

The t -improper chromatic number of random graphs

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11 September 2007
EuroComb07, Seville



Introduction

We consider the t -improper chromatic number of the Erdős-Rényi random graph.



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Introduction

We consider the t -improper chromatic number of the Erdős-Rényi random graph.

- ▶ $G_{n,p}$ — random graph with vertex set $[n] = \{1, \dots, n\}$, edges included independently with probability p .
- ▶ t -dependent set of G — a vertex subset of G which induces a subgraph of maximum degree at most t .
- ▶ t -improper chromatic number $\chi^t(G)$ of G — fewest colours needed in a t -improper colouring of G , a colouring of the vertices of G in which colour classes are t -dependent sets.

Note: $\chi^0(G) = \chi(G)$.

Improper colouring background

Cowen, Cowen and Woodall (1986) considered, for fixed $t \geq 0$, the t -improper chromatic number of planar graphs. Combined with FCT, they completely pinned down the behaviour of χ^t :

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Theorem (Cowen, Cowen and Woodall, 1986)

- ▶ *Every planar graph is 2-improperly 3-colourable,*
- ▶ \exists *planar graph that is not 1-improperly 3-colourable, and*
- ▶ \exists *planar graphs that are not t -improperly 2-colourable.*

Improper colouring basics

Proposition

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Proposition (Lovász, 1966)

$$\chi^t(G) \leq \left\lceil \frac{\Delta(G)+1}{t+1} \right\rceil.$$

The chromatic number of random graphs

(a very brief history)

We say that a property holds asymptotically almost surely (a.a.s.) if it holds with probability tending to one as $n \rightarrow \infty$.

Fix $p > 0$ and let $\gamma = \frac{2}{\ln \frac{1}{1-p}}$.

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Theorem (Grimmett and McDiarmid, 1975)

$$(1 - \varepsilon) \frac{n}{\gamma \ln n} \leq \chi(G_{n,p}) \leq (2 + \varepsilon) \frac{n}{\gamma \ln n} \text{ a.a.s.}$$

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Theorem (Bollobás, 1988, Matula and Kučera, 1990)

$$\chi(G_{n,p}) \sim \frac{n}{\gamma \ln n} \text{ a.a.s.}$$



Random improper colouring basics

Proposition

$$(1 - \varepsilon) \frac{n}{t \gamma \ln n} \leq \chi^t(\mathbf{G}_{n,p}) \leq (1 + \varepsilon) \frac{n}{\gamma \ln n} \text{ a.a.s.}$$

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Proposition

$$\chi^t(G_{n,p}) \leq (1 + \varepsilon) \frac{np}{t} \text{ a.a.s.}$$



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The upper bounds of the previous slide give the correct behaviour in nearly all choices of $t = t(n)$:

- ▶ if $t(n) \ll \ln n$, then $\chi^t(G_{n,p})$ is near $\chi(G_{n,p})$;
- ▶ if $t(n) \gg \ln n$, then $\chi^t(G_{n,p})$ is near $\Delta(G_{n,p})/t$; and
- ▶ in the intermediary case, more work is required.

Formally, ...

Main theorem

Theorem

For constant edge probability $0 < p < 1$, the following holds:

(a) *if $t(n) = o(\ln n)$, then $\chi^t(G_{n,p}) \sim \frac{n}{\gamma \ln n}$ a.a.s.;*

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For constant edge probability $0 < p < 1$, the following hold:

- (a) if $t(n) = o(\ln n)$, then $\chi^t(G_{n,p}) \sim \frac{n}{\gamma \ln n}$ a.a.s.;
- (c) if $t(n) = \omega(\ln n)$ and $t(n) = o(n)$, then $\chi^t(G_{n,p}) \sim \frac{np}{t}$ a.a.s.;
- (d) if $t(n)$ satisfies $\frac{np}{t} \rightarrow x$, where $0 < x < \infty$ and x is not integral, then $\chi^t(G_{n,p}) = \lceil x \rceil$ a.a.s.

