

Gaussian Random Fields: Spectral Measures and Fine Properties

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Fractals and Related Fields, II

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Outline

- Gaussian fields with stationary increments
- Analytic properties of Gaussian fields
- Fractal properties of Gaussian fields
- Spectral condition for strong local nondeterminism

Let $X = \{X(t), t \in \mathbb{R}^N\}$ be an \mathbb{R}^d -valued random field defined by

$$X(t) = (X_1(t), \dots, X_d(t)). \quad (1)$$

It is of interest to study the geometric properties of the following random sets:

- **Range** $X([0, 1]^N) = \{X(t) : t \in [0, 1]^N\}$.
- **Graph** $\text{Gr}X([0, 1]^N) = \{(t, X(t)) : t \in [0, 1]^N\}$.
- **Level set** $X^{-1}(x) = \{t \in \mathbb{R}^N : X(t) = x\}$.
- **Excursion set** $X^{-1}(F) = \{t \in \mathbb{R}^N : X(t) \in F\}$, $\forall F \subset \mathbb{R}^d$.
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They arise as

- scaling limits of stochastic systems;
- solutions to stochastic partial differential equations.

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- Construct covariance functions (non-negative definite functions).
- For stationary Gaussian random fields or those with stationary increments, use spectral representation theorem.

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1. Gaussian fields with stationary increments

Let $X = \{X(t), t \in \mathbb{R}^N\}$ be a centered Gaussian field with stationary increments and $X(0) = 0$. If $R(s, t) = \mathbb{E}[X(s)X(t)]$ is continuous, then $R(s, t)$ can be written as

$$R(s, t) = \int_{\mathbb{R}^N} (e^{i\langle s, \lambda \rangle} - 1)(e^{-i\langle t, \lambda \rangle} - 1) \Delta(d\lambda),$$

where $\Delta(d\lambda)$ is a Borel measure which satisfies

$$\int_{\mathbb{R}^N} \frac{\|\lambda\|^2}{1 + \|\lambda\|^2} \Delta(d\lambda) < \infty. \quad (2)$$

The measure Δ is called the *spectral measure* of X .

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It follows that X has the stochastic integral representation:

$$X(t) \stackrel{d}{=} \int_{\mathbb{R}^N} (e^{i\langle t, \lambda \rangle} - 1) \mathcal{M}(d\lambda),$$

where $\stackrel{d}{=}$ denotes equality of all finite-dimensional distributions, $\mathcal{M}(d\lambda)$ is a centered complex-valued Gaussian random measure with Δ as its control measure.

For any $h \in \mathbb{R}^N$ we have

$$\mathbb{E}(X(t+h) - X(t))^2 = 2 \int_{\mathbb{R}^N} (1 - \cos\langle h, \lambda \rangle) \Delta(d\lambda).$$

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- absolutely continuous with density $f(\lambda)$, or
- singular with fractal support (e.g., a self-similar measure), or
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Some examples

Example 1.1. For $H \in (0, 1)$, fractional Brownian motion $B^H = \{B^H(t), t \in \mathbb{R}^N\}$ with index H is a centered Gaussian random field with

$$\mathbb{E}(B^H(s)B^H(t)) = \frac{1}{2} \left(\|s\|^{2H} + \|t\|^{2H} - \|s - t\|^{2H} \right).$$

B^H is H -self-similar, isotropic and has stationary increments.

Its spectral measure has a density function

$$f_H(\lambda) = c(H, N) \|\lambda\|^{-(2H+N)},$$

where $c(H, N) > 0$ is a constant.

Example 1.2. A class of Gaussian fields can be obtained by letting spectral density functions satisfy (2) and

