

# A Wavelet-based Multifractal Analysis for scalar and vector fields: Application to developed turbulence

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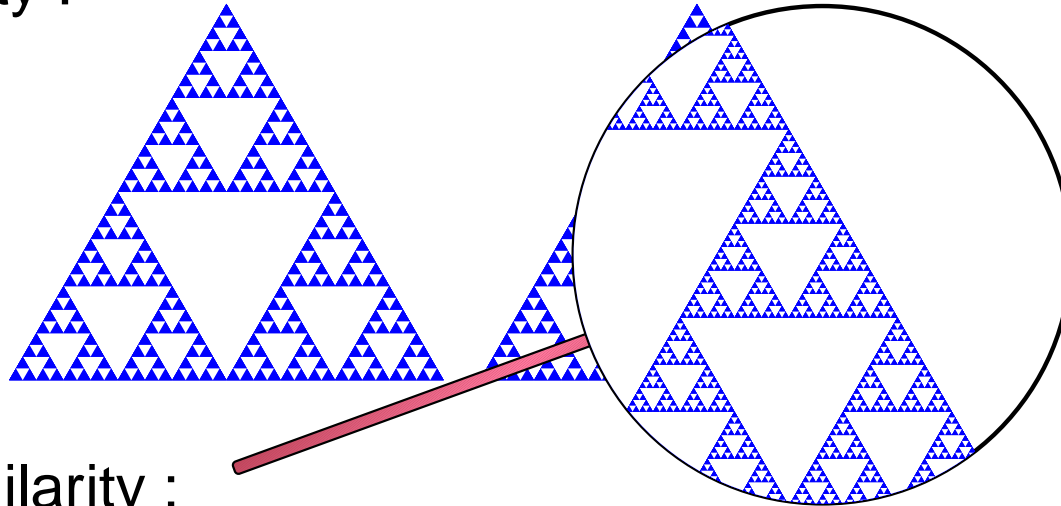
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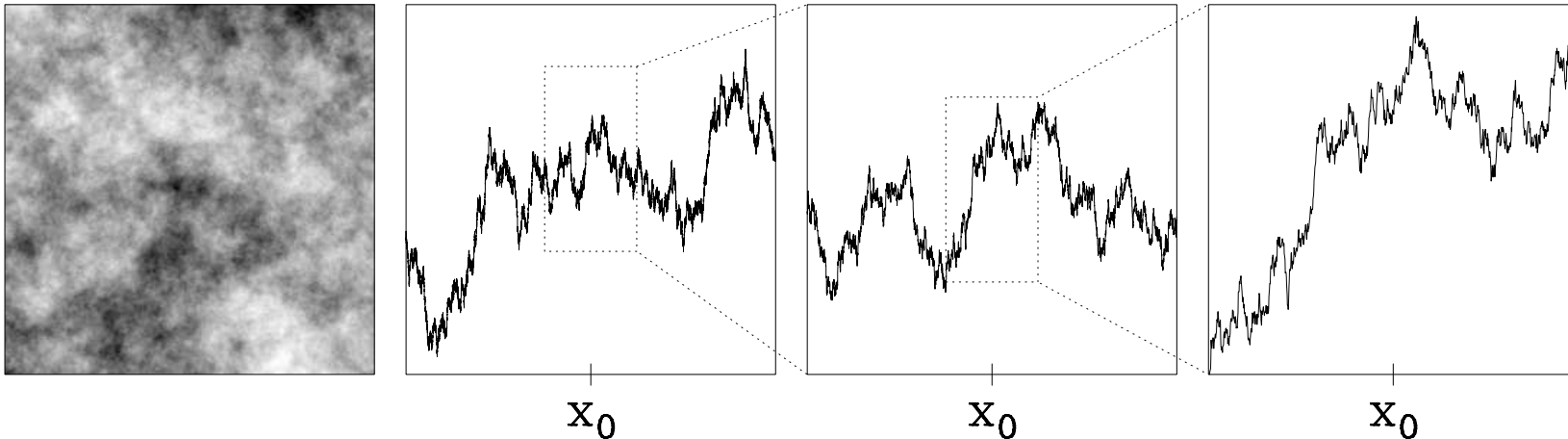
- 👉 Introduction
- 👉 Characterization of multifractal images with the 2D WTMM method
- 👉 Generalizations of the WTMM method
  - from 2D to 3D
  - from scalar to vector
- 🟢 Application to **developped turbulence** (direct numerical simulations)
  - 3D scalar case : dissipation field and enstrophy field
  - 3D vector case : velocity field and vorticity field
- 🟢 Application to **galaxy distribution** (simulation data)

# Fractal Objects : self-similarity

● exact self-similarity :



● statistical self-similarity :



$$f(x_0 + \lambda u) - f(x_0) \sim \lambda^{h(x_0)} (f(x_0 + u) - f(x_0))$$

# Self-similarity and Wavelet Transform

- **Wavelet Transform** : a mathematical microscope to study fractal objects self-similarity properties

$$\mathbf{T}_\psi[f](a, b) = \langle f, \psi_{a,b} \rangle = \frac{1}{\sqrt{a}} \int_{-\infty}^{+\infty} dx f(x) \psi^*\left(\frac{x-b}{a}\right)$$

- local self-similarity (Hölder exponent) can be seen in the wavelet transform coefficients under scaling laws (Jaffard, Mallat et coll., Holschneider and Tchamitchian) :

$$\mathbf{T}_\psi[f](a, x_0) \sim a^{h(x_0) + \frac{1}{2}}, \quad a \rightarrow 0^+$$

- **statistical** description of self-similarity :

**WTMM method** (Wavelet Transform Modulus Maxima) by Muzy, Bacry, Arnéodo (1993) for **multifractal signals**.

- many applications in bioinformatics (DNA Sequences), in turbulence, in finance, in geophysics,...

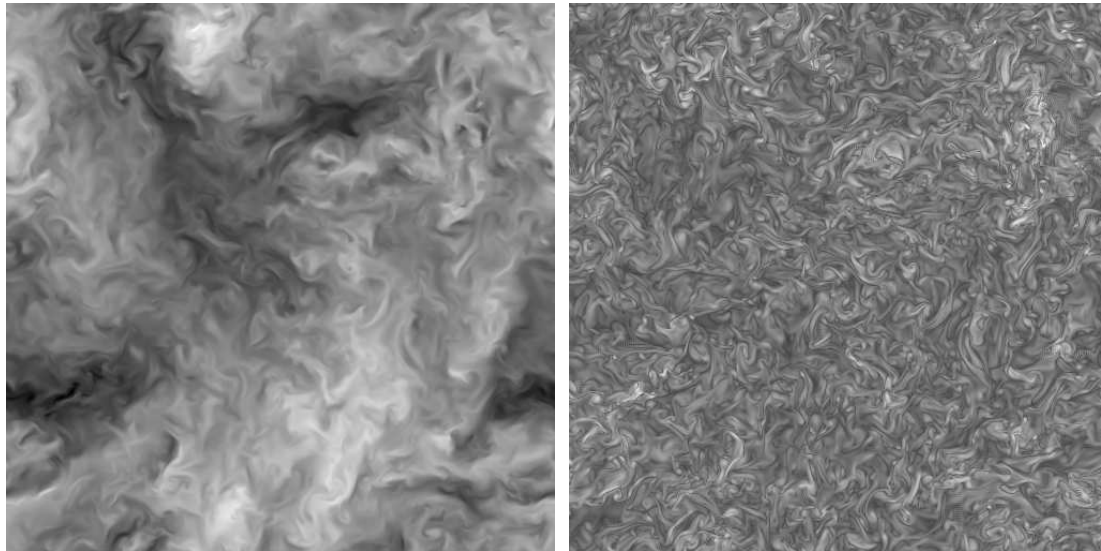
# classical multifractal formalism comes from turbulence

- The Navier-Stokes equations:

$$\partial_t \mathbf{u} + \mathbf{u} \cdot \nabla \mathbf{u} = -\frac{1}{\rho} \nabla p + \mathbf{f} + \nu \Delta \mathbf{u}, \quad +\nabla \cdot \mathbf{u} = 0 + BC + CI$$

- turbulent regime :  $\|\mathbf{u} \cdot \nabla \mathbf{u}\| \gg \|\nu \Delta \mathbf{u}\|$

- signal **highly disorganized** and structures at all scales, **unpredictable** as for details



- ☞ Statistical tools required

# Multifractal description of intermittency in turbulence

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# Multifractal description of intermittency in turbulence

● **classical multifractal formalism** comes from turbulence

☞ multifractal model for velocity : longitudinal **structure functions** based on (1D) velocity increments :

$$S_p(l) = \langle (\mathbf{e} \cdot \delta \mathbf{v}(\mathbf{r}, l \mathbf{e}))^p \rangle \sim l^{\zeta_p}, \quad p > 0$$

☞ multifractal model for (1D **surrogate**) dissipation : RSH hypothesis

$$S_p(l) \sim \langle \varepsilon_l^{p/3} \rangle l^{p/3} \sim l^{\tau_\varepsilon(p/3) + p/3}.$$

Measure of the spectra  $\tau_\varepsilon(p)$  and  $f(\alpha)$  with the **box-counting** method.

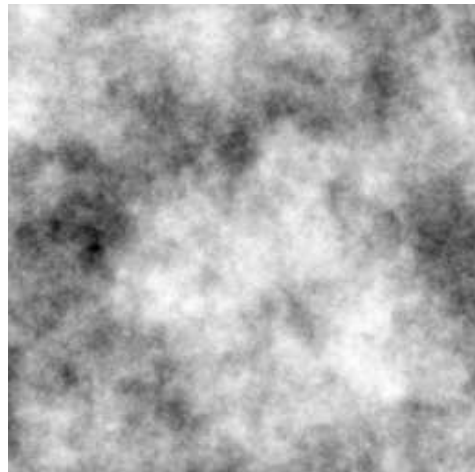
● **multifractal formalism based on WT (WTMM method)** for regularity analysis of functions, measures and distributions.

☞ generalization of the WTMM method 1D/2D → **3D**

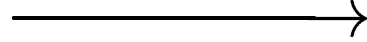
☞ generalization to **multidimensional vector fields**. First application to velocity and vorticity fields from numerical turbulent flows.

# 2D WTMM Methodology : PhD work of N. Decoster

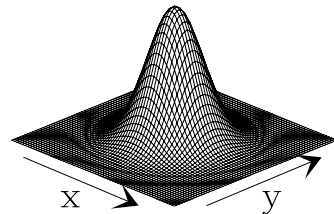
2D data :  $I$



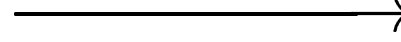
Smoothing  
scale  $a$



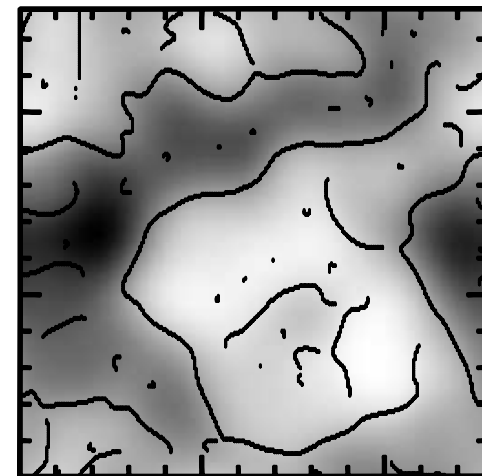
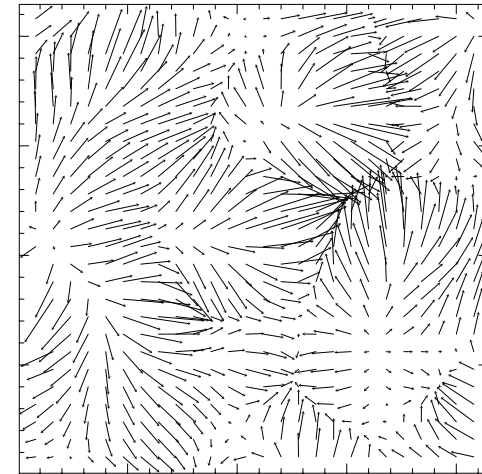
$a$



Gradient  $\nabla$



$$\begin{bmatrix} \frac{\partial}{\partial x'} & \frac{\partial}{\partial y} \end{bmatrix}$$



Wavelet Transform

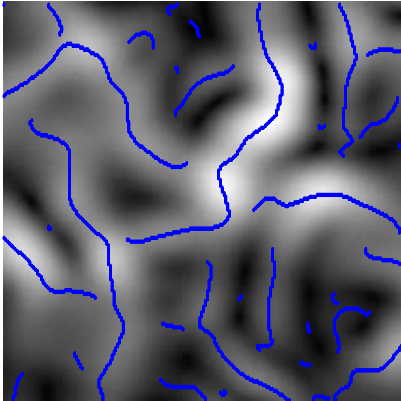
$$\mathbf{T}_\psi(\mathbf{r}, \mathbf{a}) = \begin{pmatrix} I * \frac{\partial \phi_a}{\partial x}(\mathbf{r}) \\ I * \frac{\partial \phi_a}{\partial y}(\mathbf{r}) \end{pmatrix}$$

$$\mathbf{T}_\psi(\mathbf{r}, \mathbf{a}) = \nabla(I * \phi_a)(\mathbf{r}) = (\mathcal{M}_\psi(\mathbf{r}, \mathbf{a}), \mathcal{A}_\psi(\mathbf{r}, \mathbf{a}))$$

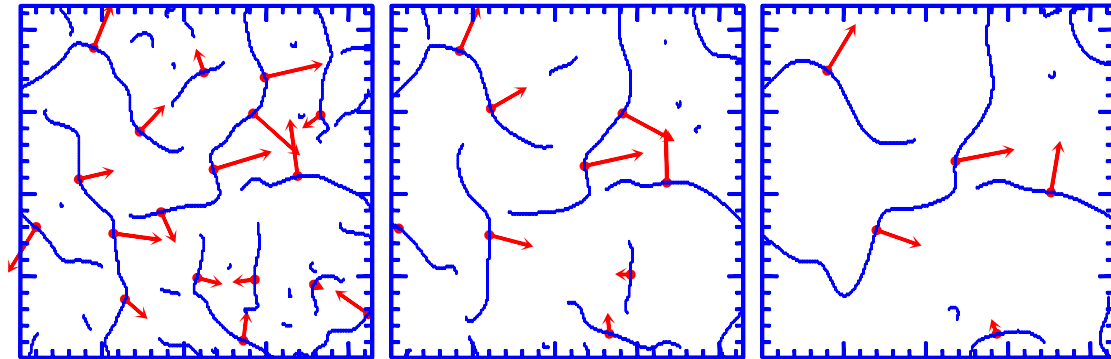


# WTMM Methodology : Skeleton

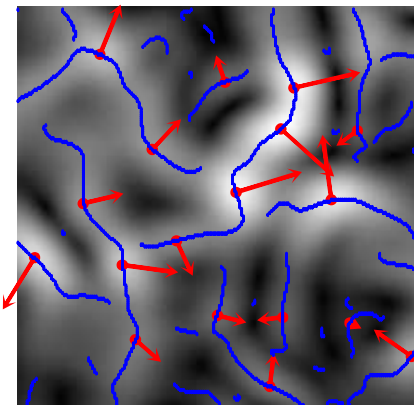
WTMM Chains



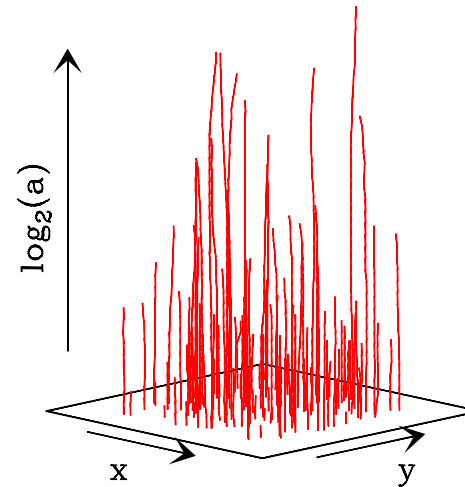
WTMM Chains at 3 different scales



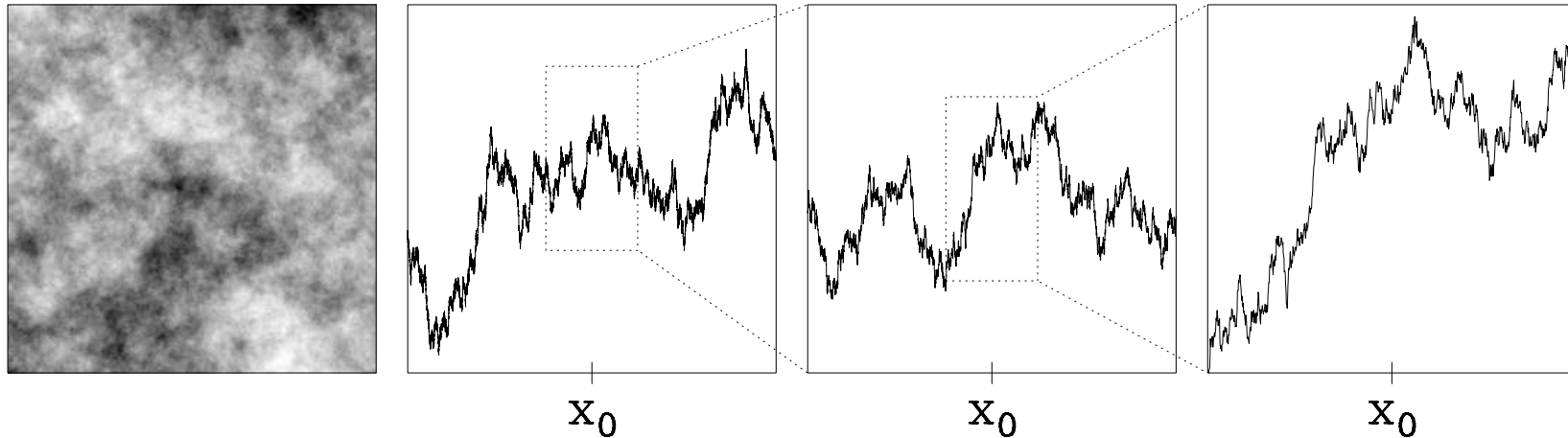
WTMMM



WTMMM linking : WT skeleton

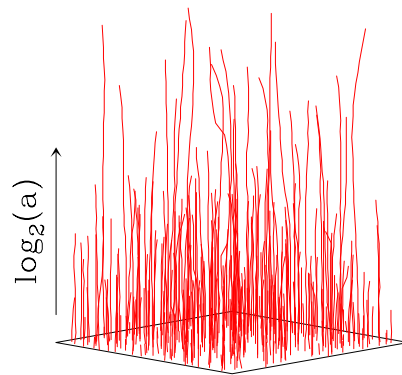


# Local roughness characterization : Hölder exponent



$$f(x_0 + \lambda u) - f(x_0) \sim \lambda^{h(x_0)} (f(x_0 + u) - f(x_0))$$

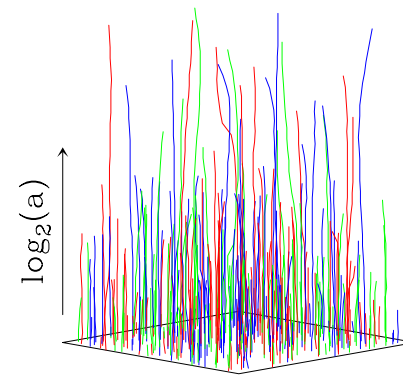
- Monofractal Image



$$\mathcal{M} \sim a^h,$$

single  $h$

- Multifractal Image

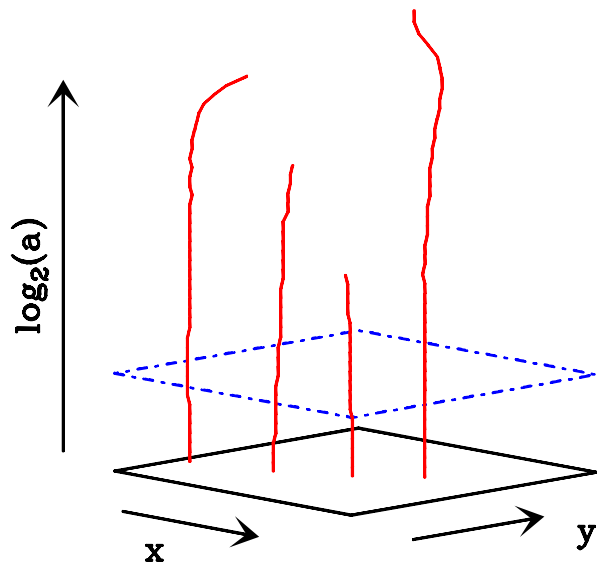


$$\mathcal{M} \sim a^h, a^h, a^h,$$
$$h \in [h_{\min}, h_{\max}]$$

# WTMM Method : multifractal formalism

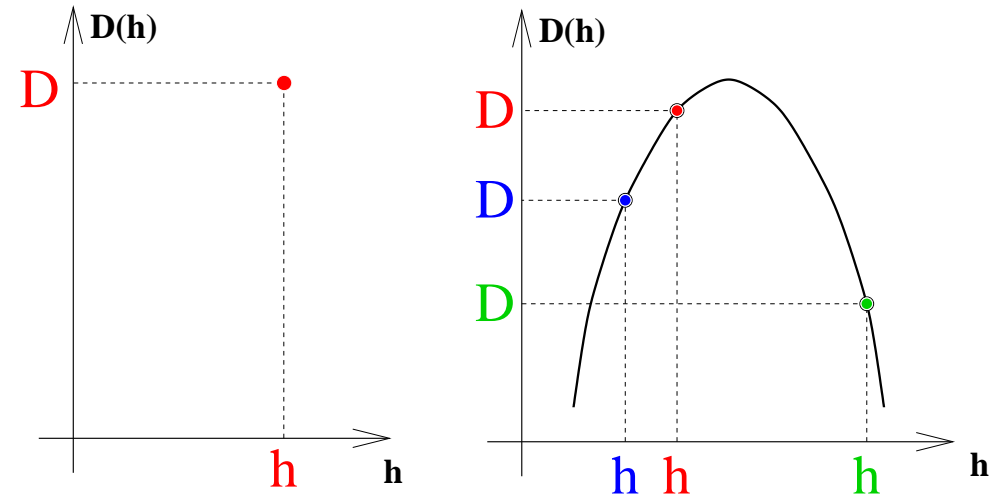
Singularity spectrum :

$$D(h) = d_H\{r \in R^d, h(r) = h\}$$



Legendre transform :

$$D(h) = \min_q (qh - \tau(q))$$



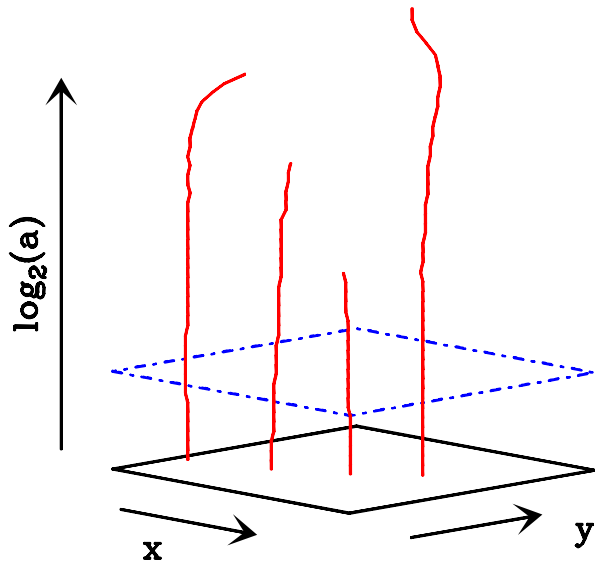
Analogy statistical physics : compute  
partition functions

$$\mathcal{Z}(q, a) = \sum_{\mathcal{L}(a)} (\mathcal{M}_\psi(r, a))^q \sim a^{\tau(q)}$$

