

## An Erdős–Révész type law of the iterated logarithm for stationary Gaussian processes<sup>\*</sup>

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Received October 11, 1991; in revised form March 23, 1992

**Summary.** Let  $\{X(t), t \geq 0\}$  be a stationary Gaussian process with  $EX(t) = 0$ ,  $EX^2(t) = 1$  and covariance function satisfying (i)  $r(t) = 1 - C|t|^\alpha + o(|t|^\alpha)$  as  $t \rightarrow 0$  for some  $C > 0$ ,  $0 < \alpha \leq 2$ ; (ii)  $r(t) = O(t^{-2\gamma})$  as  $t \rightarrow \infty$  for some  $\gamma > 0$  and (iii)  $\sup_{t \geq s} |r(t)| < 1$  for each  $s > 0$ . Put  $\xi(t) = \sup\{s: 0 \leq s \leq t, X(s) \geq (2 \log s)^{1/2}\}$ . The law of the iterated logarithm implies  $\limsup_{t \rightarrow \infty} (\xi(t) - t) = 0$  a.s. This paper gives the lower bound of  $\xi(t)$  and obtains an Erdős–Révész type LIL, i.e.,  $\liminf_{t \rightarrow \infty} (\xi(t) - t) / (t(\log t)^{(\alpha-2)/(2\alpha)} \log \log t) = -(2 + \alpha) \sqrt{\pi} / (\alpha H_\alpha (2C)^{1/\alpha})$  a.s. if  $0 < \alpha < 2$  and  $\liminf_{t \rightarrow \infty} \log(\xi(t)/t) / \log \log t = -2\pi / \sqrt{2C}$  a.s. if  $\alpha = 2$ . Applications to infinite series of independent Ornstein-Uhlenbeck processes and to fractional Wiener processes are also given.

*Mathematics Subject Classification (1991):* 60 G 15, 60 F 15

### 1 Introduction and main result

Let  $W(t)$ ,  $t \geq 0$  be a standard Wiener process and define  $\eta(t) = \sup\{s: 0 \leq s \leq t, W(s) \geq (2s \log \log s)^{1/2}\}$ ,  $t \geq 0$ . From the law of the iterated logarithm it follows immediately that

$$\limsup_{t \rightarrow \infty} \eta(t)/t = 1 \quad \text{a.s.}$$

Erdős and Révész [3] considered the lower bound of  $\eta(t)$  and obtained a new law of the iterated logarithm:

$$\liminf_{t \rightarrow \infty} \frac{(\log_2 t)^{1/2}}{\log_3 t \cdot \log t} \cdot \log \frac{\eta(t)}{t} = -C_0 \quad \text{a.s.} \tag{1.1}$$

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<sup>\*</sup> Research supported by the Fok Yingtung Education Foundation of China and by Charles Phelps Taft Postdoctoral Fellowship of the University of Cincinnati  
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for some constant  $C_0$  with  $\frac{1}{4} \leq C_0 \leq 2^{14}$ , while Shao [8] gave the exact value  $3\sqrt{\pi}$  of  $C_0$ , where  $\log_2 t = \log \log t$ ,  $\log_3 t = \log \log_2 t$ . Put

$$\bar{\eta}(t) = \sup \{s: 1 \leq s \leq t, W(s) \geq (2s \log \log s)^{1/2}\}, \quad \text{for } t \geq 1,$$

$$\hat{\eta}(t) = \sup \{s: 0 \leq s \leq t: \frac{W(e^s)}{e^{s/2}} \geq (2 \log s)^{1/2}\}, \quad \text{for } t \geq 0.$$

It is easy to see that

$$\hat{\eta}(t) = \log \bar{\eta}(e^t) \quad \text{a.s. for every } t \geq 0$$

and hence, by (1.1)

$$\liminf_{t \rightarrow \infty} \frac{(\log t)^{1/2}}{t \cdot \log_2 t} \cdot (\hat{\eta}(t) - t) = -3\sqrt{\pi} \quad \text{a.s.} \tag{1.2}$$

Clearly,  $W(e^s)/e^{s/2}$ ,  $s \geq 0$  is an Ornstein-Uhlenbeck process, a stationary Gaussian process. This promotes us to study the corresponding problem to  $\hat{\eta}(t)$  for general stationary Gaussian process.

Let  $\{X(t), t \geq 0\}$  be a real separable stationary Gaussian process with  $EX(t) = 0$  and  $EX^2(t) = 1$  for each  $t \geq 0$ . Denote the correlation function

$$r(t) = EX(t + s)X(s) \quad \text{for } s \geq 0 \text{ and } t \geq 0. \tag{1.3}$$

Consider the process

$$\xi(t) = \sup \{s: 0 \leq s \leq t, X(s) \geq (2 \log s)^{1/2}\}, \quad t \geq 0. \tag{1.4}$$

The upper class of law of the iterated logarithm implies

$$P(X(s) \geq (2 \log s)^{1/2}, \text{ i.o.}) = 1$$

under certain conditions on  $r(t)$  (cf. Qualls and Watanable [7]). Hence, we have

$$\lim_{t \rightarrow \infty} \xi(t) = \infty, \quad \text{a.s.}$$

and

$$\limsup_{t \rightarrow \infty} (\xi(t) - t) = 0 \quad \text{a.s.}$$

The aim of this paper is to give the lower bound of  $\xi(t)$ . We will state our main result in this section, while its proof is given in Sect. 2. Sections 3 and 4 will be devoted to two special Gaussian processes, infinite series of independent Ornstein-Uhlenbeck processes and fractional Wiener process, respectively.

Throughout this paper we will use the following notations:  $[x]$  denotes the integer part of  $x$ ;  $x^+ = \max(0, x)$ ;  $\log x = \ln x$  if  $x > 0$  and  $\log x = 1$  if  $x \leq 0$ , where  $\ln$  is the natural logarithm;  $\log_2 x = \log \log x$ , and  $\log_3 x = \log \log_2 x$ ;  $\sum_{i=x}^y$  stands for  $\sum_{i=[x]}^{[y]}$  and  $\max_{x \leq i \leq y}$  means  $\max_{[x] \leq i \leq [y]}$  for  $y \geq x$ .

In what follows we always assume that  $X(t)$  is a stationary Gaussian process with  $EX(t) = 0$ ,  $EX^2(t) = 1$  and correlation function  $r(t)$ .

