

Some limsup results for increments of stable processes in random scenery*

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Abstract: In this paper, we prove some limsup results for increments and lag increments of $G(t)$, which is a stable process in random scenery. The proofs rely on the tail probability estimation of $G(t)$.

Key words: Local time, Random walk in random scenery, Stable process in random scenery, Increments, Lag increments.

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INTRODUCTION

Let \mathbf{Z} , \mathbf{N}_0 and \mathbf{N} denote the sets of integers, non-negative integers and positive integers, respectively. And let $\chi = \{X_i\}_{i \in \mathbf{Z}}$ denote a sequence of real-valued random variables such that

$$EX_i = 0 \text{ and } EX_i^2 = \sigma_X^2 \quad \forall i \in \mathbf{Z}.$$

Any realization of the sequence $\{X_i\}_{i \in \mathbf{Z}}$ is called a "random scenery". Let $y = \{Y_j\}_{j \in \mathbf{N}}$ be a sequence which satisfies

$$EX_j = 0 \text{ and } EX_j^2 = \sigma_Y^2 \quad \forall j \in \mathbf{N}.$$

We will assume that the collections χ and y are defined on a common probability space and that they generate independent σ -fields. Let $S_0 = 0$

and, for each $n \in \mathbf{N}$, let $S_n = \sum_{j=1}^n Y_j$. then $S = \{S_n\}_{n \in \mathbf{N}_0}$ is a random walk on \mathbf{R} starting at $S_0 = 0$. If y is a sequence of integer-valued random variables, then S is a random walk on \mathbf{Z} . Now for each $n \in \mathbf{N}_0$, let

$$g_n = \sum_{k=0}^n X(\lfloor S_k \rfloor). \quad (1)$$

The process $G = \{g_n : n \in \mathbf{N}_0\}$ is called random walk in random scenery, on which see Kesten and Spitzer (1979) and Révész (1990) for more

details. For example, the model can be viewed as follows: if a random walker has to pay the amount of X_i dollars whenever he visits the site i , then g_n is the total amount he pays during the first n steps.

There is a continuous analogue for G introduced and analyzed by Kesten and Spitzer (1979). To describe this, let $\mathbf{Y}_\pm = \{Y_\pm(t) : t \geq 0\}$ denote two standard Brownian motions and let $\mathbf{X} = \{X(t) : t \geq 0\}$ be a strictly stable Lévy process with index α ($1 < \alpha \leq 2$). We will assume that \mathbf{Y}_+ , \mathbf{Y}_- and \mathbf{X} are defined on a common probability space and that they generate independent σ -fields. And we will define a two-sided Brownian motion $\mathbf{Y} = \{Y(t) : t \in \mathbf{R}\}$ according to the rule

$$Y(t) = \begin{cases} Y_+(t) & \text{if } t \geq 0, \\ Y_-(-t) & \text{if } t < 0. \end{cases}$$

Give a function $f: \mathbf{R} \rightarrow \mathbf{R}$, we will let

$$\int_{\mathbf{R}} f(x) dY(x) := \int_0^\infty f(x) dY_+(x) + \int_0^\infty f(-x) dY_-(x)$$

provided that both of the Itô integrals on the right-hand side are defined.

Let $\mathbf{L} = \{\mathbf{L}(t, x) : t \geq 0, x \in \mathbf{R}\}$ denote the jointly continuous version of the local time pro-

cess of \mathbf{X} , in the sense that for any measurable function $f: \mathbf{R} \rightarrow \mathbf{R}$ and for each $t \geq 0$,

$$\int_0^t f(\mathbf{X}(s)) ds = \int_{\mathbf{R}} f(a) L(t, a) da.$$

With this in mind, we can define the following process well. Let

$$G(t) := \int_{\mathbf{R}} L(t, x) dY(x) \quad \forall t \geq 0, \quad (2)$$

then $\mathbf{G} = \{G(t): t \geq 0\}$ is called a stable process in random scenery.

Remark A Since \mathbf{X} is a strictly stable Lévy process of index $\alpha (1 < \alpha \leq 2)$, it is self-similar with index α^{-1} . The process of local time \mathbf{L} inherits a scaling law from \mathbf{X} : for each $c > 0$,

$$\{\mathbf{L}(ct, x): t \geq 0, x \in \mathbf{R}\} \stackrel{d}{=} \{c^{1-1/\alpha} L(t, xc^{-1/\alpha}): t \geq 0, x \in \mathbf{R}\}.$$

Since a standard Brownian motion is self-similar with index $1/2$, it follows that G is self-similar with index $\delta = 1 - 1/(2\alpha)$.

