### HYPERGRAPH CONTAINERS

#### DAVID SAXTON AND ANDREW THOMASON

ABSTRACT. We develop a notion of containment for independent sets in hypergraphs. For every r-uniform hypergraph G, we find a relatively small collection C of vertex subsets, such that every independent set of G is contained within a member of C, and no member of C is large; the collection, which is in various respects optimal, reveals an underlying structure to the independent sets. The containers offer a straightforward and unified approach to many combinatorial questions concerned (usually implicitly) with independence.

With regard to colouring, it follows that simple r-uniform hypergraphs of average degree d have list chromatic number at least  $(1/(r-1)^2 + o(1)) \log_r d$ . For r = 2 this improves a bound due to Alon and is tight. For  $r \ge 3$ , previous bounds were weak but the present inequality is close to optimal.

In the context of extremal graph theory, it follows that, for each  $\ell$ -uniform hypergraph H of order k, there is a collection C of  $\ell$ -uniform hypergraphs of order n each with  $o(n^k)$  copies of H, such that every H-free  $\ell$ -uniform hypergraph of order n is a subgraph of a hypergraph in C, and  $\log |C| \leq cn^{\ell-1/m(H)} \log n$  where m(H) is a standard parameter (there is a similar statement for induced subgraphs). This yields simple proofs, for example, for the number of H-free hypergraphs, and for the sparsity theorems of Conlon-Gowers and Schacht. A slight variant yields a counting version of the KLR conjecture.

Likewise, for systems of linear equations the containers supply, for example, bounds on the number of solution-free sets, and the existence of solutions in sparse random subsets. Balogh, Morris and Samotij have independently obtained related results.

#### 1. INTRODUCTION

A substantial number of theorems in the literature can be phrased naturally in terms of independent sets in uniform hypergraphs, though this phraseology is not often used explicitly. An *r*-uniform hypergraph, or *r*-graph, *G* is a pair (V(G), E(G)) comprising two sets, the vertices V(G) and edges E(G) of *G*, where each edge  $e \in E(G)$  is a set of *r* elements of V(G). Hence a 2-graph is an ordinary graph. A set  $I \subset V(G)$  is *independent* if there is no edge  $e \in E(G)$  with  $e \subset I$ .

There are many questions that, on the face of it, have little to do with hypergraphs, but which can be formulated naturally in terms of the number of independent sets in some hypergraph or class of hypergraphs (examples will be given later). Nevertheless, the question *per se* of how many independent sets there can be in a graph has attracted attention only relatively recently. The maximum number of independent sets in a graph of given average degree can be determined easily via the Kruskal-Katona theorem [40, 32], but for regular graphs the maximum is harder to find: following a good estimate by Alon [1], the exact value for bipartite graphs was determined by Kahn [31] via an elegant entropy argument, and his result was extended to all graphs by Zhao [64]. There are at most

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 $(2^{d+1}-1)^{n/2d} = 2^{n/2+O(n/d)}$  independent sets in a *d*-regular graph of order *n* (that is, having *n* vertices), and this number is attained by n/2d disjoint copies of  $K_{d,d}$ .

It would be convenient for many purposes if there were at most  $2^{o(n)}$  independent sets in an *r*-graph *G* of order *n* and average degree *d*, but examples like that just cited show this hope to be a forlorn one. Nevertheless, for the applications we have in mind, it is enough to find a good collection *C* of *containers* for independent sets: this is a family of subsets of V(G) such that, for each independent set *I*, there is a set  $C \in \mathcal{C}$  with  $I \subset C$ , and  $|\mathcal{C}| = 2^{o(n)}$ . Of course, we could just take  $\mathcal{C} = \{V(G)\}$ , but this collection would not be helpful: for  $\mathcal{C}$  to be of use, a further condition is needed that each container  $C \in \mathcal{C}$  is not large, in a sense made precise later (see §3.2).

Another immediate candidate for C is the collection of maximal independent sets, but this too can be large; for example, if d is even, adding a 1-factor into the vertex classes of each  $K_{d,d}$  of the graph  $(n/2d)K_{d,d}$  produces a (d + 1)-regular graph with at least  $2^{n/4}$ maximal independent sets. (The maximum number of maximal independent sets in any graph of order n was determined by Moon and Moser [43].)

The main purpose of this paper is to show that every r-graph G of average degree d and order n does have a small collection C of containers. Typically, but not always,  $|C| \leq 2^{c_d n}$ where  $c_d$  is approximately  $d^{-1/(r-1)}$ . Results of this kind were known previously in special cases. Sapozhenko [49, 50, 51, 52, 53] treated regular 2-graphs. Containers for r-graphs were introduced and used in [55] for the restricted instance of simple regular r-graphs (a hypergraph is *simple* or *linear* if every pair of vertices lies in at most one edge). However, the most interesting applications require containers for non-regular r-graphs. Finding such containers presents significant difficulties and the method here is unrelated to that of [55]. (Nevertheless, the method of [55] is good enough to give easy proofs of some of the results here — see [58].)

We describe our main results about containers in §3. The fundamental result is Theorem 3.4 stated in §3.3. It is worth mentioning that the statement applies to all r-graphs Gbut it gives useful information only if d is large (though independently of n). In order to state the main theorem we need to introduce and motivate a couple of concepts (degree measure and the co-degree function), but their definitions are quite straightforward. This discussion all takes place in §3. The main result is, in some senses, optimal, as we shall explain.

As well as the main theorem, §3 includes two consequences of it, packaged for ready use in two different kinds of applications. These two varieties are worth emphasising, because they highlight two ways in which we might require a container C to be "not large": in one version e(G[C]) is small, which is to say that the container has only a few edges inside it, and in the other version |C| is small, meaning that the container does not have many vertices. These two situations are quite different in the way they are handled, though both are derived from the same main theorem.

The actual construction of the containers is given in §4. The construction is via an algorithm, just a few lines long. This algorithm is needed only for the proof of Theorem 3.4 and no understanding of it is required in order to apply the theorem; nevertheless the algorithm clearly lies at the heart of the whole process, and so §4 includes some discussion with the aim of illuminating what is going on.

In §5 we prove Theorem 3.4; this comes down to making some calculations that verify the performance of the algorithm. The calculations are mostly straightforward, though at one point we used a slightly more detailed argument than is necessary, in order to achieve better constants.

Having proved the main result, we proceed in §6 and §7 to derive the two consequences mentioned previously (we include as well a more technical version of one of them, useful in more sensitive applications). The optimality of the main theorem, or at least one aspect of it, is proved in §11, but a potentially better approach to the algorithm is mentioned in §12.

Before getting down to the details of the container theorem, in  $\S2$  we offer some motivation by outlining a few applications. The details of these are given in  $\S8-\S10$ .

1.1. A little notation. We use standard notation. In particular, for  $m, n \in \mathbb{N}$  we let  $[n] = \{1, \ldots, n\}$  and  $[m, n] = \{m, \ldots, n\}$ . For collections of subsets we write, for example,  $[m, n]^{(s)} = \{\sigma \subset [m, n] : |\sigma| = s\}, [m, n]^{(>s)} = \{\sigma \subset [m, n] : |\sigma| > s\}$ , and so on. As usual,  $\mathcal{P}(S)$  denotes the collection of all subsets of S; we omit parentheses where no confusion can arise, for instance writing  $\mathcal{P}[n]$  instead of  $\mathcal{P}([n])$ . If G is a hypergraph we write e(G) = |E(G)| for the number of edges of G and v(G) = |V(G)| for the number of vertices of G. If  $S \subset V(G)$  then G[S] denotes the subhypergraph of G induced by S, that is,  $G[S] = (S, E(G) \cap \mathcal{P}(S))$ .

#### 2. Some applications of containers

The purpose of this section is to highlight some results that follow from the existence of containers, in the hope of motivating the main result itself, Theorem 3.4. The applications involve list colouring, extremal graph theory, and solutions of linear equations.

The applications are of two essentially different kinds, namely those in which we require |C| to be bounded for each  $C \in C$ , and those where we require e(C) to be bounded. In fact we give only one application where |C| is bounded, namely the one about list colouring: hence Theorem 3.7 (the version of the container theorem packaged for bounds on |C|) is used only in this application.

The remaining applications require e(C) to be bounded. On the face of it they appear more numerous but this appearance is deceptive: for example, all the results concerning *H*free graphs, including those involving sparse random graphs, are actually direct corollaries of a single theorem about the class of *H*-free graphs, namely Theorem 2.3, and this theorem is the only place in the argument where the container theorem is invoked. Moreover, it is applied to just one hypergraph (more exactly, to one hypergraph G(N, H) for each *H* and *N*).

Likewise, the applications to solutions of linear equations involve translating some given problem into a question about the independent sets in a specific hypergraph G, then finding containers for this G, and then interpreting these containers back in the original context.

The list colouring application is thus rather different to the others but it is the one which originally motivated us (our early thoughts appeared in [55]), and it is the application which has shaped the algorithm that we use to construct containers.

The technical details of the list colouring and extremal graph theory applications are supplied later in §8–10. As for the arithmetical applications, we state them here in order to illustrate the use of the container theorem, but we give the details elsewhere [57], so as to maintain the focus here on the container theorem itself.

2.1. List colourings. A 2-graph G is said to be k-choosable if, whenever for each vertex  $v \in V(G)$  we assign a list  $L_v$  of k colours to v, then it is possible to choose a colour for v

from the list  $L_v$ , so that no two adjacent vertices receive the same colour. The *list chromatic* number  $\chi_l(G)$  (also called the *choice number*) is the smallest k such that G is k-choosable. If all the lists are the same then a list colouring is just an ordinary k-colouring and so  $\chi_l(G)$ is at least  $\chi(G)$ , the ordinary chromatic number of G. This natural definition was first studied by Vizing [63] and by Erdős, Rubin and Taylor [21]. One of the main discoveries of [21] is that  $\chi_l(G)$  can be much larger than  $\chi(G)$ , because  $\chi_l(K_{d,d}) = (1 + o(1)) \log_2 d$ , whereas  $\chi(K_{d,d}) = 2$ .

In fact, unlike  $\chi(G)$ ,  $\chi_l(G)$  must grow with the minimum degree of the graph G. Alon [2, 3] showed that  $\chi_l(G) \ge (1/2 + o(1)) \log_2 d$  holds for any graph G of minimum degree d.

The notion of k-choosability carries over directly to r-graphs, when it is understood that the vertex colours are chosen so that no edge has all its vertices the same colour. There is a straightforward reason, as pointed out by Alon and Kostochka [4] (see too Haxell and Pei [26]), why for  $r \geq 3$  it is not true for r-graphs G that  $\chi_{\ell}(G)$  grows with the average degree. Let F be some graph on n vertices, say  $F = (n/2)K_2$ , and let G be some r-graph each of whose edges contains an edge of F. Then  $\chi_l(G) \leq \chi_l(F)$ , so in this example  $\chi_l(G) = 2$ , whereas the average degree of G can be large. However, if we restrict to simple r-graphs the situation is different. Haxell and Pei [26] showed  $\chi_{\ell}(G) = \Omega(\log d/\log \log d)$  if G is a Steiner triple system, and Haxell and Verstraëte [27] proved that  $\chi_l(G) \geq (1 + o(1)) (\log d/5 \log \log d)^{1/2}$  for all simple d-regular 3-graphs G. Alon and Kostochka [4] showed  $\chi_l(G) \geq (\log d)^{1/(r-1)}$  for simple r-graphs G of average degree d, and in [55] it was shown that  $\chi_l(G) = \Omega(\log d)$  for simple d-regular r-graphs. We extend this to all simple r-graphs, at the same time giving a better constant.

**Theorem 2.1.** Let  $r \in \mathbb{N}$  be fixed. Let G be a simple r-graph with average degree d. Then, as  $d \to \infty$ ,

$$\chi_l(G) \ge (1+o(1)) \frac{1}{(r-1)^2} \log_r d$$

holds. Moreover, if G is regular then

$$\chi_l(G) \ge (1+o(1)) \frac{1}{r-1} \log_r d.$$

Note that, for r = 2, this improves Alon's bound [3] by a factor of 2 and is best possible. We think that the bound given for regular r-graphs might hold for general r-graphs and moreover that it too might be best possible (see §8). For colourings of non-simple r-graphs, see §12.

2.2. *H*-free graphs. An  $\ell$ -graph on vertex set [N] is said to be *H*-free if it contains no subgraph isomorphic to the  $\ell$ -graph *H*.

As far as *H*-free graphs are concerned, our main result is this: for any given  $\ell$ -graph *H*, though there are many *H*-free  $\ell$ -graphs, each of these is contained in one of a very small collection of  $\ell$ -graphs that are almost *H*-free. More exactly, there is a small collection C of  $\ell$ -graphs, each *H*-free  $\ell$ -graph being a subgraph of an  $\ell$ -graph in C, and no  $\ell$ -graph in C having more than  $o(N^{v(H)})$  copies of *H*. The main content of the theorem is that the size of C is very small. For graphs at least, Szemerédi's regularity lemma gives a collection with  $\log |\mathcal{C}| = o(N^2)$ , but the size of C in our theorem is much smaller. It is expressed in terms of a parameter m(H) that appears often in the literature.

**Definition 2.2.** For an  $\ell$ -graph H with  $e(H) \ge 2$ , let

$$m(H) = \max_{H' \subset H, e(H') > 1} \frac{e(H') - 1}{v(H') - \ell}.$$

Sometimes, H is called (strictly) balanced if the maximum is attained (uniquely) when H' = H. However, this restriction is not needed in any of our arguments and it is ignored.

We shall indicate shortly why the parameter m(H) might be expected to make an appearance here, but first we state our main theorem for H-free  $\ell$ -graphs. As usual, let ex(N, H) be the maximum number of edges in an H-free graph of order N and let  $\pi(H) = \lim_{N \to \infty} \exp(N, H) {\binom{N}{\ell}}^{-1}$ . The symbol  $\subset$  in the theorem means "is a subgraph of".

**Theorem 2.3.** Let H be an  $\ell$ -graph with  $e(H) \geq 2$  and let  $\epsilon > 0$ . For some c > 0 and for every  $N \geq c$ , there exists a collection  $\mathcal{C}$  of  $\ell$ -graphs on vertex set [N] such that

- (a) for every H-free  $\ell$ -graph I on vertex set [N], there exists  $C \in \mathcal{C}$  with  $I \subset C$ ,
- (b) for every  $\ell$ -graph  $C \in \mathcal{C}$ , the number of copies of H in C is at most  $\epsilon N^{v(H)}$ , and  $e(C) \le (\pi(H) + \epsilon) \binom{N}{\ell},$ (c)  $\log |\mathcal{C}| \le c N^{\ell - 1/m(H)} \log N,$
- (d) moreover, for every I in (a), there exists  $T = (T_1, \ldots, T_s)$  where  $T_i \subset I$ ,  $s \leq c$  and  $\sum_i e(T_i) \leq cN^{\ell-1/m(H)}$ , such that C = C(T).

The meanings of (a), (b) and (c) should be clear enough. Condition (a) is the basic property of the collection  $\mathcal{C}$ , namely that all H-free graphs are subgraphs of members of  $\mathcal{C}$ . Condition (b) is what is meant by the containers themselves being small, which is that each C contains few copies of H. This immediately implies the bound on e(C), via the supersaturation theorem of Erdős and Simonovits [55]. Condition (c) says that the collection  $\mathcal{C}$  is small.

Condition (d) should be understood in the following way. The notation C = C(T) is used to mean that C is a function of, or is determined by, T. Now T is a collection of subgraphs of I that are small, that is, have few edges. The point of condition (d) is that  $\mathcal{C}$  is therefore small, since  $|\mathcal{C}|$  is at most the number of possible objects T, which is a small number because the subgraphs comprising T are small. The bound on  $|\mathcal{C}|$  that (d) directly implies is the one given in condition (c), and for most purposes we could dispense with (d) because the bound in (c) is good enough. However, condition (d) gives slightly more information, namely that the graphs  $T_i$  comprising T are actually subgraphs of I and not just arbitrary graphs. This extra information can be just enough, in tight corners (specifically, in Lemma 10.3), to give a better result than what can be obtained by a direct use of (c) (effectively it removes the  $\log N$ ), and we retain (d) for this reason.

The existence of the collection  $\mathcal{C}$  follows straightforwardly from the results in §3, as shown in §9, by applying them to the e(H)-graph G = G(N, H), whose  $n = \binom{N}{\ell}$  vertices are the  $\ell$ -sets in [N], and whose edges are subsets of V(G) spanning a copy of H in [N]. The subsets of V(G) are then  $\ell$ -graphs with vertex set [N], and independent sets in G correspond to H-free  $\ell$ -graphs. The  $\ell$ -graphs in  $\mathcal{C}$  are simply the containers for the independent sets of Gsupplied by our main container theorem (more precisely, Corollary 3.6). In order to apply the container theorem and so obtain Theorem 2.3, all that is required is to calculate a simple parameter of G(N, H). Details are in §9.

We can now indicate why the parameter m(H) shows up in Theorem 2.3. It is well known, and not hard to check, that if  $t = o(N^{\ell-1/m(H)})$  then for some  $H' \subset H$  almost all  $\ell$ graphs on N vertices with t edges contain many fewer copies of H' than they do edges. Thus most subsets of V(G(N, H)) of size t are independent, or close to it (that is, contain many fewer edges than vertices). For reasons discussed in §3.6, this means that  $e^{\Omega(t)}$  containers are needed, and from this standpoint, Theorem 2.3 is more or less best possible. Perhaps a more convincing demonstration of optimality is that an improvement in the bound on  $|\mathcal{C}|$  in Theorem 2.3 would directly improve, say, the bound on p in Theorem 2.12, but the bound there is well known (and readily checked) to be optimal.

We remark that Theorem 2.3 can be extended so that the  $\ell$ -graphs I need not be independent: they need only have few copies of H. The extension is given in Theorem 9.2 but, again, we defer the technicalities to §9.

The remainder of this section includes further results on *H*-free  $\ell$ -graphs; as pointed out earlier, they are all (except Corollary 2.5) just consequences of Theorem 2.3.

2.3. The number of *H*-free graphs. How many *H*-free  $\ell$ -graphs are there altogether on vertex set [N]? Choosing any maximum *H*-free graph and taking all its subgraphs supplies at least  $2^{(\pi(H)+o(1))\binom{N}{\ell}}$  *H*-free graphs. But each *H*-free graph is a subgraph of a member of the collection C given by Theorem 2.3, so the total number of *H*-free graphs is at most  $|C|2^{\max_{C \in C} e(C)}$ . Now,  $\max_{C \in C} e(C) \leq (\pi(H) + o(1))\binom{N}{\ell}$ , and Theorem 2.3 shows that  $|C| = 2^{o(N^{\ell})}$ , giving the following immediate consequence.

**Corollary 2.4.** Let H be an  $\ell$ -graph. The number of H-free  $\ell$ -graphs on vertex set [N] is  $2^{(\pi(H)+o(1))\binom{N}{\ell}}$ .

In the case  $\ell = 2$ , this was proved for complete *H* by Erdős, Kleitman and Rothschild [19] and for general *H* by Erdős, Frankl and Rödl [18]. Nagle, Rödl and Schacht [45] proved it for general  $\ell$  using hypergraph regularity methods.

For  $\ell$ -graphs H which satisfy  $ex(N, H) = o(N^{\ell})$  (when  $\ell = 2$  this means H is bipartite), we have  $\pi(H) = 0$ , and Corollary 2.4 is unhelpful. Nevertheless our results can still be useful, provided appropriate information about G(N, H) is available. The simplest case is  $\ell = 2$  and  $H = K_{2,2} = C_4$ , where it is well known that  $ex(N, C_4) = (1/2 + o(1))N^{3/2}$ (Erdős, Rényi and Sós [20]), implying the trivial upper bound  $2^{O(N^{3/2} \log N)}$  for the number of  $C_4$ -free graphs. Theorem 6.3 describes what happens if we apply the main container theorem repeatedly to a hypergraph: by applying this theorem to  $G(N, C_4)$ , the following can be obtained.

# **Corollary 2.5.** The number of $C_4$ -free graphs on vertex set [N] is at most $2^{(300+o(1))N^{3/2}}$ .

We shall not prove Corollary 2.5; we state it just as an illustration of what can be derived by plugging numbers into a generic container theorem. The argument is very similar to that for the upper bound in Theorem 2.11 on the number of Sidon sets, which also relies on Theorem 6.3, and whose details can be found in [57]. The reason we do not give details for Corollary 2.5 is that Kleitman and Winston [33] obtained a finer bound, namely  $2^{(1.082+o(1))N^{3/2}}$ . The number of  $K_{s,t}$ -free graphs has been well estimated by Balogh and Samotij [7]. Recently, Morris and the first author [44], using container methods and other techniques, have shown that the number of  $C_{2k}$ -free graphs on vertex set [N] is at most  $2^{O(N^{1+1/k})}$ , where  $C_{2k}$  is the cycle of length 2k. The order of the extremal function  $ex(N, C_{2k})$  is unknown in general, though Bondy and Simonovits [9] proved  $ex(N, C_{2k}) = O(N^{1+1/k})$ . Nevertheless, it is further shown in [44] that, for some c > 0, there are more than  $2^{(1+c)ex(N,C_6)} C_6$ -free graphs of order N for infinitely many N.

2.4. Induced-*H*-free graphs. Alongside the many results about *H*-free graphs, there is a corresponding corpus about *induced H*-free graphs, that is, graphs with no induced subgraph isomorphic to *H*. The number of induced *H*-free graphs was closely estimated by Prömel and Steger [46], and there have been many subsequent refinements.

If I is an induced H-free  $\ell$ -graph, we need to ask what kind of object C must be in order that the inclusion  $I \subset C$  is helpful; if, as in Theorem 2.3, C itself is just an  $\ell$ -graph and  $I \subset C$  means I is a subgraph of C, then the induced subgraphs of I differ from those of C, which is no use. We borrow the notion of 2-coloured multigraph from [41, 62]. A 2-coloured  $\ell$ -multigraph C on vertex set [N] is a pair of edge sets  $C_R, C_B \subset [N]^{(\ell)}$ , which we call the red and the blue edge sets. Let I be an  $\ell$ -graph on [N]. Then we write  $I \subset C$ if  $E(I) \subset C_R$  and  $[N]^{(\ell)} \setminus E(I) \subset C_B$ . Thus edges in  $C_R \cap C_B$  always help towards the inclusion  $I \subset C$ . If we construct (see §9) a hypergraph akin to G(N, H) but which encodes both red and blue edges, and apply the container theorem to it, we obtain the following analogue of Theorem 2.3.

**Theorem 2.6.** Let H be an  $\ell$ -graph and let  $\epsilon > 0$ . For some c > 0 and for N sufficiently large, there exists a collection C of 2-coloured  $\ell$ -multigraphs on vertex set [N] such that

- (a) for every  $\ell$ -graph I on vertex set [N] with no induced copy of H there exists  $C \in C$  with  $I \subset C$ ,
- (b) for every  $C \in \mathcal{C}$ , the number of copies of H in C is at most  $\epsilon N^{v(H)}$ ,
- (c)  $\log |\mathcal{C}| \le c N^{\ell (v(H) \ell) / \left( \binom{v(H)}{\ell} 1 \right)} \log N.$

Note that  $(v(H) - \ell)/(\binom{v(H)}{\ell} - 1) = 1/m(K)$  where K is the complete  $\ell$ -graph of order v(H). The form of this theorem is, to an extent, reminiscent of Theorem 2.3, and it arises from the method of proof in which an induced copy of H is modelled as a red-blue colouring of K. However the value m(K) is not invariably optimal: for example, when  $\ell = 2, v(H) = 4$  and H is an induced path of length three, then the number of induced H-free graphs (sometimes known as *cographs*) is only  $N^{O(N)}$ , so these graphs themselves comprise a smaller collection of containers than that offered by the theorem.

