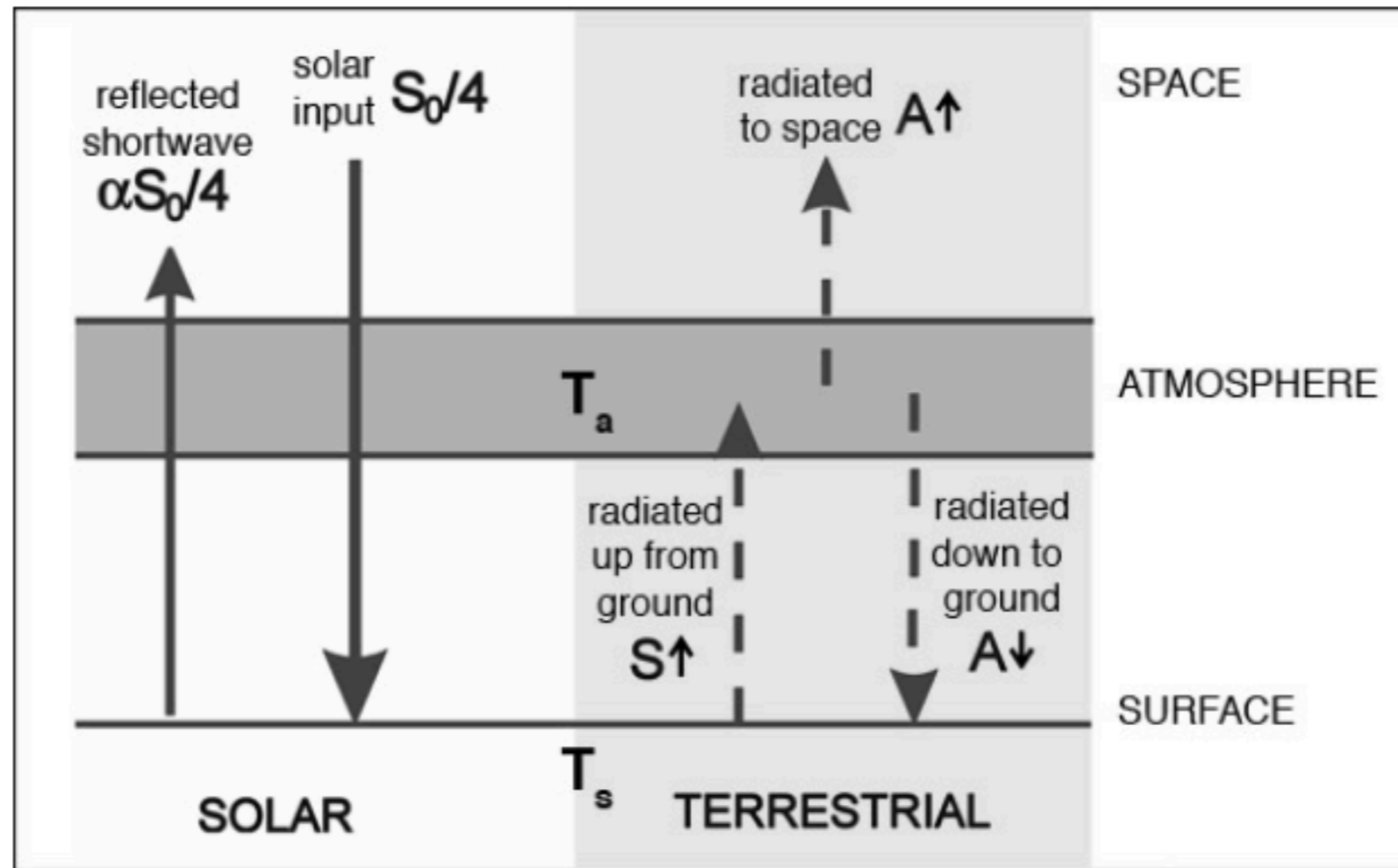
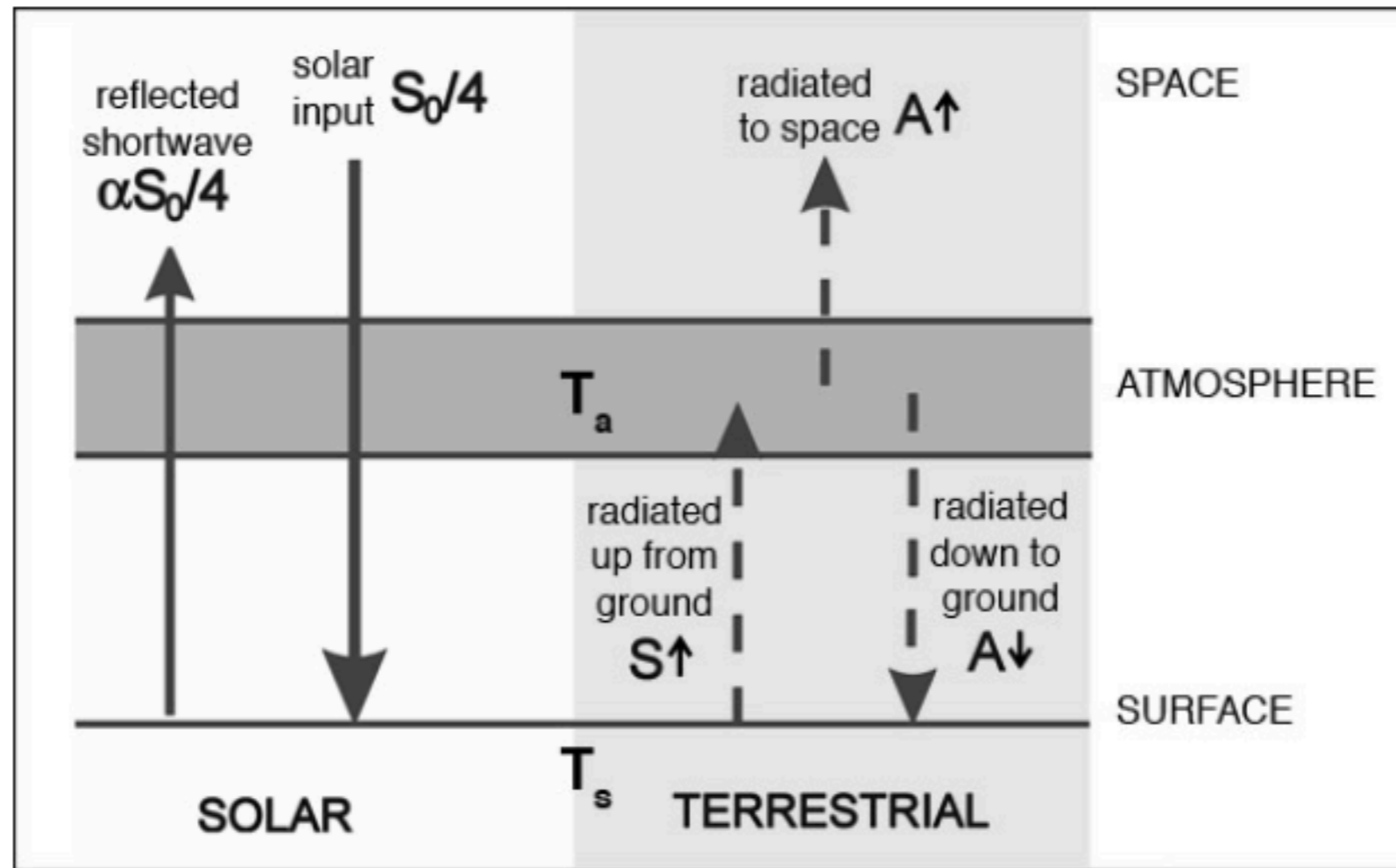


Figure 2.7: The simplest greenhouse model, comprising a surface at temperature T_s , and an atmospheric layer at temperature T_a , subject to incoming solar radiation $\frac{S_0}{4}$. The terrestrial radiation upwelling from the ground is assumed to be completely absorbed by the atmospheric layer.

Simplest Greenhouse Model



Top Energy Balance: $\frac{S_o}{4} (1 - \alpha) = \sigma T_e^4$

Sfc Energy Balance: $\frac{S_o}{4} (1 - \alpha) = \sigma (T_s^4 - T_e^4)$

Rearranging: $T_s = 2^{\frac{1}{4}} T_e = 1.19 * 255 = 303.K$
 Marshall and Plumb:AOCD

Simple 1D climate model

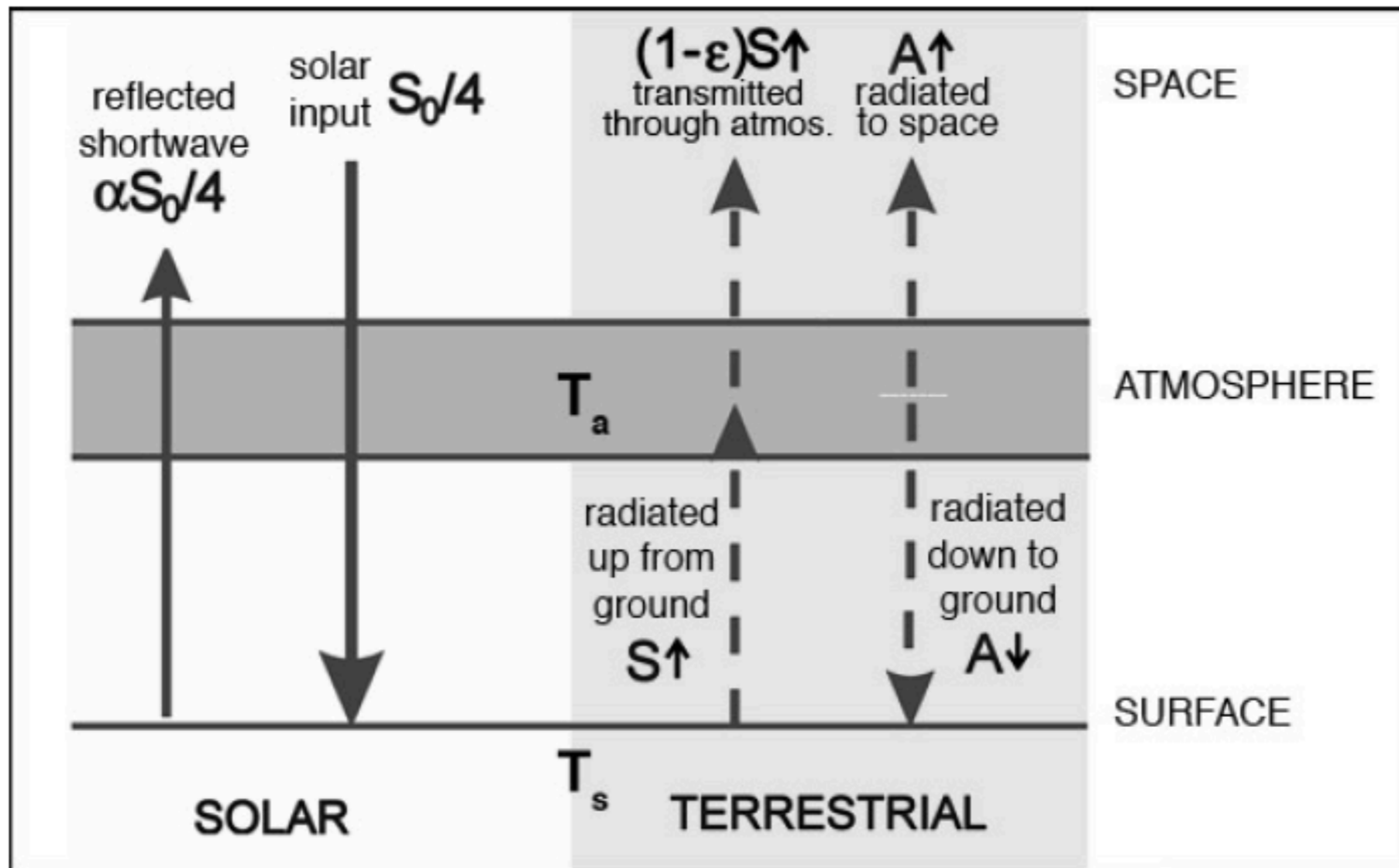
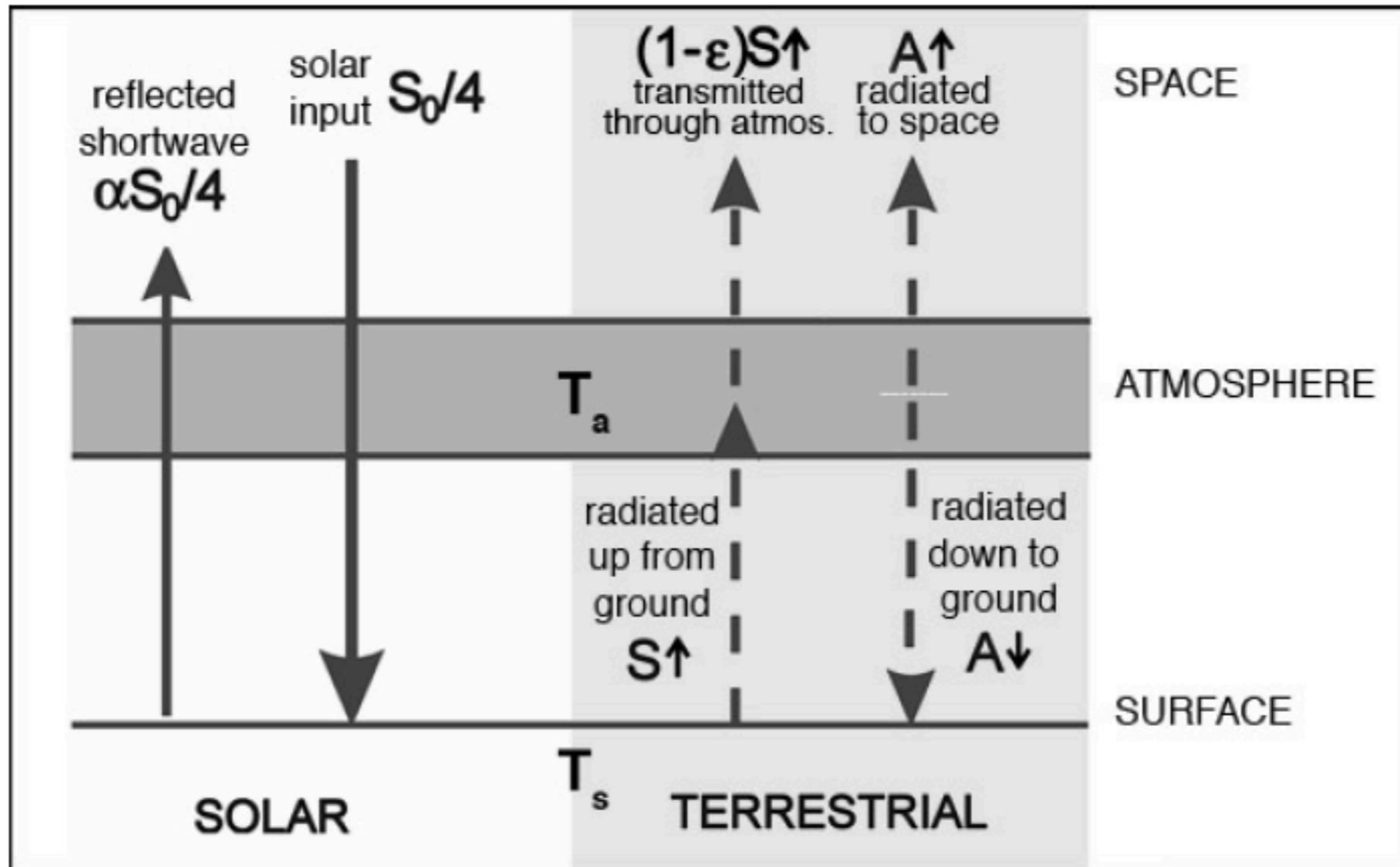


Figure 2.8: A leaky greenhouse. In contrast to Fig.2.7, the atmosphere now absorbs only a fraction, ϵ , of the terrestrial radiation upwelling from the ground.

Simple 1D climate model



$$T_s = \left(\frac{1}{1-\epsilon/2} \right)^{\frac{1}{4}} T_e = 1.136 * 255 = 289K$$

Simple 1D climate model

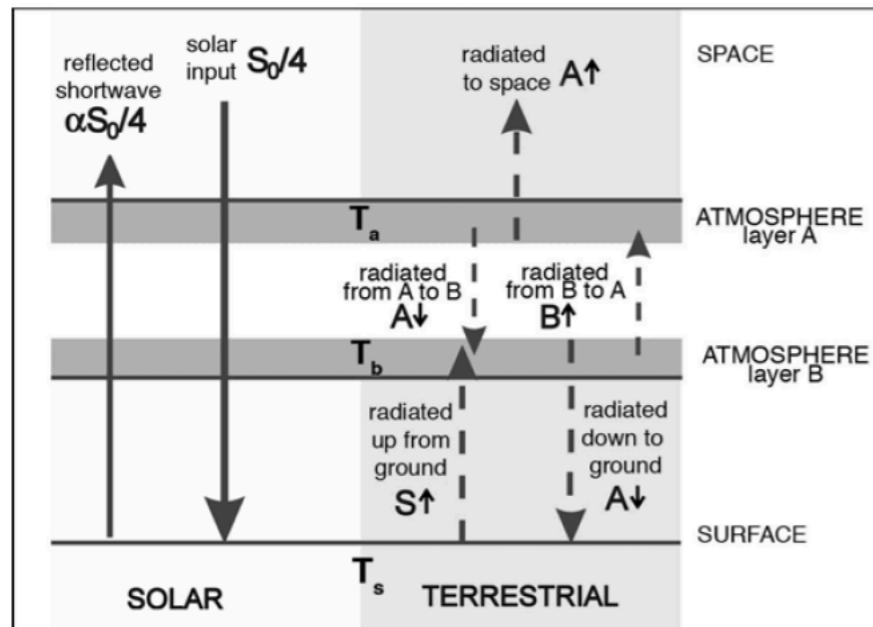


Figure 2.9: An 'opaque' greenhouse made up of two layers of atmosphere. Each layer completely absorbs the IR radiation impinging on it.

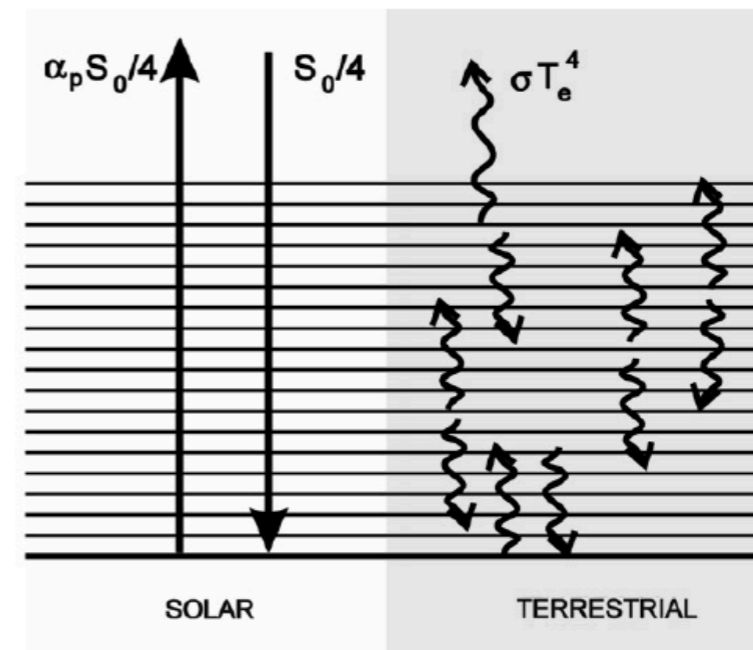


Figure 2.10: Schematic of radiative transfer model with many layers.

Introducing Climate Sensitivity

Climate sensitivity was introduced to measure the global response of the climate system to a change in the radiative forcing inputs to the system.

“To make progress on a problem with the intimidating complexity of climate change, the proper response of a scientist is to begin by considering simple questions and then add complexity as understanding is gained. The lessons drawn from these simple models must be taken seriously, but with full realization that they may not be a faithful representation of nature.”

