# Diffusion limited aggregation in the layers model

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#### Abstract

### 1 Introduction

Diffusion limited aggregation. In the classical model of Diffusion Limited Aggregation (DLA), introduced by Witten and Sander [10], [11], the process begins with a single particle cluster placed at the origin of a space, and then, one-at-a-time, particles make a random walk "from infinity" until they collide with, and stick to, the existing cluster. The process is particularly natural in Euclidean space with particles making Brownian motion, or on d-dimensional lattices. Simulations of DLA in two dimensions show tree-like figures with long branches, and Kesten [6] proved that for lattices, if d = 2 the length of these arms is at most order  $n^{2/3}$  and if d > 2, at most  $n^{2/d}$ . A similar process, Reaction Limited Aggregation (RLA), differs in that particles stick with probability less than one on each collision. Indicative publications on RLA include Ball et al [2] or Meakin and Family [9].

A distinct but related process, Internal Diffusion Limited Aggregation (IDLA), was introduced by Diaconis and Fulton [3], as a protocol for recursively building a random aggregate of particles. In IDLA particles are added to the source vertex of an infinite graph, and make

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a random walk (over occupied vertices) until they visit an unoccupied vertex at which point they halt. Thus the first particle occupies the source, and subsequent particles stick to the outside of the component rooted at the source. Hence IDLA is the opposite of DLA, in that the particles walk in the occupied vertices until they find a space, rather than walking until they collide with the occupied component. The focus in IDLA has been on the limiting shape of the component formed by the occupied vertices. The formative work by Lawler et al [7], proved that, on d-dimensional lattices the limiting shape approaches a Euclidean ball; a result subsequently refined in [1], [5] and [8], amongst others.

As pointed out by Kesten in Remark 3 of [6], the concept of releasing a particle "from infinity" is somewhat informal. In [6] this is circumvented by using the limiting distribution of the first visit to the boundary of the current component from far away, conditional on making such a visit. An alternative approach is to attach a source vertex to a large but finite graph with a designated sink vertex. This analogue of DLA for finite graphs, was previously studied by Frieze and Pegden [4] for the Boolean lattice  $\mathcal{B} = \{0,1\}^n$ . In [4] the process evolves at discrete time steps  $t = 0, 1, \ldots$ ; each of which has an associated cluster  $C_t$ . The initial cluster  $C_0$  consists of a single vertex  $\mathbf{0} = (0, \ldots, 0) \in \mathcal{B}$ , the sink vertex. The cluster  $C_t$  is produced from  $C_{t-1}$  by choosing a random decreasing walk  $\rho_t$  from the source vertex  $\mathbf{1} = (1, \ldots, 1)$ . Let v be the last vertex of the initial segment of  $\rho_t$  which is disjoint from  $C_{t-1}$ , and set  $C_t = C_{t-1} \cup \{v\}$ . The process terminates at the first time  $t_f$  when  $\mathbf{1} \in C_{t_f}$ . The current paper continues this analysis of DLA on finite graphs by considering trees, and other graphs with a layered structure.

The layers model. Let  $S_0, S_1, \ldots, S_k, S_{k+1}$  be disjoint sets of vertices, where  $S_i$  is connected to  $S_{i+1}$  by a complete bipartite graph, with edges directed from level i to level i+1, to form a layered graph G. The set  $S_i$ , at distance i from the source, forms layer i of G, at level i. We assume  $S_0$  contains a single vertex v, the source vertex of the particles, and  $S_{k+1}$  contains a single vertex z the sink vertex of the particle cluster. Initially z is the only occupied vertex. The sink vertex z is an artificial vertex added to ensure connectivity of  $C_t$ , and to maintain compatibility with the hypercube model of [4].

Let  $N_i = |S_i|$  be the size of set  $S_i$ . For i = 1, ..., k, the sizes of  $S_i$ , are either taken to be the same size,  $N_i = n/k$  (the equal layers model), growing geometrically so that  $N_i = d^i$  (the growing layers model).

To simplify notation we assume that in the equal layers model G is an (n+1)-vertex graph with layer sizes  $N_i = n/k$  for i = 1, ..., k; and that in the growing layers model,  $d^k = n$  and thus G is a  $(d^{k+1}-1)/(d-1)$ -vertex graph. Similarly for trees of height k and branching factor d, we assume that the final level is of size  $d^k = n$ .

**DLA in the layers model.** Initially, at step t = 0 all vertices of G are unoccupied. A step t consists of placing an active particle on the (unoccupied) source vertex and moving the particle forward level by level according to a random walk  $\rho_t$  from source to sink until it halts. To move forward from level i < k, the active particle currently at vertex u in level i chooses a random neighbour w in level i + 1. If the neighbour w is unoccupied the particle moves to w. If the neighbour w is occupied, the particle halts at vertex u in level i, occupies that vertex and adds a directed edge (u, w) to the edge induced component  $C_t$  rooted at the sink. As the sink z is occupied at the start, a particle reaching level i + 1 = k halts there.

At any step, the component  $C_t$  is a rooted tree with edges directed towards the root, the sink z. The process stops at a step  $t_f$ , the finish time, when the source vertex v at level zero is occupied by a halted particle. In which case there is a directed path connecting the source v to the sink z, all of whose vertices are occupied by halted particles.

On deletion of the sink z, the digraph  $D_t = C_t \setminus \{z\}$  consists of arborescences rooted at the occupied vertices in level k. Let  $v = u_0 u_1 \cdots u_k u_{k+1} = z$  be the connecting path, the path connecting source and sink at the end of the process. Perhaps the most surprising thing is that in the digraph  $D_{t_f}$ , with high probability the arborescence rooted at  $u_k$  containing the connecting path is precisely the connecting path  $v = u_0 u_1 \cdots u_k$ , and contains no other vertices.

**Theorem 1.** Let  $v = u_0 u_1 \cdots u_k u_{k+1} = z$  be the path, connecting source and sink at the end of the process. Let  $w = u_k$  be the vertex at level k on this path. The following results hold with high probability.

1. Equal layers model. For i = 1, ..., k let  $N_i = n/k$ . Let

$$T_f = [(k+1)!(n/k)^k]^{1/(k+1)} = n^{k/(k+1)}\gamma_k$$

where  $\gamma_k = O(1)$  and  $\gamma_k \to 1/e$  as  $k \to \infty$ .

Provided  $1 \le k \le \sqrt{(\log n)/(\log \log n)}$ , the finish time  $t_f$  satisfies  $T_f/\omega \le t_f \le \omega T_f$ , where  $\omega$  is any function tending to  $\infty$ .

2. Trees of branching factor d, and the growing layers model. For i = 0, ..., k let  $N_i = d^i$ . Let

$$T_f = \sqrt{k} d^{k+3/2-\sqrt{2k+2}}.$$

(a) Provided  $d \to \infty$ ,  $k \to \infty$ , (and  $k \ll d$  for trees) the finish time  $t_f$  satisfies  $T_f/(\omega d^{O(1/\sqrt{k})}) \le t_f \le \omega T_f$ .

(The extra term  $d^{O(1/\sqrt{k})}$  is there to deal with some rounding error in  $j^*$ .)

(b) At  $t_f$ , levels  $i = 1, ..., k - \lceil \sqrt{2k+2} - 1 \rceil$  contain a single occupied vertex, the vertex  $u_i$  of the the connecting path.

3. On termination, let  $v = u_0 u_1, \dots u_k z$  be the path connecting source v and sink z in  $C_{t_f}$ , and let  $D_{t_f} = C_{t_f} \setminus \{z\}$ . In either model, with high probability, the vertices  $\{u_1, \dots, u_k\}$  have in-degree one in  $C_{t_f}$ . Thus the arborescence rooted at  $u_k$  in  $D_{t_f}$  is exactly this connecting path.

The proof of Theorem 1 also gives an upper bound on the finish time of DLA for trees of branching factor at least two.

### Corollary 2. Let

$$T = d^{k+2-\sqrt{2k+2}+(1/2)\log k}$$

For  $d \geq 2$ , the expected finish time of DLA on a tree of branching factor d is O(T).

**Notation.** We use  $A_n \sim B_n$  to denote  $A_n = (1 + o(1))B_n$  and thus  $\lim_{n\to\infty} A_n/B_n = 1$ . We use  $\omega$  to denote a quantity which tends to infinity with n more slowly than any other variables in the given expression. The expression  $f(n) \ll g(n)$  indicates f(n) = o(g(n)). The inequality  $A \lesssim B$  which stands for  $A \leq (1 + o(1))B$  is used to unclutter notation in some places.

