

Using historical climate change to estimate equilibrium warming

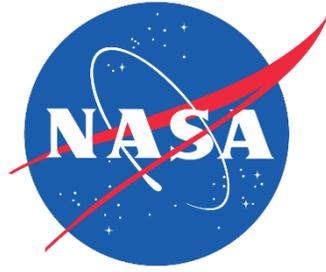
Mark Richardson^{1,*}, Kevin Cowtan², Martin B Stolpe³, Ed Hawkins⁴

¹NASA JPL, Caltech, UCLA JIFRESSE

²University of York, UK ³ETH Zurich ⁴University of Reading, UK

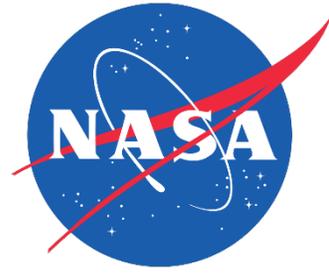
CIRA CSU seminar

*markr@jpl.caltech.edu



Equilibrium Climate Sensitivity (ECS)

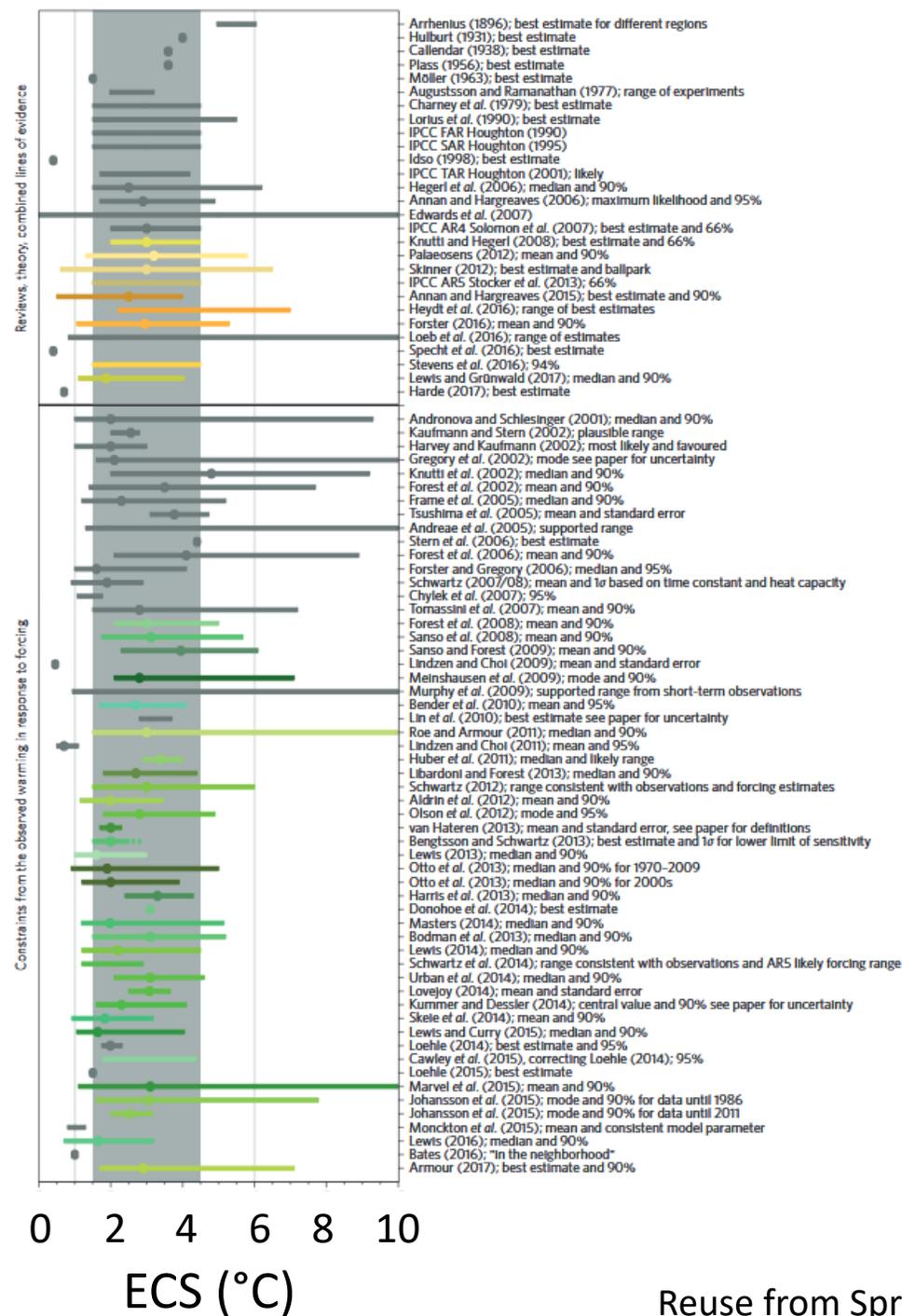
- ECS is eventual warming from doubled atmospheric CO₂
- Intergovernmental Panel on Climate Change (IPCC) 2013 report said range 1.5—4.5 °C with a best estimate of 3 °C
- Let's quickly summarise evidence then investigate the “historical energy budget” calculation approach.

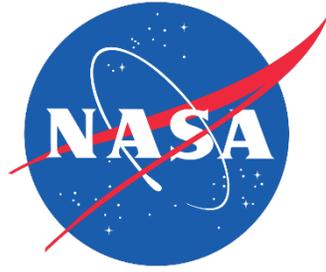


Don't look at details!

Each bar is a study.

Most cover the shaded
1.5—4.5 °C.





Paleoclimate examples:

Look at what happened in the deep past.

This isn't my business, but it's cool.

Climate Sensitivity Estimated from Temperature Reconstructions of the Last Glacial Maximum

Andreas Schmittner,^{1*} Nathan M. Urban,² Jeremy D. Shakun,³ Natalie M. Mahowald,⁴ Peter U. Clark,⁵ Patrick J. Bartlein,⁶ Alan C. Mix,¹ Antoni Rosell-Melé⁷

Assessing the impact of future anthropogenic carbon emissions is currently impeded by uncertainties in our knowledge of equilibrium climate sensitivity to atmospheric carbon dioxide doubling. Previous studies suggest 3 kelvin (K) as the best estimate, 2 to 4.5 K as the 66% probability range, and nonzero probabilities for much higher values, the latter implying a small chance of high-impact climate changes that would be difficult to avoid. Here, combining extensive sea and land surface temperature reconstructions from the Last Glacial Maximum with climate model simulations, we estimate a lower median (2.3 K) and reduced uncertainty (1.7 to 2.6 K as the 66% probability range, which can be widened using alternate assumptions or data subsets). Assuming that paleoclimatic constraints apply to the future, as predicted by our model, these results imply a lower probability of imminent extreme climatic change than previously thought.

PERSPECTIVE

doi:10.1038/nature11574

Making sense of palaeoclimate sensitivity

PALAEOSSENS Project Members*

Many palaeoclimate studies have quantified pre-anthropogenic climate change to calculate climate sensitivity (equilibrium temperature change in response to radiative forcing change), but a lack of consistent methodologies produces a wide range of estimates and hinders comparability of results. Here we present a stricter approach, to improve intercomparison of palaeoclimate sensitivity estimates in a manner compatible with equilibrium projections for future climate change. Over the past 65 million years, this reveals a climate sensitivity (in $K W^{-1} m^2$) of 0.3–1.9 or 0.6–1.3 at 95% or 68% probability, respectively. The latter implies a warming of 2.2–4.8 K per doubling of atmospheric CO_2 , which agrees with IPCC estimates.

SCIENCE ADVANCES | RESEARCH ARTICLE

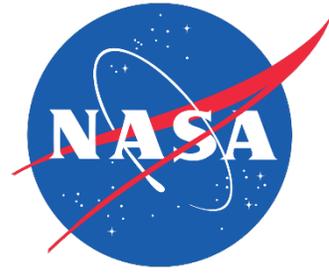
CLIMATE CHANGE

Nonlinear climate sensitivity and its implications for future greenhouse warming

Tobias Friedrich,^{1*} Axel Timmermann,¹ Michelle Tigheelaar,² Oliver Elison Timm,³ Andrey Ganopolski⁴

Global mean surface temperatures are rising in response to anthropogenic greenhouse gas emissions. The magnitude of this warming at equilibrium for a given radiative forcing—referred to as specific equilibrium climate sensitivity (S)—is still subject to uncertainties. We estimate global mean temperature variations and S using a 784,000-year-long field reconstruction of sea surface temperatures and a transient paleoclimate model simulation. Our results reveal that S is strongly dependent on the climate background state, with significantly larger values attained during warm phases. Using the Representative Concentration Pathway 8.5 for future greenhouse radiative forcing, we find that the range of paleo-based estimates of Earth's future warming by 2100 CE overlaps with the upper range of climate simulations conducted as part of the Coupled Model Intercomparison Project Phase 5 (CMIP5). Furthermore, we find that within the 21st century, global mean temperatures will very likely exceed maximum levels reconstructed for the last 784,000 years. On the basis of temperature data from eight glacial cycles, our results provide an independent validation of the magnitude of current CMIP5 warming projections.

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“Last Glacial Maximum” (cold period 19—23,000 years ago):

2.3 °C

“past 65 million years” :

2.2—4.8 °C

“784,000-year-long field reconstruction”:

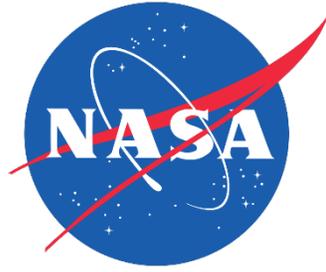
3.2 °C (full period),

4.8 °C (warm periods, like now)

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Bottom: Creative Commons Attribution License

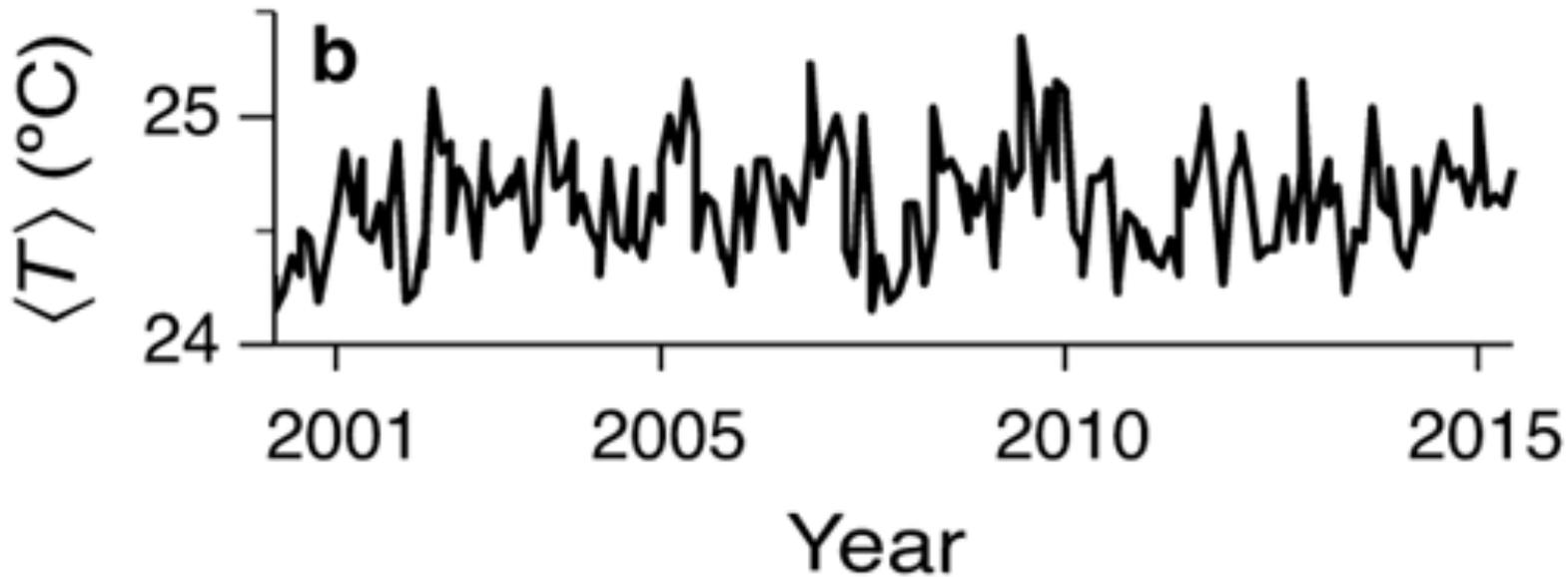
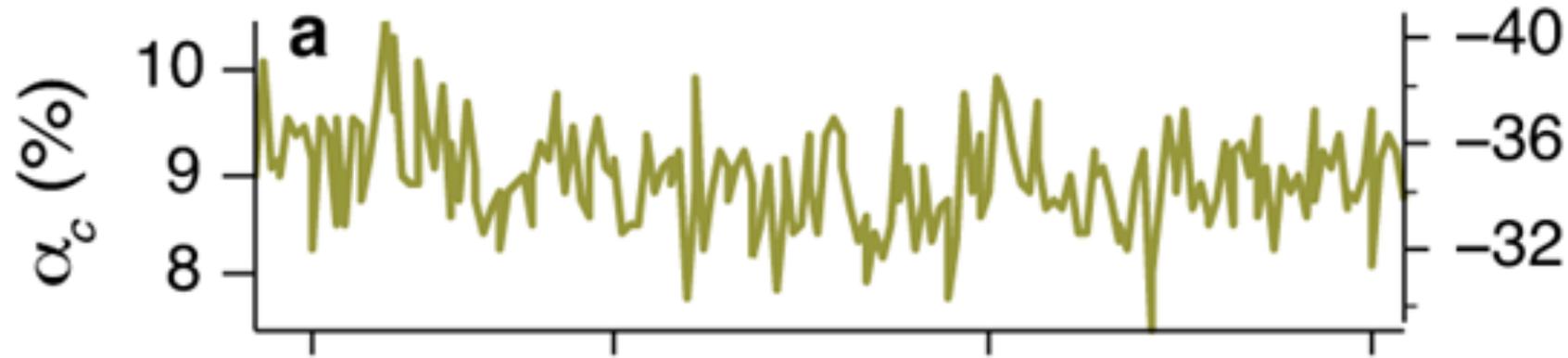
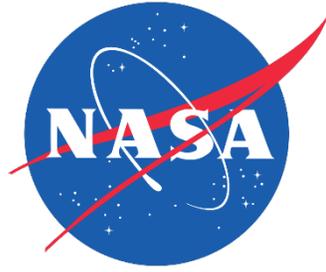


Methods using more modern data include
“emergent constraints”

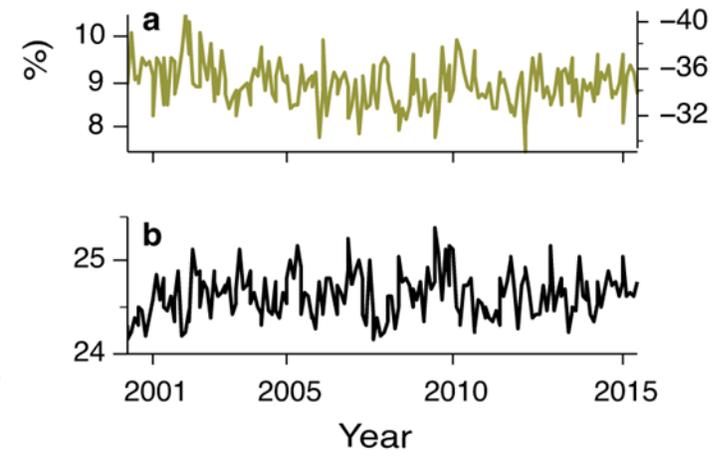
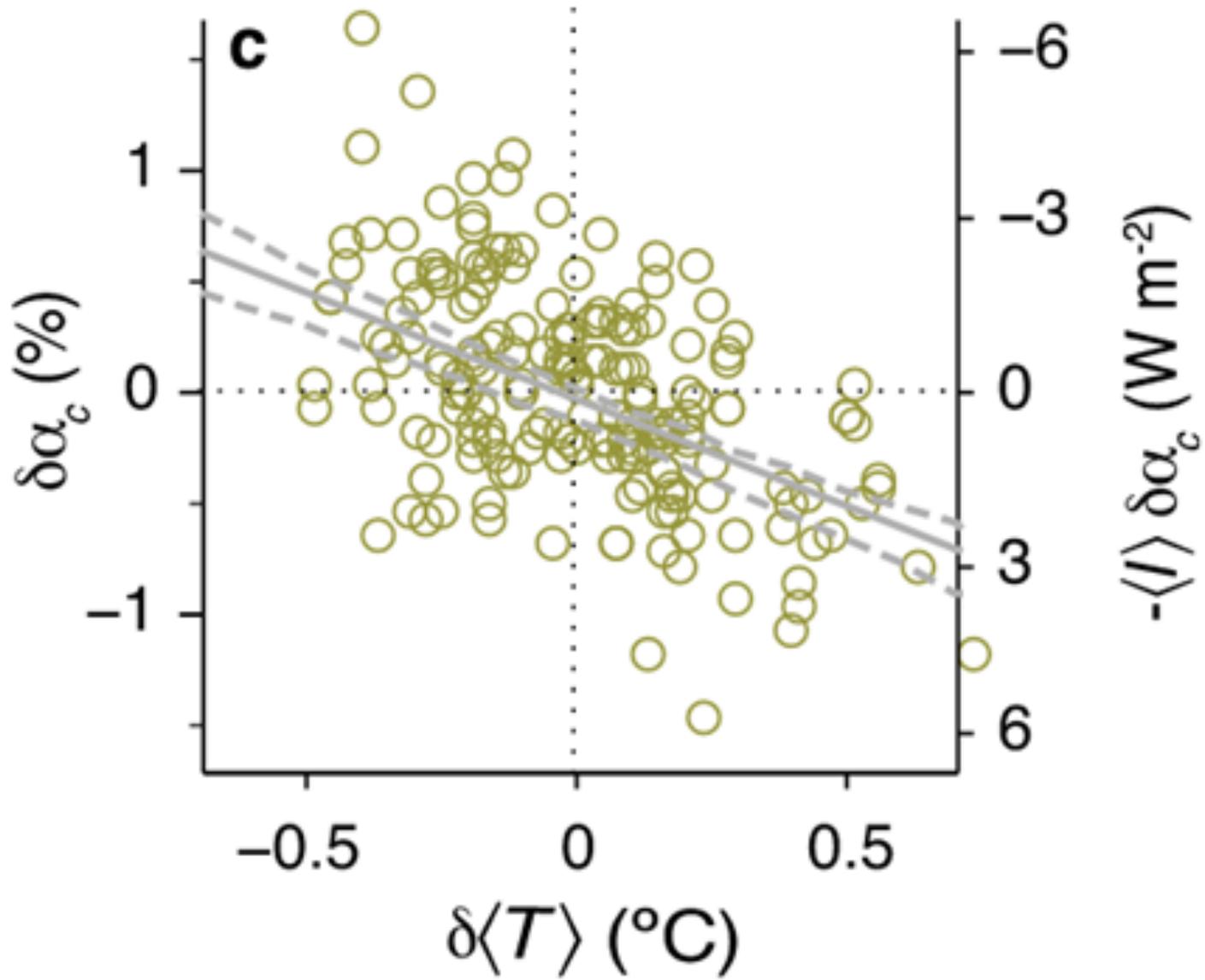
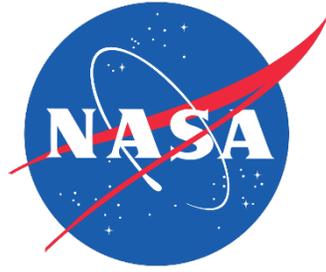
Find a process that relates to climate sensitivity
and constrain it with observations.

Using Brient & Schneider (2016, doi: [10.1175/JCLI-D-15-0897.1](https://doi.org/10.1175/JCLI-D-15-0897.1))

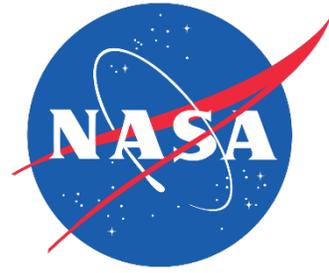
Tropical low cloud cover (CERES)



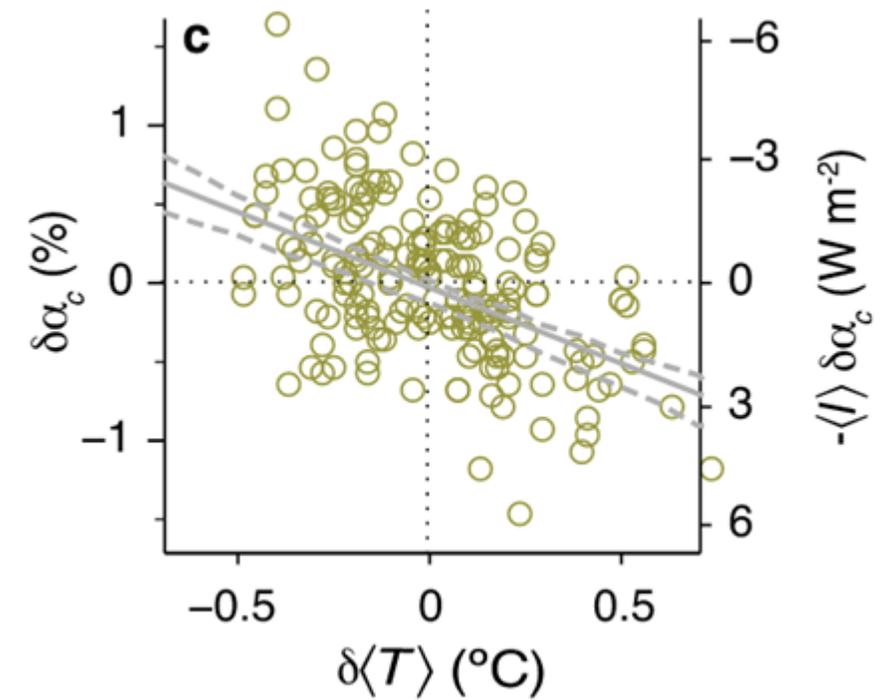
Deseasonalised SST
(ERSST)

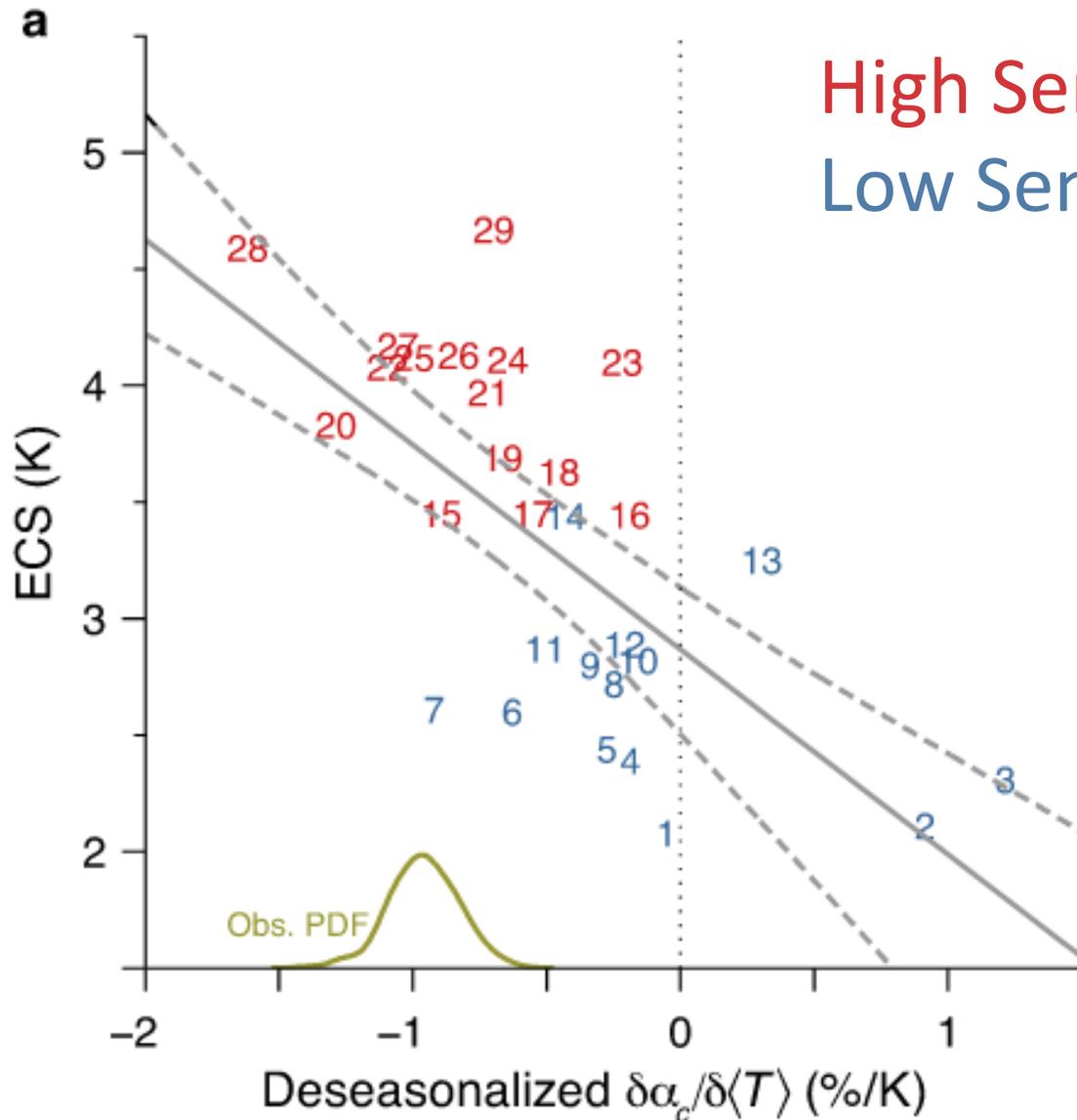
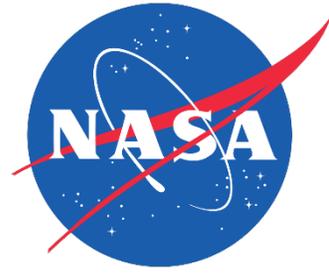


Warmer T corresponds to less low cloud reflection.

