Lower bounds on fluctuations for internal DLA

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Abstract

We consider internal diffusion limited aggregation in dimension larger than or equal to two. This is a random cluster growth model, where random walks start at the origin of the $d$-dimensional lattice, one at a time, and stop moving when reaching a site that is not occupied by previous walks.

When $n$ random walks are sent from the origin, we establish a lower bound for the inner and outer errors fluctuations of order square root of the logarithm of $n$. When dimension is larger or equal to three, this lower bound matches the upper bound recently obtained in independent works of [2] and [5]. Also, we produce as a corollary of our proof of [2], an upper bound for the fluctuation of the inner error in a specified direction.

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1 Introduction

We establish lower bounds for the inner and outer errors in internal diffusion limited aggregation (internal DLA). Internal DLA models a discrete cluster growth, and is defined as follows. Let $\Lambda$ be a subset of $\mathbb{Z}^d$ which represents the explored region at time 0. Let $N$ be an integer, and $\xi = (\xi_1, \ldots, \xi_N)$ be the initial positions of $N$ independent simple random walks $S_1, \ldots, S_N$ on $\mathbb{Z}^d$. The cluster of volume $N$, denoted $A(\Lambda, \xi)$ is obtained inductively as follows. First, $A(\Lambda, 0) = \Lambda$. Now, assume $A(\Lambda, k-1)$ is obtained, and define

$$\tau_k = \inf \{ t \geq 0 : S_k(t) \notin A(\Lambda, k-1) \}, \quad \text{and} \quad A(\Lambda, k) = A(\Lambda, k-1) \cup \{ S_k(\tau_k) \}. \quad (1.1)$$

We call explorers the random walks obeying the aggregation rule (1.1). We say that the explorer $k$ settles at time $\tau_k$. When $\Lambda = \emptyset$, we also denote $A(\emptyset, \xi)$ by $A(\xi)$.

In dimensions two and more, Lawler, Bramson and Griffeath [7] prove that in order to cover, without holes, a sphere of radius $n$, we need about the number of sites of $\mathbb{Z}^d$ in this sphere. In other words, the asymptotic shape of the cluster is a sphere. Then, Lawler in [8] shows a subdiffusive upper bound for the worse fluctuation to the spherical shape. More
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precisely, the latter result is formulated in terms of inner and outer errors, which we now introduce with some notation. We denote with $\| \cdot \|$ the euclidean norm on $\mathbb{R}^d$. For any $x$ in $\mathbb{R}^d$ and $r$ in $\mathbb{R}$, set

$$B(x, r) = \{ y \in \mathbb{R}^d : \| y - x \| < r \} \quad \text{and} \quad B(x, r) = B(x, r) \cap \mathbb{Z}^d. \quad (1.2)$$

For $\Lambda \subset \mathbb{Z}^d$, $|\Lambda|$ denotes the number of sites in $\Lambda$. The inner error $\delta_I(n)$ is such that

$$n - \delta_I(n) = \sup \left\{ r \geq 0 : B(0, r) \subset A(|B(0, n)|) \right\}. \quad (1.3)$$

Also, the outer error $\delta_O(n)$ is such that

$$n + \delta_O(n) = \inf \left\{ r \geq 0 : A(|B(0, n)|) \subset B(0, r) \right\}. \quad (1.4)$$

Note that $\delta_I$ and $\delta_O$ represent the worse fluctuations to the spherical shape.

In 2010, Asselah-Gaudilli`ere in [1, 2], and Jerison-Levine-Sheffield in [4, 5], independently showed the following upper bound.

**Theorem 1.1** When dimension $d \geq 3$, there are constants $\{\beta_d, d \geq 3\}$ such that with probability 1,

$$\limsup_{n \to \infty} \frac{\delta_I(n)}{\sqrt{\log(n)}} \leq \beta_d, \quad \text{and} \quad \limsup_{n \to \infty} \frac{\delta_O(n)}{\sqrt{\log(n)}} \leq \beta_d. \quad (1.5)$$

When dimension is 2, there is a constant $\{\beta_2\}$ such that with probability 1,

$$\limsup_{n \to \infty} \frac{\delta_I(n)}{\log(n)} \leq \beta_2, \quad \text{and} \quad \limsup_{n \to \infty} \frac{\delta_O(n)}{\log(n)} \leq \beta_2. \quad (1.6)$$

All these studies are concerned with upper bounds on $\delta_I$ and $\delta_O$, and the matter of showing that they were indeed realized remained untouched. We present now two results on lower bounds. The first one is independent from previous results. This bound which characterizes the worse fluctuations is optimal in $d \geq 3$. Morally, it says that if there are no deep holes, then long tentacles form.