Quote from Hartmann, GPC, Chapter 9, p. 229

A Quick Estimate of Climate Sensitivity

- Defined as:
$$\frac{dT_s}{dQ} = \frac{\partial T_s}{\partial Q} + \frac{\partial T_s}{\partial y_i} \frac{dy_i}{dQ}$$
- This is the response in surface temperature to a change in the energy increase.
- Now let's return to the “leaky” atmosphere model and consider the sensitivity to the emissivity.

$$\frac{\partial T_s}{\partial \epsilon} = -\frac{T_e}{2} (1 - \epsilon/2)^{-\frac{5}{4}}$$

- Suppose emissivity decreases by 1% (e.g., doubled [CO₂]), this implies a temperature change of 2.4 K

A Hierarchy of Climate Models

- **“Horses for Courses” - Jake Jacoby**
 - **Models suited for specific purposes**
- **Energy Balance Models**
 - **great for concepts, good for uncertainty**
 - **okay for feedbacks with other components**
- **Global Climate Models**
 - **great for processes/feedbacks, good for predictions, but poor for uncertainty (just too expensive)**
- **Need for Intermediate Complexity**
 - **Difficult to define these days...**

The 16th slide (via Google Images search)



EMICs: A History

Climate Dynamics (2002) 18: 579–586
DOI 10.1007/s00382-001-0200-1

M. Claussen · L. A. Mysak · A. J. Weaver · M. Crucifix
T. Fichefet · M.-F. Loutre · S. L. Weber · J. Alcamo
V. A. Alexeev · A. Berger · R. Calov · A. Ganopolski
H. Goosse · G. Lohmann · F. Lunkeit · I. I. Mokhov
V. Petoukhov · P. Stone · Z. Wang

**Earth system models of intermediate complexity: closing
the gap in the spectrum of climate system models**

EMICs: A History

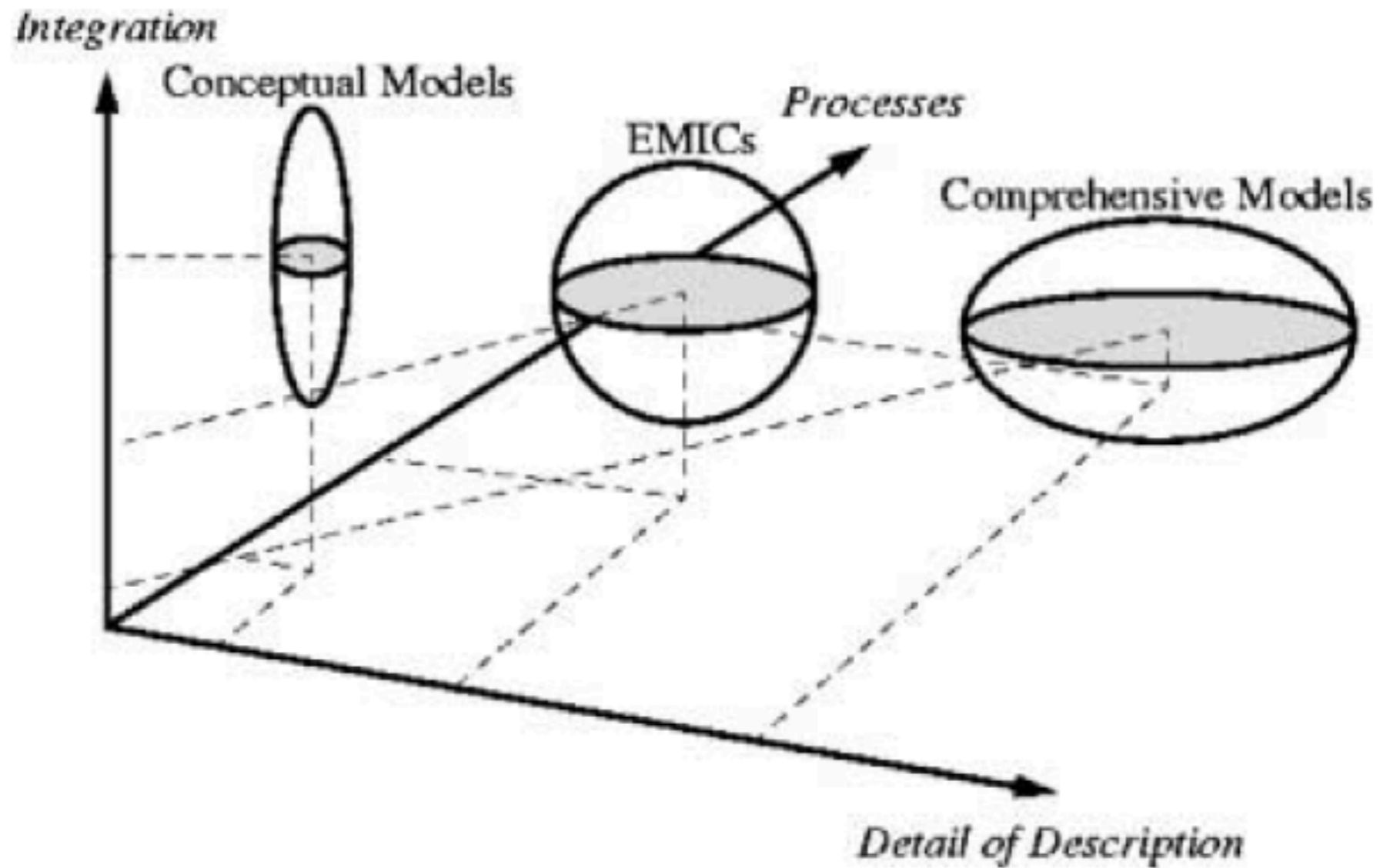


Fig. 1. Pictorial definition of EMICs. Adapted from Claussen (2000)

Claussen et al (2002)

EMICs: A History

Table 1. References to EMICs

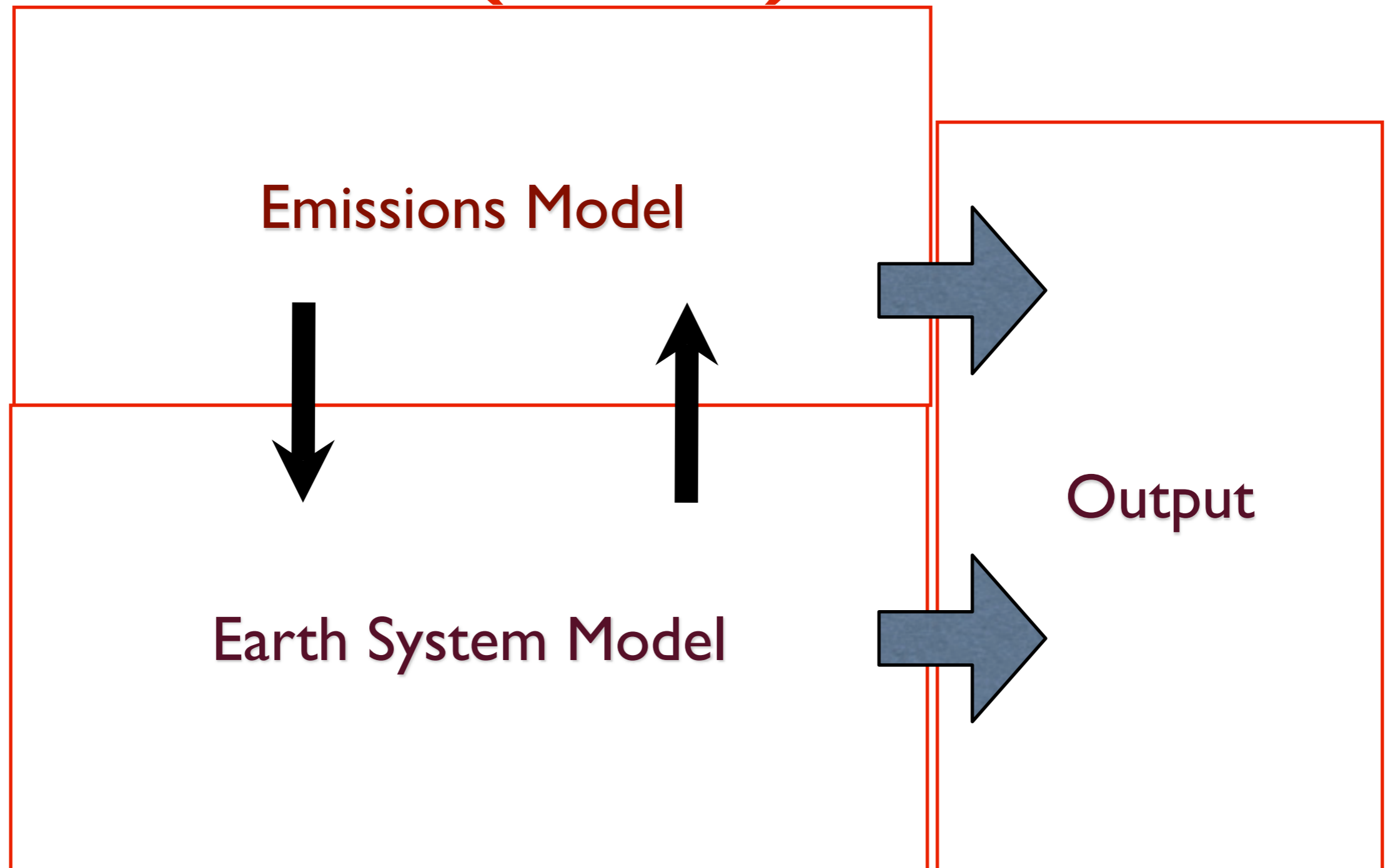
Model	Short list of references
1: Bern 2.5D	Stocker et al. (1992), Marchal et al. (1998)
2: CLIMBER-2	Petoukhov et al. (2000), Ganopolski et al. (2000)
3: EcBilt	Opsteegh et al. (1998)
4: EcBilt-CLIO	Goosse et al. (2000)
5: IAP RAS	Petoukhov et al. (1998), Handorf et al. (1999), Mokhov et al. (2000)
6: MPM	Wang and Mysak (2000), Mysak and Wang (2000)
7: MIT	Prinn et al. (1999), Kamenkovich et al. (2000)
8: MoBidiC	Crucifix et al. (2000a)(2000b)
9: PUMA	Fraedrich et al. (1998), Maier-Reimer et al. (1993)
10: Uvic	Weaver et al. (2000)
11: IMAGE 2	Alcamo (1994), Alcamo et al. (1996)

EMICs: A History

Table 2. Interactive components of the climate system being implemented into EMICs (for explanation see text)

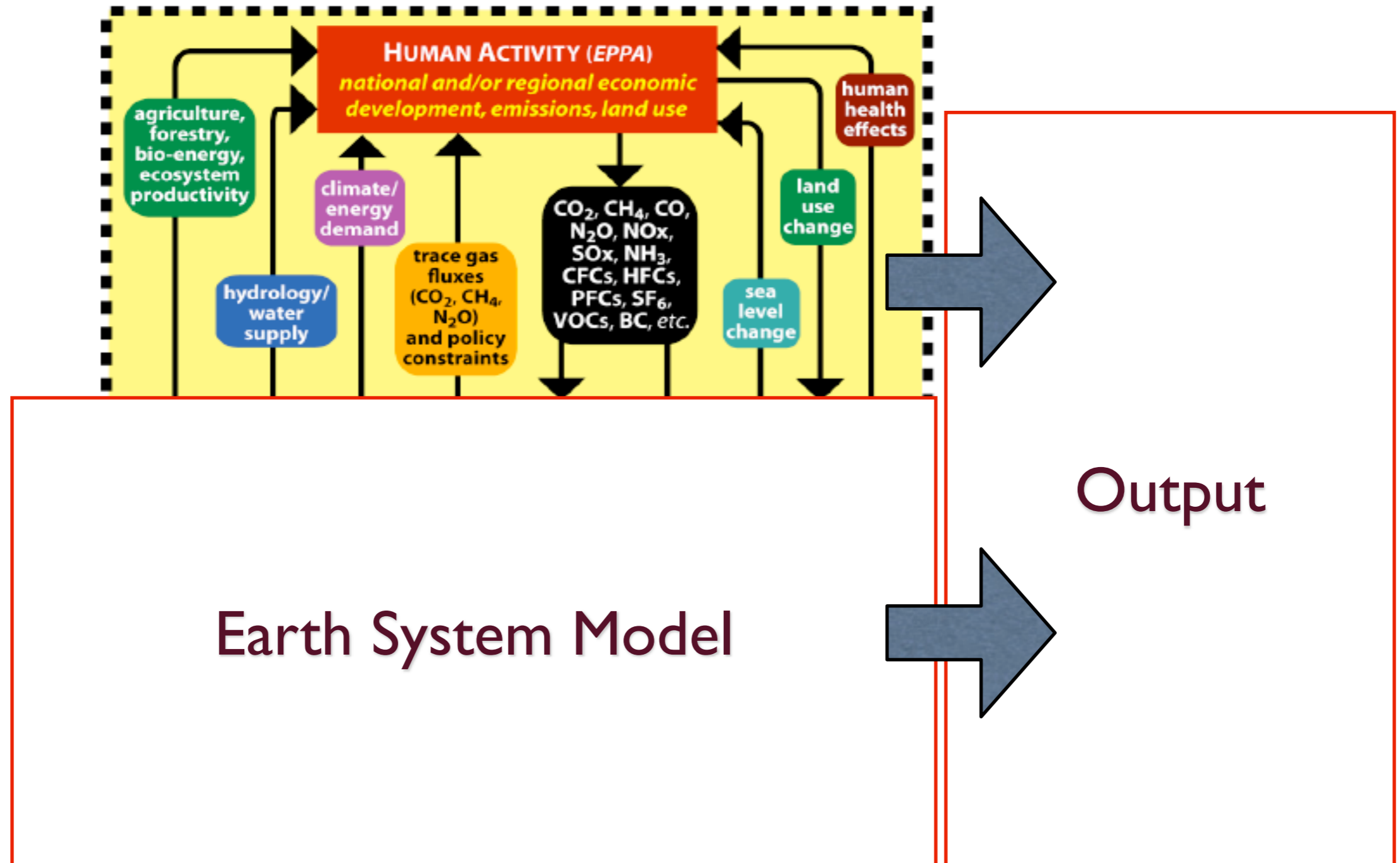
Model	Atmosphere	Ocean	Biosphere
1	EMBM, 1-D(φ)	2-D(φ, z), 3 basins	B _O , B _T
2	SDM, 2-D(φ, λ)-mL	2-D(φ, z) 3 basins	B _O , B _T , B _V
3	QG, 3-D, T21, L3	3-D, 5.6°×5.6°, L12	
4	QG, 3-D, T21-L3	3-D, 3°×3°	B _T , B _V
5	SDM, 3-D 4.5°×6°, L8	SDM, 2-D(φ, λ) 4.5°×6°, L3 fixed salinity	
6	EMBM, 1-D(φ), land/ocean boxes	2-D(φ, z), 3 basins	
7	SDM, 2-D(φ, z)/atmospheric chemistry	3-D, 4°×1.25° to 3.75°, L15	B _T
8	QG, 2-D(φ, z)-L2	2-D(φ, z), 3 basins	B _O , B _T , B _V
9	GCM, 3-D, T21, L5	3-D, 5°×5°, L11	B _O
10	DEMBM, 2-D(φ, λ)	3-D, 3.6°×1.8°, L 19	
11	DEMBM, 2-D(φ, λ)/atmospheric chemistry	2-D(φ, z), 2 basins	B _O , B _T , B _V

MIT Integrated Global System Model (IGSM2)



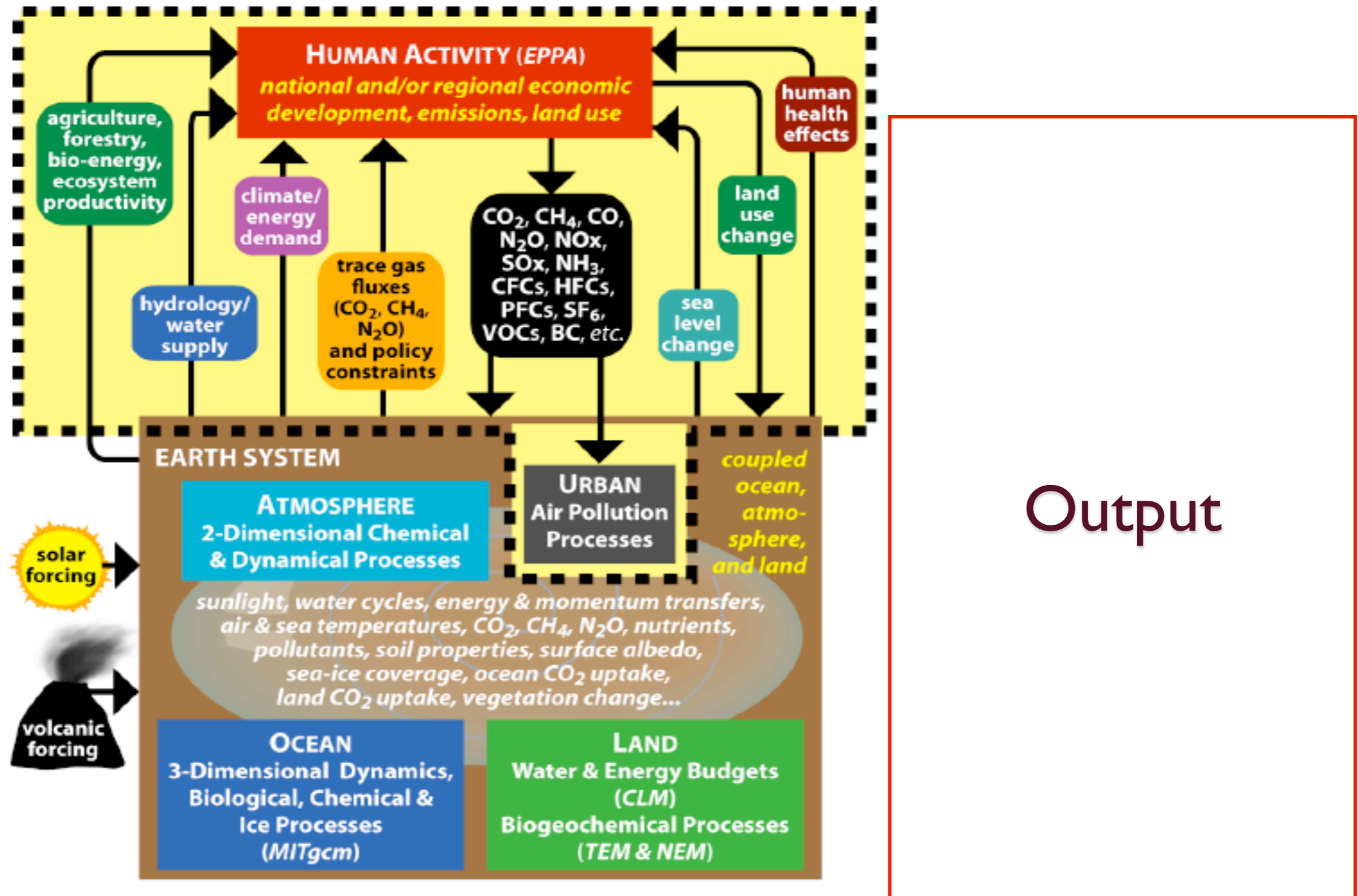
Sokolov et al. (2005, JP-Report 124)

MIT Integrated Global System Model (IGSM2)



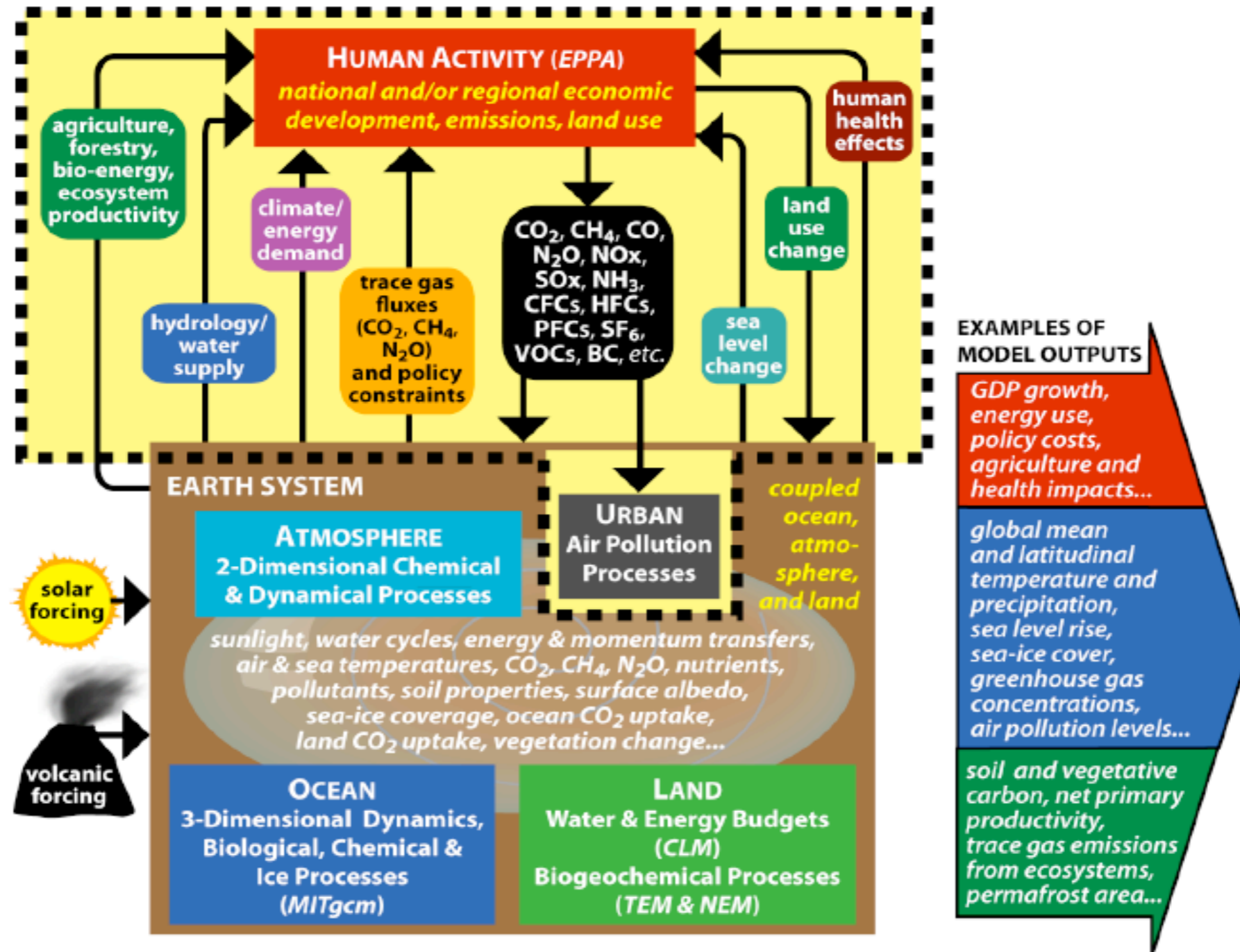
Sokolov et al. (2005, JP-Report 124)

MIT Integrated Global System Model (IGSM2)



Sokolov et al. (2005, JP-Report 124)

MIT Integrated Global System Model (IGSM2)



Sokolov et al. (2005, JP-Report 124)

MIT IGSM Overview

- **Atmospheric Model (latitude-height)**
 - **Statistical Dynamic Model**
 - **Physical and Chemical Model**
- **Ocean/Sea Ice Model (2D or 3D)**
 - **Physical, Biogeochemical, Biological Models**
- **Land Model**
 - **Physical, Biological, Biogeochemical**
 - **Terrestrial Ecosystem Model (Melillo et al.)**
- **MIT EPPA - Economic Model**

GISS Model II (8x10 grid) (basis for MIT IGSM Climate Model)

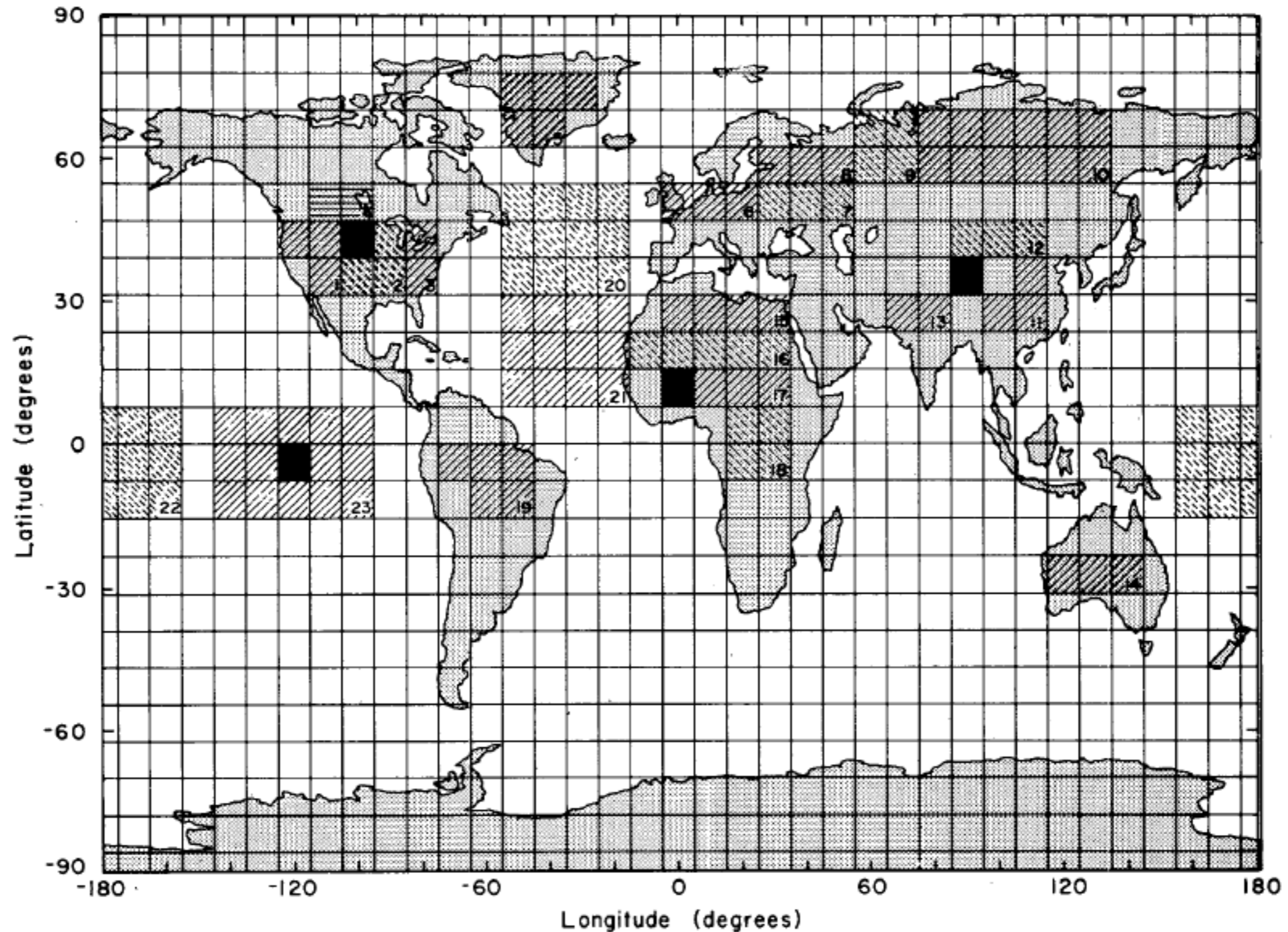


FIG. 1. Grid spacing for $8^\circ \times 10^\circ$ model. Shadings indicate one choice of regions for special monthly diagnostics. The four black regions are a particular choice of gridboxes for hourly diagnostics.

