The shape of things to come:

Why is climate sensitivity so unpredictable, and who cares anyway?

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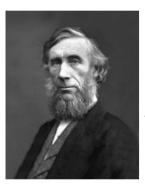
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Climate change: not a new problem



Joseph Fourier, 1827:

On the temperatures of the terrestrial sphere and interplanetary space.



John Tyndall, 1860s:

"The solar heat possesses. . . the power of crossing an atmosphere; but, when the heat is absorbed by the planet, it is so changed in quality that the rays emanating from the planet cannot get with the same freedom back into space. Thus the atmosphere admits of the entrance of the solar heat, but checks its exit; and the result is a tendency to accumulate heat at the surface of the planet."

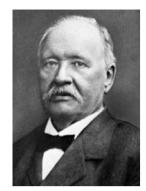


Samuel Langley, 1870s+

- Measurement of the direct effect of sun spots on terrestrial climates. Monthly Notices Roy. Astron. Soc, 1876.
- On the amount of atmospheric absorption. Amer. Journ. Sci., 1883.
- On hitherto unrecognized wavelengths. Amer. Journ. Sci., 1886.

Also Pouillet (1838); De Marchi (1895); Milankovitch (1924)...

Climate change: not a new problem



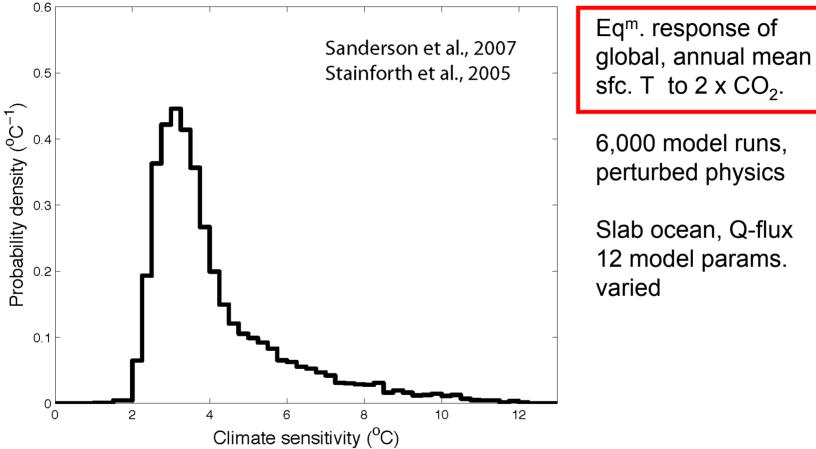
Svante Arrhenius, 1896:

THE LONDON, EDINBURGH, AND DUBLIN PHILOSOPHICAL MAGAZINE AND JOURNAL OF SCIENCE. [FIFTH SERIES.] APRIL 1896. XXXI. On the Influence of Carbonic Acid in the Air upon the Temperature of the Ground. By Prof. SVANTE ABRHENIUS

> I. Introduction : Observations of Langley on Atmospherical Absorption.

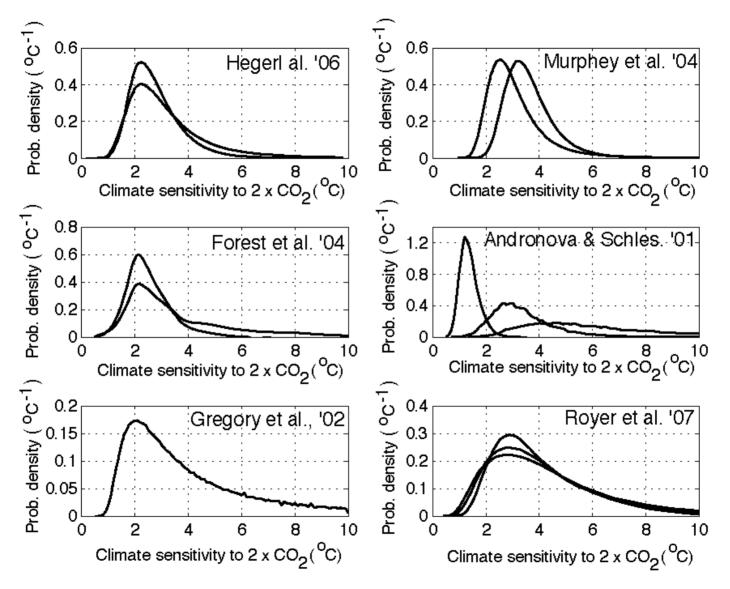
Climate sensitivity: an envelope of uncertainty

Climateprediction.net 200,000+ integrations, 31,400,000 yrs model time(!);



- Two questions:
 - 1. What governs the shape of this distribution?
 - 2. How does uncertainty in physical processes translate into uncertainty in climate sensitivity?

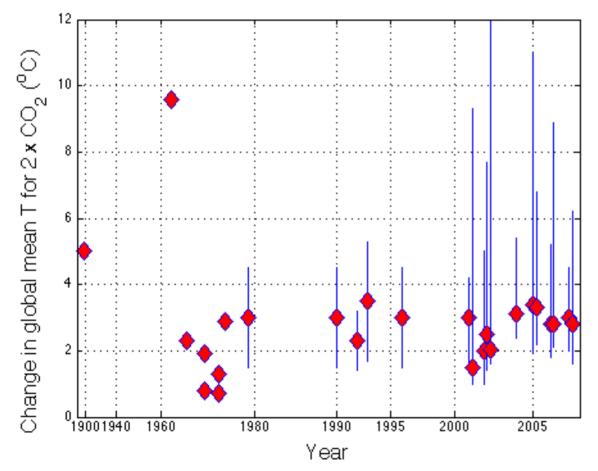
Climate sensitivity: an envelope of uncertainty



• Wide variety of models, methods, and reconstructions.

Climate sensitivity: estimates over time

Climate sensitivity = Equilibrium change in global mean, annual mean temperature given $CO_2 \rightarrow 2 \times CO_2$

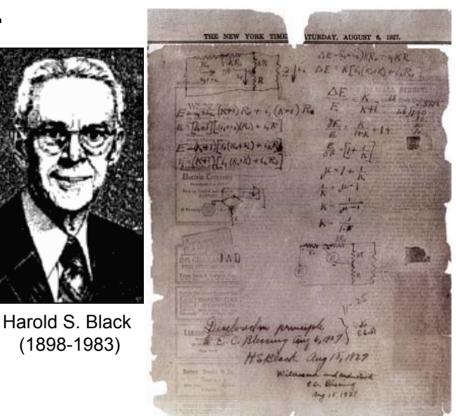


1. Arrhenius, 1896 2. Moller, 1963 3. Weatherald and Manabe, 1967 4. Manabe, 1971 5. Rasool and Schneider, 1971 6. Manabe and Weatherald, 1971 7. Sellers, 1974 8. Weare and Snell, 1974 9. NRC Charney report, 1979 10. IPCC1, 1990 11. Hoffert and Covey, 1992 12. IPCC2, 1996 13. Andronova & Schlesinger, 2001 14. IPCC3, 2001 15. Forest et al., 2002 16. Harvey & Kaufmann, 2002 17. Gregory et al., 2002 18. Murphy et al., 2004 19. Piani et al., 2005 20. Stainforth et al., 2005 21. Forest et al., 2006 22. Hegerl et al. 2006 23. IPCC4, 2007 24. Royer et al., 2007

• Why is uncertainty not diminishing with time?

Need feedback analysis!

- introduced concept of negative feedback.
- got the idea on Lackawanna ferry on his way to work.
- took nine years to get granted a patent.



- "Our patent application was treated in the same manner one would a perpetual motion machine" Black, H.S. IEEE Spectrum, 1977
- Original notes scribbled on NY Times

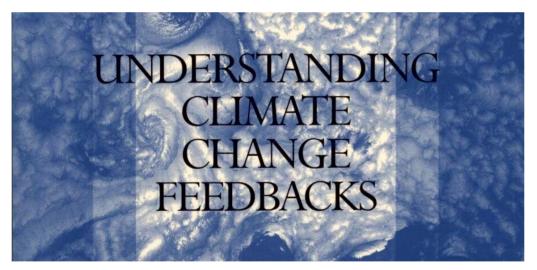
• Formalized framework for the evaluation of interactions in dynamical systems.

Feedback analysis:

Formal framework for evaluating the strength and relative importance of interactions in a dynamical system.

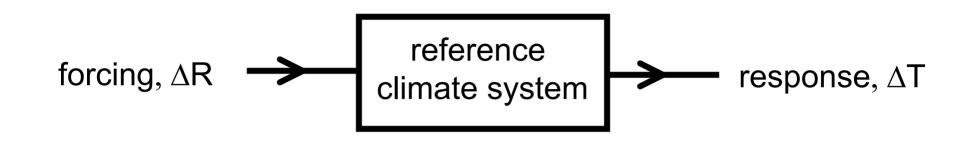
(Maxwell, 1863; Black, 1927; Cess, 1975; Charney et al., 1979; Hansen et al., 1984; Schlesinger & Mitchell, 1985)

Confusion abounds....



U.S. National Research Council report, 2003

• gets definitions of feedbacks wrong...



Climate sensitivity *parameter* defined by: $\Delta T_0 =$

 $\Delta \mathsf{T}_0 = \lambda_0 \, \Delta \mathsf{R}$

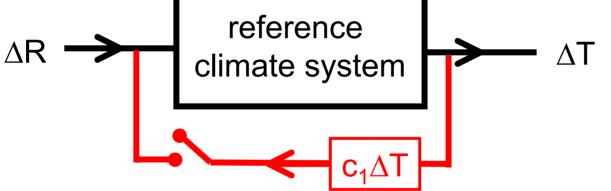
Reference climate system:

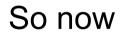
- Blackbody (i.e., no atmosphere).
- Terrestrial flux = σT^4 (Stefan-Boltzmann)
- $\lambda_0 = (4\sigma T^3)^{-1} = 0.26 \text{ K } (\text{Wm}^{-2})^{-1}$

$\Rightarrow \Delta T_0 = 1.2 \circ C$ for a doubling of CO_2



(n.b. Feedbacks are only meaningful when defined against a reference state.)

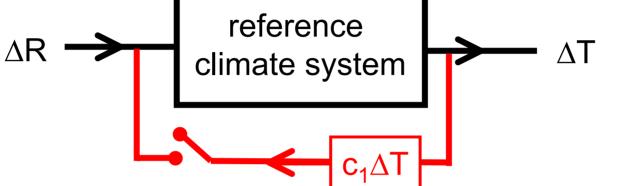




$$\Delta \mathsf{T} = \lambda_0 (\Delta \mathsf{R} + \mathbf{c_1} \Delta \mathsf{T})$$



(n.b. Feedbacks are only meaningful when defined against a reference state.)



So now
$$\Delta T = \lambda_0 (\Delta R + c_1 \Delta T)$$

Additional radⁿ forcing
due to system
response to ΔR



(n.b. Feedbacks are only meaningful when defined against a reference state.)

$$\Delta R$$
 reference
climate system ΔT

So now
$$\Delta T = \lambda_0 (\Delta R + c_1 \Delta T)$$

Rearrange for ΔT
 $\Rightarrow \quad \Delta T = \frac{\lambda_0 \Delta R}{1 - c_1 \lambda_0}$
Additional radⁿ forcing due to system response to ΔR

Feedback analysis: technobabble

Feedback factor:
$$f = c_1 \lambda_0$$

(f ∞ to fraction of output *fed back* into input)

(Gain is proportion by which system has *gained*)

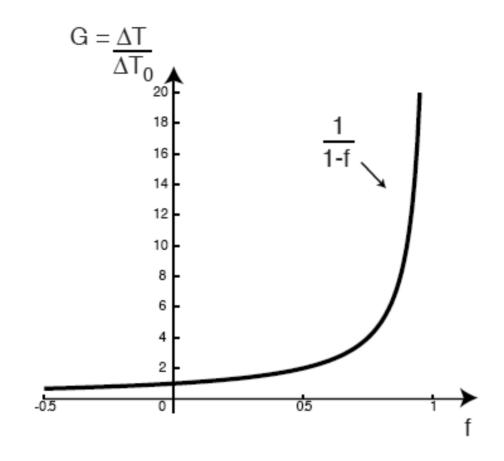
Gain = $\frac{\text{response with feedback}}{\text{response without feedback}} = \frac{\Delta T}{\Delta T_0}$

From before
$$\Delta T = \frac{\lambda_0 \Delta R}{1 - c_1 \lambda_0} = \frac{\Delta T_0}{1 - f}$$

