

# Cover time of a random graph with given degree sequence

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April 9, 2012

## Abstract

In this paper we establish the cover time of a random graph  $G(\mathbf{d})$  chosen uniformly at random from the set of graphs with vertex set  $[n]$  and degree sequence  $\mathbf{d}$ . We show that under certain restrictions on  $\mathbf{d}$ , the cover time of  $G(\mathbf{d})$  is **whp** asymptotic to  $\frac{d-1}{d-2} \frac{\theta}{d} n \log n$ . Here  $\theta$  is the average degree and  $d$  is the *effective minimum degree*.

## 1 Introduction

Let  $G = (V, E)$  be a connected graph with  $|V| = n$  vertices and  $|E| = m$  edges.

For a simple random walk  $\mathcal{W}_v$  on  $G$  starting at a vertex  $v$ , let  $C_v$  be the expected time taken to visit every vertex of  $G$ . The *vertex cover time*  $C(G)$  of  $G$  is defined as  $C(G) = \max_{v \in V} C_v$ . The vertex cover time of connected graphs has been extensively studied. It is a classic result of Aleliunas, Karp, Lipton, Lovász and Rackoff [2] that  $C(G) \leq 2m(n-1)$ . It was shown by Feige [8], [9], that for any connected graph  $G$ , the cover time satisfies  $(1 - o(1))n \log n \leq C(G) \leq (1 + o(1))\frac{4}{27}n^3$ . Between these two extremal examples, the cover time, both exact and asymptotic, has been determined for a number of different classes of graphs.

In this paper we study the cover time of random graphs  $\mathcal{G}(\mathbf{d})$  picked uniformly at random (**uar**) from the set  $\mathcal{G}(\mathbf{d})$  of simple graphs with vertex set  $V = [n]$  and degree sequence  $\mathbf{d} = (d_1, d_2, \dots, d_n)$ , where  $d_i$  is the degree of vertex  $i \in V$ . We make the following definitions:

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Let  $V_j = \{i \in V : d_i = j\}$  and let  $n_j = |V_j|$ . Let  $\sum_{i=1}^n d_i = 2m$  and let  $\theta = 2m/n$  be the average degree. We use the notations  $d_i$  and  $d(i)$  for the degree of vertex  $i$ .

Let  $0 < \alpha \leq 1$  be constant,  $0 < c < 1/8$  be constant and let  $d$  be a positive integer. Let  $\gamma = (\sqrt{\log n}/\theta)^{1/3}$ . We suppose the degree sequence  $\mathbf{d}$  satisfies the following conditions:

- (i) Average degree  $\theta = o(\sqrt{\log n})$ .
- (ii) Minimum degree  $\delta \geq 3$ .
- (iii) For  $\delta \leq i < d$ ,  $n_i = O(n^{ci/d})$ .
- (iv)  $n_d = \alpha n + o(n)$ . We call  $d$  the *effective minimum degree*.
- (v) Maximum degree  $\Delta = O(n^{c(d-1)/d})$ .
- (vi) Upper tail size  $\sum_{j=\gamma\theta}^{\Delta} n_j = O(\Delta)$ .

We call a degree sequence  $\mathbf{d}$  which satisfies conditions (i)–(vi) *nice*, and apply the same adjective to  $\mathcal{G}(\mathbf{d})$ . Basically, nice graphs are sparse, with not too many high degree vertices. Any degree sequence with constant maximum degree, and for which  $d = \delta$  is nice. The conditions hold in particular, for  $d$ -regular graphs,  $d \geq 3$ ,  $d = \delta = o(\sqrt{\log n})$ , as condition (iii) is empty. The spaces of graphs we consider are somewhat more general. The condition nice, allows for example, bi-regular graphs where half the vertices are degree  $d \geq 3$  and half of degree  $a = o(\sqrt{\log n})$ .

Conditions (i), (v), (vi) allow us to infer structural properties of  $\mathcal{G}(\mathbf{d})$  via the configuration model, in a way that is explained in Section 3.1. The effective minimum degree condition (iv), ensures that some entry in the degree sequence occurs order  $n$  times. Condition (iii) is necessary for the analysis of the random walk, as Theorem 1 does not hold when  $c > 1$ , even if the maximum degree is constant. However, the value  $c < 1/8$  in condition (iii) is somewhat arbitrary, as are the precise values in conditions (v), (vi).

It will follow from Lemma 7 that random graphs with a nice degree sequence are connected with high probability (**whp**). The following theorem gives the cover time of nice graphs.

**Theorem 1.** *Let  $G(\mathbf{d})$  be chosen **uar** from  $\mathcal{G}(\mathbf{d})$ , where  $\mathbf{d}$  is nice. Then **whp***

$$C(G(\mathbf{d})) \sim \frac{d-1}{d-2} \frac{\theta}{d} n \log n. \quad (1)$$

In this paper, the notation **whp** means with probability  $1 - n^{-\Omega(1)}$ , and  $A(n) \sim B(n)$  means  $\lim_{n \rightarrow \infty} A(n)/B(n) = 1$ .

We note that if  $d \sim \theta$ , i.e. the graph is pseudo-regular, then as long as condition (iii) holds,

$$C(G) \sim \frac{d-1}{d-2} n \log n.$$

This extends the result of [5] for random  $d$ -regular graphs.

## Structure of the paper

The proof of Theorem 1 is based on an application of (7) below. Put simply, (7) says that, if we ignore which vertices the random walk visits during the mixing time, the probability a vertex  $v$  remains unvisited in the first  $t$  steps is asymptotic to  $\exp(-\pi_v t/R_v)$ . Here  $\pi_v = d(v)/2m$  where  $d(v)$  is the degree of vertex  $v$  and  $m$  is the number of edges. The variable  $R_v$  is the expected number of returns to  $v$  during the mixing time, for a walk starting at  $v$ . To estimate  $R_v$  in Section 4.2, we describe and prove the required **whp** graph properties in Section 3. Lemma 7, proved in the Appendix establishes that nice graphs have constant conductance **whp**; which implies connectivity as asserted in the introduction. The proof that (7) is valid **whp** for  $\mathcal{G}(\mathbf{d})$  is similar to proofs in earlier papers and is given in the Appendix. The cover time  $C(G)$  in (1) is established in Section 5 as follows. Firstly an upper bound of  $(1 + o(1))C(G)$  is proved in Section 5.1. In Section 5.2 a lower bound is determined by constructing a set of vertices  $S$  such that  $\sum_{v \in S} \exp(-\pi_v t/R_v) \rightarrow \infty$  at  $t = (1 - o(1))C(G)$ .

## 2 Estimating first visit probabilities

In this section  $G$  denotes a fixed connected graph with  $n$  vertices. A random walk  $\mathcal{W}_u$  is started from a vertex  $u$ . Let  $\mathcal{W}_u(t)$  be the vertex reached at step  $t$ , let  $P$  be the matrix of transition probabilities of the walk and let  $P_u^{(t)}(v) = \mathbf{Pr}(\mathcal{W}_u(t) = v)$ . We assume that the random walk  $\mathcal{W}_u$  on  $G$  is ergodic with stationary distribution  $\pi$ , where  $\pi_v = d(v)/(2m)$ , and  $d(v)$  is the degree of vertex  $v$ .

Let  $T$  be a positive integer such that for  $t \geq T$

$$\max_{u,x \in V} |P_u^{(t)}(x) - \pi_x| \leq n^{-3}, \tag{2}$$

and let

$$\lambda = \frac{1}{KT} \tag{3}$$

for a sufficiently large constant  $K$ . The existence of such a  $T$  will follow from (20).

Considering a walk  $\mathcal{W}_v$ , starting at vertex  $v$ , let  $r_t = \Pr(\mathcal{W}_v(t) = v)$  be the probability that the walk returns to  $v$  at step  $t = 0, 1, \dots$ , and let

$$R_T(z) = \sum_{j=0}^{T-1} r_j z^j. \quad (4)$$

Given vertices  $u, v$ , let  $\mathcal{W}_u$  be a random walk starting at vertex  $u$ . For  $t \geq T$  let  $\mathbf{A}_v(t)$  be the event that  $\mathcal{W}_u$  does not visit  $v$  in steps  $T, T+1, \dots, t$ . Several versions of the following lemma have appeared previously (e.g. in [5], [6]). For completeness, a proof is given in Section 6.1 of the Appendix.

**Lemma 2.** *Let  $v \in V$  satisfy the following conditions:*

(a) *For some constant  $\psi > 0$ , we have*

$$\min_{|z| \leq 1+\lambda} |R_T(z)| \geq \psi,$$

*where  $R_T(z)$  is from (4).*

(b)  *$T\pi_v = o(1)$  and  $T\pi_v = \Omega(n^{-2})$  for all  $v \in V$ .*

Let

$$R_v = R_T(1). \quad (5)$$

Then there exists

$$p_v = \frac{\pi_v}{R_v(1 + O(T\pi_v))}, \quad (6)$$

such that for all  $t \geq T$ ,

$$\Pr(\mathbf{A}_v(t)) = \frac{(1 + O(T\pi_v))}{(1 + p_v)^t} + O(T^2\pi_v e^{-\lambda t/2}). \quad (7)$$

### 3 Graph properties

We make our **whp** calculations about properties of nice graphs in the configuration model, (see Bollobás [3]). Let  $W = [2m]$  be the set of *configuration points* and for  $i \in [n]$ , let  $W_i = [d_1 + \dots + d_{i-1} + 1, d_1 + \dots + d_i]$ . Thus  $W_i$ ,  $i = 1, \dots, n$  is a partition of  $W$ . For  $u \in W_i$ , define  $\phi : [2m] \rightarrow [n]$  by  $\phi(u) = i$ . Thus,  $|W_i| = d_i$ , and  $\phi(u)$  is the vertex corresponding to the configuration point  $u$ . Given a pairing  $F$  (i.e. a partition of  $W$  into  $m$  pairs  $\{u, v\}$ ) we obtain a multi-graph  $G_F$  with vertex set  $[n]$  and an edge  $(\phi(u), \phi(v))$  for each  $\{u, v\} \in F$ .

