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## ON THE EXTREME VALUES OF GAUSSIAN PROCESSES

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### 1. Introduction

Let us consider a separable and measurable Gaussian process<sup>(1)</sup>  $X = \{X(t), t \geq 0\}$  with mean zero and with the covariance function  $\rho(t, s) = EX(t)X(s)$ . We assume that  $\rho(t, t)$  is independent of  $t$ , say  $v (> 0)$ . The asymptotic behaviours of  $\sup_{t \in [0, T]} X(t)$  as  $T \rightarrow \infty$  have been studied by various authors [2] [3] [4] [8] [9]. For example, Pickands [8] proved that

$$(*) \quad \frac{\sup_{t \in [0, T]} X(t)}{\sqrt{2v \log T}} \xrightarrow{T \uparrow \infty} 1 \quad \text{a. s. ,}$$

under the following conditions (for stationary Gaussian process),

$$\limsup_{t \rightarrow 0} t^{-\alpha} (v - \rho(t, 0)) < \infty, \quad \text{for some } \alpha > 0,$$

and  $\lim_{t \rightarrow \infty} \rho(t, 0) = 0$ .

In this note we shall prove (\*) under certain conditions (condition A and B in Section 2) weaker than Pickands' conditions. As an application of (\*), we can prove the Hölder continuity as well as the uniform Hölder continuity for *stationary* Gaussian processes;

$$\limsup_{t \downarrow 0} \frac{|X(t) - X(0)|}{\sqrt{4(v - \rho(0, t)) \log \log \frac{1}{t}}} = 1, \quad \text{a. s. ,}$$

$$\limsup_{h \downarrow 0} \frac{\sup_{t, s \in [0, 1], |t-s|=h} |X(t) - X(s)|}{\sqrt{4(v - \rho(0, h)) \log \frac{1}{h}}} = 1, \quad \text{a. s.}$$

There are many references on this subject (see, for example, [10]). Our method, different from the usual Borel-Cantelli method, consists in making use of some transformations of path functions to reduce the behaviour of path functions near  $t=0$  to that near  $t=\infty$ .

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(1) We mean a real valued process.

**2. Results**

Let  $X = \{X(t), t \geq 0\}$  be a separable and measurable Gaussian process with  $EX(t) = 0$ ,  $\rho(t, s) = EX(t)X(s)$  and  $EX^2(t) = v (> 0)$ . We shall introduce the following two conditions:

CONDITION A. For any  $t$  and  $s$ ,

$$(1) \quad 2(v - \rho(t, s)) (= E(X(t) - X(s))^2) < \psi^2(|t - s|)$$

where  $\psi$  is a non-decreasing and continuous function on  $[0, \infty)$  such that

$$(2) \quad \int_0^\infty \psi(e^{-x^2}) dx < \infty .$$

CONDITION B.

$$(3) \quad \limsup_{T \uparrow \infty} \sup_{|t-s| > T} \rho(t, s) \leq 0 .$$

(This condition A implies the continuity of almost all sample paths, by a theorem due to X. Fernique [6]).

In Section 3, we shall prove the following theorems,

**Theorem 1.** *Under condition A, we have*

$$\limsup_{T \uparrow \infty} \frac{\sup_{t \in [0, T]} |X(t)|}{\sqrt{2v \log T}} \leq 1$$

with probability 1.

**Theorem 2.** *Under condition B, we have*

$$\limsup_{T \uparrow \infty} \frac{\sup_{t \in [0, T]} X(t)}{\sqrt{2v \log T}} \geq 1$$

with probability 1.

Therefore, if condition A as well as condition B are satisfied, we can conclude that

$$\lim_{T \uparrow \infty} \frac{\sup_{t \in [0, T]} |X(t)|}{\sqrt{2v \log T}} = \lim_{T \uparrow \infty} \frac{\sup_{t \in [0, T]} X(t)}{\sqrt{2v \log T}} = 1$$

holds with probability 1.

Suppose that  $X$  is stationary and stochastically continuous. So, the covariance function  $\gamma(t-s) = \rho(t, s)$  is expressible in the form

$$\gamma(\tau) = \int_{-\infty}^\infty e^{i\tau\lambda} dF(\lambda)$$

with a bounded measure  $dF$ , symmetric with respect to 0. Moreover, the

measure  $dF$  can be split into the continuous part  $dF_c$  and the discontinuous part  $dF_d$ ;  $dF = dF_c + dF_d$ .

