

ASYMPTOTICS FOR THE HEAT EQUATION  
IN THE EXTERIOR OF A SHRINKING COMPACT SET  
IN THE PLANE VIA BROWNIAN HITTING TIMES

ROSS G. PINSKY

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ABSTRACT. Let  $D_r = \{x \in \mathbb{R}^2 : |x| \leq r\}$  and let  $\gamma$  be a continuous, non-increasing function on  $[0, \infty)$  satisfying  $\lim_{t \rightarrow \infty} \gamma(t) = 0$ . Consider the heat equation in the exterior of a time-dependent shrinking disk in the plane:

$$\begin{aligned}u_t &= \frac{1}{2} \Delta u, \quad x \in \mathbb{R}^2 - D_{\gamma(t)}, \quad t > 0, \\u(x, 0) &= 0, \quad x \in \mathbb{R}^2 - D_{\gamma(t)}, \\u(x, t) &= 1, \quad x \in D_{\gamma(t)}, \quad t > 0.\end{aligned}$$

If there exist constants  $0 < c_1 < c_2$  and a constant  $k > 0$  such that  $c_1 t^{-k} \leq \gamma(t) \leq c_2 t^{-k}$ , for sufficiently large  $t$ , then  $\lim_{t \rightarrow \infty} u(x, t) = \frac{1}{1+2k}$ . The same result is also shown to hold when  $D_{\gamma(t)}$  is replaced by  $L_{\gamma(t)}$ , where  $L_r = \{(x_1, 0) \in \mathbb{R}^2 : |x_1| \leq r\}$ . Also, a discrepancy is noted between the asymptotics for the above forward heat equation and the corresponding backward one. The method used is probabilistic.

1. STATEMENT OF RESULTS

Let

$$D_r = \{x \in \mathbb{R}^d : |x| \leq r\}$$

and consider the heat equation in the exterior domain  $\mathbb{R}^d - D_1$ :

$$(1.1) \quad \begin{aligned}w_t &= \frac{1}{2} \Delta w, \quad (x, t) \in (\mathbb{R}^d - D_1) \times (0, \infty), \\w(x, 0) &= 0, \quad x \in \mathbb{R}^d - D_1, \\w(x, t) &= 1, \quad (x, t) \in D_1 \times [0, \infty).\end{aligned}$$

The asymptotic behavior of the solution  $w(x, t)$  for  $x \in \mathbb{R}^d - D_1$  as  $t \rightarrow \infty$  is well known:

$$\lim_{t \rightarrow \infty} w(x, t) = \begin{cases} |x|^{2-d}, & \text{if } d \geq 3, \\ 1, & \text{if } d = 1, 2. \end{cases}$$

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In particular, this shows that a heat source in the shape of a ball does not fully heat three-dimensional space, whereas a heat source in the shape of a disk does fully heat the plane.

The question we address here is this: how effective will the heating be in the plane if the heat source shrinks in time? That is, instead of having a temporally constant heat source  $D_1$ , we use a heat source  $D_{\gamma(t)}$ , where  $\gamma$  is a continuous, positive function decreasing to 0 as  $t \rightarrow \infty$ . We have the following equation:

$$(1.2) \quad \begin{aligned} u_t &= \frac{1}{2} \Delta u, \quad x \in R^2 - D_{\gamma(t)}, \quad t > 0, \\ u(x, 0) &= 0, \quad x \in R^2 - D_{\gamma(t)}, \\ u(x, t) &= 1, \quad x \in D_{\gamma(t)}, \quad t \geq 0. \end{aligned}$$

If  $\lim_{t \rightarrow \infty} u(x, t)$  exists, then it should be harmonic in  $R^2 - \{0\}$ , and since the only bounded harmonic functions in this domain are the constants, it follows that the above limit, if it exists, should be equal to a constant  $c \in [0, 1]$ . Incidentally, the same type of argument shows that the corresponding problem in higher dimensions is uninteresting: if the limit exists, it should be harmonic in  $R^d - 0$ , and dominated by  $|x|^{2-d}$ ; the only such bounded, nonnegative function is 0. A completely rigorous argument can easily be made using the maximum principle.