Choosing a pairing  $F$  uniformly at random from among all possible pairings of the points of  $W$  produces a random multi-graph  $G_F$ . Let

$$\mathcal{F}(2m) = \frac{(2m)!}{m!2^m}. \quad (8)$$

Thus  $\mathcal{F}(2m)$  counts the number of distinct pairings  $F$  of the  $2m$  points in  $W$ . Moreover the number of pairings corresponding to each simple graph  $G \in \mathcal{G}(\mathbf{d})$  is the same, so that simple graphs are equiprobable in the space of multi-graphs. Let  $\nu = \sum_i d_i(d_i - 1)/(2m)$ . Assuming that  $\Delta = o(m^{1/3})$ , (see e.g. [10]), the probability that  $G_F$  is simple is given by

$$P_S = \mathbf{Pr}(G_F \text{ is simple}) \sim e^{-\frac{\nu}{2} - \frac{\nu^2}{4}}. \quad (9)$$

Our assumption that conditions (i)–(vi) hold for  $\mathbf{d}$ , imply that  $\Delta = o(m^{1/3})$ . Also as  $\gamma = (\sqrt{\log n}/\theta)^{1/3}$ , then  $\nu = o(\sqrt{\log n})$  follows from

$$\nu \leq \frac{1}{\theta n} \left( \sum_{j=3}^{\gamma\theta} n_j j^2 + \sum_{j=\gamma\theta}^{\Delta} n_j j^2 \right) \leq \frac{1}{\theta n} (n\gamma^2\theta^2 + O(\Delta^3)) = o(\sqrt{\log n}).$$

If  $\nu = o(\sqrt{\log n})$ , then  $P_S$  in (9) is at least  $e^{-o(\log n)}$ . On the other hand, statements about graph structure we make in this paper using the configuration model fail with probability at most  $n^{-\Omega(1)}$ , which means they hold **whp** for simple graphs.

### 3.1 Structural properties of $G(\mathbf{d})$

In this section we establish the **whp** properties of nice graphs needed to estimate  $R_v$  in (5) for all  $v \in V$ .

Let  $C$  be a large constant, and let

$$\omega = C \log \log n. \quad (10)$$

A cycle or path is *small*, if it has at most  $2\omega + 1$  vertices, otherwise it is *large*. Let

$$\ell = B \log^2 n \quad (11)$$

for some large constant  $B$ . A vertex  $v$  is *light* if it has degree at most  $\ell$ , otherwise it is *heavy*. A cycle or path is *light* if all vertices are light. A light vertex  $v$  is *small* if it has degree at most  $d - 1$ .

**Lemma 3.** *Let  $\mathbf{d}$  be a nice degree sequence and let  $G(\mathbf{d})$  be chosen uniformly at random from the  $\mathcal{G}(\mathbf{d})$ . There exists  $\epsilon > 0$  constant such that with probability  $1 - O(n^{-\epsilon})$ ,*

(a) *No vertex disjoint pair of small light cycles are joined by a small light path.*

(b) No light vertex is in two small light cycles.

(c) No small cycle contains a heavy vertex or small vertex, or is connected to a heavy or small vertex by a small path.

(d) No pair of small or heavy vertices is connected by a small path.

**Proof** We note a useful inequality. For integer  $x > 0$ , let  $\mathcal{F}(2x) = \frac{(2x)!}{2^x x!}$ , as defined in (8), then

$$\frac{\mathcal{F}(\theta n - 2x)}{\mathcal{F}(\theta n)} = \frac{(\theta n - 2x)!}{\left(\frac{\theta n}{2} - x\right)! 2^{\frac{\theta n}{2} - x}} \frac{\left(\frac{\theta n}{2}\right)! 2^{\frac{\theta n}{2}}}{(\theta n)!} = \left(\prod_{i=1}^x \theta n - 2i + 1\right)^{-1} \leq \left(\frac{1}{\theta n - 2x + 1}\right)^x. \quad (12)$$

(a) Let  $S$  denote the sum over  $a, b, c$  of the expected number of subgraphs consisting of small light vertex cycles of length  $a, b$  joined by a small light vertex path of length  $c + 1$ . Then

$$S \leq \sum_{a=3}^{2\omega+1} \sum_{b=3}^{2\omega+1} \sum_{c=0}^{2\omega+1} \binom{n}{a} \binom{n}{b} \binom{n}{c} \frac{(a-1)!}{2} \frac{(b-1)!}{2} c! a b \ell^{2(a+b+c+1)} \frac{\mathcal{F}(\theta n - 2(a+b+c+1))}{\mathcal{F}(\theta n)} \quad (13)$$

**Explanation.** Choose  $a$  vertices for one cycle,  $b$  vertices for the other and  $c$  vertices for the path. Each light vertex has most  $\ell(\ell - 1)$  ways to connect to its neighbours on a given path or cycle. This explains the exponent of  $\ell$ . Choosing  $x = (a + b + c + 1) \leq 6\omega + 4$  in (12), we find  $S$  is bounded by

$$\begin{aligned} S &\leq \sum_{a=3}^{2\omega+1} \sum_{b=3}^{2\omega+1} \sum_{c=0}^{2\omega+1} n^a n^b n^c \ell^{2(a+b+c+1)} \left(\frac{1}{\theta n - (12\omega + 8)}\right)^{a+b+c+1} \\ &\leq \frac{\ell^2}{\theta n - (12\omega + 8)} \sum_a \sum_b \sum_c \left(\frac{n\ell^2}{\theta n - (12\omega + 8)}\right)^{a+b+c} \\ &= O\left(\frac{\omega^3 \ell^{12\omega+8}}{\theta n}\right) = o(1). \end{aligned} \quad (14)$$

(b) The proof for this part is similar to (a).

(c) Note that, in condition (vi), the value of  $\gamma\theta < \ell$  and thus the number,  $H$ , of heavy vertices is  $O(\Delta) = O(n^{c(d-1)/d})$ . Similarly, from condition (iii), the number of small vertices is  $O(n^{c(d-1)/d})$ . The expected number  $S$  of cycles of length  $3 \leq a \leq 2\omega + 1$  with  $a - k$  light vertices and  $k \geq 1$  heavy vertices can be bounded by the expected number of configuration pairings of cycles of this type. Thus

$$S \leq \sum_{a=3}^{2\omega+1} \sum_{k \geq 1} \binom{n}{a-k} \binom{H}{k} (a-1)! \ell^{2(a-k)} \Delta^{2k} \frac{\mathcal{F}(\theta n - 2a)}{\mathcal{F}(\theta n)}.$$

Thus, using (12), we have

$$\begin{aligned}
S &= O(1) \sum_a \sum_{k \geq 1} \binom{a}{k} n^{-k} \Delta^{3k} \ell^{2a} \\
&= O(1) \sum_a \ell^{2a} \frac{a \Delta^3}{n} \\
&= O(\omega^2) \frac{\ell^{4\omega+2} \Delta^3}{n} = o(1).
\end{aligned}$$

We next count the expected number  $S$  of cycles of lengths  $3 \leq a \leq 2\omega + 1$  containing only light vertices, which are joined to a heavy vertex by a light vertex path of length  $0 \leq b - 1 \leq 2\omega$ . This can be bounded by

$$\begin{aligned}
S &\leq \sum_{a=3}^{2\omega+1} \sum_{b=1}^{2\omega+1} \binom{n}{a} \binom{n}{b-1} \binom{H}{1} (a-1)! (b-1)! \ell^{2a+2(b-1)+1} a \Delta \frac{\mathcal{F}(\theta n - 2(a+b))}{\mathcal{F}(\theta n)} \\
&= O(\omega^2) \frac{\ell^{8\omega+4} \Delta^2}{n} = o(1).
\end{aligned}$$

(d) There are  $H = O(\Delta)$  small or heavy vertices. The expected number  $S$  of small light paths length connecting such vertices is

$$\begin{aligned}
S &\leq \sum_{a=0}^{2\omega+1} \binom{n}{a} \binom{H}{2} a! \ell^{2a} \Delta^2 \frac{\mathcal{F}(\theta n - 2(a+1))}{\mathcal{F}(\theta n)} \\
&= O\left(\frac{\omega \ell^{4\omega+2} \Delta^4}{n}\right).
\end{aligned}$$

□

For a vertex  $v$ , let  $G_v$  be the subgraph induced by the set of vertices within a distance  $\omega$  of  $v$ . As any paths or cycles contained in  $G_v$  are of length at most  $2\omega + 1$  and hence small, the following lemma is a corollary of Lemma 3.

**Lemma 4.** *Let  $G(\mathbf{d})$  be nice. Assuming the conditions (a)-(d) of Lemma 3 hold, then*

- (a) *If  $G_v$  contains a small or heavy vertex,  $G_v$  is a tree.*
- (b) *If  $G_v$  is not a tree, then  $G_v$  contains exactly one small cycle, and all vertices of  $G_v$  are light.*
- (c) *There are  $O(\ell^\omega n^{ci/d})$  vertices  $v$  such that  $G_v$  contains a small vertex of degree  $i$ .*
- (d) *There are  $O(\ell^\omega n^{2c(d-1)/d})$  vertices  $v$  such that  $G_v$  contains a heavy vertex.*

**Proof** If  $G_v$  contains a small or heavy vertex then it is a tree, and all other vertices are light. Thus  $|G_v| = O(\ell^\omega)$  for small vertices, and there are  $O(n^{ci/d})$  small vertices of degree  $i$ . If  $G_v$  contains a heavy vertex then  $|G_v| = O(\Delta \ell^\omega)$ .  $\square$

**Lemma 5.** *Let  $\mathbf{d}$  be a nice degree sequence and let  $G(\mathbf{d})$  be chosen uniformly at random from the  $\mathcal{G}(\mathbf{d})$ . For any  $\epsilon > 0$  constant, with probability  $1 - O(n^{-\epsilon})$ , there are at most  $n^{4\epsilon}$  vertices  $v$  such that  $G_v$  contains a cycle.*

**Proof** The expected number of vertices on small light cycles is at most

$$\begin{aligned} S &\leq \sum_{a=3}^{2\omega+1} \binom{n}{a} \frac{(a-1)!}{2} \ell^{2a} \frac{\mathcal{F}(\theta n - 2a)}{\mathcal{F}(\theta n)} \\ &= O(\omega \ell^{4\omega+2}). \end{aligned}$$

The probability there are more than  $n^\epsilon$  vertices on small light cycles is  $o(n^{-\epsilon/2})$ , for any  $\epsilon > 0$ . If  $G_v$  contains only light vertices, then  $|G_v| = O(\ell^\omega)$ , and thus (**whp**) there are at most  $n^{2\epsilon}$  vertices  $v$  such that  $G_v$  contains a small light cycle.  $\square$

A vertex  $v$  is *d-compliant*, if  $G_v$  is a tree, and all vertices of  $G_v$  have degree at least  $d$ . A vertex  $v$  is *d-tree-like* to depth  $h$  if the graph induced by the vertices at distance at most  $h$  from  $v$  form a  $d$ -regular tree, (i.e. all vertices on levels  $0, 1, \dots, h-1$  have degree  $d$ ).

