On the law of a triplet associated with the pseudo-Brownian bridge

Mathieu Rosenbaum, Marc Yor

On the law of a triplet associated with the pseudo-Brownian bridge

Mathieu Rosenbaum and Marc Yor

LPMA, University Pierre et Marie Curie (Paris 6)

October 27, 2013

Abstract

We identify the distribution of a natural triplet associated with the pseudo-Brownian bridge. In particular, for B a Brownian motion and T_1 its first hitting time of the level one, this remarkable law allows us to understand some properties of the process $(B_{uT_1}/\sqrt{T_1},\ u \leq 1)$ under uniform random sampling, a study started in [2].

Keywords: Brownian motion, pseudo-Brownian bridge, Bessel process, local time, hitting times, scaling, uniform sampling, Mellin transform.

1 Introduction and main results

Let $(B_t,\ t \geq 0)$ be a standard Brownian motion and $(T_a,\ a > 0)$ its first hitting times process:

$$T_a = \inf\{t,\ B_t > a\}.$$

Let U denote a uniform random variable on $[0,1]$, independent of B. In [2], the very interesting distribution of the random variable

$$\frac{B_{UT_a}}{\sqrt{T_a}}$$

is described. In particular, it is shown that this law does not depend on a, admits moments of any order and is centered. Furthermore, its density is quite remarkable, see [2] for details. In this work, our goal is to extend this study by giving some general properties of the rescaled Brownian motion up to its first hitting time of level 1, that is the process $(\alpha_u,\ u \leq 1)$ defined by

$$\alpha_u = \frac{B_{uT_1}}{\sqrt{T_1}}.$$

Here also, we will focus on the case of a uniform random sampling. Indeed, it is a very natural sampling scheme, which is known to lead to deep properties, see the seminal paper [4].

Let $(L_t,\ t \geq 0)$ be the local time of the Brownian motion at point 0 and $(\tau_l,\ l > 0)$ be the inverse local time process:

$$\tau_l = \inf\{t,\ L_t > l\}.$$

It turns out that our results on the process (α_u) will be deduced from properties obtained on the pseudo-Brownian bridge. This process was introduced in [1] and is defined by

$$(\frac{B_{u\tau_1}}{\sqrt{\tau_1}},\ u \leq 1).$$

The pseudo-Brownian bridge is equal to 0 at time 0 and time 1 and has the same quadratic variation as the Brownian motion. Thus, it shares some similarities with the Brownian bridge, which explains its name. We refer to [1] for more on this process. More precisely, we consider in this paper the triplet

$$(\frac{B_{U\tau_1}}{\sqrt{\tau_1}}, \frac{1}{\sqrt{\tau_1}}, L_{U\tau_1}),$$

where U is a uniform random variable on $[0,1]$, independent of B. Our main theorem, whose proof is given in Section 2 and where $\underset{\mathcal{L}}{=}$ denotes equality in law, is the following.

Theorem 1.1. *The following identity in law holds:*

$$(\frac{B_{U\tau_1}}{\sqrt{\tau_1}}, \frac{1}{\sqrt{\tau_1}}, L_{U\tau_1}) \underset{\mathcal{L}}{=} (\frac{1}{2}B_1, L_1, \Lambda),$$

with Λ a uniform random variable on $[0,1]$, independent of (B_1, L_1).

This result is quite remarkable in the sense that the distribution of the triplet is surprisingly simple. In particular, the marginal laws of the variables in the triplet are respectively Gaussian, absolute Gaussian and uniform distributions.

Let $(M_t,\ t \geq 0)$ be the one sided supremum of B and $(R_t,\ t \geq 0)$ be a three dimensional Bessel process starting from 0. We define the random variable γ by

$$\gamma = \sup\{t \geq 0,\ R_t = 1\}$$

and the process $(J_u,\ u \geq 0)$ by

$$J_u = \inf_{t \geq u} R_t.$$

Using Lévy's characterization of the reflecting Brownian motion and Pitman's representation of the three dimensional Bessel process, see [3] and for example [5], the aficionados of Brownian motion will easily deduce the variants of Theorem 1.1 stated in the following corollary, where U is a uniform random variable independent of B and R.

Corollary 1.1. *We have*

$$(\frac{B_{UT_1}}{\sqrt{T_1}}, \frac{1}{\sqrt{T_1}}, M_{UT_1}) \underset{\mathcal{L}}{=} (\Lambda L_1 - \frac{1}{2}|B_1|, L_1, \Lambda),$$
$$(\frac{R_{U\gamma}}{\sqrt{\gamma}}, \frac{1}{\sqrt{\gamma}}, J_{U\gamma}) \underset{\mathcal{L}}{=} (\Lambda L_1 + \frac{1}{2}|B_1|, L_1, \Lambda),$$

with Λ a uniform random variable on $[0,1]$, independent of (B_1, L_1).

