

Blurring the boundary between dynamics and physics

Tim Palmer, Peter Düben, Hugh McNamara
University of Oxford



$$\rho \left(\frac{\partial}{\partial t} + \mathbf{u} \cdot \nabla \right) \mathbf{u} = \rho \mathbf{g} - \nabla p + \nu \nabla^2 \mathbf{u}$$

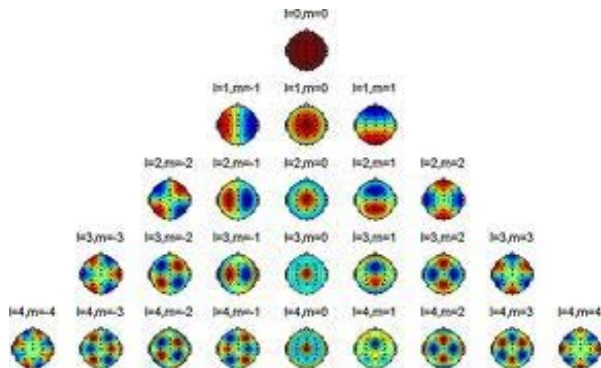
Resolved scales

The Canonical Numerical Ansatz

Unresolved scales

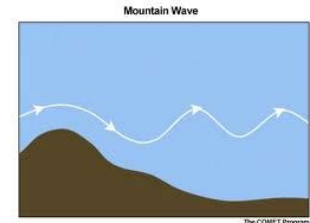
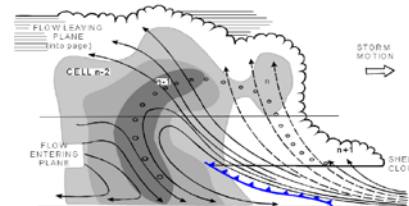
Dynamical Core

$$\zeta = \sum_{m,l} \zeta_{ml} e^{im\lambda} P_l^m(\phi)$$



Parametrisations

$$P(X_{Tr}; \alpha)$$



$$D = P$$

$$\rho \left(\frac{\partial}{\partial t} + \mathbf{u} \cdot \nabla \right) \mathbf{u} = \rho \mathbf{g} - \nabla p + \nu \nabla^2 \mathbf{u}$$

Resolved scales

Unresolved scales

Dynamical Core

$$\zeta = \sum_{m l}^{\infty} \zeta_{ml} e^{im\lambda} P_l^m(\phi)$$

- Discretisation errors
- Convergence errors
- Round-off errors

Parametrisations

$$P(X_{\text{Tr}}; \alpha)$$

- Errors in the functional form of P
- Errors in the assumed values of α

$$\rho \left(\frac{\partial}{\partial t} + \mathbf{u} \cdot \nabla \right) \mathbf{u} = \rho \mathbf{g} - \nabla p + \nu \nabla^2 \mathbf{u}$$

Resolved scales



Truncation Scale (7 to 8 orders of magnitude above viscous scale!)

Unresolved scales

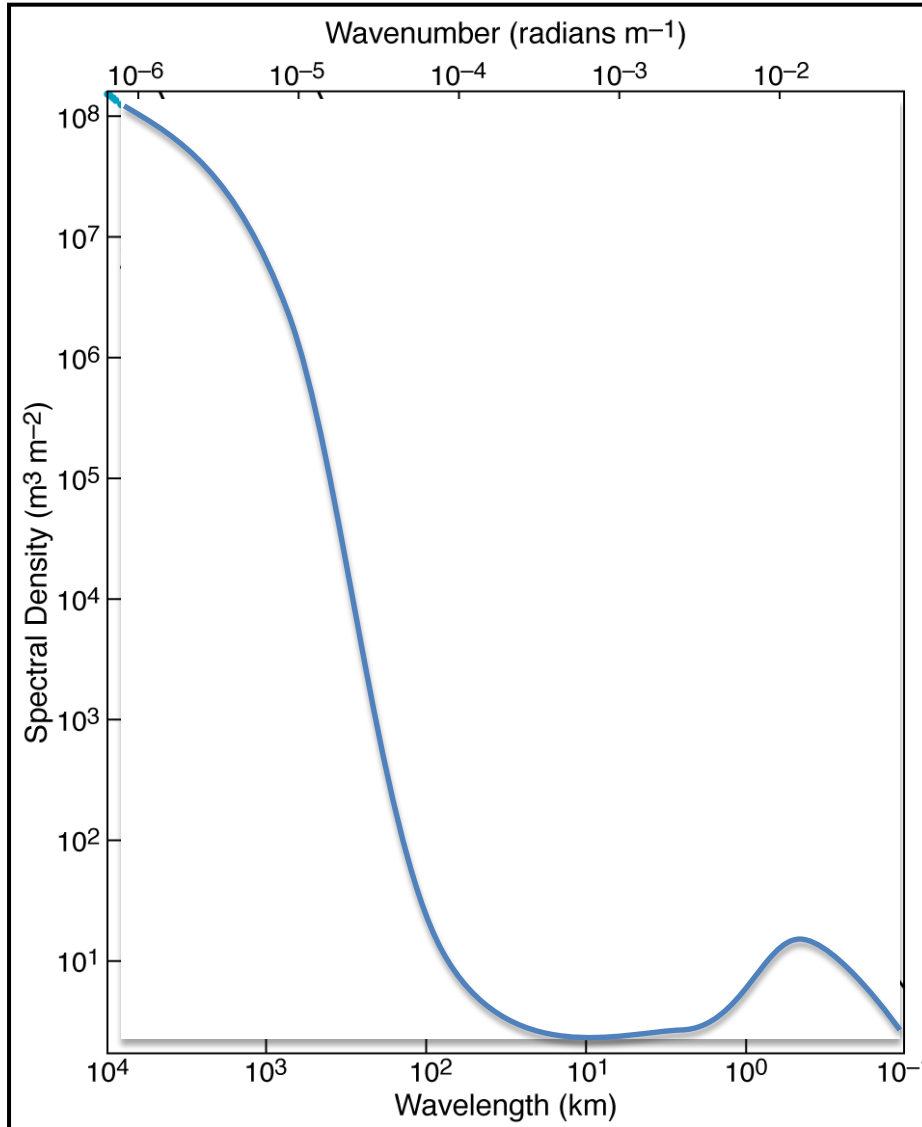


Dynamical Core



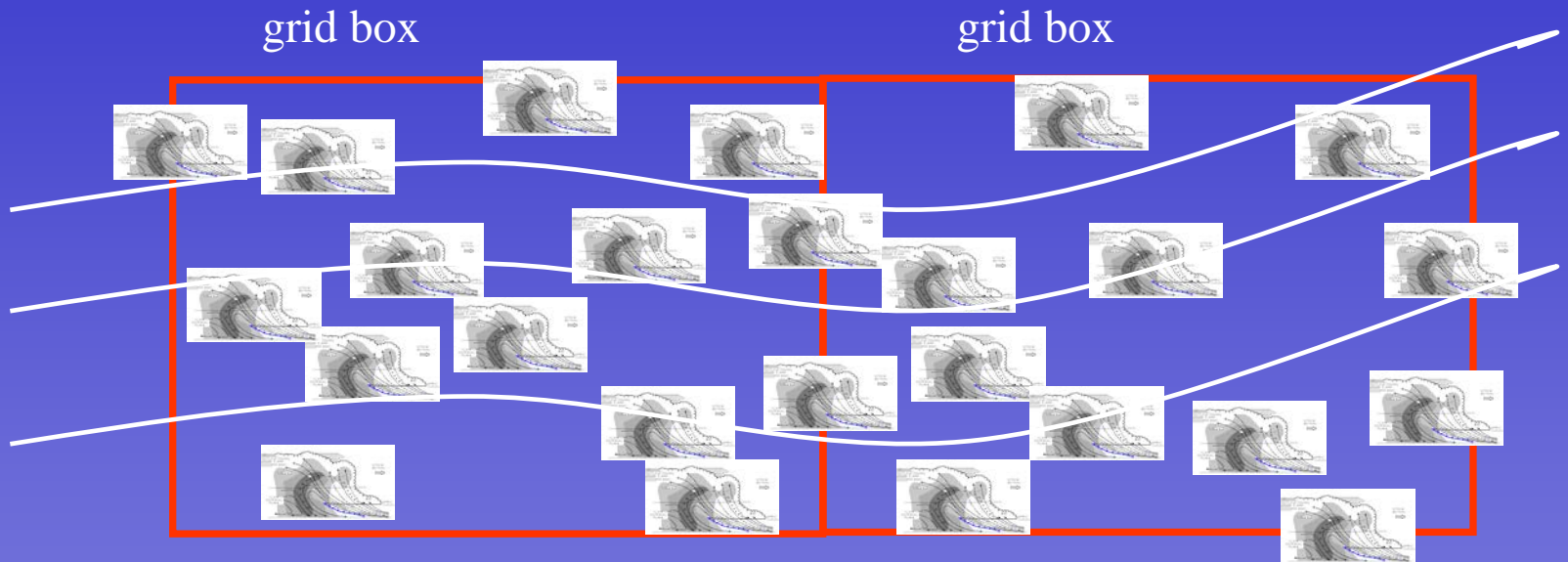
Parametrisations



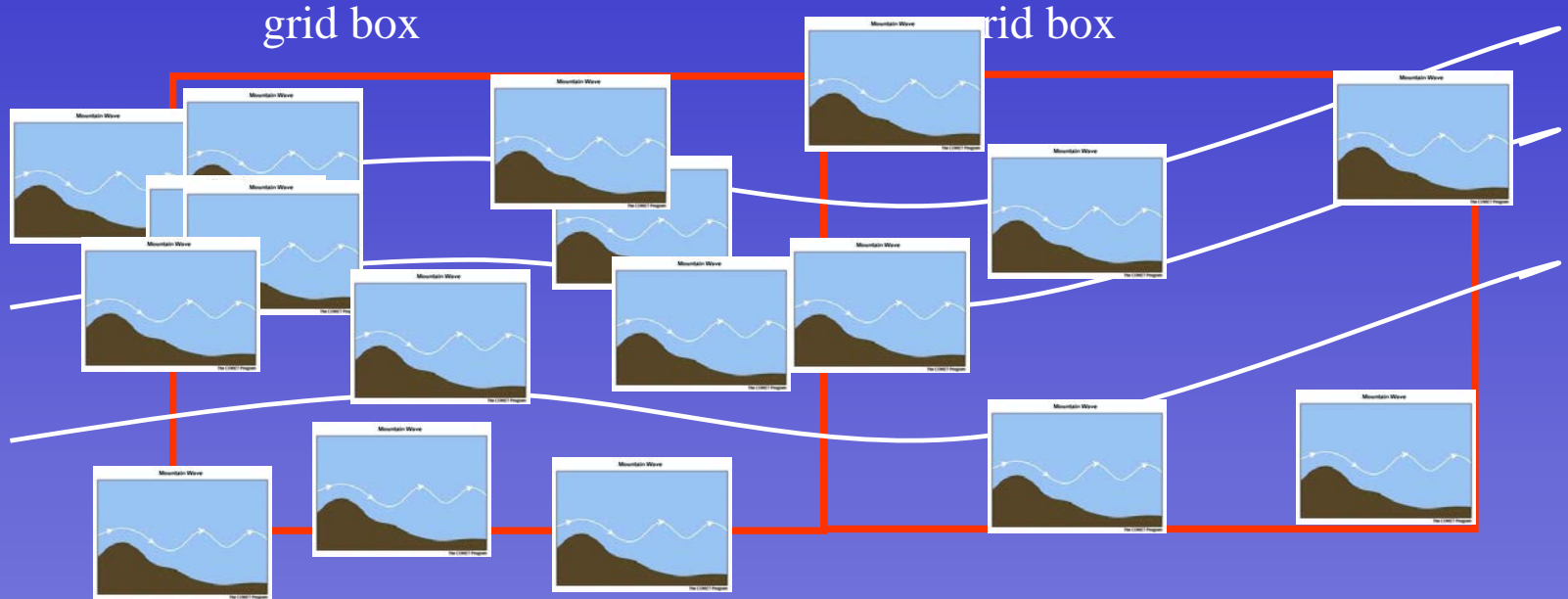


If the energy spectrum in the atmosphere was like this.....

