

# BROWNIAN MOTION REFLECTED ON BROWNIAN MOTION

Krzysztof Burdzy<sup>(1)</sup> and David Nualart<sup>(2)</sup>

**Abstract.** We study Brownian motion reflected on an “independent” Brownian path. We prove results on the joint distribution of both processes and the support of the parabolic measure in the space-time domain bounded by a Brownian path. We show that there exist two different natural local times for a Brownian path reflected on a Brownian path.

**1. Introduction.** Reflected Brownian motion in a domain with a time-varying boundary has appeared in several articles (Bass and Burdzy [BB1], Cranston and Le Jan [CLJ], El Karoui and Karatzas [EKK1,2], Knight [K] and Soucaliuc, Toth and Werner [STW]) although it had never been the main subject of study until a recent paper of Burdzy, Chen and Sylvester [BCS]. The last article contains a number of theorems on reflected Brownian motion and the corresponding heat equation in domains with smooth space-time boundaries. If the boundary of a space-time domain is of class  $C^3$  then practically all results on reflected Brownian motion in fixed domains can be proved in the new setting. [BCS] also shows that various singularities appear in domains which have rough boundaries. The critical shape of the moving boundary seems to be the square root. Analysts observed some time ago that rough boundaries cause difficulties in the parabolic potential theory. For example, the results of Hofmann and Lewis [HL] or Lewis and Murray [LM] (see also references in these papers) show that if the boundary of a time-varying domain is given by a Hölder function with exponent  $1/2$  then the parabolic measure is absolutely continuous with respect to the Lebesgue measure but this is not necessarily true in less smooth domains.

We decided to examine a particular domain with a rough time-dependent boundary, namely, a domain bounded by a Brownian path. We chose a Brownian path because Brownian trajectories are known to have rough behavior and so are qualitatively different from smooth graphs. On the other hand, there exists a large body of literature describing

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their local properties; hence there are good reasons to believe that one can prove many concrete results on Brownian motion reflected on Brownian motion. It is well known that a typical Brownian path is Hölder with exponent  $1/2 - \delta$  for any  $\delta > 0$  but fails to be Hölder with exponent  $1/2$ . Hence we will be dealing with the situation very close to the “critical modulus of continuity.”

Brownian motion reflected on Brownian motion appeared in recent papers by Soucaliuc, Toth and Werner [STW] and Burdzy, Chen and Sylvester [BCS]; we will describe some of those results later in our article. We parenthetically mention an article ([B2]) with a title which is similar to the present one, namely, “A three-dimensional Brownian path reflected on a Brownian path is a free Brownian path;” the similarity does not extend beyond the title, though.

The rest of the paper is divided into four sections. Section 2 presents two results on the distribution of reflected Brownian motion on Brownian motion. Section 3 is devoted to various local times which one can define to describe the amount of time spent by one process on the path of the other one. Section 4 gives some results on the support of the parabolic measure. Section 5 is concerned with a technical question which arose in the course of the study, namely, it contains an estimate for the non-intersection probability of Brownian paths.

**2. Results on global distributions.** Our original goal was to study local properties of reflected Brownian motion but it turned out that some global results were needed as technical elements of the proofs—we hope that they will have some interest of their own.

First we will review some facts about reflected processes. Let us consider a space-time domain with continuous deterministic moving boundary  $g(t)$ , i.e.,  $D = \{(t, x) : t \geq 0, x \leq g(t)\}$ . [BCS] contains several definitions of Brownian motion  $X_t$  in  $D$  reflected on  $g(t)$ . The definition which is most relevant to our present project uses the “Skorohod Lemma” [BCS, Lemma 3.13] or [KS, Lemma 3.6.14]. Suppose that  $B_t$  is a Brownian motion with  $B_0 \leq g(0)$ . Then there exists a unique continuous non-decreasing process  $L_t$  with  $L_0 = 0$ , such that  $X_t = B_t - L_t \leq g(t)$  for all  $t \geq 0$ , and  $L_t$  may increase only when  $X_t$  is on the boundary of the domain  $D$ , i.e.,

$$\int_0^\infty \mathbf{1}_{(-\infty, g(s))}(X_s) dL_s = 0.$$

There exists an explicit formula for the local time  $L_t$ ,

$$L_t = 0 \vee \sup_{0 \leq s \leq t} (B_t - g(t)).$$

By convention, the reflected process will stay below or at the moving boundary, unless stated otherwise. The definition applies in the obvious way to any random function  $g(t)$  which is continuous with probability 1.

In our paper all stochastic processes will be denoted by capital letters such as  $B$  or  $X$  except that we will use  $g$  to denote a random moving reflecting boundary to emphasize that, most of the time,  $g$  should be thought of as a “fixed Brownian path.”

We will combine proofs of the following two theorems into one argument.

Let  $B_t$  be a Brownian motion with  $B_0 = 0$ . Suppose that  $C_t$  is a 3-dimensional Bessel process independent of  $B_t$  and starting from 0. A process with the same distribution as  $\{(B_t + C_t)/\sqrt{2}, t \geq 0\}$  will be called a BMB-process.

**Theorem 2.1.** *Suppose that  $g(t)$  and  $B_t$  are independent Brownian motions starting from  $g(0) = B_0 = 0$ . Consider the Brownian motion  $X_t$  reflected on  $g(t)$ , obtained from  $B_t$  by the means of the Skorohod lemma.*

*Let  $L_t$  denote the local time of  $X_t$  on  $g(t)$  and let  $\xi = \inf\{t : L_t = 1\}$ . Define processes  $\tilde{X}_t = X_{\xi-t} - X_\xi$  and  $\tilde{g}(t) = g(\xi - t) - g(\xi)$ . The pair of processes  $\{(-\tilde{X}_t, -\tilde{g}(t)), t \in [0, \xi]\}$  has the same distribution as  $\{(g(t), X_t), t \in [0, \xi]\}$ .*

**Theorem 2.2.** *Let  $X_t$  be defined as in Theorem 2.1. The process  $-X_t$  is a BMB-process.*

**Proof.** Let  $Z_t^1 = (B_t - g(t))/\sqrt{2}$ ,  $Z_t^2 = (B_t + g(t))/\sqrt{2}$  and  $L_t^Z = \sup_{0 \leq s \leq t} Z_t^1$  for  $t \geq 0$ . Note that  $L_t = \sqrt{2}L_t^Z$  and

$$g(t) = (Z_t^2 - Z_t^1)/\sqrt{2}, \quad (2.1)$$

$$X_t = (Z_t^2 + Z_t^1 - 2L_t^Z)/\sqrt{2}. \quad (2.2)$$

We see that  $(g(t), X_t)$  is a deterministic function of the two-dimensional Brownian motion  $(Z_t^1, Z_t^2)$ . The processes  $-Z_t^2$  and  $-(Z_t^1 - 2L_t^Z)$  are independent and the second one is a three-dimensional Bessel process, by Pitman’s Theorem ([P]). This and (2.2) prove Theorem 2.2.

Recall from the statement of Theorem 2.1 a convention for denoting processes time-reversed at  $\xi$ , for example, we will write  $\tilde{Z}_t^1 = Z_{\xi-t}^1 - Z_\xi^1$ . The process  $Y_t = -Z_t^1 + L_t^Z$  is a reflected Brownian motion and  $L_t^Z$  is its local time at 0. By homogeneity of the Brownian excursion point process and symmetry of excursions,  $Y_t$  and  $\tilde{Y}_t$  have the same distribution, and  $-\tilde{L}_t^Z$  is the local time of  $\tilde{Y}_t$  at 0. This implies that

$$V_t \stackrel{\text{df}}{=} \tilde{Y}_t - (-\tilde{L}_t^Z) = -\tilde{Z}_t^1 + 2\tilde{L}_t^Z \quad (2.3)$$

is a Brownian motion and  $L_t^V = \sup_{0 \leq s \leq t} V_t = -\tilde{L}_t^Z$ . This, (2.1), (2.2) and (2.3) imply that

$$\tilde{g}(t) = (\tilde{Z}_t^2 - \tilde{Z}_t^1)/\sqrt{2} = (\tilde{Z}_t^2 + V_t - 2L_t^V)/\sqrt{2}, \quad (2.4)$$

$$\tilde{X}_t = (\tilde{Z}_t^2 + \tilde{Z}_t^1 - 2\tilde{L}_t^Z)/\sqrt{2} = (\tilde{Z}_t^2 - V_t)/\sqrt{2}. \quad (2.5)$$

The random time  $\xi$  may be represented as  $\xi = \inf\{t : \sqrt{2}L_t^Z = 1\} = \inf\{t : \sqrt{2}L_t^V = 1\}$ . The process  $\tilde{Z}_t^2$  is a Brownian motion stopped at  $\xi$  but otherwise independent of  $V_t$ . Theorem 2.1 follows now from the comparison of (2.1)-(2.2) with (2.4)-(2.5).  $\square$

For future reference we present in an informal way a two-dimensional representation of  $(g(t), X_t)$  and the main idea of Theorem 2.1. See [VW] for information about reflected Brownian motion with oblique angle of reflection used in this construction. Consider a half-space  $H = \{(y_1, y_2) : y_1 \geq y_2\}$  and let  $(g(t), X_t)$  be a two-dimensional Brownian motion in  $H$  with oblique reflection on the boundary, whose vector of reflection  $\mathbf{m}$  is equal to  $(0, -\sqrt{2})$ .

The process  $(g(t), X_t)$  satisfies the following two-dimensional Skorohod representation. There exist a two-dimensional Brownian motion  $(g(t), B_t)$  and a continuous additive functional  $\hat{L}_t$  of  $(g(t), B_t)$  increasing only on  $\partial H$  such that a.s. for all  $t \geq 0$ ,

$$(g(t), X_t) = (g(t), B_t) + \int_0^t \mathbf{m} d\hat{L}_s. \quad (2.6)$$

The process  $\hat{L}_t$  can be identified with the local time  $L_t^Z$  in the proof of Theorem 2.1. It follows from (2.6) that for almost every trajectory of  $g(t)$ , the process  $X_t$  is a Brownian motion reflected on  $g(t)$  in the sense of the ‘‘deterministic’’ Skorohod lemma. Indeed,  $X_t = B_t - L_t$ , where  $L_t = \sqrt{2}\hat{L}_t$ ; we always have  $B_t - L_t \leq g(t)$ ; the process  $L_t$  is continuous, non-decreasing and it does not increase when  $X_t \neq g(t)$ . The main idea of Theorem 2.1 is the observation that the process  $(\tilde{g}(t), \tilde{X}_t)$  has a representation

$$(\tilde{g}(t), \tilde{X}_t) = (\tilde{B}_t, \tilde{X}_t) + \int_0^t \mathbf{n} d\hat{L}_s, \quad (2.7)$$

where  $\tilde{B}_t$  and  $\tilde{X}_t$  are independent Brownian motions and  $\mathbf{n} = (\sqrt{2}, 0)$ . Hence, a seemingly strange identity of distributions asserted in Theorem 2.1 can be explained in a simple way by the symmetry of vectors  $\mathbf{m}$  and  $\mathbf{n}$  with respect to the line  $\{x = -y\}$ .

