



A review on wavelet transforms and their applications to MHD and plasma turbulence I

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*In collaboration with
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*Methods for Analyzing Turbulence Data
Meudon, 29 May 2015*

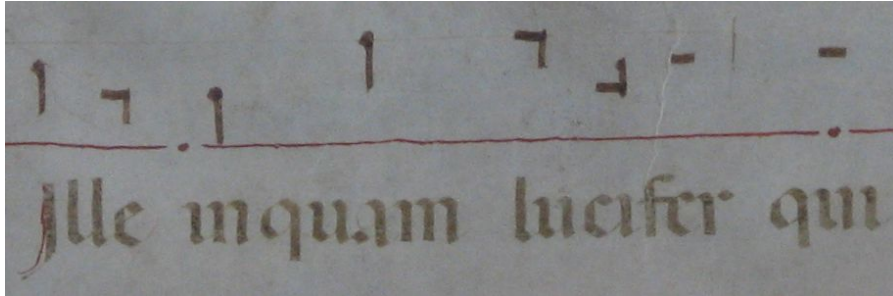
Choice of the appropriate representation

'It could be interesting, in communication theory, to represent an oscillating signal by a superposition of elementary wavelets, each of them having both a frequency and a time localization quite well defined. The useful information is indeed often carried by both the emitted frequencies and by the time structure of the signal (music is a characteristic example of that). The representation of a signal as function of time cannot exhibit the frequency content, while in contrast its Fourier analysis hides the time of emission and the duration of each elements of the signal. An adequate representation should combine the advantages of these two complementary descriptions, while providing a discrete character appropriate to the theory of communication.'

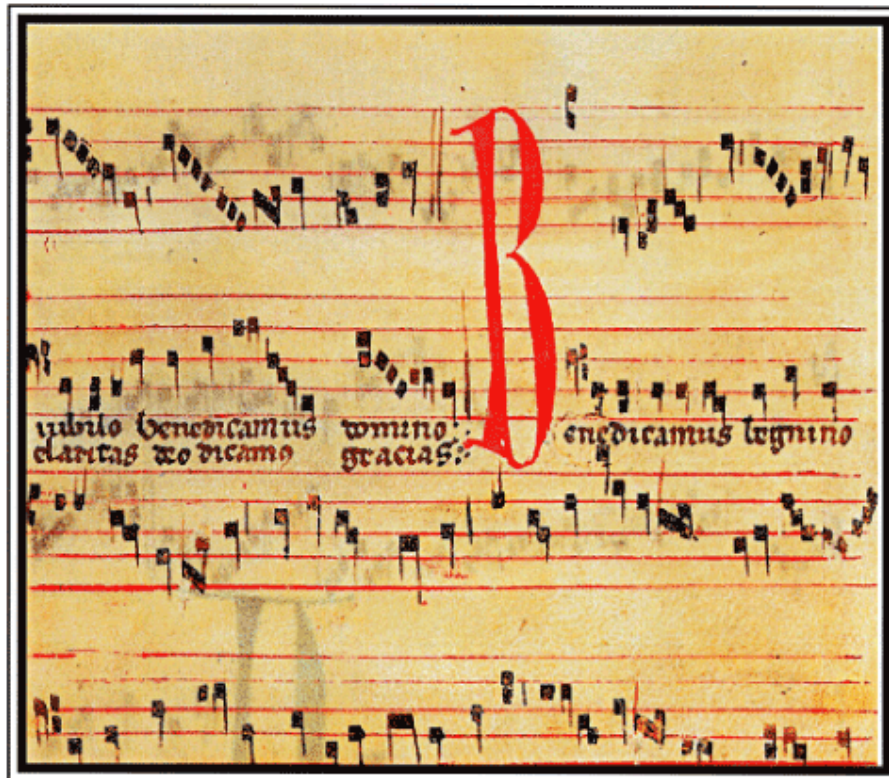
Roger Balian

Un principe d'incertitude
fort en théorie du signal
CRAS, 292, II (1981)

Representation for music



*Guido d'Arezzo
Micrologos
1025*



7 tones:

*ut re mi fa sol la si
sa re ga ma pa da ni*

12 half tones:

$$f_n = f_0 \cdot a^n$$
$$a = 2^{1/12}$$

Integral transforms

Analysis

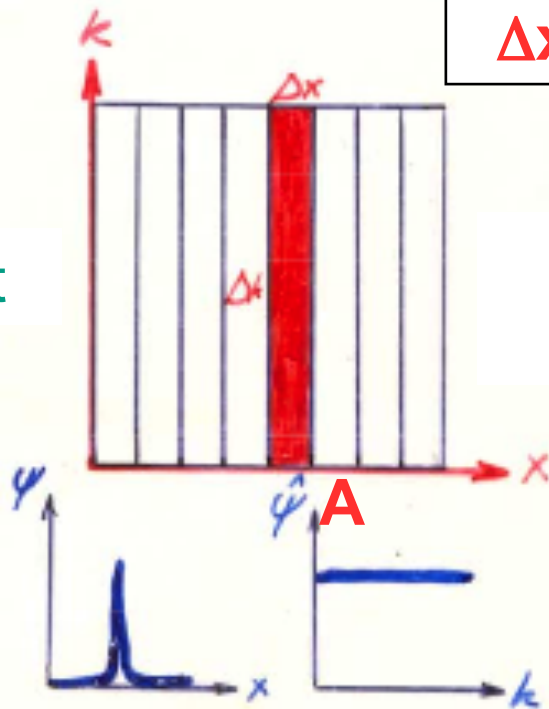
$$T_f(k) = \int f(x) \psi_k(x) dx$$

Synthesis

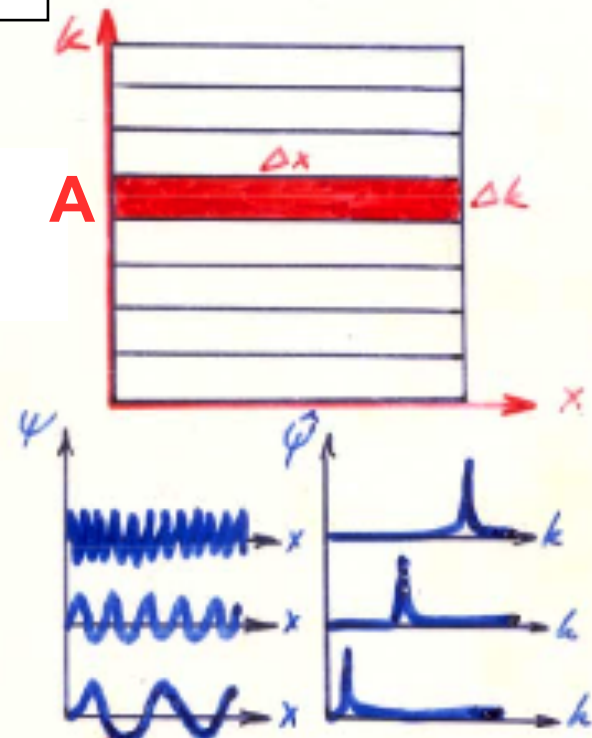
$$f(x) = \frac{1}{c} \int T_f(k) \psi_x(k) d\mu(k)$$

$$\Delta x \Delta k = A$$

Gridpoint

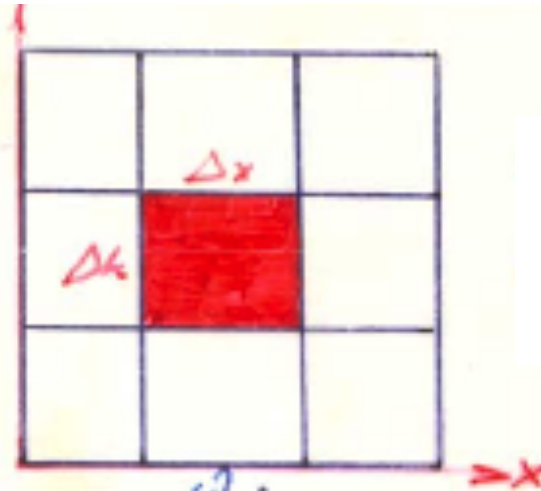


Fourier
(1807)

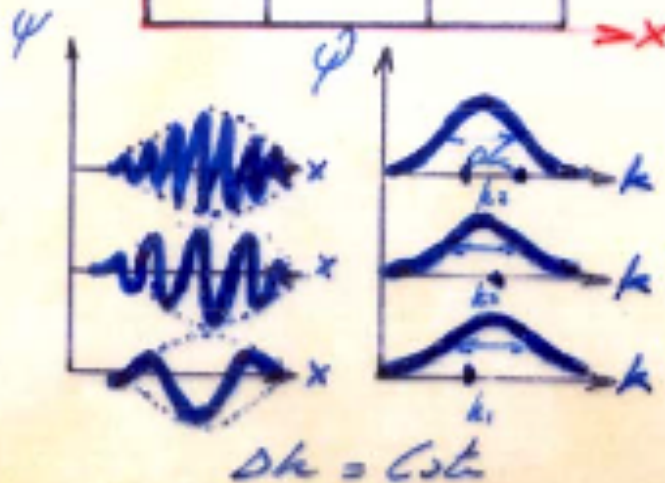
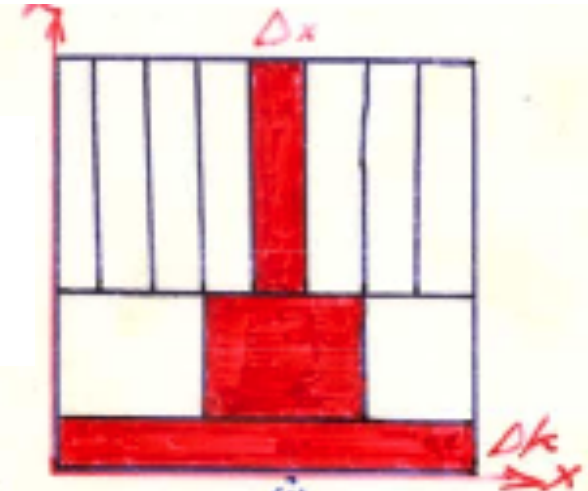


Optimal phase space tiling

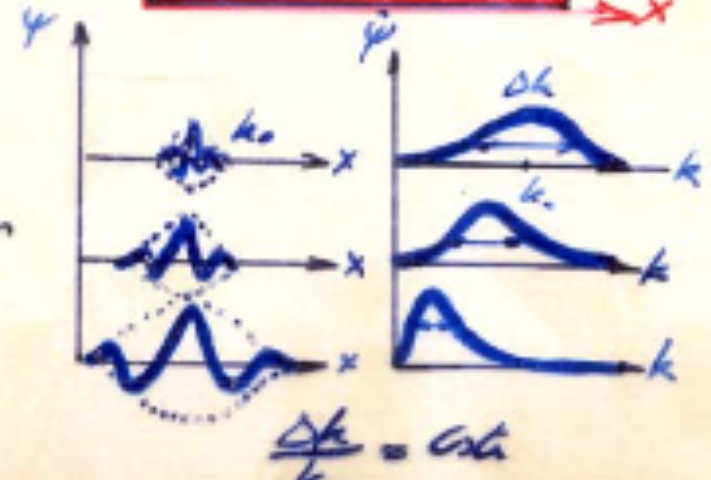
Gabor
(1946)



Wavelets
(1984)



Balian's
destruction
(1981)



Space-wavenumber
representation

$\Delta x \Delta k =$
information atom

Space-scale
representation

Choice of the analyzing wavelet

Admissibility condition

$$C_\psi = \int_0^\infty \frac{|\hat{\psi}(k)|^2}{|k|} dk < \infty$$
$$\int_{-\infty}^\infty \psi(x) dx = 0 \quad \text{or} \quad \hat{\psi}(0) = 0$$

Jean Morlet



Alex Grossmann



*Analyzing wavelet family
generated by translation (b)
and dilation (a)*

$$\psi_{a,b}(x) = \frac{1}{\sqrt{a}} \psi \left(\frac{x-b}{a} \right)$$

Grossmann and Morlet,
*Decomposition of Hardy functions into square
integrable wavelets of constant shape,*
SIAM J. math. Anal., **15**(4), 723-736, 1984

Continuous wavelet transform (CWT)

Analysis

$$\tilde{f}(a, b) = \int_{-\infty}^{\infty} f(x) \psi_{a,b}^*(x) dx$$

Synthesis

$$f(x) = \frac{1}{C_\psi} \int_0^{\infty} \int_{-\infty}^{\infty} \tilde{f}(a, b) \psi_{a,b}(x) \frac{da db}{a^2}$$

Parseval's identity

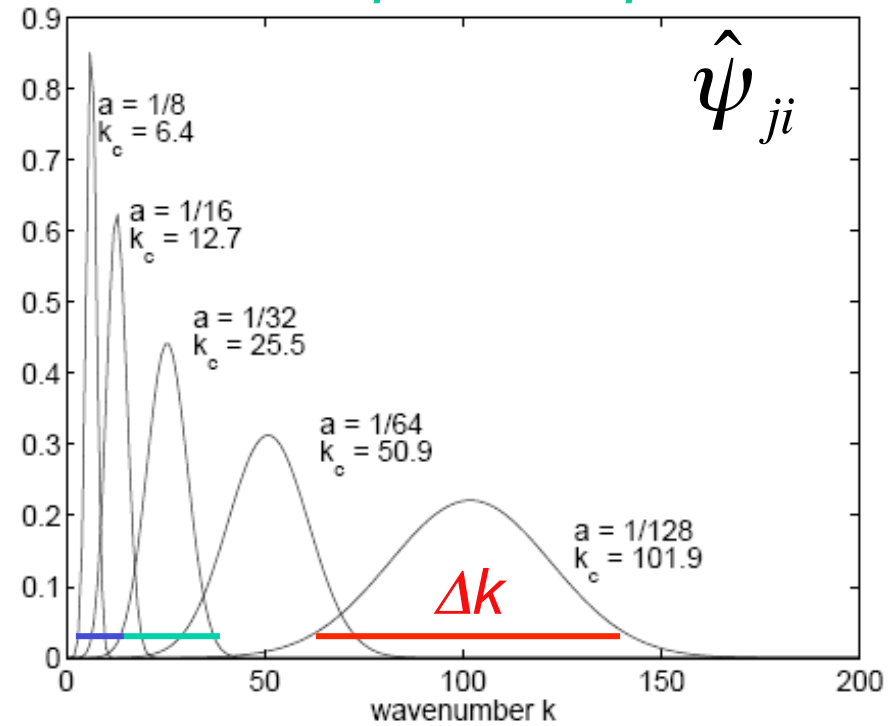
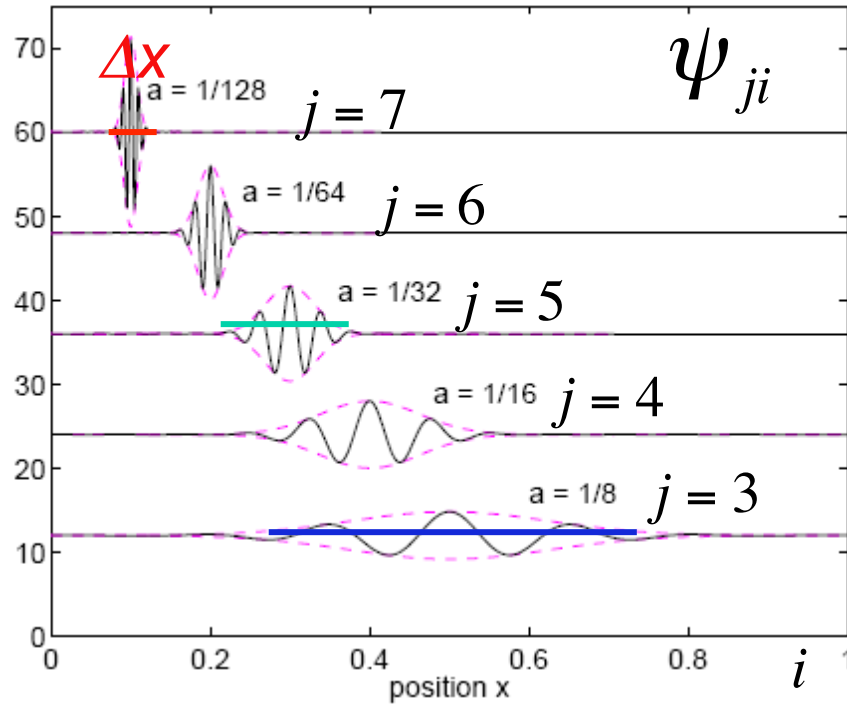
$$\langle f_1, f_2 \rangle = \int_{-\infty}^{\infty} f_1(x) f_2^*(x) dx = \frac{1}{C_\psi} \int_0^{\infty} \int_{-\infty}^{\infty} \tilde{f}_1(a, b) \tilde{f}_2^*(a, b) \frac{dadb}{a^2}$$

