Internal Diffusion Limited Aggregation on discrete groups having exponential growth

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Abstract

The Internal Diffusion Limited Aggregation has been introduced by Diaconis and Fulton in 1991. It is a growth model defined on an infinite set and associated to a Markov chain on this set. We focus here on sets which are finitely generated groups with exponential growth. We prove a shape theorem for the Internal DLA on such groups associated to symmetric random walks. For that purpose, we introduce a new distance associated to the Green function, which happens to have some interesting properties. In the case of homogeneous trees, we also get the right order for the fluctuations of that model around its limiting shape.

1 Introduction

Let \mathcal{G} be an infinite discrete set and let P be a irreducible transition matrix on that set. The matrix P allows to define a sequence of i.i.d. Markov chains $(S^n(\cdot))_{n\in\mathbb{N}^*}$ with a common starting point O(S(0) = O). The Internal Diffusion Limited Aggregation (Internal DLA) is a Markov chain A(n) $(n \in \mathbb{N})$ of increasing subsets of \mathcal{G} defined as follows:

- $A(1) = \{O\},\$
- $A(n+1) = A(n) \cup S^{n+1}(\tau_{A(n)^c}),$

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where τ_A is the hitting time of the set A.

Thus, the set A(n) grows as follows. At each discrete time n, we start the Markov chain $S^n(\cdot)$ (moving according to P) at O, wait until it leaves the previous set A(n-1), and add the first element visited outside A(n-1), to obtain A(n). Note that the Markov chain A(n) is well defined since, for all n, $\tau_{A(n)^c} < \infty$ almost surely, as A(n) is finite and $S^{n+1}(\cdot)$ is irreducible.

The name Internal DLA comes from the similitude with the (external) DLA. In the latter model, the Markov chains start "at infinity", conditioned to hit the previous cluster, and are stuck just before hitting it. See for example Lawler [9] for a precise definition and some properties.

The Internal DLA was introduced by Diaconis and Fulton in 1991 [6]. They gave the first shape theorem on \mathbb{Z} where $S^n(\cdot)$ are Simple Random Walks (nearest neighbours random walks with uniform transition probability). Later, other shape theorems were proved on \mathbb{Z}^d (Lawler, Bramson & Griffeath [9], Lawler [10], Blachère [1]). These results set that, when the random walk is centered (under some moment condition), the model has a limiting shape (see one of these for precise definition). Roughly, that means that the shape of A(n) tends to the shape of the balls of \mathbb{R}^d for a norm associated to the transition matrix P.

In the present article, we focus on groups Γ having exponential growth. It means that the balls (associated to the word metric for any finite generating set) have cardinality growing exponentially in the radius. On such groups, we consider symmetric random walks with finitely supported increment. Our main result (Theorem 3.1) is a shape theorem for the Internal DLA. The meaning of a shape theorem is to compare the random shape of the Internal DLA with a deterministic exhaustion of the group. For that purpose, we define a new distance on groups, called the hitting distance, by

$$d(x,y) \stackrel{\text{def.}}{=} -\ln(F(x,y))$$
,

where F(x,y) is the hitting probability of y for a random walk starting at x. As the Green function $G(\cdot,\cdot)$ is proportional to the function $F(\cdot,\cdot)$, the balls for that distance are in fact the sets of the form,

$$\{x \text{ s.t. } G(e, x) \ge N\},\$$

where e is the identity of the group. These sets exhaust Γ when N goes to 0.

This distance can be compared to the word distance but they are not equivalent in general. Outside its link with the Internal DLA, an interesting property of that distance is that one can deduce the exact exponential growth rate of the balls (Proposition 2.3).

The shape theorem states that the Internal DLA is close to the balls B(n) of radius Kn for the hitting distance (K is a constant that ensures that the ball B(n) contains the boundary of the ball B(n-1)). More precisely there exists some

constants C_I and C_O such that at time V(n) (=volume of B(n)) the sets of the IDLA satisfy:

$$B(n - C_I \ln n) \subset A(V(n)) \subset B(n + C_O \sqrt{n})$$

for every large n, with probability 1. This result gives the upper bounds for the fluctuations between A(V(n)) and B(n): the inner fluctuations are at most logarithmic, whence the outer fluctuations are at most of order square root of the radius. The proof of that result is adapted from the one on \mathbb{Z}^d (Lawler, Bramson & Griffeath [11] and Lawler [10]) with new ingredients related to the hitting distance and some simplifications, partially due to the setting of groups with exponential growth.

For the Simple Random Walk of homogeneous trees (where the hitting distance and the word distance are proportional) such a result was already obtained by Etheridge and Lawler (unpublished). In that setting, they also announced (in [9]) that the lower bounds for the inner and outer fluctuations were of the same order than the upper ones, but according to Lawler, a written proof of that result was missing and the lower bound for the inner error could be considered an open question. Therefore, we prove (Propositions 4.1 and 4.2) that in the case of a homogeneous tree the order of the inner fluctuation cannot be smaller of $\ln n$, while the outer fluctuation is bigger than some constant times \sqrt{n} .

The article is constructed as follows. Section 2 defines the setting of symmetric random walks on groups with exponential growth, defines the "hitting distance" and compares it to the word distance. Section 3 gives the proof of our main result, the shape theorem for the Internal DLA with upper bounds for the fluctuations. Section 4 focus on Simple Random Walk on homogeneous trees and proves the sharpness of the bounds for these examples.

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2 Symmetric random walk on discrete groups with exponential growth

Let Γ be a finitely generated group. Let \mathcal{S} be a finite symmetric set of generators. We write each element x of Γ as a word composed by these generators (letters). The minimal length among the words representing x is called the word distance |x|. Let e be the identity of Γ . According to this distance, we define the balls $B_w(n) = B_w(e, n) \stackrel{\text{def.}}{=} \{x \in \Gamma : |x| \leq n\}$ and their volume $V_w(n) \stackrel{\text{def.}}{=} \#B_w(n)$. The exponential growth property corresponds to

$$V_w(n) \ge \exp(c n)$$
,

for some constant c. Note that an exponential upper bound of the volume is always valid since the fastest volume growth occurs when there is no relation between the generators. Hence, $V_w(n) \leq 1 + (\#\mathcal{S} - 1)^n$. Note also that this property does not depend on the choice of the generating set.