... ie the world looks like this

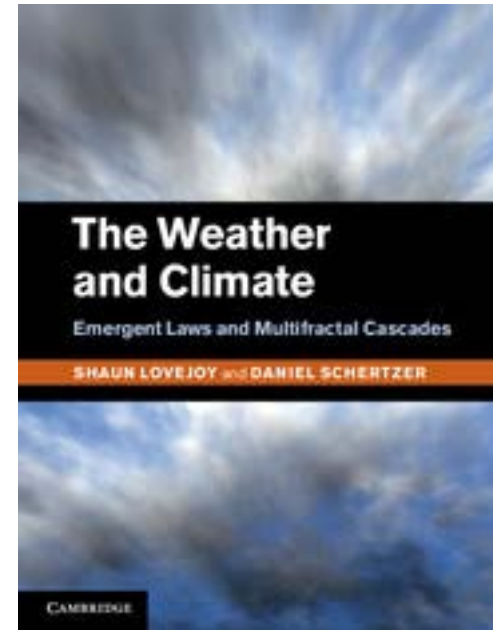
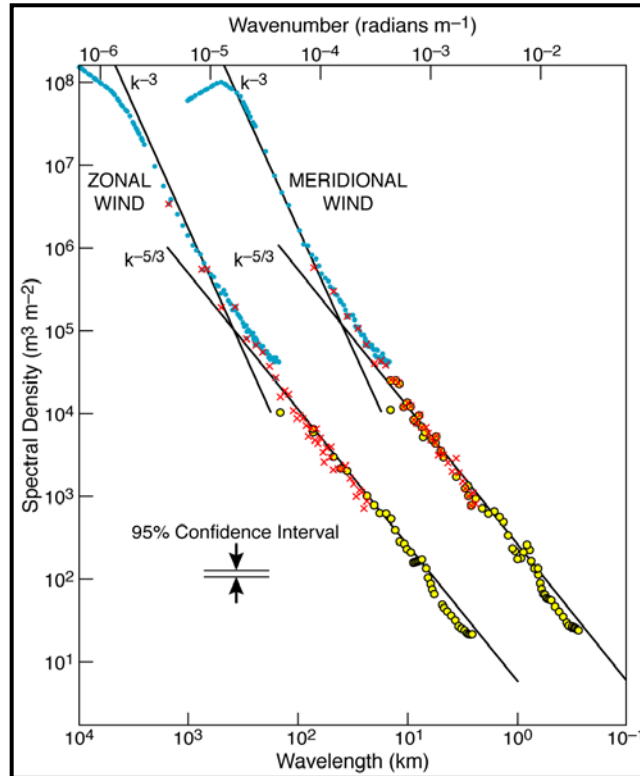


... or this

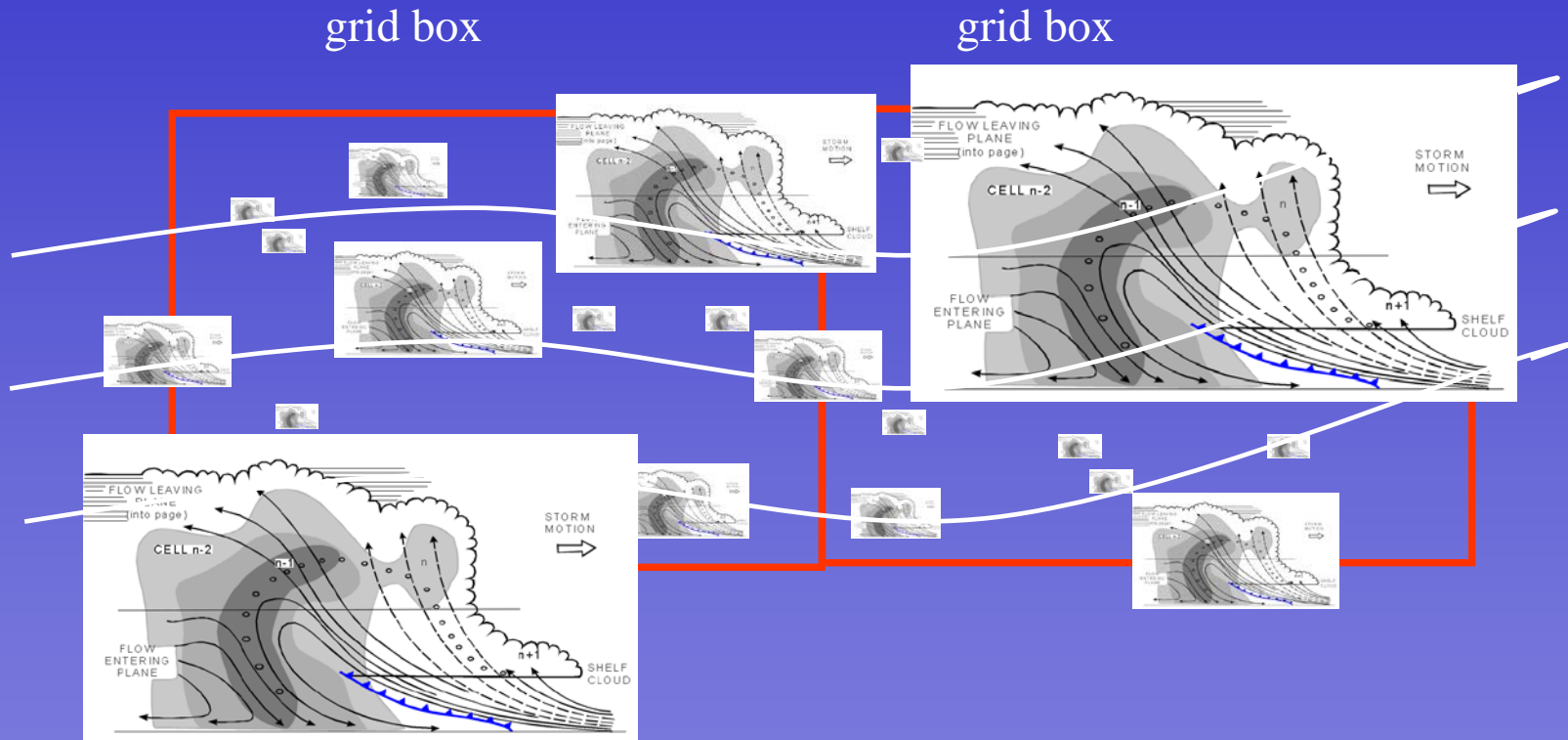


... then the Canonical Numerical Ansatz for solving the underlying PDEs would be well posed

But reality is more like this...
(Nastrom and Gage, 1985)!



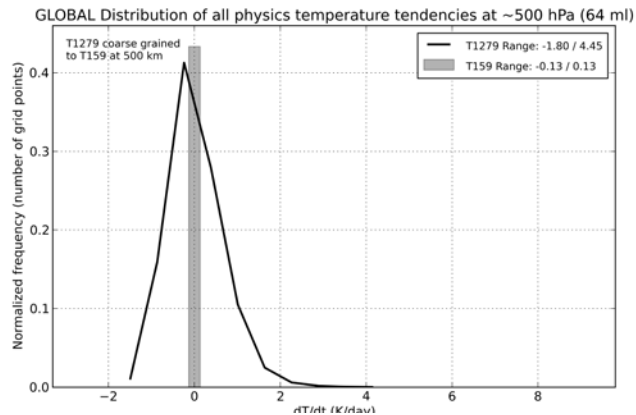
The reality of the situation



cannot be described by a simple deterministic formula

Coarse-graining (Shutts and Palmer, 2007)

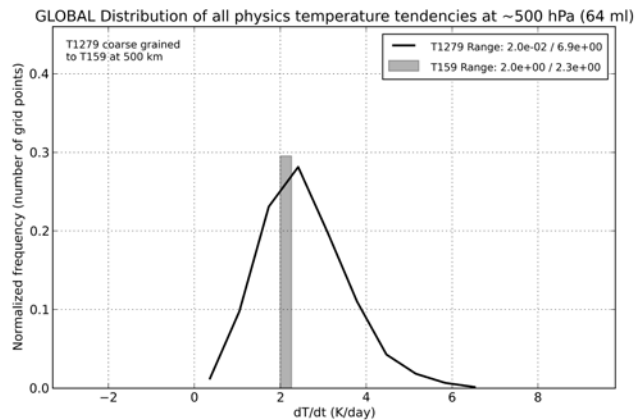
Small
tendency



Assume T1279 (16km) model = “truth”.

Assume T159 coarse-grain “model” grid.

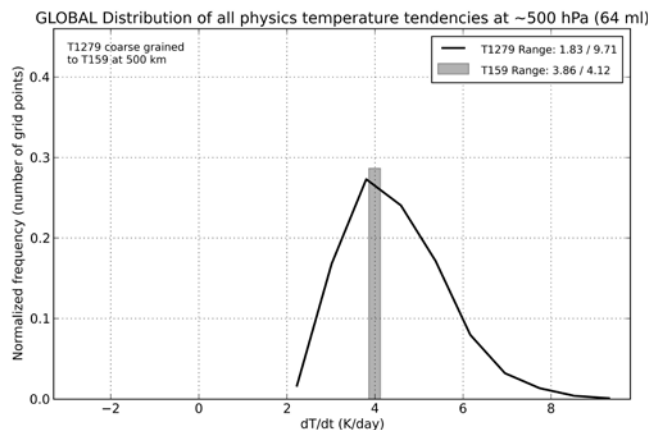
Medium
tendency



Bar= Subset of T159 total temperature parametrisation tendencies driven by T1279 coarse-grain fields.

Curve= Corresponding “true” sub-T159-scale tendency based on T1279 truth model.

