

Boundary of the Range I: Typical Behavior

Amine Asselah, Bruno Schapira

▶ To cite this version:

Amine Asselah, Bruno Schapira. Boundary of the Range I: Typical Behavior. 2015. hal-01171542v2

HAL Id: hal-01171542 https://hal.science/hal-01171542v2

Preprint submitted on 29 Oct 2015 (v2), last revised 5 Jul 2016 (v3)

HAL is a multi-disciplinary open access archive for the deposit and dissemination of scientific research documents, whether they are published or not. The documents may come from teaching and research institutions in France or abroad, or from public or private research centers. L'archive ouverte pluridisciplinaire **HAL**, est destinée au dépôt et à la diffusion de documents scientifiques de niveau recherche, publiés ou non, émanant des établissements d'enseignement et de recherche français ou étrangers, des laboratoires publics ou privés.

Boundary of the Range I: Typical Behavior.

Amine Asselah* & Bruno Schapira[†]

Abstract

We study the boundary of the range for the simple random walk on \mathbb{Z}^d in the transient regime $d \geq 3$. We show that sizes of the range and its boundary differ mainly by a martingale. As a consequence, we obtain an upper bound on the variance of order $n \log n$ in dimension three. We also establish a central limit theorem in dimension four and larger.

1 Introduction

Let $(S_n, n \ge 0)$ be a simple random walk on \mathbb{Z}^d . Its range $\mathcal{R}_n = \{S_0, \dots, S_n\}$ is a familiar object of Probability Theory since Dvoretzky and Erdös' influential paper [DE]. The object of interest in this paper is the boundary of the range

$$\partial \mathcal{R}_n = \{ x \in \mathcal{R}_n : \text{ there exists } y \sim x \text{ with } y \notin \mathcal{R}_n \},$$
 (1.1)

where $x \sim y$ means that x and y are at (graph) distance one. Our interest was triggered by a recent paper of Berestycki and Yadin [BY] which proposes a model of hydrophobic polymer in an aqueous solvent, consisting of tilting the law of a simple random walk by $\exp(-\beta|\partial\mathcal{R}_n|)$. One interprets the range as the space occupied by the polymer, and its complement as the space occupied by the solvent. Hydrophobic means that the monomers dislike the solvent, and the polymer tries to minimize the boundary of the range. The Gibbs' weight tends to minimize contacts between the monomers and the solvent, and the steric effect has been forgotten to make the model mathematically tractable. Besides its physical appeal, the model gives a central role to the boundary of the range, an object which remained mainly in the shadow until recently. To our knowledge it first appeared in the study of the entropy of the range of a simple random walk [BKYY], with the conclusion that in dimension two or larger, the entropy of the range scales like the size of the boundary of the range. Recently, Okada [Ok1] has established a law of large numbers for the boundary of the range for a transient random walk, and has obtained bounds on its expectation in dimension two.

Theorem 1.1. [Okada] Consider a simple random walk in dimension d = 2. Then

$$\frac{\pi^2}{2} \le \lim_{n \to \infty} \frac{\mathbb{E}[|\partial \mathcal{R}_n|]}{n/\log^2(n)} \le 2\pi^2,\tag{1.2}$$

where part of the result is that the limit exists. Moreover, when $d \geq 3$, almost surely

$$\lim_{n \to \infty} \frac{|\partial \mathcal{R}_n|}{n} = \mathbb{P}(\{z: \ z \sim 0\} \not\subset \mathcal{R}_{\infty} \cup \widetilde{\mathcal{R}}_{\infty}, \ H_0 = \infty), \tag{1.3}$$

where \mathcal{R}_{∞} is the range of a random walk in an infinite time horizon, and H_0 is the hitting time of 0, whereas quantities with tilde correspond to an independent copy.

^{*}LAMA, UPEC & IMéRA E-mail: amine.asselah@u-pec.fr

[†]Aix-Marseille Université, CNRS, Centrale Marseille, I2M, UMR 7373, 13453 Marseille, France. E-mail: bruno.schapira@univ-amu.fr

The thrust of our study is to show that for a transient random walk, range and boundary of the range are highly correlated objects: when one of them shrinks, so does the other. We present two ways to appreciate their similar nature. First, the sizes of a range-like set and the boundary of the range differ roughly by a martingale. To make this statement precise, we need more notation. Let $V_0 = \{z : z \sim 0\}$, be the neighbors of the origin, and for any nonempty subset V of V_0 , let $\mathcal{R}_{n,V}$ be the set of sites of \mathbb{Z}^d whose first visit occurs at some time $k \leq n$, and such that $(S_k + V_0) \cap \mathcal{R}_{k-1}^c = S_k + V$. We are now ready for our first observation.

Proposition 1.2. There is a martingale $(M_n, n \in \mathbb{N})$, adapted to the standard filtration such that for any n integer

$$|\partial \mathcal{R}_n| = \sum_{V \subset V_0} \rho_V |\mathcal{R}_{n-1,V}| + M_n + \mathcal{E}_n, \tag{1.4}$$

with $\rho_{\emptyset} = 0$ and for any non-empty V in V_0

$$\rho_{V} = \mathbb{P}(V \not\subset \mathcal{R}_{\infty}) \quad and \quad \mathbb{E}(\mathcal{E}_{n}^{2}) = \begin{cases} \mathcal{O}(n) & \text{if } d = 3\\ \mathcal{O}(\log^{3}(n)) & \text{if } d = 4\\ \mathcal{O}(1) & \text{if } d \geq 5. \end{cases}$$
(1.5)

Jain and Pruitt [JP] have established a Central Limit Theorem for the range in dimension three with a variance scaling like $n \log n$. Proposition 1.2 makes us expect that the boundary of the range has a similar behavior. Indeed, we establish the following estimate on the mean square of the martingale. This estimate is delicate, uses precise Green's function asymptotics, and the symmetry of the walk. It is our main technical contribution.

Proposition 1.3. There are positive constants C_d such that

$$Var(M_n) \le \begin{cases} C_3 n \log n & \text{if } d = 3\\ C_d n & \text{if } d \ge 4. \end{cases}$$

Also, following the approach of Jain and Pruitt [JP], we establish the following estimate on the range-like object $\mathcal{R}_{n,V}$.

Proposition 1.4. Assume that d = 3, and let V be a nonempty subset of V_0 . There is a positive constant C, such that

$$Var(|\mathcal{R}_{n,V}|) \le C \, n \log n. \tag{1.6}$$

Then, a useful corollary of Propositions 1.2, 1.3 and 1.4 is the corresponding bound for the variance of the boundary of the range in dimension 3.

Theorem 1.5. Assume that d = 3. Then, there is a positive constant C, such that

$$Var(|\partial \mathcal{R}_n|) \le C n \log n.$$
 (1.7)

This is complemented with a linear bound on the variance, which holds in dimension three and larger.

Proposition 1.6. Assume that $d \geq 3$. There are positive constants $\{c_d, d \geq 3\}$, such that

$$Var(|\partial \mathcal{R}_n|) \ge c_d n.$$
 (1.8)

The idea behind the linear lower bound (1.8) is to show that there is a *clock process* whose fluctuations are normal (on a scale square root of the time elapsed), and which is independent

of the boundary of the range process. Thus, typical fluctuations of the clock process, provoke a time change at constant boundary of the range.

Secondly, the boundary of the range has a decomposition similar to the classical Le Gall's decomposition [LG] in terms of intersection of independent ranges. This decomposition, though simple, requires more notation to be presented. For integers n, m let $\mathcal{R}(n, n+m) = \{S_k - S_n\}_{n \leq k \leq n+m}$, with the shorthand notation $\mathcal{R}_n = \mathcal{R}(0, n)$, and note that

$$\mathcal{R}(0, n+m) = \mathcal{R}(0, n) \cup (S_n + \mathcal{R}(n, n+m)).$$

Observe that $\overleftarrow{\mathcal{R}}(0,n) := -S_n + \mathcal{R}(0,n)$ and $\mathcal{R}(n,n+m)$ are independent and that by the symmetry of the walk $\overleftarrow{\mathcal{R}}(0,n)$ (resp. $\mathcal{R}(n,n+m)$) has the same law as $\mathcal{R}(0,n)$ (resp. $\mathcal{R}(0,m)$): it corresponds to the range of a walk seen backward from position S_n . Finally, note the well known decomposition

$$|\mathcal{R}(0,n+m)| = |\mathcal{R}(0,n)| + |\mathcal{R}(n,n+m)| - |\overleftarrow{\mathcal{R}}(0,n) \cap \mathcal{R}(n,n+m)|. \tag{1.9}$$

Equality (1.9) is the basis of Le Gall celebrated paper [LG] on the range of recurrent random walk. It is also a key ingredient in most work on self-intersection of random walks (see the book of Chen [C], for many references).

To write a relation as useful as (1.9) for the boundary of the range, we introduce more notation. For $\Lambda \subset \mathbb{Z}^d$, we denote $\Lambda^+ = \Lambda + \overline{V}_0$, and we define its boundary as

$$\partial \Lambda = \{ z \in \Lambda : \exists y \in \Lambda^c \text{ with } y \sim z \}.$$

Now, our simple observation is as follows.

Proposition 1.7. For any integers n, m

$$0 \ge |\partial \mathcal{R}(0, n+m)| - (|\partial \mathcal{R}(0, n)| + |\partial \mathcal{R}(n, n+m)|) \ge -Z(0, n, m), \tag{1.10}$$

with

$$Z(0,n,m) = |\overleftarrow{\mathcal{R}}(0,n)) \cap \mathcal{R}^+(n,n+m)| + |\overleftarrow{\mathcal{R}}^+(0,n) \cap \mathcal{R}(n,n+m)|. \tag{1.11}$$

We focus now on consequences of this simple decomposition. For $d \geq 3$, we define functions $n \mapsto \psi_d(n)$, with the following dimension depending growth

$$\psi_3(n) = \sqrt{n}, \qquad \psi_4(n) = \log n, \qquad \text{and for } d > 4, \qquad \psi_d(n) = 1.$$
 (1.12)

We have the following bounds for expectation and variance of the boundary of the range.

