The cover time of sparse random graphs.

Colin Cooper * Alan Frieze[†]

January 8, 2003

Abstract

We study the cover time of a random walk on graphs $G \in G_{n,p}$ when $p = \frac{c \log n}{n}, c > 1$. We prove that **whp** the cover time is asymptotic to $c \log \left(\frac{c}{c-1}\right) n \log n$.

1 Introduction

Let G = (V, E) be a connected graph, let |V| = n, and |E| = m. For $v \in V$ let C_v be the expected time taken for a simple random walk W on G starting at v, to visit every vertex of G. The cover time C_G of G is defined as $C_G = \max_{v \in V} C_v$. The cover time of connected graphs has been extensively studied. It is a classic result of Aleliunas, Karp, Lipton, Lovász and Rackoff [1] that $C_G \leq 2m(n-1)$. It is also known (see Feige [6], [7]), that for any connected graph G

$$(1 - o(1))n \log n \le C_G \le (1 + o(1))\frac{4}{27}n^3.$$

In this paper we study the cover time of the random graph, $G \in G_{n,p}$. It was shown by Jonasson [10] that **whp**

- (a) $C_G = (1 + o(1))n \log n \text{ if } \frac{np}{\log n} \to \infty.$
- (b) If c > 1 is constant and $np = c \log n$ then $C_G > (1+\alpha)n \log n$ for some constant $\alpha = \alpha(c)$.

Thus Jonasson has shown that when the expected average degree (n-1)p grows faster than $\log n$, a random graph has the same cover time **whp** as the complete graph K_n , whose cover

^{*}Department of Mathematical and Computing Sciences, Goldsmiths College, London SW14 6NW, UK

[†]Department of Mathematical Sciences, Carnegie Mellon University, Pittsburgh PA15213. Supported in part by NSF grant CCR-9818411.

time is determined by the Coupon Collector problem. Whereas, when $np = \Omega(\log n)$ this is not the case.

In this paper we sharpen Jonasson's results for the case $np = c \log n$ where $\omega = (c-1) \log n \to \infty$. This condition on ω ensures that **whp** $G_{n,p}$ is connected, (see Erdős and Rényi [5]).

Theorem 1. Suppose that $np = c \log n = \log n + \omega$ where $\omega = (c-1) \log n \to \infty$ and c = O(1). If $G \in G_{n,p}$, then **whp**

$$C_G \sim c \log \left(\frac{c}{c-1}\right) n \log n.$$

In the next section we give some properties that hold **whp** in $G_{n,p}$. In Section 3 we show that a graph with these properties has a cover time described by Theorem 1.

2 Properties of $G_{n,p}$

Let δ , Δ denote the minimum and maximum degree, and let d(u, v) denote the distance between the vertices u, v of the graph G.

Let $np = c \log n$ where c > 1. Whp $G \in G_{n,p}$ has the structural properties **P0–P7** given below. We say that a graph G with these properties is typical. The proof of the following lemma is given in the Appendix.

Lemma 1. Let $p = \frac{c \log n}{n}$ where $\omega = (c-1) \log n \to \infty$ and c = O(1). Then whp $G \in G_{n,p}$ is typical.

P0: G is connected.

P1: $\Delta(G) \leq \Delta_0 = (c+10) \log n$ and

$$\delta(G) \ge \begin{cases} 1 & c \le 1 + e^{-500} \\ \alpha \log n & c > 1 + e^{-500} \end{cases}$$

where $\alpha = \alpha^*/2$ and $\alpha^* > e^{-600}$ satisfies $c-1 = \alpha^* \log(ce/\alpha^*)$.

- **P2:** There are at most $n^{1/3}$ small vertices (i.e of degree at most $\log n/20$) and no two small vertices are within distance $\leq \frac{\log n}{(\log \log n)^2}$ of each other.
- **P3:** For $L \subseteq V$, $|L| \le 4$, let H = G L. For $S \subseteq V L$ let $e_H(S, \overline{S})$ be the number of edges of H with one end in S and the other in $\overline{S} = V (L \cup S)$.

For all $H \subseteq G$ such that $\delta(H) \ge 1$, and for all $S \subseteq V - L$, $|S| \le n/2$,

$$\frac{e_H(S,\overline{S})}{d_H(S)} \ge \frac{1}{6}.$$

P4: Let $\overline{D}(k) = n\binom{n-1}{k}p^k(1-p)^{n-1-k}$ denote the expected size of D(k) in $G_{n,p}$. Let D(k) be the number of vertices of degree k in G. Define

$$K_{0} = \{k \in [1, \Delta_{0}] : \overline{D}(k) \leq (\log n)^{-2} \}.$$

$$K_{1} = \{1 \leq k \leq 15 : (\log n)^{-2} \leq \overline{D}(k) \leq \log \log n \}.$$

$$K_{2} = \{k \in [16, \Delta_{0}] : (\log n)^{-2} \leq \overline{D}(k) \leq (\log n)^{2} \}.$$

$$K_{3} = [1, \Delta_{0}] \setminus (K_{0} \cup K_{1} \cup K_{2}).$$

P4a: If $k \in K_3$ then $\frac{1}{2}\overline{D}(k) \leq D(k) \leq 2\overline{D}(k)$, and

$$D(k) \begin{cases} = 0 & k \in K_0 \\ \leq (\log \log n)^2 & k \in K_1 \\ \leq (\log n)^4 & k \in K_2 \end{cases}$$

P4b: If $\omega \geq (\log n)^{2/3}$ then $K_1 = \emptyset$ and

$$\min\{k \in K_2\} \ge (\log n)^{1/2} \quad \text{ and } \quad |K_2| = O(\log \log n).$$

P5: The number of edges m = m(G) of G satisfies $\left| m - \frac{1}{2}cn \log n \right| \leq n^{1/2} \log n$.