A vertex  $v$  is *d-tree-regular*, if it is *d-tree-like* to depth  $h$ , *d-compliant* to depth  $\omega$  and all vertices of  $G_v$  are light. For such a vertex  $v$ , the first  $h$  levels of the BFS tree, really are a  $d$ -regular tree, and the remaining  $\omega - h$  levels can be pruned to a  $d$ -regular tree. We choose the following value for  $h$ , which depends on  $\theta$ .

$$h = \frac{1}{\log d} \log \left( \frac{\log n}{(\log \log n) \log \theta} \right) \quad (15)$$

The exact value of  $h$  is not so important. The main thing is that  $d^h \rightarrow \infty$  in Lemma 9, but not too fast in Lemma 6.

**Lemma 6.** *Let  $\mathbf{d}$  be a nice degree sequence and let  $G(\mathbf{d})$  be chosen uniformly at random from the  $\mathcal{G}(\mathbf{d})$ . There exists  $\epsilon > 0$  constant such that with probability  $1 - O(n^{-\epsilon})$ , there are  $n^{1-O(1/\log \log n)}$  *d-tree-regular* vertices.*

**Proof** Recall that  $n_d = |V_d| = \alpha n + o(n)$  for some constant  $\alpha > 0$ . We assume from Lemmas 4 and 5 that all but  $O(n^\epsilon) + O(\ell^\omega \Delta^2)$  vertices of degree  $d$  are *d-compliant*, or have a heavy vertex within distance  $\omega$ .

Let  $N_2 = 1 + d(d-1)^h$ . If  $v$  has degree  $d$  and is *d-tree-like* to depth  $h$ , then the tree of this depth rooted at  $v$  contains less than  $N_2$  vertices. We bound the probability  $P$  that a vertex



$v$  of degree  $d(v) = d$  is  $d$ -tree-like, by bounding the probability of success of the construction of a  $d$ -regular tree of depth  $h$  in the configuration model.

$$P = \Pr(\text{vertex } v \text{ is } d\text{-tree-like}) = \prod_{i=1}^{N_2-1} \frac{d(n_d - i)}{\theta n - 2i + 1} \geq \left( d \frac{n_d - N_2}{\theta n} \right)^{N_2}. \quad (16)$$

Let  $M$  count the number of  $d$ -tree-like vertices, then  $\mathbf{E}[M] = \mu = n_d P$ , and for the value of  $h$  given in (15) we have that

$$\mu = \mathbf{E}[M] = n^{1-O(1/\log \log n)}. \quad (17)$$

To estimate  $\mathbf{Var}[M]$ , let  $I_v$  be the indicator that vertex  $v$  is  $d$ -tree-like. We have

$$\mathbf{E}[M^2] = \mu + \sum_{v \in V_d} \sum_{w \in V_d, w \neq v} \mathbf{E}[I_v I_w], \quad (18)$$

and

$$\mathbf{E}[I_v I_w] = \Pr(v, w \text{ are } d\text{-tree-like}, G_v \cap G_w = \emptyset) + \Pr(v, w \text{ are } d\text{-tree-like}, G_v \cap G_w \neq \emptyset).$$

Now

$$\Pr(v, w \text{ are } d\text{-tree-like}, G_v \cap G_w = \emptyset) = \prod_{i=1}^{2N_2-2} \frac{d(n_d - i - 1)}{\theta n - 2i + 1} \leq P^2. \quad (19)$$

For any vertex  $v$ , the number of vertices  $w$  such that  $G_v \cap G_w \neq \emptyset$  is bounded from above by  $N_2 + dN_2^2$ . Using this and (19), we can bound (18) from above by  $\mu + \mu^2 + \mu(N_2 + dN_2^2)$ .

By the Chebychev Inequality, for some constant  $0 < \tilde{\epsilon} < 1$ ,

$$\Pr\left(|M - \mu| > \mu^{\frac{1}{2} + \tilde{\epsilon}}\right) \leq \frac{\mathbf{Var}[M]}{\mu^{1+2\tilde{\epsilon}}} = \frac{\mathbf{E}[M^2] - \mathbf{E}[M]^2}{\mu^{1+2\tilde{\epsilon}}} \leq \frac{\mu + \mu N_2 + \mu d N_2^2}{\mu^{1+2\tilde{\epsilon}}} = O(n^{-\epsilon}).$$

The lemma now follows from (17).  $\square$

## 4 Random walk properties

### 4.1 Mixing time

Given a graph  $G$ , the *conductance*  $\Phi(G)$  of a random walk  $\mathcal{W}_u$  on  $G$  is defined by

$$\Phi(G) = \min_{\pi(S) \leq 1/2} \frac{e(S : \bar{S})}{d(S)}$$

where  $d(S) = \sum_{v \in S} d(v)$ ,  $\pi(S) = d(S)/2m$ , and  $e(A : B)$  denotes the number of edges with one endpoint in  $A$  and the other in  $B$ . The lemma below follows by applying (9) to Lemma 12 proved in Section 6.2 of the Appendix.

**Lemma 7.** *Let  $\mathbf{d}$  be a nice degree sequence and let  $G(\mathbf{d})$  be chosen uniformly at random from the  $\mathcal{G}(\mathbf{d})$ , then with probability  $1 - O(n^{-1/9})$*

$$\Phi(G) \geq \frac{1}{100}.$$

Note that  $\Phi(G) \geq 1/100$  in Lemma 7 implies  $G(\mathbf{d})$  is connected.

We note a result from Sinclair [11], that

$$|P_u^{(t)}(x) - \pi_x| \leq (\pi_x/\pi_u)^{1/2}(1 - \Phi^2/2)^t. \quad (20)$$

Referring to Lemma 7 and (20), if we choose  $A$  sufficiently large and

$$T = A \log n \quad (21)$$

then (2) holds. There is a technical point here, in that the result (20) assumes that the walk is lazy. A lazy walk moves to a neighbour with probability  $1/2$  at any step. This assumption halves the conductance, and doubles the value of  $R_T(1)$ . Asymptotically, the cover time is also doubled by the inclusion of the lazy steps. The trajectory, and hence cover time of the underlying (non-lazy) walk can be recovered by removing the lazy steps. We will ignore the assumption in (20) for the rest of the paper; and continue as though there are no lazy steps.

## 4.2 Expected number of returns in the mixing time

**Escape probability.** Let  $v \in V$ , and  $B \subseteq V$ , and assume  $v \notin B$ . For a walk  $\mathcal{W}_v^B$  starting at  $v$ , let  $P_v(B)$  be the probability that the walk reaches  $B$  without return to  $v$ ; the *escape probability* from  $v$  to  $B$ . The value of  $P_v(B)$  is given by

$$P_v(B) = \frac{1}{d(v)R_{\text{eff}}(v, B)}, \quad (22)$$

where  $R_{\text{eff}}(v, B)$  is the *effective resistance between  $v$  and  $B$* , treating the edges as having unit resistance. If we treat  $B$  as an absorbing state, then  $f_v(B) = 1 - P_v(B)$  is the probability of a first return to  $v$  by  $\mathcal{W}_v^B$  before absorption at  $B$ ; and  $R_v(B) = 1/(1 - f_v(B)) = 1/P_v(B)$  is the expected number of returns to  $v$  before absorption at  $B$ .

The attractiveness of formula (22) is that by Rayleigh's monotonicity law, deleting edges of the graph does not decrease the effective resistance between  $v$  and  $B$ . Thus provided we do not delete any edges incident with  $v$ , such pruning cannot increase  $P_v(B)$ . See [7] for details of Rayleigh's monotonicity law, and a proof of (22).

For a vertex  $v$ , we defined  $G_v$  as the subgraph induced by the set of vertices within a distance  $\omega$  of  $v$ . Denote by  $\Gamma_v^\omega$  those vertices of  $G_v$  at distance exactly  $\omega$  from  $v$ . The following lemma relates  $R_v$  in (5) of Lemma 2 to  $R_v^* = R_v(\Gamma_v^\omega)$  obtained from (22) as described above.

**Lemma 8.** *Let  $G(\mathbf{d})$  be nice, and assume the conditions of Lemma 3 and Lemma 7 hold. Let  $\mathcal{W}_v^*$  denote a walk on  $G_v$  starting at  $v$  with  $\Gamma_v^\circ$  made into an absorbing state. Let  $R_v^* = \sum_{t=0}^{\infty} r_t^*$ , where  $r_t^*$  is the probability that  $\mathcal{W}_v^*$  is at vertex  $v$  at time  $t$ . Let  $R_v$  be given by (5), then*

$$R_v = R_v^* + o\left(\frac{1}{\log n}\right).$$

For completeness the proof of Lemma 8 is given in Section 6.3 of the Appendix. A similar proof is given in e.g. [5]. The precise value of  $R_v^*$  is given by (22). The next lemma gives some approximate bounds.

