

**SOME DISTRIBUTIONAL PROPERTIES  
OF THE BROWNIAN MOTION WITH A DRIFT  
AND AN EXTENSION OF P. LÉVY'S THEOREM**

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**Abstract.** The theorem proved by P. Lévy states that  $(\sup B - B, \sup B) \stackrel{\text{law}}{=} (|B|, L(B))$ . Here,  $B$  is the standard linear Brownian motion and  $L(B)$  is its local time in zero. In this paper, we present an extension of P. Lévy's theorem to the case of the Brownian motion with a (random) drift as well as to the case of conditionally Gaussian martingales. Besides, we give a simple proof of the equality  $2 \sup B^\lambda - B^\lambda \stackrel{\text{law}}{=} |B^\lambda| + L(B^\lambda)$  where  $B^\lambda$  is the Brownian motion with drift  $\lambda \in \mathbb{R}$ .

**Key words and phrases.** P. Lévy's theorem, local time, Brownian motion with a drift, conditionally Gaussian martingales, Skorokhod's lemma.

## 1 An Invariance Property of the Brownian Motion

1. Let  $B = (B_t)_{t \geq 0}$  be the standard linear Brownian motion and  $B^\lambda = (B_t^\lambda)_{t \geq 0}$  be the Brownian motion with a drift ( $B_t^\lambda = \lambda t + B_t$ ) where  $\lambda \in \mathbb{R}$ .

The classical theorem proved by P. Lévy (see [3], [9, ch.VI, §2, (2.3)]) states that

$$(\sup B - B, \sup B) \stackrel{\text{law}}{=} (|B|, L(B)), \quad (1)$$

i.e., the processes  $(\sup_{s \leq t} B_s - B_t, \sup_{s \leq t} B_s; t \geq 0)$  and  $(|B_t|, L_t(B); t \geq 0)$  have the same law. Here,  $L(B) = (L_t(B))_{t \geq 0}$  is the local time of  $B$  in zero:

$$L_t(B) = \lim_{\varepsilon \downarrow 0} \frac{1}{2\varepsilon} \int_0^t I(|B_s| \leq \varepsilon) ds \quad \text{a.s.}$$

It follows from (1) that

$$2 \sup B - B \stackrel{\text{law}}{=} |B| + L(B). \quad (2)$$

This “one-dimensional” property was extended in [10] to the case of the Brownian motion with a drift:

$$2 \sup B^\lambda - B^\lambda \stackrel{\text{law}}{=} |B^\lambda| + L(B^\lambda). \quad (3)$$

The paper [1] presents an extension of the “two-dimensional” P. Lévy’s result. Namely, it is proved in [1] that

$$(\sup B^\lambda - B^\lambda, \sup B^\lambda) \stackrel{\text{law}}{=} (|X^\lambda|, L(X^\lambda)). \quad (4)$$

Here,  $X^\lambda = (X_t^\lambda)_{t \geq 0}$  is the (strong) solution of the stochastic differential equation (SDE)

$$dX_t^\lambda = -\lambda \operatorname{sgn} X_t^\lambda dt + dB_t, \quad X_0^\lambda = 0 \quad (5)$$

and  $L(X^\lambda) = (L_t(X^\lambda))_{t \geq 0}$  is the local time of  $X^\lambda$  in zero.

It follows from (4) that

$$2 \sup B^\lambda - B^\lambda \stackrel{\text{law}}{=} |X^\lambda| + L(X^\lambda). \quad (6)$$

Equalities (3) and (6) taken together yield

**Theorem 1.** *For any  $\lambda \in \mathbb{R}$ ,*

$$|B^\lambda| + L(B^\lambda) \stackrel{\text{law}}{=} |X^\lambda| + L(X^\lambda). \quad (7)$$

It makes sense to give a straightforward proof of this invariance in distribution for  $|x| + L(x)$ ,  $x \in C(\mathbb{R}_+, \mathbb{R})$  when  $x = B^\lambda$  is replaced by  $x = X^\lambda$  since this property is of some interest by itself. Note that (6) + (7)  $\Rightarrow$  (3) and (3) + (7)  $\Rightarrow$  (6).

