

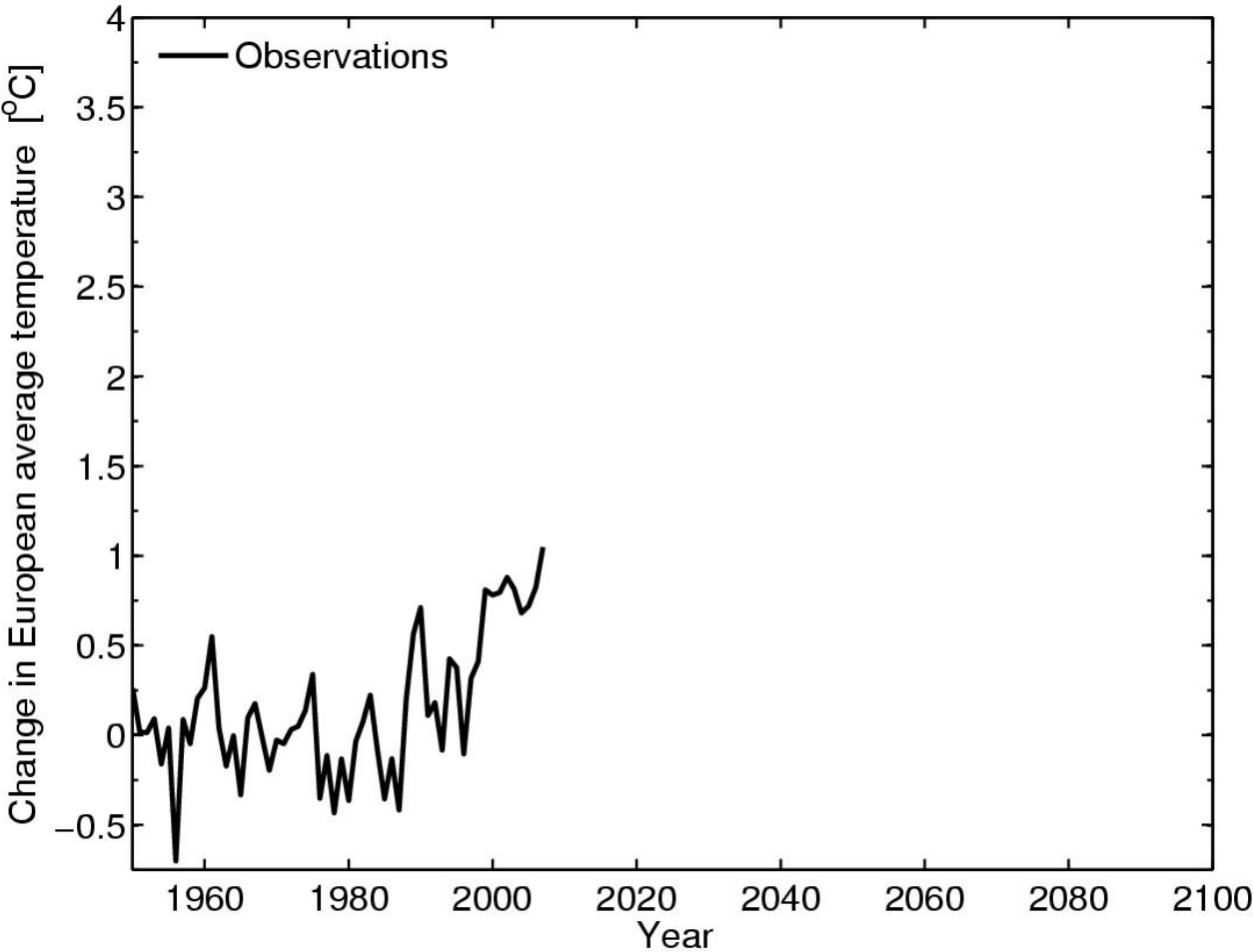
Estimation of optimal perturbations for decadal climate predictions