We clearly see that the distribution of the couple (B_1, L_1) plays an essential role in the description of the laws of the triplets in Theorem 1.1 and Corollary 1.1. Recall that for $s > 0$, the density of (B_s, L_s) at point $(x, l) \in \mathbb{R} \times \mathbb{R}^+$ is given by

$$\frac{1}{\sqrt{2\pi s^3}}(|x| + l)\exp\Big(-\frac{(l+|x|)^2}{2s}\Big). \tag{1}$$

From this expression, we deduce for fixed time $s \geq 0$ the useful factorization:

$$(|B_s|, L_s) \underset{\mathcal{L}}{=} R_s(1-U, U).$$

Using elementary computations, this last expression enables to obtain interesting consequences of Theorem 1.1 and Corollary 1.1. For example, we give in the following corollary unexpected equalities in law and independence properties for some Brownian functionals.

Corollary 1.2. *The three triplets*

$$\Big(|B_{U\tau_1}|, L_{U\tau_1}, \frac{1+2|B_{U\tau_1}|}{\sqrt{\tau_1}}\Big),$$
$$\Big(M_{UT_1} - B_{UT_1}, M_{UT_1}, \frac{1+2(M_{UT_1}-B_{UT_1})}{\sqrt{T_1}}\Big),$$
$$\Big(R_{U\gamma} - J_{U\gamma}, J_{U\gamma}, \frac{1+2(R_{U\gamma}-J_{U\gamma})}{\sqrt{\gamma}}\Big)$$

are identically distributed and their common law is that of the triplet of independent variables

$$\Big(\frac{1}{2}\Big(\frac{1}{U}-1\Big), \Lambda, R_1\Big).$$

As a general comment, we should mention that we in fact derived several ways to obtain the preceding results. We present here the approach that we think is the most easily accessible.

The rest of the paper is organized as follows. Section 2 contains the proof of Theorem 1.1. The law of the random variable $B_{UT_1}/\sqrt{T_1}$ is investigated in Section 3 and some applications of Theorem 1.1 and Corollary 1.1 can be found in Section 4. Some additional remarks are gathered in an appendix.

2 Proof of Theorem 1.1

In this section, we give the proof of our main result, Theorem 1.1.

2.1 Step 1: Introducing a Mellin type transform

First remark that using the symmetry of the Brownian motion, Theorem 1.1 is equivalent to the equality in law:

$$\Big(\Big|\frac{B_{U\tau_1}}{\sqrt{\tau_1}}\Big|, \frac{1}{\sqrt{\tau_1}}, L_{U\tau_1}\Big) \underset{\mathcal{L}}{=} \Big(\frac{1}{2}|B_1|, L_1, \Lambda\Big). \tag{2}$$

We now introduce the function $\mathcal{M}_l(a, c)$ which is defined for $a \geq 0$, $c \geq 0$ and $l \leq 1$ by

$$\mathcal{M}_l(a,c) = \mathbb{E}\Big[\Big|\frac{B_{U\tau_1}}{\sqrt{\tau_1}}\Big|^a \Big|\frac{1}{\sqrt{\tau_1}}\Big|^c 1_{\{L_{U\tau_1} \leq l\}}\Big].$$

Using properties of the Mellin transform together with a monotone class argument, we easily get that proving (2) is equivalent to show the following equality:

$$\mathcal{M}_l(a,c) = l\mathbb{E}\big[|\frac{1}{2}B_1|^a|L_1|^c\big]. \tag{3}$$

2.2 Step 2: The Mellin transform of (B_1, L_1)

In this second step, we give the Mellin transform of the couple (B_1, L_1). Having an expression for this functional is of course helpful in order to prove Equality (3).

2.2.1 The result

The Mellin transform of the couple $(|B_1|, L_1)$ is given in the following proposition.

Proposition 2.1. *For $a > 0$ and $c > 0$, we have*

$$\mathbb{E}[|B_1|^a(L_1)^c] = \frac{\Gamma(1+a)\Gamma(1+c)}{2^{\frac{a+c}{2}}\Gamma(1+\frac{a+c}{2})},$$

where Γ denotes the classical gamma function.

2.2.2 Technical Lemma

Before starting the proof of Proposition 2.1, we give a useful lemma.

