

## A MODULUS OF CONTINUITY ON THE INCREMENTS OF A TWO-PARAMETER GAUSSIAN PROCESS

KYO-SHIN HWANG AND YONG-KAB CHOI

ABSTRACT. A modulus of continuity on the increments of a two-parameter Gaussian process is obtained via estimating large deviation probability inequalities on the suprema of the Gaussian process.

### 1. Introduction and results

Let  $\{X(x, y), 0 \leq x \leq 1, 0 \leq y \leq 1\}$  be an almost surely continuous, centered two-parameter Gaussian process with  $X(0, 0) = 0$ . For two distinct points  $(x_1, y_1), (x_2, y_2)$  in  $[0, 1] \times [0, 1]$ , we always assume that  $X(x, y)$  has

$$(i) \quad E\{X(x_1, y_1) - X(x_2, y_2)\}^2 = \sigma^2(\sqrt{(x_1 - x_2)^2 + (y_1 - y_2)^2}),$$

where  $\sigma(t), t > 0$ , is a nondecreasing continuous, regularly varying function with exponent  $\gamma$  at zero for some  $0 < \gamma < 1$ , that is,  $\sigma(t) = t^\gamma L(t)$  for some  $0 < \gamma < 1$ , where  $L(\cdot)$  is a slowly varying function at zero (i.e., it is measurable, positive and  $\lim_{t \rightarrow 0} L(\lambda t)/L(t) = 1$  for all  $\lambda > 0$ ). Note that the process  $\{X(x, y)\}$  is a generalization of the two-parameter Lévy Brownian motion. For our purpose, we assume that, for any  $t > 0$ , there exist constants  $c_1, c_2 > 0$  such that

$$(ii) \quad \frac{d\sigma^2(t)}{dt} \leq c_1 \frac{\sigma^2(t)}{t} \quad \text{and} \quad \frac{d^2\sigma^2(t)}{dt^2} \leq c_2 \frac{\sigma^2(t)}{t^2}.$$

---

Received January 19, 1998. Revised June 22, 1998.

1991 Mathematics Subject Classification: 60F15, 60G17.

Key words and phrases: Wiener process, Gaussian process, regularly varying function.

This work was supported by BSRI grant 98-1405, Ministry of Education, Korea.

For example, for some  $c_0 > 0$ , let  $\sigma(t) = c_0 t^\gamma$ ,  $0 < \gamma < 1$ . Then  $\{X(t_1, t_2), 0 \leq t_1 \leq 1, 0 \leq t_2 \leq 1\}$  is a two-dimensional time parameter fractional Brownian motion of order  $2\gamma$ , and our assumptions are satisfied.

Our main result is as follows:

**THEOREM 1 (Modulus of Continuity).** *Assume that the above Gaussian process  $\{X(x, y)\}$  satisfies conditions (i) and (ii). Let  $0 < h < 1$ . Then we have*

$$\lim_{h \rightarrow 0} \sup_{0 < s \leq h} \sup_{0 \leq x \leq 1-s} \sup_{0 \leq y \leq 1} \frac{|X(x+s, y) - X(x, y)|}{\sqrt{2\sigma^2(h) \log \left\{ \left( \frac{1-h}{h} \vee 1 \right) \frac{1}{h} \right\}}} = 1 \quad \text{a.s.}$$

and

$$\lim_{h \rightarrow 0} \sup_{0 \leq x \leq 1-h} \sup_{0 \leq y \leq 1} \frac{|X(x+h, y) - X(x, y)|}{\sqrt{2\sigma^2(h) \log \left\{ \left( \frac{1-h}{h} \vee 1 \right) \frac{1}{h} \right\}}} = 1 \quad \text{a.s.,}$$

where  $m \vee n = \max\{m, n\}$ .