**Corollary.** *Let  $v_c = F_c(R^1)$ . If  $v_c$  is positive and if condition A is satisfied, then we have*

$$\limsup_{T \rightarrow \infty} \frac{\sup_{t \in [0, T]} |X(t)|}{\sqrt{2v_c \log T}} \leq 1$$

with probability 1. Moreover, if condition B is also satisfied, then

$$\lim_{T \rightarrow \infty} \frac{\sup_{t \in [0, T]} |X(t)|}{\sqrt{2v_c \log T}} = \lim_{T \rightarrow \infty} \frac{\sup_{t \in [0, T]} X(t)}{\sqrt{2v_c \log T}} = 1$$

with probability 1.

In Section 4, we shall show the following theorems for the stationary and stochastically continuous process  $X$ , using Theorems 1 and 2.

**Theorem 3.** *Let  $\sigma(t) = \sqrt{E|X(t) - X(0)|^2} > 0$ .*

*Suppose that there exist two positive constants  $\beta$  and  $L$  such that*

$$(4) \quad \frac{\sigma(ts)}{\sigma(t)} \leq Ls^\beta, \quad \text{for } t, s \in (0, 1],$$

and that

$$(5) \quad \sigma^2(t) - \sigma^2(t-h) \leq L\sigma^2(h), \quad \text{for small } t \text{ and } h.$$

Then we have

$$\limsup_{t \rightarrow 0} \frac{|X(t) - X(0)|}{\sigma(t) \sqrt{2 \log \log \frac{1}{t}}} = 1$$

with probability 1.

**Theorem 4.** *If the assumption (4) of Theorem 3 is valid and if  $\sigma^2(t)$  is concave in a small interval  $(0, \delta)$ , then we have*

$$\limsup_{h \rightarrow 0} \frac{\sup_{t, s \in [0, 1], |t-s|=h} |X(t) - X(s)|}{\sigma(h) \sqrt{2 \log \frac{1}{h}}} = 1$$

with probability 1.

### 3. Proof of Theorem 1 and 2

Without loss of generality, we may assume that  $v=1$ . Since  $X$  has continuous paths under condition A, Theorem 1 follows immediately from the statement

A. For any  $\varepsilon > 0$  and for almost all  $\omega$ , we can find a finite  $T_0(\varepsilon, \omega)$  such that, for all values of  $T$  greater than  $T_0(\varepsilon, \omega)$ , the inequality

$$\frac{\sup_{t \in [T_0, T]} |X(t, \omega)|}{\sqrt{2 \log T}} < 1 + \varepsilon$$

holds.

Let  $a(n) = [n^\varepsilon]^{(2)}$  and define  $\xi$  by

$$\xi(n, k) = X\left(n + \frac{k}{1+a(n)}\right), \quad k = 0, 1, \dots, a(n), \quad n = 0, 1, \dots$$

Using the following well-known inequality;

$$(6) \quad \frac{1}{\sqrt{2\pi}} \int_{|x| \geq c} e^{-x^2/2} dx < \frac{2}{c} e^{-c^2/2}$$

we can get

$$(7) \quad \sum_{n=0}^{\infty} \sum_{k=1}^{a(n)} P(|\xi(n, k)| \geq \sqrt{2 \log n} (1 + \varepsilon)) < \infty.$$

Therefore, using Borel-Cantelli's Lemma, we see that, for almost all  $\omega$ ,

$$\max_{k=0, \dots, a(n)} |\xi(n, k)| < \sqrt{2 \log n} (1 + \varepsilon), \quad \text{for large } n.$$

Define  $\eta$  by

$$\eta(n, k, j) = X\left(n + \frac{1}{1+a(n)} + \frac{j}{b(n)}\right) - X\left(n + \frac{k}{1+a(n)}\right),$$

$$j = 1, 2, \dots, \frac{b(n)}{1+a(n)}, \quad k = 0, 1, \dots, a(n), \quad n = 0, 1, \dots$$

where  $b(n) = (1+a(n)) \left[ \exp \frac{\varepsilon^3 (\log n)^2}{K} \right]$  and  $K = 2 \int_0^\infty \psi(e^{-x^2}) dx$ .

By virtue of condition A, we have

$$P(|\eta(n, k, j)| \geq \varepsilon \sqrt{2 \log n})$$

$$\leq P \left\{ \frac{|\eta(n, k, j)|}{D(\eta(n, k, j))} \geq \varepsilon \frac{\sqrt{2 \log n}}{D(\eta(n, k, j))} \right\} \quad \text{for } n = 1, 2, \dots,$$

$$\leq P \left\{ \frac{|\eta(n, k, j)|}{D(\eta(n, k, j))} \geq \varepsilon \frac{\sqrt{2 \log n}}{\psi(n^{-\varepsilon})} \right\},$$

where  $D(\cdot)$  stands for the standard deviation of a random variable. On the other hand, (2) implies

$$(8) \quad \psi(e^{-c^2}) \leq \frac{\sqrt{K}}{\sqrt{2c}}, \quad c < 0.$$

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(2)  $[c]$  is the integer part of  $c$ .

