

Bayesian estimation of the climate sensitivity based on a simple climate model fitted to global temperature observations

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- Joint work with
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 - ★ CICERO = Center for International Climate and Environmental Research in Oslo
- Work in progress, preliminary results

Climate sensitivity S

Definition:

Climate sensitivity = S

= The temperature increase due to a doubling of CO_2 concentrations compared to pre-industrial time (1750)

Radiative forcing

- CO_2 is only one of several factors that affect the global temperature
- Radiative forcing = The change in net irradiance into the earth relative to 1750
- Measured in Watts per square meter
- The global temperature depends on the radiative forcing
- The climate sensitivity measures the strength of this dependency

Aim of study

To estimate the climate sensitivity

- by modelling the relationship between
 - ★ estimates of radiative forcing since 1750 and
 - ★ estimates of global temperature
based on measurements since 1850
- using a climate model based on physical laws

Climate model

Could use

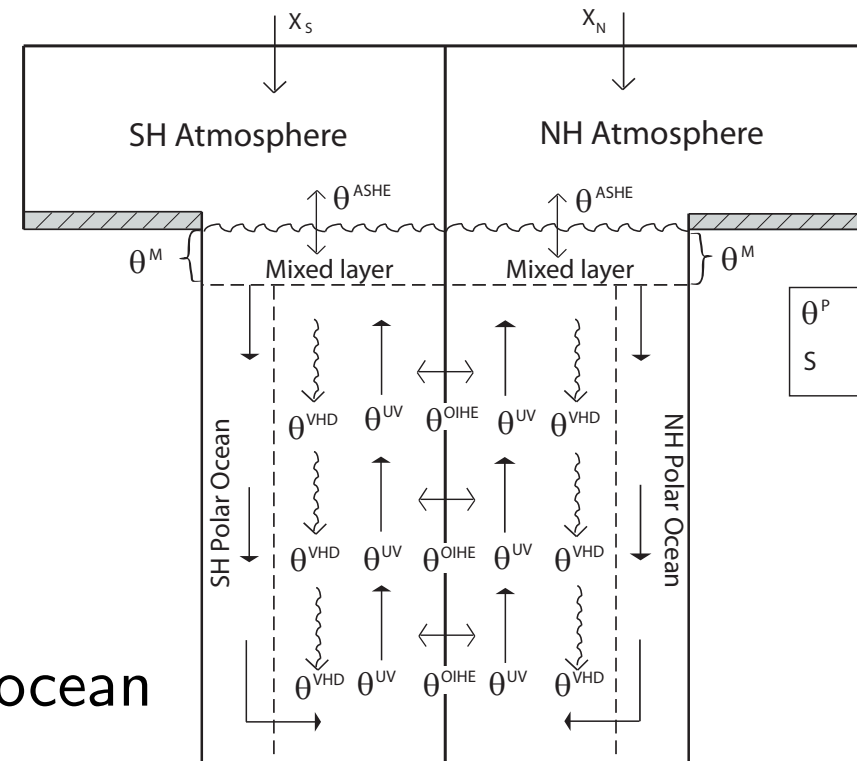
- an Atmospheric Ocean General Circulation Model, but complex and very computer intensive
- an approximation to an AOGCM, an emulator based on Gaussian processes
- a simple climate model, our approach

The “true” global state of the earth in year t

- TNH_t - Temperature at the northern hemisphere
- TSH_t - Temperature at the southern hemisphere
- OHC_t - Ocean heat content

Simple climate model

- Deterministic computer model (Schlesinger et al., 1992)
- based on
 - ★ energy balance
 - ★ upwelling diffusion ocean
- where the earth is divided into
 - ★ atmosphere and ocean
 - ★ northern and southern hemisphere
- with
 - ★ radiative forcing into the system
 - ★ energy mixing
 - * between the atmosphere and the ocean
 - * within the ocean



Simple climate model cont.

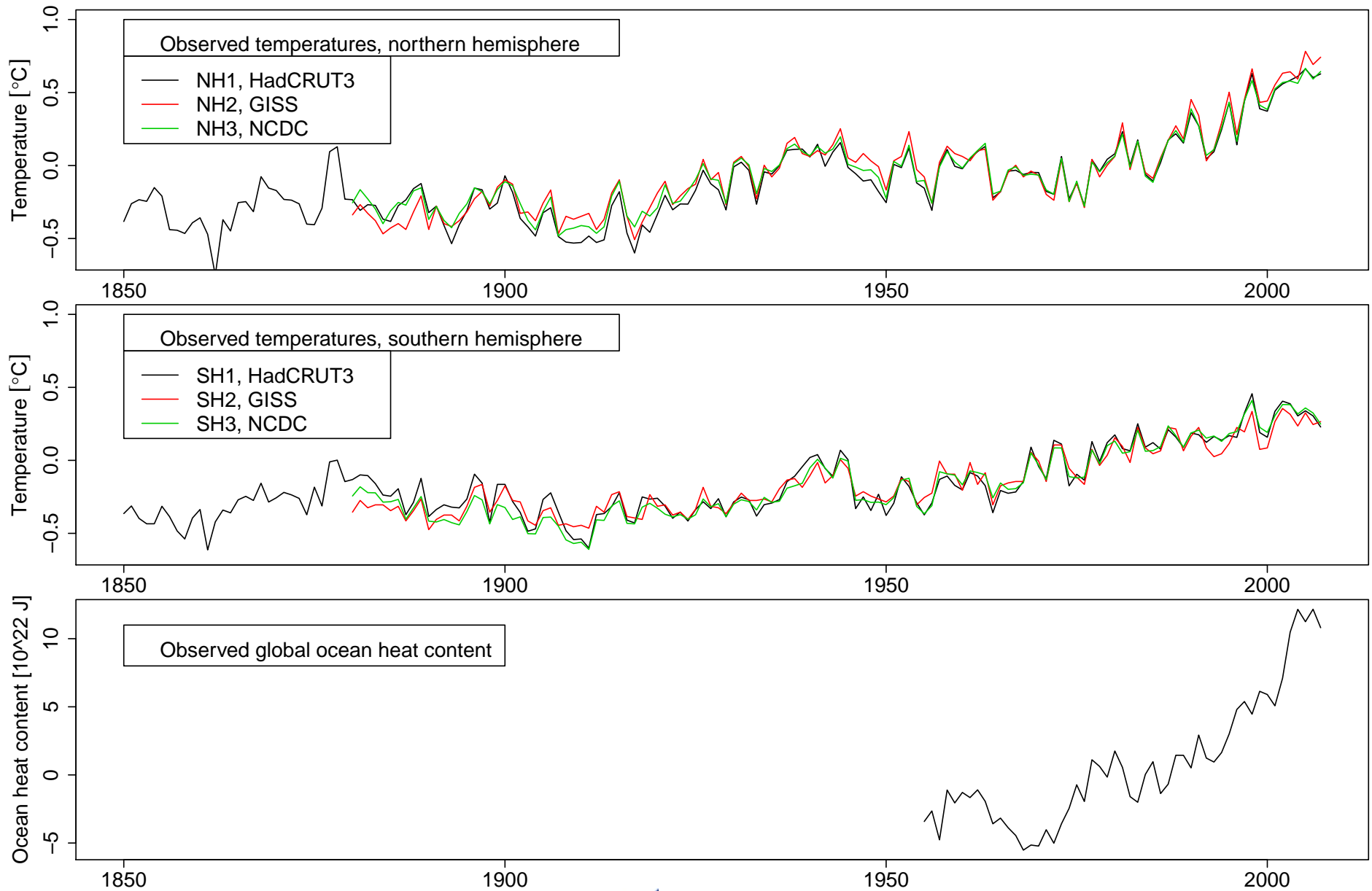
$$\mathbf{m}_t(\mathbf{x}_{1750:t}, S, \boldsymbol{\theta})$$