## 2. Proofs

The proof of Theorem 1 is mainly based on the following Lemma 1, due to Fernique [2]: Let  $\mathbb{D} = \{\mathbf{t} : \mathbf{t} = (t_1, \dots, t_N), a_i \leq t_i \leq b_i, i = 1, 2, \dots, N\}$  be a real  $N$ -dimensional parameter space. We assume that the space  $\mathbb{D}$  has the usual Euclidean norm  $\|\cdot\|$  such that

$$\|\mathbf{t} - \mathbf{s}\|^2 = \sum_{i=1}^N (t_i - s_i)^2.$$

Let  $\{X(\mathbf{t}), \mathbf{t} \in \mathbb{D}\}$  be a real-valued separable Gaussian process with  $EX(\mathbf{t}) = 0$ . Suppose that

$$0 < \sup_{\mathbf{t} \in \mathbb{D}} E(X(\mathbf{t}))^2 =: \Gamma^2 < \infty, \quad \Gamma > 0,$$

and

$$E\{X(\mathbf{t}) - X(\mathbf{s})\}^2 \leq \varphi^2(\|\mathbf{t} - \mathbf{s}\|),$$

where  $\varphi(\cdot)$  is a nondecreasing continuous function such that

$$\int_0^\infty \varphi(e^{-y^2}) dy < \infty.$$

LEMMA 1. Let  $\{X(\mathbf{t}), \mathbf{t} \in \mathbb{D}\}$  be given as the above statements. Then, for  $\lambda > 0$ ,  $x \geq 1$  and  $\mathcal{A} > 2\sqrt{N \log 2}$ , we have

$$\begin{aligned} P\left\{\sup_{\mathbf{t} \in \mathbb{D}} X(\mathbf{t}) > x\left\{\Gamma + (2\sqrt{2} + 2)\mathcal{A} \int_0^\infty \varphi(\sqrt{N}\lambda 2^{-y^2}) dy\right\}\right\} \\ \leq (4^N + \psi) \prod_{i=1}^N \left(\frac{b_i - a_i}{\lambda} \vee 1\right) e^{-x^2/2}, \end{aligned}$$

where  $\psi = \sum_{n=1}^\infty \exp\left\{\frac{1}{2} - 2^n \left(\frac{\mathcal{A}^2}{2} - 2N \log 2\right)\right\} < \infty$ .

Let us estimate an upper bound of the following large deviation probabilities by using the above Lemma 1:

LEMMA 2. For any  $\epsilon > 0$ , there exists a positive constant  $C_\epsilon$  depending on  $\epsilon$  such that the inequality

$$\begin{aligned} P\left\{\sup_{0 < s \leq h} \sup_{0 \leq x \leq 1-s} \sup_{0 \leq y \leq 1} \frac{|X(x+s, y) - X(x, y)|}{\sigma(h)} > u\right\} \\ \leq C_\epsilon \left(\frac{1-s}{h} \vee 1\right) \left(\frac{1}{h}\right) e^{-u^2/(2+\epsilon)} \end{aligned}$$

holds for all  $u > 0$  and  $0 < h < 1$ .

PROOF. For  $0 < h < 1$ , let

$$\mathbb{D}_h = \{(s, x, y) : 0 < s \leq h, 0 \leq x \leq 1 - s, 0 \leq y \leq 1\}$$

be a three dimensional space. In order to apply Lemma 1, we set

$$Z(s, x, y) = \frac{X(x+s, y) - X(x, y)}{\sigma(h)}, \quad (s, x, y) \in \mathbb{D}_h,$$

and

$$\varphi(z) = \frac{2\sigma(\sqrt{2}z)}{\sigma(h)}, \quad z > 0.$$

Clearly,  $EZ(s, x, y) = 0$  for all  $(s, x, y) \in \mathbb{D}_h$ . Since  $\sigma(\cdot)$  is nondecreasing, it follows from the assumption (i) that

$$\Gamma^2 := \sup_{(s,x,y) \in \mathbb{D}_h} E(Z(s, x, y))^2 = 1.$$

Using the elementary inequality  $(a \pm b)^2 \leq 2(a^2 + b^2)$ , we have

$$\begin{aligned} & E\{Z(s_1, x_1, y_1) - Z(s_2, x_2, y_2)\}^2 \\ & \leq \frac{2}{\sigma^2(h)} \left\{ E\{X(x_1 + s_1, y_1) - X(x_2 + s_2, y_2)\}^2 \right. \\ & \qquad \qquad \qquad \left. + E\{X(x_1, y_1) - X(x_2, y_2)\}^2 \right\} \\ & \leq \frac{4}{\sigma^2(h)} \sigma^2 \left( \sqrt{2((x_1 - x_2)^2 + (s_1 - s_2)^2 + (y_1 - y_2)^2)} \right). \end{aligned}$$