**Proposition 1.2** When dimension is three or more, set $h(k) = \sqrt{\log(k)}$ there is $\alpha_d > 0$ such that

$$\lim_{n \to \infty} P \left( \exists k > n, \delta_O(n) \geq \alpha_d h(k) \mid \delta_I(n) < \alpha_d h(n) \right) = 1. \quad (1.7)$$

When dimension is two, set $h(k) = \sqrt{\log(k) \log(\log(k))}$. There is $\alpha_2 > 0$ such that

$$\lim_{n \to \infty} P \left( \exists k > n, \delta_O(n) \geq \alpha_2 h(k) \mid \delta_I(n) < \alpha_2 h(n) \right) = 1. \quad (1.8)$$

**Remark 1.3** Note that as a consequence of (1.7), we have $\alpha_d$ positive such that

$$P \left( \delta_O(n) \geq \alpha_d \sqrt{\log(n)} \text{ or } \delta_I(n) \geq \alpha_d \sqrt{\log(n)}, \text{ i.o.} \right) = 1, \quad (1.9)$$

The second result uses the upper bound of Theorem 1.1. It says that deep holes do form.
Theorem 1.4 There are positive constants \(\{\alpha_d, d \geq 2\}\) such that when \(d \geq 3\),
\[
P(\delta_I(n) > \sqrt{\alpha_d \log(n)} \ i.o.) = 1
\]  

(1.10)

Remark 1.5 We believe that in dimension 2, for some \(\alpha_2 > 0\), almost surely both \(\delta_I(n)\) and \(\delta_O(n)\) are larger than \(\alpha_2 \log(n)\) infinitely often. However, the way of realizing this event is probably different from what we describe in \(d \geq 3\).

We present now a side result dealing with fluctuation in a given direction. The results on fluctuations for internal DLA [7, 8, 1, 2, 4, 5] all focus on inner and outer errors. The paper [6] if still of a different nature: it addresses averaged error, in a sense where inner and outer errors can cancel each other out.

Though interesting, fluctuations in a given direction remained untreated. This is presented as an open problem in Section 3 of [5], and we note that the proof of [2] yields the following bound for the inner error in a given direction.

Proposition 1.6 There are positive constants \(\{\kappa_d, d \geq 2\}\) such that for \(z\) be such that \(\|z\| < n\), we have in \(\mathbb{Z}^d\)
\[
P(z \notin A(N_n)) \leq \begin{cases} 
\exp \left( -\kappa_2 \frac{(n-\|z\|)^2}{\log(n)} \right) & \text{for } d = 2, \\
\exp \left( -\kappa_d (n - \|z\|)^2 \right) & \text{for } d \geq 3.
\end{cases}
\]  

(1.11)

The rest of the paper is organized as follows. In Section 2, we set notation and recall useful results. In Section 3, we deal with the outer error and prove Proposition 1.2. In Section 4 we deal with the inner error, and prove Theorem 1.4. Finally, in Section 5 we explain how to read Proposition 1.6 as a corollary of our previous work [2].

2 Notation and Prerequisite

Let \(S : \mathbb{N} \to \mathbb{Z}^d\) denotates a simple random walk on \(\mathbb{Z}^d\). When the initial condition is \(S(0) = z \in \mathbb{Z}^d\), its law is denoted \(P_z\). The first time \(S\) hits a domain \(\Lambda \subset \mathbb{Z}^d\), is denoted \(H(\Lambda)\), or \(H(S; \Lambda)\) to emphasize that \(S\) is the walk.

For a positive \(\gamma\), we denote by \(\rho(\gamma)\) the radius of the largest ball centered at 0 whose volume is less than \(\gamma\). In other words,
\[
\rho(\gamma) = \sup \{ n \geq 0 : |\mathbb{B}(0, n)| \leq \gamma \}.
\]  

(2.1)

The abbreviation \(b(n) = |\mathbb{B}(0, n)|\) is also handy.

We now consider a configuration \(\eta \in \mathbb{N}^{\mathbb{Z}^d}\) with a finite number of particles
\[
|\eta| := \sum_{z \in \mathbb{Z}^d} \eta(z) < \infty.
\]  

(2.2)

For \(R > 0\), \(\Lambda \subset \mathbb{Z}^d\), and \(z \in \mathbb{B}(0, R)\) (resp. \(z \in \partial \mathbb{B}(0, R)\)), we denote by \(W_R(\Lambda; \eta, z)\) the number of explorers visiting \(z\) before they exit \(\mathbb{B}(0, R)\) (resp. at the moment they reach
\[ \partial \mathbb{B}(0, R) \], when the explorers start on \( \eta \) with an explored region \( \Lambda \). The positions of explorers settled before they exit \( \mathbb{B}(0, R) \) is denoted \( A_R(\Lambda; \eta) \), which is also the support of \( W_R(\Lambda; \eta, z) \) in \( \mathbb{B}(0, R) \). Note that \( A_R(\Lambda; \eta) \) is contained in \( \mathbb{B}(0, R) \).

When we build the internal DLA cluster from \( n \) independent random walks, we also consider the unrestricted trajectories. This gives a natural coupling between explorers and independent random walks. Note that if \( \mathbb{B}(0, R) \subset \Lambda \) and \( z \in \partial \mathbb{B}(0, R) \), then \( W_R(\Lambda; \eta, z) = W_R(\mathbb{B}(0, R); \eta, z) \), and this corresponds to the number of independent random walks which exit \( \mathbb{B}(0, R) \) at site \( z \). For simplicity, we call this number \( M_R(\eta, z) \). In general, under a natural coupling, for \( z \in \mathbb{B}(0, R) \cup \partial \mathbb{B}(0, R) \),

\[ W_R(\Lambda, \eta, z) \leq M_R(\eta, z). \quad (2.3) \]