## 2 Bounds on occupancy in the layers model

For  $t \geq 0$ , let  $L_i(t)$  be the number of particles halted in level i at the end of step t. Thus  $L_i(0) = 0$  for all  $i \leq k$ , and  $t = \sum_{i=0}^k L_i(t)$ . We refer to  $L_i$  as the occupancy of level i. Note that  $L_k(t) \leq t$ , and  $L_i(t) \leq \min(t, N_i)$ .

The first step is to prove the following proposition.

#### Proposition 3. Let

$$\mu_{k-j}(t) = \frac{1}{N_k N_{k-1} \cdots N_{k-j+1}} \frac{t^{j+1}}{(j+1)!}.$$
 (1)

Assume  $t < t_f$ , and thus the process has not stopped. Provided  $\mu_{k-\ell}(t) \to \infty$  for all  $\ell \leq j$ , then  $L_{k-j}(t) \sim \mu_{k-j}(t)$  in all models.

Let  $\mathcal{H}(t) = (L_0(t), L_1(t), \dots, L_k(t))$  be the history of the process up to and including step t.

$$\mathbb{E}\left(L_{i}(t+1) \mid \mathcal{H}(t)\right) = L_{i}(t) + \frac{L_{i+1}(t)}{N_{i+1}} \prod_{j=0}^{i} \left(1 - \frac{L_{j}(t)}{N_{j}}\right), \quad i < k,$$
 (2)

$$\mathbb{E}(L_k(t+1) \mid \mathcal{H}(t)) = L_k(t) + \prod_{j=0}^k \left(1 - \frac{L_j(t)}{N_j}\right).$$
 (3)

Note that (3) follows from (2) as  $L_{k+1}(t)/N_{k+1} = 1$  for all t, and that if  $L_0(t) = 1$ , the above recurrences give  $L_i(t+1) = L_i(t)$ .

To simplify matters for small t, we can condition on the event that  $L_k(s) = s$  for  $s \leq \sqrt{N_k/\omega}$ , an event of probability  $1-1/\omega$ , and use the fact that the evolution of  $(L_0(t), L_1(t), \ldots, L_k(t))$  is Markovian. Whatever happens,  $L_i(t)$  is monotone non-decreasing in t.

### 2.1 Upper bound on occupancy at step t

The underlying random walk  $\rho_t$ , from source v to sink z defines a (v, z)-path given by  $v = u_0 u_1, ..., u_k z = v_{k+1}$ . Particle t follows this walk until halting at a vertex  $u_i$ , where  $u_{i+1}$  is the first occupied vertex encountered on the path.

Let  $B_i(t)$  denote the occupied (blocked) vertices in level i at time t. We define an upperblocked process which we use to upper bound  $L_i(t)$ . This process gives rise to sets  $\widehat{B}_i(t) \supseteq B_i(t)$  and random variables  $\widehat{L}_i(t) = |\widehat{B}_i(t)| \ge L_i(t)$ . For every vertex  $u_j$ ,  $j \le k$  on the walk  $\rho_t$ , if  $u_{j+1}$  is occupied add a vertex to  $\widehat{B}_j(t)$  as follows. If  $u_j \notin \widehat{B}_j(t)$  add  $u_j$  to  $\widehat{B}_j(t+1)$ . If  $u_j \in \widehat{B}_j(t)$  add some other  $u'_j \in S_j \setminus \widehat{B}_j(t)$  to  $\widehat{B}_j(t+1)$ .

In particular if particle t+1 halts at vertex  $u_i$  in the DLA process, then either  $u_i$  is added to both  $B_i(t+1)$  and  $\widehat{B}_i(t+1)$ , or  $u_i$  is already a member of  $\widehat{B}_i(t)$ . In either case  $B_i(t) \subseteq \widehat{B}_i(t)$  for all i and  $t \geq 0$ . It follows (among other things) that  $L_i(t) \leq \widehat{L}_i(t)$ , and  $\widehat{L}_k(t) = t$ . The next lemma gives w.h.p. bounds for  $\widehat{L}_i(t)$ .

**Lemma 4.** Let  $\mu_{k-j}(t)$  as be given by (1).

1. The expectations  $\mathbb{E} \widehat{L}_i(t)$  satisfy the recurrence

$$\mathbb{E}\left(\widehat{L}_i(t+1) \mid \widehat{H}(t)\right) = \widehat{L}_i(t) + \frac{\widehat{L}_{i+1}(t)}{N_{i+1}}.$$
(4)

- 2. If  $j^2/t = o(1)$  then  $\mathbb{E} \widehat{L}_{k-j}(t) \sim \mu_{k-j}(t)$ .
- 3. If  $t \ge t_{k-j}(\omega)$  as defined in (12) then w.h.p.  $\widehat{L}_{k-j}(t) \sim \mu_{k-j}(t)$ .

*Proof.* Equation (4) follows because the upper bound process increases the size of  $\widehat{B}_{j}(t)$  whenever the walk  $\rho_t$  contains a vertex of  $\widehat{B}_{j+1}(t)$ , this being true at all levels j = 0, ..., k. We have

$$\mathbb{E}(L_i(t+1) \mid \mathcal{H}(t)) \le L_i(t) + \frac{L_{i+1}(t)}{N_{i+1}} \le \widehat{L}_i(t) + \frac{\widehat{L}_{i+1}(t)}{N_{i+1}} = \mathbb{E}(\widehat{L}_{i+1}(t+1) \mid \widehat{\mathcal{H}}(t)).$$

Iterating (4) backwards for  $0 \le s \le t$ , and using  $\widehat{L}_i(0) = 0$ , gives

$$\mathbb{E}\,\widehat{L}_i(t) = \frac{1}{N_{i+1}} \sum_{s=0}^{t-1} \mathbb{E}\,\widehat{L}_{i+1}(s). \tag{5}$$

We claim for  $j \geq 0$  that

$$\mathbb{1}_{\{t \ge j\}} \frac{(t-j)^{j+1}}{(j+1)!} \le (N_k N_{k-1} \cdots N_{k-j+1}) \mathbb{E} \widehat{L}_{k-j}(t) \le \frac{t^{j+1}}{(j+1)!}.$$
 (6)

For given t, the induction is backwards on k-j from j=0. Now  $N_{k+1}=1$  and  $\widehat{L}_{k+1}(t)=1$  which implies that  $\widehat{L}_k(s)=s$  for  $0 \le s \le t$ . So, (6) is true when j=0 and the first non-trivial case is j=1. From (5) we see that

$$\mathbb{E}\,\widehat{L}_{k-1}(t) = \frac{1}{N_k} \sum_{s=0}^{t-1} s,\tag{7}$$

which illustrates how (6) arises from bounding this sum.

For the general induction put i = k - (j+1), and multiply (5) by  $M_{j-1} = N_k N_{k-1} \cdots N_{k-j+1}$ . Insert (6) with i + 1 = k - j into this, to give

$$\frac{1}{N_{k-j}} \sum_{s=j}^{t-1} \frac{(s-j)^{j+1}}{(j+1)!} \le M_{j-1} \mathbb{E} \widehat{L}_{k-(j+1)}(t) \le \frac{1}{N_{k-j}} \sum_{s=1}^{t-1} \frac{s^{j+1}}{(j+1)!}$$
(8)

By comparison of the sum with the related integral we have that

$$\frac{(t-1)^{m+1}}{m+1} \le 1^m + 2^m + \dots + (t-1)^m \le \frac{t^{m+1}}{m+1}.$$
 (9)

Use (9) in (8) with m = j + 1, giving

$$\frac{\mathbb{1}_{\{t \ge j+1\}}}{N_{k-j}} \frac{(t-(j+1))^{j+2}}{(j+2)!} \le M_j \mathbb{E} \widehat{L}_{k-(j+1)}(t) \le \frac{1}{N_{k-j}} \frac{t^{j+2}}{(j+2)!},$$

which completes the induction for (6). Moreover, provided  $j^2/t = o(1)$ ,

$$\mathbb{E}\,\widehat{L}_{k-j}(t) = \frac{1}{N_k N_{k-1} \cdots N_{k-j+1}} \frac{t^{j+1}}{(j+1)!} \left(1 - O\left(j^2/t\right)\right) = \mu_{k-j}(t)(1+o(1)). \tag{10}$$

This completes the proof of Lemma 4.(1) and Lemma 4.(2). Note that  $t_{k-j}(\omega) \gg (n/k)^{1/k} \gg k^2$  for  $j \geq 1$ , ensuring that  $j^2/t = o(1)$  for j, t of interest.

# 2.2 Concentration of $\widehat{L}_i(t)$ for sufficiently large t.

We now prove Lemma 4.(3).

By definition,  $\widehat{L}_k(t) = t = \mu_k(t)$ , establishing Lemma 4.(3) for j = 0.

