

Climate sensitivity in an uncertain world

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Science on Tap
October 2 2018



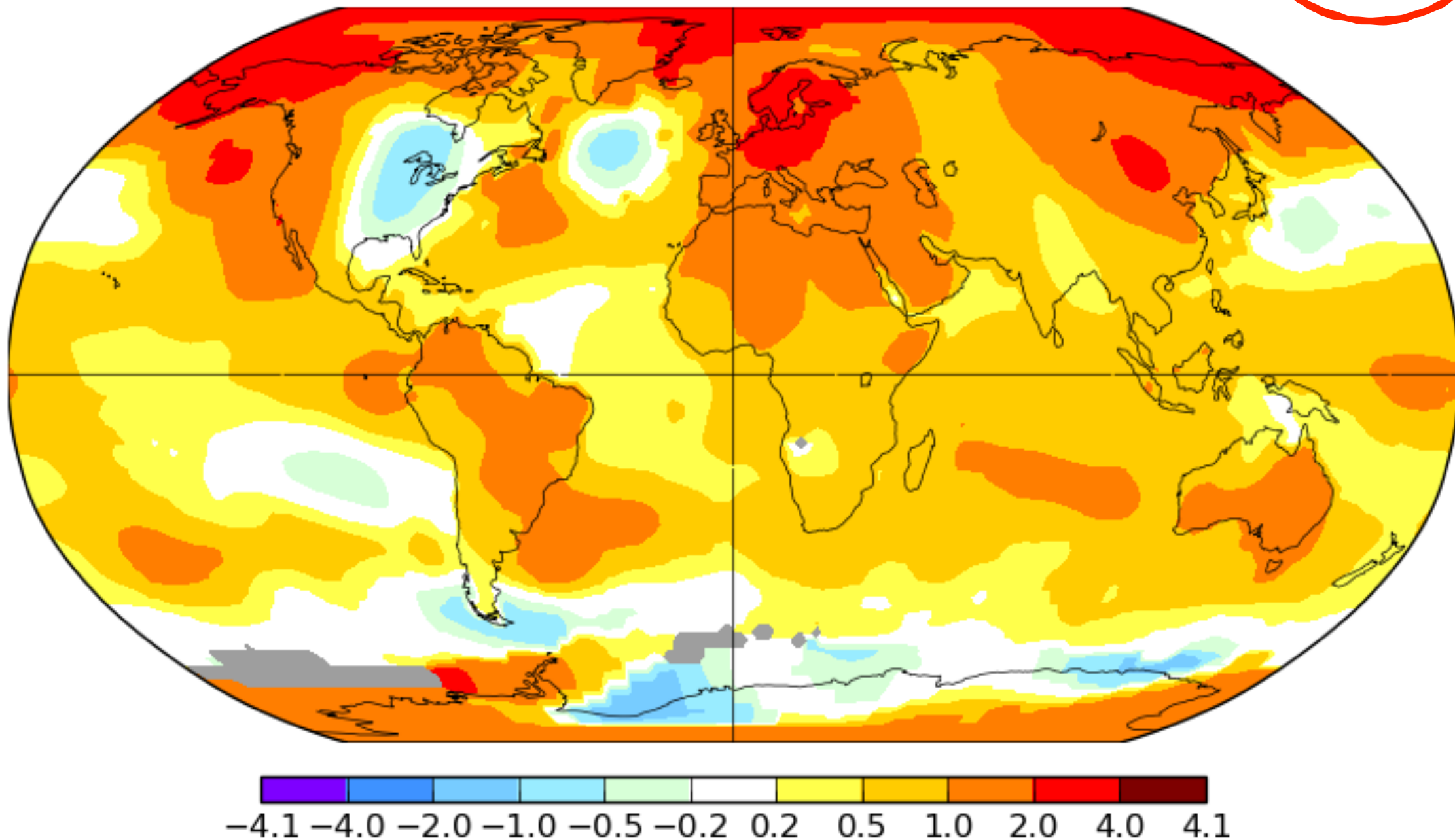
UNIVERSITY
AT ALBANY
State University of New York

2014 was the warmest year on record

Annual J-D 2014

L-OTI(°C) Anomaly vs 1951-1980

0.73



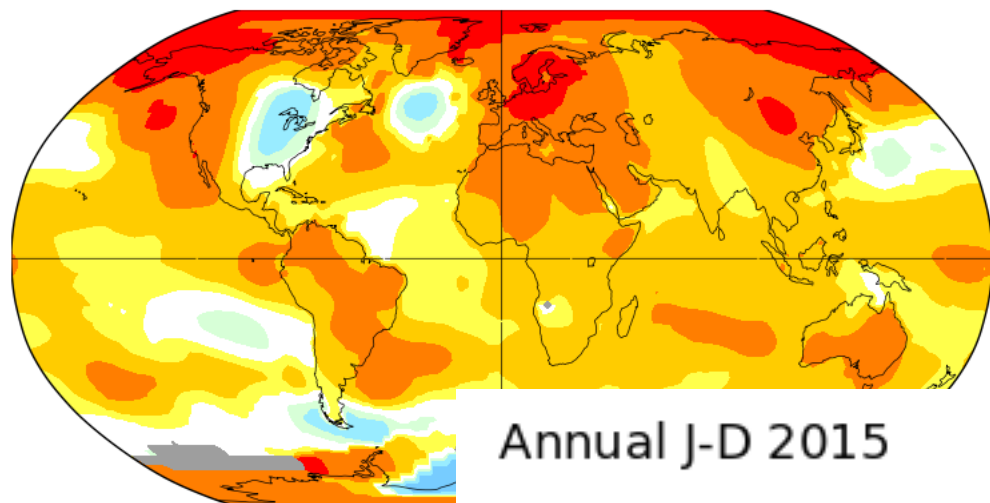
Though it did not feel that way here in the NE United States!

<https://data.giss.nasa.gov/gistemp/maps/>

Annual J-D 2014

L-OTI(°C) Anomaly vs 1951-1980

0.73

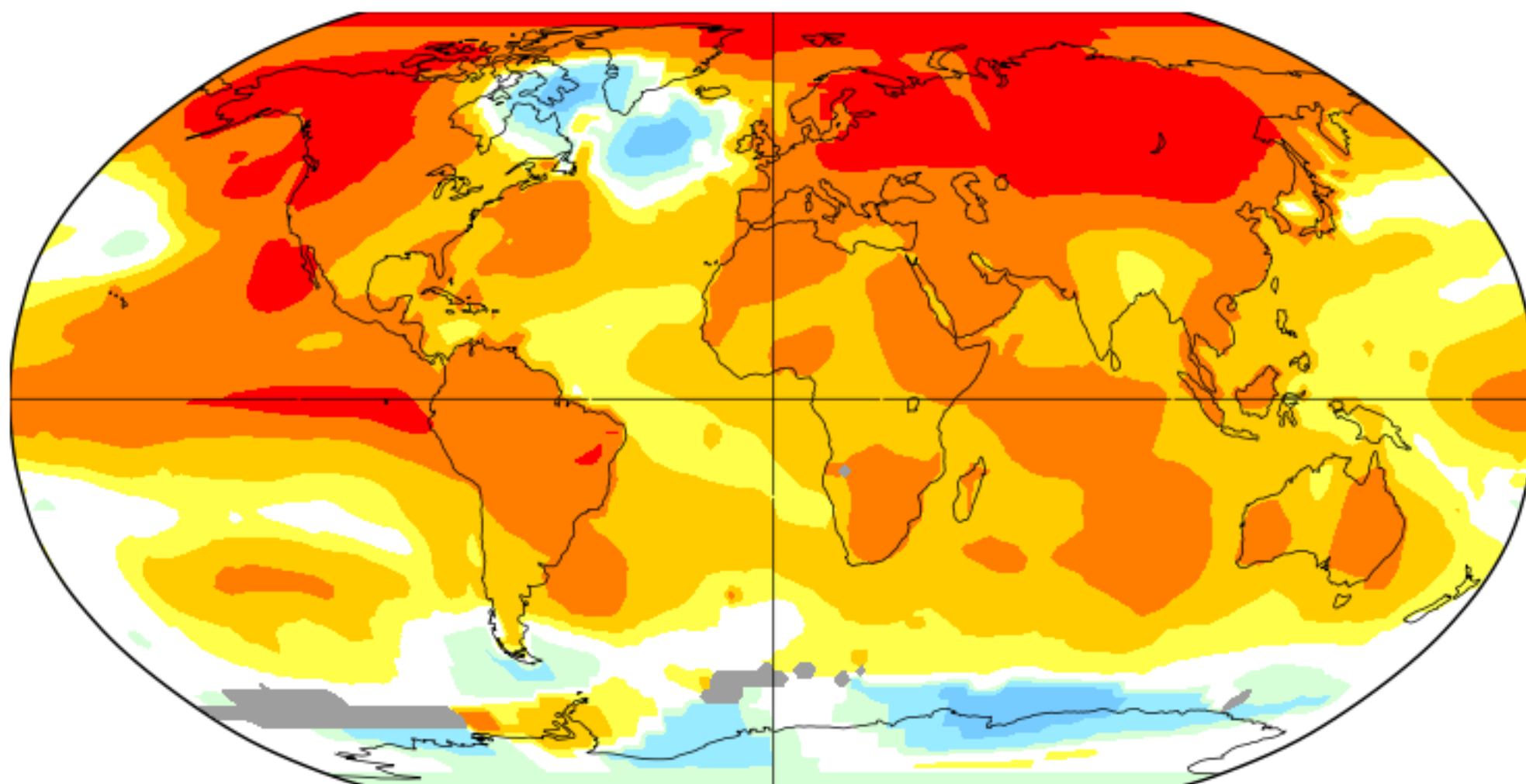


-4.1 -4.0 -2.0 -1.0 -0.5 -

Annual J-D 2015

L-OTI(°C) Anomaly vs 1951-1980

0.85



-4.1 -4.0 -2.0 -1.0 -0.5 -0.2 0.2 0.5 1.0 2.0 4.0 4.1

<https://data.giss.nasa.gov/gistemp/maps/>

Annual J-D 2014

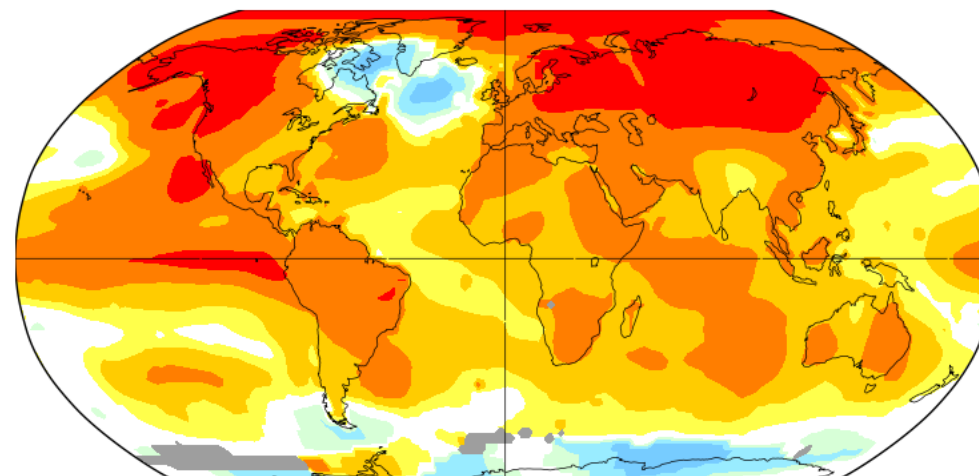
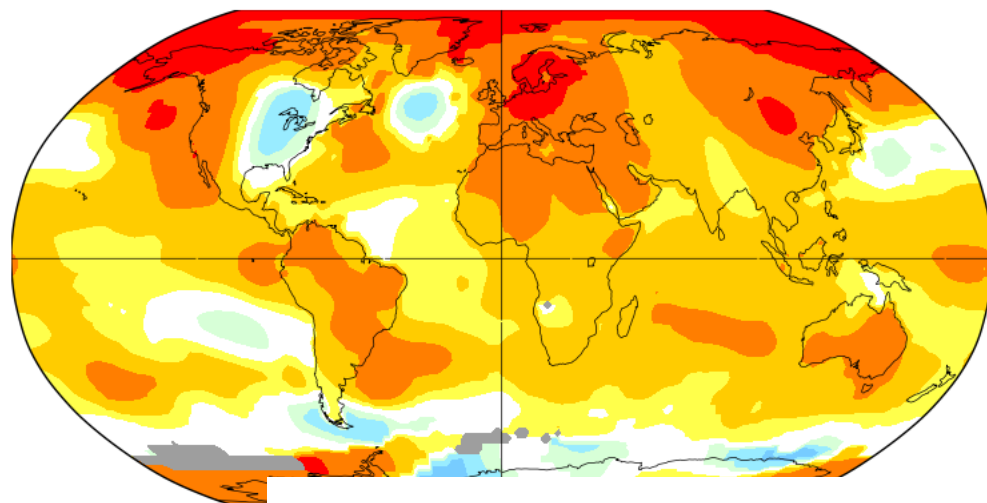
L-OTI(°C) Anomaly vs 1951-1980

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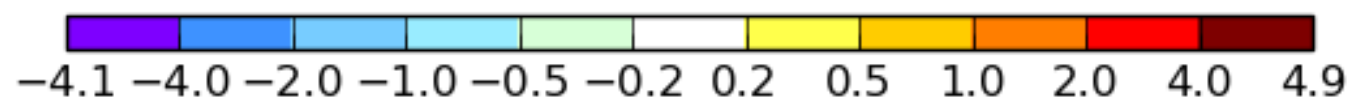
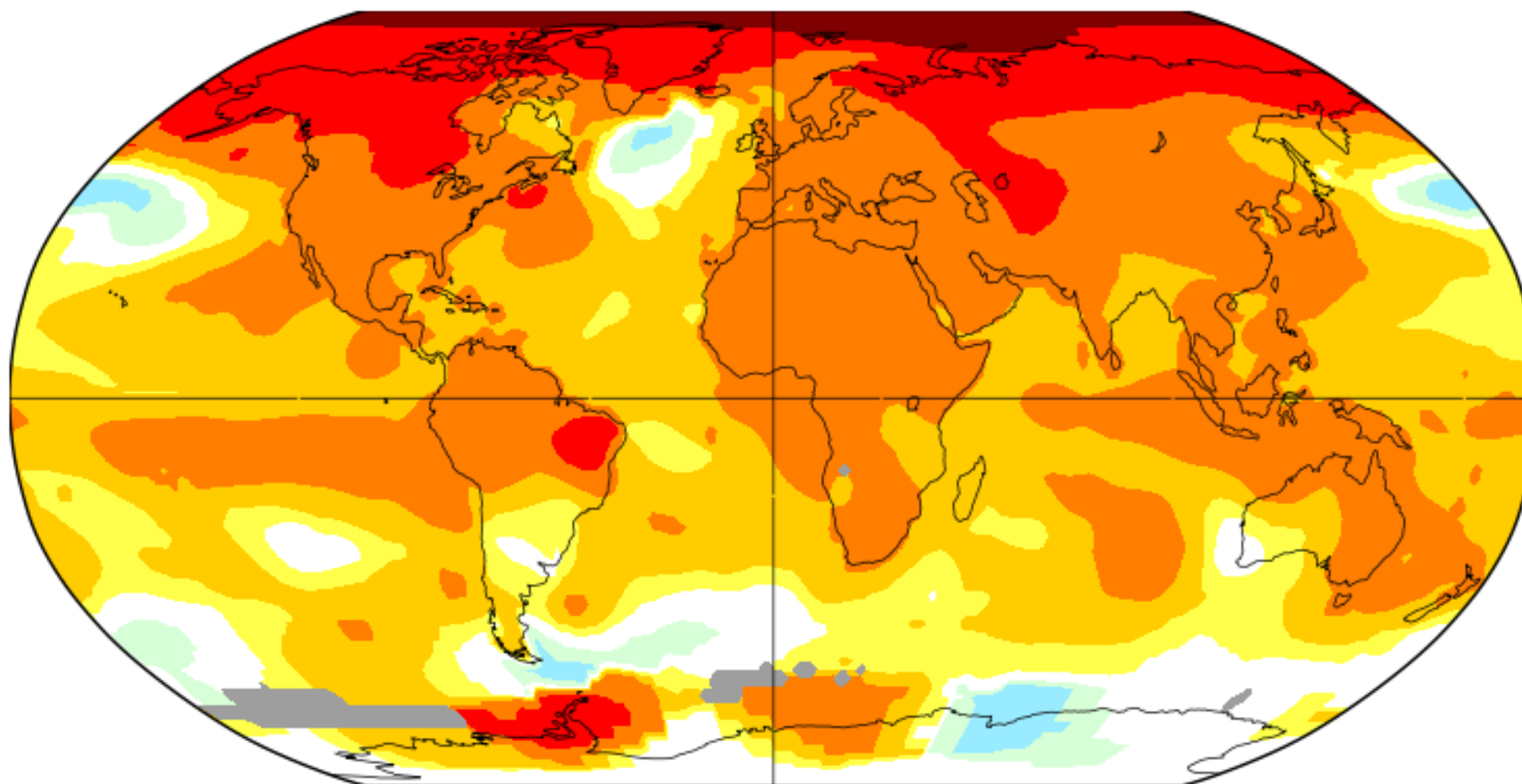


Annual J-D 2016

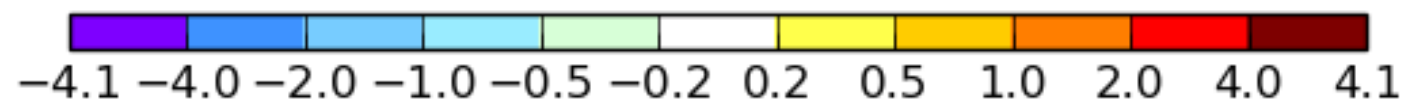
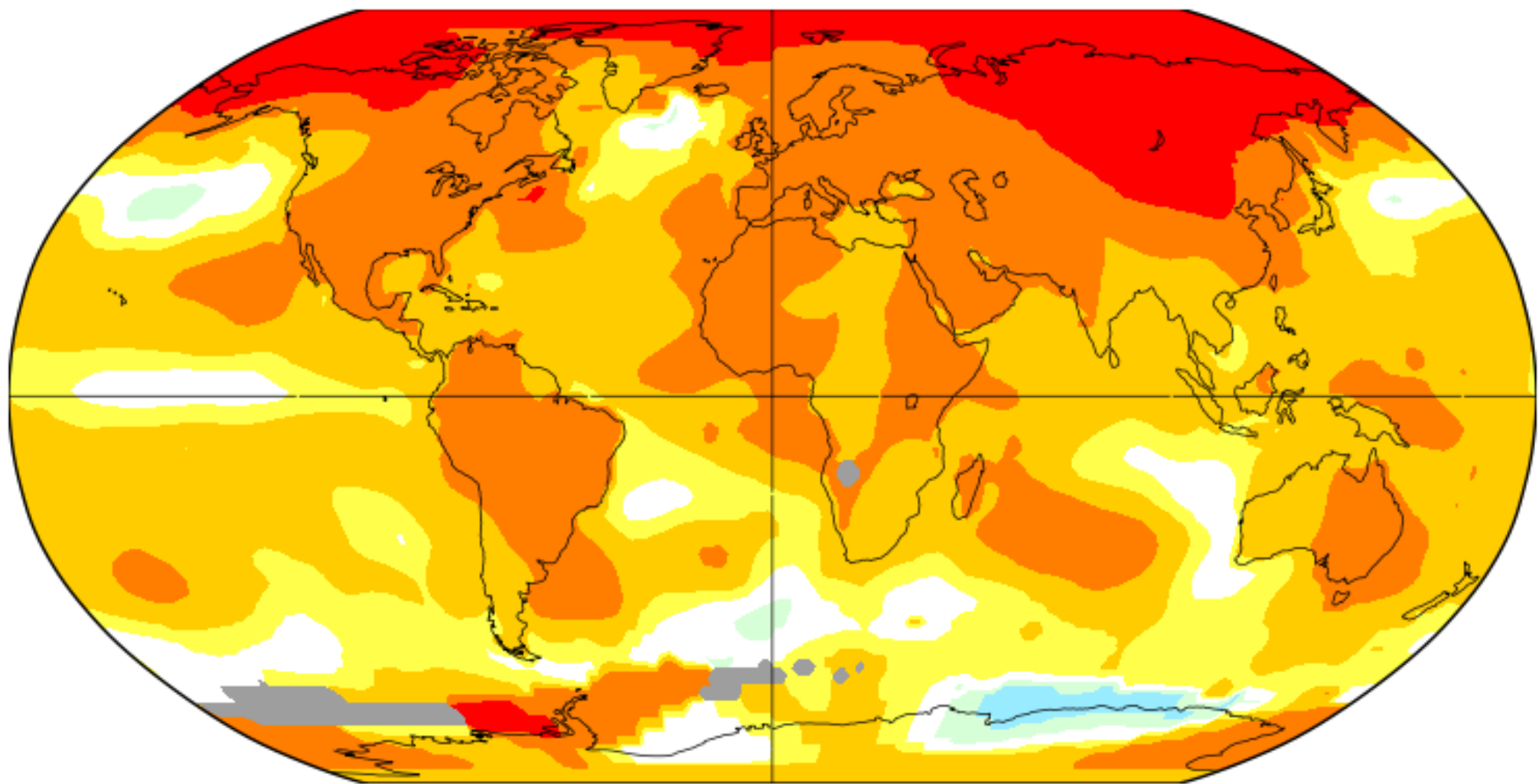
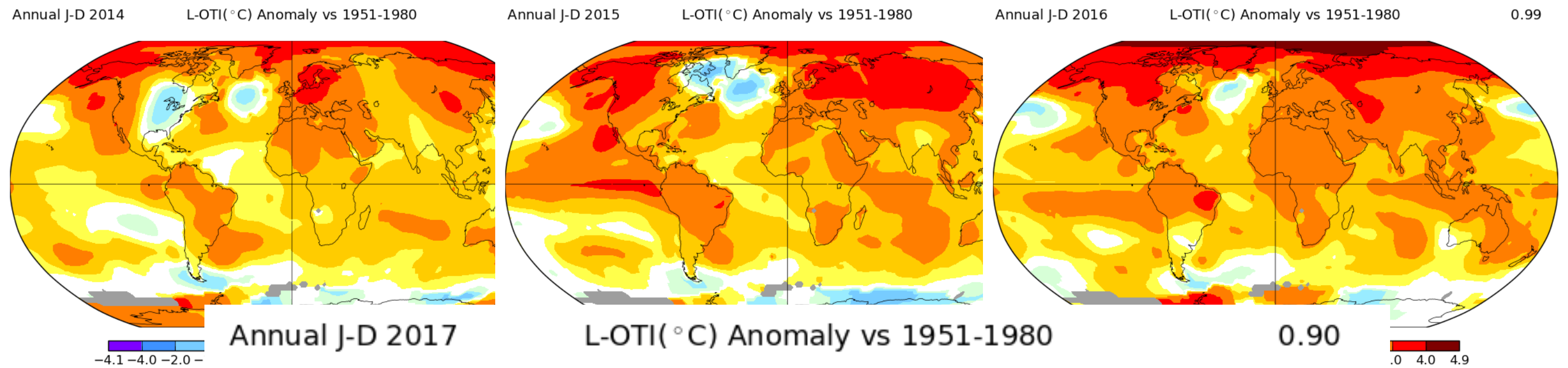
L-OTI(°C) Anomaly vs 1951-1980