Theorem 2.6 can be used to recover basic results, akin to Corollary 2.4, about the number of induced H-free  $\ell$ -graphs. In fact we can state a probabilistic version just as readily. Let  $G^{(\ell)}(N,p)$  be a random  $\ell$ -graph obtained by choosing edges independently from the complete  $\ell$ -graph  $K_N^{(\ell)}$  with probability p. Our result is stated in terms of a function  $h_p(H)$ , defined as follows (and discussed further, in slightly different terminology, in [42]). For a 2-coloured  $\ell$ -multigraph J, with vertex set [N] and having red and blue edge sets  $J_R$  and  $J_B$ , let

$$H_p(J) = -|J_R \setminus J_B| \log_2 p - |J_B \setminus J_R| \log_2(1-p).$$

The point of this definition is that, if  $J_R \cup J_B = [N]^{(\ell)}$ , then the probability that  $G^{(\ell)}(N,p)$  is a subgraph of J is  $2^{-H_p(J)}$ . Let

$$hex_p(H, N) = \min\{ H_p(J) : J_R \cup J_B = [N]^{(\ell)}, H \not\subset J \}.$$

Then we put  $h_p(H) = \lim_{N \to \infty} \max_p(H, N) {\binom{N}{\ell}}^{-1}$  (this limit exists, since an averaging argument shows  $(N - \ell) \max_p(H, N) \ge N \max_p(H, N - 1)$ ).

**Theorem 2.7.** Let 0 be constant and let <math>H be an  $\ell$ -graph. Then

 $\mathbb{P}(G^{(\ell)}(N,p) \text{ is induced-}H\text{-}free) = 2^{-(h_p(H)+o(1))\binom{N}{\ell}}.$ 

For graphs, that is,  $\ell = 2$ , this theorem was proved for p = 1/2 by Prömel and Steger [46, Theorem 1.3] and for general p by Bollobás and Thomason [8, Theorem 1.1] (clarified by Marchant and Thomason [42]). For p = 1/2 and  $\ell = 3$  it was proved by Kohayakawa, Nagle and Rödl [36] using hypergraph regularity techniques, Dotson and Nagle [15] extending this to general  $\ell$ .

It can be imagined that arguments similar to those described here could be used to obtain container results about other structures, such as tournaments.

2.5. Linear equations. Let F be either a finite field or the set of integers [N]. We consider linear systems of equations Ax = b, where A is a  $k \times r$  matrix with entries in  $F, x \in F^r$  and  $b \in F^k$ . We call such a triple (F, A, b) a  $k \times r$  linear system.

**Definition 2.8.** For a  $k \times r$  linear system (F, A, b), a subset  $I \subset F$  is *solution-free* if there is no  $x \in I^r$  with Ax = b, and ex(F, A, b) is the maximum size of a solution-free subset.

The notion of a solution-free subset is analogous to that of an H-free hypergraph in §2.2. Once again, our contribution to this topic is a container theorem for solution-free sets. It is obtained by constructing a hypergraph G whose independent sets correspond to solutionfree sets, after which a simple check of some parameter of G allows the container theorem to be applied. A precise statement, however, requires one or two technical definitions, so we omit it from here and refer the reader to [57].

Nevertheless we mention a consequence for counting solution-free subsets. For an equation Ax = b, how many solution-free subsets of F are there? A well-known instance of this question is to find the number of subsets  $S \subset [N]$  containing no solution to x + y = z; the asymptotic answer, conjectured by Cameron and Erdős [11], was given by Green [28] and by Sapozhenko [53].

For a general system, every subset of a solution-free set is itself solution-free, so there are at least  $2^{\text{ex}(F,A,b)}$  solution-free sets. For a single equation (the case k = 1), it was shown by Green [29] that there are at most  $2^{\text{ex}(F,A,b)+o(|F|)}$  solution-free subsets; Sapozhenko too [51] has results of this kind.

The same bound does not always hold for  $k \ge 2$ . If some variables are closely tied to other variables — say the equations imply that x = y — then there can be significantly more than  $2^{\text{ex}(F,A,b)}$  solution-free sets. However, a (perhaps non-standard but) natural condition on A rules out closely tied variables, and in this case the stated bound holds good.

**Definition 2.9.** We say that A has *full rank* if given any  $b \in F^k$  there exists  $x \in F^r$  with Ax = b. We then say that A is *abundant* if it has full rank and every  $k \times (r-2)$  submatrix obtained by removing a pair of columns from A still has full rank.

**Theorem 2.10.** There is a function  $f : \mathbb{N} \to \mathbb{R}$  with f(n) = o(n) such that if F is a finite field and (F, A, b) is a  $k \times r$  linear system with A abundant, then the number of solution-free subsets of F is at most  $2^{\text{ex}(F,A,b)+f(|F|)}$ .

Likewise, for each fixed integer matrix A, there is a function  $g : \mathbb{N} \to \mathbb{R}$  with g(n) = o(n)such that if ([N], A, b) is a  $k \times r$  linear system with A abundant, then the number of solutionfree subsets of [N] is at most  $2^{ex([N],A,b)+g(N)}$ .

For example, take A = (1, 1, -1) and b = (0). Theorem 2.10 says that the number of sumfree subsets of [N] is  $2^{N/2+o(N)}$ , giving a new proof of the weak form of the Cameron-Erdős conjecture, proved independently by Alon [1], by Calkin [10] and by Erdős and Granville (unpublished). Interestingly, whilst our container method for 2-graphs is closely related to arguments of Sapozhenko in [53], our derivation of the weak Cameron-Erdős conjecture is via 3-uniform hypergraphs and differs from that in [53].

Similar results hold when F is an abelian group. For the proof of Theorem 2.10 we need the fact that a subset containing few solutions is close in size to a solution-free subset. There appears to be no analogue to the simple supersaturation results that helped us at similar points in §2.2 and §2.4, so here we invoke the various removal lemmas of Shapira [60] and of Král', Serra and Vena [38, 39], extending Green's original lemma [29].

For linear systems where  $\exp(F, A, b) = o(|F|)$ , Theorem 2.10 is uninformative. One of the most prominent examples is that of Sidon sets. A set  $A \subset [n]$  is Sidon if every sum of two elements is distinct, i.e., there are no solutions to w + x = y + z with  $\{w, x\} \neq \{y, z\}$ . It is easy to see that a Sidon set has size at most  $\lceil \sqrt{2n} \rceil$ , since each of the |S|(|S|-1)/2 values x - y, where  $x, y \in S$  and y < x, are distinct and lie in  $\{1, \ldots, n-1\}$ . Erdős and Turán [23] improved this upper bound to  $|S| \leq (1 + o(1))\sqrt{n}$ , and there are examples achieving this bound.

It is natural to ask, as Cameron and Erdős did [11], how many Sidon sets there are, and the answer clearly lies between  $2^{(1+o(1))\sqrt{n}}$  and  $2^{O(\sqrt{n}\log n)}$ . Neither of these bounds, it turns out, is tight.

## **Theorem 2.11.** There are between $2^{(1.16+o(1))\sqrt{n}}$ and $2^{(55+o(1))\sqrt{n}}$ Sidon subsets of [n].

The lower bound gives a negative answer to the open question of whether there are only  $2^{(1+o(1))\sqrt{n}}$  Sidon sets. The upper bound follows from an application of Theorem 6.3, similar to that in the proof of Corollary 2.5. Kohayakawa, Lee, Rödl and Samotij [34] have obtained an upper bound of the same kind and with a better constant. For details see [57].

2.6. Sparsity. In recent times, there has been interest in the extent to which theorems holding for dense structures hold also for sparse random substructures. Our results can be applied in this context, and we give some illustrative examples involving the notions of H-free graphs and solution-free subsets already discussed.

The application of our results always fits a simple paradigm. Typically we want some statement to hold for a random substructure, with high probability; by considering an appropriate collection of containers, the fact that there are a small number of containers means that the work is reduced, via the union bound, to establishing a (generally much simpler) statement for a single container.

For example, consider a random  $\ell$ -graph  $G^{(\ell)}(N,p)$ , as defined in §2.4. Evidently there are *H*-free subgraphs of  $G^{(\ell)}(N,p)$  with  $p \exp(N,H)$  edges, but are there significantly larger *H*-free subgraphs? It was conjectured by Haxell, Kohayakawa and Luczak [24, 25], and by Kohayakawa, Luczak and Rödl [35], that if  $pN^{1/m(H)} \to \infty$  then *H*-free subgraphs of  $G^{(\ell)}(N,p)$  almost surely have at most  $(1 + o(1))p \exp(N,H)$  edges. This conjecture was recently proved by Conlon and Gowers [12] (for strictly balanced *H*) and by Schacht [59], using different methods. Our methods give an alternative proof. For each container  $C \in C$  given by Theorem 2.3, it is easily seen that, with high probability,  $G^{(\ell)}(N,p)$  contains not much more than  $pe(C) \leq (\pi(H) + o(1))p\binom{N}{\ell}$  edges of C. By the union bound this holds for all  $C \in C$ , and hence also for all H-free  $\ell$ -graphs.

**Theorem 2.12.** Let H be an  $\ell$ -graph and let  $0 < \gamma < 1$ . For some c > 0, for N sufficiently large and for  $p \ge cN^{-1/m(H)}$ , the following event holds with probability greater than  $1 - \exp\{-\gamma^3 p\binom{N}{\ell}/512\}$ :

every H-free subgraph of  $G^{(\ell)}(N,p)$  has at most  $(\pi(H) + \gamma)p\binom{N}{\ell}$  edges.

Kohayakawa, Luczak and Rödl [35] further conjectured a stability version of Theorem 2.12, proved by Conlon and Gowers [12] for strictly balanced graphs and by Samotij [48], following Schacht [59], for all graphs. They also made a stronger, technical, conjecture which has become known as the KLR conjecture, proved recently for balanced graphs by Balogh, Morris and Samotij [6]. Theorem 2.3 can be used to derive all these conjectures in a straightforward way, and indeed a counting version of Theorem 2.3 (Theorem 9.2) yields a counting version of the KLR conjecture. Because of the technical descriptions needed, and the fact that these results are consequences of Theorem 2.3, we defer further details to §10.

The same arguments can be applied to solution sets of linear equations. Here is a typical consequence.

**Theorem 2.13** (Conlon and Gowers [12], Schacht [59]). Let  $\ell \geq 3$  and  $\epsilon > 0$ . There exists a constant c > 0 such that for  $p \geq cN^{-1/(\ell-1)}$ , if  $X \subset [N]$  is a random subset chosen with probability p, then with probability tending to 1 as  $N \to \infty$ , any subset of X of size  $\epsilon |X|$ contains an arithmetic progression of length  $\ell$ .

Further examples and details can be found in [57].

#### 3. Containers

A couple of simple notions are needed for the statement of the main theorem, and we define these now. They are the co-degree function and degree measure. The co-degree function is what will determine the number  $|\mathcal{C}|$  of containers needed. The size of individual containers will be specified in terms of degree measure.

3.1. The co-degree function  $\delta(G, \tau)$ . The main difficulties in the construction of containers are already present in the case of simple hypergraphs, where the authors' original motivation lay. However the method can be adapted efficiently to any hypergraph. The number of containers we construct (and to a much lesser extent their size) depends on the way the edges overlap, but the dependence can be encapsulated by a single parameter. This parameter appears in most of the theorems.

The theorems are stated in terms of a parameter  $\tau$ , whose meaning will become clearer later, but for now it is enough to say that the number  $|\mathcal{C}|$  of containers constructed will be approximately  $2^{\tau n}$ . It is evident, then, that we shall want  $\tau$  to be as small as possible. What determines how small  $\tau$  can be is a bound on the *co-degree function*  $\delta(G, \tau)$ . This function is usually quite straightforward to compute; it is just a polynomial in  $1/\tau$  whose coefficients are expressed in terms of the edge overlaps in G. Here is the precise definition. We first define the degree of a subset of vertices, in the natural way.

**Definition 3.1.** The *degree* of a set of vertices  $\sigma \subset V(G)$  is the number of edges containing  $\sigma$ ; that is,

$$d(\sigma) = |\{ e \in E(G) : \sigma \subset e \}|.$$

If  $|\sigma| = 1$ , that is  $\sigma = \{v\}$  where  $v \in V(G)$ , we generally write d(v) instead of  $d(\{v\})$ .

We can now define the co-degree function  $\delta(G, \tau)$ .

**Definition 3.2.** Let G be an r-graph of order n and average degree d. Let  $\tau > 0$ . Given  $v \in V(G)$  and  $2 \le j \le r$ , let

$$d^{(j)}(v) = \max\left\{ d(\sigma) : v \in \sigma \subset V(G), \, |\sigma| = j \right\}.$$

If d > 0 we define  $\delta_j$  by the equation

$$\delta_j \tau^{j-1} n d = \sum_v d^{(j)}(v) \,.$$

Then the co-degree function  $\delta(G, \tau)$  is defined by

$$\delta(G,\tau) = 2^{\binom{r}{2}-1} \sum_{j=2}^{r} 2^{-\binom{j-1}{2}} \delta_j.$$

If d = 0 we define  $\delta(G, \tau) = 0$ .

The powers of 2 in the definition are rather eye-catching but they are a distraction; they are constants introduced to make Lemma 5.5 work smoothly (see the comment in §12). It does no harm for now to ignore them and to think of  $\delta(G, \tau)$  as  $\sum \delta_j$  or even as max  $\delta_j$ .

Given a hypergraph G, the degree function  $\delta(G, \tau)$  is a polynomial in  $1/\tau$  with positive coefficients (provided  $e(G) \neq 0$ ); in particular  $\delta(G, \tau)$  increases to infinity as  $\tau$  decreases to zero. One of the conditions of the main theorem, Theorem 3.4, and of most of the other theorems, is an upper bound on  $\delta(G, \tau)$ , which clearly is equivalent to a lower bound on  $\tau$ .

It is helpful to have some feel for what values  $\tau$  might take, and here are some observations intended to indicate what happens. A typical application will involve making  $\delta(G, \tau)$  less than some constant, never larger than 1/r!, so let us see what this implies for  $\tau$ .

First of all, consider the simplest case, that of an ordinary graph, when r = 2. Then  $d^{(2)}(v) = 0$  or 1, so  $\delta_2 \tau nd \leq n$ , that is,  $\delta_2 \leq 1/\tau d$ . Hence  $\delta(G, \tau) = \delta_2 \leq 1/\tau d$ . Thus  $\delta(G, \tau)$  is small provided  $\tau$  is larger than 1/d.

For general r, observe that, unless G has isolated vertices,  $d^{(j)}(v) \ge 1$  holds for all v, and so  $\delta_j \ge \tau^{1-j}/d$ . The largest of these bounds is  $\delta_r \ge \tau^{1-r}/d$  ( $\tau$  is invariably less than one) and so, for fixed r and large d, it will always be that for  $\delta(G, \tau)$  to be small we must choose  $\tau$  at least as large as  $d^{-1/(r-1)}$ .

In a simple hypergraph,  $d(\sigma) \leq 1$  holds whenever  $|\sigma| \geq 2$ , and so  $\delta_j \leq \tau^{1-j}/d$ . In this case the largest of the  $\delta_j$ 's is  $\delta_r$ , and we can make  $\delta(G, \tau)$  small by choosing  $\tau$  just a little larger than  $d^{-1/(r-1)}$ . In fact, for any hypergraph whose edges are sufficiently uniformly distributed,  $\delta_r$  is once again the  $\delta_j$  which dominates, as a simple calculation (which we omit) shows, so here again  $\delta(G, \tau)$  is small provided  $\tau$  is larger than  $d^{-1/(r-1)}$ .

Sometimes, though, the dominant  $\delta_j$  is not  $\delta_r$ . One example of this is in the case of Sidon sets: when  $|S| < n^{2/3}$  it is the value of  $\delta_2$  which is the most important (see [57]).

Another example is the hypergraph describing H-free  $\ell$ -graphs: here the most important  $\delta_j$  is determined by whichever subgraph  $H' \subset H$  achieves the maximum of  $(e(H')-1)/(v(H')-\ell)$ , and this is how m(H) enters in (see Lemma 9.3). But in each of our examples the values are easily checked.

In summary, we must always choose  $\tau \geq d^{-1/(r-1)}$ , and for simple or uniformly distributed hypergraphs the value need not be much larger. But there are applications which are far from uniformly distributed, where  $\tau$  needs to be larger and where the behaviour of  $\delta(G, \tau)$  will prove crucial.

3.2. **Degree measure.** We mentioned in the introduction that the containers must not be too large. For some applications it suffices that  $|C| \leq (1-c)|G|$  for some constant c. This is achievable for regular hypergraphs but it clearly is unattainable in general; for example, if  $G = K_{d,n-d}$  (which, for large n, has average degree close to 2d) then some container must have size at least n-d. Other applications require that the number of edges inside a container, that is, e(G[C]), is small. This is always attainable, but a bound on e(G[C]) does not of itself imply a bound on |C| suitable for the first kind of application.

We in fact measure the size of containers by what we call *degree measure*. It turns out that if the degree measure is bounded then it is possible to recover all the properties of containers that are needed.

**Definition 3.3.** Let G be an r-graph of order n and average degree d. Let  $S \subset V(G)$ . The degree measure  $\mu(S)$  of S is defined by

$$\mu(S) = \frac{1}{nd} \sum_{u \in S} d(u) \,.$$

Thus  $\mu$  is a probability measure on V(G). Note that if G is regular then  $\mu(S) = |S|/n$ , which is the *uniform* measure of S. Thus a bound on  $\mu(S)$  automatically gives a bound on |S| for regular graphs. For general graphs, obtaining a useful bound on |S| from a bound on  $\mu(S)$  is a little more indirect (Lemma 7.2).

The dependence of e(G[S]) on  $\mu(S)$  is much more straightforward, by reason of the following inequality, in which G is an r-graph of order n and average degree d:

$$e(G[S]) \le (1/r) \sum_{v \in S} d(v) = (1/r)\mu(S)nd = \mu(S)e(G).$$
(1)

Hence a bound on  $\mu(S)$  at once gives a bound on e(G[S]).

We mentioned at the outset of the paper that each container should not be too large. In a regular r-graph it is easily shown that  $|I| \leq (1 - 1/r)n$  for every independent set I. However, for general r-graphs the ratio |I|/n can be arbitrarily close to one. An important feature of degree measure is that the bound  $\mu(I) \leq 1 - 1/r$  holds for all independent sets in all r-graphs (see inequality (2) and the remark following it). So we might hope that every r-graph has containers with  $\mu(C)$  bounded away from one, and this is exactly how things turn out. A bound of this kind is enough to meet our needs. In the next subsection we state the main theorem, and afterwards, in §3.4 and §3.5, we indicate how it can yield containers either with e(G[C]) small or with |S| small.

3.3. The main theorem. The essential idea underlying the main theorem is this: there is an algorithm that, from any small set  $T \subset V(G)$  of vertices, produces another subset  $C = C(T) \subset V(G)$ . Typically, C(T) is much larger than T, but it is guaranteed that  $\mu(C)$ 

is bounded away from one. Moreover, and importantly, for any independent set I, there is some small subset  $T \subset I$  such that  $I \subset C(T)$ .

Observe now that if we define  $\mathcal{C}$  to be the collection of all sets C(T) produced from small sets T, then this collection  $\mathcal{C}$  is a collection of containers having exactly the properties we want. The construction guarantees that for each independent set I there is a container  $C \in \mathcal{C}$  with  $I \subset C$ , with  $\mu(C)$  bounded away from one, and the number  $|\mathcal{C}|$  of containers is at most the number of small sets, which gives a useable upper bound on  $|\mathcal{C}|$ .

We already introduced the parameter  $\tau$ . This parameter will measure how big the sets T must be in order for the theorem to work: essentially,  $\tau$  will be the value of  $\mu(T)$ . For regular G this means  $|T| = \tau n$ , and the number of sets of this size is (*very* approximately)  $2^{\tau n}$ , so explaining the bound  $|\mathcal{C}| \leq 2^{\tau n}$  referred to in §3.1. This is why we want  $\tau$  to be as small as possible.

The theorem includes a parameter  $\zeta$ , which is some small constant at our disposal. Often we shall take  $\zeta = 1/12r!$  but sometimes it is useful (such as in the list colouring application) to choose a smaller value. The constraint on the size of  $\tau$  in the theorem arises from the requirement that  $\delta(G, \tau) \leq \zeta$ . As discussed in §3.1, this inequality implies a lower bound on  $\tau$ . We would thus want to take  $\zeta$  as large as possible, but taking it too large spoils the bound on  $\mu(C)$ . The choice  $\zeta = 1/12r!$  generally works well. Recall from the discussion in §3.1 that  $\tau$  is then a small negative power of d. Thus, for large  $d, \tau$  will be vanishingly small compared to the constants r and  $\zeta$ .

The preceding comments should help to illuminate the main thrust of the theorem, but some further comments should be made about the detailed statement.

First of all, we shall not actually generate C from just one small set but instead from an r-tuple  $T = (T_{r-1}, \ldots, T_0) \in \mathcal{P}(I)^r$  of small sets. The principles of the remarks made above remain the same. The precise bound on the size of T is  $\mu(T_i) \leq 2\tau/\zeta$ , which (in the light of what has been said) is of order  $\tau$ .

Secondly, we use a piece of shorthand. Let  $T = (T_{r-1}, \ldots, T_1, T_0) \in \mathcal{P}([n])^r$  and let  $w \in [n]$ . Then we define  $T \cap [w] = (T_{r-1} \cap [w], \ldots, T_1 \cap [w], T_0 \cap [w])$ . The relevance of this will be discussed later.

Finally, we say that an r-graph H is b-degenerate if  $e(H[S]) \leq b|S|$  for every subset  $S \subset V(H)$ .

**Theorem 3.4.** Let G be an r-graph with vertex set [n]. Let  $\tau, \zeta > 0$  satisfy  $\delta(G, \tau) \leq \zeta$ . Then there is a function  $C : \mathcal{P}([n])^r \to \mathcal{P}([n])$ , such that, for every independent set  $I \subset [n]$  there exists  $T = (T_{r-1}, \ldots, T_0) \in \mathcal{P}(I)^r$  with

- (a)  $I \subset C(T)$ ,
- (b)  $\mu(T_0), \dots, \mu(T_{r-1}) \le 2\tau/\zeta$ ,
- (c)  $|T_0|, \ldots, |T_{r-1}| \le 2\tau n/\zeta^2$ , and
- (d)  $\mu(C(T)) \le 1 1/r! + 4\zeta + 2r\tau/\zeta$ .

Moreover if G is simple then  $C(T) \cap [w] = C(T \cap [w]) \cap [w]$  for all  $T \in \mathcal{P}([n])^r$  and  $w \in [n]$ . In fact, the above is true for all sets  $I \subset [n]$  for which either G[I] is  $\lfloor \tau^{r-1}\zeta e(G)/n \rfloor$ -degenerate or  $e(G[I]) \leq 2r\tau^r e(G)/\zeta$ .

**Remark 3.5.** The discussion preceding the theorem has hopefully helped to explain it, but a few more observations are worth making.