# WTMM Method : multifractal formalism

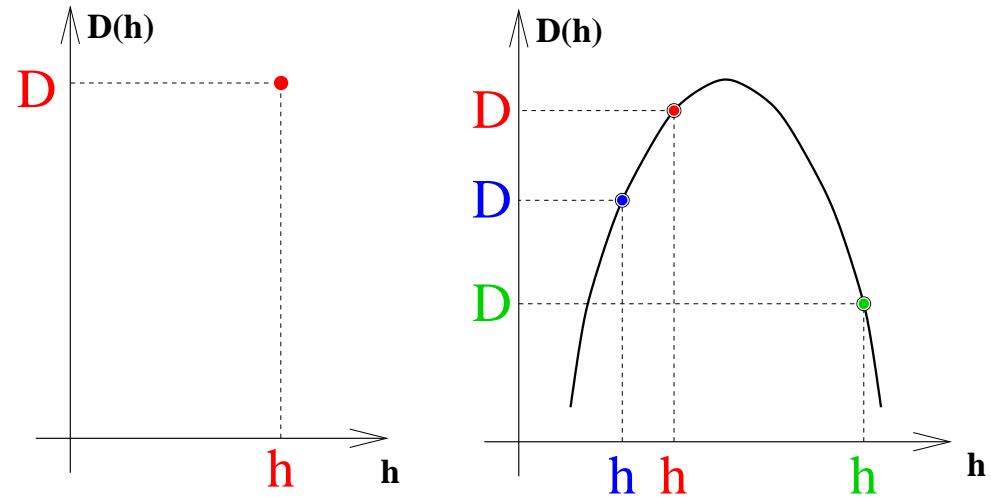
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Analogy statistical physics : compute  
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$$\mathcal{Z}(q, a) = \sum_{\mathcal{L}(a)} (\mathcal{M}_\psi(r, a))^q \sim a^{\tau(q)}$$

$$\mathcal{H}(q, a) = \sum_{\mathcal{L}(a)} \ln |\mathcal{M}_\psi(r, a)| \mathcal{W}_\psi(r, a) \sim a^{h(q)}$$

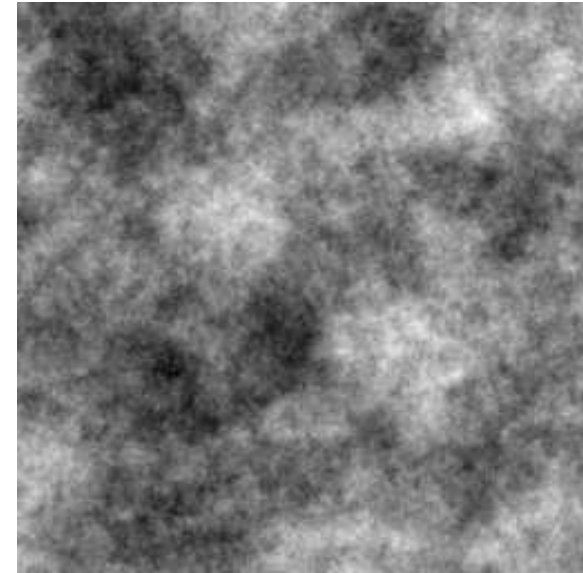
$$\mathcal{D}(q, a) = \sum_{\mathcal{L}(a)} \ln |\mathcal{W}_\psi(r, a)| \mathcal{W}_\psi(r, a) \sim a^{D(q)}$$

# Application to synthetic **monofractal** surfaces

fractional Brownian surfaces :  $B_H(r)$

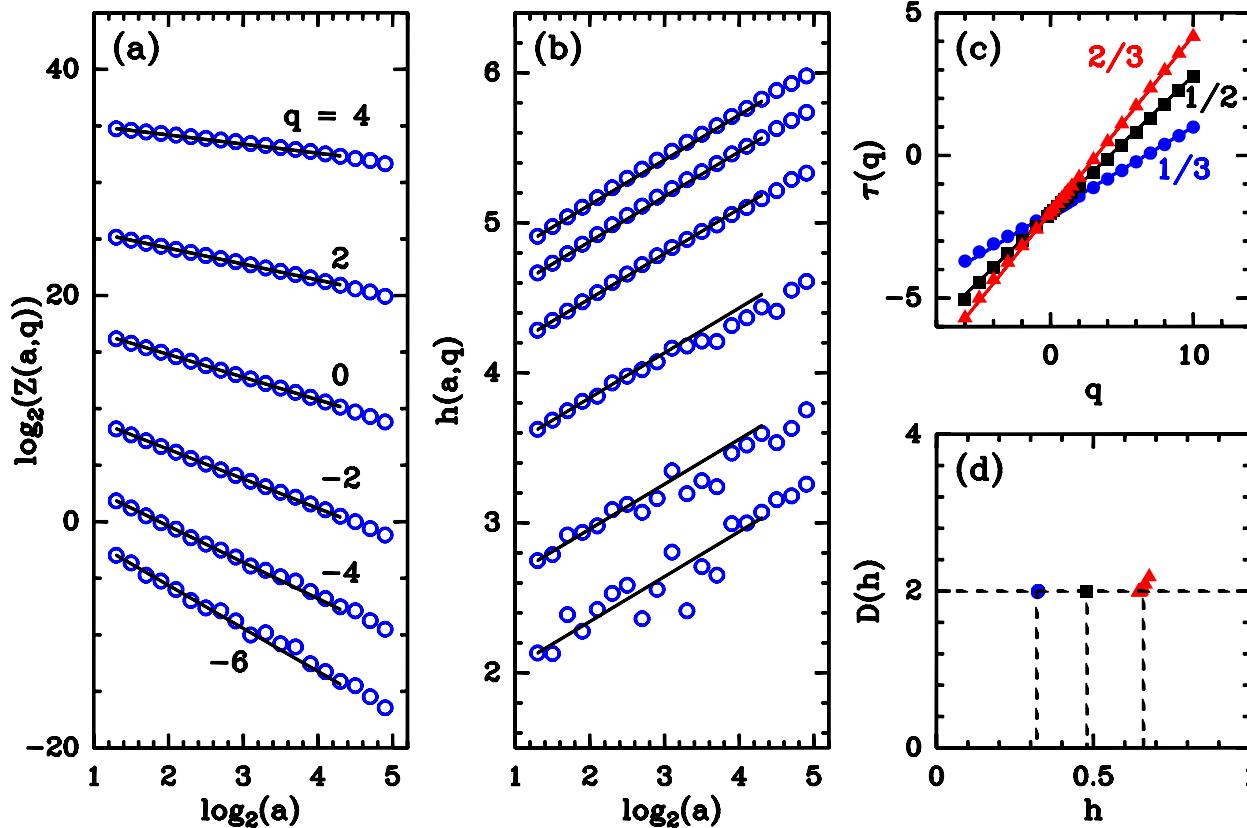
$$H = 1/3$$

- $H < 0.5$  : anti-correlated increments
- $H = 0.5$  : non-correlated increments
- $H > 0.5$  : correlated increments



Theoretical Predictions :

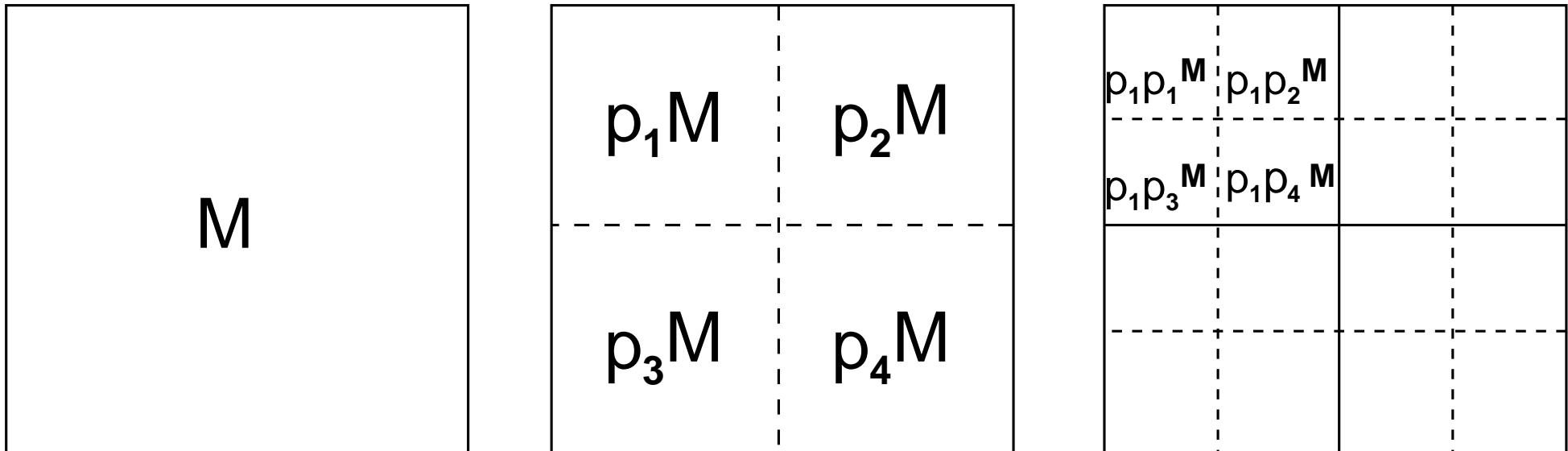
- $\tau(q)$  is linear :  
 $\tau(q) = qH - 2$
- multifractal spectrum is degenerated :  
 $D(h = H) = 2$



# Application to synthetic **multifractal** surfaces

FISC (**Fractionally Integrated Singular Cascades**) model

- simple multiplicative model : **p-model** or **multinomial model**

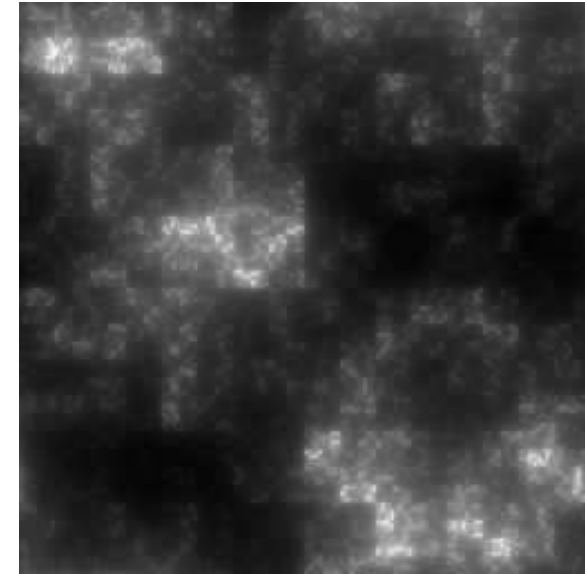


- fractional integration (Fourier domain)
- generalization :  $p$  is a random variable with  $\langle p \rangle = 1/4$  (conservative cascading process)

# Application to synthetic **multifractal** surfaces

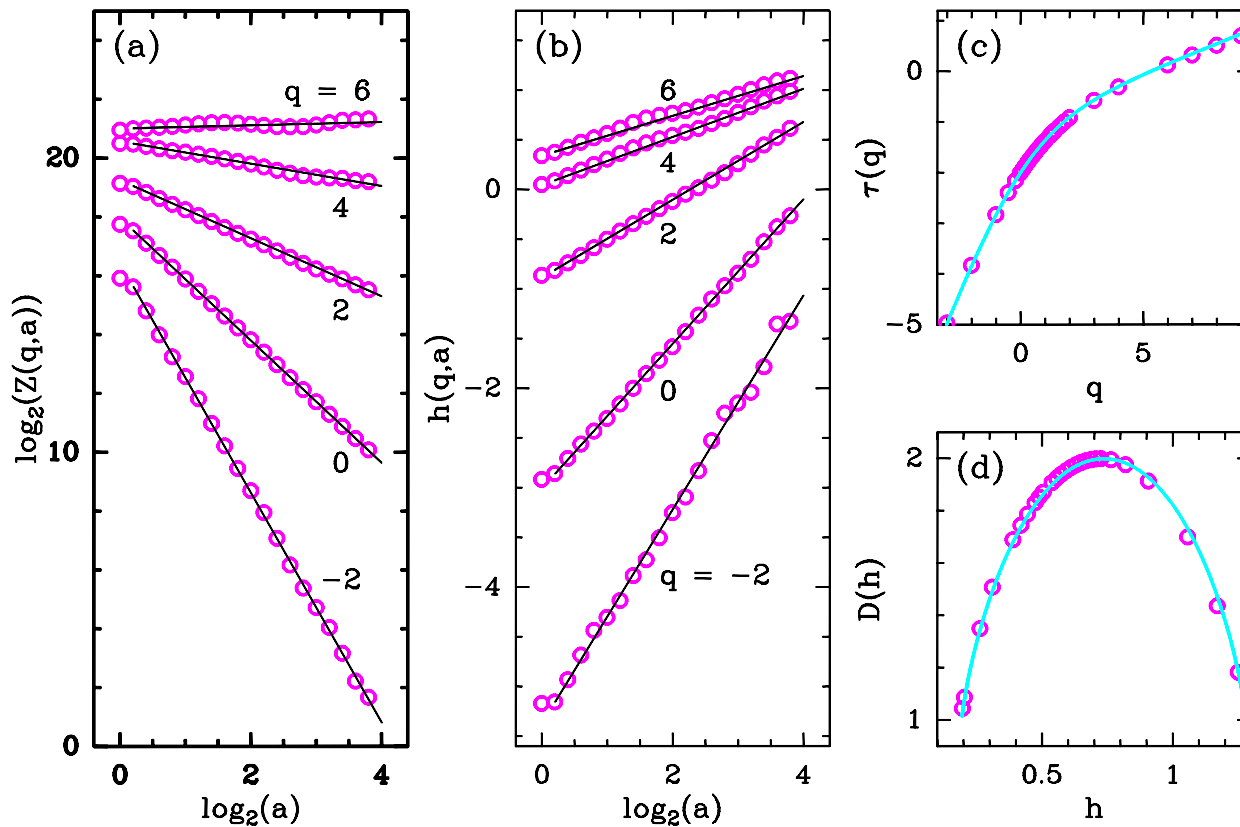
FISC

**Multifractal** (Fractionally Integrated Singular Cascades) surfaces



Theoretical predictions :

- $\tau(q)$  is non-linear  
$$\tau(q) = -2 - q(1 - H^*) - \log_2(p_1^q + p_2^q)$$
- singularity spectrum is a non-degenerated convex curve



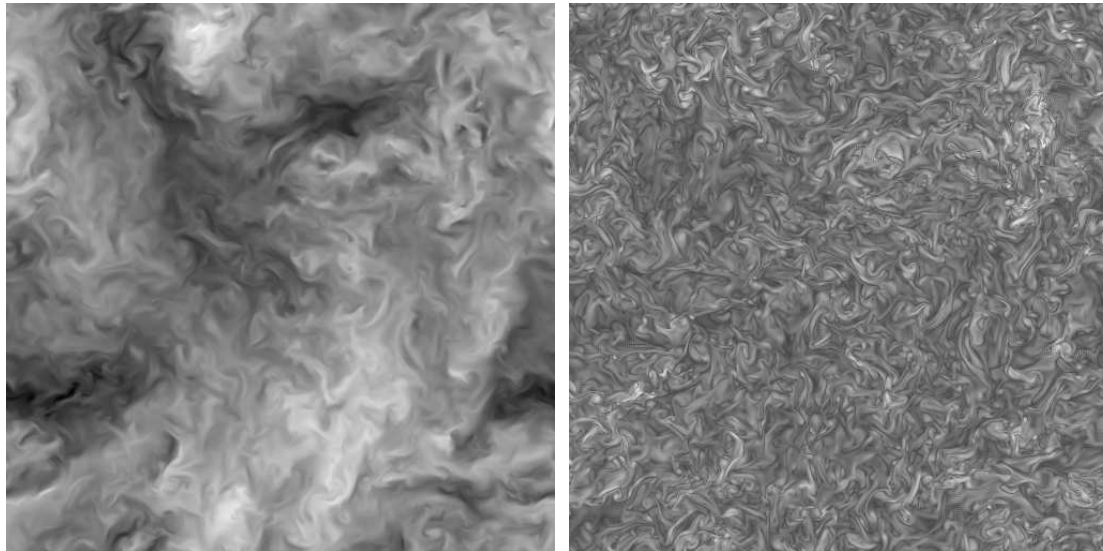
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Measure of the spectra  $\tau_\varepsilon(p)$  and  $f(\alpha)$  with the **box-counting** method.

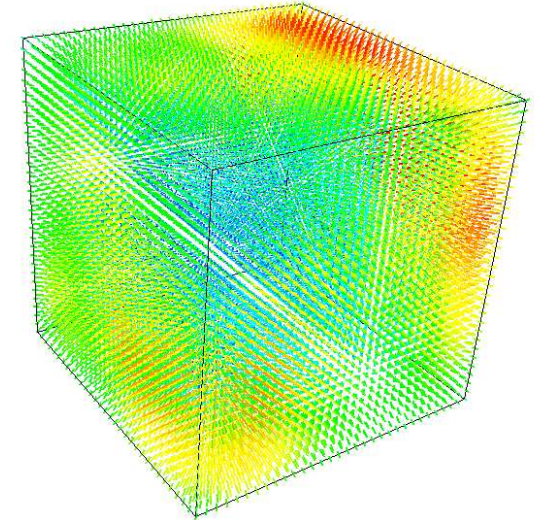
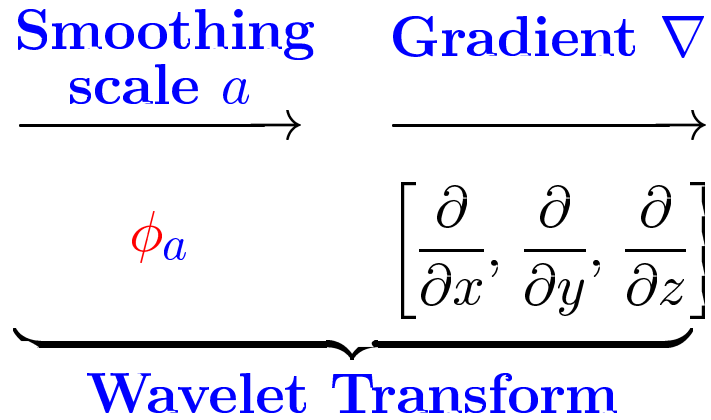
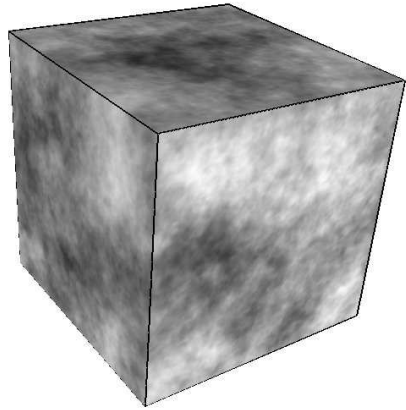
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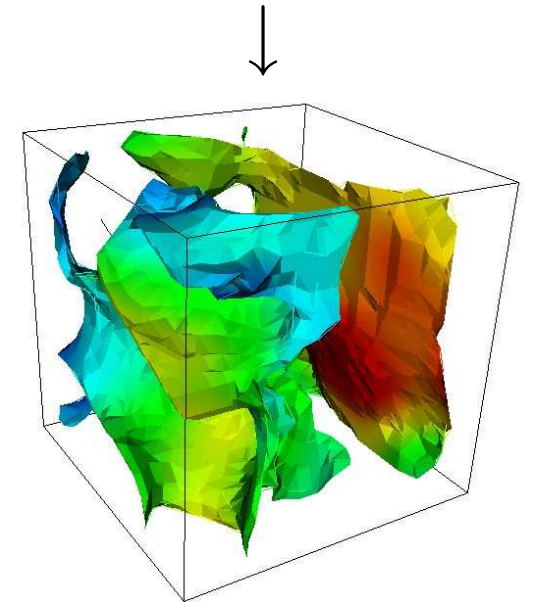
☞ generalization to **multidimensional vector fields**. First application to velocity and vorticity fields from numerical turbulent flows.

