# SHARP CONCENTRATION OF THE CHROMATIC NUMBER ON RANDOM GRAPHS $G_n$ ,

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The distribution of the chromatic number on random graphs  $G_{n,p}$  is quite sharply concentrated. For fixed p it concentrates almost surely in  $\sqrt{n} \omega(n)$  consecutive integers where  $\omega(n)$  approaches infinity arbitrarily slowly. If the average degree pn is less than  $n^{1/6}$ , it concentrates almost surely in five consecutive integers. Large deviation estimates for martingales are used in the proof.

### 1. Graphs

Given parameters n, p let  $G_{n,p}$  denote the random graph with vertex set  $[n] = \{1, ..., n\}$  and edge probability p. More precisely, let  $(\Omega, P)$  denote the discrete probability space with  $\Omega$  equal to the set of all graphs on [n] and probability P generated by

$$(1) P[\{i,j\} \in G] = p$$

these events mutually independent over all  $\{i,j\}\in[n]^2$ . For convenience we set

$$(2) d = pn$$

so that d is the average degree. We shall be particularly interested in the cases p constant and  $p=n^{-\alpha}$ ,  $0<\alpha<1$ , though our results are more general.

The chromatic number  $\chi(G)$  has been the object of much study. (See, e.g., [1-4, 7]). For constant p

(3) 
$$(1-o(1)) \frac{n \ln (1/(1-p))}{2 \ln n} \le \chi(G) \le (1+o(1)) \frac{n \ln (1/(1-p))}{\ln n}$$

while for  $p \rightarrow 0$ 

$$(4) \qquad (1-o(1))\frac{d}{2\ln d} \leq \chi(G) \leq (1+o(1))\frac{d}{\ln d}$$

almost surely. The factor of two between the bounds is most vexing. We do not here contribute directly to its elimination.

We shall show that the distribution of  $\chi(G)$  is tightly concentrated. We say a sequence X(G) with p=p(n) is concentrated in width s=s(n,p) if there exist u=u(n,p) so that

$$(5) P[u \le X(G) \le u + s] \to 1.$$

We show that s may be taken relatively small. It is both a strength and weakness of the method that u is not given explicitly.

We let  $\omega(n)$  denote a function of n approaching infinity arbitrarily slowly.

**Theorem 1.** For any  $p, \chi(G)$  is concentrated in width  $\sqrt{n} \omega(n)$ .

Theorem 2. Let  $p=n^{-\alpha}$ 

- (i) For  $0 < \alpha < 0.5$   $\chi(G)$  is concentrated in width  $s = n^{0.5 \alpha} (\ln n) \omega(n)$ (ii) For  $0.5 < \alpha < 1$   $\chi(G)$  is concentrated in width s where s is the minimal integer greater than  $(2\alpha+1)/(2\alpha-1)$ .

An extreme case occurs when  $\alpha > 5/6$ . Then five consecutive numbers suffice in that almost surely  $\chi(G) = u, u+1, u+2, u+3$  or u+4. This holds also for p = d(n)/n provided  $d(n) < n^{1/6}$ .

**Note.** By Chebyschev's Inequality any random variable X is concentrated in width  $\sqrt{\text{Var}(X)}\omega(n)$ . While using of this inequality, often called the second-moment method. has worked in some related questions it does not appear applicable to the study of the chromatic number.