And since $\Delta T = G \Delta T_0$:

$$G = \frac{1}{(1-f)}$$

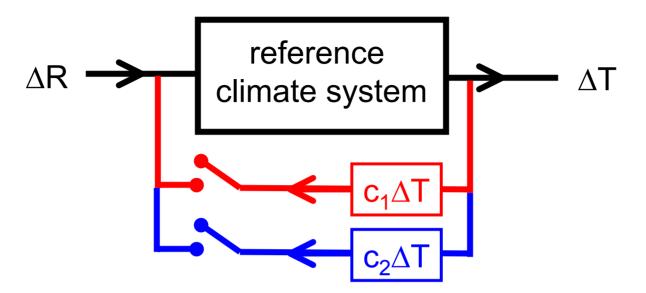
Feedbacks: gain curve



Range of possibilities:

 $\begin{array}{ll} -\infty < f < 0 & : & G < 1 \implies response \ damped \implies \mathsf{NEGATIVE} \ fdbk. \\ 0 < f < 1 & : & G > 1 \implies response \ amplified \implies \mathsf{POSITIVE} \ fdbk. \\ f > 1 & : & G \ undef. \implies \mathsf{Planet} \ explodes... \end{array}$

Feedback analysis: more than one feedback

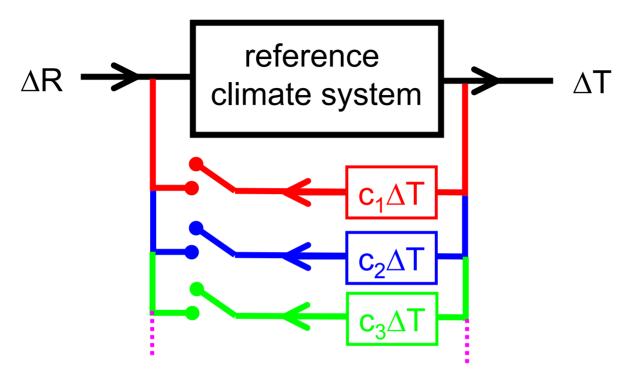


Now have $\Delta T = \lambda_0 (\Delta R + c_1 \Delta T + c_2 \Delta T)$ (two nudges)

Gives:

$$\Delta \mathsf{T} = \frac{\lambda_0}{1 - c_1 \lambda_0 - c_2 \lambda_0} \Delta \mathsf{R}$$

Feedback analysis: more than one feedback



And so in general for N feedbacks:

$$\mathbf{G} = \frac{\Delta \mathbf{T}}{\Delta \mathbf{T}_0} = \frac{1}{1 - \sum_{i=1}^{N} \mathbf{f}_i}$$

Climate feedbacks: calculating from models

Want to consider effect of variations in:

a) water vapor; b) clouds; c) sea-ice; d) snow cover; etc..

For ith climate variable:
$$c_{i}\Delta$$

$$\mathbf{c}_{i}\Delta \mathbf{T} = \delta \mathbf{R} \mathbf{)}_{j, j \neq i} = \frac{\partial \mathbf{R}}{\partial \alpha_{i}} \mathbf{)}_{j, j \neq i} \frac{\mathbf{d}\alpha_{i}}{\mathbf{d}\mathbf{T}} \Delta \mathbf{T}$$

1

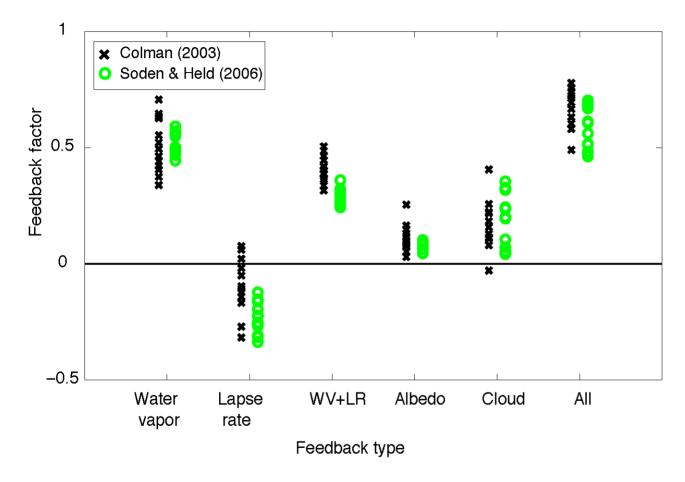
So feedback factors:

$$\mathbf{f}_{i} \approx \lambda_{0} \frac{\Delta \mathbf{R}}{\Delta \alpha_{i}} \bigg|_{j, j \neq i} \cdot \frac{\Delta \alpha_{i}}{\Delta \mathbf{T}}$$

- $\alpha_{i}\,$ can be a lumped property (like clouds, sea ice, etc.),
 - or individual model parameter (like entrainment coefficient)
 - can also calculate spatial variations in f_i if desired.

Climate feedbacks: estimating from models

From suites of GCMS:



Individual feedbacks uncorrelated among models, so can be simply combined:

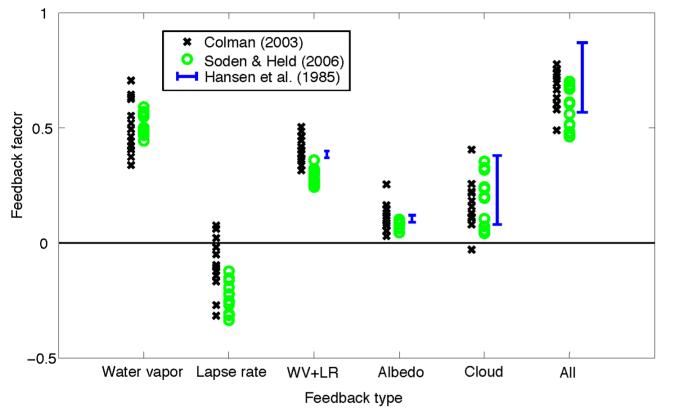
Soden & Held (2006): $\bar{f} = 0.62; \sigma_f = 0.13$

Colman (2003): $\bar{f} = 0.70; \sigma_{f} = 0.14$

 How does this uncertainty in physics translate to uncertainty in climate sensitivity?

Climate feedbacks: estimating from models

From suites of GCMS:

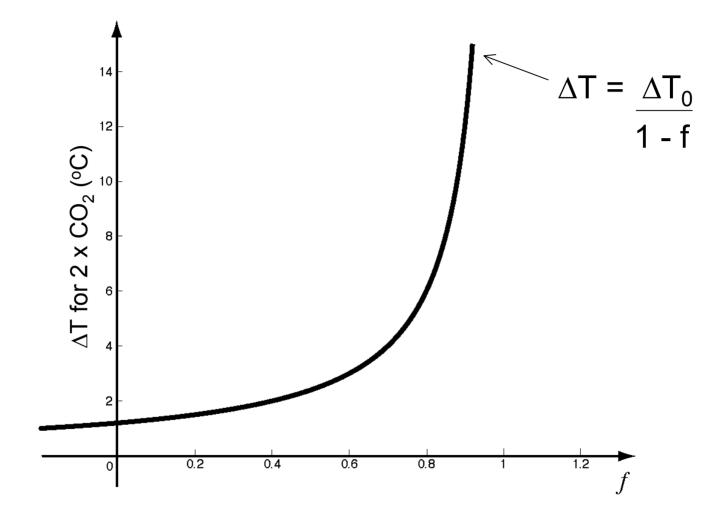


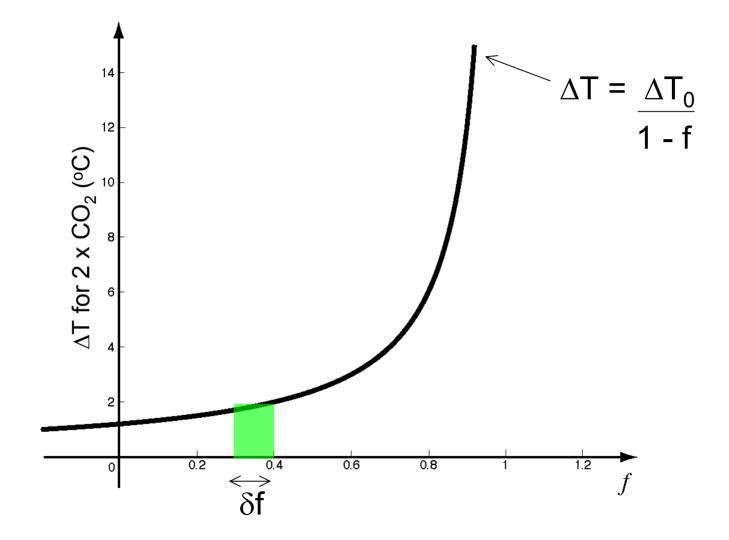
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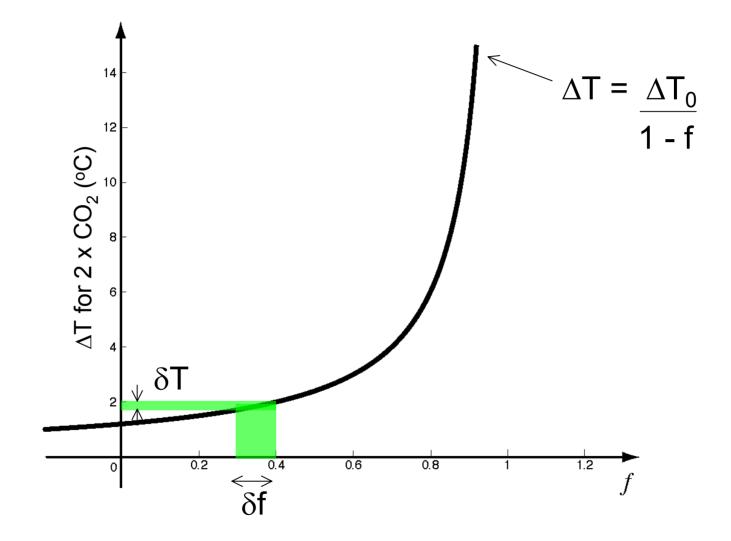
Soden & Held (2006): $\bar{f} = 0.62; \, \sigma_{\rm f} = 0.13$

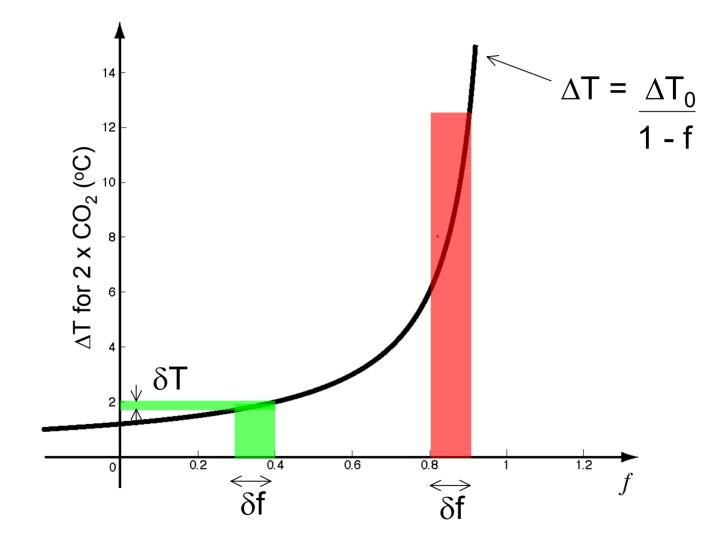
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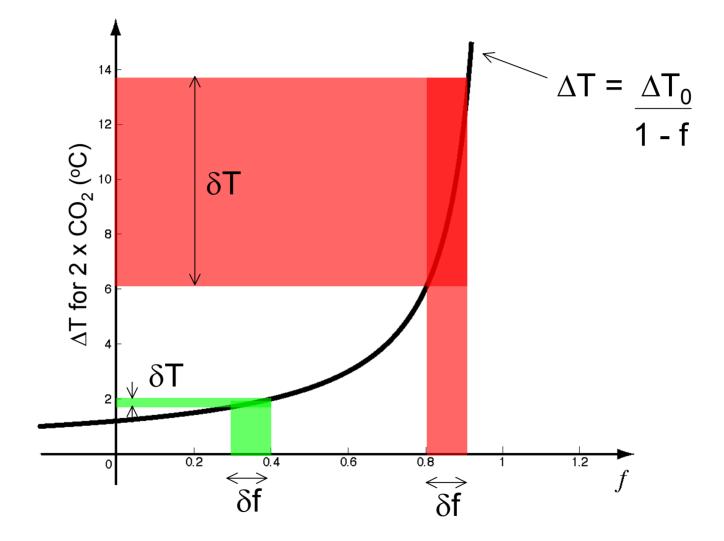
 How does this uncertainty in physics translate to uncertainty in climate sensitivity?