**Theorem 1.1** Assume that the following conditions are satisfied:

$$r(t) = 1 - C|t|^\alpha + o(|t|^\alpha) \text{ as } t \rightarrow 0 \text{ for some } C > 0, \quad 0 < \alpha \leq 2, \quad (1.5)$$

$$r(t) = O(t^{-2\gamma}) \text{ as } t \rightarrow \infty \text{ for some } \gamma > 0, \quad (1.6)$$

$$\sup_{t \geq s} |r(t)| < 1 \text{ for each } s > 0. \quad (1.7)$$

Then

$$\liminf_{t \rightarrow \infty} \frac{\xi(t) - t}{t(\log t)^{(\alpha-2)/(2\alpha)} \cdot \log_2 t} = -\frac{(2 + \alpha)\sqrt{\pi}}{\alpha H_\alpha (2C)^{1/\alpha}} \text{ a.s. if } 0 < \alpha < 2 \quad (1.8)$$

$$\liminf_{t \rightarrow \infty} \frac{\log(\xi(t)/t)}{\log_2 t} = -\frac{2\sqrt{\pi}}{H_2\sqrt{2C}} \text{ a.s. if } \alpha = 2 \quad (1.9)$$

where  $0 < H_\alpha \equiv \lim_{T \rightarrow \infty} T^{-1} \int_0^\infty e^s P(\sup_{0 \leq t \leq T} Y(t) > s) ds < \infty$ , and  $Y(t)$  is a non-stationary Gaussian process with mean  $EY(t) = -|t|^\alpha$  and covariance function  $\text{Cov}(Y(s), Y(t)) = -|t - s|^\alpha + |s|^\alpha + |t|^\alpha$ .

The value of  $H_\alpha$  is unknown except  $H_1 = 1$  and  $H_2 = 1/\sqrt{\pi}$  (cf. [5, p. 232]). (1.8) shows that for any  $t$  big enough there exists an  $s$  in

$$[t - t(\log t)^{(\alpha-2)/(2\alpha)} \cdot \log_2 t, t]$$

such that  $X(s) \geq (2 \log s)^{1/2}$  and that the length of the interval  $t(\log t)^{(\alpha-2)/(2\alpha)} \log_2 t$  is smallest possible. Moreover, the bigger the parameter  $\alpha$  is, the wider the interval will be.

## 2 Proof

We start with some preliminary lemmas.

**Lemma 2.1** Suppose  $\xi_1, \dots, \xi_n$  are standard normal variables with covariance matrix  $A^1 = (A_{ij}^1)$  and  $\eta_1, \dots, \eta_n$  similarly with covariance matrix  $A^0 = (A_{ij}^0)$ , and let  $\rho_{ij} = \max(|A_{ij}^1|, |A_{ij}^0|)$ . Further, let  $u_1, \dots, u_n$  be real numbers. Then

$$\begin{aligned} & P\left(\bigcap_{j=1}^n \{\xi_j \leq u_j\}\right) - P\left(\bigcap_{j=1}^n \{\eta_j \leq u_j\}\right) \\ & \leq \frac{1}{2\pi} \sum_{1 \leq i < j \leq n} (A_{ij}^1 - A_{ij}^0)^+ (1 - \rho_{ij}^2)^{-1/2} \exp\left(-\frac{u_i^2 + u_j^2}{2(1 + \rho_{ij})}\right). \end{aligned}$$

This is Theorem 4.2.1 of [5].

**Lemma 2.2** If (1.5) and (1.7) hold, then

$$\lim_{x \rightarrow \infty} \frac{P(\sup_{0 \leq s \leq 1} X(s) > x)}{x^{2/\alpha} \psi(x)} = C^{1/\alpha} H_\alpha \quad (2.1)$$

$$\lim_{x \rightarrow \infty} \frac{P(\max_{0 \leq j \leq x^{2/\alpha}/\theta} X(j\theta x^{-2/\alpha}) > x)}{x^{2/\alpha} \psi(x)} = C^{1/\alpha} \frac{H_\alpha(\theta)}{\theta} \quad (2.2)$$

for each  $\theta > 0$  and

$$\lim_{\theta \rightarrow 0} \frac{H_\alpha(\theta)}{\theta} = H_x \tag{2.3}$$

where  $H_x$  is defined as in Theorem 1.1 and  $\psi(x) = (2\pi)^{-1/2} x^{-1} e^{-x^2/2}$ .

This is Lemma 2.5 of [6].

From the proof of Lemma 12.2.5 of [5], one can see that the next lemma holds.

**Lemma 2.3** *If the conditions of Theorem 1.1 are satisfied, then there exist constant  $K_0$  and  $x_0$  such that*

$$\begin{aligned} P\left(\max_{0 \leq j \leq x^{2/\alpha}/\theta} X(j\theta x^{-2/\alpha}) \leq x - \theta^{\alpha/4}/x, \sup_{0 \leq s \leq 1} X(s) > x\right) \\ \leq K_0 x^{2/\alpha} \psi(x) \theta^{\frac{\alpha}{2}-1} \exp(-\theta^{-\alpha/4}/K_0) \end{aligned} \tag{2.4}$$

for each  $\theta > 0$  and  $x \geq x_0$ .

In what follows we define

$$\tilde{\alpha} = (2 + \alpha)/(2\alpha), \quad \hat{\alpha} = (2 - \alpha)/(2\alpha) \quad \text{for } 0 < \alpha \leq 2. \tag{2.5}$$

**Lemma 2.4** *Under the condition of Theorem 1.1, for each  $0 < \varepsilon < 1$ , there exist positive constants  $N$  and  $\rho$  depending only on  $\varepsilon, \alpha$  and  $\gamma$  such that*

$$\begin{aligned} P\left(\sup_{a \leq s \leq b} \frac{X(s)}{(2 \log s)^{1/2}} \leq 1\right) \\ \leq \exp\left(-\frac{(1 - \varepsilon)\alpha(2C)^{1/\alpha} H_\alpha(\log \tilde{\alpha} b - \log \tilde{\alpha}(a + 1))}{(1 + \varepsilon)(2 + \alpha)\sqrt{\pi}}\right) + Na^{-\rho} \end{aligned} \tag{2.6}$$

for each  $b \geq a + 1 \geq N$ , where  $\tilde{\alpha}$  is defined as in (2.5).