Our basic tool is a martingale known as Doob's Martingale Process. (See, e.g., [5, 6].) In general, let  $X: \Omega \to \mathbb{R}$  be a random variable and let  $\equiv_i$ ,  $0 \le i \le n$ , be a sequence of equivalence relations on  $\Omega$  reverse ordered by refinement. (That is,  $G \equiv_{i+1} G'$  implies  $G \equiv_i G'$ .) Now define  $X_0, X_1, ..., X_n$  by

(6) 
$$X_i(G) = E[X(H)|H \equiv_i G].$$

This forms a margingale—i.e.

$$(7) E[X_{i+1}|X_i] = X_i.$$

In this work we define  $\equiv_i$  by setting  $G \equiv_i H$  if for all  $1 \le x \le i$ ,  $1 \le y \le n$ ,  $\{x,y\}\in G$  iff  $\{x,y\}\in H$ . Note that  $\equiv_0$  is universal so that

$$(8) X_0(G) = E[X(G)],$$

a constant, whereas  $\equiv_n$  is equality (not just isomorphism, these are labelled graphs!) so that

$$(9) X_n = X.$$

Thus  $X_i(G)$  is the expected value of X on the random graph whose first i vertices are precisely like G's and where all edges  $\{x, y\}$ ,  $i+1 \le x, y$ , are in the graph with probability p. The sequence  $X_0, ..., X_n$  moves from no information to full information in small steps.

The following result bounds large deviations in any martingale.

**Theorem 3.** Let  $X_0, ..., X_n$  be a martingale with  $X_0$  constant and  $|X_{i+1} - X_i| \le c$  for  $0 \le i \le n$  always. Then

(10) 
$$P[|X_n - X_0| > \lambda c n^{1/2}] < 2e^{-\lambda^2/2}.$$

**Lemma 4.** If E(Y)=0,  $|Y| \le 1$ ,  $\alpha > 0$  then  $E[e^{\alpha Y}] \le (1/2)(e^{\alpha} + e^{-\alpha}) < e^{\alpha^2/2}$ .

**Proof.** Set  $f(z) = e^{\alpha z}$ . If Y has values  $z_1, ..., z_s$  with probabilities  $p_1, ..., p_s$  then

(11) 
$$E[e^{\alpha Y}] = \sum_{i=1}^{s} p_i f(z_i) \leq \sum_{i=1}^{s} p_i \left[ \left( \frac{1+z_i}{2} \right) f(1) + \left( \frac{1-z_i}{2} \right) f(-1) \right] \quad (as \ f \ convex)$$

$$= \frac{f(1) + f(-1)}{2} < e^{\alpha^2/2}$$

by examining the Taylor series.

**Proof of Theorem 3.** Take c=1,  $X_0=0$  for convenience. Set  $Y_i=X_i-X_{i-1}$ ,  $1 \le i \le n$ . Let  $\alpha > 0$ . Now

(12) 
$$E[e^{\alpha X_i}] = E[e^{\alpha Y_i} e^{\alpha X_{i-1}}]$$

$$= E[E[e^{\alpha Y_i} | X_{i-1}] e^{\alpha X_{i-1}}]$$

$$\leq e^{\alpha^2/2} E[e^{\alpha X_{i-1}}]$$

as for any  $X_{i-1}$  the Lemma assures  $E[e^{\alpha Y_i}|X_{i-1}] \leq e^{\alpha^2/2}$ . By induction

$$(13) E[e^{\alpha X_n}] \leq e^{\alpha^2 n / 2}.$$

By Markov's Inequality

(14) 
$$P[X_n > \lambda \sqrt{n}] = P[e^{\alpha X_n} > e^{\alpha \lambda \sqrt{n}}] < e^{-\alpha \lambda \sqrt{n}} E[e^{\alpha X_n}]$$
$$= \exp[\alpha^2 n/2 - \alpha \lambda \sqrt{n}] = e^{-\lambda^2/2}$$

by setting  $\alpha = \lambda n^{-1/2}$ . By symmetry the absolute value gives at most a factor of two.