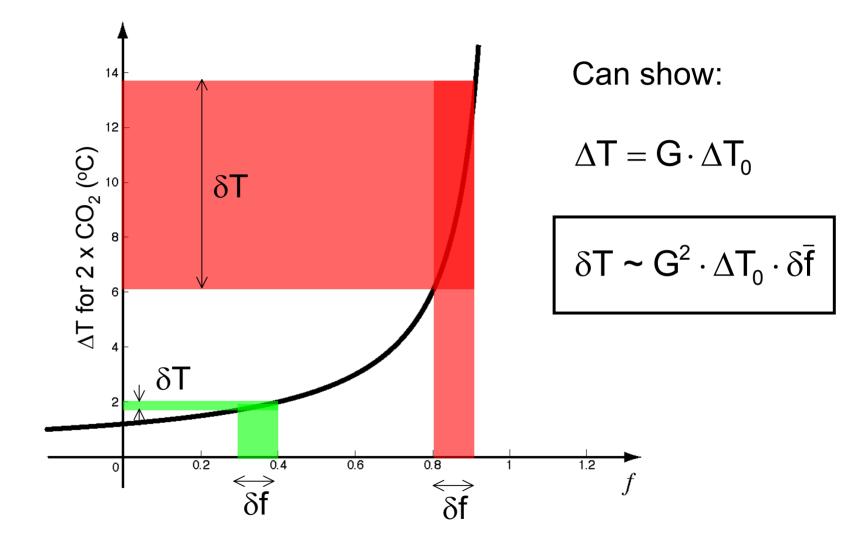












• Uncertainty in climate sensitivity strongly dependent on the gain.

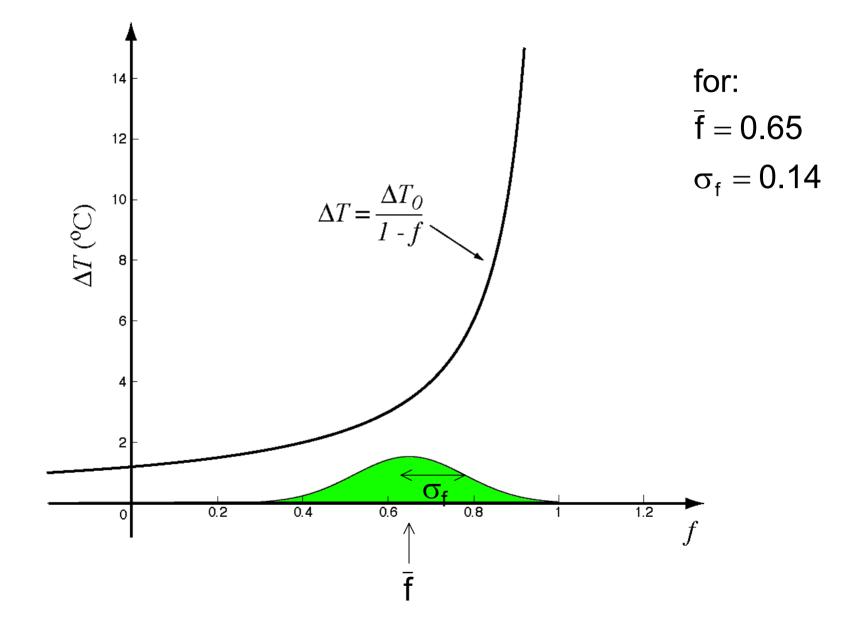
Climate sensitivity: the math

Let pdf of uncertainty
in feedbacks
$$h_f(f)$$
:
Also have:
 $\Delta T(f) = \frac{\Delta T_0}{1-f}$
So can write:
 $h_{\Delta T}(\Delta T) = h_f(f) \cdot \frac{df}{d(\Delta T)} = \frac{\Delta T_0}{\Delta T^2} \cdot h_f\left(1 - \frac{\Delta T_0}{\Delta T}\right)$
Assume Gaussian h(f):
 $h_f(f) = \frac{1}{\sigma_f \sqrt{2\pi}} \cdot \exp\left[-\frac{1}{2}\left(\frac{f-\bar{f}}{\sigma_f}\right)^2\right]$

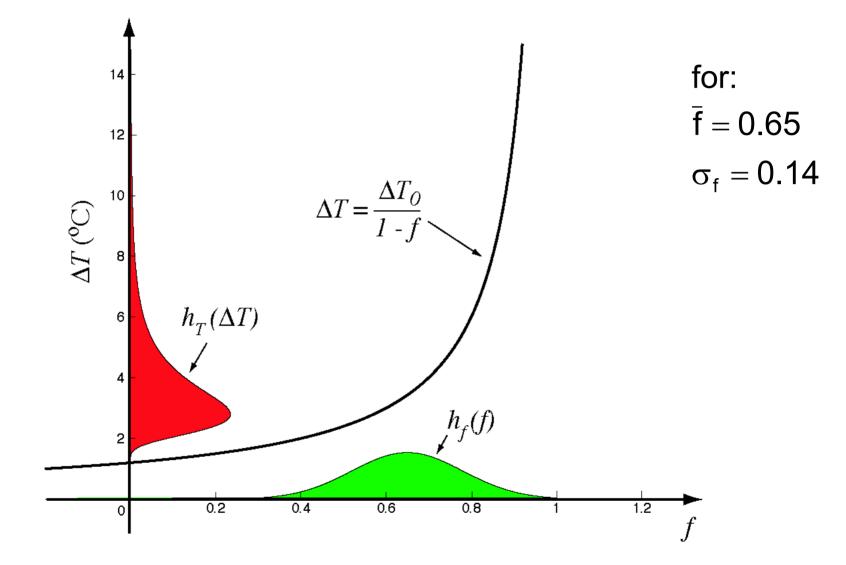
Gives

$$h_{\Delta T}(\Delta T) = \frac{1}{\sigma_{f}\sqrt{2\pi}} \cdot \frac{\Delta T_{0}}{\Delta T^{2}} \cdot \exp\left[-\frac{1}{2}\left(\frac{1-\bar{f}-\Delta T/\Delta T_{0}}{\sigma_{f}}\right)^{2}\right]$$

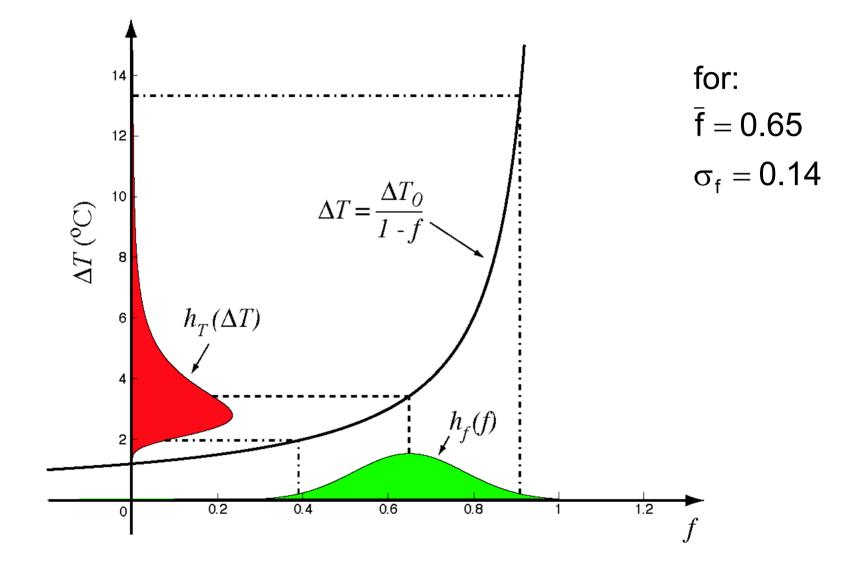
Climate sensitivity: the picture



Climate sensitivity: the picture



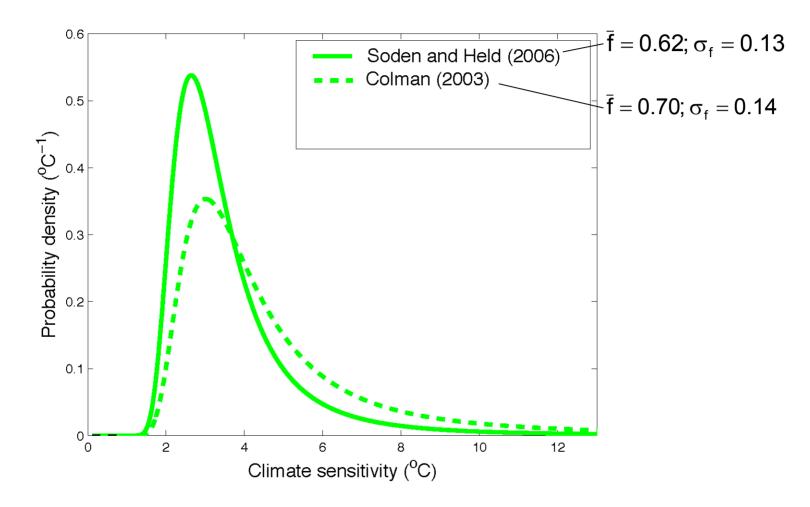
Climate sensitivity: the picture



Skewed tail of high climate sensitivity is inevitable!

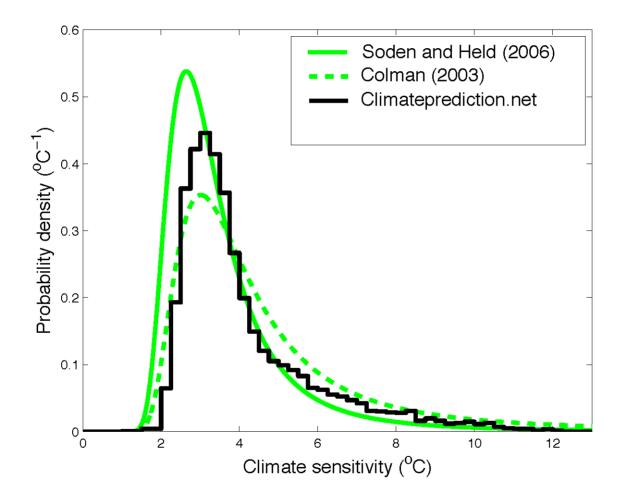
Climate sensitivity:

GCM from linear sum of feedback factors



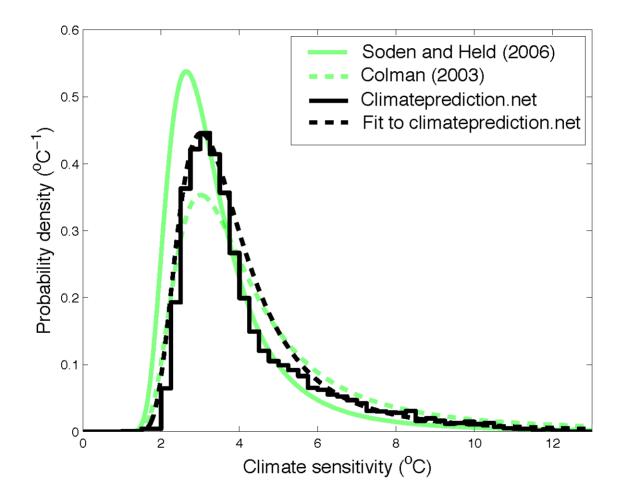
Climate sensitivity:

comparison with climateprediction.net



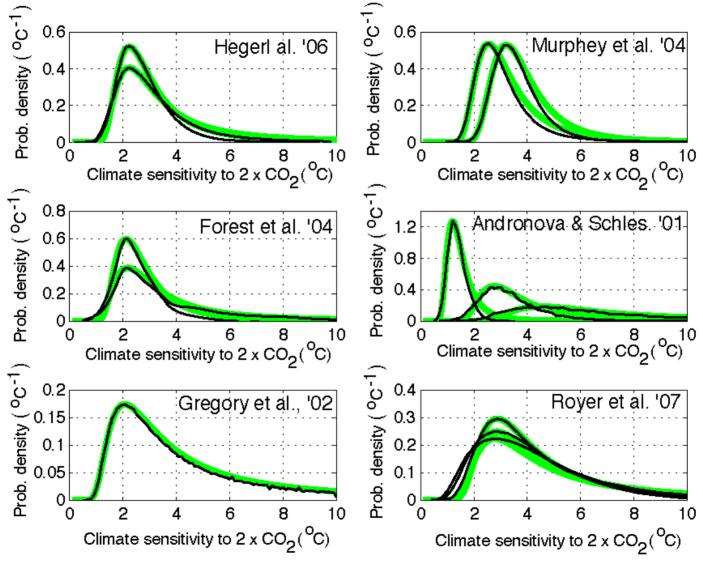
Climate sensitivity:

comparison with climateprediction.net



• GCMs produce climate sensitivity consistent with the compounding effect of essentially-linear feedbacks.