A recent paper by Soucaliuc, Toth and Werner [STW] contains strikingly counter-intuitive and beautiful theorems about Brownian motion reflected on Brownian motion.

Here is a simple version of one of their results. Suppose that  $B_t$  and  $W_t$  are independent Brownian motions with  $B_0 = W_0 = 0$ . Let  $g(t) = W_t - W_1$  for  $t \in [0, 1]$  and let  $X_t$  be the path  $B_t$  reflected on  $g(t)$  according to the Skorohod lemma. The trajectory of  $X_t$  lies above or below  $g(\cdot)$  depending on whether  $g(0) < 0$  or  $g(0) > 0$ . Then  $\{X_t, 0 \leq t \leq 1\}$  has the distribution of the standard Brownian motion. The proof given in [STW] has combinatorial nature. We have learnt (private communication) that F. Soucaliuc has a new proof based on ideas similar to (2.6) and (2.7).

**3. Two different local times on the boundary.** A natural question about any reflected Brownian motion is whether one can explicitly represent its local time on the boundary using intuitive notions. We will examine four possible definitions of the local time.

In principle, we may obtain four different continuous additive functionals in this way, although classical results on the local time of standard Brownian motion at 0 show that all “natural” definitions of Brownian local time agree. Two of the local times defined below ( $L_t^1$  and  $L_t^4$ ) are elements of analytic formulae and so they can be used, in theory at least, to calculate various probabilities. For this reason, these two local times seem to have the greatest significance—it turns out that they are different from each other.

From now on,  $g(t)$  and  $X_t$  will denote the processes constructed in Section 2. As a by-product of that construction, we have obtained our first local time,

$$L_t^1 = L_t = 0 \vee \sup_{0 \leq s \leq t} (B_t - g(t)).$$

We have shown in the proof of Theorem 2.1 that  $L_t = \sqrt{2}L_t^Z$ , and that  $L_t^Z$  is the local time at 0 of the reflected Brownian motion  $Y_t = (g(t) - X_t)/\sqrt{2}$ .

The second definition of a local time of  $X_t$  on  $g(t)$  is the following.

$$L_t^2 = \lim_{\varepsilon \rightarrow 0} \frac{1}{\varepsilon} \int_0^t \mathbf{1}_{[g(s)-\varepsilon, g(s)]}(X_s) ds = \lim_{\varepsilon \rightarrow 0} \frac{1}{\varepsilon} \int_0^t \mathbf{1}_{[0, \varepsilon]}(g(s) - X_s) ds. \quad (3.1)$$

The third definition of a local time will be based on excursion count. We will call a piece  $\{X(s), s \in (u_l, u_r)\}$  of a trajectory of reflected Brownian motion an excursion of  $X_s$  from  $g(s)$  if  $X(u_l) - g(u_l) = X(u_r) - g(u_r) = 0$  but  $X_s \neq g(s)$  for all  $s \in (u_l, u_r)$ . Let  $N_t^\varepsilon$  be the number of excursions of  $X_s$  from  $g(s)$  with  $u_l \in [0, t]$  whose maximum distance from  $g$  exceeds  $\varepsilon$ , i.e.,  $\sup_{s \in (u_l, u_r)} |g(s) - X_s| > \varepsilon$ . Let

$$L_t^3 = \lim_{\varepsilon \rightarrow 0} \varepsilon N_t^\varepsilon. \quad (3.2)$$

The existence of limits in (3.1) and (3.2) follows from well known representations of the local time of reflected Brownian motion in terms of the occupation time density and number of upcrossings, see, e.g., Definition 3.6.3 and Theorems 3.6.11, 3.6.17 and 6.2.23 of [KS]. We leave it to the reader to verify that  $L_t^1 = L_t^2 = L_t^3$ . The only subtle part of this claim is that the normalizing constants in the definitions of the three local times match properly. This is not totally obvious in view of the following remark.

**Remark 3.1.** For a fixed boundary, i.e.,  $g(t) \equiv 0$ , the normalizations of the local times do not match in the same way as in (3.1)-(3.2). We will show that  $L_t^1 = (1/2)L_t^2 = L_t^3$ . If  $g(t) \equiv 0$  then the process  $-X_t$  is a classical reflected Brownian motion. In this case, the equality of  $L_t^1$  and  $L_t^3$  follows from Theorems 3.6.17 and 6.2.23 of [KS]. Definition 3.6.3 and Theorem 3.6.17 of [KS] imply that  $L_t^2 = 2L_t^1$ . By piecing together parts of a Brownian path and a flat function one can obtain a function  $g(t)$  for which  $L_t^1$  is not a constant multiple of  $L_t^2$ . We do not know whether  $L_t^1 \equiv L_t^3$  for every (deterministic) function  $g(t)$ .

The last definition of the local time will be based on the notion of an exit system which is a rigorous approach to the excursion theory. For the original full discussion of exit systems we refer to Maisonneuve [M]. Our own presentation, although somewhat informal, has to be rather technical due to the nature of the subject. Suppose for the moment that  $g(t)$  is a fixed continuous trajectory.

An excursion law  $H^{(t,x)}$  is a  $\sigma$ -finite measure on  $C_*[0, \infty)$ , the set of functions on  $[0, \infty)$  with values in  $D = \{(t, x) : t \geq 0, x \leq g(t)\}$  which are continuous until a lifetime  $\zeta$  (finite or infinite) and are equal to the ‘‘coffin state’’  $\Delta$  on  $[\zeta, \infty)$ . Every excursion law is strong Markov on  $(0, \infty)$  with the transition probabilities of the space-time Brownian motion killed upon hitting the boundary  $g(t)$  of  $D$ .

The  $H^{(t,x)}$ -measure of the set of trajectories not starting from  $(t, x)$  is equal to 0. An exit system is a pair  $(\widehat{L}^4, \widehat{H})$  consisting of a continuous additive functional  $\widehat{L}_t^4$  of Brownian motion  $X_t$  reflected on  $g(t)$  and a family of excursion laws  $\widehat{H}^{(t,x)}$ , one for every  $(t, x)$  of the form  $(t, g(t))$ . A pair  $(\widehat{L}^4, \widehat{H})$  is called an exit system if it satisfies the exit system formula (3.3) below. Let  $\mu(t) = \inf\{s > 0 : \widehat{L}_s^4 > t\}$ ,  $\eta_t = \inf\{s > t : X(s) = g(s)\} - t$  and

$$e_t(s) = \begin{cases} X(t+s) & \text{if } 0 \leq s < \eta_t \text{ and } X_t = g(t), \\ \Delta & \text{otherwise.} \end{cases}$$

Then

$$E^{(t,x)} \sum_{0 < t < \infty} Z_t f(e_t) = E^{(t,x)} \int_0^\infty Z_t \widehat{H}^{(t,X_t)}(f) d\widehat{L}_t^4 = E^{(t,x)} \int_0^\infty Z_{\mu(t)} \widehat{H}^{(\mu(t), X_{\mu(t)})}(f) dt, \quad (3.3)$$

for all  $(t, x) \in D$ , all positive predictable processes  $Z$  and positive measurable functions  $f$  defined on  $C_*[0, \infty)$  which vanish on paths equal identically to  $\Delta$ . The existence of an exit system  $(\widehat{L}^4, \widehat{H})$ , in particular, the existence of an additive continuous functional  $\widehat{L}_t^4$ , follow from the results of Maisonneuve [M].

We note that there is a variety of exit systems, for example, if  $(\widehat{L}^4, \widehat{H})$  is an exit system, so is  $(2\widehat{L}^4, (1/2)\widehat{H})$ . We will specify a single “natural” exit system. First, we let  $A$  denote the event that an excursion  $e_t(\cdot)$  deviates from  $g$  by more than 1 unit, i.e.,

$$A = \{|g(t+s) - e_t(s)| > 1 \text{ for some } s \geq 0\}. \quad (3.4)$$

We take an arbitrary exit system  $(\widehat{L}^4, \widehat{H})$ . It is easy to see that  $\widehat{H}^{(t,x)}(A) > 0$  if and only if the excursion law  $\widehat{H}^{(t,x)}$  is not identically zero. If  $\widehat{H}^{(t,x)}$  vanishes identically, we let  $H^{(t,x)}$  be the same zero measure. For other excursion laws we let  $H^{(t,x)}(\cdot) = \widehat{H}^{(t,x)}(\cdot) / \widehat{H}^{(t,x)}(A)$  and then we define a new continuous additive functional,

$$L_t^4 = \int_0^t H^{(s,X_s)}(A) d\widehat{L}_s^4. \quad (3.5)$$

The local time  $L_t^4$  does not increase at  $(t, x)$  such that  $H^{(t,x)}$  is identically zero. In view of (3.3), the excursions of type  $A$  form a Poisson point process with intensity 1 on the  $L_t^4$ -local time scale.

We are interested in a random boundary function  $g(t)$ —a Brownian path. The results on exit systems apply to almost all paths  $g(t)$  because Brownian paths are almost surely continuous.

We will show in Theorem 3.1 that random measures  $dL_t^1$  and  $dL_t^4$  are mutually singular a.s.

Let  $C[0, \infty)$  be the family of continuous functions on  $[0, \infty)$  equipped with the topology of uniform convergence on compact intervals. We denote a generic element of  $C[0, \infty)$  by  $\omega(t)$  and we let  $\mathcal{F}_t$  be the smallest  $\sigma$ -field which makes the coordinate mappings  $\omega \rightarrow \omega(s)$  Borel-measurable for every  $s \leq t$ . We will call a set  $A \subset C[0, \infty)$  a “local property” if it belongs to  $\mathcal{F}_{0+} = \bigcap_{t>0} \mathcal{F}_t$ .

Let  $B_t$  be a Brownian motion with  $B_0 = 0$ . We will call  $A \subset C[0, \infty)$  a Brownian local property if  $A$  is a local property and  $P(\{B_t, 0 \leq t < \infty\} \in A^c) = 0$ .