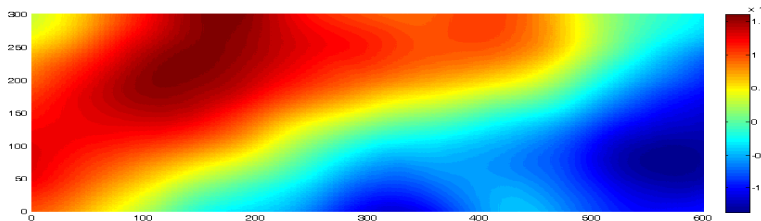
$$f(\lambda) \asymp \frac{1}{\left(\sum_{j=1}^N |\lambda_j|^{\beta_j}\right)^\gamma}, \quad \forall \lambda \in \mathbb{R}^N, \|\lambda\| \geq 1, \quad (3)$$

where $(\beta_1, \dots, \beta_N) \in (0, \infty)^N$ and

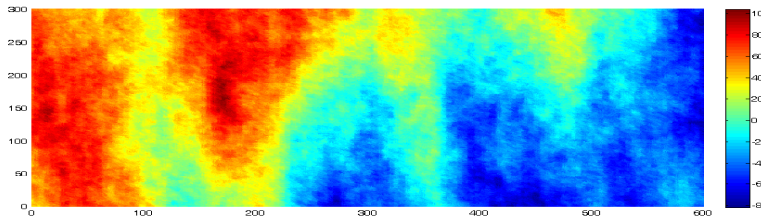
$$\gamma > \sum_{j=1}^N \frac{1}{\beta_j}.$$

Gaussian fields with such spectral densities arise in studies of stochastic fractional heat equations and in statistics.

A smooth Gaussian field



A rough Gaussian field



2. Analytic properties of Gaussian fields

Let $X = \{X(t), t \in \mathbb{R}^N\}$ be a Gaussian random field with spectral density (3). Then

- X is anisotropic and has rich geometric structures.
- $X(t)$ may be smooth in some (or all) directions, “rough” along some other (or all) directions (so that it **generates random fractals**).

Theorem 2.1 (Xue and X. 2009)

(i) If

$$\beta_j \left(\gamma - \sum_{i=1}^N \frac{1}{\beta_i} \right) > 2, \quad (4)$$

then the partial derivative $\partial X(t)/\partial t_j$ is continuous almost surely. In particular, if (4) holds for all $1 \leq j \leq N$, then almost surely $X(t)$ is continuously differentiable.

(ii) If

$$\max_{1 \leq j \leq N} \beta_j \left(\gamma - \sum_{i=1}^N \frac{1}{\beta_i} \right) \leq 2, \quad (5)$$

then $X(t)$ is not differentiable in any direction.

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Uniform modulus of continuity

Theorem 2.2 (Meerschaert, Wang and X. 2010)

Under condition (5), we have

$$\limsup_{\varepsilon \rightarrow 0} \sup_{\substack{s, t \in [0, 1]^N, \\ \rho(s, t) \leq \varepsilon}} \frac{|X(t) - X(s)|}{\rho(s, t) \sqrt{\log(1 + \rho(s, t)^{-1})}} = c_{2,1}, \quad (6)$$

where $0 < c_{2,1} < \infty$ is a constant, $\rho(s, t) = \sum_{j=1}^N |s_j - t_j|^{H_j}$, and for every $1 \leq j \leq N$,

$$H_j = \frac{\beta_j}{2} \left(\gamma - \sum_{i=1}^N \frac{1}{\beta_i} \right) \in (0, 1]. \quad (7)$$

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Ingredients for the proof

- The proof of the upper bound of (6) relies on estimate of the tail probability: For $\varepsilon > 0$ and all $x \geq c_{2,2}\varepsilon\sqrt{\log(1 + \varepsilon^{-1})}$,

$$\mathbb{P}\left\{ \sup_{\substack{s, t \in [0, 1]^N, \\ \rho(s, t) \leq \varepsilon}} |X(t) - X(s)| \geq x \right\} \leq \exp\left(-c_{2,3}\frac{x^2}{\varepsilon^2}\right).$$

- The proof of the lower bound of (6), and most of the results in this talk rely on the following **property of strong local nondeterminism (SLND)**.

Strong local nondeterminism

Under condition (5),

$$f(\lambda) \asymp \frac{1}{\left(\sum_{j=1}^N |\lambda_j|^{H_j}\right)^{Q+2}}, \quad \forall \|\lambda\| \geq 1, \quad (8)$$

where $Q = \sum_{i=1}^N \frac{1}{H_i}$.

