

Quantifying EPS forecast skill and multiscale predictability of hydrometeorological fields

Daniel SCHERTZER ¹

Shaun LOVEJOY ²

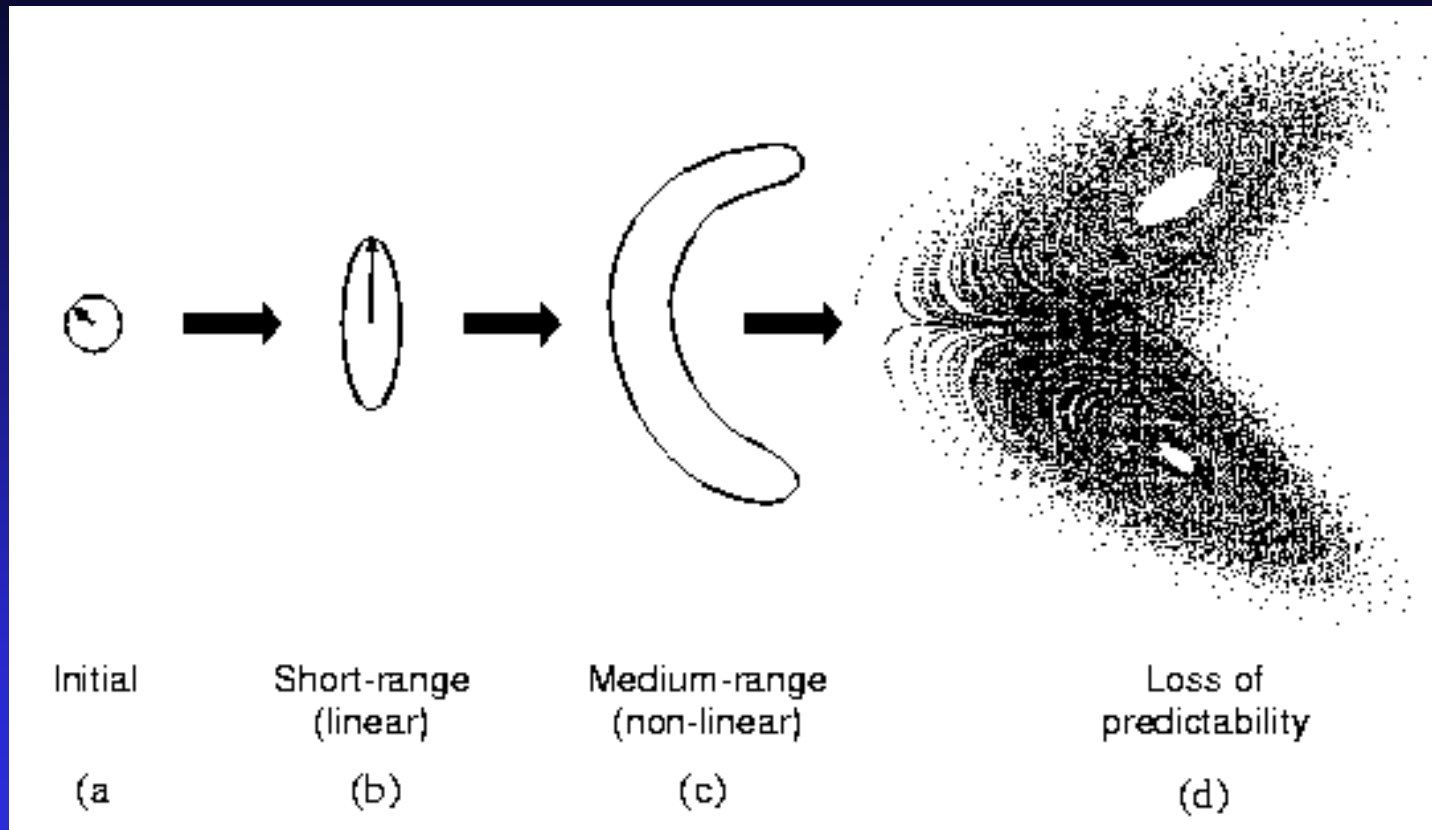
Ioulia TCHIGUIRINSKAIA ³

¹CEREVE, ENPC&Météo-France

² McGill U.

³ Sisyphe-U. Paris VI

EPS and lessons from chaos?

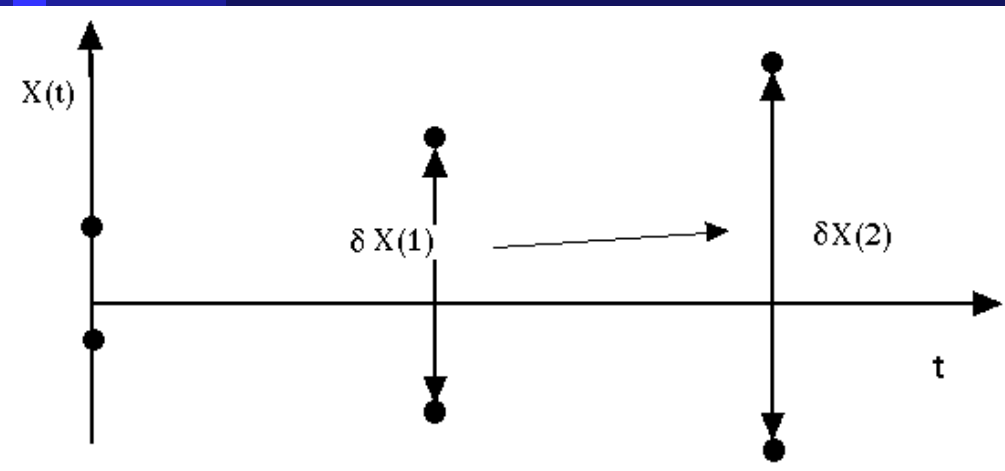


Scheme of the evolution of the empirical pdf in EPS

Phase space (according to Palmer 1999): initial ensemble (a), to (b) linear growth phase, to (c) nonlinear growth phase, to (d) loss of predictability.

Exponential error growth: $|\delta X(t)| \approx e^{\mu t} |\delta X(0)|$

■ Heuristics:



$$X(t) = G(X(t-1)) \quad X \in R$$



$$|\delta X(t)| \approx |D_{X(t-1)} G| |\delta X(t-1)|$$



$$\text{Log} [|\delta X(t)| / |\delta X(0)|] \approx \sum_{t'=0, t-1} \text{Log} (|D_{X(t')} G|) \approx t \langle |D_x G| \rangle = t \mu$$

■ Multiplicative Ergodic Theorem (Oseledec, 1968):

- if and only if:

A being the generator of the flow

- limited extension for infinite dimensions.

$$\mu = \int \text{Log}^+ (\|A(x)\|) d\rho(x) < \infty$$

EPS and Fokker-Planck equation ?

On a **finite** d-dimensional phase space

$$\{X_1, X_2, \dots, X_d\}$$

$$\dot{\underline{X}}(t) = \frac{d}{dt} \underline{X} = \underline{F}(\underline{X}, t) \Rightarrow$$

$$\frac{\partial}{\partial t} \rho(\underline{X}, t) + \sum_{i=1}^d \frac{\partial}{\partial X_i} [\dot{X}_i(t) \rho(\underline{X}, t)] = 0$$

for the pdf

$$\rho(\underline{X}, t)$$

w.r.t.

$$dX_1 dX_2 \dots dX_d$$

But noisy perturbations due to subgrid scales

$$\frac{d}{dt} \underline{X} = \underline{F}(\underline{X}, t) + \underline{f}(t)$$

If gaussian:

$$\langle f_i(t) f_j(t') \rangle = \varepsilon \delta_{i,j} \delta(t - t') \Rightarrow \frac{\partial}{\partial t} \rho(\underline{X}, t) + \sum_{i=1}^d \frac{\partial}{\partial X_i} [\dot{X}_i(t) \rho(\underline{X}, t)] - \varepsilon \Delta_X \rho(\underline{X}, t) = 0$$

If non-gaussian, e.g. Lévy type:

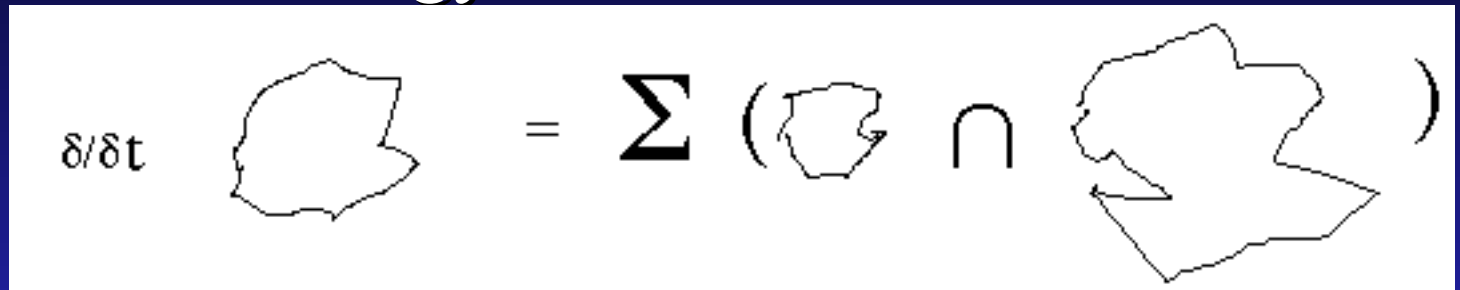
$$-\Delta_X \rightarrow (-\Delta_X)^\alpha$$

(+other fractional operators)

Power-law decorrelation:

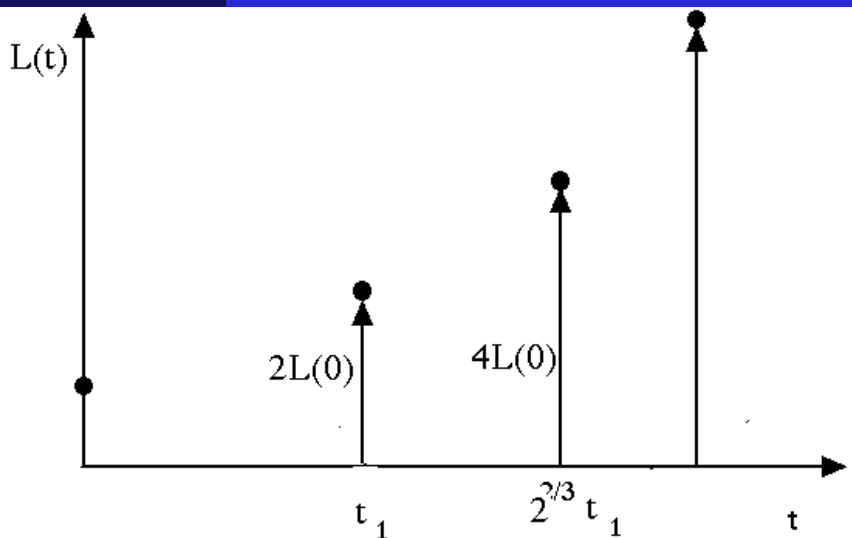
$$l \approx t^{3/2}$$

■ Phenomenology:

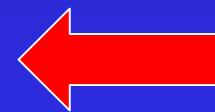


by nonlinear interactions, uncertainties grow up to larger scales

■ Heuristics :



$$l \approx t^{3/2}$$



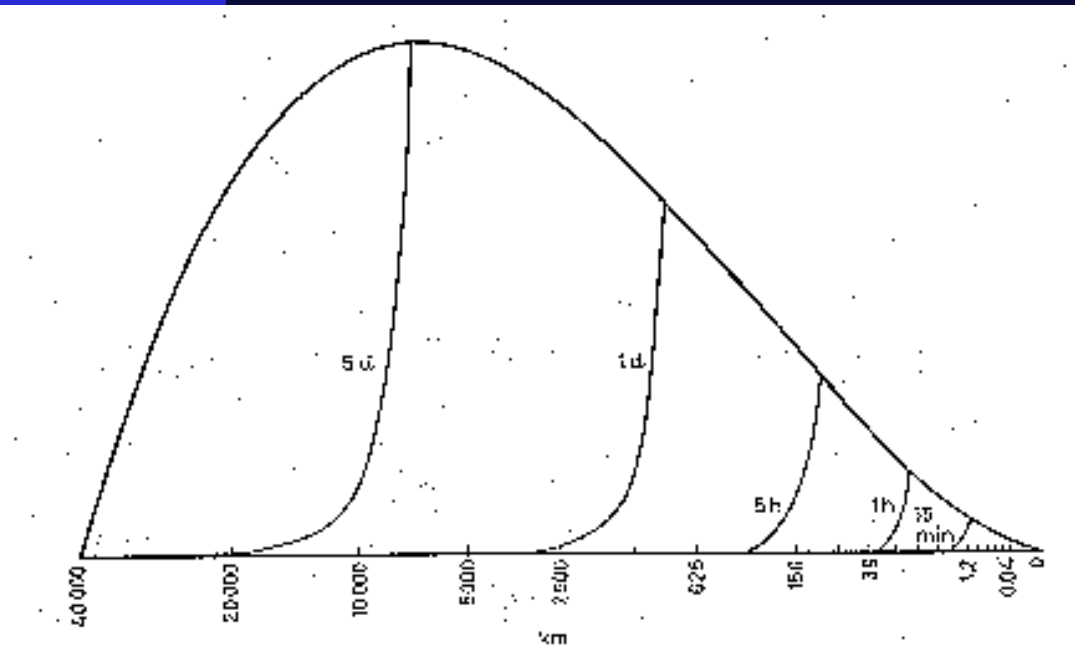
$$t \approx l / \delta u(l)$$

eddy turn-over
time

$$\varepsilon \approx \delta u(l)^3 / l$$

scale invariance of
the energy flux (K41)

Spectral analysis and closures



$$l_c \approx t^{3/2}$$

Lorenz (1969,
Leith and Kraichnan(1972),
Metais and Lesieur (1986)

Flux from correlated energy

$$e^c(\underline{x}, t) = \frac{1}{2} \underline{u}^2(\underline{x}, t) \cdot \underline{u}^1(\underline{x}, t)$$

to decorrelated energy

$$e^\Delta(\underline{x}, t) = \frac{1}{2} \left(\underline{u}^2(\underline{x}, t) - \underline{u}^1(\underline{x}, t) \right)^2$$

Similar results with:

$$l_c^{2/3} = \bar{\varepsilon}^{-1/3} t^{3/2}; \quad \bar{\varepsilon} = 10^{-3} \text{ m}^2 \text{ s}^{-3}, \quad \eta \approx 10^{-3} \text{ m};$$

Questioning the Lorenz-Leith-Kraichnan approach

- 2 fundamental comments (Lilly, 1985):
 - ◆ appraisal of Lorenz 's choice of 3D energy spectrum up to synoptic scales, initially considered as rather paradoxical:
what is the dimension of atmospheric dynamics?
 - ◆ strong criticism of the homogeneity and quasi-normal hypotheses of closures:
what are the statistics?

Richardson Diffusion law

Richardson (1926) found a scaling turbulent viscosity **up to planetary scales**:

- $\nu(L) \approx \varepsilon^{1/3} L^{4/3}$

associated to an anomalous diffusion:

- $\langle r(t)^2 \rangle \approx \varepsilon t^3$

the first quantitative law in turbulence..

