# Relationship between the fractal dimension of a fractional Brownian motion and its projection 

H. Biermé ${ }^{1}$, G. Lemineur ${ }^{2}$, A. Estrade ${ }^{1}$, R. Harba ${ }^{2}$<br>${ }^{1}$ MAPMO, Université d’Orléans BP 6759, 45067 Orléans Cedex 2<br>${ }^{2}$ LESI, Polytech’Orléans, BP 6744, 45067 Orléans Cedex 2

## Aim

- To present the relationship between the fractal dimension of nD fractional Brownian Motion ( fBm ) and that of its ( $\mathrm{n}-1$ ) D projection.
- To test this relationship on 2D synthesized fBm images.
- To apply it on 3D bone images.


## Relationship between box-counting dimension of fBm and its projections

A $n D f B m$ of $H$ index, $B_{H}$, is characterized by a self-similarity property, for all scale $\lambda: \quad\left\{B_{H}(\lambda t) ; t \in R^{n}\right\}=\left\{\lambda^{(d)} B_{H}(t) ; t \in R^{n}\right\}$.
Since $B^{H}$ is a zero mean Gaussian field with stationary increments, its distribution is completely determined by its variogram $v$ :

$$
v(t)=\frac{1}{4} E\left(\left(B_{H}(t)-B_{H}(0)\right)^{2}\right)=C_{H, n}\|t\|^{2 H}
$$

We are looking for the box-counting dimension of the $f B m$ graph, $\operatorname{dim}_{B} B_{H}$. Let $G=\left\{\left(t, B_{H}(t)\right) ; t \in[0,1]^{n}\right\} \subset R^{n+1}$ be the graph of the $f B m$.
We call $\mathrm{N}_{\delta}(\mathrm{G})$ the smallest number of sets of diameter at most $\delta$ which can cover G . As we have, for small $\delta$,

$$
\mathrm{N}_{\delta}\left(\mathrm{B}_{\mathrm{H}}\right) \approx \mathrm{C} \delta^{-\mathrm{D}},
$$

the box-counting dimension of $G$ is defined by $\operatorname{dim}_{B} B_{H}=D$.
The behaviour of the fBm variogram at small scales gives its box-counting dimension :

$$
\mathrm{v}(\mathrm{t}) \leq \mathrm{C}_{\mathrm{H}, \mathrm{n}}\|\mathrm{\| t}\|^{2 \mathrm{H}} \Rightarrow \operatorname{dim}_{\mathrm{B}} \mathrm{~B}_{\mathrm{H}} \leq \mathrm{n}+1-\mathrm{H} \text { a.s. and } \mathrm{v}(\mathrm{t}) \geq \mathrm{C}_{\mathrm{H}, \mathrm{n}}\|t\|^{2 \mathrm{H}} \Rightarrow \operatorname{dim}_{\mathrm{B}} \mathrm{~B}_{\mathrm{H}} \geq \mathrm{n}+1-\mathrm{H} \text { a.s., }
$$

then

$$
\operatorname{dim}_{B} B_{H}=n+1-H \text { a.s. }
$$

Let now have a look to the windowed X-ray transform of $B_{H}$. Fix $\alpha \in S^{n-1}$ a direction in $R^{n}$, we can define a Gaussian field with stationary increments
$\mathrm{P}_{\alpha} \mathrm{B}_{\mathrm{H}}=\left\{\mathrm{P}_{\alpha} \mathrm{B}_{\mathrm{H}}(\mathrm{s}) ; \mathrm{s} \in\langle\alpha\rangle^{\perp}\right\}$. For $\mathrm{s} \in<\alpha>^{\perp}$, we average $\mathrm{B}_{H}$ with a suitable window $\varphi$ along the line $\{\mathrm{s}+\mathrm{p} \alpha ; p \in \mathrm{R}\}$ and denote the result $\mathrm{P}_{\alpha} \mathrm{B}_{H}(\mathrm{~s})$. There exists two positive constants $C_{1}(\alpha)$ and $C_{2}(\alpha)$ such that the variogram $v_{P}$ of $P_{\alpha} B_{H}$ satisfies, for small $s \in\langle\alpha\rangle^{\perp}$,

$$
\mathrm{C}_{1}(\varphi)\|S\|^{2 \min (1, \mathrm{H}+1 / 2)} \leq \mathrm{v}_{\mathrm{P}}(\mathrm{~s}) \leq \mathrm{C}_{2}(\varphi)\| \|^{2 \min (1, \mathrm{H}+1 / 2)}
$$

Thus, $\operatorname{dim}_{\mathrm{B}} \mathrm{P}_{\alpha} \mathrm{B}_{\mathrm{H}}=\operatorname{dim}\langle\alpha\rangle^{\perp}+1-\min (1, \mathrm{H}+1 / 2)=\mathrm{n}-\min (1, \mathrm{H}+1 / 2)$.

