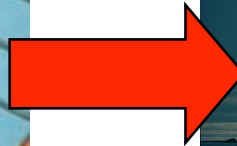


# Towards the Seamless Prediction of Weather and Climate

T.N.Palmer, ECMWF.



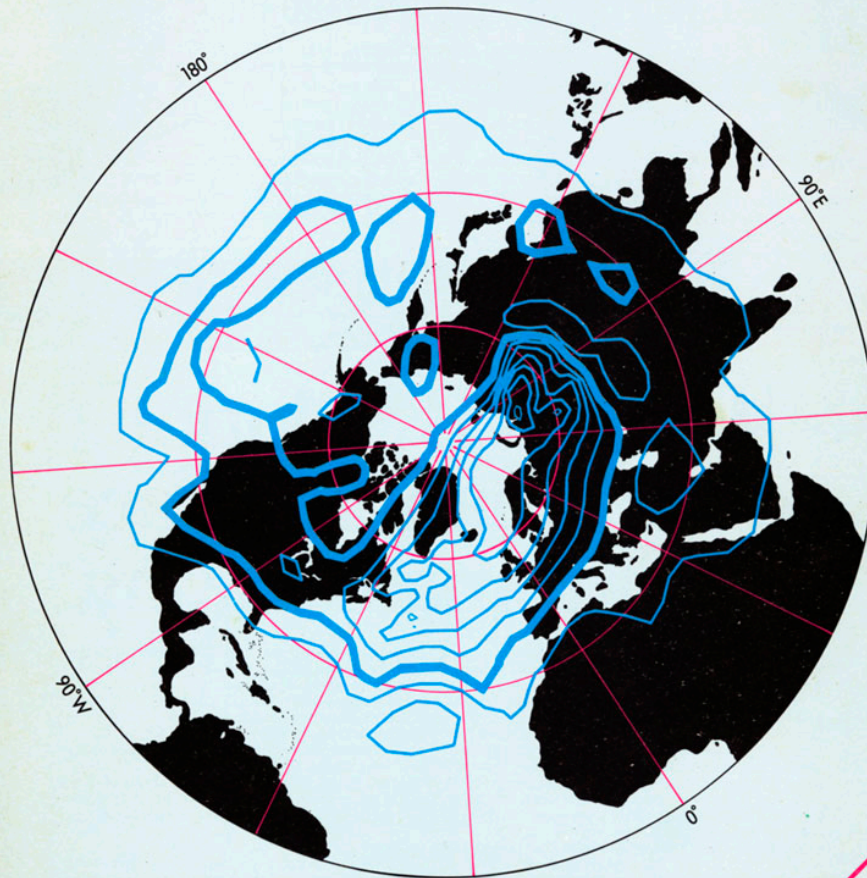
Bringing the insights and constraints of numerical weather prediction (NWP) into the climate-change arena.

With acknowledgements to :  
F.J.Doblas-Reyes, T.Jung, M.Rodwell and  
A.Weisheimer

# nature

INTERNATIONAL WEEKLY JOURNAL OF SCIENCE

Volume 305 No 5935 13-19 October 1983 £1.80 \$4.50



**PLANETARY WAVES  
IN THE STRATOSPHERE**

**BIOCHEMICAL  
INSTRUMENTS**  
product review

Climate as  
a Nonlinear  
Dynamical  
System

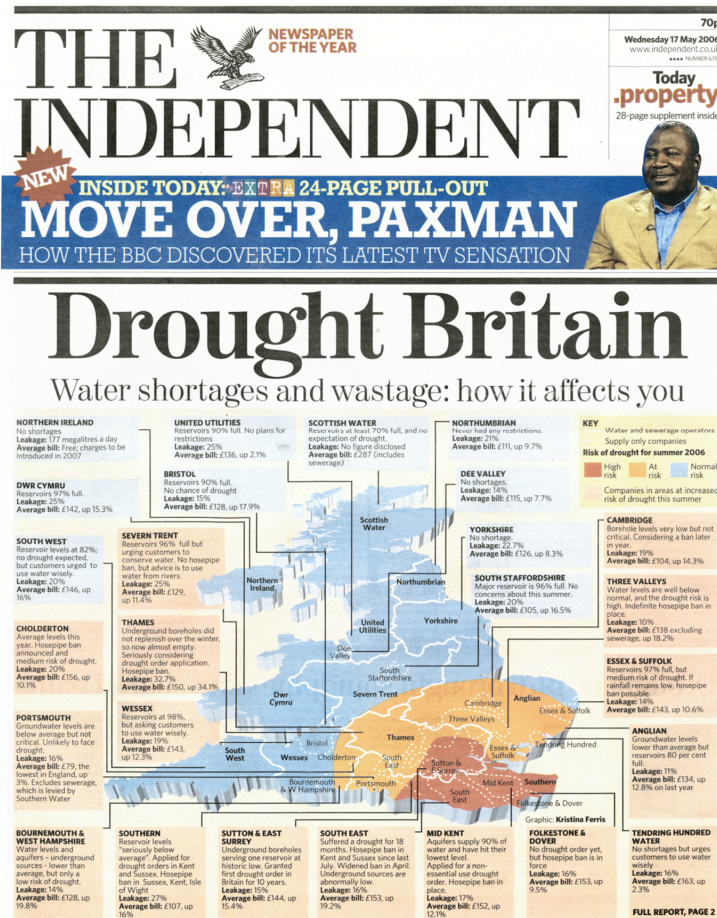
Climate Change is the  
“defining issue of our era\*”

\*Ban Ki Moon UN Secretary General

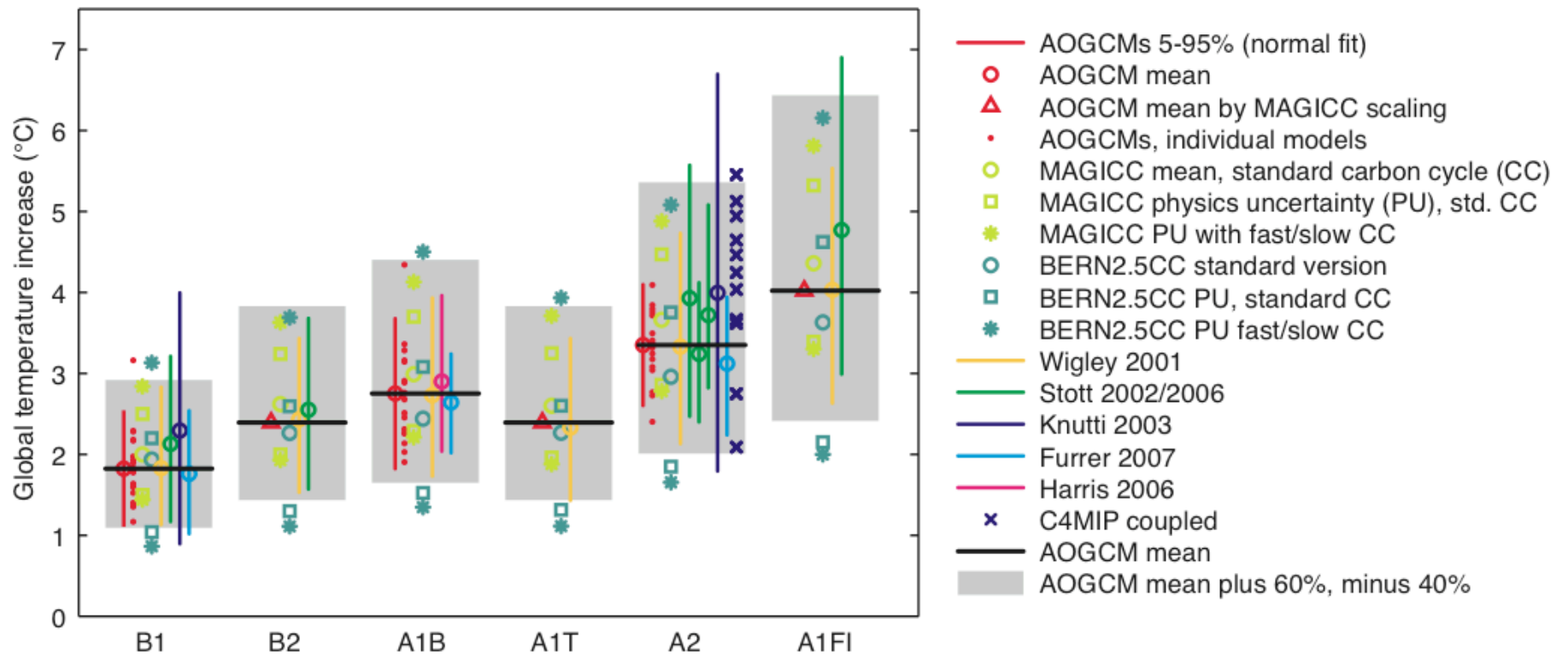
Climate change predictions provide key input to mitigation policies...



# ...and adaptation strategies



..... and yet projections of climate change, both globally and regionally, remain uncertain



# Spread of Global Warming Projections from IPCC AR4 WG1

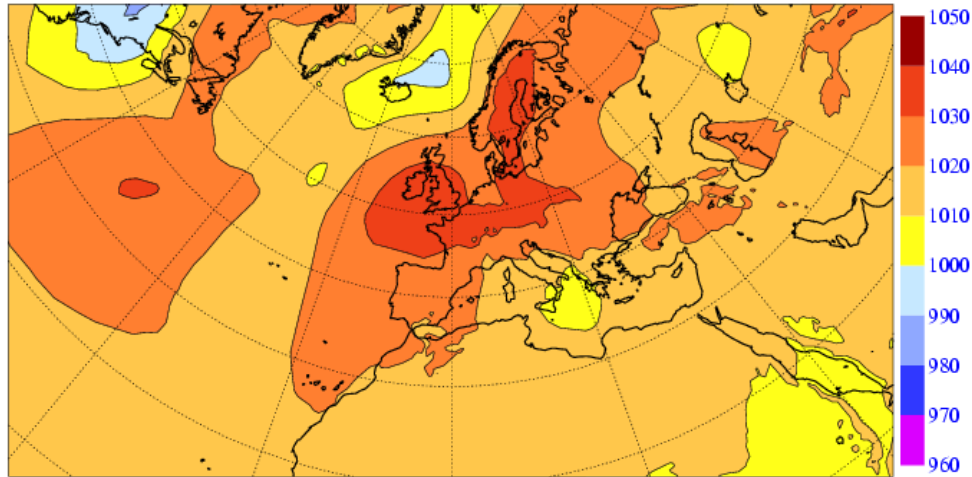
Figure 10.29



## Uncertainty of the First Kind

- Defined as the multi-model ensemble spread of climate-change projections. Has remained large since the first IPCC assessment report.
- There is a fundamental need to reduce uncertainty of the first kind. How?