- Yearly time resolution
- Output
 - ★ temperature northern hemisphere
 - ★ temperature southern hemisphere
 - ★ ocean heat content
- Input
 - ★ $\mathbf{x}_{1750:t}$ - yearly radiative forcing from 1750 until year t , separate for northern and southern hemisphere
 - ★ S - the climate sensitivity, the parameter of interest
 - ★ $\boldsymbol{\theta}$ - 6-dimensional vector of other model parameters

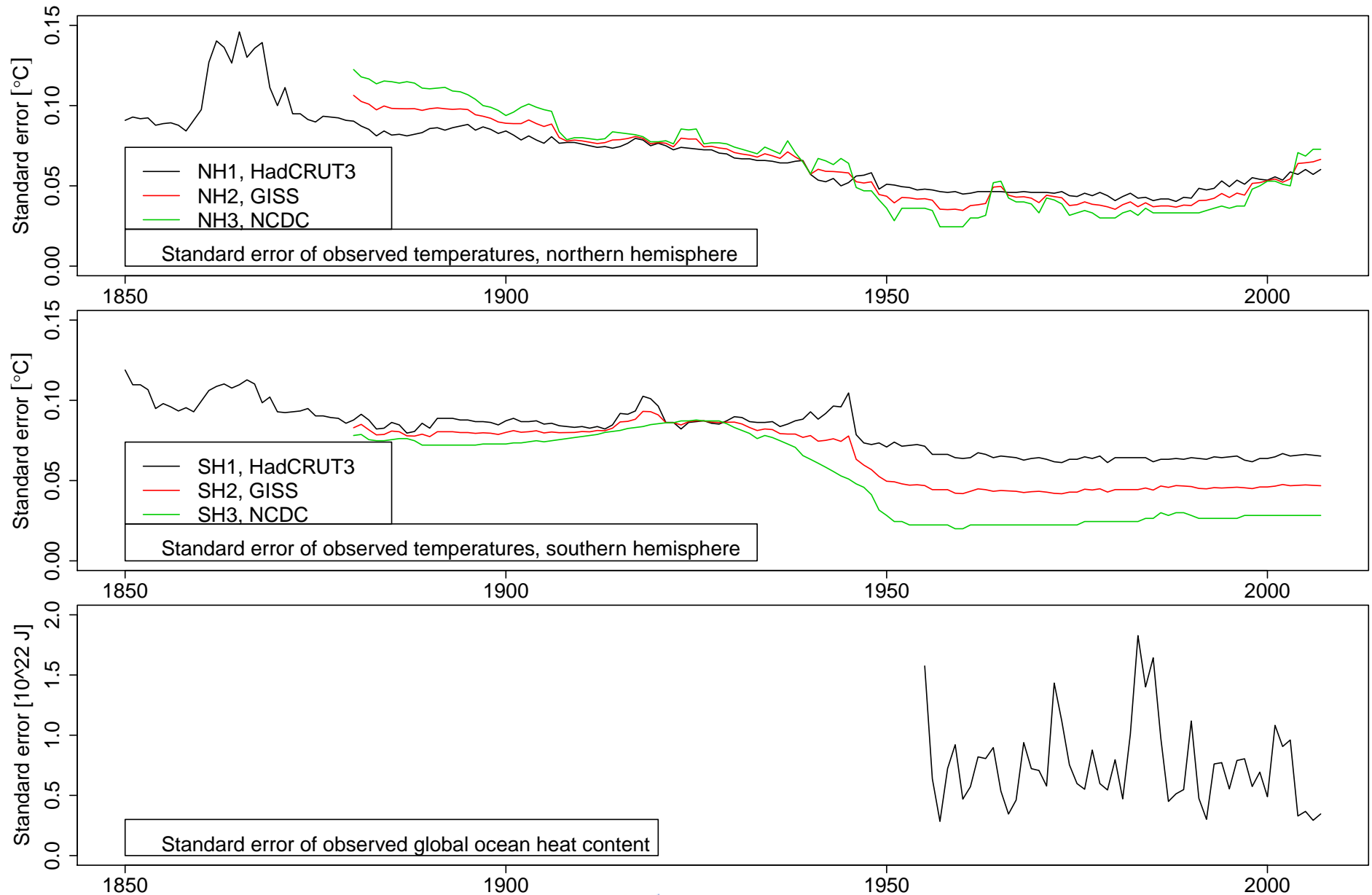
Response data

- y_t - 7-dimensional vector with yearly observed temperatures and ocean heat content
- Three pairs of series with temperature measurements for northern and southern hemisphere
 - ★ 1850-2007 (HadCRUT3, Brohan et al., 2006)
 - ★ 1880-2007 (GISS, Hansen et al. 2006)
 - ★ 1880-2007 (NCDC, Smith and Reynolds 2005)
- One series with ocean heat content measurements
 - ★ 1955-2007 (Levitus et al. 2009)
- These observations are not the truth, but are estimates of the underlying “true” global state of the earth
- s_t - 7-dimensional vector of corresponding standard errors