• Roughly speaking, the theorem says that for each I there exists  $T \subset I$  with  $\mu(T) \lesssim \tau$ ,  $I \subset C(T)$  and  $\mu(C) \lesssim 1 - 1/r!$ , provided  $\tau$  is large enough to make  $\delta(G, \tau)$  small.

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- Assertions (b) and (c) each offer different, though obviously related, ways to bound the size of T; for each bound, there are applications where it is the more convenient.
- We refer to the property  $C(T) \cap [w] = C(T \cap [w]) \cap [w]$ , which holds for simple graphs (but see §12), as the *online property*, because the construction is behaving somewhat like an online algorithm: the vertices of the container lying within the first w vertices are already determined by  $T \cap [w]$ . (Nevertheless, knowledge of the whole of G is needed to determine  $C(T \cap [w])$ .) The online property is important only for certain applications, principally Theorem 3.7. For now, the property can safely be ignored.
- The container construction method makes essentially no use of the independence of the sets I, so we include an extension to two kinds of sparse subset, where either G[I] is b-degenerate for some small b, or else  $e(G[I]) \leq bn$  for some other b. Both types of sparsity are useful. Allowing  $e(G[I]) \leq bn$  is used in Theorem 9.2, and hence Theorem 10.2; b-degeneracy is used in Theorem 8.2.

As mentioned in the introduction, Theorem 3.4 has a variety of consequences and weaker forms which are easier to apply directly. We state a couple of them now: they are developed further in  $\S6-\S7$ .

3.4. Tight containers. The first corollary is packaged for use when we want e(G[C]) to be small for each container C. This is the corollary we use to prove, say, Theorem 2.3 in §9. It makes no mention of degree measure.

The way to make e(G[C]) small is to apply the container theorem repeatedly, as follows. Suppose I is an independent set in G. Observe that Theorem 3.4 gives a container C with  $I \subset C$  and  $\mu(C) \leq 1 - c$ , where c is some positive constant (perhaps around 1/r!). By inequality (1) this means  $e(G[C]) \leq (1 - c)e(G)$ . But I is an independent subset in G[C], so we can apply Theorem 3.4 again, this time to the graph G[C], to obtain a container C' with  $I \subset C'$  and  $e(G[C']) \leq (1 - c)e(G[C]) \leq (1 - c)^2e(G[C])$ . Repeated applications allow us to obtain containers with as few edges as we wish, the only constraint being that the main theorem ceases to be effective when the number of edges remaining is very small. Of course, repeated applications increase the total number of containers, but this turns out to be inexpensive.

The following corollary is the simplest of the ones obtained in this way, in which, as usual, the size of the collection of containers is bounded by a simple function of  $\tau$ .

**Corollary 3.6.** Let G be an r-graph on vertex set [n]. Let  $0 < \epsilon, \tau < 1/2$ . Suppose that  $\tau$  satisfies  $\delta(G, \tau) \leq \epsilon/12r!$ . Then there exists a constant c = c(r), and a function  $C : \mathcal{P}([n])^s \to \mathcal{P}[n]$  where  $s \leq c \log(1/\epsilon)$ , with the following properties. Let  $\mathcal{T} = \{(T_1, \ldots, T_s) \in \mathcal{P}([n])^s : |T_i| \leq c\tau n, 1 \leq i \leq s\}$ , and let  $\mathcal{C} = \{C(T) : T \in \mathcal{T}\}$ . Then

- (a) for every independent set I there exists  $T = (T_1, \ldots, T_s) \in \mathcal{T} \cap \mathcal{P}(I)^s$  with  $I \subset C(T) \in \mathcal{C}$ ,
- (b)  $e(G[C]) \leq \epsilon e(G)$  for all  $C \in \mathcal{C}$ ,
- (c)  $\log |\mathcal{C}| \le c \log(1/\epsilon) n \tau \log(1/\tau)$ .

Moreover, (a) holds for all sets  $I \subset [n]$  for which either G[I] is  $\lfloor \epsilon \tau^{r-1} e(G)/12r!n \rfloor$ degenerate or  $e(G[I]) \leq 24\epsilon r! \tau \tau^r e(G)$ .

The main points here are again that (a) shows there is a container for each independent set, (b) shows that each container has few internal edges, and (c) bounds the size of  $|\mathcal{C}|$ .

Condition (a) contains the extra information that the container C(T) for I is constructed from T, a few small subsets of I, which can be useful sometimes, as mentioned in the discussion after Theorem 2.3.

The corollary holds provided  $\delta(G, \tau)$  is bounded above as specified. As discussed in §3.1, this implies a lower bound on  $\tau$ . In applications where it matters, the value  $c(r) = 800r!^3r$  can be taken.

Corollary 3.6 is proved in  $\S6$ , together with a finer result of this kind.

3.5. Uniformly bounded containers. Next we give a consequence of Theorem 3.4 packaged for applications when the size |C| of the container is of interest. We shall use it to prove the list colouring result, Theorem 2.1 in §2.1. The package is somewhat more subtle than Corollary 3.6. We would like a bound on |C| of the form (1-c)n for some constant c, but, as noted before in §3.2, this does not always hold, say when  $G = K_{d,n-d}$ . What can be said in such circumstances that is useful?

Given  $S \subset V(G)$ , write  $\overline{S}$  for V(G) - S and  $e(\overline{S}, S)$  for the number of edges meeting both  $\overline{S}$  and S. The sum  $\sum_{v \in S} d(v)$  counts edges inside G[S] r times each, together with edges meeting both  $\overline{S}$  and S at most r - 1 times each. Hence

$$\sum_{v \in S} d(v) - re(G[S]) \le (r-1)e(\overline{S}, S) \le (r-1)\sum_{v \notin S} d(v) = (r-1)\mu(\overline{S})nd$$

or, in other words,  $\mu(S) - re(G[S])/nd \leq (r-1)\mu(\overline{S})$ . Now  $\mu(\overline{S}) = 1 - \mu(S)$  and so we have

$$e(G[S]) \ge (\mu(S) - 1 + \frac{1}{r}) \, nd$$
. (2)

In particular, as mentioned in §3.2, if I is independent then  $\mu(I) \leq 1 - 1/r$ . Moreover if  $e(G[C]) \leq \epsilon e(G) = \epsilon nd/r$ , as in Corollary 3.6, then  $\mu(C) \leq 1 - 1/r + \epsilon/r$ . Now if G is regular, then degree and uniform measures coincide; therefore in this case Corollary 3.6 supplies containers with  $|C| \leq (1 - 1/r + \epsilon/r)n$ .

As we know, we cannot always bound |C| usefully for non-regular graphs. However, it turns out we can use a bound on  $\mu(C)$  to bound the ratio  $|C \cap [v]|/v$  away from one for some values of v and, when suitably formulated, such a bound is enough for our application to list colouring. In order to establish this bound, we shall need the online property, and so we prove the bound only for simple graphs (for which the online property holds however, see §12). An important point is that we cannot make use of iteration as in §3.4 to obtain smaller containers, because the online property does not survive iteration. (Indeed, consider the discussion in §3.4, and suppose C is determined by T and then C' by T'. For C' to have the online property we would need to be able to determine  $C' \cap [v]$  from  $T \cap [v]$ together with  $T' \cap [v]$ . But, to compute  $C' \cap [v]$ , it is necessary to know the whole of G[C], for which it is necessary to know the whole of C, and thus to know the whole of T rather than just  $T \cap [v]$ ; hence the online property fails.) Since we cannot use iteration, we have an interest in making the bound on  $\mu(C)$  in Theorem 3.4 as small as we can (see §3.6).

Let  $\mathcal{C}$  be a collection of containers for G. For each initial segment [v] of the vertex set [n], the set  $\mathcal{C}_v = \{C \cap [v] : C \in \mathcal{C}\}$  of restrictions to [v] is a collection of containers for the induced subgraph G[v]. For our application, it would be enough to find a segment [v] for which the collection  $\mathcal{C}_v$  of restrictions is well-behaved: that is,  $|C \cap [v]| \leq (1-c)v$  for each C and  $|C_v| \leq 2^{\tau v}$ . We could then work with the subgraph G[v] rather than with G. In the example of  $G = K_{d,n-d}$ , the first 2*d* vertices induce  $K_{d,d}$ , so we could take v = 2d and find good containers for G[v]. But G does not always have such a nice subgraph (see §4).

Something slightly different does work, though. Each container C = C(T) nominates a vertex v = g(C) so that both restrictions  $C \cap [v]$  and  $T \cap [v]$  are simultaneously constrained (Lemma 7.2). By the online property,  $T \cap [v]$  determines  $C \cap [v]$ . This limits the number of possible sets  $C \cap [v]$ , and it turns out to be enough for our application.

To state the precise theorem, we make a couple of technical changes to the outline just given. First, we work with tuples  $(C_1, \ldots, C_t)$  of containers rather than individual containers, since this is ultimately more efficient. Secondly, we include a lower bound k on g(C) to make sure v is not too small. In order to convert degree to uniform measure, we also ask that the vertices be ordered by decreasing degree.

**Theorem 3.7.** Let G be a simple r-graph on vertex set [n], for which the degree sequence is decreasing. Let  $0 < \zeta \leq 1/12r!$ . Suppose that  $\delta(G, \tau) \leq \zeta$ , that  $\tau \leq \zeta^2/r$ , and that  $k \in [n]$  satisfies  $\mu([k]) \leq \zeta/2r!$ . Let  $t \in \mathbb{N}$ .

Then there exists a collection  $C \subset \mathcal{P}[n]$  and a map  $g : C^t \to [k, n]$ , with the following properties:

- (a) for all independent sets I there is some  $C \in \mathcal{C}$  with  $I \subset C$ ,
- (b) for all  $v \in [n]$

 $\log |\{ (C_1 \cap [v], \dots, C_t \cap [v]) : g(C_1, \dots, C_t) = v \}| \le r \zeta^{-2} v t \tau \log(1/\tau),$ 

(c) and for all  $(C_1, \ldots, C_t) \in \mathcal{C}^t$ 

$$\frac{1}{t} \sum_{i=1}^{t} |C_i \cap [v]| \le (1 - \frac{1}{r!} + 8\zeta)v,$$

where  $v = g(C_1, ..., C_t)$ .

Moreover, (a) holds for all sets  $I \subset [n]$  for which either G[I] is  $\lfloor \tau^{r-1} \zeta e(G)/n \rfloor$ -degenerate or  $e(G[I]) \leq 2r\tau^r e(G)/\zeta$ .

The main features of this theorem are hopefully recognisable by now: (a) means each independent set is in a container, (b) means each tuple of containers nominates a vertex v, and the number of restricted containers for any nominated v is small in terms of v, and (c) means the uniform measure of the restricted containers is bounded away from one.

It is worth noting that, for regular graphs, Theorem 3.7 follows (up to the odd constant) from Corollary 3.6, because in that case, as remarked following inequality (2),  $|C| \leq (1 - 1/r + \epsilon/r)$  for every  $C \in C$ , and we can just take  $g(C_1, \ldots, C_t) = n$ .

Theorem 3.7 is proved in  $\S7$  and applied in the proof of Theorem 2.1.

3.6. **Optimality.** We conclude this section about the main theorem with some observations as to what extent it is best possible. There are two aspects to optimality: the bound on  $\mu(C)$  and the bound on  $|\mathcal{C}|$ , the latter being implied by the size of  $\tau$ .

It is easy to produce examples of G and I with  $\mu(I) = 1 - 1/r$ , and so the best bound on  $\mu(C)$  that one could hope for in general is  $\mu(C) \leq 1 - 1/r$ . The bound in Theorem 3.4 is, essentially,  $\mu(C) \leq 1 - 1/r!$ . In fact the algorithm that we use to prove Theorem 3.4 does not give containers smaller than this: there is a description in [56] and in [54] of an example of a graph G, an ordering of its vertices, and an independent set I such that the container C constructed for I satisfies  $\mu(C) \approx 1 - 1/r!$ . In this sense, Theorem 3.4 is best possible. The fact that the algorithm does achieve this bound is proved in §5.3 and, for reasons mentioned in §3.5, we put some effort into that proof, even though a shorter argument would give a useful but weaker bound.

We remark that the simple algorithm in [58] gives  $\mu(C) \leq 1 - 1/4r^2$  in one shot, but the number of containers produced is larger than here. (That method too applies only to simple hypergraphs, though this is not quite such a drawback as might at first appear.)

The more important aspect of optimality regarding Theorem 3.4 is the bound on  $|\mathcal{C}|$  implicit in the bound on  $\tau$ . Theorem 3.8 below states in what ways the bound is optimal, but before stating the theorem we explain informally what lies behind it.

A simple counting argument shows that if most sets of size t are independent, or nearly independent (having many fewer edges than vertices), then any collection C of containers must satisfy  $|C| = e^{\Omega(t)}$ , provided each container  $C \in C$  has size bounded away from n. In a graph of order n and average degree d, most sets of size  $o(nd^{-1})$  are nearly independent; for r-graphs the same is true of most sets of size  $o(nd^{-1/(r-1)})$ . This is already enough to show that the main theorem comes quite close to being best possible, given that (as remarked in §3.1)  $\tau$  is often of order  $d^{-1/(r-1)}$ .

We can say even more for certain kinds of containers. Let us call C internally generated if there is some function C(T) such that  $C = \{C(T) : T \in \mathcal{T}\}$ , where  $\mathcal{T} \subset \mathcal{P}(V(G))$ , and for every independent set I, there exists  $T \in \mathcal{T}$  with  $T \subset I \subset C(T)$ . In this case, we show that  $\mathcal{T}$  must contain sets T of size at least  $\Omega(t)$  and, moreover,  $|\mathcal{T}| = e^{\Omega(tn \log(1/t))}$ . Again, if  $\tau$  is of order  $d^{-1/(r-1)}$ , this indicates that the bounds in Corollary 3.6 are best possible to within a constant factor, and, certainly if C(T) is injective, the logarithmic factor in (c) cannot be removed, meaning (a) must be retained. (We do not assert, though, that the function C(T) given by the algorithm in this paper is injective.)

Finally, the constraint imposed on  $\tau$  by the co-degree function  $\delta(G, \tau)$  is, in a sense, also optimal. For this we assume a certain amount of symmetry in the *r*-graph *G*, say *G* is vertex and edge transitive: this condition certainly holds in several cases of interest, such as in §2.2 and §2.6 when G = G(N, H). Given an *r*-graph *G*, let  $G_j$  be the *j*-graph whose edges are the *j*-sets  $\sigma$  of maximum degree in *G*. By symmetry, every edge of *G* contains an edge of  $G_j$  and so, crucially, every independent set of  $G_j$  is independent in *G* too. Thus a collection *C* of containers for *G* furnishes a collection for  $G_j$ ; hence, by considering nearly independent sets in  $G_j$ , we can obtain lower bounds on  $|\mathcal{C}|$  in terms of the average degree of  $G_j$ . The maximum over all *j* of these bounds turns out to be exactly (up to constant factors) the size of *C* given by Corollary 3.6, determined by the constraint on  $\delta(G, \tau)$ . Hence the co-degree function is in some way capturing the right property of *G*. Readers familiar with the parameter m(H) defined in §2.2 will perhaps recognise the spirit of this argument and so sense why m(H) appears in Theorem 2.3.

**Theorem 3.8.** Let G be an r-graph of average degree d and vertex set [n]. Let  $C \subset \mathcal{P}[n]$  be such that, for every independent set I of G there is some  $C \in C$  with  $I \subset C$ . Suppose that  $|C| \leq (1-c)n$  for all  $C \in C$ . Then there is a positive constant  $\gamma = \gamma(r,c)$  such that the following hold.

- (i)  $|\mathcal{C}| > e^{\gamma nt}$ , where  $t = d^{-1/(r-1)}$ ;
- (ii) if C is internally generated (see above) and  $C = \{C(T) : T \in \mathcal{T}\}$ , then  $|T| \ge \gamma tn$ for some  $T \in \mathcal{T}$ , and  $|\mathcal{T}| \ge e^{\gamma nt \log(1/t)}$ , where  $t = d^{-1/(r-1)}$ ;
- (iii) if G is vertex and edge transitive, and  $\delta(G, \tau) \ge 1$ , then (i) and (ii) hold with  $t = \tau$ .

The proof of this theorem is given in §11.

## 4. The Algorithm

In this section we describe the method of building containers and establish the basic facts about them.

Even for 2-graphs, it is not immediate that a useful container theorem exists. The starting point for our method is the work of Sapozhenko [49, 50, 51, 52], who gave a way to build containers for regular 2-graphs. In §4.1 we describe our method for 2-graphs, which illustrates some of the essential features of the general method, though obviously not all.

This is only the starting point, of course, because a method for 2-graphs gives very little clue as to how to approach r-graphs. A method we found that works for simple regular r-graphs was described in [55] (refined in [58]), good enough to produce a good lower bound for  $\chi_l(G)$ . This would yield a similar bound for non-regular r-graphs if it were true that every such graph contains an almost regular subgraph. The requirement on the subgraph here can be made extremely weak but nevertheless it cannot be satisfied; there are examples of r-graphs where every subgraph is far from regular, somewhat along the lines of the 2graphs of Pyber, Rödl and Szemerédi [47]. A construction due to Verstraëte is described by Dellamonica and Rödl [14].

One of our requirements for a good construction is that it must satisfy the online property, which further limits the options. The method described here fulfils these needs (though see §12). It is in some ways almost opposite in approach to that in [55, 58]. We endeavour to motivate the method to the extent that we can. Nevertheless, the reason behind one or two features might become clearer after reading §5, in which the properties of the algorithm are proven.

4.1. Example for r = 2. To introduce some (but not all) of the ideas used in the algorithm, we prove Theorem 3.4 for independent sets in 2-graphs. As mentioned, for regular graphs, the strategy reduces to something close to, but not identical to, that of Sapozhenko [49, 50, 51, 52]. We have tried to make the notation as similar as possible to that used in the main algorithm, although the reader should be aware that there are some differences.

An important general feature of the method can be described immediately, which holds as well for r-graphs as for 2-graphs. The construction of a subset  $T_1$  from an independent set I, and the construction from  $T_1$  of a set which contains I, are achieved by the same algorithm, run in two slightly different modes that we call *prune mode* and *build mode*. In prune mode the algorithm receives I as input and it outputs  $T_1$ ; in build mode the algorithm receives  $T_1$  as input and it outputs a container.

Let the vertices of G be the set [n]. This just gives an ordering to the vertices: we assume no properties of the ordering. In prune mode, the algorithm begins with  $T_1 = \emptyset$ . It then examines the vertices one by one in the order  $1, \ldots, n$ ; when it reaches vertex v, it checks whether v is in I and whether v has some further property — we might say there is some "membership rule" that v must satisfy. If  $v \in I$  and v passes the membership rule, then vis added to  $T_1$ , otherwise v is not added to  $T_1$ . This is done for  $v = 1, \ldots, n$  in turn, and the algorithm then outputs  $T_1$ . In build mode, the algorithm initialises a set  $C_1 = [n]$ . It then examines each vertex v in turn. If v satisfies the membership rule, then v is removed from  $C_1$ . This is done for  $v = 1, \ldots, n$  in turn, and the algorithm then outputs  $C_1 \cup T_1$ . It will be seen that  $T_1 \subset I$ , by construction, and that  $I \subset C_1 \cup T_1$ , because the vertices left in  $C_1$  failed the membership rule, and this includes all the vertices of I except those in  $T_1$ . (Notice in passing that we did not need I to be independent.)

As stated, the build algorithm did not appear to make use of the input  $T_1$ . However, the algorithm will also construct an auxiliary structure (in the case r = 2, this is just a set) along the way. The membership rule is specified in terms of the current state of this auxiliary structure. The structure will be updated only when a vertex is in  $T_1$  and passes the rule, so both modes of the algorithm have the information to update the structure properly.

To be more specific, here is the membership rule we use for r = 2: the requirement is of course to produce small sets  $T_1$  and not too large sets  $C_1$ . Let G be a 2-graph with vertex set [n] and average degree d. Let  $\zeta > 0$ . The algorithm uses an auxiliary set  $\Gamma_1 \subset [n]$ , which is initially empty. For  $v = 1, \ldots, n$ , consider

$$F(v) = \{ w \in [v+1, n] : \{v, w\} \in E(G) \text{ and } w \notin \Gamma_1 \}.$$

We take the following as the membership rule: that  $|F(v)| \ge \zeta d(v)$ . In addition to checking the rule, if  $v \in T_1$  then the algorithm adds F(v) to  $\Gamma_1$ . Both modes of the algorithm know whether  $v \in T_1$ : prune mode because it is constructing  $T_1$ , and build mode because it is given  $T_1$ . Hence both modes construct the same  $\Gamma_1$  and both are using the same rule.

Note that, as the algorithm proceeds,  $\Gamma_1$  is just the set of vertices which have an earlier neighbour in  $T_1$ , because F(v) is precisely those vertices w for which v is an earlier neighbour but no other vertex yet in  $T_1$  is.

We now use, for the first time, the fact that I is independent. Because the vertices in  $\Gamma_1$  are neighbours of those in  $T_1 \subset I$ , we know  $I \cap \Gamma_1 = \emptyset$ . We also know that  $I \subset C_1 \cup T_1$ , the set output by build mode. So our final container is  $C(T_1) = (C_1 \cup T_1) - \Gamma_1$ . The notation  $C(T_1)$  means that  $C(T_1)$  can be constructed just from  $T_1$ .

The function  $C: \mathcal{P}[n] \to \mathcal{P}[n]$  satisfies the main points of Theorem 3.4 for r = 2. Recall from §3.1 that  $\delta(G, \tau) \leq 1/\tau d$  for a 2-graph, so take  $\tau = 1/\zeta d$ . The set  $T_0$  is not needed so take  $T_0 = \emptyset$ . We know  $I \subset C(T_1)$ , which is assertion (a) of the theorem. Now we check the measures of  $T_1$  and of  $C(T_1)$ .

Whenever a vertex v is added to  $T_1$ , the rule is satisfied, so  $\Gamma_1$  increases by at least  $\zeta d(v)$ . But  $|\Gamma_1| \leq n$ , so  $\zeta \mu(T_1) = (1/nd) \sum_{v \in T_1} \zeta d(v) \leq (1/nd) |\Gamma_1| \leq 1/d$ . Therefore  $\mu(T_1) \leq 1/\zeta d$ , which gives Theorem 3.4(b) comfortably.

Write  $C^* = C_1 - \Gamma_1$ . Every w, with  $vw \in E(C^*)$  for some v < w, lies in F(v) by definition of  $C_1$ . (Note here that  $\Gamma_1$  grows during the procedure so, at the time F(v)was defined,  $\Gamma_1$  might have been smaller than it is at the end, or in other words there might have been  $u \in F(v)$  when the rule was tested, such that u ends up in  $\Gamma_1$  and not in  $C^*$ . But this only helps.) Since  $v \in C_1$ , v failed the rule, so  $|F(v)| < \zeta d(v)$ . Therefore  $e(G[C^*]) \leq \sum_{v \in C^*} \zeta d(v) \leq \zeta n d$ . On the other hand,  $e(G[C^*]) \geq (\mu(C^*) - 1/2)n d$  by inequality (2). Thus  $\mu(C^*) \leq 1/2 + \zeta$ . Finally,  $C(T_1) = (C_1 \cup T_1) - \Gamma_1 \subset C^* \cup T_1$  and so  $\mu(C) \leq \mu(C^*) + \mu(T_1) \leq 1/2 + \zeta + 1/\zeta d$ , giving Theorem 3.4(d).

