

## COVERING THE EDGES OF A RANDOM GRAPH BY CLIQUES

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## 1. Introduction

The *clique cover number*  $\theta_1(G)$  of a graph  $G$  is the minimum number of cliques required to cover the edges of graph  $G$ . In this paper we consider  $\theta_1(G_{n,p})$ , for  $p$  constant. (Recall that in the random graph  $G_{n,p}$ , each of the  $\binom{n}{2}$  edges occurs independently with probability  $p$ ). Bollobás, Erdős, Spencer and West [1] proved that **whp** (i.e. with probability  $1-o(1)$  as  $n \rightarrow \infty$ )

$$\frac{(1-o(1))n^2}{4(\log_2 n)^2} \leq \theta_1(G_{n,.5}) \leq \frac{cn^2 \ln \ln n}{(\ln n)^2}.$$

They implicitly conjecture that the  $\ln \ln n$  factor in the upper bound is unnecessary and in this paper we prove

**Theorem 1.** *There exist constants  $c_i = c_i(p) > 0, i = 1, 2$  such that **whp***

$$\frac{c_1 n^2}{(\ln n)^2} \leq \theta_1(G_{n,p}) \leq \frac{c_2 n^2}{(\ln n)^2}.$$

**Remark 1.** A simple use of a martingale tail inequality shows that  $\theta_1$  is close to its mean with very high probability.

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## 2. Proof of Theorem 1

We write  $a_n \approx b_n$  if  $a_n/b_n \rightarrow 1$  as  $n \rightarrow \infty$ .

The lower bound is simple as the number of edges  $m$  of  $G_{n,p}$  **whp** satisfies

$$m \approx \frac{np^2}{2}$$

and the size of the largest clique  $\omega = \omega(G_{n,p})$  **whp** satisfies

$$\omega \approx 2 \log_b n$$

where  $b = 1/p$ . We may thus choose  $c_1 \approx (\ln b)^2 p/2$ .

The upper bound requires more work. Our method does not seem to yield the correct value for  $c_2$  and so we will not work hard to keep  $c_2$  small. Let  $\alpha$  be some small constant and let

$$k = \lfloor \alpha \log_b n \rfloor.$$

We consider an algorithm for randomly selecting cliques to cover the edges of  $G = G_{n,p}$ . It bears some relation to part of the algorithm described in Pippenger and Spencer [2]. At iteration  $i$  we randomly select cliques of size  $k_i = \lfloor k/i \rfloor$  none of whose edges are covered by previously chosen cliques. Our idea is to choose these cliques so that at the start of iteration  $i$  the graph  $G_i$  formed by the set  $E_i$  of edges which have not been covered behaves, for our purposes, similarly to  $G_{n,p_i}$ ,  $p_i = pe^{1-i}$ . That is it will contain about  $m_i = \binom{n}{2} p_i$  edges, it will have about  $N_i = \binom{n}{k_i} p_i^{\binom{k_i}{2}}$  cliques of size  $k_i$  and the intersection of these cliques will be similar to that for the  $k_i$ -cliques in  $G_{n,p_i}$ . In particular, in both  $G_{n,p_i}$  and  $G_i$  almost all of the edges are in about  $\zeta_i = N_i \binom{k_i}{2} / m_i$   $k_i$ -cliques.

Now in iteration  $i$  we choose a set  $\mathcal{C}_i$  of  $k_i$ -cliques from  $G_i$  to add to our cover. The available cliques are chosen independently with probability about  $1/\zeta_i$ . By our assumptions on  $G_i$ , an edge is left uncovered with probability about  $e^{-1}$ . With a bit of care we can show that our assumptions continue to hold for  $G_{i+1}$  as well.

We do this for  $i_0 = \lceil 4 \ln \ln n \rceil$  iterations. After this there are about  $\binom{n}{2} pe(\ln n)^{-4}$  uncovered edges and we can add these as cliques of size two to the cover. In iteration  $i$  we choose about  $m_i / \binom{k_i}{2} \approx ni^2 pe^{1-i} k^{-2}$  cliques and so the total number of cliques used is  $O(n^2 / \ln n)$  as required.

We now need to describe our clique choosing process a little more formally: let  $\mathcal{C}_{t,i}$  denote the set of  $t$ -cliques all of whose edges are in  $E_i$ . If

$$c_{s,j,i} = \binom{n-s}{j-s} (be^i)^{\binom{s}{2} - \binom{j}{2}},$$

then  $c_{s,j,i}$  is close to the expected number of cliques in  $\mathcal{C}_{j,i}$  which contain a particular fixed clique in  $\mathcal{C}_{s,i}$ .