# 3D scalar WTMM method

3D data

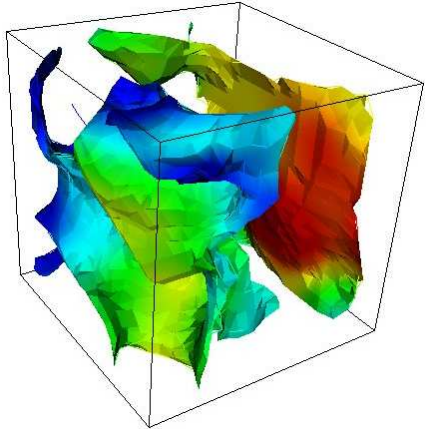


$$\mathbf{T}_\psi(\mathbf{r}, \mathbf{a}) = \begin{pmatrix} \mathbf{I} * \frac{\partial \phi_a}{\partial x}(\mathbf{r}) \\ \mathbf{I} * \frac{\partial \phi_a}{\partial y}(\mathbf{r}) \\ \mathbf{I} * \frac{\partial \phi_a}{\partial z}(\mathbf{r}) \end{pmatrix} = \nabla (\mathbf{I} * \phi_a)(\mathbf{r})$$

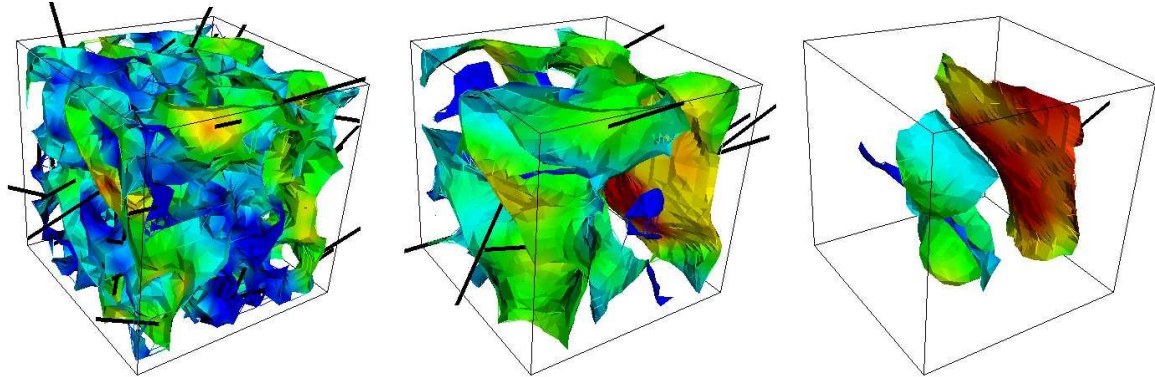


# 3D scalar WTMM method : skeleton

WTMM surfaces

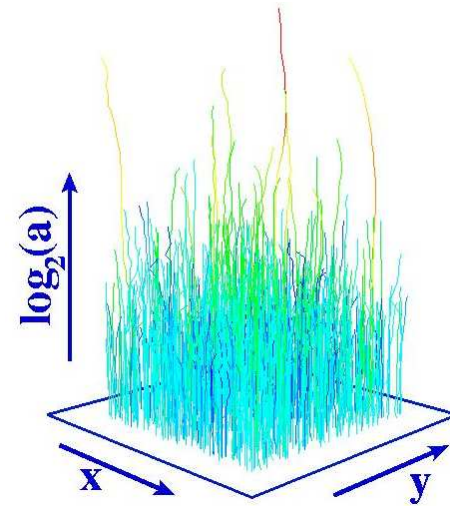
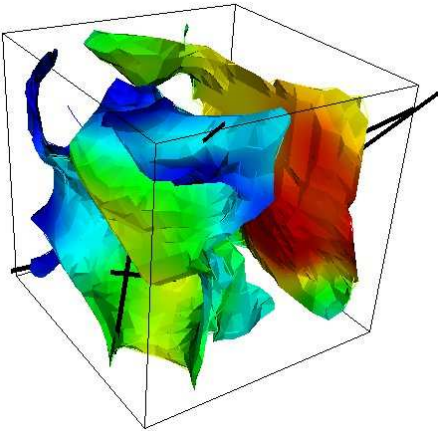


WTMM surfaces at 3 different scales

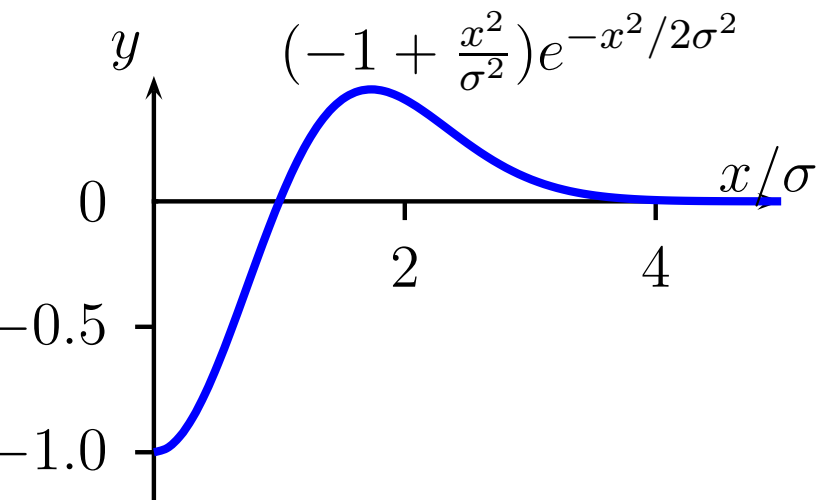
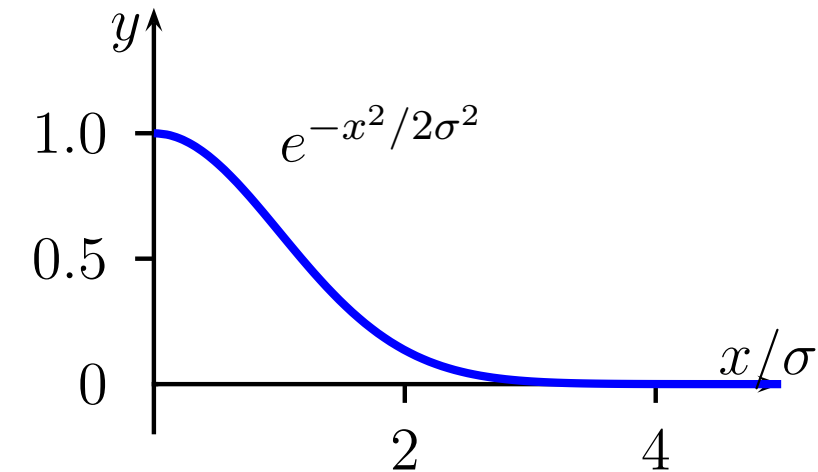


Linking **WTMMM** : WT Skeleton (projection along z)

WTMMM points



# Recursive filters in 3D



- filters with separated variables
- Approximate Gaussian filter  $e^{-x^2/2\sigma^2}$  with  $h_\sigma(x)$   
 $(a_0 \cos(\omega_0 \frac{x}{\sigma}) + a_1 \sin(\omega_0 \frac{x}{\sigma})) \exp^{-b_0 \frac{x}{\sigma}}$   
 $+ (c_0 \cos(\omega_1 \frac{x}{\sigma}) + c_1 \sin(\omega_1 \frac{x}{\sigma})) \exp^{-b_1 \frac{x}{\sigma}}$
- 4th order difference equation :  
 $y_k = n_{00}x_k + n_{11}x_{k-1} + n_{22}x_{k-2} + n_{33}x_{k-3} - d_{11}y_{k-1} - d_{22}y_{k-2} - d_{33}y_{k-3} - d_{44}y_{k-4}$

• comparison FFT/recursive filters :  
computing time decrease in 3D case:

☞ **60 % for Gaussian filter**

☞ **25 % for Mexican filter**

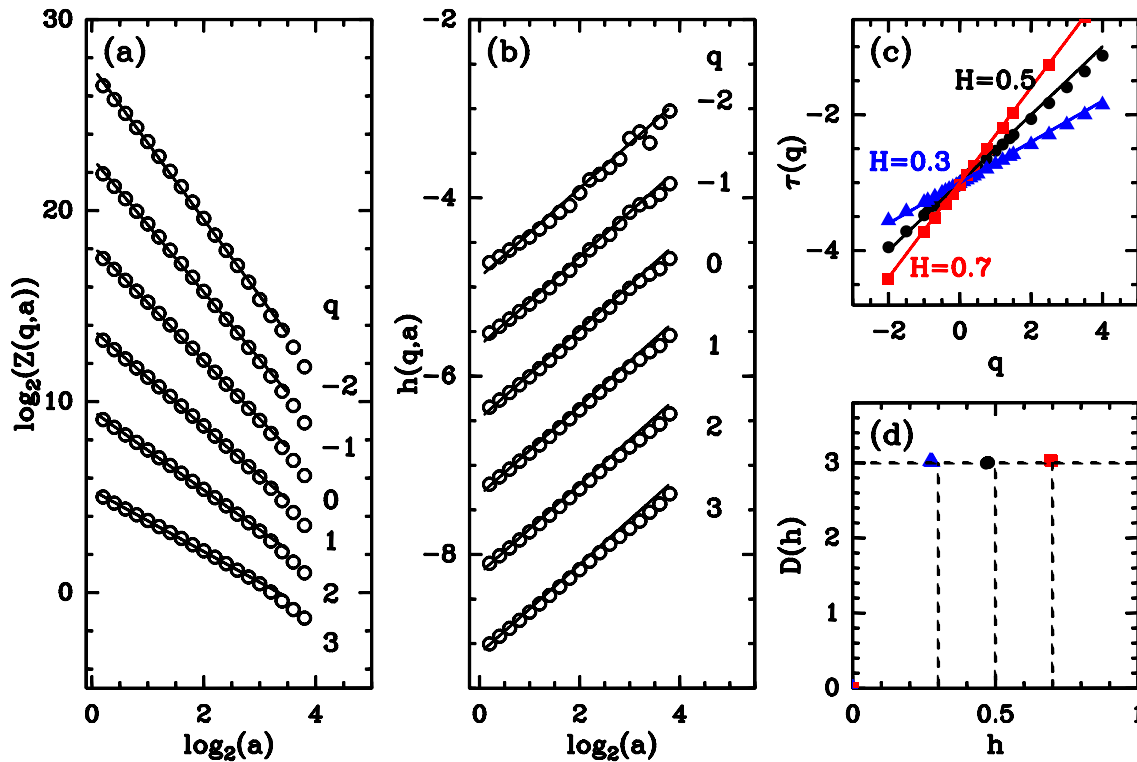
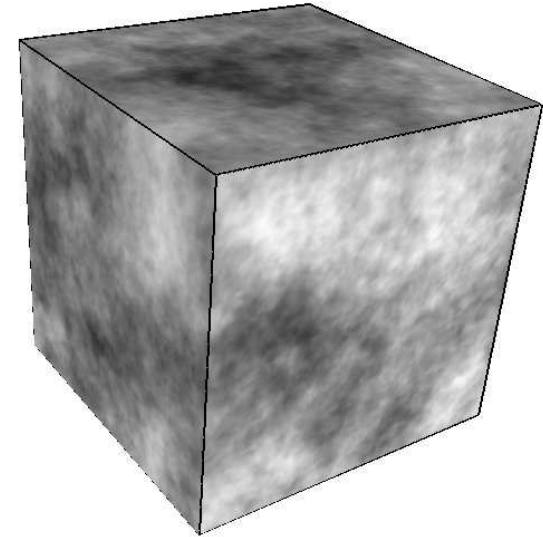
# Test-application to synthetic 3D monofractal fields

**fractional Brownian fields** :  $B_H(\mathbf{r})$

⇒  $H < 0.5$  : anti-correlated increments

⇒  $H = 0.5$  : non-correlated increments

⇒  $H > 0.5$  : correlated increments



Theoretical predictions :

⇒  $\tau(q)$  is linear:

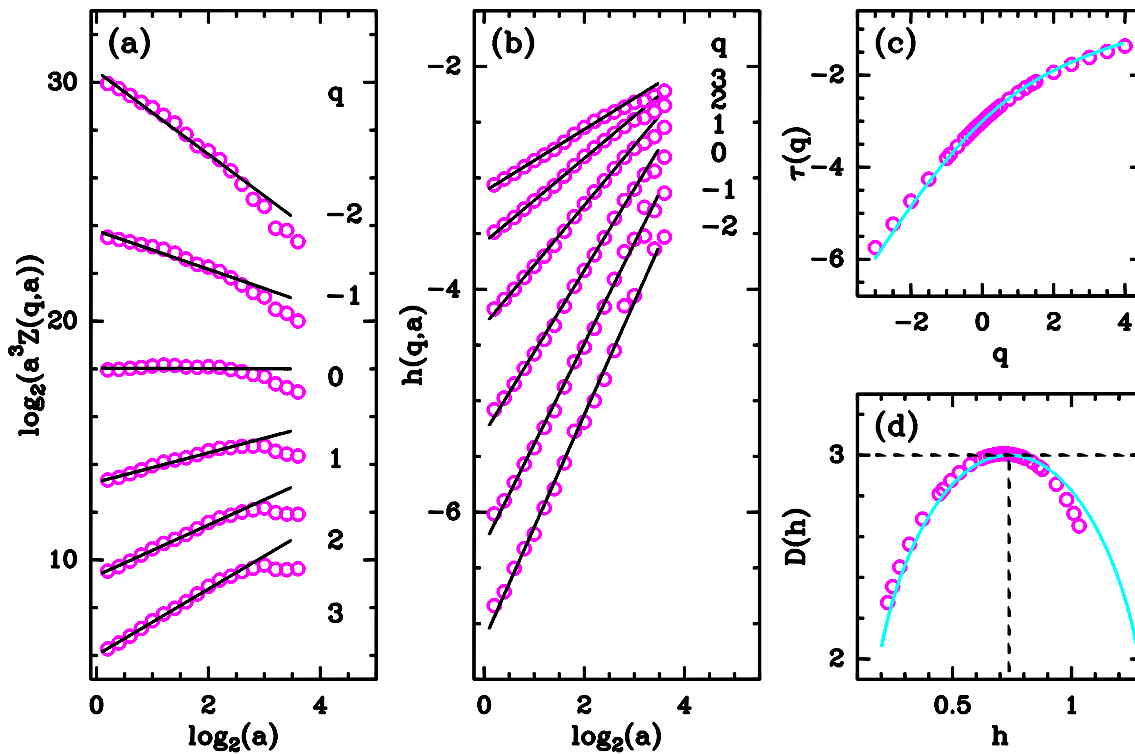
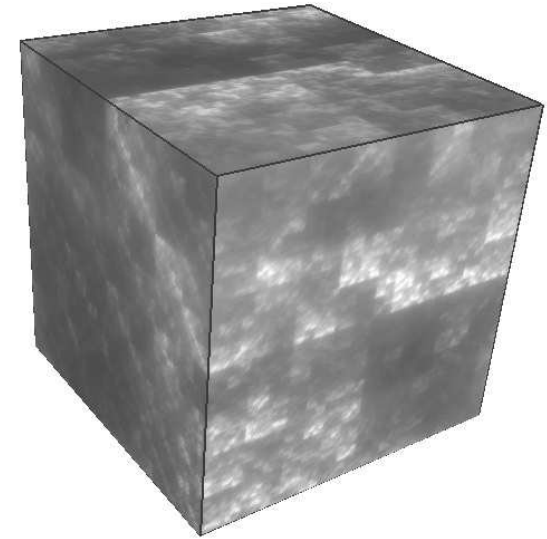
$$\tau(q) = qH - 3$$

⇒ multifractal spectrum is degenerated:

$$D(h = H) = 3$$

# Test-application to synthetic 3D multifractal fields

## 3D multifractal fields (Fractionally Integrated Singular Cascades)



### Theoretical predictions :

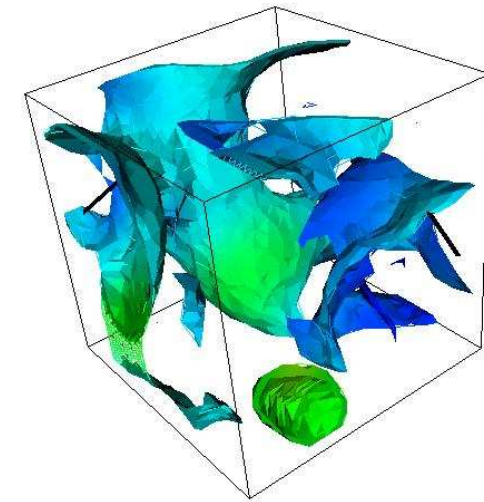
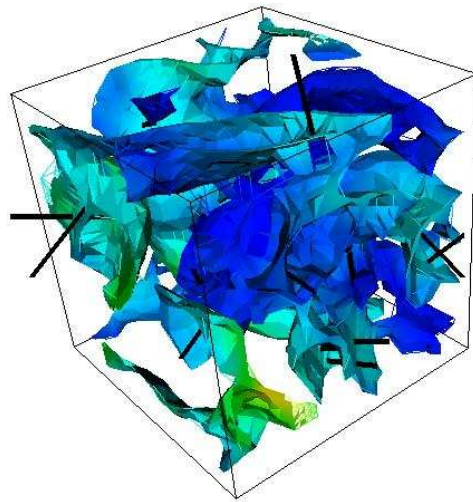
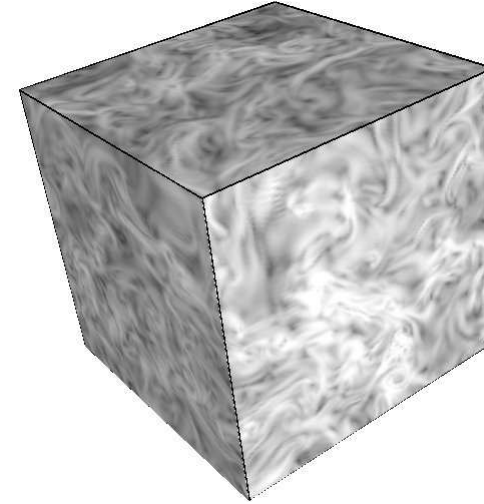
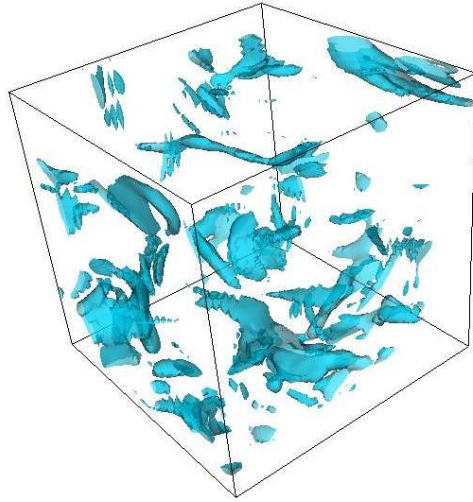
$\tau(q) = -2 - q(1 - H^*) - \log_2(p_1^q + p_2^q)$   
 with  $p_1 + p_2 = 1$

$\rightarrow$  singularity spectrum is a non-degenerated convex curve



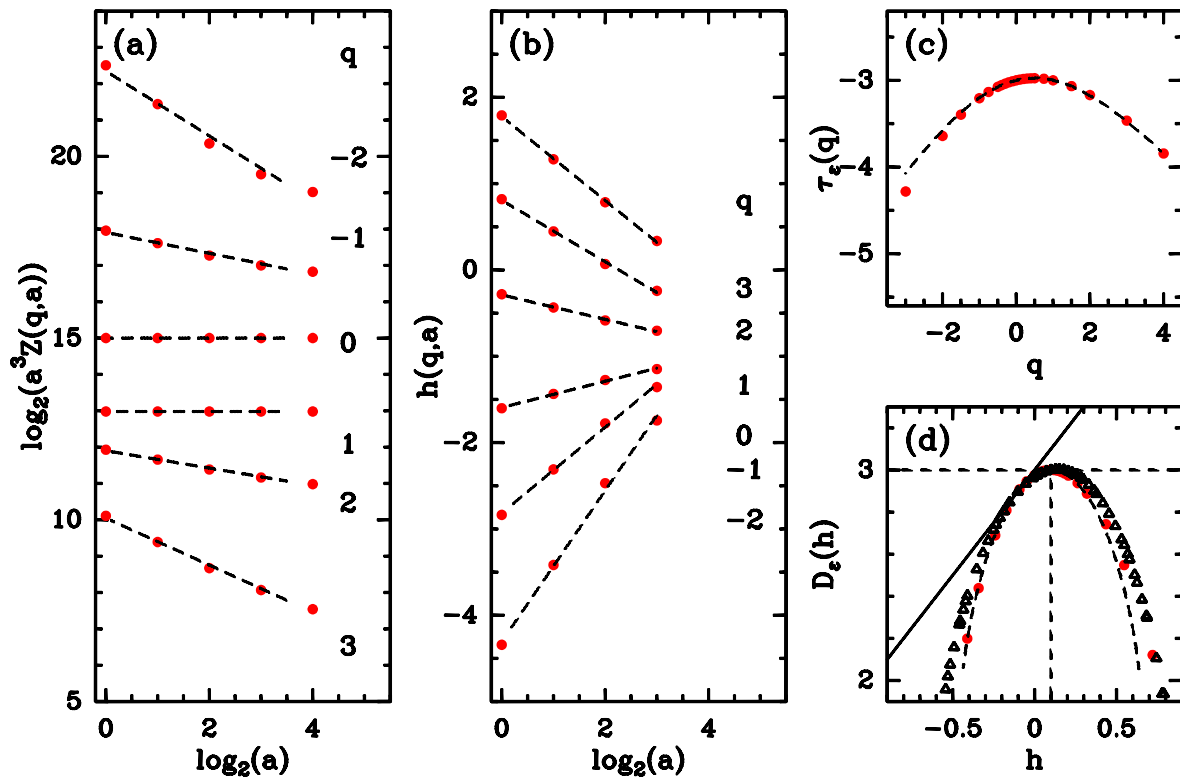
# 3D dissipation field : isotropic turbulence DNS

pseudo-spectral code,  $(512)^3$  grid,  $R_\lambda = 216$  (M. Meneguzzi)



# 3D WTMM methodology vs Box-Counting algorithms

👉 **“Box-Counting”** algorithm, binomial fit with  $p_1 = 0.3$  and  $p_2 = 0.7 \rightarrow$   
 $p_1 + p_2 = 1$  : diagnoses a **conservative** multiplicative structure



dissipation

binomial model :

$$\tau(q) = -2 - q - \log_2(p_1^q + p_2^q)$$



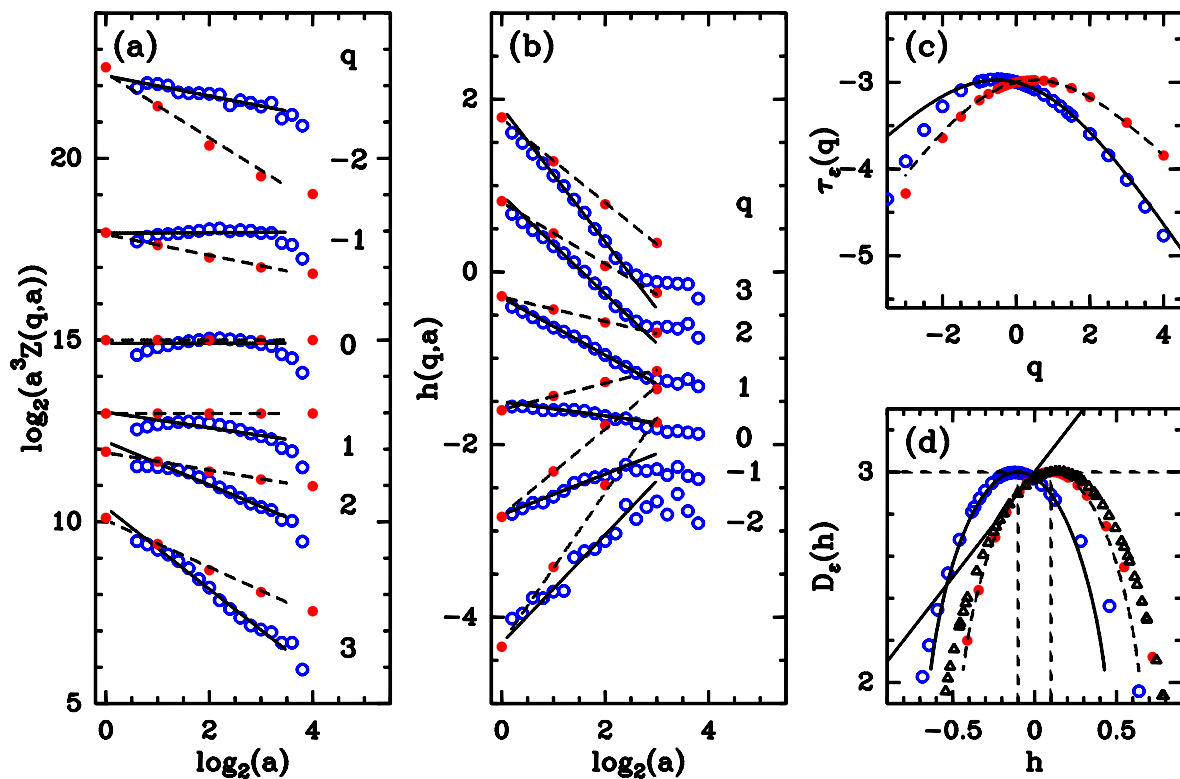
# 3D WTMM methodology vs Box-Counting algorithms