**Lemma 9.** *For a vertex  $v \in V$ , let  $\mathcal{W}_v^*$  be a walk on  $G_v$ , starting at  $v$ , and with  $\Gamma_v^\circ$  made into an absorbing state. Let  $P_v(\Gamma_v^\circ)$  be the escape probability of a walk, and let  $R_v^* = 1/P_v(\Gamma_v^\circ)$ .*

(a) *If  $v$  is  $d$ -tree-regular, then  $R_v^* = \frac{d-1}{d-2}(1+o(1))$ .*

(b) *If  $v$  is  $d$ -compliant then  $R_v^* \leq \frac{d-1}{d-2}(1+o(1))$ .*

(c) *If  $G_v$  is a tree,  $R_v^* \leq \frac{\delta-1}{\delta-2}(1+o(1))$ .*

(d) *If  $G_v$  contains a single cycle, and all vertices of  $G_v$  have degree at least  $d$ , then  $R_v^* \leq \frac{d(d-1)}{(d-2)^2}(1+o(1))$ .*

**Proof** (a)

For a biased random walk on the half-line  $(0, 1, \dots, k)$ , starting at vertex  $i$ , with absorbing states  $0, k$ , and with transition probabilities at vertices  $(1, \dots, k-1)$  of  $q = \mathbf{Pr}(\text{move left})$ ,  $p = \mathbf{Pr}(\text{move right})$ ; then

$$\mathbf{Pr}(\text{absorption at } k) = \frac{1 - (q/p)^i}{1 - (q/p)^k}. \quad (23)$$

We first project  $\mathcal{W}_v^*$  onto  $(0, 1, \dots, h)$  with  $p = \frac{d-1}{d}$  and  $q = \frac{1}{d}$ . As  $v$  is  $d$ -tree-like, the probability  $Q(h)$  of escaping from  $v$  to level  $h$  of the  $d$ -regular tree of depth  $h$  rooted at  $v$  is

$$Q(h) = \frac{1 - \frac{1}{d-1}}{1 - \left(\frac{1}{d-1}\right)^h}.$$

Thus for  $h$  given by (15),  $(d-1)^h \rightarrow \infty$  and

$$P_v(\Gamma_v^\circ) \leq Q(h) = (1 + o(1)) \frac{d-2}{d-1}.$$

On the other hand  $G_v$  is  $d$ -compliant so, by pruning, contains a  $d$ -regular subtree, and

$$P_v(\Gamma_v^\circ) \geq \frac{1 - \frac{1}{d-1}}{1 - \left(\frac{1}{d-1}\right)^\omega} = (1 + o(1)) \frac{d-2}{d-1}.$$

(b) We can find a lower bound on the escape probability as follows. Retain all edges incident with  $v$ . Working outward from the neighbours of  $v$ , prune all internal vertices of  $G_v$  down to degree  $d$ , to obtain a subtree  $\Lambda_v$  of  $G_v$  in which  $v$  has degree  $d(v)$  as in  $G_v$ . Let  $\Lambda_v^\circ$  be its leaves, and  $P_v(\Lambda_v^\circ)$  the escape probability from  $v$  to  $\Lambda_v^\circ$  in  $\Lambda_v$ . Then by considering effective resistance, in (22)

$$P_v(\Gamma_v^\circ) \geq P_v(\Lambda_v^\circ) = (1 + o(1)) \frac{d-2}{d-1}.$$

(c) If  $G_v$  is a tree, but has some vertex  $w$  of degree  $\delta \leq d(w) < d$ , then, we can prune the internal vertices of  $G_v - \{v\}$  to a  $\delta$ -regular tree. By arguments similar to (b),  $P_v(\Gamma_v^\circ) \geq (1 + o(1))(\delta - 2)/(\delta - 1)$ .

(d) If  $G_v$  contains a unique cycle, and all vertices in  $G_v$  have degree at least  $d$ , the arguments in (a) can be modified to fit this case. By assumption, there are at most two cycle edges incident with  $v$ , and  $d(v) \geq d$  so

$$P_v(\Gamma_v^\circ) \geq \frac{d-2}{d} \frac{d-2}{d-1} + \frac{2}{d} \Phi,$$

where  $\Phi \geq 0$  is the probability of no return to  $v$  given a cycle edge, or an edge on a path to a cycle was taken at  $v$ .  $\square$

At this point, a brief summary may be useful.

- $G_v$  is a vertex induced subgraph of  $G$ . Up to absorption at  $\Gamma_v^\circ$ , the boundary of  $G_v$ , a walk starting from  $v$  in  $G$  is identically coupled with a walk on  $G_v$ .
- The escape probability  $P_v(\Gamma_v^\circ)$  from  $v$  of the walk  $\mathcal{W}_v^*$  has a precise value. For  $d$ -tree-regular vertices  $v$  it can be approximated by  $P_v(\Gamma_v^\circ) = (d-2)/(d-1)(1 + O(1/d^h))$ . Our choice of  $h$  (see (15)) ensures the error term is  $o(1)$ .
- By choosing  $\omega = C \log \log n$  as in (10), and  $C$  sufficiently large,  $1/R_v$  can be written as

$$\frac{1}{R_v} = P_v(\Gamma_v^\circ) + o\left(\frac{1}{\log n}\right). \quad (24)$$

The  $o(1/\log n)$  accuracy is needed in the proof of the lower bound on the cover time.

## 5 Cover time of $G(d)$

### 5.1 Upper bound on cover time

Let  $T_G(u)$  be the time taken by the random walk  $\mathcal{W}_u$  to visit every vertex of a connected graph  $G$ . Let  $U_t$  be the number of vertices of  $G$  which have not been visited by  $\mathcal{W}_u$  at step  $t$ . We note the following:

$$C_u = \mathbf{E}[T_G(u)] = \sum_{t>0} \Pr(T_G(u) \geq t), \quad (25)$$

$$\Pr(T_G(u) \geq t) = \Pr(T_G(u) > t-1) = \Pr(U_{t-1} > 0) \leq \min\{1, \mathbf{E}[U_{t-1}]\}. \quad (26)$$

Recall from (7) that  $\mathbf{A}_s(v)$  is the event that vertex  $v$  has not been visited during steps  $T, T+1, \dots, s$ . It follows from (25), (26) that

$$C_u \leq t+1 + \sum_{s \geq t} \mathbf{E}[U_s] \leq t+1 + \sum_v \sum_{s \geq t} \Pr(\mathbf{A}_s(v)). \quad (27)$$

Let  $t_0 = \left(\frac{d-1}{d-2}\frac{\theta}{d}\right) n \log n$  and  $t_1 = (1+\epsilon)t_0$ , where  $\epsilon = o(1)$  is sufficiently large that all inequalities claimed below hold. We assume that Lemma 7 holds, and also the high probability claims of Section 3. Thus Lemma 8 and Lemma 9 give values of  $R_v$  for all  $v \in V$ . In Section 6.4 of the Appendix, we establish that Condition (a) of Lemma 2 holds. The maximum degree of any vertex is  $n^a$ ,  $a < 1$ , and  $T = A \log n$  (see (21)), so Condition (b) of Lemma 2 that  $T\pi_v = o(1)$ , holds trivially.

Recall from (6) that  $p_v = (1 + O(T\pi_v))d(v)/(\theta n R_v)$ . Thus by (7), the probability that  $\mathcal{W}_u$  has not visited  $v$  during  $[T, t]$  is given by

$$\Pr(\mathbf{A}_t(v)) = (1 + o(1))e^{-tp_v} + O(T^2\pi_v e^{-\lambda t/2}) \quad (28)$$

$$= (1 + o(1))e^{-tp_v}. \quad (29)$$

Thus

$$\begin{aligned} \sum_{t \geq t_1} (1 + o(1))e^{-tp_v} &= (1 + o(1))e^{-t_1 p_v} \sum_{(t-t_1) \geq 0} e^{-(t-t_1)p_v} \\ &= \frac{(1 + o(1))}{1 - e^{-p_v}} e^{-t_1 p_v} \\ &= O(1) \frac{\theta n R_v}{d(v)} \exp \left\{ -(1 + \Theta(\epsilon)) \frac{d(v)}{d} \frac{d-1}{d-2} \frac{\log n}{R_v} \right\}. \end{aligned} \quad (30)$$

We consider the following partition of  $V$ :

(i)  $V_A = \bigcup_v \{G_v \text{ contains a small vertex}\}$ .

- (ii)  $V_B = \bigcup_{i \geq d} \{d(v) = i : v \text{ is } d\text{-compliant}\}$ .  
(iii)  $V_C = \bigcup_{i \geq d} \{d(v) = i : G_v \text{ contains a cycle}\}$ .

*Case (i):  $G_v$  contains a small vertex.*

By Lemma 4(c) there are  $O(\ell^\omega n^{ci/d})$  vertices  $v$  for which  $G_v$  contains a vertex of degree  $i < d$ , By Lemma 9(c),  $R_v \leq (1 + o(1)) \frac{\delta-1}{\delta-2}$ . Also  $G_v$  can contain at most one small vertex of degree  $i < d$ , so  $d(v) \geq i$ . Thus (30) is bounded by

$$O(\theta n) n^{-(1+o(1)) \frac{i}{d} \frac{d-1}{d-2} \frac{\delta-2}{\delta-1}} \leq O(\theta n) n^{-(1+o(1)) \frac{i}{d} \frac{\delta(\delta-2)}{(\delta-1)^2}}.$$

The term  $\delta(\delta-2)/(\delta-1)^2 \geq 3/4$ , whereas  $c < 1/8$ . Thus

$$\sum_{\delta \leq i < d} \sum_{v \in V_i} \sum_{t \geq t_1} (1 + o(1)) e^{-tp_v} \leq O(\theta n) \sum_{\delta \leq i < d} n^{ci/d} n^{-(1+o(1))3i/4d} = o(t_1).$$

*Case (ii):  $d \leq d(v)$ ,  $v$  is  $d$ -compliant.*

Note that this includes the  $d$ -tree-regular case. For  $v \in V_B$  (30) is bounded by  $O(\theta) n^{-\Theta(\epsilon)}$ . Therefore

$$\sum_{v \in V_B} \sum_{t \geq t_1} (1 + o(1)) e^{-tp_v} \leq \sum_{v \in V_B} O(\theta) n^{-\Theta(\epsilon)} = O(\theta n) n^{-\Theta(\epsilon)} = o(t_1).$$

*Case (iii):  $d \leq d(v)$ ,  $G_v$  contains a cycle.*

These vertices  $v \in V_C$ ,  $R_v$  is given by Lemma 9(d). Thus (30) is bounded by  $O(\theta n) n^{-(1+\Theta(\epsilon)) \frac{d(d-1)}{(d-2)^2}}$ . By Lemma 4,  $|V_C| \leq n^{\epsilon'}$  where  $\epsilon' > 0$  arbitrarily small, and so we choose  $2\epsilon' < d(d-1)/(d-2)^2$ . Hence

$$\begin{aligned} \sum_{v \in V_C} \sum_{t \geq t_1} (1 + o(1)) e^{-tp_v} &= \sum_{v \in V_C} O(\theta n) n^{-(1+\Theta(\epsilon)) \frac{d(d-1)}{(d-2)^2}} \\ &= O(\theta n) n^{\epsilon'} n^{-(1+\Theta(\epsilon)) \frac{d(d-1)}{(d-2)^2}} \\ &= o(t_1). \end{aligned}$$

In each of the cases above, the term  $\sum_v \sum_{s \geq t} \Pr(\mathbf{A}_s(v)) = o(t_1)$ . Thus, from (27),  $C_u \leq (1 + o(1))t_1$  as required. This completes the proof of the upper bound on cover time of  $G(\mathbf{d})$ .  $\square$

## 5.2 Lower bound on cover time

Let  $t_2 = (1 - \epsilon)t_0$ , where  $\epsilon = o(1)$  is sufficiently large that all inequalities claimed below hold. To establish the lower bound, we exhibit a set of vertices  $S$  for which, the probability the set

$S$  is covered by a walk  $\mathcal{W}_u$  at time  $t_2$ , tends to zero. Hence  $T_G(u) > t_2$ , **whp** which implies that  $C(G) \geq t_0 - o(t_0)$ .