0.99

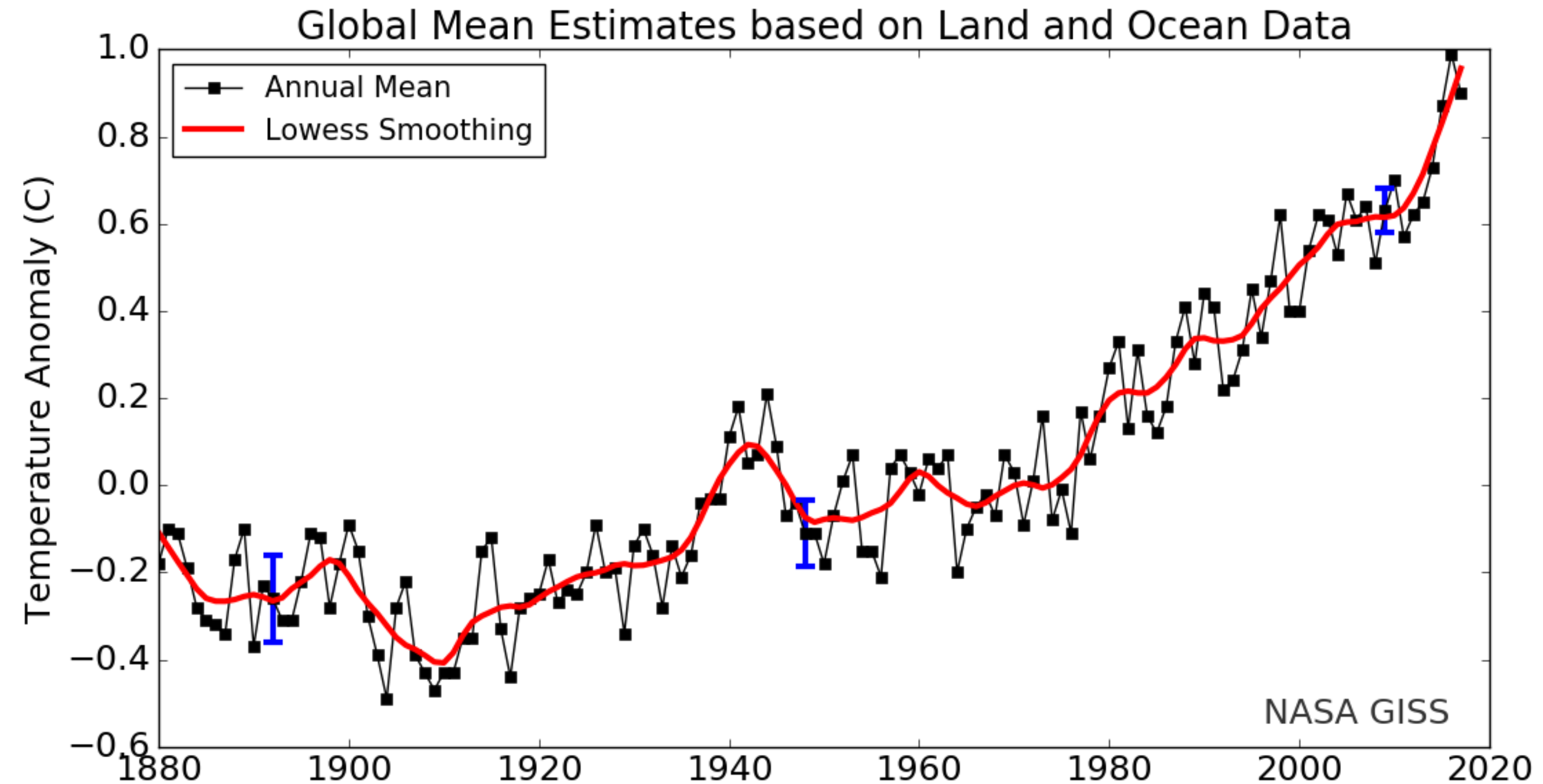
-4.1 -4.0



<https://data.giss.nasa.gov/gistemp/maps/>



<https://data.giss.nasa.gov/gistemp/maps/>

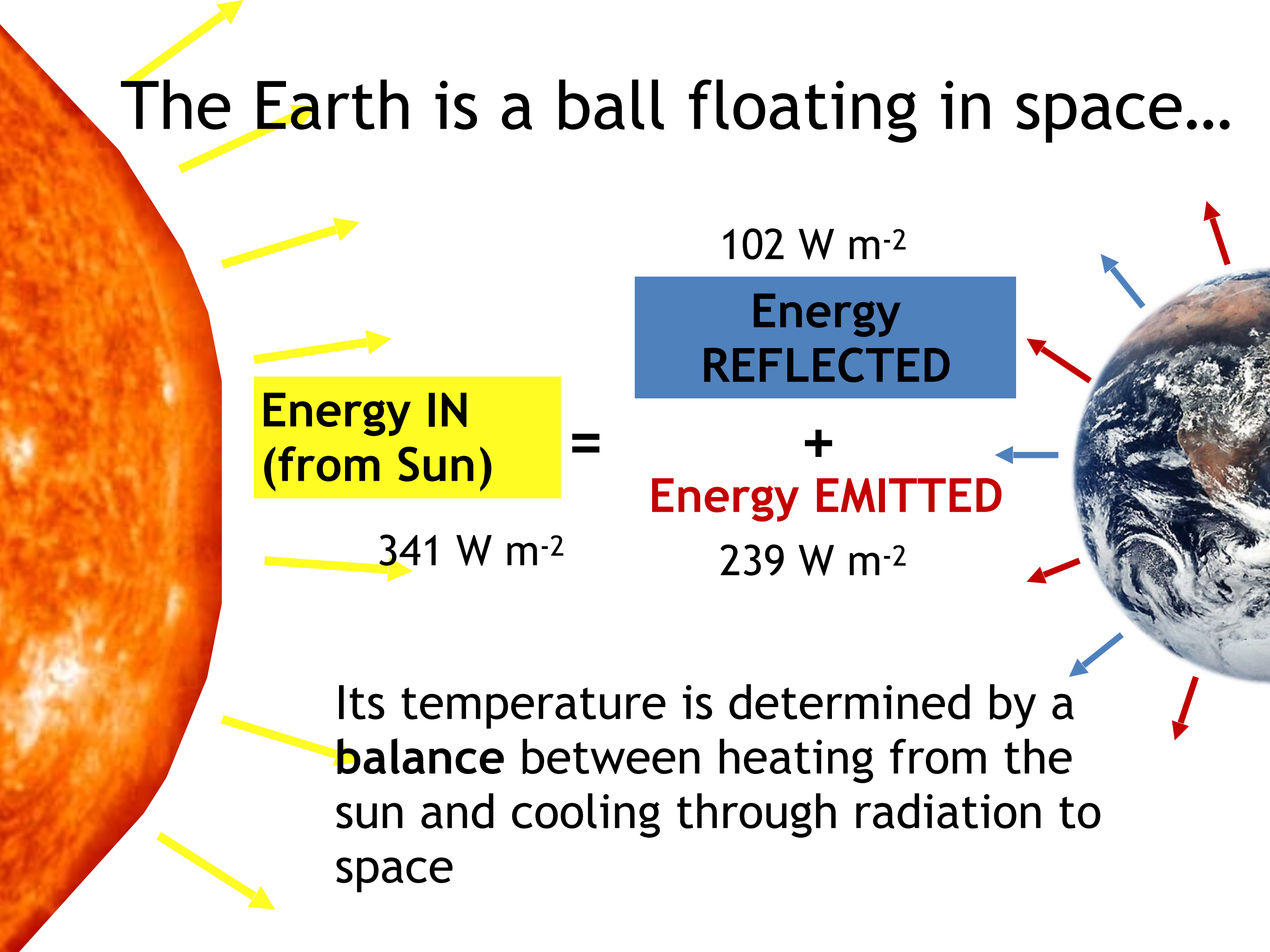


Land-ocean temperature index, 1880 to present, with base period 1951-1980. The solid black line is the global annual mean and the solid red line is the five-year lowess smooth. The blue uncertainty bars (95% confidence limit) account only for incomplete spatial sampling. [This is an update of Fig. 9a in [Hansen et al. \(2010\)](#).]

About this talk

- Look at some basic questions that are central in climate science
- **What is climate sensitivity, how (and how well) do we know what it is?**
- Weighted strongly toward some things that I'm interested in
- Not really a talk about...
 - ✱ Evidence for human influence on climate
 - ✱ Projected future climate change
 - ✱ Local and regional climate
 - ✱ Impacts of future climate change
- Pedagogical, not polemical

The Earth is a ball floating in space...



The diagram illustrates the energy balance of Earth. On the left, a large orange sun is shown with yellow arrows representing incoming solar radiation. One arrow points to a yellow box labeled 'Energy IN (from Sun)' with the value '341 W m⁻²' below it. In the center, an equals sign is followed by a blue box labeled 'Energy REFLECTED' with '102 W m⁻²' above it, and a plus sign followed by 'Energy EMITTED' in red with '239 W m⁻²' below it. On the right, a blue box labeled 'Energy REFLECTED' is shown with '102 W m⁻²' above it. To the right of the Earth, several blue arrows point away from the planet, representing outgoing infrared radiation, and several red arrows point away from the planet, representing reflected solar radiation.

**Energy IN
(from Sun)**

341 W m⁻²

102 W m⁻²

**Energy
REFLECTED**

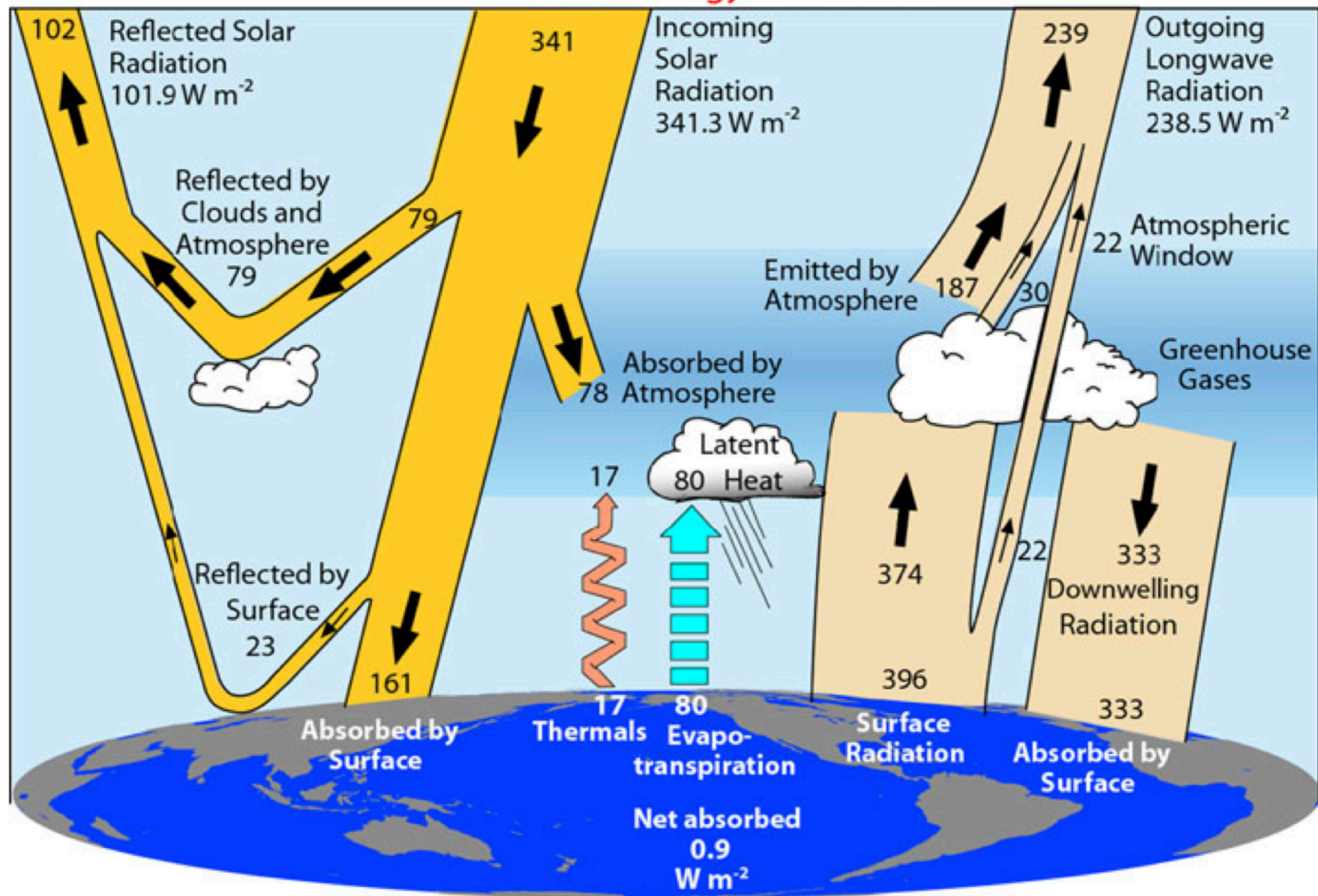
+

Energy EMITTED

239 W m⁻²

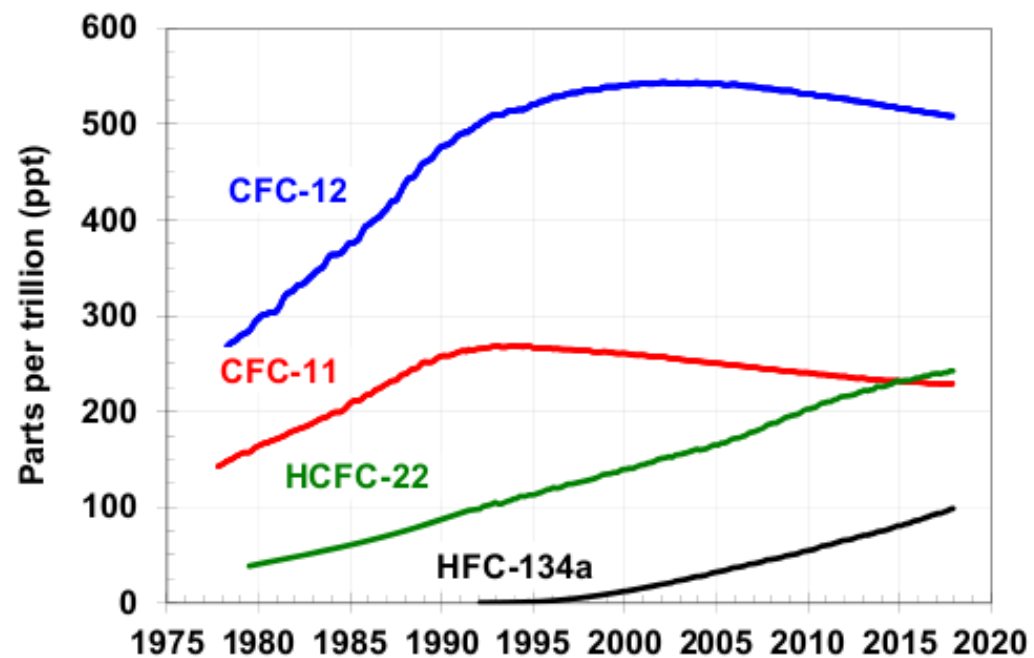
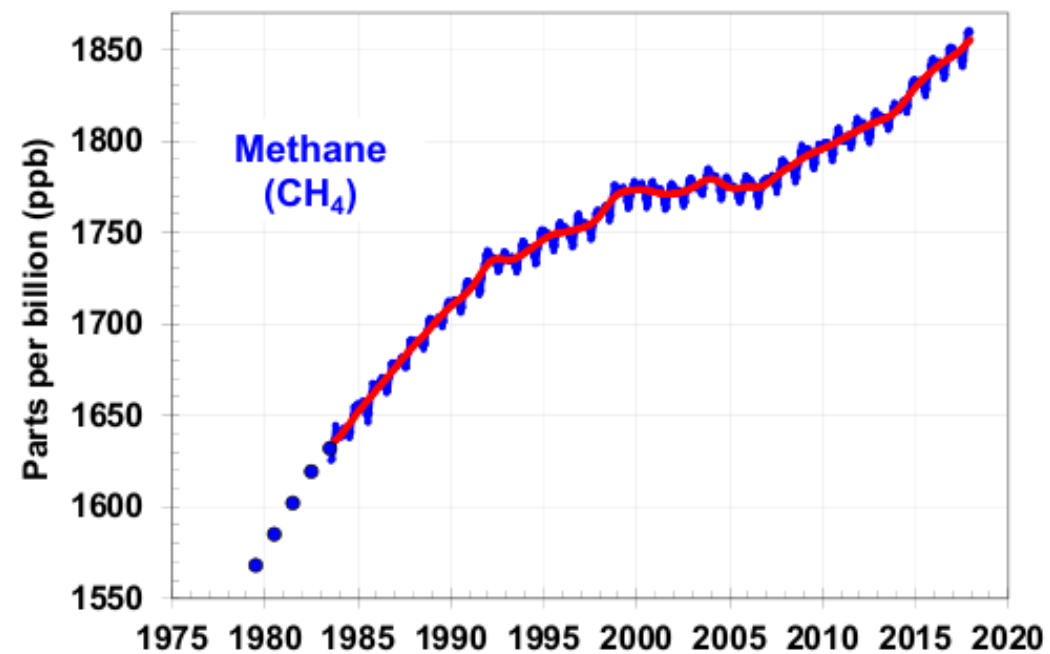
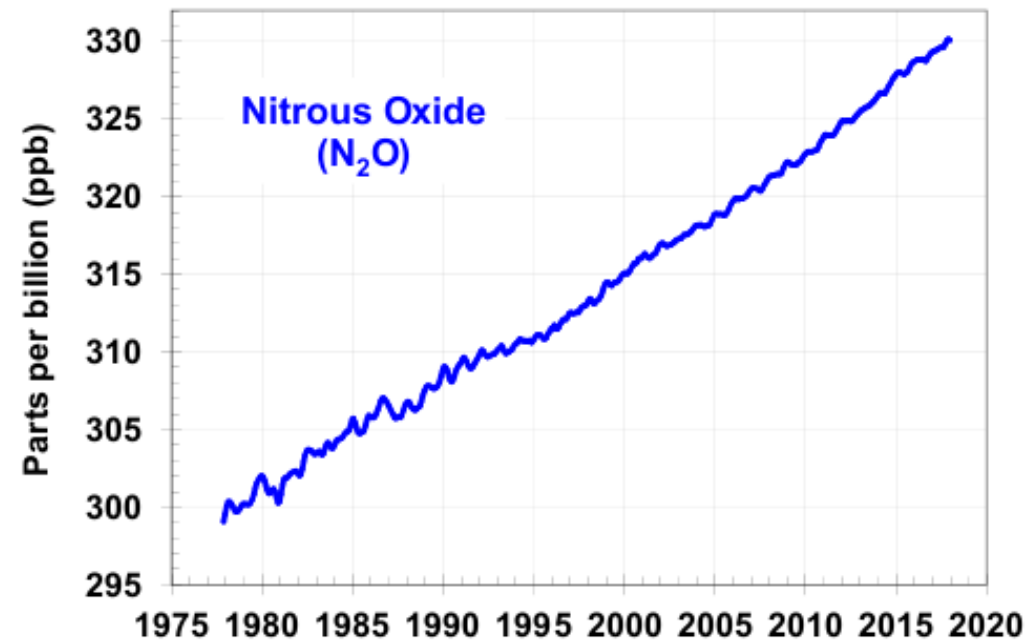
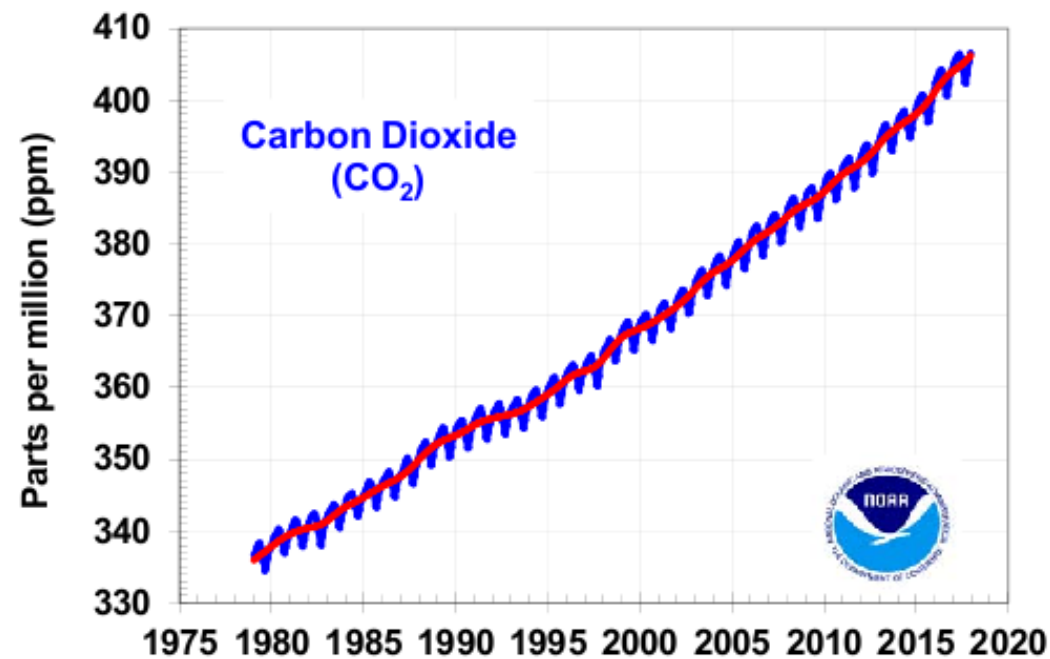
Its temperature is determined by a **balance** between heating from the sun and cooling through radiation to space