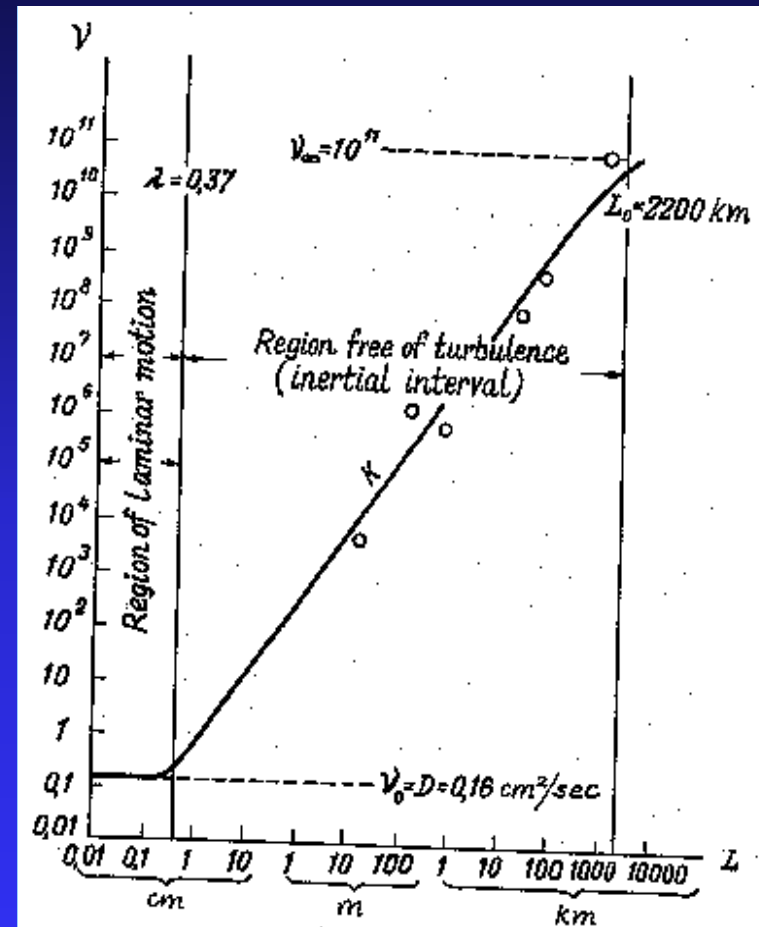
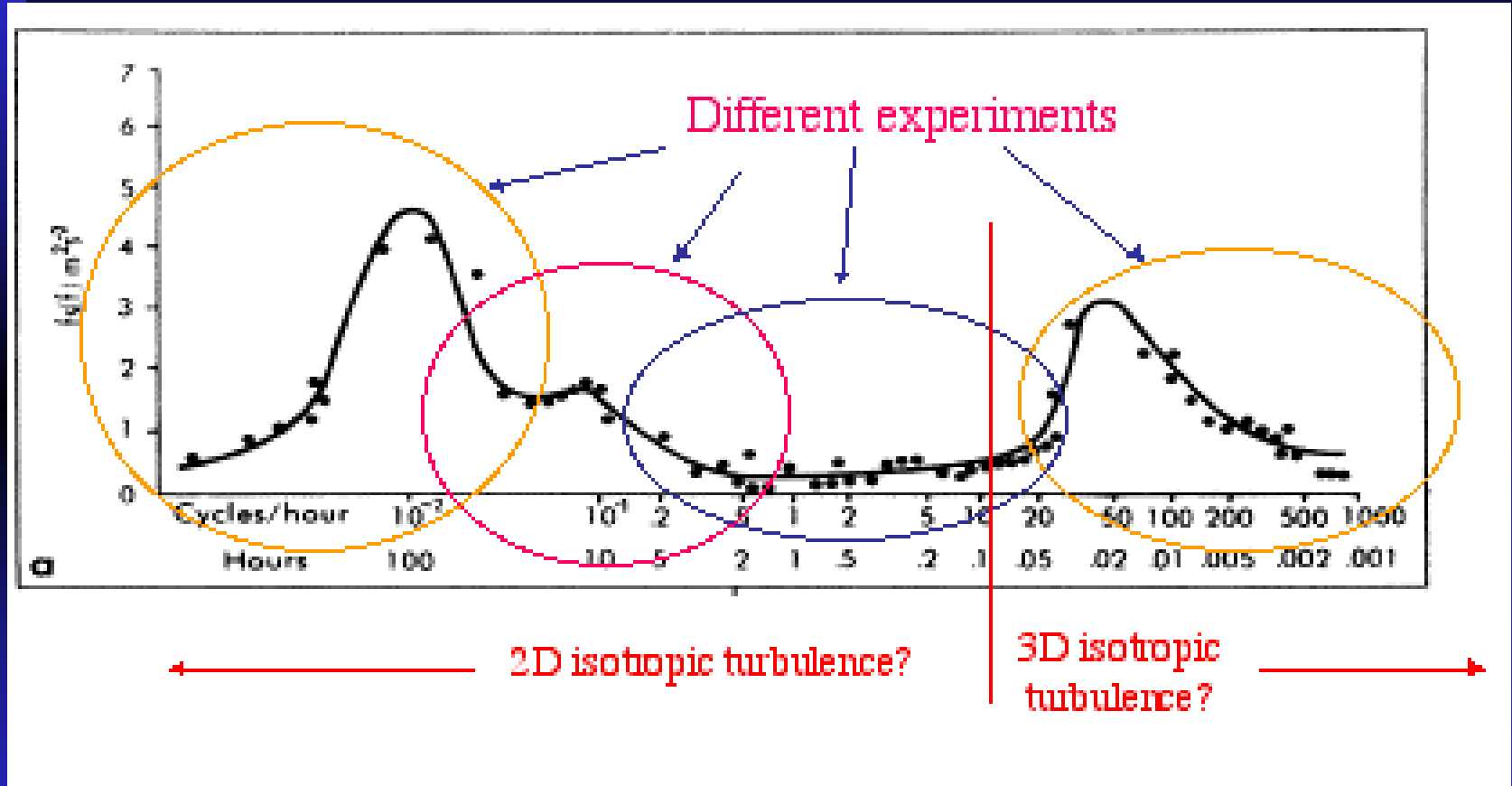


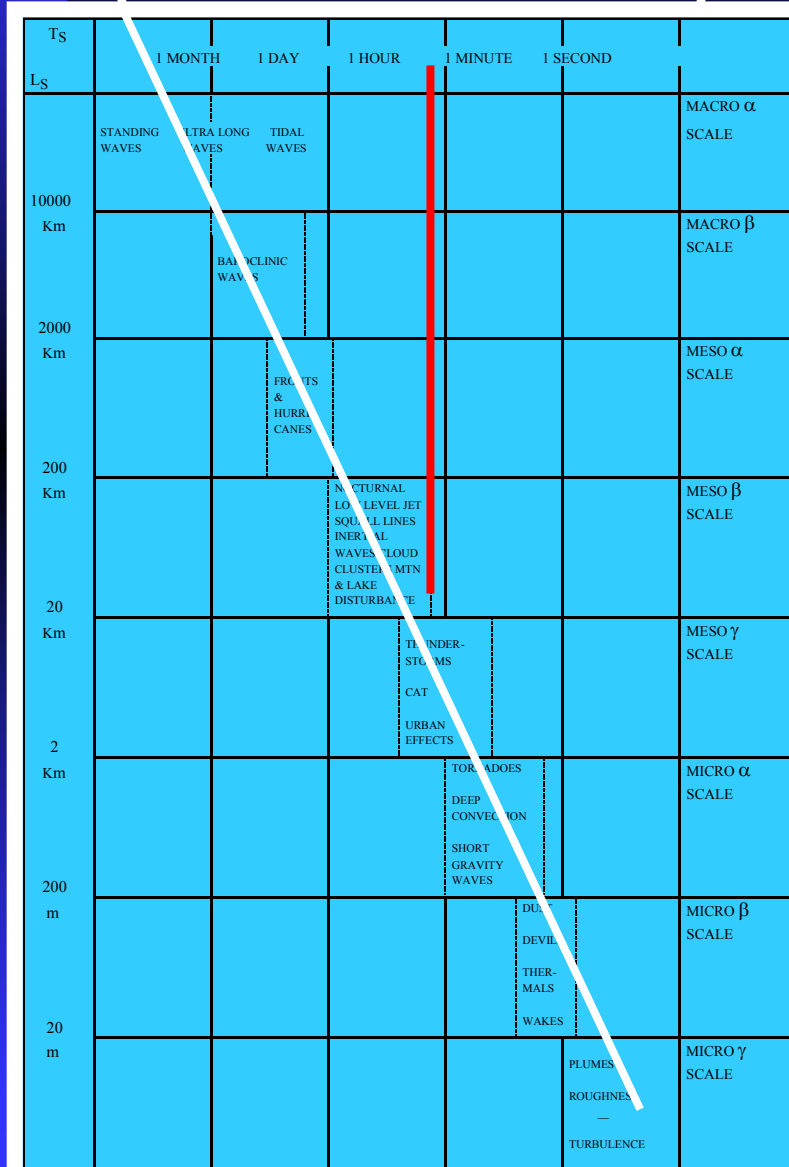
Fig. 1 The vertical diffusion coefficient $\nu(L)$ as a function of the turbulence scale L . Empirical points after Richardson.¹⁹

Van der Hoven wind spectrum (1957)



Richardson cascade is split into macro, meso, micro oscillations...

Space-Time ("Stommel") diagram



Schematic diagram showing a typical phenomenologist's view of meteorology, reproduced from Atkinson (1981) who adapted it from Orlandi (1975).

