On the Concentration of the Chromatic Number of Random Graphs

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Abstract

Shamir and Spencer proved in the 1980s that the chromatic number of the binomial random graph $G_{n,p}$ is concentrated in an interval of length at most $\omega\sqrt{n}$, and in the 1990s Alon showed that an interval of length $\omega\sqrt{n}/\log n$ suffices for constant edge-probabilities $p \in (0, 1)$. We prove a similar logarithmic improvement of the Shamir-Spencer concentration results for the sparse case $p = p(n) \to 0$, and uncover a surprising concentration 'jump' of the chromatic number in the very dense case $p = p(n) \to 1$.

Mathematics Subject Classifications: 05C15, 05C80

1 Introduction

What can we say about the chromatic number $\chi(G_{n,p})$ of an *n*-vertex binomial random graph $G_{n,p}$? From a combinatorial perspective, it is natural to ask about the typical value of $\chi(G_{n,p})$, i.e., upper and lower bounds that are close to each other. From a probabilistic perspective, it is also natural to ask about the concentration of $\chi(G_{n,p})$, i.e., how much this random variable varies. Among these two fundamental questions, significantly less is known about the concentration question that we shall study in this paper.

In a landmark paper from 1987, Shamir and Spencer [26] proved that the chromatic number $\chi(G_{n,p})$ is typically contained in an interval of length at most $\omega\sqrt{n}$, where $\omega = \omega(n)$ is an arbitrary function with $\omega \to \infty$ as $n \to \infty$, as usual. For constant edge-probabilities $p \in (0, 1)$, Alon noticed in the 1990s that this concentration interval length can be slightly improved to $\omega\sqrt{n}/\log n$, by adapting a coloring argument of Bollobás [4], see [3, Excercise 7.9.3] and Scott's note [25]. For uniform edgeprobability p = 1/2, Heckel and Riordan [9] proved in 2021 that these old concentration bounds are in fact best possible to up poly-logarithmic factors.

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Given the increasing knowledge about the concentration of $\chi(G_{n,p})$ in the dense case where $p \in (0, 1)$ is constant, it is important to clarify our understanding of the sparse case where $p = p(n) \to 0$ vanishes as $n \to \infty$. For edge-probabilities of form $p = n^{-\alpha}$ with $\alpha \in (0, 1/2)$, Shamir and Spencer proved in their 1987 paper that $\chi(G_{n,p})$ is typically contained in an interval of length at most $\omega \sqrt{np} \log n$, and a modern inspection of their proof reveals that length $\omega \sqrt{np}$ suffices. With Alon's improvement for constant p in mind, it is natural to wonder if further improvements of these sparse concentration bounds are possible.

In this paper we sharpen the concentration of $\chi(G_{n,p})$ in the sparse case $p = p(n) \to 0$, by extending Alon's logarithmic improvement to smaller edge-probabilities: Theorem 1 improves the Shamir-Spencer bound from 1987, by showing that $\chi(G_{n,p})$ is typically contained in an interval of length at most $\omega \sqrt{np}/\log n$.

Theorem 1 (Improved concentration bound). Let $\omega = \omega(n) \to \infty$ as $n \to \infty$ be an arbitrary function, and let $\delta \in (0, 1)$ be a constant. If the edge-probability p = p(n) satisfies $n^{-1/2+\delta} \ll p \leq 1-\delta$, then there is an interval of length $\lfloor \omega \sqrt{np} / \log n \rfloor$ that contains the chromatic number $\chi(G_{n,p})$ of the random graph $G_{n,p}$ with high probability, i.e., with probability tending to one as $n \to \infty$.

Our proof of Theorem 1 refines the basic ideas of Shamir, Spencer and Alon using two greedy algorithms, which enable us to bypass large deviation inequalities such as Janson's inequality via more robust Chernoff bound based arguments. Note that the concentration bound of length $\lfloor \omega \sqrt{n}p / \log n \rfloor$ is meaningful, since typically $\chi(G_{n,p}) = \Theta(np/\log(np))$ holds [4, 16, 12]. Furthermore, the restriction to $p \gg n^{-1/2+\delta}$ is close to best possible, since for $n^{-1} \ll p \ll n^{-1/2-\delta}$ the chromatic number is concentrated on two different values [17, 2, 1], whereas the bound from Theorem 1 would imply one-point concentration; see Section 2 for our more general chromatic number concentration bounds for other ranges of edge-probabilities p = p(n).

In this paper we also uncover a surprising concentration behavior of $\chi(G_{n,p})$ in the very dense case where $p = p(n) \to 1$ tends to one as $n \to \infty$: Theorem 2 shows that the typical length of the shortest interval containing $\chi(G_{n,p})$ undergoes a polynomial 'jump' around edge-probability $p = 1 - n^{-1+o(1)}$; see Figure 1.

Theorem 2 (Concentration 'jump' in the very dense case). Given $\epsilon > 0$, the following holds for the random graph $G_{n,p}$ with edge-probability p = p(n), setting $\varphi = \varphi(n,p) := n(1-p)$.

- (i) If $n^{-o(1)} \leq \varphi \ll \log n$, then no interval of length $\lfloor n^{1/2-\epsilon} \rfloor$ contains $\chi(G_{n,p})$ with high probability.
- (ii) If $\log n \ll \varphi \leqslant n^{o(1)}$, then there is an interval of length $\lfloor n^{\epsilon} \rfloor$ that contains $\chi(G_{n,p})$ with high probability.

The concentration bounds (i)–(ii) demonstrate that the typical length of the shortest interval containing $\chi(G_{n,p})$ is neither monotone nor smooth in p, which both are surprising features. We believe that the intriguing concentration jump of $\chi(G_{n,p})$ described by Theorem 2 happens infinitely many times as we vary the edge-probability $p = 1 - n^{-\Omega(1)}$; see Section 3 and Figure 1 for more details and further results.

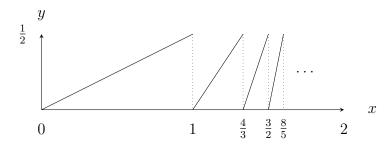


Figure 1: The exponent of the concentration interval length of very dense random graphs $G_{n,p}$: when $n^2(1-p) = n^{x+o(1)}$ with $x \in (0,2)$, then Conjecture 8 predicts that $n^{y+o(1)}$ with $y = y(x) \in [0, 1/2]$ is the length of the shortest interval that contains $\chi(G_{n,p})$ with high probability. Interestingly, this proposes that the concentration interval length of $\chi(G_{n,p})$ has infinitely many polynomial 'jumps' from $n^{1/2+o(1)}$ to $n^{o(1)}$ as we vary the edge-probability $p = 1 - n^{-2+x+o(1)}$; see Theorem 2 and Section 3 for more details.

