

WARPED INFINITELY DIVISIBLE CASCADES : BEYOND POWER LAWS

P. Chainais ,

LIMOS UMR 6158,
ISIMA - Univ. Clermont II
Clermont-Ferrand

R. Riedi ,

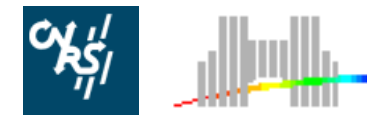
Dept of ECE,
Rice University,
Houston Texas, USA

P. Abry

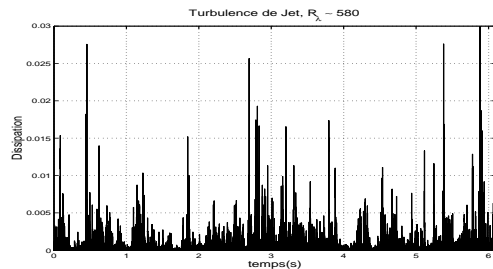
CNRS UMR 5672,
Laboratoire de Physique,
E.N.S. Lyon



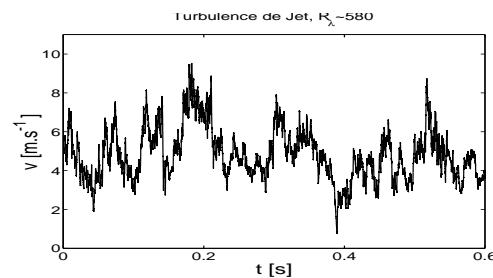
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Scaling laws in turbulence



dissipation $\mathbb{E} \varepsilon_r^q \sim r^{\tau(q)}$



velocity $\mathbb{E} |\delta v_r|^q \sim r^{q/3 + \tau(q/3)}$

ANALYSIS

multifractal formalism ($\sim \tau^{\zeta(q)}$)

log-inf. div. casc. ($H(q) \cdot n(\tau)$)

multiplicative cascades

\Rightarrow phenomenology

(Richardson's cascade)

SYNTHESIS

binomial cascades,
wavelet coeff., MRW...

multiplicative cascades

\Rightarrow algorithms

(binomial, wavelets...)

Infinitely divisible scaling laws

$$\mathbb{E} |\delta_\tau X|^q = c_q \exp[-H(q) \cdot n(\tau)]$$

✓ $n(\tau) = -\log \tau \implies \mathbb{E} |\delta_\tau X|^q = c_q \tau^{H(q)} \equiv c_q \tau^{\zeta(q)}$
(scale invariance)

✓ $n(\tau) \neq -\log \tau \implies$ towards a non-scale invariant analysis :

