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PERSISTENCE EXPONENT FOR RANDOM PROCESSES IN BROWNIAN SCENERY

FABIENNE CASTELL, NADINE GUILLOTIN-PLANTARD, AND FRÉDÉRIQUE WATBLED

ABSTRACT. We consider the one-sided exit problem for random processes in Brownian scenery, that is the asymptotic behaviour for large T, of the probability

where

$$\Delta_t = \int_{\mathbb{R}} L_t(x) \, dW(x)$$

 $\mathbb{P}\Big[\sup_{t\in[0,T]}\Delta_t\leq 1\Big]$

Here $W = \{W(x); x \in \mathbb{R}\}$ is a two-sided standard real Brownian motion and $\{L_t(x); x \in \mathbb{R}, t \geq 0\}$ is the local time of some self-similar random process with stationary increments, independent from the process W.

1. INTRODUCTION

Let $W = \{W(x); x \in \mathbb{R}\}$ be a standard two-sided real Brownian motion and $Y = \{Y(t), t \ge 0\}$ be a real-valued self-similar process of index $\gamma \in (0, 2)$ (i.e. for any c > 0, $\{Y(ct), t \ge 0\} \stackrel{(d)}{=} \{c^{\gamma}Y(t), t \ge 0\}$), with stationary increments. When it exists, we will denote by $\{L_t(x); x \in \mathbb{R}, t \ge 0\}$ a version of the local time of the process $\{Y(t); t \ge 0\}$. The process L satisfies the occupation density formula: for each bounded measurable function $f : \mathbb{R} \to \mathbb{R}$ and for each $t \ge 0$,

$$\int_0^t f(Y(s))ds = \int_{\mathbb{R}} f(x)L_t(x)dx.$$
(1)

The processes W and Y are defined on the same probability space and are assumed to be independent. We consider the random process in Brownian scenery $\{\Delta_t; t \ge 0\}$ defined as

$$\Delta_t = \int_{\mathbb{R}} L_t(x) \, dW(x).$$

The process Δ is itself a *h*-self-similar process with stationary increments, with

$$h := 1 - \frac{\gamma}{2}.$$

This process can be seen as a mixture of Gaussian processes, but it is neither Gaussian nor Markovian. It appeared independently in the mathematical literature (see [13],[3]), and in the physics literature (see [17]) where it was originally introduced to model diffusion in layered white-noise velocity field. Note indeed that Δ_t can formally be written as $\int_0^t \dot{W}(Y(s))ds$. Thus, it describes the horizontal motion of a tracer particle in a (1+1)-dimensional medium, where the motion of the particle along the vertical direction y is described by the process Y, while along the horizontal direction, the particle is driven by the white noise velocity field $\dot{W}(y)$, that depends only on the vertical coordinate y. The process { $\Delta_t; t \geq 0$ } provides a simple example

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of anomalous super-diffusion, being for instance of order $t^{3/4}$ when Y is a Brownian motion. In this paper, we are interested in the *persistence probability* of the process Δ , i.e. the asymptotic behaviour of

$$\mathcal{F}(T) := \mathbb{P}\Big[\sup_{t \in [0,T]} \Delta_t \le 1\Big]$$

as $T \to +\infty$. The study of the one-sided supremum process of random processes is a classical issue in probability. One usually gets polynomial decay of the persistence probability: $\mathcal{F}(T) \approx$ $T^{-\theta}$ for a non-negative θ , often called *persistence exponent*, or *survival exponent*. Classical examples where this exponent can be computed, include random walks or Lévy processes, and we refer the reader to the recent survey paper [1] for an account of models where the persistence exponent is known. However, there are relevant physical situations where this exponent remains unknown (see for instance [15]). The persistence exponent of the Brownian motion in Brownian scenery was studied by Redner [19, 20], and Majumdar [16]. Based on physical arguments, numerical simulations and analogy with Fractional Brownian Motion, Redner and Majumdar conjectured the value of the persistence exponent. In [6], their conjecture was proved up to logarithmic factors, when the process Y is a stable Lévy process with index $\delta \in (1, 2]$. The proof of [6] depends heavily on the increments independence of the process Y and the question raises if it is possible to compute the persistence exponent without it. The aim of this paper is to answer this question, and to provide assumptions on Y allowing to compute the persistence exponent of the random process in Brownian scenery.

- (H1): There exists a version $\{L_t(x); x \in \mathbb{R}, t \ge 0\}$ of the local time of Y;
- (H2): Y is a self-similar process of index $\gamma \in (0, 2)$;
- (H3): Y has stationary increments;
- (H4): Let $V_1 := \int L_1^2(x) dx$ be the self-intersection local time of Y. There exist a real number $\alpha > 1$, and positive constants C, c such that for any $t \ge 0$,

$$\mathbb{P}\left[V_1 \ge t\right] \le C \exp(-ct^{\alpha});$$

(H5): There exist a real number $\beta > 0$, and positive constants C, c such that for any t > 0,

$$\mathbb{P}\left[V_1 \le t\right] \le C \exp\left(-ct^{-\beta}\right).$$

Our main result is the following one.