IGSM Pressure Levels (9 levels)

Number of boxes:
 $24 \times 36 \times 9 = 7776$

Current models:
 $\sim 10^7$

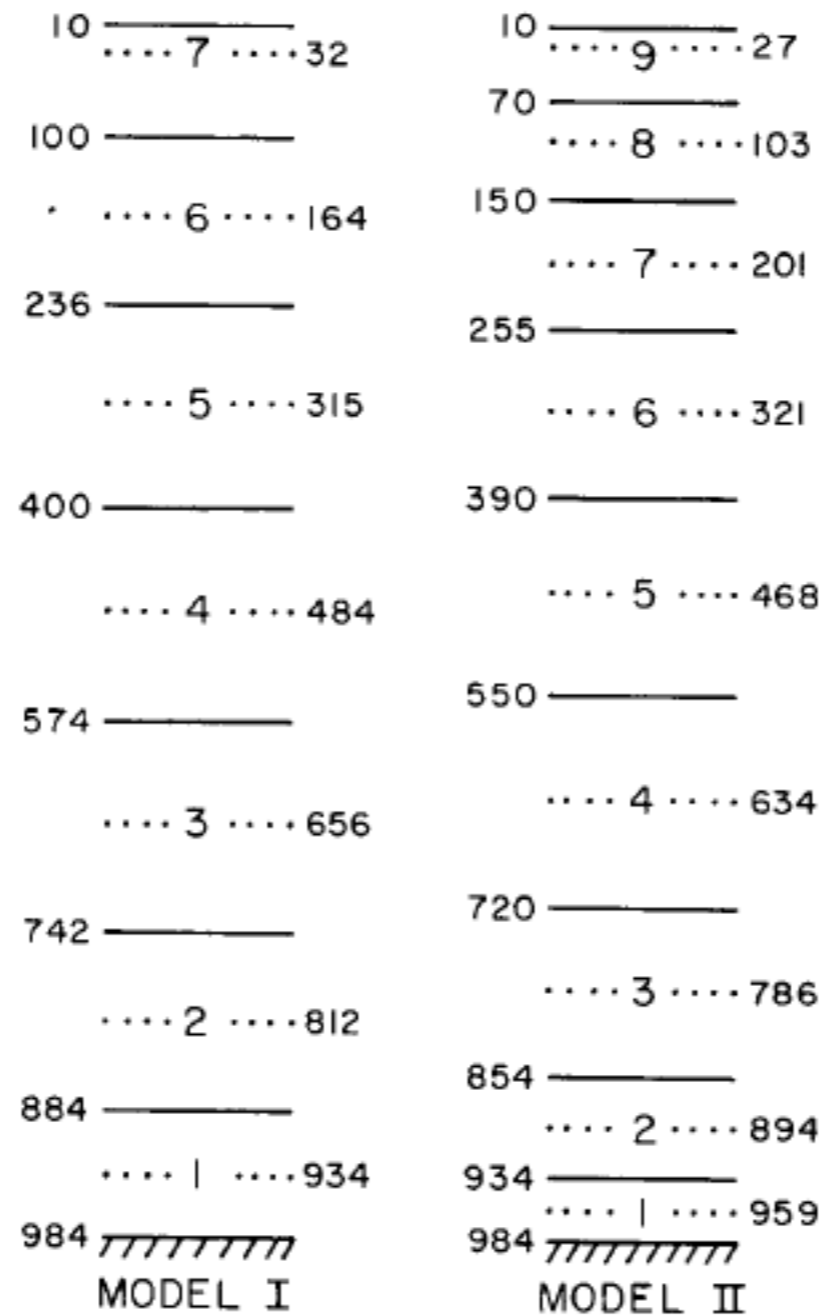


FIG. 2. Global-mean pressure levels (mb) for 7-layer Model I and 9-layer Model II. Layer edges are solid lines, and interior levels are dotted.

GISS/IGSM Model Grid Box

- Statistical
Components
of Model
- Clouds
 - Precipitation

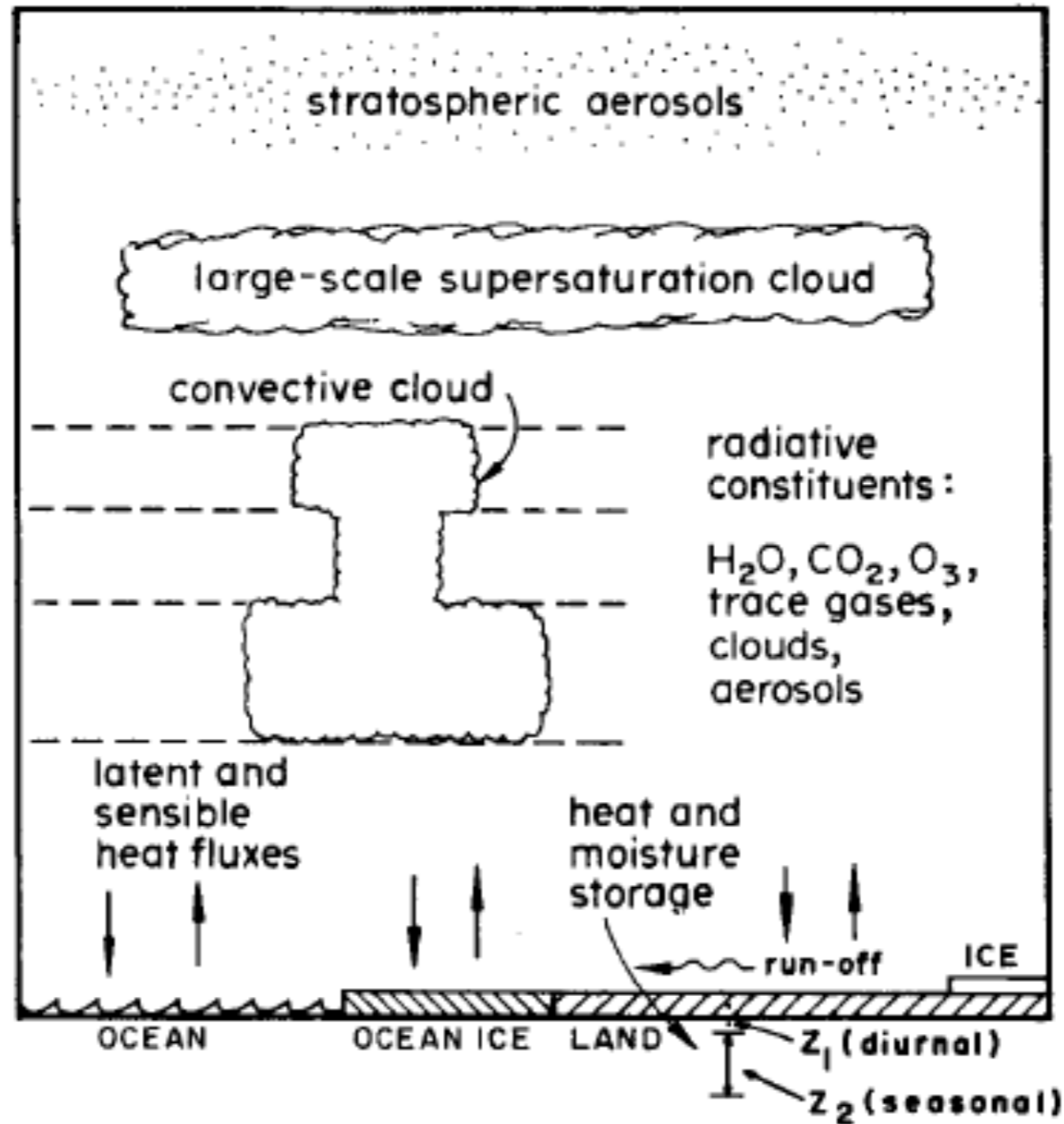


FIG. 3. Schematic illustration of model structure at a single gridbox.

GISS Model Equations (needed to include eddy fluxes)

TABLE 2. Approximate form of fundamental equations employed in computations.

	Dependent variable	Equation	Additional notation
$\frac{\partial \pi \mathbf{U}}{\partial t} = -\nabla \cdot \pi \mathbf{U} \mathbf{U} - \frac{\partial \pi \dot{\sigma} \mathbf{U}}{\partial \sigma} - \left(f + \frac{\tan \phi}{a} u \right) \mathbf{k} \times \pi \mathbf{U} - \pi \nabla \Phi - \frac{\sigma \pi}{\rho} \nabla \pi + \pi \mathbf{F}_\sigma$	U	(T5)	
$\frac{\partial \pi}{\partial t} = -\nabla \cdot \pi \mathbf{U} - \frac{\partial \pi \dot{\sigma}}{\partial \sigma}$	$\pi, \dot{\sigma}$	(T6)	
$\frac{\partial \pi \Theta}{\partial t} = -\nabla \cdot \pi \mathbf{U} \Theta - \frac{\partial \pi \dot{\sigma} \Theta}{\partial \sigma} + \frac{\pi \Theta}{c_p T} Q$	Θ	(T7)	
$\frac{\partial \pi q}{\partial t} = -\nabla \cdot \pi \mathbf{U} q - \frac{\partial \pi \dot{\sigma} q}{\partial \sigma} + \pi (E - C)$	q	(T8)	
$\frac{\partial \Phi}{\partial \sigma} = -\frac{\pi}{\rho}$	Φ	(T9)	
$p = \rho R T$	ρ	(T4)	
			<p>U horizontal velocity in σ coordinates, with components (u, v)</p> <p>$\nabla \cdot \pi \mathbf{U} \mathbf{U}$ vector with components [$\nabla \cdot (\pi \mathbf{U} u), \nabla \cdot (\pi \mathbf{U} v)$]</p> <p>$\nabla \cdot A \mathbf{U}$ horizontal divergence $\left[= \frac{1}{a \cos \phi} \left(\frac{\partial A u}{\partial \lambda} + \frac{\partial A v \cos \phi}{\partial \phi} \right) \right]$</p> <p>$\nabla A$ horizontal gradient $\left[= \frac{1}{a} \left(\frac{1}{\cos \phi} \frac{\partial A}{\partial \lambda}, \frac{\partial A}{\partial \phi} \right) \right]$</p> <p>$a$ mean planetary radius</p> <p>ϕ latitude in radians</p> <p>λ longitude in radians</p> <p>σ vertical coordinate [$= (p - p_t) / \pi$]</p> <p>E rate of evaporation</p> <p>$\dot{\sigma}$ $d\sigma/dt$</p> <p>π $p_s - p_t$, where p_s is the surface pressure and p_t the constant pressure at the model's upper boundary</p> <p>Θ potential temperature [$= T(p_{00}/p)^\kappa$, where $\kappa = R/c_p$ and $p_{00} = 1000$ mb]</p> <p>Φ geopotential [$= gz$, where z is height above sea level]</p> <p>f Coriolis parameter [$= 2\Omega \sin \phi$]</p> <p>c_p specific heat at constant pressure</p> <p>\mathbf{F}_σ horizontal frictional force</p> <p>q water vapor mixing ratio</p> <p>C rate of condensation.</p>

Development of 2D MIT Model required Eddy Flux parameterization

Development of a Two-Dimensional Zonally Averaged Statistical–Dynamical Model. Part I: The Parameterization of Moist Convection and its Role in the General Circulation

Development of a Two-Dimensional Zonally Averaged Statistical–Dynamical Model. Part II: The Role of Eddy Momentum Fluxes in the General Circulation and their Parameterization

Development of a Two-Dimensional Zonally Averaged Statistical–Dynamical Model. Part III: The Parameterization of the Eddy Fluxes of Heat and Moisture

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(Manuscript received 25 May 1989, in final form 16 November 1989)

MIT 2D Climate Model Description

- 2D statistical-dynamical atmospheric model derived from 3D GISS II AGCM (Sokolov and Stone, 1998) 46x11 (lat-height)
- Q-flux mixed layer ocean model where temperature anomalies are mixed into deep-ocean. Q-flux held fixed in transient runs.
- Adjustable model properties:
 - Climate Sensitivity (via adjustable cloud feedback)
 - Rate of deep-ocean heat uptake (via diffusivity)
 - Net Aerosol Forcing (via scattering cross-section or loading)

Wind fields from IGSM (Sokolov & Stone, 1998)

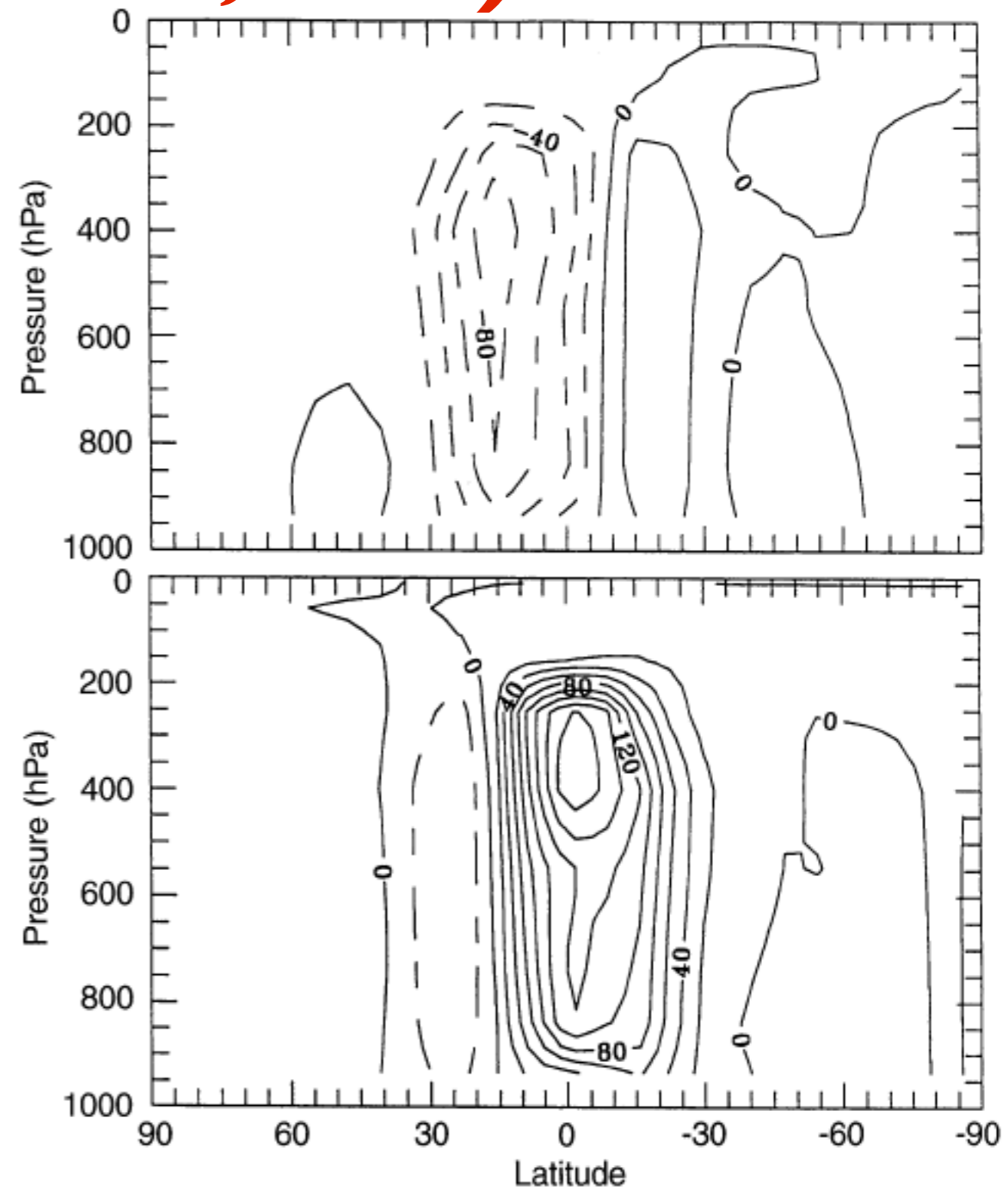
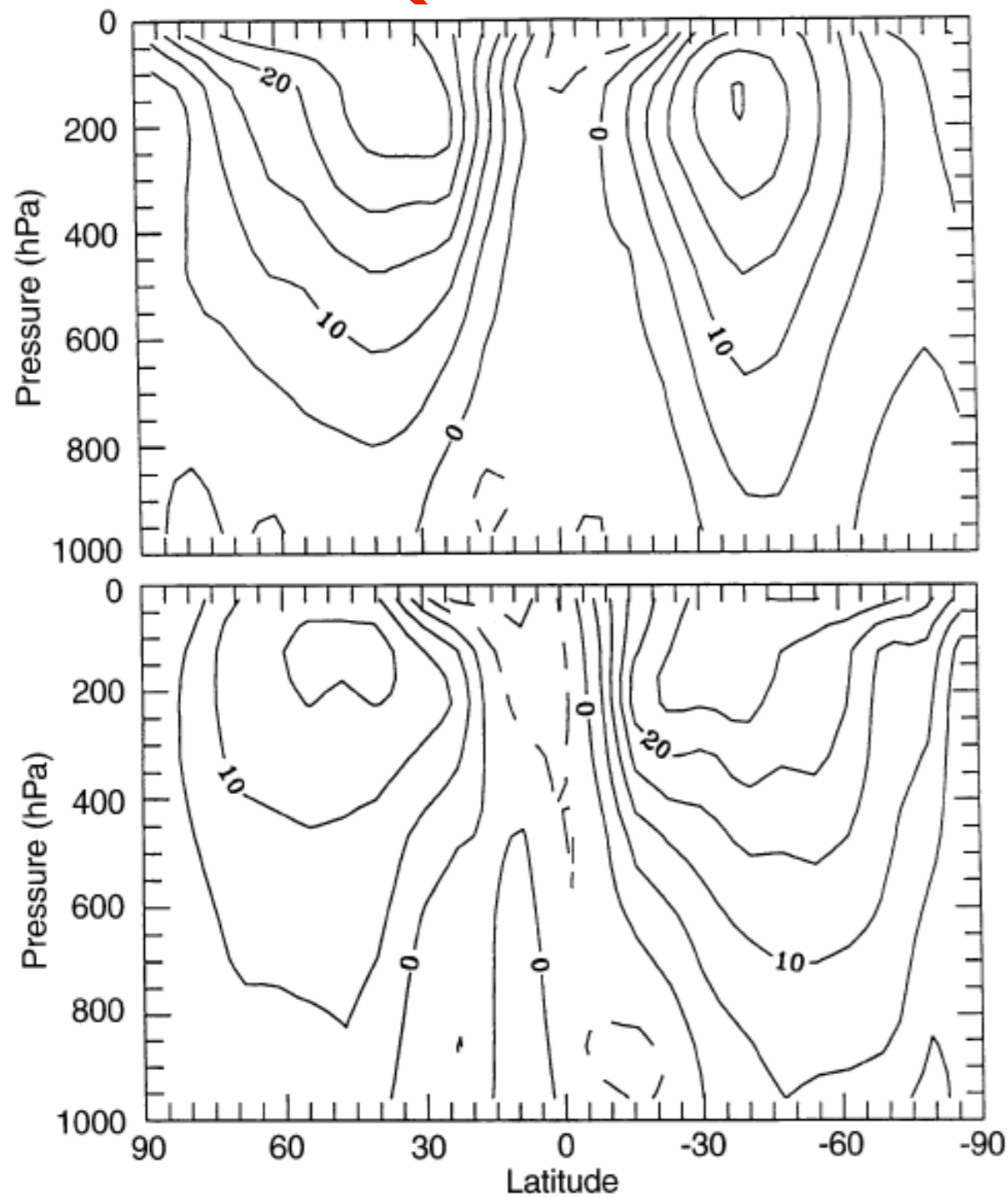