2 Concentration bounds for $\chi(G_{n,p})$

The following concentration result for the chromatic number $\chi(G_{n,p})$ generalizes Theorem 1, by removing the assumed lower bound on the edge-probability p = p(n). Interestingly, the form of the concentration bounds (1a)–(1b) changes when $\omega\sqrt{np}$ is around $\log n$ (here we lose nothing by formally ignoring the case $\omega\sqrt{np} = \Theta(\log n)$, since in that case we can then simply apply (1b) after replacing ω with $\sqrt{\omega}$, say). Note that $L = \Theta(zp/\log(zp))$ when $zp \gg (\log n)^{1+\epsilon}$, so that Theorem 3 implies Theorem 1 by rescaling ω .

Theorem 3. Let $\omega = \omega(n) \to \infty$ as $n \to \infty$ be an arbitrary function, and let $\gamma \in (0, 1)$ be a constant. If the edge-probability p = p(n) of the random graph $G_{n,p}$ satisfies 0 , $then there is an interval of length at most <math>L = L(n, p, \gamma)$ that contains the chromatic number $\chi(G_{n,p})$ with high probability, with

$$L := \begin{cases} \frac{Czp}{\log(zp/\log n)} & \text{if } zp \gg \log n, \end{cases}$$
(1a)

$$= \begin{cases} \frac{C\log n}{\log(\log n/zp)} & \text{if } zp \ll \log n, \end{cases}$$
 (1b)

where $C = C(\gamma) > 0$ is a constant and $z = z(n, \omega)$ is defined as

$$z = z(n,\omega) := \omega \sqrt{n}.$$
(2)

Theorem 3 implies that $\chi(G_{n,p})$ is typically contained in an interval of length $O(\omega\sqrt{n}/\log n)$ for constant $p \in (0, 1)$, and in an interval of length O(1) for $p \leq n^{-1/2-\delta}$. These bounds match the best-known upper bounds up to constant factors [26, 2, 25], and the best-known lower bounds up to poly-logarithmic factors [2, 9]. For $n^{-1/2} \leq p \ll 1$ we leave it as an interesting open problem whether the concentration bounds (1a)–(1b) for $\chi(G_{n,p})$ are close to best possible or not. The starting point for Theorems 1 and 3 as well as many earlier concentration proofs [26, 17, 15, 2, 25] is an observation about the chromatic number $\chi(G_{n,p})$ that can be traced back to Shamir and Spencer [26], which intuitively says (see Lemma 5) that with high probability the median of $\chi(G_{n,p})$ satisfies

$$\left|\chi(G_{n,p}) - \operatorname{Median}\left(\chi(G_{n,p})\right)\right| \leq \max_{Z \subseteq [n]: |Z| \leq z} \chi(G_{n,p}[Z]),$$
(3)

where z is defined as in (2). The induced subgraph $G_{n,p}[Z]$ has the same distribution as $G_{|Z|,p}$, so for suitable ranges of p we know [16, 12] that for a *fixed* vertex-subset $Z \subseteq [n]$ with $|Z| \approx z$ we typically have $\chi(G_{n,p}[Z]) = \Theta(zp/\log(zp))$, which suggests that Theorem 1 and (1a) are more or less the best concentration bounds we can deduce from (3).

Alon noted that for constant $p \in (0, 1)$ one can bound (3) by adapting Bollobás' analysis [4] of $\chi(G_{n,p})$ to $\chi(G_{n,p}[Z])$ for all vertex-subsets $Z \subseteq [n]$ with $|Z| \leq z$. Here the main probabilistic ingredient are large deviation inequalities such as Janson's inequality [11, 22]: these allow us to show that, with high probability, all vertex-subsets of $G_{n,p}$ with $m \approx \sqrt{n}/(\log n)$ vertices contain an independent set on $k = \Theta(\log n/p)$ vertices. To color any $G_{n,p}[Z]$, we can thus iteratively remove a largest independent set from Z and assign its vertices one new color, until at most m uncolored vertices remain, which then each obtain a new color. For constant $p \in (0, 1)$ this yields $\chi(G_{n,p}[Z]) \leq |Z|/k + m = O(zp/\log n)$ for all relevant $Z \subseteq [n]$, which together with (3) recovers the $\omega\sqrt{n}p/\log n$ concentration bound from Theorem 1 up to irrelevant constant factors.

On first sight one might think that an extra twist can extend the outlined coloring argument for $\chi(G_{n,p}[Z])$ to the sparse case $p = p(n) \to 0$ considered by Theorem 1: indeed, a simple greedy algorithm can color the remaining *m* vertices with only O(mp) colors, so for any relevant $Z \subseteq [n]$ we should overall only need

$$\chi(G_{n,p}[Z]) \leqslant \frac{|Z|}{k} + O(mp) = O\left(\frac{zp}{\log n}\right)$$
(4)

colors, which together with (3) seemingly recovers the $\omega\sqrt{np}/\log n$ concentration bound from Theorem 1 up to constant factors. Unfortunately, there is another major bottleneck we inherited from Bollobás' analysis of $\chi(G_{n,p})$: due to union bound issues¹ large deviation inequalities can only guarantee k-vertex independent sets in every m-vertex subset as long as $p \ge n^{-\sigma+o(1)}$ for some small $\sigma > 0$. To extend the range of p = p(n) we borrow ideas from Grimmett and McDiarmid's earlier analysis [6] of $\chi(G_{n,p})$ from 1975, and make Chernoff bounds the main probabilistic ingredient: these allow us to show that, with high probability, all large subgraphs of $G_{n,p}$ contain a vertex whose degree is small relative to the size of the subgraph, which enables a simple greedy algorithm to find independent sets of size $\Theta(\log n/p)$ in any m-vertex subset of $G_{n,p}$ (this was also used by Scott [25] for

¹Using large deviation inequalities inequalities the issue is that, well before $p = n^{-1/2+\delta}$, the probability that one *m*-vertex subset does not contain a *k*-vertex independent set is no longer small enough to take a union bond over all *m*-vertex subsets (no matter if ones uses Janson's inequality [11, 22], Talagrand's inequality [21], or the bounded differences inequality [19, 27]); this issue is also the reason why Bollobás' analysis of $\chi(G_{n,p})$ breaks around $p = n^{-1/3}$, see [14, Section 3] and [12, Section 7.5].

constant p). We find it surprising that a combination of the two discussed simple greedy based refinements not only yields (4) and thus Theorem 1, but also establishes the more general Theorem 3 by a refined analysis; see Section 2.1 for the details.

In our discussion of the concentration of $\chi(G_{n,p})$ we so far tacitly ignored the very dense case $p = p(n) \to 1$: this conveniently allowed us to work with independent sets of size $\Theta(\log n/p)$ instead of size $\Theta(\log_{1/(1-p)}(n))$, which in turn allowed us to write $\Theta(p)$ instead of $\log(1/(1-p)) = \sum_{k \ge 1} p^k/k$ in the numerators appearing in Theorem 1 as well as equations (1a) and (4). Theorem 4 effectively says that we only need to reverse this simplification in order to extend Theorems 1 and 3 to edge-probabilities $p = p(n) \to 1$ that tend to one.