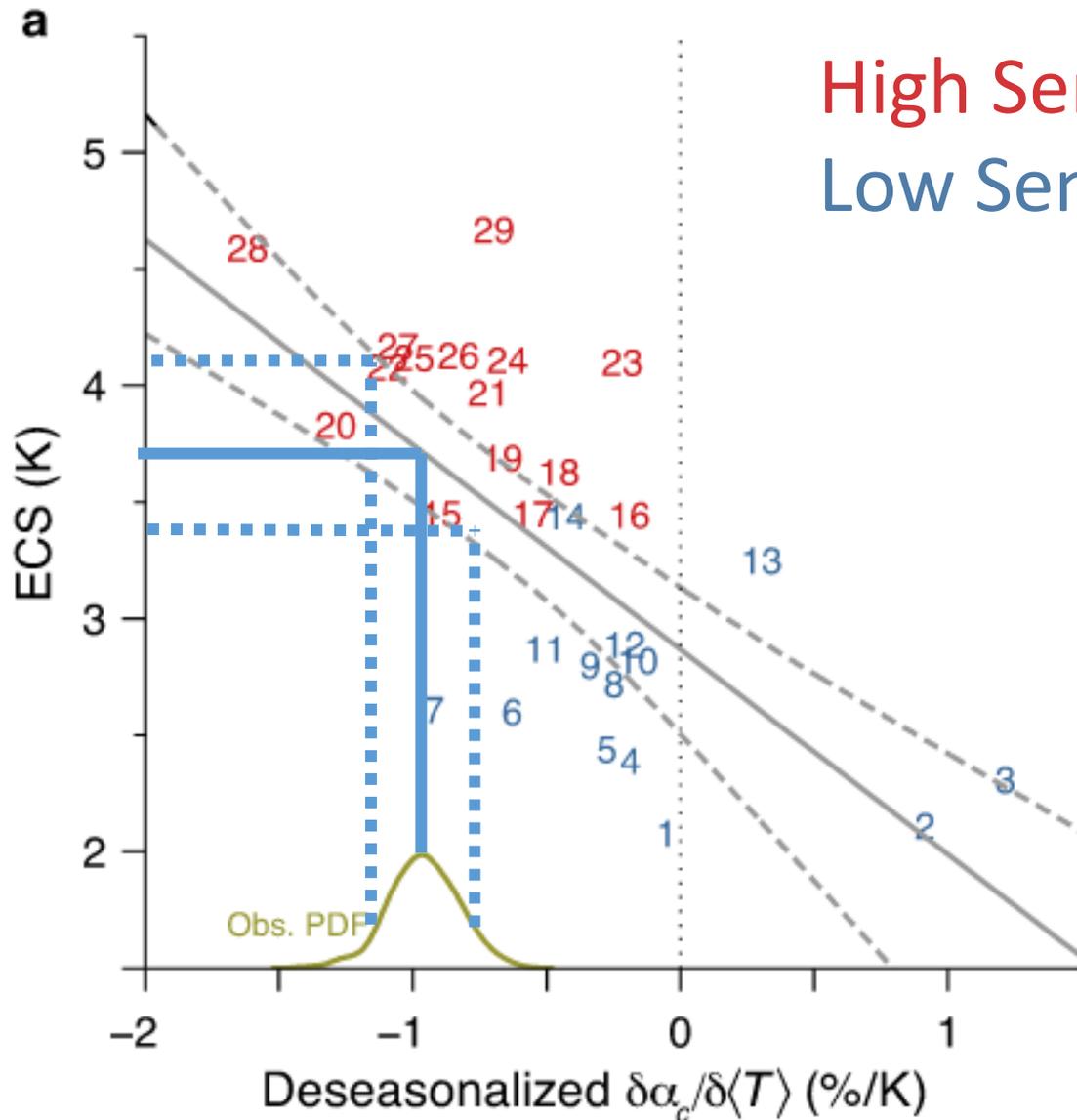
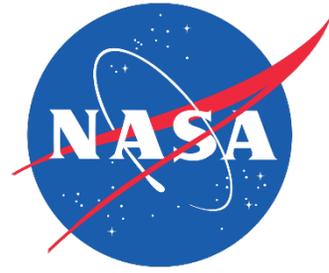


- From observations we have a probability distribution of: $\frac{\delta\alpha_c}{\delta\langle T\rangle}$
- See if this relates to ECS in CMIP5 climate models



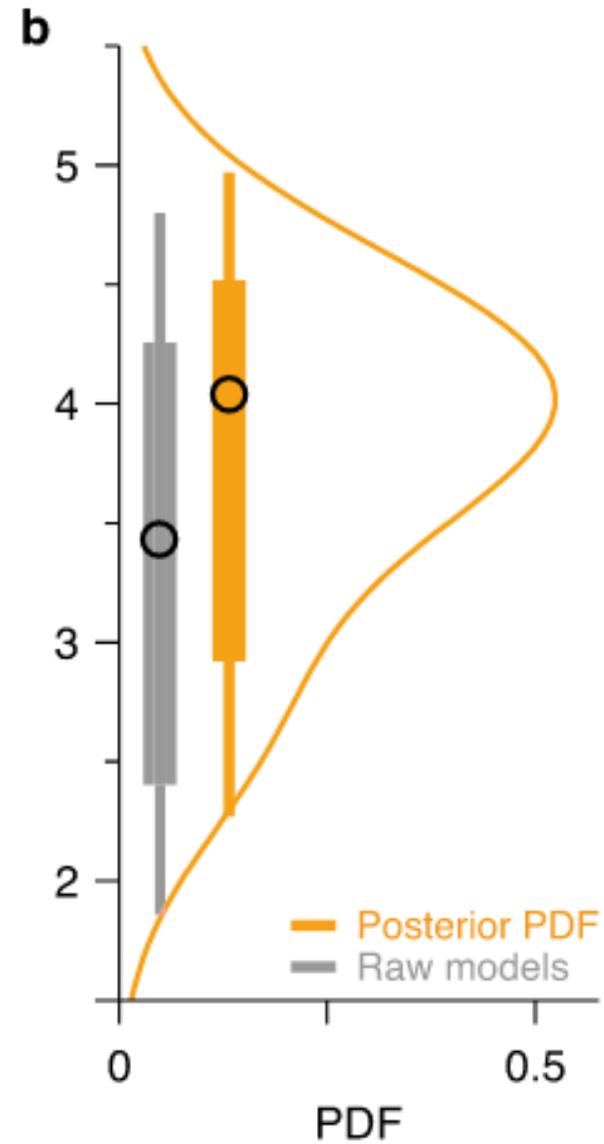
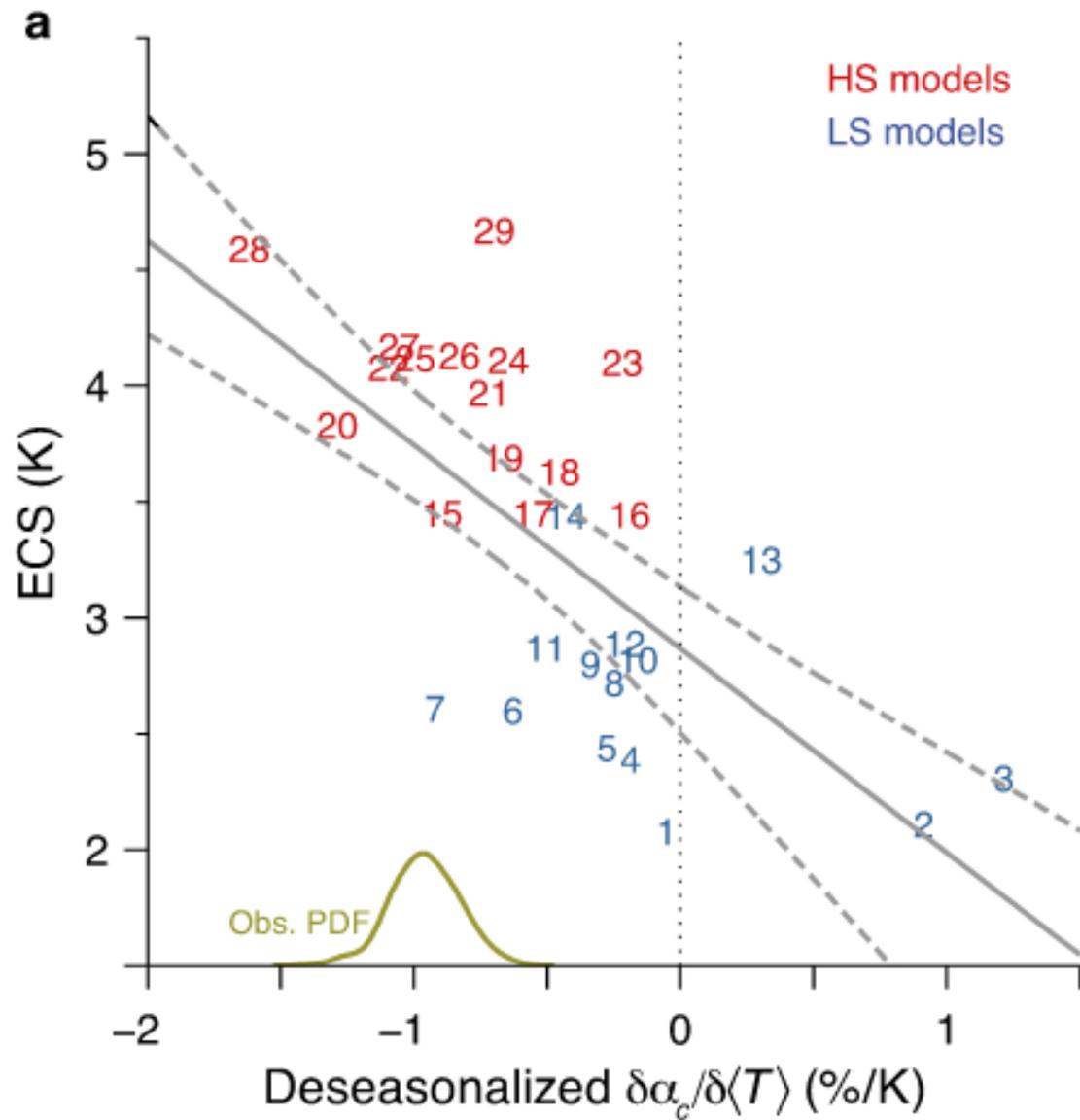
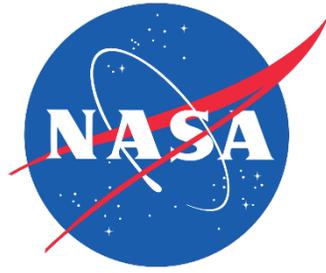


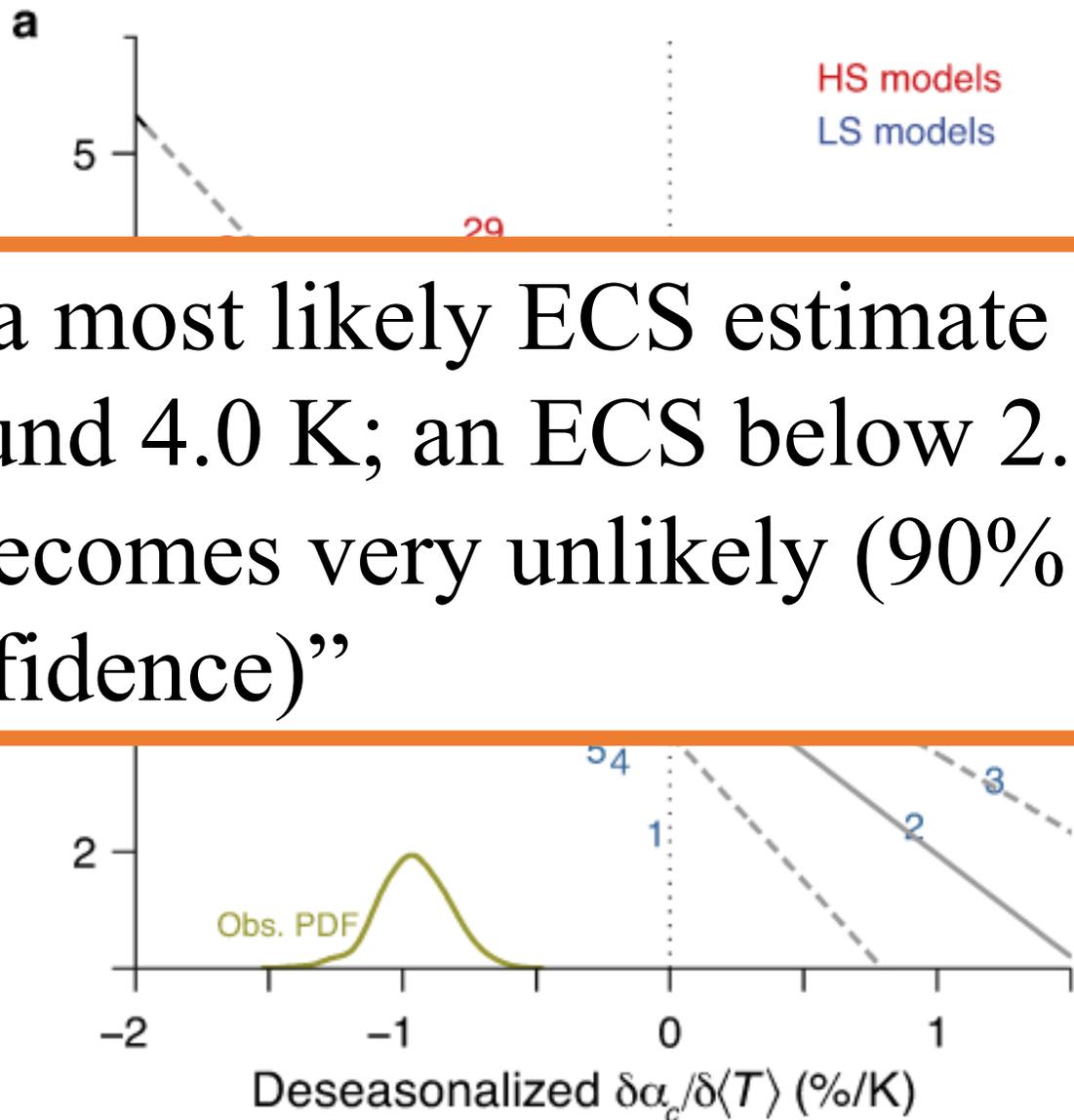
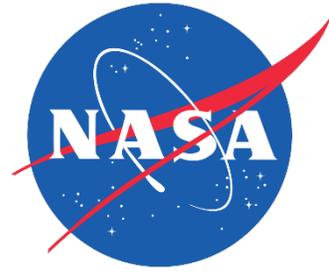
ECS is bigger when low clouds retreat a lot under warming.



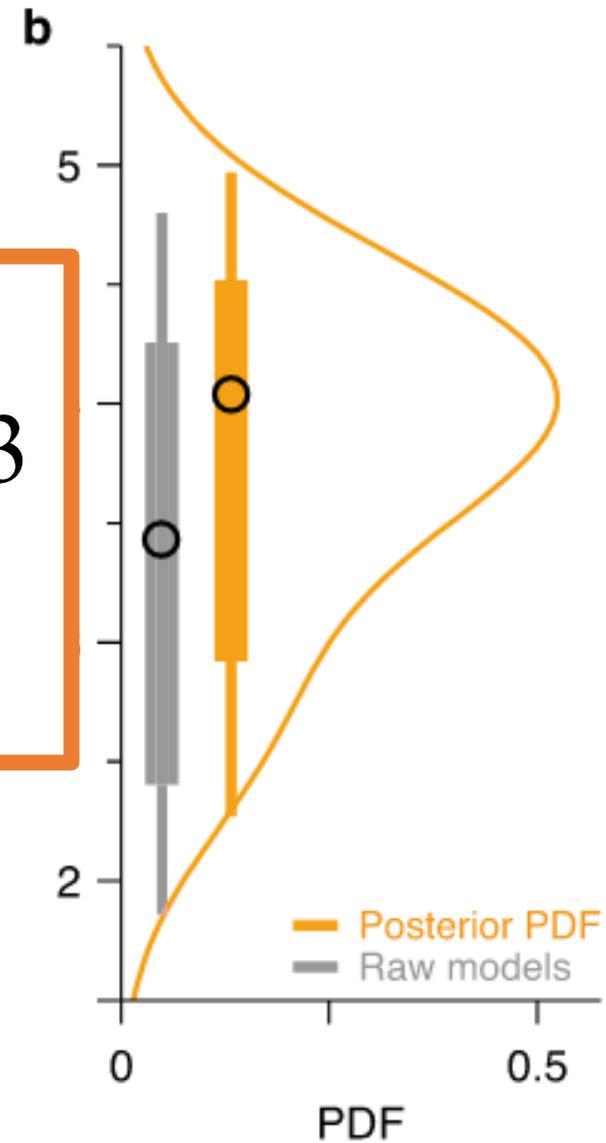
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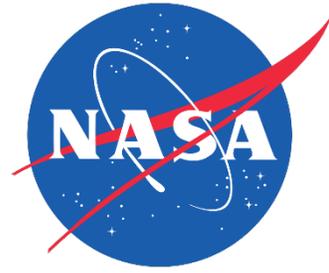
Link observed retreat to ECS using model fit with Bayes' theorem





“...a most likely ECS estimate around 4.0 K; an ECS below 2.3 K becomes very unlikely (90% confidence)”





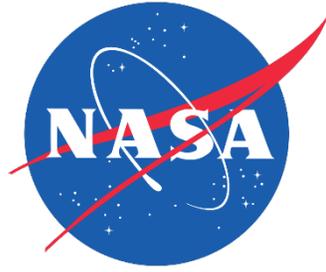
Summaries of emergent constraints:

Caldwell et al. (2018), doi: [10.1175/JCLI-D-17-0631.1](https://doi.org/10.1175/JCLI-D-17-0631.1)

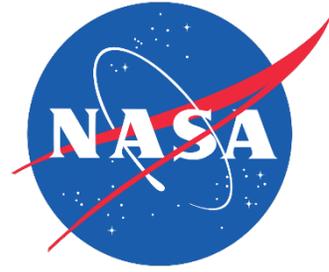
Qu et al. (2018), doi: [10.1175/JCLI-D-17-0482.1](https://doi.org/10.1175/JCLI-D-17-0482.1)

Lots, they have some issues, but usually $ECS_{\text{emergent}} > 3 \text{ }^{\circ}\text{C}$.

Cox et al. (2018, doi: [10.1038/nature25450](https://doi.org/10.1038/nature25450)) on low end from variability of historical temperatures: $2.8 \text{ }^{\circ}\text{C}$.



- Lots of ECS studies with different techniques
- Paleo spans the range
- Physics plus observation “emergent constraints” favour $ECS > 3 \text{ }^\circ\text{C}$
- What about the low end?



Linear model of Earth response

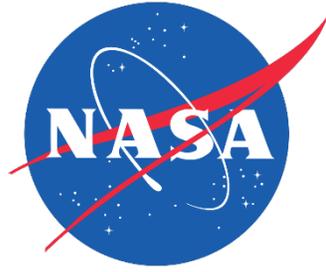
$$\Delta N(t) = \Delta F(t) - \lambda(t)\Delta T(t)$$

ΔN = Change in Earth's net heat imbalance (W m^{-2})

ΔF = Radiative forcing (W m^{-2})

λ = Feedback parameter ($\text{W m}^{-2} \text{K}^{-1}$)

ΔT = Change in temperature ($^{\circ}\text{C}$)

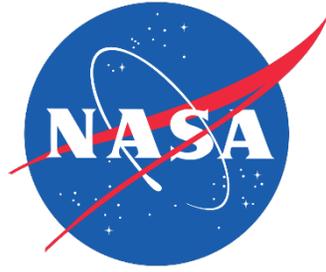


Linear model of Earth response

$$\Delta N(t) = \Delta F(t) - \lambda(t)\Delta T(t)$$

Assume:

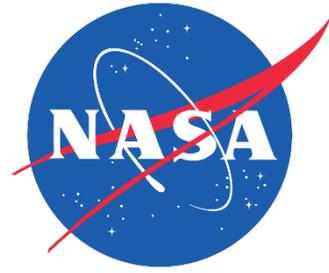
$$\lambda(t) = \lambda$$
$$\Delta N_{eq} = \Delta N(t \rightarrow \infty) = 0$$



Linear model of Earth response

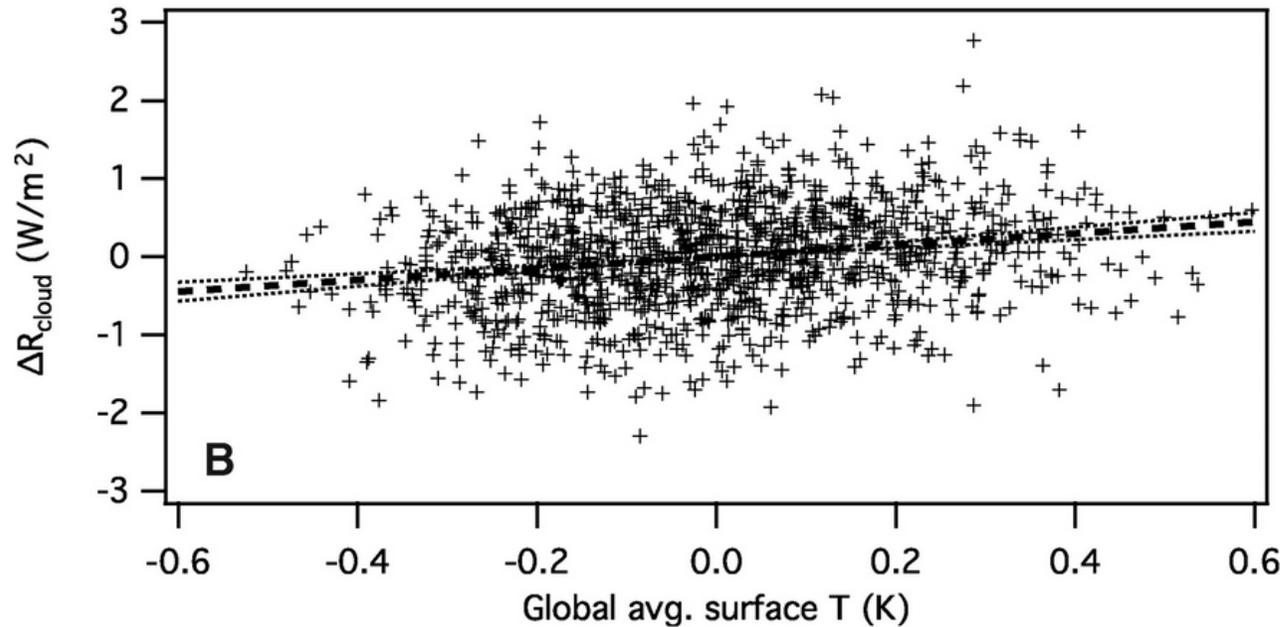
$$\lambda = \frac{\Delta F(t) - \Delta N(t)}{\Delta T(t)}$$

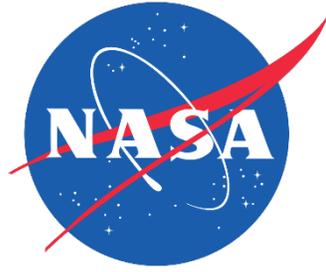
$$ECS = \Delta T_{eq,2\times CO_2} = \frac{\Delta F_{2\times CO_2}}{\lambda}$$



$$\lambda = \frac{\Delta F(t) - \Delta N(t)}{\Delta T(t)}$$

Dessler (2010): cloud feedback “likely positive”
ECS > 2 °C





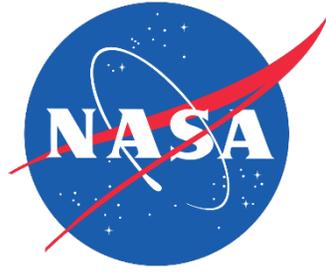
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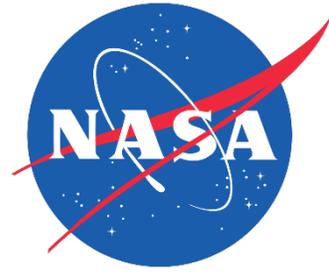
BUT THIS EQUATION IS THE SOURCE OF SEVERAL
LOW ECS ESTIMATES...

We must understand *why* some methods give
higher/lower values to better estimate ECS.

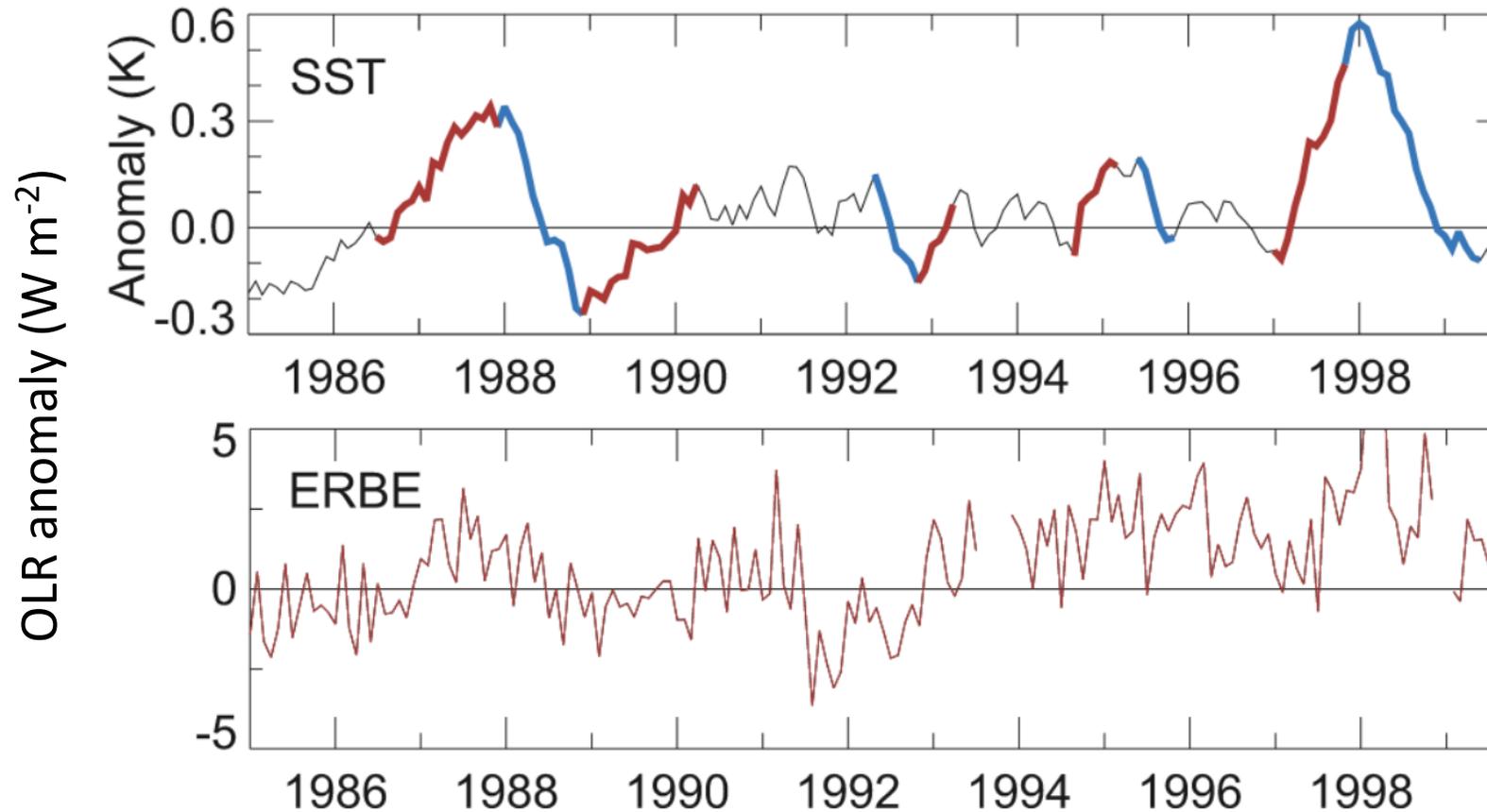
Lindzen & Choi (2009):

ECS is much lower than in models



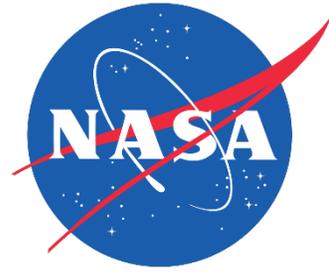


Lindzen and Choi (2009) picked ΔT and ΔN for the below coloured periods:

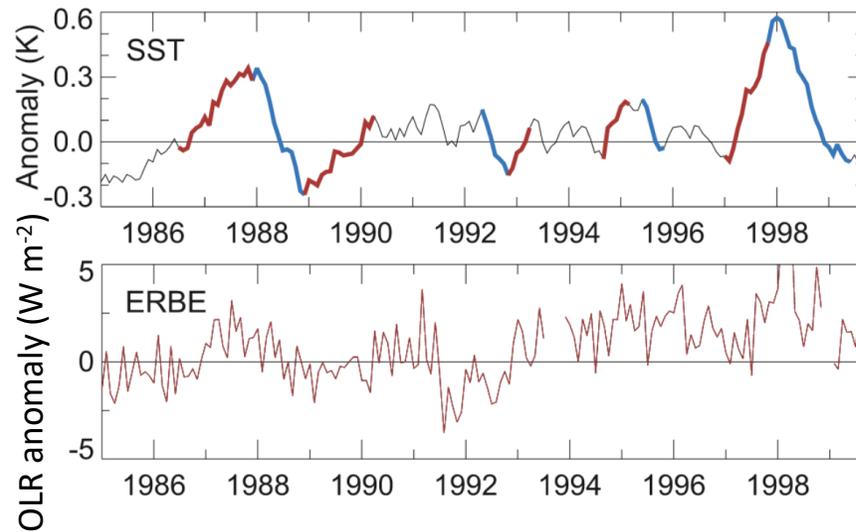


doi: 10.1029/2009GL039628

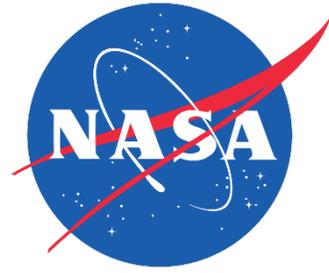
© 2009. AGU. Permission granted for academic use.