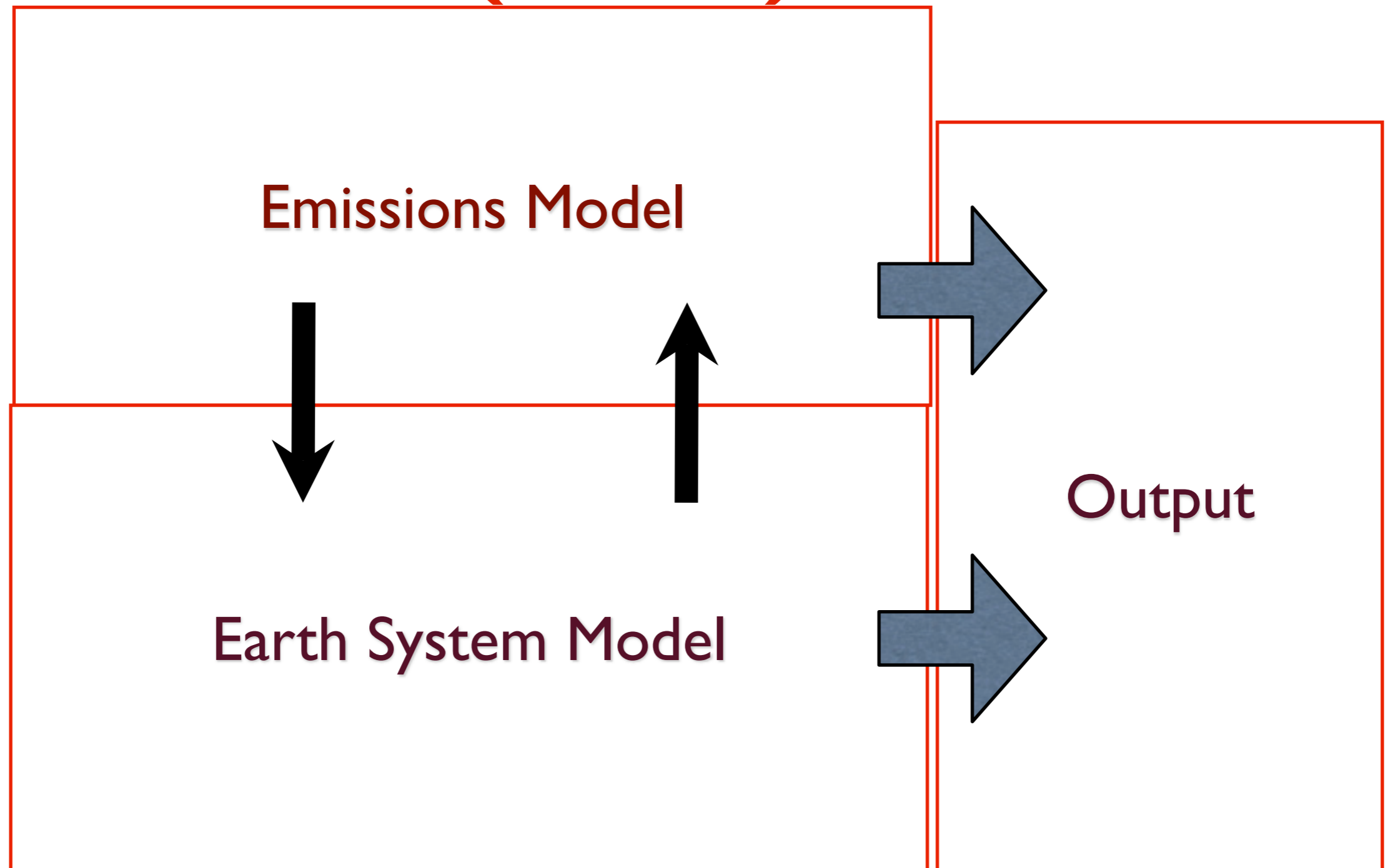
Fig. 1 Height-latitude cross section of zonal wind for DJF (top) and Fig. 2 As Fig. 1, but for meridional streamfunction JJA (bottom) from the MIT 2-D model

LUNCH BREAK

POST LUNCH BREAK

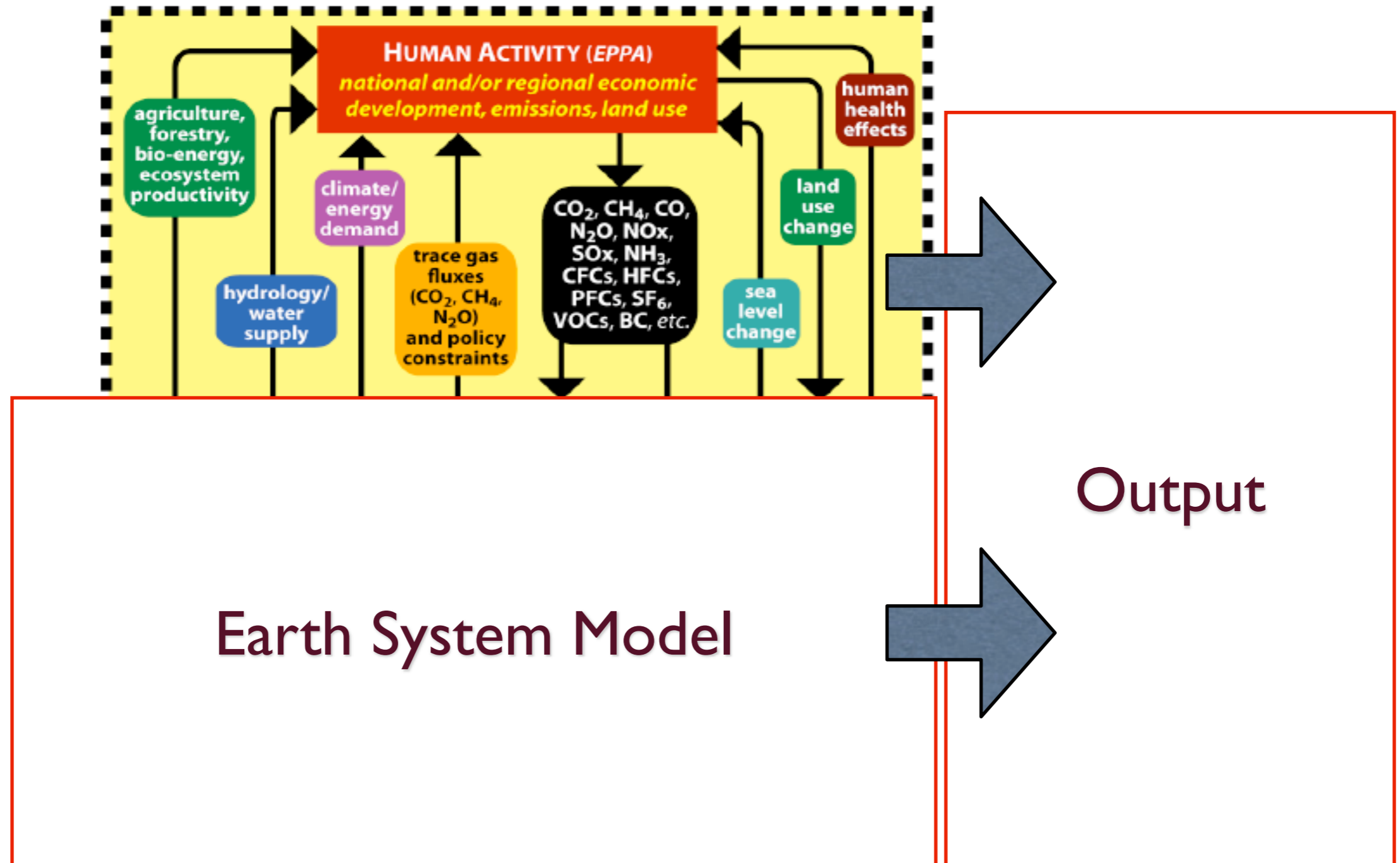
- **Using the IGSM**
 - **Uncertainty Analysis**
 - **Parameter/Property Estimation**
 - **Climate Model Evaluations**
 - **Scenario Analyses**
 - **Feedbacks and Response**

MIT Integrated Global System Model (IGSM2)



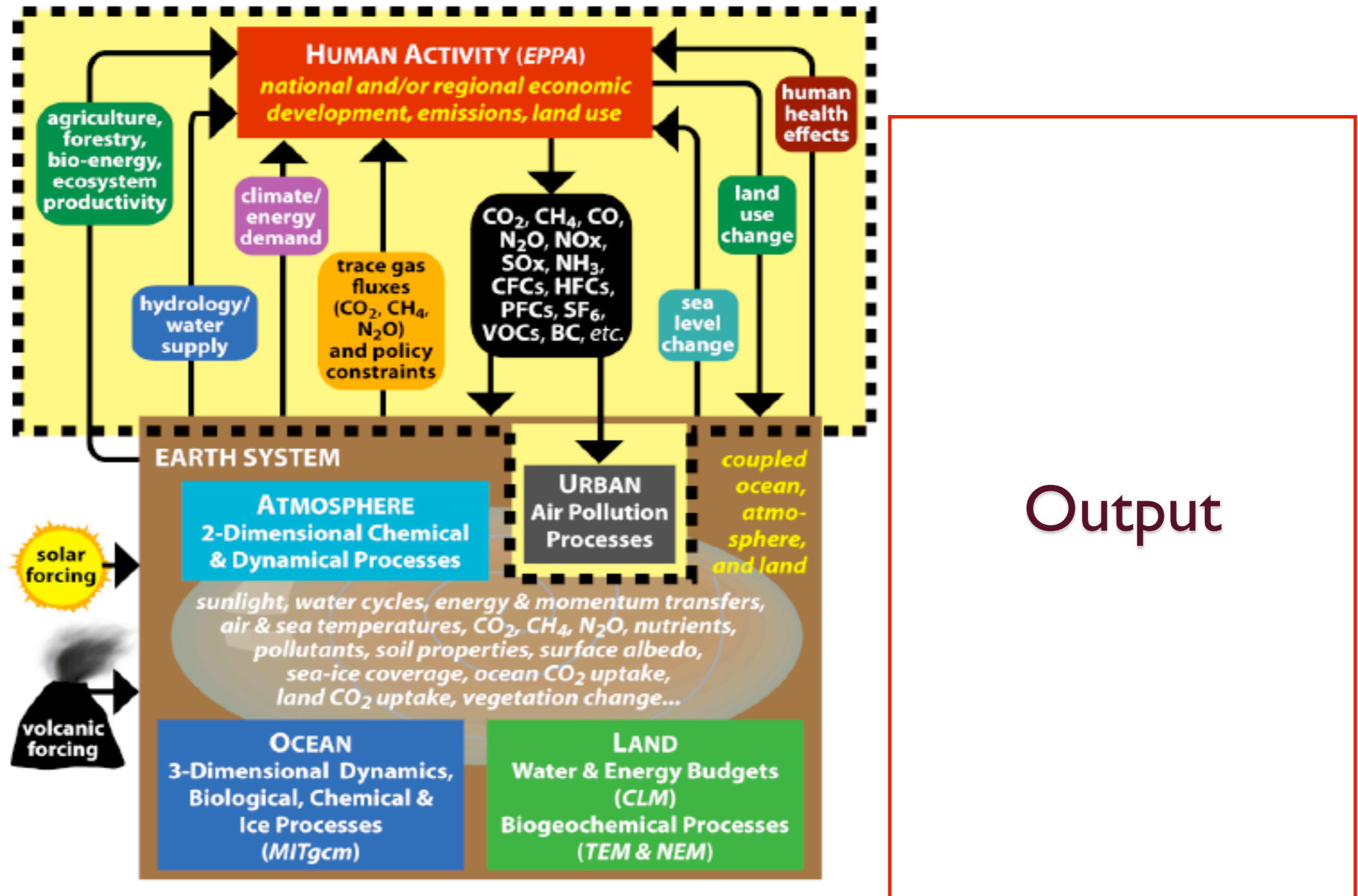
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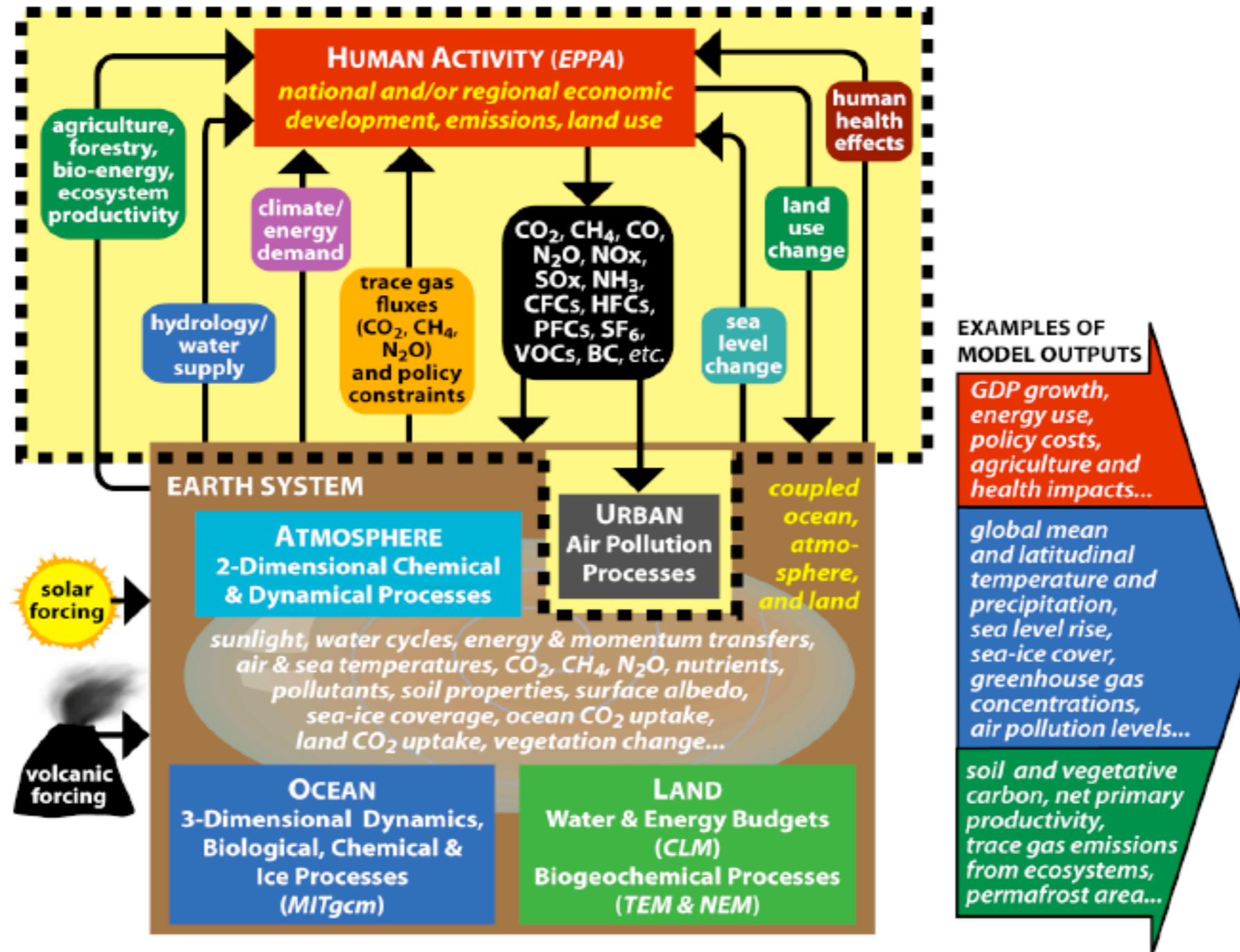
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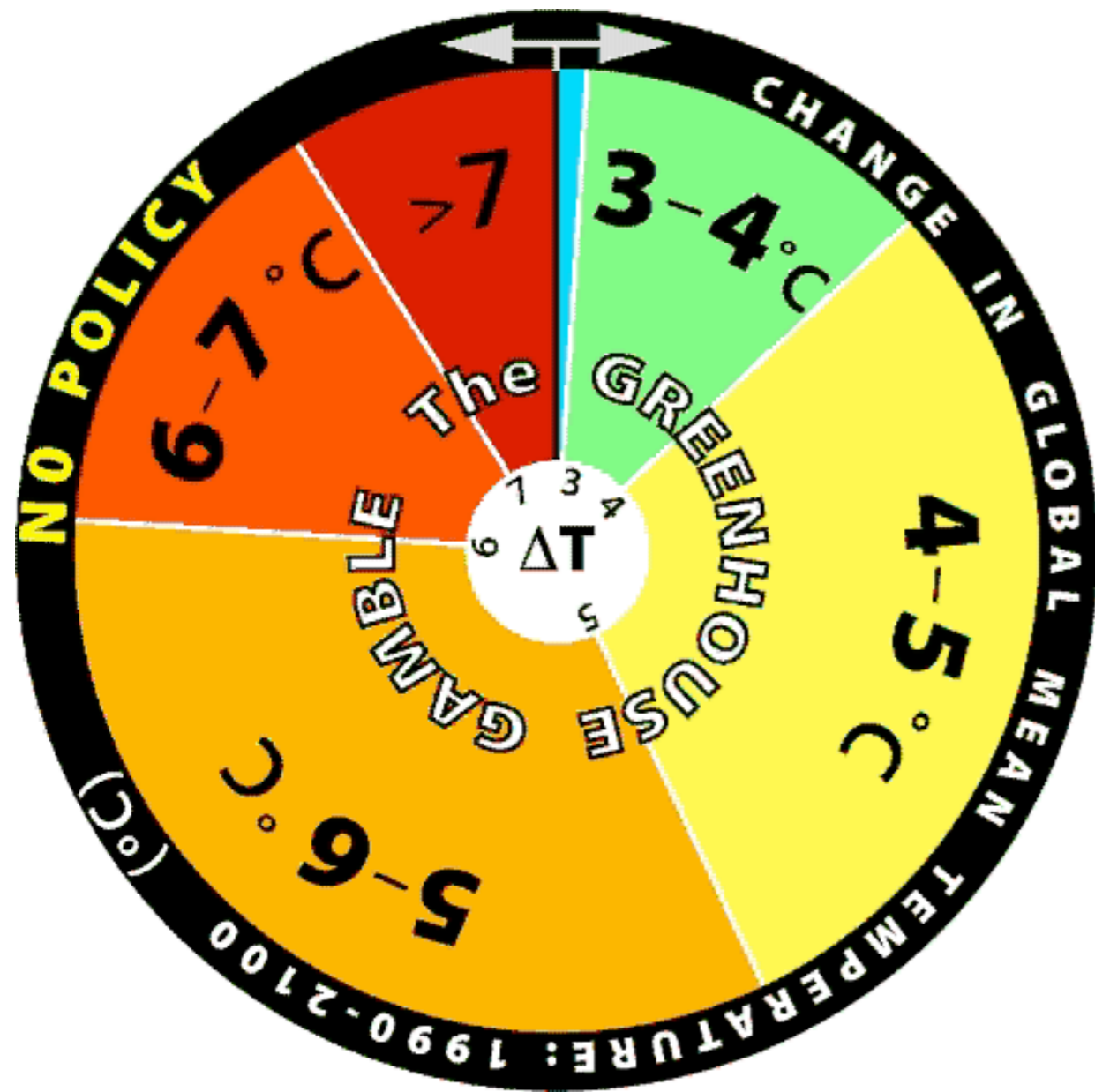
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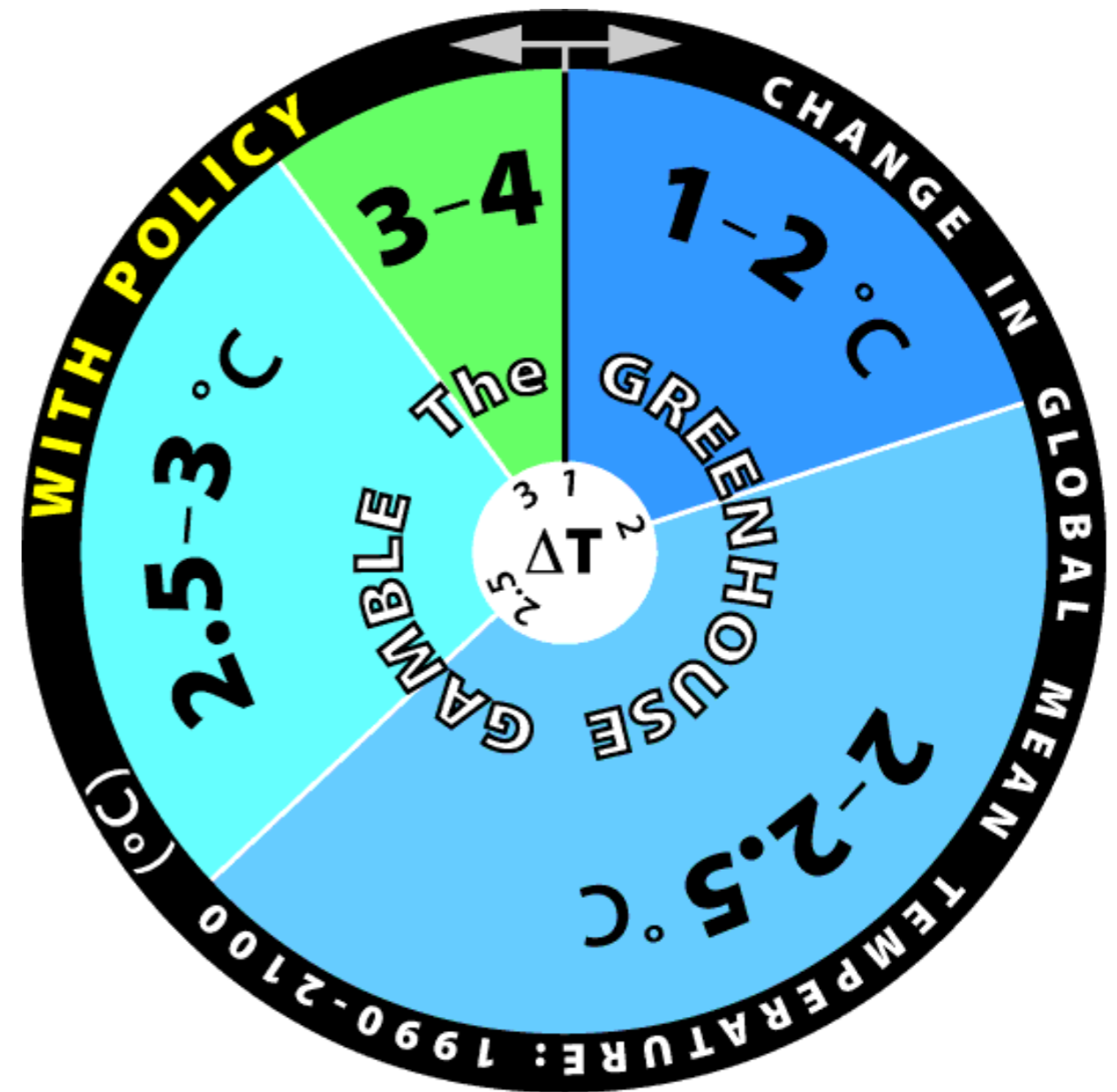
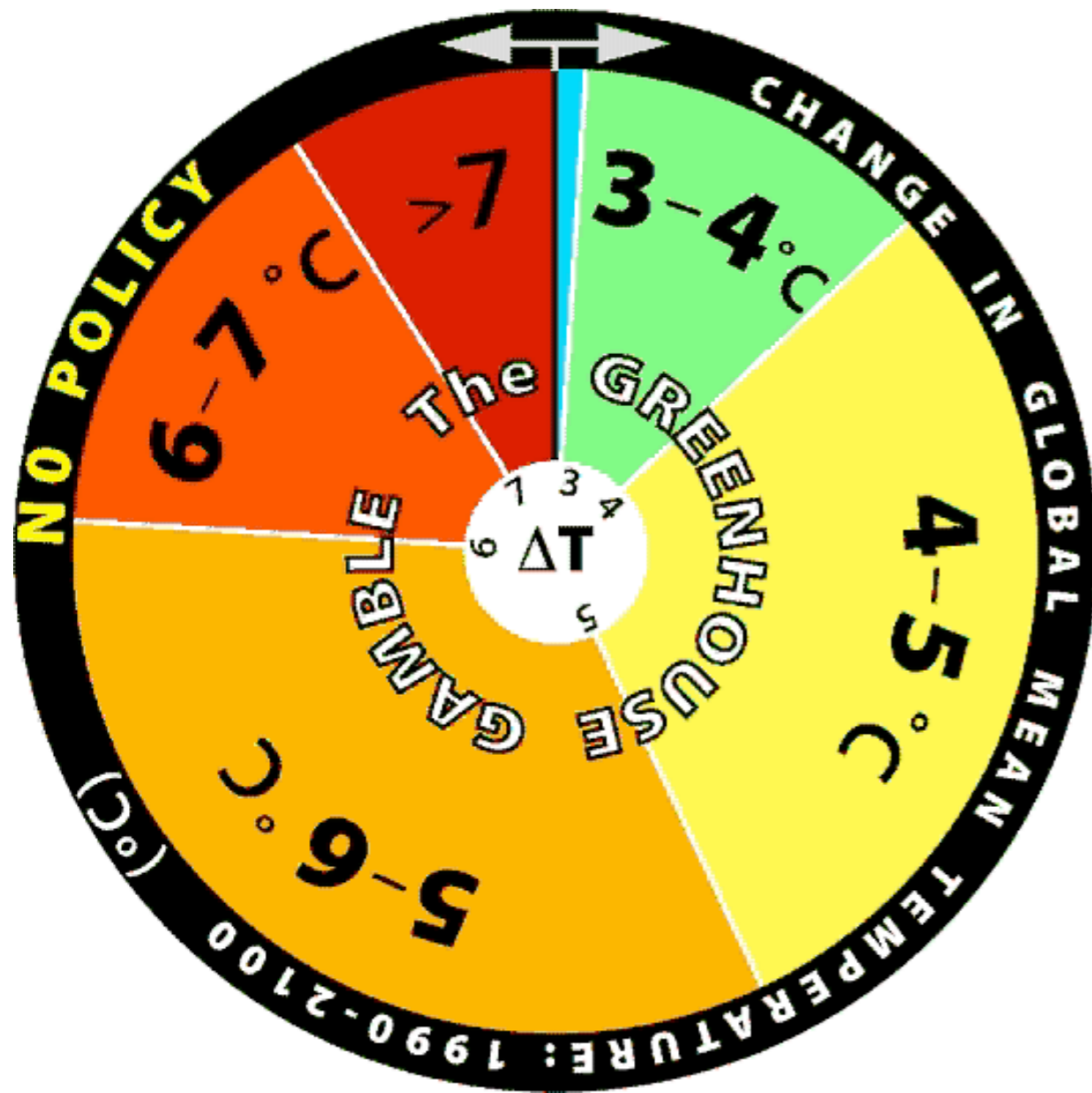
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The Greenhouse Gamble



