

# A SIMPLE PROOF OF THE DPRZ-THEOREM FOR 2d COVER TIMES.

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The  $\varepsilon$ -cover time of the two dimensional unit torus  $\mathbb{T}_2$  by Brownian motion (BM) is the time for the process to come within distance  $\varepsilon > 0$  from any point. Denoting by  $T_\varepsilon(x)$  the first time BM hits the  $\varepsilon$ -ball centered in  $x \in \mathbb{T}_2$ , the  $\varepsilon$ -cover time is thus given by

$$T_\varepsilon \equiv \sup_{x \in \mathbb{T}_2} T_\varepsilon(x). \quad (1)$$

The purpose of these short notes is to provide a concise proof of a celebrated theorem by Dembo, Peres, Rosen and Zeitouni, DPRZ for short, which settles the leading order in the small- $\varepsilon$  regime:

**Theorem 1.** (*The DPRZ-Theorem, [3]*) *Almost surely,*

$$\lim_{\varepsilon \downarrow 0} \frac{T_\varepsilon}{(\ln \varepsilon)^2} = \frac{2}{\pi}. \quad (2)$$

A key idea in the DPRZ-approach is to relate hitting times of  $\varepsilon$ -balls on  $\mathbb{T}_2$  to excursion counts between circles of mesoscopic sizes around these balls [6]; the DPRZ-proof of the theorem goes then through an involved multiscale analysis in the form of a second moment computation with truncation. We take here a similar point of view but with a number of twists which altogether lead to a considerable streamlining of the arguments. In particular, we implement the multiscale refinement of the second moment method emerged in the recent studies of Derrida's GREM and branching Brownian motion [5]. This tool brings to the fore the *true* process of covering [1] with the help of minimal infrastructure only; it also efficiently replaces the delicate tracking of points which DPRZ refer to as 'n-successful', and requires the use of finitely many scales only. All these features simplify substantially the proof of the DPRZ-theorem.

We believe the route taken here also considerably streamlines the deep DPRZ-results on late and thin/thick points of BM [2], and, what is perhaps more, it will be useful in the study of the *finer* properties. In fact, our approach carries over, *mutatis mutandis*, to these issues as well: when backed with [1], the present notes suggest that in order to address lower order corrections, one "simply" needs to increase the number of scales.

These notes are self-contained. Although, as mentioned, some key insights are taken from [3], no knowledge of the latter is assumed and detailed proofs to all statements are given.

## 1 The (new) road to the DPRZ-Theorem

We identify the unit torus  $\mathbb{T}_2$  with  $[0, 1) \times [0, 1) \subset \mathbb{R}^2$ , endowed with the metric

$$d_{\mathbb{T}_2}(x, y) = \min \{ \|x - y + (e_1, e_2)\| : e_1, e_2 \in \{-1, 0, 1\} \}.$$

We construct BM on  $\mathbb{T}_2$  by  $W_t \equiv (\hat{W}_1(t) \bmod 1, \hat{W}_2(t) \bmod 1)$ , where  $\hat{W}$  is standard BM on  $\mathbb{R}^2$ .

By monotonicity of  $T_\varepsilon$  and Borel-Cantelli Lemma, the DPRZ-Theorem steadily follows from

**Theorem 2.** For  $\delta > 0$  small enough there exist constants  $c(\delta), c'(\delta) > 0$  such that the following bounds hold for any  $0 < \varepsilon < c'(\delta)$ :

1) (upper bound)

$$\mathbb{P}\left(T_\varepsilon > (1 + \delta) \frac{2}{\pi} (\ln \varepsilon)^2\right) \leq \varepsilon^{c(\delta)}, \quad (3)$$

2) (lower bound)

$$\mathbb{P}\left(T_\varepsilon < (1 - \delta) \frac{2}{\pi} (\ln \varepsilon)^2\right) \leq \varepsilon^{c(\delta)}. \quad (4)$$

Theorem 2 will be proved by relating the natural timescale of the covering process to the excursion-counts of an embedded random walk, and a multiscale analysis of the latter which exploits some underlying, approximate hierarchical structure in the spirit of [1].

### 1.1 Scales, embedded random walks and excursion-counts

For  $R \in (0, \frac{1}{2})$  and  $K \geq 1$  we consider scales  $i = 0, 1, \dots, K$  and associate to each such scale a radius

$$r_i \equiv R \left(\frac{\varepsilon}{R}\right)^{i/K}. \quad (5)$$

BM started on  $\partial B_{r_i}$  hits  $\partial B_{r_{i+1}}$  before  $\partial B_{r_{i-1}}$  with probability  $1/2$ : by the strong Markovianity and rotational invariance, it follows that the process obtained by tracking the order in which BM visits the scales (with respect to one fixed center point and not counting multiple consecutive hits to the same scale) during one excursion from scale 1 to scale 0 is a simple random walk (SRW) started at 1 and stopped in 0. Keeping track of all BM-excursions up to some time thus yields a collection of independent SRW-excursions from 1 to 0. (The evolution of the SRW-excursions can