P6: Let $k^* = \lceil (c-1) \log n \rceil$, $V^* = \{v : d(v) = k^*\}$ and let $B^* = \{v \in V^* : dist(v, w) \le \frac{10 \log n}{(\log \log n)^2} \text{ for some } w \in V^*, w \ne v\}$. Then

$$|V^*| \ge \frac{1}{2}\overline{D}(k^*)$$
 and $|B^*| \le \frac{1}{10}\overline{D}(k^*)$.

Let $X = \{v : \delta_v \leq \alpha \log n\}$ where $\delta_v \geq 2$ is the minimum degree of a neighbour of v, excluding neighbours of degree one. Then

$$|V^* \cap X| \le \frac{1}{10}\overline{D}(k^*).$$

P7 The minimum distance between two *small* cycles of length $\leq \frac{\log n}{10 \log \log n}$ is at least $\frac{\log n}{\log \log n}$ and the minimum distance between a small vertex and a small cycle is at least $\frac{\log n}{10 \log \log n}$.

3 The cover time of a typical graph

In this section G denotes a fixed graph with vertex set [n] which satisfies $\mathbf{P0}$ - $\mathbf{P7}$ and u is some arbitrary vertex from which a walk is started. For a subgraph H of G let $W_{u,H}$ denote a random walk on H which starts at vertex u and let $W_{u,H}(t)$ denote the walk generated by the first t steps. Let $X_{u,H}(t)$ be the vertex reached at step t and let $P_{u,H}^{(t)}(v) = \mathbf{Pr}(X_{u,H}(t) = v)$. Let $\pi_{u,H}(v)$ be the steady state probability of the random walk $W_{u,H}$. For an unbiased random

walk on a connected graph H with m(H) edges, $\pi_H(v) = \pi_{u,H}(v) = \frac{d_H(v)}{2m(H)}$ where $d_H(v)$ denotes degree in H.

Our definition of typical does not rule out G being bipartite, even though $G_{n,p}$ is non-bipartite **whp** for these values of p. In which case there is no steady state distribution. We therefore assume that in such a case, at each step, the random walk does nothing with probability 1/2 and only moves to an adjacent vertex with probability 1/2. We double the expected time to cover the vertices, but the asymptotic number of non-trivial steps remains the same.

Let $H(v) = G - \{v\}$ if v is not a neighbour of a vertex w of degree 1, and let $H(v) = G - \{v, w\}$ if v has a neighbour w of degree 1. (Note that **P2** rules out a neighbour having two neighbours of degree 1). For a subgraph H let $N_H(v)$ be the neighbourhood of v in H (i.e $N_H(v) = N_G(v) \cap V(H)$). When H = G we drop the H from the above notation and often drop the u as well.

Lemma 2. Let G be typical, then there exists a sufficiently large constant K > 0 such that if $\tau_0 = K \log n$ then for all $v \in V$, and for all $u, x \in H = H(v)$, after $t \ge \tau_0$ steps

$$|P_{u,H}^{(t)}(x) - \pi_{u,H}(x)| = O(n^{-10}). \tag{1}$$

Proof The conductance Φ of the walk $W_{u,H}$ is defined by

$$\Phi(W_{u,H}) = \min_{\pi(S) \le 1/2} \frac{e_H(S : \overline{S})}{d_H(S)}.$$

It follows from **P3** that the conductance Φ of the walk $W_{u,H}$ satisfies $\Phi \geq \frac{1}{6}$. Now it follows from Jerrum and Sinclair [9] that

$$|P_{u,H}^{(t)}(x) - \pi_{u,H}(x)| = O\left(n^{1/2} \left(1 - \frac{\Phi^2}{2}\right)^t\right). \tag{2}$$

For sufficiently large K, the RHS above will be $O(n^{-10})$ at τ_0 . We remark that there is a technical point here. The result of [9] assumes that the walk is lazy, and only makes a move to a neighbour with probability 1/2 at any step. This halves the conductance but still (2) remains true. For us it is sufficient simply to keep the walk lazy for $2\tau_0$ steps until it is mixed. This is negligible compared to the cover time.

For $v \neq u \in V$, let $\mathcal{A}_t(v)$ be the event that $W_{u,G}(t)$ does not visit v.

Lemma 3.

(a) If $t > 2\tau_0$ and $\delta_v \geq 2$ then

$$\mathbf{Pr}(\mathcal{A}_{t}(v)) \leq \left(1 - \left(\left(\frac{\delta_{v} - 1}{\delta_{v}}\right)^{2} - O\left(\frac{1}{\log n}\right)\right) \frac{d(v)}{2m}\right)^{t - 2\tau_{0}} \mathbf{Pr}(\mathcal{A}_{2\tau_{0}}(v))$$

$$\mathbf{Pr}(\mathcal{A}_{t}(v)) \geq \left(1 - \left(\left(\frac{\delta_{v}}{\delta_{v} - 1}\right)^{2} + O\left(\frac{1}{\log n}\right)\right) \frac{d(v)}{2m}\right)^{t - 2\tau_{0}} \mathbf{Pr}(\mathcal{A}_{2\tau_{0}}(v))$$

(b) Suppose that $v, v' \in V^* \setminus X$ (see P6) and that $dist(v, v') > \frac{10 \log n}{(\log \log n)^2}$. Then

$$\mathbf{Pr}(\mathcal{A}_{2\tau_0}(v)\cap \mathcal{A}_{2\tau_0}(v')) \leq \left(1 - \left(1 + O\left(\frac{1}{\log n}\right)\right)\frac{k^*}{m}\right)^{t-2\tau_0}.$$