Theorem 4. Let $\omega = \omega(n) \to \infty$ as $n \to \infty$ be an arbitrary function, and let $\gamma \in (0, 1)$ be a constant. If the edge-probability p = p(n) of the random graph $G_{n,p}$ satisfies $\gamma \leq p < 1$, then there is an interval of length $\lfloor C\omega\sqrt{n}/\log_{1/(1-p)}(n) \rfloor$ that contains $\chi(G_{n,p})$ with high probability, where $C = C(\gamma) > 0$ is a constant.

This concentration bound is meaningful, since typically $\chi(G_{n,p}) = \Theta(n/\log_{1/(1-p)}(n))$ in the assumed range of p; see Section 3.3. For $p = 1 - n^{-\Theta(1)}$ the bound of Theorem 4 effectively reduces to the $\omega\sqrt{n}$ concentration bound of Shamir and Spencer, which in fact is best possible for $p = 1 - \Theta(n^{-1})$ due to Theorem 9 and the earlier work of Alon and Krivelevich [2]. Given any integer $r \ge 2$, we similarly believe that no further concentration improvements are possible for $p = 1 - \Theta(n^{-2/r})$; see Figure 1 and Conjecture 8. In Section 3 we discuss in more detail the behavior of the chromatic number $\chi(G_{n,p})$ when $p = 1 - n^{-\Omega(1)}$.

2.1 Proofs of Theorem 3 and 4

As discussed, the starting point for our proofs of Theorem 3 and 4 is the following useful observation about the concentration of $\chi(G_{n,p})$. The intuition behind Lemma 5 is that, after removing at most $z = \omega \sqrt{n}$ vertices from $G_{n,p}$, we can color the remaining vertices of $G_{n,p}$ with about M colors, where the median M = M(n,p) of $\chi(G_{n,p})$ does not depend on z or ω (which is a non-standard feature: usually a different function M is used that explicitly depends on z or ω); we defer the concentration-based proof to Appendix 4.

Lemma 5 (Chromatic number: concentration around median). For any $p = p(n) \in [0, 1]$ and $z = z(n, \omega)$ as in (2), the following holds for the random graph $G_{n,p}$. If there is a function $\Gamma = \Gamma(z, n, p)$ for which the event

$$\max_{Z\subseteq[n]:|Z|\leqslant z}\chi(G_{n,p}[Z])\leqslant\Gamma$$
(5)

holds with high probability, then with high probability we also have

$$\left|\chi(G_{n,p}) - \mathcal{M}\right| \leqslant \Gamma,\tag{6}$$

where M = M(n, p) is defined as the smallest integer with $\mathbb{P}(\chi(G_{n,p}) \leq M) \geq 1/2$.

To show concentration of $\chi(G_{n,p})$, it thus suffices to show that we can color any induced subgraph $G_{n,p}[Z]$ on $|Z| \leq z$ vertices with few colors. To efficiently color $G_{n,p}[Z]$ we exploit local sparsity by combining standard Chernoff bounds with the following two greedy-based results, whose routine proofs we defer to Appendix 4. In brief, Lemma 6 enables us to iteratively remove a large independent set from Z and assign its vertices one new color, until few vertices of Z remain, which we then color using Lemma 7. Below we write $\delta(G)$ and $\alpha(G)$ for the minimum degree of G and the size of the largest independent set of G, respectively.

Lemma 6 (Large independent sets: greedy bound). Given a graph G and parameters 0 < d < 1 < u, assume that the minimum degree satisfies $\delta(G[S]) \leq d(|S|-1)$ for all vertex-subsets $S \subseteq V(G)$ of size $|S| \geq u$. Then $\alpha(G[W]) \geq -\log_{(1-d)(1-1/u)}(|W|/u)$ for any vertex-subset $W \subseteq V(G)$ of size $|W| \geq u$.

Lemma 7 (Chromatic number: greedy bound). Given a graph G and a parameter $r \ge 0$, assume that the minimum degree satisfies $\delta(G[S]) \le r$ for all vertex-subsets $S \subseteq V(G)$. Then $\chi(G) \le r+1$.

In the remainder of this section we prove the concentration bounds of Theorems 3–4 by combining Lemma 5 with variants of the above-discussed two-phase greedy coloring argument for $G_{n,p}[Z]$.

2.1.1 Proof of Theorem 3 when $zp \gg \log n$

In the case $zp \gg \log n$ of Theorem 3, our proof strategy uses Chernoff bounds to color $G_{n,p}[Z]$ as follows. By iteratively applying Lemma 6 with $u = \Theta(\log n/p)$ and $d \approx p \gg 1/u$ to extract large independent sets from $G_{n,p}[Z]$, in the first phase the idea is to color all but O(u) many vertices of Z using $O(zp/\log(zp/\log n))$ colors, see (11) below for the details (which take into account that the largest independent sets get smaller as fewer vertices of Z remain). By applying Lemma 7 with $r = \Theta(up)$, in the second phase then idea is to then color the remaining O(u) uncolored vertices using only $r + 1 = \Theta(\log n) \ll zp/\log(zp/\log n)$ additional colors.

Proof of Theorem 3 when $zp \gg \log n$. Set $\epsilon := (1 - \gamma)/(1 + \gamma)$, $d := (1 + \epsilon)p$ as well as $u := 24\epsilon^{-2}(\log n)/p$ and $r = \lceil 2e^2up \rceil$. Note that $d < 1 \ll u$. With an eye on the minimum degree based Lemmas 6–7, let \mathcal{D} denote the event that $\delta(G_{n,p}[S]) \leq d(|S| - 1)$ for all $S \subseteq [n]$ with $|S| \geq u$, and let \mathcal{E} denote the event that $\delta(G_{n,p}[S]) \leq 2r - 1$ for all $S \subseteq [n]$ with $|S| \leq 4u$. Note that $\neg \mathcal{D}$ implies existence of $S \subseteq [n]$ with $|S| \geq u$ such that $G_{n,p}[S]$ has at least $(1 + \epsilon)p\binom{|S|}{2}$ edges. Using standard Chernoff bounds (such as [12, Theorem 2.1]) and $\epsilon^2 up/12 = 2\log n$, a routine union bound argument shows that

$$\mathbb{P}(\neg \mathcal{D}) \leqslant \sum_{u \leqslant s \leqslant n} \binom{n}{s} e^{-\epsilon^2 \binom{s}{2}p/3} \leqslant \sum_{u \leqslant s \leqslant n} \left(n e^{-\epsilon^2 u p/12}\right)^s \leqslant \sum_{s \geqslant u} n^{-s} = o(1).$$
(7)