Before stating the theorem, we derive a probabilistic representation for the solution  $u$  to (1.2). We recall a fundamental property of Brownian motion: let  $P_x$  denote Wiener measure for a standard Brownian motion  $B(t)$  in  $R^d$  starting from  $x \in R^d$  and let  $E_x$  denote the corresponding expectation. Let  $G \subset R^d$  be a domain and let  $\sigma_{G^c} = \inf\{t \geq 0 : B(t) \in G^c\}$  denote the first entrance time into  $G^c$ . If  $W(x, s)$  is twice continuously differentiable for  $x \in G$ , once continuously differentiable for  $s \in (0, t]$ , and continuous on  $\bar{G} \times (0, t] \cup G \times [0, t]$ , and if  $\frac{1}{2} \Delta W + \frac{\partial W}{\partial s}$  is bounded on  $G \times (0, t]$ , then  $W(B(s \wedge \sigma_{G^c}), s \wedge \sigma_{G^c}) - \int_0^{s \wedge \sigma_{G^c}} (\frac{1}{2} \Delta W + \frac{\partial W}{\partial s})(B(s'), s') ds'$  is a martingale under  $P_x$  for  $s \in [0, t]$ . In particular, if it so happens that  $\frac{1}{2} \Delta W + \frac{\partial W}{\partial s} \equiv 0$  in  $G \times (0, t)$ , then  $W(B(s \wedge \sigma_{G^c}), s \wedge \sigma_{G^c})$  is a martingale. Since the expectation of a martingale is constant in time, equating the value of the expectation when  $s = 0$  with the value for  $s = t$ , we obtain the formula

$$(1.3) \quad W(x, 0) = E_x W(B(t \wedge \sigma_{G^c}), t \wedge \sigma_{G^c}), \quad \text{for } x \in G.$$

Equation (1.3) leads to the well-known probabilistic representation for  $w(x, t)$ . Fix  $t > 0$  and apply (1.3) to  $W(x, s) = w(x, t - s)$  and  $G = R^d - D_1$ . From the boundary condition and the initial condition in (1.1) it follows that

$$W(B(t \wedge \sigma_{G^c}), t \wedge \sigma_{G^c}) = w(B(t \wedge \sigma_{D_1}), t - t \wedge \sigma_{D_1}) = 1$$

on the event  $\{\sigma_{D_1} < t\}$ , while

$$W(B(t \wedge \sigma_{G^c}), t \wedge \sigma_{G^c}) = w(B(t \wedge \sigma_{D_1}), t - t \wedge \sigma_{D_1}) = 0$$

on the event  $\{\sigma_{D_1} > t\}$ . Since  $P_x(\sigma_{D_1} = t) = 0$ , we obtain from (1.3) that

$$(1.4) \quad w(x, t) = P_x(\sigma_{D_1} \leq t) \quad \text{for } x \in R^d - D_1.$$

From (1.4), it follows that the dichotomy between  $\lim_{t \rightarrow \infty} w(x, t) = 1$  for  $d = 1, 2$  and  $\lim_{t \rightarrow \infty} w(x, t) < 1$  for  $d \geq 3$  which we observed above is just the dichotomy between recurrence and transience of Brownian motion.

In order to apply (1.3) to the solution  $u$  of (1.2), we define for each  $t > 0$  the stopping time