*Proof.* By Lemma 2.2, there exist  $\theta > 0$  and  $x_0$  such that

$$P\left(\max_{0 \leq j \leq x^{2/\alpha}/\theta} X(j\theta x^{-2/\alpha}) \geq x\right) \geq (1 - \varepsilon)H_\alpha C^{1/\alpha} x^{2/\alpha} \psi(x) \tag{2.7}$$

for each  $x \geq x_0$ . Put

$$\begin{aligned} b(a, \varepsilon) &= [(b - a - 1)/(1 + \varepsilon)], \\ x_i &= (2 \log(a + 1 + i(1 + \varepsilon)))^{1/2}, \quad \hat{x}_i = [x_i^{2/\alpha}/\theta], \quad i = 0, 1, \dots \end{aligned}$$

Then

$$\begin{aligned} P\left(\sup_{a \leq s \leq b} \frac{X(s)}{(2 \log s)^{1/2}} \leq 1\right) \\ \leq P\left(\max_{0 \leq i \leq (b-a-1)/(1+\varepsilon)} \sup_{a+i(1+\varepsilon) \leq s \leq a+1+i(1+\varepsilon)} \frac{X(s)}{(2 \log s)^{1/2}} \leq 1\right) \\ \leq P\left(\max_{0 \leq i \leq b(a, \varepsilon)} \sup_{a+i(1+\varepsilon) \leq s \leq a+1+i(1+\varepsilon)} \frac{X(s)}{(2 \log(a + 1 + i(1 + \varepsilon)))^{1/2}} \leq 1\right) \\ \leq P\left(\max_{0 \leq i \leq b(a, \varepsilon)} \max_{0 \leq j \leq \hat{x}_i} \frac{X(a + i(1 + \varepsilon) + j\theta x_i^{-2/\alpha})}{x_i} \leq 1\right). \end{aligned} \tag{2.8}$$

Let

$$\mathbf{X}_i = (X(a + i(1 + \varepsilon) + j\theta x_i^{-2/\alpha}), \quad 0 \leq j \leq \hat{x}_i), \quad i = 0, 1, \dots, b(a, \varepsilon),$$

$\{\mathbf{Y}_i, 0 \leq i \leq b(a, \varepsilon)\}$  be independent normal random vectors and  $\mathbf{Y}_i$  and  $\mathbf{X}_i$  have the same distribution for each  $0 \leq i \leq b(a, \varepsilon)$ . Applying Lemma 2.1 yields

$$\begin{aligned} & P\left(\max_{0 \leq i \leq b(a, \varepsilon)} \max_{0 \leq j \leq \hat{x}_i} \frac{X(a + i(1 + \varepsilon) + j\theta x_i^{-2/\alpha})}{x_i} \leq 1\right) \\ & \leq \prod_{i=0}^{b(a, \varepsilon)} P\left(\max_{0 \leq j \leq \hat{x}_i} X(a + i(1 + \varepsilon) + j\theta x_i^{-2/\alpha}) \leq x_i\right) \\ & \quad + \sum_{0 \leq i < j \leq b(a, \varepsilon)} \sum_{u=0}^{\hat{x}_i} \sum_{v=0}^{\hat{x}_j} \frac{|r(i, j, u, v)|}{\sqrt{1 - r^2(i, j, u, v)}} \exp\left(-\frac{\frac{1}{2}(x_i^2 + x_j^2)}{1 + |r(i, j, u, v)|}\right) \\ & := I_1 + I_2 \end{aligned} \tag{2.9}$$

where

$$r(i, j, u, v) = r((j - i)(1 + \varepsilon) + v\theta x_j^{-2/\alpha} - u\theta x_i^{-2/\alpha}),$$

$$I_1 = \prod_{i=0}^{b(a, \varepsilon)} P\left(\max_{0 \leq j \leq \hat{x}_i} X(a + i(1 + \varepsilon) + j\theta x_i^{-2/\alpha}) \leq x_i\right),$$

$$I_2 = \sum_{0 \leq i < j \leq b(a, \varepsilon)} \sum_{u=0}^{\hat{x}_i} \sum_{v=0}^{\hat{x}_j} \frac{|r(i, j, u, v)|}{\sqrt{1 - r^2(i, j, u, v)}} \exp\left(-\frac{\frac{1}{2}(x_i^2 + x_j^2)}{1 + |r(i, j, u, v)|}\right).$$

Put

$$r^*(s) = \sup_{t \geq s} |r(t)|, \quad s > 0. \tag{2.10}$$

Noting that

$$(j - i)(1 + \varepsilon) + v\theta x_j^{-2/\alpha} - u\theta x_i^{-2/\alpha} \geq (j - i)\varepsilon \geq \varepsilon$$

for every  $j > i, 0 \leq u \leq \hat{x}_i, 0 \leq v \leq \hat{x}_j$ , we have

$$|r(i, j, u, v)| \leq r^*((j - i)\varepsilon) \leq r^*(\varepsilon) < 1 \tag{2.11}$$

by (1.7). From (1.6) it follows that there is a  $t_0$  such that

$$r^*(t) \leq t^{-\gamma} \leq \min(1, \gamma)/4 \text{ for every } t \geq t_0. \tag{2.12}$$

In what follows, for the sake of simplicity of statement, we will use  $K$  to denote the constant which is independent of  $a$  and  $b$ , and may be different from line to line. We have

$$\begin{aligned} I_2 & \leq \frac{4}{\theta^2 \sqrt{1 - r^*(\varepsilon)}} \sum_{0 \leq i < j \leq b(a, \varepsilon)} x_i^{2/\alpha} x_j^{2/\alpha} r^*((j - i)\varepsilon) \exp\left(-\frac{x_i^2 + x_j^2}{2(1 + r^*((j - i)\varepsilon))}\right) \\ & = \frac{4}{\theta^2 \sqrt{1 - r^*(\varepsilon)}} \left( \sum_{0 < j - i \leq t_0/\varepsilon} + \sum_{j - i > t_0/\varepsilon} \right) (\cdot) \end{aligned}$$

$$\begin{aligned}
 &\leq K \left( \sum_{0 \leq i \leq b(a, \varepsilon)} x_i^{4/\alpha} \exp\left(-\frac{x_i^2}{1+r^*(\varepsilon)}\right) \right. \\
 &\quad \left. + \sum_{j-i > t_0/\varepsilon, j \leq b(a, \varepsilon)} x_i^{2/\alpha} x_j^{2/\alpha} (j-i)^{-\gamma} \exp\left(-\frac{x_i^2 + x_j^2}{2(1+\gamma/4)}\right) \right) \\
 &\leq K \left( \sum_{i=0}^{\infty} (a+i(1+\varepsilon))^{-2/(1+r^*(\varepsilon))} \log^{2/\alpha}(a+i) \right. \\
 &\quad \left. + \sum_{j-i > t_0/\varepsilon, j \leq b(a, \varepsilon)} (a+i)^{-\frac{1}{1+\gamma/4}} (a+j)^{-\frac{1}{1+\gamma/4}} \log^{1/\alpha}(a+i) \log^{1/\alpha}(a+j) \right) \\
 &\leq K \left( a^{-\frac{1-r^*(\varepsilon)}{4}} + \sum_{i=0}^{\infty} \frac{\log^{1/\alpha}(a+i)}{(a+i)^{1+\gamma/4}} \right) \\
 &\leq K(a^{-(1-r^*(\varepsilon))/4} + a^{-\gamma/6}) \\
 &\leq K a^{-\rho}
 \end{aligned}$$

where  $\rho = \min((1 - r^*(\varepsilon))/4, \gamma/6)$ .