We construct  $S$  as follows. Let  $S_d$  be the set of  $d$ -tree-regular vertices. Lemma 6 tells us that  $|S_d| = n^{1-o(1)}$ . Let  $\omega = C \log \log n$  for some large  $C$ , as in (10). Let  $S$  be a maximal subset of  $S_d$  such that the distance between any two elements of  $S$  is least  $2\omega + 1$ . Thus  $|S| = \Omega(n^{1-o(1)}/\ell^{2\omega})$ .

Let  $S(t)$  denote the subset of  $S$  which is still un-visited after step  $t$  of  $\mathcal{W}_u$ . Let  $v \in S$ , then

$$\Pr(\mathbf{A}_v(t_2)) = (1 + o(1))e^{-t_2 p_v(1-O(p_v))} + o(n^{-2}).$$

Hence

$$\mathbf{E}(|S(t_2)|) \geq (1 + o(1))|S|e^{-(1-\epsilon)t_0 p_v} - O(T) \quad (31)$$

$$= \Omega\left(\frac{n^{\epsilon/2-o(1)}}{\ell^{2\omega}}\right) \rightarrow \infty. \quad (32)$$

The term  $O(T)$  above, counts vertices of  $S$  visited during the first  $T$  steps of the walk. Let  $Y_{v,t}$  be the indicator for the event  $\mathbf{A}_t(v)$ . Let  $Z = \{v, w\} \subset S$ . We will show (below) that that for  $v, w \in S$

$$\mathbf{E}(Y_{v,t_2} Y_{w,t_2}) = \frac{1 + O(T\pi_v)}{(1 + p_Z)^{t_2}} + o(n^{-2}), \quad (33)$$

where

$$p_Z = p_v + p_w + o\left(\frac{d}{\theta n \log n}\right). \quad (34)$$

Thus

$$\mathbf{E}(Y_{v,t_2} Y_{w,t_2}) = (1 + o(1))\mathbf{E}(Y_{v,t_2})\mathbf{E}(Y_{w,t_2})$$

which implies

$$\mathbf{E}(|S(t_2)|(|S(t_2)| - 1)) \sim \mathbf{E}(|S(t_2)|)(\mathbf{E}(|S(t_2)|) - 1). \quad (35)$$

It follows from (32) and (35), that

$$\Pr(S(t_2) \neq \emptyset) \geq \frac{\mathbf{E}(|S(t_2)|)^2}{\mathbf{E}(|S(t_2)|^2)} = \frac{1}{\frac{\mathbf{E}(|S(t_2)|(|S(t_2)|-1))}{\mathbf{E}(|S(t_2)|)^2} + \mathbf{E}(|S(t_2)|)^{-1}} = 1 - o(1).$$

**Proof of (33)-(34).** Let  $\widehat{G}$  be obtained from  $G$  by merging  $v, w$  into a single vertex  $Z$ . Let  $\rho$  be the expected number of passages between  $v, w$  in  $T$  steps. By construction, as  $G_w$  is a tree, whenever the walk arrives at  $\Gamma_w^\circ$  after leaving  $v$  it will have to traverse a unique path of length  $\omega$  to reach  $w$ . Using (23) and arguments similar to Lemma 9, we find  $\rho = O(T^2/(d-1)^\omega) = o(1/\log n)$ . Thus Lemma 8 is valid for  $\widehat{G}$ .

There is a natural measure-preserving map from the set of walks in  $G$  which start at  $u$  and do not visit  $v$  or  $w$ , to the corresponding set of walks in  $\widehat{G}$  which do not visit  $Z$ . Thus

the probability that  $\mathcal{W}_u$  does not visit  $v$  or  $w$  in steps  $T\dots t$  is asymptotically equal to the probability that a random walk  $\widehat{\mathcal{W}}_u$  in  $\widehat{G}$  which also starts at  $u$  does not visit  $Z$  in steps  $T\dots t$ . The detailed argument is given in [6].

We apply Lemma 2 to  $\widehat{G}$ . The value of  $\pi_Z = 2d/\theta n$ . The vertex  $Z$  has degree  $2d$  and  $G_Z$  is otherwise  $d$ -tree-regular, as  $G_v, G_w$  are vertex disjoint. The derivation of  $R_Z^*$ , can be made as follows. The escape probability from  $Z$  to  $\Gamma_Z^\circ$  is given by

$$P_Z(\Gamma_Z^\circ) = \frac{1}{2}(P_v(\Gamma_v^\circ) + P_w(\Gamma_w^\circ)),$$

as with probability  $1/2$  the walk takes a  $v$ -edge at  $Z$  and escapes to  $\Gamma_v^\circ$  etc. Thus, as claimed in (34),

$$p_Z = \frac{\pi_Z}{R_Z}(1 + O(T\pi_Z)) = \frac{2d}{\theta n}(P_Z(\Gamma_Z^\circ) + o(1/\log n)) = p_v + p_w + o(d/\theta n \log n).$$

□

**Acknowledgement:** We thank several referees whose insight, hard work and detailed critique has helped to make this paper correct and (hopefully) more readable.

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## 6 Appendix

### 6.1 Proof of Lemma 2

#### Generating function formulation

Let  $d_t = \max_{u,x \in V} |P_u^{(t)}(x) - \pi_x|$ , and let  $T$  be such that, for  $t \geq T$

$$\max_{u,x \in V} |P_u^{(t)}(x) - \pi_x| \leq n^{-3}. \quad (36)$$

It follows from e.g. Aldous and Fill [1] that  $d_{s+t} \leq 2d_s d_t$  and so for  $k \geq 1$ ,

$$\max_{u,x \in V} |P_u^{(kT)}(x) - \pi_x| \leq \frac{2^{k-1}}{n^{3k}}. \quad (37)$$

Fix two vertices  $u, v$ . Let  $h_t = \mathbf{Pr}(\mathcal{W}_u(t) = v)$  be the probability that the walk  $\mathcal{W}_u$  visits  $v$  at step  $t$ . Let

$$H(z) = \sum_{t=T}^{\infty} h_t z^t \quad (38)$$

generate  $h_t$  for  $t \geq T$ .

Next, considering the walk  $\mathcal{W}_v$ , starting at  $v$ , let  $r_t = \mathbf{Pr}(\mathcal{W}_v(t) = v)$  be the probability that this walk returns to  $v$  at step  $t = 0, 1, \dots$ . Let

$$R(z) = \sum_{t=0}^{\infty} r_t z^t$$

generate  $r_t$ . Our definition of return includes the term  $r_0 = 1$ .

For  $t \geq T$  let  $f_t = f_t(u \rightarrow v)$  be the probability that the first visit of the walk  $\mathcal{W}_u$  to  $v$  in the period  $[T, T + 1, \dots]$  occurs at step  $t$ . Let

$$F(z) = \sum_{t=T}^{\infty} f_t z^t$$

generate  $f_t$ . Then we have

$$H(z) = F(z)R(z). \quad (39)$$

### First visit time lemma

For  $R(z)$  let

$$R_T(z) = \sum_{j=0}^{T-1} r_j z^j. \quad (40)$$

Let

$$\lambda = \frac{1}{KT} \quad (41)$$

for some sufficiently large constant  $K$ .

**Lemma 10.** *Suppose that*

(a) *For some constant  $\psi > 0$ , we have*

$$\min_{|z| \leq 1 + \lambda} |R_T(z)| \geq \psi.$$

(b)  $T\pi_v = o(1)$  and  $T\pi_v = \Omega(n^{-2})$ .

*There exists*

$$p_v = \frac{\pi_v}{R_T(1)(1 + O(T\pi_v))}, \quad (42)$$

where  $R_T(1)$  is from (40), such that for all  $t \geq T$ ,

$$f_t(u \rightarrow v) = (1 + O(T\pi_v)) \frac{p_v}{(1 + p_v)^{t+1}} + O(T\pi_v e^{-\lambda t/2}). \quad (43)$$

**Proof** Write

$$R(z) = R_T(z) + \widehat{R}_T(z) + \frac{\pi_v z^T}{1 - z}, \quad (44)$$

where  $R_T(z)$  is given by (40) and

$$\widehat{R}_T(z) = \sum_{t \geq T} (r_t - \pi_v) z^t$$

generates the error in using the stationary distribution  $\pi_v$  for  $r_t$  when  $t \geq T$ . Similarly,

$$H(z) = \widehat{H}_T(z) + \frac{\pi_v z^T}{1-z}. \quad (45)$$

Equation (37) implies that the radii of convergence of both  $\widehat{R}_T$  and  $\widehat{H}_T$  exceed  $1+2\lambda$ . Moreover, for  $Z = H, R$  and  $|z| \leq 1 + \lambda$ ,

$$|\widehat{Z}(z)| = o(n^{-2}). \quad (46)$$

Using (44), (45) we rewrite  $F(z) = H(z)/R(z)$  from (39) as  $F(z) = B(z)/A(z)$  where

$$A(z) = \pi_v z^T + (1-z)(R_T(z) + \widehat{R}_T(z)), \quad (47)$$

$$B(z) = \pi_v z^T + (1-z)\widehat{H}_T(z). \quad (48)$$

For real  $z \geq 1$  and  $Z = H, R$ , we have

$$Z_T(1) \leq Z_T(z) \leq Z_T(1)z^T.$$

Let  $z = 1 + \beta\pi_v$ , where  $\beta = O(1)$ . Since  $T\pi_v = o(1)$  we have

$$Z_T(z) = Z_T(1)(1 + O(T\pi_v)).$$

$T\pi_v = o(1)$  and  $T\pi_v = \Omega(n^{-2})$  and  $R_T(1) \geq 1$  implies that

$$A(z) = \pi_v(1 - \beta R_T(1) + O(T\pi_v))$$

It follows that  $A(z)$  has a real zero at  $z_0$ , where

$$z_0 = 1 + \frac{\pi_v}{R_T(1)(1 + O(T\pi_v))} = 1 + p_v, \quad (49)$$

say. We also see that

$$A'(z_0) = -R_T(1)(1 + O(T\pi_v)) \neq 0 \quad (50)$$

and thus  $z_0$  is a simple zero (see e.g. [4] p193). The value of  $B(z)$  at  $z_0$  is

$$B(z_0) = \pi_v(1 + O(T\pi_v)) \neq 0. \quad (51)$$

Thus,

$$\frac{B(z_0)}{A'(z_0)} = -(1 + O(T\pi_v))p_v. \quad (52)$$

Thus (see e.g. [4] p195) the principal part of the Laurent expansion of  $F(z)$  at  $z_0$  is

$$f(z) = \frac{B(z_0)/A'(z_0)}{z - z_0}. \quad (53)$$

To approximate the coefficients of the generating function  $F(z)$ , we now use a standard technique for the asymptotic expansion of power series (see e.g.[12] Th 5.2.1).