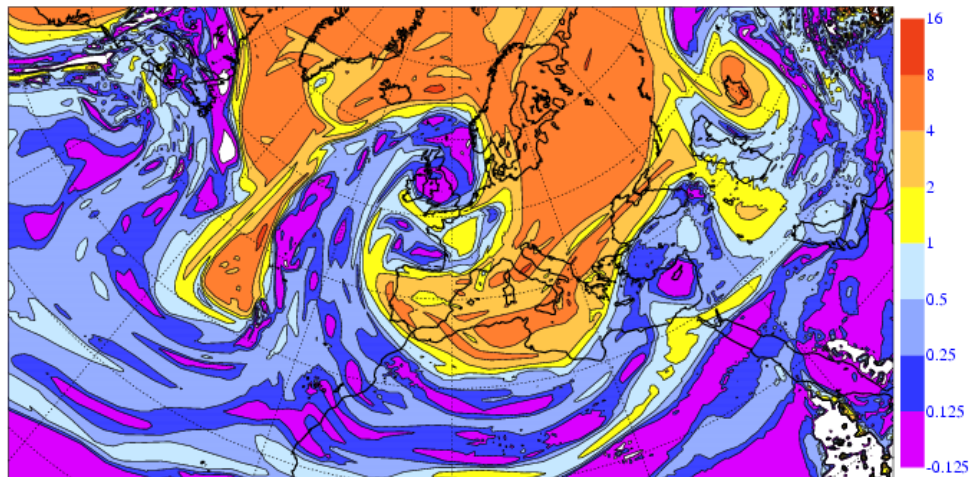
Surface Pressure



# 2005/2006 Drought

Blocking Anticyclone

Potential Vorticity on 315K

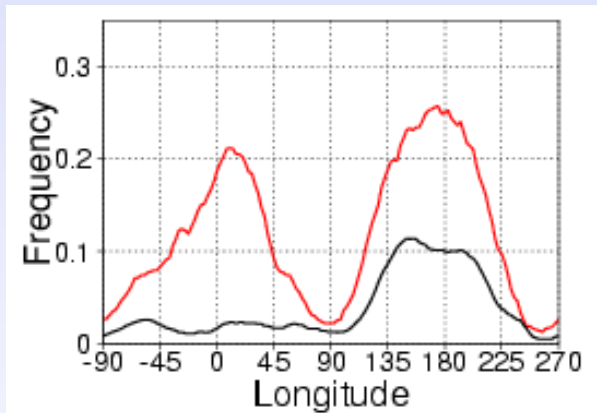


How much more frequently will blocking events occur, as a result of increased levels of CO<sub>2</sub>?

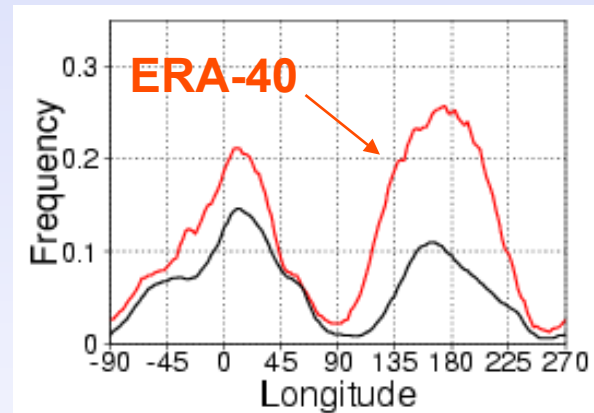
# Blocking frequency in climate models

Northern Hemisphere blocking frequency for DEMETER hindcasts  
November start, 1959-2001, 9-member ensembles  
January (third month)

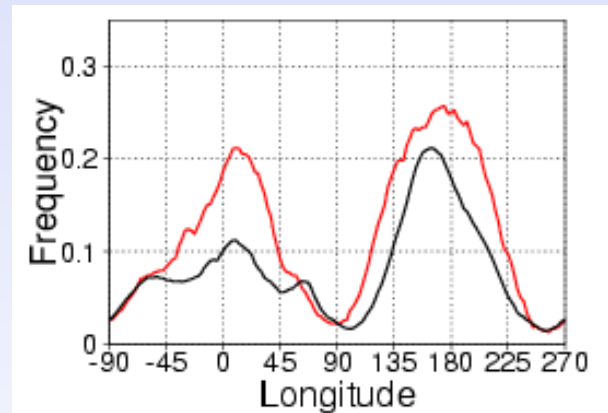
**CNRM**



**ECMWF**



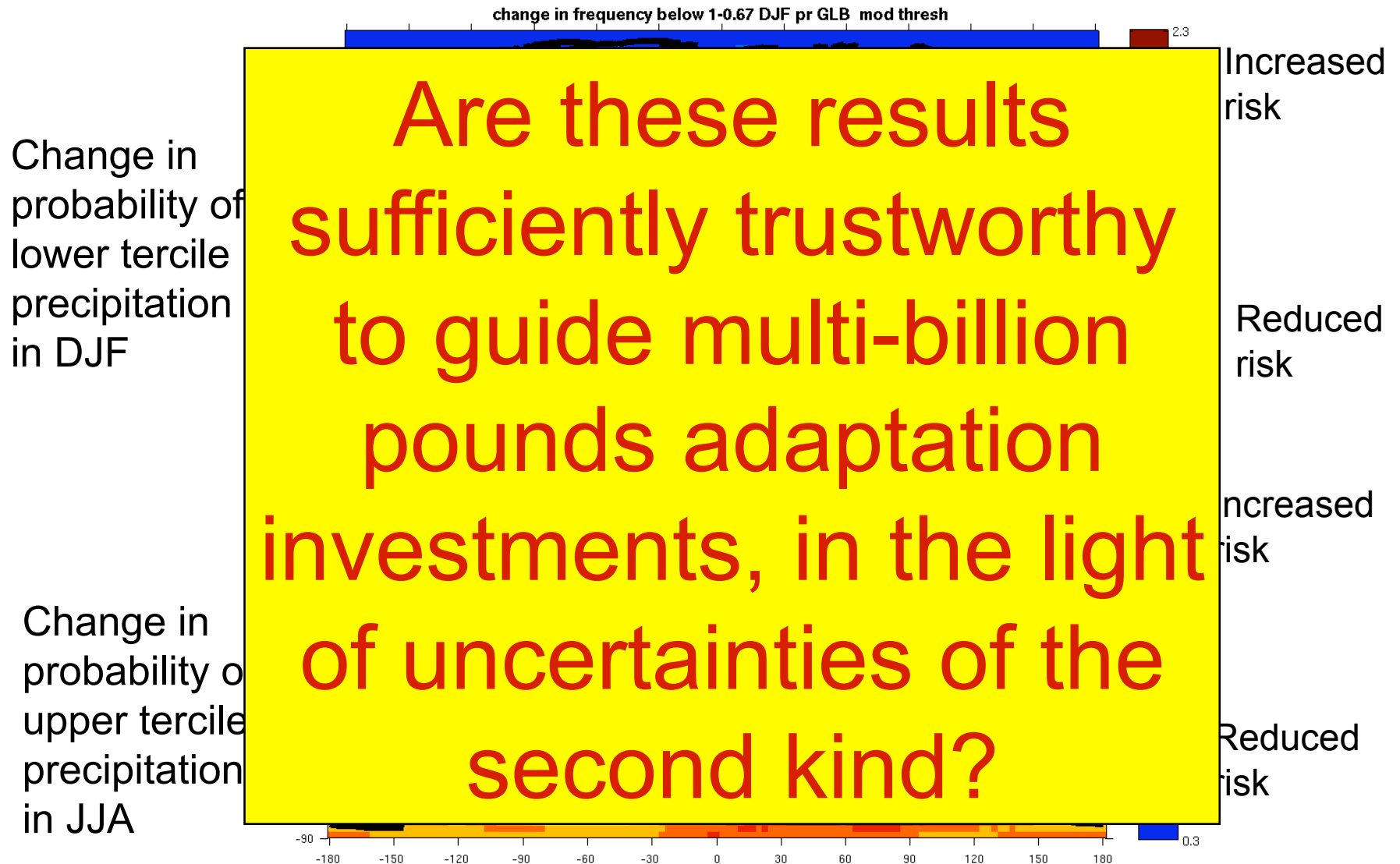
**Met Office**



## Uncertainty of the Second Kind

- Associated with biases common to all members of a multi-model ensemble of climate models, eg QBO, MJO, blocking, diurnal cycle...
- More “insidious” than uncertainty of the first kind – there is currently no agreed method to quantify, let alone reduce, the impact of uncertainties of the second kind, on climate change projections.

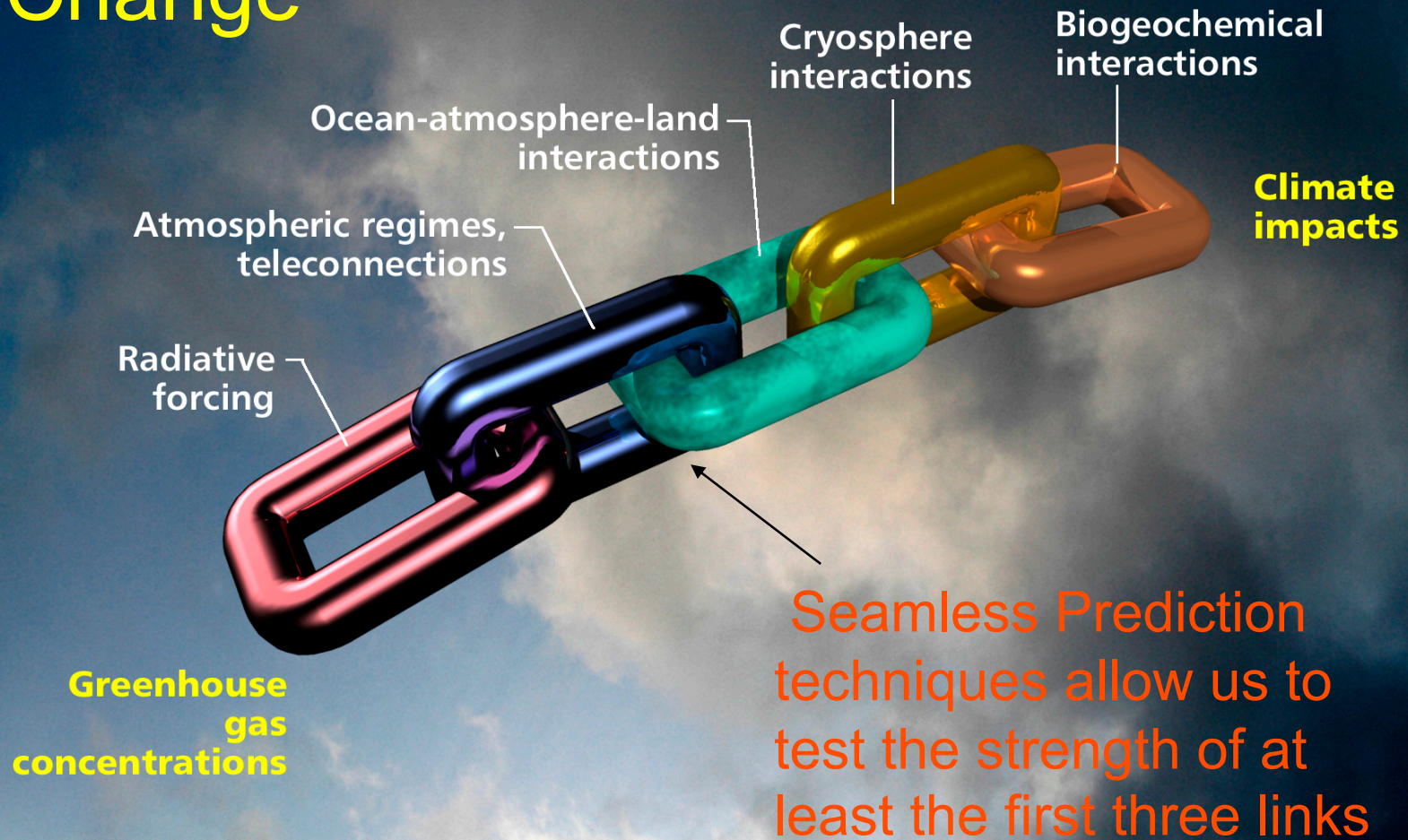
# Change in Probability of 20<sup>th</sup> Century Lower/Upper Tercile Seasonal Precipitation Under Anthropogenic Climate Change

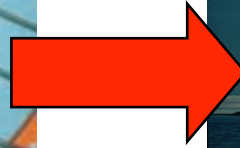


# Three Examples of the Potential Impact of Seamless Prediction Techniques

- Short-range NWP tendencies to reduce uncertainty of the first kind
- Seasonal forecast calibration techniques to quantify uncertainty of the second kind
- Stochastic parametrisation to reduce uncertainty of the second kind.