Hence, appealing to the inequality (6), we see

$$\begin{aligned} & \sum_{n=1}^{\infty} \sum_{k=0}^{a(n)} \sum_{j=1}^{b^*(n)} P(|\eta(n, k, j)| \geq \varepsilon \sqrt{2 \log n}) \\ & \leq \frac{1}{\varepsilon} \sum_{n=1}^{\infty} \frac{2\psi(n^{-\varepsilon})}{\sqrt{2 \log n}} \exp\left(\frac{\varepsilon^3(\log n)^2}{K} + \varepsilon(\log n) - \frac{\varepsilon^2 \log n}{\psi^2(n^{-\varepsilon})}\right) \end{aligned}$$

where  $b^*(n) = \frac{b(n)}{1+a(n)}$ . Combining this inequality with (8), we have

$$(9) \quad \sum_{n=1}^{\infty} \sum_{k=0}^{a(n)} \sum_{j=1}^{b^*(n)} P(|\eta(n, k, j)| \geq \varepsilon \sqrt{2 \log n}) < \infty .$$

We set  $c(p) = 2^{2^p}$  and define  $\zeta$  by

$$\begin{aligned} & \zeta(n, i, p, q, r) \\ & = X\left(n + \frac{i}{b(n)} + \frac{q}{b(n)c(p)} + \frac{r}{b(n)c(p+1)}\right) - X\left(n + \frac{i}{b(n)} + \frac{q}{b(n)c(p)}\right), \\ & \quad i = 0, 1, \dots, b(n) - 1, \quad q = 0, 1, \dots, c(p) - 1, \quad r = 1, \dots, c(p), \\ & \quad p = 1, 2, \dots, \quad n = 0, 1, \dots . \end{aligned}$$

Let  $Y(n, p) = \max_{i, q, r} |\zeta(n, i, p, q, r)|$  and  $Z(l, p) = \max_{c(l) \leq n \leq c(l+1)} Y(n, p)$ .

Then we have, for any  $h > 0$ ,

$$EZ(l, p) \leq h + \sum_{n=c(l)}^{c(l+1)} \sum_{r=1}^{c(p)} \sum_{q=0}^{c(p)-1} \sum_{i=0}^{b(n)-1} \int_h^{\infty} |x| d\mu_{\zeta(n, i, p, q, r)}(x)$$

where  $\mu_{\zeta}$  is the probability law of  $\zeta$ , ([5], Proposition 2). Hence

$$\begin{aligned} EZ(l, p) & \leq h + \sqrt{\frac{2}{\pi}} \sum_{n=c(l)}^{c(l+1)} \sum_{r, q, i} D(\zeta(n, i, p, q, r)) \exp\left(\frac{-h^2}{2D^2(\zeta(n, i, p, q, r))}\right) \\ & \leq h + b(c(l+1))c(p+1)c(l+1)\psi(1/b(c(l))c(p)) \exp\left(\frac{-h^2}{2\psi^2(1/b(c(l))c(p))}\right). \end{aligned}$$

Let  $h = h(l, p) = \sqrt{2 \log b(c(l+1))c(p+1)c(l+1)} \psi(1/b(c(l))c(p))$ .

Then, we see

$$(10) \quad EZ(l, p) \leq 2h(l, p) .$$

Recalling the definition  $b(n)$  and  $c(p)$ , we have

$$(11) \quad \sqrt{\log b(c(l+1))c(p+1)c(l+1)} \leq d(2^l + 2^{p/2}),$$

with a properly chosen constant  $d$  which may depend on  $\varepsilon$ . On the other hand, by (2),

$$\begin{aligned} (12) \quad & \sum_{p=1}^{\infty} 2^{p/2} \psi(1/b(c(l))c(p)) \leq \sum_{p=1}^{\infty} 2^{p/2} \psi(1/c(p)) \\ & \leq 3 \int_0^{\infty} \psi(2^{-x^2}) dx < \infty, \quad l = 1, 2, \dots . \end{aligned}$$