The added straight line corresponds to :

$$T \approx \varepsilon^{-1/3} L^{2/3}$$

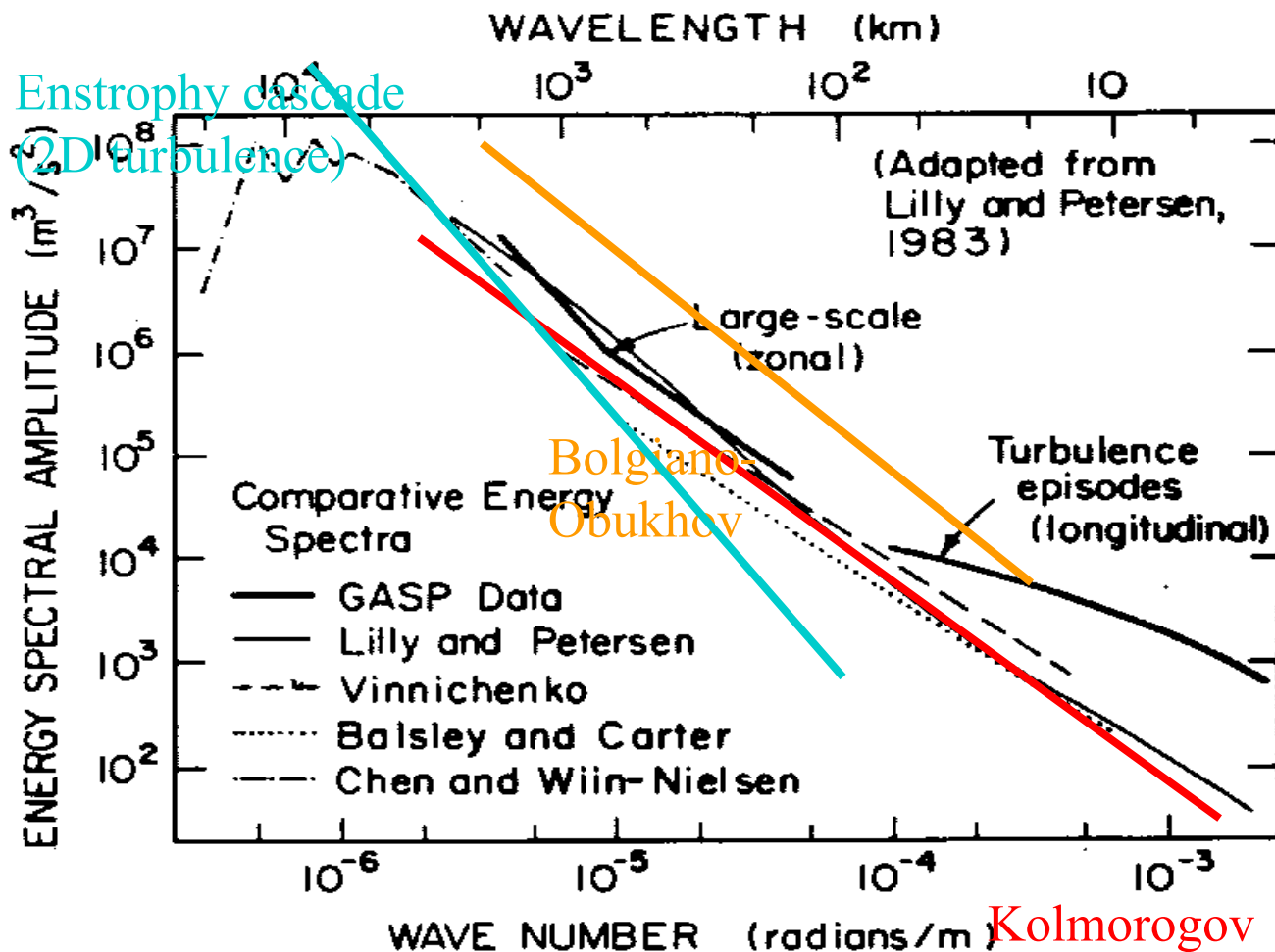
in agreement with Richardson (1926) and Kolmogorov (1941), i.e.

3D-like behavior

Schertzer et al., Fractals, 1997

The vertical line corresponds to a **2D-like large scale behavior**

Wind: the GASP Experiment



Adapted from Nastrom and Gage 1983

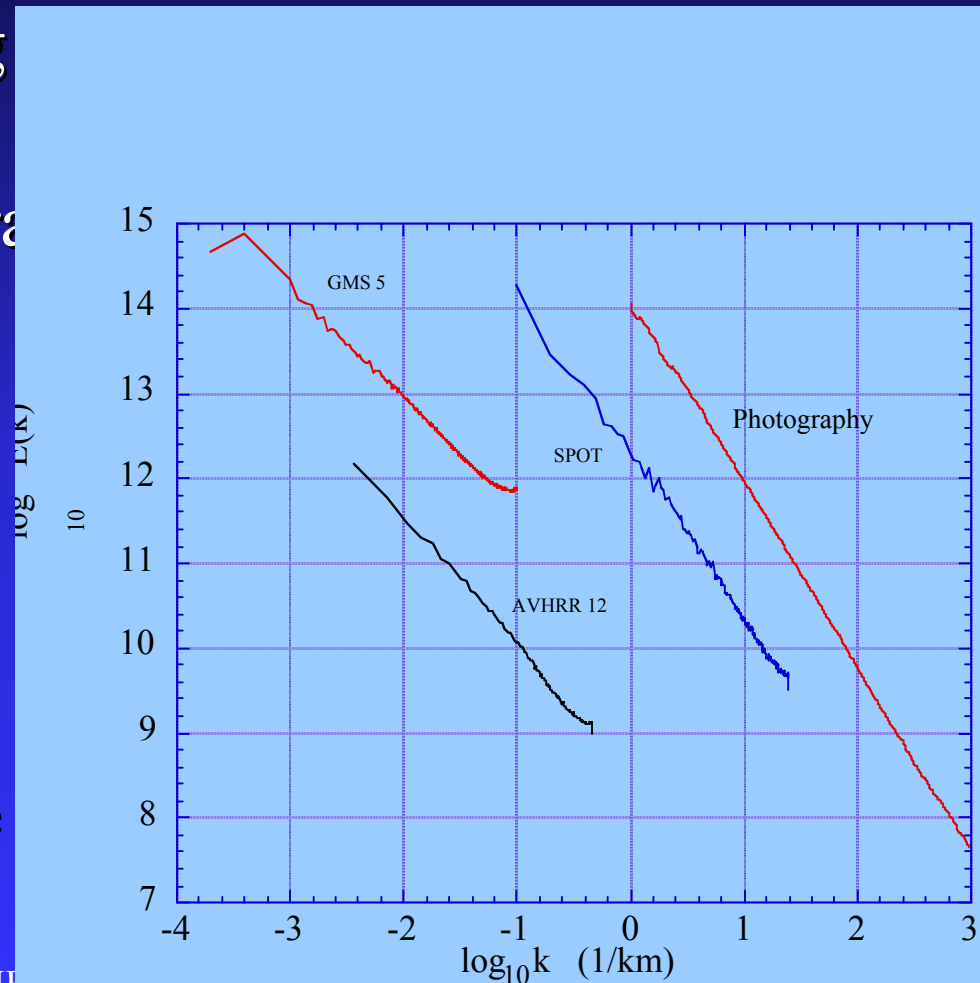
What is the outer scale of atmospheric turbulence?

Spectra of hundreds of satellite images spanning the scale range 1-5000 km, and 38 clouds spectra (1m-1km) from ground camera