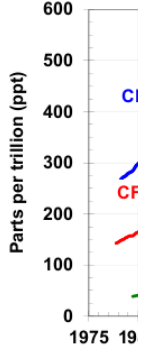
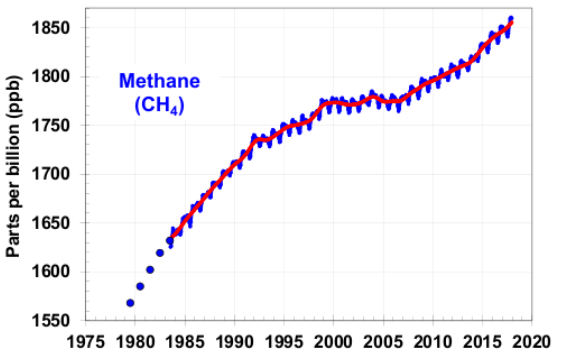
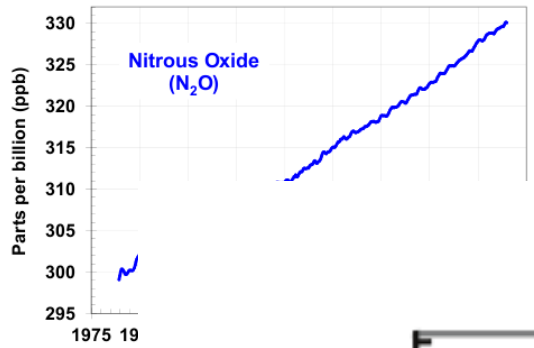
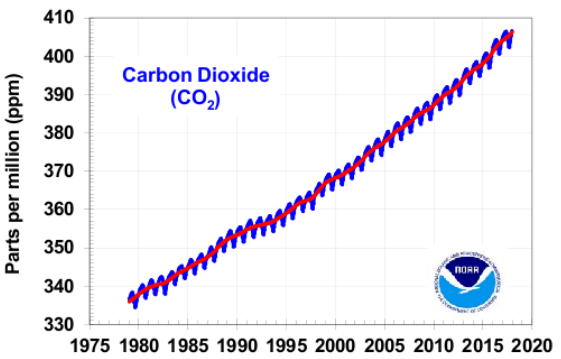
Global Energy Flows W m^{-2}



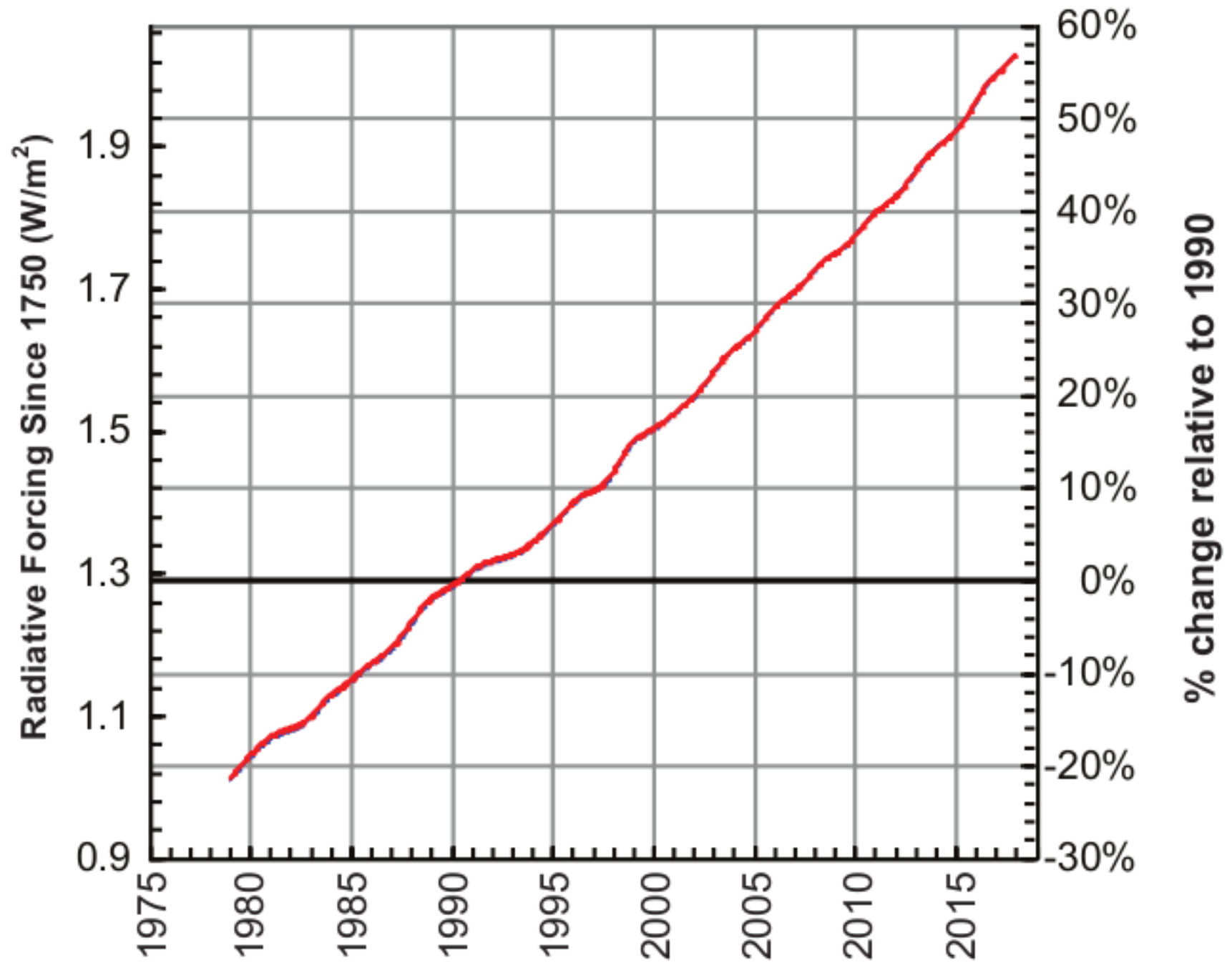
Trenberth and Fasullo (2012) Surv. Geophys.

The “forcing” — agents that cause energy to accumulate within the Earth system





Carbon Dioxide Radiative Forcing



Earth's energy budget

All values are global, annual averages, expressed as deviations from a balanced pre-industrial state

$$\begin{array}{l} \text{"Radiative forcing"} \\ \text{(new energy added to} \\ \text{the system from} \\ \text{greenhouse trapping)} \end{array} \quad \text{—} \quad \begin{array}{l} \text{"Climate response"} \\ \text{(change in emission to} \\ \text{space due to warming)} \end{array} \quad \text{=} \quad \begin{array}{l} \text{"Heat uptake"} \\ \text{(change in energy storage} \\ \text{in Earth system — mostly} \\ \text{the oceans)} \end{array}$$

Earth's energy budget

All values are global, annual averages, expressed as deviations from a balanced pre-industrial state

“Radiative forcing”
(new energy added to the system from greenhouse trapping) **—** **“Climate response”**
(change in emission to space due to warming) **=** **“Heat uptake”**
(change in energy storage in Earth system — mostly the oceans)

$$\Delta F - \lambda \Delta T = \Delta Q$$

Earth's energy budget

All values are global, annual averages, expressed as deviations from a balanced pre-industrial state

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(new energy added to the system from greenhouse trapping) **−** **“Climate response”**
(change in emission to space due to warming) **=** **“Heat uptake”**
(change in energy storage in Earth system — mostly the oceans)

$$\Delta F - \lambda \Delta T = \Delta Q$$

A key assumption: warmer planet emits more to space *(consistent with basic physics)*

λ (“lambda”) is called the Climate Feedback parameter ($\text{W m}^{-2} \text{K}^{-1}$)

Hypothetical climate change scenario

$$\Delta F - \lambda \Delta T = \Delta Q$$



- Instantly double atmospheric CO₂ — $\Delta F_{2x} = 3.7 \text{ W m}^{-2}$
- Initially $\Delta T = 0$, heat accumulates at a rate $\Delta Q = 3.7$
- Oceans warm up, surface temperature increases
- Emission to space increases a bit, energy imbalance gets smaller
- Rate of warming ΔQ gets smaller
- Eventually... Earth returns to energy balance
- $\Delta Q = 0$

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Ultimately, how much warming do we get???

$$\Delta T_{2x} = \frac{\Delta F_{2x}}{\lambda}$$

We've just calculated something called the
Equilibrium Climate Sensitivity

$$\Delta T_{2x} = \frac{\Delta F_{2x}}{\lambda}$$

**The eventual global average surface warming resulting
from a doubling of atmospheric CO₂**

We've just calculated something called the **Equilibrium Climate Sensitivity**

$$\Delta T_{2x} = \frac{\Delta F_{2x}}{\lambda}$$

**The eventual global average surface warming resulting
from a doubling of atmospheric CO₂**

Quantifying ΔT_{2x} (or ECS) is one of the central questions in climate science!

Many **impacts** of climate change are expected to scale roughly with this number.

Low sensitivity ($\Delta T_{2x} \leq 1.5^\circ\text{C}$) $\rightarrow$ mild inconvenience

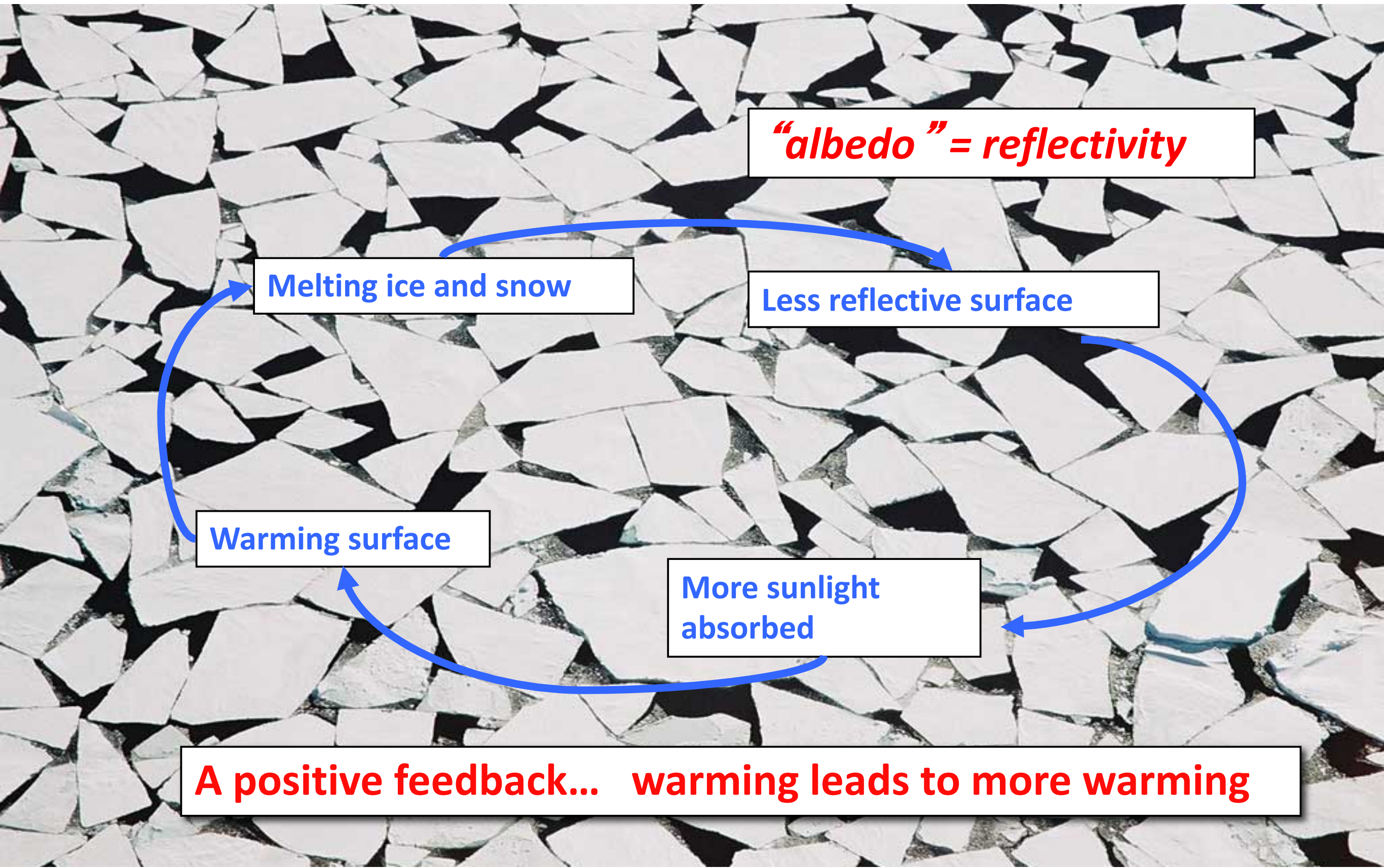
High sensitivity ($\Delta T_{2x} > 4^\circ\text{C}$) $\rightarrow$ a very different Earth for future generations

So what's the answer? How do we know it?

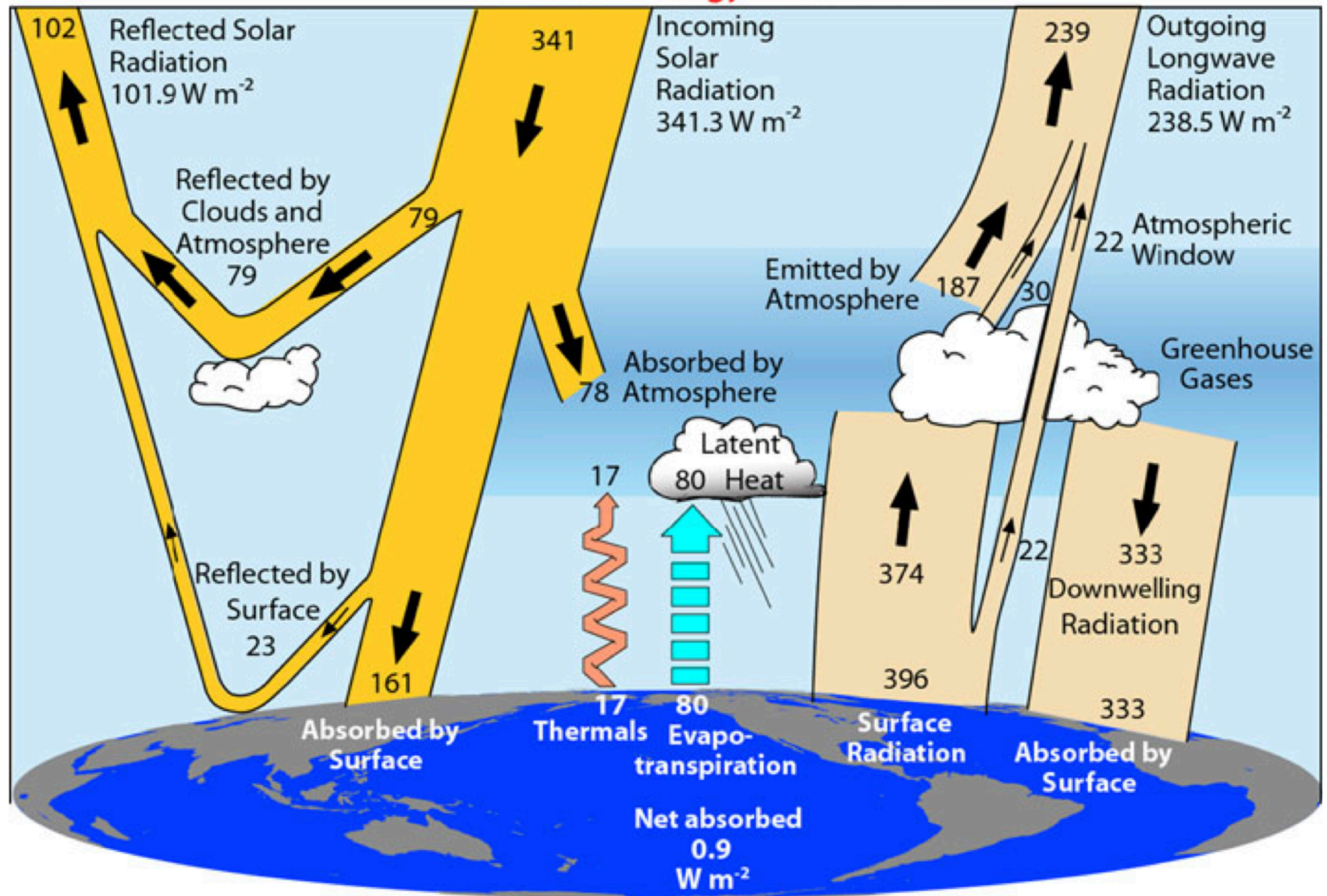
$$\Delta T_{2x} = \frac{\Delta F_{2x}}{\lambda}$$

- $\Delta F_{2x} = 3.7 \text{ W m}^{-2}$ is reasonable well known
- Almost all the uncertainty is in the value of the feedback parameter λ
- This measures the net effect of many processes that act to *amplify* (positive feedback) or *dampen* (negative feedback) the warming response.

Ice Albedo Feedback

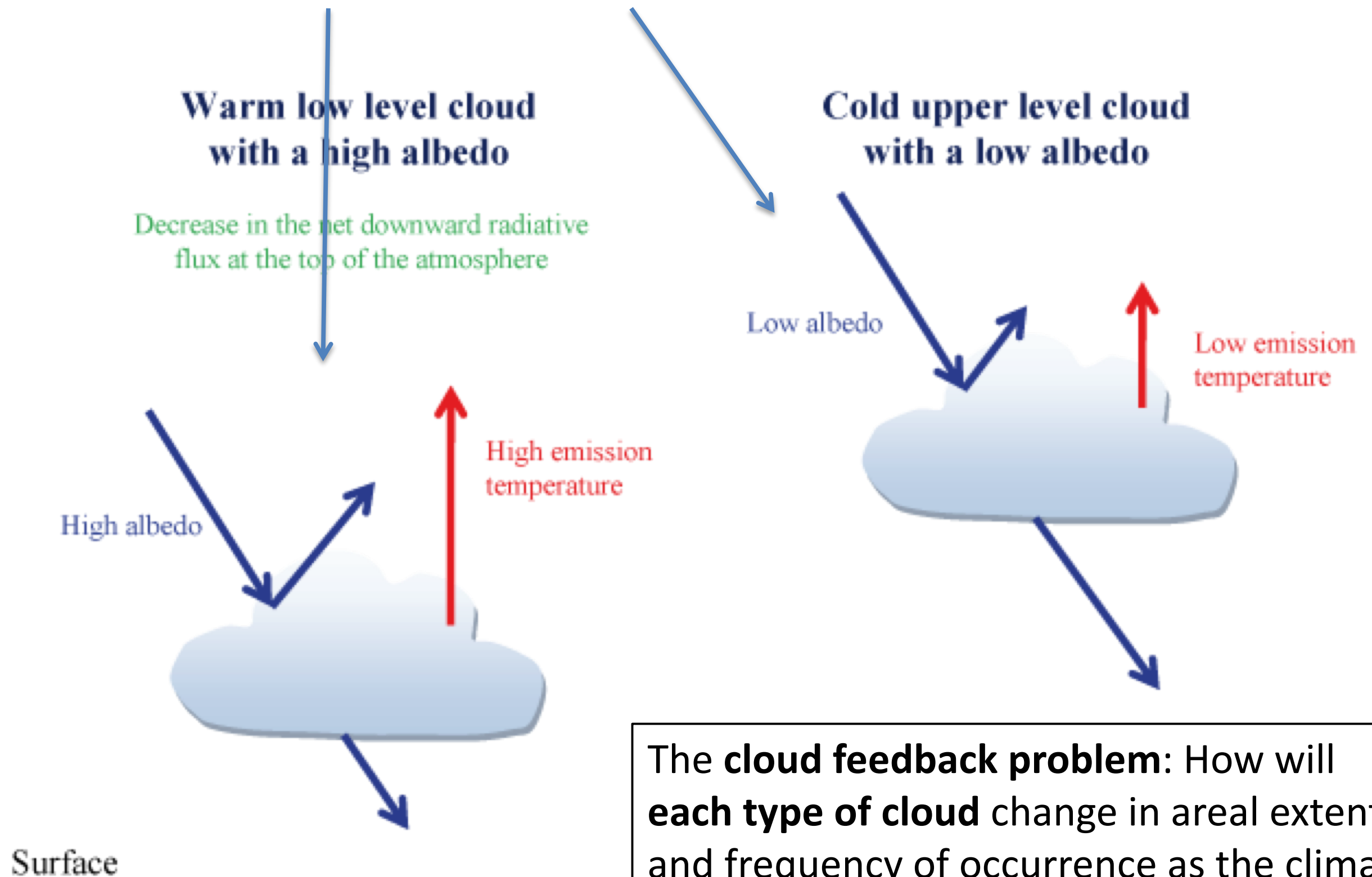


Global Energy Flows W m^{-2}



Trenberth and Fasullo (2012) Surv. Geophys.