$$\sigma_\gamma^{(t)} = \inf\{s \in [0, t] : B(s) \in D_{\gamma(t-s)}\},$$

with the convention that  $\sigma_\gamma^{(t)} = \infty$  if  $\{s \in [0, t] : B(s) \in D_{\gamma(t-s)}\}$  is empty. Fixing a  $t > 0$ , consider the function  $W(x, s) \equiv u(x, t - s)$ ,  $0 \leq s \leq t$ ,  $x \in \mathbb{R}^2 - D_{\gamma(t-s)}$ . By (1.2), we have  $(\frac{1}{2}\Delta W + \frac{\partial W}{\partial s})(x, s) \equiv 0$  for  $x \in \mathbb{R}^2 - D_{\gamma(t-s)}$  and  $0 \leq s \leq t$ . We now apply (1.4) to this choice of  $W(x, s)$  along with the stopping time  $\sigma_\gamma^{(t)}$ . The fact that the domain is time-dependent does not cause any problem. Note from the boundary condition and the initial condition in (1.2) that  $W(B(t \wedge \sigma_\gamma^{(t)}), t \wedge \sigma_\gamma^{(t)}) = u(B(t \wedge \sigma_\gamma^{(t)}), t - t \wedge \sigma_\gamma^{(t)}) = 1$  on the event  $\{\sigma_\gamma^{(t)} < t\}$ , while  $W(B(t \wedge \sigma_\gamma^{(t)}), t \wedge \sigma_\gamma^{(t)}) = u(B(t \wedge \sigma_\gamma^{(t)}), t - t \wedge \sigma_\gamma^{(t)}) = 0$  on the event  $\{\sigma_\gamma^{(t)} > t\}$ . Since  $P_x(\sigma_\gamma^{(t)} = t) = 0$ , we obtain the following probabilistic representation for the solution  $u$  of (1.2):

$$(1.5) \quad u(x, t) = P_x(\sigma_\gamma^{(t)} \leq t) = P_x(B(s) \in D_{\gamma(t-s)} \text{ for some } s \in [0, t]).$$

We will prove the following theorem.

**Theorem.** *Let  $u(x, t)$  be the solution to (1.2), where  $\gamma(t)$  is a continuous, nonincreasing function. Assume that there exist constants  $0 < c_1 < c_2$  and a constant  $k > 0$  such that  $c_1 t^{-k} \leq \gamma(t) \leq c_2 t^{-k}$  for sufficiently large  $t$ . Then*

$$(1.6) \quad \lim_{t \rightarrow \infty} u(x, t) = \frac{1}{1 + 2k}.$$

*Remark.* By a basic monotonicity property which is an obvious consequence of the maximum principle, it follows that if  $\gamma(t)$  decreases to 0 faster than any negative power of  $t$ , then  $\lim_{t \rightarrow \infty} u(x, t) = 0$ , while if  $\gamma(t)$  decreases to 0 more slowly than any negative power of  $t$ , then  $\lim_{t \rightarrow \infty} u(x, t) = 1$ .

With just a little extra work, we will prove the same result when  $D_r$  is replaced by a one-dimensional set. Let

$$L_r = \{(x_1, 0) \in \mathbb{R}^2 : |x_1| \leq r\}.$$

**Corollary.** *Let  $U(x, t)$  denote the solution to (1.2) with  $D_{\gamma(t)}$  replaced by  $L_{\gamma(t)}$ . Then (1.6) holds with  $u$  replaced by  $U$ .*

*Remark.* By the basic monotonicity property, the same result also holds if  $D_{\gamma(t)}$  or  $L_{\gamma(t)}$  is replaced by  $C_{\gamma(t)}$  where  $r \rightarrow C_r$  is a continuous map from  $[0, \infty)$  to the compact subsets of  $\mathbb{R}^2$ , satisfying  $L_r \subset C_r \subset D_r$ .

There turns out to be an interesting discrepancy between the asymptotic behavior of the solution  $u$  to the forward heat equation (1.2) and the solution to the corresponding backward heat equation. Let  $v^{(t)}(x, s)$  denote the solution to the backward heat equation

$$\begin{aligned} v_s^{(t)} + \frac{1}{2}\Delta v^{(t)} &= 0, \quad x \in \mathbb{R}^2 - D_{\gamma(s)}, \quad 0 \leq s \leq t, \\ v^{(t)}(x, t) &= 0, \quad x \in \mathbb{R}^2 - D_{\gamma(t)}, \\ v^{(t)}(x, s) &= 1, \quad x \in D_{\gamma(s)}, \quad 0 \leq s \leq t. \end{aligned}$$