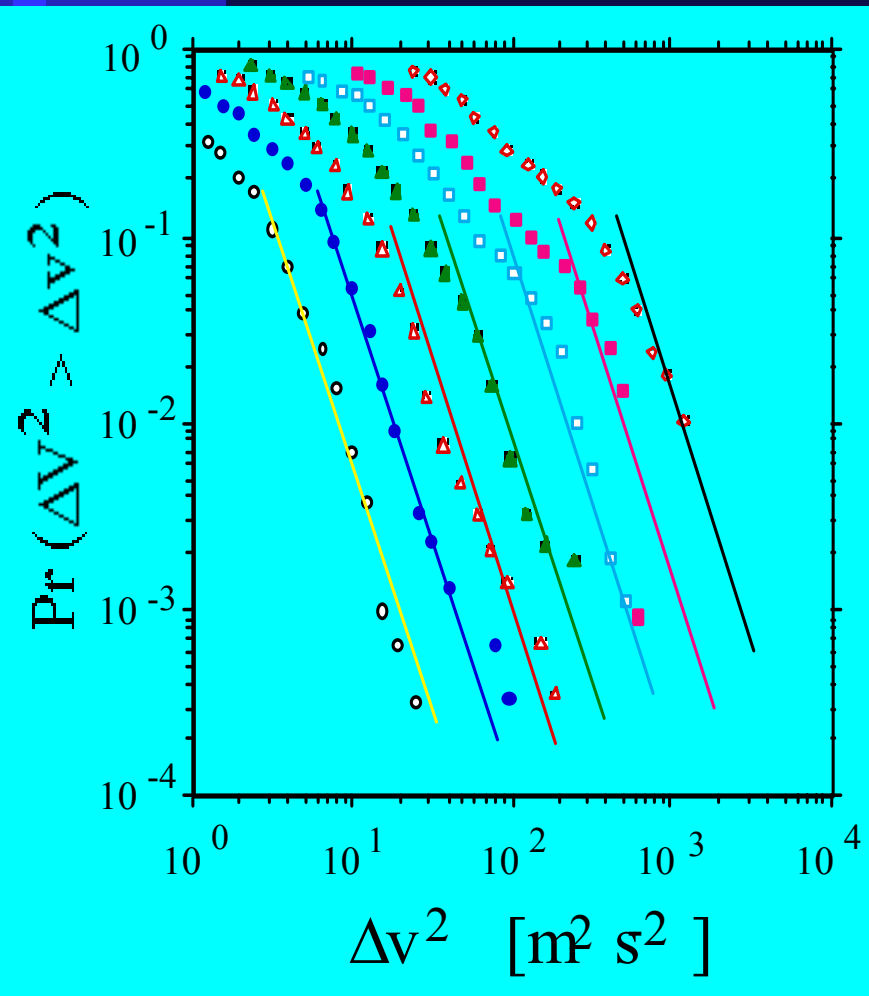
(Sachs et al, 2001)

A multifractal analysis will be more informative (see below).

$L_{\text{eff}} > 5\,000\text{ km} !!$



Landes experiment (Météo-France)



80 balloon vertical soundings:
the probability distribution of the
vertical of shears the horizontal wind,
is scaling for $\Delta z = 50, 100, 200,$
 $400, 800, 1600, 3200$ m. (*):

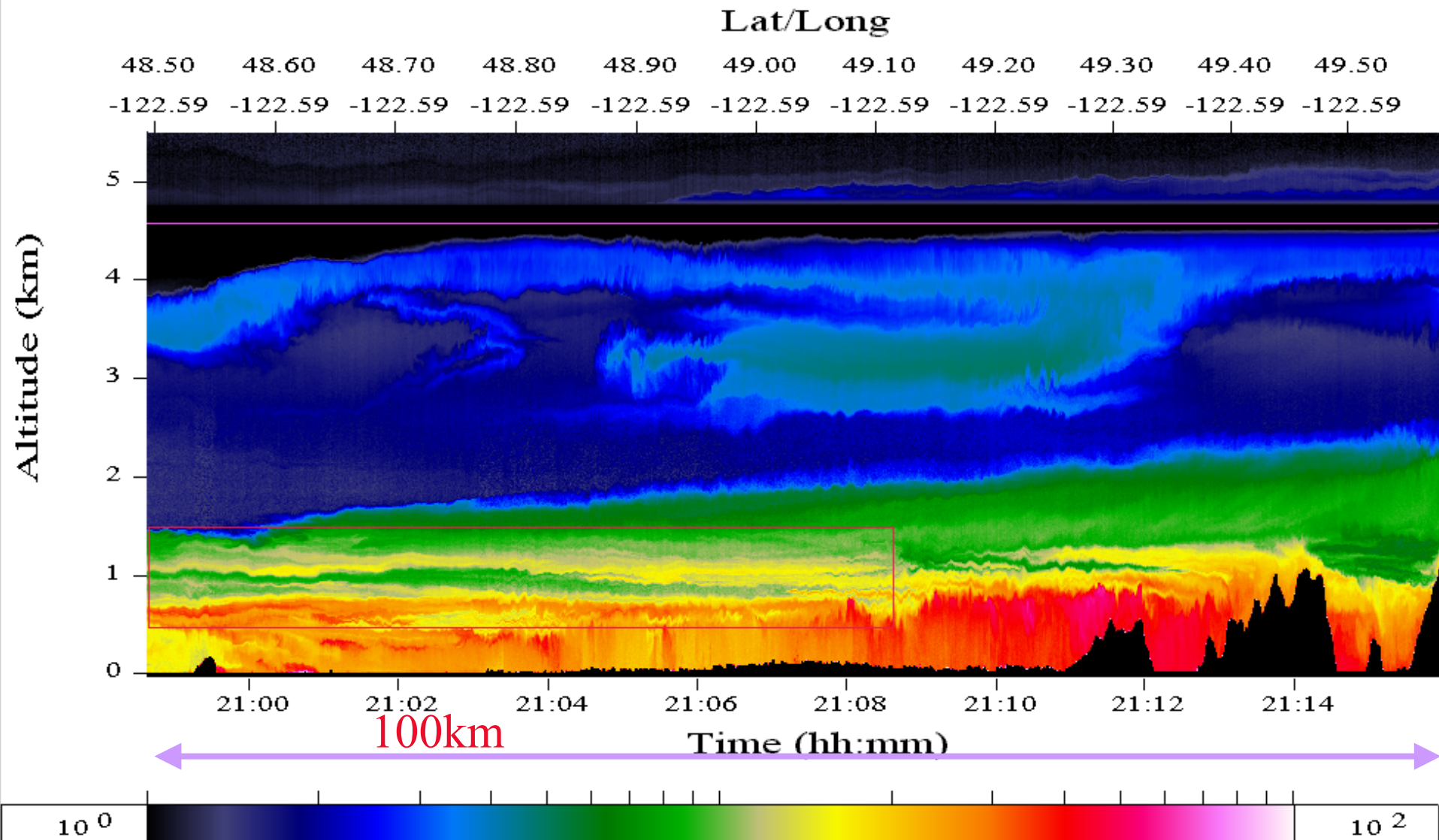
$$\Delta v_z(\Delta z) \sim \Delta z^{3/5}$$

power law tails: $q_D \approx 5$!