Let μ be a symmetric probability law whose support is \mathcal{S} . Let $(X_k)_{k\in\mathbb{N}^*}$ be a sequence of i.i.d. random variables whose common law is μ . The process

$$S(k) \stackrel{\text{def.}}{=} xX_1X_2\cdots X_k,$$

with S(0) = x, is a symmetric irreducible random walk on Γ starting at x. The process $S(\cdot)$ is then a nearest neighbors symmetric random walk on the Cayley graph $\mathcal{G} = (\Gamma, \mathcal{S})$ on which we study the Internal DLA (with $A(0) = \{e\}$). We denote \mathbb{P}^x and \mathbb{E}^x , respectively, the probability and expectation related to a random walk starting at x. When x = e, the exponent will be omitted.

The Green function G(x, y) is defined as the expected number of visits at y for a random walk starting at x:

$$G(x,y) \stackrel{\text{def.}}{=} \mathbb{E}^x \left[\sum_{k=0}^{\infty} \mathbb{I}_{\{S(k)=y\}} \right] = \sum_{k=0}^{\infty} \mathbb{P}^x [S(k)=y].$$

Since the group has an exponential growth, every random walk is transient and so the above summation is finite (G(x, y)) is actually bounded uniformly in x and y).

For a random walk $S(\cdot)$, let τ_y be the first hitting time of y:

$$\tau_y \stackrel{\text{def.}}{=} \inf\{k \ge 0 : S(k) = y\}.$$

When y is never attained, let $\tau_y = \infty$. The hitting probability of y starting at x is denoted

$$F(x,y) \stackrel{\text{def.}}{=} \mathbb{P}^x[\tau_y < \infty]$$
.

We write G(x) = G(e, x) and F(x) = F(e, x). Note that F and G are symmetric and invariant by left multiplication. In particular, G(y, y) = G(e). Therefore, the functions F and G are proportional since a straightforward computation shows

$$G(x,y) = G(y,y)F(x,y) = G(e)F(x,y)$$
. (1)

We are now ready to define the distance we will use to study the Internal DLA. For all $x, y \in \Gamma$, we define

$$d(x,y) \stackrel{\text{def.}}{=} -\ln F(x,y)$$
.

Since F(x, y) is the hitting probability of y starting at x, we call $d(\cdot, \cdot)$ the **hitting** distance.

Lemma 2.1 $d(\cdot, \cdot)$ is a left invariant distance on Γ .

Proof

As $F(x,y) \leq 1$, $d(\cdot,\cdot)$ is always non-negative.

The symmetry of the random walk and the left invariance of $F(\cdot, \cdot)$ yield the same properties for $d(\cdot, \cdot)$.

Remark also that since the random walk is symmetric and transient

$$\forall x \neq y, \ 1 > \mathbb{P}^x[\tau_x' < \infty] \ge \mathbb{P}^x[\tau_y < \infty] \mathbb{P}^y[\tau_x < \infty] = F(x, y)^2,$$

where $\tau'_x \stackrel{\text{def.}}{=} \inf\{k \geq 1 : S(k) = x\}$. Thus

$$d(x, y) = 0 \iff F(x, y) = 1 \iff x = y$$
.

Finally,

$$\mathbb{P}^x[\tau_z < \infty] \ge \mathbb{P}^x[\tau_y < \infty] \mathbb{P}^y[\tau_z < \infty]$$

leads to the triangular inequality.

Let define

$$K \stackrel{\text{def.}}{=} \max\{d(e, x) : x \in \mathcal{S}\}. \tag{2}$$

According to this distance, we define the ball of radius Kn,

$$B(x,n) \stackrel{\text{def.}}{=} \{ y \in \Gamma : d(x,y) \le Kn \}$$

and its external boundary

$$\partial B(x,n) \stackrel{\text{def.}}{=} \{ z \notin B(x,n) : \exists y \sim z \text{ and } y \in B(x,n) \}.$$

Here $y \sim z$ means that y and z are neighbours, that is $y^{-1}z \in \mathcal{S}$.

The reason why the constant K appears in the definition of the balls is to be sure that $\partial B(e,n) \subset B(e,n+1)$. Indeed, let $x \in \partial B(e,n)$ and $y \in B(e,n)$ such that $y \sim x$. Then,

$$d(e, x) \le d(e, y) + d(y, x) \le Kn + \max\{d(e, z) : z \in \mathcal{S}\} \le K(n + 1)$$
.

We write d(x) = d(e, x) and B(n) = B(e, n). To have a better understanding of this distance, we will compare it to the word distance. First remark that, for any $x \in \Gamma$, writing $x = x_1 \cdots x_{|x|}$ with all the x_i 's in S, leads to

$$d(x) \le \sum_{i=1}^{|x|} d(x_i) \le K|x|.$$
 (3)

To get a lower bound for the hitting distance, we need asymptotic estimates of the Green function. These estimates are direct consequences of the following results on the asymptotic of the iterated transition probability:

1. An off-diagonal gaussian upper bound for symmetric random walks (with finite support) on generic groups (Varopoulos [14]; Carne [4]),

$$\exists C_g > 1 \text{ s.t. } \forall x, y \in \Gamma \text{ and } k \in \mathbb{N} \quad \mathbb{P}^x[S(k) = y] \le C_g \exp(-C_g|x^{-1}y|^2/k), \quad (4)$$

2. For random walks on groups having exponential growth (Varopoulos [15]; Carlen, Kusuoka & Stroock [3]),

$$\exists C_e > 1 \text{ s.t. } \forall x, y \in \Gamma \text{ and } k \in \mathbb{N} \quad \mathbb{P}^x[S(k) = y] \le C_e \exp(-C_e k^{1/3}),$$
 (5)

3. In the special case of non-ameanable groups, the latter estimates becomes (Kesten [8])

$$\exists C_{na} > 1 \text{ s.t. } \forall x, y \in \Gamma \text{ and } k \in \mathbb{N} \quad \mathbb{P}^x[S(k) = y] \le C_{na} \exp(-C_{na}k).$$
 (6)

For an overview of such estimates and precise references, see Woess [16] or Pittet & Saloff-Coste [12].

