Blow-up lemmas for sparse graphs

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1HH by was supported by FAPESP (Proc. 2010/16526-3 and 2013/03447-6). YK was supported by FAPESP (2013/03447-6, 2013/07699-0) and CNPq (459335/2014-6, 310974/2013-5). The cooperation of the authors was supported by a joint CAPES-DAAD PROBRAL project (Proj. no. 430/15). The authors are grateful to NUMEC/USP, Núcleo de Modelagem Estocástica e Complexidade of the University of São Paulo, and Project MaCLinC/USP, for supporting this research. (July 20, 2016)
Abstract. The blow-up lemma states that a system of super-regular pairs contains all bounded degree spanning graphs as subgraphs that embed into a corresponding system of complete pairs. This lemma has far-reaching applications in Extremal Combinatorics.

We prove sparse analogues of the blow-up lemma for subgraphs of random and of pseudorandom graphs. Our main results are the following three sparse versions of the blow-up lemma: one for embedding spanning graphs with maximum degree $\Delta$ in subgraphs of $G_{n,p}$ with $p = C(\log n/n)^{\Delta}$; one for embedding spanning graphs with maximum degree $\Delta$ and degeneracy $D$ in subgraphs of $G_{n,p}$ with $p = C\Delta (\log n/n)^{2D+1}$; and one for embedding spanning graphs with maximum degree $\Delta$ in $(p, c p^\max(4, (3\Delta+1)/2))$-bijumbled graphs.

We also consider various applications of these lemmas.
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CHAPTER 1

Introduction, applications, results and proof overview

1.1. Introduction

Szemerédi’s regularity lemma [67], originally developed for the proof of Szemerédi’s celebrated result on arithmetic progressions [66], is one of the most influential tools in modern Discrete Mathematics. Numerous variants of this lemma, which is an approximate structure theorem for graphs, have been established for applications in other areas of Mathematics, such as Additive Number Theory, Information Theory, and Statistical Mechanics.

Applications of the regularity lemma include a wealth of results in such diverse areas as Extremal Combinatorics, Ramsey Theory, Property Testing, or Discrete Geometry. In such applications the regularity lemma is usually complemented by the counting lemma or the blow-up lemma. The former allows one to deduce estimates on small substructure counts from the (finitely sized) approximate structure provided by the regularity lemma. The latter, on the other hand, is powerful for establishing global structural properties. More precisely, the blow-up lemma, proved by Komlós, Szárközy and Szemerédi [46], permits the embedding of certain bounded degree spanning graphs. Alternative proofs can be found in [48, 61, 62]; for a nice introduction to the blow-up lemma and explanations about how it is used in applications see the surveys [47, 49].

One limitation of the original regularity lemma is that, because of the error terms, this lemma is suitable only for dense graphs, that is $n$-vertex graphs with $\Omega(n^2)$ edges. Nonetheless it is desirable to have equally effective tools at hand for sparse graphs. The most prominent example of why such sparse structures are of importance is without doubt the famous Green-Tao Theorem [34] on arithmetic progressions in the primes, which uses an approximate structure theorem for certain sparse subsets of the integers. Moreover, the modern branch of Extremal Combinatorics concerned with resilience results (a term coined by Sudakov and Vu [65]), which recently received much interest, investigates such sparse graphs.

Analogues of the regularity lemma that also work in a sparse setting, that is, for $n$-vertex graphs with $o(n^2)$ edges, do exist [41, 64]. However, a corresponding counting lemma simply fails to be true in general (see, e.g., [42]). This impediment can be overcome by posing additional restrictions on the graphs under consideration. The existing counterexamples are known not to occur in random or certain pseudorandom graphs, and it was a major breakthrough when recently counting lemmas could finally be established in these settings: Counting lemmas for subgraphs of random graphs were proved in [19, 26, 63], and for subgraphs of pseudorandom graphs in [25].

What was so far missing in this effort to transfer these tools to the sparse setting was a sparse version of the blow-up lemma. An important step in this direction was taken in [44], where an embedding lemma for bounded degree graphs on $cn$ vertices for some very small constant $c$ in $n$-vertex subgraphs of random graphs was established. In [22] it was then shown that methods developed in [13] can be
used to prove a blow-up type result for embedding almost spanning bipartite graphs. Moreover, in [18] a sparse embedding lemma for the special case of spanning triangle factors was proved. Analogues of these partial results for pseudorandom graphs are not known. But in [56] the importance of a generalisation of the blow-up lemma to pseudorandom graphs was acknowledged.

In this paper we provide this missing piece and establish several sparse versions of the blow-up lemma for random and for pseudorandom graphs. We also discuss a variety of relatively straightforward applications of these lemmas and indicate more intricate applications, which will appear elsewhere.

Organisation. We first motivate our blow-up lemmas in Section 1.2 by collecting various applications of these lemmas, some of which will be proved in Chapter 6 and some of which will be proved in future papers. In Section 1.3 we then provide our blow-up lemmas together with the necessary notation. We also state regularity inheritance lemmas which are necessary in applications. In Section 1.4 we provide an outline of the proofs of the blow-up lemmas. In Chapters 2–5 we give the proofs of our blow-up lemmas. We will describe the purposes of these various chapters in more detail in the proof outline (Section 1.4). We give the proofs of our applications in Chapter 6, and finish off with some concluding remarks in Chapter 7.

Notational remarks. We will routinely omit floor and ceiling signs when they do not affect the argument. All logarithms are taken to base 2.

1.2. Applications

In this section we collect a number of applications of our main results, establishing new structural properties of random and of pseudorandom graphs. We defer proofs of all these results to Chapter 6. The random graph model we work with in this paper is the binomial model $G_{n,p}$, where each potential edge is included in a graph with $n$ vertices independently with probability $p = p(n)$. If $G_{n,p}$ has some property with probability tending to 1 as $n$ tends to infinity, we say $G_{n,p}$ has this property asymptotically almost surely, abbreviated a.a.s.

The study of pseudorandom graphs was initiated by Thomason [68], who asked for a set of easy deterministic properties enjoyed by $G_{n,p}$ a.a.s. which by themselves imply many of the complex structural properties we know to hold for $G_{n,p}$. The pseudorandomness notion we use in our blow-up lemma is closely related to the notion suggested by Thomason, and is among the most widely used ones by now (for example, the sparse counting lemma in [25] is developed for this notion as well).

We say a graph $\Gamma$ is $(p, \beta)$-bijumbled if for all subsets $X, Y \subseteq V(\Gamma)$ we have
\[
|e_{\Gamma}(X,Y) - p|X||Y| \leq \beta \sqrt{|X||Y|},
\]
where $e_{\Gamma}(A,B)$ is the number of edges in $\Gamma$ with one endvertex in $A$ and the other endvertex in $B$. The random graph $G_{n,p}$ is with high probability $(p, \beta)$-bijumbled with $\beta = O(\sqrt{pn})$, which justifies this definition.

Another class of pseudorandom graph we shall refer to in the applications are $(n, d, \lambda)$-graphs. These have been studied extensively and are a special case of bijumbled graphs. For a graph $\Gamma$ let $\lambda_1 \geq \lambda_2 \geq \cdots \geq \lambda_n$ be the eigenvalues of the adjacency matrix of $\Gamma$. We call $\lambda(\Gamma) := \max\{|\lambda_2|, |\lambda_n|\}$ the second eigenvalue of $\Gamma$. An $(n, d, \lambda)$-graph $\Gamma$ is a $d$-regular graph on $n$ vertices with $\lambda(\Gamma) \leq \lambda$. The connection between $(n, d, \lambda)$-graphs and bijumbled graphs is provided by the well-known expander mixing lemma (see, e.g., [17]), which states that if $\Gamma$ is an $(n, d, \lambda)$-graph, then
\[
|e_{\Gamma}(A,B) - \frac{d}{2}|A||B| \leq \lambda(\Gamma) \sqrt{|A||B|}
\]
for all disjoint subsets $A, B \subseteq V(\Gamma)$. This implies that $\Gamma$ is $(\frac{d}{2}, \lambda(G))$-bijumbled.
1.2.1. Universal graphs. Let $\mathcal{H}$ be a set of graphs. We say $G$ is $\mathcal{H}$-universal if $H \subseteq G$ for each $H \in \mathcal{H}$. Two particular classes of interest are $\mathcal{H}(n, \Delta)$, the $n$-vertex graphs with maximum degree $\Delta$, and $\mathcal{H}(n, d, \Delta)$, the $n$-vertex graphs with maximum degree $\Delta$ and degeneracy $d$.

It is interesting to ask when random, or quasirandom, graphs are $\mathcal{H}$-universal. An easy corollary of Lemma 1.21 is that for $\Delta \geq 2$, the random graph $G_{n,p}$ is a.a.s. $\mathcal{H}(n, \Delta)$-universal if $p \geq C\left(\frac{\log n}{n}\right)^{1/(2\Delta)}$, reproving a result of Dellamonica, Kohayakawa, Rödl and Ruciński [29] (for $\Delta \geq 3$) and Kim and Lee [40] (for $\Delta = 2$), the proof being similar to that of Theorem 1.1 below. One would expect that $G_{n,p}$ is already $\mathcal{H}(n, d, \Delta)$-universal for some rather smaller $p$ if $d$ is much less than $\Delta$. For $d = 1$, i.e. for bounded degree trees, Montgomery [58] announced that one may take $p = \frac{\text{polylog } n}{n}$, which is optimal up to log factors. For $d \geq 2$, the best previous result, due to Ferber, Nenadov and Peter [31] is that we can take $p = \omega\left(\Delta^{12}n^{-1/4d} \log^{3} n\right)$. We prove the following strengthening.

**Theorem 1.1.** For each $d \geq 2$, $\Delta \in \mathbb{N}$ and each $\gamma > 0$ there exists $C$ such that if $p \geq C\left(\frac{\log n}{n}\right)^{1/(2\Delta d)}$, the random graph $G_{n,p}$ is a.a.s. $\mathcal{H}(n, d, \Delta)$-universal. Furthermore, if $p \geq C\left(\frac{\log n}{n}\right)^{1/(2\Delta)}$ the random graph $G_{(1+\gamma)n,p}$ is a.a.s. $\mathcal{H}(n, d, \Delta)$-universal.

The almost spanning universality result for $G_{(1+\gamma)n,p}$ with $p \geq C\left(\frac{\log n}{n}\right)^{1/2d}$ should be compared to the recent result of Conlon, Ferber, Nenadov and Škorić [27], who showed that the random graph $G_{(1+\gamma)n,p}$ is $\mathcal{H}(n, \Delta)$-universal when $p = \omega(n^{-1/(\Delta-1)} \log^{2} n)$ for $\Delta \geq 3$. Our result is better for families $\mathcal{H}(n, d, \Delta)$ with $\Delta \geq 2d + 1$.

It is also interesting to ask how big $e(G)$ must be for an $\mathcal{H}(n, d, \Delta)$-universal graph $G$. Alon and Capalbo showed that the correct answer is $\Theta(n^{2-2/\Delta})$ in [12] for the case $d = \Delta$ (i.e. for the class $\mathcal{H}(n, \Delta)$), and that if one further insists that $v(G) = n$ then the correct answer is still $O(n^{2-2/\Delta} \log^{1/\Delta} n)$ in [11] (where the extra polylog-factor is believed to be unnecessary). Theorem 1.1 provides sparser $\mathcal{H}(n, d, \Delta)$-universal graphs when $d \leq \Delta/4$.

Finally, we are able to show that sufficiently pseudorandom graphs are $\mathcal{H}(n, \Delta)$-universal. This was first suggested by Krivelevich, Sudakov and Szabó [56] who gave a condition on $\lambda$ for $(n, d, \lambda)$-graphs to contain a triangle factor. In earlier work [3] we were able to improve on their condition, showing that there is $\varepsilon > 0$ such that $(p, c\varepsilon^{5/2}n)$-bijumbled $n$-vertex graphs $G$ with minimum degree $\frac{1}{2}pn$ actually contain the square of a Hamilton cycle (which implies the theorem of Krivelevich, Sudakov and Szabó with a better value for $\lambda$). For the class $\mathcal{H}(n, \Delta)$, to the best of our knowledge the only existing universality result is that obtained by applying the original blow-up lemma of Komlós, Sárközy and Szemerédi [46], which is possible when $\beta \leq n^{-\Omega(\Delta^{-2})}$. We can at least achieve the correct power of $\Delta$ in the exponent.

**Theorem 1.2.** For each $\Delta \geq 2$ there exists $c > 0$ such that for any $p > 0$, if $\beta \leq c p^{\max(4, \Delta/2 + 1/2)} n$, any $n$-vertex $(p, \beta)$-bijumbled graph $G$ with $\delta(G) \geq \frac{1}{2}pn$ is $\mathcal{H}(n, \Delta)$-universal.

We note that Alon and Bourgain [10] showed that the ‘Cayley-sums’ $G$ of any multiplicative subgroup $U$ of a finite field $\mathbb{F}_{q}$, where $V(G) = U$ and $uv \in E(G)$ whenever $u + v \in U$, is an $(|U|, d, \sqrt{q})$-graph. Thus the above theorem shows that sufficiently large multiplicative subgroups of finite fields contain all ‘bounded-degree’ additive patterns; see [3] or [10] for a more detailed discussion.

\footnote{In fact they proved a universality result in terms of a constraint on the ‘maximum average degree’; in the class of graphs with degeneracy $d$ this quantity is between $d$ and $2d$.}
1.2.2. Partition universality. Given a set $H$ of graphs, the graph $G$ is $r$-partition universal for $H$ if in any $r$-colouring of the edge-set $E(G)$ there is a colour class which is $H$-universal. Kohayakawa, Rödl, Schacht and Szemerédi [44] showed that for each $r$ and $\Delta$ there exists $C$ such that if $p \geq C \left( \frac{\log n}{n} \right)^{1/2\Delta}$ then $G_{n,p}$ is a.a.s. $r$-partition universal for $\mathcal{H}(n, \Delta)$. This result is again an easy corollary of Lemma 1.21, proved along the same lines as the following lower bound for $r$-partition universality for $\mathcal{H}(n, d, \Delta)$.

**Theorem 1.3.** For each $r, d, \Delta \in \mathbb{N}$ there exists $C$ such that if $p \geq C \left( \frac{\log n}{n} \right)^{1/(2d)}$, then a.a.s. $G_{n,p}$ is $r$-partition universal for $\mathcal{H}(n, d, \Delta)$.

Kohayakawa, Rödl, Schacht and Szemerédi [44] also asked if a sufficiently pseudorandom graph is $r$-partition universal for $\mathcal{H}(n, \Delta)$. Using Lemma 1.25 we can answer this in the affirmative.

**Theorem 1.4.** For each $r, \Delta \in \mathbb{N}$ there exists $c > 0$ such that if $p > 0$ and $G$ is an $(n/c)$-vertex graph which is $(p, cp^{\max(4, (3\Delta+1)/2)})$-bijumbled, then $G$ is $r$-partition universal for $\mathcal{H}(n, \Delta)$.

We note that a corollary to Theorem 1.3 (which also follows from the proof in [44]) is the following Folkman-type result, where by a Folkman-type we refer to the existence of certain $\mathcal{F}$-free graphs that are Ramsey.

**Corollary 1.5.** For each $r \in \mathbb{N}$ and $\Delta \geq 2$, there exists a $K_{2\Delta}$-free graph which is $r$-partition universal for $\mathcal{H}(n, \Delta)$.

1.2.3. Maker-Breaker games. The Maker-Breaker $H$-game on $K_n$ with bias $b$ is the following game. Maker and Breaker take turns to colour the edges of $K_n$ with respectively red and blue. In each turn, Maker colours one edge, while Breaker colours $b$ edges (and edges may not be recoloured). Maker’s aim is to create a red copy of $H$, while Breaker’s aim is to prevent Maker from creating a red $H$. This class of games has been studied extensively; see the book of Beck [20] and a recent survey of Krivelevich [50]. If Breaker wins the $H$-game on $K_n$ with bias $b$, then obviously Breaker also wins with bias $b + 1$; it follows that there is a threshold bias $b$ for each game which is the smallest $b$ such that Breaker wins the $H$-game.

If either player, playing a randomised strategy, wins a given game with positive probability against perfect play from the opponent, then they have a deterministic winning strategy. Bednarska and Łuczak [21] used this observation to determine the order of magnitude of the threshold bias for the $H$-game on $K_n$ for each fixed graph $H$, showing that it is $\Theta(p^{-1})$ for the threshold probability $p$ at which $G_{n,p}$ contains $H$ ‘robustly’ in the sense that removing any $\varepsilon(H)pn^2$ edges fails to destroy all copies of $H$. However for $H$ depending on $n$, much less is known. The critical bias for the Hamiltonicity game, with $H = C_n$, was determined only recently by Krivelevich [52], while for more general bounded degree graphs it was only known that the critical bias tends to infinity as $n$ tends to infinity\(^2\). Using a recent result of Ferber, Krivelevich and Naves [32], which informally shows that Maker can a.a.s. make a subgraph of $G_{n,p}$ with minimum degree very close to $pn$, we can show that in fact Maker can win the $\mathcal{H}(n, \Delta)$-universality game on $K_{(1+\varepsilon)n}$ (i.e., Maker’s graph contains simultaneously each $H \in \mathcal{H}(n, \Delta)$) against a polynomially growing bias, and that this bias can be increased on the class $\mathcal{H}(n, d, \Delta)$ if $d$ is small enough. For the subclasses $\mathcal{H}'(n, \Delta)$ and $\mathcal{H}''(n, d, \Delta)$ of triangle-free graphs in $\mathcal{H}(n, \Delta)$ and $\mathcal{H}(n, d, \Delta)$ respectively, we can even win on $K_n$.

\(^2\)The article [16] only explicitly shows that the critical bias is at least two, but Krivelevich [51] has explained that the techniques there would work for any constant $b$ if $n$ is large enough.
Theorem 1.6. For each \(d, \Delta \in \mathbb{N}\) and \(\delta > 0\) there exists \(c > 0\) with the following properties. If \(b \leq c \left( \frac{n}{\log n} \right)^{1/\Delta}\) then Maker wins the \(H(n, \Delta)\)-universality game on \(K_{(1+\delta)n}\) with bias \(b\). Furthermore, with the same bias Maker wins the \(H'(n, \Delta)\)-universality game on \(K_n\).

If \(b \leq c \left( \frac{n}{\log n} \right)^{1/(2d)}\) then Maker wins the \(H(n, d, \Delta)\)-universality game on \(K_{(1+\delta)n}\) with bias \(b\). If we insist on the stronger \(b \leq c \left( \frac{n}{\log n} \right)^{1/(2d+1)}\), then Maker wins the \(H'(n, d, \Delta)\)-game on \(K_n\).

We note that the proof of Theorem 1.6 uses a randomised strategy for Maker which succeeds with high probability against any strategy of Breaker, but this strategy does not seem to be easy to derandomise. It would be interesting to find explicit deterministic Maker strategies for these games, even for substantially smaller bias.

In [4] we strengthen the above result, proving the following theorem in which the 'triangle-free' restriction is removed.

Theorem 1.7 (Allen, Böttcher, Kohayakawa, Naves, Person [4]). For each \(d, \Delta \in \mathbb{N}\) and \(\delta > 0\) there exists \(c > 0\) with the following properties. If \(b \leq c \left( \frac{n}{\log n} \right)^{1/\Delta}\) then Maker wins the \(H(n, \Delta)\)-universality game on \(K_n\) with bias \(b\).

If \(b \leq c \left( \frac{n}{\log n} \right)^{1/(2d+1)}\) then Maker wins the \(H(n, d, \Delta)\)-universality game on \(K_n\) with bias \(b\).

1.2.4. Resilience of low-bandwidth graphs. The Bollobás-Komlós conjecture, now the Bandwidth Theorem, proved by Böttcher, Schacht and Taraz [24], generalises (up to a small error term) several results dealing with spanning graphs in extremal graph theory. To state it, we need the concept of bandwidth \(bw(H)\), defined as the smallest natural \(b\) such that there is a labelling of the vertices of \(H\) by \(1, \ldots, v(H)\) such that \(|i-j| \leq b\) whenever \(ij\) is an edge in that ordering.

Theorem 1.8 (Bandwidth Theorem [24]). For each \(\gamma > 0\) and \(\Delta\) there exists \(\beta > 0\) such that for all sufficiently large \(n\) the following holds. If \(H\) is any \(n\)-vertex graph with \(\Delta(H) \leq \Delta\) and \(bw(H) \leq \beta n\), and \(G\) is any \(n\)-vertex graph with \(\delta(G) \geq (\frac{\chi(H)-1}{\chi(H)}) \cdot \gamma n\), then \(H \subseteq G\).

Böttcher, Kohayakawa and Taraz [22] proved a ‘resilience version’ of this theorem for almost-spanning bipartite graphs \(H\), working in typical random graphs \(G_{n,p}\) with \(p \geq C \left( \frac{\log n}{n} \right)^{1/\Delta}\). Huang, Lee and Sudakov [36] proved another resilience version which allows for non-bipartite \(H\), but only works in random graphs with \(p\) constant. We can prove the following common extension\(^3\) of these results.

Theorem 1.9. For each \(\gamma > 0\) and \(\Delta\) there exist \(\beta > 0\) and \(C\) such that if \(p \geq C \left( \frac{\log n}{n} \right)^{1/\Delta}\), then a.a.s. \(\Gamma = G_{n,p}\) has the following property. If \(H\) is any \((1-\gamma)n\)-vertex graph with \(\Delta(H) \leq \Delta\) and \(bw(H) \leq \beta n\), and \(G\) is any \(n\)-vertex subgraph of \(\Gamma\) with \(\delta(G) \geq (\frac{\chi(H)-1}{\chi(H)}) \cdot \gamma pn\), then \(H \subseteq G\).

In [1] the following strengthening of this result will be proved, showing that we can actually have \(v(H) = n - C p^{-2}\) if \(p \leq 1/\log n\), which is optimal, and for some \(H\) even \(v(H) = n\). Furthermore a similar resilience statement for sufficiently bijumbled graphs, and an improvement when \(H\) has small degeneracy, are obtained. It is significantly harder to obtain these stronger results, and the proof of Theorem 1.9 is a good illustration of how one can apply our blow-up lemmas to obtain resilience results, so we give the proof of Theorem 1.9 in Section 6.4.

\(^3\)Actually Huang, Lee and Sudakov allowed for spanning embeddings of some graphs \(H\), so that the result here is not quite an extension of their theorem, though the result in [1] is.
Theorem 1.10 (Allen, Böttcher, Ehrenmüller and Taraz [1]). For each \( \gamma > 0 \), \( \Delta \geq 2 \), and \( k \geq 1 \), there exist constants \( \beta > 0 \) and \( C > 0 \) such that the following holds a.a.s. for \( \Gamma = G(n,p) \) if \( p \geq C \left( \frac{\log n}{n} \right)^{1/\Delta} \). Let \( G \) be a spanning subgraph of \( \Gamma \) with \( \delta(G) \geq \left( \frac{\Delta - 1}{2} + \gamma \right) pn \), and let \( H \) be a \( k \)-colourable graph on \( n \) vertices with \( \Delta(H) \leq \Delta \), bandwidth at most \( \beta n \), and such that there are at least \( C \max\{p^{-2}, p^{-1} \log n\} \) vertices in \( V(H) \) that are not contained in any triangles of \( H \). Then \( G \) contains a copy of \( H \).

In [1] we also obtain an improvement on this (working for smaller values of \( p \)) for graphs \( H \) whose degeneracy is significantly less than their maximum degree, a version working when \( \Gamma \) is a bijumbled graph rather than a random one, and a resilience result for \( F \)-factors in \( G(n,p) \) which allows for much smaller \( p \). All these results depend upon the blow-up lemmas proved here; we refer the interested reader to [1] for the precise statements.

1.2.5. Robustness of the Bandwidth Theorem. Robustness is an alternative measure of 'how strongly' an extremal theorem holds, proposed by Krivelevich, Lee and Sudakov [54]. They showed that if \( G \) is a 'Dirac graph', i.e., a graph with \( \delta(G) \geq \frac{1}{2} \chi(G) \), then there is \( C \) such that for \( p \geq C \frac{\log n}{n} \), a.a.s. the graph \( G_p \) obtained by keeping edges of \( G \) independently with probability \( p \) is Hamiltonian. Here we show a similar result for the Bandwidth Theorem.

Theorem 1.11. For each \( \gamma > 0 \) and \( \Delta \geq 2 \) there exist \( \beta > 0 \) and \( C > 0 \) such that if \( p \geq C \left( \frac{\log n}{n} \right)^{1/\Delta} \), the following holds. If \( H \) is any \( n \)-vertex graph with \( \Delta(H) \leq \Delta \) and \( \text{bw}(H) \leq \beta n \), and \( G \) is any \( n \)-vertex graph with \( \delta(G) \geq \left( \frac{\chi(H) - 1}{\chi(H)} + \gamma \right)n \), then a.a.s. \( H \subseteq G_p \).

We give a fairly detailed sketch proof of this result in Section 6.5. We note that it also follows from the main result of [6]: in fact, the stronger statement (which does not follow from our sketch proof) that under the conditions of Theorem 1.11, \( G_p \) is universal for all \( n \)-vertex graphs \( H \) with \( \Delta(H) \leq \Delta \) and \( \text{bw}(H) \leq \beta n \) is a consequence of the result in [6].

1.3. Statements of the results

The full version of our blow-up lemmas are technically complex and so, in order to make their statement more compact, we will introduce a number of lengthy definitions. To provide motivation for these definitions we first state a simplified version of our blow-up lemma for random graphs, which is useful only in a few applications.

We will then provide the full version of the blow-up lemma for \( G_{n,p} \) in Section 1.3.2, and turn to a blow-up lemma for embedding degenerate graphs into subgraphs of \( G_{n,p} \) in Section 1.3.3. Finally, in Section 1.3.4 we present the blow-up lemma for bijumbled graphs.

1.3.1. A simplified version. The dense blow-up lemma is an embedding lemma for super-regular pairs. Before we can formulate it we need some definitions.

Let \( G = (V,E) \) be a graph and \( A \) and \( B \) be disjoint subsets of \( V \). We also write \( V(G) \) and \( E(G) \) for the set of vertices and edges of \( G \), respectively, and \( v(G) \) and \( e(G) \) for the sizes of these sets. For a vertex \( v \in V \) we denote by \( N_G(v;A) \) the set of neighbours of \( v \) in \( A \), that is, those vertices \( u \in A \) with \( vu \in E \). We write \( \deg_G(v;A) \) for its cardinality \( |N_G(v;A)| \). We write \( e_G(A) \) for the number of edges in \( G \) with both vertices in \( A \) and \( e_G(A,B) \) for the number of edges in \( G \) with one vertex in \( A \) and one in \( B \). Given \( X \subseteq V(G) \), we write \( N(X;A) := \bigcup_{v \in X} N_G(v;A) \) for the joint neighbourhood in \( A \) of the vertices. We let \( N^*_G(X;A) := \bigcap_{v \in X} N_G(v;A) \) denote
the common neighbourhood in $A$ of vertices from $X$, and write $\deg_G(X; A)$ for its size $|N^+_G(X; A)|$. We will often omit the set brackets in $N^+_G(v_1, \ldots, v_t; A)$, writing instead $N^+_G(v_1, \ldots, v_t; A)$. If the graph $G$ is clear from the context we sometimes omit it in the subscripts. Furthermore, if we omit the set $A$ we intend $A = V(G)$.

The density of the pair $(A, B)$ is $d_G(A, B) = e_G(A, B)/|A||B|$. Let $\varepsilon, d \geq 0$. The pair $(A, B)$ is called $(\varepsilon, d)$-regular (in $G$) if we have $d(A', B') \geq d - \varepsilon$ for all $A' \subseteq A$ with $|A'| \geq \varepsilon|A|$ and $B' \subseteq B$ with $|B'| \geq \varepsilon|B|$.

**Remark 1.12.** Observe that this differs from the usual definition of $\varepsilon$-regularity in that we only require a lower bound on $d(A', B')$. This is sometimes called dense in the literature, in particular in [44]. Note also that with our definition the density of an $(\varepsilon, d)$-regular pair $(A, B)$ is only lower-bounded by $d - \varepsilon$. This, though non-standard, has the advantage (over the usual definition where we would require the density of $(A, B)$ to be lower-bounded by $d$) that large subpairs of $(\varepsilon, d)$-regular pairs are $(\varepsilon', d)$-regular (rather than $(\varepsilon', d - \varepsilon')$-regular). We have to use a different regularity parameter, which is unavoidable, but at least we can stick to $d$ for density throughout, and we advocate the use of this definition in future.

We shall use (sparse versions of) this regularity concept whenever we work with random graphs in this paper. But for pseudorandom graphs we unfortunately need the usual regularity definition with an upper bound on $d(A', B')$ as well. Whenever the distinction between the two different concepts is essential, which is usually not the case, we will explicitly state which version we are using, calling the former ‘lower-regularity’ and the latter ‘full-regularity’. In most of the paper however, we will refer to both versions as ‘regularity’. This will help us to unify much of the proofs of the blow-up lemmas for random graphs and of pseudorandom graphs. Why it is necessary to use these two different regularity variants is explained in Section 1.3.4.

An $(\varepsilon, d)$-regular pair $(A, B)$ is called $(\varepsilon, d)$-super-regular if for every $u \in A$ we have $\deg_G(u; B) \geq (d - \varepsilon)|B|$ and for every $v \in B$ we have $\deg_G(v; A) \geq (d - \varepsilon)|A|$. The dense blow-up lemma, first proved by Komlós, Sárközy and Szemerédi [46], then states the following. Let $G$ be a graph formed from a collection of $(\varepsilon, d)$-super-regular pairs with density $d \gg \varepsilon$, and let $G^*$ be obtained from $G$ by replacing the super-regular pairs with complete bipartite graphs. If $H$ is a graph with maximum degree $\Delta$ which embeds into $G^*$, then $H$ embeds into $G$.

This notion of regularity is not meaningful for sparse graphs, because the definition above implies that any pair with $o(n^2)$ edges is regular (with $d = 0$). For obtaining a meaningful sparse version of regularity and super-regularity we need to relate the density of a pair to the overall density of the graph under study. This can be obtained the density in the definitions above with the $p$-density for a suitable $p$.

For $0 < p < 1$, the $p$-density of the pair $(A, B)$ is $d_{G,p}(A, B) = e_G(A, B)/(p|X||Y|)$. In other words, we simply scale the usual density by $p$. Sparse regular pairs are then defined as follows. The pair $(A, B)$ is $(\varepsilon, d, p)$-regular (in $G$) if we have $d_{G,p}(A', B') \geq d - \varepsilon$ for all $A' \subseteq A$ with $|A'| \geq \varepsilon|A|$ and $B' \subseteq B$ with $|B'| \geq \varepsilon|B|$. Similarly, we define a pair to be a sparse super-regular pair if it is a sparse regular pair and satisfies a minimum degree condition.

The setting we will work with in this subsection is when the graph $G$ into which we want to embed is a subgraph of a random graph $\Gamma$. As mentioned above, for pseudorandom graphs we will work with a slightly different notion of regularity (see Section 1.3.4). The parameter $p$ will usually be (a constant factor away from) the density of the ambient graph $\Gamma$.

**Definition 1.13** (Sparse super-regularity). A pair $(A, B)$ in $G \subseteq \Gamma$ is called $(\varepsilon, d, p)$-super-regular (in $G$) if it is $(\varepsilon, d, p)$-regular and for every $u \in A$ and $v \in B$
we have
\[
\deg_G(u; B) > (d - \varepsilon) \max\{p|B|, \deg_{\Gamma}(u; B)/2\}.
\]
\[
\deg_G(v; A) > (d - \varepsilon) \max\{p|A|, \deg_{\Gamma}(v; A)/2\}.
\]

We remark that the term \((d - \varepsilon)p|B|\) is a natural lower bound in the minimum degree condition (2) by the following fact, which easily follows from the definition of regularity (and which we shall use throughout the paper).

**Fact 1.14.** Let \((A, B)\) be an \((\varepsilon, d, p)\)-regular pair. Less than \(\varepsilon|A|\) vertices of \(A\) have less than \((d - \varepsilon)p|B|\) neighbours in \(B\).

Observe though that in (2) we have the additional term \((d - \varepsilon)\deg_{\Gamma}(u; B)/2\), which dominates when \(u\) has an exceptionally high \(\Gamma\)-degree into \(B\). The reason why this is necessary will become clear in a little while after we discuss the need for regularity inheritance. Note also that it follows from the definition above that in a super-regular pair \((A, B)\) we have \(\deg_{\Gamma}(u; B) \geq (d - \varepsilon)p|B|\) for each \(u \in A\), since \(G \subseteq \Gamma\).

Unfortunately, this straightforward extension of the concept of super-regular pairs to the sparse setting is not on its own enough for a sparse blow-up lemma. To see this, suppose \(A, B\) and \(C\) are equally-sized vertex sets, and each pair is super-regular. We would like to find that there is a spanning triangle factor, but it is possible that for some \(a \in A\) there are no edges between \(N_G(a; B)\) and \(N_G(a; C)\). Similarly, if \(A, B, C, D\) form a 4-cycle of super-regular pairs in \(G\), it is nevertheless possible that for some \(a \in A\) there is no 4-cycle in \(G\) using \(a\) and one vertex of each of the other three sets. These problems do not occur for dense \((\varepsilon, d, p)\)-super-regular pairs for the following reason. In the first example, since \(N_G(a; B)\) is of size at least \((d - \varepsilon)|B|\) and similarly for \(N_G(a; C)\), the so-called slicing lemma guarantees that \((N_G(a; B), N_G(a; C))\) is \((2\varepsilon/d, d)\)-regular, that is, it inherits regularity from the pair \((B, C)\) and thus in particular contains edges. A similar argument deals with the second example. The slicing lemma can be generalised to the sparse setting, and again easily follows from the definition of regularity.

**Lemma 1.15 (Slicing lemma).** Let \((A, B)\) be an \((\varepsilon, d, p)\)-regular pair and \(A' \subseteq A, B' \subseteq B\) be sets of sizes \(|A'| \geq \alpha|A|, |B'| \geq \alpha|B|\). Then \((A', B')\) is \((\varepsilon/\alpha, d, p)\)-regular.

But this unfortunately does not solve the problem indicated above because the size of \(N_G(a; B)\) is only lower bounded by \((d - \varepsilon)p|B|\), which can be tiny compared to \(B\).

To remedy this, in our blow-up lemmas we will explicitly require that the pair \((N_{\Gamma}(a; B), C)\), and similarly \((N_{\Gamma}(a; B), N_{\Gamma}(a; C))\), ‘inherit’ regularity, that is, they are themselves regular pairs. Since \(\deg_{G}(a; B)\) is a large fraction of \(\deg_{\Gamma}(a; B)\) by super-regularity, the slicing lemma does show that regularity of \((N_G(a; B), C)\) follows from regularity of \((N_{\Gamma}(a; B), C)\), and similarly for \((N_G(a; B), N_G(a; C))\). Note that this is the point where we require the second term in (2).

Requiring this ‘inheritance of regularity’ is reasonable because it is known that in \((\varepsilon, d, p)\)-regular pairs contained in random graphs \(\Gamma\), for almost all vertices \(a\), the pairs \((N_{\Gamma}(a; B), C)\) and similarly \((N_{\Gamma}(a; B), N_{\Gamma}(a; C))\) do inherit sparse regularity in this way. This phenomenon was studied in [43, 33, 44]. In this paper (see Section 1.3.5), we prove this statement with tight bounds on what ‘almost all vertices’ means.

**Definition 1.16 (Regularity inheritance).** Let \(A, B\) and \(C\) be vertex sets in \(G \subseteq \Gamma\), where \(A\) and \(B\) are disjoint and \(B\) and \(C\) are disjoint, but we do allow \(A = C\). We say that \((A, B, C)\) has one-sided \((\varepsilon, d, p)\)-inheritance if for each \(u \in A\) the pair \((N_{\Gamma}(u, B), C)\) is \((\varepsilon, d, p)\)-regular.

If in addition \(A\) and \(C\) are disjoint, then we say that \((A, B, C)\) has two-sided \((\varepsilon, d, p)\)-inheritance if for each \(u \in A\) the pair \((N_{\Gamma}(u, B), N_{\Gamma}(u, C))\) is \((\varepsilon, d, p)\)-regular.
1.3. Statements of the Results

In this definition we take neighbourhoods in the ambient graph \( \Gamma \) instead of \( G \) because this turns out to be easier to handle in applications (where we need to take care of the vertices in regular pairs that do not satisfy these inheritance conditions).

The simplified version of our sparse blow-up lemma in random graphs now states that if we are given an equitable vertex partition \( V_1 \cup \ldots \cup V_r \) of a graph \( G \subseteq \Gamma \) such that all pairs are super-regular and all triples have one- and two-sided regularity inheritance, then any \( r \)-partite graph \( H \) with bounded maximum degree that can be embedded into the complete \( r \)-partite graph with partition classes \( V_1, \ldots, V_r \) can also be embedded into \( G \). Here, a partition of a set is called equitable if each pair of partition classes differ in size by at most one.

**Lemma 1.17** (Simple blow-up lemma for \( G_{n,p} \)). For all \( \Delta \geq 2, r \in \mathbb{N} \) and \( d > 0 \) there exist \( \varepsilon > 0 \) and \( C \) such that if \( p \geq C(\log n/n)^{1/\Delta} \), then the random graph \( \Gamma = G_{n,p} \) a.a.s. has the following property. Let \( G \subseteq \Gamma \) and \( H \) with \( \Delta(H) \leq \Delta \) be graphs on \( n \) vertices. Suppose that \( H \) has an \( r \)-colouring with colour classes \( X_1, \ldots, X_r \) and that \( V(G) = V_1 \cup \cdots \cup V_r \) is an equitable partition with \( |V_i| \geq |X_i| \) for all \( i \in [r] \) and such that the following conditions hold.

(i) \( (V_i, V_j) \) is \((\varepsilon, d, p)\)-super-regular in \( G \) for each \( i, j \in [r] \) with \( i \neq j \).

(ii) \( (V_i, V_j, V_k) \) has one-sided \((\varepsilon, d, p)\)-inheritance for each \( i, j, k \in [r] \) with \( i \neq j \) and \( j \neq k \).

(iii) \( (V_i, V_j, V_k) \) has two-sided \((\varepsilon, d, p)\)-inheritance for each \( i, j, k \in [r] \) with \( i \neq j \), \( j \neq k \), and \( k \neq i \).

Then \( H \) is a subgraph of \( G \).

Let us briefly comment on the lower bound \( p \geq C(\log n/n)^{1/\Delta} \) on the probability for which our result works. We do not believe that this bound is in general best possible, though for \( \Delta \in \{2, 3\} \) it is optimal up to the log-factor. However, it matches the best known current lower bound \([29]\) for \( p \) such that \( G_{n,p} \) is universal for bounded degree spanning \( H \). This universality result is easily implied by Lemma 1.17 with \( G = \Gamma \). The problem of improving on \([29]\) has been prominent in random graph theory for a few years, and seems to be hard. For a more detailed discussion, see Section 7.1.

The restriction \( \Delta \geq 2 \) is necessary for our proof as written. The statement above is true for \( \Delta = 1 \), when in fact we do not require conditions (ii) or (iii), but to see this it is easiest to verify Hall’s condition in \( G \) directly rather than to modify our proof.

What is the difference between this simplified blow-up lemma and our full-strength blow-up lemma for random graphs, Lemma 1.21? Firstly, in the latter we do not require the partition of \( G \) to be equitable but allow the partition classes to differ in size by a constant factor.

Secondly, we do not require all pairs in the partition to be super-regular (or have regularity inheritance). Instead we will introduce the concept of a ‘reduced graph’ which encodes where we have super-regular pairs in our partition. In fact we will even have two reduced graphs \( R \) and \( R' \subseteq R \) where the former represents regular pairs and the latter super-regular pairs, the reason for which will become clear later.

Thirdly, we do not require two-sided regularity inheritance everywhere in the partition of \( G \), but only in certain cases where triangles of \( H \) need to be embedded. This is helpful in some applications, for example in Theorem 1.6 we can use the Ferber-Krivelevich-Naves strategy \([32]\) to win Maker-Breaker games with spanning bounded degree triangle-free graphs. This strategy does not allow Maker to win (for example) the spanning triangle factor game, ultimately because of a failure of two-sided regularity inheritance.
Fourthly, the $\varepsilon$ we can choose for our full-scale blow-up lemma does not depend on $r$, but only on the maximum degree of the reduced graph $R'$. This is a difference also in the dense setting (that is, when $p = 1$) to the blow-up lemma of [46]. In typical applications of the latter it is necessary to apply this blow-up lemma several times to embed a spanning $H$ since this lemma only applies to small subgraphs of the reduced graph (because $\varepsilon$ depends on $r$). It is then necessary to use so-called image restrictions and some manual embedding to connect up the pieces of $H$, which technically complicates the proofs. Our blow-up lemma on the other hand avoids this and is formulated with the intention that in applications it typically only needs to be applied once, which should make it simpler to use.

Finally, our blow-up lemma permits so-called image restrictions. Roughly speaking, image restriction means specifying, for certain vertices of $H$, small subsets of $G$ into which these vertices are to be embedded. As explained in the last paragraph we believe that in the dense setting one usually does not need these image restrictions with our blow-up lemma. However in the sparse setting they can be useful, as there may be a few vertices which do not satisfy the super-regularity or inheritance conditions required by the blow-up lemma in any sparse-regular partition of $G$. Hence we need to embed some $H$-vertices on these vertices ‘by hand’ before applying the blow-up lemma. These pre-embedded $H$-vertices then create image restrictions for their neighbours. Indeed in [1] precisely this approach is used to prove Theorem 1.10.

1.3.2. Random graphs. As explained in the previous subsection one of the differences of our blow-up lemma to the simplified version stated there is that it uses reduced graphs to specify where edges can be embedded in the graph $G$ (which is a subgraph of a random graph $\Gamma$). We will now first define this concept and explain what we require of the graph $H$ that we want to embed to be ‘compatible’ with such reduced graphs.

The setting in which we work is as follows. Let $G$ and $H$ be two graphs, on the same number of vertices, given with partitions $V = \{V_i\}_{i \in [r]}$ and $X = \{X_i\}_{i \in [r]}$ of their respective vertex sets. We call the parts $V_i$ of $G$ clusters. We say that $V$ and $X$ are size-compatible if $|V_i| = |X_i|$ for all $i \in [r]$. Moreover, for $\kappa \geq 1$ we say that $(G,V)$ is $\kappa$-balanced if there exists $m \in \mathbb{N}$ such that we have $m \leq |V_i| \leq \kappa m$ for all $i,j \in [r]$ (and thus for all $X_i \in X$ if $V$ and $X$ are size-compatible). Our goal will be to embed $H$ into $G$ respecting these partitions.

As mentioned before, we will have two reduced graphs $R$ and $R' \subseteq R$, where $R'$ represents super-regular pairs and $R$ regular pairs. More precisely, we require the following properties of $R$ and $R'$ and the partitions $V$ and $X$ of $G$ and $H$.

DEFINITION 1.18 (Reduced graphs and one-sided inheritance). Let $R$ and $R'$ be graphs on $r$ vertices.

- $(H,X)$ is an $R$-partition if each part of $X$ is empty, and whenever there are edges of $H$ between $X_i$ and $X_j$, the pair $ij$ is an edge of $R$,
- $(G,V)$ is an $(\varepsilon,d,p)$-regular $R$-partition if for each edge $ij \in R$ the pair $(V_i,V_j)$ is $(\varepsilon,d,p)$-regular.

In this case we also say that $R$ is a reduced graph of the partition $V$.

- $(G,V)$ is $(\varepsilon,d,p)$-super-regular on $R'$ if for every $ij \in E(R')$ the pair $(V_i,V_j)$ is $(\varepsilon,d,p)$-super-regular.

Suppose now that $(G,V)$ is an $(\varepsilon,d,p)$-regular $R'$-partition.

- $(G,V)$ has one-sided inheritance on $R'$ if $(V_i,V_j,V_k)$ has one-sided $(\varepsilon,d,p)$-inheritance for every $ij,jk \in E(R')$. 

We occasionally also use these concepts when we only work on an induced subgraph of $G$, that is, for a pair $(G, V')$ where $V'$ is a partition of a subset of the vertices of $G$.

As in Lemma 1.17, we do require the case $i = k$ in the definition of one-sided inheritance.

It remains to describe where in our partitions we require two-sided inheritance. For this we first need to define so-called ‘buffer sets’ of vertices in $H$, containing ‘potential buffer vertices’. The purpose of these buffer sets is that a subset of these vertices, later to be called ‘buffer vertices’, will be embedded last (for more detailed explanations see Section 1.4). For this to work we require the edges emanating from these vertices and their neighbours to be assigned to the super-regular pairs given by $R'$ (and not to other pairs of $R$). It is only when a potential buffer vertex is contained in a triangle that we require two-sided inheritance on $R'$.

The buffer sets can be chosen by the user of the blow-up lemma. Moreover, we stress that they do not make our blow-up lemma less powerful in the dense setting than the blow-up lemma of [46], because the latter requires all edges of $H$ to be assigned to super-regular pairs.

**Definition 1.19** (Buffer sets and two-sided inheritance). Let $R' \subseteq R$ be graphs on $r$ vertices, and $(H, \mathcal{X})$ be an $R$-partition and $(G, V)$ a size-compatible $(\varepsilon, d, p)$-regular $R$-partition. We say the family $\tilde{\mathcal{X}} = \{ \tilde{X}_i \}_{1 \leq i \leq r}$ of subsets $\tilde{X}_i \subseteq X_i$ is an $(\alpha, R')$-buffer for $H$ if

- $|\tilde{X}_i| \geq \alpha |X_i|$ for all $i \in [r]$ and
- for each $i \in [r]$ and each $x \in \tilde{X}_i$, the first and second neighbourhood of $x$ go along $R'$, that is, for each $xy, yz \in E(H)$ with $y \in X_i$ and $z \in X_k$ we have $ij \in R'$ and $jk \in R'$.

We also call the vertices in $\tilde{X}$ potential buffer vertices. Moreover, $(G, V)$ has two-sided inheritance on $R'$ for $\tilde{X}$ if

- $(V_i, V_j, V_k)$ has two-sided $(\varepsilon, d, p)$-inheritance whenever there is a triangle $x_i, x_j, x_k$ in $H$ with $x_i \in \tilde{X}_i$, $x_j \in \tilde{X}_j$, and $x_k \in \tilde{X}_k$.

We remark that we shall later also occasionally refer to the set of actual buffer vertices as buffer, when it is clear from the context which set we mean.

Finally, our blow-up lemma allows image restrictions. These generalise the image restrictions permitted in the dense blow-up lemma. However, in the sparse setting the necessary conditions become somewhat more involved. The idea is as follows. Suppose we wish to embed a graph $H^*$ into a graph $G^*$. Unfortunately $G^*$ does not meet the conditions of our blow-up lemma, typically because regularity inheritance fails. We find a subgraph $G$ of $G^*$ which does meet the conditions of our blow-up lemma, and ‘pre-embed’ some vertices of $H^*$ onto the vertices $V(G^*) \setminus V(G)$. This leaves the induced subgraph $H$ of $H^*$ to embed into $G$. The image restrictions then originate from these pre-embedded vertices: If $x \in X_i \subseteq V(H)$ has neighbours $\{z_1, \ldots, z_k\}$ in $V(H^*) \setminus V(H)$ which are pre-embedded to $\{u_1, \ldots, u_k\} = J_x \subseteq V(\Gamma) \setminus V(G)$, then $J_x$ restricts the embedding of $x$ to $I_x = N^G_{\tilde{\mathcal{X}}}(J_x; V_i)$. In the following definition we do not explicitly refer to the graphs $H^*$ and $G^*$, but only to abstract restricting sets $J_x$, so that we do not need to include the graphs $H^*$ and $G^*$ in our blow-up lemma. For the same reason we take neighbourhoods in $\Gamma$ instead of $G^*$ in this definition. In addition, to simplify notation, we define an image restriction set $I_x$ for each vertex $x$ of $H$. For most vertices $x$, however, this set is the trivial set $I_x = X_i$ where $X_i$ is the part of $X$ containing $x$.

**Definition 1.20** (Image restrictions). Let $R$ be a graph on $r$ vertices, and $(H, \mathcal{X})$ be an $R$-partition and $(G, V)$ a size-compatible $(\varepsilon, d, p)$-regular $R$-partition,
where $G \subseteq \Gamma$. Let $\mathcal{I} = \{I_x\}_{x \in V(H)}$ be a collection of subsets of $V(G)$, called image restrictions, and $\mathcal{J} = \{J_x\}_{x \in V(H)}$ be a collection of subsets of $V(\Gamma) \setminus V(G)$, called restricting vertices. We say that $\mathcal{I}$ and $\mathcal{J}$ are a $(\rho, \zeta, \Delta, \Delta_J)$-restriction pair if the following properties hold for each $i \in [r]$ and $x \in X_i$.

(a) The set $X_i^* \subseteq X_i$ of image restricted vertices in $X_i$, that is, vertices such that $I_x \neq V_i$, has size $|X_i^*| \leq \rho |X_i|$.

(b) If $x \in X_i^*$, then $I_x \subseteq N^*_x(J_x; V_i)$ is of size at least $\zeta(dp)^{|J_x|}|V_i|$.

(c) If $x \in X_i^*$, then $|J_x| + \deg_H(x) \leq \Delta$ and if $x \notin X_i^*$, then $J_x = \emptyset$.

(d) Each $\Gamma$-vertex appears in at most $\Delta_J$ of the sets of $\mathcal{J}$.

(e) We have $|N^*_x(J_x; V_i)| = (p \pm \varepsilon p)^{|J_x|}|V_i|$.

(f) If $x \in X_i^*$, for each $xy \in E(H)$ with $y \in X_j$, the pair $(N^*_x(J_x; V_i), N^*_y(J_y; V_j))$ is $(\varepsilon, d, p)$-regular in $G$.

This definition does indeed generalise the dense image restrictions of [46], since (a) is one of the conditions of [46], while if $p = 1$ and so $\Gamma$ is the complete graph, (b) reduces to the other condition of [46]. We can set $J_x = \emptyset$ for all $x$, so that (c)–(e) become trivial, and (f) follows since $\mathcal{X}$ is an $R$-partition and $\mathcal{V}$ is an $(\varepsilon, d, p)$-regular $R$-partition. In the sparse setting the conditions amount to requiring that pre-embedded vertices creating image restrictions are embedded on ‘typical vertices’.

We can now formulate our blow-up lemma for random graphs.

**Lemma 1.21 (Blow-up lemma for $G_{n,p}$).** For all $\Delta \geq 2$, $\Delta_R$, $\Delta_J$, $\alpha, \zeta, d > 0$, $\kappa > 1$ there exist $\varepsilon, \rho > 0$ such that for all $r_1$ there is a $C$ such that for

$$p > C \left( \frac{\log n}{n} \right)^{1/\Delta}$$

the random graph $\Gamma = G_{n,p}$ a.a.s. satisfies the following.

Let $R$ be a graph on $r \leq r_1$ vertices and let $R' \subseteq R$ be a spanning subgraph with $\Delta(R') \leq \Delta_{R'}$. Let $H$ and $G \subseteq \Gamma$ be graphs with $\kappa$-balanced size-compatible vertex partitions $X = \{X_i\}_{i \in [r]}$ and $V = \{V_i\}_{i \in [r]}$, respectively, which have parts of size at least $m \geq n/(\kappa r_1)$. Let $\mathcal{X} = \{\tilde{X}_i\}_{i \in [r]}$ be a family of subsets of $V(H)$, $\mathcal{I} = \{I_x\}_{x \in V(H)}$ be a family of image restrictions, and $\mathcal{J} = \{J_x\}_{x \in V(H)}$ be a family of restricting vertices. Suppose that

- **(BUL1)** $\Delta(H) \leq \Delta$, $(H, \mathcal{X})$ is an $R$-partition, and $\tilde{X}$ is an $(\alpha, R')$-buffer for $H$,
- **(BUL2)** $(G, \mathcal{V})$ is an $(\varepsilon, d, p)$-regular $R$-partition, which is $(\varepsilon, d, p)$-super-regular on $R'$, has one-sided inheritance on $R'$, and two-sided inheritance on $R'$ for $\tilde{X}$,
- **(BUL3)** $\mathcal{I}$ and $\mathcal{J}$ form a $(\rho, \zeta, \Delta, \Delta_J)$-restriction pair.

Then there is an embedding $\psi: V(H) \rightarrow V(G)$ such that $\psi(x) \in I_x$ for each $x \in H$.

**1.3.3. Degenerate graphs.** We next present a version of our blow-up lemma for random graphs which allows for smaller edge probabilities in $G_{n,p}$ if we want to embed graphs $H$ whose maximum degree is much larger than their degeneracy. The degeneracy $\text{deg}(H)$ of a graph $H$ is the smallest integer $\ell$ such that $H$ is $\ell$-degenerate, that is, each induced subgraph of $H$ has minimum degree at most $\ell$. Equivalently, there is an order of $V(H)$ such that each vertex has at most $\ell$ neighbours before that vertex in the order. For example, trees have degeneracy 1 and planar graphs have degeneracy at most 5. We remark that (a variant of) the degeneracy determines the exponent in the probability $p$, but we nevertheless require the maximum degree of $H$ to be bounded by a constant $\Delta$ in the lemma below. The constant in the probability $p$ depends on $\Delta$. For this blow-up lemma we use the same notion of regularity, super-regularity, reduced graphs, inheritance, buffer sets and image restrictions as for Lemma 1.21.
Given an order $\tau$ on $V(H)$ and a family $\mathcal{J}$ of image restricting vertices, we define
$$\pi^\tau(x) := |J_x| + \left|\{y \in N_H(x) : \tau(y) < \tau(x)\}\right|.$$ 
In other words, $\pi^\tau(x)$ denotes the number of neighbours of $x$ that come before $x$, including the restricting vertices $J_x$.

Quantifying the dependence on $\tau$ of the probability we can work with is somewhat complex. Firstly, for a vertex $x$ we distinguish whether it has neighbours $y$ succeeding $x$ in the order $\tau$, in which case we will need to maintain one-sided inheritance properties when embedding $x$, or even two such neighbours $y$ and $z$ which form an edge, in which case we will need to maintain two-sided inheritance, or not.

Secondly, we put stricter requirements on $x$ if it has some potential buffer vertices in its neighbourhood (since we need to maintain certain properties to embed the buffer vertices; see $(\text{ORD1})$ and $(\text{ORD3})$). Thirdly, we put even stricter requirements on vertices $x$ which are image restricted or have preceding neighbours far away from $x$ (see $(\text{ORD2})$). Since this is a very restrictive condition however, we allow a very small set $X^e$ of exceptional vertices which are exempted from this rule.

**Definition 1.22** ($(D,p,m)$-bounded order). Let $H$ be a graph given with buffer sets $\tilde{X}$ and a restriction pair $\mathcal{I} = \{I_i\}_{i \in [r]}$ and $\mathcal{J} = \{J_i\}_{i \in [r]}$. Let $\tilde{X} \cup X$. Let $\tau$ be an ordering of $V(H)$ and $X^e \subseteq V(H)$. Then $\tau$ is a $(D,p,m)$-bounded order for $H$, $\tilde{X}$, $\mathcal{I}$ and $\mathcal{J}$ with exceptional set $X^e$ if the following conditions are satisfied for each $x \in V(H)$.

1. If $\pi^\tau(x) \leq D_x$, and if $x \in N(\tilde{X})$ even $\pi^\tau(x) \leq D_x - 1$. Finally, if $x \in \tilde{X}$ we have $\deg(x) \leq D$.

2. One of the following holds:
   - $x \in X^e$.
   - $\pi^\tau(x) \leq \frac{1}{2}D$.
   - $x$ is not image restricted and every neighbour $y$ of $x$ with $\tau(y) < \tau(x)$ satisfies $\tau(x) - \tau(y) \leq p^{\pi^\tau(y)m}$.

3. If $x \in N(\tilde{X})$ then all but at most $D - 1 - \max_{x \in X^e} \pi^\tau(z)$ neighbours $y$ of $x$ with $\tau(y) < \tau(x)$ satisfy $\tau(x) - \tau(y) \leq p^{Dm}$.

In order to obtain the best possible value of $p$, our aim is always to find an order $\tau$ which is $(D,p,m)$-bounded and minimises the value of $D$. To give some intuition about what typically is possible, we refer to some of the results of Section 1.2.

We can typically obtain $D \leq 2 \deg(H) + 1$, taking $\tau$ to be a degeneracy order for $H$. This gives us $\pi^\tau(x) \leq \deg(x)$, so that $(\text{ORD2})$ and $(\text{ORD3})$ hold trivially (the former with room to spare, the latter not). Furthermore, most of $(\text{ORD1})$ is trivially satisfied (though one needs to observe that if $\deg(H) = 1$ then $H$ contains no triangle). The only point which is unclear is the restriction $\deg(x) \leq D$ for $x \in \tilde{X}$. In practice, as seen in Theorems 1.1 and 1.6, in applications one often can obtain this. The reason is that any graph $H$ contains many vertices of degree at most $2 \deg(H)$, and typically we can find a partition of $H$ in which these vertices are well-distributed among the parts.

In the event that we do not require a spanning embedding, but can afford to leave a small fraction of vertices in each part uncovered, we can obtain the slightly stronger $D \leq 2 \deg(H)$, as in Theorems 1.4 and 1.6. The reason for this is that
we can ‘pad’ an almost-spanning $H$ by adding isolated vertices to obtain a spanning $H'$ to which we apply Lemma 1.23 (see the proof of Theorem 1.4 for details). We again use a degeneracy order $\tau$, but this time let $X$ be the isolated vertices.

When we have a degeneracy order $\tau$ with the extra (bandwidth-type) property that all edges go between vertices very close together in the order, we can typically choose $D \leq \text{deg}(H)+3$. The reason is that in this case $(\text{ORD2})$ and $(\text{ORD3})$ are automatically satisfied, and we only need to worry about $(\text{ORD1})$. Again, we need to be able to choose potential buffer vertices of degree at most $\text{deg}(H)+3$, but in applications this is usually possible. This applies, for example, in the case that $H$ is an $F$-factor.

So far we did not mention the set $X^e$, or image restrictions. In applications often only very few vertices need to be image restricted, and we can typically put all of them in $X^e$. This is, for example, critical to obtaining a resilience result for bounded-degree trees with $p = C \left( \frac{\log n}{n} \right)^{1/3}$ in [1].

In general, the exceptional set $X^e$ gives us the possibility to specify a small set of vertices in $H$ for which the value of $\pi^\tau$ does not have to be bounded by $D/2$. This is for example useful when a few vertices have been embedded ‘by hand’ before the use of the blow-up lemma, creating image restrictions. Since we cannot usually select these vertices according to the degeneracy order of $H$, as a consequence a few image restricted vertices will have more neighbours embedded ‘by hand’ than if $H$ was embedded in the degeneracy order. These vertices should then be put in $X^e$.

Moreover, in $(\text{ORD3})$ for each buffer neighbour $x$ we allow a few (depending on $D$ and $\tau$) exceptions to the rule that embedded neighbours of $x$ need to be close to $x$ in the order. This is useful because vertices $y$ of exceptionally high degree often have to come relatively early in the order to satisfy $(\text{ORD1})$ and $(\text{ORD2})$, and hence cannot necessarily be close to $x$.

We can now state our blow-up lemma for degenerate graphs. The only difference to the previous blow-up lemma is that we ask for a $(D, p, \varepsilon n/r_1)$-bounded order of $H$, that we allow only vertices with degree at most $D$ as potential buffer vertices, and that the exponent in the bound on $p$ is determined by $D$. We remark that for some graphs $H$, for example $\ell$-regular graphs, which have degeneracy $\ell$, this bound on $p$ is worse than the bound in Lemma 1.21. However for trees or planar graphs it is often much better.

**Lemma 1.23** (Blow-up lemma for $G_{n,p}$ to embed degenerate graphs). For all $\Delta \geq 2$, $\Delta R_1$, $\Delta J$, $D$, $\alpha, \zeta, d > 0$, $\kappa > 1$ there exist $\varepsilon, \rho > 0$ such that for all $r_1$ there is a $C$ such that for

$$p \geq C \left( \frac{\log n}{n} \right)^{1/D}$$

the random graph $\Gamma = G_{n,p}$ a.a.s. satisfies the following.

Let $R$ be a graph on $r \leq r_1$ vertices and let $R' \subseteq R$ be a spanning subgraph with $\Delta(R') \leq \Delta R$. Let $H$ and $G \subseteq \Gamma$ be graphs with $\kappa$-balanced, size-compatible vertex partitions $X = \{X_i\}_{i \in [r]}$ and $V = \{V_i\}_{i \in [r]}$, respectively, which have parts of size at least $m \geq n/(\kappa r_1)$. Let $X = \{X_i\}_{i \in [r]}$ be a family of subsets of $V(H)$, $I = \{I_x\}_{x \in V(H)}$ be a family of image restrictions, and $J = \{J_x\}_{x \in V(H)}$ be a family of restricting vertices. Let $\tau$ be an order of $V(H)$ and $X^e \subseteq V(H)$ be a set of size $|X^e| \leq \varepsilon p^{\max_{x \in X^e} \pi^\tau(x)} n/r_1$. Suppose that

(DBUL1) $\Delta(H) \leq \Delta$, $(H, X)$ is an $R$-partition, and $\bar{X}$ is an $(\alpha, R')$-buffer for $H$,  
(DBUL2) $(G, V)$ is an $(\varepsilon, d, p)$-regular $R$-partition, which is $(\varepsilon, d, p)$-super-regular on $R'$, has one-sided inheritance on $R'$, and two-sided inheritance on $R'$ for $\bar{X}$, 
(DBUL3) $I$ and $J$ form a $(\rho, \zeta, \Delta J, \Delta J)$-restriction pair.
(DBUL4) \( \tau \) is a \( (D, p, \varepsilon n/r_1) \)-bounded order for \( H, \tilde{X}, I, J \) with exceptional set \( X^\varepsilon \). Then there is an embedding \( \psi : V(H) \to V(G) \) such that \( \psi(x) \in I_x \) for each \( x \in H \).

Let us briefly indicate how this blow-up lemma performs in practice. As we see in for example Theorem 1.6, we can use it to obtain spanning embeddings in some situations with \( D = 2 \text{degen}(H) + 1 \), thus with \( p \geq C \left( \frac{\log n}{n} \right)^{1/(2 \text{degen}(H)+1)} \). In [1] we obtain the promised version of Theorem 1.10 for degenerate graphs, using the same value of \( D \). However to obtain this result, we have to pre-embed some vertices of \( H \), before applying Lemma 1.23, and we thus have to make some alterations to the degeneracy order to obtain an order \( \tau \). The neighbours of the pre-embedded vertices then threaten to destroy \((D, p, m)\)-boundedness, and it is critical that we can put them into \( X^\varepsilon \).

As a second example, if the degeneracy order (or something close to it) on \( H \) happens to have the property that all edges go only a short distance, then conditions \((\text{ORD3})\) and \((\text{ORD2})\) become trivially true, and we can obtain much stronger results. For example, in [1] we give a resilience result for \( F\)-factors in \( G(n, p) \) with \( p \geq C \left( \frac{\log n}{n} \right)^{1/(\text{degen}(F)+3)} \), that is, using Lemma 1.23 with \( D = \text{degen}(F) + 3 \). In this application we do not need any exceptions, and can set \( X^\varepsilon = \emptyset \).

### 1.3.4. Bijumbled graphs

Finally, we provide a blow-up lemma for embedding bounded degree graphs into subgraphs of sufficiently bijumbled graphs. As indicated earlier, in bijumbled graphs we use a stronger notion of regularity, which in addition to the lower bound also requires an upper bound on the edge density of subpairs.

**Definition 1.24 (Regularity in bijumbled graphs).** In bijumbled graphs, we say that \((X, Y)\) is \((\varepsilon, d, p)\)-regular if there is a \(d' \geq d\) such that for any \(X' \subseteq X\) with \(|X'| \geq \varepsilon |X|\) and \(Y' \subseteq Y\) with \(|Y'| \geq \varepsilon |Y|\), we have \(d_p(X', Y') = d' \pm \varepsilon\). When we want to be it clear that we are working with this regularity concept we also call such a pair \((\varepsilon, d, p)\)-fully-regular.

The reason why we use lower-regularity in random graphs is that the regularity inheritance we use in random graphs (see Section 2.1.3) provides only lower-regular pairs. On the other hand, we use full-regularity in bijumbled graphs because the regularity inheritance we prove for bijumbled graphs (again, see Section 2.1.3) requires fully-regular pairs. Unfortunately, we do not know, in either case, how to prove a regularity inheritance statement which works with the ‘other’ version of regularity.

All other parts of our proofs work for both lower-regularity and full-regularity. Since we would like to use many of these parts for random graphs as well as for bijumbled graphs, we let regular pairs mean lower-regular pairs whenever we work in random graphs, and fully-regular pairs whenever we work in bijumbled graphs.

In particular, super-regularity, reduced graphs, inheritance, buffer sets and image restrictions for bijumbled graphs are defined exactly as for random graphs, once the regularity concept there is replaced with the regularity concept defined here. Our blow-up lemma for bijumbled graphs then has analogous requirements and conclusions as Lemma 1.21, with the only exception that the image restrictions allowed here are much weaker: if \( H \) has maximum degree \( \Delta \) we can only image restrict about a \( pA \)-fraction of the vertices in any given partition class of \( H \), rather than a small constant fraction.

**Lemma 1.25 (Blow-up Lemma for bijumbled graphs).** For all \( \Delta \geq 2, \Delta_{R'}, \Delta_J, \alpha, \zeta, d > 0, \kappa > 1 \) there exist \( \varepsilon, \varrho > 0 \) such that for all \( r_1 \) there is a \( c > 0 \) such that if \( p > 0 \) and

\[
\beta \leq cp^{\alpha \max(4, (3\Delta + 1)/2)}r_1
\]
any \((p, \beta)\)-bijumbled graph \(\Gamma\) on \(n\) vertices satisfies the following.