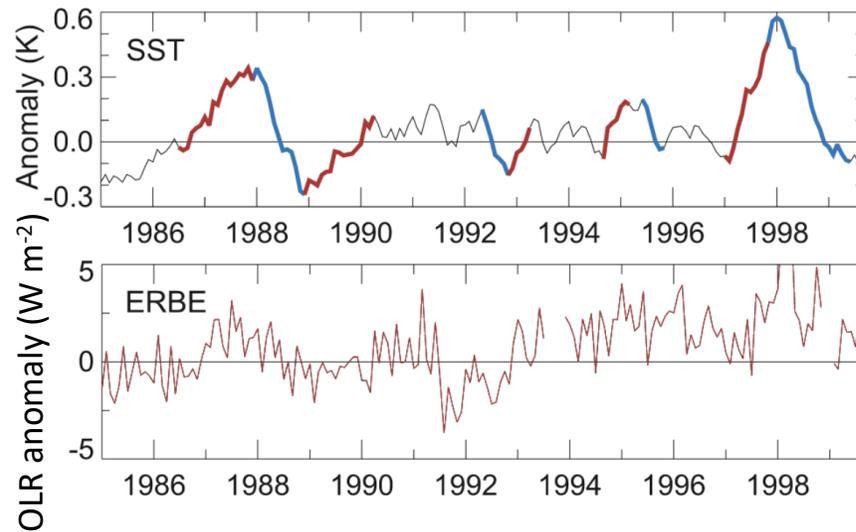
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“ERBE data appear to demonstrate a climate sensitivity of about 0.5°C ”



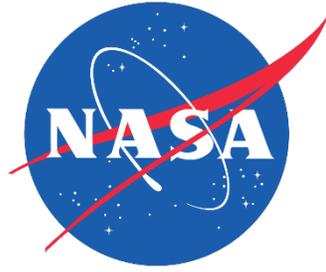
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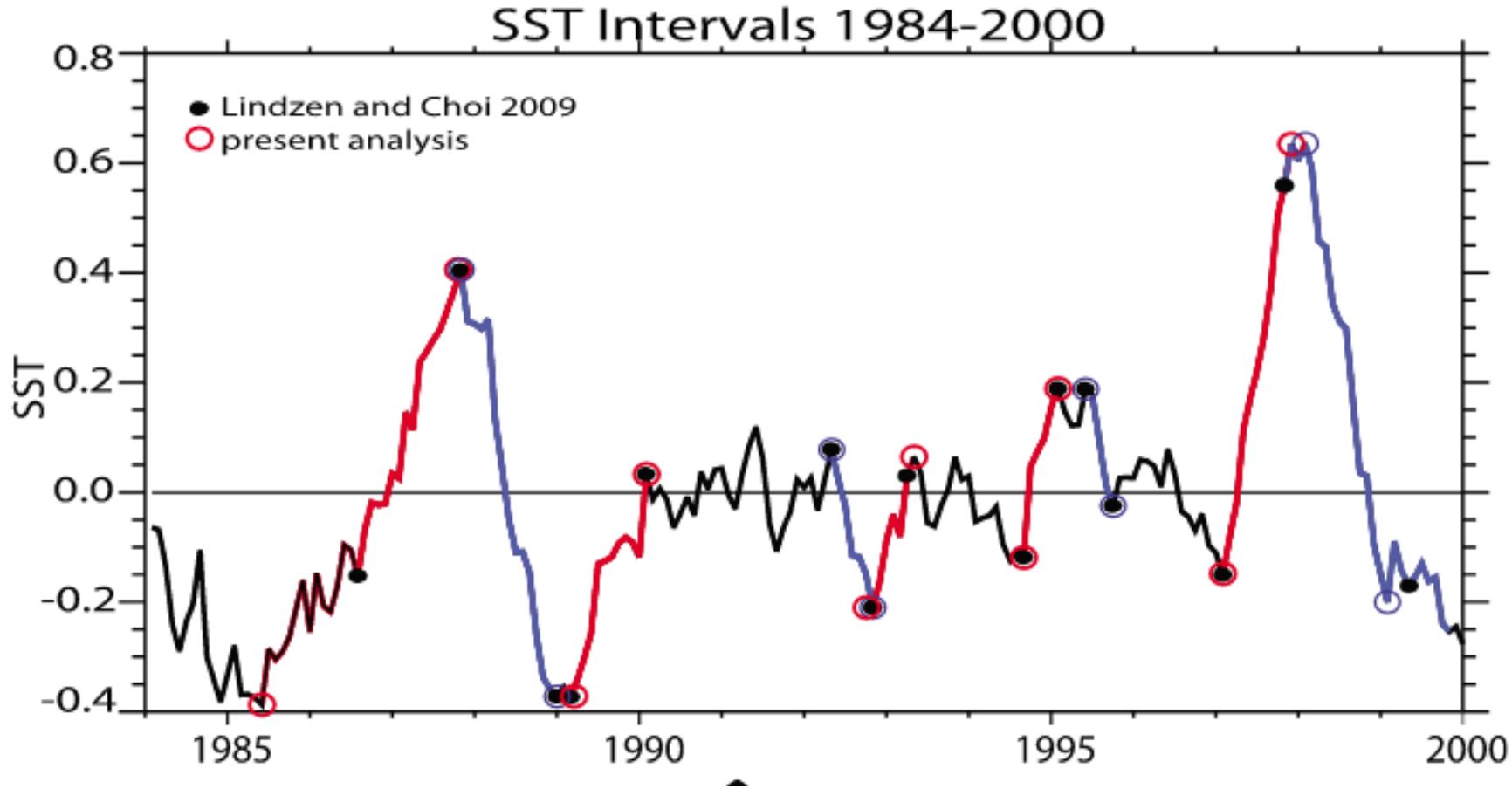
But how and why did they pick those coloured periods?

doi: 10.1029/2009GL039628

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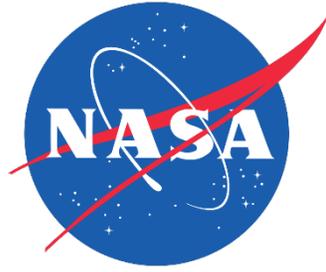
BUT:



Trenberth, Wong & O'Dell (2010) : “[Lindzen & Choi’s 2009] results are neither robust nor meaningful, as small sensible changes in the dates bounding their warming and cooling intervals entirely change the conclusions.”

Lindzen & Choi (2009):

If you select time periods in just the right way, then
ECS is much lower than in models



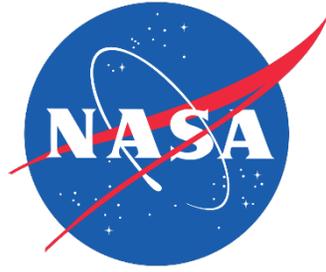
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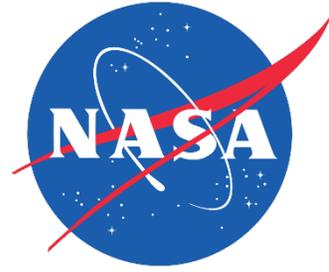
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Spencer & Braswell (2011):

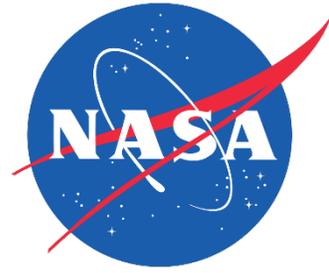
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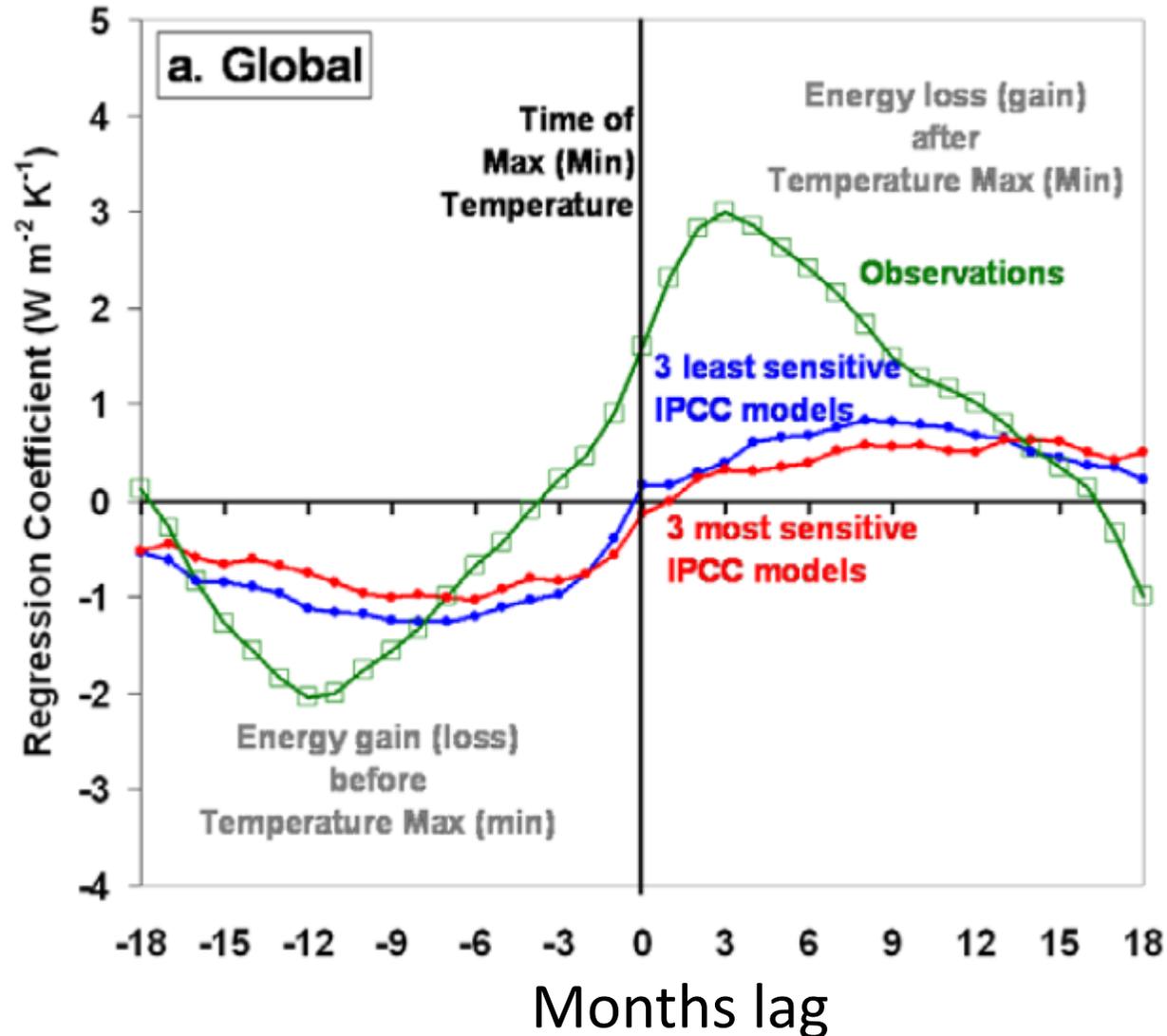


ScienceDaily.com – “Earth’s atmosphere may be more efficient at releasing energy to space than climate models indicate, satellite data suggest”

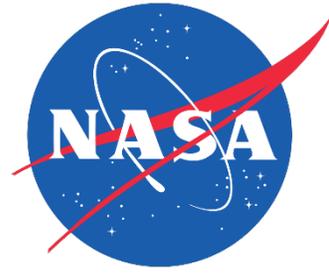
Forbes – “New NASA Data Blow Gaping Hole in Global Warming Alarmism”



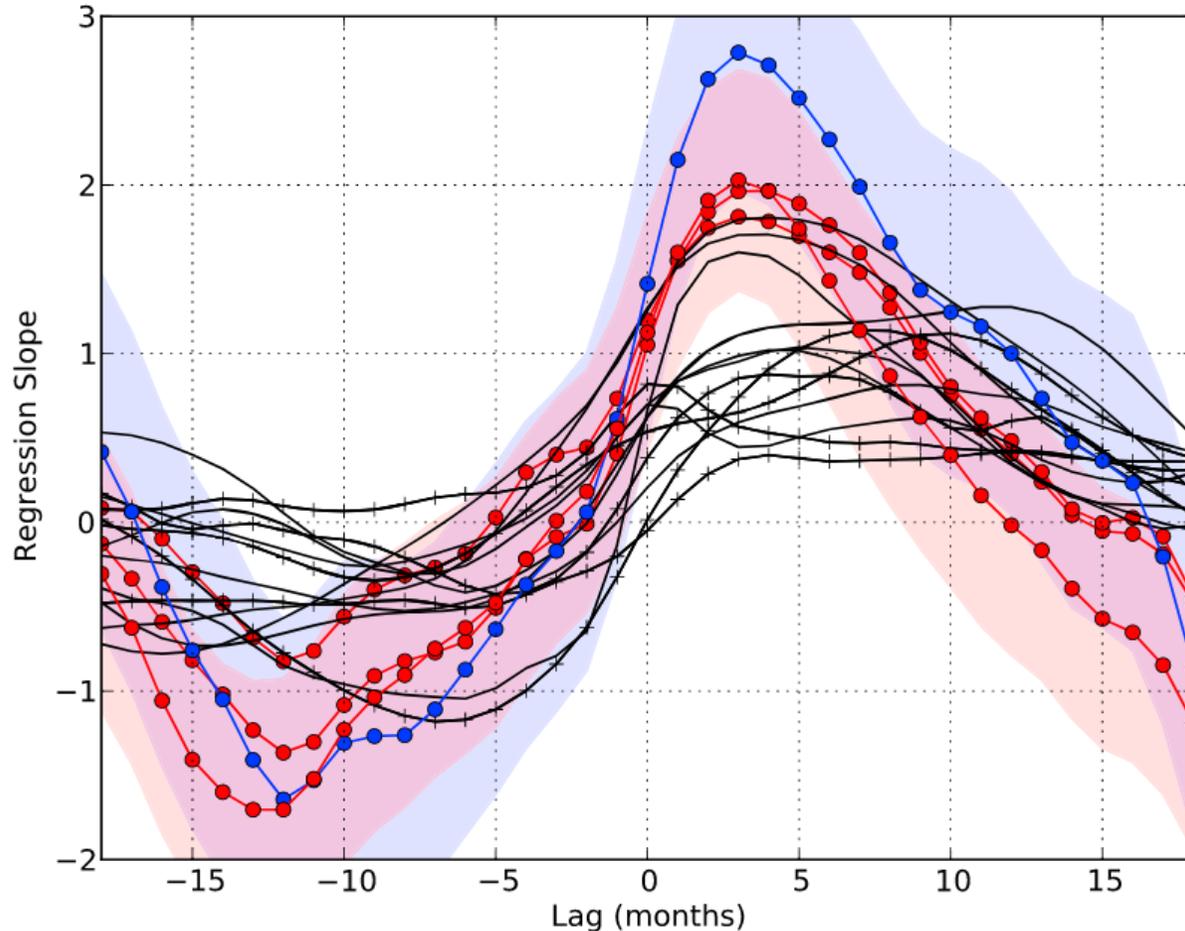
Spencer and Braswell (2011)



“...satellite-based metric ...depart substantially in the direction of lower climate sensitivity ...we find that, with traditional methods, it is not possible to accurately quantify this discrepancy”



Dessler (2011):



Colours = obs \pm uncertainty
Black lines = models

Original paper looked at 14 models, plotted just 6. Only plotted $\frac{1}{4}$ of obs datasets.

“[Plotting all] of the data provide a much different conclusion.”

Lindzen & Choi (2009):

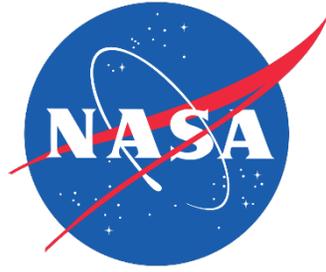
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Spencer & Braswell (2011):

If you remove most of the model and obs, then

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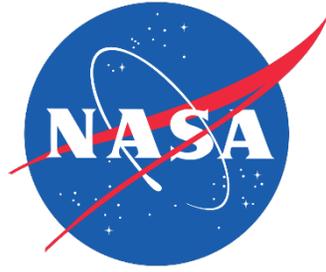
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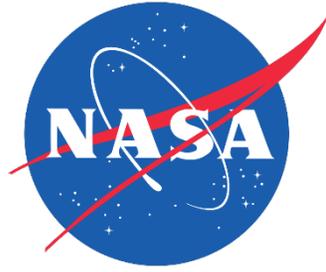
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Monckton et al. (2015)

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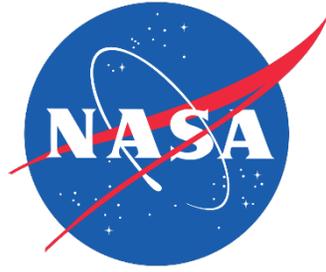




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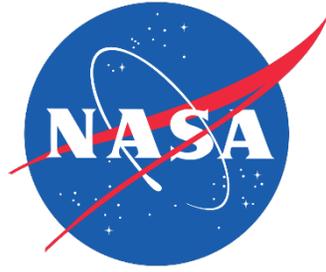
“Peer-reviewed pocket-calculator climate model exposes serious errors in complex computer models” – Phys.org

“Is climate change really that dangerous? Predictions are ‘very greatly exaggerated’, claims study” – The Daily Mail



Monckton et al. (2015)

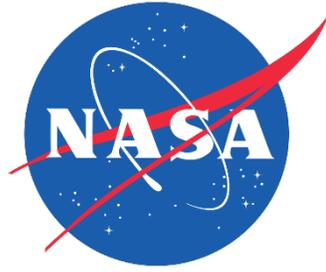
“The model indicates...that, since feedbacks are likely to be net-negative, a better estimate is 1.0 K;”



Monckton et al. (2015)

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“The simple model has only five tunable parameters:”

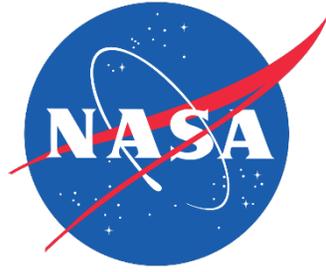


Monckton et al. (2015)

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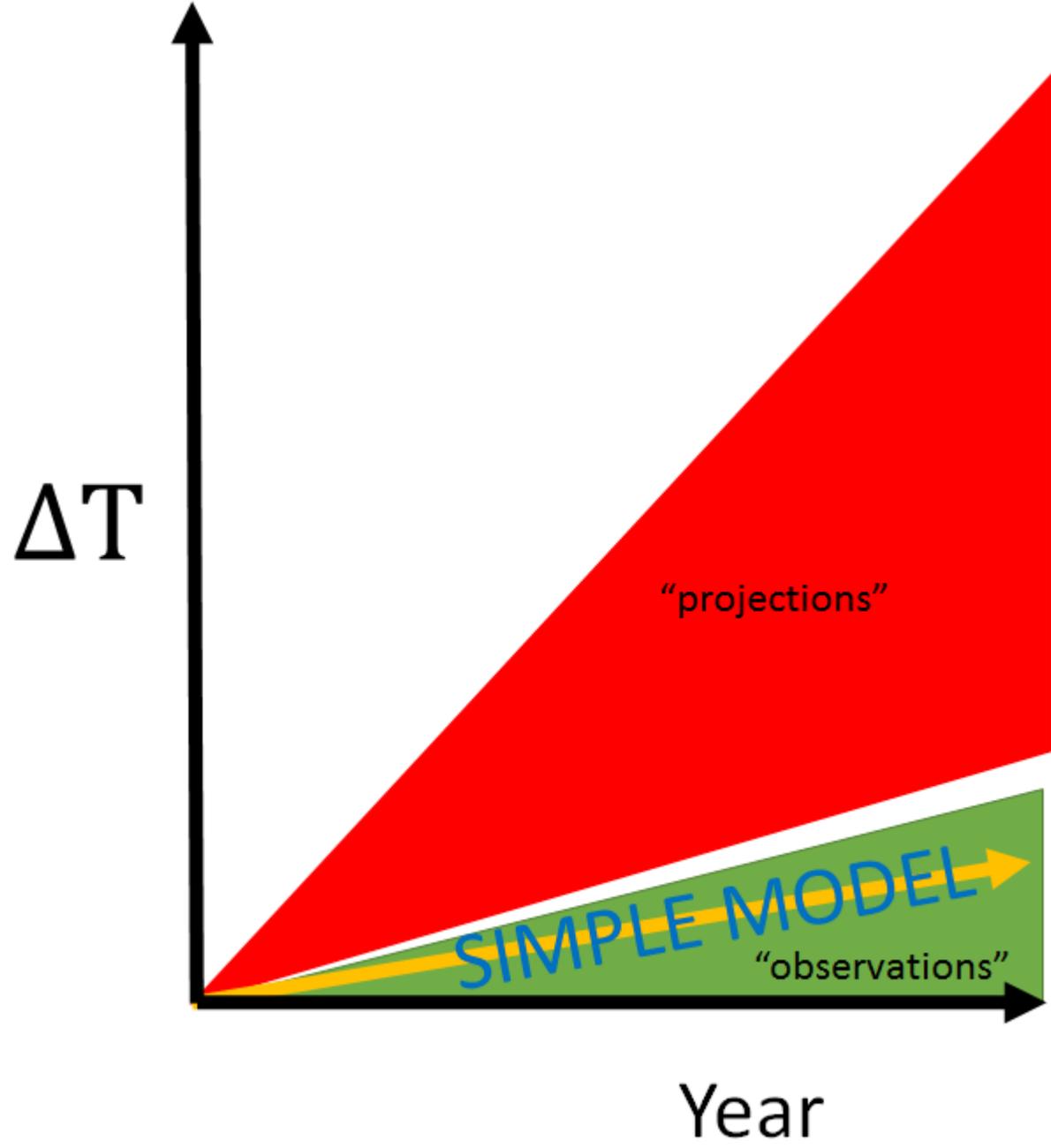
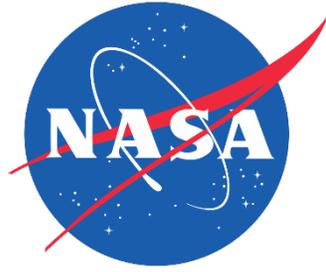
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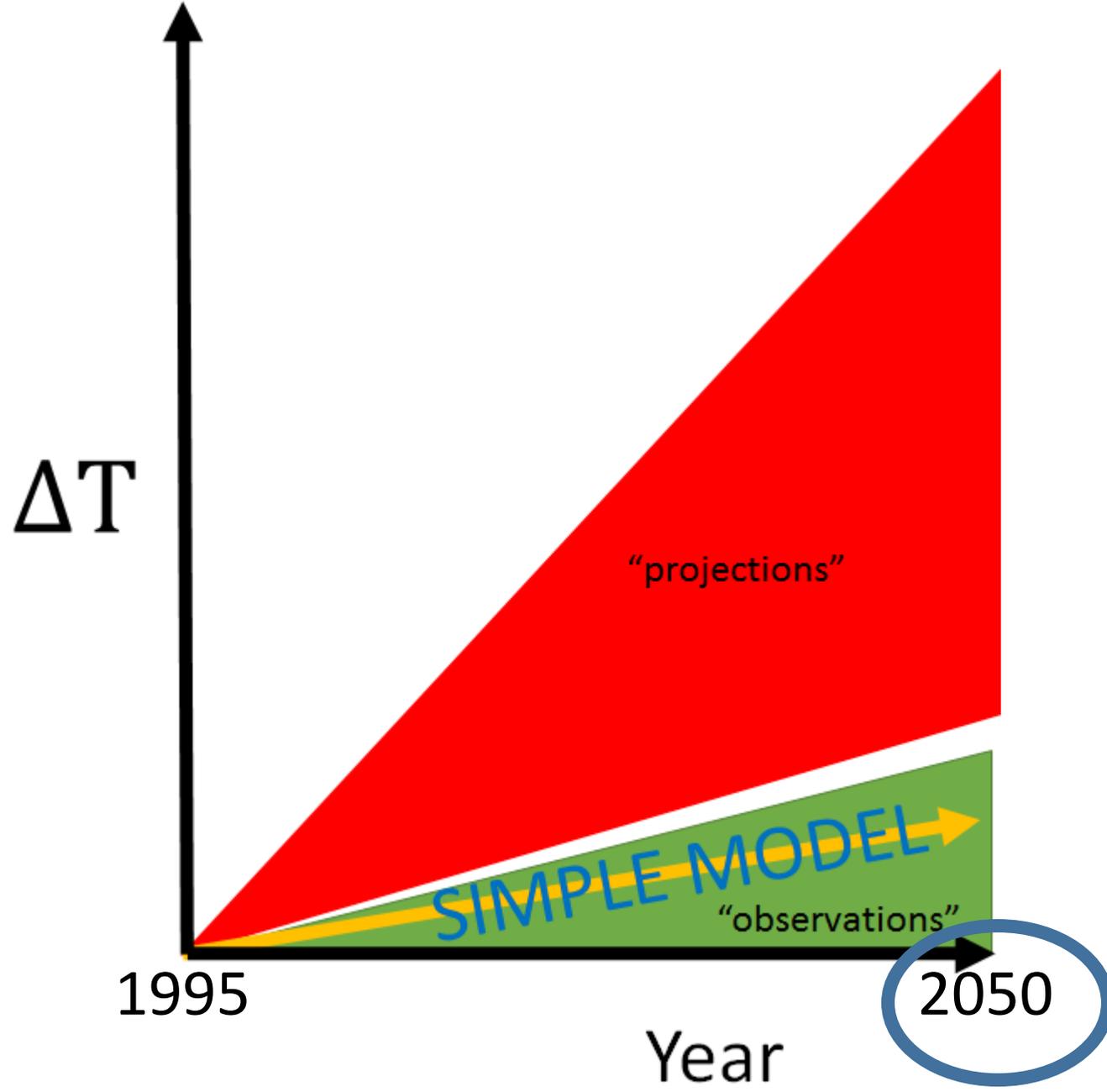
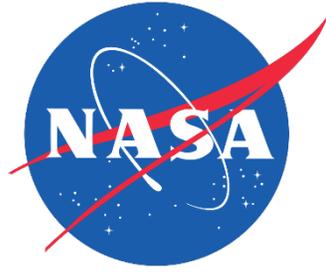
“In [Fig. 5](#), a regime of temperature stability is represented by $g_{\infty} \leq +0.1$, the maximum value allowed by process engineers designing electronic circuits intended not to oscillate under any operating conditions.”