Large
tendency

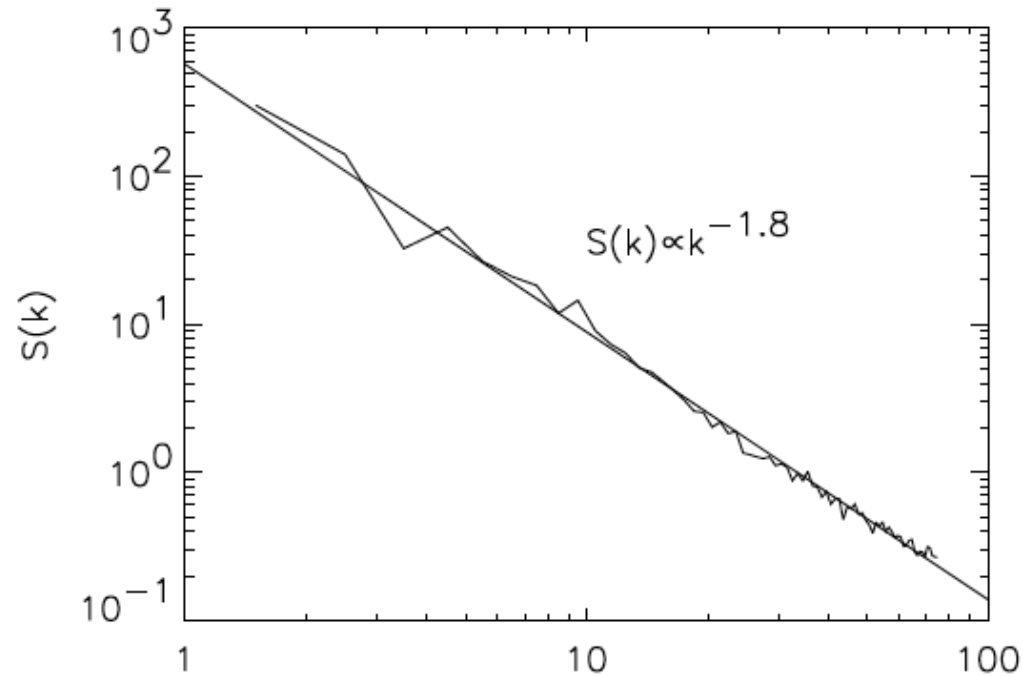


le when the parametrisations think the sub-grid pdf is a thin hat function, the reality is a much broader pdf.

The standard deviation increases with parametrised tendency – consistent with multiplicative stochasticity.

Callado-Palares and Shutts, 2013.

Earth's Topography has Power Law Structure Too

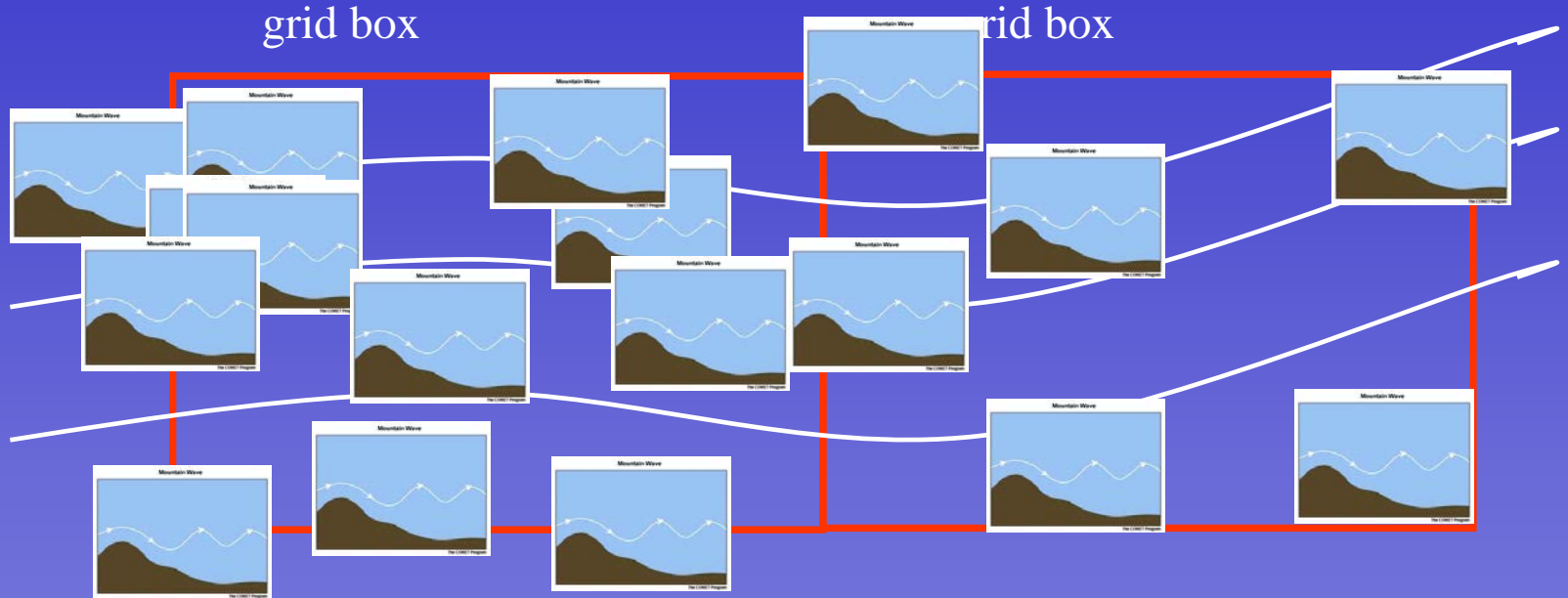


Why is topography fractal?

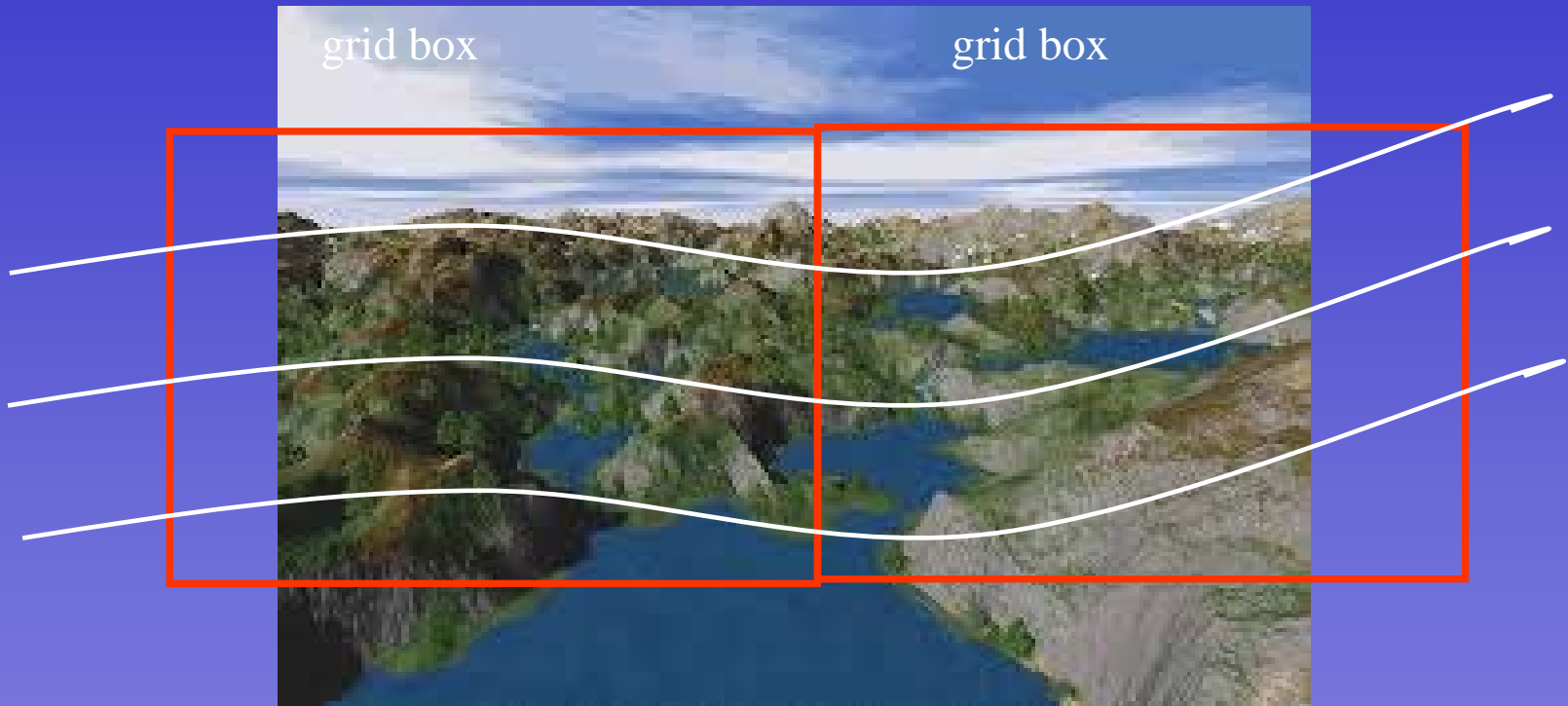
Jon D. Pelletier
Department of Geological Sciences, Snee Hall, Cornell University, Ithaca, New York

Figure 4. Average power spectrum S as a function of wave number k for one dimensional transects of the surface generated with the RSOS model. A least square fit to the logarithms of the ordinate and abscissa yield a slope of -1.81 indicating that $S(k) \propto k^{-1.81}$.

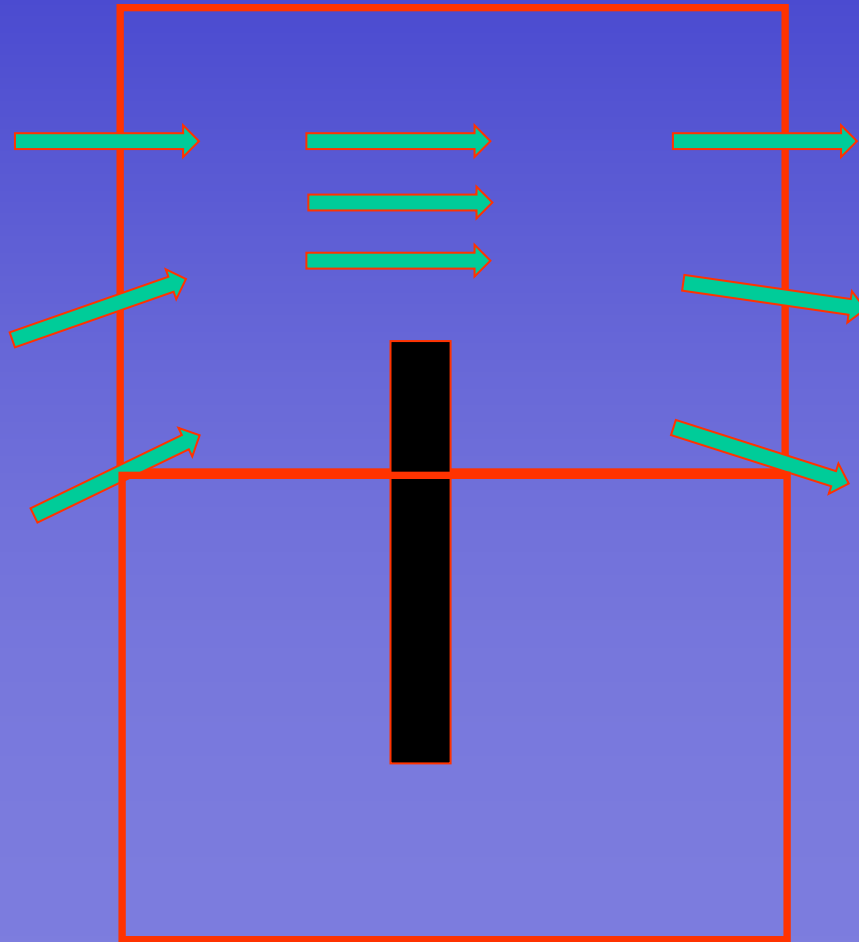
... ie not like this



... but this



grid box





From Schertzer and
Lovejoy, 1993

$$\rho \left(\frac{\partial}{\partial t} + \mathbf{u} \cdot \nabla \right) \mathbf{u} = \rho \mathbf{g} - \nabla p + \nu \nabla^2 \mathbf{u}$$

Resolved scales

Unresolved scales

Dynamical Core

$$\zeta = \sum_{m l} \zeta_{ml} e^{im\lambda} P_l^m(\phi)$$

Parametrisations

$$P(X_{\text{Tr}}; \alpha)$$

The Canonical Numerical Ansatz – ie the deterministic delineation into “resolved” and “parametrised” scales - is itself a (the?) major source of model error.



