THE CHROMATIC NUMBER OF RANDOM INTERSECTION GRAPHS

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Abstract

We study problems related to the chromatic number of a random intersection graph \( G(n, m, p) \). We introduce two new algorithms which colour \( G(n, m, p) \) with almost optimum number of colours with probability tending to 1 as \( n \to \infty \). Moreover we find a range of parameters for which the chromatic number of \( G(n, m, p) \) asymptotically equals its clique number.

Keywords: random intersection graphs, chromatic number, colouring algorithms.

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1. Introduction

Let \( V = \{v_1, \ldots, v_n\} \) be a set of vertices and \( W = \{w_1, \ldots, w_m\} \) be an auxiliary set of objects. Each vertex \( v \in V \) is assigned a subset of objects \( W_v \subseteq W \). The intersection graph generated by a family of sets \( \{W_v \subseteq W : v \in V\} \) is the graph with vertex set \( V \) in which two vertices \( v \) and \( v' \) are connected by an edge if their sets of objects \( W_v \) and \( W_{v'} \) intersect (i.e., \( W_v \cap W_{v'} \neq \emptyset \)). An intersection graph generated by a family of sets \( \{W_v \subseteq W : v \in V\} \) is a random intersection graph if for each \( W_v \) the set \( v \in V \) is chosen at random.

In this paper we will study the asymptotic (as \( n \to \infty \)) properties of a random intersection graph \( G(n, m, p) \) introduced by Karoński, Scheinerman and Singer-Cohen in [10]. In a random intersection graph \( G(n, m, p) \) for each vertex \( v \in V \) and each object \( w \in W \) we have \( w \in W_v \) with probability \( p \) independently of all other vertices and objects. In what follows we will use the

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asymptotic notation \( o(\cdot) \), \( \ll \) and \( \gg \) consistently with [9]. Moreover we will write that a random graph on \( n \) vertices \( G_n \) has property \( \mathcal{A} \) with high probability if 

\[
\Pr \{ G_n \text{ has property } \mathcal{A} \} \to 1 \text{ as } n \to \infty.
\]

In [10], among others, it was shown that for some range of parameters the asymptotic behaviour of \( G(n, m, p) \) differs from this of a classical random graph \( G(n, \hat{p}) \) (with the vertex set \( V \), in which each edge appears independently with probability \( \hat{p} \)). The subsequent article by Fill, Scheinerman and Singer-Cohen [6] (see also [14]) determined a range of parameters for which both models \( G(n, m, p) \) and \( G(n, \hat{p}) \) are equivalent. It appears that for many graph properties there exists a range of parameters for which \( G(n, m, p) \) and \( G(n, \hat{p}) \) with the same edge density have the property with asymptotically (as \( n \to \infty \)) the same probability and outside which \( G(n, m, p) \) and \( G(n, \hat{p}) \) differ. We will be particularly interested in some of those properties related to the chromatic number of \( G(n, m, p) \). We will mainly concentrate on analysing algorithmic aspects of the problem.

The classical results concerning \( G(n, \hat{p}) \) ([4, 5, 8, 12, 16], see also Chapter 7 in [9]) show that as \( 1/n \ll \hat{p} \ll 1 \) with high probability \( \chi(G(n, \hat{p})) = (1 + o(1)) \frac{n}{\alpha(G(n, \hat{p}))} \), where \( \chi(G) \) is the chromatic number of \( G \) and \( \alpha(G) \) is its independence (stability) number. Therefore for sparse \( G(n, \hat{p}) \) with \( \hat{p} \) above the phase transition threshold, a trivial relation \( \chi(G(n, \hat{p})) \geq n/\alpha(G(n, \hat{p})) \) is with high probability asymptotically tight. Moreover, for \( 1/n \ll \hat{p} \ll 1 \), the greedy algorithm (or its slight modification) with high probability uses \( (1 + o(1))2\chi(G(n, \hat{p})) \) colours to colour properly \( G(n, \hat{p}) \). Last but not least, no polynomial time algorithm which with high probability uses at most \( (2 - \delta)G(n, \hat{p}) \) colours (for some constant \( \delta > 0 \)) has been found so far. For an extensive discussion on the problem see [7] or Chapter 7 in [9].