Let \(R\) be a graph on \(r \leq r_1\) vertices and let \(R' \subseteq R\) be a spanning subgraph with \(\Delta(R') \leq \Delta_R\). Let \(H\) and \(G\) be graphs given with \(\kappa\)-balanced, size-compatible vertex partitions \(X = \{X_i\}_{i \in [r]}\) and \(V = \{V_i\}_{i \in [r]}\), respectively, which have parts of size at least \(m \geq n/(\kappa r_1)\). Let \(\tilde{X} = \{\tilde{X}_i\}_{i \in [r]}\) be a family of subsets of \(V(H)\), \(\mathcal{I} = \{I_x\}_{x \in V(H)}\) be a family of image restrictions, and \(\mathcal{J} = \{J_x\}_{x \in V(H)}\) be a family of restricting vertices. Suppose that

\[(JBUL1)\] \(\Delta(H) \leq \Delta\), \(\langle H, X\rangle\) is an \(R\)-partition, and \(\tilde{X}\) is an \((\alpha, R')\)-buffer for \(H\),

\[(JBUL2)\] \((G, Y)\) is an \((\varepsilon, d, p)\)-regular \(R\)-partition, which is \((\varepsilon, d, p)\)-super-regular on \(R'\), and has one-sided inheritance on \(R'\), and two-sided inheritance on \(R'\) for \(\tilde{X}\).

\[(JBUL3)\] \(\mathcal{I}\) and \(\mathcal{J}\) form a \((gp^\Delta, \zeta, \Delta, \Delta_f)\)-restriction pair.

Then there is an embedding \(\psi\) : \(V(H) \rightarrow V(G)\) such that \(\psi(x) \in I_x\) for each \(x \in H\).

We remark that we do not believe that the bound on \(\beta\) in this result is even close to being best possible (this is discussed further in the concluding remarks, Section 7.1.2). However, this blow-up lemma is the first general embedding result which allows the embedding of spanning structures in bijumbled graphs (see for example Theorem 1.2). Previously such embedding results were known only for some special subgraphs \(H\), such as Hamilton cycles [55], triangle factors [56], or powers of Hamilton cycles [3]. Moreover, our blow-up lemma allows for resilience results in pseudorandom graphs. Previously, the only results for large subgraphs in this direction that we are aware of deal with cycles. Specifically, Sudakov and Vu [65] found the local resilience of \((n, d, \lambda)\)-graphs with respect to Hamiltonicity, Dellamonica, Kohayakawa, Marciniszyn, and Steger [28] found the local resilience of bijumbled graphs with respect to containing long cycles, and Krivelevich, Lee and Sudakov [53], with a stronger bijumbledness requirement, found the local resilience with respect to pancyclicity (containing cycles of all lengths).

1.3.5. Inheritance of regularity. The final results we would like to highlight are the regularity inheritance lemmas for random graphs we mentioned earlier. These are necessary not only in many applications of our sparse blow-up lemmas, such as [1], but are also useful in other random graph contexts, for example in [2, 7].

Both statements rely crucially on the regularity inheritance work of Gerke, Kohayakawa, Rödl and Steger [33], and also use ideas from [44].

**Lemma 1.26** (One-sided lower-regularity inheritance in \(G_{n,p}\)). For each \(0 < \varepsilon' < d\) there are \(\varepsilon_0 > 0\) and \(C\) such that for all \(0 < \varepsilon < \varepsilon_0\) and \(0 < p < 1\), a.a.s. \(\Gamma = G_{n,p}\) has the following property. Let \(G \subseteq \Gamma\) be a graph and \(X, Y\) be disjoint subsets of \(V(\Gamma)\). If \((X, Y)\) is \((\varepsilon, d, p)\)-lower-regular in \(G\) and

\[|X| \geq C \max(p^{-2}, p^{-1} \log n)\quad\text{and}\quad|Y| \geq C p^{-1} \log n,\]

then for at most \(Cp^{-1} \log n\) vertices \(z \in V(\Gamma)\) the pair \((N_{\Gamma}(z; X), Y)\) is not \((\varepsilon', d, p)\)-lower-regular in \(G\).

**Lemma 1.27** (Two-sided lower-regularity inheritance in \(G_{n,p}\)). For each \(0 < \varepsilon' < d\) there are \(\varepsilon_0 > 0\) and \(C\) such that for all \(0 < \varepsilon < \varepsilon_0\) and \(0 < p < 1\), a.a.s. \(\Gamma = G_{n,p}\) has the following property. Let \(G \subseteq \Gamma\) be a graph and \(X, Y\) be disjoint subsets of \(V(\Gamma)\). If \((X, Y)\) is \((\varepsilon, d, p)\)-lower-regular in \(G\) and

\[|X|, |Y| \geq C \max(p^{-2}, p^{-1} \log n),\]

then for at most \(C \max(p^{-2}, p^{-1} \log n)\) vertices \(z \in V(\Gamma)\) the pair \((N_{\Gamma}(z; X), N_{\Gamma}(z; Y))\) is not \((\varepsilon', d, p)\)-lower-regular in \(G\).
These two results are quite similar to [44, Proposition 15]. The crucial difference are the bounds on the sizes of $X$ and $Y$ as well as the bound on the number of vertices $z$ that do not preserve regularity. We remark that the bounds given in [44, Proposition 15] are (roughly) equivalent to our bounds when $p$ is as small as possible in our blow-up lemmas, that is $p = \Theta((\log n/n)^{1/\Delta})$, but for bigger $p$ our results are stronger.

We would like to stress that these results give sharp bounds, up to the constant $C$, on the number of vertices which may fail to inherit lower-regularity. for any $p < \frac{1}{2}$. We prove both theorems, and that they are sharp, in Section 2.1.3.

It is worth noting that the construction showing that $\Omega(p^{-1} \log n)$ vertices may fail to inherit lower-regularity in both lemmas uses a lower-regular pair $(X, Y)$ in which $|X| = o(n)$. It would be particularly interesting to know if there is a version of Lemma 1.27 in which we insist on $|X|, |Y| = \Omega(n)$ but can strengthen the conclusion by saying that only $Cp^{-2}$ vertices fail to inherit lower-regularity. If such a statement is true, we would obtain correspondingly stronger bounds in Theorem 1.10.

Finally, we note that corresponding inheritance lemmas (which take as input and give as output fully-regular pairs) for bijumbled graphs were first proved by Conlon, Fox and Zhao [25]. The improved versions of these lemmas which we state in Section 2.1.3 were proved in [5].

1.4. Proof overview

The following is a high-level overview of the proofs of our blow-up lemmas. In these proofs we merge ideas that were also used in proofs of dense blow-up lemmas [23, 46, 61] and of the sparse embedding lemma in [44] with many new ingredients. Of course there are differences between the proofs of Lemmas 1.21, 1.23, and 1.25 but at the level of this overview we will avoid mentioning most of them.

In each of our blow-up lemmas we want to embed a graph $H$, which is given with a partition $X$ and potential buffer sets $\tilde{X}$, into a graph $G \subseteq \Gamma$ with a compatible super-regular partition $V$, possibly with some image restrictions, and where $\Gamma$ is a sparse random or bijumbled graph. To avoid technical details, we will in this overview largely ignore the image restrictions; these turn out not to play a large role in the proofs. Our embedding strategy for $H$ is comprised of different randomised embedding procedures. Before we can apply them though, we need to prepare the graphs $H$ and $G$ by subpartitioning their partition classes suitably (details of what we need can be found in Sections 2.3.2 and 2.3.3).

We start by subpartitioning the partition classes of $H$ such that any pair of vertices in a part of the new partition is at distance at least ten, which is possible by a trick first used by Alon and Füredi [15], and in a blow-up lemma setting by Rödl and Ruciński [61], that relies on the Hajnal-Szemerédi Theorem (Theorem 2.3). Having only distant vertices in a part will provide sufficient independence of these vertices in our randomised embedding procedures. This idea is also used in [23].

Next, we subpartition each cluster $V_i$ of $G$ randomly into several parts

$$V_i = V_i^{\text{main}} \cup V_i^{\text{buf}} \cup V_i^{\text{out}} \cup V_i^{\text{in}}.$$ 

The first of these sets is large (of size $(1 - 3\mu)|V_i|$ for some small $\mu$), while the remaining three are much smaller (of size $\mu|V_i|$). Because the subpartitionings were performed randomly, subparts of super-regular pairs (given by $R^i$) maintain super-regularity. The reason for doing this is that we perform the embedding in stages, and these stages require separate parts. This idea is also used in [23].

We also partition $X_i$ into a large part $X_i^{\text{main}}$ (of size at most $(1 - 4\mu)|X_i|$) and a small part $X_i^{\text{buf}}$ (of size $4\mu|X_i|$), called the set of buffer vertices. The latter set is
required to be a subset of the potential buffer vertices $\bar{X}_i$ in $X_i$. We also require various extra properties of $X_i^{\text{buf}}$. In the proof of the random graphs blow-up lemma, Lemma 1.21, we also include another small subpart $X_i^c$ in this partitioning step. This is due to the fact that in this proof we have to treat the case that most of the vertices $X_i$ are contained in copies of $K_{\Delta+1}$ specially (in order to obtain the claimed bound on the probability $p$). We need the sets $V_i^c$ to embed this small subpart.

At this point $H$ and $G$ are prepared for the embedding. We now describe our randomised embedding approach. Firstly, we make use of a randomised greedy algorithm (RGA) to embed $X_i^{\text{main}}$ into $V_i^{\text{main}}$ (see Section 3.2 for the simplest version used in the proof of Lemma 1.21). An algorithm of this type was also used by Komlós, Sárközy and Szemerédi [46] to prove the dense blow-up lemma. The RGA embeds the sets $X_i^{\text{main}}$ vertex by vertex, in each step avoiding some bad vertices of $G$ and embedding the current vertex $x \in X_i^{\text{main}}$ into its so-called candidate set. If $\psi$ is the partial embedding of $H$ into $G$ that we have constructed so far, then the candidate set $C(x) \subseteq V_i$ is the set of vertices adjacent in $G$ to each $\psi(y)$ such that $y$ is an embedded neighbour of $x$ in $H$. However, some of the vertices in $C(x)$ may have been used as images for other vertices already, so we let $A(x) = C(x) \setminus \text{Im}(\psi)$ be the available candidate set for $x$. Obviously, $x$ has to be embedded into $A(x)$ in order to obtain an embedding.

In order to succeed with this strategy, we have to maintain certain properties for the partial embedding $\psi$, which we call good partial embedding properties (see Section 2.3.5). We remark that the properties of a restriction pair ensure that the trivial partial embedding, which we have at the beginning when no vertices of $H$ are embedded yet, is a good partial embedding. This fact means that in the proofs we usually do not need to distinguish between vertices which are and are not image restricted (and it also justifies that we ignore image restrictions in this overview).

Among other properties, such as regularity properties, in a good partial embedding we require that for each unembedded $x \in X_i$ the set $C(x)$ is large, that is of size $\Omega(d'p'|V_i^{\text{main}}|$) where $\ell$ is the number of already embedded neighbours of $y$. In order to maintain this property we need to avoid certain bad vertices $B(x)$ when embedding $x$. More precisely, when we embed $x$ this leads to a change of the candidate sets $C(y)$ of unembedded neighbours $y$ of $x$. The set $B(x)$ contains the vertices $v \in V_i$ that would lead to some $C(y)$ becoming small (or have some other bad properties that would prevent us from maintaining a good partial embedding; see Section 2.3.7). The RGA then embeds $x$ uniformly at random into $V_i^{\text{main}} \cap A(x) \setminus B(x)$.

That we choose the images randomly helps us in several ways. Most immediately, we can use it to show that the sets $C(y)$ for unembedded vertices $y \in X_i^{\text{main}} \cup X_i^{\text{buf}}$ are ‘uniformly’ distributed over $V_i^{\text{main}}$. In particular, this means that they tend not to be contained entirely in $\text{Im}(\psi)$, which, together with the fact that $B(y)$ is always small, implies that the set $V_i^{\text{main}} \cap A(y) \setminus B(y)$ to which we wish to embed a future $y \in X_i^{\text{main}}$ is usually not small.

Unfortunately though, the RGA will not succeed in embedding every vertex of $X_i^{\text{main}}$, because occasionally we will come across a vertex $x$ such that $V_i^{\text{main}} \cap A(x) \setminus B(x)$ is small. We put such vertices $x$ in a queue $X_i^q \subseteq X_i^{\text{main}}$. The use of such a queue appears already in [46]. The ‘uniform’ distribution of the sets $C(x)$ allows us to show that $X_i^q$ remains much smaller than the so far untouched $V_i^n$. We will show that this allows us to embed $X_i^q$ into $V_i^n$, maintaining a good partial embedding. We note that in the proof of the random graphs blow-up lemma, Lemma 1.21, we perform this embedding after we have embedded all other vertices of $X_i^{\text{main}}$, using a matching strategy similar to that of Kohayakawa, Rödl, Schacht and Szemerédi [44]. In the proofs of the other two blow-up lemmas, Lemma 1.23 and 1.25, however,
we embed the vertices of $X^{\text{main}}$ in the given order, again using a random greedy strategy. In either case, the underlying idea is to show that the only reason why we might fail to embed $X_i^{q}$ into $V_i^{q}$ is that $G$ contains a ‘dense spot’, contradicting the fact that our random or bijumbled $\Gamma$ does not contain such dense spots.

At this point, all vertices of $X^{\text{main}}$ are embedded, and it remains to embed the carefully chosen buffer vertices $X^{\text{buf}}_i$. Since any two buffer vertices are at distance at least 10 in $H$, in particular all neighbours of buffer vertices have been embedded, so the candidate set $C(x)$ of a buffer vertex $x$ will not change anymore, and it suffices to find a system of distinct representatives for the available candidate sets $A(x)$, which we achieve by verifying Hall’s condition. Thus, for each set $Y \subseteq X^{\text{buf}}_i$, we need to show that the union $W$ of the available candidate sets for $y \in Y$ satisfies $|W| \geq |Y|$. We separate three cases: $Y$ is either a small fraction of $X^{\text{buf}}_i$, or most of $X^{\text{buf}}_i$, or somewhere intermediate. In the first case we use essentially the same argument as when embedding the queue vertices. In the last case we show that the ‘uniform’ distribution of the sets $C(x)$ for $x \in X^{\text{buf}}_i$ implies that $W$ is most of $V_i \setminus \text{Im}(\psi)$.

To handle the remaining case when $Y$ contains most or all of $X^{\text{buf}}_i$, we show that the RGA gives us an extra property: for every vertex $v \in V_i$ there are many vertices $x \in X^{\text{buf}}_i$ such that $v$ is a candidate for $x$. This is the point in the proof where we use the super-regularity and inheritance properties that $G$ satisfies on the subgraph $R'$ of $R$. We also remark that in the case of the random graphs blow-up lemma, Lemma 1.21, we cannot in general establish this property: When most of the vertices in $X_i$ are contained in copies of $K_{\Delta+1}$ we obtain the claimed feature only for most vertices instead of all vertices $v \in V_i$. Hence in this case, to recover the special property, we perform an additional embedding stage, using the sets $X^{c}_i$ and $V^{c}_i$, to fix these buffer defects (see Section 3.4). This additional embedding stage also uses the super-regularity and inheritance properties of $R'$. We do not use these properties elsewhere in the proof. Finally, using again that $\Gamma$ and hence $G$ does not have ‘dense spots’ we can show that the above described extra property implies $|W| \geq |Y|$ as desired.

Summarising, our blow-up lemma proofs contain three main embedding procedures: the random greedy algorithm, the queue embedding, and the embedding of the buffer vertices. In the proof of Lemma 1.21 we also perform an additional embedding procedure for fixing buffer defects.

One important difference between the proofs of Lemmas 1.21 and 1.25, and of Lemma 1.23, is that in the former we choose the order of embedding vertices in the RGA, whereas in the latter an order is given to us. We change it only by moving the vertices $X^{\text{buf}}_i$ to the end of the order.

### 1.4.1. Organisation of the proofs of the blow-up lemmas.

In the following Section 2.1 we give a collection of probabilistic and graph-theoretic tools which we will need in the proofs, and in the applications, of our blow-up lemmas.

In Section 2.2 we formulate a number of deterministic properties which are enjoyed by random graphs (asymptotically almost surely) and bijumbled graphs for certain parameter ranges. In the proofs of our blow-up lemmas, we will assume only these properties of our ambient graph $\Gamma$; we will make neither further probabilistic calculations about the random graph $G_{n,p}$ nor further use of bijumbledness.

In Section 2.3 we describe the general setup for the proofs of our blow-up lemmas, including the setup of constants, the preparation of the graphs $G$ and $H$, and the definition of candidate sets, good partial embeddings, bad vertices, and related concepts. Apart from the selection of vertices $X^c$ (which we do in the proof
of Lemma 1.21, Section 3.1) this section covers the above proof overview up to the point at which ‘H and G are prepared for the embedding’.

In Section 2.4 we give four lemmas relating to random greedy algorithms. The first of these generates an order on $V(H)$ which we use for the RGAs in the proofs of Lemmas 1.21 and 1.25. The next three are statements about the typical behaviour of a class of RGAs, one of which (Lemma 2.27) we use in the analysis of all three of our RGAs, and the other two of which we use in the RGAs for Lemmas 1.21 and 1.25.

Chapter 3 constitutes the remainder of the proof of Lemma 1.21. In Section 3.1 we formulate four lemmas which describe the results of the four embedding procedures we use in the proof. These are the RGA (Lemma 3.1), the queue embedding (Lemma 3.2), the fixing of buffer defects (Lemma 3.3) and the buffer embedding (Lemma 3.4). We prove the last of these (which is short) in Section 3.1 and round the section off by deducing Lemma 1.21. In Section 3.2 we state exactly the RGA (Algorithm 1) whose typical output is Lemma 3.1, and analyse it to prove that lemma. We note that, although we mentioned in the overview above the ‘extra property’ of the RGA that allows for the buffer embedding when discussing that phase of the proof, we prove the ‘extra property’ holds in this section, so that everything we need from the RGA is contained in Lemma 3.1 (and the corresponding point also holds for our other two RGA lemmas). In Section 3.3 we state our (much simpler) queue embedding procedure, and show that it suffices for Lemma 3.2. Finally, in Section 3.4 we explain how to deal with buffer defects caused by our inability to handle clique buffers in the RGA, giving Algorithm 2 whose output yields Lemma 3.3.

Chapter 4 contains the corresponding parts of the proof of Lemma 1.25. We prove Lemma 1.25 before Lemma 1.23 because we use some ideas from the proof of the former, in a simpler situation, to prove the latter. In Section 4.1 we state Lemma 4.1, which describes the typical output of the RGA we use here, and proceed to deduce Lemma 1.25. Note that we do not have any separate queue embedding and buffer defect fixing steps in the proof of Lemma 1.25: the former is done by the RGA, while the latter is unnecessary as this RGA typically does not create buffer defects. We do need to analyse the buffer embedding, but this is done in the proof of Lemma 1.25 directly as we will not be able to re-use the argument. In Section 4.2 we state the RGA we use (Algorithm 3) and give the analysis showing its typical output is as described in Lemma 4.1. Note that although the RGA handles the embedding of both the queue and non-queue vertices of $X_{\text{main}}$, the analyses we need for the two types of vertex are quite different.

Finally, Chapter 5 completes the proof of Lemma 1.23. In Section 5.1, we state Lemma 5.1, which again describes the typical output of the RGA we use, and deduce Lemma 1.23 from it. In this deduction, we can re-use Lemma 3.4 which tells us that the buffer embedding succeeds. In Section 5.2, we state Algorithm 4, the RGA algorithm we use, and analyse it to prove Lemma 5.1. This is the most complicated RGA we use, and its analysis is correspondingly difficult. It embeds each of queue vertices, non-queue vertices and exceptional vertices (in $X^e$) of $X_{\text{main}}$, and we need to analyse each separately (though $X^e$ turns out to be very easy). Most of the ideas used in this part were seen previously in the proof of Lemma 4.1. We also need to analyse the embedding of neighbours of buffer vertices by the RGA in order to show that the buffer embedding is possible, and the ideas we use for this were mainly seen in Lemmas 2.28 and 2.29.
CHAPTER 2

Tools, preparation and setup

2.1. Tools

2.1.1. Probabilistic inequalities. We need the following forms of Chernoff’s inequality and of the hypergeometric inequality. Recall that if \( X \) is hypergeometrically distributed with parameters \( N, n, m \), then \( \mathbb{E}X = \frac{mn}{N} \).

**Theorem 2.1** (Corollary 2.4, Theorems 2.8 and 2.10 from [37]). Let \( X \) be a random variable which is either the sum of a collection of independent Bernoulli random variables, or is hypergeometrically distributed. Then we have for \( \delta \in (0, 3/2) \)

\[
P(X > (1 + \delta)\mathbb{E}X) < e^{-\delta^2\mathbb{E}X/3} \quad \text{and} \quad P(X < (1 - \delta)\mathbb{E}X) < e^{-\delta^2\mathbb{E}X/3}.
\]

We also have, for any \( t \geq 6\mathbb{E}X \),

\[
P(X \geq \mathbb{E}X + t) \leq e^{-t}.
\]

We will often want to bound above the sum of a sequence of Bernoulli random variables \( Y_1, \ldots, Y_n \) coming from some process which are not independent, but which have the following sequentially dependent structure. Suppose \( \mathcal{H}_0, \ldots, \mathcal{H}_n \) are increasing ‘histories’ of the process, that is, information on all the random choices made in the process up to given increasing times in the process. Suppose that for each \( 1 \leq i \leq n \), the value of the random variable \( Y_i \) is determined by \( \mathcal{H}_i \), and that we have functions \( p_i = p_i(\mathcal{H}_{i-1}) \) such that \( \mathbb{E}[Y_i | \mathcal{H}_{i-1}] \leq p_i \) holds almost surely, that is, with probability 1. If \( \sum_{i=1}^n p_i \) is almost surely bounded above by \( x \), then the following lemma claims the same upper tail bound on \( \sum_{i=1}^n Y_i \) holds as we would get from Theorem 2.1 if the \( Y_i \) were independent and the sum of their expectations were \( x \), and also gives the same lower tail bound as Theorem 2.1 under similar conditions.

It is convenient to phrase this lemma in terms of a sequence of partitions \( \mathcal{F}_0, \ldots, \mathcal{F}_n \), each refining the previous, of a probability space \( \Omega \). For the connection to processes and histories, observe that any finite stochastic process is associated with the finite probability space of all possible outcomes, with the probability measure coming from the process. The possible histories of the process up to any given time \( t \) naturally give a partition of this probability space, and two histories up to an earlier and later time give two partitions, the first refined by the second. Thus the following lemma indeed applies to sequentially dependent random variables as described above.

We note that this lemma could be phrased in terms of a filtration and random variables measurable with respect to elements of the filtration, which might be more familiar to readers with a background in probability. That we do not use this notation here is purely to avoid defining these concepts. Furthermore, we remark that the lemma is essentially a super/submartingale inequality, proved in the standard way.\(^1\)

However, we did not find this particular inequality in the literature, so give a proof from first principles here.

**Lemma 2.2** (Sequential dependence lemma). Let \( \Omega \) be a finite probability space, and \( \mathcal{F}_0, \ldots, \mathcal{F}_n \) be partitions of \( \Omega \), with \( \mathcal{F}_{i-1} \) refined by \( \mathcal{F}_i \) for each \( i \in [n] \). For each

\[^1\text{We would like to thank Oliver Riordan and Ori Gurel-Gurevich for pointing this out to us.} \]
of $F$, we conclude by induction that $f$ is constant on each part $F$. Let $x$ be a real number, $\delta \in (0,3/2)$, and $X = Y_1 + \cdots + Y_n$.

(a) If $\sum_{i=1}^n p_i \leq x$ holds almost surely, and $\mathbb{E}[Y_i|\mathcal{F}_{i-1}] \leq p_i$ holds almost surely for all $i \in [n]$, then

$$\mathbb{P}(X > (1 + \delta)x) < e^{-\delta^2 x/3}.$$  

(b) If $\sum_{i=1}^n p_i \geq x$ holds almost surely, and $\mathbb{E}[Y_i|\mathcal{F}_{i-1}] \geq p_i$ holds almost surely for all $i \in [n]$, then

$$\mathbb{P}(X < (1 - \delta)x) < e^{-\delta^2 x/3}.$$  

Proof. We start with (a). We shall first show by induction on $n$ that if $\sum_{i=1}^n p_i \leq x$ holds almost surely and $\mathbb{E}[Y_i|\mathcal{F}_{i-1}] \leq p_i$ holds almost surely for all $i \in [n]$, then we have the following bound on the moment generating function of $X$ for every $u \geq 0$:

$$\mathbb{E}e^{uX} = \mathbb{E} \prod_{i=1}^n e^{uY_i} \leq \left(1 - \frac{x}{n} + \frac{x}{n} e^u\right)^n.$$  

For $n = 1$ we have $\mathbb{P}(Y_1 = 1|\mathcal{F}_0) \leq p_1 \leq x$ almost surely, hence $\mathbb{P}(Y_1 = 1) \leq x$ almost surely. We conclude that indeed

$$\mathbb{E}e^{uY_1} = 1 - \mathbb{P}(Y_1 = 1) + e^u\mathbb{P}(Y_1 = 1) = 1 - x + xe^u,$$

because $1 - t + te^u$ is non-decreasing in $t$.

For $n \geq 2$, we shall use for each $F \in \mathcal{F}_1$ with $\mathbb{P}(F) > 0$, the induction hypothesis applied to the $n - 1$ random variables $p_{12}F, \ldots, p_nF$ and the $n - 1$ random variables $Y_{12}F, \ldots, Y_{1n}F$. This is possible because, since $\mathbb{P}(F) > 0$, we have $\sum_{i=2}^n p_i|F \leq x - p_1|F$ almost surely and $\mathbb{E}[Y_i|F,\mathcal{F}_{i-1}] \leq p_i|F$ almost surely for each $2 \leq i \leq n$. We conclude by induction that

$$\mathbb{E} \prod_{i=1}^n e^{uY_i} = \sum_{F \in \mathcal{F}_1} \mathbb{P}(F) e^{uY_1|F} \mathbb{E} \prod_{i=2}^n e^{uY_i|F} \leq \sum_{F \in \mathcal{F}_1} \mathbb{P}(F) e^{uY_1|F} \left(1 - \frac{x - p_1|F}{n-1} + \frac{x - p_1|F}{n-1} e^u\right)^{n-1}.$$  

Further, because $Y_1$ is constant on each part $F$ of $\mathcal{F}_1$ we have for each $F_0 \in \mathcal{F}_0$ that

$$\sum_{F \in \mathcal{F}_1, F \subseteq F_0} \mathbb{P}(F) e^{uY_1|F} = \mathbb{P}(F_0) e^{uY_1|F_0}$$

$$= \mathbb{P}(F_0)(1 - \mathbb{P}(Y_1 = 1|F_0) + e^u\mathbb{P}(Y_1 = 1|F_0)) \leq \mathbb{P}(F_0)(1 - p_1|F_0 + e^u p_1|F_0),$$

where the inequality uses that $1 - t + te^u$ is non-decreasing in $t$. Because $p_1$ is constant on each part $F_0$ of $\mathcal{F}_0$ it thus follows from (4) that

$$\mathbb{E} \prod_{i=1}^n e^{uY_i} \leq \sum_{F_0 \in \mathcal{F}_0} \sum_{F \in \mathcal{F}_1, F \subseteq F_0} \mathbb{P}(F) e^{uY_1|F} \left(1 - \frac{x - p_1|F_0}{n-1} + \frac{x - p_1|F_0}{n-1} e^u\right)^{n-1} \leq \sum_{F_0 \in \mathcal{F}_0} \mathbb{P}(F_0)(1 - p_1|F_0 + e^u p_1|F_0) \left(1 - \frac{x - p_1|F_0}{n-1} + \frac{x - p_1|F_0}{n-1} e^u\right)^{n-1}.$$  

Since $f(t) = \ln(1 - t + te^u)$ is a concave function in $t$ it follows from Jensen’s inequality that

$$f(t) + (n-1)f\left(\frac{x-t}{n-1}\right) \leq n \cdot f\left(\frac{1}{n} \left(t + (n-1)\frac{x-t}{n-1}\right)\right) = n \cdot f\left(\frac{x}{n}\right),$$

where
for any $t \in [0,1]$. Substituting this, with $t = p_1|F_0$ for each $F_0 \in F_0$, into (5) yields (3), as desired.

We are now in a position to apply Bernstein’s method of applying Markov’s inequality to the moment generating function. Since $e^{uz}$ is strictly increasing in $z$ for each $u > 0$, we have

$$
P(X > (1+\delta)x) = \mathbb{P}(e^{uX} > e^{(1+\delta)ux}) \leq \left(\mathbb{E}e^{uX}\right) e^{-(1+\delta)ux} \leq \left(1 - \frac{\delta}{x} + \frac{x}{\delta} e^u\right)^n e^{-(1+\delta)ux},
$$

if $u > 0$, where the first inequality is Markov’s inequality. We may assume $(1+\delta)x < n$, since otherwise $\mathbb{P}(X > (1+\delta)x) = 0$ and the lemma statement is trivial. Thus we can choose $u$ such that $e^u = \frac{(1+\delta)(n-x)}{n(1+\delta)x}$, plug that into the right hand side of (6) and obtain

$$
P(X > (1+\delta)x) \leq \left(1 + \frac{\delta x}{n(1+\delta)x}\right)^n \left(\frac{(1+\delta)(n-x)}{n(1+\delta)x}\right)^{(1+\delta)x} \leq e^{\delta x} (1+\delta)^{-(1+\delta)x} \leq e^{-\delta^2 x/3},
$$

where the last inequality uses $(1+\delta)\ln(1+\delta) - \delta \geq \delta^2/3$. This proves (a).

The proof of (b) is similar. By the analogous calculation, we have

$$
\mathbb{E}e^{u(n-X)} \leq \left(1 - \frac{n-x}{n} + \frac{n-x}{e^u}\right)^n,
$$

and again Bernstein’s method, this time choosing $e^u = \frac{n-x+\delta x}{(n-x)(1-\delta)}$, gives the desired result for $\delta \in (0,1)$:

$$
P(X < (1-\delta)x) = \mathbb{P}(n-X > n-x + \delta x) = \mathbb{P}(e^{u(n-X)} > e^{u(n-x+\delta x)})
\leq \mathbb{E}e^{u(n-X)} e^{-u(n-x+\delta x)} \leq \left(1 - \frac{n-x}{n} + \frac{n-x}{e^u}\right)^n e^{-u(n-x+\delta x)}
= \left(1 + \frac{n-x}{n} \left(\frac{n-x+\delta x}{(n-x)(1-\delta)}\right)\right)^n \left(\frac{n-x+\delta x}{(n-x)(1-\delta)}\right)^{-(n-x+\delta x)}
= (1-\delta)^{-n} \left(\frac{n-x+\delta x}{(n-x)(1-\delta)}\right)^{-(n-x+\delta x)} = (1-\delta)^{-(1-\delta)x} \left(\frac{n-x+\delta x}{n-x}\right)^{-(n-x+\delta x)}
\leq (1-\delta)^{-(1-\delta)x} \left(1 + \frac{\delta x}{n-x}\right)^{-(n-x)} \leq (1-\delta)^{-(1-\delta)x} e^{-\delta x} \leq e^{-\delta^2 x/3}.
$$

Note that in this case we only need to consider $\delta \in (0,1)$ as the probability of $X < 0$ is zero, so this gives (b).

\begin{flushright}
\Box
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\subsection{2.1.2 Equitable partitions.}

The Hajnal-Szemerédi Theorem states that any graph $F$ has a $k$-colouring with equitable colour classes for every $k \geq (\Delta(F) + 1)$.

\begin{theorem}[Hajnal-Szemerédi \cite{35}]
Given any graph $F$ and $k \geq \Delta(F) + 1$, there is an equitable partition $V(F) = V_1 \cup \cdots \cup V_k$ such that each part is independent.
\end{theorem}

For preparing the graph $H$ for the embedding in the proofs of our blow-up lemmas, we require an equitable partitioning result similar to the Hajnal-Szemerédi Theorem (which we shall apply to an auxiliary graph $F$ defined for each part of the given partition of $H$). The difference is that we want to specify a subset $X$ of the vertices of $F$ and obtain an equitable partition of $F$ that also equitably partitions $X$ (this subset will be the buffer set). Since for us it is not essential to obtain a sharp bound on the number of parts required, this result is not difficult to deduce from the Hajnal-Szemerédi Theorem.
LEMMA 2.4. Given any graph $F$ and a subset $X$ of $V(F)$, there is an equitable partition $V(F) = V_1 \cup \ldots \cup V_{8\Delta(F)}$ such that each part is independent and the sets $V_i \cap X$ form an equitable partition of $X$.

PROOF. We may assume without loss of generality that $|X| \leq v(F)/2$, since otherwise we could replace $X$ with $V(F) \setminus X$. We now first use the Hajnal-Szemerédi Theorem to get an equitable partition of $F[X]$ into $8\Delta(F)$ independent parts. We proceed by adding the remaining vertices of $F$ to these parts, maintaining their independence, to obtain a partition $V_1 \cup \ldots \cup V_{8\Delta(F)}$ of $V(F)$. Among all possible such partitions we choose the one which is as equitable as possible. Observe that in this process it is always possible to put every vertex into some part: for any vertex, there are at most $\Delta(F)$ parts to which it cannot be added, so at least $7\Delta(F)$ parts exist to which it can be added to form an independent set.

We are now done if there exist no pair of parts $V_i, V_j$ whose sizes differ by two or more. Suppose $V_1$ is the smallest part. Observe that there are at most $\Delta(F)|V_1|$ vertices in $V(F) \setminus X$ which have a neighbour in $V_1$. Since $8\Delta(F)|V_1| \leq v(F)$ it follows that there are at least $\frac{7}{8}v(F)$ vertices in $V(f) \setminus X$ which have no neighbour in $V_1$.

Assume for contradiction that there are parts $V_i$ with $|V_i| \geq |V_1| + 2$. Observe that for each such $i$ every vertex of $V_i \setminus X$ is adjacent to some vertex of $V_1$, or we would be able to move a vertex from $V_i \setminus X$ to $V_1$ and obtain a more equitable partition. Hence the at least $\frac{7}{8}v(F)$ vertices in $V(F) \setminus X$ which have no neighbour in $V_1$ are all contained in parts of size at most $|V_i| + 1 \leq 2|V_1|$, and thus lie in at least $\frac{7}{8}\Delta(F)$ different parts. Now fix a set $V_j$ with $|V_j| \geq |V_1| + 2$, and let $v$ be any vertex of $V_j \setminus X$ (which exists because we started with an equitable partition of $X$). Since $v$ has at most $\Delta(F) < \frac{7}{8}\Delta(F)$ neighbours, there is a set $V_k$ which contains no vertex adjacent to $v$, and which contains a vertex $w$, not in $X$, with no neighbours in $V_1$. We replace $V_j$ with $V_j \setminus \{v\}$, $V_k$ with $V_k \cup \{v\} \setminus \{w\}$, and $V_1$ with $V_1 \cup \{w\}$. The result is a more equitable partition, a contradiction.

2.1.3. Regularity inheritance. In our blow-up lemmas we require certain regularity inheritance properties. In this subsection we shall make the two different regularity concepts explicit and states that for most vertices $z \in Z$ the pair $(N_F(z; X), Y)$ inherits full-regularity.

Recall that, according to our definition, regularity in random graphs and regularity in bijumbled graphs are different things: lower-regularity in the former and full-regularity in the later. Recall also that the sole reason for this difference is that we can only establish regularity inheritance for the respective version, that is, that the lemmas provided in this subsection require lower-regularity when they concern random graphs and full-regularity when they concern bijumbled graphs. Accordingly, in this subsection we shall make the two different regularity concepts explicit and talk about lower-regularity and full-regularity for clarity.

For subgraphs $G$ of bijumbled graphs $\Gamma$ we can simply rely on the following inheritance lemmas from [5]. These lemmas consider three disjoint vertex sets $X, Y, Z$ such that $(X, Y)$ forms a fully-regular pair. The first lemma is a one-sided regularity inheritance statement and states that for most vertices $z \in Z$ the pair $(N_F(z; X), Y)$ inherits full-regularity.

LEMMA 2.5 (One-sided regularity inheritance in bijumbled graphs [5, Lemma 3]).

For each $\varepsilon', d > 0$ there are $\varepsilon_0, c_0 > 0$ such that for all $0 < \varepsilon \leq \varepsilon_0$ and $0 < p < 1$ the following holds. Let $G \subseteq \Gamma$ be graphs and $X, Y, Z$ be disjoint subsets of $V(\Gamma)$. If
(X, Y) is $(\varepsilon, d, p)$-fully-regular in $G$ and $\Gamma$ is
\[
(p, c_0 p^2 (\log \frac{1}{p})^{-1/2} \sqrt{\max \{\|X\|Y|, |X||Z|\}})\text{-bijumbled}
\]
then for at most $\varepsilon'|Z|$ vertices $z$ of $Z$ the pair $(N_\Gamma(z; X), Y)$ is not $(\varepsilon', d, p)$-fully-
regular in $G$.

The second lemma requires stronger bijumbledness for establishing two-sided
regularity inheritance. It states that for at most $z \in Z$ the pair $(N_\Gamma(z; X), N_\Gamma(z; Y))$
inherits full-regularity.

**Lemma 2.6** (Two-sided regularity inheritance in bijumbled graphs [5, Lemma 4]).

For each $\varepsilon', d > 0$ there are $\varepsilon_0, c_0 > 0$ such that for all $0 < \varepsilon < \varepsilon_0$ and $0 < p < 1$
the following holds. Let $G \subseteq \Gamma$ be graphs and $X, Y, Z$ be disjoint subsets of $V(\Gamma)$. If $(X, Y)$ is $(\varepsilon, d, p)$-fully-regular in $G$ and $\Gamma$ is
\[
(p, c_0 p^3 \sqrt{\max \{\|X||Y|, |X||Z|, |Y||Z|\}})\text{-bijumbled}
\]
then for at most $\varepsilon'|Z|$ vertices $z$ of $Z$ the pair $(N_\Gamma(z; X), N_\Gamma(z; Y))$ is not $(\varepsilon', d, p)$-
fully-regular in $G$.

We note that the regularity inheritance lemmas in random graphs, Lemmas 1.26
and 1.27, stated in Section 1.3.5, are stated in a different form to the ones for
bijumbled graphs above, not requiring any bound on $p$, but rather having $p$ appear
in the estimates for the sizes of $X$ and $Y$ as well as the number of vertices not
preserving regularity. We state them in this form because this seems most suitable
when we want to use them in applications of our blow-up lemma. For bijumbled
graphs on the other hand we decided not to switch to this form since we would
obtain more complicated conditions in this case (bounding the products of set sizes
such as $|X||Y|$, instead of the sets themselves).

In the proofs of Lemmas 1.26 and 1.27 we follow the approach of [44],
and rely on the following result of Gerke, Kohayakawa, Rödl and Steger [33] stating
that the vast majority of subpairs $(X', Y)$ of a lower-regular pair $(X, Y)$ in any
graph inherit lower-regularity. Unfortunately this result becomes false if we replace
lower-regularity by full-regularity, and it is precisely this reason why we need to
work with lower-regularity in random graphs.

**Theorem 2.7** (Theorem 3.6 from [33]). For any $d, \beta, \varepsilon' > 0$ there exist $\varepsilon_0 > 0$
and $C$ such that for any $0 < \varepsilon < \varepsilon_0$ and $0 < p < 1$, if $(X, Y)$ is an $(\varepsilon, d, p)$-
lower-regular pair in a graph $G$, then the number of sets $X' \subseteq X$ with $|X'| = w \geq C/p$
such that $(X', Y)$ is an $(\varepsilon', d, p)$-lower-regular pair in $G$ is at least
\[
(1 - \beta^w) \left(\frac{|X|}{w}\right).
\]

The following is an analogous statement for subpairs $(X', Y')$ of $(X, Y)$ and is
an immediate consequence of Theorem 2.7.

**Corollary 2.8** (Corollary 3.8 from [33]). For any $d, \beta, \varepsilon' > 0$ there exist $\varepsilon_0 > 0$
and $C$ such that for any $0 < \varepsilon < \varepsilon_0$ and $0 < p < 1$, if $(X, Y)$ is an
$(\varepsilon, d, p)$-lower-regular pair in a graph $G$, then the number of pairs $X' \subseteq X$ and
$Y' \subseteq Y$ with $|X'| = w_1 \geq C/p$ and $|Y'| = w_2 \geq C/p$ such that $(X', Y')$ is an
$(\varepsilon', d, p)$-lower-regular pair in $G$ is at least
\[
(1 - \beta^\min(w_1,w_2)) \left(\frac{|X|}{w_1}\right)\left(\frac{|Y|}{w_2}\right).
\]

The proof idea for both of our random graphs inheritance lemmas is the following.
First, we will show that if $(X, Y)$ is a counterexample, that is, a lower-regular pair
not satisfying the conclusion of either lemma, then there is a counterexample $(X^*, Y^*)$
with both parts having the minimum sizes allowed and such that the ‘offending’
vertices $z$ are outside $X^* \cup Y^*$. We will then show that this latter structure is
unlikely to exist in \( G_{n,p} \). It may not be obvious what we gain by reducing \((X,Y)\) to \((X^*, Y^*)\); what we gain is that there are not too many edges of \( \Gamma \) between \( X^* \) and \( Y^* \), and hence not too many possible lower-regular subgraphs of \( \Gamma \). This allows us to take a union bound over all possible choices, which would fail if we attempted it with the original \((X,Y)\).

In this proof we shall use the following two lemmas. The first lemma assumes we are given a regular pair together with a collection of \( \ell \) subpairs that are substantially sparser than the regular pair, and asserts that we can scale down the sizes of this pair and the subpairs.

**Lemma 2.9.** For each \( d, \varepsilon', \delta > 0 \) there exist \( \varepsilon_0 \) and \( C \) such that for all \( 0 < \varepsilon < \varepsilon_0 \) the following holds. Let \((X,Y)\) be an \((\varepsilon,d,p)\)-lower-regular pair in a graph \( G \), and suppose \(|X| \geq m \geq C p^{-1-\delta} \) for \( i \in \{0,1\} \). Further, let \( X_j \subseteq X \) and \( Y_j \subseteq Y \) with \( j \in [\ell] \) form \( \ell \) subpairs of \((X,Y)\) for some \( \ell \in [m] \), such that \(|X_j| \geq \delta p^i |X| \) and \( d_p(X_j, Y_j) < d - \delta \).

Then there are \( X^* \subseteq X \) with \(|X^*| = m \) and \( X^*_j \subseteq X \cap X^* \) with \(|X^*_j| \geq \delta^2 p^{i} m/6 \) for each \( j \in [\ell] \), such that \((X^*, Y)\) is \((\varepsilon',d,p)\)-lower-regular in \( G \) and \( d_p(X^*_j, Y_j) < d - \delta/2 \) for each \( j \in [\ell] \).

**Proof.** Given \( d,\varepsilon, \delta > 0 \), let \( \varepsilon_0 \) and \( C' \) be returned by Theorem 2.7 with input \( d, \varepsilon', \delta \), and let \( C \geq C' \) be such that \((x^2/C) \exp(-\delta^2 x/36) < \frac{1}{2} \) for each \( x \geq 1 \).

Given \( 0 < \varepsilon < \varepsilon_0 \) and \( \delta > 0 \), and given an \((\varepsilon,d,p)\)-lower-regular pair \((X,Y)\) with \(|X| \geq m \geq C p^{-1-\delta} \), let \( X^* \subseteq X \) be chosen uniformly at random from the \( m \)-sized subsets of \( X \). By Theorem 2.7, with probability at least \( \frac{1}{2} \), the pair \((X^*,Y)\) is \((\varepsilon'\,d,p)\)-lower-regular.

Given any pair \((X_j,Y_j)\) with \( d_p(X_j,Y_j) < d - \delta \), there exists a subset \( X_j'' \) of \( X_j \) consisting of at least \( \delta |X_j|/3 \) vertices each of whose degree into \( Y_j \) is smaller than \((d-\delta/2)p|Y_j|\), as otherwise we would have

\[
e(X_j \setminus X_j'', Y_j) \geq (1 - \delta/3)|X_j|(d - \delta/2)p|Y_j| > (d - \delta)p|X_j||Y_j| > \varepsilon(X_j, Y_j),
\]
a contradiction. Let \( X_j'' = X_j'' \cap X^* \), then \( d_p(X_j'', Y_j) < d - \delta/2 \), by definition of \( X_j'' \).

The quantity \(|X_j''|\) is hypergeometrically distributed with mean at least \( \delta^2 p^{i} m/3 \) since \(|X_j''| \geq \delta |X_j|/3 \) and \(|X_j| \geq \delta p^i |X| \geq \delta p^i m \). By Theorem 2.1, the probability that \(|X_j''| < \delta^2 p^{i} m/6 \) is at most \( \exp(-\delta^2 p^{i} m/36) \).

By choice of \( m \), we have \((p' m)^2 \geq C p^{-1} p' m \geq C m \), so setting \( x = p' m \geq 1 \), by choice of \( C \) we have \((x^2/C) \exp(-\delta^2 x/36) < \frac{1}{2} \). Thus taking a union bound, we conclude that with probability at least

\[
1 - \frac{1}{2} - m \exp(-\delta^2 p^{i} m/36) > 0
\]
all of the above good events occur, giving the conclusion of the lemma. \( \square \)

The next easy lemma shows that adding a few vertices to an \((\varepsilon,d,p)\)-lower-regular pair does not destroy regularity. We note that the corresponding result for an \((\varepsilon,d,p)\)-fully-regular pair would require knowing that \( G \) does not contain dense spots.

**Lemma 2.10.** Let \( 0 < \varepsilon < \frac{1}{10} \). Let \( G \) be a graph and let \( U', V' \subseteq V(G) \) be disjoint sets such that \((U', V')\) is \((\varepsilon,d,p)\)-lower-regular in \( G \). If \( U \supseteq U' \) with \(|U| \leq (1 + \frac{1}{10} \varepsilon^2)|U'| \) and \( V \supseteq V' \) with \(|V| \leq (1 + \frac{1}{10} \varepsilon^2)|V'| \) are disjoint, then \((U,V)\) is \((2\varepsilon,d,p)\)-lower-regular in \( G \).
Proof. Let $X$ and $Y$ be arbitrary subsets of $U$ and $V$ with $|X| \geq 2\varepsilon|U|, |Y| \geq 2\varepsilon|V|$. We need to show that $e_G(X, Y) \geq (d - 2\varepsilon)p|X||Y|$. Since $(U', V')$ is $(\varepsilon, d, p)$-lower-regular in $G$, and since $2\varepsilon - \frac{1}{10} \varepsilon^3 > \varepsilon$, we have

\[
e_G(X \cap U', Y \cap V') \geq (d - \varepsilon)p|X \cap U'||Y \cap V'| \geq (d - \varepsilon)(1 - \frac{1}{10} \varepsilon^3)^2 p|X||Y| \geq (d - 2\varepsilon)p|X||Y|,
\]
as desired. \hfill \Box

We now prove our one-sided regularity inheritance lemma, Lemma 1.26.

Proof of Lemma 1.26. Given $0 < \varepsilon', d$ we set $\delta_1 = (\varepsilon')^2/24$, $\delta_2 = \delta_1^2/24$ and $\beta = 2^{-10\varepsilon' p^{-3}}$. Now Theorem 2.7 with input $d, \beta, \delta_2$ returns constants $\varepsilon_2$ and $C_2$. We now let $\varepsilon_1$ and $C_1$ be the output of Lemma 2.9 for input $d$, $\varepsilon_2$ and $\delta_1$. We let $\varepsilon_3$ and $C_3$ be the output of Lemma 2.9 for input $d$, $\varepsilon_1$ and $\varepsilon'/2$. Finally, we set

$C' = \max(1600, C_1, 2C_2, C_3), \quad C = 2000(\varepsilon')^{-3} C'$ and $\varepsilon_0 = \frac{1}{10}\min\{\varepsilon_1, \varepsilon_2, \varepsilon_3\}$.

Suppose that $p \geq Cn^{-1/2}$ (since otherwise $|X| > n$ and the claim of the lemma vacuously holds).

We first show that a.a.s. $\Gamma = G_{n, p}$ has the following property:

$(\star)$ For any disjoint $X, Y \subseteq V(\Gamma)$ both of size at least $24p^{-1} \log n$ we have $e_\Gamma(X, Y) = (1 \pm \frac{1}{2})p|X||Y|$, and for any $Z \subseteq V(\Gamma)$ of size at least $12p^{-1} \log n$ we have $e_\Gamma(Z) \leq p|Z|^2$.

Indeed, using Theorem 2.1 and a union bound over the choices of $X$ and $Y$, the failure probability of the first assertion is at most

\[
\sum_{|X|, |Y| \geq 24p^{-1}\log n} n^{\max(|X|, |Y|)} d \exp\left(-\frac{p|X||Y|}{12}\right) \leq \sum_{|X|, |Y|} 2 \cdot 2^{2\max(|X|, |Y|)} \log n \exp\left(-2\max(|X|, |Y|) \log n\right) \leq 2n^2 \cdot 2^{2\exp(-2)} \cdot \log^2 n,
\]
where the first inequality uses $|X|, |Y| \geq 24p^{-1} \log n$ and the second max$(|X|, |Y|) \geq \log n$. Since $2^2 \exp(-2) < 1$, this tends to zero as $n$ tends to infinity. Similarly, the failure probability of the second assertion is at most

\[
\sum_{|Z| \geq 12p^{-1}\log n} n^{|Z|} \cdot e^{-p|Z|^2/6} \leq \sum_{|Z|} 2^{|Z|} \log n e^{-2|Z| \log n} < n \cdot e^{-n \log^2 n},
\]
which tends to zero as $n$ tends to infinity.

Our proof proceeds by contradiction. The following claim states that, provided $(\star)$ holds, if the conclusion of Lemma 1.26 does not hold for any pair of sets $X$ and $Y$ and graph $G \subseteq \Gamma$, then there exists a minimum-sized bad example. The proof of Claim 2.11 combines Lemmas 2.9 and 2.10, and we provide its proof immediately after we derive the conclusion of Lemma 1.26.

Claim 2.11. Suppose that $(\star)$ holds. Suppose further that we have $G \subseteq \Gamma$ and disjoint $X, Y \subseteq V(\Gamma)$ such that $(X, Y)$ is $(\varepsilon, d, p)$-lower-regular in $G$, with $|X| \geq C\max(p^{-2}, p^{-1}\log n)$ and $|Y| \geq C p^{-1} \log n$, and that there is a set $Z$ of vertices of $\Gamma$ with $|Z| = C' p^{-1} \log n$ such that for each $z \in Z$ the pair $(N_\Gamma(z; X), Y)$ is not $(\varepsilon', d, p)$-lower-regular in $G$.

Then there are disjoint subsets $X^*, Y^*, Z^*$ of subsets of $V(G)$ with $|X^*| = \frac{1}{2} C\max(p^{-2}, p^{-1}\log n), |Y^*| = \frac{1}{2} C p^{-1} \log n$, and with $|Z^*| = 10^{-6}(\varepsilon')^C C p^{-1} \log n$, and
such that \((X^*, Y^*)\) is \((\varepsilon_2, d, p)\)-lower-regular in \(G\), and for each \(z \in Z^*\) we have \(\deg_G(z; X^*) = (1 \pm \frac{1}{3})p|X^*|\), but \((N_T(z; X^*), Y^*)\) is not \((\delta_2, d, p)\)-lower-regular in \(G\).

We now prove that the structure which Claim 2.11 provides is unlikely to be in \(G_{n,p}\). To that end, we first fix arbitrary disjoint sets \(X^*, Y^*\) and \(Z^*\) of sizes \(\frac{1}{2}C\max(p^{-2}, p^{-1}\log n), \frac{1}{4}Cp^{-1}\log n\) and \(10^{-6}(c')^3 \cdot Cp^{-1}\log n\) respectively. We now reveal the edges of \(G_{n,p}\) between \(X^*\) and \(Y^*\). If there are more than \(\frac{3}{2}p|X^*||Y^*|\) such edges, then \((\star)\) does not hold and there is nothing to prove, so we assume from now on that there are at most \(\frac{2}{3}p|X^*||Y^*|\) such edges. We fix an arbitrary subgraph \(G^*\) of these edges. In the event that \(G^*\) is not \((\varepsilon_2, d, p)\)-lower-regular, there is nothing to prove, so we assume from now on that it is \((\varepsilon_2, d, p)\)-lower-regular.

Next, we reveal the edges of \(G_{n,p}\) from \(Z^*\) to \(X^*\). In the event that for some \(z \in Z^*\) we have \(|N_T(z) \cap X^*| \neq (1 \pm \frac{1}{2})p|X^*|\), the conclusion of Claim 2.11 does not hold, and there is nothing to prove. We therefore assume that this does not occur. Thus for each \(z \in Z^*\), the set \(N_T(z) \cap X^*\) is a uniformly random subset of \(X^*\) of size \(|N_T(z) \cap X^*|\), and the latter random variable is supported on \(\{1 \pm \frac{1}{2}p|X^*|\}\).

By Theorem 2.7, with inputs \(d, \beta, 2d_2\), the probability of choosing a \(t\)-element subset of \(X^*\) uniformly at random and discovering that it does not induce with \(Y^*\) a \((\delta_2, d, p)\)-lower-regular subpair of \(G^*\) is at most \(\beta^t\). In particular, the probability that \((N_T(z) \cap X^*, Y^*)\) is not \((\delta_2, d, p)\)-lower-regular is at most

\[
\sum_{t=p|X^*|/2}^{3p|X^*|/2} \mathbb{P}(|N_T(z) \cap X^*| = t) \beta^t \leq \beta^{|X^*|/2}.
\]

Since these events are independent for different \(z \in Z^*\), the probability that \((N_T(z) \cap X^*, Y^*)\) is not \((\delta_2, d, p)\)-lower-regular for all \(z \in Z^*\) is at most \(\beta^{|X^*|/2} |Z^*|/2\).

Taking a union bound over the at most \(n|X^*|+|Y^*|+|Z^*|\) choices of \(X^*, Y^*\) and \(Z^*\), and the at most \(2\beta^{|X^*|/2} |Z^*|/2\) choices of \(G^*\), we conclude that with probability at least

\[
2^{-\varepsilon_2} \max(p^{-2}, p^{-1}\log n) \log n
\]

the graph \(G_{n,p}\) satisfies both \((\star)\) and the conclusion of Claim 2.11.

Therefore we can bound the probability that the claim of Lemma 1.26 fails by

\[
\mathbb{P}[(\star) \text{ fails in } G_{n,p}] + 2^{-\varepsilon_2} \max(p^{-2}, p^{-1}\log n) \log n = o(1).
\]

To complete the proof of Lemma 1.26, we now provide the proof of Claim 2.11.

**Proof of Claim 2.11.** Let \(X' = X \setminus Z\) and \(Y' = Y \setminus Z\). Observe that \(|Y'| \geq (1 - \frac{\varepsilon}{C})|Y|\) and \(|X'| \geq (1 - \frac{\varepsilon}{C})|X|\). Because \(C > 2\varepsilon\) it follows that \((X', Y')\) is \((2\varepsilon, d, p)\)-lower-regular in \(G\) by the slicing lemma, Lemma 1.15. Now the number of vertices \(z \in Z\) with \(|N_T(z; X')| < \frac{1}{2}p|X'|\) is at most \(24p^{-1}\log n\) by property \((\star)\). Also using property \((\star)\), the number of edges in \(Z\) is at most \(p|Z|^2\), and so at most \(|Z|/2\) vertices have more than \(4p|Z|\) neighbours in \(Z\). Let \(Z'\) consist of those vertices of \(Z\) with at most \(4p|Z|\) neighbours in \(Z\) and at least \(\frac{1}{2}p|X'|\) neighbours in \(X'\). Then we have \(|Z'| \geq \frac{1}{2}|Z| - \frac{2p}{3}|Z| \geq \frac{1}{4}|Z|\) by choice of \(C'\).

Now let \(z\) be any vertex in \(Z'\). We would like to apply Lemma 2.10 to argue that lower-regularity of the pair \((N_T(z; X'), Y')\) implies lower regularity of the pair
\((N_T(z; X), Y)\). For this we first need to check that
\[
|N_T(z; X \setminus X')| \leq \frac{1}{10}(\epsilon')^3 |N_T(z; X')|.
\]
Indeed, by definition \(z\) has at most \(4p|Z|\) neighbours in \(Z\), so
\[
|N_T(z; X \setminus X')| \leq 4p|Z| = 4C'|\log n|,
\]
and also by definition \(z\) has at least
\[
\frac{1}{2}p|X'| \geq \frac{1}{4}p|X| \geq \frac{1}{4}C \max \left(p^{-1}, \log n \right)
\]
neighbours in \(X'\). By choice of \(C\), (7) follows as required. Similarly, \(|Y'| \leq \frac{C'}{\log n} |Y| \leq \frac{1}{10}(\epsilon')^3 |Y'|\). Hence we can apply Lemma 2.10 and conclude that if \((N_T(z; X'), Y')\) is an \((\epsilon', d, p)\)-lower-regular pair in \(G\), then \((N_T(z; X), Y)\) is \((\epsilon', d, p)\)-lower-regular in \(G\). By definition of \(Z\) the latter is not the case, so we conclude that for each \(z \in Z'\) the pair \((N_T(x; X'), Y')\) is not \((\frac{1}{2}\epsilon', d, p)\)-lower-regular in \(G\).

Let \(Z''\) be a set of \(t = \frac{1}{2}C'p^{-1} \log n\) vertices \(z \in Z'\) with \(|N_T(z) \cap X'| = (1 \pm \frac{1}{4})|X'|\). It is possible to choose such a set because by (\(\ast\)) all but at most \(48p^{-1} \log n\) of the at least \(\frac{1}{4}|Z| \geq \frac{1}{2}Cp^{-1} \log n\) vertices of \(Z\) have this property. Now for each \(z \in Z''\), since \((N_T(z; X'), Y')\) is \((\frac{1}{2}\epsilon', d, p)\)-lower-regular, there are sets \(X'_z \subseteq N_T(z; X')\) and \(Y'_z \subseteq Y'\) such that \(d_p(X'_z, Y'_z) < d - \frac{1}{2}\epsilon'\) and with \(|Y'_z|, \geq \epsilon' |Y'|\) and \(|X'_z| \geq \frac{1}{2}\epsilon' |N_T(z; X')| \geq \frac{1}{2} \epsilon' p |X'|\), where the last inequality uses the definition of \(Z''\).

We now shall apply Lemma 2.9 twice, first to scale down \(Y'\), and then to scale down \(X'\). So first we apply this lemma with input \(d, \varepsilon_1, \varepsilon'/2\) in place of \(\delta_1\) and \(i = 0\), \(m = \frac{1}{2}C'p^{-1} \log n \geq C_3p^{-1} \log n\), to the pair \((Y', X')\) and the subpairs \((Y_z', X_z')\). This is possible because \(|Y'| \geq \frac{1}{2} |Y| \geq \frac{1}{2} C'p^{-1} \log n = m\) and \(|X'_z| \geq \frac{1}{2} \epsilon' |Y'|\) for each \(z \in Z''\), because \(|Z''| = t \leq m\), and because \((Y', X')\) is \((\epsilon, d, p)\)-lower-regular with \(2\varepsilon \leq \varepsilon_3\) and we have \(d_p(Y'_z, X'_z) < d - \frac{1}{2}\epsilon'\) for each \(z \in Z''\). We conclude that there is a set \(Y^* \subseteq Y'\) of size \(m\), and for each \(z \in Z''\) a subset \(Y_z^*\) of \(Y_z' \cap Y^*\) of size at least \((\epsilon'/2)^2 m / 6 = \delta_1 |Y^*|\), such that \((Y^*, X')\) is \((\epsilon_1, d, p)\)-lower-regular and \(d_p(Y_z^*, X_z') < d - \epsilon'/4 < d - \delta_1\) for each \(z \in Z''\).

In the second application of Lemma 2.9 we use input \(d, \varepsilon_2, \delta_1\), and \(i = 1\), \(m' = \frac{1}{2}C \max \left(p^{-2}, \log^{-1} n\right) \geq C_4p^{-2} \log n\) and apply this lemma to the pair \((X', Y^*)\) and the subpairs \((X'_z, Y_z^*)\). This is possible because \(|X'| \geq \frac{1}{2} |X| \geq \frac{1}{2} C \max \left(p^{-2}, \log^{-1} n\right) = m'\) and \(|X'_z| \geq \frac{1}{2} \epsilon p |X'| \geq \delta_1 p |X'|\) for each \(z \in Z''\), because \(|Z''| = t \leq m'\), and because \((X', Y^*)\) is \((\varepsilon_1, d, p)\)-lower-regular and we have \(d_p(X'_z, Y'_z) < d - \delta_1\) for each \(z \in Z''\). Thus, we obtain a set \(X^* \subseteq X'\) of size \(m'\), and subsets \(X'_z^*\) of \(X'_z \cap X^*\) of size at least \(\delta_2 p m' / 6 \geq \delta_2 p |X'|\) for each \(z \in Z'\), such that \((X^*, Y^*)\) is \((\varepsilon_2, d, p)\)-lower-regular and for each \(z \in Z''\) we have \(d_p(X'_z^*, Y_z^*) < d - \delta_1 / 2 < d - \delta_2\).

Finally, let \(Z^*\) be the set of vertices \(z\) in \(Z''\) such that \(|N_T(z; X^*)| = (1 \pm \frac{1}{4}) |X^*|\). By (\(\ast\)) we have \(|Z^*| \geq |Z''| - 48p^{-1} \log n = \frac{1}{2} C p^{-1} \log n - 48p^{-1} \log n \geq \frac{1}{16} C p^{-1} \log n = \frac{1}{16} C' p^{-1} \log n > 10^{-5} (\epsilon')^3 m\). This proves the claim, since \(d_p(X'_z^*, Y_z^*) < d - \delta_1 / 2 < d - \delta_2\) for each \(z \in Z^*\) verifies that the pairs \((N_T(z; X^*), Y^*)\) are not \((\delta_2, d, p)\)-lower-regular, because \(X_z^* \subseteq X_z \cap X^* \subseteq N_T(z; X^*) \cap X^* \subseteq Y_z^* \subseteq Y_z \cap Y^* \subseteq Y^*\), and because \(|Y_z^*| \geq \delta_1 |Y^*|\), and because \(|N_T(z; X^*)| \leq \frac{1}{2} p |X^*|\) also implies that \(|X_z^*| \geq \delta_2'^2 p |X^*| / 6 \geq \delta_2 |N_T(z; X^*)|\).

The proof of Lemma 1.27 is very similar, so we shall be brief.

Proof of Lemma 1.27 (sketch). We can begin as in the previous proof (in particular the choice of constants remains the same). We replace Theorem 2.7 with
2.2. Deterministic properties of the ambient graph

Corollary 2.8. We use $m = m' = \frac{1}{2} C \max(p^{-2}, p^{-1} \log n)$. Note that we require $m \geq \frac{1}{2} C p^{-2}$ since the expected neighbourhood of a vertex of $G_{n,p}$ in a set of size $m$ is $pm$, and this needs to be large enough to apply Corollary 2.8. We require of the vertices $Z''$ that they have also $(1 \pm \frac{1}{2}) p |Y'|$ neighbours in $Y'$ (and hence we now lose $96 p^{-1} \log n$ vertices). The sets $Y''$ now have size at least $\varepsilon' |Y'|/2$, and both applications of Lemma 2.9 are therefore with $i = 1$. In order to obtain the set $Z^*$ we remove in addition vertices which do not have $(1 \pm \frac{1}{2}) p |Y'|$ neighbours in $Y^*$. However since we remove only at most $192 p^{-1} \log n$ vertices, and since $C$ is large, we still have $|Z^*| > m/2$.

The previous proof with the changes as above leads us to the following conclusion. It is enough to show that $\Gamma = G_{n,p}$ a.a.s. does not contain a ‘bad object’ consisting of sets $X^*$ and $Y^*$ each of size $m$, which form a $(\varepsilon_2, d, p)$-lower-regular pair in a subgraph $G$ of $\Gamma$, together with a set $Z^*$ of $10^{-6} (\varepsilon')^3 m$ vertices of $\Gamma$ such that $|N_{\Gamma}(z \cap X^*)|, |N_{\Gamma}(z \cap Y^*)| = (1 \pm \frac{1}{2}) pm$ and $\{N_{\Gamma}(z) \cap X^*, N_{\Gamma}(z) \cap Y^*\}$ is not $(\delta_2, d, p)$-lower-regular for each $z \in Z^*$.

The proof that this ‘bad object’ does not exist in $G_{n,p}$ is essentially identical. We use Corollary 2.8 to deduce that conditioning on $(\star)$ the probability for (fixed) $X^*, Y^*, Z^*$ and $G$ that the neighbourhood of a given $z \in Z^*$ induces an irregular pair is still at most $\beta p^{m/2}$. The union bound over choices of $X^*, Y^*, Z^*$ and $G$ now yields that the probability of existence of the ‘bad object’ is at most

$$n^{m+m+\frac{2}{2} p^{m/2} p m/2 \beta^{10^{-6} (\varepsilon')^3 p m^2} = n^{5m/2 - 2p m^2 - 2 \delta^2 - 2p m^2},$$

and since $m \geq \frac{1}{2} C p^{-1} \log n$, this tends to zero as $n$ tends to infinity. $\square$

We conclude this section by sketching constructions which show that Lemmas 1.26 and 1.27 are sharp for $p < \frac{1}{2}$.