# A Nonlinear Perspective on Climate Change





# 1. Reducing Uncertainties of the First Kind Using 6hr NWP tendencies.

**T.N.Palmer and P.J.Webster**, 1994: Towards a unified approach to climate and weather prediction. Global Change. European Commission, EUR 15158en, 265-281.

**Rodwell, M.J. and T.N.Palmer**, 2007: Using numerical weather prediction to assess climate models. Q.J.R. Meteorol.Soc., 133, 129-146.



http://www.newscientist.com - Soaring global warming 'can't be r

File Edit View Go Bookmarks Tools Window Help

NewScientist.com

HOME | NEWS | EXPLORE BY SUBJECT | LAST WORD | SUBSCRIBE | SEARCH | ARCHIVE  
| RSS | JOBS

### Soaring global warming 'can't be ruled out'

19:03 26 January 2005  
NewScientist.com news service  
Jenny Hogan

The Earth may be much more sensitive to global warming than previously thought, according to the first results from a massive distributed-computing project.

The project tested thousands of climate models and found that some produced a world that warmed by a huge 11.5°C when atmospheric carbon dioxide concentrations reached the levels expected to be seen later this century.

This extreme result is surprising because it lies far outside the 1.4°C to 4.5°C range predicted by the Intergovernmental Panel on Climate Change (IPCC) for the same CO<sub>2</sub>-level increase - a doubling of CO<sub>2</sub> concentration from pre-industrial times. But it is possible the IPCC range was wrong because its estimate is based on just a handful of different computer models.



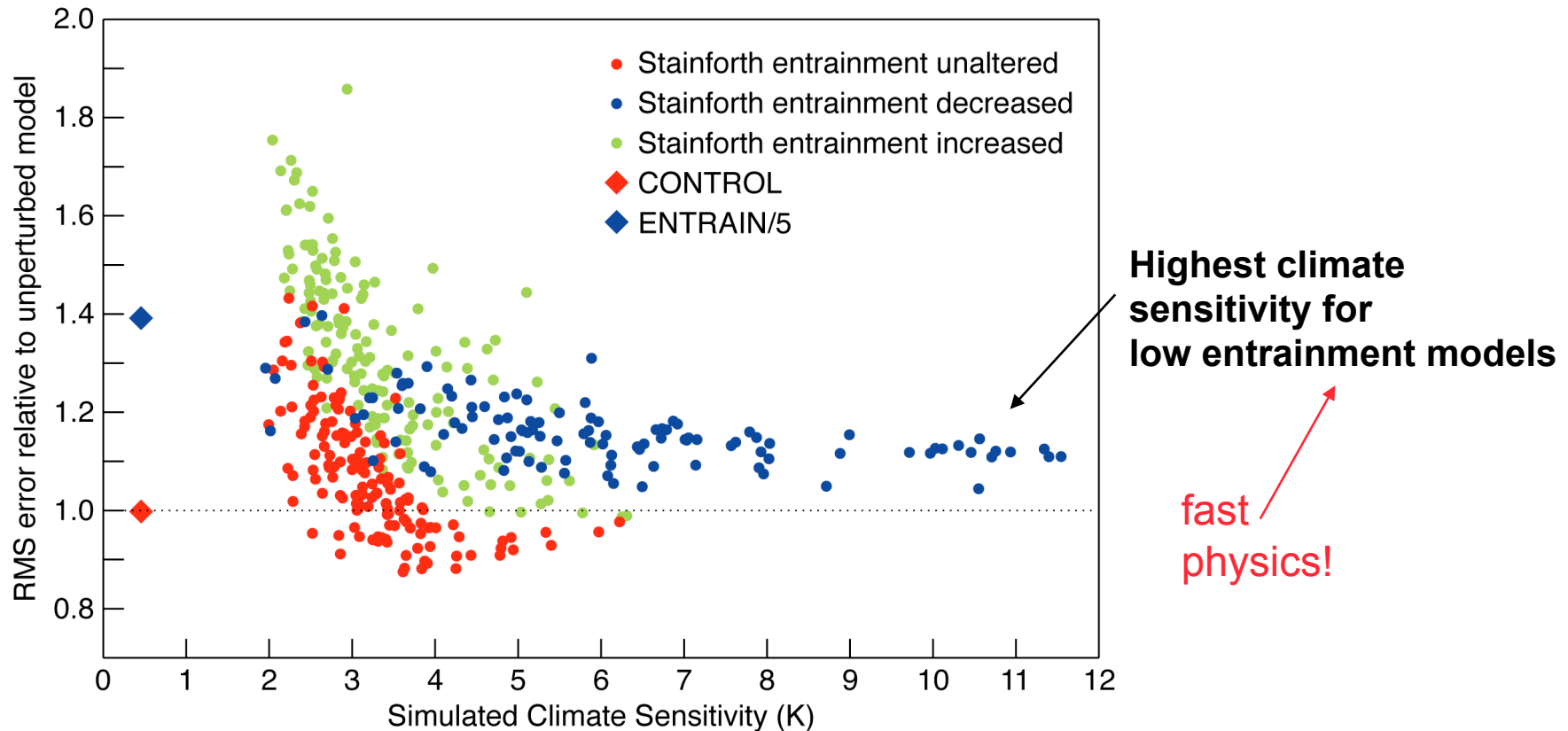
[Enlarge image](#)

The climate modelling software divides the Earth's surface into boxes hundreds of kilometres square (image: Climateprediction.net)

Click to Print

“There are no obvious problems with the high temperature models, Stainforth says.... The uncertainty at the upper end has exploded, says team-member Myles Allen.”

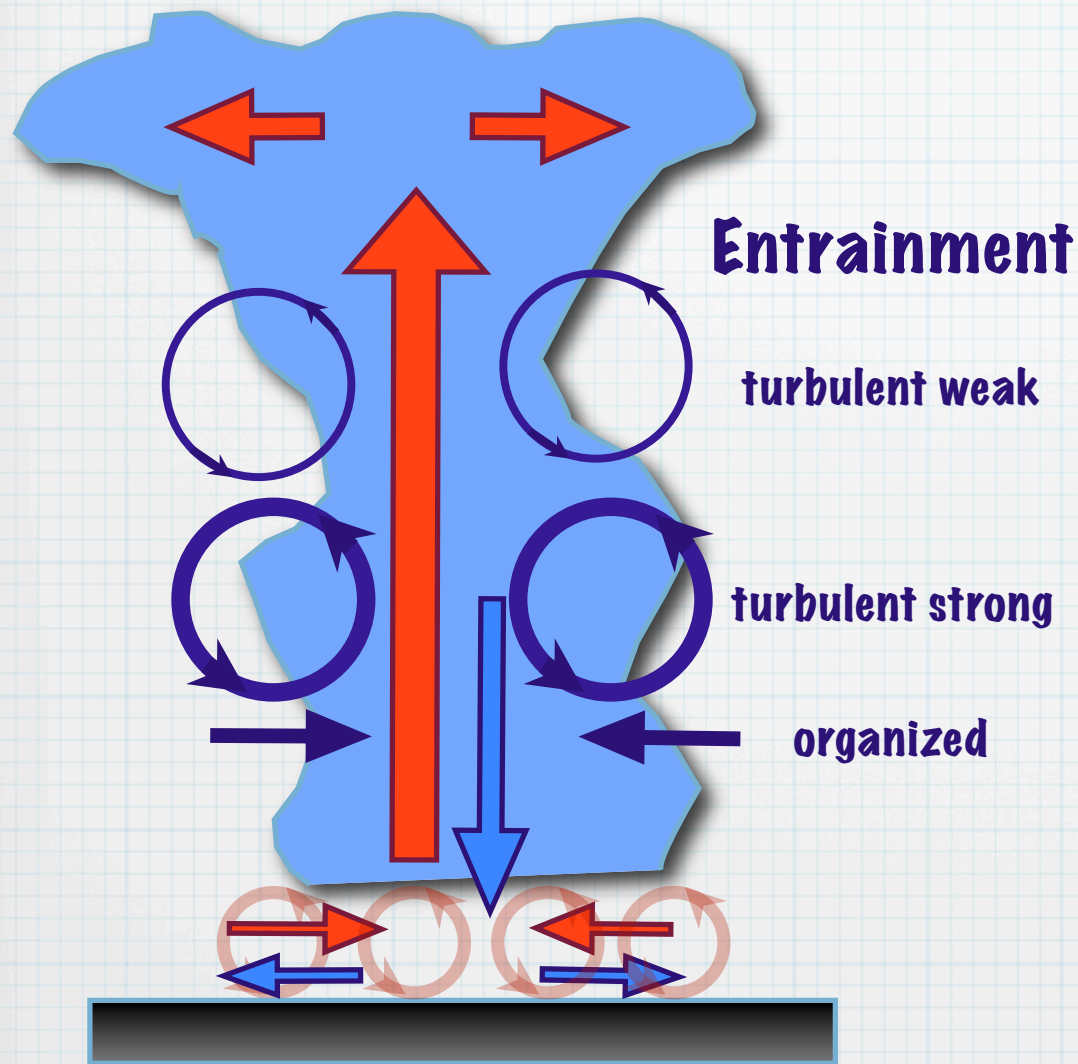
## Climate: Error vs Sensitivity



**Circles:** AGCM + Mixed-Layer model results from Stainforth et al. (2005) show combined RMSE of 8 year mean, annual mean  $T_{2m}$ , SLP, precipitation and ocean-atmosphere sensible+latent heat fluxes (equally weighted and normalised by the control).

**Diamonds:** AGCM results from Rodwell & Palmer (2006) show RMSE from 39 year mean, annual mean  $T_{850}$ , SLP and precipitation (equally weighted and normalised by the control).