Lemma 2.1. *Let $\mathcal{E}$ and $\mathcal{E}'$ be two independent standard exponential variables, independent of B. The following equality in law holds:*

$$(|B_{2\mathcal{E}}|, L_{2\mathcal{E}}) \underset{\mathcal{L}}{=} (\mathcal{E}, \mathcal{E}').$$

This result is in fact quite well known. However, for sake of completeness we give its proof here.

Proof. To prove lemma 2.1, we compute the Mellin transform of the couple $(|B_{2\mathcal{E}}|, L_{2\mathcal{E}})$. Let $a > 0$ and $c > 0$. From the law of the couple $(|B_t|, L_t)$ obtained from (1), we get

$$\begin{aligned}
\mathbb{E}[|B_{2\mathcal{E}}|^a|L_{2\mathcal{E}}|^c] &= \frac{1}{2}\int_0^{+\infty} dt\mathrm{e}^{-t/2}\mathbb{E}[|B_t|^a|L_t|^c] \\
&= \frac{1}{2}\int_0^{+\infty} dt\mathrm{e}^{-t/2}\int_0^{+\infty}dx\int_0^{+\infty}dl\sqrt{\frac{2}{\pi t^3}}x^a l^c(x+l)\mathrm{e}^{-\frac{(l+x)^2}{2t}} \\
&= \int_0^{+\infty}dx\int_0^{+\infty}dl x^a l^c\int_0^{+\infty}\frac{dt}{\sqrt{2\pi t^3}}\mathrm{e}^{-t/2}(x+l)\mathrm{e}^{-\frac{(l+x)^2}{2t}}.
\end{aligned}$$

Using the expression of the density of an inverse Gaussian random variable, it is easily seen that the integral in (dt) is equal to $\exp\big(-(x+l)\big)$. Therefore, the remaining double integral is equal to $\Gamma(1+a)\Gamma(1+c)$. This gives the result since

$$\mathbb{E}[|\mathcal{E}|^a|\mathcal{E}'|^c] = \Gamma(1+a)\Gamma(1+c).$$

□

2.2.3 Proof of Proposition 2.1

We now give the proof of Proposition 2.1. First remark that, by scaling, we have

$$\sqrt{2\mathcal{E}}(|B_1|, L_1) \underset{\mathcal{L}}{=} (|B_{2\mathcal{E}}|, L_{2\mathcal{E}}).$$

Thus, using Lemma 2.1, we get

$$\mathbb{E}[(2\mathcal{E})^{\frac{a+c}{2}}]\mathbb{E}[|B_1|^a L_1^c] = \mathbb{E}[|\mathcal{E}|^a|\mathcal{E}'|^c] = \Gamma(1+a)\Gamma(1+c).$$

We eventually obtain the result since

$$\mathbb{E}[(2\mathcal{E})^{\frac{a+c}{2}}] = 2^{\frac{a+c}{2}}\Gamma(1+\frac{a+c}{2}).$$

2.3 Step 3: End of the proof of Theorem 1.1

According to Proposition 2.1, in order to prove Theorem 1.1, it suffices to show that

$$\mathcal{M}_l(a,c) = l\frac{\Gamma(1+a)\Gamma(1+c)}{2^{\frac{a+c}{2}}\Gamma(1+\frac{a+c}{2})2^a}. \tag{4}$$

To this purpose, let us write $\mathcal{M}_l(a,c)$ under the form

$$\begin{aligned}\mathcal{M}_l(a,c) &= \mathbb{E}\big[\frac{1}{\tau_1^{\frac{a+c}{2}+1}}\int_0^{\tau_1} ds|B_s|^a 1_{\{L_s\le l\}}\big]\\ &= \mathbb{E}\big[\frac{1}{\tau_1^{\frac{a+c}{2}+1}}\int_0^{\tau_l} ds|B_s|^a\big].\end{aligned}$$

Then, using the definition of the Gamma function and a change of variable, we easily get

$$\mathcal{M}_l(a,c) = \frac{1}{2^{\frac{a+c}{2}}\Gamma(1+\frac{a+c}{2})}\int_0^{+\infty} d\mu\mu^{1+a+c}\mathbb{E}\big[\int_0^{\tau_l} ds|B_s|^a e^{-\frac{\mu^2}{2}\tau_1}\big]. \tag{5}$$