☞ “**Box-Counting**” algorithm, binomial fit with  $p_1 = 0.3$  and  $p_2 = 0.7 \rightarrow$

$p_1 + p_2 = 1$  : diagnoses a **conservative** multiplicative structure

☞ “**3D WTMM**” method reveals a **non-conservative** multiplicative structure :

binomial fit with  $p_1 = 0.36$  and  $p_2 = 0.80 \Rightarrow p_1 + p_2 \neq 1$



dissipation

binomial model :

$$\tau(q) = -2 - q - \log_2(p_1^q + p_2^q)$$

remark:

$$p_1 = \frac{p_1}{p_1 + p_2}$$

$\Rightarrow$  inconsistent

**box-counting !**

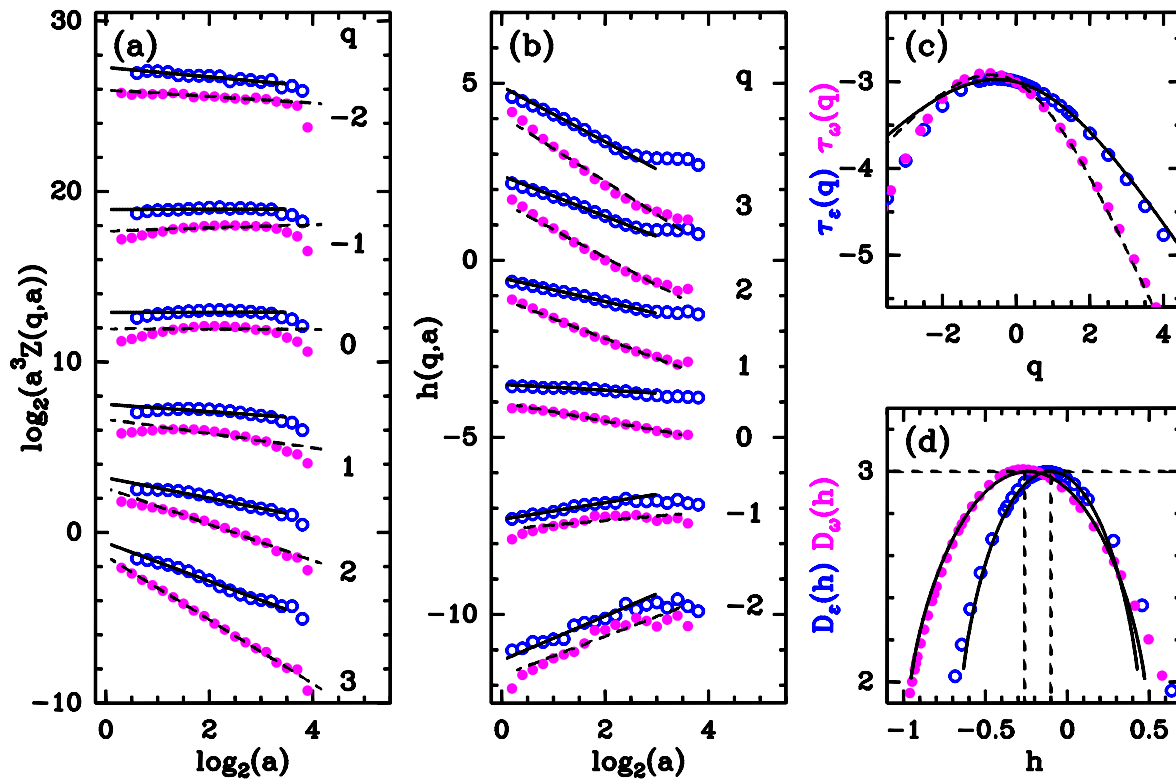
# Comparative 3D WTMM analysis of dissipation and enstrophy

➔ **Dissipation, non-conservative** multiplicative structure :

binomial fit with  $p_1 = 0.36$  and  $p_2 = 0.80 \Rightarrow p_1 + p_2 = 1.16$

➔ **Enstrophy, non-conservative** multiplicative structure :

binomial fit with  $p_1 = 0.38$  and  $p_2 = 0.94 \Rightarrow p_1 + p_2 = 1.32$

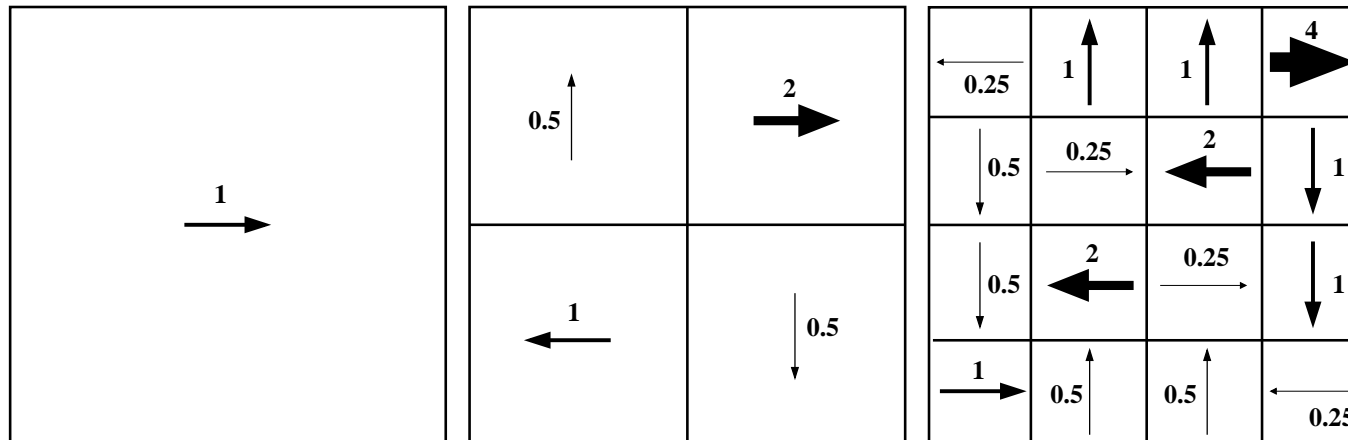


Intermittency coefficients:

● **Dissipation :**  
 $C_2 \sim 0.22$

● **Enstrophy :**  
 $C_2 \sim 0.30$

# Self-similar multifractal vector-valued measure (2D case)



➔ Falconer and O'Neil (1995)

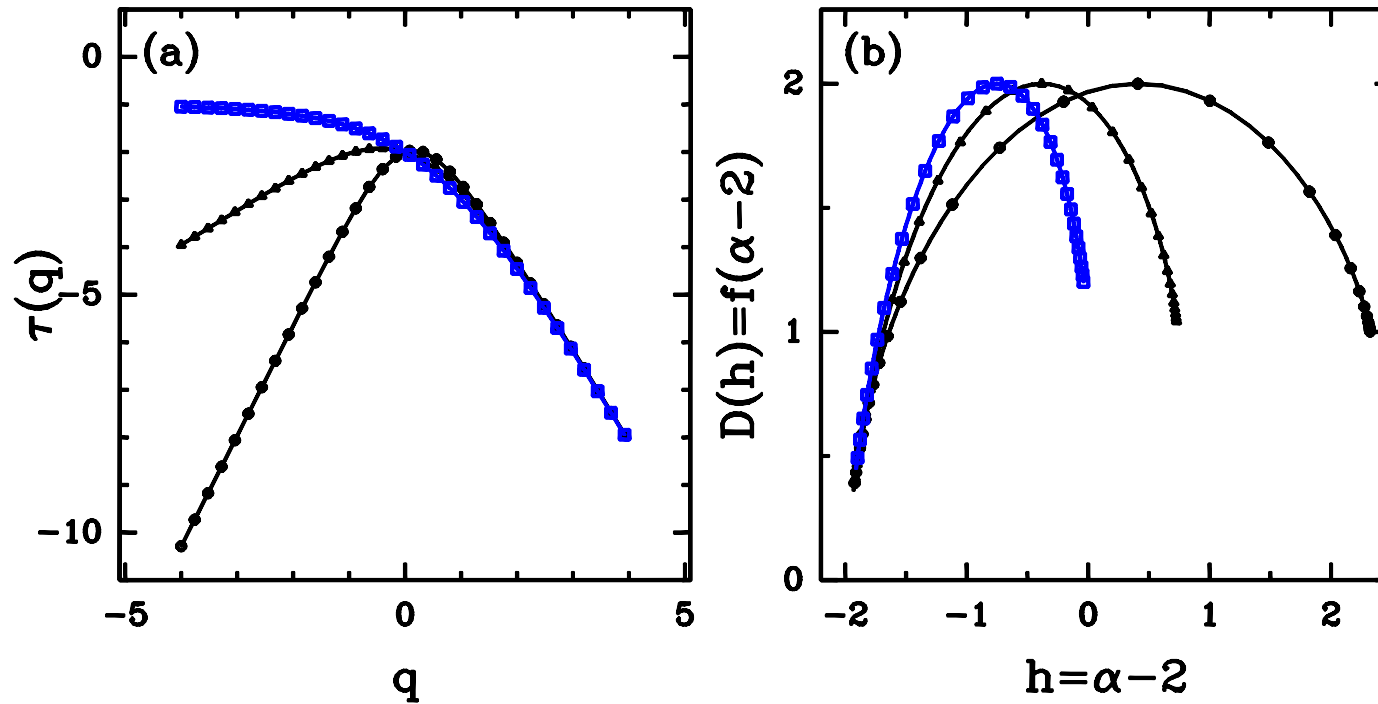
➔ scalar measure  $\{r : \lim_{l \rightarrow 0} \frac{\log \mu(B(r,l))}{\log l} = \alpha\}, \quad \alpha = h + 2$

$$\mathcal{Z}(q, l) = \sum_i \mu_i^q(l) \sim l^{\tau_\mu(q)}$$

➔ vector-valued measure  $\left\{ r : \lim_{l \rightarrow 0} \frac{\log \int_{B(r,l)} \|\Phi_l \mu(s)\| d\mathcal{L}_d(s)}{\log l} = \alpha \right\}, \quad \alpha = h + 2$

$$\mathcal{Z}(q, l) = \sum_i \|\Phi_l \mu\|_i^q \sim l^{\tau_\mu(q)}$$

# Self-similar multifractal vector-valued measure (2D case)



→  $\tau_{\mu}(q) = -\frac{\log(p_1^q + p_2^q + p_3^q + p_4^q)}{\log 2}$

→  $D_{\mu}(h) = f_{\mu}(\alpha - 2) = \inf_q(qh - \tau_{\mu}(q)).$

# Tensorial wavelet transform (2D case)

1. Tensorial wavelet transform of field  $\mathbf{V} = (V_1, V_2)$  :

$$\mathbb{T}_{\psi}[\mathbf{V}](\mathbf{b}, \mathbf{a}) = (\mathbb{T}_{\psi_i}[V_j](\mathbf{b}, \mathbf{a})) = \begin{pmatrix} \mathbb{T}_{\psi_1}[V_1] & \mathbb{T}_{\psi_1}[V_2] \\ \mathbb{T}_{\psi_2}[V_1] & \mathbb{T}_{\psi_2}[V_2] \end{pmatrix}$$

$$\mathbb{T}_{\psi_i}[V_j](\mathbf{b}, \mathbf{a}) = \mathbf{a}^{-2} \int d^2x r \psi_i(\mathbf{a}^{-1}(\mathbf{r} - \mathbf{b})) V_j(\mathbf{r}), j = 1, 2$$

2. Direction of greatest variation of vector field :

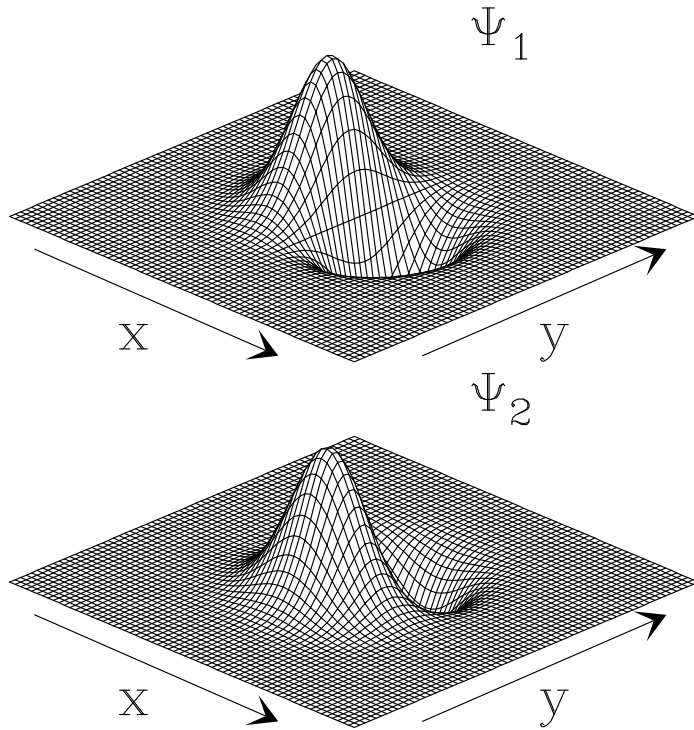
$$|\mathbb{T}_{\psi}[\mathbf{V}]| = \sup_{\mathbf{C} \neq 0} \frac{||\mathbb{T}_{\psi}[\mathbf{V}] \cdot \mathbf{C}||}{||\mathbf{C}||}$$

3. Singular value decomposition of WT tensor:

$$\mathbb{T}_{\psi}[\mathbf{V}] = (\mathbf{G}) \cdot \begin{pmatrix} \sigma_{\max} & 0 \\ 0 & \sigma_{\min} \end{pmatrix} \cdot (\mathbf{D})^T$$

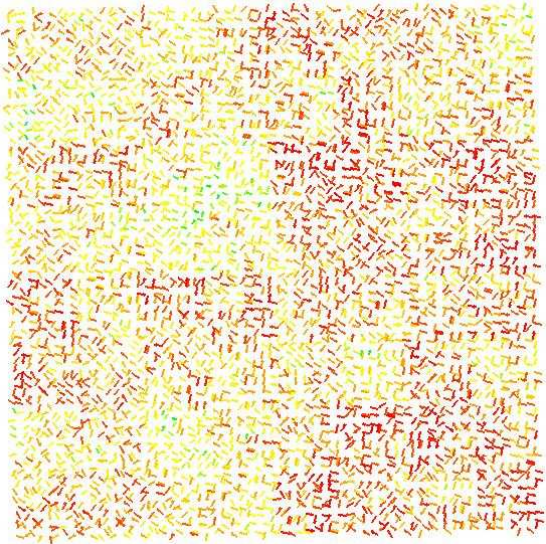
4. Tensorial wavelet transform :

$$\mathbb{T}_{\psi, \max}[\mathbf{V}](\mathbf{b}, \mathbf{a}) = \sigma_{\max} \mathbf{G} \sigma_{\max}$$



# Tensorial 2D WTMM methodology

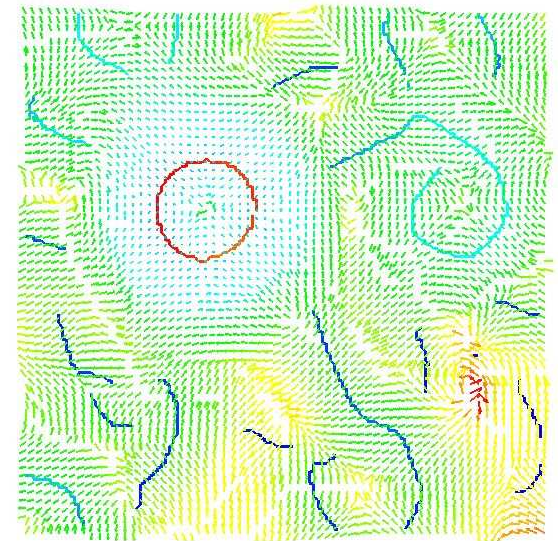
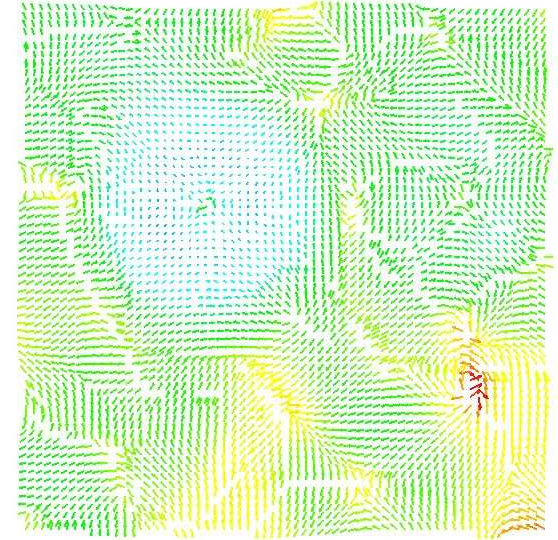
Data



Tensorial wavelet transform



$$\mathbf{T}_{\psi, \max}[\mathbf{V}](\mathbf{b}, a) = \sigma_{\max} \mathbf{G}_{\sigma_{\max}}$$



**Modulus Maxima**  $\sigma_{\max}$  chains of tensorial wavelet transform at scale  $a$  :

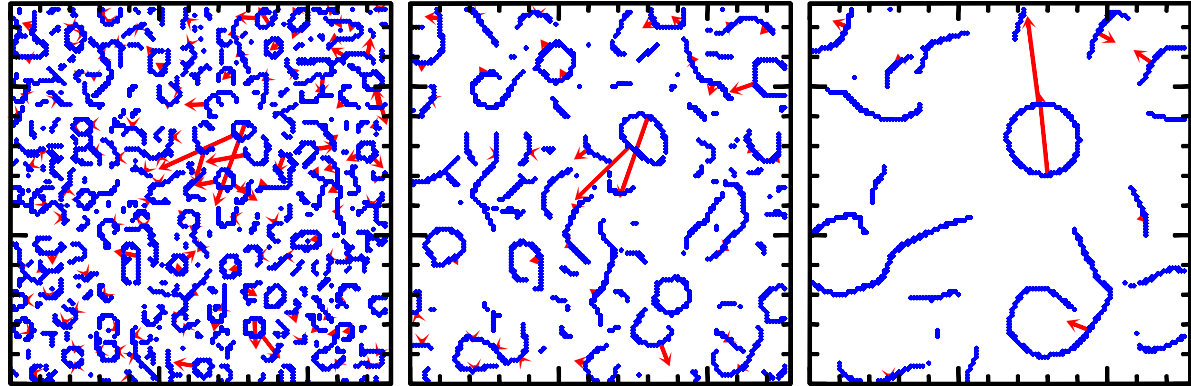
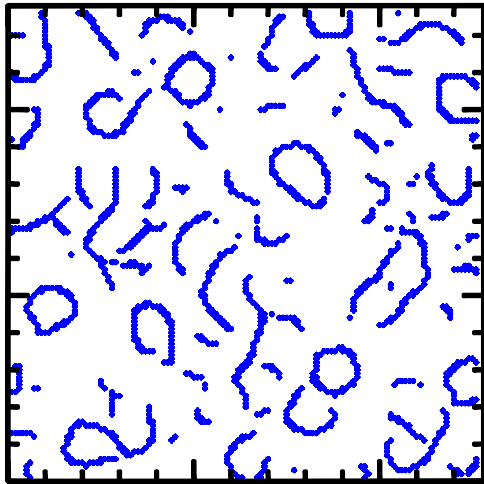
$$\left\{ (b, a) / \frac{\partial \sigma_{\max}}{\partial G_{\max}} = 0 \quad \text{et} \quad \frac{\partial^2 \sigma_{\max}}{\partial G_{\max}^2} < 0 \right\}$$



# Tensorial 2D WTMM methodology: Skeleton

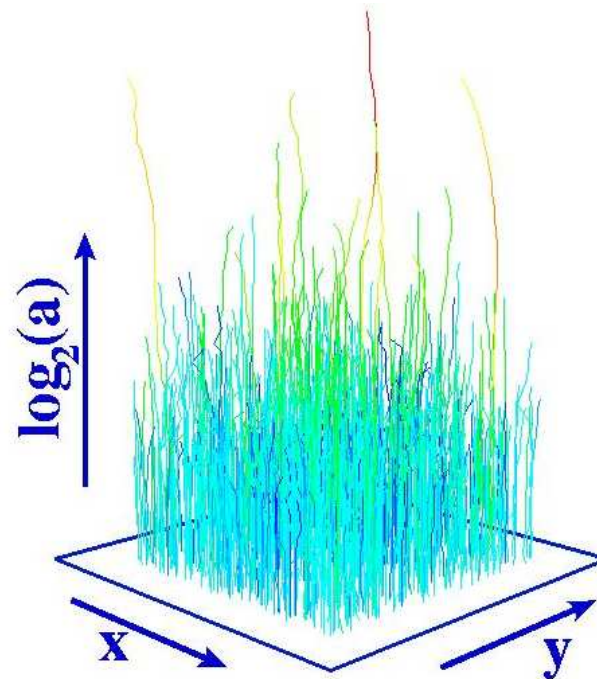
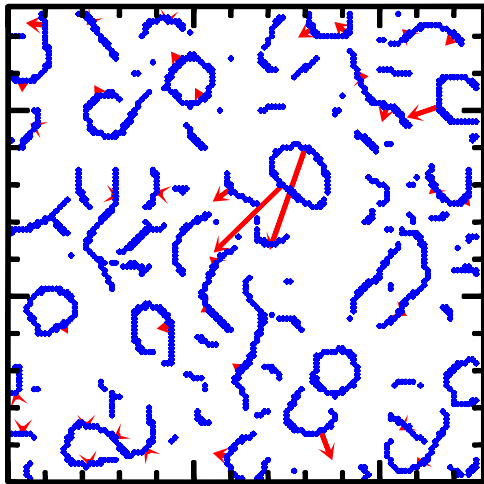
## WTMM Chains at 3 different scales