Observations



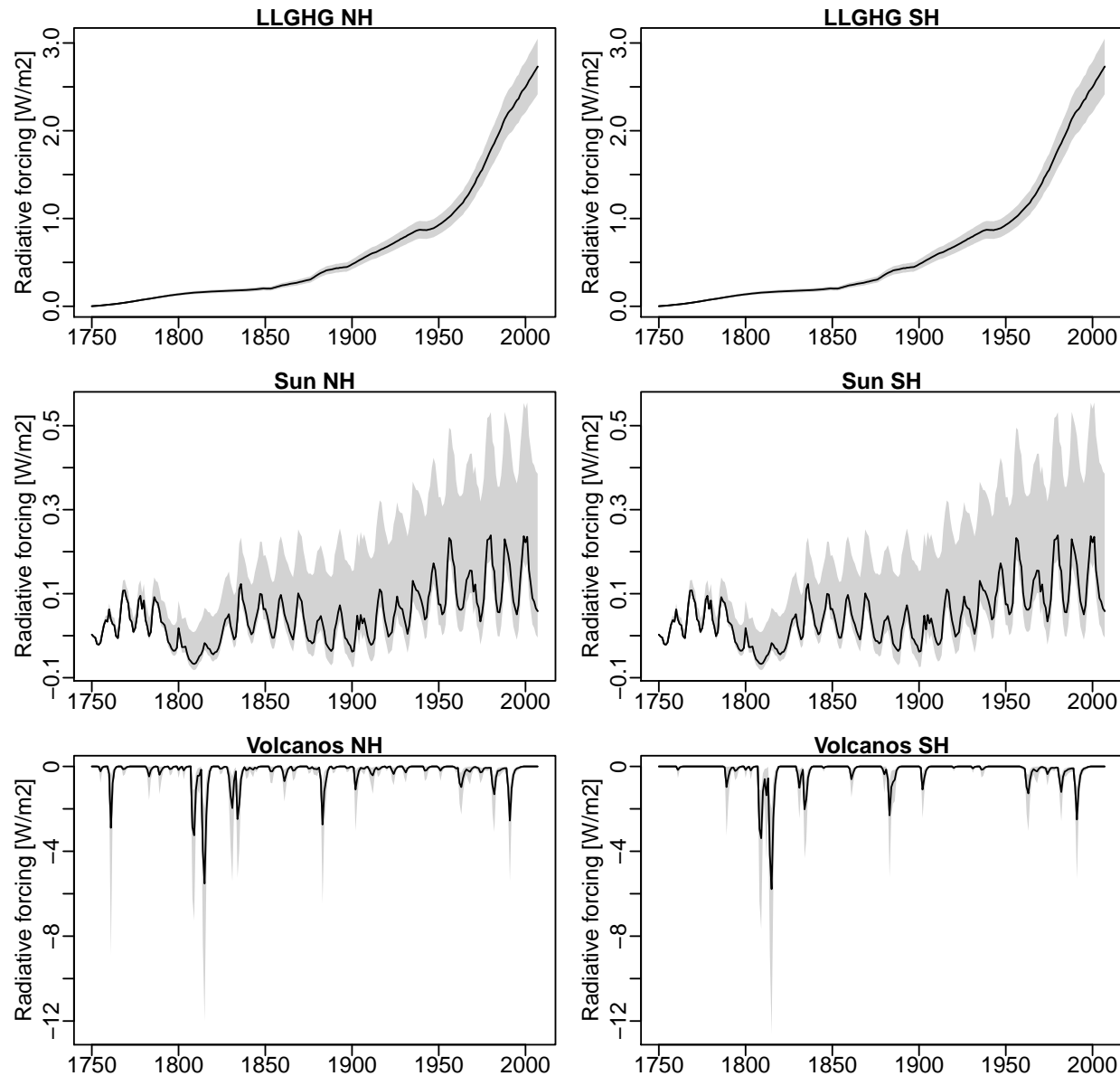
Standard errors



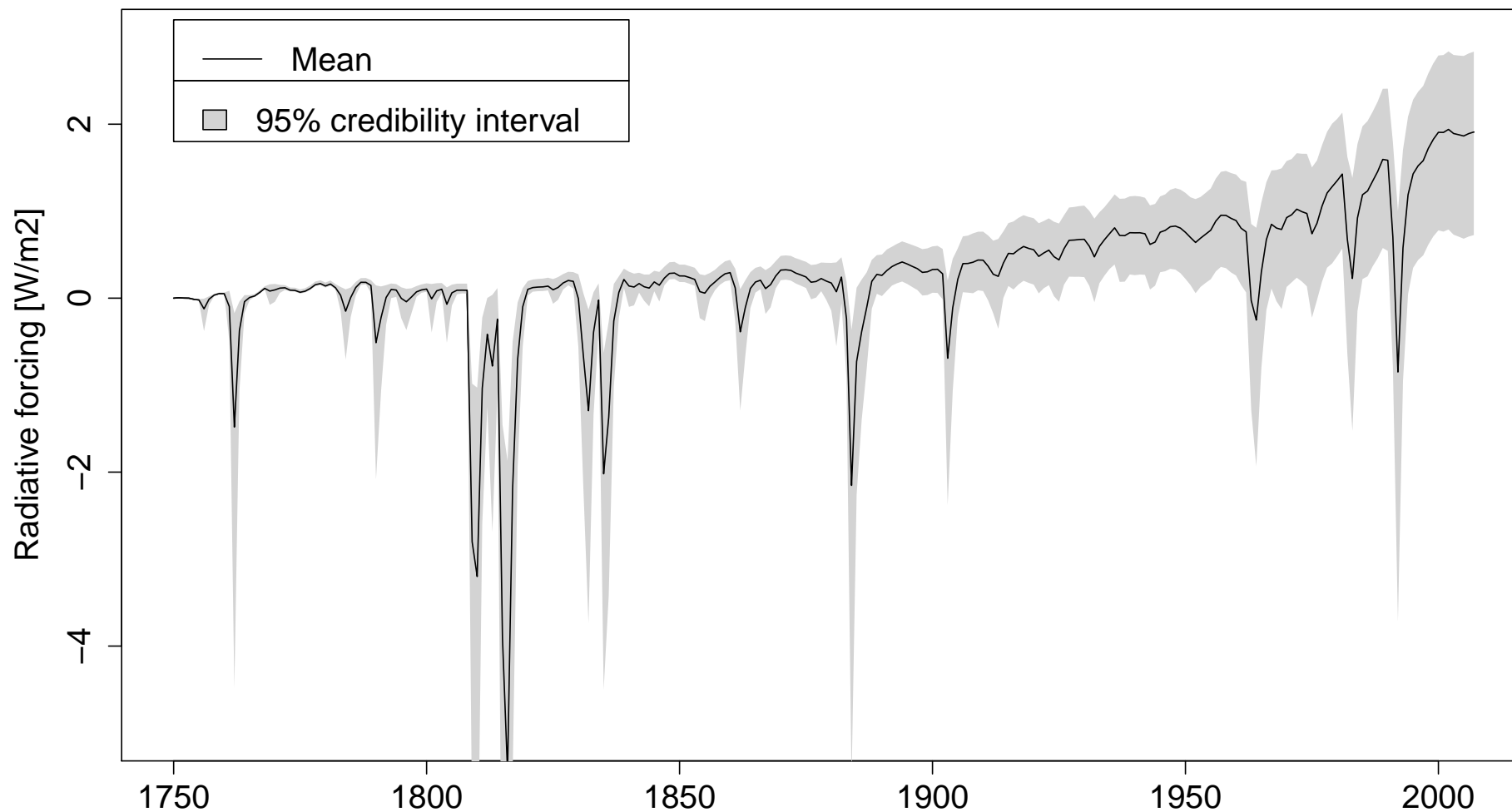
Radiative forcing

- We will specify our best knowledge about historical radiative forcing as prior distributions of 9 independent components, based on temperature-independent estimates of each component, including uncertainties
 - ★ long-lived greenhouse gases
 - ★ direct aerosols
 - ★ indirect aerosols
 - ★ solar radiation
 - ★ volcanoes
 - ★ land use
 - ★ tropospheric ozone
 - ★ stratospheric ozone
 - ★ stratospheric H_2O

Priors of components of radiative forcing



Prior of total radiative forcing



Model for “true” global state of the earth

$$\mathbf{g}_t = (TNH_t, TSH_t, OHC_t)^T$$

Combined deterministic + stochastic model

$$\mathbf{g}_t = \mathbf{m}_t(\mathbf{x}_{t:1750}, S, \boldsymbol{\theta}) + \mathbf{n}_t^m$$

- \mathbf{n}_t^m : model error, dimension 3

Model for observations

$$\mathbf{y}_t = \mathbf{A}\mathbf{g}_t + \beta_0 + \mathbf{n}_t^o$$

