

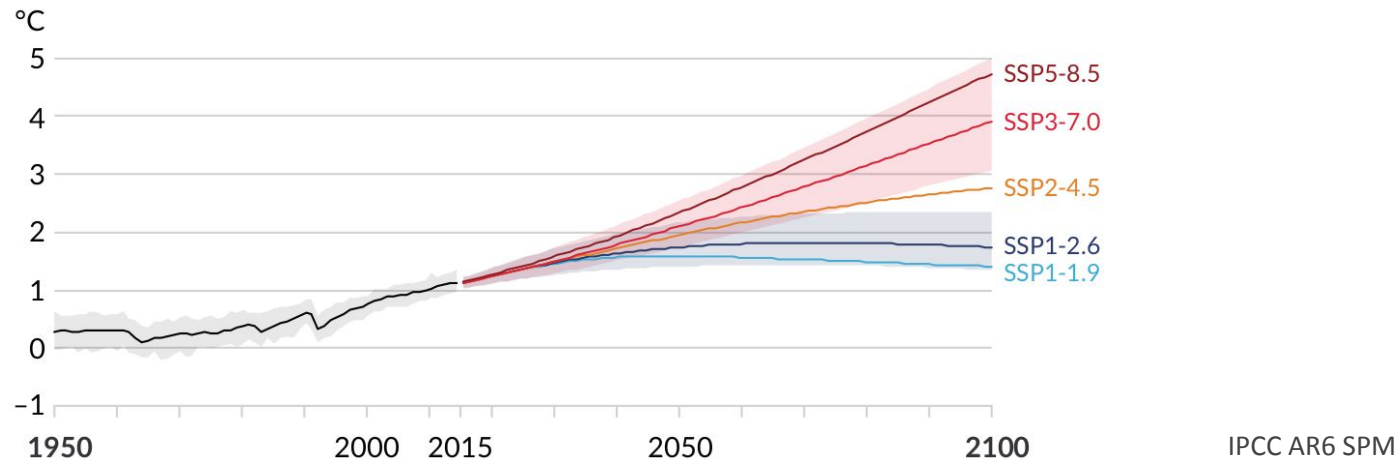
Separating Uncertainties in Projections of Future Climate – Importance, Lessons, and Solutions

Lukas Brunner | CliMatters Workshop | February 3rd 2022

Many thanks to all collaborators!

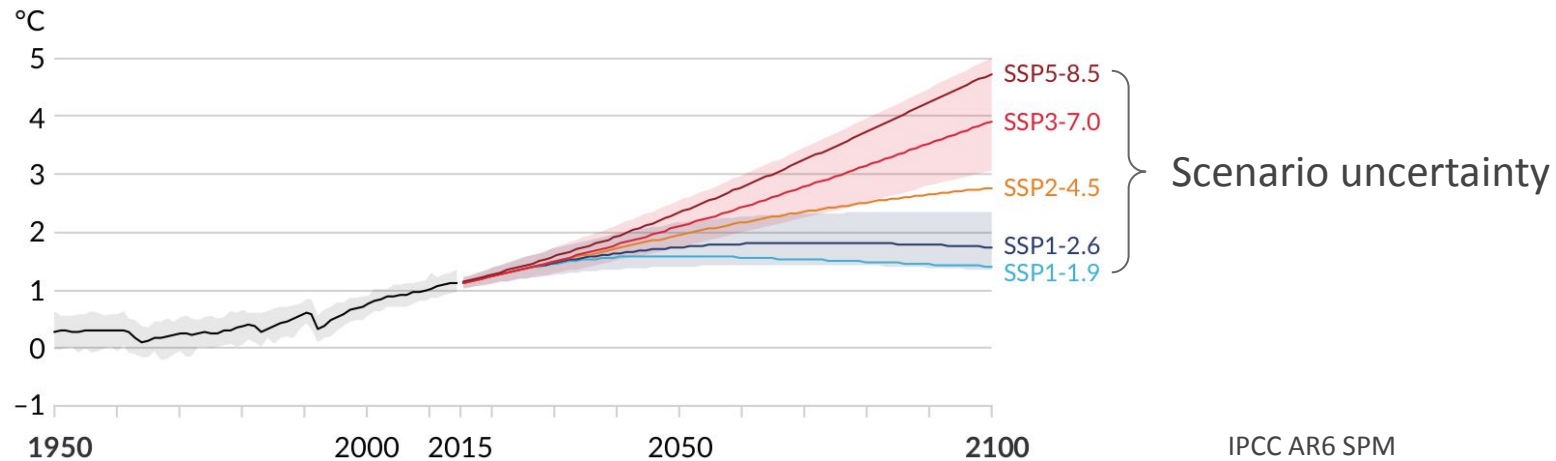
Sources of uncertainty in projections of future climate