Proof (a) Fix $w \neq v$ and $y \in N_H(v)$. Let $\mathcal{W}_k(y)$ denote the set of walks in H(v) which start at w, finish at y, are of length $2\tau_0$ and which leave a vertex in the neighbourhood $N_H(v)$ exactly k times. (Note that the walk can leave $y \in N_H(v)$ without necessarily leaving $N_H(v)$). Let $\mathcal{W}_k = \bigcup_y \mathcal{W}_k(y)$ and let $W = (w_0, w_1, \ldots, w_{2\tau_0}) \in \mathcal{W}_k(y)$. Let

$$\rho_W = \frac{\mathbf{Pr}(X_{w,G}(s) = w_s, s = 0, 1, \dots, 2\tau_0)}{\mathbf{Pr}(X_{w,H}(s) = w_s, s = 0, 1, \dots, 2\tau_0)}.$$
(3)

Then

$$1 \ge \rho_W \ge \left(\frac{\delta_v - 1}{\delta_v}\right)^k.$$

This is because

$$\frac{\mathbf{Pr}(X_{w,H}(s) = w_s \mid X_{w,H}(s-1) = w_{s-1})}{\mathbf{Pr}(X_{w,G}(s) = w_s \mid X_{w,G}(s-1) = w_{s-1})} = \begin{cases} 1 & w_{s-1} \notin N_G(v) \\ \frac{d_G(w_{s-1})}{d_G(w_{s-1}) - 1} & w_{s-1} \in N_G(v) \end{cases}$$

If $\mathcal{E} = \{X_{w,G}(\tau) \neq v, 0 \leq \tau \leq 2\tau_0\}$ then

$$\mathbf{Pr}(\mathcal{E}) = \sum_{k\geq 0} \sum_{W\in\mathcal{W}_k} \mathbf{Pr}(W_{w,G}(2\tau_0) = W)$$

$$= \sum_{k\geq 0} \sum_{W\in\mathcal{W}_k} \rho_W \mathbf{Pr}(W_{w,H}(2\tau_0) = W)$$

$$\geq \sum_{k\geq 0} p_k \left(\frac{\delta_v - 1}{\delta_v}\right)^k$$

where

$$p_k = \sum_{W \in \mathcal{W}_k} \mathbf{Pr}(W_{w,H}(2\tau_0) = W) = \mathbf{Pr}(W_{w,H}(2\tau_0) \in \mathcal{W}_k).$$

We will show later that

$$p_0 + p_1 + p_2 \ge 1 - O((\log n)^{-1})$$
 (4)

which immediately implies that

$$\mathbf{Pr}(\mathcal{E}) \ge p_0 + p_1 \left(1 - \frac{1}{\delta_v} \right) + p_2 \left(1 - \frac{1}{\delta_v} \right)^2 \ge \left(1 - \frac{1}{\delta_v} \right)^2 - O((\log n)^{-1}).$$

Now fix y and write

$$\mathbf{Pr}(X_{w,G}(2\tau_0) = y \mid \mathcal{E}) = \sum_{k \geq 0} \sum_{W \in \mathcal{W}_k(y)} \mathbf{Pr}(W_{w,G}(2\tau_0) = W) \mathbf{Pr}(\mathcal{E})^{-1}$$
$$= \sum_{k \geq 0} \sum_{W \in \mathcal{W}_k(y)} \rho_W \mathbf{Pr}(W_{w,H}(2\tau_0) = W) \mathbf{Pr}(\mathcal{E})^{-1}.$$

Now if

$$p_{k,y} = \frac{\mathbf{Pr}(W_{w,H} \in \mathcal{W}_k(y))}{\mathbf{Pr}(X_{w,H}(2\tau_0) = y)}$$

$$= \mathbf{Pr}(W_{w,H}(2\tau_0) \text{ leaves a vertex of } N_H(v) \text{ } k \text{ times } | X_{w,H}(2\tau_0) = y)$$

then

$$\sum_{k>0} p_{k,y} \left(\frac{\delta_v - 1}{\delta_v} \right)^k \le \frac{\mathbf{Pr}(X_{w,G}(2\tau_0) = y \mid \mathcal{E})}{\mathbf{Pr}(X_{w,H}(2\tau_0) = y)} \le \mathbf{Pr}(\mathcal{E})^{-1}.$$

We will show later that

$$p_{0,y} + p_{1,y} + p_{2,y} \ge 1 - O((\log n)^{-1}) \tag{5}$$

and so

$$\left(\frac{\delta_v - 1}{\delta_v}\right)^2 - O\left(\frac{1}{\log n}\right) \le \left|\frac{\mathbf{Pr}(X_{w,G}(2\tau_0) = y \mid \mathcal{E})}{\mathbf{Pr}(X_{w,H}(2\tau_0) = y)}\right| \le \left(\frac{\delta_v}{\delta_v - 1}\right)^2 + O\left(\frac{1}{\log n}\right).$$

Taking w as $X_{u,G}(t-2\tau_0-1)$, and conditioning on $\mathcal{A}_{t-2\tau_0-1}(v)$, we deduce that

$$\left(\frac{\delta_v - 1}{\delta_v}\right)^2 - O\left(\frac{1}{\log n}\right) \le \left|\frac{\mathbf{Pr}(X_{u,G}(t-1) = y \mid \mathcal{A}_{t-1}(v))}{\mathbf{Pr}(X_{w,H}(2\tau_0) = y)}\right| \le \left(\frac{\delta_v}{\delta_v - 1}\right)^2 + O\left(\frac{1}{\log n}\right).$$