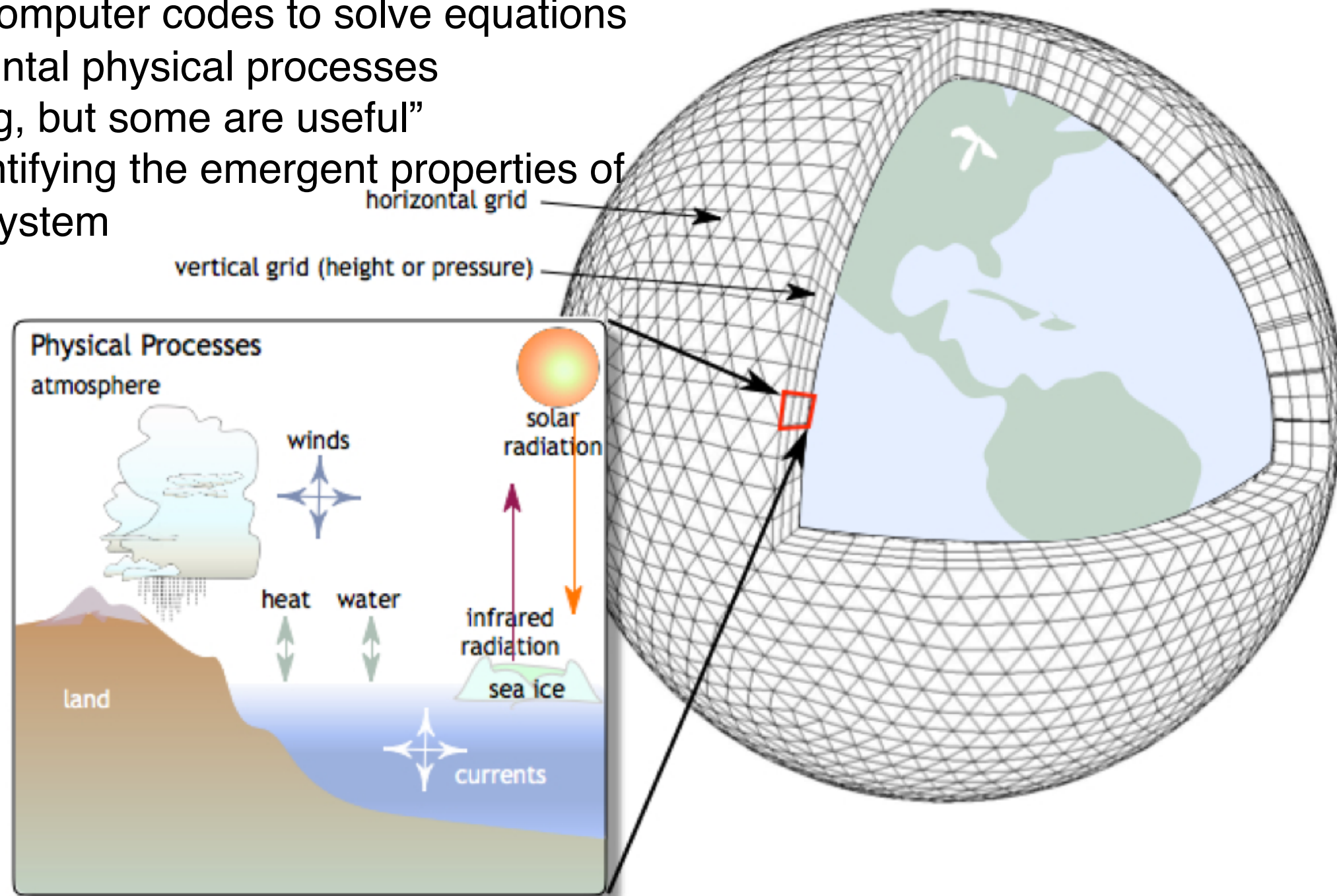
Clouds **cool** and **warm** the planet's surface



The **cloud feedback problem**: How will **each type of cloud** change in areal extent and frequency of occurrence as the climate warms???? *Enormously complicated*

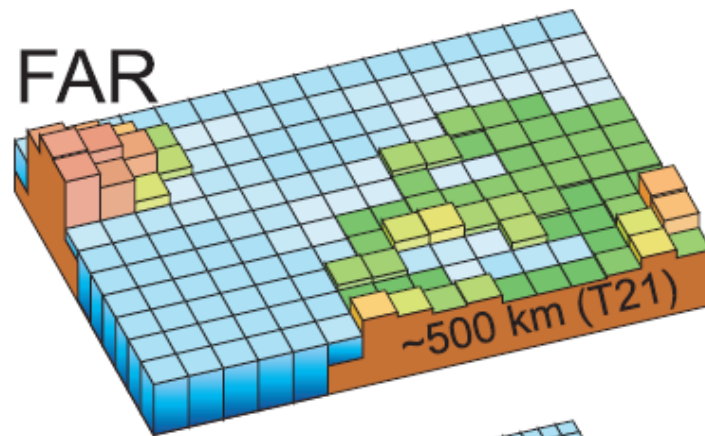
Global climate models...

- Extremely complex computer codes to solve equations representing fundamental physical processes
- “All models are wrong, but some are useful”
- Our best way of quantifying the emergent properties of the complex climate system

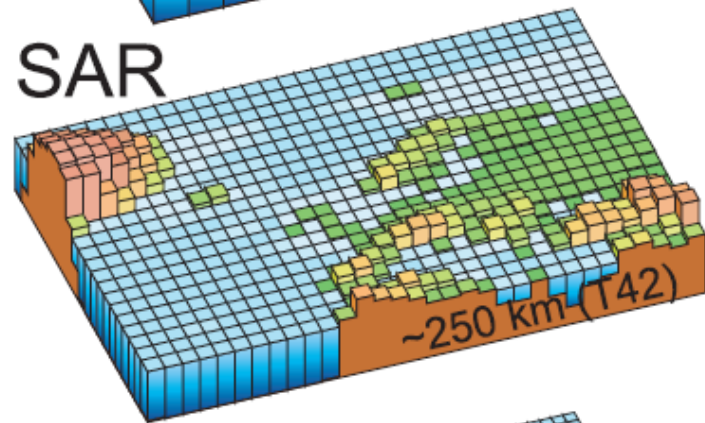


Lots of **different models** with slight differences in the way processes are represented. That's a good thing! Lets us sample some of the **uncertainty** in our **climate projections**.

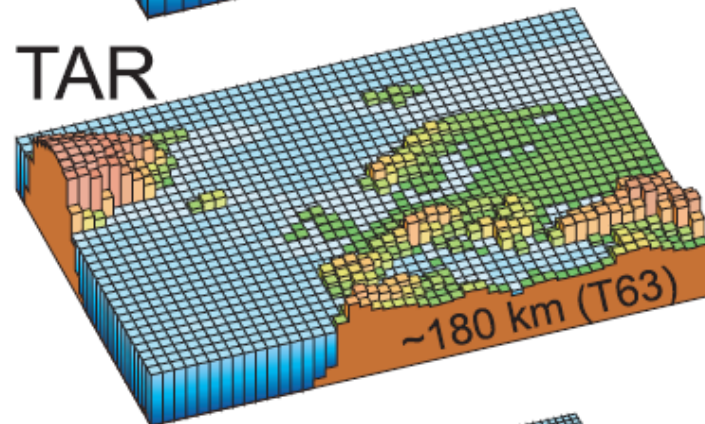
Progress in climate modeling (1)



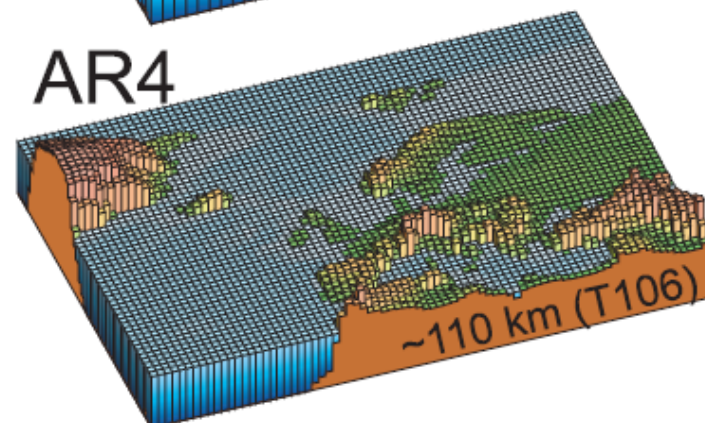
1990



1995



2001



2007

Increasing model resolution typically means **more simulation** and **less parameterization**

Climate models require enormous computer power!



↑
Somewhere in Japan

→
Somewhere in Albany



Higher resolution + more processes resolved + longer simulations + bigger ensembles =
more computing resources needed

History of climate sensitivity estimates

$$\Delta T_{2x} = \frac{\Delta F_{2x}}{\lambda}$$

- 1979 (NAS “Charney report”): $3^{\circ}\text{C} \pm 1.5^{\circ}\text{C}$
- 1990 (IPCC first assessment): 1.5°C to 4.5°C , best guess = 2.5°C
- 2001 (IPCC third assessment): “likely in the range of 1.5 to 4.5°C ”
- 2007 (IPCC fourth assessment): “very likely is greater than 1.5°C and **likely to lie in the range 2 to 4.5°C** , with a most likely value of about 3°C ”
- 2013 (IPCC fifth assessment): “**likely in the range 1.5°C to 4.5°C** (high confidence), extremely unlikely less than 1°C (high confidence), and very unlikely greater than 6°C (medium confidence)”

Knowledge and confidence increasing

History of climate sensitivity estimates

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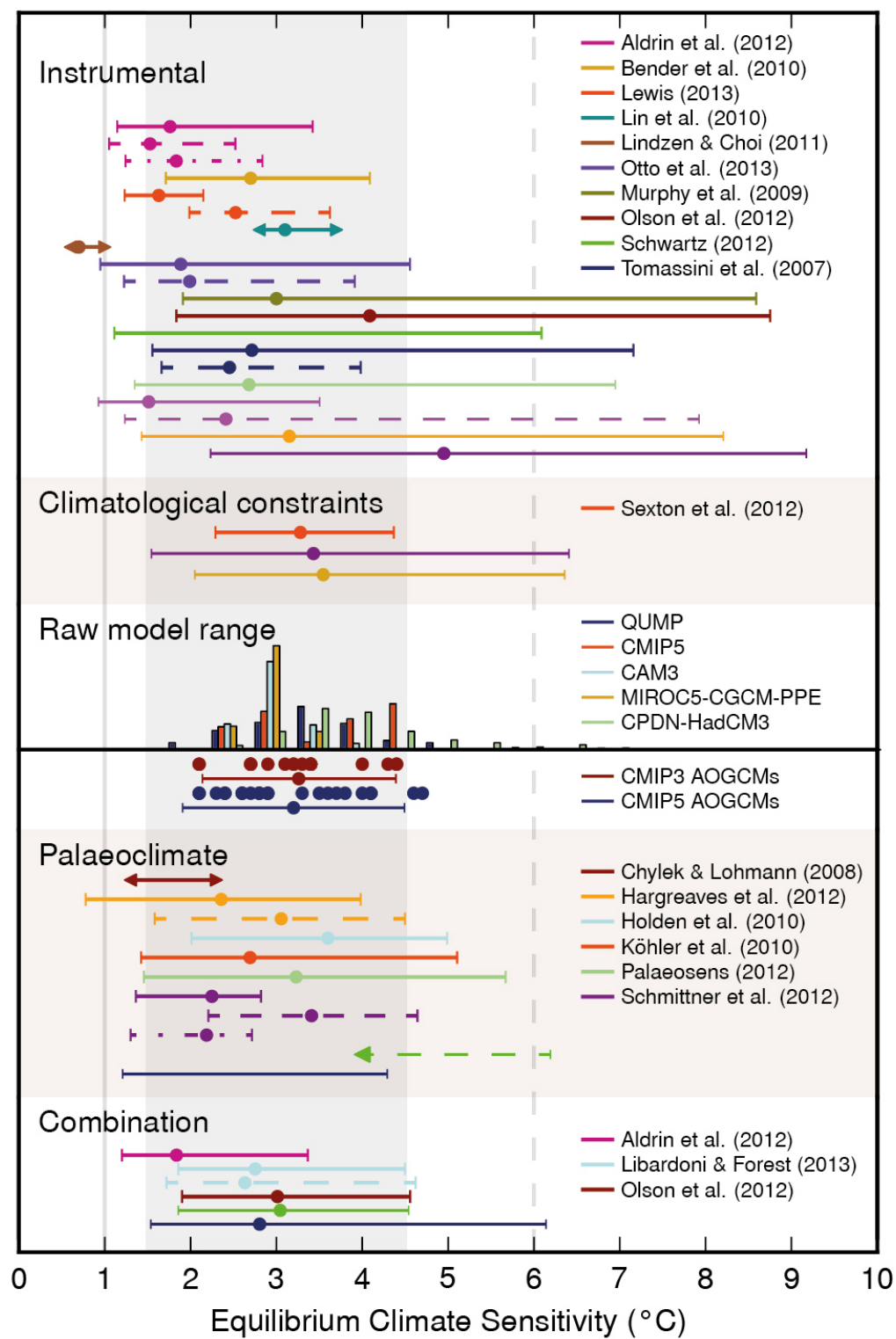
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Knowledge and confidence increasing

Why did low sensitivity become more likely between 2007 and 2013?
Is this correct?

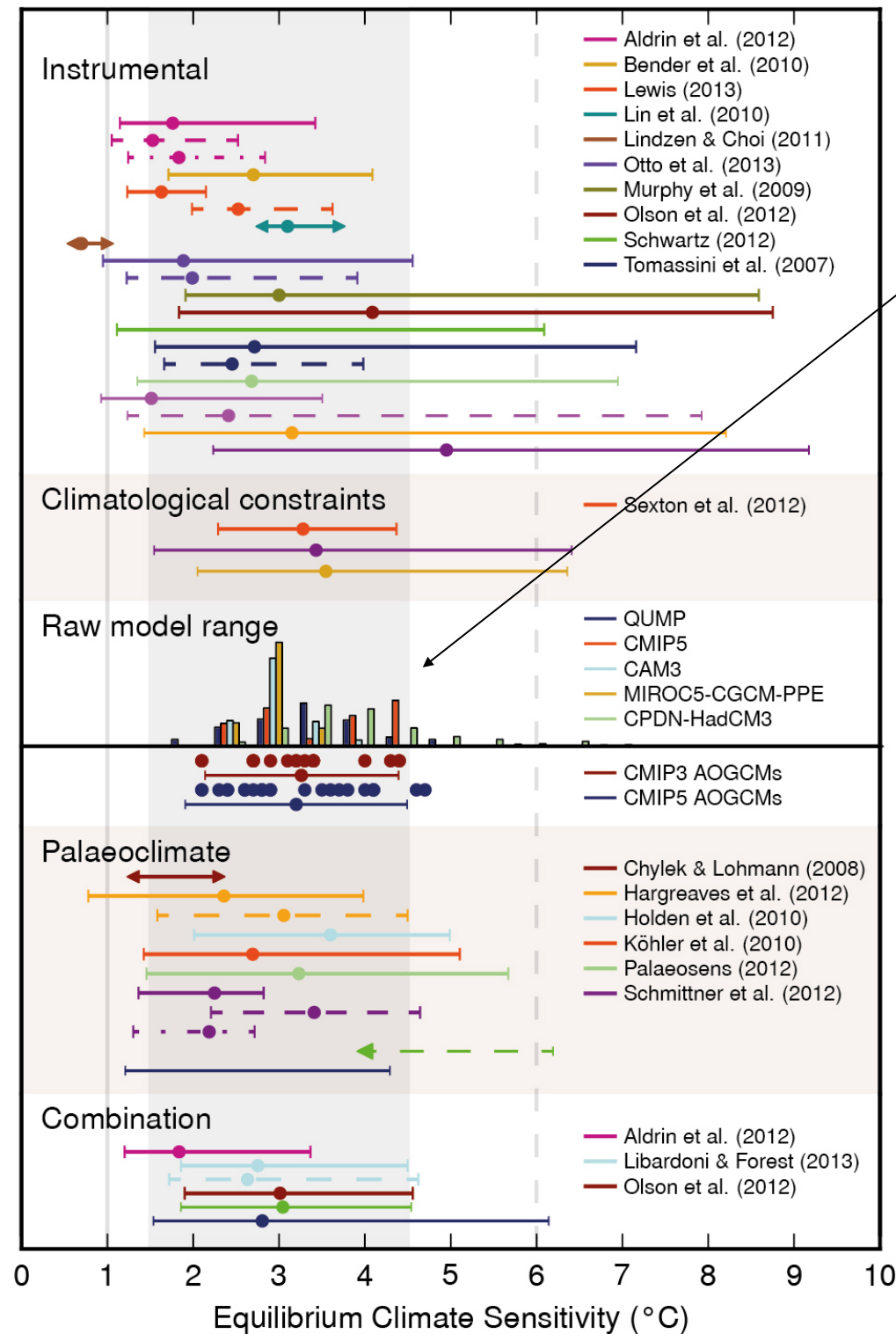
IPCC report (2013): multiple lines of evidence for climate sensitivity

$$\Delta T_{2x} = \frac{\Delta F_{2x}}{\lambda}$$



Box 12.2, Figure 1 | Probability density functions, distributions and ranges for equilibrium climate sensitivity, based on Figure 10.20b plus climatological constraints shown in IPCC AR4 (Meehl et al., 2007b; Box 10.2, Figure 1), and results from CMIP5 (Table 9.5). The grey shaded range marks the *likely* 1.5°C to 4.5°C range, and the grey solid line the *extremely unlikely* less than 1°C, the grey dashed line the *very unlikely* greater than 6°C. See Figure 10.20b and Chapter 10 Supplementary Material for full caption and details. Labels refer to studies since AR4. Full references are given in Section 10.8.

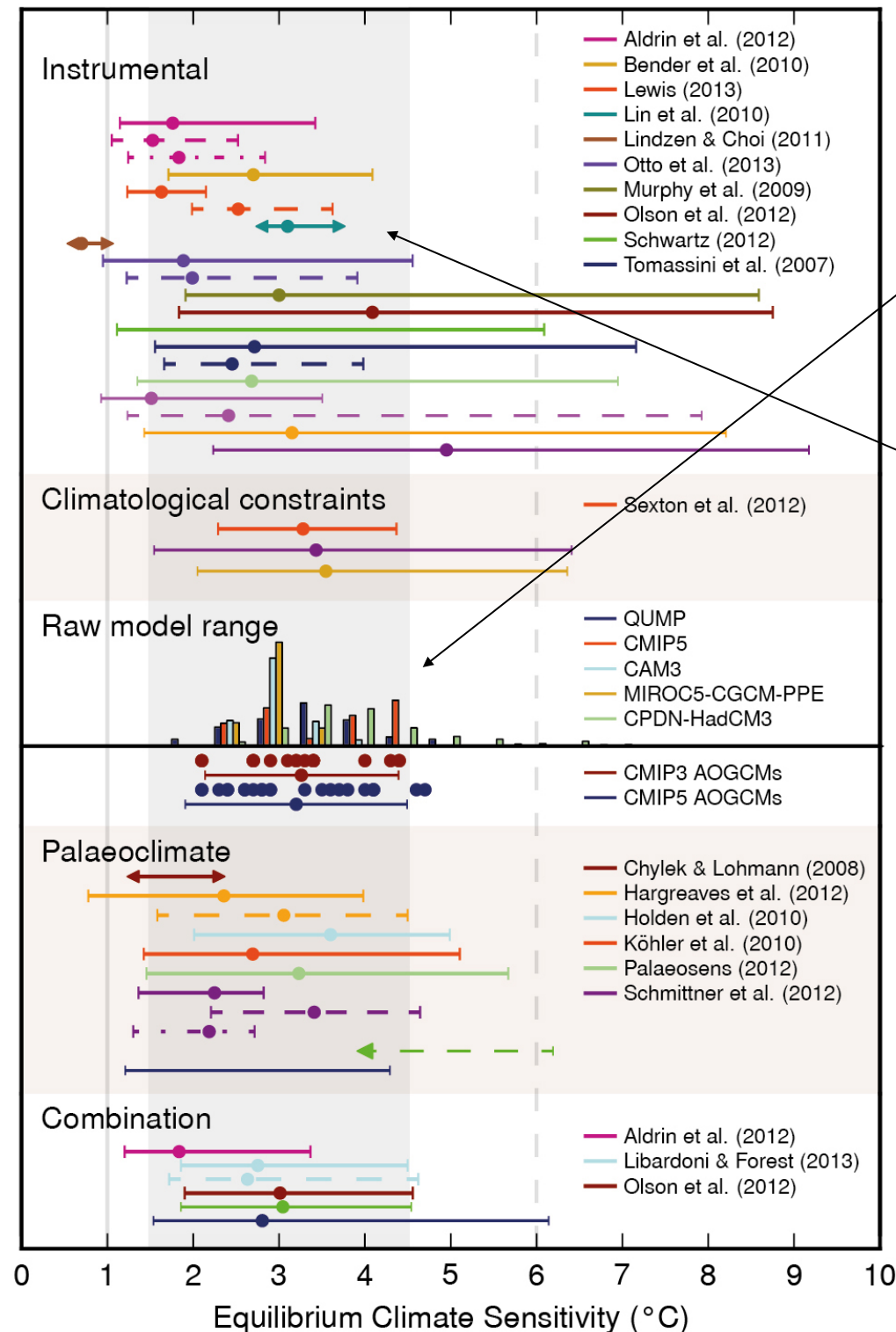
IPCC report (2013): multiple lines of evidence for climate sensitivity $\Delta T_{2x} = \frac{\Delta F_{2x}}{\lambda}$



Predictions from climate models consistently medium to high sensitivity

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IPCC report (2013): multiple lines of evidence for climate sensitivity $\Delta T_{2x} = \frac{\Delta F_{2x}}{\lambda}$



Predictions from climate models consistently medium to high sensitivity

But some studies using the instrumental (observed) climate record argued for lower sensitivity

Box 12.2, Figure 1 | Probability density functions, distributions and ranges for equilibrium climate sensitivity, based on Figure 10.20b plus climatological constraints shown in IPCC AR4 (Meehl et al., 2007b; Box 10.2, Figure 1), and results from CMIP5 (Table 9.5). The grey shaded range marks the *likely* 1.5°C to 4.5°C range, and the grey solid line the *extremely unlikely* less than 1°C, the grey dashed line the *very unlikely* greater than 6°C. See Figure 10.20b and Chapter 10 Supplementary Material for full caption and details. Labels refer to studies since AR4. Full references are given in Section 10.8.