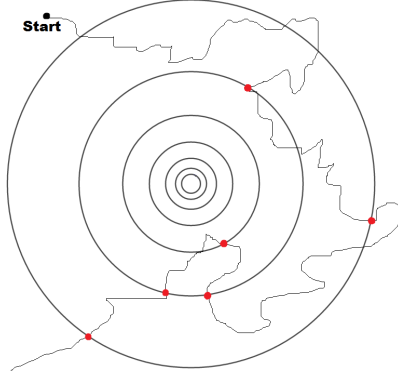


Figure 1: Reading off the SRW excursions  $1 \rightarrow 0$  and  $1 \rightarrow 2 \rightarrow 1 \rightarrow 0$

be unambiguously read off the BM-path, see Figure 1). For  $x \in \mathbb{T}_2$ , we set

$$D_n(x) \equiv \text{time at which } W \text{ completes the } n\text{-th excursion from } \partial B_{r_1}(x) \text{ to } B_{r_0}^c(x). \quad (6)$$

**Proposition 1.** (Concentration of excursion-counts) For  $\delta, R \in (0, \frac{1}{2})$  and  $x \in \mathbb{T}_2$ , it holds

$$\mathbb{P}\left(D_n(x) \geq (1 + \delta) n \frac{1}{\pi} \ln \frac{r_0}{r_1}\right) \leq \exp\left(-n \left(\frac{\delta^2}{8} + o_{r_1}(1)\right)\right) \quad (7)$$

$$\mathbb{P}\left(D_n(x) \leq (1 - \delta) n \frac{1}{\pi} \ln \frac{r_0}{r_1}\right) \leq \exp\left(-n \left(\frac{\delta^2}{4} + o_{r_1}(1)\right)\right) \quad (8)$$

for all  $n \in \mathbb{N}$  as  $r_1 \rightarrow 0$ .

Proposition 1 will bear fruits when combined with the following

**Proposition 2.** (First moment of hitting times) *There exists an universal constant  $C > 0$ , such that*

$$\left| \mathbb{E}_y[\tau_{B_r(x)}] - \frac{1}{\pi} \ln \frac{d_{\mathbb{T}_2}(x, y)}{r} \right| \leq C \quad (9)$$

for all  $x \in \mathbb{T}_2$ ,  $r > 0$  and  $y \in \mathbb{T}_2 \setminus B_r(x)$ . Also

$$\mathbb{E}_y[\tau_{B_r^c(x)}] = \frac{r^2 - d_{\mathbb{T}_2}(x, y)^2}{2} \quad (10)$$

for all  $x \in \mathbb{T}_2$ ,  $r \in (0, \frac{1}{2})$  and  $y \in B_r(x)$ .

Propositions 1 and 2 make precise the intuition that  $D_n(x) \approx n \mathbb{E}_{B_{r_0}}[\tau_{B_{r_1}}]$ , allowing in particular to switch from the natural timescale to the excursion-counts. Armed with the above results, which will be proved in Section 2.1, we discuss the main steps behind Theorem 2. The upper bound is easy: we address that first.

Here and below,  $L_\varepsilon$  will denote the square lattice of mesh size  $\lceil \varepsilon^{-1} \rceil^{-1}$ . Remark that  $|L_\varepsilon| \approx \varepsilon^{-2}$ .

## 1.2 The upper bound

We will show that, with overwhelming probability, at time

$$t_\varepsilon(\delta) \equiv (1 + \delta) \frac{2}{\pi} (\ln \varepsilon)^2, \quad (11)$$

each  $\varepsilon$ -ball with center on  $L_\varepsilon$  has been hit by BM and extend this to the entire torus thereafter.

**Lemma 1.** *For  $\delta > 0$  small enough there exist constants  $c, c' > 0$  depending on  $\delta$  only such that*

$$\mathbb{P}(\exists x \in L_\varepsilon \text{ such that } T_\varepsilon(x) > t_\varepsilon(\delta)) \leq \varepsilon^c \quad (12)$$

holds for all  $0 < \varepsilon < c'$ .

*Proof.* We set

$$n_\varepsilon(\delta) = -(1 + \delta/2) 2K \ln(\varepsilon), \quad (13)$$

which is slightly larger than the typical amount of excursions at time  $t_\varepsilon(\delta)$ . For an  $\varepsilon$ -ball to be avoided up to some time: either *i*) BM needs to complete less than  $n_\varepsilon(\delta)$  excursions from scale 1 to scale 0 in that time or *ii*) scale  $K$ , corresponding to the  $\varepsilon$ -ball, has to be avoided for at least  $n_\varepsilon(\delta)$  many excursions. Therefore setting

$$\mathcal{I}(x) \equiv \text{number of the first excursion from } \partial B_{r_1}(x) \text{ to } B_{r_0}^c(x) \text{ that hits } B_{r_K}(x).$$

we have

$$\mathbb{P}(\exists x \in L_\varepsilon \text{ s.t. } T_\varepsilon(x) > t_\varepsilon(\delta)) \leq \mathbb{P}(\exists x \in L_\varepsilon \text{ s.t. } \mathcal{I}(x) > n_\varepsilon(\delta) \text{ or } D_{n_\varepsilon(\delta)}(x) \geq t_\varepsilon(\delta)). \quad (14)$$