(a) Global surface temperature change relative to 1850–1900



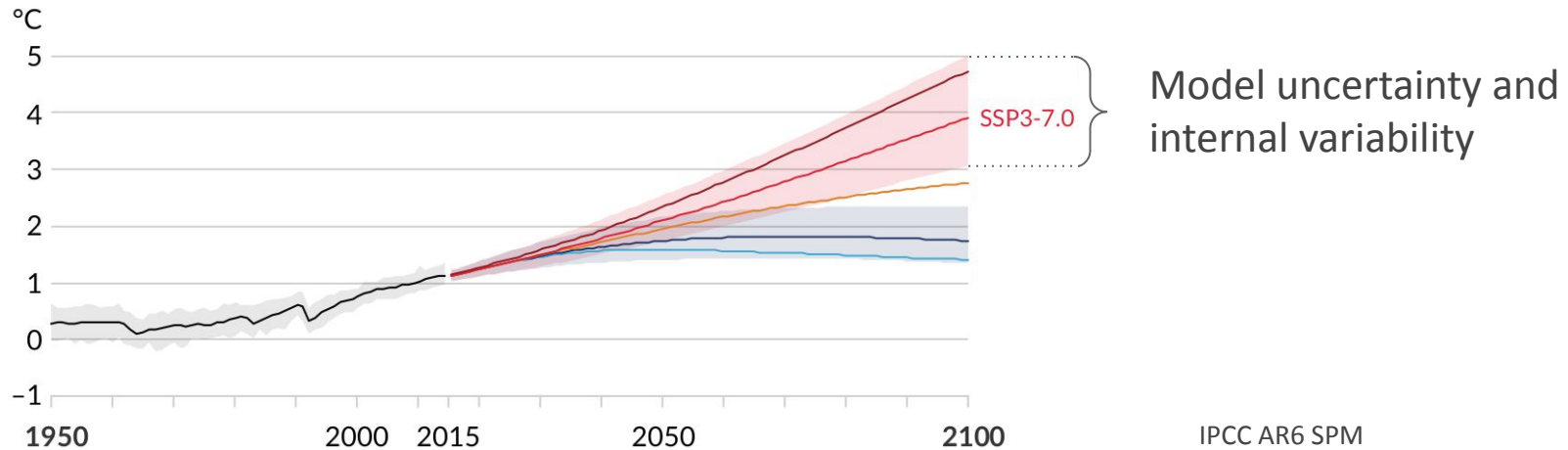
Sources of uncertainty in projections of future climate

(a) Global surface temperature change relative to 1850–1900



Sources of uncertainty in projections of future climate

(a) Global surface temperature change relative to 1850–1900





Interactions between uncertainties

Interaction between uncertainties I: robust changes

The multi-model mean potentially hides the interaction between model uncertainty and internal variability

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unpublished
figure

Precipitation change in the period 2081-2100 relative to 1986-2005 (SRES-A2 / RCP8.5 / SSP5-8.5). Brunner et al. (in preparation)



Interaction between uncertainties I: robust changes

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figure

Robustness is a combination of the change signal in each model and the difference between models → **Model uncertainty**

Each model can show **significant** change → **internal variability**

Areas with low robustness but high significance are marked as **inconsistent**

Precipitation change in the period 2081-2100 relative to 1986-2005 (SRES-A2 / RCP8.5 / SSP5-8.5). Brunner et al. (in preparation)



Interactions between uncertainties II: some lessons from EUCP

Challenge: compare different methods to constrain uncertainty in future changes

Comparing Methods to Constrain Future European Climate Projections Using a Consistent Framework

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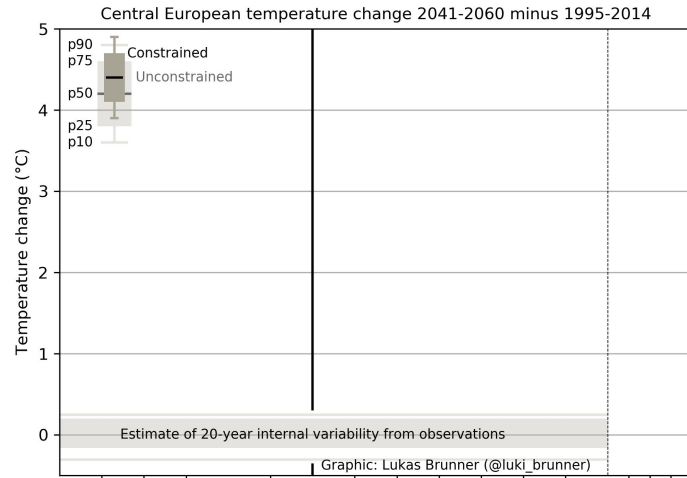
^e Fondazione Centro Euro-Mediterraneo sui Cambiamenti Climatici, Bologna, Italy