It will be convenient to define

$$I_i(G) = \{H \colon H \equiv_i G\}$$

(16) 
$$P^{l}(H) = \frac{P(H)}{\sum\limits_{J \in I_{l}(H)} P(J)}.$$

Here  $P^i$  represents a conditional probability with edges involving 1, ..., i fixed. Given G, H with  $G \equiv_i H$  we define a bijection

(17) 
$$\varphi \colon I_{i+1}(G) \to I_{i+1}(H)$$

by setting  $\varphi(J)=K$  if for all  $1 \le x < y \le n$  with  $x \ne i+1$ ,  $\{x,y\} \in J$  iff  $\{x,y\} \in K$ . That is,  $\varphi(J)$  is given by changing those edges  $\{i+1,y\}$ , i+1 < y, to conform to H. The bijection  $\varphi$  preserves conditional probability

(18) 
$$\varphi(J) = K \Rightarrow P^{i+1}(J) = P^{i+1}(K)$$

as  $P^{i+1}$  depends only on edges  $\{x, y\}$  with i+1 < x < y.

**Theorem 5.** Let  $X: \Omega \rightarrow \mathbb{R}$  be the chromatic number. Then

$$|X_{i+1}(G) - X_i(G)| \le 1.$$

An intuitive view may be useful.  $X_i(G)$  is the expected value of X(G) given knowledge of points 1, ..., i. Now we are told about point i+1. But a single point can only affect chromatic number by at most one!

Proof. Doing average of averages

(20) 
$$X_{i}(G) = \sum_{H \equiv_{i} G} X_{i+1}(H) P^{i}(H)$$
so
(21) 
$$X_{i+1}(G) - X_{i}(G) = \sum_{H \equiv_{i} G} [X_{i+1}(G) - X_{i+1}(H)] P^{i}(H).$$
Now
(22) 
$$X_{i+1}(G) = \sum_{J \equiv_{i+1} G} X(J) P^{i+1}(J)$$

(23) 
$$X_{i+1}(H) = \sum_{K \equiv_{i+1} H} X(K) P^{i+1}(K).$$

Employing the bijection  $\varphi$  and (18)

(24) 
$$X_{i+1}(H) - X_{i+1}(G) = \sum_{J \ge \dots, G} (X(\varphi J) - X(J)) P^{i+1}(J).$$

Suppose X(J)=c. We may c-color  $\varphi J - \{i+1\} = J - \{i+1\}$ . We then may (c+1)-color  $\varphi J$  using the extra color on i+1, if necessary. Hence  $X(\varphi J) \le c+1$ . By symmetry

$$(25) |X(\varphi J) - X(J)| \leq 1.$$

Applying this to (24)

$$|X_{i+1}(H) - X_{i+1}(G)| \le 1$$

and applying this to (21)

$$|X_{i+1}(G) - X_i(G)| \le 1$$

as desired.

**Proof of Theorem 1.** Let p arbitrary,  $\lambda > 0$ . Theorem 3 gives

(28) 
$$P[X(G) - E[X(G)] > \lambda \sqrt{n}] < 2e^{-\lambda^2/2}.$$

**Note.** A concentration result similar to Theorem 1 can be obtained for a random function X whenever  $X(G) - X(\varphi G)$  is small.

#### 2. Sparse graphs

While Theorem 1 holds for all p its usefulness declines as p decreases in view of the known bounds (3). For sparse graphs we modify the function X. We set  $p = n^{-\alpha}$  with  $0 < \alpha < 1$ .

Let  $\varepsilon > 0$  be arbitrarily small and fixed. Let u = u(n, p) be the minimal integer such that

(29) 
$$P[\chi(G) \leq u] > \varepsilon.$$

From (4)

(30) 
$$u > \frac{(1 - o(1))d}{2 \ln d}.$$

Our object is to show that with s given by Theorem 2

$$(31) P[\gamma(G) \le u+s] \to 1.$$

For any G and any map  $\psi: [n] \rightarrow [u]$  we let flaws $(G, \psi)$  denote the number of edges  $\{i, j\} \in G$  with  $\psi(i) = \psi(j)$ . Set

(32) 
$$X(G) = \min \text{ flaws } (G, \psi),$$

the minimum over all  $\psi$ :  $[n] \rightarrow [u]$ . Then X(G) = 0 if and only if  $\chi(G) \le u$  so

$$(33) P[X \le u] \ge \varepsilon.$$

Let  $X_0, X_1, ..., X_n$  denote the Doob Martingale generated by X. Let  $\deg(G, i)$  denote the degree of point i in graph G. Call G normal if

(34) 
$$\deg(G, i) \leq 5d$$
, all  $i \in G$ .