Monckton et al. (2015)

“[The model’s] output proves to be broadly consistent with observation, while the now-realized projections of the general-circulation models have proven to be relentlessly exaggerated.”





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If you select time periods in just the right way, then

ECS is much lower than in models

Spencer & Braswell (2011):

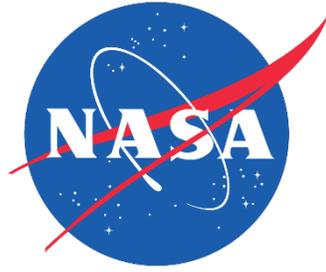
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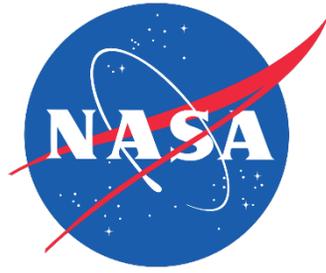
ECS is much lower than in models

Monckton et al. (2015)

If you make up parameters and data

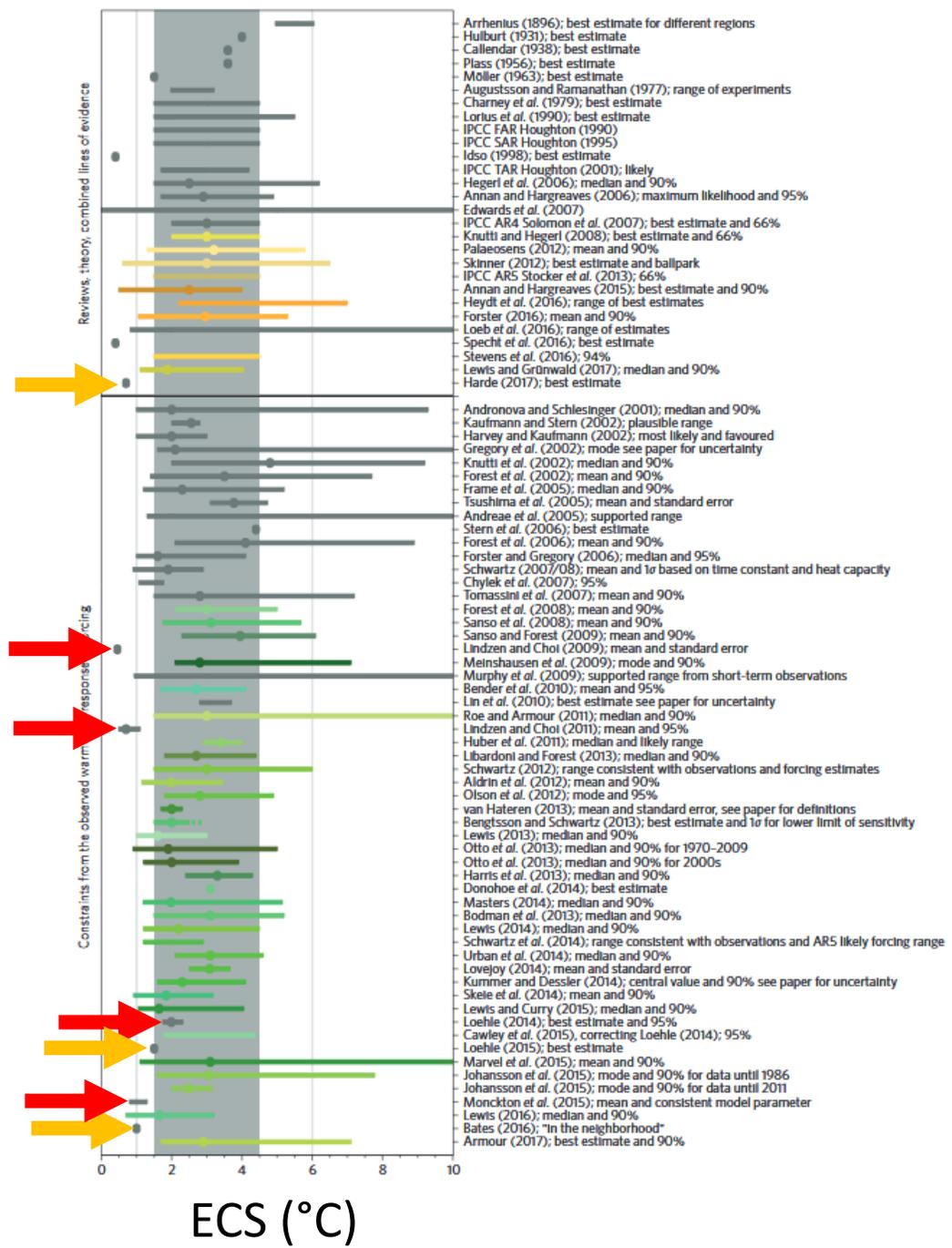
ECS is much lower than in models

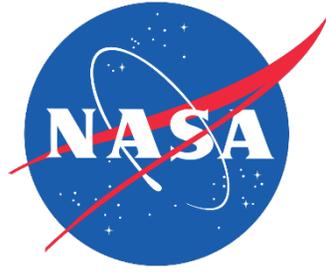




Red published comments show errors

Orange use same or similar techniques to those known to be flawed.





Lindzen & Choi (2009):

If you select time periods in just the right way, then

ECS is much lower than in models

Spencer & Braswell (2011):

If you remove most of the model and obs, then

ECS is much lower than in models

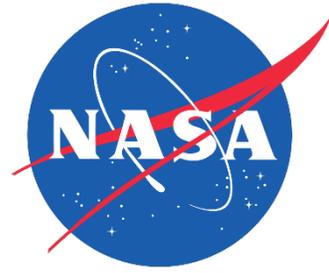
Monckton et al. (2015)

If you make up parameters and data

ECS is much lower than in models

Lewis & Curry (2018)

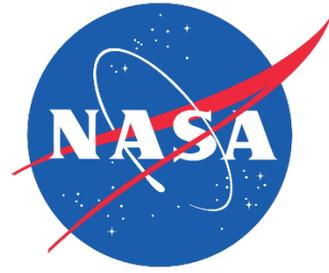
ECS is quite a bit lower than in models



“IPCC Overestimates Climate Sensitivity: Study” –
Competitive Enterprise Institute

“Some More Insensitivity about Global Warming”
– Cato Institute

“Is Climate Alarmist Consensus about to Shatter?”
– Heartland Institute



How does this paper compare?

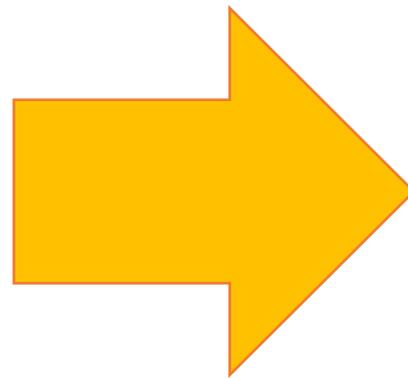
$$ECS = \Delta F_{2\times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$

$\Delta F_{2\times CO_2}$ 3.81 W m⁻²

ΔT 0.80 °C

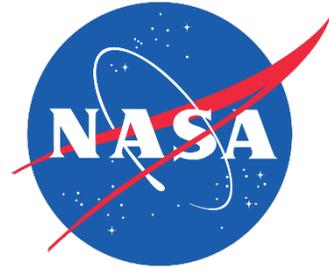
ΔF 2.52 W m⁻²

ΔN 0.50 W m⁻²



ECS 1.50 °C

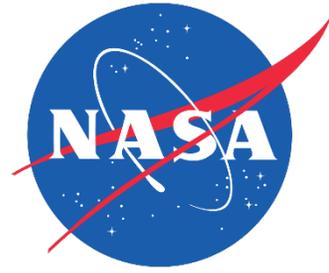
5—95 %: 1.05—2.45 °C



How does this paper compare?

$$ECS = \Delta F_{2\times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$

Come from models

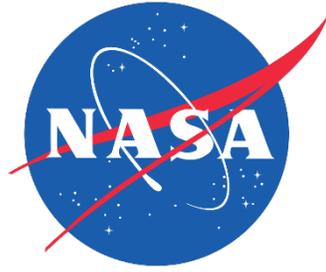


How does this paper compare?

$$ECS = \Delta F_{2 \times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$

Come from models

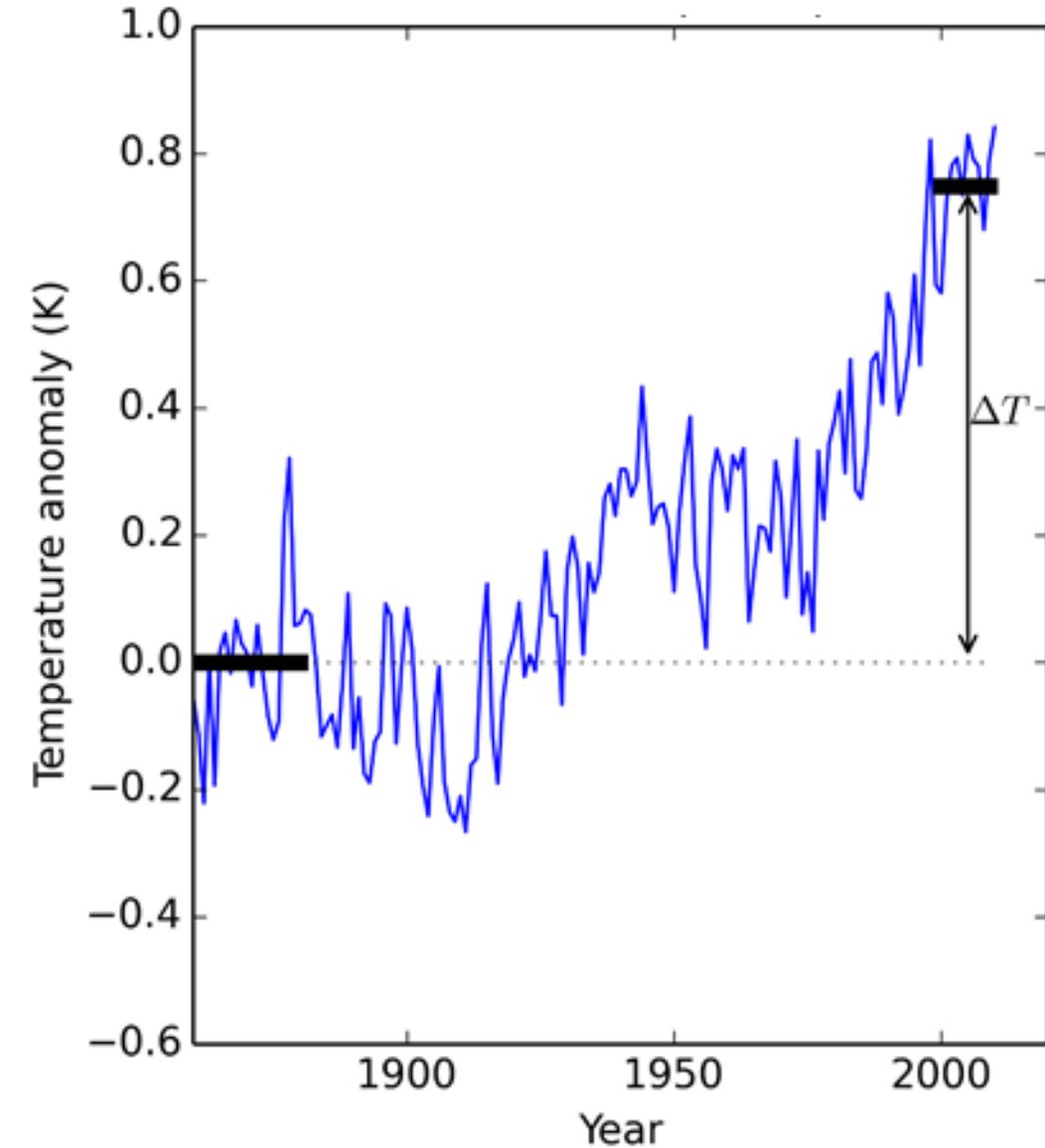
Let's look at the ΔT and ΔN to see if they explain the difference between Lewis & Curry and IPCC



$$ECS = \Delta F_{2 \times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$

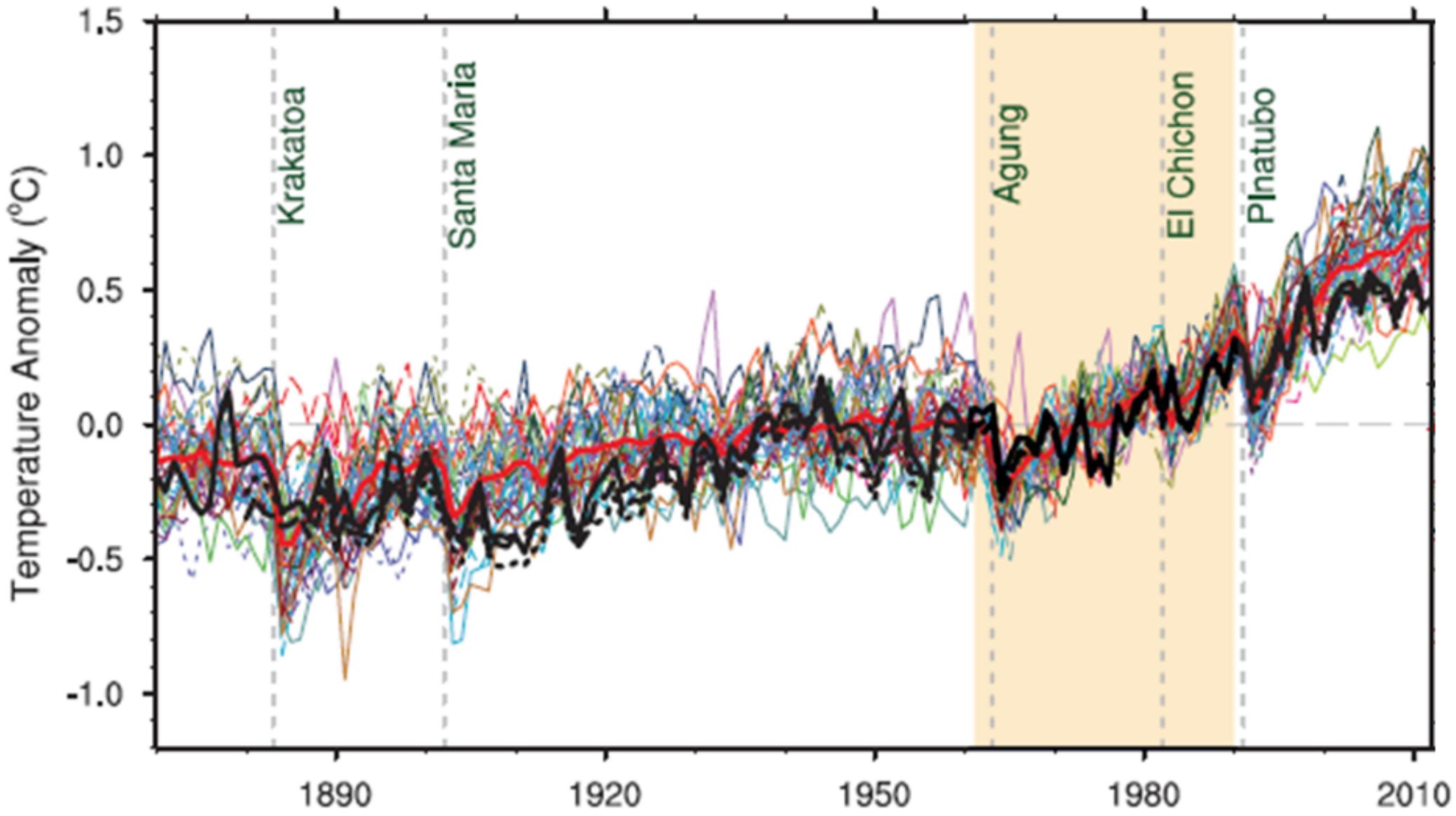
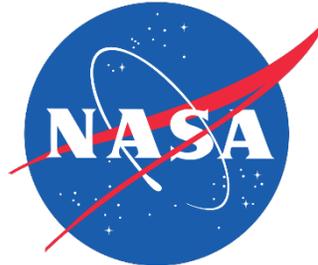
Lewis & Curry use:

$$\Delta T = \overline{T_{2007-2016}} - \overline{T_{1869-1882}}$$

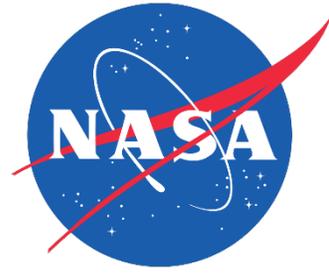


IPCC AR5 Figure 9-08,
CMIP5 vs HadCRUT4 (obs)

$$ECS = \Delta F_{2 \times CO2} \frac{\Delta T}{\Delta F - \Delta N}$$



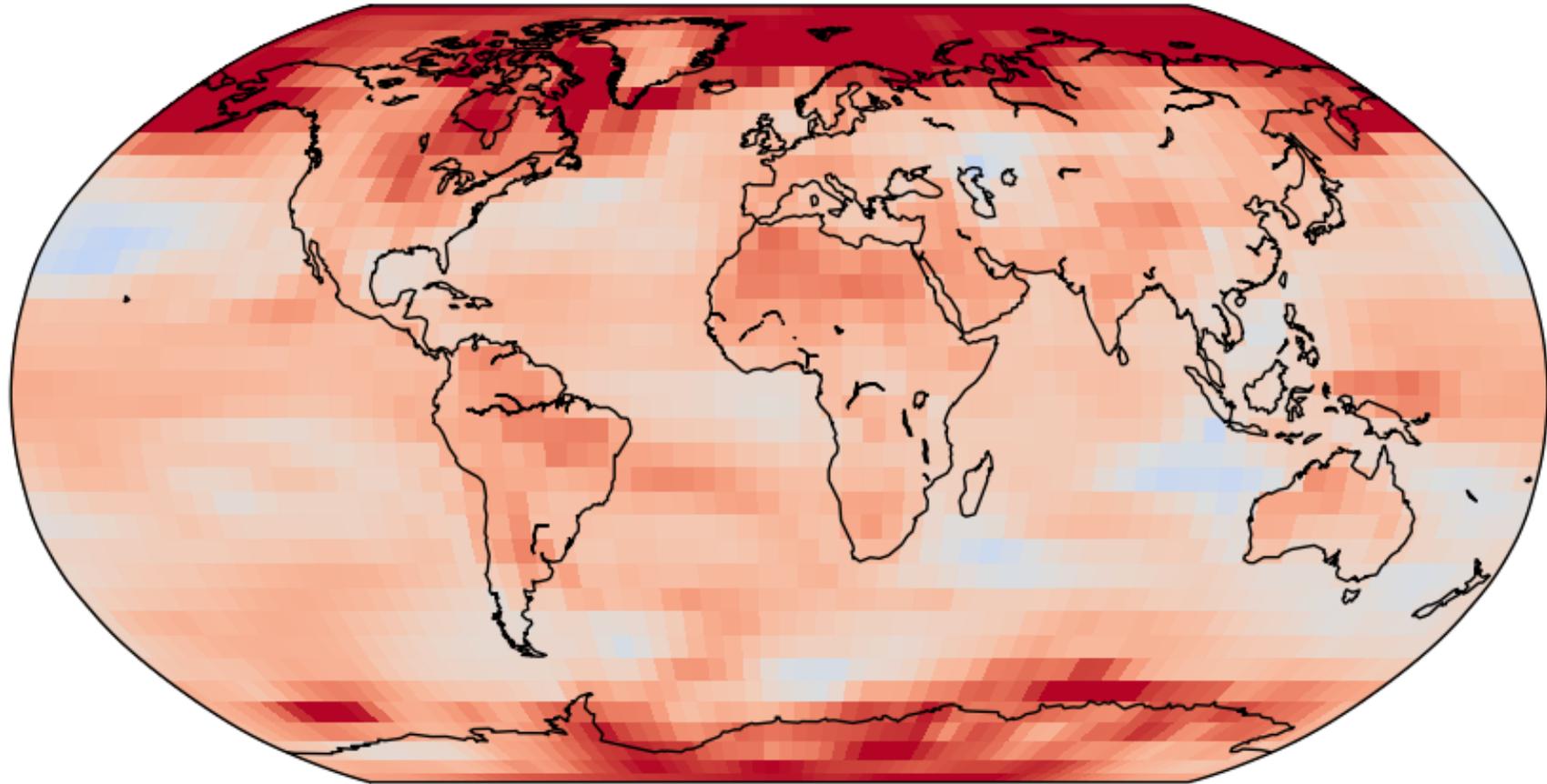
$$ECS = \Delta F_{2\times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$



1 Model:

1861—2005
warming:

0.66 °C

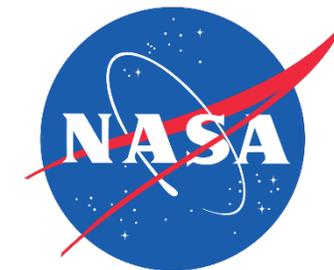


Temperature change (°C)



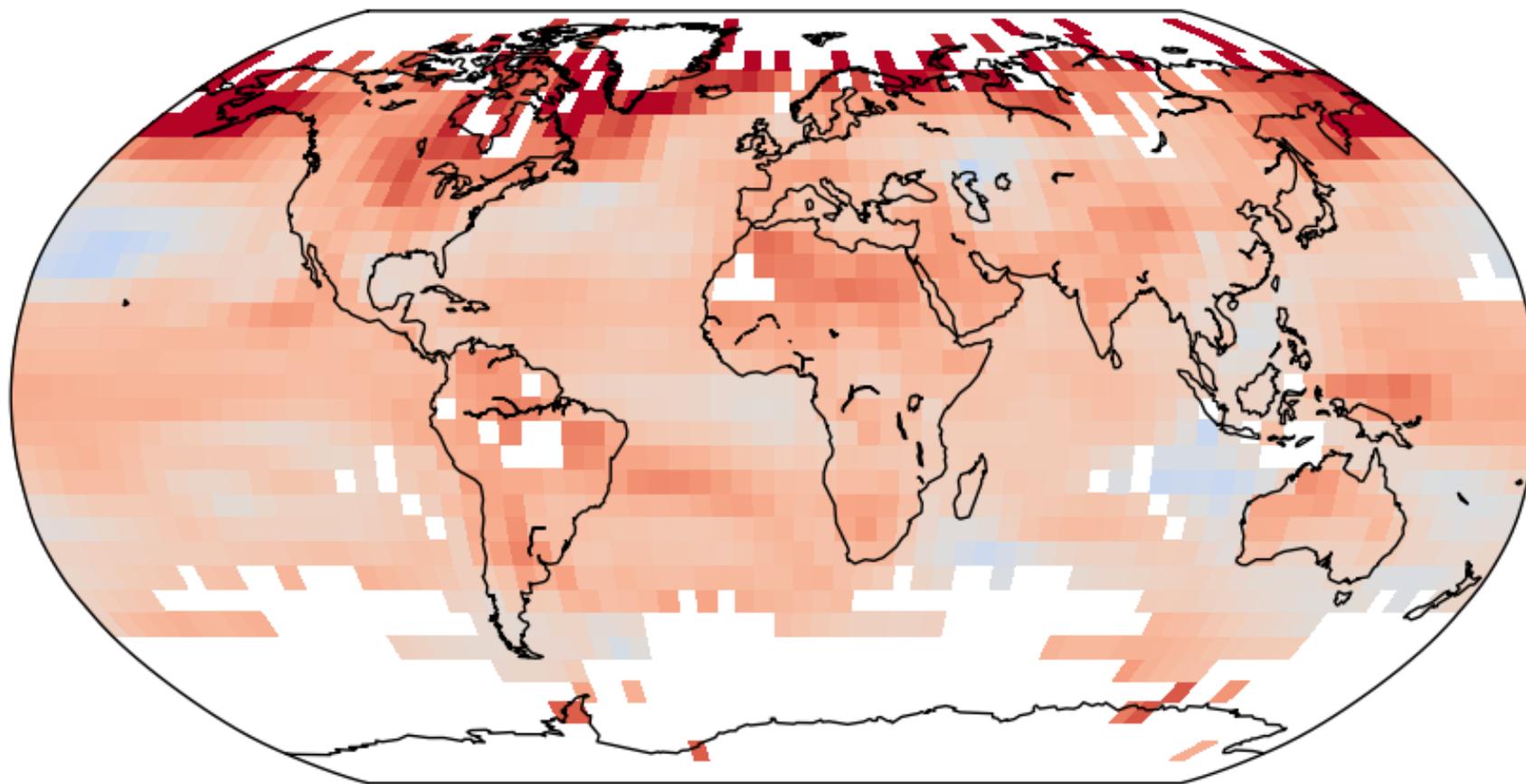
Measured
areas 1996—
2005

$$ECS = \Delta F_{2 \times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$

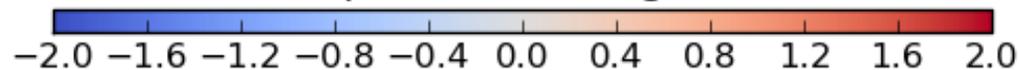


1861—2005
warming:

0.59 °C



Temperature change (°C)