# One key parameter in a convection parametrisation is the entrainment-rate parameter

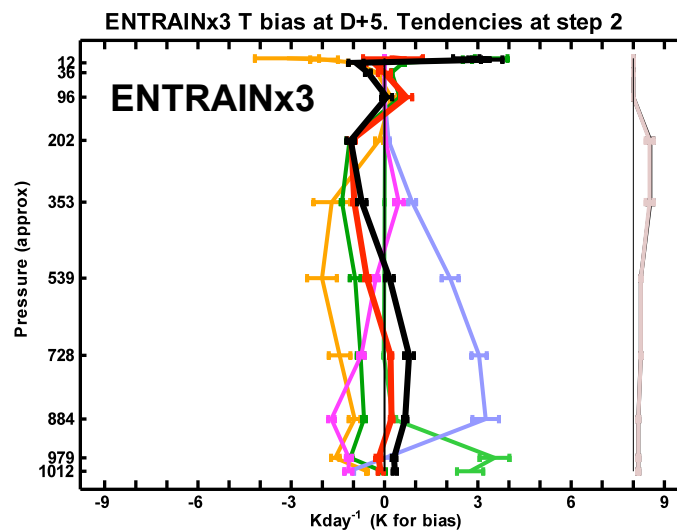
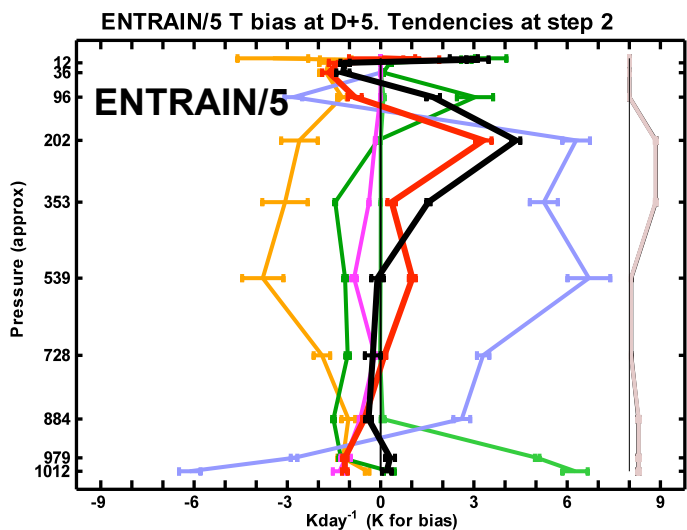
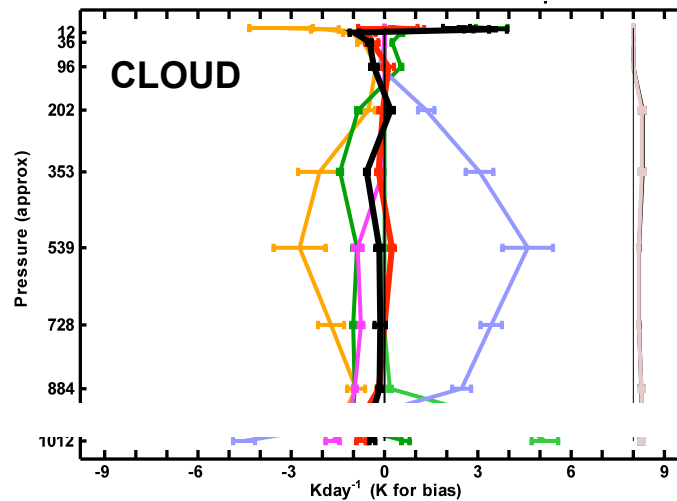
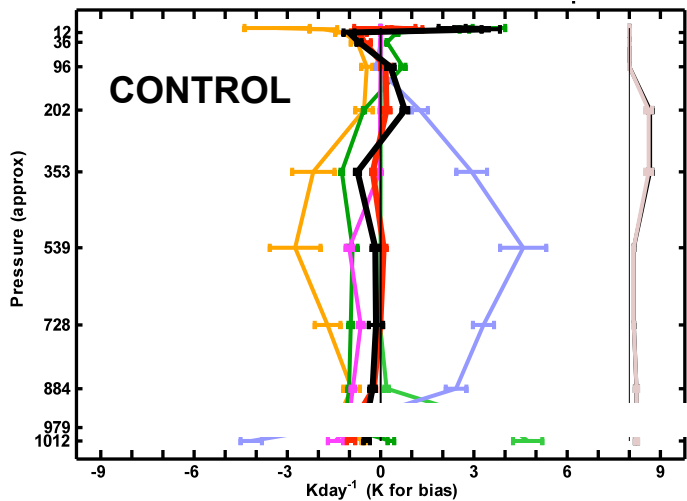
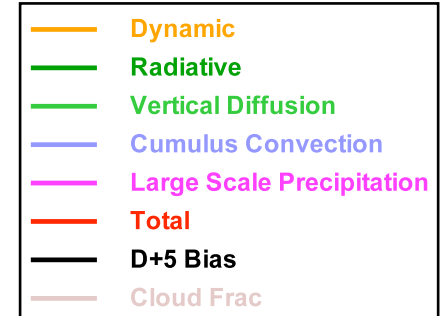


## Entrainment

- mixes environmental air into convective clouds
- is caused by **turbulence** and/or **organized inflow**
- thereby **reduces the difference of cloud to environment**, which is the fuel the cloud thrives on
- strength of its effect depends on entrainment rate** (model parameter) and **difference in properties of cloud and environment**
- high entrainment rate** and/or very **dry environment** -> **shallow clouds**
- low entrainment rate** and/or very **moist environment** -> **deep clouds**

# January 2005 Initial T Tendencies

Caveat: Not same model as Stainforth et al.



ENTRAIN/5 and ENTRAINx3 are out of balance

# Advantages of short-range tendency over conventional methods for assessing climate models

- Climate variability is dominated by a few EOFs (eg NAO, PNA, ENSO ...etc)
- Hence although individual parametrisations represent specific physical processes, their impact on climate can be degenerate, eg different parametrisations having similar responses
- Leads to classic problem of “compensating errors”
- Hence very difficult to assess what is the “best” set of parametrisations, eg by tuning to 20<sup>th</sup> Century climate.

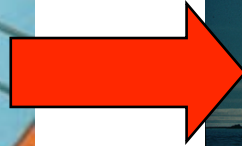
## Diagnosing Climate Error by Short-Range NWP Tendency

$$\dot{T} = \dot{T}_{dynamics} + \dot{T}_{long-wave\ radiation} + \dot{T}_{short-wave\ radiation} + \dot{T}_{convection} + \dots$$

Because they represent different physical processes, initial tendencies are approximately orthogonal, ie

$$\|T\|^2 \approx \|T_{dynamics}\|^2 + \|T_{long-wave\ radiation}\|^2 + \|T_{short-wave\ radiation}\|^2 + \|T_{convection}\|^2 + \dots$$

Hence reducing error in the norm of any one tendency will reduce error in the total tendency. Less possibility of compensating error.

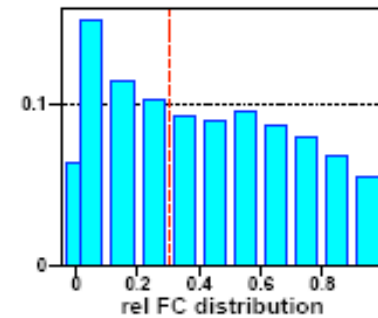
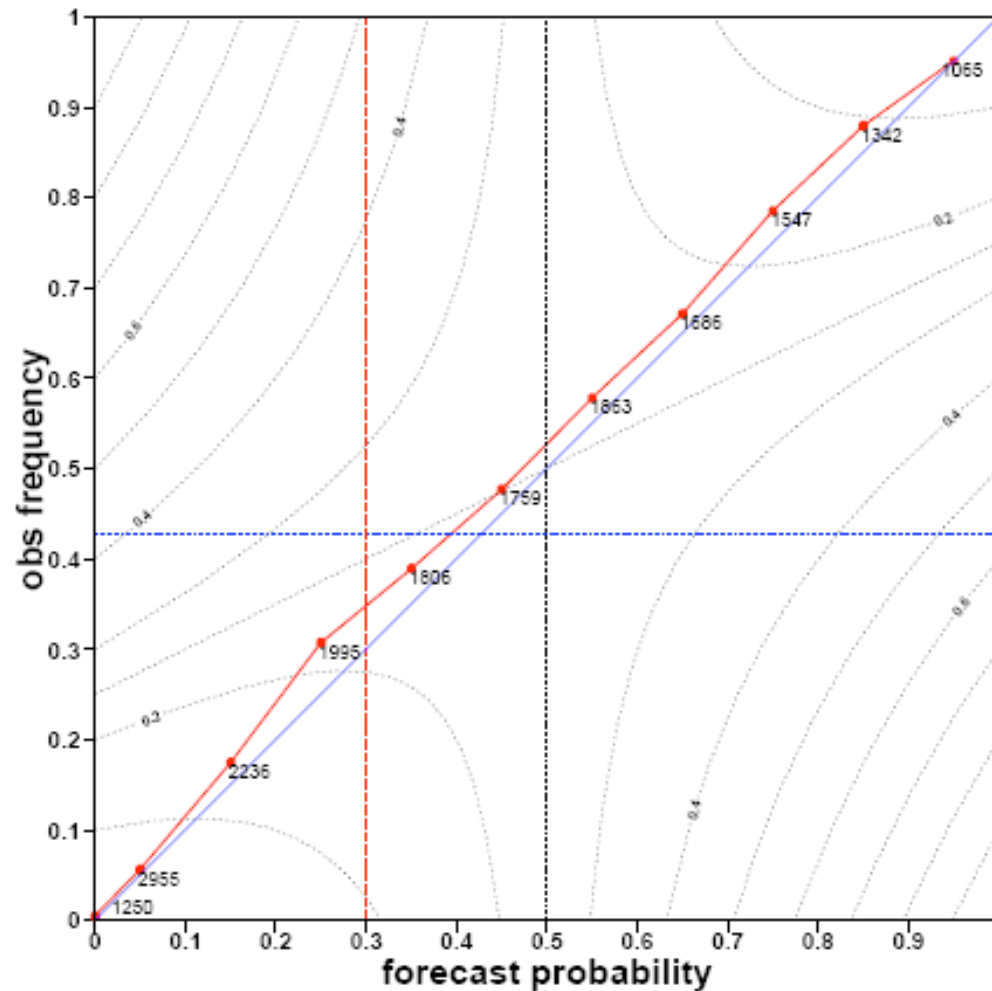


## 2. Quantifying Uncertainties of the Second Kind Using Seasonal Forecast Reliability Diagrams

Palmer, T.N., F.J. Doblas-Reyes, A. Weisheimer and M.J. Rodwell.  
2008: Towards Seamless Prediction: Calibration of Climate-Change  
Projections Using Seasonal Forecasts. BAMS, 89, 459-470,

# EPS Day 6 Precip Europe March-May 2008

Mar08-May08 t + 144 Europe an 24h-precip gt 1 mm  
BrSc = 0.157 LCB rSkSc= 0.40 Uncertainty= 0.245