The problems related to the chromatic number of \( G(n, m, p) \) were first studied by Behrisch, Taraz and Ueckerdt [2] and then by Nikoletseas, Raptopoulos and Spirakis [13]. In [13] the authors analysed the problem of colourability of almost all vertices (i.e., all but at most \( o(n) \) vertices) while in [2] two colouring algorithms were studied. Each of the algorithms analysed in [2], for some range of parameters, with high probability colour \( G(n, m, p) \) with \( \omega(G(n, m, p)) \) colours. Here by \( \omega(G) \) we denote the clique number of a graph \( G \). In particular in [2] the following two theorems were shown.

**Theorem 1.** Let \( m \) = \( n^\beta \) with \( \beta > 0 \) fixed and \( mp \ll \sqrt{mp} \) (i.e., \( mp^2 \ll \frac{1}{n} \)). Then \( G(n, m, p) \) can with high probability be coloured optimally in linear time and \( \chi(G(n, m, p)) = \omega(G(n, m, p)) \).

If \( mp^2 = o(1) \) then the probability that two vertices are connected by an edge in \( G(n, m, p) \) equals \( 1 - (1 - p^2)^m = (1 + o(1))mp^2 \). Moreover \( mp^2 = \frac{1}{n} \) is a threshold function for phase transition in \( G(n, m, p) \) [1]. Therefore Theorem 1 describes the structure of \( G(n, m, p) \) far below phase transition threshold. For
\(G(n, \hat{p})\) this case is trivial since then with high probability in \(G(n, \hat{p})\) all components are trees. The proof of Theorem 1 basis on the fact that for \(mp^2 \ll 1/n\), \(\mathcal{G}(n, m, p)\) with high probability has no induced cycles of length greater than 3.

**Theorem 2.** Let \(m = n^\beta\) with \(0 < \beta < 1\) fixed and \(mp \ll \frac{1}{m \ln m}\). Then \(\mathcal{G}(n, m, p)\) can with high probability be coloured optimally in linear time. Moreover, for \(np > \ln^4 n\) we have with high probability

\[
\chi(\mathcal{G}(n, m, p)) = \omega(\mathcal{G}(n, m, p)) = (1 + o(1))np.
\]

Theorems 1 and 2 show that if \(mp\) is ,,small” then with high probability

\[
\chi(\mathcal{G}(n, m, p)) = \omega(\mathcal{G}(n, m, p)),
\]

i.e., a trivial relation \(\chi(\mathcal{G}(n, m, p)) \geq \omega(\mathcal{G}(n, m, p))\) is tight. Moreover these results give algorithms which, in this range of parameters, use exactly \(\chi(\mathcal{G}(n, m, p))\) colours. A natural question arises, for what choice of \(n, m\) and \(p\) with high probability

\[
\chi(\mathcal{G}(n, m, p)) = (1 + o(1))\omega(\mathcal{G}(n, m, p))
\]

and when with high probability

\[
\chi(\mathcal{G}(n, m, p)) = (1 + o(1))\frac{n}{\alpha(\mathcal{G}(n, m, p))}.
\]

Moreover one could ask whether, in the case where with high probability (1) is fulfilled, there exists a simple algorithm using almost the optimum number of colours. As the chromatic number and the independence number are tightly related, based on results determining \(\alpha(\mathcal{G}(n, m, p))\) [15] one could state the following conjecture.

**Conjecture 1.** Let \(m \geq n^\beta\) for some constant \(\beta \in (0, 1)\) and \(\frac{1}{n} \ll mp^2 \ll 1\).

(i) If \(m = o(n)\) and \(mp \ll \ln \frac{n}{m}\) then with high probability

\[
\chi(\mathcal{G}(n, m, p)) = (1 + o(1))\omega(\mathcal{G}(n, m, p))
\]

and there exists a polynomial time algorithm which with high probability colours \(\mathcal{G}(n, m, p)\) with \((1 + o(1))\chi(\mathcal{G}(n, m, p))\) colours.