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Exploiting $r \ge 2e^2up > 2\log n$, we similarly see that

$$\mathbb{P}(\neg \mathcal{E}) \leq \sum_{1 \leq s \leq 4u} \binom{n}{s} \binom{\binom{s}{2}}{sr} p^{sr} \\
\leq \sum_{1 \leq s \leq 4u} \left[n \left(\frac{4upe}{2r} \right)^r \right]^s \leq \sum_{s \geq 1} \left(ne^{-r} \right)^s \leq \sum_{s \geq 1} n^{-s} = o(1).$$
(8)

We henceforth assume that the events \mathcal{D} and \mathcal{E} both hold. Consequently, if a vertex subset $W \subseteq [n]$ has size at least $|W| \ge 2u$, then using Lemma 6 as well as $1/u \ll d =$ $(1+\epsilon)p \le (1+\epsilon)\gamma$ and $-\log(1-d) \le d/(1-d)$ it follows that the induced subgraph $G_{n,p}[W]$ contains an independent set of size at least

$$\alpha\left(G_{n,p}[W]\right) \ge -\log_{(1-d)(1-1/u)}\left(|W|/u\right) \ge \frac{\log\left(|W|/u\right)}{-2\log(1-d)} \ge \frac{\log\left(|W|/u\right)}{Ap} =: I\left(|W|\right) \quad (9)$$

for a suitable constant $A = A(\epsilon, \gamma) \in (0, \infty)$. In order to color any vertex subset $Z \subseteq [n]$ of size at most $|Z| \leq z$, among the so-far uncolored vertices of Z we iteratively choose a largest independent set and assign its vertices one new color, until a set $R \subseteq Z$ of at most $|R| \leq 4u$ uncolored vertices remains. Applying Lemma 7 to $G_{n,p}[R]$, in view of the event \mathcal{E} and $zp/\log n \gg 1$ we then color R using at most

$$\chi(G_{n,p}[R]) \leq 2r = \Theta(\epsilon^{-2}\log n) \ll \frac{zp}{\log(zp/\log n)}$$
 (10)

many colors. To bound the number of colors used for $Z \setminus R$, for a fixed integer $i \ge 0$ we shall first bound the number of independent sets chosen while the number of uncolored vertices from Z is between $z2^{-(i+1)}$ and $z2^{-i}$, and then sum over all feasible integers $i \ge 0$. Namely, after recalling $u = \Theta(\epsilon^{-2}(\log n)/p)$ and $zp \gg \log n$, it follows in view of (9) that the procedure colors $Z \setminus R$ using at most

$$\chi\left(G_{n,p}[Z \setminus R]\right) \leqslant \sum_{i \ge 0: z2^{-i} \ge 4u} \frac{z2^{-i}}{I\left(z2^{-(i+1)}\right)} \leqslant \frac{Azp}{\log(z/u)} \sum_{i \ge 0} \frac{(i+2)}{2^i} \leqslant \frac{O(Azp)}{\log(zp/\log n)}$$
(11)

many colors, where the second inequality exploits $z2^{-i} \ge 4u$ to infer $\log(z2^{-(i+1)}/u) \ge \log(z/u)/(i+2)$. Combining (10)–(11) with Lemma 5 and estimates (7)–(8) then implies (1a) for suitable $C = C(\gamma) > 0$.

2.1.2 Proof of Theorem 3 when $zp \ll \log n$

In the remaining case $zp \ll \log n$ of Theorem 3, we can directly bound $\chi(G_{n,p}[Z])$ using Lemma 7.

Proof of Theorem 3 when $zp \ll \log n$. Set $r := \left\lceil 4 \log n / \log\left(\frac{\log n}{zp}\right) \right\rceil$ and C := 16, say. Let \mathcal{E} denote the event that $\delta(G_{n,p}[S]) \leq 2r - 1$ for all $S \subseteq [n]$ with $|S| \leq z$. Noting that $zp \ll \log n$ implies $zp/r \ll \sqrt{zp/\log n}$, similar to (8) it follows via a standard

union bound argument that

$$\mathbb{P}(\neg \mathcal{E}) \leqslant \sum_{1 \leqslant s \leqslant z} \binom{n}{s} \binom{\binom{s}{2}}{sr} p^{sr}$$
$$\leqslant \sum_{1 \leqslant s \leqslant z} \left[n \left(\frac{zpe}{2r} \right)^r \right]^s \leqslant \sum_{s \geqslant 1} \left[n \left(\frac{zp}{\log n} \right)^{r/2} \right]^s \leqslant \sum_{s \geqslant 1} n^{-s} = o(1).$$

We henceforth assume that the event \mathcal{E} holds. Applying Lemma 7, we then color any $Z \subseteq [n]$ with $|Z| \leq z$ using at most $\chi(G_{n,p}[Z]) \leq 2r$ many colors, which together with Lemma 5 implies (1b) for $r \geq 2$ (in which case the ceiling in the definition of rcauses no major rounding issues).

In the remaining case r = 1 it is easy to see that $p \ll n^{-4}$ holds (with room to spare), so that $G_{n,p}$ has with high probability no edges. Hence $\chi(G_{n,p})$ is concentrated on one value, which trivially establishes (1b).

2.1.3 Proof of Theorem 4

The proof strategy for Theorem 4 is similar but simpler to Section 2.1.1. Namely, by iteratively applying Lemma 6 to extract independent sets of size $\Theta(\log_{1/(1-p)}(n))$ from $G_{n,p}[Z]$, we first color all but at most $n^{1/3}$ many vertices of Z using $O(z/\log_{1/(1-p)}(n))$ colors, and then trivially color the remaining uncolored vertices using at most $n^{1/3} \ll z/\log_{1/(1-p)}(n)$ additional colors (by giving each vertex a new color).

Proof of Theorem 4. Set q := 1-p, d := 1-q/2 as well as $u := 96(\log n)/q$ and $m := n^{1/3}$. If $q \leq n^{-1/9}$ then Lemma 5 implies the claimed bound in Theorem 4 by noting that $\chi(G_{n,p}[Z]) \leq |Z| \leq z \leq 9z \log(1/q)/\log n$, so we henceforth assume that $q \geq n^{-1/9}$. Let \mathcal{D} denote the event that $\delta(G_{n,p}[S]) \leq d(|S|-1)$ for all $S \subseteq [n]$ with $|S| \geq u$. Note that $\neg \mathcal{D}$ implies existence of $S \subseteq [n]$ with $|S| \geq u$ such that $G_{n,p}[S]$ has at most $q/2 \cdot \binom{|S|}{2}$ non-edges. Similarly to (7), using 1-p=q and $uq/48=2\log n$ it routinely follows that

$$\mathbb{P}(\neg \mathcal{D}) \leqslant \sum_{u \leqslant s \leqslant n} \binom{n}{s} e^{-\binom{s}{2}q/12} \leqslant \sum_{u \leqslant s \leqslant n} \left(n e^{-uq/48}\right)^s \leqslant \sum_{s \geqslant u} n^{-s} = o(1).$$
(12)