For a clique  $S \in \mathcal{C}_{s,i}$  we let

$$X_{S,j,i} = |\{C \in \mathcal{C}_{j,i} : C \supseteq S\}|$$

and for integer  $s \geq 0$ ,

$$X_{s,j,i}^* = \max\{X_{S,j,i} : S \in \mathcal{C}_{s,i}\}.$$

**Algorithm COVER**

**begin**

$E_1 := E(G_{n,p}); \mathcal{C}_{COVER} := \emptyset;$

**for**  $i = 1$  **to**  $i_0$  **do**

**begin**

A: independently place each  $C \in \mathcal{C}_{\lfloor k/i \rfloor, i}$  into  $\mathcal{C}_{COVER}$  with probability

$$X_{2, \lfloor k/i \rfloor, i}^{*-1};$$

B: for each  $u \in E_i$  which is not covered by a clique in Step A, add  $u$   
(as a clique of size 2) to  $\mathcal{C}_{COVER}$  with probability  $\varrho_u$  where

$$e^{-1} - X_2^{*-1} = \left(1 - \frac{1}{X_2^*}\right)^{X_u} (1 - \varrho_u),$$

$$X_2^* = X_{2, \lfloor k/i \rfloor, i}^* \text{ and } X_u = X_{u, \lfloor k/i \rfloor, i}.$$

**end**

$\mathcal{C}_{COVER} := \mathcal{C}_{COVER} \cup E_{i_0+1}.$

**end**

Observe first that the definition of  $\varrho_u$  assumes that  $X_2^*$  is large (which it is **whp**) and so

$$\begin{aligned} \left(1 - \frac{1}{X_2^*}\right)^{X_u} &\geq \left(1 - \frac{1}{X_2^*}\right)^{X_2^*} \\ &\geq e^{-1} - X_2^{*-1}, \end{aligned}$$

and  $\varrho_u$  is properly defined.

The following lemma contains the main core of the proof:

**Lemma 1.** *Let  $\mathcal{E}_i$  refer to the following two conditions:*

(a)  $X_{S,j,i} \leq (1 + \beta_i)c_{s,j,i}, \quad 0 \leq s \leq j \leq k/i \text{ and } S \in \mathcal{C}_{s,i},$

where  $\beta_i = in^{-1/4},$

(b)  $X_{u,j,i} \geq (1 - \beta_i)c_{2,j,i}, \quad u \in E_i \text{ and } 2 \leq j \leq k/i$

for all but at most  $in^{15/8}$  edges.

Then

$$(1) \quad \Pr(\mathcal{E}_1) = 1 - o(n^{-1}),$$

$$(2) \quad \Pr(\mathcal{E}_{i+1} \mid \mathcal{E}_i) \geq 1 - O(n^{-1/8}).$$

We defer the proof of the lemma to the next section and show how to use it to prove Theorem 1. Observe first that

$$(3) \quad \frac{c_{s+1,j,i}}{c_{s,j,i}} = \left( \frac{j-s}{n-s} \right) (be^i)^s,$$

and

$$(4) \quad c_{s,j,i} \geq n^{7/8}$$

when  $\alpha$  is small and  $0 \leq s < j \leq k/i$ .

Next let  $Y_i$  and  $Z_i$  denote the number of  $\lfloor k/i \rfloor$ -cliques and edges respectively added to  $\mathcal{C}_{COVER}$  in iteration  $i$ .

$$(5) \quad \mathbb{E}(Y_i \mid \mathcal{E}_i) = \mathbb{E} \left( \frac{X_{0,\lfloor k/i \rfloor,i}^*}{X_{2,\lfloor k/i \rfloor,i}^*} \mid \mathcal{E}_i \right) \leq (1 + o(1)) \frac{c_{0,\lfloor k/i \rfloor,i}}{c_{2,\lfloor k/i \rfloor,i}} \approx \frac{n^2 i^2}{bk^2 e^i},$$

on using (3).