Numerics Group



Physics Group

Not my
problem!



Not my
problem!



dreamstime.com

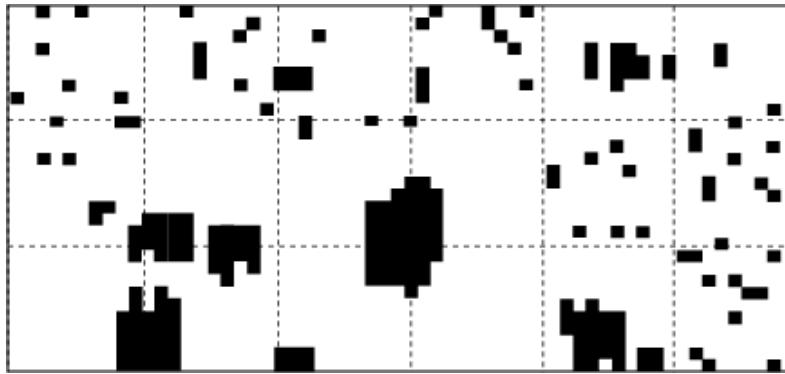
Dynamical
Core

“Physics”
Computationally cheap
stochastic-dynamic
model providing
specific realisations of
sub-grid processes



Not such a “brick wall” interface. Only makes sense
in an ensemble context. But forecasts should only
made in an ensemble context in any case!

Stochastic Cellular Automaton for Convection



Palmer 1997

Probability of an “on” cell
proportional to CAPE and
number of adjacent “on” cells
– “on” cells feedback to the
resolved flow

Stochastic Cellular Automata

A stochastic parameterization for deep convection using cellular automata

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European Centre for Medium-Range Weather Forecasts
Europäisches Zentrum für mittelfristige Wettervorhersage
Centre européen pour les prévisions météorologiques à moyen terme

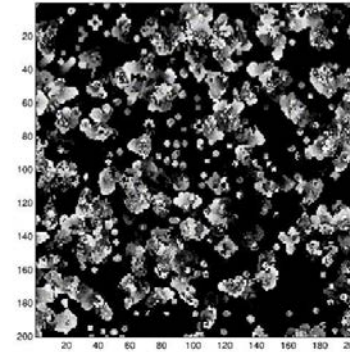


Figure 1: Example of a Cellular Automaton following the rules of Conway's game of life

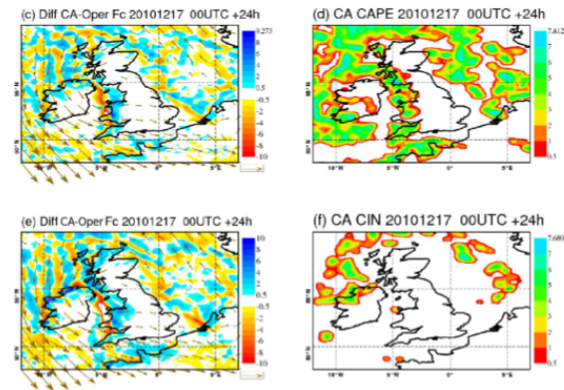
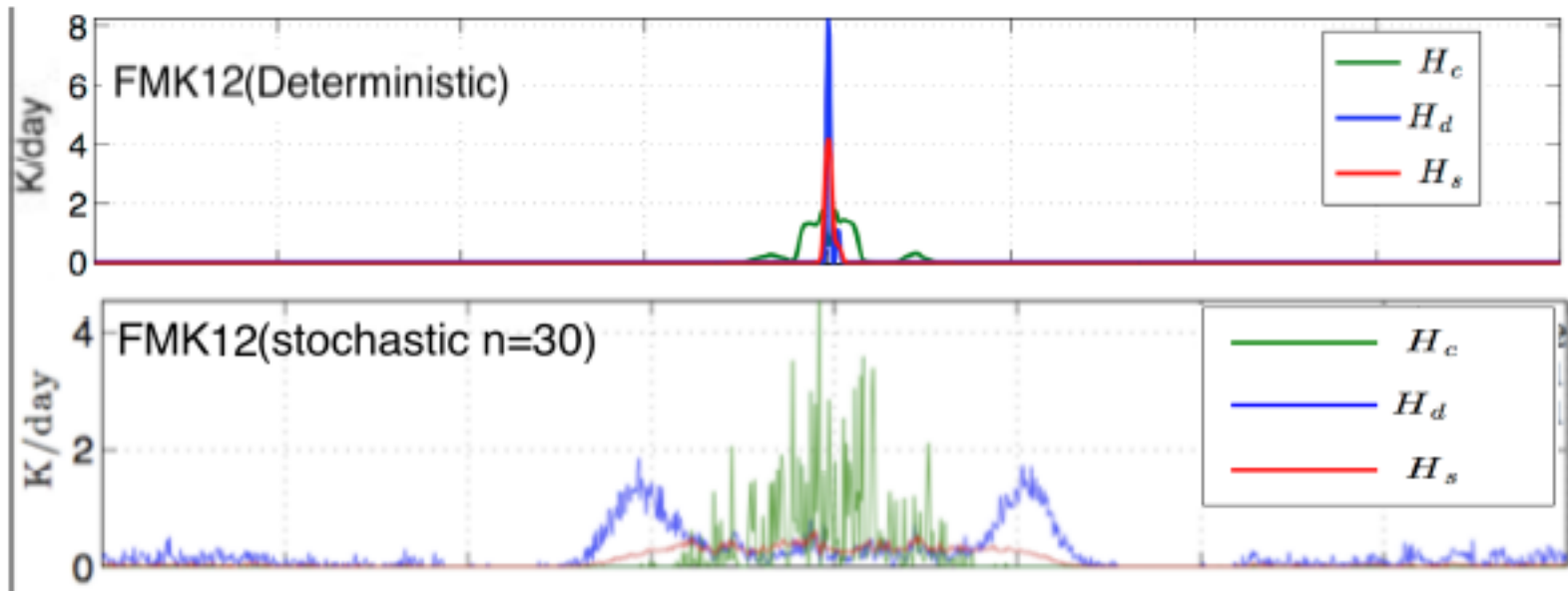


Figure 5: 24 hour accumulated precipitation (mm) on 17 December 2010 over the British Isles and Western Europe as observed by the OPERA radar network (a), the operational 24 hour deterministic IFS forecast at spectral resolution T1279=16 km (b), difference (mm) between the operational forecast and forecast using CA with CAPE seeding (c), the corresponding CA pattern for CAPE seeding (number of lives) (d), difference (mm) between the operational forecast and forecast using CA with CIN seeding (e), and corresponding CA pattern for CIN seeding (number of lives) (f)

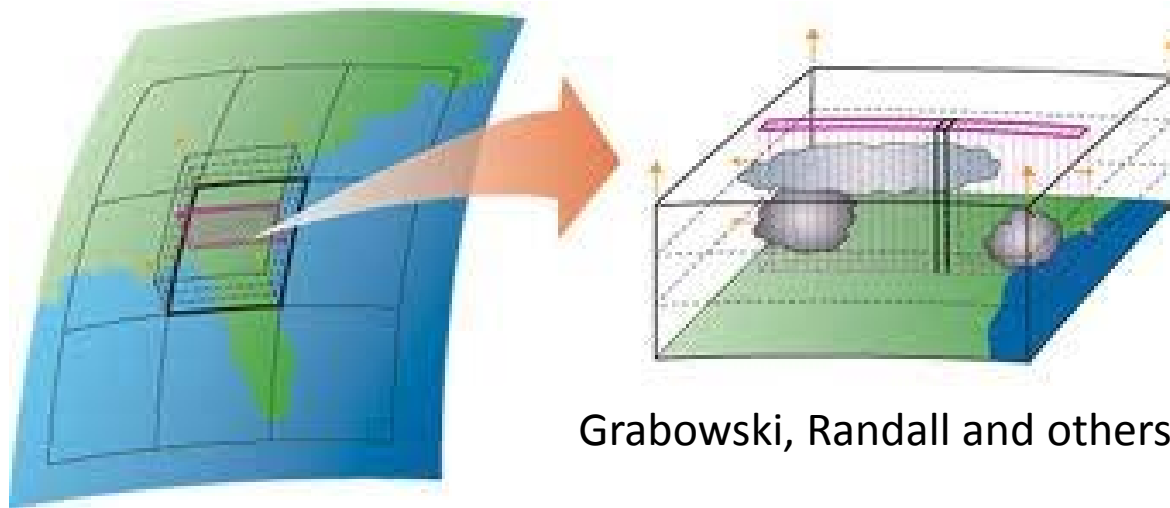
Stochastic and Deterministic Multicloud parameterizations for tropical convection

Yevgeniy Frenkel · Andrew J. Majda ·
Boualem Khouider



Stochastic multicloud model based on a Markov chain lattice model. An extension of an Ising-type spin-flip model used for phase transitions in material science

Superparameterization



Grabowski, Randall and others.

Efficient stochastic superparameterization for geophysical turbulence

Ian Grooms¹ and Andrew J. Majda¹

Department of Mathematics, and Center for Atmosphere Ocean Science, Courant Institute of Mathematical Sciences, New York University, New York, NY 10012

Contributed by Andrew J. Majda, February 7, 2013 (sent for review January 3, 2013)

Efficient computation of geophysical turbulence, such as occurs in the atmosphere and ocean, is a formidable challenge for the following reasons: the complex combination of waves, jets, and vortices; significant energetic backscatter from unresolved small scales to resolved large scales; a lack of dynamical scale separation between large and small scales; and small-scale instabilities, conditional on the large scales, which do not saturate. Nevertheless, efficient methods are needed to allow large ensemble simulations

The initial successes of SP, given the drastic simplification of the large–small coupling and of the small-scale dynamics, suggest that further computational savings might be had, without decreasing performance, by making further simplifications of the small-scale dynamics. Xing et al. (6) pursued this line of reasoning by developing sparse space–time SP algorithms using embedded domains that do not fill the spatiotemporal grid of the

Stochastic Parametrization and Model Uncertainty

Palmer, T.N., R. Buizza, F. Doblas-Reyes,
T. Jung, M. Leutbecher, G.J. Shutts,
M. Steinheimer, A. Weisheimer

