

Sample Path Properties of Anisotropic Gaussian Random Fields

Let $X = \{X(t), t \in \mathbb{R}^n\}$ be a Gaussian random field with values in \mathbb{R}^d . For simplicity, we assume the coordinate processes X_1, \dots, X_d are i.i.d. Then many sample path properties of X can be determined by the following function

$$\sigma^2(s, t) = \mathbb{E}[(X_1(t) - X_1(s))^2].$$

If we have knowledge ~~about~~^{on} the dependence structure of X , then more precise information about the sample functions of X can be derived.

In these lectures, we will consider three conditions on X :

(C1)

$$K_1 \sum_{i=1}^N |s_i - t_i|^{2H_i} \leq \sigma^2(s, t) \leq K_2 \sum_{i=1}^N |s_i - t_i|^{2H_i}$$

(C2) $\text{Var}(X(t) | X(s)) \geq K_3 \sum_{i=1}^N |t_i - s_i|^{2H_i}$.

(C3) $\forall \varepsilon > 0$
 $\exists K_4 > 0$ s.t. $\forall n \geq 1, u \in (\varepsilon, +\infty)^N, t^1, \dots, t^n \in (\varepsilon, +\infty)^N$,
 $\text{Var}(X(t) | X(t^1), \dots, X(t^n)) \geq K_4 \sum_{i=1}^N \min_{j=0,1,\dots,n} \{ |t_i^j - t_i^j|^{2H_i} \}$

where $t^0 = 0$.

We show: (C1) \Rightarrow sample path continuity and Hausdorff dimension

(C1) + (C2) \Rightarrow level sets and existence of local times

(C1) + (C3) \Rightarrow Joint continuity of the local times.

Examples of Anisotropic Gaussian Random fields

↳ Fractional Brownian sheet: $B^H = \{B^H(t), t \in \mathbb{R}_+^N\}$ is \mathbb{R}^d -valued, with mean zero and covariance function given by

$$\mathbb{E}(B_j^H(t) B_k^H(s)) = \delta_{j,k} \prod_{\ell=1}^N \frac{1}{2} (|t_\ell|^{2H_\ell} + |s_\ell|^{2H_\ell} - |t_\ell - s_\ell|^{2H_\ell}) \quad \forall j, k=1, \dots, d.$$

So the coordinate processes are i.i.d.

$\{B_j^H(t), t \in \mathbb{R}_+^N\}$ has the following two stochastic integral representations

$$(i) \quad B_j^H(t) = c_H \int_{\mathbb{R}^N} \prod_{\ell=1}^N \left[(t_\ell - r_\ell)_+^{H_\ell - \frac{1}{2}} - (-r_\ell)_+^{H_\ell - \frac{1}{2}} \right] W(dr)$$

where $a_+ = \max\{a, 0\}$, and W is the standard Brownian sheet.

$$(ii) \quad B_j^H(t) = c'_H \int_{\mathbb{R}^N} \prod_{\ell=1}^N \frac{e^{it_\ell r_\ell} - 1}{i\lambda_\ell^{H_\ell + \frac{1}{2}}} \tilde{W}(d\lambda)$$

where \tilde{W} is a complex Gaussian measure with λ_N as its control measure. Leb.

Exercise ① verify the following scaling property: $\forall c > 0$

$$\{B^H(ct), t \in \mathbb{R}^N\} \stackrel{d}{=} \left\{ c^{\frac{N}{2H_1}} B^H(t), t \in \mathbb{R}^N \right\} \Rightarrow \text{OSS}$$

② verify (i) and (ii) above

- All three conditions (C1), (C2) and (C3) hold for B^H ; see Ayache and Xiao (2005), Mu and Xiao (2006), Xiao and Zhang (2002)
- 2. } Gaussian random fields with stationary increments
- } Additive fractional Brownian motion

Let $X_1, \dots, X_N = \{X_N(t), t \in \mathbb{R}^1\}$ be independent fractional Brownian motions in \mathbb{R}^d with Hurst indices H_1, \dots, H_N , respectively. Then the (N, d) -random field $\{X(t), t \in \mathbb{R}^N\}$ defined by

$$X(t) = X_1(t_1) + \dots + X_N(t_N)$$

is an operator self-similar Gaussian random field with stationary increments.
 Kono (1975) studied some asymptotic properties of such X .

~~3. Gaussian random fields with stationary increments~~

3. Random string processes of Muller and Tribe (2002).

Let $\{u_t(x), t \geq 0, x \in \mathbb{R}\}$ be the solution of the SPDE:

$$\frac{\partial u_t(x)}{\partial t} = \frac{\partial^2 u}{\partial x^2} + \dot{w},$$

where $\dot{w}(x, t)$ is a space-time white noise, with appropriate initial conditions.

Let $X = \{X(t, x), t \geq 0, x \in \mathbb{R}\}$ be defined by

$$X(t, x) = (X_1(t, x), \dots, X_d(t, x)),$$

where X_1, \dots, X_d are independent copies of $\{u_t(x), t \geq 0, x \in \mathbb{R}\}$.

$$\mathbb{E}[(u_t(x) - u_s(y))^2] \asymp |x - y| + |t - s|^{1/2}$$

It can be shown that conditions (C1) and (C2) hold for X .

§ 1. Modulus of continuity of Anisotropic Gaussian Random Fields.

- For fractional Brownian sheet, see Ayache and Xiao (2005), using wavelet expansion.

- More generally, we may apply the following extension of Garsia's lemma

Arnold and Imkeller (1996) due to Funaki, Kikuchi and Potthoff (2006), Khoshnevisan (Pri-C)

To state their theorem, we need some notation: Let $I \subseteq \mathbb{R}^N$ be a fixed interval. For any $s, t \in I$, let

$$q(s, t) = \sum_{i=1}^N |s_i - t_i|^{2H_i},$$

(Where $0 < H_1 \leq \dots \leq H_N \leq 1$ are constants) For any $s \in I$, denote

$$V_s(r) = \{ t \in I : q(s, t) \leq r \}$$

Theorem 1.1. Suppose that X is a continuous mapping from I to \mathbb{R} . If there exist two strictly increasing functions ψ and p on \mathbb{R}_+ with $\psi(0) = p(0) = 0$ and $\lim_{u \rightarrow \infty} \psi(u) = \infty$ such that

$$B := \int_I \int_I \psi \left(\frac{|X(s) - X(t)|}{p(q(s, t))} \right) ds dt < \infty$$

then for all $s, t \in I$, we have

$$|X(s) - X(t)| \leq K_1 \max_{z \in \{s, t\}} \int_0^{q(s, t)} \psi^{-1} \left(\frac{4B}{\lambda_N (V_z(u))^2} \right) p(u) du,$$

where $K_1 > 0$ is a constant depending on N and (H_1, \dots, H_N) only.