Consider the stopping time

$$\sigma_\gamma = \inf\{t \geq 0 : B(t) \in D_{\gamma(t)}\},$$

and define

$$v(x, t) = P_x(\sigma_\gamma \leq t).$$

An analysis similar to that used to obtain the probabilistic representation of  $u$  in (1.5) shows that  $v(x, t) = v^{(t)}(x, 0)$ . (Use (1.3) with  $W(x, s) = v_s^t(x, s)$ ,  $0 \leq s \leq t$ , and use  $\sigma_\gamma$  in place of  $\sigma_{G^c}$ .) An old result of Spitzer [3] indicates that for  $|x| > \gamma(0)$ ,

$$(1.7) \quad P_x\left(\frac{|B(t)|}{\gamma(t)} \leq 1 \text{ for some } t \geq 0\right) = 1 \text{ if and only if } \int^\infty \frac{1}{t|\log \gamma(t)|} dt = \infty.$$

In terms of  $v^{(t)}$ , (1.7) states that

$$(1.8) \quad \begin{aligned} \lim_{t \rightarrow \infty} v^{(t)}(x, 0) &= \lim_{t \rightarrow \infty} P_x(B(s) \in D_{\gamma(s)} \text{ for some } s \in [0, t]) = 1 \\ &\text{if and only if } \int^\infty \frac{1}{t|\log \gamma(t)|} dt = \infty. \end{aligned}$$

Note, for example, that the integral in (1.8) will be infinite if  $\gamma(t) \geq t^{-\log \log t}$ , but not if  $\gamma(t) \leq t^{-(\log \log t)^{1+\epsilon}}$ . In contrast, by Theorem 1, it follows that

$$(1.9) \quad \begin{aligned} &\lim_{t \rightarrow \infty} u(x, t) \\ &= \lim_{t \rightarrow \infty} P_x(B(s) \in D_{\gamma(t-s)}, \text{ for some } s \in [0, t]) \end{aligned} \begin{cases} = 1, & \text{if } \lim_{t \rightarrow \infty} t^k \gamma(t) = \infty, \\ & \text{for all } k > 0, \\ \neq 1, & \text{if } \lim_{t \rightarrow \infty} t^k \gamma(t) = 0, \\ & \text{for some } k > 0. \\ = 0, & \text{if } \lim_{t \rightarrow \infty} t^k \gamma(t) = 0, \\ & \text{for all } k > 0. \end{cases}$$

Thus, there is a discrepancy between the asymptotic behavior of the solutions to the forward and backward heat equations. In probabilistic terms, there is a discrepancy in the asymptotic behavior of the hitting times of a shrinking disk, depending on whether the shrinking occurs in forward time or backward time. In particular, note that if  $\gamma(t)$  decays to 0 faster than any negative power of  $t$ , but  $\int^\infty \frac{1}{t|\log \gamma(t)|} dt = \infty$ , as occurs for example if  $\gamma(t) = t^{-\log \log t}$ , then (1.8) is equal to 1 while (1.9) is equal to 0.

## 2. PROOF OF THE THEOREM AND COROLLARY

Since everything concerning the Theorem is radially symmetric, we will use the notation  $P_r$ ,  $r \geq 0$ , instead of  $P_x$ ,  $x \in R^2$ , where  $r = |x|$ . When we turn to the proof of the Corollary, we will return to the notation  $P_x$ . We will need the following Lemma.

**Lemma.** Let  $\tau_a = \inf\{t \geq 0 : |X(t)| = a\}$ . For  $l_1, l_2$  satisfying  $0 < l_1 < \frac{1}{2} < l_2$ , and for  $0 < a \leq b \leq \frac{1}{2}t^{l_2}$ , there exists a universal constant  $\lambda > 0$  such that

$$\begin{aligned} & \frac{\log b - \log a}{l_2 \log t - \log a} - \frac{1}{\lambda} \exp(-\lambda t^{2l_2-1}) \\ & \leq P_b(\tau_a > t) \leq \frac{\log b - \log a}{l_1 \log t - \log a} + \frac{1}{\lambda} \exp(-\lambda t^{1-2l_1}). \end{aligned}$$