**Theorem 6.** If G is normal

$$|X_{i+1}(G) - X_i(G)| \le 11d/u.$$

**Proof.** Equations (21), (24) and bijection  $\varphi$  are as before. We must bound  $|X(J)-X(\varphi J)|$ , noting that J,  $\varphi J$  differ only at point i+1 where they are identical to G and H respectively. Let  $\psi$  be the coloring of J with X(J) flaws. Deleting i+1,  $\psi$  is a coloring of  $J-\{i+1\}=\varphi J-\{i+1\}$  with at most X(J) flaws. Point i+1 is adjacent to  $\deg(H,i+1)$  other vertices in J. Color i+1 so that at most  $\deg(H,i+1)/u$  of these points have that color. This gives a  $\psi^*$  with

(36) 
$$X(\varphi J) \leq \text{flaws}(\varphi J, \psi^*) \leq X(J) + [\deg(H, i+1)/u].$$

Similarly

$$(37) X(J) \leq X(\varphi J) + [\deg(G, i+1)/u]$$

and thus

(38) 
$$|X(\varphi J) - X(J)| \leq u^{-1} [\deg(G, i+1) + \deg(H, i+1)].$$

As this holds for all J = I + I G, (24) gives

$$|X_{i+1}(J) - X_{i+1}(G)| \le u^{-1} [\deg(G, i+1) + \deg(H, i+1)]$$

and now (21) gives

(40)

$$|X_{i+1}(G) - X_i(G)| \le u^{-1} \Big[ \sum_{H \equiv_i G} \deg(G, i+1) P^i(H) + \sum_{H \equiv_i G} \deg(H, i+1) P^i(H) \Big].$$

The first sum is simply  $\deg(G, i+1) \le 5d$ . The second sum is the conditional expected value

(41) 
$$E[\deg(H, i+1)|H \equiv_i G]$$

which is the number of  $j \le i$  with  $\{j, i+1\} \in G$  plus p(n-i-1) (i.e., the random edges) so at most  $\deg(G, i+1) + pn \leq 6pn$ . Together

$$|X_{i+1}(G) - X_i(G)| \le 11 pn/u = 11 d/u. \quad \blacksquare$$

Most G are normal and for these G,  $X_{i+1}-X_i$  is never large. We modify Theorem 3 to allow for a small probability of a large change. In application the r below shall be extremely small.

**Theorem 7.** Let  $X_0, ..., X_n$  be a martingale with  $X_0$  constant and such that

- (i) With probability  $\geq 1-r$ ,  $|X_{i+1}-X_i| \leq c$  for  $0 \leq i < n$ .
- (ii) Always  $|X_{i+1} X_i| \le n$ . Suppose  $nr^{1/2} \le c$ . Then

(43) 
$$P[|X_n - X_0| > (2c)\lambda n^{1/2} + n^2 r^{1/2}] < 2e^{-\lambda^2/2} + 2nr^{1/2}.$$

**Proof.** Let  $F_i$  be the event " $|X_{i+1}-X_i| > c$ ". We define a new martingale  $Y_0, \ldots$ ...,  $Y_n$ . Set  $Y_0 = X_0$ . Assume  $Y_i$  has been defined and let  $p = P[F_i|X_i]$ .

Case 1.  $p \ge r^{1/2}$ . We "terminate" the martingale by setting  $Y_{i+1} = Y_i$  and  $Y_i = Y_i$ for all  $j, i+1 \le j \le n$ .