First, for Lemma 1.26, we fix a set $Z$ of $c p^{-1} \log n$ vertices. It is not hard to verify that for any $\delta > 0$, provided $c > 0$ is small enough, on revealing the edges of $\Gamma = G_{n,p}$ we a.a.s. find a set $X$ of $n^{1-\delta}$ vertices such that each vertex of $Z$ has exactly one neighbour in $X$. We now fix a set $Y$ of $\frac{3}{4} n$ vertices disjoint from $X$. It is easy to check that a.a.s. $(X, Y) \in (\varepsilon, \frac{1}{2}, p)$-regular in $\Gamma$ for any $\varepsilon > 0$. But for any $z \in Z$ and any $0 < \varepsilon' < d \leq \frac{1}{4}$, the pair $\{N_{\Gamma}(z) \cap X, Y\}$ is not $(\varepsilon', \nu, p)$-lower-regular, because this would require that the one vertex $z \in Z$ is adjacent to at least $1 - \varepsilon)' |Y| > \frac{1}{10} pm$ vertices of $Y$, whereas a.a.s. the maximum degree of $\Gamma$ is less than $\frac{1}{10} pm$.

For Lemma 1.27, when $p \gg 1/ \log n$ we perform a similar construction, selecting two sets $X$ and $Y$ with $|X|, |Y| = n^{1-\delta}$ such that each $z \in Z$ has one neighbour in each of $X$ and $Y$. It is again easy to check that a.a.s. for at least half the vertices $z \in Z$ the pair $(N_{\Gamma}(z) \cap X, N_{\Gamma}(z) \cap Y)$ has density zero in $\Gamma$ (that is, the at most one edge which could be present is not), so in particular is not $(\varepsilon', \nu, p)$-lower-regular for any $0 < \varepsilon' < d$.

Finally, for each $\varepsilon > 0$ there exists $p_\varepsilon > 0$ such that for $0 < p < p_\varepsilon$, Lee and Sudakov [36, Proposition 6.3] construct a subgraph $G$ of $\Gamma$ with minimum degree $(1 - \varepsilon)|p m$ in which $\Omega(p^{-2})$ vertices are not in triangles. This construction in particular shows that $\Omega(p^{-2})$ vertices can fail to inherit regularity in Lemma 1.27.

2.2. Deterministic properties of the ambient graph

In this section we introduce the deterministic properties which we require of our ambient graphs $\Gamma$. We then also prove that random and bijumbled graphs have (a subset of) these properties, in Lemma 2.17 and Lemma 2.18, respectively. These are exactly the properties which we shall use in our blow-up lemma proofs.
Our first deterministic property asserts that most vertices of $\Gamma$ have close to the expected degree into any given reasonably large subset of vertices. This property will be used for both random and bijumbled graphs.

**Definition 2.12 (Neighbourhood size property $\text{NS}(\epsilon, T, \Delta)$).** Given $\epsilon > 0$ and integers $T$ and $\Delta$, we say that the graph $\Gamma$ has property $\text{NS}(\epsilon, T, \Delta)$ if the following is true for some $p$ such that $\Gamma$ has density $(1 \pm \epsilon)p$. For any set $W$ of size at least $\epsilon p^{-1} v(\Gamma)/T^2$, there are at most $\epsilon p^{-1} v(\Gamma)/T^2$ vertices $v$ outside $W$ such that $\deg_{\Gamma}(v; W) \neq (1 \pm \epsilon)p|W|$. 

Our next property concerns one-sided and two-sided regularity inheritance. Based on the regularity inheritance lemmas established in Section 2.1.3 we shall also establish this property for random as well as bijumbled graphs. Recall our convention though, that in random graphs regular pairs are lower-regular pairs and in bijumbled graphs they are fully-regular pairs.

**Definition 2.13 (Regularity inheritance property $\text{RI}(\epsilon, (\epsilon_{a,b}, \epsilon', d, T, \Delta))$.** Given $0 < \epsilon < \epsilon' < d$ and integers $T$, $\Delta$, we say that $\Gamma$ has property $\text{RI}(\epsilon, (\epsilon_{a,b}, \epsilon', d, T, \Delta))$ if we have $\epsilon_{a,b} \in [\epsilon, \epsilon']$ for each $0 \leq a, b \leq \Delta - 1$ such that the following holds for some $p$ such that $\Gamma$ has density $(1 \pm \epsilon)p$. 

If $0 \leq a \leq \Delta - 2$ and $0 \leq b \leq \Delta - 1$, if $G \subseteq \Gamma$ and if $X, Y \subseteq V(\Gamma)$ are disjoint sets with 

$$|X| \geq \epsilon' p^{\Delta-2} \frac{v(\Gamma)}{T^2} \quad \text{and} \quad |Y| \geq \epsilon' p^{\Delta-1} \frac{v(\Gamma)}{T^2},$$

and if $(X, Y)$ is $(\epsilon_{a,b}, d, p)$-regular pair in $G$, then $(N_\Gamma(v; X), Y)$ is $(\epsilon_{a+1,b}, d, p)$-regular in $G$ for all but at most $\epsilon p^{\Delta-1} v(\Gamma)/T^2$ vertices $v \in V(\Gamma) \setminus (X \cup Y)$. Furthermore, if additionally $b \leq \Delta - 2$ and 

$$|Y| \geq \epsilon' p^{\Delta-2} \frac{v(\Gamma)}{T^2},$$

then the pair $(N_\Gamma(v; X), N_\Gamma(v; Y))$ is $(\epsilon_{a+1,b+1}, d, p)$-regular in $G$ for all but at most $\epsilon p^{\Delta-2} v(\Gamma)/T^2$ vertices $v \in V(\Gamma) \setminus (X \cup Y)$.

The next property concerns the count of certain stars in $\Gamma$, and we shall call it congestion property (following [44], where a very similar property was used). We only establish this property for random graphs. For bijumbled graphs it does not hold (with any reasonable choice of parameters). Given a graph $\Gamma$, a set $U \subseteq V(\Gamma)$ and a collection $\mathcal{F}$ of pairwise disjoint $\ell$-sets in $V(\Gamma)$, we define the congestion graph $\text{CG}(\Gamma, U, \mathcal{F})$ to be the bipartite graph with vertex sets $U$ and $\mathcal{F}$ with $uF$ an edge of $\text{CG}(\Gamma, U, \mathcal{F})$ if $u \in U$ is a common neighbour in $\Gamma$ of the vertices in $F \in \mathcal{F}$.

**Definition 2.14 (Congestion property $\text{CON}(\rho, T, \Delta)$).** Given $\rho > 0$ and integers $T$ and $\Delta$, we say that $\Gamma$ has property $\text{CON}(\rho, T, \Delta)$ if the following statement is true for some $p$ such that $\Gamma$ has the density $(1 \pm \rho)p$. For each $1 \leq \ell \leq \Delta$, each $U \subseteq V(\Gamma)$ and each collection $\mathcal{F}$ of pairwise disjoint $\ell$-sets in $V(\Gamma) \setminus U$, if we have $|U| \leq |\mathcal{F}| \leq \rho v(\Gamma)$, then 

$$e(\text{CG}(\Gamma, U, \mathcal{F})) \leq 7p^\ell |U||\mathcal{F}| + \rho p^\ell \cdot \frac{v(\Gamma)}{T} |\mathcal{F}|.$$

The congestion condition will help us to verify Hall’s condition (on some linearly sized set) in order to embed many vertices at a time. However in the proof of our blow-up lemma for degenerate graphs, Lemma 1.23, we cannot use this strategy. Instead we will embed the vertices one by one in the given order. For this purpose we need to have a ‘local’ version of the congestion property, which we will want to apply to sets $U$ in the common $\Gamma$-neighbourhood of already embedded vertices. We remark that for such small sets $U$ the bound on the number of edges in the
congestion graph given by the congestion property becomes trivial. The following local congestion property is designed to give a useful bound in this case. Again, we shall only establish this property for random graphs.

**Definition 2.15 (Local congestion property $\text{LCON}(\varepsilon, T, \Delta)$).** Given $\varepsilon > 0$ and integers $T$ and $\Delta$, we say that $\Gamma$ has property $\text{LCON}(\varepsilon, T, \Delta)$ if the following statement is true for some $p$ such that $\Gamma$ has density $(1 \pm \varepsilon)p$. For each $i, \ell \geq 1$ with $i + \ell \leq \Delta$, each $U \subseteq V(\Gamma)$ of size at least $\varepsilon p^i v(\Gamma)/T^2$ and each collection $F$ of pairwise disjoint $\ell$-sets in $V(\Gamma) \setminus U$ we have

$$e(\text{CG}(\Gamma, U, F)) \leq 7p^i |U| \max \left(\varepsilon |U|, |F|\right).$$

Since for bijumbled graphs we cannot use the congestion property or local congestion property we introduce the following final deterministic property which we shall use there instead. This property is a strengthening of the neighbourhood size property, which also takes smaller sets $W$ in which we measure neighbourhoods into consideration, and further requires the number of exceptional vertices outside $W$ to be smaller. We call it ‘lopsided’ to distinguish it from property $\text{NS}(\varepsilon, T, \Delta)$, in which the maximum number of failing vertices and the minimum size of the set into which neighbourhoods are taken are the same.

**Definition 2.16 (Lopsided neighbourhood size property $\text{LNS}(\varepsilon, T, \Delta)$).** Given $\varepsilon > 0$ and integers $T$ and $\Delta$, we say that $\Gamma$ has property $\text{LNS}(\varepsilon, T, \Delta)$ if the following is true for some $p$ such that $\Gamma$ has density $(1 \pm \varepsilon)p$. For each $0 \leq j \leq \Delta - 1$ and any set $W$ of size at least $\varepsilon p^{2\Delta + j}v(\Gamma)/T^2$, there are at most $\varepsilon p^{2\Delta + j - 1}v(\Gamma)/T^2$ vertices $v$ outside $W$ such that $\deg_{\Gamma}(v; W) \neq (1 \pm \varepsilon)p|W|$.

### 2.2.1. Random graphs.

In this subsection we will show that $G_{n,p}$ has the neighbourhood size property, regularity inheritance property, congestion property and local congestion property by proving the following lemma.

**Lemma 2.17 (Deterministic properties of $G_{n,p}$).** For every $\Delta \geq 2$ and $d, \varepsilon' > 0$ there exist $\varepsilon > 0$ and $\varepsilon_{a,b} > 0$ for each $0 \leq a, b \leq \Delta - 1$ such that for every $T$ and $g > 0$ there exists $C > 0$ such that if $p \geq C(\log n/n)^{1/\Delta}$ then $G_{n,p}$ a.a.s. has

1. $\text{NS}(\varepsilon, T, \Delta)$,
2. $\text{RI}(\varepsilon, (\varepsilon_{a,b}), \varepsilon', d, T, \Delta)$,
3. $\text{CON}(g, T, \Delta)$,
4. $\text{LCON}(\varepsilon, T, \Delta)$.

The proof of part (a) is standard. Part (b) follows from the regularity inheritance lemmas, Lemmas 1.26 and 1.27. The proof of part (c) follows arguments from [44] (but is slightly different), and part (d) is proved similarly. For completeness we give the details.

**Proof of Lemma 2.17.** Given $\Delta \geq 1$ and $d, \varepsilon' > 0$, we assume without loss of generality that $(1 - \varepsilon')^\Delta > 1/2$. We choose $\varepsilon_{a,b}$ for each $0 \leq a \leq \Delta - 1$ and $0 \leq b \leq \Delta - 1$ as follows. We set $\varepsilon_{\Delta - 1, \Delta - 1} = \varepsilon'/2$ and we define the other $\varepsilon_{a,b}$ inductively. For each $a$ and $b$, we require that $\varepsilon_{a,b}$ is smaller than the $\varepsilon_0$ returned by Lemma 1.26 with input $\varepsilon_{a+1, b}/2$ and $d$ (provided $a < \Delta - 1$), and that returned with input $\varepsilon_{a, b+1}/2$ and $d$ (provided $b < \Delta - 1$), and also than the $\varepsilon_0$ returned by Lemma 1.27 with input $\varepsilon_{a+1, b+1}/2$ and $d$ (provided $a < \Delta - 1, b < \Delta - 1$). Let $\varepsilon$ be the minimum of the $\varepsilon_{a,b}$. Note that we then have $\varepsilon_{a,b} \leq \varepsilon_{b,a}$ for each $a, b$. Given $T$, let $C_1$ be the maximum of all of the constants $C$ returned by the above applications of Lemmas 1.26 and 1.27. Further, given $g$ set

$$C = \frac{100T^2 C_1 \Delta}{\varepsilon^4 g}.$$
Now let \( p \geq C(\log n/n)^{1/\Delta} \). Let \( \Gamma = G_{n,p} \). By the Chernoff bound in Theorem 2.1, since \( p > 1/n \), a.a.s. \( \Gamma \) has density \((1 \pm \min(\varepsilon, \delta))p\). In the following we condition on this occurring.

Proof of (a): Using the Chernoff bound again, if \( X \) and \( Y \) are any disjoint sets of size at least \( 6e^{-2}p^{-1} \log n \), then the probability that \( e(X, Y) \neq (1 \pm \varepsilon)p|X||Y| \) is at most \( 2\exp(-\varepsilon^2 p|X||Y|/3) \). It follows that the probability that there exist such \( X \) and \( Y \) is at most

\[
\sum_{|X||Y| \geq 6e^{-2}p^{-1} \log n} n^{\left|X|+|Y|\right|} \cdot 2 \exp\left( -\frac{\varepsilon^2 p|X||Y|}{3} \right).
\]

\[
\leq \sum_{|X||Y|} 2 \cdot 2^{\max(|X||Y|) \log n} \exp\left( -2 \max(|X|, |Y|) \log n \right)
\]

\[
\leq 2n^2 \left( 2^{\log n} \right)^{\log^2 n},
\]

where the first inequality uses \( \min(|X|, |Y|) \geq 6e^{-2}p^{-1} \log n \). It follows that a.a.s. any such pair of sets in \( \Gamma \) has density \((1 \pm \varepsilon)p\). We condition on this in the following.

Now suppose that \( X \) is any set of size at least \( \varepsilon p^{\Delta-1} n/T^2 \) in \( \Gamma \), and \( Y \) is the set of vertices outside \( X \) with fewer than \((1 - \varepsilon)p|X|\) neighbours in \( X \). Since the density of \((X,Y)\) is less than \((1 - \varepsilon)p\), and since by choice of \( C \) we have \( \varepsilon p^{\Delta-1} n/T^2 > 6e^{-2}p^{-1} \log n \), we conclude that \( |Y| < 6e^{-2}p^{-1} \log n < \varepsilon p^{\Delta-1} n/(2T^2) \) (again by choice of \( C \)). The same argument bounds the number of vertices outside \( X \) with more than \((1 + \varepsilon)p|X|\) neighbours in \( X \), so \( \Gamma \) has NS\((\varepsilon, \delta, \Delta)\) as desired.

Proof of (b): In this proof we shall apply the regularity inheritance lemmas, Lemmas 1.26 and 1.27, less than \( 3\Delta^2 \) times. Hence we can assume that the less than \( 3\Delta^2 \) properties asserted to hold a.a.s. for \( G_{n,p} \) by these applications of the lemmas hold simultaneously a.a.s for \( G_{n,p} \).

Given \( 0 \leq a \leq \Delta - 2 \) and \( 0 \leq b \leq \Delta - 1 \), suppose that \((X,Y)\) is an \((\varepsilon_{a,b}, d, p)\)-lower-regular pair in a subgraph \( G \) of \( \Gamma \) with \(|X| \geq \varepsilon' p^{\Delta-2} n/T^2 \) and \(|Y| \geq \varepsilon' p^{\Delta-1} n/T^2 \). Let \( Z \) be the set of vertices outside \( X \cup Y \) such that \((N_T(z) \cap X, N_T(z) \cap Y)\) is not \((\varepsilon_{a+1,b+1}, d, p)\)-lower-regular in \( G \). By choice of \( p \), we have \( \varepsilon' p^{\Delta-2} n/T^2 \geq C_1 p^{-2} \log n \) and \( \varepsilon' p^{\Delta-1} n/T^2 \geq C_1 p^{-1} \log n \). Therefore we can apply Lemma 1.26 to conclude that

\[
|Z| \leq C_1 p^{-1} \log n \leq \frac{2}{\varepsilon''} C p^{-1} \log n < \varepsilon p^{\Delta-1} n/T^2,
\]

as desired.

Similarly, given \( 0 \leq a \leq \Delta - 2 \) and \( 0 \leq b \leq \Delta - 2 \), suppose that \((X,Y)\) is an \((\varepsilon_{a,b}, d, p)\)-lower-regular pair in a subgraph \( G \) of \( \Gamma \) with \(|X| \geq \varepsilon' p^{\Delta-2} n/T^2 \) and \(|Y| \geq \varepsilon' p^{\Delta-2} n/T^2 \). Let \( Z \) be the set of vertices outside \( X \cup Y \) such that \((N_T(z) \cap X, N_T(z) \cap Y)\) is not \((\varepsilon_{a+1,b+1}, d, p)\)-lower-regular in \( G \). By choice of \( p \), we have \( \varepsilon' p^{\Delta-2} n/T^2 \geq C_1 p^{-2} \log n \). Therefore we can apply Lemma 1.27 to conclude that

\[
|Z| \leq C_1 \max(p^{-2}, p^{-1} \log n) \leq \frac{2}{\varepsilon''} C \max(p^{-2}, p^{-1} \log n) < \varepsilon p^{\Delta-2} n/T^2,
\]

again as desired.

Proof of (c): Given \( 1 \leq i \leq \Delta \), a set \( U \subseteq V(\Gamma) \), and a family \( F \) of pairwise disjoint \( \ell \)-sets in \( V(\Gamma) \setminus U \) with \(|U| \leq |F| \leq \ell |V(\Gamma)| \), the graph \( CG(\Gamma, U, F) \) is a random bipartite graph with edge probability \( p^\ell \) and with parts \( U \) and \( F \). So the expected number of edges of \( CG(\Gamma, U, F) \) is \( p^\ell |U||F|/T \). By the Chernoff bound in Theorem 2.1 the probability that

\[
|e(CG(\Gamma, U, F))| > 7p^\ell |U||F| + \varepsilon p^\ell n|F|/T
\]

is at most

\[
\exp\left( -\frac{\varepsilon p^\ell n|F|}{T} \right).
\]
If $|F| = m \leq gn$, then since $|U| \leq |F|$ we have $|U| \leq m$. Taking a union bound over $\ell$, over $m$, over the at most $n^{m+1}$ choices for $U$ and its size, and over the at most $n^{\Delta m}$ choices for $F$, we see that the probability of failure of $\text{CON}(g, T, \Delta)$ is at most

$$
\sum_{\ell=1}^{\Delta} \sum_{m=1}^{gn} n^{m+1} n^{\Delta m} \exp \left( -gp^{\ell} / T \right) \leq \Delta \sum_{m=1}^{gn} \exp \left( 2\Delta m \log n - gp^{\Delta m} / T \right)
$$

$$
\leq \Delta \sum_{m=1}^{gn} \exp \left( - (\Delta + 3) \log n \right) < \Delta n \exp \left( - (\Delta + 2) \log n \right) < 1 / n,
$$

where the second inequality uses $p^{\Delta n} \geq C \log n$ and the choice of $C$. We conclude that $G_{n,p}$ a.a.s. has $\text{CON}(g, T, \Delta)$ as desired.

Proof of (d): Given integers $i, \ell \geq 1$ such that $i + \ell \leq \Delta$, a set $U \subseteq V(\Gamma)$ of size at least $e \log n / T$, and a family of pairwise disjoint $\ell$-sets $F$ in $V(\Gamma) \setminus U$, the graph $\text{CG}(\Gamma, U, F)$ is a random bipartite graph with edge probability $p^\ell$ and parts $U$ and $F$. The expected number of edges it contains is therefore $p^\ell |U| |F|$.

We separate two cases. If $|F| \leq \varepsilon |U|$, then we have $Ee(\text{CG}(\Gamma, U, F)) \leq e p^\ell |U|^2$, so by Theorem 2.1 the probability of $e(\text{CG}(\Gamma, U, F)) \geq 7e p^\ell |U|^2$ is at most

$$
\exp \left( -6e p^\ell |U|^2 \right) \leq \exp \left( -6e p^{\ell + \varepsilon |F| / T^2} \right) \leq \exp(-3\Delta |U| \log n).
$$

Taking the union bound over the at most $\Delta^2$ choices of $i$ and $\ell$, at most $n^2$ choices of $|U|$ and $|F| \leq |U|$, the at most $n^{\ell+1}|U|$ choices of $U$ and the at most $n^{\Delta |U|}$ choices of $F$, we see that the probability that there exists such a choice with $e(\text{CG}(\Gamma, U, F)) \geq 7e p^\ell |U|^2$ is at most

$$
\Delta^2 n^2 n^{(\Delta+1)|U|} \exp(-3\Delta |U| \log n)
$$

which tends to zero as $n$ tends to infinity.

Next we consider the case $|F| \geq \varepsilon |U|$. Again by Theorem 2.1 the probability of $e(\text{CG}(\Gamma, U, F)) \geq 7e p^\ell |U||F|$ is at most

$$
\exp \left( -6e p^\ell |U||F| \right) \leq \exp \left( -6e p^{\ell + \varepsilon |F| / T^2} \right) \leq \exp(-3\varepsilon^{-1} \Delta |F| \log n).
$$

Taking the union bound over the at most $\Delta^2$ choices of $i$ and $\ell$, at most $n^2$ choices of $|U|$ and $|F|$, the at most $n^{\ell+1}|F|$ choices of $U$ and the at most $n^{\Delta |F|}$ choices of $F$, we see that the probability that there exists such a choice with $e(\text{CG}(\Gamma, U, F)) \geq 7e p^\ell |U||F|$ is at most

$$
\Delta^2 n^2 n^{(\Delta+1)/\varepsilon |F|} \exp(-3\varepsilon^{-1} \Delta |F| \log n)
$$

which tends to zero as $n$ tends to infinity. We conclude that $G_{n,p}$ a.a.s. has $\text{LCON}(\varepsilon, T, \Delta)$ as desired. \hfill $\Box$

2.2.2. Bijumbled graphs. In this subsection we will verify the neighbourhood size property, the regularity inheritance property, and the lopsided neighbourhood size property for bijumbled graphs, as stated in the following lemma.

Lemma 2.18 (Deterministic properties of bijumbled graphs). For each $\Delta \in \mathbb{N}$ and $d, \varepsilon' > 0$ there exist $\varepsilon_{a,b} > 0$ for $0 \leq a, b \leq \Delta$ and $\varepsilon > 0$ such that for each $T \in \mathbb{N}$ there is $c > 0$ such that if $p > 0$ then any $(p, \beta)$-bijumbled graph on $n$ vertices has

(a) $\text{NS}(\varepsilon, T, \Delta + 1)$ if $\beta \leq cp^{\Delta+1} n$,

(b) $\text{RI}(\varepsilon, (\varepsilon_{a,b}, \varepsilon', d, T, \Delta + 1)$ if $\beta \leq cp^{\Delta+2} n$,

(c) $\text{LNS}(\varepsilon, T, \Delta)$ if $\beta \leq cp^{(3\Delta+1)/2} n$.

Let us briefly justify why we cannot use the congestion property in bijumbled graphs. Indeed, blowing up a bijumbled graph by a factor $\Delta$—that is, replacing vertices with independent $\Delta$-sets and edges with complete bipartite graphs—degrades
the bijumbledness only slightly, but in this blow up the congestion condition fails badly: If we choose \( \ell = \Delta \) and as \( \Delta \)-tuples the blow ups of vertices, then the congestion property fails by a factor \( p^{\Delta - 1} \).

In the proof of Lemma 2.18, Part (b) is a consequence of the regularity inheritance lemmas, Lemmas 2.5 and 2.6. Parts (a) and (c) use the following easy consequences of the definition of bijumbled graphs.

**Proposition 2.19.** Let \( \Gamma \) be a \((p, \beta)\)-bijumbled graph on \( n \) vertices and \( W \subseteq V(\Gamma) \).

(a) If \( \varepsilon \geq 2\beta \frac{1}{2} \), then \( e(\Gamma) = (1 + \varepsilon)p\binom{n}{2} \).

(b) \( \deg_G(v; W) \neq (1 + \varepsilon)p|W| \) for at most \( 2\beta^2/(\varepsilon^2p^2|W|) \) vertices \( v \in V(\Gamma) \setminus W \).

**Proof.** For any balanced cut \( U \cup U' = V(\Gamma) \) we have

\[
epsilon(U, U') = p|U||U'| + \beta \sqrt{|U||U'|} = \left(1 + 2\beta \frac{1}{2}\right)p\binom{n}{2} = (1 + \varepsilon)p\binom{n}{2}.
\]

Since the density of \( \Gamma \) equals the average density over all balanced cuts of \( \Gamma \), part (a) follows.

For (b) let \( U \) be the set of \( v \in V(\Gamma) \setminus W \) with \( \deg_G(v; W) > (1 + \varepsilon)p|W| \). We have

\[
(1 + \varepsilon)p|U||W| < e(U, W) \leq p|U||W| + \beta \sqrt{|U||W|}
\]

and hence \( |U| < \beta^2/(\varepsilon^2p^2|W|) \). Similarly, the number of \( v \in V(\Gamma) \setminus W \) with \( \deg_G(v; W) < (1 - \varepsilon)p|W| \) is also at most \( \beta^2/(\varepsilon^2p^2|W|) \) and (b) follows. \( \square \)

We can now prove Lemma 2.18.

**Proof of Lemma 2.18.** Given \( \Delta \geq 2 \) and \( d, \varepsilon' > 0 \), we assume \((1 - \varepsilon')^\Delta > 1/2\). We choose \( \varepsilon_{a,b} \) for each \( 0 \leq a \leq \Delta \) and \( 0 \leq b \leq \Delta \) as follows. We set \( \varepsilon_{\Delta, \Delta} = \varepsilon'/2 \) and we define the other \( \varepsilon_{a,b} \) inductively. For each \( a \) and \( b \), we require that \( \varepsilon_{a,b} \) is smaller than the \( \varepsilon_0 \) returned by Lemma 2.5 with input \( \varepsilon_{a+1,b}/2 \) and \( d \) (provided \( a < \Delta \)), and that returned with input \( \varepsilon_{a,b+1}/2 \) and \( d \) (provided \( b < \Delta \)), and the \( \varepsilon_0 \) returned by Lemma 2.6 with input \( \varepsilon_{a+1,b+1}/2 \) and \( d \) (provided \( a < \Delta, b < \Delta \)). Let \( \varepsilon \) be the minimum of the \( \varepsilon_{a,b} \). Note that we then have \( \varepsilon_{a,b} = \varepsilon_{h,a} \) for each \( a, b \). Let \( \varepsilon' \) be the minimum of all the constants \( \varepsilon_0 \) returned by the applications of Lemmas 2.5 and 2.6. Given \( T \), let

\[
e = \frac{1}{2} \varepsilon'^2 \varepsilon' T^{-2}.
\]

Let \( \Gamma \) be a \((p, \beta)\)-bijumbled graph on \( n \) vertices with \( \beta \leq cp^{\Delta - 1}n \). By Proposition 2.19(a) the graph \( \Gamma \) has density \((1 + \varepsilon)p\).

Proof of (a): Let \( W \subseteq V(\Gamma) \) be a set with \(|W| \geq \varepsilon p^{\Delta} n / T^2 \). By Proposition 2.19(b) the number of vertices \( v \in V(\Gamma) \setminus W \) such that \( \deg_G(v; W) \neq (1 + \varepsilon)p|W| \) is at most

\[
\frac{2\beta^2}{\varepsilon^2p^2 \cdot \varepsilon p^{\Delta - 1} n T^{-2}} \leq \frac{2\varepsilon^2T^2 p^{\Delta} n}{\varepsilon^3} \leq \varepsilon p^{\Delta} \frac{n}{T^2},
\]

and hence \( \Gamma \) has NS(\( \varepsilon, T, \Delta + 1 \)).

Proof of (c): Assume \( \beta \leq cp^{(\Delta+1)/2} \). Let \( 0 \leq j \leq \Delta \) and \( W \subseteq V(\Gamma) \) be a set with \(|W| \geq \varepsilon p^{\Delta - j} n / T^2 \). By Proposition 2.19(b) the number of vertices \( v \in V(\Gamma) \setminus W \) such that \( \deg_G(v; W) \neq (1 + \varepsilon)p|W| \) is at most

\[
\frac{2\beta^2}{\varepsilon^2p^2 \cdot \varepsilon p^{\Delta - j - 1} n T^{-2}} \leq \frac{2\varepsilon^2T^2 p^{\Delta - j - 1} n}{\varepsilon^3} \leq \varepsilon p^{\Delta - j - 1} \frac{n}{T^2},
\]

and hence \( \Gamma \) has LNS(\( \varepsilon, T, \Delta \)).

Proof of (b): We shall prove that \( \Gamma \) has RI(\( \varepsilon, (\varepsilon_{a,b}), \varepsilon', d, T, \Delta + 1 \)) if \( \beta \leq \varepsilon p^{\Delta - 2} n \) by contradiction. Given \( 0 \leq a \leq \Delta - 1 \) and \( 0 \leq b \leq \Delta \), suppose that \((X,Y)\) is an \((\varepsilon_{a,b}, d, p)\)-fully-regular pair in \( G \subseteq \Gamma \) with \(|X| \geq \varepsilon' p^{\Delta - 1} n / T^2 \) and \(|Y| \geq \varepsilon' p^\Delta n / T^2 \).
Let $Z$ be the set of vertices $z \in V(\Gamma) \setminus (X \cup Y)$ such that $(N_T(z) \cap X, Y)$ is not $(\varepsilon_{a+1,b,d,p})$-fully-regular in $G$. Assume for contradiction that $|Z| \geq \varepsilon p^{\Delta-1} n/T^2$. Then we have $\max (|X||Y|, |X||Z|) \geq \varepsilon p^{2\Delta-2} n^2/T^3$. Therefore, by choice of $c$ and since $(\log_2 \frac{1}{p})^{-1} > p$, we see that

$$\beta \leq c p^{\Delta+2} n \leq c' \sqrt{\varepsilon \varepsilon' p^{\Delta+2} n^2/T^2} < c' p^2 (\log_2 \frac{1}{p})^{-1/2} \sqrt{\max (|X||Y|, |X||Z|)} ,$$

and thus, $\Gamma$ is bijumbled enough to apply Lemma 2.5, whose statement contradicts the assumption on $Z$.

Similarly, given $0 \leq a,b \leq \Delta - 1$, suppose that $(X, Y)$ is an $(\varepsilon_{a,b,d,p})$-fully-regular pair in $G \subseteq \Gamma$ with $|X| \geq \varepsilon' p^{\Delta-1} n/T^2$ and $|Y| \geq \varepsilon' p^{\Delta-1} n/T^2$. Let $Z$ be the set of vertices $z \in V(\Gamma) \setminus (X \cup Y)$ such that $(N_T(z) \cap X, N_T(z) \cap Y)$ is not $(\varepsilon_{a+1,b+1, d,p})$-fully-regular in $G$. Again, assuming for contradiction that $|Z| \geq \varepsilon' p^{\Delta-1} n/T^2$, we have

$$\max (|X||Y|, |Y||Z|, |X||Z|) \geq (c')^2 p^{2\Delta-2} n^2/T^4,$$

so that by choice of $c$, we see that

$$\beta \leq c p^{\Delta+2} n \leq c' \sqrt{c' \varepsilon' p^{\Delta+2} n^2/T^2} < c' p^3 \sqrt{\max (|X||Y|, |Y||Z|, |X||Z|)} ,$$

hence the conclusion of Lemma 2.6 contradicts the assumption on $Z$. \hfill \Box

2.3. Preprocessing

This section constitutes the first step towards the proofs of our blow-up lemmas. More or less, one can think of this section as showing that, for each of our blow-up lemmas, it suffices to prove a corresponding blow-up lemma in which several extra conditions are required. This reduction, as mentioned in the proof overview (Section 1.4) involves refining the partitions $\mathcal{X}_{BL}$ of $H$ and $\mathcal{V}_{BL}$ of $G$ given by the user of the blow-up lemma to obtain new partitions $\mathcal{X}$ of $H$ and $\mathcal{V}$ of $G$, which in turn entails replacing the reduced graphs $R_{BL}$ and $R'_{BL}$ with new graphs $R$ and $R'$, and altering various constants mentioned in the blow-up lemmas. It is convenient to keep using the letter choices seen in our blow-up lemmas for these altered partitions, graphs and constants in the proofs to come, so in this section we use the suffix BL for objects mentioned in one of our blow-up lemmas which we will alter for the proof, following the style of the previous sentence.

In addition to formally defining ‘blow-up lemma with several extra conditions’, which we refer to as the general setup for our proofs, we define the concept of good partial embedding, which is a central concept of our proofs. We show that obtaining the general setup implies that the trivial partial embedding (that is, with no vertices embedded) is a good partial embedding.

This section is structured as follows. In Section 2.3.1 we collect and describe, mainly as reference for the reader, the various constants that appear in our blow-up lemma proofs. Sections 2.3.2 and 2.3.3 then list the properties we require of the refined partitions of $H$ and $G$, respectively. In Section 2.3.4 we prove that partitions with these desired properties can be obtained from partitions supplied to our blow-up lemmas. In Section 2.3.5 we define some further key concepts used in our embedding procedures, such as candidate sets, available candidate sets and good partial embeddings. These concepts were mentioned already in the proof overview. In Section 2.3.6 the general setup that we shall use in our proofs. Finally, in Section 2.3.7, we define the notion of bad vertices which will, as explained in the proof overview, be vertices avoided in our embedding process, and prove Lemma 2.25 which states that most vertices in candidate sets are not bad.
2.3.1. Constants. In the following chapters, sections and lemmas we will use various constants. These are listed below, with their meaning. Firstly, we have the following constants which are chosen by the user of our blow-up lemmas.

- $\Delta$ is the maximum degree of $H$.
- $\Delta_J$ is the maximum number of times a vertex of $\Gamma$ is a restricting vertex, that is, appears in a set $J_x$ for $x \in V(H)$.
- $D$ is a constant governed by the degeneracy of $H$, which is only used in the degenerate graphs blow-up lemma.
- $d$ is the minimum $p$-density of regular pairs.

The user of our blow-up lemmas further chooses the constants $\Delta_{R',R}$, $\alpha_{BL}$, $\zeta_{BL}$, $\kappa_{BL}$ and $r_{1BL}$ (the last after being given $\varepsilon_{BL}$ and $\varrho_{BL}$). However our preprocessing in Lemma 2.22 changes the values of these constants (and the reduced graphs $R$ and $R'$), giving the following constants which we use in the rest of the paper.

- $\Delta_{R'}$ is the maximum degree of the reduced graph $R'$, which captures super-regular pairs.
- $\alpha$ is the fraction of a part required to be potential buffer vertices.
- $\zeta$ is the minimum relative size of any image restriction set $I_x$.
- $\kappa$ is the balancing factor (greater than 1), bounding $|V_i|/|V_j|$ for any pair of clusters of $G$ (and hence parts of $H$).
- $r_1$ is the upper bound on the number of clusters.

The blow-up lemmas guarantee the existence of the following constants.

- $\varepsilon_{BL}$ is the regularity we require in the regular partition provided by the user of the blow-up lemma.
- $\varrho_{BL}$ is the fraction of vertices in each part of the partition of $H$ supplied by the user which may be image restricted.
- $C$ is large, only appears in the random graphs blow-up lemmas and is the constant factor in the bound on the probability $p$.
- $c$ is small, only appears in the bijumbled graphs blow-up lemma and is the constant factor in the bound on the bijumbledness error term $\beta$.

Furthermore, the most important additional constants appearing in the proofs of the blow-up lemmas are the following.

- $\mu$ is the fraction of each cluster of $G$ contained in each of the “small” sets of the partition of $G$, thus $|V^a_i| = |V^s_i| = |V^buf_i| = \mu|V_i|$ for each $i$.
- $\vartheta$ comes into the maximum fraction $\varrho \vartheta$ of each part of $H$ which may be image restricted. It is equal to either zero (proving Lemmas 1.21 and 1.23) or $\Delta$ (proving Lemma 1.25).
- $\varrho$ appears in the maximum fraction $\varrho \vartheta$ of each part of $H$ which may be image restricted. We also use $\varrho$ for a second purpose (which does not conflict with the first one): it is the fraction of non-image restricted vertices in $X_i$ which may enter the queue during the RGA without causing the RGA to fail. Hence in total at most $2\varrho |X_i|$ vertices of $X_i$ may enter the queue during the RGA.
- $\varepsilon'$ is the global worst case regularity appearing in the proof of the blow-up lemmas. That is, whenever we use the fact that a pair is $(\cdot,d,p)$-regular in $G$, it will be at worst $(\varepsilon',d,p)$-regular.
- $\varepsilon_{a,b}$ is the worst case regularity in the proofs between underlying restriction sets of adjacent vertices in $H$ with respectively $a$ and $b$ previously embedded neighbours (see Section 2.3.5 for definitions of these terms).
\( \varepsilon \) is the initial regularity after pre-processing, and also controls the deterministic properties we require of \( \Gamma \).

In order to make our proofs work it is enough to have constants in the size order
\[
0 < \varepsilon \ll \varepsilon_{a,b} \leq \varepsilon' \ll \varrho \ll \mu \ll \alpha, \zeta, d, \Delta^{-1}, \Delta_{R'}^{-1}, \Delta_{J}^{-1}, D^{-1}, \kappa^{-1}
\]
and
\[
0 < c, C^{-1} \ll r_{1}^{-1}, \varepsilon
\]
where by \( x \ll y \) we mean that there is a non-decreasing function \( f : (0, 1] \rightarrow (0, 1] \)
such that our proof works if \( 0 < a \leq f(b) \). Observe that the constants on the right hand
sides are (effectively) chosen by the user of the blow-up lemma. We remark that one can then
safely read this paper assuming that, for example, any function of \( \varrho \) appearing in the
proofs tending to zero with \( \varrho \) is much smaller than any function of \( \mu \) appearing in the
proofs tending to zero with \( \mu \). However, for the convenience of the reader wishing to verify
the proofs, we specify our constants (more or less) explicitly in each of the following results. We have made no attempt to optimise the values we give.

### 2.3.2. The partition of \( H \).

In Section 2.3.6 we will refine the partition \( X^{BL} \) of \( H \) given as input to one of our
blow-up lemmas to obtain a partition \( X = \{X_{i}\}_{i \in[r]} \) of \( V(H) \) with some additional
properties, which we shall need in our proofs. We shall also refine the given partition of \( G \) (see Section 2.3.3 for the properties we require from this refined partition of \( G \)). In this subsection we will define the properties
that we require from the refined partition of \( H \). To put this definition into context
(and provide some explanations on how it fits to the refinement of the partition
of \( G \) in the subsequent subsection) we first need some explanations concerning our
strategy in the proofs of the blow-up lemmas.

We shall (in a series of steps, some of which are performed in this preprocessing,
and some of which are performed on later parts of the proof) construct subsets
\( X_{i}^{main} \), \( X_{i}^{q} \), \( X_{i}^{c} \) and \( X_{i}^{buf} \) of \( X_{i} \). We define
\[
X_{i}^{main} = \bigcup_{i \in [r]} X_{i}^{main}, \quad X_{i}^{q} = \bigcup_{i \in [r]} X_{i}^{q}, \quad X_{i}^{c} = \bigcup_{i \in [r]} X_{i}^{c}, \quad X_{i}^{buf} = \bigcup_{i \in [r]} X_{i}^{buf}.
\]
Here, \( X_{i}^{buf} \subseteq \bar{X}_{i} \) contains the buffer vertices, which will be chosen from the
set of potential buffer vertices in this preprocessing step in Section 2.3.6. The set \( X_{i}^{c} \) is
only used in the proof of the random graphs blow-up lemma, Lemma 1.21; in
the other proofs we set \( X_{i}^{c} = \emptyset \). The vertices in \( X_{i}^{c} \) shall form the reserved cliques
which we will use to fix so-called buffer defects (see Section 3.1 for more details).

We choose this set of vertices at the beginning of the proof of Lemma 1.21. Once we
chose \( X_{i}^{buf} \) and \( X_{i}^{c} \), we set \( X_{i}^{main} := X_{i} \setminus (X_{i}^{buf} \cup X_{i}^{c}) \). These sets will have sizes
\( |X_{i}^{buf}| = 4\mu|X_{i}| \) and \( |X_{i}^{c}| \leq \mu|X_{i}| \) and hence \( (1 - 5\mu)|X_{i}| \leq |X_{i}^{main}| \leq (1 - 4\mu)|X_{i}| \).
The set \( X_{i}^{q} \subseteq X_{i}^{main} \) will form the queue and only gets chosen during the embedding,
in the random greedy algorithm; it will also be of size at most \( \mu |X_{i}| \). We remark
that in the following definition we shall only refer to the sets \( X_{i} \) and \( X_{i}^{buf} \). We only
close to mention the sets \( X_{i}^{q} \) and \( X_{i}^{c} \) here as well because we need them to motivate
the refinement of the partition of \( G \) in the next subsection.

The properties that we require of our refined partition and the buffer vertices
are collected in the following definition. We remark that the reduced graphs \( R \)
and \( R' \) used for this refined partition are blow-ups of the reduced graphs \( R_{BL} \) and \( R'_{BL} \) provided to the respective blow-up lemma. These properties bound the number of buffer vertices per part, and state that the first and second
neighbourhoods of buffer vertices lie along edges of \( R' \). They also assert that any pair of vertices within
one part \( X_{i} \), but also any pair of buffer vertices, and any pair consisting of a buffer
vertex and an image restricted vertex, are far apart. Note that this implies that
$H[X_i, X_j]$ forms a matching for any pair of parts $X_i$ and $X_j$, and it also implies that buffer vertices may not be image restricted. We further require that all vertices in any one $X^b_i$ have the same degree, and distinguish between clique buffers and non-clique buffers. The latter will be important for the selection of reserved cliques in the proof of Lemma 1.21 (Section 3.1).

**Definition 2.20 (Good $H$-partition).** We say that a partition $\mathcal{X} = \{X_i\}_{i \in [r]}$ of $V(H)$ is a good $H$-partition for reduced graphs $R' \subseteq R$ on vertex sets $[r]$, with buffer $\bar{X}$, buffer vertices $X^b \subseteq \bigcup \bar{X}$ and image restricting vertices $\mathcal{J} = \{J_x\}_{x \in V(H)}$, if $\mathcal{X}$ is an $R$-partition and the following conditions are satisfied for each $i, j, k \in [r]$.

1. $(H_1)$ $(H, \mathcal{X})$ is an $R$-partition and $\bar{X}$ is an $(\alpha, R')$-buffer for $H$.
2. $(H_2)$ $\mathcal{H}(x, y) \geq 10$ and $J_x \cap J_y = \emptyset$ for each $x, y \in X_i$ with $x \neq y$.
3. $(H_3)$ $|X^b_i| = 4\mu|X_i|$, and $|\{x \in X_i : x \in N(X^b_i)\}| \leq 4\kappa \Delta^R H_i|X_i|$.
4. $(H_4)$ $\mathcal{H}(x, y) \geq 5$ for each $x, y \in X^b_i$ with $x \neq y$.
5. $(H_5)$ $\mathcal{H}(x, y) \geq 3$ for each $x \in X^b_i$ and each image restricted $y$.
6. $(H_6)$ All vertices of $X^b_i$ have degree $b$ for some $0 \leq b \leq \Delta$. We then call $X^b_i$ a degree-$b$ buffer.
7. $(H_7)$ Either all or none of the vertices of $X^b_i$ are in copies of $K_{\Delta+1}$. We then call $X^b_i$ a clique buffer or non-clique buffer, respectively.
8. $(H_8)$ If $X^b_i$ is a clique buffer, then at least $\frac{1}{2\Delta+4}\alpha|X_i|$ vertices of $\bar{X}_i$ are in copies of $K_{\Delta+1}$ which do not contain either vertices of $X^b_i$ or image restricted vertices.

Note that in this definition we do need to know about image restricting vertices (since they play a rôle in $(H_2)$), which are in $\Gamma$, and we do need to know which vertices of $H$ are image restricted (since they play a rôle in $(H_5)$), but we do not need to know anything about $G$.

### 2.3.3. The partition of $G$

We next define the properties we require from our refined partition of $G$. Firstly, we shall subdivide each cluster of the partition $\mathcal{V}_{BL}$ of $G$, which is given as input to one of our blow-up lemmas, into several new clusters to obtain a new partition $V$ matching the refined partition of $H$. Further, as also mentioned in the proof overview, we shall partition each new cluster $V_i$ into parts $V^m_i, V^q_i, V^c_i$ and $V^b_i$. Again, we set

$$
V^m = \bigcup_{i \in [r]} V^m_i, \quad V^q = \bigcup_{i \in [r]} V^q_i, \quad V^c = \bigcup_{i \in [r]} V^c_i, \quad V^b = \bigcup_{i \in [r]} V^b_i.
$$

As previously indicated, each vertex $x \in V(H)$ will be embedded into the cluster $V_i$ such that $x \in X_i$. In the rest of the paper, we often refer to $V_i$ simply by $V(x)$, to $V^m_i$ by $V^m(x)$, and so on, to avoid the use of indices whenever these are not important.

As explained before, most of the vertices from $V^m_i$ will be used to embed most vertices of $X^m_i$, while some vertices from $V^q_i$ will be used to embed the remainder, which constitutes the clique reservoir $V^c_i$. We also have the set $V^b_i$, the clique reservoir, that we will use in the proof of the random graphs blow-up lemma to embed some of the vertices in $V^c_i$. Some vertices of each $X^c_i$ may need to be used to fix so-called buffer defects, and these could be embedded anywhere in $V_i$. All other vertices of each $X^c_i$ get embedded to $V^c_i$. Finally, the vertices from $V^b_i$ together with all remaining vertices of $V^m_i \cup V^q_i \cup V^c_i$, that is, vertices that were not used when embedding $X^m_i \cup X^q_i \cup X^c_i$, will be used for embedding the buffer vertices $X^b_i$.

The properties we require of a good $G$-partition are as follows. It must be a regular $R$-partition, and furthermore the one-sided and two-sided inheritance properties of $\mathcal{V}_{BL}$ on $R'_{BL}$ must be transferred to $V$ on $R'$. The super-regularity properties of $\mathcal{V}_{BL}$ on $R'_{BL}$ must be transferred to give minimum degree conditions for
each of the parts $V_i^{\text{main}}, V_i^q$ and $V_i^c$. Finally, we need to create new image restriction sets $I$ from the supplied $I^{\text{BL}}$ which with the (unchanged) $J$ form a restriction pair.

**Definition 2.21.** (Good $G$-partition). Given a good $H$-partition of $V(H)$ for reduced graphs $R' \subseteq R$ on vertex sets $[r]$, with buffer $\hat{X}$ and buffer vertices $X_{i}^{\text{buf}} \subseteq \hat{X}$, with image restrictions $I = \{I_x\}_{x \in V(H)}$, and restricting vertices $J = \{J_x\}_{x \in \nu(V)}$, we say that a partition $\mathcal{V} = \{V_i\}_{i \in [r]}$ of $V(G)$ with a partition $V_i = V_i^{\text{main}} \cup V_i^q \cup V_i^c \cup V_{i}^{\text{buf}}$ of each cluster is a good $G$-partition for $R' \subseteq R$, if the following conditions are satisfied.

$(G1)$ For each $i$ we have $|V_i^{\text{main}}| = (1 - 3\mu)|V_i|$ and $|V_i^q| = |V_i^c| = |V_i^{\text{buf}}| = \mu|V_i|$.

$(G2)$ $(G, \mathcal{V})$ is an $(\epsilon, d, p)$-regular $R'$-partition, which has one-sided inheritance on $R'$, and two-sided inheritance on $R'$ for $\hat{X}$.

$(G3)$ For each $ij \in R'$ and $v \in V_i$, we have
\[
\deg_G(v; V_j^{\text{main}}) \geq (1 - 3\mu)(d - \epsilon) \max \left(p|V_j|, \deg_G(v; V_j)/2 \right),
\]
\[
\deg_G(v; V_j^q) \geq \mu(d - \epsilon) \max \left(p|V_j|, \deg_G(v; V_j)/2 \right),
\]
\[
\deg_G(v; V_j^c) \geq \mu(d - \epsilon) \max \left(p|V_j|, \deg_G(v; V_j)/2 \right).
\]

$(G4)$ For each $x \in X_i$, we have
\[
|I_x \cap V_i^{\text{main}}| \geq (1 - \epsilon)(1 - 3\mu)|I_x|,
\]
\[
|I_x \cap V_i^q|, |I_x \cap V_i^c|, |I_x \cap V_i^{\text{buf}}| \geq (1 - \epsilon)\mu|I_x|,
\]
\[
|N^*(J_x; V_i^{\text{main}})| = (1 \pm \epsilon)(1 - 3\mu)p^1|V_i|, \text{ and}
\]
\[
|N^*(J_x; V_i^q)|, |N^*(J_x; V_i^c)|, |N^*(J_x; V_i^{\text{buf}})| = (1 \pm \epsilon)\mu p^{1/2}|V_i|.
\]

$(G5)$ $I$ and $J$ form a $(gp^0, \zeta, \Delta, \Delta_J)$-restriction pair for the partitions $\hat{X}$ and $\mathcal{V}$.

$(G6)$ For each $i$, the number of image restricted vertices in $X_i$ is at most $gp^0|X_i|$. Note that the sizes of the sets $V_i^{\text{main}}, V_i^q, V_i^c, V_i^{\text{buf}}$ are not be the same as the sizes of the sets $X_i^{\text{main}}, X_i^q, X_i^c$ and $X_i^{\text{buf}}$. For example we will have $|X_i^{\text{main}}| = (1 - 4\mu)|X_i| < (1 - 3\mu)|V_i| = |V_i^{\text{main}}|$, and $|X_i^q|, |X_i^c| \leq 2\mu|X_i|$, which is much smaller than $\mu|V_i|$. Thus, the subsets $X_i^{\text{main}}, X_i^q, X_i^c$ of $X_i$ will be significantly smaller than the corresponding set in $V_i$, while the set $X_i^{\text{buf}}$ is larger than $V_i^{\text{buf}}$. We also note that we do require two-sided inheritance for $\hat{X}$ in $(G2)$, not just for $X_i^{\text{buf}}$. In the proof of Lemma 1.21, when dealing with buffer defects, we will make use of this stronger statement.

### 2.3.4. Obtaining a good $H$-partition and a good $G$-partition

The following lemma proves that we can obtain good partitions of $H$ and $G$ from the partitions provided to our blow-up lemmas. The proof of this lemma is straightforward, though not short since several conditions must be checked. Briefly, the idea is that we will draw an auxiliary graph $F_i$ on each part $X_i^{\text{BL}}$ of $\hat{X}^{\text{BL}}$ with edges joining pairs of vertices which are at distance less than 10 or share an image restricting vertex. We will apply Lemma 2.4 to each of these graphs together with the set $\hat{X}$ to obtain a partition $\hat{X}$ and reduced graphs $R$ and $R'$ satisfying the first three conditions of a good $H$-partition, and then construct the sets $X_i^{\text{buf}}$ greedily to obtain the remaining properties. We will then randomly refine the partition $V_i^{\text{BL}}$ to obtain a matching partition $V$, and further randomly split each part $V_i$ of $V$ into $V_i^{\text{main}}, V_i^q, V_i^c$ and $V_i^{\text{buf}}$. We will use Theorem 2.1 to obtain concentration for various set sizes in the randomly chosen parts, which in particular show that with high probability we obtain the desired good $G$-partition. We stress that although $p$ with a lower bound reminiscent of our random graph blow-up lemmas makes an appearance in the lemma statement, this does not mean we are going to assume
Γ is a random graph. We will simply need to know that the quantity $p^n$ is large compared to $\log n$.

Lemma 2.22 (Good partitions lemma). For all positive integers $\Delta$, $\Delta_{R'}$, $\Delta_J$, $r_1$, $\theta$ and $b$, all $\kappa > 2$, all $\alpha, \zeta, d, \varepsilon > 0$, and all sufficiently small $\mu, \varrho > 0$, there exists $C$ such that the following holds whenever $p \geq C\left(\frac{\log n}{n}\right)^{1/b}$ and $n > C$. Let $\delta = \frac{1}{8}(\Delta + \Delta_J)^{-10}$ and $r \leq r_1$.

Let $R_{BL}$ be a graph on $r_{BL} = \delta r$ vertices and let $R'_{BL} \subseteq R_{BL}$ be a spanning subgraph with $\Delta(R'_{BL}) \leq \Delta_{R'_{BL}} = \delta \Delta_R$. Let $H$ and $G \subseteq \Gamma$ be graphs with $\frac{1}{\kappa}$-balanced size-compatible vertex partitions $X^{BL} = \{X_i^{BL}\}_{i \in [r_{BL}]}$ and $V^{BL} = \{V_i^{BL}\}_{i \in [r_{BL}]}$, respectively, which have parts of size at least $\frac{\theta}{\kappa}$.

Furthermore, we have $\Delta(\delta_{BL})_{i \in [r_{BL}]}$ in $F$ than $10$ in $x$ between $\delta \Delta_{R'}$. Observe that $\Delta(\delta_{BL})_{i \in [r_{BL}]}$ is a random graph. We will simply need to know that the quantity $p^n$ is large compared to $\log n$.

For each $i$, $i \in [r_{BL}]$, we define a graph $F_i$ on the vertex set $X'^{BL}_i$ by putting an edge between $x$ and $x'$ whenever either $J_x \cap J_{x'} \neq \emptyset$ or $x$ and $x'$ are at distance less than $10$ in $H$. Observe that $\Delta(F_i) \leq (\Delta + \Delta_J)^{10}$. We now apply Lemma 2.4 to $F_i$ with the vertex partition $X'^{BL}_i$. This returns an equitable partition of the vertices $X'^{BL}_i$ into independent sets in $F_i$, which also partitions $X'^{BL}_i$ equitably. We do this for each $i \in [r_{BL}]$ to obtain a partition $X$ of $V(H)$ into $(\Delta + \Delta_J)^{10}r_{BL} = \delta^{-1}r_{BL} = r$ parts and a family $X = \{X_i\}_{i \in [r]}$. We let $R'$ be the graph obtained from $R_{BL}$ by replacing each vertex with an independent set on $(\Delta + \Delta_J)^{10} = \delta^{-1}$ vertices, and each edge with a complete bipartite graph between the corresponding sets. We obtain $R'$ similarly from $R'_{BL}$. Thus $(H, X)$ is an $R$-partition, satisfying (H2) by construction. Furthermore, we have $\Delta(R') = \delta^{-1}\Delta(R'_{BL}) \leq \Delta_R$.

Since each $X'^{BL}_i$ has size at least $\frac{\theta}{\kappa}$ and is equipartitioned into $\delta^{-1}$ parts, we see that, because $n$ is chosen sufficiently large, we have $|X_i| \geq \frac{n}{(\kappa r_{BL})}$ for each $i \in [r]$. Given $i, j \in [r]$, let $i', j' \in [r_{BL}]$ be such that $X_i \subseteq X'^{BL}_{i'}$ and $X_j \subseteq X'^{BL}_{j'}$. Then we have $|X_i| \leq \delta |X'^{BL}_{i'}| + 1 \leq \frac{\delta}{2} |X'^{BL}_{i'}| + 1 \leq \frac{\delta}{2}(|X_j| + 1) + 1 \leq \kappa |X_j|$ where the final inequality is since $|X_j| \geq \frac{n}{(\kappa r_{BL})}$ and $n$ is sufficiently large. Thus $X$ is $\kappa$-balanced.
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Given $i \in [r]$ let $i' \in [r_{BL}]$ be such that $X_i \subseteq X_{i'}^{BL}$. Then we have

$$|\tilde{X}_i| \geq \delta |X_{i'}^{BL}| - 1 \geq 2\alpha \delta |X_{i'}^{BL}| - 1 \geq 2\alpha (|X_i| - 1) - 1 \geq \alpha |X_i|$$

where again the final inequality is since $n$ is sufficiently large. Thus $\tilde{X}_i$ forms an $(\alpha, R')$-buffer for $H$.

Before we verify the remaining good $H$-partition properties, it is convenient to check $(G6)$. By assumption $(iii)$ of the lemma, at most $\frac{1}{2} \delta g^{\rho} |X_{i'}^{BL}|$ vertices of any given $X_{i'}^{BL}$ are image restricted. Thus for any $i$ such that $X_i \subseteq X_{i'}^{BL}$, at most

$$\frac{1}{2} \delta g^{\rho} |X_{i'}^{BL}| \leq \frac{1}{2} g^{\rho} (|X_i| + 1) \leq g^{\rho} |X_i|$$

vertices of $X_i$ are image restricted, as required for $(G6)$.

We proceed with properties $(H3)$–$(H8)$ by choosing buffer vertices $X_i^{buf}$ from the potential buffer vertices $\tilde{X}_i$ for each $i \in [r]$ sequentially. For a given $i \in [r]$, we first decide on the degree of the buffer and whether or not it is a clique buffer. We split the vertices of $\tilde{X}_i$ into $\Delta + 2$ subsets, one for each of the possible degrees in $\{0, \ldots, \Delta - 1\}$, one for vertices of degree $\Delta$ not in copies of $K_{\Delta + 1}$, and one for vertices in copies of $K_{\Delta + 1}$. We take a largest one $S_i$ of these sets, and will choose $X_i^{buf}$ within it. We do this greedily, one vertex at a time, always picking a vertex at distance at least five from previously chosen buffer vertices, and at distance at least three from any image restricted vertices. We now justify that it is possible to do this, that is, that we can pick the desired $4\mu |X_i|$ vertices in $S_i$ without running out of vertices in $S_i$.

Since all neighbours (respectively, second neighbours) in $H$ of vertices in $\tilde{X}_i$ are in clusters $V_j$ with $ij \in R'$ (respectively, $V_k$ with $ijk \in R'$), by $(G6)$, verified above, it follows that the number of image restricted vertices at distance 2 or less to a vertex of $\tilde{X}_i$ is at most $(\Delta^2 + \Delta R_i + 1) g_{\rho} |X_i|$. Each of these vertices has at most one neighbour or second neighbour in $\tilde{X}_i'$, since the distance between any two vertices of $X_i$ is at least ten. Similarly, if $x$ is in some $\tilde{X}_k$ and is at distance four or less from some vertex of $\tilde{X}_i$, then the path between $x$ and $\tilde{X}_i$ follows edges of $R'$, and there is only one vertex of $\tilde{X}_i$ which is at distance four or less from $x$. Thus the total number of vertices of $\tilde{X}_i$ which are at distance four or less from some vertex of $X_i^{buf}$ is at most

$$(1 + \Delta R_i + \Delta^2 + \Delta^3 + \mu \|X_i\|) \cdot 4\mu |X_i|.$$

Since $|S_i| \geq \frac{1}{\Delta^2 + 2} |\tilde{X}_i|$, it follows that at each step the number of vertices in $S_i \subseteq \tilde{X}_i$ we have to choose from is at least

$$\frac{1}{\Delta^2 + 2} \alpha |X_i| - 3 \Delta^2 + \Delta R_i + \Delta^3 + \mu \|X_i\|,$$

which is greater than $\frac{1}{\Delta^2 + 2} \alpha |X_i|$ by choice of $\alpha$ and $\mu$. Thus we can always choose the desired vertices of $X_i^{buf}$, giving $(H4)$–$(H7)$. In the event that $X_i^{buf}$ is a clique buffer, we have at the end at least $\frac{\Delta - 1}{\Delta^2 + 2} \alpha |X_i|$ vertices of $\tilde{X}_i$ remaining which are in copies of $K_{\Delta + 1}$ (by definition of $S_i$) that do not contain either vertices in $X_i^{buf}$ or image restricted vertices, giving $(H8)$.

Finally, neighbours of $X_i^{buf}$ in $X_j$ must be in parts $X_j$ such that $ij \in R'$. Since any two vertices in $X_j$ are at distance at least ten, each such vertex has at most one neighbour in $X_j$ and so at most $\Delta R_i - 4\mu |X_i|$ vertices of $X_i$ are in $N(X_i^{buf})$, completing $(H3)$. This establishes that we have a good $H$-partition.

We now construct a corresponding good $G$-partition. The construction is simple: for each $V_i^{BL} \in V^{BL}$ we choose, uniformly at random, an equitable partition into $\delta^{-1}$ parts, which parts we associate (arbitrarily) to subparts of $X_i^{BL}$ of corresponding size, obtaining $V$. Note that this is possible since $|X_{buf}| = |V_i^{BL}|$. We then, for each $i \in [r]$, choose uniformly at random a partition of $V_i$ into one set of size $(1 - 3\mu)|V_i|$.
and three of size \( |V_i| \), which form \( V_i^{\text{main}} \), \( V_i^q \), \( V_i^c \) and \( V_i^{\text{buf}} \) respectively. For each \( x \in V_i \), we let \( I_x = I_x^{BL} \cap V_i \).

By construction \( \mathcal{X} \) and \( \mathcal{V} \) are size-compatible, so what remains is to show that \( \mathcal{V} \) is with positive probability a good \( G \)-partition. In fact, we will show this holds asymptotically almost surely. Recall that we have already established \((G6)\), and by construction \( (G1) \) holds, so that the remaining conditions are \((G2)-(G5)\).

Observe that the sets \( V_i \), \( V_i^{\text{main}} \), \( V_i^q \), \( V_i^c \) and \( V_i^{\text{buf}} \) are all distributed as the uniform random set of that size within \( V_i^{BL} \), where \( V_i \subseteq V_i^{BL} \), though they are of course not independent. It follows that for any set \( U \subseteq V_i^{BL} \), the sizes of each of the sets \( V_i \cap U, V_i^{\text{main}} \cap U, V_i^q \cap U, V_i^c \cap U \) and \( V_i^{\text{buf}} \cap U \) are hypergeometrically distributed, each with (since \( n \) is sufficiently large) expectation at least \( \frac{\mu \delta}{2} |U| \). In particular, if \( |U| \geq 10^3 \varepsilon^{-2} \log n \), then the probability that any given one of these sizes is not within a \((1 \pm \frac{\varepsilon}{10})\)-factor of its expectation is by Theorem 2.1 at most \( 2 e^{-\varepsilon^2 |U|/300} \leq n^{-2} \).

We now give various sets \( U \) to which we will apply this observation, and verify that each is sufficiently large. For each \( i,j \in R' \) and each \( v \in V_j \), by construction of \( R' \) and since \((G, V^{BL})\) is super-regular on \( R^{BL}_v \), we have

\[
\deg_G(v; V_i^{BL}) \geq \deg_G(v; V_i^{BL}) \geq (d-\varepsilon) \max \{ p|V_i^{BL}|, \frac{1}{2} \deg_G(v; V_i^{BL}) \} \geq 10^3 \varepsilon^{-2} \log n ,
\]

where the final inequality holds by choice of \( p \) and \( C \). We can thus take \( U \) to be \( N^{BL}(v; V_i^{BL}) \) or \( N_G(v; V_i^{BL}) \).

For each \( x \in X_i \), where \( X_i \subseteq X_i^{BL} \), since \((I^{BL}, J)\) form a \((\frac{1}{2} \delta \varepsilon, 2\varepsilon, \Delta, \Delta_J)\)-restriction pair, we have

\[
|N_I^*(J_x; V_i^{BL})| \geq |I_x^{BL}| \geq (\varepsilon dp)|J_x^{BL}| \geq 10^3 \varepsilon^{-2} \log n ,
\]

where the final inequality comes from the assumption \( |J_x| \leq b \) and our choice of \( p \) and \( C \). We can thus also take \( U \) to be \( N_I^*(J_x^{BL}; V_i^{BL}) \) or \( I_x \).

In total, we have at most \( 5r \) randomly chosen sets in which we are interested, and at most \( 4n \) choices of \( U \) with which we intersect the random sets. Taking the union bound, we see that a.a.s. each of these intersections has size within a \((1 \pm \frac{\varepsilon}{10})\)-factor of its expectation. In particular, there exists a partition \( \mathcal{V} \) in which each of these intersections is within a \((1 \pm \frac{\varepsilon}{10})\)-factor of its expectation. We fix such a partition, and prove that it is the desired good \( G \)-partition.

We begin with \((G2)\). Suppose \( i, j \in R \). Then \( V_i \) and \( V_j \) come from two parts \( V_i^{BL} \) and \( V_j^{BL} \) of \( V^{BL} \) which form a \((\frac{1}{2} \delta \varepsilon, d, p)\)-regular pair, and we have \( |V_i| > \frac{1}{2} \delta |V_i^{BL}| \) and \( |V_j| > \frac{1}{2} \delta |V_j^{BL}| \). It follows from the definition that \( (V_i, V_j) \) is \((\varepsilon, d, p)\)-regular, so \((G, \mathcal{V})\) forms an \((\varepsilon, d, p)\)-regular \( R \)-partition. Given now \( i,j,k \in R' \), with \( V_j \subseteq V_k \) and \( V_k \subseteq V_i \), and any \( v \in V_i \), we have by construction of \( \mathcal{V} \) that \( |N_G(v; V_j)| \geq \frac{1}{2} \delta |N(v; V_i^{BL})| \) and \( |V_k| > \frac{1}{2} \delta |V_k^{BL}| \). Since the pair \((N_G(v; V_j), V_k^{BL})\) is \((\frac{1}{2} \delta \varepsilon, d, p)\)-regular in \( G \), it follows from the definition that the pair \((N_G(v; V_j), V_k)\) is \((\varepsilon, d, p)\)-regular in \( G \), so that we have two-sided inheritance on \( R' \). A similar argument shows that if also \( j,k \in R' \) and there is a triangle of \( H \) with one vertex in each of \( \tilde{X}_i, X - j \) and \( X_k \), then also \((N_G(v; V_j), N_G(v; V_k))\) is \((\varepsilon, d, p)\)-regular in \( G \), so that we have two-sided inheritance on \( R' \) for \( \mathcal{X} \).