Furthermore, by (8), with a properly chosen constant  $d'$

$$(13) \quad \psi(1/b(c(l))c(p)) \leq d'(2^{2l} + 2^p)^{-1/2}.$$

Using the inequality, for  $\alpha \in (0, 1)$ ,

$$x + y \geq x^\alpha y^{1-\alpha} + x^{1-\alpha} y^\alpha \geq x^\alpha y^{1-\alpha}, \quad \text{for } x < 0, y < 0,$$

we have,

$$\psi(1/b(c(l))c(p)) \leq d' 2^{-(2/3)l} 2^{-(1/6)p}.$$

Hence, combining this with (11) and (12), we have

$$\sum_{p=1}^{\infty} h(l, p) \leq dd' 2^{l/3} \sum_{p=1}^{\infty} 2^{-p/6} + 3d \int_0^{\infty} \psi(2^{-x^2}) dx.$$

Therefore, by (10),

$$\sum_{p=1}^{\infty} 2^{-l/2} \sum_{p=1}^{\infty} EZ(l, p) < \infty.$$

For any  $\varepsilon > 0$ ,

$$\begin{aligned} P\left(\sum_{p=1}^{\infty} Y(n, p) \geq \varepsilon \sqrt{2 \log n}, \quad \text{for some } n \in (c(l), \dots, c(l+1))\right) \\ \leq P\left(\sum_{p=1}^{\infty} Z(l, p) \geq \varepsilon \sqrt{2 \log c(l)}\right) \\ \leq \frac{\sum_{p=1}^{\infty} EZ(l, p)}{\varepsilon \sqrt{2 \log c(l)}}. \end{aligned}$$

Hence, we have

$$(14) \quad \sum_{j=1}^{\infty} P\left(\sum_{p=1}^{\infty} Y(n, p) \geq \varepsilon \sqrt{2 \log n}, \quad \text{for some } n \in (c(l), \dots, c(l+1))\right) < \infty.$$

Since  $X$  has continuous paths, for  $t \in \left[ n + \frac{k}{1+a(n)} + \frac{j}{b(n)}, n + \frac{k}{1+a(n)} + \frac{j+1}{b(n)} \right]$ ,

$$|X(t)| \leq \sum_{p=1}^{\infty} Y(n, p) + |\eta(n, k, j)| + |\xi(n, k)|.$$

Therefore, recalling (7) (9) and (14), for almost all  $\omega$ , we can choose a finite  $N_0(\omega)$  so that, for  $n = N_0(\omega), N_0(\omega) + 1, \dots$

$$\sup_{t \in [n, n+1]} |X(t, \omega)| \sqrt{2 \log n} (1 + 3\varepsilon).$$

This completes the proof of statement A.

To prove Theorem 2, it is enough to show the statement.

B. For any  $\varepsilon > 0$ , we can find a finite  $T_0(\varepsilon)$  such that the inequality

$$\liminf_{N \uparrow \infty} \frac{\max_{j=1, \dots, N} X(jT_0, \omega)}{\sqrt{2 \log N}} > 1 - \varepsilon$$

holds, with probability 1.

Without any difficulty, we can carry out the same method as in [8] (pp. 203–204). We take  $T$  so that  $\sup_{|t-s|>T} \rho(t, s) < \varepsilon$ . Let  $\{\xi, \eta_n, n=1, 2, \dots\}$  be a system of independent Gaussian random variables with  $E\xi = E\eta_n = 0$ ,  $E\xi^2 = \varepsilon$  and  $E\eta_n^2 = 1 - \varepsilon$ . Put  $Y_t = \xi + \eta_t$ . Then

$$(15) \quad EY_t^2 = EX^2(T) = 1$$

and

$$EY_t Y_s \geq EX(IT) X(jT).$$

On the other hand, let  $R(= \{r_{ij}\})$  be a  $N \times N$  symmetric positive definite matrix with 1's along the diagonal. Define

$$Q(c; \{r_{ij}\}) \equiv \int_{-\infty}^c \int_{-\infty}^c \frac{1}{(2\pi)^{N/2} \sqrt{\det R}} \exp\left(-\frac{1}{2}(x_1, \dots, x_N) R^{-1}(x_1, \dots, x_N)^t\right) dx_1 \dots dx_N.$$

Then  $Q(c; \{r_{ij}\})$  is an increasing function of the arguments  $\{r_{ij}\}$ , ([2], p. 508). Combining this with (15), we get

$$(16) \quad P\left(\max_{k=1, \dots, N} X(Tk) \leq c\right) \leq P\left(\max_{k=1, \dots, N} Y_k \leq c\right).$$

