1993

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RANDOM GRAPH BY CLIQUES

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December 2, 1993

1   Introduction

The *clique cover number* $\theta_1(G)$ of a graph G is the minimum number of cliques required to cover the vertices 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 $n \to \infty$)

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

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

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

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

Remark 1: a simple use of a martingale tail inequality shows that θ_1 is close to its mean with very high probability.

2 Proof of Theorem 1

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

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 α 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 $\lfloor k/i \rfloor$ none of whose edges are covered by previously chosen cliques. We do

<div align="center">2</div>

this for $i_0 = \lceil 4 \ln \ln n \rceil$ iterations. At the start of iteration i there is a set E_i of edges which have not yet been covered. Let $C_{t,i}$ denote the set of t-cliques all of whose edges are in E_i. At iteration i we choose randomly cliques from $C_{i,\lfloor k/i \rfloor}$ with probability p_i (to be defined). p_i is chosen so that for most of the edges of E_i, the number of chosen cliques which contain that edge is approximately distributed as a Poisson random variable with mean close to one. (A few edges may be in few cliques and so we have to choose some random edges as well.) The process is thus designed so that for $u \in E_i$,

$$\mathbf{Pr}(u \in E_{i+1} \mid u \in E_i) \approx e^{-1}.$$

So 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 $C_{j,i}$ which contain a particular fixed clique in $C_{s,i}$.

We now need to describe our clique choosing process a little more formally: for a clique $S \in C_{s,i}$ we let

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

and for integer $s \geq 0$,

$$X_{s,j,i} = \max\{X_{S,j,i} : S \in C_{s,i}\}.$$

Algorithm COVER

begin

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

    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 ρ_u where

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

$X_2 = X_{2,\lfloor k/i \rfloor, i}$ 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 ρ_u assumes that X_2 is large (which it is **whp**) and so

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

and ρ_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)

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

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

(b)

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

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

Then

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

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

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

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

and

$$c_{s,j,i} \geq n^{j/2} \tag{4}$$

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

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.

$$
\begin{aligned}
\mathbf{E}(Y_i \mid \mathcal{E}_i) &= \mathbf{E}\left(\frac{X_{0,\lfloor k/i \rfloor,i}}{X_{2,\lfloor k/i \rfloor,i}} \,\Big|\, \mathcal{E}_i\right) \\
&\leq (1 + o(1))\frac{c_{0,\lfloor k/i \rfloor,i}}{c_{2,\lfloor k/i \rfloor,i}} \\
&\approx \frac{n^2 i^2}{bk^2 e^i},
\end{aligned}
\tag{5}
$$

on using (3)

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

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

Thus

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

and so

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

Now a simple calculation gives

$$\rho_u = O\left(\frac{X_2 - X_u}{X_2}\right) \tag{7}$$

and so

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

Thus

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

and so

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

and

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

Also

$$\begin{aligned}
\mathbf{Pr}(u \in E_{i+1} \mid u \in E_i) &= \left(1 - \frac{1}{X_2}\right)^{X_*}(1 - \rho_u) \\
&< e^{-1}.
\end{aligned}$$

Thus

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

and

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

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

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 $x_1 \in N_c(S)$, $x_2 \in N_c(S \cup \{x_1\}), \ldots, x_{j-s} \in N_c(S \cup \{x_1, x_2, \ldots, x_{j-s-1}\})$. The number of choices ν_t for x_t given $x_1, x_2, \ldots, x_{t-1}$ is distributed as $Bin(n - (s - t + 1), p^{s+t-1})$. Thus for $o \leq \epsilon \leq 1$

$$\mathbf{Pr}\left(\left|\frac{\nu_t}{(n-s-t+1)p^{s+t-1}} - 1\right| \geq \epsilon\right) \leq 2\exp\left\{-\frac{\epsilon^2(n-s-t+1)p^{s+t-1}}{3}\right\}$$

$$\leq 2\exp\{-\epsilon^2 n^{1-\alpha}/4\}.$$

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

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

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 $e_1, e_2, \ldots, e_t \in E_i$. Then*