Wavelet representation

Physical space

$$\Delta x \Delta k > C$$

Spectral space

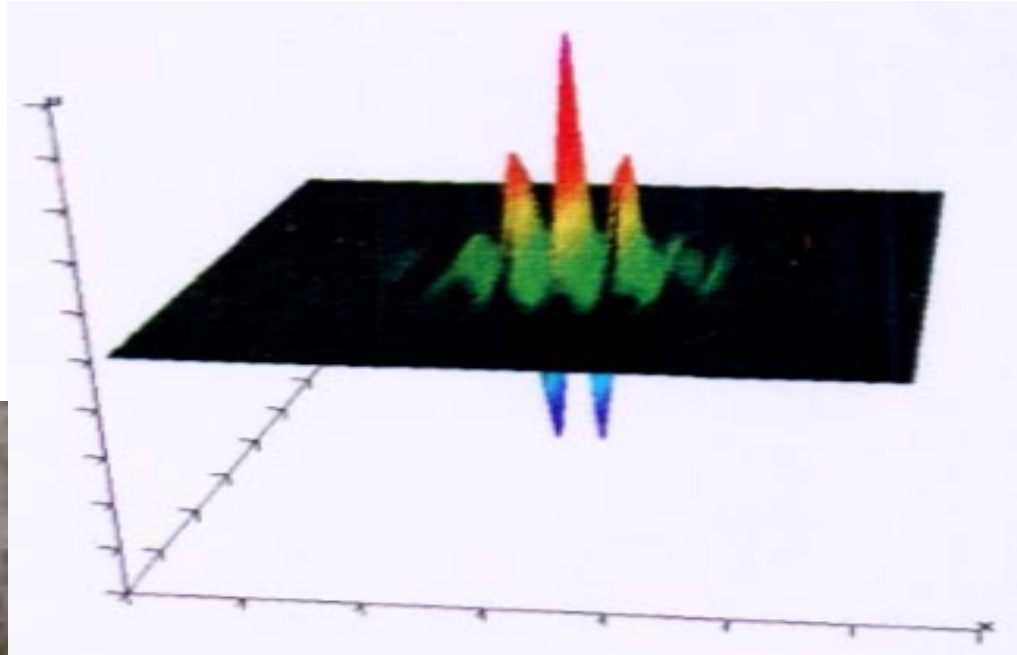


Farge,
Wavelet transforms and
their applications to turbulence
Ann. Rev. Fluid Mech., **92**, 1992

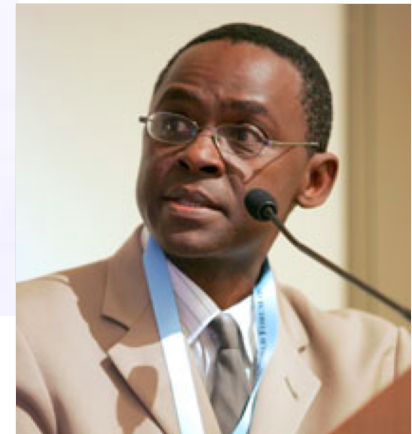
Farge and Schneider,
Wavelets: application to turbulence,
Encyclopedia of Mathematical Physics,
Springer, 408-42, 2006

2D continuous wavelet transform

Jean-Pierre
Antoine



Romain
Murenzi



2D Morlet mother wavelet

*The wavelet family is generated
by translating, dilating and rotating
the 2D mother wavelet*

Analyzing wavelet

Field to analyze

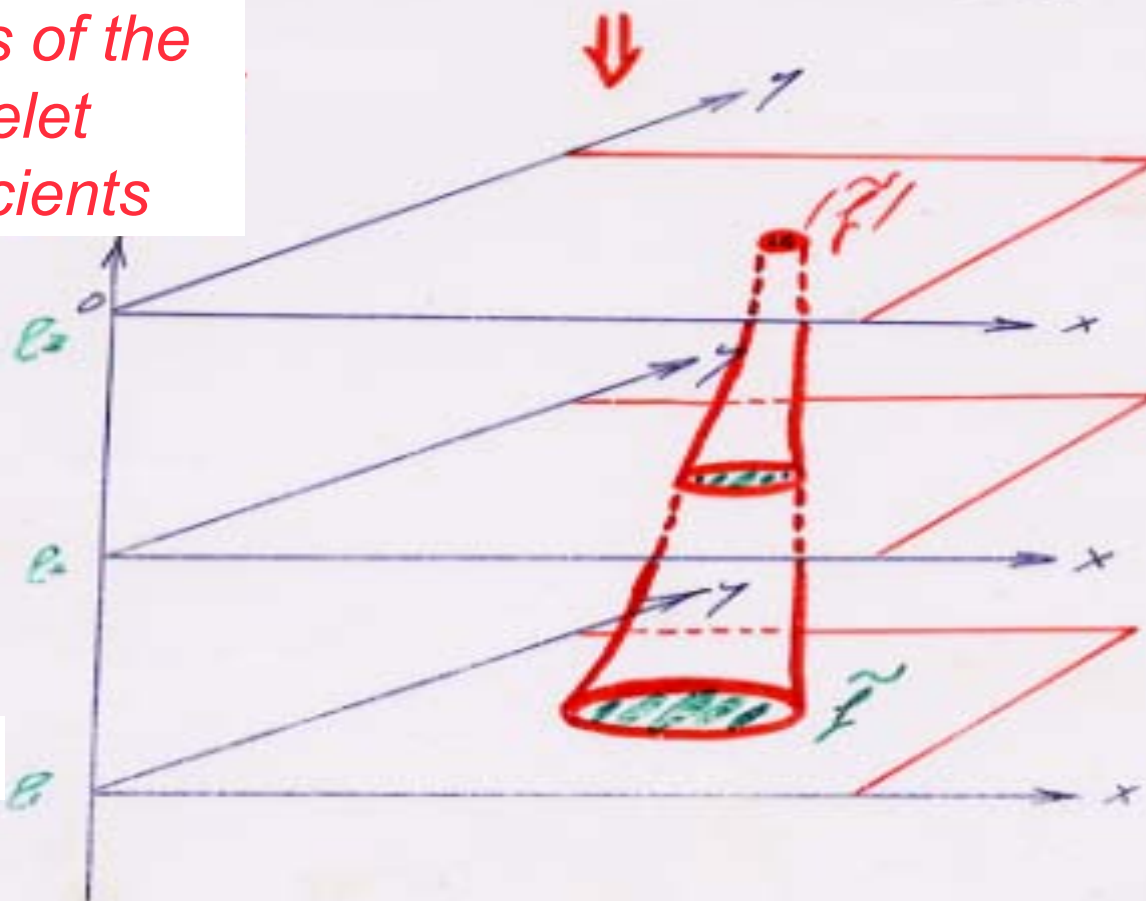


Modulus of the
wavelet
coefficients

Small scale



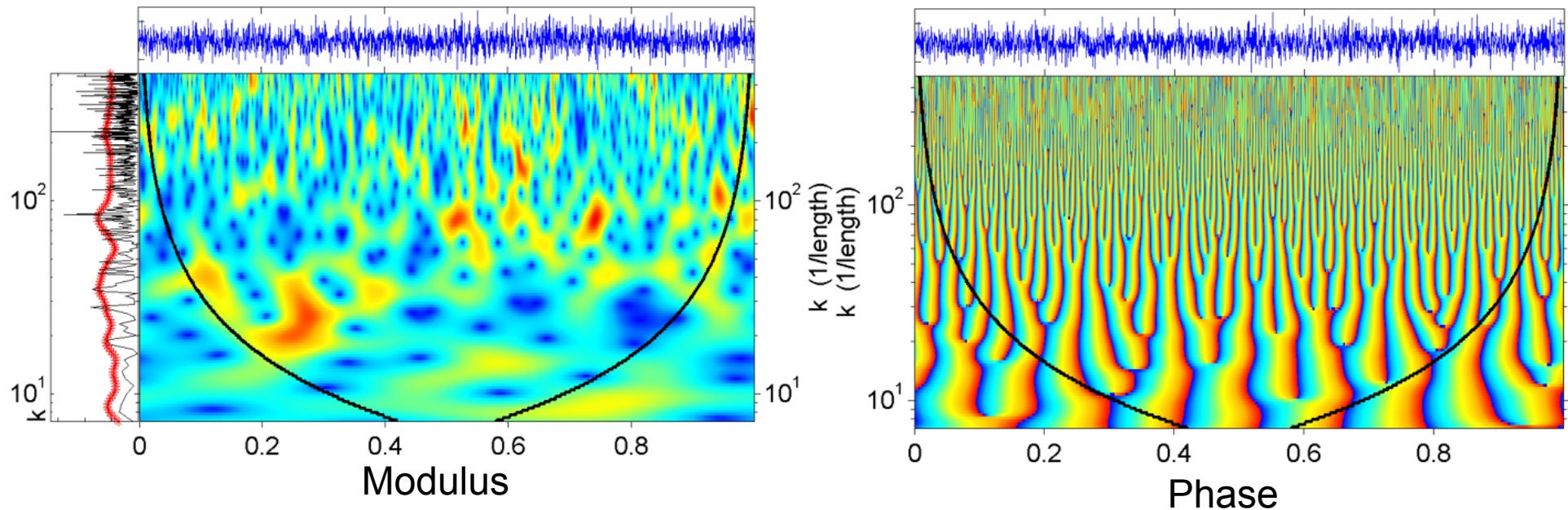
Large scale



Logarithm
of the scale

Reproducing kernel of the CWT

The CWT of a Gaussian white noise reveals its reproducing kernel

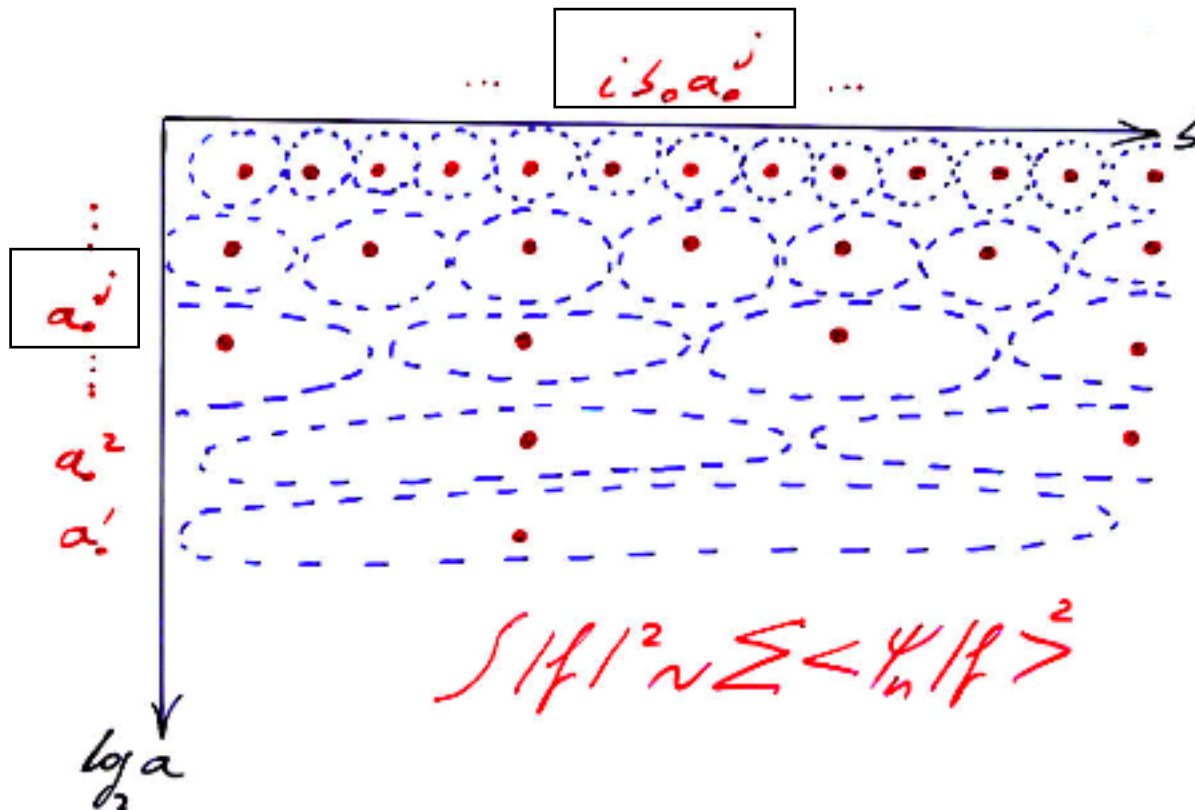


This is the correlation between the wavelets which corresponds to the redundancy between the coefficients

$$K(b', a', b, a) = \langle \psi_{b' a'} | \psi_{ba} \rangle$$

Wavelet frame

We can then select a finite number of wavelets restricted to a discrete grid optimally chosen such that the wavelet family associated to this grid constitutes a quasi-orthogonal basis \Rightarrow a wavelet frame



For example
for Marr wavelet

we need

$$a_0 = 2^{1/2}$$

$$b_0 = 1/2$$

Orthogonal wavelet transform

Wavelet analysis :

$$\tilde{f}_{ji} = \langle \psi_{ji} | f \rangle \quad \text{with} \quad \psi_{ji} = 2^{j/2} \psi(2^j x - i)$$

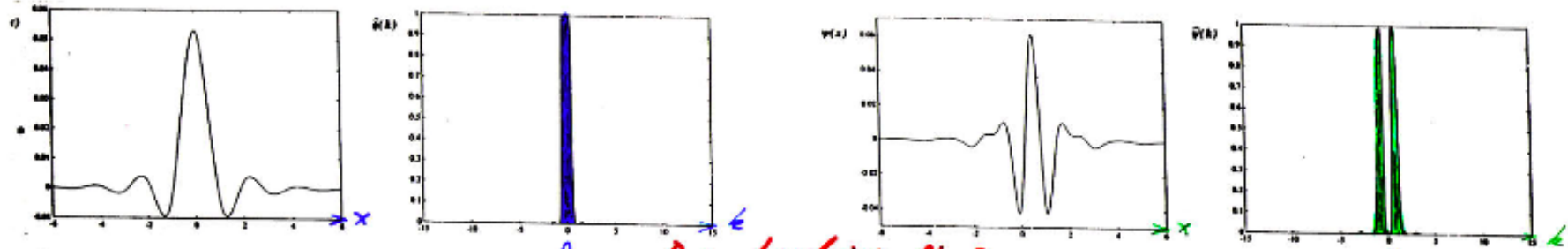
Wavelet synthesis :

$$f = \sum_{ji} \langle \psi_{ji} | f \rangle \psi_{ji}$$

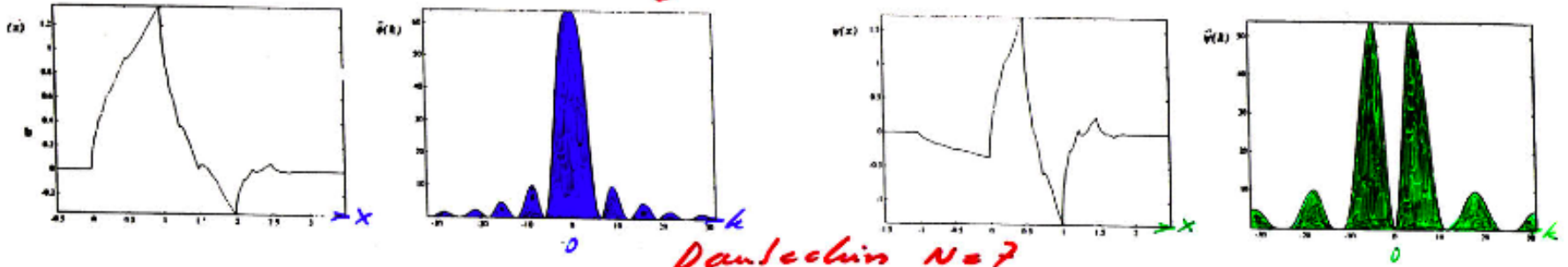
A signal sampled on N points is wavelet analyzed and synthesized in CN operations if one uses compactly-supported wavelets computed from a quadratic mirror filter of length M .