Lemma 2.2 Let Γ be a group having exponential growth, then $\exists C > 0$ such that

$$G(x) \le \exp(-C|x|^{1/2}) \ \forall x \in \Gamma.$$

Proof

Using (4) and (5),

$$G(x) = \sum_{k=0}^{\infty} \mathbb{P}[S(k) = x] = \sum_{k \le |x|^{3/2}} \mathbb{P}[S(k) = x] + \sum_{k > |x|^{3/2}} \mathbb{P}[S(k) = x]$$

$$\leq \sum_{k \le |x|^{3/2}} C_g \exp(-C_g |x|^2 / k) + \sum_{k > |x|^{3/2}} C_e \exp(-C_e k^{1/3})$$

$$\leq C_g |x|^{3/2} \exp(-C_g |x|^{1/2}) + 15|x| \exp(-C_e |x|^{1/2})$$

$$\leq \exp(-C|x|^{1/2}),$$

for some constant C. The before last inequality comes from the fact that for any sufficiently large N

$$\sum_{k=N+1}^{\infty} \exp(-C_e k^{1/3}) \leq \int_{N}^{+\infty} \exp(-C_e t^{1/3}) dt$$

$$\leq \int_{N}^{+\infty} \frac{d(-6t^{2/3} \exp(-C_e t^{1/3}))}{dt} dt = -6N^{2/3} \exp(-C_e N^{1/3})$$

Actually, Revelle [13] proved independently Lemma 2.2.

Note that when the group is non-ameanable, using (6), this upper bound of the Green function becomes:

$$G(x) \le \exp(-C|x|). \tag{7}$$

The above lemma assures that G(x) tends to 0 as |x| goes to infinity. By (1), so does F(x). Lemma 2.2 leads also to

$$d(x) \ge K_1 |x|^{1/2} \,, \tag{8}$$

for some constant K_1 .

Actually, as soon as the Green function decreases exponentially (with respect to the word distance), these distances are equivalent. This is the case when Γ is non-ameanable (7) and also for some ameanable groups (for instance the Lamplighter group $\mathbb{Z} \wr \mathbb{Z}_2$, Brofferio & Woess [2]).

Remark that (3) and (8) are sharp in the sense that examples exist where the Green function decreases as $\exp(-C|x|)$ or $\exp(-C|x|^{1/2})$ depending on the direction toward infinity (for instance $\mathbb{Z} \wr \mathbb{Z}$, Revelle [13])

As $B_w(n) \subset B(n)$, the size V(n) of the balls B(n) grows exponentially, but we prove that the constant in the exponential (lower and upper) bound is exactly the constant K which appeared in the definition of the balls.

Proposition 2.3 There exist two constants C_a and C_b such that for every integer n,

$$C_a \exp(Kn) \le V(n) \le C_b n^3 \exp(Kn)$$
.

Proof

The proof of the upper bound relies on the following estimate taken from Revelle [13]: for any finite set A,

$$\sum_{x \in A} G(x) \le (16C_e^{-2} + 1)(\ln \# A)^3. \tag{9}$$

By sake of completeness, we give the computation that leads to (9). The estimate (5) (which introduces the constant C_e) yields

$$\sum_{x \in A} G(x) \leq \sum_{k=0}^{\infty} \sum_{x \in A} \mathbb{P}[S(k) = x]$$

$$= \sum_{k \leq C_e^{-3}(\ln \# A)^3 + 1} \sum_{x \in A} \mathbb{P}[S(k) = x] + \sum_{k > C_e^{-3}(\ln \# A)^3 + 1} \sum_{x \in A} \mathbb{P}[S(k) = x]$$

$$\leq C_e^{-3}(\ln \# A)^3 + 1 + \#A \sum_{k > C_e^{-3}(\ln \# A)^3 + 1} C_e \exp(-C_e k^{1/3})$$

$$\leq (16C_e^{-2} + 1)(\ln \# A)^3.$$

When A = B(n), the definition of the ball B(n) and (9) give

$$#B(n) \cdot \exp(-Kn) \le \sum_{x \in B(n)} G(x) \le (16C_e^{-2} + 1)(\ln #B(n))^3$$
 (10)

which implies, since $(\ln \#B(n))^3 \leq (\#B(n))^{1/2}$, that #B(n) is at most an exponential of n. Applying once more (10) leads to the $C_b n^3 \exp(Kn)$ upper bound, for some positive constant C_b .

For the lower bound, remark that

$$\sum_{x \in \partial B(n-1)} G(x) \le (\#\partial B(n-1)) \exp(-K(n-1)).$$

The left hand side is larger than (or equal to) 1 since the random walk is transient and it must go through $\partial B(n-1)$. So with $C_a \stackrel{\text{def.}}{=} \exp(-K)$,

$$V(n) \ge \#\partial B(n-1) \ge C_a \exp(Kn)$$
.

Remark 2.4 For non-ameanable groups, the upper bound in the volume growth becomes $C_b n \exp(Kn)$. Indeed, using the specific estimate of the return probability (6), the upper bound in (9) becomes a constant times $\ln \# A$.

3 Limiting shape for the Internal DLA

Let us now state our main result, a comparison between the shape of the Internal DLA and the balls for the hitting distance.

Theorem 3.1 Let $A(\cdot)$ be the Internal DLA on a finitely generated group of exponential growth, associated to a symmetric random walk with finitely supported increment. Then, for any constants $C_I > 3/K$ and $C_O > 2$,

$$\mathbb{P}[\exists n_0 \text{ s.t. } \forall n > n_0, B(n - C_I \ln n) \subset A(V(n)) \subset B(n + C_O \sqrt{n})] = 1.$$

This result give an upper bound for the two random variables $\delta_I(n)$ and $\delta_O(n)$ describing the inner and outer fluctuation of A(n) with respect to B(n)

$$\delta_I(n) \stackrel{\text{def.}}{=} n - \inf\{d(z)/K : z \notin A(V(n))\},$$

 $\delta_O(n) \stackrel{\text{def.}}{=} \sup\{d(z)/K : z \in A(V(n))\} - n.$

An alternative definition is that $B(n-\delta_I(n))$ is the largest ball included in A(V(n)), and $B(n+\delta_O(n))$ is the smallest ball which contains A(V(n)). Theorem 3.1, states

that for any constants $C_I > 3/K$ and $C_O > 2$, the events $\{\delta_I(n) > C_I \ln n\}$ and $\{\delta_O(n) > C_O \sqrt{n}\}$ occur only finitely often, with probability 1.

We split the proof into two parts, one regarding the inner error and the other regarding the outer error.