Therefore

$$\mathbf{Pr}(\mathcal{A}_{t}(v) \mid \mathcal{A}_{t-1}(v)) \geq 1 - \left(\left(\frac{\delta_{v}}{\delta_{v}-1}\right)^{2} + O\left(\frac{1}{\log n}\right)\right) \sum_{y \in N_{H}(v)} P_{w,H(v)}^{(2\tau_{0})}(y) \frac{1}{d(y)}$$

$$= 1 - \left(\left(\frac{\delta_{v}}{\delta_{v}-1}\right)^{2} + O\left(\frac{1}{\log n}\right)\right) \sum_{y \in N_{H}(v)} \left(\frac{d(y)-1}{2m-2d(v)} + O\left(\frac{1}{n^{10}}\right)\right) \frac{1}{d(y)}$$

$$\geq 1 - \left(\left(\frac{\delta_{v}}{\delta_{v}-1}\right)^{2} + O\left(\frac{1}{\log n}\right)\right) \left(\frac{d(v)}{2m} - \frac{1}{2m-2d(v)} \sum_{y \in N_{H}(v)} \frac{1}{d(y)}\right)$$

$$= 1 - \left(\left(\frac{\delta_{v}}{\delta_{v}-1}\right)^{2} + O\left(\frac{1}{\log n}\right)\right) \frac{d(v)}{2m}.$$

Here we use **P2** to see that $\sum_{y \in N_H(v)} \frac{1}{d(y)} \leq \frac{40 d(v)}{\log n}$

Similarly,

$$\mathbf{Pr}(\mathcal{A}_t(v) \mid \mathcal{A}_{t-1}(v)) \le 1 - \left(\left(\frac{\delta_v - 1}{\delta_v} \right)^2 - O\left(\frac{1}{\log n} \right) \right) \frac{d(v)}{2m}$$

and the lemma follows immediately.

Proof of (4,5). Clearly, we only need to prove (5) and so fix $y \in N_H(v)$.

Let W(a, b, t) denote the set of walks in H from a to b of length t and for $W \in W(a, b, t)$ let $\mathbf{Pr}(W) = \mathbf{Pr}(W_{a,H}(t) = W)$. Then for $x \in V(H)$ we have

$$\mathbf{Pr}(X_{w,H}(\tau_0) = x \mid X_{w,H}(2\tau_0) = y) = \sum_{\substack{W_1 \in \mathcal{W}(w,x,\tau_0) \\ W_2 \in \mathcal{W}(x,y,\tau_0)}} \frac{\mathbf{Pr}(W_1)\mathbf{Pr}(W_2)}{\mathbf{Pr}(\mathcal{W}(w,y,2\tau_0))}$$

$$= \pi_{x,H}^{-1} \sum_{\substack{W_1 \in \mathcal{W}(w,x,\tau_0) \\ W_2 \in \mathcal{W}(x,y,\tau_0)}} \frac{\mathbf{Pr}(W_1)\pi_{x,H}\mathbf{Pr}(W_2)}{\mathbf{Pr}(\mathcal{W}(w,y,2\tau_0))}$$

and with W_3 equal to the reversal of W_2 ,

$$= \pi_{x,H}^{-1} \pi_{y,H} \sum_{\substack{W_1 \in \mathcal{W}(w,x,\tau_0) \\ W_3 \in \mathcal{W}(y,x,\tau_0)}} \frac{\mathbf{Pr}(W_1)\mathbf{Pr}(W_3)}{\mathbf{Pr}(\mathcal{W}(w,y,2\tau_0))}$$

$$= \frac{\pi_{x,H}^{-1} \pi_{y,H}}{\mathbf{Pr}(\mathcal{W}(w,y,2\tau_0))} \mathbf{Pr}(\mathcal{W}(w,x,\tau_0)) \mathbf{Pr}(\mathcal{W}(y,x,\tau_0))$$

$$= \frac{\pi_{x,H}^{-1} \pi_{y,H}}{\mathbf{Pr}(\mathcal{W}(w,y,2\tau_0))} (\pi_{x,H} - O(n^{-10}))^2$$

$$= \pi_{x,H} - O(n^{-9} \log n).$$

It follows that the variation distance between $X_{w,H}(\tau_0)$ and a vertex chosen from the steady state distribution π_H is $O(n^{-8}\log n)$. Now given $x=X_{w,H}(\tau_0),\ W_{w,H}(\tau_0)$ is a random walk of length τ_0 from w to x and $W_2=(x=X_{w,H}(\tau_0),X_{w,H}(\tau_0+1),\ldots,y=X_{w,H}(2\tau_0))$ is a random walk of length τ_0 from x to y. For $W\in\bigcup_{\xi}\mathcal{W}(\xi,y,\tau_0)$ let $\mathbf{Q}(W)$ be the probability that $(y,X_{w,H}(2\tau_0-1),\ldots,X_{w,H}(\tau_0))=W$. Then we have

$$\mathbf{Q}(W) = (1 + O(n^{-8}\log n)) \frac{\pi_{x,H} \mathbf{Pr}(W^{reverse})}{\mathbf{Pr}(\mathcal{W}(x, y, \tau_0))}$$

$$= (1 + O(n^{-8}\log n)) \frac{\pi_{y,H} \mathbf{Pr}(W)}{\mathbf{Pr}(\mathcal{W}(x, y, \tau_0))}$$

$$= (1 + O(n^{-8}\log n)) \frac{\pi_{y,H} \pi_{x,H} \mathbf{Pr}(W)}{\pi_{x,H} \mathbf{Pr}(\mathcal{W}(x, y, \tau_0))}$$

$$= (1 + O(n^{-8}\log n)) \frac{\pi_{y,H} \pi_{x,H} \mathbf{Pr}(W)}{\pi_{y,H} \pi_{x,H} \mathbf{Pr}(W)}$$

Thus if $W = (w_1, w_2, ..., w_{\tau_0})$ then

$$\mathbf{Q}(W \mid X_{w,H}(\tau_0) = w_1) = (1 + O(n^{-8}\log n)) \frac{\mathbf{Pr}(W)}{\mathbf{Pr}(W(y, x, \tau_0))}$$

and so the distribution of $W_2^{reverse}$ is within variation distance $O(n^{-8} \log n)$ of that of a random walk of length τ_0 from y to a vertex x chosen with distribution π_H .