We henceforth assume that the event \mathcal{D} holds. Consequently, if a vertex subset $W \subseteq [n]$ has size at least $|W| \ge m$, then using Lemma 6 together with $m/u \gg n^{1/6}$ as well as $1/u \ll p = 1 - q$ and $q \le 1 - \gamma$ it follows that the induced subgraph $G_{n,p}[W]$ contains an independent set of size at least

$$\alpha\left(G_{n,p}[W]\right) \ge -\log_{\left[q/2\cdot(1-1/u)\right]}\left(|W|/u\right) \ge \frac{\log(m/u)}{\log(2/q^2)} \ge \frac{\log n}{A\log(1/q)} =: k$$

for a suitable constant $A = A(\gamma) \in (0, \infty)$. In order to color any vertex subset $Z \subseteq [n]$ of size at most $|Z| \leq z$, among the so-far uncolored vertices of Z we iteratively choose an independent set of size at least k and assign its vertices one new color, until at most m

uncolored vertices remain, and then use one new color for each remaining vertex. In view of $|Z| \leq z$ and $m \ll z/k$ it follows that this procedure colors Z using at most

$$\chi(G_{n,p}[Z]) \leqslant \frac{|Z|}{k} + m \leqslant \frac{2z}{k} \leqslant \frac{2Az}{\log_{1/q}(n)}$$

many colors, which together with Lemma 5 and estimate (12) completes our proof with $C := \max\{4A, 18\}$.

3 Concentration of $\chi(G_{n,p})$: the very dense case

We conclude by discussing the behavior of the chromatic number $\chi(G_{n,p})$ in the very dense case $1 - p = n^{-\Omega(1)}$. Here $\chi(G_{n,p})$ is closely linked to the size $\alpha = \alpha(G_{n,p})$ of the largest independent set of $G_{n,p}$. Namely, inspired by [9], it seems plausible that a near-optimal coloring can be obtained by first picking as many vertex-disjoint independent sets of size α as possible, and then covering (almost all of) the remaining vertices with independent sets of size $\alpha - 1$. More concretely, for $n^{-2/r} \ll 1 - p \ll n^{-2/(r+1)}$ we expect to have about

$$\mu_{r+1} = \mu_{r+1}(n, p) := \binom{n}{r+1} (1-p)^{\binom{r+1}{2}}$$
(13)

many vertices in independent sets of largest size $\alpha = \alpha(G_{n,p}) = r + 1$ (which in fact are mostly vertex-disjoint). Since $\mu_{r+1} = o(n)$, the near-optimal coloring heuristic then suggests that $\chi(G_{n,p}) \approx o(n) + n/r \approx n/r$, as we shall make rigorous in Section 3.3. Furthermore, since in our coloring heuristic we pick almost all independent sets of size r+1, whose number is well-known to fluctuate by about $\sqrt{\mu_{r+1}}$ in $G_{n,p}$ (see [23]), it then becomes plausible that $\chi(G_{n,p})$ should also vary by this amount, as formalized by the following conjecture.

Conjecture 8. If the edge-probability p = p(n) satisfies $(\log n)^{1/\binom{r}{2}} n^{-2/r} \ll 1 - p \leq (1+o(1))n^{-2/(r+1)}$ for some integer $r \geq 1$, then the following holds, for any $\epsilon > 0$ and any function $\omega = \omega(n) \to \infty$ as $n \to \infty$.

- (i) There is an interval of length $|\omega\sqrt{\mu_{r+1}}|$ that contains $\chi(G_{n,p})$ with high probability.
- (ii) No interval of length $\lfloor c_{\sqrt{\mu_{r+1}}} \rfloor$ contains $\chi(G_{n,p})$ with probability at least $\epsilon + o(1)$, where $c = c(\epsilon, r) > 0$.

Note that $\sqrt{\mu_{r+1}}$ increases from $\Theta(1)$ to $\Theta(\sqrt{n})$ as 1 - p increases from $n^{-2/r}$ to $n^{-2/(r+1)}$, so Conjecture 8 predicts that the concentration interval length of $\chi(G_{n,p})$ is neither monotone nor smooth in p. Furthermore, Conjecture 8(i) proposes a refinement of the well-known $\omega\sqrt{n}$ concentration bound of Shamir and Spencer [26]. More interestingly, by varying $r \ge 1$, Conjecture 8 also predicts that the concentration interval length changes infinitely many times from $n^{1/2+o(1)}$ to $(\log n)^{\Theta(1)}$ as 1 - p decreases from $n^{-o(1)}$ to $n^{-2+o(1)}$; see also Figure 1. This intriguing behavior differs conceptually from very recent predictions [9] for constant $p \in (0, 1)$.

The remainder of this section is organized as follows. In Section 3.1 we prove Conjecture 8 for r = 1, and in Section 3.2 we establish Theorem 2, i.e., show that the concentration interval length of $\chi(G_{n,p})$ undergoes a polynomial 'jump' around $1 - p = n^{-1+o(1)}$. In Section 3.3 we then prove that, under the assumptions of Conjecture 8, the chromatic number is typically $\chi(G_{n,p}) = (1 + o(1))n/r$, as predicted above.

3.1 Concentration result: proof of Conjecture 8 for r = 1

The following result verifies the case r = 1 of Conjecture 8: it states that $\chi(G_{n,p})$ is concentrated on an interval of length about $\sqrt{n} \cdot \sqrt{n(1-p)}$ when $n^{-2} \ll 1-p = O(1/n)$. In this range of p, Theorem 9(i) refines the well-known $\omega\sqrt{n}$ concentration bound of Shamir and Spencer [26]. Theorem 9(ii) extends observations of Alon and Krivelevich [2] and Bollobás [5] for p = 1 - 1/(10n) and $n^{-2} \ll 1 - p \ll n^{-3/2}$, respectively.

Theorem 9. If p = p(n) satisfies $n^{-2} \ll 1 - p \leq (D + o(1))/n$ for some constant $D \in (0, \infty)$, then for q := 1 - p the following holds, for any $\epsilon > 0$ and any function $\omega = \omega(n) \to \infty$ as $n \to \infty$.

- (i) There is an interval of length $|\omega n \sqrt{q}|$ that contains $\chi(G_{n,p})$ with high probability.
- (ii) No interval of length $\lfloor cn\sqrt{q} \rfloor$ contains $\chi(G_{n,p})$ with probability at least $\epsilon + o(1)$, where $c = c(\epsilon, D) > 0$.