**P r o o f o f T h e o r e m 1.** It follows from P. Lévy’s theorem (1) that for any  $\lambda \in \mathbb{R}$  and  $t \geq 0$ ,

$$\mathbb{E} e^{\lambda B_t} = \mathbb{E} e^{\lambda(L_t(B) - |B_t|)}. \quad (8)$$

The following “conditional” version of equality (8) is the key point in the proof of Theorem 1: for any  $t \geq 0$ ,

$$\mathbb{E} \left[ e^{\lambda B_t} \middle| \mathcal{F}_t^{R(B)} \right] = \mathbb{E} \left[ e^{\lambda(L_t(B) - |B_t|)} \middle| \mathcal{F}_t^{R(B)} \right] \quad \text{a.s.} \quad (9)$$

Here,  $R(B) = (R_t(B))_{t \geq 0}$ ,  $R_t(B) = |B_t| + L_t(B)$  and  $\mathcal{F}_t^{R(B)} = \sigma(R_s(B); s \leq t)$ . Furthermore,

$$\mathbb{E} \left[ e^{\lambda B_t} \middle| \mathcal{F}_t^{R(B)} \right] = \frac{\operatorname{sh} \lambda R_t(B)}{\lambda R_t(B)}. \quad (10)$$

In order to prove (9) and (10), we first note that the symmetry considerations lead to the equality

$$\mathbb{E} \left[ e^{\lambda B_t} \middle| \mathcal{F}_t^{|B|} \right] = \frac{1}{2} \left( e^{\lambda |B_t|} + e^{-\lambda |B_t|} \right) \left( = \operatorname{ch} \lambda |B_t| \right)$$

where  $\mathcal{F}_t^{|B|} = \sigma(|B_s|; s \leq t)$ . This, combined with the inclusion  $\mathcal{F}_t^{R(B)} \subseteq \mathcal{F}_t^{|B|}$ , implies that

$$\mathbb{E} \left[ e^{\lambda B_t} \middle| \mathcal{F}_t^{R(B)} \right] = \frac{1}{2} \mathbb{E} \left[ e^{\lambda |B_t|} + e^{-\lambda |B_t|} \middle| \mathcal{F}_t^{R(B)} \right]. \quad (11)$$

It follows from [8] (see also [9, ch.VI, §3, (3.6)]) that the conditional distribution  $\operatorname{Law}(|B_t| | \mathcal{F}_t^{R(B)})$  is the uniform distribution on  $[0, R_t(B)]$ . Therefore,

$$\frac{1}{2} \mathbb{E} \left[ e^{\lambda |B_t|} + e^{-\lambda |B_t|} \middle| \mathcal{F}_t^{R(B)} \right] = \frac{1}{2\lambda R_t(B)} \left( e^{\lambda R_t(B)} - e^{-\lambda R_t(B)} \right).$$

This, together with (11), proves (10).

Similarly, we deduce that

$$\begin{aligned} \mathbb{E} \left[ e^{\lambda(L_t(B) - |B_t|)} \Big| \mathcal{F}_t^{R(B)} \right] &= e^{\lambda R_t(B)} \mathbb{E} \left[ e^{-2\lambda|B_t|} \Big| \mathcal{F}_t^{R(B)} \right] = \\ &= \frac{1}{2\lambda R_t(B)} \left( e^{\lambda R_t(B)} - e^{-\lambda R_t(B)} \right) = \frac{\text{sh } \lambda R_t(B)}{\lambda R_t(B)}. \end{aligned} \quad (12)$$

Combining (10) and (12), we obtain the desired equality (9).