^f The Abdus Salam International Centre for Theoretical Physics, Trieste, Italy

^g Royal Netherlands Meteorological Institute, De Bilt, Netherlands

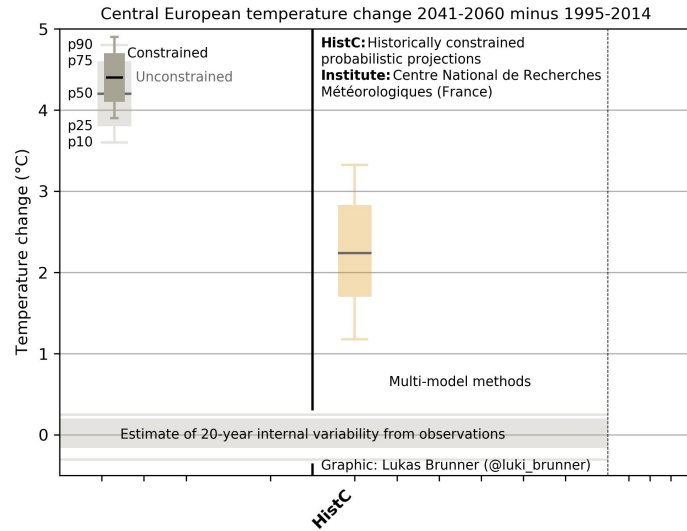
^h CNRM, Université de Toulouse, Météo-France, CNRS, Toulouse, France

Comparing constrained projections from different methods



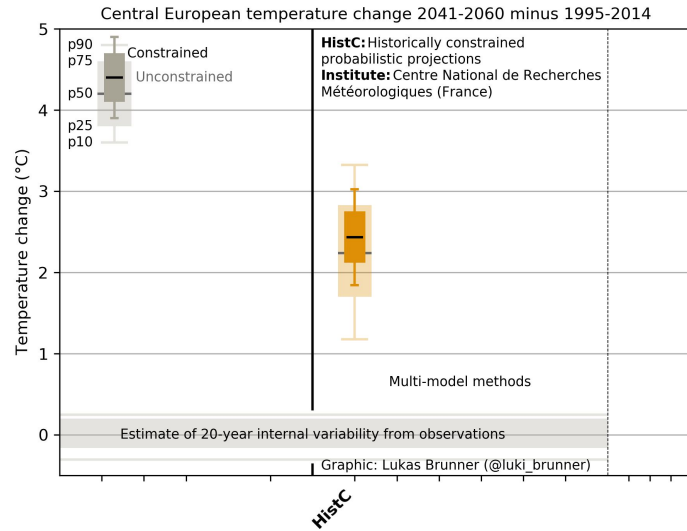
Brunner et al. (2020)

Comparing constrained projections from different methods



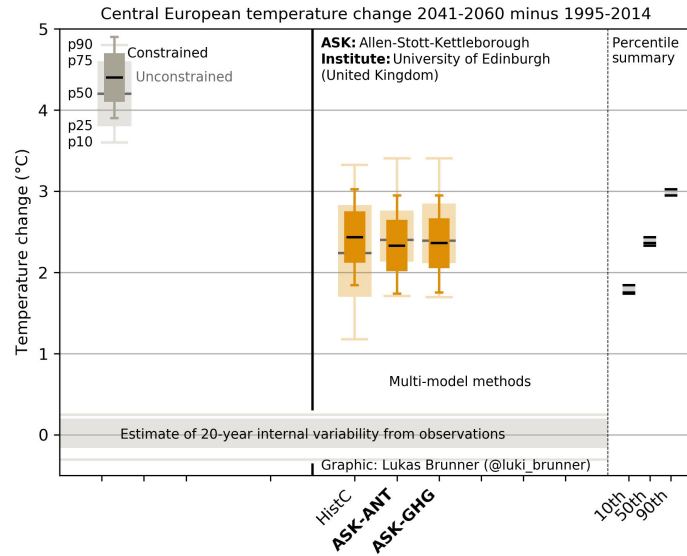
Brunner et al. (2020)

Comparing constrained projections from different methods



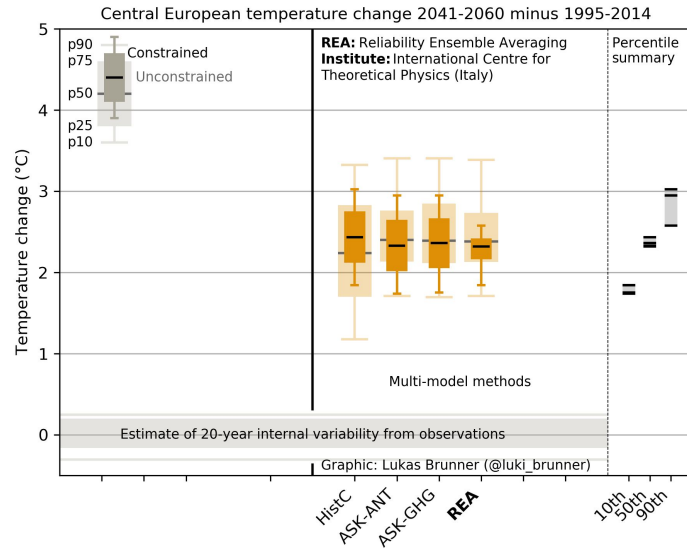
Brunner et al. (2020)