Theorem 1. Assume (H1) to (H5) hold. There exists a constant c > 0, such that for large enough T,

$$T^{-\gamma/2}(\ln T)^{-c} \le \mathbb{P}\Big[\sup_{t \in [0,T]} \Delta_t \le 1\Big] \le T^{-\gamma/2}(\ln T)^{+c}.$$
 (2)

The paper is organized as follows. Section 2 provides three examples of processes satisfying (H1) to (H5), including the stable Lévy process with index $\delta \in (1, 2]$, and thus generalizing the result of [6]. Section 3 states some useful properties of the process Δ . Section 4 is devoted to the proof of Theorem 1.

2. Some examples

2.1. A sufficient condition for (H5). In this section we assume that (H1) holds and we provide a sufficient condition on $\{Y(t), t \ge 0\}$ allowing to check (H5). Let us first compare V_1 with $\max_{x \in [0,1]} |Y(x)|$. By noting that

 $s \in [0,1]$

$$1 = \int_{\{|x| \le \max_{s \in [0,1]} |Y(s)|\}} L_1(x) dx \le V_1^{1/2} \sqrt{2 \max_{s \in [0,1]} |Y(s)|},$$

we deduce that

$$\frac{1}{2\max_{s\in[0,1]}|Y(s)|} \le V_1.$$
(3)

Lemma 2.1. Assume that (H1) holds and that there exist positive constants C and c, and a real number $\beta > 0$ such that for all t > 0,

$$\mathbb{P}\left[\max_{s\in[0,1]}|Y(s)|\geq t\right]\leq C\exp\left(-ct^{\beta}\right)\,.$$

Then, (H5) holds true.

Proof. It is a direct consequence of (3) that for all t > 0,

$$\mathbb{P}\left[V_1 \le t\right] \le \mathbb{P}\left[\max_{s \in [0,1]} |Y(s)| \ge \frac{1}{2t}\right] \le C \exp(-c2^{-\beta}t^{-\beta}).$$

2.2. Stable Lévy processes. A process $Y = \{Y(t), t \ge 0\}$ is a strictly stable Lévy process with index $\delta \in (1, 2]$ if it is a process with stationary and independent increments, such that Y(0) = 0 and for any $t \ge 0$ and $u \in \mathbb{R}$,

$$\mathbb{E}[e^{iuY(t)}] = \exp\left\{-c|u|^{\delta}t\left(1+i\zeta\,\operatorname{sgn}(u)\tan(\pi\delta/2)\right)\right\},\tag{4}$$

where $\zeta \in [-1, 1]$, and c is a positive scale parameter.

Lemma 2.2. The stable Lévy process with index $\delta \in (1, 2]$ satisfies (H1) to (H5) with $\gamma = \frac{1}{\delta}$, $\alpha = \delta$, and $\beta = \frac{\delta}{2\delta - 1}$.

Proof. It is immediate that the Lévy process satisfies (H2) and (H3). A continuous version of the local time exists since $\delta > 1$ (this was proved by Boylan [4]), so it satisfies (H1). Moreover, Corollary 5.6 in [14] states that there exist positive constants C and ξ s.t. for every t > 0,

$$\mathbb{P}[V_1 \ge t] \le C e^{-\xi t^{\delta}},$$

which gives (H4). To prove (H5) we let $V_t := \int L_t^2(x) dx$ be the self-intersection local time of Y up to time t for $t \ge 0$, and we show that there exists l > 0 such that

$$\mathbb{P}\left[V_t \le 1\right] \le e^{-lt} \text{ for } t \text{ large enough.}$$
(5)

As $\{V_t, t \ge 0\}$ is self-similar of index $\frac{2\delta-1}{\delta}$, (5) implies that for every positive ε small enough,

$$\mathbb{P}[V_1 \le \varepsilon] = \mathbb{P}[V_{\varepsilon^{-\frac{\delta}{2\delta - 1}}} \le 1] \le e^{-l\varepsilon^{-\frac{\delta}{2\delta - 1}}}$$

which gives (H5). To prove (5) we show that the function defined on \mathbb{R}^+ by

$$f(t) = \log \mathbb{P}[V_t \le 1]$$

is subadditive. Let us fix s, t in $[0, +\infty)$. We consider the process $Y^{(s)} := \{Y_u^{(s)}, u \ge 0\}$ defined by

$$Y_u^{(s)} = Y_{u+s} - Y_s$$

its local time $\{L_u^{(s)}(x); x \in \mathbb{R}, u \ge 0\}$, and its self-intersection local time $\{V_u^{(s)}, u \ge 0\}$. Note that

$$L_{t+s}(x) = L_s(x) + L_t^{(s)}(x - Y_s).$$

Hence

$$V_{t+s} \ge V_s + V_t^{(s)}.$$

The process Y being a Lévy process, $V_t^{(s)}$ and V_s are independent and $V_t^{(s)}$ has the same law as V_t . Therefore

$$\mathbb{P}[V_{t+s} \le 1] \le \mathbb{P}[V_s \le 1; V_t^{(s)} \le 1] = \mathbb{P}[V_s \le 1]\mathbb{P}[V_t \le 1].$$

Thus f is subadditive, which implies that $\frac{f(t)}{t}$ converges, as $t \to +\infty$, towards a limit $-l = \inf_{t>0} \frac{f(t)}{t}$ (see [11]). It remains to show that l is strictly positive. As -l is less or equal than f(1), it is enough to show that $\mathbb{P}[V_1 > 1] > 0$. But thanks to (3) and Proposition 10.3 of [9],

$$\mathbb{P}[V_1 > 1] \ge \mathbb{P}\left[\max_{s \in [0,1]} |Y(s)| < \frac{1}{2}\right] > 0,$$

and this concludes the proof of (5).