One influential study: observations of global energy budget cannot rule out low sensitivity, e.g. $\Delta T_{2x} \sim 1.5^\circ\text{C}$

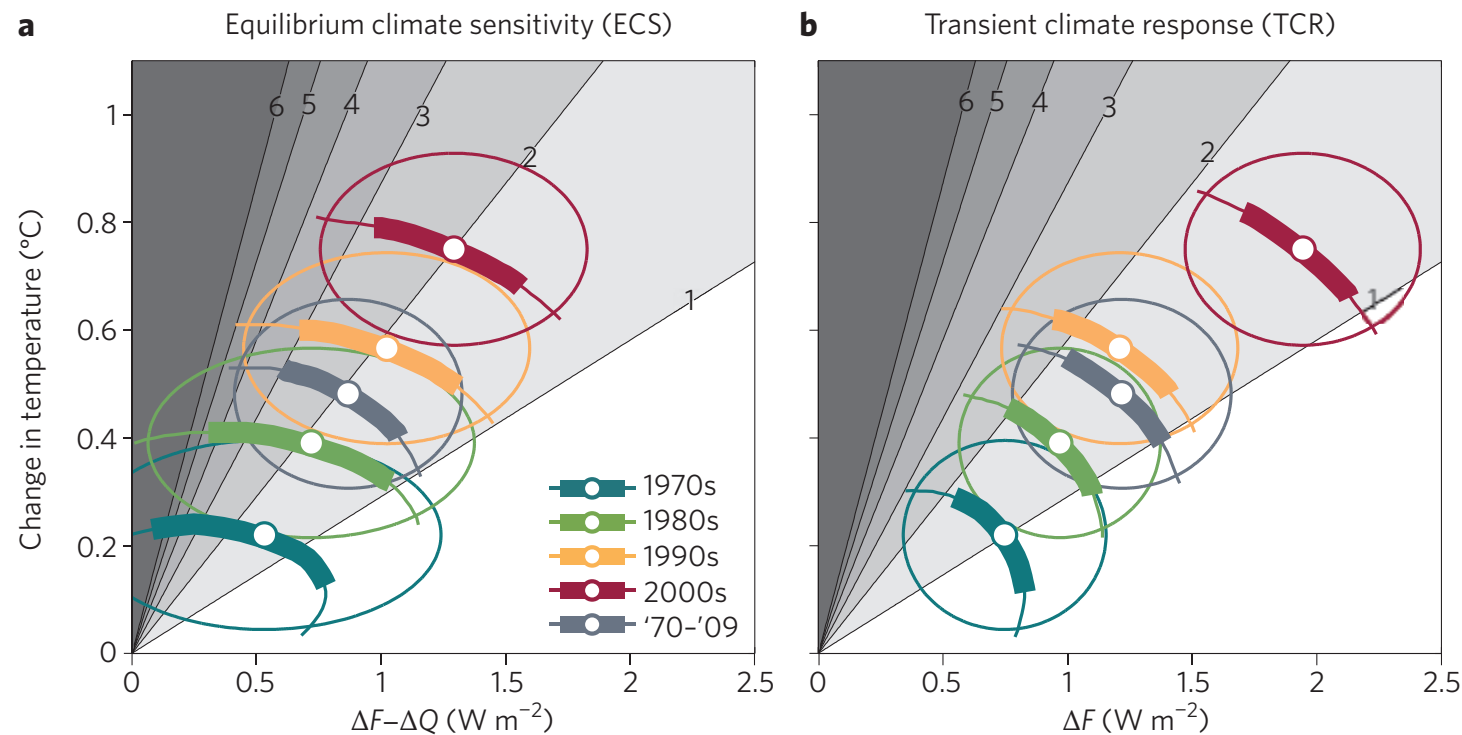


Figure 1 | Observations of the global energy budget and their implications. Observations of the global mean temperature change plotted against change in forcing minus heat uptake ($\Delta F - \Delta Q$) for the equilibrium climate sensitivity (ECS) (**a**) and against ΔF for the transient climate response (TCR) (**b**), for each of the four decades 1970s, 1980s, 1990s and 2000s and for the 40-year period 1970–2009. Ellipses represent likelihood contours enclosing 66% two-dimensional confidence regions; best-fit points of maximum likelihood are indicated by the circles; and the curved thick and thin lines represent the 17–83% and 5–95% confidence intervals of the resulting one-dimensional likelihood profile in ECS (or TCR), respectively. All time periods are referenced to 1860–1879, including a small correction in ΔQ to account for disequilibrium in this reference period¹⁴. Straight contours show iso-lines of ECS (**a**) and TCR (**b**), calculated using a best-fit value of F_{2x} of 3.44 W m^{-2} (also adjusted for fast feedbacks)¹⁰. Uncertainty in F_{2x} is assumed to be correlated with forcing uncertainty in long-lived greenhouse gases¹⁰. To avoid dependence on previous assumptions¹⁶, we report results as likelihood-based confidence intervals.

**Otto et al. (2013), *Nature Geosci.*
doi:10.1038/ngeo1836**

The “Energy Budget” method

using observations of global planetary energy budget to calculate climate sensitivity

Assume the observed period obeys our basic climate model

$$\Delta F - \lambda \Delta T = \Delta Q$$

We can quantify these (global averages)
from observations

Historical radiative forcing is reasonably
well known (greenhouse gases,
anthropogenic aerosols, volcanoes, etc)

The “Energy Budget” method

using observations of global planetary energy budget to calculate climate sensitivity

Then from historical observations we can calculate (with some uncertainty) the historical feedback parameter

$$\lambda_{obs} = \frac{\Delta F_{obs} - \Delta Q}{\Delta T_{obs}}$$

The “Energy Budget” method

using observations of global planetary energy budget to calculate climate sensitivity

$$\lambda_{obs} = \frac{\Delta F_{obs} - \Delta Q}{\Delta T_{obs}}$$

Since Climate Sensitivity is given by $\Delta T_{2x} = \frac{\Delta F_{2x}}{\lambda}$

it then follows that, using historical observations, we can infer

$$\Delta T_{2x} = \frac{\Delta T_{obs} \Delta F_{2x}}{(\Delta F_{obs} - \Delta Q)}$$

Empirical measure of climate sensitivity, independent of climate models! (??)

The “Energy Budget” method

using observations of global planetary energy budget to calculate climate sensitivity

$$\Delta T_{2x} = \frac{\Delta T_{obs} \Delta F_{2x}}{(\Delta F_{obs} - \Delta Q)}$$

*The method is simple and satisfying...
so simple that we often forget that it DOES depend on a climate model*

Specifically, the assumption is that $\lambda_{2x} = \lambda_{obs}$

**.. the feedbacks for short-term climate variability ARE THE SAME
as the feedbacks governing long-term greenhouse warming**

But this is not true!

*Big thrust of recent research: understanding how and why
lambda (i.e. the feedback) varies in time*

**An uncertain world: many possible simulated climatic histories.
Reality chose one particular path**

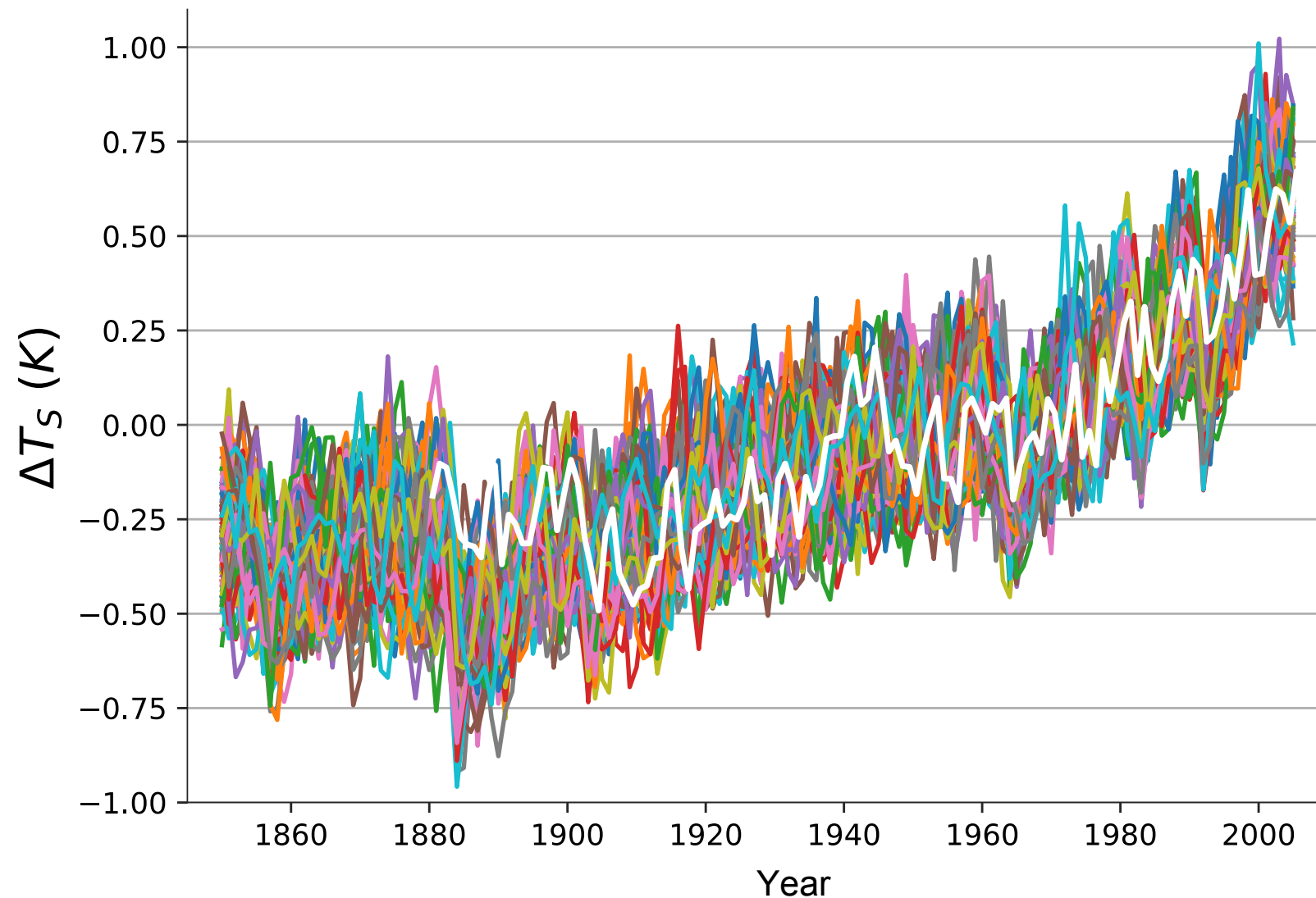
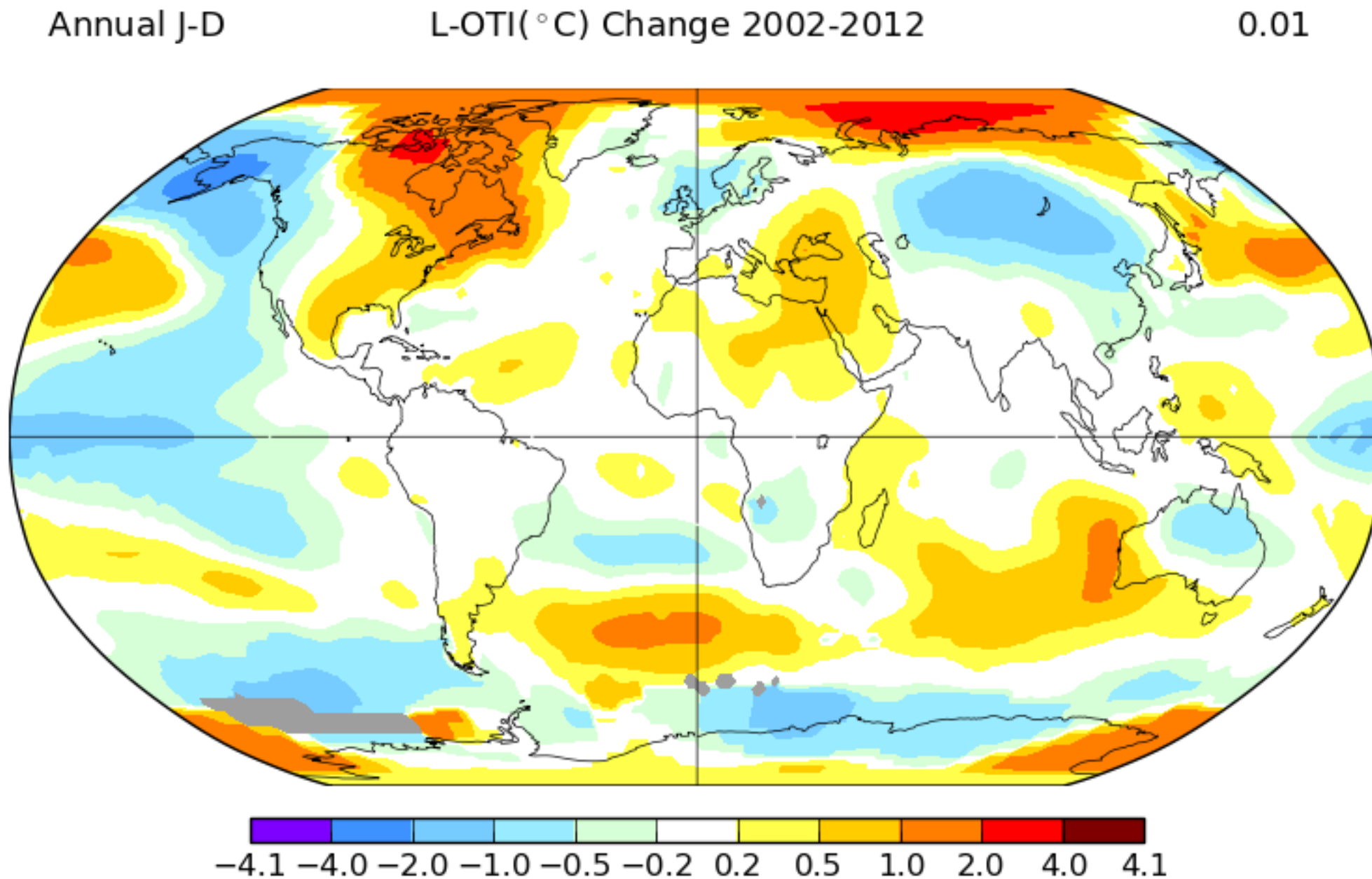


Figure 1. Plot of annual and global average surface temperature from the 100 members of the MPI-ESM1.1 ensemble (colored lines), along with the GISTEMP measurements (Hansen et al., 2010) (white line). Temperatures are referenced to the 1951–1980 average.

**Dessler et al. (2018), *ACP*
doi:10.5194/acp-2017-1236**

Temperature trends, 2002 - 2012 (the “hiatus” period)



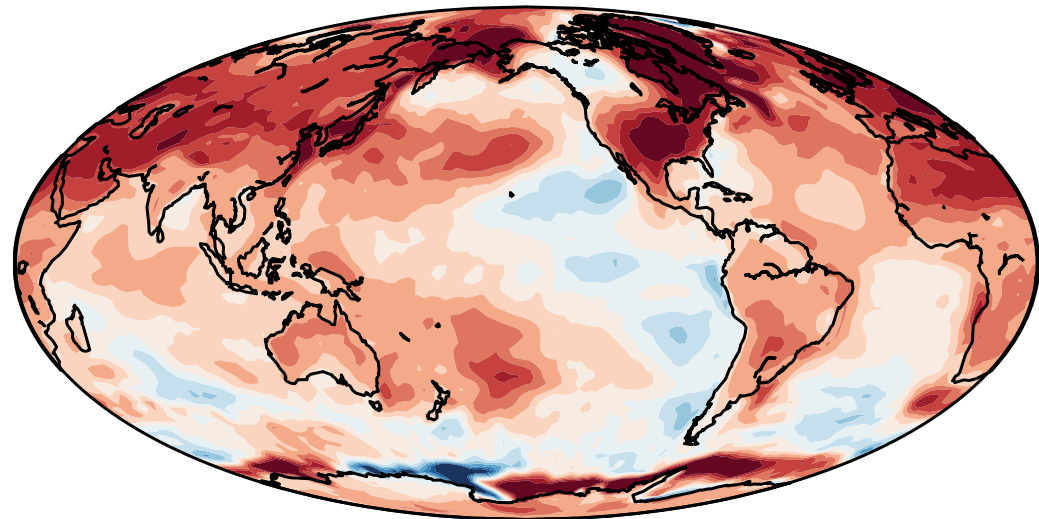
Anomalous cooling in central and east tropical Pacific

<https://data.giss.nasa.gov/gistemp/maps/>

Climate models did not “predict” the hiatus.

a

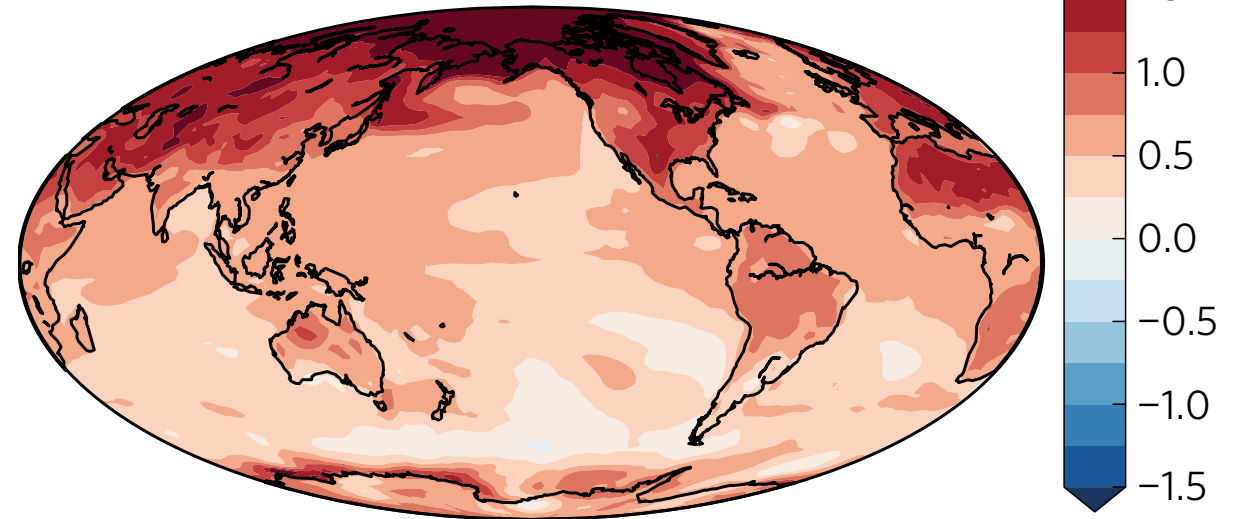
AMIP T_s trend



Observations

b

CMIP5-historical T_s trend

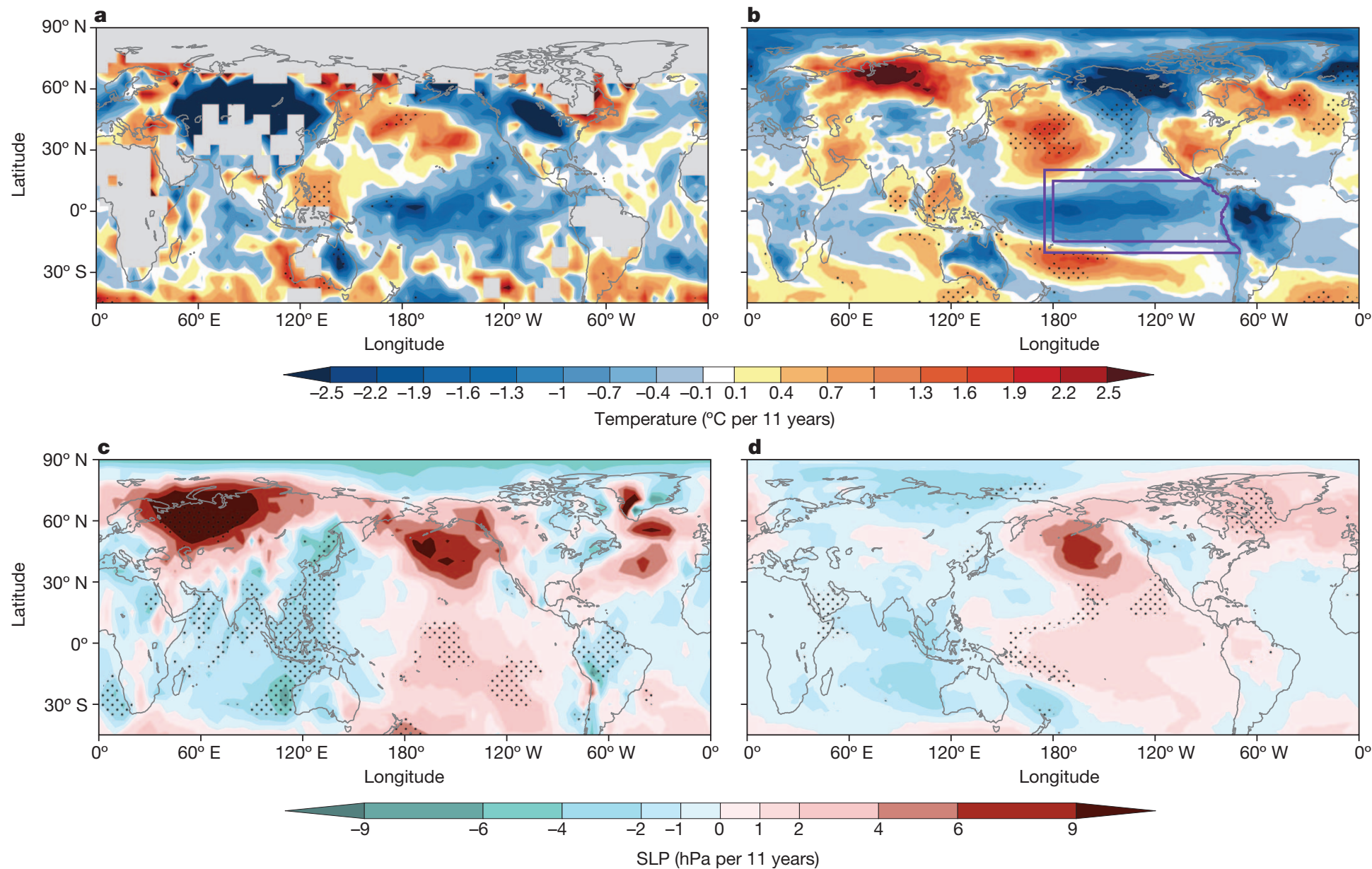


Models

**Averaging over many simulations gives the “forced” response —
the variability is not as predictable**

**Zhou et al. (2016), *Nature Geosci.*
doi:10.1038/NGEO2828**

A clever use of complex climate model as laboratory:

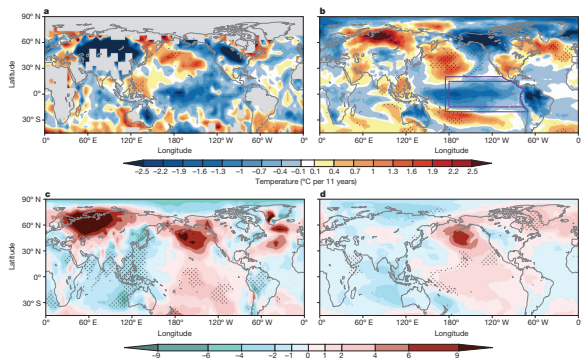


Simulations where the surface temperature is pinned to observations only with the tropical Pacific box