- \mathbf{A} : 7x3 matrix copying the northern and southern temperatures 3 times, to compare model with observations
- β_0 : intercept, accounts for different reference periods, dimension 7
- \mathbf{n}_t^o : observational (measurement) error, dimension 7
- Can be difficult to separate model error and observational error, i.e. careful with too strict interpretation

Model error

We assume \mathbf{n}_t^m is VAR(1) (vector autoregressive process of order 1)

$$\mathbf{n}_t^m = \Phi^m \mathbf{n}_{t-1}^m + \boldsymbol{\varepsilon}_t^m$$

Φ^m is diagonal

$$\boldsymbol{\varepsilon}_t^m \sim N(\mathbf{0}, \Sigma^m)$$

elements of $\boldsymbol{\varepsilon}_t^m$ are correlated

Observational error

- We assume \mathbf{n}_t^o is a scaled VAR(1)

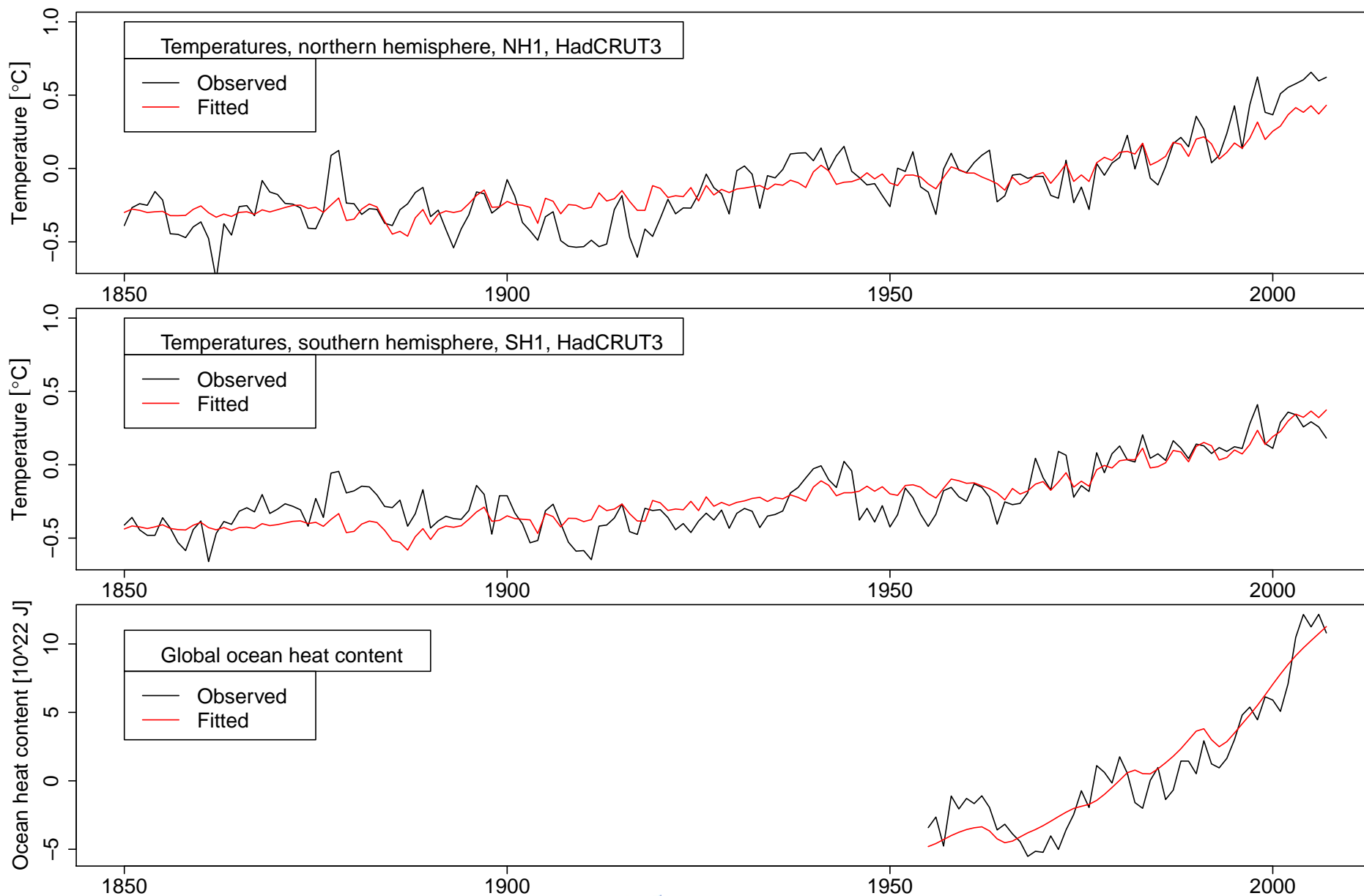
$$\mathbf{n}_t^o = \text{diag}(\mathbf{s}_t) \mathbf{n}_t^{o*}$$

where \mathbf{s}_t is the vector of known observational standard errors
and \mathbf{n}_t^{o*} is VAR(1)