Thus the distribution of a random walk of length $2\tau_0$ from w to y and that of $W_1, W_3^{reversed}$ is $O(n^{-8} \log n)$ where W_1, W_3 are obtained by (i) choosing x from the steady state distribution and then (ii) choosing a random walk W_1 from w to x and a random walk W_3 from y to x. Furthermore, the variation distance between the distribution of W_1 and a random walk of length τ_0 from w is $O(n^{-9})$. Similarly, the variation distance between distribution of W_3 and a random walk of length τ_0 from y is $O(n^{-9})$.

Now consider W_1 and let Z_t be the distance of $X_{w,H}(t)$ from v. We observe from **P2** and **P7** that except for at most one value $\bar{a} \in J = [1, \frac{\log n}{2(\log \log n)^2}]$ we have

$$\mathbf{Pr}(Z_{t+1} = a + 1 \mid Z_t = a) \ge 1 - \frac{20}{\log n}, \qquad a \in I \setminus \bar{a}.$$

and this will enable us to prove

$$\mathbf{Pr}(W_1 \text{ or } W_3 \text{ make a } return \text{ to } N_H(v)) = O(1/\log n) \tag{6}$$

and this implies (5). (Note that a move from $N_H(v)$ to $N_H(v)$ has to be counted as a return here.)

To prove (6), let t_0 be the first time that W_1 visits $N_H(v)$. We have to estimate the probability that W_1 returns to $N_H(v)$ later on and so we can assume w.l.o.g. that $w \in N_H(v)$ i.e. $Z_0 = 1$.

It follows from P2 and P7 that

$$\mathbf{Pr}(Z_i = i + 1, i = 1, \dots, 6 \mid Z_0 = 1) \ge \left(1 - \frac{40}{\log n}\right)^6.$$
 (7)

To check this consider two possiblilites:

- (a) There is no small vertex in the ≤ 7 neighbourhood N_7 of v. Since there is at most one edge joining two vertices in N_7 , we see that $\mathbf{Pr}(Z_{i+1} > Z_i) = 1 \frac{40}{\log n}$ for $i = 1, \ldots, 6$ and (7) follows.
- (b) On the other hand, if there is a small vertex x in N_7 then with probability $\geq 1 \frac{20}{\log n}$ the first move from w takes us further away from x and (7) follows as before.

If $Z_3=4$ and there is a return to $N_H(v)$ then there exists $\tau \leq \tau_0$ such that $Z_{\tau}=4, Z_{\tau+1}=3$ and $Z_{\tau+2}\leq 3$. If there is no small vertex within distance 4 of v then **P2** and **P7** imply

$$\mathbf{Pr}(\exists \tau \le \tau_0: \ Z_{\tau} = 4, Z_{\tau+1} = 3, Z_{\tau+2} \le 3) = O\left(\frac{\tau_0}{(\log n)^2}\right). \tag{8}$$

If there is a unique small vertex within distance 4 of v and $Z_6 = 7$ and there is a return to $N_H(v)$ then there exists $\tau \leq \tau_0$ such that $Z_\tau = 7, Z_{\tau+1} = 6$ and $Z_{\tau+2} = 5$ (no small cycles close to v now). We can then argue as in (8) that the probability of this $O\left(\frac{\tau_0}{(\log n)^2}\right)$. This completes the proof of part (a) of the lemma.

(b) We simply run through the proof as in (a), replacing v by v, v': $H = H(v, v') = G - \{v, v'\}$, $N_H(v, v') = N_G(v) \cup N_G(v')$. The proof of (5) remains valid because v, v' are far apart. \square

3.1 The upper bound on cover time

From here on, A_1, A_2, \ldots are a sequence of unspecified positive constants.

Let $t_0 = \lceil 2m \log \frac{c}{c-1} \rceil$. We now prove for typical graphs, that for any vertex $u \in V$

$$C_u \le t_0 + o(m). \tag{9}$$

Let $T_G(u)$ be the time taken to visit every vertex of G by the random walk W_u . Let U_t be the number of vertices of G which have not been visited by W_u at step t. We note the following:

$$\mathbf{Pr}(T_G(u) > t) = \mathbf{Pr}(U_t > 0) \le \min\{1, \mathbf{E} U_t\},\tag{10}$$

$$C_u = \mathbf{E} T_G(u) = \sum_{t>0} \mathbf{Pr}(T_G(u) > t)$$
(11)

It follows from (10,11) that for all t

$$C_u \le t + \sum_{s>t} \mathbf{E} \ U_s = t + \sum_{v \in V} \sum_{s>t} \mathbf{Pr}(\mathcal{A}_s(v)).$$
 (12)

Now, by Lemma 3, for $s > 2\tau_0$,

$$\begin{aligned} \mathbf{Pr}(\mathcal{A}_{s}(v)) & \leq \left(1 - \left(\left(\frac{\delta_{v} - 1}{\delta_{v}}\right)^{2} - \frac{A_{1}}{\log n}\right) \frac{d(v)}{2m}\right)^{s - 2\tau_{0}} \mathbf{Pr}(\mathcal{A}_{2\tau_{0}}(v)) \\ & \leq \exp\left(-\frac{sd(v)}{2m}\left(1 - \frac{A_{2}}{\log n}\right)\right), & \text{if } \delta_{v} \geq \alpha \log n \end{aligned}$$

where α is as in **P1**.