Now recall that the process τ_l is a subordinator with Laplace exponent at point $\lambda \in \mathbb{R}^{+*}$ equal to $\sqrt{2\lambda}$, see [5]. Therefore,

$$\begin{aligned}\mathbb{E}\big[\int_0^{\tau_l} ds|B_s|^a e^{-\frac{\mu^2}{2}\tau_1}\big] &= \mathbb{E}\big[\int_0^{\tau_l} ds|B_s|^a e^{-\frac{\mu^2}{2}\tau_l-\mu(1-l)}\big]\\ &= e^{-\mu}\mathbb{E}\big[\int_0^{\tau_l} ds|B_s|^a e^{-\frac{\mu^2}{2}\tau_l-\mu(|B_{\tau_l}|-L_{\tau_l})}\big].\end{aligned}$$

Then, using Lévy's theorem, see [5], together with the optional stopping theorem, we get

$$\mathbb{E}\big[\int_0^{\tau_l} ds|B_s|^a e^{-\frac{\mu^2}{2}\tau_1}\big] = e^{-\mu}\mathbb{E}\big[\int_0^{+\infty} ds|B_s|^a e^{\mu(L_s-|B_s|)-\frac{\mu^2 s}{2}}1_{\{L_s\le l\}}\big].$$

With the help of the joint law of $(|B_s|, L_s)$ given in (1), we obtain that this last quantity is also equal to

$$e^{-\mu}\int_0^{+\infty} dx\int_0^l dm x^a e^{\mu(m-x)}2\int_0^{+\infty}\frac{ds}{\sqrt{2\pi s^3}}(m+x)e^{-\frac{(m+x)^2}{2s}}e^{-\frac{\mu^2 s}{2}}.$$

Using again the expression of the density of an inverse Gaussian random variable, we see that the integral in (ds) is equal to $\exp\big(-\mu(x+m)\big)$. Consequently, we obtain

$$\begin{aligned}\mathbb{E}\Big[\int_0^{\tau_l} ds|B_s|^a \mathrm{e}^{-\frac{\mu^2}{2}\tau_1}\Big] &= 2\mathrm{e}^{-\mu}\int_0^{+\infty} dx \int_0^l dm x^a \mathrm{e}^{-2\mu x} \\ &= 2l\mathrm{e}^{-\mu}\frac{\Gamma(1+a)}{(2\mu)^{a+1}}.\end{aligned}$$

Plugging this equality into Equation (5) gives

$$\mathcal{M}_l(a,c) = \frac{l}{2^{\frac{a+c}{2}}\Gamma(1+\frac{a+c}{2})}\frac{\Gamma(1+a)\Gamma(1+c)}{2^a},$$

which is the desired identity (4).

3 The properties of the law of α revisited

In this section, we focus on the random variable

$$\alpha = \frac{B_{UT_1}}{\sqrt{T_1}},$$

with U a uniform random variable on $[0,1]$ independent of B.

3.1 Two equivalent characterizations of the distribution of α

In [2], we describe in term of its density the law of the variable α. In particular, we show that this variable is centered. This characterization of the distribution of α is obtained thanks to the computation of the Mellin transform of the positive and negative parts of α. More precisely, we have for $m > 0$

$$\begin{aligned}\mathbb{E}[(\alpha_+)^m] &= \mathbb{E}[|N|^m]2\int_0^1 dz \frac{z^{1+m}}{(1+2z)} \\ \mathbb{E}[(\alpha_-)^m] &= \mathbb{E}[|\frac{N}{2}|^m](\frac{\log(3)}{2}).\end{aligned}$$

This leads to the following description of the law of α, which is of course equivalent to that given in [2].

Proposition 3.1. *Let N and Z be two independent random variables with N standard Gaussian and Z with density*

$$\frac{2z}{(1-\frac{\log(3)}{2})(1+2z)}1_{\{0<z<1\}}.$$

We have the following equalities in law:

$$\begin{aligned}(\alpha|\alpha>0) &\overset{\mathcal{L}}{=} |N|Z \\ (-\alpha|\alpha<0) &\overset{\mathcal{L}}{=} \frac{1}{2}|N| \\ \mathbb{P}[\alpha>0] &= 1-\frac{\log(3)}{2}.\end{aligned}$$

We wish to compare this characterization with the one obtained from Corollary 1.1, namely

$$\alpha \underset{\mathcal{L}}{=} \Lambda L_1 - \frac{1}{2}|B_1|. \tag{6}$$

We first note that the centering property is easily recovered since

$$\mathbb{E}[\Lambda L_1 - \frac{1}{2}|B_1|] = \frac{1}{2}\mathbb{E}[L_1 - |B_1|] = 0.$$