By Markov inequality and union bound

$$(14) \leq \sum_{x \in L_\varepsilon} \mathbb{P}(\mathcal{I}(x) > n_\varepsilon(\delta)) + \mathbb{P}(D_{n_\varepsilon(\delta)}(x) \geq t_\varepsilon(\delta)). \quad (15)$$

The probability that  $n_\varepsilon(\delta)$  independent excursions of a SRW starting in 1 all hit 0 before  $K$  is given by  $(1 - 1/K)^{n_\varepsilon(\delta)}$ , while the second probability on the r.h.s of (15) is estimated by Proposition 1. This shows that the above is at most

$$|L_\varepsilon| \left[ \left(1 - \frac{1}{K}\right)^{n_\varepsilon(\delta)} + \exp\left(-\frac{\delta^2}{72} n_\varepsilon(\delta)\right) \right] \leq \varepsilon^\delta (1 + o_\varepsilon(1)), \quad (16)$$

for  $K$  large enough, the last inequality since  $1 - \frac{1}{K} \leq e^{-1/K}$ , and  $|L_\varepsilon| \approx \varepsilon^{-2}$ .  $\square$

Coming back to the upper bound in Theorem 2,

$$\mathbb{P}(T_\varepsilon > t_\varepsilon(\delta)) = \mathbb{P}(\exists x \in \mathbb{T}_2 : T_\varepsilon(x) > t_\varepsilon(\delta)) \leq \mathbb{P}(\exists x \in L_{\varepsilon/10} : T_{\varepsilon/10}(x) > t_\varepsilon(\delta)), \quad (17)$$

the last step using that any  $\varepsilon$ -ball contains a ball of radius  $\varepsilon/10$  with center in  $L_{\varepsilon/10}$ . For  $\varepsilon > 0$  small enough depending on  $\delta$  we have  $t_\varepsilon(\delta) \geq t_{\varepsilon/10}(\delta/2)$ , therefore it holds that

$$(17) \leq \mathbb{P}(\exists x \in L_{\varepsilon/10} : T_{\varepsilon/10}(x) > t_{\varepsilon/10}(\delta/2)). \quad (18)$$

Applying Lemma 1 with  $\varepsilon/10$  for  $\varepsilon$  and  $\delta/2$  yields the upper bound in Theorem 2.

### 1.3 The lower bound

We show that with overwhelming probability there exists  $x \in \mathbb{T}_2$  with avoided  $\varepsilon$ -ball at time

$$t = t(\varepsilon, \delta) \equiv (1 - \delta)^4 \frac{2}{\pi} (\ln \varepsilon)^2. \quad (19)$$

Theorem 2 will then follow immediately by considering<sup>1</sup>  $\hat{\delta} \equiv 1 - (1 - \delta)^4$ . Set

$$n(j) = n(j; \varepsilon, \delta, K) \equiv -2K(1 - \delta)^j \ln \varepsilon, \quad (j \in \mathbb{N}). \quad (20)$$

With  $\tau_r \equiv \tau_r(x)$  denoting the first time BM hits the  $r$ -ball around  $x \in \mathbb{T}_2$ , we define the events

$$\mathcal{R} \equiv \bigcap_{x \in L_\varepsilon} \{D_{n(3)}(x) > t\} \quad \text{and} \quad (21)$$

$$\mathcal{R}^x \equiv \{\tau_{r_1} < \tau_{r_K}\} \cap \{\text{At most } n(2) \text{ excursions } [\delta k] \rightarrow [\delta k] - 1 \text{ during first } n(3) \text{ excursions } 1 \rightarrow 0\}. \quad (22)$$

For  $n \in \mathbb{N}$  and  $l \in \{1, \dots, K-1\}$ , let

$$\mathcal{N}_l^x(n) \equiv \text{number of excursions of } W \text{ from } \partial B_{r_l}(x) \text{ to } \partial B_{r_{l+1}}(x) \text{ within the first } n \text{ excursions from } \partial B_{r_l}(x) \text{ to } \partial B_{r_{l-1}}(x) \text{ after time } \tau_{r_1}. \quad (23)$$

For  $x \in \mathbb{T}_2$  define the events

$$A^x \equiv \bigcap_{l=\lceil \delta K \rceil}^{K-1} A_l^x, \quad \text{where} \quad A_l^x \equiv \left\{ \mathcal{N}_l^x \left( n \left( 1 - \frac{l}{K} \right)^2 \right) \leq n \left( 1 - \frac{l+1}{K} \right)^2 \right\}. \quad (24)$$

The events  $A, \mathcal{R}$  are motivated by the following observations. First, it can be checked via Doob's h-transform that the expected number of excursions from  $l$  to  $l+1$  performed by a SRW started at 1 and stopped at 0 and conditioned not to hit  $K$ , is approximately  $[1 - (l+1)/K]^2$ . The events  $A^x$  thus describe the natural avoidance strategy of scale  $K$  by  $n$  independent such SRW, which is in turn equivalent to specifying the avoidance strategy of an  $\varepsilon$ -ball. Second, we claim that

$$\mathcal{R} \cap \mathcal{R}^x \cap A^x \subset \{B_\varepsilon(x) \text{ is not hit up to time } t\}. \quad (25)$$