– turbulence Castaing et al. 1990

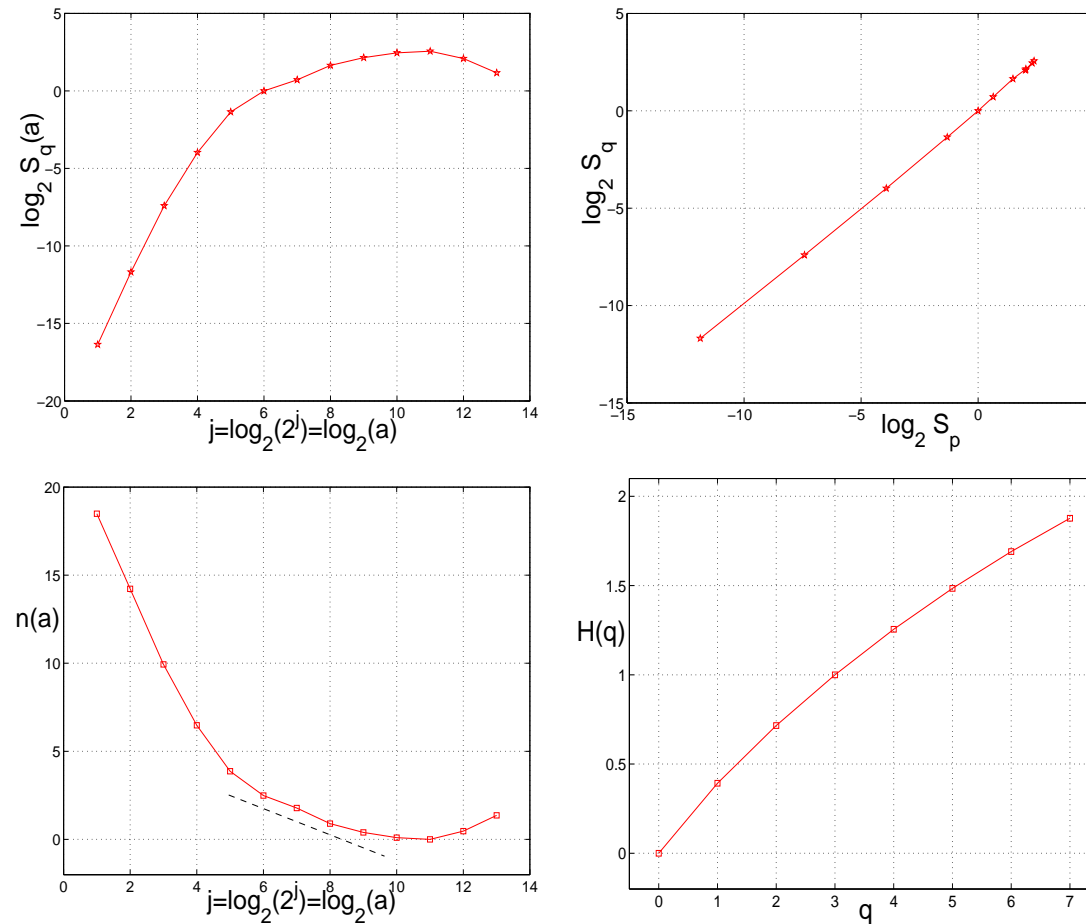
– internet traffic Veitch et al. 2000

✓ multiplicative interpretation

**How to build a signal
with infinitely divisible scaling laws ?**

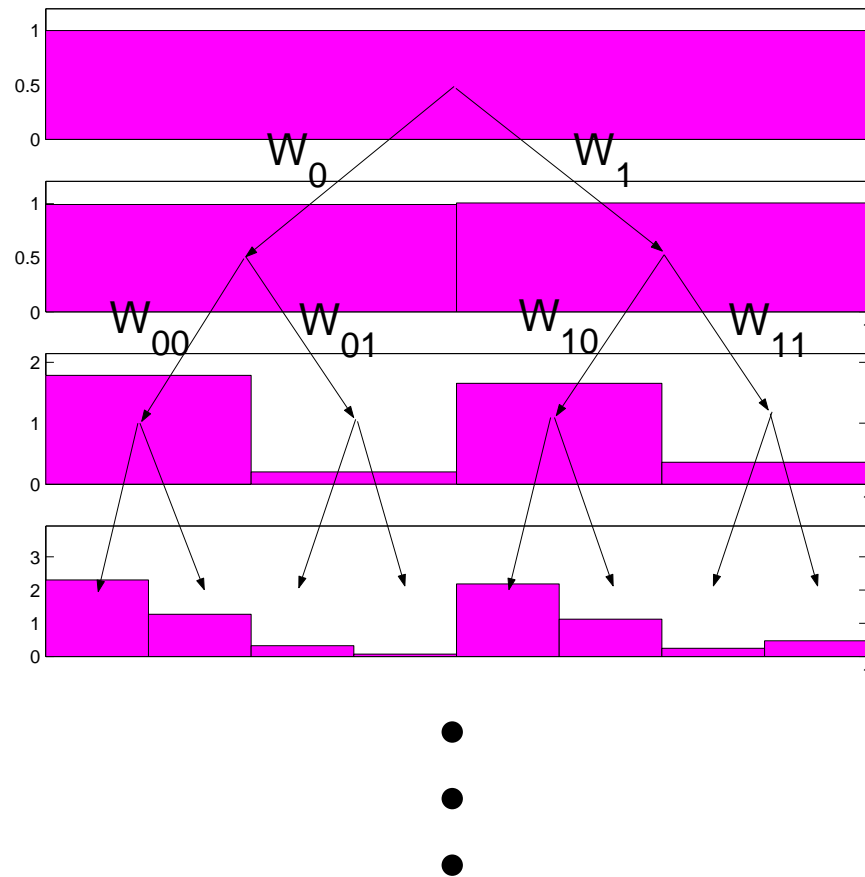
with "nice properties" : *stationary increments*, etc...

Example : ID scaling in turbulence

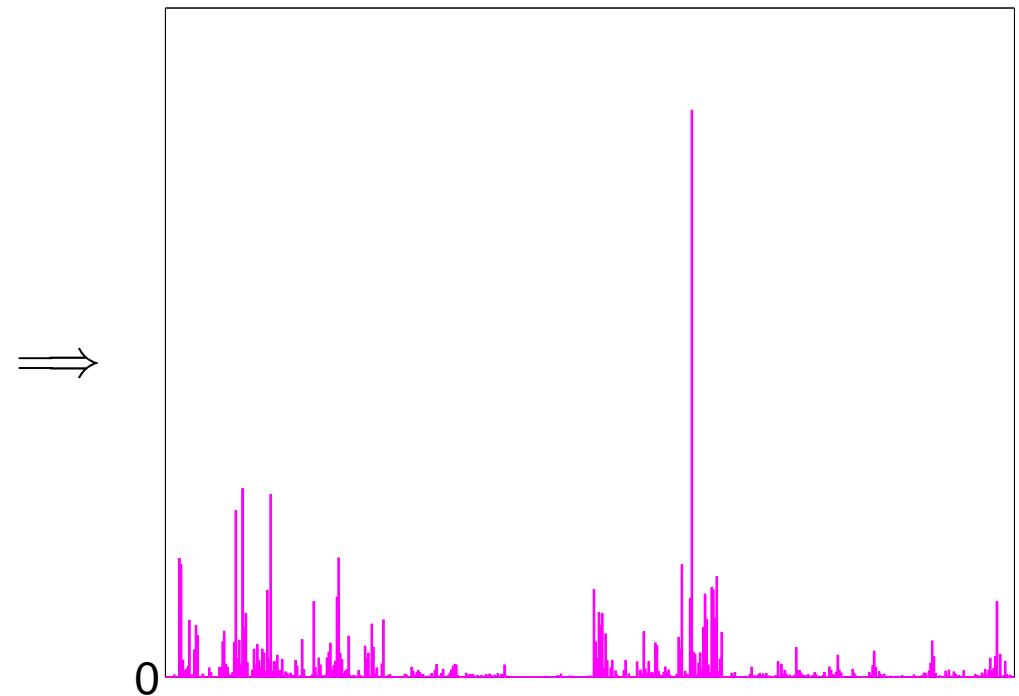


Binomial cascades

tree structure

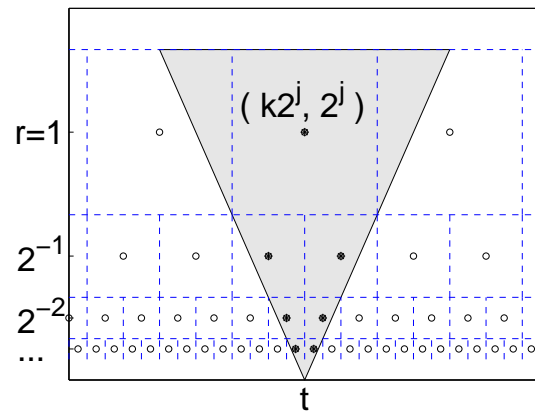


irregular signal



From binomial to infinitely divisible cascades

Binomial

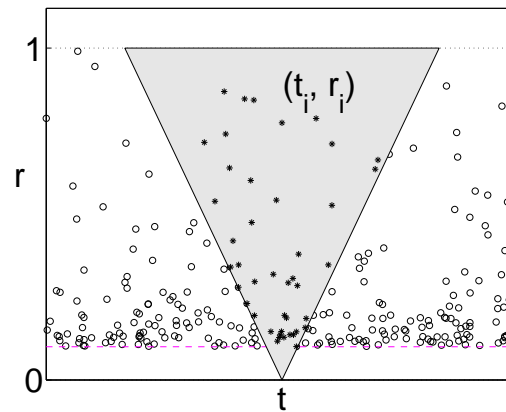


$$Q_r(t) = \prod_j \Lambda_j(t),$$

$$r = 2^j \text{ only}$$

Mandelbrot 1974

Compound Poisson



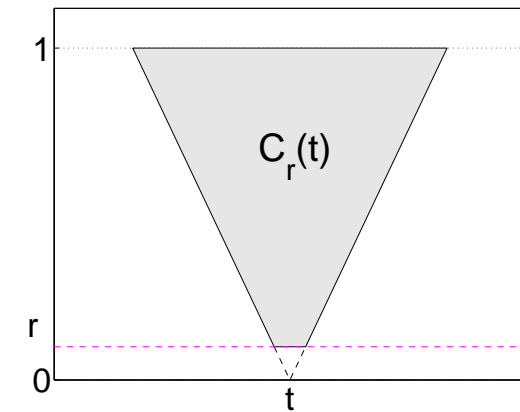
$$Q_r(t) \propto \prod_{(t_i, r_i)} W_i$$