Thus we get

$$\begin{aligned} & E\{Z(s_1, x_1, y_1) - Z(s_2, x_2, y_2)\}^2 \\ & \leq \varphi^2 \left( \sqrt{2((s_1 - s_2)^2 + (x_1 - x_2)^2 + (y_1 - y_2)^2)} \right). \end{aligned}$$

Since  $\sigma(\cdot)$  is a regularly varying function, we can show that for any  $\epsilon > 0$ , there exists a small  $c = c(\epsilon) > 0$  such that

$$(2\sqrt{2} + 2)\mathcal{A} \int_0^\infty \varphi(\sqrt{3}ch2^{-y^2}) dy < \epsilon/8$$

(cf. Lemma 4.5 in Choi [1]). Let  $u = \omega(1 + (\epsilon/8))$ ,  $\omega \geq 1$ , then we have, by Lemma 1,

$$\begin{aligned} & P \left\{ \sup_{0 < s \leq h} \sup_{0 \leq x \leq 1-s} \sup_{0 \leq y \leq 1} \frac{|X(x + s, y) - X(x, y)|}{\sigma(h)} > u \right\} \\ & \leq 2P \left\{ \sup_{(s,x,y) \in \mathbb{D}_h} Z(s, x, y) \right. \\ & \qquad \qquad \qquad \left. > \omega \left( 1 + (2\sqrt{2} + 2)\mathcal{A} \int_0^\infty \varphi(\sqrt{3}ch2^{-y^2}) dy \right) \right\} \\ & \leq C_\epsilon \left( \frac{1-s}{h} \vee 1 \right) \left( \frac{1}{h} \right) e^{-u^2/(2+\epsilon)}, \end{aligned}$$

where  $C_\epsilon$  is a positive constant depending on  $\epsilon > 0$ . In case  $0 < u \leq 1$ , the result is obvious if we take  $C_\epsilon$  large enough. □

The following lemma is easily obtained by a combined modification of Corollary 4.2.4 in Leadbetter et al. ([3], p. 84) and Lemma 4.4 in Choi ([1], p. 199):

LEMMA 3. Let  $\{Y_{ij}, i, j = 1, 2, \dots, n\}$  be jointly standardized normal random variables with covariance  $(Y_{ij}, Y_{i'j'}) = \Lambda_{ij}^{i'j'}$  such that

$$\delta := \max_{(i,j) \neq (i',j')} |\Lambda_{ij}^{i'j'}| < 1.$$

Let the subsequences  $\{l_i; i = 1, 2, \dots, k\}$  and  $\{l_j; j = 1, 2, \dots, h\}$  of integers be such that  $1 \leq l_1 < l_2 < \dots < l_k \leq n$  and  $1 \leq l_1 < l_2 < \dots < l_h \leq n$  with  $h, k \leq n$ . Denote  $r_{ij}^{i'j'} = \Lambda_{l_i l_j}^{l_{i'} l_{j'}}$ . Assume that

$$|r_{ij}^{i'j'}| < |i - i'|^{-\nu} |j - j'|^{-\nu} < 1, \quad i \neq i', j \neq j'$$

for some  $\nu > 0$ , and let  $u = \{(2 - \eta) \log(hk)\}^{1/2}$ , where  $0 < \eta < (1 - \delta)\nu / (1 + \nu + \delta)$ . Then for any real number  $u$  we have

$$P\left\{ \max_{1 \leq i \leq k} \max_{1 \leq j \leq h} Y_{l_i l_j} \leq u \right\} \leq \{\Phi(u)\}^{kh} + c(kh)^{-\delta_0},$$

where  $\delta_0 = \{\nu(1 - \delta) - \eta(1 + \delta + \nu)\} / \{(1 + \nu)(1 + \delta)\} > 0$  and  $c = c(\delta)$  is a constant independent of  $n, u, k$  and  $h$ , and  $\Phi(u) = \int_{-\infty}^u \frac{1}{\sqrt{2\pi}} e^{-y^2/2} dy$ .