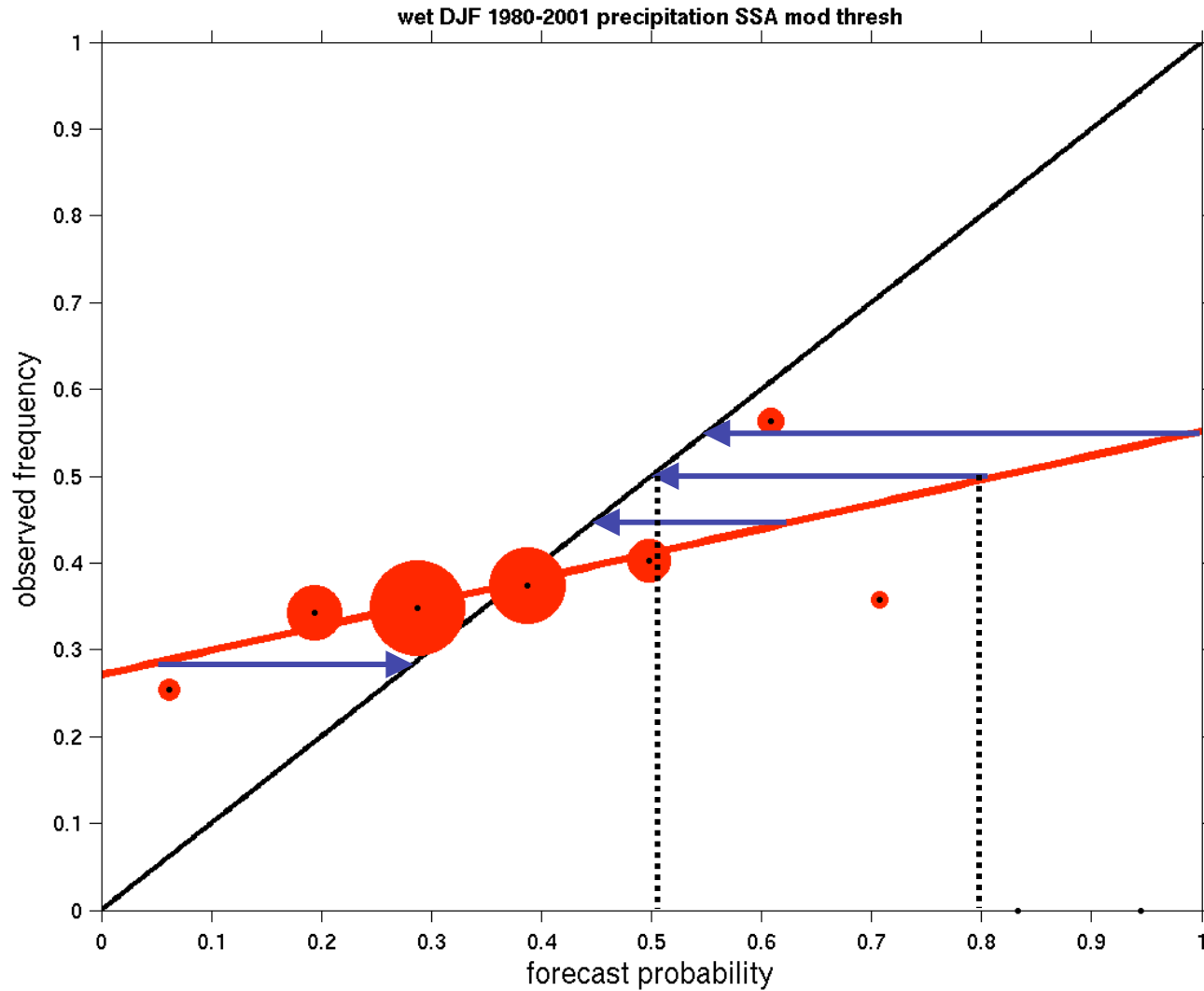
B(S)S\_RSL= 0.088( 0.36)  
B(S)S\_REL= 0.001( 1.00)

..... sample clim

----- clim 84-93



# Calibrating Probabilistic Forecasts

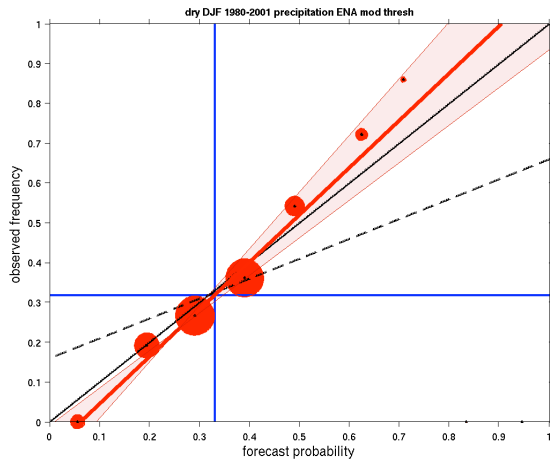


In regions where a multi-model seasonal forecasts are unreliable (due to uncertainties of the second kind) then the climate change signal may similarly be unreliable.

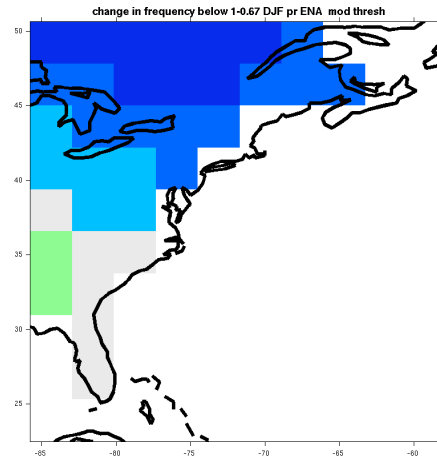
Use seasonal forecasts to calibrate climate change projections of precipitation?

# DEMETER Multi-Model Seasonal Predictions

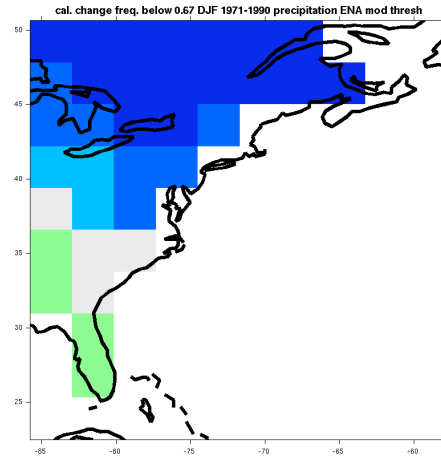
## Eastern North America dry DJF



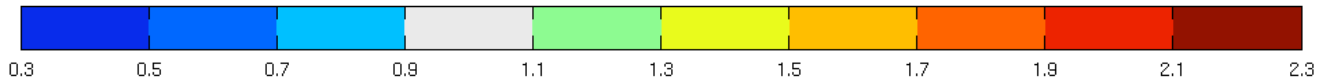
DEMETER  
Reliability



AR4 Uncalibrated

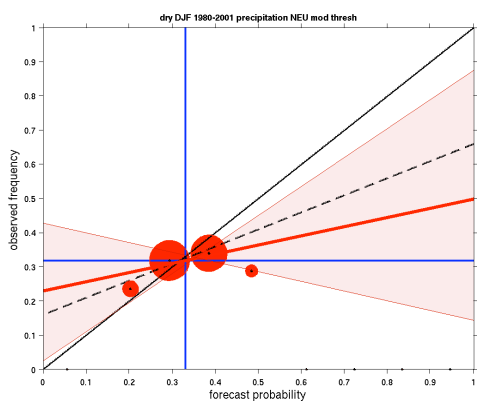
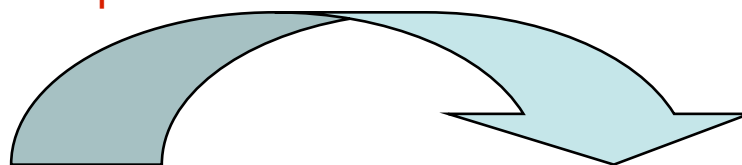


AR4 Calibrated

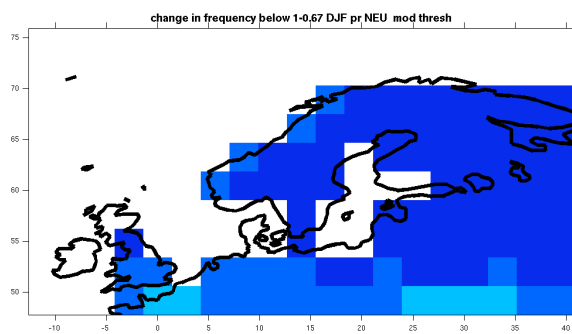


# Northern Europe dry DJF

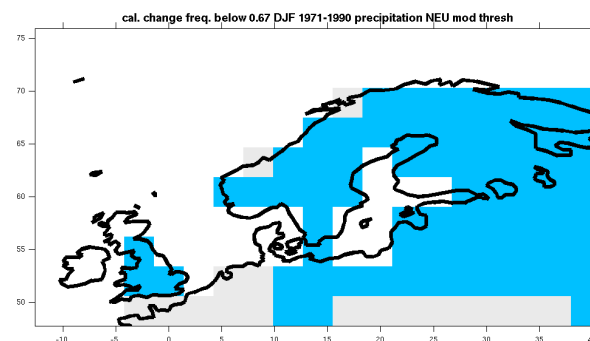
Reliability of seasonal forecasts is poor, hence discount the strong AR4 probabilities



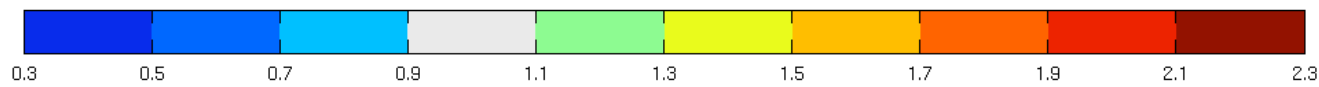
DEMETER  
Reliability



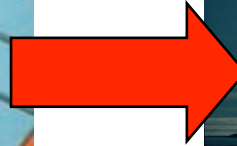
AR4 Uncalibrated



AR4 Calibrated



Would better decisions be made (eg with regard to regional infrastructure investments for climate adaptation) using the calibrated probabilities compared with the uncalibrated probabilities?



### 3. Reducing uncertainty of the second kind: stochastic-dynamic parametrisation

**Palmer, 2001:** A nonlinear dynamical perspective on model error: a proposal for non-local stochastic-dynamic parametrisation in weather and climate prediction models. QJ, 127, 279-304.

Medium-Range Ensemble Prediction  
Systems must include  
representations of model uncertainty  
in order that ensemble forecasts are  
not overconfident.