We have not completely proved Theorem 3.4 for r = 2, since we have not shown condition (c): indeed, at present the bound on the degree measure of  $T_1$  does not imply a bound on  $|T_1|$ , since this requires a lower bound on the degrees of the vertices of the graph. One way round this is to amend the membership rule so as to allow  $v \in T_1$  only if  $d(v) \ge \zeta d$ . It can be shown that this increases  $\mu(C(T_1))$  by only  $\zeta$ , and all vertices in  $T_1$  now have degree at least  $\zeta d$ , so  $|T_1|\zeta d \leq \sum_{v \in T_1} d(v) = \mu(T_1)nd$ . Condition (c) then follows from (b).

We did not check the online property, but that is not hard to do. Nor did we check the cases where I is not independent, but, as has been seen, the independence of I was barely used. In fact, the set  $T_0$  is there to take care of these cases.

4.2. The general algorithm. For general r-graphs we use the same method of an algorithm running in prune/build mode as used in 4.1, but we need a membership rule that will handle edges of size r > 2. Moreover, if the r-graph is not simple, we need to handle overlapping edges carefully.

Given an r-graph G with vertex set [n], we in fact run the algorithm r-1 times; we label these runs by s = r - 1, r - 2, ..., 1 in turn. In run s, the algorithm has input I and output  $T_s$  (in prune mode) or input  $T_s$  and output  $C_s$  (in build mode). Along the way it builds an auxiliary multigraph  $P_s$ . Here,  $P_s$  is s-uniform but multiple edges are allowed; in other words  $E(P_s)$  is a multiset. The multigraph  $P_s$  is constructed from  $P_{s+1}$ , which is supplied by the previous run of the algorithm. For the first run, when s = r - 1, we supply  $P_r = G$ .

Each edge  $\{u_{s-1}, u_{s-2}, \ldots, u_0\} \in E(P_s)$  with  $u_{s-1} < u_{s-2} < \cdots < u_0$  will come from an edge  $\{v_{r-1}, v_{r-2}, \cdots, v_s, u_{s-1}, u_{s-2}, \ldots, u_0\} \in E(G)$ , where  $v_{r-1} < \cdots < v_s < u_{s-1}$  and  $v_j \in T_j, r-1 \ge j \ge s$ . Equivalently, each edge of  $P_s$  is an edge of  $P_{s+1}$  whose first vertex, which is in  $T_s$ , has been removed. The reason  $P_s$  is defined as a multigraph, even if G itself does not have multiple edges, is so that distinct edges of G give rise to distinct edges of  $P_s$ . As will be seen, this allows more vertices to be added to  $T_s$ , which in turn conveys more information about the independent set.

The multigraph  $P_1$  is 1-uniform: its edges are sets containing single vertices. If  $\{u_0\} \in E(P_1)$  then there is an edge  $\{v_{r-1}, v_{r-2}, \cdots, v_1, u_0\} \in E(G)$  with  $v_j \in T_j, r-1 \ge j \ge 1$ . So, if I is an independent set and the sets  $T_j$  are chosen within I, as they will be, then  $u_0 \notin I$ , and so the container C can be chosen from vertices not in  $E(P_1)$ . Note here that  $E(P_1)$  is playing the role that  $\Gamma_1$  did in §4.1. Our first aim, then, is to ensure that  $E(P_1)$  is as large as possible, and to this end we attempt to make  $E(P_s)$  large for each s. However this aim has to be balanced against keeping the sets  $T_s$  small.

Hence we shall choose a parameter  $\tau$ , so that, roughly speaking,  $T_s$  will comprise a proportion  $\tau$  of the vertices (in degree measure), and we aim to design the algorithm so that, ideally, the size of  $E(P_s)$  will be roughly  $\tau$  times the size of  $E(P_{s+1})$ . This means the average degree of  $P_s$  will typically be around  $\tau^{r-s}d$ . The parameter  $\tau$  is the same as that discussed in §3.1 and the constraint  $\tau \geq d^{-1/(r-1)}$  described there is precisely what is needed to ensure that  $E(P_1)$  contains something worthwhile.

However not every edge of G with its first r-s vertices in  $T_{r-1}, \ldots, T_s$  will be admitted as an edge of  $P_s$ , but only a selection of these. We do not allow edges into  $P_s$  if they increase the degree of some vertex, or the degree of some subset  $\sigma \subset [n]$ , beyond some agreed threshold. We define the degree of  $\sigma$  in the multigraph  $P_s$  to be

$$d_s(\sigma) = |\{e \in E(P_s) : \sigma \subset e\}|,$$

where we are counting edges with multiplicity in the multiset  $E(P_s)$ . (Naturally we may write  $d_s(v)$  instead of  $d_s(\{v\})$  if  $v \in [n]$ .) There are several reasons for wanting to bound the degrees in  $P_s$ . One reason is the hope of keeping the vertex degrees near to  $\tau^{r-s}$  times the degrees in G, so that degree measure in  $P_s$  relates to measure in G; in particular, small sets of vertices cannot account for most of the edges of  $P_s$  unless those sets have large measure in G. A second reason for controlling degrees of subsets is that only by doing so can we restrain the degrees of vertices at later stages: this comes out in the proof of Lemma 5.2.

So we proceed in the following way. We begin with  $P_r = G$ , and then apply the algorithm below to construct  $P_s$  from  $P_{s+1}$  using  $T_s$ , with s taking the values  $r - 1, r - 2, \ldots, 1$  in turn. During the application of the algorithm, the degrees  $d_s(\sigma)$  in  $P_s$  will grow, as edges are added. We denote by  $\Gamma_s$  the collection of vertices and subsets whose degrees have reached their bound, and we do not permit the addition to  $P_s$  of any edge which contains a current member of  $\Gamma_s$ . The set  $\Gamma_s$  will grow too during the construction. We remark that, as defined in §4.1,  $\Gamma_1$  was the the set of vertices in  $P_1$  that have positive degree: in the general algorithm we have the option of specifying a larger threshold for entry into  $\Gamma_1$ .

Two real numbers are included in the input to the algorithm. The parameter  $\tau$  is the more important and has already been discussed. The parameter  $\zeta$  is a small constant, used in the rule to decide membership of  $T_s$ .

As in §4.1, the membership rule involves a collection F of edges in  $P_s$  with first vertex v, the rule being:  $d(v) \ge \zeta d$  and  $|F| \ge \zeta \tau^{r-s-1} d(v)$ . Further, if  $v \in T_s$ , then F is added to  $E(P_s)$ , and the set  $\Gamma_s$  is updated appropriately. In the general algorithm F is a multiset rather than a set, to maintain the condition that edges of  $P_s$  correspond to different edges of G.

Again as in §4.1, the independence of the set I is not actually used by the algorithm, and it is useful to define the algorithm for general subsets  $I \subset [n]$ .

Algorithm

INPUT an r-graph G on vertex set [n]an (s + 1)-multigraph  $P_{s+1}$  on vertex set [n]parameters  $\tau, \zeta > 0$ in prune mode a subset  $I \subset [n]$ in build mode a subset  $T_s \subset [n]$ 

OUTPUT an s-multigraph  $P_s$  on vertex set [n]in prune mode a subset  $T_s \subset [n]$ in build mode a subset  $C_s \subset [n]$ 

put  $E(P_s) = \emptyset$  and  $\Gamma_s = \emptyset$ in prune mode put  $T_s = \emptyset$ in build mode put  $C_s = [n]$ 

for v = 1, 2, ..., n do:

let  $F = \{f \in [v+1, n]^{(s)} : \{v\} \cup f \in E(P_{s+1}), \text{ and } \forall \sigma \in \Gamma_s \ \sigma \not\subset f \}$ [here F is a multiset with multiplicities inherited from  $E(P_{s+1})$ ] in prune mode if  $d(v) \ge \zeta d$  and  $|F| \ge \zeta \tau^{r-s-1} d(v)$  and  $v \in I$ , add v to  $T_s$ in build mode if  $d(v) \ge \zeta d$  and  $|F| \ge \zeta \tau^{r-s-1} d(v)$ , remove v from  $C_s$ if  $v \in T_s$  then add F to  $E(P_s)$ for each  $u \in [v+1,n]$ , if  $d_s(u) > \tau^{r-s} d(u)$ , add  $\{u\}$  to  $\Gamma_s$ for each  $\sigma \in [v+1,n]^{(>1)}$ , if  $d_s(\sigma) > 2^s \tau d_{s+1}(\sigma)$ , add  $\sigma$  to  $\Gamma_s$  The algorithm therefore adds to  $P_s$  s-edges which, with  $v \in T_s$  as first vertex, form an edge of  $P_{s+1}$  and which do not contain (at that moment) any subset in  $\Gamma_s$ . The degree threshold for a vertex entering  $\Gamma_s$  is in terms of its degree d(u) in the original graph G, whereas for a larger subset  $\sigma$  it is in terms of its degree in  $P_{s+1}$ ; this difference is for technical reasons arising in the proof of Lemma 5.5. See §12 for further comment.

The basic feature of prune/build modes using a membership rule is now invoked: if  $T_s$  is constructed from I by running the algorithm in prune mode, and then  $C_s$  is constructed from  $T_s$  by re-running the algorithm in build mode, then  $I \subset C_s \cup T_s$  holds. Another option for a container arises if I is independent. We noted earlier that, in this case, if  $\{u_0\} \in E(P_1)$  then  $u_0 \notin I$ . In particular, if I is independent and  $\{u_0\} \in \Gamma_1$  then  $u_0 \notin I$ , because  $u_0$  is a vertex whose degree in  $P_1$  has risen above some (non-negative) threshold so  $\{u_0\} \in E(P_1)$ . Abusing notation slightly, we write  $I \subset [n] - \Gamma_1$ . Therefore each of  $C_s \cup T_s$ ,  $1 \leq s \leq r - 1$ , and  $[n] - \Gamma_1$  is a container for I; our aim is to ensure that at least one of these is a good container, meaning that its size is not close to [n].

Here then is a way of viewing the operation of the algorithm. If  $\Gamma_1$  is large then  $[n] - \Gamma_1$ is a good container for I. If  $\Gamma_1$  is not large then, since the degrees in  $P_1$  are bounded, the average degree of  $P_1$  must be small. But  $P_r = G$ , whose average degree is not small, so there must be some s for which  $P_{s+1}$  has large average degree (of order  $\tau^{r-s-1}d$ ) but  $P_s$ has small average degree (much smaller than  $\tau^{r-s}d$ ). Since the degrees are bounded, there must have been plenty of vertices of  $P_{s+1}$  which could have contributed edges to  $E(P_s)$ but did not do so. Why did they not do so? Only because they are not in I and so not available for  $T_s$ . These are exactly the vertices which are removed from  $C_s$ : hence for this value of s,  $C_s \cup T_s$  will be a good container for I.

We add an observation here about simple graphs G, for use when considering the online property (Lemma 4.4). The set  $\Gamma_s$  is non-uniform in order to handle sets  $\sigma$  for which  $d_s(\sigma)$ has become too large. If G is simple it is unnecessary to cater for such a possibility. If  $|\sigma| \geq 2$  then  $\sigma$  can appear in at most one edge of G, so  $d_s(\sigma)$  will be zero or one. Formally, for the way the algorithm is stated,  $\sigma$  will quite likely be inserted into  $\Gamma_s$  as soon as it appears in some edge of F (because  $2^s \tau^{r-s}$  is likely to be less than one). However this has no effect on the subsequent construction of  $P_s$  because  $\sigma$  will never again appear in F. Hence, for simple graphs, the appearance in the algorithm of  $\sigma$  with  $|\sigma| \geq 2$  can be ignored.

#### 4.3. Properties of the construction. We are thus led to two important definitions.

**Definition 4.1.** Let G be an r-graph on vertex set [n] and let  $I \subset [n]$ . Let  $\tau, \zeta > 0$ . Let  $T_{r-1}, \ldots, T_1$  be the sets constructed by repeated applications of the algorithm in prune mode. Let  $B = \{v \in [n] : d(v) < \zeta d\}$ . Let  $T_0 = I \cap (\Gamma_1 \setminus B)$ . Then we define

$$T(G, I, \tau, \zeta) = (T_{r-1}, \dots, T_1, T_0) \in \mathcal{P}(I)^r.$$

The *r*-tuple *T* is the fruit of running the algorithm in prune mode, from which the container for *I* will be built. As noted earlier, if *I* is an independent set then  $I \cap \Gamma_1 = \emptyset$ , so  $[n] - \Gamma_1$  is a container for *I*. We shall in fact use the slightly larger container  $[n] - (\Gamma_1 \setminus B)$ , but the difference is negligible because  $\mu(B) < \zeta$ . The introduction of *B* ensures that, just as the vertices in  $T_{r-1}, \ldots, T_1$  have degree at least  $\zeta d$ , so do those in  $T_0$ . This will be needed to prove Theorem 3.4(c).

Now comes the main definition — that of containers.

**Definition 4.2.** Let G be an r-graph on vertex set [n] and let  $T = (T_{r-1}, \ldots, T_1, T_0) \in \mathcal{P}([n])^r$ . Let  $\tau, \zeta > 0$ . Let  $C_{r-1}, \ldots, C_1$  be constructed by repeated applications of the algorithm in build mode, using  $T_{r-1}, \ldots, T_1$ . Let  $B = \{v \in [n] : d(v) < \zeta d\}$ . Let  $C_0 = [n] - (\Gamma_1 \setminus B)$ . The container  $C(G, T, \tau, \zeta)$  is then

$$C(G,T,\tau,\zeta) = (C_{r-1} \cap C_{r-2} \cap \cdots \cap C_1 \cap C_0) \cup T_{r-1} \cup T_{r-2} \cdots \cup T_1 \cup T_0.$$

**Lemma 4.3.** If  $T = T(G, I, \tau, \zeta)$  then  $I \subset C(G, T, \tau, \zeta)$ .

*Proof.* We noted earlier that  $I \subset C_s \cup T_s$  for s > 0. Moreover,  $I \subset C_0 \cup T_0$  by definition, since  $C_0 = [n] - (\Gamma_1 \setminus B)$  and  $T_0 = I \cap (\Gamma_1 \setminus B)$ . Hence  $I \subset C(G, T, \tau, \zeta)$ .

Before computing the size of the containers  $C(G, T, \tau, \zeta)$  and the number of them, we check the online property, namely that  $C(G, T, \tau, \zeta) \cap [w]$  is determined just by  $T \cap [w]$ . Recall that we are asserting the online property only for simple graphs.

**Lemma 4.4.** Let G be a simple r-graph on vertex set [n] and let  $T \in \mathcal{P}([n])^r$ . Then, for each  $w \in [n]$ ,  $C(G, T, \tau, \zeta) \cap [w] = C(G, T \cap [w], \tau, \zeta) \cap [w]$  holds.

*Proof.* The tuple T supplies sets  $T_{r-1}, \ldots, T_1$  as inputs for the algorithm in build mode, which produces sets  $C_{r-1}, C_{r-2}, \ldots, C_1$  and  $\Gamma_1$ . The set  $T_0$  is supplied directly by T, and we take  $C_0 = [n] - (\Gamma_1 \setminus B)$ . Let  $T' = T \cap [w]$ , where  $T' = (T'_{r-1}, \ldots, T'_1, T'_0)$  and  $T'_s = T_s \cap [w]$ ,  $0 \le s \le r-1$ . Let  $C'_{r-1}, C'_{r-2}, \ldots, C'_1, C'_0$  and  $\Gamma'_1$  be the corresponding sets produced when the inputs to the algorithm are  $T'_{r-1}, \ldots, T'_1, T'_0$ . We need to show that  $C'_s \cap [w] = C_s \cap [w]$ for all s; Definition 4.2 then shows  $C(G, T, \tau, \zeta) \cap [w] = C(G, T \cap [w], \tau, \zeta) \cap [w]$ .

Let  $P_r, \ldots, P_1$  be the multigraphs used during the runs of the algorithm with the original inputs, and  $P'_r, \ldots, P'_1$  those used with the truncated inputs. The crucial point is that, though  $P_s$  and  $P'_s$  might have different edges inside the vertex set [w + 1, n], they are otherwise identical; that is, if e is an s-edge with  $e \cap [w] \neq \emptyset$ , then  $e \in E(P_s)$  if and only if  $e \in E(P'_s)$  (with the same multiplicity). This can be seen for  $s = r, r - 1, \ldots, 1$  in turn, given that  $P_r = P'_r = G$ . Consider the run of the algorithm building  $P_s$ , as v runs through  $v = 1, \ldots, w$ . As mentioned at the end of §4.2, subsets  $\sigma$  with  $|\sigma| \ge 2$  have no effect because G is simple. By induction on v, the singletons  $\{u\}$  in  $\Gamma_s$  are the same as those in  $\Gamma'_s$  while  $v \in [w]$ , and the set F is the same for  $P_s$  as for  $P'_s$ , because F is defined by edges in  $P_{s+1}$ whose first vertex lies in [w]. Hence the membership rule, the set of singletons  $\{u\}$  added to  $\Gamma_s$ , and the set of edges F added to  $E(P_s)$ , are the same in  $P'_s$  and  $P_s$ . In particular,  $C'_s \cap [w] = C_s \cap [w]$  for  $s \ge 1$ . For the same reasons,  $\Gamma'_1 \cap [w] = \Gamma_1 \cap [w]$ . The set B is defined in terms of G itself, and so  $C'_0 \cap [w] = [w] - (\Gamma'_1 \setminus B) = [w] - (\Gamma_1 \setminus B) = C_0 \cap [w]$ .

#### 5. Container calculations

In this section we estimate the measure of the tuples  $T(G, I, \tau, \zeta)$  and of the containers  $C(G, T, \tau, \zeta)$ , thereby proving Theorem 3.4.

5.1. Degrees and co-degrees. Before making these estimates we need information on how large the degrees can be in  $P_s$ . The intention behind the set  $\Gamma_s$  is to prevent degrees being much larger than the target degrees, namely  $\tau^{r-s}d(u)$  for the vertex u; after the degree of u attains this level, no further edges containing u are added to  $P_s$ . However, when a vertex u enters  $\Gamma_s$ , it does so because some multiset F has been added to  $E(P_s)$ . Since F can include many edges that contain u, the degree  $d_s(u)$  can increase significantly in one step, from an initial value at most the target value  $\tau^{r-s}d(u)$  to something much larger. The extent of this problem depends ultimately on the way the edges of G overlap each other. The reason  $\Gamma_s$  is defined the way it is in the algorithm, is to keep control of the degree problem without increasing  $\tau$  more than is necessary. This definition lies at the heart of the efficiency of the algorithm. Control of the degrees can be expressed succinctly in terms of the co-degree function  $\delta(G, \tau)$ .

First we need a small calculation.

**Lemma 5.1.** For  $2 \leq s \leq r$  and  $2 \leq j \leq s$ , let  $a_s^{(j)}$  be given by the equations  $a_r^{(j)} = \delta_j$ and  $a_s^{(j)} = 2^s a_{s+1}^{(j)} + a_{s+1}^{(j+1)}$  for s < r, where  $\delta_j$  was defined in Definition 3.2. Then  $a_s^{(2)} \leq 4^{2-s}\delta(G,\tau)$  holds for  $s \geq 2$ .

Proof. Since  $a_s^{(2)} \ge 2^s a_{s+1}^{(2)} \ge 4a_{s+1}^{(2)}$ , it is enough to prove that  $a_2^{(2)} \le \delta(G, \tau)$ . Now by dint of the definition it is clear that  $a_s^{(j)}$  is a linear combination of the numbers  $\delta_{j+\ell}$ ,  $\ell \ge 0$ . We claim that the coefficient of  $\delta_{j+\ell}$  in  $a_s^{(j)}$  is at most  $2^{\binom{r}{2} - \binom{s+\ell}{2} + \ell}$ . This is certainly true if s = r, since the only positive coefficient is that of  $\delta_j$  (i.e.  $\ell = 0$ ). For s < r we may prove the claim on the assumption that it is true for s + 1. If  $\ell = 0$  then the coefficient of  $\delta_{j+\ell}$  in  $a_{s+1}^{(j+1)}$  is zero, and the claim follows because  $2^{\binom{r}{2} - \binom{s}{2}} = 2^s 2^{\binom{r}{2} - \binom{s+1}{2}}$ . If  $\ell \ge 1$  we have

$$2^{s} 2^{\binom{r}{2} - \binom{s+1+\ell}{2} + \ell} + 2^{\binom{r}{2} - \binom{s+\ell}{2} + \ell - 1} = 2^{\binom{r}{2} - \binom{s+\ell}{2} + \ell} \left[ 2^{-\ell} + 2^{-1} \right] \le 2^{\binom{r}{2} - \binom{s+\ell}{2} + \ell}$$

and the claim follows in this case too. Hence the claim always holds, and so

$$a_{2}^{(2)} \leq 2^{\binom{r}{2}} \sum_{\ell=0}^{r-2} 2^{-\binom{\ell+2}{2}+\ell} \delta_{2+\ell} = 2^{\binom{r}{2}-1} \sum_{j=2}^{r} 2^{-\binom{j-1}{2}} \delta_{j} = \delta(G,\tau),$$

by definition of  $\delta(G, \tau)$ .

Here is the main lemma about degrees in  $P_s$ , and it shows the role of the co-degree function  $\delta(G, \tau)$  in the analysis of the algorithm. As explained in §4, we would ideally like  $\sum_{u \in U} d_s(u) \approx \tau^{r-s} \mu(U) nd$  for each subset  $U \subset [n]$ . The lemma shows that this holds as an upper bound, with a small error expressed in terms of  $\delta(G, \tau)$ .

**Lemma 5.2.** Let G be an r-graph on vertex set [n] with average degree d. Let  $P_r = G$  and let  $P_{r-1}, \ldots, P_1$  be the multigraphs constructed by some run of the algorithm, either in build mode or in prune mode. Then

$$\sum_{u \in U} d_s(u) \le \left(\mu(U) + 4^{1-s}\delta(G,\tau)\right)\tau^{r-s} \, nd$$

holds for all subsets  $U \subset [n]$  and for  $1 \leq s \leq r$ .

Proof. Recall that, as the algorithm proceeds, an element enters the set  $\Gamma_s$  when its degree exceeds some threshold: for a vertex, when  $d_s(u) > \tau^{r-s}d(u)$ , and for a larger set when  $d_s(\sigma) > 2^s \tau d_{s+1}(\sigma)$ . Let  $u \in U$ . If  $u \notin \Gamma_s$  then  $d_s(u) \leq \tau^{r-s}d(u)$ . If  $u \in \Gamma_s$  then u was added to  $\Gamma_s$  after some other vertex v was inspected and some multiset F was added to  $E(P_s)$ , raising  $d_s(u)$  beyond  $\tau^{r-s}d(u)$ . After u was added to  $\Gamma_s$ ,  $d_s(u)$  did not change. Hence  $\sum_{u \in U} d_s(u) \leq \tau^{r-s} \sum_{u \in U} d(u)$  plus the extra contribution from the multisets F. It is these extra contributions that we must now examine and bound in terms of  $\delta(G, \tau)$ . To do this, we must consider all the elements  $\sigma \in \Gamma_s$ , not just the vertices. Each of these enters  $\Gamma_s$  when its degree exceeds its threshold by a little extra. These extras percolate down to form the extra for the vertex u, in a way that Lemma 5.1 is designed to capture.