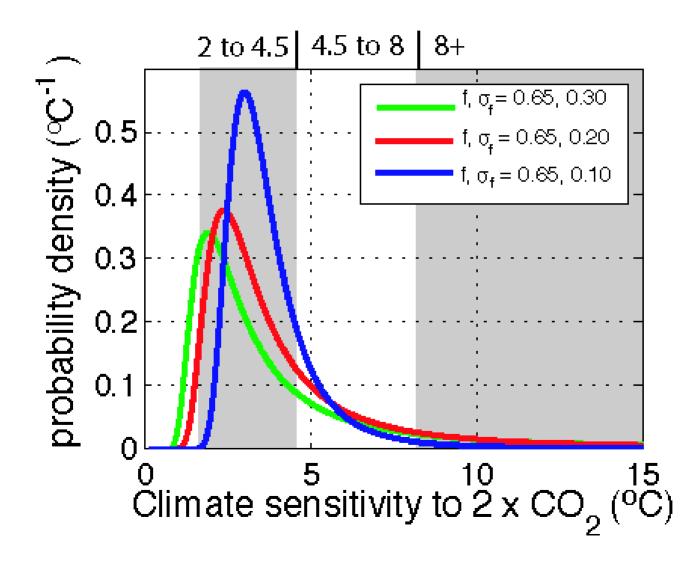
Climate sensitivity: comparison with studies



• $h_{\Delta T}(\Delta T)$ works pretty well.

Climate sensitivity: can we do better?

• How does uncertainty in climate sensitivity depend on σ_{f} ?



Climate sensitivity: can we do better?

\bar{f}, σ_{f}	2 to 4.5 °C	4.5 to 8 °C		> 8 °C	
0.65, 0.3	29%		14%	13%	science is here
0.65, 0.2	43%		18%	12%	
0.65, 0.1	55%		20%	8%	
0.65, 0.05	95%		5%	0%	_ need to get here!

- Not much change as a function of $\sigma_{\rm f}$

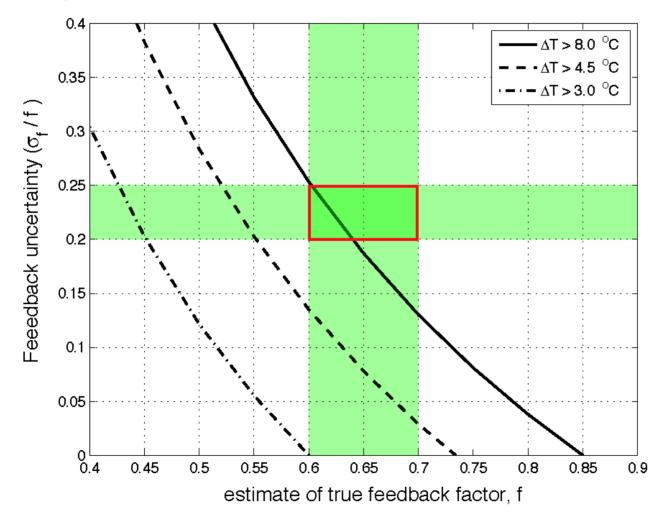
Climate sensitivity: can we do better?

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- Not much change as a function of $\sigma_{\rm f}$

Climate sensitivity: can we do better?

• Combination of mean feedback and uncertainty at which a given climate sensitivity can be rejected.



• Need to get cross hairs below a given line to reject that ΔT with 95% confidence

Summary:

- Climate change is unpredictable because climate change is inescapable.
 - -Uncertainty is inherent is a system where the feedbacks are substantially positive.
 - -Fat tail of the possibility of extreme climate sensitivity is inevitable, and has severe policy & planning consequences (e.g., Weitzman, 2008)
- The unpredictability of climate is predictable.

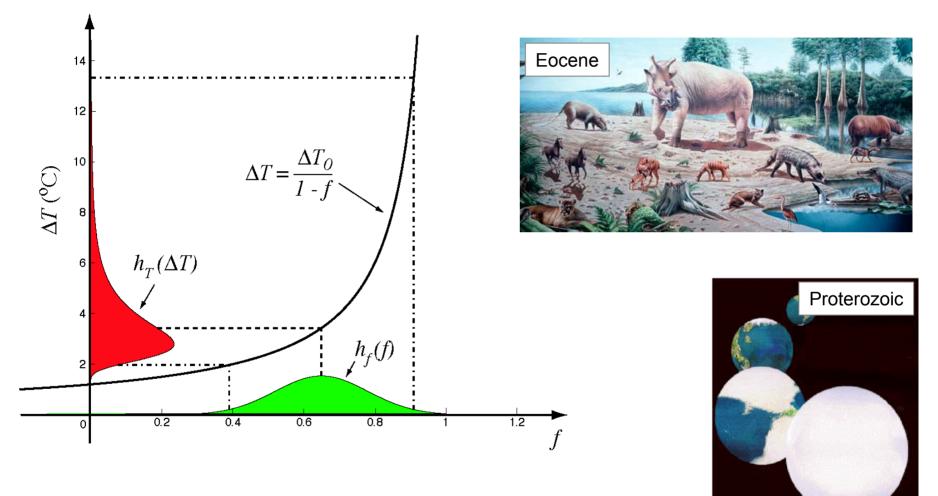
-Compounding effect of essentially linear feedbacks dominates system sensitivity.

• If you know the feedback factors, and their uncertainties don't need10⁴ GCMs (or 10⁷ model years!).

-Results suggest a simple relationship between forcing, feedbacks, and response

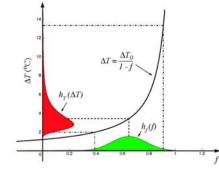
Paleoclimate speculations?

• What if feedback strengths change as a function of mean state?



• Dramatic changes in physics are not necessary for dramatic changes in climate sensitivity!

Can it be this simple?



- What is right about these ideas?
 - Very likely accounts for skewed tail of climate sensitivity pdfs.
 - From a modeling perspective, reducing uncertainties model parameters have limited effect on reducing uncertainty in climate sensitivity.
- What is wrong about these ideas?
 - h(f) cannot strictly be Gaussian.
 - feedback framework is a linear analysis in a very nonlinear world.
 - conclusions come from a modeling perspective. Observations of what actually happens have not been used!

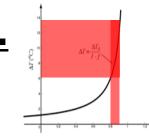
Climate sensitivity: other approaches.

Using observations (Allen et al., 2006)

Estimate λ from global energy budget:

$$\lambda(\Delta R - \Delta Q) \sim \Delta T$$

 ΔR = climate forcing; ΔQ = energy imbalance; ΔT = temperature change



Climate sensitivity: other approaches.

Using observations (Allen et al., 2006)

Estimate λ from global energy budget:

 $\begin{array}{l} \underline{Example: \ for \ present \ day.} \\ \Delta T = 0.65 \pm 0.025 \ ^{\circ}C; \\ \underline{\Delta Q = 0.85 \pm 0.08 \ W \ m^{-2};} \\ \underline{\Delta R = 1.8 \pm 0.42 \ W \ m^{-2}} \end{array}$

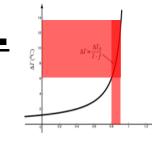
Lessons:

- Uncertainties in forcing dominate and still produce skewed tails.
- Forcing uncertainty comes from solar variability, volcanoes, aerosols, etc.
- True for any past climate reconstruction & also for modern...

 ΔR = climate forcing; ΔQ = energy imbalance; ΔT = temperature change

TM and a decompressor is picture.

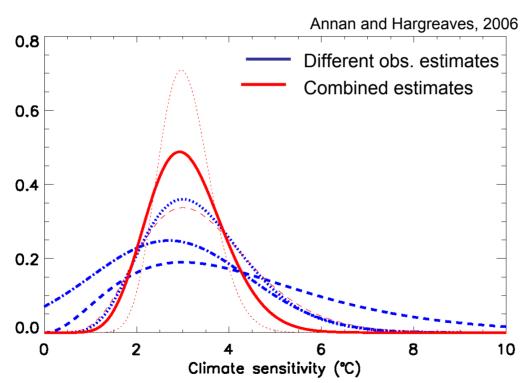




Climate sensitivity: other approaches.

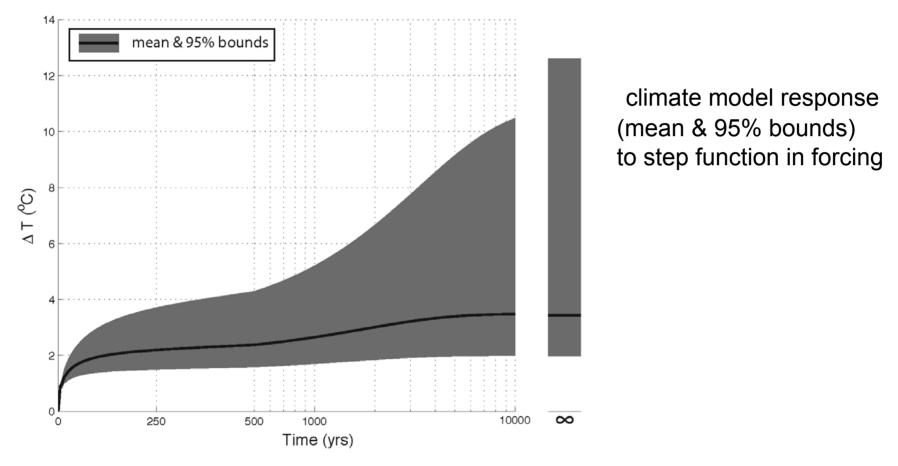
Combining different estimates (e.g. Annan & Hargreaves, 2006; Crucifix, 2006; Sherwood & Forest, 2007)

• In principle, climate sensitivity can be derived from multiple time intervals (little ice age, last ice age, modern, etc)



Bayesian estimates depend very sensitively on prior assumptions and the independence of different information.

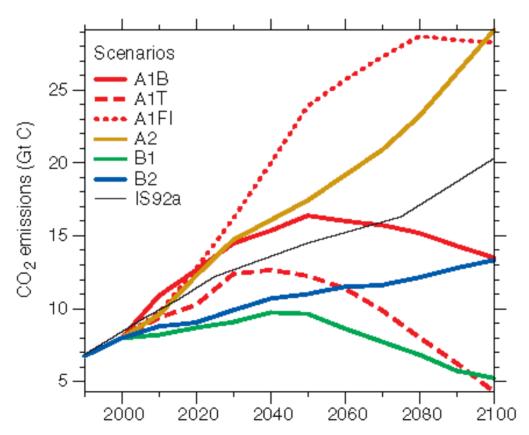
However, climate sensitivity is an equilibrium measure of climate, and climate change is a time dependent problem...



 It takes a very long time for the full pdf of climate response to be realized

However, climate sensitivity is an equilibrium measure of climate, and climate change is a time dependent problem...

IPCC CO₂ emissions scenarios,

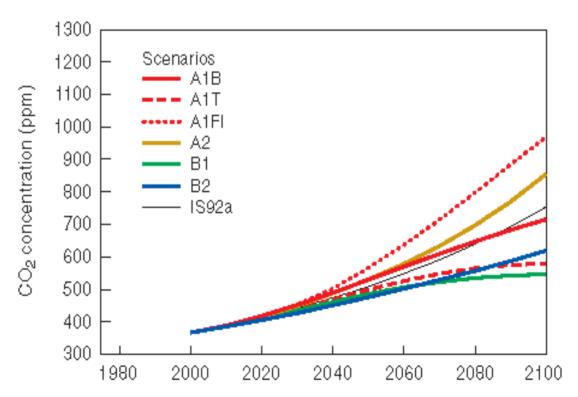


IPCC, 2001

However, climate sensitivity is an equilibrium measure of climate, and climate change is a time dependent problem...

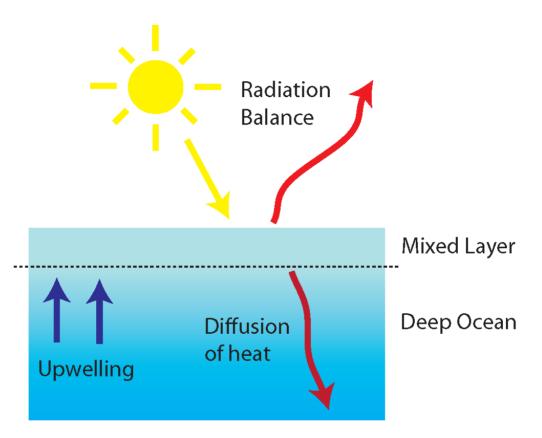
IPCC CO₂ concentration scenarios:

IPCC, 2001



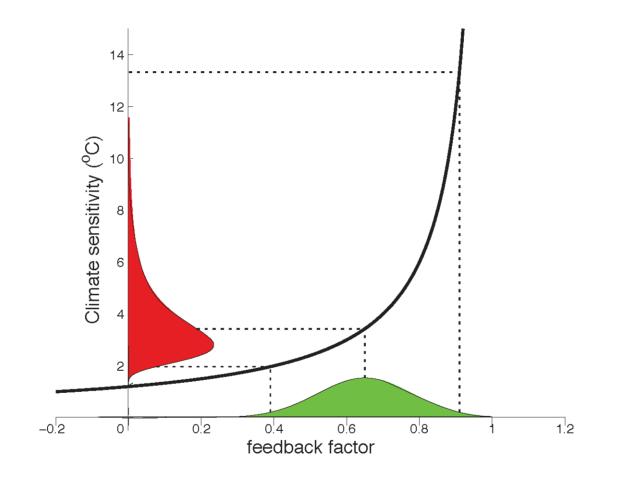
How does the envelope of response evolve in time?