Proposition 1.8. When dimension is larger than or equal to three, there are constants C_d such that for any integer n

$$\frac{C_d \psi_d(n)}{n} \ge \frac{E[|\partial \mathcal{R}_n|]}{n} - \lim_{k \to \infty} \frac{E[|\partial \mathcal{R}_k|]}{k} \ge 0. \tag{1.13}$$

Assume now that the dimension is four or larger. Then, the limit of $Var(|\partial \mathcal{R}_n|)/n$ exists, is positive, and

$$\left| \frac{Var(|\partial \mathcal{R}_n|)}{n} - \lim_{k \to \infty} \frac{Var(|\partial \mathcal{R}_k|)}{k} \right| \le \frac{C_d \sqrt{n} \psi_d(n)}{n}, \tag{1.14}$$

Moreover, a standard Central Limit Theorem holds for $|\partial \mathcal{R}_n|$.

Remark 1.9. We have stated our results for the simple random walk, but they hold, with similar proofs, for walks with symmetric and finitely supported increments.

Okada obtains also in [Ok1] a large deviation principle for the upper tail (the probability that the boundary be larger than its mean), and in [Ok2] he studies the most frequently visited sites of the boundary, and proves results analogous to what is known for the range.

In a companion paper [AS], we obtain large deviations for the lower tail, and provide applications to phase transition for a properly normalized Berestycki-Yadin's polymer model.

The rest of the paper is organized as follows. In Section 2, we fix notation, recall known results on the Green's function, and prove a result about covering a finite subset. In Section 3, we establish the Martingale decomposition of Proposition 1.2 and prove Proposition 1.3. We prove Proposition 1.6 in Section 4. In Section 5, we present the dyadic decomposition for the boundary of the range and deduce Proposition 1.8, using Le Gall's argument. Finally in the Appendix, we prove Proposition 1.4.

2 Notation and Prerequisites

For any $y, z \in \mathbb{Z}^d$, we denote by ||z - y|| the Euclidean norm between y and z, and by $\langle y, z \rangle$ the corresponding scalar product. Then for any r > 0 we denote by B(z, r) the ball of radius r centered at z:

$$B(z,r) := \{ y \in \mathbb{Z}^d : ||z - y|| \le r \}.$$

For $x \in \mathbb{Z}^d$, we let \mathbb{P}_x be the law of the random walk starting from x. For Λ a subset of \mathbb{Z}^d we define the hitting time of Λ as

$$H_{\Lambda} := \inf\{n \ge 1 : S_n \in \Lambda\}.$$

Note that in this definition we use the convention to consider only times larger than or equal to one. At some point it will also be convenient to consider a shifted version, so we also define for $k \geq 2$,

$$H_{\Lambda}^{(k)} := \inf\{n \ge k : S_n \in \Lambda\}. \tag{2.1}$$

We will need bounds on the heat kernel, so let us recall a standard result:

$$\mathbb{P}(S_n = z) \le C \frac{1}{n^{d/2}} \exp(-c||z||^2/n) \quad \text{for all } z \text{ and } n \ge 1,$$
(2.2)

for some positive constants c and C (see for instance [HSC]). Now we recall also the definition and some basic properties of Green's function. For $u, v \in \mathbb{Z}^d$, Green's function is

$$G(u,v) = \mathbb{E}_u \Big[\sum_{n>0} \mathbb{I} \{ S_n = v \} \Big],$$

and we use extensively the well-known bound (see [LL, Theorem 4.3.1]):

$$G(0,z) = \mathcal{O}\left(\frac{1}{1 + ||z||^{d-2}}\right). \tag{2.3}$$

We also consider Green's function restricted to a set $A \subset \mathbb{Z}^d$, which for $u, v \notin A$ is defined by

$$G_A(u, v) = \mathbb{E}_u \Big[\sum_{n=0}^{H_{A^c} - 1} \mathbb{I} \{ S_n = v \} \Big].$$

We recall that G_A is symmetric (see [LL, Lemma 4.6.1]):

$$G_A(u,v) = G_A(v,u)$$
 for all $u,v \in \mathbb{Z}^d$,

and that G is also invariant by translation of the coordinates: G(u, v) = G(0, v - u). Also, for $n \ge 0$,

$$G_n(u,v) = \mathbb{E}_u \Big[\sum_{k=0}^n \mathbb{I} \{ S_n = v \} \Big].$$

It is well known (use (2.3) and Theorem 3.6 of [LG]) that for ψ_d defined in (1.12), we have, for some positive constants $\{C_d, d \geq 3\}$

$$\sum_{z \in \mathbb{Z}^d} G_n^2(0, z) \le C_d \psi_d(n). \tag{2.4}$$

We can now state the main result of this Section.

Lemma 2.1. Let Λ be a finite subset of \mathbb{Z}^d , and fix $z \in \Lambda$. Then, there is a constant $c(\Lambda)$, such that for any two neighboring sites $y \sim y'$,

$$\mathbb{P}_{y}(\Lambda \subset \mathcal{R}_{\infty}) - \mathbb{P}_{y'}(\Lambda \subset \mathcal{R}_{\infty}) = c(\Lambda) \frac{\langle y' - y, y - z \rangle}{\|y - z\|^{d}} + \mathcal{O}\left(\frac{1}{\|y - z\|^{d}}\right). \tag{2.5}$$

Moreover,

$$c(\Lambda) = \frac{1}{dv_d} \sum_{\substack{x \in \Lambda, x' \notin \Lambda \\ x' \land x}} \mathbb{P}_{x'} (H_{\Lambda} = \infty) \mathbb{P}_x (\Lambda \subset \mathcal{R}_{\infty}),$$

where v_d denote the volume of the unit ball in \mathbb{R}^d .

Proof. First, since Λ is finite, we can always assume that y and y' do not belong to Λ . Now by a first entry decomposition

$$\mathbb{P}_{y}(\Lambda \subset \mathcal{R}_{\infty}) = \sum_{x \in \Lambda} \mathbb{P}_{y}(S_{H_{\Lambda}} = x, H_{\Lambda} < \infty) \mathbb{P}_{x}(\Lambda \subset \mathcal{R}_{\infty}). \tag{2.6}$$

Next, fix $x \in \Lambda$ and transform the harmonic measure into the restricted Green's function (see for instance [LL, Lemma 6.3.6]):

$$\mathbb{P}_y\big(S_{H_\Lambda} = x, \ H_\Lambda < \infty\big) = \frac{1}{2d} \sum_{v \in \Lambda^c, \, v \sim x} G_{\Lambda^c}(y, v) = \frac{1}{2d} \sum_{v \in \Lambda^c, \, v \sim x} G_{\Lambda^c}(v, y).$$

Note also (see [LL, Proposition 4.6.2]) that

$$G_{\Lambda^c}(v,y) = G(v,y) - \mathbb{E}_v \Big[\mathbb{1}\{H_{\Lambda} < \infty\} G(S_{H_{\Lambda}},y) \Big].$$

Therefore,

$$\mathbb{P}_{y}(S_{H_{\Lambda}} = x) - \mathbb{P}_{y'}(S_{H_{\Lambda}} = x) = \frac{1}{2d} \sum_{v \in \Lambda^{c}, v \sim x} (G(v, y) - G(v, y')) \\
- \frac{1}{2d} \sum_{v \in \Lambda^{c}, v \sim x} \mathbb{E}_{v} \left[\mathbb{I}\{H_{\Lambda} < \infty\} \left(G(S_{H_{\Lambda}}, y) - G(S_{H_{\Lambda}}, y') \right) \right].$$
(2.7)

Now, we have ([LL, Corollary 4.3.3]) the expansion for any $z \in \Lambda$

$$G(z,y) - G(z,y') = \frac{2}{v_d} \frac{\langle y' - y, y - z \rangle}{\|y - z\|^d} + \mathcal{O}\left(\frac{1}{\|y - z\|^d}\right). \tag{2.8}$$

Combining (2.6), (2.7) and (2.8) we obtain the result (2.5).

3 Martingale Decomposition

In this Section, we establish Proposition 1.2, as well as Proposition 1.3 dealing with the variance of the martingale.

3.1 Definition of the martingale and proof of Proposition 1.2

For V nonempty subset of V_0 and $k \geq 0$, let

$$I_{k,V} = \mathbb{I}\{S_k \notin \mathcal{R}_{k-1} \text{ and } (S_k + V_0) \cap \mathcal{R}_k^c = S_k + V\},$$

and

$$J_{k,V} = \mathbb{I}\{(S_k + V) \nsubseteq \{S_j, j \ge k\}\}.$$

Then for $n \geq 1$, define

$$\mathcal{R}_{n,V} = \{ S_k : I_{k,V} = 1, \ k \le n \},$$

and

$$M_{n,V} = \sum_{k=0}^{n-1} I_{k,V} \left(\mathbb{E}[J_{k,V} \mid \mathcal{F}_n] - \mathbb{E}[J_{k,V} \mid \mathcal{F}_k] \right),$$

with $(\mathcal{F}_k, k \geq 0)$ the standard filtration of the walk. Note that by construction $(M_{n,V}, n \geq 0)$ is a martingale. Moreover, the Markov property gives

$$\mathbb{E}[J_{k,V} \mid \mathcal{F}_k] = \mathbb{E}[J_{k,V}] = \mathbb{P}(V \nsubseteq \mathcal{R}_{\infty}) = \rho_V,$$

for all $k \geq 0$, where \mathcal{R}_{∞} is the range of the whole path of the walk. In particular the second sum in the definition of $M_{n,V}$ is precisely equal to $\rho_V |\mathcal{R}_{n-1,V}|$. Now for any $k \leq n$, write

$$J_{k,n,V} = \mathbb{I}\{(S_k + V) \not\subseteq \{S_k, \dots, S_n\}\}.$$

Then define

$$\partial \mathcal{R}_{n,V} = \{ S_k : I_{k,V} J_{k,n,V} = 1, \ k \le n \},$$
 (3.1)

and

$$\mathcal{E}_{n,V} := \sum_{k=0}^{n-1} I_{k,V} \{ J_{k,n,V} - \mathbb{E}[J_{k,V} \mid \mathcal{F}_n] \}.$$

With these definitions we get

$$|\partial \mathcal{R}_{n,V}| = M_{n,V} + \rho_V |\mathcal{R}_{n-1,V}| + \mathcal{E}_{n,V}$$
 for all $V \subset V_0$.