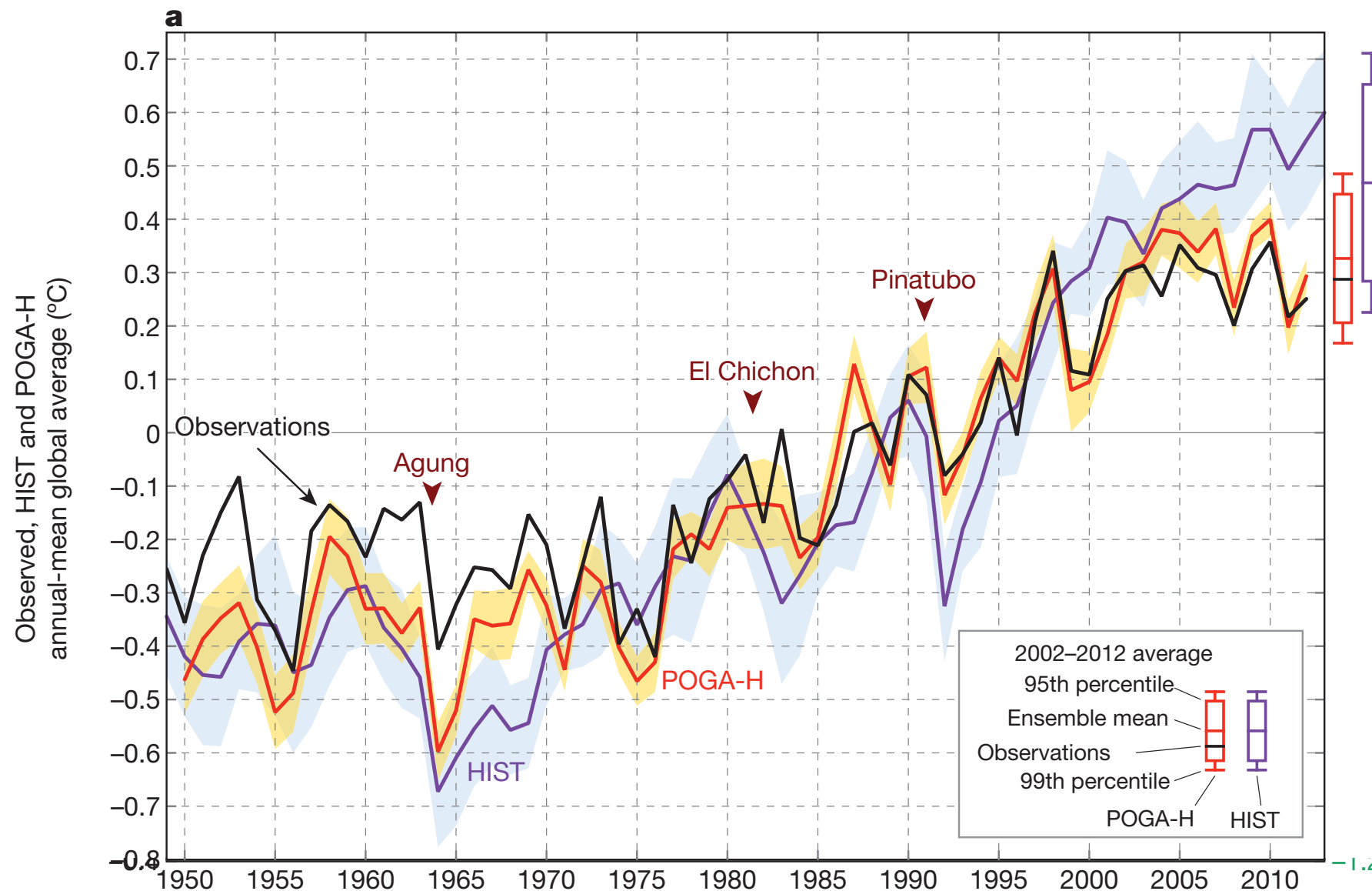
Figure 2 | Observed and simulated trend patterns in boreal winter for 2002–2012. a and b show near-surface temperature and c and d show SLP from observations (a and c) and POGA-H (b and d) in DJF. Grey shading represents

missing values. Stippling indicates regions exceeding 95% statistical confidence. Purple boxes in b show the restoring region of POGA experiments.

**Kosaka and Xie (2013), *Nature*
doi:10.1038/nature12534**



As goes the tropical Pacific, so goes the world...



**Kosaka and Xie (2013), *Nature*
doi:10.1038/nature12534**

Low clouds (reflectors!!) respond strongly to East Pacific temperatures

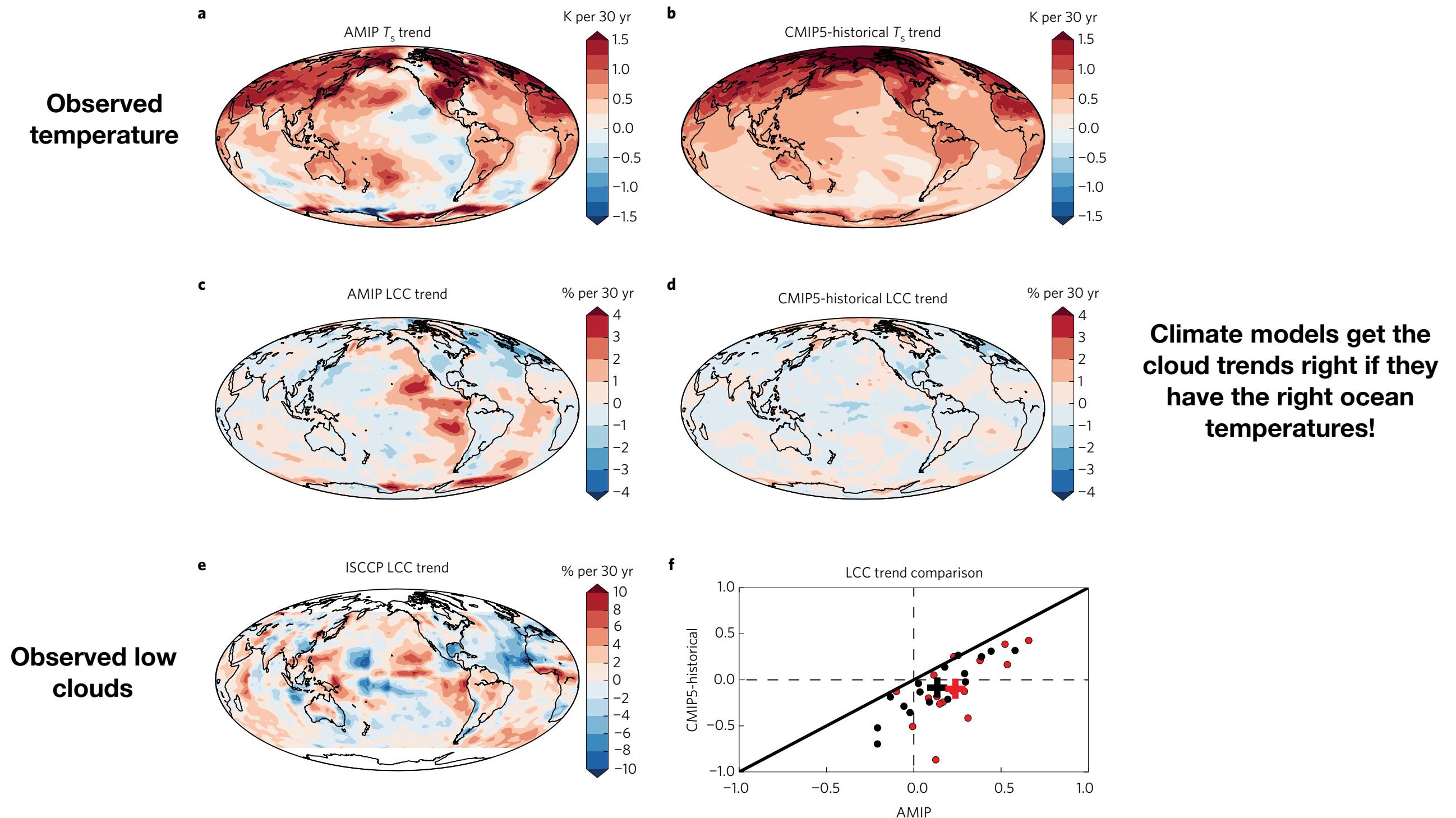
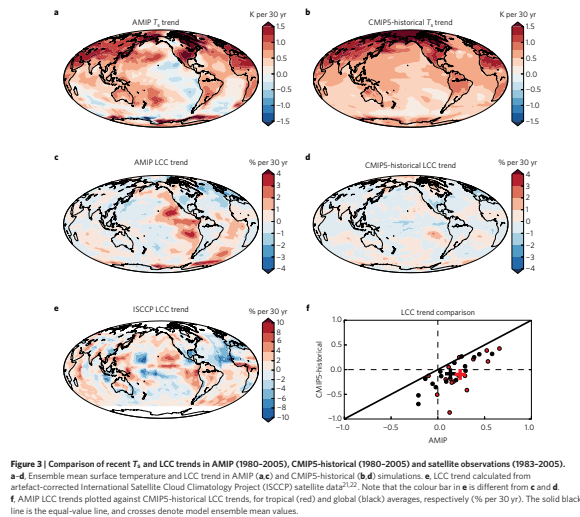


Figure 3 | Comparison of recent T_s and LCC trends in AMIP (1980–2005), CMIP5-historical (1980–2005) and satellite observations (1983–2005).

a–d, Ensemble mean surface temperature and LCC trend in AMIP (**a,c**) and CMIP5-historical (**b,d**) simulations. **e**, LCC trend calculated from artefact-corrected International Satellite Cloud Climatology Project (ISCCP) satellite data^{21,22}. Note that the colour bar in **e** is different from **c** and **d**. **f**, AMIP LCC trends plotted against CMIP5-historical LCC trends, for tropical (red) and global (black) averages, respectively (% per 30 yr). The solid black line is the equal-value line, and crosses denote model ensemble mean values.

Zhou et al. (2016), Nature Geosci.
doi:10.1038/NCEO2828

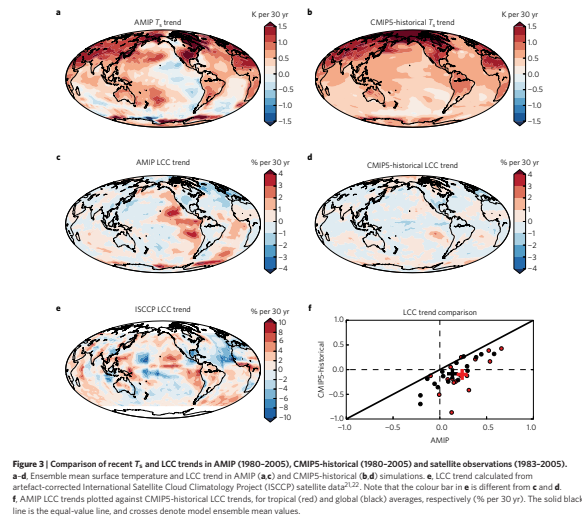
Historical warming/cooling patterns are not good analogs for future warming...



Recent history, especially the “hiatus” in the 2000s:

- suppressed warming of East Pacific
 - *not entirely understood, but due to slowly evolving ocean circulation*
- Increased marine low cloud cover
 - *Suppressed surface warming stabilizes the atmosphere, makes low clouds*
- More reflection of sunlight
- slower global warming
- smaller inferred climate sensitivity

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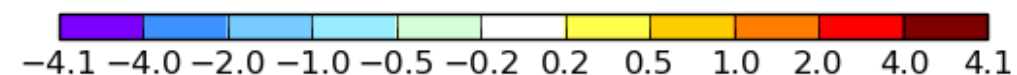
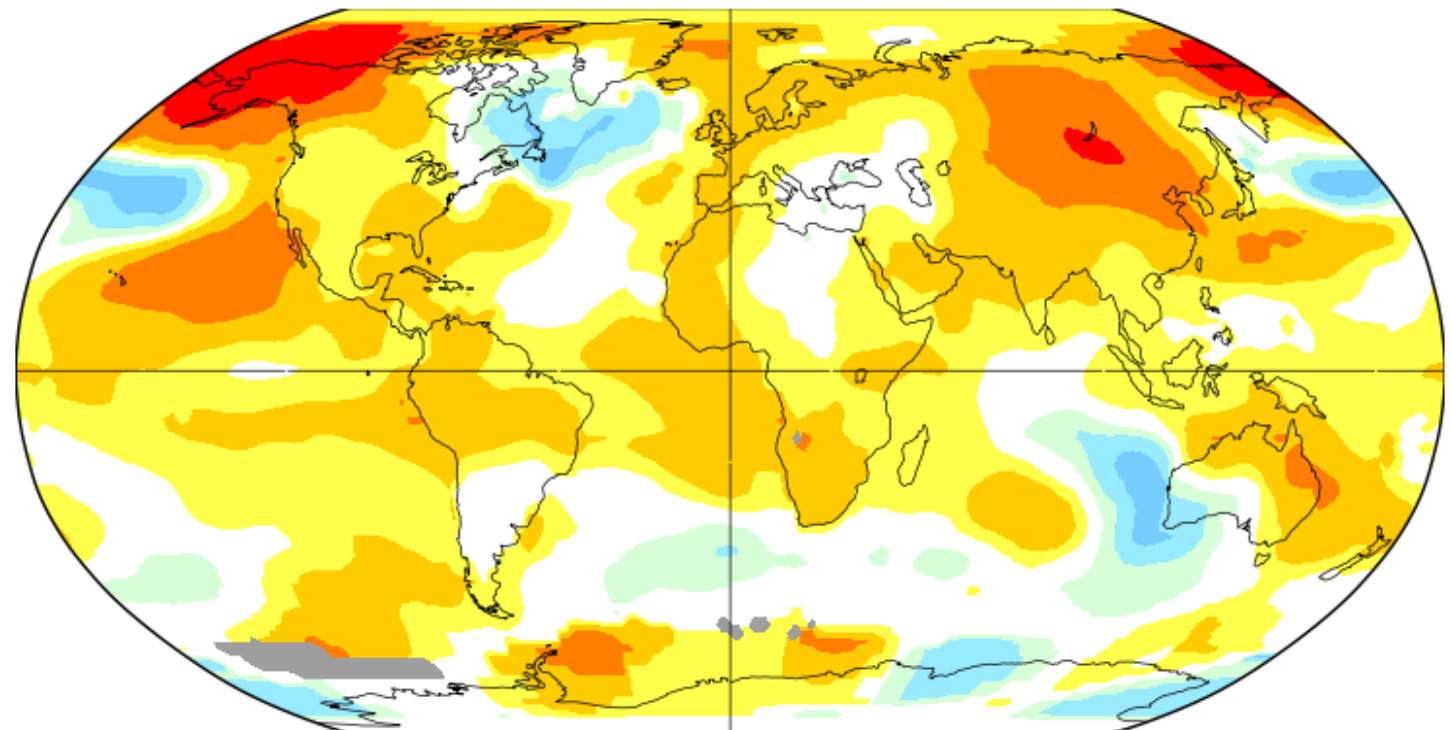
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- Increased marine low cloud cover
 - *Suppressed surface warming stabilizes the atmosphere, makes low clouds*
- More reflection of sunlight
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- smaller inferred climate sensitivity

There is no reason to expect this pattern to continue... and it is not continuing

Annual J-D

L-OTI(°C) Change 2012-2017

0.37



<https://data.giss.nasa.gov/gistemp/maps/>

Using the climate models to differentiate between past and future sensitivity...

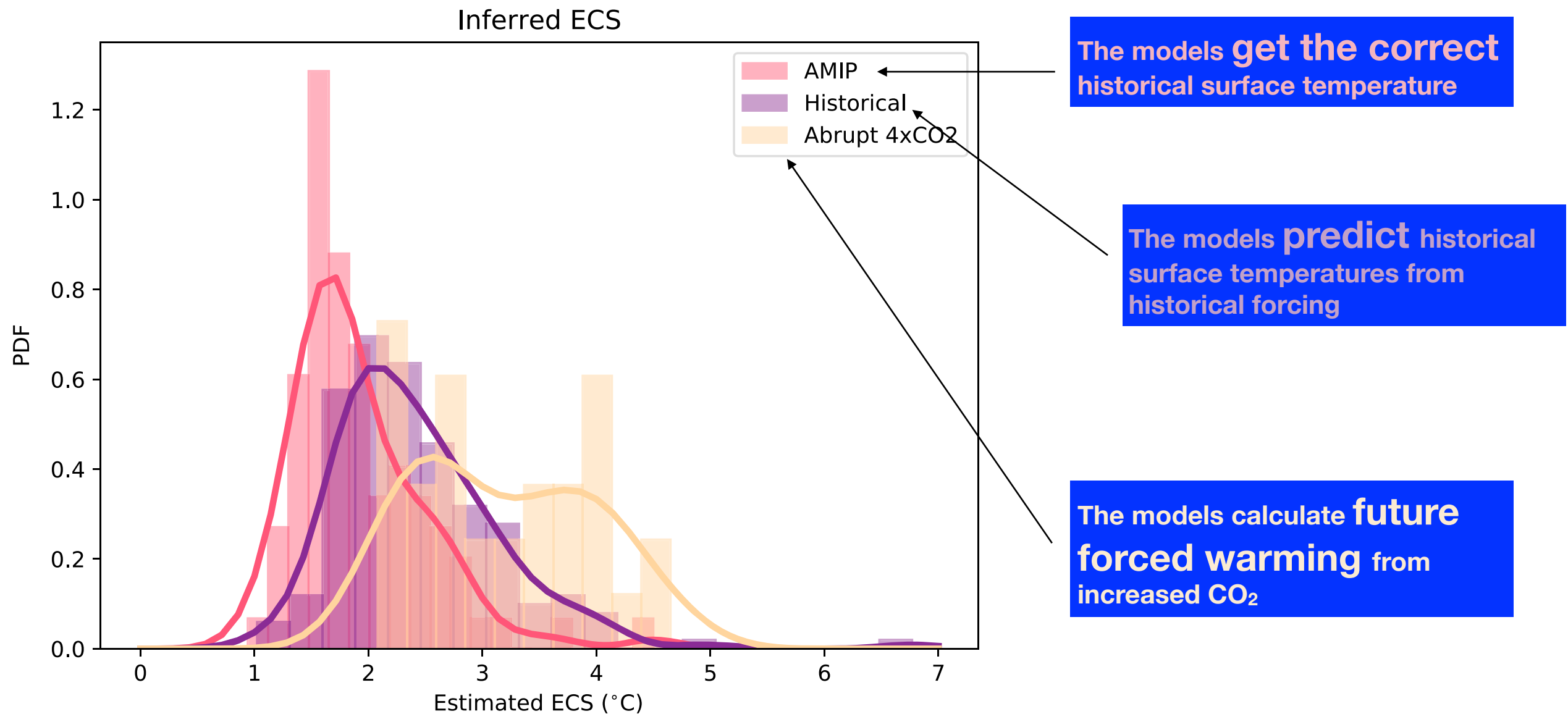


Figure 1. Equilibrium climate sensitivities inferred from amip (pink), historical (purple), and long-term (yellow) simulations, with kernel density estimates overplotted for visual clarity.

Reinterpreting observations of Earth's energy budget

$$\Delta T_{2x} = \frac{\Delta T_{obs} \Delta F_{2x}}{S(\Delta F_{obs} - \Delta Q)}$$

Using state-of-the-art climate models to account for differences in the feedbacks affecting historical and future climate change

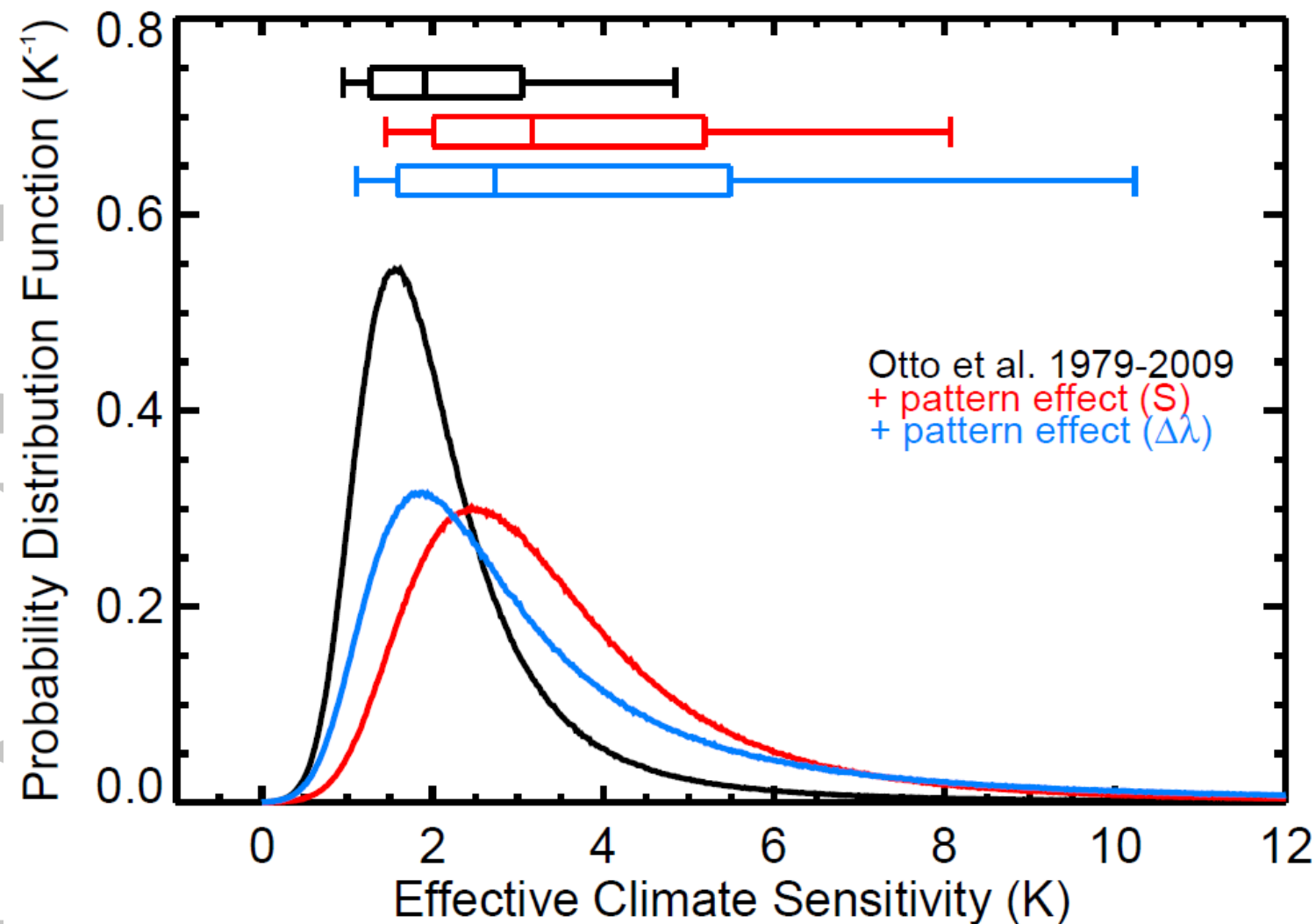


Figure 3: Comparison of the EffCS probability distribution function from a historical energy budget constraint (Otto et al, 2013), before (black) and after (colours) accounting for the pattern effect between historical climate change and abrupt-4xCO₂. ‘Red’ accounts for the pattern effect by scaling the historical feedback parameter λ_{hist} by the ratio ($S=\lambda_{4xCO_2}/\lambda_{amip}$) of the feedbacks found in the *amip-piForcing* and *abrupt-4xCO₂* simulations. ‘Blue’ accounts for the pattern effect by adding the difference in feedbacks ($\Delta\lambda=\lambda_{4xCO_2}-\lambda_{amip}$) to λ_{hist} (see Section 4 and Table 1). Box plots show the 5-95% confidence interval (end bars), the 17-83% confidence interval (box ends) and the median (line in box).