Proof. The main idea is that the study of study of $\chi(G_{n,p})$ effectively reduces to the study of the maximum matching in the complement. Writing M for the maximum size (counting number of edges) of a matching in the complement of $G_{n,p}$, we have $\chi(G_{n,p}) \leq M + (n - 2M) = n - M$. Furthermore, for any proper coloring of $G_{n,p}$ with color classes $(I_s)_{1 \leq s \leq \chi(G_{n,p})}$ we have $n = \chi(G_{n,p}) + \sum_s (|I_s| - 1)$. Defining X_i as the number of independent sets of size i in $G_{n,p}$, for $Y := \sum_{i \geq 3} iX_i$ it then readily follows that

$$n - M - Y \leqslant \chi(G_{n,p}) \leqslant n - M.$$
(14)

Since $\mathbb{E}X_{i+1}/\mathbb{E}X_i = \mu_{i+1}/\mu_i \leq nq^i = o(q)$ for $i \geq 3$, it routinely follows that $\mathbb{E}Y = (1 + o(1))3\mu_3 = \Theta(n^3q^3) = o(n\sqrt{q})$. Applying Markov's inequality to Y, from (14) it follows that with high probability

$$\chi(G_{n,p}) = n - M - o(n\sqrt{q}). \tag{15}$$

Gearing up towards applying a combinatorial version of Talagrand's inequality, note that $M \ge s$ can be certified by s non-edges of $G_{n,p}$, and that adding or removing an edge from $G_{n,p}$ changes M by at most one. Since $\mathbb{E}M \le {n \choose 2}q$, a standard application of Talagrand's inequality (such as [21, Theorem 2]) thus yields

$$\mathbb{P}(|M - \mathbb{E}M| \ge \omega n\sqrt{q}/3) \le 2 \cdot \exp\left(-\Theta\left(\frac{\omega^2 n^2 q}{\mathbb{E}M + \omega n\sqrt{q}}\right)\right) = o(1),$$

which together with (15) completes the proof of case (i).

Finally, the remaining case (ii) follows immediately from (15) and Lemma 10 below (since the complement of $G_{n,p}$ has the same distribution as $G_{n,q}$, where $qn \leq D + o(1)$ and $n\sqrt{q} \gg 1$ hold by assumption).

Lemma 10. Let $q = q(n) \in [0,1]$ satisfy $n\sqrt{q}e^{-qn} \gg 1$, and define M as the maximum size (counting number of edges) of a matching in $G_{n,q}$. Then for any $\epsilon > 0$ there is a constant $d = d(\epsilon) > 0$ such that for any interval I of length $\lfloor dn\sqrt{q}e^{-qn} \rfloor$ we have $\mathbb{P}(M \in I) \leq \epsilon + o(1)$.

Proof-Outline of Lemma 10. The heuristic idea is that fluctuations in the number of isolated edges is a source of fluctuations of M. Defining $G_{n,q}^L \subseteq G_{n,q}$ as the induced subgraph of $G_{n,q}$ consisting of all connected components of size at least three, we set $Y := n - |V(G_{n,q}^L)|$. We also define X_1 as the number of isolated edges in $G_{n,q}$, and define X_2 as the maximum size of a matching in $G_{n,q}^L$. The main idea is then to show that, after conditioning on a fixed graph $G_{n,q}^L = G$ with $Y \approx \mathbb{E} Y$, the number X_1 of isolated edges fluctuates by $\Theta(\sqrt{\mathbb{E} X_1}) = \Theta(n\sqrt{q}e^{-qn})$. Since X_2 is determined by the graph G we conditioned on, this then allows us to show that $M = X_1 + X_2$ fluctuates by the same order of magnitude. See Appendix 5 for the full technical details of the proof of Lemma 10 (which are rather tangential to the other arguments here).

Remark 11 (Extension of Theorem 9). After refining the deviation from $cn\sqrt{q}$ to $cn\sqrt{q}e^{-nq}$ for suitable $c = c(\epsilon) > 0$, the above proof of Theorem 9 (ii) carries over under the weaker assumption $n\sqrt{q}e^{-nq} \gg \max\{1, (nq)^3\}$, which in particular holds for when $n^{-1} \ll nq \ll \log n$, say.

3.2 Polynomial concentration jump: proof of Theorem 2

The following simple lemma proves a weak version of Conjecture 8(i) for $r \ge 2$, with deviation $\omega \sqrt{\mu_{r+1}}$ replaced by $\omega \mu_{r+1}$. For our purposes, the crux is that Lemma 12 establishes concentration of $\chi(G_{n,p})$ on $n^{o(1)}$ values near the transition points $1-p = n^{-2/r+o(1)}$. By combining the case r = 2 of Lemma 12 with Remark 11 we thus immediately establish Theorem 2, which essentially says that the typical length of the shortest interval containing $\chi(G_{n,p})$ undergoes a steep transition from $\Omega(n^{1/2-\epsilon})$ down to $O(n^{\epsilon})$ around edge-probability $p = 1 - n^{-1+o(1)}$, as illustrated by Figure 1 and predicted by Conjecture 8. This polynomial concentration 'jump' is intriguing, and we remark that a related phenomenon also occurs in [20, 18].

Lemma 12. Let $\omega = \omega(n) \to \infty$ as $n \to \infty$ be an arbitrary function, and let $r \ge 2$ be an integer. If p = p(n) satisfies $(\log n)^{1/\binom{r}{2}} n^{-2/r} \ll 1 - p \ll (\log n) n^{-2/(r+1)}$, then there is an interval of length $\lfloor \omega \mu_{r+1} \rfloor$ that contains $\chi(G_{n,p})$ with high probability. Furthermore, for $1 - p = n^{-2/r}x$ we have $\omega \mu_{r+1} \le \omega x^{\binom{r+1}{2}}$.

Proof. Using the assumed lower bound on q := 1 - p, a celebrated result of Johansson, Kahn and Vu [13, Theorem 2.1] implies that the complement of $G_{n,p}$ with high probability contains $\lfloor n/r \rfloor$ vertex-disjoint cliques of size r, meaning that we can color all

but $n - r\lfloor n/r \rfloor \leq r$ vertices of $G_{n,p}$ using at most n/r colors. By coloring the remaining uncolored vertices with distinct new colors, it follows that with high probability

$$\chi(G_{n,p}) \leqslant n/r + r. \tag{16}$$

To obtain a lower bound on $\chi(G_{n,p})$ we shall use a variant of the argument leading to (14), again writing X_i for the number of independent sets of size i in $G_{n,p}$. Indeed, for any proper coloring of $G_{n,p}$ with color classes $(I_s)_{1 \leq s \leq \chi(G_{n,p})}$ we have $n = \sum_s |I_s| \leq$ $Y + r \cdot \chi(G_{n,p})$ for $Y := \sum_{i>r} iX_i$, so that

$$\chi(G_{n,p}) \ge (n-Y)/r. \tag{17}$$

Using the assumed upper bound on q = 1 - p we infer that $\mathbb{E}X_{i+1}/\mathbb{E}X_i \leq nq^i = o(q)$ for $i \geq r+1$ (with room to spare), and so it routinely follows that $\mathbb{E}Y = \Theta(\mu_{r+1})$. Finally, in view of (16)–(17), now an application of Markov's inequality to Y readily completes the proof.

The deviation $\omega \mu_{r+1}$ in Lemma 12 can be further reduced using a two-round exposure argument (in some range of p), but a proof of the conjectured deviation $\omega \sqrt{\mu_{r+1}}$ seems to require additional ideas.