Noting that  $X(\cdot)$  is a stationary process, we derive from (2.7) that

$$\begin{aligned}
 I_1 &= \prod_{i=0}^{b(a, \varepsilon)} P\left(\max_{0 \leq j \leq \hat{x}_i} X(j\theta x_i^{-2/\alpha}) \leq x_i\right) \\
 &\leq \exp\left(-\sum_{i=0}^{b(a, \varepsilon)} P\left(\max_{0 \leq j \leq \hat{x}_i} X(j\theta x_i^{-2/\alpha}) > x_i\right)\right) \\
 &\leq \exp\left(-\sum_{i=0}^{b(a, \varepsilon)} (1-\varepsilon) C^{1/\alpha} H_\alpha x_i^{2/\alpha} \psi(x_i)\right) \\
 &= \exp\left(- (1-\varepsilon) C^{1/\alpha} H_\alpha \sum_{i=0}^{b(a, \varepsilon)} \frac{(2\log(a+1+i(1+\varepsilon)))^2}{\sqrt{2\pi}(a+1+i(1+\varepsilon))}\right) \\
 &= \exp\left(-\frac{(1-\varepsilon)(2C)^{1/\alpha} H_\alpha}{2\sqrt{\pi}} \sum_{i=0}^{b(a, \varepsilon)} \frac{\log^{\hat{\alpha}}(a+1+i(1+\varepsilon))}{a+1+i(1+\varepsilon)}\right)
 \end{aligned}$$

provided that  $a$  is sufficiently large, where  $\hat{\alpha}$  is defined as in (2.5). An elementary calculation implies

$$\begin{aligned}
 \sum_{i=0}^{b(a, \varepsilon)} \frac{\log^{\hat{\alpha}}(a+1+i(1+\varepsilon))}{a+1+i(1+\varepsilon)} &\geq \int_0^{b-a-1} \frac{\log^{\hat{\alpha}}(a+1+y(1+\varepsilon))}{a+1+y(1+\varepsilon)} dy \\
 &= \frac{2\alpha}{(1+\varepsilon)(2+\alpha)} (\log^{\hat{\alpha}} b - \log^{\hat{\alpha}}(a+1)).
 \end{aligned}$$

Hence

$$I_1 \leq \exp\left(-\frac{(1-\varepsilon)\alpha(2C)^{1/\alpha} H_\alpha}{(1+\varepsilon)(2+\alpha)\sqrt{\pi}} (\log^{\hat{\alpha}} b - \log^{\hat{\alpha}}(a+1))\right). \tag{2.15}$$

Putting the above inequalities together yields that (2.6) holds true. This proves the lemma.

**Lemma 2.5** *Under the condition of Theorem 1.1, for each  $0 < \varepsilon < 1$ , there exist positive constants  $N$  and  $\tau$  depending only on  $\varepsilon$ ,  $\alpha$  and  $\gamma$  such that*

$$\begin{aligned} & P\left(\bigcap_{0 \leq i \leq [b-a]} \left\{ \max_{0 \leq j \leq y_i^{2/\alpha}/\theta_i} X(a+i+j\theta_i y_i^{-2/\alpha}) < y_i - \theta_i^{\alpha/4}/y_i \right\}\right) \\ & \geq \frac{1}{4} \exp\left(-\frac{(1+\varepsilon)\alpha(2C)^{1/\alpha} H_\alpha (\log^2 b - \log^2 a)}{(2+\alpha)\sqrt{\pi}}\right) - Na^{-\tau} \end{aligned} \quad (2.16)$$

for each  $b \geq a+1 \geq N$ , where  $y_i = (2\log(a+i))^{1/2}$ ,  $\theta_i = \log^{-8/\alpha}(a+i)$ ,  $\tilde{a}$  is as in (2.5).

*Proof.* Put

$$\bar{y}_i = y_i - \theta_i^{\alpha/4}/y_i, \quad \hat{y}_i = [y_i^{2/\alpha}/\theta_i], \quad i = 0, 1, \dots$$

Applying Lemma 2.1, one can find that

$$\begin{aligned} & P\left(\bigcap_{0 \leq i \leq [b-a]} \left\{ \max_{0 \leq j \leq y_i^{2/\alpha}/\theta_i} X(a+i+j\theta_i y_i^{-2/\alpha}) < y_i - \theta_i^{\alpha/4}/y_i \right\}\right) \\ & \geq \prod_{i=0}^{[b-a]} P\left(\max_{0 \leq j \leq \hat{y}_i} X(a+i+j\theta_i y_i^{-2/\alpha}) < \bar{y}_i\right) \\ & \quad - \frac{1}{2\pi} \sum_{0 \leq i < j \leq b-a} \sum_{u=0}^{\hat{y}_i} \sum_{v=0}^{\hat{y}_j} \frac{(\tau(i, j, u, v))^+}{\sqrt{1-\tau^2(i, j, u, v)}} \exp\left(-\frac{\frac{1}{2}(\bar{y}_i^2 + \bar{y}_j^2)}{1+|\tau(i, j, u, v)|}\right) \\ & := J_1 - J_2, \end{aligned} \quad (2.17)$$

where  $\tau(i, j, u, v) = -r(j-i+v\theta_j y_j^{-2/\alpha} - u\theta_i y_i^{-2/\alpha})$ .

Clearly, for  $j \geq i+2$ ,  $0 \leq u \leq \hat{y}_i$  and  $0 \leq v \leq \hat{y}_j$ , by (1.7)

$$|\tau(i, j, u, v)| \leq r^*(j-i-1) \leq r^*(1) < 1. \quad (2.18)$$

On the other hand, by (1.5), there exists a constant  $0 < t_1 < 1$  such that

$$r(t) \geq 1 - C|t|^\alpha/2 > 0 \text{ for every } 0 \leq t \leq t_1.$$

Hence

$$(\tau(i, j, u, v))^+ = 0, \quad \text{if } j = i+1, 1 + v\theta_j y_j^{-2/\alpha} - u\theta_i y_i^{-2/\alpha} \leq t_1 \quad (2.19)$$

and

$$|\tau(i, j, u, v)| \leq r^*(t_1) < 1 \quad \text{if } j = i+1, 1 + v\theta_j y_j^{-2/\alpha} - u\theta_i y_i^{-2/\alpha} > t_1. \quad (2.20)$$

Therefore, by (2.18), (2.19) and (2.20) we obtain

$$\begin{aligned} J_2 & \leq \sum_{0 \leq i \leq b-a, j=i+1} \sum_{u=0}^{\hat{y}_i} \sum_{v=0}^{\hat{y}_j} \frac{1}{\sqrt{1-r^*(t_1)}} \exp\left(-\frac{\frac{1}{2}(\bar{y}_i^2 + \bar{y}_j^2)}{1+r^*(t_1)}\right) \\ & \quad + \sum_{0 \leq i+2 \leq j \leq b-a} \sum_{u=0}^{\hat{y}_i} \sum_{v=0}^{\hat{y}_j} \frac{r^*(j-i-1)}{\sqrt{1-r^*(1)}} \exp\left(-\frac{\frac{1}{2}(\bar{y}_i^2 + \bar{y}_j^2)}{1+r^*(j-i-1)}\right). \end{aligned}$$