**Andrews et al. (2018), *GRL*
doi:10.1029/2018GL078887**

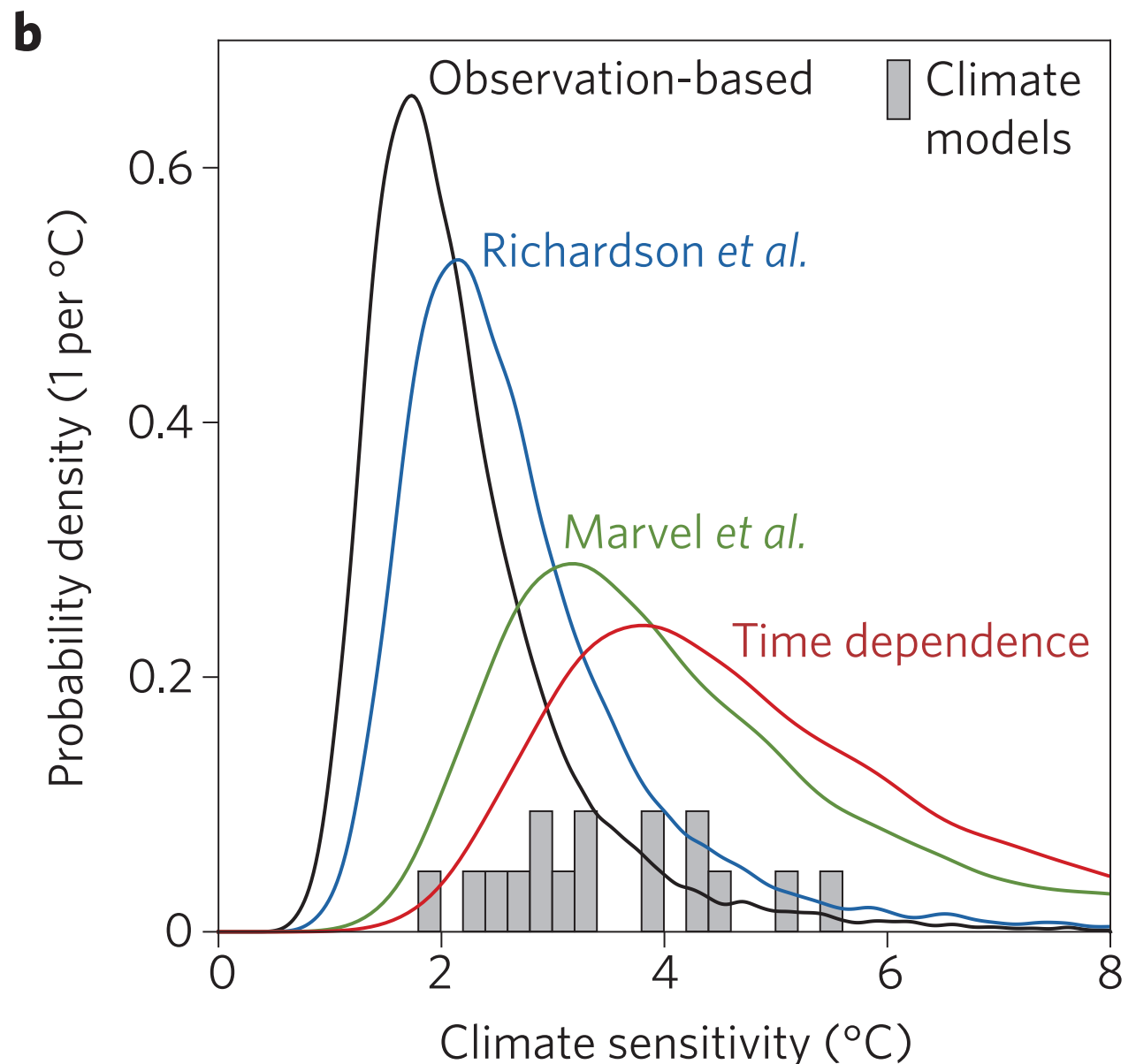
Making best use of Earth observations to infer climate sensitivity... is a work in progress!

The simplest budget-based estimates are almost surely biased low — for a variety of reasons!

Almost every serious attempt to correct these estimates using detailed knowledge of the workings of the climate system yield higher sensitivities.

There is no compelling evidence that current climate models are too sensitive.

(also... we cannot rule out very high sensitivity through these arguments)



**Armour (2016), *Nature Clim. Change*
doi:10.1038/nclimate3079**

Conclusion: *learning the right lessons about the future from the past*

- Climate change was/will be a combination of **forced warming** (more predictable) and complex regional variability (less predictable)
- Recent observations tell us about one (only) out of many possible histories
- There is (unfortunately) **no such thing as a measurement of climate sensitivity** — only fitting observations to climate models!
- **Naively fitting** to the **simplest model** of the global energy balance leads us astray in several ways:
 - **Pattern effects:** parts of the world that warm the slowest (East Pacific, Southern Ocean) are (coincidentally?) also regions associated with strongly amplifying feedbacks (clouds and ice)
 - Sensitivity **inferred from any short-term period of warming** is likely to be **biased low**. Future loss of marine cloud will give additional warming.
 - The “**hiatus**” **period of the 2000s** is especially problematic! Anomalous cooling of East Pacific really exacerbated this low bias. This period was simply **not a good proxy for future global warming**.
 - Other ongoing work: teasing apart effects of **greenhouse gases** from **aerosols** on past warming, biases in global temperature datasets, etc etc etc.
- Detailed studies, informed by nuanced physical understanding of climatic processes, show that **a lukewarm future** (e.g. sensitivity < 2°C) **is very unlikely**.

Extra slides

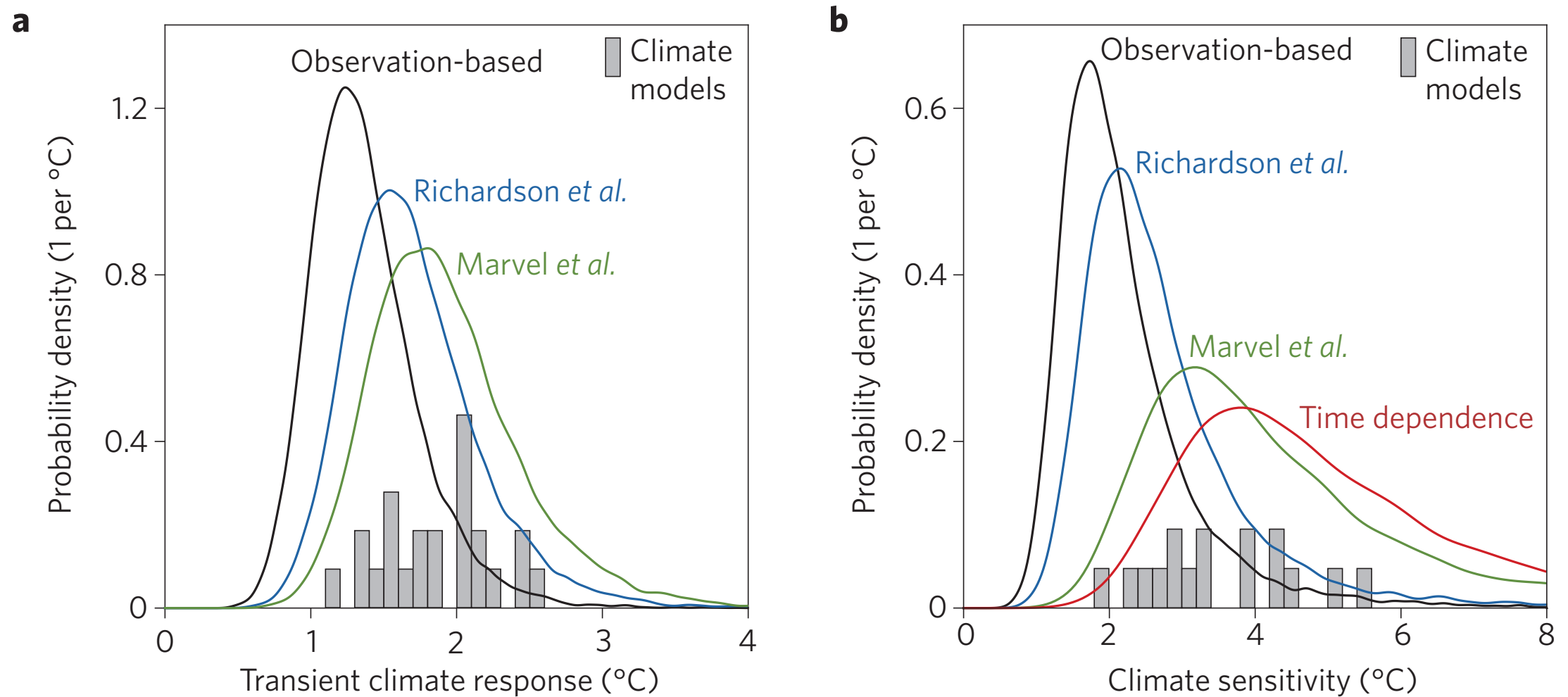
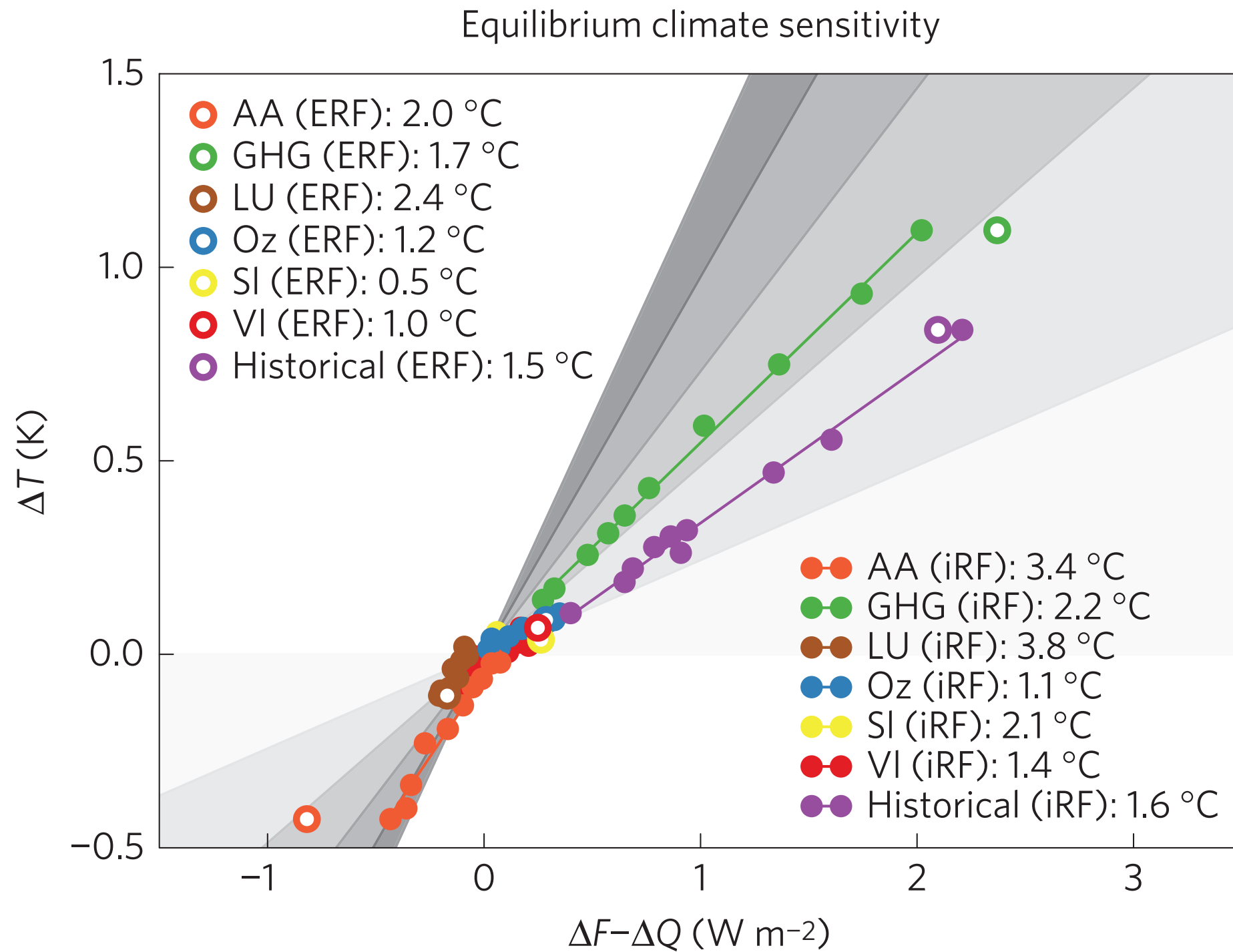
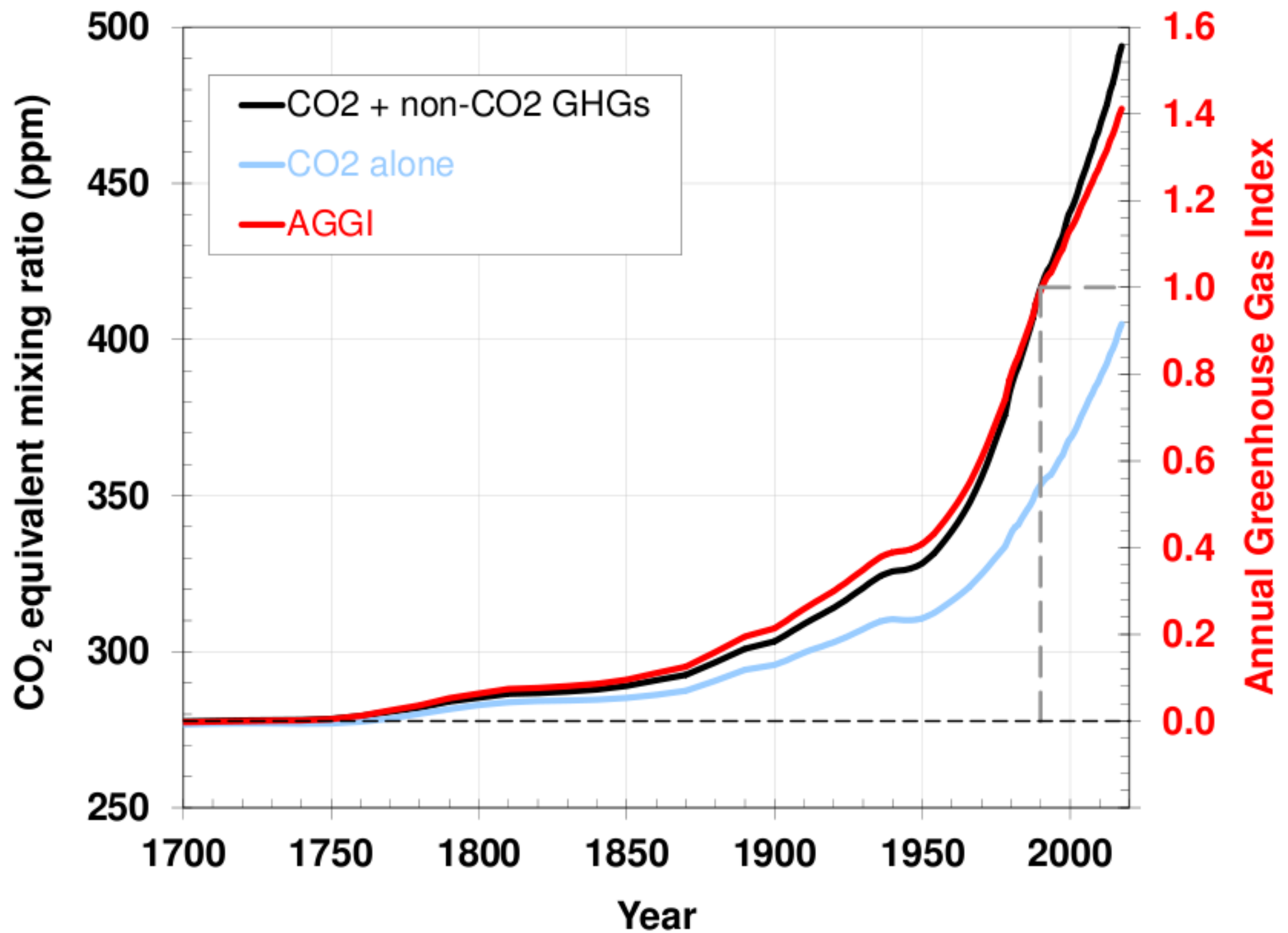


Figure 1 | Probability distribution of climate response to forcings. **a**, Transient climate response estimated from observations¹ (black), and its revision following Richardson *et al.*³ (blue) then following Marvel *et al.*⁶ (green). **b**, As with **a** but for climate sensitivity, with an additional revision for climate sensitivity appearing smaller than its true value⁷⁻¹¹ (red). Histogram of climate model values shown in grey.

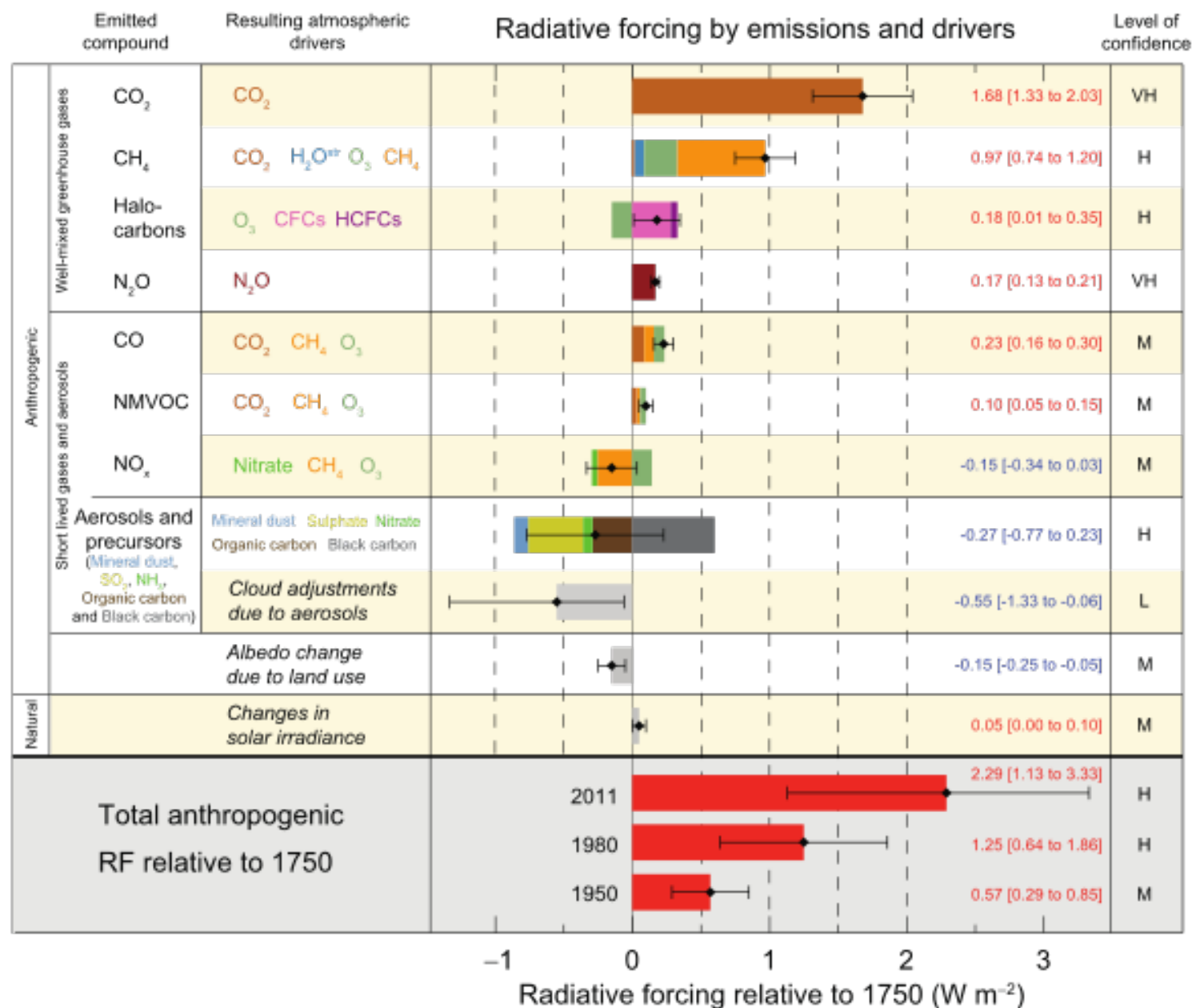
**Armour (2016), *Nature Clim. Change*
doi:10.1038/nclimate3079**



**Marvel et al. (2015), *Nature Clim. Change*
doi:10.1038/NCLIMATE2888**



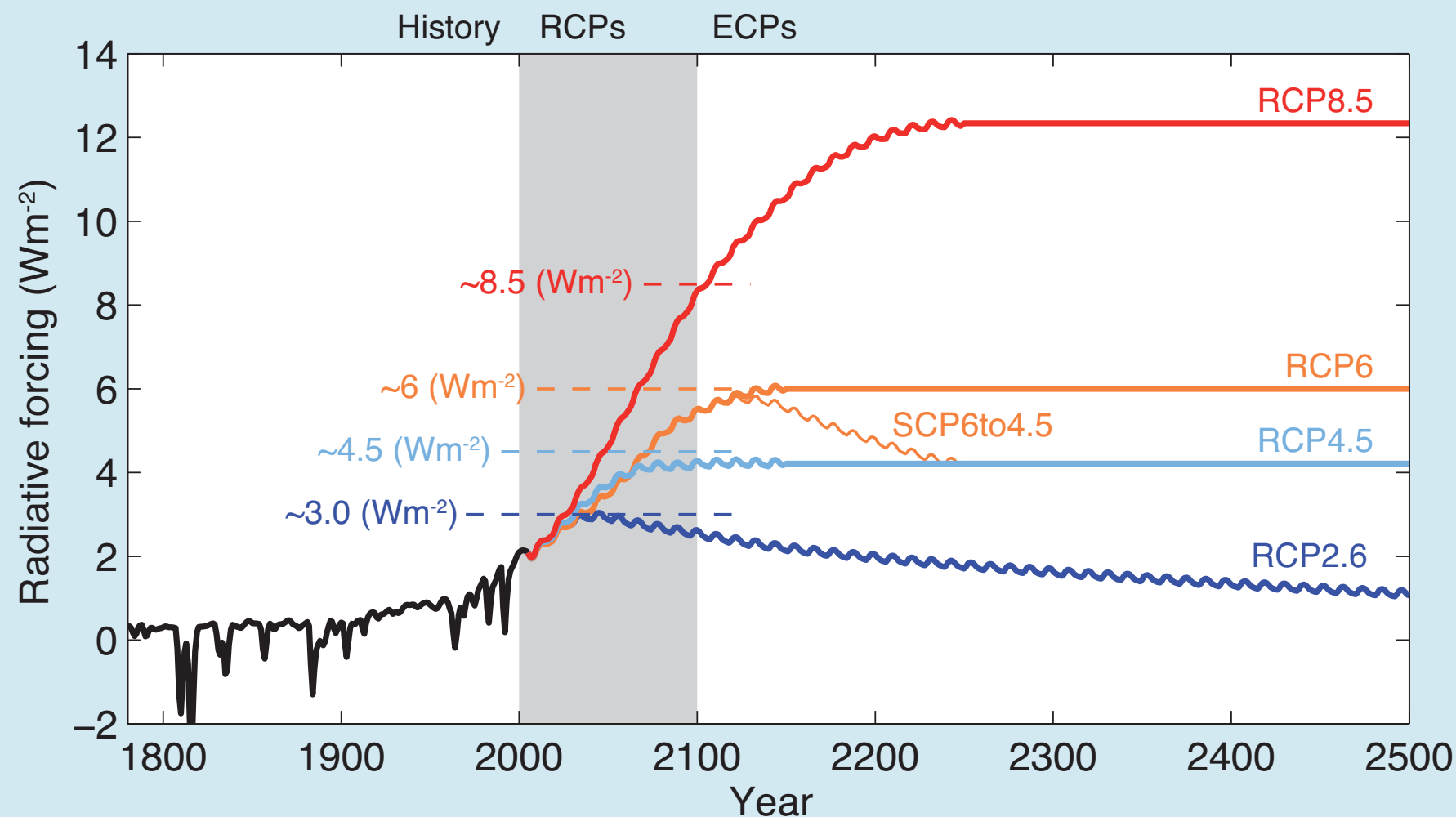
“Radiative forcing”: sources of energy to the climate system



The total energy input into the climate system by human activities over the industrial era is about 2.3 W m⁻².