### 2.1 The abelian property

Diaconis and Fulton [3] allow explorers to start on distinct sites, and show that the law of the cluster is invariant under permutation of the order in which explorers are launched. This invariance is named the abelian property. As a consequence, one can realize the cluster by sending many exploration waves. Let us illustrate this observation by building \( A(\emptyset, (n+m)\delta_0) \) in three waves, since we need later this very example. The first wave consists in launching \( n \) explorers, and they settle in \( A_1 = A(\emptyset, n\delta_0) \), which we call the cluster after the first wave. Then, we launch \( m \) explorers that we color green for simplicity, and if they reach \( \partial \mathbb{B}(0, R) \) before settling, then we stop them on \( \partial \mathbb{B}(0, R) \). The settled green explorers make up the cluster \( A_R(A_1; m\delta_0) \). The cluster after the second wave is then

\[ A_2 = A_1 \cup A_R(A_1; m\delta_0). \]

For \( z \in \partial \mathbb{B}(0, R) \), we call \( \zeta_R \) the configuration of the green explorers stopped on \( \partial \mathbb{B}(0, R) \). In other words,

\[ \zeta_R = \{ W_R(A_1; m\delta_0, z), \ z \in \partial \mathbb{B}(0, R) \}. \]

Then, the cluster after the third wave is obtain as we launch the stopped green explorers, and

\[ A_3 = A_2 \cup A(A_2; \zeta_R). \]

The abelian properties implies that \( A(\emptyset, (n+m)\delta_0) \) equals in law to \( A_3 \). It is convenient to think of the growing cluster as evolving in discrete time, where time counts the number of exploration waves.

### 2.2 On the harmonic measure.

We first recall a well known property of Poisson variables.
Lemma 2.1 Let \( \{U_n, n \in \mathbb{N}\} \) be an i.i.d. sequence with value in a set \( E \), and \( \{E_1, \ldots, E_n\} \) a partition of \( E \). Then, if \( X \) is an independent Poisson random variable of parameter \( \lambda \), and if
\[
X_i = \sum_{n \leq X} \mathbb{1}_{U_n \in E_i},
\]
then \( \{X_i, i = 1, \ldots, n\} \) are independent Poisson variables with \( E[X_i] = \lambda \times P(U \in E_i) \).

Now, for \( z \in \mathbb{Z}^d \setminus \{0\} \), let
\[
\Sigma(z) = \partial \mathbb{B}(0, ||z||),
\]
and note that \( z \) belongs to \( \Sigma(z) \) since there is \( z' \in \mathbb{B}(0, ||z||) \) and \( ||z - z'|| = 1 \) (see Lemma 2.1 of [1]). When we start independent random walks on \( \eta \), and for \( z \neq 0, h > 0 \), and \( \Lambda \subset \Sigma(z) \), we denote by \( N_z(\eta, \Lambda, h) \) the number of walks initially on \( \eta \) which hit \( \Sigma(z) \) on \( \Lambda \), and then visit \( z \) before exiting \( \mathbb{B}(0, ||z|| + h) \). As a consequence of Lemma 2.1 we have

Corollary 2.2 Let \( X \) be a Poisson variable of parameter \( \lambda \), and \( \eta = X \delta_0 \). For any \( z \neq 0, h, h' > 0 \), and \( \Lambda, \Lambda' \subset \Sigma(z) \) with \( \Lambda \cap \Lambda' = \emptyset \), we have that \( N_z(\eta, \Lambda, h) \) and \( N_z(\eta, \Lambda', h') \) are independent Poisson variables, and
\[
E[N_z(\eta, \Lambda, h)] = \lambda \times \mathbb{P}_0 \left( S(H(\Sigma(z))) \in \Lambda, H(z) < H(\mathbb{B}(0, ||z|| + h)) \right)
\]
\[
= \lambda \sum_{y \in \Lambda} \mathbb{P}_0 \left( S(H(\Sigma(z)) = y) \right) \mathbb{P}_y \left( H(z) < H(\mathbb{B}(0, ||z|| + h)) \right).\tag{2.5}
\]

By combining well known asymptotics of the harmonic measure with Corollary 2.2 we obtain the following lemma.

Lemma 2.3 Assume that dimension is two or more. Let \( \eta \) be as in Corollary 2.2. For \( z \neq 0 \), and \( R > 0 \) let \( \Lambda = \mathbb{B}(z, R) \cap \Sigma(z) \), and \( \Lambda' = \Sigma(z) \setminus \Lambda \). There is \( \kappa > 0 \), independent of \( z \), and \( R \) such that
\[
P(N_z(\eta, \Lambda, \infty) = 0, N_z(\eta, \Lambda', R) = 0) \leq \exp \left( -\kappa \frac{\lambda R}{||z||^{d-1}} \right).\tag{2.6}
\]

Proof. Since \( N_z(\eta, \Lambda, \infty) \) and \( N_z(\eta, \Lambda', R) \) are independent Poisson variables, we have
\[
P(N_z(\eta, \Lambda, \infty) = 0, N_z(\eta, \Lambda', R) = 0) = \exp \left( -E[N_z(\eta, \Lambda, \infty)] - E[N_z(\eta, \Lambda', R)] \right). \tag{2.7}
\]