t_i : uniform \Rightarrow stationary

r_i : $1/r^2 \Rightarrow$ scaling

Barral & Mandelbrot 2002

Infinitely Divisible



$$Q_r(t) \propto \exp M(C_r(t))$$

CONTINUOUS

MULT. CASCADE

Schmitt & Marsan 2001

Muzy & Bacry 2002

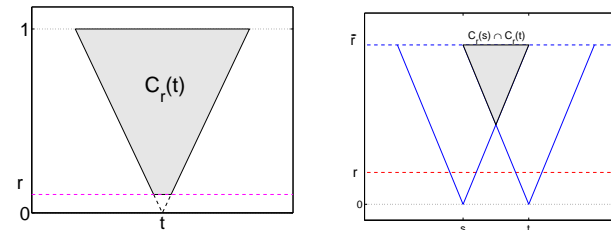
Ch. & Riedi & Abry 2003

Infinitely divisible noise (IDC noise)

$\mathbf{M}(\mathcal{C}_r(\mathbf{t}))$: moment gen. func. = $\exp[-m(\mathcal{C}_r) \rho(q)]$ (inv. éch. = $r\rho(q)$)

- G = infinitely divisible distr., moment gen. func. $\tilde{G}(q) = e^{-\rho(q)}$,
- positive measure $dm(t, r)$ [i. é. = $\frac{(1+\delta(1-r))dt dr}{r^2}$],

$$Q_r(t) = \frac{\exp[M(\mathcal{C}_r(t))]}{\mathbb{E}[\exp M(\mathcal{C}_r(t))]}$$



$\implies Q_r(t)$ is stationary

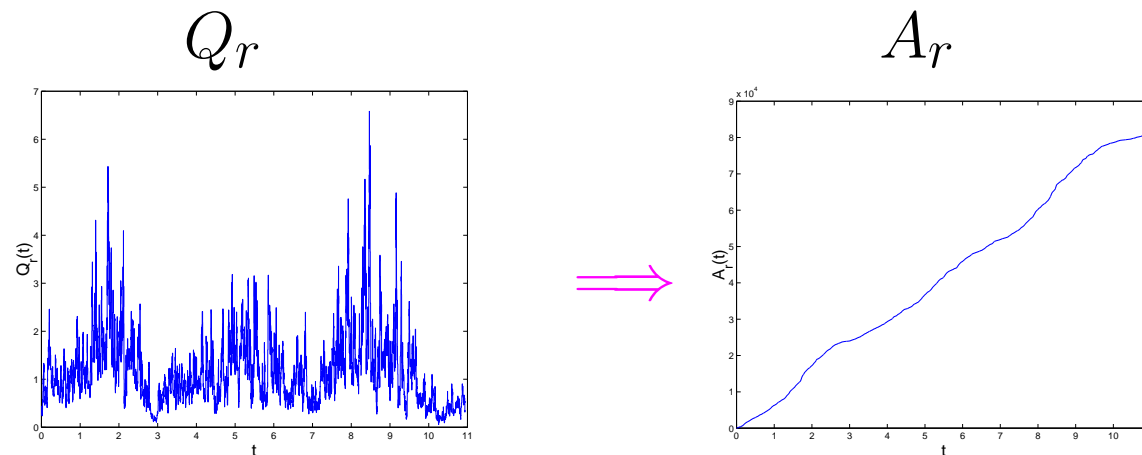
$$\varphi(q) = \rho(q) - q\rho(1) \implies \mathbb{E}[Q_r^q] = \exp[-\varphi(q)m(\mathcal{C}_r)]$$

$$[\text{i.é. } \mathbb{E}[Q_r^q] = e^{-\varphi(q)} \cdot r^{\varphi(q)}]$$

Infinitely divisible motion (IDC motion)

Pb : Q_r degenerates when $r \rightarrow 0 \dots$

Solution : $A_r(t) = \int_0^t Q_r(s) ds \implies \mathbb{E}A_r(t) = t$



$A(t) = \lim_{r \rightarrow 0} A_r(t) \dots$ has stationary increments

and $\mathbb{E}\delta_\tau A^q \sim \tau^q \exp[-\varphi(q)m(\mathcal{C}_\tau)]$

IDC cascades obey IDC scaling

$$\text{locally averaged dissipation} \quad \equiv \quad \varepsilon_{\tau}(t) = \frac{[A(t + \tau) - A(t)]}{\tau}$$

$$[\text{in practice : } \varepsilon_{\tau}^{(r)}(t) = \frac{1}{\tau} \int_t^{t+\tau} Q_r(t) dt]$$

Scaling of moments \equiv Inf. Div. scaling laws

$$\mathbb{E} \varepsilon_{\tau}^q \sim \exp[-\varphi(q) \cdot m(\mathcal{C}_{\tau})] \equiv \exp[-\mathbf{H}(q) \cdot \mathbf{n}(\tau)]$$

\implies **synthesis of a turbulent dissipation with Inf. Div. scaling**

Infinitely divisible random walk (IDC random walk)

fractional **B**rownian **m**otion B_H , $A(t)$ an Inf. Div. measure,

$$V_H(t) = B_H(A(t)), \quad t \in \mathbb{R}^+$$

Rappel $\left\{ \begin{array}{l} B_H \text{ has stationary increments} \\ \forall t \in \mathbb{R}^+, \mathbb{E}|B_H(t)|^q = t^{qH} \cdot \mathbb{E}|B_H(1)|^q, \end{array} \right.$

V_H has stationary increments,

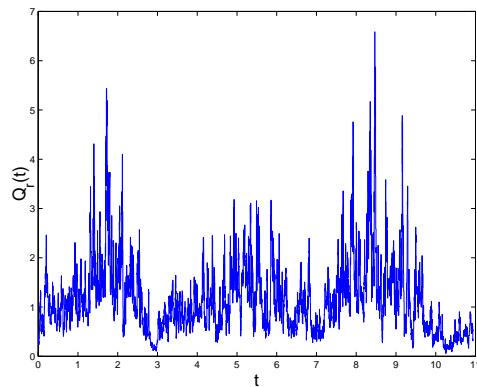
positive/negative fluctuations,

and

$$\mathbb{E}|\delta_\tau V_H|^q \sim \tau^{qH} \exp[-\varphi(qH)m(\mathcal{C}_\tau)]$$

