Model Imperfections

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With acknowledgement for contributions with no imperfections to:

Judith Berner, Roberto Buizza, Paco Doblas-Reyes, Renate Hagedorn, Thomas Jung, Glenn Shutts



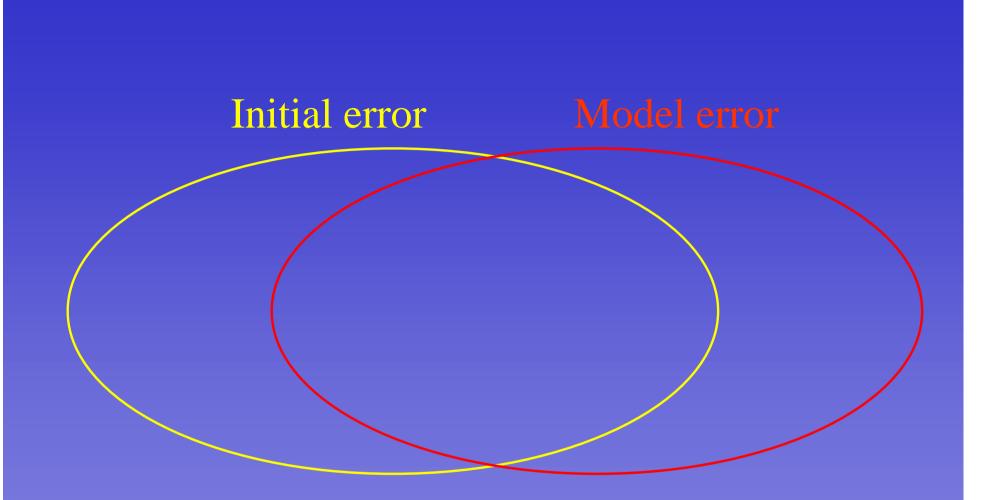
Imperfect forecasts are due to:

Initial imperfections

Model imperfections

But...





These are not disjoint sets!



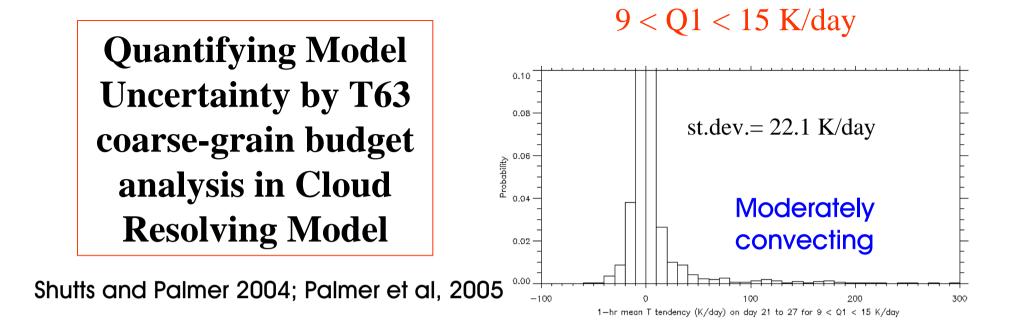
Initial condition – a value of the forecast model's state vector obtained by assimilating observations (typically point or pixel-scale) into the model.

•Observations which are influenced by scales of motion that are not well represented by the model's equations of motion (up to several grid lengths) can't be assimilated accurately.

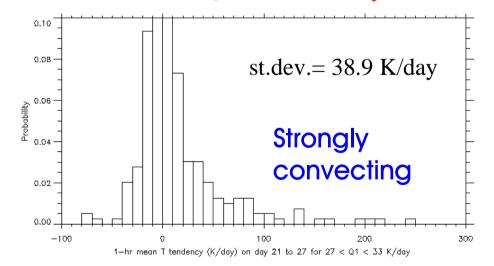
•Such model-induced initial error doesn't partition well into J_b and J_o .

-Singular vectors may sample some of this modelinduced initial uncertainty.





27 < Q1 < 33 K/day



pdfs from T tendency sub-samples selected according to their Q1 range.

tendency data drawn from 7 fields (each with 8192 x 128 points) at z=5 km and 24 hours apart