We prove below that  $F(z) = f(z) + g(z)$ , where  $g(z)$  is analytic in  $C_\lambda = \{|z| \leq 1 + \lambda\}$  and that  $M = \max_{z \in C_\lambda} |g(z)| = O(T\pi_v)$ .

Let  $a_t = [z^t]g(z)$ , then (see e.g.[4] p143),  $a_t = g^{(t)}(0)/t!$ . By the Cauchy Inequality (see e.g. [4] p130) we see that  $|g^{(t)}(0)| \leq Mt!/(1 + \lambda)^t$  and thus

$$|a_t| \leq \frac{M}{(1 + \lambda)^t} = O(T\pi_v e^{-t\lambda/2}).$$

As  $[z^t]F(z) = [z^t]f(z) + [z^t]g(z)$  and  $[z^t]1/(z - z_0) = -1/z_0^{t+1}$  we have

$$[z^t]F(z) = \frac{-B(z_0)/A'(z_0)}{z_0^{t+1}} + O(T\pi_v e^{-t\lambda/2}). \quad (54)$$

Thus, we obtain

$$[z^t]F(z) = (1 + O(T\pi_v)) \frac{p_v}{(1 + p_v)^{t+1}} + O(T\pi_v e^{-t\lambda/2}),$$

which completes the proof of (43).

Now  $M = \max_{z \in C_\lambda} |g(z)| \leq \max |f(z)| + \max |F(z)| = O(T\pi_v) + \max |F(z)|$ , where  $F(z) = B(z)/A(z)$ . On  $C_\lambda$  we have, using (46)-(48),

$$|F(z)| \leq \frac{O(\pi_v)}{\lambda |R_T(z)| - O(T\pi_v)} = O(T\pi_v).$$

We now prove that  $z_0$  is the only zero of  $A(z)$  inside the circle  $C_\lambda$  and this implies that  $F(z) - f(z)$  is analytic inside  $C_\lambda$ . We use Rouché's Theorem (see e.g. [4]), the statement of which is as follows: *Let two functions  $\phi(z)$  and  $\gamma(z)$  be analytic inside and on a simple closed contour  $C$ . Suppose that  $|\phi(z)| > |\gamma(z)|$  at each point of  $C$ , then  $\phi(z)$  and  $\phi(z) + \gamma(z)$  have the same number of zeroes, counting multiplicities, inside  $C$ .*

Let the functions  $\phi(z), \gamma(z)$  be given by  $\phi(z) = (1 - z)R_T(z)$  and  $\gamma(z) = \pi_v z^T + (1 - z)\widehat{R}_T(z)$ .

$$|\gamma(z)|/|\phi(z)| \leq \frac{\pi_v(1 + \lambda)^T}{\lambda\psi} + \frac{|\widehat{R}_T(z)|}{\psi} = o(1).$$

As  $\phi(z) + \gamma(z) = A(z)$  we conclude that  $A(z)$  has only one zero inside the circle  $C_\lambda$ . This is the simple zero at  $z_0$ .  $\square$

**Corollary 11.** For  $t \geq T$  let  $\mathbf{A}_t(v)$  be the event that  $\mathcal{W}_u$  does not visit  $v$  in steps  $T, T+1, \dots, t$ . Then, under the assumptions of Lemma 10,

$$\Pr(\mathbf{A}_t(v)) = \frac{(1 + O(T\pi_v))}{(1 + p_v)^t} + O(T^2\pi_v e^{-\lambda t/2}).$$

**Proof** We use Lemma 10 and

$$\Pr(\mathbf{A}_t(v)) = \sum_{\tau > t} f_\tau(u \rightarrow v).$$

□

## 6.2 Proof of conductance bound in Lemma 7

By the conductance of a configuration  $C$ , we mean the conductance of a random walk on the underlying multi-graph  $M(C)$ . It is however, the configurations we sample  $\mathbf{uar}$  in the proof of Lemma 12.

**Lemma 12.** Let  $\mathbf{d} = (d_1, d_2, \dots, d_n)$  be a sequence of natural numbers, satisfying  $\min d_i \geq 3$  and  $\theta \leq n^{1/4}$ . With probability  $1 - o(n^{-1/9})$  the conductance  $\Phi$  of a  $\mathbf{uar}$  sampled configuration  $C(\mathbf{d})$  satisfies  $\Phi \geq 0.01$ .

**Proof** Let  $F(a) = a!/((a/2)!2^{(a/2)})$ . With this notation,

$$\frac{F(b)F(a-b)}{F(a)} = \frac{\binom{a/2}{b/2}}{\binom{a}{b}} = O(1) \left(\frac{b}{a}\right)^{b/2} \left(1 - \frac{b}{a}\right)^{(a-b)/2}. \quad (55)$$

For any  $S \subseteq V$  let  $d(S)$  denote the sum of the degrees of the vertices of  $S$ . A set  $S$  is *small* if  $d(S) \leq (\theta n)^{1/4}$ . A set is *large* if  $(\theta n)^{1/4} \leq d(S) \leq \theta n/2$ . Let  $8/9 < \beta < 1$  be a positive constant.

**SMALL SETS** ( $\delta|S| \leq d(S) \leq (\theta n)^{1/4}$ ).

Let  $N(s, \beta)$  be the expected number of small sets  $S$  of size  $s$  with at least  $\beta d(S)$  induced edges.

$$N(s, \beta) = \sum_S \binom{d(S)}{\beta d(S)} \frac{F(\beta d(S))F(\theta n - \beta d(S))}{F(\theta n)}. \quad (56)$$

Noting that  $\binom{L}{k} \leq (Le/k)^k$  and using (55) with  $\delta s \leq d(S) \leq (\theta n)^{1/4}$  and  $\delta \geq 3$  we find

$$\begin{aligned}
N(s, \beta) &\leq O(1) \sum_S \left( \frac{d(S)e}{\beta d(S)} \right)^{\beta d(S)} \left( \frac{\beta d(S)}{\theta n} \right)^{\beta d(S)/2} \left( 1 - \frac{\beta d(S)}{\theta n} \right)^{(\theta n - \beta d(S))/2} \\
&\leq O(1) \sum_S \left( \frac{e}{\beta} \right)^{\beta d(S)} \left( \frac{\beta d(S)}{\theta n} \right)^{\beta d(S)/2} \\
&\leq O(1) \left( \frac{ne}{s} \left( \frac{e^2}{\beta(\theta n)^{3/4}} \right)^{3\beta/2} \right)^s \\
&= O((e^4 n^{-(9\beta/8-1)})^s).
\end{aligned}$$

Thus

$$\sum_{\substack{|S|=s \\ S \text{ Small}}} N(s, \beta) = O(n^{-(9\beta/8-1)}).$$

**LARGE SETS**  $((\theta n)^{1/4} \leq d(S) \leq \theta n/2)$ .

Let  $N(s, \beta)$  be the expected number of large sets  $S$  of size  $s$  inducing at least  $\beta d(S)$  edges. As before,  $N(s, \beta)$  is given by (56). Let  $d(S) = \alpha \theta n$  where  $0 < \alpha \leq 1/2$ . Let  $\varepsilon = 1 - \beta$ . We note the following approximation:

$$\binom{d(s)}{\beta d(S)} = \binom{\alpha \theta n}{\beta \alpha \theta n} = \frac{O(1)}{\sqrt{\varepsilon \beta \alpha \theta n}} \frac{1}{\beta^{\beta \alpha \theta n} \varepsilon^{\varepsilon \alpha \theta n}}.$$

Thus

$$N(s, \beta) \leq \sum_S \frac{O(1)}{\sqrt{\varepsilon \beta \alpha \theta n}} \left( \frac{(\alpha \beta)^{\alpha \beta} (1 - \alpha \beta)^{1 - \alpha \beta}}{(\varepsilon^\varepsilon \beta^\beta)^{2\alpha}} \right)^{\frac{\theta n}{2}} = \sum_S f(S). \quad (57)$$

Let  $s = cn$ . We henceforth assume that we choose the value  $\alpha = \alpha^*$  which maximizes  $f(S)$  for  $|S| = cn$ . With this convention we can write

$$N(cn, \beta) \leq \frac{O(1)}{\sqrt{\varepsilon \beta c(1-c) \alpha \theta n^2}} \left( \left( \frac{(\alpha \beta)^{\alpha \beta} (1 - \alpha \beta)^{1 - \alpha \beta}}{(\varepsilon^\varepsilon \beta^\beta)^{2\alpha}} \right)^{\frac{\theta}{2}} \frac{1}{c^c (1-c)^{1-c}} \right)^n. \quad (58)$$

We split the proof for large sets into two parts: Those sets for which  $\alpha \leq 1/\theta$  and those for which  $1/\theta \leq \alpha \leq 1/2$ .