The next lemma is easily verified by the same techniques as in the proof of Lemma 4.6 in Choi ([1], p. 204):

LEMMA 4. Let  $h$  be in  $0 < h_{n+1} \leq h < h_n < 1, n \in \mathbb{N}$ , where  $\mathbb{N}$  denotes the set of positive integers and  $h_n$  is decreasing. Then we have

$$\begin{aligned} & \liminf_{h \rightarrow 0} \sup_{0 \leq x \leq 1-h} \sup_{0 \leq y \leq 1} \frac{|X(x+h, y) - X(x, y)|}{\sqrt{2\sigma^2(h) \log\left\{\left(\frac{1-h}{h} \vee 1\right) \frac{1}{h}\right\}}} \\ & \geq \liminf_{n \rightarrow \infty} \sup_{0 \leq x \leq 1-h_n} \sup_{0 \leq y \leq 1} \frac{|X(x+h_n, y) - X(x, y)|}{\sqrt{2\sigma^2(h_n) \log\left\{\left(\frac{1-h_n}{h_n} \vee 1\right) \frac{1}{h_n}\right\}}} \quad \text{a.s.} \end{aligned}$$

Now we are ready to prove Theorem 1:

PROOF OF THEOREM 1. We first prove that

$$(2.1) \quad \limsup_{h \rightarrow 0} \sup_{0 < s \leq h} \sup_{0 \leq x \leq 1-s} \sup_{0 \leq y \leq 1} \frac{|X(x+s, y) - X(x, y)|}{\sqrt{2\sigma^2(h) \log \left\{ \left( \frac{1-h}{h} \vee 1 \right) \frac{1}{h} \right\}}} \leq 1 \quad \text{a.s.}$$

Let  $h$  be in  $h_{n+1} \leq h \leq h_n$ ,  $n \in \mathbb{N}$ , where  $h_n$  is decreasing and will be specified later on. Note that

$$(2.2) \quad \begin{aligned} & \sup_{0 < s \leq h} \sup_{0 \leq x \leq 1-s} \sup_{0 \leq y \leq 1} \frac{|X(x+s, y) - X(x, y)|}{\sqrt{2\sigma^2(h) \log \left\{ \left( \frac{1-h}{h} \vee 1 \right) \frac{1}{h} \right\}}} \\ & \leq \sup_{0 < s \leq h_n} \sup_{0 \leq x \leq 1-s} \sup_{0 \leq y \leq 1} \frac{|X(x+s, y) - X(x, y)|}{\sqrt{2\sigma^2(h_n) \log \left\{ \left( \frac{1-h_n}{h_n} \vee 1 \right) \frac{1}{h_n} \right\}}} \frac{\sigma(h_n)}{\sigma(h_{n+1})}. \end{aligned}$$

For any small  $\epsilon > 0$ , take  $T > (1 + \epsilon)/(2\epsilon)$ , and set  $h_n = (n + 1)^{-T}$ ,  $n \in \mathbb{N}$ . Then we have

$$(2.3) \quad \lim_{n \rightarrow \infty} \frac{\sigma(h_{n+1})}{\sigma(h_n)} = 1.$$

Indeed, since  $\sigma(\cdot)$  is regularly varying,

$$\begin{aligned} 1 & \geq \frac{\sigma(h_{n+1})}{\sigma(h_n)} = \frac{\sigma\left(\left(1 - \frac{1}{n+2}\right)^T h_n\right)}{\sigma(h_n)} \\ & = \left(1 - \frac{1}{n+2}\right)^{T\gamma} \frac{L\left(\left(1 - \frac{1}{n+2}\right)^T h_n\right)}{L(h_n)} \rightarrow 1 \end{aligned}$$

as  $n \rightarrow \infty$ . On the other hand, it follows from Lemma 2 that, for any  $\epsilon > 0$ ,