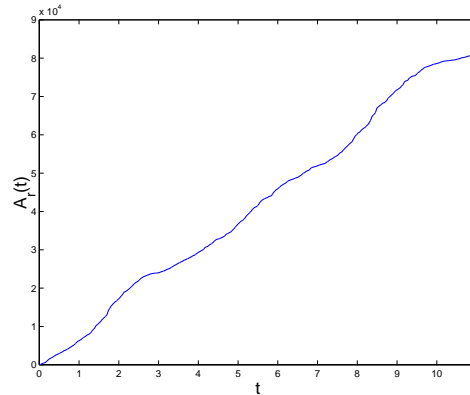
In summary...

noise
(density)



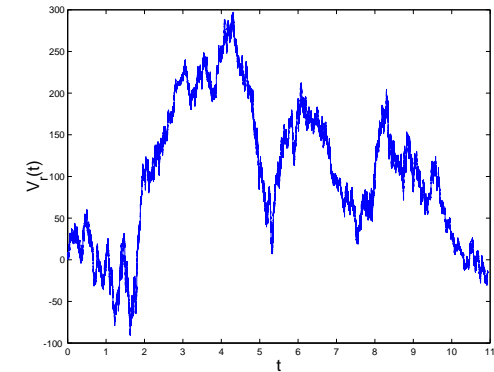
$$\int Q_r$$

motion
(measure)



$$B_H(A)$$

random walk



$$\mathbb{E}[Q_r^q]$$

$$\equiv$$

$$\exp[-\varphi(q)m(\mathcal{C}_r)]$$

$$\mathbb{E}\delta_\tau A^q$$

$$\mathfrak{N}$$

$$\tau^q \exp[-\varphi(q)m(\mathcal{C}_\tau)]$$

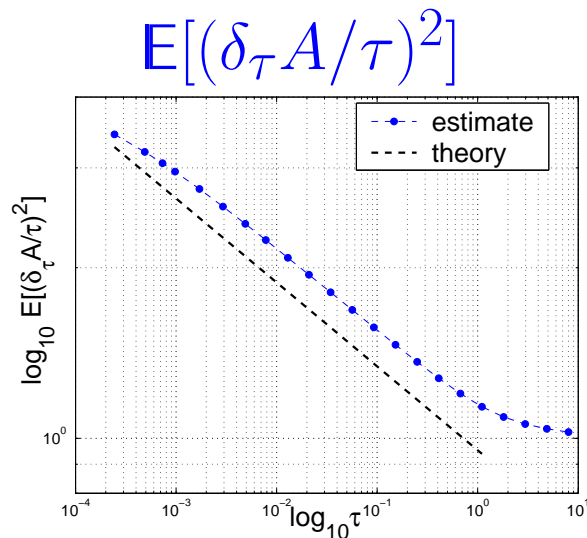
$$\mathbb{E}|\delta_\tau V_H|^q$$

$$\mathfrak{N}$$

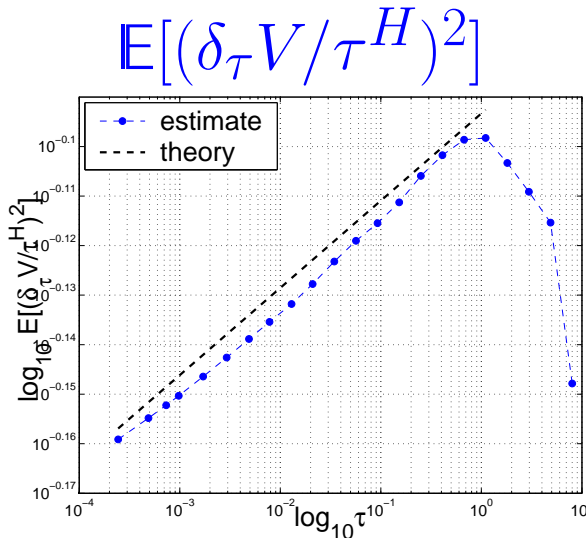
$$\tau^{qH} \exp[-\varphi(qH)m(\mathcal{C}_\tau)]$$

- continuous time ($t \in \mathbb{R}^+$), stationary increments, continuous scale invariance, $\forall \varphi(q)$ of an Inf. Div. distribution,
- **MATLAB** procedures