We now turn to the proof of the invariance property (7). Let  $P_B$ ,  $P_{B^\lambda}$  and  $P_{X^\lambda}$  be the distributions of the processes  $B$ ,  $B^\lambda$  and  $X^\lambda$  on the canonical path space  $C(\mathbb{R}_+, \mathbb{R})$ . We will use  $P_B|_{\mathcal{F}_t}$ ,  $P_{B^\lambda}|_{\mathcal{F}_t}$  and  $P_{X^\lambda}|_{\mathcal{F}_t}$  to denote the restrictions of these measures to the  $\sigma$ -fields  $\mathcal{F}_t = \sigma(x_s; s \leq t)$ ,  $t \geq 0$  where  $(x_t)_{t \geq 0}$  is the canonical process on  $C(\mathbb{R}_+, \mathbb{R})$ .

It is well known (see, for example, [4, ch.7, §2]) that

$$\frac{d(P_{B^\lambda}|_{\mathcal{F}_t})}{d(P_B|_{\mathcal{F}_t})}(B) = e^{\lambda B_t - \frac{\lambda^2}{2}t}. \quad (13)$$

Taking into account *Tanaka's formula*,

$$|B_t| = \int_0^t \text{sgn } B_s dB_s + L_t(B)$$

(see [9, ch.VI, §1]), we obtain

$$\frac{d(P_{X^\lambda}|_{\mathcal{F}_t})}{d(P_B|_{\mathcal{F}_t})}(B) = e^{-\lambda \int_0^t \text{sgn } B_s dB_s - \frac{\lambda^2}{2}t} = e^{\lambda(L_t(B) - |B_t|) - \frac{\lambda^2}{2}t}. \quad (14)$$

We will use  $P_{R(B)}$ ,  $P_{R(B^\lambda)}$  and  $P_{R(X^\lambda)}$  to denote the images of  $P_B$ ,  $P_{B^\lambda}$  and  $P_{X^\lambda}$  under the map

$$C(\mathbb{R}_+, \mathbb{R}) \ni x \longmapsto R(x) \in C(\mathbb{R}_+, \mathbb{R})$$

where  $R_t(x) = |x_t| + L_t(x)$ . Set  $\mathcal{F}_t^R = \sigma(R_s(x); s \leq t)$ . It can easily be checked that ( $P_B$  - a.s.)

$$\frac{d(P_{R(B^\lambda)}|_{\mathcal{F}_t^R})}{d(P_{R(B)}|_{\mathcal{F}_t^R})}(R(B)) = \mathbb{E} \left[ \frac{d(P_{B^\lambda}|_{\mathcal{F}_t})}{d(P_B|_{\mathcal{F}_t})}(B) \Big| \mathcal{F}_t^{R(B)} \right] \quad (15)$$

and

$$\frac{d(P_{R(X^\lambda)}|_{\mathcal{F}_t^R})}{d(P_{R(B)}|_{\mathcal{F}_t^R})}(R(B)) = \mathbb{E} \left[ \frac{d(P_{X^\lambda}|_{\mathcal{F}_t})}{d(P_B|_{\mathcal{F}_t})}(B) \Big| \mathcal{F}_t^{R(B)} \right]. \quad (16)$$

Combining (13), (14) and (9), we see that the right-hand sides in (15) and (16) coincide ( $P_B$  - a.s.) and

$$\frac{d(P_{R(B^\lambda)}|_{\mathcal{F}_t^R})}{d(P_{R(B)}|_{\mathcal{F}_t^R})}(R(B)) = \frac{d(P_{R(X^\lambda)}|_{\mathcal{F}_t^R})}{d(P_{R(B)}|_{\mathcal{F}_t^R})}(R(B)) = e^{-\frac{\lambda^2}{2}t} \frac{\text{sh } \lambda R_t(B)}{\lambda R_t(B)}. \quad (17)$$

Therefore,  $P_{R(B^\lambda)} = P_{R(X^\lambda)}$ . This completes the proof of Theorem 1.

**2. Remark.** 1) It follows from (17) that  $R(B^\lambda) \stackrel{\text{law}}{=} R(B^{-\lambda})$  and  $R(X^\lambda) \stackrel{\text{law}}{=} R(X^{-\lambda})$ .