Summing up $M_{n,V}$ over nonempty subsets of V_0 we obtain another martingale

$$M_n = \sum_{V \subset V_0} M_{n,V},$$

and an error term

$$\mathcal{E}_n = \sum_{V \subset V_0} \mathcal{E}_{n,V}.$$

Since $\partial \mathcal{R}_n$ is the disjoint union of $\partial \mathcal{R}_{n,V}$'s over nonempty $V \subset V_0$, we obtain the first part (1.4) of Proposition 1.2.

Now we prove (1.5). First note that for any $k \leq n-1$,

$$|J_{k,n,V} - \mathbb{E}[J_{k,V} \mid \mathcal{F}_n]| \le \mathbb{P}_{S_n}(H_{S_k + V_0} < \infty) = \mathcal{O}\left(\frac{1}{1 + \|S_n - S_k\|^{d-2}}\right).$$

Then by using the invariance of the walk by time inversion, we get

$$\mathbb{E}[\mathcal{E}_{n,V}^2] = \mathcal{O}\left(\sum_{k,k' \le n} \mathbb{E}\left[\frac{1}{(1 + \|S_k\|^{d-2})(1 + \|S_{k'}\|^{d-2})}\right]\right). \tag{3.2}$$

Moreover, by using the heat kernel bound (2.2), we arrive at

$$\mathbb{E}\left[\frac{1}{(1+\|S_k\|^{d-2})^2}\right] = \begin{cases} \mathcal{O}(1/k) & \text{if } d=3\\ \mathcal{O}((\log k)/k^2) & \text{if } d=4\\ \mathcal{O}(k^{-d/2}) & \text{if } d \ge 5. \end{cases}$$
(3.3)

The desired result follows by using Cauchy-Schwarz.

3.2 Variance of the Martingale

We establish here Proposition 1.3. Let us notice here that our proof works for M_n only, and not for all the $M_{n,V}$'s. If we set for $n \geq 0$,

$$\Delta M_n = M_{n+1} - M_n,$$

then, Proposition 1.3 is a direct consequence of the following result.

$$\mathbb{E}[(\Delta M_n)^2] = \begin{cases} \mathcal{O}(\log n) & \text{if } d = 3\\ \mathcal{O}(1) & \text{if } d \ge 4. \end{cases}$$
 (3.4)

The proof of (3.4) is divided in three steps. The first step brings us to a decomposition of ΔM_n as a finite combination of simpler terms (3.6), plus a rest whose L^2 -norm we show is negligible. In the second step, we observe that when we gather together some terms (3.10), their L^2 -norm takes a particularly nice form (3.11). Finally in the third step we use these formula and work on it to get the right bound.

Step 1. In this step, we just use the Markov property to write ΔM_n in a nicer way, up to some error term, which is bounded by a deterministic constant. Before that, we introduce some more notation. For $k \leq n$, set

$$I_{k,n,V} = \mathbb{I}\{(S_k + V_0) \cap \{S_k, \dots, S_n\}^c = S_k + V\}.$$

The Markov property and the translation invariance of the walk show that for all $k \leq n$

$$\mathbb{E}[J_{k,V} \mid \mathcal{F}_n] = \sum_{V \cap V' \neq \emptyset} I_{k,n,V'} \, \mathbb{P}_{S_n - S_k} \left(V \cap V' \not\subseteq \mathcal{R}_{\infty} \right).$$

Note that $I_{k,n,V'} \neq I_{k,n+1,V'}$ means that S_{n+1} and S_k are neighbors. However, the number of k indices such that S_{n+1} and S_k are neighbors and $I_{k,V} = 1$ is at most 2d. Therefore the number of indices k satisfying $I_{k,V} \neq 0$ and $I_{k,n,V'} \neq I_{k,n+1,V'}$, for some V', is bounded by 2d. As a consequence, by using also that terms in the sum defining $M_{n,V}$ are bounded in absolute value by 1, we get

$$\Delta M_n = \sum_{k=0}^{n-1} \sum_{V \cap V' \neq \emptyset} I_{k,V} I_{k,n,V'} \Big\{ \mathbb{P}_{S_{n+1} - S_k} \left(V \cap V' \not\subseteq \mathcal{R}_{\infty} \right) - \mathbb{P}_{S_n - S_k} \left(V \cap V' \not\subseteq \mathcal{R}_{\infty} \right) \Big\} + r_n, \quad (3.5)$$

with $|r_n| \leq 2d + 1$. Lemma 2.1 is designed to deal with the right hand side of (3.5), with the result that

$$\Delta M_n = \sum_{k=0}^{n-1} \sum_{V \cap V' \neq \emptyset} c(V \cap V') I_{k,V} I_{k,n,V'} \frac{\langle S_{n+1} - S_n, S_n - S_k \rangle}{1 + \|S_n - S_k\|^d} + \mathcal{O}(B_n),$$
(3.6)

with

$$B_n = \sum_{z \in \partial \mathcal{R}_n} \frac{1}{1 + \|S_n - z\|^d}.$$
 (3.7)

Step 2. The term B_n of (3.7) can be bounded as follows. By using first the invariance of the law of the walk by time inversion, we can replace the term $S_n - z$ by z. Then we write

$$\mathbb{E}[B_n^2] = \mathbb{E}\Big[\left(\sum_{z \in \partial \mathcal{R}_n} \frac{1}{1 + \|z\|^d}\right)^2\Big] = \sum_{z, z' \in \mathbb{Z}^d} \frac{1}{(1 + \|z\|^d)(1 + \|z'\|^d)} \mathbb{P}(z \in \partial \mathcal{R}_n, z' \in \partial \mathcal{R}_n).$$
(3.8)

Then by assuming for instance that $||z|| \le ||z'||$ (and $z \ne z'$), and by using (2.3) we obtain

$$\mathbb{P}(z \in \partial \mathcal{R}_n, z' \in \partial \mathcal{R}_n) \leq \mathbb{P}(H_z < \infty, H_{z'} < \infty)
\leq 2G(0, z)G(z, z') = \mathcal{O}\left(\frac{1}{1 + ||z||^{d-2}||z' - z||^{d-2}}\right).$$
(3.9)

Therefore

$$\mathbb{E}[B_n^2] = \mathcal{O}\left(\sum_{1 \le ||z|| < ||z'||} \frac{1}{||z||^{2d-2} ||z'||^d ||z' - z||^{d-2}}\right).$$

Next we can for instance divide the last sum into two parts and we immediately get:

$$\sum_{1 \le \|z\| < \|z'\|} \frac{1}{\|z\|^{2d-2} \|z'\|^d \|z' - z\|^{d-2}} \\
= \sum_{1 \le \|z\| < \|z'\| \le 2\|z\|} \frac{1}{\|z\|^{2d-2} \|z'\|^d \|z' - z\|^{d-2}} + \sum_{1 \le 2\|z\| < \|z'\|} \frac{1}{\|z\|^{2d-2} \|z'\|^d \|z' - z\|^{d-2}} \\
= \mathcal{O}\left(\sum_{1 \le \|z\| < \|z'\| \le 2\|z\|} \frac{1}{\|z\|^{3d-2} \|z' - z\|^{d-2}} + \sum_{1 \le 2\|z\| < \|z'\|} \frac{1}{\|z\|^{2d-2} \|z'\|^{2d-2}}\right) = \mathcal{O}(1).$$

Now, it remains to bound the main term in (3.6). For two nonempty subsets U and U' of V_0 , write $U \sim U'$, if there exists an isometry of \mathbb{Z}^d sending U onto U'. This of course defines an equivalence relation on the subsets of V_0 , and for any representative U of an equivalence class, we define

$$\widetilde{I}_{k,n,U} = \sum_{V \cap V' \sim U} I_{k,V} I_{k,n,V'},$$

and

$$H_{n,U} = \sum_{k=0}^{n-1} \widetilde{I}_{k,n,U} \frac{\langle S_{n+1} - S_n, S_n - S_k \rangle}{1 + \|S_n - S_k\|^d}.$$

Note that since the function $c(\cdot)$ is invariant under isometry, we can rewrite the main term in (3.6) as

$$\sum_{U} c(U) H_{n,U}. \tag{3.10}$$

Then observe that for any U,

$$\mathbb{E}[H_{n,U}^2 \mid \mathcal{F}_n] = \left\| \sum_{k=0}^{n-1} \widetilde{I}_{k,n,U} \frac{S_n - S_k}{1 + \|S_n - S_k\|^d} \right\|^2.$$

Moreover, since the law of the walk is invariant under time inversion, and since for any path S_0, \ldots, S_n , and any k, the indicator $\widetilde{I}_{k,n,U}$ is equal to 1 if and only if it is also equal to 1 for the reversed path S_n, \ldots, S_0 , we get

$$\mathbb{E}[H_{n,U}^2] = \mathbb{E}[\|\mathcal{H}_{n,U}\|^2],\tag{3.11}$$

where

$$\mathcal{H}_{n,U} := \sum_{z \in \partial \widetilde{\mathcal{R}}_{n,U}} \frac{z}{1 + \|z\|^d}, \quad \text{with} \quad \partial \widetilde{\mathcal{R}}_{n,U} := \left\{ S_k : \widetilde{I}_{k,n,U} = 1, \ k \le n - 1 \right\}.$$
 (3.12)