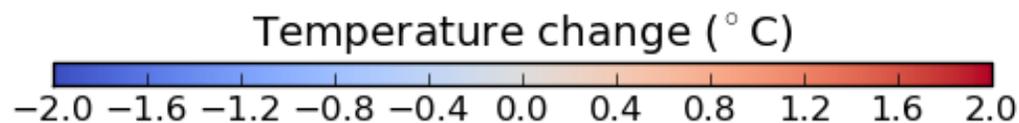
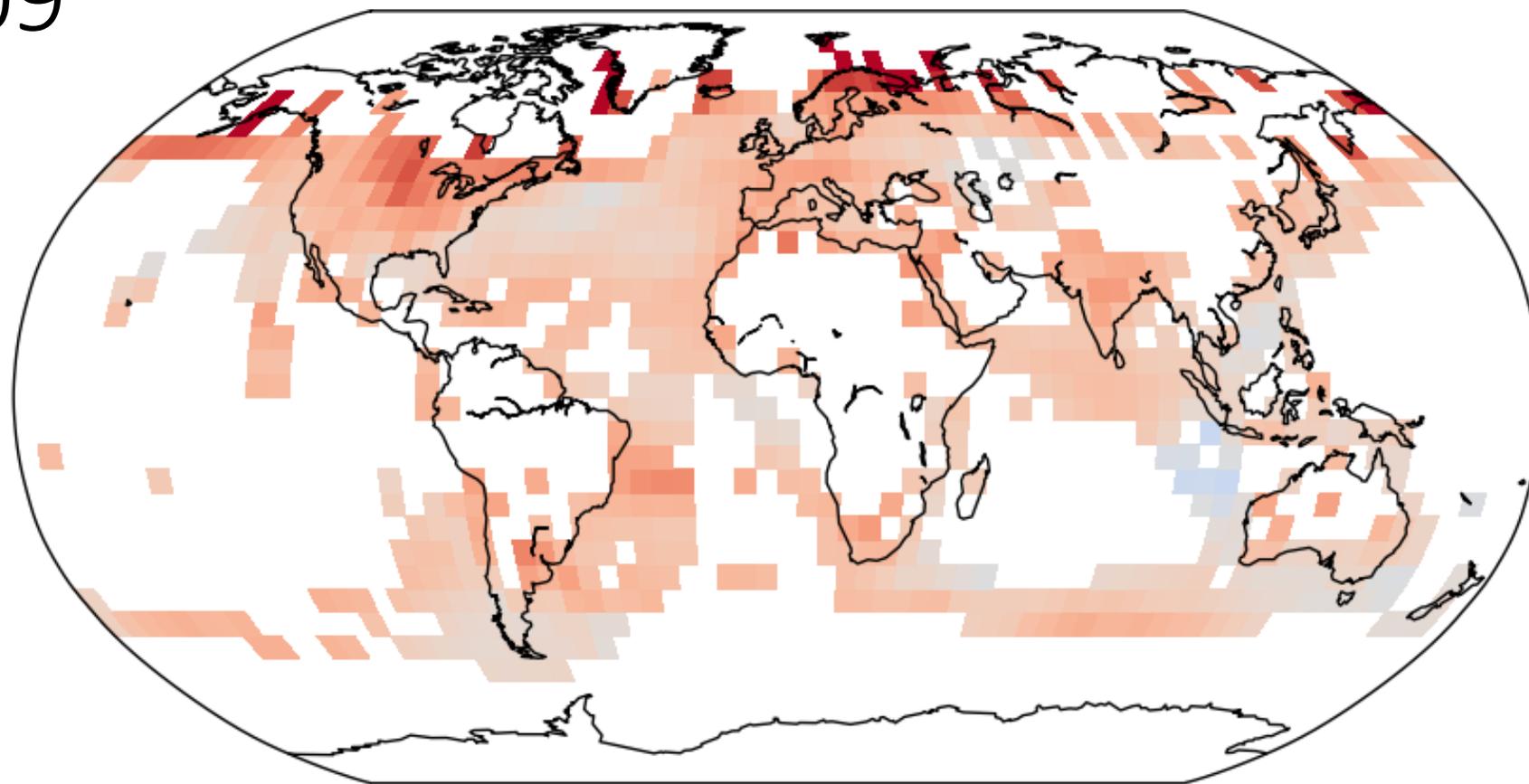
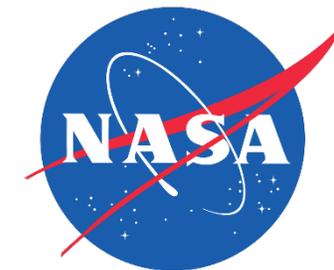


Measured
areas
1900—1909

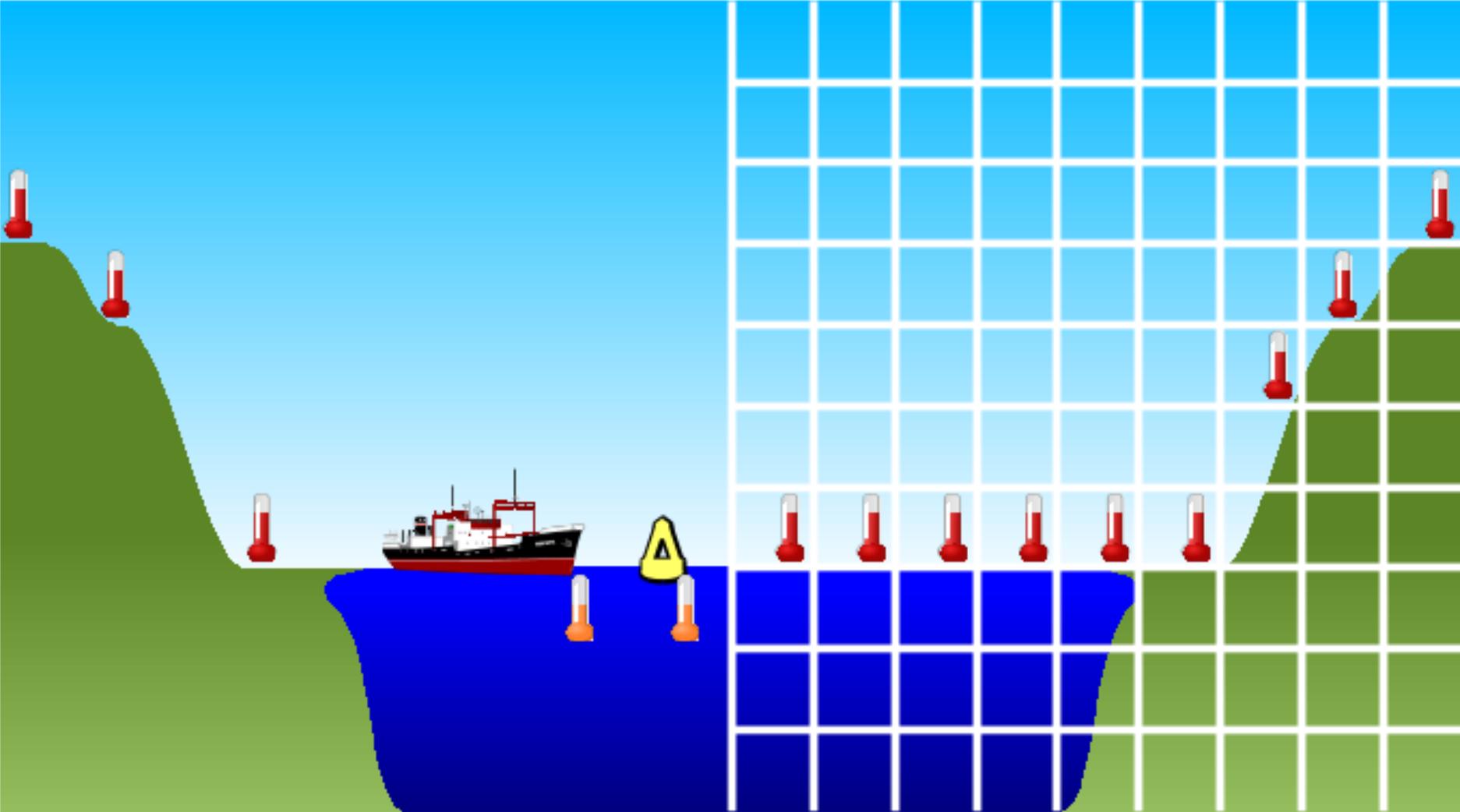
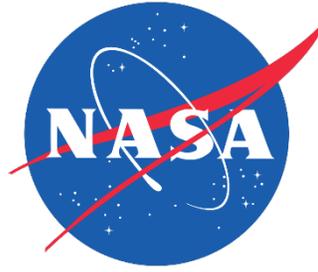
1861—2009
warming:

0.53 °C

$$ECS = \Delta F_{2\times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$

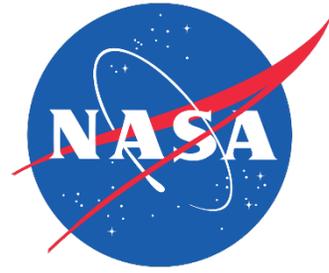


$$ECS = \Delta F_{2\times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$



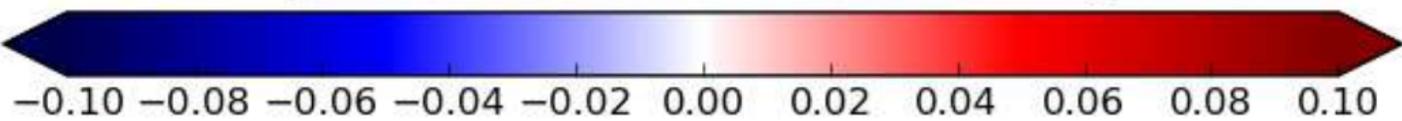
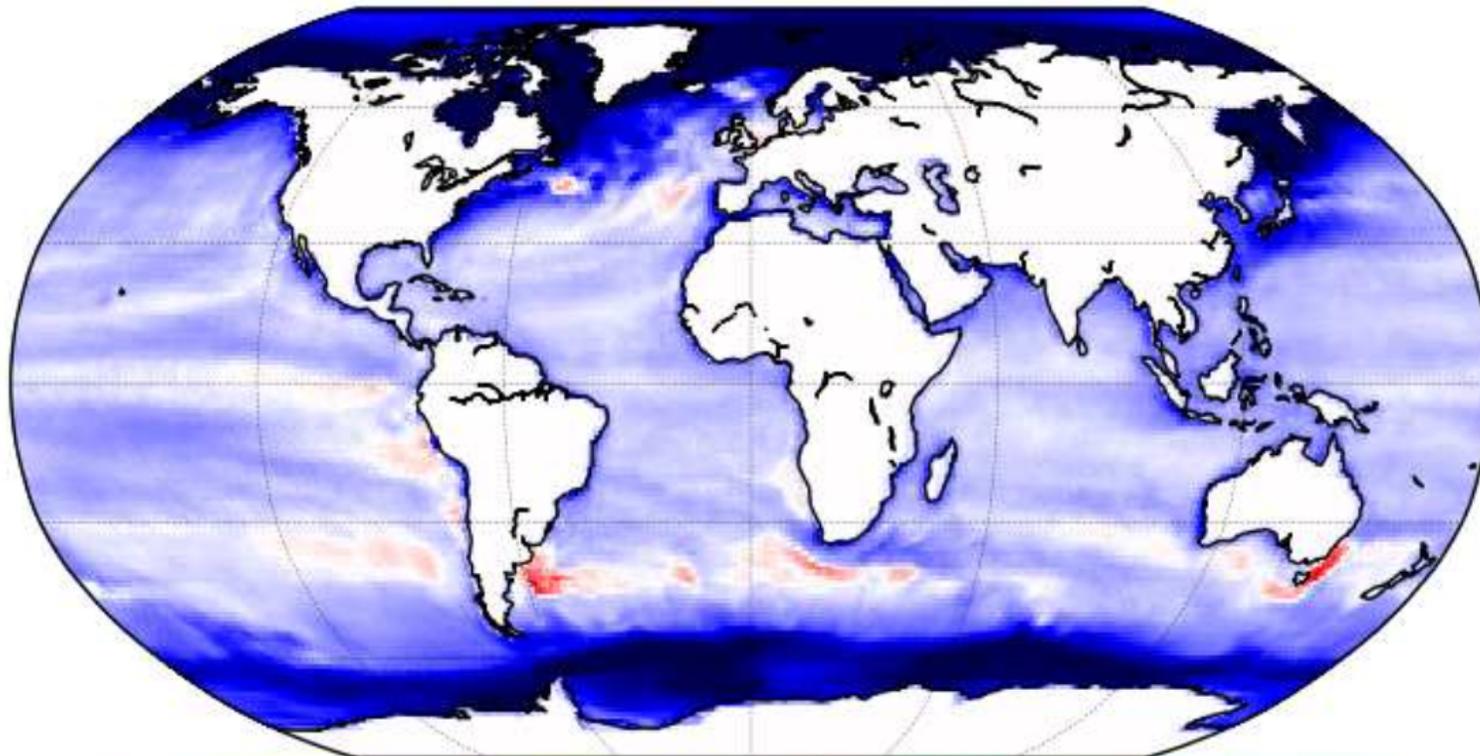
Observations

Models



$$ECS = \Delta F_{2\times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$

(b) Trend difference, blended temperature anomalies

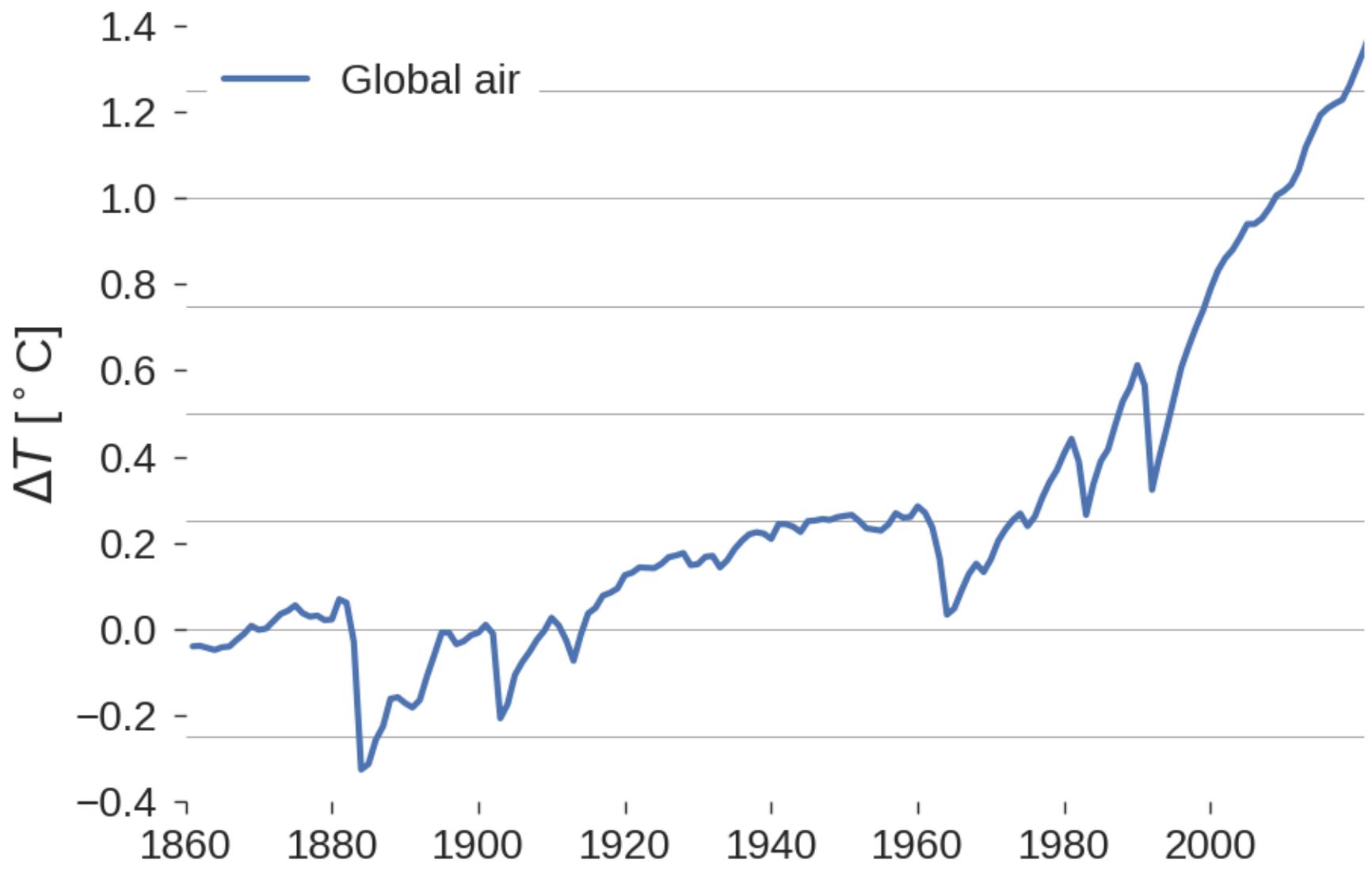
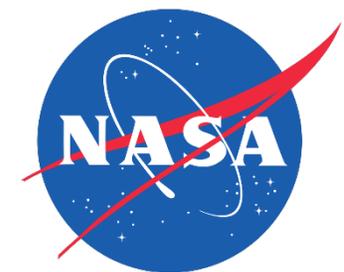


K/decade

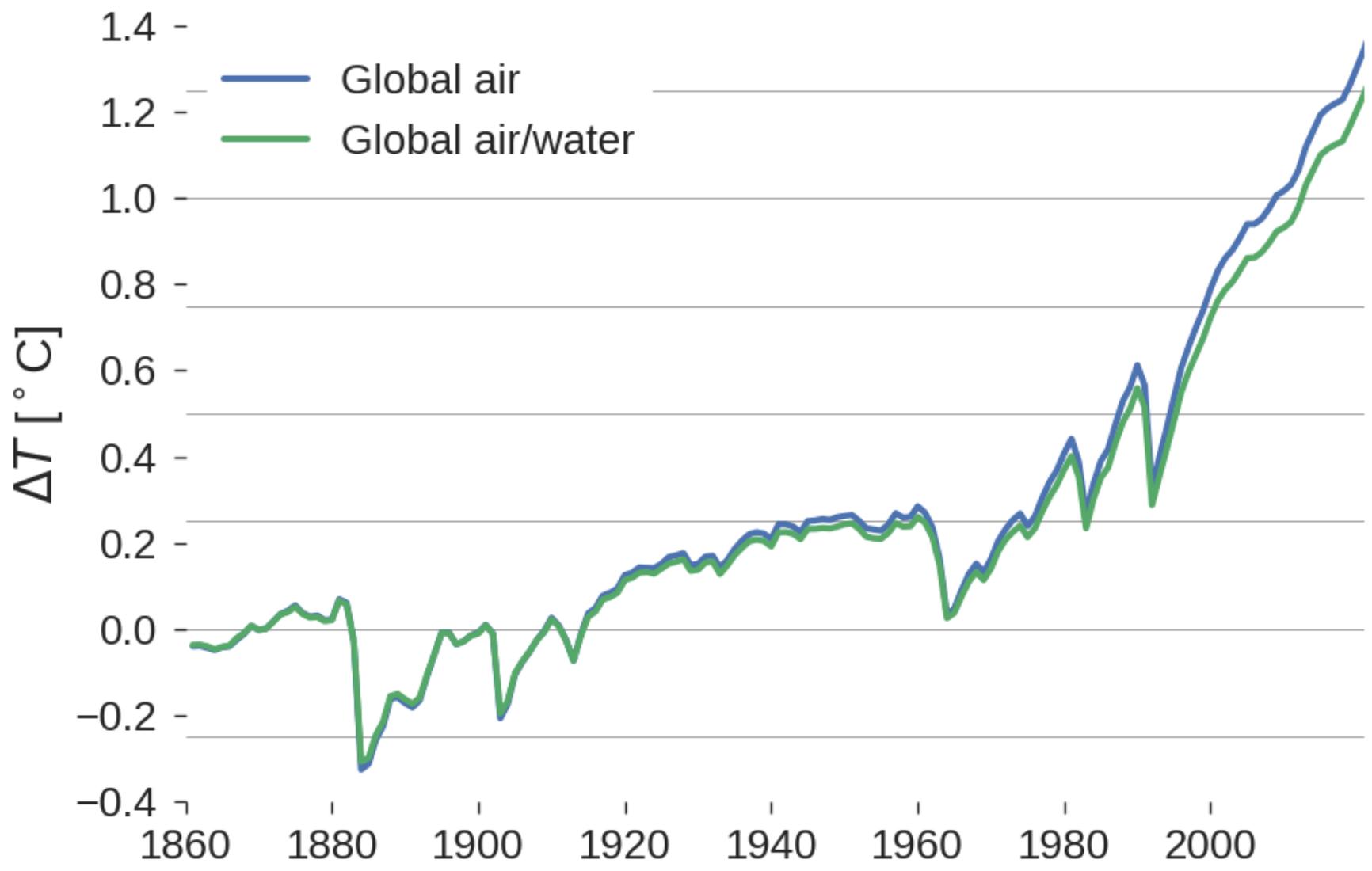
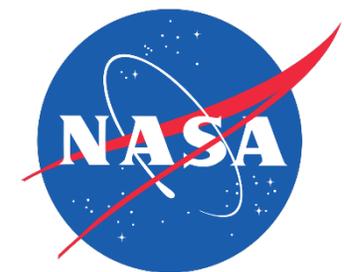
1985—2014 CMIP5
mean ocean minus
air trend

SSTs warm less.

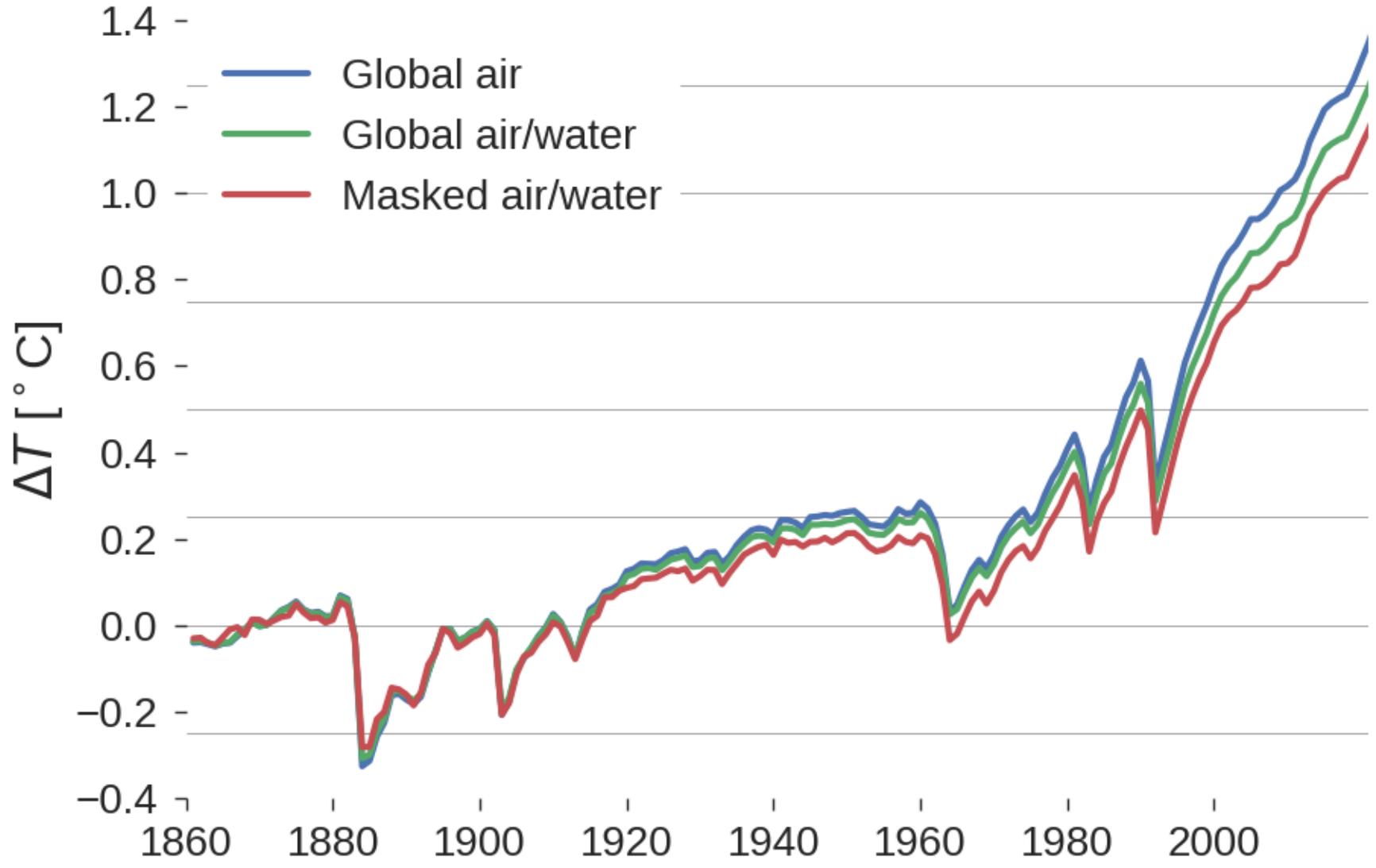
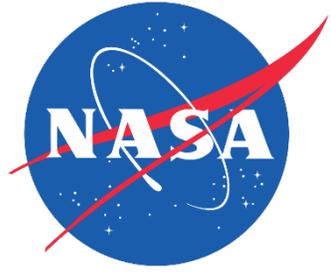
$$ECS = \Delta F_{2\times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$



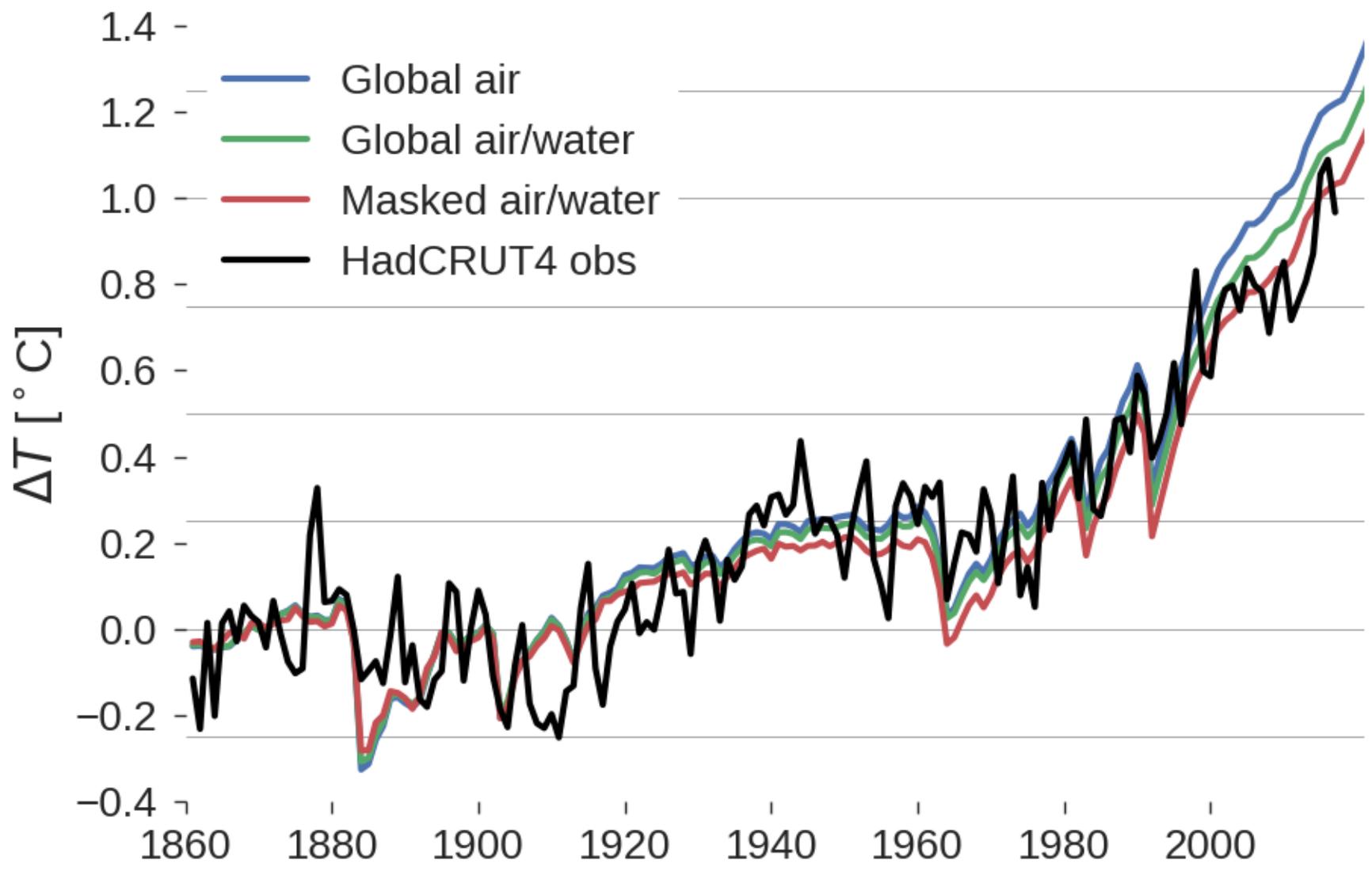
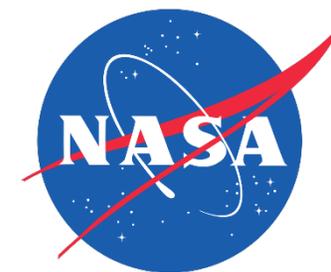
$$ECS = \Delta F_{2\times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$