For any  $\varepsilon'$ , ( $0 < \varepsilon' < 1$ ), we have

$$(17) \quad \begin{aligned} & \sum_{n=1}^{\infty} P\left(\max_{k=1, \dots, 2^n} Y_k \leq \sqrt{2(1-\varepsilon) \log 2^n} (1-\varepsilon')\right) \\ & \leq \sum_{n=1}^{\infty} P\left(\xi \leq -\frac{\varepsilon'}{2} \sqrt{2(1-\varepsilon) \log 2^n}\right) \\ & \quad + \sum_{n=1}^{\infty} P\left(\max_{k=1, \dots, 2^n} \eta_k \leq \left(1 - \frac{\varepsilon'}{2}\right) \sqrt{2(1-\varepsilon) \log 2^n}\right) < \infty, \end{aligned}$$

by the inequality of (6). Therefore, using (16) and (17), we have

$$\liminf_{N \uparrow \infty} \frac{\max_{k=1, \dots, N} X(kT)}{\sqrt{2 \log N}} > \sqrt{1-\varepsilon} (1-\varepsilon'), \quad \text{a.s.}$$

Since  $\varepsilon'$  is arbitrary, we get statement B.

To prove Corollary, we shall express  $X$  by the sum of mutually independent Gaussian processes so that

$$(18) \quad X(t) = \xi(t) + \sum_{n=0}^{\infty} \eta_n \cos \lambda_n t + \sum_{n=0}^{\infty} \zeta_n \sin \lambda_n t$$



where  $E\xi(t)=E\eta_n=E\zeta_n=0$  and the stationary Gaussian process  $\xi$  has the continuous spectral measure  $dF_c$ , ([7]). We define  $X_k(t)$  by

$$(19) \quad X_k(t) = X(t) - \sum_{n=0}^{k-1} \eta_n \cos \lambda_n t - \sum_{n=0}^{k-1} \zeta_n \sin \lambda_n t .$$

Then it is easily seen that

$$E |X_k(t) - X_k(s)|^2 \leq E |X(t) - X(s)|^2$$

and the process  $X_k$  also satisfies condition A. Therefore, by Theorem 1, we have

$$\limsup_{T \uparrow \infty} \frac{\sup_{t \in [0, T]} |X_k(t)|}{\sqrt{2v_k \log T}} \leq 1, \quad \text{a.s.},$$

where  $v_k = EX_k^2(t)$ . Since almost all sample paths of  $\sum_{n=0}^{k-1} \eta_n \cos \lambda_n t + \sum_{n=0}^{k-1} \zeta_n \sin \lambda_n t$  are bounded functions, we have

$$\limsup_{T \uparrow \infty} \frac{\sup_{t \in [0, T]} |X_k(t)|}{\sqrt{2v_k \log T}} = \limsup_{T \uparrow \infty} \frac{\sup_{t \in [0, T]} |X(t)|}{\sqrt{2v_k \log T}} \quad \text{a.s.}$$

Therefore, we obtain the former half of Corollary, since  $v_k$  tends to  $v_c$ .

As to the latter half, condition B implies

$$\liminf_{T \uparrow \infty} \frac{\sup_{t \in [0, T]} |X(t)|}{\sqrt{2v_c \log T}} \geq \sqrt{\frac{v}{v_c}} \geq 1$$

by Theorem 2. Hence, we have  $v=v_c$ . Therefore under conditions A and B, we complete the proof of Corollary.

#### 4. Proof of Theorem 3 and 4

To prove Theorem 3, we shall firstly derive the following inequality from assumption (4),

$$(20) \quad \limsup_{t \downarrow 0} \frac{|X(t) - X(0)|}{\sigma(t) \sqrt{2 \log \frac{1}{t}}} \leq 1, \quad \text{a.s.}$$

We shall introduce an auxiliary Gaussian process  $Y$  by

$$Y(n+t) = \frac{X(2^{-n} - t2^{-n-1}) - X(0)}{\sigma(2^{-n} - t2^{-n-1})}, \quad t \in [0, 1], \quad n=0, 1, \dots .$$

Since  $X$  has continuous paths by (4), ([1], [6]),  $Y$  is also a continuous Gaussian process with  $EY(t)=0$  and  $EY^2(t)=1$ . Moreover, using (4), we have

$$\begin{aligned} E|Y(n+t) - Y(n+s)|^2 &= \frac{\sigma^2((t-s)2^{-n-1}) - (\sigma(2^{-n} - t2^{-n-1}) - \sigma(2^{-n} - s2^{-n-1}))^2}{\sigma(2^{-n} - t2^{-n-1})\sigma(2^{-n} - s2^{-n-1})} \\ &\leq \frac{\sigma^2((t-s)2^{-n-1})}{\sigma(2^{-n} - t2^{-n-1})\sigma(2^{-n} - s2^{-n-1})} \\ &\leq L^2(t-s)^{2\beta}, \quad \text{for } t, s \in [0, 1]. \end{aligned}$$