Corollary 1.2. Let $X = \{X(t), t \in \mathbb{R}^N\}$ be a centered Gaussian r.f. satisfying

$$K_2 \sum_{i=1}^N |t_i - s_i|^{2H_i} \mathbb{E} (X(t) - X(s))^2 \leq K_3 \sum_{i=1}^N |t_i - s_i|^{2H_i} \text{ for all } s, t \in I$$

then there exist a random variable A and a constant K_4 such that

$$\sup_{s, t \in I} |X(s) - X(t)| \leq A \cdot \sum_{i=1}^N |t_i|$$

Corollary 1.2 Let $X = \{X(t), t \in I\}$ be a centered Gaussian random field.

If there exist a constant $K_2 > 0$ and $(H_1, \dots, H_N) \in (0, 1]^N$ such that

$$\mathbb{E}[(X(s) - X(t))^2] \leq K_2 \sum_{i=1}^N |s_i - t_i|^{2H_i}, \quad \forall s, t \in I.$$

Then there exists a random variable A with finite moments such that

$$|X(s) - X(t)| \leq A \left(\sum_{i=1}^N |s_i - t_i|^{H_i} \right) \sqrt{\log \left(1 + \left(\sum_{i=1}^N |s_i - t_i|^{H_i} \right)^2 \right)}, \quad \forall s, t \in I.$$

Proof. We choose $\psi(x) = e^{\frac{x^2}{4K_2} - 1}$, $p(x) = \sqrt{x}$, then

$$\begin{aligned} \mathbb{E} \int_I \int_I \psi \left(\frac{|X(s) - X(t)|}{\sqrt{q(s,t)}} \right) ds dt &\leq \int_I \int_I \left(\int_{\mathbb{R}} e^{-\frac{x^2}{4\sigma^2}} \frac{1}{\sqrt{2\sigma}} e^{-\frac{x^2}{2\sigma^2}} dx \right) ds dt \\ &:= K < \infty, \quad \text{where } \sigma^2 = \mathbb{E}(X(s) - X(t))^2. \end{aligned}$$

Hence, Theorem 1.1 implies that $\forall s, t \in I$,

$$\begin{aligned} |X(t) - X(s)| &\leq A \cdot \max_{z \in \{s, t\}} \int_0^{q(s,t)} \sqrt{\log(1+u)} \frac{du}{\sqrt{u}} \\ &\leq A \sqrt{q(s,t) \cdot \log(1+q(s,t))} \end{aligned}$$

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Note $\lambda_n(V_n(u)) = \lambda_n \{t \in I: q(t, z) \leq u\} \asymp u^{\frac{1}{H_1} \wedge \dots \wedge \frac{1}{H_N}}$.

§2. Hausdorff dimension of the range and graph

Let $X = \{X(t), t \in \mathbb{R}^N\}$ be a centered Gaussian random field with values in \mathbb{R}^d , $\forall t \in (0,1)^N$. We assume the coordinate processes X_1, \dots, X_d are i.i.d and there are constants $K_4, K_5 > 0$ and $(H_1, \dots, H_N) \in (0,1)^N$ such that

$$K_4 \sum_{i=1}^N |s_i - t_i|^{2H_i} \leq \mathbb{E} [X(t) - X(s)]^2 \leq K_5 \sum_{i=1}^N |s_i - t_i|^{2H_i}$$

for all $s, t \in [0,1]^N$. For convenience, we assume $0 < H_1 \leq H_2 \leq \dots \leq H_N < 1$.

We now determine the Hausdorff dimensions of the random sets:

$$X([0,1]^N) = \{X(t) : t \in [0,1]^N\} \subseteq \mathbb{R}^d,$$

$$\text{Gr}X([0,1]^N) = \{(t, X(t)) : t \in [0,1]^N\} \subseteq \mathbb{R}^{N+d}.$$

Theorem 2.1 Under the above assumptions, we have

$$\dim_H X([0,1]^N) = \min \left\{ d, \sum_{i=1}^N \frac{1}{H_i} \right\} \quad \text{a.s.} \quad (2.1)$$

$$\begin{aligned} \dim_H \text{Gr}X([0,1]^N) &= \min \left\{ \sum_{j=1}^k \frac{H_k}{H_j} + N - k + (1 - H_k)d, 1 \leq k \leq N, \sum_{i=1}^N \frac{1}{H_i} \right\} \\ &= \begin{cases} \sum_{i=1}^N \frac{1}{H_i} & \text{if } \sum_{i=1}^N \frac{1}{H_i} \leq d \\ \sum_{j=1}^k \frac{H_k}{H_j} + N - k + (1 - H_k)d & \text{if } \sum_{i=1}^N \frac{1}{H_i} \leq d < \sum_{j=1}^k \frac{1}{H_j} \end{cases} \end{aligned}$$

We divide the proof into two parts: (i) proof of the upper bounds, (ii) proof of the

lower bounds. For any $F \subseteq \mathbb{R}^d$,

(i) In order to prove $\dim_H F \leq a$, it is sufficient to show that for any $\gamma > a$, $\forall \varepsilon > 0$, there exist a sequence of balls (cubes) $\{C_n\}$ with $\text{diam } C_n \leq \varepsilon$ such that

$$F \subseteq \bigcup_n C_n \quad \text{and} \quad \sum_{n=1}^{\infty} (\text{diam}(C_n))^\gamma \leq K, \quad (2.3)$$

where $K > 0$ does not depend on ε .

The upper bounds in Theorem 2.1 follow from Lemma 2.3 below and Corollary 1.2.

→ We assume $H_1 \leq H_2 \leq \dots \leq H_N$

Lemma 2.2. Let $f: [0,1]^N \rightarrow \mathbb{R}^d$ be a function satisfying the following conditions: there exist $(H_1, \dots, H_N) \in (0,1)$ and a constant $K > 0$ such that

$$|f(s) - f(t)| \leq K \sum_{i=1}^N |s_i - t_i|^{H_i} \quad \text{for all } s, t \in [0,1]^N. \quad (24)$$

Then

$$\dim_H f([0,1]^N) \leq \min \left\{ d, \sum_{i=1}^N \frac{1}{H_i} \right\} \quad (25)$$

$$\dim_H \text{Gr}f([0,1]^N) \leq \min \left\{ \sum_{i=1}^N \frac{1}{H_i}; \sum_{j=1}^k \frac{H_k}{H_j} + N - k + (1 - H_k)d, 1 \leq k \leq N \right\} \quad (26)$$

Proof. ~~I will only~~ ^{first} prove (25). Since $f([0,1]^N) \subseteq \mathbb{R}^d$, we have

$\dim_H f([0,1]^N) \leq d$. Hence it remains to prove

$$\dim_H f([0,1]^N) \leq \sum_{i=1}^N \frac{1}{H_i}. \quad (27)$$

$\left[\forall \gamma > \sum_{j=1}^N \frac{1}{H_j} \right]$ For any $n \geq 2$, we divide $[0,1]^N$ into m_n sub-rectangles $\{R_{ni}\}$ with sides parallel to the axes and side-lengths $n^{-\frac{1}{H_i}}$ ($i=1, 2, \dots, N$) respectively. Then

$$m_n \leq K n^{\sum_{i=1}^N \frac{1}{H_i}}. \quad (28)$$

and $f([0,1]^N)$ can be covered by $f(R_{ni})$ ($1 \leq i \leq m_n$). By (25), we have

$$\text{diam } f(R_{ni}) \leq K n^{-1}. \quad (29)$$

For any $\gamma > \sum_{j=1}^N \frac{1}{H_j}$, we have

$$\sum_{i=1}^{m_n} (\text{diam } f(R_{ni}))^\gamma \leq K n^{\sum_{i=1}^N \frac{1}{H_i}} \cdot n^{-\gamma} \leq K$$

this implies $\dim_H f([0,1]^N) \leq \gamma$. Since $\gamma > \sum_{j=1}^N \frac{1}{H_j}$ is arbitrary, (27) follows.