3.3 The typical value

Finally, an approximate coloring argument similar to Section 3.2 also allows us to determine the with high probability asymptotics of the chromatic number $\chi(G_{n,p})$ for most edge-probabilities of form $1 - p = n^{-\Omega(1)}$, settling a recent conjecture of Isaev and Kang, see [10, Conjecture 1.2].

Theorem 13. If the edge-probability p = p(n) satisfies $n^{-2/r} \ll 1 - p \ll n^{-2/(r+1)}$ for some fixed integer $r \ge 1$, then with high probability $\chi(G_{n,p}) = (1 + o(1))n/r$.

Proof. For the lower bound we exploit that the argument leading to (17) again gives $\chi(G_{n,p}) \ge (n-Y)/r$ and $\mathbb{E}Y = \Theta\left(n^{r+1}(1-p)^{\binom{r+1}{2}}\right)$. Since $1-p \ll n^{-2/(r+1)}$ implies $\mathbb{E}Y = o(n)$, now an application of Markov's inequality to Y readily shows that with high probability $\chi(G_{n,p}) \ge (1-o(1))n/r$.

We also include the upper bound argument from [10] for completeness. Namely, using $1-p \gg n^{-2/r}$, an old result of Ruciński [24, Theorem 4] implies that the complement of $G_{n,p}$ with high probability contains (1 - o(1))n/r vertex-disjoint cliques of size r, meaning that we can color all but o(n) vertices of $G_{n,p}$ using at most n/r colors. By coloring the remaining uncolored vertices with distinct new colors, it follows that with high probability $\chi(G_{n,p}) \leq n/r + o(n) \leq (1 + o(1))n/r$, completing the proof.

Note that the estimate $\chi(G_{n,p}) = (1+o(1))n/r$ from Theorem 13 can also be rewritten as

$$\chi(G_{n,p}) = \frac{(1+o(1))n}{\left\lfloor 2\log_{1/(1-p)}(np) \right\rfloor}.$$
(18)

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Interestingly, the same expression (18) also gives the with high probability asymptotics of $\chi(G_{n,p})$ when the edge-probability p = p(n) satisfies $1/n \ll p \leq 1 - n^{-o(1)}$, see [4, 16, 10]. We leave it as an intriguing open problem to determine the asymptotics of $\chi(G_{n,p})$ for edge-probabilities of the form $p = 1 - \Theta(n^{-2/r})$, in which case the behavior of related graph parameters also remains open, see [7, Section 4.2].

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4 Appendix: proofs of Lemmas 5–7

Proof of Lemma 5. Define $\lambda = \lambda(n, p, \omega)$ as the smallest integer with $\mathbb{P}(\chi(G_{n,p}) \leq \lambda) \geq 1/\omega$. Let Y denote the minimum size of a vertex subset $Z \subseteq [n]$ with $\chi(G_{n,p}[[n] \setminus Z]) \leq \lambda$. By definition of λ we have

$$\mathbb{P}(\chi(G_{n,p}) < \lambda) \leqslant 1/\omega \leqslant \mathbb{P}(Y=0).$$
(19)

Using the vertex-exposure approach to the bounded difference inequality, we write $Y = Y(X_1, \ldots, X_n)$ where the independent auxiliary variables $X_i := \{(i, j) \in E(G_{n,p}) : i < j\}$ contain the edges of $G_{n,p}$ between vertex *i* and vertices $\{i+1,\ldots,n\}$. Note that changing a single X_i can change Y by at most one. In view of (19), now a routine application of the bounded differences inequality (such as [12, Corollary 2.7]) to Y yields

$$1/\omega \leqslant \mathbb{P}(Y=0) = \mathbb{P}(Y \leqslant \mathbb{E} Y - \mathbb{E} Y) \leqslant \exp\left(\frac{-(\mathbb{E}Y)^2}{2n}\right),$$

which implies that $\mathbb{E}Y \leq \sqrt{2n \ln \omega}$. Since $z = \omega \sqrt{n} \gg \sqrt{2n \ln \omega}$, by again applying the bounded differences inequality to Y it then follows that

$$\mathbb{P}(Y \ge z) \leqslant \mathbb{P}(Y \ge \mathbb{E}Y + \sqrt{2n \ln \omega}) \leqslant 1/\omega,$$

which together with estimate (19) and the 'with high probability' event (5) implies that

$$\mathbb{P}(\lambda \leqslant \chi(G_{n,p}) \leqslant \lambda + \Gamma) \ge 1 - 2/\omega - \mathbb{P}(\text{event } (5) \text{ fails}) = 1 - o(1).$$
(20)

Combining the definition of the median M = M(n, p) with (19) and (20), it deterministically follows that

$$\lambda \leqslant M \leqslant \lambda + \Gamma$$

for all sufficiently large n, which together with (20) then completes the proof of Lemma 5.

Proof of Lemma 6. Given a vertex subset $W \subseteq V(G)$ of size $|W| \ge u$, we construct an independent set greedily: set $W_0 = W$ and, for $i \ge 1$, pick $w_i \in W_{i-1}$ with minimal degree in $G[W_{i-1}]$ and set

 $W_i = \{ v \in W_{i-1} : v \text{ not adjacent to } w_i \}.$

We terminate as soon as W_j is empty, in which case we obtain an independent set $\{w_1, \ldots, w_j\} \subseteq W$. If $|W_{i-1}| \ge u$ holds, then we know that w_i has degree at most $d(|W_i| - 1)$ in $G[W_i]$, implying that

$$|W_i| \ge (1-d)(|W_{i-1}|-1) \ge (1-d)(1-1/u)|W_{i-1}|.$$

It follows that W_i is non-empty as long as

$$i - 1 \leq -\log_{(1-d)(1-1/u)} (|W|/u) =: I(W|),$$

so we terminate with an independent set $\{w_1, \ldots, w_j\} \subseteq W$ of size $j \ge \lfloor I(|W|) + 1 \rfloor \ge I(|W|)$.

Proof of Lemma 7. We apply induction on the number of vertices. The base case |V(G)| = 1 is trivial. For the induction step |V(G)| > 1, we pick a vertex v of minimum degree, inductively color the subgraph G - v (obtained by removing v) using r + 1 colors, and color v with a color not used by its at most r neighbors.