Lemma 2.3 (X. 2009)

\exists constant $c_{2,4} > 0$ such that for all $n \geq 1$ and $u, t^1, \dots, t^n \in [0, 1]^N$,

$$\text{Var}\left(X(u) \mid X(t^1), \dots, X(t^n)\right) \geq c_{2,4} \min_{0 \leq k \leq n} \rho(u, t^k)^2, \quad (9)$$

where $t^0 = 0$.

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Law of the iterated logarithm

Theorem 2.4 (Meerschaert, Wang and X., 2010)

Assume condition (5) holds, then for every $t_0 \in \mathbb{R}^N$

$$\limsup_{r \rightarrow 0} \frac{\max_{\rho(0,h) \leq r} |X(t_0 + h) - X(t_0)|}{r(\log \log 1/r)^{1/2}} = c_{2,5}, \quad a.s.,$$

where $0 < c_{2,5} < \infty$ is a constant.

This describes the **largest** local oscillation of $X(t)$, which is useful for estimating the **packing measure** of the range and graph.

Chung's law of the iterated logarithm

Theorem 2.5 (Luan and X., 2010)

Assume condition (5) holds, then for every $t_0 \in \mathbb{R}^N$

$$\liminf_{r \rightarrow 0} \frac{\max_{\rho(0,h) \leq r} |X(t_0 + h) - X(t_0)|}{r(\log \log 1/r)^{-1/Q}} = c_{2,6}, \quad a.s.,$$

where $0 < c_{2,6} < \infty$ is a constant.

This describes the **smallest** local oscillation of $X(t)$, which will be useful for estimating the **Hausdorff measure** of the range and graph.

3. Fractal properties of Gaussian fields

Let $X = \{X(t), t \in \mathbb{R}^N\}$ be an (N, d) Gaussian random field defined by (1). That is,

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

We assume that X_1, \dots, X_d are independent copies of a real-valued Gaussian field X_0 .

We consider first the fractal properties of the range $X([0, 1]^N)$, or more generally, $X(E)$, where $E \subseteq [0, 1]^N$ is a Borel set.

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3.1 Exact Hausdorff measure of the ranges

Let $\sigma^2(s, t) = \mathbb{E} (X_0(s) - X_0(t))^2$ and ρ be the metric on \mathbb{R}^N defined by $\rho(s, t) = \sum_{j=1}^N |s_j - t_j|^{H_j}$ for $s, t \in \mathbb{R}^N$.

Theorem 3.1 (Ayache and X. 2005; X. 2009)

If

$$c_{3,1} \rho^2(s, t) \leq \sigma^2(s, t) \leq c_{3,2} \rho^2(s, t), \quad \forall s, t \in [0, 1]^N$$

for some constants $c_{3,1} > 0$ and $c_{3,2} < \infty$. Then

$$\dim_{\text{H}} X ([0, 1]^N) = \min \left\{ d; \sum_{j=1}^N \frac{1}{H_j} \right\}, \quad \text{a.s.}$$

Question: Is there a function φ such that

$$0 < \mathcal{H}_\varphi(X([0, 1]^N)) < \infty \quad \text{a.s.}?$$

Let $B^H = \{B^H(t), t \in \mathbb{R}^N\}$ be a fractional Brownian motion in \mathbb{R}^d with Hurst index $H \in (0, 1)$. Talagrand (1995) showed that, **if $N < Hd$, then**

$$0 < \mathcal{H}_\varphi(B^H([0, 1]^N)) < \infty \quad \text{a.s.},$$

where

$$\varphi(r) = r^{\frac{N}{H}} \log \log(1/r).$$

Theorem 3.2 (Luan and X., 2010)

Assume that X_0 has stationary increments with spectral density $f(\lambda)$ which satisfies (8).

(i). If $\sum_{j=1}^N \frac{1}{H_j} > d$, then a.s. $X([0, 1]^N)$ has positive Lebesgue measure (interior points).