Since  $Y_i$  is binomially distributed, we see using standard bounds on the tails of the binomial, that

$$\Pr \left( Y_i \geq \frac{2n^2 i^2}{bk^2 e^i} \mid \mathcal{E}_i \right) \leq n^{-1}.$$

Thus

$$\Pr \left( \sum_{i=1}^{i_0} Y_i \geq \sum_{i=1}^{i_0} \frac{2n^2 i^2}{bk^2 e^i} \mid \mathcal{E}_0 \right) = O \left( \frac{i_0}{n^{1/8}} \right),$$

and so

$$(6) \quad \Pr \left( \sum_{i=1}^{i_0} Y_i \geq \sum_{i=1}^{i_0} \frac{2n^2 i^2}{bk^2 e^i} \right) = o(1).$$

Now a simple calculation gives

$$(7) \quad \varrho_u = O \left( \frac{X_2^* - X_u}{X_2^*} \right)$$

and so

$$\mathbb{E}(Z_i \mid \mathcal{E}_i) = O(in^{15/8} + \beta_i |E_i|) = O(n^{15/8} \ln n).$$

Thus

$$\Pr(Z_i \geq n^{31/16} \mid \mathcal{E}_i) = O(n^{-1/16} \ln n)$$

and so

$$\Pr(\exists 1 \leq i \leq i_0 : Z_i \geq n^{31/16} \mid \mathcal{E}_0) = O(n^{-1/16} (\ln n)^2)$$

and

$$(8) \quad \Pr\left(\sum_{i=1}^{i_0} Z_i \geq i_0 n^{31/16}\right) = o(1).$$

Also

$$\Pr(u \in E_{i+1} \mid u \in E_i) = \left(1 - \frac{1}{X_2^*}\right)^{X_u} (1 - \varrho_u) < e^{-1}.$$

Thus

$$E(|E_{i_0+1}|) = O\left(\frac{n^2}{(\ln n)^4}\right)$$

and

$$(9) \quad \Pr\left(|E_{i_0+1}| \geq \frac{n^2}{(\ln n)^3}\right) = o(1).$$

Theorem 1 follows from (6), (8) and (9) and

$$|\mathcal{C}_{COVER}| = \sum_{i=1}^{i_0} Y_i + \sum_{i=1}^{i_0} Z_i + |E_{i_0+1}|.$$

As we only use estimates for  $X_{0, \lfloor k/i \rfloor, i}^*$  and  $X_{2, \lfloor k/i \rfloor, i}^*$  the reader may wonder why it is necessary to prove Lemma 1(a) for  $0 \leq s \leq j \leq k/i$ . The reason is simply that the lemma is proved by induction and we use a stronger induction hypothesis than the needed outcome.

### 3. Proof of Lemma 1

If  $s=j$  then  $X_{S,j,i} = c_{s,j,i} = 1$  and so we can assume  $s < j$  from now on.

Let us first consider  $\mathcal{E}_1$ . Fix a set  $S$  of size  $s$ ,  $0 \leq s \leq k$ . Assume it forms a clique in  $G$ . This does not condition any edges not contained in  $S$ . For a set  $T$  let  $N_c(T)$  denote the set of common neighbours of  $T$  in  $G$ . We can enumerate the set of  $j$ -cliques containing  $S$  as follows: choose  $x_1 \in N_c(S)$ ,  $x_2 \in N_c(S \cup \{x_1\})$ ,  $\dots$ ,  $x_{j-s} \in N_c(S \cup \{x_1, x_2, \dots, x_{j-s-1}\})$ . The number of choices  $\nu_t$

for  $x_t$  given  $x_1, x_2, \dots, x_{t-1}$  is distributed as  $\text{Bin}(n - (s - t + 1), p^{s+t-1})$ . Thus for  $0 \leq \varepsilon \leq 1$

$$\begin{aligned} \Pr \left( \left| \frac{\nu_t}{(n - s - t + 1)p^{s+t-1}} - 1 \right| \geq \varepsilon \right) &\leq 2 \exp \left\{ -\frac{\varepsilon^2(n - s - t + 1)p^{s+t-1}}{3} \right\} \\ &\leq 2 \exp \{-\varepsilon^2 n^{1-\alpha}/4\}. \end{aligned}$$

Putting  $\varepsilon = n^{-1/3}$  we see that since there are  $n^{O(\ln n)}$  choices for  $x_1, x_2, \dots, x_{j-s}$ ,

$$\Pr \left( \left| \frac{X_{S,j,0}}{c_{s,j,0}} - 1 \right| \geq n^{-1/3+o(1)} \right) \leq \exp\{-n^{1/4}\}.$$

There are  $n^{O(\ln n)}$  choices for  $S$  and (1) follows.

Assume now that  $\mathcal{E}_i$  holds. We first prove

**Lemma 2.** Suppose  $e_1, e_2, \dots, e_t \in E_i$ . Then

$$\Pr(e_t \in E_{i+1} \mid e_1, e_2, \dots, e_{t-1} \in E_{i+1}) = e^{-1} \left( 1 + O\left(\frac{t \ln n}{n}\right) \right)$$

uniformly for  $1 \leq t \leq n^{1/2}$ .