Now we turn to the proof of (26). We will show that there are several different ways to cover $\text{Gr}f([0,1]^N)$, each of which leads to an upper bound for $\dim_H \text{Gr}f([0,1]^N)$.

For each fixed integer $n \geq 2$, we have

$$\text{Gr}f([0,1]^N) \subseteq \bigcup_{i=1}^{m_n} R_{ni} \times f(R_{ni})$$

and $\text{diam}(R_{ni} \times f(R_{ni})) \leq K n^{-1}$

This implies $\dim_{\mathbb{H}} \text{Grf}([0,1]^N) \leq \sum_{k=1}^N \frac{1}{H_k}$. (2.10)

We fix an integer $1 \leq k \leq N$. observe that each $R_{n,k} \times f(R_{n,k})$ can be covered by $l_{n,k}$ cubes in \mathbb{R}^{N+d} of sides $n^{-\frac{1}{H_k}}$, where

$$l_{n,k} \leq K \cdot n^{\sum_{j=1}^N (\frac{1}{H_k} - \frac{1}{H_j})} \cdot n^{(H_k-1)d} \quad (2.11)$$

Hence $\text{Grf}([0,1]^N)$ can be covered by $m_k \times l_{n,k}$ cubes in \mathbb{R}^{N+d} with sides $n^{-\frac{1}{H_k}}$. ~~Since~~ Since $\forall \delta > \sum_{j=1}^k \frac{H_k}{H_j} + N - k + (1 - H_k)d$, (2.8) and (2.11) imply

$$\sum_{i=1}^{m_k} \sum_{j=1}^{l_{n,k}} (n^{-\frac{1}{H_k}})^{\delta} \leq K n^{\frac{1}{H_k} (\sum_{j=1}^k \frac{H_k}{H_j} + N - k + (1 - H_k)d - \delta)} \rightarrow 0$$

This yields $\dim_{\mathbb{H}} \text{Grf}([0,1]^N) \leq \sum_{j=1}^k \frac{H_k}{H_j} + N - k + (1 - H_k)d$. (2.12)

It is clear that (2.6) follows from (2.10) and (2.12). #

The conditions of Lemma 2.2 can be relaxed.

[Lemma 2.2'] Let $f: [0,1]^N \rightarrow \mathbb{R}^d$ be a function satisfying the following condition: there exists $(H_1, \dots, H_N) \in (0,1)^N$

Lemma 2.3' Let $f: [0,1]^N \rightarrow \mathbb{R}^d$ be a function and $(H_1, \dots, H_N) \in (0,1)^N$ satisfying $H_1 \leq \dots \leq H_N$. If for any $\epsilon > 0$, there exists a constant $K > 0$ such that

$$|f(s) - f(t)| \leq K \sum_{k=1}^N |t_k - s_k|^{H_k - \epsilon}, \quad \forall s, t \in [0,1]^N$$

Then (2.5) and (2.6) still hold.

(ii) Proof of the lower bounds.

Method \forall Borel set $F \subseteq \mathbb{R}^d$, in order to prove $\dim_{\mathbb{H}} F \geq \delta$, it is sufficient to find a ~~probability~~ ^{Borel} measure σ on F such that

$$\int_F \int_F \frac{\sigma(dx) \sigma(dy)}{|x-y|^{\delta}} < \infty$$

this follows from Frostman's theorem.

Exercises: Let C be the ternary Cantor set. show that $\dim_H C = \frac{\log 2}{\log 3}$

- prove (2.17) (a simple substitution)
- prove Lemma 2.5.

The lower bounds in Theorem 2.1 follow from the following proposition

Proposition 2.4. Let $X = \{X(t), t \in \mathbb{R}^N\}$ be a Gaussian random field in \mathbb{R}^d with i.i.d. components and $(H_1, \dots, H_N) \in (0,1)^N$ satisfy $H_1 \leq \dots \leq H_N$. Assume for some $\varepsilon \in (0,1)$, there exists a constant $K > 0$ such that

$$\mathbb{E}[(X_1(t) - X_1(s))^2] \geq K \sum_{i=1}^N |t_i - s_i|^{2H_i} \quad \text{for all } s, t \in [\varepsilon, 1]^N \quad (2.13)$$

then almost surely

$$\dim_H X([\varepsilon, 1]^N) \geq \min \left\{ d, \sum_{i=1}^N \frac{1}{H_i} \right\} \quad (2.14)$$

$$\dim_H \text{Gr} X([\varepsilon, 1]^N) \geq \min \left\{ \sum_{i=1}^N \frac{1}{H_i}; \sum_{j=1}^k \frac{H_k}{H_j} + N - k + (1 - H_k)d, 1 \leq k \leq N \right\} \quad (2.15)$$

Proof. Since, for any $\varepsilon > 0$, $\dim_H X([\varepsilon, 1]^N) \geq \dim_H X([\varepsilon, 1]^N)$, it is sufficient to show $\dim_H X([\varepsilon, 1]^N) \geq \min \left\{ d, \sum_{i=1}^N \frac{1}{H_i} \right\}$ a.s.

Let σ be the image measure of $\lambda_N \llcorner [\varepsilon, 1]^N$ under the mapping $t \rightarrow X(t)$. Then σ is a random Borel measure on $X([\varepsilon, 1]^N)$. We only need to show

$$\mathbb{E} \int_{[\varepsilon, 1]^N} \int_{[\varepsilon, 1]^N} \frac{1}{|X(s) - X(t)|^\gamma} ds dt < \infty \quad (2.16)$$

for all $0 < \gamma < \min \left\{ d, \sum_{i=1}^N \frac{1}{H_i} \right\}$.

By Fubini's theorem and (2.13), ~~the left-hand side of (2.16) is at most~~

$$\begin{aligned} \mathbb{E} \int_{[\varepsilon, 1]^N} \int_{[\varepsilon, 1]^N} \frac{1}{\left(\sum_{i=1}^N |t_i - s_i|^{2H_i} \right)^{\gamma/2}} \cdot \mathbb{E} \left(\frac{1}{|\xi|^\gamma} \right) ds dt \\ \leq K \int_{[\varepsilon, 1]^N} \int_{[\varepsilon, 1]^N} \frac{1}{\left(\sum_{i=1}^N |t_i - s_i|^{2H_i} \right)^{\gamma/2}} ds dt \end{aligned}$$

where we have used the fact that, for any standard d -dimensional normal vector ξ and any $\gamma \in (0, d)$, $\mathbb{E}(|\xi|^{-\gamma}) < \infty$.