Hence,

$$\begin{aligned} E|Y(n) - Y(n-s)|^2 &= E|Y(n-1+1) - Y(n-1+1-s)|^2 \\ &\leq L^2s^{2\beta}, \quad \text{for } s \in [0, 1]. \end{aligned}$$

Therefore, we have

$$E|Y(t) - Y(s)|^2 \leq 4L^2|t-s|^{2\beta}, \quad \text{for } |t-s| \leq 1.$$

On the other hand  $E|Y(t) - Y(s)|^2 \leq 4$ . Hence,  $Y$  satisfies condition A. So, Theorem 1 tells us that

$$\limsup_{T \rightarrow \infty} \frac{\max_{t \in [0, T]} |Y(t)|}{\sqrt{2 \log T}} \leq 1, \quad \text{a.s.},$$

holds. Therefore

$$\limsup_{T \rightarrow \infty} \frac{|Y(T)|}{\sqrt{2 \log T}} \leq 1, \quad \text{a.s.}$$

Hence, we have

$$\limsup_{t \rightarrow 0} \frac{|X(t) - X(0)|}{\sigma(t)\sqrt{2 \log \varphi(t)}} \leq 1, \quad \text{a.s.},$$

where  $\varphi$  is defined by  $\varphi(2^{-n} - \tau 2^{-n-1}) = n + \tau$  for  $\tau \in [0, 1]$  and  $n = 0, 1, \dots$ . Since

$$(21) \quad \left( \log \log \frac{1}{t} \right) / \log \varphi(t) \rightarrow 1, \quad \text{as } t \rightarrow 0,$$

we obtain (20).

By virtue of (4) and (5), we shall show the converse inequality of (20). For  $n < m$ , we have

$$\begin{aligned} EY(n+t)Y(m+s) &= \frac{1}{2} \frac{\sigma^2(2^{-n} - t2^{-n-1}) + \sigma^2(2^{-m} - s2^{-m-1}) - \sigma^2(2^{-n} - t2^{-n-1} - 2^{-m} + s2^{-m-1})}{\sigma(2^{-n} - t2^{-n-1})\sigma(2^{-m} - s2^{-m-1})} \\ &\leq \frac{1}{2}(L+1) \frac{\sigma^2(2^{-m} - s2^{-m-1})}{\sigma(2^{-n} - t2^{-n-1})\sigma(2^{-m} - s2^{-m-1})} \\ &\leq \text{const. } 2^{-\beta(m-n)}. \end{aligned}$$

So,  $Y$  satisfies condition B. Hence, for  $\varepsilon > 0$  and for almost all  $\omega$ , we can choose a finite  $T_0(\varepsilon, \omega)$  so that, for any  $T$  greater than  $T_0$ , the inequality

$$\frac{\max_{t \in [0, T]} Y(t, \omega)}{\sqrt{2 \log T}} > 1 - \varepsilon$$

holds. For any  $v$  smaller than  $\varphi^{-1}(T_0)^{(3)}$ , the inequality

$$\max_{u \in [v, 1]} \frac{X(u, \omega) - X(0, \omega)}{\sigma(u) \sqrt{2 \log \varphi(v)}} > 1 - \varepsilon$$

holds. Since  $\frac{X(u, \omega) - X(0, \omega)}{\sigma(u)}$  is continuous on  $(0, 1]$ , for  $\delta > 0$ , we can take  $S_0(\omega)$ , smaller than  $\varphi^{-1}(T_0)$ , so that, for any  $s$  smaller than  $S_0(\omega)$ ,

$$\max_{u \in [s, \delta]} \frac{X(u, \omega) - X(0, \omega)}{\sigma(u) \sqrt{2 \log \varphi(u)}} > 1 - \varepsilon .$$

Therefore, for any  $\delta > 0$ , and for almost all  $\omega$ ,

$$\sup_{u \in (0, \delta]} \frac{X(u, \omega) - X(0, \omega)}{\sigma(u) \sqrt{2 \log \varphi(u)}} > 1 - \varepsilon .$$

Combining this with (21), we get the converse inequality of (20).