It remains to compute expected values. To estimate \( E[N_z(\eta, \Lambda, \infty)] \), we recall that there is a constant \( c_d \) such that
\[
\mathbb{P}_y(H(z) < \infty) \leq \frac{c_d}{1 + ||y - z||^{d-2}}.\tag{2.8}
\]

Using (2.5), there is a constant \( c \)
\[
E[N_z(\eta, \Lambda, \infty)] \leq \sum_{y \in \Lambda} \lambda \mathbb{P}_0 \left( S(H(\Sigma(z)) = y) \right) \frac{c_d}{1 + ||y - z||^{d-2}}
\]
\[
\leq c \frac{\lambda}{2 ||z||^{d-1}} \left( 1 + \sum_{k=1}^{R} \frac{k^{d-2}}{1 + k^{d-2}} \right) \leq c \frac{\lambda R}{||z||^{d-1}}.\tag{2.9}
\]
Now, to estimate $E [N_\gamma (\eta, \eta', R)]$, we recall Lemma 5(b) of [4], which states that for some constant $c'_d$, for $y \in \Sigma(z)$

$$P_y (H(z) < H(B^c(0, \|z\| + R))) \leq \frac{c'_d R^2}{\|z - y\|^d}$$  \hspace{1cm} (2.10)

Using (2.5) and (2.10), there is a constant $c'$ such that

$$E [N_\gamma (\eta, \eta', R)] \leq \sum_{y \in \Sigma(z) \setminus B(z, R)} \lambda \mathbb{P}_y \left(S(H(\Sigma(z))) = y\right) \frac{c'_d R^2}{\|z - y\|^d}$$

$$\leq \frac{c'}{2} \frac{\lambda R^2}{\|z\|^d} \sum_{k=1}^{2\|z\| + 2} \frac{k^{d-2}}{k^d} \leq \frac{c'}{2} \frac{\lambda R}{\|z\|^{d-1}}$$  \hspace{1cm} (2.11)

Combining (2.9), and (2.11), we obtain the desired result.

\section{The outer error}

In this section, we prove Proposition 1.2. Let us explain the proof in dimension three or more, and explain in Remark 3.1 of Step 2 how we adapt the proof to dimension two.

For positive reals $\alpha$ and $\gamma$, to be chosen later, we set $h(n) = \alpha \sqrt{\log(n)}$ and $L(n) = \gamma \sqrt{\log(n)}$ for estimates on the outer and inner fluctuations. Even though we eventually take $h(n) = L(n)$, it is useful to keep in mind their distinct nature. The limit (1.9) follows if for some small $\gamma = \alpha$, we have

$$\lim_{n \to \infty} P \left( \exists k \geq n, \delta_O(k) \geq h(k) \mid \delta_I(n) < L(n) \right) = 1. \hspace{1cm} (3.1)$$

Indeed,

$$P \left( \exists k \geq n, \delta_O(k) \geq h(k) \text{ or } \delta_I(n) \geq L(n) \right)$$

$$\geq P \left( \delta_I(n) \geq L(n), \text{ or } \exists k \geq n, \delta_O(k) \geq h(k) \right)$$

$$\geq P \left( \delta_I(n) \geq L(n) \right) + P \left( \exists k \geq n, \delta_O(k) \geq h(k) \mid \delta_I(n) < L(n) \right) P \left( \delta_I(n) < L(n) \right)$$

$$\geq 1 - P \left( \exists k \geq n, \delta_O(k) \geq h(k) \mid \delta_I(n) < L(n) \right) P \left( \delta_I(n) < L(n) \right). \hspace{1cm} (3.2)$$

We now prove (3.1). For integer $n$, assume that $A(b(n)\delta_0)$ is realized. Let $X_n$ be a Poisson random variable with parameter $\lambda_n = |B(0, n + h(n)) \setminus B(0, n)|$. We realize the cluster $A((b(n) + X_n)\delta_0)$ through three exploration waves, as explained in Section 2.1. After the first wave with $b(n)$ explorers, we launch $X_n$ explorers, the green ones, and stop them on $\Sigma := \partial B(0, n - L(n))$. Under the event $\{\delta_I(n) < L(n)\}$, note that for $z \in \Sigma$

$$W_{n-L(n)}(A(b(n)\delta_0); X_n\delta_0, z) = M_{n-L(n)}(X_n\delta_0, z).$$
The configuration of stopped random walks on $\Sigma$ is denoted $\zeta$. Note that $\zeta$ is independent of $A(b(n)\delta_0)$. For $z \in \Sigma$, we call $\text{cov}(z)$ the event that the $\zeta(z)$ green explorers starting on $z$ produce a cluster $A(\emptyset, \zeta(z)\delta_z)$ which satisfies
\[ A(\emptyset, \zeta(z)\delta_z) \cap \mathbb{B}^c(0, n + 4h(n)) \neq \emptyset. \] (3.3)

Note that the green explorers contributing to $\text{cov}(z)$ start on the positions of the random walks stopped on $z \in \Sigma$, which are associated with the green explorers. Assume, for a moment that when $\text{cov}(z)$ happens, there is a tentacle of $A((b(n) + X_n)\delta_0)$ which protrudes $\mathbb{B}(0, n + 4h(n))$. Assume also that under condition on $X_n$, we have
\[ n + 4h(n) \geq R_n + h(R_n), \quad \text{where} \quad R_n = \rho(b(n) + X_n). \] (3.4)

We would deduce that $\delta_O(R_n) \geq h(R_n)$.