Examples of orthogonal wavelets

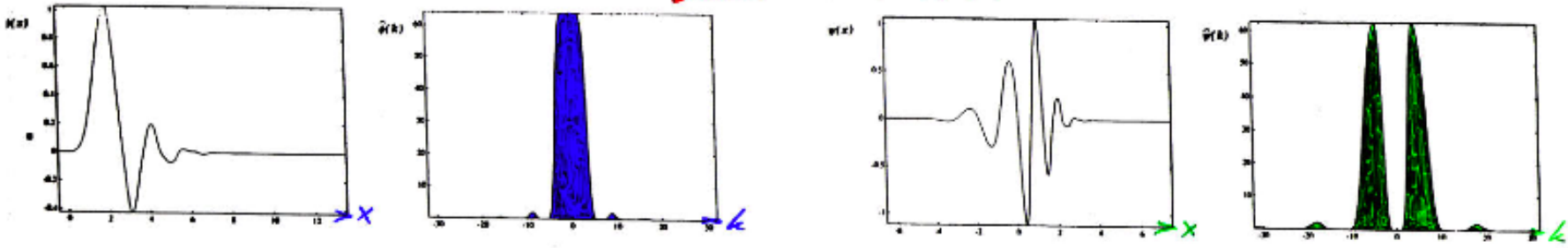
Bath-Lemanic



Daubechies N=2



Daubechies N=7



ϕ
Scaling functions

$|\hat{\phi}|$

ψ

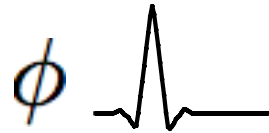
Mother wavelets

$|\hat{\psi}|$

2D orthogonal wavelets

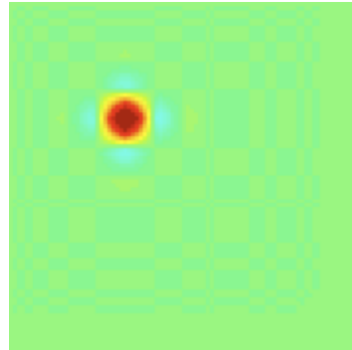
Scaling function

Wavelet



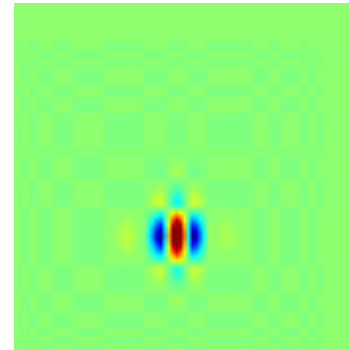
$$\phi(x)\phi(y)$$

Coarse
approximation



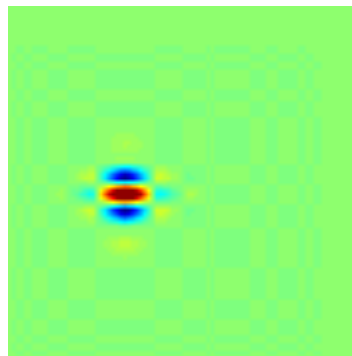
$$\psi(x)\phi(y)$$

Horizontal
details



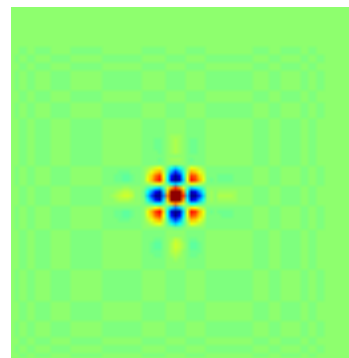
$$\phi(x)\psi(y)$$

Vertical
details



$$\psi(x)\psi(y)$$

Diagonal
details



3D orthogonal wavelets

A 3D vector field $v(\mathbf{x})$ sampled on $N = 2^{3J}$ equidistant grid points

$\psi_\lambda(\mathbf{x})$ 3D wavelet \rightarrow orthogonal wavelet series

$$v(\mathbf{x}) = \sum_{\lambda \in \Lambda} \tilde{v}_\lambda \psi_\lambda(\mathbf{x}), \quad \tilde{v}_\lambda = \langle v, \psi_\lambda \rangle$$

$$\Lambda = \{ \lambda = (j, i_n, \mu)_{n=1,2,3}, \dots, J-1, i_n = 0, \dots, 2^j - 1, n = 1, 2, 3, \text{ and } \mu = 1, \dots, 7 \}$$

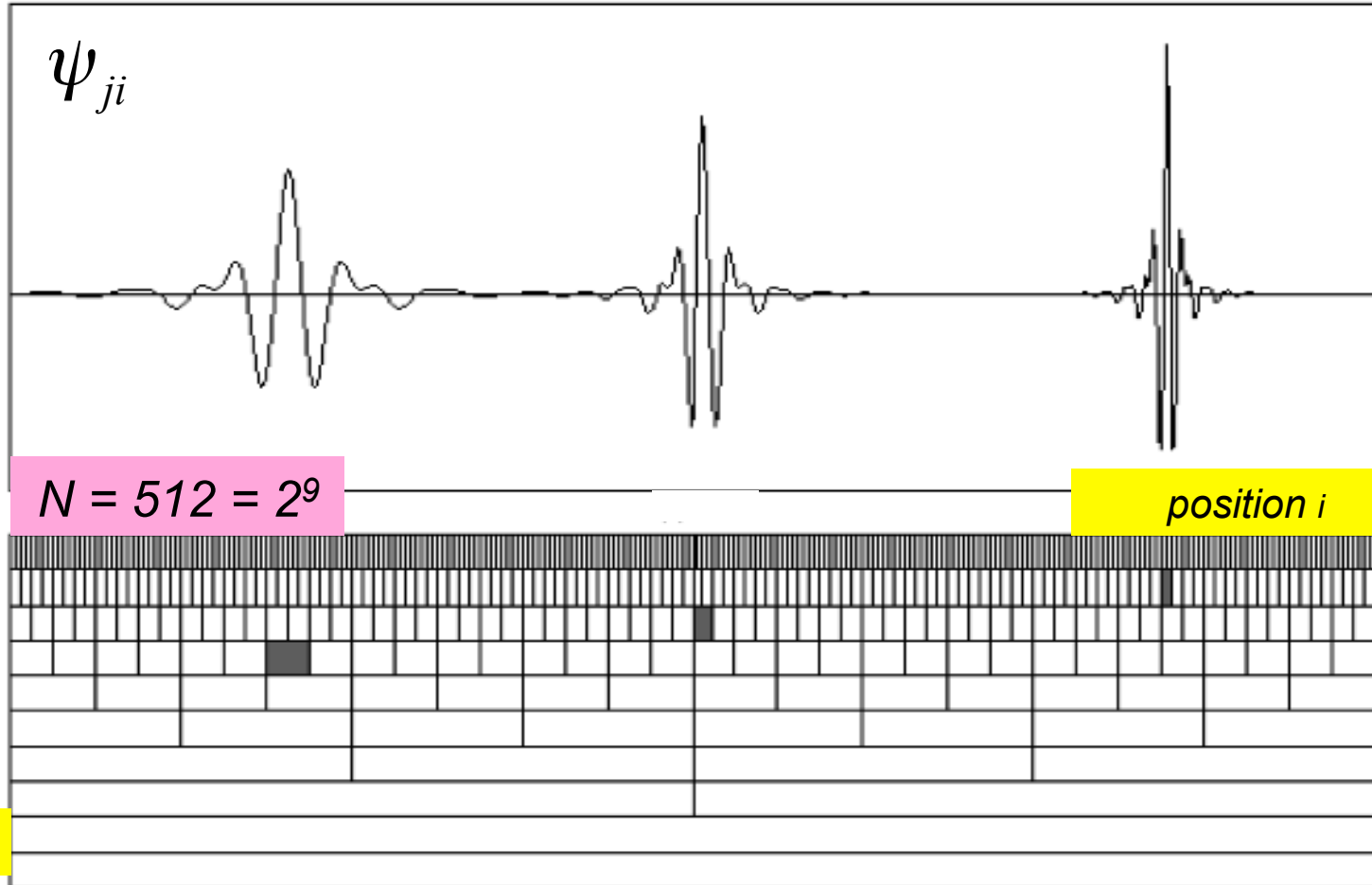
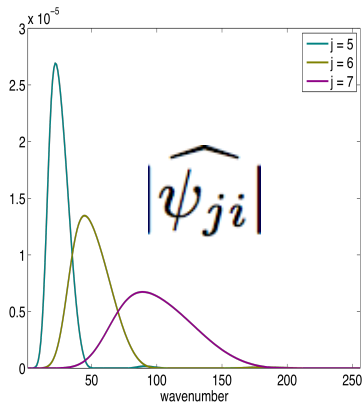
$N_j = 7 \times 2^{3j}$, wavelet coefficients at a scale indexed by j

- fast algorithm with linear complexity
- no redundancy between the coefficients

We use Coifman 12 wavelet
compactly supported with four vanishing moments.

Orthogonal wavelet representation

Wavelets



Wavelet coefficients

$$\tilde{f}_{ji} = \langle \psi_{ji} | f \rangle$$

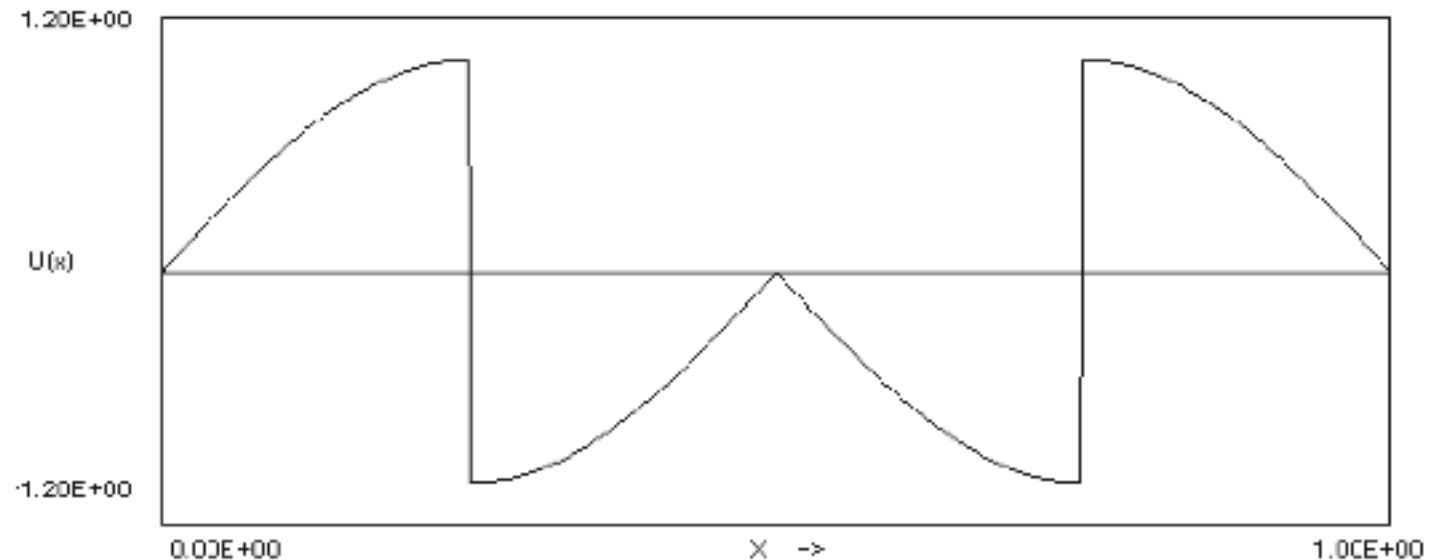
scale j

position i

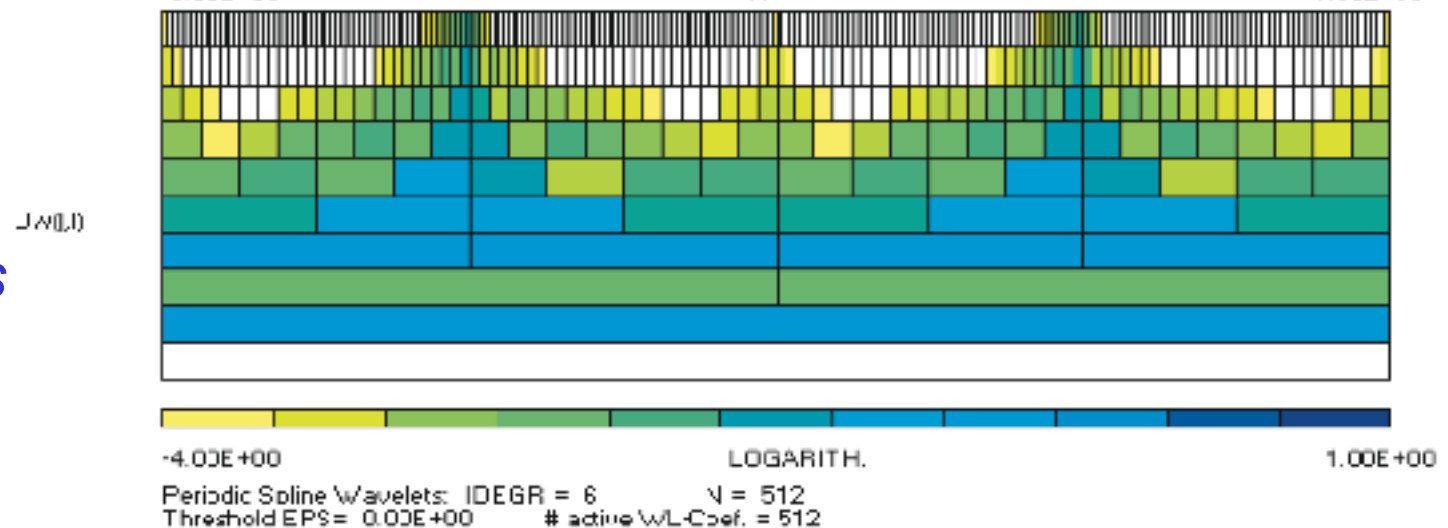
Mallat,
A wavelet tour of
signal processing, 3rd edition,
Academic Press, 2008