Comparing constrained projections from different methods



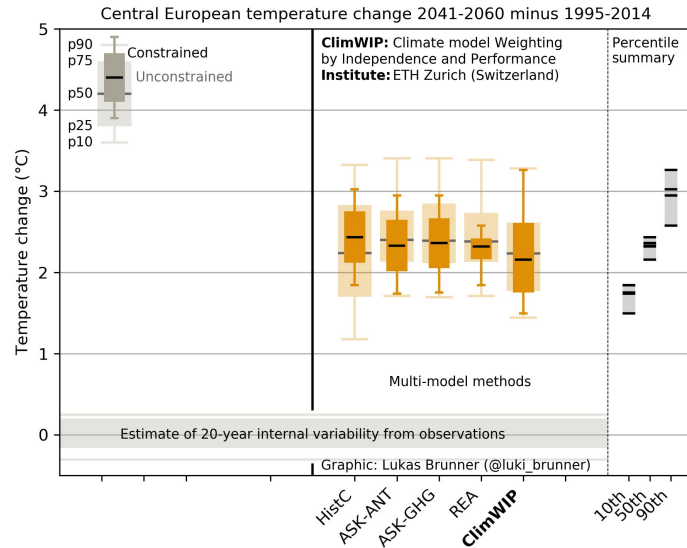
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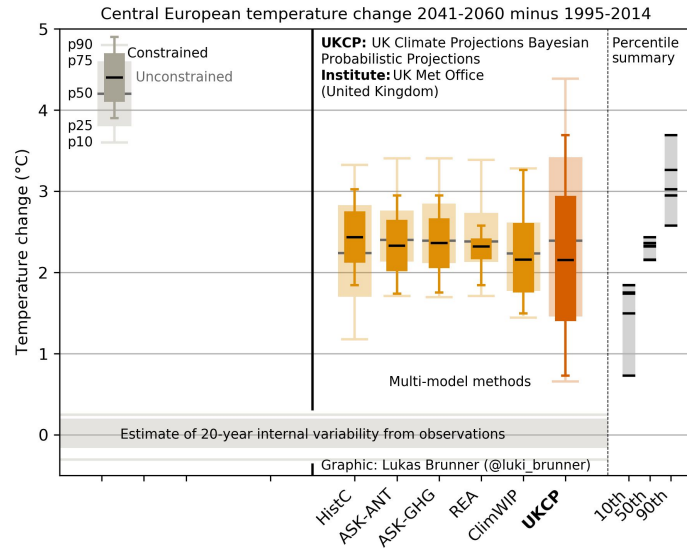
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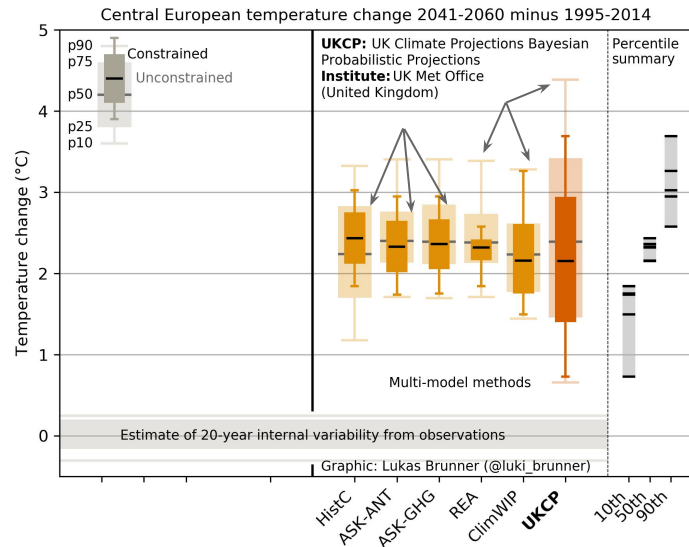
Brunner et al. (2020)

Comparing constrained projections from different methods



Brunner et al. (2020)

Comparing constrained projections from different methods

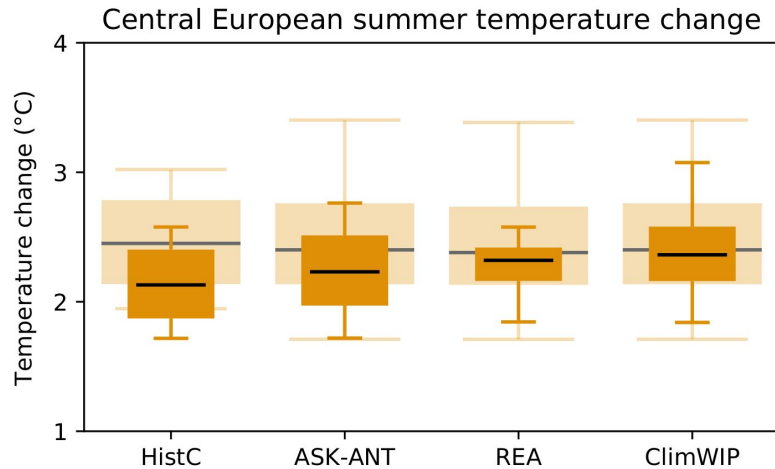


Why are the **unconstrained** projection uncertainties different?