*Proof.* Since  $P_b(\tau_a > \tau_c) = \frac{\log b - \log a}{\log c - \log a}$ , for  $a < b < c$  [2, p. 38], we have for  $t, l > 0$  such that  $b < t^l$ ,

$$P_b(\tau_a > t) \leq P_b(\tau_a > \tau_{t^l}) + P_b(\tau_{t^l} > t) = \frac{\log b - \log a}{l \log t - \log a} + P_b(\tau_{t^l} > t)$$

and

$$P_b(\tau_a > t) \geq P_b(\tau_a > \tau_{t^l}) - P_b(\tau_{t^l} < t) = \frac{\log b - \log a}{l \log t - \log a} - P_b(\tau_{t^l} < t).$$

The Brownian scaling property gives

$$P_b(\tau_{t^l} > t) = P_{bt^{-l}}(\tau_1 > t^{1-2l}) \text{ and } P_b(\tau_{t^l} < t) = P_{bt^{-l}}(\tau_1 < t^{1-2l}).$$

If  $l \in (0, \frac{1}{2})$ , then  $P_{bt^{-l}}(\tau_1 > t^{1-2l}) \leq \frac{1}{\lambda} \exp(-\lambda t^{1-2l})$  for some  $\lambda > 0$  [1, Theorem 3.6.1], while if  $l > \frac{1}{2}$  and  $b \leq \frac{1}{2}t^l$ , then  $P_{bt^{-l}}(\tau_1 < t^{1-2l}) \leq \frac{1}{\lambda} \exp(-\lambda t^{2l-1})$  for some  $\lambda > 0$  [1, Theorem 2.2.2]. The Lemma follows from these estimates.  $\square$

*Proof of the Theorem.* By the basic monotonicity property alluded to in the Remark following the Theorem, we can assume without loss of generality that  $\gamma(t) = c(1+t)^{-k}$  for some  $c > 0$  and  $k > 0$ . In light of (1.5), to prove the Theorem we must show that

$$(2.1) \quad \lim_{t \rightarrow \infty} P_b(|B(s)| \leq c(1+t-s)^{-k} \text{ for some } s \in [0, t]) = \frac{1}{1+2k}.$$

Using the right-hand inequality in the Lemma and the monotonicity of  $c(1+t)^{-k}$ , we have for large  $t$ ,

$$(2.2) \quad \begin{aligned} & P_b(|B(s)| \leq c(1+t-s)^{-k} \text{ for some } s \in [0, t]) \geq P_b(\tau_{c(1+t)^{-k}} \leq t) \\ & = 1 - P_b(\tau_{c(1+t)^{-k}} > t) \geq 1 - \frac{\log b - \log c + k \log(1+t)}{l_1 \log t - \log c + k \log(1+t)} - \frac{1}{\lambda} \exp(-\lambda t^{1-2l_1}). \end{aligned}$$

Letting  $t \rightarrow \infty$  in (2.2) and then letting  $l_1$  increase to  $\frac{1}{2}$ , we obtain

$$(2.3) \quad \liminf_{t \rightarrow \infty} P_b(|B(s)| \leq c(1+t-s)^{-k} \text{ for some } s \in [0, t]) \geq \frac{1}{1+2k}.$$

Applying the Markov property at time  $\frac{t}{2}$ , and using the monotonicity of  $c(1+t)^{-k}$  again, we have

$$(2.4) \quad P_b(|B(s)| > c(1+t-s)^{-k} \text{ for all } s \in [0, t]) \geq E_b(P_{|B(\frac{t}{2})|}(\tau_c > \frac{t}{2}); \tau_{c(1+\frac{t}{2})^{-k}} > \frac{t}{2}).$$

A direct calculation shows that there exists a constant  $C > 0$ , depending on  $b$ , such that for any  $\epsilon > 0$ ,

$$(2.5) \quad P_b(|B(\frac{t}{2})| \leq \epsilon t^{\frac{1}{2}}) \leq C\epsilon^2 \text{ for all } t > 0.$$