To continue, we will make use of the following fact: $\forall \alpha \in (0, 1)$ and ~~any~~ $p > \frac{1}{2\alpha}$,

$$\int_{-\infty}^{\infty} \frac{dx}{|x|^{2\alpha + p}} \leq K(A)^{-(p - \frac{1}{2\alpha})} \quad (2.17)$$

Using (2.17) repeatedly, we have $\varepsilon_j < \infty$. This proves (2.16).

In order to prove (2.15), we need the following lemma.

Lemma 2.5. Let α, β and η be positive constants. For $A > 0$ and $B > 0$,

let
$$J := J(A, B) = \int_0^1 \frac{dx}{(A+x^\alpha)^\beta (B+x)^\eta}.$$

Then there exists a finite constant $K > 0$, depending on α, β, η only, such that the following holds for all $A, B > 0$ satisfying $A^{\frac{1}{\alpha}} \leq K \cdot B$:

(i) If $\alpha\beta > 1$, then

$$J \leq K \cdot \frac{1}{A^{\beta - \frac{1}{\alpha}} B^\eta}$$

(ii) If $\alpha\beta = 1$, then

$$J \leq K \frac{1}{B^\eta} \log(1 + BA^{-\frac{1}{\alpha}})$$

(iii) If $0 < \alpha\beta < 1$, and $\alpha\beta + \eta \neq 1$, then

$$J \leq K \left(\frac{1}{B^{\alpha\beta + \eta - 1}} + 1 \right)$$

Proof. Omitted. Cf. Ayache and Xiao (2005)

Now we prove the lower bound in (2.15). Since $\dim_{\mathbb{H}} \text{Gr}X([0,1]^N) \geq \dim_{\mathbb{H}} X([0,1]^N)$ always holds, we only need to consider the case when

$$\sum_{j=1}^{k-1} \frac{1}{H_j} \leq d < \sum_{j=1}^k \frac{1}{H_j} \quad \text{for some } 1 \leq k \leq N.$$

Here and in the sequel, $\sum_{j=1}^0 \frac{1}{H_j} = 0$.

Let $0 < \varepsilon < 1$ and $0 < \delta < \sum_{j=1}^k \frac{H_k}{H_j} + N - k + (1 - H_k)d$ be fixed, but arbitrary constants. We assume δ is close to the right-end point so that $\delta \in (N - k + d, N - k + d + 1)$. In order to show $\dim_{\mathbb{H}} \text{Gr}X([0,1]^N) \geq \delta$ a.s., it is sufficient to verify

$$G_\delta = \int_{[0,1]^N} \int_{[0,1]^N} \mathbb{E} \left[\frac{1}{(|1-s-t|^2 + |X(s)-X(t)|^2)^{\delta/2}} \right] ds dt < \infty \quad (2.18)$$

Denote $\sigma^2(s,t) = \mathbb{E}[(X_1(s) - X_1(t))^2]$. We write for all $s \neq t$,

$$\mathbb{E} \left[\frac{1}{(|s-t|^2 + |X(t) - X(s)|^2)^{r/2}} \right] = \mathbb{E} \left[\frac{1}{\sigma(s,t)^r \left(\frac{|s-t|^2}{\sigma^2(s,t)} + |\xi|^2 \right)^{r/2}} \right]$$

$$\leq K \frac{1}{\sigma(s,t)^d |s-t|^{r-d}},$$

where the last inequality follows from the fact that $\forall r > d$, $\forall a \in \mathbb{R}$ and $\xi \sim N(0, I)$,

$$\mathbb{E} \left[\frac{1}{(a^2 + |\xi|^2)^{r/2}} \right] \leq K a^{-(r-d)}. \quad (\text{verify it!})$$

Hence
$$G_r \leq K \int_0^1 dt_0 \dots \int_0^1 \frac{1}{\left(\sum_{i=1}^N t_i^{H_i} \right)^d \left(\sum_{i=1}^N t_i \right)^{r-d}} dt_1.$$

Applying Lemma 2.5 repeatedly, one can verify $G_r < \infty$. This finishes the proof of (2.18), and, ~~hence~~ hence, Proposition 2.4.

Exercise ~~Finish~~ Finish the last part of the proof.

Further questions arise, for example,

- (1) What if we replace $[0,1]^N$ by an arbitrary Borel set $E \subseteq (0,\infty)^N$?
- (2) What about the packing dimension of $X(E)$ and $\text{Gr} X(E)$?
- (3) Instead of Hausdorff dimension, can we consider the exact Hausdorff and packing measures of $X([0,1]^N)$ and $\text{Gr} X([0,1]^N)$?

Answering these questions requires better understanding of the random field X .

§3. Level sets and Local times

For any $x \in \mathbb{R}^d$, let $L_x = \{t \in (0, \infty)^N : X(t) = x\}$ be the x -level set of $X = \{X(t), t \in \mathbb{R}^N\}$. We are interested in the following two questions:

- (i) when is $L_x \neq \emptyset$ with positive probability? (X hits points)
- (ii) when $L_x \neq \emptyset$, what is $\dim_H L_x = ?$

We assume the following conditions:

(3.1) $\forall \varepsilon > 0, \exists K_1, K_2$ such that

$$\mathbb{E}(X_1(t) - X_1(s))^2 \leq K_2 \sum_{i=1}^N |t_i - s_i|^{2H_i} \quad \forall s, t \in [\varepsilon, 1]^N$$

(3.2) $\mathbb{E}(X_1(t)^2) \geq K$ for all $t \in [\varepsilon, 1]^N$ and

$$\text{Var}(X_1(t) | X_1(s)) \geq K_1 \sum_{i=1}^N |t_i - s_i|^{2H_i} \quad \forall s, t \in [\varepsilon, 1]^N.$$

Remarks.

① Note that $\mathbb{E}(X_1(t) - X_1(s))^2 \geq \text{Var}(X_1(t) | X_1(s))$, so (3.2) implies

$$\mathbb{E}(X_1(t) - X_1(s))^2 \geq K_1 \sum_{i=1}^N |t_i - s_i|^{2H_i}$$

② Since $\text{detcov}(X_1(s), X_1(t)) = \text{Var}(X_1(s)) \cdot \text{Var}(X_1(t) | X_1(s))$, (3.1) and (3.2)

imply that

$$\text{detcov}(X_1(s), X_1(t)) \geq K \sum_{i=1}^N |t_i - s_i|^{2H_i}, \quad \forall s, t \in [\varepsilon, 1]^N$$

Theorem 3.1 Assume the above conditions hold. ~~We~~ We have

(i) If $\sum_{i=1}^N \frac{1}{H_i} < d$, then for every $x \in \mathbb{R}^d$, $L_x = \emptyset$ a.s.