2.3. Fractional Brownian motion. The Fractional Brownian motion of Hurst index $H \in (0,1)$ is the real centered Gaussian process $\{B_H(t), t \ge 0\}$ with covariance function

$$\mathbb{E}[B_H(t)B_H(s)] = \frac{1}{2}(t^{2H} + s^{2H} - |t - s|^{2H}).$$

Lemma 2.3. The Fractional Brownian motion with Hurst index $H \in (0,1)$ satisfies (H1) to (H5) with $\gamma = H$, $\alpha = 1/H$, $\beta = 2$.

Proof. It follows readily from the definition that B_H is self-similar with index H and has stationary increments. The existence of a jointly continuous version of its local time process for $H \in (0,1)$ is a classical fact (see for instance paragraph 22 in [10]). Theorem 1 of [12] asserts that there exists $\lambda_0 > 0$ such that for every $0 < \lambda < \lambda_0$,

$$\mathbb{E}\left[e^{\lambda V_1^{1/H}}\right] < \infty \,, \tag{6}$$

which implies (H4) with $\alpha = 1/H$. Finally, (H5) follows from Lemma 2.1 and Fernique's estimation ([8], Theorem 4.1.1): there exists $c_F > 0$ such that for any $x \ge \sqrt{5}$,

$$\mathbb{P}[\max_{s \in [0,1]} |B_H(s)| \ge c_F x] \le 10 \int_x^{+\infty} e^{-\frac{v^2}{2}} dv.$$
(7)

This implies (H5) with $\beta = 2$.

2.4. Iterated Brownian motion. Let $\{B(x), x \in \mathbb{R}\}$ be a two-sided real standard Brownian motion, and let $\{\tilde{B}(t), t \ge 0\}$ be a real standard Brownian motion, independent of $\{B(x), x \in \mathbb{R}\}$. The iterated Brownian motion is the process $\{Y(t), t \ge 0\}$ defined by

$$Y(t) = B(B(t)) \,.$$

Lemma 2.4. The iterated Brownian motion satisfies (H1) to (H5) with $\gamma = 1/4$, and $\alpha = \beta = 4/3$.

Proof. The self-similarity and increments stationarity of the iterated Brownian motion are direct consequences of the self-similarity and increments independence of the Brownian motion. The existence and joint continuity of the local times of iterated Brownian motion were proved in

[5, 7]. Let us prove (H5). For x > 0,

$$\mathbb{P}\left[\max_{s\in[0,1]}|Y(s)| \ge x\right] \le \mathbb{P}\left[\max_{s\in[0,1]}\left|\tilde{B}(s)\right| \ge x^{2/3}\right] + \mathbb{P}\left[\max_{|u|\le x^{2/3}}|B(u)|\ge x\right] \\
\le \mathbb{P}\left[\max_{s\in[0,1]}\left|\tilde{B}(s)\right| \ge x^{2/3}\right] + \mathbb{P}\left[\max_{u\in[0,x^{2/3}]}|B(u)|\ge x\right] \\
\le 2\mathbb{P}\left[\max_{s\in[0,1]}|B(s)|\ge x^{2/3}\right] \\
\le 4\mathbb{P}\left[\max_{s\in[0,1]}B(s)\ge x^{2/3}\right] = 4\mathbb{P}\left[|B(1)|\ge x^{2/3}\right].$$
(8)

This proves (H5) with $\beta = 4/3$ using Lemma 2.1.

To prove (H4), we use the uniform norm on the local times of iterated Brownian motion proved in Lemma 4 of [21]: there exists a constant K > 0 such that for any even integer n,

$$\sup_{x \in \mathbb{R}} \mathbb{E}\left[(L_1(x))^n \right] \le K^n (n!)^{3/4} \,. \tag{9}$$

Since $\int_{\mathbb{R}} L_1(x) dx = 1$, Hölder's inequality implies that for any integer n,

$$V_1^n = \left(\int_{\mathbb{R}} L_1(x)^2 \, dx\right)^n \le \int_{\mathbb{R}} L_1(x)^{n+1} \, dx = \int_{-\max_{s \in [0,1]} |Y(s)|}^{\max_{s \in [0,1]} |Y(s)|} L_1(x)^{n+1} \, dx$$