Let us do the calculation. By analogy with Definition 3.2 we define

$$d_s^{(j)}(u) = \max \{ d_s(\sigma) : u \in \sigma \in [n]^{(j)} \}$$

for  $j \ge 2$ , where here it is the final values of these quantities that are used — that is, we measure these quantities in the output multigraph  $P_s$ .

When s = r the lemma is true by definition of  $\mu(U)$ , so from now on we assume  $s \leq r-1$ . Let  $u \in [n]$ ; then  $d_s^{(j)}(u) = d_s(\sigma)$  for some  $\sigma \in [n]^{(j)}$  with  $u \in \sigma$ . If  $\sigma \notin \Gamma_s$  then  $d_s(\sigma) \leq 2^s \tau d_{s+1}(\sigma)$ . If  $\sigma \in \Gamma_s$  then  $\sigma$  was added to  $\Gamma_s$  after some vertex  $v \notin \sigma$  was inspected and F was added to  $E(P_s)$ . Before this took place,  $d_s(\sigma) \leq 2^s \tau d_{s+1}(\sigma)$  held; since the number of edges of F containing  $\sigma$  was at most  $d_{s+1}(\sigma \cup \{v\})$ , we have, in both cases,

$$d_s^{(j)}(u) = d_s(\sigma) \le 2^s \tau d_{s+1}(\sigma) + d_{s+1}(\sigma \cup \{v\}) \le 2^s \tau d_{s+1}^{(j)}(u) + d_{s+1}^{(j+1)}(u).$$
(3)

We claim that

$$\sum_{u \in [n]} d_s^{(j)}(u) \le a_s^{(j)} \tau^{r-s+j-1} n d_s$$

where  $a_s^{(j)}$  was defined in Lemma 5.1. Indeed, for s = r the claim (with equality) is just the definition of  $\delta_j$ , and for  $s \leq r - 1$  it follows immediately by induction (on r - s) from inequality (3) and the definition of  $a_s^{(j)}$ . Hence, for  $s \geq 1$ , we have by Lemma 5.1

$$\sum_{u \in [n]} d_{s+1}^{(2)}(u) \le 4^{1-s} \tau^{r-s} nd \,\delta(G, \tau) \,. \tag{4}$$

Now let  $u \in U$ . As mentioned before, either  $d_s(u) \leq \tau^{r-s} d(u)$  or u was added to  $\Gamma_s$  after some vertex v was inspected and F was added to  $E(P_s)$ . Since F has at most  $d_{s+1}(\{u,v\})$  edges containing u, the degree of u in  $P_s$  is at most  $\tau^{r-s} d(u) + d_{s+1}(\{u,v\})$ . Now  $d_{s+1}(\{u,v\}) \leq d_{s+1}^{(2)}(u)$  so, using (4), we have

$$\sum_{u \in U} d_s(u) \le \sum_{u \in U} \left( \tau^{r-s} d(u) + d_{s+1}^{(2)}(u) \right) \le \tau^{r-s} \mu(U) n d + 4^{1-s} \tau^{r-s} n d \, \delta(G, \tau) \,,$$

which establishes the lemma.

5.2. The measure of the sets  $T_s$ . We now estimate the measures of the sets  $T_s$ . Ideally, they would have degree measure at most  $\tau$ , or, more exactly,  $\tau/\zeta$ . In fact such a bound does hold with a small error determined by  $\delta(G, \tau)$ .

**Lemma 5.3.** Let  $I \subset [n]$  and  $T = T(G, I, \tau, \zeta) = (T_{r-1}, \ldots, T_1, T_0)$ . Then  $\mu(T_s) \leq (\tau/\zeta)(1 + \delta(G, \tau))$  for  $1 \leq s \leq r - 1$ .

*Proof.* The set  $T_s$  is output when the algorithm is run in prune mode. During the run of the algorithm, each vertex v which enters  $T_s$  contributes a set F of at least  $\zeta \tau^{r-s-1} d(v)$  edges to  $E(P_s)$ . But the total size of  $E(P_s)$  is limited, because the degrees in  $P_s$  are constrained. Writing d for the average degree of G, Lemma 5.2 yields

$$\zeta \tau^{r-s-1} \mu(T_s) nd = \sum_{v \in T_s} \zeta \tau^{r-s-1} d(v) \le e(P_s) \le \sum_{u \in [n]} d_s(u)$$
$$\le \tau^{r-s} nd(1 + 4^{1-s} \delta(G, \tau))$$

and this proves the lemma.

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The set  $T_0$  needs a different argument. As noted before,  $T_0 = \emptyset$  if I is independent.

**Lemma 5.4.** Let G be an r-graph on vertex set [n] with average degree d. Let  $I \subset [n]$  and  $T = T(G, I, \tau, \zeta) = (T_{r-1}, \ldots, T_1, T_0)$ . If  $e(G[I]) \leq bn$  where  $b \leq 2\tau^r d/\zeta$ , then  $\mu(T_0) \leq 2\tau/\zeta$ . If G[I] is b-degenerate where  $b \leq \zeta \tau^{r-1} d/r$ , then  $\mu(T_0) \leq (\tau/\zeta)(1 + \delta(G, \tau))$ .

Proof. Recall that  $T_0 = I \cap (\Gamma_1 \setminus B)$ . So, for each  $v \in T_0$ ,  $d_1(v) > \tau^{r-1}d(v)$  holds because  $v \in \Gamma_1$ . Here the degree  $d_1(v)$  is in the multigraph  $P_1$ . Let  $J = T_{r-1} \cup \cdots \cup T_1 \cup T_0 \subset I$ . Recall that distinct 1-edges  $\{v\}$  in the multigraph  $P_1$  correspond to distinct r-edges  $\{v_{r-1}, \ldots, v_1, v\} \subset I$  with  $v_{r-1} < \cdots < v_1 < v$  and  $v_s \in T_s$ , so these edges lie in G[J]. It follows that  $\tau^{r-1}\mu(T_0)nd \leq \sum_{v \in T_0} d_1(v) \leq e(G[J])$ .

If  $e(G[I]) \leq bn$  then  $\tau^{r-1}\mu(T_0)nd \leq e(G[J]) \leq e(G[I]) \leq bn \leq 2\tau^r dn/\zeta$ , so  $\mu(T_0) \leq 2\tau/\zeta$ . If G[I] is b-degenerate, then  $e(G[J]) \leq b|J|$ . Recall from the construction of the sets  $T_s$ ,  $s \geq 1$ , in prune mode that  $d(v) \geq \zeta d$  for  $v \in T_s$ . For  $v \in T_0$  we have  $d(v) \geq \zeta d$  because  $v \notin B$ . Thus  $d(v) \geq \zeta d$  for all  $v \in J$ . So

$$\tau^{r-1}\mu(T_0)nd \le e(G[J]) \le b|J| \le \frac{b}{\zeta d} \sum_{v \in J} d(v) = \frac{b}{\zeta d}\mu(J)nd \le \frac{\tau^{r-1}}{r}\mu(J)nd$$

Therefore  $r\mu(T_0) \le \mu(J) \le \mu(T_{r-1}) + \dots + \mu(T_0)$ , and so  $(r-1)\mu(T_0) \le \mu(T_{r-1}) + \dots + \mu(T_1) \le (r-1)(\tau/\zeta)(1 + \delta(G, \tau))$  by Lemma 5.3.

5.3. The measure of the container  $C(G, T, \tau, \zeta)$ . We now prove the crucial fact that the measure of the container  $C(G, T, \tau, \zeta)$  is bounded above by some constant less than one. This can be established with a fairly simple argument, but just a little more care yields a bound close to 1 - 1/r!, which is best possible, as described in §3.6.

It is in order to achieve this bound that the number  $2^s$  appears in the algorithm, in the condition  $d_s(\sigma) > 2^s \tau d_{s+1}(\sigma)$  for entry of  $\sigma$  into  $\Gamma_s$ . Hence it is that powers of 2 appear in the definition of  $\delta(G, \tau)$ , having permeated there via Lemma 5.2. The condition can be relaxed to  $d_s(\sigma) > k \tau d_{s+1}(\sigma)$  for some smaller value of k, with some slight reduction in the constants in the definition of  $\delta(G, \tau)$ , but at the expense of a weaker bound on  $\mu(C(G, T, \tau, \zeta))$ . See §12 for further comment.

**Lemma 5.5.** Let  $T = (T_{r-1}, \ldots, T_0) \in \mathcal{P}([n])^r$ . Then

$$\mu(C(G,T,\tau,\zeta)) \leq 1 - \frac{1}{r!} + \frac{15}{4}\zeta + \frac{1}{4}\delta(G,\tau) + \sum_{s=0}^{r-1}\mu(T_s).$$

Proof. Recall from Definition 4.2 that  $C_0 = [n] - (\Gamma_1 \setminus B)$ , where  $B = \{v \in [n] : d(v) < \zeta d\}$ and d is the average degree of G. Recall too that  $C_{r-1}, \ldots, C_1$  are constructed by the algorithm in build mode. Let  $C = C_{r-1} \cap \cdots \cap C_0$ . We define  $D_1 = [n] - (C \setminus B)$ . Now  $\mu(C) \leq 1 - \mu(D_1) + \mu(B)$  and  $\mu(B) < \zeta$ . By Definition 4.2,  $C(G, T, \tau, \zeta) = C \cup T_{r-1} \cup \cdots \cup T_0$ . So to prove the lemma it is enough to prove that  $\mu(D_1) \geq 1/r! - 11\zeta/4 - \delta(G, \tau)/4$ .

We first outline the argument, before filling in the details. We have  $C_0 = [n] - (\Gamma_1 \setminus B)$ , so  $(C_0 \setminus B) \cap \Gamma_1 = \emptyset$ . Since  $C \subset C_0$ , this means  $(C \setminus B) \cap \Gamma_1 = \emptyset$ , so  $\Gamma_1 \subset [n] - (C \setminus B) = D_1$ . We extend  $D_1$  to a partition  $D_1, \ldots, D_r$  of [n] as follows:

$$D_1 = [n] \setminus (C \setminus B) \quad \text{where } \Gamma_1 \subset D_1 \text{ and } B \subset D_1$$
  

$$D_2 = \{v \in [n] : \{v\} \in \Gamma_2, v \notin D_1\}$$
  

$$D_3 = \{v \in [n] : \{v\} \in \Gamma_3, v \notin (D_1 \cup D_2)\}$$
  

$$\vdots$$
  

$$D_{r-1} = \{v \in [n] : \{v\} \in \Gamma_{r-1}, v \notin (D_1 \cup \cdots \cup D_{r-2})\}$$
  

$$D_r = [n] \setminus (D_1 \cup \cdots \cup D_{r-1}).$$

We aim to bound  $\mu(D_s)$  above, for  $s \ge 2$ , in terms of  $\mu(D_{s-1}), \ldots, \mu(D_1)$ . By induction, this means  $\mu(D_s)$  is bounded above in terms of  $\mu(D_1)$ , and, since  $\mu([n]) = 1$  and  $D_1, \ldots, D_r$  partition [n], we obtain a lower bound on  $\mu(D_1)$  as desired. It is convenient to define  $D_{\le s} = D_1 \cup \cdots \cup D_{s-1}, D_{\le s} = D_s \cup D_{\le s}$  and so on. Notice that  $C \setminus B = D_{\ge 2}$ .

To find the desired upper bound for  $\mu(D_s)$  in terms of  $\mu(D_{< s})$ , we look at edges of  $P_s$ . The point of the definition of  $D_s$  is that if  $v \in D_s$  then  $\{v\} \in \Gamma_s$  so  $d_s(v) \ge \tau^{r-s} d(v)$ ; thus the number edges of  $P_s$  meeting  $D_s$  can be bounded below in terms of  $\mu(D_s)$ . As we shall explain, we expect very few edges of  $P_s$  to lie inside  $D_{\geq s}$ , so nearly all edges meeting  $D_s$ meet  $D_{<s}$  too, and since the number of edges meeting  $D_{<s}$  is bounded above in terms of  $\mu(D_{< s})$  by Lemma 5.2, we are done. So the fundamental point of the proof is that  $D_{\geq s}$ should contain few edges of  $P_s$ .

Define the following trio of subsets of the edges of  $P_s$ , for each  $s \ge 2$ :

$$X_s = \{ f \in E(P_s) : |f \cap D_{\langle s}| \ge 1, |f \cap D_{\rangle s}| \ge 2 \}$$
  

$$Y_s = \{ f \in E(P_s) : f \subset D_{\geq s} \}$$
  

$$Z_s = \{ f \in Y_s : \sigma \subset f \text{ for some } \sigma \in \Gamma_{s-1}, |\sigma| \ge 2 \}.$$

Here,  $Y_s$  is the set previously discussed of edges inside  $D_{\geq s}$ ; if  $Y_s$  is empty, or small, then the above sketch proof works. We come to  $X_s$  and  $Z_s$  shortly.

Suppose first that G is a simple r-graph. As noted in §4.2, the sets  $\sigma$  with  $|\sigma| \geq 2$  play no role, and can be deleted from the algorithm; in particular  $Z_s = \emptyset$ . Consider an edge in  $Y_s$ . It has a first vertex v, where  $v \in D_{\geq s}$ . Now  $D_{\geq 2} = C \setminus B \subset C = C_{r-1} \cap \cdots \cap C_0$ , so in particular  $v \in C_{s-1} \setminus B$ . By definition of  $C_{s-1} \setminus B$ , all but  $\zeta \tau^{r-s} d(v)$  edges of  $P_s$  with first vertex v meet  $\Gamma_{s-1}$  and so meet  $D_{\leq s}$ , by definition of the  $D_i$  (note this is true even if s = 2). This means there are at most  $\zeta \tau^{r-s} d(v)$  edges in  $Y_s$  with first vertex v, so  $Y_s$  is indeed relatively tiny, as desired.

If G is not simple, then the argument of the previous paragraph yields that  $Y_s \setminus Z_s$  is tiny, so we need worry only if  $Z_s$  is large. But this would mean there are many  $\sigma \in \Gamma_{s-1}$ with  $d_s(\sigma)$  large, which, by definition of  $\Gamma_{s-1}$ , implies  $d_{s-1}(\sigma)$  is large. This turns out to give rise to many edges in  $X_{s-1}$ . However, we shall see below in (5) that, unlike edges in  $Y_s$ , edges in  $X_s$  only *improve* the original estimate for  $\mu(D_s)$  in terms of  $\mu(D_{< s})$ . So we trade off a loss in  $\mu(D_s)$  caused by  $Y_s$ , that is, by  $Z_s$ , for a gain in  $\mu(D_{s-1})$  caused by  $X_{s-1}$ . The relative trade-off can be weighted in favour of  $X_{s-1}$  by the 2<sup>s</sup> term in the definition of  $\Gamma_s$  in the algorithm, and this is precisely the reason for its appearance.

Now we can start the proof. We count edges in  $P_s$  but take into account both  $X_s$  and  $Y_s$ . We can take  $d_s(v) \ge \tau^{r-s} d(v)$  for all  $v \in D_s$ : for s < r this is because  $\{v\} \in \Gamma_s$ , and for s = r it holds trivially. Note that  $X_s \cap Y_s = \emptyset$  and each member of  $E(P_s) \setminus Y_s$  meets  $D_{<s}$ . So, for  $s \ge 2$ ,

$$\begin{aligned} \tau^{r-s}\mu(D_s)nd &\leq \sum_{v \in D_s} d_s(v) \\ &= \sum_{f \in E(P_s)} |f \cap D_s| \\ &= \sum_{f \in E(P_s) \setminus (X_s \cup Y_s)} |f \cap D_s| + \sum_{f \in X_s} |f \cap D_s| + \sum_{f \in Y_s} |f \cap D_s| \\ &\leq (s-1)|E(P_s) \setminus (X_s \cup Y_s)| + (s-3)|X_s| + s|Y_s| \\ &= (s-1)|E(P_s) \setminus Y_s| - 2|X_s| + s|Y_s| \\ &\leq (s-1) \sum_{v \in D_{< s}} d_s(v) - 2|X_s| + s|Y_s| \\ &\leq (s-1)\tau^{r-s} nd \left(\mu(D_{< s}) + 4^{1-s}\delta(G,\tau)\right) - 2|X_s| + s|Y_s|, \end{aligned}$$
(5)

where the last line employs Lemma 5.2. This is the bound on  $\mu(D_s)$  that we want.

For convenience, we further define the numbers  $x_s, y_s, z_s$  by  $|X_s| = x_s \tau^{r-s} nd$ ,  $|Y_s| = y_s \tau^{r-s} nd$  and  $|Z_s| = z_s \tau^{r-s} nd$ . Observe that  $X_2 = \emptyset$  because edges in  $X_s$  have at least three vertices, and  $Z_2 = \emptyset$  because  $\Gamma_1$  contains no  $\sigma$  with  $|\sigma| = 2$ . Thus we have the initial conditions  $x_2 = z_2 = 0$ .

Tidying up (5), we obtain  $\mu(D_s) \leq (s-1)(\mu(D_{\leq s}) + 4^{1-s}\delta(G,\tau)) - 2x_s + sy_s$ . Adding  $\mu(D_{\leq s}) = \mu(D_{\leq s-1})$  to each side gives

$$\mu(D_{\leq s}) \leq s\mu(D_{\leq s-1}) - 2x_s + sy_s + (s-1)4^{1-s}\delta(G,\tau)$$

for each  $s \ge 2$ . Multiplying this inequality by 1/s! and summing over  $s = 2, \ldots, r$ , noting that  $\mu(D_{\le r}) = 1$ ,  $D_{\le 1} = D_1$  and  $x_2 = 0$ , we obtain

$$\frac{1}{r!} \le \mu(D_1) - 2\sum_{s\ge 3} \frac{x_s}{s!} + \sum_{s\ge 2} \frac{y_s}{(s-1)!} + \frac{1}{4}\delta(G,\tau),$$
(6)

where we used  $\sum_{s \ge 2} 4^{1-s} (s-1)/s! < 1/4$ .

Let  $s \ge 2$  and let  $f \in Y_s \setminus Z_s$ . If f contains a subset  $\sigma \in \Gamma_{s-1}$  then  $|\sigma| = 1$ , say  $\sigma = \{u\}$ . But  $\{u\} \in \Gamma_{s-1}$  implies  $u \in D_{\le s}$  by definition of  $D_{s-1}$  (even if s = 2), which contradicts  $f \in Y_s$ . Thus f contains no member of  $\Gamma_{s-1}$ . Let v be the first vertex of f. Now  $v \in D_{\ge s} \subset D_{\ge 2} = C \setminus B$ , so  $v \in C_{s-1} \setminus B$ . By the construction of  $C_{s-1}$ , v is the first vertex of fewer than  $\zeta \tau^{r-s} d(v)$  edges of  $P_s$  that contain no member of  $\Gamma_{s-1}$ , so it is the first vertex of fewer than  $\zeta \tau^{r-s} d(v)$  edges in  $Y_s \setminus Z_s$ . Therefore  $|Y_s| - |Z_s| \le \sum_{v \in D_{\ge s}} \zeta \tau^{r-s} d(v) = \zeta \tau^{r-s} \mu(D_{\ge s}) nd$ . Hence  $y_s - z_s \le \zeta \mu(D_{\ge s}) \le \zeta$ . In particular  $y_2 \le \zeta$ , because  $z_2 = 0$ .

Let  $s \ge 3$  and put  $S = \{\sigma \in \Gamma_{s-1} : |\sigma| \ge 2, \sigma \subset D_{\ge s}\}$ . By definition of  $Z_s$ , each member of  $Z_s$  contains a member of S. Let F be the set of edges of  $P_{s-1}$  that contain a member of S. Then each edge in F contains at least two vertices of  $D_{\ge s}$ ; therefore  $F \subset X_{s-1} \cup Y_{s-1}$ .

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Hence

$$z_s \tau^{r-s} nd = |Z_s| \leq \sum_{\sigma \in S} d_s(\sigma)$$
  
$$\leq \sum_{\sigma \in S} \frac{1}{\tau 2^{s-1}} d_{s-1}(\sigma) \quad \text{by definition of } \Gamma_{s-1}$$
  
$$= \frac{1}{\tau 2^{s-1}} \sum_{f \in F} |\{\sigma \in S : \sigma \subset f\}|$$
  
$$\leq \frac{1}{\tau} |F| \quad \text{since } |f| = s - 1 \text{ for each } f \in F$$
  
$$\leq \frac{1}{\tau} |X_{s-1} \cup Y_{s-1}| = (x_{s-1} + y_{s-1}) \tau^{r-s} nd.$$

Thus  $z_s \leq x_{s-1} + y_{s-1}$  for  $s \geq 3$ . Since  $y_s \leq z_s + \zeta$  this means  $y_s \leq x_{s-1} + y_{s-1} + \zeta$ ; by repeating and applying both  $x_2 = 0$  and  $y_2 \leq \zeta$ , this yields  $y_s \leq x_{s-1} + x_{s-2} + \cdots + x_3 + (s-1)\zeta$  for  $s \geq 3$ . The inequality holds for s = 2 also. Substituting this inequality into inequality (6) we obtain

$$\frac{1}{r!} \le \mu(D_1) + \sum_{s \ge 3} x_s \left( -\frac{2}{s!} + \sum_{j=s}^{r-1} \frac{1}{j!} \right) + \zeta \sum_{s \ge 2} \frac{1}{(s-2)!} + \frac{1}{4} \delta(G,\tau).$$

The coefficient of  $x_s$  is negative, and so  $1/r! \le \mu(D_1) + 11\zeta/4 + \delta(G,\tau)/4$ , which is what we needed to prove.

5.4. **Proof of Theorem 3.4.** As expected, our choices for T and C(T) in Theorem 3.4 will usually be  $T = T(G, I, \tau, \zeta)$  and  $C(T) = C(G, T, \tau, \zeta)$ .

Proof of Theorem 3.4. Notice that the theorem is trivial if  $\zeta \geq 1/4r!$ , since in that case the function C(T) = [n] works, with  $T = (\emptyset, \dots, \emptyset)$  representing all *I*. Recall too from §3.1 that  $\delta(G, \tau) \to 0$  as  $\tau \to \infty$ . Hence the condition  $\delta(G, \tau) \leq \zeta$  is satisfiable by making  $\tau$ large enough, although if  $\tau \geq \zeta/2r$  the theorem is similarly trivial.

In the remaining cases we take  $T = T(G, I, \tau, \zeta)$  and  $C(T) = C(G, T, \tau, \zeta)$ . Then assertion (a) of the theorem holds because of Lemma 4.3, and the online property holds for simple graphs because of Lemma 4.4.

Let d be the average degree of G. By Lemmas 5.3 and 5.4, assertion (b) holds for sets I for which G[I] is  $\lfloor \tau^{r-1}\zeta e(G)/n \rfloor$ -degenerate, that is, b-degenerate with  $b \leq \zeta \tau^{r-1}d/r$ , using the fact that  $\delta(G,\tau) \leq \zeta \leq 1$ . Likewise, (b) holds for sets I for which  $e(G[I]) \leq 2r\tau^r e(G)/\zeta$ , that is,  $e(G[I]) \leq bn$  with  $b \leq 2\tau^r d/\zeta$ . Either of these implies (b) for independent sets I, by taking b = 0.

For every  $v \in T_s$  we have  $d(v) \ge \zeta d$ : for  $s \ge 1$  this holds by the definition of the algorithm, and for s = 0 it holds by the definition of  $T_0$ . Hence  $|T_s| \le (1/\zeta d) \sum_{v \in T_s} d(v) = (n/\zeta)\mu(T_s)$ . Thus (c) follows from (b).