Academic example

*Function
sampled on
 $N = 512$
grid-points*



*$N = 512$
wavelet
coefficients*

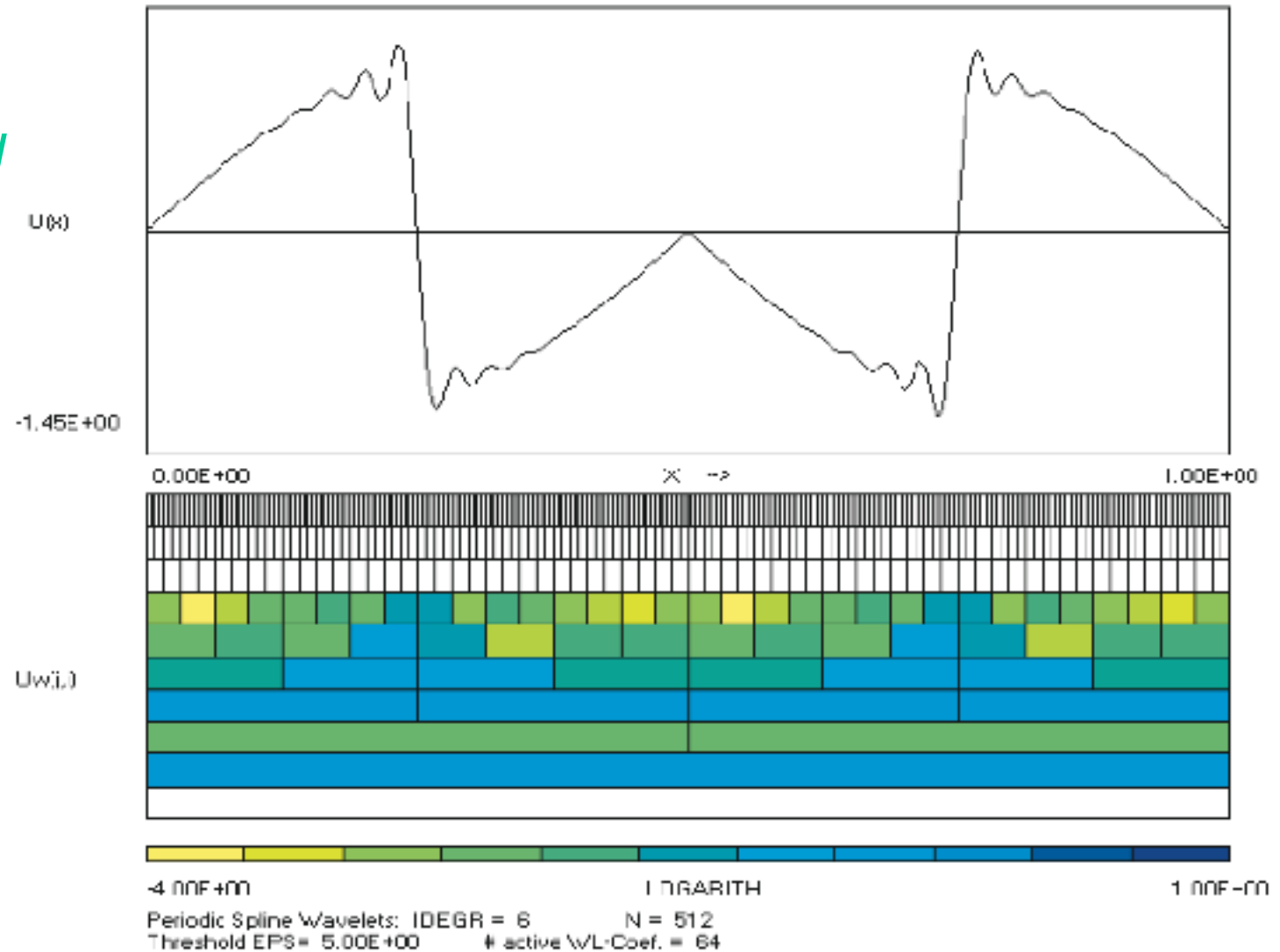


Linear approximation

Function reconstructed from 64 wavelet coefficients

64 wavelet coefficients such that

$$j \leq 6$$

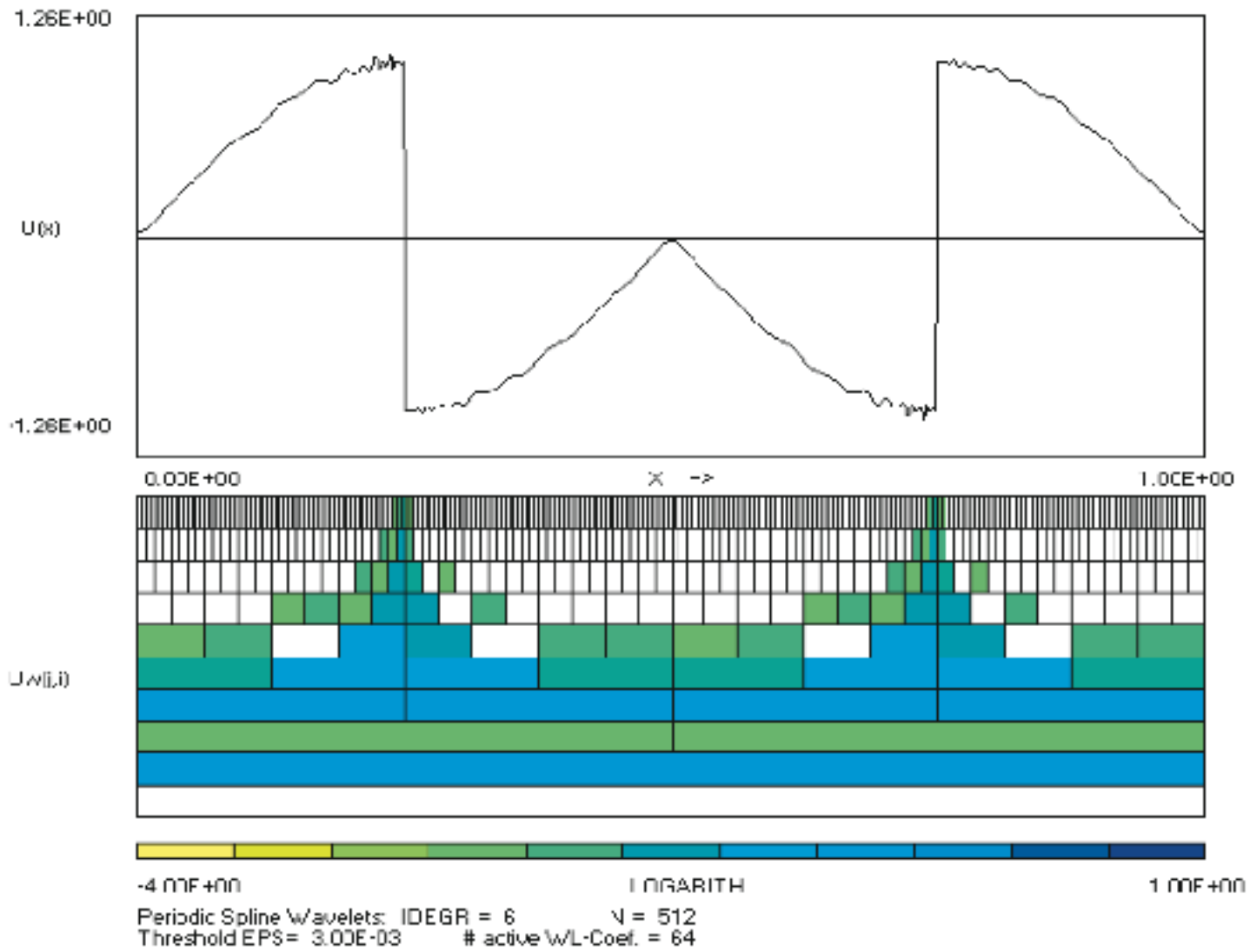


Nonlinear approximation

Function reconstructed from 64 wavelet coefficients

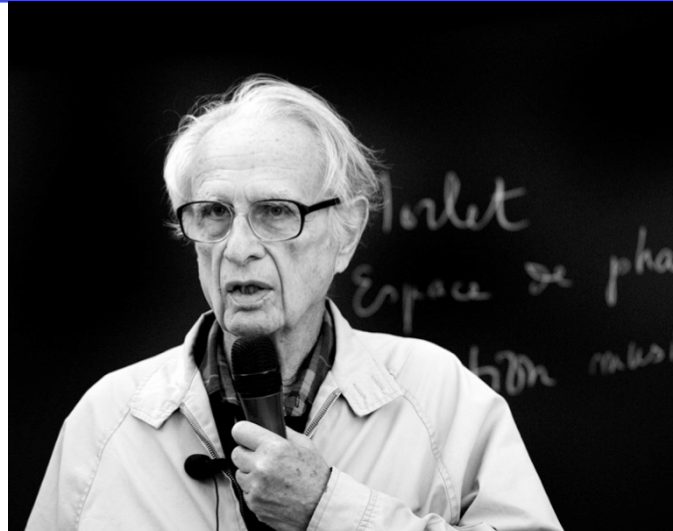
64 wavelet coefficients such that

$$|\tilde{f}_{ji}| > \epsilon$$



Wavelet analysis

Grossmann



Meyer



Daubechies



Mallat



Continuous / orthogonal wavelets

Analyzing functions are
 translates and dilates
 of an oscillating function (of zero mean)



Well localized in both space and wavenumber

$$\tilde{f}(l, \vec{x}) = \langle \psi_{l, \vec{x}} | f \rangle$$

Continuous wavelets

$$\psi_{l, \vec{x}}(x') = \frac{1}{l^{n/2}} \psi\left(\frac{\vec{x}' - \vec{x}}{l}\right)$$

- Translates and dilates
 vary continuously
- Redundant representation

- Coefficients are easy to read
- Unfold in both space and scale
- Good for analysis

Orthogonal wavelets

$$\psi_{j, i}(x') = 2^{j/2} \psi(2^j x' - i)$$

- Translates and dilates are
 on a discrete dyadic grid
- Orthogonal basis

- Coefficients not easy to read
- sampled on a dyadic grid
- For filtering and compression

How to extract coherent structures?

Since there is **not yet a universal definition of coherent structures** which emerge out of turbulent fluctuations due to the nonlinear interactions, **we adopt an apophetic method** :
instead of defining what they are, we define what they are not.

For this we propose the minimal statement :
'Coherent structures are not noise'



Extracting coherent structures becomes a **denoising problem**,
not requiring any hypotheses on the structures themselves
but only on the noise to be eliminated.

Choosing the **simplest hypothesis** as a first guess,
we suppose we want to eliminate an **additive Gaussian white noise**,
and for this we use a **nonlinear wavelet filtering**.

*Farge, Schneider et al.,
Phys. Fluids, 15 (10), 2003*

*Azzalini, Farge, Schneider,
ACHA, 18 (2), 2005*

Denoising using wavelets

Gaussian **white noise** is by definition **equidistributed** among all **modes** and the amplitude of the coefficients is given by its r.m.s., whatever the functional basis one considers.

Therefore the **coefficients of a noisy signal whose amplitudes are larger than the r.m.s. of the noise belong to the denoised signal**. This procedure corresponds to **nonlinear filtering**.

The advantage of performing such a nonlinear filtering using the wavelet representation is that **the wavelet coefficients preserve the space locality**, since wavelets are functions localized in both physical and spectral space.

Since we do not know *a priori* the r.m.s. of the noise, we have proposed an **iterative procedure** which takes as first guess the r.m.s. of the noisy signal.

Azzalini, M. F., Schneider, 2005
Appl. Comput. Harmonic Analysis, **18** (2)

Wavelet denoising algorithm

Aphatic method :

- no hypothesis on the structures,
- *only hypothesis on the noise,*
- *simplest hypothesis as our first choice.*

Hypothesis on the noise :

$$f_n = f_d + n$$

- n Gaussian white noise,
- $\langle f_n^2 \rangle$ variance of the noisy signal,
- N number of coefficients of f_n .

Wavelet decomposition :

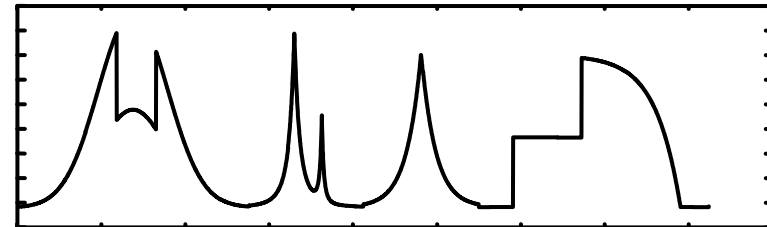
$$\tilde{f}_{ji} = \langle f | \psi_{ji} \rangle \quad \begin{array}{l} j \text{ scale,} \\ i \text{ position} \end{array}$$

Estimation of the threshold :

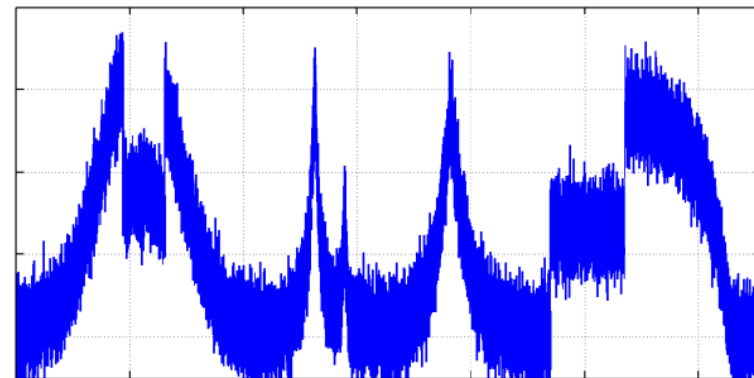
$$\varepsilon_n = \sqrt{2 \langle f_n^2 \rangle \ln(N)}$$

Wavelet reconstruction :

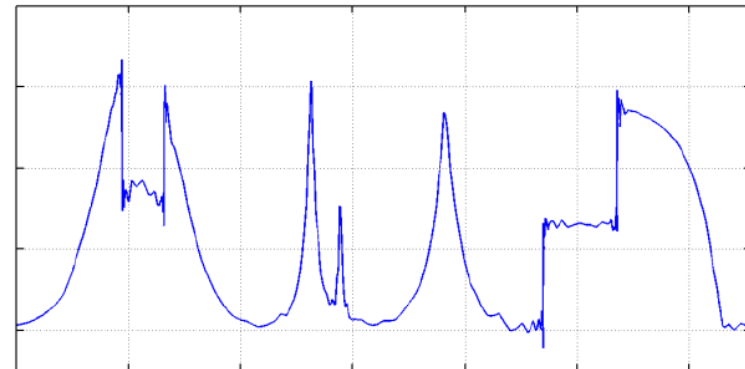
$$f_d = \sum_{j,i: |\tilde{f}_{ji}| > \varepsilon_n} \tilde{f}_{ji} \psi_{ji}$$



f



f_n

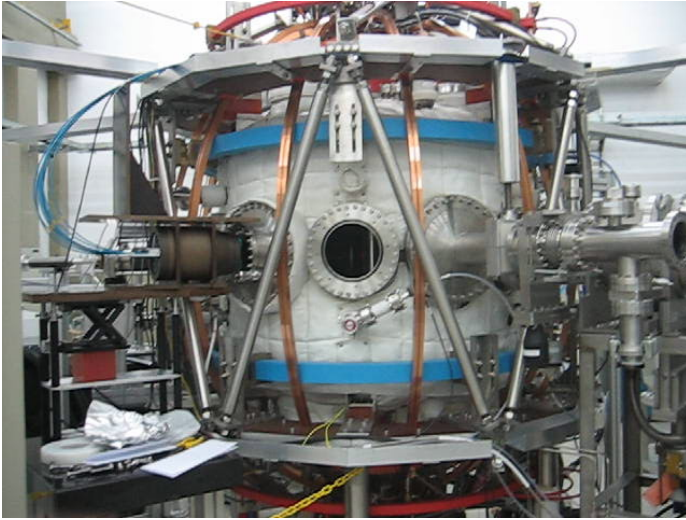


f_d

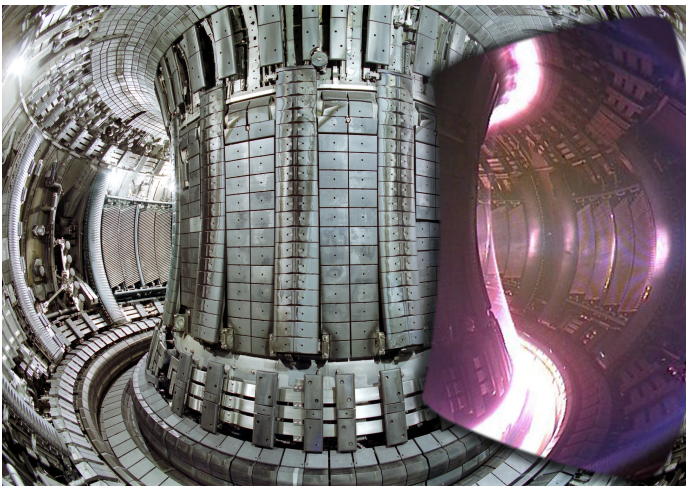
Donoho, Johnstone,
Biometrika, **81**, 1994

Azzalini, M. F., Schneider,
ACHA, **18** (2), 2005

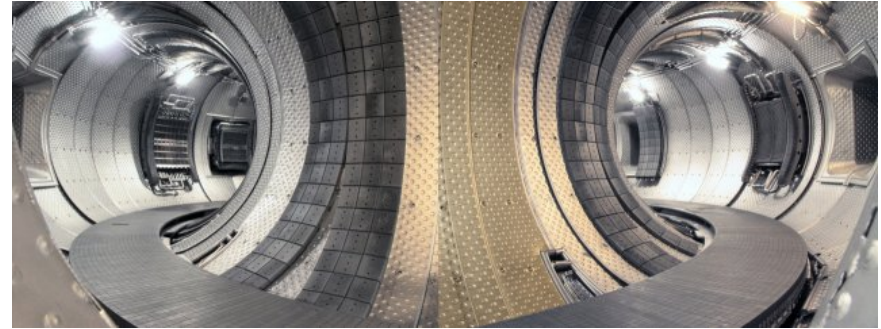
1. Application to plasma turbulence in tokamaks



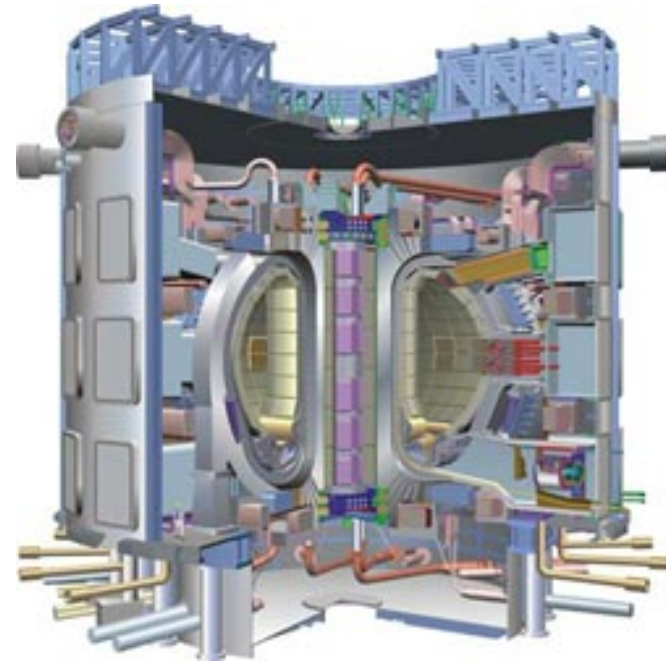
ETE, INPE (Brazil)



JET, Culham (Europe)



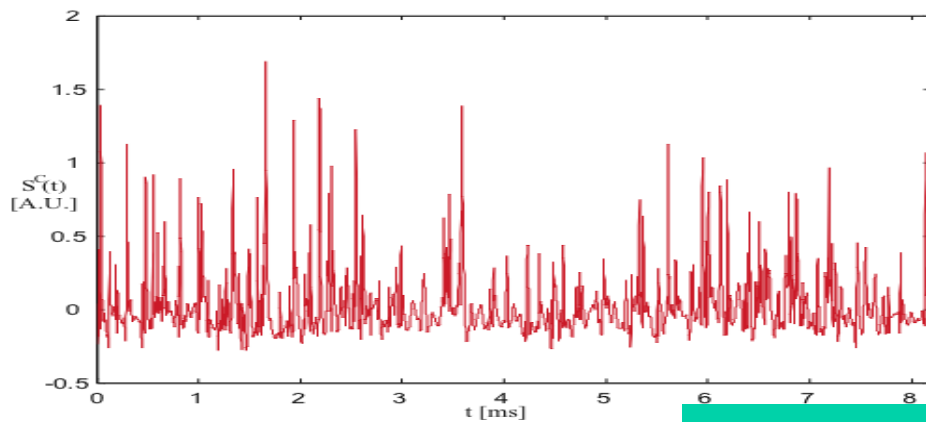
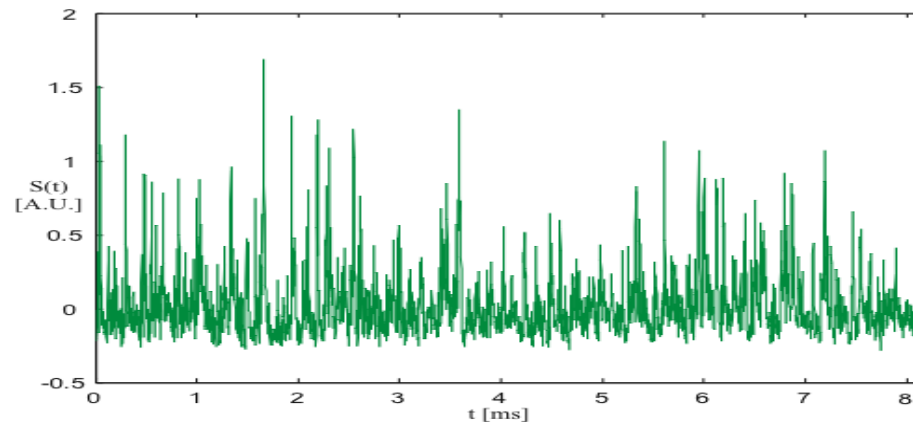
Tore-Supra, Cadarache (France)