* see also Endlich et al., JAS, 1969,

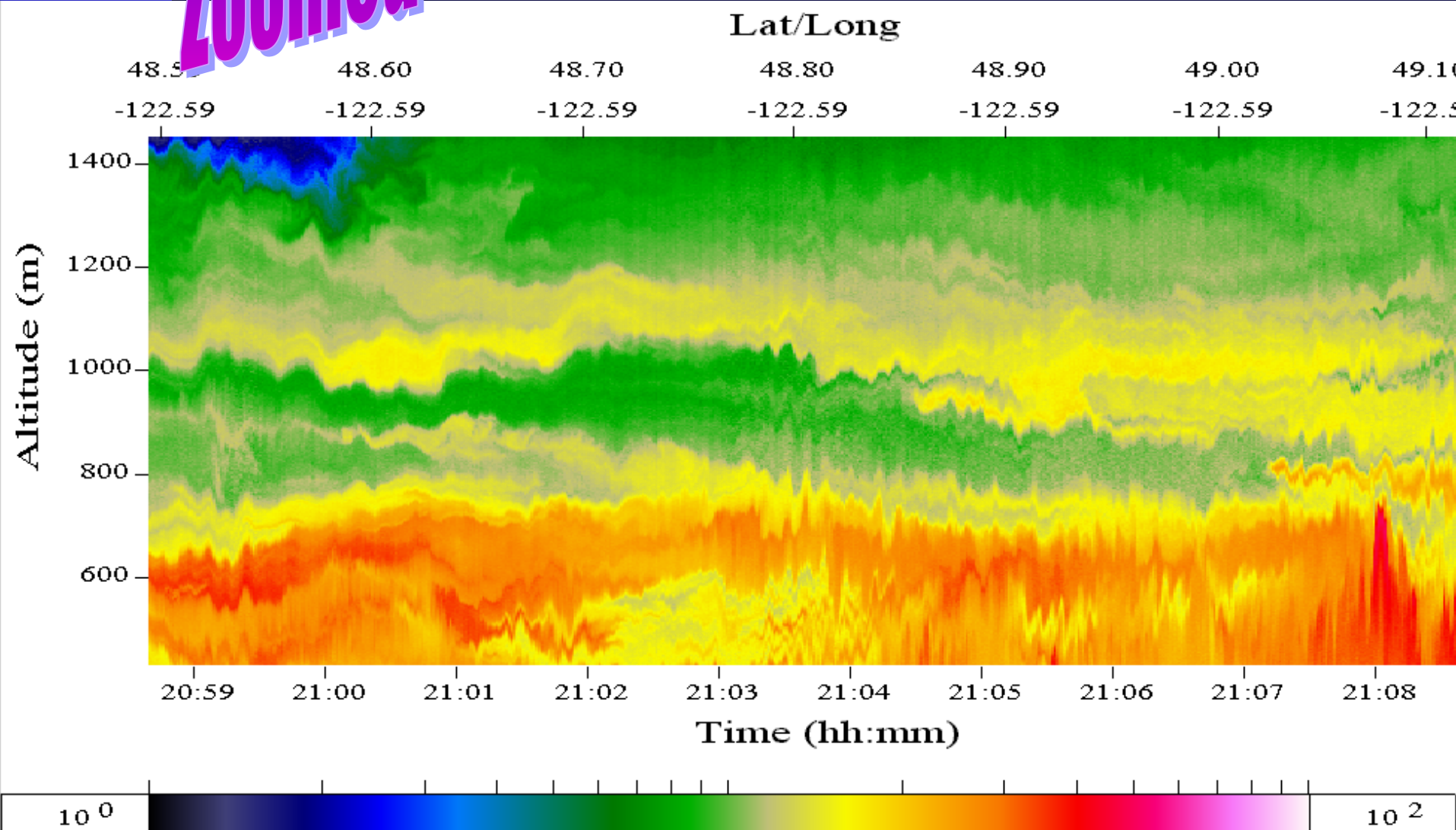
HEPEX, Reading Adelfang, JAS, 1971

AERIAL Data on August 14 (Line 5 N-S) During Pacific2001



AERIAL Data on August 14 (Line 5) During Pacific2001

Zoomed



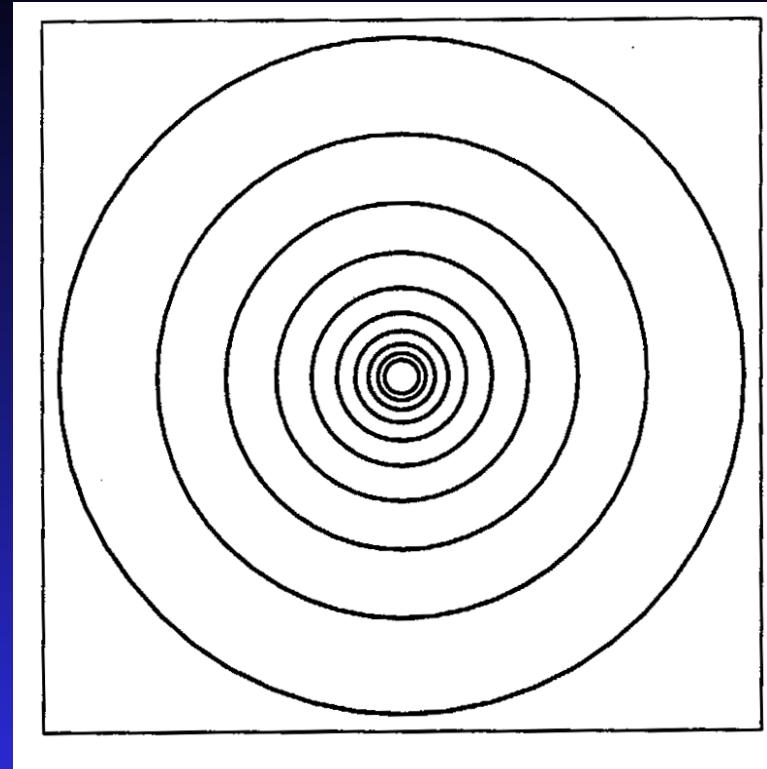
The physical scale function predicted by the 23/9D model of atmospheric dynamics

- stratification
- flattening of structures at larger scales

$$\|\Delta r\| = l_s \left(\left(\frac{\Delta y}{l_s} \right)^2 + \left(\frac{\Delta x}{l_s} \right)^2 + \left(\frac{\Delta z}{l_s} \right)^{2/H_z} \right)^{1/2}$$