Then from **P4**,

$$\mathbf{E} \ U_s \le T_3(s) + T_1(s) + T_2(s) + T_X(s) \tag{13}$$

where

$$T_3(s) = 2\sum_{k=1}^{n-1} n \binom{n-1}{k} p^k (1-p)^{n-1-k} e^{-\frac{sk}{2m} \left(1 - \frac{A_2}{\log n}\right)},$$

$$T_i(s) = \sum_{k \in K_i} D(k) e^{-\frac{sk}{2m} \left(1 - \frac{A_2}{\log n}\right)}, \qquad i = 1, 2$$

and

$$T_X(s) = \sum_{v \in X} \left(1 - \left(\left(\frac{\delta_v - 1}{\delta_v} \right)^2 - O\left(\frac{1}{\log n} \right) \right) \frac{d(v)}{2m} \right)^{s - 2\tau_0}$$

$$\leq 2 \sum_{v \in X} \exp\left\{ - \left(\left(\frac{\delta_v - 1}{\delta_v} \right)^2 - \frac{A_3}{\log n} \right) \frac{sd(v)}{2m} \right\}.$$

Now for $\gamma > 0$,

$$\sum_{s=t_0+1}^{\infty} e^{-\gamma s} \le \gamma^{-1} e^{-\gamma t_0}. \tag{14}$$

Let $\lambda = \frac{t_0}{2m} \left(1 - \frac{A_2}{\log n} \right)$. Applying (14) we get

$$\sum_{s=t_{0}+1}^{\infty} T_{3}(s) \leq 3m \sum_{k=1}^{n-1} \frac{n}{k} \binom{n-1}{k} p^{k} (1-p)^{n-k-1} e^{-k\lambda}$$

$$\leq 6 \frac{m}{p} e^{\lambda} \sum_{k=1}^{n-1} \binom{n}{k+1} p^{k+1} (1-p)^{n-k-1} e^{-(k+1)\lambda}$$

$$< 7 \frac{m}{p} \frac{c}{c-1} \left(1-p+p e^{-\lambda}\right)^{n}$$

$$\leq 7 \frac{mn}{(c-1)\log n} e^{-np+np e^{-\lambda}}$$

$$\leq 8 \frac{m e^{2A_{2}}}{(c-1)\log n}$$

$$= o(m). \tag{15}$$

We have used the estimation,

$$npe^{-\lambda} \le (c \log n) \left(\frac{c-1}{c}\right) \left(1 + \frac{1}{c-1}\right)^{A_2/\log n}$$

 $\le (1 + O(n^{-1}))((c-1)\log n) \left(1 + \frac{2A_2}{(c-1)\log n}\right).$

Note that we have used $(c-1)\log n \to \infty$ to get the second line.

Continuing we get

$$\sum_{s=t_0+1}^{\infty} T_1(s) \leq A_4 m \sum_{k \in K_1} \frac{(\log \log n)^2}{k} e^{-k\lambda}$$

$$= o(m)$$
(16)

since either (i) $\omega \geq (\log n)^{2/3}$ and $K_1 = \emptyset$ or (ii) $\omega < (\log n)^{2/3}$ and $e^{\lambda} \geq (1 - o(1))(\log n)^{1/3}$.

$$\sum_{s=t_0+1}^{\infty} T_2(s) \leq A_5 m \sum_{k \in K_2} \frac{(\log n)^4}{k} e^{-k\lambda}$$

$$= o(m)$$
(17)

since either (i) $\omega \ge (\log n)^{2/3}$ and $\min\{k \in K_2\} \ge (\log n)^{1/2}$ and $|K_2| = O(\log \log n)$ or (ii) $\omega < (\log n)^{2/3}$ and $e^{\lambda} \ge (1 - o(1))(\log n)^{1/3}$.

Note now that $\delta_v \geq 2$ and if $v \in X$ (see **P6**) then from **P2** $d(v) \geq \log n/20$. Thus

$$\sum_{s=t_0+1}^{\infty} T_X(s) \leq \sum_{s=t_0+1}^{\infty} \sum_{v \in X} \exp\left\{-\frac{sd(v)}{10m}\right\}$$

$$\leq \sum_{v \in X} \frac{10m}{d(v)} \exp\left\{-\frac{t_0 d(v)}{10m}\right\}$$

$$\leq \sum_{v \in X} \frac{200m}{\log n} \exp\left\{-\frac{t_0 \log n}{200m}\right\} \qquad \text{by } \mathbf{P2}$$

$$\leq \sum_{v \in X} \frac{200m}{\log n} \left(\frac{c-1}{c}\right)^{\log n/201}$$

$$= o(m) \qquad (18)$$

since either (i) $c \ge 1 + e^{-500}$ and $X = \emptyset$ or (ii) $c < 1 + e^{-500}$, in which case we use $(c-1)/c \le e^{-500}$.

As $C_G = \max_{u \in V} C_u$, the upper bound on C_G now follows from (9), (13), (15), (16), (17), (18) and (12) with $t = t_0$.

3.2 The lower bound on cover time

For any vertex u, we can find a set of vertices S such that at time $t_1 = t_0(1 - \epsilon)$, $\epsilon \to 0$, the probability the set S is covered by the walk W_u tends to zero. Hence $T_G(u) > t_1$ whp which implies that $C_G \ge (1 - o(1))t_0$.

We construct S as follows. Let k^*, V^*, B^* be as defined in Property **P6**.

Let $S^* = V^* \setminus (B^* \cup X)$ and let

$$\epsilon = \frac{10}{(c-1)\log c/(c-1)} \frac{\log\log n}{\log n} = o(1) \text{ and } \delta = \frac{(\log n)^3}{|S^*|}.$$

Note that

$$\overline{D}(k^*) = \Omega\left(\frac{n^{(c-1)\ln(c/(c-1))}}{\sqrt{(c-1)\log n}}\right) = \Omega((\log n)^a)$$
(19)

for any constant a > 0. Then **P6** implies that $|S^*| = \Omega((\log n)^a)$ for any constant a > 0.

