# **Detecting multifractality under heavy tails** Danijel Grahovac and Nikolai. N. Leonenko

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# Introduction

 $\{X(t)\}\$  is said to be multifractal if it has stationary increments and there exist functions c(q) and  $\tau(q)$  such that

 $E|X(t)|^q = c(q)t^{\tau(q)}, \quad ext{for all } t \in [0, T], q \in [q_-, q_+],$ 

for some T > 0 and  $q_{-}, q_{+} \in \mathbb{R}$ .  $\tau(q)$  is called the scaling function.  $\tau$  is always concave and if  $\{X(t)\}$  is H-s.s., then  $\tau(q) = Hq$ . Legendre transform of  $\tau$  is called the multifractal spectrum

 $d(h) = \inf_{q} \left( hq - \tau(q) + 1 \right),$ 

#### Statistical methods for detecting multifractal behavior

Partition function (or empirical structure function) - dividing the interval [0, T] into N blocks of length  $\Delta t$ 

$$S_q(T,\Delta t) = rac{1}{N}\sum_{i=1}^N |X(i\Delta t) - X((i-1)\Delta t)|^q$$

If  $\{X(t)\}$  is multifractal, then from the definition

n 
$$ES_q(T,\Delta t)= au(q)\ln\Delta t+\ln c(q).$$

# Asymptotic scaling function and spectrum



#### Example 1

(1)

Reproducing analysis from Fisher, Calvet, Mandelbrot [1997] and Calvet and Fisher

Suppose  $X_1, \ldots, X_T$  is a sample observed at discrete equally spaced time instants from a stochastic process  $\{X(t), t \ge 0\}$ . It is enough to consider only sampling at time instants  $1, 2, \ldots n$ .

Empirical scaling function - for each value q > 0 estimate  $\tau(q)$  as the slope in the simple linear regression of  $\ln S_q(T, \Delta t)$  on  $\ln \Delta t$ 

$$\hat{\tau}_{N,T}(q) = \frac{\sum_{i=1}^{N} \ln \Delta t_i \ln S_q(n, \Delta t_i) - \frac{1}{N} \sum_{i=1}^{N} \ln \Delta t_i \sum_{j=1}^{N} \ln S_q(n, \Delta t_i)}{\sum_{i=1}^{N} (\ln \Delta t_i)^2 - \frac{1}{N} \left(\sum_{i=1}^{N} \ln \Delta t_i\right)^2},$$

where  $1 \leq \Delta t_i \leq T$  for i = 1, ..., N. If  $\hat{\tau}$  is nonlinear, then one can suspect the existence of multifractal scaling.

Estimated spectrum - estimated as Legendre transform of  $\hat{\tau}$ .

### Assumptions

We say process  $\{X(t)\}$  is of type  $\mathfrak{L}$  if it satisfies  $P Y_t = X(t) - X(t-1), t \in \mathbb{N}$  is a strictly stationary sequence having a heavy-tailed marginal distribution with index  $\alpha$ , i.e.

$$P(|Y_1| > x) = rac{L(x)}{x^{lpha}}$$

where L(t), t > 0 is a slowly varying function  $L(tx)/L(x) \to 1$  as  $x \to \infty$ , for every t > 0.

 $\triangleright$  (Y<sub>t</sub>) satisfies strong mixing property with an exponentially decaying rate

$$a( au) = \sup_{t \geq 0} \sup_{A \in \mathcal{F}_t, B \in \mathcal{F}^{t+ au}} |P(A \cap B) - P(A)P(B)| = O(e^{-b au}), ext{ as } au o \infty$$

where  $\mathcal{F}_t = \sigma\{Y_s, s \leq t\}, \ \mathcal{F}^{t+\tau} = \sigma\{Y_s, s \geq t+\tau\}$  $\triangleright EY_t = 0$  when  $\alpha > 1$ 

Examples: all Lévy processes with X(1) heavy-tailed (e.g.  $\alpha$ -stable Lévy processes, Student Lévy process), cumulative sum of stationary processes like Ornstein-Uhlenbeck (OU) type processes or diffusions with heavy-tailed marginal distributions We are interested in the rate of growth of the partition function  $\frac{\ln S_q(T,T^s)}{\ln T}$ 

[2002] of DM/USD exchange rate data with Student Lévy process

Figure: Scaling function of the data - Figure 6. from Calvet and Fisher [2002]



Figure: Estimated spectrum of the data -Figure 7. from Calvet and Fisher [2002]



Figure: Estimated scaling function of generated Student Lévy process



Figure: Estimated spectrum of generated Student Lévy process



# Example 2

Comparison: 5307 daily closing values of S&P500 stock market index collected in the period from January 1, 1980 until December 31, 2000 vs. same length sample path of

# Theorem (Asymptotic behaviour of the partition function)

If  $\{X(t)\}$  is of type  $\mathfrak{L}$ , then for q > 0 and every  $s \in (0, 1)$ 

 $\frac{\ln S_q(T, T^s)}{\ln T} \xrightarrow{P} R_\alpha(q, s) := \begin{cases} \frac{sq}{\alpha}, & \text{if } q \leq \alpha \text{ and } \alpha \leq 2, \\ s + \frac{q}{\alpha} - 1, & \text{if } q > \alpha \text{ and } \alpha \leq 2, \\ \frac{sq}{2}, & \text{if } q \leq \alpha \text{ and } \alpha > 2, \\ \max\left\{s + \frac{q}{\alpha} - 1, \frac{sq}{2}\right\}, & \text{if } q > \alpha \text{ and } \alpha > 2, \end{cases}$ 

as  $T \to \infty$ , where  $\stackrel{P}{\to}$  stands for convergence in probability. Relation (1) holds approximately.

# Theorem (Asymptotic behaviour of the scaling function)

Suppose  $\Delta t_i$  is of the form  $T^{\frac{i}{N}}$  for i = 1, ..., N. Then, for every q > 0,

$$\lim_{N o\infty} \mathop{\mathrm{plim}}_{T o\infty} \hat{ au}_{N,T}(q) = au_\infty(q),$$

where plim stands for limit in probability and

$$au_{\infty}(q) = egin{cases} rac{q}{lpha}, & ext{if } 0 < q \leq lpha \& lpha \leq 2, \ 1, & ext{if } q > lpha \& lpha \leq 2, \ rac{q}{2}, & ext{if } 0 < q \leq lpha \& lpha \leq 2, \ rac{q}{2}, & ext{if } 0 < q \leq lpha \& lpha > 2, \ rac{q}{2} + rac{2(lpha - q)^2(2lpha + 4q - 3lpha q)}{2(2lpha + 2q)^2}, & ext{if } q > lpha \& lpha > 2. \end{cases}$$

Student Lévy process with  $X(1) \stackrel{d}{=} T(2.5, 0.0072, 0)$ 

Figure: Estimated scaling function of S&P 500 index with  $\tau_{\infty}$  for  $\alpha = 2.5$ 



Figure: Estimated spectrum of S&P 500 index



Figure: Estimated scaling function of generated Student Lévy process



Figure: Estimated spectrum of generated Student Lévy process

![](_page_0_Figure_58.jpeg)

# Acknowledgments and references

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![](_page_0_Picture_64.jpeg)

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![](_page_0_Picture_66.jpeg)