RESULTS AND PROOFS

Before we state the results, we shall give a notion. Throughout this paper we denote $\log x = \ln \max\{e, x\}$. And let \mathbf{G} be a stable process in random scenery, defined by Eq. (2). $\alpha (1 < \alpha \leq 2)$ is the index of the Lévy process \mathbf{X} .

Csáki et al. (1999) and Zhang (2001a, b) proved that g_n can be approximated by $G(t)$ under some suitable conditions. This work is aimed at studying the increments of $G(t)$. Here are our main theorems.

Theorem 1 Let $h > 0$ be a real number, we have

$$\limsup_{h \rightarrow 0} \sup_{0 \leq t \leq T-h} \sup_{0 < s \leq h} \frac{|G(t+s) - G(t)|}{h^\delta (\log(1/h)/\gamma)^{1/\beta}} \leq 1 \text{ a. s.}, \quad (3)$$

where $\beta = 2\alpha/(1 + \alpha)$, $\delta = 1 - 1/(2\alpha)$ and $\gamma = \gamma(\alpha)$ is a positive real number whose value will be defined in Lemma 4.

Theorem 2 Let a_T be a continuous function of T , $0 < a_T \leq T$, then

$$\limsup_{T \rightarrow \infty} \sup_{0 \leq t \leq T-a_T} \sup_{0 < s \leq a_T} \beta_T |G(t+s) - G(t)| \leq 1 \text{ a. s.}, \quad (4)$$

where $\beta_T^{-1} = a_T^\delta \{(\log(T/a_T) + \log \log T)/$

$\gamma\}^{1/\beta}$.

If we have also (I) $\lim_{T \rightarrow \infty} \frac{a_T}{T} = \rho$, $0 < \rho \leq 1$, then

$$\limsup_{T \rightarrow \infty} \beta_T |G(T + a_T) - G(T)| = 1 \text{ a. s.}, \quad (5)$$

$$\limsup_{T \rightarrow \infty} \sup_{0 \leq t \leq T-a_T} \beta_T |G(t + a_T) - G(t)| = 1 \text{ a. s.}, \quad (6)$$

$$\limsup_{T \rightarrow \infty} \sup_{0 < s \leq a_T} \beta_T |G(T + s) - G(T)| = 1 \text{ a. s.}, \quad (7)$$

$$\limsup_{T \rightarrow \infty} \sup_{0 \leq t \leq T-a_T} \sup_{0 < s \leq a_T} \beta_T |G(t + s) - G(t)| = 1 \text{ a. s.} \quad (8)$$

Theorem 3 We have

$$\limsup_{T \rightarrow \infty} \sup_{0 < t \leq T} |G(T) - G(T-t)|/d(T, t) = 1 \text{ a. s.}, \quad (9)$$

$$\limsup_{T \rightarrow \infty} \sup_{0 < t \leq T} \sup_{t \leq s \leq T} |G(s) - G(s-t)|/d(T, t) = 1 \text{ a. s.}, \quad (10)$$

where $d(T, t) = t^\delta \{(\log(T/t) + \log \log t)/\gamma\}^{1/\beta}$.

The proofs of the theorems depend on the following lemmas.

Lemma 4 (Khoshnevisan and Lewis, 1998) Let $\lambda \geq 0$, then there exists a positive real number $\gamma = \gamma(\alpha)$, such that

$$\lim_{\lambda \rightarrow \infty} \lambda^{-\beta} \log P \{G(1) \geq \lambda\} = -\gamma. \quad (11)$$

Lemma 5 For any $\epsilon > 0$, there exists a constant $C = C(\alpha, \epsilon) > 0$ such that the inequality

$$P \left\{ \sup_{0 \leq t \leq T-h} \sup_{0 < s \leq h} |G(t+s) - G(t)| \geq h^\delta \lambda \right\} \leq Ch^{-1} \exp \{-(1-\epsilon)\gamma\lambda^\beta\} \quad (12)$$

holds for every positive λ and $h < 1$, where $\gamma = \gamma(\alpha)$ is a positive real number which is specified by Lemma 4.