<http://globalchange.mit.edu/resources/gamble/>

The Greenhouse Gamble



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Major Climate Projection Uncertainties

Consider the energy balance equation for the global-mean surface temperature anomaly (ΔT):

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

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Change in global
mean heat content

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Change in global
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Future
Forcings

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Net Feedbacks
 $\lambda = 1/S$

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Flux of heat
into deep-
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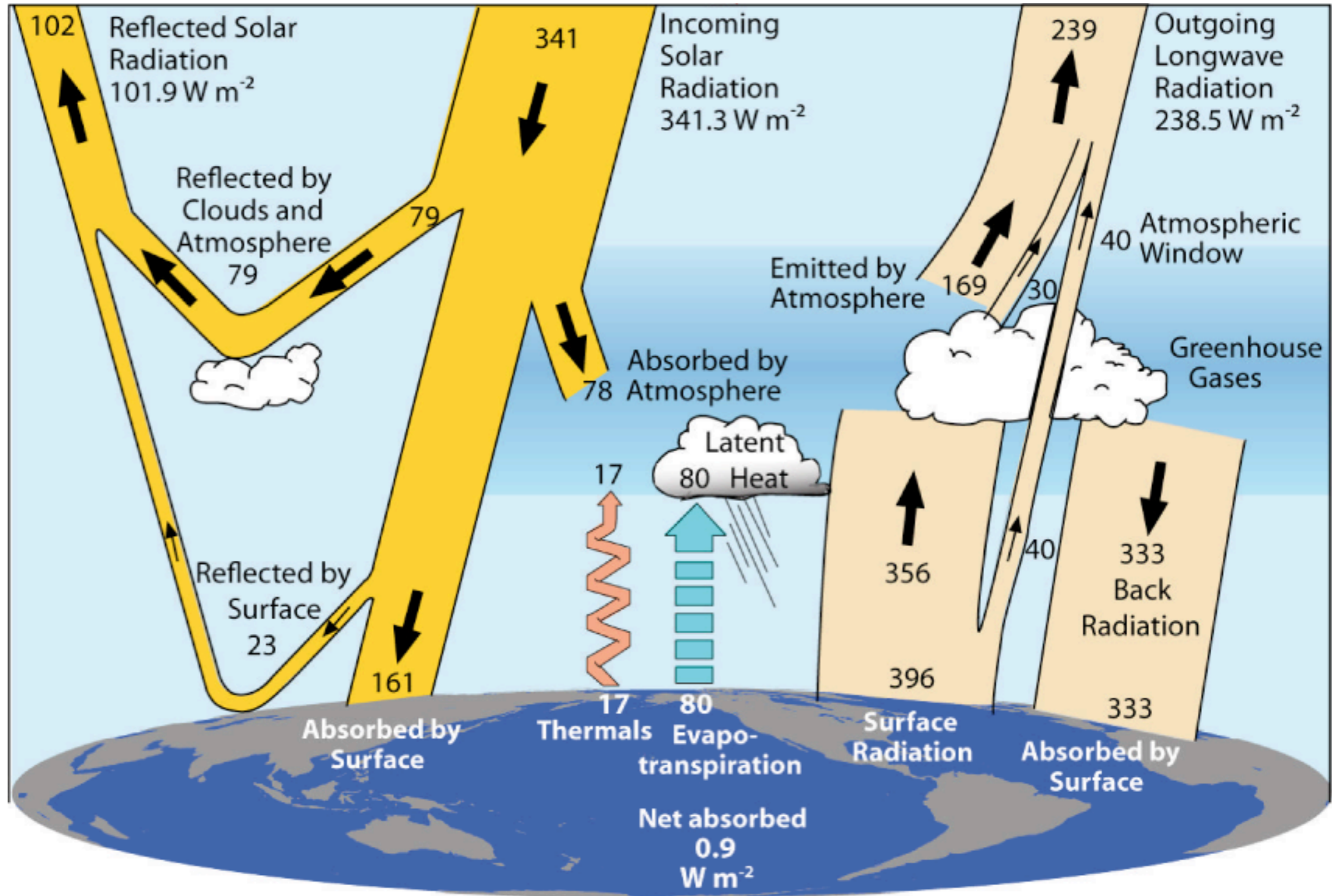
Net Feedbacks
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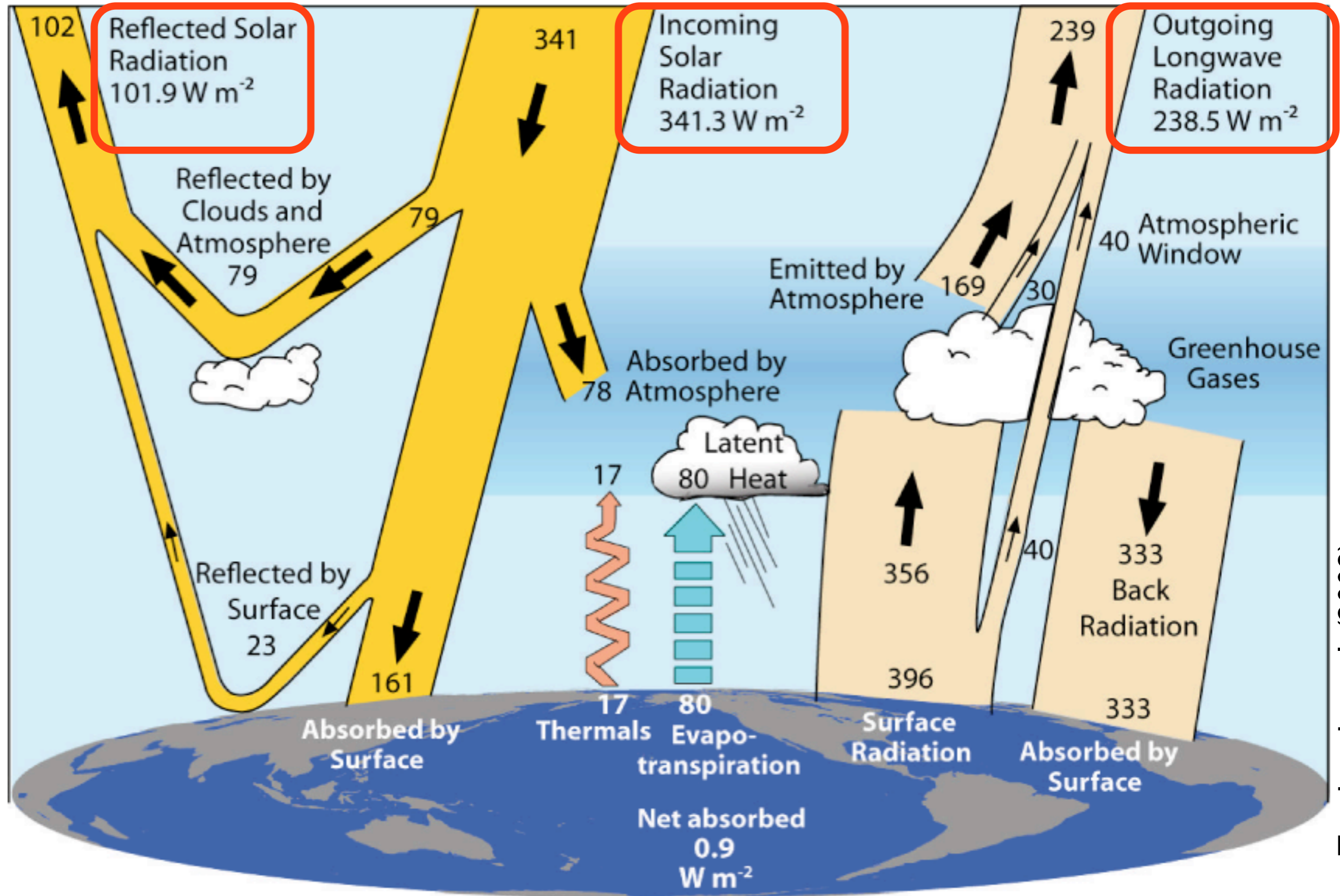
Conceptually: This is a good framework for organizing where the uncertainty exists.

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.

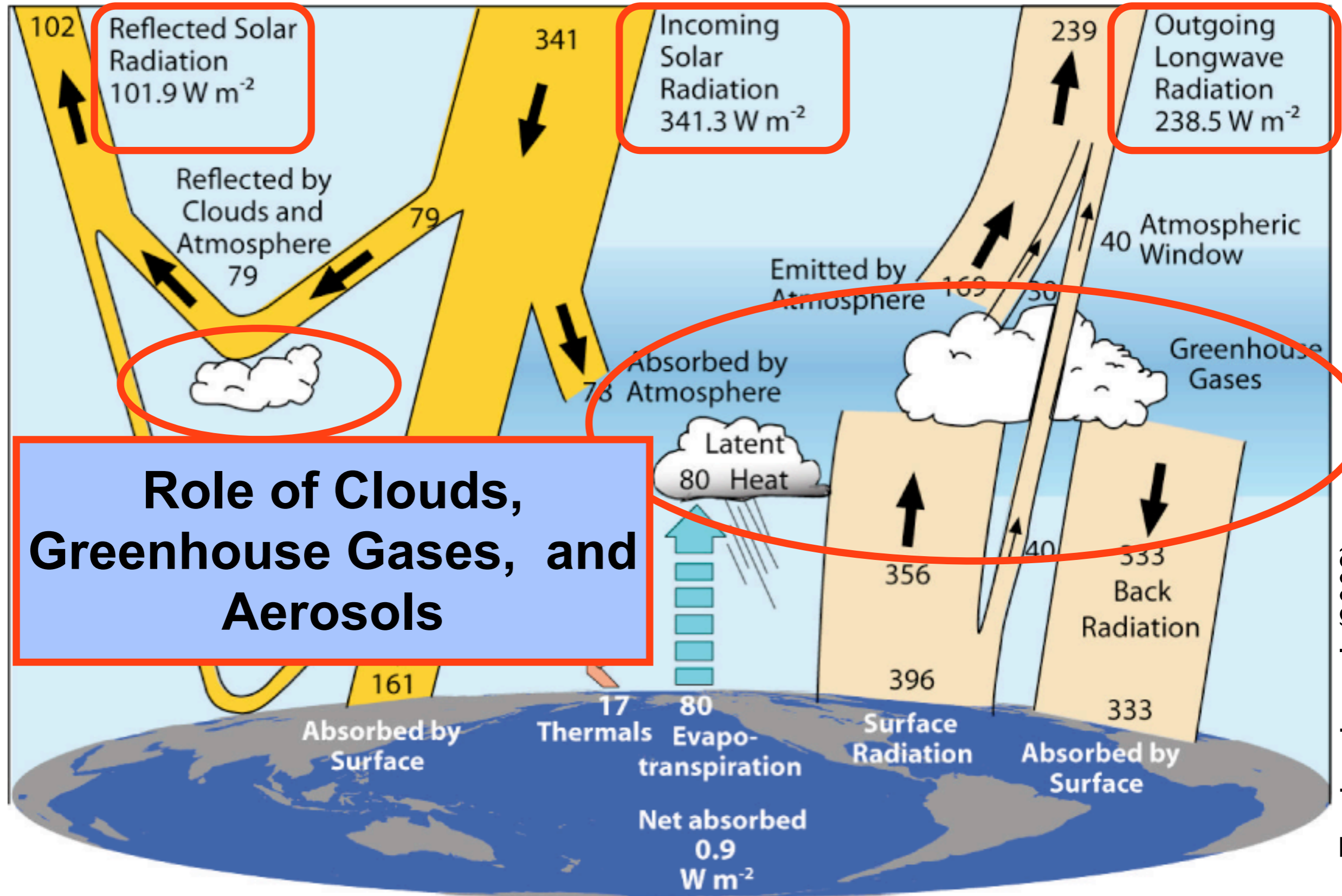
Global Climate System Energy Fluxes



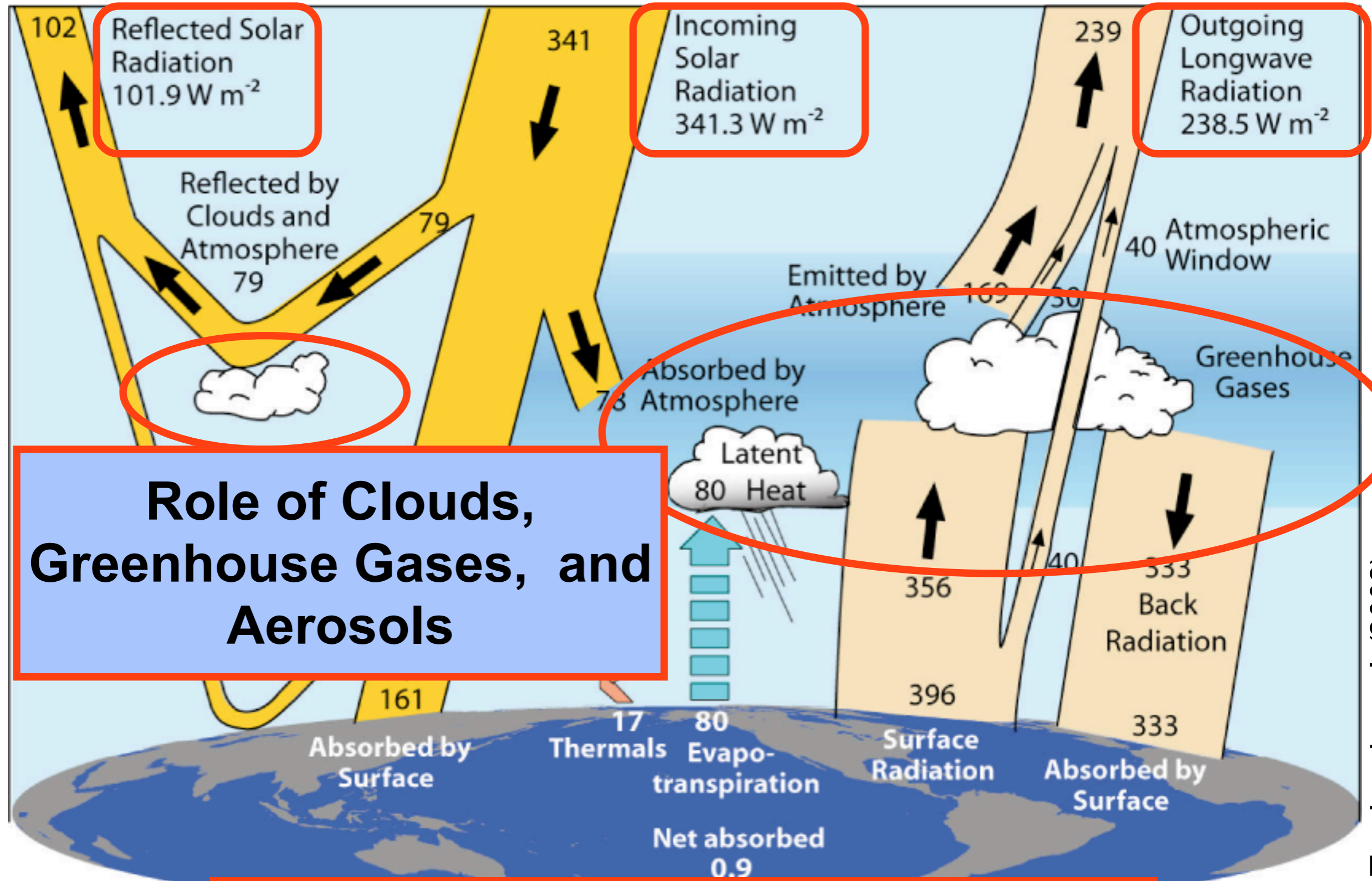
Global Climate System Energy Fluxes



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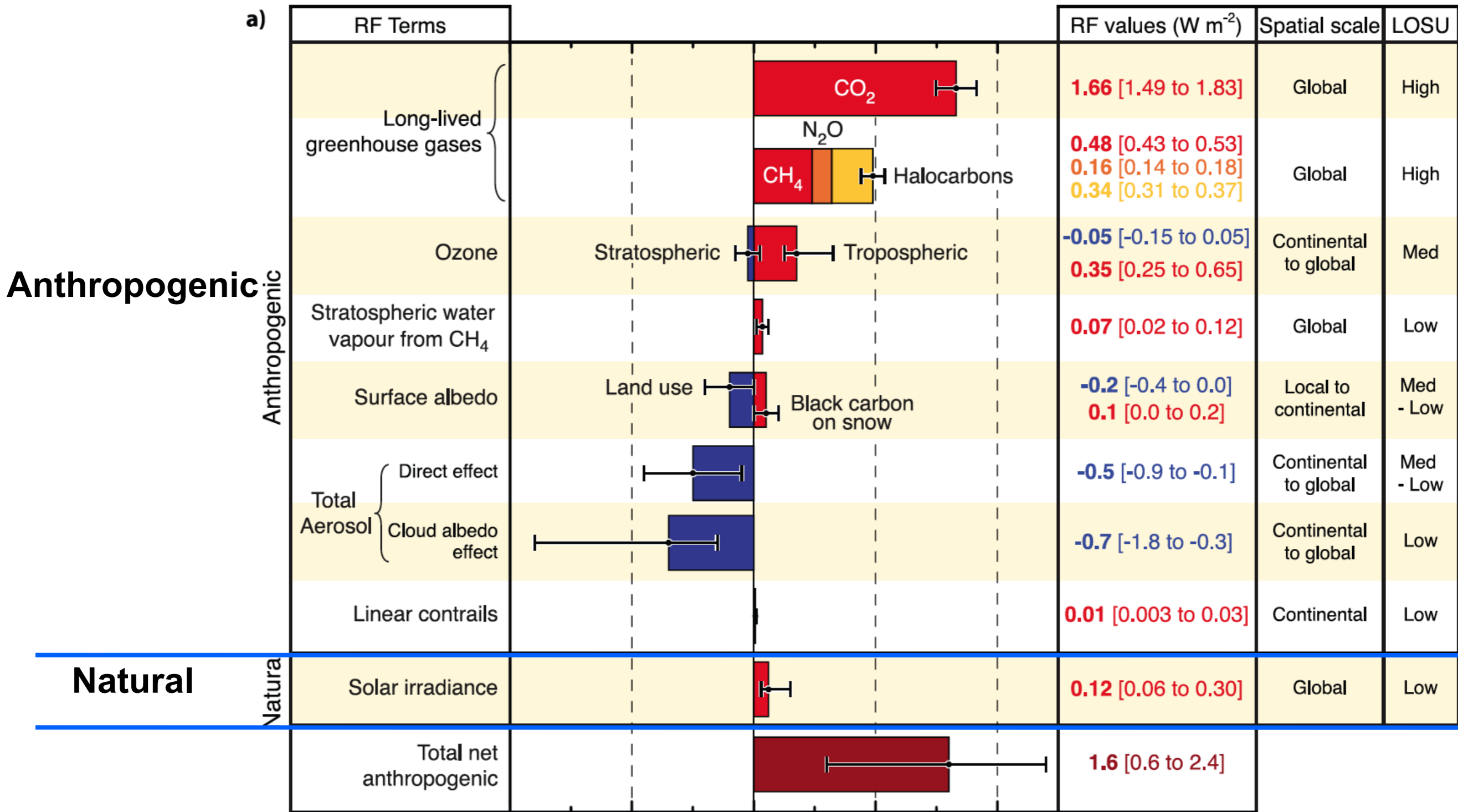