2) The results of [8] (see also [9, ch.VI, §3, (3.5)]) imply that the process  $2 \sup B - B$  has the same law as the *three-dimensional Bessel process*  $\text{Bes}(3)$  started at zero. The infinitesimal generator  $\mathcal{A}_\nu$  of the  $\nu$ -dimensional Bessel process  $\text{Bes}(\nu)$  is given by

$$\mathcal{A}_\nu = \frac{\nu - 1}{2r} \frac{d}{dr} + \frac{1}{2} \frac{d^2}{dr^2}.$$

Thus, for  $\nu = 3$ , we have

$$\mathcal{A}_3 = \frac{d}{dr} + \frac{1}{2} \frac{d^2}{dr^2}.$$

This implies that the process  $\text{Bes}(3)$  is a (strong) solution of the SDE

$$dr_t = \frac{1}{r_t} dt + d\beta_t, \quad r_0 = 0$$

where  $\beta = (\beta_t)_{t \geq 0}$  is the Brownian motion.

It follows from (2) that the process  $R(B) = |B| + L(B)$  is also a version of the Bessel process  $\text{Bes}(3)$ .

3) It was proved in [10] that the process  $2 \sup B^\lambda - B^\lambda$  is a diffusion process with the following infinitesimal generator:

$$\mathcal{A}_3^\lambda = \lambda \coth \lambda r \frac{d}{dr} + \frac{1}{2} \frac{d^2}{dr^2}.$$

Equalities (3) and (6) imply that each of the processes  $|B^\lambda| + L(B^\lambda)$  and  $|X^\lambda| + L(X^\lambda)$  is a diffusion process with infinitesimal generator  $\mathcal{A}_3^\lambda$ .

4) Due to equalities (15) and (17), the process

$$M_t = e^{-\frac{\lambda^2}{2}t} \frac{\text{sh } \lambda R_t(B)}{\lambda R_t(B)}$$

is a  $(\mathcal{F}_t^{R(B)}, P_{R(B)})$ -martingale. Set  $\tau_a = \inf\{t \geq 0 : R_t(B) = a\}$ ,  $a > 0$ . It is clear that  $P(\tau_a < \infty) = 1$  and the family of random variables  $\{M_{t \wedge \tau_a}, t \geq 0\}$  is uniformly integrable. Applying the *optional stopping theorem*, we get  $\mathbb{E} M_{\tau_a} = 1$ . This is equivalent to the well-known property

$$\mathbb{E} e^{-\frac{\lambda^2}{2}\tau_a} = \frac{\lambda a}{\text{sh } \lambda a}.$$

By comparison, for  $\sigma_a = \inf\{t : |B_t| \geq a\}$ , one has

$$\mathbb{E} e^{-\frac{\lambda^2}{2}\sigma_a} = \frac{1}{\text{ch } \lambda a}.$$

It is obvious that  $\tau_a \leq \sigma_a$ .

## 2 Some Extensions of P. Lévy's Theorem

1. The proof of property (4) was given in [1]. This proof is based on the application of *Girsanov's theorem* to the processes  $B^\lambda$  and  $X^\lambda$ , *Tanaka's formula* to  $|B|$  and *P. Lévy's theorem* (1) to the process  $B$ . However, it turns out that property (4) can be deduced directly from *Skorokhod's lemma* which is formulated below. This lemma is usually used to prove P. Lévy's theorem (see, for example, [9, ch.VI, §2, (2.1)]). Moreover, Skorokhod's lemma makes it possible to give new extensions of property (1) (see Theorems 3 and 4 below).

**Lemma (Skorokhod).** *Let  $y = (y_t)_{t \geq 0}$  be a continuous function such that  $y_0 \geq 0$ . There exists a unique pair of functions  $(x, l) = (x_t, l_t)_{t \geq 0}$  such that*

- (a)  $x = y + l$ ,
- (b)  $x \geq 0$ ,
- (c)  $l = (l_t)_{t \geq 0}$  is increasing, continuous, vanishing at zero, and the corresponding measure  $dl_t$  is carried by  $\{s \geq 0 : x_s = 0\}$ .