Case 2.  $p < r^{1/2}$  and the martingale has not been previously terminated. Define

$$X_{i+1}^{\bullet} = \begin{cases} X_i & \text{if } F_i \\ X_{i+1} & \text{if } \neg F_i. \end{cases}$$

Then

(45) 
$$E[X_{i+1}^*|X_i] = E[X_{i+1}|X_i] + E[X_{i+1}^* - X_{i+1}|X_i] = X_i + A_i$$

where  $A_i = E[X_i - X_{i+1}|X_i, F_i] \cdot P[F_i|X_i]$  so that

(46) 
$$|A_i| \leq E[|X_i - X_{i+1}||X_i, F_i] P[F_i|X_i] \leq nr^{1/2}$$

by (ii) and the assumption of Case 2. Finally set

(47) 
$$Y_{i+1} = Y_i + [X_{i+1}^* - X_i - A_i].$$

The expectation  $Y_{i+1} - Y_i$  conditioned upon  $Y_i$  (which is determined by  $X_{\leq i}$ ) is zero, in view of Case 1 and (45). Thus  $Y_0, ..., Y_n$  is indeed a martingale. Moreover

$$(48) |Y_{i+1}-Y_i| \le c+A_i \le 2c always.$$

[The gambling analogy may be useful here. A gambler plays a series of fair games and  $X_{i+1}-X_i$  is the amount he wins in the (i+1)-st game. If he is about to play a "wild" game (with  $P[|X_{i+1}-X_i|>c]>r^{1/2}$ ) then we adjust by quitting. If the game is not wild we change all big wins and losses  $(|X_{i+1}-X_i|>c)$  to zero  $(X_{i+1}^*=X_i)$  and making an adjustment  $A_i$  on all payoffs to keep the game fair.] Call G supernormal if Case 1 never occurs and  $F_i$  never occurs. For supernormal G

$$(49) Y_n = X_n - (A_0 + \dots + A_{n-1})$$

so that

$$|Y_n - X_n| \le n^2 r^{1/2}.$$

We bound

(51) 
$$P[G \text{ is not supernormal}] \leq \sum_{i=1}^{n} P[F_i] + \sum_{i=1}^{n} P[P[F_i|X_i] > r^{1/2}].$$

Now

(52) 
$$r \ge P[F_i] \ge r^{1/2} P[P[F_i|X_i] > r^{1/2}]$$

SO

(53) 
$$P[P[F_i|X_i] > r^{1/2}] < r^{1/2}$$

and

(54) 
$$P[G \text{ is not supernormal}] \leq nr + nr^{1/2} \leq 2nr^{1/2}.$$

We apply Theorem 3 to the martingale  $Y_0, ..., Y_n$ . Using (48),

(55) 
$$P[|Y_n - Y_0| > (2c)\lambda n^{1/2}] < 2e^{-\lambda^2/2}.$$

Now  $X_0 = Y_0$  so using (50)

$$P[|X_n - X_0| > (2c)\lambda n^{1/2} + 2n^2 r^{1/2}] \ll$$

(56) 
$$\ll P[G \text{ is not supernormal}] + P[|Y_n - Y_0| > 2c\lambda n^{1/2}] < 2e^{-\lambda^2/2} + 2nr^{1/2}$$

as desired.

In our case

(57) 
$$c = 11d/n \le 23 \ln d \le 23 \ln n$$

(58) 
$$r = P[G \text{ is not normal}] \le nP[\deg(G, i) \ge 5pn].$$

But  $\deg(G, i)$  has Binomial Distribution  $Z \sim B(n, p)$ . We use the result

(59) 
$$P[Z > pn + K] < e^{-K^2/2pn}.$$

With k=4pn

(60) 
$$r \le n e^{-8pn} = n e^{-8n^{1-\alpha}}$$

with  $p=n^{-\alpha}$ . As r is so small  $3n^3r^{1/2}\ll 1$  and so, as  $X_n=X$  and  $X_0=E(X)$ 

(61) 
$$P[|X-EX| > \omega(n)n^{1/2} \ln n] \to 0.$$

Combining (61) and the defining property (33) of u:

$$(62) EX < n^{1/2} (\ln n) \omega(n).$$

Applying (61) once more in the opposite direction

(63) 
$$P[X > 2n^{1/2}(\ln n)\omega(n)] \to 0.$$

That is, almost surely G may be u-colored with less than  $n^{1/2}(\ln n)\omega(n)$  flaws.