Let us first give some definitions used throughout the proof. Let $S^{j}(\cdot)$ be the jth random walk that defines the Internal DLA A(V(n)). Recall that all the $S^{j}(\cdot)$ start from e. Moreover, we let this random walk running even after it adds a new point to the cluster A(j-1). This adding time will then be the stopping time:

$$\sigma^j \stackrel{\text{def.}}{=} \inf\{k \ge 0 : S^j(k) \notin A(j-1)\}.$$

We also define the hitting time of a point z by the (unstopped) random walk S^{j} :

$$\tau_z^j \stackrel{\text{def.}}{=} \inf\{k \ge 0 : S^j(k) = z\}.$$

3.1 Proof of the inner part

We will prove that there exists a constant C_I such that, with probability 1, for all n large enough

$$B(n - C_I \ln n) \subset A(V(n))$$
.

We start our proof by computing an upper bound for the probability that a given point $z \in B(n)$ does not belong to the Internal DLA at time V(n). For that purpose, we define the following indicator random variable

$$N^j = N^j(z) \stackrel{\text{def.}}{=} \mathbb{1}_{\{\tau_z^j \le \sigma^j\}},$$

which can be decomposed as $N_j = M_j - L_j$ where

$$M^j = M^j(z) \stackrel{\mathrm{def.}}{=} \ \mathbbm{1}_{\{\tau^j_z < \infty\}} \ \ \mathrm{and} \ \ L^j = L^j(z) \stackrel{\mathrm{def.}}{=} \ \mathbbm{1}_{\{\sigma^j \leq \tau^j_z < \infty\}} \,.$$

Observe that z does not belong to A(V(n)) if and only if $N^j=0$ for every j. So, for any $\lambda>0$,

$$\mathbb{P}[z \not\in A(V(n))] = \mathbb{P}\left[\sum_{j=1}^{V(n)} N^j = 0\right] \leq \mathbb{E}\left[e^{-\lambda \sum_{j=1}^{V(n)} N^j}\right]$$

$$= \mathbb{E}\left[e^{-\lambda \sum_{j=1}^{V(n)} M^j + \lambda \sum_{j=1}^{V(n)} L^j}\right]$$

$$\leq \mathbb{E}\left[e^{-2\lambda \sum_{j=1}^{V(n)} M^j}\right]^{1/2} \times \mathbb{E}\left[e^{2\lambda \sum_{j=1}^{V(n)} L^j}\right]^{1/2}.$$

As the M^{j} 's are i.i.d. indicator random variables,

$$\mathbb{E}\left[e^{-2\lambda\sum_{j=1}^{V(n)}M^{j}}\right] = \prod_{j=1}^{V(n)}\mathbb{E}\left[e^{-2\lambda M^{j}}\right] = \prod_{j=1}^{V(n)}(1 - (1 - e^{-2\lambda})F(z))$$

$$= (1 - (1 - e^{-2\lambda})F(z))^{V(n)} \le \exp(-(1 - e^{-2\lambda})F(z)V(n)).$$

Observe that the L^j 's are not independent, but they can be bounded from above by random variables which are independent when knowing where the random walk S^j adds to the cluster. More precisely, let define the hitting time of z after the adding time σ^j ,

$$\widetilde{\tau_z}^j \stackrel{\text{def.}}{=} \inf\{k \ge \sigma^j \text{ s.t. } S^j(k) = z\}.$$

Then,

$$L^j \leq \widetilde{L}^j \stackrel{\text{def.}}{=} \mathbb{1}_{\{\widetilde{\tau_z}^j < \infty\}}.$$

Let \mathcal{G}_n be the σ -algebra generated by $S^j(k \wedge \sigma^j)$ for all $j \leq V(n)$ and $k \in \mathbb{N}$, that is the σ -algebra containing all the information on the V(n) random walks before they add to the cluster. As the \widetilde{L}^j 's are independent conditioning by \mathcal{G}_n ,

$$\mathbb{E}\left[e^{2\lambda\sum_{j=1}^{V(n)}L^{j}}\right] \leq \mathbb{E}\left[e^{2\lambda\sum_{j=1}^{V(n)}\tilde{L}^{j}}\right] = \mathbb{E}\left[\mathbb{E}\left[e^{2\lambda\sum_{j=1}^{V(n)}\tilde{L}^{j}} \middle| \mathcal{G}_{n}\right]\right] \\
= \mathbb{E}\left[\prod_{j=1}^{V(n)}\mathbb{E}\left[e^{2\lambda\tilde{L}^{j}} \middle| \mathcal{G}_{n}\right]\right].$$

Since the \widetilde{L}^{j} 's are indicator random variables,

$$\mathbb{E}\left[e^{2\lambda\sum_{j=1}^{V(n)}L^{j}}\right] \leq \mathbb{E}\left[\prod_{j=1}^{V(n)}(1+(e^{2\lambda}-1)\mathbb{E}[\widetilde{L}^{j}\mid\mathcal{G}_{n}])\right] \leq \mathbb{E}\left[e^{(e^{2\lambda}-1)\sum_{j=1}^{V(n)}\mathbb{E}[\widetilde{L}^{j}\mid\mathcal{G}_{n}]}\right].$$

Finally,

$$\sum_{j=1}^{V(n)} \mathbb{E}\left[\widetilde{L}^{j} \middle| \mathcal{G}_{n}\right] = \sum_{j=1}^{V(n)} \mathbb{E}\left[\mathbb{1}_{\{\widetilde{\tau}_{z}^{j} < \infty\}} \middle| \mathcal{G}_{n}\right] = \sum_{j=1}^{V(n)} \mathbb{P}^{S^{j}(\sigma^{j})}[\widetilde{\tau}_{z}^{j} < \infty]$$
$$= \sum_{y \in A(V(n))} F(y, z) = \sum_{y \in A(V(n))} F(z, y).$$

The definition of the hitting distance implies that for any $y \in B(z, n)$ and any $y' \notin B(z, n)$, F(z, y) > F(z, y'). Thus, since #A(V(n)) = V(n) = #B(z, n),

$$\sum_{y \in A(V(n))} F(z, y) \le \sum_{y \in B(z, n)} F(z, y) = \sum_{y \in B(n)} F(y).$$
 (11)

Putting things together, we get that for any $\lambda > 0$, n and $z \in B(n)$,

$$\mathbb{P}[z \not\in A(V(n))] \le \exp\left(-C_{\lambda}V(n)F(z) + C_{\lambda}' \sum_{y \in B(n)} F(y)\right). \tag{12}$$

with $C_{\lambda} = \frac{1 - e^{-2\lambda}}{2}$ and $C'_{\lambda} = \frac{e^{2\lambda} - 1}{2}$.

