

Convex Rearrangements

For a process $X(t)$, $t \geq 0$ with stationary increments, observed at times $0, 1/N, \dots, (N-1)/N$ define increments

$$Y_{k,N} = X((k+1)/N) - X(k/N), \quad k = 0, 1, \dots, N-1. \quad (1)$$

Let, further $Y_{0:N} \leq Y_{1:N} \leq \dots \leq Y_{(N-1):N}$ denote the order statistic of Y 's from (1).

A *convex rearrangement* of $X(t)$ is defined as

$$\mathcal{V}X_N(t) = X(0) + \sum_{i=0}^{\lfloor Nt \rfloor - 1} Y_{i:N} + (Nt - \lfloor Nt \rfloor) Y_{\lfloor Nt \rfloor : N},$$

The sequence $Y_{k,N}$, $k = 0, 1, \dots, N-1$ is stationary with variance $\sigma_N^2 = \mathbb{E}Y_{1,N}^2$ and covariance $\gamma_N(h) = \mathbb{E}Y_{k,N} Y_{k+h,N}$.

Davydov (1998), Davydov and Thilly (2000), and Phillippe and Thilly (2002) investigated the asymptotic behavior of normalized $\mathcal{V}X_N(t)$ for broad classes of Gaussian and α -stable processes.

Theorem 0.1 (DAVYDOV AND THILLY, 2000)

Let for each $h \in \{1, 2, \dots, N-1\}$

$$[\text{CONDITION } \mathcal{A}] \quad \left| \frac{\gamma_N(h)}{\sigma_N^2} \right| \leq \min\{1, K \cdot (\ln h)^{-1-\tau}\},$$

for some fixed positive constants K and τ . Denote by

$$L(t) = \int_0^t \Phi^{-1}(s) ds = -\frac{1}{\sqrt{2\pi}} \exp\left\{-\frac{1}{2} (\Phi^{-1}(t))^2\right\},$$

where Φ is the c.d.f. of a standard normal distribution. Then, when $N \rightarrow \infty$,

$$\frac{\mathcal{V}X_N(t)}{b_N} \rightarrow L(t), \text{ a.s.}$$

where $b_N = N\sigma_N$.

Example 0.1 Let $B_H(t)$ be the fractional Brownian motion with the exponent H and $\mathcal{V}B_{H,N}(t)$ its convex rearrangement. Since $\sigma_N^2 = \mathbb{E}Y(1, N)^2 = \mathbb{E}B_H(1/N)^2 = C \cdot (1/N)^{2H}$. Then, the condition \mathcal{A} in Theorem 0.1 is satisfied and when $N \rightarrow \infty$ almost surely on $[0,1]$,

$$\frac{1}{N^{1-H} \sqrt{C}} \mathcal{V}B_{H,N}(t) \rightarrow L(t). \quad (2)$$

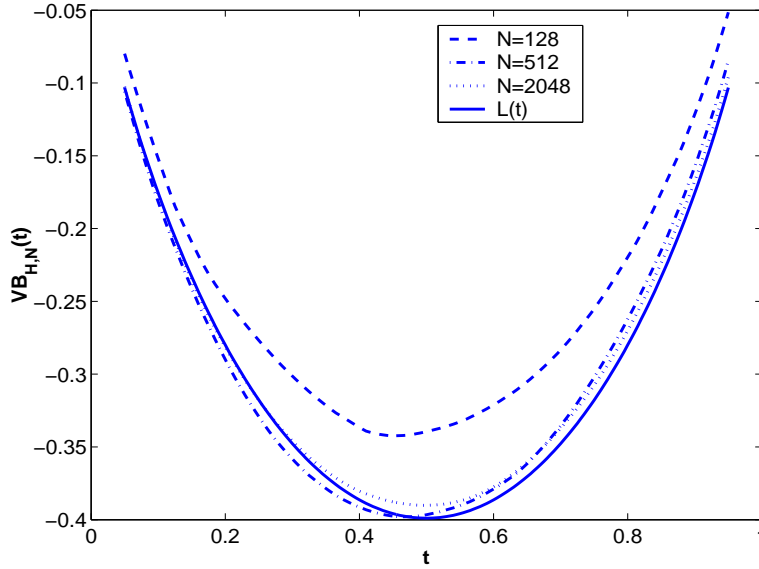


Figure 1: Approximation of $\int_0^t \Phi^{-1}(s)ds$ (solid line) by normalized $\mathcal{V}B_{H,N}$ for $H = 0.4$ and $N = 128$ (–), 512 (–.), and 2048. (..)

By solving (2) with respect to H , one can propose a family of consistent estimators of H , indexed by a continuous parameter $t_0 \in$

$(0, 1)$. That opens an interesting question of design which t_0 to select.