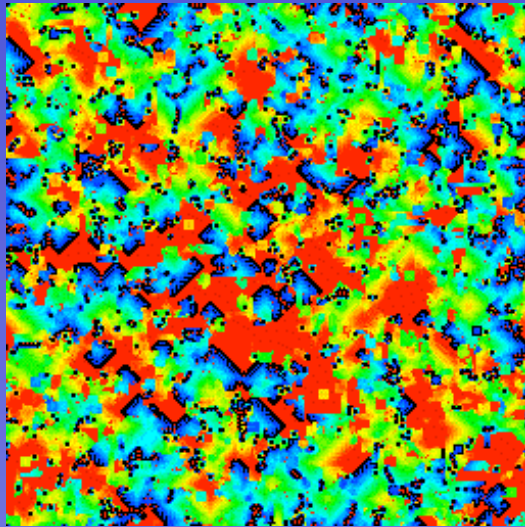
To do this, a stochastic  
representation of random sub-grid  
model uncertainties has been  
developed.

$$\dot{T} = \dot{T}_{\text{dynamics}} + \dot{T}_{\text{long-wave radiation}} + \dot{T}_{\text{short-wave radiation}} + \dot{T}_{\text{convection}} + \dots$$

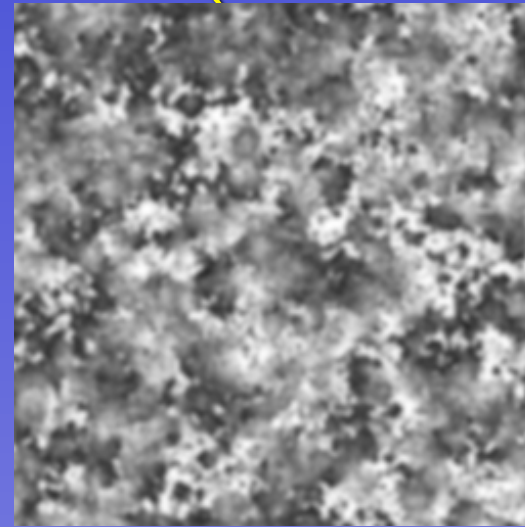
Represent each sub-grid tendency as a pdf  
(Buizza et al, 1999)



# Cellular Automaton Stochastic Backscatter Scheme (CASBS)



smooth  
→  
scale



Cellular Automaton state

streamfunction forcing shape  $\Psi$   
function

$$\frac{\partial \psi}{\partial t} = \Psi(x, y) \cdot \sqrt{rD}$$

$D$  = sub-grid energy dissipation due to numerical diffusion,  
mountain drag and convection

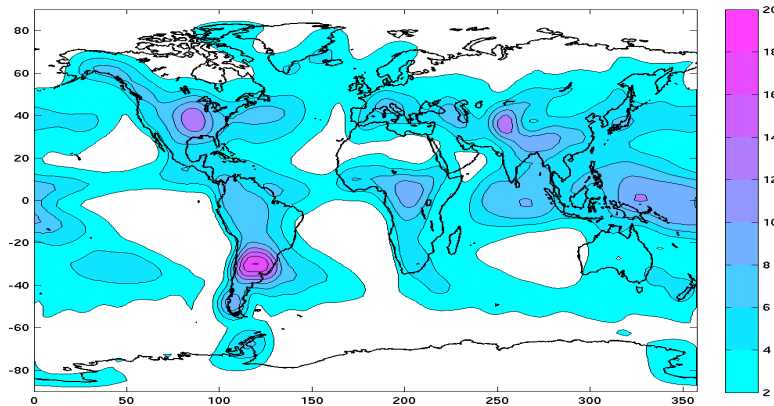
$r$  = backscatter parameter

G.Shutts, 2005

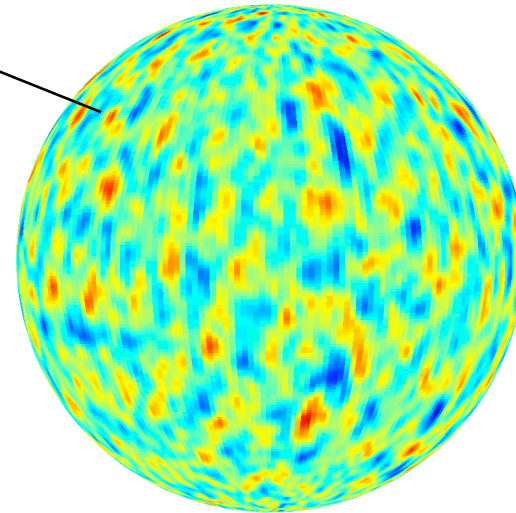
# Stochastic Spectral Backscatter Scheme (SPBS)

Rationale: A fraction of the dissipated energy is scattered upscale and acts as streamfunction forcing for the resolved-scale flow (LES) (cf Shutts and Palmer 2004, Shutts 2005)

$$\Delta\psi^* \propto \sqrt{D}\psi'$$



**Total Dissipation rate from numerical dissipation, convection, gravity/mountain wave drag.**



**Spectral Markov chain: temporal and spatial correlations prescribed**

Berner et al, 2008

Because of the nonlinearities of climate, this stochastic representation of model uncertainty **can also reduce systematic biases in climate models.**

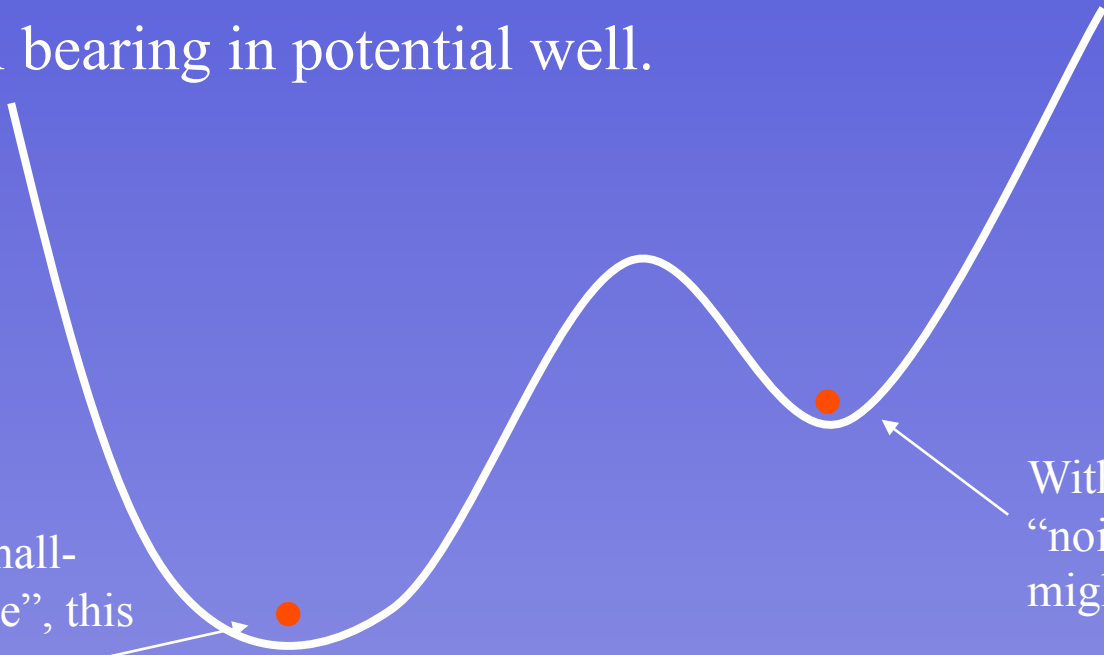
# Can adding noise change the mean state of a system?

Eg ball bearing in potential well.

Without small-scale “noise”, this regime is too dominant

Without small-scale “noise”, this minimum might be inaccessible

With noise, the mean state shifts to the right



PHILOSOPHICAL  
TRANSACTIONS  
— OF —  
THE ROYAL  
SOCIETY

A

MATHEMATICAL, PHYSICAL  
& ENGINEERING SCIENCES

ISSN 1364-503X

volume 366

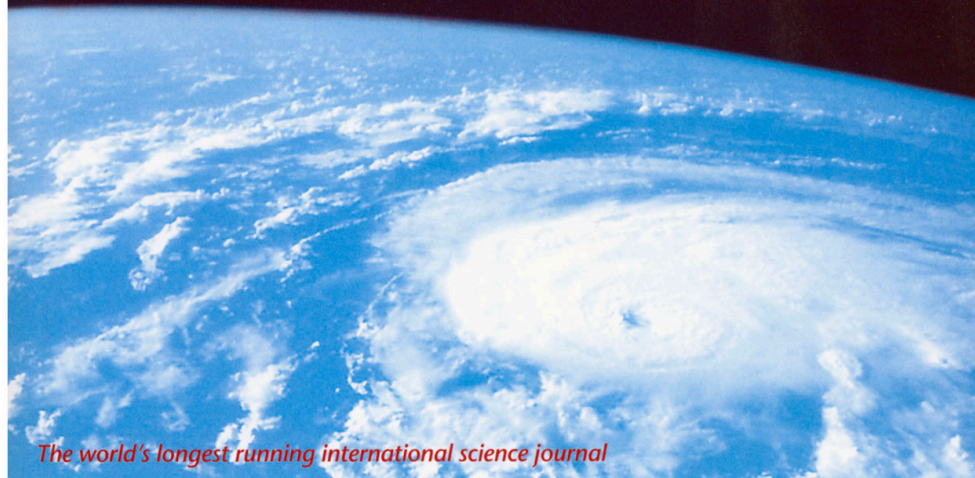
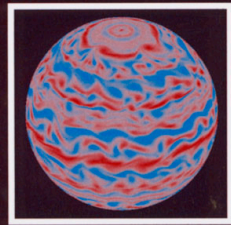
number 1875

pages 2419–2641

*In this issue*

***Stochastic physics and climate modelling***

*Papers of a Theme Issue compiled and edited by Tim Palmer and Paul Williams*



*The world's longest running international science journal*



Royal Society **Publishing**

*Informing the science  
of the future*

**28 July 2008**

If an Earth-System model purports to be a comprehensive tool for predicting climate, it should be capable of predicting the uncertainty in its predictions.

The governing equations of Earth-System models should be inherently probabilistic.

“I believe that the ultimate climatic models....will be stochastic, ie random numbers will appear somewhere in the time derivatives.” Lorenz (1975)

Lorenz E.N. 1975. Climatic Predictability. In “The Physical Basis of Climate and Climate Modelling”. WMO GARP Publication Series No 16. World Meteorological Organisation. Geneva: 265 pp.