5 Appendix: proof of Lemma 10

Proof of Lemma 10. We keep the setup from the proof-outline in Section 3.1. In particular, we define $G_{n,q}^L \subseteq G_{n,q}$ as the induced subgraph of $G_{n,q}$ consisting of all connected components of size at least three, define X_1 as the number of isolated edges in $G_{n,q}$, and define X_2 as the maximum size (counting number of edges) of a matching in $G_{n,q}^L$. Turning to the typical number $Y := n - |V(G_{n,q}^L)|$ of vertices in components of size at most two, note that the expected number of isolated vertices in $G_{n,q}$ is

$$\lambda_0 := n(1-q)^{n-1} \sim ne^{-qn}$$

and that the expected number of isolated edges in $G_{n,q}$ is

$$\lambda_1 := \binom{n}{2} q(1-q)^{2n-4} \sim \frac{n^2 q e^{-2qn}}{2},$$

where we use the shorthand $a_n \sim b_n$ for $a_n = (1 + o(1))b_n$ to avoid clutter. Note that $\mathbb{E} Y = \lambda_0 + 2\lambda_1$. By a routine second moment calculation, for suitable $\epsilon_n = o(1)$ it follows that with high probability

$$Y = (1 \pm \epsilon_n)(\lambda_0 + 2\lambda_1). \tag{21}$$

Deferring the choice of the sufficiently small constant $d = d(\epsilon) > 0$, set

$$N := \lfloor dn\sqrt{q}e^{-qn} \rfloor \sim d\sqrt{2\lambda_1} \gg 1.$$
(22)

Since $M = X_1 + X_2$, with X_2 determined by $G_{n,q}^L$, for any interval I of length N it follows that

$$\mathbb{P}(M \in I) \leqslant \mathbb{P}\left(\text{estimate (21) fails}\right) + \sum_{G} \mathbb{P}\left(X_1 + X_2 \in I \mid G_{n,q}^L = G\right) \mathbb{P}(G_{n,q}^L = G)$$

$$\leqslant o(1) + \max_{G,J} \mathbb{P}\left(X_1 \in J \mid G_{n,q}^L = G\right),$$
(23)

where J is taken over all intervals of length N, and G is taken over all graphs for which (i) all connected components all have size at least three and (ii) its number of vertices n-Y is compatible with (21).

To complete the proof, it suffices to show that the final probability appearing in (23) is at most ϵ for sufficiently large n. To this end we henceforth fix a graph G that occurs in the maximum of (23) described above, and let J be any interval of length N that maximizes $\mathbb{P}(X_1 \in J \mid G_{n,q}^L = G)$. We define $G_{n,q}^S \subseteq G_{n,q}$ as the induced subgraph of $G_{n,q}$

consisting of all connected components of size at most two. Note that $G_{n,q}^S$ has Y vertices, with $Y = n - |V(G_{n,q}^L)|$ determined by $G_{n,q}^L$. After conditioning on $G_{n,q}^L = G$, a moment's thought reveals that $G_{n,q}^S$ has the same distribution as $G_{Y,q}$ conditioned on all connected components having size at most two. Since there are $\frac{1}{u!} \prod_{0 \le i < u} {Y-2i \choose 2}$ graphs on Y vertices that consist of u isolated edges and Y - 2u isolated vertices, it follows for any integer $0 \le m \le Y/2 - 1$ that

$$\Phi_{G}(m) := \frac{\mathbb{P}(X_{1} = m \mid G_{n,q}^{L} = G)}{\mathbb{P}(X_{1} = m + 1 \mid G_{n,q}^{L} = G)}$$

$$= \frac{\frac{1}{m!} \prod_{0 \le i < m} {Y - 2i \choose 2} \cdot q^{m} (1 - q)^{{Y \choose 2} - m}}{\frac{1}{(m+1)!} \prod_{0 \le i \le m} {Y - 2i \choose 2} \cdot q^{m+1} (1 - q)^{{Y \choose 2} - (m+1)}} = \frac{(1 - q)(m+1)}{q{Y - 2m \choose 2}}.$$
(24)

To get a handle on these probabilities, first note that $\lambda_0 = \Omega(\lambda_1)$ follows from $\lambda_1/\lambda_0 = \Theta(nqe^{-nq}) = O(1)$. Using (21) we then deduce $\sqrt{q} Y = \Theta(\sqrt{q}\lambda_0) = \Theta(n\sqrt{q}e^{-qn}) \gg 1$ as well as

$$\Phi_G(0) = \frac{\Theta(1)}{(\sqrt{q} Y)^2} \ll 1$$
 and $\Phi_G(Y/2 - 1) = \frac{\Theta(\sqrt{q} Y)}{q^{3/2}} \gg 1$,

so by the intermediate value theorem there exists a real $0 < m_0 < Y/2 - 1$ satisfying $\Phi_G(m_0) = 1$. Noting that $\Phi_G(m)$ is a strictly increasing function, it now is routine to check that $m_0 \sim \lambda_1$, since for any $m \sim \lambda_1$ we have $Y - 2m \sim \lambda_0 + o(\lambda_1) \sim \lambda_0$ and $\Phi_G(m) \sim \lambda_1/(q\lambda_0^2/2) \sim 1$. Using $\Phi_G(m_0) = 1$ together with $Y - 2m_0 \sim \lambda_0 = \Omega(\lambda_1)$ and $m_0 \sim \lambda_1$, for any integer $m = m_0 + i$ with $i = o(\lambda_1)$ we similarly infer that

$$\Phi_G(m) = \frac{\Phi_G(m_0+i)}{\Phi_G(m_0)} = \frac{\binom{y-2m_0}{2}}{\binom{y-2m_0-2i}{2}} \frac{m_0+i+1}{m_0+1} = \left(1 + \frac{2i}{y-2m_0-2i+O(1)}\right)^2 \left(1 + \frac{i}{m_0+1}\right) = 1 + \frac{\Theta(i)}{\lambda_1}.$$
(25)

Since the probability ratio $\Phi_G(m)$ from (24) is a strictly increasing function, the interval J must intersect $\{ [m_0] - 1, \ldots, [m_0] + 1 \}$ in at least one element, because otherwise we could increase the probability $\mathbb{P}(X_1 \in J \mid G_{n,q}^L = G)$ by shifting the interval Jby plus or minus one (contradicting maximality). Furthermore, by definition of m_0 , we have $\Phi_G(m) \leq 1$ for all $m \leq m_0$. By choosing the constant $d = d(\epsilon) > 0$ sufficiently small, for all integers $m \in J$ and $1 \leq s \leq \lfloor 2/\epsilon \rfloor$ it follows in view of (24), (25) and (22) that

$$\frac{\mathbb{P}(X_1 = m \mid G_{n,q}^L = G)}{\mathbb{P}(X_1 = m + 2Ns \mid G_{n,q}^L = G)} = \prod_{0 \le i < 2sN} \Phi_G(m) \\
\leq \left(1 + \frac{\Theta(sN)}{\lambda_1}\right)^{2Ns} \\
\leq e^{O(s^2d^2)} \\
\leq 2.$$
(26)

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Since the interval J has length N, by applying (26) to each $m \in J$ it then follows that

$$2/\epsilon \cdot \mathbb{P}(X_1 \in J \mid G_{n,q}^L = G) \leqslant \sum_{1 \leqslant s \leqslant \lceil 2/\epsilon \rceil} 2 \mathbb{P}(X_1 - 2Ns \in J \mid G_{n,q}^L = G) \leqslant 2$$

which together with estimate (23) completes the proof of Lemma 10, as discussed. \Box