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- (a) if $t(n) = o(\ln n)$, then $\chi^t(G_{n,p}) \sim \frac{n}{\gamma \ln n}$ a.a.s.;
- (b) if $t(n) = \Theta(\ln n)$, then $\chi^t(G_{n,p}) = \Theta(\frac{n}{\ln n})$ a.a.s.;
- (c) if $t(n) = \omega(\ln n)$ and $t(n) = o(n)$, then $\chi^t(G_{n,p}) \sim \frac{np}{t}$ a.a.s.;
- (d) if $t(n)$ satisfies $\frac{np}{t} \rightarrow x$, where $0 < x < \infty$ and x is not integral, then $\chi^t(G_{n,p}) = \lceil x \rceil$ a.a.s.

The t -dependence number

We bound a related parameter, the *t -dependence number* $\alpha^t(G)$ of G — the size of a largest t -dependent set in G .
Note: $\alpha^0(G) = \alpha(G)$.

Proposition

$$\chi^t(G) \geq \frac{|V(G)|}{\alpha^t(G)}.$$

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$$\chi^t(G_{n,p}) \geq \frac{n}{\alpha^t(G_{n,p})}.$$



Proof sketch: $t(n) = o(\ln n)$

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“ \geq ” uses $\chi^t \geq \frac{n}{\alpha^t}$ and a first moment estimate of α^t .

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- ▶ Let $k = k(n) = \lceil \frac{1}{1-\varepsilon} \gamma \ln n \rceil$ and let X be the number of t -dependent sets of size k in $G_{n,p}$.
- ▶ We show that $\mathbf{E}(X) \rightarrow 0$.

Proof sketch: $t(n) = o(\ln n)$

The crucial estimate is as follows:

- ▶ Let $g(k, t)$ be the number of graphs on $[k] = \{1, \dots, k\}$ with **average** degree at most t . The expected number of t -dependent k -sets is at most

$$\binom{n}{k} (1-p)^{\binom{k}{2} - \frac{tk}{2}} g(k, t)$$

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- ▶ Since a graph on k vertices with average degree d' has $kd'/2$ edges,

$$g(k, t) \leq \sum_{s=0}^{tk/2} \binom{\binom{k}{2}}{s}.$$

Proof sketch: $t(n) = \Omega(\ln n)$

Theorem

If $t(n) = \Theta(\ln n)$, then there exist constants $C, C' > 0$ such that $C \frac{n}{\ln n} \leq \chi^t(\mathbf{G}_{n,p}) \leq C' \frac{n}{\ln n}$ a.a.s.

Theorem

If $t(n) = \omega(\ln n)$ and $\varepsilon > 0$ fixed, then $(1 - \varepsilon) \frac{np}{t} \leq \chi^t(\mathbf{G}_{n,p}) \leq \lceil (1 + \varepsilon) \frac{np}{t} \rceil$ a.a.s.

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For both of these results,

“ \leq ” follows from $\chi^t \leq \lceil (\Delta + 1)/(t + 1) \rceil$ and

“ \geq ” the first moment estimate of α^t relies on passing from maximum to average degree as well as a Chernoff bound.



$$t(n) = \Theta(\ln n)$$

Suppose $t(n) \sim \tau \ln n$ and $k(n) = \kappa \ln n$ for constants $\tau, \kappa > 0$.

By large deviation techniques (cf. Dembo and Zeitouni (1998)), we can better estimate the expected number of t -dep. k -sets:

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Lemma

The expected number of *average* t -dependent k -sets is *at most*

$$\exp\left(k \ln n \left(1 - \frac{\kappa}{2} \Lambda^*\left(\frac{\tau}{\kappa}\right) + o(1)\right)\right)$$

where $\Lambda^*(x) = x \ln \frac{x}{\rho} + (1-x) \ln \frac{1-x}{1-\rho}$.

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The expected number of t -dependent k -sets is

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where $\Lambda^*(x) = x \ln \frac{x}{\rho} + (1-x) \ln \frac{1-x}{1-\rho}$.



$$t(n) = \Theta(\ln n)$$

- ▶ If $1 - \frac{\kappa}{2} \Lambda^* \left(\frac{\tau}{\kappa} \right) < 0$, then $\alpha^t(G_{n,p}) \leq \kappa \ln n$ and $\chi^t(G_{n,p}) \geq \frac{n}{\kappa \ln n}$ a.a.s.
- ▶ If $1 - \frac{\kappa}{2} \Lambda^* \left(\frac{\tau}{\kappa} \right) > 0$, then the expected number of t -dependent k -sets goes to infinity;

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- ▶ If $1 - \frac{\kappa}{2} \Lambda^* \left(\frac{\tau}{\kappa} \right) > 0$, then the expected number of t -dependent k -sets goes to infinity; *moreover*, setting $j(n) \sim \frac{n}{\kappa \ln n}$, the expected number of t -improper j -colourings goes to infinity.