The role of the ocean

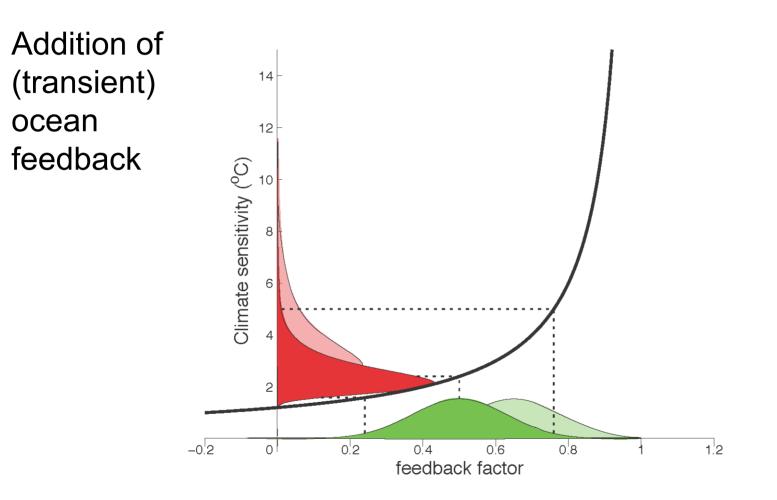


• The ocean heat uptake acts as a (transient) negative feedback.

The role of the ocean

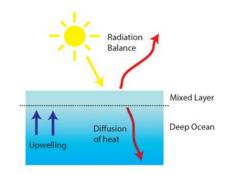


The role of the ocean



• Ocean -ve feedback strongly reduces the width of the envelope

The role of the ocean



• Equations for a climate model.....

Mixed layer:
$$\rho Ch_{ml} \frac{\partial T'_{ml}}{\partial t} + \frac{T'_{ml}}{\lambda} - \kappa \frac{\partial T'}{\partial z} \Big|_{z=0} = \Delta R_f(t)$$

Deep ocean:
$$\frac{\partial T'}{\partial t} = \chi \frac{\partial^2 T'}{\partial z^2} - w \frac{\partial T'}{\partial z}$$

- Forget the equations, the point is...
 - Climate feedbacks combine simply, so...
 - Can integrate a simple climate model to propagate the full range of uncertainty in feedbacks, ocean heat uptake, forcing etc...

The role of the ocean

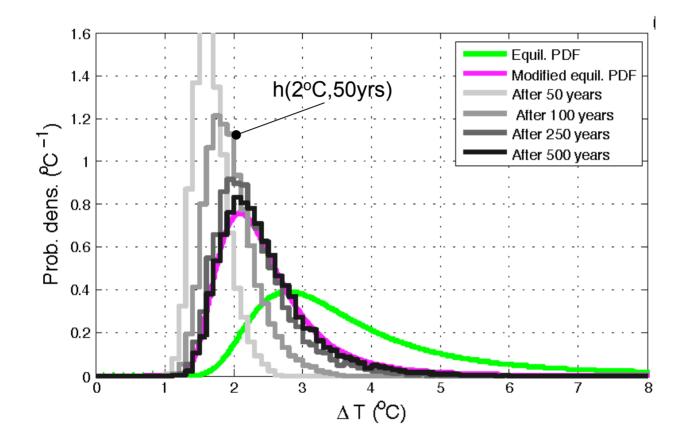
Response of temperature to a doubling of CO₂

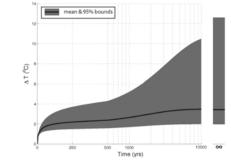
— Eq^m. prob. distr.

Eq^m. prob. distr.
 modified by -ve
 ocean feedback

QuickTime[™] and a MPEG-4 Video decompressor are needed to see this picture.

Why does the tail grow so slowly?

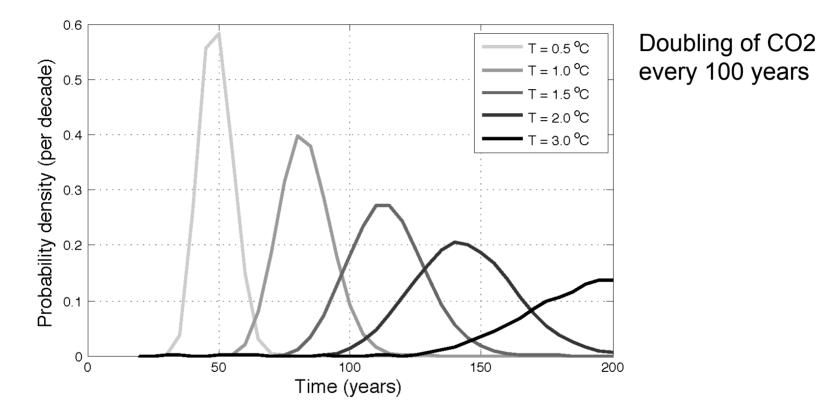




Let h(T,t) = probability density at some (T,t).

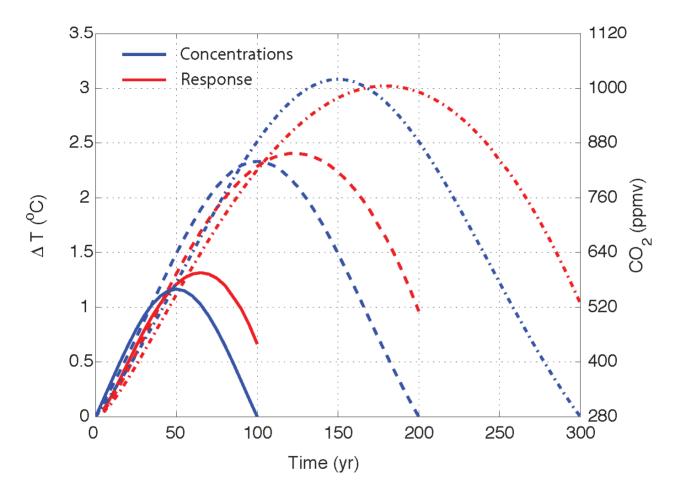
What governs how h(T,t) varies with time? (can talk about later...)

1. What is the likelihood of reaching a given temperature at a given time?



• The larger the temperature contemplated, the more uncertain it is when that temperature will be reached *(policy implications?)*.

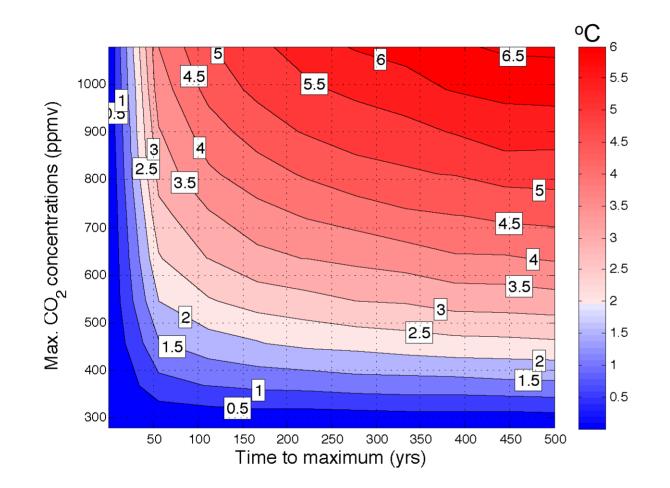
2. Which uncertainties matter most?



Concentration scenarios controlled by:

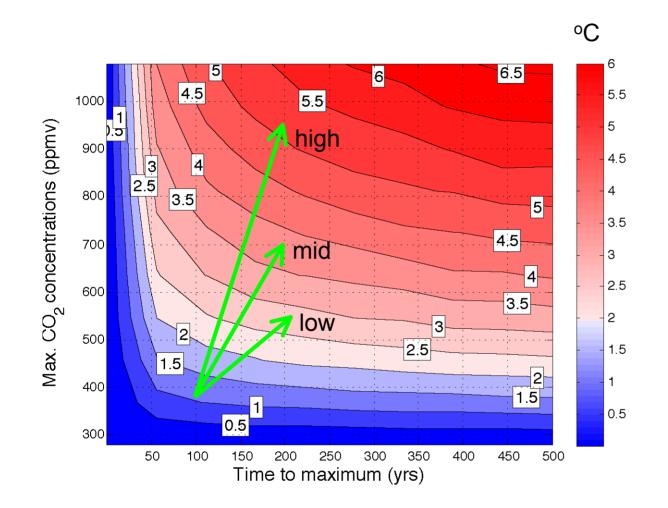
- maximum concentration.
- time to maximum.

2. Which uncertainties matter most?



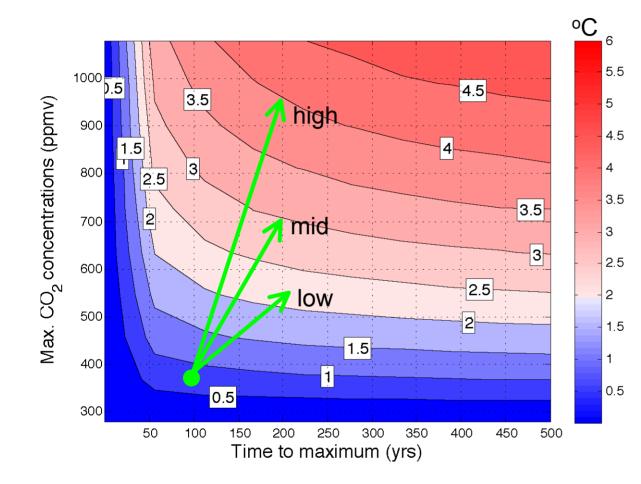
• If you are above the line, there is a 1 in 20 chance of seeing that climate change.

2. Which uncertainties matter most?



• All IPCC emissions scenarios yield a significant risk of dangerous climate change...

2. Which uncertainties matter most?

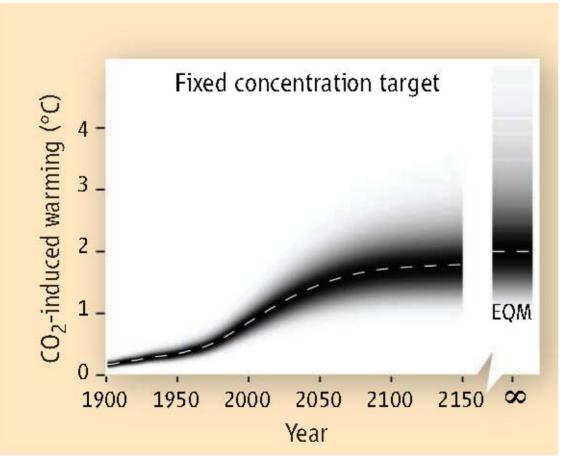


What happens In you halve the uncertainty in all climate model parameters?

- Getting smarter about climate, and reducing uncertainty, does not help that much.
- Uncertainty in emissions (and eventual concentrations) dominates.

2. Why even care about climate sensitivity? (sense and sensitivity...)

 Constraining climate sensitivity not terribly relevant for predicting climate change...

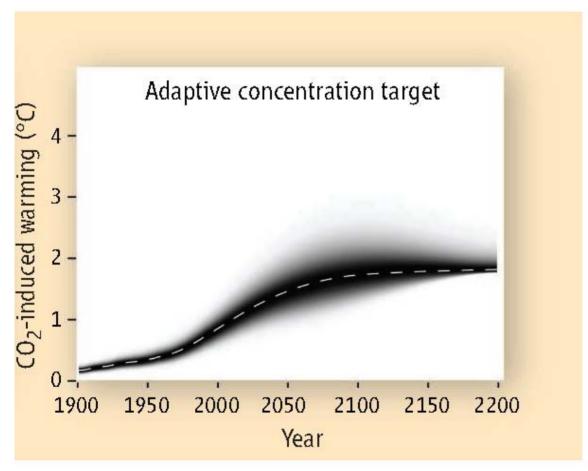


(Allen and Frame, 2007) Stabilization target of 450 ppm at 2100

High end sensitivities take a long, long time to be realized...