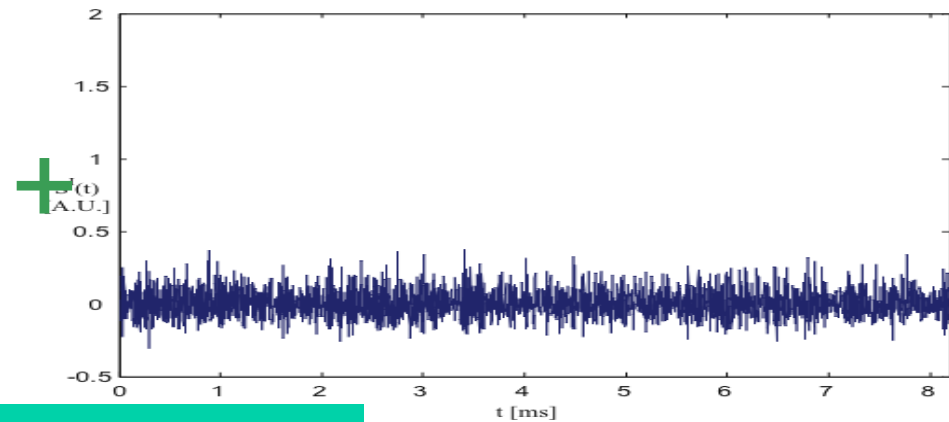
ITER (World)

Extraction of coherent structures SOL

Ion density fluctuations measured by a fast reciprocating Langmuir probe in the SOL of the tokamak Tore Supra
(Pascal Devynck, Tore-Supra, CEA-Cadarache)



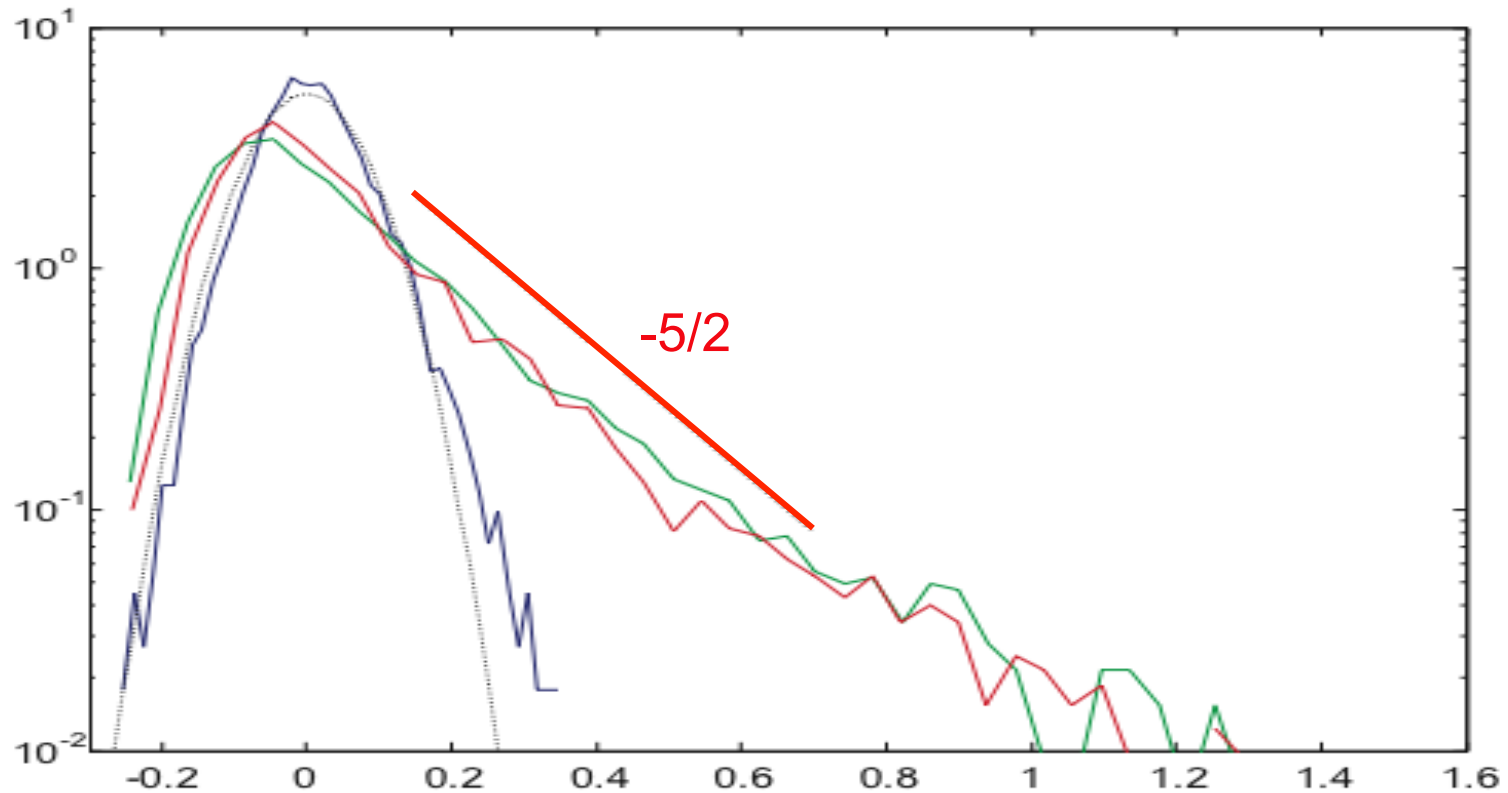
Coherent



Incoherent

Farge, Schneider & Devynck
Phys. Plasmas, 13, 042304, 2006

PDF of the density fluctuations

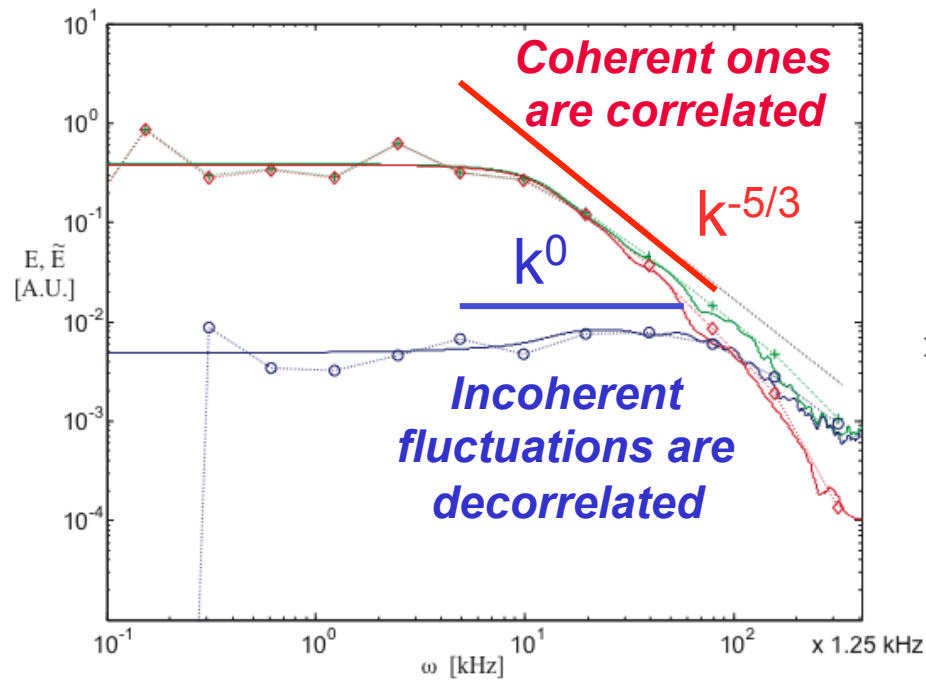


Total fluctuations = coherent + incoherent fluctuations

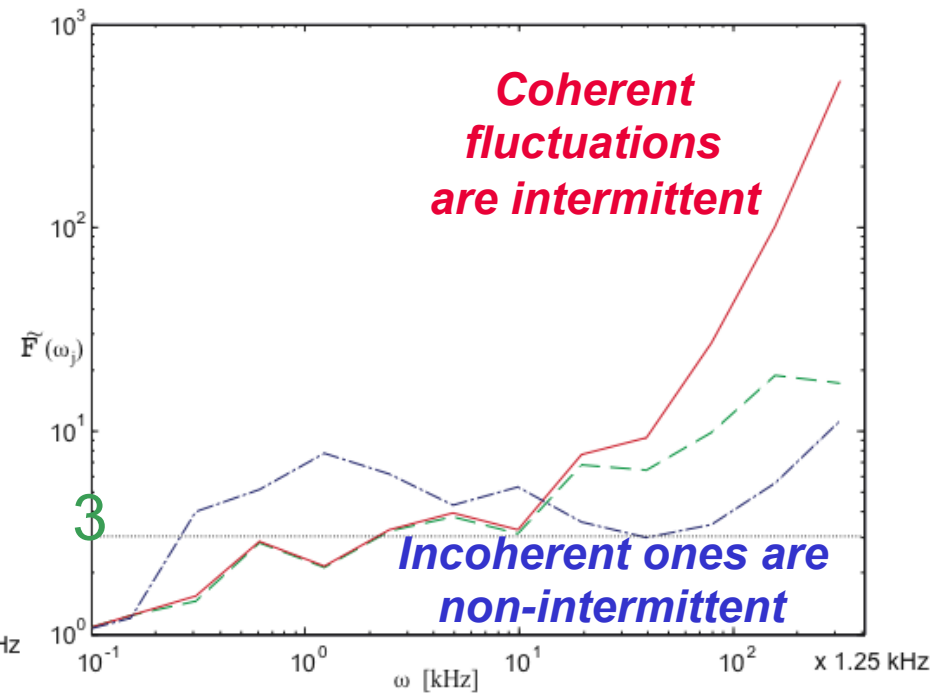
*Farge, Schneider & Devynck,
Phys. Plasmas, 13, 2006*

Correlation and intermittency

Scalogram
(stabilized periodogram)



Flatness versus scale
(from wavelet coefficients)