Ed Hawkins

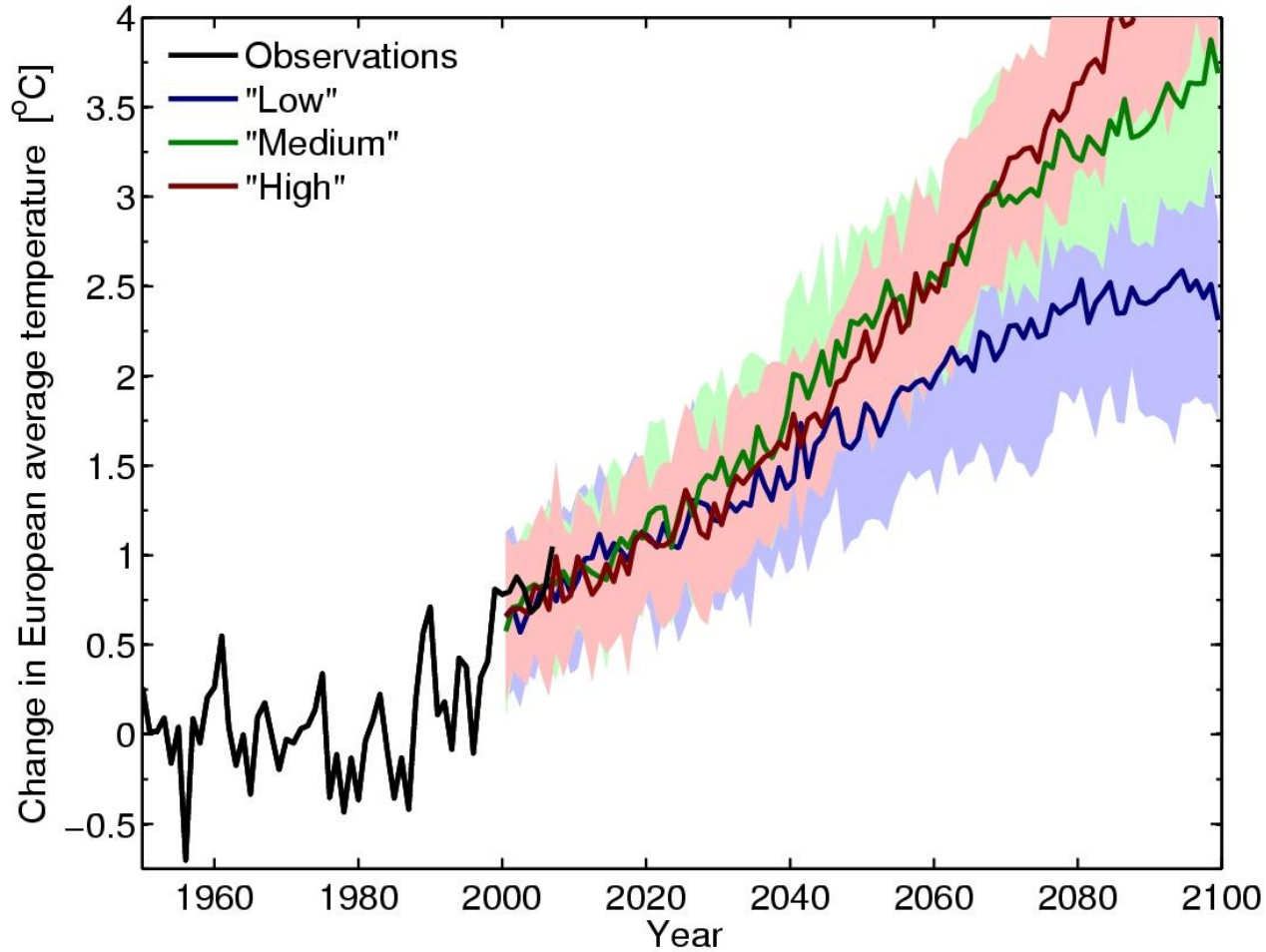
Thanks to: Rowan Sutton

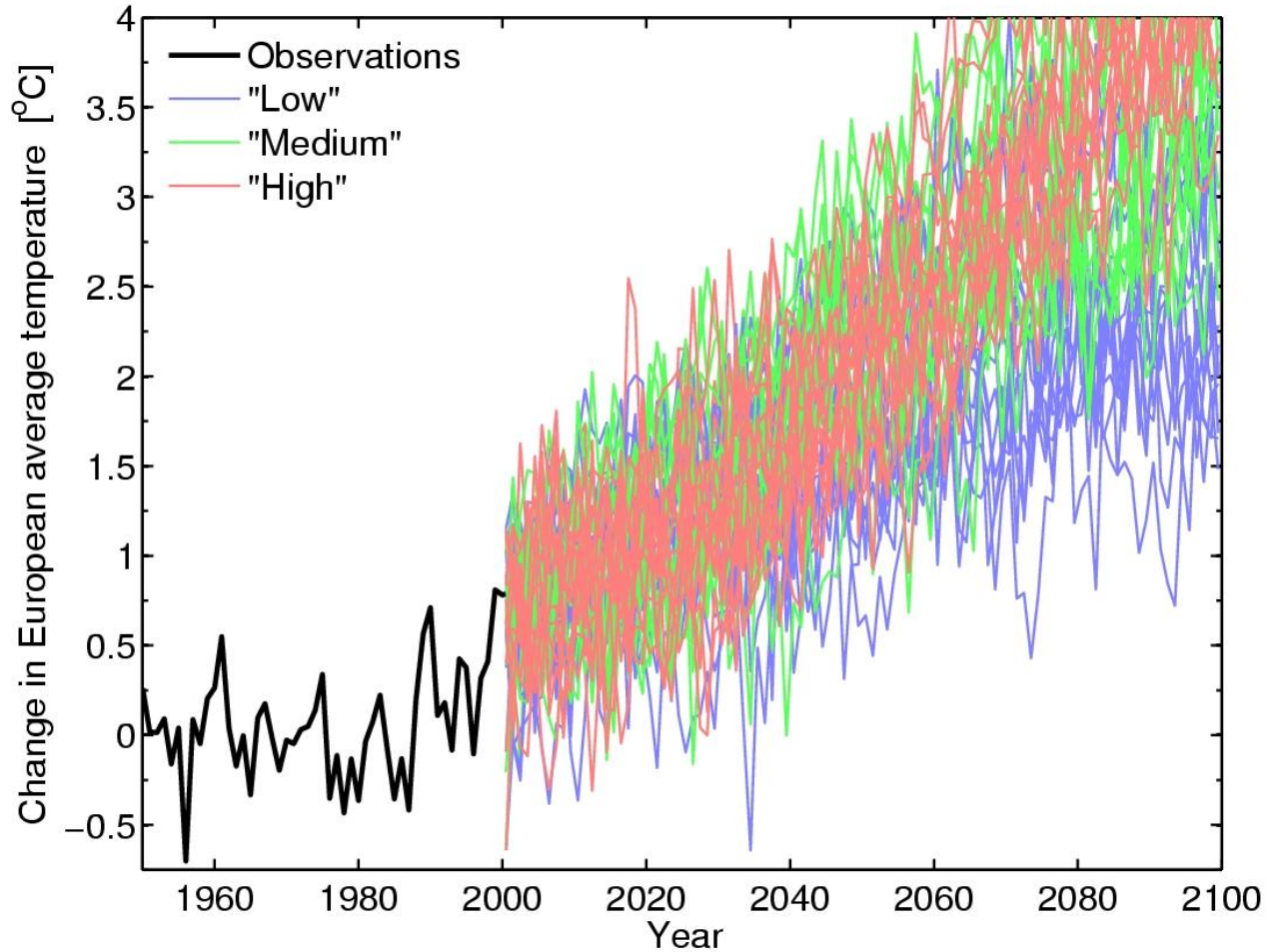
DSWC09 workshop

EU temperature projections

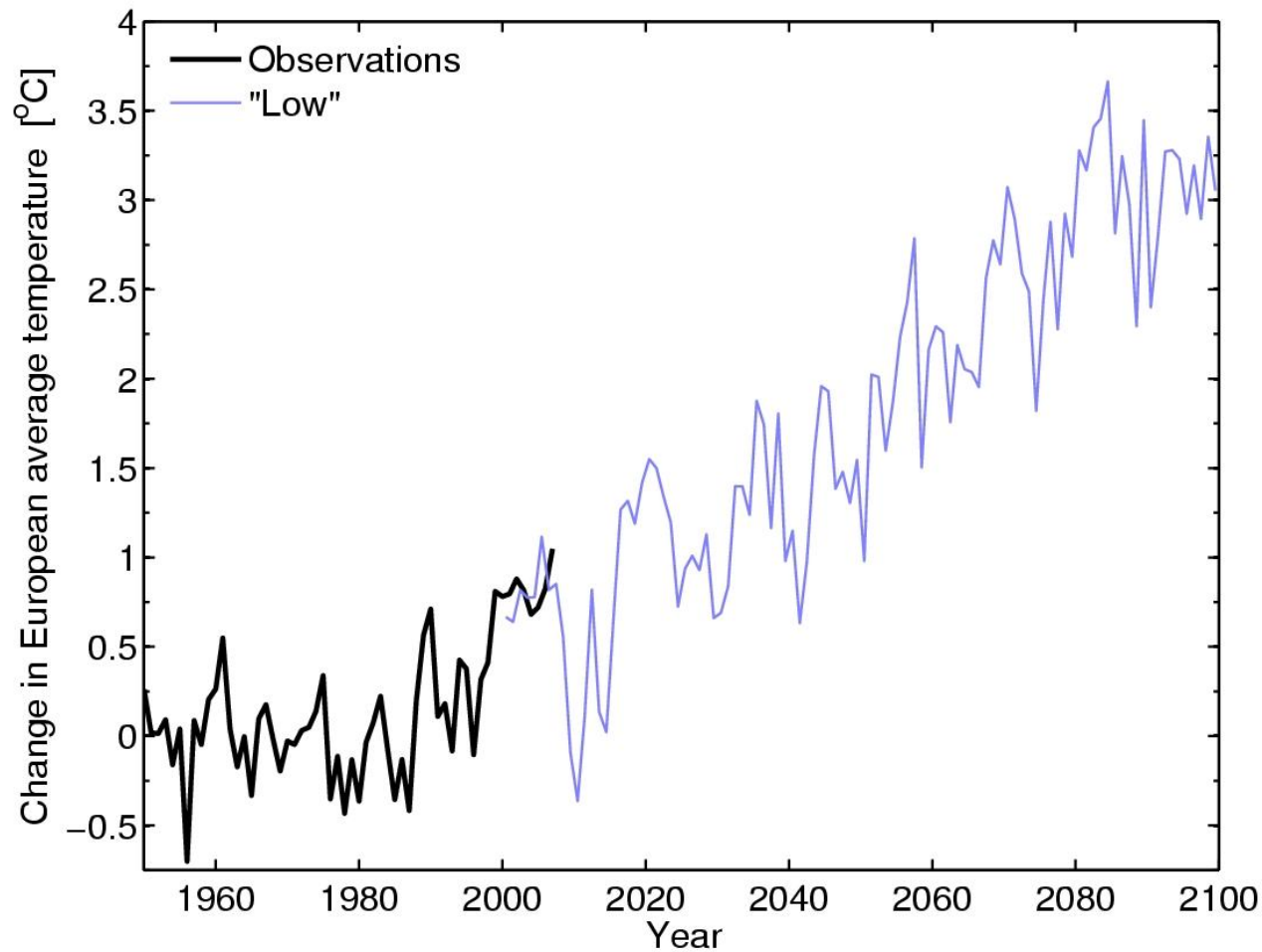