We remark that the inequality (12) is valid for every transient groups whatever its growth rate. For groups of exponential growth, Proposition 2.3 enables to estimate the last term in the exponential. Thus, if $z \in B(n - C_I \ln n)$ (that is $F(z) \ge n^{KC_I} e^{-Kn}$), using (9),

$$\mathbb{P}[z \notin A(V(n))] \le e^{-C_{\lambda}V(n)n^{KC_I}e^{-K_n} + c \cdot C_{\lambda}'(\ln V(n))^3} \le e^{-C_{\lambda} \cdot C_a n^{KC_I} + c' \cdot C_{\lambda}' n^3} \le e^{-C_{\lambda} \cdot C_a n^3},$$

for $C_I > 3/K$ and n large enough. Then,

$$\sum_{z \in B(n - C_I \ln n)} \mathbb{P}[z \notin A(V(n))] \leq V(n - C_I \ln n) e^{-C_{\lambda} \cdot C_a n^3}$$

$$\leq C_b (n - C_I \ln n)^3 e^{Kn - C_I \ln n - C_{\lambda} \cdot C_a n^3}$$

which is summable. Thus the result follows from Borel-Cantelli Lemma.

3.2 Proof of the outer part

To get the upper bound for the outer error in Theorem 3.1, we will estimate the decay of the expected number of points in A(V(n)) that lies within the spheres $\partial B(n+p)$ for $p \geq 1$. Checking that this decay is exponential in p^2 will lead to the desired result.

We define

$$Z_p(j) \stackrel{\text{def.}}{=} \#(A(j) \cap \partial B(n+p)) = \sum_{i=1}^j \mathbb{1}_{S^i(\sigma^i) \in \partial B(n+p)},$$

and $\nu_p(j) \stackrel{\text{def.}}{=} \mathbb{E}[Z_p(j)]$. If a random walk adds to the cluster in $\partial B(n+p+1)$, that implies it has stayed within the cluster before. It means that

$$\{S^i(\sigma^i)\in\partial B(n+p+1)\}\subset\{\exists x\in\partial B(n+p)\cap A(i-1)\ \text{s.t.}\ \tau^i_x<\infty\}\,.$$

Therefore,

$$\nu_{p+1}(j) \leq \sum_{i=1}^{j} \mathbb{P}[\exists x \in \partial B(n+p) \cap A(i-1) \text{ s.t. } \tau_{x}^{i} < \infty]$$

$$\leq \sum_{i=1}^{j} \mathbb{E}\left[\sum_{x \in \partial B(n+p) \cap A(i-1)} \mathbb{I}_{\tau_{x}^{i} < \infty}\right]$$

$$\leq \sum_{i=1}^{j} \mathbb{E}\left[Z_{p}(i-1) \max_{x \in \partial B(n+p)} \mathbb{P}[\tau_{x} < \infty]\right],$$

by independence between S^i and A(i-1). Then,

$$\nu_{p+1}(j) \le \sum_{i=1}^{j} \exp(-K(n+p))\nu_p(i-1).$$
 (13)

For any fixed j, iterating (13) leads to

$$\nu_p(j) \le \exp(-Kn(p-1))\exp(-Kp(p-1)/2)\frac{j^p}{p!}$$

by a simple induction on p started from $\nu_1(j) \leq j$ and using

$$\sum_{i=1}^{j} \frac{(i-1)^p}{p!} \le \frac{j^{p+1}}{(p+1)!}.$$

Finally, as $p! \ge p^p \exp(-p)$,

$$\nu_p(j) \le \exp(-Kn(p-1))\exp(-Kp(p-1)/2)\exp(p)p^{-p}j^p$$
.

Thus, for j = V(n) and p larger than some constant independent of n,

$$\nu_p(V(n)) \le n^{3p} \exp\left(-K\left(\frac{p^2}{3} - n\right)\right)$$
.

Therefore, for $C_O > 2$ and $p = C_O \sqrt{n}$,

$$\mathbb{P}[A(V(n)) \not\subset B(n + C_O \sqrt{n})] \leq \mathbb{P}[Z_{C_O \sqrt{n}}(V(n)) \geq 1]$$

$$\leq \nu_{C_O \sqrt{n}}(V(n))$$

$$\leq \exp\left(-K\left(\frac{C_O^2}{4} - 1\right)n\right),$$

for n large enough.

Thus, by the Borel-Cantelli Lemma, with probability 1, for all n large enough,

$$A(V(n)) \subset B(n + C_O\sqrt{n})$$
.

4 Lower bounds for the errors when ${\mathcal G}$ is the homogeneous tree ${\mathbb T}_q$

In this section we will give lower bounds for the inner and outer fluctuations in the specific case of Simple Random Walks on the homogeneous tree.

Let \mathcal{G} be the homogeneous tree \mathbb{T}_q of degree $q \geq 3$. When q is even, \mathbb{T}_q is the Cayley graph of the free group $F_{q/2}$. When q is odd, \mathbb{T}_q is the Cayley graph of the free product of $F_{(q-1)/2}$ by \mathbb{Z}_2 . On such a graph, most of the estimates we need are known exactly (Dynkin & Maljutov [7] and Cartier [5]) for the Simple Random Walk (nearest neighbours with uniform transition probability). In particular since

$$F(z) = (q-1)^{-|z|},$$

the hitting distance and the word distance are proportional and the ball $B_w(n)$ coincides with B(n) exactly thanks to the use of the constant $K = \ln(q-1)$. Thus we have

$$V(n) = \frac{q(q-1)^n - 2}{q-2}$$
 and $\#\partial B(n) = q(q-1)^n$.