Spread-Skill for Three Operational Ensemble Forecast Systems

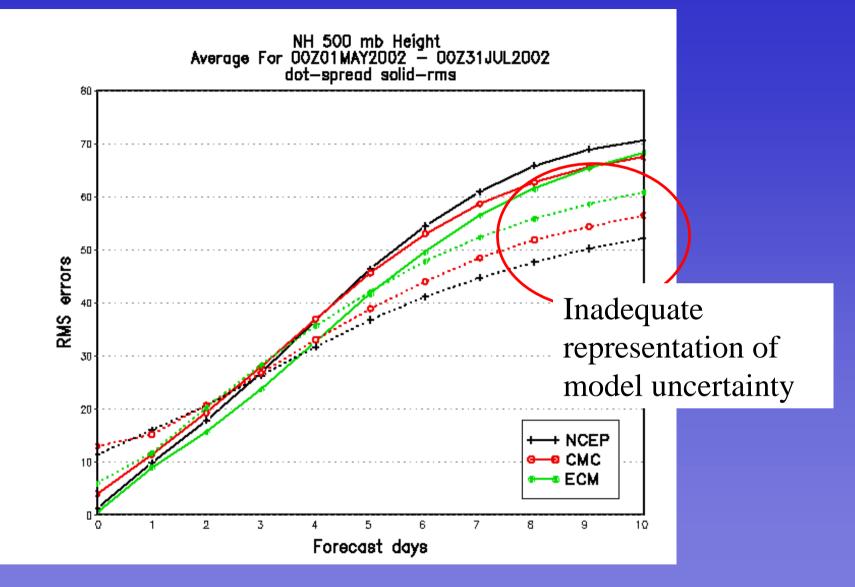
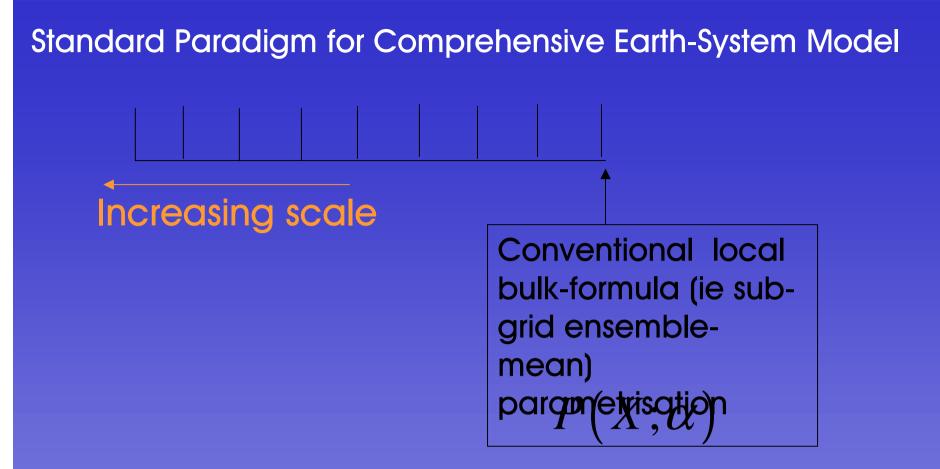


Figure 6. May-June-July 2002 average RMS error of the ensemble-mean (solid lines) and ensemble standard deviation (dotted lines) of the EC-EPS (green lines), the MSC-EPS (red lines) and the NCEP-EPS (black lines). Values refer to the 500 hPa geopotential height over the northern hemisphere latitudinal band 20°-80°N.



Representing model uncertainty *within* this general framework:

- Sample over different GCMs (eg DEMETER, CMIP)
- Sample over different parametrisations *P* (eg Houtekamer, 1996)

Sample over different parameters α (eg Murphy et al, 2004; Stainforth et al 2005)





Development of a European Multi-Model Ensemble System for Seasonal to Interannual Prediction



DEMETER Multi-model ensemble system

• 7 global coupled ocean-atmosphere climate models

Partner	Atmosphere	Ocean
ECMWF		HOPE
	IFS	OPA 8.3
	ARPEGE	OPA 8.1
	ARPEGE	OPA 8.3
	ECHAM-4	OPA 8.2
	ECHAM-5	MPI-OM1
	HadCM3	HadCM3

9 member ensembles
ERA-40 initial conditions
SST and wind perturbations
4 start dates per year
6 months hindcasts

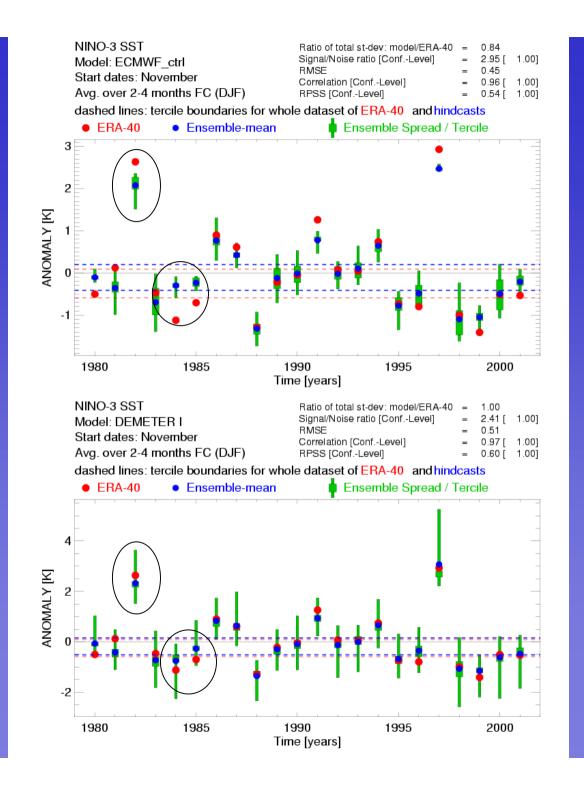
Hindcast production for: 1980-2001 (1958-2001)