Therefore,

$$\mathbb{E}\left[V_{1}^{n}\right] \leq \int_{\mathbb{R}} \mathbb{E}\left[L_{1}(x)^{n+1}\mathbf{1}_{|x|\leq\max_{s\in[0,1]}|Y(s)|}\right] dx$$

$$\leq \int_{\mathbb{R}} \sqrt{\mathbb{E}\left[L_{1}(x)^{2(n+1)}\right]} \sqrt{\mathbb{P}\left[\max_{s\in[0,1]}|Y(s)|\geq|x|\right]} dx$$

$$\leq K^{n+1} \left((2n+2)!\right)^{3/8} \int_{\mathbb{R}} \sqrt{\mathbb{P}\left[\max_{s\in[0,1]}|Y(s)|\geq|x|\right]} dx, \qquad (10)$$

where we used (9) in the last inequality. By (8), the integral in (10) is finite. We deduce then from Stirling's formula that there exits a constant C > 0 such that for any integer n,

$$\mathbb{E}\left[V_1^n\right] \le C^n n^{\frac{3}{4}n}$$

Hence, for any t > 0, and any integer n,

$$\mathbb{P}\left[V_1 \ge t\right] \le t^{-n} C^n n^{\frac{3}{4}n} \,.$$

Optimizing over the values of n leads to take $n = \left[e^{-1}\left(\frac{t}{C}\right)^{4/3}\right]$, so that $t^{-n}C^n n^{\frac{3}{4}n} \simeq \exp\left(-\frac{3}{4e}\left(\frac{t}{C}\right)^{4/3}\right)$ for large t. This proves that the iterated Brownian motion satisfies (H4) with $\alpha = 4/3$.

3. Auxiliary statements on $(\Delta_t, t \ge 0)$

For a certain class of stochastic processes $\{X_t; t \ge 0\}$ (to be specified below), Molchan [18] proved that the asymptotic behavior of

$$\mathbb{P}\Big[\sup_{t\in[0,T]}X_t\leq 1\Big]$$

is related to the quantity

$$I(T) := \mathbb{E}\left[\left(\int_0^T e^{X_t} dt\right)^{-1}\right].$$

We refer to [2] where the relationship between both quantities is clearly explained as well as the heuristics.

Theorem 3.1 (Statement 1, [18]). Let $\{X_t; t \ge 0\}$ be a continuous process, self-similar with index $\gamma > 0$, with stationary increments s.t. for every $\theta > 0$,

$$\mathbb{E}\Big[\exp\left(\theta \max_{t\in[0,1]}|X_t|\right)\Big] < +\infty$$

Then, as $T \to +\infty$,

$$\mathbb{E}\left[\left(\int_0^T e^{X_t} dt\right)^{-1}\right] = HT^{-(1-\gamma)} \Big(\mathbb{E}\Big[\max_{t\in[0,1]} X_t\Big] + o(1)\Big).$$

By applying this result to our random process Δ we get

Proposition 3.2. Assume (H1) to (H4) hold. For any $\gamma \in (0, 2)$, as $T \to +\infty$,

$$\mathbb{E}\left[\left(\int_0^T e^{\Delta_t} dt\right)^{-1}\right] = \left(1 - \frac{\gamma}{2}\right) T^{-\gamma/2} \Big(\mathbb{E}\Big[\max_{t \in [0,1]} \Delta_t\Big] + o(1)\Big).$$

Before proving Proposition 3.2 we first establish three useful inequalities concerning the process $\{\Delta_t, t \ge 0\}$. First we show that the process satisfies the Kolmogorov-Centsov criterion, which allows us to assume that its sample paths are continuous (Lemma 3.3). Then we show that the process satisfies a maximal inequality (Lemma 3.4), and next we provide an exponential upper tail bound for Δ_1 (Lemma 3.5).

Lemma 3.3. Assume (H1) to (H4) hold. Then for every $a \ge 1$,

$$(\mathbb{E}|\Delta_t - \Delta_s|^a)^{1/a} = C(a)|t - s|^h, \ t, s \ge 0$$
(11)

where $C(a) \leq ca^{\nu}$ with $\nu := \frac{1}{2} \left(1 + \frac{1}{\alpha}\right)$. In particular the process satisfies the Kolmogorov-Centsov criterion of continuity.

Proof. The increments of the process Δ being stationary, by self-similarity, we have for every $t, s \geq 0$,

$$\mathbb{E}[|\Delta_t - \Delta_s|^a] = |t - s|^{ha} \mathbb{E}[|\Delta_1|^a] \leq |t - s|^{ha} \mathbb{E}[|\Delta_1|^{2([a]+1)}]^{a/(2([a]+1))}$$

From the formula of the even moments of a centered Gaussian law, we can derive the even moments of the random variable Δ_1 , namely, for any $m \in \mathbb{N}$,

$$\mathbb{E}[\Delta_1^{2m}] = \mathbb{E}\Big[\mathbb{E}[\Delta_1^{2m}|Y]\Big] = \mathbb{E}[V_1^m] \ \frac{(2m)!}{2^m m!}.$$

First of all, from Stirling's formula, for m large enough, we have

$$\frac{(2m)!}{2^m m!} \le C \left(\frac{2}{e}\right)^m m^m$$

for some constant C > 0. Moreover, we deduce from (H4) that

$$\mathbb{E}[V_1^m] = \int_0^{+\infty} m x^{m-1} \mathbb{P}[V_1 \ge x] dx$$

$$\leq \int_0^{+\infty} C m x^{m-1} \exp(-cx^\alpha) dx = \frac{Cm}{\alpha} c^{-m/\alpha} \Gamma\left(\frac{m}{\alpha}\right).$$
(12)

It is now easy to derive (11). In particular taking $a > \frac{1}{h}$ we get the Kolmogorov-Centsov criterion of continuity.