Moreover, the second moment can also be computed without difficulty. Indeed,

$$\mathbb{E}[(\Lambda L_1 - \frac{1}{2}|B_1|)^2] = \frac{1}{3}\mathbb{E}[L_1^2] - \frac{1}{2}\mathbb{E}[L_1|B_1|] + \frac{1}{4}.$$

Then, using for example the formula for the Mellin transform of the couple $(|B_1|, L_1)$ given in Proposition 2.1, we get

$$\mathbb{E}[(\Lambda L_1 - \frac{1}{2}|B_1|)^2] = \frac{1}{3}.$$

We now show that Proposition 3.1 and Equation (6) match. This is deduced from the following elementary description of the random variable A defined by

$$A = \Lambda U - \frac{1}{2}(1-U),$$

with Λ and U two independent uniform random variables on $[0,1]$.

Proposition 3.2. *Let Z be the random variable defined in Proposition 3.1 and V a uniform variable on $[0,1]$, independent of Z. We have*

$$(A|A>0) \underset{\mathcal{L}}{=} VZ$$
$$(-A|A<0) \underset{\mathcal{L}}{=} \frac{1}{2}V$$
$$\mathbb{P}[A>0] = 1 - \frac{\log(3)}{2}.$$

Proof. Let f be a positive measurable function. The density of A can be computed directly as follows. We have

$$\begin{aligned}\mathbb{E}[f(A)] &= \int_0^1 d\lambda \mathbb{E}\big[f\big(\lambda U - \frac{1}{2}(1-U)\big)\big] \\ &= \int_{\mathbb{R}} dx f(x) u(x),\end{aligned}$$

with

$$u(x) = \mathbb{E}\big[\frac{1}{U} 1_{\{\frac{1+2x}{3} \le U \le 1+2x\}}\big].$$

Now, note that on the one hand

- if $-\frac{1}{2} \le x \le 0$,

$$u(x) = \int_{\frac{1+2x}{3}}^{1+2x} \frac{du}{u} = \log(3),$$

- if $0 \leq x \leq 1$,

$$u(x) = \int_{\frac{1+2x}{3}}^{1} \frac{du}{u} = \log(\frac{3}{1+2x}),$$

- if $x \leq -\frac{1}{2}$ or $x > 1$,

$$u(x) = 0.$$

On the other hand, it is easily seen that the density of VZ is

$$\frac{1}{1-\frac{\log(3)}{2}} \log\big(\frac{3}{1+2x}\big).$$

This ends the proof of Proposition 3.2. □

Let us now start from Equation (6) and recover Proposition 3.1. From Equation (6), with our usual notation, we get

$$\alpha \underset{\mathcal{L}}{=} R_1 A.$$

Thus,

$$(\alpha | \alpha > 0) \underset{\mathcal{L}}{=} (R_1 A | A > 0).$$

From Proposition 3.2, we obtain

$$(R_1 A | A > 0) \underset{\mathcal{L}}{=} R_1 V Z \underset{\mathcal{L}}{=} |N| Z,$$

which gives the first result in Proposition 3.1. The two other results are proved similarly, with the help of Proposition 3.2.

3.2 A warning and some developments around Proposition 3.1

Since $1/\sqrt{T_1}$ is distributed as $|N|$, from Proposition 3.1, it may be tempting to think that

$$(B_{UT_1} | B_{UT_1} > 0)$$

is distributed as Z and is independent of T_1. However, this is wrong. This incited us to look at the joint law of $1/\sqrt{T_1}$ and B_{UT_1}. Indeed, although encoded in Corollary 1.1, it may deserve an explicit presentation which we give in the following theorem.

Theorem 3.1.

- *Let $p \geq 0$. For ϕ a positive measurable function, the following formulas hold:*

$$\mathbb{E}\big[\frac{1}{(\sqrt{T_1})^p} \phi(B_{UT_1}) 1_{\{B_{UT_1} > 0\}}\big] = c_p \int_0^1 db \phi(b) \big(1 - \frac{1}{(3-2b)^{p+1}}\big)$$
$$\mathbb{E}\big[\frac{1}{(\sqrt{T_1})^p} \phi(B_{UT_1}) 1_{\{B_{UT_1} < 0\}}\big] = c_p \int_{-\infty}^0 dx \phi(x) \big(\frac{1}{(1-2x)^{p+1}} - \frac{1}{(3-2x)^{p+1}}\big),$$

where $c_p = \mathbb{E}[|N|^p] = \frac{\Gamma(1+p)}{2^{p/2}\Gamma(1+p/2)}$.