We next consider  $\widehat{L}_{k-j}(t)$  for  $1 \leq j \leq k-1$ . The random variable  $\widehat{L}_{k-j}(t)$  is obtained by choosing a random vertex  $u \in S_{k-j}$  irrespective of the current occupancy of u, and choosing a random neighbour  $w \in S_{k-j+1}$ . If w is occupied then  $\widehat{L}_{k-j}(t+1) = \widehat{L}_{k-j}(t) + 1$ . Thus,  $\widehat{L}_{k-j}(t+1) = \widehat{L}_{k-j}(t) + Q_{k-j}(t)$  where  $\mathbb{P}(Q_{k-j}(t) = 1) = (\widehat{L}_{k-j+1}(t)/N_{k-j})$  independently of any previous outcomes.

In particular,  $\mathbb{E} Q_{k-1}(t) = t/N_k$ , and by equation (7),  $\mathbb{E} \widehat{L}_{k-1}(t) = t(t-1)/(2N_k) \sim \mu_{k-1}(t)$ . By Hoeffding's Inequality,

$$\mathbb{P}\left(\widehat{L}_{k-1}(t) \notin \mu_{k-1}(t) \left(1 \pm \frac{1}{\omega}\right)\right) \le 2 \exp\left\{-\frac{(1-o(1))t^2}{6\omega^2 N_k}\right\} \le 2e^{-\omega} \tag{11}$$

if  $t \geq 2\omega^{3/2}\sqrt{2N_k}$ .

For  $j \ge 0$  we proceed as follows: let  $t_1 = t_1(k-j)$  be such that  $\mu_{k-j}(t_1) = 1$ . Thus  $t_1(k) = 1$  and

$$t_1(k-j) = [(j+1)!N_kN_{k-1}\cdots N_{k-j+1}]^{1/(j+1)}.$$

The fact that  $t_1(k-j+1) \ll t_1(k-j)$  so that  $\mu_{k-j+1}(t_1(k-j))$  is sufficiently large, is a model dependent calculation given in subsequent sections. The statement of Lemma 4 assumes the truth of this.

Let 
$$t_k(\omega) = 1$$
, and for  $j \ge 1$ , let  $t_{k-j}(\omega) = (4\omega^3)^{1/(j+1)} t_1(k-j)$  so that 
$$t_{k-j}(\omega) = (4\omega^3)^{1/(j+1)} [(j+1)! N_k N_{k-1} \cdots N_{k-j+1}]^{1/(j+1)}, \tag{12}$$

and  $\mu_{k-j}(t_{k-j}(\omega)) = 4\omega^3$ .

Let  $\mathcal{E}_j$  denote the event that  $\widehat{L}_{k-j}(t) \in \mu_{k-j}(t)(1 \pm 1/\omega)$  for  $t \geq t_{k-j}(\omega)$ . We claim that  $\mathbb{P}(\mathcal{E}_j) \geq 1 - 2je^{-\omega}$  and we verify this by induction.

For j = 0,  $\mathbb{P}(\mathcal{E}_0) = 1$ , and for j = 1,  $\mathbb{P}(\mathcal{E}_1) \ge 1 - 2e^{-\omega}$  follows from equation (11). For  $j \ge 2$ , by Hoeffding's inequality, we have that for  $t \ge t_{k-j}(\omega)$ ,

$$\mathbb{P}(\neg \mathcal{E}_j \mid \mathcal{E}_{j-1}) \le 2 \exp\left\{-\frac{\mu_{k-j}(t)}{3\omega^2} \left(1 - \delta\right)\right\} \le 2 \exp\left\{-\frac{\mu(t_{k-j}(\omega))}{3\omega^2} \left(1 - o(1)\right)\right\} \le 2e^{-\omega},$$

where  $\delta = O(1/\omega + j^2/t)$ , includes the correction from (10).

We complete the induction, on the assumption that  $\log k \ll \omega$ , via

$$\mathbb{P}(\neg \mathcal{E}_j) \le \mathbb{P}(\neg \mathcal{E}_j \mid \mathcal{E}_{j-1}) + \mathbb{P}(\neg \mathcal{E}_{j-1}) \le 2e^{-\omega} + 2(j-1)e^{-\omega} = 2je^{-\omega} = o(1). \tag{13}$$

### 2.3 Lower bound on occupancy at step t

A lower bound  $\widetilde{L}_i(t)$  on  $L_i(t)$  can be found as follows. We consider a process in which a particle halts at level i whenever it chooses an out neighbour in  $\widehat{B}_{i+1}(t)$ . Let  $L_j^*(t)$  bound  $\widehat{L}_j(t)$  from above w.h.p. Let  $\widetilde{L}_i(t)$  be obtained by replacing L by  $L^*$  in in the bracketed terms on the RHS of (2)–(3). Then  $\widetilde{L} \leq L \leq \widehat{L}$  w.h.p., and we have the following recurrences.

$$\mathbb{E}\widetilde{L}_i(t+1) = \mathbb{E}\widetilde{L}_i(t) + \frac{\mathbb{E}\widetilde{L}_{i+1}(t)}{N_{i+1}} \prod_{j=0}^i \left(1 - \frac{L_j^*(t)}{N_j}\right), \tag{14}$$

$$\mathbb{E}\widetilde{L}_k(t+1) = \mathbb{E}\widetilde{L}_k(t) + \prod_{j=0}^k \left(1 - \frac{L_j^*(t)}{N_j}\right). \tag{15}$$

The solution to these recurrences is given in Section 3.2 for the equal layers model, and in Section 4 for the growing layers model.

## 3 Analysis of DLA in the equal layers model

The first subsection below shows that provided k is not too large, then w.h.p. there exists a large gap in the number of steps between the time for which  $\mathbb{E} \widehat{L}_{k-j} = 1$  with all lower values zero, and the time for which  $\mathbb{E} \widehat{L}_{k-(j+1)} = 1$ . This gap allows  $\widetilde{L}_{k-j}$  to increase and become concentrated, whilst at the same time all lower values remain zero w.h.p. This allows us to conclude that provided  $\mu_i(t)$  is sufficiently large,  $L_i(t)$  is concentrated around  $\mu_i(t)$ . We use this to conclude that  $L_0(t) = 0$  until  $L_1(t) \to \infty$  suitably fast, and thus estimate the finish time of the DLA process in the equal layers model.

# 3.1 Evolution of the state vector $\widehat{L}$ in the equal layers model.

Let  $\widehat{L} = (\widehat{L}_0, \widehat{L}_1, \dots, \widehat{L}_k)$  be the state vector of the upper-blocked process. The entries in  $\widehat{L}$  are non-negative integers, and if  $\widehat{L}_i = 0$ , then  $\widehat{L}_{i-1} = 0$ . Let

$$T_f = [(k+1)!N_1 \cdots N_k]^{1/(k+1)},$$
 (16)

where  $N_1 = \cdots = N_k = n/k$ . It will be shown below that  $T_f$  is asymptotically the expected finish time. The following argument for  $t \leq \omega T_f$  proves there is a large enough gap t'' - t between  $\mu_{k-j}(t) = 1$  and  $\mu_{k-(j+1)}(t'') = 1$  for  $\widehat{L}_{k-j}(t'')$  to be concentrated.

Define  $\beta = \beta_k = N_k/\omega T_f$ . As

$$[(k+1)!]^{1/(k+1)} = k\Theta(1+k^{3/2k}) = k\Theta(1),$$

then

$$\beta = \frac{N_k}{\omega T_f} = \frac{n/k}{\omega [(k+1)!(n/k)^k]^{1/(k+1)}} = \Theta(1) \frac{1}{\omega k} \left(\frac{n}{k}\right)^{1/(k+1)}.$$
 (17)

We can assume that  $\beta \geq (\log n)^{k/2}$ . This will be true if  $\omega \leq \log n$  since we assume that  $k \leq \sqrt{(\log n)/\log \log n}$ .