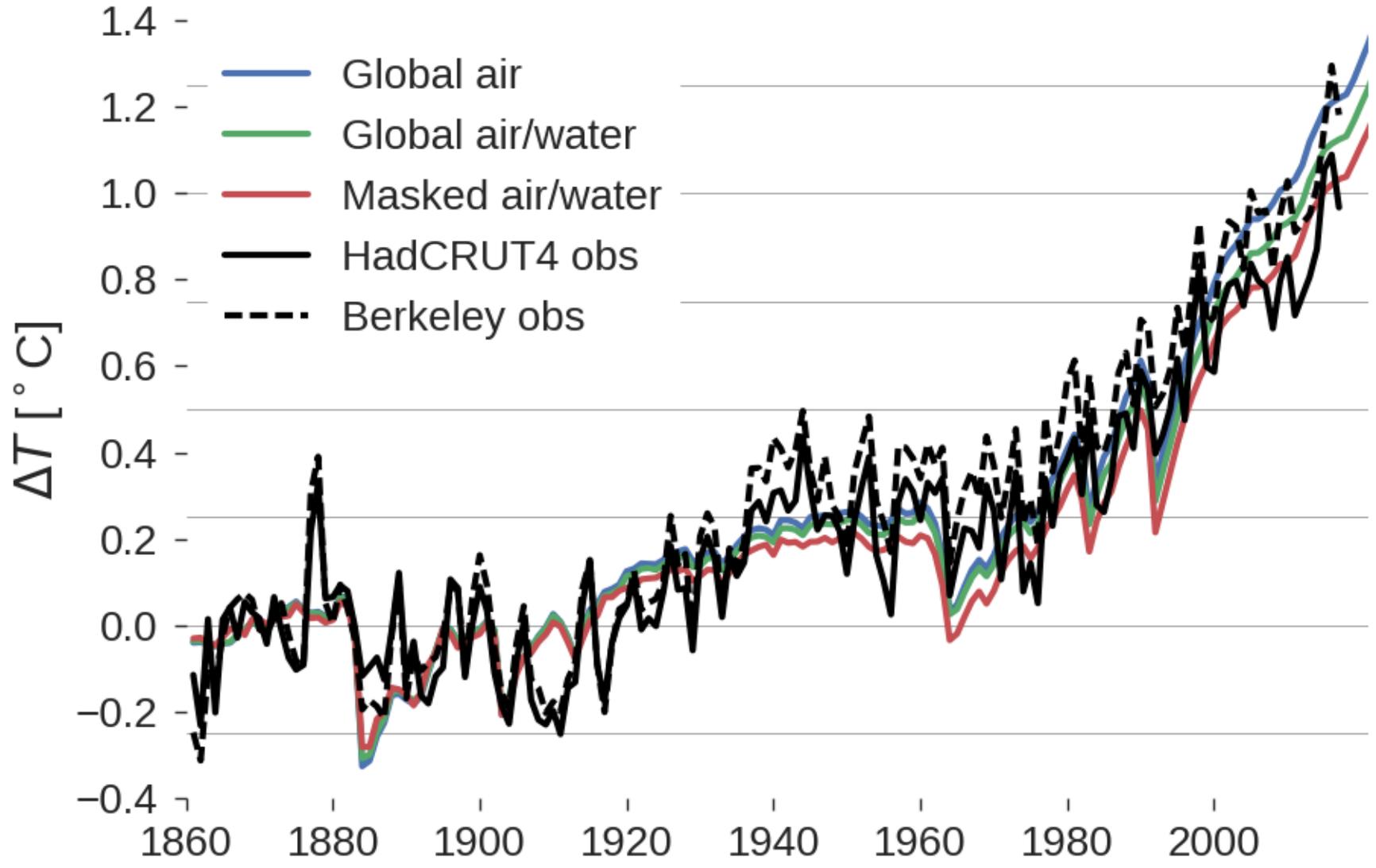
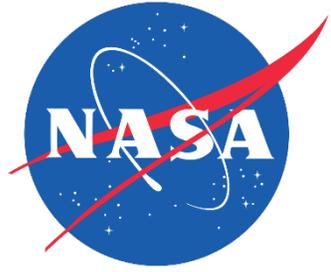
$$ECS = \Delta F_{2 \times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$

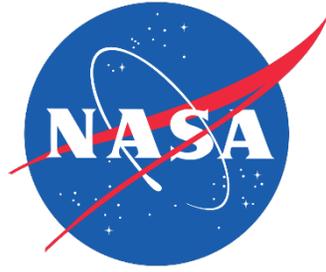


$$ECS = \Delta F_{2 \times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$



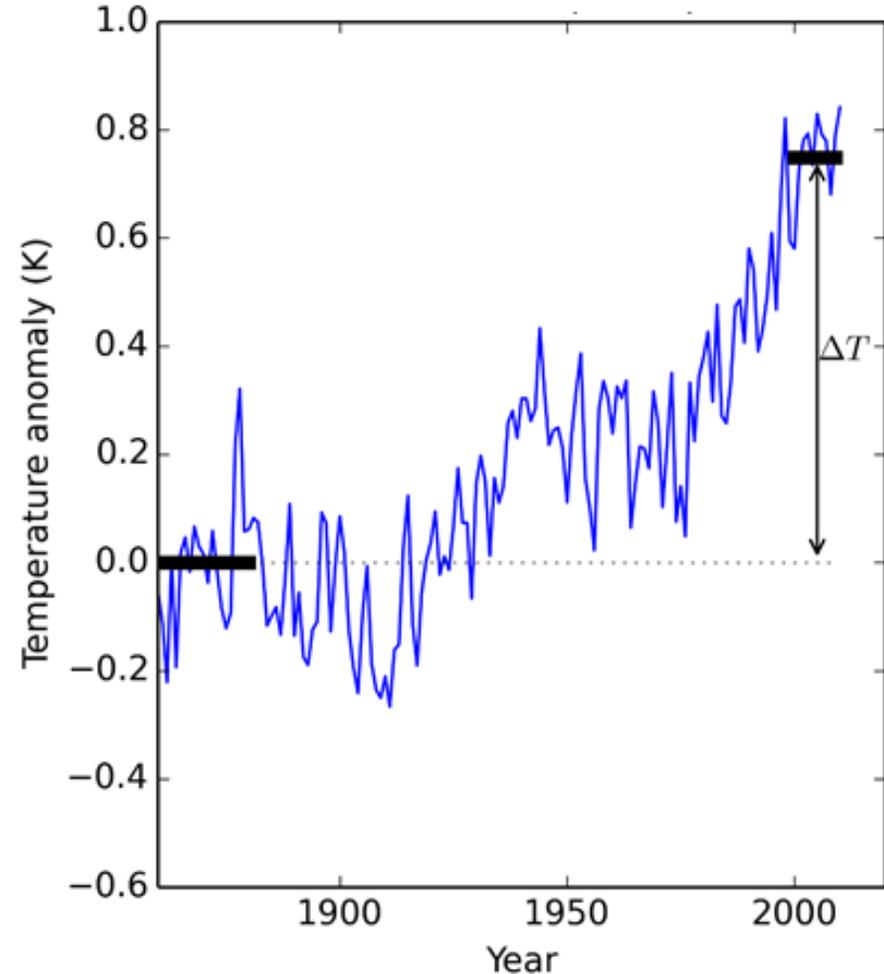
$$ECS = \Delta F_{2 \times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$

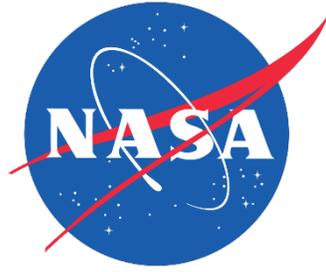




Consistent comparison

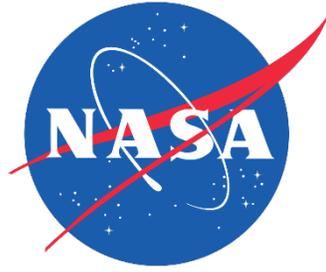
- ΔT ~20 % higher with global coverage + reduced air/water blending bias
- Berkeley Earth minimizes these and says 0.95 °C
- Lewis & Curry used HadCRUT4 0.80 °C for main result





Consistent comparison ΔN

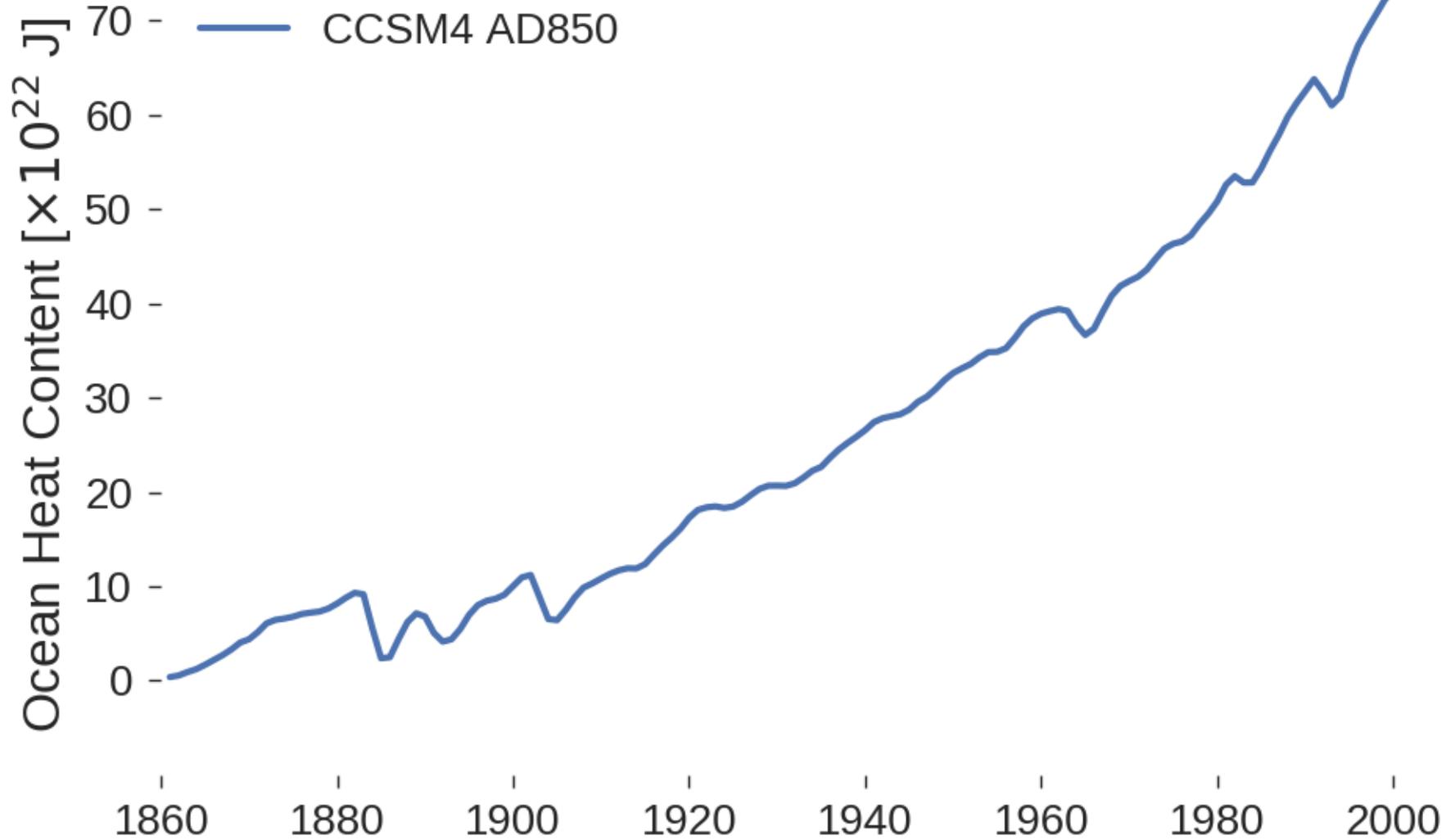
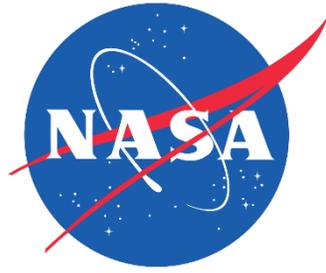
- ΔN is the energy imbalance
- Mostly ocean heat, generally from Argo floats.
- **WE DO NOT HAVE 1869—1882 DATA... where does it come from?**



Early period heat uptake

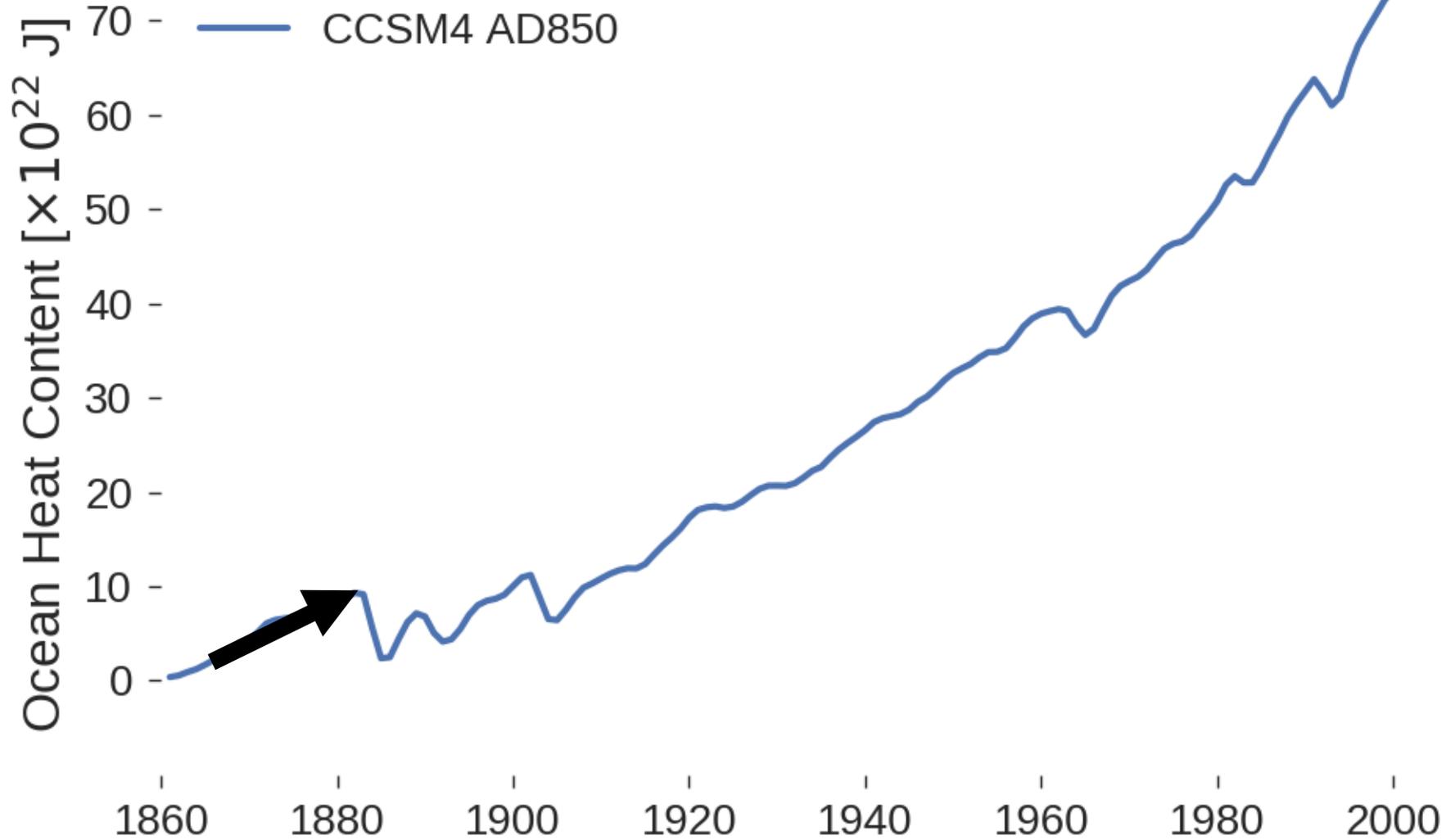
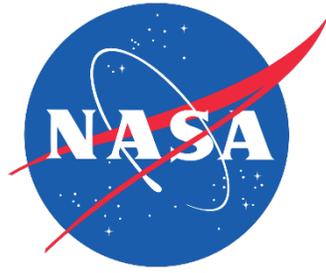
- Lewis & Curry take $\overline{N_{1869-1882}}$ from a CCSM4 model run that began in AD850.
- Climate models are “spun up” in a control run, often without volcanoes
- Small, missing volcanic cooling so ocean is *slightly* out of balance
- The moment volcanoes erupt, deep ocean cools.

Source of Lewis & Curry early N



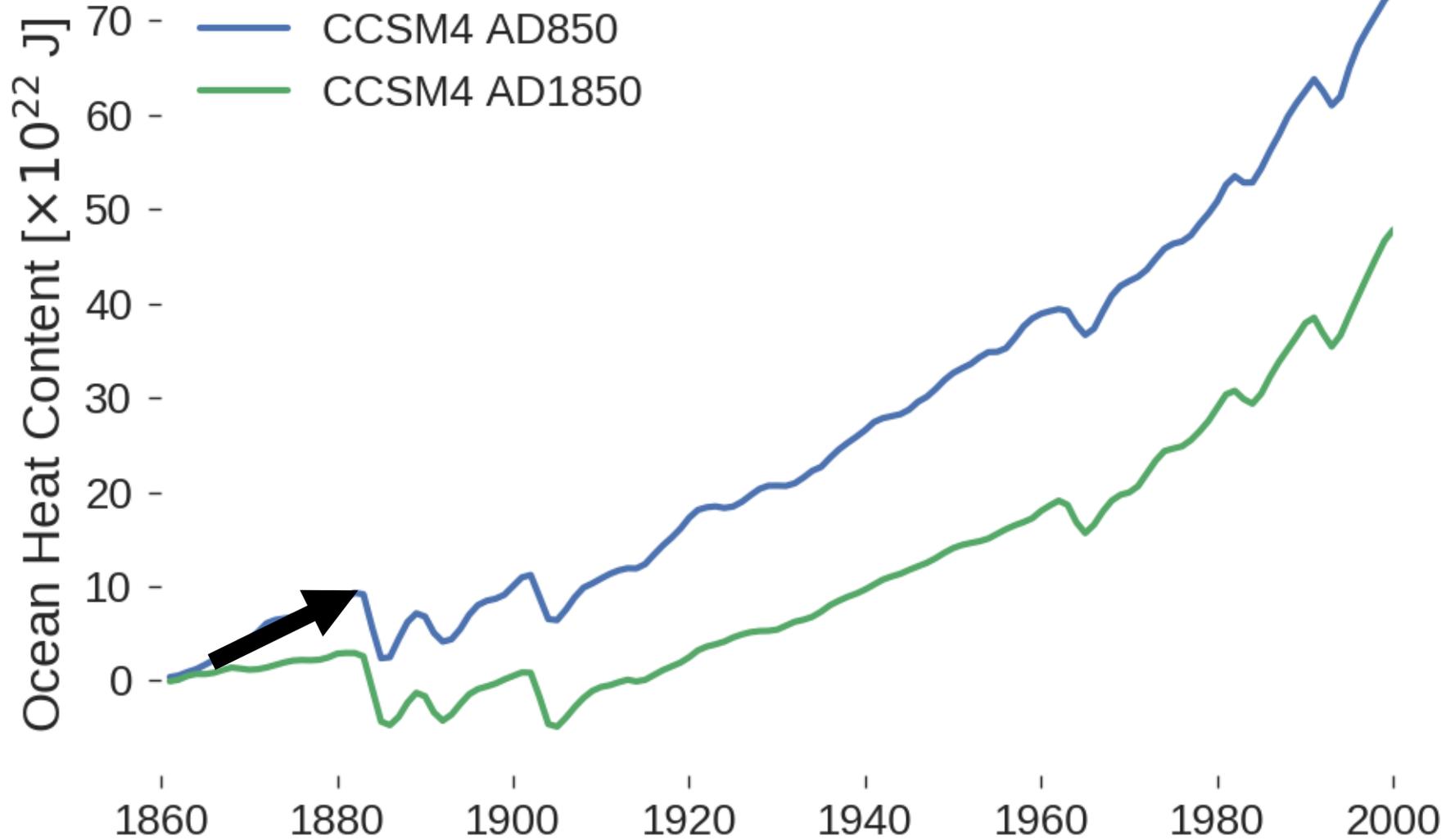
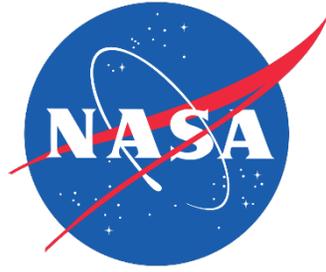
$$ECS \propto \frac{\Delta T}{\Delta F - \Delta N}$$

Source of Lewis & Curry early N



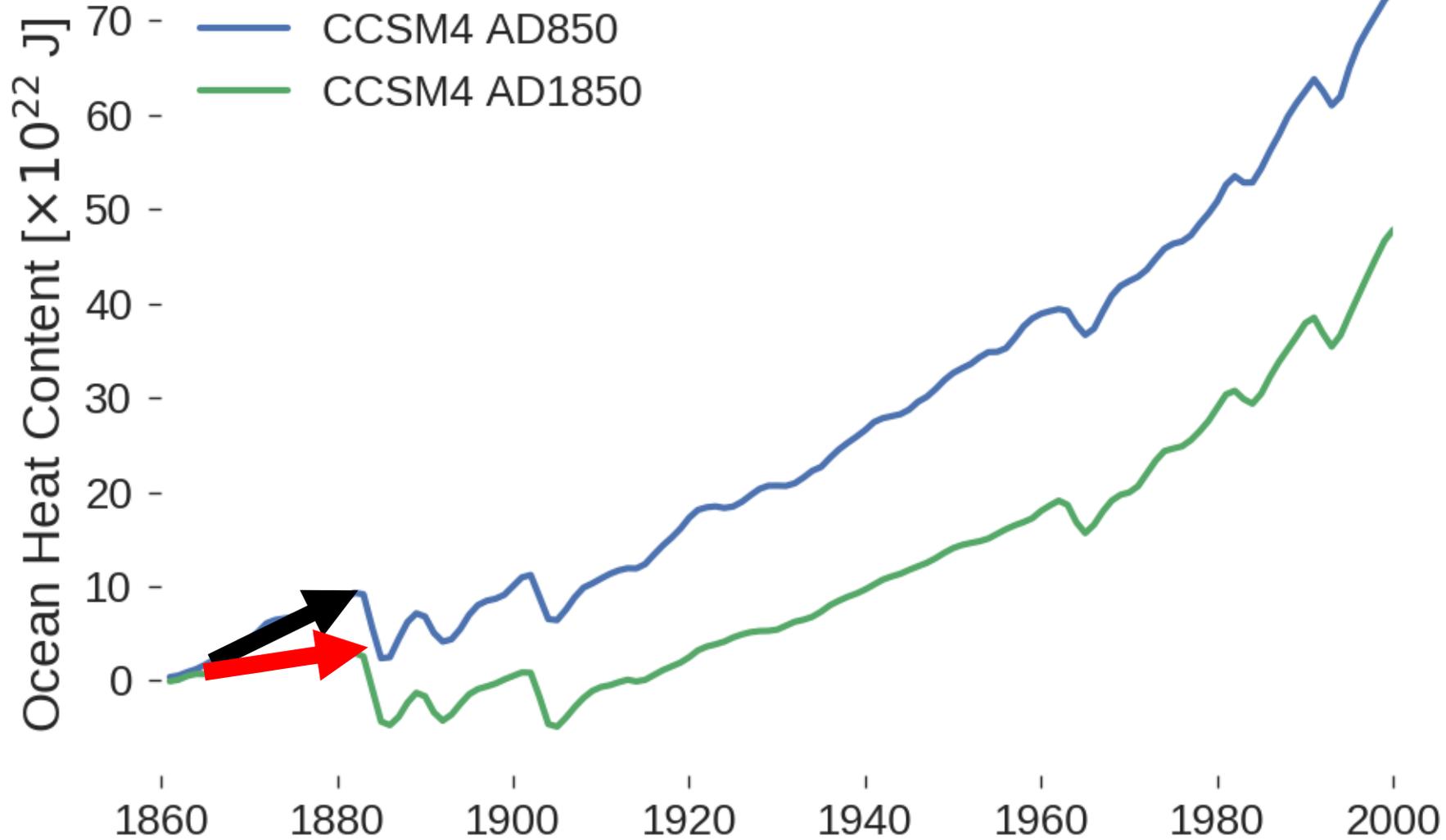
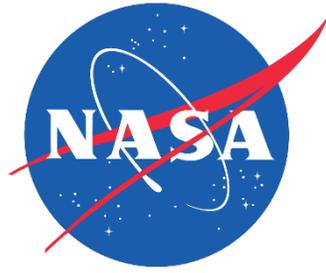
$$ECS \propto \frac{\Delta T}{\Delta F - \Delta N}$$

Source of Lewis & Curry early N



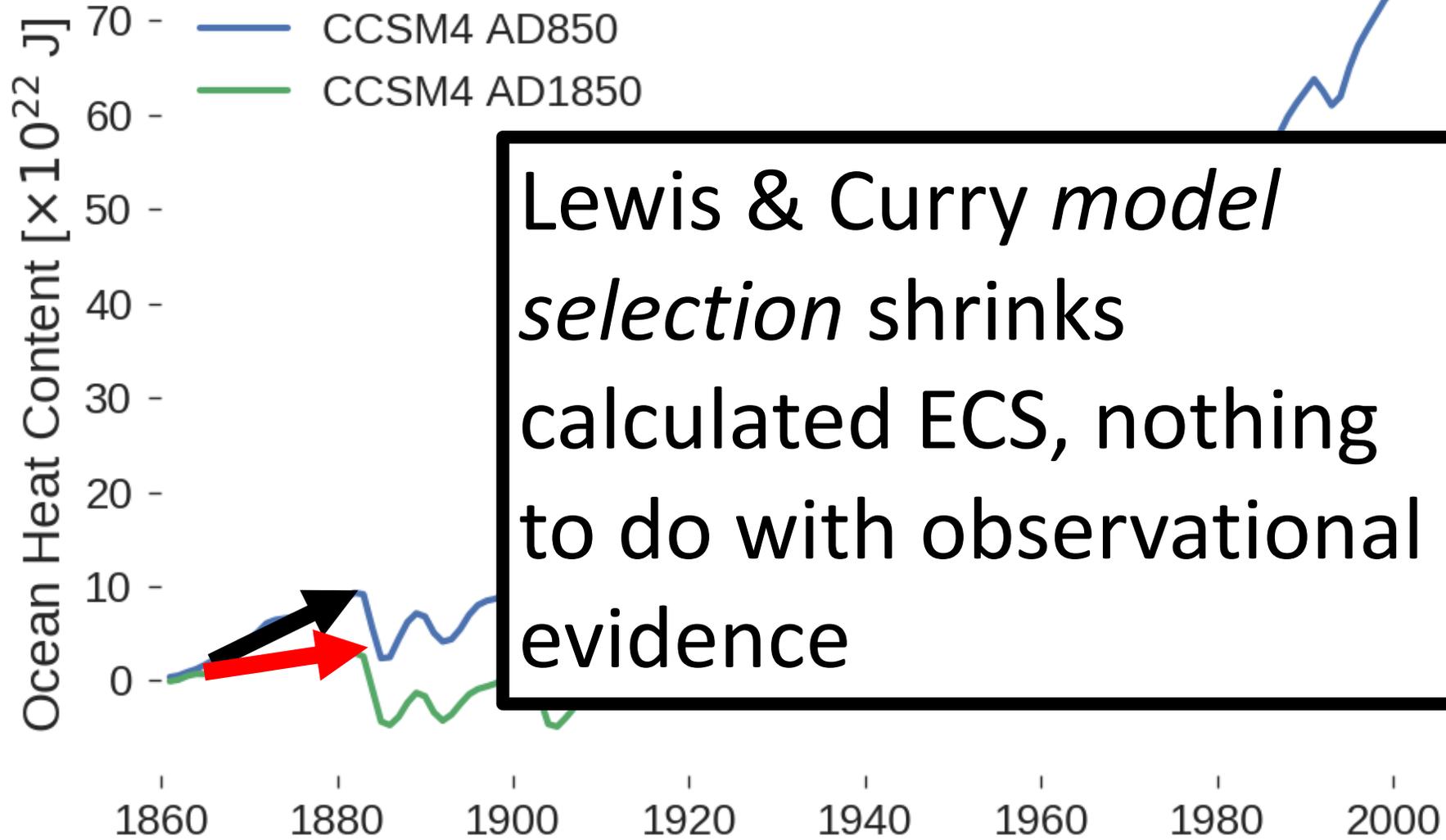
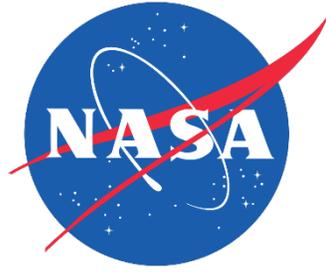
$$ECS \propto \frac{\Delta T}{\Delta F - \Delta N}$$