Proof By Lemma 4 and Remark A, we obtain obviously, for any $\epsilon > 0$, there exists a positive real number $\lambda_0 = \lambda_0(\alpha, \epsilon)$, such that

$$P \{ |G(t+h) - G(t)| \geq h^\delta \lambda \} \leq 2 \exp \{-(1-\epsilon/2)\gamma\lambda^\beta\} \quad (13)$$

holds for every positive $h < 1$ and $\lambda \geq \lambda_0$.

Using again the above notations, for a positive real number t and integer r , let $t_r = [2^r t]/2^r = \sum_{j=0}^r (\epsilon_j(t)/2^j)$, where $t =$

$\sum_{j=0}^{\infty} (\varepsilon_j(t)/2^j)$, $\varepsilon_0(t) = 0, 1, 2, \dots$; $\varepsilon_k(t) = 0, 1$, for $k = 1, 2, \dots$, and $\varepsilon_k(t)$ should not be identically 1 from some k on. Also write $R = 2^r$. Clearly, for each $\omega \in \Omega$, and s, t, r fixed, We have

$$|G(t+s) - G(t)| \leq |G((t+s)_r) - G(t_r)| + \sum_{j=0}^{\infty} |G((t+s)_{r+j+1}) - G((t+s)_{r+j})| + \sum_{j=0}^{\infty} |G(t_{r+j+1}) - G(t_{r+j})|.$$

Observe that

$$\sup_{0 \leq t \leq 1-h} \sup_{0 < s \leq h} |(t+s)_r - t_r| \leq h + R^{-1},$$

$$\sup_{0 \leq t \leq 1-h} \sup_{0 < s \leq h} |(t+s)_{r+j+1} - (t+s)_{r+j}| \leq 2^{-(r+j+1)},$$

$$\sup_{0 \leq t \leq 1-h} \sup_{0 < s \leq h} |t_{r+j+1} - t_{r+j}| \leq 2^{-(r+j+1)},$$

for any positives $h < 1$, and non-negative integers r, j . So by Eq. (13) and the definition of the stochastic process $\mathbf{G} = \{G(t) : t \geq 0\}$, we have

$$P \left\{ \sup_{0 \leq t \leq 1-h} \sup_{0 < s \leq h} |G(t+s) - G(t)| \geq (h + R^{-1})^\delta u + 2 \sum_{j=0}^{\infty} (2^{-(r+j+1)})^\delta x_j \right\} \leq 2R(Rh + 1) \exp \left\{ -(1 - \varepsilon/2) \gamma u^\beta \right\} + 8R \sum_{j=0}^{\infty} 2^j \exp \left\{ -(1 - \varepsilon/2) \gamma x_j^\beta \right\}$$

for any positive $u \geq \lambda_0$, $x_j \geq \lambda_0$. Put $x_j = \left(\frac{j}{(1 - \varepsilon/2)\gamma} + u^\beta \right)^{1/\beta}$, then

$$P \left\{ \sup_{0 \leq t \leq 1-h} \sup_{0 < s \leq h} |G(t+s) - G(t)| \geq u((h + R^{-1})^\delta + 2(R^{-1})^\delta \sum_{j=0}^{\infty} (2^{-j-1})^\delta) + 2(R^{-1})^\delta \sum_{j=0}^{\infty} \left(\frac{j}{(1 - \varepsilon/2)\gamma} \cdot 2^{-\frac{2\alpha-1}{1+\alpha}(j+1)} \right)^{1/\beta} \right\} \leq (2R(Rh + 1) + 8R \sum_{j=0}^{\infty} (2/e)^j) \exp \left\{ -(1 - \varepsilon/2) \gamma u^\beta \right\}.$$

Put $R = 2^r$ such that $R \leq k/h < 2R$, where k is a positive constant which is large enough and will be specified later on. Then

$$2R(Rh + 1) + 8R \sum_{j=0}^{\infty} (2/e)^j \leq 2kh^{-1}(k + 1)$$

$$+ 8kh^{-1} \sum_{j=0}^{\infty} (2/e)^j = : Ah^{-1}k(k + 1) = : Ch^{-1},$$

where A is a positive constant which does not depend on k . Here $C = C(k)$ is a positive constant only relying on k . In the sequel, C only indicates a constant, which can take different values in different places. Also,