The function  $l$  is moreover given by

$$l_t = \sup_{s \leq t} (-y_s \vee 0).$$

For the proof of this lemma, see [11], [9, ch.VI, §2, (2.1)].

**2.** We now turn to the extension of P. Lévy's theorem to the Brownian motion with a *random* drift.

First, let us consider the SDEs

$$dB_t^{(a)} = a(t, B^{(a)}) dt + dB_t, \quad B_0^{(a)} = 0 \quad (18)$$

and

$$dX_t^{(a)} = -a(t, Y^{(a)}) \operatorname{sgn} X_t^{(a)} dt + dB_t, \quad X_0^{(a)} = 0. \quad (19)$$

Here,  $a = a(t, x)$  is a bounded process on  $C(\mathbb{R}_+, \mathbb{R})$  that is predictable with respect to the canonical filtration  $\mathcal{F}_t = \sigma(x_s, s \leq t)$ , and

$$Y_t^{(a)} = - \int_0^t \operatorname{sgn} X_s^{(a)} dX_s^{(a)}. \quad (20)$$

For a constant drift  $a(t, x) \equiv \lambda$ , the solution of (18) is the Brownian motion with drift  $\lambda$  while SDE (19) transforms to (5).

**Lemma 2.** *There is weak existence and uniqueness in law for SDE (18) as well as for system (19)-(20).*

**P r o o f.** First, we prove the existence of a weak solution for system (19)-(20). Consider the canonical path space  $C(\mathbb{R}_+, \mathbb{R}^2)$  with the canonical process  $(x, y) = (x_t, y_t)_{t \geq 0}$  and the filtration  $\mathcal{F}_t = \sigma(x_s, y_s; s \leq t)$ . Let  $P$  be a probability measure on  $\mathcal{F}_\infty = \sigma(x_s, y_s; s \geq 0)$  such that  $(x_t)_{t \geq 0}$  is the Brownian motion with respect to  $P$  and

$$y_t = -(P) \int_0^t \operatorname{sgn} x_s dx_s$$

where the symbol  $(P)$  denotes that the stochastic integral is constructed with respect to  $P$ .

Let us define the measures  $P'_t$  on the  $\sigma$ -fields  $\mathcal{F}_t (t \geq 0)$  by

$$\frac{dP'_t}{dP_t}(x, y) = M_t = \exp \left\{ -(P) \int_0^t a(s, y) dx_s - \frac{1}{2} \int_0^t a^2(s, y) ds \right\} \quad (21)$$

where  $P_t = P|_{\mathcal{F}_t}$ . In view of the boundedness of  $a(s, y)$ , *Novikov's criterion* (see [6], [4, ch.6, §2]) implies that  $M_t$  is a  $(\mathcal{F}_t, P)$ -martingale. Thus,  $P'_t|_{\mathcal{F}_s} = P'_s$  for  $s \leq t$ . Therefore, there exists a probability measure  $P'$  on  $\mathcal{F}_\infty$  such that  $P'|_{\mathcal{F}_t} = P'_t$  for any  $t \geq 0$  (see [12, p. 34]).

As the stochastic integral remains unchanged with the (locally) equivalent change of measure (see [5, Lemme III.2.]), we have

$$y_t = -(P) \int_0^t \operatorname{sgn} x_s dx_s = -(P') \int_0^t \operatorname{sgn} x_s dx_s.$$

This equality, combined with *Girsanov's theorem* (see [2, ch.III, §3b, (3.11)]), proves that the process  $(x_t, y_t)_{t \geq 0}$  is a solution of system (19)-(20) on the filtered probability space  $(C(\mathbb{R}_+, \mathbb{R}^2), \mathcal{F}_t, P')$ .