Note that $\partial \widetilde{\mathcal{R}}_{n,U}$ reads also

$$\partial \widetilde{\mathcal{R}}_{n,U} = \{ S_k : k \le n - 1, (S_k + V_0) \cap \mathcal{R}_n^c \sim U \}.$$

Therefore, we only need to prove that for any U,

$$\mathbb{E}[\|\mathcal{H}_{n,U}\|^2] = \begin{cases} \mathcal{O}(\log n) & \text{if } d = 3\\ \mathcal{O}(1) & \text{if } d \ge 4. \end{cases}$$
(3.13)

Step 3. First note that

$$\mathbb{E}[\|\mathcal{H}_{n,U}\|^2] = \sum_{z,z' \in \mathbb{Z}^d} \frac{\langle z, z' \rangle}{(1 + \|z\|^d)(1 + \|z'\|^d)} \mathbb{P}\left(z \in \partial \widetilde{\mathcal{R}}_{n,U}, z' \in \partial \widetilde{\mathcal{R}}_{n,U}\right). \tag{3.14}$$

In dimension 4 or larger, (3.13) can be established as follows. First Cauchy-Schwarz inequality gives

$$|\langle z, z' \rangle| \leq ||z|| \, ||z'||,$$

for all z, z'. Then by using again the standard bound on Green's functions as in (3.9), we get

$$\mathbb{E}[\|\mathcal{H}_{n,U}\|^2] = \mathcal{O}\left(\sum_{1 \le \|z\| < \|z'\|} \|z\|^{3-2d} \|z'\|^{1-d} \|z - z'\|^{2-d}\right) = \mathcal{O}(1),$$

as desired. We consider now the more difficult case d=3. Actually, since it might be interesting to see what changes in dimension 3, we keep the notation d in all formula as long as possible. Note that if $z \in \partial \widetilde{\mathcal{R}}_{n,U}$, then in particular $||z|| \leq n$ and H_z is finite. Therefore the restriction of the sum in (3.14) to the set of z, z' satisfying $||z|| \leq ||z'|| \leq 2||z||$ is bounded (in absolute value) by

$$\sum_{\|z\| \le \|z'\| \le 2\|z\| \le 2n} \frac{2\|z\|^2}{(1+\|z\|^d)^2} \mathbb{P}\left(H_z < \infty, H_{z'} < \infty\right). \tag{3.15}$$

Moreover, as we have already recalled, for any $z \neq z'$, with $||z|| \leq ||z'||$,

$$\mathbb{P}(H_z < \infty, H_{z'} < \infty) \le 2G(0, z)G(z, z') = \mathcal{O}\left(\frac{1}{\|z\|^{d-2}\|z - z'\|^{d-2}}\right).$$

Therefore, the sum in (3.15) is bounded above (up to some constant) by

$$\sum_{1 \le ||z|| < ||z'|| \le 2||z|| \le 2n} ||z||^{4-3d} ||z - z'||^{2-d} = \mathcal{O}\left(\sum_{1 \le ||z|| \le n} ||z||^{6-3d}\right) = \mathcal{O}(\log n).$$

Thus it only remains to bound the sum in (3.14) restricted to the z and z' satisfying $||z'|| \ge 2||z||$. To this end observe that if z' is visited first,

$$\sum_{1 \le 2\|z\| \le \|z'\| \le n} \frac{|\langle z, z' \rangle|}{(1 + \|z\|^d)(1 + \|z'\|^d)} \mathbb{P}(H_{z'+V_0} < H_z < \infty)$$

$$= \mathcal{O}\left(\sum_{1 \le 2\|z\| \le \|z'\| \le n} \|z\|^{1-d} \|z'\|^{3-2d} \|z - z'\|^{2-d}\right)$$

$$= \mathcal{O}\left(\sum_{1 \le 2\|z\| \le \|z'\| \le n} \|z\|^{1-d} \|z'\|^{5-3d}\right)$$

$$= \mathcal{O}(\log n),$$

where for the first equality we used in particular Cauchy-Schwarz inequality and again the standard bound on Green's function, and for the second one, we used that when $||z'|| \geq 2||z||$, we have $||z'|| \approx ||z - z'||$. Thus in (3.14) we can also add in all the probabilities the events $H_z < H_{z'+V_0}$. Now for any z, call

$$\tau_z := \inf\{n \ge H_z : S_n \notin B(z, ||z||/2)\},\$$

and

$$\sigma_z := \inf\{n \ge \tau_z : S_n \in z + V_0\}.$$

Then define the event

$$E_{z,n,U} := \left\{ z \in \partial \widetilde{\mathcal{R}}_{n,U} \right\} \cap \left\{ \sigma_z = \infty \right\}.$$

Observe next that if $1 \le ||z|| \le ||z'||/2$,

$$\mathbb{P}\left(z \in \partial \widetilde{\mathcal{R}}_{n,U}, \, z' \in \partial \widetilde{\mathcal{R}}_{n,U}, \, (E_{z,n,U} \cap E_{z',n,U})^c\right) = \mathcal{O}\left(\frac{1}{\|z\|^{2d-4}\|z'\|^{d-2}}\right).$$

Therefore, similar computations as above, show that in (3.14), we can replace the events $z, z' \in \partial \widetilde{\mathcal{R}}_{n,U}$ respectively by $E_{z,n,U}$ and $E_{z',n,U}$. Finally a last simplification of the problem is to remove the indices n. More precisely define

$$E_{z,U} := \{ H_z < \infty \} \cap \{ V_0 \cap \{ S_0 - z, \dots, S_{\tau_z} - z \}^c \sim U \} \cap \{ \sigma_z = \infty \}.$$

The easiest part is to replace $E_{z,n,U}$ by $E_{z,U}$, since it suffices to observe that

$$E_{z,n,U} \cap E_{z,U}^c \cap E_{z',n,U} \cap \{H_z < H_{z'+V_0}\} = \emptyset.$$

The analogue for z' is more delicate, but one can argue as follows. Recall that we assume $1 \le ||z|| \le ||z'||/2$, and let $k \le n$. By using (2.3) and (2.2), we get for a constant C > 0 (and

recalling also the definition (2.1) for some positive constants c and C,

$$\mathbb{P}_{z}\left(E_{z',k,U} \cap E_{z',U}^{c}\right) \leq \mathbb{P}_{z}\left(H_{z'} \leq k \leq H_{z'+V_{0}}^{(k+1)}\right) \\
\leq C \mathbb{E}_{z}\left[\mathbb{I}\{H_{z'} \leq k\} \frac{1}{1 + \|S_{k} - z'\|^{d-2}}\right] \\
\leq C \sum_{i=1}^{k} \mathbb{P}_{z}(S_{i} = z') \mathbb{E}\left(\frac{1}{1 + \|S_{k-i}\|^{d-2}}\right) \\
\leq C \sum_{i=1}^{k} \frac{e^{-c\|z'\|^{2}/i}}{i\sqrt{i}} \frac{1}{1 + \sqrt{k-i}} \\
\leq C \frac{1}{\|z'\|\sqrt{k}} e^{-c\|z'\|^{2}/(2k)},$$

where for the second and third lines we used the Strong Markov Property, and Cauchy-Schwarz and (3.3) for the fourth one. In particular

$$\sup_{k\geq 1} \mathbb{P}_z \left(E_{z',k,U} \cap E_{z',U}^c \right) = \mathcal{O} \left(\frac{1}{\|z'\|^2} \right).$$

Then it follows by using again the Markov property, that

$$\mathbb{P}\left(E_{z,U}, E_{z',n,U} \cap E_{z',U}^{c}, H_{z} < H_{z'+V_{0}}\right) \leq \mathbb{P}\left(E_{z',n,U} \cap E_{z',U}^{c}, H_{z} < H_{z'+V_{0}}\right) \\
= \sum_{k \leq n} \mathbb{P}(H_{z} = n - k) \,\mathbb{P}_{z}\left(E_{z',k,U} \cap E_{z',U}^{c}\right) \\
= \mathcal{O}\left(\frac{1}{\|z\| \|z'\|^{2}}\right).$$

Thus indeed we can replace the event $z' \in \partial \widetilde{\mathcal{R}}_{n,U}$ by $E_{z',U}$ in (3.14). Now in the remaining sum, we gather together the pairs (z,z') and (z,-z'), and we get, using Cauchy-Schwarz again,

$$\left| \sum_{1 \leq 2||z|| \leq ||z'|| \leq n} \frac{\langle z, z' \rangle}{(1 + ||z||^d)(1 + ||z'||^d)} \mathbb{P}\left(E_{z,U}, E_{z',U}, H_z < H_{z'+V_0}\right) \right|$$

$$\leq \sum_{1 \leq 2||z|| \leq ||z'|| \leq n} \frac{2}{||z||^{d-1} ||z'||^{d-1}} \left| \mathbb{P}\left(E_{z,U}, E_{z',U}, H_z < H_{z'+V_0}\right) - \mathbb{P}\left(E_{z,U}, E_{-z',U}, H_z < H_{-z'+V_0}\right) \right|.$$
(3.16)

Then for any $1 \le ||z|| \le ||z'||/2$, we have

$$\mathbb{P}\left(E_{z,U}, E_{z',U}, H_z < H_{z'+V_0}\right) \\
= \sum_{y \in \overline{\partial} B(z, ||z||/2)} \mathbb{P}\left(E_{z,U}^*, H_z < H_{z'+V_0}, S_{\tau_z} = y\right) \mathbb{P}_y\left(H_{z+V_0} = \infty, E_{z',U}\right), \quad (3.17)$$

where by $\overline{\partial}B(z,||z||/2)$ we denote the set of points which are not in B(z,||z||/2) but have a neighbor in this ball, and where

$$E_{z,U}^* := \{V_0 \cap \{S_0 - z, \dots, S_{\tau_z} - z\}^c \sim U\}.$$

Now for any $y \in \overline{\partial} B(z, ||z||/2)$, and $||z'|| \ge 2||z||$, by using again (2.3) we get

$$\mathbb{P}_y\left(H_{z+V_0} = \infty, E_{z',U}\right) = \mathbb{P}_y\left(E_{z',U}\right) - \mathcal{O}\left(\frac{1}{\|z\|^{d-2}\|z'\|^{d-2}}\right). \tag{3.18}$$

Moreover, the same argument as in the proof of Lemma 2.1, shows that if y and y' are neighbors,

$$\mathbb{P}_{y}\left(E_{z',U}\right) = \mathbb{P}_{y'}\left(E_{z',U}\right) + \mathcal{O}\left(\frac{1}{\|z' - y\|^{d-1}}\right).$$

Therefore if $1 \le ||z|| \le ||z'||/2$ and $y \in \overline{\partial}B(z,||z||/2)$ we get

$$\mathbb{P}_{y}\left(E_{z',U}\right) = \mathbb{P}_{0}\left(E_{z',U}\right) + \mathcal{O}\left(\frac{\|y\|}{\|z' - y\|^{d-1}}\right) = \mathbb{P}_{0}\left(E_{z',U}\right) + \mathcal{O}\left(\frac{\|z\|}{\|z'\|^{d-1}}\right). \tag{3.19}$$

On the other hand, by symmetry, for any U,

$$\mathbb{P}_0\left(E_{z',U}\right) = \mathbb{P}_0\left(E_{-z',U}\right).$$

By combining this with (3.16), (3.17), (3.18) and (3.19), we obtain (3.13) and conclude the proof of (3.4).