EU temperature projections

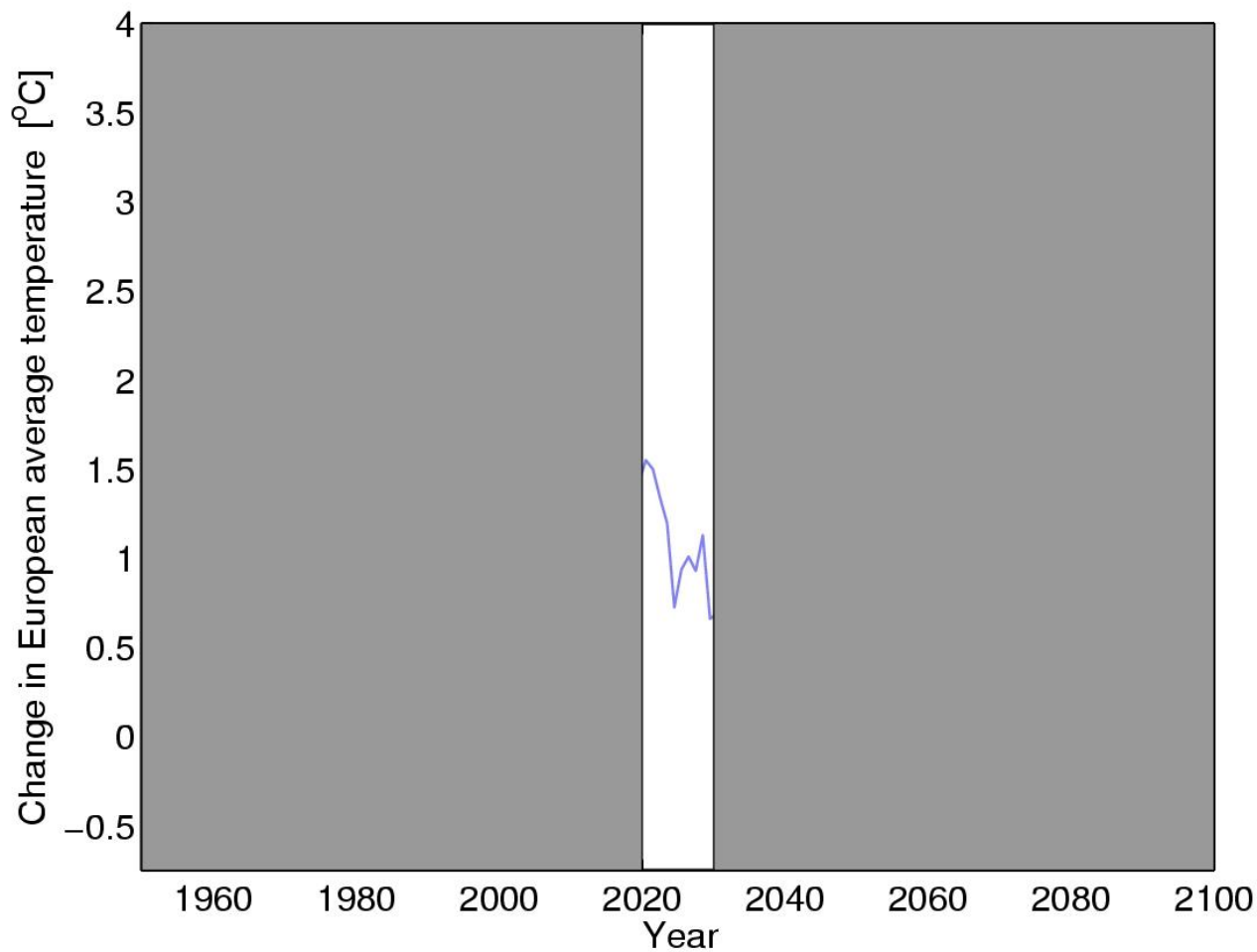




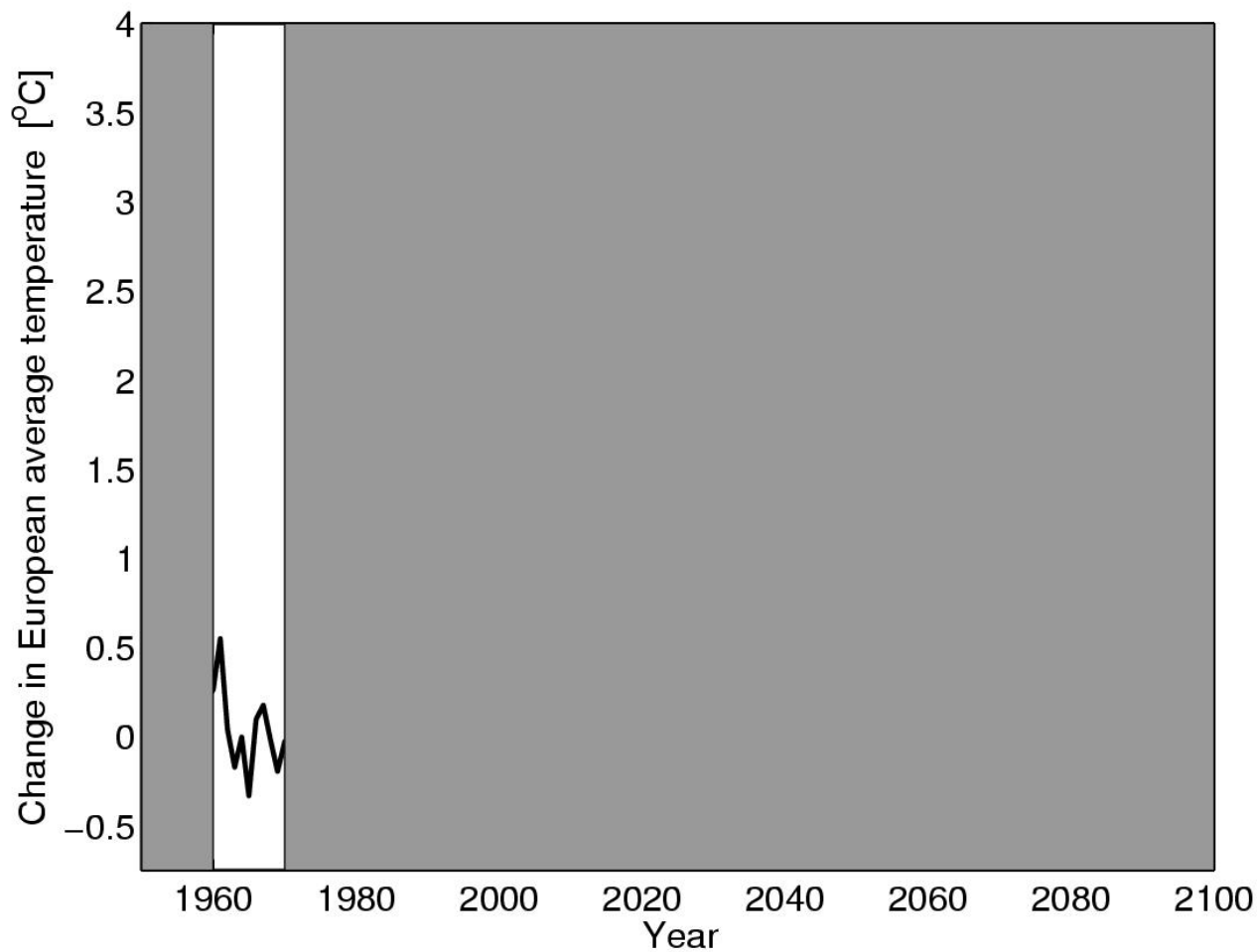
“Ensembles contain information”



“Ensembles contain information”