Let the self-similar process $X(t)$, $t \geq 0$ satisfies the hypothesis \mathcal{A} in Theorem 0.1. Let $t_0 \in (0, 1)$ and let

$$\hat{E}(N, t_0) = 1 - \frac{\log |\mathcal{V}X_N(t_0)/\sqrt{C}L(t_0)|}{\log n}.$$

Then $\hat{E}(N, t_0) \longrightarrow H$, *a.s.*, when $N \rightarrow \infty$.

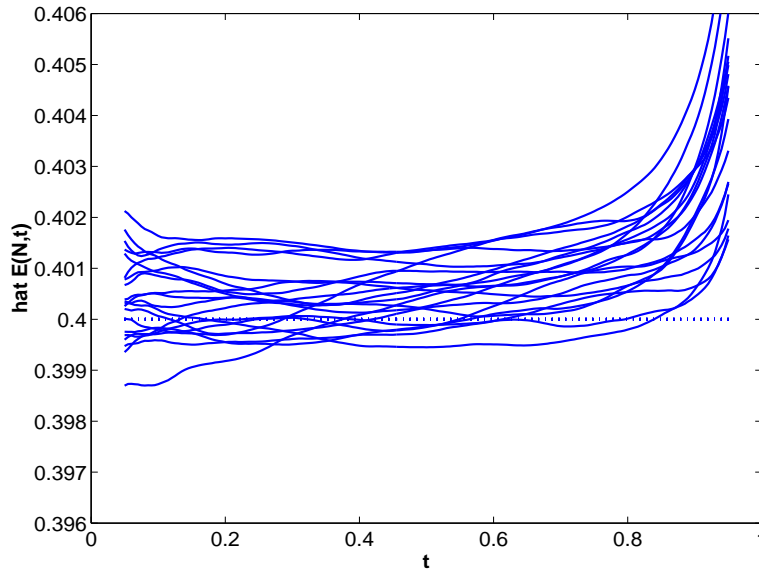


Figure 2: Estimators $\hat{E}(N, t)$ for $H = 0.4$, and $N = 2096$

Let $X(t)$, $t \geq 0$ be a selfsimilar process satisfying \mathcal{A} and let \mathbf{g} be the filter. If the convex rearrangements are applied on the \mathbf{g} -filtered process $Y\left(\frac{k}{N}\right) = \sum_{n=0}^{\ell} g_n X\left(\frac{k-n}{N}\right)$, $k = 0, 1, \dots, N-1$ the resulting polygonal process $Y_N(t) = Y\left(\frac{k}{N}\right) + (Nt-k) \left[Y\left(\frac{k-1}{N}\right) - Y\left(\frac{k}{N}\right) \right]$, satisfies the condition \mathcal{A} as well, and

$$\frac{1}{b_N} \mathcal{V}Y_N(t) \longrightarrow L(t), \text{ a.s., when } N \rightarrow \infty, \quad (3)$$

where $b_N = N\sigma_N(\mathbf{g})$. The filter \mathbf{g} of length ℓ has to be selected in such a way so that the variance $\sigma_N(\mathbf{g}) = \mathbb{E} \left| Y\left(\frac{k-1}{N}\right) - Y\left(\frac{k}{N}\right) \right|^2$ is not 0. Sufficient condition is that $f(s) = \sum_{i,j=0}^{\ell} g_i g_j |i-j|^s$ does not vanish for $s \in (0, 2)$ for the selected filter \mathbf{g} .

The double filtering of $X(t)$ in (3) can be replaced by a single filter \mathbf{g} with properties $\sum_n n^r g_n = 0$, $0 \leq r \leq p-1$ and $\sum_n n^p g_n \neq 0$. Indeed, only the existence of first vanishing moment allows the decomposition of \mathbf{g} to convolution of $\{1, -1\}$ and a finite filter \mathbf{u} of length $\ell - 1$.

Thus denote \mathbf{g} -convex rearrangement of $X(t)$ as

$$\mathcal{V}X_{\mathbf{g},N}(t) = X(0) + \sum_{i=0}^{\lfloor Nt \rfloor - 1} Y_{\mathbf{g},(i:N)} + (Nt - \lfloor Nt \rfloor) Y_{\mathbf{g},(\lfloor Nt \rfloor:N)},$$

where $\{Y_{\mathbf{g},(0:N)}, \dots, Y_{\mathbf{g},(N-1:N)}\}$ is order statistics for the sequence $Y_{\mathbf{g}}\left(\frac{k}{N}\right) = \sum_{n=0}^{\ell} g_n X\left(\frac{k-n}{N}\right)$, for $k = 0, 1, \dots, N-1$.

Let \mathbf{g}^d be dilation of filter \mathbf{g} obtained by inserting $d-1$ zeros between non-zero filter taps. For example, for $\mathbf{g} = \{1 \ -2 \ 1\}$ the 3-dilated filter is $\mathbf{g} = \{1 \ 0 \ 0 \ -2 \ 0 \ 0 \ 1\}$.