The following definition is related to BMB-processes defined in the previous section. Suppose that  $C_t$  is a 3-dimensional Bessel process independent of  $B_t$  and starting from 0. We will say that  $A \subset C[0, \infty)$  is a BMB local property if  $A$  is a local property and

$$P(\{(B_s + C_s)/\sqrt{2}, 0 \leq s \leq \infty\} \in A^c) = 0.$$

By Blumenthal's 0-1 Law, if a local property  $A$  is not a Brownian local property then  $A^c$  is; the same holds for BMB local properties.

For a deterministic function  $g(t) : [0, \infty) \rightarrow \mathbf{R}$  let  $g_s^+(t) = g(s+t) - g(s)$  for  $t \geq 0$ , and let  $g_s^-(t) = g(s-t) - g(s)$  for  $t \leq s$  and  $g_s^-(t) = g(0) - g(s)$  for  $t \geq s$ .

Recall the pair of processes  $(g(t), X_t)$  defined in Section 2.

**Theorem 3.1.** (i) Fix any Brownian local property  $A_1$  and any BMB local property  $A_2$ . For almost every trajectory of  $g(t)$  the following is true. With probability 1, the random measure  $dL_t^1$  does not charge the set of  $s$  such that  $g_s^+(\cdot) \in A_1^c$  or  $g_s^-(\cdot) \in A_2^c$ .

(ii) If  $A$  is a BMB local property then, for almost all trajectories of  $g(t)$ , the measure  $dL_t^4$  does not charge the set of  $s$  such that  $g_s^+(\cdot) \in A^c$  or  $g_s^-(\cdot) \in A^c$ , a.s.

(iii) There exist a Brownian local property  $A_1$  and a BMB local property  $A_2$  such that  $A_1 \cap A_2 = \emptyset$ .

(iv) With probability 1, the random measures  $dL_t^1$  and  $dL_t^4$  are mutually singular.

**Proof.** Consider a Brownian local property  $A_1$ . Fix any  $a > 0$  and let  $\xi_a = \inf\{t : L_t^1 = a\}$ . The random time  $\xi_a$  is a stopping time for  $Y_t = (g(t), X_t)$  so the process  $\{g(t + \xi_a) - g(\xi_a), t \geq 0\}$  is a Brownian motion. This implies that with probability 1, the function  $\{g(t + \xi_a) - g(\xi_a), t \geq 0\}$  has the Brownian local property  $A_1$ . Hence, for almost every trajectory  $g(t)$ , we have  $\{g(t + \xi_a) - g(\xi_a), t \geq 0\} \in A_1$  a.s. For the rest of this paragraph, we fix a trajectory of  $g(t)$  for which the last statement is true. By Fubini's theorem, with probability 1,  $\{g(t + \xi_a) - g(\xi_a), t \geq 0\} \in A_1$  for almost all  $a > 0$ . Let  $U$  be a random variable with the exponential distribution, independent of the processes considered so far. Consider  $R = \xi_U$  and note that a.s.,  $R$  does not belong to the set of  $s$  such that  $\{g(t + s) - g(s), t \geq 0\} \in A_1^c$ . If the random measure  $dL_t^1$  charged with positive probability a set of times  $s$  such that  $\{g(t + s) - g(s), t \geq 0\} \in A_1^c$  then  $R$  would belong to  $A_1^c$  with positive probability, which is not the case. We conclude that  $dL_t^1$  does not charge the set of  $s$  such that  $g_s^+(\cdot) \in A_1^c$ .

Next consider a BMB local property  $A_2$ . We will assume without loss of generality that  $X_0 = g(0) = 0$ . Let  $\tilde{X}_t = X_{\xi_1 - t} - X_{\xi_1}$ ,  $\tilde{g}(t) = g(\xi_1 - t) - g(\xi_1)$  and  $\tilde{L}_t^1 =$

$-(L_{\xi_1-t}^1 - L_{\xi_1}^1)$ . According to Theorem 2.1,  $\{(-\tilde{X}_t, -\tilde{g}(t)), t \in [0, \xi_1]\}$  has the same distribution as  $\{(g(t), X_t), t \in [0, \xi_1]\}$ . Fix some  $a \in (0, 1)$  and set  $\tilde{\xi}_{1-a} = \inf\{t : \tilde{L}_t^1 = 1 - a\}$ . Note that  $\xi_1 - \xi_a = \tilde{\xi}_{1-a}$ . It is easy to see that  $\tilde{L}_t^1$  is adapted to the filtration generated by  $(-\tilde{X}_t, -\tilde{g}(t))$ , for example, using the equivalent definition (3.1) of the local time. Hence,  $\tilde{\xi}_{1-a}$  is a stopping time for  $(-\tilde{X}_t, -\tilde{g}(t))$ . Since  $\{-\tilde{X}_t, t \in [0, \xi_1]\}$  is a Brownian motion and  $\{-\tilde{g}(t), t \in [0, \xi_1]\}$  is a Brownian motion reflected on  $-\tilde{X}_t$ , the same is true of  $\{-\tilde{X}_{\tilde{\xi}_{1-a}+t} + \tilde{X}_{\tilde{\xi}_{1-a}}, t \in [0, \xi_1 - \tilde{\xi}_{1-a}]\}$  and  $\{-\tilde{g}(\tilde{\xi}_{1-a} + t) + \tilde{g}(\tilde{\xi}_{1-a}), t \in [0, \xi_1 - \tilde{\xi}_{1-a}]\}$ , by the strong Markov property applied at  $\tilde{\xi}_{1-a}$ . It follows from Theorem 2.2 that  $\{\tilde{g}(\tilde{\xi}_{1-a} + t) - \tilde{g}(\tilde{\xi}_{1-a}), t \in [0, \xi_1 - \tilde{\xi}_{1-a}]\}$  is a BMB process and so  $\{\tilde{g}(\tilde{\xi}_{1-a} + t) - \tilde{g}(\tilde{\xi}_{1-a}), t \in [0, \xi_1 - \tilde{\xi}_{1-a}]\} \in A_2$ , a.s. This is equivalent to saying that, for a fixed  $a \in (0, 1)$ , a.s.,  $\{g(\xi_a - t) - g(\xi_a), t \in [0, \xi_a]\} \in A_2$ . We can use this to deduce that  $dL_t^1$  does not charge the set of  $s$  such that  $g_s^-(\cdot) \in A_2^c$ , just like in the first part of the proof.

We will next analyze the support of  $dL_t^4$ . Recall the definition (3.5) of  $L_t^4$ . In a sense, we have eliminated from the support of  $L_t^4$  all times  $s$  from which an excursion cannot start. More precisely, if we can show that excursions of  $X_t$  from  $g(t)$  start only at points having a certain property, it will follow that the measure  $dL_t^4$  must be supported on the set of points with this property.

Suppose that  $A$  is a BMB local property.

The end of an excursion of  $-\tilde{g}(t)$  from  $-\tilde{X}_t$  corresponds in an obvious way to the start of an excursion of  $X_t$  from  $g(t)$ . Fix any rational number  $u > 0$  and consider the end  $T_u$  of an excursion of  $-\tilde{g}(t)$  from  $-\tilde{X}_t$  straddling  $u$  (there might be no such excursion, for example, if  $-\tilde{g}(u) = -\tilde{X}_u$  or  $u > \xi_1$ ; in such a case we let  $T_u = \xi_1$ ). The random variable  $T_u$  is a stopping time relative to  $(-\tilde{X}_t, -\tilde{g}(t))$  so  $\{\tilde{g}(T_u + t) - \tilde{g}(T_u), t \in [0, \xi_1 - T_u]\} \in A$ , a.s., just like in the earlier part of the proof. The same argument as before proves that  $dL_t^4$  does not charge the set of  $s$  such that  $g_s^-(\cdot) \in A^c$ .

Recall the notation from the proof of Theorem 2.1. An excursion of  $X_t$  from  $g(t)$  starting at time  $s$  corresponds to an excursion of  $Y_t^1$  from 0 starting at the same time, in view of (2.1) and (2.3). The local properties of  $\{Y_{t+s}^1 - Y_s^1, t \geq 0\}$  are those of the 3-dimensional Bessel process. Hence,  $\{(Y_{t+s}^1 - Y_s^1, Z_{t+s}^2 - Z_s^2), t \geq 0\}$  has the local properties of a pair of independent processes—a 3-dimensional Bessel process and a Brownian motion (see Theorem 5.2 in [B1]). Since the time  $s$  is the starting point of an excursion, there exists  $\delta > 0$  such that for  $t \in [0, \delta]$ ,

$$g(s+t) - g(s) = K_{s+t}^1 - K_s^1 = (Y_{t+s}^1 - Y_s^1 + Z_{t+s}^2 - Z_s^2)/\sqrt{2}.$$

This shows that if  $s$  is a starting point of an excursion then  $g_s^+(\cdot) \in A$  a.s. The proof of

(ii) is complete.

Finally we will prove part (iii). Let

$$A_1 = \left\{ \omega : \liminf_{t \downarrow 0} \frac{\omega(t)}{\sqrt{2t \log \log(1/t)}} = -1 \right\},$$

and

$$A_2 = \left\{ \omega : \liminf_{t \downarrow 0} \frac{\omega(t)}{\sqrt{2t \log \log(1/t)}} \geq -\frac{1}{\sqrt{2}} \right\}.$$

Obviously,  $A_1 \cap A_2 = \emptyset$ . The classical local Law of Iterated Logarithm for Brownian motion implies that  $A_1$  is a local Brownian property. It will now suffice to show that  $A_2$  is a BMB local property.

Recall the definition of a BMB-process. We take a Brownian motion  $B_t$  and an independent 3-dimensional Bessel process  $C_t$ , both starting from 0. Then  $\{(B_t + C_t)/\sqrt{2}, t \geq 0\}$  is a BMB-process. Since  $C_t \geq 0$  for all  $t \geq 0$ , a.s., we have using the LIL,

$$\liminf_{t \downarrow 0} \frac{(B_t + C_t)/\sqrt{2}}{\sqrt{2t \log \log(1/t)}} \geq \liminf_{t \downarrow 0} \frac{1}{\sqrt{2}} \frac{B(t)}{\sqrt{2t \log \log(1/t)}} = -\frac{1}{\sqrt{2}}.$$

This means that  $A_2$  is a BMB local property. The proof of (iii) is complete.