Completely similar to the estimation of  $I_2$ , we can arrive that there exist positive constants  $K$  and  $\tau$  such that

$$J_2 \leq K a^{-\tau} \tag{2.21}$$

for every  $a$  sufficiently large. Using Lemma 2.2, one can also obtain that

$$\begin{aligned} J_1 &= \prod_{i=0}^{\lfloor b-a \rfloor} P\left(\max_{0 \leq j \leq \hat{y}_i} X(j\theta_i y_i^{-2/\alpha}) < \bar{y}_i\right) \\ &\geq \prod_{i=0}^{\lfloor b-a \rfloor} \left(1 - P\left(\sup_{0 \leq s \leq 1} X(s) \geq \bar{y}_i\right)\right) \\ &\geq \frac{1}{4} \exp\left(-\frac{(1+\varepsilon)\alpha(2C)^{1/\alpha} H_\alpha}{(2+\alpha)\sqrt{\pi}} \left(\log^{\tilde{\alpha}} b - \log^{\tilde{\alpha}} a\right)\right) \end{aligned} \tag{2.22}$$

provided that  $a$  is sufficiently large, along the same line of the proof of  $I_1$  in Lemma 2.4. This proves (2.16), by (2.17), (2.21) and (2.22).

*Proof of Theorem 1.1* We formulate the proof in three steps.

*Step 1* Assume  $0 < \alpha < 2$ . Then

$$\liminf_{t \rightarrow \infty} \frac{\zeta(t) - t}{t \log^{-\hat{\alpha}} t \cdot \log_2 t} \geq -(1 + 2\varepsilon)^2 C_1 \quad \text{a.s.} \tag{2.23}$$

for every  $0 < \varepsilon < \frac{1}{4}$ , where  $C_1 = (2 + \alpha)\sqrt{\pi}/(\alpha H_\alpha(2C)^{1/\alpha})$ ,  $\hat{\alpha}$  is as in (2.5).

*Proof.* Put

$$t_k = \exp(k^{1/\tilde{\alpha}}), \quad s_k = t_k - (1 + 2\varepsilon)C_1 t_k \log^{-\hat{\alpha}} t_k \cdot \log_2 t_k, \quad k = 1, 2, \dots$$

Then

$$\begin{aligned} \log^{\tilde{\alpha}} t_k - \log^{\tilde{\alpha}} s_k &\sim \log^{\tilde{\alpha}} t_k - (\log t_k - (1 + 2\varepsilon)^2 C_1 (\log t_k)^{-\hat{\alpha}} \cdot \log_2 t_k)^{\tilde{\alpha}} \\ &\sim \frac{(1 + 2\varepsilon)^2 (2 + \alpha) C_1}{2\alpha} \log_2 t_k. \end{aligned} \tag{2.24}$$

Noting that

$$\{\zeta(t) \leq a\} = \left\{ \sup_{a < s \leq t} \frac{X(s)}{(2 \log s)^{1/2}} < 1 \right\}$$

for every  $0 < a < t$ , and using Lemma 2.4 and (2.24), we obtain that

$$\begin{aligned} &P\left(\frac{\zeta(t_k) - t_k}{t_k \log^{-\hat{\alpha}} t_k \cdot \log_2 t_k} \leq -(1 + 2\varepsilon)C_1\right) \\ &= P\left(\sup_{s_k < s \leq t_k} \frac{X(s)}{(2 \log s)^{1/2}} < 1\right) \\ &\leq \exp\left(-\frac{(1 - \varepsilon)\alpha(2C)^{1/\alpha} H_\alpha}{(1 + \varepsilon)(2 + \alpha)\sqrt{\pi}} (\log^{\tilde{\alpha}} t_k - \log^{\tilde{\alpha}}(s_k + 1))\right) + N s_k^{-\rho} \end{aligned}$$



$$\begin{aligned} &\leq \exp\left(-\frac{(1-\varepsilon)(2C)^{1/\alpha}(1+2\varepsilon)^2 C_1 \log_2 t_k}{(1+2\varepsilon)2\sqrt{\pi}}\right) + 2Nt_k^{-\rho} \\ &\leq \exp\left(-\frac{(1+\varepsilon/2)(2+\alpha)\log_2 t_k}{2\alpha}\right) + 2Nt_k^{-\rho} \\ &\leq 2k^{-(1+\varepsilon/2)} \end{aligned}$$

for every  $k$  sufficiently large. Hence, by the Borel-Cantelli lemma, we have

$$\liminf_{k \rightarrow \infty} \frac{\xi(t_k) - t_k}{t_k \log^{-\hat{\alpha}} t_k \cdot \log_2 t_k} \geq -(1+2\varepsilon)^2 C_1 \quad \text{a.s.} \tag{2.25}$$

Since  $\xi(t)$  is a non-decreasing random function of  $t$ , for every  $t_k \leq t \leq t_{k+1}$ , we have

$$\begin{aligned} \frac{\xi(t) - t}{t \log^{-\hat{\alpha}} t \cdot \log_2 t} &\geq \frac{\xi(t_k) - t_{k+1}}{t_k \log^{-\hat{\alpha}} t_k \cdot \log_2 t_k} \\ &= \frac{\xi(t_k) - t_k}{t_k \log^{-\hat{\alpha}} t_k \cdot \log_2 t_k} - \frac{t_{k+1} - t_k}{t_k \log^{-\hat{\alpha}} t_k \cdot \log_2 t_k}. \end{aligned} \tag{2.26}$$

An elementary calculation implies

$$\lim_{k \rightarrow \infty} \frac{t_{k+1} - t_k}{t_k \log^{-\hat{\alpha}} t_k \cdot \log_2 t_k} = 0 \tag{2.27}$$

which together with (2.26) yields

$$\liminf_{t \rightarrow \infty} \frac{\xi(t) - t}{t \log^{-\hat{\alpha}} t \cdot \log_2 t} = \liminf_{k \rightarrow \infty} \frac{\xi(t_k) - t_k}{t_k \log^{-\hat{\alpha}} t_k \cdot \log_2 t_k} \quad \text{a.s.} \tag{2.28}$$

This proves (2.23), by (2.25) and (2.28).

*Step 2* Assume  $0 < \alpha < 2$ . Then

$$\liminf_{t \rightarrow \infty} \frac{\xi(t) - t}{t \log^{-\hat{\alpha}} t \cdot \log_2 t} \leq -(1-\varepsilon)C_1 \quad \text{a.s.} \tag{2.29}$$

for every  $0 < \varepsilon < (2-\alpha)/8 < \frac{1}{4}$ .