Source of Lewis & Curry early N

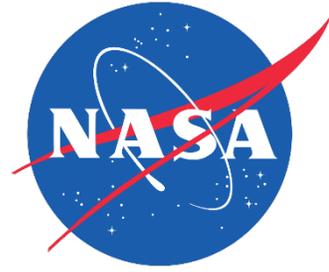


$$ECS \propto \frac{\Delta T}{\Delta F - \Delta N}$$

Source of Lewis & Curry early N



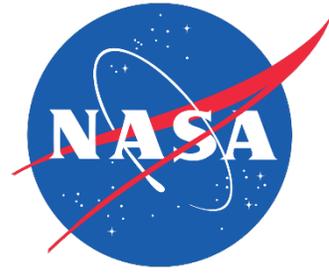
$$ECS \propto \frac{\Delta T}{\Delta F - \Delta N}$$



Let's do emergent constraint

- Use Lewis & Curry (2018) method on every CMIP5 model that has outputs

$$ECS_{hist} = \Delta F_{2 \times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$

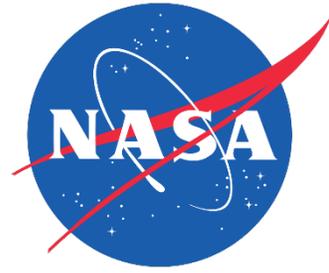


Let's do emergent constraint

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Forster et al. (2013)



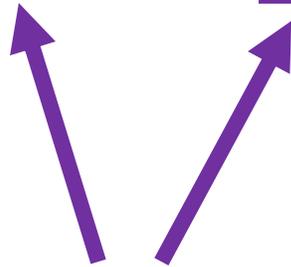
Let's do emergent constraint

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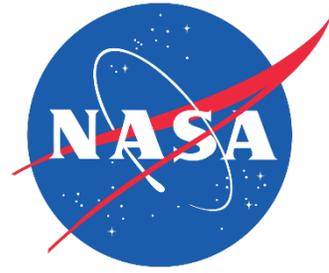
CMIP5 sampled like observations



$$ECS_{hist} = \Delta F_{2 \times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$



Forster et al. (2013)



Let's do emergent constraint

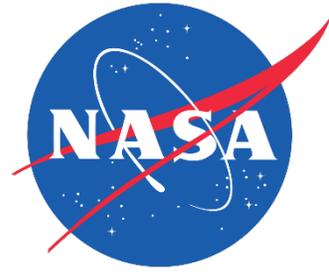
- Use Lewis & Curry (2018) method on every CMIP5 model that has outputs

CMIP5 sampled like observations

$$ECS_{hist} = \Delta F_{2 \times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$

2007—2016 = CMIP5 output

Forster et al. (2013)



Let's do emergent constraint

- Use Lewis & Curry (2018) method on every CMIP5 model that has outputs

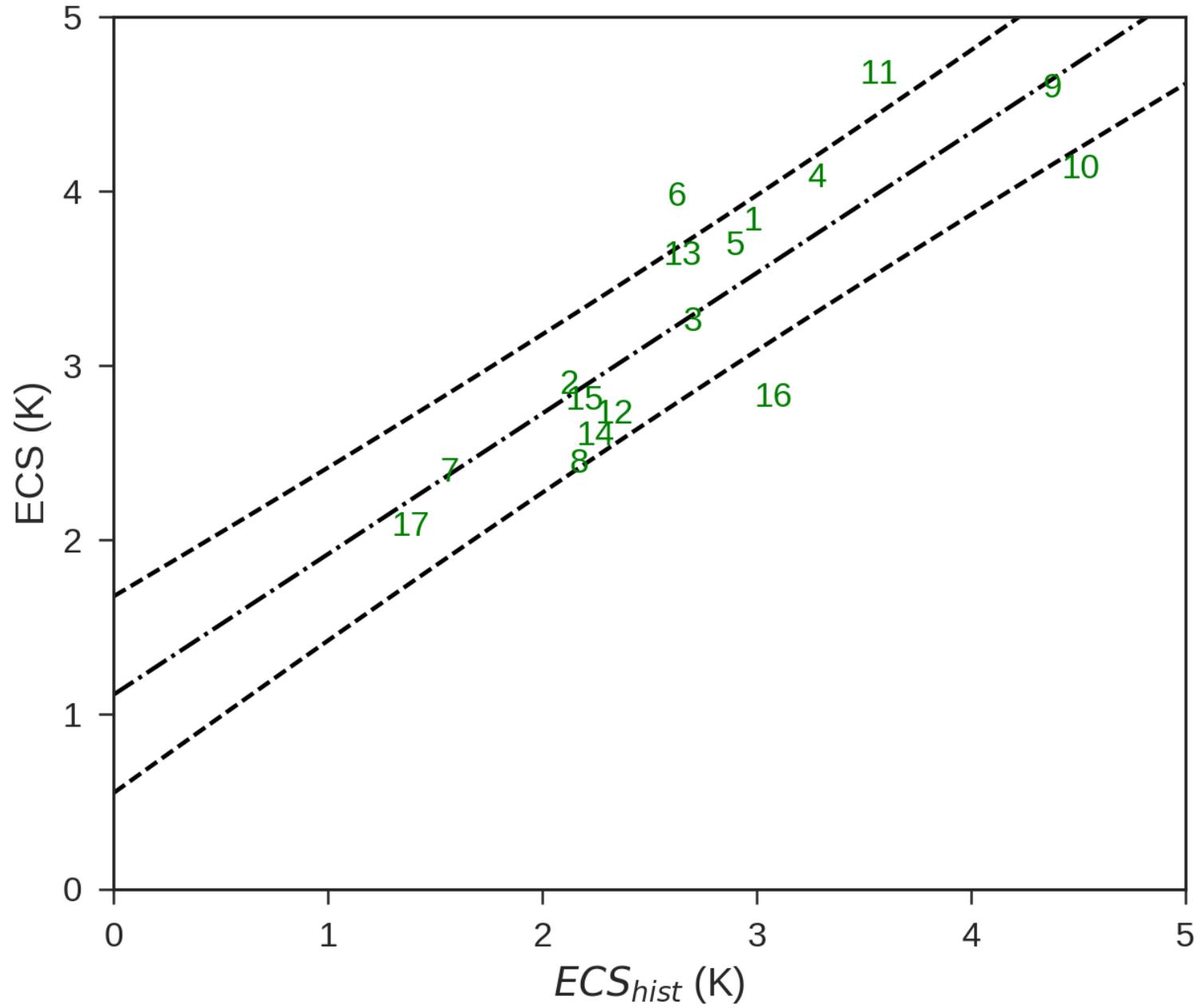
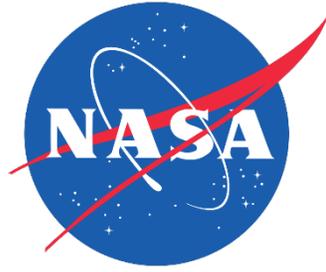
CMIP5 sampled like observations

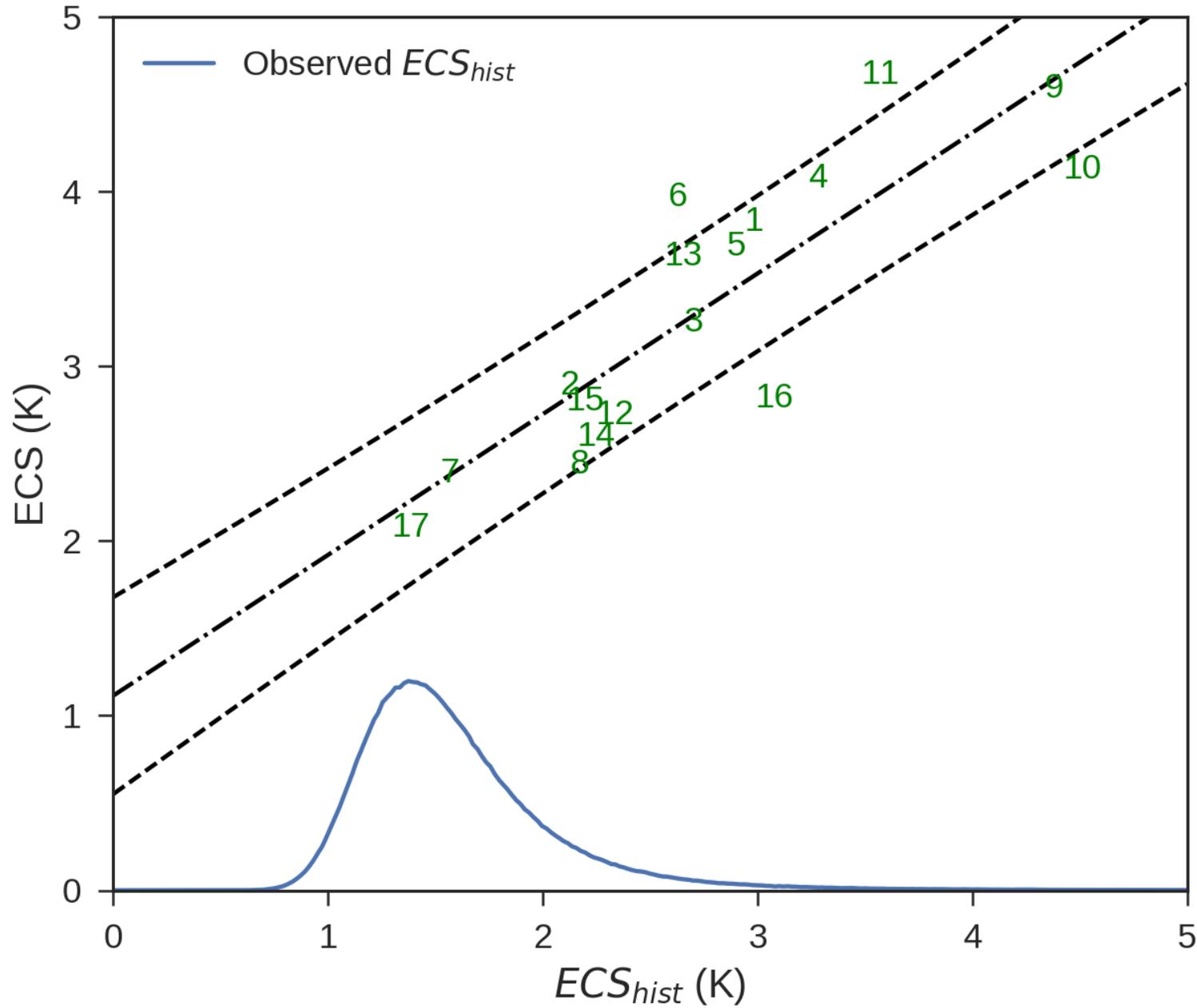
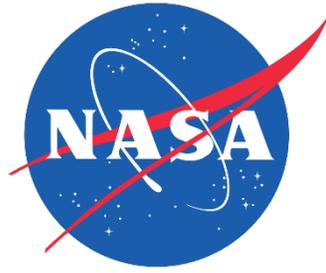
$$ECS_{hist} = \Delta F_{2 \times CO_2} \frac{\Delta T}{\Delta F - \Delta N}$$

2007—2016 = CMIP5 output

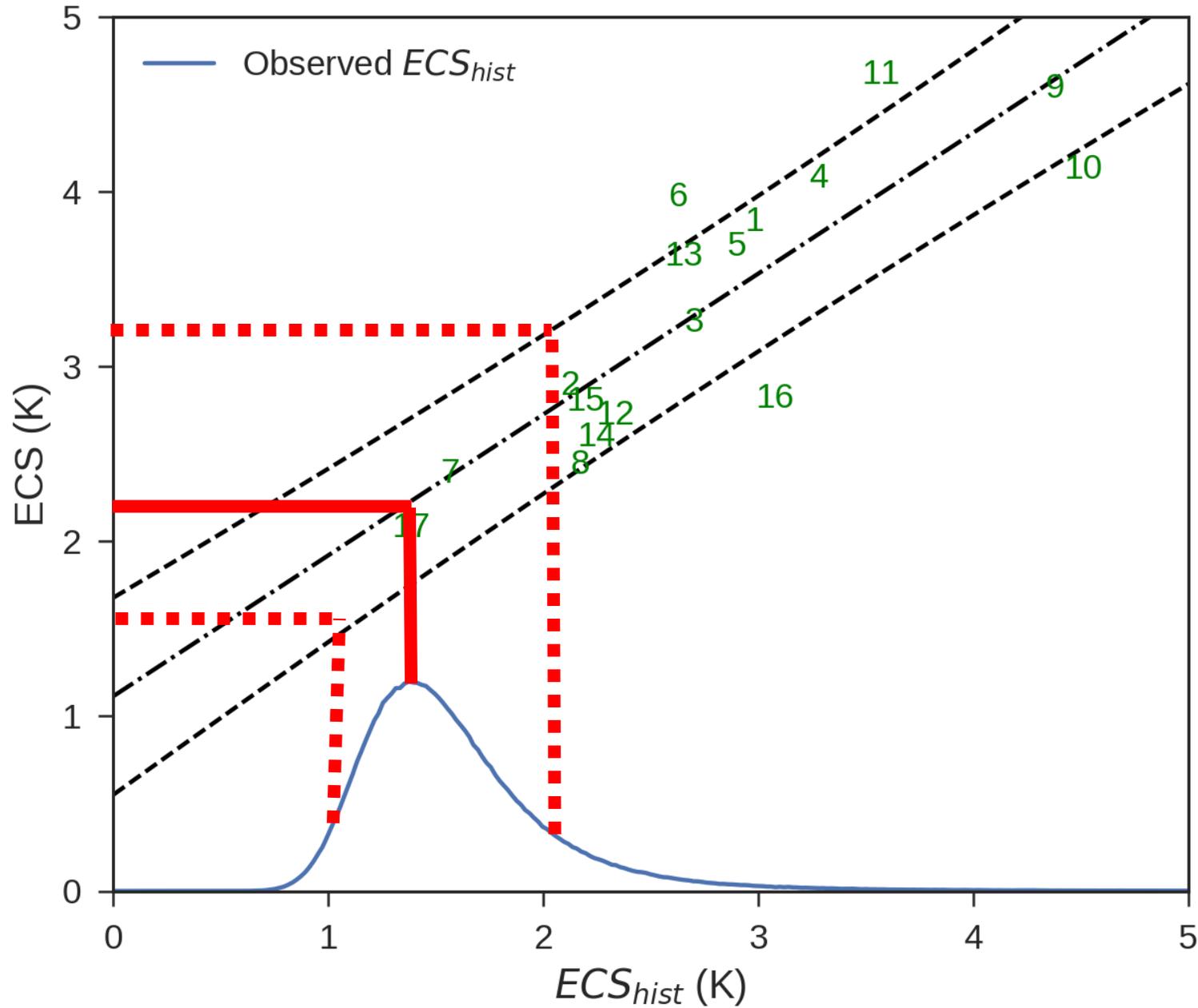
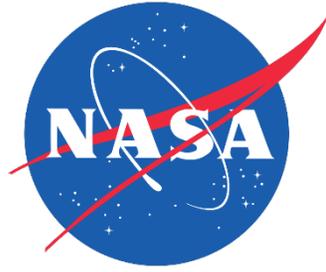
1869—1882 = Scaled CCSM4, like in Lewis & Curry

Forster et al. (2013)

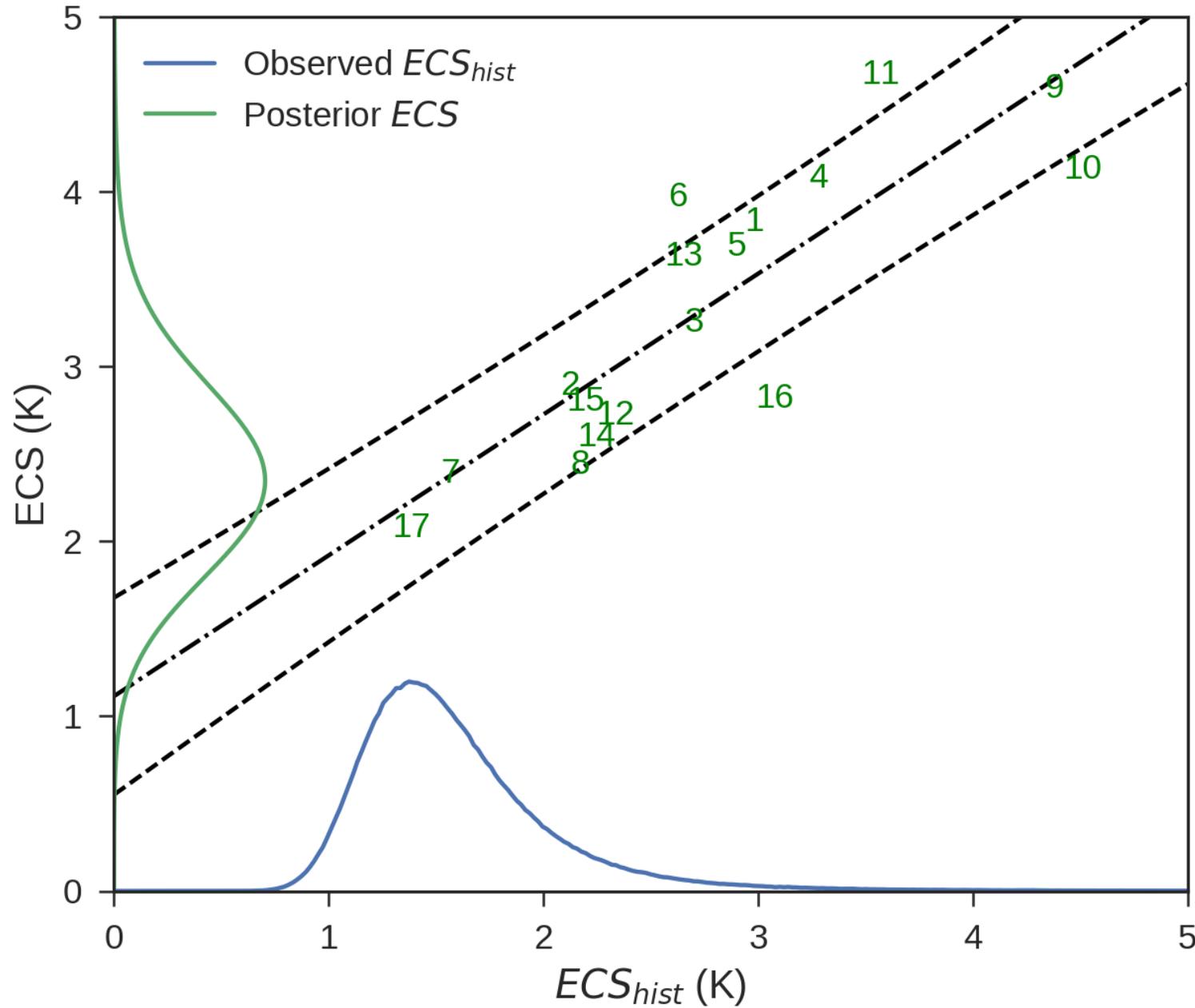
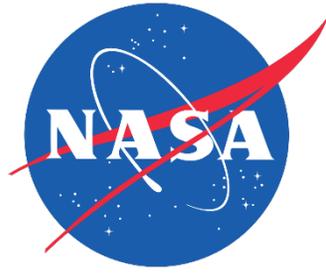




HadCRUT4 ECS_{hist}
1.5 ° (1.1—2.4 °C)

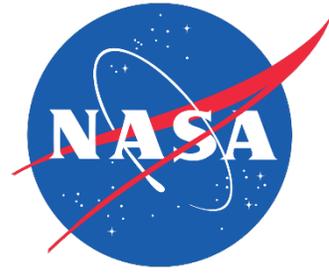


Apply Bayes' theorem with $P(ECS/ECS_{hist})$ from CMIP5 fit

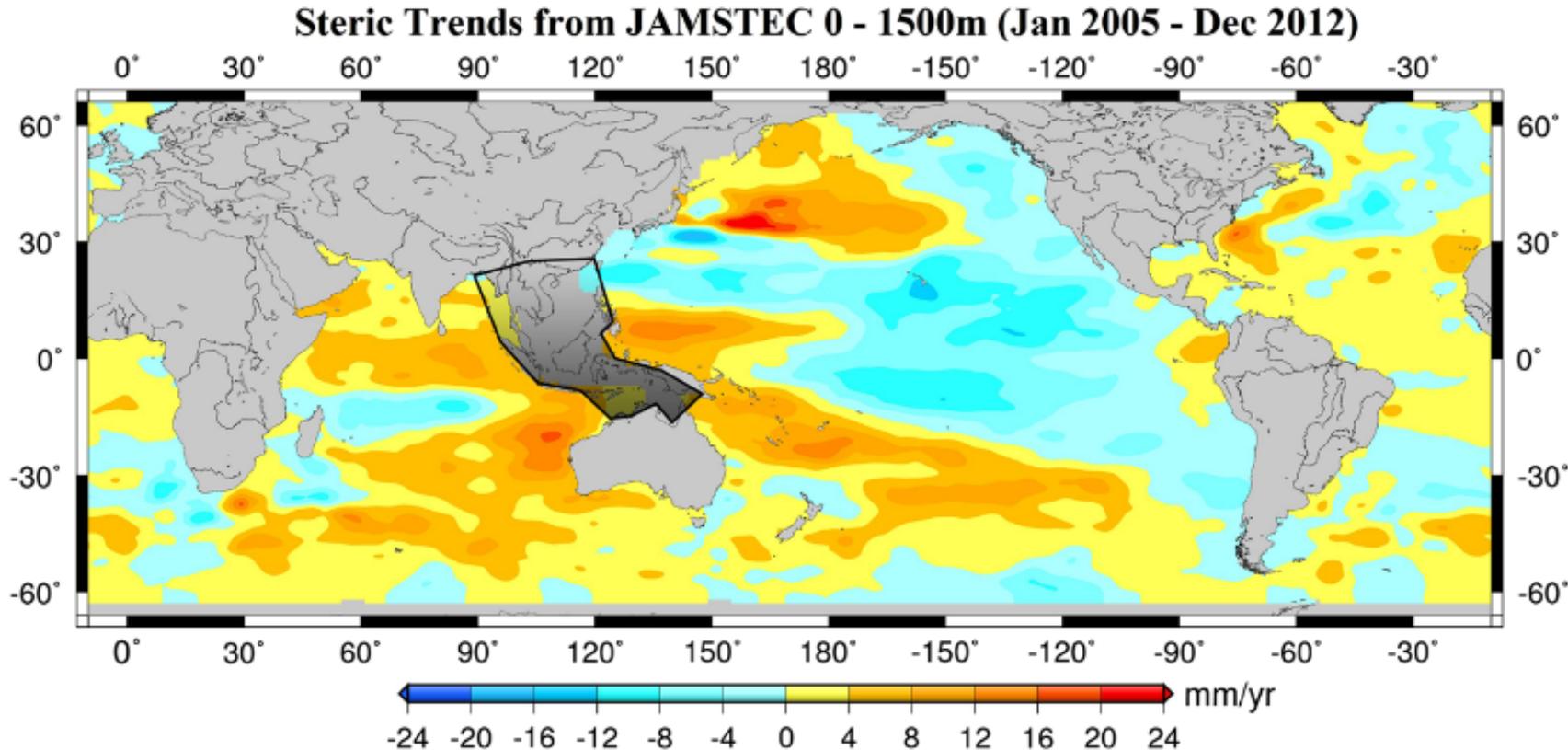


Posterior ECS from
HadCRUT4:
2.4 °C (1.5—3.4 °C)

(i.e. +60 % from
methodology)

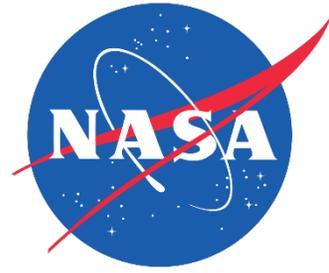


Recent heat imbalance from Argo floats



$$ECS \propto \frac{\Delta T}{\Delta F - \Delta N}$$

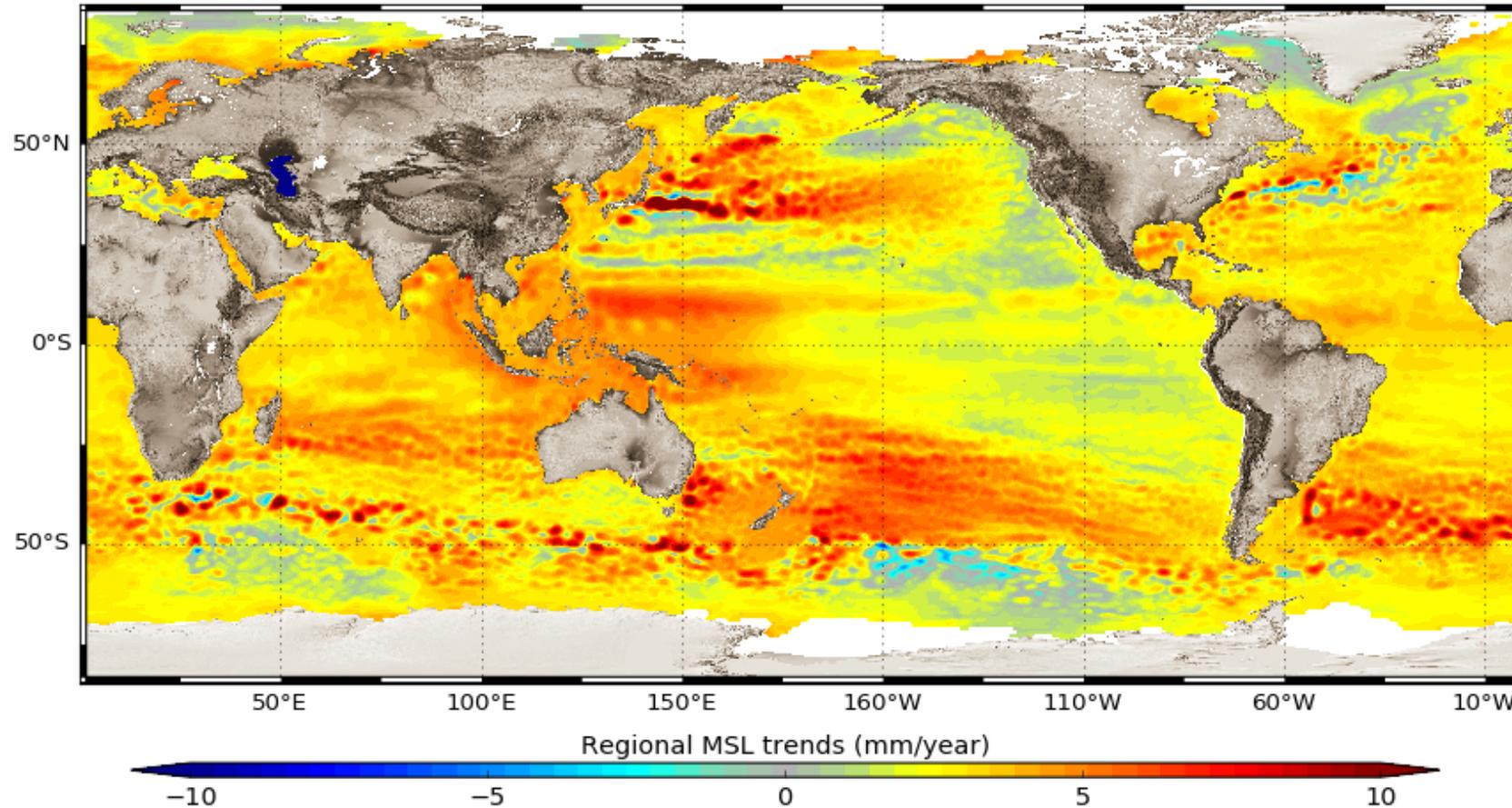
Fig. 10 Steric trend map based on Jamstec data over January 2005–December 2012 showing the Argo data gap in the Indonesian region and the contours (*black line*) of the area considered in this study to estimate—using the ORAS4 reanalysis—its contribution to the global mean steric trend