2. Why even care about climate sensitivity? (sense and sensitivity...)

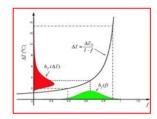
Constraining climate sensitivity not terribly relevant for predicting climate change... (Allen and Frame, 2007)



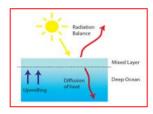
Concentration target adjusted at 2050.

In the face of uncertain information, adaptation is the answer!

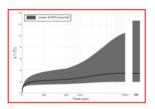
Conclusions II:



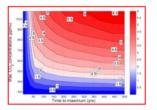
• We face a practical limit to the predictability of climate sensitivity - the fat tail is inevitable...

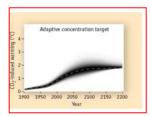


• Ocean heat uptake acts as a strong buffer.



• Growth of the fat tail is very slow.

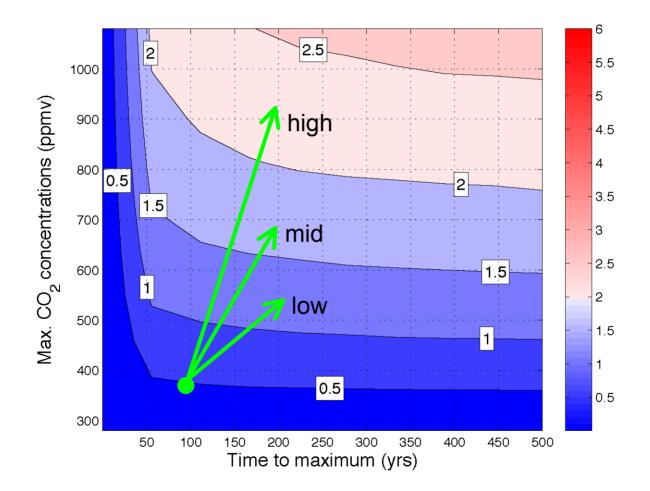




- Uncertainty in emissions swamps uncertainty in climate feedbacks.
- Flexibility is key!

The role of the ocean

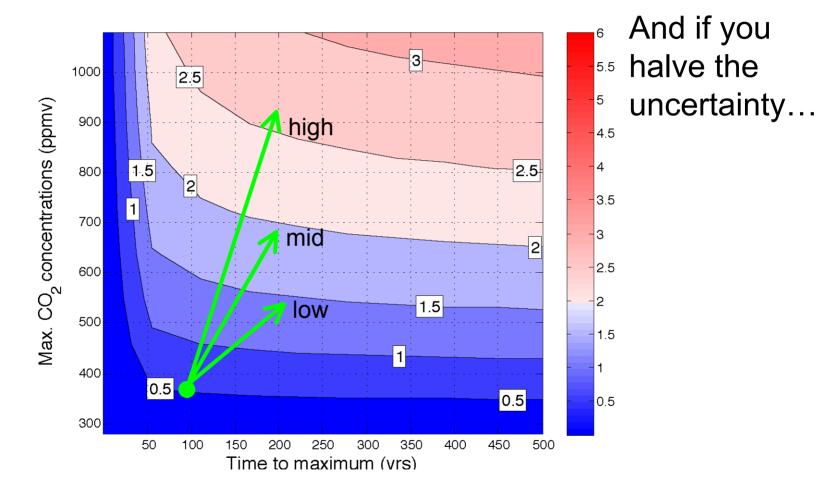
95% likelihood climate change



if you are above the line you have at least a 19 in 20 chance of that climate change

The role of the ocean

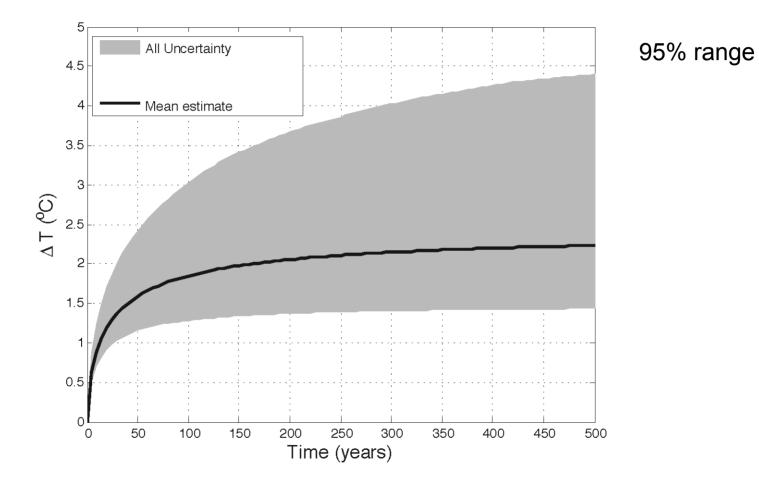
95% likelihood of climate change



if you are above the line you have at least a 19 in 20 chance of that climate change

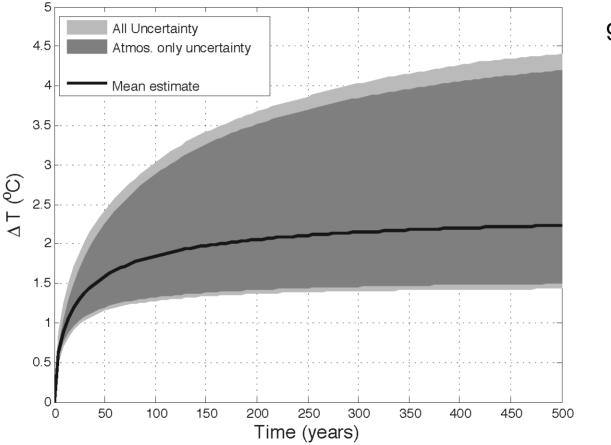
What kind of uncertainties matter for projections?

Response to a step-function doubling of CO₂



What kind of uncertainties matter for projections?

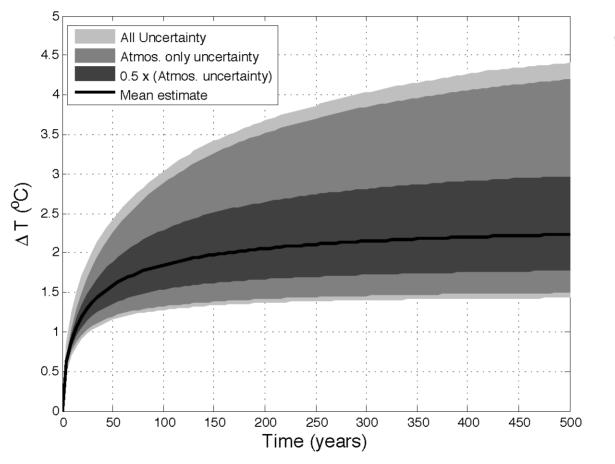
Response to a step-function doubling of CO₂



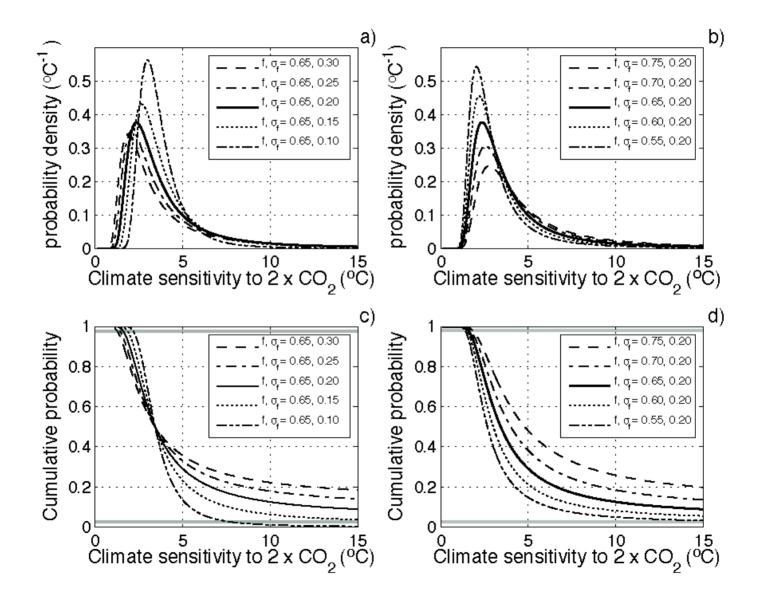
95% range

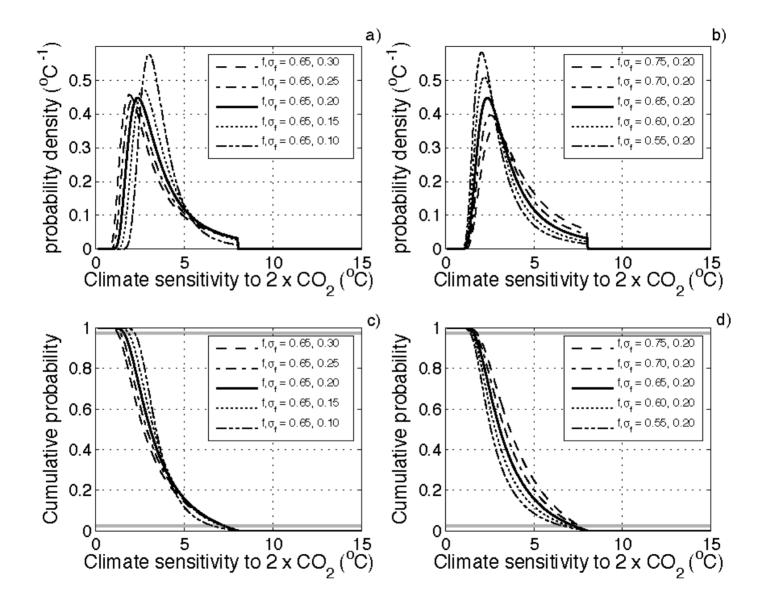
What kind of uncertainties matter for projections?

Response to a step-function doubling of CO₂

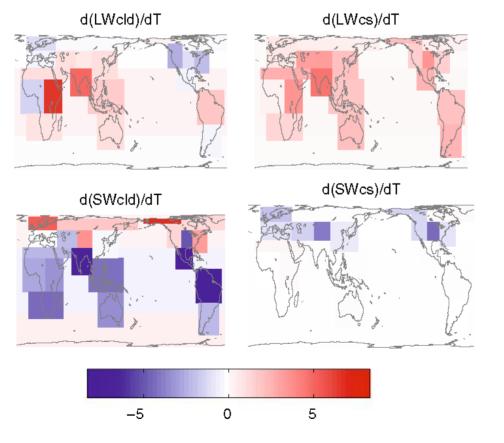


95% range





- cloud entrainment parameter has biggest impact on climate sensitivity in *climateprediction.net* ensemble.
- entrainment \downarrow , upper level moisture \uparrow , clear sky greenhouse \uparrow



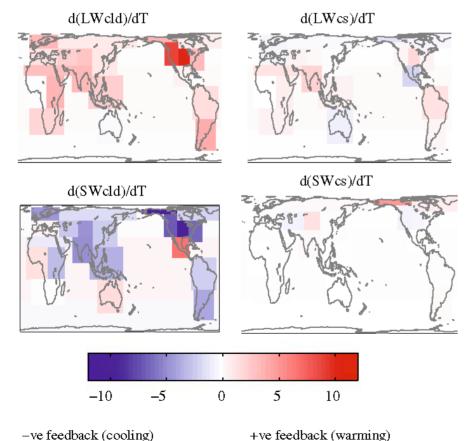
Surface radⁿ tendencies assoc. with entrainment

+ve feedback (warming)

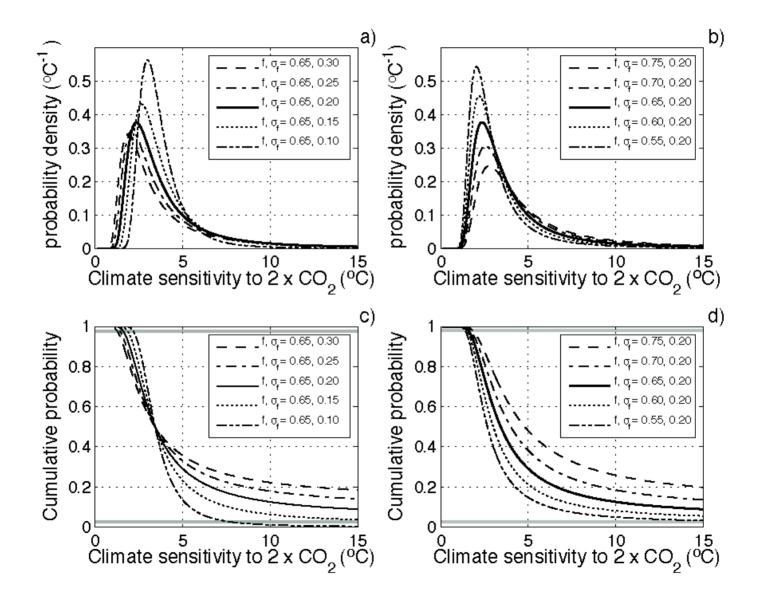
-ve feedback (cooling)