Isotropic function

$H_z=1$

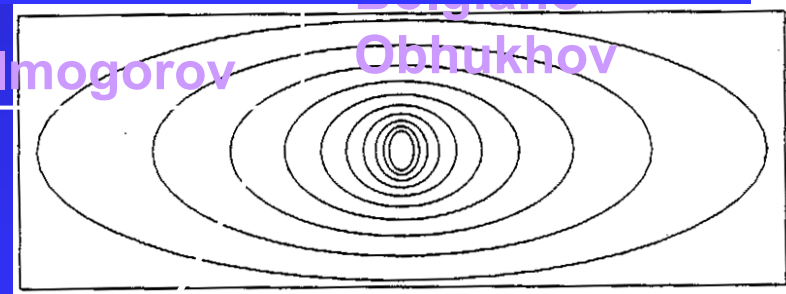


Anisotropic physical scale function

Kolmogorov

Bergano Obhukhov

$H_z=5/9$

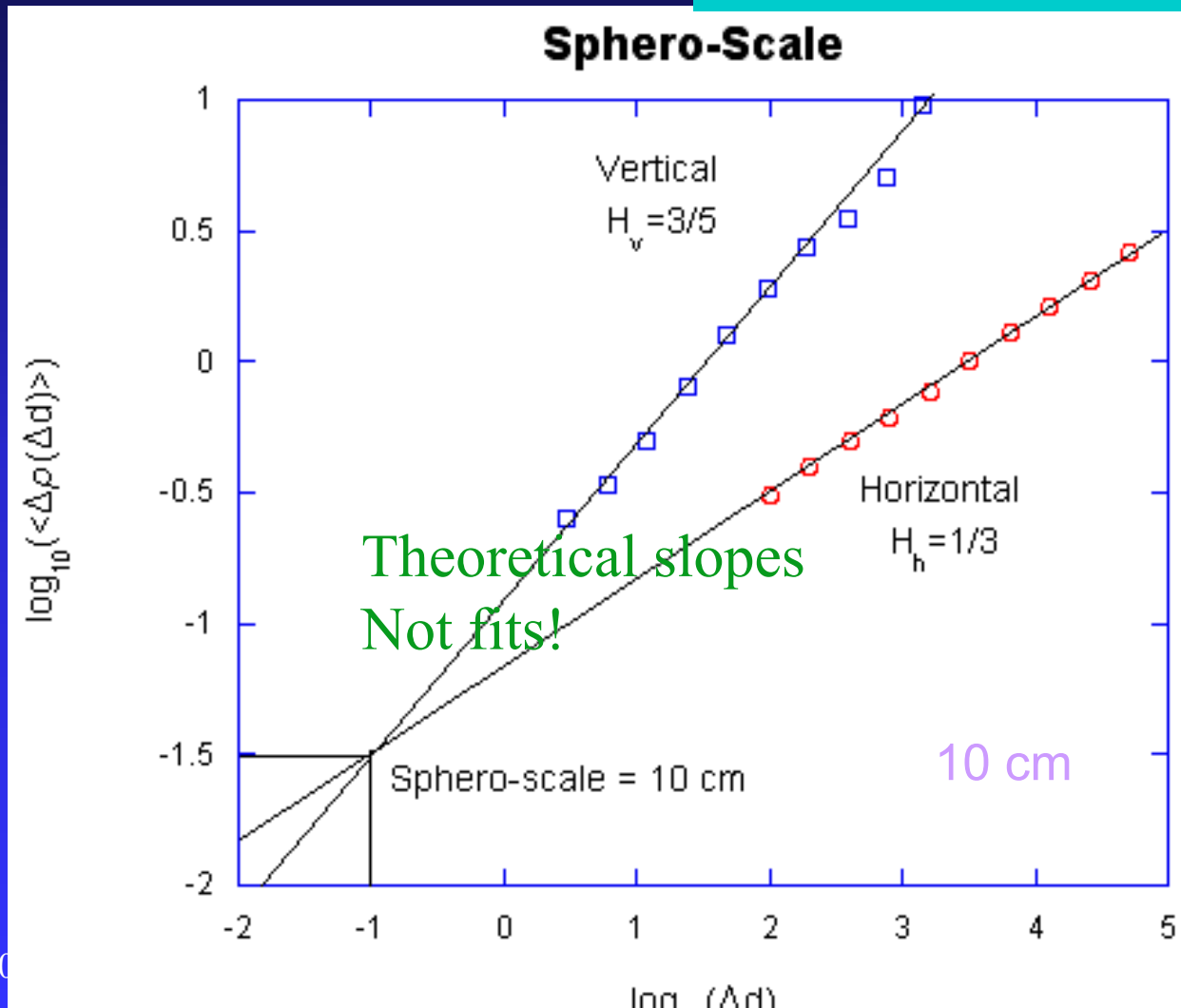


Sphero-scale

Direct Confirmation of physical scale from lidar passive scalars

Generalized Corrsin-Obhukov law

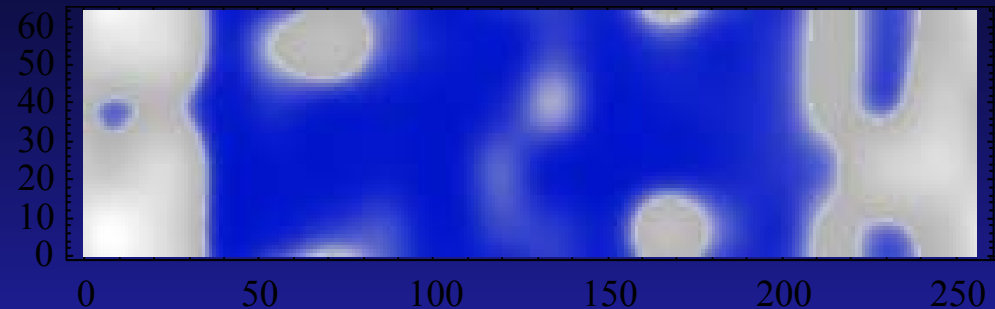
$$\Delta\rho(\Delta r) = \chi^{1/2} \varepsilon^{-1/6} \|\Delta r\|^{1/3}$$



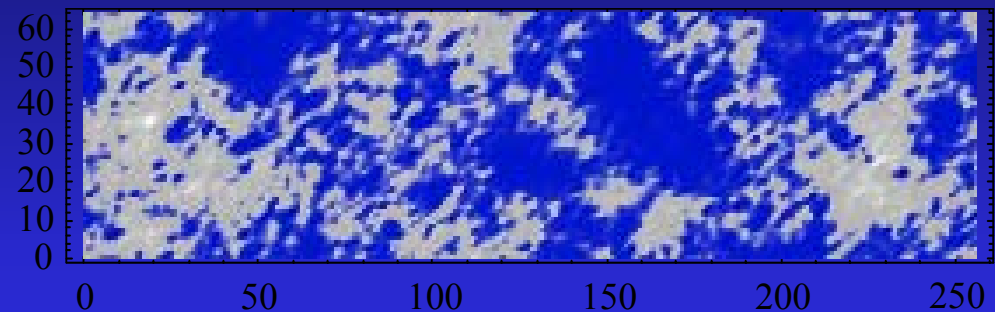
Lilley, L+S,
Strawbridge
2003

Cascade model vs. cascade of models

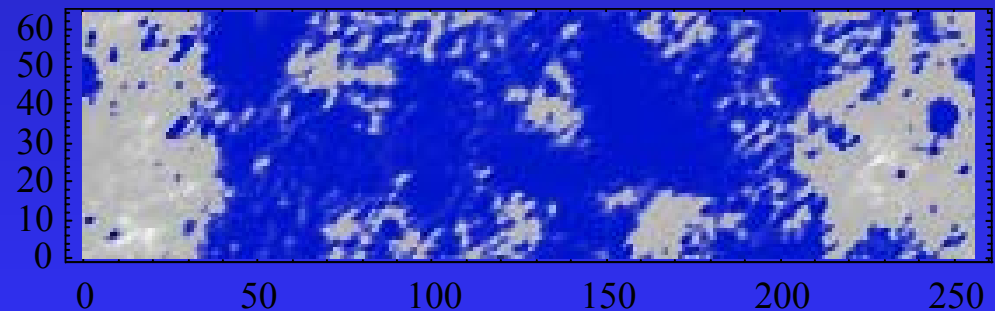
Contributions from l_1 to l_2



Contributions from l_2 to l_3



One parameter
multiplicative group
 \Rightarrow Global result:
contributions from l_1 to l_3



Self-similar multifractal cloud

QuickTime™ and a
Animation decompressor
are needed to see this picture.

Downscaling (blow up) factor 1000; $\alpha=1.8$, $C_1=0.05$, $H=1/3$

Downscaling of stratified clouds

If $l_s=1\text{m}$, then at start: 8km wide, end of zoom: 2.5m wide (factor 32000)

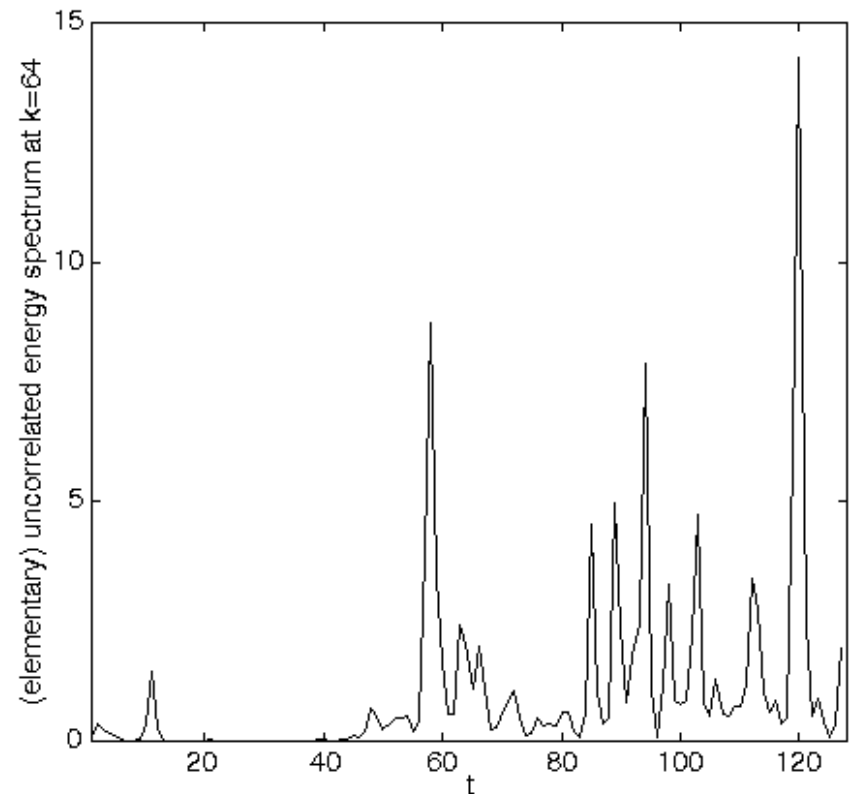
QuickTime™ and a
Animation decompressor
are needed to see this picture.

$\alpha=1.8$, $C1=0.05$, $H=0.33$, $d=7/9$, $c=-2/9$, $e=0.$, $f=0$, $l_s=2^{-5}$ to (about 2^{10} at the 108th image (factor 1.1 enlargements). There are 4×4 subpixels, $2^7 \times 2^8$.