We now prove the uniqueness in law for system (19)-(20). Let  $Q'$  be a probability measure on  $(C(\mathbb{R}_+, \mathbb{R}^2), \mathcal{F}_\infty)$  corresponding to an arbitrary weak solution of this system. Define the measures  $Q_t$  on the  $\sigma$ -fields  $\mathcal{F}_t (t \geq 0)$  by

$$\frac{dQ_t}{dQ'_t}(x, y) = \exp \left\{ (Q') \int_0^t a(s, y) dx_s + \frac{1}{2} \int_0^t a^2(s, y) ds \right\}. \quad (22)$$

Arguing as above, we see that there exists a probability measure  $Q$  on  $\mathcal{F}_\infty$  such that  $Q|\mathcal{F}_t = Q_t$  for any  $t \geq 0$ . By Girsanov's theorem, the process  $(x_t)_{t \geq 0}$  is  $(\mathcal{F}_t, P)$ -Brownian motion. Furthermore,

$$y_t = -(Q') \int_0^t \operatorname{sgn} x_s dx_s = -(Q) \int_0^t \operatorname{sgn} x_s dx_s.$$

Thus,  $Q = P$ . With the equality

$$(Q') \int_0^t a(s, y) dx_s = (Q) \int_0^t a(s, y) dx_s = (P) \int_0^t a(s, y) dx_s,$$

formulas (21) and (22) imply that

$$\frac{dP'_t}{dP_t} = \left( \frac{dQ_t}{dQ'_t} \right)^{-1} = \frac{dQ'_t}{dQ_t}, \quad t \geq 0.$$

Hence,  $Q' = P'$ . This proves the uniqueness in law for system (19)-(20).

The weak existence and the uniqueness in law for SDE (18) are proved in the same way.

**Theorem 3.** *Let  $B^{(a)}$  and  $X^{(a)}$  be solutions of SDEs (18)-(20). Then,*

$$\left( \sup B^{(a)} - B^{(a)}, \sup B^{(a)} \right) \stackrel{\text{law}}{=} \left( |X^{(a)}|, L(X^{(a)}) \right), \quad (23)$$

*i.e., the processes  $(\sup_{s \leq t} B_s^{(a)} - B_t^{(a)}, \sup_{s \leq t} B_s^{(a)}; t \geq 0)$  and  $(|X^{(a)}|, L(X^{(a)}); t \geq 0)$  have the same law.*

**P r o o f.** By Tanaka's formula,

$$|X_t^{(a)}| = \int_0^t \operatorname{sgn} X_s^{(a)} dX_s^{(a)} + L_t(X^{(a)}) = -Y_t^{(a)} + L_t(X^{(a)}).$$

Applying Skorokhod's lemma, we derive that  $L_t(X^{(a)}) = \sup_{s \leq t} Y_s^{(a)}$ . Thus,

$$|X_t^{(a)}| = \sup_{s \leq t} Y_s^{(a)} - Y_t^{(a)}$$

and obviously,

$$\left( \sup_{s \leq t} Y_s^{(a)} - Y_t^{(a)}, \sup_{s \leq t} Y_s^{(a)}; t \geq 0 \right) = \left( |X_t^{(a)}|, L_t(X^{(a)}); t \geq 0 \right). \quad (24)$$

Further,

$$\begin{aligned} Y_t^{(a)} &= - \int_0^t \operatorname{sgn} X_s^{(a)} dX_s^{(a)} = \int_0^t a(s, Y^{(a)}) ds - \int_0^t \operatorname{sgn} X_s^{(a)} dB_s = \\ &= \int_0^t a(s, Y^{(a)}) ds + \tilde{B}_t. \end{aligned} \quad (25)$$

It follows from P. Lévy's theorem (see [9, ch.IV, §3, (3.6)]) that the process  $\tilde{B}_t = - \int_0^t \operatorname{sgn} X_s^{(a)} dB_s$  is the standard linear Brownian motion. It is obvious that  $\tilde{B}$  is a martingale with respect to the filtration  $\mathcal{F}^{Y^{(a)}} = (\mathcal{F}_t^{Y^{(a)}})_{t \geq 0}$ . Thus,  $Y^{(a)}$  is a solution of SDE (18). The uniqueness in law for this SDE implies that  $Y^{(a)} \stackrel{\text{law}}{=} B^{(a)}$ . Now, the desired result follows from (24).