(ii) If \(n = O(m)\) or \(m = o(n)\) and \(mp \gg \ln \frac{n}{m}\) then with high probability

\[
\chi(\mathcal{G}(n, m, p)) = (1 + o(1))\frac{n}{\omega(\mathcal{G}(n, m, p))}.
\]

We will propose two new algorithms which colour intersection graphs. The analysis of them shows that Conjecture 1 is true in two cases. The first one is \(np \geq 2 \ln n\) and \(mp = o(\frac{\ln \ln n}{\ln n})\). The second case is \(\ln n \ll mp \ll \ln \frac{n}{m \ln m}\) and \(np \gg \ln n\). Therefore we considerably sharpen Theorem 2 and propose new ideas which possibly might be used for other values of \(n, m\) and \(p\).
Recall that \( V = \{v_1, \ldots, v_n\} \) and \( W = \{w_1, \ldots, w_m\} \). In what follows we use the following notation.

\[
V_w = \{v \in V : w \in W\}, \quad \text{for } w \in W;
\]
\[
X_i = |V_{v_i}|, 1 \leq i \leq m;
\]
\[
Y_i = |W_{w_i}|, 1 \leq i \leq n.
\]

Note that in \( G(n,m,p) \) sets \( W_{v_i} (1 \leq i \leq n) \) are independent and \( Y_i (1 \leq i \leq n) \) are independent with the binomial distribution \( \text{Bin}(m,p) \). Moreover for all \( w \in W \) sets \( V_w = \{v \in V : w \in W\} \) are independent and their sizes \( X_i (1 \leq i \leq m) \) are independent with the binomial distribution \( \text{Bin}(n,p) \). By colouring greedily we will mean using the algorithm which, given a palette of colours \( \{k_1, k_2, \ldots\} \) and a graph with an ordered set of vertices \( \{v_1, v_2, \ldots, v_n\} \), colours properly vertices one by one, using each time the available colour with the smallest index (i.e., for each vertex, the algorithm uses the colour with the smallest index chosen from the set of colours which have not been used to colour the neighbours of the considered vertex).

2. Result for \( mp \ll \ln \ln n / \ln n \).

**Algorithm 1**

INPUT: an intersection graph \( G \) with an ordered vertex set \( V \) generated by a family of sets \( \{W_v : v \in V\} \) and an infinite palette of colours;

OUTPUT: a proper vertex colouring of \( G \);

1. \( V_2 := \{v \in V : |W_v| \geq 2\} \);

2. Colour greedily the vertices of \( G[V_2] \), the subgraph of \( G \) induced on \( V_2 \);

3. Colour greedily the remaining vertices of \( G \).

**Theorem 3.** Let \( p \in (0,1) \) and let \( n, m \) be positive integers. For any \( \varepsilon \in (0,1) \) and any \( C > 1 + \varepsilon \) there exists \( c \in (0,1) \) such that if

\[
np \geq C \ln n \quad \text{and} \quad mp \leq c \frac{\ln \ln n}{\ln n},
\]

then for large \( n \) with probability at least \( 1 - 3n^{-\varepsilon} \) Algorithm 1 colours \( G(n,m,p) \) with \( \omega(G(n,m,p)) \) colours.
Proof. Let \{k_1, k_2, \ldots\} be a palette of colours. Let \(G\) be an intersection graph generated by a family of sets \(\{W_v \subseteq W, v \in V\}\). The greedy colouring in Step 2 of Algorithm 1 uses at most \(\Delta(G[V_2]) + 1\) colours. Moreover for any \(i\) and \(j, i \neq j\), in \(G\) there is no edge between \(V_{w_i} \setminus V_2\) and \(V_{w_j} \setminus V_2\). Therefore, after Step 2, the largest index of the colours from \(\{k_1, k_2, \ldots\}\) used to colour vertices from \(V_{w_i}\) \((w_i \in W)\) is at most the maximum of \(\Delta(G[V_2]) + 1\) and \(X_i\). Thus Algorithm 1 uses at most \(\max\{\Delta(G[V_2]) + 1, X_1, \ldots, X_m\}\) colours. On the other hand \(\max_{1 \leq i \leq m} X_i \leq \omega(G)\). Therefore to prove the theorem it is enough to prove that under conditions of the theorem in \(G(n, m, p)\) with probability at least \(1 - 3n^{-\varepsilon}\) we have

\[
\Delta(G(n, m, p) [V_2]) < \max_{1 \leq i \leq m} X_i.
\]