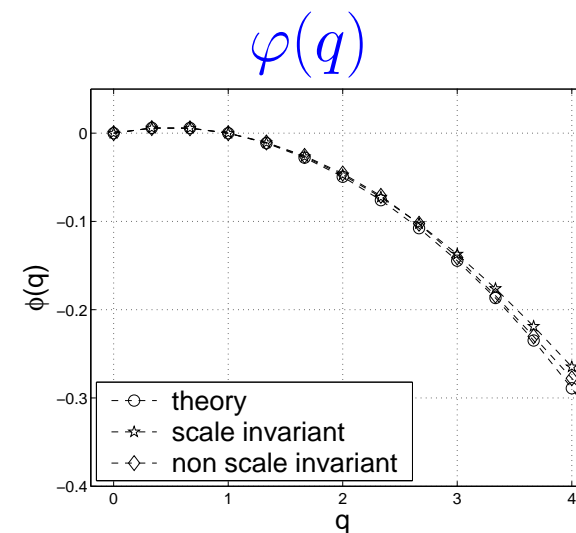
Scale invariance and power laws



$\varphi(2) \log \tau$



$\varphi(2H) \log \tau$

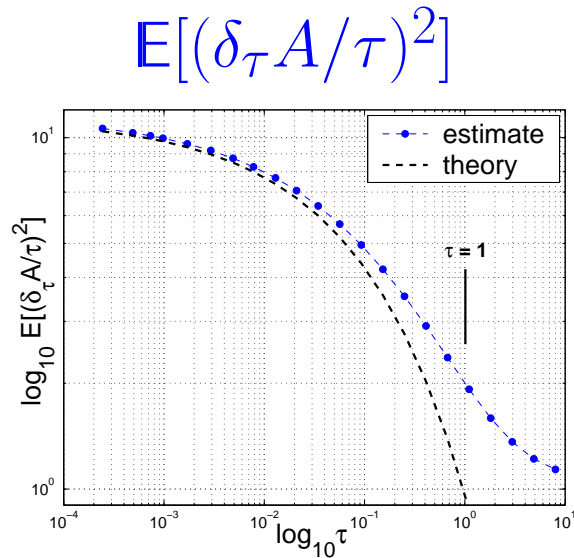


linear behaviours in log-log diagrams

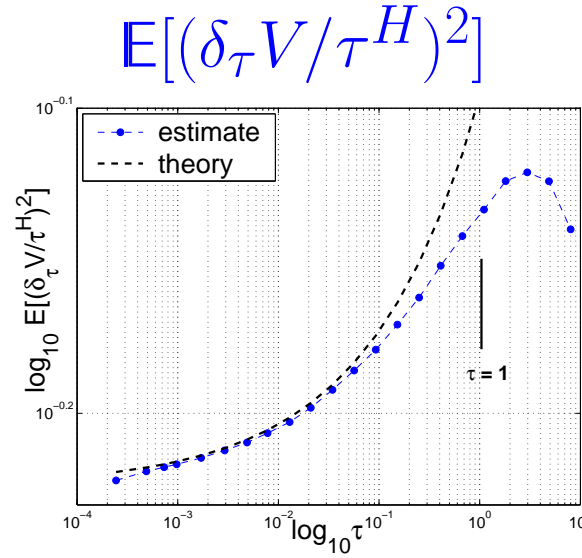


POWER LAWS

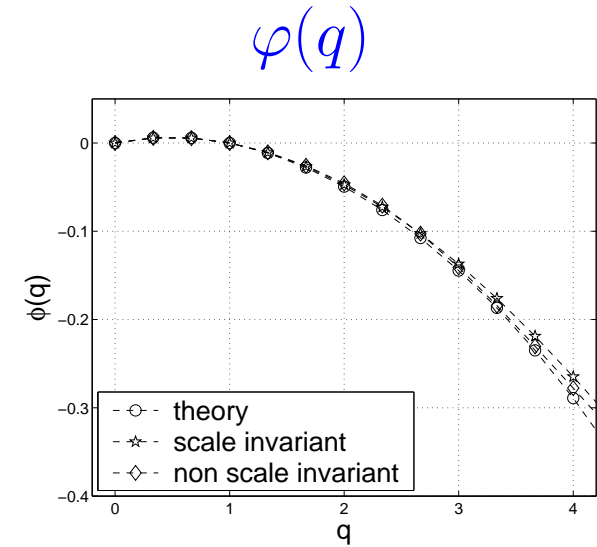
Beyond power laws...



$$-\varphi(2)m(\mathcal{C}_\tau)$$



$$-\varphi(2H)m(\mathcal{C}_\tau)$$



non-linear behaviours in log-log diagrams

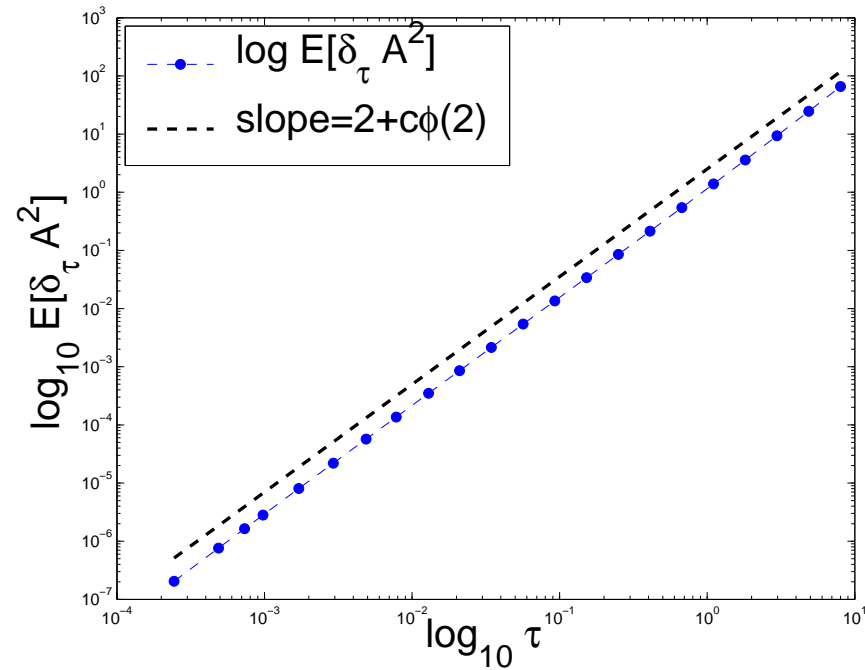


~~POWER LAWS~~

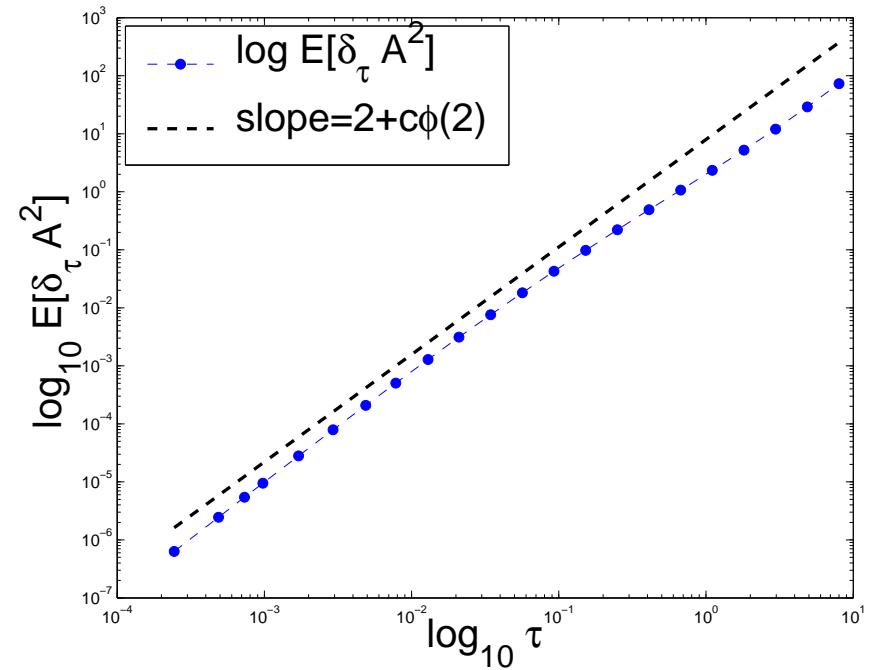
warped cascade

$E\delta_\tau A^2$ multifractal / warped

MULTIFRACTAL



WARPED



A simple example

$$dm(t, r) = \frac{dt dr}{r^{2+\beta}}$$

yields

$$m(\mathcal{C}_\tau) \underset{\tau \rightarrow 0}{\sim} \begin{cases} \tau^{-\beta} & \text{if } \beta > 0, \\ -\ln \tau & \text{if } \beta = 0, \\ \frac{1 - \tau^{-\beta}}{\beta} \rightarrow C & \text{if } \beta < 0. \end{cases}$$

▼ **Multifractal Analysis** $\implies \lim_{\tau \rightarrow 0}$

then the case $\beta < 0$ presents no interest for multifractal analysis ($\sim \tau^q$).