We now proceed through four steps. First, we show that (3.3) implies that the final cluster is not inside $\mathbb{B}(0, n + 4h(n))$. Secondly, we estimate the cost of producing a tentacle realizing $\text{cov}(z)$. Then, we establish conditions ensuring (3.4). Finally, we show that for an appropriate choice of $\alpha$, one event $\text{cov}(z)$ realizes for some $z \in \Sigma$.

**Step 1: Coupling.** By coupling, it is easy to see that for any subset $\Lambda$, and $z \in \Sigma$
\[ A(\emptyset; \zeta(z)\delta_z) \subset \Lambda \cup A(\Lambda; \zeta(z)\delta_z) \subset \Lambda \cup A(\Lambda; \zeta). \] (3.5)

If we denote by $A_2$ the cluster after the second exploration wave (see Section 2.1), then we have with an equality in law
\[ A_2 \cup A(A_2; \zeta) = A((b(n) + X_n)\delta_0). \] (3.6)

Now, using (3.5), (3.6), and (3.3), we conclude that the final cluster is not in $\mathbb{B}(0, n + 4h(n))$.

**Step 2: Long tentacles.** To produce $\text{cov}(z)$, we first bring a number of green explorers at $z$ proportional to $h(n)$, and force them to make a tentacle normal to $\Sigma$ at $z$, with a height $4h(n) + \underline{L}(n)$. More precisely, draw unit cubes centered on the points of the sequence
\[ x_n = \left(\|z\| + n\right) \frac{z}{\|z\|} \in \mathbb{R}^d \quad \text{(and} \quad \|x_n\| = \|z\| + n). \]

Each such cube contains at least a site of $\mathbb{Z}^d$, say $z_n$. Note that $\|z_n - z_{n-1}\| \leq 2\sqrt{d} + 1$, so that we can exhibit a sequence $\{z = y_1, y_2, \ldots, y_N\}$ of nearest neighbors in $\mathbb{Z}^d$ such that $\|y_N - z\| \geq 4h(n) + \underline{L}(n)$, with $N \leq c(4h(n) + \underline{L}(n))$ for some constant $c$ independent of $n$. Now, if $M_{n-L(n)}(X_n\delta_0, z) \geq N$, and if we launch the green explorers stopped on $z$ and force them to walk along the sequence $\{y_1, y_2, \ldots, y_N\}$, with the $k$-th explorer settling on $y_k$, then we realize $\text{cov}(z)$, and its probability is larger than
\[ \left(\frac{1}{2d}\right)^{\sum_{k=1}^{N} k} \geq \exp \left(-c(4h(n) + \underline{L}(n))^2\right). \]
Remark 3.1 In dimension 2, there is a better strategy to build a tentacle. Since we believe that it yields an estimate which is not optimal, we do not give the full proof, but give enough details of the construction so that the interested reader can easily fill the details.

We first bring a larger number of green explorers at \( z \), about \( \frac{h(n)}{\log^2(h(n))} \times h(n) \). The probability of so doing is larger than

\[
\exp \left(-C \frac{h(n)}{\log^2(h(n))} \log \left( \frac{h(n)}{\log^2(h(n))} \right) \times h(n) \right) \geq \exp \left(-c \frac{h^2(n)}{\log(h(n))} \right),
\]

for two positive constants \( C, c \). Then, the explorers are forced to fill sequentially cylindrical compartments of a telescope-like domain that we now describe. Let \( R \) be the integer part of \( 4h(n) + L(n) \), and divide \( B(z, R) \) into \( R \) shells of length \( h_1, \ldots, h_R \) with for \( i = 1, \ldots, R \)

\[
h_i = \frac{R}{i \times \log(R)}.
\]

Choose the sequence of \( R \) points of \( \mathbb{R}^d \)

\[
\forall i \in \{1, \ldots, R\} \quad x_i = (\|z\| + h_i) \frac{z}{\|z\|}.
\]

There is \( z_i \in \partial B(z, h_i) \) with \( \|z_i - x_i\| \leq 2 \). Now, assume we have brought \( n_i := A\pi h_i^2 + 1 \) (for a fixed constant \( A \)) explorers in \( B(z_i, h_i/4) \). Then, with a probability larger than \( 1 - 1/h_i^2 \), they cover the ball \( B(z_i, h_{i+1}) \) by Lemma 1.3 of \([2]\). Now, if we bring additional explorers in \( B(z_i, h_i/4) \), they can reach \( B(z_{i+1}, h_{i+1}/4) \) with a positive probability, say \( \exp(-\kappa) \): indeed, they only need to escape a cube-like domain centered on \( z_i \), of side-length \( h_{i+1}/4 \) on the side which contains \( z_{i+1} \). The cost of the scenario, for which we only described the \( i \)-th step, is therefore of order

\[
\prod_{i=1}^{R-1} \left( 1 - \frac{1}{h_i^2} \right) \times e^{-\kappa \sum_{i=2}^{R} n_i} \times e^{-\kappa \sum_{i=3}^{R} n_i} \cdots \times e^{-\kappa n_R} \geq C \exp \left(-\kappa \sum_{i=2}^{R} (i-1)n_i \right),
\]

for positive constants \( \kappa, C \). Note that with the choice of \( h_i \) in \((3.8)\), and \( n_i = A\pi h_i^2 + 1 \), we have with \( L(n) = h(n) \) and for a positive constant \( \kappa' \)

\[
\sum_{i=2}^{R} (i-1)n_i \geq \kappa' \frac{h^2(n)}{\log(h(n))}.
\]