Finally, property (d) follows from Lemma 5.5 and assertion (b), so we are done.  $\Box$ 

## 6. TIGHT CONTAINERS

We turn now to the first of our packaged versions of the container theorem, Corollary 3.6, which supplies containers with e(G[C]) small. The way to obtain sparser containers by repeated applications of the container theorem was discussed in §3.4, and here we calculate

what is achievable. Given an independent set I in the r-graph G, we apply the container theorem to obtain a container C with  $I \subset C$ . We then apply the container theorem again, this time to G[C], to obtain a sparser container C', then apply the theorem to G[C'], and so on, until the container is as sparse as we need, or the average degree in the container is so small that a further application of the container theorem yields no information.

The only point that needs consideration is how much effort we are willing, or able, to put into the calculation of the codegree function  $\delta(G[C], \tau)$  at each stage. This function determines how small  $\tau$  can be and hence how many (or few) containers are built. Evidently the degree  $d(\sigma)$  of some set  $\sigma \subset V(G)$  is no larger in G[C] than it is in G, so the simplest approach is just to use the original  $\delta(G, \tau)$  as an upper bound for  $\delta(G[C], \tau)$ . This works well for a limited number of iterations and it is the basis of the proof of Corollary 3.6.

However there are applications where the number of iterations is large — growing with n (examples are Corollary 2.5 and Theorem 2.11), and where care is needed in keeping track of the codegree function. In such circumstances, Theorem 6.3 can be used; Corollary 3.6 is then a special case of this theorem.

We begin with a simple lemma to help count the number of containers being generated by iteration.

**Lemma 6.1.** There are at most  $\exp\{s\theta n(1+\log(1/\theta))\}\$  s-tuples of subsets  $T_1, \ldots, T_s \subset [n]$ with  $|T_1| + \ldots + |T_s| \leq s\theta n$ , where  $0 \leq \theta \leq 1$ .

Proof. Let there be  $N_j$  such s-tuples with  $|T_1| + \ldots + |T_s| = j$ . We wish to bound  $N = N_0 + N_1 + \ldots + N_{\lfloor s\theta n \rfloor}$ . The generating function for the numbers of subsets of [n] of size i is  $(1+x)^n$ . Hence the coefficient of  $x^j$  in  $((1+x)^n)^s$  is the number of ways to choose sets  $T_1, \ldots, T_s$  of sizes  $t_1, \ldots, t_s$  such that  $t_1 + \cdots + t_s = j$ ; in other words  $N_0 + N_1 x + N_2 x^2 + \ldots = ((1+x)^n)^s$ . Therefore, since  $0 \le \theta \le 1$ , we have  $\theta^{s\theta n} N \le (1+\theta)^{ns} \le e^{s\theta n}$ .

The next theorem is a version of Theorem 3.4 stripped of references to degree measure. It is this theorem that we shall apply iteratively.

**Theorem 6.2.** Let G be an r-graph on vertex set [n]. Suppose that  $\delta(G, \tau) \leq 1/12r!$ , where  $0 < \tau < 1/2$ . Then there exists a collection  $C \subset \mathcal{P}[n]$  such that

- (a) for every independent set I there exists  $T = (T_{r-1}, \ldots, T_0) \in \mathcal{P}(I)^r$  with  $I \subset C(T) \in \mathcal{C}$  and  $|T_i| \leq 288r!^2 \tau n$ ,
- (b)  $\log |\mathcal{C}| \le 288 r!^2 rn\tau \log(1/\tau)$ , and
- (c)  $e(G[C]) \leq (1 1/2r!)e(G)$  for all  $C \in \mathcal{C}$ .

Moreover, (a) holds for all sets  $I \subset [n]$  for which either G[I] is  $\lfloor \tau^{r-1}e(G)/12r!n \rfloor$ degenerate or  $e(G[I]) \leq 24r!r\tau^r e(G)$ .

Proof. Let  $\zeta = 1/12r!$ . We may assume that  $\tau \leq \zeta^2/r$ , since otherwise we may take  $T_{r-1}, \ldots, T_0$  to be a partition of I into sets of size at most n/r and C(T) = I, in which case the constraints in the theorem are easily satisfied (since  $|\mathcal{C}| \leq 2^n$ : here we used  $\tau < 1/2$ ). Apply Theorem 3.4 to G with  $\zeta = 1/12r!$ . For each set I we have  $T = (T_{r-1}, \ldots, T_0)$  and a container C(T) satisfying properties (a)–(d) of that theorem. Take  $\mathcal{C}$  to be the collection of all such C. Since  $\tau \leq \zeta^2/r$ , we have  $2r\tau/\zeta \leq 2\zeta$ , so  $\mu(C) \leq 1 - 1/r! + 6\zeta = 1 - 1/2r!$ . It follows from inequality (1) that  $e(G[C]) \leq (1 - 1/2r!)e(G)$ .

Hence (a) and (c) of the present theorem are satisfied and it remains to check (b). Theorem 3.4 tells us that each container C is specified by sets  $T_0, \ldots, T_{r-1}$  each of size at

most  $\theta n$ , where  $\theta = 2\tau/\zeta^2 = 288r!^2\tau \le 2/r \le 1$ . By Lemma 6.1 we have

$$\log |\mathcal{C}| \le r\theta n (1 + \log(1/\theta)) \le r\theta n \log(1/\tau) = 288r!^2 rn\tau \log(1/\tau),$$

establishing (b) and completing the proof.

Repeated applications of Theorem 6.2 lead to the next theorem. The rather technical appearance is the natural consequence of retaining conditions on the codegree function at each stage, so that, if information on this function is available, then use can be made of it.

**Theorem 6.3.** Let G be an r-graph on vertex set [n]. Let  $e_0 \leq e(G)$ . Suppose that, for each  $U \subset [n]$  with  $e(G[U]) \geq e_0$ , the function  $\tau(U)$  satisfies  $\tau(U) < 1/2$  and  $\delta(G[U], \tau(U)) \leq 1/2$ 1/12r!. For  $e_0 \leq m \leq e(G)$  define

$$f(m) = \max\{-|U|\tau(U)\log\tau(U): U \subset [n], \ e(G[U]) \ge m\}$$
  
$$\tau^* = \max\{\tau(U): U \subset [n], \ e(G[U]) \ge e_0\}$$

Let  $k = \log(e_0/e(G))/\log(1-1/2r!)$ . Then there exists a collection  $\mathcal{C} \subset \mathcal{P}[n]$  such that

- (a) for every independent set I there exists  $T = (T_1, \ldots, T_s) \in \mathcal{P}(I)^s$  with  $I \subset C(T) \in$  $C, |T_i| \leq 288r!^2 \tau^* n \text{ and } s \leq (k+1)r,$
- (b)  $e(G[C]) \leq e_0 \text{ for all } C \in C$ , (c)  $\log |\mathcal{C}| \leq 288r!^2 r \sum_{0 \leq i < k} f(e_0/(1-1/2r!)^i)$ .

Moreover, (a) holds for all  $I \subset [n]$  for which either G[I] is  $|\tau(U)^{r-1}e(G[U])/12r!|U||$ degenerate or  $e(G[I]) \leq 24r!r\tau(U)^r e(G[U])$ , for all  $U \subset [n]$  with  $e(G[U]) \geq e_0$ .

*Proof.* We will show that for all t with  $e_0 \le t \le e(G)/(1-1/2r!)$ , there exists a collection  $\mathcal{C}_t \subset \mathcal{P}[n]$  satisfying conditions (a)–(c), where the constant  $e_0$  has been replaced by t in (a)–(c), and k is replaced by  $k(t) = \log(t/e(G))/\log(1-1/2r!)$ .

When  $t \ge e(G)$ , we may take  $C_t = \{[n]\}$ . Otherwise, suppose t < e(G). It is enough to show that  $C_t$  exists provided  $\mathcal{D} = C_{t/(1-1/2r!)}$  exists. Each  $D \in \mathcal{D}$  is specified by a tuple  $T' = (T_1, \ldots, T_{s'})$  with  $s' \le (k(t/(1-1/2r!))+1)r = k(t)r$ . If  $e(G[D]) \le t$ , let  $\mathcal{C}_t(D) = \{D\}$ . Otherwise, apply Theorem 6.2 with  $\tau = \tau(D) \leq \tau^*$  to the r-graph G[D], and let  $\mathcal{C}_t(D)$  be the collection of containers given by the theorem. Then put  $\mathcal{C}_t = \bigcup_{D \in \mathcal{D}} \mathcal{C}_t(D)$ .

If  $C \in \mathcal{C}_t(D)$  then C is specified completely by T', together with the r-tuple appearing in condition (a) of Theorem 6.2 if the theorem was applied. Hence C is specified completely by a tuple of size at most (k(t) + 1)r, so satisfying condition (a). If  $D \in \mathcal{D}$  then either  $e(G[D]) \leq t$  in which case  $|\mathcal{C}_t(D)| = 1$ , or e(G[D]) > t in which case

$$\log |\mathcal{C}_t(D)| \le 288r!^2 r |D|\tau(D) \log(1/\tau(D)) \le 288r!^2 r f(t).$$

Hence

$$\log |\mathcal{C}_t| \le \log |\mathcal{D}| + 288r!^2 r f(t) \le 288r!^2 r \sum_{0 \le i < k(t)} f(t/(1 - 1/2r!)^i).$$

Finally for  $C \in \mathcal{C}_t(D)$ , note that  $e(G[C]) \leq t$ , since if e(G[D]) > t then by condition (c) of Theorem 6.2  $e(G[C]) \le (1 - 1/2r!)e(G[D]) \le t$ . 

For certain applications the technical detail of Theorem 6.3 is not needed; what is required is a simple statement that a few iterations will produce a container with a negligible proportion of the original edges. Such a statement was presented earlier as Corollary 3.6.

Proof of Corollary 3.6. Let  $e_0 = \epsilon e(G)$ . Observe that for  $U \subset [n]$ , if  $e(U) \ge \epsilon e(G)$  then  $\delta(G[U], \tau) \le \delta(G, \tau)/\epsilon \le 1/12r!$ . Therefore we may apply Theorem 6.3 to the graph G with  $e_0 = \epsilon e(G)$  and  $\tau(U) = \tau$  for all U. Then  $\tau^* = \tau$  and  $f(m) = n\tau \log(1/\tau)$ . Hence we obtain a collection  $\mathcal{C}$  satisfying conditions (a) and (b) of the corollary, and

$$\log |\mathcal{C}| \le 288r!^2 r \left( 1 + \frac{\log \epsilon}{\log(1 - 1/2r!)} \right) n\tau \log(1/\tau),$$

giving condition (c).

#### 7. Uniformly bounded containers

Theorem 3.4 provides containers of degree measure bounded away from one. In this section we seek containers of uniform measure bounded away from one.

For regular hypergraphs, the results of §6 suffice, as pointed out in §3.5. However for non-regular hypergraphs we need something else. For reasons outlined in §3.5, we consider initial intervals  $[v] \subset [n]$ , and look for an interval such that  $|C \cap [v]|$  is bounded away from v. There will in fact be many such intervals, as the next lemma shows. This is the basic lemma which translates information about  $\mu$ -measure into information about uniform measure. In the lemma, S is a multiset, so  $\mu(S)$ ,  $|S \cap [v]|$  and so on have their natural interpretations counting with multiplicities.

**Lemma 7.1.** Let  $\mu : [n] \to \mathbb{R}$  be a measure with  $\mu(1) \ge \mu(2) \ge \cdots \ge \mu(n)$ , and let  $S \subset [n]$  be a multiset. Let  $W = \{v : |S \cap [v]| \ge \alpha v\}$ . Then

$$\alpha \,\mu(W) \,\leq \,\mu(S)$$

holds for all  $\alpha \geq 0$ .

*Proof.* Let  $W = \{w_1, \ldots, w_k\}$  where k = |W| and  $w_1 < w_2 < \ldots < w_k$ . Define the numbers  $s_1, \ldots, s_k$  by  $s_1 = |S \cap [w_1]|$  and  $s_i = |S \cap [w_{i-1} + 1, w_i]|$  for  $i \ge 2$ . Then we have  $\mu(S \cap [w_{i-1} + 1, w_i]) \ge s_i \mu(w_i)$ , because  $\mu(1) \ge \mu(2) \ge \cdots \ge \mu(n)$ . Therefore

$$\mu(S) \geq \mu(S \cap [w_1]) + \mu(S \cap [w_1 + 1, w_2]) + \dots + \mu(S \cap [w_{k-1} + 1, w_k])$$
  

$$\geq s_1 \mu(w_1) + s_2 \mu(w_2) + \dots + s_k \mu(w_k)$$
  

$$= \sum_{i=1}^k \alpha \mu(w_i) + \sum_{i=1}^k (s_1 + \dots + s_i - \alpha i)(\mu(w_i) - \mu(w_{i+1})),$$
  

$$= \alpha \mu(W) + \sum_{i=1}^k (s_1 + \dots + s_i - \alpha i)(\mu(w_i) - \mu(w_{i+1})),$$
(7)

where  $\mu(w_{k+1})$  is defined to be zero. Now  $|S \cap [w_i]| = s_1 + \cdots + s_i$  holds for  $1 \le i \le k$ , and so  $s_1 + \cdots + s_i \ge \alpha w_i$ , because  $w_i \in W$ . In particular,  $s_1 + \cdots + s_i \ge \alpha i$ , since  $w_i \ge i$ . Moreover  $\mu$  is a decreasing function, so each summand in (7) is non-negative, and the lemma follows.

In fact we shall need not just that  $|C \cap [v]|$  is bounded for a single container C but that the average  $(1/t) \sum_{i=1}^{t} |C_i \cap [v]|$  is bounded for a collection  $C_1, \ldots, C_t$ . We shall, at the same time, be interested in the sets  $T_1, \ldots, T_s$  used to construct these containers, and we need to find a v for which the average of the  $|T_j \cap [v]|$  is simultaneously under control. For technical reasons arising when we come to the application, very small values of v will be of no use, so we set a lower bound on its value. The next lemma prepares the way. **Lemma 7.2.** Let  $\mu$  be a probability measure on [n] with  $\mu(1) \ge \mu(2) \ge \cdots \ge \mu(n) \ge 0$ . Let  $T_1, \ldots, T_s, C_1, \ldots, C_t$  be subsets of [n], with  $\mu(T_i) \le \lambda$  for  $1 \le i \le s$  and  $\mu(C_j) \le 1 - c - \eta$  for  $1 \le j \le t$ , where  $c, \eta > 0$ . Suppose moreover that  $k \in [n]$  and  $\mu([k]) \le \eta c$ . Then there exists  $v \in [k, n]$  with

$$\frac{1}{s}\sum_{i=1}^{s}|T_i\cap[v]| < \frac{\lambda}{\eta}v \qquad and \qquad \frac{1}{t}\sum_{i=1}^{t}|C_i\cap[v]| < (1-c)v$$

*Proof.* Let  $U = \{v : \sum_{i=1}^{s} |T_i \cap [v]| \ge s\lambda v/\eta\}$ . Writing S for the multiset which is the disjoint union of  $T_1, \ldots, T_s$ , so that  $\mu(S) \le s\lambda$  and  $|S \cap [v]| = \sum_{i=1}^{s} |T_i \cap [v]|$ , we can apply Lemma 7.1 with  $\alpha = s\lambda/\eta$  to obtain  $\mu(U) \le \mu(S)/\alpha \le \eta$ .

In like manner, let  $W = \{v : \sum_{i=1}^{t} |C_i \cap [v]| \ge t(1-c)v\}$ . Writing now S for the multiset which is the disjoint union of  $C_1, \ldots, C_t$ , so that  $\mu(S) \le t(1-c-\eta)$  and  $|S \cap [v]| = \sum_{i=1}^{t} |C_i \cap [v]|$ , we apply Lemma 7.1 with  $\alpha = t(1-c)$  to obtain  $\mu(W) \le t(1-c-\eta)/\alpha = 1-\eta/(1-c)$ .

It follows that  $\mu(U \cup W \cup [k]) \leq \eta + 1 - \eta/(1-c) + \eta c < 1$ , so there exists  $v \in [n]$  not contained in  $U \cup W \cup [k]$ . This v satisfies the conditions of the corollary (indeed, with  $v \in [k+1,n]$ ).

We can now prove the main result about containers and uniform measure, namely Theorem 3.7, which was discussed in §3.5. The idea of the proof is roughly as follows. Theorem 3.4 supplies a set of containers. For each tuple  $(C_1, \ldots, C_t)$  of these containers we use Lemma 7.2 to nominate a vertex  $v = g(C_1, \ldots, C_t)$  so that the restrictions to [v] of the  $C_i$  and of their generating sets  $T_j$  are simultaneously bounded in uniform measure. The online property means that the restrictions  $C_i \cap [v]$  are determined by the  $T_j \cap [v]$ , which are small, and so the number of restricted containers is small.

Proof of Theorem 3.7. Apply Theorem 3.4 to G to obtain a collection C of containers C(T) for  $T = (T_{r-1}, \ldots, T_0) \in \mathcal{P}([n])^r$ . By assumption,  $\tau \leq \zeta^2/r$ , and so  $2r\tau/\zeta \leq 2\zeta$ . Therefore  $\mu(C(T)) \leq 1 - 1/r! + 6\zeta$ .

Let  $(C_1, \ldots, C_t) \in \mathcal{C}^t$ , where  $t \in \mathbb{N}$ . Each  $C_i$  is specified by an *r*-tuple of sets  $T_j$ , so the whole collection  $(C_1, \ldots, C_t)$  is specified by rt sets which, after re-labelling, we call  $T_1, \ldots, T_{rt}$ , with  $\mu(T_i) \leq 2\tau/\zeta$  for  $1 \leq i \leq rt$ . Let  $c = 1/r! - 8\zeta$ , so (since  $\zeta \leq 1/12r!$ ) c > 1/4r!. Let  $\eta = 2\zeta$ . By assumption,  $\mu([k]) \leq \zeta/2r!$ , so  $\mu([k]) < \eta c$ . Hence the conditions of Lemma 7.2 are satisfied with s = rt and  $\lambda = 2\tau/\zeta$ , and so there exists  $v \in [k, n]$  with

$$\frac{1}{s}\sum_{i=1}^{s}|T_{i}\cap[v]| < \frac{\tau}{\zeta^{2}}v \quad \text{and} \quad \frac{1}{t}\sum_{i=1}^{t}|C_{i}\cap[v]| < (1-\frac{1}{r!}+8\zeta)v.$$

Define  $g(C_1, \ldots, C_t) = v$ . Then (a) and (c) of the theorem are satisfied.

To obtain (b), we need that the containers have the online property: in other words, the t-tuple  $(C_1 \cap [v], \ldots, C_t \cap [v])$  is determined by  $T_1 \cap [v], \ldots, T_s \cap [v]$ . This online property is guaranteed by Theorem 3.4. Hence the size of the set  $Z = \{(C_1 \cap [v], \ldots, C_t \cap [v]) : g(C_1, \ldots, C_t) = v\}$  is bounded by the number of tuples  $(T_1 \cap [v], \ldots, T_s \cap [v])$ . Now  $\sum_{i=1}^s |T_i \cap [v]| < s\theta v$ , where  $\theta = \tau/\zeta^2 < 1$ . So by Lemma 6.1

$$\log |Z| \le s\theta v (1 + \log(1/\theta)) \le s\theta v \log(1/\tau) = \zeta^{-2} v tr\tau \log(1/\tau),$$

which completes the proof.

#### 8. LIST COLOURINGS

In [55], a lower bound for the list colouring number of a regular hypergraph was proved. Theorem 2.1 of that paper, based on a simple probabilistic argument, gave a bound of approximately  $(\log k)/\log(1/c)$  provided there is a collection C of containers for the independent sets, with  $|C| \leq (1-c)n$  for each  $C \in C$  and with  $|C| \leq e^{n/k}$ . The proof fails to work for a general hypergraph because it is not possible to find containers of bounded size.

As mentioned in §3.5, Corollary 3.6 supplies suitable containers for regular hypergraphs, and the number of containers is fewer than in [55]. This gives a direct improvement on the result of [55]. However, to obtain a similar result for general hypergraphs we must make use of Theorem 3.7.

It is worth recapping briefly the simple argument of [55], because it explains the basis of what follows though without the technicalities. It also gives a clear illustration of why containers are useful.

Let G be an r-graph with vertex set [n]. Let [t] be some set of colours and let  $\mathcal{L} = \{L_u : u \in [n], L_u \subset [t]\}$  be a collection of colour lists, one for each vertex, with  $|L_u| = \ell$  for each  $u \in [n]$ . A colouring of G is a choice function  $f : [n] \to [t]$  with  $f(u) \in L_u$  such that no edge is monochromatic. If we can find a collection  $\mathcal{L}$  with no colouring, then  $\chi_l(G) > \ell$ , which is our goal. We choose the lists  $L_u$  at random from a palette [t] with t around  $\ell^2$  (so choosing with replacement is much the same as choosing without). If the lists admit a choice function f, then, for each colour  $i \in [t]$ , the set of vertices with f(u) = i is independent. Thus there exists a collection of independent sets  $(I_1, \ldots, I_t)$  with  $u \in I_{f(u)}$  for all  $u \in [n]$ . We say that  $\mathcal{L}$  is compatible with  $(I_1, \ldots, I_t)$  if such a choice function f exists with  $u \in I_{f(u)}$  for all u; in other words, the graph can be coloured so that all the vertices receiving colour i lie within  $I_i, 1 \leq i \leq t$ . Notice that we did not specify that  $I_i$  is precisely the set of vertices u with f(u) = i, only that it contains them all.

Let  $\mathcal{I}$  be the collection of independent sets. It follows that if we can find a collection  $\mathcal{L}$  compatible with no tuple  $(I_1, \ldots, I_t) \in \mathcal{I}^t$ , then we have shown  $\chi_l(G) > \ell$ . We say that such an  $\mathcal{L}$  is  $\mathcal{I}$ -incompatible. Suppose now that  $|I| \leq (1-c)n$  for all  $I \in \mathcal{I}$ . Roughly speaking (precise calculations come in the proof of Lemma 8.1), given a tuple  $(I_1, \ldots, I_t)$ , an average vertex u will lie in at most (1-c)t of the  $I_i$ , so the probability that  $L_u$  contains a colour f(u) with  $u \in I_{f(u)}$  is at most  $1 - c^\ell \leq e^{-c^\ell}$ . Hence the probability of  $\mathcal{L}$  being compatible with a given tuple  $(I_1, \ldots, I_t)$  is at most  $e^{-nc^\ell}$ , and so the probability that  $\mathcal{L}$  fails to be  $\mathcal{I}$ -incompatible is at most  $|\mathcal{I}|^t e^{-nc^\ell}$ . If this were less than one then there would exist an  $\mathcal{I}$ -incompatible collection  $\mathcal{L}$ . Unfortunately  $|\mathcal{I}|$  can be as large as  $2^{\Omega(n)}$  and the approach yields nothing.

However, the same argument can be made with the containers  $\mathcal{C}$  in place of the independent sets  $\mathcal{I}$ ; for each independent set  $I_j$  above there must be a  $C_j$  containing it, and for a choice function to work there must be a tuple  $(C_1, \ldots, C_t)$  with which f is compatible, meaning  $u \in C_{f(u)}$  for each  $u \in [n]$ . We now want  $\mathcal{L}$  to be  $\mathcal{C}$ -incompatible, that is, compatible with no  $(C_1, \ldots, C_t)$ , and, assuming  $|C_i| \leq (1-c)n$  for all i, the probability of this failing is at most  $|\mathcal{C}|^t e^{-nc^\ell}$ . If  $|\mathcal{C}| = 2^{\tau n}$  with  $\tau = d^{-1/(r-1)}$ , then this probability is less than one for some  $\ell$  with  $\ell = \Omega(\log d)$ , which is therefore a lower bound for  $\chi_l(G)$ .