Both of \((G3)\) and \((G4)\) follow directly from the construction of \( \mathcal{V} \) together with, respectively, the \((\frac{1}{2} \delta \varepsilon, d, p)\)-super-regularity of \((G, V^{BL})\) on \( R^{BL}_v \) and that \((I^{BL}, J)\) forms a \((\frac{1}{2} \delta \varepsilon, 2\varepsilon, \Delta, \Delta_J)\)-restriction pair. The same holds for everything except \((f)\) in the definition of a restriction pair for \((G5)\), since \((G6)\) was already established. Given an image restricted \( x \in X_i \subseteq X_i^{BL} \) and an edge \( xy \in E(H) \).
with \( Y \in X_j \subseteq X_j^{BL} \), the pair \( (N_T^y(J_y;V_j), N_T^y(J_y;V_j)) \) is \((\frac{1}{2} \delta \varepsilon, d, p)\)-regular by assumption. Since \( |N_T^y(J_y;V_j)| \geq \frac{1}{2} |N_T^y(J_y;V_j)| \), and similarly for \( y \), it follows from the definition that \( (N_T^y(J_y;V_j), N_T^y(J_y;V_j)) \) is \((\varepsilon, d, p)\)-regular, completing the verification of \((G5)\). \( \square \)

2.3.5. Underlying restrictions, candidates and good partial embeddings.
Suppose \( \psi \) is a partial embedding of \( H \) into \( G \)—that is, an injective graph homomorphism with domain \( \text{Dom}(\psi) \subseteq V(H) \).

We call a vertex \( x \in H \) embedded if \( x \in \text{Dom}(\psi) \), and otherwise unembedded. Of course this is with respect to \( \psi \), but this will always be clear from the context and we will not in future mention it. The set of embedded neighbours of \( x \) is

\[ \Pi(x) := \{ \psi(y) : xy \in E(H), y \in \text{Dom}(\psi) \} \]

and we define

\[ \pi(x) := |\Pi(x)| \text{ and } \pi^*(x) = |\Pi(x)| + |J_x|. \]

Thus, \( \pi^*(x) \) counts the number of already embedded neighbours of \( x \) by the partial embedding \( \psi \), including the ‘neighbours’ from \( J_x \) (if any).

Given an unembedded vertex \( x \in X_i \), we define the underlying restriction set, candidate set and available candidate set of \( x \) by

\[ U(x) := V_i \cap N_T^x(\Pi(x) \cup J_x), \]
\[ C(x) := I_x \cap N_T^x(\Pi(x)), \text{ and } \]
\[ A(x) := C(x) \setminus \text{Im}(\psi), \]

respectively. See Figure 1 for an example in which \( x \) and \( y \) are embedded and their common neighbour \( z \) is not.

\[ \text{Figure 1. Underlying restriction sets, candidate sets, and available candidate sets. The thick lines are edges of } G, \text{ and the wavy lines of } \Gamma. \]

Recall that \( G \) is a subgraph of \( \Gamma \) (being either a random or a bijumbled graph), which is typically far from being complete. The underlying set signifies thus the vertices that could in principle be used for embedding (if \( G = \Gamma \)), whereas the candidate sets denote the vertices that are possible in \( G \), and the available candidate sets only consider those vertices from the candidate sets which haven’t been previously taken as images by the vertices embedded earlier (in \( \text{Im}(\psi) \)). We also refer to vertices
from $C(x)$ as candidate vertices for $x$. Further, we define $U^\text{main}(x) := U(x) \cap V^\text{main}$, $U^q(x) := U(x) \cap V^q$, $U^c(x) := U(x) \cap V^c$ and $U^\text{buf}(x) := U(x) \cap V^\text{buf}$, and similarly the respective subsets of the candidate and available candidate sets of $x$, illustrated in Figure 2.

Finally we define the underlying restriction graph between $Y \subseteq X_i$ and $Z \subseteq V_i$ to be the bipartite graph with parts $Y$ and $Z$ and edges $yz$ whenever $z \in U(y)$. We define similarly the candidate graph between $Y \subseteq X_i$ and $Z \subseteq V_i$ to be the bipartite graph with parts $Y$ and $Z$ and edges $yz$ whenever $z \in C(y)$. If $x \in V(H)$ and $v \in C(x)$ we say $v$ is candidate for $x$.

We call $\psi$ a good partial embedding if the following conditions hold.

(GPE1) For each $x \in \text{Dom}(\psi)$ we have $\psi(x) \in I_x$.

(GPE2) For each unembedded $x \in X_i$ we have

$$|U(x)| = (p \pm \epsilon p)^{\pi(x)}|V_i|,$$

$$|U^\text{main}(x)| = (1 - 3\mu)(p \pm \epsilon p)^{\pi(x)}|V_i|,$$

and

$$|U^q(x)|, |U^c(x)|, |U^\text{buf}(x)| = \mu(p \pm \epsilon p)^{\pi(x)}|V_i|.$$

(GPE3) For each unembedded $x$ we have

$$|C(x)| \geq (1 - \epsilon')(dp - \epsilon' p)^{\pi(x)}|I_x|,$$

$$|C^\text{main}(x)| \geq (1 - \epsilon')(1 - 3\mu)(dp - \epsilon' p)^{\pi(x)}|I_x|,$$

and

$$|C^q(x)|, |C^c(x)|, |C^\text{buf}(x)| \geq (1 - \epsilon')\mu(dp - \epsilon' p)^{\pi(x)}|I_x|.$$

(GPE4) For each $xy \in E(H)$ with $x$ and $y$ unembedded, the pair $(U(x), U(y))$ is $(\epsilon\pi(x, \pi(y), d, p))$-regular in $G$.

The intention of this definition is that a good partial embedding is ‘locally’ good—if $x$ is unembedded, then almost all of $C(x)$ consists of vertices $v$ such that if $v$ is not in $\text{Im}(\psi)$, then $\psi \cup \{x \to v\}$ is a good partial embedding (This assertion is proved in Lemma 2.25).

It is important to observe that the trivial partial embedding, in which no vertices are embedded, is not automatically a good partial embedding: image restricted vertices might destroy any of (GPE2)–(GPE4). However, if we are provided with a
good $H$-partition and a corresponding good $G$-partition, then from the definitions the trivial partial embedding is a good partial embedding.

Finally, in much of the rest of this paper we will want to consider not just one partial embedding, but a sequence $\psi_0, \ldots$ of them. We will want to refer to sets $A_t^{\text{main}}(x), C_t^{\text{main}}(x)$ and so on, and quantities $\pi(x)$ and so on, with reference to each of these. We will do this by following the convention that a subscript $t$ attached to any of these means it is with reference to $\psi_t$. Thus we write $A_t^{\text{main}}(x)$ for the set $A^{\text{main}}(x)$ with respect to $\psi_t$, and similarly $C_t^{\text{main}}(x), \Pi_t(x), \pi_t(x), \Pi_t^*(x), U_t(x), C_t(x), A_t(x)$ and so on.

### 2.3.6. The general setup.

In this subsection we put forward the following notion of the General Setup. This notion encapsulates many of the definitions and constant choices of this whole section, which we will need as assumptions in many of the lemmas to come, and which we therefore wish to give a convenient name to.

**Definition 2.23 (General Setup).** When we say that we assume the General Setup we assert that constant choices satisfying the conditions of Section 2.3.1 have been made, that there exist $\kappa$-balanced size-compatible partitions $X$ and $V$ of $V(H)$ and $V(G)$ respectively whose parts are of size at least $\frac{n}{m^{1+\epsilon}}$, together with reduced graphs $R$ and $R'$ with $\Delta(R') \leq \Delta R'$, an image restriction pair $(I, J)$, an $(\alpha, R')$-buffer $X$, buffer vertices $X^{\text{buf}}$, and sets $V^{\text{main}}, V^q, V^c$ and $V^{\text{buf}}$, that the partition of $V(H)$ is a good $H$-partition, and that the partition of $G$ is a good $G$-partition.

We stress again that these conditions in particular imply directly from the definitions that the trivial partial embedding is a good partial embedding.
2.3.7. Bad vertices. As mentioned in the proof overview, we will embed vertices of $H$ into $G$ one at a time, choosing each vertex in a way that allows for the future continuation of the embedding. This ‘allowing for future continuation’ comes in two parts, ‘local’ conditions which broadly say that we embed a vertex of $H$ in a way which allows for the embedding of other vertices nearby in $H$, and ‘global’ conditions which say that the embedding of many (distant) vertices of $H$ do not tend to cause problems for any given vertex. We handle the latter probabilistically, and it is the main work of this paper. The purpose of this subsection is to define precisely what we require locally, by specifying which vertices of $G$ are ‘bad’ for $x \in V(H)$, i.e. not suitable for embedding $x$ to, and to show that these are always a small set of vertices (in comparison to the set $C(x)$). In fact, the definition of ‘badness’ also depends on a set $Q \subseteq V(H)$. In the RGA (randomised greedy algorithm, outlined in the proof overview, Section 1.4), $Q$ will be the current queue, while at other times $Q$ will be $V(H)$.

**Definition 2.24** (Bad vertices with respect to $\psi$ and $Q$; badness condition). Let $\psi$ be a good partial embedding and $Q \subseteq V(H)$ be a set of unembedded vertices. The vertex $v$ is called bad for $x$ with respect to $\psi$ and $Q$ if the extension $\psi \cup \{x \rightarrow v\}$ is not a good partial embedding or there is an unembedded neighbour $y$ of $x$ not in $Q$ such that

$$\deg_G(v; A^\text{main}(y)) < (d - \varepsilon')p|A^\text{main}(y)|. \quad (8)$$

When $\psi$ and $Q$ are clear from the context (as they always will be) we let $B(x)$ be the set of vertices in $C(x)$ which are bad for $x$ with respect to $\psi$ and $Q$. We will refer to (8) as the badness condition.

The following lemma provides control over the bad vertices with respect to a good partial embedding.

**Lemma 2.25.** We assume the General Setup. Let $\psi$ be a good partial embedding and let $Q \subseteq V(H)$ be a set of vertices such that for each unembedded $x \in V(H) \setminus Q$ we have

$$|A^\text{main}(x)| \geq \frac{1}{2} \mu(d - \varepsilon')p\pi^*(x)|V^\text{main}(x)|. \quad (9)$$

Given an unembedded $x \in V(H)$, let $D$ be such that the following holds for all unembedded $y, z \in V(H)$.

(i) $D \geq \pi^*(x) + 1$.

(ii) If $xy \in E(H)$ then $D \geq \pi^*(y) + 1$.

(iii) If $xy, yz \in E(H)$ then $D \geq \pi^*(y) + 2, \pi^*(z) + 1$.

(iv) If $xy, yz, zx \in E(H)$ then $D \geq \pi^*(x) + 2, \pi^*(y) + 2, \pi^*(z) + 2$.

Then the following hold.

(a) If all neighbours of $x$ are embedded then no vertex in $C(x) \setminus \text{Im}(\psi)$ is bad for $x$ with respect to $\psi$ and $Q$.

(b) Suppose that $\Gamma$ satisfies $\text{NS}(\varepsilon, r_1, D)$ and $\text{RI}(\varepsilon, (\varepsilon_{a,b}), \varepsilon', d, r_1, D)$. Then at most $20\Delta^2 \varepsilon'p\pi^*(x)|V(x)|$ vertices of $C(x)$ are bad for $x$ with respect to $\psi$ and $Q$.

**Proof.** We require

$$\mu \leq \frac{1}{4}, \quad \varepsilon' \leq \frac{\mu d 10000k4\Delta}{\epsilon^2} \quad \text{and} \quad \varepsilon \leq \frac{d \varepsilon'}{\kappa}.$$  

For (a), observe that all the conditions for a vertex $v$ to be in $B(x)$ are trivially false.

For part (b), we need to consider all of the possible reasons $v$ could be bad for $x$ with respect to $\psi$ and $Q$. It could be that there is some unembedded neighbour $y \in V(H) \setminus Q$ of $x$ such that badness condition (8) holds. The pair $(U(x), U(y))$ is
\( (\varepsilon', d, p) \)-regular in \( G \) by \((GPE4)\). By assumption and because \((GPE2)\) holds for \( y \) with respect to \( \psi \) we have

\[
|A^{\text{main}}(y)| \geq \frac{1}{2} \mu (d - \varepsilon') \pi^*(y) p \pi^*(y) |V^{\text{main}}(y)| = \frac{1}{2} \mu (1 - 3\mu)(d - \varepsilon') \pi^*(y) p \pi^*(y) |V(y)|
\]

\[
\geq \frac{1}{2} \mu (1 - 3\mu)(d - \varepsilon') \pi^*(y) p \pi^*(y) \frac{|U(y)|}{(1 + \varepsilon) \pi^*(y) p \pi^*(y)} \geq \varepsilon' |U(y)|.
\]

So badness condition \((8)\) holds for at most \(\varepsilon' |U(x)| \leq 2\varepsilon' p \pi^*(x) |V(x)| \) vertices of \( C(x) \), where we use \((GPE2)\) and \((1 - \varepsilon)^3 \leq 2\). Since \( x \) has at most \( \Delta \) unembeded neighbours, in total there are at most \(2\Delta \varepsilon' p \pi^*(x) |V(x)| \) vertices of \( C(x) \) such that the badness condition \((8)\) holds for some unembedded neighbour of \( x \).

It could be that \( \psi \cup \{x \to v\} \) is not a good partial embedding. Since \( v \in C(x) \subseteq I_x \) (\(GPE1\)) cannot fail. We next consider \((GPE3)\). Since \( \psi \) is a good partial embedding, \((GPE3)\) cannot fail for any vertex \( y \) which is not a neighbour of \( x \) (because the candidate sets of these vertices do not change). So let \( y \) be any neighbour of \( x \) in \( H \). Recall that \( (U(x), U(y)) \) is \((\varepsilon', d, p)\)-regular in \( G \) by \((GPE4)\). We have

\[
|C(y)| \geq (1 - \varepsilon')(dp - \varepsilon'p) \pi^*(y) |I_y| \geq (1 - \varepsilon')(dp - \varepsilon'p) \pi^*(y) \pi(y) |V(y)|
\]

\[
\geq (1 - \varepsilon')(dp - \varepsilon'p) \Delta (1 + \varepsilon)^{-\Delta} \pi(y) \geq \varepsilon' |U(y)|.
\]

Since \( C(x) \subseteq U(x) \), there are at most \(\varepsilon'|U(x)| \leq 2\varepsilon' p \pi^*(x) |V_i| \) vertices \( v \in C(x) \) such that \( |C(y) \cap N_G(v)| < (d - \varepsilon') p |C(y)| \). Since \( (d - \varepsilon') p |C(y)| \geq (1 - \varepsilon')(dp - \varepsilon'p) \pi^*(y) |I(y)| \), there are at most \(2\varepsilon' p \pi^*(x) |V(x)| \) vertices \( v \in C(x) \) such that the first condition in \((GPE3)\) fails for \( y \) with respect to \( \psi \cup \{x \to v\} \). We can analogously argue for each of the other four conditions. Since \( x \) has at most \( \Delta \) unembedded neighbours, we conclude that for all but at most \(10\Delta \varepsilon' p \pi^*(x) |V_i| \) vertices \( v \in C(x) \) we have that \((GPE3)\) is satisfied with respect to \( \psi \cup \{x \to v\} \).

Now we turn to \((GPE2)\). Again, it suffices to consider neighbours \( y \) of \( x \). As \((GPE2)\) holds with respect to \( \psi \), we see that \( |U(y)| \geq (p - \varepsilon p) \pi^*(y) |V(y)| \geq \varepsilon p^{D-1}/r_1 \) since \( |V(y)| \geq n/((kr_1)) \) and by assumption \((ii)\). It follows that we can use \(\text{NS}(\varepsilon, r_1, D)\) and assumption \((i)\) to conclude that all but at most \(\varepsilon p^{D-1}/r_1 \) \( (d - \varepsilon' p) \pi^*(x) |V(x)| \) vertices \( v \in C(x) \) are such that \( |U(y) \cap N_T(v)| = (1 + \varepsilon) p |U(y)| = (p \pm \varepsilon p) \pi^*(y) + |V(x)| \), again using \((GPE2)\) for \( y \) with respect to \( \psi \). Again, we can argue analogously for the other four conditions, and so in total for all but at most \(5\Delta \varepsilon' p \pi^*(x) |V_i| \) vertices \( v \in C(x) \) we have that \((GPE2)\) is satisfied with respect to \( \psi \cup \{x \to v\} \).

It remains to consider \((GPE4)\). Again, we only need to consider edges \( yz \in E(H) \) between unembedded vertices such that \( y \in N_H(x) \) or \( z \in N_H(x) \), or both. Consider first the case that only \( y \in N_H(x) \). The case that only \( z \in N_H(x) \) follows analogously. Since \((GPE2)\) is satisfied with respect to \( \psi \) we have \( |U(y)| \geq (p - \varepsilon p) \pi^*(y) |V(y)| \) and \( |U(z)| \geq (p - \varepsilon p) \pi^*(z) |V(z)| \), which together with assumption \((iii)\) implies \( |U(y)| \geq (p - \varepsilon p)^{D-2}/(kr_1) \) \( \geq \varepsilon' p^{D-2}/r_1^2 \) and \( |U(z)| \geq \varepsilon' p^{D-1}/r_1^2 \). Hence, since \( \Gamma \) satisfies \(\text{RI}(\varepsilon, \varepsilon' d, r_1, D)\) and the pair \((U(y), U(z))\) is \((\varepsilon \pi^*(y), \varepsilon \pi^*(z), d, p)\)-regular in \( G \) because \((GPE4)\) is satisfied with respect to \( \psi \), we get that there are at most \(\varepsilon p^{D-1}/r_1^2 \) \( \leq \varepsilon' \pi^*(x) |V(x)| \) vertices \( v \in C(x) \) such that \( |N_T(v) \cap U(y), U(z)| \) is \((\varepsilon \pi^*(y) + 1, \varepsilon \pi^*(z) + 1, d, p)\)-regular in \( G \). Analogously, in the second case that \( y, z \in N_H(x) \) we can use assumption \((iv)\) to conclude that there are at most \(\varepsilon p^{D-2}/r_1^2 \) \( \leq \varepsilon' \pi^*(x) |V_i| \) vertices \( v \in C(x) \) such that \( |N_T(v) \cap U(y), U(z)| \) is \((\varepsilon \pi^*(y) + 1, \varepsilon \pi^*(z) + 1 + 1, d, p)\)-regular in \( G \). So, in total there are at most \(\Delta \varepsilon' p \pi^*(x) |V(x)| \) vertices \( v \in C(x) \) such that \((GPE4)\) is not satisfied with respect to \( \psi \cup \{x \to v\} \).
Summing up, there are at most
\[2 \Delta \varepsilon' p^\pi(x) |V(x)| + 10 \Delta \varepsilon' p^\pi(x) |V(x)| + 5 \Delta \varepsilon' p^\pi(x) |V(x)| + \Delta^2 \varepsilon' p^\pi(x) |V(x)| \leq 20 \Delta^2 \varepsilon' p^\pi(x) |V(x)|\]
vertices \(v \in C(x)\) such that \(v\) is bad for \(x\) with respect to \(\psi\) and \(Q\).

\[\square\]

2.4. The behaviour of random greedy algorithms

In this section we give four lemmas which show good properties of the various random greedy algorithms we use in this paper. Roughly speaking, each of our RGAs (of which we give three, one for each of Lemmas 1.21, 1.23 and 1.25) has the property that we embed vertices sequentially, at each time choosing an image of the next vertex uniformly at random in a not-too-small subset of its candidate vertices, and this, together with the General Setup, is all we need to know about our RGAs for the purposes of this section. Nevertheless, the reader who wishes to see a concrete example of an RGA to which the following lemmas apply should look at Algorithm 1 (Section 3.2), which is the simplest of our RGAs.

The first lemma constructs the vertex ordering which we shall use in the proofs of Lemmas 1.21 and 1.25. In Lemma 1.23 the vertex ordering is supplied. The point of the ordering we use is that we need to analyse the way neighbours of (non-clique) buffer vertices are embedded during the RGA. Putting them first in the order, with each neighbourhood coming as an interval in the order, makes this possible. Even then, for Lemma 1.21 we need to insist that in degree-\(\Delta\) buffers the final two neighbours of a buffer vertex are not adjacent, thus their embeddings do not affect each other much. Ultimately, the fact that we cannot expect these nice conditions from the order supplied to Lemma 1.23 is the reason why we cannot obtain stronger bounds on \(p\) in that lemma. Note that in the following lemma we do not need to assume the General Setup, but we nevertheless use names for the parts of \(H\) corresponding to those from the General Setup as we will use the lemma.

**Lemma 2.26 (Vertex order for the RGA).** Let \(H\) be a graph with \(\Delta(H) \leq \Delta\) and \(V(H) = X^\text{buf} \cup X^\text{main} \cup X^c\) such that \(N(X^\text{buf}) \subseteq X^\text{main}\) and each pair of vertices in \(X^\text{buf}\) has distance at least ten. Then there exists an ordering \(\tau\) of \(X^\text{main}\) with the following properties.

(a) For all \(x \in X^\text{main} \setminus N(X^\text{buf})\) and \(y \in N(X^\text{buf})\) we have \(\tau(x) > \tau(y)\).

(b) For all \(x \in X^\text{buf}\) we can enumerate \(N(x)\) as \(y_1, \ldots, y_b\) such that \(\tau(y_{j+1}) = \tau(y_j) + 1\) for all \(j \in [b-1]\). Moreover, if \(\deg_H(x) = \Delta\) and \(x\) is not in a copy of \(K_{\Delta+1}\) then \(y_{\Delta+1} y_{\Delta} \not\in E(H)\).

(c) The neighbours of non-clique buffer vertices come before the neighbours of the clique-buffer vertices.

**Proof.** We separate the vertices in \(X^\text{buf}\) into two classes, those not in copies of \(K_{\Delta+1}\) and those in copies of \(K_{\Delta+1}\). We take any enumeration \(X^\text{buf} = \{x_1, \ldots, x_{|X^\text{buf}|}\}\) with the vertices not in copies of \(K_{\Delta+1}\) coming first. We now construct the ordering \(\tau\) as follows. We start with the empty ordering. For each \(i\) in succession such that \(x_i \in X^\text{buf}\) is not in a copy of \(K_{\Delta+1}\), we append to \(\tau\) the vertices \(N_H(x_i)\) in some order such that if \(\deg_H(x_i) = \Delta\) then the last two vertices in \(\tau\) of \(N_H(x_i)\) are not adjacent. Note that if \(\deg_H(x) = \Delta\) then there is a pair of non-adjacent vertices in \(N_H(x)\) because \(x_i\) is not in a copy of \(K_{\Delta+1}\). Next, for each \(i\) in succession such that \(x_i \in X^\text{buf}\) is in a copy of \(K_{\Delta+1}\) we append to \(\tau\) the vertices \(N_H(x_i)\) in an arbitrary order. Finally, we append to \(\tau\) any remaining vertices of \(X^\text{main}\) in an arbitrary order. \(\square\)
The next lemma states that provided we embed vertices uniformly at random into not-too-small sets, the candidate sets of unembedded vertices are likely to be ‘distributed uniformly’. In this lemma, as will be the case for much of the rest of the paper, we have a sequence $\psi_0, \ldots$ of good partial embeddings, and we want to talk about the various sets and quantities defined in Section 2.3.5, such as candidate sets, with respect to each of these good partial embeddings. As mentioned in Section 2.3.5, we will follow the convention that (for example) $C_t(x)$ is the candidate set of $x$ with respect to $\psi_t$.

**Lemma 2.27 (Uniform distribution of candidate sets).** We assume the General Setup. Suppose that for some $T$ we have a sequence $\psi_0, \psi_1, \ldots, \psi_T$ of good partial embeddings, where $\psi_0$ is the trivial partial embedding with empty domain and each $\psi_t$ is obtained from $\psi_{t-1}$ by embedding some vertex $x$ in $V(H) \setminus \text{Dom}(\psi_{t-1})$ to a uniform random vertex from a subset of $C_{t-1}(x)$ of size at least $\frac{\mu(\delta^a/20\Delta)}{\delta^a}$. For every set $W \subseteq V_i$ of size at least $\theta|\psi_i|$, the number of vertices $x \in X_i \setminus X^*_i$ (i.e. vertices which are not image restricted) such that there exists $t = t(x)$ when $x$ is unembedded and we have

$$|C_t(x) \cap W| < (dp - \varepsilon')^{\pi^*_t(x)}|W|$$

is at most $\theta|\psi_i|$.

This lemma corresponds to the ‘Main Lemma’ of [46], and its proof is similar in spirit. The proof of this lemma exploits the fact that when a vertex $y$ is embedded, condition (9) might become true for at most one vertex from $X_i \setminus X^*_i$ (since by (H2) the vertices in $X_i$ have distance at least 10). Moreover, by the condition of always embedding into not too small subsets and by the regularity property (GPE4), the vertex $y$ makes an unlucky choice with a very small probability, so that many such choices are extremely unlikely.

**Proof of Lemma 2.27.** We require

$$\varepsilon' < \min \left( 4^{-\Delta}d^2 \theta, \frac{\mu(\delta^a/20\Delta)}{\delta^a} 2^{-4/\theta} \right) \quad \text{and} \quad 2^{-n/(\kappa r + 1)} < 1.$$

Fix a set $W \subseteq V_i$ of size at least $\theta|\psi_i|$, and a set $X' \subseteq X_i \setminus X^*_i$ of size $\theta|\psi_i|$. We aim to show that the probability of the following event is at most $2^{-n/\kappa r + 1}$. For each $x \in X'$ there is a $t = t(x)$ such that $x$ satisfies (9). The desired result then follows by taking the union bound over all $x$, subsets $W$ of $V_i$, and $X'$ of $X_i$.

If $x$ is any vertex of $X'$, then we have $C_0(x) = V_i$ since $x$ is not image restricted, thus $|C_0(x) \cap W| = |W|$ and hence (9) is false for $x$. If there is a $t(x)$ such that $x$ satisfies (9), then we can fix $t$ to be the smallest integer such that (9) is true for $t + 1$ and $x$. Since the candidate set of $x$ changes only when a neighbour of $x$ is embedded, it follows that the vertex $y$ that is embedded to create $x_{t+1}$ from $\psi_t$ is a neighbour of $x$ in $H$ and thus $\pi^*_{t+1}(x) = \pi^*_t(x) + 1$. Moreover, since equation (9) becomes true for $x$, the vertex $y$ is embedded to a vertex $w$ such that

$$\deg_{G_t}(w; C_t(x) \cap W) < ((d - \varepsilon')p)|C_t(x) \cap W|,$$

as otherwise we would still have

$$|C_{t+1}(x) \cap W| \geq ((d - \varepsilon')p)|C_t(x) \cap W| \geq ((d - \varepsilon')p)^{\pi^*_t(x)}|W|.$$

Since $\psi_t$ is a good partial embedding, by (GPE4) the pair $(U_t(x), U_t(y))$ is an $(\varepsilon', d, p)$-regular pair in $G$, and by (GPE2) we have $|U_t(x)| = (p \pm \varepsilon')^{\pi^*_t(x)}|V_i|$, and $|U_t(y)| = (p \pm \varepsilon')^{\pi^*_t(y)}|V(y)|$. Since (9) is false for $x$ at time $t$, we have
\[ |C_t(x) \cap W| \geq \left( (d - \varepsilon') p \right)^{\pi_t(x)} |W|, \]

and so

\[ |C_t(x) \cap W| \geq \left( \frac{d - \varepsilon'}{1 + \varepsilon'} \right)^{\pi_t(x)} g[U_t(x)] \geq \varepsilon'|U_t(x)|, \]

by the requirements on the constants at the beginning of the proof. We conclude by \((\varepsilon', d, p)\)-regularity of \((U_t(x), U_t(y))\) that at most \(\varepsilon'|U_t(y)| \leq 2\varepsilon'p^{\pi_t(y)}|V(y)|\) vertices \(w\) of \(U_t(y)\) satisfy (10).

Since \(\psi_{t+1}\) is created by embedding \(y\) uniformly at random into a subset of \(C_t(y) \subseteq U_t(y)\) of size at least \(\frac{1}{10\mu\zeta(dp)}\pi_t^2(y)|V(y)|\), the probability of embedding \(y\) to a vertex \(w\) satisfying (10), conditioning on any history, is at most

\[ \frac{2\varepsilon'p^{\pi_t(y)}|V(y)|}{10\mu\zeta(dp)} \pi_t(y)|V(y)| \leq 20\varepsilon'\mu^{-1}\zeta^{-1}d^{-\Delta}. \]

(11)

Next we argue that the probability that for each \(x \in X'\) there is a first time \(t = t(x)\) such that \(x\) satisfies (9) is at most

\[ (20\Delta\varepsilon'\mu^{-1}\zeta^{-1}d^{-\Delta})^{6|X'|} \leq 2^{-4|X'|}. \]

Let us denote this event by \(\mathcal{E}_{X'}\). Observe that \(\mathcal{E}_{X'}\) is split into \(\Delta^{|X'|}\) events (since \(\Delta(H) \leq \Delta\)) by specifying for each \(x \in X'\) a neighbour \(y_x\) of \(x\) whose embedding occurs at time \(t(x)\). Let \((y_x)_{x \in X'}\) be any such assignment, and \(\mathcal{E}_{X',(y_x)}\) be the corresponding event. We aim to bound \(\mathbb{P}(\mathcal{E}_{X',(y_x)})\).

By (H2) the vertices of \(X_i\) are at distance at least ten in \(H\), so the vertices \((y_x)_{x \in X'}\) are distinct. The corresponding conditional probabilities thus multiply, and we have

\[ \mathbb{P}(\mathcal{E}_{X',(y_x)}) \leq (20\varepsilon'\mu^{-1}\zeta^{-1}d^{-\Delta})^{6|X'|} \]

by (11). Applying the union bound over the events \(\mathcal{E}_{X',(y_x)}\) we conclude

\[ \mathbb{P}(\mathcal{E}_{X'}) \leq (20\Delta\varepsilon'\mu^{-1}\zeta^{-1}d^{-\Delta})^{|X'|} \leq 2^{-4|X'|}. \]

where the final inequality is by choice of \(\varepsilon'\).

Taking the union bound over the at most \(2^{|V_i|} = 2^{|X_i|}\) choices of \(W\) in \(V_i\) and \(2^{|X_i|}\) choices of \(X'\) in \(X_i\), we see that the probability that, for any fixed \(i\), there exist subsets \(W\) of \(V_i\) and \(X'\) of \(X_i\), of sizes at least \(g[V_i]\) and \(g[X_i]\) respectively, such that each vertex \(x\) of \(X'\) satisfies (9) at some time \(t\), is at most \(2^{|X_i|}2^{-4|X_i|} = 2^{-2|X_i|}\).

Now we have \(|X_i| \geq n/(\kappa r_1)\), and \(n \geq n_0\) where \(n_0\) is chosen large enough such that \(2^{-n_0/(\kappa r_1)} r_1 \leq 1\). Thus taking the union bound over the at most \(r_1\) choices of \(i\), we conclude that the probability that there exists \(i\), and a subset \(W\) of \(V_i\) such that there are \(g[X_i]\) vertices \(x\) of \(X_i \setminus X_i^*\) each of which satisfies (9) at some time \(t\), is at most \(2^{-n/(\kappa r_1)} r_1\) as desired. \(\square\)

The next lemma shows that, again provided we embed vertices uniformly into not-too-small sets, we do not tend to cover vertex neighbourhoods in \(G\) disproportionately fast. Specifically, if \(ij \in E(R')\) then, by (G3), each \(v \in V_i\) has a large \(G\)-neighbourhood in \(V_j\). At some time \(T\) when only a small fraction of each part of \(H\) has been embedded, it is very likely that less than half of this \(G\)-neighbourhood is in the image \(\text{Im}(\psi_T)\) of the current partial embedding. One should think of this as: early on in the embedding process, the minimum degree conditions (G3) provided by super-regularity are preserved.

The idea behind Lemma 2.28 is as follows. As discussed in the proof overview (Section 1.4), we will need our RGAs to guarantee that each \(v \in V_i\) is a candidate for many vertices \(x \in X_i^{\text{out}}\) in order to complete the embedding. This means we need it to be not too unlikely that \(N_H(x)\) is embedded to \(N_G(v)\) for any given \(x \in X_i^{\text{out}}\), and as a first step to showing this it is necessary to show that we have not covered \(N_G(v)\) with embedded vertices before we get around to embedding \(N_H(x)\).
In Lemmas 1.21 and 1.25, we prove this by embedding $N(X^{bf})$ first and applying Lemma 2.28. In the proof of Lemma 1.25 we set $B = \Delta + 1$ and will show that this is enough for Lemma 2.28 to handle all of $N(X^{bf})$. In the proof of Lemma 1.21, by contrast, we set $B = \Delta$, which turns out to be good enough to handle buffers of degree up to and including $\Delta$ which are not clique buffers. This is one of the reasons why we have to handle clique buffers differently.

**Lemma 2.28** (Preservation of super-regularity). We assume the General Setup. Suppose that $\Gamma$ has NS$(\varepsilon, r_1, B)$. Suppose that for some $T$ we have $\psi_0, \psi_1, \ldots, \psi_T$ a sequence of good partial embeddings, where $\psi_0$ is the trivial partial embedding with empty domain and each $\psi_t$ is obtained from $\psi_{t-1}$ by embedding one vertex $x_t \in V(H)$ to a uniform random vertex from a subset of $C_{t-1}(x_t)$ of size at least $\frac{1}{p_0}(dp)^{\Delta - t - 1}(x_t)|V(x_t)|$. Suppose furthermore that for each $t$ the vertex $x_t$ has at most $B - 2$ neighbours in $\text{Dom}(\psi_{t-1})$, and that for each $i \in [r]$ we have $|\text{Dom}(\psi_T) \cap X_i| \leq \frac{8\mu_2 \Delta R}{\text{deg}_G(v)}|X_i|$.

Then with probability at least $1 - \exp(-\varepsilon pm/r_1)$, for each vertex $v \in V_i$ and $j$ such that $ij \in E(R')$ we have $|N_G(v; V^\text{main}_j) \setminus \text{Im}(\psi_T)| \geq \frac{1}{2} \text{deg}_G(v; V_j)$.

The proof of Lemma 2.28 involves estimating the probability of embedding $x_t$ to $N_G(v)$, conditioned on $\psi_{t-1}$. We show that either this probability is small, or that a previously embedded neighbour of $x_t$ was embedded ‘badly’, which is guaranteed to be a low-probability event. In either case, for $x_t$ to be embedded to $N_G(v)$ an unlikely event must occur. For $N_G(v)$ to be substantially filled up, many of these events, which are sequentially dependent, have to occur. Lemma 2.2 shows this is unlikely enough to take a union bound over all choices of $v$.

**Proof.** We require

$$\mu < \frac{d^2}{1000n \Delta R}, \quad \varepsilon < \frac{\mu d}{1000} \quad \text{and} \quad r_1 \exp(-\varepsilon pm/r_1) < 1.$$ 

Fix $ij \in E(R')$ and a vertex $v \in V_i$. We first estimate the conditional probability that $x_t \in X_j \cap \text{Dom}(\psi_T)$ is embedded to $N_G(v) \cap V^\text{main}(x_t)$, given the history. Because $x_t$ is embedded uniformly at random to a subset of $C_{t-1}(x_t)$ of size at least $\frac{1}{p_0}(dp)^{\Delta - t - 1}(x_t)|V_j|$, this probability is at most

$$\frac{|U^\text{main}(x_t) \cap N_G(v)|}{\frac{1}{p_0}(dp)^{\Delta - t - 1}(x_t)|V_j|}.$$ 

We are therefore interested in estimating the numerator. We separate two cases: when $|U^\text{main}_{t-1}(x_t) \cap N_G(v)| \leq (p + \varepsilon p)^{\Delta - t - 1}(x_t)\text{deg}_G(v; V_j)$, and otherwise. In the former case (12) is at most $20d^{-\Delta \text{deg}_G(v; V_j)}|V_j|$. By (G3) we have $\text{deg}_G(v; V_j) \geq (1 - 3\mu)(d - \varepsilon)p|V_j|$, so we have

$$20d^{-\Delta \text{deg}_G(v; V_j)}|V_j| \geq 10d^{-\Delta + 1}p.$$ 

In the latter case, i.e. $|U^\text{main}_{t-1}(x_t) \cap N_G(v)| > (p + \varepsilon p)^{\Delta - t - 1}(x_t)\text{deg}_G(v; V_j)$ but (with $\pi^t(x_t) = \pi^{t-1}(x_t) + 1$)

$$\text{deg}_G(v; y_t; U^\text{main}_{t-1}(x_t) \cap N_G(v)) > (p + \varepsilon p)^{\Delta - t - 1}(x_t) + 1 \text{deg}_G(v; V_j).$$ 

Let $W$ be a superset of $U^\text{main}_{t-1}(x_t) \cap N_G(v)$ of size $(p + \varepsilon p)^{\Delta - t - 1}(x_t)\text{deg}_G(v; V_j) > (d - \varepsilon)p^{B-1}|V_j|$, where the inequality uses the fact that $\pi^t_{t-1}(x_t) \leq B - 2$ and (G3). By property NS$(\varepsilon, r_1, B)$, which $\Gamma$ satisfies, we see that there are at most $\frac{p^{B-1}n}{r_1}$
vertices \( Z \) of \( \Gamma \) which have this many neighbours in \( W \) and thus satisfy (14). Now \( \pi_{t-1}(y) \leq B - 2 \) by the conditions of the lemma. We conclude that the probability of embedding \( y \) to a vertex of \( Z \), conditioning on the history, is at most

\[
\frac{\exp(r_{t-1}/n)}{\prod_{j} (d_j)^{n-|V(y)|}} \leq 10 \varepsilon d^{-\Delta} \kappa p < 10d^{-\Delta + 1} p,
\]

where we use the fact \( |V(y)| \geq n/(\kappa r_1) \) and the choice of \( \varepsilon \).

We define a sequence of Bernoulli random variables \( Y_1, \ldots, Y_T \) as follows. Given the embedding \( \psi_{t-1} \), if \( x_t \) is a neighbour of a vertex in \( X_t \cap \text{Dom}(\psi_T) \) none of whose previous neighbours were badly embedded, and \( x_t \) is badly embedded, we set \( Y_t = 1 \). If \( x_t \) is in \( X_t \), none of its previous neighbours were badly embedded, and \( x_t \) is embedded to \( N_G(v; V_j) \), we set \( Y_t = 1 \). Otherwise we set \( Y_t = 0 \). By assumption, the total number of \( Y_t \) which are not deterministically zero (that is, are in \( X_t \cap \text{Dom}(\psi_T) \) or are neighbours of such a vertex) is at most \( 8 \mu \kappa \Delta_{R'}(\Delta + 1)|X_t| \). Observe furthermore that for all of the \( Y_t \) which are not deterministically zero, we have just shown that, in view of (13) and (15), \( Y_t \) is one with probability at most

\[
20d^{-\Delta} \frac{\deg_G(v; V^\text{main}_{[j]})}{|V_j|} \quad \text{conditioning on the history up to, but not including, embedding } x_t.
\]

This history determines \( Y_{t-1} \), so that the \( Y_1, \ldots, Y_T \) are sequentially dependent. It follows that we can apply Lemma 2.2, with

\[
x = 8 \mu \kappa \Delta_{R'}(\Delta + 1)|X_t| \cdot 20d^{-\Delta} \frac{\deg_G(v; V_j)}{|V_j|} = 160 \mu \kappa d^{-\Delta} \Delta_{R'}(\Delta + 1) \deg_G(v; V_j)
\]

and \( \delta = 1 \), to show that

\[
P(Y_1 + \cdots + Y_T \geq 2x) \leq 2 \exp \left(-\frac{2x}{3} \right) \leq \exp \left(-\frac{2p}{r_1} \right),
\]

where the final inequality is by choice of \( \varepsilon \) and since \( \deg_G(v; V_j) \geq (1 - 3\mu) (d - \varepsilon)p|V_j| \) by (G3). By choice of \( \mu \), we conclude that with probability at most \( \exp(-2\varepsilon pn/r_1) \) we have \( Y_1 + \cdots + Y_T \geq \frac{1}{2} \deg_G(v; V_j) \). Now observe that for any vertex to be embedded to \( N_G(v; V^\text{main}_{[j]} \) one of these variables \( Y_1, \ldots, Y_T \) must be one, and since vertices of \( X_t \) are at distance at least ten in \( H \) (by (H2)), no \( Y_t \) can be responsible for two different vertices of \( X_t \) being embedded to \( N_G(v; V^\text{main}_{[j]} \). Thus \( Y_1 + \cdots + Y_T \) is an upper bound for \( \text{Im}(\psi_T) \cap N_G(v; V^\text{main}_{[j]} \). Taking a union bound over the choices of \( j \) and of \( v \), and using (G3), we see that with probability at least

\[
1 - r_1 n \exp(-2\varepsilon pn/r_1) > 1 - \exp(-\varepsilon pn/r_1)
\]

the statement of the lemma holds. \( \Box \)

Our final lemma in this section complements the above lemma, showing that provided \( v \in V_i \) does not have \( N_G(v) \) more than half covered by \( \text{Im}(\psi) \) early in the embedding, it is reasonably likely that \( N_H(x) \) is embedded to \( N_G(v) \) for any given \( x \in X_i^\text{buf} \). As with the previous lemma, it contains a parameter \( B \) which will be either set to \( \Delta \) (in the proof of Lemma 1.21) or to \( \Delta + 1 \) (in the proof of Lemma 1.25), and again in the former case the consequence is that we cannot use it to deal with clique buffers (that is, when \( N_H(x) \) is a copy of \( K_\Delta \)).

**Lemma 2.29** (Probable buffer embedding). We assume the General Setup. Suppose that \( B \in \{\Delta, \Delta + 1\} \). Suppose that \( \Gamma \) has \( N_S(\varepsilon, r_1, B) \) and \( R_1(\varepsilon, \varepsilon, r_1, B) \). Given \( v \in V_i \) and \( x \in X_i^\text{buf} \), suppose that \( \psi_0 \) is a good partial embedding in which no vertex at distance two or less from \( x \) is embedded, and suppose \( Q_0 \subseteq X^\text{main} \). Suppose further that \( |N_G(v; V^\text{main}_{[j]}) \cap \text{Im}(\psi_0)| \geq \frac{1}{2} \deg_G(v; V_j) \) for each \( j \) such that \( ij \in E' \). Let \( y_1, \ldots, y_k \) be the neighbours of \( x \), and suppose that if \( b = B \) then \( y_{k-1}y_k \) is not an edge of \( H \).

Suppose that \( \psi_1, \ldots, \psi_{k-1} \) are good partial embeddings and \( Q_1, \ldots, Q_k \) are subsets of \( X^\text{main} \). We write \( \alpha_i(y) \) for \( A(y) \) with respect to \( \psi_i \), and so on. We write \( \beta_i(y) \) for the set of bad vertices for \( y \) with respect to \( Q_i \). Suppose that \( \psi_i \) is obtained from \( \psi_{i-1} \)
by embedding \( y_t \) uniformly at random into \( A_{t-1}^{\text{main}}(y_t) \setminus B_{t-1}(y_t) \) for each \( t = 1, \ldots, b \), and suppose that for each \( t = 1, \ldots, b \) we have
\[
Q_t = Q_{t-1} \cup \{ z \in X^{\text{main}} \setminus \text{Dom}(\psi_t) : |A_t^{\text{main}}(z)| < \frac{\mu(d - \varepsilon')\pi_t(z)p\pi_t(z)|V^{\text{main}}(z)|}{1000\kappa\Delta^2} \}.
\]

Then with probability at least \( (d^3p/10)^b \), we have \( \psi_0(\mathcal{N}_G(x)) \subseteq \mathcal{N}_G(v) \).

The proof of this lemma is quite long, but much of it is ‘bookkeeping’ in the style of Lemma 2.25. Briefly, the idea is as follows. In \( \psi_0 \), no neighbour of any \( y_t \) is embedded and hence each \( y_t \) has candidate set \( C_0(y_t) = V(y_t) \). We know that \( \mathcal{N}_G(v; V(y_t)) \) is not covered by \( \text{Im}(\psi_0) \), so if we choose a uniform random vertex of \( A_0(y_t) \) the probability of choosing a member of \( \mathcal{N}_G(v) \) is at least \( dp/4 \). However, the random greedy algorithm does not choose a random vertex of \( A_0(y_t) \), but rather of \( A_0(y_t) \setminus B_0(y_t) \). Thus we have to show that \( B_0(y_t) \) does not cover too much of \( \mathcal{N}_G(v) \), we can do using properties (G.2) and (G.3) of the partition of \( G \), and \( NS(\varepsilon, r_1, B) \) and \( R(I(\varepsilon, (\varepsilon_{a,b}), \varepsilon', d, r_1, B) \) which we assume of \( \Gamma \), in much the same way as the proof of Lemma 2.25. In addition we have to show that some extra properties are preserved which allow us to analyse the embedding of \( y_2, \ldots, y_b \), which we can show is likely in a similar way. We will refer to a good partial embedding with these extra properties as a buffer partial embedding. We conclude that the probability of embedding \( y_t \) to \( \mathcal{N}_G(v) \) and maintaining a buffer partial embedding is at least \( d^3p/10 \), conditioning on the history, for each \( j \). The conditional probabilities multiply, giving the desired result.

\textbf{Proof.} We require
\[
\mu \leq \frac{1}{5}, \quad \varepsilon < (\varepsilon')^2, \quad \varepsilon' \leq \frac{\mu d\Delta \xi}{1000\kappa\Delta^2} \quad \text{and} \quad (d - \varepsilon')^{10\Delta} > \frac{1}{2}d^{10\Delta}.
\]

For each \( 0 \leq t \leq b \) and each \( y_t \), we define \( \hat{A}_t^{\text{main}}(y_t) := A_t^{\text{main}}(y_t) \cap \mathcal{N}_G(v) \) and for any vertex \( x' \in X \) we set \( \hat{U}_t(x') := U_t(x') \cap \mathcal{N}_G(v) \).

Observe that since \( y_t \) is embedded to a vertex of \( C_{t-1}(y_t) \setminus B_{t-1}(y_t) \) for each \( t = 1, \ldots, b \) we automatically maintain the property that \( \psi_t \) is a good partial embedding for each \( t \) (see Definition 2.24 of bad vertices). We formulate five additional conditions on \( \psi_t \) for \( t = 0, \ldots, b \) which we refer to as a buffer partial embedding, which allow us to show a lower bound on the desired probability inductively.

\textbf{(BPE1)} We have \( \psi_t(y_1), \ldots, \psi_t(y_b) \in \mathcal{N}_G(v) \),

\textbf{(BPE2)} \( |\hat{U}_t(y_k)| = (p + \varepsilon'p)\pi_t(y_k) \deg_G(v, V(y_k)) \) for \( k = t + 1, \ldots, b \),

\textbf{(BPE3)} \( |\hat{A}_t^{\text{main}}(y_k)| > \frac{1}{2}(dp - \varepsilon'p)\pi_t(y_k) \deg_G(v, V(y_k)) \) for \( k = t + 1, \ldots, b \).

\textbf{(BPE4)} for unembedded \( y_k \) and \( y_k \), with \( y_ky_k \in E(H) \) we have \( (\hat{U}_t(y_k), \hat{U}_t(y_k)) \) is

\textbf{(BPE5)} for unembedded \( y_k \) and \( z \) with \( y_kz \in E(H) \) the pair \( (\hat{U}_t(y_k), \hat{U}_t(z)) \) is

Observe that these conditions are all satisfied for \( \psi_0 \), i.e. \( \psi_0 \) is a buffer partial embedding. Indeed, \( \textbf{(BPE1)} \) is vacuously satisfied. \( \textbf{(BPE2)} \) holds by definition of \( \hat{U} \) and since no neighbours of any \( y_k \) are embedded. \( \textbf{(BPE3)} \) holds by the definition of \( \hat{A}_t^{\text{main}} \), the assumption on \( \deg_G(v, V(x_k)) \) of the lemma, and since no neighbours of any \( y_k \) are embedded. If \( y_k \) and \( y_k \) are adjacent in \( H \) then \( xy_ky_k \) is a triangle in \( H \) with \( x \in X^{\text{buf}} \subseteq X_\gamma \), so by \( \textbf{(G.2)} \) we have \( \textbf{(BPE4)} \). Finally, again since \( x \in X^{\text{buf}} \subseteq X_\gamma \), by \( \textbf{(G.2)} \) we have \( \textbf{(BPE5)} \).

For each \( t = 0, \ldots, b - 1 \) we let \( P_t \) be the set of poor vertices \( u \) in \( \hat{A}_t^{\text{main}}(y_{t+1}) \) such that if \( \psi_t(y_{t+1}) = u \) then \( \psi_t \) is not a buffer partial embedding.

We now show that, for any \( t = 1, \ldots, b \), given a buffer partial embedding \( \psi_{t-1} \), the probability that \( \psi_t \) is a buffer partial embedding is at least \( d^3p/10 \). This clearly
yields, by multiplying the conditional probabilities, the lower bound of \((d^3 p/10)^b\) claimed in the statement of the lemma. For the analysis it will suffice to show that only a tiny proportion of vertices from \(A_{t-1}^\text{main}(y_t)\) are in \(B_{t-1}(y_t)\) or \(I_{t-1}\). These estimates follow along the lines of Lemma 2.25. Properties \((BPE1)-(BPE5)\) are crafted in a way that allows the inductive verification below.

We first give some lower bounds on set sizes. If \(\psi_t\) is a good partial embedding, then we have for an unembedded vertex \(z\):

\[
|U_t^\text{c}(z)| \geq |C_t^\text{n}(z)| \geq (1 - \varepsilon') \mu(dp - \varepsilon' p) \tau^\Omega(z) \zeta|V(z)| \geq \varepsilon' p \tau^\Omega(z)n/r_1 \tag{16}
\]

where the second inequality is by \((GPE3)\) and by \((G5)\), and the third because \(|V(z)| \geq n/(\kappa r_1)\) and by choice of \(\varepsilon'\). The same lower bound also holds for \(U_t(z), C_t(z)\) et cetera since these sets are all by \((GPE2)\) and \((GPE3)\) at least as large. If in addition \(\psi_t\) is a buffer partial embedding, then for \(y_k\) with \(k > t\) we have

\[
|\tilde{U}_t(y_k)| \geq |\hat{A}_{t}^\text{main}(y_k)| \geq \frac{1}{2} (dp - \varepsilon' p) \tau^\Omega(y_k) \deg_G(v; V(y_k)) \geq \varepsilon' p \tau^\Omega(y_k) n/r_1 \tag{17}
\]

where the second inequality is by \((BPE3)\) and the third is by \((G3)\) and choice of \(\varepsilon'\). We also have for \(k > t\), using \((BPE2), (BPE3), (GPE2), (G3)\) and the choice of \(\varepsilon\) and \(\varepsilon'\) that

\[
|\hat{A}_{t}^\text{main}(y_k)| \geq \max \left( \frac{1}{4} d^3 p |U_t(y_k)|, \frac{1}{4} d^3 |\tilde{U}_t(y_k)| \right) \geq \varepsilon'|\tilde{U}_t(y_k)|. \tag{18}
\]

Estimating \(|\hat{A}_{t-1}^\text{main}(y_t) \cap B_{t-1}(y_t)|\). Suppose \(\psi_{t-1}\) is a buffer partial embedding. In the following we are going to estimate the number of vertices \(w\) from \(\hat{A}_{t-1}^\text{main}(y_t)\) which turn out to be bad with respect to \(\psi_{t-1}\) and \(Q_{t-1}\). For that we will consider all of the possible reasons.

First assume that there is some unembedded neighbour \(z \in V(H) \setminus Q_{t-1}\) of \(y_t\) such that the ‘badness condition’

\[
\deg_G(w; A^\text{main}(z)) < (d - \varepsilon') p |A^\text{main}(z)| \tag{19}
\]

holds for \(w\) and \(z\). The pair \((\tilde{U}_{t-1}(y_t), U_{t-1}(z))\) is \((\varepsilon, \tau_{t-1}(y_t), \tau_{t-1}(z), d, p)\)-regular in \(G\) by \((BPE5)\). Because \(z \notin Q_{t-1}\) we have that

\[
|\hat{A}_{t-1}^\text{main}(z)| \geq \frac{1}{2} \mu(d - \varepsilon') \tau_{t-1}(z) \rho \tau_{t-1}(z)|V^\text{main}(z)| \geq \frac{1}{2} \mu(1 - 3\mu)(d - \varepsilon') \tau_{t-1}(z) \rho \tau_{t-1}(z)|V(z)| \geq \varepsilon'|\tilde{U}_{t-1}(z)|,
\]

where the first inequality is by choice of \(Q_{t-1}\) in the statement of the lemma, the second inequality is by \((GPE2)\) and the last by choice of \(\varepsilon'\). Therefore the badness condition \((19)\) holds for at most

\[
\varepsilon' \tau_{t-1}(y_t) |\tilde{U}_{t-1}(y_t)| \leq 4d^3 \varepsilon' |\hat{A}_{t-1}^\text{main}(y_t)|
\]

vertices \(w\) of \(\hat{A}_{t-1}^\text{main}(y_t)\). Since \(y_t\) may have at most \(\Delta\) unembedded neighbours, in total there are at most \(4\Delta d^3 \varepsilon' |\hat{A}_{t-1}^\text{main}(y_t)|\) vertices \(w\) of \(\hat{A}_{t-1}^\text{main}(y_t)\) such that the badness condition \((19)\) holds for some unembedded neighbour of \(y_t\).

Next we need to estimate the number of vertices \(w\) from \(\hat{A}_{t-1}^\text{main}(y_t)\) such that \(\psi_{t-1} \cup \{y_t \rightarrow w\}\) is not a good partial embedding (i.e. doesn’t satisfy properties \((GPE1)-(GPE4)\)). First observe that since \(w \in \hat{A}_{t-1}^\text{main}(y_t) \subseteq C_{t-1}(y_t) \subseteq I_{t-1}\), \((GPE1)\) cannot fail.

Next we turn to \((GPE2)\). It is sufficient to consider unembedded neighbours \(z\) of \(y_t\), so \(\tau_{t-1}(z) \leq \Delta - 1\). Since \(\psi_{t-1}\) is a good partial embedding, by \((16)\) we have

\[
|U_{t-1}(z)| \geq \varepsilon' \rho^{\Delta - 1}n/r_1.
\]

Therefore, by the neighbourhood size property \(NS(\varepsilon, r_1, B)\), at most \(\varepsilon' \rho^{\Delta - 1}n/r_1^2\) vertices from \(\hat{A}_{t-1}^\text{main}(y_t)\) violate condition \((GPE2)\) for \(U_{t-1}(z)\). Further we have \(\tau_{t-1}(y_t) \leq B - 2\) since \(y_t\) has at most \(y_1, \ldots, y_{B-2}\) as embedded
neighbours (if \(b = B\) then by assumption \(y_{b}y_{b-1}\) is not an edge of \(H\)), so since \(\varepsilon < (\varepsilon')^2\) we have
\[
\varepsilon p^{B-1}n/r_1^2 \leq \varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_t)|.
\]
We can argue analogously for the other four conditions of (GPE2) obtaining in total that for all but at most \(5\Delta\varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_t)|\) vertices \(w\) from \(\hat{A}^{\text{main}}_{t-1}(y_t)\), (GPE2) is satisfied with respect to \(\psi_{t-1} \cup \{y_t \to w\}\).

Now we consider (GPE3). Let \(z\) be any unembedded neighbour of \(y_t\). Because \(z\) is at distance at most 2 from \(x \in X^{\text{inf}}\), by (H5) \(z\) is not image restricted, so \(I_z = V(z)\). Since (GPE2) and (GPE3) hold for \(\psi_{t-1}\) we have:
\[
|C_{t-1}(z)| \geq (1 - \varepsilon')(dp - \varepsilon'p)^{t-1}(z)|V(z)| \geq \varepsilon'|U_{t-1}(z)|.
\]
By (BPE5), \((\hat{U}_{t-1}(y_t), U_{t-1}(z))\) is \((\varepsilon\pi_{t-1}(y_t), \pi_{t-1}(z), d, p)\)-regular in \(G\). So, there are at most
\[
\varepsilon\pi_{t-1}(y_t) \hat{U}_{t-1}(y_t) \leq 4d^\Delta \varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_t)|
\]
vertices \(w\) in \(\hat{U}_{t-1}(y_t)\) such that \(|C_{t-1}(z) \cap N_G(w)| < (d - \varepsilon'p)|C_{t-1}(z)|\). Similarly we argue for each of the other four conditions. Thus, since \(y_t\) has at most \(\Delta\) unembedded neighbours, in total for all but at most \(20\Delta d^\Delta \varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_t)|\) vertices \(w \in \hat{A}^{\text{main}}_{t-1}(y_t)\) we have that (GPE3) is satisfied with respect to \(\psi_{t-1} \cup \{y_t \to w\}\).

It remains to consider (GPE4). Again, we only need to consider edges \(zz' \in E(H)\) between unembedded vertices such that \(z \in N_H(y_t)\). Consider first the case that \(z' \notin N_H(y_t)\). In this case we have \(\pi^*(z') \leq \Delta - 2, \pi^*(z') \leq \Delta - 1\) and \(\pi_{t-1}(y_t) \leq \Delta - 2\) (since both \(x\) and \(z\) are unembedded neighbours of \(y_t\)). Since \(\Gamma\) satisfies property \(\overline{\text{RI}}(\varepsilon, (\varepsilon_a, b), \varepsilon', d, r_1, B)\), using (16) applied to \(\psi_{t-1}\) and the fact that by (GPE4) the pair \((U_{t-1}(z), U_{t-1}(z'))\) is \((\varepsilon\pi_{t-1}(z), \pi_{t-1}(z'), d, p)\)-regular in \(G\), we see that there are at most
\[
\varepsilon p^\Delta n/r_1^2 \leq \varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_t)|
\]
vertices \(w \in \hat{A}^{\text{main}}_{t-1}(y_t)\) such that the pair \((N_{\Gamma}(w) \cap U_{t-1}(z), U_{t-1}(z'))\) is not \((\varepsilon\pi_{t-1}(z) + 1, \pi_{t-1}(z'), d, p)\)-regular in \(G\).

In the case that \(z, z' \in N_H(y_t)\) we have \(\pi_{t-1}(y_t) \leq B - 3\). This requires a little explanation. Observe that \(y_t\) has unembedded neighbours \(z\) and \(z'\), so \(\pi_{t-1}(y_t) \leq \Delta - 2\), and if \(B = \Delta + 1\) then we are done. If \(B = \Delta\), then observe that the embedded neighbours of \(y_t\) are contained in \(\{y_1, \ldots, y_{t-1}\}\). It follows that if \(t \leq \Delta - 2\), we have \(\pi_{t-1}(y_t) \leq \Delta - 3\) as desired. If \(t = \Delta\), then \(x, z\) and \(z'\) are distinct (since the only unembedded neighbour of \(x\) is \(y_t\)) and we again have \(\pi_{t-1}(y_t) \leq \Delta - 3\). It remains to consider the case \(t = \Delta - 1\). Again we are done if \(x, z\) and \(z'\) are distinct. If however (without loss of generality) we have \(x = z\), then \(z'\) is an unembedded neighbour of \(x\) which is not \(y_t\); in other words, we have \(z' = y_\Delta\). But we assumed that \(y_\Delta y_\Delta\) is not an edge of \(H\), contradicting the assumption \(y_t z' \in H\). We have thus justified \(\pi_{t-1}^*(y_t) \leq B - 3\), so by \(\overline{\text{RI}}(\varepsilon, (\varepsilon_a, b), \varepsilon', d, r_1, B)\) we conclude that there are at most
\[
\varepsilon p^{B-2} n/r_1^2 \leq \varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_t)|
\]
vertices \(w \in \hat{A}^{\text{main}}_{t-1}(y_t)\) such that \((N_{\Gamma}(w) \cap U_{t-1}(z), N_{\Gamma}(w) \cup U_{t-1}(z'))\) is not \((\varepsilon\pi_{t-1}(z) + 1, \pi_{t-1}(z') + 1, d, p)\)-regular. So, in total there are at most \(\Delta^2 \varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_t)|\) vertices \(w \in \hat{A}^{\text{main}}_{t-1}(y_t)\) such that (GPE4) is not satisfied with respect to \(\psi_{t-1} \cup \{y_t \to w\}\).

So far we have seen an estimate on \(|\hat{A}^{\text{main}}_{t-1}(y_t) \cap B_{t-1}(y_t)|\): at most
\[
4\Delta d^\Delta \varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_t)| + 5\Delta \varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_t)| + 20\Delta d^\Delta \varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_t)| + \Delta^2 \varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_t)|
\]
vertices \(w \in \hat{A}^{\text{main}}_{t-1}(y_t)\) are such that \(w\) is bad for \(y_t\) with respect to \(\psi_{t-1}\) and \(Q_{t-1}\).
Estimating $|P_t - 1|$. Observe that by definition, if $\psi_{t-1}$ is a buffer partial embedding then for each $w \in \hat{A}^{\text{main}}_{t-1}(y_t)$ the embedding $\psi_{t-1} \cup \{y_t \to w\}$ has (BPE1). For $t = b$ the remaining properties (BPE2)-(BPE5) are trivial, so we now on assume $t \leq b - 1$.

For (BPE2), let $k \in \{t + 1, \ldots, b\}$ be such that $y_k y_k \in E(H)$. Observe that $\pi^*_{t-1}(y_k) \leq t - 1 \leq b - 2 \leq B - 2$. Therefore, by NS($\varepsilon$, $r_1$, $B$) and (17), for all but at most $\varepsilon_p^{B-1} n/r_1^2 \leq \varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_k)|$ vertices $w$ we have $|N_T(w) \cap \hat{U}_{t-1}(y_k)| = (1 + \varepsilon') \rho|\hat{U}_{t-1}(y_k)|$.

To estimate the number of vertices $w$ that do not preserve (BPE3), we only need to consider those $y_k$ with $k \geq t + 1$ such that $y_k y_k \in E(H)$. We use that $(\hat{U}_{t-1}(y_k), \hat{U}_{t-1}(y_k))$ is $(\varepsilon', d, p)$-regular in $G$ by (BPE4). By (18) we have $|\hat{A}^{\text{main}}_{t-1}(y_k)| \geq \varepsilon'|\hat{U}_{t-1}(y_k)|$, so there are at most

$$\varepsilon'|\hat{U}_{t-1}(y_k)| \leq 4d^{-\Delta} \varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_k)|$$

vertices $w$ in $\hat{A}^{\text{main}}_{t-1}(y_k)$ such that $\deg_{G}(w, \hat{A}^{\text{main}}_{t-1}(y_k)) < (d - \varepsilon') \rho|\hat{A}^{\text{main}}_{t-1}(y_k)|$.

Next we consider (BPE5). Let $k \in \{t + 1, \ldots, b\}$, and $z \in V(H)$ be such that $y_k z \in E(H)$. By (BPE5) the pair $(\hat{U}_{t-1}(y_k), \hat{U}_{t-1}(z))$ is $(\varepsilon_{t-1}(y_k), \pi_{t-1}^*(y_k), d, p)$-regular in $G$. There are three cases to consider: $y_k y_k \in E(H)$, $y_k z \in E(H)$, and both.

In the first case (i.e. $y_k y_k \in E(H)$, $y_k z \notin E(H)$), we have $t < B - 1$ because $y_k y_k y_k$ is not an edge of $H$, so we conclude $\pi^*_{t-1}(y_k), \pi^*_{t-1}(y_k) \leq t - 1 \leq B - 3$, and $\pi^*_{t-1}(z) \leq \Delta - 1$. By (16) and (17) applied to $\psi_{t-1}$, and $\mathrm{RI}(\varepsilon, (\varepsilon_{a,b}), \varepsilon', d, r_1, B)$, we see that for all vertices $w$ of $\hat{A}^{\text{main}}_{t-1}(y_k)$ but at most

$$\varepsilon_p^{B-1} n/r_1^2 \leq \varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_k)|,$$

the pair $(N_T(w) \cap \hat{U}_{t-1}(y_k), \hat{U}_{t-1}(z))$ is $(\varepsilon_{t-1}(y_k), \pi_{t-1}^*(y_k) + 1, \pi_{t-1}(z), d, p)$-regular in $G$.

In the second case (i.e. $y_k y_k \notin E(H)$, $y_k z \in E(H)$), we have $\pi^*_{t-1}(y_k), \pi^*_{t-1}(y_k) \leq t - 1 \leq B - 2$, and $\pi^*_{t-1}(z) \leq t - 1 \leq \Delta - 2$ and similarly, by $\mathrm{RI}(\varepsilon, (\varepsilon_{a,b}), \varepsilon', d, r_1, B)$, all but at most $\varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_k)|$ vertices $w$ of $\hat{A}^{\text{main}}_{t-1}(y_k)$ are such that $(\hat{U}_{t-1}(y_k), N_T(w) \cap \hat{U}_{t-1}(z))$ is $(\varepsilon_{t-1}(y_k), \pi_{t-1}(z) + 1, d, p)$-regular in $G$.

In the final case (i.e. $y_k y_k \in E(H)$, $y_k z \in E(H)$), again we have $t < B - 1$, hence $\pi^*_{t-1}(y_k), \pi^*_{t-1}(y_k) \leq B - 3$, and $\pi^*_{t-1}(z) \leq \Delta - 2$, and again all but at most $\varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_k)|$ vertices $w$ of $\hat{A}^{\text{main}}_{t-1}(y_k)$ are such that $(N_T(w) \cap \hat{U}_{t-1}(y_k), N_T(w) \cap \hat{U}_{t-1}(z))$ is $(\varepsilon_{t-1}(y_k), \pi_{t-1}(z) + 1, d, p)$-regular in $G$.

In total, we see that for all but at most $\Delta^2 \varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_k)|$ vertices $w$ of $\hat{A}^{\text{main}}_{t-1}(y_k)$, the partial embedding $\psi_{t-1} \cup \{y_k \to w\}$ has (BPE5).