*Case of  $\alpha \leq 1/\theta$ .*

We need to remove the dependence on  $c$  in the right hand side of the expression (58) for  $N(cn, \beta)$ . We first deal with the square root term. Since  $\frac{1}{n} \leq c \leq \frac{(n-1)}{n}$ , we have that  $c(1-c) \geq \frac{n-1}{n^2}$  and so

$$c(1-c) \alpha \theta n^2 \geq \frac{n-1}{n^2} (\theta n)^{1/4} n \geq (\theta n)^{1/4} / 2.$$

Therefore, as  $\beta, \varepsilon$  are positive constants,

$$\frac{1}{\sqrt{\varepsilon\beta c(1-c)\alpha\theta n^2}} = \frac{O(1)}{(\theta n)^{1/8}}.$$

We next consider the main term of (58). For  $0 \leq x \leq 1/2$ , the function

$$g(x) = x^x(1-x)^{1-x}$$

satisfies,  $g(0) = 1$  and is monotonically decreasing with minimum  $g(1/2) = 1/2$ .

Since  $d(S) \geq 3s$ , and  $s = cn$ , from  $d(S) = \alpha\theta n$  we deduce that  $c \leq \alpha\theta/3$ . As  $\alpha \leq 1/\theta$  then  $c \leq \alpha\theta/3 \leq 1/3$ . Therefore  $g(c) \geq g(\alpha\theta/3)$ , and we can replace  $c$  by  $\alpha\theta/3$  in (58). Hence

$$\begin{aligned} N(cn, \beta) &= \frac{O(1)}{(\theta n)^{1/8}} \left( \frac{(\alpha\beta)^{\alpha\beta\theta/2}(1-\alpha\beta)^{1-\alpha\beta\theta/2}}{(\alpha\theta/3)^{\alpha\theta/3}(1-\alpha\theta/3)^{1-\alpha\theta/3}} \frac{(1-\alpha\beta)^{\theta/2-1}}{(\varepsilon^\varepsilon\beta^\beta)^{\alpha\theta}} \right)^n \\ &= \frac{O(1)}{(\theta n)^{1/8}} (\phi(\alpha, \beta, \theta))^n. \end{aligned}$$

We next maximize  $\phi(\alpha, \beta, \theta)$ . Let  $h(x, y) = (yx)^x(1-yx)^{1-x}$  for  $0 < x, y \leq 1$ . Considering  $h(x, y)$  as a function of  $y$ , there is a unique maximum at  $y = 1$ , given by

$$\begin{aligned} \frac{\partial}{\partial y} \log(h(x, y)) &= x \left( \frac{1}{y} - \frac{1-x}{1-yx} \right) = 0, \\ \frac{\partial^2}{\partial y^2} \log(h(x, y)) &= -x \left( \frac{1}{y^2} + \frac{x(1-x)}{(1-yx)^2} \right) < 0. \end{aligned}$$

Therefore  $h(x, y) < h(x, 1) = g(x)$ . So  $h(\alpha\beta\theta/2, 2/\theta) < g(\alpha\beta\theta/2) < g(\alpha\theta/3)$ . Hence

$$\phi(\alpha, \beta, \theta) \leq \frac{(1-\alpha\beta)^{\theta/2-1}}{(\varepsilon^\varepsilon\beta^\beta)^{\alpha\theta}}.$$

We prove below, that

$$\frac{\partial}{\partial \theta} \left\{ \frac{(1-\alpha\beta)^{\theta/2-1}}{(\varepsilon^\varepsilon\beta^\beta)^{\alpha\theta}} \right\} < 0. \quad (59)$$

Since  $\theta \geq \delta \geq 3$ , we have that

$$\frac{(1-\alpha\beta)^{\theta/2-1}}{(\varepsilon^\varepsilon\beta^\beta)^{\alpha\theta}} \leq \frac{e^{-\alpha\beta/2}}{(\varepsilon^\varepsilon\beta^\beta)^{3\alpha}} \leq \lambda^\alpha,$$

where  $\lambda < 0.7$ , provided  $\beta \geq 0.99$ .

Now since  $\alpha\theta n \geq (\theta n)^{1/4}$  for large sets, and  $\theta \leq n^{1/4}$  by conditions of the lemma, we have that  $\alpha n \geq n^{1/16}$ . Thus

$$\begin{aligned} N(cn, \beta) &= \frac{O(1)}{(\theta n)^{1/8}} (\phi(\alpha, \beta, \theta))^n \\ &= O(\lambda^{n^{1/16}}). \end{aligned}$$

As  $s = cn$  can take at most  $n$  values we have that  $\sum N(cn, \beta) = O(n\lambda^{n^{1/16}})$ .

**Proof of (59).**

$$\frac{\partial}{\partial \theta} \left\{ \frac{(1 - \alpha\beta)^{\theta/2-1}}{(\varepsilon^\varepsilon \beta^\beta)^{\alpha\theta}} \right\} = \frac{1}{1 - \alpha\beta} \left( \frac{(1 - \alpha\beta)^{\frac{1}{2}}}{(\varepsilon^\varepsilon \beta^\beta)^\alpha} \right)^\theta \log \left( \frac{(1 - \alpha\beta)^{\frac{1}{2}}}{(\varepsilon^\varepsilon \beta^\beta)^\alpha} \right).$$

Let

$$f(\alpha, \beta) = \frac{(1 - \alpha\beta)}{(\varepsilon^\varepsilon \beta^\beta)^{2\alpha}}.$$

When  $\alpha = 0$ ,  $f(\alpha, \beta) = 1$ . We prove that, for  $\beta \geq 0.99$ ,  $f(\alpha, \beta) < 1$  for  $\alpha > 0$ , which will establish the result. Note that

$$\frac{\partial}{\partial \alpha} f(\alpha, \beta) = \frac{-1}{(\varepsilon^\varepsilon \beta^\beta)^{2\alpha}} (\beta + (1 - \alpha\beta) \log(\varepsilon^\varepsilon \beta^\beta)^2). \quad (60)$$

Consider

$$\begin{aligned} \frac{d}{d\beta} \{ \log(\varepsilon^\varepsilon \beta^\beta)^2 + \beta \} &\equiv \frac{d}{d\beta} \{ \log((1 - \beta)^{1-\beta} \beta^\beta)^2 + \beta \} \\ &= 2 \log \left( \frac{\beta}{1 - \beta} \right) + 1. \end{aligned}$$

For  $\beta > \frac{1}{2}$ , the last line above is positive, and thus  $\log(\varepsilon^\varepsilon \beta^\beta)^2 > -\beta$ . It follows that (60) is negative, as required.

*Case of  $1/\theta \leq \alpha \leq 1/2$ .*

Continuing to evaluate  $N(s, \beta)$  as before, and referring to  $f(S)$  as given by the right hand side term of (57), let

$$A(\alpha) = \frac{(\alpha\beta)^{\alpha\beta} (1 - \alpha\beta)^{1-\alpha\beta}}{(\varepsilon^\varepsilon \beta^\beta)^{2\alpha}}.$$

Thus

$$\begin{aligned} \log(A(\alpha)) &= (\alpha\beta) \log((\alpha\beta)) + (1 - \alpha\beta) \log(1 - \alpha\beta) - 2\alpha \log(\varepsilon^\varepsilon \beta^\beta), \\ \frac{\partial}{\partial \alpha} \log(A(\alpha)) &= \beta \log(\alpha\beta) - \beta \log(1 - \alpha\beta) - 2 \log(\varepsilon^\varepsilon \beta^\beta). \end{aligned}$$

Setting  $\frac{\partial}{\partial \alpha} \log(A(\alpha)) = 0$  gives

$$\alpha = \frac{\varepsilon^{2\varepsilon/\beta} \beta}{1 + \varepsilon^{2\varepsilon/\beta} \beta^2}.$$

Let  $\alpha_0$  be the solution to this when  $\beta = 0.99$ . Thus  $\alpha_0 \approx 0.477$ . Also,

$$\frac{\partial^2}{\partial \alpha^2} \log(A(\alpha)) = \beta \left( \frac{1}{\alpha} + \frac{\beta}{1 - \alpha\beta} \right) > 0$$



hence the stationary point  $\alpha_0$  is a minima. As  $\theta \geq 3$  and by inspection,  $A(0.5) < A(1/3)$  then  $A(\alpha_0) \leq A(1/\theta)$ . We can use  $\alpha^* = 1/\theta$  as the value of  $\alpha$  maximizing  $A(\alpha)$  in the range  $1/\theta \leq \alpha \leq 1/2$ . It follows that

$$\begin{aligned} \sum_{\substack{S \text{ Large} \\ \alpha \geq 1/\theta}} f(S) &= \left( \frac{1}{\sqrt{\theta n}} \right) 2^n (A(1/\theta))^{\frac{\theta n}{2}} \\ &= O(1) 2^n \left( \frac{(\beta/\theta)^{\frac{\beta}{2}} (1 - \beta/\theta)^{\frac{1}{2}(\theta - \beta)}}{\varepsilon^\varepsilon \beta^\beta} \right)^n. \end{aligned}$$

Let

$$T(\theta) = \left( \frac{\beta}{\theta} \right)^\beta \left( 1 - \frac{\beta}{\theta} \right)^{\theta - \beta},$$

then

$$\frac{\partial}{\partial \theta} \log(T(\theta)) = \log \left( \frac{\theta - \beta}{\theta} \right).$$

Thus  $T(\theta)$  is monotone decreasing in  $\theta$ , and so  $T(\theta) \leq T(3)$ . Finally

$$\begin{aligned} \sum N(s, \beta) &\leq O(n) 2^n \left( \frac{(\beta/3)^{\frac{\beta}{2}} (1 - \beta/3)^{\frac{1}{2}(3 - \beta)}}{\varepsilon^\varepsilon \beta^\beta} \right)^n \\ &= O(n (0.8)^n). \end{aligned}$$

This completes the proof of the lemma. □

### 6.3 Proof of Lemma 8

For convenience, we restate the lemma.