Thus, the probability of bringing \( h^2(n)/\log^2(n) \) explorers in \( z \), and the probability of building a tentacles of height \( 5h(n) \) is larger than

\[
\exp \left(-\kappa_2 \frac{h^2(n)}{\log(h(n))} \right),
\]

for some positive constant \( \kappa_2 \).
Step 3: Bounding $X_n$. We impose that $X_n \leq 2\lambda_n$.

$$X_n \leq 2\left| \mathcal{B}(0, n + h(n)) \setminus \mathcal{B}(0, n) \right| \leq \left| \mathcal{B}(0, n + 2h(n)) \setminus \mathcal{B}(0, n) \right| \implies X_n + b(n) \leq \left| \mathcal{B}(0, n + 2h(n)) \right| \implies R_n \leq n + 2h(n). \tag{3.11}$$

The conclusion of (3.11) implies also that $h(R_n) \leq 2h(n)$, and this implies (3.4). Note that since $X_n$ is a Poisson variable of mean $\lambda_n$, there is a constant $c$, such that

$$P(X_n > 2\lambda_n) \leq \exp \left(-ch(n)n^{d-1}\right). \tag{3.12}$$

Step 4: Many possible tentacles. We summarize Step 1 to Step 3, as establishing that

$$\{X_n \leq 2\lambda_n, \exists z \in \Sigma \text{ cov}(z), \delta_I(n) < L(n)\} \subset \{\delta_O(R_n) \geq h(R_n), \delta_I(n) < L(n)\} \subset \{\exists k > n, \delta_O(k) \geq h(k), \delta_I(n) < L(n)\}. \tag{3.13}$$

The key observation now is that $\{\text{cov}(z), z \in \Sigma\}$ are independent Poisson variables which are also independent of $\delta_I(n)$. Indeed, $\{\text{cov}(z), z \in \Sigma\}$ deals with the green explorers, whereas $\delta_I(n)$ depends on the $b(n)$ explorers which we launch first. Taking probability on both sides of (3.13), and dividing by $P(\delta_I(n) < L(n))$, we obtain

$$P(\exists k > n, \delta_O(k) \geq h(k)|\delta_I(n) < L(n)) \geq P(X_n \leq 2\lambda_n, \exists z \in \Sigma \text{ cov}(z)). \tag{3.14}$$

Thus,

$$P(\exists k > n, \delta_O(k) \geq h(k)|\delta_I(n) < L(n)) \geq P(\exists z \in \Sigma \text{cov}(z)) - P(X_n > 2\lambda_n). \tag{3.15}$$

Now, note that for some positive constants $\kappa, \kappa'$

$$P \left( \bigcup_{z \in \Sigma} \text{cov}(z) \right) = 1 - \prod_{z \in \partial B} \left( 1 - P(\text{cov}(z) \cap \{\zeta(z) > c(4h(n) + L(n))\}) \right) \geq 1 - \exp \left(-\kappa'n^{d-1} \exp \left(-\kappa(h^2(n) + L^2(n))\right)\right). \tag{3.16}$$

Now, there is $\alpha > 0$ such that for $h(n) \leq \alpha \sqrt{\log(n)}$, and $L(n) = h(n)$

$$\lim_{n \to \infty} n^{d-1} \exp \left(-\kappa(h^2(n) + h(n))\right) = \infty. \tag{3.17}$$

We now combine (3.15), (3.13), (3.16) and (3.17) to conclude the proof. In dimension 2, the estimate (3.16) has to be replaced with (3.10).

4 The inner error

In this section, we prove Theorem 1.4. We show that in the process of going from a cluster of volume $b(n)$ to one of volume $b(2n)$, chances tend to one as $n$ tends to infinity, that
there appears a cluster $A$ whose inner error is larger than $\alpha \sqrt{\log(\rho(A))}$ for some positive $\alpha$ independent of $n$. To do so, we launch many exploration waves, each one is made up of a Poisson number of explorers.

We proceed inductively. For positive reals $\alpha, \beta$, to be chosen later, we set $h(n) = \alpha \sqrt{\log(n)}$, and $\bar{L}(n) = \beta \sqrt{\log(n)}$. $\bar{L}(n)$ will refer to an outer radius. Let $\{G_n, n \geq 0\}$ denotes the natural filtration associated with the evolution by waves.

First, we launch $b(n)$ explorers. Assume that explorers of wave $k-1$ have been launched, and are settled. Knowing $G_{k-1}$, the size of the $k$-th wave, denoted $X_k$, is a Poisson variable of parameter

$$\lambda(k) = |\mathbb{B}(0, R_{k-1} + 2h(R_{k-1})) \setminus \mathbb{B}(0, R_{k-1})|, \quad \text{where} \quad R_{k-1} = \rho(b(n) + X_1 + \cdots + X_{k-1}). \quad (4.1)$$

Since $R_{k-1}$ is of order $n$, each wave fills approximately a peel of width $2h(n)$, and $n/2h(n)$ waves fill approximately $\mathbb{B}(0, 2n)$. We prove in this section that for an appropriate $\alpha$

$$\lim_{n \to \infty} P \left( \bigcup_{1 \leq k < n/2h(n)} \{ \delta_I(R_k) > \alpha \sqrt{\log(R_k)} \} \right) = 1. \quad (4.2)$$

We now proceed in estimating the probability of observing a deep hole after each exploration waves. We set $A_k = \{ \delta_I(R_k) > \alpha \sqrt{\log(R_k)} \}$.