CO₂ from fossil fuel combustion is the single largest culprit.

Figure SPM.5 | Radiative forcing estimates in 2011 relative to 1750 and aggregated uncertainties for the main drivers of climate change. Values are global average radiative forcing (RF¹⁴), partitioned according to the emitted compounds or processes that result in a combination of drivers. The best estimates of the net radiative forcing are shown as black diamonds with corresponding uncertainty intervals; the numerical values are provided on the right of the figure, together with the confidence level in the net forcing (VH – very high, H – high, M – medium, L – low, VL – very low). Albedo forcing due to black carbon on snow and ice is included in the black carbon aerosol bar. Small forcings due to contrails (0.05 W m⁻², including contrail induced cirrus), and HFCs, PFCs and SF₆ (total 0.03 W m⁻²) are not shown. Concentration-based RFs for gases can be obtained by summing the like-coloured bars. Volcanic forcing is not included as its episodic nature makes it difficult to compare to other forcing mechanisms. Total anthropogenic radiative forcing is provided for three different years relative to 1750. For further technical details, including uncertainty ranges associated with individual components and processes, see the Technical Summary Supplementary Material. [8.5; Figures 8.14–8.18; Figures TS.6 and TS.7]



Box 1.1, Figure 1 | Total RF (anthropogenic plus natural) for RCPs and extended concentration pathways (ECP)—for RCP2.6, RCP4.5, and RCP6, RCP8.5, as well as a supplementary extension RCP6 to 4.5 with an adjustment of emissions after 2100 to reach RCP4.5 concentration levels in 2250 and thereafter. Note that the stated RF levels refer to the illustrative default median estimates only. There is substantial uncertainty in current and future RF levels for any given scenario. Short-term variations in RF are due to both volcanic forcings in the past (1800–2000) and cyclical solar forcing assuming a constant 11-year solar cycle (following the CMIP5 recommendation), except at times of stabilization. (Reproduced from Figure 4 in Meinshausen et al., 2011.)

Global mean temperature changes

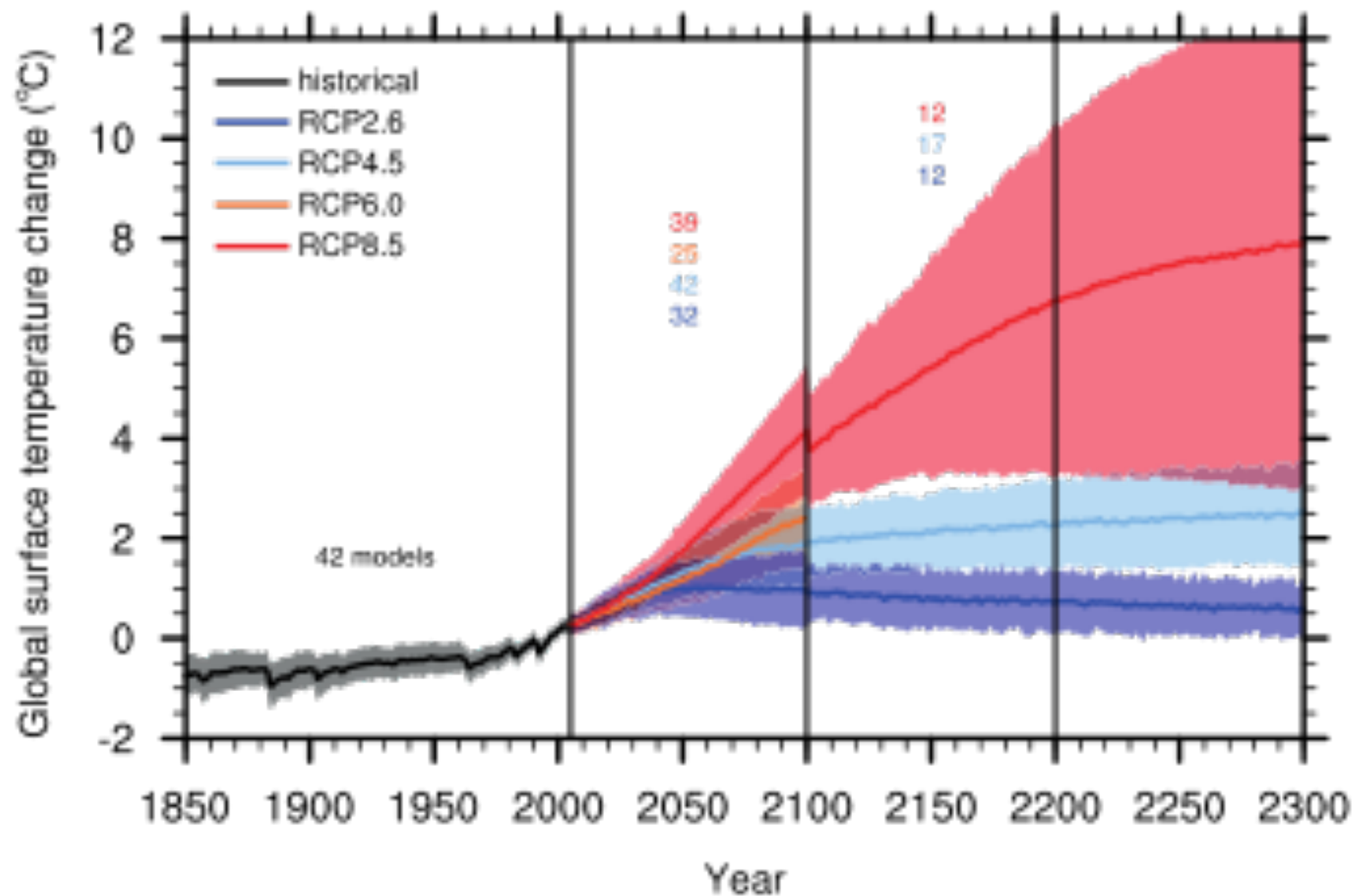


Figure 12.5 | Time series of global annual mean surface air temperature anomalies (relative to 1986–2005) from CMIP5 concentration-driven experiments. Projections are

Annual mean surface air temperature change (RCP4.5: 2081-2100)



42 different possible futures
under one particular
radiative forcing scenario

Figure 12.9 | Surface air temperature change in 2081–2100 displayed as anomalies with respect to 1986–2005 for RCP4.5 from one ensemble member of each of the concentration-driven models available in the CMIP5 archive.

50-year warming trends simulated in a single climate model...
each one of these scenarios is equally likely!

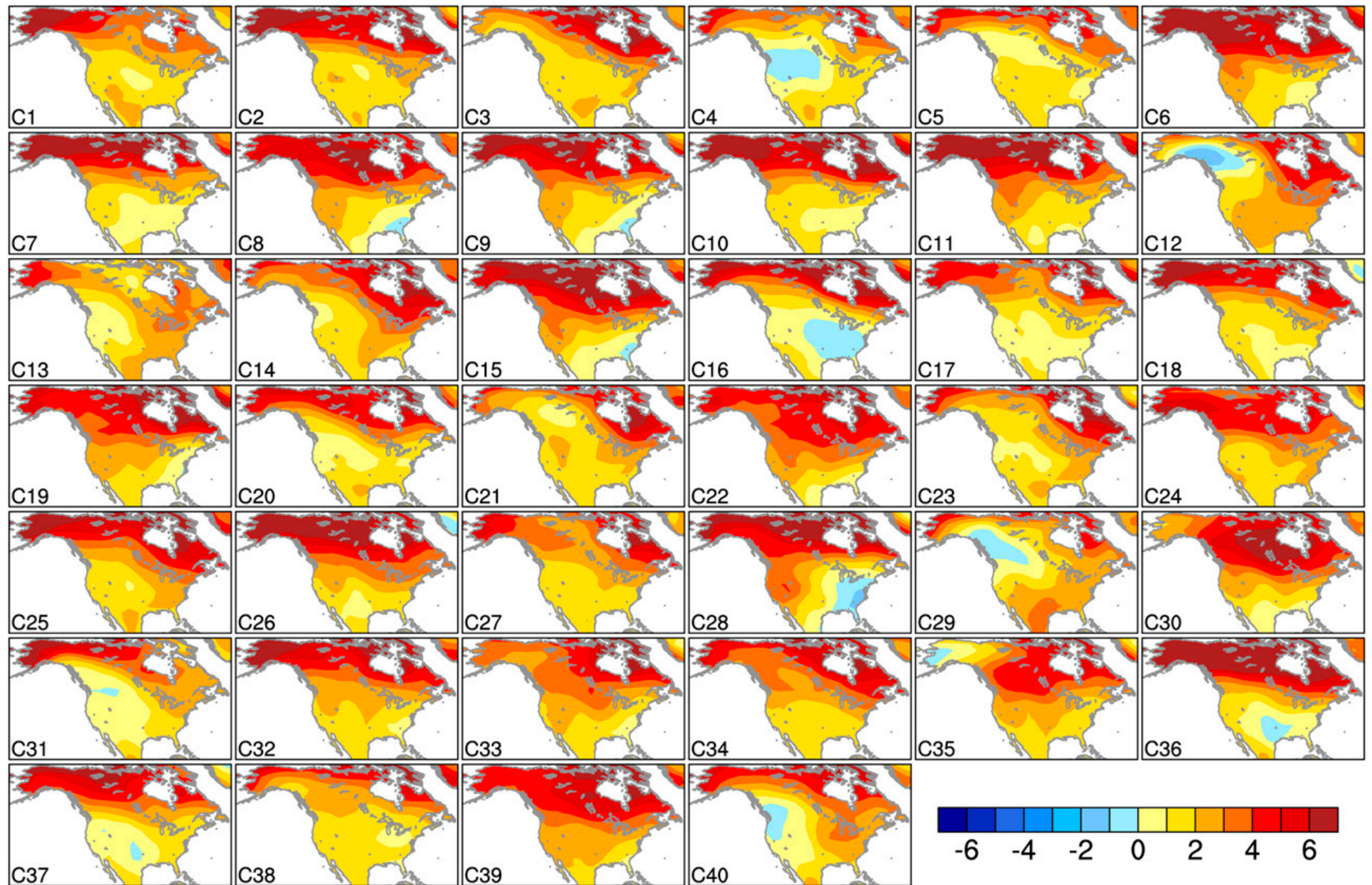


FIG. 1. Winter SAT trends [2010–60; $^{\circ}\text{C} (51 \text{ yr})^{-1}$] from each of the 40 CCSM3 ensemble members.

Deser et al. (2014), *J. Climate*
doi:10.1175/JCLI-D-13-00451.s1

Averaging over all these 40 simulations gives the “forced warming” — the part of the future change attributable to increased greenhouse gases

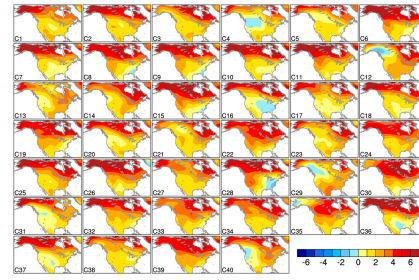
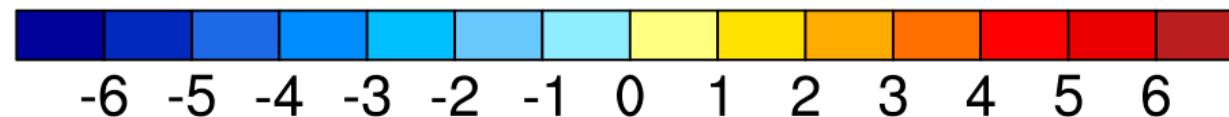
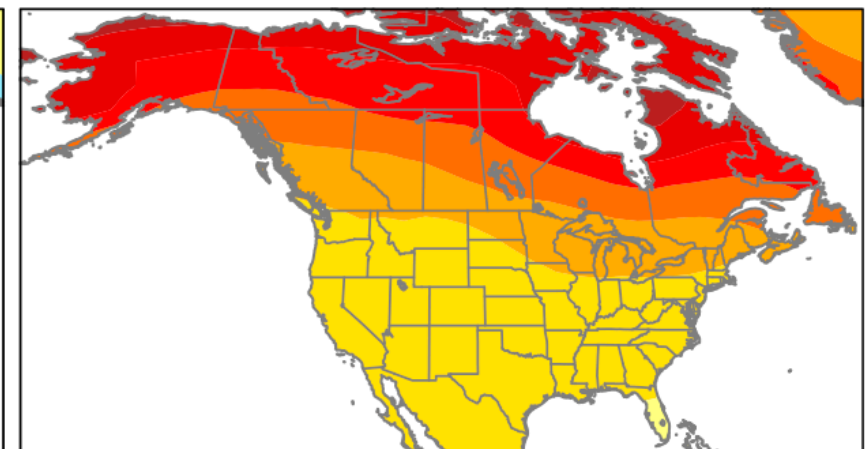
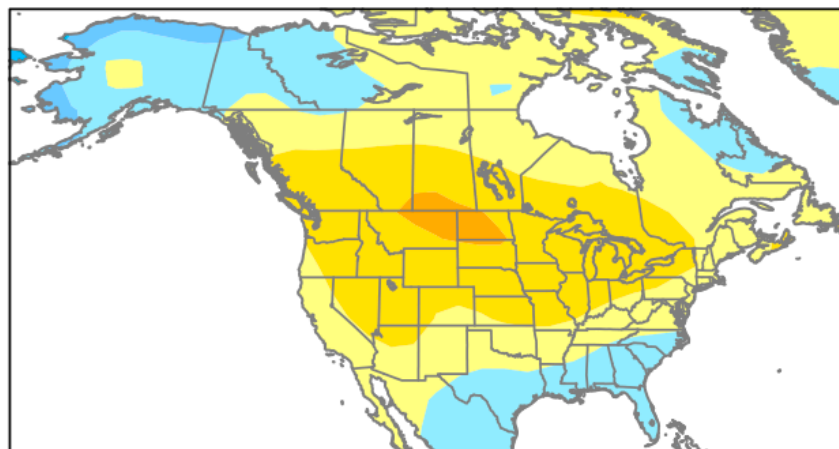
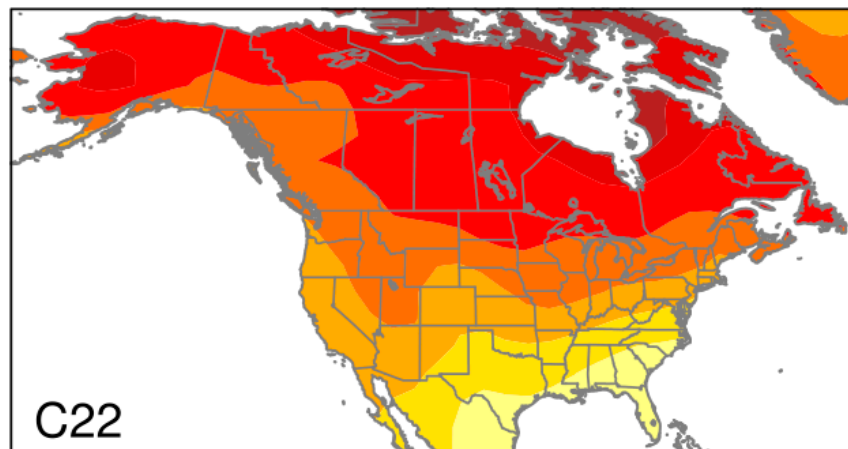
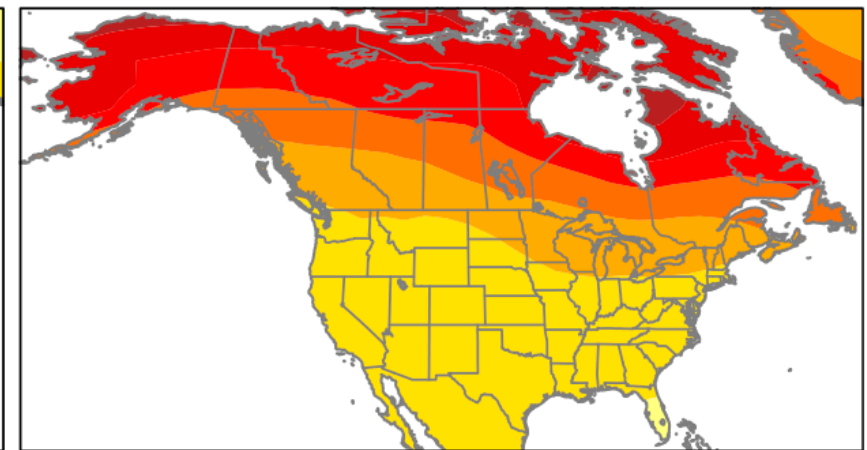
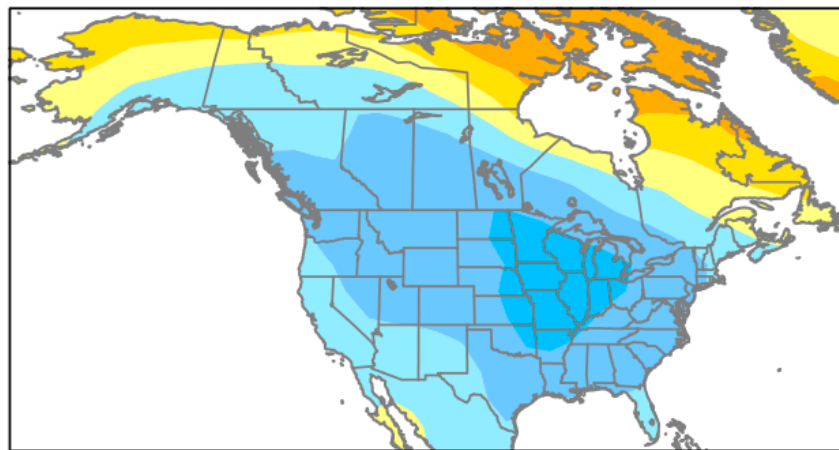
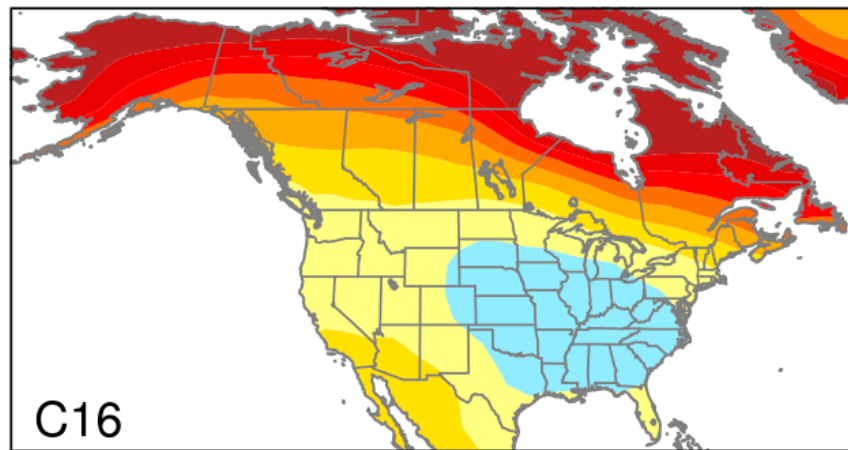


FIG. 1. Winter SAT trends [2010–60 °C (°F) yr⁻¹] from each of the 40 CCSM ensemble members.

a) Winter SAT Trends
Total

Internal

Forced



Deser et al. (2014), *J. Climate*
doi:10.1175/JCLI-D-13-00451.s1