**Lemma 13.** *Let  $\mathcal{W}_v^*$  denote the walk on  $G_v$  starting at  $v$  with  $\Gamma_v^\circ$  made into an absorbing state. Let  $R_v^* = \sum_{t=0}^{\infty} r_t^*$  where  $r_t^*$  is the probability that  $\mathcal{W}_v^*$  is at vertex  $v$  at time  $t$ . There exists a constant  $\zeta \in (0, 1)$  such that*

$$R_v = R_v^* + O(\zeta^\omega).$$

**Proof** We bound  $|R_v - R_v^*|$  by using

$$R_v - R_v^* = \sum_{t=0}^{\omega} (r_t - r_t^*) + \sum_{t=\omega+1}^T (r_t - r_t^*) - \sum_{t=T+1}^{\infty} r_t^*. \quad (61)$$

Case  $t \leq \omega$ . When a particle starting from  $v$  is absorbed at  $\Gamma_v^\circ$ , this is at distance  $\omega$  from  $v$ . Thus for  $t < \omega$ ,  $r_t^* = r_t$ , and

$$\sum_{t=0}^{\omega} (r_t - r_t^*) = 0. \quad (62)$$

Case  $\omega + 1 \leq t \leq T$ . Using (20) with  $x = u = v$  and  $\zeta = (1 - \Phi^2/2) < 1$ , we have for  $t \geq \omega$ , that  $r_t = \pi_v + O(\zeta^t)$ . Since  $\Delta = O(n^a)$ ,  $a < 1$ , we have  $T\pi_v = o(\zeta^\omega)$  and so

$$\sum_{t=\omega+1}^T |r_t - r_t^*| = \sum_{t=\omega+1}^T r_t \leq \sum_{t=\omega+1}^T (\pi_v + \zeta^t) = O(\zeta^\omega). \quad (63)$$

Case  $t \geq T + 1$ . It remains to estimate  $\sum_{t=T+1}^\infty r_t^*$ . We upper bound  $r_t^*$  by a probability  $\sigma_t$  as follows. Assume first that  $G_v$  is a tree. Consider an unbiased random walk  $X_0^{(b)}, X_1^{(b)}, \dots$  starting at  $|b| < a \leq \omega$  on the infinite line  $(\dots, -a, \dots, -1, 0, 1, \dots, a, \dots)$ .  $X_m^{(b)}$  is the sum of  $m$  independent  $\pm 1$  random variables. The central limit theorem implies that there exists a constant  $c > 0$  such that

$$\Pr(|X_{ca^2}^{(0)}| < a) \leq e^{-1/2}. \quad (64)$$

Now for any  $t$  and  $b$  with  $|b| < a$ , we have

$$\Pr(|X_\tau^{(b)}| < a, \tau = 0, \dots, t) \leq \Pr(|X_\tau^{(0)}| < a, \tau = 0, \dots, t) \quad (65)$$

which is justified with the following game: We have two walks,  $A$  and  $B$  coupled to each other, with  $A$  starting at position 0 and  $B$  at position  $b$ , which, w.l.o.g, we shall assume is positive. The walk is a simple random walk which comes to a halt when either of the walks hits an absorbing state (that being,  $-a$  or  $a$ ). Since they are coupled,  $B$  will win iff they drift  $(a - b)$  to the right from 0 and  $A$  will win iff they drift  $-a$  to the left from 0. Given the symmetry of the walk,  $B$  has a higher chance of winning.

For  $t > T$ , we define  $\sigma_t$  by

$$\sigma_t = \Pr(|X_\tau^{(0)}| < a, \tau = 0, 1, \dots, t) \leq (e^{-1/2})^{\lfloor t/(ca^2) \rfloor}. \quad (66)$$

The paths from  $v$  to  $\Gamma_v^\circ$  in the tree satisfy  $a \leq \omega$ , and so

$$\sum_{t=T+1}^\infty \sigma_t \leq \sum_{t=T+1}^\infty e^{-t/(3c\omega^2)} \leq \frac{e^{-T/(3c\omega^2)}}{1 - e^{-1/(3c\omega^2)}} = O(\omega^2 e^{-\Theta(\frac{\log n}{\omega^2})}) = O(\zeta^\omega)$$

We now turn to the case where  $G_v$  contains a unique light cycle  $C$ . Let  $x$  be the furthest vertex of  $C$  from  $v$  in  $G_v$ . This is the only possible place where the random walk is more likely to get closer to  $v$  at the next step. We can see this by considering the breadth first construction of  $G_v$ . Thus we can compare our walk with random walk on  $[-a, a]$  where there is a unique value  $x < a$  such that only at  $\pm x$  is the walk more likely to move towards the origin and even then this probability is at most  $2/3$ . Using results (64), (65) for the unbiased walk on the line, we have

$$\Pr(\exists \tau \leq ca^2 : |X_\tau^{(b)}| \geq x) \geq 1 - e^{-1/2}.$$

The probability the particle walks from  $x$  to  $a$  without returning to the cycle is at least  $1/3(a-x)$ . Thus

$$\Pr(\exists \tau \leq ca^2 : |X_{\tau+a-x}^{(b)}| \geq a) \geq (1 - e^{-1/2})/3a \geq \frac{13}{100a},$$

and so

$$\sigma_t = \Pr(|X_\tau^{(0)}| < a, \tau = 0, 1, \dots, t) \leq (1 - 13/(100a))^{\lfloor t/(2ca^2) \rfloor} \leq e^{-t/(20ca^3)}. \quad (67)$$

As  $a \leq \omega$ ,

$$\sum_{t=T+1}^{\infty} \sigma_t \leq \sum_{t=T+1}^{\infty} e^{-t/(20c\omega^3)} \leq \frac{e^{-T/(20c\omega^3)}}{1 - e^{-1/(20c\omega^3)}} = O\left(\omega^3 e^{-O(\frac{\log n}{\omega^3})}\right) = O(\zeta^\omega)$$

□

## 6.4 Condition (a) of Lemma 2

**Lemma 14.** *For  $|z| \leq 1 + \lambda$ , there exists a constant  $\psi > 0$  such that  $|R_T(z)| \geq \psi$ .*

**Proof** As in Lemma 8, we consider the walk  $\mathcal{W}_v^*$  on  $G_v$ , starting from  $v$ , and with absorption at  $\Gamma_v^\circ$ . For this walk, let  $\beta_t$  be the probability of a first return to  $v$  at step  $t$ , and let  $r_t^*$  be the probability of a return to  $v$  at step  $t$ .

Let  $\beta(z) = \sum_{t=1}^T \beta_t z^t$ , let  $\alpha(z) = 1/(1 - \beta(z))$ , and write  $\alpha(z) = \sum_{t=0}^{\infty} \alpha_t z^t$ . Thus  $\alpha_t$  is the probability of a return to  $v$  at time  $t$  for a walk  $\mathcal{W}_v^\dagger$ , all of whose excursions from  $v$  are length at most  $T$ . Observe that  $\alpha_t \leq r_t^* \leq r_t$ . We shall prove below that the radius of convergence of  $\alpha(z)$  is at least  $1 + \Omega(1/\omega^3)$ .

We can write

$$\begin{aligned} R_T(z) &= \alpha(z) + Q(z) \\ &= \frac{1}{1 - \beta(z)} + Q(z), \end{aligned} \quad (68)$$

where  $Q(z) = Q_1(z) + Q_2(z)$ , and

$$\begin{aligned} Q_1(z) &= \sum_{t=0}^T (r_t - \alpha_t) z^t \\ Q_2(z) &= - \sum_{t=T+1}^{\infty} \alpha_t z^t. \end{aligned}$$

We note that  $Q(0) = 0$ ,  $\alpha(0) = 1$  and  $\beta(0) = 0$ .

We will show below that

$$|Q_2(z)| = o(1) \tag{69}$$

for  $|z| \leq 1 + 2\lambda$  and thus the radius of convergence of  $Q_2(z)$  (and hence  $\alpha(z)$ ) is greater than  $1 + \lambda$ . This will imply that  $|\beta(z)| < 1$  for  $|z| \leq 1 + \lambda$ , so that the expression (68) is well defined. For suppose there exists  $z_0$  such that  $|\beta(z_0)| \geq 1$ . Then  $\beta(|z_0|) \geq |\beta(z_0)| \geq 1$  and we can assume (by scaling) that  $\beta(|z_0|) = 1$ . We have  $\beta(0) < 1$  and so we can assume that  $\beta(|z|) < 1$  for  $0 \leq |z| < |z_0|$ . But as  $\rho$  approaches 1 from below, (68) is valid for  $z = \rho|z_0|$  and then  $|R_T(\rho|z_0|)| \rightarrow \infty$ , contradiction.

Recall that  $\lambda = 1/KT$ . Clearly  $\beta(1) \leq 1$  and so for  $|z| \leq 1 + \lambda$

$$\beta(|z|) \leq \beta(1 + \lambda) \leq \beta(1)(1 + \lambda)^T \leq e^{1/K}.$$

Using  $|1/(1 - \beta(z))| \geq 1/(1 + \beta(|z|))$  we obtain

$$|R_T(z)| \geq \frac{1}{1 + \beta(|z|)} - |Q(z)| \geq \frac{1}{1 + e^{1/K}} - |Q(z)|. \tag{70}$$

We now prove that  $|Q(z)| = o(1)$  for  $|z| \leq 1 + \lambda$  and the lemma will follow.

Turning our attention first to  $Q_1(z)$ , we have

$$|Q_1(z)| \leq (1 + \lambda)^T |Q_1(1)| \leq e^{2/K} \sum_{t=0}^T |r_t - \alpha_t| \tag{71}$$

From (62), (63) of the proof of Lemma 8, we see that  $\sum_{t=0}^T |r_t - \alpha_t| = o(1)$ , hence  $|Q_1(z)| = o(1)$ .

We now consider  $Q_2(z)$ . As in Lemma 8, let  $r_t^*$  be the probability that a walk  $\mathcal{W}_v^*$  on  $G_v$  starting at  $v$  has not been absorbed at  $\Gamma_v^\circ$  by step  $t$ . Then  $\alpha_t \leq r_t^* \leq \sigma_t$ , so

$$|Q_2(z)| \leq \sum_{t=T+1}^{\infty} \sigma_t |z|^t,$$

In the case where  $G_v$  is a tree we can use (66) to prove that the radius of convergence of  $Q_2(z)$  is at least  $e^{1/(3c\omega^2)} > 1 + 1/(3c\omega^2) > 1 + 2\lambda$ , where  $\omega = \log \log \log n$  is given in (10), and  $\lambda = O(1/\log n)$ . So for  $|z| \leq 1 + \lambda$ ,

$$|Q_2(z)| \leq \sum_{t=T+1}^{\infty} e^{\lambda t - t/(3c\omega^2)} = o(1).$$

In the case that  $G_v$  contains a unique cycle, we can use (67) to see that the radius of convergence of  $Q_2(z)$  is at least  $e^{\frac{1}{20c\omega^3}} > 1 + 2\lambda$ . So for  $|z| \leq 1 + \lambda$ ,

$$|Q_2(z)| \leq \sum_{t=T+1}^{\infty} e^{\lambda t - t/(20c\omega^3)} = o(1).$$

□