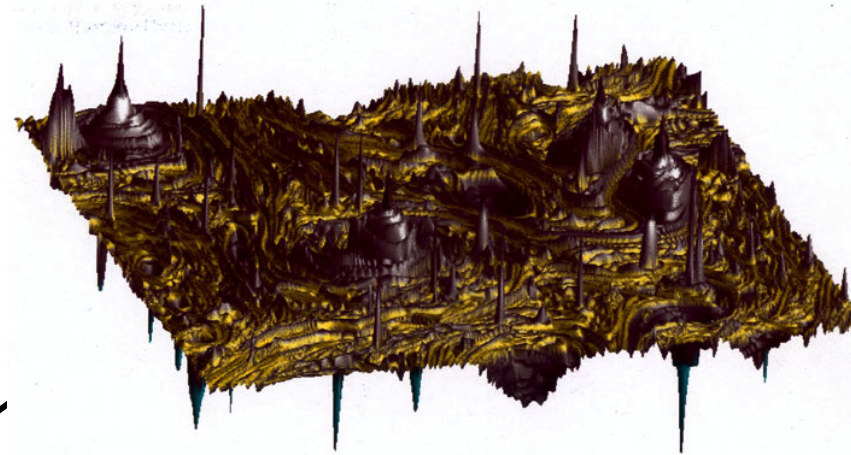


Total fluctuations = coherent + incoherent fluctuations

2. Applications in 2D fluid turbulent flows

DNS
 $N=512^2$

0.2 % of coefficients
99.8 % of kinetic energy
93.6 % of enstrophy

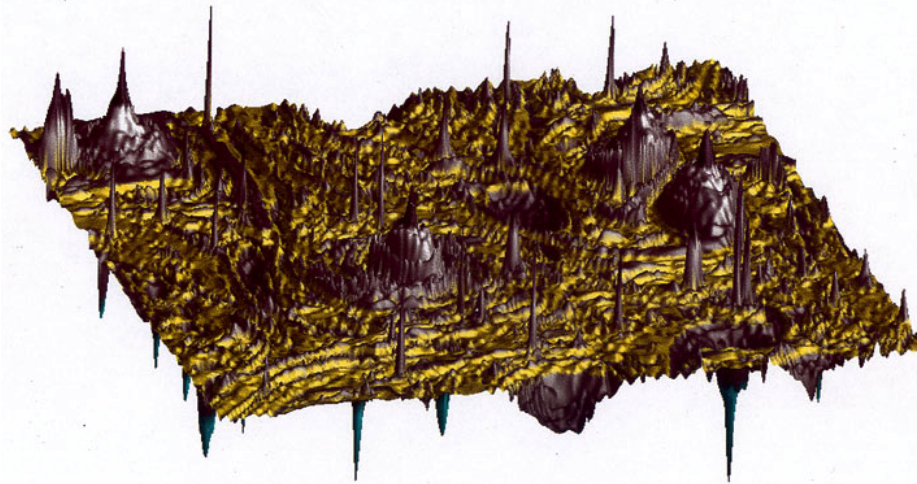


99.8 % of coefficients
0.2 % of kinetic energy
6.4 % of enstrophy

Coherent flow

Total flow

Incoherent flow



+



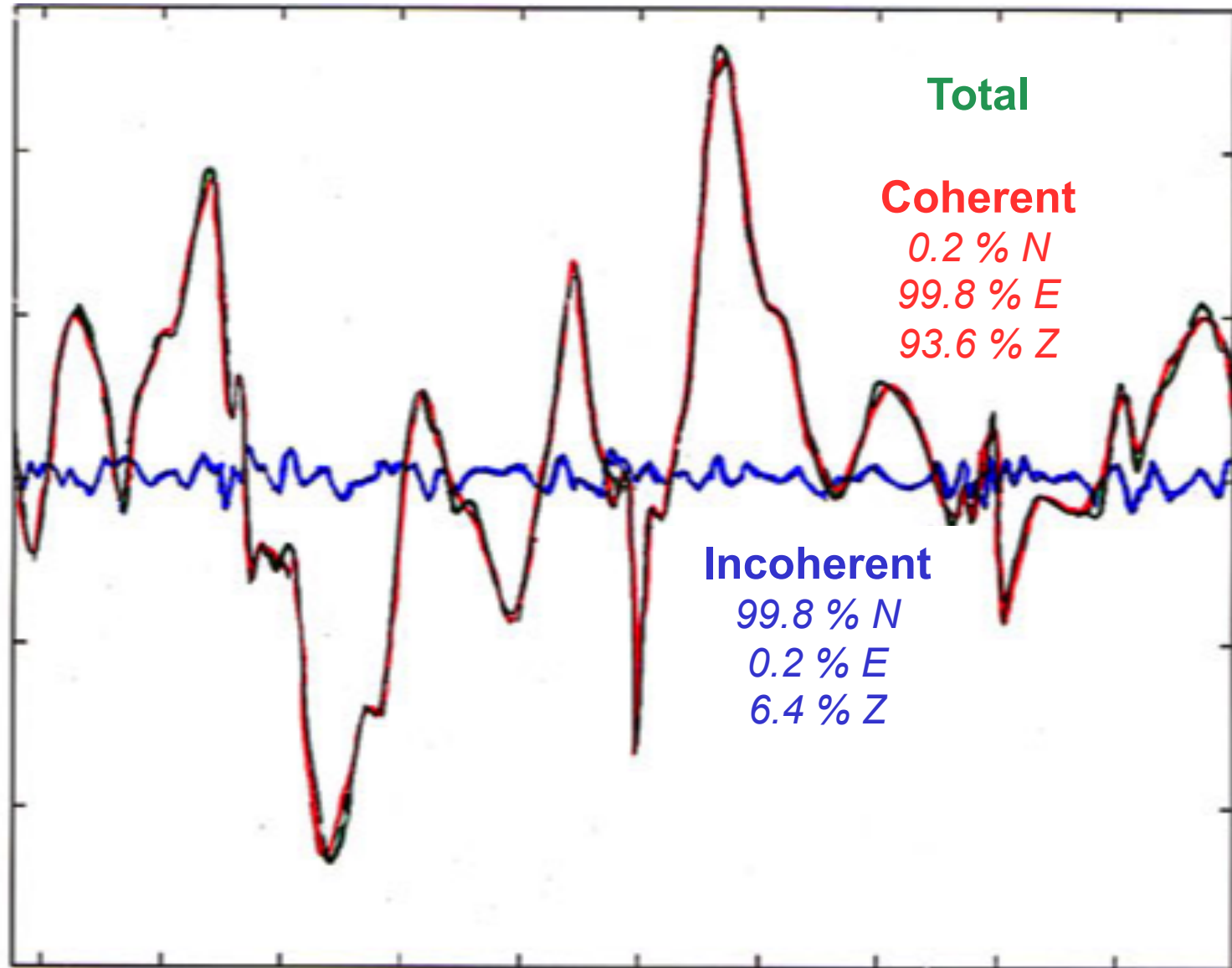
$+\omega_{min}$  $+\omega_{max}$

1D cut of the vorticity field

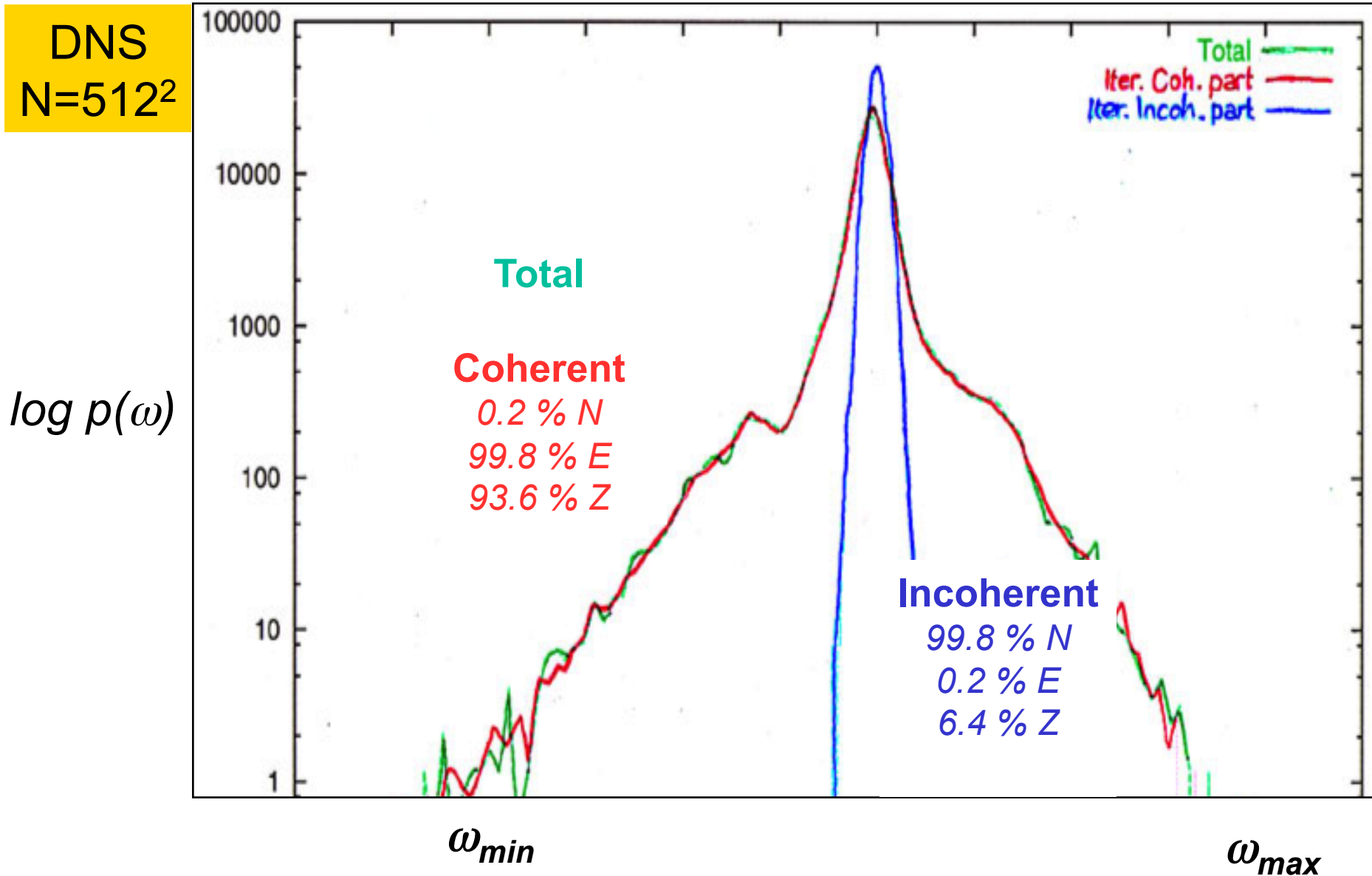
DNS
N=512²

$$\omega_t = \omega_c + \omega_i$$

$$Z_t = Z_c + Z_i$$



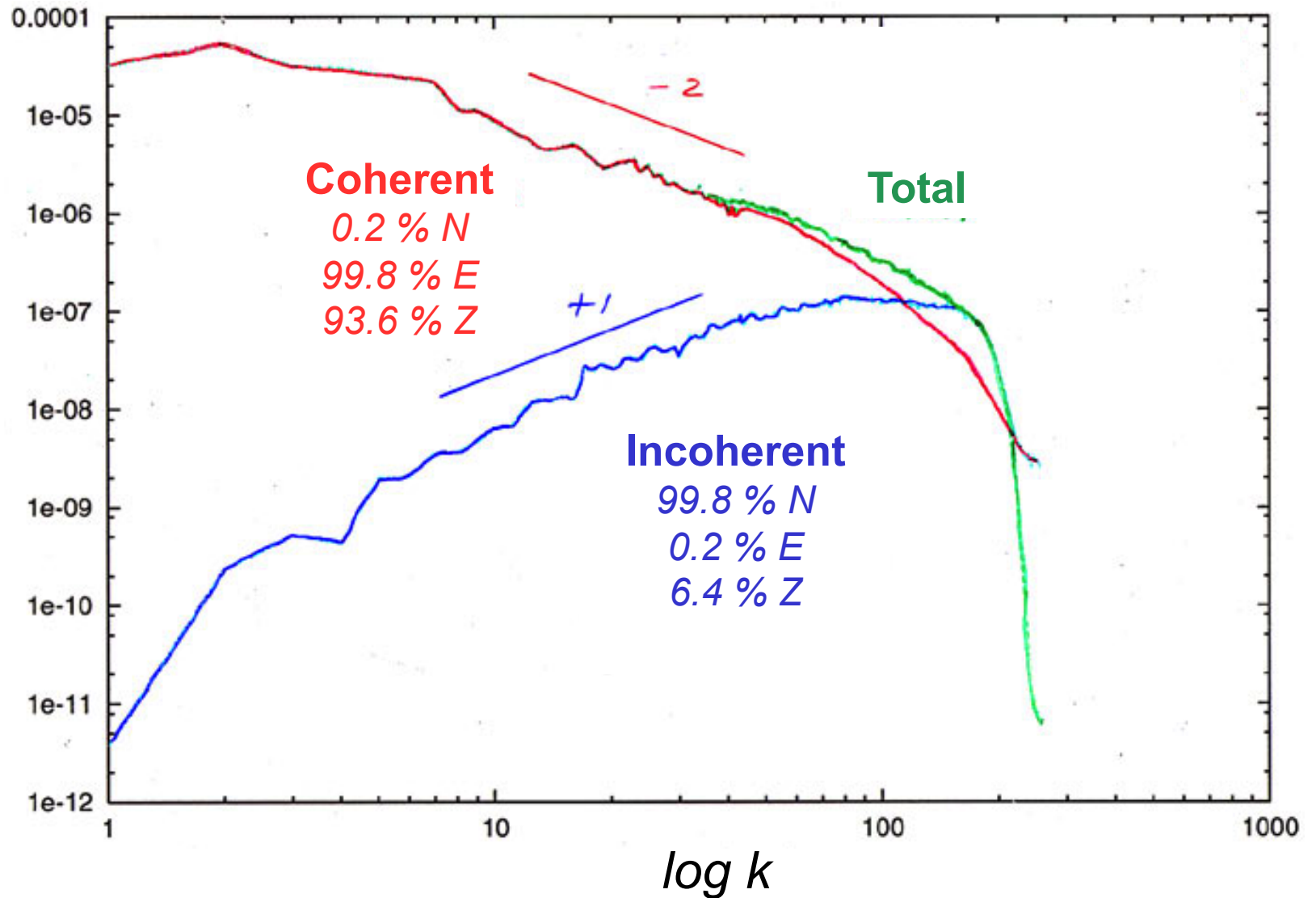
PDF of vorticity



Enstrophy spectrum

DNS
 $N=512^2$

$\log Z(k)$



A posteriori proof of coherence

DNS
N=512²

Coherent structures are
locally (in space and time)
steady solutions of Euler equation,
thus, for 2D flows :

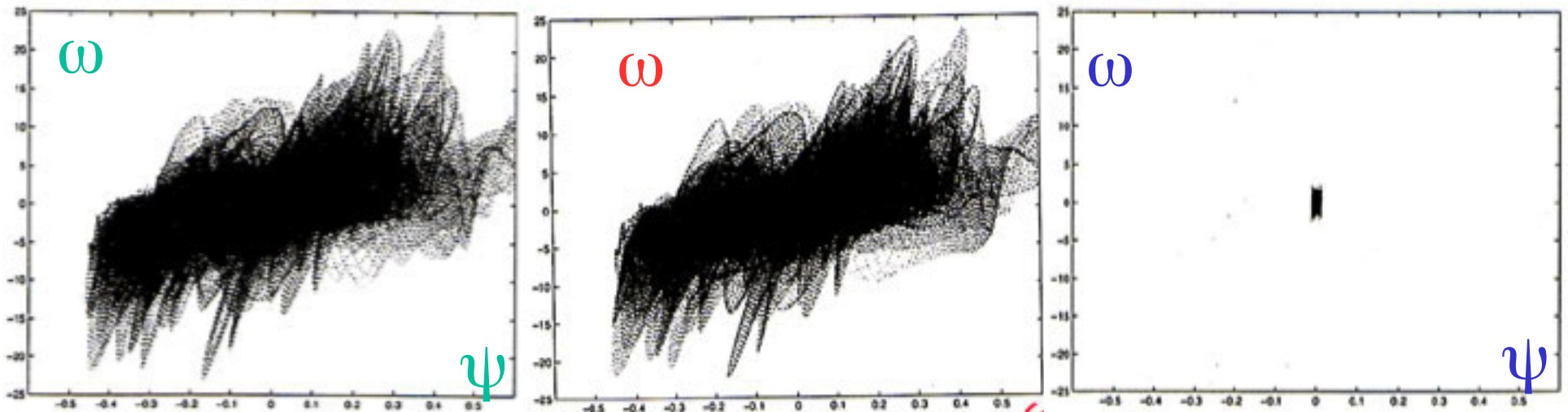
$$\omega = \sinh(\psi)$$

Arnold, 1965,
Joyce & Montgomery, 1973
Robert & Sommeria, 1991

Total

Coherent

Incoherent

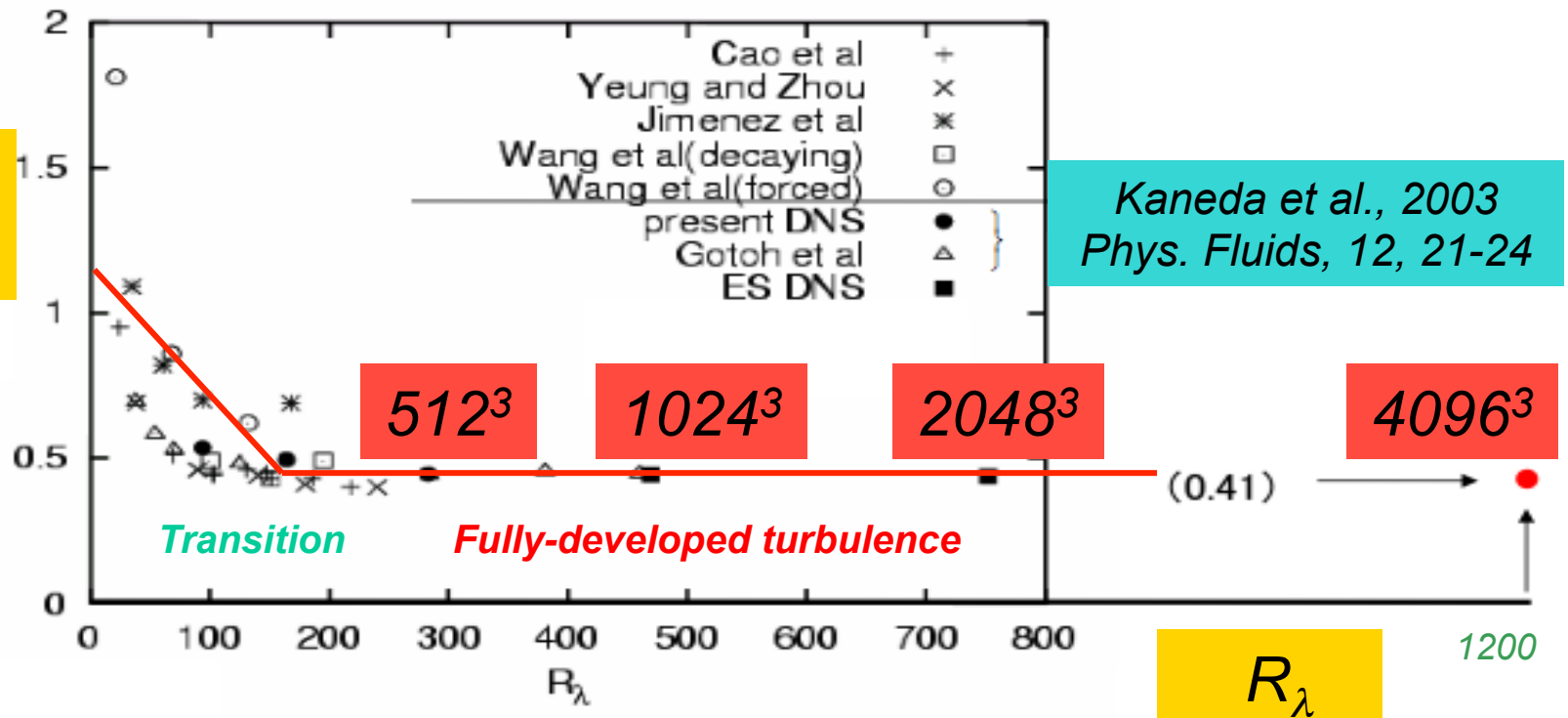


3. Application to 3D fluid turbulence

Normalized energy dissipation $\alpha \rightarrow ?$

$$\alpha = \epsilon L / u'^3 \quad \text{as } \nu \rightarrow 0, \text{ or } Re \rightarrow \infty$$

Dissipation rate α



Dissipation rate is independent of viscosity \Rightarrow turbulent dissipation
 How turbulent dissipation differs from viscous dissipation?