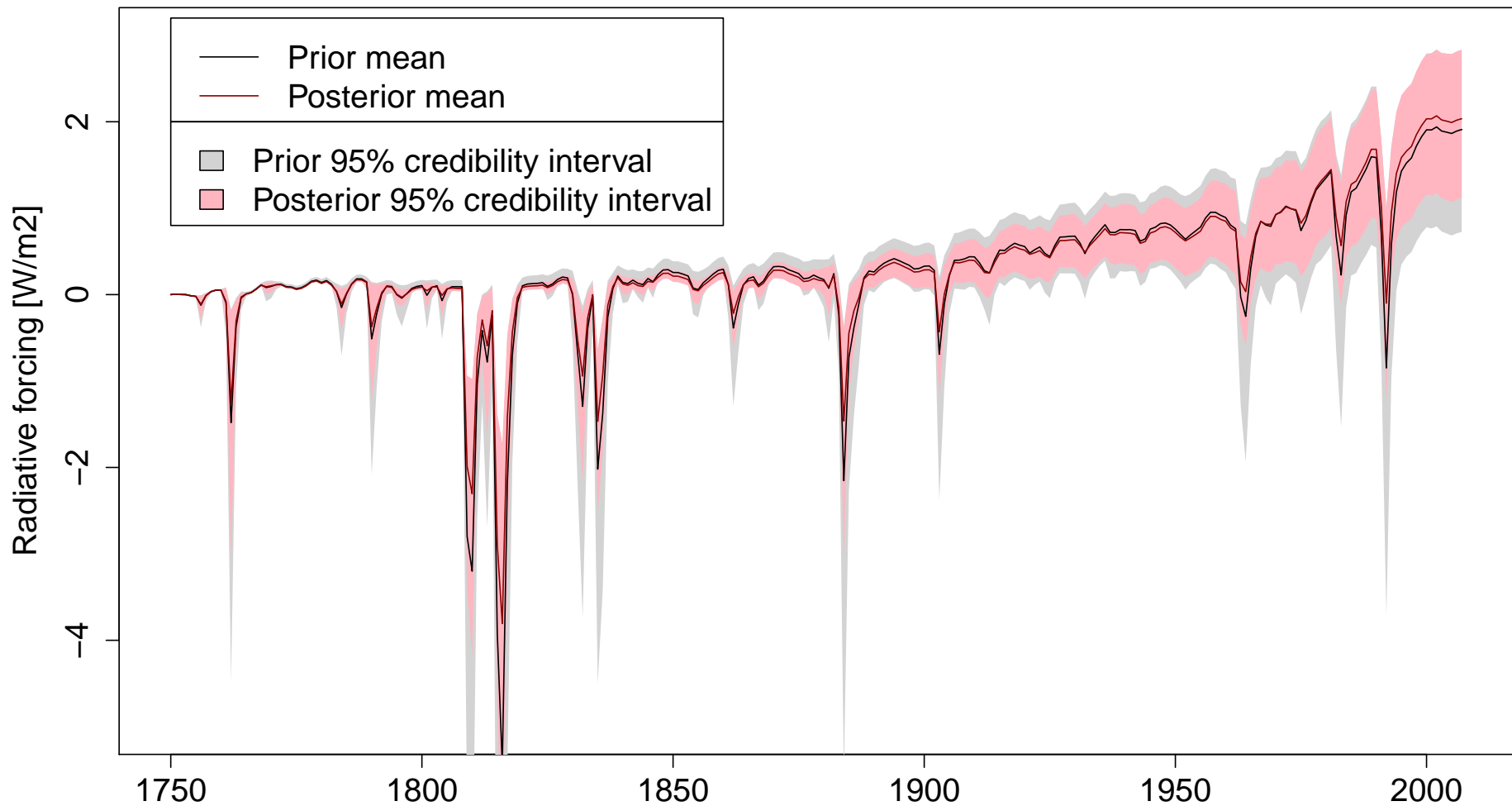
Estimation

- Bayesian approach (Kennedy and O'Hagan 2001), using MCMC
- Vague prior for S
- Informative priors for $\mathbf{x}_{t:1750}$ and θ
- Vague priors for other parameters