**Proof.**

$$\begin{aligned} (10) \quad \Pr(e_t \in E_{i+1} \mid e_1, e_2, \dots, e_{t-1} \in E_{i+1}) &\geq \Pr(e_t \in E_{i+1}) \\ &= \left(1 - \frac{1}{X_2^*}\right)^{X_u} (1 - \varrho_u) = e^{-1} - X_2^{*-1}. \end{aligned}$$

Here  $u = e_t$ ,  $X_u = X_{u, [k/i], i}$  and  $X_2^* = X_{2, [k/i], i}^*$  and inequality (10) follows from the fact that knowing  $e_1, e_2, \dots, e_{t-1} \in E_{i+1}$  tells us that certain cliques (and edges) were not chosen for  $\mathcal{C}_{COVER}$ . On the other hand

$$\begin{aligned} (11) \quad \Pr(e_t \in E_{i+1} \mid e_1, e_2, \dots, e_{t-1} \in E_{i+1}) &\leq \left(1 - \frac{1}{X_2^*}\right)^{X_u - tX_3^*} (1 - \varrho_u) \\ &= (e^{-1} - X_2^{*-1}) \left(1 - \frac{1}{X_2^*}\right)^{tX_3^*} = e^{-1} \left(1 + O\left(\frac{tX_3^*}{X_2^*}\right)\right), \end{aligned}$$

where  $X_3^* = X_{3, [k/i], i}^*$ . If  $\mathcal{E}_i$  holds then  $X_3^*/X_2^* = O(\ln n/n)$ .

Inequality (11) follows from the fact that  $e_t = u$  lies in at least  $X_u - (t-1)X_3^*$  cliques which contain none of  $e_1, e_2, \dots, e_{t-1}$ . This in turn arises from a two term inclusion-exclusion inequality and the fact that  $e_t$  and  $e_i$  together lie in at most  $X_3^*$  cliques, for  $1 \leq i \leq t-1$ . ■

Now fix a set  $S \in \mathcal{C}_{s,i}$  and let  $X = X_{S,j,i+1}$  for some  $j \leq k/(i+1)$ . Condition on  $S \in \mathcal{C}_{s,i+1}$ . Let  $\mathcal{C}_{S,j,i} = \{C \in \mathcal{C}_{j,i} : C \supseteq S\}$ . Then

$$\begin{aligned} \mathbb{E}(X) &= \sum_{C \in \mathcal{C}_{S,j,i}} \Pr(C \in \mathcal{C}_{j,i+1} \mid S \in \mathcal{C}_{s,i+1}) \\ (12) \quad &= X_{S,j,i} \exp \left\{ \binom{s}{2} - \binom{j}{2} \right\} \left( 1 + O \left( \frac{j^4 \ln n}{n} \right) \right), \end{aligned}$$

on using Lemma 2.

We are going to use the Markov inequality

$$(13) \quad \Pr(X \geq x) \leq \frac{\mathbb{E}((X)_r)}{(x)_r}$$

where  $(x)_r = x(x-1)(x-2)\dots(x-r+1)$  and  $r = \lfloor n^{3/8} \rfloor$ .

Let  $\mathcal{B}(\ell_2, \ell_3, \dots, \ell_r) = \{(C_1, C_2, \dots, C_r) : (i) \ C_t \neq C_{t'} \text{ for } t \neq t', (ii) \ C_t \in \mathcal{C}_{S,j,i}, (iii) \ |\mathcal{C}_t \cap (C_1 \cup C_2 \cup \dots \cup C_{t-1})| = s + \ell_t, \text{ for } t, t' = 2, 3, \dots, r\}$ . Then

$$\mathbb{E}((X)_r) = \sum_{\ell_2, \ell_3, \dots, \ell_r} \sum_{\mathcal{B}(\ell_2, \ell_3, \dots, \ell_r)} \Pr(C_1, C_2, \dots, C_r \in \mathcal{C}_{j,i+1} \mid S \in \mathcal{C}_{s,i+1}).$$

From (12)

$$\Pr(C_1 \in \mathcal{C}_{j,i+1} \mid S \in \mathcal{C}_{s,i+1}) = \exp \left\{ \binom{s}{2} - \binom{j}{2} \right\} \left( 1 + O \left( \frac{j^4 \ln n}{n} \right) \right)$$

and

$$\begin{aligned} &\Pr(C_t \in \mathcal{C}_{j,i+1} \mid C_1, C_2, \dots, C_{t-1} \in \mathcal{C}_{j,i+1}) \\ &= \exp \left\{ \binom{s + \ell_t}{2} - \binom{j}{2} \right\} \left( 1 + O \left( \frac{j^4 \ln n}{n} \right) \right) \\ &= \exp \left\{ \binom{s + \ell_t}{2} - \binom{s}{2} \right\} \frac{c_{s,j,i+1}}{c_{s,j,i}} \left( 1 + O \left( \frac{j^4 \ln n}{n} \right) \right). \end{aligned}$$