Let

\[
\psi(x) = x \ln x - x + 1.
\]

By Chernoff’s inequality (see for example Theorem 2.1 in [9]) for any random variable \(X\) with the binomial distribution with mean \(\mu\) we have

\[
\Pr\{X \geq x\mu\} \leq \exp(-\psi(x) \mu) \quad \text{for } x > 1;
\]

\[
\Pr\{X \leq x\mu\} \leq \exp(-\psi(x) \mu) \quad \text{for } x \in (0, 1).
\]

Let \(\varepsilon > 0, C > (1 + \varepsilon), \text{ and } np \geq C \ln n\). Moreover, define \(A \in (0, 1)\) and \(c \in (0, 1)\) to be such that

\[
\psi(A) = \frac{(1 + \varepsilon) \ln n}{np} \quad \text{and} \quad A \ln c - c + 1 = 0.
\]

Note that \(\frac{(1 + \varepsilon) \ln n}{np} < 1\), i.e., \(A\) is well defined.

Moreover (2) implies \(m \leq n\). Therefore by (6) and (7)

\[
\Pr\left\{ \forall_{1 \leq i \leq m} X_i > Anp \right\} \geq 1 - \Pr\left\{ \exists_{1 \leq i \leq m} X_i \leq Anp \right\}
\]

\[
\geq 1 - m \exp(-(1 + \varepsilon) \ln n) = 1 - n^{-\varepsilon}.
\]

Thus

\[
\Pr\left\{ \max_{1 \leq i \leq m} X_i > Anp \right\} \geq 1 - n^{-\varepsilon}.
\]

Now we will find the upper bound on \(\Delta(G(n, m, p) [V_2])\). For \(v_i \in V_2\), let \(Z_i\) \((1 \leq i \leq n)\) be the number of neighbours of vertex \(v_i \in V\) which are contained...
in $V_2$. Moreover let $Z_i = 0$ for $v_i \notin V_2$. Given $Y_i = d$ ($d \geq 2$) $Z_i$ has the binomial distribution with parameters $n-1$ and

$$\left(1 - (1-p)^d\right) - dp(1-p)^{m-1} \leq dmp^2.$$ 

Therefore, under condition $Y_i = d$ ($d \geq 2$) $Z_i$ is stochastically dominated by random variable $Z^*_{i,d}$ with the binomial distribution $\text{Bin}(n,dmp^2)$.

By (5), for large $n$,

$$\Pr \left\{ Y_i \geq \left\lceil \frac{\ln n}{\ln \ln n} \right\rceil \right\} \leq \exp \left( -\frac{\ln n}{\ln \ln n} \cdot \ln \left( \frac{\ln n}{\ln \ln n} \right) + \frac{\ln n}{\ln \ln n} - mp \right)$$

$$\leq \exp \left( -\frac{\ln n}{\ln \ln n} \cdot (2\ln \ln n - 2\ln \ln n) + \frac{\ln n}{\ln \ln n} - mp \right)$$

$$\leq \exp \left( -2\ln n \left( 1 - \frac{\ln \ln n}{\ln \ln n} - \frac{1}{2\ln \ln n} + \frac{mp}{2\ln n} \right) \right)$$

$$\leq \exp(-\frac{n\ln n}{\ln n} - 1 - \varepsilon) = n^{-1-\varepsilon}.$$ 

In the second inequality we have used the fact that by definition we have $c < 1$, i.e., $mp \leq \frac{\ln n}{\ln \ln n} \leq \frac{\ln n}{\ln \ln n}$.