$$u((h + R^{-1})^\delta + 2(R^{-1})^\delta \sum_{j=0}^{\infty} (2^{-j-1})^\delta) + 2(R^{-1})^\delta \sum_{j=0}^{\infty} \left(\frac{j}{(1 - \varepsilon/2)\gamma} 2^{-\frac{2\alpha-1}{1+\alpha}(j+1)} \right)^{1/\beta} \leq uh^\delta((1 + 2/k)^\delta + 2(2/k)^\delta \sum_{j=0}^{\infty} (2^{-j-1})^\delta) + 2h^\delta(2/k)^\delta \sum_{j=0}^{\infty} \left(\frac{j}{(1 - \varepsilon/2)\gamma} 2^{-\frac{2\alpha-1}{1+\alpha}(j+1)} \right)^{1/\beta} = : h^\delta(u((1 + 2/k)^\delta + G/k^\delta) + B/k^\delta),$$

where G, B are positive constants not relying on k . Now letting $\lambda = u((1 + 2/k)^\delta + G/k^\delta) + B/k^\delta$, then

$$P \left\{ \sup_{0 \leq t \leq 1-h} \sup_{0 < s \leq h} |G(t+s) - G(t)| \geq h^\delta \lambda \right\} \leq Ch^{-1} \exp \left\{ -(1 - \varepsilon/2) \gamma u^\beta \right\} \leq Ch^{-1} \exp \left\{ -(1 - \varepsilon) \gamma \lambda^\beta \right\},$$

where the last inequality follows from the inequality

$$u = \frac{\lambda - B/k^\delta}{(1 + 2/k)^\delta + G/k^\delta} \geq \frac{\lambda}{((2 - \varepsilon)/(2 - 2\varepsilon))^{1/\beta}}$$

for any given $\varepsilon > 0$ upon taking k large enough. This proves our lemma with $\lambda \geq \lambda_0(\varepsilon)$, while it is trivially true for $\lambda \in (0, \lambda_0)$. Since, in the latter case, the right-hand side of Eq. (12) is larger than one for C big enough.

Lemma 6 For any $\varepsilon > 0$, there exists a constant $C = C(\alpha, \varepsilon) > 0$ such that the inequality

$$P \left\{ \sup_{0 \leq t \leq 1-h} \sup_{0 < s \leq h} |G(t+s) - G(t)| \geq h^\delta \lambda \right\} \leq CTh^{-1} \exp \left\{ -(1 - \varepsilon) \gamma \lambda^\beta \right\}$$

holds for every positive λ , T and $h < T$, where $\gamma = \gamma(\alpha)$ is a positive real number which is specified by Lemma 4.

Proof This lemma follows from Eq. (12) and the following observation: for any fixed $T > 0$, we have

$$\{G(t) : 0 \leq t \leq T\} \stackrel{d}{=} \{T^\delta G(t/T) : 0 \leq t \leq T\}.$$

Remark The proofs of Lemma 5 and Lemma 6 come from the idea of Csörgö and Révész

(1981).

Proof of Theorem 1 Let $A_h = \sup_{0 \leq t \leq 1-h} \sup_{0 < s \leq h} |G(t+s) - G(t)|$. We apply Eq. (12) with $\lambda = (1 + 3\epsilon)(\log(1/h)/\gamma)^{1/\beta}$, for any $0 < \epsilon < 1/3$, then

$$P \left\{ \frac{A_h}{h^\delta (\log(1/h)/\gamma)^{1/\beta}} \geq 1 + 3\epsilon \right\} \leq Ch^{-1} \cdot \exp \{ - (1 - \epsilon)(1 + 3\epsilon)^\beta \log(1/h) \} \leq Ch^\epsilon.$$

Take $T > 1/\epsilon$, and let $h_n = n^{-T}$, for $n = 1, 2, \dots$, then

$$\sum_{n=1}^{\infty} P \left\{ \frac{A_{h_n}}{h_n^\delta (\log(1/h_n)/\gamma)^{1/\beta}} \geq 1 + 3\epsilon \right\} \leq C \sum_{n=1}^{\infty} n^{-T\epsilon} < \infty$$

and the Borel-Cantelli lemma implies that

$$\limsup_{n \rightarrow \infty} \frac{A_{h_n}}{h_n^\delta (\log(1/h_n)/\gamma)^{1/\beta}} < 1 + 3\epsilon \text{ a.s.}$$

Let us take $h_{n+1} < h \leq h_n$, for $n = 1, 2, \dots$, then for each $\omega \in \Omega$, and $0 < \epsilon < 1/3$, we have

$$\begin{aligned} \limsup_{h \rightarrow 0} \frac{A_h}{h^\delta (\log(1/h)/\gamma)^{1/\beta}} &\leq \\ \limsup_{n \rightarrow \infty} \frac{A_{h_n}}{h_{n+1}^\delta (\log(1/h_n)/\gamma)^{1/\beta}} &= \\ \limsup_{n \rightarrow \infty} \frac{A_{h_n}}{h_n^\delta (\log(1/h_n)/\gamma)^{1/\beta}} \left(\frac{h_n}{h_{n+1}} \right)^\delta &\leq \\ 1 + 3\epsilon \text{ a.s.} \end{aligned}$$

Whence we obtain Eq. (3).