Observed and fitted response values



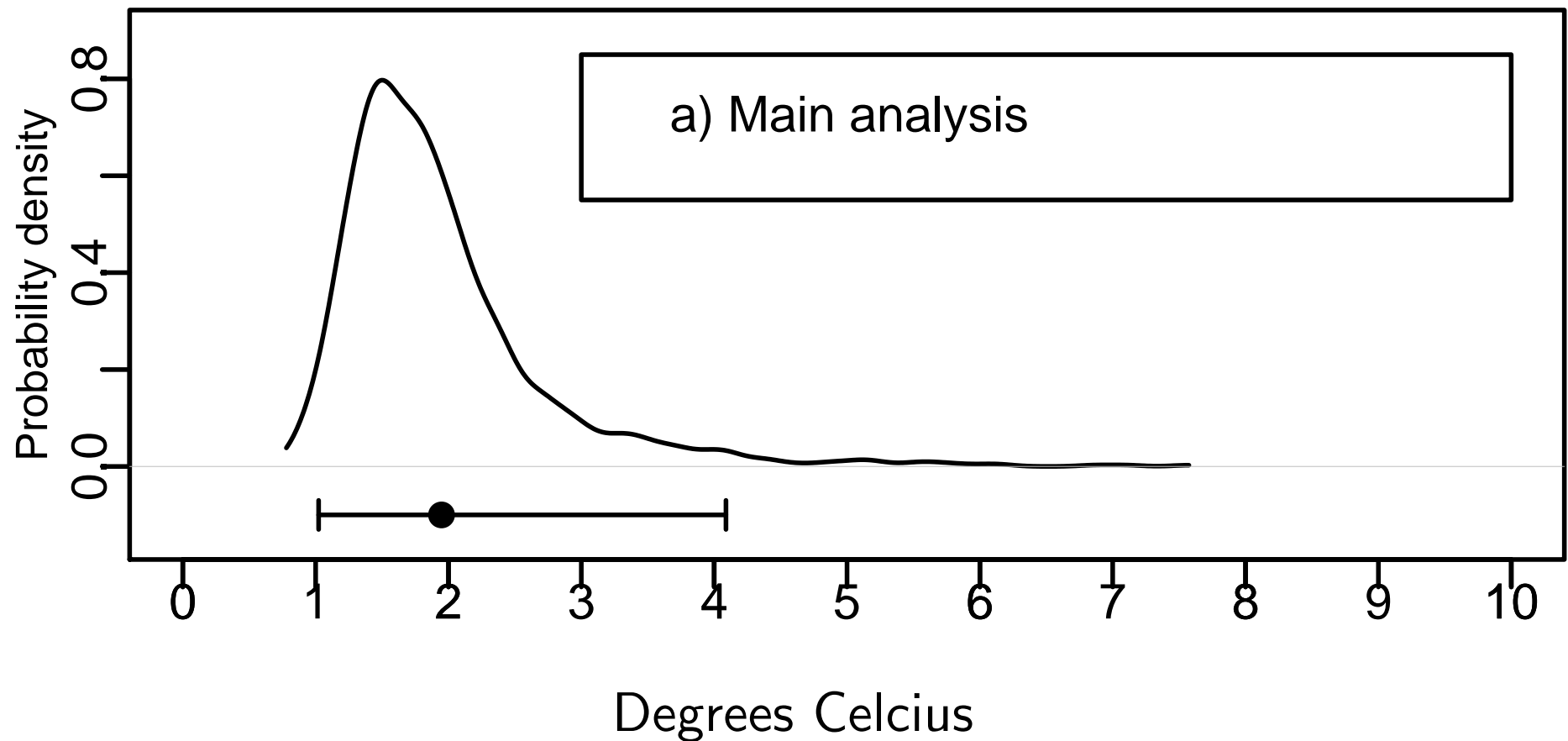
Posterior vs. prior of radiative forcing



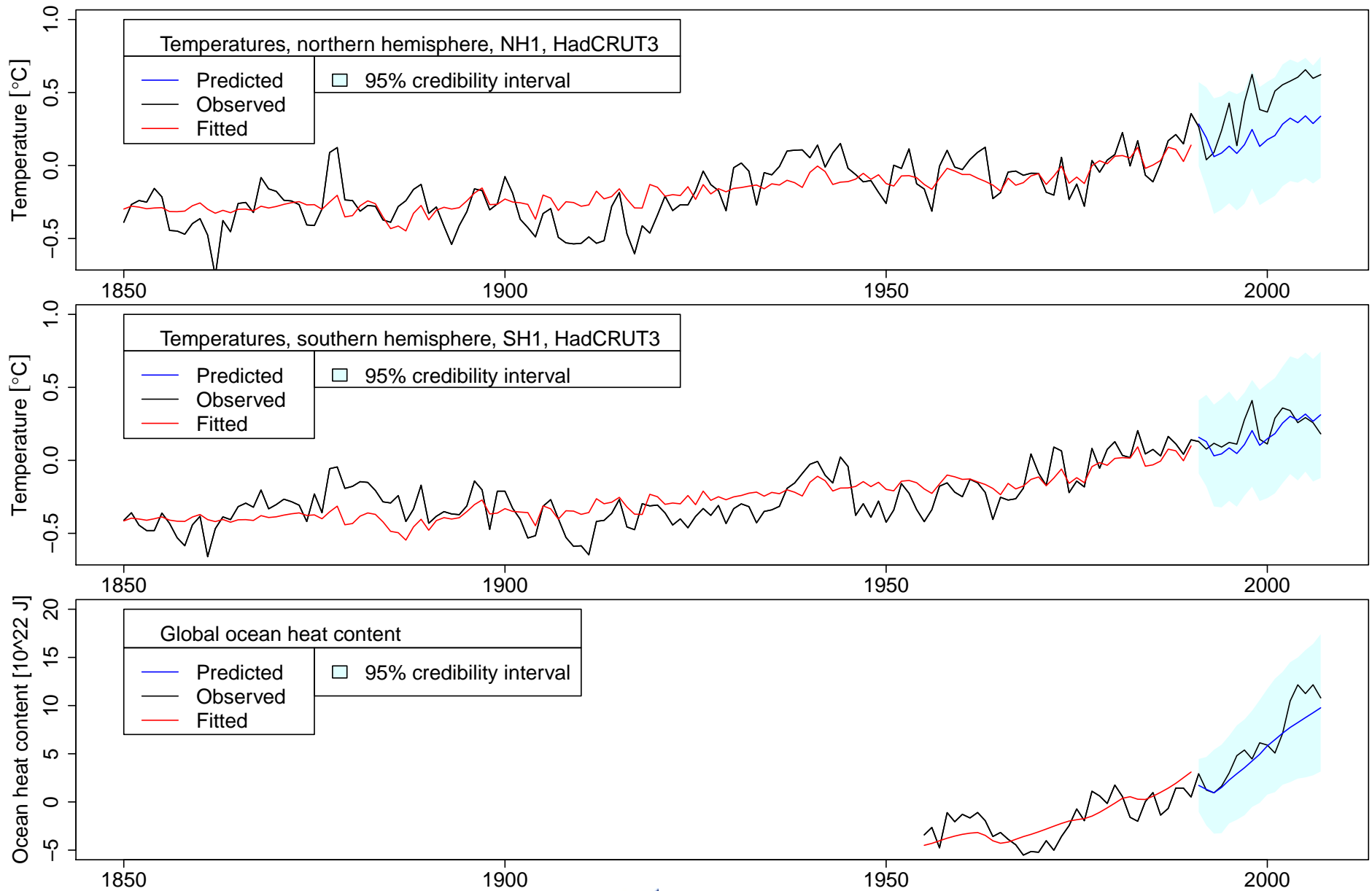
Model and observational errors

- Positive autocorrelations - ϕ 's between 0.5 and 0.7
- Positive correlations between temperature errors
- Temperature errors and ocean heat content errors are uncorrelated