Finally, we handle (BPE4). Let $k, \ell \in \{t + 1, \ldots, b\}$ be such that $y_k y_\ell \in E(H)$. We have $\pi^*_{t-1}(y_k), \pi^*_{t-1}(y_\ell), \pi^*_{t-1}(y_k) \leq t - 1 \leq B - 3$. By (BPE4), the pair $(\hat{U}_{t-1}(y_k), \hat{U}_{t-1}(y_\ell))$ is $(\varepsilon_{t-1}(y_k), \pi_{t-1}(y_k), d, p)$-regular in $G$. Without loss of generality we may assume $y_k y_\ell \in E(H)$, and again there are two cases to consider depending on whether $y_k y_\ell \in E(H)$ or not. As before, using (17), by $\mathrm{RI}(\varepsilon, (\varepsilon_{a,b}), \varepsilon', d, r_1, B)$, at most

$$\varepsilon_p^{B-2} n/r_1^2 \leq \varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_\ell)|$$

vertices $w \in \hat{A}^{\text{main}}_{t-1}(y_k)$ are such that $(N_T(w) \cap \hat{U}_{t-1}(y_k), N_T(w) \cap \hat{U}_{t-1}(y_\ell))$ is not $(\varepsilon_{t-1}(y_k), \pi_{t-1}(y_k) + 1, d, p)$-regular in $G$. The other case follows similarly, and we conclude that for all but at most $\Delta^2 \varepsilon' |\hat{A}^{\text{main}}_{t-1}(y_k)|$ vertices $w$ of $\hat{A}^{\text{main}}_{t-1}(y_k)$, the partial embedding $\psi_{t-1} \cup \{y_k \to w\}$ has (BPE4).
2.4. THE BEHAVIOUR OF RANDOM GREEDY ALGORITHMS

Summing up, there are at most
\[
4\Delta d^{-}\varepsilon' |\hat{A}_{t-1}^{\text{main}}(y_t)| + 5\Delta d^{-}\varepsilon' |\hat{A}_{t-1}^{\text{main}}(y_t)| + 20\Delta d^{-}\varepsilon' |\hat{A}_{t-1}^{\text{main}}(y_t)| + \Delta^2 \varepsilon' |\hat{A}_{t-1}^{\text{main}}(y_t)|
\]
\[
+ \Delta \varepsilon' |\hat{A}_{t-1}^{\text{main}}(y_t)| + 4\Delta d^{-}\varepsilon' |\hat{A}_{t-1}^{\text{main}}(y_t)| + \Delta^2 \varepsilon' |\hat{A}_{t-1}^{\text{main}}(y_t)|
\]
\[
\leq 50\Delta^2 d^{-}\varepsilon' |\hat{A}_{t-1}^{\text{main}}(y_t)|
\]

vertices \( w \in \hat{A}_{t-1}^{\text{main}}(y_t) \) such that \( w \) is bad for \( y_t \) with respect to \( \psi_{t-1} \) and \( Q_{t-1} \) or \( w \in P_{t-1} \).

Now we can estimate the probability that \( \psi_t \) is a buffer partial embedding, conditioning on the history and on \( \psi_{t-1} \) being a buffer partial embedding. This event occurs if \( y_t \) is embedded to a member of \( \hat{A}_{t-1}^{\text{main}}(y_t) \setminus (B_{t-1}(y_t) \cup P_{t-1}) \). The number of such vertices is at least
\[
|\hat{A}_{t-1}^{\text{main}}(y_t)| - 50\Delta^2 d^{-}\varepsilon' |\hat{A}_{t-1}^{\text{main}}(y_t)| \geq \frac{1}{2}|\hat{A}_{t-1}^{\text{main}}(y_t)| \quad \geq \frac{1}{2}(dp - \varepsilon' p)\pi_{t-1}(y_t) + 1|V(y_t)|,
\]
while \( y_t \) is embedded into a set of size at most \( U_{t-1}(y_t) \), which by \( (GPE2) \) has size at most \( (p + \varepsilon p)\pi_{t-1}(y_t)|V(y_t)| \). We see that the desired conditional probability is at least
\[
\frac{1}{2}(dp - \varepsilon' p)\pi_{t-1}(y_t) + 1|V(y_t)| \geq \frac{d\Delta p}{10}.
\]

The statement of the lemma follows since the conditional probabilities multiply. \( \square \)
Proof of the blow-up lemma for random graphs

3.1. Main lemmas and the proof of the blow-up lemma

In this section we divide the proof of Lemma 1.21 into four lemmas, which correspond to our four different embedding stages described in Section 1.4. The first three of these lemmas will be proved in the subsequent three sections. The proof of the fourth lemma is short and hence given here. At the end of this section we show how these four lemmas together with Lemma 2.22 imply Lemma 1.21.

The first of our lemmas encapsulates the outcome after applying the randomised greedy algorithm (RGA) which tries to embed $X_{\text{main}}$ into $V_{\text{main}}$. How this RGA operates is explained in the proof of this lemma in Section 3.2. The lemma claims the existence of a good partial embedding with certain deterministic properties which we require and which the RGA with high probability produces.

**Lemma 3.1 (RGA lemma).** We assume the General Setup. Suppose that $\Gamma$ has $\text{NS}(\varepsilon, r_1, \Delta)$ and $\text{RI}(\varepsilon, (\varepsilon a, b), \varepsilon', d, r_1, \Delta)$. Then there is a good partial embedding $\psi_{\text{RGA}}$ with the following properties. For each $i$, let $X^q_i := X_{\text{main}}^i \setminus \text{Dom}(\psi_{\text{RGA}})$.

Then the following hold for each $i$.

1. **(RGA 1)** All neighbours of all buffer vertices are embedded by $\psi_{\text{RGA}}$.
2. **(RGA 2)** Every vertex in $X_i \cap \text{Dom}(\psi_{\text{RGA}})$ is embedded to $V_{\text{main}}^i$ by $\psi_{\text{RGA}}$.
3. **(RGA 3)** We have $|X^q_i| \leq 2\rho |X_i|$. 
4. **(RGA 4)** For every set $W \subseteq V_i$ of size at least $\rho |V_i|$, there are at most $\rho |X_i| / 2$ candidates in $W$.
5. **(RGA 5)** If $X^b_i$ is not a clique buffer, then every vertex in $V_i$ is a candidate for at least $\mu (\frac{d\Delta p}{100}) |X_i| / 2$ vertices of $X^b_i$.

Note that we do not have that all vertices in $V_i$ are candidate for many vertices of $X^b_i$ if $X^b_i$ is a clique buffer, and actually it may not be true. But it follows from point (RGA 4) that only at most $\rho |V_i|$ vertices of $V_i$ can fail to be candidate for many vertices of $X^b_i$. This is the ‘technical complication’ mentioned in the proof overview. It is due to this complication that we will need to utilize the sets $V_i^c$ in Section 3.4.

The following lemma, proved in Section 3.3, allows us to embed the queue vertices into the sets $V_i^q$. Recall that in the proof overview we mentioned that the proof that we can embed the queue vertices relies on showing that failure to do so would imply the existence of a ‘dense spot’ in $G$ which cannot even be present in $\Gamma$. The congestion condition $\text{CON}(\rho, r_1, \Delta)$ which we assume of $\Gamma$ constitutes the statement that $\Gamma$ has no ‘dense spot’.

**Lemma 3.2 (Queue embedding lemma).** We assume the General Setup. Suppose that $\Gamma$ has $\text{NS}(\varepsilon, r_1, \Delta)$, $\text{RI}(\varepsilon, (\varepsilon a, b), \varepsilon', d, r_1, \Delta)$ and $\text{CON}(\rho, r_1, \Delta)$. Let $\psi$ be a good partial embedding whose image is disjoint from the sets $V_i^q$, and suppose that for each $i$ we have a set $X^q_i \subseteq X_i$ of size at most $2\rho |X_i|$. Then there is a good partial embedding $\psi_q$ extending $\psi$ such that $\text{Dom}(\psi_q) \setminus \text{Dom}(\psi) = X^q$ and $\text{Im}(\psi_q) \setminus \text{Im}(\psi) \subseteq V^q_i$. 

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Recall that in the conclusion of Lemma 3.1 we were not able to guarantee that all vertices of \( V_i \) are candidate for many vertices of \( X_i^{buf} \) in the event that \( X_i^{buf} \) is a clique buffer. In order to ‘fix’ this problem, we will, before we begin the embedding, ‘reserve’ some copies \( K_i \) of \( K_{\Delta+1} \), each of which has a vertex in \( X_i \). We put the vertices of all such reserved cliques into the set \( X^c \), which as we mentioned in Section 2.3.2 is disjoint from \( X^{main} \) and \( X^{buf} \) (full details will be given in the proof of Lemma 1.21 below). After we embed the queue using Lemma 3.2 we may find that there are still some vertices of \( V_i \) which are not in the image of our current good partial embedding \( \psi \) and which are not candidate for many vertices of \( X_i^{buf} \).

We embed the cliques \( K_i \), using these vertices and the vertices of the clique reservoir \( V^c \), to yield a good partial embedding \( \psi_{good} \) in which only the vertices of \( X_i^{buf} \) remain to be embedded and in which every vertex of \( V_i \) which is not in the image of \( \psi_{good} \) is candidate for many vertices of \( X_i^{buf} \). The following lemma, which we prove in Section 3.4, states that, given appropriate sets of reserved cliques, this embedding is possible. Note that the above sketch is a slight oversimplification: we may actually need to embed a few vertices of \( X_i^{buf} \) in this step as well.

**Lemma 3.3 (Embedding reserved cliques).** We assume the General Setup. Suppose that \( \Gamma \) has NS(\( \varepsilon, r_1, \Delta \)), RI(\( \varepsilon, (\varepsilon_{a,b}, \varepsilon', d, r_1, \Delta) \)) and CON(\( q, r_1, \Delta \)). Let \( \psi \) be a good partial embedding whose image is disjoint from the sets \( V^c_i \). Suppose that for each \( i \), either

(a) \( X_i^{buf} \) is a degree-\( b \) non-clique buffer with \( K_i = \emptyset \), and all vertices in \( V_i \) are candidate for at least \( \mu(d^{\Delta+1}p/100)^b|X_i| \) vertices in \( X_i^{buf} \), or

(b) \( X_i^{buf} \) is a clique buffer with \( |K_i| = 2q|X_i| \), and all but at most \( q|V_i| \) vertices of \( V_i \) are candidate for at least \( \mu(dp)\Delta|X_i|/2 \) vertices in \( X_i^{buf} \).

Finally suppose that for each \( i \in [r] \), where \( X_i^{buf} \) is a clique buffer, we have a set \( K_i \) of \( 2q|X_i| \) reserved cliques \( K_{\Delta+1} \), and for \( K \in K_i \) and \( x, y \in K \) we have \( x \in X_j \) and \( y \in X_k \) where \( jk \in R' \). Moreover, let the set \( X_i^c \) of vertices \( x \in X_i \) contained in some reserved clique from \( K_i \) satisfy: \( |X_i^c| \leq 2\varepsilon(\Delta R + 1)|q|X_i| \).

Then there is a good partial embedding \( \psi_{good} \) extending \( \psi \) such that for each \( i \), where \( X_i^{buf} \) is a degree-\( b \) buffer, the following hold.

(FIN1) \( X_i^c \subseteq \text{Dom}(\psi_{good}) \).

(FIN2) For each \( x \in X_i^{buf} \setminus \text{Dom}(\psi_{good}) \) we have \( |A^{buf}(x)| \geq \mu(dp)^b|V_i|/4 \).

(FIN3) Each \( v \in V_i \setminus \text{Im}(\psi_{good}) \) is candidate for at least \( \mu(d^{\Delta+1}p/100)^b|X_i| \) of the vertices \( X_i^{buf} \setminus \text{Dom}(\psi_{good}) \).

Recall that \( A^{buf}(x) = A(x) \cap V^{buf}(x) = C(x) \cap V^{buf}(x) \setminus \text{Im}(\psi) \).

Finally, we give a lemma which we use to show that we can embed the reserved vertices. This lemma is proved using Hall’s condition, as we outlined in the proof overview. We will use it again in the proof of Lemma 1.23, which is why we do not give an explicit constant in (CPM3).

**Lemma 3.4 (Embedding buffer vertices).** We assume the General Setup. Suppose that \( \Gamma \) has CON(\( p, r_1, \Delta \)). Suppose that \( \psi \) is a good partial embedding, in which all the vertices \( N(X_i^{buf}) \) are embedded, and that for some \( i \) we have a subset \( X_i^r \) of \( X_i^{buf} \), and a subset \( V_i' \) of \( V_i \setminus \text{Im}(\psi) \) with \( |X_i^r| = |V_i'| \). Suppose \( X_i^{buf} \) is a degree-\( b \) buffer, and for some \( \delta > 0 \) we have

(CPM1) for each \( x \in X_i^r \) we have \( |C(x) \cap V_i'| \geq \mu(dp)^b|V_i|/4 \).

(CPM2) for every set \( W \subseteq V_i' \) of size at least \( q|V_i| \), there are at most \( q|X_i| \) vertices in \( X_i^r \) which do not have a candidate in \( W \), and

(CPM3) each \( v \in V_i' \) is candidate for at least \( \delta dp|X_i| \) of the vertices \( X_i^r \).

Suppose further that \( q \leq \frac{\text{fin}_b}{100} \). Then there is a good partial embedding \( \psi' \) extending \( \psi \) such that \( \text{Dom}(\psi') = \text{Dom}(\psi) \cup X_i^r \) and \( \text{Im}(\psi') = \text{Im}(\psi) \cup V_i' \).
Proof. Let \( Y \subseteq X'_i \) be non-empty, and let \( U \) be the set of vertices in \( V'_i \) which are candidate for some member of \( Y \). We wish to verify Hall’s condition, i.e. show that \(|U| \geq |Y|\). We separate three cases.

First, suppose \( 0 < |Y| \leq \frac{|X_i|}{\varrho} \). If \(|U| < |Y|\), then we can take a subset \( Y' \) of \( Y \) of size \(|U|\). Since by \((CPM1)\) each vertex in \( Y' \) has at least \( \mu(dp)^{|V_i|}/4 \) candidates in \( V'_i \), which must lie in \( U \), the number of edges in the candidate graph between \( Y' \) and \( U \) is at least
\[
\frac{1}{2} \mu(dp)^{|V_i|}|Y'| - \frac{1}{2} \mu(dp)^{|V_i|}|Y'| + \frac{1}{2} \mu(dp)^{|V_i|}|Y'|
\begin{align*}
\geq & \frac{\mu dp}{8\varrho} |Y'||U| + \frac{\mu dp}{8\varrho} |Y'| \\
\geq & 7p^{|Y'||U|} + \frac{\mu dp}{8\varrho} |Y'|/r_1,
\end{align*}
\]
where the second line comes from substituting \( |V_i| = |X_i| > |Y'|/\varrho = |U|/\varrho \) and the fact \(|V_i| \geq n/(\kappa r_1)\), respectively, and the third follows from the choice of \( \varrho \).

The candidate graph between \( Y' \) and \( U \) is a subgraph of the underlying restriction graph between the same sets, i.e. if \( uv \) is an edge then \( u \in U(y) \) (for the definitions see Section 2.3.5). We set \( \mathcal{F} = \{ \psi(N_H(y)) : y \in Y' \} \) and it is clear that the congestion graph \( CG(\Gamma, U, \mathcal{F}) \) is isomorphic to this underlying restriction graph. Hence \( e(CG(\Gamma, U, \mathcal{F})) > 7p^{|Y'||U|} + \frac{\mu dp}{8\varrho} |Y'|/r_1 \), in contradiction to \( CON(\varrho, r_1, \Delta) \).

We conclude that \(|U| \geq |Y|\) in this case.

Second, if \( |X_i| < |Y| \leq |X'_i| - \varrho |X_i| = |V'_i| - \varrho |V_i| \) and \(|U| < |Y|\), then \(|V'_i \setminus U| > \varrho |V_i|\), so by \((CPM2)\) there are at most \( \varrho |X_i| \) vertices of \( X'_i \) which do not have candidates in \( V'_i \setminus U \). In particular there is a vertex of \( Y \) with candidates in \( V'_i \setminus U \), in contradiction to the definition of \( U \). We conclude that \(|U| \geq |Y|\) in this case as well.

Finally, suppose \(|Y| > |X'_i| - \varrho |X_i| = |V'_i| - \varrho |V_i|\). The vertices \( V'_i \setminus U \) are candidates only for vertices in \( X'_i \setminus Y \), and each vertex in \( V'_i \setminus U \) is a candidate for at least \( \delta p^{|X_i|} \) vertices in \( X'_i \setminus Y \) by \((CPM3)\). If \(|U| < |Y|\), then \(|V'_i \setminus U| > |X'_i \setminus Y|\), and we can take a set \( W \subseteq V'_i \setminus U \) of size \(|X'_i \setminus Y|\). Now the number of edges in the candidate graph between \( W \) and \( X'_i \setminus Y \) is at least
\[
\delta p^{|X_i||W|} > \frac{1}{2} \delta p^{|W|} \left( \frac{1}{2} |X'_i \setminus Y||W| + \frac{n}{\kappa r_1} |X'_i \setminus Y| \right)
\begin{align*}
\geq & 7p^{|X'_i \setminus Y||W|} + \frac{\mu dp}{8\varrho} |X'_i \setminus Y|/r_1,
\end{align*}
\]
by essentially the same calculation as in the first case, using \(|X_i| = |V_i| \geq n/(\kappa r_1)\).

Taking \( \mathcal{F}' = \{ \psi(N_H(x)) : x \in X'_i \setminus Y \} \), the number of edges in the corresponding graph \( CG(\Gamma, W, \mathcal{F}') \) is at least this quantity, in contradiction to \( CON(\varrho, r_1, \Delta) \).

This completes the verification of Hall's condition, so there is a partial embedding \( \psi' \) extending \( \psi \) with \( \text{Dom}(\psi') = \text{Dom}(\psi) \cup X'_i \) in which the vertices \( X'_i \) are embedded to \( V'_i \). It is trivially the case, since \( X_{\text{buf}} \) is independent and all the vertices \( N(X_{\text{buf}}) \) are embedded in \( \psi' \), that \( \psi' \) is a good partial embedding.

We are now ready to prove Lemma 1.21. Briefly, the proof will go as follows. We will apply Lemma 2.22 to find partitions of \( G \) and \( H \), and graphs \( R \) and \( R' \), satisfying the General Setup. We choose, for each \( i \) such that \( X_i \) is a clique buffer, in succession, a set \( \mathcal{K}_i \) of \( 2\varrho |X_i| \) copies of \( K_{\Delta+1} \) each with a vertex in \( \tilde{X}_i \), none of which contain vertices in \( X_{\text{buf}} \) or image restricted vertices and none of which are in other sets \( \mathcal{K}_j \). We let \( X_i \) be the set of vertices in cliques \( \mathcal{K}_i \) for \( i \in [r] \). Now we are in a position to apply in succession Lemmas 3.1, 3.2 and 3.3 to obtain a good partial embedding \( \psi_{\text{good}} \) in which the only unembedded vertices are in \( X_{\text{buf}} \) and which satisfies the conclusions of Lemma 3.3. Since \( X_{\text{buf}} \) is independent in \( H \), we can embed the remaining vertices of \( X_{\text{buf}} \) to \( V_i \) for each \( i \) without affecting the properties of any other vertices. We do this using Lemma 3.4.
Proof of Lemma 1.21. First we choose constants as follows. Given $\Delta$, $\Delta^{BL}$, $\Delta_j$ integers, $\alpha^{BL}, \kappa^{BL}$ and $d > 0$, and $\kappa^{BL} > 1$, we set $\vartheta = 0$, $\Delta^{R'} = 8(\Delta + \Delta_j)^{10}A^{BL}_{\vartheta}$, $\alpha = \frac{1}{10}B^{BL}$, $\zeta = \frac{1}{2}B^{BL}$ and $\kappa = 2B^{BL}$. We now choose $\mu, q$ and $\epsilon > 0$ satisfying the conditions in Lemmas 2.22, 2.25, 3.4, 3.1, 3.2, 3.3, the last three of which are proved in the following three sections. For convenience we provide here sufficient choices:

$$
\mu < \frac{\mu^2d^{\Delta^2+1}}{1320\kappa\Delta R'}, \quad \vartheta < \frac{\mu^2d^{\Delta^2+1}}{250\kappa\Delta R'}, \quad \epsilon > \frac{\mu\zeta\vartheta d^\Delta}{32\kappa^2\Delta^2\Delta R'}.
$$

Now for input $\Delta$, $d$ and $\epsilon'$, Lemma 2.17 returns constants $\alpha, \beta$ and $\kappa > 0$. Here we additionally require that $\epsilon < (\epsilon')^2$. We let $\epsilon^{BL} = \frac{1}{10}(\Delta + \Delta_j)^{-10}\epsilon$ and $\vartheta^{BL} = \frac{1}{10}(\Delta + \Delta_j)^{-10}\epsilon$.

Now Lemma 1.21 returns $\epsilon^{BL}$ and $\vartheta^{BL}$. Given $r_1^{BL}$ we let $r_1 = 8(\Delta + \Delta_j)^{10}r^{BL}_1$. We choose $C$ sufficiently large for Lemma 2.17 with input $\Delta$, $\epsilon'$ and $r_1$. Given $p \geq C\left(\frac{\rho B}{\alpha}\right)^{\lambda}^{1/\alpha}$, Lemma 2.17 states that a.a.s. $\Gamma = G_{n,p}$ has properties $NS(\epsilon, r_1, \Delta)$, $R(\epsilon, (\epsilon,a), \epsilon', d, r_1, \Delta)$ and $CON(\vartheta, r_1, \Delta)$ respectively. From now on we will assume $\Gamma$ is an $n$-vertex graph which satisfies these three properties.

Given a graph $R^{BL}$ on $r_1^{BL}$ vertices, and a spanning subgraph $R^{BL}_v$ with $\Delta(R^{BL}) \leq \Delta^{BL}$, and graphs $H$ and $G \subseteq \Gamma$ with vertex partitions $X^{BL}$ and $Y^{BL}$, families of image restrictions $\mathcal{I}^{BL}$ and of image restricting vertices $\mathcal{J}^{BL}$, and a family of potential buffer vertices $\mathcal{X}^{BL}$, suppose that the conditions of Lemma 1.21 are satisfied. Then Lemma 2.22 gives (with $b_{1,22} = \Delta$) a graph $R$ on $r \leq r_1$ vertices, a spanning subgraph $R'$ with $\Delta(R') \leq \Delta^{R'}$, and $\kappa$-balanced size-compatible partitions $X$ and $V$ of $H$ and $G$ respectively, each part having size at least $n/(\kappa r_1)$, together with a family $\mathcal{X}$ of potential buffer vertices and $\mathcal{I}$ of image restrictions, subsets $X^\text{buf}_i$ of $X_i$ for each $i \in [r]$, and partitions $V_\psi = V^\text{main}_i \cup V^\text{ext}_i \cup V^\psi \cup Y^\text{buf}_i$ for each $i \in [r]$ which satisfy the General Setup.

Now for each $i \in [r]$ such that $X^\text{buf}_i$ is a clique buffer, in succession, we do the following. We choose $2g|X_i|$ vertices from $X_i$ which are in copies of $K_{\Delta+1}$ that do not contain image restricted vertices, vertices of $X^\text{buf}$, or vertices of $X^\text{c}$. We add these copies of $K_{\Delta+1}$ to $K_i$, and their vertices to $X^\text{c}$. Observe that this is possible since $(H8)$ guarantees that each $X_i$ contains at least $\frac{1}{2\Delta+2}\alpha|X_i|$ vertices in copies of $K_{\Delta+1}$ whose vertices are neither image restricted nor in $X^\text{buf}$. Moreover, by $(H1)$ the edges of these cliques lie along $R'$, so the number of these cliques which are chosen for some $K_j$ with $j < i$ is at most $\Delta^{R'}2g|X_i| < \frac{\alpha|X_i|}{2\Delta+2}$ by choice of $\vartheta$.

We obtain the following properties.

**RSC1** The sets $\{K_i\}_{i \in [r]}$ are pairwise disjoint, and if $X^\text{buf}_i$ is a clique buffer, then $K_i$ contains $2g|X_i|$ cliques, each with one vertex in $X_i$, otherwise it is empty.

**RSC2** For each $i \in [r], K \in K_i$ and $x, y \in K$ we have $x \in X_j$ and $y \in X_k$ where $jk \in R'$.

**RSC3** For each $i \in [r], X^\psi_i$ of vertices $x \in X_i$ contained in some reserved clique satisfies:

$$
|X^\psi_i| \leq 2\kappa(\Delta^{R'} + 1)g|X_i|.
$$

Notice that (RSC2) holds by $(H1)$, since the first and second neighbours of vertices from $X_i$ go along the edges of $R'$ by the definition of $(\alpha, R')$-buffer.

We let $X^\text{main} = VH \setminus (X^\text{buf} \cup X^\text{c})$. We now begin the embedding of $H$ into $G$.

By Lemma 3.1, there is a good partial embedding $\psi^{\text{RGA}}$ with properties $(\text{RGA1})$–$(\text{RGA5})$ stated in that lemma. Letting $X^\text{q} = X^\text{main} \setminus \text{Dom}(\psi^{\text{RGA}})$, by $(\text{RGA3})$ we have $|X^\text{q}_i| \leq 2g|X_i|$ for each $i$, and by $(\text{RGA2})$ we see that $\text{Im}(\psi^{\text{RGA}})$ is disjoint from each set $V^\text{q}_i$, so the conditions of Lemma 3.2 are met. Feeding $\psi^{\text{RGA}}$ into Lemma 3.2 we obtain a good partial embedding $\psi$ extending $\psi^{\text{RGA}}$ whose domain is $X^\text{main}$ and whose image is contained in $V^\text{main} \cup V^\text{q}$. By $(\text{RGA1})$ all neighbours of all buffer
3.2. THE RANDOM GREEDY ALGORITHM

vertices are in Dom(ψRGA), and therefore the candidate sets of all buffer vertices are the same with respect to ψRGA and to ψq. In particular ψq satisfies (RGA4) and (RGA5).

We now verify the conditions for Lemma 3.3. Since ψq extends ψRGA and the vertices Im(ψq) \ \Im(ψRGA) are embedded in the sets Vq it follows by (RGA2) that Im(ψq) is disjoint from the sets \( \bigcup V \) and \( \bigcup V_{buf} \). By (RGA5), if \( X_{buf} \) is not a clique buffer then all vertices in \( V_\ell \) are candidates for at least \( \mu(d^2p/100)^h|X_\ell| \) vertices in \( X_{buf} \) and \( \mathcal{K}_i \) is empty by (RSC1).

If on the other hand \( X_{buf} \) is a clique buffer, then \(|\mathcal{K}_i| = 2\varrho|X_\ell|\). Now let \( W \) be the set of vertices in \( V_\ell \) which are candidates for fewer than \( \mu(dp)^\Delta|X_\ell|/2 \) vertices of \( X_{buf} \). We will show that \(|W| < \varrho|V_\ell|\). If \(|W| \geq \varrho|V_\ell|\), then by (RGA4) there are at most \( \varrho|X_\ell| \) vertices in \( X_{buf} \) with fewer than \( (dp)^\Delta|W|/2 \) candidates in \( W \). It follows that the average over \( v \in W \) of the number of vertices of \( X_{buf} \) for which \( v \) is a candidate, is at least

\[
\frac{1}{|W|} \left(|X_{buf}^i| - \varrho|X_\ell|\right) \frac{(dp)^\Delta|W|}{2} > \frac{1}{2} \mu(dp)^\Delta|X_\ell|,
\]

where we used the facts \(|X_{buf}^i| = 4\mu|X_\ell|\) and \( \varrho < \mu \) in the inequality. Since at least one \( v \in W \) attains at least the average, we have a contradiction to the definition of \( W \).

The conditions of Lemma 3.3 are thus satisfied, and by applying it we obtain a good partial embedding \( \psi_{good} \). Since \( \psi_{good} \) extends \( \psi_q \) and all sets \( X_i^\ell \) are contained in Dom(\( \psi_{good} \)) by (FIN1), we conclude that the only vertices remaining unembedded are in \( X_{buf} \).

Finally, for \( i \in [\tau] \), let \( X_i^\ell = X_{buf} \setminus \Dom(\psi_{good}) \) and let \( X_i^\ell \) be a degree-\( b \) buffer for some \( b \). Let \( V_i^\ell = V_i \setminus \Im(\psi_{good}) \). Because \(|X_i| = |V_i|\) we have \(|X_i^\ell| = |V_i^\ell|\). Because \( \psi_{good} \) satisfies (FIN2), we have (CPM1). Because \( \psi_{good} \) satisfies (RGA4), in particular we have (CPM2). Finally, because \( \psi_{good} \) satisfies (FIN3) we have (CPM3) with \( \delta = \mu(d^2/100)^h \). Thus by Lemma 3.4 there is an embedding \( \psi' \) extending \( \psi_{good} \) which embeds \( X_i^\ell \) to \( V_i^\ell \). Repeating this for each \( i \in [\tau] \), which we may do since \( X_{buf} \) is an independent set in \( H \), we obtain the desired embedding of \( H \) into \( G \). □

3.2. The random greedy algorithm

In this section we describe the random greedy algorithm (RGA) and prove that it produces a partial embedding which satisfies the assertions of the RGA lemma, Lemma 3.1, with high probability. This is Algorithm 1 below. It embeds vertices \( x \) of \( H \) sequentially, following an order \( \tau \) given by Lemma 2.26. In doing so, it builds up a sequence \( \psi_0, \ldots \) of good partial embeddings, and a queue of vertices which it will not embed; we let \( Q_t \) be the queue at time \( t \) (i.e. corresponding to \( \psi_t \)). Recall that (as mentioned in Section 2.3.5) by \( A^\text{main}(x) \) we mean the set \( A^\text{main}(x) \) with reference to the partial embedding \( \psi_t \). We let \( B_t(x) \) denote the set of bad vertices (Section 2.3.7) with respect to \( \psi_t \) and \( Q_t \). As mentioned in the proof overview (Section 1.4), to create \( \psi_{t+1} \) from \( \psi_t \) we embed some vertex \( x \) uniformly at random into the set \( A^\text{main}(x) \setminus B_t(x) \), the set of available candidate vertices in \( X^\text{main} \) minus the bad vertices; and we add \( y \) to the queue if the set \( A^\text{main}(y) \setminus B_t(y) \) gets small.

Note that exactly \( t \) vertices are embedded in \( \psi_t \), though these vertices are not necessarily the first \( t \) vertices of \( \tau \); vertices in \( Q_t \) are skipped. The queue set \( Q_{t\text{RGA\#end}} \) at the time \( t_{\text{RGA\#end}} \) when the RGA terminates will then form the queue \( X^n \) mentioned in Lemma 3.1.

3.2.1. Proof of the RGA lemma. The proof that Algorithm 1 a.a.s. produces a good partial embedding with the properties required in Lemma 3.1 is now quite straightforward: most of the work is to check that the conditions of the various
lemmas in Section 2.4 are met. The critical point is to show that certain invariants
(see Claim 3.5 below) are maintained.

Proof of Lemma 3.1. We require
\[ \rho < \mu \leq \frac{1}{100\kappa \Delta \delta} , \quad \epsilon' \leq \frac{\mu(\delta d - \epsilon)\epsilon}{1000\kappa \Delta^2 \delta^2 \Delta \delta} , \quad \epsilon \leq \frac{\epsilon'}{\kappa d} \]
and
\[ p \geq 10r_1 \epsilon^{-1} \left( \frac{\log n}{n} \right)^{1/\Delta} , \quad n > \kappa r_1^2 . \]

First, we apply Lemma 2.26 to obtain an ordering \( \tau \) of the vertices of \( H \) with the stated properties \((a)-(c)\).

Given this ordering \( \tau \) we run the random greedy algorithm, Algorithm 1, and in the following we show that \( \psi_{RGA} = \psi_{RGA_{\text{main}}} \) satisfying properties \((RGA1)-(RGA5)\) of Lemma 3.1.

We first show that some invariants are maintained (deterministically) throughout the algorithm. Recall that \( \pi_t(x) := \pi_t(x) + |J_x| \), where \( \pi_t(x) \) denotes the number of the embedded neighbours of \( x \) at time \( t \). We could obtain a stronger statement here than \((INV3)\) below, but we will not need it, so we give this statement in order to be able to re-use the same invariants in the proofs of the other two blow-up lemmas.

Claim 3.5. The following hold at each time \( t \) in the running of Algorithm 1.

\( INV1 \) \( \psi_t \) is a good partial embedding.

\( INV2 \) For each \( x \in X_{\text{main}} \setminus \text{Dom}(\psi_t) \), either \( x \in Q_t \) or we have
\[ |A_t(\text{main})(x)| \geq \frac{1}{4} \mu(d - \epsilon')\pi_t(x)^2 |V_{\text{main}}(x)| . \]

\( INV3 \) When we embed \( x \) to create \( \psi_{t+1} \), we do so uniformly at random into a set of size at least
\[ \frac{1}{1000} \mu |dp| \pi_t(x)^2 |V(x)| . \]

Proof. The invariant \((INV1)\) is maintained by definition of \( A_t(x) \) and \( B_t(x) \). Observe that immediately before reaching the Repeat line, the queue is updated by adding precisely any vertices which would fail \((INV2)\), so that this invariant too holds by definition. Finally, by choice of \( \epsilon' \) and by \((INV2)\), if the vertex \( x \) is
embedded to create $\psi_{t+1}$, then we have $|A_{t+1}^{\text{main}}(x)| \geq \frac{1}{4} \mu(dp)\tau^*(x)|V(x)|$. Now, if all neighbours of $x$ are embedded under $\psi_t$, then by (a) we have (INV3), while otherwise we apply Lemma 2.25(b) with $D = \Delta$, for which the required assumptions on $\Gamma$ hold, to bound $|B_t(x)|$. By choice of $\varepsilon'$ we conclude that (INV3) holds.

We now begin to verify the properties of Lemma 3.1 are met. Property (RGA 2), stating that all vertices in $X_t \cap \text{Dom}(\psi_{\text{RGA}})$ are embedded to $V_t^{\text{main}}$, holds by the definition of the algorithm.

**Claim 3.6.** Property (RGA 1) holds. Moreover, let $x$ be a buffer vertex and $y$ its neighbour whose embedding creates $\psi_{t+1}$, then

$$|A_{t+1}^{\text{main}}(y)| \geq \frac{3}{4} (dp - \varepsilon' p)\tau^*(y)|V(y)|.$$  \hspace{1cm} (20)

**Proof.** We require $\mu \leq 1/(100\kappa\Delta_R^\gamma)$. To show property (RGA 1), i.e., that all neighbours of buffer vertices are in $\text{Dom}(\psi_{\text{RGA}})$, it is sufficient to prove that none of the neighbours of buffer vertices are in $Q_{\text{RGAend}}$. This, however, is clear. Indeed, suppose that $y_1, \ldots, y_b$ with $b \leq \Delta$ are the neighbours of a buffer vertex $x$ appearing in this order in $\tau$. Suppose the embedding of $y_1$ creates $\psi_{t+1}$. By (H4), buffer vertices are at distance at least five in $H$, hence, the neighbours of the vertices have distance at least three, and by (H5) they are not image restricted. Therefore, at time $t$, the available candidate set of each $y_j$ is $A_{t+1}^{\text{main}}(y_j) = V_t^{\text{main}}(y_j) \setminus \text{Im}(\psi_t)$. The size of $A_{t+1}^{\text{main}}(y_j)$ is by (H3), (G1) and choice of $\mu$ at least

$$|V_t^{\text{main}}(y_j)| - 4\kappa\Delta_R\mu|V(y_j)| = (1 - 3\mu - 4\kappa\Delta_R\mu)|V(y_j)| \geq \frac{3}{4} |V(y_j)|.$$  

It follows that $y_j$ is not added to $Q_t$ for any $t \leq t_1$. Now the vertices $y_1, \ldots, y_b$ are embedded consecutively, and since they are neighbours of $x$, by (H2), into distinct clusters of $G$. By Definition 2.24 (of bad vertices with respect to $\psi$ and $Q$), for each time $t'$ with $t_1 \leq t' \leq t$, where $y_j$ is embedded to create $\psi_{t+1}$, we have

$$|A_{t'}^{\text{main}}(y_j)| \geq \frac{3}{4} (dp - \varepsilon' p)\tau^*(y_j)|V(y_j)|,$$

giving (20) and that $y_j$ never enters $Q_t$, as desired.

Observe that because Algorithm 1 preserves (INV3), the conditions of Lemma 2.27 are met. Thus with probability at least $1 - 2^{-n/(\kappa\tau^*)}$, the following event $\mathcal{E}_{1,2.27}$ holds. For every $i \in [r]$ and $W \subseteq V_i$ with $|W| \geq g|V_i|$, the number of vertices $x \in X_t \setminus X^*_t$ such that for some time $t$ (at which $x$ is unembedded) we have $|C_t(x) \cap W| < (dp - \varepsilon' p)\tau^*(x)|W|$ is at most $g|X_i|$.

Suppose that $\mathcal{E}_{1,2.27}$ holds. Then properties (RGA 3) and (RGA 4) hold. To see that (RGA 3) holds, set $W := V_t^{\text{main}} \setminus \text{Im}(\psi_{\text{RGAend}})$. By (H3) and (G1), we have $|X_t^{\text{main}}| \leq (1 - \mu)|X_i|$ and $|V_t^{\text{main}}| = (1 - 3\mu)|V_i|$. We conclude $|W| \geq \mu|V_i|$. Suppose that $x \in Q_{\text{RGAend}}$. Then there is a first time $t$ at which $x \in Q_t$. Since we have $A_{t+1}^{\text{main}}(x) \supseteq C_t(x) \cap W$, by the construction of $Q_t$ in Algorithm 1, we have

$$|C_t(x) \cap W| < \frac{1}{2} \mu(dp - \varepsilon' p)\tau^*(x)|V_t^{\text{main}}(x)| < (dp - \varepsilon')\tau^*(x)|W|$$

so that $x$ satisfies condition (9) of Lemma 2.27. Since $|W| > g|V_i|$, and because $\mathcal{E}_{1,2.27}$ holds, the number of $x \in X_t^{\text{main}} \cap Q_{\text{RGAend}}$ which are in $X_t \setminus X_t^*$ is at most $g|X_i|$. By (G6), we have $|X_t^*| \leq g|X_i|$, so $|X_t^{\text{main}} \cap Q_{\text{RGAend}}| \leq g|X_i| + |X_t^*| \leq 2g|X_i|$. By definition of Algorithm 1, the vertices $X_t^{\text{main}} \cap Q_{\text{RGAend}}$ are precisely the vertices of $X_t^{\text{main}}$ not in $\text{Dom}(\psi_{\text{RGAend}})$, giving (RGA 3) as desired.

Property (RGA 4) follows from $\mathcal{E}_{1,2.27}$ by observing that $X_t^{\text{buf}} \subseteq X_t$ and that when $X_t^{\text{buf}}$ is a degree-$b$ buffer we have $(dp - \varepsilon' p)\tau_{\text{RGAend}}^*(x) \geq (dp)^b/2$ by choice of $\varepsilon'$.

To complete the proof of Lemma 3.1 we now verify that a.a.s. after Algorithm 1 finishes, the property (RGA 5) is satisfied. By (a) and (c) of Lemma 2.26, the set
of vertices in $N(X^\text{buf})$ which are not neighbours of clique buffer vertices is an initial segment of $T$ vertices of $\tau$. We first show that Lemma 2.28 applies to the sequence of good partial embeddings produced by Algorithm 1 on this initial segment. Observe that by Claim 3.6, all vertices in this initial segment satisfy (20) and none enters the queue. By (20), Lemma 2.25, and the choice of $\varepsilon'$, each such $y$, embedded to create $v_{r+1}$, is embedded into a subset of $C_i(y)$ of size at least

$$\frac{2}{5}(dp - \varepsilon'p)^{\pi_i}|V(y)| - 20\Delta^2\varepsilon'p^{\pi_i}|V(y)| > \frac{1}{10}(dp)^{\pi_i(y)}|V(y)|.$$ 

Furthermore, for each such $y$ we have $\pi_i(y) \leq \Delta - 2$ by Lemma 2.26(b). Finally, by (H3) we have $|\text{Dom}(\psi_\tau) \cap X_i| \leq 4\mu_k\Delta R_i|X_i|$ for each $i \in [r]$. This justifies that we can apply Lemma 2.28 with $V$ vertex $\tau$, and since $q_{\text{main}} - q_{\text{buf}}$ is the set of bad vertices for $\tau$, specifically, which would violate property $\text{CON}(\phi)$, therefore try to find a system of distinct representatives $Q$ is the set of bad vertices for $\tau$, in the order $x_1 \leq x_2 \leq \ldots \leq x_q$. Let $H$ be a graph with $\Delta = \Delta_0$, $\deg(x) = \deg(y)$ for each $x, y \in V$, and $\deg(x)$ is embedded in $\psi_\tau(y)$.

Thus, by Lemma 2.29, the probability that $y_j \in E(R')$, we have $|N_G(v; V_{\text{main}}) \setminus \text{Im}(\psi_\tau)| \geq \frac{1}{2} \deg_G(v; V_j)$. We assume from now on that $X^\text{buf}$ is a non-clique degree-$b$ buffer, and we fix a vertex $v \in V_i$. We would like to estimate the probability that (RGA 5) fails for $v$.

To that end, first fix $x \in X^\text{buf}$, and let $y_1, \ldots, y_b$ be an enumeration of $N_H(x)$ in the order $\tau$. We now justify that if $E_{1,2,5}$ holds, then we are in a position to apply Lemma 2.29 with $B = \Delta$. Recall that no vertices of $N(X^\text{buf})$ enter the queue, by Claim 3.6, so that $\psi_\tau(y_i)$ is the good partial embedding created by the embedding of $y_i$ (and so on). By (H4), and since Dom($\psi_\tau(y_i-1)$) $\subseteq N(X^\text{buf})$, no vertices at distance two or less from $x$ in $H$ are embedded in $\psi_\tau(y_i-1)$. Note that Lemma 2.26(b) states that if $b = \Delta$ then $y_b-1 y_b$ is not an edge of $H$.

Algorithm 1 creates the sequence $\psi_\tau(y_1)-1, \ldots, \psi_\tau(y_b)$ of good partial embeddings, and the sequence of queue sets $Q_{\tau(y_1)-1}, \ldots, Q_{\tau(y_b)}$, according to the requirements of Lemma 2.29. Thus, by Lemma 2.29, the probability that $N_H(x)$ is embedded to $N_G(v)$, conditioning on $\psi_\tau(y_i)$ and that $|N_G(v; V_{\text{main}}) \setminus \text{Im}(\psi_\tau(y_i-1))| \geq \frac{1}{2} \deg_G(v; V_j)$ for each $j$ such that $y_j \in E(R')$, is at least $(d^2p/10)^b$.

Let $x_1, \ldots, x_m$ be an enumeration of $X^\text{buf}$ according to the order on $N(X^\text{buf})$ given by $\tau$. Let $Y_1, \ldots, Y_m$ be Bernoulli random variables with $Y_j = 1$ if either $N_H(x_j)$ is embedded by $\psi_{\text{RGA,main}}$ to $N_G(v)$ or we witness a failure of $E_{1,2,5}$ before the first neighbour of $x_j$ is embedded. Then we have just shown that $Y_j$ has probability at least $(d^2p/10)^b$ of being one, conditioned on the history up to, but not including, the embedding of the first vertex of $N_H(x_j)$. This history determines $Y_{j-1}$, so we can apply Lemma 2.2, with $x = 4\mu|X_i|(d^2p/10)^b$ and $\delta = \frac{1}{2}$, to conclude that $Y_1 + \cdots + Y_m \geq 2\mu|X_i|(d^2p/10)^b$ with probability at least $1 - \exp(-\mu|X_i|(d^2p/10)^b/3)$. Taking the union bound over all $v \in V(G)$ we see that with probability at least $1 - n\exp(-\varepsilon_p^2n/r_1) > 1 - 1/n$ (where the inequality is by choice of $p$ and since $b \leq \Delta$), either we witness a failure of $E_{1,2,5}$, or (RGA 5) holds.

Putting together the three probability bounds, we conclude that a.a.s. the events $E_{1,2,7}, E_{1,2,8}$ and (RGA 5) all hold, proving Lemma 3.1.

### 3.3. Queue embedding

In this section we prove Lemma 3.2. The idea is as follows. For each $i \in [r]$ in succession, we embed $X_i^\text{buf}$ into $V_n^\text{buf}$, maintaining a good partial embedding. The way we do this is as follows. We need to embed each $x \in X_i^\text{buf}$ into $C_i(x) \setminus B(x)$, where $B(x)$ is the set of bad vertices for $x$ with respect to the current good partial embedding and $Q = V(H)$. (This choice of $Q$ is made only so that we are able to apply Lemma 2.25 without having to verify the at this time pointless badness condition (8).) We therefore try to find a system of distinct representatives $v_x \in C_i(x) \setminus B(x)$ for each $x \in X_i^\text{buf}$, which we do by verifying Hall’s condition. In turn, we verify Hall’s condition by showing that its failure implies the existence of a ‘dense spot’ in $G$, which is too dense for $\Gamma$, specifically, which would violate property $\text{CON}(\rho, r_1, \Delta)$ which $\Gamma$
satisfies. At this point the reader should be concerned that embedding \( x' \in X_i^q \) to \( \psi_{x'} \) could change \( B(x) \)—but observe that since \( x \) and \( x' \) are at distance at least ten in \( H \), this does not occur.

**Proof of Lemma 3.2.** We require

\[
\mu \leq \frac{1}{6}, \quad \theta \leq \frac{\mu \zeta d^\Delta}{2000k \Delta}, \quad \epsilon' \leq \frac{\mu \zeta d^\Delta}{100004^3 \Delta^2} \quad \text{and} \quad \epsilon \leq \frac{\epsilon'}{kd}.
\]

Let \( \psi_i \) be a good partial embedding whose image is disjoint from \( V^q \), and suppose that for each \( i \in [r] \) the set \( X_i^q \) has size at most \( 2\theta |X_i| \). We now define a sequence of good partial embeddings \( \psi_1, \ldots, \psi_r \) with \( \text{Dom}(\psi_i) = \text{Dom}(\psi_0) \cup \bigcup_{j=1}^{i} X_j^q \) and \( \psi_i(x) \in V_j^q \) for each \( x \in X_j^q \) and \( 1 \leq j \leq t \). We let \( C_i(x) \) be the candidate set of \( x \) with respect to \( \psi_i \), and so on, and let \( B_i(x) \) be the set of bad vertices for \( x \) with respect to \( \psi_i \) and \( Q = V(H) \).

Suppose that for some \( 1 \leq t \leq r \) we have constructed \( \psi_{t-1} \) as above. We let \( \{x_i\}_{x \in X_t^q} \) be a system of distinct representatives for the sets \( \{ C_i^q \setminus B_i^q \}_{x \in X_t^q} \), and we set \( \psi_t = \psi_{t-1} \cup \{ x \rightarrow \psi_i : x \in X_t^q \} \). We need to prove that this system of distinct representatives exists, and that the resulting \( \psi_t \) is a good partial embedding.

To see that the system of distinct representatives exists, we verify Hall’s condition. Let \( X \subseteq X_t^q \) be non-empty, and let \( U = \bigcup_{x \in X} C_i^q \setminus B_i^q \). Then we need to show \( |U| \geq |X| \). Assume for a contradiction that \( |U| < |X| \) holds. By averaging, there is \( b \in \{0, \ldots, \Delta\} \) such that we find a subset \( X_b \) of (not necessarily all) vertices \( x \) in \( X \) with \( \pi_{t-1}(x) = b \) of size exactly \( \frac{1}{2^\Delta + 1}|U| \). Each \( x \in X_b \) has \( C_i^q(x) \geq (1 - \epsilon') \mu(dp - \epsilon') \delta_j \cdot ||I_x||V_i|| \) by \((GPE3)\), and \( |I_x| \geq \zeta(dp - \epsilon') \delta_j \cdot ||V_i|| \) by \((G5)\). We conclude, by Lemma 2.25, that

\[
|U^q \setminus B_i^q(x)| \geq \frac{1}{2} \mu \zeta \delta_j dp ||V_i|| - 20 \Delta^2 \epsilon' \delta_j ||V_i|| \geq \frac{1}{2} \mu \zeta \delta_j dp ||V_i||.
\]

by choice of \( \epsilon' \). In particular, we see that for each \( x \in X_b \) we have \( |U^q \setminus B_i^q(x) \cap U| \geq \frac{1}{2} \mu \zeta \delta_j dp ||V_i|| \). We therefore have

\[
\sum_{x \in X_b} |U^q \setminus B_i^q(x) \cap U| \geq |X_b| \cdot \frac{1}{2} \mu \zeta \delta_j dp ||V_i||.
\]

Since we would like to resort to the congestion property \( \text{CON}(\theta, r_1, \Delta) \) we eventually need to pass to a subset \( U' \) of \( U \) of size \( |X_b| \). Picking uniformly at random a subset \( U' \) of \( U \) of size \( |X_b| = \frac{|U|}{2^\Delta + 1} \), we have

\[
\mathbb{E}_{U'} \sum_{x \in X_b} |U^q \setminus B_i^q(x) \cap U'| \geq \frac{|X_b|}{2^\Delta + 1} \cdot \frac{1}{2} \mu \zeta \delta_j dp ||V(x)||,
\]

and so in particular there is a subset \( U' \) of \( U \) of size \( |X_b| \) such that

\[
\sum_{x \in X_b} |U^q \setminus B_i^q(x) \cap U'| \geq \frac{|X_b|}{2^\Delta + 1} \cdot \frac{1}{2} \mu \zeta \delta_j dp ||V_i||.
\]

We now apply \( \text{CON}(\theta, r_1, \Delta) \) with \( F = \{ \psi_{t-1}(\Pi_{t-1}(x) \cup J_x) : x \in X_b \} \) and the set \( U' \). Note that \( |U'| = |F| \leq 2|\Gamma||X_b|/(\Delta + 1) \leq \theta |V(\Gamma)| \), so that we can do this. We conclude that

\[
\epsilon(CG(\Gamma, U', F)) \leq 7 \delta_j ||U'|| ||F|| + \theta \delta_j n ||F|| r_1.
\]

But the edges of this congestion graph are precisely the pairs \( xu \) such that \( x \in X_b \) and \( u \in U^q \setminus B_i^q(x) \cap U' \), so we have

\[
\frac{1}{2^\Delta + 1} \mu \zeta \delta_j dp ||V_i|| \leq 14 \theta \delta_j ||X_b|| ||V_i|| + \theta \epsilon \delta_j ||X_b|| ||V_i||,
\]

where we use \( |X_b| = |U'| \leq 2\theta |V_i|| ||V_i|| \leq n/(\kappa r_1) \). This is a contradiction by choice of \( \theta \). We conclude that the desired system of distinct representatives exists.
Now we show that \( v_\psi \) is a good partial embedding. Since the \( v_x \) are distinct, \( v_\psi \) is injective. Since \( C_{t-1}^q(x) \subseteq I_x \) for each \( x \), we have \((GPE1)\) for each \( x \in X_t^q\). If \((GPE3)\) or \((GPE2)\) were to fail for some \( y \), then \( y \in N_H(x) \) for some \( x \in X_t^q\). Since by \((H2)\), vertices of \( X_t^q \) are at distance at least 10 in \( H \), this \( x \) is unique. But by definition of \( B_{t-1}(x) \) the vertex \( v_x \) is not bad for \( x \) with respect to \( v_{t-1} \), i.e. this case does not occur. Finally, if \((GPE4)\) fails for some \( yz \in E(H) \), then again at least one of \( y \) and \( z \) (possibly both) is a neighbour of some \( x \in X_t^q\). Again by \((H2)\) this \( x \) is unique, and again by definition of \( B_{t-1}(x) \), \( v_x \) is not bad for \( x \) with respect to \( v_{t-1} \), so this case too does not occur. Thus \( v_\psi \) is a good partial embedding as desired.

By induction on \( t \), the final \( v_t := v_\psi \) is a good partial embedding satisfying the conclusion of Lemma 3.2.

### 3.4. Fixing buffer defects

To prove Lemma 3.3 we need to describe how we embed the reserved cliques. The basic idea is as follows: if \( X_{3i}^\text{buf} \) is a clique buffer, and a vertex \( v \) of \( V_i \) is a candidate for too few vertices in \( X_{3i}^\text{buf} \) and not in the image of \( \psi \) (we call such a vertex ‘poor’), then we will embed a clique \( K \) using \( v \) and some further vertices of \( V^* \). The only difficulty is that some of these ‘poor’ vertices \( v \) may lie in \( V_{3i}^\text{buf} \) and we risk destroying the hard-earned property that every vertex \( X_{3i}^\text{buf} \) has many candidates. In order to deal with this, we will need to embed some vertices \( X_{3i}^\text{buf} \) as well—at which point further vertices of \( V_i \) may become ‘poor’ and require embedding, and so on.

Before we prove Lemma 3.3 we state two auxiliary lemmas. The first justifies that this ‘and so on’ terminates without eating up too many vertices, i.e. that we can find for each \( i \) small subsets \( P_i \) of \( V_i \) and \( D_i \) of \( X_{3i}^\text{buf} \) such that every vertex \( x \in X_{3i}^\text{buf} \setminus D_i \) has many candidates in \( V_{3i}^\text{buf} \setminus P_i \), and every vertex of \( V_i \setminus P_i \) is a candidate for many vertices of \( X_{3i}^\text{buf} \setminus D_i \).

**Lemma 3.7.** We assume the General Setup. Suppose that \( g < \mu d^\Delta/(250\kappa) \) and that \( \Gamma \) has \( \text{CON}(g, \kappa_1, \Delta) \). Suppose \( \psi \) is a good partial embedding, and \( X_{3i}^\text{buf} \) is a clique buffer. Suppose furthermore that all but at most \( g|V_i| \) vertices of \( V_i \) are candidates for at least \( \mu(dp)^\Delta|X_i|/2 \) vertices of \( X_{3i}^\text{buf} \). Then there are subsets \( P_i \) of \( V_i \) and \( D_i \) of \( X_{3i}^\text{buf} \) with the following properties.

(\( IB1 \)) We have \( |P_i| < 2g|V_i| \) and \( |D_i| < 2g|X_i| \).

(\( IB2 \)) Each vertex of \( X_{3i}^\text{buf} \setminus D_i \) has at least \( \mu(dp)^\Delta|V_i|/4 \) candidates in \( V_{3i}^\text{buf} \setminus P_i \).

(\( IB3 \)) Each vertex of \( V_i \setminus P_i \) is a candidate for at least \( \mu(dp/100)^\Delta|X_i| \) of the vertices \( X_{3i}^\text{buf} \setminus D_i \).

(\( IB4 \)) Each vertex in \( D_i \) has at least \( \mu(dp)^\Delta|V_i|/4 \) candidates in \( P_i \cap V_{3i}^\text{buf} \).

The conclusion \((IB4)\) will be used to show that we can embed \( D_i \) into \( P_i \cap V_{3i}^\text{buf} \).

**Proof of Lemma 3.7.** We start with \( D_i = \emptyset \) and \( P_i \) being the set of vertices which fail \((IB3)\). We sequentially add vertices to \( D_i \) and \( P_i \) which fail \((IB2)\) and \((IB3)\) respectively (note that the property of failing either condition is monotone), until either there are no failing vertices to add to either set or one of the two sets reaches \( 2g|V_i| \) vertices (or more). Observe that any vertex \( x \) of \( D_i \) has by \((GPE3)\) at least \( (1 - \varepsilon')\mu(d - \varepsilon')^\Delta p^\Delta|V_i| \) candidates in \( V_{3i}^\text{buf} \). Since \( x \) must have violated \((IB2)\) at some point, by choice of \( \varepsilon' \) it therefore has at least \( \mu(dp)^\Delta|V_i|/4 \) candidates in \( P_i \cap V_{3i}^\text{buf} \), establishing \((IB4)\).

In the process of building \( D_i \) and \( P_i \) we call a vertex \( v \in V_i \setminus P_i \) poor if \( v \) is a candidate for less than \( \mu(dp/100)^\Delta|X_i| \) vertices of \( X_{3i}^\text{buf} \setminus D_i \). If \( P_i \) reaches \( 2g|V_i| \) vertices first, let \( P \) be a subset of \( P_i \) and \( D \) a superset of \( D_i \), each of size exactly \( 2g|V_i| \). Since at most \( g|V_i| \) vertices in \( V_i \) are poor with respect to \( X_{3i}^\text{buf} \), at least \( g|V_i| \) of the
vertices in $P$ were added to $P_1$ because they became poor; i.e., they are candidate for at least $\mu(dp)\Delta|X_i|/2$ vertices of $X_i^\text{buf}$, but for less than $\mu(dp/100)\Delta|X_i|$ of the vertices $X_i^\text{buf}\setminus D_i$. It follows that each candidate for at least $\mu(dp)\Delta|X_i|/4$ vertices of $D_i$, and so there are at least $g[V_i]\mu(dp)\Delta|X_i|/4$ edges in the candidate graph between $P$ and $D$. Now

$$\frac{1}{2}g\mu(dp)\Delta|X_i||V_i| = \frac{1}{8}g\mu(dp)\Delta|X_i||V_i| + \frac{1}{8}g\mu(dp)\Delta|X_i||V_i|$$

$$> \frac{\mu\Delta}{8\kappa_1}p\Delta|P||D| + \frac{\mu\Delta}{8\kappa_1}p\Delta n|D|$$

$$> 7p^2\Delta|P||D| + gp^2 n|D|/r_1,$$

where the first inequality comes from the sizes of $P$ and $D$, and the second from the choice of $g$ sufficiently small. The last line is in contradiction to the congestion condition $\text{CON}(\varepsilon, r_1, \Delta)$, since the candidate graph between $P$ and $D$ is a subgraph of $\text{CG}(\Gamma, P, F_D)$ with $F_D := \{\psi(N_H(x)) : x \in D\}$.

If $D_i$ reaches $2g|V_i|$ vertices first, we define similarly $P$ a superset of $P_1$ and $D$ a subset of $D_i$, each of size $2g|V_i|$. Each vertex of $D$ has at least $\mu(dp)\Delta|V_i|/4$ candidates in $P$ by (IB4), and thus the candidate graph between $P$ and $D$ has at least $2g|V_i|\mu(dp)\Delta|V_i|/4$ edges, which is larger than the previous case and so also gives a contradiction to $\text{CON}(\varepsilon, r_1, \Delta)$. This establishes (IB1).

We conclude that the process terminates for lack of failing vertices, i.e. with the desired sets. □

The second lemma simply gives a condition on a vertex $v \in V_i$ under which we can find a clique $K_{\Delta+1}$ containing $v$ in $G$ whose remaining vertices lie in $V^c\setminus\text{Im}(\psi)$.

**Lemma 3.8.** We assume the General Setup. Suppose that $\varepsilon < \varepsilon' < (d/32)\mu/(2k\Delta^2)$, and that $\Gamma$ has $\text{NS}(\varepsilon, r_1, \Delta)$, $\text{RI}(\varepsilon, (\varepsilon, d, \rho), \varepsilon', d, r_1, \Delta)$ and $\text{CON}(\varepsilon, r_1, \Delta)$. Let $i, j_1, \ldots, j_\Delta$ form a copy of $K_{\Delta+1}$ in $R'$ and $v$ be a vertex of $V_i$. Suppose that $\psi$ is any good partial embedding in which $v$ is unembedded, and such that $v$ has at least

$$\frac{\mu d}{2} \max (p|V_j^c|, \deg_{\Gamma}(v; V_j^c))$$

(21) neighbours in $V^c_{j_k}\setminus\text{Im}(\psi)$ in $G$ for each $1 \leq k \leq \Delta$. Suppose that there is a copy of $K_{\Delta+1}$ in $H$ using a vertex of $X_i$ and a vertex of each of $X_{j_1}, \ldots, X_{j_\Delta}$. Then there are vertices $v_k \in V^c_{j_k}\setminus\text{Im}(\psi)$ for $1 \leq k \leq \Delta$, such that $v, v_1, \ldots, v_\Delta$ form a copy of $K_{\Delta+1}$ in $G$.

**Proof.** We choose the vertices $v_1, \ldots, v_\Delta$ in succession. When we choose $v_s$, if $s < \Delta$ we require that for each $s + 1 < k \leq \Delta$ we have

$$|N^c_{\Gamma}(v, v_1, \ldots, v_s) \cap V_{j_k}| = (p \pm \varepsilon p)^s \deg_{\Gamma}(v; V_{j_k})$$

and

$$|N^c_{\Gamma}(v, v_1, \ldots, v_s) \cap (V^c_{j_k}\setminus\text{Im}(\psi))| \geq (\frac{dp}{4})^s \max (p|V_{j_k}|, \deg_{\Gamma}(v; V_{j_k})) .$$

If $s < \Delta - 1$ we further require that for each $s + 1 \leq k < k' \leq \Delta$ the pair

$$|N^c_{\Gamma}(v, v_s, \ldots, v_{s+1}) \cap V_{j_k}, N^c_{\Gamma}(v, v_1, \ldots, v_s) \cap V_{j_{k'}}|$$

(24)

is $(\varepsilon_s, d, p)$-regular.

Note that these conditions are satisfied, with $s = 0$, before we have chosen any vertices. The first is a tautology, the second is the assumption on $v$ in the lemma statement, and the third is a statement that $V_i$ has two-sided inheritance with respect to $V_{j_k}$ and $V_{j_{k'}}$, which holds since we assumed there is a triangle of $X$ using one vertex of each of $X_i$, $X_{j_k}$ and $j_{k'}$, and by (H1) and (G2).

Suppose we have chosen vertices $v_1, \ldots, v_{s-1}$ so far and that we have $s \leq \Delta - 1$. Because $\Gamma$ has $\text{NS}(\varepsilon, r_1, \Delta)$, at step $s$ the number of vertices $w$ failing the first
condition, i.e. condition (22) with \( v_s = w \), is at most
\[
\Delta \varepsilon p^{\Delta - 1} |V(\Gamma)| / r_1^2 < \left( \frac{d\varepsilon}{4} \right)^{s-1} \frac{\mu_d}{12} \max (p|V_{J^*}|, \deg_{\Gamma}(v; V_{J^*})).
\]
By \((\varepsilon_{s-1,s-1}, d, p)\)-regularity of the pair
\[
(N^*_p(v, v_1, \ldots, v_{s-1}) \cap V_{J^*}, N^*_p(v, v_1, \ldots, v_{s-1}) \cap V_{J^*}),
\]
and using an upper bound (22) on \( N^*_p(v, v_1, \ldots, v_{s-1}) \cap V_{J^*} \), at most
\[
\Delta \varepsilon (p + \varepsilon p)^{s-1} \deg_{\Gamma}(v; V_{J^*}) < \left( \frac{d\varepsilon}{4} \right)^{s-1} \frac{\mu_d}{6} \max (p|V_{J^*}|, \deg_{\Gamma}(v; V_{J^*}))
\]
vertices \( w \in V^c \setminus \Im(\psi) \) fail the second condition (condition (23) with \( v_s = w \)). Finally, if \( s \leq \Delta - 2 \) and since \( \Gamma \) has \( RI(\varepsilon, (\varepsilon_{s,b}), \varepsilon', d, r_1, \Delta) \), at most
\[
\Delta^2 \varepsilon p^{\Delta - 2} |V(\Gamma)| / r_1^2 < \left( \frac{d\varepsilon}{4} \right)^{s-1} \frac{\mu_d}{32} \max (p|V_{J^*}|, \deg_{\Gamma}(v; V_{J^*}))
\]
vertices fail the third condition, condition (24). Thus there are at least
\[
\left( \frac{d\varepsilon}{4} \right)^{s-1} \frac{\mu_d}{32} \max (p|V_{J^*}|, \deg_{\Gamma}(v; V_{J^*})) > 0
\]
vertices which satisfy all three conditions, and we can choose \( v_s \) for each \( 1 \leq s \leq \Delta \) as desired.

Furthermore observe that for \( s = \Delta \) it suffices to choose an arbitrary vertex from \( N^*_G(v, v_1, \ldots, v_{\Delta - 1}) \cap (V^c_\psi \setminus \Im(\psi)) \). The lemma follows.

We can now prove Lemma 3.3. We construct the desired embedding in two steps. First, we embed each \( D_i \) into \( P_i \cap V^b_i \), which we can do by verifying Hall's condition using \((IB4)\) and the congestion condition \( CON(g, r_1, \Delta) \). We then have to cover the remaining poor vertices using reserved cliques, and embed any left-over reserved cliques at the end of this process. We do this by sequentially by applying Lemma 3.8 to find a destination for a reserved clique, in the first case letting \( v \) be the next poor vertex and in the second case choosing \( v \) to be some vertex of \( V^c \).

Note that in this step, we need to maintain the degree condition (21) of Lemma 3.8. We have this condition initially by \((G3)\), and we will see that to maintain it, it suffices to choose at each step a 'most dangerous' vertex \( v \), that is, one minimising the parameter mindeg we define in the proof.

**Proof of Lemma 3.3.** We assume that \( \Gamma \) has \( NS(\varepsilon, r_1, \Delta), RI(\varepsilon, (\varepsilon_{a,b}), \varepsilon', d, r_1, \Delta) \) and \( CON(g, r_1, \Delta) \). We further assume
\[
\theta < \min \left( \frac{\mu_d \Delta}{250\varepsilon} , \frac{\mu_d}{64\varepsilon(|\Delta R'| + 1)} \right) \quad \text{and} \quad \varepsilon < \varepsilon' < \frac{d\Delta}{32\varepsilon |\Delta \Delta R'|}.
\]

For each \( i \) such that \( X^b_i \) is a clique buffer, we let \( P_i \) and \( D_i \) be as given by Lemma 3.7 (the assumptions on constants above imply the requirements of that lemma and of Lemma 3.8). If \( X^b_i \) is not a clique buffer, then we let \( P_i = D_i = \emptyset \).