Research Department

October 8, 2009

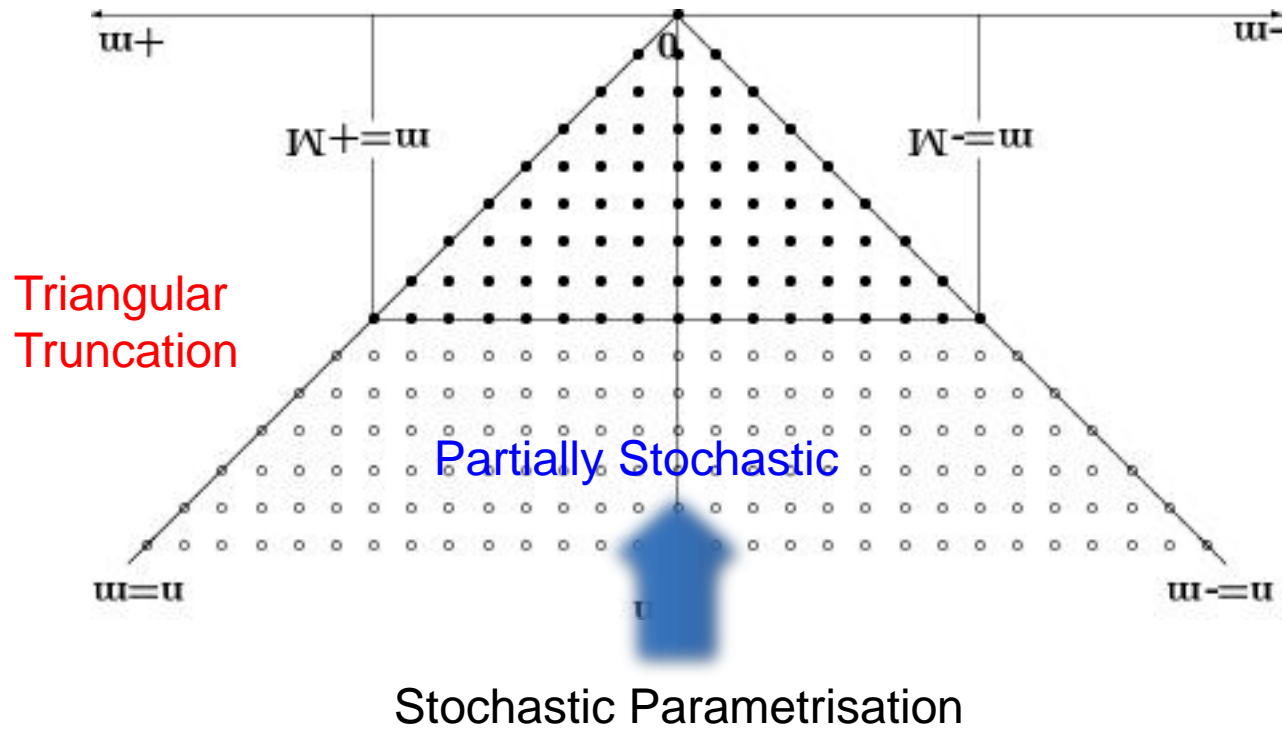
This paper has not been published and should be regarded as an Internal Report from ECMWF.
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European Centre for Medium-Range Weather Forecasts
Europäisches Zentrum für mittelfristige Wettervorhersage
Centre européen pour les prévisions météorologiques à moyen terme

- Improved forecast reliability
- Reduced systematic error

Originally based on CA
pattern generators,
now spectral.

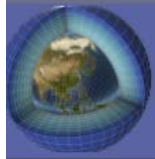


Are we “over-engineering” our dynamical cores by using double-precision bit-reproducible computations for high wavenumbers, thereby making them inefficient for evolution to high resolution?

Towards the cloud-resolved model



Infrastructure Strategy for the European Earth System Modelling Community 2012-2022



4.1 A grand challenge: Towards 1 km global resolution

A “grand challenge” for the longer term is to develop global climate models which resolve convective scale motions (nominally around 1km horizontal resolution). Although ostensibly this challenge is only about resolution, ENES believes that addressing this challenge will also support nearly all of the other scientific goals outlined earlier.

Possible for NWP by 2030? For climate change predictions, we cannot not wait that long!



Is degrading the dynamical core as we approach the truncation scale a credible route to global cloud resolution (< 1km) by 2020?

Less precise numerics, more reliable forecasts!

Floating point numbers

- Most computers follow the IEEE 754 standard

$$x = (-1)^s \cdot c \cdot b^q$$

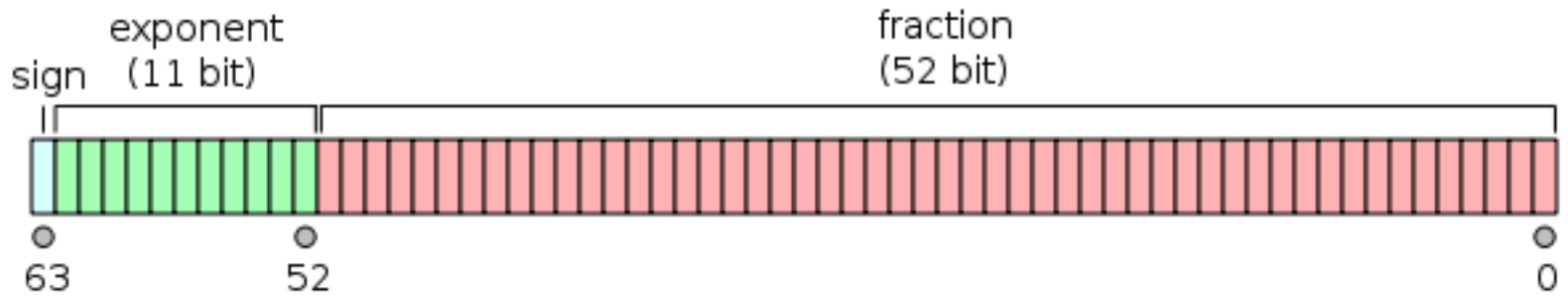
$$-12.345 = (-1)^1 \cdot 12345 \cdot 10^{-3}$$

s sign
c significand (coefficient)
b base
q exponent

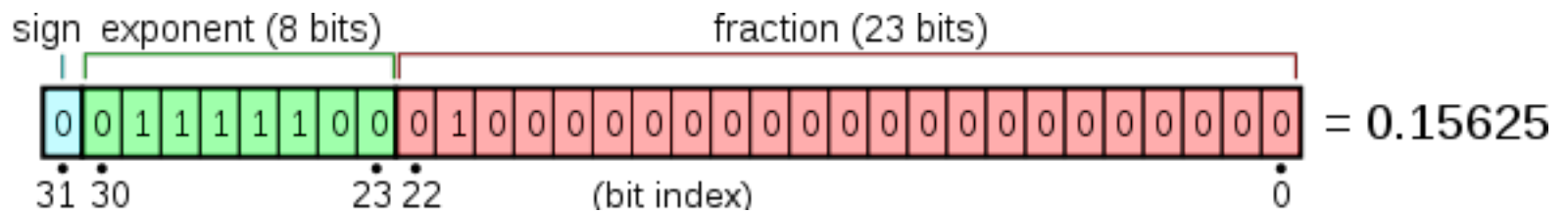
- Examples

Name	Size	Decimal digits	Minimum number	Maximum number
half precision	2 Bytes	3.3	10^{-5}	10^4
single precision	4 Bytes	7.2	10^{-38}	10^{38}
double precision	8 Bytes	16.0	10^{-308}	10^{308}
quadruple precision	16 Bytes	34.0	10^{-4932}	10^{4932}

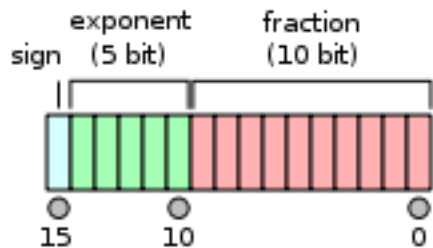
NB. 1/32 precision = 1 bit = on/off (a cellular automaton)!!



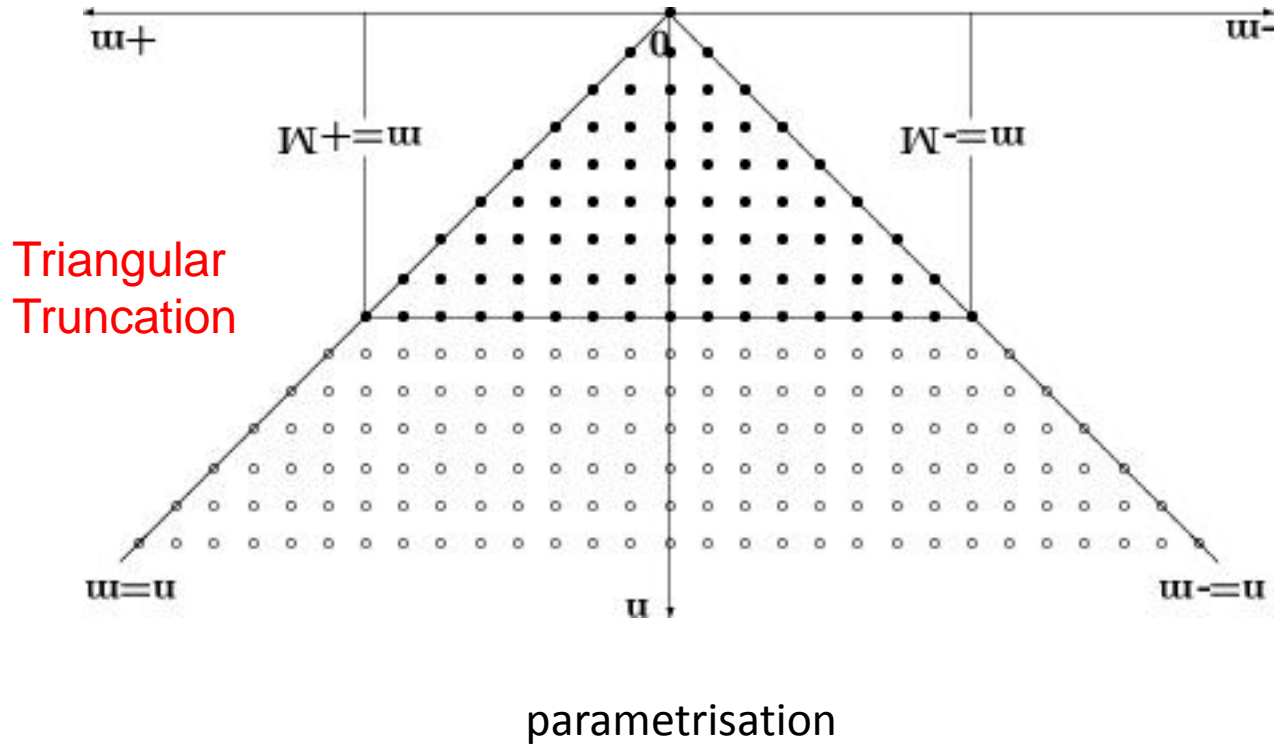
Double Precision



Single Precision



Half Precision



Triangular
Truncation



Reduced Precision
arithmetic

Motivation

- Move less information

```
real(kind=8) :: a    ! I am 8 Bytes  
real(kind=4) :: b    ! I am 4 Bytes
```

- Fit more information into cache
- Lower precision arithmetic is faster

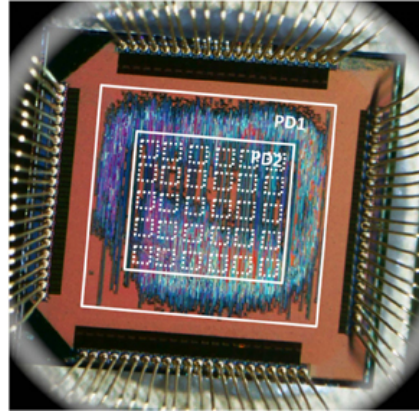
```
a = a+a-a*a*a    ! Wow, time flies!  
b = b+b-b*b*b    ! That was fast!
```

Superefficient inexact chips

<http://news.rice.edu/2012/05/17/computing-experts-unveil-superefficient-inexact-chip/>



Krishna Palem.
Rice, NTU
Singapore



In terms of speed, energy consumption and size, inexact computer chips like this prototype, are about 15 times more efficient than today's microchips.

