

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

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ABSTRACT. Let $D_r = \{x \in R^2 : |x| \leq r\}$ and let γ be a continuous, nonincreasing 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:

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

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

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

and consider the heat equation in the exterior domain $R^d - D_1$:

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

1991 *Mathematics Subject Classification.* 35K05, 35B40, 60J65.

Key words and phrases. heat equation, planar Brownian motion, hitting times, modulus of Brownian motion, large time asymptotics.

This research was supported by the Fund for the Promotion of Research at the Technion and by the V.P.R. Fund.

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

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

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 in this note 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 γ 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 nonnegative function is 0. A completely rigorous argument can be made easily 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 $\Delta W + \frac{\partial W}{\partial s}$ is bounded on $G \times (0, t]$, then $W(B(s \wedge \sigma_{G^c}), s \wedge \sigma_{G^c}) - \int_0^{s \wedge \sigma_{G^c}} (\Delta W + \frac{\partial W}{\partial s'})(B(s'), s') ds'$ is a martingale under P_x for $s \in [0, t]$. In particular, if it so happens that $\Delta W + \frac{\partial W}{\partial s} \equiv 0$

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:

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

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

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

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 $W(x, s) \equiv u(x, t - s)$, $0 \leq s \leq t$, $x \in R^2 - D_{\gamma(t-s)}$. By (1.2), we have $(\Delta W + \frac{\partial W}{\partial s})(x, s) \equiv 0$, for $x \in R^2 - D_{\gamma(t-s)}$ 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

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

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 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

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

$$v^{(t)}(x, t) = 0, \quad x \in R^2 - D_{\gamma(t)},$$

$$v^{(t)}(x, s) = 1, \quad x \in D_{\gamma(s)}, \quad 0 \leq s \leq t.$$

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 σ_γ in place of σ_{G^c} .) An old result of Spitzer [3] indicates that for $|x| > \gamma(0)$,

$$(1.7) \quad P_x\left(\frac{|B(t)|}{\gamma(t)} \leq 1, \text{ for some } t \geq 0\right) = 1 \text{ if and only if } \int_0^\infty \frac{1}{t|\log \gamma(t)|} dt = \infty.$$

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

$$(1.8) \quad \begin{aligned} \lim_{t \rightarrow \infty} v^{(t)}(x, 0) &= \lim_{t \rightarrow \infty} P_x(B(s) \in D_{\gamma(s)} \text{ for some } s \in [0, t]) = 1 \\ \text{if and only if } &\int_0^\infty \frac{1}{t|\log \gamma(t)|} dt = \infty. \end{aligned}$$

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

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

Thus, there is a discrepancy between the asymptotic behavior of the solutions to the forward and the 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 $\int_0^\infty \frac{1}{t|\log \gamma(t)|} dt = \infty$, 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 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 , $x \in \mathbb{R}^2$, 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*

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

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 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 righthand 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

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

Using the lower bound in the Lemma to estimate the two probabilities on the righthand 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 ϵ 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). \square

Proof of 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

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

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 lefthand 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 righthand 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 $\inf_{y \in R^2: |y|=\frac{1}{2}} P_y(\tau_{L_1} < \tau_1) \geq \rho$. 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 ∂D_a before hitting L_a , and this event occurs with probability no greater than $1 - \rho$. Then, starting from ∂D_a , the Brownian motion has a probability $q_{a,c}$ of returning to $\partial D_{\frac{a}{2}}$ before reaching ∂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 ∂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 righthand side of (2.15) proves (2.9). \square

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