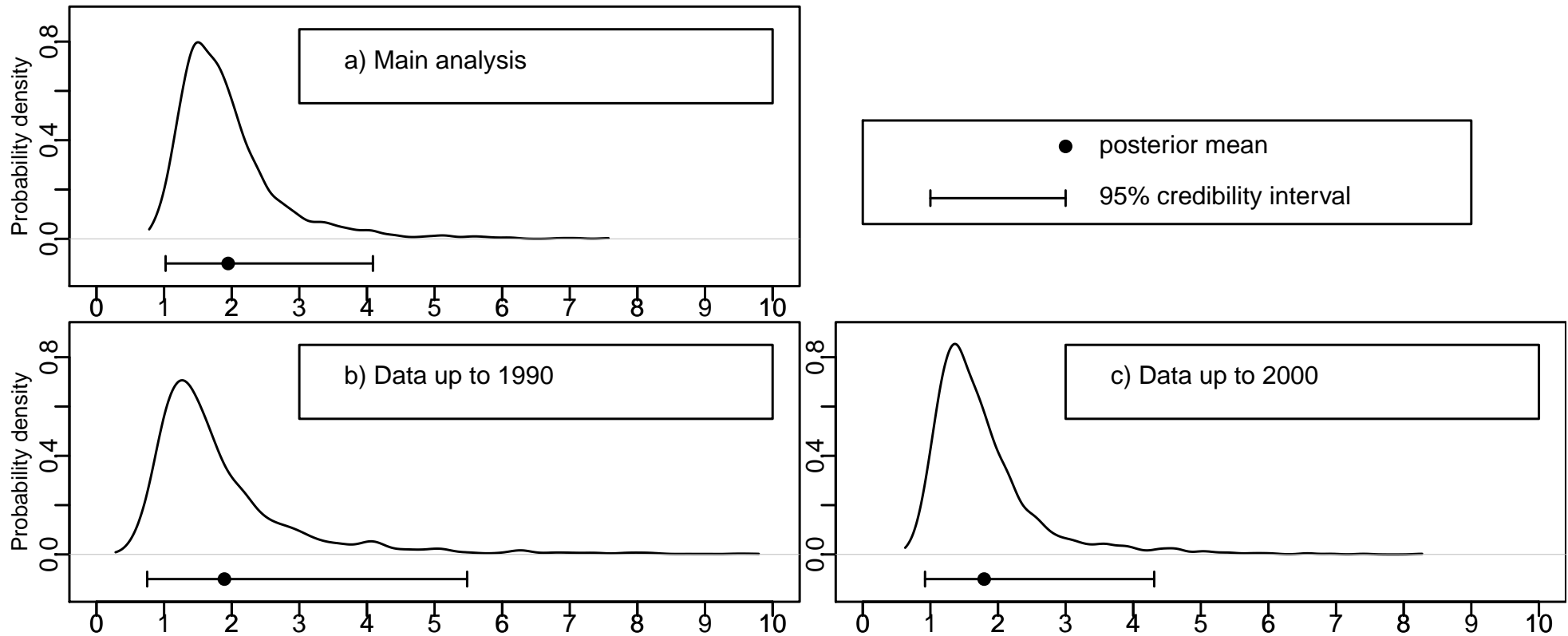
Posterior of the climate sensitivity S



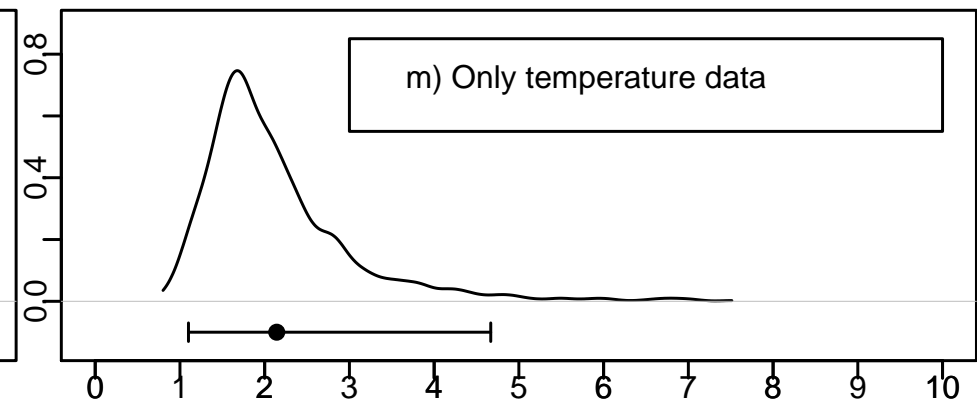
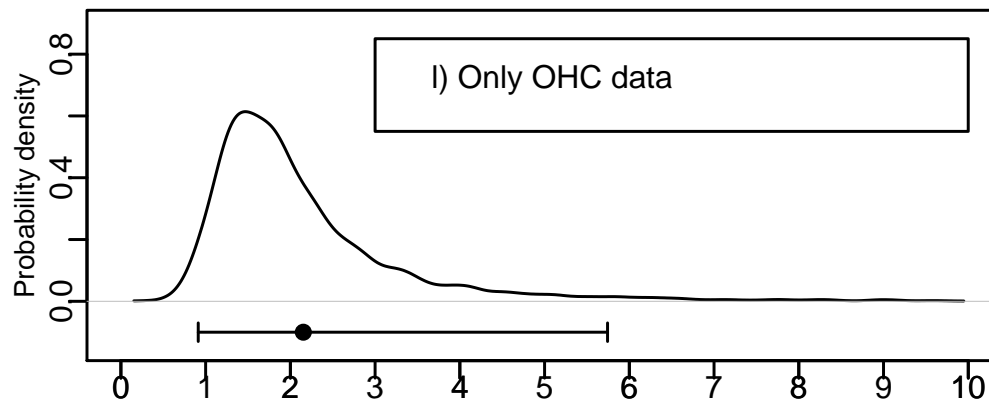
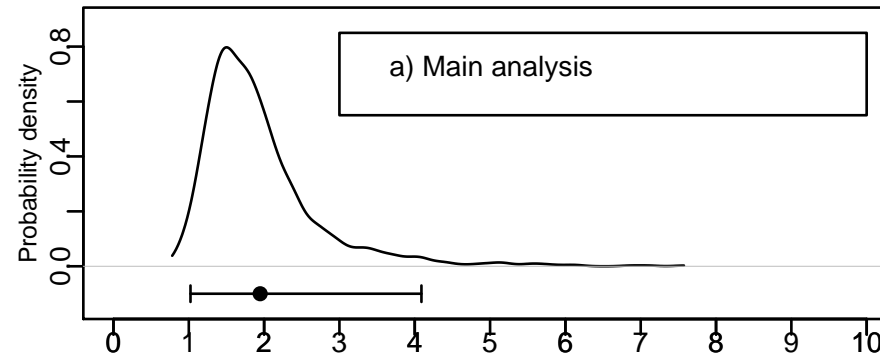
Re-estimation 1850-1990 + prediction 1991-2007