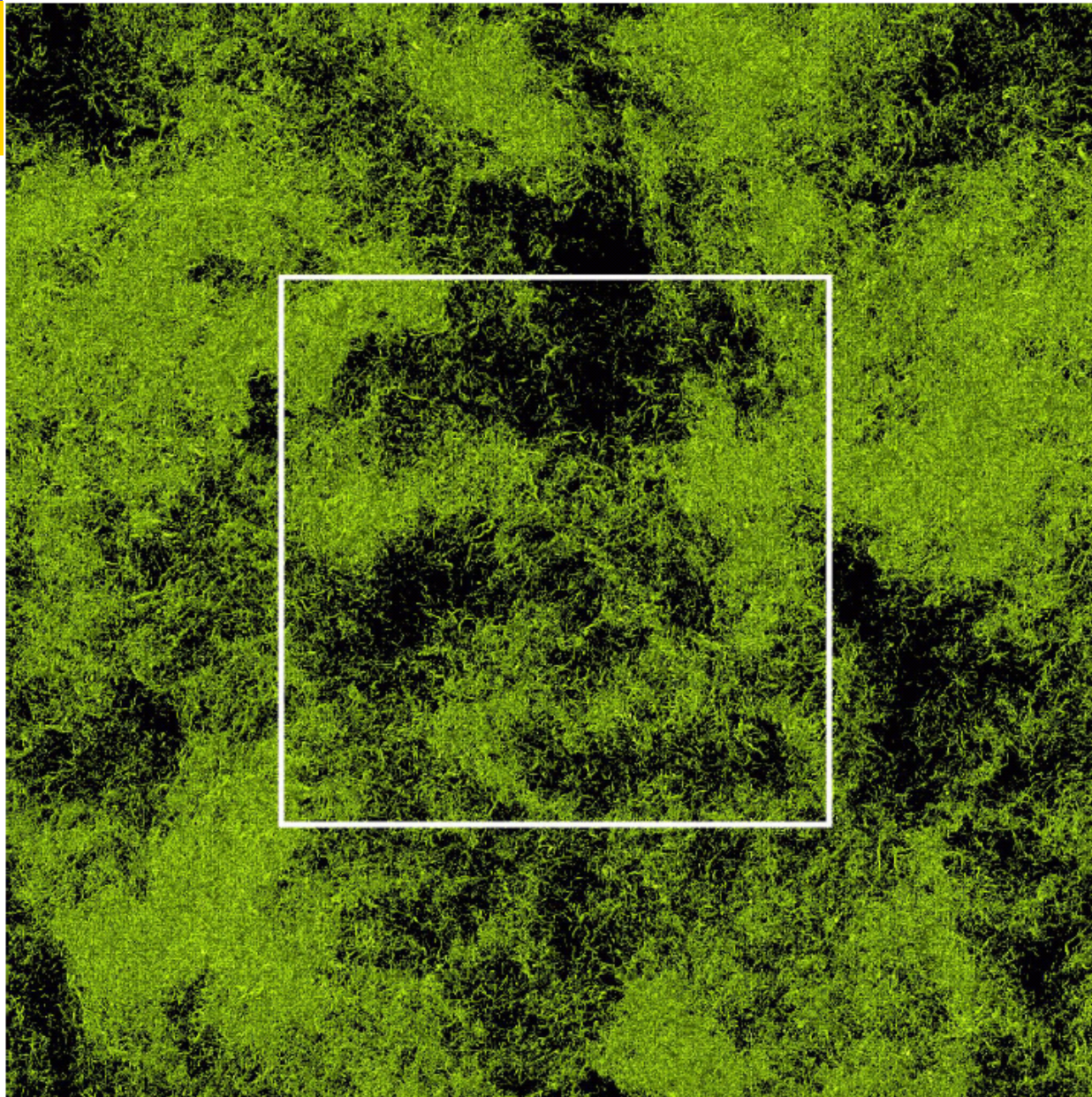
3D homogeneous isotropic turbulence

Resolution
 $N=2048^3$

L ,
integral
scale

*Computed
in 2002
on ES1
14 Tflops
10 Tbytes*

*Kaneda,
Ishihara
et al., 2003,
Phys. Fluids,
12, 21-24*

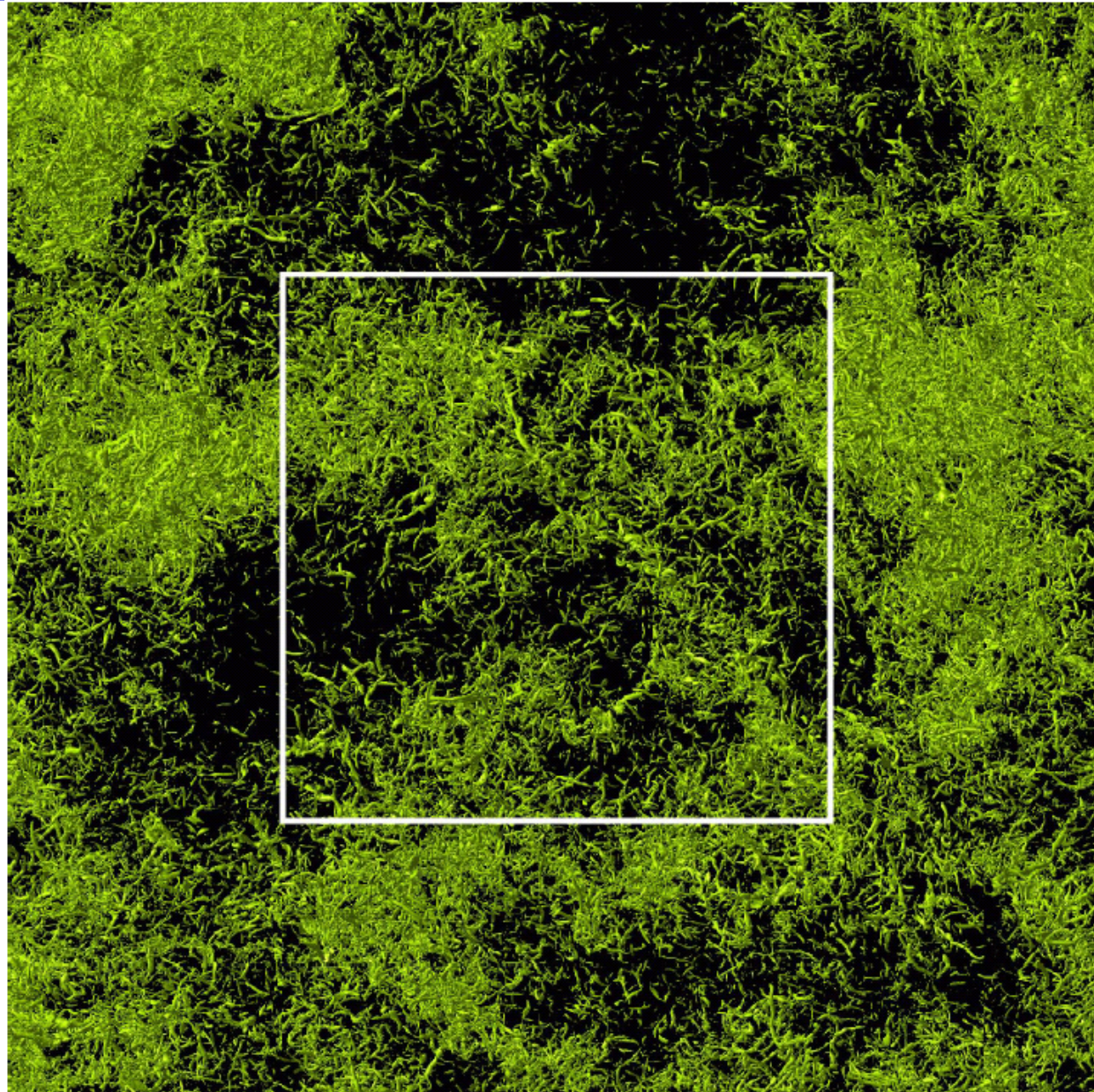


2π

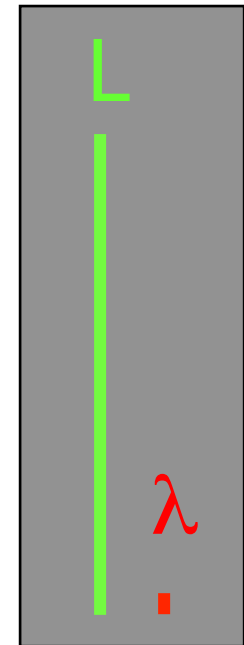
L

Zoom (sub-cube 1024^3)

Resolution
 $N=2048^3$



L ,
Integral
scale



λ ,
Taylor
micro-
scale

*Kaneda,
Ishihara
et al., 2003,
Phys. Fluids,
12, 21-24*

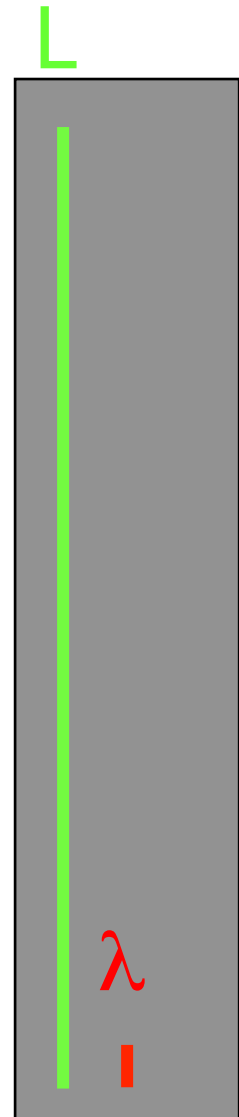
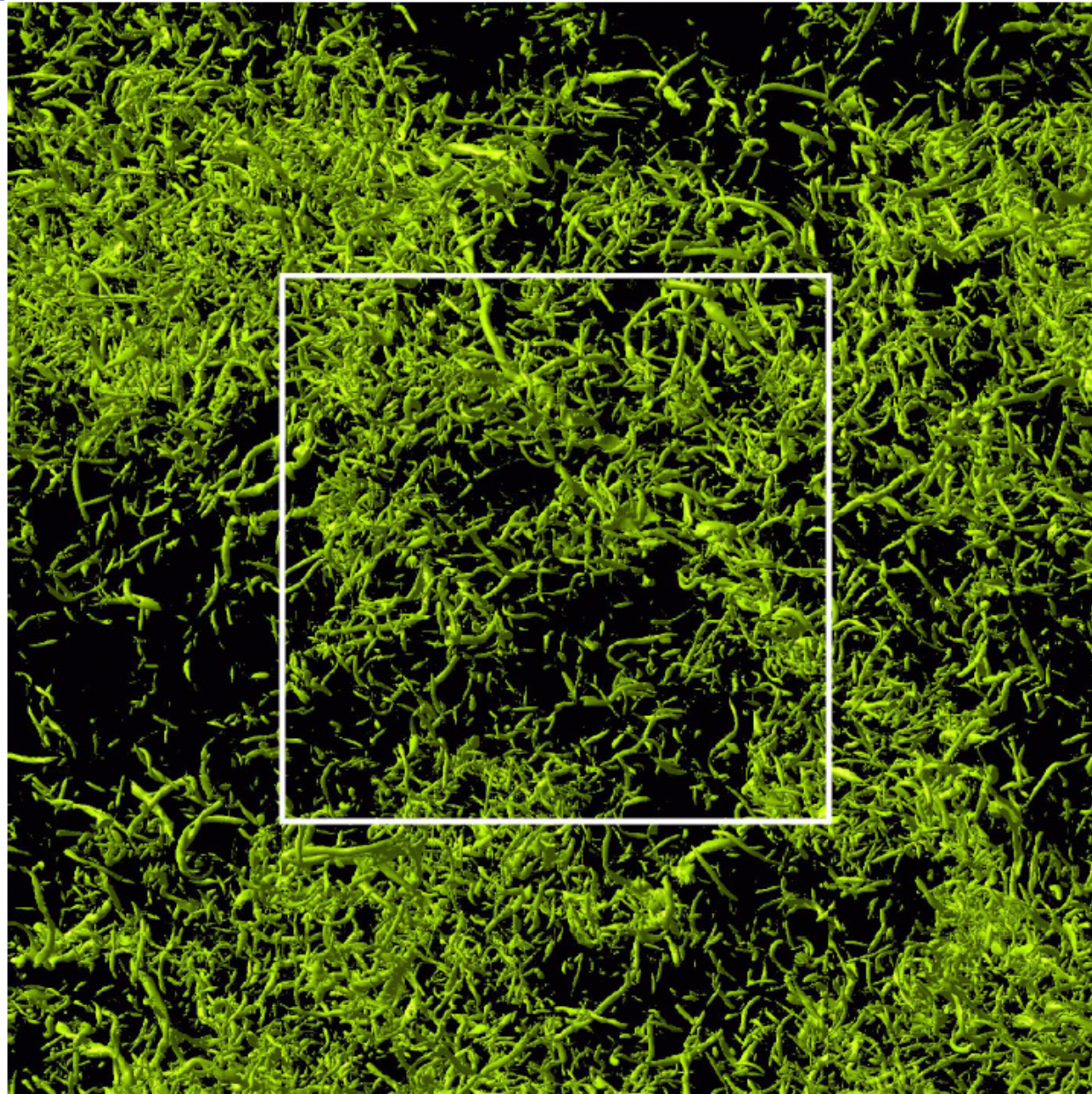
Zoom (sub-cube 512^3)

Resolution
 $N=2048^3$

L ,
integral
scale

λ ,
Taylor
macro-
scale

*Kaneda,
Ishihara
et al., 2003,
Phys. Fluids,
12, 21-24*



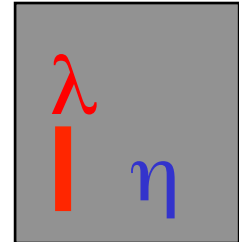
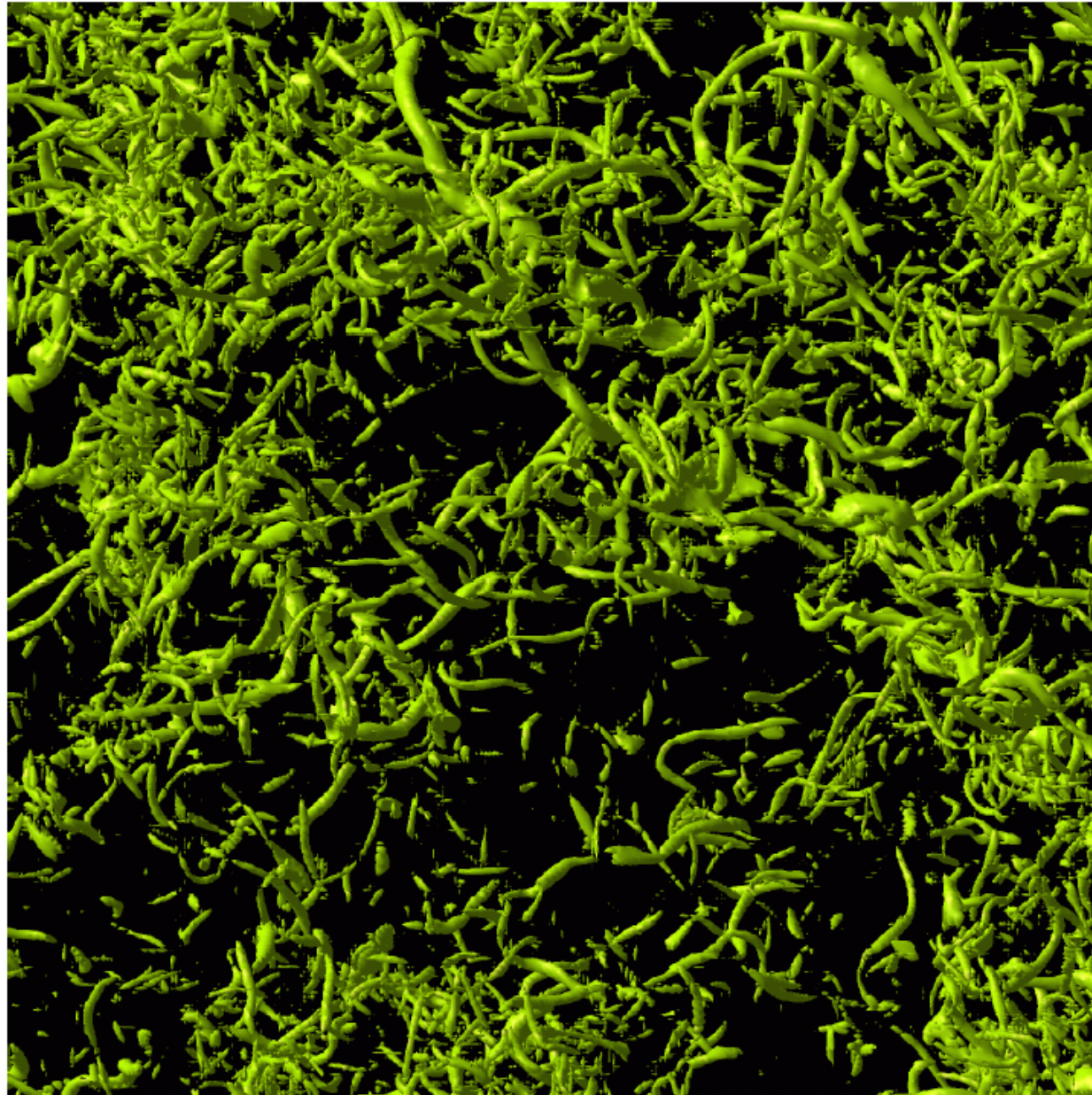
Zoom (sub-cube 256^3)

Resolution
 $N=2048^3$

λ ,
Taylor
macro-
scale

η ,
Kolmogorov
dissipative
scale

*Kaneda,
Ishihara
et al., 2003,
Phys. Fluids,
12, 21-24*



Zoom (sub-cube 128^3)

DNS
 $N=2048^3$

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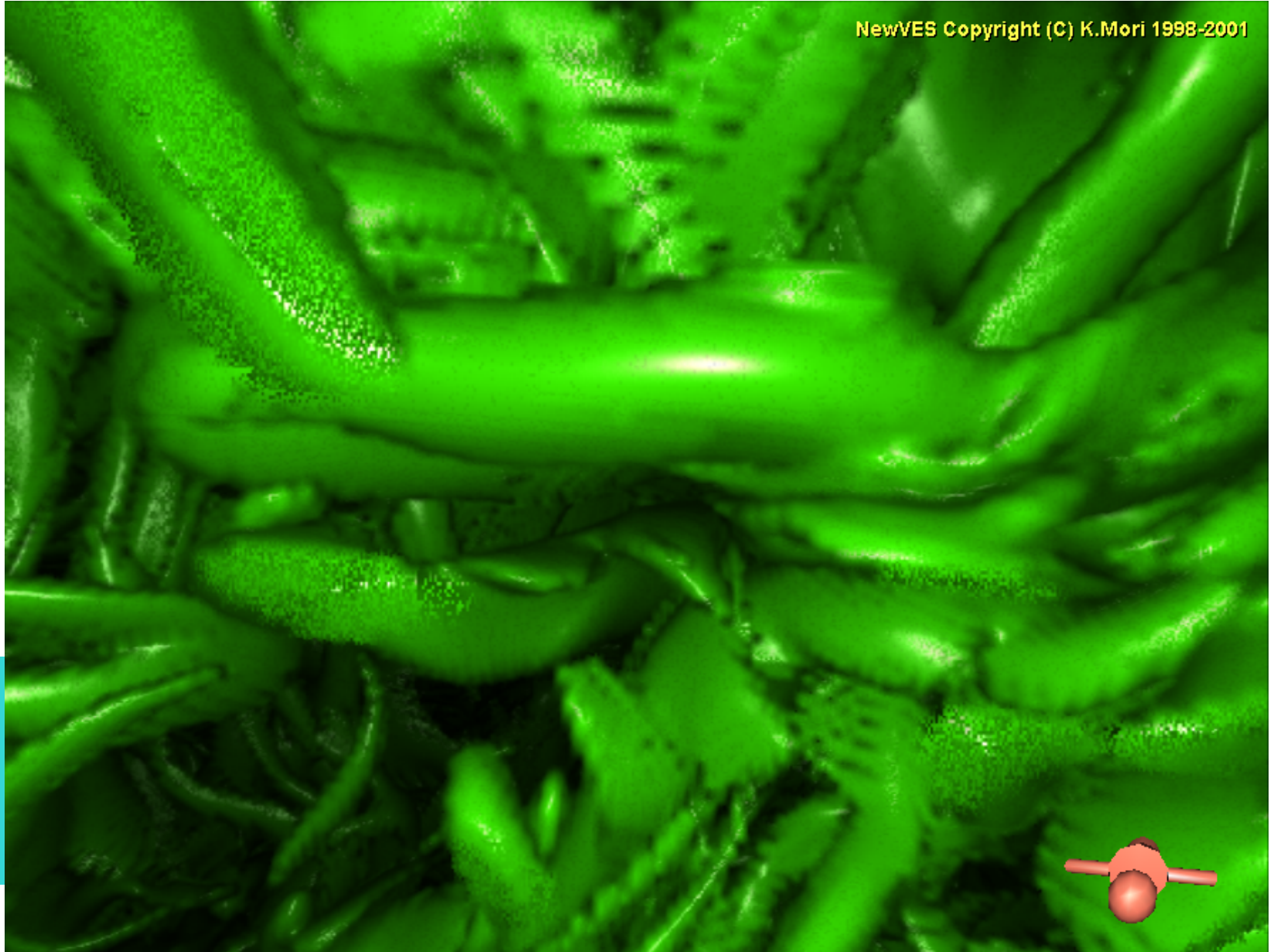