WTMM Chains



**WTMMM** linking: **WT Skeleton**

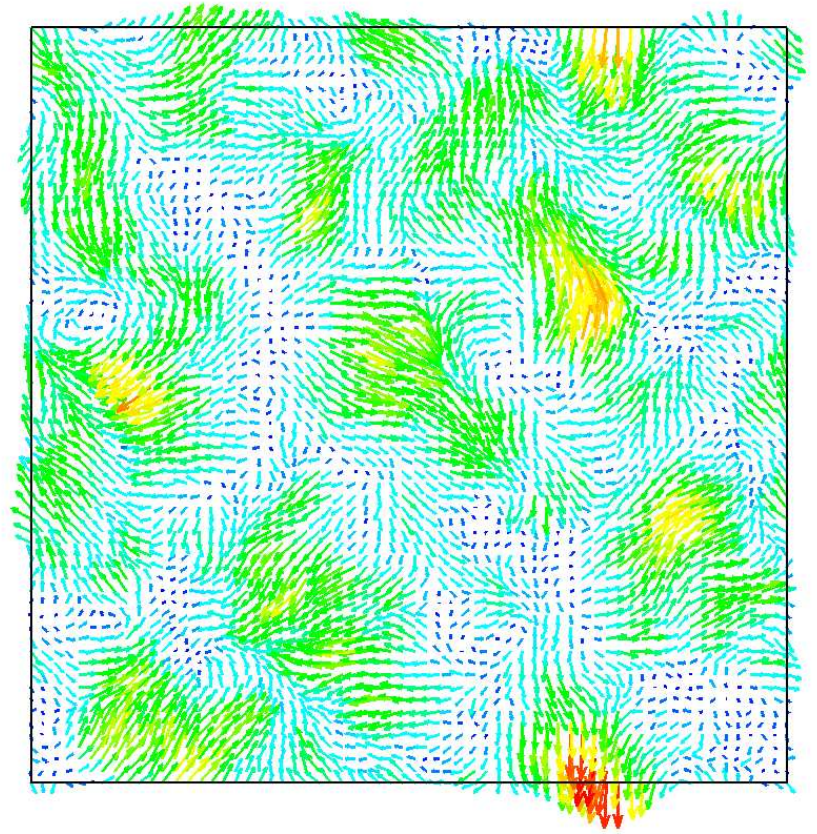
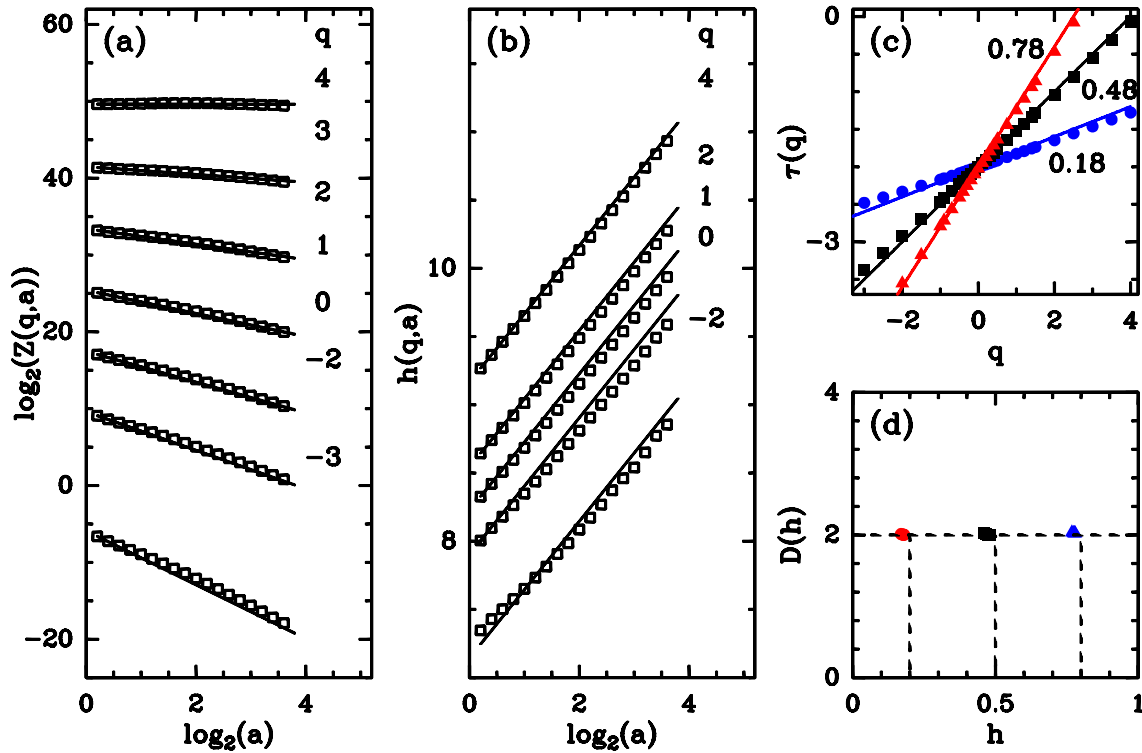
WTMMM



# Monofractal 2D vector fields

fractional Brownian fields :  $B_H(\mathbf{r})$

👉 Spectral method simulation



Theoretical predictions :

● linear  $\tau(q)$ :  $\tau(q) = qH - 2$

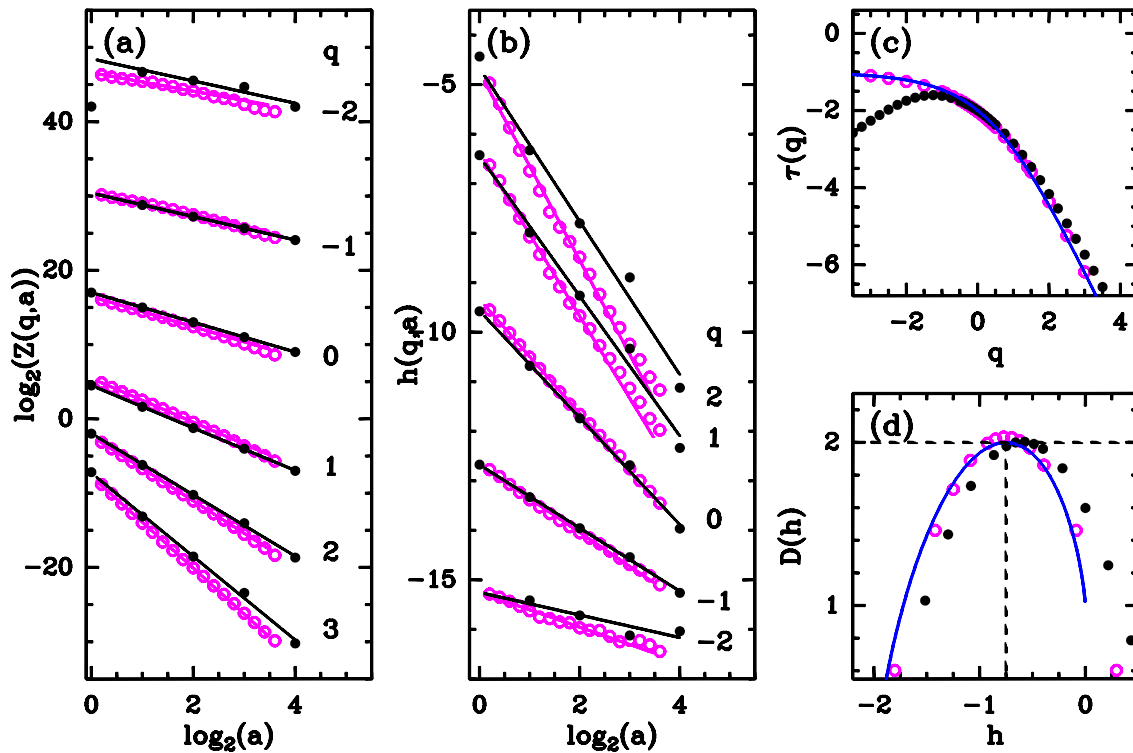
● degenerated singularity spectrum:

$$D(h = H) = 2$$



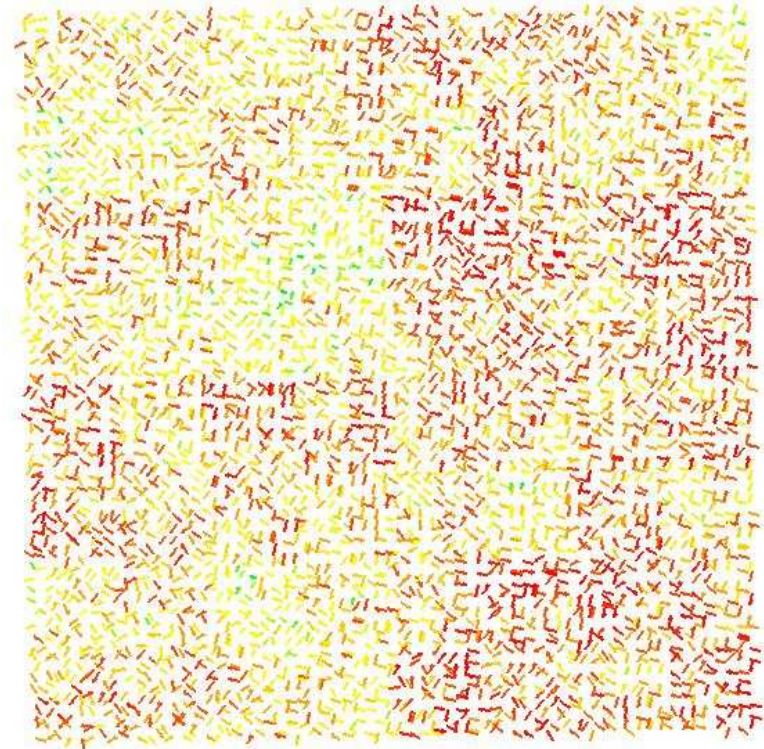
# 2D self-similar multifractal vector-valued measures

Self-similar **multifractal** vector-valued measures  
(Falconer and O'Neil's **model**)



● vectorial box-counting

○ vectorial 2D WTMM method



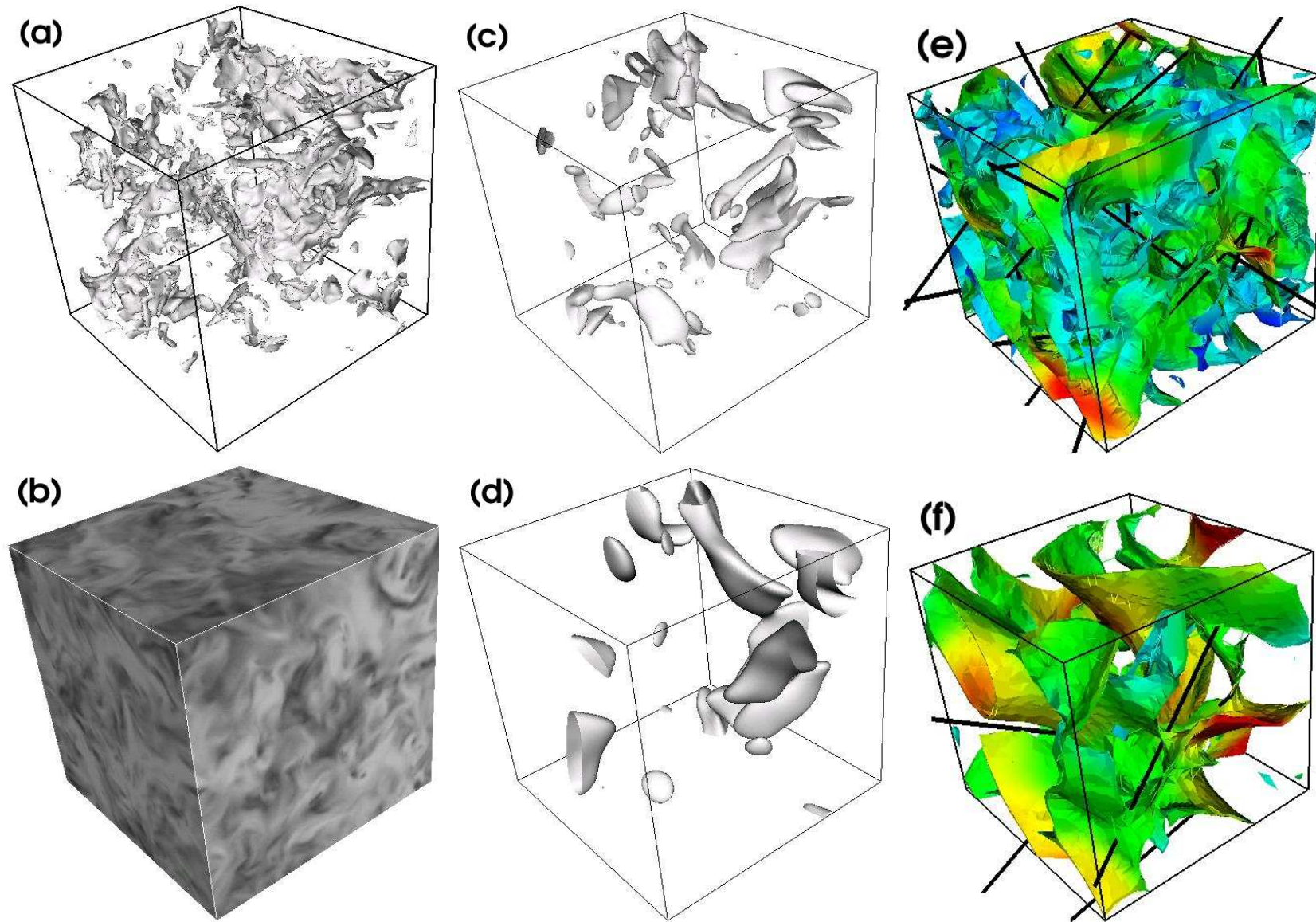
Theoretical predictions:

●  $\tau(q) = -\log_2(p_1^q + p_2^q + p_3^q + p_4^q)$

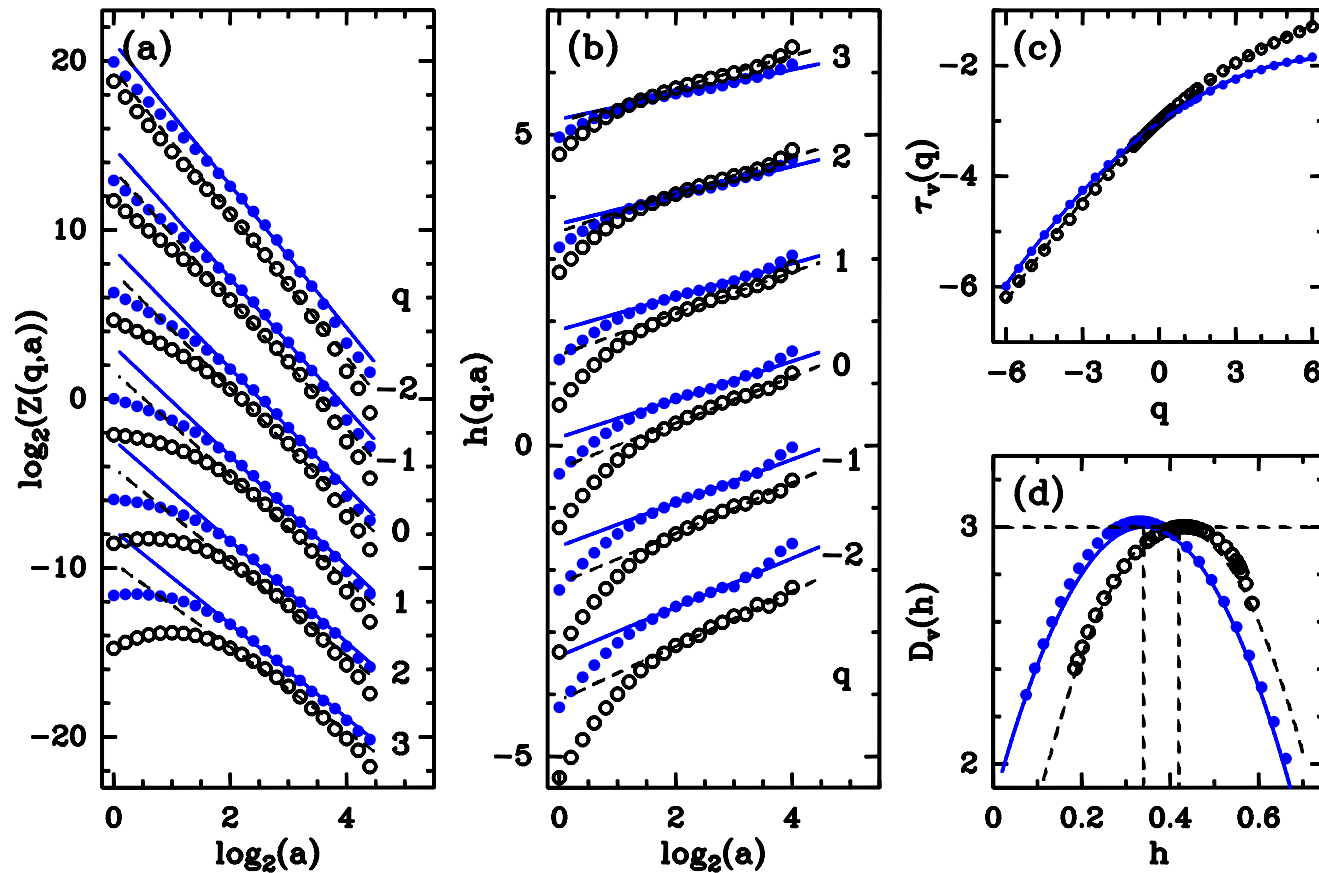
$p_1 = p_4 = 0.5$ ,  $p_2 = 2$  and  $p_3 = 1$

● vectorial box-counting is less accurate

# Tensorial 3D WTMM method: turbulent velocity field ( $R_\lambda = 140$ )



# Tensorial 3D WTMM method: singularity spectrum of velocity



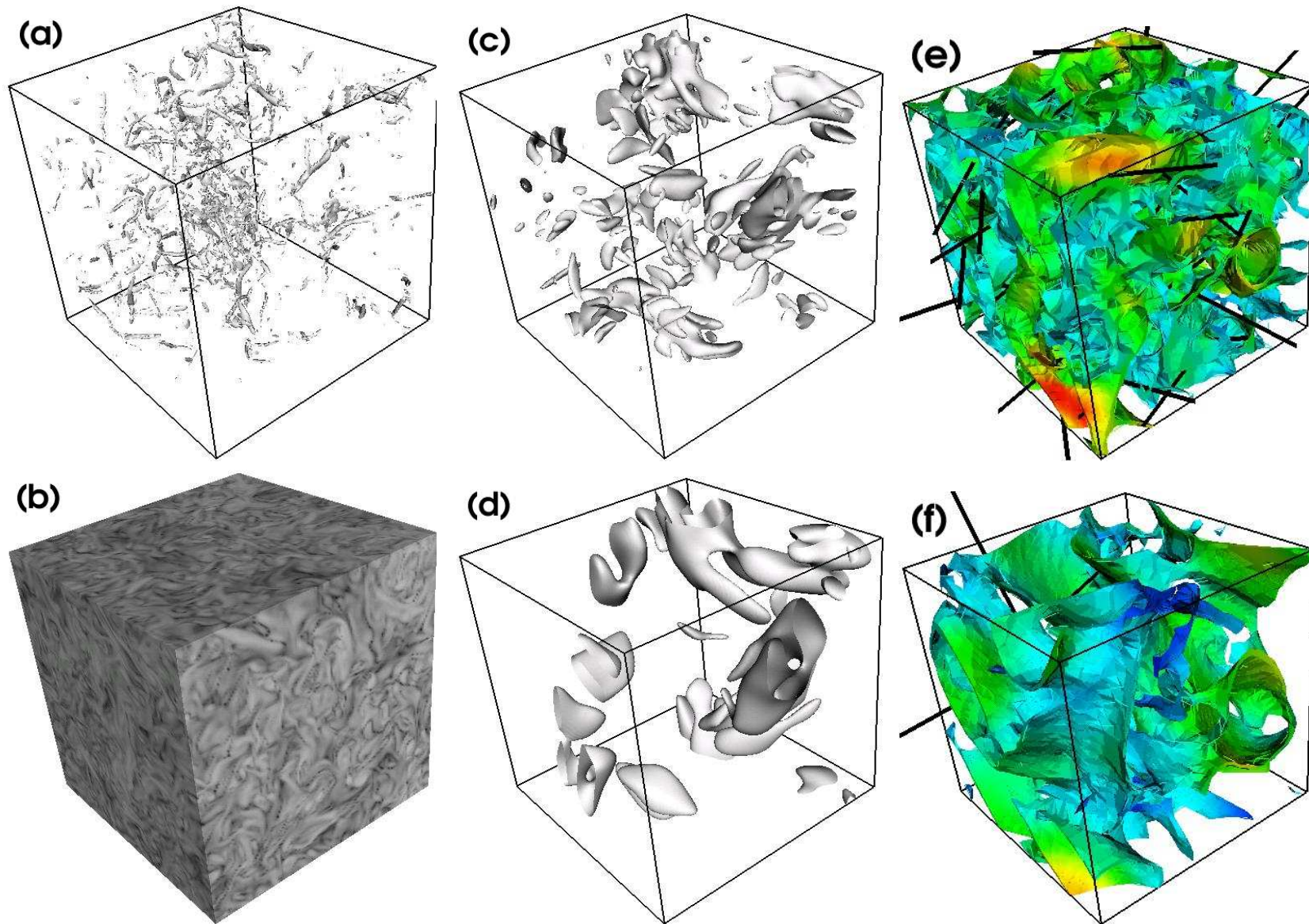
- parabolic fit :  $\tau(q) = -C_0 - C_1 q - C_2 \frac{q^2}{2}$
- intermittency coefficient  $C_2 = 0.049 \pm 0.004$

1D increments method:

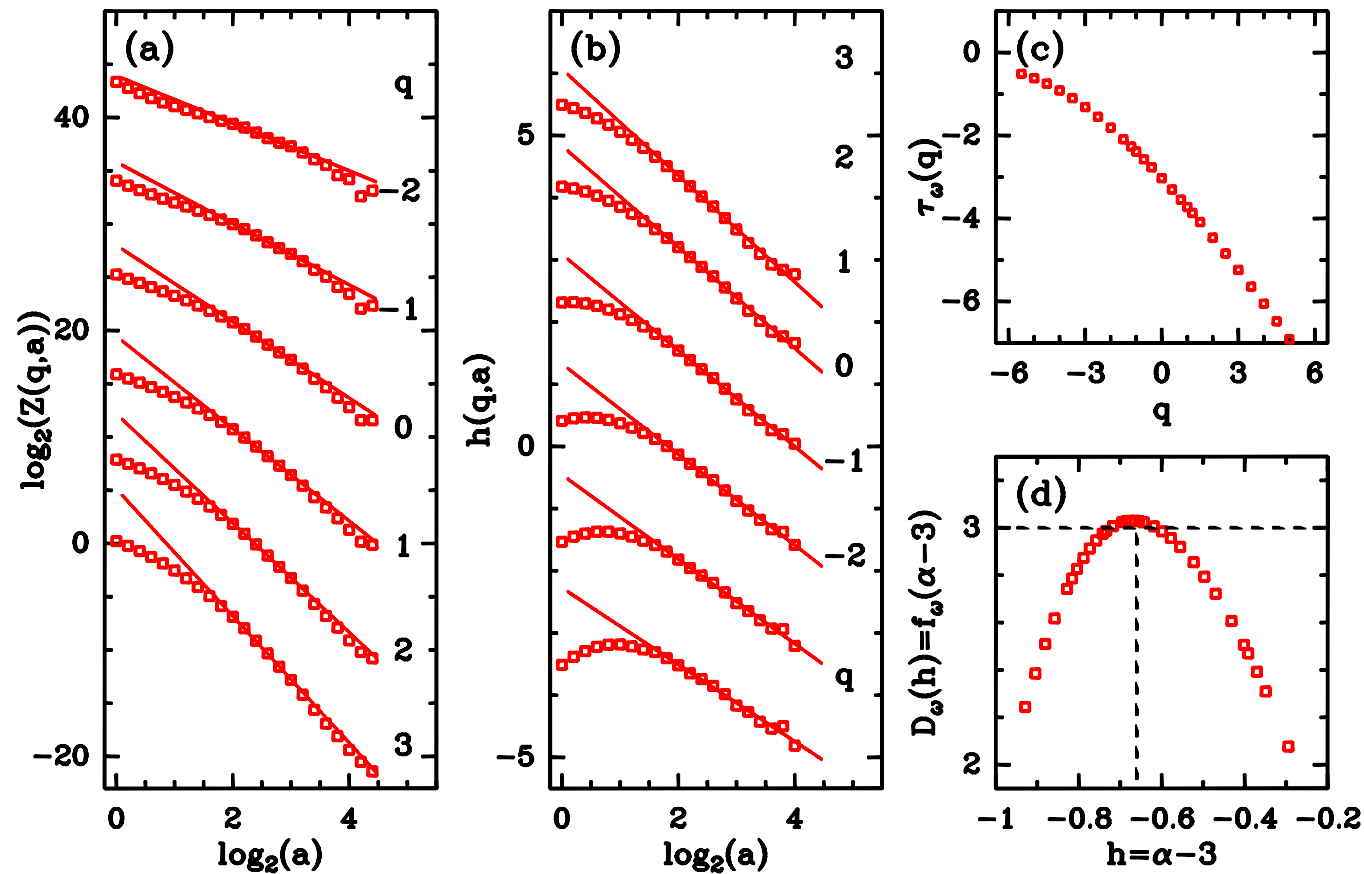
- longitudinal :  $C_2(\delta v_L) \sim 0.025$
- transverse :  $C_2(\delta v_T) \sim 0.040$