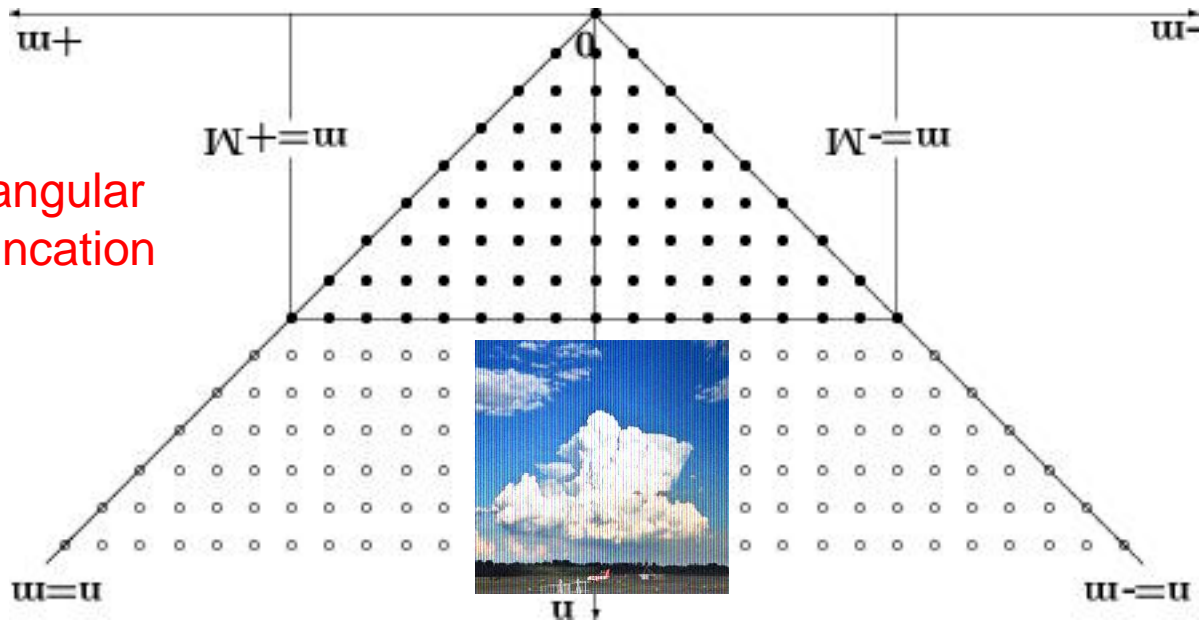


This comparison shows frames produced with video-processing software on traditional processing elements (left), inexact processing hardware with a relative error of 0.54 percent (middle) and with a relative error of 7.58 percent (right). The inexact chips are smaller, faster and consume less energy. The chip that produced the frame with the most errors (right) is about 15 times more efficient in terms of speed, space and energy than the chip that produced the pristine image (left).

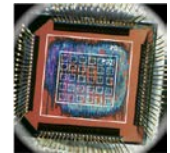


Towards the Stochastic Dynamical Core?

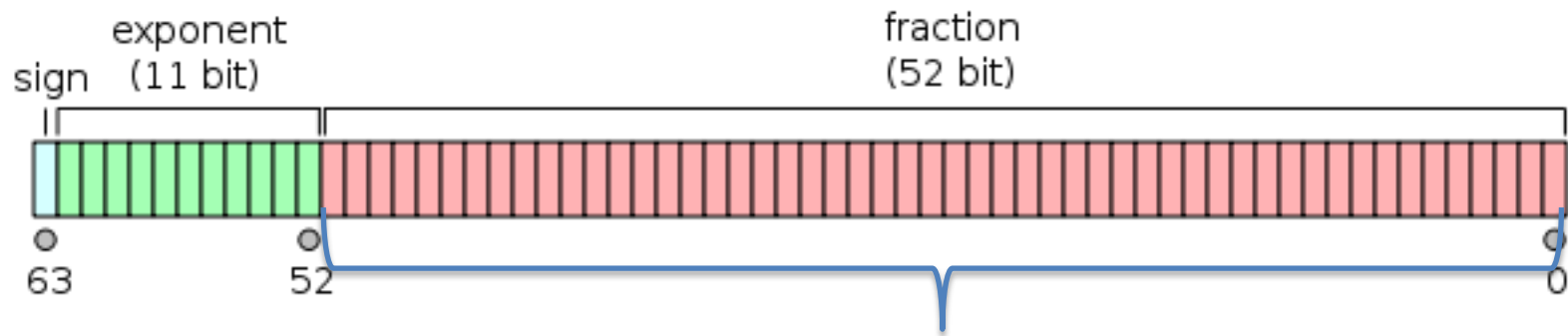
Triangular Truncation



Inexactness of chip



Emulator of Stochastic Chip



10% probability of bit flip = 90% reduction in power consumption by chips



Experiments with the Lorenz '96 System (i)

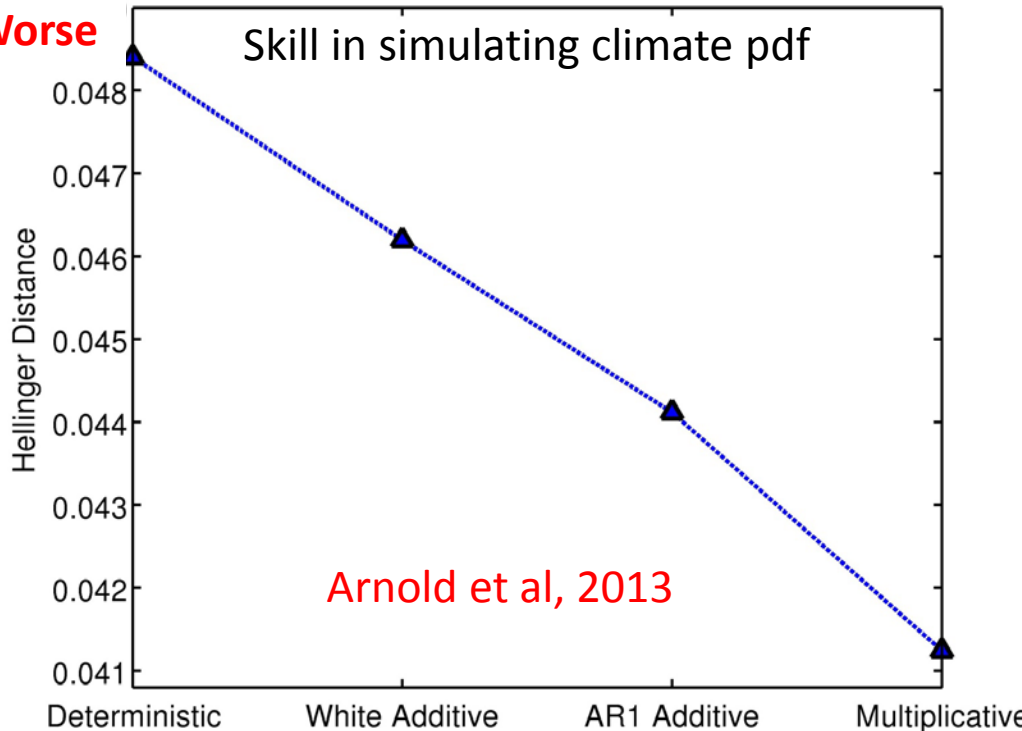
$$\frac{dX_k}{dt} = -X_{k-1} (X_{k-2} - X_{k+1}) - X_k + F - \frac{hc}{b} \sum_{j=J(k-1)+k}^{kJ} Y_j$$

$$\frac{dY_j}{dt} = -cbY_{j+1} (Y_{j+2} - Y_{j-1}) - cY_j + \frac{hc}{b} X_{\text{int}[(j-1)/J+1]}$$

Assume Y unresolved

Approximate sub-grid tendency by U

Worse Skill in simulating climate pdf



Deterministic: $U = U_{\text{det}}$
 Additive: $U = U_{\text{det}} + e_{w,r}$
 Multiplicative: $U = (1+e_r) U_{\text{det}}$

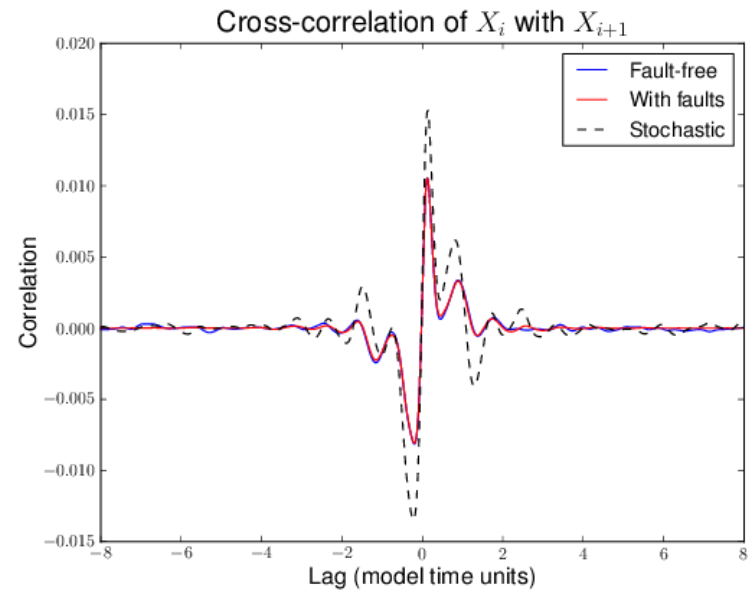
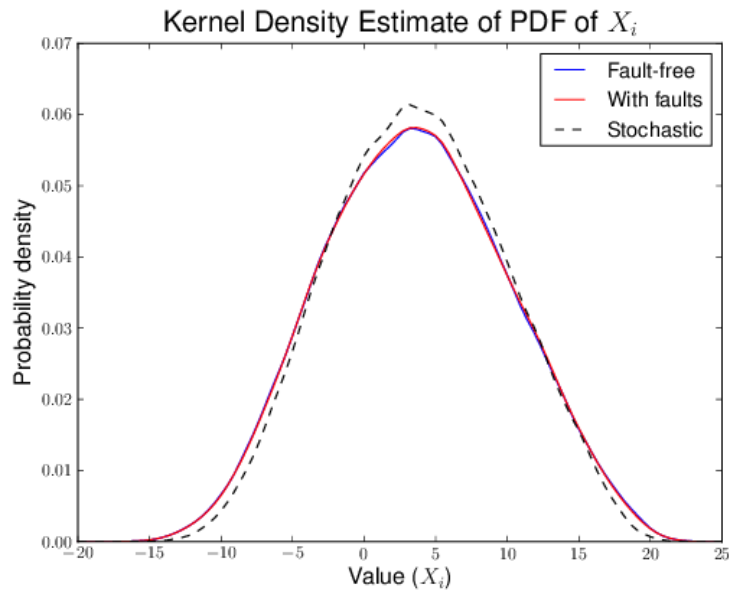
Where:

U_{det} = cubic polynomial in X

$e_{w,r}$ = white / red noise

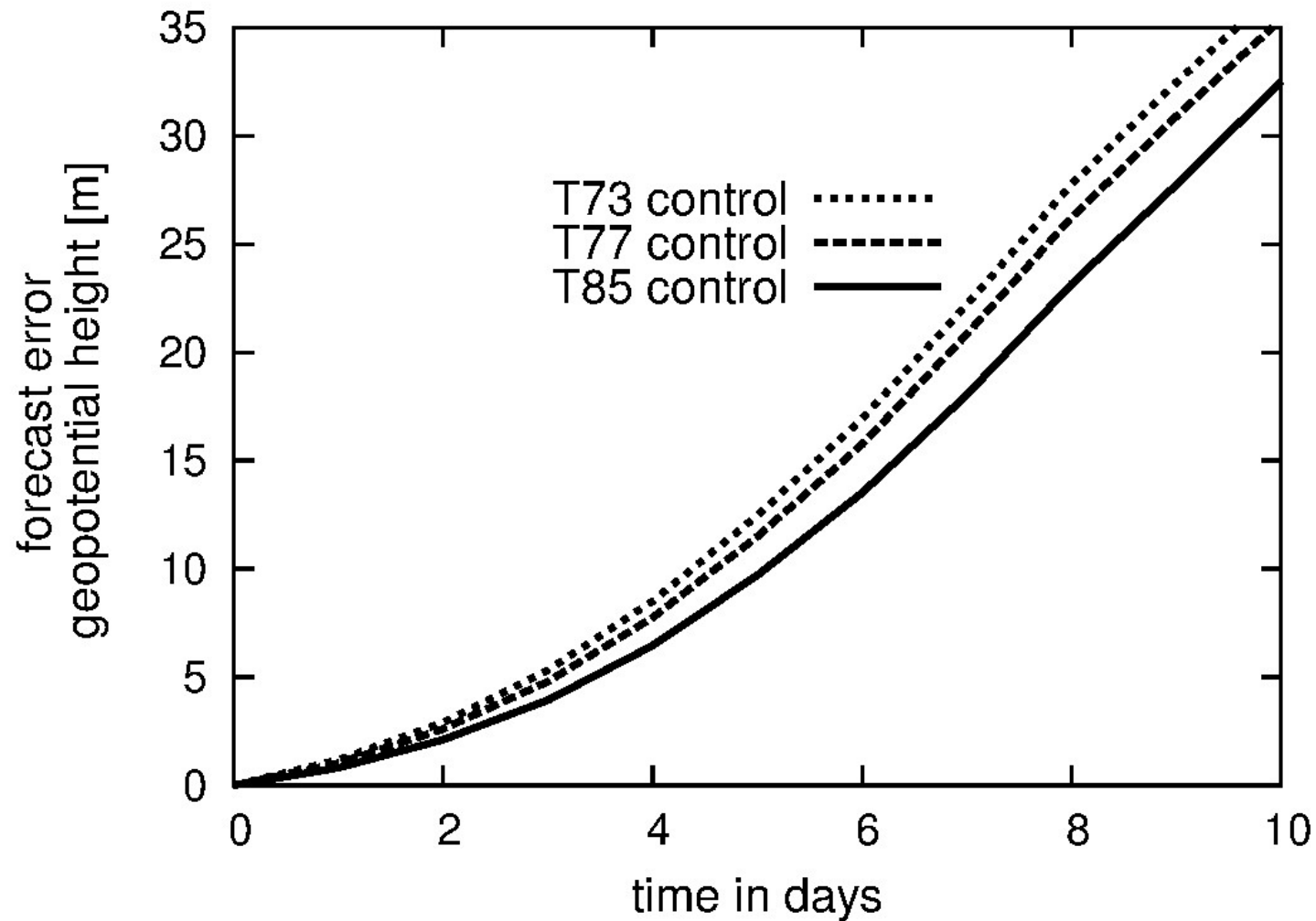
Fit parameters from full model

20% fault rate on Y variables



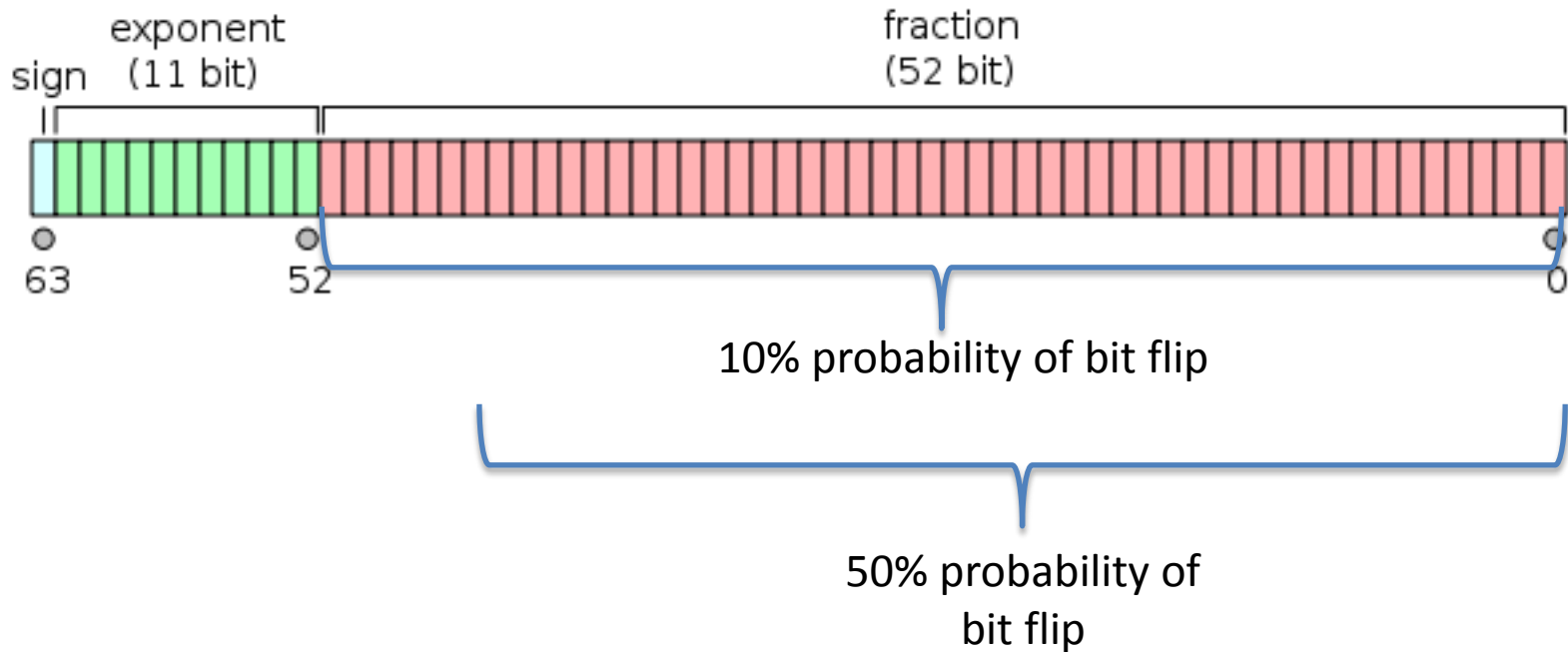
Imprecise L96 is more accurate than
parametrised L96

Weather forecasts with imprecise processing



Truth = T159 integration. 500hPa Geopotential height rms error.

Emulator of Stochastic Chip/Reduced Precision on T85 spectral model



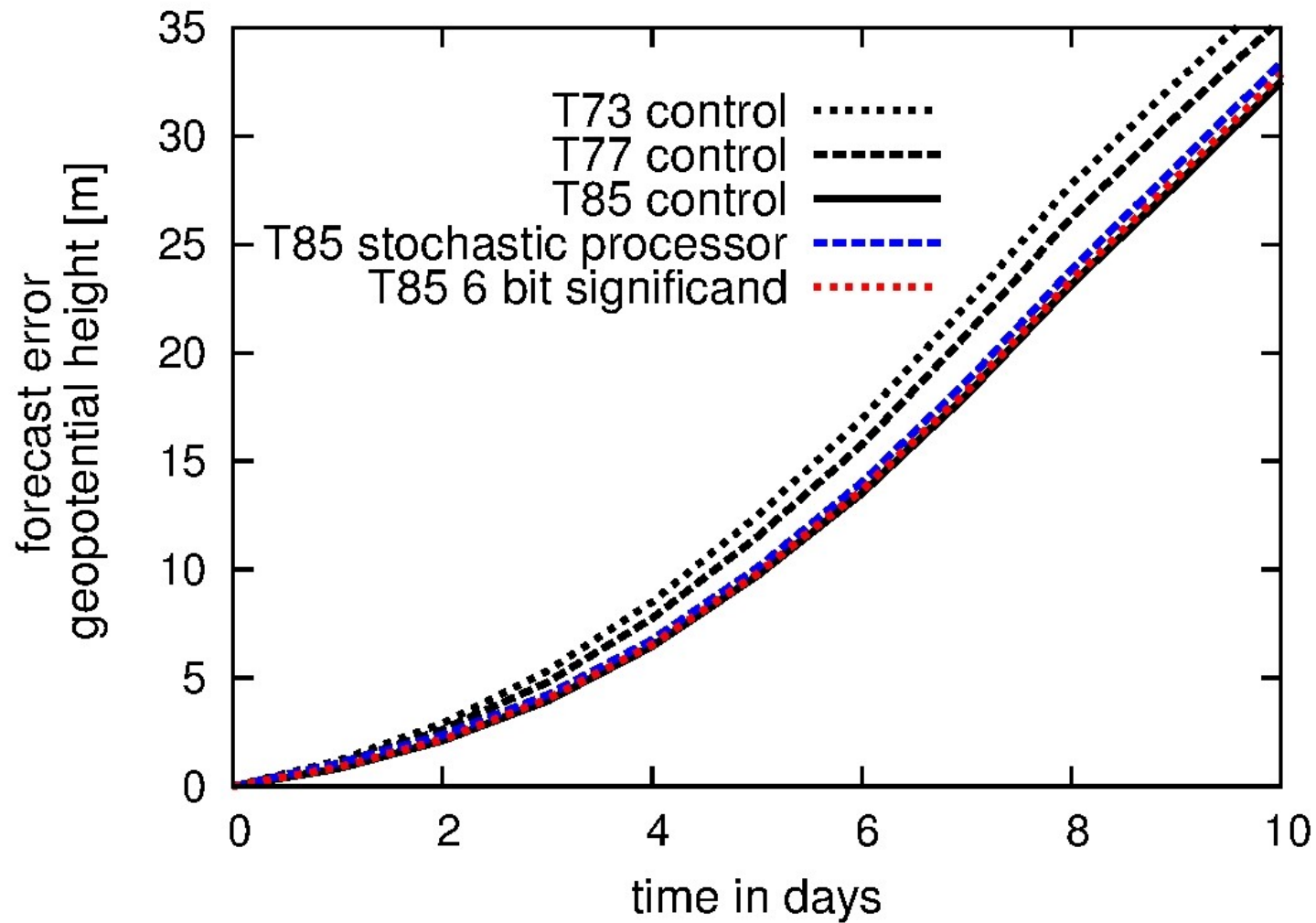
The emulator is used on 50% of numerical workload:

All floating point operations in grid point space

All floating point operations in the Legendre transforms between wavenumbers 31 and 85.