We have for any  $t \leq \omega T_f$ , using that  $N_{k-j} = N_k = n/k$  in the equal layers model, that

$$\mu_{k-(j+1)}(t) = \frac{t}{(j+2)N_{k-j}} \cdot \mu_{k-j}(t) \le \frac{1}{j\beta}\mu_{k-j}(t), \tag{18}$$

and similarly

$$\mu_{k-(j+\ell)}(t) = \frac{t}{(j+\ell+1)N_{k-\ell-1}} \cdot \mu_{k-(j+\ell)-1}(t) \le \frac{1}{(j\beta)^{\ell}} \mu_{k-j}(t), \quad \text{for } 1 \le \ell \le k-j.$$
 (19)

Consider  $\mathbb{E} \widehat{L}_{k-j}(t)$ . When  $\mu_{k-j}(t) \sim 1$ , then  $\mathbb{E} \widehat{L}_{k-j}(t) \sim 1$  and the Markov inequality implies that w.h.p.  $\widehat{L}_{k-j}(t) \in I_{\omega} = [0, 1, ..., \omega]$ . From (18)–(19), when  $\mu_{k-j}(t) \sim 1$ ,

$$\mu_{k-(j+\ell)}(t) \le \frac{1+o(1)}{(j\beta)^{\ell}}, \quad \text{for } 1 \le \ell \le k-j,$$
 (20)

and thus w.h.p.  $\widehat{L}_0 = 0, \widehat{L}_1 = 0, \dots, \widehat{L}_{k-(j+1)} = 0.$ 

By arguments similar to that for (18), (19), we see that

$$\mu_{k-j+\ell}(t) \ge \mu_{k-j}(t) \ j(j-1)\cdots(j-\ell+1) \beta^{\ell}$$
 for  $\ell \le j$ .

So if  $\mu_{k-j}(t) \sim 1$  then for  $\ell \leq j$ ,  $\mu_{k-j+\ell}(t) \to \infty$  and  $\widehat{L}_{k-j+\ell}$  is equal to  $(1+o(1))\mathbb{E}\widehat{L}_{k-j+\ell}$  w.h.p. (As already remarked at the end of Section 2.1,  $k^2/t = o(1)$  for t of interest here and we can invoke Lemma 4.) Thus at times t such that  $\mu_{k-j}(t) \sim 1$  (implying that  $t \leq \omega T_f$ ), w.h.p., the state vector  $\widehat{L}$  is such that  $\mu_{k-\ell} \to \infty$  for  $\ell \leq j-1$ , and

$$\widehat{L}(t) = (0, \dots, 0, \ \widehat{L}_{k-j} \in I_{\omega}, \ (1 + o(1))\mu_{k-j+1}, \ (1 + o(1))\mu_{k-j+2}, \dots, \ (1 + o(1))\mu_k).$$
 (21)

Define t' > t so that  $\mathbb{E} \widehat{L}_{k-j}(t') \sim \mu_{k-j}(t') = \omega$  where  $\omega \to \infty$  but  $\omega \ll \beta$ . It follows from (1) that  $t' \sim \omega^{1/(j+1)}t$ . Using  $T_f = \gamma_k n^{k/(k+1)}$  from Proposition 5.1, for j < k we have  $t' \le \omega T_f$  which follows from  $t' = \Theta(1)\omega^{1/(j+1)}N_k^{j/(j+1)} \ll \omega T_f$  for  $\omega$  growing sufficiently slowly,

$$\mu_{k-(j+1)}(t') = \frac{t'}{(j+2)N_k} \omega \le \frac{\omega}{\beta}, \qquad \mu_{k-j}(t') = \omega, \qquad \mu_{k-j+1}(t') = \frac{(j+1)N_k}{t'} \omega \ge \omega\beta.$$

Choosing  $\log \log n \le \omega \ll \beta$ , we have that w.h.p.  $\widehat{L}_{k-(j+1)}(t') = 0$  as before and  $\mu_{k-j+1} \ge \omega \log n$ . But now

$$\widehat{L}(t') = (0, ..., 0, (1 + o(1))\mu_{k-j}, (1 + o(1))\mu_{k-j+1}, ..., (1 + o(1))\mu_k).$$
(22)

This condition persists until t'' such that  $\mu_{k-(j+1)}(t'') \sim 1$ . Equating  $\mu_{k-j}(t) \sim 1 \sim \mu_{k-(j+1)}(t'')$  and using  $t \leq \omega T_f$  gives

$$(t'')^{j+2} \sim t^{j+1}(j+2)N_k \implies t'' \ge t\left(\frac{N_k}{t}\right)^{1/(j+2)} \ge t\beta^{1/k} \ge t(\log n)^{1/2},$$

by our choice of  $\beta$ . We conclude that  $t'' \geq (\log n)^{1/2}t$ , so the claimed gap exists.

### 3.2 Lower bound on occupancy in the equal layers model.

Let  $L_j^*(t) = \omega \mu_j(t)$ . By (21), (22) of the previous section, this is a w.h.p. upper bound on  $\widehat{L}_j(t)$  for  $t \leq \omega T_f$ . Referring to (14)–(15), we claim that

$$\prod_{i=0}^{i} \left( 1 - \frac{L_j^*(t)}{N_j} \right) \ge 1 - \sum_{i=0}^{i} \left( \frac{L_j^*(t)}{N_j} \right) \ge 1 - \frac{iL_i^*(t)}{N_i} = 1 - o(1).$$
 (23)

Note first that as  $T_f = O(kn^{k/(k+1)})$  and  $N_k = n/k$ , then provided  $k \leq \sqrt{\log n/\log \log n}$  and  $\omega$  grows sufficiently slowly,

$$\frac{k\omega^2 T_f}{N_k} = O\left(\frac{\omega^2 k^{2+1/(k+1)}}{n^{1/(k+1)}}\right) = o(1); \tag{24}$$

and as  $N_{k-j+1} = N_k$  and  $\mu_{k-j}(t)/\mu_{k-j+1}(t) = t/(j+1)N_{k-j+1} = o(1)$  by (24), it follows that

$$\mu_{k-j}(t)/N_{k-j} \le t/N_k. \tag{25}$$

Thus (23) follows from (24), (25) and

$$\frac{jL_j^*(t)}{N_j} = \frac{j\omega\mu_j(\omega T_f)}{N_j} \le \frac{k\omega\mu_k(\omega T_f)}{N_k} \le \frac{k\omega^2 T_f}{N_k} = o(1). \tag{26}$$

Let  $t_i(\omega)$  be given by (12). For those  $i \leq k$ , and  $t \geq t_i(\omega)$ , so that  $\mu_i(t) \to \infty$  suitably fast, we obtain that  $\mathbb{E} \widetilde{L}_i(t) \sim \mu_i(t) \sim \mathbb{E} \widehat{L}_i(t)$  and thus w.h.p.,

$$\mathbb{E} L_{k-j}(t+1) \sim \mathbb{E} \widehat{L}_{k-j}(t+1) \sim \frac{t^{j+1}}{(j+1)!} \frac{1}{N_k N_{k-1} \cdots N_{k-j+1}}.$$
 (27)

As  $L_i(t) \leq \widehat{L}_i(t)$  by construction, and w.h.p.  $\widehat{L}_i(t) \leq (1 + o(1))\mathbb{E}\widehat{L}_i(t)$ , this implies  $L_i(t) \leq (1 + o(1))\mu_i(t)$  w.h.p., if  $\mu_i(t) \to \infty$ .

At step t + 1, equation (23) implies that, in the DLA-process, particle t + 1 arrives at level i with probability (1 - o(1)). It halts at this level with probability  $L_{i+1}(t)/N_{i+1} = o(1)$ , (see (2), (25)), and thus w.h.p. all but t(1 - o(1)) particles arrive at level k, the o(1) term being given in (26).

It is an induction on  $\ell = k - 1, \ldots, i$  in that order, that provided  $t \geq t_{\ell}(\omega)$ , then  $\mu_{\ell}(t) \rightarrow \infty$  sufficiently fast, the values  $L_k(t), L_{k-1}(t), \ldots, L_i(t)$  are concentrated in the lower tail, according to the arguments in Section 2.2. Thus  $L_i(t) \sim \mu_{i+1}(t)$  w.h.p..

In the case that  $\mathbb{E} \widehat{L}_i(t) = o(1)$  then a fortiori  $L_i(t) = 0$  w.h.p., and if  $\mathbb{E} \widehat{L}_i(t) = O(1)$  then w.h.p.  $L_i(t) \leq \omega \mu_i(t)$ . We can thus argue that a gap theorem similar to that in Section 3.1 holds for the state vector  $L = (L_0, L_1, \ldots, L_k)$ .

In particular, and most importantly, let  $t_1 = t_1(0)$  be such that  $\mu_0(t_1) = 1$ , and thus  $t_1 \sim T_f$ . Let  $t' = t_1/\omega$ . Then w.h.p.,  $L_0(t') = O(1/\omega)$  and  $\mathbb{E} L_1(t') \geq \beta/\omega$  where  $\beta$  is given in (17). It follows that for  $t \geq t'$ ,  $L_1(t)$  is concentrated around  $\mu_1(t)$ .

This completes the proof of Proposition 3 for the equal layers model.