4 Lower Bound on the Variance

In this section, we prove Proposition 1.6. When a simple random walk makes double backtracks, that is $S_{n-1} = S_{n-3}$ and $S_n = S_{n-2}$, then the range (and its boundary) remain constant during steps $\{n-2, n-1, n\}$. With this observation in mind, a lower bound on the variance is obtained as we decompose the simple random walk into two independent processes: a clock process counting the number of double-backtracks, and a trajectory without double-backtrack.

4.1 Clock Process

We construct by induction a no-double backtrack walk $(\widetilde{S}_n, n \in \mathbb{N})$. First, $\widetilde{S}_0 = 0$, and \widetilde{S}_1 and $\widetilde{S}_2 - \widetilde{S}_1$ are chosen uniformly at random among the elements of V_0 (the set of neighbors of the origin). Next, assume that \widetilde{S}_k has been defined for all $k \leq 2n$, for some $n \geq 1$. Let $\mathcal{N}_2 = \{(x, y) : x \sim 0 \text{ and } y \sim x\}$ and choose (X, Y) uniformly at random in $\mathcal{N}_2 \setminus \{(\widetilde{S}_{2n-1} - \widetilde{S}_{2n}, \widetilde{S}_{2n-2} - \widetilde{S}_{2n})\}$. Then set

$$\widetilde{S}_{2n+1} = \widetilde{S}_{2n} + X$$
 and $\widetilde{S}_{2n+2} = \widetilde{S}_{2n} + Y$. (4.1)

Thus, the walk \widetilde{S} has no double-backtrack. Note that by sampling uniformly in the whole of \mathcal{N}_2 we would have generated a simple random walk (SRW). Now, to build a SRW out of \widetilde{S} , it is enough to add at each even time a geometric number of double-backtracks. The geometric law is given by

$$\mathbb{P}(\xi = k) = (1 - p)p^k \quad \text{for all } k \ge 0, \tag{4.2}$$

with $p = 1/(2d)^2$. Note that the mean of ξ is equal to p/(1-p). Now, consider a sequence $(\xi_n, n \ge 1)$ of i.i.d. random variables distributed like ξ and independent of \widetilde{S} . Then define

$$\widetilde{N}_0 = \widetilde{N}_1 = 0$$
 and $\widetilde{N}_k := \sum_{i=1}^{[k/2]} \xi_i$ for $k \ge 2$. (4.3)

A SRW can be built out from \widetilde{S} and \widetilde{N} as follows. First, $S_i = \widetilde{S}_i$ for i = 0, 1, 2. Then, for any integer $k \geq 1$

$$S_{2i-1} = \widetilde{S}_{2k-1} \quad \text{and} \quad S_{2i} = \widetilde{S}_{2k} \quad \text{for all } i \in [k + \widetilde{N}_{2(k-1)}, k + \widetilde{N}_{2k}].$$

This implies that if $\widetilde{\mathcal{R}}$ is the range of \widetilde{S} and $\partial \widetilde{\mathcal{R}}$ its boundary, then for any integer k

$$\mathcal{R}_{k+2\widetilde{N}_k} = \widetilde{\mathcal{R}}_k \quad \text{and} \quad \partial \mathcal{R}_{k+2\widetilde{N}_k} = \partial \widetilde{\mathcal{R}}_k.$$
 (4.4)

4.2 A Law of Large Numbers and some consequences

Recall that Okada [Ok1] proved a law of large numbers for $|\partial \mathcal{R}_n|$, see (1.3), and call ν_d the limit of $|\partial \mathcal{R}_n|/n$. Since \widetilde{N}_n/n also converges almost surely toward p/[2(1-p)], we deduce from (4.4) that

$$\frac{|\partial \widetilde{\mathcal{R}}_n|}{n} \longrightarrow \frac{\nu_d}{1-p} \quad \text{almost surely.} \tag{4.5}$$

Let us show now another useful property. We claim that for any $\alpha > 0$,

$$\lim_{r \to \infty} \mathbb{P}(|(\mathcal{R}'_{\infty})^{+} \cap \widetilde{\mathcal{R}}_{r}^{+}| \ge \alpha r) = 0, \tag{4.6}$$

where \mathcal{R}'_{∞} is the total range of another simple random walk independent of $\widetilde{\mathcal{R}}$. To see this recall that the process (\mathcal{R}_n) is increasing, and therefore using (4.4) one deduce

$$\mathbb{P}\big(|(\mathcal{R}'_{\infty})^{+} \cap \widetilde{\mathcal{R}}_{r}^{+}| \geq \alpha r\big) \leq \mathbb{P}\big(\widetilde{N}_{r} \geq \frac{p}{1-p}r\big) + \mathbb{P}\big(|(\mathcal{R}'_{\infty})^{+} \cap \mathcal{R}_{Cr}^{+}| \geq \alpha r\big),$$

with C = 2p/(1-p) + 1. The first term on the right-hand side goes to 0, in virtue of the law of large numbers satisfied by \widetilde{N} , and the second one also as can be seen using Markov's inequality and the estimate:

$$\mathbb{E}[|(\mathcal{R}'_{\infty})^{+} \cap \mathcal{R}^{+}_{Cr}|] \leq \sum_{x,y \in \overline{V}_{0}} \sum_{z \in \mathbb{Z}^{d}} G(0,z+x) G_{Cr}(0,z+y) = \mathcal{O}(\sqrt{r}\log r),$$

which follows from (2.3) and [LG, Theorem 3.6].

A consequence of (4.6) is the following. Define $c = \nu_d/[2(1-p)]$. We have that for k large enough, any $t \ge 1$, and $r \ge \sqrt{k}$

$$\mathbb{P}(|\partial \widetilde{\mathcal{R}}_k| \ge t) \ge \frac{1}{2} \implies \mathbb{P}(|\partial \widetilde{\mathcal{R}}_{k+r}| \ge t + cr) \ge \frac{1}{4}, \tag{4.7}$$

and also

$$\mathbb{P}(|\partial \widetilde{\mathcal{R}}_k| \le t) \ge \frac{1}{2} \implies \mathbb{P}(|\partial \widetilde{\mathcal{R}}_{k-r}| \le t - cr) \ge \frac{1}{4}. \tag{4.8}$$

To see this first note that the set-inequality (1.10) holds as well for $\widetilde{\mathcal{R}}$. Hence, with evident notation

$$|\partial \widetilde{\mathcal{R}}_{k+r}| \ge |\partial \widetilde{\mathcal{R}}_r| + |\partial \widetilde{\mathcal{R}}(r, k+r)| - 2|\widetilde{\mathcal{R}}^+(r, r+k) \cap \widetilde{\mathcal{R}}_r^+|.$$

Since $\widetilde{\mathcal{R}}(r,r+k)$ is independent of $\widetilde{\mathcal{R}}_r$ and $\widetilde{\mathcal{R}}_\infty$ has the same law as \mathcal{R}_∞ , this gives

$$|\partial \widetilde{\mathcal{R}}_{k+r}| \ge |\partial \widetilde{\mathcal{R}}_r| + |\partial \widetilde{\mathcal{R}}(r, k+r)| - 2|(\mathcal{R}'_{\infty})^+ \cap \widetilde{\mathcal{R}}_r^+|, \tag{4.9}$$

with \mathcal{R}'_{∞} a copy of \mathcal{R}_{∞} independent of $\widetilde{\mathcal{R}}$. Now (4.5), (4.6) and (4.9) immediately give (4.7) and (4.8).