Lemma 3.4. Assume (H1) holds. Let $T, x \ge 0$. Then

$$\mathbb{P}[\max_{s \in [0,T]} \Delta_s \ge x | Y] \le 2\mathbb{P}[\Delta_T \ge x | Y],\tag{13}$$

$$\mathbb{P}[\max_{s \in [0,T]} \Delta_s \ge x] \le 2\mathbb{P}[\Delta_T \ge x].$$
(14)

Proof. Conditionally to the process Y, the process $(\Delta_t)_{t\geq 0}$ is a centered Gaussian process on \mathbb{R} with covariance function

$$\mathbb{E}\left[\Delta_s \Delta_t | Y\right] = \int_{\mathbb{R}} L_s(x) L_t(x) dx.$$

Moreover for any $t \ge s \ge 0$,

$$\begin{split} \mathbb{E}\Big[\Delta_t^2|Y\Big] - \mathbb{E}\Big[\Delta_s^2|Y\Big] - \mathbb{E}\Big[(\Delta_t - \Delta_s)^2|Y\Big] &= \int_{\mathbb{R}} (L_t^2(x) - L_s^2(x) - (L_t(x) - L_s(x))^2)dx \\ &= \int_{\mathbb{R}} 2L_s(x)(L_t(x) - L_s(x))dx \ge 0, \end{split}$$

hence applying Proposition 2.2 in [14], we deduce the inequality (13). By integrating we obtain the maximal inequality (14). \Box

Lemma 3.5. Assume (H1) and (H4) hold. There exist C > 0 and $\delta > 0$ such that for any x > 0,

$$\mathbb{P}[\Delta_1 \ge x] \le C \exp(-\delta x^{\frac{2\alpha}{1+\alpha}}).$$
(15)

Proof. Conditionally to the process Y, the random variable Δ_1 is a real centered Gaussian variable with variance V_1 . For each $u \in \mathbb{R}$, let

$$\Phi(u) = \frac{1}{\sqrt{2\pi}} \int_u^{+\infty} e^{-\frac{s^2}{2}} ds.$$

Then for x > 0 and $\theta > 0$,

$$\mathbb{P}[\Delta_1 \ge x] = \int_0^{+\infty} \Phi(xz^{-1/2}) \mathbb{P}_{V_1}(dz) = \int_0^{x^{\theta}} \Phi(xz^{-1/2}) \mathbb{P}_{V_1}(dz) + \int_{x^{\theta}}^{+\infty} \Phi(xz^{-1/2}) \mathbb{P}_{V_1}(dz) \\ \le \Phi(x^{1-\frac{\theta}{2}}) + \mathbb{P}[V_1 \ge x^{\theta}],$$

where we used that the function $z \mapsto \Phi(xz^{-1/2})$ is non decreasing with values in [0, 1].

Using (H4) and the classical inequality

$$\Phi(u) \le \frac{1}{\sqrt{2\pi u}} e^{-u^2/2}$$
 for every $u > 0$,

and choosing $\theta = \frac{2}{1+\alpha}$, we obtain that for x large enough,

$$\mathbb{P}[\Delta_1 \ge x] \le \frac{x^{-\alpha/(1+\alpha)}}{\sqrt{2\pi}} e^{-\frac{1}{2}x^{2\alpha/(1+\alpha)}} + Ce^{-cx^{2\alpha/(1+\alpha)}} \le Ce^{-\delta x^{2\alpha/(1+\alpha)}},$$

with $\delta = \min(c, 1/2)$. This implies (15) for every x > 0.

Proof of Proposition 3.2. It follows easily from assumptions (H1), (H2), (H3), that the process $\{\Delta_t; t \ge 0\}$ is self-similar with index $h := 1 - \frac{\gamma}{2}$, with stationary increments. By Lemma 3.3 we can assume that it is continuous. Hence, it is enough to prove that for every $\theta > 0$,

$$\mathbb{E}\Big[\exp\left(\theta \max_{t\in[0,1]} |\Delta_t|\right)\Big] < +\infty.$$
(16)

Let $\theta > 0$. We have

$$\mathbb{E}\Big[\exp\left(\theta \max_{t\in[0,1]} |\Delta_t|\right)\Big] = \int_0^\infty \mathbb{P}\Big[\exp\left(\theta \max_{t\in[0,1]} |\Delta_t|\right) \ge \lambda\Big] d\lambda$$
$$\leq 2 + \int_2^\infty \mathbb{P}\Big[\max_{t\in[0,1]} |\Delta_t| \ge \frac{\ln(\lambda)}{\theta}\Big] d\lambda.$$

Since the process $\{\Delta_t; t \ge 0\}$ is symmetric and satisfies the maximal inequality (14) of Lemma 3.4,

$$\mathbb{P}\Big[\max_{t\in[0,1]} |\Delta_t| \ge \frac{\ln(\lambda)}{\theta}\Big] \le 2 \mathbb{P}\Big[\max_{t\in[0,1]} \Delta_t \ge \frac{\ln(\lambda)}{\theta}\Big]$$
$$\le 4 \mathbb{P}[\Delta_1 \ge (\ln\lambda)/\theta].$$

We apply the inequality (15) of Lemma 3.5, and since the function $\lambda \to \exp(-\delta((\ln \lambda)/\theta)^{2\alpha/(1+\alpha)})$ is integrable at infinity for any $\alpha > 1$, the assertion (16) follows.