Remark in fact that on  $\mathcal{R}^x$ , the ball  $B_\varepsilon(x)$  is not hit before  $\partial B_{r_1}(x)$ , hence the  $\varepsilon$ -ball can only be hit in an excursion from  $B_{r_1}$  to  $B_{r_0}$ .  $\mathcal{R}$  ensures that there are at most  $n(3)$ -excursions before time  $t$ . Therefore, on  $\mathcal{R}^x \cap \mathcal{R}$ , there are at most  $n(2)$  excursions from scale  $[\delta K] \rightarrow [\delta K] - 1$  at time  $t$ . But on  $A^x$ , none of these excursions reaches scale  $K$ , hence the  $\varepsilon$ -ball is not hit, and (25) holds.

In light of (25), and in view of the lower bound in Theorem 2, estimates on the probabilities of the  $\mathcal{R}, A$ -events are needed. This information is provided by Lemma 2 and 3 below, whose proofs are deferred to Section 2.2. Concerning the  $\mathcal{R}$ -event we state

<sup>1</sup>This is notationally convenient, but holds no deeper meaning.

**Lemma 2.** For all  $\delta > 0$  and large enough  $K = K(\delta) \in \mathbb{N}$  there exist constants  $\kappa, \kappa' > 0$  depending on  $\delta, K$  only such that

$$\inf_{x \in L_\varepsilon \setminus B_{r_1}(W_0), \varepsilon \in (0, \kappa')} \mathbb{P}(\mathcal{R}^x), \mathbb{P}(\mathcal{R}) \geq 1 - \varepsilon^\kappa. \quad (26)$$

Concerning the  $A$ -events,

**Lemma 3.** (One-point estimates) For  $K$  large,  $\varepsilon > 0$  small enough (depending on  $\delta$ )

$$\varepsilon^{2-1.99\delta} \leq \mathbb{P}(A^x) \leq \varepsilon^{2-2.01\delta}, \quad (27)$$

Coming back to the lower bound, restricting to the set  $L_\varepsilon^* \equiv L_\varepsilon \setminus B_{r_1}(W_0)$  yields that

$$\begin{aligned} \mathbb{P}\left(\sup_{x \in \mathbb{T}_2} T_\varepsilon(x) > t\right) &\geq \mathbb{P}(\exists x \in L_\varepsilon^* \text{ such that } B_\varepsilon(x) \text{ is not hit up to time } t) \\ &\stackrel{(25)}{\geq} \mathbb{P}(\mathcal{R} \text{ and } \exists x \in L_\varepsilon^* \text{ such that } \mathcal{R}^x \cap A^x) \\ &\geq \frac{\mathbb{E}[\#\{x \in L_\varepsilon^* : \mathcal{R}^x \cap A^x\}]^2}{\mathbb{E}[\#\{x \in L_\varepsilon^* : \mathcal{R}^x \cap A^x\}^2]} - \mathbb{P}(\mathcal{R}^c), \end{aligned} \quad (28)$$

by Paley-Zygmund inequality. Rotational invariance and strong Markovianity imply that  $\mathcal{R}^x$  and  $A^x$  are independent, hence the above is *at least*

$$\left[ \sum_{x \in L_\varepsilon^*} \mathbb{P}(\mathcal{R}^x) \mathbb{P}(A^x) \right]^2 / \left[ \sum_{x, y \in L_\varepsilon^*} \mathbb{P}(A^x \cap A^y) \right] - \mathbb{P}(\mathcal{R}^c). \quad (29)$$

We now analyse the denominator. First, remark that for  $d_{\mathbb{T}_2}(x, y) > 2r_{\lceil \delta K \rceil - 1}$ , the  $A$ -events decouple: in fact, they are rotationally invariant and depend on disjoint excursions, hence the strong Markov property yields  $\mathbb{P}(A^x \cap A^y) = \mathbb{P}(A^x) \mathbb{P}(A^y)$ . Shortening

$$\mathcal{A} \equiv \sum_{x \in L_\varepsilon^*} \mathbb{P}(A^x), \quad \mathcal{B} \equiv \sum_{x, y \in L_\varepsilon} \mathbb{1}_{\{d_{\mathbb{T}_2}(x, y) \leq 2r_{\lceil \delta K \rceil - 1}\}} \mathbb{P}(A^x \cap A^y),$$

by Lemma 2 and the exact decoupling we thus have that

$$\begin{aligned} (29) &\geq (1 - \varepsilon^\kappa)^2 \frac{\mathcal{A}^2}{\mathcal{A}^2 + \mathcal{B}} - \varepsilon^\kappa \geq (1 - \varepsilon^\kappa)^2 \left(1 - \frac{\mathcal{B}}{\mathcal{A}^2}\right) - \varepsilon^\kappa \\ &\geq (1 - \varepsilon^\kappa)^2 \left(1 - \frac{\mathcal{B}}{\varepsilon^{-3.96\delta}}\right) - \varepsilon^\kappa, \end{aligned} \quad (30)$$