▲ **LID Analysis** $\implies \cancel{\lim_{\tau \rightarrow 0}}$

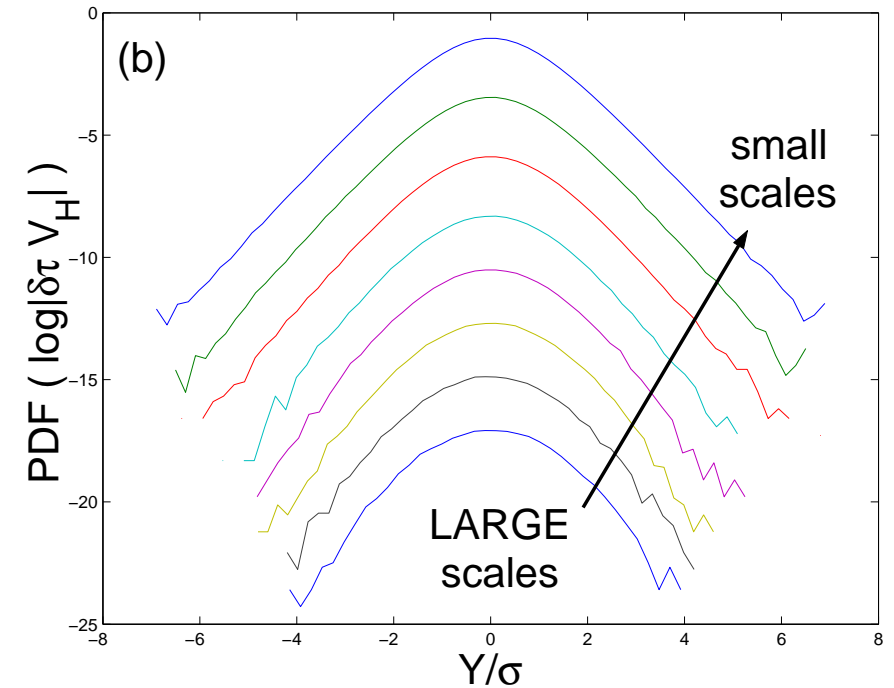
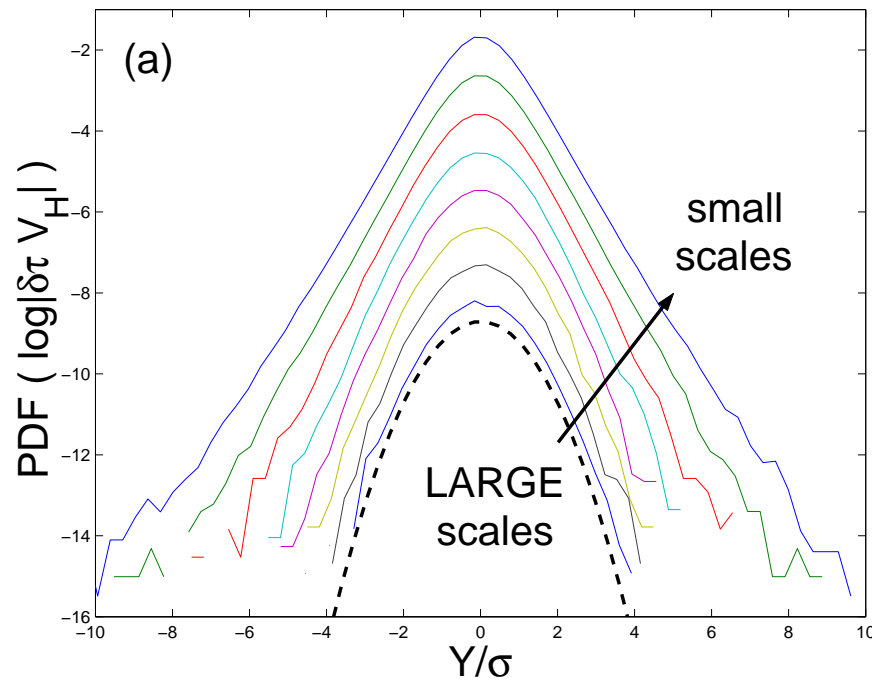
interest = **evolution through the scales**