Moreover for $d \leq \left\lceil \frac{\ln n}{\ln \ln n} \right\rceil$ by (4), (5), and (7) we get

$$\Pr \left\{ Z^*_{i,d} \geq Anp \right\} \leq \exp \left( -dnmp^2 \varphi \left( \frac{A}{dmp} \right) \right)$$

$$= \exp \left( -dnmp^2 \left( \frac{A}{dmp} \ln \frac{A}{dmp} - \frac{A}{dmp} + 1 \right) \right)$$

$$= \exp \left( -np(A \ln A - A \ln dmp - A + dmp) \right)$$

$$\leq \exp \left( -np\varphi(A) + np(A \ln(dmp) - dmp + 1) \right)$$

$$\leq \exp \left( -(1+\varepsilon) \ln n + np(A \ln c - c + 1) \right)$$

$$= \exp \left( -(1+\varepsilon) \ln n \right) = n^{-1-\varepsilon}.$$ 

In the first inequality we used (5). The second last line follows by (7) and by the fact that function $f(x) = A \ln x - x + 1$ is increasing for $x \leq A$ (recall that $dmp \leq c \leq A$). In order to obtain the last line, (7) was used.

Thus

$$\Pr \left\{ \Delta(G(n,m,p) \mid V_2) \geq Anp \right\} = \Pr \left\{ \exists \ 1 \leq i \leq n \ Z_i \geq Anp \right\} \leq n \Pr \left\{ Z_1 \geq Anp \right\}$$
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\[ n \leq \sum_{d=2}^{\ln n} \Pr \{ Z_{1,d} \geq Anp \} \Pr \{ Y_1 = d \} + n \Pr \left\{ Y_1 \geq \left\lfloor \frac{\ln n}{\ln \ln n} \right\rfloor \right\} \leq 2n^{-\varepsilon}. \]

Therefore by (8) and (9) with probability at least \( 1 - 3n^{-\varepsilon} \)

\[ \Delta(\mathcal{G}(n,m,p)[V_2]) < Anp < \max_{1 \leq i \leq m} X_i \leq \omega(\mathcal{G}(n,m,p)), \]

i.e., (3) is fulfilled.

3. Result for \( mp \gg \ln \ln n \)

The result presented in this section will partly rely on Theorem 1.2 from [11].

The result from [11] concerns a uniform random intersection graph in which, for all \( v \in V \), \( W_v \) is chosen independently and uniformly at random from all \( D \)-element subsets of \( W \). In [11] it was shown that, under some conditions on \( D \), \( n \) and \( m \), a greedy random algorithm colours a uniform random intersection graph with almost the optimum number of colours.

We will use the fact that the algorithm analysed in [11] may be easily rewritten to work on any random intersection graph in which, for each \( v \in V \), \( |W_v| \leq D \) and \( W_v \) is chosen independently and uniformly at random from all subsets of \( W \) of size \( |W_v| \). Namely each vertex \( v \in V \) would independently choose additional \( D - |W_v| \) vertices from \( W \setminus W_v \) to obtain a uniform random intersection graph. Then one would run the algorithm described in [11] on the constructed uniform random intersection graph. We will call the above described algorithm \( \text{Algorithm}[D] \).

**Lemma 4** (Corollary of Theorem 1.2 from [11]). Let \( D, m, n \geq 2 \), \( \bar{D} = \frac{Dn}{m} \), and \( V_{\leq D} := \{ v \in V : |W_v| \leq D \} \). For any \( \delta > 0 \) there is a constant \( c_\delta > 0 \) such that the following holds. Suppose that

\[ D \leq c_\delta \ln \left( \frac{m}{\ln D} \right) \quad \text{and} \quad D \leq c_\delta \ln \left( \frac{\bar{D}}{\ln m} \right). \]

Then \( \text{Algorithm}[D] \) properly colours all vertices of \( \mathcal{G}(n,m,p)[V_{\leq D}] \) with at most \( \lceil \bar{D}(1 + \delta) \rceil \) colours with probability at least \( 1 - 2m^{-1} - 2\bar{D}^{-1} \).

**Remark 5.** Let \( D \geq 2 \) and \( m \geq n^\beta \) for some \( \beta \in (0,1) \). Then by a simple calculation we get that for large \( n \)

\[ \ln \left( \frac{\bar{D}}{\ln m} \right) \geq \ln \left( \frac{n}{m \ln m} \right) \]
and
\[
\ln \left( \frac{m}{\ln D} \right) = \ln \left( \left( \frac{n}{m \ln m} \right)^{\beta} \right) + \ln \left( \frac{m^{1+\beta}(\ln m)^{\beta}}{n^{\beta} \ln D} \right) \geq \beta \ln \left( \frac{n}{m \ln m} \right).
\]

Algorithm 2

INPUT: an intersection graph \( G \) with an ordered vertex set \( V \) generated by a family of sets \( \{W_v : v \in V\} \);
an infinite palette of colours;
integer \( D \);

OUTPUT: a proper vertex colouring of \( G \);

1. \( V_{\leq D} := \{v \in V : |W_v| \leq D\} \);
2. Colour \( G[V_{\leq D}] \) using Algorithm\([D]\);
3. Colour greedily remaining vertices of \( G \).