the last step by Lemma 3 and using that  $|L_\varepsilon| \geq \varepsilon^{-2+0.01\delta}$ . It thus remains to analyze the  $\mathcal{B}$ -term: by regrouping terms according to the distance,

$$\mathcal{B} \leq \sum_{i=\lceil \delta K \rceil - 2}^K \sum_{x, y \in L_\varepsilon} \mathbb{1}_{\{d_{\mathbb{T}_2}(x, y) \in [r_{i+1}, r_i]\}} \mathbb{P}(A^x \cap A^y). \quad (31)$$

To get a handle on the two-points probabilities appearing in (31), we follow the recipe from [5, Sec. 3.1.1 p. 97-98], exploiting the approximate hierarchical structure which underlies the excursion-counts, and which is best explained with the help of a picture, see Figure 2 below. First, the circles associated to  $x, y$  on small scales  $i$  (left) are almost identical and so are the excursion counts; this suggests that  $A_i^x \cap A_i^y$  is well represented by  $A_i^x$  alone. Dropping one of the events is an estimate by worst case scenario known in this context as "REM approximation". For larger  $i$  (middle) this approximation is not sharp, but only a single scale can fall into this case as we can choose  $\varepsilon$  arbitrarily small for given  $K$ . Choosing  $K$  large makes the influence of a single

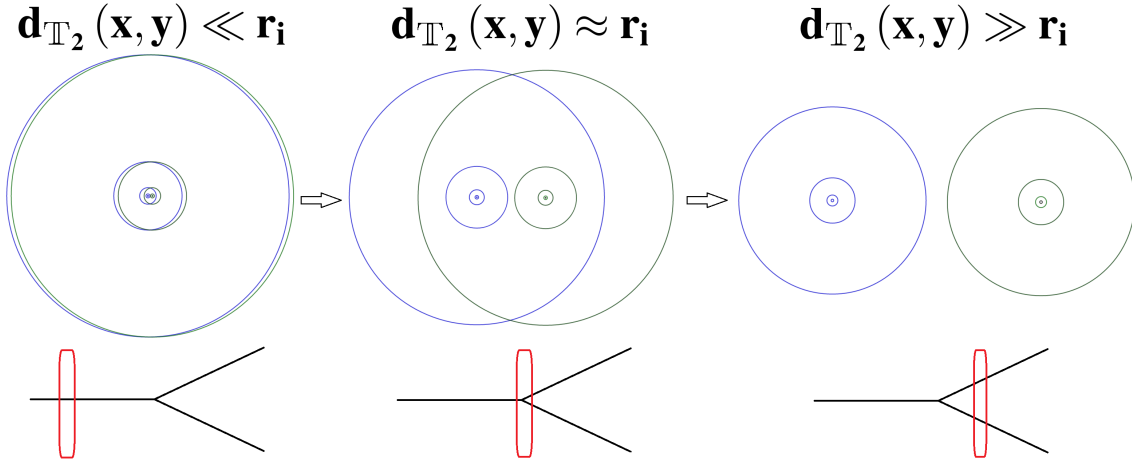


Figure 2: Common branch on small scales (left) and decoupling on large scales (right).

scale comparatively small. For  $i$  large (right), balls are disjoint, which by rotational invariance and strong Markovianity yields independent excursion counts. Such approximate tree-structure of excursion counts is summarized in the lower picture, with the red box corresponding to the scale at hand. By these considerations, for  $i \geq \lceil \delta K \rceil - 2$  and  $d_{\mathbb{T}_2}(x, y) \in [r_{i+1}, r_i]$ , we write

$$\begin{aligned}
\mathbb{P}(A^x \cap A^y) &= \mathbb{P}\left(\bigcap_{l=\lceil \delta K \rceil}^{K-1} A_l^x \cap \bigcap_{l=\lceil \delta K \rceil}^{K-1} A_l^y\right) \\
&\leq \mathbb{P}\left(\bigcap_{l=\lceil \delta K \rceil, l \neq i, i+1}^{K-1} A_l^x \cap \bigcap_{l=i+1}^{K-1} A_l^y\right) \quad (\text{"REM approximation"}) \quad (32) \\
&= \prod_{l=\lceil \delta K \rceil, l \neq i, i+1}^{K-1} \mathbb{P}(A_l^x) \prod_{l=i+1}^{K-1} \mathbb{P}(A_l^y) \quad (\text{exact decoupling}) \\
&\leq \varepsilon^{4-2.01\delta-2\frac{i+1}{K}} \quad (\text{Lemma 3 / one-point estimates}).
\end{aligned}$$

There are at most  $2\varepsilon^{-4}\pi r_i^2$  pairs of points on  $L_\varepsilon$  with distance at most  $r_i$ : using that  $r_i \leq \varepsilon^{i/K}$ , and (32) in (31) we get

$$\mathcal{B} \leq \sum_{i=\lceil \delta K \rceil - 2}^K 2\pi \varepsilon^{-2.01\delta - \frac{4}{K}} \leq \varepsilon^{-2.02\delta}. \quad (33)$$