(ii) If $\sum_{i=1}^N \frac{1}{H_i} > d$, then for every $x \in \mathbb{R}^d$ and $0 < \varepsilon < 1$, with

positive probability,

$$\dim_H(L_x \cap [\varepsilon, 1]^N) = \min \left\{ \sum_{j=1}^k \frac{H_k}{H_j} + N - k - H_k d, 1 \leq k \leq N \right\}$$

$$= \sum_{j=1}^k \frac{H_k}{H_j} + N - k - H_k d \quad \text{if } \sum_{j=1}^{k-1} \frac{1}{H_j} \leq d < \sum_{j=1}^k \frac{1}{H_j}.$$

(3.3)

Remark The case $\sum_{k=1}^N \frac{1}{H_k} = d$ is unsolved in general. For X being the Brownian sheet or (N, d) fractional Brownian motion, Orey and Pruitt (1973), Talagrand (1978) proved $L_X = \emptyset$ a.s., respectively.

Proof of Theorem 3.1 For any $n \geq 2$, we divide the interval $[0, 1]^N$ into $m_n \leq n^{\sum_{k=1}^N \frac{1}{H_k}}$ sub-rectangles $R_{n,i}$ of side-lengths $n^{-\frac{1}{H_k}}$, $k=1, 2, \dots, N$. Let $\delta \in (0, 1)$ be fixed, and let $\tau_{n,i}$ be the lower-left vertex of $R_{n,i}$. Then

$$\begin{aligned} \mathbb{P}\{x \in X(R_{n,i})\} &\leq \mathbb{P}\left\{\max_{s, t \in R_{n,i}} |X(s) - X(t)| \leq n^{-(1-\delta)}, x \in X(R_{n,i})\right\} \\ &\quad + \mathbb{P}\left\{\max_{s, t \in R_{n,i}} |X(s) - X(t)| > n^{-(1-\delta)}\right\} \\ &\leq \mathbb{P}\left(|X(\tau_{n,i}) - x| \leq n^{-(1-\delta)}\right) + e^{-cn^{2\delta}} \\ &\leq K \cdot n^{-(1-\delta)d} \end{aligned} \tag{3.4}$$

where, to derive the second inequality, we have used an estimate of the tail probability of the supremum of Gaussian process (cf. e.g. Talagrand, 1993). Hence, if $\sum_{k=1}^N \frac{1}{H_k} < d$, we can choose $\delta > 0$ such that $(1-\delta)d > \sum_{k=1}^N \frac{1}{H_k}$.

Let $N_n = \#$ of rectangles $R_{n,i}$ such that $x \in X(R_{n,i})$. It follows from (3.4) that

$$\mathbb{E}(N_n) \leq K n^{\sum_{k=1}^N \frac{1}{H_k}} \cdot n^{-(1-\delta)d} \rightarrow 0 \text{ as } n \rightarrow \infty$$

Since N_n is integer-valued, Fatou's lemma implies that a.s. $N_n = 0$ for infinitely many n integers $n \geq 1$. This proves $L_X = \emptyset$ a.s.

Now we assume $\sum_{k=1}^N \frac{1}{H_k} > d$ and define a covering $\{R_{n,i}'\}$ of $L_X \cap [0, 1]^N$

by

$$R_{n,i}' = \begin{cases} R_{n,i} & \text{if } x \in X(R_{n,i}) \\ \emptyset & \text{otherwise} \end{cases}$$

~~Hence~~, For every $k \in \{1, 2, \dots, N\}$, ~~the interval~~ $R_{n,i}'$ can be covered by

is $n^{\sum_{j=k+1}^N (H_k^{-1} - H_j^{-1})}$ cubes of side-length $n^{-H_k^{-1}}$. Thus, we can cover the level set $L_x \cap [\varepsilon, 1]^N$ by a sequence of cubes of side-length $n^{-\frac{1}{H_k}}$. Let $\delta \in (0, 1)$ be an arbitrary constant and let

$$\gamma = \sum_{j=1}^k \frac{H_k}{H_j} + N - k - H_k(1-\delta)d$$

It follows from (3.4) that

$$\begin{aligned} & \mathbb{E} \left[\sum_{i=1}^{m_n} n^{\sum_{j=k+1}^N (H_k^{-1} - H_j^{-1})} \cdot (n^{-H_k^{-1}})^{\gamma} \cdot \mathbb{1}_{\{x \in X(R_{n,i})\}} \right] \\ & \leq K n^{\sum_{i=1}^N H_k^{-1} + \sum_{j=k+1}^N (H_k^{-1} - H_j^{-1}) - \gamma H_k^{-1} - (1-\delta)d} = K \end{aligned}$$

Using Fatou's lemma again, we obtain that the γ -dimensional Hausdorff measure of $L_x \cap [\varepsilon, 1]^N$ is finite a.s. This implies $\dim_H(L_x \cap [\varepsilon, 1]^N) \leq \gamma$ a.s. Letting $\delta \downarrow 0$, we obtain

$$\dim_H(L_x \cap [\varepsilon, 1]^N) \leq \min_{1 \leq k \leq N} \left\{ \sum_{j=1}^k \frac{H_k}{H_j} + N - k - H_k d \right\} \quad \text{a.s.}$$

To prove the reverse inequality, we assume $\sum_{j=1}^{k-1} \frac{1}{H_j} \leq d < \sum_{j=1}^k \frac{1}{H_j}$ for some $1 \leq k \leq N$. Let $\delta > 0$ be a small constant such that

$$\gamma := \sum_{j=1}^k \frac{H_k}{H_j} + N - k - H_k(1+\delta)d > N - k$$

If we can prove that there is a constant $K > 0$, independent of δ , such that

$$\mathbb{P} \left\{ \dim_H(L_x \cap [\varepsilon, 1]^N) \geq \gamma \right\} \geq K \quad (3.5)$$

Then, by letting $\delta \downarrow 0$, we have

$$\mathbb{P} \left\{ \dim_H(L_x \cap [\varepsilon, 1]^N) \geq \sum_{j=1}^k \frac{H_k}{H_j} + N - k - H_k d \right\} \geq K > 0.$$

The proof of (3.5) is based on the capacity argument due to Kahane (1985). Similar methods have been applied by Adler (1981), Testard (1986) and Xiao (1995).

Let \mathcal{M}_r^+ be the space of all non-negative measures on \mathbb{R}_+^N with finite

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δ -energy. i.e. $\int_{\mathbb{R}^N} \int_{\mathbb{R}^N} |x-y|^{-\delta} \mu(dx) \mu(dy) < \infty$. It is known that M_δ^+ is a complete metric space under the metric

$$\|\mu\|_\delta = \int_{\mathbb{R}^N} \int_{\mathbb{R}^N} \frac{\mu(ds) \mu(dt)}{|s-t|^\delta} \quad (\text{verify this!})$$

see Adler (1981). Exercise Convergence in M_δ^+ implies weak convergence?