Let

$$\xi_n \stackrel{\text{def.}}{=} \min\{k \ge 0 \text{ s.t. } S(k) \in \partial B(n)\},$$

be the exit time from the ball B(n). Then, as we deal with Simple Random Walks,

$$\mathbb{P}[S(\xi_n) = z] = \frac{1}{q(q-1)^n} \cdot \delta_{z \in \partial B(n)}.$$

4.1 Lower bound for the outer error

Proposition 4.1 For any constant $C_o < (\ln(q-1)/\ln q)^{1/2}$,

$$\mathbb{P}[A(V(n)) \cap \partial B(n + C_o\sqrt{n}) = \emptyset \text{ infinitely often}] = 0$$

Proof

By Theorem 3.1, for any constant C_I sufficiently large, if one defines the increasing sequence of events

$$\Omega_{n_0} \stackrel{\text{def.}}{=} \{ \forall n \geq n_0, B(n - C_I \ln(n) + 1) \subset A(V(n)) \} .$$

One has

$$\lim_{n_0 \to \infty} \mathbb{P}[\Omega_{n_0}] = 1. \tag{14}$$

Suppose that Ω_{n_0} occurs and fix $n > n_0$. Denote $R \stackrel{\text{def.}}{=} n - 1 - C_I \ln(n - 1)$. Let the first V(n-1) random walks build A(V(n-1)), which covers B(R). Then we consider

the next V(n)-V(n-1) random walks and stop them as soon as they reach $\partial B(R)$. Let N_x be the number of random walks that are stopped in $x \in \partial B(R)$. Letting the stopped random walks start again we can reconstruct the original Internal DLA. The same random walks construct new smaller Internal DLA's, $A_x(N_x)$, that start in every $x \in \partial B(R)$ and are built up using N_x random walks. We state that if $B(R) \subset A(V(n-1))$ then

$$A(V(n)) \supset A(V(n-1)) \bigcup_{x \in \partial B(R)} A_x(N_x).$$

Thus, if there exists $x \in \partial B(R)$ such that $A_x(N_x) \cap \partial B(R+r) \neq \emptyset$, then $A(V(n)) \cap \partial B(R+r) \neq \emptyset$. Observe also that the random walks that start on $\partial B(R)$ are independent and, thus, all the $\{A_x(N_x)\}_{x \in \partial B(R)}$ are independent once one knows all the N_x , that is they are independent for the probability $\mathbb{P}[\cdot \mid \sigma_R]$. Here σ_R is the σ -algebra generated by the $S^i(k \wedge \xi_R)$ for all $i \leq N \stackrel{\text{def.}}{=} V(n) - V(n-1)$ and all $k \in \mathbb{N}$, that is the σ -algebra containing all the information on the N random walks until they hit $\partial B(R)$.

For all $x \in \partial B(R)$, there exits a path C_x of (graph) length r from x to $\partial B(R+r)$. Note that

$${A_x(N_x) \cap \partial B(R+r) = \emptyset} \subset {\mathcal{C}_x \not\subset A_x(N_x)}.$$

We split the N_x random walks that start at x into $\lfloor \frac{N_x}{r} \rfloor$ packets of size r ($\lfloor \cdot \rfloor$ is the lower integer part), the rest being ignored. The probability that all the random walks of one packet follow exactly \mathcal{C}_x when they start, is $(q^{-r})^r$. By definition of the Internal DLA, this event guaranty that \mathcal{C}_x is contained in $A_x(N_x)$. Thus, since all packets are independent

$$\mathbb{P}[\mathcal{C}_x \not\subset A_x(N_x) \mid \sigma_R] \le (1 - q^{-r^2})^{\left\lfloor \frac{N_x}{r} \right\rfloor} \le (1 - q^{-r^2})^{\frac{N_x}{r} - 1}.$$

Finally, for $n > n_0$,

$$\mathbb{P}[\{A(n) \cap \partial B(R+r) = \emptyset\} \quad \cap \quad \Omega_{n_0}] \\
\leq \quad \mathbb{P}[\forall x \in \partial B(R), A_x(N_x) \cap \partial B(R+r) = \emptyset] \\
= \quad \mathbb{E}\left[\prod_{x \in \partial B(R)} \mathbb{P}[A_x(N_x) \cap \partial B(R+r) = \emptyset \mid \sigma_R]\right] \\
\leq \quad \mathbb{E}\left[\prod_{x \in \partial B(R)} (1 - q^{-r^2})^{\frac{N_x}{r} - 1}\right] \\
= \quad \mathbb{E}\left[(1 - q^{-r^2})^{\sum_{x \in \partial B(R)} \frac{N_x}{r} - 1}\right] \\
= \quad (1 - q^{-r^2})^{\frac{V(n) - V(n-1)}{r} - \#\partial B(R)} \\
\leq \quad \exp\left(-q^{-r^2}\left(\frac{V(n) - V(n-1)}{r} - \#\partial B(R)\right)\right).$$

Remark that taking $r = C_o \sqrt{n} + 1 + C_I \ln(n-1)$, with a constant $C_o < (\ln(q-1)/\ln q)^{1/2}$ and $C_I > \frac{1}{2}$, we get

$$\exp\left(-q^{-r^2}\left(\frac{V(n)-V(n-1)}{r}-\#\partial B(R)\right)\right) \le \exp(-\exp(cn)),$$

for some constant c > 0. Then, by Borel-Cantelli Lemma,

$$\mathbb{P}[\{A(V(n)) \cap \partial B(n + C_o\sqrt{n}) \text{ i.o.}\} \cap \Omega_{n_0}] = 0.$$

Letting n_0 goes to infinity leads (by (14)) to the result.

4.2 Lower bound for the inner error

For the inner error, we prove a weaker result than the one on the outer error. Namely, we get that infinitely often, the inner error is of order $\ln n$. It assures the sharpness of the order $\ln n$ for a general upper bound but does not prevent the inner error to be sometimes smaller. Thus our result is somehow weaker than the one announced by Etheridge and Lawler.

Proposition 4.2 For any constant $C_i < 1/(2 \ln(q-1))$,

$$\mathbb{P}[A(V(n))^c \cap B(n - C_i \ln n) \neq \emptyset \text{ infinitely often}] = 1.$$

Proof

Let $N \stackrel{\text{def.}}{=} V(n)$ and $R \stackrel{\text{def.}}{=} n - 1 - C_i \ln n$. Then the event $\{A(N)^c \cap B(R+1) \neq \emptyset\}$ contains the event that there exists a point on the sphere of radius R attained by none of N independent random walks.