$$\begin{aligned} & P \left\{ \sup_{0 < s \leq h_n} \sup_{0 \leq x \leq 1-s} \sup_{0 \leq y \leq 1} \frac{|X(x+s, y) - X(x, y)|}{\sqrt{2\sigma^2(h_n) \log \left\{ \left( \frac{1-h_n}{h_n} \vee 1 \right) \frac{1}{h_n} \right\}}} > \sqrt{1 + \epsilon} \right\} \\ & \leq C_\epsilon \left( \frac{1-s}{h_n} \vee 1 \right) \frac{1}{h_n} \left\{ \left( \frac{1-h_n}{h_n} \vee 1 \right) \frac{1}{h_n} \right\}^{-2(1+\epsilon)/(2+\epsilon)} \\ & \leq C_\epsilon \left( \frac{1}{h_n} \right)^{-2\epsilon/(1+\epsilon)} = C_\epsilon (n+1)^{-T(2\epsilon/(1+\epsilon))}. \end{aligned}$$

Since  $T(2\epsilon/(1 + \epsilon)) > 1$  for given  $\epsilon > 0$ , the series

$$\sum_n P\left\{ \sup_{0 < s \leq h_n} \sup_{0 \leq x \leq 1-s} \sup_{0 \leq y \leq 1} \frac{|X(x+s, y) - X(x, y)|}{\sqrt{2\sigma^2(h_n) \log\left\{\left(\frac{1-h_n}{h_n} \vee 1\right) \frac{1}{h_n}\right\}}} > \sqrt{1 + \epsilon} \right\}$$

is convergent. The Borel-Cantelli lemma implies that

(2.4)

$$\limsup_{n \rightarrow \infty} \sup_{0 < s \leq h_n} \sup_{0 \leq x \leq 1-s} \sup_{0 \leq y \leq 1} \frac{|X(x+s, y) - X(x, y)|}{\sqrt{2\sigma^2(h_n) \log\left\{\left(\frac{1-h_n}{h_n} \vee 1\right) \frac{1}{h_n}\right\}}} \leq 1 \quad \text{a.s.}$$

Combining (2.2), (2.3) and (2.4), we obtain (2.1).

Next we prove that

(2.5)

$$\liminf_{h \rightarrow 0} \sup_{0 \leq x \leq 1-h} \sup_{0 \leq y \leq 1} \frac{|X(x+h, y) - X(x, y)|}{\sqrt{2\sigma^2(h) \log\left\{\left(\frac{1-h}{h} \vee 1\right) \frac{1}{h}\right\}}} \geq 1 \quad \text{a.s.}$$

It is obvious that there exists  $\beta > 0$  big enough such that

$$\left(\frac{1-h}{h} \vee 1\right) \frac{1}{h} \geq \left(\log \frac{1}{h}\right)^\beta$$

for  $h > 0$  sufficiently small. For such  $h$ , one can choose large integer  $N$  such that

$$\left(\log \frac{1}{h}\right)^\beta > N^2.$$

Define nonnegative integers  $\nabla_1(h)$  and  $\nabla_2(h)$  by

$$\nabla_1(h) = \left\lfloor \frac{1-h}{Nh} \right\rfloor, \quad \nabla_2(h) = \left\lfloor \frac{1}{Nh} \right\rfloor$$

for  $h > 0$  sufficiently small. Then there exists a constant  $c > 0$  such that

(2.6)

$$\nabla_1(h)\nabla_2(h) \geq c\left(\log \frac{1}{h}\right)^\beta$$

provided  $h > 0$  is small enough. For  $i = 0, 1, 2, \dots, \nabla_1(h)$  and  $j =$

$0, 1, 2, \dots, \nabla_2(h)$ , we also define

$$Z(i, j) = X(Nhi + h, Nhj) - X(Nhi, Nhj).$$

Then  $Z(i, j)/\sigma(h)$  are standard normal random variables. It follows that, for any  $0 < \epsilon' < \epsilon < 1$  and small  $h$ ,

$$\begin{aligned} (2.7) \quad & P \left\{ \sup_{0 \leq x \leq 1-h} \sup_{0 \leq y \leq 1} \frac{X(x+h, y) - X(x, y)}{\sigma(h)} \right. \\ & \left. < \sqrt{2(1-\epsilon) \log \left\{ \left( \frac{1-h}{h} \vee 1 \right) \frac{1}{h} \right\}} \right\} \\ & \leq P \left\{ \max_{0 \leq i \leq \nabla_1(h)} \max_{0 \leq j \leq \nabla_2(h)} \frac{Z(i, j)}{\sigma(h)} < \sqrt{2(1-\epsilon') \log(\nabla_1(h)\nabla_2(h))} \right\}. \end{aligned}$$