Thus, since  $P_b(\tau_c > \frac{t}{2})$  is increasing in  $b$  for  $b > c$ , we obtain from (2.4) and (2.5) that for any  $\epsilon > 0$  and sufficiently large  $t$

$$(2.6) \quad \begin{aligned} P_b(|B(s)| > c(1+t-s)^{-k} \text{ for all } s \in [0, t]) \\ \geq P_b(\tau_{c(1+\frac{t}{2})^{-k}} > \frac{t}{2})P_{\epsilon t \frac{1}{2}}(\tau_c > \frac{t}{2}) - C\epsilon^2. \end{aligned}$$

Using the lower bound in the Lemma to estimate the two probabilities on the right-hand side of (2.6), we have for sufficiently large  $t$ ,

$$(2.7) \quad \begin{aligned} P_b(|B(s)| > c(1+t-s)^{-k} \text{ for all } s \in [0, t]) \\ \geq \left( \frac{\log b - \log c + k \log(1 + \frac{t}{2})}{l_2 \log \frac{t}{2} - \log c + k \log(1 + \frac{t}{2})} - \frac{1}{\lambda} \exp(-\lambda t^{2l_2-1}) \right) \\ \times \left( \frac{\log \epsilon + \frac{1}{2} \log t - \log c}{l_2 \log \frac{t}{2} - \log c} - \frac{1}{\lambda} \exp(-\lambda t^{2l_2-1}) \right) - C\epsilon^2. \end{aligned}$$

Letting  $t \rightarrow \infty$  in (2.7), then letting  $\epsilon$  decrease to 0 and  $l_2$  decrease to  $\frac{1}{2}$  gives

$$(2.8) \quad \liminf_{t \rightarrow \infty} P_b(|B(s)| > c(1+t-s)^{-k} \text{ for all } s \in [0, t]) \geq \frac{2k}{1+2k}.$$

Now (2.1) follows from (2.3) and (2.8). □

*Proof of the Corollary.* We now return to the notation  $P_x$ ,  $x \in R^2$ . Let  $\tau_{L_a} = \inf\{t \geq 0 : B(t) \in L_a\}$ . We will show that there exists a constant  $K > 0$  such that for  $0 < a < |x| < c$ ,

$$(2.9) \quad \frac{\log |x| - \log a}{\log c - \log a} \leq P_x(\tau_{L_a} > \tau_c) \leq \frac{\log |x| - \log \frac{a}{2} + K}{\log c - \log \frac{a}{2}}.$$

Using (2.9), it follows immediately from the proof of the Lemma that

$$(2.10) \quad \begin{aligned} \frac{\log |x| - \log a}{l_2 \log t - \log a} - \frac{1}{\lambda} \exp(-\lambda t^{2l_2-1}) \\ \leq P_x(\tau_{L_a} > t) \leq \frac{\log |x| - \log \frac{a}{2} + K}{l_1 \log t - \log \frac{a}{2}} + \frac{1}{\lambda} \exp(-\lambda t^{1-2l_1}). \end{aligned}$$

One now proves the Corollary just as the Theorem was proved, using the estimate (2.10) in place of the estimate in the Lemma.

It remains to prove (2.9). The left-hand inequality in (2.9) of course follows trivially since  $P_x(\tau_{L_a} > \tau_c) \geq P_x(\tau_a > \tau_c) = \frac{\log |x| - \log a}{\log c - \log a}$ . For the right-hand inequality, we begin by using the strong Markov property to write

$$(2.11) \quad P_x(\tau_{L_a} < \tau_c) \geq E_x(P_{B(\tau_{\frac{a}{2}})}(\tau_{L_a} < \tau_c); \tau_{\frac{a}{2}} < \tau_c).$$