The existence of  $\mathcal{L}$ , contingent on the existence of a suitable set of containers  $\mathcal{C}$ , is proved in detail in the next lemma. The main difference between the lemma and the preceding sketch is that we cannot assume  $|C_i| \leq (1-c)n$  for each container, and instead we must use the properties of C given by Theorem 3.7.

**Lemma 8.1.** Let  $0 < \epsilon, c < 1$ . Then there exists  $k_0 = k_0(\epsilon, c)$ , such that the following property holds for all  $k > k_0$ .

Let  $\ell = \lfloor (1-\epsilon) \log k / \log(1/c) \rfloor$  and let  $t = \lfloor 2\ell^2/c \rfloor$ . Let n > k and let  $\mathcal{C} \subset \mathcal{P}[n]$ . Suppose that there is a map  $g : \mathcal{C}^t \to [k, n]$ , such that

(a) 
$$\frac{1}{t} \sum_{i=1}^{t} |C_i \cap [v]| \le (1-c)v$$

holds for every  $(C_1, \ldots, C_t) \in \mathcal{C}^t$ , where  $v = g(C_1, \ldots, C_t)$ . Suppose moreover that

(b) 
$$|\{ (C_1 \cap [v], \dots, C_t \cap [v]) : g(C_1, \dots, C_t) = v \}| \le e^{vt/k}$$

holds for all  $v \in [n]$ . Then there is a collection of lists  $\{L_u : u \in [n]\}$ , each of size  $|L_u| = \ell$ , which is C-incompatible.

*Proof.* For each  $u \in [n]$ , let  $L_u \in [t]^{(\ell)}$  be a subset of [t] of size  $\ell$  chosen uniformly and independently at random, and let  $\mathcal{L} = \{L_u : u \in [n]\}$  be the collection of lists. We need to show that, with positive probability,  $\mathcal{L}$  is compatible with no tuple  $(C_1, \ldots, C_t) \in \mathcal{C}^t$ .

Given some  $(C_1, \ldots, C_t) \in \mathcal{C}^t$ , then  $\mathcal{L}$  is compatible with  $(C_1, \ldots, C_t)$  if there is a choice function  $f: [n] \to [t]$  with  $u \in C_{f(u)}$  for all u. We define, for each  $u \in [n]$ , the set of colours

$$B_u = B_u(C_1, \dots, C_t) = \{i \in [t] : u \in C_i\}.$$

We can find a choice function if, and only if, we can select  $f(u) \in L_u \cap B_u$  for each  $u \in [n]$ ; in other words, if  $L_u \cap B_u \neq \emptyset$ . Hence we shall prove the theorem by showing that, with positive probability, for every tuple  $(C_1, \ldots, C_t)$  there is some  $u \in [n]$  with  $L_u \cap B_u = \emptyset$ .

In fact, we claim something stronger: with positive probability,  $\mathcal{L}$  rejects every tuple  $(C_1, \ldots, C_t)$ , meaning that there is some  $u \in [v]$  with  $L_u \cap B_u = \emptyset$ , where  $v = g(C_1, \ldots, C_t)$ . Notice that the event that  $(C_1, \ldots, C_t)$  is rejected depends only on  $\mathcal{L}$  and on the tuple  $(C_1 \cap [v], \ldots, C_t \cap [v])$ ; it is because the conditions of the theorem give information about this tuple that we work with the stronger claim.

To establish the claim, fix for the time being some tuple  $(C_1, \ldots, C_t)$  and let  $v = g(C_1, \ldots, C_t)$ . Let  $u \in [v]$  and write  $\mathbf{1}_{u \in C_i}$  for the indicator that  $u \in C_i$ . By condition (a) of the theorem, we have

$$\sum_{u \in [v]} |B_u| = \sum_{u \in [v]} \sum_{i=1}^t \mathbf{1}_{u \in C_i} = \sum_{i=1}^t |C_i \cap [v]| \le (1-c)vt.$$
(8)

Let  $p_u$  be the probability that  $L_u \cap B_u = \emptyset$ , or equivalently  $L_u \subset [t] \setminus B_u$ . Then

$$p_u = \mathbb{P}(L_u \cap B_u = \emptyset) = {\binom{z_u}{\ell}} {\binom{t}{\ell}}^{-1}$$
 where  $z_u = t - |B_u|$ .

We note here that  $\ell \ge 1$  if  $k_0$  is large enough and thus  $ct > \ell$ . Write z for the average of the values  $z_u$  for  $u \in [v]$ . Then inequality (8) yields  $vz = \sum_u t - |B_u| \ge vct$ . So we have

$$\sum_{u \in [v]} p_u = \sum_{u \in [v]} {\binom{z_u}{\ell}} {\binom{t}{\ell}}^{-1} \ge v {\binom{z}{\ell}} {\binom{t}{\ell}}^{-1} \\ \ge v {\binom{ct}{\ell}} {\binom{t}{\ell}}^{-1} \ge v(c - (\ell - 1)/t)^{\ell}.$$

Since  $\ell \geq 1$  we have  $(\ell - 1)/t \leq (\ell - 1)/(2\ell^2/c - 1) \leq c/2\ell$ , and so  $(c - (\ell - 1)/t)^\ell \geq c^\ell (1 - 1/2\ell)^\ell \geq c^\ell/2$ . Hence the probability that  $\mathcal{L}$  fails to reject  $(C_1, \ldots, C_t)$  is

$$\mathbb{P}(B_u \cap L_u \neq \emptyset \text{ for all } u \in [v]) = \prod_{u \in [v]} (1 - p_u)$$
$$\leq \exp\{-\sum_{u \in [v]} p_u\} \leq \exp\{-vc^{\ell}/2\}$$

As mentioned, the event that  $(C_1, \ldots, C_t)$  is not rejected depends only on the tuple  $(C_1 \cap [v], \ldots, C_t \cap [v])$  and, by condition (b) of the theorem, there are at most  $\exp\{vt/k\}$  of these tuples as  $(C_1, \ldots, C_t)$  ranges over  $\mathcal{C}^t$ . Hence if we fix v and write  $P_v$  for the probability that there is some tuple  $(C_1, \ldots, C_t)$  with  $v = g(C_1, \ldots, C_t)$  which is not rejected, then, recalling the definitions  $\ell = |(1 - \epsilon) \log k / \log(1/c)|$  and  $t = |2\ell^2/c|$ , we have

$$P_{v} \leq \exp\{vt/k - vc^{\ell}/2\}$$

$$\leq \exp\left\{\frac{v}{2k}\left[\frac{4}{c}\left(\frac{(1-\epsilon)\log k}{\log 1/c}\right)^{2} - k^{\epsilon}\right]\right\} \qquad \text{since } c^{\ell} \geq k^{\epsilon-1}$$

$$\leq \exp\left\{-\frac{v}{2k}k^{\epsilon/2}\right\} \qquad \text{if } k_{0}, \text{ and so } k, \text{ is large enough}$$

$$\leq \exp\left\{-\frac{1}{2}v^{\epsilon/2}\right\} \qquad \text{since } k \leq v$$

$$\leq v^{-2} \qquad \text{if } k_{0}, \text{ and so } v \geq k_{0}, \text{ is large enough}.$$

Finally, if we consider all tuples  $(C_1, \ldots, C_t) \in C^t$ , the probability that one of them is not rejected is at most

$$\sum_{v \in [k,n]} P_v \le \sum_{v \ge k} v^{-2} < 1$$

if  $k_0$  is large. This establishes our claim and so proves the lemma.

We can now prove our main result about list colouring. The proof follows by feeding Theorem 3.7 into Lemma 8.1 (for regular graphs we use Corollary 3.6 instead of Theorem 3.7). The lower bound on  $\chi_l(G)$  given by Lemma 8.1 is  $(1 + o(1)) \log k / \log(1/c)$ . Comparing condition (b) in Theorem 3.7 with that in Lemma 8.1 shows that k is not far from  $\zeta^2/\tau$ , and we know that  $\tau$  for simple graphs can be roughly  $d^{-1/(r-1)}$ . This explains where the log d in the theorem comes from.

To get the best result, we want the number c in Lemma 8.1 to be as large as possible, which, by comparing Lemma 8.1(a) with Theorem 3.7(c) means making  $\zeta$  small (unlike in other applications where typically  $\zeta = 1/12r!$  is a good choice.) However if  $\zeta$  is too small

then k becomes small. For these reasons we choose  $\zeta = \zeta(d)$  so that, as  $d \to \infty$ , then  $\zeta = o(1)$  and  $\zeta = d^{o(1)}$ , the exponent here being negative.

Proof of Theorem 2.1. As explained in the preceding discussion, we take  $\zeta = \zeta(d)$  so that, as  $d \to \infty$ , then  $\zeta = o(1)$  and  $\zeta = d^{o(1)}$ . Let  $\tau = d^{-1/(r-1)}\zeta^{-3}$ . We now check that the conditions of Theorem 3.7 are satisfied. Certainly  $\zeta \leq 1/12r!$  because  $\zeta = o(1)$ . Also, recalling Definition 3.2, we have  $d^{(j)}(v) \leq 1$  by simplicity and  $\delta_j = \sum_v d^{(j)}(v)/\tau^{j-1}nd \leq \zeta^2 = o(\zeta)$ , so  $\delta(G, \tau) \leq \zeta$ . Moreover  $\tau \leq \zeta^2/r$  because  $\zeta = d^{o(1)}$ .

Let  $k = \lfloor \zeta^3 / \tau \log(1/\tau) \rfloor$ . Then  $\log k = (1/(r-1) + o(1)) \log d$ . Let  $v \in V(G)$ . For each  $u \in V(G)$ , at most one edge contain both u and v, and so  $d(v) \leq n$ . It follows that  $\mu([k]) \leq (1/nd)kn \leq \zeta/2r!$ . This completes the check of the conditions of Theorem 3.7. Therefore there exists a collection C of containers for the independent sets of G, satisfying properties (b) and (c) of Theorem 3.7, and since  $\zeta^{-2}r\tau \log(1/\tau) < 1/k$  it follows that conditions (a) and (b) of Lemma 8.1 are satisfied, with  $c = 1/r! - 8\zeta \geq (1 + o(1))r^{-(r-1)}$ .

Consequently there are lists of size  $(1 + o(1)) \log k / \log(1/c)$  that are not C-compatible, which is to say lists of size at least  $(1/(r-1)+o(1)) \log d / \log(1/c) \ge (1/(r-1)^2+o(1)) \log_r d$ . Since C is a set of containers for the independent sets of G, the first claim of the theorem follows.

The proof for regular graphs is similar, except that in Lemma 8.1 we are able to take c = 1/r + o(1). To achieve this we make use of Corollary 3.6 instead of Theorem 3.7. With  $\tau$ ,  $\zeta$  and k defined as before, we can take  $\epsilon = \zeta$  in Corollary 3.6 because  $\delta(G, \tau) = o(\zeta)$ . We obtain a collection C of containers such that  $e(G[C]) \leq \zeta e(G) = o(e(G))$  for all  $C \in C$ . Because G is regular this implies, as mentioned after inequality (2), that  $|C| \leq (1 - 1/r + o(1))n$  where n = |G|. We can now apply Lemma 8.1 by defining  $g(C_1, \ldots, C_t) = n$  for all  $(C_1, \ldots, C_t)$ ; note that condition (b) of the theorem is satisfied because, by Corollary 3.6,  $\log |\mathcal{C}| = O(\log(1/\epsilon)n\tau \log(1/\tau)) < n/k$ . The remainder of the proof is the same.

The bound given for r-graphs of average degree d is weaker than that for regular r-graphs because we only had containers of measure 1 - 1/r! available, rather than 1 - 1/r. Probably this is an artifact of our algorithm, and  $\chi_l(G) \ge (1/(r-1) + o(1)) \log_r d$  holds for r-graphs of average degree d.

Observe that, since the proof uses Corollary 3.6 instead of Theorem 3.7 for regular graphs, it is not necessary to impose the condition of simplicity in the regular case. The proof in fact works provided  $d^{(j)}(v) \leq d^{(r-j)/(r-1)+o(1)}$  as  $d \to \infty$  for every  $v \in V(G)$  and every  $2 \leq j \leq r$  (recall Definition 3.2), since  $\tau$  can then be chosen to ensure  $\delta(G, \tau) \leq \zeta$ , and the bound on  $d^{(2)}$  implies a bound on the maximum degree which in turn bounds  $\mu([k])$ . This implies a theorem of Alon and Kostochka [5] in the case of regular hypergraphs.

As far as non-simple regular graphs go, the bound is tight. Indeed, let K(r,m) be the complete *r*-partite *r*-graph with *m* vertices in each class. Suppose that lists of size  $\ell$  are given to the vertices. Randomly choose, for each colour in the palette, a vertex class on which that colour is forbidden to be used; then the expected number of vertices with no available colour is  $rmr^{-\ell}$  which is less than one if  $\ell > 1 + \log_r m$ , and so  $\chi_l \leq$  $2 + \log_r m$  (see Haxell and Verstraëte [27]). This graph is *d*-regular where  $d = m^{r-1}$  so  $\chi_l \leq 2 + (1/(r-1))\log_r d$ . Note that  $d^{(j)}(v) = m^{r-j} = d^{(r-j)/(r-1)}$ .

It is not hard to construct an *m*-regular simple subgraph G of K(r, m), and so (putting d = m) we have simple d-regular r-graphs with  $\chi_l \leq 2 + \log_r d$ . Quite possibly  $\chi_l \leq 2$ 

 $2 + (1/(r-1))\log_r d$  in this case too, because a subgraph of G with  $d^{1-1/(r-1)}$  vertices in each class is likely to be very sparse, and a random colouring might be repairable if  $rdr^{-\ell} < d^{1-1/(r-1)}$ , or  $\ell > 1 + (1/(r-1))\log_r d$ . But this argument is far from rigorous.

As an illustration of the use of containers for non-independent sets we finish with the next result.

**Theorem 8.2.** Let G be a graph with average degree d. Then, for each  $u \in V(G)$  there is a list  $L_u$  of  $(1+o(1)) \log_2 d$  colours, such that it is not possible to choose a colour  $c(u) \in L_u$ with the vertices of each colour spanning a planar graph.

*Proof.* We follow the proof of Theorem 2.1 with r = 2, except we use a set  $\mathcal{C}$  of containers for those subsets I for which G[I] is planar. Since a planar graph is 5-degenerate, we can apply Theorem 3.7 and continue with the proof exactly as before, provided  $5 \leq \tau d\zeta/r$ . But  $\tau = d^{-1}\zeta^{-3}$  so this condition holds comfortably.

It is possible to extend the colouring results here to non-simple r-graphs — see §12.

#### 9. H-FREE GRAPHS

In this section we prove Theorem 2.3. In fact we will show a slight strengthening of it. We will apply the container theorem given by Corollary 3.6 to the following hypergraph, whose independent sets correspond to H-free  $\ell$ -graphs on vertex set [N].

**Definition 9.1.** Let H be an  $\ell$ -graph. Let r = e(H). The r-graph G(N, H) has vertex set  $[N]^{(\ell)}$ , where  $B = \{v_1, ..., v_r\} \in V(G)^{(r)}$  is an edge whenever B, considered as an  $\ell$ -graph with vertices in [N] and with r edges, is isomorphic to H.

We re-emphasise that all our results about *H*-free graphs are simple consequences of Theorem 2.3, and that this theorem is itself just a restatement, in graphical language, of the container theorem applied to the hypergraph G(N, H). As already mentioned, *H*-free graphs are precisely independent subsets of G(N, H), and the graphs that contain the *H*-free graphs, which themselves contain few *H*-free graphs, are precisely the containers given by Corollary 3.6. In order to apply the corollary to G(N, H), all that is needed is to estimate  $\delta(G(N, H), \tau)$ . The easy calculation is carried out in Lemma 9.3. As discussed in §3.6, the outcome is more or less optimal, for every *H*.

That said, we do permit ourselves a variation on this theme. For certain purposes, it turns out that we want to work, not with all possible copies of H, but with only a subset of them of particular interest. For example, we may care only about copies of H whose vertices are aligned with some partition of [N]; this is the case for the KLR conjecture (Theorem 10.2). It will be seen that all that is needed here is to apply the container theorem to a subgraph of G(N, H) rather than to G(N, H) itself. Hence the next theorem generalizes Theorem 2.3 by allowing this. At the same time we also strengthen the theorem by providing a collection of containers for graphs that are not necessarily H-free, but nonetheless contain few copies of H. This is allowed by the container theorem, in which the sets I do not have to be completely independent.

The notation  $\widetilde{G} \subset G(N, H)$  means  $\widetilde{G}$  is a subgraph of G(N, H), and in every case of interest  $V(\widetilde{G}) = V(G(N, H)) = [N]^{(\ell)}$ . The edges of  $\widetilde{G}$  thus represent a subcollection of the copies of H on vertex set [N]. So if  $A \subset [N]^{(\ell)}$ , that is, if A is an  $\ell$ -graph on vertex set [N], then the induced e(H)-graph  $\widetilde{G}[A]$  corresponds to all copies of H in the collection  $\widetilde{G}$  that

are present in A, and  $e(\tilde{G}[A])$  is the number of copies of H that are both in the collection  $\tilde{G}$  and present in A.

Recall that  $\pi(H) = \lim_{N \to \infty} \exp(N, H) {N \choose \ell}^{-1}$ .

**Theorem 9.2.** Let H be an  $\ell$ -graph with  $e(H) \geq 2$  and let  $\epsilon > 0$ . There exists c > 0 such that the following is true. Let  $N \geq c$ . Let  $\widetilde{G} \subset G(N, H)$  with  $e(\widetilde{G}) \geq \epsilon N^{v(H)}$ . Let q satisfy  $N^{-1/m(H)} \leq q \leq 1/c$ . Then there exists a collection  $\mathcal{C}$  of  $\ell$ -graphs on vertex set [N] such that

- (a) for every  $\ell$ -graph  $I \subset [N]^{(\ell)}$  with  $e(\widetilde{G}[I]) < q^{e(H)}N^{v(H)}$ , there exists  $C \in \mathcal{C}$  with  $I \subset C$ ,
- (b) every  $C \in \mathcal{C}$  satisfies  $e(\widetilde{G}[C]) \leq \epsilon N^{v(H)}$ , and moreover if  $\widetilde{G} = G(N,H)$  then  $e(C) \leq (\pi(H) + \epsilon) {N \choose \ell}$ ,
- (c)  $\log |\mathcal{C}| \le cq N^{\ell} \log N$ .
- (d) moreover, for every I in (a), there exists  $T = (T_1, \ldots, T_s)$  where  $T_i \subset I$ ,  $s \leq c$  and  $\sum_i e(T_i) \leq cqN^{\ell}$ , such that C = C(T),

Theorem 2.3 follows immediately from Theorem 9.2 by taking  $\tilde{G} = G(N, H)$  and  $q = N^{-1/m(H)}$ .

All that remains before applying the container theorem to G(N, H), or more generally to some dense subgraph  $\tilde{G}$ , is to calculate  $\delta(\tilde{G}, \tau)$ .

**Lemma 9.3.** Let H be an  $\ell$ -graph with  $r = e(H) \ge 2$  and let  $\gamma \le 1$ . Let N be sufficiently large. Let  $\widetilde{G} \subset G(N, H)$  with  $e(\widetilde{G}) = \alpha N^{v(H)}$  for some  $\alpha > 0$ . Then

$$\delta(\widetilde{G}, \gamma^{-1} N^{-1/m(H)}) \le 2^{r^2} v(H)! \gamma/\alpha.$$

Proof. Consider  $\sigma \subset [N]^{(\ell)}$  (so  $\sigma$  is both a set of vertices of  $\widetilde{G}$  and an  $\ell$ -graph on vertex set [N]). The degree of  $\sigma$  in  $\widetilde{G}$  is at most the number of ways of extending  $\sigma$  to an  $\ell$ -graph isomorphic to H. If  $\sigma$  as an  $\ell$ -graph is not isomorphic to any subgraph of H, then clearly  $d(\sigma) = 0$ . Otherwise, let  $v(\sigma)$  be the number of vertices in  $\sigma$  considered as an  $\ell$ -graph, so there exists  $V \subset [N]$ ,  $|V| = v(\sigma)$  with  $\sigma \subset V^{(\ell)}$ . Edges of  $\widetilde{G}$  containing  $\sigma$  correspond to copies of H in  $[N]^{(\ell)}$  containing  $\sigma$ , each such copy given by a choice of  $v(H) - v(\sigma)$  vertices in [N] - V and a permutation of the vertices of H. Hence for N sufficiently large,

$$d(\sigma) \le v(H)! \binom{N - v(\sigma)}{v(H) - v(\sigma)} \le v(H)! N^{v(H) - v(\sigma)}$$

For  $v \in V(G)$  and  $1 \leq j \leq r$ , the quantity  $d^{(j)}(v)$  is the maximum of  $d(\sigma)$  over all  $\sigma \subset [N]^{(\ell)}$  with  $v \in \sigma$  and  $|\sigma| = j$ . Thus

$$d^{(j)}(v) \le v(H)! N^{v(H)-f(j)}$$
, where  $f(j) = \min_{H' \subset H, e(H')=j} v(H')$ .

Let  $\tau = \gamma^{-1} N^{-1/m(H)}$ . Since  $\sum_{v} d^{(1)}(v) = \alpha r N^{v(H)}$ , for  $2 \leq j \leq e(H)$  we have

$$\delta_j = \frac{\sum_v d^{(j)}(v)}{\tau^{j-1} \alpha r N^{v(H)}} \le (v(H)!/r\alpha) \tau^{1-j} N^{\ell-f(j)} \le (v(H)!/r\alpha) N^{\ell-f(j)+(j-1)/m(H)} \gamma.$$

By definition of f(j) and m(H),  $\ell - f(j) + (j-1)/m(H) \leq 0$ . Hence  $\delta_j \leq (v(H)!/r\alpha)\gamma$ and so

$$\delta(G,\tau) = 2^{\binom{r}{2}-1} \sum_{j=2}^{r} 2^{-\binom{j-1}{2}} \delta_j \le 2^{r^2} (v(H)!/\alpha) \gamma$$

as claimed.

A well-known supersaturation theorem bounds the number of edges in containers.

**Proposition 9.4** (Erdős and Simonovits [22]). Let H be an  $\ell$ -graph and let  $\epsilon > 0$ . There exists  $N_0$  and  $\eta > 0$  such that if C is an  $\ell$ -graph on  $N \ge N_0$  vertices containing at most  $\eta N^{v(H)}$  copies of H then  $e(C) \le (\pi(H) + \epsilon) {N \choose \ell}$ .