Now for each \( v \) in succession, we find a matching in the candidate graph between \( D_i \) and \( P_i \cap V^b_i \). To do this we simply verify Hall's condition, using the congestion condition and \((IB4)\). Specifically, suppose we have a non-empty \( W \subseteq D_i \), and \( Z \subseteq P_i \) is the set of vertices in \( P_i \) which are candidate for some \( w \in W \). We need to verify that \(|W| \leq |Z| \). If this were false, then we could let \( Z' \) be a superset of \( Z \), and the number of edges in the candidate graph between \( W \) and \( Z' \) is by \((IB4)\) at least \( \mu(dp)^{\Delta} |V_i||W|/4 \). Notice that \(|D_i| \leq 2|P_i|V_i| \) by \((IB1)\). Now we have
\[
\frac{1}{2} \mu(dp)^{\Delta} |V_i||W| = \frac{1}{2} \mu(dp)^{\Delta} |V_i||W| + \frac{1}{2} \mu(dp)^{\Delta} |V_i||W|
\]
\[
> \frac{\mu(d\varepsilon}{16\varepsilon} \mu(dp)^{\Delta} |Z'||W| + \frac{\mu(d\varepsilon}{8\varepsilon} dp^{\Delta} n||W|
\]
\[
> 7dp^{\Delta} |Z'||W| + g\mu(dp)^{\Delta} |W| / r_1,
\]
which is in contradiction to $\text{CON}(\rho, r_1, \Delta)$, applied with $Z'$ and the family $F = \{\psi(N_H(w)) : w \in W\}$ to the congestion graph $CG(\Gamma, Z', F)$. We conclude that Hall’s condition holds and hence the desired matching exists. We let $\psi'$ be the embedding obtained by embedding each vertex of $D_i$ to its matching partner in $P_i \cap V_i^{\text{out}}$. Observe that $\psi'$ is a good partial embedding since all neighbours of the buffer vertices are already embedded under $\psi$, and thus $(\text{GPE}1)$–$(\text{GPE}4)$ trivially hold for $\psi'$.

Finally, we provide an algorithm below to embed the cliques $K_i$ for each $i \in [r_1]$. Given a vertex $v \in V_i$ and a partial embedding $\psi$ we define the parameter

$$\mindeg_{\psi}(v) := \min_{j:i \in R'} \deg_{G'}(v; V_j^c \setminus \text{Im}(\psi)) = \max (p|V_j|, \deg_{G'}(v; V_j)),$$

which we will use in the algorithm below, Algorithm 2. Observe that we have initially $\mindeg_{\psi}(v) \geq \mu(d-\varepsilon)/2$ for all $v \in V_i$ with $\deg_{R'}(i) > 0$, which follows from $(G3)$.

**Algorithm 2: Removing poor vertices**

$t := 0$; $\psi_0 := \psi'$;

while $\bigcup P_i \setminus \text{Im} (\psi_t) \neq \emptyset$ do

choose $v \in \bigcup P_i \setminus \text{Im} (\psi_t)$ minimising $\mindeg_{\psi_t}(v)$;

set $i$ such that $v \in V_i$;

choose an unembedded clique $K_H$ in $K_i$;

choose a clique $K_G$ containing $v$ and one vertex from each $V_j^c \setminus \text{Im}(\psi_t)$ such that $X_j$ intersects $K_H$ and $j \neq i$;

set $\psi_{t+1} := \psi_t \cup \{K_H \rightarrow K_G\}$;

$t := t + 1$;

end

while $\bigcup K_i \neq \emptyset$ do

choose an unembedded clique $K_H$ in $\bigcup K_i$;

set $i$ such that $K_H \in K_i$;

choose $v \in \bigcup V_i^c \setminus \text{Im}(\psi_t)$ minimising $\mindeg_{\psi_t}(v)$;

choose a clique $K_G$ containing $v$ and one vertex from each $V_j^c \setminus \text{Im}(\psi_t)$ such that $X_j$ intersects $K_H$ and $j \neq i$;

set $\psi_{t+1} := \psi_t \cup \{K_H \rightarrow K_G\}$;

$t := t + 1$;

end

return $\psi_{\text{good}} := \psi_t$;

We claim that the algorithm runs correctly and the finally returned $\psi_{\text{good}}$ is the desired good partial embedding. It remains to justify this claim.

Roughly speaking each of the vertices from $P_i$ can serve us as an image for each of the $2\rho|X_i|$ cliques from $K_i$ to embed the corresponding vertex into (although some vertices of $P_i$ may be used as ‘intermediate’ ones). This is done during the first while-loop and the notion of $\mindeg_{\psi}(v)$ is responsible for successful completion. Note that $|P_i| < 2\rho|V_i|$ holds by $(IB1)$ and $|K_i| = 2\rho|X_i|$, so if the algorithm doesn’t fail then all the vertices from $\bigcup P_i$ will be embedded. In the second while-loop we have to take care of the remaining unembedded cliques in $\bigcup K_i$. Moreover, since the algorithm embedded only onto vertices $\bigcup P_i \cup V_i^c$, the desired properties $(\text{FIN}2)$ and $(\text{FIN}3)$ are now guaranteed by $(IB2)$ and $(IB3)$. 

3.4. Fixing Buffer Defects
Therefore, it remains only to show that there is no failure while choosing a clique $K_G$ in some iteration. By Lemma 3.8, this is guaranteed if we can show that at each time $t$ the vertex $v \in V_t$ chosen has at least
\[
\frac{nd}{r} \max \left( p |V_j|, \deg_{\Gamma}(v; V_j) \right)
\]neighbours in each set $V_j \setminus \Im(\psi_t) \cup P_j$ such that $ij \in R'$.

By the assumption of the lemma, for each $i$, the set $X_i$ of vertices $x \in X_i$ contained in some reserved clique from $\bigcup_j K_j$ satisfies $|X_i| \leq 2\kappa(\Delta_R + 1)|g|X_i|$. It follows that for any $t$ we have $|\Im(\psi_t) \cap V_j| \leq 2\kappa(\Delta_R + 1)|g|X_j|$. Now we make the following observation, which will be crucial in the final argument. Suppose $U$ is a subset of some $V_j$ of size at most $2\kappa(\Delta_R + 1)|g|X_j|$, then the number of vertices in $\Gamma$ which have more than $4\kappa(\Delta_R + 1)\|\psi_t\|X_j|$ neighbours in $U$ is at most $\epsilon p^{\Delta-1}|V(\Gamma)|/r^2$. This is a direct consequence of the neighbourhood size property $\NS(\epsilon, r_1, \Delta)$ of $\Gamma$ and it can be applied to sets $U = \Im(\psi_t) \cap V_j$.

Each vertex $v$ of $V_t$ has at least
\[
\mu(d - \epsilon) \max \left( p |V_j|, \deg_{\Gamma}(v; V_j) \right) > \frac{nd}{r} \max \left( p |V_j|, \deg_{\Gamma}(v; V_j) \right)
\]neighbours in $V_j$ by $(G3)$. Since $4\kappa(\Delta_R + 1)\|\psi_t\|X_j| < \mu dp |V_j|/16$, it follows that for any $t$ there are at most $\Delta_R \epsilon p^{\Delta-1}|V(\Gamma)|/r^2$ vertices in $V_t$ with fewer than
\[
\frac{nd}{r} \max \left( p |V_j|, \deg_{\Gamma}(v; V_j) \right)
\]neighbours in any set $V_j \setminus \Im(\psi_t)$ such that $ij \in R'$, i.e. with $\mindeg_{\psi_t}(v) < 3\mu d/16$. In what comes we derive a contradiction to this.

At each time, at most one vertex is embedded in any given $V_j$. So if there is a time $t$ at which a vertex $v \in V_t$ is chosen with fewer than
\[
\frac{nd}{r} \max \left( p |V_j|, \deg_{\Gamma}(v; V_j) \right)
\]neighbours in some set $V_j \setminus \Im(\psi_t)$ such that $ij \in R'$, then at each of the preceding $\mu dp |V_j|/16$ times $t'$, $v$ had $\mindeg_{\psi_t'}(v) < 3\mu d/16$. Hence each of the at least $\mu dpn/(16\kappa r_1)$ vertices $v'$ chosen at these times has
\[
\mindeg_{\psi_t'}(v') \leq \mindeg_{\psi_t'}(v') < 3\mu d/16.
\]Since there are at most $r_1$ clusters $V_t$ in $G$, and since $\frac{ndp}{16\kappa r_1} > r_1 \Delta_R \epsilon p^{\Delta-1}|V(\Gamma)|/r^2$, this gives more vertices with $\mindeg_{\psi_t'}(v') < 3\mu d/16$ than exist in $G \subseteq \Gamma$, which contradiction completes the proof. \qed
CHAPTER 4

Proof of the blow-up lemma for bijumbled graphs

4.1. The RGA lemma and the proof of the blow-up lemma

In this section we prove Lemma 1.25, conditional on the random greedy lemma (Lemma 4.1) which we prove in Section 4.2.

The proof is quite similar to that of Lemma 1.21, and indeed we make use of the same General Setup. The difference is that we can no longer use property $\text{CON}(\varrho, r_1, \Delta)$ as a bijumbled graph $\Gamma$ need not satisfy it (see the discussion in Section 2.2). Instead, we ‘replace’ it with the lopsided neighbourhood size property $\text{LNS}(\varepsilon, r_1, \Delta)$.

In the next section we will use a somewhat different random greedy strategy than Algorithm 1 to prove the following lemma. However, we will re-use the auxiliary lemmas proved in Section 2.4 in its proof.

Lemma 4.1. We assume the General Setup. Suppose further that $\Gamma$ has properties $\text{NS}(\varepsilon, r_1, \Delta + 1)$, $\text{RI}(\varepsilon, (\varepsilon a, b), \varepsilon', d, r_1, \Delta + 1)$ and $\text{LNS}(\varepsilon, r_1, \Delta)$, and that at most $\varrho p \Delta |X_i|$ vertices in each $X_i$ are image restricted. Then there is a good partial embedding $\psi_{\text{RGA}}$ such that the following hold for each $i$.

- $(\text{PRGA}_1)$ Every vertex in $X_{i}^{\text{main}}$ is embedded to $V_{\text{main}} \cup V_{q}$ by $\psi_{\text{RGA}}$, and no vertex in $X_{i}^{\text{buf}}$ is embedded.
- $(\text{PRGA}_2)$ Every vertex in $V_i$ is a candidate for at least $\mu (d \Delta p/100)^b |X_i|$ vertices of $X_{i}^{\text{buf}}$.
- $(\text{PRGA}_3)$ For every set $W \subseteq V_i$ of size at least $\varrho |V_i|$, there are at most $\varrho |X_i|$ vertices in $X_{i}^{\text{buf}}$ with fewer than $\frac{1}{2} (dp)^b |W|$ candidates in $W$.
- $(\text{PRGA}_4)$ For all $x \in X_{i}^{\text{buf}}$ we have
  \[ \sum_{y \in X_{i}^{\text{buf}} : \left| U_{\text{buf}}(y) \cap C_{\text{buf}}(x) \right| > (p+\varepsilon p)^b |C_{\text{buf}}(x)|} \frac{|U_{\text{buf}}(y) \cap C_{\text{buf}}(x)|}{|C_{\text{buf}}(y)|} \leq \frac{|C_{\text{buf}}(x)|}{20}. \]
- $(\text{PRGA}_5)$ For each $i$, for all but at most $\varepsilon' p^2 |V_i|$ vertices $v \in V_i$, there are at most $\varepsilon' p^2 |V_i|$ vertices $v' \in V_i$ such that
  \[ |\{x \in X_{i}^{\text{buf}} : v, v' \in C(x)\}| > 24 \mu \Delta^2 (20 \mu^{-1} \zeta^{-1} d^{-\Delta}) b p^{2h} |X_i|. \]

Comparing this lemma to Lemma 3.1 (RGA lemma), one notices that we demand more of the NS and RI pseudorandomness properties. The reason for doing this is that we can then make use of Lemmas 2.28 and 2.29 for all buffer vertices, whether or not they are in copies of $K_{\Delta+1}$, and thus there is no exception for clique buffers in $(\text{PRGA}_2)$. Recall that Lemma 2.28 asserts that while embedding a small initial segment (in particular the neighbours of the buffer vertices), the neighbourhoods of other vertices are reduced by at most a factor of $1/2$ in size, while Lemma 2.29 gives a lower bound on the (conditional) probability of the event that a given vertex from $G$ can be used as an image of a particular buffer vertex. This simplifies our proof substantially—we no longer need to fix buffer defects—and in this case has
very little effect on the bijumbledness we ultimately require of \( \Gamma \) (it only affects the result in the case \( \Delta = 2 \)).

Property \((PRGA 3)\) corresponds to \((RGA 4)\). We also see that we ask for an additional pseudorandomness property \( LNS(\varepsilon, r_1, \Delta) \), which we require in order to embed all vertices (the algorithm does generate a queue, but it also embeds it) and also in order to obtain \((PRGA 4)\) and \((PRGA 5)\), which then form the ‘no dense spots’ property we require in order to complete the embedding, which rôle was played in the proof of Lemma 1.21 by the CON property. Somewhat informally, \((PRGA 4)\) asserts that typically candidate sets intersect as if they are random sets, while property \((PRGA 5)\) states the same about intersections of the neighbourhoods of vertices from \( V(G) \) in the candidate graph.

4.1.1. Outline of the proof of Lemma 1.25. The proof of Lemma 1.25 now looks quite similar to the proof of Lemma 1.21. Again, we use Lemma 2.22 to obtain the General Setup. However this time we do not choose any reserved cliques, whether or not we have buffer vertices in cliques. We note that although we could state a version of Lemma 4.1 which asked only for \( NS(\varepsilon, r_1, \Delta) \) and \( RL(\varepsilon, (\varepsilon a, b), \varepsilon', d, r_1, \Delta) \) rather than with \( \Delta + 1 \), and which made exceptions for clique buffers, this would not improve our eventual bijumbledness requirement on \( \Gamma \), except for \( \Delta = 2 \), since for \( \Delta \geq 3 \) the requirement is determined by \( LNS(\varepsilon, r_1, \Delta) \). We apply Lemma 4.1 to obtain a good partial embedding whose domain is \( X^{\text{main}} \) with the additional properties stated there. We then have only to embed \( X^{\text{buf}} \). Again, we do this one part at a time, by verifying Hall’s condition, and again we separate the verification into small, large and medium-sized subsets of \( X^{\text{buf}} \). Again, the ‘medium-sized’ case is dealt with quickly using \((PRGA 3)\).

However now we approach the ‘small’ and large cases differently, since we no longer have access to the CON property. In the ‘small’ case (Claim 4.2), our approach is the following. Given \( X \subseteq X^{\text{buf}} \), we try to construct an embedding of \( X \) into \( V_i^{\text{buf}} \) such that each \( x \) is embedded into \( C(x) \). Although this embedding is not part of the embedding we will finally give, its existence verifies Hall’s condition for \( X \). The way we will construct the embedding is simply to embed \( X \) vertex by vertex, at each step choosing for \( x \) uniformly at random a so far unused vertex of \( C^{\text{buf}}(x) \). Since \( \psi_{\text{RGA}} \) has \((GPE 3)\) we know \( C^{\text{buf}}(x) \) is always reasonably large, and we will show that with high probability it is never more than half covered. Let us briefly explain how this works. We assume that no candidate sets are more than half covered before reaching \( x \). When we embedded some vertex \( y \), the probability of embedding it to \( C(x) \) was at most

\[
\frac{|U^{\text{buf}}(y) \cap C^{\text{buf}}(x)|}{\frac{1}{2}|C^{\text{buf}}(y)|}
\]

since (by assumption) we embed \( y \) into a set of size at least \( \frac{1}{2}|C^{\text{buf}}(y)| \), of which certainly at most \( |U^{\text{buf}}(y) \cap C^{\text{buf}}(x)| \) are in \( C^{\text{buf}}(x) \). But \((PRGA 4)\) gives us an upper bound for the sum of these probabilities over all \( y \in X^{\text{buf}} \) such that the probability is ‘large’, and it is easy to account for the \( y \in X^{\text{buf}} \) such that the probability is not large. We can thus apply Lemma 2.2 to conclude that it is unlikely that \( C^{\text{buf}}(x) \) was more than half covered, and by the union bound over all \( x \in X^{\text{buf}} \) we conclude that with high probability we never fail.

For the ‘large’ case, Claim 4.3, we use a similar argument, but using \((PRGA 2)\) and \((PRGA 5)\) instead of \((GPE 3)\) and \((PRGA 4)\).

4.1.2. Proof of Lemma 1.25. Now we give the full details outlined above.

**Proof of Lemma 1.25.** First we choose constants as follows. Given \( \Delta, \Delta^{BL}_R, \Delta_J \) integers, \( a^{BL}, b^{BL} \) and \( d > 0 \), and \( \kappa^{BL} > 1 \), we set \( \vartheta = \Delta, \Delta_r = 8(\Delta + \Delta_J)^{10} a^{BL} \).
\[
\alpha = \frac{1}{2} \alpha_{\text{BL}}, \quad \zeta = \frac{1}{2} \zeta_{\text{BL}} \quad \text{and} \quad \kappa = 2 \kappa_{\text{BL}}.
\]
We now choose
\[
\mu = \frac{\alpha d^n}{20000n \Delta^2 \epsilon^2}, \quad \theta = \frac{\mu^2 \epsilon^2}{20000 \Delta^2}, \quad \varphi = \frac{1}{100n \sqrt{\Delta}}
\]
and
\[
\varphi' = \frac{\mu^2 \epsilon^2}{20000 \Delta^2} n^{-1/\varphi}.
\]
Now for input \( \Delta, d \) and \( \epsilon' \), Lemma 2.18 returns constants \( \epsilon_{a,b} \) and \( \epsilon_{12,18} > 0 \). We set
\[
\epsilon = \min\left(\frac{2^{10} \Delta d (\epsilon')^2}{\epsilon_{a,b}}, \epsilon_{12,18}\right).
\]
We let \( \epsilon_{\text{BL}} = \frac{1}{10} (\Delta + \Delta) - 10 \epsilon \) and \( \theta_{\text{BL}} = \frac{1}{10} (\Delta + \Delta') - 10 \theta \).

Now Lemma 1.25 returns \( \epsilon_{\text{BL}} \) and \( \theta_{\text{BL}} \). Given \( r_{\text{BL}} \) we let \( r_1 = 8(\Delta + \Delta)10r_{\text{BL}} \).
We choose \( c \) sufficiently small for Lemma 2.18 with input \( \Delta, d, \epsilon' \) and \( T_r = r_1 \).
We require
\[
p \geq 10^{6}(\Delta + \Delta)10^{r_{\text{BL}}} \left(\frac{\log n}{n}\right)^{1/(2\Delta)},
\]
where the condition on the constant in front of \( \left(\frac{\log n}{n}\right)^{1/(2\Delta)} \)-term comes from Lemma 2.22, while the term \( \left(\frac{\log n}{n}\right)^{1/(2\Delta)} \) is required for Lemma 4.1. We remark however that this lower bound is not a restriction since a somewhat stronger lower bound on \( p \) of the form \( \Omega(n^{1/2-1}) \) for any \( (p, c(p)n) \)-bijumbled graph follows by considering the inequality (1) which defines bijumbledness for, say, sets \( X = \{v\} \) and \( Y = V(G) \setminus \{v\} \cup N_G(v) \).

Let \( \Gamma \) be an \( n \)-vertex, \((p, c(p)_{\text{max}}(4\Delta + 2 + 2)\text{BL}) \)-bijumbled graph. Then Lemma 2.18 states that \( \Gamma \) has properties \( \text{NS}(\epsilon, r_1, \Delta + 1) \), \( \text{RI}(\epsilon, (\epsilon, a, b), \epsilon', d, r_1, \Delta + 1) \) and \( \text{LNS}(\epsilon, r_1, \Delta) \) respectively. From now on we will simply assume \( \Gamma \) is an \( n \)-vertex graph which satisfies these three properties.

Given a graph \( R^\text{BL} \) on \( r_{\text{BL}} \leq r^\text{BL} \) vertices, a spanning subgraph \( R' \text{BL} \) with \( \Delta(R^\text{BL}) \leq \Delta'(R') \), and graphs \( H \) and \( G \subseteq \Gamma \) with vertex partitions \( X^\text{BL} \) and \( Y^\text{BL} \), a family of potential buffer vertices \( X^\text{BL} \), and a \( (\theta_{\text{BL}}, \phi_{\text{BL}}, \zeta_{\text{BL}}, \Delta, \Delta') \)-restriction pair \( T^\text{BL} \), \( J \), suppose that the conditions of Lemma 1.25 are satisfied. Then Lemma 2.22 (with \( b_{12,22} = \Delta \)) gives a graph \( R \) on \( r \leq r_{\text{BL}} \) vertices, a spanning subgraph \( R' \) with \( \Delta(R') \leq \Delta(r) \), and \( \kappa \)-balanced size-compatible partitions \( \mathcal{X} \) and \( \mathcal{Y} \) of \( H \) and \( G \) respectively, each part having size at least \( n/(\kappa r_1) \), together with a family \( \mathcal{X} \) of potential buffer vertices and \( \mathcal{T} \) of image restrictions, subsets \( \mathcal{X}_{i}^\text{buf} \) of \( \mathcal{X} \), for each \( i \in [r] \), and partitions \( V_i = V^\text{main}_i \cup V^\text{q}_i \cup V^\text{c}_i \cup V^\text{buf}_i \) for each \( i \in [r] \) which satisfy the General Setup.

We let \( X^\text{main} = V(H) \setminus X^\text{buf} \). We now begin the embedding of \( H \) into \( G \). By Lemma 4.1, there is a good partial embedding \( \psi_{\text{RGA}} \) of \( X^\text{main} \) into \( V^\text{main} \cup V^\text{q} \) with properties \( (\text{PRGA} 1)-(\text{PRGA} 5) \) stated in that lemma.

Our aim now is to complete the embedding of \( H \) into \( G \) by finding for each \( i \) a matching in the available candidate graph between \( V'_i = V_i \setminus \text{Im}(\psi_{\text{RGA}}) \) and \( X^\text{buf}_i \).
Let us assume that \( X^\text{buf}_i \) is a degree-\( b \) buffer. To show there is a matching, we will verify Hall’s condition, so let \( Y \) be a non-empty subset of \( X^\text{buf}_i \). Let \( U \) be the set of vertices in \( V'_i \) which are candidate for at least one vertex of \( Y \), then our aim is to show \( |U| \geq |Y| \). We split this into three cases, the harder two of which are done in the following claims.

**Claim 4.2.** If \( 0 < |Y| \leq b|X_i| \), then we have \( |U| \geq |Y| \).

**Proof.** By \( (H5) \) no \( y \in Y \) is image restricted, so because \( \psi_{\text{RGA}} \) is a good partial embedding, by \( (\text{GPE3}) \) for each \( y \in Y \) we have
\[
|C^\text{buf}(y)| \geq (1 - \epsilon') \mu(d - \epsilon') p^b |V_i| \geq \frac{1}{2} \mu(p^b)|V_i|,
\]
where the second inequality is by choice of \( \epsilon' \). Let \( y_1, \ldots, y_j |Y| \) be an enumeration of \( Y \). Now for each \( j = 1, \ldots, |Y| \) we choose \( v_j \) uniformly at random from \( C^\text{buf}(y_j) \) \( (\text{GPE3}) \) for each \( y \in Y \) we have
\{v_1, \ldots, v_{j-1}\}$ if this is possible; if not, we say $v_j$ does not exist. Our aim is to show that this does not occur, in other words that we obtain $|Y|$ distinct vertices of $|U|$, verifying the claim.

Let $\mathcal{H}_0$ be the empty history, and for each $1 \leq j \leq |Y|$, let $\mathcal{H}_j$ be the history of this process up to and including the choice of $v_j$. We claim that a.s.

$$|C_{buf}(y_j) \setminus \{v_1, \ldots, v_{j-1}\}| \geq \frac{1}{2}|C_{buf}(y_j)|$$

(25)

holds for each $j$. For a given $j$ we define random variables $Z_{\ell}^{(j)}$, $\ell \in [j-1]$, as follows. We set $Z_{\ell}^{(j)} = 1$ if $v_\ell \in C_{buf}(y_j)$ and $|C_{buf}(y) \setminus \{v_1, \ldots, v_{\ell-1}\}| \geq \frac{1}{2}|C_{buf}(y)|$ hold, and $Z_{\ell}^{(j)} = 0$ otherwise. For any history $\mathcal{H}_{\ell-1}$ of this process we can bound the conditional expectation of $Z_{\ell}^{(j)}$ by

$$E(Z_{\ell}^{(j)} | \mathcal{H}_{\ell-1}) = \mathbb{P}(Z_{\ell}^{(j)} = 1 | \mathcal{H}_{\ell-1})$$

$$\leq \frac{|C_{buf}(y) \cap C_{buf}(y_j) \setminus \{v_1, \ldots, v_{\ell-1}\}|}{|C_{buf}(y) \setminus \{v_1, \ldots, v_{\ell-1}\}|} \leq \frac{|U_{buf}(y) \cap C_{buf}(y_j)|}{\frac{1}{2}|C_{buf}(y)|}.$$  (26)

Observe that $\sum_{\ell=1}^{j-1} Z_{\ell}^{(j)}$ is exactly the number of those vertices $v_\ell$ among $v_1, \ldots, v_{\ell-1}$ which lie in $C_{buf}(y_j)$ and were chosen from a set of size at least $\frac{1}{2}|C_{buf}(y)|$. To obtain an upper bound on the expectation of $\sum_{\ell=1}^{j-1} Z_{\ell}^{(j)}$, we sum up (26) for all $\ell \in [j-1]$. We split this sum according to whether $|U_{buf}(y) \cap C_{buf}(y_j)| \leq (p + \varepsilon)p|C_{buf}(y_j)|$ or not. In the former case, each summand is by (GPE2) and (GPE3) at most

$$2(p + \varepsilon)p|\mu(p + \varepsilon)p\mathbb{P}(|V_i|) \leq 4d^{-\Delta}p^b,$$

where the inequality is by choice of $\varepsilon$ and $\varepsilon'$, while (PRGA4) bounds the sum over the remaining terms. We get

$$\sum_{\ell=1}^{j-1} E[Z_{\ell}^{(j)} | \mathcal{H}_{\ell-1}] \leq \sum_{\ell=1}^{j-1} \frac{|U_{buf}(y) \cap C_{buf}(y_j)|}{10|C_{buf}(y)|} \leq 4(j - 1)d^{-\Delta}p^b + \frac{|C_{buf}(y_j)|}{10} \leq \frac{|C_{buf}(y)|}{10}$$

where the last inequality uses $j \leq |Y| \leq \theta|X_i|$, (GPE3) and the choice of $\theta$ and $\varepsilon$. Using Lemma 2.2 with $\delta = 1$, we conclude that the probability of the event $\sum_{\ell=1}^{j-1} Z_{\ell}^{(j)} > \frac{1}{2}|C_{buf}(y_j)|$ is at most $e^{-|C_{buf}(y_j)|/15} \leq e^{-\mu d^p|V_i|/30}$.

Observe that the event $\sum_{\ell=1}^{j-1} Z_{\ell}^{(j)} > \frac{1}{2}|C_{buf}(y_j)|$ contains the event that (25) fails at $j$, given that it did not fail for smaller $j$. Thus, taking a union bound over the $|Y| \leq n$ values of $j$, we see that the probability that (25) fails at any stage is at most

$$n \cdot e^{-\mu d^p|V_i|/30},$$

which is smaller than one for all sufficiently large $n$ by choice of $p$.

We conclude that there is a positive probability of choosing $|Y|$ distinct vertices in $U$, so $|U| \geq |Y|$ as desired.

We stress that although the proof of Claim 4.2 constructs an embedding, this embedding is not part of the final embedding of $H$ into $G$, and exists only to verify the claim. The same goes for the following claim.

**Claim 4.3.** If $|Y| > |X_{buf}^1| - \theta|X_i|$, then $|U| \geq |Y|$.

**Proof.** As in the proof of Lemma 1.21, the vertices $V'_i \setminus U$ are candidates only for vertices of $X_{buf}^1 \setminus Y$, and since $|V'_i| = |X_{buf}^1|$, what we need to show is that $|V'_i \setminus U| \leq |X_{buf}^1 \setminus Y|$.
Let $v \in V'_i$ be a vertex and let $C(v)$ denote those vertices $x$ from $X_{i \text{buf}}$ for which $v$ is a candidate, i.e. $C(v)$ denotes the neighbours of $v$ in the candidate graph between $V'_i \setminus U$ and $X_{i \text{buf}} \setminus Y$. By (PRGA2), for each vertex $v$ of $V'_i \setminus U$ we have $|C(v)| \geq \mu(d^{2}p/100)^{b}|X_{i}|$. By (PRGA5), for all but at most $\varepsilon^{p^{2b}}|V_{i}|$ vertices $v$ of $V'_i \setminus U$, we have
\[
\left\{ v' \in V_i : \left| \{ x \in X_{i \text{buf}} : v, v' \in C(x) \} \right| > 24\mu\Delta^{2}\left( \frac{20}{\mu C_{\text{prga}}} \right)^{b}p^{2b}|X_{i}| \right\} \leq \varepsilon^{p^{2}}|V_{i}|. \tag{27}
\]
We choose an ordering $v_1, \ldots, v_{|V'_i \setminus U|}$ of $V'_i \setminus U$ which puts the vertices $v$ failing (27) first. We choose, for each $j = 1, \ldots, |V'_i \setminus U|$, a vertex $x_j$ uniformly at random from the vertices of $X_{i \text{buf}} \setminus \{x_1, \ldots, x_{j-1}\}$ for which $v_j$ is a candidate if this is possible; if not, we say $x_j$ does not exist. As in the previous claim, our aim is to show that a.a.s. this does not occur.

Let $\mathcal{H}_0$ be the empty history, and for each $1 \leq j \leq |V'_i \setminus U|$, let $\mathcal{H}_j$ be the history of this process up to and including choosing $x_j$. We claim that, with positive probability, at each step $i$ we choose from a set of size at least $\frac{1}{2}|C(v_{i})| \geq \frac{1}{2}\mu(d^{2}p/100)^{b}|X_{i}|$.

We introduce random variables $Z^{(j)}_{t}$ for $j \in [|V'_i \setminus U|]$ and $t \in [j-1]$ as follows. We set $Z^{(j)}_{t} = 1$ if $x_{t} \in C(v_{j})$ and $|C(v_{j}) \setminus \{x_{1}, \ldots, x_{t-1}\}| \geq \frac{1}{2}|C(v_{t})|$ hold, and we set $Z^{(j)}_{t} = 0$ otherwise. Observe that $\sum_{t=1}^{j-1} Z^{(j)}_{t}$ is exactly the number of vertices $x_k$ among $x_1, \ldots, x_{t-1}$ which lie in $C(v_{j})$ and which were chosen from a set of size at least $\frac{1}{2}|C(v_{k})|$. Thus, it suffices to show that with positive probability, for every $j$ we have
\[
\sum_{t=1}^{j-1} Z^{(j)}_{t} < \frac{1}{2}\mu(d^{2}p/100)^{b}|X_{i}|. \tag{28}
\]
If $v_j$ is a vertex failing (27), then $j \leq \varepsilon^{p^{2}}|V_{i}| < \frac{1}{2}\mu(d^{2}p/100)^{b}|X_{i}|$, where the second inequality is by choice of $\varepsilon'$. Thus, even if all vertices $x_1, \ldots, x_{j-1}$ happen to be candidates for $v_j$, we still have (28). Next we assume that $v_j$ satisfies (27). For each of the at most $\varepsilon'\mu^{p^{2}}|X_{i}|$ vertices $v_j$ such that
\[
\left| \{ x \in X_{i \text{buf}} : v_{j}, v_{t} \in C(x) \} \right| > 24\mu\Delta^{2}(20\mu^{-1}\xi^{-1}d^{-3})^{b}p^{2b}|X_{i}|,
\]
the conditional expectation $E(Z^{(j)}_{t} | \mathcal{H}_{t-1})$ can be bounded from above by 1, while for the remaining vertices, it is at most
\[
\frac{\left| \{ x \in X_{i \text{buf}} : v_{j}, v_{t} \in C(x) \} \right|}{\frac{1}{2}|C(v_{t})|} \leq \frac{24\mu\Delta^{2}(20\mu^{-1}\xi^{-1}d^{-3})^{b}p^{2b}|X_{i}|}{\frac{1}{2}\mu(d^{2}p/100)^{b}|X_{i}|} \leq 48\Delta^{2}\left( \frac{2000}{\mu C_{\text{prga}}} \right)\Delta^{b}.
\]
This yields
\[
\sum_{t=1}^{j-1} E[Z^{(j)}_{t} | \mathcal{H}_{t-1}] \leq \varepsilon' \mu^{p^{2}}|X_{i}| + |V'_i \setminus U| \cdot 48\Delta^{2}\left( \frac{2000}{\mu C_{\text{prga}}} \right)\Delta^{b} \leq \frac{1}{2}\mu(d^{2}p/100)^{b}|X_{i}|,
\]
where the last inequality is because $|V'_i \setminus U| \leq \varrho|X_{i}|$ by choice of $\varrho$ and $\varepsilon'$. As before, we apply Lemma 2.2 for each $j \in [|V'_i \setminus U|]$ with $\delta = 1$ and $(Z^{(j)}_{t})_{t \in [j-1]}$. We thus may bound the probability that $v_j$ is candidate for more than $\frac{1}{2}|C(v_{i})| \geq \frac{1}{2}\mu(d^{2}p/100)^{b}|X_{i}|$ of the vertices $x_1, \ldots, x_{j-1}$ by at most $e^{-\mu(d^{2}p/100)^{b}|X_{i}|/12}$. Taking a union bound over the at most $\varrho|X_{i}|$ choices of $j$ we see that with probability $1 - \varrho|X_{i}| \cdot e^{-\mu(d^{2}p/100)^{b}|X_{i}|/12} > 0$ (where the inequality is by choice of $p$), at each step we choose from a set of size at least $\frac{1}{2}\mu(d^{2}p/100)^{b}|X_{i}|$ as claimed. In particular, we succeed in choosing $|V'_i \setminus U|$ distinct vertices of $X_{i \text{buf}} \setminus Y$, so as desired we have $|U| \geq |Y|$. \qed

The final case is to show that $\varrho|X_{i}| < |Y| \leq |X_{i \text{buf}}| - \varrho|X_{i}| = |V'_i \setminus U| - \varrho|V_{i}|$ and $|U| < |Y|$ is a contradiction. In this case, we have $|V'_i \setminus U| > \varrho|V_{i}|$. By (PRGA3)
there are at most $\varrho|X_t|$ vertices of $X_t$ with fewer than $\frac{1}{2}(dp)^b|V_t| \setminus U| \setminus U$, so in particular there is a vertex of $Y$ with candidates in $V_t \setminus U$, in contradiction to the definition of $U$.

This completes the verification of Hall’s condition, so we can extend $\psi$ to an embedding $\psi$ of $H$ into $G$ as desired, completing the proof of Lemma 1.25. □

4.2. Proof of the bijumbled graphs RGA lemma

The proof of Lemma 4.1 is very similar to the proof of Lemma 3.1 (RGA lemma), and indeed we can reuse the auxiliary lemmas of Section 2.4. We again use a random greedy algorithm which we show produces the desired embedding with high probability. The difference from Algorithm 1 is that when at time $t$ we reach a vertex $x$ in the order $\tau$ (given again by Lemma 2.26) which is in the queue $Q_t$, we do not skip it, but instead we embed it into $V^q$. Thus, we will always have $t = \tau(x) - 1$. The embedding will be done uniformly at random into the set $A_t(x) \setminus B_t(x)$. We remind, that $B_t(x)$ is the set of bad vertices with respect to $\psi_t$ and $Q_t$, that is, the vertices $v$ such that the extension $\psi_t \cup \{x \to v\}$ is not a good partial embedding or there is an unembedded neighbour $y$ of $x$ not in $Q_t$, such that $\deg_{\psi_t}(y; A_t^{\main}(y)) < (d - \varepsilon)p|A_t^{\main}(y)|$ holds (see Definition 2.24).

The embedding algorithm we use is the following Algorithm 3 below. Observe

Algorithm 3: Another random greedy algorithm

**Input:** $G \subseteq \Gamma$ and $H$ with partitions satisfying the General Setup; an ordering $\tau$ on $X^{\main}$

$t := 0$ ;

$\psi_0 := \emptyset$ ;

$Q_0 := \{x \in V(H) : |I_x| < \frac{1}{2}\mu(d - \varepsilon)^{|J_x|}p|J_x||V^{\main}(x)|\}$ ;

repeat

let $x \in X^{\main} \setminus \Dom(\psi_t)$ be the next vertex in the order $\tau$ ;

if $x \in Q_t$ and $|A_t^x(x) \setminus B_t(x)| < \frac{1}{2}\mu\zeta(dp)^{\pi_1(x)}|V(x)|$ then

halt with failure ;

end

choose $v$ uniformly at random in

$\begin{cases} 
A_t^x(x) \setminus B_t(x) & \text{if } x \notin Q_t ; \\
A_t^x(x) \setminus B_t(x) & \text{if } x \in Q_t ; 
\end{cases}$

$\psi_{t+1} := \psi_t \cup \{x \to v\}$ ;

$Q_{t+1} := Q_t$ ;

forall the $y \in X^{\main} \setminus \Dom(\psi_{t+1})$ do

if $|A_t^{\main}(y)| < \frac{1}{2}\mu(d - \varepsilon)^{\pi_t^{\main}(y)}p^{\pi_t^{\main}(y)}|V^{\main}(y)|$ then

$Q_{t+1} := Q_{t+1} \cup \{y\}$ ;

end

end

$t := t + 1$ ;

until $\Dom(\psi_t) = X^{\main}$ ;

$t_{\text{RGAend}} := t$ ;

that there is an important difference from Algorithm 1. Algorithm 3 can fail, and we will have to prove that with high probability it does not. This means proving that sets $A_t^x(x)$ do not get small, or equivalently that sets $C_t^x(x)$ are never substantially covered by $\Dom(\psi_t)$. However there is also an important similarity: provided the algorithm has not yet halted with failure, it maintains the same invariants as Algorithm 1 (see Claim 4.4).
Observe that (in contrast to Algorithm 1) if Algorithm 3 has not yet failed at time \( t \), then all of the first \( t \) vertices in the order \( \tau \) have been embedded. Thus \( \pi^\tau_{(x)−1}(x) \) is always equal to \( \pi^\tau(x) \), where

\[
\pi^\tau(x) := \{|y ∈ N_H(x) : \tau(y) < \tau(x)\} + |J_x|.
\]

Much of the work of proving Lemma 4.1 is contained in the auxiliary lemmas in Section 2.4 which we use again here. What remains is to justify that Algorithm 3 with high probability does not halt with failure and does give a good partial embedding with properties \((PRGA\,4)\) and \((PRGA\,5)\).

**Proof of Lemma 4.1.** We require

\[
\begin{align*}
\mu &\leq \frac{\mu^\Delta}{1320n\Delta n}, \\
\epsilon' &\leq \frac{\mu^\Delta d^\Delta}{1000\Delta^2\Delta^\Delta n^2}2^{-4/\epsilon}, \\
p &\geq \frac{1000\Delta^\Delta n^2}{\mu^\Delta d^\Delta \epsilon^\Delta n^{2000\epsilon n/1000}}\left(\log \frac{n}{\epsilon}\right)^{1/2}.
\end{align*}
\]

Recall that we assumed that \( \Gamma \) possesses properties \( NS(\varepsilon, r, 1, \Delta + 1) \) and \( \text{LNS}(\varepsilon, r, 1, \Delta) \.

Let \( \tau \) be an order on \( V(H) \) given by Lemma 2.26. We run Algorithm 3.

**Claim 4.4.** The following hold at each time \( t \) in the running of Algorithm 3.

\[(INV\,1) \quad \psi_t \text{ is a good partial embedding,} \]
\[(INV\,2) \quad \text{Either } |A_{\text{main}}^t(x)| \geq \frac{1}{2}\mu(d − \epsilon')\pi^\tau_{(x)}(x)|V_{\text{main}}^t(x)| \text{ or } x \in Q_t.\]
\[(INV\,3) \quad \text{When we embed } x \text{ to create } \psi_{t+1}, \text{ we do so uniformly at random into a set of size at least } \frac{1}{10}\mu\zeta(dp)^\tau_{(x)}|V(x)|.\]

**Proof.** Algorithm 3 maintains \((INV\,1)\) and \((INV\,2)\) by definition. Since \( \frac{1}{2}\mu(d − \epsilon')\pi^\tau_{(x)}(x)p^\Delta |V_{\text{main}}^t(x)| \geq \frac{1}{10}\mu\zeta(dp)^\tau_{(x)}|V(x)| \) by choice of \( \epsilon' \) for each \( 0 ≤ b ≤ \Delta \), by Lemma 2.25 it also maintains \((INV\,3)\).

As in the proof of Lemma 3.1, the conditions for Lemma 2.27 are met, so we conclude that with probability at least \( 1 − 2^{-n/(\kappa r^4)} \) at each time \( t \) when Algorithm 3 is running and for each \( i ∈ [r] \) we have \( |Q_t \cap X^i| ≤ g|X^i| + |X^i| ≤ 2g|X^i| \), where the final inequality is by \((G6)\). In order to show that Algorithm 3 runs successfully, we need to show that the ‘halt with failure’ line is a.a.s. never reached, i.e. that if \( x ∈ Q_{\tau(x)−1} \) then \( |A_{\tau(x)−1}^q(x) \setminus B_{\tau(x)−1}^q(x)| \) is at least \( \frac{1}{10}\mu\zeta(dp)^\tau_{(x)−1}(x)|V_i| \). Since we know by Lemma 2.25 that \( B_{\tau(x)−1}(x) \) is small and by \((GPE3)\) that \( C_{\tau(x)−1}(x) \) is large, what we want is to show that \( \text{Im}(\psi_{t−1}) \) covers only a small fraction of \( C_{\tau(x)−1}(x) \).

Observe that we embed the vertices \( Q_t \cap X_i \) into \( V_i^q \) in a random procedure which is very similar to the embedding strategy we used in the proof of Claim 4.2 (verification of Hall’s condition for small sets of buffer vertices). However here a complication is that we do not know ‘in advance’ what \( C_{\pi^\tau(x)−1}(x) \) will be when embedding earlier vertices to \( V_i^q \). But we do know that it will be contained in the \( \Gamma \)-neighbourhood in \( V_i \) of some collection of \( \pi^\tau(x) ≤ \Delta \) vertices, and that it will be large. Thus, it suffices to prove the following. For each \( i \), each \( 1 ≤ \ell ≤ \Delta \) and \( v_1, \ldots, v_\ell \) such that \( N_{v_\ell}(v_1, \ldots, v_\ell; V_i) \) is large, \( \text{Im}(\psi_t) \) never covers much of \( N_{\tau(x)}^q(v_1, \ldots, v_\ell; V_i) \). We can prove this statement by using the same analysis as in the proof of Lemma 1.21, given a ‘sum condition’ similar to \((PRGA\,4)\). Thus our next task is to show such a statement is likely to hold. We will also show that if Algorithm 3 does not halt with failure then it is likely to have \((PRGA\,4)\), since the arguments are similar.
CLAIM 4.5. Suppose that for each \( i \in [r] \) and each time \( t \) we have \(|Q_i \cap X_i| \leq 2p|X_i|\). Then a.a.s. the following statements hold.

For any \( t \), any \( 1 \leq \ell \leq \Delta \), any \( i \in [r] \) and any vertices \( v_1, \ldots, v_\ell \) such that

\[
\deg_G(v_1, \ldots, v_\ell; V_i) \geq \varepsilon p' |V_i|
\]

we have

\[
\sum_{y \in X^\text{main}_{\tau(y) \cap Q_i}} \frac{|U_{\tau(y)-1}(y) \cap N^*_t(v_1, \ldots, v_\ell; V_i)|}{|A^q_{\tau(y)-1}(y) \setminus B_{\tau(y)-1}(y)|} \leq \frac{\mu \zeta d^\Delta \deg_G(v_1, \ldots, v_\ell; V_i)}{20}.
\] (29)

For any \( 1 \leq \ell \leq \Delta \), any \( i \in [r] \) and any vertices \( v_1, \ldots, v_\ell \) such that

\[
\deg_G(v_1, \ldots, v_\ell; V_i^\text{buf}) \geq (dp - \varepsilon' p')^\ell |V_i^\text{buf}|,
\]

we define \( X'_i \) to be the set

\[
\{ y \in X_i^\text{buf} : |U_i^\text{buf}(y) \cap N_G^*(v_1, \ldots, v_\ell; V_i^\text{buf})| > (p + \varepsilon p) \deg_G(v_1, \ldots, v_\ell; V_i^\text{buf}) \}
\]

and have

\[
\sum_{y \in X'_i} \frac{|U_i^\text{buf}(y) \cap N^*_t(v_1, \ldots, v_\ell; V_i^\text{buf})|}{|C^*_t(y)|} \leq \frac{\deg_G(v_1, \ldots, v_\ell; V_i^\text{buf})}{20}.
\] (30)

The idea of the proof of (29) is as follows. The denominators on the left hand side of (29) are by (INV3) never much smaller than they ‘should’ be, so the main task is to show that the numerators do not tend to be too large. To show this, we consider the evolution of \( U_i(y) \cap N^*_t(v_1, \ldots, v_\ell; V_i) \) for some fixed \( y \) as \( t \) increases. Since \( J_y = \emptyset \), at first this set has the size we expect, namely it is all of \( N^*_t(v_1, \ldots, v_\ell; V_i) \).

Each time a neighbour of \( y \) is embedded, we expect that the set size shrinks by a factor roughly \( p \). If this is the case for each neighbour, the size at \( t = \tau(y) - 1 \) is roughly a \( p^{\pi^*(y)} \)-factor times its original size, which turns out to be a good enough bound for (29). If not, there is some first time when we embed a neighbour of \( y \), say the \( s \)th neighbour, ‘badly’, that is, the set size does not shrink by a factor roughly \( p \). We say we fail at step \( s \). At worst, it could be that the set size does not thereafter change, so that it stays roughly a \( p^{\pi^*(y)-1} \)-factor times its original size. In this case we ‘lose’ a \( p^{\pi^*(y)-1} \)-factor. But the LNS property tells us that the probability of failing at step \( s \) is less than \( p^{\pi^*(y)-1} \). Heuristically, this gets us back the lost factor; to make it rigorous, we apply Lemma 2.2.

PROOF. We require

\[
\theta \leq \frac{\mu \zeta^2 d^\Delta}{2000}, \quad \varepsilon \leq \min \left(2^{-\Delta} \varepsilon', \frac{\theta \mu \zeta d^\Delta}{20 \kappa \Delta} \right) \quad \text{and} \quad \rho \geq \frac{20 \Delta^4 \kappa r^2}{\theta}, \quad \left(\frac{\log n}{n}\right)^{1/\Delta}.
\]

We start with (29). Given \( 1 \leq \ell \leq \Delta \), let \( v_1, \ldots, v_\ell \) be vertices of \( \Gamma \) such that the set \( U := N^*_t(v_1, \ldots, v_\ell; V_i) \) has size at least \( \varepsilon p' |V_i| \). By (INV3) we have the lower bound

\[
|A^q_{\tau(y)-1}(y) \setminus B_{\tau(y)-1}(y)| \geq \frac{1}{\mu} \deg_G(v_1, \ldots, v_\ell; V_i) |V_i|
\]

for the denominator in each summand of (29), and the difficulty is to upper bound the numerator. Consider the running of Algorithm 3. At any time \( t \leq \tau(y) - 1 \), the set \( U^*_t(y) \cap U \) is the \( \Gamma \)-neighbourhood in \( U \) of the \( \pi^*_t(y) \) embedded vertices from \( N_H(y) \), and in a truly random set we would thus ‘expect’ to find that

\[
|U_t(y) \cap U| \leq (p + \varepsilon p)^{\pi^*_t(y)} |U|.
\] (31)

If this inequality remains true up to \( t = \tau(y) - 1 \), then since \(|J_y| = 0 \) by assumption, we have \( \pi^*_t(y) = \pi^*(y) \), and (31) gives an upper bound good enough for the
We have \( \pi \) step \( \leq y \) number of vertices \( z \) (which has at least one unembedded neighbour and hence at most \( \Delta \)). Therefore, by LNS(\( \epsilon \)), we know that the number of such vertices \( z \) which is at least \( \Delta - 1 \) embedded to a vertex \( \epsilon \) in \( \Gamma \) with ‘too many’ neighbours in \( U_{\tau(z)} - 1(y) \cap U \). Let \( W \) be a superset of \( U_{\tau(z)} - 1(y) \cap U \) of size \((p + \epsilon p)^{s-1}|U|[|]\). By choice of \( \epsilon \) we have \( |W| \geq \epsilon p^{2s-1}|V[|\). Because \( y \) fails at step \( s \) we see that \( z \) is embedded to a vertex \( v \) of \( \Gamma \) with \( \deg_{\Gamma}(v; W) > (p + \epsilon p)^{s}|U| = (p + \epsilon p)|W| \). By LNS(\( \epsilon, r_1, \Delta \)) we know that the number of such vertices \( v \) in \( \Gamma \) is at most \( \epsilon p^2 |V[|/|\). Therefore, by (INV3), the probability of embedding \( z \) (which has at least one unembedded neighbour and hence at most \( \Delta - 1 \) embedded neighbours) to such a vertex \( v \), conditioning on the history up to but not including embedding \( z \), is at most

\[
\frac{\epsilon p^2 |V[|/|\mu \zeta(d)|^{2\Delta - 1}|V[|/10 \leq \frac{\theta}{2\Delta} p^{\Delta - s + 1} \leq \frac{\theta}{2\Delta} p^{\tau(y) - s + 1},
\]

where the first inequality is by choice of \( \epsilon \) and the second since \( \tau(y) \leq \Delta \).

We can restate this as: the probability that \( y \) fails at step \( s = \tau(y) - j + 1 \), conditioning on the history up to but not including the embedding of the vertex at step \( s \), is at most \( \frac{\epsilon p^2}{2\Delta} p^j \) for each \( 1 \leq j \leq \tau(y) \). It follows that the expected number of vertices \( y \in X_1 \) which fail at step \( \tau(y) - j + 1 \) is at most \( \frac{\epsilon p^2}{2\Delta} p^j |X[| \), and by Lemma 2.2, applied with \( \delta = 1 \), the probability that more than \( \frac{\epsilon p^2}{2\Delta} p^j |X[| \) vertices \( y \in X_1 \) fail at step \( \tau(y) - j + 1 \) is at most \( \frac{\epsilon p^2}{2\Delta} p^j |X[| \). The probability that, for each \( 1 \leq j \leq \Delta \), at most \( \frac{\epsilon p^2}{2\Delta} p^j |X[| \) vertices \( y \in X_1 \) fail at step \( \tau(y) - j + 1 \) is thus at least

\[
1 - \Delta \exp \left(-\frac{\epsilon p^2}{2\Delta} p^j |X[| \right).
\]

Suppose this good event occurs, i.e. for each \( j \), at most \( \frac{\epsilon p^2}{2\Delta} p^j |X[| \) vertices \( y \in X_1 \) fail at step \( s = \tau(y) - j + 1 \). Since a vertex \( y \) failing at step \( j \) satisfies

\[
|U_{\tau(y)} - 1(y) \cap U| \leq (p + \epsilon p)^{\tau(y) - j - 1} |U|[|]\)
\]

we have

\[
\sum_{y \in X_1} \frac{|U_{\tau(y)} - 1(y) \cap U|}{|U_{\tau(y)} - 1(y) \cap U| > (p + \epsilon p)^{\tau(y) - j}(v)|U|[|]} \leq \sum_{y \in X_1} \frac{(p + \epsilon p)^{\tau(y) - j}|U|}{10 \mu \zeta(d)(p)^{\tau(y)} |V[|} \leq \sum_{j=1}^{\Delta} \frac{(p + \epsilon p)^{\tau(y) - j} |U|}{10 \mu \zeta(d)(p)^{\tau(y)} |V[|} \leq \frac{20 |U|}{\mu \zeta(d)} \leq \frac{\mu \zeta(d)}{|U|},
\]

(33)
where the last two inequalities are by choice of $\varepsilon$ and $\rho$ respectively. On the other hand, because $|X^\text{main}_i \cap Q_t| \leq 2q|X_i|$, we have

$$
\sum_{y \in X^\text{main}_i \cap Q_t: |U_{\tau(y)-1}(y) \cap U| \leq (p+\varepsilon)p^{\pi^*}(y)|U|} \frac{|U_{\tau(y)-1}(y) \setminus B_{\tau(y)-1}(y)|}{|A^n_{\tau(y)-1}(y) \setminus B_{\tau(y)-1}(y)|} \leq \frac{(1 + \varepsilon)\Delta |U| \cdot 2q|X_i|}{\frac{1}{\mu} \mu \zeta^\Delta |V_i|} \leq \frac{40q|U|}{\mu \zeta^\Delta |U|} \leq \frac{\mu \zeta^\Delta |U|}{40},
$$

and putting these together we conclude (29).

It remains to bound the probability that for some vertices $v_1, \ldots, v_t$ with $\deg_G(v_1, \ldots, v_t; V) \geq \varepsilon \varepsilon' |V|$, the inequality (29) is violated, in other words that the good event mentioned above fails. There are $r \leq r_1$ choices of $i$, and at most $n + n^2 + \cdots + n^{\Delta} \leq \Delta n^\Delta$ choices of $v_1, \ldots, v_t$, so by the union bound and (32) the probability of such a bad event occurring is at most

$$r_1 \Delta n^\Delta \cdot 2\Delta \exp \left( - \frac{\mu \zeta^\Delta}{20} n^\Delta \right)
$$

which tends to zero as $n$ tends to infinity by choice of $p$.

We now come to (30). This inequality looks very much like (33) above, and its proof is almost identical to the proof of (33). The differences are that we consider $G$-neighbourhoods in $V^\text{buf}_i$ not $\Gamma$-neighbourhoods in $V_i$ and that we have a lower bound on $C_i^\text{buf}(x)$ from (GPE3) rather than on $|A^n_i(x) \setminus B_i(x)|$ from (INV3). Since this lower bound is larger, the same constant choices work. We omit the details. □

We note that it is this claim which is most responsible for our eventual bijumbledness requirement on $\Gamma$. Although the proof may seem quite wasteful—we assume that if one neighbour is embedded badly then potentially all future neighbours can be embedded badly without any further penalty, which seems unreasonable—we were not able to make it work with any weaker condition than LNS($\varepsilon$, $r_1$, $\Delta$) for general graphs $H$. Our analysis is in some sense tight for vertices $y$ which fail at step $\pi^*(y)$, when the final neighbour $z$ of $y$ coming before $y$ in $\pi$ is embedded. What still seems unreasonable is that all the vertices in $\Gamma$ which have exceptionally high degree to $U_{\tau(z)-1}(y) \cap U$ turn out to be in $A_{\tau(z)-1}(z)$, which is the worst case we are effectively using in our proof. We expect that a more careful analysis, possibly involving some modification to Algorithm 3, would allow one to improve on this and thus improve on the bijumbledness requirement of Lemma 1.25.

We can now use Lemma 2.2 to complete the proof that the sets $C_i^\text{r}(x)$ do not get covered by $\text{Im}(\psi_i)$, much as in Claim 4.2 (verification of Hall’s condition for small sets of buffer vertices).

**Claim 4.6.** A.a.s. for each $x \in X^\text{main}$, at each time $t \leq \tau(x) - 1$ and before the termination of Algorithm 3, we have $|C_i^\text{r}(x) \cap \text{Im}(\psi_i)| \leq \frac{1}{2}|C_i^\text{r}(x)|$.

**Proof.** We require

$$
\varepsilon \leq \varepsilon' \leq 2^{-2\Delta}, \quad \rho \leq \frac{\mu \zeta^\Delta}{30} \quad \text{and} \quad p \geq \frac{1000q}{\mu \zeta^\Delta} \left( \frac{\log n}{n} \right)^{1/\Delta}.
$$

Suppose that the conclusions of Claim 4.5 hold (in particular (29) holds). Given $i \in [r]$ and $x \in X^\text{main}_i$, if $t \leq \tau(x) - 1$ and Algorithm 3 has not terminated before time $t$, then by (GPE3) and (G5) we have $|C_i^\text{r}(x)| \geq (1 - \varepsilon')\mu \zeta (dp - \varepsilon') \pi^*(x)|V_i|$. Since $C_i^\text{r}(x) \subseteq U_i(x) = N^\text{r}_i(v_1, \ldots, v_t; V)$ for some vertices $v_1, \ldots, v_t$ with $t = \pi^*_i(x) \leq \Delta$,
we conclude by (29) that
\[
\sum_{y \in \mathcal{X}^{\text{main}} \cap \mathcal{Q}_i : \tau(y) \leq t} \frac{|U_{\tau(y)-1}(y) \cap C_i^\infty(y)|}{|A_{\tau(y)-1}(y) \setminus B_{\tau(y)-1}(y)|} \leq \frac{\mu \zeta d^3 |U_i(x)|}{20}.
\]

Observe that the summand in the above inequality is an upper bound for the probability that \( y \) is embedded to \( C_i^\infty(y) \) for \( y \in \mathcal{Q}_i \), conditioning on the history up to but not including the embedding of \( y \). Since the probability that \( y \) is embedded to \( C_i^\infty(y) \) is zero if \( y \) is not in \( \mathcal{Q}_i \), we are in a position to apply Lemma 2.2. This lemma, with \( \delta = 1 \), tells us that the probability that more than \( \frac{1}{10} \mu \zeta d^3 |U_i(x)| \) vertices \( y \) of \( \mathcal{X}^{\text{main}} \) with \( J_y = \emptyset \) are embedded to \( C_i^\infty(y) \) is at most \( \exp \left( -\frac{1}{100} \mu \zeta d^3 |U_i(x)| \right) \). If this ‘bad’ event does not occur, then by (GPE2), (GPE3) and (G5), and because the number of vertices in \( \mathcal{X}^{\text{main}} \) with \( J_y \neq \emptyset \) is by (G6) at most \( \gamma p\zeta \Delta |X_i| \), we have the desired statement:
\[
|C_i^\infty(y) \cap \text{Int}(\psi_t)| \leq \frac{1}{10} \mu \zeta d^3 |U_i(x)| + \gamma p\zeta |X_i| \leq \frac{1}{2} (1 - \epsilon') \mu \zeta (dp - \epsilon' p) \tau_i(x) |V_i|.
\]

The probability that the conclusions of Claim 4.5 fail to hold, or that any of the above ‘bad’ events occur, is at most \( o(1) + \tau_i n^2 \cdot \exp \left( -\frac{1}{100} \mu \zeta d^3 (p - \epsilon' p) \Delta n \right) \), which tends to zero as \( n \) tends to infinity by choice of \( p \), completing the proof. \( \square \)

If the conclusions of Lemma 2.27, Claim 4.5 and Claim 4.6 hold (which we think of as being good events), then by (GPE3) and Lemma 2.25 we have
\[
|A_{\tau(x)-1}(x) \setminus B_{\tau(x)-1}(x)| \geq \frac{1}{2} \mu \zeta (dp - \epsilon' p) \tau(x) |V_i| - 20 \Delta^2 \epsilon' p \tau(x) |V_i|
\]
and the right hand side is by choice of \( \epsilon' \) at least \( \frac{1}{10} \mu \zeta (dp) \tau(x) |V_i| \). In other words, the ‘halt with failure’ line of Algorithm 3 is never reached. Since each of the three good events a.a.s. occurs, the algorithm a.a.s. completes successfully. Now (PRGA1) is guaranteed by successful completion of Algorithm 3, and (PRGA3) is as (RGA4) in the proof of Lemma 3.1 implied by the good event of Lemma 2.27, while (PRGA4) is given by the good event of Claim 4.5, specifically taking (30) with \( t = t_{\text{RGAend}} \) and vertices \( \psi_t(N_H(x)) \) for each \( x \in \mathcal{X}_r^{\text{buf}} \).

Next we establish that (PRGA2) a.a.s. holds. We apply Lemma 2.28 with \( B = \Delta + 1 \) and \( T = \tau(z) \) where \( z \) is the last vertex in \( N(X_r^{\text{buf}}) \). Since \( N(X_r^{\text{buf}}) \) forms the initial segment of \( \tau \) by Lemma 2.26(a), we see that all vertices \( y \) embedded up to time \( T \) have \( \tau(y) \leq \Delta + 1 \leq B - 2 \), and by identical logic as in Claim 3.6 none of these vertices enter the queue. Therefore, by (GPE3), Lemma 2.25 and choice of \( \epsilon' \), each vertex \( y \) from \( N(X_r^{\text{buf}}) \) is embedded uniformly at random to a set of size at least \( \frac{1}{10} (dp) \tau(y) |V(y)| \). Finally, by (H3) we have \( |N(X_r^{\text{buf}}) \cap X_i| \leq 4 \Delta \zeta \mu |X_i| \) for each \( i \in [r] \). This justifies that the conditions of Lemma 2.28 are met, so we conclude that a.a.s. the following event \( E_{1225} \) occurs. For each \( v \in V_i \) and \( j \) such that \( ij \in E(R^i) \) we have \( |N_G(v; V_j^{\text{main}}) \setminus \text{Int}(\psi_T)| \geq \frac{1}{2} \deg_G(v; V_j^{\text{main}}) \).

We are now in a position to apply Lemma 2.29, again with \( B = \Delta + 1 \). Observe that this time, if \( E_{1225} \) occurs, it applies to any \( x \in X_r^{\text{buf}} \) and \( v \in V_i \), whether or not \( x \) is in a clique buffer. The deduction that a.a.s. the embedding \( \psi_{t_{\text{RGAend}}} \) has (PRGA2) follows exactly as in the proof of Lemma 3.1 (RGA lemma), and we do not repeat it.

It remains to establish (PRGA5). For this we require the following claim, which shows that if vertices \( v \) and \( v' \) are common candidates for too many \( x \in X_r^{\text{buf}} \) for (PRGA5), then the reason is that they have an exceptionally large common \( \Gamma \)-neighbourhood in some cluster.

Claim 4.7. Asymptotically almost surely at the termination of Algorithm 3 the following holds. For each \( i \) and pair of vertices \( v, v' \in V_i \) such that \( \deg^\Gamma_G(v, v'; V_j) \leq \frac{1}{10} \frac{1}{20} \mu \zeta d^3 (dp - \epsilon' p) \Delta n \)
(p + εp)^2|V_j| whenever ij \in R', if b is such that X_i^\text{buf} is a degree-b buffer, we have

\[ |\{x \in X_i^\text{buf} : v, v' \in C(x)\}| \leq 24\mu\Delta^2(20\mu^{-1}\zeta^{-1}d^{-\Delta})^b\epsilon^2p^b|X_i|. \]  

(34)

To prove this we show that for each x \in X_i^\text{buf}, however previous vertices are embedded, it is not too likely that N_H(x) is embedded to N_T^*(v, v'), and apply Lemma 2.2 and the union bound to deduce (34) for all i, v, and v'. In turn, to prove the desired upper bound on the probability of embedding N_H(x) to N_T^*(v, v'), we analyse the embedding of the vertices N_H(x) = \{y_1, \ldots, y_b\} one by one. We would expect that in each case roughly a \epsilon^2p^2-proportion of U(y_i) is contained in N_T^*(v, v'), and if this is the case at each step we obtain the desired upper bound. If not, the reason is that a previously embedded vertex—which must be one of the y_j, since we have not yet embedded any other neighbouring vertices—was ‘badly embedded’. Using the LNS property, we show this is an unlikely event and again obtain the desired upper bound. In fact, this is a slight oversimplification: we have to separate the cases that exactly one vertex is embedded badly (which implies we still have only at worst a probability p of embedding future vertices of N_H(x) to N_T^*(v, v')) or that more than one vertex is embedded badly, in which case we might have probability 1 of embedding future vertices of N_H(x) to N_T^*(v, v'), but this is counterbalanced by the unlikelihood of embedding two vertices badly.

**Proof.** We require

\[ \epsilon \leq \min(2^{-\Delta}, \tfrac{\mu\epsilon\Delta}{100\Delta}) \quad \text{and} \quad p \geq 20\mu^{-1}n^{1/(2\Delta)}. \]

Let v, v' \in V_i be such that \text{deg}_F(v, v', V_j) \leq (p + \epsilon)p^2|V_j| for each j such that ij \in R'. Suppose that X_i^\text{buf} is a degree-b buffer and b \in [\Delta].

Since any two vertices of X_i^\text{buf} are at distance at least ten in H by (H2), it is enough to show for any one vertex x \in X_i^\text{buf}, by considering the embedding of N_H(x), that the probability of v, v' \in C(x), conditioning on the history up to but not including the embedding of N_H(x), is at most 3\Delta^2(20\mu^{-1}\zeta^{-1}d^{-\Delta})^b\epsilon^2p^b. If we show this, then, since the vertices of N_H(x) are embedded consecutively, by Lemma 2.2, applied with \delta = 1 and using |X_i^\text{buf}| = 4\mu|X_i|, we see that (34) holds with probability at least 1 - \exp(-4\mu\epsilon^2(20\mu^{-1}\zeta^{-1}d^{-\Delta})^b\epsilon^2p^b|X_i|), and taking a union bound over at most \epsilon^2n choices of v, v' we conclude that by choice of p the conclusion of the claim a.a.s. holds as desired.