Evolution of probability density functions

MULTIFRACTAL

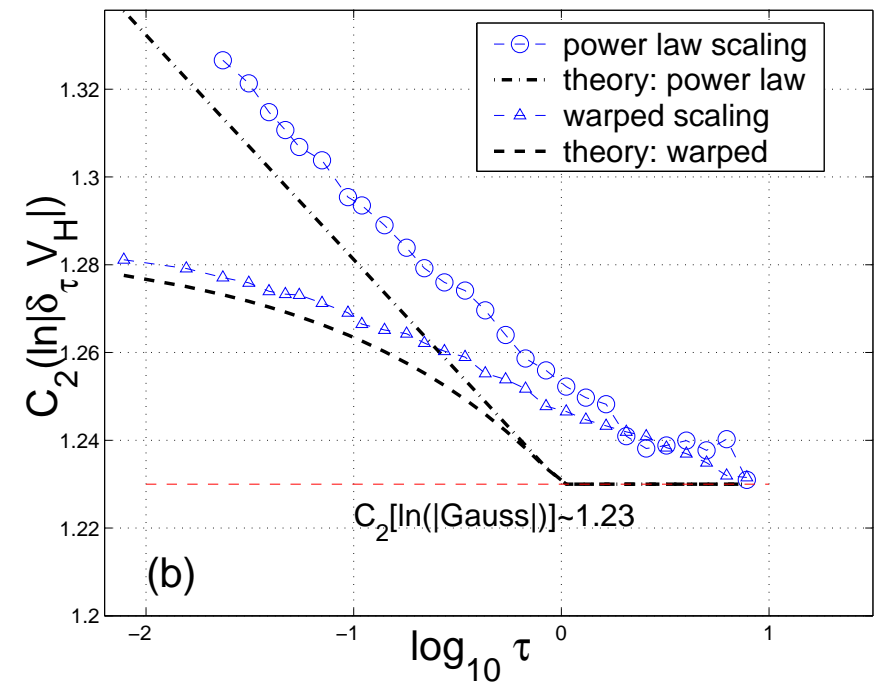
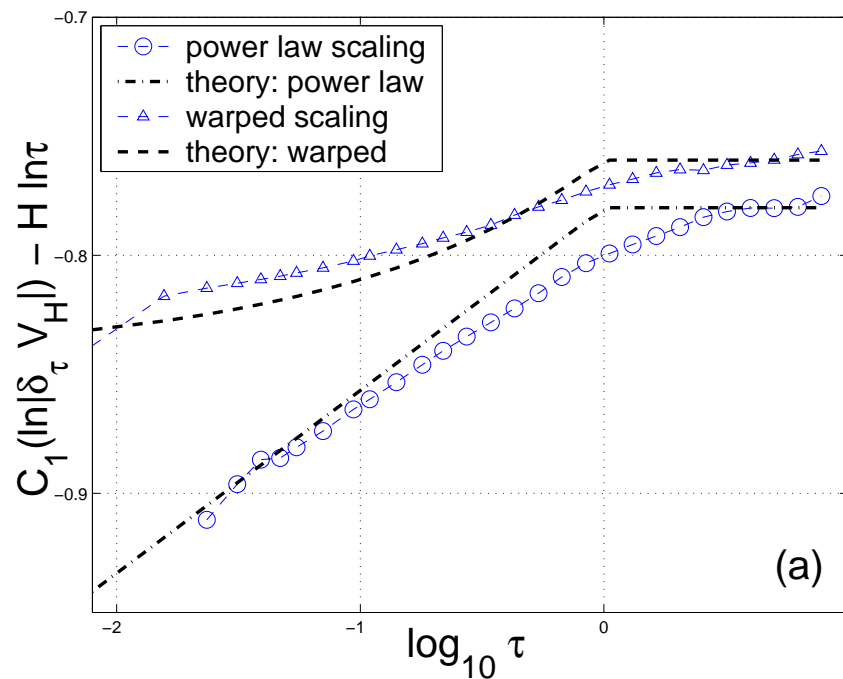
\neq

WARPED ($\beta = -0.4$)



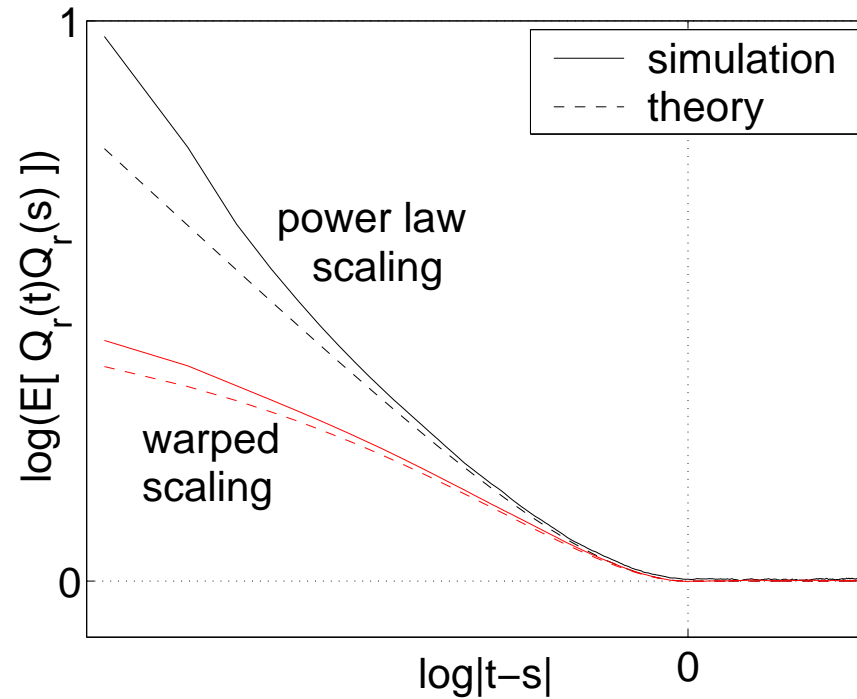
from gaussian to *non gaussian* towards small scales
(ex : kurtosis...)

Cumulants of log(increments of V_H)

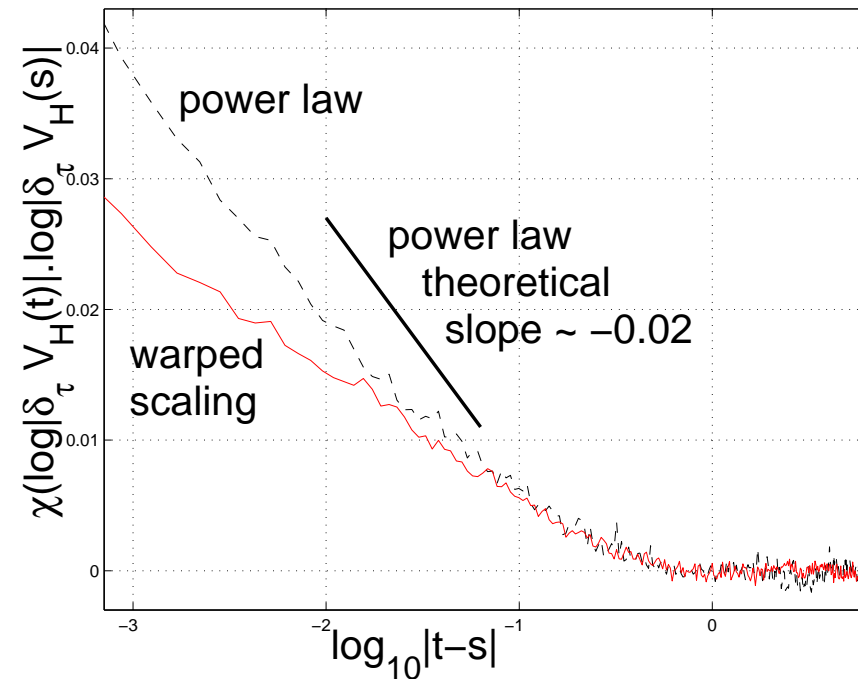


$$(a) C_1 - H \ln \tau \simeq -H \varphi'(0) m(\mathcal{C}_\tau) \quad (b) C_2 \simeq -H^2 \varphi''(0) m(\mathcal{C}_\tau)$$