(ii). If $\sum_{j=1}^N \frac{1}{H_j} < d$, then \exists constants $c_{3,3} > 0$ and $c_{3,4} < \infty$

$$c_{3,3} \leq \mathcal{H}_\varphi(X([0, 1]^N)) \leq c_{3,4} \quad \text{a.s.},$$

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Proof of $\mathcal{H}_\varphi(X([0, 1]^N)) \geq c$

Let μ be the occupation measure of X , defined by

$$\mu(B) = \lambda_N \{t \in [0, 1]^N : X(t) \in B\}, \quad \forall B \in \mathcal{B}(\mathbb{R}^d).$$

Lemma 3.3

For every $t_0 \in [0, 1]^N$, let $T_{t_0}(r) = \mu(B(X(t_0), r))$. Then

$$\limsup_{r \rightarrow 0} \frac{T_{t_0}(r)}{r^{\sum_{j=1}^N \frac{1}{H_j}} \log \log 1/r} \leq c < \infty, \quad \text{a.s.}$$

Sketch of Proof. Since

$$T_{t_0}(r) = \int_{[0,1]^N} \mathbf{1}_{\{\|X(t) - X(t_0)\| \leq r\}} dt,$$

we use Fubini's theorem to get

$$\begin{aligned} \mathbb{E}(T(r)) &= \int_{[0,1]^N} \mathbb{P} \{ \|X(t) - X(t_0)\| < r \} dt \\ &\leq \int_{[0,1]^N} \min \left\{ 1, \left(\frac{r}{\rho(t, t_0)} \right)^d \right\} dt \\ &\leq c r^Q, \end{aligned}$$

where $Q = \sum_{j=1}^N H_j^{-1}$.

For all $n \geq 2$,

$$\begin{aligned} & \mathbb{E} (T_{t_0}(r)^n) \\ &= \int_{[0,1]^{Nn}} \mathbb{P} \{ \|X(t^j) - X(t_0)\| \leq r, 1 \leq j \leq n \} dt^1 \cdots dt^n \end{aligned}$$

Using the **property of SLND**, conditioning and induction, we have

$$\mathbb{E} (T_{t_0}(r)^n) \leq c^n n! r^{Qn}.$$

This leads to Lemma 3.3.

Proof of $\mathcal{H}_\varphi(X([0, 1]^N)) \leq c$

The covering method was invented by Talagrand (1995), which is different from that for Markov processes.

In our case, we show

- For most of points $t_0 \in [0, 1]^N$, there is a sequence $r_n \downarrow 0$, such that $X(B_\rho(t_0, r_n))$ can be covered by a ball of radius $cr(\log \log 1/r)^{-1/Q}$. These are **good points**.
- For **bad points** $t_0 \in [0, 1]^N$, we can cover $X(B_\rho(t_0, r))$ by a ball of radius $c r \sqrt{\log 1/r}$.

The property of SLND plays an important role in the proof.

Question: What about the case $\sum_{j=1}^N \frac{1}{H_j} = d$?

3.2 Hausdorff dimension of $X(E)$

To determine $\dim_{\mathbb{H}} X(E)$ for an arbitrary Borel set $E \subset \mathbb{R}^N$, it is not enough to use $\dim_{\mathbb{H}} E$.

One needs to use a Hausdorff-type dimension, $\dim_{\mathbb{H}}^{\rho} E$, on the metric space (\mathbb{R}^N, ρ) .

Theorem 3.4 (Wu and X., 2007; X., 2009)

For any Borel set $E \subseteq \mathbb{R}^N$, we have

$$\dim_{\mathbb{H}} X(E) = \min\{d, \dim_{\mathbb{H}}^{\rho} E\}, \quad \text{a.s.} \quad (10)$$

A uniform version of (10) is proved by Wu and Xiao (2009). Its proof relies on SLND.

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3.3 Hausdorff dimension of $X^{-1}(F)$

Theorem 3.5 (Biermé, Lacaux and X., 2009)

For any Borel set $F \subseteq \mathbb{R}^d$ such that $\sum_{j=1}^N \frac{1}{H_j} > d - \dim_{\text{H}} F$, $\dim_{\text{H}} X^{-1}(F)$ equals

$$\min_{1 \leq k \leq N} \left\{ \sum_{j=1}^k \frac{H_k}{H_j} + N - k - H_k(d - \dim_{\text{H}} F) \right\}$$

$$= \sum_{j=1}^k \frac{H_k}{H_j} + N - k - H_k(d - \dim_{\text{H}} F),$$

$$\text{if } \sum_{j=1}^{k-1} \frac{1}{H_j} \leq d - \dim_{\text{H}} F < \sum_{j=1}^k \frac{1}{H_j}.$$