4. Proof of Theorem 1

Aurzada's proof of the lower bound of $\mathbb{P}\left[\sup_{t\in[0,T)}B_H(t)\leq 1\right]$ (see [2]), rests on both following arguments: the Fractional Brownian motion satisfies the hypothesis of Theorem 3.1, and it satisfies the inequality (valid for *a* large enough)

$$(\mathbb{E}|B_H(t) - B_H(s)|^a)^{1/a} = C(a)|t - s|^H, \ t, s \ge 0$$
(17)

with $C(a) \leq ca^{\nu}$, for some c and $\nu > 0$. Our random process Δ satisfies the hypothesis of Theorem 3.1, and we showed in Lemma 3.3 that it satisfies the inequality (17). Therefore the proof of Aurzada ([2]) allows us to derive the lower bound in Theorem 1.

As in [18] and [2], the main idea in the proof of the upper bound in (2), is to bound I(T) from below by restricting the expectation to a well-chosen set of paths.

Conditionally to Y, the process $\{\Delta_t; t \ge 0\}$ is a centered Gaussian process such that for every $0 \le t < s$,

$$\mathbb{E}[\Delta_t \Delta_s | Y] = \int_{\mathbb{R}} L_t(x) L_s(x) \, dx \ge 0,$$
$$\mathbb{E}[\Delta_t(\Delta_s - \Delta_t) | Y] = \int_{\mathbb{R}} L_t(x) (L_s(x) - L_t(x)) \, dx \ge 0$$

since $t \to L_t(x)$ is a.s. increasing for all $x \in \mathbb{R}$. It follows then from Slepian's lemma, that for every $0 \le u < v < w$ and every real numbers a, b,

$$\mathbb{P}\left[\sup_{t\in[u,v]}\Delta_t \le a, \ \sup_{t\in[v,w]}\Delta_t \le b\Big|Y\right] \ge \mathbb{P}\left[\sup_{t\in[u,v]}\Delta_t \le a\Big|Y\right] \mathbb{P}\left[\sup_{t\in[v,w]}\Delta_t \le b\Big|Y\right]$$
(18)

$$\mathbb{P}\left[\sup_{t\in[u,v]}\Delta_t \le a, \sup_{t\in[v,w]}(\Delta_t - \Delta_v) \le b \middle| Y\right] \ge \mathbb{P}\left[\sup_{t\in[u,v]}\Delta_t \le a \middle| Y\right] \mathbb{P}\left[\sup_{t\in[v,w]}(\Delta_t - \Delta_v) \le b \middle| Y\right].$$
(19)

Let $a_T = (\ln T)^a$, where a > 0 will be chosen later; let $\alpha_T := \frac{\ln T}{1 + \ln T}$, which belongs to (0, 1)and set $\beta_T = \sqrt{V_{a_T}}$ where $V_{a_T} := \int_{\mathbb{R}} L^2_{a_T}(x) dx$. Let us define the random function

$$\phi(t) := \begin{cases} 1 & \text{for } 0 \le t < a_T, \\ 1 - \beta_T & \text{for } a_T \le t \le T, \end{cases}$$

which is Y-measurable. Clearly, we have

$$\mathbb{E}\left[\left(\int_{0}^{T} e^{\Delta_{t}} dt\right)^{-\alpha_{T}} \left|Y\right] \ge \left(\int_{0}^{T} e^{\phi(t)} dt\right)^{-\alpha_{T}} \mathbb{P}\left[\forall t \in [0, T], \Delta_{t} \le \phi(t) \left|Y\right]\right]$$

By Slepian's lemma (see (18)), we have

$$\mathbb{P}\Big[\forall t \in [0,T], \Delta_t \le \phi(t) \Big| Y\Big] \ge \mathbb{P}\Big[\forall t \in [0,a_T], \Delta_t \le 1 \Big| Y\Big] \mathbb{P}\Big[\forall t \in [a_T,T], \Delta_t \le 1 - \beta_T \Big| Y\Big].$$

Remark that

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$$\mathbb{P}\Big[\forall t \in [a_T, T], \Delta_t \leq 1 - \beta_T \Big| Y\Big] \geq \mathbb{P}\Big[\Delta_{a_T} \leq -\beta_T; \forall t \in [a_T, T], \Delta_t - \Delta_{a_T} \leq 1 \Big| Y\Big] \\
\geq \mathbb{P}\Big[\Delta_{a_T} \leq -\beta_T \Big| Y\Big] \mathbb{P}\Big[\forall t \in [a_T, T], \Delta_t - \Delta_{a_T} \leq 1 \Big| Y\Big],$$

by Slepian's lemma (see (19)). Note that

$$\mathbb{P}[\Delta_{a_T} \le -\beta_T | Y] = \Phi(\beta_T V_{a_T}^{-1/2}) = \Phi(1).$$