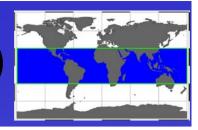
ECMWF

DEMETER

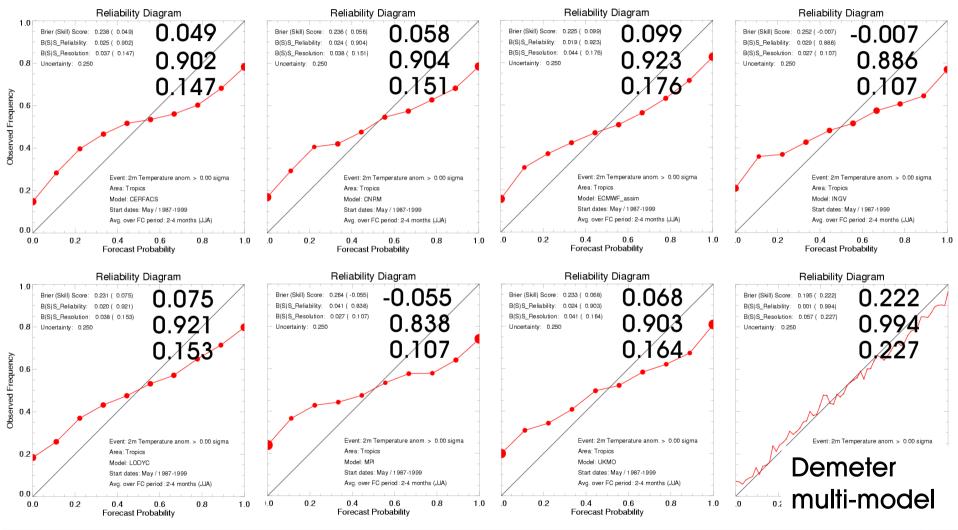
Palmer et al, 2004; Hagedorn et al 2005







Reliability: 2m-Temp.>0



Palmer et al, 2004 BAMS; Hagedorn et al 2005 Tellus.

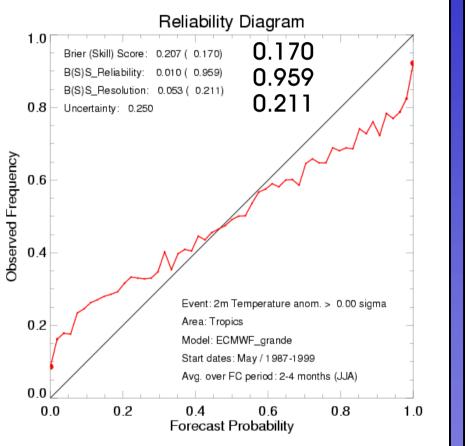
Reliability: 2m-Temp.>0

Reliability Diagram

Brier (Skill) Score: 0.195 (0.222)

0.222

1.0



0.994 B(S)S_Reliability: 0.001 (0.994) B(S)S Resolution: 0.057 (0.227) 0.227 0.8 - Uncertainty: 0.250 **Observed Frequency** 0.6 0.4 Event: 2m Temperature anom. > 0.00 sigma Area: Tropics 0.2 Model: DEMETER II Start dates: May / 1987-1999 Avg. over FC period: 2-4 months (JJA) 0.0 0.2 0.0 0.4 0.8 1.0 0.6 Forecast Probability