### 3.3 Finish time of DLA in the equal layers model.

**Proposition 5.** For  $k \geq 2$ , let G be an equal layers graph with level sizes  $N_i = n/k$ , i = 1, ..., k. Let  $T_f$  be given by (16), then

- 1.  $T_f = n^{k/(k+1)}e^{-1}(1 + O(\log k/k))$
- 2. With high probability, the finish time  $t_f$  of the DLA process in G satisfies  $T_f/\omega \leq t_f \leq \omega T_f$ , where  $\omega \to \infty$  arbitrarily slowly.

*Proof.* Note that

$$\frac{(k+1)!}{k^k} \sim \frac{1}{k^k} \sqrt{2\pi} e^{-(k+1)} k^{k+3/2} (1+1/k)^{k+3/2} = e^{-(k+1)} \Theta_k,$$

where  $\Theta_k \sim (e^{1-1/k+O(1/k^2)}\sqrt{2\pi}k^{3/2})$ . Thus

$$T_f = \left[ (k+1)! N_1 \cdots N_k \right]^{1/(k+1)} = \left( \frac{(k+1)!}{k^k} n^k \right)^{1/(k+1)} = n^{\frac{k}{k+1}} e^{-1} (\Theta_k)^{1/(k+1)}, \tag{28}$$

where  $(\Theta_k)^{1/k+1} = e^{O(\log k/k})$  is bounded for  $k \geq 2$  and tends to one as  $k \to \infty$ .

The choice of  $T_f$  is such that  $\mathbb{E} \widehat{L}_0(T_f) \sim 1$ , indeed from (10)

$$\mathbb{E}\,\widehat{L}_0(T_f) = 1 + O\left(\frac{k^2}{T_f}\right) = 1 + O\left(\frac{k^2}{n^{k/(k+1)}}\right),\tag{29}$$

which is 1 + o(1), provided  $k = o(\sqrt{n})$  say. Moreover  $\mathbb{E} \widehat{L}_0(T_f/\omega) \sim 1/\omega^{k+1}$  proving that  $t_f = \Omega(T_f)$  w.h.p.

The value of  $\mathbb{E} \widehat{L}_{k-j}(t)$  is monotone decreasing with increasing j. This follows from (18) because  $T_f = o(N_k)$ , making  $\beta$  large, see (28). We next investigate the concentration of  $L_1(T_f)$ , which continues the discussion in the last paragraph of the previous section.

$$\mathbb{E} L_1(T_f) \sim \mu_1(T_f) \sim \frac{(T_f)^k}{k!(n/k)^{k-1}} = \frac{1}{k!} [(k+1)!)^k (n/k)]^{\frac{1}{k+1}} = \Theta(n^{1/(k+1)} k^{-3/2(k+1)}). \quad (30)$$

For concentration of  $\widehat{L}_1(T_f)$  we require  $\mathbb{E} L_1(T_f) \to \infty$ , which occurs whenever  $k = o(\log n)$ . At this point in the process, even if  $L_0(T_f) = 0$ , the expected waiting time for a particle to hit  $L_1(T_f)$  is  $O(n^{k/(k+1)}/k) = O(T_f)$ , so w.h.p. this will occur by time  $\omega T_f$ , completing the proof of Proposition 5, and hence Theorem 1.1.

### 3.4 Existence of a unique connecting path component.

We regard occupied vertices as coloured either red or blue, with all occupied vertices in level k coloured blue.

A particle halts at vertex u in level i, if it chooses an edge uw to an occupied neighbour w in level i+1. We consider this edge uw as being directed from u to w in the component rooted at the sink. If u is the first in-neighbour of w then u is coloured blue. If however w already has an in-neighbour u', then u, u' and all other in-neighbours are (re)coloured red. At any step, the red vertices in a level are those with siblings, and the blue ones are the unique in-neighbour of some vertex in the next level. The process halts when there is a directed path of occupied vertices  $v = u_0 u_1 \cdots u_k$  from the source to level k.

**Lemma 6.** With high probability, the path  $v = u_0u_1 \cdots u_k = w$  from the source to level k is blue and thus the in-arborescence of halted particles rooted at  $w = u_k$  is exactly this (v, w)-path.

*Proof.* As before, let  $B_i$  be the set of occupied vertices in level i, where  $L_i = |B_i|$ . As each  $u \in B_i$  has a unique out-neighbour, the subset  $Out(B_i)$  of  $B_{i+1}$  with at least one in-neighbour has size at most  $L_i$ .

Let  $\mathbb{1}_{\{k-j,s\}}$  be the indicator that particle s halts in level k-j and is coloured red due to a pre-existing sibling. In this case particle s has chosen an out-neighbour in the existing set  $\operatorname{Out}(B_{k-j}) \subseteq B_{k-j+1}$ , and thus

$$\mathbb{E} \mathbb{1}_{\{k-j,s\}} \leq \frac{\mathbb{E} \widehat{L}_{k-j}(s)}{(n/k)} \sim \frac{\mu_{k-j}(s)}{(n/k)}.$$

If  $Z_{k-j}(t)$  is the number of red vertices in level k-j, then

$$\mathbb{E} Z_{k-j}(t) \lesssim 2 \sum_{s=1}^{t} \frac{\mu_{k-j}(s)}{(n/k)} = \frac{2}{(n/k)} \sum_{s=1}^{t} \frac{s^{j+1}}{(j+1)!(n/k)^{j}} \sim 2 \frac{t^{j+2}}{(j+2)!(n/k)^{j+1}} = 2\mu_{k-(j+1)}(t). \quad (31)$$

The factor 2 covers the case where the pre-existing sibling was blue but is recoloured red.

Define a random variable  $\widehat{Z}_{k-j}(t) = Z_{k-j}(t) + Q_{k-j}(t) \geq Z_{k-j}(t)$ , where  $Q_{k-j}$  is Bernoulii  $Be(p_{k-j})$ , with  $p_{k-j} = (\widehat{L}_{k-j} - |\operatorname{Out}(B_{k-j})|)/(n/k)$ . As  $\mu_i \to \infty$  at  $t_f$  for all  $i \geq 1$ , for  $i \geq 2$ ,  $\widehat{Z}_i(t_f) = (1 + o(1))\mathbb{E}\widehat{Z}_i(t_f)$  w.h.p., since  $Z_i(t_f) \leq \widehat{Z}_i(t_f) \leq \widehat{L}_i(t_f)$ . On the other hand, by definition  $\mathbb{E}L_0(T_f) = 1$ , so by the Markov Inequality  $Z_1(t_f) \leq \widehat{Z}_1(t_f) \leq \omega$ .

Consider the colour of the path  $v = u_0 u_1 u_2 \cdots u_k$  at  $t \leq t_f$ . Vertex v is blue as it is the first neighbour of  $u_1$ . Let  $R_i(s)$  be the red vertices in level i at step s. Vertex  $u_1$  was chosen u.a.r. so, as  $\mu_0(T_f) \sim 1$  (see (29)),  $\mu_1(T_f) = \Theta(1) n^{1/(k+1)}$  (see (30)), and  $t_f \geq T_f/\omega$ ,

$$\mathbb{P}(u_1 \in R_1(t)) = \frac{Z_1(t)}{L_1(t)} \le \frac{\omega}{\mu_1(t)} \le \frac{\omega^2}{n^{1/(k+1)}}.$$
 (32)

Let  $s_1$  be the step at which  $u_1 = u_1(s_1)$  was occupied by a halted particle. This happened because  $s_1$  chose an occupied out-neighbour  $u_2 = u_2(s_2)$ , where  $s_2 < s_1$ . If  $u_2$  is red at step t, then either it became red before step  $s_1$ , or at some later step. As  $u_2$  was chosen u.a.r. by the particle  $s_1$  at  $u_1$  from among the  $L_2(s_1)$  occupied vertices in level 2,

$$\mathbb{P}(u_2 \in R_2(s_1)) = \frac{Z_2(s_1)}{L_2(s_1)} \lesssim 2\omega \frac{\mu_1(s_1)}{\mu_2(s_1)} = \frac{2\omega s_1}{k(n/k)} = \frac{2\omega s_1}{n}.$$

The factor of 2 is from (31) and  $\omega$  covers lack of concentration of  $\widehat{Z}_2$  or  $L_2$ , if  $s_1$  is early. If  $u_3$  is the chosen out-neighbour of  $u_2$ ,

$$\mathbb{P}(u_2 \in R_2(t) \backslash R_2(s_1)) \le \frac{1}{(n/k)} \sum_{\tau = s_1 + 1}^{t} \mathbb{E} \, \mathbb{1}_{\{u_3 \text{ chosen at } \tau\}} = \frac{t - s_1}{(n/k)}.$$

Thus

$$\mathbb{P}(u_2 \in R_2(t)) \le \frac{\omega kt}{n},$$

and similarly for  $u_3, \ldots, u_{k-1}$ . Thus

$$\mathbb{P}((v, u_k)\text{-path is blue}) \ge \prod_{i=1}^{k-1} \left(1 - \frac{\omega kt}{n}\right) = 1 - O\left(\frac{\omega k^2 t}{n}\right),$$

where  $t \leq t_f \leq \omega n^{k/(k+1)}$ . Provided  $\omega k^2/n^{1/(k+1)} = o(1)$ , which holds for  $3k \log k \leq n$ , the path from the source to vertex  $w = u_k$  level k is blue, and thus the in-arborescence root at w is exactly this path.