# Tensorial 3D WTMM method: turbulent velocity field ( $R_\lambda = 140$ )

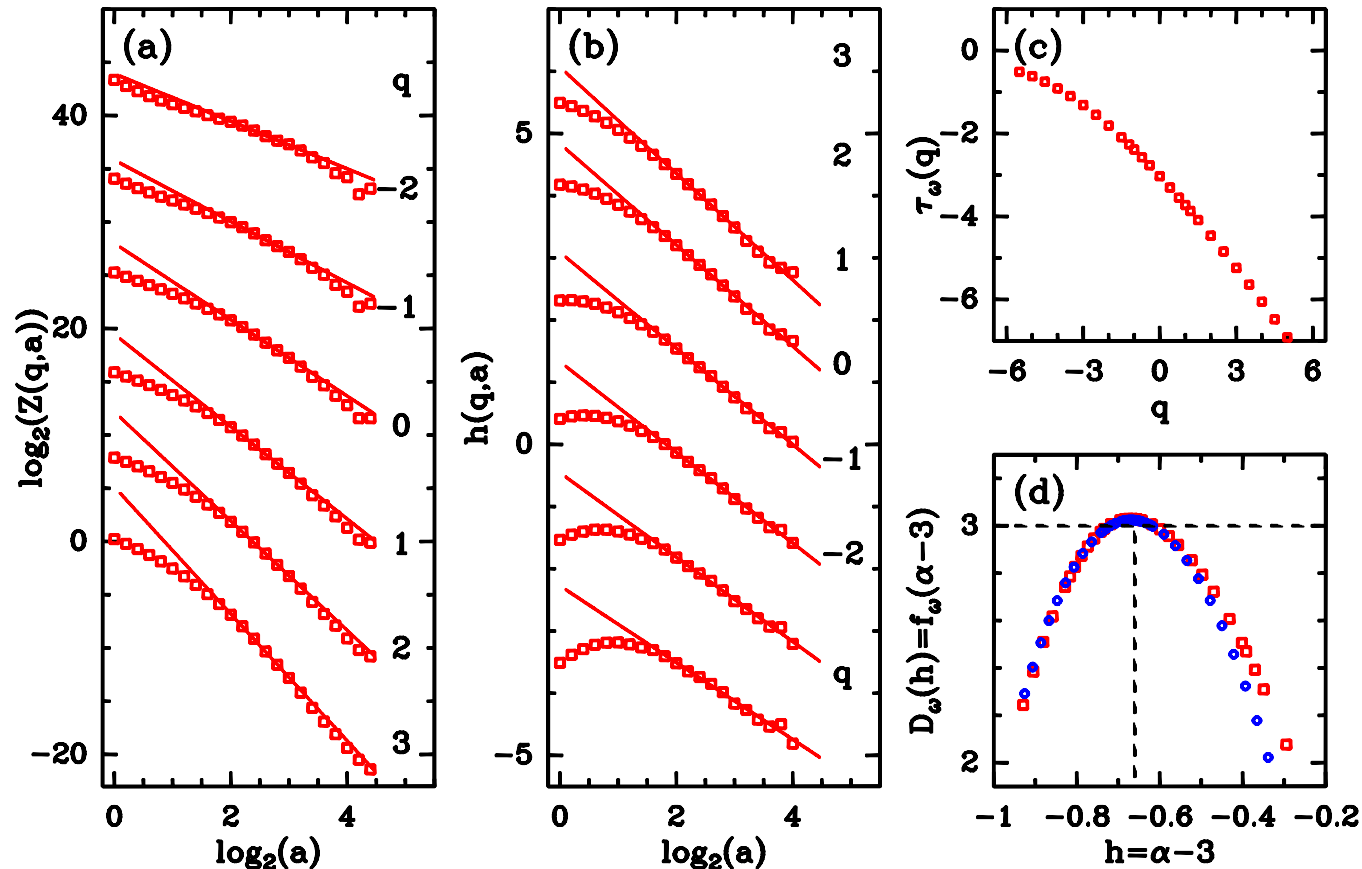


# Tensorial 3D WTMM method: singularity spectrum of vorticity



□ vorticity

# Tensorial 3D WTMM method: singularity spectrum of vorticity



□ vorticity

○  $D_v(h + 1)$  spectrum translated velocity

⇒ same 3D intermittency coefficient !

## assessment:

- WTMM multifractal analysis: moving towards **vector fields**

## outlooks :

- ☞ better understanding of the information embedded in the **WT tensor**.
- ☞ **identification of coherent structures** in turbulence using **WT tensor**'s smallest singular value: vorticity filaments or sheets.
- ☞ others applications : astrophysics (interstellar medium, interstellar turbulence), MHD, geophysics, ...

## Thanks :

- E. Lévêque, Laboratoire de Physique, ENS Lyon (Turbulent flows DNS).

## References :

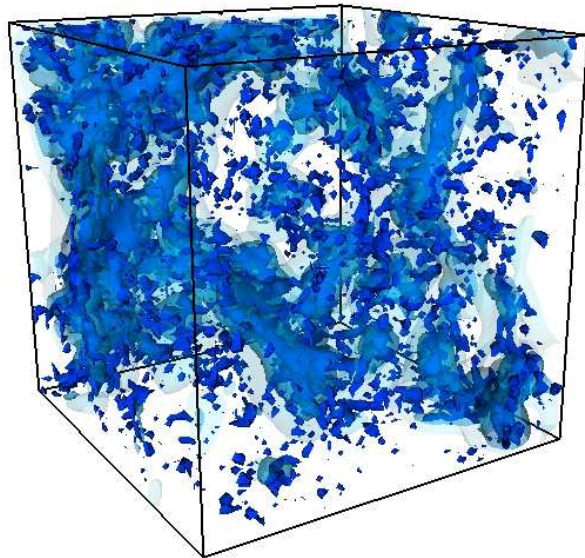
- P. Kestener and A. Arneodo, *Phys. Rev. Lett.*, **91:194501**, 2003.
- P. Kestener and A. Arneodo, *Phys. Rev. Lett.*, **93:044501**, 2004.



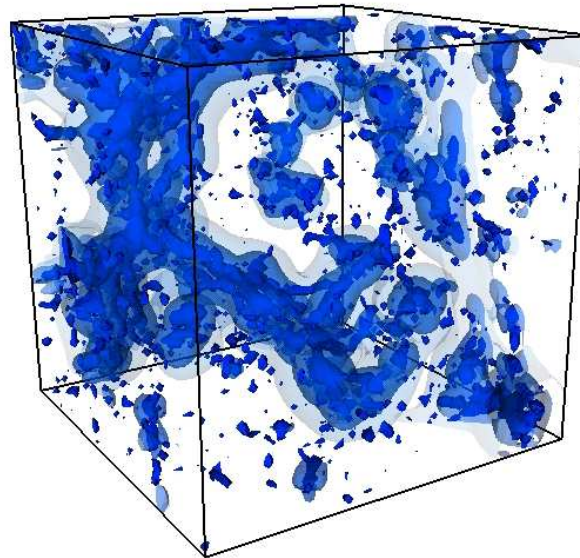
## Preliminary results :

Isosurface plots of data and 3D WT modulus (Gaussian filtering) :

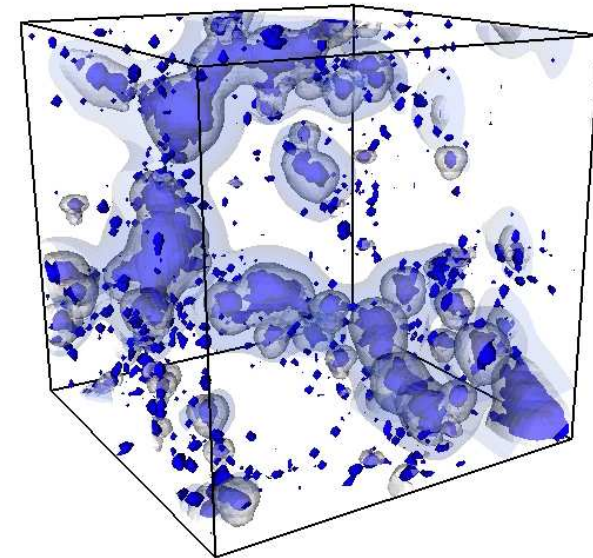
$z = 5$



$z = 2$

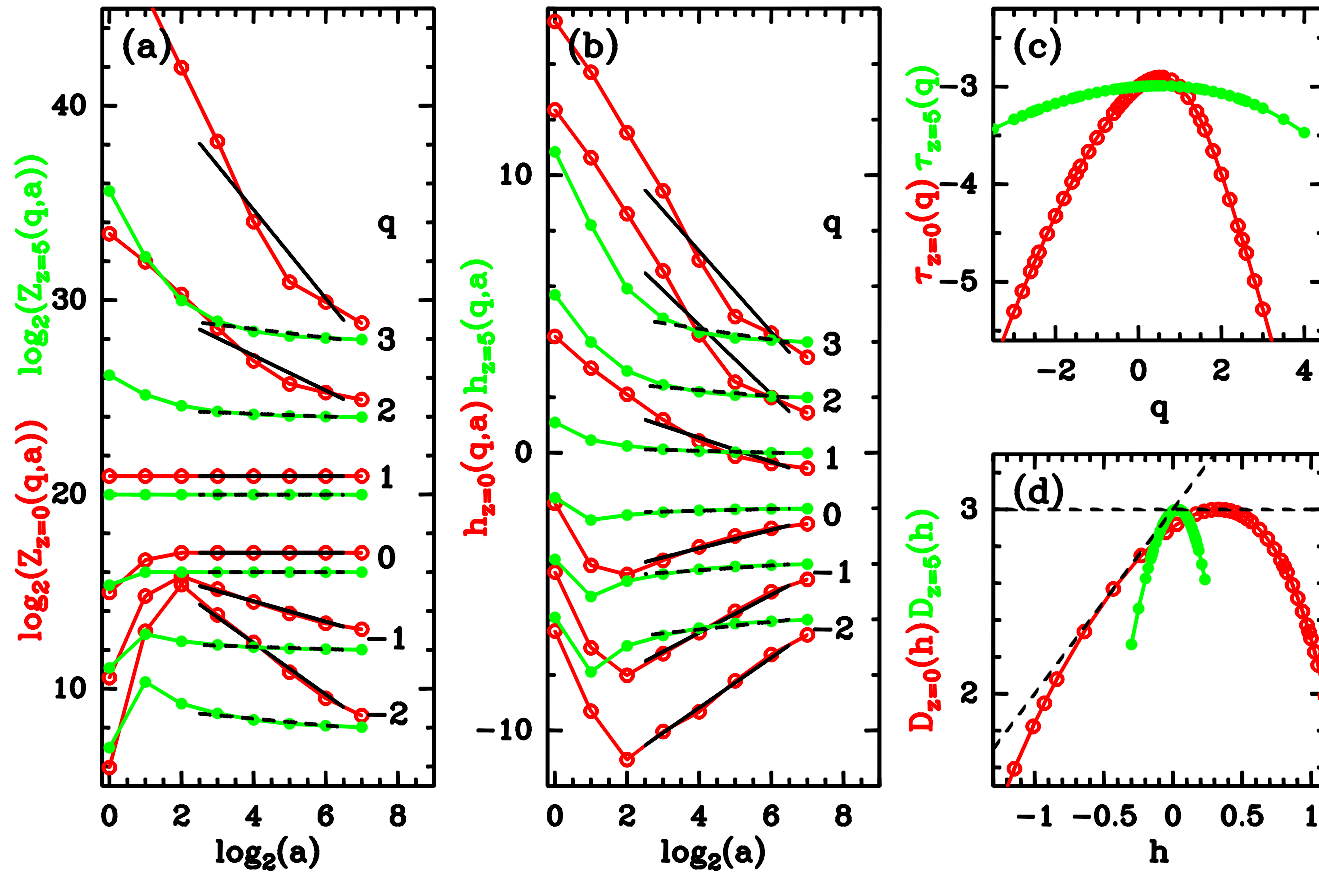


$z = 0$



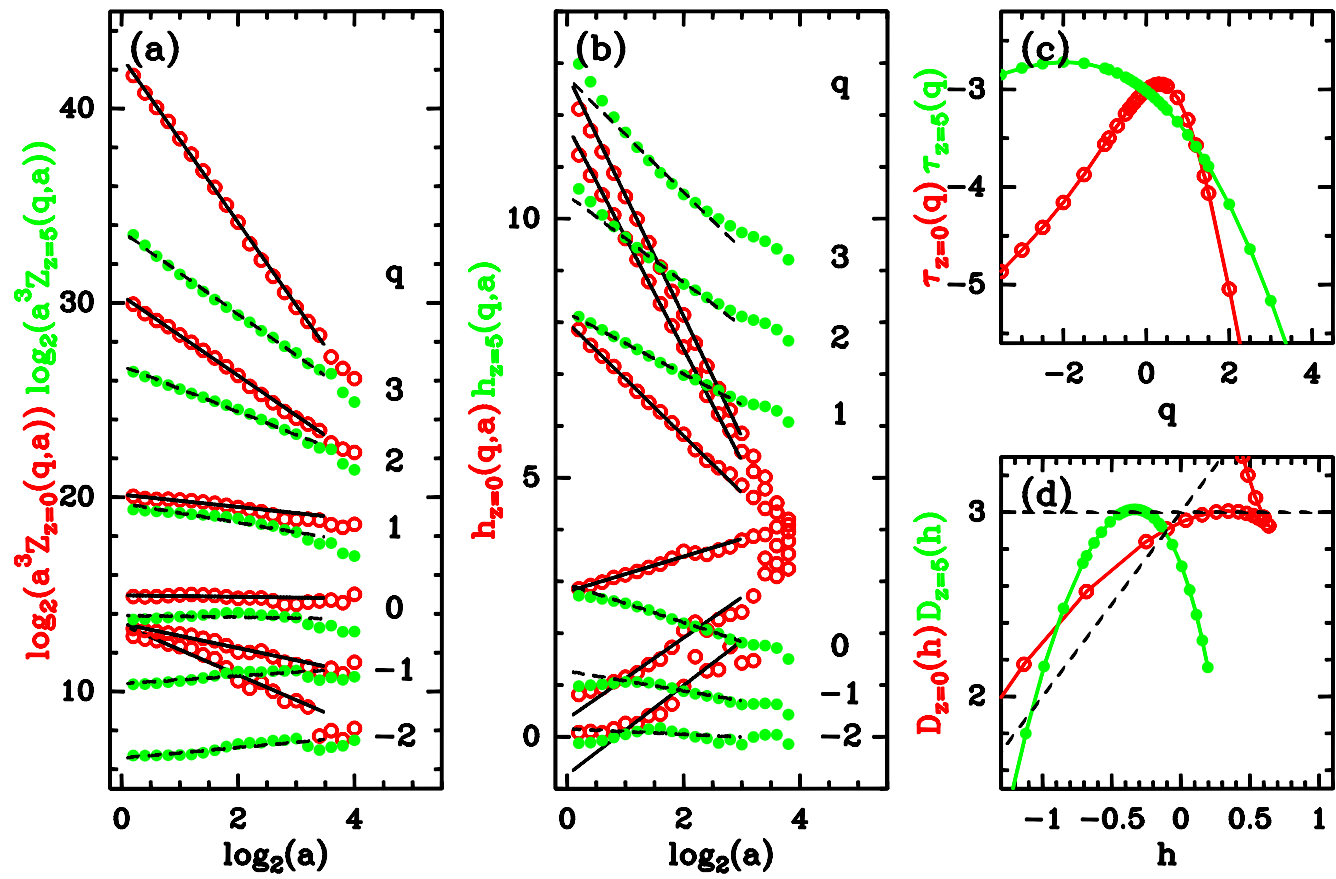


# 3D Box-Counting method: singularity spectrum ( $z = 0$ and $z = 5$ )



- NO CLEAR SCALING !!!
- perhaps scaling at large scales ??

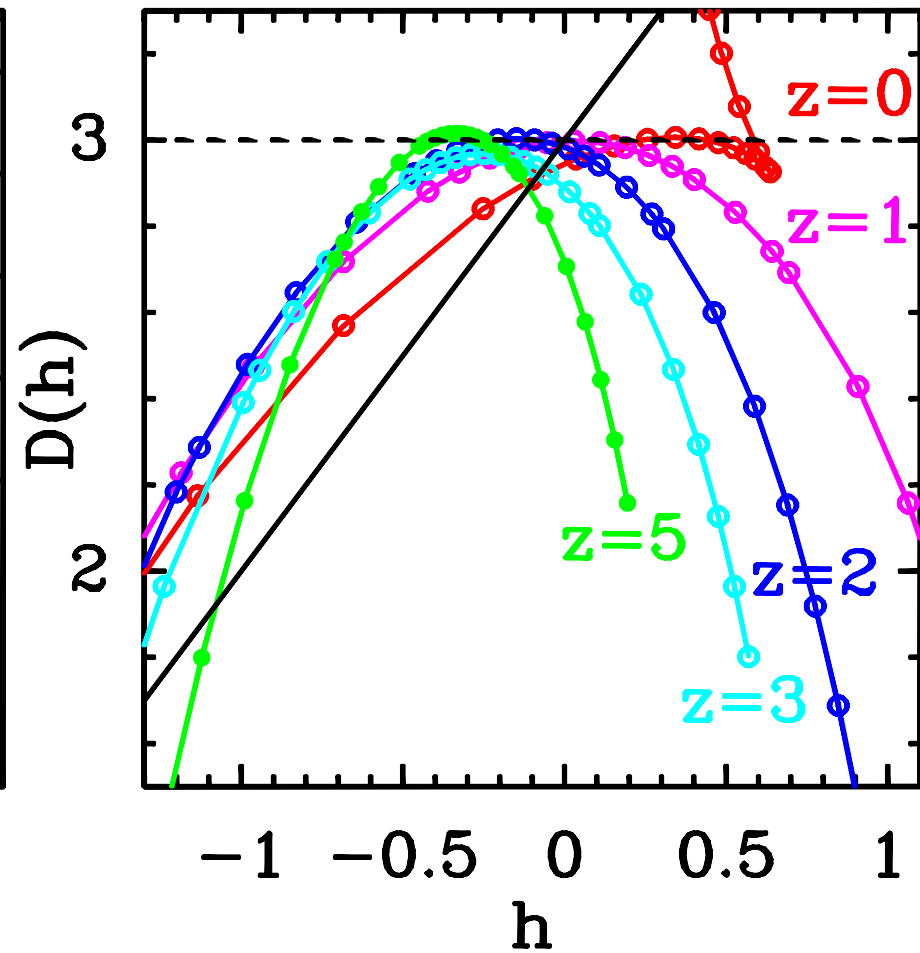
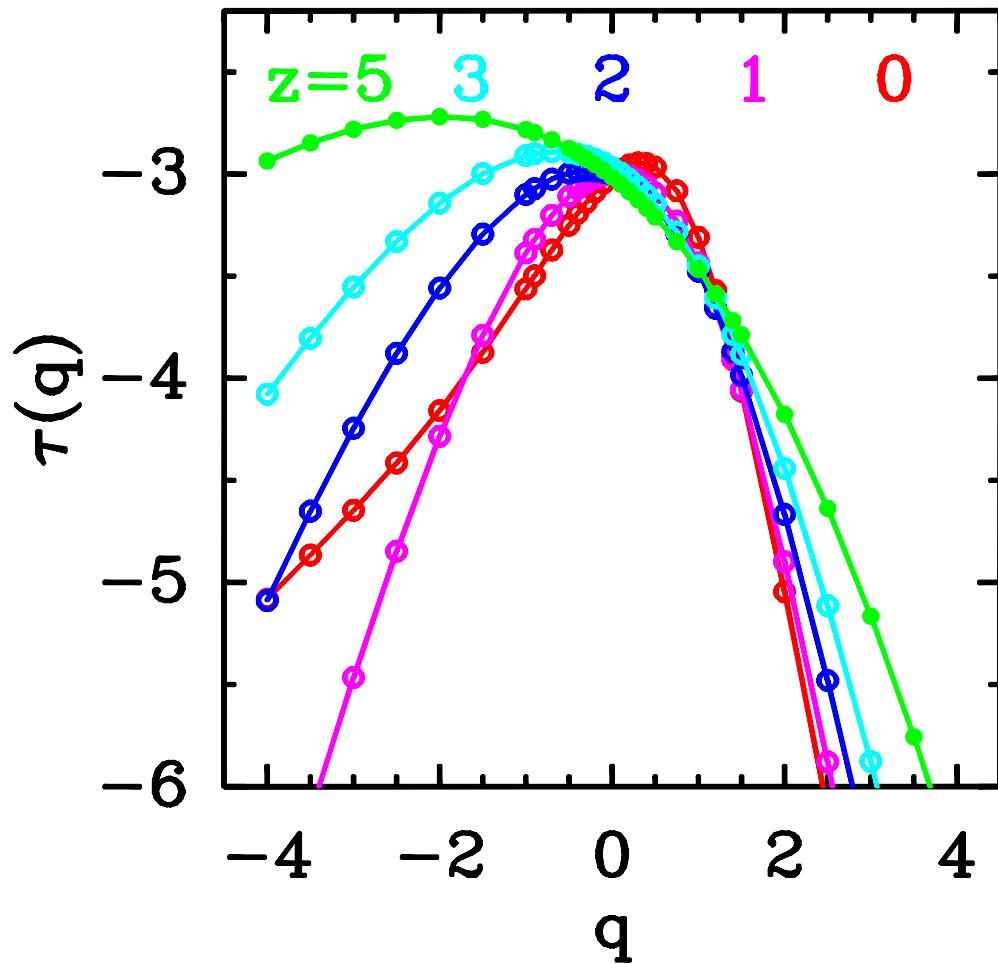
# Scalar 3D WTMM method: singularity spectrum ( $z = 0$ and $z = 5$ )



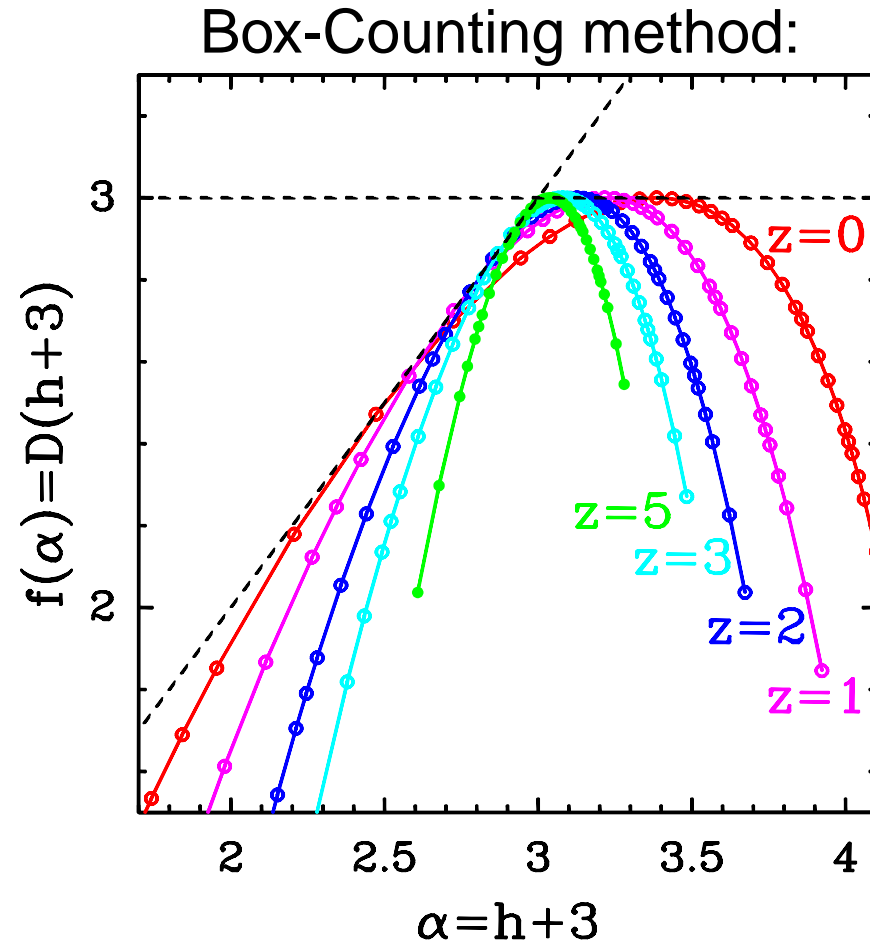
- parabolic fit:  $\tau(q) = -C_0 - C_1 q - C_2 \frac{q^2}{2}$
- intermittency coefficients  $C_2 = 1.32 \pm 0.05$   $C_2 = 0.22 \pm 0.02$

# Scalar 3D WTMM method: **singularity spectra** of galaxy distribution

Scalar 3D WTMM method:



# Box-Counting method: **singularity spectra** of galaxy distribution



- Box-counting intermittency coefficients are smaller than those computed using 3D WTMM method !!! (To be continued).