Moreover, it is easy to check that when T goes to infinity,

$$\int_0^T e^{\phi(t)} dt = O(a_T + Te^{-\beta_T})$$

In the following C is a constant whose value may change but does not depend on T. Then we can write that for T large enough

$$\mathbb{E}\left[\left(\int_{0}^{T} e^{\Delta_{t}} dt\right)^{-\alpha_{T}} \left|Y\right] \ge C(a_{T} + Te^{-\beta_{T}})^{-\alpha_{T}} \mathbb{P}\left[\sup_{t \in [0, a_{T}]} \Delta_{t} \le 1 \left|Y\right] \mathbb{P}\left[\sup_{t \in [a_{T}, T]} (\Delta_{t} - \Delta_{a_{T}}) \le 1 \left|Y\right]\right].$$
(20)

Next we use the maximal inequality (13) of Lemma 3.4 to write

$$\mathbb{P}\Big[\sup_{t \in [0, a_T]} \Delta_t \le 1 | Y] = 1 - \mathbb{P}\Big[\sup_{t \in [0, a_T]} \Delta_t \ge 1 | Y] \ge 1 - 2\mathbb{P}\big[\Delta_{a_T} \ge 1 | Y\big] = \mathbb{P}[|Z| \le V_{a_T}^{-1/2} | Y]$$

where Z is a Gaussian variable $\mathcal{N}(0,1)$ independent of Y, from which we deduce that there exists a constant c > 0 such that

$$\mathbb{P}\Big[\sup_{t \in [0, a_T]} \Delta_t \le 1 | Y \Big] \ge c \min(V_{a_T}^{-1/2}, 1).$$
(21)

Injecting (21) into (20) we get that for T large enough,

$$\mathbb{P}\Big[\sup_{t\in[a_T,T]} (\Delta_t - \Delta_{a_T}) \le 1 | Y \Big] \le C \mathbb{E}\left[\left(\int_0^T e^{\Delta_t} dt \right)^{-\alpha_T} | Y \right] (a_T + T e^{-V_{a_T}^{1/2}})^{\alpha_T} \max(V_{a_T}^{1/2}, 1).$$

By integrating and using successively Hölder's inequality with $p_T = \frac{1}{\alpha_T}$, $\frac{1}{q_T} + \frac{1}{p_T} = 1$, Jensen's inequality, the inequality $(x + y)^{\alpha_T} \le x^{\alpha_T} + y^{\alpha_T}$ for x, y > 0, and Proposition 3.2, we get that for T large enough,

$$\begin{aligned} \mathbb{P}\Big[\sup_{t\in[a_{T},T]}(\Delta_{t}-\Delta_{a_{T}}) &\leq 1\Big] \\ &\leq C\mathbb{E}\left[\left(\int_{0}^{T}e^{\Delta_{t}}dt\right)^{-1}\right]^{1/p_{T}}\mathbb{E}\left[\left(a_{T}^{\alpha_{T}}\max(V_{a_{T}}^{1/2},1)+T^{\alpha_{T}}e^{-\alpha_{T}}V_{a_{T}}^{1/2}\max(V_{a_{T}}^{1/2},1)\right)^{q_{T}}\right]^{1/q_{T}} \\ &\leq CT^{-\frac{\gamma}{2p_{T}}}\|f_{1}+f_{2}\|_{q_{T}} \leq CT^{-\frac{\gamma}{2p_{T}}}(\|f_{1}\|_{q_{T}}+\|f_{2}\|_{q_{T}}), \end{aligned}$$
with

with

$$f_1 = a_T^{\alpha_T} \max(V_{a_T}^{1/2}, 1), \quad f_2 = T^{\alpha_T} e^{-\alpha_T V_{a_T}^{1/2}} \max(V_{a_T}^{1/2}, 1),$$

The lefthand term is greater than the quantity we want to bound from above, since by stationarity,

$$\mathbb{P}\Big[\sup_{t\in[a_T,T]}(\Delta_t - \Delta_{a_T}) \le 1\Big] = \mathbb{P}\Big[\sup_{t\in[0,T-a_T]}\Delta_t \le 1\Big] \ge \mathbb{P}\Big[\sup_{t\in[0,T]}\Delta_t \le 1\Big].$$

Concerning the righthand term, we recall that $\alpha_T = \frac{1}{p_T} = \frac{\ln T}{1 + \ln T}$ and $\frac{1}{q_T} = 1 - \alpha_T = \frac{1}{1 + \ln T}$. Hence, when T goes to infinity, $T^{-\frac{\gamma}{2p_T}} \leq CT^{-\frac{\gamma}{2}}$. Therefore,

$$\mathbb{P}\Big[\sup_{t\in[0,T]}\Delta_t \le 1\Big] \le CT^{-\frac{\gamma}{2}}(\|f_1\|_{q_T} + \|f_2\|_{q_T}).$$
(22)

It remains to prove that $||f_1||_{q_T}$ and $||f_2||_{q_T}$ are bounded by logarithmic terms.