# 4 Analysis of DLA in the growing layers model

**Proposition 7.** Let G be a growing layers graph with level sizes  $N_i = d^i$ , for i = 0, ..., k. Let

$$T_f = \sqrt{k} \ d^{k+3/2 - \sqrt{2k+2}}. (33)$$

Let  $j = \max\{i : (2k+2) \ge i(i+1)\}$ , and let  $j^* = \sqrt{2k+2} - 1$  be the solution to 2k+2 = i(i+1).

- 1. Provided  $d \to \infty$ ,  $k \to \infty$ , and  $j = j^*$  the finish time  $t_f$  satisfies  $T_f/\omega \le t_f \le \omega T_f$ , where  $\omega \to \infty$  arbitrarily slowly.
- 2. If  $j \neq j^*$  the error in finish time  $t_f$  above introduced by ignoring rounding is  $d^{-O(1/\sqrt{k})}$ , and thus  $T_f d^{-O(1/\sqrt{k})}/\omega \leq t_f \leq \omega T_f$ .
- 3. At  $t_f$ , levels  $i = 1, ..., k \lceil \sqrt{2k+2} 1 \rceil$  contain a single occupied vertex, the vertex  $u_i$  of the the connecting path.

*Proof.* The size  $N_i$  of layer i is defined as  $d^i$ . It follows that the product of the set sizes in the denominator of  $\mu_{k-j}(t)$  in (1) is given by

$$N_k N_{k-1} \cdots N_{k-j+1} = d^k d^{k-1} \cdots d^{k-j+1} = d^{kj-j(j-1)/2}$$

and thus (1) becomes

$$\mu_{k-j}(t) = \frac{t^{j+1}}{(j+1)! \ d^{kj-j(j-1)/2}}.$$
(34)

Assume d is sufficiently large. The upper bound  $\mathbb{E}\widehat{L}_{k-j}$  is obtained in Section 2. However, a problem can arise in the growing layers model in the upper bound calculations. The value of  $\mu_{k-j}$  can decrease with increasing j and then (anomalously) increase again. This is presumably because the recurrence used to establish it assumes  $\mu_{k-\ell} \to \infty$  for all  $\ell < j$ , which is not the case.

Let t be such that  $\mu_{k-j}(t) = 1$ , and let  $t_1 = t_1(k-j)$  be  $\lceil t \rceil$ , so that  $\mu_{k-j}(t) \sim 1$  at step  $t_1$ . From (34),

$$\mu_{k-j}(t_1) = \frac{t_1^{j+1}}{(j+1)! \ d^{kj-j(j-1)/2}} \sim 1 \quad \Longrightarrow \quad t_1 \sim [(j+1)!]^{1/(j+1)} \ d^{\frac{2kj-j(j-1)}{2(j+1)}}. \tag{35}$$

From (1)

$$\frac{\mu_{k-j+1}(t)}{\mu_{k-j}(t)} = \frac{(j+1)d^{k-j+1}}{t},$$

so setting  $\mu_{k-j}(t_1(k-j)) \sim 1$  gives

$$\mu_{k-j+1}(t_1) \sim \frac{j+1}{[(j+1)!]^{1/j+1}} d^{k-j+1-\frac{2kj-j(j-1)}{2(j+1)}}$$

$$\sim \frac{e}{(2\pi(j+1))^{1/2(j+1)}} d^{\frac{2k+2-j(j+1)}{2(j+1)}}.$$
(36)

The leading term on the RHS is bounded, and the exponent of d on the RHS is positive provided 2k + 2 > j(j + 1), which ensures that  $\mathbb{E} \widehat{L}_{k-j+1}(t)$  is sufficiently large close to  $t_1(k-j)$ .

What value of k-j maximizes the step  $t_1=t_1(k-j)$  at which  $\mu_{k-j}(t)\sim 1$ ? Write the exponent of d on the RHS of (35) as f(j)/2 where

$$f(j) = \frac{2kj - j(j-1)}{(j+1)} = (2k+2) - j - \frac{2k+2}{j+1}.$$

The maximum of f(j) occurs when  $(j^* + 1)^2 = (2k + 2)$ , giving  $f(j^*) = (j^*)^2$ . The (not necessarily integer) value of  $j^*$ ,  $k - j^*$  and  $d^{f(j^*)/2}$  are

$$j^* = \sqrt{2k+2} - 1, \qquad k - j^* = k + 1 - \sqrt{2k+2}, \qquad d^{f(j^*)/2} = d^{k+3/2 - \sqrt{2k+2}}.$$
 (37)

Integer solutions to  $(j^*+1)^2 = (2k+2)$  exist for some values of k. Let a be a positive integer, and put  $j^*+1=2a$  so that  $k=2a^2-1$ ; this holds for example for k=1,5,17,31. In the case that  $j^*$  is not integer, we address the rounding error at the end of this section.

Ignoring rounding for now, we evaluate  $t_1 = t_1(k-j)$ , at  $j = j^*$  where  $f(j) = j^2$ , to find

$$t_1 \sim [(j+1)!]^{1/(j+1)} d^{f(j)/2}$$

$$\sim e^{-1} (\sqrt{2\pi})^{1/(j+1)} (j+1)^{1+1/2(j+1)} d^{j^2/2}$$

$$= C_{k-j} \sqrt{2k+2} d^{k+3/2-\sqrt{2k+2}},$$
(38)

where  $C_{k-j} = O(1)$ . Note that  $t_1 = \Theta(T_f)$ . From (34),

$$\frac{\mu_{k-(j+1)}(t)}{\mu_{k-j}(t)} = \frac{t}{(j+2)d^{k-j}}$$

so that at  $t_1$ , for some C, C' = O(1),

$$\mu_{k-j^*}(t_1) \sim 1, \qquad \mu_{k-(j^*-1)}(t_1) = Cd^{1/2}, \qquad \mu_{k-(j^*+1)}(t_1) = C'd^{1/2}.$$

At first this seems confusing, as one would expect to have  $\mu_{k-(j^*-1)}(t_1) = Cd^{1/2} = o(1)$  here. But assuming  $d^{1/2} \to \infty$ ,  $\mathbb{E} \widehat{L}_{k-j^*+1} = C''d^{1/2}$ , which is the last value at which the condition  $\mu_{k-j+1} \to \infty$  is valid in the recurrence from k-j+1 to k-j; and is where the assumption in Proposition 3 breaks down.

Note that at  $t_1/\omega$ ,  $\mu_{k-j^*}(t_1/\omega) = O(\omega^{-(j^*+1)})$  from the definition of  $t_1(k-j^*)$ . Thus w.h.p.  $\widehat{L}_{k-j^*}(t_1/\omega) = o(1)$  in the upper process. Consequently all levels  $k-\ell$ ,  $\ell \geq j^*$ , are zero, w.h.p. in both the upper and the DLA process. In which case, as occupancy is monotone non-decreasing, this holds in the state vector L(t) of the DLA process, for all  $t \leq t_1/\omega$  and all  $\ell \geq j^*$ .