Also,

$$\begin{aligned} |\mathcal{B}(\ell_2, \ell_3, \dots, \ell_r)| &\leq \prod_{t=1}^r \left( \binom{(t-1)j - s}{\ell_t} X_{s+\ell_t,j,i}^* \right) \\ &\leq \prod_{t=1}^r (rj)^{\ell_t} (1 + \beta_i) \left( \frac{b^{s+\ell_t} j e^{i(s+\ell_t)}}{n} \right)^{\ell_t} c_{s,j,i}. \end{aligned}$$

Hence,

$$\begin{aligned} \frac{E((X)_r)}{c_{s,j,i+1}^r} &\leq \left(1 + O\left(\frac{(\ln n)^4 r}{n}\right)\right) \sum_{\ell_2, \ell_3, \dots, \ell_r} \prod_{t=1}^r (1 + \beta_i) \left(\frac{e^{(\ell_t + 2s - 1)/2} r j^2 (be^i)^{s + \ell_t}}{n}\right)^{\ell_t} \\ &\leq \left(1 + O\left(\frac{(\ln n)^4 r}{n}\right)\right) (1 + \beta_i)^r \sum_{\ell_2, \ell_3, \dots, \ell_r} \left(\frac{rk^2 e^{3k} b^{2k}}{n}\right)^{\ell_2 + \dots + \ell_t} \\ (15) \quad &\leq (1 + rn^{-3/4})(1 + \beta_i)^r, \end{aligned}$$

for  $\alpha$  sufficiently small.

Hence, using (13),

$$\begin{aligned} \text{by (15)} \quad \Pr(X \geq (1 + \beta_{i+1})c_{s,j,i+1}) &\leq \frac{2(1 + \beta_i)^r c_{s,j,i+1}^r}{((1 + \beta_{i+1})c_{s,j,i+1})^r}, \\ \text{using (4)} \quad &\leq 3 \left(\frac{1 + \beta_i}{1 + \beta_{i+1}}\right)^r, \\ &\leq 3 \exp\left\{-\frac{r(\beta_{i+1} - \beta_i)}{1 + \beta_{i+1}}\right\} = \exp\{-n^{1/8 - o(1)}\}. \end{aligned}$$

There are  $n^{O(\ln n)}$  choices for  $S$  and  $j$  and so part (a) of the lemma is proven. It remains only to deal with  $X_{u,j,i+1}$  for an edge  $u \in E_i$ . It follows from (12) that if  $X = X_{u,j,i+1}$  then

$$(16) \quad E(X) = X_{u,j,i} \exp\left\{\left\{\binom{s}{2} - \binom{j}{2}\right\} \left(1 + O\left(\frac{j^4 \ln n}{n}\right)\right)\right\},$$

and from (15) that

$$(17) \quad E(X(X-1)) \leq \left(1 + \frac{2}{n^{3/4}}\right) c_{2,j,i+1}^2.$$

Suppose now that  $X_{u,j,i} \geq (1 - \beta_i)c_{2,j,i}$ . Then (16) and (17) imply that

$$(18) \quad \Pr(X \leq (1 - \beta_{i+1})c_{2,j,i+1}) \leq 3n^{-1/4}.$$

Now let  $Z_{i+1}$  denote the number of edges  $u \in E_{i+1}$  for which  $X_{u,j,i+1} \leq (1 - \beta_{i+1})c_{2,j,i+1}$  and  $\hat{Z}_{i+1}$  those  $u$  counted in  $Z_{i+1}$  for which  $X_{u,j,i} \geq (1 - \beta_i)c_{2,j,i}$ . Then

$$Z_{i+1} \leq Z_i + \hat{Z}_{i+1}$$

and from (18)

$$E(\hat{Z}_{i+1} \mid \mathcal{E}_i) \leq 3|E_i|n^{-1/4}.$$

So

$$\Pr(Z_{i+1} \geq (i+1)n^{15/8} \mid \mathcal{E}_i) \leq \Pr(\hat{Z}_{i+1} \geq n^{15/8} \mid \mathcal{E}_i) = O(n^{-1/8}).$$

This completes the proof of Lemma 1.



## References

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