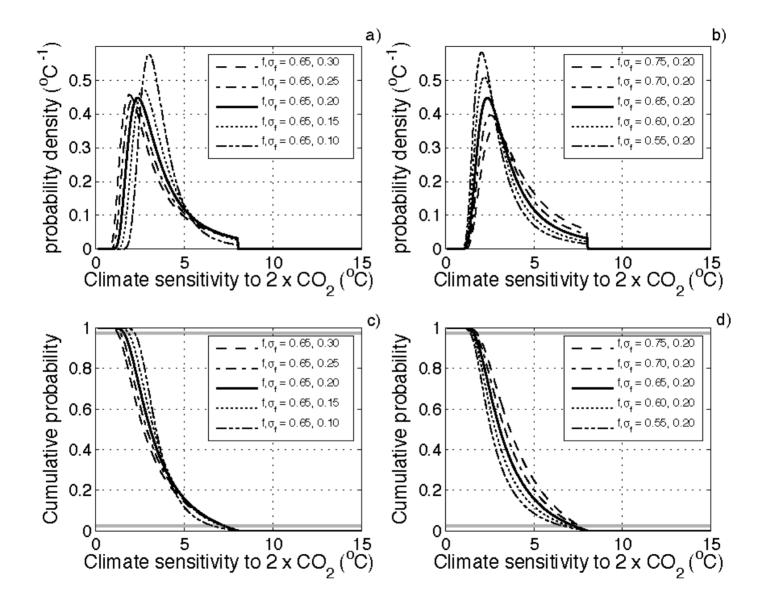
• ice fall speed has 2nd biggest impact on climate sensitivity in *climateprediction.net* ensemble.

• fall speed \downarrow , clouds/humidity \uparrow , greenhouse effect \uparrow

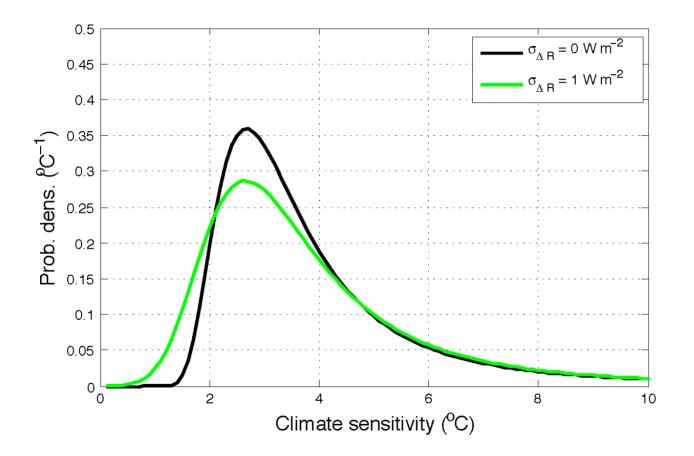


Surface radⁿ Tendencies assoc. with fall speed



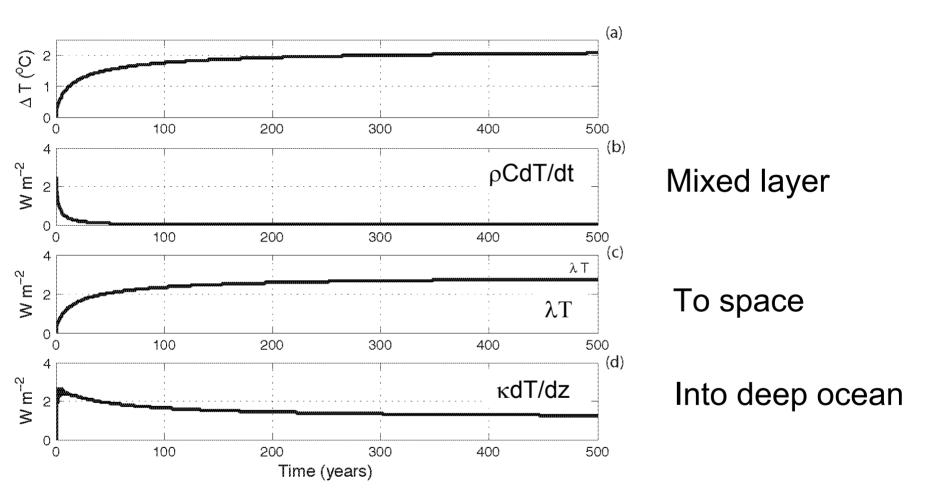


Effect of introducing uncertainty in the forcing on equilibrium climate response



Makes lower climate response more likely

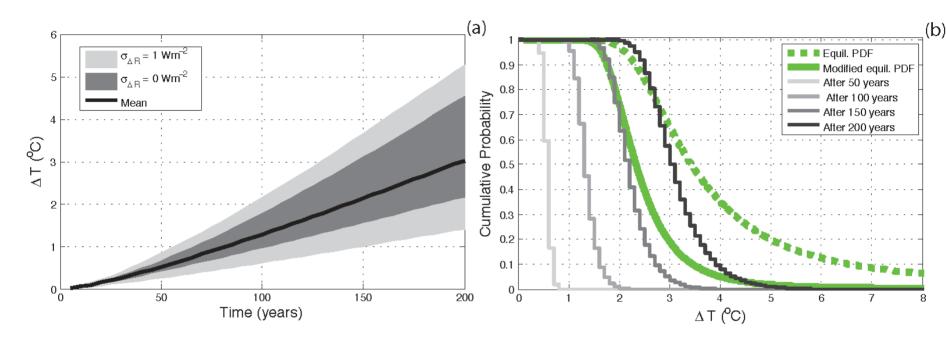
Evolution of the three terms in the energy balance in response to a step function in forcing



• Warming term rapidly diminshes to near zero...

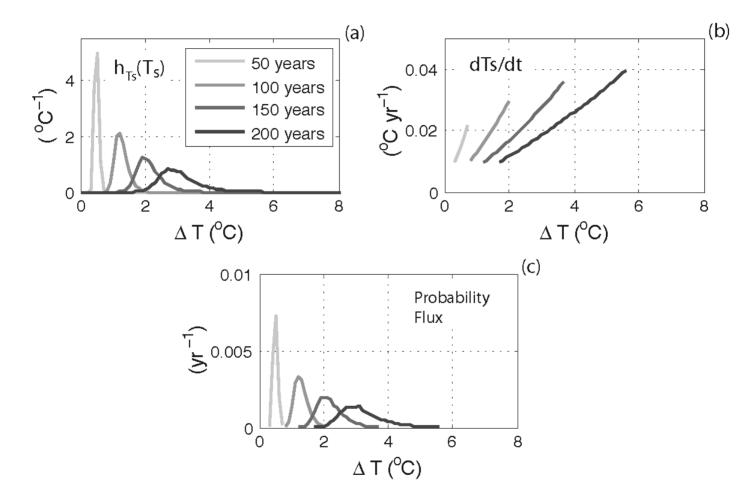
Response to ramp forcing

CO2 doubling every 100 years



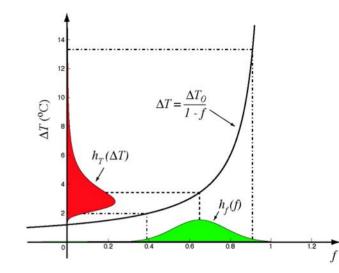
Response to ramp forcing

CO2 doubling every 100 years



What's right about this?

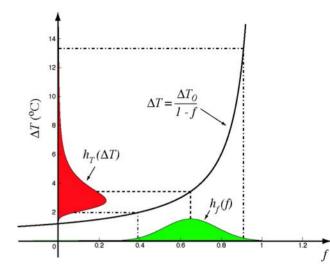
• Very likely accounts for skewed tail of climate sensitivity pdfs.



• From a modeling perspective, reducing uncertainties model parameters have limited effect on reducing uncertainty in climate sensitivity.

What's wrong about this?

- h(f) cannot strictly be Gaussian.
 not a big deal, any reasonable h(f) will do.
- feedback framework is a linear analysis in a very nonlinear world.
- conclusions come from a modeling perspective. Observations of what actually happens have not been used!



Where does our uncertainty in f come from?

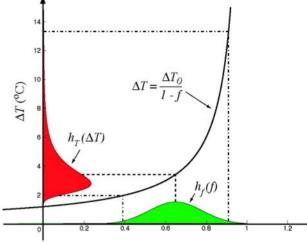
- 1. Ignorance?!
- 2. Nonlinearities in climate feedbacks From basic analysis: $\Delta R = \frac{dR}{dT} \Delta T + O(\Delta T^2)$

But can take quadratic terms...

$$\Delta R = \frac{dR}{dT} \Delta T + \frac{1}{2} \frac{d^2 R}{dT^2} \Delta T^2 + O(\Delta T^3)$$

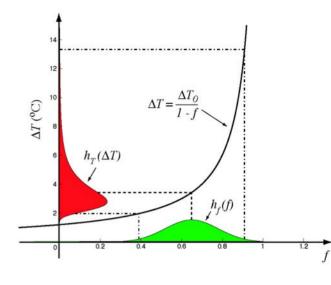
giving...

$$G = \frac{1}{1 - f - \frac{\Delta T}{2} \frac{df}{dT}}$$



Where does our uncertainty in f come from?

- 2. Nonlinearities in climate feedbacks.
- Stefan-Boltzmann, Clausius-Clapeyron nonlinearities give δf ~0.02 for ΔT~ 4°C.



• Colman et al. (1997) nonlinearities in water vapor, clouds, and lapse rate feedbacks, giving $\delta f \sim 0.1$ for $\Delta T = 4^{\circ}C$.

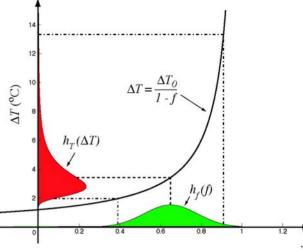
Where does our uncertainty in f come from?

3. Climate sensitivity varies with mean state.

- Senior and Mitchell (2000) climate

 sensitivity increases 40% under a global warming scenario.
- Boer and Yu (2003) climate sensitivity <u>decreases</u> 20%.
- Crucifix (2006) different models have very different changes in sensitivity between LGM and modern climates.

- 4. Chaotic climate system.
- Lea et al. (2005); Knight et al. (2007) suggest small but identifiable effects.



Other approaches:

Using observations (Allen et al., 2006)

Estimate λ from global energy budget:

 $\lambda \sim \frac{\Delta T}{\Delta R - \Delta Q} \qquad \begin{array}{l} \Delta R = \text{climate forcing;} \\ \Delta Q = \text{energy imbalance;} \\ \Delta T = \text{temperature change} \end{array}$

Example: for present day.

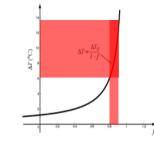
- $\Delta T=0.65\pm0.025 \text{ °C}; \ \Delta Q=0.85\pm0.08 \text{ W m}^{-2}; \ \Delta R=1.8\pm0.42 \text{ W m}^{-2}$
- Uncertainties in forcing dominate and still produce skewed tails.

Combining different estimates

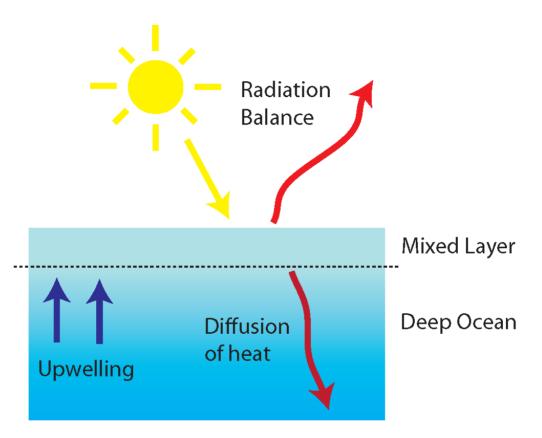
(e.g. Annan & Hargreaves, 2006; Crucifix, 2006; Sherwood & Forest, 2007)

Bayesian estimates:-

depends very sensitively on prior assumptions, and the independence of different information.

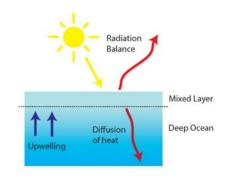


The role of the ocean



• The ocean heat uptake acts as a (transient) negative feedback.