For  $\ell < j^*$ , as in the equal layers model we will use (23) to obtain (27). To do this we check conditions (24) and (25) for  $t \le t_f$ . Provided  $k \to \infty$ , but  $k = o((\log n / \log \log n)^2)$ ,

$$\frac{k\omega^2 T_f}{N_k} = \frac{\omega^2 k^{3/2} d^{k+3/2 - \sqrt{2k+2}}}{d^k} = \omega^2 k^{3/2} d^{3/2 - \sqrt{2k+2}} = o(1); \tag{39}$$

and as  $N_{k-j} = d^{k-j}$ , and  $j^* + 1 = \sqrt{2k+2}$ ,

$$\frac{\mu_{k-\ell}(t)}{\mu_{k-\ell+1}(t)} = \frac{t}{(\ell+1)d^{k-\ell+1}} \le \omega \sqrt{k} d^{1/2-\sqrt{2k+2}+\ell} \le \omega \sqrt{k} d^{1/2-\sqrt{2k+2}+j^*-1} = \omega \sqrt{k} d^{-3/2},$$

provided  $t \leq \omega T_f$ . Thus provided  $d > k\omega^2$ ,

$$\frac{\mu_{k-\ell}(t)}{N_{k-\ell}} \leq \frac{\omega \sqrt{k}}{d^{3/2}} \frac{\mu_{k-\ell+1}(t)}{N_{k-\ell}} = \frac{\omega \sqrt{k}}{d^{1/2}} \frac{\mu_{k-\ell+1}(t)}{N_{k-\ell+1}} \leq \frac{\mu_{k-\ell+1}(t)}{N_{k-\ell+1}}.$$

As  $\mu_k(t) = t$ , it follows that

$$\mu_{k-j}(t)/N_{k-j} \le t/N_k,\tag{40}$$

and thus (26) holds. As before, let  $t_i(\omega)$  be given by (12). For those  $i \leq k$ , and  $t \geq t_i(\omega)$ , so that  $\mu_i(t) \to \infty$  suitably fast, we obtain that  $\mathbb{E} \widetilde{L}_i(t) \sim \mu_i(t) \sim \mathbb{E} \widehat{L}_i(t)$ , and we have  $L_{k-\ell}(t) \sim \widehat{L}_{k-\ell}(t) \sim \mu_{k-\ell}(t)$ .

Let  $t_0$  be the first step at which  $L_{k-j^*}(t) = 1$ . Then either  $t_0 \le t_1$ , or, as the probability a particle halts at level  $k - j^*$  is

$$\phi = \frac{L_{k-(j^*-1)}(t_1)}{N_{k-(j^*-1)}} \sim \frac{Cd^{1/2}}{d^{k-j^*+1}} = \frac{C}{d^{k+3/2-\sqrt{2k+2}}};$$

the probability this does not occur in a further  $t_1$  steps is, see (38),

$$(1-\phi)^{t_1} \le e^{-t_1\phi} = e^{-(1+o(1))C\sqrt{2k+2}} = o(1),$$

assuming  $k \to \infty$ . As  $\widehat{L}_{k-j^*}(\max(t_0, t_1)) \le \omega$ , w.h.p.; we assume that  $L_{k-j^*}(\max(t_0, t_1)) \le \omega$ , and for  $\ell > j^*$ ,  $L_{k-\ell} = 0$ .

Let u be the vertex in level  $k-j^*$  containing the unique particle halted at  $t_0$ . Construct a path back from u to the source as follows. Wait until a particle halts at  $w_{k-j^*-1}$  in level  $k-j^*-1$  by choosing edge  $w_{k-j^*-1}u$ . The expected time for this is  $d^{k-j^*}$ . In a further expected time  $d^{k-j^*-1}$ , the path will extend backwards, as a particle will halt in level  $k-j^*-2$  by choosing edge to  $w_{k-j^*-1}$  etc. Thus in a further

$$T = d^{k-j^*} + d^{k-j^*-1} + \dots + d = d^{k-j^*} \left( \frac{1 - 1/d^{k-j^*}}{1 - 1/d} \right) = \Theta(d^{k-j^*}) = \Theta(d^{k+1-\sqrt{2k+2}})$$

expected steps there will be a path  $vw_1 \cdots w_{k-j^*-1}u$  of halted particles extending from the source v to vertex u thus stopping the DLA process (if it has not already halted). This path should be unique, as the expected time for it to branch backwards at any level i is  $d^i \gg d^{i-1}$  if  $d \to \infty$ .

On the other hand the time to create another halted particle in level  $k - j^*$  has expected value

$$\frac{1}{\phi} = \Theta(d^{k+3/2 - \sqrt{2k+2}}) = \Theta(Td^{1/2}).$$

On the assumption that  $d \to \infty$ , in at most  $\omega(t_1 + T)$  steps the process has halted w.h.p., as claimed.

The effect of rounding error in  $j^*$ . In the case that  $j^*$  is not integer, we require the maximum j such that 2k + 2 > j(j + 1). Clearly  $j = \lfloor j^* \rfloor$  satisfies 2k + 2 > j(j + 1), but what about  $j = \lceil j^* \rceil$ ? Put  $j = j^* + \varepsilon$ . Further analysis, not given here, shows that the condition  $2k + 2 \ge j(j + 1)$ , is satisfied by  $j = \lceil j^* \rceil$  up to some  $\varepsilon \in (1/2, 1)$ .

Let  $j = \max\{i : (2k+2) \ge i(i+1)\}$  and suppose that  $j = \lfloor j^* \rfloor$  so that  $j^* = j + \varepsilon$ . Let  $T_M = \Theta(t_1(k-j))$  be given by

$$T_M = \sqrt{k} d^{\frac{1}{2}(2k+2-j-(2k+2)/(j+1))}$$

be a revised estimate of the order of the halting time, where  $T_M \leq T_f$  as  $j^*$  maximizes  $T_f$ . As  $j^* = \sqrt{2k+2} - 1$ ,  $T_f = \sqrt{k}d^{(j^{*2}/2)}$ , see (37), and  $2k+2-j^*-(j^*)^2 = \sqrt{2k+2}$ ,

$$\begin{split} \frac{T_M}{T_f} = & d^{\frac{1}{2}\left((2k+2) - j^* + \varepsilon - \frac{2k+2}{j^* + 1 - \varepsilon} - (j^*)^2\right)} \\ = & d^{\frac{1}{2}\left(\sqrt{2k+2} - \frac{\sqrt{2k+2}}{1 - \varepsilon/(\sqrt{2k+2})} + \varepsilon\right)} \\ = & d^{-\frac{\varepsilon^2}{2\sqrt{2k+2}}(1 + O(1/\sqrt{k}))} \end{split}$$

Choosing  $j = \lceil j^* \rceil = j^* + \varepsilon$  gives the same result. Thus the effect of rounding is to alter  $T_f$  by  $\Theta(1)d^{-O(1/\sqrt{k})}$ .

Existence of a unique connecting path component. Finally, we prove that the arborescence rooted at level k containing the connecting path from source to sink, consists uniquely of that path. The proof is similar to Lemma 6 for the equal layers model. At  $t_f$ , w.h.p. there is a unique path from level  $k - j^*$  to level zero, so that level  $k - j^* + 1$  plays the role of level one. By analogy with (30),

$$\mathbb{P}(u_{k-j^*+1} \in R_{k-j^*+1}(t)) = \frac{Z_{k-j^*+1}(t)}{L_{k-j^*+1}(t)} \le \frac{\omega}{\mu_{k-j^*+1}(t)} \le O\left(\frac{\omega^2}{d^{1/2}}\right),$$

where we used an earlier result that  $\mu_{k-(j^*-1)}(t_1) = Cd^{1/2}$ .

Thus as  $t_f \leq \omega t_1(j^*)$ , where  $j^* = \sqrt{2k+2} - 1$  and  $t_1(k-j^*)$  is given by (38)

$$\mathbb{P}((v, u_k) - \text{path is blue}) \ge \left(1 - \frac{\omega^2}{d^{1/2}}\right) \prod_{j=1}^{j^*-2} \left(1 - \frac{\omega t_f}{d^{k-j}}\right) = 1 - O\left(\frac{\omega^2}{d^{1/2}}\right) - O\left(\frac{\omega\sqrt{k}}{d^{3/2}}\right).$$

## 5 DLA on trees with large branching factor

Let G be a labelled tree with branching factor d and final level k, such that  $d^k = n$ . As before, the source of the particles is the unique vertex v at level zero, an artificial sink vertex z at level k + 1 is attached to the vertices at level k, and  $S_i$  is the set of vertices in level i.

Let  $j^* = \sqrt{2k+2} - 1$ , and let  $T_f$  be given by

$$T_f = \sqrt{k} \ d^{k+3/2 - \sqrt{2k+2}}. (41)$$

The average number of particles arriving at a vertex x at level  $k-j^*$  by step  $T_f$  is order  $\sqrt{kd}$ , and the distance  $j^*$  to level k is less than this. However, the average number of particles passing through a vertex w at level  $k-j^*+1$  by step  $T_f$  is

$$\frac{T_f}{d^{k-j}} = \sqrt{\frac{k}{d}} = o(1),\tag{42}$$

provided  $k \ll d$ , which we assume to be true. Thus the event that  $j^*$  particles have arrived at w is an upper tail event. There should be few vertices w with a path to level k consisting of  $j^*$  such halted particles. We assume henceforth that  $j \leq j^* - 1$ , and look for such paths.