Let  $r(i, i', j, j') = \text{correlation}(Z(i, j), Z(i', j'))$ ,  $i \neq i'$ ,  $j \neq j'$ , and set  $l = |i - i'| \geq 1$ ,  $m = |j - j'| \geq 1$ . Using the relation  $ab = \frac{1}{2}(a^2 + b^2 - (a - b)^2)$ , it follows from the condition (ii) that

$$\begin{aligned} & |\text{covariance}(Z(i, j), Z(i', j'))| = |E\{Z(i, j)Z(i', j')\}| \\ & = |E\{X((Ni+1)h, Njh)X((Ni'+1)h, Nj'h) \\ & \quad - X((Ni+1)h, Njh)X(Ni'h, Nj'h) \\ & \quad - X(Nih, Njh)X((Ni'+1)h, Nj'h) \\ & \quad + X(Nih, Njh)X(Ni'h, Nj'h)\}| \\ & = \frac{1}{2} \left| \sigma^2 \left( \sqrt{(Nl+1)^2 h^2 + (Nm)^2 h^2} \right) - \sigma^2 \left( \sqrt{(Nl)^2 h^2 + (Nm)^2 h^2} \right) \right. \\ & \quad \left. - \left( \sigma^2 \left( \sqrt{(Nl)^2 h^2 + (Nm)^2 h^2} \right) - \sigma^2 \left( \sqrt{(Nl-1)^2 h^2 + (Nm)^2 h^2} \right) \right) \right| \\ & = \frac{1}{2} \left| \int_{\sqrt{(Nl)^2 h^2 + (Nm)^2 h^2}}^{\sqrt{(Nl+1)^2 h^2 + (Nm)^2 h^2}} d\sigma^2(x) - \int_{\sqrt{(Nl-1)^2 h^2 + (Nm)^2 h^2}}^{\sqrt{(Nl)^2 h^2 + (Nm)^2 h^2}} d\sigma^2(x) \right| \\ & \leq c \frac{\sigma^2 \left( \sqrt{(Nl+1)^2 h^2 + (Nm)^2 h^2} \right) h^2}{(Nl-1)^2 h^2 + (Nm)^2 h^2}, \end{aligned}$$



where  $c > 0$  is a constant. Hence we have, for any small  $h$ ,

$$\begin{aligned} |r(i, j, i', j')| &\leq c \frac{\sigma^2 \left( h \sqrt{(Nl + 1)^2 + (Nm)^2} \right)}{\sigma^2(h) \left( (Nl - 1)^2 + (Nm)^2 \right)} \\ &\leq c \frac{(Nl + 1)^2 + (Nm)^2}{(Nl - 1)^2 + (Nm)^2} \left( (Nl + 1)^2 + (Nm)^2 \right)^{-(1-\gamma)} \\ &\leq c \left( (Nl - 1)^2 + (Nm)^2 \right)^{-(1-\gamma)} < (l^2 + m^2)^{-(1-\gamma)} \\ &\leq (2lm)^{-(1-\gamma)} < (lm)^{-\nu}, \end{aligned}$$

where  $N$  is sufficiently large and  $0 < \nu = 1 - \gamma$ . To estimate the upper bound for the last inequality of (2.7), let us now apply Lemma 3 for

$$\begin{aligned} Y_{i,l_j} &= \frac{Z(i, j)}{\sigma(h)}, \quad i = 0, 1, 2, \dots, \nabla_1(h); \quad j = 0, 1, 2, \dots, \nabla_2(h), \\ |r_{ij}^{i'j'}| &= |r(i, j, i', j')| < (lm)^{-\nu}, \quad l = |i - i'| \geq 1, \quad m = |j - j'| \geq 1, \\ u = u(h) &= \{(2 - \eta) \log(\nabla_1(h)\nabla_2(h))\}^{1/2} \quad \text{and} \quad \eta = 2\epsilon' < \frac{(1 - \delta)\nu}{1 + \nu + \delta}. \end{aligned}$$