**On the holes left after wave $k-1$.** Observe that by definition, on $A_{k-1}^c$

$$\mathbb{B}(0, R_{k-1} - h(R_{k-1})) \subset A(b(n) + X_1 + \cdots + X_{k-1}),$$

which implies that

$$\left( \mathbb{B}(0, R_{k-1}) \setminus \mathbb{B}(0, R_{k-1} - h(R_{k-1})) \cup \partial \mathbb{B}(0, R_{k-1}) \right) \cap A(b(n) + X_1 + \cdots + X_{k-1}) \neq \emptyset. \quad (4.3)$$

Choose any $Z_k$ in the intersection of the non-empty set of (4.3), and note that

$$R_{k-1} - h(R_{k-1}) \leq \|Z_k\| \leq R_{k-1} + 1. \quad (4.4)$$

Recall that we have defined $\Sigma(Z_k) := \partial \mathbb{B}(0, \|Z_k\|)$. We launch the $X_k$ explorers, that we name the *green explorers*, and we stop them as they reach $\Sigma(Z_k)$. The green explorers which settle before reaching $\Sigma(Z_k)$ play no role here, and we bound the number of green explorers stopped on some region $\Lambda \in \Sigma(Z_k)$, by the number of corresponding random walks exiting $\Sigma(Z_k)$ on $\Lambda$. Thus, if we choose

$$\Lambda_k = \mathbb{B}(Z_k, \bar{L}(R_k)) \cap \Sigma(Z_k), \quad \text{and} \quad \Lambda'_k = \Sigma(Z_k) \setminus \Lambda_k, \quad (4.5)$$

and if we denote

$$I_k = \{ N_{Z_k}(X_k \delta_0, \Lambda_k, \infty) = 0, \ N_{Z_k}(X_k \delta_0, \Lambda'_k, 7\bar{L}(R_{k-1})) = 0 \} \quad (4.6)$$
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then, on the event $I_k$, green explorers either exit a ball of radius $R_{k-1} - h(R_{k-1}) + 7\bar{L}(R_{k-1})$, or do not visit $Z_k$. In other words,

$$I_k \subset \{ R_k - \delta(I)(R_k) < \|Z_k\| \} \cup \{ R_k + \delta(O)(R_k) \geq \|Z_k\| + 7\bar{L}(R_{k-1}) \}.$$ 

In order to conclude that $\{ \delta(I)(R_k) \geq h(R_k) \}$ or $\{ \delta(O)(R_k) \geq \bar{L}(R_k) \}$, we need to find conditions on $X_k$ that guarantee that

$$R_k - \|Z_k\| \geq h(R_k), \text{ and } \|Z_k\| - R_k + 7\bar{L}(R_{k-1}) \geq \bar{L}(R_k). \quad (4.7)$$

**Conditions on $X_k$ fulfilling (4.7).** We call

$$C_k = \{ 2\lambda(k) \geq X_k \} \cap \left\{ X_k \geq \frac{2}{3}\lambda(k) \right\}.$$ 

On the one hand, if $X_k \geq \frac{2}{3}\lambda(k)$, then

$$X_k \geq \frac{2}{3}\left| \mathbb{B}(0, R_{k-1} + 2h(R_{k-1})) \setminus \mathbb{B}(0, R_{k-1}) \right|$$

$$\geq \left| \mathbb{B}(0, R_{k-1} + \frac{4}{3}h(R_{k-1})) \setminus \mathbb{B}(0, R_{k-1}) \right| \implies R_k \geq R_{k-1} + \frac{4}{3}h(R_{k-1}). \quad (4.8)$$

On the other hand, if $X_k \leq 2\lambda_k$, then

$$X_k \leq 2\left| \mathbb{B}(0, R_{k-1} + 2h(R_{k-1})) \setminus \mathbb{B}(0, R_{k-1}) \right|$$

$$\leq \left| \mathbb{B}(0, R_{k-1} + 4h(R_{k-1})) \setminus \mathbb{B}(0, R_{k-1}) \right| \implies R_k \leq R_{k-1} + 4h(R_{k-1}). \quad (4.9)$$

Now, for $x$ large enough, the following implication is obvious

$$x \leq y + 4h(y) \implies h(x) \leq \frac{4}{3}h(y) - 1, \text{ and } \bar{L}(x) \leq 2\bar{L}(y). \quad (4.10)$$

If $n$ is large enough, (4.10) and (4.9) imply that $h(R_k) \leq \frac{4}{3}h(R_{k-1}) - 1$, which in turn, with (4.8), yields

$$R_k \geq R_{k-1} - 1 + h(R_k). \quad (4.11)$$

Also, $\bar{L}(R_k) \leq 2\bar{L}(R_{k-1})$, $\bar{L}(R_{k-1}) \geq h(R_{k-1})$, in combination with (4.9), imply that

$$R_{k-1} - h(R_{k-1}) + 7\bar{L}(R_{k-1}) - \bar{L}(R_k) \geq R_{k-1} + 4h(R_{k-1}) \geq R_k. \quad (4.12)$$

Thus, if $C_k \cap A_{k-1}$ holds, then (4.11), (4.11) and (4.12) imply that conditions (4.7) holds.