Let  $X_w = X_w(k-j, \ell, t)$  be the indicator for the event that exactly  $\ell$  particles reach vertex w in level k-j at or before step t. On the assumption that all levels i < k-j are unoccupied at t,

$$\mathbb{E} X_w = {t \choose \ell} \left(\frac{1}{d^{k-j}}\right)^{\ell} \left(1 - \frac{1}{d^{k-j}}\right)^{t-\ell}.$$
 (43)

Let  $X = \sum_{w \in S_{k-j}} X_w$  be the number of vertices at level k-j which have exactly  $\ell = j+1$  particles reaching them. Thus  $\mathbb{E} X = d^{k-j} \mathbb{E} X_w$ , and

$$\begin{split} \mathbb{E}\,X(X-1) = & (d^{k-j})(d^{k-j}-1)\binom{t}{2\ell}\binom{2\ell}{\ell}\left(\frac{1}{d^{k-j}}\right)^\ell\left(\frac{1}{d^{k-j}}-1\right)^\ell\left(1-\frac{2}{d^{k-j}}\right)^{t-2\ell} \\ = & \left(1+O\left(\frac{\ell}{d^{k-j}}\right)\right)(\mathbb{E}\,X)^2. \end{split}$$

Thus, as  $X \leq t_f/\ell$  deterministically,

$$\mathbf{Var}X = \mathbb{E}X + O\left(\frac{\ell}{d^{k-j}}\right) (\mathbb{E}X)^2 = (1 + O(\sqrt{k/d}) \mathbb{E}X = (1 + o(1))\mathbb{E}X. \tag{44}$$

Let  $\ell = j+1$ . The probability the particles arriving at w form a path  $w = w_0 w_1 w_2 \cdots w_j = u$  of halted particles from level k-j to level k is obtained as follows. Let  $1 \le s_0 \le s_1 \le s_2 \le \cdots \le s_j \le t$  be the steps (and particle labels) of the halted particles forming this path. Thus  $s_0$  reaches  $w_j = u$  an event of probability  $1/d^j$ ,  $s_1$  reaches  $w_{j-1}$  and picks u as a neighbour, an event also of probability  $1/d^j$  and so on, and finally  $s_j$  picks  $w_1$  as a neighbour and halts at w, an event of probability 1/d. Thus

$$\mathbb{P}(\text{path } w_0 w_1 \cdots w_j \mid (j+1) \text{ arrivals at } w) = \frac{1}{d^j} \frac{1}{dd^2 \cdots d^j}.$$

We call such a path exact if the halted particles on the path are the only particles in the sub-tree from level k-j to level k rooted at w. Let  $Y_w$  be the number of exact paths from w to level k at step t. As there are  $d^j$  choices for u given w, we have

$$\mathbb{E} Y_{w} = d^{j} \frac{1}{d^{j}} \frac{1}{dd^{2} \cdots d^{j}} {t \choose j+1} \left( \frac{1}{d^{k-j}} \right)^{j+1} \left( 1 - \frac{1}{d^{k-j}} \right)^{t-(j+1)}$$

$$= (1 - O(j^{2}/t)) \frac{1}{d^{k} \cdots d^{k-j}} \frac{t^{j+1}}{(j+1)!} e^{-t/d^{k-j}}$$

$$\sim \frac{t^{j+1}}{(j+1)!} \frac{1}{d^{k} \cdots d^{k-j}} = \frac{\mu_{k-j}(t)}{d^{k-j}},$$
(45)

where  $\mu_{k-j}(t)$  is given by (34), and we used the assumption (42) that  $j \leq j^* - 1$ . Let  $Y(k-j, j+1, t) = \sum Y_w$  be the number of exact paths from level k-j to k at step t. There are  $d^{k-j}$  choices for w, so  $\mathbb{E} Y \sim \mu_{k-j}(t)$ .

The value  $t_1(k-j^*)$  from (38) satisfies  $t_1 = \Theta(T_f)$  where  $T_f$  is from (33). Using (36) and (37) we obtain that

$$\mu_{k-j^*+1}(T_f) = \Theta(d^{1/2}),$$

so 
$$\mathbb{E} Y(k-j^*+1,j+1,t_1(k-j^*) \sim \mu_{k-j^*+1}(t_1) = \Theta(d^{1/2}).$$

Comparing (43) and (46), we see that  $\mathbb{E}Y = \mathbb{E}X/d^{j(j+1)/2}$ , and thus  $\mathbb{E}X = \Theta(d^{1/2+j(j+1)/2})$ . By the Chebychev inequality, w.h.p.  $X \sim \mathbb{E}X(k-j,j+1,t)$ , see (44). The events of exact path construction at vertices w,w' of level k-j are independent, given exactly j+1 particles arrive at each of w,w'. As  $\mathbb{E}Y(k-j,j+1,t) = \Theta(d^{1/2})$  where  $d \to \infty$ , then, for some small  $\varepsilon > 0$ , w.h.p.  $(1-\varepsilon)\mathbb{E}Y \leq Y \leq (1+\varepsilon)\mathbb{E}Y$ . Conditional on this,  $Y = \Theta(d^{1/2})$ .

Next consider that  $\ell \geq j+2$  particles reach vertex w, so that the path is not exact, as at least one other halted particle occupies the sub-tree rooted at w. Let  $Q(k-j,\ell,t)$  be the number of these events. Then, provided (42) holds,

$$\sum_{\ell \ge j+2} \mathbb{E} Q(k-j,\ell,t) = \Theta(1) \frac{t}{d^{k-j}} \mathbb{E} Y = O(1) \sqrt{\frac{k}{d}} \mathbb{E} Y = o(\mathbb{E} Y).$$

From Section 4, at time  $t_1(k-j^*)/\omega$ ,  $\mu_{k-j^*}(t_1(k-j^*)/\omega) = (1/\omega)^{j^*+1}$  and we assume levels  $1, ..., k-j^*$  are empty. At  $t_1$  there are  $\Theta(d^{1/2})$  paths from k-j to level k, of which  $o(d^{1/2})$  are not exact. Either one of these paths has extended back to some x in  $k-j^*$  or level  $k-j^*$  is currently unoccupied. In the latter case, any particles which passed via x in between times  $t_1/\omega$  and  $t_1$  must have chosen some out neighbour of x other than w. The probability a given particle extends some exact path from level  $k-j^*+1$  to  $k-j^*$  is order  $d^{1/2}/d^{k-j^*+1}$ . Let T(hit) be the time for the first such extension to happen. Then

$$\mathbb{E} T(\text{hit}) = d^{k-j^*+1/2} = d^{k+3/2-\sqrt{2k+2}} = O(T_f).$$

By analogy with the analysis in Section 4, w.h.p. this selected exact path of occupied vertices from  $k - j^*$  to k will extend back to the source v, in

$$T = O(d^{k-j^*}) = O(d^{k+1-\sqrt{2k+2}}) = O\left(\frac{\mathbb{E}T(\text{hit})}{\sqrt{d}}\right)$$

expected steps; whereas the expected wait for a second exact path to extend is  $\mathbb{E}T(\text{hit})$ . This completes the proof of Theorem 1 for trees.

## 5.1 An upper bound on finish time for trees with any $d \ge 2$

### Proposition 8. Let

$$T = d^{k+3/2 - \sqrt{2k+2} + 1/2 \log_d k}.$$

Then  $\omega d^{1/2}T$  is a w.h.p. upper bound on the finish time of DLA in any tree with branching factor  $d \geq 2$ , and height k.

Let j be such that  $d^{k-j} \geq T$  so that  $e^{-T/d^{k-j}} = \Theta(1)$  in the calculation of  $\mathbb{E} Y$  (see (45)). Let h be integer such that

$$k - j = k - j^* + h \ge k + 1 - \sqrt{2k + 2} + h \ge k + 3/2 - \sqrt{2k + 2} + (1/2)\log_d k$$

so that  $h = \lceil 1/2 + 1/2 \log_d k \rceil$ . In Section 5 we assumed that  $k \ll d$  so that h = 1. At  $T \sim t_1(k-j^*)$  from the growing layers model

$$\frac{\mu_{k-j^*+h}}{\mu_{k-j^*}} = \Theta(1)2^{h/2}d^{h^2/2}.$$

This ratio is  $\omega(1)$ , as either  $d \to \infty$  or if d is constant, then  $\log_d k = \log_d \log_d n$ . Either some path of halted particles already extends to a level i where  $i < k - j^* + h$ , or all vertices in these levels are unoccupied, and w.h.p. there exists a (not necessarily exact) path of length  $j^* - h + 1$ . Pick the longest path P(w), where w in level  $k - j^* + \ell$  is the final vertex on this path. It takes at most  $\Theta(d^{k-j^*+h})$  expected steps to extend P(w) back to the source. Failing this either the path from w to v is now blocked internally in this time by some longer path, or some other disjoint path P' has become longer. Either way, w.h.p. the process has halted in a further  $\omega d^{k-j^*+h} = \omega d^{1/2}T$  steps.

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