Let fix r < R, for every $z \in \partial B(R)$, we define its class as

$$[z] \stackrel{\text{def.}}{=} \partial B(R) \cap B(z, 2r)$$
,

and let $A_R \stackrel{\text{def.}}{=} \bigcup_{i=1}^N [S^i(\xi_R)]$ be the set of all classes that are hit by the N random walks when they first exit B(R). Then we have

$$\mathbb{P}[A(N)^{c} \cap B(R) \neq \emptyset]$$

$$\geq \mathbb{P}[\exists z \in \partial B(R) \text{ s.t. } \forall i \leq N, \ \tau_{z}^{i} = \infty]$$

$$\geq \mathbb{P}[\exists z \in \partial B(R) \text{ s.t. } \forall i \leq N, \ \tau_{z}^{i} = \infty \text{ and } A_{R} \subsetneq \partial B(R)]]$$

$$= \sum_{A \subsetneq \partial B(R)} \mathbb{P}[\exists z \in \partial B(R) \text{ s.t. } \forall i \leq N, \ \tau_{z}^{i} = \infty \mid A_{R} = A] \mathbb{P}[A_{R} = A].$$

To get a lower bound for the above conditional probability we will need, once the set A is fixed, to choose a suitable point called z_A on $\partial B(R)$ that has few chances to

be hit by the random walks after ξ_R^i . For that purpose, we construct the following algorithm. Suppose $A \subsetneq \partial B(R)$ is a fixed set, then we will write a "Mouse algorithm" whose aim is to find the most secure place z_A on $\partial B(R)$ for a mouse knowing that #A cats will start simple random walks (hunts) from each of the points in A.

The mouse starts from the origin and proceeds toward $\partial B(R)$ choosing the direction of the subtree with the fewest cats. Denote (m_j) (j = 0 to R) the successive positions of the mouse, that is the steps of our algorithm:

- $m_0 = e$;
- Choose m_{j+1} inside $\{y \sim m_j : |y| = j+1 \text{ and } B(y, R-j) \cap A \text{ is minimal}\}.$

Let $z_A \stackrel{\text{def.}}{=} m_{R+1}$. It is easy to check that $z_A \notin A$. Recall that σ_R is the σ -algebra generated by the $S^i(k \wedge \xi_R)$ for all $i \leq N$ and all $k \in \mathbb{N}$, that is the σ -algebra containing all the information on the N random walks before they hit $\partial B(R)$. Then,

$$\begin{split} \mathbb{P}[\exists z \in \partial B(R) \quad \text{s.t.} \quad \forall i \leq N, \ \tau_z^i = \infty \mid A_R = A] \\ & \geq \quad \mathbb{P}[\forall i \leq N, \ \tau_{z_A}^i = \infty \mid A_R = A] \\ & = \quad \mathbb{P}[\forall i \leq N, \ \xi_R^i \leq \tau_{z_A}^i = \infty \mid A_R = A] \\ & = \quad \mathbb{E}\left[\mathbb{P}[\forall i \leq N, \ \xi_R^i \leq \tau_{z_A}^i = \infty \mid \sigma_R] \mid A_R = A\right] \\ & = \quad \mathbb{E}\left[\prod_{i=1}^N \mathbb{P}[\xi_R^i \leq \tau_{z_A}^i = \infty \mid \sigma_R] \mid A_R = A\right] \\ & = \quad \mathbb{E}\left[\prod_{i=1}^N \mathbb{P}^{S^i(\xi_R)}[\tau_{z_A} = \infty] \mid A_R = A\right] . \end{split}$$

By transience of the random walk, $\mathbb{P}^{S^i(\xi_R)}[\tau_{z_A} = \infty] = 1 - F(S^i(\xi_R), z_A)$ has a uniform (in all parameter) strickly positive lower bound. Thus, there exists a constant c > 0 such that

$$\mathbb{E}\left[\prod_{i=1}^{N} \mathbb{P}^{S^{i}(\xi_{R})}[\tau_{z_{A}} = \infty] \middle| A_{R} = A\right] \geq \mathbb{E}\left[\exp\left(-c\sum_{i=1}^{N} F(S^{i}(\xi_{R}), z_{A})\right) \middle| A_{R} = A\right]$$

$$\geq \exp\left(-c\sum_{i=1}^{N} \mathbb{E}\left[F(S^{i}(\xi_{R}), z_{A}) \middle| A_{R} = A\right]\right)$$

$$= \exp\left(-cN\mathbb{E}\left[F(S^{1}(\xi_{R}), z_{A}) \middle| A_{R} = A\right]\right).$$

By definition of the Mouse algorithm, we easily check that, for every j,

$$T_j \stackrel{\text{def.}}{=} B(z_A, 2(R+1-j)) \cap A = B(m_j, R+1-j) \cap A,$$

and that its cardinality is $\#T_j \leq \#A(q-1)^{-j}$. Remark also that since A must be a union of sets of the type $\partial B(R) \cap B(z,2r)$ with $z \in \partial B(R)$ (otherwise $\mathbb{P}[A_R = A] = 0$), $z_A \notin A$ implies $d(z_A, A) > 2r$. Then,

$$\mathbb{E}\left[F(S^{1}(\xi_{R}), z_{A}) \mid A_{R} = A\right] = \sum_{j=0}^{R-r} \mathbb{E}\left[\mathbb{1}_{S^{1}(\xi_{R}) \in T_{j} \setminus T_{j+1}} F(S^{1}(\xi_{R}), z_{A}) \mid A_{R} = A\right]$$

$$= \sum_{j=0}^{R-r} \mathbb{E}\left[\mathbb{1}_{S^{1}(\xi_{R}) \in T_{j} \setminus T_{j+1}} (q-1)^{-2(R+1-j)} \mid A_{R} = A\right]$$

$$= \sum_{j=0}^{R-r} (q-1)^{-2(R+1-j)} \mathbb{P}\left[S^{1}(\xi_{R}) \in T_{j} \setminus T_{j+1} \mid A_{R} = A\right].$$

Remark that the event $A_R = A$ is invariant by any bijection γ of $\partial B(R)$ such that $\gamma(A) = A$ and for any $x \in \partial B(R)$, $[\gamma(x)] = \gamma([x])$. Thus, the law of $S^1(\xi_R)$ under the condition $A_R = A$ is the uniform distribution on A. Thus,

$$\mathbb{P}\left[S^1(\xi_R) \in T_j \backslash T_{j+1} \mid A_R = A\right] = \frac{\#(T_j \backslash T_{j+1})}{\#A} \le \frac{\#T_j}{\#A}.$$