Then the last inequality of (2.7) is less than or equal to

$$\{\Phi(u(h))\}^{(\nabla_1(h)+1)(\nabla_2(h)+1)} + c(\nabla_1(h)\nabla_2(h))^{-\delta_0}$$

for some  $\delta_0 > 0$ . Thus we have

$$\begin{aligned} P \left\{ \sup_{0 \leq x \leq 1-h} \sup_{0 \leq y \leq 1} \frac{X(x+h, y) - X(x, y)}{\sqrt{2\sigma^2(h) \log\left\{ \left( \frac{1-h}{h} \vee 1 \right) \frac{1}{h} \right\}}} < \sqrt{1-\epsilon} \right\} \\ (2.8) \quad &\leq \exp(-c(\nabla_1(h)\nabla_2(h))^\epsilon) + c(\nabla_1(h)\nabla_2(h))^{-\delta_0} \\ &\leq c(\nabla_1(h)\nabla_2(h))^{-\delta_0}, \end{aligned}$$

where  $c$  is a constant, which may differ in lines. Let  $h_n = \exp(-n^a)$ ,  $0 < a < 1$ ,  $n \in \mathbb{N}$ . Note that the  $\beta$  in (2.6) can be taken sufficiently large so that  $\beta > 1/(a\delta_0)$ . Thus (2.8) implies that

$$\begin{aligned} P \left\{ \sup_{0 \leq x \leq 1-h_n} \sup_{0 \leq y \leq 1} \frac{X(x+h_n, y) - X(x, y)}{\sqrt{2\sigma^2(h_n) \log\left\{ \left( \frac{1-h_n}{h_n} \vee 1 \right) \frac{1}{h_n} \right\}}} < \sqrt{1-\epsilon} \right\} \\ \leq c \left( \log \frac{1}{h_n} \right)^{-\beta\delta_0} = cn^{-a\beta\delta_0} \end{aligned}$$

and

$$\sum_n P \left\{ \sup_{0 \leq x \leq 1-h_n} \sup_{0 \leq y \leq 1} \frac{X(x+h_n, y) - X(x, y)}{\sqrt{2\sigma^2(h_n) \log \left\{ \left( \frac{1-h_n}{h_n} \vee 1 \right) \frac{1}{h_n} \right\}}} < \sqrt{1-\epsilon} \right\} < \infty.$$

By the Borel-Cantelli lemma, we obtain

$$\liminf_{n \rightarrow \infty} \sup_{0 \leq x \leq 1-h_n} \sup_{0 \leq y \leq 1} \frac{X(x+h_n, y) - X(x, y)}{\sqrt{2\sigma^2(h_n) \log \left\{ \left( \frac{1-h_n}{h_n} \vee 1 \right) \frac{1}{h_n} \right\}}} \geq \sqrt{1-\epsilon} \quad \text{a.s.}$$

Let  $h$  be in  $h_{n+1} \leq h < h_n$ . It follows from Lemma 4 that the result (2.5) holds. The inequalities (2.1) and (2.5) complete the proof of Theorem 1.  $\square$

### References

- [1] Y. K. Choi, *Erdős-Rényi type laws applied to Gaussian processes*, J. Math. Kyoto Univ. **31** (1991), no. 1, 191–217.
- [2] X. Fernique, *Continuité des processus Gaussiens*, C. R. Acad. Sci. Paris t. **258** (1964), 6058–6060.
- [3] M. R. Leadbetter, G. Lindgren and H. Rootzen, *Extremes and Related Properties of Random Sequences and Processes*, Springer-Verlag, New York, 1983.

Department of Mathematics  
 College of Natural Science  
 Gyeongsang National University  
 Chinju 660-701, Korea  
*E-mail:* mathykc@nongae.gsnu.ac.kr