Now for $v, w \neq u$ let $\mathcal{A}_t(v, w)$ be the event that W has not visited v or w by step t. Let $Q \subseteq S^*$ be given by

$$Q = \{ v \in S^* : \mathbf{Pr}(\mathcal{A}_{2\tau_0}(v)) < 1 - \delta, \text{ or } \mathbf{Pr}(\mathcal{A}_{2\tau_0}(v, w)) < (1 - \delta)^2, \text{ for some } w \in S^* \}.$$

Now in time $2\tau_0$, W can visit at most $2\tau_0 + 1$ vertices and so

$$\sum_{v \in V} \mathbf{Pr}(\overline{A}_{2\tau_0}(v)) \le 2\tau_0 + 1 \text{ and } \sum_{v,w \in V} \mathbf{Pr}(\overline{A}_{2\tau_0}(v,w)) \le \binom{2\tau_0 + 1}{2}.$$

Thus

$$|Q| \le \frac{2\tau_0 + 1}{\delta} + \frac{2\tau_0(2\tau_0 + 1)}{2(1 - (1 - \delta)^2)} = o(|S^*|).$$

Therefore, if $S = S^* \setminus Q$,

$$|S| \ge \frac{\overline{D}(k^*)}{3}.$$

Let S(t) denote the subset of S which has not been visited by W by time t. Now

$$\mathbf{E} |S(t)| \ge \sum_{v \in S} \left(1 - \left(1 + \frac{A_6}{\log n} \right) \frac{k^*}{2m} \right)^{t - 2\tau_0} \mathbf{Pr} (\mathcal{A}_{2\tau_0}(v)).$$

Setting $t = t_1$ we have

$$\mathbf{E} |S(t_1)| = \Omega \left(\frac{n^{(c-1)\log c/(c-1)}}{\sqrt{(c-1)\log n}} \exp\left(-\frac{k^*}{2m}t_1\right) \right)$$

$$= \Omega \left(\frac{n^{\epsilon(c-1)\log c/(c-1)}}{\sqrt{(c-1)\log n}} \right)$$

$$= \Omega((\log n)^9). \tag{20}$$

Let $Y_{v,t}$ be the indicator for the event that $W_u(t)$ has not visited vertex v at time t. As $v, w \in S$ are not adjacent, and have no common neighbours, when we delete v, w the total degree of H(v, w) is 2m - 2d(v) - 2d(w), and $d(v) = d(w) = k^*$. It follows from Lemma 3(b) that for $v, w \in S$

$$\mathbf{E}\left(Y_{v,t_1}Y_{w,t_1}\right) \leq \left(1 - \left(1 + O\left(\frac{1}{\log n}\right)\right)\frac{k^*}{m}\right)^{t_1 - 2\tau_0}$$

$$\leq (1 + o(1))\mathbf{E}\left(Y_{v,t_1}\mathbf{E}\left(Y_{w,t_1}\right)\right). \tag{21}$$

It follows therefore that

$$\mathbf{Pr}(S(t_1) \neq 0) \geq \frac{(\mathbf{E} |S(t_1)|)^2}{\mathbf{E} |S(t_1)|^2} = \frac{1}{\frac{\mathbf{E}|S_{t_1}|(|S_{t_1}|-1)}{(\mathbf{E}|S(t_1)|)^2} + (\mathbf{E} |S_{t_1}|)^{-1}} = 1 - o(1)$$

from (20) and (21).

Acknowledgement We thank Johan Jonasson for pointing out a significant error in earlier draft.

References

- [1] R. Aleliunas, R.M. Karp, R.J. Lipton, L. Lovász and C. Rackoff, Random Walks, Universal Traversal Sequences, and the Complexity of Maze Problems. *Proceedings of the 20th Annual IEEE Symposium on Foundations of Computer Science* (1979) 218-223.
- [2] B.Bollobás, Random graphs (Second edition), Cambridge University Press (2001).
- [3] B.Bollobás, T.Fenner and A.M.Frieze, An algorithm for finding Hamilton paths and cycles in random graphs, *Combinatorica* 7 (1987) 327-341.
- [4] C. Cooper and A.M. Frieze, Crawling on web graphs, to appear in *Proceedings of STOC* 2002.
- [5] P. Erdős and A. Rényi, On random graphs I, Publ. Math. Debrecen 6 (1959) 290-297.
- [6] U. Feige, A tight upper bound for the cover time of random walks on graphs, Random Structures and Algorithms 6 (1995) 51-54.
- [7] U. Feige, A tight lower bound for the cover time of random walks on graphs, Random Structures and Algorithms 6 (1995) 433-438.
- [8] S.Janson, T.Łuczak and A.Ruciński, Random graphs, Wiley (2000).
- [9] M. Jerrum and A. Sinclair. The Markov chain Monte Carlo method: an approach to approximate counting and integration. In Approximation Algorithms for NP-hard Problems.
 (D. Hochbaum ed.) PWS (1996) 482-520
- [10] J. Jonasson, On the cover time of random walks on random graphs, Combinatorics, Probability and Computing, 7 (1998), 265-279.

4 Appendix: Typical graph properties

A proof of **P0**,**P1** can be found in Bollobás [2] or Janson, Łuczak and Ruciński [8]. A proof of **P2** can be found in [3].

P3: Case of $1 \le s = |S| \le n/(c \log n)$.

We first prove that **whp** $e_G(S, S) \leq s \log \log n$. Now

$$\mathbf{Pr}(\exists S : e_G(S, S) \ge s \log \log n)$$

$$\le \binom{n}{s} \binom{\binom{s}{2}}{s \log \log n} p^{s \log \log n} \le \left(\frac{ne}{s}\right)^s \left(\frac{spe}{2 \log \log n}\right)^{s \log \log n}$$

$$\le \exp\left(-s \left(\log \log n \cdot \log \left(\frac{2n \log \log n}{cse \log n}\right) - \log \frac{ne}{s}\right)\right)$$

$$= o(n^{-2}).$$

By property $\mathbf{P0}$ and the definition of H, both G and H contain no isolated vertices and hence d(S) > 0. We write $e(S, \overline{S})/d(S) = 1 - 2e(S, S)/d(S)$. Partition S into sets S_1 and S_2 , where S_1 are the vertices of S of degree at most $(\log n)/10$. Let T_1 be the neighbour set of S_1 in S_2 and let T_2 be the neighbour set of S_1 in \overline{S} . By property $\mathbf{P1}$ the set S_1 induces no edges, and the neighbours of vertices of S_1 are distinct. Thus

$$\frac{2e_H(S,S)}{d_H(S)} \leq \frac{2(|T_1| + |S_2| \log \log n)}{2|T_1| + |T_2| + |S_2|((\log n)/10 - |L|)} \\
\leq \frac{2 + \log \log n}{(\log n)/10} = o(1).$$

Now use

$$d_H(S) = e_H(S, \overline{S}) + 2e_H(S, S).$$

Case of $n/(c \log n) \le s \le n/2$.