**3.** In all the cases above, the "basic" process is expressed by the Brownian motion  $B = (B_t)_{t \geq 0}$ . In what follows, the "generating" process will be a *conditionally Gaussian*

(with respect to the filtration  $\mathcal{F}^{\langle M \rangle} = (\mathcal{F}_t^{\langle M \rangle})_{t \geq 0}$ ) martingale  $M = B_{\langle M \rangle} = (B_{\langle M \rangle_t})_{t \geq 0}$  for which  $B$  and the quadratic variation  $\langle M \rangle$  are *independent* processes (see [2, ch.II, (6.2)], [7]). In [13] such martingales are called *Ocone martingales*.

**Theorem 4.** *Let  $M = B_{\langle M \rangle}$  be a conditionally Gaussian martingale with independent processes  $B$  and  $\langle M \rangle$ . Set*

$$M_t^\lambda = \lambda \langle M \rangle_t + M_t.$$

Let  $X^\lambda$  denote a solution of the SDE

$$dX_t^\lambda = -\lambda \operatorname{sgn} X_t^\lambda d\langle M \rangle_t + dM_t, \quad X_0^\lambda = 0. \quad (26)$$

Then,

$$(\sup M^\lambda - M^\lambda, \sup M^\lambda) \stackrel{\text{law}}{=} (|X^{(a)}|, L(X^{(a)})). \quad (27)$$

**P r o o f.** We first prove that SDE (26) has a solution. Let  $B = (B_t)_{t \geq 0}$  be the initial Brownian motion and  $Z^\lambda = (Z_t^\lambda)_{t \geq 0}$  be the solution of the SDE

$$dZ_t^\lambda = -\lambda \operatorname{sgn} Z_t^\lambda dt + dB_t, \quad Z_0^\lambda = 0.$$

This equation (compare with (5)) has the unique strong solution (see [14]). Due to the properties of the time-change ( $t \mapsto \langle M \rangle_t$ ), we have

$$Z_{\langle M \rangle_t}^\lambda = -\lambda \int_0^{\langle M \rangle_t} \operatorname{sgn} Z_s^\lambda ds + B_{\langle M \rangle_t} = -\lambda \int_0^t \operatorname{sgn} Z_{\langle M \rangle_s}^\lambda d\langle M \rangle_s + M_t.$$

Thus, the process  $X_t^\lambda = Z_{\langle M \rangle_t}^\lambda$  is a solution of (26).

Set  $Y_t^\lambda = -\int_0^t \operatorname{sgn} X_s^\lambda dX_s^\lambda$  (compare with (20)). Then,

$$(\sup_{s \leq t} Y_s^\lambda - Y_t^\lambda, \sup_{s \leq t} Y_s^\lambda; t \geq 0) = (|X_t^\lambda|, L_t(X^\lambda); t \geq 0) \quad (28)$$

which can be proved by the same argument as (24). Besides, thanks to (26), we have

$$Y_t^\lambda = -\int_0^t \operatorname{sgn} X_s^\lambda dX_s^\lambda = \lambda \langle M \rangle_t - \int_0^t \operatorname{sgn} X_s^\lambda dM_s.$$

The following statement is true for martingales  $M = B_{\langle M \rangle}$  with independent  $B$  and  $\langle M \rangle$  (see [7, Theorem A, Lemma (2.5)]). *If  $\varepsilon = (\varepsilon_t)_{t \geq 0}$  is a predictable process taking values  $\pm 1$ , then,*

$$\left( \langle M \rangle_t, \int_0^t \varepsilon_s dM_s; t \geq 0 \right) \stackrel{\text{law}}{=} (\langle M \rangle_t, M_t; t \geq 0).$$

Consequently,  $Y^\lambda \stackrel{\text{law}}{=} M^\lambda$ . This, together with (28), completes the proof of (27).

**Corollary 5.** *If  $M = B_{\langle M \rangle}$  is a conditionally Gaussian martingale with independent  $B$  and  $\langle M \rangle$ , then*

$$(\sup M - M, \sup M) \stackrel{\text{law}}{=} (|M|, L(M)).$$

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