Role of Clouds, Greenhouse Gases, and Aerosols

Role of Land and Oceans

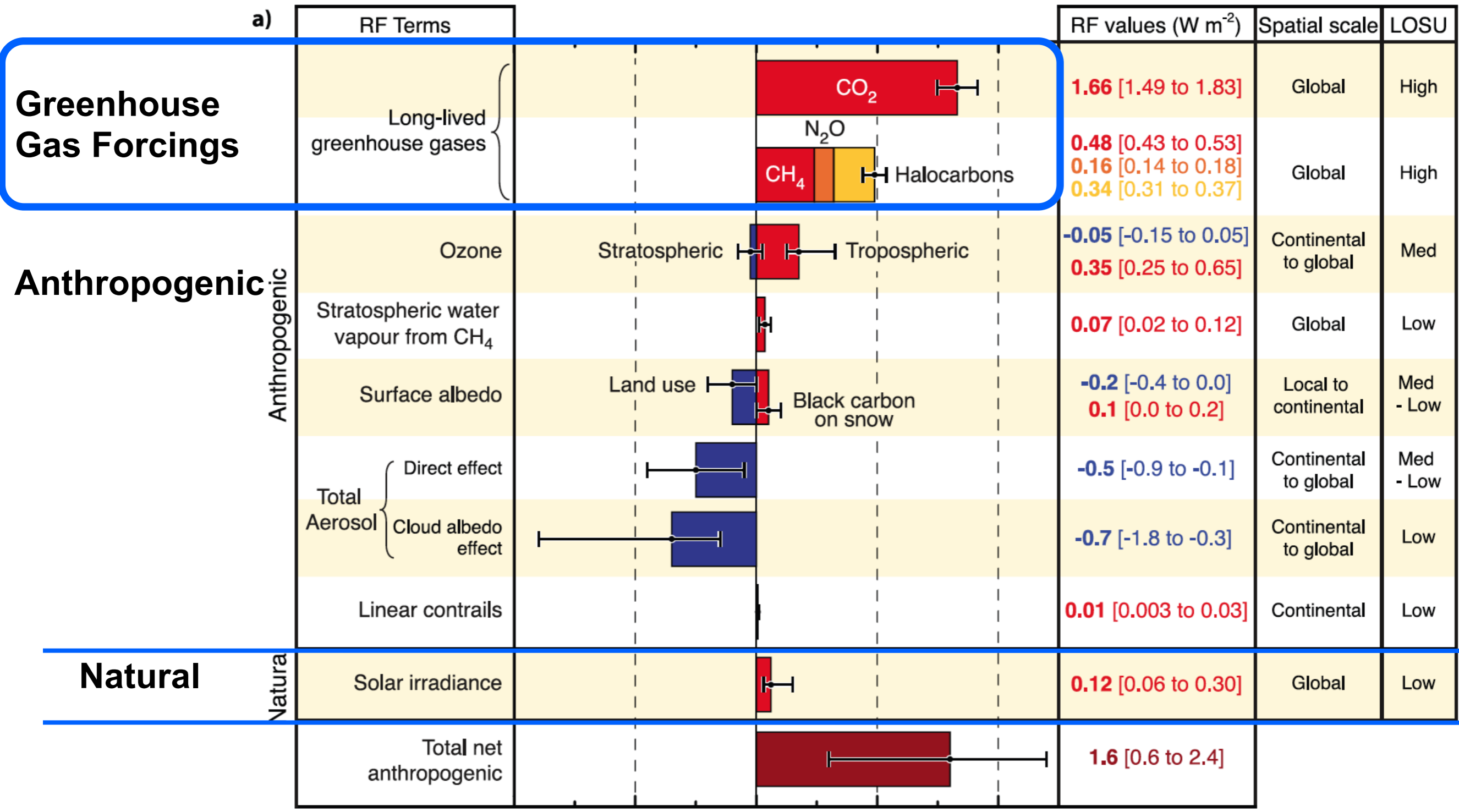
Historical Climate Forcing Factors

GLOBAL MEAN RADIATIVE FORCINGS



Historical Climate Forcing Factors

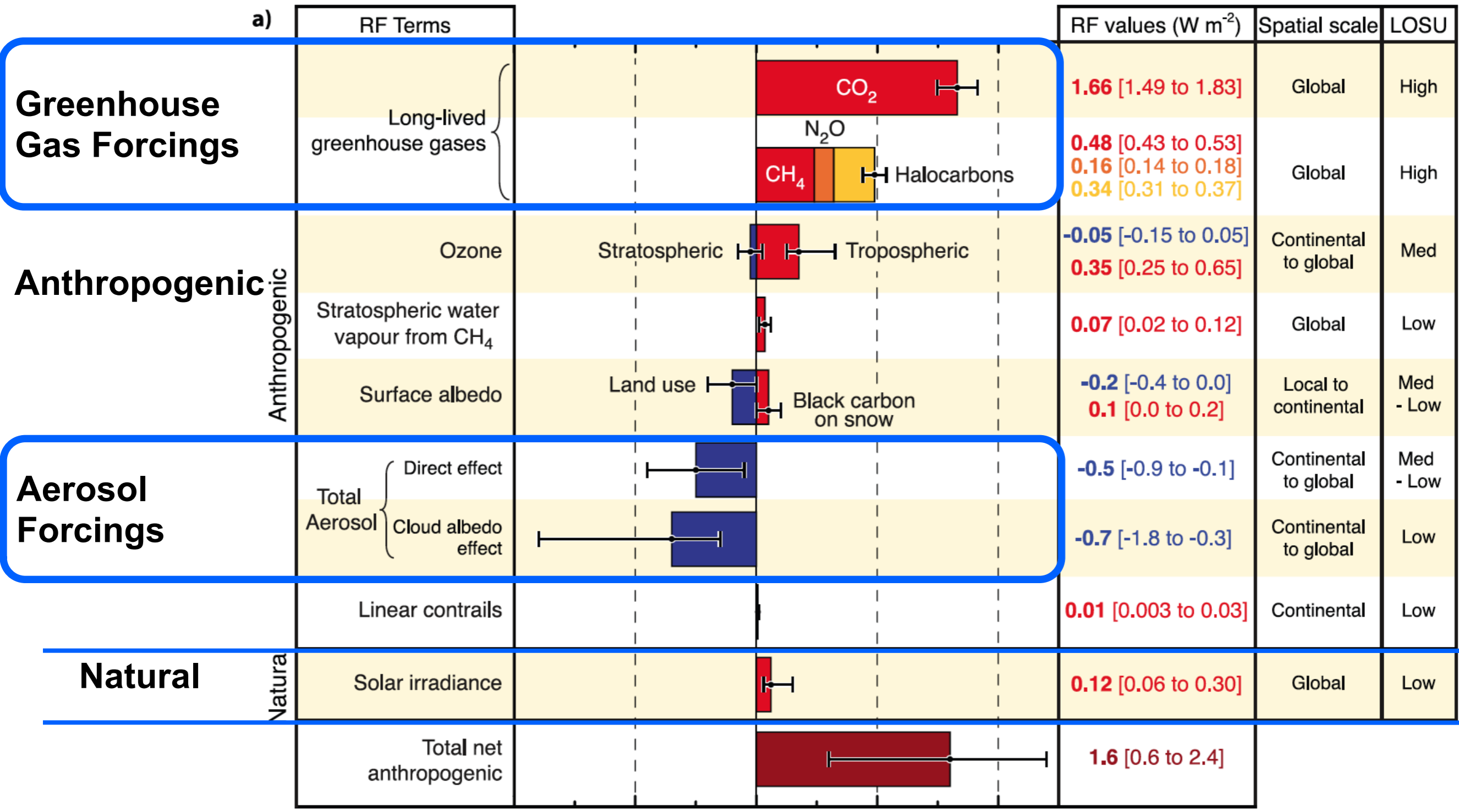
GLOBAL MEAN RADIATIVE FORCINGS



NB: Volcanic activity not included

Historical Climate Forcing Factors

GLOBAL MEAN RADIATIVE FORCINGS

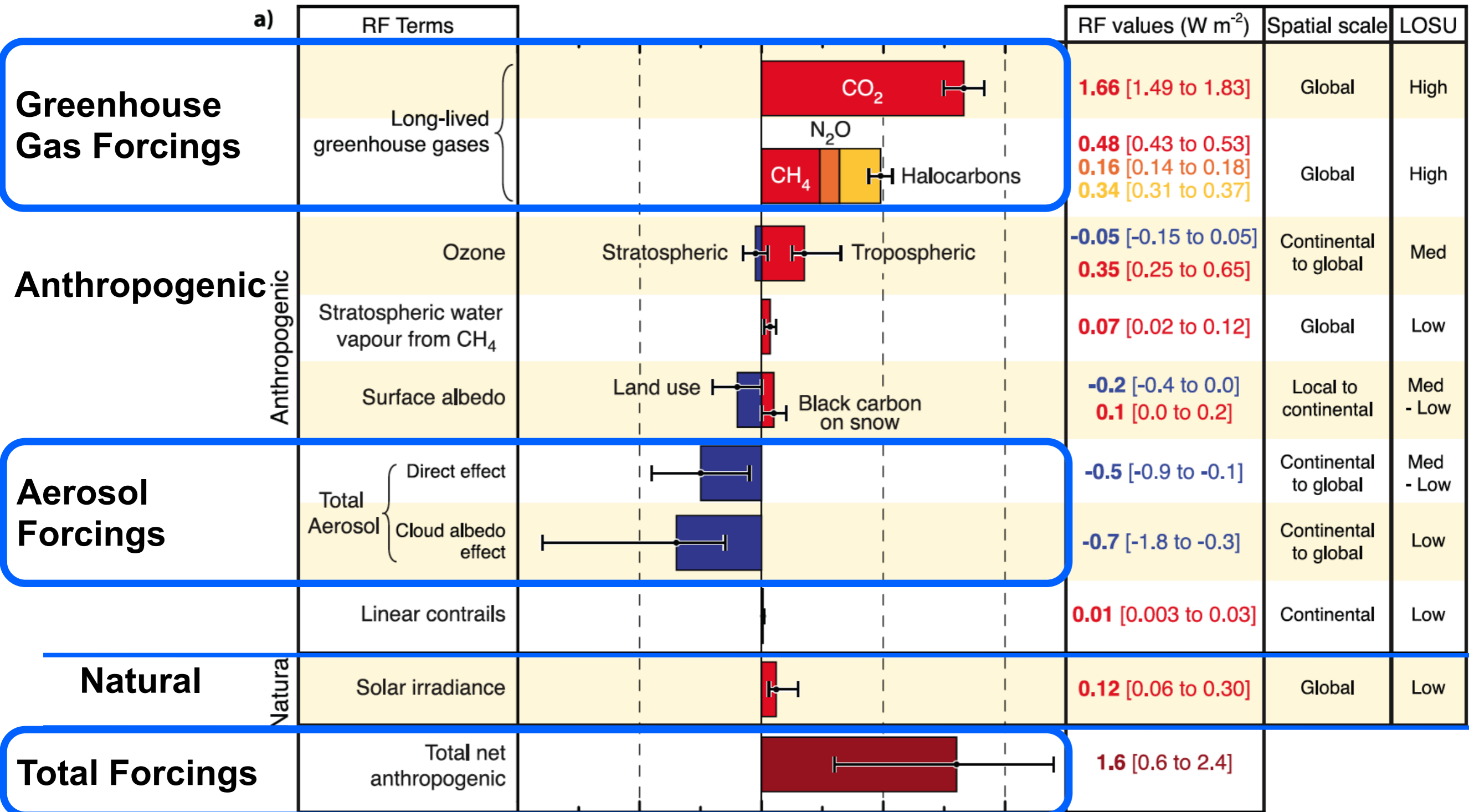


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Radiative Forcing ($W m^{-2}$)

Historical Climate Forcing Factors

GLOBAL MEAN RADIATIVE FORCINGS

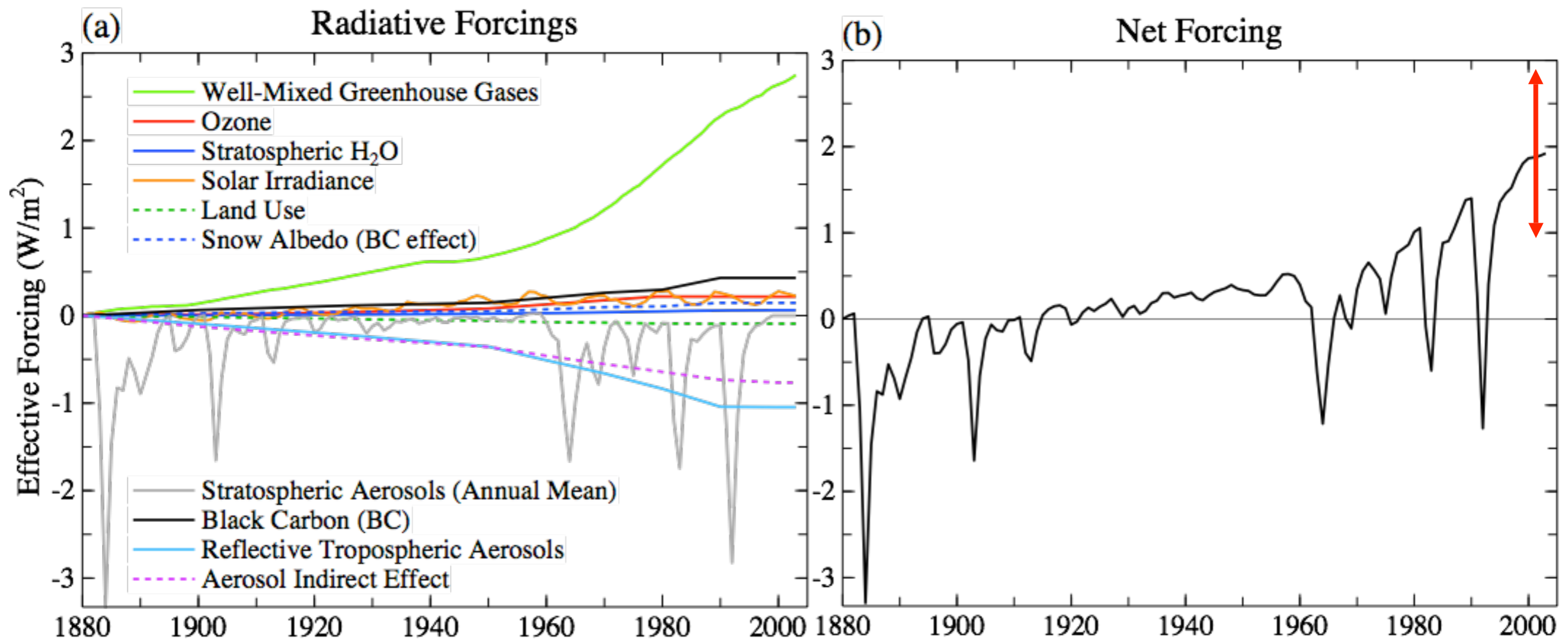


NB: Volcanic activity not included

Radiative Forcing ($W m^{-2}$)

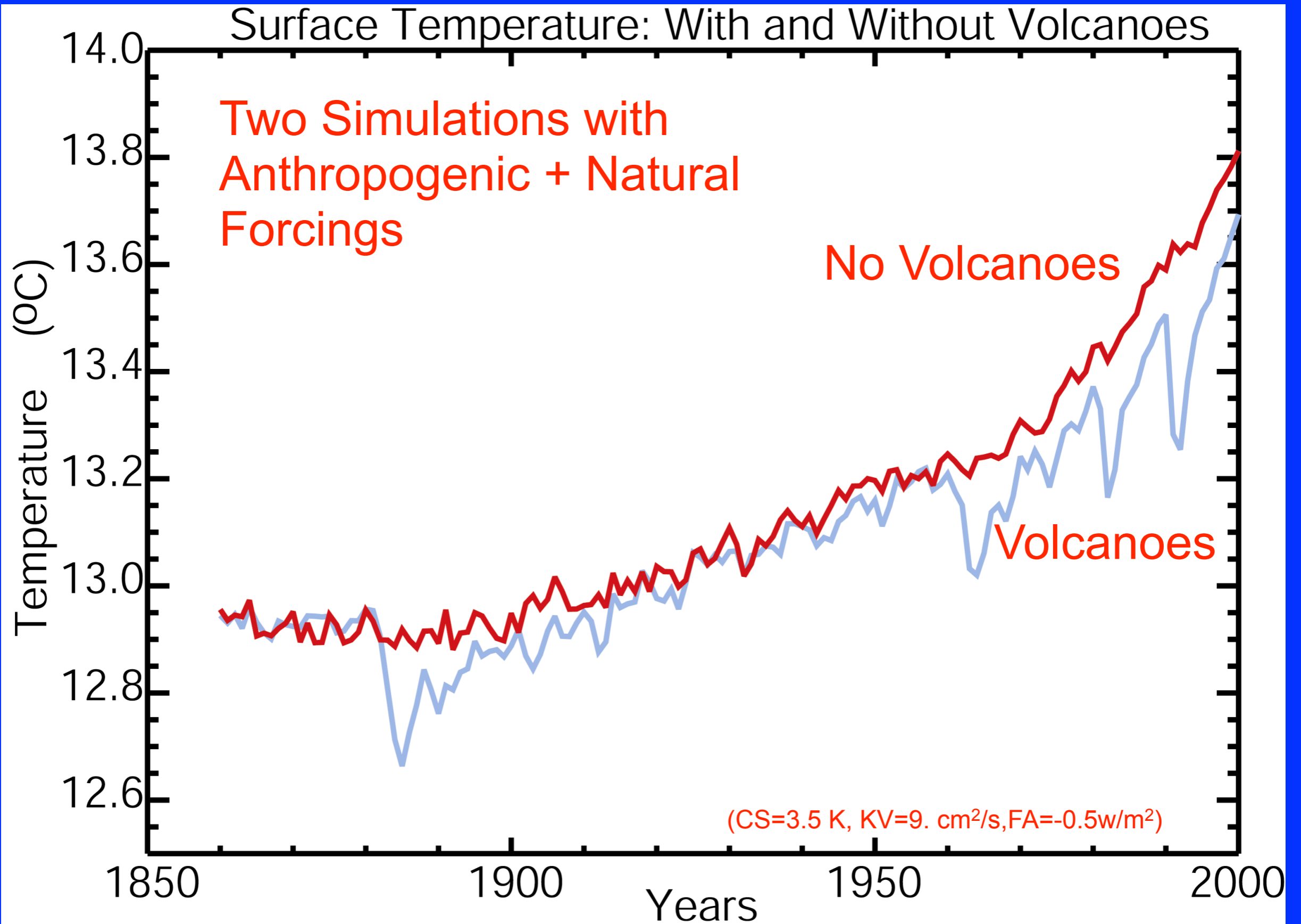
Historical Climate Forcing Factors

Estimated Time-series of Radiative Forcings
Total Uncertainty = +/- 1 W/m²



From NASA GISS: <http://data.giss.nasa.gov/modelforce/>

Major concern is getting the forcings right.

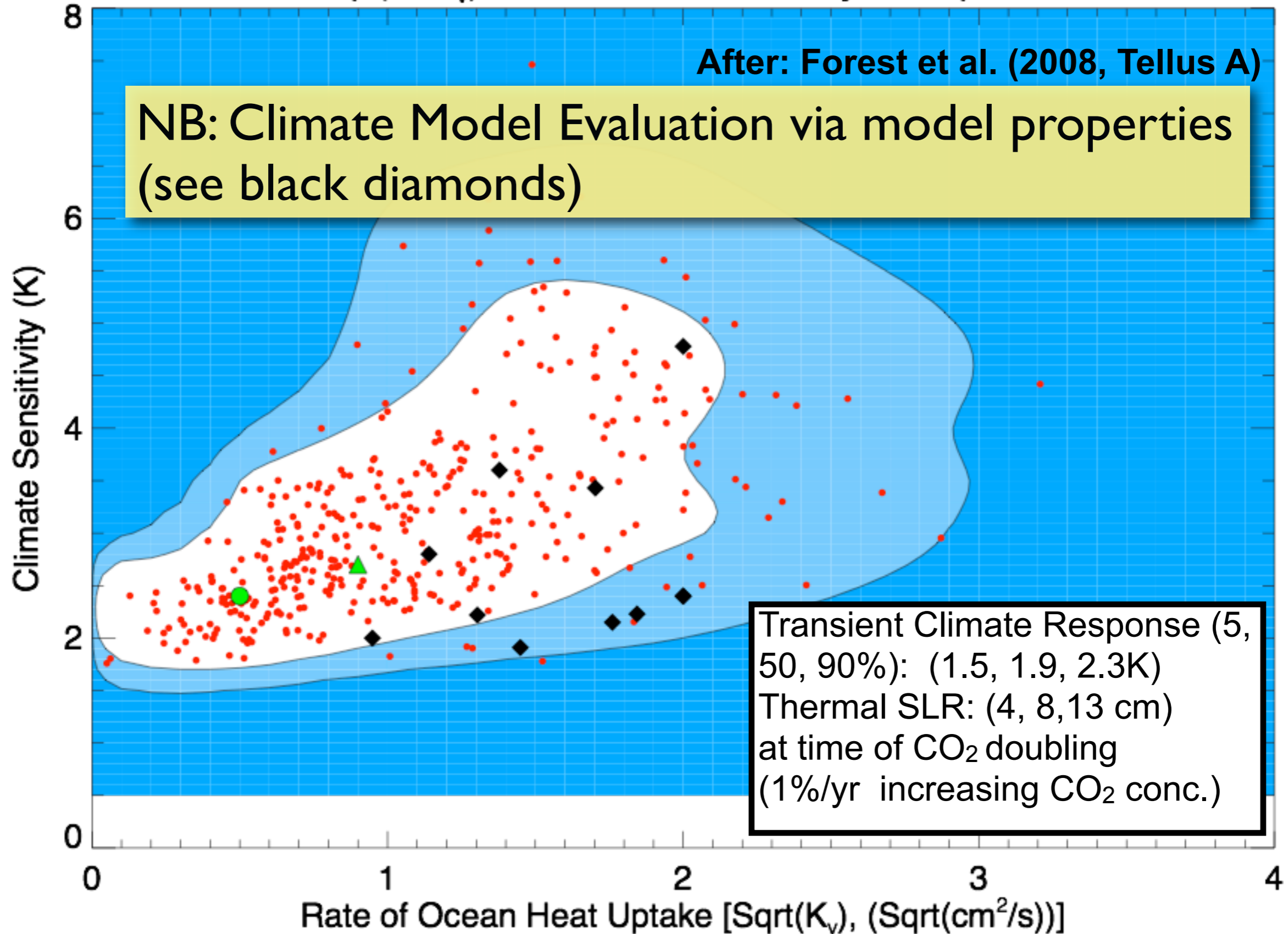


Latin Hypercube Sample (n=250)

$p(S, K_v)$: IGSM2 Uncertainty Sample

After: Forest et al. (2008, Tellus A)