# Cancellation exponent basics

- Signed singular measure :

$$\forall A, \exists B \subset A / \mu_s(A)\mu_s(B) < 0$$

(ex : magnetic field component, ...)

- Cancellation exponent :  $\kappa = \lim_{\varepsilon \rightarrow 0} \frac{\ln \sum_i |\mu(I_{i,\varepsilon})|}{\ln(1/\varepsilon)}$ ,

where  $\{I_{i,\varepsilon}\}$  is a tiling of measure's support.

- using the wavelet transform:

$$\kappa = -D_F - \tau(q = 1)$$

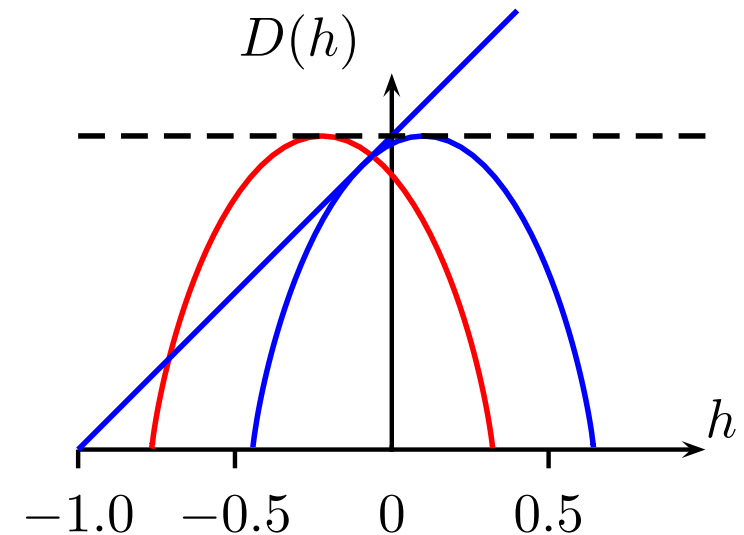
$\Rightarrow$  choice of the analyzing wavelet.

- link to the notion of **conservativity** of a multiplicative cascade:

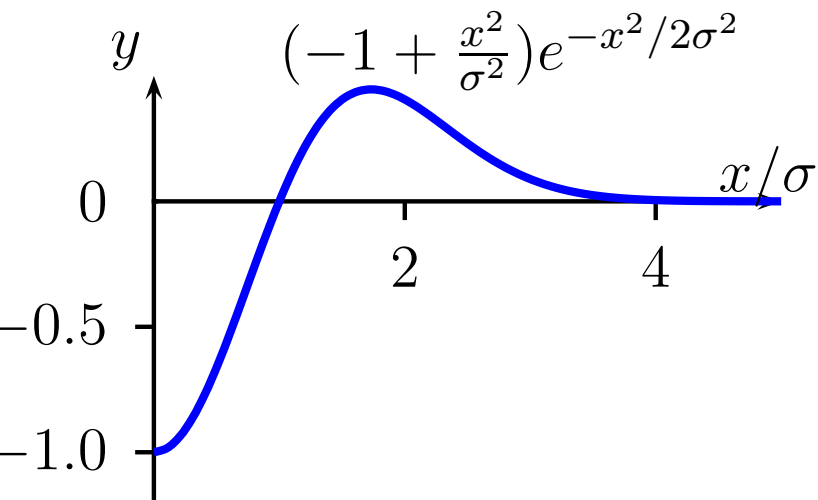
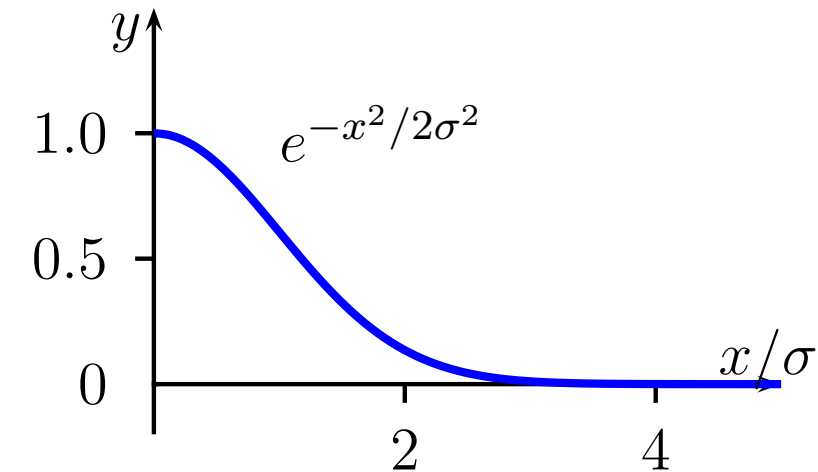
$$\kappa = -D_F - \tau(q = 1) = \frac{\ln \langle M \rangle}{\ln b} : \text{transfert rate of the measure from scale } a \text{ to scale } a/b$$

- **conservative** cascade  $\iff \kappa = 0$

- **non-conservative** cascade  $\iff \kappa \neq 0$



# Recursive filter technics in 3D

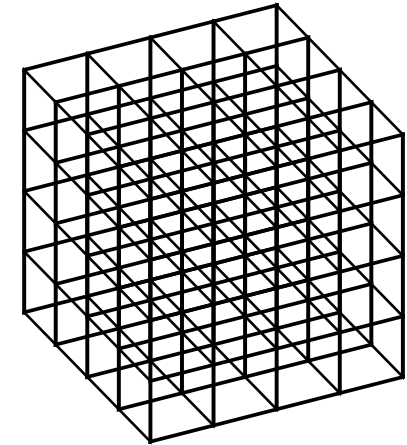
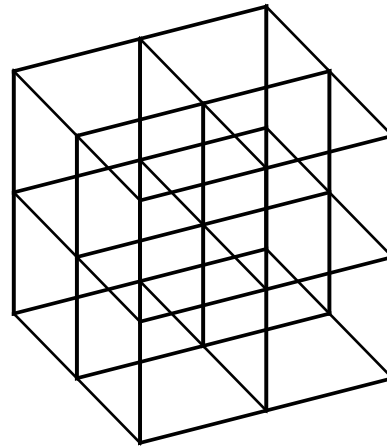
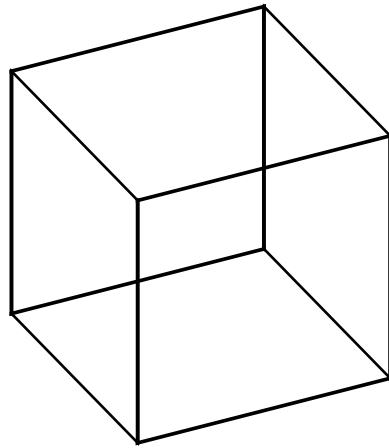


- Gaussian filter  $e^{-x^2/2\sigma^2}$  approximated by  $h_\sigma(x)$**   
 $(a_0 \cos(\omega_0 \frac{x}{\sigma}) + a_1 \sin(\omega_0 \frac{x}{\sigma})) \exp^{-b_0 \frac{x}{\sigma}}$   
 $+ (c_0 \cos(\omega_1 \frac{x}{\sigma}) + c_1 \sin(\omega_1 \frac{x}{\sigma})) \exp^{-b_1 \frac{x}{\sigma}}$
- Coefficients  $a_i, b_i, c_i$  and  $\omega_i$  are estimated by minimizing the relative quadratic error:**

$$\epsilon^2 = \frac{\sum_{i=1}^{10\sigma} (g_\sigma(i) - h_\sigma(i))^2}{\sum_{i=1}^{10\sigma} g_\sigma(i)^2}$$

- 4th order recursive equation:**  
 $y_k = n_{00}x_k + n_{11}x_{k-1} + n_{22}x_{k-2} + n_{33}x_{k-3} - d_{11}y_{k-1} - d_{22}y_{k-2} - d_{33}y_{k-3} - d_{44}y_{k-4}$
- computing time decrease in 3D: 60 % for Gaussian filter and 25 % for Mexican filter**

# Box-counting algorithms : multifractal measure



- $\mu$ : probability measure whose support  $E \subset \mathbb{R}^d$
  - singularity exponent:  $\alpha(x) = \lim_{l \rightarrow 0} \frac{\log \mu(\mathcal{B}(l, x))}{\log l}$
  - method: tile support of the measure with boxes of size  $l_i = L/2^i$
  - partition functions:  $S_q(l_i) = \sum_{\mu(\mathcal{B}) \neq 0} [\mu(\mathcal{B})]^q = \langle \mu^q \rangle$
- multifractal spectrum:  $\tau(q) = \lim_{l \rightarrow 0} \frac{\log S_q(l)}{\log l}$ ;

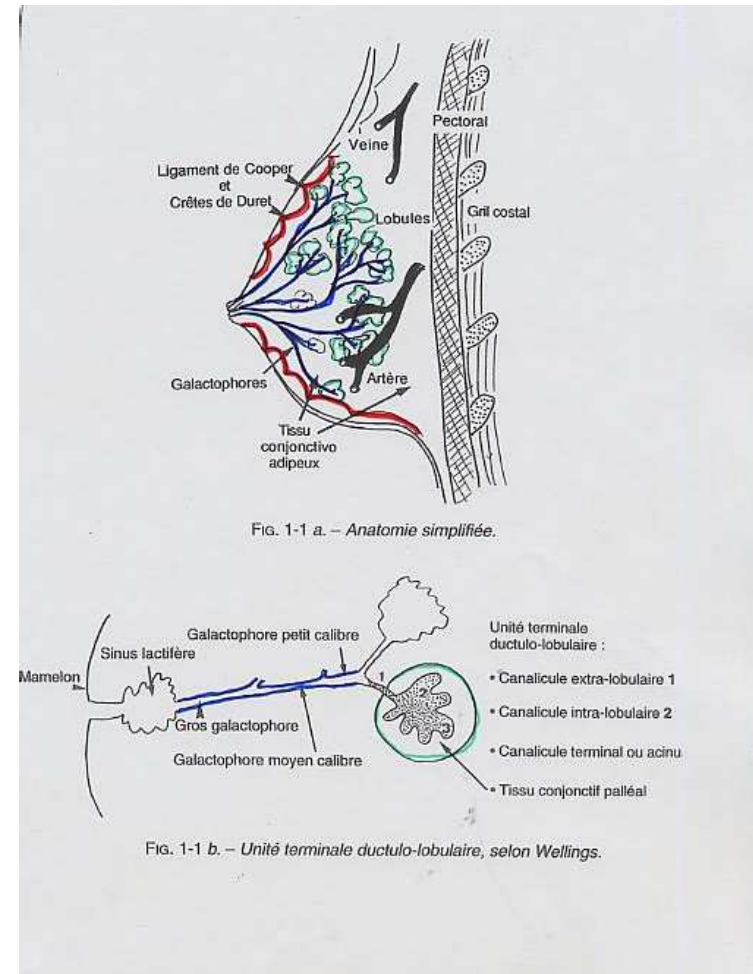
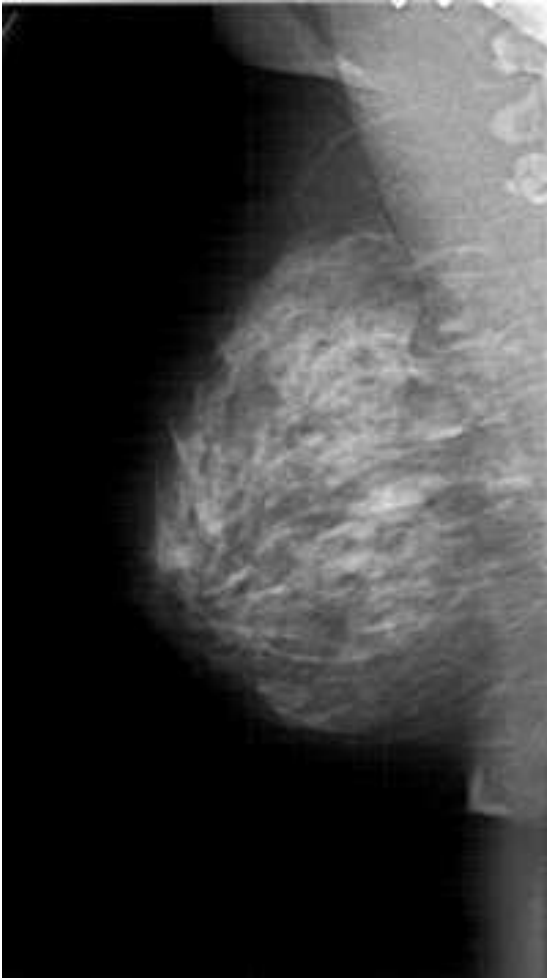
$$f(\alpha) = D(h = \alpha - d) = \min_q (\alpha q - d q - \tau(q))$$





# Mammography and breast anatomy

Goals : using WTMM method to **diagnosis help** of breast cancer



# What is breast cancer ?

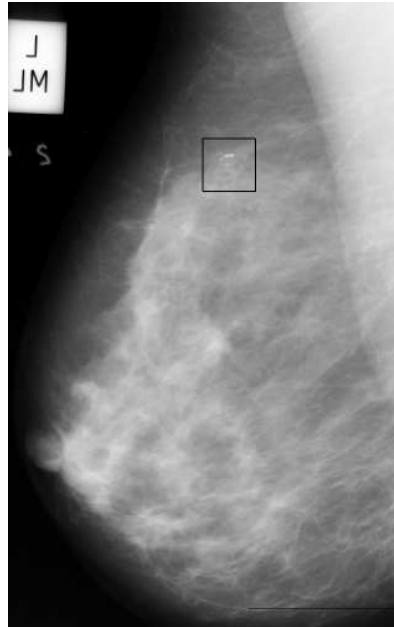
---

- malignant tumor of mammal gland
- incidence : 30000 new case each year in France
- prevention is very difficult (as opposed to lung cancer)
- hereditarity : 5 to 10 % only (BRCA1/2 genes)
- forecast depends on the tumoral volume at diagnosis  
⇒ **SCREENING** using mammography

# Radiological anomalies



● Opacities



● Calcifications



● Architectural  
distorcions

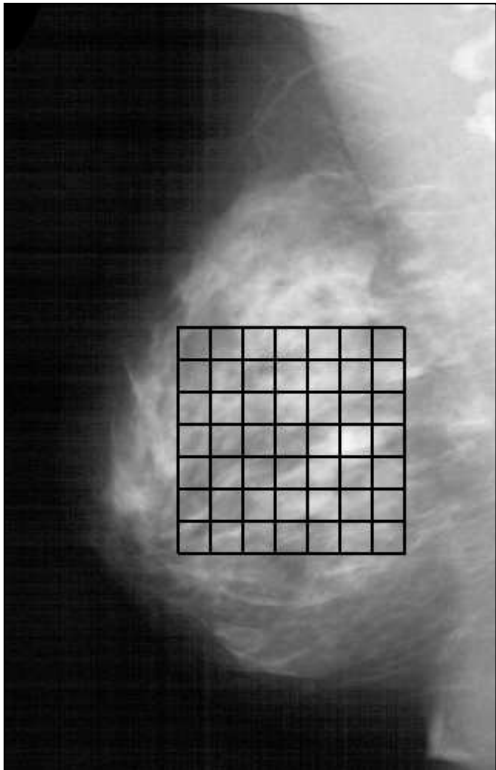
# Digitalized mammographies : texture analysis

- dense breasts : more difficult to diagnose
- only 2 classes of monofractal properties

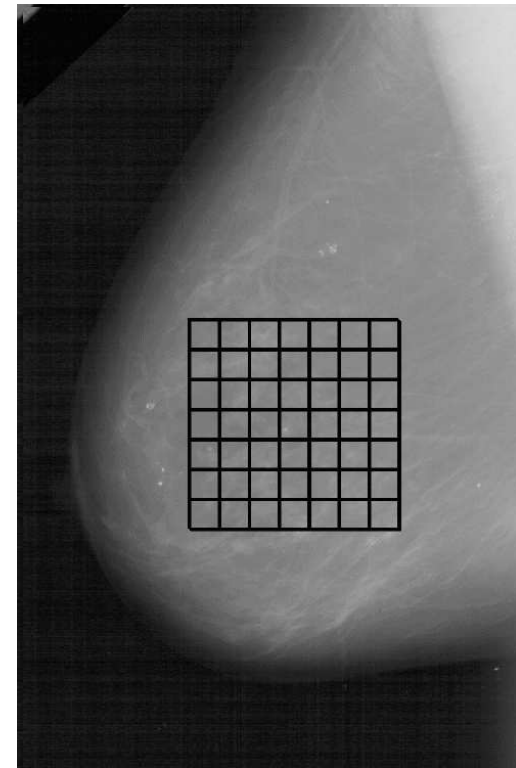
**Digital Database for Screening Mammography:**