Part (iv) follows directly from (i)-(iii). □

When the reflecting boundary is sufficiently smooth, the time-reversed reflected Brownian path has a distribution absolutely continuous with respect to reflected Brownian path run in the reversed direction; see Theorem 3.12 of [BCS] for a rigorous statement of the result. Our Theorem 3.1 (i) implies that if we reverse in time a path of Brownian motion reflected on a (fixed) path  $g(t)$  of Brownian motion then the resulting trajectory will be fundamentally different from the Brownian path running in the negative direction of time and reflected on the same (fixed) Brownian boundary  $g(t)$ . This is because the definition of the local time  $L_t^2$  is “time-reversible,” i.e., if (3.1) is used to define a random measure  $dL_t^2$  and the same definition is applied to the time-reversed path, then the measure  $dL_t^2$  and its time-reversed counterpart are supported on the same subset of the time axis. However, the points in the support of  $dL_t^2 = dL_t^1$  have different properties to the left and right.

One would like to characterize local times  $L_t^1$  and  $L_t^4$  by specifying their Revuz measures (see [R]). This does not seem to be an easy task for several reasons. First, the results contained in Theorem 3.1 show that the Revuz measures of  $L_t^1$  and  $L_t^4$  are not equal to the

“projection”  $\lambda$  of the Lebesgue measure on  $g(t)$ , defined by  $\lambda(\{(t, g(t)) : s \leq t \leq u\}) = u - s$  for all  $s \leq u$ . The second reason why it may be hard to find an explicit characterization for the Revuz measures of  $L_t^1$  and  $L_t^4$  is that the transition density for  $X_t$  is unbounded near (some points of) the reflecting boundary  $g(t)$ —see our next theorem. Good estimates for the transition densities of the Brownian motion seem to be the key of the Revuz measure-type characterization for the local time spent on a curve by (non-reflecting) Brownian motion (see Remark 2.4 of [BB2]).

**Theorem 3.2.** *For definiteness, assume that  $g(0) = X_0 = 0$ . Almost all trajectories  $g(t)$  have the following property. Let  $u(t, x) = u_g(t, x)$  be defined by*

$$u(t, x)dx = P(X_t \in dx \mid g(\cdot)).$$

*Every non-empty open interval of the real line contains a point  $s$  such that*

$$\limsup_{x \uparrow g(s)} u(s, x) = \infty.$$

**Proof.** It follows from the results of Davis [D], using the invariance of Brownian path properties under time-reversal, that for almost all trajectories of  $g(t)$ , and every non-empty open interval  $(t_1, t_2)$ , there exist  $s \in (t_1, t_2)$  and  $\delta > 0$  such that

$$g(s - t) - g(s) > (1/2)\sqrt{t} \quad \text{for all } 0 < t < \delta.$$

Because of the lack of domain monotonicity for the heat equation solution singularities and the form of results stated in [BCS], we will use the following obviously weaker assertion; there exist  $s \in (t_1, t_2)$  and  $\delta > 0$  such that

$$g(s - t) - g(s) > \sqrt{t} |\log t|^{-1} \quad \text{for all } 0 < t < \delta. \quad (3.6)$$

Fix some trajectory  $g(t)$  and  $s > 0$  satisfying (3.6) and let  $g_1(t)$  be a continuous function such that  $g_1(0) = 0$ ,  $g_1(t) \leq g(t)$  for all  $t$ , and

$$g_1(s - t) - g_1(s) = \sqrt{t} |\log t|^{-1} \quad \text{for all } 0 < t < \delta.$$

Let  $X_t^1$  be the Brownian motion reflected on  $g_1(t)$ , i.e., inside the space-time domain  $D = \{(t, x) : t \geq 0, x \leq g_1(t)\}$ . Assume that  $X_0^1 = 0$  and let  $u_1(t, x)$  be its density, i.e.,

$u_1(t, x)dx = P(X_t^1 \in dx)$ . Recall the process  $X_t$  conditioned on our “fixed” trajectory  $g(t)$ . Since  $g_1(t) \leq g(t)$ , we may assume, in view of Corollary 3.15 of [BCS], that  $X_t^1 \leq X_t$  for all  $t$  a.s. This implies that

$$P(X_s^1 > y) \leq P(X_s > y), \quad (3.7)$$

for  $y < g(s)$ . By Theorem 4.5 of [BCS], there are no heat atoms on the path  $g(t)$  so  $P(X_s = g(s)) = 0$ . This and (3.7) imply that

$$\int_y^{g(s)} u_1(s, x)dx \leq \int_y^{g(s)} u(s, x)dx, \quad (3.8)$$

for every  $y < g(s)$ . By Theorem 4.10 (ii) and Lemma 4.8 (i) of [BCS] and the remarks following that lemma,  $\limsup_{x \uparrow g(s)} u_1(s, x) = \infty$ . Lemma 4.8 (ii) and (iii) of [BCS] and the remarks following it imply that in fact  $\lim_{x \uparrow g(s)} u_1(s, x) = \infty$ . This and (3.8) easily imply that  $\limsup_{x \uparrow g(s)} u(s, x) = \infty$ . Note, however, that our argument does not imply that  $\lim_{x \uparrow g(s)} u(s, x) = \infty$ . We leave it as an open problem to determine whether there exist points  $s$  with this property.  $\square$

In the language of [BCS], a point  $(s, g(s))$  with the property stated in Theorem 3.2 would be called a “singularity” of the heat equation solution. Another type of singularity is when a point  $(s, g(s))$  is a “heat atom,” i.e., when  $P(X_s = g(s)) > 0$ . It has been proved in [BCS] that with probability 1, there are no heat atoms on the paths of Brownian motion  $g(t)$ . Singularities seem to be much harder to understand than heat atoms in view of the results presented in [BCS]. In particular, there is no domain monotonicity for singularities. This means that there exist deterministic functions  $g(t), g_1(t), g_2(t)$  and a time  $s > 0$  such that  $g_1(s) = g(s) = g_2(s)$ ,  $g_1(t) \leq g(t) \leq g_2(t)$  for  $t \leq s$ , the heat equation has a singularity at  $(s, g(s))$  in the space-time domain bounded by  $g(t)$  but there is no singularity at the same point in the domains corresponding to  $g_1(t)$  and  $g_2(t)$ .

We end this section with a theorem stated without proof. Part of Theorem 3.1 applies not only in the context of Brownian motion reflected on Brownian motion but it can also be applied to the excursion theory of the standard (non-reflected) Brownian motion from the graph of an independent Brownian motion. See [BB2] for results on the local time of Brownian motion on deterministic rough curves.

The local times  $L_t^2, L_t^3$  and  $L_t^4$  have their analogues in the non-reflecting context but  $L_t^1$  does not. Let  $\widehat{g}(t)$  and  $\widehat{X}_t$  be independent Brownian motions. It is not necessary to

assume that  $\widehat{g}(0) \geq \widehat{X}_0$ . Upon close inspection of the definitions of  $L_t^2, L_t^3$  and  $L_t^4$ , it turns out that they apply verbatim to the case when the processes  $g(t)$  and  $X_t$  are independent Brownian motions. Hence, we let  $\widehat{L}_t^2, \widehat{L}_t^3$  and  $\widehat{L}_t^4$  be the local times of  $\widehat{X}_t$  on  $\widehat{g}(t)$ , defined in the same way as  $L_t^2, L_t^3$  and  $L_t^4$  were defined relative to  $g(t)$  and  $X_t$ .

**Theorem 3.3.** (i) *The limits in the definitions of  $\widehat{L}_t^2$  and  $\widehat{L}_t^3$  exist a.s. It follows that for almost every path  $\widehat{g}(t)$ , the convergence in these expressions holds with (conditional) probability 1.*

(ii) *With probability 1,  $2\widehat{L}_t^2 = \widehat{L}_t^3$  for all  $t \geq 0$ .*

(iii) *With probability 1, the random measures  $d\widehat{L}_t^2$  and  $d\widehat{L}_t^4$  are mutually singular.*

**Proof.** The theorem can be proved using the same techniques as applied earlier in this section. Specifically, one has to analyze local time on and the excursions from the diagonal line made by the two-dimensional Brownian motion  $(\widehat{g}(t), \widehat{X}_t)$ . The details are left to the reader.  $\square$

**4. Support of the parabolic measure.** Consider a (deterministic) continuous function  $g(t)$  and the corresponding space-time domain  $D = \{(t, x) : 0 \leq t \leq 1, x \leq g(t)\}$ . Suppose that  $g(0) = 0$  and let  $X_t$  be the reflected Brownian motion in  $D$  with  $X_0 = -1$ . Let  $\tau = \inf\{t \in (0, 1) : X_t = g(t)\}$  and  $\sigma = \sup\{t \in (0, 1) : X_t = g(t)\}$ , with the convention that  $\inf \emptyset = 1$  and  $\sup \emptyset = 0$ . The parabolic measure  $\mu_\tau$  on the boundary of  $D$  is defined by  $\mu_\tau(A) = P(X_\tau \in A)$ . The last exit measure is given by  $\mu_\sigma(A) = P(X_\sigma \in A)$ . We will examine the “projections” of both measures on  $\mathbf{R}$ , namely,  $\mu_\tau^{\mathbf{R}}(K) = \mu_\tau(\{(t, g(t)) : t \in K\})$  and  $\mu_\sigma^{\mathbf{R}}(K) = \mu_\sigma(\{(t, g(t)) : t \in K\})$ , for  $K \subset \mathbf{R}$ .

Now assume that  $g(t)$  is a Brownian motion. The measures  $\mu_\tau^{\mathbf{R}}$  and  $\mu_\sigma^{\mathbf{R}}$  have different support, a.s. This follows from Theorem 3.1 (iii) and the following result.

**Theorem 4.1.** *Fix any Brownian local property  $A_1$  and any BMB local property  $A_2$ . For almost every trajectory of  $g(t)$  the following is true. With probability 1, the measure  $\mu_\tau^{\mathbf{R}}$  does not charge the set of  $s$  such that  $g_s^+(\cdot) \in A_1^c$  or  $g_s^-(\cdot) \in A_2^c$ . The measure  $\mu_\sigma^{\mathbf{R}}$  does not charge the set of  $s$  such that  $g_s^+(\cdot) \in A_2^c$  or  $g_s^-(\cdot) \in A_1^c$ .*

**Proof.** The starting points of excursions were analyzed in the proof of Theorem 3.1 (ii). The last assertion of the present theorem follows from that argument because  $\sigma$  is the starting point of an excursion of  $X_t$  from  $g(t)$ .