4.3 Lower Bound

First, by using (1.13), there is a positive constant $C_0 > 2d$, such that

$$|\mathbb{E}[|\partial \mathcal{R}_n|] - \nu_d n| \le C_0 \sqrt{n} \quad \text{for all } n \ge 1.$$
(4.10)

Take k_n to be the integer part of (1-p)n. We have either of the two possibilities

(i)
$$\mathbb{P}(|\partial \widetilde{\mathcal{R}}_{k_n}| \le \nu_d n) \ge \frac{1}{2}$$
 or (ii) $\mathbb{P}(|\partial \widetilde{\mathcal{R}}_{k_n}| \ge \nu_d n) \ge \frac{1}{2}$. (4.11)

Assume for instance that (i) holds, and note that (ii) would be treated symmetrically. Define, $i_n = [(1-p)(n-A\sqrt{n})]$, with $A = 3C_0/(c(1-p))$, and note that using (4.8)

$$\mathbb{P}(|\partial \widetilde{\mathcal{R}}_{i_n}| \le \nu_d n - 3C_0 \sqrt{n})) \ge \frac{1}{4},\tag{4.12}$$

for n large enough. Now set

$$\mathcal{B}_n = \left\{ \frac{2\widetilde{N}_{i_n} - 2\mathbb{E}[\widetilde{N}_{i_n}]}{\sqrt{n}} \in [A+1, A+2] \right\}.$$

Note that there is a constant $c_A > 0$, such that for all n large enough

$$P(\mathcal{B}_n) \ge c_A. \tag{4.13}$$

Moreover, by construction.

$$\mathcal{B}_n \subset \left\{ i_n + 2\widetilde{N}_{i_n} \in [n, n + 3\sqrt{n}] \right\}. \tag{4.14}$$

Now using the independence of \widetilde{N} and $\partial \widetilde{\mathcal{R}}$, (4.4), (4.12), (4.13) and (4.14), we deduce that

$$\mathbb{P}(\exists m \in \{0, \dots, 3\sqrt{n}\} : |\partial \mathcal{R}_{n+m}| \le \nu_d n - 3C_0 \sqrt{n}) \ge \frac{c_A}{4}.$$

Then one can use the deterministic bound:

$$|\partial \mathcal{R}_n| \le |\partial \mathcal{R}_{n+m}| + 2dm,$$

which holds for all $n \ge 1$ and $m \ge 0$. This gives

$$\mathbb{P}(|\partial \mathcal{R}_n| \le \nu_d n - 2C_0 \sqrt{n}) \ge \frac{c_A}{4},$$

which implies that $Var(|\partial \mathcal{R}_n|)/n \geq C_0^2 c_A/4 > 0$, using (4.10).

5 On Le Gall's decomposition

In this Section, we establish Propositions 1.6, 1.7 and 1.8.

5.1 Mean and Variance

Inequality (1.10) holds since

$$z \in \partial \mathcal{R}(0,n) \setminus (S_n + \mathcal{R}(n,n+m))^+ \cup \partial (S_n + \mathcal{R}(n,n+m)) \setminus \mathcal{R}^+(0,n) \Longrightarrow z \in \partial \mathcal{R}(0,n+m).$$

Define

$$X(i,j) = |\partial \mathcal{R}(i,j)|$$
 and $\overline{X}(i,j) = X(i,j) - \mathbb{E}[X(i,j)].$

Observe that in (1.10) the deviation from linearity is written in terms of an intersection of two independent range-like sets. This emphasizes the similarity between range and boundary of the range. Now (1.10) implies the same inequalities for the expectation.

$$0 \ge \mathbb{E}[X(0, n+m)] - (\mathbb{E}[X(0, n)] + \mathbb{E}[X(n, n+m)]) \ge -\mathbb{E}[Z(0, n, m)]. \tag{5.1}$$

Combining (1.10) and (5.1), we obtain our key (and simple) estimates

$$|\overline{X}(0, n+m) - (\overline{X}(0, n) + \overline{X}(n, n+m))| \le \max(Z(0, n, m), \mathbb{E}[Z(0, n, m)]). \tag{5.2}$$

If $||X||_p = (\mathbb{E}[X^p])^{1/p}$, then using the triangle inequality, we obtain for any p > 0,

$$\left| \|\overline{X}(0,n+m)\|_{p} - \|\overline{X}(0,n) + \overline{X}(n,n+m)\|_{p} \right| \le \|Z(0,n,m)\|_{p} + \|Z(0,n,m)\|_{1}. \tag{5.3}$$

The deviation from linearity of the centered p-th moment will then depend on the p-th moment of Z(0, n, m). We invoke now Hammersley's Lemma [HA], which extends the classical subadditivity argument in a useful manner.

Lemma 5.1. [Hammersley] Let (a_n) , (b_n) , and (b'_n) be sequences such that

$$a_n + a_m - b'_{n+m} \le a_{m+n} \le a_n + a_m + b_{n+m}$$
 for all m and n . (5.4)

Assume also that the sequences (b_n) and (b'_n) are positive and non-decreasing, and satisfy

$$\sum_{n>0} \frac{b_n + b_n'}{n(n+1)} < \infty. \tag{5.5}$$

Then, the limit of a_n/n exists, and

$$-\frac{b'_n}{n} + 4\sum_{k>2n} \frac{b'_k}{k(k+1)} \ge \frac{a_n}{n} - \lim_{k\to\infty} \frac{a_k}{k} \ge +\frac{b_n}{n} - 4\sum_{k>2n} \frac{b_k}{k(k+1)}.$$
 (5.6)

We obtain now the following moment estimate.

Lemma 5.2. For any integer k, there is a constant C_k such that for any n, m integers,

$$E[Z^{k}(0,n,m)] \le C_{k}(\psi_{d}^{k}(n)\psi_{d}^{k}(m))^{1/2}.$$
(5.7)

Recall that ψ_d is defined in (1.12).

Proof. Observe that Z(0, n, m) is bounded as follows.

$$Z(0, n, m) \leq 2 \sum_{z \in \mathbb{Z}^d} \mathbb{I}\{z \in \mathcal{R}^+(0, n) \cap (S_n + \mathcal{R}^+(n, n+m))\}$$

$$\leq 2 \sum_{z \in \mathbb{Z}^d} \mathbb{I}\{z \in \overleftarrow{\mathcal{R}}^+(0, n) \cap \mathcal{R}^+(n, n+m)\}$$

$$\leq 2 \sum_{x \in \overleftarrow{V}_0} \sum_{y \in \overleftarrow{V}_0} \sum_{z \in \mathbb{Z}^d} \mathbb{I}\{z + x \in \overleftarrow{\mathcal{R}}^+(0, n), z + y \in \mathcal{R}^+(n, n+m)\}.$$

We now take the expectation of the k-th power, and use the independence of $\overline{\mathcal{R}}^+(0,n)$ and $\mathcal{R}(n,n+m)$

$$\begin{split} E[Z^k(0,n,m)] &\leq 2^k \sum_{x_1,y_1 \in \overline{V}_0} \dots \sum_{x_k,y_k \in \overline{V}_0} \sum_{z_1,\dots,z_k} \mathbb{E}\left[\prod_{i=1}^k \mathbb{I}\{z_i + x_i \in \overleftarrow{\mathcal{R}}(0,n)\} \mathbb{I}\{z_i + y_i \in \mathcal{R}(n,n+m)\}\right] \\ &\leq 2^k \sum_{x_1,y_1 \in \overline{V}_0} \dots \sum_{x_k,y_k \in \overline{V}_0} \sum_{z_1,\dots,z_k} \mathbb{P}\left(H_{z_i + x_i} < n, \ \forall i = 1,\dots,k\right) \mathbb{P}\left(H_{z_i + y_i} < m, \ \forall i = 1,\dots,k\right) \\ &\leq 2^k |\overline{V}_0|^{2k} \Big(\sum_{z_1,\dots,z_k} \mathbb{P}\left(H_{z_i} < n, \ \forall i = 1,\dots,k\right)^2\Big)^{1/2} \Big(\sum_{z_1,\dots,z_k} \mathbb{P}\left(H_{z_i} < m, \ \forall i = 1,\dots,k\right)^2\Big)^{1/2} \\ &\leq 2^k |\overline{V}_0|^{2k} k! \Big(\sum_{z_1,\dots,z_k} G_n^2(0,z_1) \dots G_n^2(z_{k-1},z_k)\Big)^{1/2} \Big(\sum_{z_1,\dots,z_k} G_n^2(0,z_1) \dots G_m^2(z_{k-1},z_k)\Big)^{1/2} \\ &\leq C_k \Big(\psi_d^k(n) \psi_d^k(m)\Big)^{1/2}, \end{split}$$

which concludes the proof.

Henceforth, and for simplicity, we think of ψ_d of (1.12) rather as $\psi_3(n) = \mathcal{O}(\sqrt{n})$, $\psi_4(n) = \mathcal{O}(\log(n))$ and for $d \geq 5$, $\psi_d(n) = \mathcal{O}(1)$ (in other words, we aggregate in ψ_d innocuous constants). As an immediate consequence of (5.1) and Lemma 5.2, we obtain for any $n, m \in \mathbb{N}$,

$$\mathbb{E}[|\partial \mathcal{R}_n|] + \mathbb{E}[|\partial \mathcal{R}_m|] - \max(\psi_d(n), \psi_d(m)) \le \mathbb{E}[|\partial \mathcal{R}_{n+m}|] \le \mathbb{E}[|\partial \mathcal{R}_n|] + \mathbb{E}[|\partial \mathcal{R}_m|]. \tag{5.8}$$

The inequalities of (5.8) and Hammersley's Lemma imply that the limit of $\mathbb{E}[|\partial \mathcal{R}_n|]/n$ exists and it yields (1.13) of Proposition 1.8.