- *The joint law of* $(1/\sqrt{T_1}, B_{UT_1})$ *admits the density* h *defined on* $\mathbb{R}^{+*} \times (-\infty, 1]$ *by*

$$h(z,x) = \sqrt{\frac{2}{\pi}}\big(\mathrm{e}^{-z^2/2} - \mathrm{e}^{-(3-2x)^2 z^2/2}\big)1_{\{z>0,\ 0<x<1\}} + \sqrt{\frac{2}{\pi}}\big(\mathrm{e}^{-(1-2x)^2 z^2/2} - \mathrm{e}^{-(3-2x)^2 z^2/2}\big)1_{\{z>0,\ x<0\}}.$$

- *The law of* B_{UT_1} *admits the density* k *defined on* $(-\infty, 1)$ *by*

$$k(x) = \frac{2(1-x)}{3-2x}1_{\{0<x<1\}} + \frac{2}{(1-2x)(3-2x)}1_{\{x<0\}}.$$

Remark that from Theorem 3.1, we get that

$$(1 - B_{UT_1} | B_{UT_1} > 0)$$

is distributed as Z. We now give the proof of Theorem 3.1.

Proof. To prove the first part of Theorem 3.1, we use the fact that

$$\mathbb{E}\big[\frac{1}{(\sqrt{T_1})^p}\phi(B_{UT_1})1_{\{B_{UT_1}>0\}}\big]$$

is equal to

$$\frac{1}{2^{p/2}\Gamma(1+p/2)}\int_0^{+\infty} d\mu\mu^{1+p}\mathbb{E}\big[\int_0^{T_1} ds\phi(B_s)1_{\{B_s>0\}}\mathrm{e}^{-\frac{\mu^2}{2}T_1}\big].$$

From Proposition 3.1 in [2] (or the computation of I_μ in the same paper), the above expectation is equal to

$$\int_0^1 db\phi(b)\frac{1}{\mu}(\mathrm{e}^{-\mu} - \mathrm{e}^{-\mu(3-2b)}).$$

Then, using Fubini's theorem and integrating in μ yields the first formula. The second formula is obtained in the same manner.

For the proof of Part 2 in Theorem 3.1, in order to obtain the density at point (z,x) in $\mathbb{R}^{+*} \times (0,1)$, $h(z,x)$, we note that from Part 1 of Theorem 3.1, the formula

$$\mathbb{E}\big[f(\frac{1}{\sqrt{T_1}}, B_{UT_1})1_{\{B_{UT_1}>0\}}\big] = \mathbb{E}[f(|N|, V)] - \int_0^1 \frac{db}{3-2b}\mathbb{E}\big[f\big(\frac{|N|}{3-2b}, b\big)\big],$$

with V a uniform variable on $[0,1]$ independent of N, holds for every function f of the form

$$f(z,b) = z^p\phi(b),$$

with ϕ some positive measurable function. An application of the monotone class theorem yields the validity of the above formula for every positive measurable function f. Then, a simple change of variables gives the first formula in Part 2. The second formula is proved likewise.

To prove Part 3 in Theorem 3.1, it suffices to take $p = 0$ in Part 1.

□

4 Applications

In this section, we give two applications of Theorem 1.1 and Corollary 1.1.

4.1 A family of centered Brownian functionals

In [2], we established that the variable H defined by

$$H = \frac{1}{T_1^{3/2}} \int_0^{T_1} ds B_s$$

admits moments of all orders and is centered. This centering property is equivalent to that of the random variable

$$\alpha = \frac{B_{UT_1}}{\sqrt{T_1}},$$

where U is a uniform random variable on $[0,1]$, independent of B. In fact, Theorem 1.1 and Corollary 1.1 enable to build families of centered functionals involving the Brownian motion and its first hitting time of level 1, the local time at point 0 and its inverse process, and the running maximum. We have the following theorem, in which H_1 is equal to H.

Theorem 4.1. *For any $p \geq 1$, the random variables H_p and H'_p are centered, with*

$$H_p = \frac{1}{T_1^{p/2+1}} \int_0^{T_1} ds \big((\frac{p+1}{2p^2} - 1) M_s^p + B_s M_s^{p-1} \big)$$

and

$$H'_p = \frac{1}{\tau_1^{p/2+1}} \int_0^{\tau_1} ds \big(\frac{p+1}{2p^2} L_s^p - |B_s| L_s^{p-1} \big).$$

Proof. First remark that from Lévy's equivalence theorem, H_p and H'_p have the same law. Then, obviously the expectation of H'_p is that of

$$(\frac{p+1}{2p^2}) \frac{L^p_{U\tau_1}}{(\sqrt{\tau_1})^p} - \frac{|B_{U\tau_1}| (L_{U\tau_1})^{p-1}}{(\sqrt{\tau_1})^p}.$$