Applying this estimate to (30) and putting  $\hat{\delta} \equiv 1 - (1 - \delta)^4$  we therefore see that

$$\mathbb{P}\left(\sup_{x \in \mathbb{T}_2} T_\varepsilon(x) > (1 - \hat{\delta}) \frac{2}{\pi} (\ln \varepsilon)^2\right) \geq 1 - \varepsilon^{\hat{c}}, \quad (34)$$

for  $\hat{c} \equiv \frac{1}{2} \min\{\kappa, 1.94\delta\}$ , settling the lower bound of Theorem 2.

## 2 Proofs

### 2.1 Hitting times and excursion-counts

The study of hitting times for BM is closely related to Green's functions. Estimates on the torus have however proofs which are either opaque or hard to find: we include here an elementary treatment based on Fourier analysis for the reader's convenience.

**Lemma 4.** *The function*

$$F(x, y) \equiv G_x(y) - \frac{1}{2\pi} \ln d_{\mathbb{T}_2}(x, y), \quad \text{where} \quad G_x(y) \equiv - \sum_{p \in 2\pi\mathbb{Z}^2 \setminus \{0\}} \frac{1}{|p|^2} e^{ip(x-y)} \quad (35)$$

is bounded on  $\mathbb{T}_2^2 \setminus \{(x, x) : x \in \mathbb{T}_2\}$ .

*Proof.* It suffices to consider  $y$  in a small neighborhood of  $x$ , as otherwise the result is trivial. So let  $z \equiv x - y$  and assume that  $2|z_1| \geq |z|$  (swapping coordinates otherwise). We have

$$\left| \sum_{\substack{p \in 2\pi\mathbb{Z}^2 \setminus \{0\} \\ |p| > |z|^{-1}}} \frac{1}{|p|^2} e^{ipz} \right| = \left| \sum_{\substack{p \in 2\pi\mathbb{Z}^2 \setminus \{0\} \\ |p| > |z|^{-1}}} \frac{1}{1 - e^{i2\pi z_1}} \frac{1}{|p|^2} \left( e^{ipz} - e^{i(p+(2\pi, 0))z} \right) \right|. \quad (36)$$

Shifting the difference from the exponential to  $|p|^{-2}$  by collecting terms with the same exponent, and by the triangle inequality, one obtains boundedness uniformly over  $z \neq 0$  in a small enough neighborhood of 0. The extra terms due to the boundary of the summation domain are easily shown to be bounded. By combining the summand  $p$  and  $-p$  we see that sums of this form are real valued. Therefore

$$\sum_{\substack{p \in 2\pi\mathbb{Z}^2 \setminus \{0\} \\ |p| \leq |z|^{-1}}} \frac{1}{|p|^2} e^{ipz} = \sum_{\substack{p \in 2\pi\mathbb{Z}^2 \setminus \{0\} \\ |p| \leq |z|^{-1}}} \frac{1}{|p|^2} \cos(pz). \quad (37)$$

Since  $|pz| \leq 1$  for all summands contained in this sum we can estimate  $\cos(x) \leq 1 - x^2/4$ . Hence

$$\left| G_x(y) - \sum_{\substack{p \in 2\pi\mathbb{Z}^2 \setminus \{0\} \\ |p| \leq |z|^{-1}}} \frac{1}{|p|^2} \right| \quad (38)$$

is uniformly bounded for  $y$  in a small neighborhood of  $x$ . The claim of Lemma 4 then follows by rearranging summands into groups  $C_j \equiv \{p \in 2\pi\mathbb{Z}^2 \setminus \{0\} : |p|^2 \in ((j-1)^3, j^3]\}$ , estimating  $|p|^{-2}$  by best/worst case scenario within each group, and using that  $|C_j| = \frac{3}{4\pi} j^2 + O(j^{3/2})$ .  $\square$

*Proof of Proposition 2: first moment of hitting times.* Let  $\mu(y) \equiv \mathbb{E}_y[\tau_{B_r(x)}]$ . For  $\Delta$  the Laplacian with periodic boundary condition on  $\mathbb{T}_2$  we have Poisson's equation  $\Delta\mu = -2$  on  $\mathbb{T}_2 \setminus B_r(x)$  with  $\mu = 0$  on  $B_r(x)$ . Plainly,

$$G_x(y) \equiv - \sum_{p \in 2\pi\mathbb{Z}^2 \setminus \{0\}} \frac{1}{|p|^2} e^{ip(x-y)} \quad (39)$$

is a Green function, i.e. solution of  $\Delta G_x = 1 - \delta_x$  on the torus. In particular,  $\mu + 2G_x$  is harmonic on  $\mathbb{T}_2 \setminus B_r(x)$ . By the maximum principle, and since  $\mu \equiv 0$  on  $\partial B_r(x)$ ,

$$2 \inf_{z \in \partial B_r(x)} G_x(z) \leq \mu(y) + 2G_x(y) \leq 2 \sup_{z \in \partial B_r(x)} G_x(z) \quad (40)$$

holds. It follows from Lemma 4 that  $\mu(y) - \frac{1}{\pi} \ln[d_{\mathbb{T}_2}(x, y)/r]$  is bounded, and the first claim (9) is proved. The second claim (10) is elementary as we can identify the ball on  $\mathbb{T}_2$  with the ball in  $\mathbb{R}^2$  and exploit rotational invariance to solve Poisson's equation explicitly.  $\square$