Theorem 6. Let \( p \in (0,1) \) and \( m \geq n^\beta \) for some constant \( \beta \in (0,1) \). For all \( \delta \in (0,1) \) there exists \( c \) and \( C \) such that if
\[
\frac{1}{c} \ln \ln n \leq mp \leq c \ln \left( \frac{n}{m \ln m} \right) \quad \text{and} \quad np \geq C \ln n
\]
and
\[
D = \left\lceil \left( 1 + \frac{\delta}{5} \right) mp \right\rceil + 1,
\]
then for large \( n \) with probability at least \( 1 - 2m^{-1} - 2(np)^{-1} - 2n^{-0.2} \), Algorithm 2 colours \( G(n,m,p) \) with at most \((1+\delta)np\) colours and
\[
\chi(G(n,m,p)) \leq (1+\delta)np.
\]

In the proof we will use the following lemma.

Lemma 7 (Lemma 1 in [3]). Let \( d_i \) and \( d_j \) be positive integers and \( 0 < \xi < 1 \) be such that \( d_i + d_j \leq \xi m \). Then the probability that \( v_i \) and \( v_j \) are neighbours in \( G(n,m,p) \) under the condition that \( |W_{v_i}| = d_i \) and \( |W_{v_j}| = d_j \) is
\[
\frac{d_id_j}{m} \left( 1 - \frac{d_id_j}{m} \right) \leq \Pr \left\{ W_{v_i} \cap W_{v_j} \neq \emptyset \mid |W_{v_i}| = d_i, |W_{v_j}| = d_j \right\}
\leq \frac{d_id_j}{m} \left( 1 + \frac{2}{(1 - \xi) \frac{d_id_j}{m}} \right).
\]

Proof of Theorem 6. Let
\[
c = \min \left\{ \frac{1}{8}, \frac{\psi (1 + \delta/5)}{2}, \frac{\beta \xi^{4/5}}{(1 + \delta/4)} \right\},
\]

\[
(11)
\]
where $\psi(\cdot)$ is defined in (4) and $c_\delta/4$ is as in Lemma 4.

Let $D = [(1 + \delta/5)mp] + 1$. By assumption $c^{-1}\ln\ln n \leq mp$, therefore $D \leq (1 + \delta/4)mp$ for large $n$. Moreover, let

\[ V_{\leq D} = \{ v_i \in V : Y_i \leq D \}, \]
\[ V_{> D} = \{ v_i \in V : D < Y_i \leq \ln n \}. \]

In what follows we will prove that

(12) $\Pr\{\big(V_{\leq D} \cup V_{> D}\big) = V\} \geq 1 - n^{-0.2}$

and

(13) $\Pr\left\{ \Delta(G(n, m, p) \mid V_{> D}) \leq \frac{\delta}{4}np \right\} \geq 1 - n^{-0.2}$.

By the definition of $D$ and (11), for large $n$, we have

\[ D \leq \left(1 + \frac{\delta}{4}\right) mp \leq \left(1 + \frac{\delta}{4}\right)c\ln\frac{n}{m\ln m} \leq \beta c_\delta/4 \ln\frac{n}{m\ln m}. \]

Thus by Remark 5 and by Lemma 4 with probability at least $1 - 2m^{-1} - 2(np)^{-1}$ ALGORITHM[D] colours $G(n, m, p) \mid V_{\leq D}$ with at most

\[ \left(1 + \frac{\delta}{4}\right)\frac{Dn}{m} \leq \left(1 + \frac{\delta}{4}\right)^2 np \leq \left(1 + \frac{3\delta}{4}\right)np \]

colours. This combined with (12) and (13) shows that for large $n$ with probability at least $1 - 2m^{-1} - 2(np)^{-1} - 2n^{-0.2}$ ALGORITHM 2 uses at most $(1 + \delta)np$ colours to colour $G(n, m, p)$. We are left with showing (12) and (13).