We define a sequence of random positive measures μ_n on the Borel sets C of $[\varepsilon, 1]^N$ by

$$\begin{aligned} \mu_n(C) &= (2\pi n)^{d/2} \cdot \int_C \exp\left(-\frac{n|x(t)-x|^2}{2}\right) dt \\ &= \int_C \int_{\mathbb{R}^d} \exp\left(-\frac{|\xi|^2}{2n} + i\langle X(t)-x, \xi \rangle\right) d\xi dt \end{aligned} \quad (3.6)$$

We will show the sequence $\{\mu_n\}$ has a subsequence that converges to a Borel measure μ on $[\varepsilon, 1]^N$, with positive probability. The measure μ must be supported on $L_X \cap [\varepsilon, 1]^N$.

It follows from Kahane (1985), Testard (1986) that, if there are constant $k_1 > 0$, $k_2 > 0$ such that

$$(3.7) \quad \mathbb{E}(\|\mu_n\|) \geq k_1, \quad \mathbb{E}(\|\mu_n\|^2) \leq k_2, \quad \text{where } \|\mu_n\| = \mu_n([\varepsilon, 1]^N)$$

$$(3.8) \quad \mathbb{E}(\|\mu_n\|_\delta) < \infty$$

then there is a subsequence of $\{\mu_n\}$, say $\{\mu_{n_k}\}$, such that $\mu_{n_k} \rightarrow \mu$ in M_δ^+ and μ is strictly positive with probability $\geq k_1^2/(2k_2)$. In this case, it follows (3.6) that the measure μ has its support in $L_X \cap [\varepsilon, 1]^N$ almost surely. Hence Frostman's theorem yields (3.5) with $K = k_1^2/(2k_2)$.

It remains to verify (3.7) and (3.8). By Fubini's theorem we have

$$\begin{aligned} \mathbb{E}(\|\mu_n\|) &= \int_{[\varepsilon, 1]^N} \int_{\mathbb{R}^d} \exp\left(-\frac{|\xi|^2}{2n}\right) \cdot \mathbb{E} \exp(i\langle X(t)-x, \xi \rangle) d\xi dt \\ &= \int_{[\varepsilon, 1]^N} \int_{\mathbb{R}^d} \exp\left(-\frac{1}{2}(n^{-1} + \sigma^2(t))|\xi|^2\right) \cdot e^{-i\xi \cdot x} d\xi dt \\ &= \int_{[\varepsilon, 1]^N} \left(\frac{2\pi}{n^{-1} + \sigma^2(t)}\right)^{d/2} \exp\left(-\frac{|x|^2}{2(n^{-1} + \sigma^2(t))}\right) dt \end{aligned}$$

$$\geq \int_{[\varepsilon, 1]^N} \left(\frac{2\pi}{H\sigma^2(t)} \right)^{d/2} \exp\left(-\frac{|x|^2}{2\sigma^2(t)}\right) dt = K_1$$

Denote by I_{2d} the identity matrix of order $2d$, and by $\text{cov}(X(s), X(t))$ the covariance matrix of $(X(s), X(t))$. Let $\Gamma = n^2 I_{2d} + \text{cov}(X(s), X(t))$. Then

$$\begin{aligned} \mathbb{E}(\|H_n\|^2) &= \int_{[\varepsilon, 1]^N} \int_{[\varepsilon, 1]^N} \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} e^{i\langle \xi, x \rangle} \exp\left(-\frac{1}{2}(\xi, \eta) \Gamma^{-1}(\xi, \eta)\right) d\xi d\eta ds dt \\ &= \int_{[\varepsilon, 1]^N} \int_{[\varepsilon, 1]^N} \frac{(2\pi)^d}{\sqrt{\det \Gamma}} \exp\left(-\frac{1}{2}(x, x) \Gamma^{-1}(x, x)\right) ds dt \\ &\leq \int_{[\varepsilon, 1]^N} \int_{[\varepsilon, 1]^N} \frac{(2\pi)^d}{\sqrt{\det \text{cov}(X(s), X(t))}} ds dt \end{aligned}$$

Note that

$$\begin{aligned} \det \text{cov}(X(s), X(t)) &= \det \text{cov}(X_1(s), X_1(t))^d = [\text{Var}(X(s)) \cdot \text{Var}(X(t)|X(s))]^d \\ &\geq K \left(\sum_{\lambda=1}^N |\lambda_2 - t_2|^{2H_2} \right)^d \end{aligned}$$

Therefore

$$\mathbb{E}(\|H_n\|^2) \leq K \int_{[\varepsilon, 1]^N} \int_{[\varepsilon, 1]^N} \frac{ds dt}{\left(\sum_{\lambda=1}^N |\lambda_2 - t_2|^{2H_2} \right)^{d/2}} = K_2 < \infty.$$

Since $d < \frac{N}{\sum_{\lambda=1}^N \frac{1}{H_2}}$.

Similarly, we can verify that

$$\begin{aligned} \mathbb{E}(\|H_n\|_g) &= \int_{[\varepsilon, 1]^N} \int_{[\varepsilon, 1]^N} \frac{ds dt}{|s-t|^g} \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} e^{-i\langle \xi, x \rangle} \exp\left(-\frac{1}{2}(\xi, \eta) \Gamma^{-1}(\xi, \eta)\right) d\xi d\eta \\ &\leq K \int_{[\varepsilon, 1]^N} \int_{[\varepsilon, 1]^N} \frac{1}{\left(\sum_{\lambda=1}^N |\lambda_2 - t_2| \right)^g \left(\sum_{\lambda=1}^N |\lambda_2 - t_2|^{2H_2} \right)^{d/2}} ds dt \\ &\leq K \int_0^1 \dots \int_0^1 \frac{1}{\left(\sum_{\lambda=1}^N t_2 \right)^g \left(\sum_{\lambda=1}^N t_2^{2H_2} \right)^{d/2}} dt. \end{aligned}$$

By using Lemma 25 repeatedly, we can show the last integral is convergent. This finishes the proof of Theorem 3.1. #

Now we turn to the existence of local times of X . As a matter of fact, the random measure μ constructed in the proof of theorem 3.1 is a local time $L(x, \cdot)$.

Let us recall ~~the~~ definition of local time of a function: $X: I \rightarrow \mathbb{R}^d$ where $I \subseteq \mathbb{R}^N$ is a Borel set. The occupation μ_x is defined by

$$\mu_x(C) = \lambda_N \{ t \in I : X(t) \in C \} \quad \forall C \in \mathcal{B}(\mathbb{R}^d)$$

If $\mu_x \ll \lambda_d$, then we say X has a local time on I and denote by

$$L(x, I) = \frac{d\mu_x}{d\lambda_d}(x)$$

The local time theory for Gaussian processes, ~~introduced~~ ^{initiated} by Berman 67-71 the local time. See Geman and Horowitz (1980) for more information

Theorem 3.2 Under the conditions of Theorem 3.1, for any rectangle $I \subseteq (0, \infty)^N$, X has a local time $L(\cdot, I) \in L^2(\mathbb{R}^d \times \Omega)$ if and only if $\sum_{i=1}^N \frac{1}{H_i} > d$. In the latter case, $L(x, I)$ admits the following L^2 -representation

$$L(x, I) = (2\pi)^{-d} \int_{\mathbb{R}^d} e^{-i\langle y, x \rangle} \int_I e^{i\langle y, X(s) \rangle} ds dy \quad (3.9)$$

Proof Let μ_x be the occupation measure of X defined above. It follows from theorem 21.9 in Geman and Horowitz (1980) that X has a local time $L(\cdot, I) \in L^2(\lambda \times \mathbb{P})$ if and only if

$$\int_{\mathbb{R}^d} \int_I \int_I \mathbb{E} e^{i\langle \frac{y}{\|y\|}, X(s) - X(t) \rangle} ds dt d\frac{y}{\|y\|} < \infty$$

Note that, the left hand side equals

$$\int_I \int_I \frac{ds dt}{(\mathbb{E}(X_1(s) - X_1(t))^2)^{d/2}} \leq K \int_I \int_I \frac{ds dt}{\left(\sum_{i=1}^N |s_i - t_i|^{2H_i} \right)^{d/2}}$$

which is finite if and only if $\sum_{i=1}^N \frac{1}{H_i} > d$.