*Proof.* Let

$$\begin{aligned} b_k &= \exp(k^{(1+\varepsilon^2)/\hat{\alpha}}), \quad a_k = b_k - (1-2\varepsilon)C_1 b_k \log^{-\hat{\alpha}} b_k \cdot \log_2 b_k, \\ y_{k,i} &= (2\log(a_k + i))^{1/2}, \quad \theta_{k,i} = \log^{-8/\alpha}(a_k + i), \\ \bar{y}_{k,i} &= y_{k,i} - \theta_{k,i}^{3/4}/y_{k,i}, \quad \hat{y}_{k,i} = y_{k,i}^{2/\alpha}/\theta_{k,i}, \\ E_k &= \left\{ \xi(b_k) \leq a_k \right\} = \left\{ \sup_{a_k < s \leq b_k} \frac{X(s)}{(2\log s)^{1/2}} < 1 \right\}, \\ A_k &= \bigcap_{0 \leq i \leq [b_k - a_k]} \left\{ \max_{0 \leq j \leq \hat{y}_{k,i}} X(a_k + i + j\theta_{k,i} y_{k,i}^{-2/\alpha}) \leq \bar{y}_{k,i} \right\}. \end{aligned}$$

It suffices to show that

$$P(E_n \text{ i.o.}) = 1. \quad (2.30)$$

Clearly, for  $m \geq 1$

$$\begin{aligned} & P\left(\bigcup_{k=m}^{\infty} A_k\right) \leq P\left(\bigcup_{k=m}^{\infty} E_k\right) + \sum_{k=m}^{\infty} P(A_k E_k^c) \leq P\left(\bigcup_{k=m}^{\infty} E_k\right) \\ & + \sum_{k=m}^{\infty} \sum_{i=0}^{b_k - a_k} P\left(\max_{0 \leq j \leq \hat{y}_{k,i}} X(a_k + i + j\theta_{k,i} y_{k,i}^{-2/\alpha}) \leq \bar{y}_{k,i}, \sup_{0 \leq s \leq 1} X(a_k + i + s) \geq y_{k,i}\right) \\ & = P\left(\bigcup_{k=m}^{\infty} E_k\right) + \sum_{k=m}^{\infty} \sum_{i=0}^{b_k - a_k} P\left(\max_{0 \leq j \leq \hat{y}_{k,i}} X(j\theta_{k,i} y_{k,i}^{-2/\alpha}) \leq \bar{y}_{k,i}, \sup_{0 \leq s \leq 1} X(s) \geq y_{k,i}\right) \\ & := P\left(\bigcup_{k=m}^{\infty} E_k\right) + L_m. \end{aligned} \quad (2.31)$$

Using Lemma 2.3, we have, for some  $K_0 > 0$

$$\begin{aligned} L_m & \leq K_0 \sum_{k=m}^{\infty} \sum_{i=0}^{b_k - a_k} y_{k,i}^{2/\alpha} \psi(y_{k,i}) \theta_{k,i}^{\alpha/2 - 1} \exp\left(-\frac{\theta_{k,i}^{-\alpha/4}}{K_0}\right) \\ & \leq K \sum_{k=m}^{\infty} \sum_{i=0}^{b_k - a_k} \log^{9\hat{\alpha}}(a_k + i) \cdot (a_k + i)^{-1} \exp\left(-\frac{\log^2(a_k + i)}{K_0}\right) \\ & \leq K \sum_{k=m}^{\infty} \sum_{i=0}^{\infty} \log^{9\hat{\alpha}}(a_k + i) \cdot (a_k + i)^{-3} \\ & \leq K \sum_{k=m}^{\infty} a_k^{-1} \\ & \leq Km^{-4} \end{aligned}$$

provided  $m$  is large enough. Therefore

$$\lim_{m \rightarrow \infty} L_m = 0$$

and

$$\lim_{m \rightarrow \infty} P\left(\bigcup_{k=m}^{\infty} E_k\right) \geq \lim_{m \rightarrow \infty} P\left(\bigcup_{k=m}^{\infty} A_k\right).$$

To finish the proof of (2.30), we only need to show that

$$P(A_n \text{ i.o.}) = 1. \quad (2.32)$$

Similarly to (2.24), we have

$$\log^{\tilde{\alpha}} b_k - \log^{\tilde{\alpha}} a_k \sim \frac{(1 - 2\varepsilon)(2 + \alpha)C_1}{2\alpha} \log_2 b_k.$$

Now from Lemma 2.5 it follows that

$$\begin{aligned} P(A_k) &\geq \frac{1}{4} \exp\left(-\frac{(1+\varepsilon)\alpha(2C)^{1/\alpha}H_\alpha(\log \tilde{a}_k - \log \tilde{a}_k)}{(2+\alpha)\sqrt{\pi}}\right) - Na_k^{-\tau} \\ &\geq \frac{1}{4} \exp\left(-\frac{(1+2\varepsilon)(2C)^{1/\alpha}(1-2\varepsilon)C_1 \log_2 b_k}{2\sqrt{\pi}}\right) - 2Nb_k^{-\tau} \\ &\geq \frac{1}{8} k^{-(1-\varepsilon^4)} \end{aligned}$$

for every  $k$  sufficiently large. Hence

$$\sum_{k=1}^{\infty} P(A_k) = \infty \tag{2.33}$$

and (2.32) will be implied by

$$\liminf_{n \rightarrow \infty} \frac{\sum_{1 \leq k \neq l \leq n} P(A_k A_l)}{(\sum_{k=1}^n P(A_k))^2} \leq 1 \tag{2.34}$$

by the general form of the Borel-Cantelli lemma (cf. [9]). We below verify that (2.34) is satisfied. It is easy to see that

$$\frac{a_{k+1} - b_k}{b_{k+1} - b_k} \sim 1. \tag{2.35}$$

Applying Lemma 2.1, we get that for  $k < l$

$$\begin{aligned} P(A_k A_l) &\leq P(A_k)P(A_l) \\ &+ \sum_{i=0}^{b_k - a_k} \sum_{j=0}^{\hat{y}_{k,i}} \sum_{u=0}^{b_l - a_l} \sum_{v=0}^{\hat{y}_{l,u}} \frac{\bar{\tau}(i, j, u, v)}{\sqrt{1 - \bar{\tau}(i, j, u, v)}} \exp\left(-\frac{\frac{1}{2}(\bar{y}_{k,i}^2 + \bar{y}_{l,u}^2)}{1 + \bar{\tau}(i, j, u, v)}\right) \\ &:= P(A_k)P(A_l) + C_{k,l} \end{aligned} \tag{2.36}$$