We now show that the probability of v, v' \in C(x) is at most 3\Delta^2(20\mu^{-1}\zeta^{-1}d^{-\Delta})^b\epsilon^2p^b conditioning on any \psi_T where T = \tau(x) - |N_H(x)| - 1 is the time immediately before the first vertex of N_H(x) is embedded. Observe that v, v' \in C(x) can occur only if N_H(x) is embedded into N_T^*(v, v'). If y \in N_H(x) \cap X_j, then the probability that y is embedded into N_T^*(v, v') is, using (INV3), at most

\[ \frac{|U_{\tau(y)-1}(y) \cap N_T^*(v, v')|}{\frac{1}{\mu}(d\mu)^{\tau(y)}|V_j|}. \]

Thus, we would like to show that the numerator is at most (p + \epsilon)p^2 + \tau(y)|V(y)| for all y \in N_H(x), since by multiplying conditional probabilities 10\mu^{-1}\zeta^{-1}(p + \epsilon)p^2 we would obtain the desired upper bound. Of course, this does not always happen, since it may be the case that a neighbour z of y in H with \tau(z) < \tau(y) was embedded ‘badly’, that is, to a vertex with more than

\[ (p + \epsilon)p|U_{\tau(z)-1}(y) \cap N_T^*(v, v')| \]

neighbours in U_{\tau(z)-1}(y) \cap N_T^*(v, v'). Because \tau satisfies Lemma 2.26(a), any such z is in N(X_i^\text{buf}). Since x is at distance at most 2 from z in H, and at distance at least 5 from any other vertex of X_i^\text{buf} by (H4), we see that z \in N_H(x). Thus, if exactly
\[ s \text{ many neighbours } z \text{ of } y \text{ with } \tau(z) < \tau(y) \text{ are embedded badly then we see that with (35) the following inequality} \]
\[ |U_{\tau(y)}(y) \cap N_1^{\epsilon}(v, v')| \leq (p + \epsilon p)^{\tau(y) - s}|N_1^{\epsilon}(v, v')| \leq (p + \epsilon p)^{\tau(y) + 2 - s}|V(y)| \]
holds. We separate in what follows three cases.

First, no vertices in \( N_H(x) \) are embedded badly. Then the probability that all \( b \) vertices in \( N_H(x) \) are embedded into \( N_1^{\epsilon}(v, v') \), conditioning on \( \psi_T \), is at most \((20\mu^{-1} \varepsilon p^2 d^{-\Delta} \rho^2)^b\) by choice of \( \varepsilon \).

Second, exactly one vertex in \( N_H(x) \) is embedded badly. We let the vertices of \( N_H(x) \) be \( y_1, \ldots, y_b \) in order of \( \tau \). Let us suppose that \( y_1 \) is the vertex embedded badly. Then the probability that the first \( \ell - 1 \) vertices of \( N_H(x) \) are embedded into \( N_1^{\epsilon}(v, v') \), conditioning on \( \psi_T \), is at most \((20\mu^{-1} \varepsilon p^2 d^{-\Delta} \rho^2)^{\ell - 1}\) by choice of \( \varepsilon \) (since these are not badly embedded and conditional probabilities multiply).

We now estimate the probability of \( y_1 \) being embedded badly, conditioning on \( \psi_T \) and on the embeddings of the previous \( \ell - 1 \) vertices not being bad. Observe that, since there has been no previous bad embedding, we have for each \( \ell' > \ell \)
\[ |U_{\tau(y_\ell)}(y_\ell) \cap N_1^{\epsilon}(v, v')| \leq (p + \epsilon p)^{\tau(y_\ell) - 1}|V(y_\ell)|, \]
and since \( \pi_{\tau(y_\ell)}(y_\ell) \leq \ell - 1 \leq \Delta - 2 \), the right hand side is, by choice of \( \varepsilon \), at least \( \epsilon p^2 n_{v_1} \). By LNS\( (\varepsilon, r_1, \Delta) \) the number of vertices in \( \Gamma \) with more than \((p + \epsilon p)^{\tau(y_\ell) - 1}|V(y_\ell)| \) neighbours into \( U_{\tau(y_\ell)}(y_\ell) \cap N_1^{\epsilon}(v, v') \) is at most \( \epsilon p^{2 \Delta - 1} n_{v_1} \). Since \( \pi(y_1) \leq \ell - 1 \), by (INV3) the probability that \( y_1 \) is embedded to such a bad vertex is at most
\[ \frac{\Delta \epsilon p^{2 \Delta - 1} n_{v_1}}{\mu \zeta(dp)^{\Delta - 1}|V(y_\ell)|} / 10 \leq p^{2 \Delta - \ell} \]
where the inequality is by choice of \( \varepsilon \).

Finally, the probability that the last \( b - \ell \) vertices are embedded into \( N_1^{\epsilon}(v, v') \), conditioning on the previous embeddings, is at most \((20\mu^{-1} \varepsilon p^2 d^{-\Delta} \rho^2)^{b - \ell}\) by choice of \( \varepsilon \). Indeed, for each vertex \( \ell' > \ell \) we have by (36) that
\[ |U_{\tau(y_\ell)}(y_\ell) \cap N_1^{\epsilon}(v, v')| \leq (p + \epsilon p)^{\tau(y_\ell) + 1}|V(y_\ell)| \]
since we condition on exactly one neighbour of \( y_\ell \) being embedded badly. This gives the conditional probability at most \( 20\mu^{-1} \varepsilon p^2 d^{-\Delta} \rho^2 \). The conditional probabilities multiply, and taking the union bound over the choices of \( \ell \), we see that the probability of this case occurring and all vertices of \( N_H(x) \) being embedded to \( N_1^{\epsilon}(v, v') \), conditioning on \( \psi_T \), is at most
\[ \Delta \cdot (20\mu^{-1} \varepsilon p^2 d^{-\Delta} \rho^2)^{\ell - 1} \cdot p^{2 \Delta - \ell} \cdot (20\mu^{-1} \varepsilon p^2 d^{-\Delta} \rho^2)^{b - \ell} \leq \Delta (20\mu^{-1} \varepsilon p^2 d^{-\Delta} \rho^2)^{b - \ell - 2} \leq \Delta^2 (20\mu^{-1} \varepsilon p^2 d^{-\Delta} \rho^2)^{b} \]
where the last inequality follows since \( \Delta \geq 2 \) and \( b \leq \Delta \).

In the third case, at least two vertices in \( N_H(x) \) are embedded badly. Suppose that the first two badly embedded vertices are the \( j \)th and \( k \)th vertices. Then the same logic as above, in particular the inequality (36), tells us that the probability that this case occurs and all vertices of \( N_H(x) \) are embedded to \( N_1^{\epsilon}(v, v') \) is at most
\[ \Delta^2 \cdot (20\mu^{-1} \varepsilon p^2 d^{-\Delta} \rho^2)^{j - 1} \cdot p^{2 \Delta - j} \cdot (20\mu^{-1} \varepsilon p^2 d^{-\Delta} \rho^2)^{k - j - 1} \cdot p^{2 \Delta - k} \leq \Delta^2 (20\mu^{-1} \varepsilon p^2 d^{-\Delta} \rho^2)^{b - 2} p^{4 \Delta - 3} \]
Putting these three cases together, and using the fact that \( 4\Delta - 3 \geq 2b \) since \( \Delta \geq 2 \) and \( b \leq \Delta \), we conclude that the probability that \( v, v' \in C(x) \) is at most \( 3\Delta^2 (20\mu^{-1} \varepsilon p^2 d^{-\Delta} \rho^2)^{b} \) as desired.
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Suppose now that the good event of Claim 4.7 holds. For each $i \in [r]$, there may be ‘exceptional’ vertices $v \in V_i$ which have more than $(p + \varepsilon)p|V_j|$ neighbours in $V_j$ for some $j$ such that $ij \in R'$, but by NS($\varepsilon, r_1, \Delta + 1$) there are at most $\Delta_{R'} \varepsilon p^2 n / r_1 \leq \varepsilon' p^2 |V_i|$ such vertices. If $v$ is not exceptional—that is, it has at most $(p + \varepsilon)|V_j|$ neighbours in each $V_j$ with $ij \in R'$—then again there may be ‘exceptional’ vertices $v' \in V_i$ such that $v'$ has more than $(p + \varepsilon)p^2 |V_j|$ common neighbours with $v$ in $V_j$ for some $j$ such that $ij \in R'$, but again by NS($\varepsilon, r_1, \Delta + 1$) there are at most $\Delta_{R'} \varepsilon p^2 n / r_1 \leq \varepsilon' p^2 |V_i|$ such. Because the good event of Claim 4.7 holds, for non-exceptional pairs of vertices $v, v'$ we have the bound given in (PRGA 5). Because the good event of Claim 4.7 holds a.a.s., property (PRGA 5) is a.a.s. obtained. This completes the proof of Lemma 4.1. □
CHAPTER 5

Improved bounds for degenerate graphs

5.1. The RGA lemma and the proof of the blow-up lemma

In this section we prove Lemma 1.23. We have already seen most of the ideas in the proofs of Lemmas 1.21 and 1.25. We use the same General Setup, and we continue to obtain it using Lemma 2.22. However, rather than defining an order $\tau$ putting the buffer vertices $X^{buf}$ at the end and their neighbours at the front, as we did in the proofs of Lemmas 3.1 and 4.1, we are supplied with an order $\tau$ which we will modify only in that we move $X^{buf}$ to the end of the order. In particular, the neighbours of buffer vertices need not appear early in $\tau$. The reason for this is that moving vertices in the order could result in substantially increasing $\pi^\tau(x)$ for some vertices $x$. It is easy to check that moving $X^{buf}$ to the end of the order only increases $\pi^\tau(x)$ for $x \in X^{buf}$, thus preserves conditions (ORD1)–(ORD3) for non-buffer vertices.

We will use a modification of Algorithm 3 to perform this embedding. The modification consists of handling the exceptional vertices $X^e$ differently, which allows us to deal with a few vertices $x$ with $\pi^\tau(x)$ significantly larger than normal; such vertices appear in applications. We can show that this algorithm succeeds in embedding all the vertices of $X^{main}$, in the order $\tau$, and that it returns a good partial embedding with the properties detailed in the lemma below with positive probability. Since we will have $X^{main} = V(H) \setminus X^{buf}$, we can complete the proof using Lemma 3.4 to embed the buffer vertices in much the same way as in the proof of Lemma 1.21.

Note that property LCON which appears in the following lemma plays the same role as LNS in the proof of Lemma 4.1. It is easier to work with, and allows for linearly many image restrictions (but it is not in general true in bijumbled graphs).

**Lemma 5.1.** We assume the General Setup. Suppose that an (exceptional) set $X^e$ with $|X^e| \leq \frac{1}{2} \epsilon_p \max_{x \in X^e} \pi^\tau(x) n/\ell_1$ is given. Suppose that $D \geq 1$, and $\tau$ is a $(D,p,\frac{1}{2} \epsilon n/\ell_1)$-bounded order for $H$, $\tilde{X}$, $I$ and $J$ with this exceptional set $X^e$. Suppose furthermore that the vertices $X^{buf}$ form the final segment of $\tau$. Suppose that $\Gamma$ has properties $\text{NS}(\epsilon, \ell_1, D)$, $\text{RI}(\epsilon, (\epsilon, a, b, \epsilon'), \ell_1, \ell_1, D)$ and $\text{LCON}(\epsilon, \ell_1, D)$. Then there is a good partial embedding $\psi_{RGA}$ of $H$ into $G$ with the following properties for each $i \in [r]$. Let $b$ be such that $X^{buf}_i$ is a degree-$b$ buffer.

**DRGA1** Every vertex in $X^{main}_i$ is embedded to $V^{main}_i \cup V^{q}_i \cup V^{c}_i$ by $\psi_{RGA}$, and no vertex in $X^{buf}_i$ is embedded.

**DRGA2** For every set $W \subseteq V_i$ of size at least $\ell |V_i|$, there are at most $\ell |X_i|$ vertices in $X^{buf}_i$ with fewer than $(dp)^b |W|/2$ candidates in $W$.

**DRGA3** Every vertex in $V_i$ is a candidate for at least 

$$4^{-10 \Delta^3} 2^{-1000D^2 \mu^{-1} \gamma^{-1} \Delta^{-1} D^2 \mu^{-1} \Delta^{-3} \mu_p^b |X_i|}$$

vertices of $X^{buf}_i$. 


We will prove this lemma in Section 5.2. Assuming Lemma 5.1, we are in a position to prove Lemma 1.23. This amounts to a verification that the conditions of Lemma 1.23 suffice to apply Lemmas 2.17, 2.22, 3.4 and 5.1.

Proof of Lemma 1.23. First we choose constants as follows. Given $\Delta, \Delta_{R^*}, \Delta_J$ and $D$ integers, $\alpha^{BL}, \zeta^{BL}$ and $d > 0$, and $\epsilon^{BL} > 1$, we set $\vartheta = \epsilon^{BL}$, $\Delta_{R^*} = 8(\Delta + \Delta_J)^{10}\Delta_{R^*}$, $\alpha = \frac{1}{2}\delta^{BL}$, $\zeta = \frac{1}{2}\delta^{BL}$ and $\kappa = 2\kappa^{BL}$. We now choose

$$
\mu = \frac{\alpha}{200000\Delta^{\alpha}\Delta^{10}(\Delta + 2)}, \quad \varrho = \frac{\mu^2\zeta^{d+2}}{100006(n+1)} 4^{-2000\Delta^{3}\mu^{-1}\zeta^{-1}d^{-d}}
$$

and

$$
\epsilon' = \frac{\mu^2\zeta^{d+1}}{1000\Delta^{3}\kappa^{2}} 2^{-280(D+1)\mu^{-1}\zeta^{-1}d^{-d}}.
$$

Now for input $D$ and $d, \epsilon'$, Lemma 2.17 returns constants $(\epsilon_{a,b})$ and $\epsilon_{L,(b)}>0$. We set

$$
\epsilon = \min \left( \epsilon_{L,(b)}, \frac{\varrho}{\kappa} \right).
$$

We let $\epsilon^{BL} = \frac{1}{10}(\Delta + \Delta_J)^{-10}\epsilon$ and $\varrho^{BL} = \frac{1}{10}(\Delta + \Delta_J)^{-10}\varrho$. Now Lemma 1.23 returns $\epsilon^{BL}$ and $\varrho^{BL}$. Given $p^{BL}$, we let $r_1^{BL} = 8(\Delta + \Delta_J)^{10}p^{BL}$. We choose $C$ sufficiently large for Lemma 2.17 with input $D, d', \epsilon'$, $r_1$ and $\varrho$, and such that

$$
C \geq \frac{10^8 4^{10\Delta^3} (\Delta + \Delta_J)^4 \kappa r_1^2 2^{280(D+1)\mu^{-1}\zeta^{-1}d^{-d}}}
$$

Given $p \geq C (\log n)^{1/D}$, Lemma 2.17 (items (a), (b), (c) and (d)) states that a.a.s. $\Gamma = G_{p,r}$ has properties NS$(\epsilon,r_1,D), R_I(\epsilon,\epsilon_{a,b}), \epsilon', d, r_1, D., CON(\varrho,r_1,D)$ and LCON$(\epsilon,r_1,D)$ respectively. From now on we will assume $\Gamma$ is an $n$-vertex graph which satisfies these four properties.

Given a graph $R_{BL}$ on $r_{BL} \leq r_1^{BL}$ vertices, and a spanning subgraph $R_{BL}$ with $\Delta(R_{BL}) \leq \Delta_{R^*}$, and $H$ and $G \subseteq \Gamma$ with vertex partitions $X^{BL}$ and $Y^{BL}$, families of image restrictions $Z^{BL}$ and of image restricting vertices $J$, a family of potential buffer vertices $X^{BL}$, and an exceptional set $X^e$, suppose that the conditions of Lemma 1.23 are satisfied. Then Lemma 2.22 gives a graph $R$ on $r \leq r_1$ vertices, a spanning subgraph $R'$ with $\Delta(R') \leq \Delta_{R^*}$, and $\kappa$-balanced size-compatible partitions $X$ and $Y$ of $H$ and $G$ respectively, each part having size at least $n/(\kappa r_1)$, together with a family $X$ of potential buffer vertices and $X$ of image restrictions, subsets $X^e_{i}$ for each $i \in [r]$, and partitions $V_i = V_i^{main} \cup V_i^{sq} \cup V_i^{e} \cup V_i^{buf}$ for each $i \in [r]$, which satisfy the General Setup.

We now modify the provided order $\tau^{BL}$ on $V(H)$ by moving the set $X^{buf}$ to the end of the order, obtaining a new order $\tau$. Since $\epsilon^{BL}/r_1^{BL} = \frac{1}{2}\epsilon/n$, the order $\tau$, which is $(D,p,\epsilon^{BL}/r_1^{BL})$-bounded for $H, X^{BL}$, $Z^{BL}$ and $J$ with the exceptional set $X^e$ of size at most $\epsilon^{BL}p^\max_{x \in X} \pi^*(x)n/r_1^{BL} = \frac{1}{2}\epsilon p^\max_{x \in X} \pi^*(x)n/r_1$, is also $(D,p,\frac{1}{2}\epsilon/n)$-bounded for $H, X, I$ and $J$.

We let $X^{main} = V(H) \setminus X^{buf}$. We now begin the embedding of $H$ into $G$. By Lemma 5.1, there is a good partial embedding $\psi^{RGA}$ with the properties stated in that lemma. By (GPE3), condition (CPM1) of Lemma 3.4 is satisfied, while condition (CPM2) holds by (DRGA2) and (CPM3) follows from (DRGA3) with

$$
\delta = 4^{-10\Delta^3} 2^{-1000D^2\mu^{-1}\zeta^{-1}d^{-d}} d^{2D^2} \Delta^{-3} \mu.
$$

Therefore, we can find an embedding $\psi'$ extending $\psi^{RGA}$ which embeds $X^{buf}$ to $V_i \setminus \text{Im}(\psi^{RGA})$. Repeating this for each $i \in [r]$, which we may do since $X^{buf}$ is an independent set in $H$, we obtain the desired embedding of $H$ into $G$. \qed
5.2. Proof of the degenerate graph RGA lemma

We prove Lemma 5.1 by analysing Algorithm 4 below. The analysis is quite similar to what we saw before in the proofs of Lemmas 3.1 and 4.1 (indeed, the main difference is that we are more careful to bound powers of $p$ using $D$ rather than just $\Delta$), so we will be brief and highlight the differences. The only difference between Algorithm 3 and Algorithm 4 is that vertices of $X^e$ are embedded into $V^c$ rather than $V_{\text{main}}$ or $V^q$.

**Algorithm 4:** Random greedy algorithm for degenerate graphs

```
Input: $G \subseteq \Gamma$ and $H$ with partitions satisfying the General Setup; an ordering $\tau$ on $X_{\text{main}}$

$t := 0$ ;
$\psi_0 := \emptyset$ ;
$Q_0 := \{ x \in V(H) : |I_x| < \frac{1}{2} \mu(d-\varepsilon)^{\lfloor |J_x|/p \rfloor} |V_{\text{main}}(x)| \}$ ;
repeat

let $x \in X_{\text{main}} \setminus \text{Dom}(\psi_t)$ be the next vertex in the order $\tau$  ;
if $x \in Q_t \setminus X^e$ and $|A_t^\tau(x) \setminus B_t(x)| < \frac{1}{2} \mu \zeta(d \mu \zeta) |\pi_t(x) \setminus V(x)|$ then
| halt with failure ;
end
if $x \in Q_t \setminus X^e$ and $|A_t^\tau(x) \setminus B_t(x)| < \frac{1}{2} \mu \zeta(d \mu \zeta) |\pi_t(x) \setminus V(x)|$ then
| halt with failure ;
end

choose $v$ uniformly at random in

\[
\begin{cases}
A_t^\tau(x) \setminus B_t(x) & \text{if } x \notin (Q_t \cup X^e) \\
A_t^\tau(x) \setminus B_t(x) & \text{if } x \in Q_t \setminus X^e \\
A_t^\tau(x) \setminus B_t(x) & \text{if } x \in X^e
\end{cases}
\]

；
$\psi_{t+1} := \psi_t \cup \{ x \to v \}$ ;
$Q_{t+1} := Q_t$ ;
forall the $y \in X_{\text{main}} \setminus \text{Dom}(\psi_{t+1})$ do
if $|(A^\tau_{t+1}(y)) \setminus B_t(x)| < \frac{1}{2} \mu(d-\varepsilon)^{\lfloor \pi_{t+1}(y) \rfloor} p \pi_{t+1}(y) |V_{\text{main}}(y)|$ then
| $Q_{t+1} := Q_{t+1} \cup \{ y \}$ ;
end
end

$t := t + 1$ ;
until Dom($\psi_t$) = $X_{\text{main}}$;

$t_{\text{RGAend}} := t$ ;
```

We will see that (ORD1) (see Definition 1.22 of $(D, p, m)$-bounded order) is precisely what we need to make Lemma 2.25 work with properties NS($\varepsilon, r_1, D$) and RI($\varepsilon, (e_a, b), \varepsilon', d, r_1, D$), so allowing us to prove that $B_t(x)$ is always much smaller than $A_t(x)$. As in the proof of Lemma 4.1, our first task is to show that the algorithm a.a.s. completes successfully. Again by Lemma 2.27 we can show that the queue remains small. We can also show, using (ORD2), that all vertices $x \in X_t \setminus X^e$ which enter the queue have $\pi^e(x) \leq D/2$. We embed the vertices of $X^e \cap X_t$ which enter the queue greedily into $V^c$, and there are so few such vertices that this is guaranteed to succeed. Property LCON($\varepsilon, r_1, D$) turns out to be what we need to verify that the queue embedding of the remaining vertices a.a.s. is successful, using the same strategy as in the proof of Lemma 4.1.
At this stage we have (DRGA1) simply because the algorithm completes, while (DRGA2) follows from Lemma 2.27. It remains to prove (DRGA3), which is where we need to use (ORD3). Here we deviate from the strategy we saw previously. We can no longer assume that neighbors of buffer vertices appear early in \(\tau\), and thus we have to prove that for any \(v \in V(G)\), even towards the end of the embedding, it is still reasonably likely that neighbors of buffer vertices are embedded to \(N_G(v)\). We will see (Claim 5.6 below) that properties NS(\(\varepsilon, r_1, D\)), RI(\(\varepsilon, (\varepsilon_a, b)\), \(\varepsilon', d, r_1, D\)) and LCON(\(\varepsilon, r_1, D\)) allow us to show that a.a.s. given any \(\ell \leq D - \max_{x \in X_{\text{main}} \setminus X_{\text{main}}} \pi^*(x)\) vertices \(v_1, \ldots, v_\ell\) of \(G\) such that \(N_G^*(v_1, \ldots, v_\ell; V_{\text{main}})\) is small, the set \(N_G^*(v_1, \ldots, v_\ell; V_{\text{main}})\) is never completely filled by \(I_m(v)\). The idea of the proof remains similar to that of Lemma 2.28. We show, as there, that when the first small fraction of the vertices in the order \(\tau\) are embedded, only at most half of \(N_G^*(v_1, \ldots, v_\ell; V_{\text{main}})\) is covered. But then we repeat this, showing that the next small fraction of \(\tau\) covers only at most half of what remains, and so on, so that all vertices are embedded what remains is an exponentially small, bounded away from zero, fraction of the original \(N_G^*(v_1, \ldots, v_\ell; V_{\text{main}})\). Once we have shown this, completing the proof that \(\tau_H(x)\) is not too unlikely to be embedded to \(N_G(v)\) for any \(x \in X_{\text{buf}}\) and \(v \in V(x)\) can be done along similar lines to the proof of Lemma 2.29.

Proof of Lemma 5.1. We require

\[
\mu \leq \frac{1}{\delta}, \quad \rho \leq \frac{\mu^2 \Delta^2 d^2 D}{10000(\Delta + 1)}, \quad \varepsilon' \leq \frac{\mu \varepsilon d^2}{1000\Delta^2 \varepsilon^{-1} \kappa^{-1} \rho^{-1}} 2^{-280(D + 1) \mu^{-1} \varepsilon^{-1} d^{-D}}, \quad \varepsilon \leq \frac{d \varepsilon'}{n}
\]

and

\[
p \geq \frac{10000 \varepsilon d^2 \Delta \kappa^{-1} \rho^{-1} \mu^{-1}}{\varepsilon \mu \varepsilon d^2 \Delta \kappa^{-1} \rho^{-1} \mu^{-1} (\log n)^{1/2}}.
\]

We run Algorithm 4, using the order \(\tau\) supplied to Lemma 5.1. We claim that we can apply Lemma 2.25 to bound \(B_{\tau(x)}(x)\). To see this, we consider three cases.

If \(x\) has no neighbors after \(\tau(x)\) in \(\tau\), then Lemma 2.25 states that \(B_{\tau(x)}(x) = \emptyset\) without requiring any property of \(\Gamma\).

If \(x\) has a neighbor \(y\) with \(\tau(x) < \tau(y)\), then (ORD1) states that \(\pi^*(x) \leq D - 1\), so in particular \(\pi_{\tau(x)-1}^*(x) \leq D - 1\), satisfying (i). Furthermore, (ORD1) states that \(\pi^*(y) \leq D\), so \(\pi_{\tau(x)-1}^*(y) \leq D - 1\) since \(x\) is not embedded, satisfying (ii).

Now suppose that \(xy, yz \in E(H)\) for some unembedded \(y\) and \(z\). There are two cases to consider. First, if \(\tau(y) < \tau(z)\) then (ORD1) states that \(\pi^*(y) \leq D - 1\) and \(\pi^*(z) \leq D\). Since \(x\) and \(y\) are unembedded at time \(\tau(x) - 1\), we thus have \(\pi_{\tau(x)-1}^*(y) \leq D - 2\) and \(\pi_{\tau(x)-1}^*(z) \leq D - 1\), as required by (iii). Second, if \(\tau(y) > \tau(z)\), then \(\pi_{\tau(x)-1}^*(y) \leq \pi^*(y) \leq D - 1\) (BY (ORD1)), while since \(\pi^*(y) \leq D\) and both the vertices \(x\) and \(z\) are unembedded, we have \(\pi_{\tau(x)-1}^*(y) \leq D - 2\), again as required by (iii).

Lastly, if there are unembedded vertices \(y\) and \(z\) such that \(xy, yz, xz\) are all edges of \(H\), then we claim \(\pi_{\tau(x)-1}^*(x) = \pi_{\tau(x)-1}^*(y) = \pi_{\tau(x)-1}^*(z) \leq D - 2\). To see this, suppose \(\tau(y) < \tau(z)\). By (ORD1) we have \(\pi^*(x) \leq D - 2\), \(\pi^*(y) \leq D - 1\) and \(\pi^*(z) \leq D\). Since \(x\) and \(y\) are unembedded, we have \(\pi_{\tau(x)-1}^*(y) \leq D - 1 - 1 = D - 2\), and \(\pi_{\tau(x)-1}^*(z) \leq D - 2\). Thus condition (iv) is satisfied.

We conclude that \(|B_{\tau(x)}(x)| \leq 20D^2 \varepsilon' p^\tau(x) |V(x)|\) for each \(x \in X_{\text{main}}\).

As before, the first task is to show that certain invariants are maintained. These are identical to the invariants seen in the previous two RGA lemmas, but are repeated here for the reader’s convenience.

**Claim 5.2.** The following hold at each time \(t\) in the running of Algorithm 4.

- **(INV1)** \(\psi_t\) is a good partial embedding.
- **(INV2)** Either \(|A_t^\text{main}(x)| \geq \frac{1}{4} \mu (d - \varepsilon') \pi_t^*(x) p^\tau(x) |V_{\text{main}}(x)|\) or \(x \in Q_t\).
(INV3) When we embed \( x \) to create \( \psi_{i+1} \), we do so uniformly at random into a set of size at least \( \frac{1}{2} \mu(d - \epsilon')p^{\tau}(x)|V(x)| \).

**Proof.** Algorithm 4 maintains (INV1) and (INV2) by definition. Since \( \frac{1}{2} \mu(d - \epsilon')p^{\tau}(x) - 20\Delta \epsilon' p^{\tau}(x) > \frac{1}{2} \mu(d'p^{\tau}(x))p^{\tau}(x) \) by choice of \( \epsilon' \) for each \( x \in X^\text{main} \), by Lemma 2.25 it also maintains (INV3).

As in the previous proofs, the conditions of Lemma 2.27 are met, so we conclude that with probability at least \( 1 - 2^{-n/(\kappa r^3)} \), for each \( i \in [r] \) and time \( t \) when Algorithm 4 is running, we have \( |Q_t \cap X_i| < q_i |X_i| + |X_i^\ast| \leq 2p_i |X_i| \) by (G6).

We now want to show that Algorithm 4 a.a.s. does not halt with failure. Since we know that \( |\beta((\tau - 1)(x))| < 20\Delta \epsilon' p^{\tau}(x)|V(x)| \), by (GPE3) and (G5) it is enough to show that a.a.s. we have

\[
\begin{align*}
|A_0^\tau(x)_{-1}(x)| &\geq \frac{1}{2}|C_0^\tau(x)_{-1}(x)| \quad \text{for} \quad x \in X^\text{main} \cap Q_{\tau}(x)_{-1} \setminus X^c, \quad \text{and} \\
|A_c^\tau(x)_{-1}(x)| &\geq \frac{1}{2}|C_c^\tau(x)_{-1}(x)| \quad \text{for} \quad x \in X^\text{main} \cap Q_{\tau}(x)_{-1} \setminus X^c.
\end{align*}
\]

The latter of these is easy, since we embed only vertices of \( X^c \) into \( V^c \), and \( |X^c| \leq \frac{1}{2} \mu p^{\max(x)\pi(x)n/r_1} \), which by (GPE3) and choice of \( \epsilon \) is smaller than \( \frac{1}{2} |C_c^\tau(x)_{-1}(x)| \). We now establish the former inequality.

First, we show that we need only concern ourselves with \( x \) such that \( \pi(x) \leq D/2 \).

**Claim 5.3.** For each \( x \in X^\text{main} \setminus X^c \), if \( \pi(x) > D/2 \) then there is no time \( t \) such that \( x \in Q_t \).

**Proof.** We require \( \epsilon' \leq \frac{1}{2} d^2 \mu \) and \( \epsilon \leq \kappa^{-1} \epsilon' \).

Suppose \( \pi(x) > D/2 \) and \( x \not\in X^c \). Let \( i \) be such that \( x \in X_i \). By (ORD2), \( x \) is not image restricted, and all \( \pi(x) \) neighbours of \( x \) which precede \( x \) in \( \tau \) occur in the \( \frac{1}{2} \mu p^{\pi(x)n/r_1} \leq \epsilon' p^{\pi(x)}|X_i| \) places of \( \tau \) before \( x \). Because \( x \) is not image restricted, at time \( t \leq \tau(x) - \epsilon' p^{\pi(x)}|X_i| \) we have \( C_{\tau}(x)|V_i^\text{main} \setminus \im(\psi_i) \). Because \( |X_i| = |V_i^\text{main}| - \mu |V_i| \), it follows that \( |A_{\tau}^\text{main}(x)| \geq \mu |V_i| \), and in particular \( x \) does not enter \( Q_t \) at time \( t \).

Now suppose \( t = \tau(x) - \epsilon' p^{\tau}(x)|X_i| + s \) for some 0 \( \leq s \leq \epsilon' p^{\tau}(x)|X_i| - 1 \). We claim that

\[
|A_{\tau+1}^\text{main}(x)| \geq (dp - \epsilon' p^{\tau}(x))\mu |V_i| - s.
\]

Indeed, for \( s = 0 \) we have \( \pi(x) = 0 \) and the statement \( |A_{\tau}^\text{main}(x)| \geq \mu |V_i| \) was established above. Now suppose the statement holds for some \( s \geq 0 \). At time \( t = \tau(x) - \epsilon' p^{\pi(x)}|X_i| + s \) we embed a vertex \( y \) to obtain \( \psi_{i+1} \). If \( y \) is neither a neighbour of \( x \) nor in \( X_i \), then \( A_{\tau+1}^\text{main}(x) = A_{\tau+1}^\text{main}(x) \) and the statement holds for \( s + 1 \). If \( y \) is a neighbour of \( x \), then we embed \( y \) to a vertex of \( A_t(y) \setminus B_t(y) \), and in particular to a vertex \( v \) such that \( \deg_G(v; A_{\tau}^\text{main}(x)) \geq (d - \epsilon') p |A_{\tau}^\text{main}(x)| \). We conclude that

\[
|A_{\tau+1}^\text{main}(x)| \geq (d - \epsilon') p ((dp - \epsilon' p^{\tau}(x))\mu |V_i| - s) > (dp - \epsilon' p^{\tau}(x))\mu |V_i| - s - 1.
\]

as desired. Finally, if \( y \in X_i \), then we have \( A_{\tau}^\text{main}(x) = A_{\tau}^\text{main}(x) \setminus \{y\} \), and again

\[
|A_{\tau+1}^\text{main}(x)| \geq (dp - \epsilon' p^{\tau}(x))\mu |V_i| - s - 1.
\]

Thus for each \( \tau(x) - \epsilon' p^{\tau}(x)|X_i| \leq t \leq \tau(x) - 1 \) we have

\[
|A_{\tau}^\text{main}(x)| \geq (dp - \epsilon' p^{\tau}(x))\mu |V_i| - \epsilon' p^{\tau}(x)|X_i| \geq \frac{1}{2} (dp - \epsilon' p^{\tau}(x))\mu |V_i|,
\]

where the final inequality is by choice of \( \epsilon' \). It follows that \( x \) does not satisfy the condition to enter \( Q_t \). Since we embed \( x \) at time \( \tau(x) - 1 \), we conclude that as desired there is no time \( t \) such that \( x \in Q_t \).
5.2. PROOF OF THE DEGENERATE GRAPH RGA LEMMA

As in the proof of Lemma 4.1 the next step is to establish a ‘sum condition’. In fact, we establish almost the same inequality as (29) in Claim 4.5 (the right hand side is identical, while the left hand side sum runs over all vertices outside $X^c$, including those $y$ with $J_y \neq \emptyset$), though we use a different analysis to do so.

CLAIM 5.4. Suppose that $\Gamma$ has LCON($\varepsilon, r_1, D$). Then the following statement holds.

For any $t$, any $1 \leq \ell \leq D/2$, any $i \in [r]$ and any vertices $v_1, \ldots, v_{\ell}$ such that
\[ \deg_T(v_1, \ldots, v_{\ell}; V_i) \geq \varepsilon' p^j |V_i| \]
we have
\[ \sum_{y \in X^\text{main} \cap Q_i \setminus X^c} \frac{|U_{\pi_{\tau}(y)-1}(y) \cap N^*_1(y; v_1, \ldots, v_{\ell}; V_i)|}{|A_{\pi_{\tau}(y)-1}(y) \setminus B_{\pi_{\tau}(y)-1}(y)|} \leq \frac{\mu \zeta d^\Delta \deg_T(v_1, \ldots, v_{\ell}; V_i)}{20}. \quad (37) \]

The proof of this amounts to checking that the images under our embedding of the $D$-tuples $(x; j, \tau)$ for $y \in X^\text{main} \cap Q_i \setminus X^c$ form a family of sets to which we can apply the LCON property. This property gives us the desired inequality.

PROOF. We require $\rho \leq \frac{d^2 \zeta^2 d^2 2^\Delta}{10000 (\Delta + 1)}$, $\varepsilon' \leq \mu \zeta (d/4)^D$ and $\varepsilon \leq \frac{\varepsilon'}{\kappa}$. Since the left hand side of (37) is increasing in $t$, we may assume $t$ is the time at which Algorithm 4 ends.

Given $1 \leq \ell \leq D/2$ and $i \in [r]$, let $v_1, \ldots, v_{\ell}$ be vertices such that $U := N^*_1(v_1, \ldots, v_{\ell}; V_i)$ has size at least $\varepsilon' p^j |V_i|$. By choice of $\varepsilon$ and $\varepsilon'$, we have $|U| \geq \varepsilon p^j n/r_1$. It follows that we can apply LCON($\varepsilon, r_1, D$) to bound the number of edges in $CG(\Gamma, U, F)$ for any $1 \leq j \leq D/2$ and family $F$ of pairwise disjoint $j$-sets in $V(\Gamma \setminus U)$. We set
\[ F_j := \{ \Pi_{\tau(x)}^{-1}(x) \cup J_x : x \in X^\text{main} \cap Q_i \setminus X^c, \pi^\tau(x) = j, \tau(x) \leq t \}. \]

Observe that $F_j$ is indeed a family of $j$-sets by definition of $\pi^\tau(x)$, that these sets are pairwise disjoint because no vertex of $H$ has more than one neighbour in $X_i$ by (H2), and element $F$ of $F_j$ intersects $U \subseteq V_i$ because these sets $F$ are images under $\psi$ of neighbours of vertices in $X_i$. Because $\varepsilon |U|, |X^\text{main} \cap Q_i| \leq 2 \rho |X_i|$, by choice of $\varepsilon$ and by LCON($\varepsilon, r_1, D$) we have
\[ e(CG(\Gamma, U, F_j)) \leq 7 \rho^j |U| (2 \rho |X_i|) = 14 \rho \rho^j |U| |X_i|. \]

By definition of $U_i(x)$ (see Section 2.3.5) we have the following equality
\[ e(CG(\Gamma, U, F_j)) = \sum_{x \in X^\text{main} \cap Q_i \setminus X^c: \pi^\tau(x) = j, \tau(x) \leq t} |U_{\tau(x)-1}(x) \cap U| \]
and hence
\[ \sum_{x \in X^\text{main} \cap Q_i \setminus X^c: \pi^\tau(x) = j, \tau(x) \leq t} |U_{\tau(x)-1}(x) \cap U| \leq 14 \rho \rho^j |U| |X_i| \]
for each $1 \leq j \leq D/2$. Observe that the $j = 0$ case of the same inequality is trivially true (indeed with 14 replaced by 2, though we will not use this): each vertex of $X^\text{main} \cap Q_i$ contributes at most $|U|$ to the sum. Now by (INV3) we have
\[ |A_{\pi_{\tau}(x)}^0(x) \setminus B_{\pi_{\tau}(x)}(x)| \geq \frac{1}{10} \mu \zeta (\rho p)^{\pi^\tau(x)} |V_i| \]
for each $x \in X_t^{\text{main}} \cap Q_t$, with $\tau(x) \leq t$. Finally, since any vertex $x$ in $X_t^{\text{main}} \cap Q_t$ satisfies $\pi^\tau(x) \leq D/2$ by Claim 5.3, we conclude that

$$\sum_{x \in X_t^{\text{main}} \cap Q_t \setminus X^e : \tau(x) \leq t} |\Delta^q_{\tau(x)}(x) \cap U| \leq \sum_{j=0}^{D/2} \sum_{x \in X_t^{\text{main}} \cap Q_t : \pi^\tau(x) = j, \tau(x) \leq t} \frac{14g|U||X_t|}{10\mu\zeta(dp)^{\pi^\tau(x)}|V_t|} \leq \frac{14(1 + D/2)g|U||X_t|}{10\mu\zeta(dp)^{D/2}|V_t|} \leq \frac{\mu dp^2}{20} |U|,$$

where the final inequality is by choice of $g$, as desired. \hfill \square

Now we have

**Claim 5.5.** A.a.s. for each $x \in X^{\text{main}} \setminus X^e$, at each time $t \leq \tau(x) - 1$ and before the termination of Algorithm 4, we have $|C_t^G(x) \cap \text{Im}(\psi_i)| < \frac{1}{2}|C_t^G(x)|$.

**Proof.** The proof of Claim 4.6, if we replace $X^{\text{main}}$ with $X^{\text{main}} \setminus X^e$, and replace (29) with (37), without handling image restricted vertices specially (since condition (37) deals with them), gives this claim verbatim. \hfill \square

As in the proof of Lemma 4.1, the conclusion of Claim 5.5 holds a.a.s., which implies that Algorithm 4 completes successfully. Again, the successful running implies (DRGA1), and the good event of Lemma 2.27 holding implies (DRGA2).

We would like to emphasise at this point that we have established all the desired conclusions of Lemma 5.1 except (DRGA3), and we have so far not used condition (ORD3), and not used the second part of (ORD1) (which states $\tau(x) \leq D_x - 1$ for $x \in N(\bar{X})$). It follows that one can establish a version of Lemma 5.1 omitting (DRGA3) given an order $\tau$ which need only satisfy the first part of (ORD1) (which states $\tau(x) \leq D_x$ for all $x \in V(H)$) and (ORD2). We will return to this point in Section 7.1.3.

We now turn to proving (DRGA3). First, we show that $G$-common neighbourhoods of at most $D - \max_{x \in X^{\text{main}} \setminus X^e} \pi^\tau(x)$ vertices which are large do not get completely filled at any time in the running of Algorithm 4. Note that the strange-looking number of vertices whose $G$-neighbourhoods we control comes from (ORD3). When we use this condition, we will want to control a $G$-common neighbourhood of a vertex $v \in V_t$ (which we eventually want to show is likely to be candidate for many vertices of $X^{\text{bifi}}$) together with the embedded images of some neighbours of a vertex $x \in N(X^{\text{bifi}})$. What this condition says is that we have the desired control for all the neighbours of $x$ which are embedded before $\tau(x) - \frac{1}{2} \rho D_q/r_1$; we will see that neighbours embedded after this time can be dealt with easily. This lemma replaces Lemma 2.28 which we used in the proofs of the previous two RGA lemmas.

**Claim 5.6.** Suppose that $\Gamma$ has LCON($\varepsilon, r_1, D$). Then a.a.s. the following holds at each time $t$. Given any $i \in [r]$, any $1 \leq \ell \leq D - \max_{x \in X^{\text{main}} \setminus X^e} \pi^\tau(x)$, and any vertices $v_1, \ldots, v_\ell$ of $G$, the following hold. If $\deg_G(v_1, \ldots, v_\ell; V_t^{\text{main}}) \geq (dp/2)^\ell |V_t^{\text{main}}|$, then

$$|N^*_G(v_1, \ldots, v_\ell; V_t^{\text{main}}) \cap \text{Im}(\psi_i)| \geq (1 - 4\mu)2^{-280(D+1)}\mu^{-1} \ell^{-1} \ell^{-1} \deg_G(v_1, \ldots, v_\ell; V_t^{\text{main}}).$$

If $\deg_G(v_1, \ldots, v_\ell; V_t^{\text{main}}) \geq (dp/2)^\ell |V_t^{\text{main}}|$, then

$$|N^*_G(v_1, \ldots, v_\ell; V_t^{\text{main}}) \setminus \text{Im}(\psi_i)| \geq \frac{1}{2} \deg_G(v_1, \ldots, v_\ell; V_t^{\text{main}}).$$

(38)

The proof of (38), as discussed at the beginning of this section, roughly amounts to repeating the argument of Lemma 2.28 several times. We show that, for some small constant $\eta$, each successive interval consisting of an $\eta$-fraction of the vertices $X_t^{\text{main}}$ is likely to cover less than half of whatever of $N^*_G(v_1, \ldots, v_\ell; V_t^{\text{main}})$ was
uncovered before embedding that interval. The argument for each given interval is morally similar to that seen in Lemma 2.28, though here we take a short-cut by using property LCON rather than NS, which simplifies the calculations.

**Proof.** Let \( h = \max_{x \in X^\text{main} \setminus X^e} \pi^T(x) \). We require

\[
\mu \leq \frac{1}{8}, \quad \varepsilon < \frac{1}{7} 2^{-280(D+1) \mu^{-2} \zeta^{-1} d^{-D}} \left( \frac{d}{2} \right)^{D+1}, \quad \theta \leq \frac{\mu^2 \zeta d^D}{1000(D + 1)}, \quad p \geq \frac{1000 \Delta^2}{\mu \zeta d^D} \left( \frac{\log n}{\eta} \right)^{1/2}.
\]

We first prove inequality (38). Set \( \eta = \frac{\mu \zeta d^D}{280(D + 1 \mu)} \). Given \( i \in [r] \) and \( 1 \leq \ell \leq D - h \), suppose that \( v_1, \ldots, v_t \in V(\Gamma) \) are such that \( N^*_G(v_1, \ldots, v_t; V^\text{main}) \geq (dp/2)\ell |V^\text{main}| \). We split the vertices of \( X^\text{main} \setminus X^e \) into intervals \( \text{Int}_{1, \ldots, \text{Int}_{1/n}} \) of equal size, with the first being the first \( \eta |X^\text{main} \setminus X^e| \) vertices in the order \( \tau \), and so on.

We now aim to show that for any fixed set \( \mathcal{V} \subseteq \mathcal{V} \) of size at least \( \varepsilon \rho |V| \) and any \( 1 \leq j \leq 1/\eta \), conditioning on the embedding up to the last vertex of \( \text{Int}_{1, \ldots, j} \), with high probability at most \( \frac{1}{2} |U| \) vertices of \( \text{Int}_j \) are embedded to \( U \). To that end, for each \( i \in \text{Int}_j \), let \( \mathcal{H}_{i,j} \) denote the history up to but not including the embedding of \( x \). By definition of \( h \), all vertices \( x \in \text{Int}_j \) have \( \pi^T(x) \leq h \), so we split the vertices \( x \) of \( \text{Int}_j \) up into \( h + 1 \) classes according to \( \pi^T(x) \). We apply LCON \((\varepsilon, r_1, D) \) with the set \( U \) and the family \( \mathcal{F}_x \) of \( s \)-sets given by the embedded images of \( N^*_H(x) \) for \( x \in \text{Int}_j \) with \( \pi^T(x) = s \). Since \( \varepsilon |U|, |\mathcal{F}_x| \leq \eta |X^\text{main}_i| \), we obtain the inequality

\[
\sum_{x \in \text{Int}_{j+1}, \pi^T(x) = s} |U_{\tau(x) - 1}(x) \cap U| \leq 7p^s |U| |\eta| X^\text{main}_i.
\]

Now each \( x \in \text{Int}_j \) with \( \pi^T(x) = s \) is by \((\text{INV}3)\) embedded uniformly at random into a set of size at least \( \frac{1}{\eta} \mu \zeta (dp)^s |V_i| \). We thus have

\[
\sum_{x \in \text{Int}_{j+1}} P(x \text{ is embedded to } U, |\mathcal{H}_{x,j}|) \leq (h + 1) \frac{70 \eta |U|}{\mu \zeta d^D} \leq (D + 1) \frac{70 \eta |U|}{\mu \zeta d^D}
\]

and by Lemma 2.2 with \( \delta = 1 \) we see that the probability that more than

\[
140(D + 1) \mu \zeta^{-1} d^{-D} |U| \leq \frac{1}{2} |U|
\]

of the vertices in \( \text{Int}_{j+1} \) are embedded to \( U \) is at most

\[
\exp \left( - (D + 1) \frac{70 \eta |U|}{3 \mu \zeta d^D} \right) \leq n^{D-1},
\]

where the inequality is because \( |U| \geq \varepsilon p^D \eta |V_i| \) and by choice of \( p \). This is what we wanted to show about \( U \), and we will now proceed to use it for various choices of \( U \).

Let \( U_0 = N^*_G(v_1, \ldots, v_t; V^\text{main}_i) \), and \( U_j = U_{j-1} \setminus \psi_{t,j+1}(\text{Int}_j) \) for each \( j \geq 1 \). In the event that \( |U_j| \geq \frac{1}{2} |U_{j-1}| \) for each \( j \), we have

\[
|U_j| \geq 2^{-1/\delta} (d/2)^j p^s |V^\text{main}_i| \geq \varepsilon p^s |V_i|
\]

for each \( j \), where the final inequality is by choice of \( \varepsilon \). In other words, each \( U_{j-1} \) is large enough for the above probabilistic calculation to be valid, so taking a union bound over the \( 1/\eta \) choices of \( j \), we see that with probability at least \( 1 - \eta^{-1} n^{-D-1} \) we have

\[
|N^*_G(v_1, \ldots, v_t; V^\text{main}_i) \setminus \text{Im}(\psi_{t,j})| \geq (1 - 4 \mu) 2^{1-j} \deg_G(v_1, \ldots, v_t; V^\text{main}_i)
\]

for each \( 1 \leq j \leq 1/\eta + 1 \). Taking a union bound over the at most \( r_1 \) choices of \( i \in [r] \) and \( n + n^2 + \cdots + n^D \) choices of \( v_1, \ldots, v_t \) we see that the desired event of the lemma holds with probability at least \( 1 - r_1 \eta^{-1} n^{-D-1} \), which tends to one as \( n \) tends to infinity.
The proof of (39) is simpler. Observe that \(|X_i^0| \leq 2p|X_i| < \eta|X_i^{\text{main}}|\), and \(\deg_G(v_i, \ldots, v_i; V_i^0) \geq (dp/2)|V_i^0| \geq \varepsilon p|V_i|\), so that the analysis above, with only one “interval” \(X_i^0 \setminus X^c\), gives the desired result.

Finally, we are in a position to prove (DRGA3), which we do in the following claim. The proof is quite similar to that of Lemma 2.29, which amounts to show that any vertex \(v \in V_i\) is likely to be a candidate vertex for “reasonably” many of the buffer vertices \(X_i^{\text{buf}}\). More precisely, our strategy is to fix a vertex \(v \in V_i\) and show that the inequality (40) below, which encapsulates these reasonable bounds, holds with sufficiently high probability to apply the union bound over all \(v \in V(G)\). To do this, we introduce the concept of an \((x \rightarrow v)\)-buffer partial embedding for \(x \in X_i^{\text{buf}}\). As in the proof of Lemma 2.29, this is a good partial embedding which has a few extra properties that let us show that it is not too unlikely that \(N_H(x)\) is embedded to \(N_G(v)\). As there, the proof that it is not too unlikely that at each step Algorithm 4 maintains an \((x \rightarrow v)\)-buffer partial embedding is mainly “bookkeeping” and long but not very hard. However, there is an important difference. It is no longer useful to simply multiply the conditional probabilities of maintaining an \((x \rightarrow v)\)-buffer partial embedding in order to estimate the conditional probability that \(N_H(x)\) is embedded to \(N_G(v)\). This is because the vertices \(N_H(x)\) are no longer embedded as a segment of \(\tau\), so that Lemma 2.2 is not applicable directly to these products of conditional probabilities. Nevertheless we show below that we can still apply Lemma 2.2, several times and with some extra care, to obtain the desired bounds.

**Claim 5.7.** The following holds a.a.s. for each \(i \in [r]\). Let \(b\) be such that \(X_i^{\text{buf}}\) is a degree-\(b\) buffer. Then for each \(v \in V_i\) we have

\[
|\{x \in X_i^{\text{buf}} : v \in C_{\text{RGAend}}(x)\}| \geq 4^{-10} \Delta^3 2^{-1000D^2\mu^{-1}d^{-2}d^2D^2} \Delta^{-3} \mu^b |X_i|.
\]

**Proof.** We require

\[
\mu \leq \frac{1}{8}, \quad \varepsilon' < \frac{\mu \varepsilon d^2}{10000D^{2}D^2} \cdot 2^{-280(D+1)} \mu^{-1}d^{-D}, \quad \varepsilon \leq \varepsilon',
\]

and

\[
p > 10000 \cdot 4^{-10} \Delta^3 \Delta^{-1} \mu^{-1}d^{-1} \Delta^{-1} \cdot 2^{280(D+1)} \mu^{-1}d^{-1} \Delta^{-1} \left(\frac{\log n}{n}\right)^{1/D}.
\]

Let \(\beta = (1 - 4\mu)2^{-280(D+1)} \mu^{-1}d^{-D}\). It is convenient to consider only vertices of \(X_i^{\text{buf}}\) which are far from \(X^c\); since \(X^c\) is very small, doing so does not exclude many vertices of \(X_i^{\text{buf}}\).

Given \(x \in X_i^{\text{buf}}\) which is at distance greater than three from any vertex of \(X^c\), and \(v \in V_i\), we say that a good partial embedding \(\psi\) is an \((x \rightarrow v)\)-buffer partial embedding if the following hold.

**BPE1** For each \(y \in \text{Dom}(\psi) \cap N_H(x)\) we have \(\psi(y) \in N_G(v)\).

**BPE2** For each unembedded \(y \in N_H(x)\) we have

\[
U(y) \cap N_H(y) = (p \pm \varepsilon p)^{\varepsilon \Delta} \deg_{\Gamma}(v; V(y)),
\]

\[
U^{\text{main}}(y) \cap N_H(y) = (p \pm \varepsilon p)^{\varepsilon \Delta} \deg_{\Gamma}(v; V^{\text{main}}(y))
\]

and

\[
U^{\text{main}}(y) \cap N_H(y) = (p \pm \varepsilon p)^{\varepsilon \Delta} \deg_{\Gamma}(v; V^{\text{main}}(y))
\]

**BPE3** For each unembedded \(y \in N_H(x)\) we have

\[
C(y) \cap N_G(y) \geq (dp - \varepsilon p)^{\varepsilon \Delta} \deg_{\Gamma}(v; V(y))
\]

and

\[
C^{\text{main}}(y) \cap N_G(y) \geq (dp - \varepsilon p)^{\varepsilon \Delta} \deg_{\Gamma}(v; V^{\text{main}}(y))
\]

**BPE4** For each unembedded \(y, z \in N_H(x)\) with \(yz \in E(H)\), the pair \((U(y) \cap N_H(v), U(z) \cap N_H(v))\) is \((\varepsilon p^{\Delta}(y), p^{\Delta}(z), d, p)\)-regular in \(G\).
For each unembedded $y \in N_H(x)$ and $z \in N_H(y)$, the pair $(U(y) \cap N_G(v), U(z))$ is $(\varepsilon_\pi(y), \pi(z), d, p)$-regular in $G$.

Much as in the proof of Lemma 2.29, the empty partial embedding $\psi_0$ is an $(x \rightarrow v)$-buffer partial embedding. Indeed, (BPE5) is trivially true since by (H5) neighbours of buffer vertices are not image restricted. For (BPE4) and (BPE5), since buffer vertices are by (H5) at distance at least three from image restricted vertices the sets $U(y)$ and $U(z)$ are equal to $V(y)$ and $V(z)$ respectively, and the required regularity is thus given by (G2), since $X_{b}^{\text{buf}} \subseteq \tilde{X}_t$.

Given an $(x \rightarrow v)$-buffer partial embedding $\psi$ and an unembedded vertex $y$, we let $P(y)$ be the set of poor vertices for $y$, namely those vertices $u \in C(y)$ such that either $\psi \cup \{y \rightarrow u\}$ is not an $(x \rightarrow v)$-buffer partial embedding, or such that there is an unembedded $z \in N_H(y) \cap N_H(x)$ such that either of the following conditions hold:

$$\deg_G(u; A^\text{main}(z) \cap N_G(v)) \leq (d - \varepsilon')p|A^\text{main}(z) \cap N_G(v)| \quad \text{or}$$

$$\deg_G(u; A^4(z) \cap N_G(v)) \leq (d - \varepsilon')p|A^4(z) \cap N_G(v)|.$$  \hspace{1cm} (41)

Observe that if $y$ is at distance four or greater from $x$ in $H$, then $P(y)$ is always empty. This has two important consequences. First, we are about to talk about the probability of no vertices being embedded to poor vertices: this is really a condition on at most $3\Delta^4$ vertex embeddings. Second, it means that if $x, x' \in X_{i}^{\text{buf}}$ by (H2) the distance between $x$ and $x'$ is at least ten, so that any given vertex embedding affects at most one of whether we have an $(x \rightarrow v)$-buffer partial embedding or an $(x' \rightarrow v)$-buffer partial embedding.

Before we try to estimate the probability of maintaining an $(x \rightarrow v)$-buffer partial embedding, we strengthen the conclusion of Claim 5.6 to cover the ‘last few vertices’ before embedding $y \in N_H(x)$.

**FACT 5.8.** For a given $x \in X_{i}^{\text{buf}}$ at distance greater than three from $X^e$ and a fixed $v \in V(x)$, provided that up to time $t$ no vertex has been embedded to a poor vertex (with respect to an $(x \rightarrow v)$-buffer partial embedding), for each $y \in N_H(x)$ with $\tau(y) > t$, the following a.a.s. hold:

$$|A^{\text{main}}_i(y) \cap N_G(v)| \geq \frac{1}{2}(dp/2)^{\pi^*_i(y)} \deg_G(v; V^{\text{main}}(y))$$

and

$$|A^{4}_i(y) \cap N_G(v)| \geq \frac{1}{2}(dp/2)^{\pi^*_i(y)} \deg_G(v; V^q(y)).$$  \hspace{1cm} (42)

We will see that Claim 5.6, together with (ORD3), show that this fact holds provided $t \leq \tau(y) - \frac{1}{2}ep^Dn/r_1$. The number of vertices embedded in the remaining $\frac{1}{2}ep^Dn/r_1$ is too small to significantly fill up either of these sets, so we need only show that embedding neighbours of $y$ does not adversely affect (42). This is guaranteed by condition (41) in the definition of poor vertices.

**PROOF.** Let $h = \max_{z \in X^\text{main} \setminus X^e} \pi^*_i(z)$. Since $X_{i}^{\text{buf}} \subseteq \tilde{X}$ and $y \in N_H(x)$ we infer by (ORD3) that all but at most $s$ neighbours $z$ of $y$ satisfy $\tau(y) - \tau(z) \leq p^D \cdot \frac{1}{2}n/r_1$ for some $s \leq D - 1 - h$ (recall that $\tau$ is a $(D, p, \frac{1}{2}n/r_1)$-bounded order). Therefore, if $t \leq \tau(y) - \frac{1}{2}ep^Dn/r_1$ then $\pi^*_i(y) \leq s \leq D - 1 - h$, so that $\Pi^*_i(y) \cup \{v\}$ is by (BPE4), (G3) and (G1) a set of $s + 1 \leq D - h$ vertices in $G$ with at least $(dp/2)^sV^q(y)$ common neighbours in $V^q(y)$ and at least $(dp/2)^{s+1}V^{\text{main}}(y)$ common neighbours in $V^{\text{main}}(y)$.

Therefore, by Claim 5.6, we have

$$|A^{\text{main}}_i(y) \cap N_G(v)| \geq \beta(dp/2)^{\pi^*_i(y)} \deg_G(v; V^{\text{main}}(y))$$

and

$$|A^{4}_i(y) \cap N_G(v)| \geq \frac{1}{2}(dp/2)^{\pi^*_i(y)} \deg_G(v; V^q(y)).$$
which gives (42) for this range of $t$. Now suppose that
\[ t = \tau(y) - \varepsilon p^D : \frac{1}{2}n/r_1 + \ell \]
for some $0 \leq \ell < \frac{1}{2}p^D n/r_1$. We claim that
\[
|A^\text{main}(y) \cap N_G(v)| \geq \beta(dp/2)^{\pi_1^*(y)} \deg_G (v; V^{\text{main}}(y)) - \ell \quad \text{and}
|A^q(y) \cap N_G(v)| \geq \frac{1}{2}(dp/2)^{\pi_1^*(y)} \deg_G (v; V^q(y)) - \ell.
\] (43)
Indeed, we have just established that (43) holds for $\ell = 0$. Given $\ell \geq 1$, suppose that (43) holds for $\ell - 1$, and consider the vertex $z$ embedded at time $\tau(z) = \tau(y) - \varepsilon p^D : \frac{1}{2}n/r_1 + \ell$.

If $z$ is not a neighbour of $y$, then its embedding decreases either of the sets on the left hand side of (43) by at most one, so the inequality continues to hold. If $z$ is a neighbour of $y$, then by (41) in the definition of $P(z)$ we have
\[
|A^q_{\tau(z)}(y) \cap N_G(v)| \geq (d - \varepsilon')p|A^q_{\tau(z)-1}(y) \cap N_G(v)|
\geq \frac{1}{2}(dp/2)^{\pi_1^*(y)} \deg_G (v; V^q(y)) - (d - \varepsilon')p\ell
\]
where the second inequality is by choice of $\varepsilon'$ and (43). A similar inequality holds for $|A^\text{main}_{\tau(z)}(y) \cap N_G(v)|$. Thus again (43) continues to hold. We conclude that (43) holds for all $0 \leq \ell < \frac{1}{2}p^D n/r_1$.

Now since $x \in X^\text{bad}$, we have $\tau(x) > \tau(y)$, so that by (ORD1) we have $\pi^*_1(y) \leq \pi^*_1(y) \leq 2 - 1$. By (G3) and choice of $\varepsilon$, we conclude that (42) holds for all $t < \tau(y)$ as desired.

We now continue with the proof of Claim 5.7 by estimating the probability of not choosing a poor vertex at a given step in Algorithm 4.

**FACT 5.9.** Suppose that $\psi$ is a good partial embedding generated together with a queue set $Q$ by Algorithm 4 which is also an $(x \to v)$-buffer partial embedding, where $x$ is at distance greater than three from any vertex of $X^c$, and (42) holds for $\psi$ and all unembedded $y$. Given an unembedded $y$, suppose that $u$ is a vertex chosen uniformly at random in $A^\text{main}(y) \cup B(y)$ (if $y \notin Q \cup X^c$), in $A^q(y) \cup B(y)$ (if $y \in Q \cup X^c$) or in $A^q(y) \cup B(y)$ (if $y \in X^c$). Then the following hold.

- If $\text{dist}_H(x, y) > 3$ then $P(u \notin P(y)) = 1$.
- If $\text{dist}_H(x, y) \in \{2, 3\}$ then $P(u \notin P(y)) \geq \frac{1}{2}$, and
- If $\text{dist}_H(x, y) = 1$ then $P(u \notin P(y)) \geq \beta d^{D-1} 4^{D^2} p$.

The proof of this fact is quite similar to the analysis in Lemma 2.29. Note that the use we make of $\psi$ and $Q$ being generated by Algorithm 4 is that the invariants of Claim 5.2 hold; we do not in the following proof perform any analysis of the probabilistic process generating $\psi$ and $Q$.

**PROOF.** If $\text{dist}_H(x, y) > 3$ then $P(y) = \emptyset$ and the claim is trivial.

By (INV3) we embed $y$ uniformly at random into a set of size at least $\frac{1}{10} \mu_\zeta(dp)^{\pi^*(y)} |V(y)|$. Thus in order to show the second item we simply need to establish that $P(y) \cup B(y)$ is small compared to this set; this can be established using NS($\varepsilon, r_1, D$), RI($\varepsilon, (\varepsilon_{a, b}, \varepsilon', d, r_1, D)$ and the buffer and good partial embedding properties, as in the proof of Lemma 2.29. Note that to do this we need to use the fact, given by (ORD1), that $\pi^*(z) \leq D - 2 - 1$ for each $z \in N(X)$, in other words that $\pi^*(z) \leq D - 2$ for all $z \in N_H(x)$ and $\pi^*(z) \leq D - 3$ if there is a triangle $zuv'$ in $H$ with $\tau(u), \tau(w') > \tau(z)$. Note further that in order to establish that few vertices fail (41), we need to know that $A^\text{main}(z) \cap N_G(v)$ and $A^q(z) \cap N_G(v)$ are not small in comparison to $U(z) \cap N_F(v)$. This follows from (42) since $z \in N_H(x)$.
Finally, we embed \( y \) uniformly at random into a subset of \( U(y) \), which by \((GPE2)\) has size at most \( 2p^{\pi^*(y)}|V(y)| \). If \( xy \in H \), we need to show that it is not too unlikely that \( y \) is embedded to \( N_G(v) \). By \((42)\), \((G3)\) and choice of \( \beta \), we see that
\[
|A_{\text{main}}^\beta(y) \cap N_G(v)|, |A_i^\beta(y) \cap N_G(v)| \geq \frac{\beta}{2}(dp/2)^{\pi^*(y)}|V(y)|.
\]
It is thus enough to show that at most half of these vertices are in \( P(y) \cup B(y) \), which again we can do using \( \text{NS}(\varepsilon, r_1, D), \text{RI}(\varepsilon, (\varepsilon, \varepsilon, D), \varepsilon, d, r_1, D) \) and the buffer and good partial embedding properties, as in the proof of Lemma 2.29. Again, we use the fact \( \pi^*(z) \leq D_z - 1 \) for \( z \in N_H(x) \) in this verification.

We now continue with the proof of Claim 5.7. We fix \( i \) and \( v \in V_i \). We have established that for any given vertex \( x \) of \( X_i \) far from \( X^e \), the probability that \( N_H(x) \) is embedded to \( N_G(v) \) is not too small. Furthermore, this probability is more or less given by the embeddings of vertices at distance at most three from \( x \); and distinct vertices of \( X_i \) are at distance at least ten by \((H2)\), so that no vertex is within distance three of two distinct \( x, x' \) of \( X_i \). It should not be surprising that this is enough to show that indeed with very high probability a reasonable fraction of the \( x \in X_i \) have \( N_H(x) \) embedded to \( N_G(v) \); this is what we ultimately want. However, because the sets \( N_H(x) \) interleave each other in the order \( \pi \), we cannot simply apply Lemma 2.2. We now embark on taking the ‘extra care’ mentioned before the Claim to complete the proof.

For each \( x \in X_i \), let \( M_x \) be the set of vertices of \( H \) at distance one, two or three from \( x \). We say \( x \) survives at step \( j \) if after the embedding of the \( j \)th vertex of \( M_x \) in the order \( \pi \), we still have an \( (x \to v) \)-buffer partial embedding. We will use Lemma 2.2 to show that a reasonable fraction of \( x \in X_i \) survive at step 1. We would like then to use Lemma 2.2 again to show that (because a reasonable fraction of vertices survive at step 1) a reasonable fraction of vertices survive at step 2, and so on.