$$\|f_1\|_{q_T} = a_T^{\alpha_T} \mathbb{E} \left[V_{a_T}^{q_T/2} \mathbf{1}_{V_{a_T} \ge 1} + \mathbf{1}_{V_{a_T} \le 1} \right]^{1/q_T} \le a_T^{\alpha_T} \left(\mathbb{E} \left[V_{a_T}^{q_T} \right]^{1/(2q_T)} + 1 \right).$$

By (H2), $V_{a_T} \stackrel{\mathcal{L}}{=} a_T^{2-\gamma} V_1$. Therefore,

$$\mathbb{E}\left[V_{a_{T}}^{q_{T}}\right]^{1/(2q_{T})} = a_{T}^{1-\frac{\gamma}{2}} \mathbb{E}\left[V_{1}^{q_{T}}\right]^{1/(2q_{T})}$$

Recall from (12) that,

$$\mathbb{E}[V_1^m] \le Cc^{-m/\alpha} \frac{m}{\alpha} \Gamma\left(\frac{m}{\alpha}\right) \text{ for every } m \in \mathbb{N},$$

so that using Stirling's formula, it is easy to show that for T large enough

$$\mathbb{E}[V_1^{q_T}]^{1/2q_T} \le C(\ln T)^{\frac{1}{2\alpha}}$$

We conclude that for T large enough

$$\|f_1\|_{q_T} \le C(\ln T)^{a(2-\frac{\gamma}{2})+\frac{1}{2\alpha}}.$$
(23)

Let us now turn our attention to $||f_2||_{q_T}$.

$$\|f_2\|_{q_T} \leq T^{\alpha_T} \mathbb{E} \left[e^{-2q_T \alpha_T V_{a_T}^{1/2}} \right]^{1/2q_T} \mathbb{E} \left[\max(V_{a_T}, 1)^{q_T} \right]^{1/2q_T} \\ \leq CT (\ln T)^{a(1-\frac{\gamma}{2})+\frac{1}{2\alpha}} \mathbb{E} \left[e^{-2q_T \alpha_T a_T^{1-\frac{\gamma}{2}} \sqrt{V_1}} \right]^{1/2q_T}.$$

Let us note $\lambda_T = 2q_T \alpha_T a_T^{1-\frac{\gamma}{2}} = 2(\ln T)^{1+a(1-\frac{\gamma}{2})}$. Then using (H5),

$$\mathbb{E}\left[e^{-\lambda_T V_1^{1/2}}\right] \le \int_0^{+\infty} \mathbb{P}\left[V_1 \le \frac{u^2}{\lambda_T^2}\right] e^{-u} du$$
$$\le C \int_0^{+\infty} e^{-c\frac{\lambda_T^{2\beta}}{u^{2\beta}}} e^{-u} du,$$

for some constants c and C. We perform the change of variable $u = \lambda_T^{\frac{2\beta}{1+2\beta}} v$ in the preceding integral. This yields

$$\begin{split} \mathbb{E}\left[e^{-\lambda_{T}V_{1}^{1/2}}\right] &\leq C\lambda_{T}^{2\beta/(1+2\beta)} \int_{0}^{+\infty} e^{-\lambda_{T}^{\frac{2\beta}{1+2\beta}} \left(v+cv^{-2\beta}\right)} dv \\ &\leq C\lambda_{T}^{2\beta/(1+2\beta)} \int_{1}^{+\infty} e^{-\lambda_{T}^{\frac{2\beta}{1+2\beta}} v} \, dv + C\lambda_{T}^{2\beta/(1+2\beta)} \int_{0}^{1} e^{-\lambda_{T}^{\frac{2\beta}{1+2\beta}} cv^{-2\beta}} \, dv \\ &\leq Ce^{-\lambda_{T}^{\frac{2\beta}{1+2\beta}}} + Ce^{-c\lambda_{T}^{\frac{2\beta}{1+2\beta}}} \\ &\leq Ce^{-c\lambda_{T}^{\frac{2\beta}{1+2\beta}}} \,, \end{split}$$

for some other constants c and C. This leads to

$$||f_2||_{q_T} \le CT(\ln T)^{a(1-\frac{\gamma}{2})+\frac{1}{2\alpha}} e^{-c(\ln T)^{\frac{2\beta}{1+2\beta}(1+a(1-\frac{\gamma}{2}))-1}}$$

We choose a such that $a(1-\frac{\gamma}{2}) > 1+\frac{1}{\beta}$. Then, $Te^{-c(\ln T)^{\frac{2\beta}{1+2\beta}(1+a(1-\frac{\gamma}{2}))-1}}$ remains bounded, and when T goes to infinity, we get

$$||f_2||_{q_T} \le C(\ln T)^c$$
, (24)

for some constant $c > 1 + \frac{1}{\beta} + \frac{1}{2\alpha}$. From (22), (23) and (24) we deduce that for T large enough,

$$\mathbb{P}\big[\sup_{t\in[0,T]}\Delta_t\leq 1\big]\leq CT^{-\frac{\gamma}{2}}(\ln T)^c,$$

with $c > \frac{\beta+1}{\beta} \frac{4-\gamma}{2-\gamma} + \frac{1}{2\alpha}$.

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