To prove Theorem 4, we shall fix a positive  $\varepsilon$  arbitrarily and define by

$$\begin{aligned} \xi(n, k, l) &= \left( X\left(\frac{k+l}{b(n)}\right) - X\left(\frac{k}{b(n)}\right) \right) / \sigma\left(\frac{l}{b(n)}\right) \\ l &= 1, 2, \dots, a(n), \quad k = 0, 1, \dots, b(n), \quad n = 1, 2, \dots, \end{aligned}$$

where  $a(n) = [2^{ne}]$  and  $b(n) = 2^n a(n)$ . Using (6), we have

$$(22) \quad \sum_{n=1}^{\infty} \sum_{k=0}^{b(n)} \sum_{l=1}^{a(k)} P(|\xi(n, k, l)| \geq (1 + \varepsilon) \sqrt{2 \log 2^n}) < \infty .$$

Define a continuous Gaussian process  $\{Y(s), s \geq 0\}$  by

$$\begin{aligned} Y(N(n, k) + t) &= \begin{cases} \left( X\left(\frac{k+t}{b(n)}\right) - X\left(\frac{k}{b(n)}\right) \right) / \sigma\left(\frac{1}{b(n)}\right), & \text{for } t \in [0, 1], \\ \left( X\left(\frac{k+2-t}{b(n)}\right) - X\left(\frac{k}{b(n)}\right) \right) / \sigma\left(\frac{1}{b(n)}\right), & \text{for } t \in [1, 2], \end{cases} \\ & \quad k = 0, 1, \dots, b(n), \quad n = 1, 2, \dots, \end{aligned}$$

where  $N(n, k) = 2 \sum_{j=1}^{n-1} b(j) + 2k$ . Then, using (4), we have

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(3)  $\varphi^{-1}$  means the inverse function of  $\varphi$ .

$$\begin{aligned}
 & E | Y(N(n, k)+t) - Y(N(n, k)+s) |^2 \\
 &= \frac{\sigma^2((t-s)/b(n))}{\sigma^2(1/b(n))} \leq L^2 |t-s|^{2\beta}, \quad \text{for } t, s \in [0, 1].
 \end{aligned}$$

Hence

$$E | Y(N(n, k)+t) - Y(N(n, k)+s) |^2 \leq L^2 |t-s|^{2\beta}, \quad \text{for } t, s \in [0, 2].$$

Since,  $Y(2j)=0$ , for  $j=0, 1, 2, \dots$ , we get

$$(23) \quad E | Y(u) - Y(v) |^2 \leq 4L^2 |u-v|^{2\beta}$$

and

$$(24) \quad E | Y(u) |^2 \leq L^2.$$

Let  $\eta$  be a standard Gaussian variable which is independent to  $\{Y(u), u \geq 0\}$ . We shall define a Gaussian process  $Z$  by

$$Z(u) = Y(u) + \sqrt{1+L^2-EY^2(u)} \eta$$

and show that  $Z$  satisfies condition A.

$$\begin{aligned}
 (25) \quad & E(Z(u) - Z(v))^2 \\
 & \leq 4L^2 |u-v|^{2\beta} + (1+L^2-EY^2(v)) \left| \sqrt{1 + \frac{EY^2(v) - EY^2(u)}{1+L^2-EY^2(v)}} - 1 \right|^2.
 \end{aligned}$$

By (23) and (24), we have

$$|EY^2(u) - EY^2(v)| \leq 4L^2 |u-v|^\beta.$$

Hence, using the inequality  $|\sqrt{1+x}-1| \leq |x|$  for  $|x| \leq 1$ , we see that the second term of the right side of (25) is less than  $16L^4 |u-v|^{2\beta}$  for  $|u-v| \leq (4L^2)^{-1/\beta}$ . So, the process  $Z$  satisfies condition A. Therefore, by  $EZ(t)=0$  and  $EZ^2(t)=L^2+1$ , we have

$$\limsup_{T \uparrow \infty} \frac{\max_{u \in [0, T]} |Z(u)|}{\sqrt{L^2+1} \sqrt{2 \log T}} \leq 1, \quad \text{a.s.}$$

This implies that

$$\limsup_{T \uparrow \infty} \frac{\max_{u \in [0, T]} |Y(u)|}{\sqrt{L^2+1} \sqrt{2 \log T}} \leq 1, \quad \text{a.s.},$$

because the second component of  $Z$  is bounded in  $u$ , for almost all  $\omega$ . Recalling the definition of  $Y$ , we have

$$\limsup_{n \uparrow \infty} \frac{\max_{k=0, \dots, b(n)} \max_{t \in [0, T]} \left| X \left( \frac{k+t}{b(n)} \right) - X \left( \frac{k}{b(n)} \right) \right|}{\sqrt{2 \log N(n+1, 0)} \sigma \left( \frac{1}{b(n)} \right)} \leq \sqrt{L^2+1}, \quad \text{a.s.}$$