The expected value of $e_H(S, \overline{S})$ is at least $\mu = s(n-s-4)p$. Thus from Chernoff bounds, for fixed s,

$$\mathbf{Pr}(\exists S : e_H(S, \overline{S}) \le \mu/2) \le \binom{n}{s} e^{-\frac{c}{9} \frac{s(n-s-4)}{n} \log n}$$

$$\le \exp\left(-s\left(\frac{c}{18}\log n - \log\frac{ne}{s}\right)\right)$$

$$= o(n^{-2}).$$

We note that $\mathbf{E} d_H(S) = 2\binom{s}{2}p + s(n-s-|L|)p$. Thus

$$\mathbf{Pr}(\exists S : d_H(S) \ge \frac{3}{2} \mathbf{E} \ d_H(S)) \le \binom{n}{s} e^{-\frac{c}{20} s \log n}$$

$$\le \exp\left(-s \left(\frac{c}{20} \log n - \log \frac{ne}{s}\right)\right)$$

$$= o(n^{-2}).$$

Thus

$$\frac{e_H(S, \overline{S})}{d_H(S)} \ge \frac{\frac{1}{2}s(n-s-4)p}{\frac{3}{2}(2\binom{s}{2}p + s(n-s)p)} \ge \frac{1}{6}.$$

P4a: First observe that

$$\mathbf{Pr}(\exists k \in K_0: \ D(k) > 0) \le \sum_{k \in K_0} \overline{D}(k) = \frac{|K_0|}{(\log n)^2} = O\left(\frac{1}{\log n}\right).$$

Then

$$\mathbf{Pr}(\exists k \in K_1: \ D(k) > (\log \log n)^2) \le \sum_{k \in K_1} \frac{\overline{D}(k)}{(\log \log n)^2} = O\left(\frac{1}{\log \log n}\right).$$

Similarly,

$$\mathbf{Pr}(\exists k \in K_2: \ D(k) > (\log n)^4) \le \sum_{k \in K_2} \frac{\overline{D}(k)}{(\log n)^4} = O\left(\frac{1}{\log n}\right).$$

A simple calculation gives that for our range of values of p

$$\mathbf{E}\left(D(k)(D(k)-1)\right) = \overline{D}(k)^2 \left(1 + O\left(\frac{\log n}{n}\right)\right).$$

Thus

$$\mathbf{Var}D(k) = \overline{D}(k) \left(1 + O\left(\frac{\overline{D}(k)\log n}{n}\right)\right).$$

Applying the Chebychef inequality we see that

$$\mathbf{Pr}(D(k) \le \frac{1}{2}\overline{D}(k) \text{ or } D(k) \ge 2\overline{D}(k)) \le \frac{4\mathbf{Var}D(k)}{\overline{D}(k)^2} = \frac{4}{\overline{D}(k)}\left(1 + O\left(\frac{\log n}{n}\right)\right).$$

So, as $|K_3| = O(\log n)$,

$$\mathbf{Pr}(\exists k \in K_3: \ D(k) \leq \frac{1}{2}\overline{D}(k) \text{ or } D(k) \geq 2\overline{D}(k)) = O\left(\frac{1}{\log n}\right).$$

P4b: The sequence $(\overline{D}(k), k \ge 0)$ is unimodal and

$$\frac{\overline{D}(k+1)}{\overline{D}(k)} \sim \frac{c \log n}{k+1} \quad \text{when } k = O(\log n).$$
 (22)

Moreover, for $k \leq \Delta_0$ there is a positive constant A = A(k) such that

$$\overline{D}(k) \sim An^{1-c} \left(\frac{ce \log n}{k}\right)^k \frac{1}{k^{1/2}}.$$
 (23)

Suppose first that there exists $k \in K_1 \cup K_2$ such that $k < (\log n)^{1/2}$. It follows from (23) that $c-1 < (\log n)^{-1/3}$, for if $c-1 \ge (\log n)^{-1/3}$ and $k < (\log n)^{1/2}$ then $\overline{D}(k) = o((\log n)^{-2})$.

Now suppose that $k \in K_2$ implies $k \ge (\log n)^{1/2}$. Observe from (23) that both $\overline{D}(\lfloor (c-c^{1/3})\log n \rfloor)$ and $\overline{D}(\lfloor (c+c^{1/3})\log n \rfloor)$ are much greater than $(\log n)^2$ Thus either $k \le n$

 $(c-c^{1/3})\log n$ or $k \ge (c+c^{1/3})\log n$. In either case, we see from iterating (22) that $|K_2| = O(\log \log n)$.

P5: This follows immediately from Chernoff bounds.

P6: From (19) we see that $\lceil (c-1)\log n \rceil \in K_3$ for c constant. That $|V^*| \geq \frac{1}{2}\overline{D}(k^*)$ now follows from **P3**. Now $|B^*| \leq |\{(v,w) \in (V^*)^2 : dist(v,w) \leq d = \frac{10\log n}{(\log\log n)^2}\}|$. Therefore

$$\mathbf{E} |B^*| \le \overline{D}(k^*)^2 \sum_{k=1}^d n^k p^{k+1} = o(\overline{D}(k^*)),$$

and the second part of P6 follows from the Markov inequality. The third part is a similar first moment calculation.

P7: A proof of similar results can be found in [3].