Finally, when $\sum_{i=1}^N \frac{1}{H_i} > d$, (3.9) follows from the Plancherel theorem. #

§4. Joint continuity of the Local times

In order to study the regularity of the local times of X , we further assume ^{X has} the following "sectorial local nondeterminism".

(C3). $\forall \varepsilon > 0$, there exists a constant $K > 0$ such that for any integer $n \geq 1$, $u \in [\varepsilon, 1]^N$, $t^1, \dots, t^n \in (\varepsilon, 1]^N$, we have

$$\text{Var}(X(u) | X(t^1), \dots, X(t^n)) \geq K \sum_{l=1}^N \min \left\{ |u_l - t_l^j|^{2H_l} : 0 \leq j \leq n \right\}$$

where $t_l^0 = 0$ for $l=1, 2, \dots, N$.

$$\bar{H} = (H_1, \dots, H_N)$$

Under this condition, we say X is sectorially locally nondeterministic with index \bar{H} .

We will not deal with the question whether or when a Gaussian random field is "sectorially locally nondeterministic". Khoshnevisare and Xiao (2004) established this property for the Brownian sheet, Wu and Xiao (2006) proved that fractional Brownian sheets have this property.

Theorem 4.1. Let $X = \{X(t), t \in \mathbb{R}^N\}$ be an (N, d) Gaussian random field with i.i.d. components. If X_l is sectorially locally nondeterministic with index (H_1, \dots, H_N) such that $0 < H_1 \leq \dots \leq H_N < 1$ and $\sum_{l=1}^N \frac{1}{H_l} > d$. Then X has a jointly continuous local time. That is, the function $(x, t) \mapsto L(x, [\langle \varepsilon \rangle, \langle \varepsilon \rangle + t])$ is continuous.

The proof of theorem 4.1 is based on moment estimates for $L(x, B)$, $L(x+y, B) - L(x, B)$ and the multiparameter version of the Kolmogorov's continuity lemma.

Lemma 4.2 Suppose the conditions of theorem 4.1 hold. There is a constant $K_1 > 0$ such that for all $B = [a, a + \langle r \rangle]$, where $a \in (\varepsilon, \infty)^d$, all $x \in \mathbb{R}^d$ and $n \geq 1$,

$$\mathbb{E} [L(x, B)^n] \leq K_1^n (n!)^N r^{nH_1 (\sum_{l=1}^N \frac{1}{H_l} - d)} \quad (4.1)$$

(3.9)

Proof. It follows from $\left[\begin{array}{l} \text{see also} \\ (25.5) \text{ in Geman and Horowitz (1980)} \end{array} \right]$ that

$$\begin{aligned} \mathbb{E}[L(x, B)^n] &= (2\pi)^{-nd} \int_{B^n} \int_{\mathbb{R}^{nd}} \exp(-i \sum_{j=1}^n \langle u^j, x \rangle) \cdot \mathbb{E} \exp(i \sum_{j=1}^n \langle u^j, x(t^j) \rangle) \\ &\leq (2\pi)^{-nd} \int_{B^n} \int_{\mathbb{R}^{nd}} \exp\left[-\frac{1}{2} \text{Var}\left(\sum_{j=1}^n \langle u^j, x(t^j) \rangle\right)\right] d\bar{u} d\bar{t} \\ &= \int_{B^n} \frac{(2\pi)^{-nd/2}}{\sqrt{\det \text{cov}(x(t^1), \dots, x(t^n))}} d\bar{t} \\ &= (2\pi)^{-\frac{nd}{2}} \int_{B^n} \frac{d\bar{t}}{[\det \text{cov}(x_1(t^1), \dots, x_1(t^n))]^{\frac{d}{2}}} \end{aligned}$$

For any Gaussian vector (Z_1, \dots, Z_n) , we have

$$\det \text{cov}(Z_1, \dots, Z_n) = \text{Var}(Z_1) \prod_{j=2}^n \text{Var}(Z_j | Z_1, \dots, Z_{j-1}).$$

It follows from this identity and the sectorial local nondeterminism that

$$\begin{aligned} \det \text{cov}(x_1(t^1), \dots, x_1(t^n)) &\geq \prod_{j=2}^n \text{Var}(x(t^j) | x(t^1), \dots, x(t^{j-1})) \\ &\geq K^n \prod_{j=2}^n \left(\sum_{k=1}^N \min\{|t_k^j - t_k^{i}|^{2H_k} : 0 \leq i \leq j-1\} \right) \end{aligned}$$

Hence

$$\mathbb{E}[L(x, B)^n] \leq K^n \int_{B^n} \prod_{j=2}^n \left(\sum_{k=1}^N \min\{|t_k^j - t_k^i|^{2H_k} : 0 \leq i \leq j-1\} \right)^{\frac{d}{2}} d\bar{t}$$

We integrate in the order $dt^n dt^{n-1} \dots dt^2 dt^1$. First consider the integral

$$\mathcal{I}_n = \int_{B^n} \frac{dt^n}{\left(\sum_{k=1}^N \min\{|t_k^n - t_k^i|^{2H_k} : 0 \leq i \leq n-1\} \right)^{\frac{d}{2}}},$$

where t^1, \dots, t^{n-1} are fixed, with distinct coordinates (the set of such points has full $(n-1)d$ -dimensional Lebesgue measure)

We introduce N permutations π_1, \dots, π_N of $\{1, 2, \dots, n-1\}$ such that for every $l = 1, \dots, N$,

$$t_l^{\pi_1(l)} < t_l^{\pi_2(l)} < \dots < t_l^{\pi_N(l)}$$

For convenience, we denote $t_l^{\pi_1(l)} = a_l$ and $t_l^{\pi_N(l)} = a_l + r$ [recall $B = [a, a+r]$]

For every sequence $(i_1, \dots, i_N) \in \{1, \dots, n-1\}^N$, let $\tau_{(i_1, \dots, i_N)} = (t_1^{\pi_{i_1}}, \dots, t_N^{\pi_{i_N}})$ be the "center" of the rectangle

$$I_{i_1, \dots, i_N} = \prod_{l=1}^N \left[t_l^{\pi_{i_l}} - \frac{1}{2} (t_l^{\pi_{i_l}} - t_l^{\pi_{i_l-1}}), t_l^{\pi_{i_l}} + \frac{1}{2} (t_l^{\pi_{i_l}} - t_l^{\pi_{i_l-1}}) \right]$$

with the convention that the left-end point of the interval is a_l whenever $i_l = 1$ and the interval is closed and its right-end is $a_l + r$ whenever $i_l = n-1$. Thus the rectangles $\{I_{i_1, \dots, i_N}\}$ form a partition of $B = [a, a+r]$.