On the other hand,  $\frac{\log N(n+1, 0)}{\log 2^n}$  tends to  $1+\varepsilon$  when  $n$  tends to  $\infty$ . Therefore, for almost all  $\omega$ , there is  $n_0(\omega)$  such that, for any integer  $n$  greater than  $n_0(\omega)$ , the inequality

$$(26) \quad \max_{k=0, \dots, b(n), t \in [0, 1]} \left| X\left(\frac{k+t}{b(n)}, \omega\right) - X\left(\frac{k}{b(n)}, \omega\right) \right| \leq (1+2\varepsilon)\sqrt{L^2+1} \sigma\left(\frac{1}{b(n)}\right) \sqrt{2 \log 2^n}$$

holds. On the other hand, we have

$$\frac{\sigma\left(\frac{1}{b(n)}\right)}{\sigma(\tau)} \leq L2^{\beta-\beta\varepsilon n}, \quad \text{for } \tau \in [2^{-n-1}, 2^{-n}].$$

Moreover, for small positive  $\tau$ , we take integer  $n$  and  $i$  so that

$$2^{-n-1} < \tau \leq 2^{-n} \quad \text{and} \quad \frac{i}{b(n)} \leq \tau < \frac{i+1}{b(n)}.$$

Then, we have, by the concavity of  $\sigma^2$ ,

$$\frac{\sigma\left(\frac{i}{b(n)}\right)}{\sigma(\tau)} \leq 1,$$

and, for any positive  $s(<1-\tau)$ ,

$$\begin{aligned} |X(s+i)-X(s)| &\leq \max_{j=0, \dots, b(n)} \max_{l=1, \dots, a(n)} |\xi(n, j, l)| \sigma(i/b(n)) \\ &+ 2 \max_{k=0, \dots, b(n)} \max_{u \in [0, 1]} \left| X\left(\frac{k+u}{b(n)}\right) - X\left(\frac{k}{b(n)}\right) \right|. \end{aligned}$$

Therefore, appealing to (22) and (26), we see that, for almost all  $\omega$ ,

$$(27) \quad |X(s+i)-X(s)| \leq (1+2\varepsilon)\sigma(\tau)\sqrt{2 \log \frac{1}{\tau}}, \quad \text{for small } \tau.$$

We shall derive the converse inequality of (27) from the concavity of  $\sigma^2(t)$ . Define a separable Gaussian process  $Y$  by

$$\begin{aligned} Y(2^n+k+t) &= \frac{K((k+1)2^{-n})-X(k2^{-n})}{\sigma(2^{-n})}, \quad t \in [0, 1], \\ k &= 0, 1, \dots, 2^n-1, \quad n = 1, 2, \dots. \end{aligned}$$

Then, by the convexity of the covariance function of  $X$ , we have

$$\begin{aligned}
 & EY(2^l+k+t)Y(2^m+j+s) \\
 &= \frac{1}{\sigma(2^{-l})\sigma(2^{-m})} \{ \gamma(k2^{-l}-j2^{-m}) - \gamma(k2^{-l}-(j+1)2^{-m}) \\
 &\quad - \gamma((k+1)2^{-l}-j2^{-m}) + \gamma((k+1)2^{-l}+(j+1)2^{-m}) \} \leq 0, \\
 &\qquad\qquad\qquad \text{for } (j+1)2^{-m} \leq k2^{-l}.
 \end{aligned}$$

Hence  $Y$  satisfies condition B. So, for any  $\varepsilon > 0$  and for almost all  $\omega$ , there exists an integer  $n_0(\varepsilon, \omega)$  such that

$$\max_{m=1, \dots, n} \frac{\max_{k=0, \dots, 2^m-1} |X((k+1)2^{-m}, \omega) - X(k2^{-m}, \omega)|}{\sigma(2^{-m}) \sqrt{2 \log(2^{n+1}-2)}} > 1 - \varepsilon,$$

for  $n \geq n_0(\varepsilon, \omega)$ .

Hence, for any integer  $l$ ,

$$\sup_{m=l, l+1, \dots} \frac{\max_{k=0, \dots, 2^m-1} |X((k+1)2^{-m}) - X(k2^{-m})|}{\sigma(2^{-m}) \sqrt{2 \log 2^m}} > 1 - \varepsilon, \quad \text{a.s.}$$

Consequently, we have the following required inequality

$$\limsup_{h \downarrow 0} \sup_{t, s \in [0, 1], |t-s|=h} \frac{|X(t) - X(s)|}{\sigma(h) \sqrt{2 \log \frac{1}{h}}} > 1 - \varepsilon, \quad \text{a.s.}$$

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