*Kaneda,
Ishihara
et al., 2003,
Phys. Fluids,
12, 21-24*

Zoom (sub-cube 64^3)

DNS
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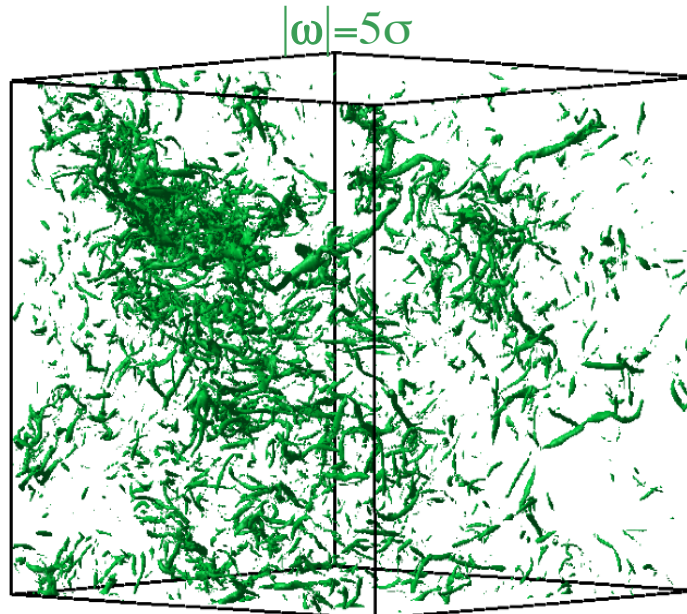
*Kaneda,
Ishihara
et al., 2003,
Phys. Fluids,
12, 21-24*

Extraction of coherent structures in 3D flows

DNS
 $N=2048^3$

Coherent vorticity

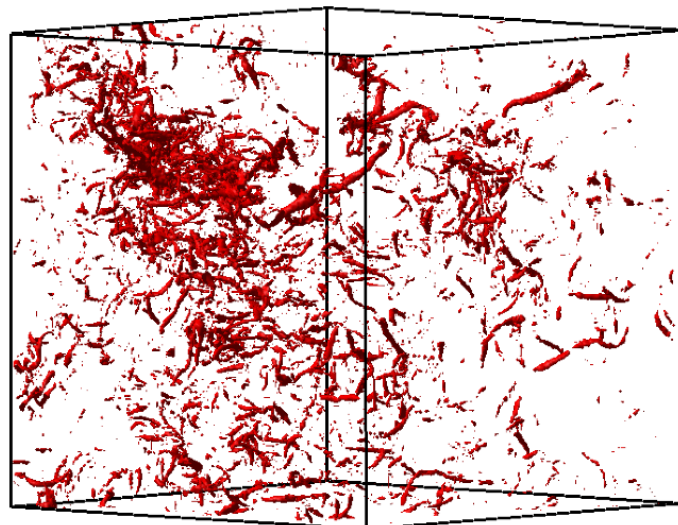
2.6 % N coefficients
80% enstrophy
99% energy



with $\sigma=(2Z)^{1/2}$

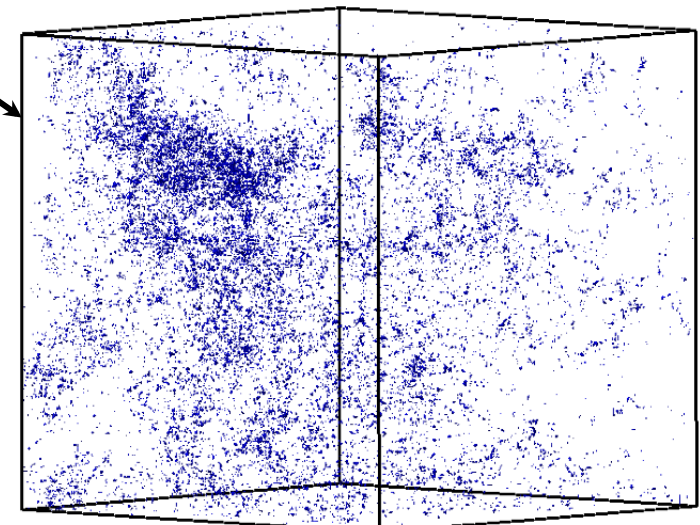
Incoherent vorticity

97.4 % N coefficients
20 % enstrophy
1% energy



Total vorticity

+

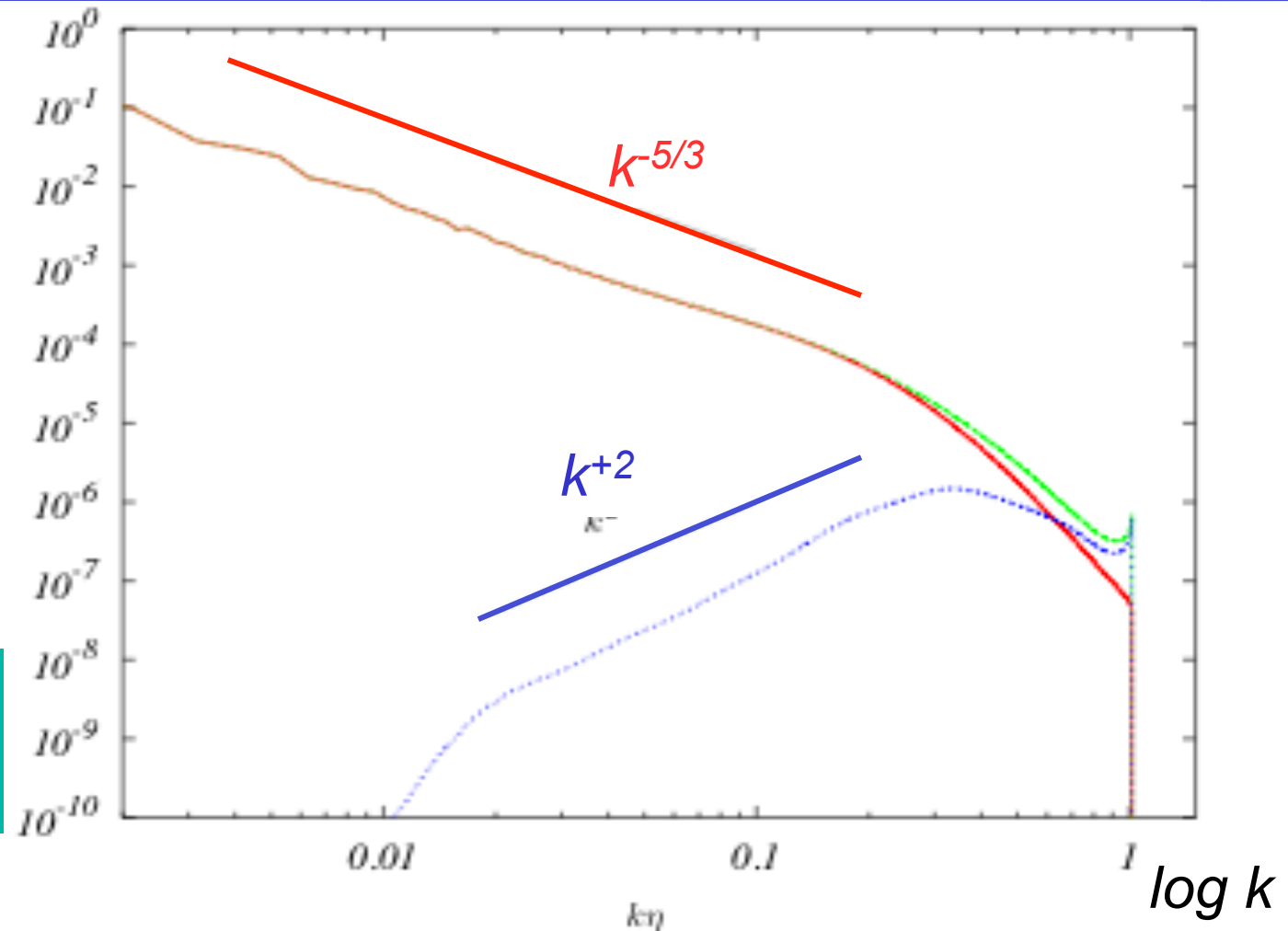


Okamoto, Yoshimatsu,
Schneider, Farge,
Kaneda, 2007,
Phys. Fluids, **19**, 1159

Energy spectrum

DNS
 $N=2048^3$

$\log E(k)$



Okamoto, Yoshimatsu,
Farge, Schneider,
Kaneda, 2007
Phys. Fluids, 19(11)

2.6 % N coefficients
80% enstrophy
99% energy

Multiscale Coherent
 $k^{-5/3}$ scaling, i.e.
long-range correlation

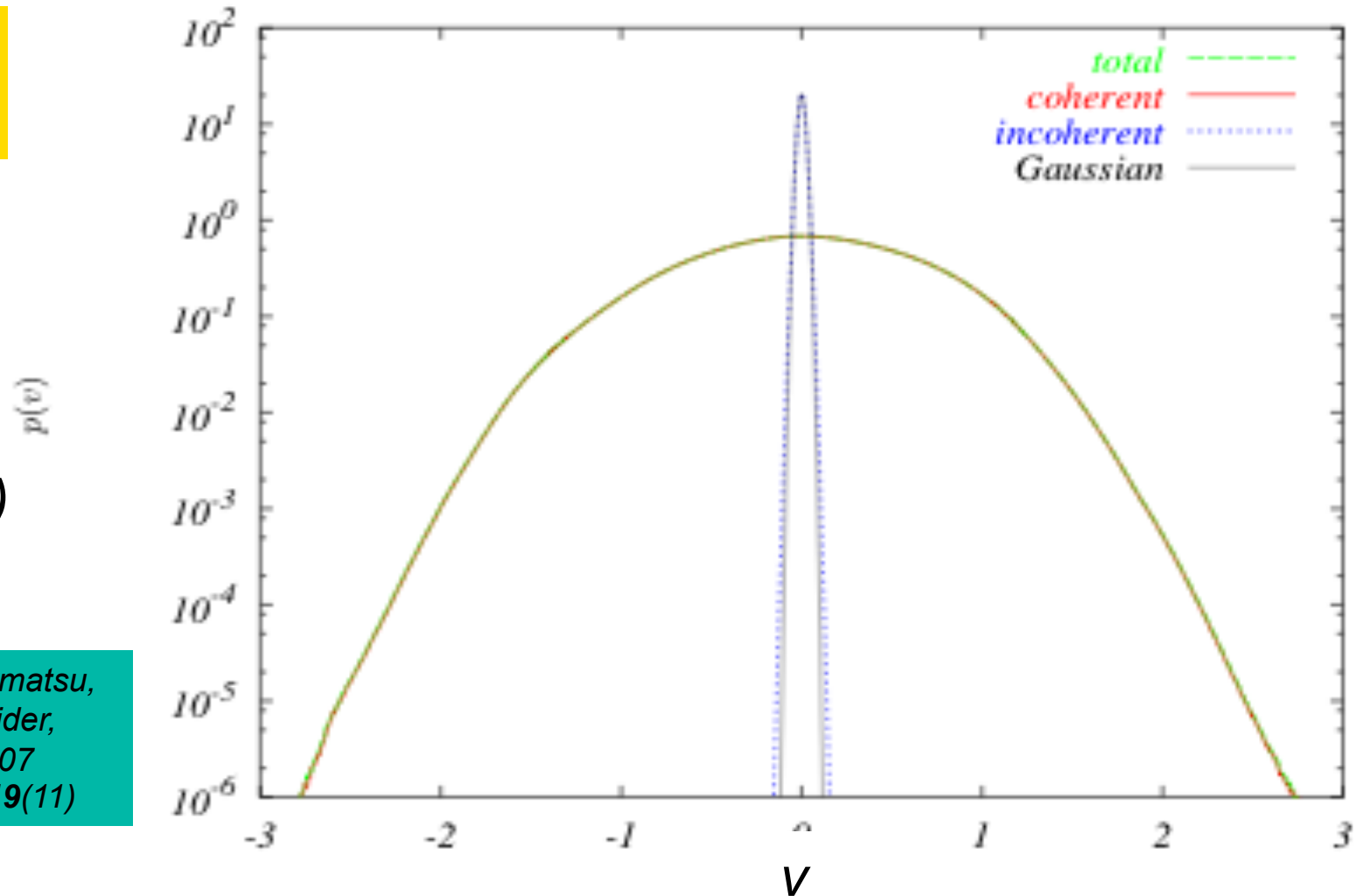
Multiscale Incoherent
 k^{+2} scaling, i.e.
energy equipartition

PDF of velocity

DNS
 $N=2048^3$

$\log p(v)$

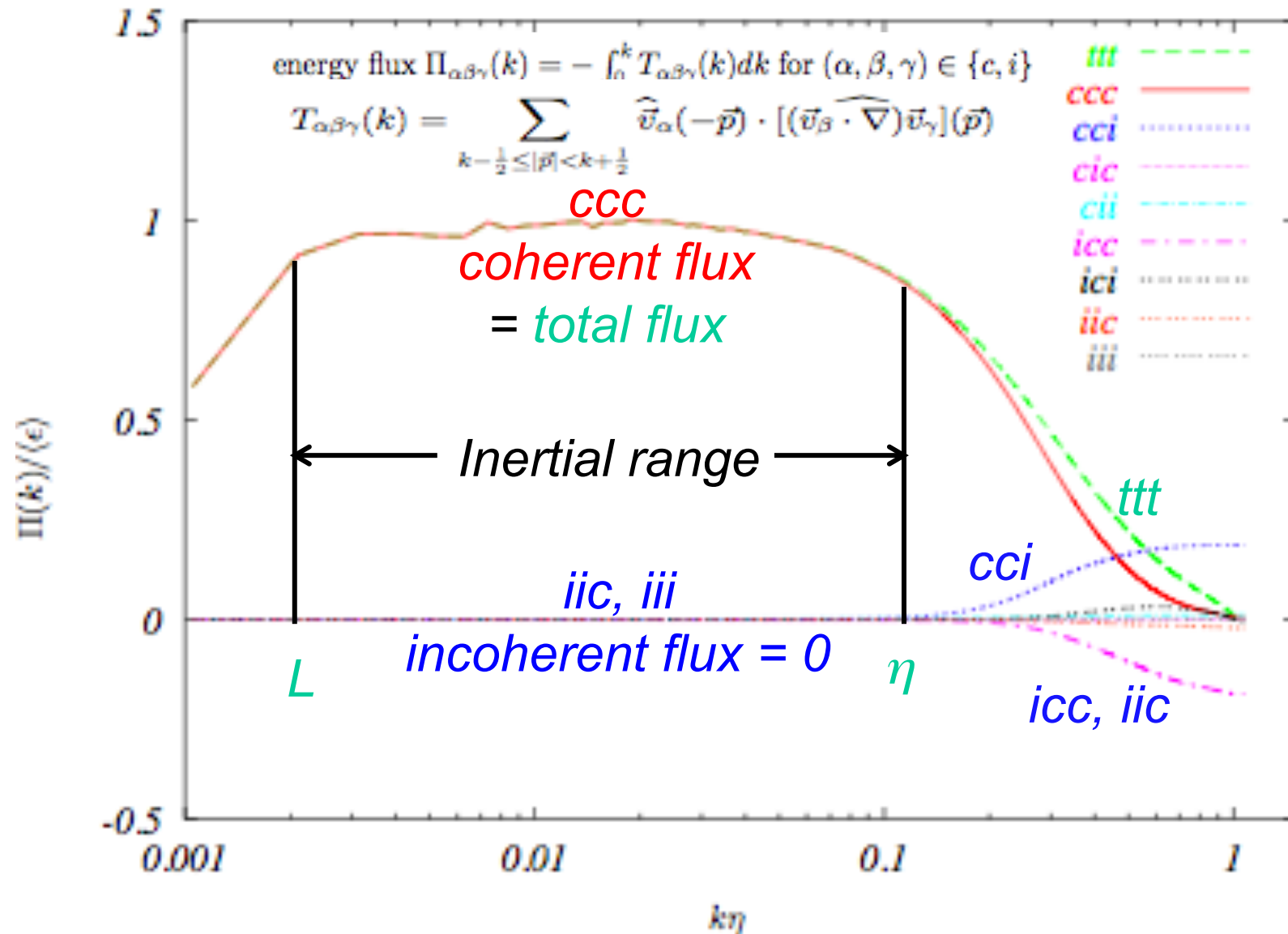
Okamoto, Yoshimatsu,
Farge, Schneider,
Kaneda, 2007
Phys. Fluids, 19(11)



2.6 % N coefficients
99% energy

The total and coherent flows have the same extrema.
The incoherent flow has a Gaussian PDF,
therefore its effect should be easy to model

Nonlinear transfers and energy fluxes



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Marie Farge, 1992

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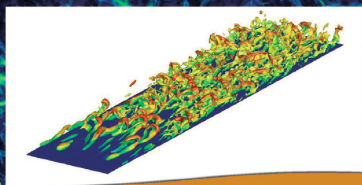
Encyclopedia of Mathematical Physics, Elsevier, 408-419

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