where

$$\begin{aligned} \bar{\tau}(i, j, u, v) &= |r(a_l + u + v\theta_{l,u}y_{l,u}^{-2/\alpha} - a_k - i - j\theta_{k,i}y_{k,i}^{-2/\alpha})| \\ &\leq r^*(a_l - b_k - 1) \\ &\leq r^*(\frac{1}{2}(b_l - b_{l-1})) \\ &\leq (b_l - b_{l-1})^{-\gamma} \leq \min(1, \gamma)/4 \end{aligned}$$

by (2.35) and (1.6), for every  $k, l$  ( $k < l$ ) sufficiently large. Therefore

$$\begin{aligned} C_{k,l} &\leq K \sum_{i=0}^{b_k - a_k} \sum_{u=0}^{b_l - a_l} \hat{y}_{k,i} \hat{y}_{l,u} (b_l - b_{l-1})^{-\gamma} \exp\left(-\frac{\log(a_k + i) + \log(a_l + u)}{1 + \gamma/4}\right) \\ &\leq K b_k^{\gamma/4} \log^{9/\alpha} b_k \cdot b_l^{\gamma/4} \log^{9/\alpha} b_l \cdot (b_l - b_{l-1})^{-\gamma} \\ &\leq K \exp(-\gamma l^{\alpha/4}/8) \end{aligned}$$

for some constant  $K$  not depending on  $k$  and  $l$ , and for every  $k < l$  sufficiently large. Hence we have

$$\sum_{0 \leq k < l < \infty} C_{k,l} < \infty. \quad (2.37)$$

Now (2.34) follows from (2.36), (2.37) and (2.33). This proves (2.30) and so does (2.29).

*Step 3* If  $\alpha = 2$ , then

$$\liminf_{t \rightarrow \infty} \frac{\log(\xi(t)/t)}{\log_2 t} \geq -(1 + 2\varepsilon)^2 C_2 \quad \text{a.s.} \quad (2.38)$$

and

$$\liminf_{t \rightarrow \infty} \frac{\log(\xi(t)/t)}{\log_2 t} \leq -(1 - 2\varepsilon) C_2 \quad \text{a.s.} \quad (2.39)$$

for every  $0 < \varepsilon < \frac{1}{4}$ , where  $C_2 = 2\sqrt{\pi}/(H_2\sqrt{2C}) = 2\pi/\sqrt{2C}$ .

*Proof.* Put

$$t_k = e^k, \quad s_k = t_k \exp(- (1 + 2\varepsilon)^2 C_2 \log_2 t_k), \quad k = 1, 2, \dots$$

Proceeding the same way as that of the proof of (2.25), one can obtain that

$$\liminf_{k \rightarrow \infty} \frac{\log(\xi(t_k)/t_k)}{\log_2 t_k} \geq -(1 + 2\varepsilon)^2 C_2 \quad \text{a.s.} \quad (2.40)$$

On the other hand, it is clear that

$$\liminf_{t \rightarrow \infty} \frac{\log(\xi(t)/t)}{\log_2 t} = \liminf_{k \rightarrow \infty} \frac{\log(\xi(t_k)/t_k)}{\log_2 t_k} \quad \text{a.s.} \quad (2.41)$$

since  $\lim_{k \rightarrow \infty} \log(t_{k+1}/t_k)/\log_2 t_k = 0$ . This proves (2.38), by (2.40) and (2.41).

Let

$$b_k = \exp(k^{1+\varepsilon^2}), \quad a_k = b_k \exp(- (1 - 2\varepsilon) C_2 \log_2 b_k), \quad k = 1, 2, \dots$$

Noting that

$$\frac{a_{k+1} - b_k}{a_{k+1}} \sim 1,$$

along the same line of the proof of (2.30), we also have

$$\liminf_{k \rightarrow \infty} \frac{\log(\xi(b_k)/b_k)}{\log_2 b_k} \leq -(1 - 2\varepsilon) C_2 \quad \text{a.s.}$$

This proves (2.39).

Now the proof of Theorem 1.1 is completed.

### 3 Infinite series of independent Ornstein-Uhlenbeck processes

In this section we consider a special stationary Gaussian process, infinite series of independent Ornstein-Uhlenbeck processes. A real valued stationary Gaussian process  $\{X(t), -\infty < t < \infty\}$  is called an Ornstein-Uhlenbeck process with

coefficients  $\gamma$  and  $\lambda$  ( $\gamma \geq 0, \lambda > 0$ ) if  $EX(t) = 0$  and  $EX(s)X(t) = (\gamma/\lambda)\exp(-\lambda|t-s|)$ . Let  $Y(t) = (X_1(t), X_2(t), \dots)$ , where  $\{X_i(t), -\infty < t < \infty\}$  are independent Ornstein-Uhlenbeck processes with coefficients  $\gamma_i$  and  $\lambda_i$  ( $i = 1, 2, \dots$ ). The process  $Y(\cdot)$  was first studied by Dawson [2] as the stationary solution of the infinite array of stochastic differential equations

$$dX_i(t) = -\lambda_i X_i(t)dt + \sqrt{2\gamma_i}dW_i(t), \quad i = 1, 2, \dots,$$

where  $\{W_i(t), -\infty < t < \infty\}$  are independent standard Wiener processes. The path properties of  $Y(\cdot)$  have been extensively investigated by various authors during the past two decades. We refer to Csáki et al. [1] and references therein.

We concern here with the infinite sums of  $Y(\cdot)$ . Assume

$$0 < \Gamma_0^2 = \sum_{i=1}^{\infty} \frac{\gamma_i}{\lambda_i} < \infty. \tag{3.1}$$

Define

$$X(t) = \frac{1}{\Gamma_0} \sum_{i=1}^{\infty} X_i(t), \tag{3.2}$$

$$\sigma^2(t) = \frac{1}{\Gamma_0^2} \sum_{i=1}^{\infty} \frac{\gamma_i}{\lambda_i} (1 - e^{-\lambda_i t}), \quad t \geq 0. \tag{3.3}$$

It is clear that  $\{X(t), t \geq 0\}$  is a stationary Gaussian process with  $EX(t) = 0$ ,  $EX^2(t) = 1$  and covariance function

$$r(t) = EX(t+s)X(s) = 1 - \sigma^2(t) \quad \text{for } s, t \geq 0. \tag{3.4}$$

**Theorem 3.1** *Let  $\{X(t), t \geq 0\}$  be the infinite sums of independent Ornstein-Uhlenbeck processes, defined as in (3.2). Put  $\xi(t) = \sup\{s: 0 \leq s \leq t, X(s) \geq (2 \log s)^{1/2}\}$ ,  $t \geq 0$ . Let  $\sigma^2(t)$  be as in (3.3). Assume that there exist  $0 < \alpha \leq 1$ ,  $C > 0$  and  $\delta > 0$  such that*

$$\sum_{i=1}^{\infty} \frac{\gamma_i}{\min(\lambda_i, \lambda_i^{1+\delta})} < \infty, \tag{3.5}$$

$$\lim_{t \downarrow 0} \frac{\sigma^2(t)}{t^\alpha} = C. \tag{3.6}$$

Then (1.8) holds.