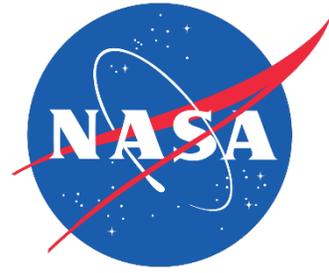
Sea level trends from altimetry '91—18

Multi-Mission Sea Level Trends

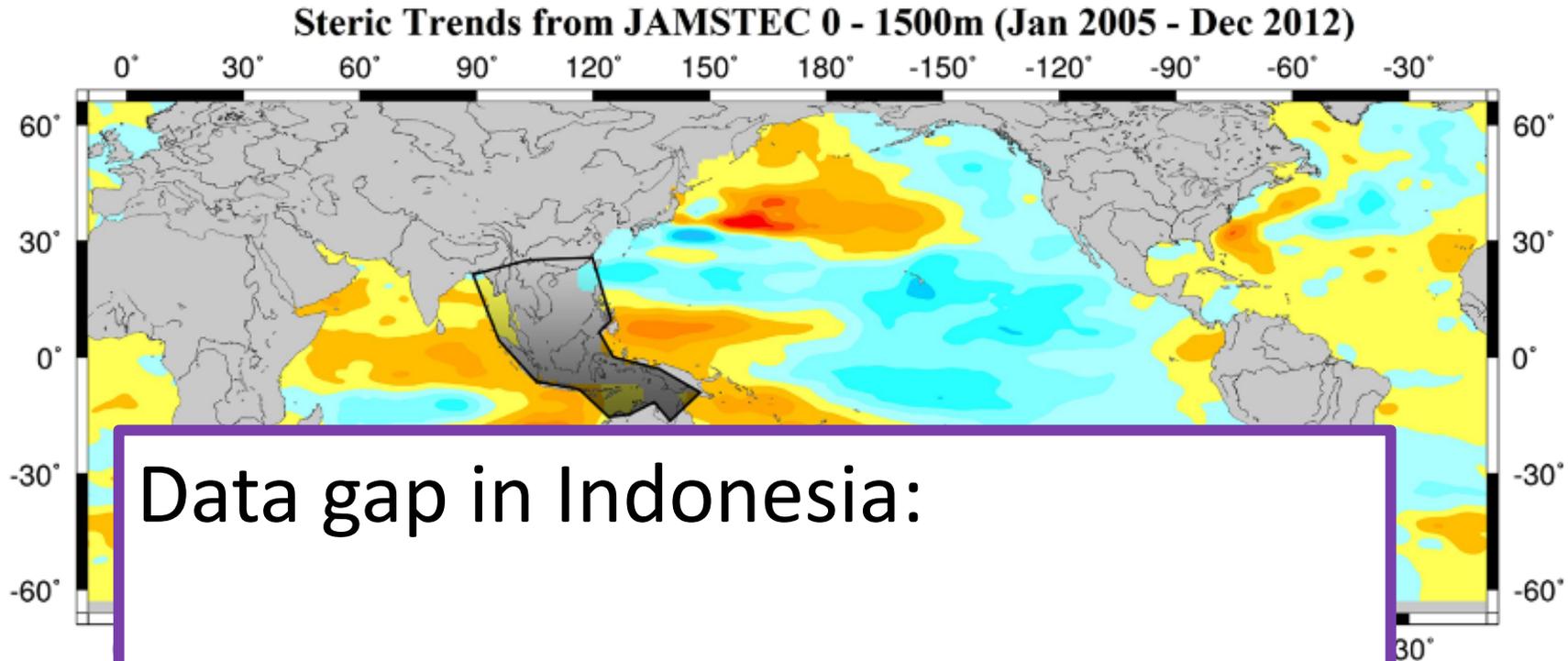
Period: Sep-1992 to Jan-2018



$$ECS \propto \frac{\Delta T}{\Delta F - \Delta N}$$



Recent heat imbalance from Argo floats



$$ECS \propto \frac{\Delta T}{\Delta F - \Delta N}$$

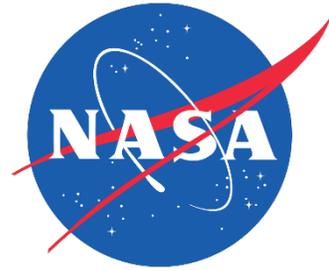
Data gap in Indonesia:
Gravimetry + altimetry suggests heat uptake here worth $\sim 0.13 \text{ W m}^{-2}$ globally

Fig. 10
gap in the
using the

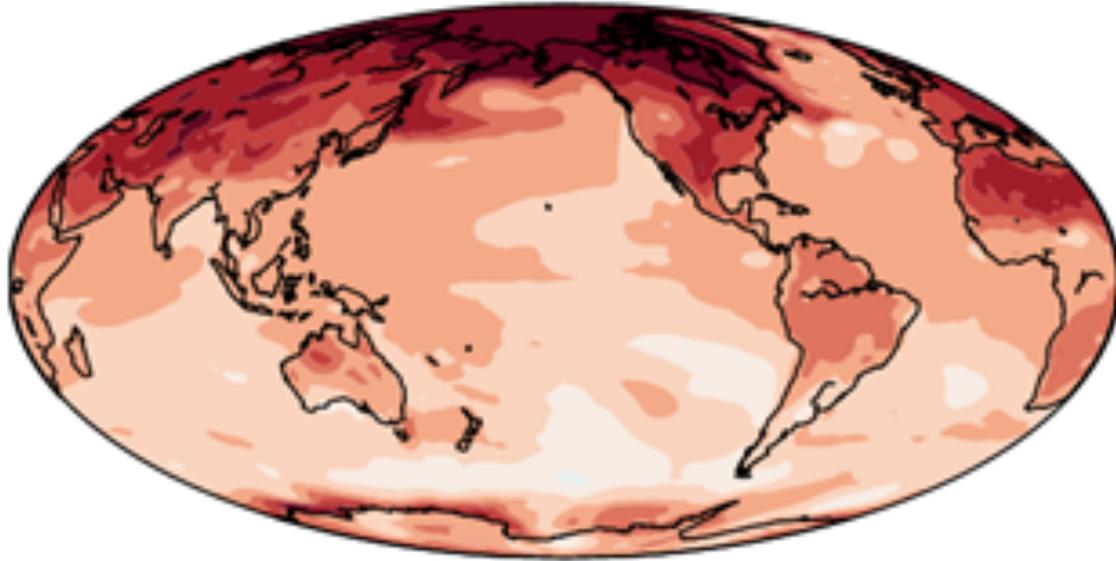
the Argo data
to estimate—

1980—2005

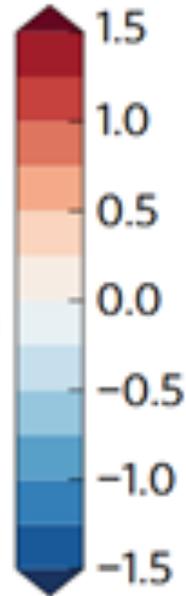
Zhou, Zelinka & Klein <http://dx.doi.org/10.1038/ngeo2828>



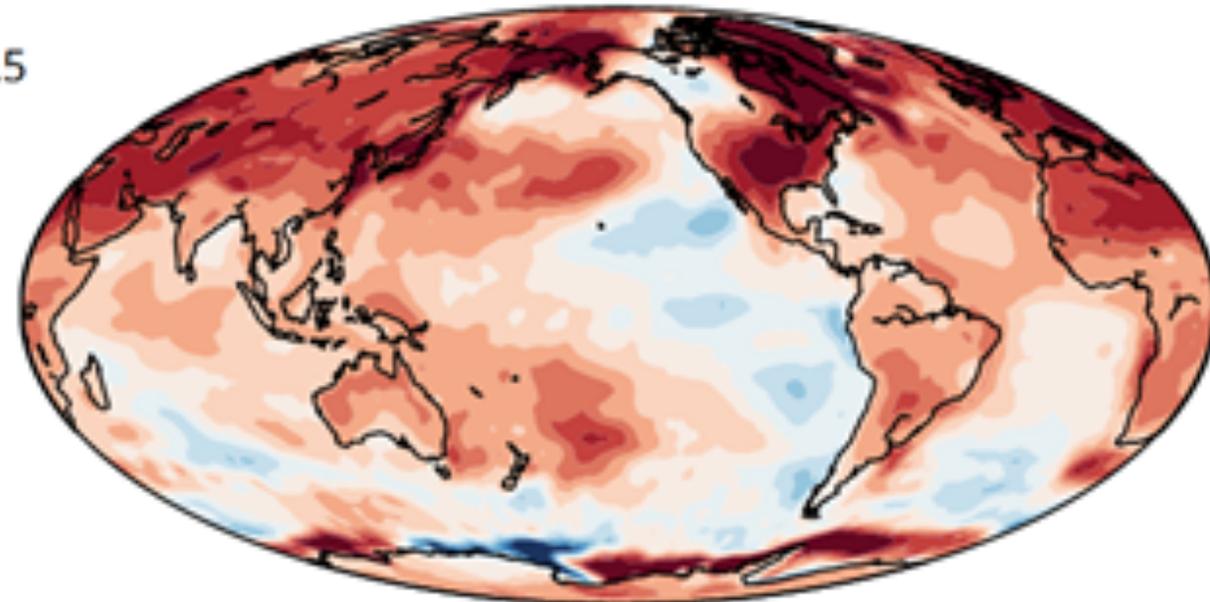
CMIP5-historical T_s trend



K per 30 yr

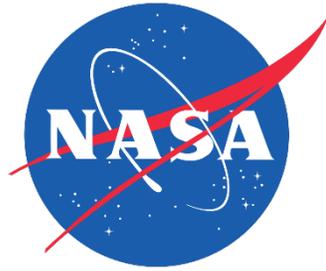


AMIP T_s trend

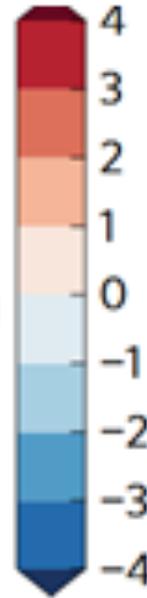
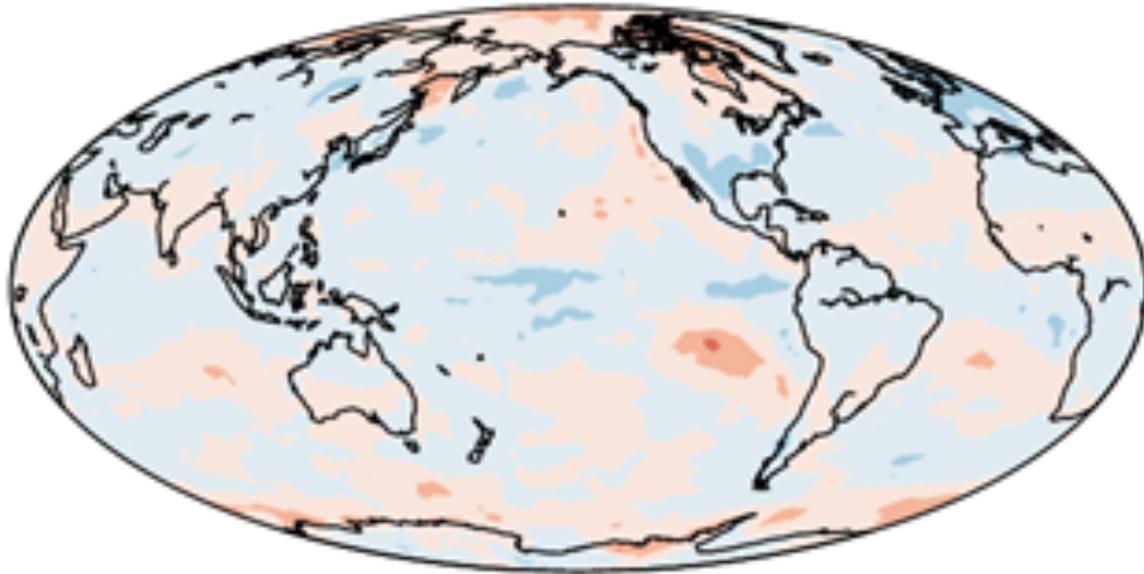


1980—2005

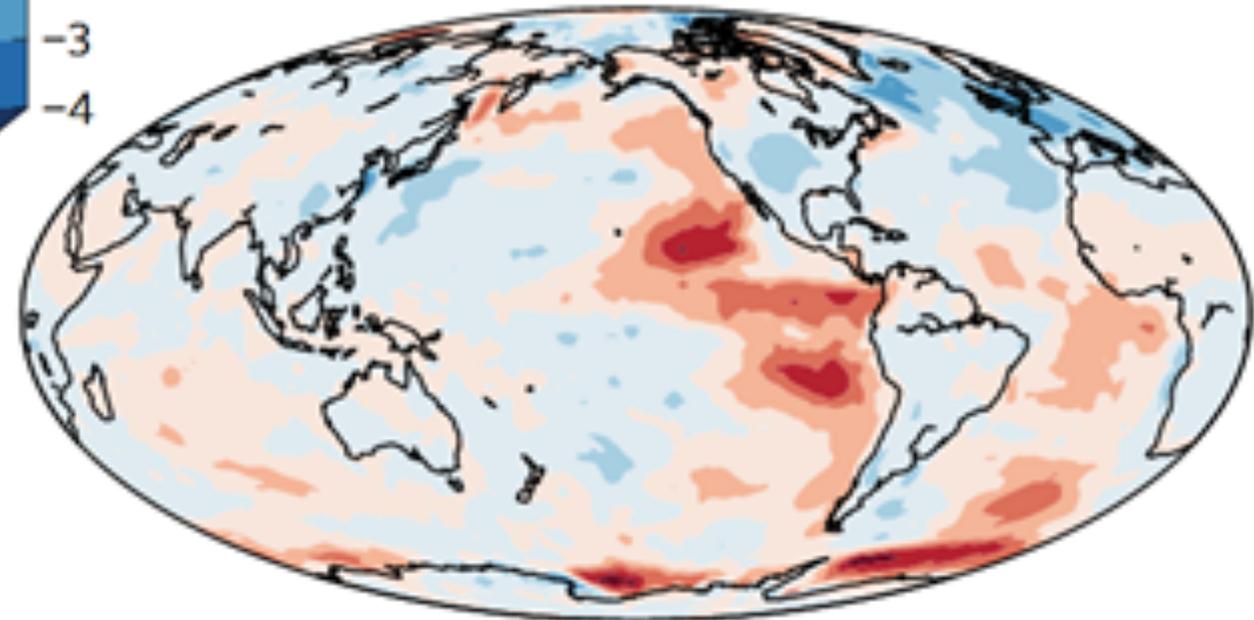
Zhou, Zelinka & Klein <http://dx.doi.org/10.1038/ngeo2828>



CMIP5-historical LCC trend

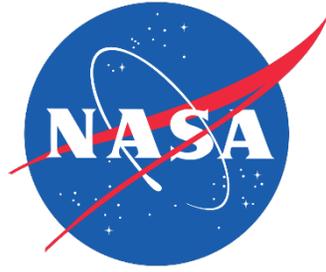


AMIP LCC trend

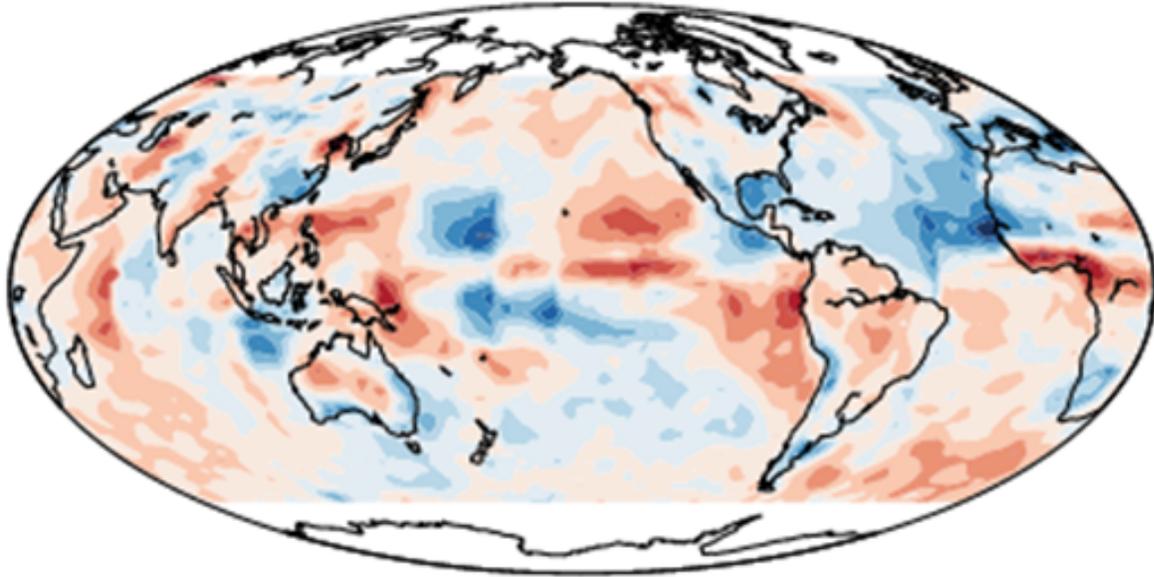


1980—2005

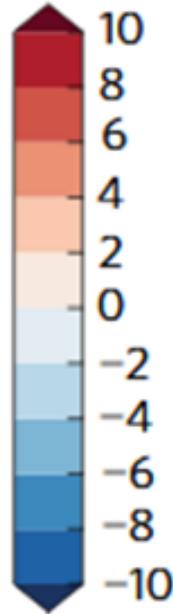
Zhou, Zelinka & Klein <http://dx.doi.org/10.1038/ngeo2828>



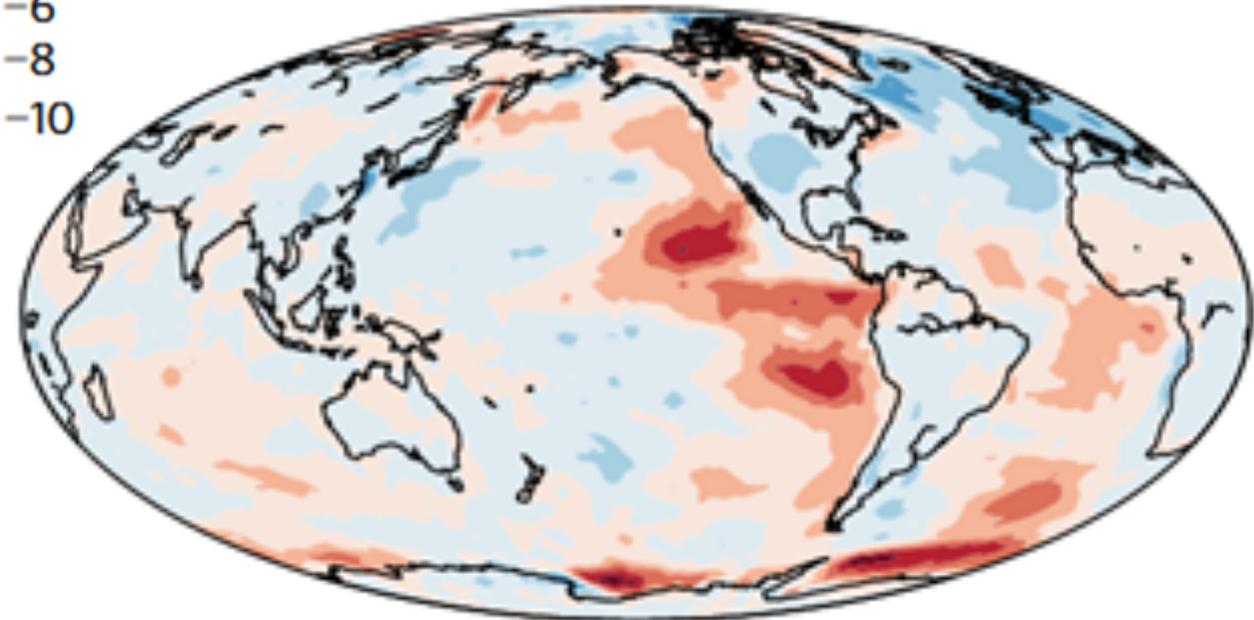
ISCCP LCC trend



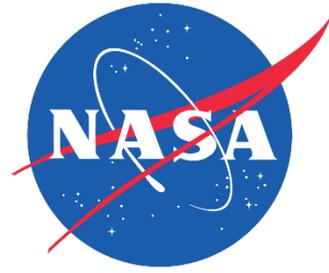
% per 30 yr



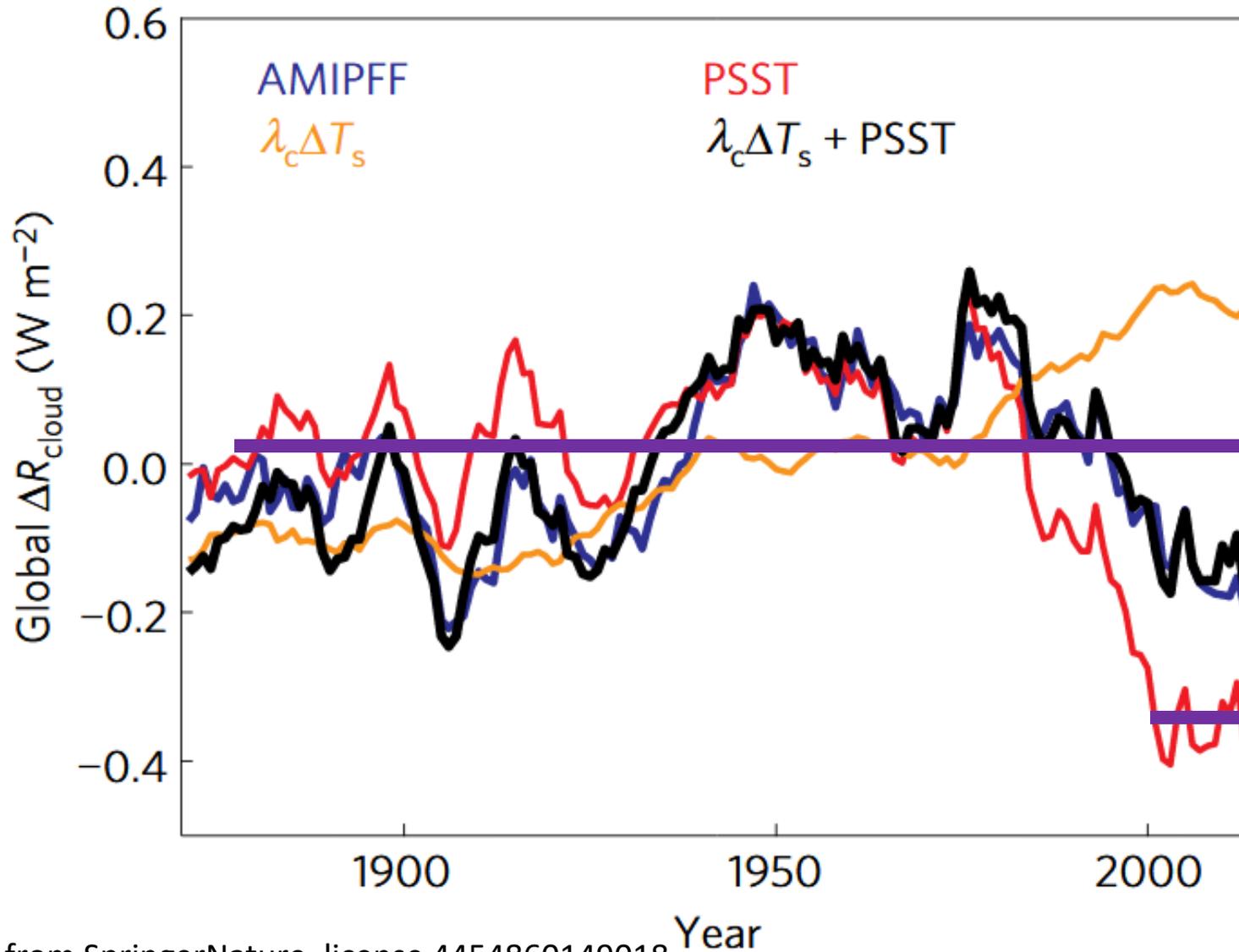
AMIP LCC trend



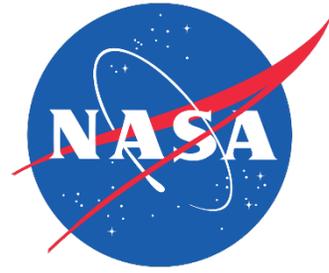
RED is “pattern difference from average”



a



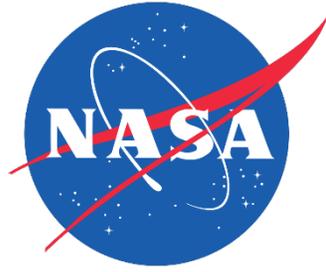
ΔN due to
cloud pattern
 $>0.35 \text{ W m}^{-2}$



- Lewis & Curry $ECS_{hist} = 1.5 \text{ }^\circ\text{C}$ maps to $ECS_{true} = 2.4 \text{ }^\circ\text{C}$
- We must:
 - 1) Make sure heat uptake data aren't biased
 - 2) Work out cloud pattern effects
 - 3) Improve ΔF comparison

Dieng or Zhou ΔN effects change ECS_{true} to 2.5—3.0 $^\circ\text{C}$

ΔF could be bigger shift.



Conclusions

- When you see an energy budget study on ECS, look for three things:
 - How do they select their data and time periods?
 - Do they assume $\lambda = \text{constant}$?
 - Do they test their *exact calculation* on climate models and check if it works when we know answer?
- Doing this for Lewis & Curry suggests 2.4 °C+, not 1.5 °C