Note that $mp \leq c\ln n/(m\ln m) \leq c\ln n$ (for large $n$), i.e., $\ln n/mp \geq c^{-1}$. Moreover $x^{-1}\psi(x)$ is increasing for $x \geq 1$. Therefore by (5) and (11), for large $n$ and any $1 \leq i \leq n$

\[ \Pr\{Y_i \geq \ln n\} \leq \exp\left(-\ln n \cdot \frac{mp}{\ln n} \cdot \psi\left(\frac{\ln n}{mp}\right)\right) \]
\[ \leq \exp\left(-\ln n \cdot \frac{\psi(\psi^{-1})}{e^{-1}}\right) \leq \exp\left(-\ln n \cdot \frac{\psi(8)}{8}\right) \leq n^{-1.2}. \]

Thus

\[ \Pr\left\{ \exists 1 \leq i \leq n \ Y_i \geq \ln n \right\} \leq \sum_{i=1}^{n} \Pr\{Y_i \geq \ln n\} \leq n^{-0.2}, \]

which implies (12).
Now we focus on showing (13). For \(v_i, v_j \in V\), denote by \(v_i \sim v_j\) the event that \(v_i\) and \(v_j\) are neighbours in \(G(n, m, p)\) (i.e., \(W_{v_i} \cap W_{v_j} \neq \emptyset\)). Recall that \(Y_i\) and \(Y_j\) are independent random variables. For \(d_i, d_j \leq \ln n\) and large \(n\) we have \(d_i + d_j \leq m/2\) and \(d_i d_j / m \leq 1/2\). Therefore by Lemma 7

\[
\Pr \{v_i \sim v_j \cap \{D \leq Y_j \leq \ln n\} | Y_i = d_i\} \leq \frac{3p \ln n}{\ln n}.
\]

Recall that \(Y_1, Y_2, \ldots, Y_n\) are independent random variables and one may choose \(W_{v_i}\) by first choosing the value \(Y_i\) and given \(Y_i = d_i\), choosing \(W_{v_i}\) uniformly at random from all \(d_i\)-element subsets of \(W\) (each independently of all other vertices in \(V\)). Therefore, given \(Y_i = d_i\), \(D \leq d_i \leq \ln n\), degree of vertex \(v_i\) in \(G(n, m, p)[V_{\geq D}]\) has the binomial distribution

\[
\text{Bin}(n-1, \Pr \{v_i \sim v_j \cap \{D \leq Y_j \leq \ln n\} | Y_i = d_i\}),
\]

which is stochastically dominated by a random variable \(Z^*\) with the binomial distribution \(\text{Bin}(n, 3p/\ln n)\).
Let $n$ be large enough to have $\delta \geq 24/\ln n$. If $C \geq 25/\delta$ and $D \leq d_i \leq \ln n$ then by (5) and the monotonicity of $\psi(x)/x$

$$
\Pr\left\{ Z^* \geq \frac{\delta}{4} np \right\} \leq \exp\left( -\frac{3np}{\ln n} \cdot \frac{\delta \ln n}{12} \cdot \frac{\psi\left(\frac{\delta \ln n}{12}\right)}{\frac{\delta \ln n}{12}} \right)
$$

$$
\leq \exp\left( -\frac{C\delta \ln n}{4} \cdot \frac{\psi(2)}{2} \right) \leq n^{-1.2}.
$$

Therefore

$$
\Pr\left\{ \Delta(G(n,m,p) [V \geq D]) \geq \frac{\delta}{4} np \right\} \leq \sum_{1 \leq i \leq n} \Pr\left\{ Z^* \geq \frac{\delta}{4} np \right\} \leq n^{-0.2},
$$

which implies (13).

**References**


doi:10.1017/S0963548398003459

doi:10.1137/130942292

doi:10.1007/BF01375472

doi:10.1007/978-3-642-03816-7_51

doi:10.1002/rsa.20356

doi:10.1016/j.tcs.2014.01.006

doi:10.1016/0196-6774(84)90003-8

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