Variance of X(0,n). Let us write (5.3) for p=2

$$\left| \|\overline{X}(0, n+m)\|_{2} - \|\overline{X}(0, n) + \overline{X}(n, n+m)\|_{2} \right| \le 2\|Z(0, n, m)\|_{2}. \tag{5.9}$$

Now, the independence of $\overline{X}(0,n)$ and $\overline{X}(n,n+m)$ gives

$$\|\overline{X}(0,n) + \overline{X}(n,n+m)\|_{2}^{2} = \|\overline{X}(0,n)\|_{2}^{2} + \|\overline{X}(0,m)\|_{2}^{2}.$$
(5.10)

By taking squares on both sides of (5.9) and using (5.10), we obtain

$$\|\overline{X}(0, n+m)\|_{2}^{2} \leq \|\overline{X}(0, n)\|_{2}^{2} + \|\overline{X}(0, m)\|_{2}^{2} + 4\|\overline{X}(0, n) + \overline{X}(n, n+m)\|_{2}\|Z(0, n, m)\|_{2} + 4\|Z(0, n, m)\|_{2}^{2},$$

$$(5.11)$$

and

$$\|\overline{X}(0,n)\|_{2}^{2} + \|\overline{X}(0,m)\|_{2}^{2} \leq \|\overline{X}(0,n+m)\|_{2}^{2} + 4\|\overline{X}(0,n+m)\|_{2}\|Z(0,n,m)\|_{2} + 4\|Z(0,n,m)\|_{2}^{2}.$$
(5.12)

Now, using (5.11) and Lemma 5.2 with k=2, we have for n=m and for $\varepsilon>0$,

$$\operatorname{Var}(X(0,2n)) \le (1+\varepsilon)2\operatorname{Var}(X(0,n)) + (1+\frac{1}{\varepsilon})\psi_d^2(n). \tag{5.13}$$

We iterate this inequality L times with L the integer part of $\log n/\log 2$ to obtain for some constant C independent of n, L

$$\operatorname{Var}(X(0,n)) \leq C(1+\varepsilon)^{L} 2^{L} + C'_{d}(1+\frac{1}{\varepsilon}) \sum_{k=1}^{L} (1+\varepsilon)^{k-1} 2^{k-1} \psi_{d}^{2}(\frac{n}{2^{k}})$$

$$\leq C n \log^{2}(n) \quad \text{when we choose} \quad \varepsilon = \frac{1}{L}.$$
(5.14)

Second, we use the rough bound of (5.14) as an a priori bound for the upper and lower bounds respectively (5.11) and (5.12) for Var(X(0,2n)), in order to apply Hammersley's Lemma with $b_n = \sqrt{n} \log n \times \psi_d(n)$. In dimension four or more we do fulfill the hypotheses of Hammersley's Lemma, which in turn produces the improved bound $\text{Var}(|\partial \mathcal{R}_n|) \leq Cn$, and then again we can use Hammersley's Lemma with a smaller $b_n = \sqrt{n}\psi_d(n)$ which eventually yields (1.14) of Proposition 1.8. The fact that the limit of the normalized variance is positive follows from Proposition 1.6.

5.2 Central Limit Theorem

The principle of Le Gall's decomposition is to repeat dividing each strand into smaller and smaller pieces producing independent boundaries of *shifted* ranges. For two reals s, t let [s], [t] be their integer parts and define X(s,t) = X([s], [t]). For l and k integer, let $X_k^{(l)} = X((k-1)n/2^l, kn/2^l)$. Let also $Z_k^{(l)}$ be the functional of the two strands obtained by dividing the k-th strand after l-1 divisions. Thus, after L divisions, with $2^L < n$, we get

$$\sum_{i=1}^{2^L} X_i^{(L)} - \sum_{l=1}^{L} \sum_{i=1}^{2^{l-1}} Z_i^{(l)} \le X(0,n) \le \sum_{i=1}^{2^L} X_i^{(L)}.$$

The key point is that $\{X_i^{(L)},\ i=1,\ldots,2^L\}$ are independent, and have the same law as $X(0,n/2^L)$ or $X(0,n/2^L+1)$. Now, we define the error term $\mathcal{E}(n)>0$ as

$$\overline{X}(0,n) = \sum_{i=1}^{2^L} (X_i^{(L)} - \mathbb{E}[X_i^{(L)}]) - \mathcal{E}(n),$$

and (1.10) implies that

$$\mathbb{E}[\mathcal{E}(n)] \le \sum_{l=1}^{L} 2^{l} \psi_d(\frac{n}{2^{l}}).$$

Note that we can choose L growing to infinity with n, but such that $\mathbb{E}[\mathcal{E}(n)]/\sqrt{n}$ goes to 0.

We now use a remark of Le Gall [LG, Remark (iii) p.503]. In order to show that Lindeberg's condition for the triangular array $\{X_i^{(L)},\ i=1,\ldots,2^L\}$ is satisfied, we estimate the fourth moment of X(0,n) using the previous decomposition and a sub-additivity argument. Using (5.3) with p=4, we have

$$\|\overline{X}(0, n+m)\|_{4} \le \left(\left(\mathbb{E}[\overline{X}^{4}(0, n)] + 6\mathbb{E}[\overline{X}^{2}(0, n)] \mathbb{E}[\overline{X}^{2}(0, m)] + \mathbb{E}[\overline{X}^{4}(0, m)] \right) \right)^{1/4} + \sqrt{\psi_{d}(n)\psi_{d}(m)}.$$

Take m = n, to obtain

$$\|\overline{X}(0,2n)\|_{4} \leq \left(2\mathbb{E}[\overline{X}^{4}(0,n)] + 6\mathbb{E}[\overline{X}^{2}(0,n)]^{2}\right)^{1/4} + \psi_{d}(n)$$
$$\leq 2^{1/4}\|\overline{X}(0,n)\|_{4} + 6^{1/4}\|\overline{X}(0,n)\|_{2} + \psi_{d}(n).$$

Define $A_n = ||\overline{X}(0, n)||_4 / (||\overline{X}(0, n))||_2$, then

$$A_{2n} \le \left(\frac{2^{1/4}}{2^{1/2}}A_n + \frac{6^{1/4}}{2^{1/2}}\right) \frac{2^{1/2} \|\overline{X}(0,n)\|_2}{\|\overline{X}(0,2n)\|_2} + \frac{\psi_d(n)}{\|\overline{X}(0,2n)\|_2}.$$

The inequality (5.2) is valid in any dimension. It is however only in dimension larger than three that we have a lower bound for the variance. Note that (5.2) yields

$$\left|\overline{X}(0,n) + \overline{X}(n,m)\right| \le \left|\overline{X}(0,m)\right| + Z(0,n,m).$$

With m = 2n, we obtain

$$\left(2\mathbb{E}[\overline{X}^2(0,n)]\right)^{1/2} \le \|\overline{X}(0,2n)\|_2 + \psi_d(n).$$

Thus,

$$\frac{2^{1/2} \|\overline{X}(0,n)\|_2}{\|\overline{X}(0,2n)\|_2} \le 1 + \frac{\psi_d(n)}{\|\overline{X}(0,2n)\|_2}.$$

This implies that

$$A_{2n} \le \frac{1 + o(1)}{2^{1/4}} A_n + \frac{6^{1/4}}{2^{1/2}} (1 + o(1)).$$

and therefore (A_n) is bounded. We deduce that Lindeberg's condition for triangular arrays is satisfied (see [F] page 530, or [HH]), and as a consequence the Central Limit Theorem holds.

Appendix

A Estimates on Ranges

In this section, we prove Proposition 1.4. We first introduce some other range-like sets allowing us to use the approach of Jain and Pruitt [JP]. Recall that the sets $\mathcal{R}_{n,V}$ are disjoints, and for $U \subset V_0$, define

$$\overline{\mathcal{R}}_{n,U} := \bigcup_{V \supset U} \mathcal{R}_{n,V}. \tag{A.1}$$

Next for $U \subset V_0$, define

$$\alpha(U) = |\overline{\mathcal{R}}_{n,U}| - \mathbb{E}(|\overline{\mathcal{R}}_{n,U}|)$$
 and $\beta(U) = |\mathcal{R}_{n,U}| - \mathbb{E}(|\mathcal{R}_{n,U}|).$

The definition (A.1) yields

$$\alpha(U) = \sum_{V \supset U} \beta(V),$$

and this relation can be inverted as follows:

$$\beta(V) = \sum_{U\supset V} (-1)^{|U\setminus V|} \alpha(U).$$

As a consequence, for $V \subset V_0$,

$$\operatorname{Var}(|\mathcal{R}_{n,V}|) = \mathbb{E}\left(\beta^2(V)\right) \le 2^{|V_0 \setminus V|} \sum_{U \supset V} \mathbb{E}\left(\alpha^2(U)\right).$$

We will see below that each $\overline{\mathcal{R}}_{n,V}$ has the same law as a range-like functional that Jain and Pruitt analyze by using a last passage decomposition, after introducing some new variables. But let us give more details now. So first, we fix some $V \subset V_0$, and for $n \in \mathbb{N}$, set $Z_n^n = 1$, and

$$Z_{i} = \mathbf{1} \left(\{ S_{i+k} \notin (S_{i} + \overline{V}) \mid \forall k \geq 1 \} \right) \quad \forall i \in \mathbb{N},$$

$$Z_{i}^{n} = \mathbf{1} \left(\{ S_{i+k} \notin (S_{i} + \overline{V}) \mid \forall k = 1, \dots, n-i \} \right) \quad \forall i < n$$

$$W_{i}^{n} = Z_{i}^{n} - Z_{i} \quad \forall i \leq n,$$

where

$$\overline{V} = V \cup \{0\}.$$

A key point in this decomposition is that Z_n and Z_i^n are independent. Now, define

$$\underline{\mathcal{R}}_{n,V} = \sum_{i=0}^{n-1} Z_i^n, \tag{A.2}$$

and note that since the increments are symmetric and independent, $\overline{\mathcal{R}}_{n,V}$ and $\underline{\mathcal{R}}_{n,V}$ are equal in law. Now, equality (A.2) reads as

$$\underline{\mathcal{R}}_{n,V} = \sum_{i=0}^{n-1} Z_i + \sum_{i=0}^{n-1} W_i^n,$$

and thus by Cauchy-Schwarz,

$$\operatorname{Var}(\underline{\mathcal{R}}_{n,V}) \le 2 \sum_{i=1}^{n-1} \operatorname{Var}(Z_i) + 4 \sum_{i=1}^{n-1} \sum_{i=0}^{j-1} \operatorname{Cov}(Z_i, Z_j) + 4 \sum_{i=1}^{n-1} \sum_{i=0}^{j} \mathbb{E}\left(W_i^n W_j^n\right). \tag{A.3}$$