From Theorem 1.1, this random variable has the same law as

$$(\frac{p+1}{2p^2}) \Lambda^p L_1^p - \frac{1}{2} |B_1| \Lambda^{p-1} L_1^{p-1}.$$

The expectation of this last quantity is equal to

$$\frac{1}{2p^2} \mathbb{E}[L_1^p] - \frac{1}{2p} \mathbb{E}[|B_1| L_1^{p-1}].$$

Using for example the Mellin transform of the couple $(|B_1|, L_1)$ given in Proposition 2.1, it is easily seen that this expression is equal to zero (one may also use the martingale property of $\frac{1}{p} L_t^p - |B_t| L_t^{p-1}$, see [5]). □

4.2 On the law of $R_{U\gamma}/\sqrt{\gamma}$

We now focus on the distribution of $R_{U\gamma}/\sqrt{\gamma}$. From Corollary 1.1, we know that

$$\frac{R_{U\gamma}}{\sqrt{\gamma}} \underset{\mathcal{L}}{=} \Lambda L_1 + \frac{1}{2}|B_1| \underset{\mathcal{L}}{=} R_1 A',$$

with

$$A' = \Lambda U + \frac{1}{2}(1-U).$$

There is the following description of the laws of A' and $R_{U\gamma}/\sqrt{\gamma}$.

Proposition 4.1. *The law of A' admits the density l given by*

$$l(a) = \log\big(\frac{1}{|2a-1|}\big)1_{\{0<a<1\}}.$$

Consequently, the law of $R_{U\gamma}/\sqrt{\gamma}$ admits the following density:

$$\sqrt{\frac{2}{\pi}}x^2 \int_1^{+\infty} dy y \exp(-\frac{x^2y^2}{2}) l(\frac{1}{y}).$$

Proof. The density of A' is obtained thanks to straightforward computations. Let f be a positive measurable function. Using the density of a three dimensional Bessel variable R_1, see [5], we have

$$\mathbb{E}[f(R_1A')] = \sqrt{\frac{2}{\pi}} \int_0^{+\infty} dr r^2 \mathrm{e}^{-r^2/2} \mathbb{E}[f(rA')].$$

Now, using the density of A', we get

$$\begin{aligned}
\mathbb{E}[f(R_1A')] &= \sqrt{\frac{2}{\pi}} \int_0^{+\infty} dr r^2 \mathrm{e}^{-r^2/2} \int_0^r \frac{dx}{r} l(\frac{x}{r}) f(x) \\
&= \int_0^{+\infty} dx f(x) \sqrt{\frac{2}{\pi}} \int_x^{+\infty} dr r \mathrm{e}^{-r^2/2} l(\frac{x}{r}) \\
&= \int_0^{+\infty} dx f(x) \sqrt{\frac{2}{\pi}} x^2 \int_1^{+\infty} dy y \mathrm{e}^{-x^2y^2/2} l(\frac{1}{y}).
\end{aligned}$$

□

5 Conclusion and future work

In this paper, we establish the law of a triplet associated with the pseudo-Brownian bridge. This process has been introduced in [1] and is defined as

$$\big(\frac{B_{u\tau_1}}{\sqrt{\tau_1}},\ u \le 1\big).$$

In particular, this enables us to understand in depth some properties of the random variable

$$\alpha = \frac{B_{UT_1}}{\sqrt{T_1}}$$

studied in [2]. In a forthcoming work, we intend to develop some consequences of the obtained results for the Brownian bridge and the Brownian meander.

Appendices

A A simple proof for the joint law of $(1/\sqrt{\tau_1}, L_{U\tau_1})$

The fact that

$$(\frac{1}{\sqrt{\tau_1}}, L_{U\tau_1}) \overset{\mathcal{L}}{=} (L_1, \Lambda)$$

can obviously be deduced from Theorem 1.1. However, interestingly, we can give a simple proof for this equality in law. Indeed, for $\lambda \geq 0$ and $l < 1$, we have

$$\begin{aligned}\mathbb{E}[\mathrm{e}^{-\lambda\tau_1} 1_{\{L_{U\tau_1} \leq l\}}] &= \mathbb{E}[\frac{1}{\tau_1}\int_0^{\tau_1} ds 1_{\{L_s \leq l\}} \mathrm{e}^{-\lambda\tau_1}] \\ &= \mathbb{E}[\frac{\tau_l}{\tau_1}\mathrm{e}^{-\lambda\tau_1}].\end{aligned}$$