"Warped" autocorrelation function



$$E[Q_r(t)Q_r(s)]$$



$$\text{Autocorrelation of } \ln |\delta_\tau V_H|$$

Intermediate conclusions

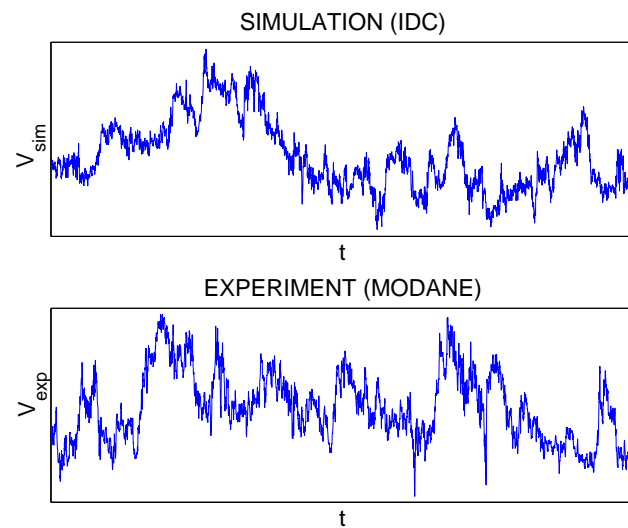
- ✓ we can *synthesize* Inf. Div. models,
- ✓ **warped cascades** : *a first controlled departure from power laws,*
- ✓ the effect is weak for V_H,
- ✓ **applications** : turbulence, internet traffic, finance...

Next ?

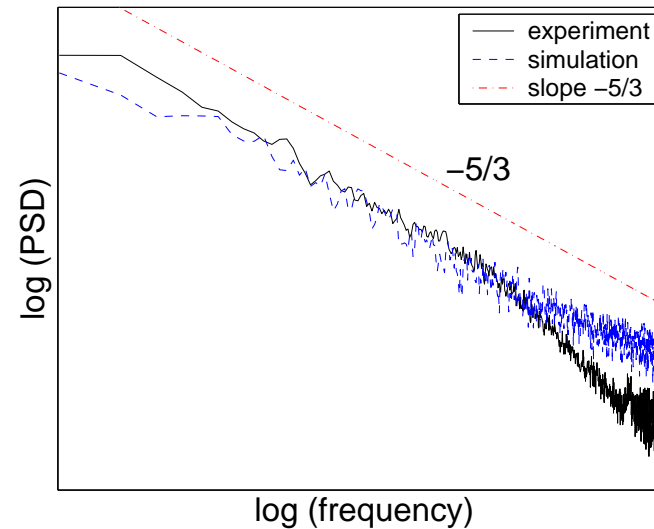
- ✓ farther from power laws : is it possible ?...
- ✓ other ways than IDC ? (pb of integration...)
- ✓ skewness ?
- ✓ analysis : estimation, prediction...
- ✓ multidimensional/multivariate cascades
- ✓ a toolbox with synthesis algorithms.

IDC vs She-Lévêque model and "Modane" turbulent signal

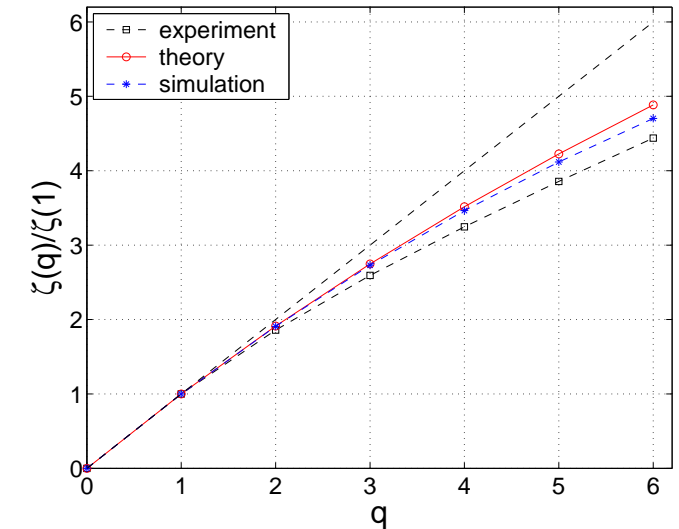
\equiv Inf. Div. random walk V_H versus velocity v



V_H versus Modane



Spectra



Exponents $\zeta(q)$

\implies artificial signal obeying the She-Lévêque model

De l'auto-similarité aux Cascades Infiniment Divisibles

$$[\delta_\tau X(t) = X(t + \tau) - X(t)]$$

➡ AUTO-SIMILARITÉ

$$\mathbf{E} |\delta_\tau X|^q = \mathbf{E} |X(1)|^q \cdot |\tau|^{qH}$$

ex : f.B.m., Linear Fractional Stable Motion...

➡ FORMALISME MULTIFRACTAL

$$\mathbf{E} |\delta_\tau X|^q = c_q \tau^{\zeta(q)}$$

ex : casc. binomiales, casc. aléa. ondelette, multifractal random walk (MRW)...

➡ CASCADES INF. DIVISIBLES

$$\mathbf{E} |\delta_\tau X|^q = c_q \exp[-H(q) \cdot n(\tau)]$$

- $H(q) \equiv \zeta(q)$ si $n(\tau) = -\log \tau$,
(multifractal)
- a priori $n(\tau) \neq -\log \tau$.

ex : ... ???

A stochastic equation for the Log-Normal case

Notation : $Q_r \equiv \varepsilon_\lambda$ with $\lambda = \frac{1}{r} \gg 1$

$$\varepsilon_\lambda(t) = \lambda^{-\mu/2} \exp \left(\mu^{1/2} \int_{t+1-\lambda}^t (t+1-u)^{-1/2} dB(u) \right)$$

⇓

$$\begin{cases} d\varepsilon_\lambda(t) = \sqrt{\mu} \varepsilon_\lambda(t) \left(dB(t) + \frac{1}{2}(1 - W(t))dt \right) \\ W(t) = \sqrt{\mu} \int_{t+1-\lambda}^t (t+1-u)^{-3/2} dB(u) \end{cases}$$

[F.G. Schmitt, Eur. Phys. J. B, 2003]