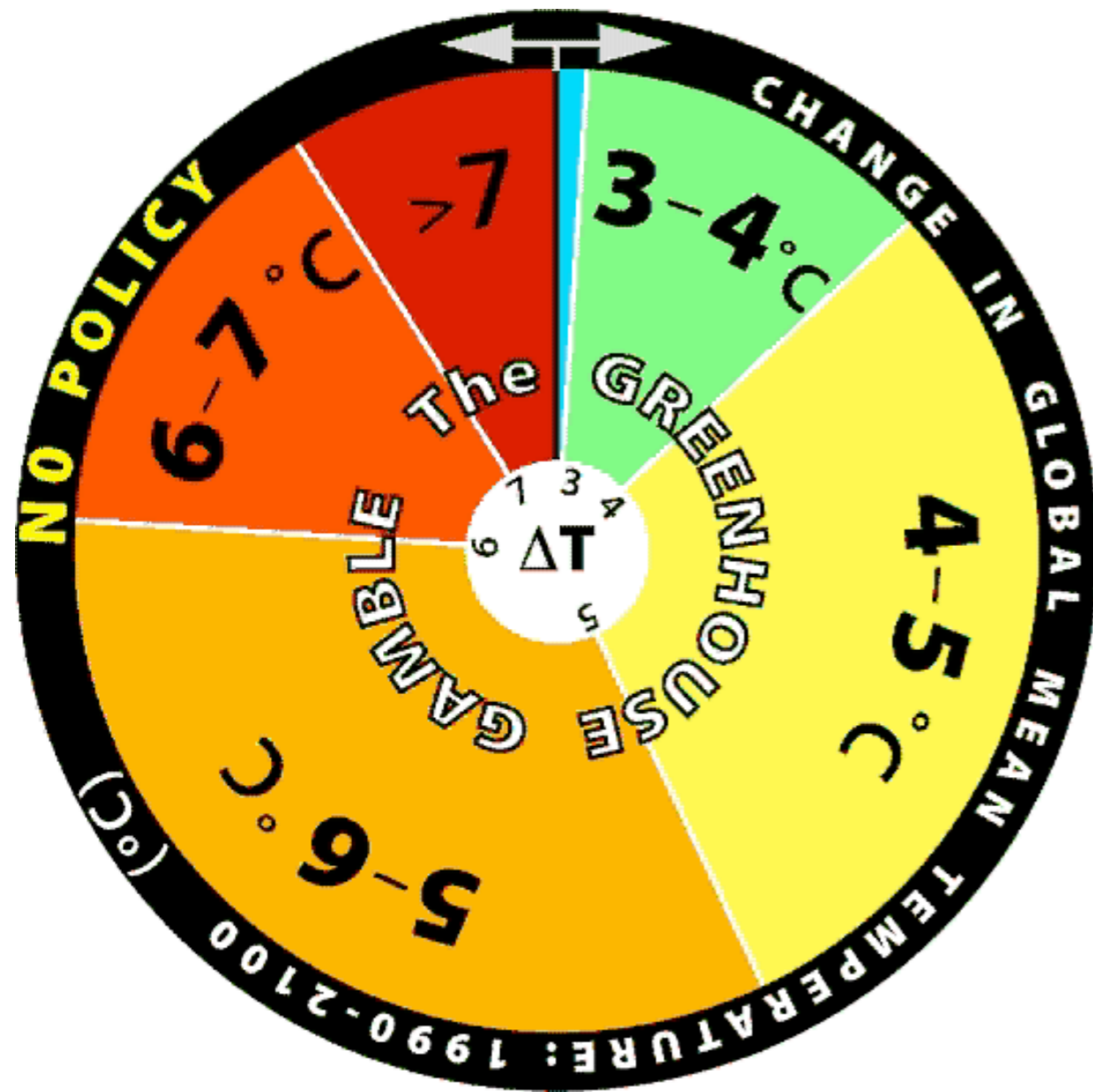
NB: Climate Model Evaluation via model properties
(see black diamonds)



Included Uncertainties

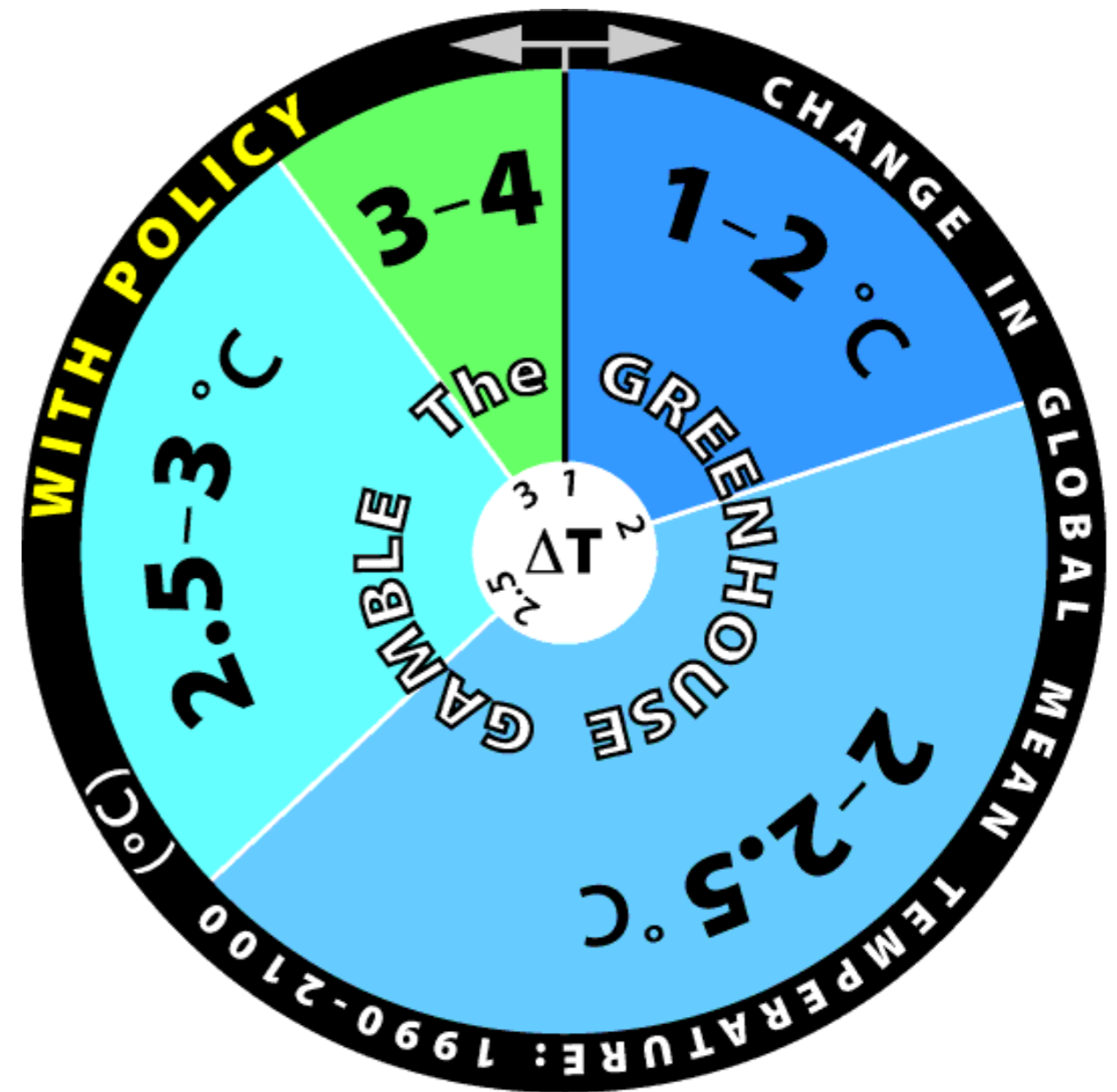
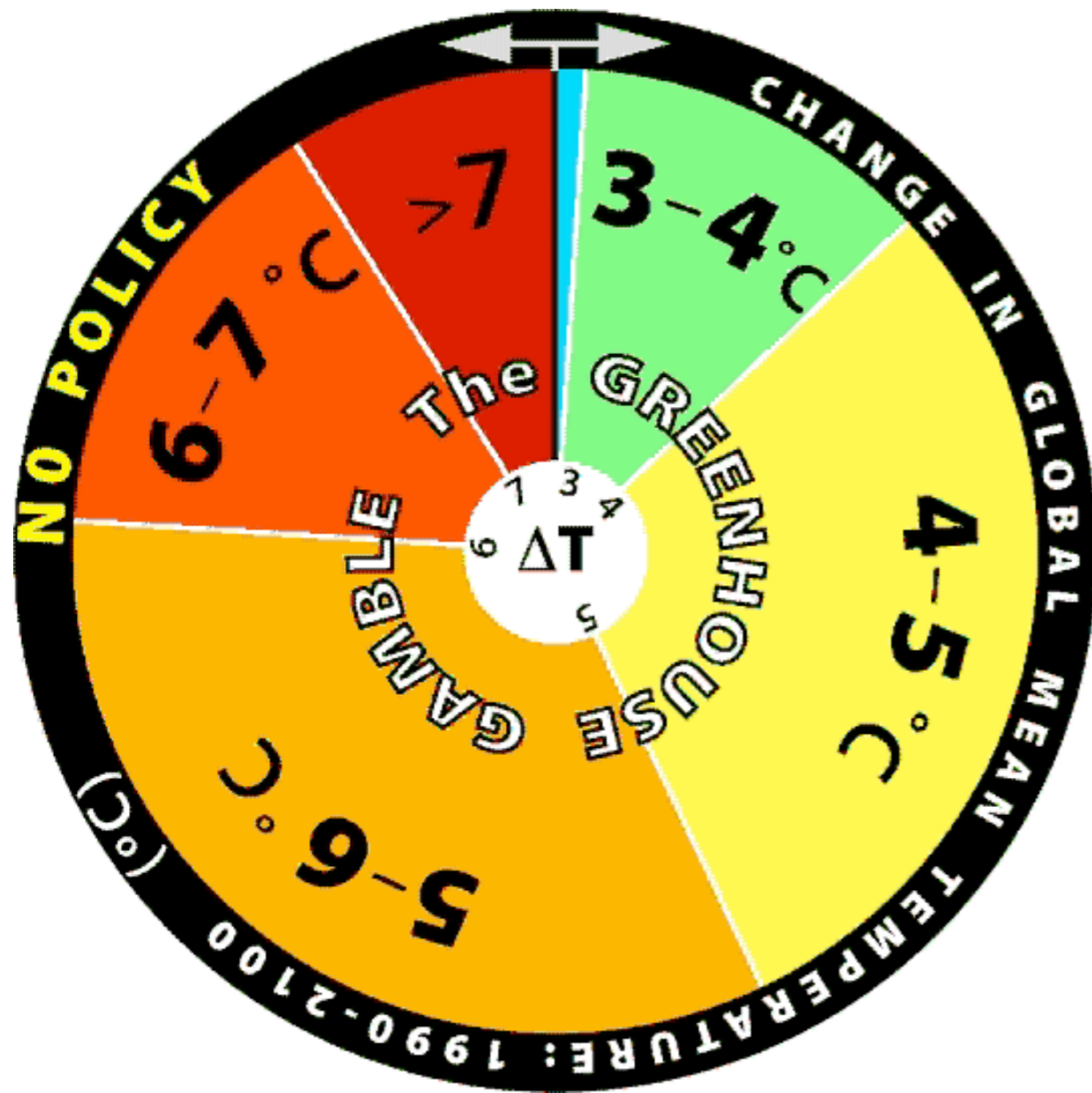
- **Emissions Uncertainty from MIT EPPA4**
 - Population: 6-13 billion, Energy Resources, Efficiency/Technology
- **Climate System Response**
(Calibrated in Forest et al. 2008)
 - Climate Sensitivity
 - Rate of Heat uptake by Deep Ocean
 - Radiative Forcing Strength of Aerosols
- **Carbon Cycle Uncertainty:**
 - CO₂ Fertilization Effect on Ecosystem
 - Rate of Carbon Uptake by Deep-Ocean
- **Trends in Precip. Freq. on CH₄ + N₂O**
(Statistics scaled using by AR4 model trends)

The Greenhouse Gamble



<http://globalchange.mit.edu/resources/gamble/>

The Greenhouse Gamble



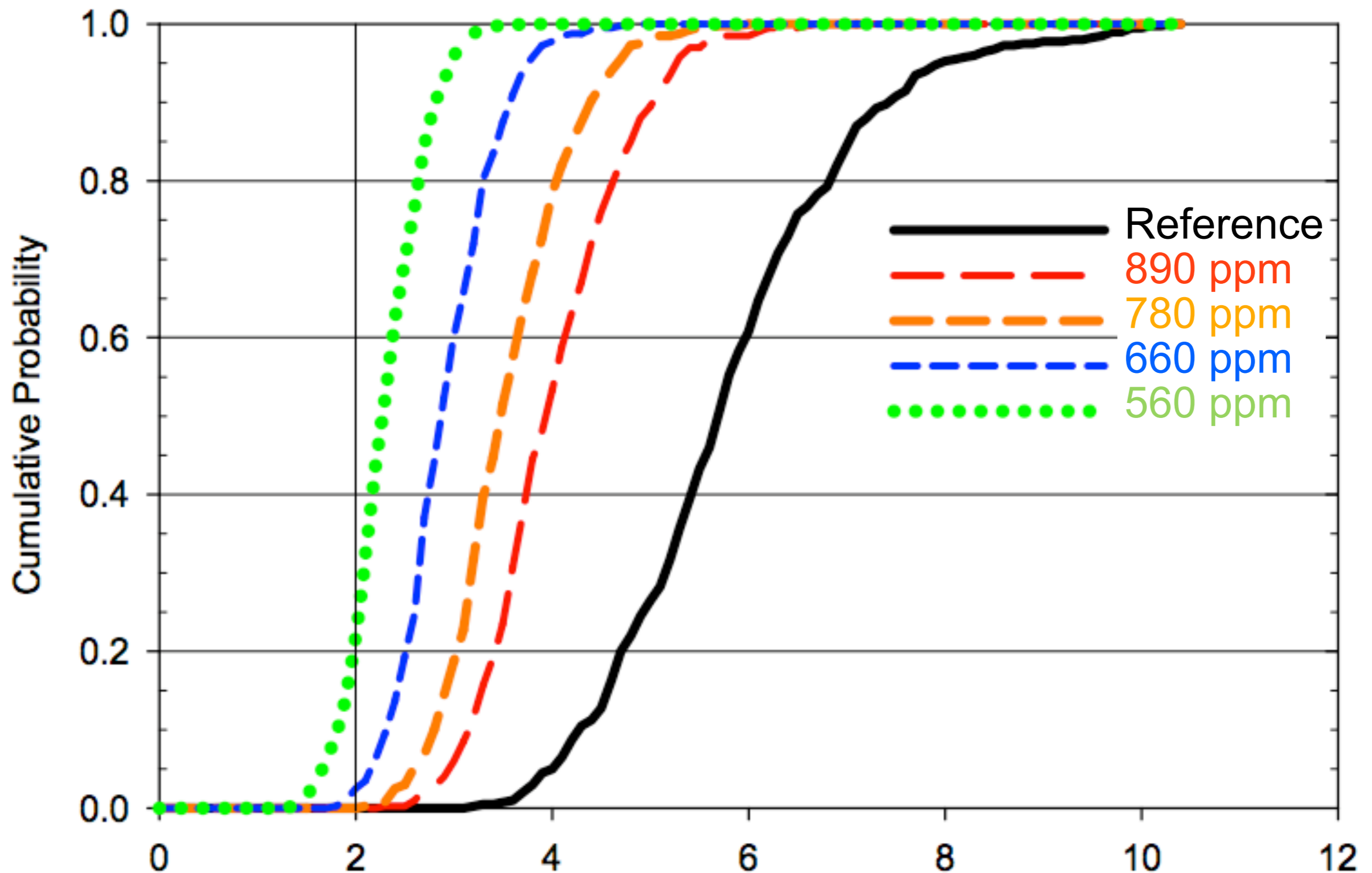
<http://globalchange.mit.edu/resources/gamble/>

Risks of Global Mean Temperature Increase 1990-2100

	$\Delta T > 2^{\circ}\text{C}$	$\Delta T > 4^{\circ}\text{C}$	$\Delta T > 6^{\circ}\text{C}$
No Policy	400 in 400	17 in 20	1 in 4
Stabilize at 750	400 in 400	1 in 4	1 in 400
Stabilize at 650	97 in 100	7 in 100	<1 in 400
Stabilize at 550	8 in 10	1 in 400	<1 in 400
Stabilize at 450	1 in 4	<1 in 400	<1 in 400

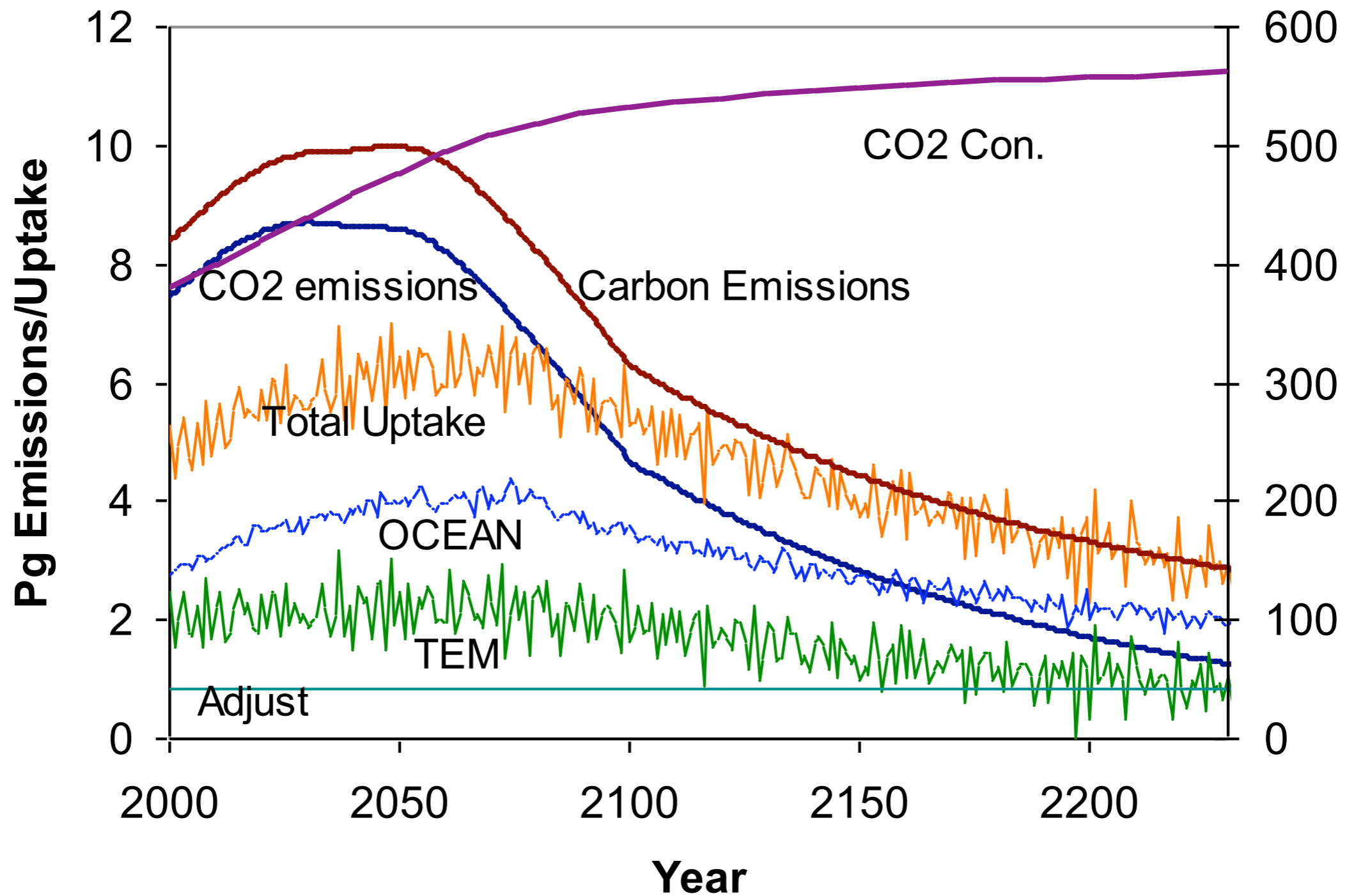
Sources: Sokolov et al. (2009, J. Climate); Webster et al. (2009, MIT JP Report 180)

Cumulative Probability of T_s in 2100 above Pre-industrial Temperature



Stabilization Scenario Example

Uptake and Emissions



Extra Slides

From: Sokolov et al. (2003, J. Climate)