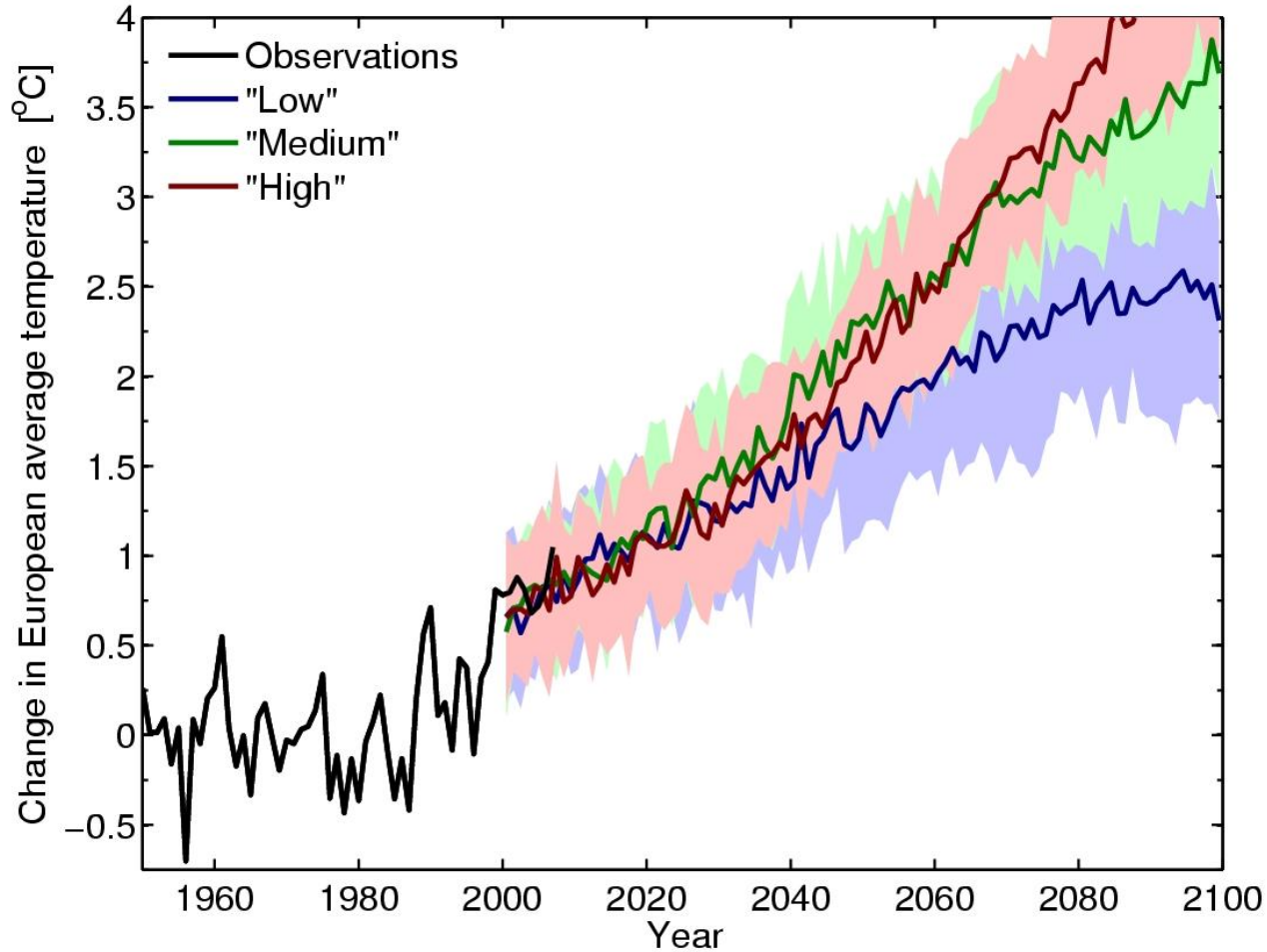


“Ensembles contain information”

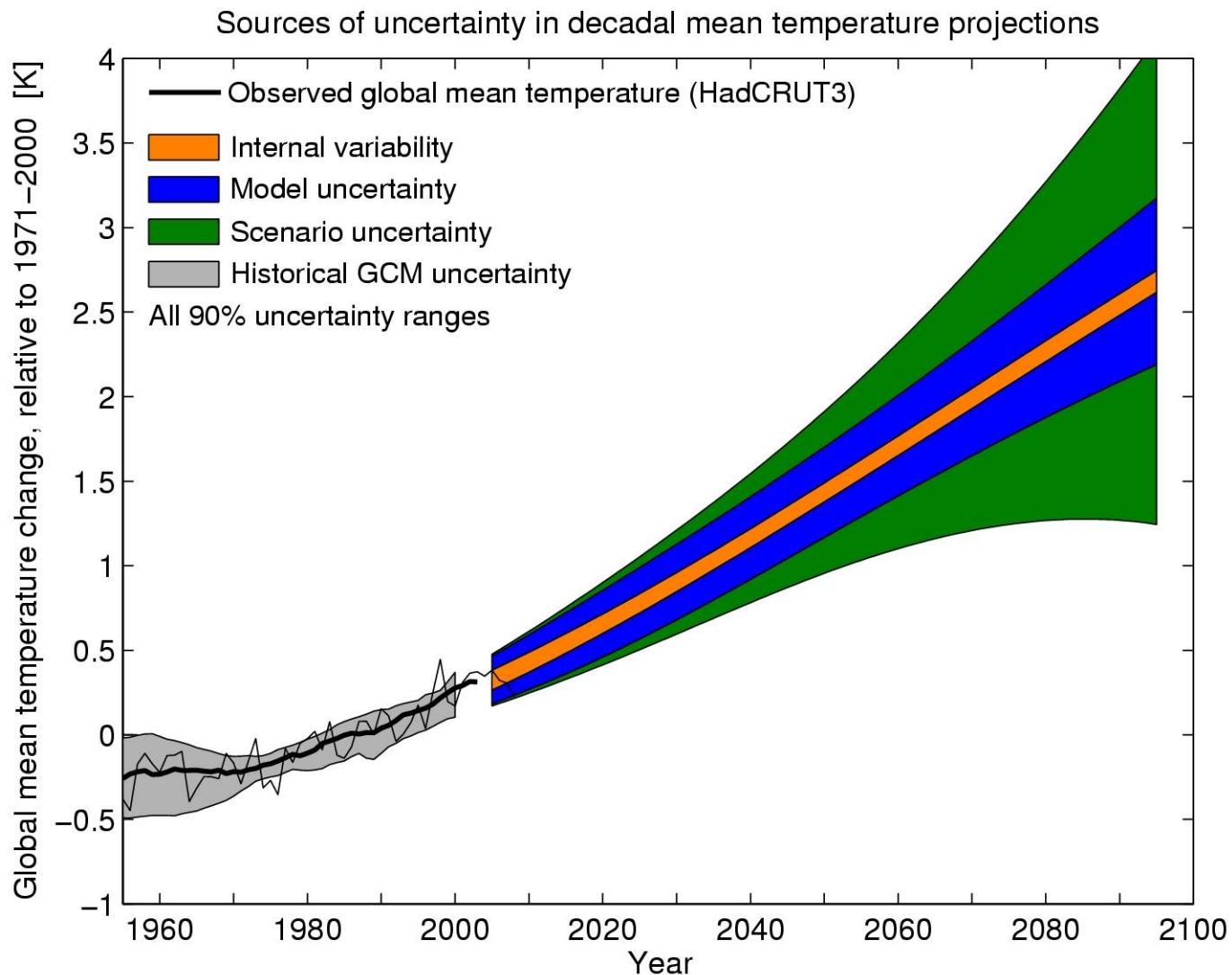


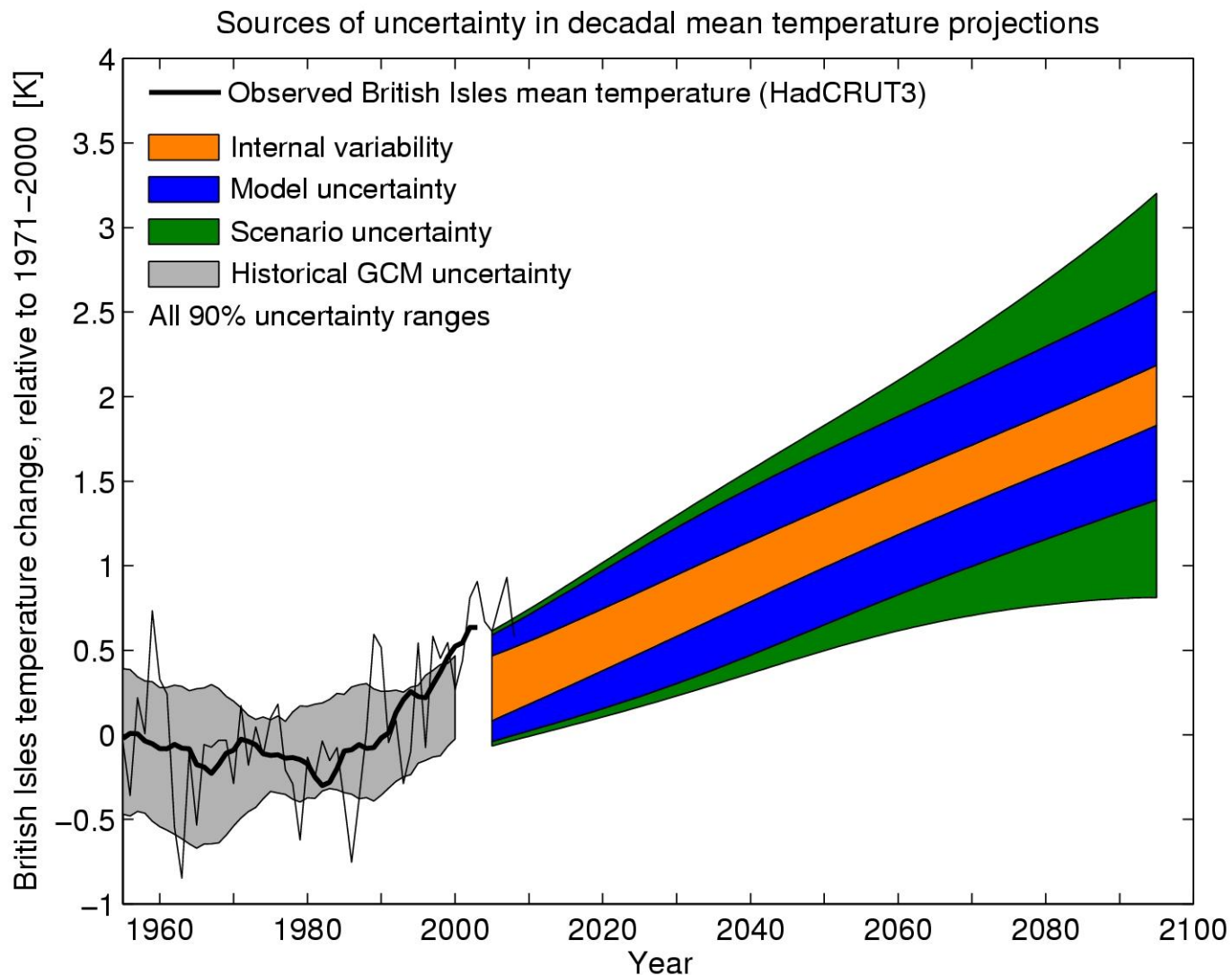
“Ensembles contain information”