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- ▶ If $1 - \frac{\kappa}{2} \Lambda^* \left(\frac{\tau}{\kappa} \right) < 0$, then $\alpha^t(G_{n,p}) \leq \kappa \ln n$ and $\chi^t(G_{n,p}) \geq \frac{n}{\kappa \ln n}$ a.a.s.
- ▶ If $1 - \frac{\kappa}{2} \Lambda^* \left(\frac{\tau}{\kappa} \right) > 0$, then the expected number of t -dependent k -sets goes to infinity; *moreover*, setting $j(n) \sim \frac{n}{\kappa \ln n}$, the expected number of t -improper j -colourings goes to infinity.

Conjecture

Let κ be the unique value satisfying $\kappa > \tau/p$ and $\frac{\kappa}{2} \Lambda^* \left(\frac{\tau}{\kappa} \right) = 1$.
Then $\chi^t(G_{n,p}) \sim \frac{n}{\kappa \ln n}$ a.a.s.

[It is routine to check that there exists a unique $\kappa > \tau/p$ such that $\frac{\kappa}{2} \Lambda^* \left(\frac{\tau}{\kappa} \right) = 1$.]

χ^t for sparse random graphs

Theorem (Łuczak, 1991)

Suppose $0 < p(n) < 1$ and $p(n) = o(1)$. Set $d(n) = np(n)$.

If $\varepsilon > 0$, *then there exists*
constant d_0 such that, if $d(n) \geq d_0$, then
 $(1 - \varepsilon) \frac{d}{2 \log d} \leq \chi(\mathbf{G}_{n,p}) \leq (1 + \varepsilon) \frac{d}{2 \log d}$ *a.a.s.*

χ^t for sparse random graphs

Theorem

Suppose $0 < p(n) < 1$ and $p(n) = o(1)$. Set $d(n) = np(n)$.

- (a) If $\varepsilon > 0$ and $t(n) = t_0$ for some fixed $t_0 \geq 0$, then there exists constant d_0 such that, if $d(n) \geq d_0$, then
- $$(1 - \varepsilon) \frac{d}{2^{\log d}} \leq \chi(G_{n,p}) \leq (1 + \varepsilon) \frac{d}{2^{\log d}} \text{ a.a.s.}$$
- (b) If $d(n) = \omega(1)$ and $t(n) = o(\ln d)$, then $\chi^t(G_{n,p}) \sim \frac{d}{2^{\ln d}}$ a.a.s.

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Suppose $0 < p(n) < 1$ and $p(n) = o(1)$. Set $d(n) = np(n)$.

- (a) If $\varepsilon > 0$ and $t(n) = t_0$ for some fixed $t_0 \geq 0$, then there exists constant d_0 such that, if $d(n) \geq d_0$, then $(1 - \varepsilon) \frac{d}{2^{\log d}} \leq \chi(G_{n,p}) \leq (1 + \varepsilon) \frac{d}{2^{\log d}}$ a.a.s.
- (b) If $d(n) = \omega(1)$ and $t(n) = o(\ln d)$, then $\chi^t(G_{n,p}) \sim \frac{d}{2^{\ln d}}$ a.a.s.

Furthermore, if $d(n) = \omega(\sqrt{\ln n})$, then the following hold:

- (c) if $t(n) = \Theta(\ln d)$, then $\chi^t(G_{n,p}) = \Theta\left(\frac{d}{\ln d}\right)$ a.a.s.;
- (d) if $t(n) = \omega(\ln d)$ and $t(n) = o(d)$, then $\chi^t(G_{n,p}) \sim \frac{d}{t}$ a.a.s.;
- (e) if $t(n)$ satisfies $\frac{d}{t} \rightarrow x$, where $0 < x < \infty$ and x is not integral, then $\chi^t(G_{n,p}) = \lceil x \rceil$ a.a.s.