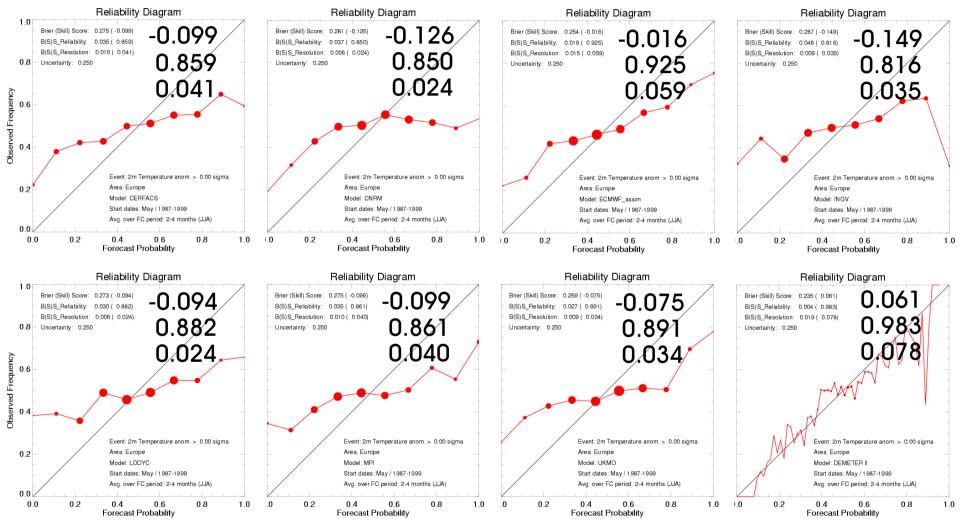
single-model (54 members)

multi-model



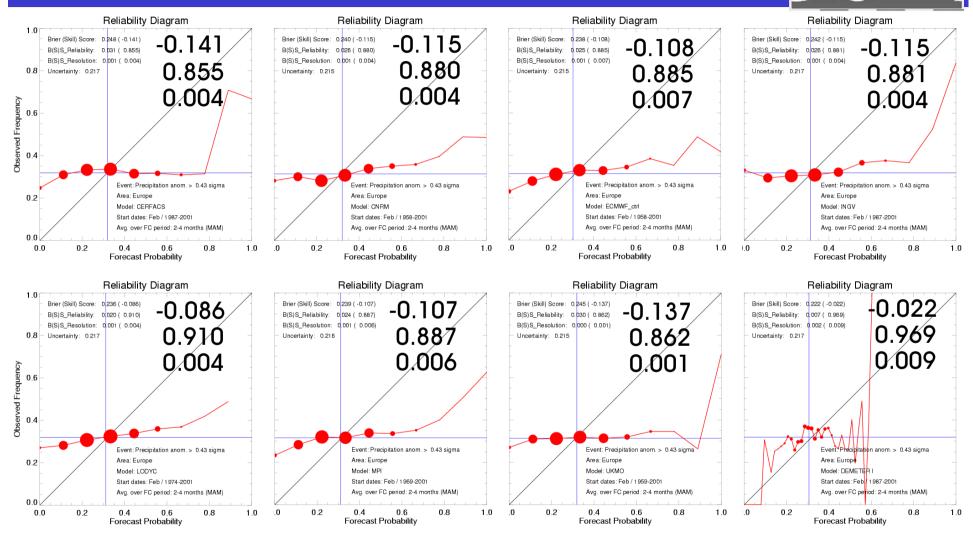
Reliability: 2m-Temp.>0





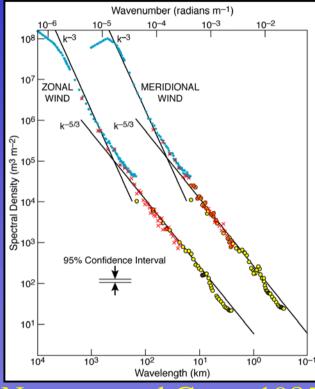
E



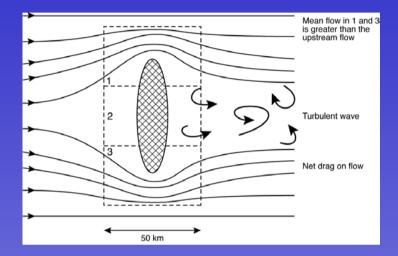


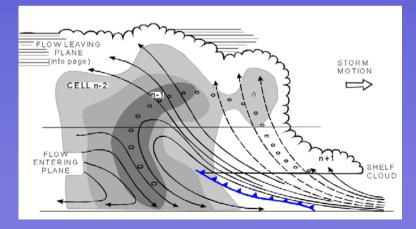


Are there inherent shortcomings to the multimodel representation of uncertainty? Yes!



Nastrom and Gage, 198







Stochastic Paradigm for Comprehensive Earth-System Model

Increasing scale

Stochastic parametrisation

 $P(X;\alpha,e)$

e is spatio-temporally correlated stochastic noise, sampling some prescribed sub-grid PDF (eg based on a CRM).