- different models used
- handling of multiple ensemble members
- internal variability included
- parameter uncertainty included
- calculation of percentiles (Gaussian assumption)

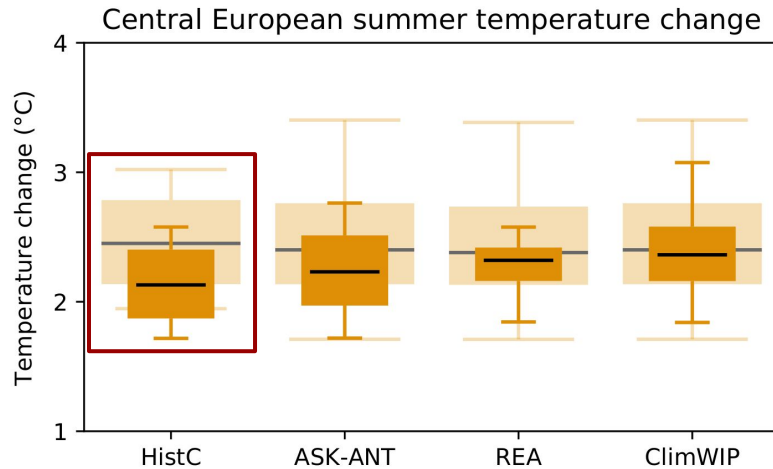
Brunner et al. (2020)

Consistency versus methods available



- ✓ same model pool
- ✓ internal variability included
- ✓ no parameter uncertainty

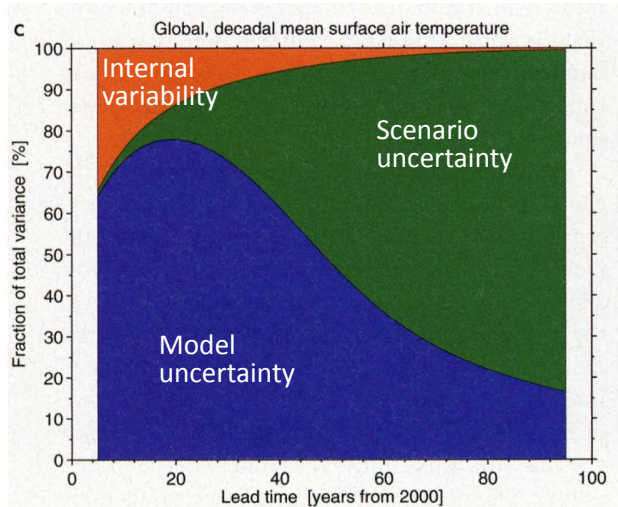
Consistency versus methods available



- ✓ same model pool
- ✓ internal variability included
- ✓ no parameter uncertainty
- ✗ calculation of percentiles
- ✗ separation of internal variability and model uncertainty

Separating uncertainties

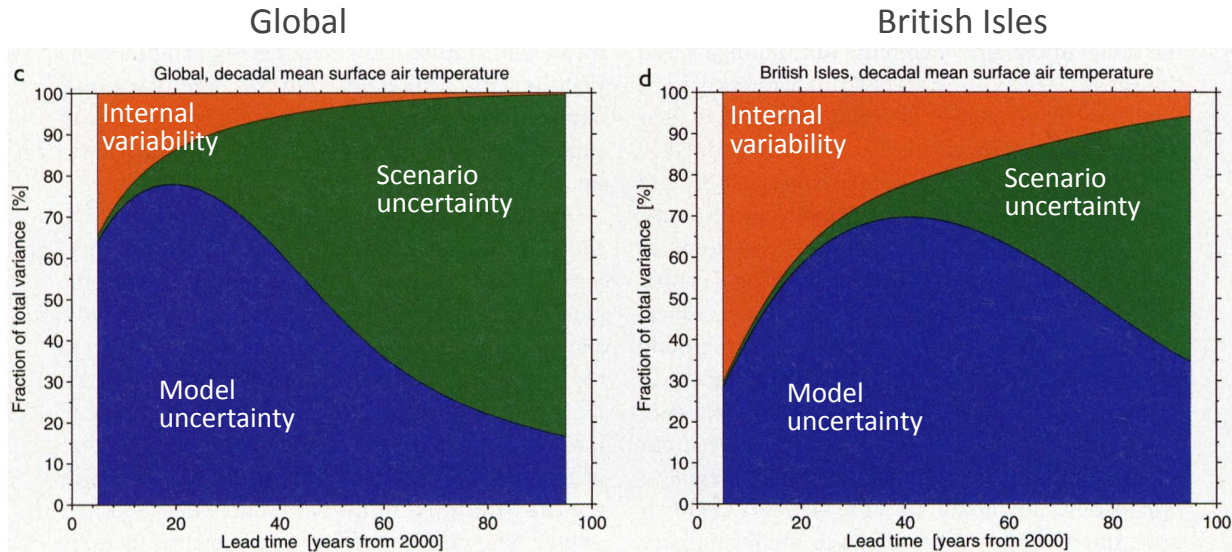
Fractional contributions to total uncertainty (CMIP3)