*Proof.* It is easy to find that (1.7) is satisfied since  $\sigma^2(t)$  is a positive non-decreasing function for  $t > 0$ . By Theorem 1.1, it suffices to verify that (1.6) is satisfied. Notice that

$$\begin{aligned} r(t) &= \frac{1}{\Gamma_0^2} \sum_{i=1}^{\infty} \frac{\gamma_i}{\lambda_i} e^{-\lambda_i t} \\ &= \frac{1}{\Gamma_0^2} \left( \sum_{i: \lambda_i \geq t^{-1/2}} + \sum_{i: \lambda_i < t^{-1/2}} \right) \frac{\gamma_i}{\lambda_i} e^{-\lambda_i t} \\ &\leq \frac{1}{\Gamma_0^2} \left( \sum_{i=1}^{\infty} \frac{\gamma_i}{\lambda_i} \exp(-t^{1/2}) + \sum_{i: \frac{1}{\lambda_i} > t^{1/2}} \frac{\gamma_i}{\lambda_i} \right) \\ &\leq \exp(-t^{1/2}) + \frac{1}{\Gamma_0^2} \sum_{i=1}^{\infty} \frac{\gamma_i}{\lambda_i^{1+\delta}} \cdot t^{-\delta/2} \\ &\leq Kt^{-\delta/2} \end{aligned}$$

for some constant  $K$  and for every  $t \geq 2$ . This proves (1.6), as desired.

**Corollary 3.1** *Let  $\{X(t), t \geq 0\}$  and  $\{\xi(t), t \geq 0\}$  be defined as in Theorem 3.1. Assume for some  $\delta > 0$*

$$\sum_{i=1}^{\infty} \frac{\gamma_i}{\min(1, \lambda_i^{1+\delta})} < \infty . \tag{3.7}$$

Then

$$\liminf_{t \rightarrow \infty} \frac{\xi(t) - t}{t \log^{-1/2} t \log_2 t} = -\frac{3\sqrt{\pi}}{2\Gamma_1} \text{ a.s.} \tag{3.8}$$

where  $\Gamma_1 = \Gamma_0^{-2} \sum_{i=1}^{\infty} \gamma_i$ .

*Proof.* Clearly, (3.7) implies (3.5) and

$$\sum_{i=1}^{\infty} \gamma_i < \infty .$$

The latter yields

$$\lim_{t \downarrow 0} \frac{\sigma^2(t)}{t} = \Gamma_1 .$$

Now (3.8) follows from Theorem 3.1 with  $\alpha = 1$ ,  $C = \Gamma_1$  and  $H_1 = 1$ .

#### 4 Fractional Wiener process

Let  $\{Z(t), t \geq 0\}$  be a fractional Wiener process of order  $\alpha$ , i.e., a centred Gaussian process with stationary increments and variance  $EZ^2(t) = t^\alpha$ , where  $0 < \alpha < 2$ . Consider

$$\eta(t) = \sup \{s: 0 \leq s \leq t, Z(s) \geq (2s^\alpha \log_2 s)^{1/2}\}, \quad t \geq 0 . \tag{4.1}$$

By the upper class of increments for  $Z(t)$  (cf. [4]), one has

$$\lim_{t \rightarrow \infty} \eta(t) = \infty \text{ a.s.}$$

and

$$\limsup_{t \rightarrow \infty} (\eta(t) - t) = 0 \text{ a.s.}$$

The following theorem presents the lower bound of  $\eta(\cdot)$ .

**Theorem 4.1** *We have*

$$\liminf_{t \rightarrow \infty} \frac{(\log_2 t)^{(2-\alpha)/(2\alpha)} \cdot \log(\eta(t)/t)}{\log t \cdot \log_3 t} = -\frac{(2+\alpha)\sqrt{\pi}}{\alpha H_\alpha} \text{ a.s.} \tag{4.2}$$

*Proof.* Let

$$X(t) = Z(e^t)/e^{\alpha t/2}, \quad t \geq 0 .$$

Then,  $EX(t) = 0$ ,  $EX^2(t) = 1$  and

$$EX(s)X(t) = \frac{1}{2}(e^{\alpha(t-s)/2} + e^{\alpha(s-t)/2} - |e^{(t-s)/2} - e^{(s-t)/2}|^\alpha).$$

Hence,  $\{X(t), t \geq 0\}$  is a stationary Gaussian process with correlation function

$$r(t) = \frac{1}{2}(e^{\alpha t/2} + e^{-\alpha t/2} - (e^{t/2} - e^{-t/2})^\alpha) \text{ for } t \geq 0. \tag{4.3}$$

It is not difficult to find that

$$r(t) - 1 \sim \frac{1}{2}t^\alpha \text{ as } t \downarrow 0, \tag{4.4}$$

$$r(t) \sim e^{-\frac{\alpha}{2}t} + \alpha e^{-(1-\frac{\alpha}{2})t} \text{ as } t \rightarrow \infty \tag{4.5}$$

and

$$\sup_{t \geq s} |r(t)| < 1 \text{ for every } s > 0. \tag{4.6}$$

Therefore, by Theorem 1.1

$$\liminf_{t \rightarrow \infty} \frac{(\log t)^{(2-\alpha)/(2\alpha)}(\zeta(t) - t)}{t \cdot \log_2 t} = -\frac{(2+\alpha)\sqrt{\pi}}{\alpha H_\alpha} \text{ a.s.} \tag{4.7}$$

where  $\zeta(t)$  is defined as in (1.4). Let

$$\bar{\eta}(t) = \sup \{s : 1 \leq s \leq t : Z(s) \geq (2s^\alpha \log_2 s)^{1/2}\} \text{ for } s \geq 1.$$

Then

$$\zeta(t) = \log \bar{\eta}(e^t) \text{ a.s. for every } t \geq 0. \tag{4.8}$$

Consequently, we have, by (4.7)

$$\liminf_{t \rightarrow \infty} \frac{(\log_2 t)^{(2-\alpha)/(2\alpha)} \cdot \log(\bar{\eta}(t)/t)}{\log t \cdot \log_3 t} = -\frac{(2+\alpha)\sqrt{\pi}}{\alpha H_\alpha} \text{ a.s.} \tag{4.9}$$

This proves (4.2) by (4.9) and the fact that  $|\bar{\eta}(t) - \eta(t)| \leq 1$  for every  $t \geq 1$ .

*Acknowledgements.* The author thanks an associate editor and the editor for their helpful comments.

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