<http://marathon.csee.usf.edu/Mammography/Database.html>

**Dense breast**

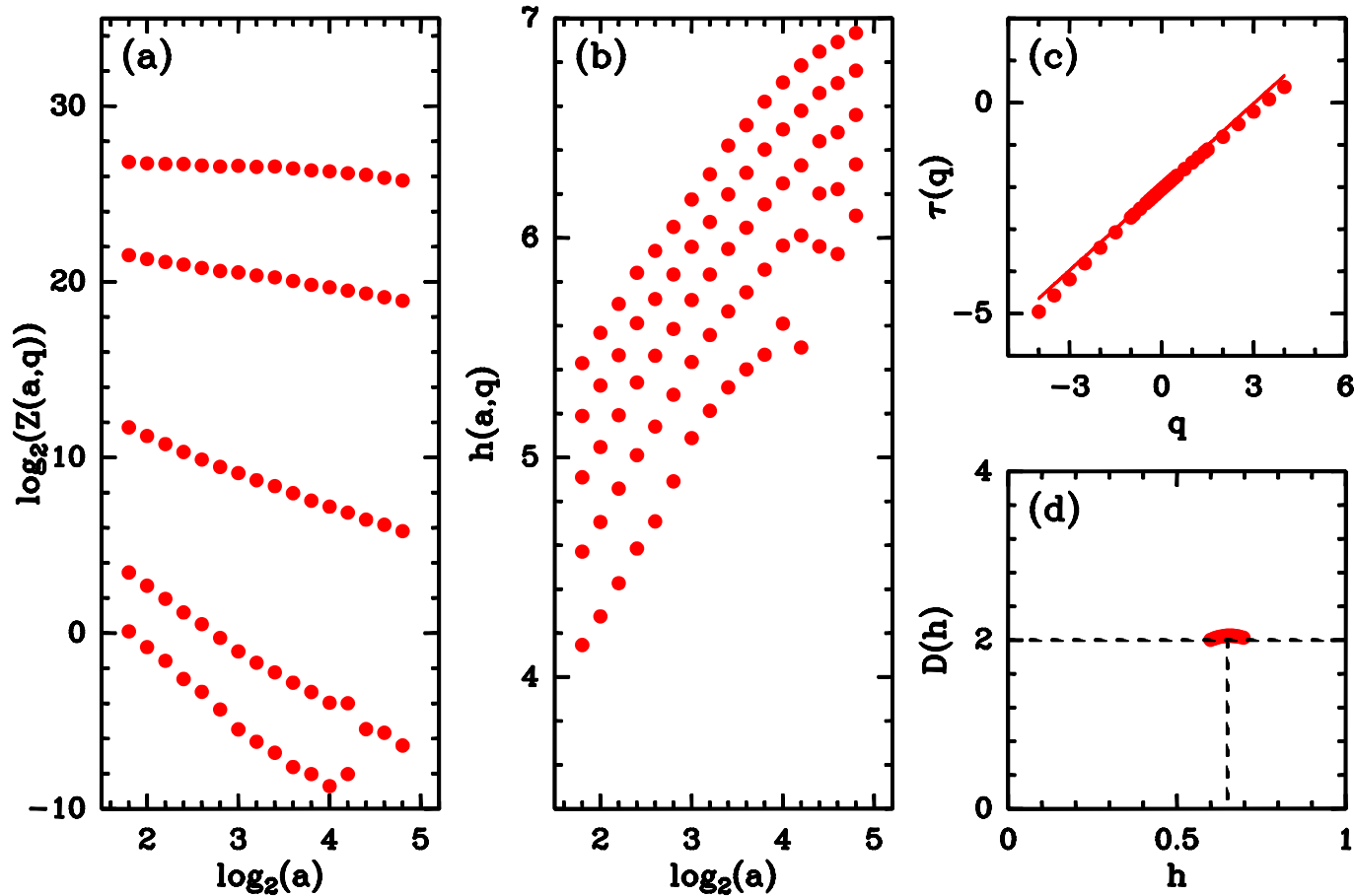


**Fatty breast**



# Application of 2D WTMM methodology in mammography

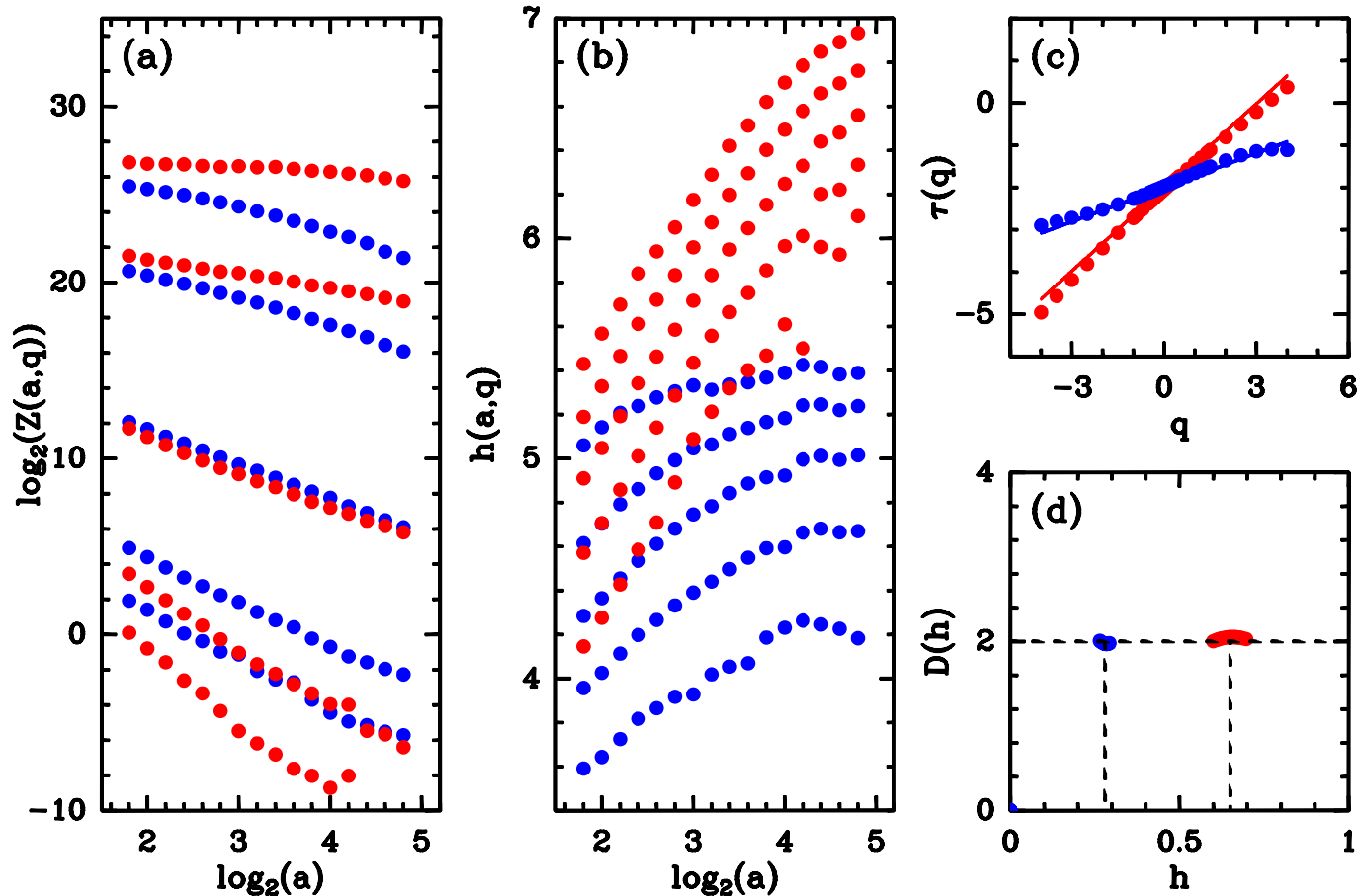
Tissue classification : **dense**



● **Dense breast :**  
monofractal,  $H = 0.65$   
**persitent** correlations

# Application of 2D WTMM methodology in mammography

## Tissue classification : dense vs fatty



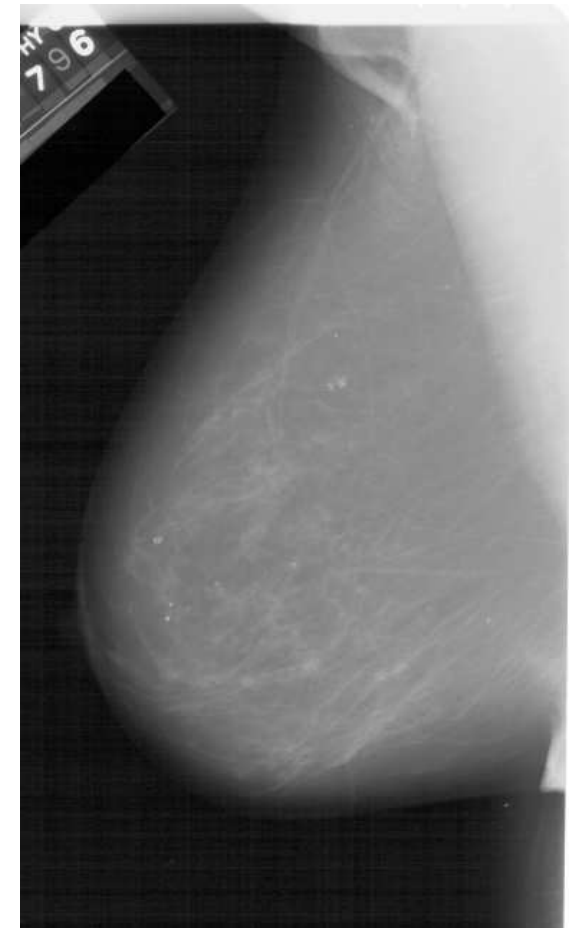
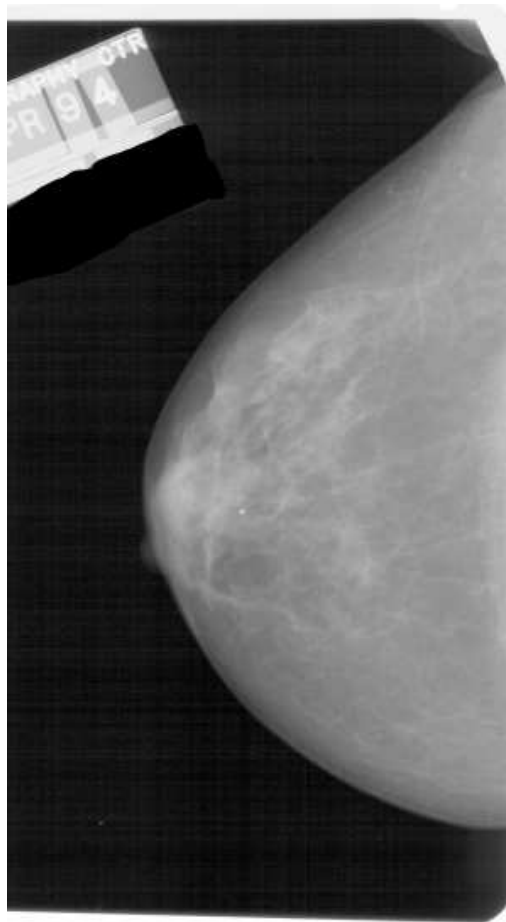
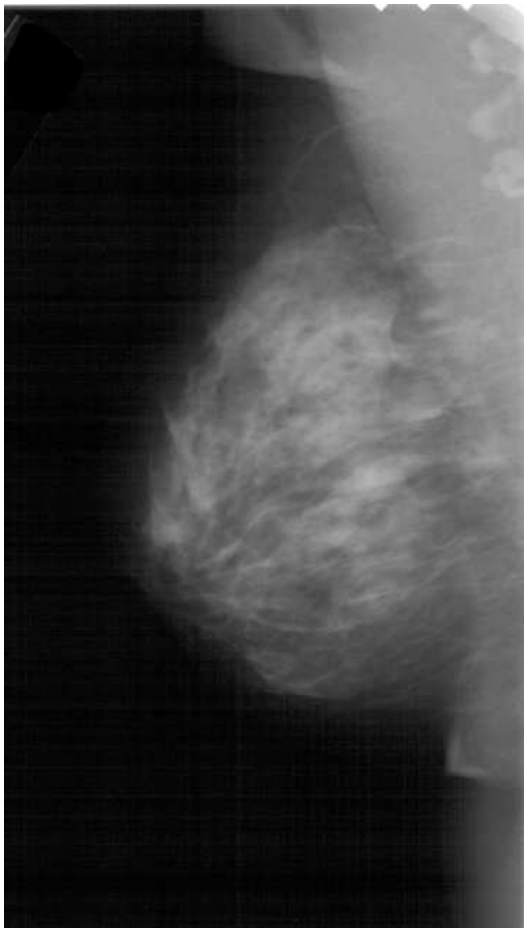
- Dense breast :**  
monofractal,  $H = 0.65$   
persistent correlations
- Fatty breast :**  
monofractal,  $H = 0.30$   
anti-persistent correlations



# Application to digitalized mammographies

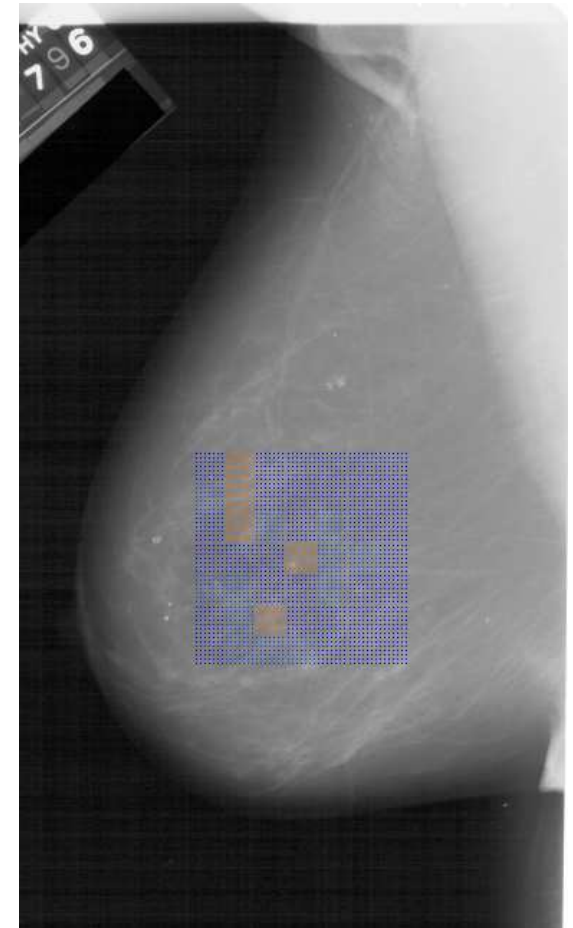
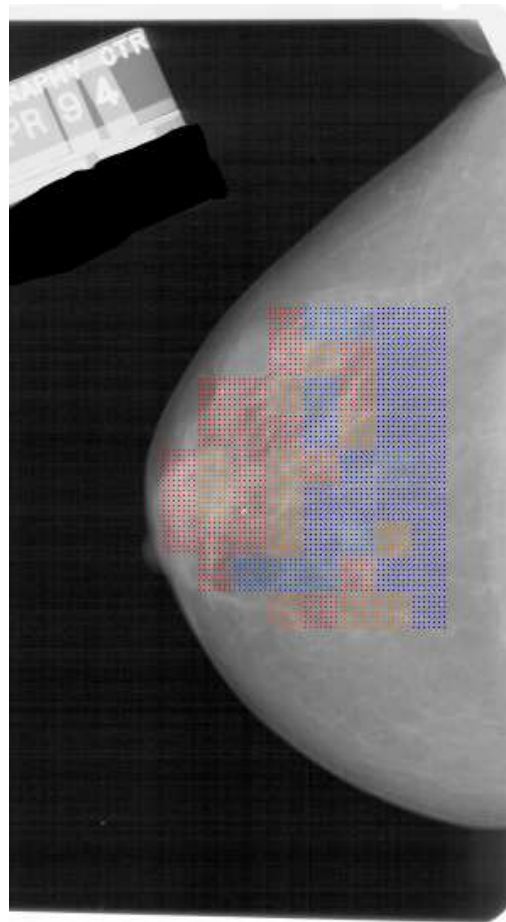
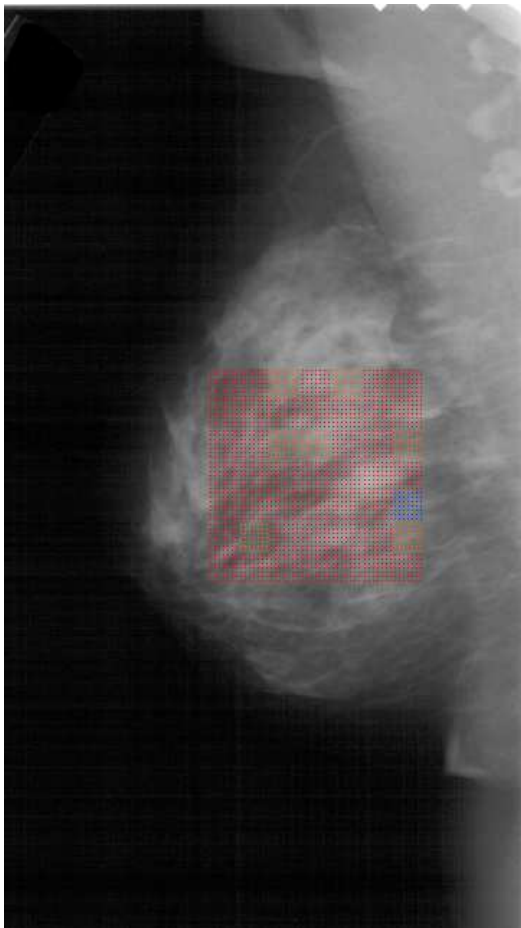
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Colored Maps :  
segmentation of **dense**  $h > 0.52$  areas and **fatty**  $h < 0.38$  areas



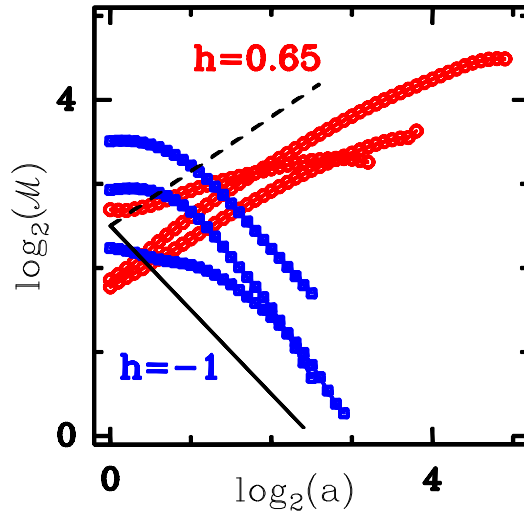
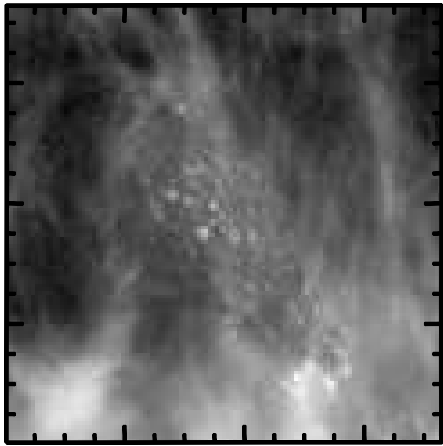
# Application to digitalized mammographies

Colored Maps :  
segmentation of **dense**  $h > 0.52$  areas and **fatty**  $h < 0.38$  areas



# Microcalcifications detection

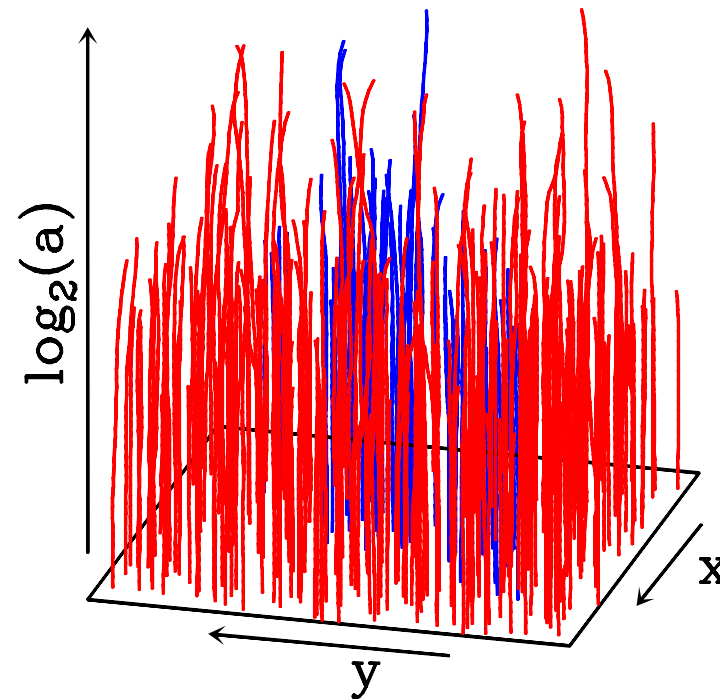
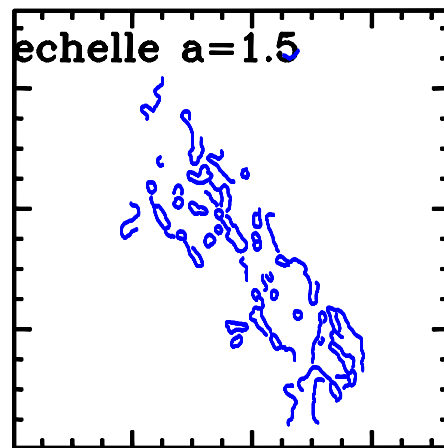
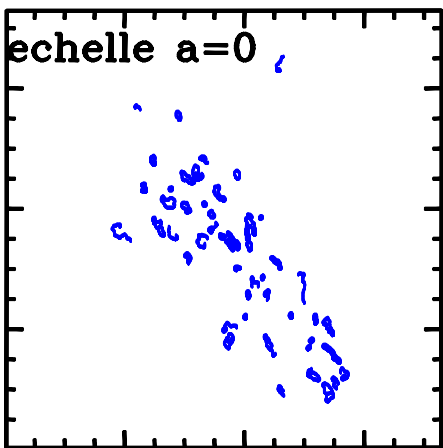
Segmentation of WT skeleton lines :  
microcalcifications vs background texture



● Background lines

● Microcalcifications

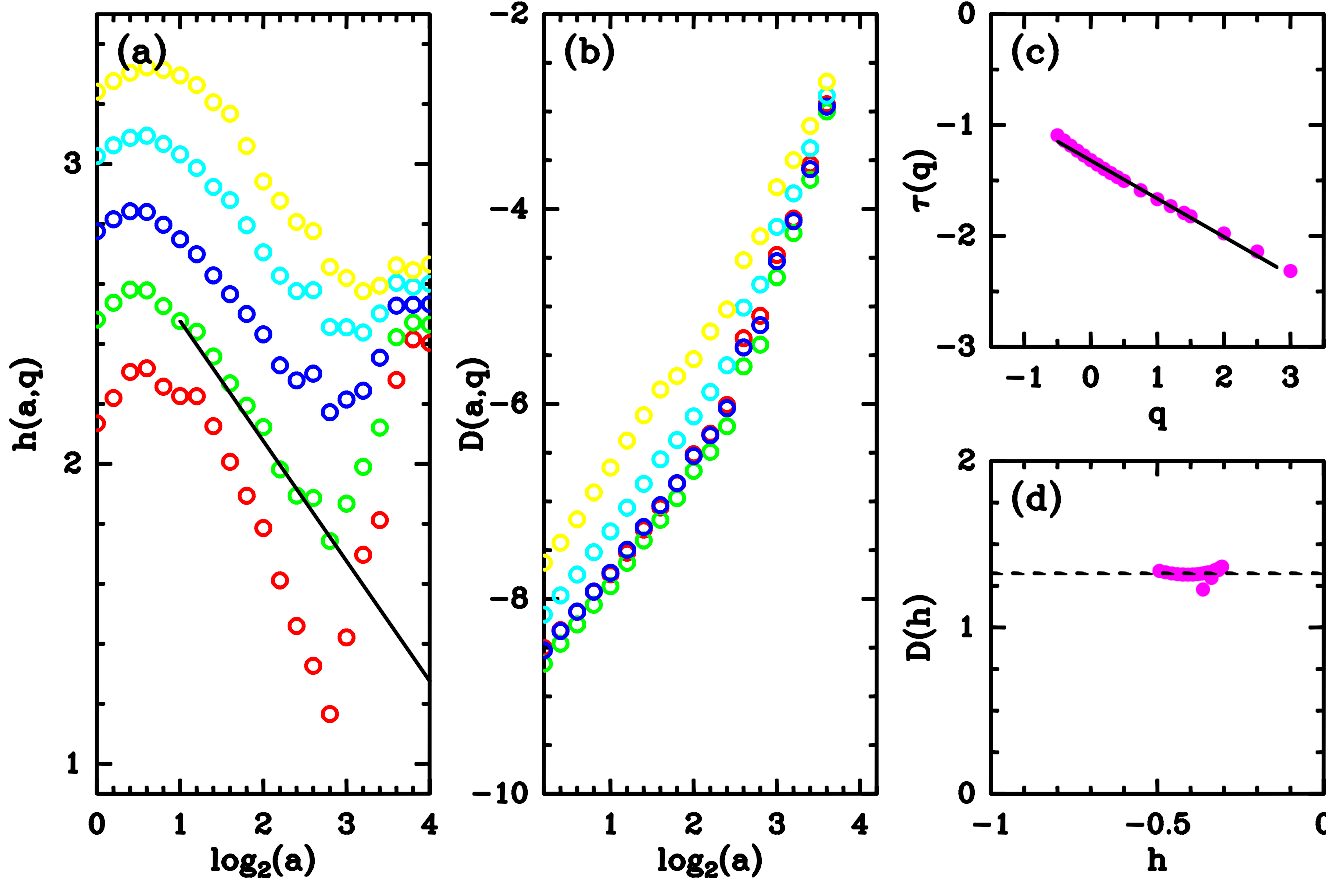
almost-punctual objects behave like  
'Dirac' shapes ( $h = -1$ )



# Cluster of microcalcifications

Study of microcalcification spatial distribution

Partition functions :



$$D_F = 1.3$$

observation :

fractal ramification of  
cluster of  
microcalcifications  
( $1 < D_F < 2$ ) seems to be  
correlated to the  
pathology's malignancy

# Conclusions and prospects (1)

---

- the **2D WTMM method** provides a framework for an automated measure of the breast **radio-density** and for studying the **fractal geometry of clusters of microcalcifications**.
- further study is necessary to validate quantitatively how far measuring the fractal dimension  $D_F$  could improve computer-aided diagnosis systems **benign/malignant**