In order to carry this out, it is convenient not in fact to look at all \( x \in X_i \), but only at a subset \( Y \) in which the probability of surviving at each step \( j \) (as given by Claim 5.9) does not depend on the particular vertex in \( Y \) but only on \( j \). We now construct such a set \( Y \). For any \( x \in X_i \), at distance greater than three from \( X^e \), there are at most \( \Delta^3 + \Delta^2 + \Delta < 3\Delta^3 \) vertices of \( H \) at distance at most 3 from \( x \) in \( H \). We can therefore associate to each \( x \in X_i \) at distance greater than three from \( X^e \) a 0–1 vector of length at most \( 3\Delta^3 \), taking the value 0 at place \( j \) if the \( j \)th vertex \( y \) of \( M_x \) (in the order \( \pi \)) is not a neighbour of \( x \), and 1 if \( y \) is a neighbour of \( x \). There are at most \( 3\Delta^3 \) choices for this vector; we fix a most common choice \( c = (c_j) \) of length \(|c| \), and let \( Y \) be the subset of vertices of \( X_i \) at distance at least three from \( X^e \) associated to this most common choice. By construction, by \((H3)\), by choice of \( \varepsilon \) and since \( |X^e| \leq \frac{1}{2}\varepsilon p^{D_1} n/r_1 \) and \(|X_i| \geq n/kr_1 \), we have
\[
|Y| \geq \frac{1}{6}\Delta^{-3} 2^{-3\Delta^3}|X_i|.
\]

For each \( x \in Y \) and \( 1 \leq j \leq |c| \), let \( \mathcal{H}_{x,j} \) be the history of Algorithm 4 up to the point immediately before embedding the \( j \)th vertex of \( M_x \). For each \( x \in Y \), we create a collection of Bernoulli random variables \( A_{x,1} \) for \( x \in Y \), which are set equal to one if either \( x \) survives at step 1 or \((42)\) has failed before embedding the first element of \( M_x \), and zero otherwise. These random variables have a natural order given by the order \( \pi \) on the first elements (in \( \pi \)) of the \( M_x \) for \( x \in Y \), and they are in this order sequentially dependent. We define
\[
s_1 = \begin{cases} \frac{1}{2} \beta d^{D_1+1} 4^{-D p} |Y| & \text{if } c_1 = 1, \\ \frac{1}{2} |Y| & \text{if } c_1 = 0. \end{cases}
\]
Now for each $A_{x,1}$, either Fact 5.9 gives us a lower bound on $E[A_{x,1}|\mathcal{H}_{x,1}]$, or (42) has failed before we embed the first vertex of $M_x$, in which case $E[A_{x,1}|\mathcal{H}_{x,1}] = 1$. We conclude, by definition of $s_1$, that

$$\sum_{x \in Y} E[A_{x,1}|\mathcal{H}_{x,1}] \geq 2s_1.$$ 

Thus we can use Lemma 2.2, with $\delta = 1/4$, to show that with very high probability we have $\sum_{x \in Y} A_{x,1} \geq s_1 + 1$ (We will fill in the missing quantitative details later). We denote this good event by $A_1$. Observe that if (42) does not fail for any $y$ or $t$ (which is a.a.s. the case) then $A_1$ holding says that at least $s_1 + 1$ vertices survive at step 1.

There remains a small technical difficulty in continuing the programme outlined above. We would like to count vertices surviving at step 2, and use Lemma 2.2 and a lower bound on the sum of conditional expectations given by the product of $s_1$ and the probability bound from Fact 5.9 to show that the result is likely to be large. But it is possible that less than $s_1$ vertices survive at step 1, so that this bound on the sum of conditional expectations does not hold almost surely (it only holds with very high probability).

To get around this problem, we use the following trick. We say $x$ is dangerous at step 1 if the following holds. Immediately after embedding the first vertex of $M_x$ in Algorithm 4, the number of vertices $x'$ in $Y$ such that $A_{x,1}$ is certainly equal to one (i.e. such that the first vertex of $M_{x'}$ was embedded to a vertex which is neither poor nor bad, so $x'$ survives at step 1), plus the number of vertices $x''$ in $Y$ such that the first vertex of $M_{x''}$ has not yet been embedded, is at most $s_1$. In other words, we say $x$ is dangerous at step 1 if after embedding the first vertex of $M_x$ we already know that $A_1$ does not occur.

We now define Bernoulli random variables $A_{x,2}$ for $x \in Y$, set equal to one if either $x$ survives at step 2, or $x$ is dangerous at step 1, or (42) has failed for some $y$ prior to embedding the second vertex of $M_x$, and zero otherwise. The point of this definition is that it gives us an a priori lower bound on the sum of conditional expectations of the $A_{x,2}$, as we require to apply Lemma 2.2, but nevertheless with very high probability the sum of the $A_{x,2}$ does simply count the number of vertices in $Y$ which survive at step 2, because with very high probability (42) does not fail and $A_1$ does occur, so no vertex is dangerous at step 1. Again, this collection of random variables has a natural order given by the order $\tau$ on the second vertices of the $M_x$, and again in this order the random variables are sequentially dependent. Note that (as required for this last to be true) we do know whether $x$ is dangerous at step 1 before we embed the second vertex of $M_x$. This justifies that we can apply Lemma 2.2 to estimate the sum of the $A_{x,2}$.

We now complete this programme. We need to give integers $s_j$ for $2 \leq j \leq |c|$, which are our desired lower bounds on the number of vertices in $Y$ surviving at step $j$, to define the events $A_j$ and the concept of dangerous at step $j$ for $j \geq 2$, and define the random variables $A_{x,j}$ for $3 \leq j \leq |c|$. We also need to say explicitly what ‘very high probability’ is in our applications of Lemma 2.2.

For each $2 \leq j \leq |c|$, we set

$$s_j = \begin{cases} \frac{1}{2}d^{D+1}4^{-D}ps_{j-1} & \text{if } c_j = 1 \\ \frac{1}{2}s_{j-1} & \text{if } c_j = 0 \end{cases}.$$ 

We let the event $\sum_{x \in Y} A_{x,j} \geq s_j + 1$ be $A_j$. We say $x \in Y$ is dangerous at step $j$ if immediately after embedding the $j$th vertex of $M_x$, the number of $x' \in Y$ such that $A_{x',j}$ is certainly equal to one, plus the number of $x' \in Y$ such that the $j$th vertex of $M_{x'}$ has not yet been embedded, is at most $s_j$. Again, this means that
after embedding the $j$th vertex of $M_x$, we already know $A_j$ does not occur. Finally, for each $j \geq 3$ we define the Bernoulli random variables $A_{x,j}$ for $x \in Y$, set equal to one if $x$ survives at step $j$, or is dangerous at step $j - 1$, or (42) has failed before embedding the $j$th vertex of $M_x$, and zero otherwise.

Before we continue, we observe that

$$s_1 \geq s_2 \geq \cdots \geq s_{|e|} \geq \left(\frac{1}{4}\right)^{\Delta^2 + \Delta^3} \left(\frac{1}{2}\beta d^{D+1} - D\right)^b |Y|$$

where $b$ is the degree of the buffer $X_{buf}^b$, i.e. deg$_H(x) = b$ for each $x \in X_{buf}^b$; this number exists by $(H6)$. The reason is simply that there are at most $\Delta^2 + \Delta^3$ vertices of $H$ at distance two or three in $H$ from any given $x \in Y$, and $b$ neighbours of $x$. In particular, by choice of $p$, by our lower bound on $|Y|$ and since $n$ is sufficiently large, we have $3s_j/2 > s_j + 1$ for each $1 \leq j \leq |c|$.

We now show that with very high probability $A_j$ occurs for each $1 \leq j \leq |c|$. Recall that by Fact 5.9, we have the lower bound $\sum_{x \in Y} \mathbb{E}[A_{x,1} | \mathcal{H}_{x,1}] \geq 2s_1$. By Lemma 2.2 with $\delta = 1/4$, and the observation $3s_j/2 > s_j + 1$, the probability that $A_1$ fails is thus at most $e^{-s_1/24}$.

Now for $j \geq 2$, for each $x \in Y$, either $x$ survived at step $j - 1$ and (42) has not failed before embedding the $j$th vertex of $M_x$, in which case the expectation of $A_{x,j}$ conditioned on $\mathcal{H}_{x,j}$ is at least the quantity given in Fact 5.9, or $x$ was dangerous at step $j - 1$ or (42) failed before embedding the $j$th vertex of $M_x$, in which case we have $\mathbb{E}[A_{x,h} | \mathcal{H}_{x,j}] = 1$, or none of these occur, in which case we have $\mathbb{E}[A_{x,h} | \mathcal{H}_{x,j}] \geq 0$. Furthermore, by definition of the $A_{x,j}$ we know (a priori, before beginning the embedding) that one of the first three cases occurs for at least $s_{j-1}$ of the vertices $Y$, so that we have $\sum_{x \in Y} \mathbb{E}[A_{x,j} | \mathcal{H}_{x,j}] \geq 2s_j$ by definition of $s_j$. Applying Lemma 2.2, with $\delta = 1/4$, and since $3s_j/2 > s_j + 1$, we conclude that the probability that $A_j$ fails is at most $e^{-s_j/24}$.

By definition, if (42) never fails and each event $A_1, \ldots, A_{|e|}$ occurs, then no vertex is dangerous at any step. Thus the number of vertices of $Y$ surviving at step $|e|$ is at least $s_{|e|}$. By definition, if $x$ survives at step $|e|$ then $N_H(x)$ is embedded to $N_G(v)$, so that $v$ is a candidate for $x$. Since $s_{|e|}$ is by (45), (44) and $(H3)$ at least

$$\left(\frac{1}{4}\right)^{\Delta^2 + \Delta^3} \left(\frac{1}{2}\beta d^{D+1} - D\right)^b \cdot \frac{1}{6}\Delta^{-32\beta^3 - 3\Delta^3} \cdot 4|\mathcal{H}_x|$$

we conclude that (40) holds for $v$, as desired. The probability that any one of the events $A_1, \ldots, A_{|e|}$ fails is at most $e^{-s_{|e|}/24} < n^{-D-1}$ by (45), (44), $(H3)$, since $|X_i| \geq n/\kappa r_1$ and by choice of $p$. Taking the union bound over the at most $3\Delta^3$ events $A_j$, and the at most $n$ choices of $v$, and since a.a.s. (42) never fails by Fact 5.8, we conclude that a.a.s. (40) holds for all $v \in V_i$ for each $i \in [r]$, completing the proof of Claim 5.7.

The good event of Claim 5.7 holding gives $(DRGA3)$. Since a.a.s. the good events of each of the above claims and lemmas hold, this completes the proof of Lemma 5.1.
CHAPTER 6

Proofs of applications

In this chapter we prove the various theorems listed in Section 1.2. Where a stronger result is proved elsewhere, we only sketch the proofs.

6.1. Universal graphs

We begin by showing that \( G_{n,p} \) is universal for spanning bounded-degree graphs with degeneracy \( d \) when \( p \geq C \left( \frac{\log n}{n} \right)^{1/(2d+1)} \).

**Proof of Theorem 1.1.** Observe that any \( d \)-degenerate \( n \)-vertex graph has at most \( dn \) edges and so contains at least \( \frac{n}{2d+1} \) vertices of degree at most \( 2d \). We will apply Lemma 1.23 with input \( \Delta \), with \( \Delta = \frac{\zeta}{16\Delta} \), \( \zeta = 1 \), density \( \frac{1}{2} \), and \( \kappa = 2 \). Lemma 1.23 returns \( \varepsilon > 0 \), and for input \( r_1 = 8\Delta \) also \( C \). Choose \( \varepsilon' \ll \varepsilon \) and suppose \( C \) is also large enough for Lemma 2.17(a) with input \( \varepsilon', 8\Delta \) and 2. Let \( p \geq C \left( \frac{\log n}{n} \right)^{1/(2d+1)} \). Fix an equipartition \( V_1, \ldots, V_{3\Delta} \) of \([n]\). Now \( G_{n,p} \) a.a.s. satisfies the good event of Lemma 1.23. By Lemma 2.17(a) it a.a.s. has property \( NS(\varepsilon', 8\Delta, D) \). Finally, using Theorem 2.1 and the union bound it is easy to check that a.a.s. for each \( i \) and \( v \in V_i \), the vertex \( v \) has \( \left( 1 \pm \varepsilon \right) p |V_j| \) neighbours in \( V_j \) for each \( j \neq i \). Fix a graph \( \Gamma \) with each of these properties. Property \( NS(\varepsilon', 8\Delta, D) \) implies that any pair of disjoint subsets of \( V(\Gamma) \), each of size at least \( \frac{pn}{D} \), is \( (\varepsilon', \frac{1}{2}, p) \)-regular, so letting \( R = R' = K_{3\Delta} \) we have an \( (\varepsilon', \frac{1}{2}, p) \)-regular \( R \)-partition which has super-regularity and one- and two-sided inheritance on \( R' \).

Given \( H \) on \( n \) vertices with degeneracy \( d \) and maximum degree \( \Delta \), let \( X_1, \ldots, X_{3\Delta} \) be an equipartition of \( V(H) \) into independent sets, with each \( X_i \) containing at least \( \frac{n}{8\Delta(2d+1)} |X_1| \) vertices of degree at most \( 2d \). This equipartition exists by Lemma 2.4. We designate the vertices of degree at most \( 2d \) as potential buffer vertices. We do not image restrict any vertices. Then (perhaps after reordering) the \( X_i \) form an \( R \)-partition of \( V(H) \) which is size-compatible with the \( V_i \). We let \( \tau \) be a degeneracy order on \( V(H) \). We move all buffer vertices to the end of the ordering \( \tau \). Observe that this slightly changed ordering satisfies conditions (ORD1)–(ORD3) and thus is \( (D, p, \varepsilon' n/r_1) \)-bounded. Then the conditions of Lemma 1.23 are satisfied, so \( H \subseteq \Gamma \) as desired.

To obtain the second statement, we work identically except that we add to each part of \( V(H) \) isolated vertices to obtain a size-compatible \( R \)-partition, and we designate these isolated vertices as potential buffer vertices instead of the low-degree vertices in \( H \). Then Lemma 1.23, applied with \( D = 2d, p = C \left( \frac{\log n}{n} \right)^{1/D} \) and all other parameters staying as before, gives the desired stronger conclusion. \( \square \)

It is quite easy to use Lemma 1.25 to show that sufficiently pseudorandom graphs are universal for bounded-degree graphs.

**Sketch proof of Theorem 1.2.** Given \( G \), we take a random equipartition of \( V(G) \) into \( \Delta + 1 \) clusters \( V_1, \ldots, V_{\Delta+1} \). For each \( v \in V(G) \) and \( i \), the quantity \( \deg(v; V_i) \) is hypergeometrically distributed with mean at least \( \frac{1}{2} |V_i| \), so by Theorem 2.1 and the union bound we see that with positive probability we have
deg(e; V_i) ≥ \frac{1}{t} p |V_i| for each r and i. We fix such a partition, and let \( R = R' = K_{\Delta+1} \).

Using Theorem 2.3 we can for any n-vertex graph \( H \) with \( \Delta(H) ≤ \Delta \) find a size-compatible \( R \)-partition of \( V(H) \), and applying Lemma 1.25 with all vertices designated as potential buffer vertices, with \( d = \frac{1}{t} \) and with no vertices image restricted, we see that \( H ⊆ G \) as desired.

6.2. Partition universality

We now prove that \( G_{n,p} \) is a.a.s. \( r \)-partition universal for \( \mathcal{H}(n, d, \Delta) \) provided \( p ≥ C\left(\frac{\log n}{n}\right)^{1/(2d)} \), which improves on the result of Kohayakawa, Rödl, Schacht and Szemerédi [44] for graphs with \( d ≤ \Delta/2 \). A proof of the result of [44] can be obtained along very similar lines.

We need two by now classical results of Extremal Combinatorics: Ramsey’s Theorem and Turán’s Theorem. The Ramsey number \( R(k) \) all but at most \( (2) \) is \( R(k) \) and Szekeres [60] proved these numbers exist, while the following quantitative statement is due to Erdős and Szekeres [30].

**Theorem 6.1 (Erdős and Szekeres [30]).** For any \( k \) and \( t \) we have \( R_k(K_t) ≤ k^{kt} \).

Turán, generalising a result of Mantel [57], proved the following.

**Theorem 6.2 (Turán [69]).** For any \( r ≥ 3 \) the unique \( K_r \)-free graph with most edges is the complete balanced \((r−1)\)-partite graph on \( n \) vertices.

In our applications we will solely use that an \( n \)-vertex \( K_r \)-free graph has at most \( \frac{(r−2)n^2}{2(r−1)} + o(n^2) \) edges.

Finally we state the version of the sparse regularity lemma for many colours that we are going to apply. We also say that a graph \( G \) with density \( p \) is \((\eta, D)\)-upper-uniform if for all disjoint sets \( U \) and \( W \) of cardinality at least \( \eta v(G) \) we have \( e_G(U, W) ≤ Dp|U||W| \).

**Lemma 6.3 (Sparse regularity lemma, coloured version [41]).** For any real \( D, \varepsilon > 0 \), integers \( k \) and \( t_0 \), there exist \( \eta > 0 \) and \( T \) such that any graph \( G \) of edge density \( p \) and on at least \( t \) vertices, which is \((\eta, D)\)-upper-uniform and whose edges are coloured with \( k \), colours admits a partition of \( V(G) \) into \( V_1, \ldots, V_t \) with the following properties:

1. \( t_0 ≤ t ≤ T \);
2. \( ||V_i|−|V_j|| ≤ 1 \) for all \( i, j \);
3. all but at most \( 2t^2 \) pairs \((V_i, V_j)\) are \((\varepsilon, d, p)\)-regular in each of the \( k \) colours for some possibly different \( d \).

The assumption on \((\eta, D)\)-upper uniformity is easily seen to be satisfied for any moderate \( \eta \) and \( D > 1 \) by our random graph \( G_{n,p} \) and by the bijumbled graphs that we will be considering. The partition asserted by Lemma 6.3 is called \( \varepsilon \)-regular. We define a coloured multigraph \( R(\delta) \) on \([t]\) (associating each \( i ∈ [t] \) with the class \( V_i \)) as follows. We put an edge \( ij \) into \( E(R) \) in colour \( c \) if the pair \((V_i, V_j)\) is \((\varepsilon, \delta, p)\)-regular in colour \( c \) in \( G \).

After these preparations we can complete the proof.

**Proof of Theorem 1.3.** We apply Lemma 1.23 with input \( \Delta \), with \( \Delta_{R'} = 8\Delta \), \( D = 2d \) (except when \( d = 1 \) when we set \( D = 2 \)), \( α = \frac{1}{2}, \zeta = 1, \) density \( \frac{\log n}{n} \), and \( κ = 1 \). Lemma 1.23 returns \( \varepsilon > 0 \), which we suppose is small enough for the application of Turán’s Theorem below. We let \( r_1 \) be large enough for the \( k \)-coloured sparse regularity lemma with input \( \varepsilon \) and also for the applications of Turán’s and Ramsey’s Theorem below, and obtain \( C ≥ r_1 \) from Lemma 1.23. Give \( p \), we generate \( Γ = G_{Cn,p} \),
suppose it satisfies the good event of Lemma 1.23 and take any \( k \)-colouring of its edges. We apply the sparse regularity lemma for coloured graphs to this coloured graph. We obtain a coloured reduced graph with at most \( \varepsilon \)-fraction of pairs not forming edges, in which we find a \( 4\Delta+1 \)-vertex clique where all edges are present by Turán’s Theorem, and in that a monochromatic \((\Delta+1)\)-vertex clique by Ramsey’s Theorem. Thus we have \( \Delta+1 \) equal-sized clusters \( V_1, \ldots, V_{\Delta+1} \), each of size at least \( n \), such that there is a colour \( c \) in which each pair of clusters is \((\varepsilon, \frac{1}{2\Delta}, p)\)-regular. Let \( G \) be the graph of colour \( c \) edges, let \( R = K_{\Delta+1} \) and \( R' \) be the empty graph.

Given \( H \in \mathcal{H}(n, \Delta) \), by Theorem 2.3 we can find an equipartition of \( V(H) \) into \( \Delta+1 \) independent sets. We ‘pad’ each set by adding isolated vertices to obtain a size-compatible \( R \)-partition \( X_1, \ldots, X_{\Delta+1} \), and designate the isolated vertices as potential buffer vertices which come last in the degeneracy order \( \tau \) of \( H \). Then the result follows by Lemma 1.23, with \( \tau \) being \((D, p, \varepsilon n/r)\)-bounded order on \( V(H) \).

**Sketch proof of Theorem 1.3.** The same approach as for Theorem 1.3 works, replacing Lemma 1.23 with Lemma 1.25. It is easy to check that the upper-uniformity condition of Lemma 6.3 is satisfied for \( \Gamma \) with the bijumbledness condition of the theorem.

**Sketch proof of Corollary 1.5.** We follow the same approach as in the proof of Theorem 1.3, with the exceptions that we use \( p = \Theta(\frac{\log n}{n}) \), that we use Lemma 1.21 instead of Lemma 1.23, and that after generating \( \Gamma = G_{n, p} \) which satisfies the good event of Lemma 1.21, we form \( G \) by deleting a minimum number of edges to remove all copies of \( K_{2\Delta} \). The expected number of copies of \( K_{2\Delta} \) in \( G_{n, p} \) is \( p^{\binom{2\Delta}{2}} n^{2\Delta} = o(n) \), so by Markov’s inequality we see that a.a.s. we delete \( o(n^2) \) edges of \( \Gamma \) to form \( G \). This \( G \) is the claimed \( r \)-partition universal graph for \( \mathcal{H}(n, \Delta) \). The remainder of the proof of Theorem 1.3 proves the partition universality, since the number of edges deleted is too small to destroy regularity of any pair of clusters.

### 6.3. Maker-Breaker games

The proof of Theorem 1.6 uses the result of Ferber, Krivelevich and Naves [32], which informally says that if \( p \ll b^{-1} \) is not too small, there is a \((\Delta+1)\)-bounded order on \( K_n \) which gives Maker a subgraph of \( G_{n, p} \) with minimum degree very close to \( pn \). This reduces the proof to showing a \((1 - \varepsilon^*)pn / (\Delta+1) \) neighbourhood in both \( \Gamma \) and \( G \) in each set \( V_i \).

**Sketch proof of Theorem 1.6.** We sketch the proof for \( \mathcal{H}'(n, \Delta) \)-universality first. We take an equipartition of \([n]\) into \( \Delta+1 \) clusters \( V_1, \ldots, V_{\Delta+1} \). We let \( \varepsilon > 0 \) be sufficiently small for Lemma 1.21, with no vertices image restricted, and \( p \) sufficiently large and we set \( \varepsilon^* \ll \varepsilon \). Now a.a.s. Maker obtains a subgraph \( G \) of \( \Gamma = G_{n, p} \) with \( \delta(G) \geq (1 - \varepsilon^*)pn \) by following the strategy of Ferber, Krivelevich and Naves [32]. It is easy to check, using Chernoff’s inequality, that each vertex has \((1 + 2\varepsilon^*)pn / (\Delta+1) \) neighbours in both \( \Gamma \) and \( G \) in each set \( V_i \).

Since the total number of edges of \( \Gamma \) which are not in \( G \) leaving any vertex is at most \( 4\varepsilon^*pn \), the total number of such edges between any two vertex sets \( X \) and \( Y \) is at most \( 4\varepsilon^*pn |X| \). An easy application of the Chernoff bound shows that \((X, V_i)\) is \((\varepsilon^*, \frac{1}{2\Delta}, p)\)-regular in \( \Gamma \) for each \( V_i \) and disjoint vertex set \( X \) with \( |X| \geq pn / (\Delta+1) \), and since \( \varepsilon^* \ll \varepsilon \) we conclude that any such \((X, V_i)\) is also \((\varepsilon, \frac{1}{2\Delta}, p)\)-regular in \( G \).

\(^{1}\text{Formally, one can state this as: there is a randomised Maker strategy which generates a graph } \Gamma \text{ from the distribution } G_{n, p} \text{, and a spanning subgraph } G \text{ of edges obtained by Maker, which a.a.s. has minimum degree } (1 - \varepsilon)pn.\)
Letting $R = R' = K_{\Delta+1}$, we see that $G$ is $(\varepsilon, \frac{1}{2}, p)$-super-regular on $R'$ and has one-sided inheritance on $R'$. Given any graph $H \in \mathcal{H}'(n, \Delta)$ we use Theorem 2.3 to find an equipartition of $V(H)$ into $\Delta + 1$ independent sets $X_1, \ldots, X_{\Delta+1}$. Then the conditions of Lemma 1.21 are satisfied, so we have $H \subseteq G$ as desired.

For the almost-spanning $\mathcal{H}(n, \Delta)$-universality, we repeat the same argument, replacing $[n]$ with $[(1 + \delta)n]$ and ‘padding’ each equipartition class of $H$ with independent vertices to be size-compatible with $V$. We take these independent vertices to be the potential buffer vertices, so that two-sided inheritance is not needed.

For the degeneracy statements, we replace Lemma 1.21 with Lemma 1.23, taking respectively $D = 2d$ for the almost-spanning universality and $D = 2d + 1$ for the spanning universality. Again, similar to Theorem 1.3, it is easy to verify that the degeneracy order of $H \in \mathcal{H}(n, d, \Delta)$ is an appropriately ‘bounded’ order in the sense of Definition 1.22.

In both cases, since we have shown that Maker has a randomised strategy which wins with positive probability against any strategy of Breaker, it follows that Breaker does not have a winning strategy. Since finite draw-free games are determined, we conclude that Maker does have a winning strategy. □

Note that, by an analysis similar to the proof of Theorem 1.3, one can show that Maker also succeeds with this strategy in making a graph which is $r$-partition universal for $\mathcal{H}(cn, \Delta)$ (respectively, for $\mathcal{H}(cn, d, \Delta)$) for some small $c > 0$, matching the density of the best known constructions.

6.4. Resilience of low-bandwidth graphs

For this section we need the minimum degree form of the sparse regularity lemma, which we quote from [22], the paper in which the bipartite case of Theorem 1.9 is proved. To state it we need to define two concepts. First, an $\varepsilon$-equipartition of a vertex set $V$ is a partition $V = V_0 \cup \ldots \cup V_r$ such that $|V_0| \leq \varepsilon|V|$ and $|V_1| = \cdots = |V_r|$. Second, if $G$ is a graph with vertex set $V$, then the $(\varepsilon, d, p)$-lower-regular reduced graph of $G$, with respect to a given $\varepsilon$-equipartition $V = V_0 \cup \ldots \cup V_r$, is the graph on $[r]$ with edges $ij$ corresponding to $(\varepsilon, d, p)$-lower-regular pairs $(V_i, V_j)$ in $G$.

**Lemma 6.4** ([22], Lemma 4.4). For all $\vartheta \in [0, 1]$, $\varepsilon > 0$ and every integer $r_0$, there exists $r_1 \geq 1$ such that for all $d \in [0, 1]$ the following holds a.a.s. for $\Gamma = G_{n,p}$ if $\log^4 n / (pn) = o(1)$. Let $G = (V,E)$ be a spanning subgraph of $\Gamma$ with $\deg_G(v) \geq \alpha \deg_{\Gamma}(v)$ for all $v \in V$. Then there is an $\varepsilon$-equipartition of $G$ with $(\varepsilon, d, p)$-lower-regular reduced graph $R$ of minimum degree $\delta(R) \geq (\vartheta - d - \varepsilon)\nu(R)$, and $r_0 \leq \nu(R) \leq r_1$.

Using this, we can sketch the proof of Theorem 1.9. The strategy consists of modifying the argument in [24], the paper in which the Bandwidth Theorem was proved. We will not state formally the lemmas from that paper which we require; the reader not familiar with that argument will wish to read the following sketch in conjunction with Section 2 of [24], in which the lemmas are formally stated and their use outlined. The changes to their strategy we make are as follows. Their ‘Lemma for $G$’ is replaced with Lemma 6.4 (since we do not need most of the properties of the ‘Lemma for $G$’ in this setting), and Theorem 1.8 finds a ‘backbone graph’ in the resulting $(\varepsilon, d, p)$-reduced graph $R$. We can use their ‘Lemma for $H$’ as written, and in this setting it gives a partition of $V(H)$ which is directly suitable to apply Lemma 1.21, yielding the desired embedding of $H$ into $G$.

**Proof of Theorem 1.9.** Given $\gamma > 0$ and $\Delta$, we choose $d \ll \gamma$ and $r_0 \gg \Delta$. We will apply Lemma 1.21 with input $\Delta$, $\Delta' = \Delta + 1$, $\alpha = \gamma/2$, $\zeta = 1$, $d$ and
\[ \kappa = 2. \text{Lemma 1.21 returns } \vartheta, \varepsilon > 0, \text{of which we are only interested in } \varepsilon. \text{We assume, without loss of generality, that } \varepsilon \ll d. \text{We let } r_1 \text{ be returned by Lemma 6.4 for input } \vartheta = \max_{2 \leq r \leq \Delta + 1} \left( \frac{1}{\varepsilon} + \gamma \right), \varepsilon \text{ and } r_0. \text{Finally, we choose } \beta \ll r_1^{-1}, \text{and let } C \text{ be returned by Lemma 1.21 for input } r_1. \]

Now, given \( p \geq C \left( \log \frac{n}{\varepsilon} \right)^{1/\Delta} \), we generate \( \Gamma = G_{n,p} \), and assume it satisfies the conditions of Lemmas 1.21 and 6.4 for the parameters given above. Let a graph \( H \) on \((1 - \gamma)n\) vertices with \( \Delta(H) \leq \Delta \) and \( \text{bw}(H) \leq \beta n \) be given. Let \( r = \chi(H) \leq \Delta + 1. \) Let a spanning subgraph \( G \) of \( \Gamma \) with minimum degree \( \left( \frac{\varepsilon}{\beta} + \gamma \right) \beta n \) be given.

We apply Lemma 6.4 to \( G \), obtaining an \((\varepsilon, d, p)\)-reduced graph \( R \), with \( r_0 \leq v(R) \leq r_1 \), and minimum degree at least \( \left( \frac{\varepsilon}{\beta} + \frac{\gamma}{2} \right) v(R) \). By Theorem 1.8, we can find in \( R \) a spanning ‘backbone graph’. This consists of a collection of vertex-disjoint copies of \( K_r \), which come in a linear order, such that between one copy of \( K_r \) and the next there is a copy of \( K_{r,r} \) with a perfect matching removed. This graph is \( r \)-colourable and has maximum degree \( 3(r - 1) \leq 3\Delta \) and bandwidth at most \( 2r \leq 2\Delta \), so that Theorem 1.8 is indeed applicable provided \( r_0 \) is sufficiently large compared to \( \Delta \).

It is quite easy to find a homomorphism from \( H \) to the backbone graph. We simply divide \( V(H) \) up into intervals in the bandwidth order, and map successive intervals of \( V(H) \) to successive copies of \( K_r \) in the backbone graph, choosing vertices of each copy of \( K_r \) according to a fixed \( r \)-colouring of \( H \). The role of the bandwidth restriction here is to ensure that edges of \( H \) either lie within one interval, or go from one interval to the next, so that we only need edges in \( R \) from one copy of \( K_r \) in the backbone graph to the next in order to obtain a homomorphism. The point of fixing an \( r \)-colouring of \( H \) is that the \( i \)th vertex in one copy of \( K_r \) and that in the next are not adjacent in the backbone graph (since a perfect matching was removed from the \( K_{r,r} \) between them) and we need to ensure that no edge of \( H \) will be assigned to have one endpoint in each.

Unfortunately, this is not quite enough: the colour classes of \( H \) could be quite unbalanced, so that the homomorphism we have just described maps many more vertices of \( H \) to some vertices of \( R \) than others. In order to repair this, we need to ‘rebalance’, which requires that each copy of \( K_r \) in the backbone graph extends to a copy of \( K_{r+1} \) using some other vertex of \( R \) (which may be anywhere in the backbone graph). The ‘Lemma for \( H' \)’ of [24] now states that given a backbone graph whose \( r \)-cliques extend to copies of \( K_{r+1} \), and \( H' \), there is a homomorphism \( f \) from \( V(H) \) to \( V(R) \) in which each vertex of \( R \) is the image of approximately the same number of vertices of \( H \). Since \( \varepsilon \ll \gamma \), since \( v(H) \leq (1 - \gamma)n \) and by choice of \( \beta \) sufficiently small, the ‘approximately’ in this statement in particular guarantees \( |f^{-1}(i)| \leq (1 - \frac{\gamma}{2})|V_i| \) for each \( i \in V(R) \).

It remains only to verify that the conditions of Lemma 1.21 can be met in order to find an embedding of \( H \) into \( G \). The idea is simple: we ‘pad’ \( H \) by adding \( n - v(H) \) isolated vertices, and give a partition \( X = (X_i)_{i \in V(R)} \) of \( V(H) \) in which \( X_i \) consists of \( f^{-1}(i) \) together with isolated vertices such that \( |V_i| = |X_i| \). We designate the isolated vertices in each \( X_i \) as the potential buffer vertices \( \tilde{X}_i \). It then follows that the empty graph \( R' \) on \( V(R) \) with no edges, together with these potential buffer vertices, gives us a \((\gamma/2, R')\)-buffer for \( H \). By construction, the partition \( X \) is an \( R' \)-partition of \( H \). By definition, \( \tilde{V} \) is an \((\varepsilon, d, p)\)-regular \( R' \)-partition of \( G \), and the super-regularity and inheritance properties required of \( R' \) are satisfied vacuously. Finally, we do not image restrict any vertices of \( H \), so that the restriction pair properties are satisfied vacuously. Thus Lemma 1.21 gives us the desired embedding of \( H \) into \( G \).

\[ \square \]

We stress that the main difficulty in the proof of Theorem 1.8 is to obtain a spanning embedding; an almost-spanning embedding is much easier. It is similarly,
and for similar reasons, much harder to prove Theorem 1.10 than Theorem 1.9. It is also worth noting that in the proof given in [24], there is a substantial amount of routine technical work to do in between obtaining size-compatible partitions of $H$ and $G$ and using the blow-up lemma to get an embedding of $H$ into $G$, which is encapsulated in the ‘partial embedding lemma’. This work is necessary because the blow-up lemma of [46] cannot be applied to the entire reduced graph. Our Lemma 1.21 can be applied to the entire reduced graph, and thus replaces both the partial embedding lemma and the blow-up lemma of [46].

6.5. Robustness of the Bandwidth Theorem

As with the proof of Theorem 1.9, the proof of Theorem 1.11 amounts to modifying the proof of Theorem 1.8. However this time we need to rely on rather more of the machinery built up in [24]. Again, the reader not familiar with the argument there will wish to read this sketch in conjunction with Section 2 of [24].

**Sketch proof of Theorem 1.11.** We choose constants as in [24, ‘Proof of Theorem 2’], with the exception that we obtain $\varepsilon'$ from Lemma 1.21 and $\varepsilon$ from Theorem 2.7 and Corollary 2.8 for input $\varepsilon'$ rather than from the blow-up lemma of [46] and the ‘partial embedding lemma’ of [24]. We then follow the proof given there up to the point at which in [24] the first vertices of $H$ are embedded using the partial embedding lemma. Let us recap what this amounts to. We are given graphs $G$ and $H$ satisfying the conditions of Theorem 1.11. We apply the ‘Lemma for $G$’ of [24], which first returns a partition of $V(G)$ into parts $V_{i,j}'$, where $1 \leq i \leq k$ and $1 \leq j \leq \chi(H)$, with the following properties\(^2\). First, $k \leq K_0 = K_0(\gamma, \chi(H))$ is bounded in terms of $\gamma$ and $\chi(H)$. Second, the $(\varepsilon, d, 1)$-reduced graph $R$ of this partition, whose vertex set is $[k] \times [\chi(H)]$ matching the indices of the partition, has minimum degree at least $(\frac{\chi(H)-1}{\chi(H)} + \frac{\gamma}{2})v(R)$. Third, $R$ contains a spanning backbone graph; that is, if $|i - i'| \leq 1$ and $j \neq j'$ then $(i, j)(i', j') \in E(R)$. Fourth, the parts $V_{i,j}'$ and $V_{i',j'}'$ differ in size by at most one, and each part has size between $(1 - \varepsilon)n / k\chi(H)$ and $2n / k\chi(H)$.

Now we apply the ‘Lemma for $H$’ of [24], with the reduced graph $R$ and the integer partition of $n$ given by the $|V_{i,j}'|$. This gives us a homomorphism $f$ from $H$ to $R$, and a set of ‘special vertices’ $Z \subseteq V(H)$, with the following properties, which depend on a quantity $\xi$ satisfying $\beta \ll \xi \ll \varepsilon, K_0^{-1}$. First, $|Z| \leq K_1 \chi(H) n$. Second, for each $(i, j) \in V(R)$ we have $|f^{-1}(i, j)| = |V_{i,j}'| \pm \xi n$. Third, if $uw \in H$ has neither endpoint in $Z$, then the first coordinates of $f(u)$ and $f(w)$ are equal, in other words $u$ and $v$ are mapped to vertices of the same clique in $R$. We set $X_{i,j} = f^{-1}(i, j)$ for each $(i, j) \in R$. By construction, the resulting partition $X$ is an $R$-partition of $H$.

Next, we return to the ‘Lemma for $G$’, which guarantees, given the sizes of the parts $|X_{i,j}|$ satisfying the above properties, a partition $V$ of $V(G)$ with parts $V_{i,j}$ for $(i, j) \in V(R)$ which has the following properties. First, $|V_{i,j}| = |X_{i,j}|$ for each $(i, j) \in V(R)$. Second, $R$ is an $(\varepsilon, d, 1)$-reduced graph for $G$ with respect to $V$. Third, $G$ is super-regular on the graph $R'$ whose edges are $(i, j)(i', j')$ for $i \in [k]$ and $1 \leq j < j' \leq \chi(H)$, in other words on the $K_{\chi(H)}$-factor in the backbone graph.

This is the point at which, in [24], the embedding of $H$ into $G$ begins. It is worth remarking that, because Lemma 1.21 applies to the entire reduced graph, we could complete their proof by simply verifying the conditions of Lemma 1.21 for $p = 1$ (much as we do below), rather than needing the technical work of the partial embedding lemma.

\(^2\)The ‘Lemma for $G$’ does not explicitly return this partition, but it is convenient for the explanation to mention its existence; it also does not explicitly give the upper bound on sizes of the parts, but this follows from the proof.
Recall that we wish to show that $H$ is a subgraph of $G_p$, where $p \geq C \left(\frac{\log n}{n}\right)^{1/\Delta}$ for some suitably large $C$. Observe that $E(G_p)$ is distributed as $E(G) \cap E(G_{n,p})$. We now generate $\Gamma = G_{n,p}$. Asymptotically almost surely, the good event of Lemma 1.21 occurs. We need to verify that $H$ and $G_p = G \cap \Gamma$ a.a.s. satisfy the conditions of Lemma 1.21. We begin with $H$. Recall that $\mathcal{X}$ is an $R$-partition of $H$. By choice of $\xi$, it is 4-balanced. Also by choice of $\xi$, less than half of the vertices of any given $X_{i,j}$ are at distance two or less from $Z$. We let $\tilde{X}_{i,j}$ be the vertices of $X_{i,j}$ at distance three or more from $Z$. This gives us a $(\frac{1}{2}, R')$-buffer for $H$.

We now need to show that a.a.s. $V$ is an $(\varepsilon', d, p)$-regular $R'$-partition of $G \cap \Gamma$, and that a.a.s. it is $(\varepsilon', d, p)$-super-regular and has one- and two-sided inheritance on $R'$. The first of these is an easy consequence of Theorem 2.1 (Chernoff’s inequality) and the fact that $V$ is an $(\varepsilon, d, p)$-regular $R$-partition of $G$. We can simply take the union bound over the at most $2^n$ choices of pairs of subsets of $V(G)$ which we need to have density at least $(d - \varepsilon')p$. Since $G$ has super-regularity on $R'$, again using Theorem 2.1 and taking the union bound over the choices of vertices in $V(G)$ and $V(R')$, we see that a.a.s. $\Gamma \cap G$ has super-regularity on $R'$. Next we show that a.a.s. if $v \in V_i$ and $ij, ik \in R'$, so $(N_{\Gamma \cap G}(v; V_j), V_k)$ is $(\varepsilon', d, p)$-regular in $G$. Indeed, by Theorem 2.7, with $\beta = \frac{1}{4}$, the probability that this fails is at most $2^{-pn}$, so that we can take a union bound over all choices of $i, j, k$ and $v \in V_i$. Similarly, using Corollary 2.8, if also $ik \in R'$, a.a.s. $(N_{\Gamma \cap G}(v; V_j), N_{\Gamma \cap G}(v; V_k))$ is $(\varepsilon', d, p)$-regular in $G$.

We do not image restrict any vertices of $H$, so the restriction pair condition of Lemma 1.21 is vacuously satisfied. Thus, applying Lemma 1.21, we see $H \subseteq G \cap \Gamma$ as desired. \hfill \square
CHAPTER 7

Concluding remarks

7.1. Optimality of our main results

7.1.1. Lemma 1.21. In the case $\Delta = 2$ and $H$ contains a triangle, our result is optimal up to the log factor, since it is well known that when $p = o(n^{-1/2})$ one can delete all triangles from $G_{n,p}$ by removing only $o(pn^2)$ edges—so any ‘blow-up-type’ statement will be false.

When $\Delta = 3$, the statement of Lemma 1.21 is optimal up to the log factor. In the event that $H$ is a spanning $K_4$-factor, for small $c > 0$ and $p = cn^{-1/3}$, it is typically possible to find a subgraph $G$ of $G_{4n,p}$ with the following properties. There is a partition $V(G) = V_1 \cup \cdots \cup V_4$ with each set of size $n$. Letting $R = R' = K_4$, the partition is an $(\varepsilon, 1/2, p)$-regular $R$-partition, which is super-regular, with one- and two-sided inheritance on $R'$. However there is a vertex of $G$ which is in no $K_4$, and thus there is no $K_4$-factor covering $G$.

To construct $G$, we let $V_1, \ldots, V_4$ be an equipartition of $[4n]$, fix a vertex $v \in V_1$, reveal $G_{4n,p}$, remove all edges within each part, and then remove from $N_G(v)$ all triangles by deleting a minimum number of edges. It is not hard to check using Theorem 2.1 that a.a.s. before this final step, we have an $(\varepsilon, 3/4, p)$-regular $R$-partition with super-regularity and one- and two-sided inheritance on $R'$. It is similarly easy to check that a.a.s. before the final step the degree of each vertex in $V_i$ to any other $V_j$ is close to $pn$, and any pair of vertices has about $p^2n$ common neighbours in other parts. Thus the expected number of triangles in $N_G(v)$ is $p^3(pm)^3 = 6n$, and the actual number is somewhat concentrated, so that a.a.s. in the final step we delete at most $2\varepsilon^3n$ edges. This is far too small to destroy one-sided inheritance, and by choice of $c$ too small to destroy two-sided inheritance. It remains to show that we have not removed too many edges from any one vertex. But the expected number of edges removed from $u \in N_G(v)$ is at most $p(p^2n)^2 = cp^2n$; the actual number is by Theorem 2.1 exponentially concentrated, and hence a.a.s. we remove $o(pn)$ edges from each $u$, so do not destroy super-regularity.

Perhaps inserting an extra condition into the statement, such as insisting that all vertex neighbourhoods contain $\Omega(p^6n^3)$ triangles, would allow one to prove a blow-up lemma which allows for embedding a $K_4$-factor down to the natural limit $p = n^{-2/5}$. However it would not always be possible to obtain such a condition. In [4] it is shown that Breaker wins the $K_4$-factor game on $K_n$ with a bias $13n^{1/3}$ (Lemma 1.21 is used to show that Maker wins when the bias is $c(n \log n)^{1/3}$ for some small $c > 0$). It follows that any ‘extra condition’ is one which Maker cannot guarantee to obtain with bias $13n^{1/3}$.

Finally, for $\Delta \geq 4$ one might hope that the statement of our blow-up lemma remains true down to $p = Cn^{-2/(\Delta+3)}$, this being the point at which one can generalise the above construction (and therefore the point at which the statement provably does not guarantee a $K_{\Delta+1}$-factor). Perhaps, optimistically, one might hope that there are some natural extra conditions which even allow $p \geq Cn^{-2/(\Delta+2)}$, this being the point at which one can remove all copies of $K_{\Delta+1}$ by deleting a tiny fraction of the edges of $G_{n,p}$.
However we believe that improving upon our result is likely to be very challenging. Even in the (much simpler and well-studied) setting of trying to prove $H(n, \Delta)$-universality of $G_{n,p}$, there has been no improvement upon what can be obtained from Lemma 1.21. For almost-spanning universality, Conlon, Ferber, Nenadov and Skorić [27] could improve on Lemma 1.21, showing that the random graph $G_{(1+\gamma)n,p}$ is a.a.s. $H(n, \Delta)$-universal when $p = \omega(n^{-1/(\Delta-1)} \log^3 n)$ for $\Delta \geq 3$. But the improvement in the exponent one would desire is of order $\Delta^{-1}$ not $\Delta^{-2}$.

7.1.2. Lemma 1.25. We do not believe the bijumbledness requirement of Lemma 1.25 is optimal. Most of the proof would work with $(p, Cp^{\Delta+1} n)$-bijumbledness, but we were not able to find a way to avoid the LNS property which requires the stronger condition. Nevertheless we conjecture that $(p, Cp^{\Delta+1} n)$-bijumbledness suffices (we expect that the extra factor of $p$ can be gained by using reserved cliques as in the proof of Lemma 1.21). It is still not clear that this would be optimal. For $\Delta = 2$ we need $(p, Cp^2 n)$-bijumbledness, since Alon [9] constructed a $(p, Cp^2 n)$-bijumbled graph without triangles. It is a believable conjecture (see for example Conlon, Fox and Zhao [25]) that there are $(p, Cp^2 n)$-bijumbled graphs without $K_{\Delta+1}$ for every $\Delta$, in which case the same requirement would be necessary for a blow-up lemma. It is possible, however (conjectured in [25], but the contrary is conjectured by Kohayakawa, Rödl, Schacht and Skokan [45]) that copies of $K_{\Delta+1}$ cannot be guaranteed in regular subgraphs of $(p, \beta)$-bijumbled graphs at this point, but instead that $\beta \leq Cp^{\Delta+1} n$ is required (it is proved in [25] that at this point one can guarantee copies of $K_{\Delta+1}$).

Note that Lemma 1.25 does not permit linearly many image restrictions. Only about a $p^2$-fraction of vertices in each part may be image restricted. In many applications this is not a problem (see for example [1]), but it could well cause a problem for some applications. We could modify the proof strategy substantially in order to have some control over linearly many vertices of each $X_i$. The modification we would make is the following. We would permit the user of the blow-up lemma to specify $g|X_i|$ ‘pre-embedded’ vertices in each part $X_i$ which are to be embedded first (before even the neighbours of buffer vertices). The user is then permitted to embed these vertices sequentially, subject to four conditions. First, the result must be a good partial embedding $\psi$. Second, each vertex must be embedded to a uniform random vertex from a set of size at least $\frac{1}{50} \mu_G(dp)^{\Delta-1} n/(\kappa r_1)$. Third, for each $jk \in R$ and $v \in V_j$, at most $\frac{1}{100} \Delta \deg_G(v; V_k)$ neighbours of $v$ in $V_k$ may be in $\Im(\psi)$. Fourth, the total number of vertices in $X_i$ with pre-embedded neighbours is at most $g|X_i|$. We would then follow the proof strategy of Lemma 1.25 to embed the remainder of $H$ into $G$, treating the neighbours of pre-embedded vertices as ‘image restricted’. Note that this does not automatically resolve the problem with Claim 4.5, since the partition of $V(G)$ considered at this point is not the same as the partition the user of the blow-up lemma supplies. But it is easy to modify Lemma 2.22 to show that all the sets under consideration are with high probability evenly distributed by the random equipartition, and then the proof of Claim 4.5 does go through. Checking the full details of this approach, and for that matter using the resulting blow-up lemma, seems likely to be non-trivial, but it could potentially allow for stronger theorems.

7.1.3. Lemma 1.23. We made some efforts to prove as flexible a statement as we could. Observe that even without saying anything about the structure of $H$ beyond its degeneracy, the result is ‘almost’ as powerful as Lemma 1.21, in that a $d$-degenerate $n$-vertex graph with bounded maximum degree, which we can handle with $p \approx n^{-1/(2d+1)}$, can contain almost $dn$ edges, comparable to a $2d$-regular graph for which Lemma 1.21 would require $p \approx n^{-1/2d}$. 
7.2. Algorithmic embedding

In the event that we only need an almost-spanning embedding, we can take
the potential buffer vertices in each part to be isolated, and hence \( D = 2d \).
We can then embed \( d \)-degenerate graphs with \( p \approx n^{-1/(2d)} \), matching the performance of Lemma 1.21. Finally, if \( H \) is an \( F \)-factor we can take \( D = d + 3 \), in which case the performance of Lemma 1.23 substantially improves, working with \( p \approx n^{-1/(d+3)} \).

This last case is one in which we can improve Lemma 1.23. Recall from the
proof of Lemma 5.1 that we only require \((\text{ORD})3\) and the condition \( \pi^*(x) \leq D_e - 1 \)
for \( x \in N(\tilde{X}) \) within \((\text{ORD1})\) in order to prove \((\text{DRGA3})\). When embedding an
\( F \)-factor (provided \( R' \) is suitable, for example \( R' = K_{|F|} \)), we do not really need
this last condition, as \((\text{DRGA2})\) shows that only a few vertices in each cluster of \( G \)
can fail \((\text{DRGA3})\), and we can use an argument similar to Lemma 3.3 to deal with
them. This allows us to reduce the required \( D \) by one.

In particular, for each \( s \leq t \) one can prove a blow-up lemma which embeds a
\( K_{s,t} \)-factor when \( p \geq C(\log n/n)^{1/s} \). This is almost optimal, since the 2-density of \( K_{s,t} \)
is \( \frac{st-1}{s+t-2} \), which approaches \( s \) as \( t \) becomes large; when \( p \) is below \( n^{-s(t-1)/(st-1)} \),
one can delete a very small fraction of the edges of \( G(n,p) \) to destroy all copies of
\( K_{s,t} \).

7.2. Algorithmic embedding

The proofs of our blow-up lemmas give polynomial-time randomised algorithms
which with high probability construct the embeddings we prove exist. It is quite
tedious to check the details, but we provide a sketch for the interested reader of how
one can do this.

The main difficulty concerns the certification of sparse-regular pairs. Alon, Duke,
Lefmann, Rödl and Yuster [8] showed that determining if a given bipartite graph
is \( \varepsilon \)-fully-regular is co-NP-complete, but that there is a polynomial-time algorithm
which either certifies \( \varepsilon \)-full-regularity, or returns a witness to the failure of \( \varepsilon' \)-full-
regularity, for some \( \varepsilon' \) which may be much smaller than \( \varepsilon \) but does not depend
on the number of vertices in the regular pair. For sparse graphs, a corresponding
polynomial-time certification algorithm was given by Alon, Coja-Oghlan, Hán, Kang,
Rödl and Schacht [14] which either certifies \((\varepsilon,d,p)\)-full-regularity or returns a
witness to the failure of \((\varepsilon',d,p)\)-full-regularity.

Unfortunately, we do not know of any such algorithm in the literature for lower-
regular pairs, so we now sketch a certification algorithm for lower-regularity which
works in subgraphs of random graphs. Given as input \( \varepsilon, d, p \) and a bipartite graph,
which must be `bounded’ (see [14]), we apply the algorithmic sparse regularity lemma of [14]
with regularity parameter \( \frac{1}{100} \varepsilon^3 \). This returns a partition of each side of the
bipartite graph into approximately equal numbers of parts, which approximately
equipartition each side, even if the bipartite graph itself is very unbalanced (we may
want, for example, to know whether a bipartite graph with parts of size \( pn \) and
\( p^2 n \) respectively is lower-regular). We choose \( 0 < \varepsilon' < \frac{1}{100} \varepsilon^3 \) lower bounding the fraction of either side contained in any one part of the partition. Now if any pair of
parts in this partition has density less than \((d - \varepsilon') p \) it is a witness to a failure of
\((\varepsilon', d, p)\)-lower-regularity. If not, we claim the bipartite graph is \((\varepsilon, d, p)\)-lower-regular;
the choice of regularity parameter in the use of sparse regularity ensures that there
are too few irregular pairs to seriously affect densities between large sets. We note
that the requirement of `boundedness’ is needed for the algorithmic sparse regularity
lemma, and that this boundedness holds a.a.s. in subgraphs of typical random graphs
(and is implied by the NS property which we require in any case). In contrast to the
certification algorithm for sparse full-regularity, the dependency of \( \varepsilon' \) on \( \varepsilon \) here is
very poor: there is a tower-type relationship, which appears iterated in the constant
dependencies of the algorithmic versions of Lemmas 1.21 and 1.23.
For either random or bijumbled ambient graphs \( \Gamma \), given a certification algorithm and inheritance lemmas, one can prove a variant of the RI property (see Section 2.2) in which not only do typical vertex neighbourhoods inherit (either version of) regularity, but they do so certifiably. We follow the proofs more or less as in Section 2.2, except that at each step, where we need certifiable \((\epsilon, d, p)\)-regularity for some \(\epsilon\), we obtain \(\epsilon'\) from the certification algorithm and then let \(\epsilon''\) be returned by our inheritance lemmas for input \(\epsilon'\). Now a typical vertex neighbourhood is then \((\epsilon'', d, p)\)-regular, so that the certification algorithm will certify it to be \((\epsilon, d, p)\)-regular as desired.

We now sketch how one can use this to obtain algorithmic versions of our blow-up lemmas.

It is necessary to check that the algorithm implicit in the preprocessing (Lemma 2.22) is a randomised polynomial time algorithm. This follows since the Hajnal-Szemerédi Theorem (Theorem 2.3) has an algorithmic version \([39]\), and since the proof of Lemma 2.4 can then easily be made constructive. Furthermore, a failure of the randomised partitioning to produce a good \(G\)-partition can be detected in polynomial time.

It is further necessary to check that each of our RGA algorithms (Algorithms 1, 3 and 4) can be carried out in polynomial time. This amounts to checking that the various sets that appear in these algorithms can be constructed in polynomial time. For most of these sets, this is obviously possible. However, in order to construct the bad sets \(B_i(x)\) we need the certifiable regularity inheritance discussed above.

In proving Lemma 1.21 we give (implicitly) an algorithm for embedding queue vertices. Again, to run this algorithm we need to construct the bad sets \(B(x)\) and this requires certifiable regularity inheritance. We also use the fact that the bipartite matching problem can be solved in polynomial time. We have yet another algorithm for fixing buffer defects. The proof of Lemma 3.7 constructs the sets \(P_i\) and \(D_i\) of this algorithm in polynomial time, while the remainder, Algorithm 2, is trivially polynomial time.

Finally, each of our blow-up lemmas is completed by embedding the buffer vertices. This amounts to a bipartite matching problem, and can be solved in polynomial time.

It seems reasonable to believe that it is possible to derandomise our RGA algorithms and preprocessing algorithm, which would yield polynomial time algorithms for constructing each of the claimed embeddings. Certainly Komlós, Sárközy and Szemerédi \([48]\) were able to derandomise their original (RGA-based) proof of the dense blow-up lemma. However we did not attempt to check whether their methods suffice in our case.

### 7.3. Directed graphs

Although our blow-up lemmas as written apply to undirected graphs, we can also apply them to subdigraphs of random directed graphs (or bijumbled directed graphs). We give for illustration the directed statement corresponding to Lemma 1.21. In order to state this, we define the undirection of a digraph \(\overrightarrow{G}\) to be the graph \(G\) with \(uv \in E(G)\) if and only if either \(\overrightarrow{uv}\) or \(\overrightarrow{vu}\) is an arc of \(\overrightarrow{G}\). The terms which we defined for undirected graphs (such as \(\overrightarrow{R}\)-partition, buffer, and so on) are taken as applying to the undirections of all digraphs, with the exception that when we talk about an \(\overrightarrow{R}\)-partition of \(\overrightarrow{G}\) or \(\overrightarrow{H}\) we mean in addition that all arcs of \(\overrightarrow{G}\) or \(\overrightarrow{H}\) go in the direction specified by the arcs of \(\overrightarrow{R}\).

**Theorem 7.1.** For all \(\Delta, \Delta_R, \Delta_J, \alpha, \zeta, d > 0, \kappa > 1\) there exist \(\epsilon, \varrho > 0\) such that for all \(r_1\) there is a \(C\) such that for \(q > C(\log n/n)^{1/\Delta}\) the following holds. Let
Then there is an embedding $\psi$ (as opposed to only of vertices), since this is necessary (even in the dense case) for $G$ to be a subgraph of the undirection of $R$. Suppose that $(G)$ bound on $v$ proved that the answer is ‘yes’ if the restriction and trivially the answer is ‘no’ if $\Delta + 2$ is replaced by $\Delta + 1$. Nešetřil and Rödl \cite{nesetriel1994} graph $G$, the following problems.

We have seen that the answer is ‘yes’ if $\Delta + 2$ is replaced by $2\Delta$ (Corollary 1.5), beyond the question of improving on our main results, we would like to pose several open problems.

**Problem 7.2.** Is it true that for each $r, \Delta \geq 2$ and $n$ there exists a $K_{\Delta+2}$-free graph $G$ which is $r$-partition universal for $\mathcal{H}(n, \Delta)$, with $v(G) = O(n)$?

We have seen that the answer is ‘yes’ if $\Delta + 2$ is replaced by $2\Delta$ (Corollary 1.5), and trivially the answer is ‘no’ if $\Delta + 2$ is replaced by $\Delta + 1$. Nešetřil and Rödl \cite{nesetriel1994} proved that the answer is ‘yes’ if the restriction $v(G) = O(n)$ is removed, but the bound on $v(G)$ coming from their proof is a fast-growing function of $n$.\[p = 2q - q^2\]. The random directed graph $\overrightarrow{G} = \overrightarrow{G}_{n,q}$ asymptotically almost surely satisfies the following.

Let $H$ be a digraph on $r \leq r_1$ vertices without cycles of length 2 and let $R'$ be a subgraph of the undirection of $H$ with $\Delta(R') \leq \Delta'$. Let $\overrightarrow{H}$ and $\overrightarrow{G} \subseteq \overrightarrow{G}$ be digraphs given with $r$-balanced, size-compatible vertex partitions $\mathcal{X} = \{X_i\}_{i \in [r]}$ and $V = \{V_i\}_{i \in [r]}$ with parts of size at least $m \geq n/(r\delta_1)$. Let $I = \{I_x\}_{x \in V(\overrightarrow{H})}$ be a family of image restrictions, and $J = \{J_x\}_{x \in V(\overrightarrow{H})}$ be a family of restricting vertices. Suppose that

- (BUL1) The undirection of $\overrightarrow{H}$ has maximum degree at most $\Delta$, $(\overrightarrow{H}, \mathcal{X})$ is an $\overrightarrow{R}$-partition, and $\mathcal{X} = \{X_i\}_{i \in [r]}$ is an $(\alpha, R')$-buffer for $\overrightarrow{H}$,
- (BUL2) $(\overrightarrow{G}, \mathcal{V})$ is an $(\epsilon, d, p)$-regular $\overrightarrow{R}$-partition, which is $(\epsilon, d, p)$-super-regular on $R'$, has one-sided inheritance on $R'$, and two-sided inheritance on $R'$ for $\mathcal{X}$,
- (BUL3) $I$ and $J$ form a $(p, \zeta, \Delta, \Delta, J)$-restriction pair.

Then there is an embedding $\psi: V(\overrightarrow{H}) \to V(\overrightarrow{G})$ such that $\psi(x) \in I_x$ for each $x \in \overrightarrow{H}$.

This theorem is a corollary of Lemma 1.21. We simply work with the undirections of all digraphs, which satisfy the conditions of Lemma 1.21, and observe that an embedding $\psi$ of the undirection of $\overrightarrow{H}$ into the undirection of $\overrightarrow{G}$ such that $\psi(x) \in V_i$ for each $x \in X_i$ is by definition of an $\overrightarrow{R}$-partition automatically an embedding of $\overrightarrow{H}$ into $\overrightarrow{G}$. Observe that we chose $p$ so that the undirection of $\overrightarrow{G}_{n,q}$ is $G_{n,p}$.

With rather more work, we believe we could allow $\overrightarrow{H}$ to contain 2-cycles. More generally, we could allow $G$ to be coloured from a palette of at most $\kappa$ colours, define $R$ to be the multicoloured graph (with edges permitted to have several colours) corresponding to relatively dense sparse-regular pairs in the given colour, and supply a coloured graph $H$ with a coloured graph homomorphism to $R$, which we require to be embedded with edges going to edges of $G$ with the correct colour. We believe that such coloured versions of all three of our blow-up lemmas can be proved, following the strategies given in this paper. However to do so requires appropriate modifications to several definitions and recalculation of various parameters. We see no reason why this should cause difficulty, but we did not check the details.

**7.4. Hypergraphs**

It seems likely that the techniques developed in this paper will be very helpful for proving a blow-up lemma for uniform hypergraphs which works relative to sparse random or pseudorandom (appropriately defined) hypergraphs. Ideally, one might hope for a version of the hypergraph blow-up lemma of Keevash \cite{keevash2017} which allows for ‘image restrictions’ of vertex tuples of size up to $k - 1$ in $k$-uniform hypergraphs (as opposed to only of vertices), since this is necessary (even in the dense case) for some applications.

**7.5. Open problems**

Beyond the question of improving on our main results, we would like to pose the following problems.

**Problem 7.2.** Is it true that for each $r, \Delta \geq 2$ and $n$ there exists a $K_{\Delta+2}$-free graph $G$ which is $r$-partition universal for $\mathcal{H}(n, \Delta)$, with $v(G) = O(n)$?

We have seen that the answer is ‘yes’ if $\Delta + 2$ is replaced by $2\Delta$ (Corollary 1.5), and trivially the answer is ‘no’ if $\Delta + 2$ is replaced by $\Delta + 1$. Nešetřil and Rödl \cite{nesetriel1994} proved that the answer is ‘yes’ if the restriction $v(G) = O(n)$ is removed, but the bound on $v(G)$ coming from their proof is a fast-growing function of $n$. \[p = 2q - q^2\]. The random directed graph $\overrightarrow{G} = \overrightarrow{G}_{n,q}$ asymptotically almost surely satisfies the following.

Let $\overrightarrow{H}$ be a digraph on $r \leq r_1$ vertices without cycles of length 2 and let $R'$ be a subgraph of the undirection of $\overrightarrow{H}$ with $\Delta(R') \leq \Delta'$. Let $\overrightarrow{H}$ and $\overrightarrow{G} \subseteq \overrightarrow{G}$ be digraphs given with $r$-balanced, size-compatible vertex partitions $\mathcal{X} = \{X_i\}_{i \in [r]}$ and $V = \{V_i\}_{i \in [r]}$ with parts of size at least $m \geq n/(r\delta_1)$. Let $I = \{I_x\}_{x \in V(\overrightarrow{H})}$ be a family of image restrictions, and $J = \{J_x\}_{x \in V(\overrightarrow{H})}$ be a family of restricting vertices. Suppose that

- (BUL1) The undirection of $\overrightarrow{H}$ has maximum degree at most $\Delta$, $(\overrightarrow{H}, \mathcal{X})$ is an $\overrightarrow{R}$-partition, and $\mathcal{X} = \{X_i\}_{i \in [r]}$ is an $(\alpha, R')$-buffer for $\overrightarrow{H}$,
- (BUL2) $(\overrightarrow{G}, \mathcal{V})$ is an $(\epsilon, d, p)$-regular $\overrightarrow{R}$-partition, which is $(\epsilon, d, p)$-super-regular on $R'$, has one-sided inheritance on $R'$, and two-sided inheritance on $R'$ for $\mathcal{X}$,
- (BUL3) $I$ and $J$ form a $(p, \zeta, \Delta, \Delta, J)$-restriction pair.

Then there is an embedding $\psi: V(\overrightarrow{H}) \to V(\overrightarrow{G})$ such that $\psi(x) \in I_x$ for each $x \in \overrightarrow{H}$.

This theorem is a corollary of Lemma 1.21. We simply work with the undirections of all digraphs, which satisfy the conditions of Lemma 1.21, and observe that an embedding $\psi$ of the undirection of $\overrightarrow{H}$ into the undirection of $\overrightarrow{G}$ such that $\psi(x) \in V_i$ for each $x \in X_i$ is by definition of an $\overrightarrow{R}$-partition automatically an embedding of $\overrightarrow{H}$ into $\overrightarrow{G}$. Observe that we chose $p$ so that the undirection of $\overrightarrow{G}_{n,q}$ is $G_{n,p}$.

With rather more work, we believe we could allow $\overrightarrow{H}$ to contain 2-cycles. More generally, we could allow $G$ to be coloured from a palette of at most $\kappa$ colours, define $R$ to be the multicoloured graph (with edges permitted to have several colours) corresponding to relatively dense sparse-regular pairs in the given colour, and supply a coloured graph $H$ with a coloured graph homomorphism to $R$, which we require to be embedded with edges going to edges of $G$ with the correct colour. We believe that such coloured versions of all three of our blow-up lemmas can be proved, following the strategies given in this paper. However to do so requires appropriate modifications to several definitions and recalculation of various parameters. We see no reason why this should cause difficulty, but we did not check the details.
Problem 7.3. For what $p$ is $G_{n,p}$ typically $H(n, \Delta)$-universal? Typically $H(n, d, \Delta)$-universal? Does the answer change substantially if $G_{n,p}$ is replaced with $G_{Cn,p}$ for $C$ large? or for $C$ slightly larger than one?

Problem 7.4. For what $\beta$ are $(p, \beta)$-bijumbled $n$-vertex graphs $G$ with minimum degree $\frac{1}{2}pn$ always $H(n, \Delta)$-universal? Does the answer change if we allow $G$ to have $Cn$ vertices for $C$ large? or for $C$ slightly larger than one?

Problem 7.5. Do there exist graphs $G$ which are $r$-partition universal for $H(n, \Delta)$ with only $O(n^{2-2/\Delta})$ edges?

Problem 7.6. For what bias $b$ can Maker win the $H(n, \Delta)$-universality game on $K_n$? on $K_{Cn}$ for $C$ large? for $C$ slightly larger than one?

Although there exist universal graphs with $n^{2-2/\Delta}$ edges (Alon and Capalbo [11]), Maker certainly cannot make them with a bias $\Omega(n^{2/\Delta})$, since Maker requires $b = O(n^{2/(\Delta+2)})$ in order to make just one copy of $K_{\Delta+1}$.
Bibliography