Posteriors for reduced data

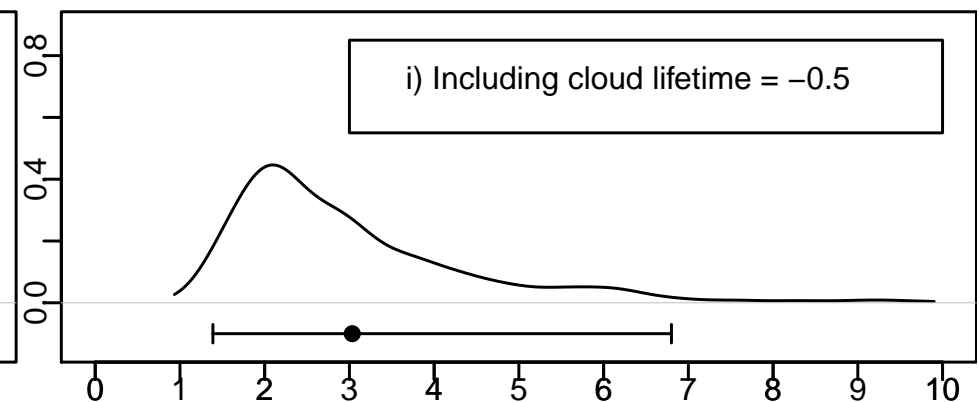
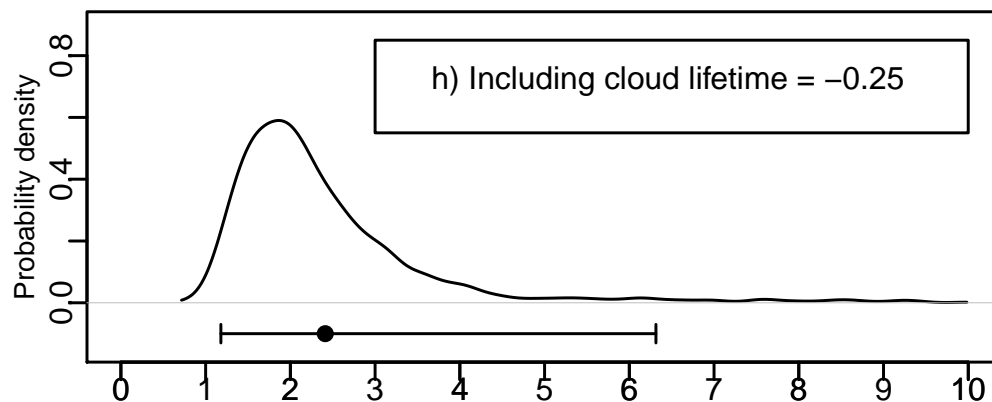
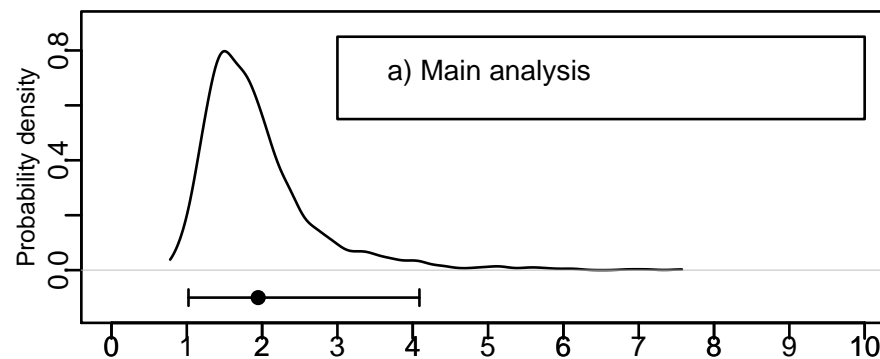


Posteriors for only temperature or OHC



Posteriors including cloud lifetime effect

- Aerosols change the lifetime of clouds
- Cloud lifetime is *not* included a radiative forcing by IPCC (2007)
- But plays a similar role in our approach



Further work

- Include the cloud lifetime effect with uncertainty
- Update priors of forcing,
more precise estimates of historical forcing will be available soon
- Include data from 2008 and 2009

Thank you for your attention!



Other approaches i)

Tommasini, Reichert, Kunsch, Buser, Knutti and Borsuk (2009) in
Applied Statistics

$$y_t = \mathbf{m}_t(\mathbf{x}_{t:1750}, S, \boldsymbol{\theta}) + \mathbf{n}_t^o$$

- Similar model \mathbf{m}_t
- Only one temperature series
- No autocorrelation in observational error \mathbf{n}_t^o for temperature
- No intercept
- No model error term, all model error due to error in forcing
- No informative prior for forcing

$$x_t = \mu_t + \phi_t = \text{mean} + \text{random}$$

Other approaches ii)

Sanso and Forest (2009) in Applied Statistics

- Medium complex climate model - computer intensive
- Emulator
- Response is aggregated temperature data
 - ★ $\text{mean}(1946-1955) - \text{mean}(1905-1995)$,
 $\text{mean}(1956-1965) - \text{mean}(1905-1995)$,
 $\text{mean}(1966-1975) - \text{mean}(1905-1995)$,
 $\text{mean}(1976-1985) - \text{mean}(1905-1995)$,
 $\text{mean}(1986-1995) - \text{mean}(1905-1995)$
 - ★ Earth divided into four zonal bands
 - ★ 20 dimensional vector

Other approaches iii)

Gregory, Raper, Stott and Rayner (2002) in Journal of Climate

$$S = 3.71 \cdot \Delta T / (\Delta RF - \Delta OHC)$$

- ΔT = mean temp 1957-1994 - mean temp 1861-1900
- ΔRF = mean RF 1957-1994 - mean RF 1861-1900
- ΔOHC = mean OHC 1957-1994 - mean OHC 1861-1900