# Seamless Prediction and Decadal Forecasting

---

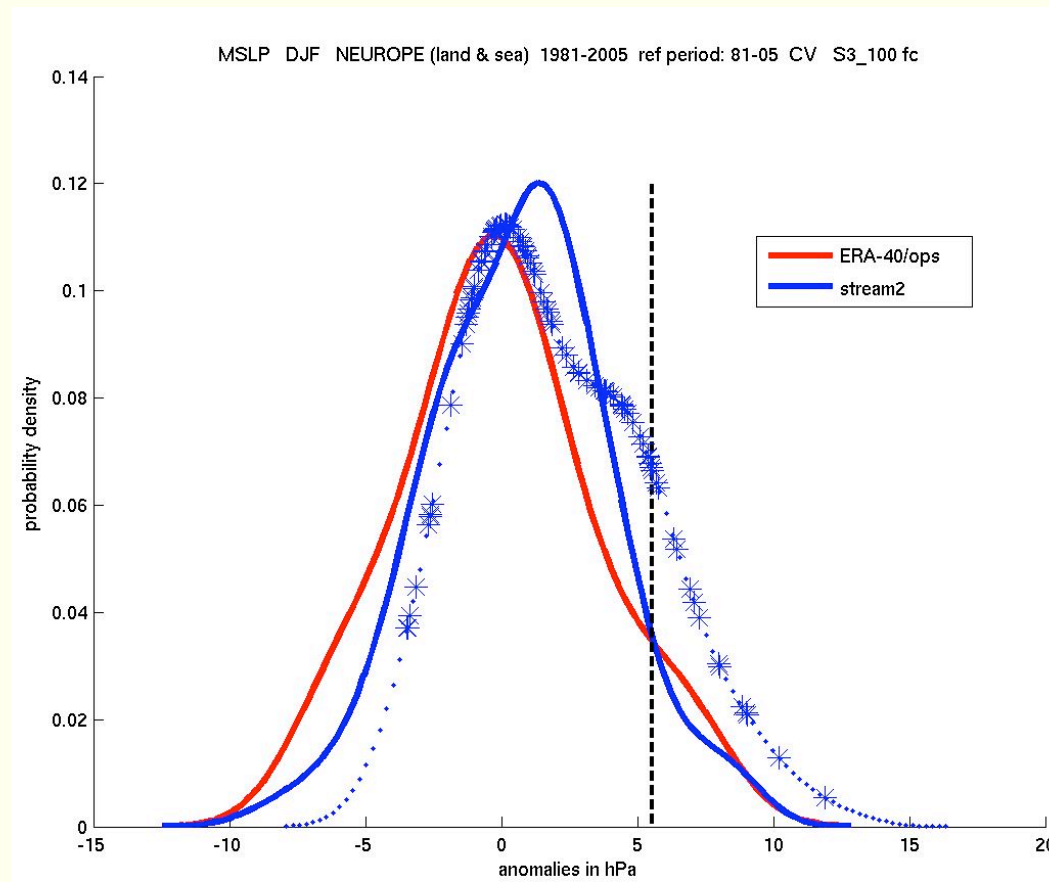
- Decadal prediction provides a meeting ground between the weather and climate communities
  - ❖ Impact of accurate initial conditions (weather)
  - ❖ Impact of greenhouse-gas scenarios (climate)and therefore a focus for seamless prediction studies.
- Possible contribution to AR5?



# Return to Winter of 2005/6

- How good were the ECMWF seasonal forecasts for Northern Europe?
- Could these predictions have been improved if the a) stratospheric polar vortex, b) the tropics, had been successfully forecast?

Northern Europe (land & sea) DJF 2005/06 sea level pressure



coupled forecast  
for DJF 2005/06  
with 100 ensemble members

# Relaxation Formulation

ECMWF model:

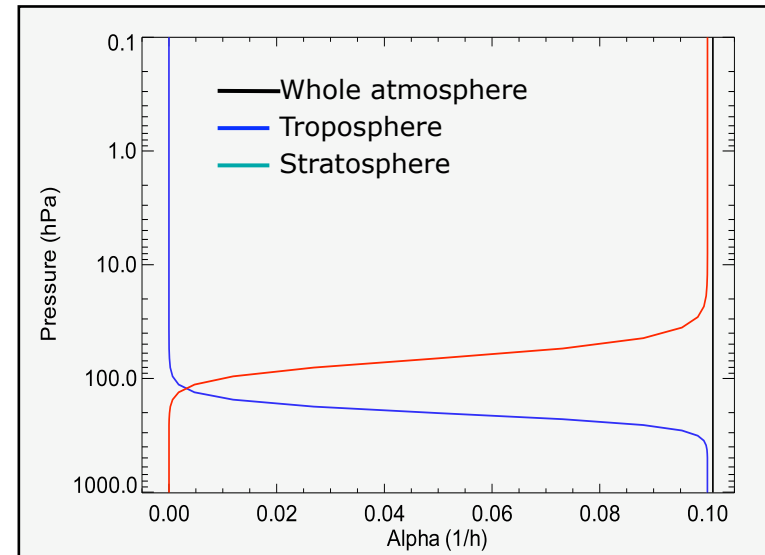
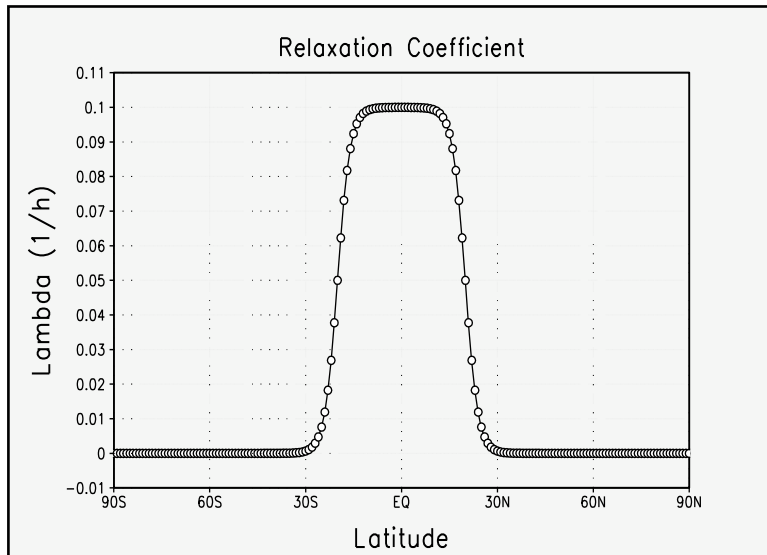
$$\frac{\partial \mathbf{x}}{\partial t} = M(\mathbf{x})$$

ECMWF model with relaxation:

$$\frac{\partial \mathbf{x}}{\partial t} = M(\mathbf{x}) - \lambda(\mathbf{x} - \mathbf{x}^{ref})$$

- Relaxations coefficient,  $\lambda$ , depends on longitude, latitude and height.
- Relaxation for  $u$ ,  $v$ ,  $T$  and  $\ln p_s$  (same  $\lambda$ )
- $\mathbf{x}^{ref}$  is based on (interpolated) ERA-40 data.

# Relaxation Regions



# Experimental Setup

- Model version 32R1 (5/06–5/11 2007)
- T<sub>L</sub>95 (210 km) with 60 vertical levels
- Initial and boundary conditions as well as  $\mathbf{x}^{\text{ref}}$ : Operational analyses (interpolated to T<sub>L</sub>95L60)
- Ensemble: 17 members (2005111612/to/2005112012/by/6hrs)
- Period of interest: 1/12/05 bis 28/02/06

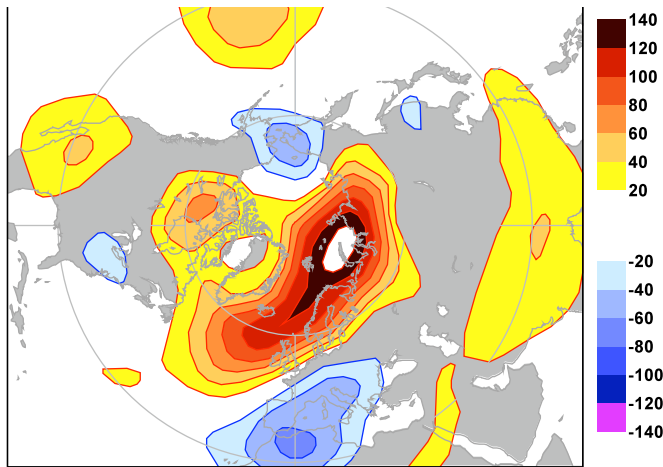
- Calibration run (16. November, 1990-2006)

- Control ensemble (observed SST/sea ice)

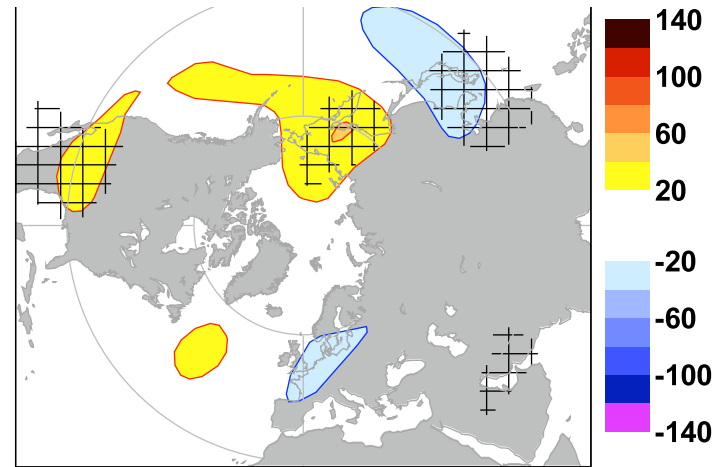
- Relaxation ensembles (various regions)