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Corollary 3.6

If $\sum_{j=1}^N \frac{1}{H_j} > d$, then for every $x \in \mathbb{R}^d$,

$$\dim_{\mathbb{H}} X^{-1}(x) = \min_{1 \leq k \leq N} \left\{ \sum_{j=1}^k \frac{H_k}{H_j} + N - k - H_k d \right\}$$

$$= \sum_{j=1}^k \frac{H_k}{H_j} + N - k - H_k d,$$

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

Question: Find an exact Hausdorff measure function for $X^{-1}(x)$.

Borel measure on $X^{-1}(x)$: local times of X

Theorem 3.7 (Ayache, Wu and X. 2008; Wu and X., 2011)

If $\sum_{j=1}^N \frac{1}{H_j} > d$, then

- X has a local time $L(x, t)$ which is continuous for $(x, t) \in \mathbb{R}^d \times \mathbb{R}^N$.
- for every $x \in \mathbb{R}^d$, the measure $L(x, \cdot)$ is supported by $X^{-1}(x)$ and for all $t \in \mathbb{R}^N$ and $r > 0$ small enough,

$$L(x, B(t, r)) \leq c_{3,5} r^\beta (\log \log(1/r))^\eta := \psi(r),$$

where $\beta = \dim_{\text{H}} X^{-1}(x)$ and η is a constant.

This implies $X^{-1}(x)$ has positive ψ -Hausdorff measure.

Borel measure on $X^{-1}(x)$: local times of X

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4. Spectral condition for strong local nondeterminism

For any $\lambda \in \mathbb{R}^N$ and $h > 0$, denote by $C(\lambda, h)$ the cube with side-length $2h$ and center λ , i.e.,

$$C(\lambda, h) = \{x \in \mathbb{R}^N : |x_j - \lambda_j| \leq h, j = 1, \dots, N\}.$$

Let $L^2(C(0, T))$ be the subspace of $g \in L^2(\mathbb{R}^N)$ whose support is contained in $C(0, T)$.

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Theorem 4.1 (Luan and X., 2010)

Let $\{Y(t), t \in \mathbb{R}^N\}$ be a real, centered Gaussian field with stationary increments and $Y(0) = 0$. If for some $h > 0$ the spectral measure Δ of Y satisfies

$$\begin{aligned} 0 < \liminf_{\|\lambda\| \rightarrow \infty} \rho(0, \lambda)^{Q+2} \Delta(C(\lambda, h)) \\ &\leq \limsup_{\|\lambda\| \rightarrow \infty} \rho(0, \lambda)^{Q+2} \Delta(C(\lambda, h)) < \infty, \end{aligned} \quad (11)$$

then for any $T > 0$ such that $ThN < \log 2$, for all $u, t^1, \dots, t^n \in C(0, T)$,

$$\text{Var}\left(Y(u) \mid Y(t^1), \dots, Y(t^n)\right) \geq c_{4,1} \min_{0 \leq k \leq n} \rho(u, t^k)^2,$$

Examples

Example 4.1. Let $\{\xi_n, n \in \mathbb{Z}^N\}$ and $\{\eta_n, n \in \mathbb{Z}^N\}$ be two independent sequences of i.i.d. $N(0, 1)$ random variables. Let

$$Z(t) = \sum_{n \in \mathbb{Z}^N} a_n (\xi_n \cos \langle n, t \rangle + \eta_n \sin \langle n, t \rangle), \quad t \in \mathbb{R}^N,$$

where $\{a_n, n \in \mathbb{Z}^N\}$ is a sequence of real numbers such that

$$a_n^2 \asymp \frac{1}{\left(\sum_{j=1}^N |n_j|^{H_j}\right)^{Q+2}}.$$

Then the Gaussian field $Y(t) = Z(t) - Z(0)$ has **the same local properties** as the Gaussian fields considered in the previous sections.

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Other examples include

- the solution of fractional stochastic heat equation on the circle (at fixed time t); see Tindel, Tudor and Viens (2003, 2004), Nualart and Viens (2009).
- spherical fractional Brownian motion, Istas (2005).

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Thank you