The random variable  $\tau$  is a stopping time for the process  $(g(t), X_t)$ . The argument in the first part of the proof of Theorem 3.1 (i) applies to any stopping time in place of  $\xi_a$  so it shows that  $\{g(\tau + t) - g(\tau), t \geq 0\} \in A_1$  a.s. This means that the measure  $\mu_\tau^{\mathbf{R}}$  does not charge the set of  $s$  such that  $g_s^+(\cdot) \in A_1^c$ .

It remains to examine the behavior of  $g(t)$  just before time  $\tau$ . This has been done in the proof of Theorem 3.1 (ii).  $\square$

Theorem 4.1 can be combined with Theorem 2.1 to give the following informal description of path behavior near the endpoints of excursions of  $X_t$  from  $g(t)$ . Suppose that  $g(t)$  is a Brownian motion starting from  $g(0) = 0$ , and  $X_t$  is a Brownian motion reflected on  $g(t)$ , constructed from an independent Brownian motion  $B_t$ . Let  $s = \sup\{t < 1 : X_t = g(t)\}$  and  $u = \inf\{t > 1 : X_t = g(t)\}$ . With probability 1,  $s < 1 < u$ . The times  $s$ ,  $1$ , and  $u$  cut the trajectories into eight pieces. We will think about them as paths emanating from  $X_s$  and  $X_u$ , i.e.,  $\{X_{s-t} - X_s, 0 \leq t \leq s\}$ ,  $\{g(s-t) - g(s), 0 \leq t \leq s\}$ ,  $\{X_{s+t} - X_s, 0 \leq t \leq 1-s\}$ ,  $\{g(s+t) - g(s), 0 \leq t \leq 1-s\}$ ,  $\{X_{u-t} - X_u, 0 \leq t \leq u-1\}$ ,  $\{g(u-t) - g(u), 0 \leq t \leq u-1\}$ ,  $\{X_{u+t} - X_u, 0 \leq t < \infty\}$ ,  $\{g(u+t) - g(u), 0 \leq t < \infty\}$ . The first and last of these paths have Brownian local properties; the other 6 paths have BMB local properties. See Figure 1.

Figure 1.

We will use Theorem 4.1 to obtain information about the size of the support of  $\mu_\tau^{\mathbf{R}}$

and  $\mu_\sigma^{\mathbf{R}}$ . The results of Lewis and Murray [LM] and Hofmann and Lewis [HL] show that the parabolic measure  $\mu_\tau^{\mathbf{R}}$  is absolutely continuous with respect to the Lebesgue measure if the boundary function  $g(t)$  is Hölder with exponent  $1/2$ . Brownian paths barely fail to be Hölder with exponent  $1/2$  but it is known that their modulus of continuity is the function  $h(\delta) = c\sqrt{2\delta \log(1/\delta)}$  for any  $c > 1$  (see [KS], Section 2.9.F). Of course, there exist (deterministic) functions  $g(t)$  which are not Hölder with exponent  $1/2$  but for which  $\mu_\tau^{\mathbf{R}}$  is absolutely continuous with respect to the Lebesgue measure.

We will show that this is not the case for a typical Brownian path. In addition, we will give an “upper bound” for the exact Hausdorff measure of the support of  $\mu_\tau^{\mathbf{R}}$ .

Let us recall the definition of the Hausdorff measure. Suppose that  $\phi : (0, 1) \rightarrow \mathbf{R}$  is an increasing continuous function with  $\phi(\delta) \rightarrow 0$  as  $\delta \downarrow 0$ , and such that for some  $c > 0$  and all  $\delta \in (0, 1)$  we have  $\phi(2\delta) < c\phi(\delta)$ . For a family  $\mathcal{K}$  of bounded sets  $K \subset \mathbf{R}$ , we let  $|\mathcal{K}| = \sup\{\text{diam}(K) : K \in \mathcal{K}\}$ . Then we define the  $\phi$ -Hausdorff measure of a set  $M$  by

$$\phi\text{-}m(M) = \liminf_{\varepsilon \downarrow 0} \left\{ \sum_{K \in \mathcal{K}} \phi(\text{diam}(K)) : |\mathcal{K}| < \varepsilon, M \subset \bigcup_{K \in \mathcal{K}} K \right\}.$$

**Theorem 4.2.** *Fix any  $\gamma < \infty$  and let  $\phi(\delta) = \delta|\log(1/\delta)|^\gamma$ . Then for almost all Brownian trajectories  $g(t)$  the following is true. There exist sets  $M_\tau$  and  $M_\sigma$  such that  $\phi\text{-}m(M_\tau) = \phi\text{-}m(M_\sigma) = 0$ , and  $\mu_\tau^{\mathbf{R}}(M_\tau^c) = \mu_\sigma^{\mathbf{R}}(M_\sigma^c) = 0$ .*

**Proof.** Suppose that  $B_t$  is a standard Brownian motion with  $B_0 = 0$ . We will first estimate the probability that  $B_t \leq ((1 + \delta)/\sqrt{2})\sqrt{2t \log \log(1/t)}$  for a small fixed  $\delta > 0$  and all  $t \in (\varepsilon, 1/(2e))$ .

The process  $X_t = e^{t/2}B(e^{-t})$  is Ornstein-Uhlenbeck with drift  $-x/2$  and standard quadratic variation. We will analyze excursions of  $X_t$  from 0 and the local time  $\ell_t$  of  $X_t$  at 0. The excursion measure of those Brownian excursions whose lifetime is greater than  $t$  is equal to  $c_1 t^{-1/2}$ , for some constant  $c_1$ . One can use this and an explicit representation of  $X_t$  in terms of Brownian motion to see that the excursion measure of  $X_t$ -excursions with lifetime greater than  $t$  is bounded by  $c_2 \exp(-t/2)$ . If  $\alpha_s = \inf\{t : \ell_t > s\}$  then  $\alpha_s$  is a subordinator whose Lévy measure has exponentially decaying tail. Easy calculations (see [Be], p. 72) then show that  $E \exp(\lambda \alpha_s) \leq \exp(c_3 s)$ , for  $\lambda < 1/2$ , where  $c_3 = c_3(\lambda) < \infty$ . This implies that for  $c_3 = c_3(1/4)$ ,

$$\begin{aligned} P(\ell_t \leq (1/8c_3)t) &= P((1/4)\alpha_{(1/8c_3)t} \geq t/4) \leq e^{-t/4} E \exp((1/4)\alpha_{(1/8c_3)t}) \\ &\leq e^{-t/4} \exp(c_3(1/8c_3)t) = e^{-t/8}. \end{aligned} \tag{4.1}$$

The scale function for the diffusion  $X_t$  is equal to  $S(x) = \int_0^x \exp(y^2/2)dy$ , so if  $X_t$  starts from  $X_0 = \varepsilon$ , the chance that it will hit  $x$  before hitting 0 is equal to

$$S(\varepsilon)/S(x) = \frac{\int_0^\varepsilon \exp(y^2/2)dy}{\int_0^x \exp(y^2/2)dy}.$$

By dividing this quantity by  $\varepsilon$  and letting  $\varepsilon \rightarrow 0$ , we obtain the excursion measure of those excursions which hit level  $x$  before their lifetime—it is equal to  $(\int_0^x \exp(y^2/2)dy)^{-1}$ . Hence the probability that there are no excursions with height exceeding  $x$  starting before the local time exceeds  $(1/8c_3)t$  is equal to  $\exp\left(- (1/8c_3)t (\int_0^x \exp(y^2/2)dy)^{-1}\right)$ . For  $x$  of the form  $x = \sqrt{\gamma \log t}$ , with  $\gamma < 2$ , the last expression is bounded by  $\exp(-c_4 t^{c_5})$ , for some  $c_4, c_5 > 0$ . This and (4.1) show that, for large  $t$ ,

$$P\left(\sup_{1 \leq s \leq t} X_s / \sqrt{\gamma \log s} \leq 1\right) \leq P\left(\sup_{1 \leq s \leq t} X_s \leq \sqrt{\gamma \log t}\right) \leq \exp(-c_6 t^{c_7}),$$

for some  $c_6, c_7 \in (0, 1)$ . Translating this back to the language of Brownian motion  $B_t$ , we obtain

$$P\left(\sup_{e^{-y} \leq s \leq 1/2} B_s / \sqrt{\gamma s \log \log(1/s)} \leq 1\right) \leq \exp(-c_6 y^{c_7}).$$

We will apply this formula with  $e^{-y} = 1/\sqrt{m}$ , where  $m$  is a large integer. Suppose  $0 < \delta < \sqrt{2} - 1$ . Then

$$P\left(\sup_{1/\sqrt{m} \leq s \leq 1/2} \frac{B_s}{((1+\delta)/\sqrt{2})\sqrt{2s \log \log(1/s)}} \leq 1\right) \leq \exp(-c_8 (\log m)^{c_9}). \quad (4.2)$$

Standard estimates ([KS], Problem 2.9.22) show that for arbitrarily large  $c_{10} > 0$  and large  $m$ ,

$$\begin{aligned} P\left(\sup_{0 < t \leq 1/m} B_t \geq c_4 \sqrt{(2/m) \log \log m}\right) &= 2P\left(B_{1/m} \geq c_{10} \sqrt{(2/m) \log \log m}\right) \\ &\leq 2 (\log m)^{-c_{10}^2}. \end{aligned} \quad (4.3)$$

For large  $m$  we have

$$\begin{aligned} ((1+\delta)/\sqrt{2})\sqrt{(2/\sqrt{m}) \log \log \sqrt{m}} \\ > ((1+\delta/2)/\sqrt{2})\sqrt{(2/\sqrt{m}) \log \log \sqrt{m}} + c_{10}\sqrt{(2/m) \log \log m}, \end{aligned}$$

so for large  $m$  and all  $t \geq 1/\sqrt{m}$ ,

$$\begin{aligned} & ((1 + \delta)/\sqrt{2})\sqrt{2t \log \log(1/t)} \\ & > ((1 + \delta/2)/\sqrt{2})\sqrt{(2/\sqrt{m}) \log \log \sqrt{m}} + c_{10}\sqrt{(2/m) \log \log m}. \end{aligned}$$