For any  $t > 0$  and  $y \in R^2$  with  $|y| = \frac{1}{2}$ , we have  $P_y(\tau_{L_1} < t \wedge \tau_1) > 0$ . This can be proved in any number of ways; for instance, by applying the Stroock-Varadhan support theorem [1, Theorem 2.6.1], or by applying the reflection principle [2] to the second coordinate of the Brownian motion and using the independence of the two components. Since Brownian motion is a Feller process, it follows that  $P_y(\tau_{L_1} < t \wedge \tau_1)$  is continuous in  $y$  [1, Theorem 1.3.1]; thus,

$$\inf_{y \in R^2: |y| = \frac{1}{2}} P_y(\tau_{L_1} < t \wedge \tau_1) > 0,$$

and a fortiori there exists a  $\rho > 0$  such that  $\inf_{y \in \mathbb{R}^2: |y|=\frac{1}{2}} P_y(\tau_{L_1} < \tau_1) \geq \rho$ . By Brownian scaling it then follows that

$$(2.12) \quad P_y(\tau_{L_a} < \tau_a) \geq \rho > 0 \text{ for } |y| = \frac{a}{2} \text{ and all } a > 0.$$

Also,

$$(2.13) \quad P_y(\tau_{\frac{a}{2}} < \tau_c) = \frac{\log a - \log c}{\log \frac{a}{2} - \log c} \equiv q_{a,c} \text{ for } |y| = a.$$

We use (2.12), (2.13) and the strong Markov property to estimate  $P_y(\tau_{L_a} < \tau_c)$  for  $|y| = \frac{a}{2}$ . Consider the event  $\tau_{L_a} > \tau_c$  under  $P_y$  with  $|y| = \frac{a}{2}$ . In order for this event to occur, first of all, starting from  $y \in \partial D_{\frac{a}{2}}$ ,  $B(t)$  must hit  $\partial D_a$  before hitting  $L_a$ , and this event occurs with probability no greater than  $1 - \rho$ . Then, starting from  $\partial D_a$ , the Brownian motion has a probability  $q_{a,c}$  of returning to  $\partial D_{\frac{a}{2}}$  before reaching  $\partial D_c$  (during which time it may hit  $L_a$ , but we ignore this), in which case it gets another chance, starting from  $\partial D_{\frac{a}{2}}$ , to hit  $L_a$  before hitting  $\partial D_a$ , etc. This reasoning gives the estimate

$$(2.14) \quad P_y(\tau_{L_a} > \tau_c) \leq \sum_{n=0}^{\infty} (1 - \rho)^{n+1} q_{a,c}^n (1 - q_{a,c}) = \frac{(1 - q_{a,c})(1 - \rho)}{1 - (1 - \rho)q_{a,c}} \equiv Q_{a,c} \text{ for } |y| = \frac{a}{2}.$$

From (2.11), (2.14), and the fact that  $P_x(\tau_{\frac{a}{2}} < \tau_c) = \frac{\log |x| - \log c}{\log \frac{a}{2} - \log c}$ , we have

$$(2.15) \quad \begin{aligned} P_x(\tau_{L_a} > \tau_c) &\leq 1 - E_x(P_{B(\tau_{\frac{a}{2}})}(\tau_{L_a} < \tau_c)); \tau_{\frac{a}{2}} < \tau_c \leq 1 - (1 - Q_{a,c}) \frac{\log |x| - \log c}{\log \frac{a}{2} - \log c} \\ &= \frac{\log |x| - \log \frac{a}{2} + Q_{a,c}(\log c - \log |x|)}{\log c - \log \frac{a}{2}}. \end{aligned}$$

From (2.13) and (2.14), we have

$$(2.16) \quad Q_{a,c} = \frac{(1 - \rho) \log 2}{\log 2 + \rho \log \frac{c}{a}}.$$

It follows from (2.16) that there exists a constant  $K > 0$  such that  $Q_{a,c} \log c \leq K$ . Substituting this estimate in the right-hand side of (2.15) proves (2.9).  $\square$

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DEPARTMENT OF MATHEMATICS, TECHNION-ISRAEL INSTITUTE OF TECHNOLOGY, HAIFA, 32000 ISRAEL

*E-mail address:* pinsky@techunix.technion.ac.il