# Z500 Anomalies (DJF 2005/06)

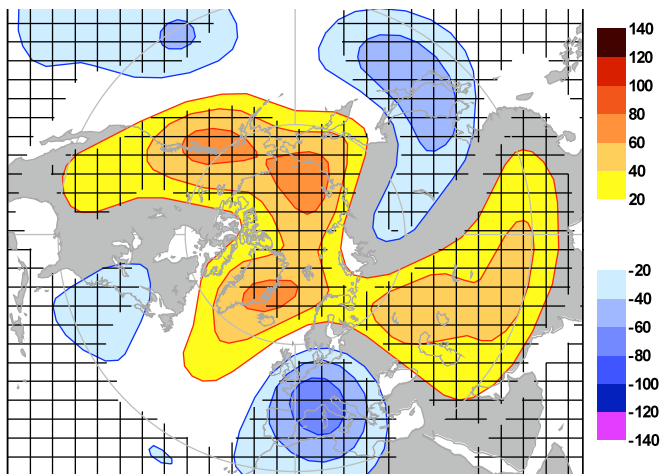
Observations



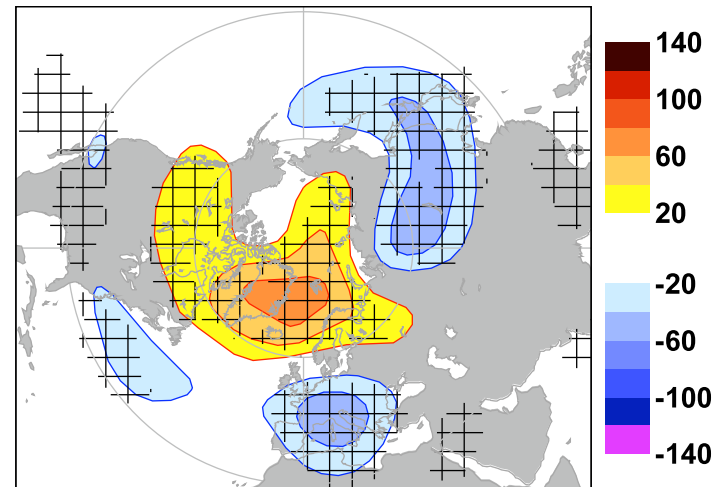
Control Integration



Relaxation: Tropics

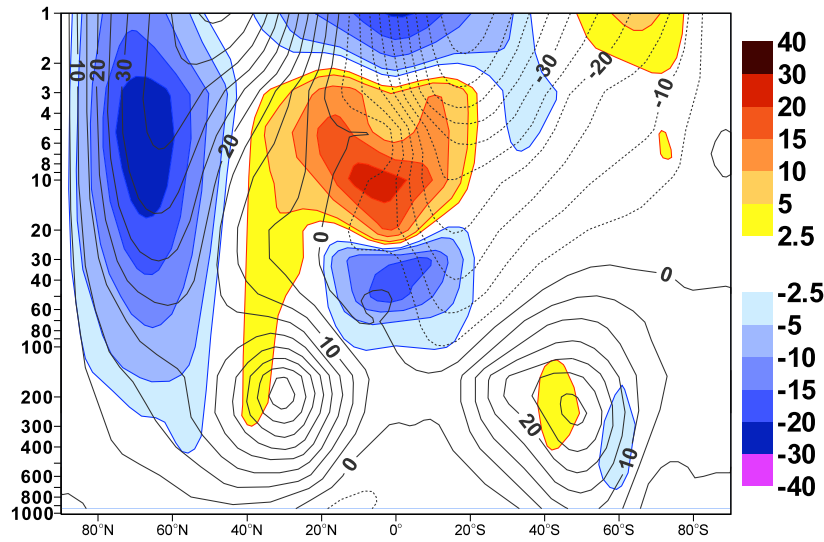


Relaxation: Polar Vortex

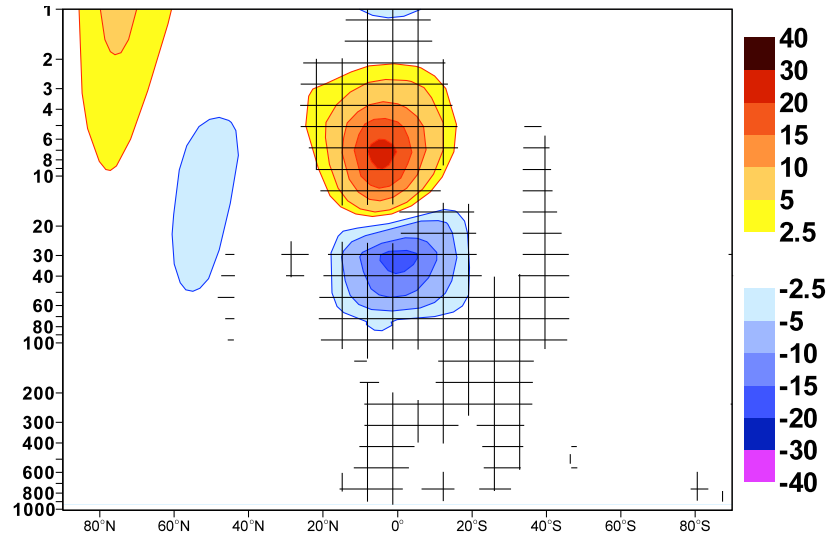


# [u]-Anomalies (DJF 2005/06)

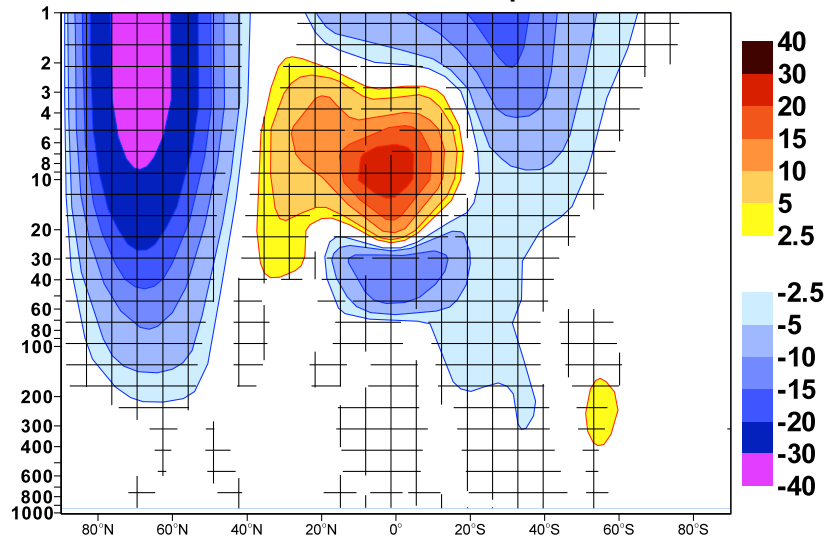
## Observations



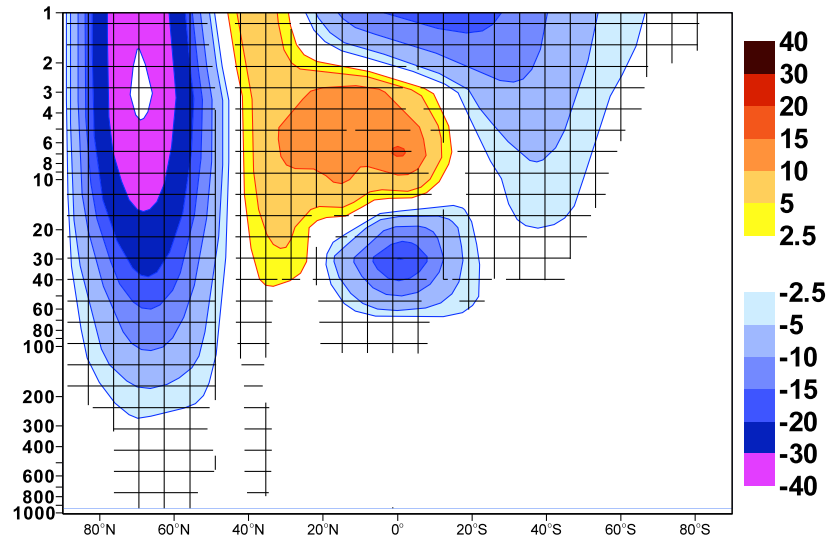
## Control Integration



## Relaxation: Tropics

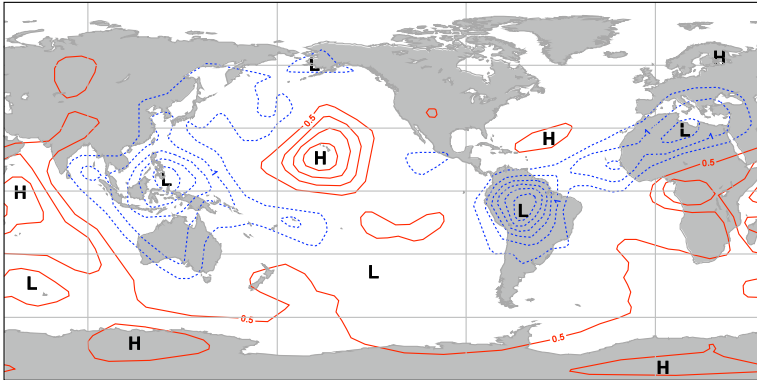


## Relaxation: Polar Vortex

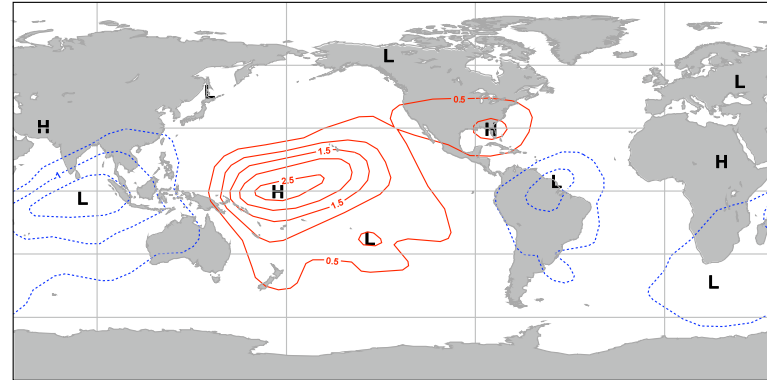


# X Anomalies 200-300hPa (DJF 2005/06)

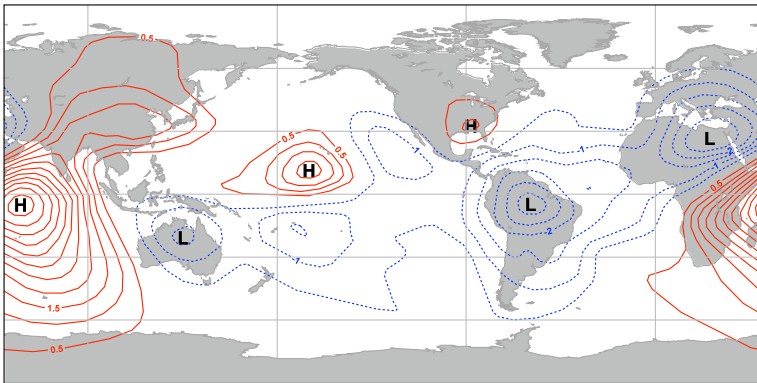
Observations



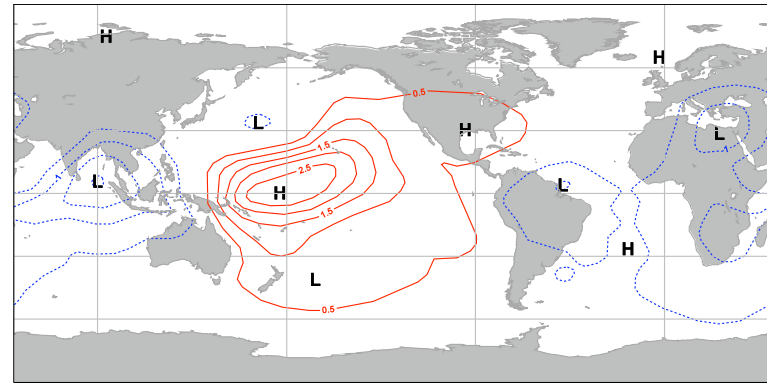
Control Integration



Relaxation: Tropics



Relaxation: Polar Vortex



Notice: "Observed" anomaly =  $OD(05/06) - E4(\text{Climate})$ , that is, different model formulation + vertical resolution





## Conclusions (I)

---

- Climate change projections are still uncertain
  - ❖ Uncertainty of the first kind (multi-model spread)
  - ❖ Uncertainty of the second kind (common model deficiencies)
- Seamless prediction allows the insights and constraints of numerical weather forecasting to be brought to bear on the climate-change prediction problem
- Decadal prediction is the natural meeting ground between weather and climate communities. Fertile area for 2-way interaction



## Conclusions (II)

---

- Middle atmosphere plays a role in tropospheric climate simulation, but need for major improvements in our ability to simulate the tropics remains a priority.
  - ❖ Need to study how well our ability to simulate climate is improved using convectively-resolved models.
  - ❖ Need for dedicated petaflop computing resources for climate (WCRP Climate Modelling Summit)