Theorem 0.2

Let

$$\mathbb{D}(N, d_1, d_2, t_0) = \frac{\mathcal{V}X_{\mathbf{g}^{d_2},N}(t_0)}{\mathcal{V}X_{\mathbf{g}^{d_1},N}(t_0)}. \quad (4)$$

Then for any $t_0 \in [0, 1]$, and integers d_1 , and d_2 ,

$$\frac{\log |\mathbb{D}(N, d_1, d_2, t_0)|}{\log(d_2/d_1)} \longrightarrow H, \text{ a.s.}$$

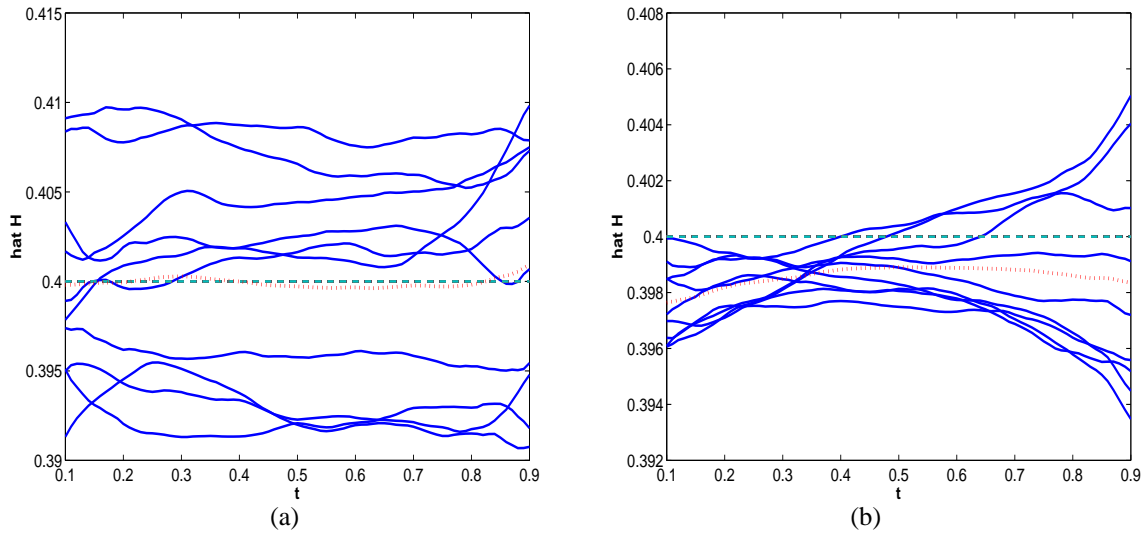


Figure 3: Estimation of H by (4). See the figconcre3.m

APPENDIX: MATLAB PROGRAMS

```

function filtd = dilate_filter(filt,k)
%-----
% Dilate Filter by k zeros between the original
% taps, k integer > 0.
%-----
newlength = (k+1)*length(filt)-k; filtd = zeros(1,newlength);
filtd(1:(k+1):newlength)=filt;
%-----
% Brani 10/15/02

```

```

function v = conre(t, Proc, fi)
%-----
% Convex Rearrangement
% Proc can be, for example MakeFBMNew(1024, 1/3, 1);
n = length(Proc);
difs =filter(fi,1,Proc);
difs = difs(length(fi):n);
nn = length(difs);
diffs(1:nn)=sort(difs);
upper = floor(nn.*t); v=[];
for i = 1:length(t)
    v = [v sum( difs(1:(upper(i)-1)))+ ...
        ((nn.*t(i) - upper(i)).*difs(floor(nn.*t(i))))];
end

```

```

%-----
close all
clear all
family = 'Symmlet'; par = 4;
k1 = 2; k2 = 3; H= 0.4; seed = 1;
n = 128 * 1024;
b = (0.1:0.01:0.9);
baseline = [];
for seed = 1:5
%-----
a = MakeFBMNew(n, H, seed);
filt = MirrorFilt(MakeONFilter(family, par));
filtd1 = dilate_filter(filt,k1);
filtd2 = dilate_filter(filt,k2);
y1 = conre(b,a, filtd1);
y2 = conre(b,a, filtd2);
z = y2./y1;
zz = log2( abs(z) )./log2((k2+1)/(k1+1));
plot(zz)
    hold on
plot(H*ones(length(zz)))
    hold on
baseline = [baseline; zz];
end
plot(mean(baseline), 'r-')

```

REFERENCES

- Davydov Yu., Thilly E., Convex rearrangements of Gaussian processes. Theory of Probability and its Applications.
- Philippe, A. and Thilly, E., Identification of locally self-similar Gaussian process by using convex rearrangements, Methodology and Computing in Applied Probability Vol 4, N2, p 195-207, (2002)