**On a deep hole in one shell.** We choose an integer $k < n/2h(n)$. We have seen that knowing $A_{k-1}$

$$I_k \cap C_k \subset A_k \cup \{ \delta_0(R_k) \geq \bar{L}(R_k) \}. \quad (4.13)$$

Taking conditional probabilities on both sides of (4.13), we obtain,

$$\mathbb{I}_{A_{k-1}} \mathbb{P}\left( A_k \cap C_k \mid G_{k-1} \right) = \mathbb{I}_{A_{k-1}} \left( \mathbb{P}(C_k) - \mathbb{P}(A_k \cap C_k \mid G_{k-1}) \right)$$

$$\leq \mathbb{I}_{A_{k-1}} \left( \mathbb{P}(C_k) - \mathbb{P}(I_k \cap C_k \mid G_{k-1}) + \mathbb{P}\left( \{ \delta_0(R_k) \geq \bar{L}(R_k) \} \mid G_{k-1} \right) \right)$$

$$\leq \mathbb{P}\left( \{ \delta_0(R_k) \geq \bar{L}(R_k) \} \mid G_{k-1} \right) + \mathbb{I}_{A_{k-1}} (1 - \mathbb{P}(I_k \mid G_{k-1})). \quad (4.14)$$
Now, we invoke Lemma 2.3 with \( \|z\|, R \) and \( \lambda \) respectively of order \( n, \bar{L}(n) \), and \( h(n)n^{d-1} \). As a consequence, we have on \( C_k \cap A_k^c \) for a constant \( \kappa \),

\[
\inf_{k \leq n/2h(n)} P(I_k \mid G_{k-1}) \geq \exp(-\kappa h(n)\bar{L}(n)). \tag{4.15}
\]

If we denote \( N \) for the integer part of \( n/2h(n) \), and proceed inductively, we obtain

\[
P(\bigcup_{k \leq N} A_k) - P(\bigcap_{k \leq N} C_k) \geq -P(\forall k \leq N, A_k^c \cap C_k)
\geq -E \left[ \mathbb{1}_{\forall k < N, A_k^c \cap C_k} P(A_k^c \cap C_k \mid G_{N-1}) \right]
\geq -E \left[ \mathbb{1}_{\forall k < N, A_k^c \cap C_k} P(\{\delta_0(R_N) \geq \bar{L}(R_N)\} \mid G_{N-1}) + 1 - P(I_N \mid G_{N-1}) \right]
\geq -P(\{\delta_0(R_N) \geq \bar{L}(R_N)\}) - (1 - \exp(-\kappa h(n)\bar{L}(n)))P(\forall k < N, A_k^c \cap C_k)
\geq - \sum_{k \leq N} P(\{\delta_0(R_k) \geq \bar{L}(R_k)\}) - (1 - \exp(-\kappa h(n)\bar{L}(n)))^N.
\]

Thus,

\[
P(\bigcup_{k \leq N} A_k) \geq 1 - \sum_{k \leq N} (P(\{\delta_0(R_k) \geq \bar{L}(R_k)\}) + P(C_k^c)) - (1 - \exp(-\kappa h(n)\bar{L}(n)))^N. \tag{4.16}
\]

Now, we have established in [2], that for \( \beta \) large enough, the probability of \( \{\delta_0(R_k) \geq \bar{L}(R_k)\} \) decays faster than any power in \( n \), whereas the fact that \( X_k \) is Poisson implies that for some constant \( c \), we have \( P(C_k^c) \leq \exp(-c h(n)n^{d-1}) \). The last term on the last display of (4.16) tends to 0 if

\[
\lim_{n \to \infty} \frac{n}{2h(n)} \exp(-\kappa h(n)\bar{L}(n)) = \infty. \tag{4.17}
\]

In dimension 3 or more, (4.17) holds for \( \alpha \) small enough.

## 5 Proof of Proposition 1.6

The proof is a direct corollary of formula (3.11) of [2]. We consider actually tiles of size 1, that is site of \( \mathbb{Z}^d \). Inequality (3.8) of [2] shows that for some constant \( c_d \) (depending only on dimension) and \( R = \|z\| < n \), we have

\[
E[W_R(\emptyset, N\delta_0, z)] \geq c_d(n - \|z\|). \tag{5.1}
\]

Inequality (3.10) of [2] is written a little differently as

\[
P(W_R(\emptyset, N\delta_0, z) = 0) \leq \begin{cases} \exp(-\lambda \kappa_2(n - \|z\|) + \lambda^2 c_d \log(n)) & \text{for } d = 2, \\ \exp(-\lambda \kappa_d(n - \|z\|) + \lambda^2 c_d^2) & \text{for } d \geq 3. \end{cases} \tag{5.2}
\]

As we optimize (5.2) in \( \lambda > 0 \), we obtain (1.11).
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