Proof of Theorem 2 This will be given in two steps.

Step1: Let $A(T) = \sup_{0 \leq t \leq T - a_T} \sup_{0 < s \leq a_T} |G(t+s) - G(t)|$. We will prove

$$\limsup_{T \rightarrow \infty} A(T) \leq 1 \text{ a.s.} \quad (15)$$

Proof For any $0 < \epsilon < 1/3$, take a positive real number θ such that $1 < \theta < 1 + \epsilon$, write

$$\begin{aligned} A_k &:= \{T: \theta^k \leq T/a_T < \theta^{k+1}\}, \\ A_{kj} &:= \{T: \theta^j \leq a_T < \theta^{j+1}, T \in A_k\}, \end{aligned}$$

for any integers $k \geq 0$ and j . We have

$$\begin{aligned} \limsup_{T \rightarrow \infty} A(T) &= \limsup_{k+j \rightarrow \infty} \sup_{T \in A_k} A(T) \leq \limsup_{k+j \rightarrow \infty} \\ &\sup_{0 \leq t \leq \theta^{j+1}} \sup_{0 < s \leq \theta^{j+1}} \frac{\theta^\delta |G(t+s) - G(t)|}{(\theta^{j+1})^\delta ((\log(\theta^k \log \theta^{k+j}))/\gamma)^{1/\beta}}. \end{aligned} \quad (16)$$

On the other hand, by Lemma 6, we obtain

$$P \left\{ \sup_{0 \leq t \leq \theta^{j+1}} \sup_{0 < s \leq \theta^{j+1}} \frac{|G(t+s) - G(t)|}{(\theta^{j+1})^\delta ((\log(\theta^k \log \theta^{k+j}))/\gamma)^{1/\beta}} \geq 1 + 3\epsilon \right\} \leq C\theta^{k+1} \exp \{ - (1 - \epsilon)(1 + 3\epsilon)^\beta \} \log(\theta^k \log \theta^{k+j}) \leq C\theta^{-k\epsilon} j^{-(1+\epsilon)} (\log \theta)^{-(1+\epsilon)}.$$

Observing

$$\sum_{j=1}^{\infty} \sum_{k=1}^{\infty} \theta^{-k\epsilon} j^{-(1+\epsilon)} < \infty,$$

together with Eq. (16), we proof Eq. (15), by the Borel-Cantelli lemma and letting $\theta \rightarrow 1$.

Step 2: Let $B(T) = \beta_T (G(T + a_T) - G(T))$. Suppose that Condition (I) of Theorem 2 is fulfilled, then

$$\limsup_{T \rightarrow \infty} B(T) \geq 1 \text{ a.s.} \quad (17)$$

Proof Let $T_k = \exp \{k^p\}$ for $k \geq 1$, where $p > 1$ will be specified later on. And for any positive $\epsilon < 1$, define $\phi = ((1 - \epsilon)/\gamma)^{1/\beta}$, this implies that $\gamma\phi^\beta = 1 - \epsilon$.

It is easy to see that the proof of Eq. (17) can be reduced to the following result: for each $0 < \epsilon < 1$,

$$\limsup_{k \rightarrow \infty} \xi_k \geq 0 \text{ a.s.}, \quad (18)$$

where $\xi_k = \frac{G(T_k + a_{T_k}) - G(T_k)}{a_{T_k}^\delta} - \phi(\log(T_k/a_{T_k}) + \log \log T_k)^{1/\beta}$, for $k \geq 1$.