EU temperature projections



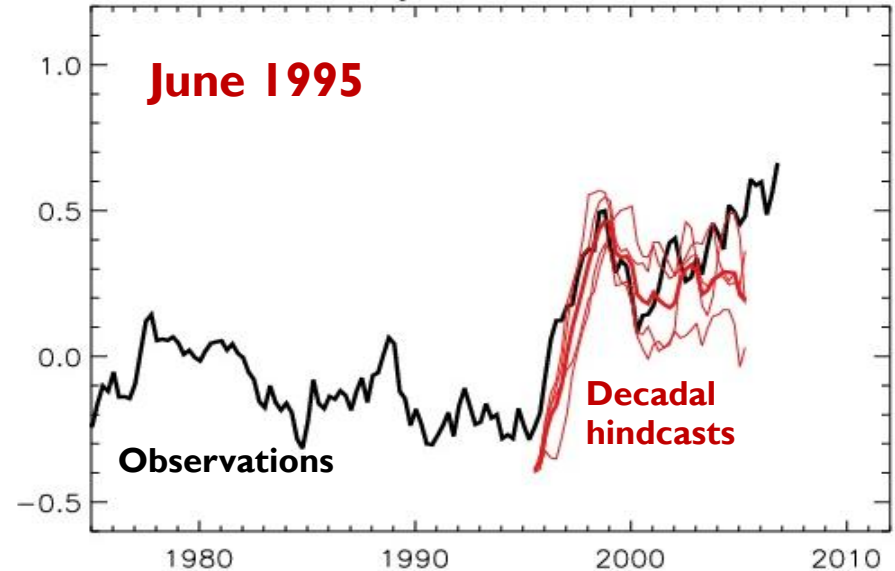
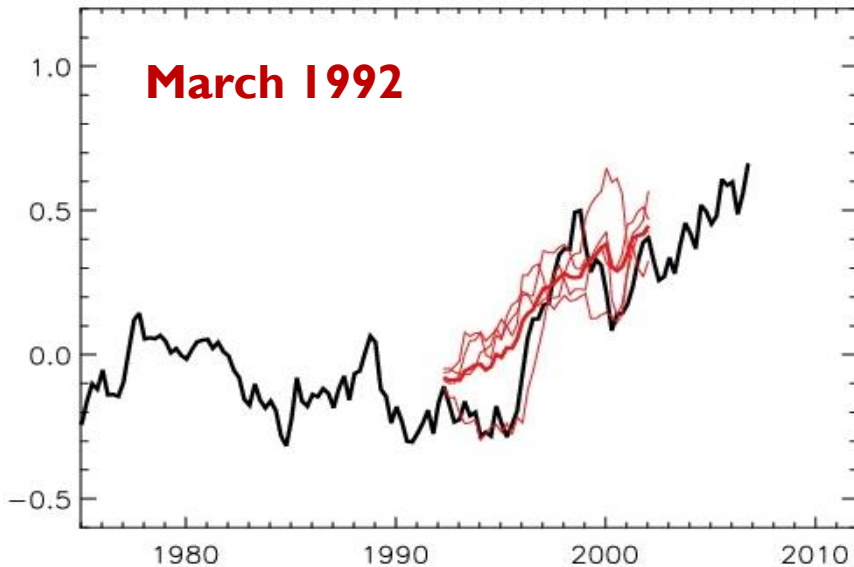
“Ensembles contain information”





From: Jon Robson, Rowan Sutton, Doug Smith

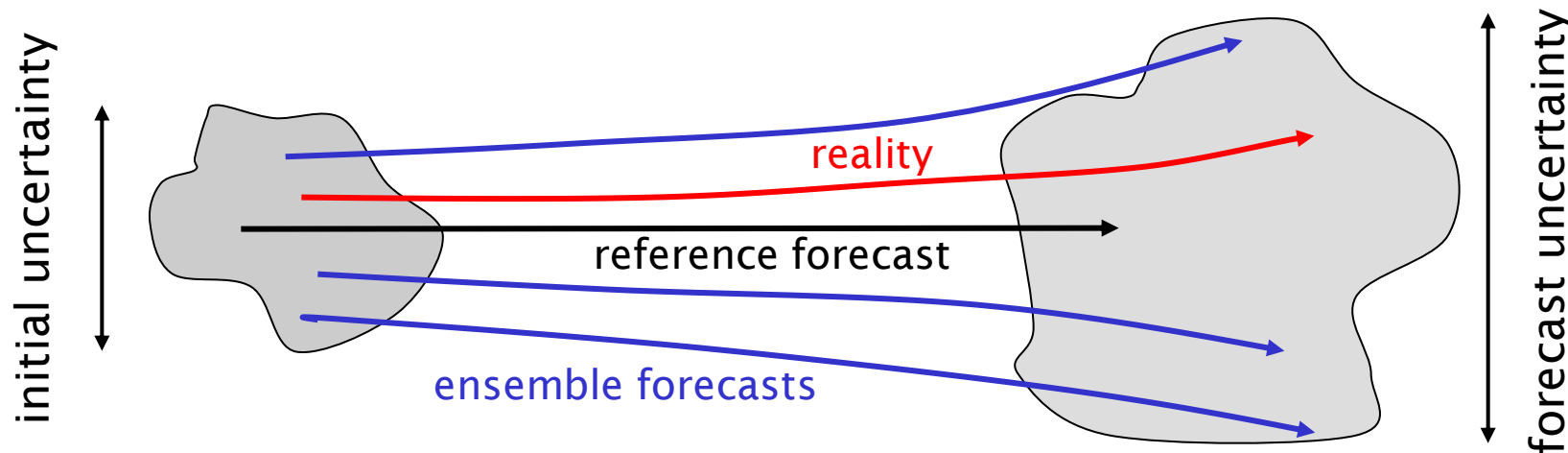
North Atlantic upper ocean heat content



Can potentially use decadal predictions to:

- validate (or not) GCMs on climate timescales,
- learn about model errors at a process level
- reduce uncertainty in predictions

- ⦿ Long ‘memory’ in the ocean provides potential predictability on decadal timescales
- ⦿ Decadal climate predictions are now being made (for IPCC AR5)
 - initialised from ocean state to try and predict both the response to radiative forcings *and* the internal variability component of climate
- ⦿ But, large uncertainties in our knowledge of current ocean state
 - need to efficiently sample this initial condition uncertainty
- ⦿ Methods developed for numerical weather prediction can be extended and adapted for decadal climate predictions
 - requires identifying perturbations relevant for climate timescales



Optimal perturbations for decadal predictions are:

- ⦿ perturbations which grow most rapidly
- ⦿ consistent with the observational uncertainties
- ⦿ average over weather 'noise'
- ⦿ useful for:
 - efficient perturbations in ensemble forecasts
 - identifying regions where *additional* observations would be most valuable to improve predictions

Using two different methods for the Atlantic Ocean in the HadCM3 GCM:

1. Linear Inverse Modelling (LIM) e.g. Penland & Sardeshmukh (1995)

(computationally inexpensive, initial condition independent)

2. Climatic Singular Vectors (CSVs) e.g. Kleeman et al. (2003)

(expensive, estimated for each initial condition separately)