*Proof of Proposition 1: concentration of excursion-counts.* By Kac's moment formula [4],

$$\mathbb{E}_x[\tau_A^i] \leq i! \sup_{x \in \mathbb{T}} \mathbb{E}_x[\tau_A]^i, \quad A \subset \mathbb{T} \text{ closed.} \quad (41)$$

By monotone convergence, Taylor-expanding the exponential function, and by the above estimate,

$$\mathbb{E}_x \left[ e^{\theta \tau_A} \right] \leq 1 + \theta \mathbb{E}_x [\tau_A] + \sum_{i=2}^{\infty} \left( \theta \sup_{x \in \mathbb{T}} \mathbb{E}_x [\tau_A] \right)^i \leq \exp \left( \theta \mathbb{E}_x [\tau_A] + 2\theta^2 \sup_{x \in \mathbb{T}} \mathbb{E}_x [\tau_A]^2 \right) \quad (42)$$

for  $0 < \theta < \frac{1}{2} \left( \sup_{x \in \mathbb{T}} \mathbb{E}_x [\tau_A] \right)^{-1}$ . Using  $e^{-x} \leq 1 - x + x^2$  for positive  $x$  gives

$$\mathbb{E}_x \left[ e^{-\theta \tau_A} \right] \leq 1 - \theta \mathbb{E}_x [\tau_A] + \theta^2 \sup_{x \in \mathbb{T}} \mathbb{E}_x [\tau_A]^2 \leq \exp \left( -\theta \mathbb{E}_x [\tau_A] + \theta^2 \sup_{x \in \mathbb{T}} \mathbb{E}_x [\tau_A]^2 \right). \quad (43)$$

Consider  $\tau^{(i \leftarrow)}$  the time it takes  $W$  to get from  $\partial B_{r_1}(x)$  to  $B_{r_0}^c(x)$  the  $i$ -th time;  $\tau^{(i \rightarrow)}$  the time  $W$  needs to get from  $\partial B_{r_0}(x)$  to  $B_{r_1}(x)$  the  $i$ -th time after  $B_{r_1}(x)$  has been hit the first time and  $\tau_{r_1}$  the time it takes  $W$  to get from the starting point to  $\partial B_{r_1}(x)$ . Now by definition we have

$$D_n(x) = \tau_{r_1} + \sum_{i=1}^{n-1} \tau^{(i \rightarrow)} + \sum_{i=1}^n \tau^{(i \leftarrow)}. \quad (44)$$

Exponential Markov inequality gives for any  $t, \theta > 0$

$$\mathbb{P}(D_n(x) \geq t) \leq e^{-\theta t} \mathbb{E} \left[ e^{\theta D_n(x)} \right] \quad (45)$$

Using (44), by strong Markovianity and estimating by worst starting points this is

$$\leq e^{-\theta t} \left( \sup_{z \in \mathbb{T}_2} \mathbb{E}_z \left[ e^{\theta \tau_{r_1}} \right] \right) \left( \sup_{z \in B_{r_0}(x)} \mathbb{E}_z \left[ e^{\theta \tau^{(1 \rightarrow)}} \right] \right)^{n-1} \left( \sup_{z \in B_{r_1}(x)} \mathbb{E}_z \left[ e^{\theta \tau^{(1 \leftarrow)}} \right] \right)^n \quad (46)$$

Using (42) with  $\theta = -\frac{\pi \delta}{4 \ln r_1}$ , and applying Proposition 2, we obtain

$$\begin{aligned} \sup_{z \in \mathbb{T}_2} \mathbb{E}_z \left[ e^{\theta \tau_{r_1}} \right] &\leq e^{\frac{\delta}{4} + \frac{\delta^2}{8} + o_{r_1}(1)} \\ \sup_{z \in B_{r_0}(x)} \mathbb{E}_z \left[ e^{\theta \tau^{(1 \rightarrow)}} \right]^{n-1} &\leq e^{(n-1) \left( \frac{\delta}{4} + \frac{\delta^2}{8} + o_{r_1}(1) \right)} \\ \text{and } \sup_{z \in B_{r_1}(x)} \mathbb{E}_z \left[ e^{\theta \tau^{(1 \leftarrow)}} \right]^n &\leq e^{n o_{r_1}(1)}. \end{aligned} \quad (47)$$