Total uncertainty =
Internal variability +
Scenario uncertainty +
Model uncertainty

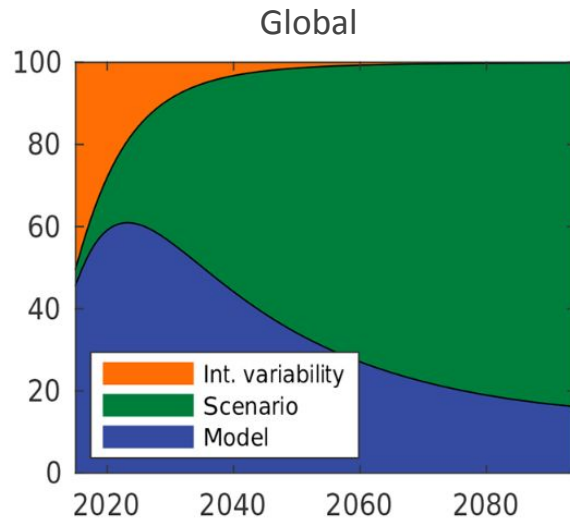
Hawkins and Sutton (2009)

Fractional contributions to total uncertainty (CMIP3)



Hawkins and Sutton (2009)

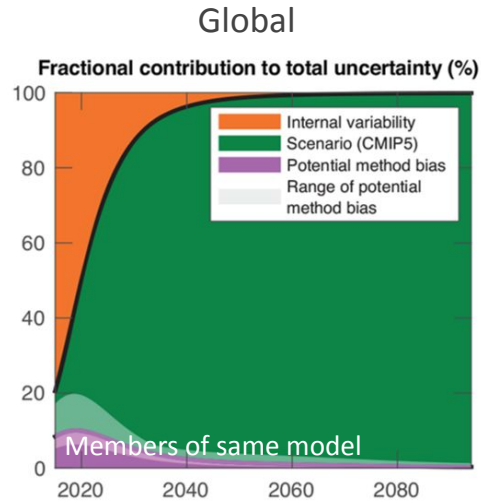
Fractional contributions to total uncertainty (CMIP5)



Lehner et al. (2020)

Quantifying method bias using large ensembles

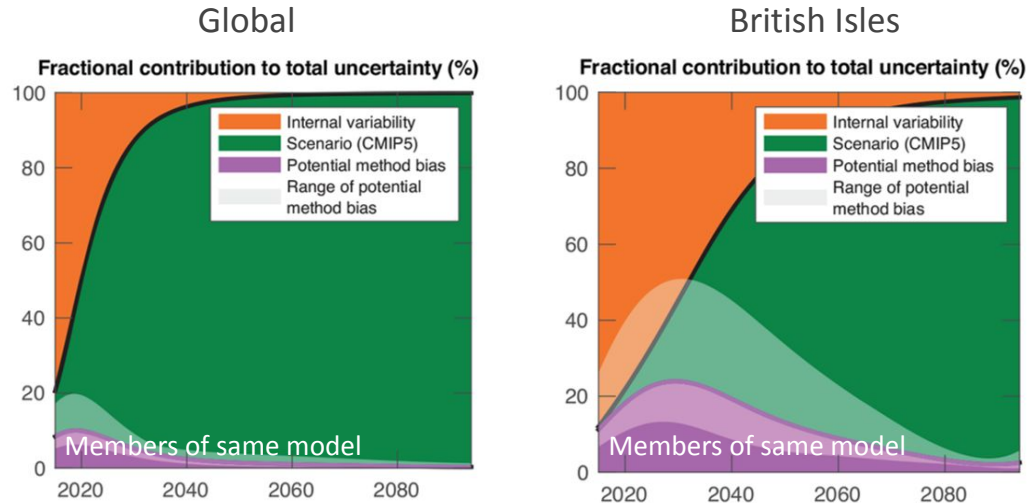
The model uncertainty for a single model should be zero by definition!



Lehner et al. (2020)

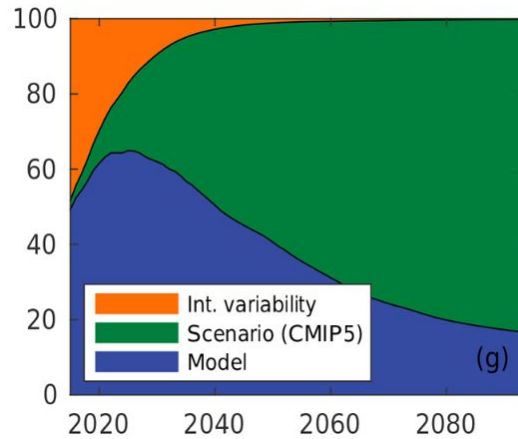
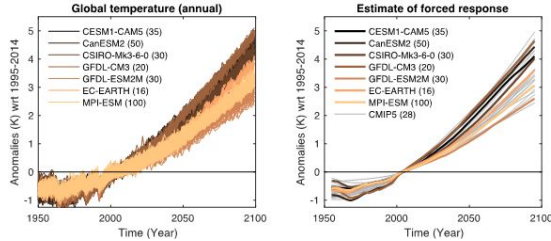
Quantifying method bias using large ensembles

The model uncertainty for a single model should be zero by definition!