Now, consider in general (τ_l) a subordinator and denote by ψ its Laplace exponent. Thus, we have

$$\begin{aligned}\mathbb{E}[\frac{\tau_l}{\tau_1}\mathrm{e}^{-\lambda\tau_1}] &= \mathbb{E}[\tau_l \int_0^{+\infty} dt \mathrm{e}^{-(t+\lambda)\tau_1}] \\ &= \int_0^{+\infty} dt \mathbb{E}[\tau_l \mathrm{e}^{-(t+\lambda)\tau_l}] \mathrm{e}^{-(1-l)\psi(t+\lambda)}.\end{aligned}$$

Using the fact that the Laplace exponent is differentiable on $\mathbb{R}^{+*}$, we get

$$\begin{aligned}\mathbb{E}[\frac{\tau_l}{\tau_1}\mathrm{e}^{-\lambda\tau_1}] &= \int_0^{+\infty} dt l\psi'(t+\lambda)\mathrm{e}^{-l\psi(t+\lambda)}\mathrm{e}^{-(1-l)\psi(t+\lambda)} \\ &= l\mathrm{e}^{-\psi(\lambda)}.\end{aligned}$$

This proves the independence of $1/\sqrt{\tau_1}$ and $L_{U\tau_1}$ and the fact that $L_{U\tau_1}$ is uniformly distributed. The equality in law

$$\frac{1}{\sqrt{\tau_1}} \overset{\mathcal{L}}{=} L_1$$

is easily obtained by scaling.

B On a one parameter family of random variables including α

In this section of the appendix, we consider the family of variables defined for $0 < c \leq 1$ by

$$\alpha_c = \Lambda L_1 - c|B_1|,$$

as an extension of our study of

$$\alpha \underset{\mathcal{L}}{=} \alpha_{1/2} \underset{\mathcal{L}}{=} B_{UT_1}/\sqrt{T_1}.$$

The variables α_c, although less natural than $\alpha_{1/2}$, enjoy some similar remarkable properties. Indeed, Proposition 3.2 and Proposition 3.1 admit the following extensions.

Proposition B.1. *Let $0 < c \leq 1$ and $C = 1/c$. Let Λ and U be two independent uniform variables on $[0,1]$ and*

$$A_c = \Lambda U - c(1-U).$$

We have

$$\begin{aligned}
(A_c | A_c > 0) &\underset{\mathcal{L}}{=} V Z_C \\
(-A_c | A_c < 0) &\underset{\mathcal{L}}{=} cV \\
\mathbb{P}[A_c > 0] &= 1 - c\log(1+C),
\end{aligned}$$

where V and Z_C are independent, with V uniform on $[0,1]$ and Z_C a random variable with density given by

$$\frac{C}{1 - c\log(1+C)} \frac{dzz}{(1+Cz)} 1_{\{0<z<1\}}.$$

Proposition B.2. *Let $0 < c \leq 1$. The following equalities in law hold.*

$$\begin{aligned}
(\alpha_c | \alpha_c > 0) &\underset{\mathcal{L}}{=} |N| Z_C \\
(-\alpha_c | \alpha_c < 0) &\underset{\mathcal{L}}{=} c|N| \\
\mathbb{P}[\alpha_c > 0] &= 1 - c\log(1+C).
\end{aligned}$$

Proof. To establish Proposition B.1, we simply compute the density of A_c. Proposition B.2 ensues since

$$\alpha_c \underset{\mathcal{L}}{=} R_1 A_c$$

and

$$R_1 V \underset{\mathcal{L}}{=} |N|,$$

with the same notation as previously. □

References

[1] P. Biane, J.-F. Le Gall, and M. Yor. Un processus qui ressemble au pont brownien. In *Séminaire de Probabilités XXI*, pages 270–275. Springer, 1987.

[2] R. Elie, M. Rosenbaum, and M. Yor. On the expectation of normalized brownian functionals up to first hitting times. *arXiv preprint arXiv:1310.1181*, 2013.

[3] J. W. Pitman. One-dimensional brownian motion and the three-dimensional bessel process. *Advances in Applied Probability*, 7:511–526, 1975.

[4] J. W. Pitman. Brownian motion, bridge, excursion, and meander characterized by sampling at independent uniform times. *Electronic Journal of Probability*, 4(11):1–33, 1999.

[5] D. Revuz and M. Yor. *Continuous martingales and Brownian motion*, volume 293. Springer, 1999.