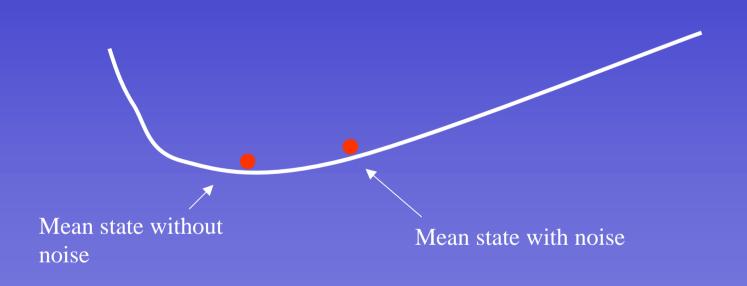
Fundamental conceptual difference from standard paradigm: at each timestep, stochastic parametrisation generates a specific realisation, and not the mean of some putative ensemble of sub-grid circulations. Possible benefits of stochastic parametrisation:More complete representation of model uncertainty

•Reduction in model systematic error (noiseinduced drift)

•More accurate estimate of internal climate variability (cf detection/attribution studies)



Could stochastically sampling the probability distribution of the sub-grid tendency, rather than always sampling the mode of the distribution, make a difference?





Without small-scale "noise", this minimum might be inaccessible

Without smallscale "noise", this regime is too dominant



ECMWF Stochastic Physics Scheme

$$\dot{X} = D + P + e$$
$$e = \varepsilon P$$

Buizza, Miller and Palmer, 1999; Palmer 2001

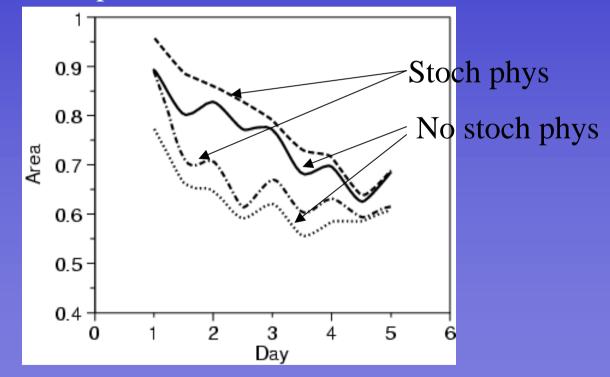
•ε is a stochastic variable, drawn from a uniform distribution in [-0.5, 0.5], constant over time intervals of 6hrs and over 10x10 lat/long boxes

•Multiplicative form of noise consistent with coarse-grained budget analysis from cloud-resolved model.



<u>Stochastic Physics has a positive impact on</u> <u>medium-range EPS skill</u>

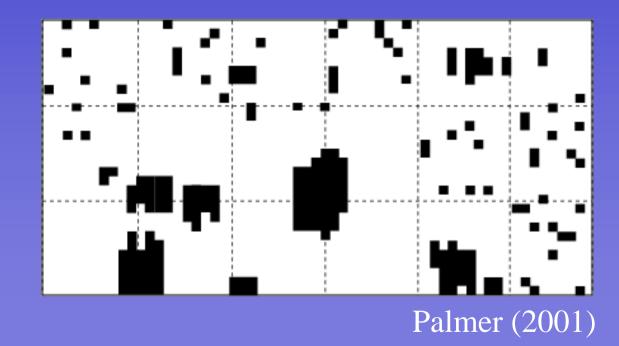
Area under ROC curve. E: precip>40mm/day. Winter- top curves. Summer – bottom curves



Buizza et al, 1999



Instead of "pure" stochastic forcing, consider a stochastic-dynamic representation of subgridscale – eg cellular automaton (Wolfram 2002) – can represent both chaotic (stochastic) and coherent (eg soliton, wave-like) processes.

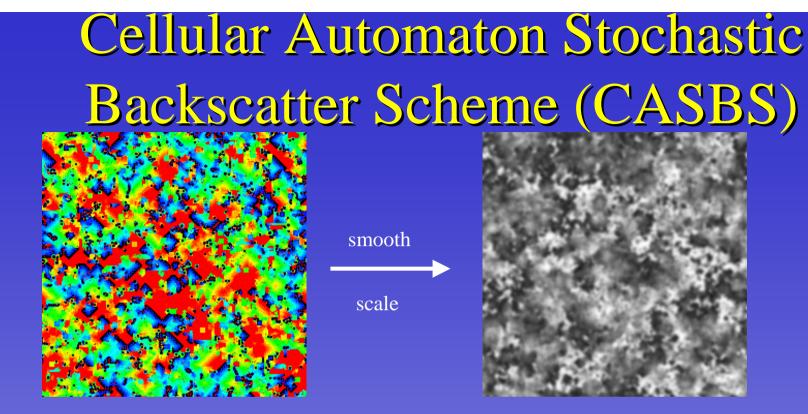




A stochastic-dynamic paradigm for Earth-System Models

Increasing scale

Coupled over a range of scales Computationally-cheap nonlinear stochasticdynamic model, providing specific realisations of subgrid motions rather than ensemble-mean sub-grid effects