With  $t = (1 + \delta) n \frac{1}{\pi} \ln \frac{r_0}{r_1}$ , and by the above estimates, (46) reads

$$\mathbb{P} \left( D_n(x) \geq (1 + \delta) n \frac{1}{\pi} \ln \frac{r_0}{r_1} \right) \leq e^{-n \left( \frac{\delta}{4} + \frac{\delta^2}{8} + o_{r_1}(1) \right)} e^{n \left( \frac{\delta}{4} + \frac{\delta^2}{8} + o_{r_1}(1) \right)}, \quad (48)$$

settling (7). As for (8): for any  $n \in \mathbb{N}$  and  $\theta > 0$  we have

$$\mathbb{P}(D_n(x) \leq t) \leq e^{\theta t} \mathbb{E} e^{-\theta D_n(x)} \leq e^{\theta t} \mathbb{E} \left[ e^{-\theta \tau^{(1 \rightarrow)}} \right]^{n-1}. \quad (49)$$

Choosing  $\theta = \frac{\pi \delta}{2 \ln r_1}$  and  $t = (1 - \delta) n \frac{1}{\pi} \ln \frac{r_0}{r_1}$ , applying (43) together with Proposition 2 yields the second claim and concludes the proof of Proposition 1.  $\square$



## 2.2 Estimates for $\mathcal{R}$ and $A$

*Proof of Lemma 2.* For  $x \in L_\varepsilon^*$ ,  $\{\tau_{r_1} < \tau_{r_K}\}$  almost surely. By rotational invariance and strong Markovianity, the number of excursions from scale  $\lceil \delta K \rceil$  to scale  $\lceil \delta K \rceil - 1$  in different excursions from scale 1 to scale 0 are independent of each other. The number of excursions from scale  $\lceil \delta K \rceil$  to scale  $\lceil \delta K \rceil - 1$  in one excursion from scale 1 to scale 0 is distributed like the product of a Bernoulli distributed and an independent geometrically distributed random variable, both with parameter  $\lceil \delta K \rceil^{-1}$ . (This product has expectation 1). By Cramér's theorem,

$$\begin{aligned} & \mathbb{P}(\text{more than } n(2) \text{ times } \lceil \delta K \rceil \rightarrow \lceil \delta K \rceil - 1 \text{ in the first } n(3) \text{ excursions } 1 \rightarrow 0) \\ & \leq \exp\left(-n(3)I\left(\frac{1}{1-\delta}\right)\right) = \varepsilon^{2K(1-\delta)^3 I\left(\frac{1}{1-\delta}\right)}, \end{aligned} \quad (50)$$

with  $I$  the rate function of a Bernoulli( $1/\lceil \delta K \rceil$ )  $\times$  geometric( $1/\lceil \delta K \rceil$ ). It follows that  $\mathbb{P}((\mathcal{R}^x)^c)$  vanishes polynomially in  $\varepsilon$  for fixed  $\delta$  and  $K$ . Taking the complement yields the first claim.

By Proposition 1 we have

$$\mathbb{P}(D_{n(3)}(x) \leq t) \leq \varepsilon^{2K(1-\delta)^3(\delta^2/4 + o_{r_1}(1))}, \quad (51)$$

which vanishes faster then, say,  $\varepsilon^3$  for  $K$  sufficiently large. The second claim thus follows by union bound over all  $x \in L_\varepsilon$  on the complements.  $\square$

*Proof of Lemma 3.* The number of times a SRW goes from  $l$  to  $l+1$  before going from  $l$  to  $l-1$  is geo( $1/2$ )-distributed. Therefore  $\mathcal{N}_l^x(n)$  is, by strong Markovianity and rotational invariance, the sum of  $n$  independent geo( $1/2$ )-distributed r.v.'s. Hence by Cramér's theorem

$$\begin{aligned} \mathbb{P}(A^x) &= \prod_{l=\lceil \delta K \rceil}^{K-1} \mathbb{P}(A_l^x) = \prod_{l=\lceil \delta K \rceil}^{K-1} \exp\left(-n\left(1 - \frac{l}{K}\right)^2 I\left(\frac{(1 - \frac{l+1}{K})^2}{(1 - \frac{l}{K})^2}\right) + o_\varepsilon(n)\right) \\ &= \exp\left(-\frac{n}{K^2} \sum_{l=\lceil \delta K \rceil}^{K-1} (K-l)^2 I\left(\left(1 - \frac{1}{K-l}\right)^2\right) + o_\varepsilon(n)\right), \end{aligned} \quad (52)$$

where  $I(x) = x \ln(x) - (1+x) \ln\left(\frac{1+x}{2}\right)$  is the geo( $1/2$ )-rate function. Using  $I(1) = I'(1) = 0$  and  $I''(1) = \frac{1}{2}$  one quickly obtains  $j^2 I\left(\left(1 - 1/j\right)^2\right) = 1 + o_j(1)$  as  $j \rightarrow \infty$ , and therefore

$$\mathbb{P}(A^x) = \exp\left(-\frac{n}{K}(1-\delta)(1 + o_K(1)) + o_\varepsilon(n)\right) = \varepsilon^{2(1-\delta)(1+o_K(1))+o_\varepsilon(1)}, \quad (53)$$

concluding the proof of the Lemma.  $\square$

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