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

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

Proof

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

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

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

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where $X_3 = X_{3,\lfloor k/i \rfloor,i}$. If $\mathcal{E}_i$ holds then $X_3/X_2 = O(\ln n/n)$. $\qquad\square$

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

$$
\begin{aligned}
\mathbf{E}(X) &= \sum_{C \in \mathcal{C}_{S,j,i}} \mathbf{Pr}(C \in \mathcal{C}_{j,i+1} \mid S \in \mathcal{C}_{s,i+1}) \\
&= X_{S,j,i} \exp\left\{\binom{s}{2} - \binom{j}{2}\right\}\left(1 + O\left(\frac{j^4 \ln n}{n}\right)\right),
\end{aligned}
\tag{11}
$$

on using Lemma 2.

We are going to use the Markov inequality

$$
\mathbf{Pr}(X \geq x) \leq \frac{\mathbf{E}((X)_r)}{(x)_r}
\tag{12}
$$

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

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

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

From (11)

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

and

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

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Also,

$$|\mathcal{B}(\ell_2, \ell_3, \ldots, \ell_r)| \leq \prod_{t=1}^{r} \left(\binom{(t-1)j - s}{\ell_t} X_{s+\ell_t, j, i} \right)$$

$$\leq \prod_{t=1}^{r} (rj)^{\ell_t} (1+\beta_i) \left(\frac{b^{s+\ell_t} j e^{i(s+\ell_t)}}{n} \right)^{\ell_t} c_{s,j,i}.$$

Hence,

$$\frac{\mathbf{E}((X)_r)}{c_{s,j,i+1}^r} \leq \left(1 + O\left(\frac{(\ln n)^4 r}{n} \right) \right) \sum_{\ell_2, \ell_3, \ldots, \ell_r} \prod_{t=1}^{r} (1+\beta_i) \left(\frac{e^{(\ell_t + 2s - 1)/2} r j^2 (be^i)^{s+\ell_t}}{n} \right)^{\ell_t}$$

$$\leq \left(1 + O\left(\frac{(\ln n)^4 r}{n} \right) \right) (1+\beta_i)^r \sum_{\ell_2, \ell_3, \ldots, \ell_r} \left(\frac{r k^2 e^{3k} b^{2k}}{n} \right)^{\ell_2 + \cdots + \ell_t} \tag{13}$$

$$\leq (1 + rn^{-3/4})(1+\beta_i)^r, \tag{14}$$

for α sufficiently small.

Hence, using (12),

$$\mathbf{Pr}(X \geq (1+\beta_{i+1})c_{s,j,i+1}) \leq \frac{2(1+\beta_i)^r c_{s,j,i+1}^r}{((1+\beta_{i+1})c_{s,j,i+1})_r}, \qquad \text{by (14)}$$

$$\leq 3 \left(\frac{1+\beta_i}{1+\beta_{i+1}} \right)^r, \qquad \text{using (4)}$$

$$\leq 3 \exp \left\{ -\frac{r(\beta_{i+1} - \beta_i)}{1+\beta_{i+1}} \right\}$$

$$= \exp\{-n^{1/4 - o(1)}\}.$$

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 (11) that if $X = X_{u,j,i+1}$ then

$$\mathbf{E}(X) = X_{u,j,i} \exp \left\{ \binom{s}{2} - \binom{j}{2} \right\} \left(1 + O\left(\frac{j^4 \ln n}{n} \right) \right), \tag{15}$$

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and from (14) that

$$\mathbf{E}(X(X-1)) \leq \left(1 + \frac{2}{n^{3/4}}\right) c_{2,j,i+1}^2.$$ (16)

Suppose now that $X_{u,j,i} \geq (1 - \beta_i)c_{2,j,i}$. Then (15) and (16) imply that

$$\mathbf{Pr}(X \leq (1 - \beta_{i+1})c_{2,j,i+1}) \leq 3n^{-1/4}.$$ (17)

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 (17)

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

So

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

this completes the proof of Lemma 1.

References

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[2] N.Pippenger and J.Spencer, *Asymptotic behaviour of the chromatic index for hypergraphs*, Journal of Combinatorial Theory A (1989) 24-42.

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