It follows that if for some  $s \in [0, 1/m]$  and all  $t \in (1/m, 1/2]$  we have

$$B_t - B_s \leq ((1 + \delta/2)/\sqrt{2})\sqrt{2(t-s) \log \log(1/(t-s))}$$

then either

$$B_t \leq ((1 + \delta)/\sqrt{2})\sqrt{2t \log \log(1/t)}$$

for all  $t \in [1/\sqrt{m}, 1/2]$  or

$$B_t \geq c_{10}\sqrt{(2/m) \log \log m}$$

for some  $t \in (0, 1/m]$ . Since  $c_{10}$  is arbitrarily large, we have for any  $c_{11} < \infty$  and sufficiently large  $m$ , using (4.2) and (4.3),

$$\begin{aligned} & P\left(\exists s \in [0, 1/m] \forall t \in (0, 1/2] : B_{s+t} - B_s \leq ((1 + \delta/2)/\sqrt{2})\sqrt{2t \log \log(1/t)}\right) \\ & \leq P\left(\exists s \in [0, \frac{1}{m}] \forall t \in (\frac{1}{m}, 1/2] : B_t - B_s \leq \left(\frac{1 + \delta/2}{\sqrt{2}}\right)\sqrt{2(t-s) \log \log(1/(t-s))}\right) \\ & \leq \exp(-c_8(\log m)^{c_9}) + 2(\log m)^{-c_{10}^2} \leq (\log m)^{-c_{11}}. \end{aligned} \quad (4.4)$$

Let  $N_m$  be the number of intervals of the form  $(j/m, (j+1)/m]$ ,  $j = 0, 1, \dots, m-1$ , such that for some  $s \in (j/m, (j+1)/m]$  and all  $t \in (0, 1/2]$  we have

$$B_{s+t} - B_s \leq ((1 + \delta/2)/\sqrt{2})\sqrt{2t \log \log(1/t)}. \quad (4.5)$$

Then (4.4) implies that  $EN_m \leq m(\log m)^{-c_{11}}$ , for large  $m$ . We have

$$P(N_m \geq m(\log m)^{-c_{11}+1}) \leq \frac{EN_m}{m(\log m)^{-c_{11}+1}} \leq \frac{1}{\log m}.$$

We take  $m_j$  to be the integer part of  $1 + e^{j^2}$  so that

$$P(N_{m_j} \geq m_j(\log m_j)^{-c_{11}+1}) \leq \frac{1}{\log m_j} \leq \frac{1}{j^2}.$$

By the Borel-Cantelli Lemma, with probability 1, only a finite number of events  $\{N_{m_j} \geq m_j(\log m_j)^{-c_{11}+1}\}$  occur.

Let  $W$  be the set of all  $s \in (0, 1]$  such that (4.5) holds for all  $t \in (0, 1/2]$ . The set  $W$  is covered by  $N_m$  intervals of length  $1/m$ . For sufficiently large  $j$ ,  $W$  is covered by at most  $m_j(\log m_j)^{-c_{11}+1}$  intervals of length  $1/m_j$ . Let  $\phi(r) = r(\log(1/r))^{c_{11}-2}$  and note that

$$m_j(\log m_j)^{-c_{11}+1}\phi(1/m_j) = \frac{1}{\log m_j}.$$

Since  $1/\log m_j \rightarrow 0$  as  $j \rightarrow \infty$ , the definition of the Hausdorff measure shows that  $\phi\text{-}m(W) = 0$ . This is true for an arbitrarily large  $c_{11}$ .

Let  $W_n$  be the set of all  $s \in (0, 1]$  such that (4.5) holds for all  $t \in (0, 1/n]$ . For any fixed integer  $n \geq 2$ , one can show that  $\phi\text{-}m(W_n) = 0$ , a.s., in a way completely analogous to the case  $n = 2$  presented above. It follows that, with probability 1,  $\phi\text{-}m\left(\bigcup_{n \geq 2} W_n\right) = 0$ . By symmetry and time-reversal, if we let  $\widehat{W}_n$  be the set of all  $s \in (0, 1]$  such that

$$B_{s-t} - B_s \geq -((1 + \delta/2)/\sqrt{2})\sqrt{2t \log \log(1/t)}$$

holds for all  $t \in (0, s \wedge 1/n]$ , then  $\phi\text{-}m\left(\bigcup_{n \geq 2} \widehat{W}_n\right) = 0$ , a.s. It will suffice to find sets  $M_\tau \subset \bigcup_{n \geq 2} \widehat{W}_n$  and  $M_\sigma \subset \bigcup_{n \geq 2} \widehat{W}_n$  such that  $\mu_\tau^{\mathbf{R}}(M_\tau^c) = \mu_\sigma^{\mathbf{R}}(M_\sigma^c) = 0$ .

The event

$$A_2 = \left\{ \omega : \liminf_{t \downarrow 0} \frac{\omega(t)}{\sqrt{2t \log \log(1/t)}} \geq -\frac{1}{\sqrt{2}} \right\}$$

is one of the BMB local properties according to the proof of Theorem 3.1 (iii). Let  $M_\tau = M_\sigma = \{s \in (0, 1] : g_s^-(\cdot) \in A_2\}$ . By Theorem 4.1, the measures  $\mu_\tau^{\mathbf{R}}$  and  $\mu_\sigma^{\mathbf{R}}$  do not charge the set of  $s$  such that  $g_s^-(\cdot) \in A_2^c$ , i.e.,  $\mu_\tau^{\mathbf{R}}(M_\tau^c) = \mu_\sigma^{\mathbf{R}}(M_\sigma^c) = 0$ . One can easily verify that  $M_\tau = M_\sigma \subset \bigcup_{n \geq 2} \widehat{W}_n$ .  $\square$

**5. Conditional non-intersection probabilities.** Recall various local times  $L_t^1, L_t^2, L_t^3$  and  $L_t^4$  from Section 3. This section is an attempt to shed some light on the mysterious fact that  $L_t^3$  is not a constant multiple of  $L_t^4$  although both local times are defined in terms of excursions of  $X_t$  from  $g(t)$ .

First we will argue that  $L_t^3$  has been defined by counting the “wrong” excursions. Recall the event  $A = \{|g(t+s) - e_t(s)| > 1 \text{ for some } s \geq 0\}$  from (3.4) and other notation used in the definition of  $L_t^4$ . The excursion laws  $H^{(t,x)}$  have been normalized so that  $H^{(t,x)}(A) = 1$ , unless  $H^{(t,x)}$  is identically equal to zero. For  $\varepsilon \in [0, 1]$ , let  $M_\varepsilon$  be the set of  $(t, x)$  such that

$$P(\inf\{s > t : |X_t - g(t)| = 1\} < \inf\{s > t : X_t = g(t)\} \mid X_t = x) = \varepsilon.$$

Let

$$A_\varepsilon = \{e_t(s) \in M_\varepsilon \text{ for some } s \geq 0\}.$$

By the strong Markov property of  $H^{(t,x)}$  applied at the hitting time of  $M_\varepsilon$ , we have  $H^{(t,x)}(A_\varepsilon) = 1/\varepsilon$ . It follows that the excursions of  $X_t$  from  $g(t)$  which hit  $M_\varepsilon$  arrive according to a Poisson point process with intensity  $1/\varepsilon$ , on the time scale defined by the local time  $L_t^4$ . Let  $\widehat{N}_s^\varepsilon$  be the number of excursions of  $X_t$  from  $g(t)$  which started before time  $s$  and which hit the set  $M_\varepsilon$ . Then standard techniques show that for a sequence  $\varepsilon_n$  which decreases to 0 sufficiently fast we have  $L_t^4 = \lim_{n \rightarrow \infty} \varepsilon_n \widehat{N}_t^{\varepsilon_n}$ , for all  $t \geq 0$ , a.s. Hence, one can define  $L_t^4$  using excursions but one has to count excursions which hit  $M_\varepsilon$ , not excursions which depart from  $g(t)$  by  $\varepsilon$  units. Since

$$\lim_{n \rightarrow \infty} \varepsilon_n \widehat{N}_t^{\varepsilon_n} = L_t^4 \neq L_t^1 = \lim_{n \rightarrow \infty} \varepsilon_n N_t^{\varepsilon_n},$$

it is clear that most excursions included in one count do not make the other list. It seems that for a typical excursion which departs  $\varepsilon$  units from  $g(t)$ , the probability that it will hit  $M_1$  is much different from  $\varepsilon$ . One may wonder what this probability is.

For technical reasons, we will not address the last question but a different one which we nevertheless believe is “morally equivalent” to it. We will estimate the probability that Brownian motion starting  $\varepsilon$  units away from  $g(t)$  will not hit the trajectory of  $g(t)$  for one unit of time. The answer, of course, depends on the trajectory of  $g(t)$ . We will try to quantify this dependence.

It will be convenient to present the next result using a product probability space  $\Omega = C[0, 1] \times C[0, 1]$ . Its generic element will be denoted  $\omega = (\omega_1, \omega_2)$ . We will use  $g(t)$  and  $X_t$  to denote the canonical stochastic processes, i.e.,  $g(t)(\omega) = \omega_1(t)$  and  $X_t(\omega) = \omega_2(t)$ .

**Theorem 5.1.** *Let  $P^\varepsilon$  be a probability measure on  $\Omega$  which makes  $g(t)$  and  $X_t$  independent Brownian motions with  $g(0) = 0$  and  $X_0 = -\varepsilon$ .*

(i) *For some  $0 < c_1 < c_2 < \infty$  and all  $\varepsilon \in (0, 1)$ ,*

$$c_1 \varepsilon \leq P^\varepsilon(X_t \neq g(t), 0 \leq t \leq 1) \leq c_2 \varepsilon.$$

(ii) *Let*

$$q(\omega_1) = P^\varepsilon(X_t \neq g(t), 0 \leq t \leq 1 \mid g(\cdot) = \omega_1).$$

*Then for any  $\gamma_1, \gamma_2 > 0$  with  $\gamma_1 + \gamma_2 < 1$ , and all sufficiently small  $\varepsilon > 0$ ,*

$$P^\varepsilon(q > \varepsilon |\log \varepsilon|^{-\gamma_1}) < |\log \varepsilon|^{-\gamma_2}. \quad (5.1)$$

The proof of the above result hinges on the estimate in the next theorem. That theorem is concerned with distributions of stochastic processes which are not necessarily absolutely continuous with respect to each other. We will give a meaning to the Radon-Nikodym derivative for such measures. Suppose that  $Q$  and  $P$  are probability measures. The measure  $Q$  can be represented as  $Q_1 + Q_2$ , the sum of two non-negative measures, such that  $Q_1$  is absolutely continuous with respect to  $P$  and  $Q_2$  is singular with respect to  $P$ . Then we define the Radon-Nikodym derivative  $dQ/dP$  as  $dQ_1/dP$ . Note that the integral of  $dQ/dP$  with respect to  $P$  may take any value in  $[0, 1]$ .