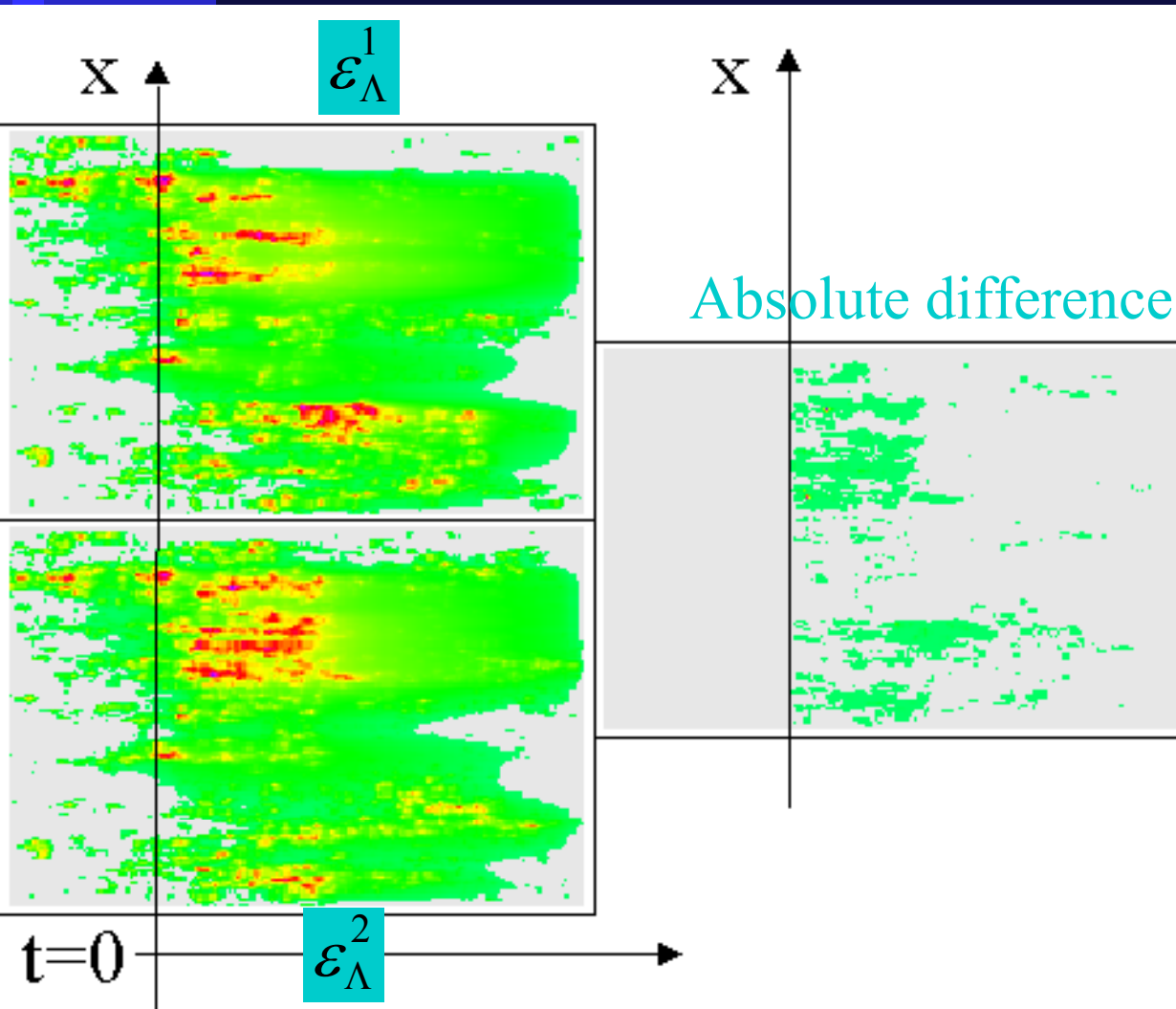
Multifractal rain simulation

Fondamental role of intermittency:
Intermittent puffs of decorrelation

QuickTime™ and a
Animation decompressor
are needed to see this picture.



Multifractal predictability limits



Generation of intense Differences, due to Growing independence between the two fields at small scales..

Multifractality of the joint field

$$\varepsilon_{\Lambda}^1 \varepsilon_{\Lambda}^2$$

$$\langle \varepsilon_{\Lambda}^q \rangle \approx \Lambda^{K(q)} \Rightarrow$$

$$C^{(q)}(\varepsilon_{\Lambda}^1, \varepsilon_{\Lambda}^2) = \langle (\varepsilon_{\Lambda}^1 \varepsilon_{\Lambda}^2)^q \rangle / \langle \varepsilon_{\Lambda}^1 \rangle^q \langle \varepsilon_{\Lambda}^2 \rangle^q$$

$$\propto \lambda(t)^{K(q,2)}; K(q,2) = K(2q) - 2K(q)$$

Hint:
the 2 fields remain identical on a decreasing scale ratio:
whereas independent for smaller scales.

$$\lambda(t) \approx \text{Min} \left[\Lambda, (T/t)^{\frac{1}{1-H_t}} \right] \quad (t > 0)$$

($1 - H_t$ = dynamical exponent
T = outer time scale)

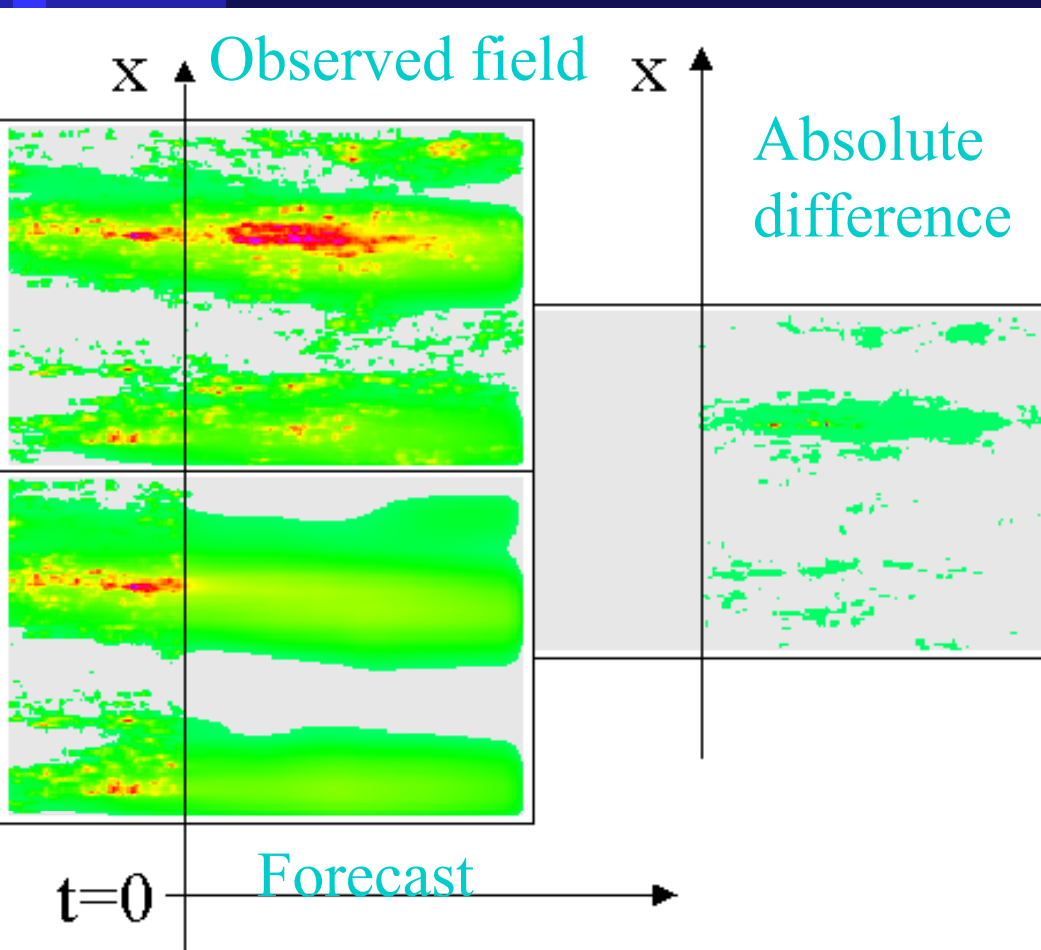
$$\text{Pr}(\varepsilon_{\Lambda}^q > \Lambda^{\gamma}) \approx \Lambda^{c(\gamma)} \Rightarrow \text{Pr} \left((\varepsilon_{\Lambda}^1 \varepsilon_{\Lambda}^2) > \lambda(t)^{\gamma} \langle \varepsilon_{\Lambda}^1 \rangle \langle \varepsilon_{\Lambda}^2 \rangle \right) \approx \Lambda^{c(\gamma,2)}$$

$$c(\gamma), K(q)$$

$$c(\gamma, 2), K(q, 2)$$

are (merely) dual for the Legendre transform !

Multifractal forecast and deterministic sub-grid modeling



Future component of the generator deterministically prescribed to preserve the mean field: rapid loss of small structures!

This confirms and explain why Stochastic sub-grid modeling Can do much better than deterministic ones

(Buizza *et al.*, 1999; Houtekamer *et al.*, 1996)!

Quantitative statements with the help of :

$$C^{(q)}(\varepsilon_{\Lambda}^F, \varepsilon_{\Lambda}^0)$$

Conclusions

- EPS, a radical change in weather forecasting: deterministic framework abandoned for a probabilistic framework!
- However, do our current deterministic models can generate a reliable pdf (Liouville or FP ‘ scenarios ’) ?
- fundamental importance of a wide range of time and **space** scales, as well as of intensities:
 - ➡ no characteristic time of predictability,
 - ➡ well defined power-law predictability decay,
 - ➡ optimize prediction w.r.t. them ?
- Let 's talk... and work together !