The role of the ocean



Mixed layer:
$$\rho Ch_{ml} \frac{\partial T'_{ml}}{\partial t} + \frac{T'_{ml}}{\lambda} - \kappa \frac{\partial T'}{\partial z}\Big|_{z=0} = \Delta R_f(t)$$

Deep ocean:

$$\frac{\partial T'}{\partial t} = \chi \frac{\partial^2 T'}{\partial z^2} - W \frac{\partial T'}{\partial z}$$

ī

The role of the ocean

Nondimensionalized:

Mixed layer:

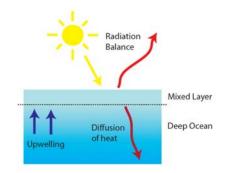
$$\left. X \frac{\partial T}{\partial t} + (1 - f_a) T - f_o \frac{\partial T}{\partial z} \right|_{z=0} = 1$$

Deep ocean:

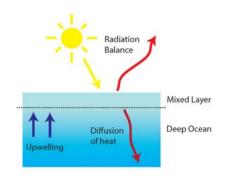
$$\frac{\partial T'}{\partial t} = \frac{\partial^2 T'}{\partial z^2} - \frac{\partial T'}{\partial z}$$

Solution depends three nondimensional parameters f_a, X, & f_o

$$X = \frac{\tau_{ml}}{\tau_o} = \frac{\text{mixed layer response time}}{\text{deep ocean mixing time}} \approx 10^{-2}$$



The role of the ocean



$$\left. X \frac{\partial T}{\partial t} + (1 - f_a)T - f_o \frac{\partial T}{\partial z} \right|_{z=0} = 1$$

Deep ocean:

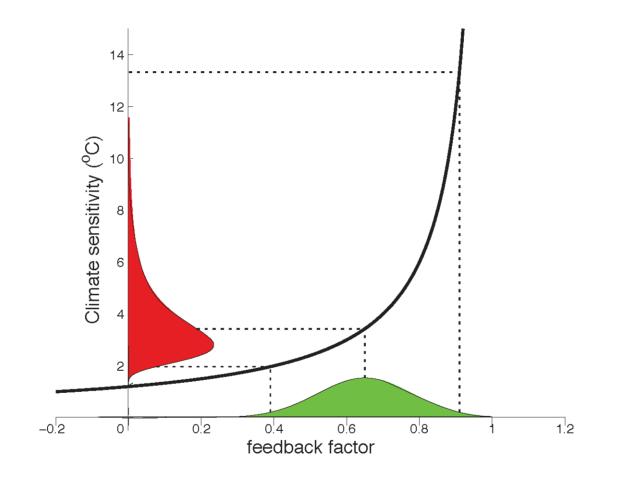
$$\frac{\partial T'}{\partial t} = \frac{\partial^2 T'}{\partial z^2} - \frac{\partial T'}{\partial z}$$

Solution depends three nondimensional parameters f_a , X, & f_o

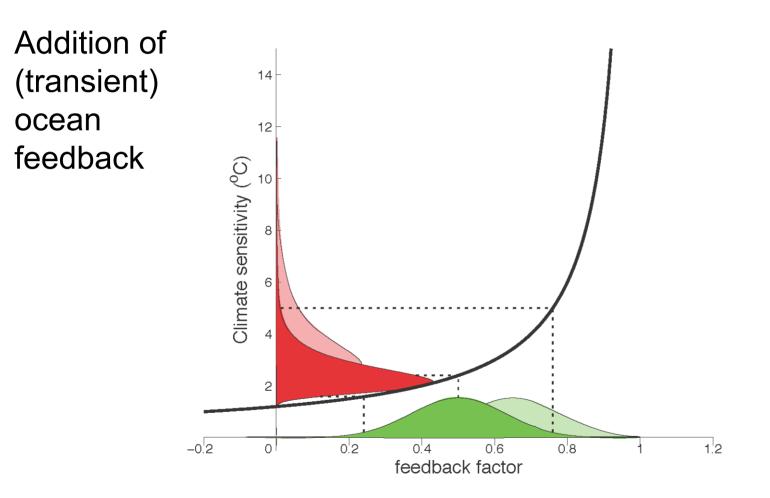
$$f_o = \rho C \lambda_o w =$$
 ocean feedback factor ≈ -0.15

w = upwelling rate

The role of the ocean



The role of the ocean



• Ocean -ve feedback strongly reduces the width of the envelope

The role of the ocean

Analytical solution allows for extremely efficient Monte Carlo computation of the effect of parameter uncertainties

<u>Parameter</u>	<u>Mean</u>	<u>1σ</u>
Atmospheric feedback	0.65	0.15
Upwelling rate	4 m yr ⁻¹	1.5 m yr ^{_1}
Mixed layer depth	75 m	25 m
Ocean diffusivity	1.5 cm ² s ⁻¹	0.5 cm ² s ⁻¹

The role of the ocean

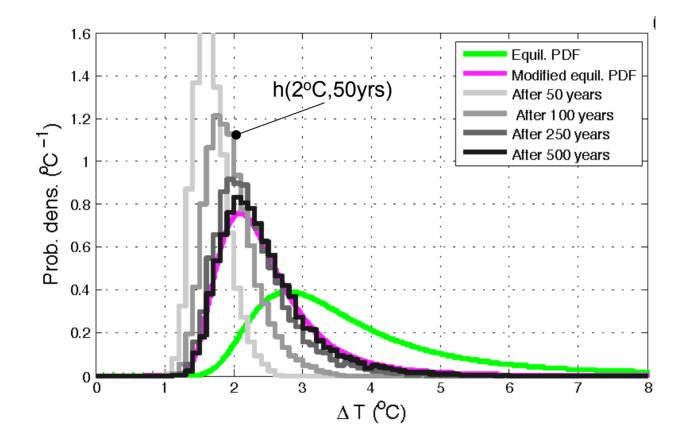
Response of temperature to a doubling of CO₂

— Eq^m. prob. distr.

Eq^m. prob. distr.
 modified by -ve
 ocean feedback

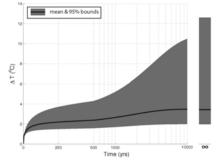
QuickTime[™] and a MPEG-4 Video decompressor are needed to see this picture.

Why does the tail grow so slowly?



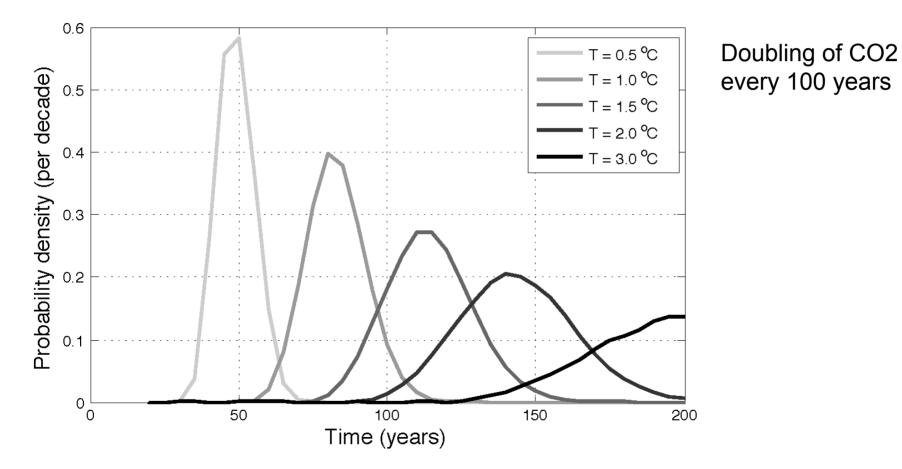
Let h(T,t) = probability density at some (T,t).

What governs how h(T,t) varies with time?

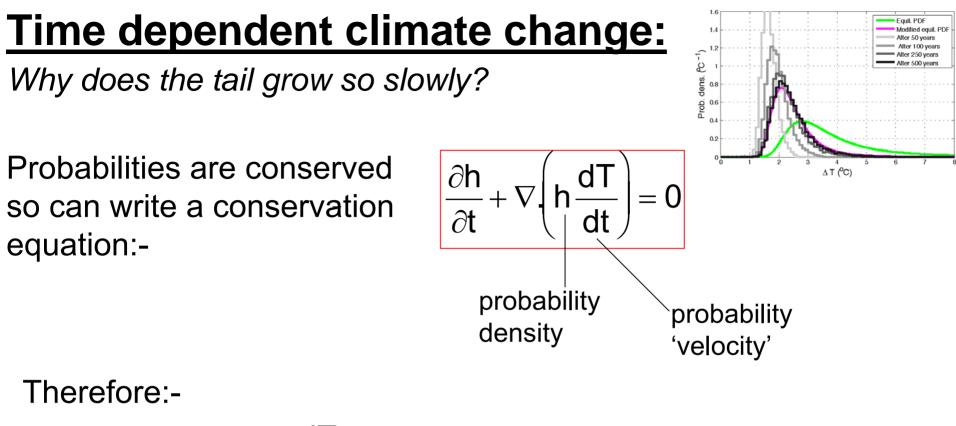


Response to a ramp forcing

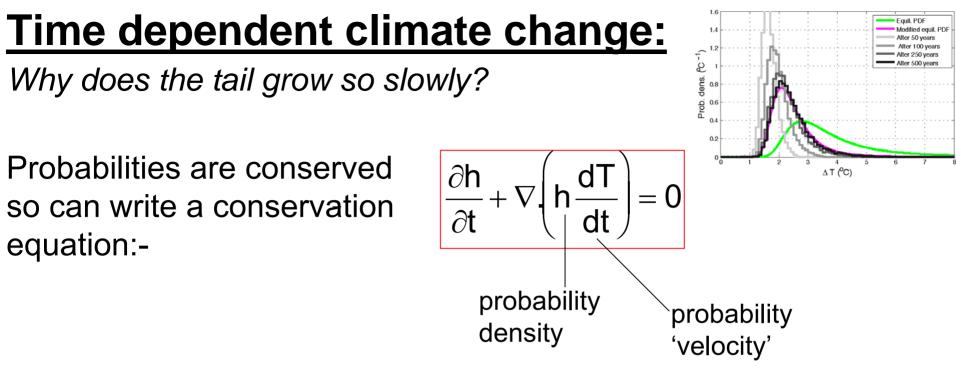
□ What is the likelihood of reaching a given temperature at a given time?



• The larger the temperature contemplated, the more uncertain it is when that temperature will be reached.



$$h(T,t) \times \frac{dT}{dt}$$
 = flux of probabilities to higher T



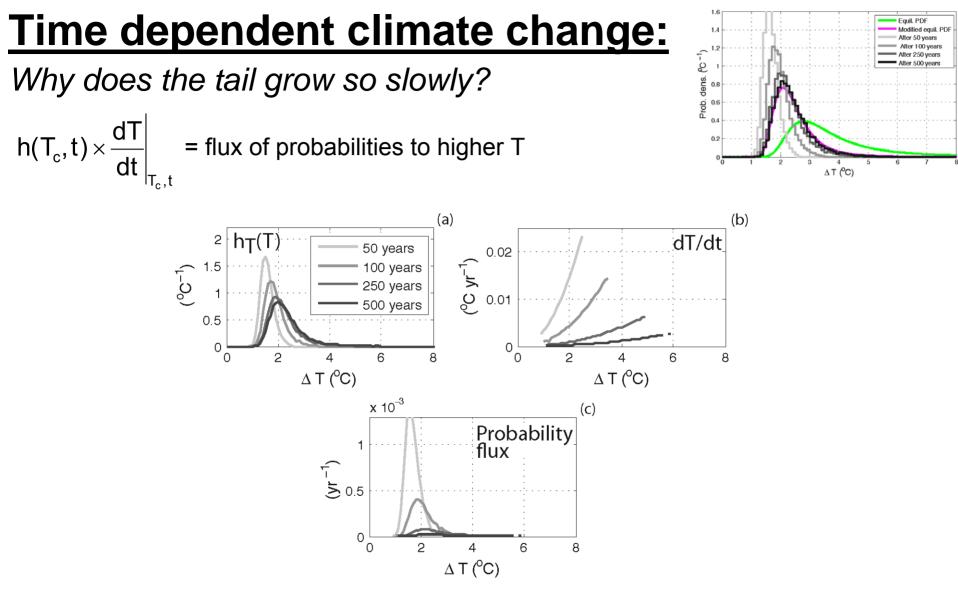
Can integrate from T_c to ∞ :

Т

$$\frac{dp_{cum}}{dt}\bigg|_{T>T_{c},t} = h(T_{c},t) \times \frac{dT}{dt}\bigg|_{T_{c},t}$$

(since $h(\infty,t)=0$)

$$p_{cum}|_{T>T_{c},t}$$
 = cumulative probability of T >T_c



• Flux in the tail diminishes quickly to low (but non zero) values.