**Theorem 5.2.** *Suppose that  $B_t$  and  $X_t$  are standard Brownian motions and  $Y_t$  is a process which does not take positive values. Assume that  $B_t, X_t$  and  $Y_t$  are independent and  $B_0 = X_0 = Y_0 = -\varepsilon < 0$ . If  $Q_1$  denotes the distribution of  $\{(B_t + Y_t)/\sqrt{2}, t \in [0, 1]\}$  on  $C[0, 1]$  and  $Q_2$  is the analogous distribution of  $\{(B_t + X_t)/\sqrt{2}, t \in [0, 1]\}$  then for any  $\gamma_1, \gamma_2 > 0$  with  $\gamma_1 + \gamma_2 < 1$ , for sufficiently small  $\varepsilon > 0$  we have*

$$Q_2(dQ_1/dQ_2 > |\log \varepsilon|^{-\gamma_1}) < |\log \varepsilon|^{-\gamma_2}.$$

**Remark 5.1.** Before we prove Theorems 5.1 and 5.2, we would like to describe a seemingly natural strategy to obtain a good estimate of  $dQ_1/dQ_2$  in Theorem 5.2 which in fact leads to a vicious circle of ideas. We will only sketch the argument and leave the details to the reader.

We will use the notation of Theorem 5.2 except that the processes will start above the axis—in other words, we will flip them to the positive side. Let  $Y_t$  be a Brownian motion starting from  $\varepsilon > 0$  under a probability measure  $P$ . It is a standard exercise to check that if

$$M_t = \frac{Y_t}{Y_0} \mathbf{1}_{\{Y_s > 0, 0 \leq s \leq t\}}, \quad (5.2)$$

and  $Q$  is defined by  $dQ/dP = M_t$  on  $C[0, t]$  then  $Y_t$  is a three-dimensional Bessel process ([RY], page 446) under  $Q$ .

Consider processes  $B_t$  and  $Y_t$ . Suppose that under  $Q_1$  these processes are independent,  $B_0 = Y_0 = \varepsilon > 0$ ,  $B_t$  is a Brownian motion and  $Y_t$  is a three-dimensional Bessel process. Let  $Q_2$  be a probability measure which makes  $B_t$  and  $Y_t$  independent Brownian motions

starting from  $\varepsilon$ . Since  $B_t$  is independent of  $Y_t$  and has the same distribution under both measures,

$$\frac{dQ_1}{dQ_2}((B_s, Y_s), 0 \leq s \leq t) = \frac{dQ_1}{dQ_2}(Y_s, 0 \leq s \leq t) = M_t.$$

Let  $W_t = (B_t + Y_t)/\sqrt{2}$  and  $R_t = (-B_t + Y_t)/\sqrt{2}$ . The goal of Theorem 5.2 is to find an estimate for the following Radon-Nikodym derivative,

$$\begin{aligned} \frac{dQ_1}{dQ_2}(W_s, 0 \leq s \leq t) &= E^{Q_2}(M_t \mid W_s, 0 \leq s \leq t) \\ &= E^{Q_2}\left(\frac{Y_t}{Y_0} \mathbf{1}_{\{Y_s > 0, 0 \leq s \leq t\}} \mid W_s, 0 \leq s \leq t\right) \\ &= E^{Q_2}\left(\frac{W_t + R_t}{W_0 + R_0} \mathbf{1}_{\{W_s + R_s > 0, 0 \leq s \leq t\}} \mid W_s, 0 \leq s \leq t\right) \\ &= \frac{1}{2\varepsilon} E^{Q_2}\left((W_t + R_t) \mathbf{1}_{\{W_s + R_s > 0, 0 \leq s \leq t\}} \mid W_s, 0 \leq s \leq t\right). \end{aligned} \quad (5.3)$$

Under  $Q_2$ , the processes  $W_t$  and  $R_t$  are independent Brownian motions so  $W_t + R_t$  takes mostly moderate values. Hence it is natural to expect that the last line in (5.3) is as hard to estimate as

$$P^{Q_2}(W_s + R_s > 0, 0 \leq s \leq t \mid W_s, 0 \leq s \leq t).$$

This conditional probability is analogous to the one which is estimated in Theorem 5.1 (ii). The estimate in Theorem 5.1 (ii) is based on an estimate of  $dQ_1/dQ_2$  in Theorem 5.2 and thus the vicious circle of ideas is closed. We will use an argument unrelated to the Girsanov theorem to prove Theorem 5.2.

**Proof of Theorem 5.2.** Let  $B_t$  denote the standard Brownian motion with  $B_0 = 0$ . We will need several parameters whose values will be specified later. Suppose  $\alpha \in (0, 1)$ , consider a small  $\varepsilon > 0$ , and let  $k$  be the largest integer with  $\alpha^{k/2} \geq \varepsilon$ . Let  $c > 0$  and  $a = c\sqrt{\log k}$ . We will write  $t_n = \alpha^n$ . Using the well known formula for the distribution of the maximum of Brownian motion and (4.1), we obtain for all  $n \leq k$ , assuming  $k$  is sufficiently large,

$$P\left(\max_{0 \leq s \leq t_n} B_s \geq a\sqrt{t_n}\right) \leq \frac{\sqrt{2}}{c\sqrt{\pi \log k}} k^{-c^2/2}. \quad (5.4)$$

We will assume that  $c^2/2 > 1$  from now on. We obtain from (5.4), for some fixed  $\beta_1 > 0$  and large  $k$ ,

$$P\left(\bigcup_{0 \leq n \leq k} \left\{ \max_{0 \leq s \leq t_n} B_s \geq a\sqrt{t_n} \right\}\right) \leq \frac{\sqrt{2}}{c\sqrt{\pi \log k}} k^{1-c^2/2} \leq |\log \varepsilon|^{-\beta_1}. \quad (5.5)$$

Next consider  $b \in (0, 1)$ . Let  $\delta, \delta_1 > 0$  be arbitrarily small constants. We have

$$\sqrt{2}\varepsilon < \delta ab\sqrt{t_{n-1} - t_n}$$

for large  $k$  and all  $n \leq k$ . We obtain for large  $k$ , using the lower bound in (4.1),

$$\begin{aligned} & P\left(\max_{t_n \leq s \leq t_{n-1}} B_s - B_{t_n} > \sqrt{2}\varepsilon + ab\sqrt{t_{n-1} - t_n}\right) \\ & \geq 2P(B_1 > (1 + \delta)ab) \\ & \geq \frac{(1 - \delta_1)\sqrt{2}}{(1 + \delta)bc\sqrt{\pi \log k}} k^{-(1 + \delta)^2 b^2 c^2 / 2}. \end{aligned}$$

This implies for large  $k$ ,

$$\begin{aligned} & P\left(\bigcup_{1 \leq n \leq k} \left\{ \max_{t_n \leq s \leq t_{n-1}} B_s - B_{t_n} > \sqrt{2}\varepsilon + ab\sqrt{t_{n-1} - t_n} \right\}\right) \\ & \geq 1 - \left(1 - \frac{(1 - \delta_1)\sqrt{2}}{(1 + \delta)bc\sqrt{\pi \log k}} k^{-(1 + \delta)^2 b^2 c^2 / 2}\right)^k \\ & \geq 1 - \exp\left(-\frac{(1 - \delta_1)\sqrt{2}}{(1 + \delta)bc\sqrt{\pi \log k}} k^{1 - (1 + \delta)^2 b^2 c^2 / 2}\right). \end{aligned} \quad (5.6)$$

We will choose the parameters so that  $1 - (1 + \delta)^2 b^2 c^2 / 2 > 0$ .

It is time for us to specify the values of the parameters or, to be more precise, the ranges of their values. First we fix some  $b \in (1/\sqrt{2}, 1)$ . Next we choose  $\alpha \in (0, 1)$  so small that

$$1 < \sqrt{\frac{1}{\alpha} - 1} - \frac{1}{b\sqrt{2\alpha}}.$$

Set  $\gamma = \sqrt{\frac{1}{\alpha} - 1} - \frac{1}{b\sqrt{2\alpha}}$ , and choose  $\delta > 0$  such that  $\delta < \min(\gamma - 1, \frac{1}{b} - 1)$ . We find some  $c$  satisfying

$$\max\left(\frac{\sqrt{2}}{b\gamma}, \sqrt{2}\right) < c < \frac{\sqrt{2}}{b(1 + \delta)}.$$

Finally we choose  $c_1$  such that  $\sqrt{2} < c_1 < cb\gamma$ , and let  $d = c_1\sqrt{\log k}$ .

The following inequality is completely analogous to (5.5). We have replaced the maximum with the minimum,  $a$  with  $-d$  and  $c$  with  $c_1$ .

$$P\left(\bigcup_{0 \leq n \leq k} \left\{ \min_{0 \leq s \leq t_n} B_s \leq -d\sqrt{t_n} \right\}\right) \leq \frac{\sqrt{2}}{c_1\sqrt{\pi \log k}} k^{1 - c_1^2 / 2}. \quad (5.7)$$

Assume that the following events hold

$$\bigcup_{1 \leq n \leq k} \left\{ \max_{t_n \leq s \leq t_{n-1}} B_s - B_{t_n} > \sqrt{2}\varepsilon + ab\sqrt{t_{n-1} - t_n} \right\} \quad (5.8)$$

and

$$\bigcap_{0 \leq n \leq k} \left\{ \min_{0 \leq s \leq t_n} B_s > -d\sqrt{t_n} \right\}. \quad (5.9)$$

Then for some  $n \in \{1, 2, \dots, k\}$  and  $s \in [t_n, t_{n-1}]$ ,

$$\begin{aligned} B_s &= (B_s - B_{t_n}) + B_{t_n} > \sqrt{2}\varepsilon + ab\sqrt{t_{n-1} - t_n} - d\sqrt{t_n} \\ &= \sqrt{2}\varepsilon + ab\sqrt{\alpha^{n-1} - \alpha^n} - d\sqrt{\alpha^n} \\ &= \sqrt{2}\varepsilon + \sqrt{\alpha^{n-1}}(ab\sqrt{1 - \alpha} - d\sqrt{\alpha}) \\ &\geq \sqrt{2}\varepsilon + \sqrt{t_{n-1}}a/\sqrt{2}. \end{aligned}$$

We see that if the events (5.8) and (5.9) occurred then the following event holds,

$$\bigcup_{1 \leq n \leq k} \left\{ \max_{t_n \leq s \leq t_{n-1}} B_s > \sqrt{2}\varepsilon + (a/\sqrt{2})\sqrt{t_{n-1}} \right\}.$$

Recall that  $c_1 > \sqrt{2}$  and  $1 - (1 + \delta)^2 b^2 c^2 / 2 > 0$ . Then, in view of (5.6) and (5.7), for some fixed  $\delta_2, \beta_2 > 0$  and large  $k$ ,

$$\begin{aligned} &P \left( \bigcup_{1 \leq n \leq k} \left\{ \max_{t_n \leq s \leq t_{n-1}} B_s > \sqrt{2}\varepsilon + (a/\sqrt{2})\sqrt{t_{n-1}} \right\} \right) \\ &\geq 1 - \exp \left( -\frac{(1 - \delta_1)\sqrt{2}}{(1 + \delta)bc\sqrt{\pi \log k}} k^{1 - (1 + \delta)^2 b^2 c^2 / 2} \right) - \frac{\sqrt{2}}{c_1 \sqrt{\pi \log k}} k^{1 - c_1^2 / 2} \\ &\geq 1 - k^{1 - (1 - \delta_2)c_1^2 / 2} \\ &\geq 1 - |\log \varepsilon|^{-\beta_2}. \end{aligned} \quad (5.10)$$

Let us examine possible values of exponents  $\beta_1$  and  $\beta_2$  in (5.5) and (5.10). We can choose  $b$  very close to  $1/\sqrt{2}$  and so  $c$  can be very close to 2.