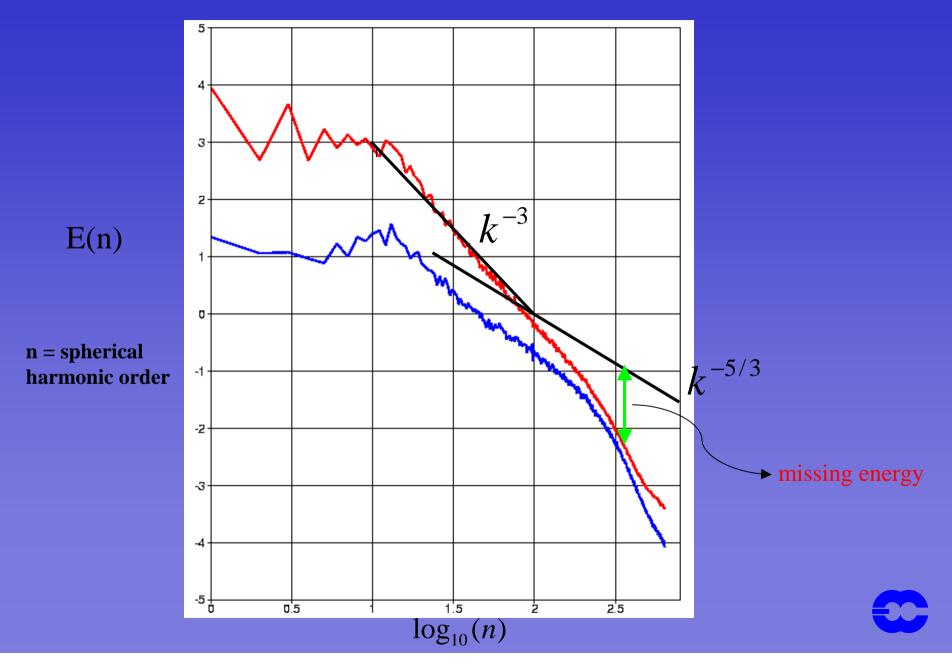
Cellular Automaton state

streamfunction forcing shape Ψ function

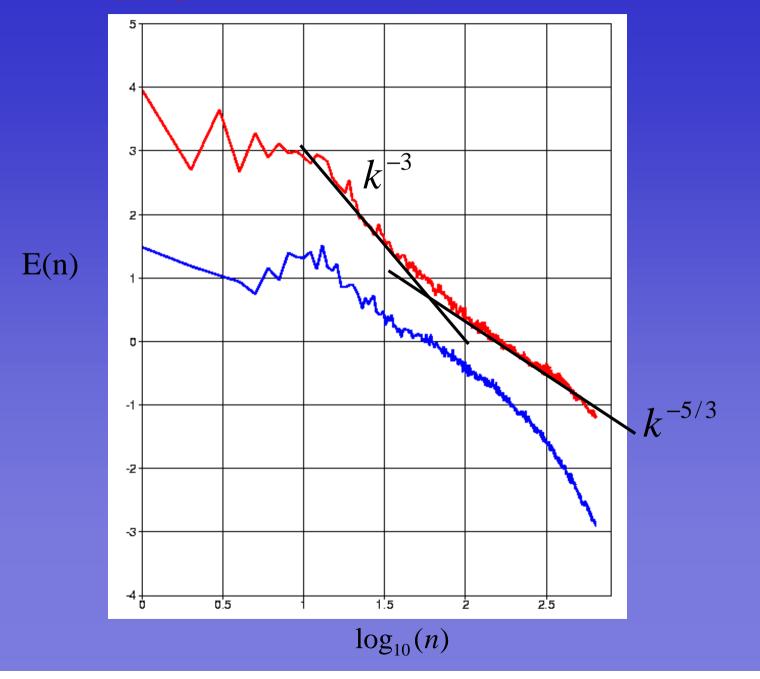
$$\frac{\partial \psi}{\partial t} = \alpha \cdot \Psi(x, y) \cdot \sqrt{D}$$

D = sub-grid energy dissipation due to numerical diffusion, mountain drag and convection $\alpha =$ dimensional parameter G.Shutts, ECMWF Tech Memo

Energy spectrum in 5-day T799 run



Energy spectrum in 5-day T799 run with CASBS



Model Integrations

6 month runs from 1 October 1962-2001

Cy26r3 TL95L60 atmosphere only (with and without CASBS)