Reduce dimensionality by representing ocean variability in HadCM3 control run (1 100 years) with leading 3d EOFs:

$$\text{GCM: } \frac{dy}{dt} = F(\mathbf{y})$$

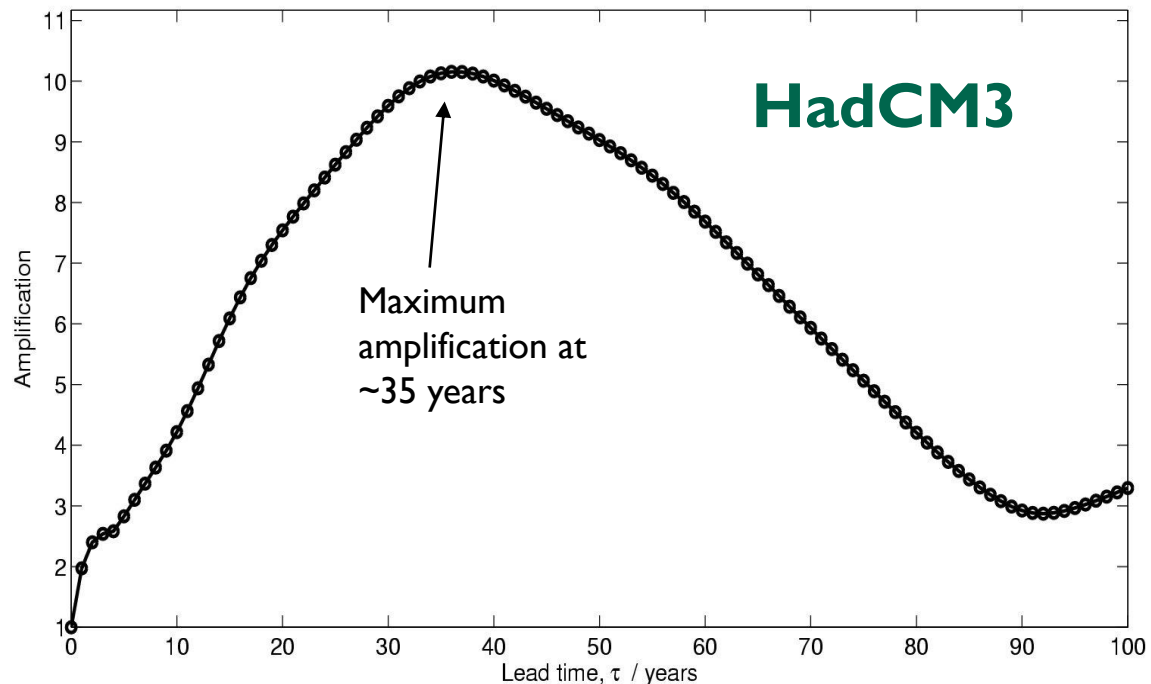
\mathbf{y} represents ocean data

$$\text{LIM: } \frac{d\mathbf{x}}{dt} = \mathbf{B}\mathbf{x} + \xi$$

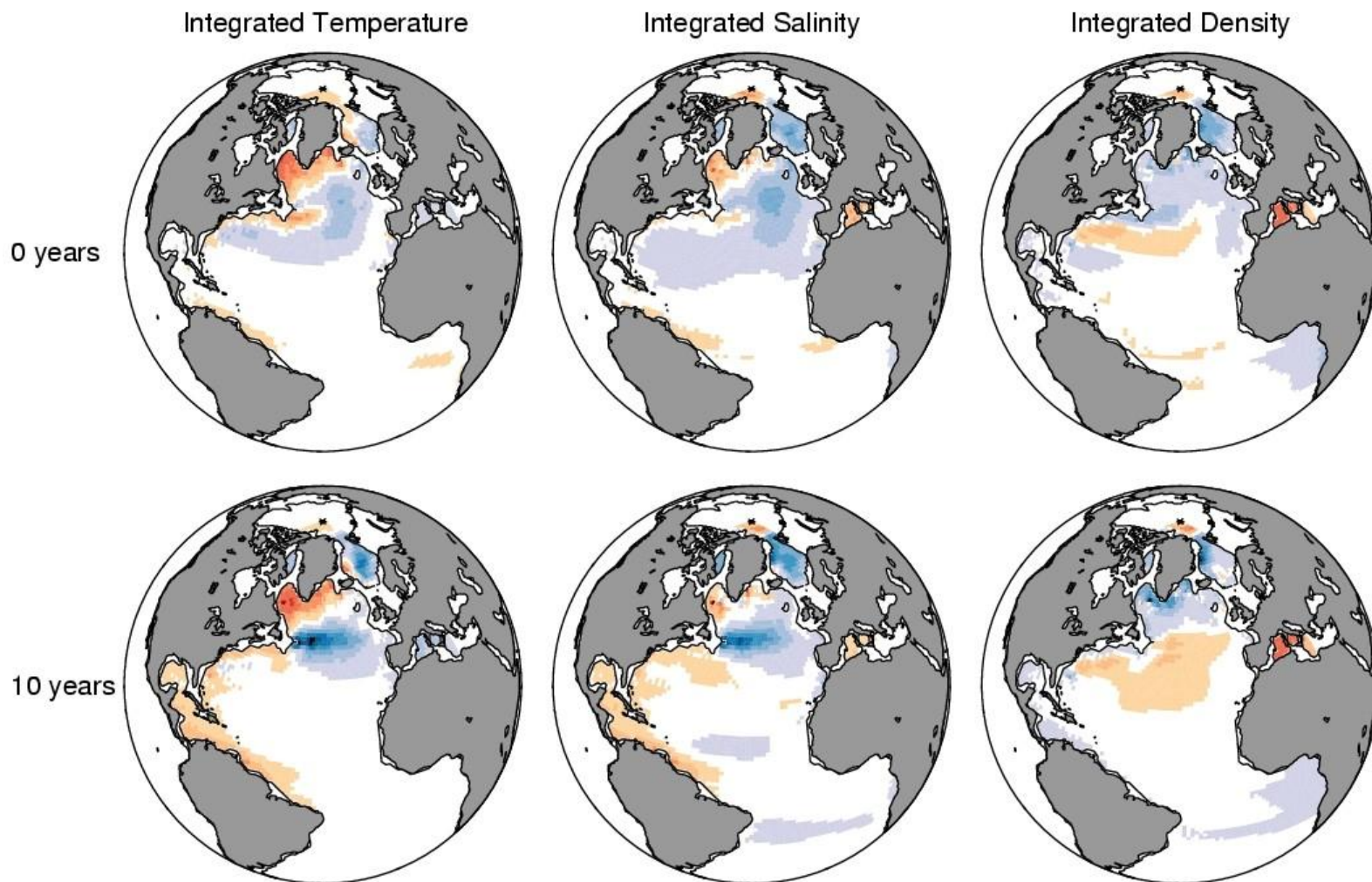
\mathbf{x} represents leading PCs

LIM forecast model:

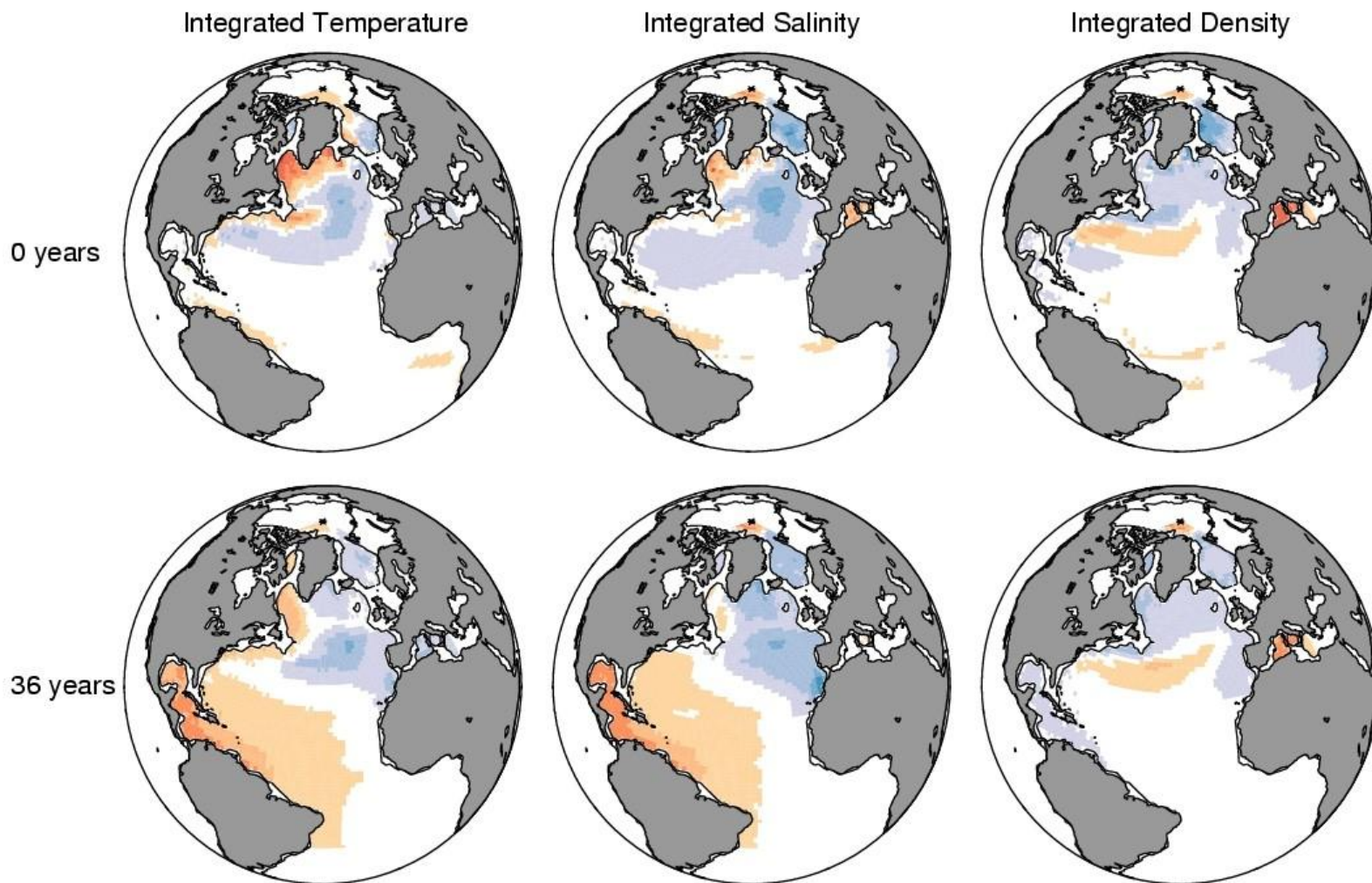
$$\mathbf{x}(t + \tau) = \mathbf{P}_\tau \mathbf{x}(t)$$



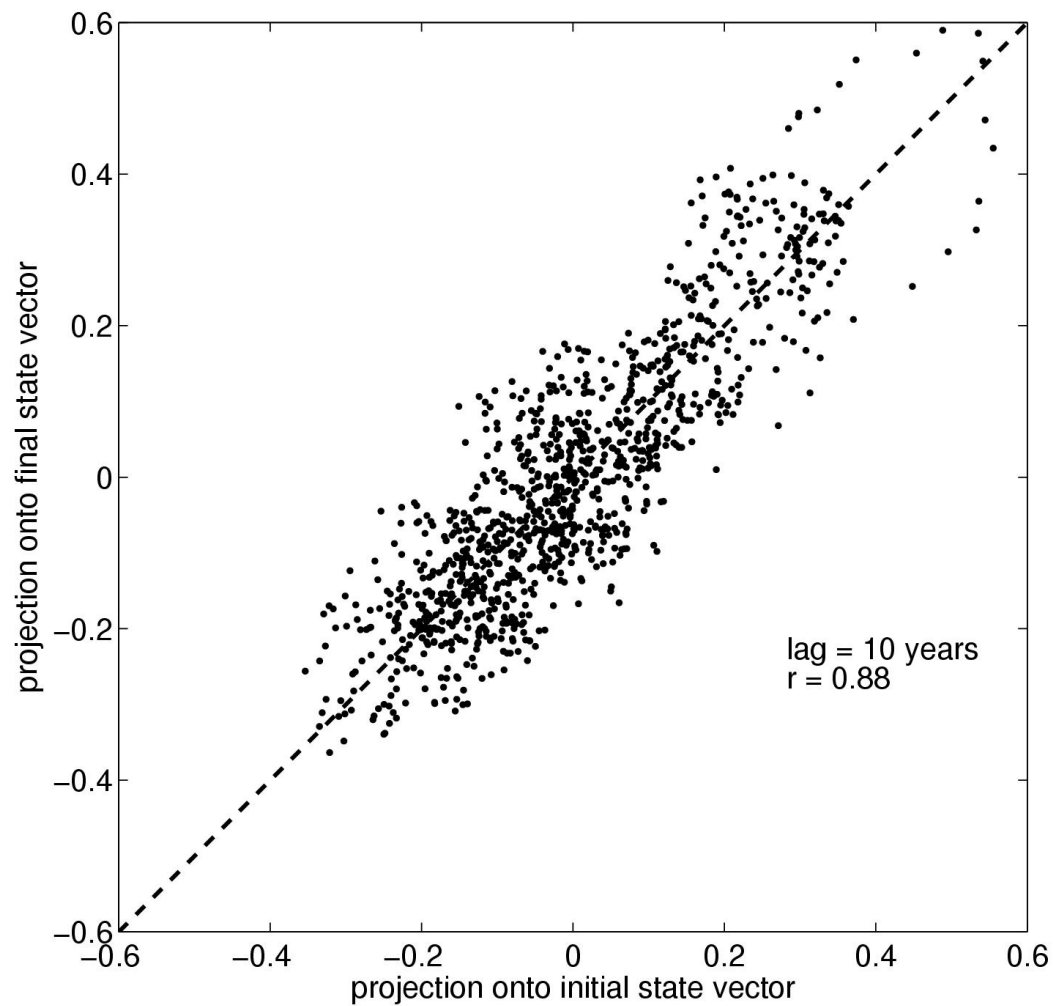
Ocean variables integrated to 1500m depth



Ocean variables integrated to 1500m depth



Does the linear model work?

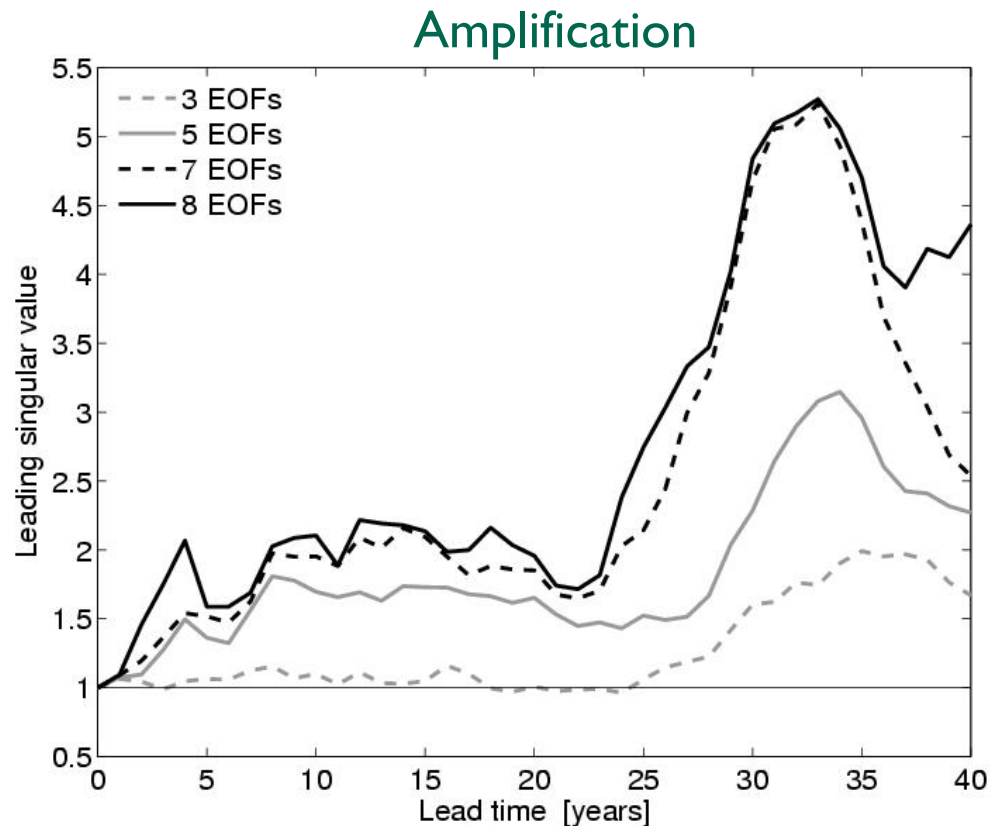


Climatic Singular Vectors (CSVs)