## Application to 2D synthesized fBm

We synthesized 2D fBm images with a size of $725 \times 725$ pixels and performed their projection following the vertical axis. Points are regularly spaced with a step of $2^{-10}$ We then estimated their box-counting dimensions.


2D fBm ( $\mathrm{H}=0.2$ ).


1D signal of its projection.

|  | $\mathrm{H}=0,2$ | $\mathrm{H}=0,2$ | $\mathrm{H}=0,3$ | $\mathrm{H}=0,3$ | $\mathrm{H}=0,4$ | $\mathrm{H}=0,4$ |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: |
| $\operatorname{Dim}_{\mathrm{B}} \mathrm{B}_{\mathrm{H}}=2+1-\mathrm{H}$ | 2,7605 | 2,768 | 2,6967 | 2,6964 | 2,622 | 2,603 |
| $\operatorname{Dim}_{\mathrm{B}} \mathrm{BB}_{\mathrm{H}}=1+1-(\mathrm{H}+1 / 2)$ | 1,3556 | 1,2728 | 1,3427 | 1,2906 | 1,1736 | 1,2155 |
| Error $=3 / 2-\operatorname{Dim}_{\mathrm{B}} \mathrm{B}_{\mathrm{H}}+\operatorname{Dim}_{\mathrm{B}} \mathrm{PB}_{\mathrm{H}}$ | 0,0951 | 0,0048 | 0,146 | 0,0942 | 0,0516 | 0,1125 |

We estimated the dimensions by regression of $\log \left(\mathrm{N}_{\delta}\right)$ for $\delta=2^{-4}, 2^{-5}, 2^{-6}, 2^{-7}$. The fractal dimensions evaluated suffer from excessive bias. It is probably a consequence of the fact that $\mathrm{N}_{\delta} \approx \mathrm{C}^{-\mathrm{D}}$ for a constant C , rather than $\mathrm{N}_{\delta} \approx \delta^{-\mathrm{D}}$.

## Application to 3D bone images

Human bone is composed of trabecular bone whose modifications could lead to osteoporotic fractures. It is a porous material ordered in trabeculae of $100 \mu \mathrm{~m}$ on average thickness and separated of 300 to $500 \mu \mathrm{~m}$ on average. Direct 3D evaluation of trabecular bone structure could not be done in clinical practice due to the complexity, price, low resolution and accessibility of the devices. So the main alternative to the 3D method is the conventional bone radiography. It would be very interesting to link 2D parameters measured on radiographs with 3D ones.


The bone volume is given by a binary random field $B:[0,1]^{3} \rightarrow\{0,1\}$. We can define the random set $K=\left\{t \in[0,1]^{3} ; B(t)=1\right\} \subset R^{3}$. We assume that $K$ is a self-similar set of index $\mathrm{H}<1 / 2$ such that $\operatorname{dim}_{\mathrm{B}} \mathrm{K}=3-\mathrm{H}$. We obtain a radiograph of the bone by performing a windowed X-ray transform in the direction $\alpha \in \mathrm{S}^{2}$ of B denoted $\mathrm{P}_{\alpha} \mathrm{B}=\left\{\mathrm{P}_{\alpha} \mathrm{B}(\mathrm{s}) ; \mathrm{s} \in\langle\alpha\rangle^{\perp}\right\}$. We conjecture the graph of $\mathrm{P}_{\alpha} \mathrm{X}$ has a box-counting dimension $\operatorname{dim}_{\mathrm{B}} \mathrm{P}_{\alpha} \mathrm{B}=2+1-(\mathrm{H}+1 / 2)$.


We tested the relation on 8 bone samples. The images were acquired with a high resolution micro tomograph and have a size of 256 pixels $^{3}$ with a pixel size of $12 \mu \mathrm{~m}$. We computed the projections by just summing the image following the height axis.

| $\operatorname{Dim}_{\mathrm{B}} \mathrm{B}=3-\mathrm{H}$ | 2,6772 | 2,6538 | 2,6281 | 2,8047 | 2,7057 | 2,6961 | 2,5694 | 2,8398 |
| :--- | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $\operatorname{Dim}_{\mathrm{B}} \mathrm{PB}=2+1-(\mathrm{H}+1 / 2)$ | 2,366 | 2,2994 | 2,2792 | 2,2988 | 2,2676 | 2,2884 | 2,2999 | 2,341 |
| Error=1/2-Dim ${ }_{B} \mathrm{~B}+\operatorname{Dim}_{\mathrm{B}} \mathrm{PB}$ | 0,1888 | 0,1456 | 0,1511 | $-0,0059$ | 0,0619 | 0,0923 | 0,2305 | 0,0012 |