Next for i < j < n, we have (recall the definition (2.1))

$$\begin{split} \mathbb{E}\left(W_i^n\,W_j^n\right) &=& \mathbb{P}\left(n < H_{S_i + \overline{V}}^{(i+1)} < \infty, \; n < H_{S_j + \overline{V}}^{(j+1)} < \infty\right) \\ &=& \sum_{x \neq \overline{V}} \mathbb{P}(S_{j-i} = x, \; H_{\overline{V}} > j-i)\,\mathbb{P}_x\big(n-j < H_{\overline{V}} < \infty, \; n-j < H_{x + \overline{V}} < \infty\big), \end{split}$$

where for the second equality we just used the Markov property and translation invariance of the walk. Therefore

$$\sum_{i=0}^{j} \mathbb{E}\left(W_{i}^{n} W_{j}^{n}\right) \leq \sum_{x \notin \overline{V}} G_{j}(0, x) \, \mathbb{P}_{x}\left(H_{\overline{V}} < \infty, \ n - j < H_{x + \overline{V}} < \infty\right) \\
\leq \sum_{u, z \in \overline{V}} \sum_{x \notin \overline{V}} G_{j}(0, x) \, \mathbb{P}_{x}\left(H_{y} < \infty, \ n - j < H_{x + z} < \infty\right).$$

Then Lemma 4 of [JP] shows that

$$\sum_{i=0}^{j} \mathbb{E}\left(W_{i}^{n} W_{j}^{n}\right) = \begin{cases} \mathcal{O}\left(\sqrt{\frac{j}{n-j}}\right) & \text{if } d = 3\\ \mathcal{O}\left(\frac{\log j}{n-j}\right) & \text{if } d = 4\\ \mathcal{O}\left((n-j)^{1-d/2}\right) & \text{if } d \geq 5, \end{cases}$$

and thus

$$\sum_{j=1}^{n-1} \sum_{i=0}^{j} \mathbb{E}\left(W_i^n W_j^n\right) = \begin{cases} \mathcal{O}(n) & \text{if } d = 3\\ \mathcal{O}((\log n)^2) & \text{if } d = 4\\ \mathcal{O}(1) & \text{if } d \ge 5. \end{cases}$$
(A.4)

Now, for i < j < n, by using that Z_i^j and Z_j are independent, we get

$$Cov(Z_i, Z_j) = -Cov(W_i^j, Z_j).$$

On the other hand, similarly as above,

$$\mathbb{E}(W_i^j Z_j) = \mathbb{P}\left(n < H_{S_i + \overline{V}}^{(i+1)} < \infty, H_{S_j + \overline{V}}^{(j+1)} = \infty\right)$$

$$= \sum_{x \notin \overline{V}} \mathbb{P}(S_{j-i} = x, H_{\overline{V}} > j - i) \mathbb{P}_x \left(H_{\overline{V}} < \infty, H_{x + \overline{V}} = \infty\right).$$

Since in addition,

$$\mathbb{E}(Z_i) = \mathbb{P}_x \left(H_{x \perp \overline{V}} = \infty \right)$$
 for all x ,

and

$$\mathbb{E}(W_i^j) = \sum_{x \notin \overline{V}} \mathbb{P}\left(S_{j-i} = x, H_{\overline{V}} > j - i\right) \mathbb{P}_x \left(H_{\overline{V}} < \infty\right),$$

we deduce that

$$\operatorname{Cov}(Z_i, Z_j) = \sum_{x \notin \overline{V}} \mathbb{P}(S_{j-i} = x, H_{\overline{V}} > j - i) b_V(x),$$

with

$$b_{V}(x) := \mathbb{P}_{x} \left(H_{\overline{V}} < \infty \right) \, \mathbb{P}_{x} \left(H_{x+\overline{V}} = \infty \right) - \mathbb{P}_{x} \left(H_{\overline{V}} < \infty, \ H_{x+\overline{V}} = \infty \right). \tag{A.5}$$

Now we need the following equivalent of Lemma 5 of [JP].

Lemma A.1. For any $V \subset V_0$, and $x \notin \overline{V}$,

$$b_V(x) = \mathbb{P}_x (H_{\overline{V}} < H_{r+\overline{V}} < \infty) \mathbb{P}_x (H_{\overline{V}} = \infty) + \mathcal{E}(x, V),$$

with

$$\mathcal{E}(x,V) := \sum_{z \in x + \overline{V}} \mathbb{P}_x \left(S_{H_{x + \overline{V}}} = z, \ H_{x + \overline{V}} < H_{\overline{V}} \right) \left(\mathbb{P}_z \left(H_{\overline{V}} < \infty \right) - \mathbb{P}_x \left(H_{\overline{V}} < \infty \right) \right).$$

Moreover,

$$|\mathcal{E}(x,V)| = \mathcal{O}\left(\frac{1}{\|x\|^{d-1}}\right).$$

Assuming this lemma for a moment, we get

$$a_{j} = \sum_{i=0}^{j-1} \operatorname{Cov}(Z_{i}, Z_{j}) = \sum_{i=0}^{j-1} \sum_{x \notin \overline{V}} \mathbb{P}(S_{j-i} = x, H_{\overline{V}} > j - i) b_{V}(x)$$

$$= \mathcal{O}\left(\sum_{x \notin \overline{V}} \frac{G_{j}(0, x)}{\|x\|^{d-1}}\right) = \mathcal{O}\left(\sum_{1 \leq \|x\| \leq j} \frac{1}{\|x\|^{2d-3}}\right)$$

$$= \begin{cases} \mathcal{O}(\log j) & \text{if } d = 3\\ \mathcal{O}(1) & \text{if } d \geq 4, \end{cases}$$

from which we deduce that

$$\sum_{j=0}^{n-1} a_j = \begin{cases} \mathcal{O}(n\log n) & \text{if } d = 3\\ \mathcal{O}(n) & \text{if } d \ge 4. \end{cases}$$
(A.6)

Then Proposition 1.4 follows from (A.3), (A.4) and (A.6).

Proof of Lemma A.1. Note first that

$$\begin{array}{lcl} b_V(x) & = & \mathbb{P}_x \big(H_{\overline{V}} < \infty, \ H_{x+\overline{V}} < \infty \big) - \mathbb{P}_x \big(H_{\overline{V}} < \infty \big) \mathbb{P}_x \big(H_{x+\overline{V}} < \infty \big) \\ & = & \mathbb{P}_x \big(H_{\overline{V}} < H_{x+\overline{V}} < \infty \big) + \mathbb{P}_x \big(H_{x+\overline{V}} < H_{\overline{V}} < \infty \big) - \mathbb{P}_x \big(H_{\overline{V}} < \infty \big) \mathbb{P}_x \big(H_{x+\overline{V}} < \infty \big). \end{array}$$

Moreover,

$$\begin{split} \mathbb{P}_x \big(H_{x+\overline{V}} < H_{\overline{V}} < \infty \big) &= \sum_{z \in x+\overline{V}} \mathbb{P}_x \big(S_{H_{x+\overline{V}}} = z, \ H_{x+\overline{V}} < H_{\overline{V}} \big) \mathbb{P}_z \big(H_{\overline{V}} < \infty \big) \\ &= \mathbb{P}_x \big(H_{x+\overline{V}} < H_{\overline{V}} \big) \mathbb{P}_x (H_{\overline{V}} < \infty) + \mathcal{E}(x,V). \end{split}$$

The first assertion of the lemma follows. The last assertion is then a direct consequence of standard asymptotics on the gradient of the Green's function (see for instance [LL, Corollary 4.3.3]).

Remark. By adapting the argument in [JP] we could also prove that in dimension 3, $Var(|\underline{\mathcal{R}}_{n,V}|) \sim \sigma^2 n \log n$, for some constant $\sigma > 0$, and then obtain a central limit theorem for this modified range. However it is not clear how to deduce from it an analogous result for $|\mathcal{R}_{n,V}|$, which would be useful in view of a potential application to the boundary of the range.

Acknowledgements. We would like to thank Gregory Maillard for discussions at an early stage of this work. We also thank Pierre Mathieu for mentioning us a gap in a previous version, where we omitted to show that the limiting term in (1.14) was nonzero. A.A. received support of the A*MIDEX grant (ANR-11-IDEX-0001-02) funded by the French Government "Investissements d'Avenir" program.

References

- [AS] A. Asselah and B. Schapira, Boundary of the Range II: Lower Tail Large Deviations.
- [BKYY] I. Benjamini, G. Kozma, A. Yadin, A. Yehudayoff, Entropy of random walk range, Ann. Inst. Henri Poincaré Probab. Stat. 46 (2010), 1080–1092.
- [BY] N. Berestycki and A. Yadin, Condensation of random walks and the Wulff crystal, arXiv:1305.0139.
- [C] X. Chen, Random Walk Intersections, Large Deviations and Related Topics. Mathematical Surveys and Monographs, Vol 157, 2009, AMS.
- [DE] A. Dvoretzky, P. Erdós, Some problems on random walk in space, Proceedings of the Second Berkeley Symposium on Mathematical Statistics and Probability, 1950, pp. 353–367. University of California Press, Berkeley and Los Angeles, (1951).
- [F] **W. Feller**, An introduction to probability theory and its applications, II, New York, Wiley, 1961.
- [HSC] W. Hebisch, L. Saloff-Coste, Gaussian estimates for Markov chains and random walks on groups, Ann. Probab. 21, (1993), 673–709.
- [HH] P. Hall, C. C. Heyde, Martingale limit theory and its application, Probability and Mathematical Statistics. Academic Press, Inc., New York-London, (1980), xii+308 pp.
- [HA] J. M. Hammersley, Generalization of the Fundamental Theorem on Subadditive Functions, Proc. Cambridge Phil. Society, 58, (1962), 235–238.
- [JP] N. C. Jain, W. E. Pruitt, The range of transient random walk, J. Analyze Math. 24 (1971), 369–393.

- [LL] G. F. Lawler and V. Limic, Random walk: a modern introduction. Cambridge Studies in Advanced Mathematics, 123. Cambridge University Press, Cambridge, (2010). xii+364 pp.
- [LG] J.-F. Le Gall, Propriétés d'intersection des marches aléatoires. I. Convergence vers le temps local d'intersection, Comm. Math. Phys. 104 (1986), 471–507.
- [Ok1] I. Okada The inner boundary of random walk range, to appear in J. Math. Soc. Japan.
- [Ok2] I. Okada Frequently visited site of the inner boundary of simple random walk range, arXiv:1409.8368.