Therefore,

$$\mathbb{E}\left[F(S^{1}(\xi_{R}), z_{A}) \mid A_{R} = A\right] \leq \sum_{j=0}^{R-r} \frac{(q-1)^{-2(R+1-j)}}{(q-1)^{j}} \leq (q-1)^{-R-r}.$$

Finally,

$$\begin{split} \mathbb{P}[\exists z \in \partial B(R) \text{ s.t. } \forall i \leq N, \, \tau_z^i = \infty] & \geq \sum_{A \varsubsetneq \partial B(R)} \exp\left(-cN(q-1)^{-R-r}\right) \mathbb{P}[A_R = A] \\ & = \exp\left(-cN(q-1)^{-R-r}\right) \mathbb{P}[A_R \varsubsetneq \partial B(R)] \,. \end{split}$$

Remark that $\mathbb{P}[A_R \subsetneq \partial B(R)]$ is the probability that N balls uniformly distributed into $M = q(q-1)^{R-r}$ boxes leave at least one empty box. By Lemme 4.3, that will be prove at the end of the section,

$$\mathbb{P}[A_R \subsetneq \partial B(R)] \ge 1 - \exp\left(-\frac{q(q-1)^{R-r}}{2} \exp\left(-4N\frac{(q-1)^{r-R}}{q}\right)\right).$$

When $N = V(n) \le q(q-1)^n$, $R = n - C_i \ln n$ and $r = C_1 \ln n$ (the constant C_1 will be fixed below), we easily check that

$$\exp\left(-\frac{q(q-1)^{R-r}}{2}\exp\left(-4N\frac{(q-1)^{r-R}}{q}\right)\right) \\ \leq \exp\left(-\frac{q \, n^{-(C_1+C_i)\ln(q-1)}}{2}\exp\left(n\ln(q-1) - 4n^{(C_1+C_i)\ln(q-1)}\right)\right).$$

So, with $C_1 = \frac{1}{2\ln(q-1)}$, for any constant $C_i < \frac{1}{2\ln(q-1)}$,

$$\mathbb{P}[A_R \subsetneq \partial B(R)] \xrightarrow[n \to \infty]{} 1.$$

With the same constants C_1 and C_i , as $C_1 > C_i$,

$$\exp\left(-cN(q-1)^{-R-r}\right) \ge \exp\left(-c \cdot q \cdot n^{-(C_1-C_i)\ln(q-1)}\right),$$

also tends to 1 when n goes to infinity. Finally,

$$\mathbb{P}[\exists z \in \partial B(n - C_i \ln n) \text{ s.t. } \forall i \leq V(n), \, \tau_z^i = \infty] \xrightarrow[n \to \infty]{} 1.$$

Thus $\mathbb{P}[A(V(n))^c \cap B(n-C_i \ln n) \neq \emptyset] \to 1$ which completes the proof by Fatou Lemma.

We now prove the estimate on the multinomial law that we used in the proof.

Lemma 4.3 Let put independently N balls into M boxes $(M \ge 3)$ with uniform probability. Let $\mathbb{P}_{N,M}$ denote the probability law of that process. Then,

$$\mathbb{P}_{N,M}[Each\ box\ contains\ at\ least\ one\ ball] \leq \exp\left(-\frac{M}{2}\exp\left(-4\frac{N}{M}\right)\right)$$
.

Proof

Let (b_k) (k = 1 to M) denote the M boxes and (a_i) (i = 1 to N) the N balls. Let also define

$$\begin{array}{cccc} p_{N,M} & \stackrel{\mathrm{def.}}{=} & \mathbb{P}_{N,M}[\forall k \leq M, \; \exists i \leq N \; \; \mathrm{s.t.} \; \; a_i \in b_k] \; ; \\ q_{N,M} & \stackrel{\mathrm{def.}}{=} & \mathbb{P}_{N,M}[\exists i \leq N \; \; \mathrm{s.t.} \; \; a_i \in b_1] \; ; \\ I_1 & \stackrel{\mathrm{def.}}{=} & \{i \; \; \mathrm{s.t.} \; \; a_i \in b_1\} \; . \end{array}$$

Therefore,

$$\begin{split} p_{N,M} &= \sum_{I \neq \emptyset} \mathbb{P}_{N,M}[\{\forall k \neq 1 \,\exists i \not\in I \, \text{ s.t. } a_i \in b_k\} \cap I_1 = I] \\ &= \sum_{I \neq \emptyset} \mathbb{P}_{N,M}[\forall k \neq 1 \,\exists i \not\in I \, \text{ s.t. } a_i \in b_k \mid I_1 = I] \times \mathbb{P}_{N,M}[I_1 = I] \\ &= \sum_{I \neq \emptyset} p_{N-\#I,M-1} \times \mathbb{P}_{N,M}[I_1 = I] \\ &\leq \sum_{I \neq \emptyset} p_{N,M-1} \times \mathbb{P}_{N,M}[I_1 = I] \\ &= p_{N,M-1} \cdot q_{N,M} \,. \end{split}$$

The third equality comes from the fact that conditioning by $I_1 = I$, the N - #I remaining balls are uniformly distributed among the M-1 remaining boxes. The inequality comes from the monotonicity in N of $p_{N,M}$. By induction,

$$p_{N,M} \le \prod_{j=2}^{M} q_{N,j} \cdot p_{N,1} .$$

As $p_{N,1} = 1$ and $q_{N,j} = 1 - \left(1 - \frac{1}{j}\right)^N$,

$$p_{N,M} \leq \prod_{j=\lfloor (M+1)/2 \rfloor}^{M} q_{N,j} \leq \prod_{j=\lfloor (M+1)/2 \rfloor}^{M} \left(1 - \left(1 - \frac{2}{M}\right)^{N}\right)$$

$$\leq \left(1 - \left(1 - \frac{2}{M}\right)^{N}\right)^{M/2} \leq \left(1 - \exp\left(-c\frac{2N}{M}\right)\right)^{M/2}$$

$$\leq \exp\left(-\frac{M}{2}\exp\left(-c\frac{2N}{M}\right)\right),$$

for some constant c > 0 (c = 2 for instance, since $M \ge 3$).

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