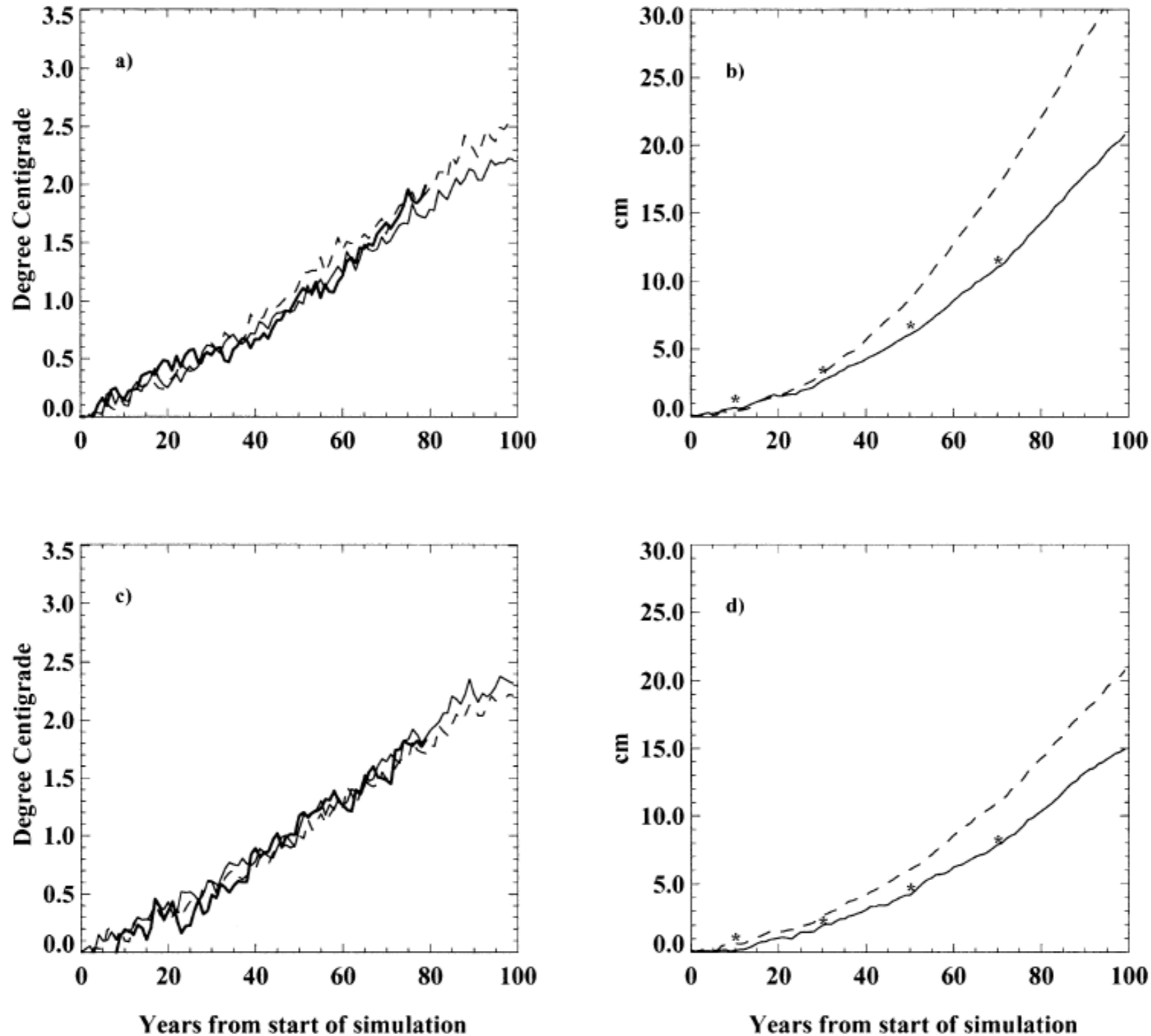
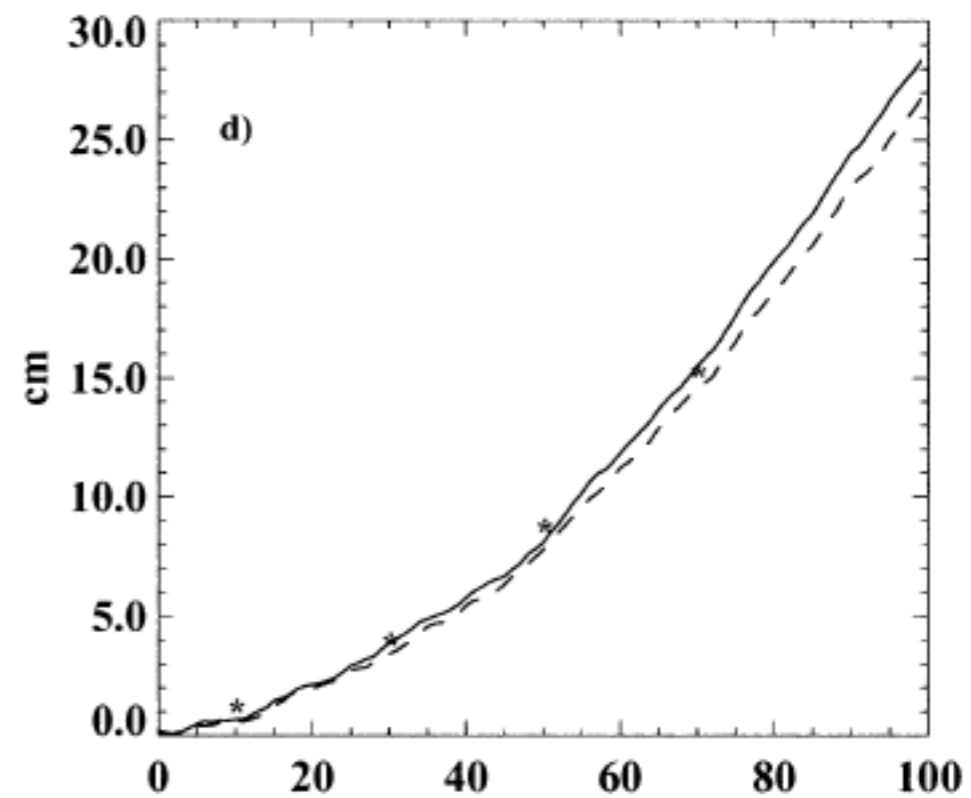
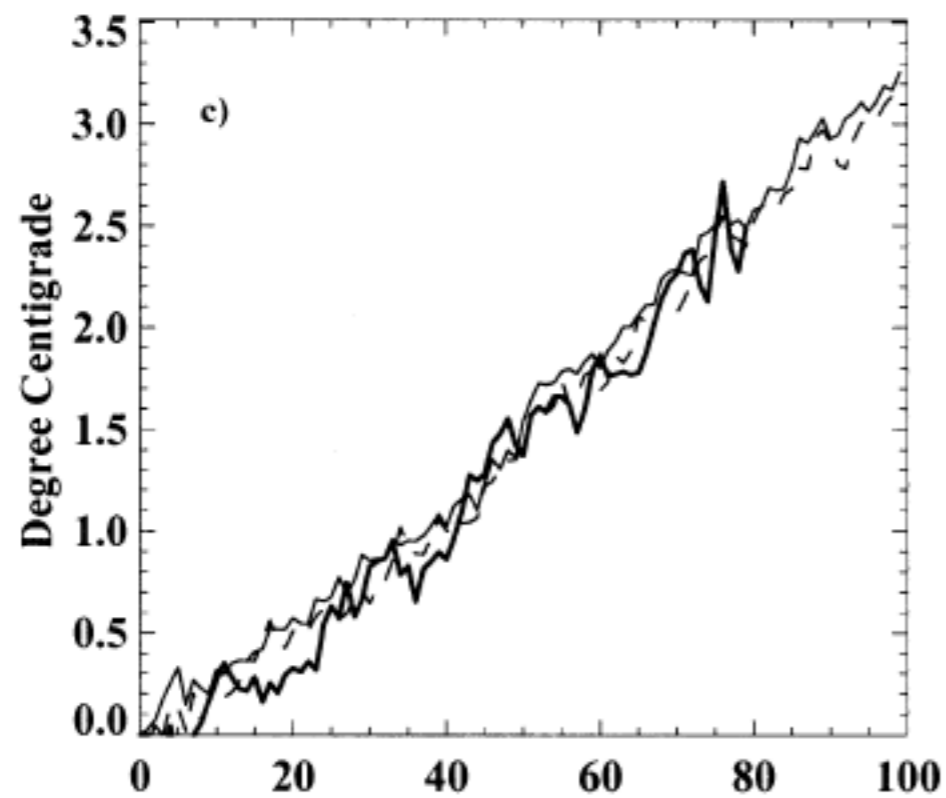
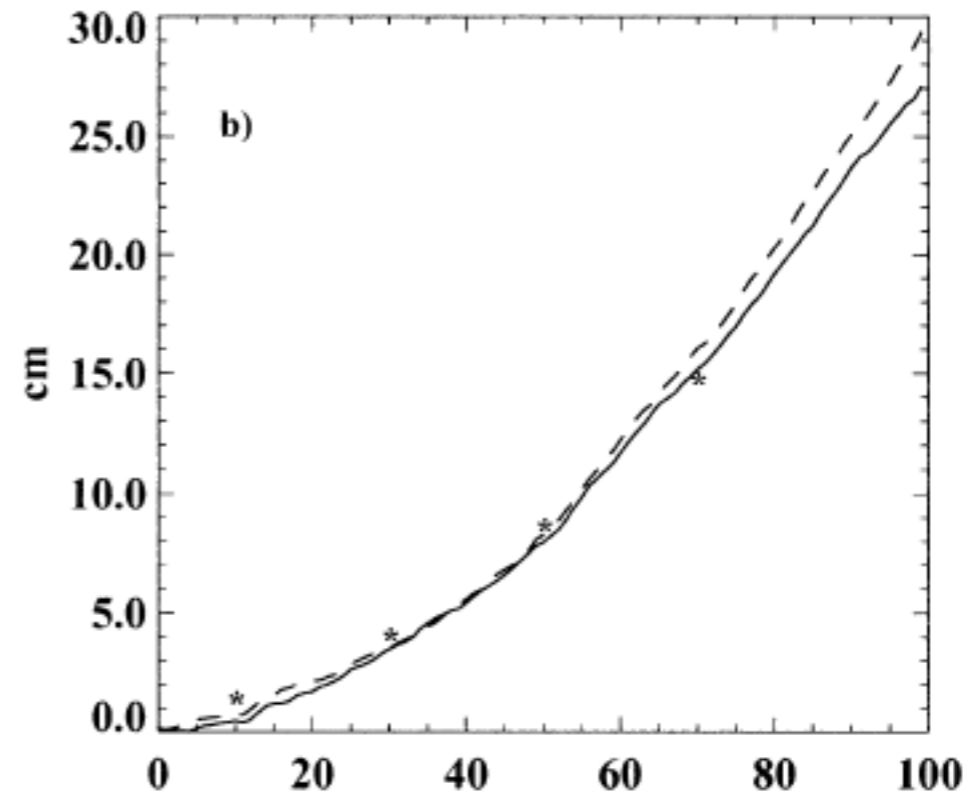
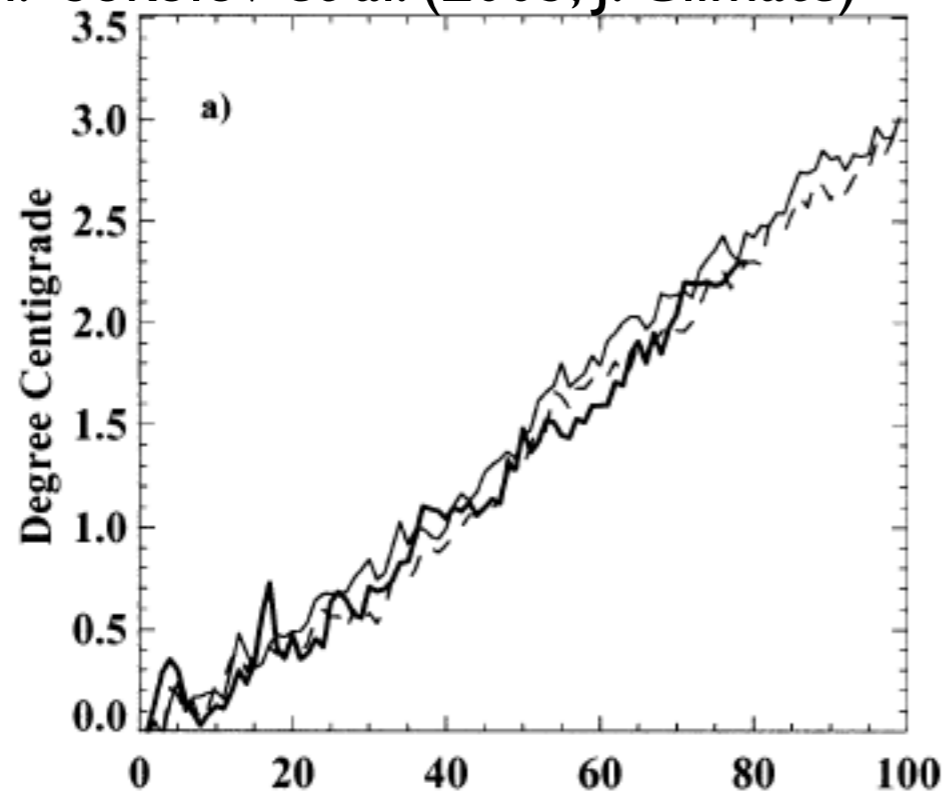


FIG. 2. Changes of annual mean global mean surface air temperature and sea level (thermal expansion) in simulations with the (a), (b) MRI1 and (c), (d) ECHAM3/LSG AOGCMs and in simulations with the versions of the MIT 2D climate model with effective (thin solid lines) and equilibrium (dashed lines) climate sensitivities. Data from CMIP2 simulations with AOGCMs are shown by thick solid line (SAT) and by asterisks (sea level). Unfortunately, while changes in SAT from these simulations are available on an annual basis, sea level rise due to thermal expansion of the ocean is not. The data required to calculate thermal expansion were saved as a 20-yr mean for four consecutive segments of the simulations. In this study we used data on sea level rise for these four periods provided by S. Raper (Raper et al. 2002).

From: Sokolov et al. (2003, J. Climate)

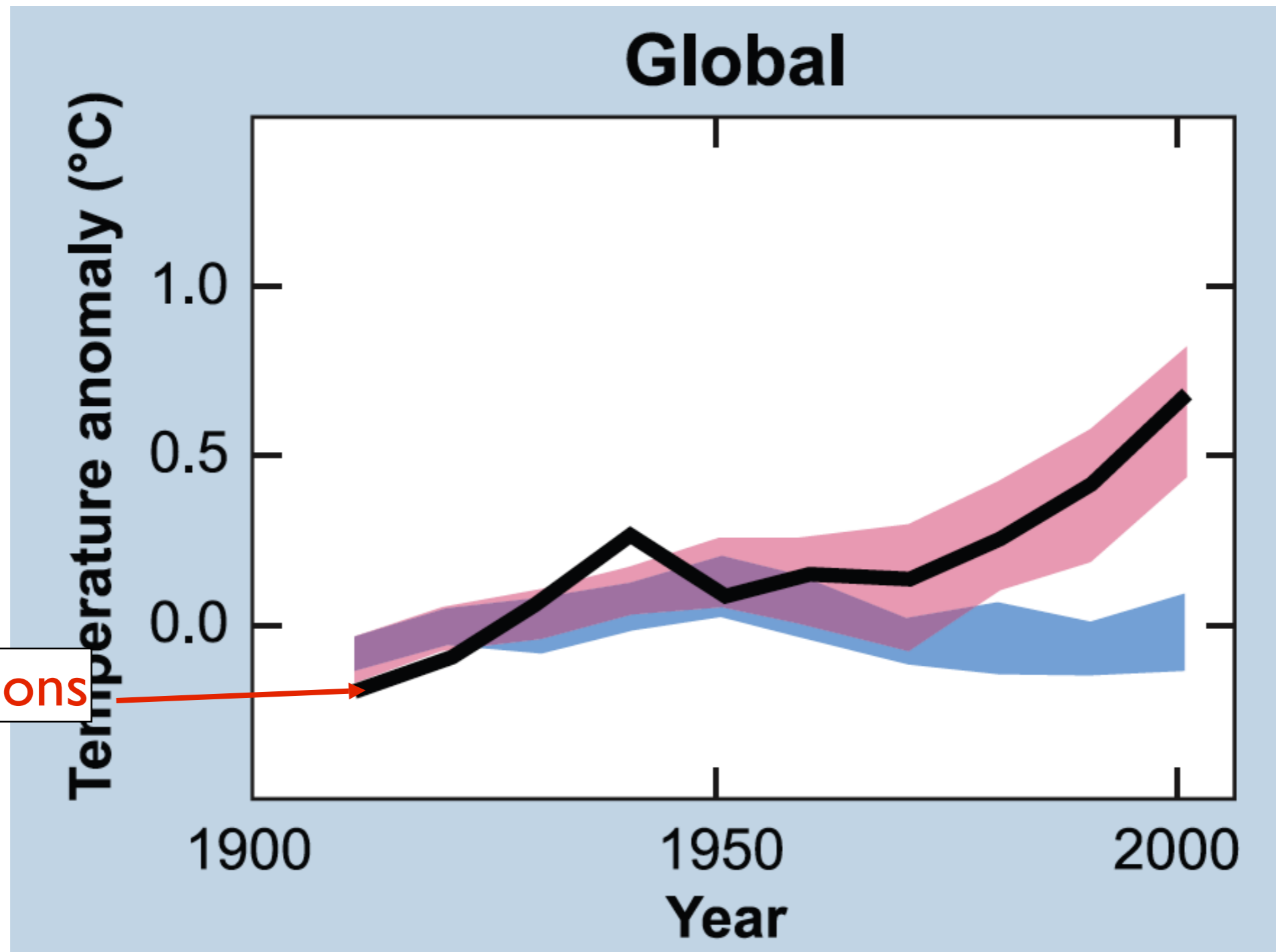


Years from start of simulation

Years from start of simulation

FIG. 3. As in Fig. 2 but for the (a), (b) CSIRO and (c), (d) GFDL R15 AOGCMs.

Climate Change Observations and Climate Model Hindcasts



Observations



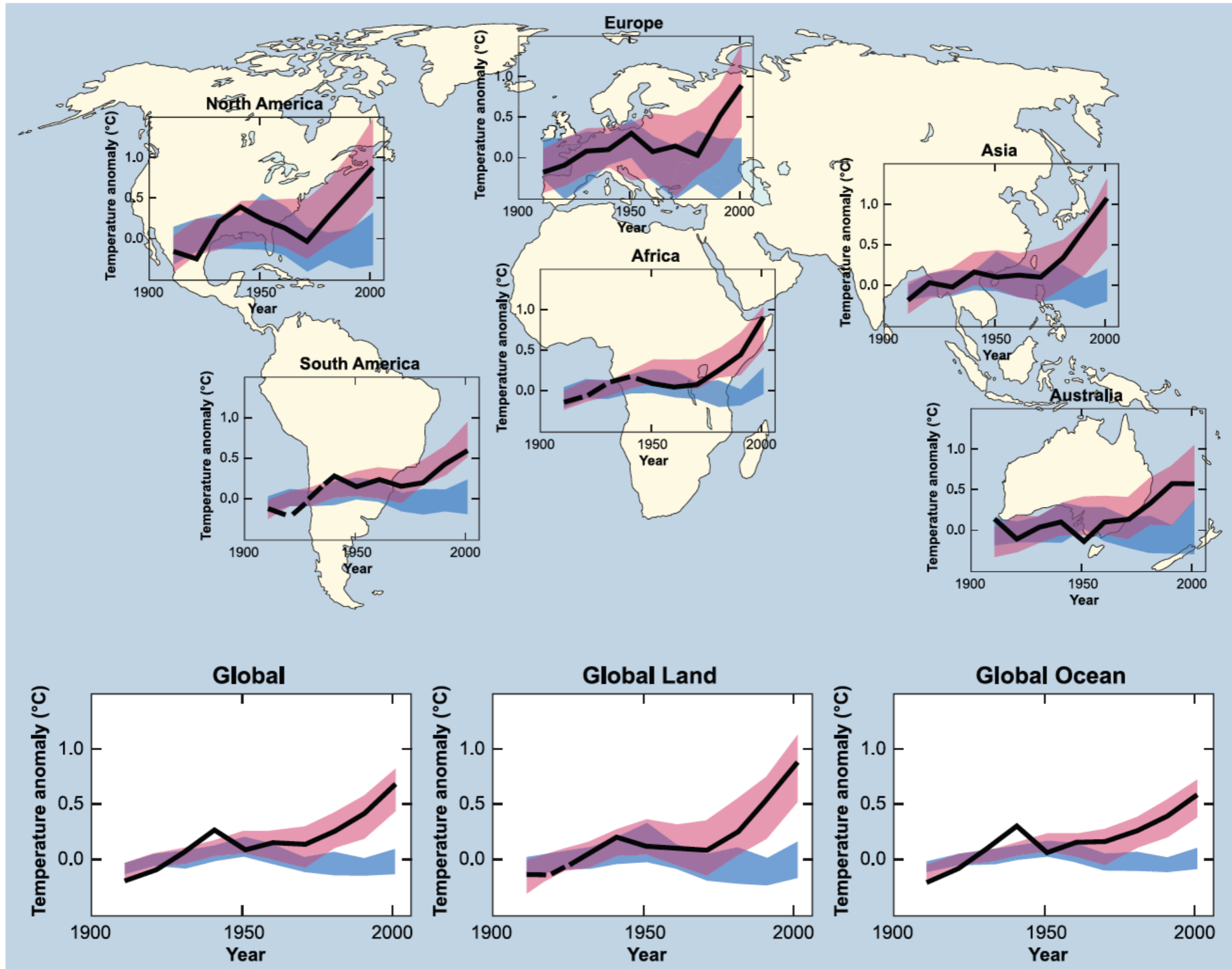
Models using only natural forcings (n~21)



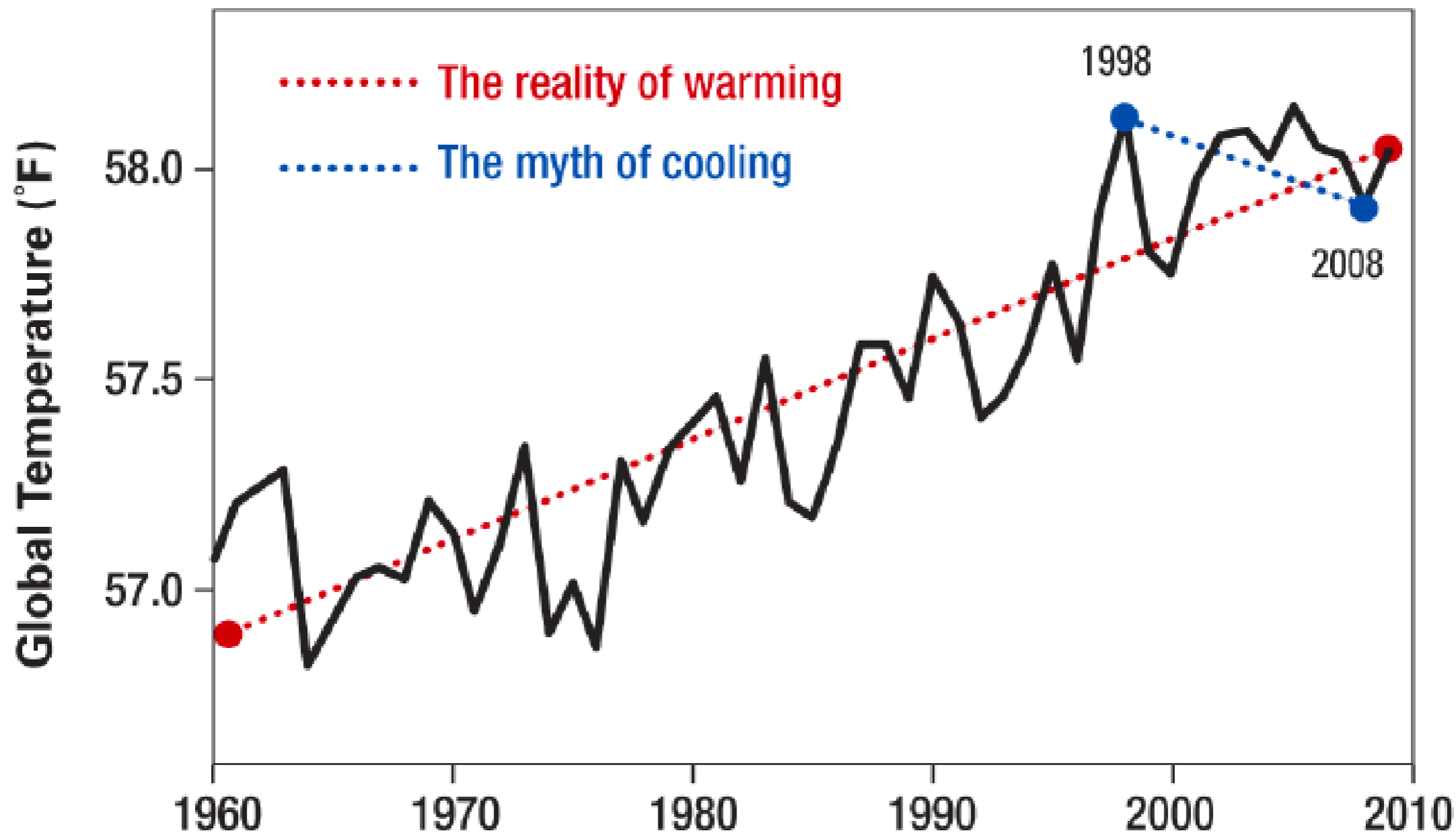
Models using natural and anthropogenic forcings

Climate Change Observations and Climate Model Hindcasts

GLOBAL AND CONTINENTAL TEMPERATURE CHANGE



Surface Temperature Records



Global Surface Temperature Over the Last Fifty Years

The year 1998 was particularly warm and has been used to falsely claim that the following decade has seen little change or a cooling in temperature. Red shows the correct trend from 1960 through 2008, blue is an erroneous trend over ten years resulting from "cherry-picking" the start and finish dates. Source: NOAA/NCDC data¹, design idea K. Hayhoe.

Surface Temperature Records

Global Land–Ocean Temperature Index