Cost approx that of T73

Weather forecasts with imprecise processing

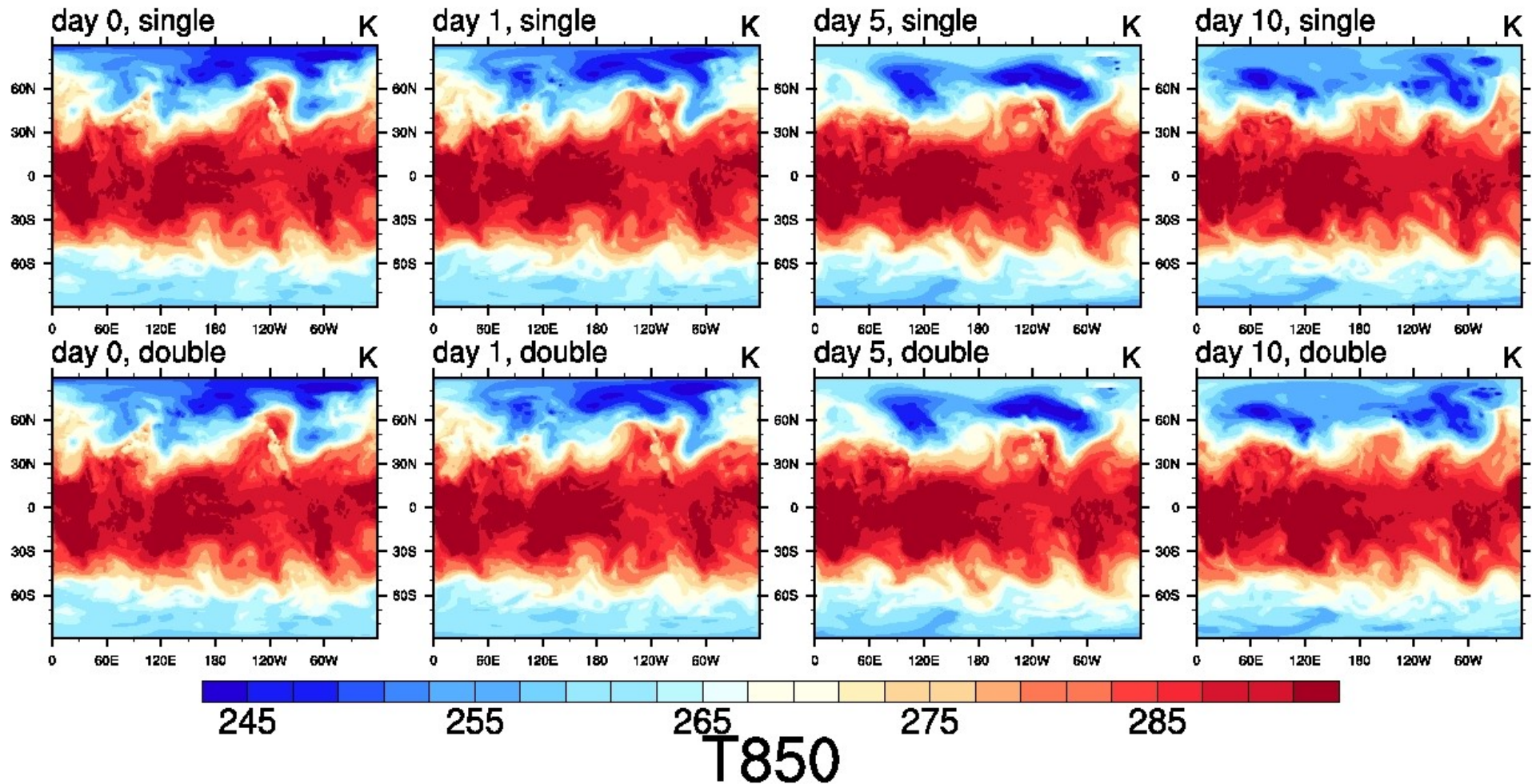


Would the IFS work in single precision?

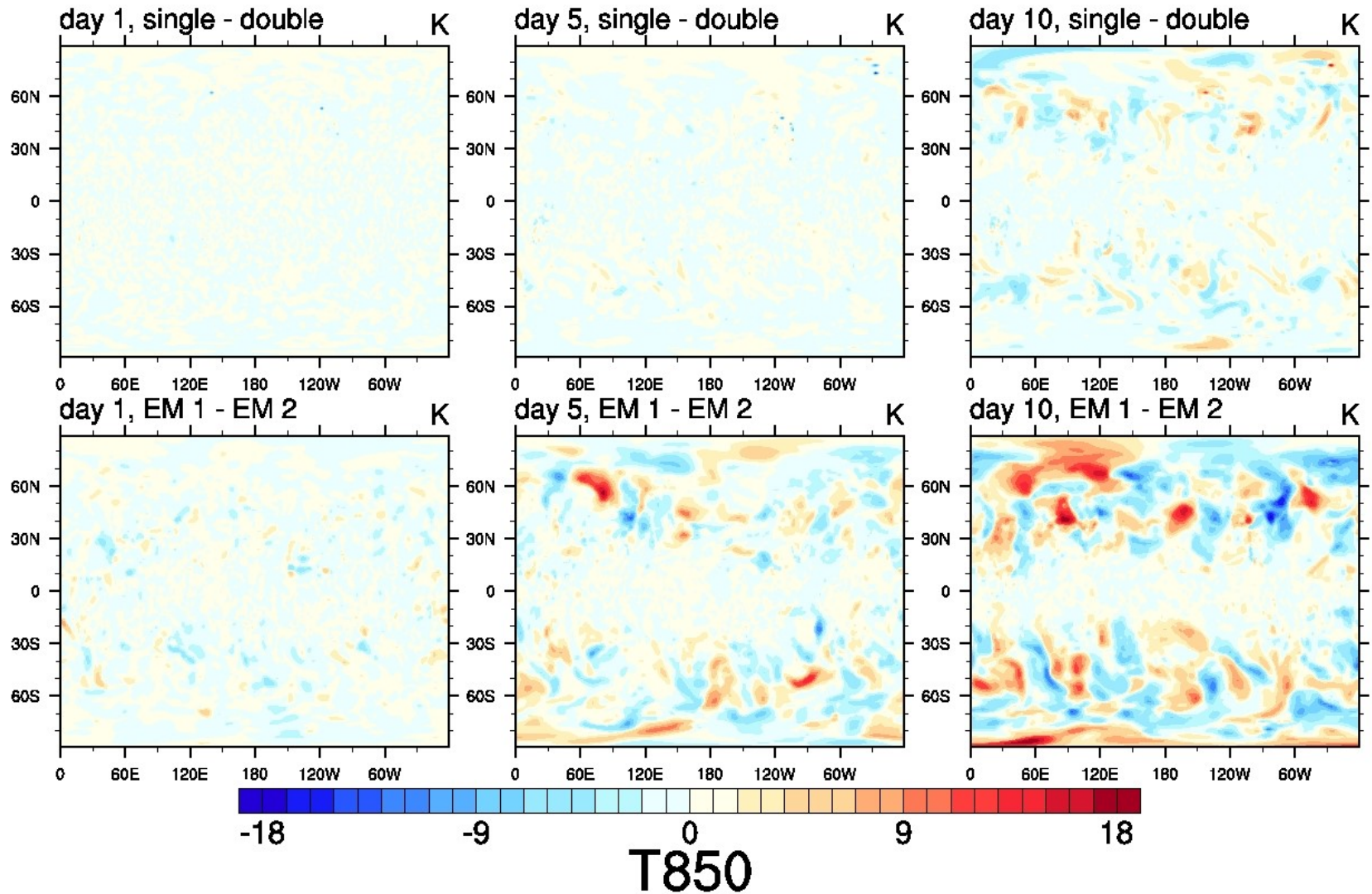
Approach:

- Using OpenIFS, (nearly) all of the double precision numbers have been replaced by single precision floating point numbers.
- We perform a weather forecast at T159 resolution with double and with single precision and compare the results.

Would the IFS work in single precision?



Would the IFS work in single precision?



Top row: Differences between the double and the single precision simulation.

Bottom row: Differences between two ensemble members for a T159 IFS forecast with SPPT

Could we run leg2 EPS
at higher resolution
using single precision
arithmetic?

In a presentation at ECMWF on **Challenges in Application Scaling in an Exascale Environment**, IBM's Chief Engineer for HPC, Don Grice, noted that:

“Increasingly there will be a tension between energy efficiency and error detection”,

and asked whether :

“...there needs to be a new software construct which identifies critical sections of code where the right answer must be produced”

([http://www.ecmwf.int/newsevents/meetings/workshops/2010/high performance computing 14th/index.html](http://www.ecmwf.int/newsevents/meetings/workshops/2010/high_performance_computing_14th/index.html))

In the context of NWP/Climate models

- Which parts of the code need to be precise and which parts not?
- Where can we drop the need for precise determinism?
- Is a discriminating approach to the use of precision/imprecision, determinism/stochasticity, a credible route for evolution to ultra-high resolution (eg <1km) – and hence more reliable weather and climate forecasts - by 2020?

20 Years Ago

Dynamics

Parametrisation

$O(100\text{km})$

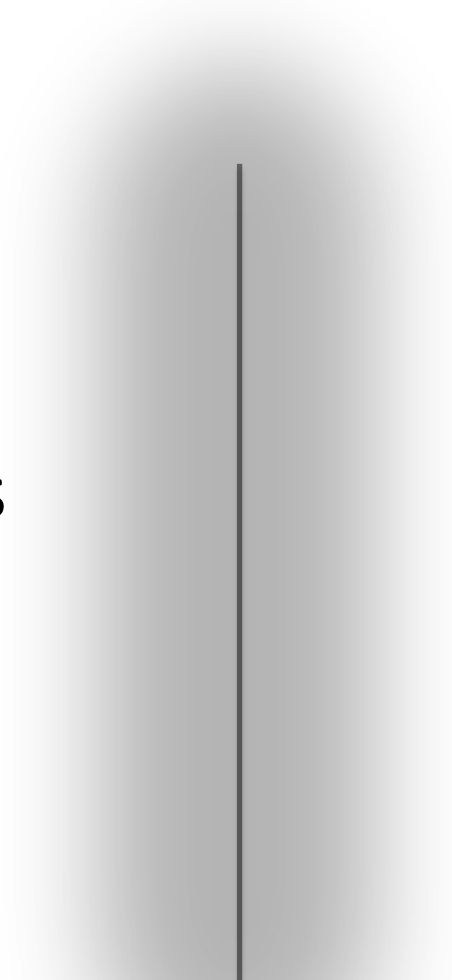
A diagram consisting of a vertical black line. To the left of the line is the word 'Dynamics' and to the right is the word 'Parametrisation'. Below the line is the text 'O(100km)'. The text '20 Years Ago' is positioned at the top center of the image.

Now

Dynamics

Parametrisation

$O(10\text{km})$



By 2020?

Dynamics

Parametrisation

$O(1\text{km})$