For every $t^n \in B$, let I_{i_1, \dots, i_N} be the unique rectangle containing t^n .

then the sectorial local non-determinism of X yields

$$\text{Var}(X_t(t^n) | X_s(t^i), i \leq n-1) \geq K \sum_{l=1}^N |t_l^n - t_l^{\pi_{i_l}}|^{2H_l}$$

Hence we have

$$\begin{aligned} I_n &\leq K \sum_{(i_1, \dots, i_N)} \int_{I_{i_1, \dots, i_N}} \frac{dt^n}{\left(\sum_{l=1}^N |t_l^n - \tau_{(i_1, \dots, i_N)}^l|^{2H_l} \right)^{d/2}} \\ &\stackrel{(n!)}{\leq} K n^N \int_0^r \dots \int_0^r \frac{ds_1 \dots ds_N}{\left(\sum_{l=1}^N s_l^{2H_l} \right)^{d/2}} \\ &\stackrel{\text{does not matter}}{=} K n^N \cdot r^N \int_0^1 \dots \int_0^1 \frac{dt_1 \dots dt_N}{\left(\sum_{l=1}^N r^{2H_l} t_l^{2H_l} \right)^{d/2}} \\ &\leq K n^N r^{H_l \left(\sum_{j=1}^N \frac{1}{H_j} \right) - d} \end{aligned}$$

Repeating this procedure, we derive

$$\mathbb{E}(L(x, B)^n) \leq K^n (n!)^N \cdot r^{n H_l \left(\sum_{j=1}^N \frac{1}{H_j} \right) - d}$$

#

Lemma 4.2. Assume the conditions of theorem 4.1 hold. Then, for any $\gamma \in (0, 1)$ small, there exists a positive and finite constant K_2 such that for all $B = [a, a + \langle r \rangle]$, $x, y \in \mathbb{R}^d$ and all even numbers $n \geq 2$,

$$\mathbb{E} \left[(L(x, B) - L(y, B))^n \right] \leq K_2^n (n!)^N |x - y|^{\frac{n}{N} \gamma} r^{n H_1 \left(\sum_{j=1}^N \frac{1}{H_j} - d - \gamma \right)}$$

Proof. It is known that for every even integer $n \geq 2$

$$\mathbb{E} \left[(L(x, B) - L(y, B))^n \right] = (2\pi)^{-nd} \int_{B^n} \int_{\mathbb{R}^{nd}} \frac{r^n}{j!} \left(e^{-i \langle u^j, x \rangle} - e^{-i \langle u^j, y \rangle} \right) \cdot \mathbb{E} \exp \left(i \sum_{j=1}^n \langle u^j, X(t^j) \rangle \right) d\bar{u} d\bar{t} \quad (4.2)$$

where $\bar{u} = (u^1, \dots, u^n)$, $\bar{t} = (t^1, \dots, t^n)$ and each $u^j \in \mathbb{R}^d$, $t^j \in (\varepsilon, \infty)^N$. The details that lead to (4.2) are explained in Geman and Horowitz (1980); see also Pitt (1978)

Consider the non-decreasing function $\Delta(u) = \min\{1, |u|^\gamma\}^{\frac{1}{\langle \alpha \rangle \langle r \rangle}}$ and the elementary inequality:

$$|e^{iu} - 1| \leq 2\Delta(u), \quad \forall u \in \mathbb{R}$$

By the triangle inequality,

$$\begin{aligned} |e^{-i \langle u^j, x \rangle} - e^{-i \langle u^j, y \rangle}| &\leq \sum_{k=1}^d |e^{-i u_k^j (x_k - y_k)} - 1| \\ &\leq 2 \sum_{k=1}^d \Delta(u_k^j (x_k - y_k)). \end{aligned}$$

It follows that

$$\begin{aligned} \mathbb{E} \left[(L(x, B) - L(y, B))^n \right] &\leq (2\pi)^{-nd} \cdot 2^n \sum' \int_{B^n} \int_{\mathbb{R}^{nd}} \frac{r^n}{m!} \Delta(u_{k_m}^m (x_{k_m} - y_{k_m})) \cdot \\ &\quad \cdot \exp \left(-\frac{1}{2} \text{Var} \left(\sum_{j=1}^n \langle u^j, X(t^j) \rangle \right) \right) d\bar{u} d\bar{t} \end{aligned}$$

$$\leq K^n \sum' \int_{B^n} d\bar{t} \frac{r^n}{m!} \left\{ \int_{\mathbb{R}^{nd}} \Delta(u_{k_m}^m (x_{k_m} - y_{k_m}))^n \cdot \exp \left(-\frac{1}{2} \text{Var} \left(\sum_{j=1}^n \langle u^j, X(t^j) \rangle \right) \right) d\bar{u} \right\}^{1/n}$$

where the last inequality follows from Hölder's inequality.

The summation \sum' is taken over all sequences $(k_1, \dots, k_n) \in \{1, 2, \dots, d\}^n$ and

Fix $\bar{k} = (k_1, \dots, k_n) \in \{1, \dots, d\}^n$, and $t^1, \dots, t^n \in B$, we proceed to estimate the integral

$$M = \int_{\mathbb{R}^{nd}} \Delta \left(\frac{u_{k_m}}{k_m} (x_m - y_m) \right) \exp \left[-\frac{1}{2} \text{Var} \left(\sum_{j=1}^n \langle u^j, X(t^j) \rangle \right) \right] d\bar{u}$$

By a lemma of Cuzick and Dupreez (1982),

$$M = \frac{(2\pi)^{(nd-1)/2}}{(\det \text{cov}(X(t^1), \dots, X(t^n)))^{1/2}} \int_{\mathbb{R}^n} \Delta^n \left(\frac{x_m - y_m}{\sigma_j(\bar{k})} z \right) e^{-\frac{z^2}{2}} dz,$$

where $\sigma_j^2(\bar{k})$ is the conditional variance of $X_{k_m}(t^j)$ given $X_k(t^i)$ ($l \neq k_m$ and $1 \leq i \leq n$, or $l = k_m$ and $i \neq j$).

Since X_1, \dots, X_d are i.i.d., we have

$$\sigma_j^2(\bar{k}) = \text{Var}(X_1(t^j) | X_1(t^i), i \neq j) \geq k \cdot \sum_{i=1}^n \min \{ |t^j - t^i|, i \neq j \}$$

The rest of the proof follows similar arguments in Khoshnevisan and Xiao (2004).

We omit the details.