By the definition of T_k , there exists an integer $N_1 \geq 1$ large enough such that $T_k + a_{T_k} \leq 2T_k \leq T_{k+1}$, for each $k \geq N_1$. Thus, by Proposition 3.1 of Khoshnevisan and Lewis (1998), the collection of random variables $\{\xi_k, k \geq N_1\}$ is pairwise positively quadrant dependent. So it would suffice to establish items (a) and (b) of Proposition 4.1 of Khoshnevisan and Lewis (1998). Using Lemma 4, we have

$$\lim_{k \rightarrow \infty} \frac{\log P \{ \xi_k \geq 0 \}}{\log((T_k \log T_k)/a_{T_k})} = -\gamma\phi^\beta = -(1 - \epsilon).$$

Let $1 - \epsilon < q < 1$, then there exists an integer $N_2 \geq N_1$ such that, for each $k \geq N_2$, we get

$$P \{ \xi_k \geq 0 \} \geq (a_{T_k}/(T_k \log T_k))^q.$$

And let $0 < h < \rho \leq 1$, by condition (I), we know we can take an integer $N_3 \geq N_2$, which satisfies, for each $k \geq N_3$, $a_{T_k}/T_k \geq h$. Hence

taking $1 < p \leq 1/q$, we obtain

$$\sum_{k=N_3}^{\infty} (a_{T_k} / (T_k \log T_k))^q \geq C \sum_{k=N_3}^{\infty} k^{-pq} \geq C \sum_{k=N_3}^{\infty} k^{-1} = \infty,$$

which verifies (a).

And clearly $\text{Var}((G(T_j + a_{T_j}) - G(T_j)) / a_{T_j}^{\delta}) < \infty$, for $j = 1, 2, \dots$. Put $0 < \lambda := 1/(1 + h) < 1$ in Theorem 5.2 of Khoshnevisan and Lewis (1998), then, there exists a positive constant $C = C(\alpha, \lambda)$, such that

$$\text{Cov}(\xi_j, \xi_k) = \text{Cov}\left(\frac{G(T_j + a_{T_j}) - G(T_j)}{a_{T_j}^{\delta}}, \frac{G(T_k + a_{T_k}) - G(T_k)}{a_{T_k}^{\delta}}\right) \leq C \left(\frac{T_j + a_{T_j}}{T_k + a_{T_k}}\right)^{1/(2\alpha)},$$

for $N_3 \leq j < k$. While

$$\frac{T_j + a_{T_j}}{T_k + a_{T_k}} \leq \frac{2T_j}{T_k} = 2 \exp\left\{-p \int_j^k x^{p-1} dx\right\} \leq 2 \exp\{-pj^{p-1}(k-j)\}.$$

For $j \geq N_3$, let $b_j = \exp\{-pj^{p-1}/(2\alpha)\}$, thus

$$\sum_{N_3 \leq j < k < \infty} \text{Cov}(\xi_j, \xi_k) \leq C \sum_{j=N_3}^{\infty} \sum_{k=j+1}^{\infty} b_j^{(k-j)} \leq C \frac{1}{1-b_1} \sum_{j=N_3}^{\infty} b_j < \infty,$$

which verifies (b). Thus Eq. (18) is proved.

Proof of Theorem 3 Similarly to the proof in Theorem 1.1.3 of Lin and Lu (1992), it is easy to see

$$\limsup_{T \rightarrow \infty} \sup_{0 < t \leq T} \sup_{t \leq s \leq T} |G(s) - G(s-t)| / d(T, t) \leq 1 \quad \text{a. s.} \tag{19}$$

On the other hand, using a law of iterated logarithm for stable processes in random scenery (see Khoshnevisan and Lewis, 1998), we get

$$\limsup_{T \rightarrow \infty} \sup_{0 < t \leq T} |G(T) - G(T-t)| / d(T, t) \geq \limsup_{T \rightarrow \infty} G(T) / (T^{\delta} (\gamma^{-1} \log \log T)^{1/\beta}) = 1 \quad \text{a. s.},$$

together with Eq. (19), we obtain our results immediately.

A NOTE

Let $\mathbf{B} = \{B(t) : t \geq 0\}$ and $\mathbf{W} = \{W(x) : x \in \mathbf{R}\}$ be independent real-valued standard Brownian motions with $B(0) = W(0) = 0$. Now define the process $\mathbf{G}' = \{G'(t) : t \geq 0\}$, which will be called Brownian motion in Brownian scenery, by

$$G'(t) = \int_{\mathbf{R}} L'(t, x) dW(x) \quad \forall t \geq 0,$$

where $\{L'(t, x) : t \geq 0, x \in \mathbf{R}\}$ is the jointly continuous version of the local time process of \mathbf{B} . Then by the result of Csáki et al. (1999), we get the following corollary immediately.

Corollary 7 If we replace \mathbf{G} by \mathbf{G}' in Eqs. (3)-(10), then the above results hold with $\gamma = (81/32)^{1/3}$.

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