Lehner et al. (2020)

Estimating the forced response directly using large ensembles



Lehner et al. (2020)

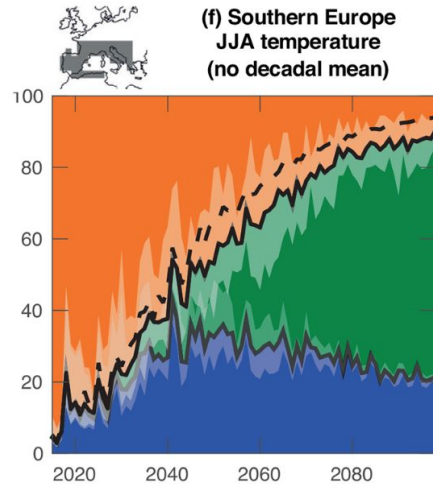
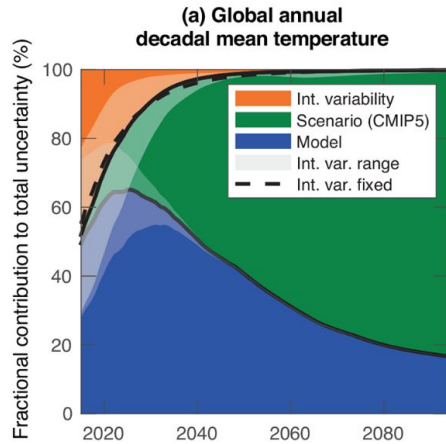
Summary and Conclusions

- 3 main sources of uncertainty in projections of future climate
 - scenario uncertainty
 - **model uncertainty**
 - **internal variability**
- Model uncertainty & internal variability can be hard to separate but their interaction is important, e.g., for
 - **robustness of future changes**
 - **consistent comparison of expected changes**
- Large ensembles (SMILEs) can help to better quantify the contributions from different sources of uncertainty

Literature

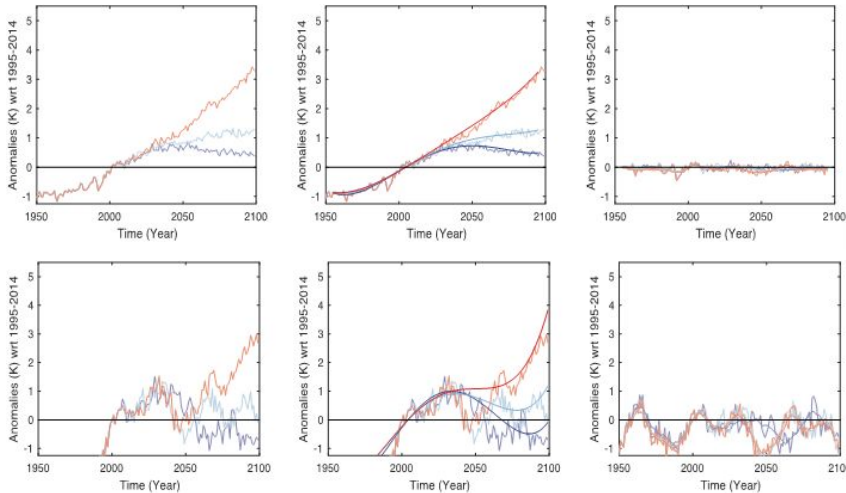
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- Hawkins, E., & Sutton, R. (2009). The Potential to Narrow Uncertainty in Regional Climate Predictions. *Bulletin of the American Meteorological Society*, 90(8), 1095–1108. <https://doi.org/10.1175/2009BAMS2607.1>
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Estimating changes in internal variability



Potential method bias

Is a 4th order polynomial appropriate to estimate the forced response?

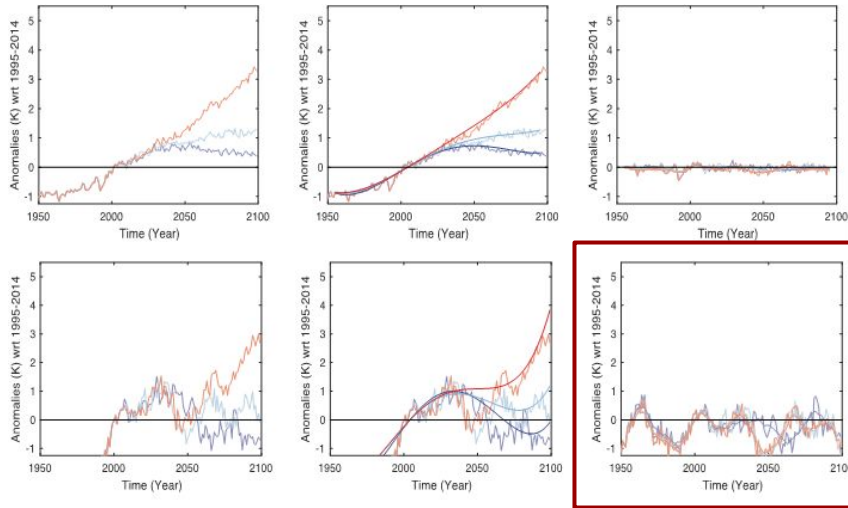


Global, decadal mean temperature

Southern Ocean, decadal mean
temperature. Courtesy: FL

Potential method bias

Is a 4th order polynomial appropriate to estimate the forced response?