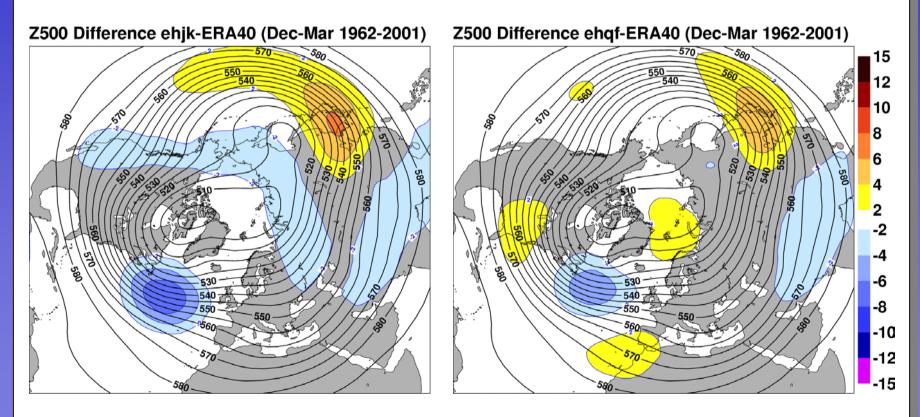
> Analysis on DJFM



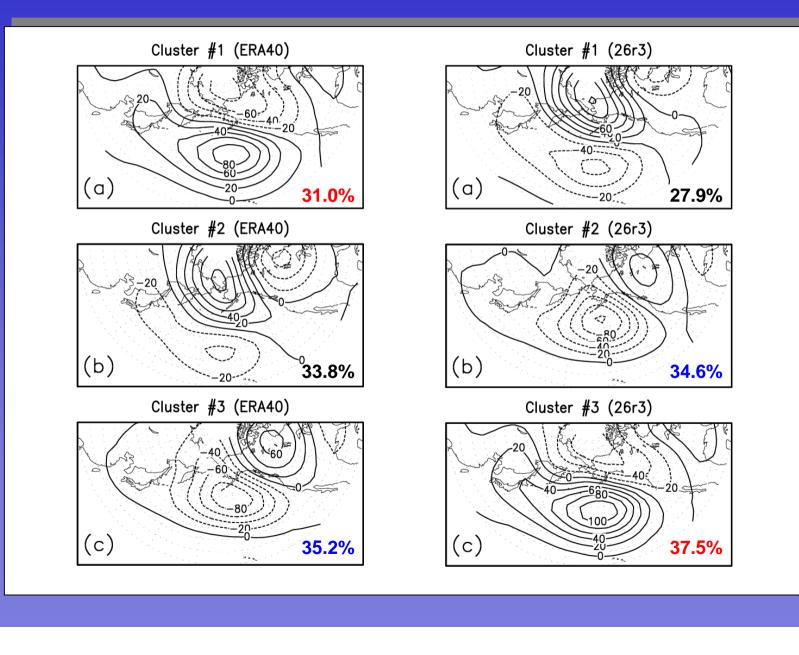
Systematic Z500 Error

Control

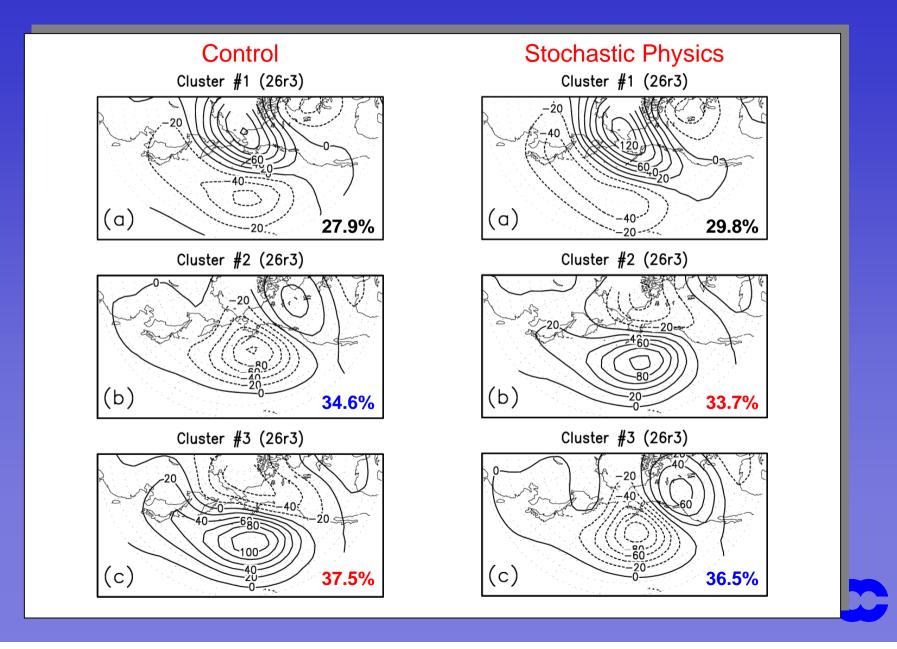
Stochastic Physics



Weather Regimes: ERA-40 vs control



Weather Regimes: Impact of Stochastic Physics



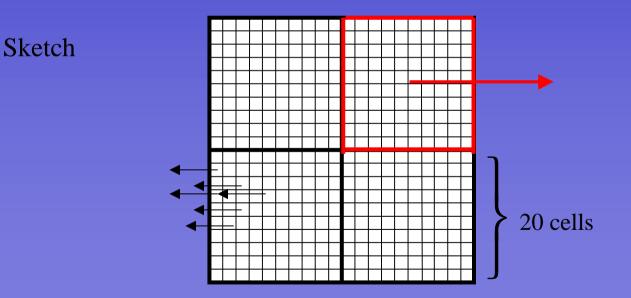
Develop the CA rules to incorporate meteorology. For example

- •Agglomerations (organised convection) have larger probability of continued life
- •Advection of individual on-cells by trade wind (through grid-box boundaries)
- •Eastward propagation of "probability waves" coupled to on-state cells (cf moist Kelvin-wave dynamics; MJO)

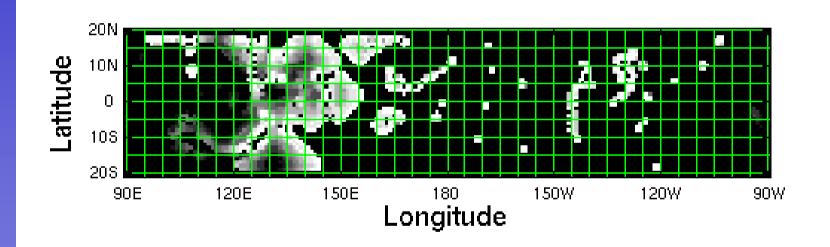


Multiscale-CA

- Small-scale cells evolve according to rules and propagate/advect to the west
- Intermediate-scale cells can be on/off and propagate to the east
- Fertile cells in small-scale CA can only be born if intermediate-scale cell is 'on'