Build a tangent linear propagator matrix (**P**) using an
“ensemble of ensembles” run from a single initial condition

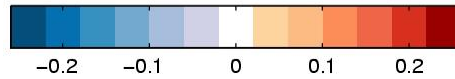
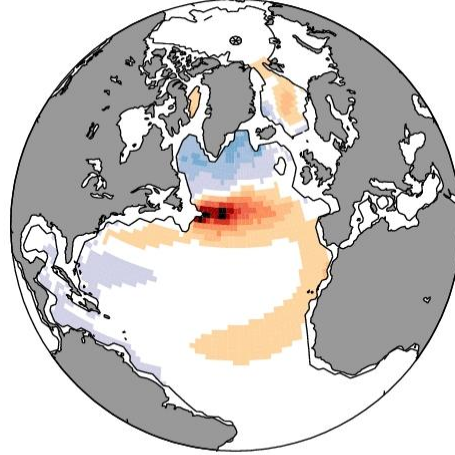
- control ensemble
- 8 EOF perturbed ensembles
- 16 members each
- run for 40 years
- further ensembles to test optimal perturbations

Total: ~7000 years with
HadCM3, or ~20 CPU-years

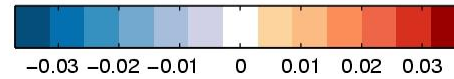
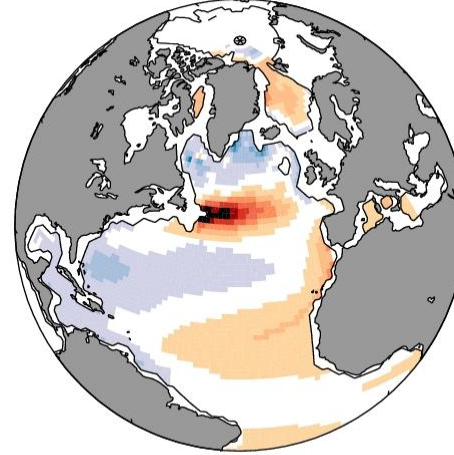


Optimal
perturbation

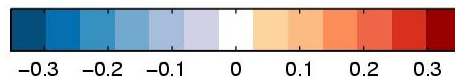
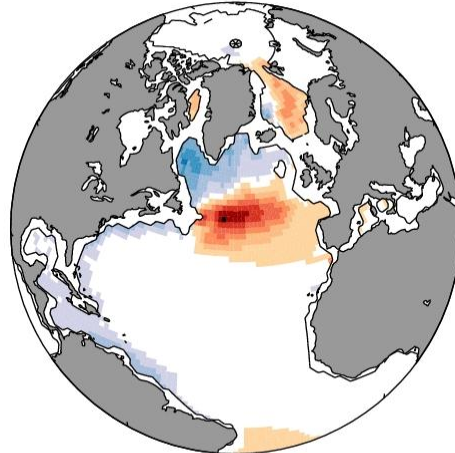
Integrated temperature [K]



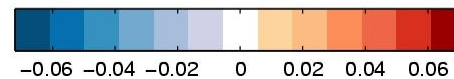
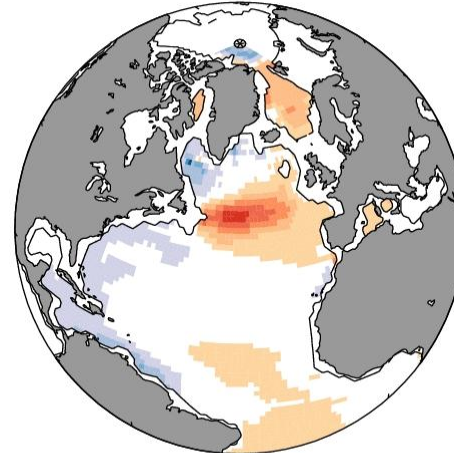
Integrated salinity [psu]



Integrated temperature [K]



Integrated salinity [psu]



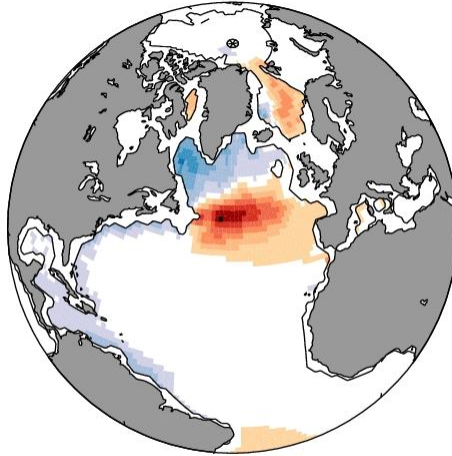
Linear model
prediction
for 10 years
later

Note changed
colour scales!

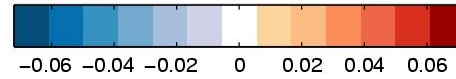
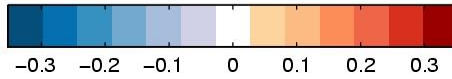
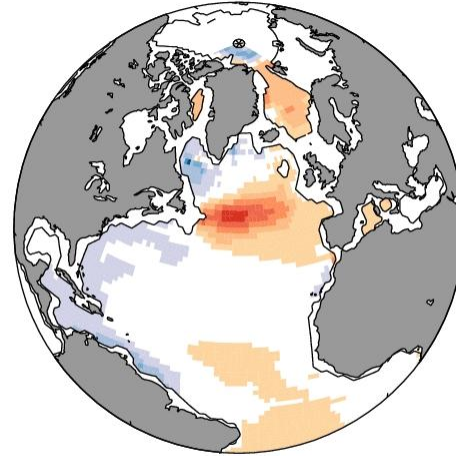
Does the CSV work?

Linear
model
prediction:

Integrated temperature [K]



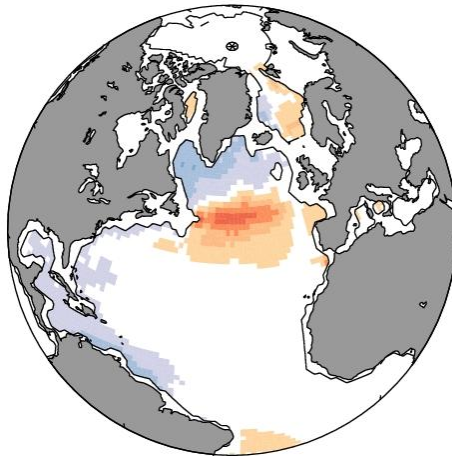
Integrated salinity [psu]



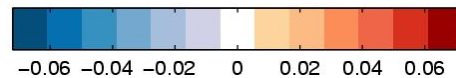
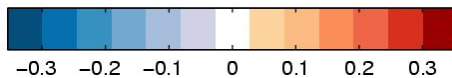
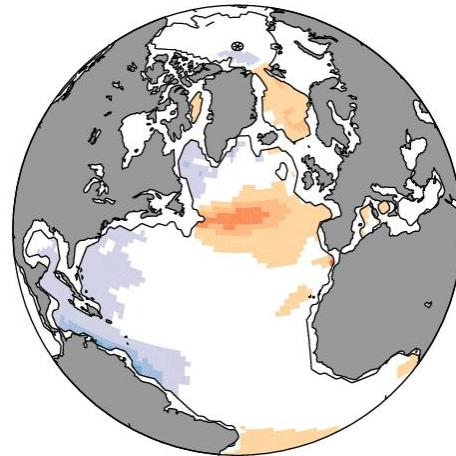
After 10
years

Actual:

Integrated temperature [K]



Integrated salinity [psu]



- ⦿ Decadal predictions can help test climate models at a process level
- ⦿ Demonstrated two methods for estimating optimal perturbations for decadal climate predictions
- ⦿ In HadCM3, both methods show significant amplification
 - maximum growth after ~35 years
 - largest growing perturbations located in far North Atlantic
 - other models show similar features (e.g. Tziperman et. al. 2009)
- ⦿ These approaches have great potential to guide development of both:
 - efficient decadal ensemble forecasting systems
 - optimal ocean observing systems