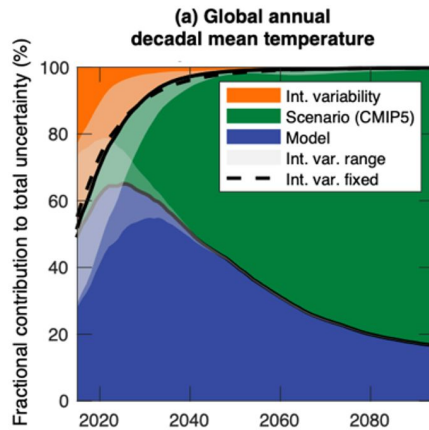


Global, decadal mean temperature

Southern Ocean, decadal mean temperature. Courtesy: FL

Estimating the amplitude of internal variability

What is the correct value for internal variability?

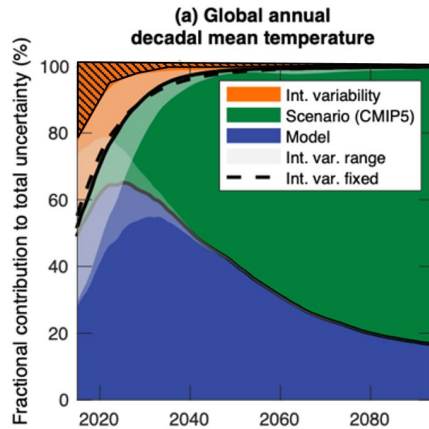


Different models give different estimates of internal variability

Lehner et al. (2020)

Estimating the amplitude of internal variability

What is the correct value for internal variability?

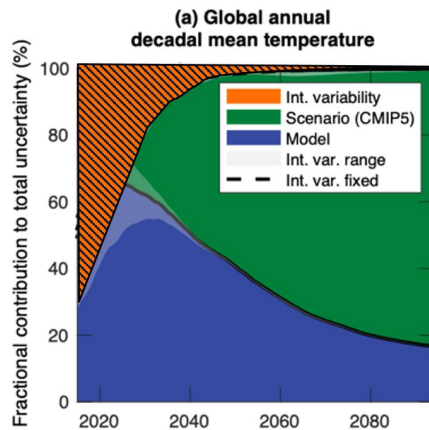


Model with low internal variability
(EC-Earth)

Lehner et al. (2020)

Estimating the amplitude of internal variability

What is the correct value for internal variability?

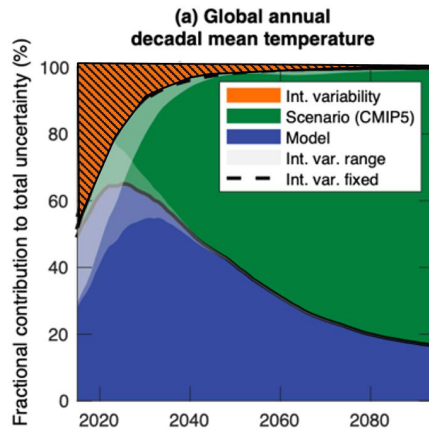


Model with high internal variability
(GFDL-CM3)

Lehner et al. (2020)

Estimating the amplitude of internal variability

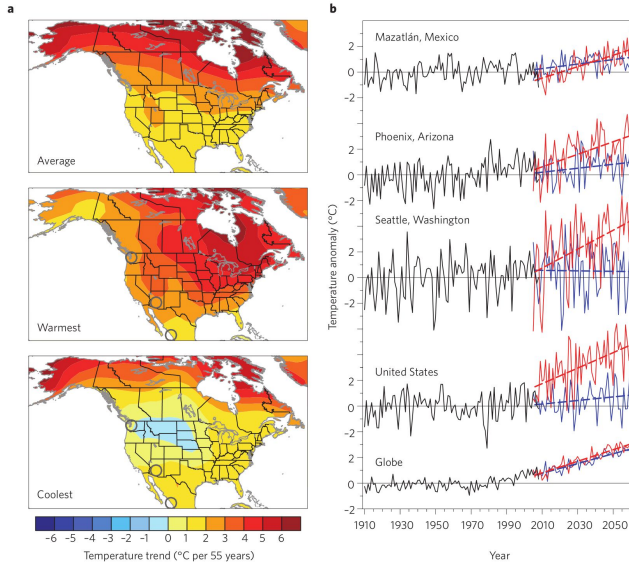
What is the correct value for internal variability?



Multi-model mean internal variability

Lehner et al. (2020)

Effects of internal variability



Locally internal variability can lead to differences in trends even within one model

Deser et al. (2012)