It follows that (5.5) holds for any  $\beta_1 < 1$ . Given any  $b \in (1/\sqrt{2}, 1)$ ,  $\alpha$  can be chosen very close to 0 and then  $c_1$  can be made arbitrarily large. We conclude that (5.10) holds for any  $\beta_2 < \infty$ .

Now recall processes  $B_t, X_t$  and  $Y_t$  from the statement of the theorem. The processes  $B_t$  and  $X_t$  are standard Brownian motions and  $Y_t$  is a process which does not take positive

values. We have assumed that  $B_t, X_t$  and  $Y_t$  are independent and  $B_0 = X_0 = Y_0 = -\varepsilon < 0$ . We immediately obtain from (5.5) that for  $\beta_1 < 1$  and small  $\varepsilon > 0$ ,

$$P \left( \bigcup_{0 \leq n \leq k} \left\{ \max_{0 \leq s \leq t_n} (B_s + Y_s)/\sqrt{2} \geq (a/\sqrt{2})\sqrt{t_n} \right\} \right) \leq |\log \varepsilon|^{-\beta_1}. \quad (5.11)$$

Since  $(B_t + X_t)/\sqrt{2}$  is a standard Brownian motion starting from  $-\sqrt{2}\varepsilon$ , (5.10) implies

$$P \left( \bigcup_{1 \leq n \leq k} \left\{ \max_{t_n \leq s \leq t_{n-1}} (B_s + X_s)/\sqrt{2} \geq (a/\sqrt{2})\sqrt{t_{n-1}} \right\} \right) \geq 1 - |\log \varepsilon|^{-\beta_2}, \quad (5.12)$$

for any fixed  $\beta_2 < \infty$  and small  $\varepsilon > 0$ .

Consider the following event,

$$A = \bigcup_{1 \leq n \leq k} \left\{ \omega \in C[0, 1] : \max_{t_n \leq s \leq t_{n-1}} \omega_s \geq (a/\sqrt{2})\sqrt{t_{n-1}} \right\}.$$

Recall measures  $Q_1$  and  $Q_2$  from the statement of the theorem. We can rewrite (5.11) and (5.12) as

$$Q_1(A) \leq |\log \varepsilon|^{-\beta_1}$$

and

$$Q_2(A) \geq 1 - |\log \varepsilon|^{-\beta_2}.$$

Let  $B = \{dQ_1/dQ_2 > |\log \varepsilon|^{-\gamma_1}\}$  and assume that  $Q_2(A \cap B) \geq |\log \varepsilon|^{-\gamma_2}$ . Then

$$|\log \varepsilon|^{-\beta_1} \geq Q_1(A) \geq Q_1(A \cap B) \geq \int_{A \cap B} \frac{dQ_1}{dQ_2}(\omega) dQ_2(\omega) \geq |\log \varepsilon|^{-\gamma_1 - \gamma_2}. \quad (5.13)$$

We can take  $\beta_1$  arbitrarily close to 1, in particular, we can assume that  $\gamma_1 + \gamma_2 < \beta_1$  because  $\gamma_1 + \gamma_2 < 1$ . Then (5.13) gives us a contradiction for  $\varepsilon > 0$  small and we have to conclude that  $Q_2(A \cap B) < |\log \varepsilon|^{-\gamma_2}$ . Hence,

$$\begin{aligned} Q_2(dQ_1/dQ_2 > |\log \varepsilon|^{-\gamma_1}) &= Q_2(B) = Q_2(A \cap B) + Q_2(A^c \cap B) \\ &\leq Q_2(A \cap B) + Q_2(A^c) < |\log \varepsilon|^{-\gamma_2} + |\log \varepsilon|^{-\beta_2}. \end{aligned}$$

The exponent  $\beta_2$  can be taken arbitrarily large so

$$Q_2(dQ_1/dQ_2 > |\log \varepsilon|^{-\gamma_1}) < |\log \varepsilon|^{-\gamma_2},$$

for any pair  $\gamma_1, \gamma_2$  with  $\gamma_1 + \gamma_2 < 1$ , and sufficiently small  $\varepsilon > 0$ .  $\square$

**Proof of Theorem 5.1.** (i) The process  $Z_t = (X_t - g(t))/\sqrt{2}$  is a Brownian motion starting from  $-\varepsilon/\sqrt{2}$ . The event in question is equivalent to  $A = \{Z_t < 0, 0 \leq t \leq 1\}$ . By the gambler's ruin problem, the probability that  $Z_t$  will hit  $-1$  before hitting  $0$  is equal to  $\varepsilon/\sqrt{2}$ . Standard arguments can be used to show that this implies that  $P(A)$  is of order  $\varepsilon$ .

(ii) In addition to  $Z_t$  defined in part (i) of the proof, we will need  $W_t = (X_t + g(t))/\sqrt{2}$ . Under  $P^\varepsilon$ , the processes  $Z_t$  and  $W_t$  are independent Brownian motions starting from  $-\varepsilon/\sqrt{2}$ . Let  $Q^\varepsilon$  denote the probability measure  $P^\varepsilon$  conditioned by  $A$ , the event defined in part (i). Under  $Q^\varepsilon$ , the processes  $Z_t$  and  $W_t$  are independent,  $W_t$  is a Brownian motion starting from  $-\varepsilon/\sqrt{2}$  and  $-Z_t$  is a three-dimensional Bessel process; hence  $Z_t$  is a non-positive process. Note that  $g_t = (W_t - Z_t)/\sqrt{2}$ .

The estimate in part (i) of the theorem yields

$$\begin{aligned} P^\varepsilon(q > c_2\varepsilon|\log\varepsilon|^{-\gamma_1}) &< P^\varepsilon(P^\varepsilon(A|g) > P^\varepsilon(A)|\log\varepsilon|^{-\gamma_1}) \\ &= P^\varepsilon\left(E_{P^\varepsilon}\left(\frac{dQ^\varepsilon}{dP^\varepsilon} \mid g\right) > |\log\varepsilon|^{-\gamma_1}\right). \end{aligned} \quad (5.14)$$

Consider the probabilities

$$Q_1 = Q^\varepsilon \circ \left(\frac{W+Z}{\sqrt{2}}\right)^{-1}$$

and

$$Q_2 = P^\varepsilon \circ \left(\frac{W+Z}{\sqrt{2}}\right)^{-1}.$$

Note that  $Q_1$  is absolutely continuous with respect to  $Q_2$ . For any Borel set  $G$  in  $C[0, \infty)$  we have

$$\begin{aligned} \int_{\{g \in G\}} E_{P^\varepsilon}\left(\frac{dQ^\varepsilon}{dP^\varepsilon} \mid g\right) dP^\varepsilon &= Q^\varepsilon(g \in G) = Q^\varepsilon\left(\frac{W-Z}{\sqrt{2}} \in G\right) \\ &= Q^\varepsilon\left(\frac{W+Z}{\sqrt{2}} \in -G - \varepsilon\right) = Q_1(-G - \varepsilon) \\ &= \int_{-G - \varepsilon} \frac{dQ_1}{dQ_2} dQ_2 = \int_{\{g \in G\}} \frac{dQ_1}{dQ_2}(-\omega - \varepsilon) dP^\varepsilon(\omega). \end{aligned}$$

Hence, a.s.,

$$E_{P^\varepsilon}\left(\frac{dQ^\varepsilon}{dP^\varepsilon} \mid g\right)(\omega) = \frac{dQ_1}{dQ_2}(-\omega - \varepsilon)$$

and we deduce

$$P^\varepsilon\left(E_{P^\varepsilon}\left(\frac{dQ^\varepsilon}{dP^\varepsilon} \mid g\right) > |\log\varepsilon|^{-\gamma_1}\right) = Q_2\left(\frac{dQ_1}{dQ_2} > |\log\varepsilon|^{-\gamma_1}\right). \quad (5.15)$$

Theorem 5.2 can be applied to measures  $Q_2$  and  $Q_1$ , and from (5.14) and (5.15) we obtain for any fixed  $\gamma_1, \gamma_2 > 0$  with  $\gamma_1 + \gamma_2 < 1$ , and small  $\varepsilon > 0$ ,

$$P^\varepsilon (q > c_2 \varepsilon |\log \varepsilon|^{-\gamma_1}) < Q_2 \left( \frac{dQ_1}{dQ_2} > |\log \varepsilon|^{-\gamma_1} \right) < |\log \varepsilon|^{-\gamma_2}.$$

By slightly adjusting the values of  $\gamma_1$  and  $\gamma_2$  we can eliminate  $c_2$  from the formula. This completes the proof of (5.1).  $\square$

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Krzysztof Burdzy  
 Department of Mathematics  
 University of Washington  
 Box 354350 Seattle, WA 98195-4350, USA  
 e-mail: burdzy@math.washington.edu

David Nualart  
 Facultat de Matemàtiques  
 Universitat de Barcelona  
 Gran Via 585, 08007 Barcelona, Spain  
 e-mail:nualart@mat.ub.es