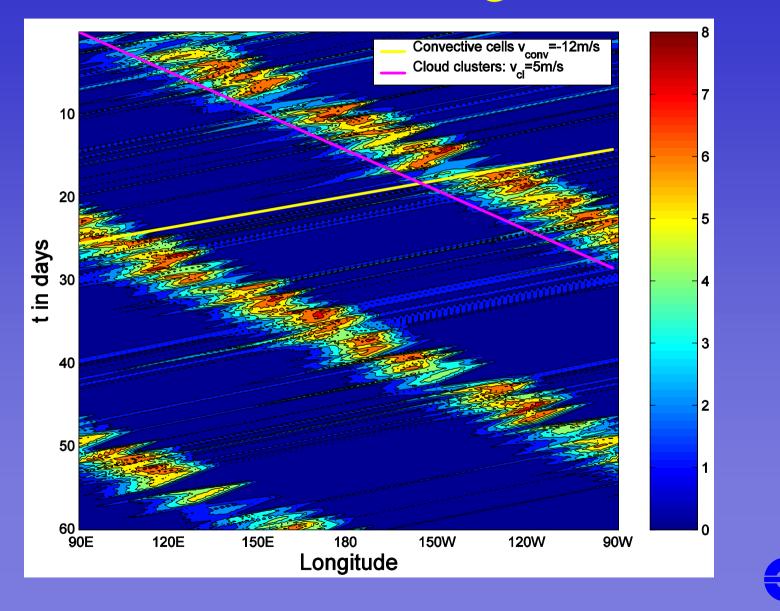








Hovmöller diagram



ENSEMBLES

- (EU-FP6 Integrated Project, Successor of DEMETER)
- Inter-compare performance of:
- Multi-model;
- Perturbed parameter;
- Stochastic-dynamic parametrisation;
- in coordinated seasonal and decadal timescale integrations.

ENSEMBLES will provide substantial input into the new WCRP COPES (Coordinated Observation and Prediction of the Earth System) initiative.



CONCLUSIONS and FUTURE DIRECTIONS

Representing model uncertainty is still at a primitive stage. Much important work could be done, eg by analysis of coarse-grained budgets from cloud-resolving models.

•A study intercomparing different representations of model uncertainty:

Multi-model ensembles

Multi-parameter ensembles

Stochastic parametrisation

could/should be undertaken for 15-day forecast timescales under TIGGE auspices ("THORPEX meets COPES").