**Lemma 8.** Almost surely every subset of G with  $k \le n^{1/2}(\ln n)\omega(n)$  vertices may be s-colored where s is given by Theorem 2.

**Proof.** If G does not have this property there will be a minimal set V of vertices which is not s-colorable with  $|V| = k < n^{1/2}(\ln n)\omega(n)$ . Then each  $x \in V$  must be degree in V of a least s-1 as otherwise  $V - \{x\}$  may be s-colored and then x can be given one of these colors. Hence V must have at least k(s-1)/2 edges. Then

(64) 
$$P[G \text{ does not have property}] < \sum {n \choose k} \left( \frac{k \choose 2}{\frac{k(s-1)}{2}} \right) p^{k(s-1)/2}.$$

The sum is over all  $k < n^{1/2}(\ln n)\omega(n)$ . The factors are the number of k-sets, the choices for the k(s-1)/2 edges, and the probability that those edges are in G. We bound

(65) 
$$\binom{n}{k} < \binom{ne}{k}^k, \left( \frac{\binom{k}{2}}{2} \right) \sim \left[ \frac{k}{s-1} \right]^{k(s-1)/2}.$$

If s is such that

$$(66) \qquad \frac{ne}{k} \left[ \frac{kp}{s-1} \right]^{(s-1)/2} < 1$$

then the sum will be small. It suffices to have  $k=n^{1/2+o(1)}$ , i.e.,

(67) 
$$n^{1/2+o(1)} \left[ \frac{n^{1/2+o(1)}p}{s-1} \right]^{(s-1)/2} < 1.$$

For  $p=n^{-\alpha}$  with  $\alpha < 1/2$ , (67) holds with  $s=n^{1/2-\alpha+o(1)}$ , or, more precisely, with  $s=n^{1/2}(\ln n)\omega(n)p$ . For  $p=n^{-\alpha}$  with  $\alpha > 1/2$ , (67) gives that:  $(1/2-\alpha)((s-1)/2)+1/2<0$  and the minimal integer s with that property will suffice.

What about  $\chi(G)$ . Almost surely G can be u-colored with  $\leq n^{1/2}(\ln n)\omega(n)$  flaws. Let V be the set of vertices in flaws so that G-V is u-colored. Now V can be s-colored, together this gives a (u+s)-coloring of G, completing the proof of Theorem 2.

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#### References

- [1] B. Bollobás and P. Erdős, Cliques in random graphs, Math. Proc. Cambridge Phil. Soc., 80 (1976). 419—427.
- [2] W. F. DE LA VEGA, On the Chromatic number of sparge random graphs, in: Graph theory and combinatorics (ed. B. Bollobás) Academic Press, London, 1984, 321—328.
- [3] P. Erdős and J. Spencer, *Probabilistic Methods in Combinatorics*, Academic Press, New York, 1974.
- [4] G. GRIMMET and C. McDiarmid, On coloring random graphs, *Math. Proc. Cambridge Phil. Soc.*, 77 (1985), 313—324.
- [5] S. Karlin and H. Taylor, First Course in Stochastic Processes, 2nd ed., Academic Press, New York, 1975.
- [6] V. MILMAN and G. SCHECHTMAN, Asymptotic theory of normed linear spaces, Lecture Notes in Mathematics, Springer.
- [7] E. Shamr and R. Uffal, Sequential and distributed graph coloring algorithms with performance analyses in random graph spaces. *J. of Algorithms*, 5 (1984), 488—501.

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