Proof of Theorem 9.2. In what follows, c is taken to be sufficiently large (depending on  $\epsilon$  and H). Let  $\eta = \eta(\epsilon, H)$  be given by Proposition 9.4, and let  $\beta = \min\{\epsilon, \eta\}$ . Recall that r = e(H). Apply Corollary 3.6 to  $\widetilde{G}$  with  $\tau = \sqrt{cq}$  and with  $\beta$  playing the role of  $\epsilon$  in the corollary. Thus  $\sqrt{c}N^{-1/m(H)} \leq \tau$ , and so Lemma 9.3 implies that  $\delta(\widetilde{G}, \tau) \leq \beta/12r!$  if c is large. Moreover  $\tau \leq 1/\sqrt{c} < 1/2$  if c is large. Hence the conditions of Corollary 3.6 are satisfied; denote by  $\tilde{c}$  the constant c appearing in the corollary. The collection of containers C satisfies the following.

- For every I with  $e(\widetilde{G}[I]) \leq 24\beta r! r\tau^r e(\widetilde{G})$ , there exists some  $C \in \mathcal{C}$  with  $I \subset C$ . This implies condition (a) of the present theorem, since  $q^{e(H)}N^{v(H)} = (\tau/\sqrt{c})^r N^{v(H)} \leq (\tau/\sqrt{c})^r (1/\epsilon) e(\widetilde{G}) \leq 24\beta r! r\tau^r e(\widetilde{G})$  provided that c is sufficiently large.
- For each  $C \in \mathcal{C}$ , we have  $e(\widetilde{G}[C]) \leq \beta e(\widetilde{G}) \leq \beta N^{v(H)}$ . In the case that  $\widetilde{G} = G(N, H)$ , Proposition 9.4 implies  $e(C) \leq (\pi(H) + \epsilon) {N \choose \ell}$ , because we chose  $\beta \leq \eta$ . This gives condition (b).
- The size of the collection is  $\log |\mathcal{C}| \leq \tilde{c} \log(1/\beta) {N \choose \ell} \tau \log(1/\tau)$ , which gives condition (c), again provided that c is sufficiently large.
- Finally, for every set I as above, there exists  $T = (T_1, \ldots, T_s) \in \mathcal{P}(I)^r$  such that  $C = C(T), |T_i| \leq \tilde{c}\tau {N \choose \ell}$ , and  $s \leq \tilde{c}\log(1/\beta)$ . This implies condition (d) of the present theorem, provided that c is sufficiently large.

This completes the proof.

We now prove the theorems about induced H-free graphs that were stated in §2.4. As mentioned there, the proofs are very similar to those just given for H-free graphs, so we shall sketch the details. The crucial difference is that we need to consider containers not in G(N, H) but in another hypergraph that captures induced copies of H.

We already discussed in §2.4 how the notion of 2-coloured multigraphs can help. We say that a 2-coloured  $\ell$ -multigraph J on vertex set [N] is *entire* if  $J_R \cup J_B = [N]^{(\ell)}$ . One can think of J as representing a class of  $\ell$ -graphs on vertex set [N], in each of which the edges of  $J_R \setminus J_B$  are present, the edges of  $J_B \setminus J_R$  are absent, but edges in  $J_R \cap J_B$  can be present or absent.

Let  $r = \binom{v(H)}{\ell}$ . Let  $G_i(N, H)$  be the *r*-graph whose vertex set is two copies of  $[N]^{(\ell)}$ , denoted by  $V_R$  and  $V_B$  (vertices in  $V_R$  correspond to  $\ell$ -edges and vertices in  $V_B$  correspond to non- $\ell$ -edges), and whose edges correspond to induced copies of H; thus  $f \in (V_R \cup V_B)^{(r)}$  is an edge of  $G_i(N, H)$  whenever  $f \cap V_R$  and  $f \cap V_B$  are the edges and non-edges, respectively, of an  $\ell$ -graph isomorphic to H with vertices in [N]. Note that every induced-H-free  $\ell$ graph  $I \subset [N]^{(\ell)}$  corresponds to an independent set of  $G_i(N, H)$ , namely the set  $I_R \subset V_R$ corresponding to the edges of I together with the set  $I_B \subset V_B$  corresponding to non-edges of I. Observe that, regarded as a 2-coloured multigraph with edges  $I_R$  and  $I_B$ , I is entire. In general, every subgraph of  $G_i(N, H)$  corresponds to a 2-coloured  $\ell$ -multigraph on vertex set [N], and if any such subgraph C contains an independent set I representing an induced H-free graph as just described, then C is entire.

The graph  $G_i(N, H)$  has very similar properties to those of G(N, K) where  $K = K_{v(H)}^{(\ell)}$  is the complete  $\ell$ -graph on v(H) vertices. In particular, for fixed  $\tau$ , the  $\delta_j$  for  $G_i(N, H)$  differ by only a constant factor from those for G(N, K). Let  $m = m(K) = \left(\binom{v(H)}{\ell} - 1\right)/(v(H) - \ell)$ . Then Lemma 9.3, or the calculation in its proof, shows that  $\delta(G_i(N, H), N^{-1/m}/\epsilon) = O(\epsilon)$ . We are now ready to establish Theorem 2.6.

Proof of Theorem 2.6. We mimic the proof of Theorem 9.2 by applying Corollary 3.6, but this time to the graph  $G_i(N, H)$ . We just noted that  $\delta(G_i(N, H), N^{-1/m}/\epsilon) = O(\epsilon)$ , so we can choose  $\tau = O(N^{-1/m}/\epsilon)$  so that the conditions of Corollary 3.6 are satisfied. The corollary yields a collection  $\mathcal{C} \subset \mathcal{P}(V_R \cup V_B)$ , where each  $C \in \mathcal{C}$  is identified with a 2coloured  $\ell$ -multigraph in the natural way. The properties of  $\mathcal{C}$  claimed in Theorem 2.6 follow directly from those provided by the corollary.

Next we derive Theorem 2.7. To do so, we need a suitable version of Proposition 9.4. Recall the definition of the function  $H_p$  given in §2.4.

**Lemma 9.5** (Supersaturation for induced  $\ell$ -graphs). Let H be an  $\ell$ -graph and let  $0 < \epsilon, p < 1$ . There exists  $N_0$  and  $\eta > 0$  such that if C is an entire 2-coloured  $\ell$ -multigraph on  $N \ge N_0$  vertices containing at most  $\eta N^{v(H)}$  copies of H then  $H_p(C) \ge (h_p(H) - \epsilon) {N \choose \ell}$ .

Proof. Since  $H_p(C) \geq 0$  for all C, we may that assume  $h_p(H) \geq \epsilon$ , the lemma being otherwise trivial. By the definition of  $h_p(H)$ , there exists some  $m \geq 1$  such that every entire 2-coloured  $\ell$ -multigraph D on m vertices with  $H_p(D) < (h_p(H) - \epsilon/2) {m \choose \ell}$  contains a copy of H. Let  $\mathcal{M} \subset [N]^{(m)}$  be the collection of m-sets M such that C[M] contains H. Each edge of C appears in C[M] for  ${n-\ell \choose m-\ell}$  sets  $M \in [N]^{(m)}$ . By considering the contribution of each edge to  $H_p(C)$  and to  $H_p(C[M])$  we see that

$$H_p(C)\binom{N-\ell}{m-\ell} = \sum_{M \in [N]^{(m)}} H_p(C[M]) = \sum_{M \in \mathcal{M}} H_p(C[M]) + \sum_{M \in [N]^{(m)} - \mathcal{M}} H_p(C[M])$$
$$\geq 0 + \left(\binom{N}{m} - |\mathcal{M}|\right) (h_p(H) - \epsilon/2)\binom{m}{\ell}$$

Suppose that C contains at most  $\eta N^{v(H)}$  copies of H for some  $\eta > 0$ . Each copy of H is contained in  $\binom{N-v(H)}{m-v(H)}$  subgraphs  $C[M], M \in \mathcal{M}$ , and hence

$$|\mathcal{M}| \le \eta N^{v(H)} \binom{N - v(H)}{m - v(H)} \le \eta N^m \le \frac{\epsilon}{2h_p(H)} \binom{N}{m},$$

where the last inequality holds if  $\eta$  is small and N is large. Dividing through by  $\binom{N-\ell}{m-\ell}$  in the above expression for  $H_p(C)$ , we obtain

$$H_p(C) \ge (1 - \epsilon/2h_p(H))(h_p(H) - \epsilon/2)\binom{N}{\ell} \ge (h_p(H) - \epsilon)\binom{N}{\ell},$$

as claimed.

Proof of Theorem 2.7. Recall from §2.4 the definition of  $hex_p(H, N)$ . Take an entire 2coloured  $\ell$ -multigraph J satisfying  $H \not\subset J$  and  $H_p(J) = hex_p(H, N) = (h_p(H) + o(1)) {N \choose \ell}$ . The probability that  $G^{(\ell)}(N, p)$  is induced H-free is at least the probability that  $G^{(\ell)}(N, p) \subset J$ , which equals  $2^{-H_p(J)}$ , giving the lower bound in the theorem. Now let  $\epsilon > 0$  and let  $\eta$  be given by Lemma 9.5. Let  $\mathcal{C}$  be the collection of 2-coloured  $\ell$ -multigraphs given by Theorem 2.6 satisfying  $|\mathcal{C}| = 2^{o(N^{\ell})}$  and for every  $C \in \mathcal{C}$ , the number of copies of H in C is at most  $\eta N^{v(H)}$ . Let  $\mathcal{C}' \subset \mathcal{C}$  consist of those  $C \in \mathcal{C}$  that are entire. By Lemma 9.5, for each  $C \in \mathcal{C}'$  we have  $H_p(C) \ge (h_p(H) - \epsilon) {N \choose \ell}$ . Since every induced-H-free graph on vertex set [N] is contained in some  $C \in \mathcal{C}'$ ,

$$\mathbb{P}(G^{(\ell)}(N,p) \text{ is induced-}H\text{-}\text{free}) \leq \sum_{C \in \mathcal{C}'} 2^{-H_p(C)} \leq 2^{-(h_p(H) - \epsilon + o(1))\binom{N}{\ell}}.$$

But  $\epsilon > 0$  was arbitrary and this completes the proof of Theorem 2.7.

#### 10. Sparsity

In this section we prove Theorem 2.12 and related theorems. We remark once again that there are no further applications of a container theorem here; we just use Theorem 2.3 or the slightly more technical Theorem 9.2 together with some straightforward probabilistic arguments. (In the same way, sparse arithmetical results such as Theorem 2.13 and those obtained in [57] follow from a theorem analogous to Theorem 2.3 about solution-free sets.)

Note that the condition  $p \ge cN^{-1/m(H)}$  in Theorem 2.12 is tight up to the value of c. Indeed, if  $p = o(N^{-1/m(H)})$ , it is readily checked that for some subgraph  $H' \subset H$  with m(H') = m(H), the expected number of copies of H' is much less than the number of edges, and removing very few edges will result in an H-free subgraph.

As a further illustration of the paradigm described in §2.6, we prove two other conjectures of Kohayakawa, Luczak and Rödl [35]. The first of these has already been proved, by Conlon and Gowers [12] for strictly balanced graphs and by Samotij [48], following Schacht [59], for all graphs. It states that, for non-bipartite H, not only does every H-free subgraph I of a random graph have at most  $(1 + o(1))p\pi(H) {N \choose 2}$  edges, but in the case that I has close to  $p\pi(H) {N \choose 2}$  edges, it can be made  $(\chi(H) - 1)$ -partite by removing a small number of edges.

**Theorem 10.1.** Let H be a 2-graph with  $\pi(H) > 0$  and let  $0 < \gamma < 1$ . There exist constants  $\epsilon, c > 0$  such that for N sufficiently large and for  $p \ge cN^{-1/m(H)}$ , the following is true. Let  $E_0$  be the event that there exists an H-free subgraph  $I \subset G(N, p)$  with  $e(I) \ge (1 - \frac{1}{\chi(H) - 1} - \epsilon)p\binom{N}{2}$  which cannot be made  $(\chi(H) - 1)$ -partite by removing at most  $\gamma p\binom{N}{2}$  edges. Then  $\mathbb{P}(E_0) \le \exp\{-\epsilon^2 p\binom{N}{2}\}$ .

The dense (p = 1) version of this theorem is the stability theorem of Erdős and Simonovits [16, 17, 61]; indeed this theorem states that *every* sufficiently dense *H*-free graph *I* can be made  $(\chi(H) - 1)$ -partite in the way described. Theorem 10.1 is therefore the assertion that a similar phenomenon holds with high probability in sparser random graphs.

The other conjecture from [35], sometimes known as the KLR conjecture, has a more technical statement. Let G be a graph. For  $U, W \subset V(G)$ , write  $E_G(U, W) \subset E(G)$  for the set of edges of G with one vertex in U and one vertex in W. Let  $e_G(U, W) = |E_G(U, W)|$  and write  $d_G(U, W) = e_G(U, W)/(|U||W|)$  for the edge density. For  $0 < \eta, p \le 1$ , say that the pair (U, W) is  $(\eta, p)$ -regular if for every  $U' \subset U$  with  $|U'| \ge \eta |U|$  and  $W' \subset W$  with  $|W'| \ge \eta |W|$ , the edge density satisfies

$$|d_G(U', W') - d_G(U, W)| \le \eta p.$$

This extends the notion of regularity to sparse graphs of density p.

Let H be a graph on vertex set [h]. In what follows,  $V_1, \ldots, V_h$  is a partition of [N] = [hn], where each part has size  $|V_i| = n$ . Let G be a graph on vertex set [N]. We say that G is  $(H, \eta, p)$ -regular if for every pair  $(V_i, V_j)$  with  $\{i, j\} \in E(H)$ , the bipartite subgraph of Gbetween  $V_i$  and  $V_j$  is  $(\eta, p)$ -regular. A canonical copy of H in G (whether regular or not) is a set of vertices  $v_1, \ldots, v_h$  with  $v_i \in V_i$  such that  $\{v_i, v_j\}$  is an edge of G whenever  $\{i, j\} \in E(H)$ ; we say that such a copy of H is aligned to the partition  $V_1, \ldots, V_h$ . Denote by  $i_H(G)$  the number of canonical copies of H in G. We say that G is H-free if it does not contain any canonical copies of H; that is,  $i_H(G) = 0$ . Finally, denote by G = G(n, M, H)a graph chosen uniformly at random from all h-partite graphs with parts  $V_1, \ldots, V_h$ , having  $e_G(V_i, V_j) = M$  if  $\{i, j\} \in E(H)$  and  $e_G(V_i, V_j) = 0$  otherwise.

We shall be interested in whether G = G(n, M, H) is  $(H, \eta, p)$ -regular, where we shall always take  $p = M/n^2$ . (This may seem like a strict requirement in the definition of regularity. However, the value of p does not matter up to a constant factor, since it is only the value of  $\eta p$  that is used in the definition, and so  $\eta$  may be adjusted appropriately.) In the case when  $M = \Omega(n^2)$ , that is, when the graph G is dense, and in addition Gis  $(H, \eta, p)$ -regular, then the well-known embedding lemma states that G must contain a canonical copy of H. We would like to extend this to the sparse case, when  $p \ll 1$ . In fact, Luczak [37] showed that when p = o(1), then there exist graphs G that are  $(H, \eta, p)$ -regular and are H-free. The KLR conjecture states that although such examples exist, there are very few of them; few enough so that a typical random graph does not contain any such example, and thus with high probability every  $(H, \eta, p)$ -regular subgraph of a random graph contains a canonical copy of H. Even this claim will fail if p is really small, and indeed examples similar to those mentioned earlier show that we must require  $p \gg n^{-1/m(H)}$ .

The counting lemma is a strengthening of the embedding lemma; it says that, for constant p, small  $\eta$  and large n, we have not just  $i_H(G) > 0$  but  $i_H(G) = (1 + o(1))n^{v(H)}p^{e(H)}$ . One could hope to generalize the counting lemma too to the sparse setting. The hypergraph container methods do not seem appropriate for establishing such a precise count, but nonetheless they are enough to establish something weaker, namely that there are very few  $(H, \eta, p)$ -regular graphs with  $o(n^{v(H)}p^{e(H)})$  copies of H.

The next theorem verifies the KLR conjecture, and further gives the analogous result for the weak counting lemma.

**Theorem 10.2.** Let H be a graph and let  $\alpha > 0$ . There exists c > 0, such that for n sufficiently large and  $M \ge cn^{2-1/m(H)}$ , if G = G(n, M, H) is chosen at random, then

$$\mathbb{P}\left( G \text{ is } (H, \frac{1}{c}, \frac{M}{n^2}) \text{-regular and } i_H(G) \leq \frac{1}{c} n^{v(H)} \left(\frac{M}{n^2}\right)^{e(H)} \right) \leq \alpha^M.$$

This theorem with the stronger constraint  $i_H(G) = 0$  is what is often referred to as the KLR conjecture; it was proved for balanced H by Balogh, Morris and Samotij [6]. As mentioned, the weak counting lemma for  $(H, \eta, p)$ -regular subgraphs of a random graph follows from Theorem 10.2 by the union bound over all possible bad subgraphs. Conlon, Gowers, Samotij and Schacht [13] proved this lemma directly (i.e., they showed that the number of H in every  $(H, \eta, p)$ -regular subgraph has the correct order of magnitude with high probability); moreover, for strictly balanced H, they obtained the precise counting lemma (i.e., the number of H is  $(1 + o(1))n^{v(H)}p^{e(H)}$  with high probability). We turn now to the proofs of the theorems. As mentioned, the derivation of the theorems is straightforward once an appropriate container theorem is available, and the arguments here are routine (similar to those in [6]), but we include details for completeness.

We already indicated in §2.6 how to prove Theorem 2.12. Each *H*-free graph *I* contains no more than  $(\pi(H) + o(1)) {N \choose \ell}$  edges, and with high probability  $G^{(\ell)}(N,p)$  contains not much more than  $(\pi(H) + o(1))p {N \choose \ell}$  of these edges. This is not in itself enough to prove Theorem 2.12 because there are too many independent sets. But the argument is valid with containers *C* instead of independent sets *I*, and the theorem then does follow via the union bound, because there are few containers.

Very similar arguments are used to prove Theorems 10.1 and Theorem 10.2. For Theorem 10.1 we show that the containers that matter are close to being  $(\chi(H) - 1)$ -partite, so a randomly chosen subgraph has the same property. For Theorem 10.2 we show that the containers must contain a very sparse subset, from which G = G(n, M, H) is very unlikely to have chosen many edges, and hence is unlikely to be regular. The union bound then finishes the job.

For each application we need the following probabilistic lemma. (Strictly speaking, Theorem 10.2 uses a modification where the hypergeometric distribution is used instead of the binomial.) This lemma is the place where the condition  $T \subset I$  appearing in the container theorem actually matters, and so is important for that reason. It is phrased in a slightly cumbersome way because of the need to cover each of our required applications, but the principle is simple. If a randomly chosen subset  $X \subset V(G)$  contains an unexpectedly large independent set, then X meets some container C(T) in more vertices than expected. This event is unlikely for two independent reasons: it requires both that  $T \subset X$  and that  $X \cap (C(T) - T)$  be large. Both of these contribute to making the overall probability small.

**Lemma 10.3.** Given  $0 < \nu < 1$  and  $s \ge 1$ , there is a constant  $\phi = \phi(\nu, s)$  such that the following holds. Let L be a set, |L| = n, and let  $\mathcal{I} \subset \mathcal{P}(L)$ . Let  $t \ge 1$ , let  $\phi t/n \le p \le 1$  and let  $\nu n/2 \le d \le n$ . Suppose for each  $I \in \mathcal{I}$  there exists both  $T_I = (T_1, \ldots, T_{s'}) \in \mathcal{P}(I)^{s'}$  and  $D = D(T_I) \subset L$ , where  $s' \le s$ ,  $\sum_i |T_i| \le t$  and  $|D(T_I)| \le d$ . Let  $X \subset L$  be a random subset where each element is chosen independently with probability p. Then

$$\mathbb{P}\left(|D(T_I) \cap X| > (1+\nu)pd \text{ for some } I \subset X, I \in \mathcal{I}\right) \le \exp\{-\nu^2 pd/32\}.$$
(9)

*Proof.* Consider  $I \in \mathcal{I}$  and  $T = T_I = (T_1, \ldots, T_{s'})$ . Let  $J(T) = T_1 \cup \cdots \cup T_{s'}$ . Let  $E_T$  be the event that

$$J(T) \subset X$$
 and  $|D(T) \cap X| \ge (1+\nu)pd$ .

The event  $E_T$  is contained in  $F_T \cap G_T$ , where  $F_T$  is the event that  $J(T) \subset X$  and  $G_T$  is the event that  $|(D(T) - J(T)) \cap X| \ge (1 + \nu)pd - |J(T)|$ . Since  $F_T$  and  $G_T$  are independent,  $\mathbb{P}(E_T) \le \mathbb{P}(F_T)\mathbb{P}(G_T)$ . Choose a set  $D' \supset D(T)$  with |D'| = d. Note that  $|J(T)| \le t \le pn/\phi \le 2pd/\phi\nu \le \nu pd/2$  if  $\phi$  is large. Hence, using standard estimates for the binomial random variable Bin(n, p) (e.g., [30, Corollary 2.3]),

$$\mathbb{P}(G_T) = \mathbb{P}(|(D(T) - J(T)) \cap X| \ge (1 + \nu)pd - |J(T)|) \\ \le \mathbb{P}(|D' \cap X| \ge (1 + \nu)pd - |J(T)|) \\ \le \mathbb{P}(|D' \cap X| \ge (1 + \nu/2)pd) = \mathbb{P}(\text{Bin}(d, p) \ge (1 + \nu/2)pd) \\ \le \exp\{-\nu^2 pd/16\}.$$

Note that  $\mathbb{P}(F_T) = p^{|J(T)|}$ . Given some set  $J \subset L$  with |J| = j, there are at most  $2^{sj}$  tuples T such that J(T) = J, because, for each  $i \in J$ , there are at most  $2^s$  ways to specify which of the subsets  $T_1, \ldots, T_{s'}$  contain i. Let  $x = pn/t \ge \phi$ , so  $t \le 2pd/x\nu$ . If  $\phi$  is large we may assume p(n-t) > t, so, summing over the possible sizes of J,

$$\sum_{T} \mathbb{P}(F_T) \le \sum_{j=0}^{t} \binom{n}{j} 2^{sj} p^j \le (t+1) \left(\frac{ne2^s p}{t}\right)^t \le (xe^2 2^s)^t \le (xe^2 2^s)^{\frac{2pd}{x\nu}} \le \exp\{\nu^2 pd/32\}$$

holds if  $\phi$ , and therefore x, is large. If there exists  $I \subset X$ ,  $I \in \mathcal{I}$  with  $|D(T_I) \cap X| \ge (1+\nu)pd$ , then the event  $E_{T_I}$  holds. Hence the probability in (9) is bounded by

$$\sum_{T} \mathbb{P}(F_T) \mathbb{P}(G_T) \le \exp\{\nu^2 p d/32\} \exp\{-\nu^2 p d/16\} \le \exp\{-\nu^2 p d/32\}$$

as claimed.

Proof of Theorem 2.12. Let  $\mathcal{I}$  be the set of H-free  $\ell$ -graphs on vertex set [N]. Let  $\epsilon = \gamma/4$  and  $L = [N]^{(\ell)}$ . For  $I \in \mathcal{I}$ , let  $T = T_I$ , C = C(T) and  $c' = c(H, \epsilon)$  be given by Theorem 2.3. Our aim is to apply Lemma 10.3 with D(T) = C(T) and

$$\nu = \gamma/2, \quad d = (\pi(H) + \epsilon) \binom{N}{\ell}, \quad s = c', \quad t = c' N^{\ell - 1/m(H)}$$

The conditions of Lemma 10.3 then hold with  $n = \binom{N}{\ell}$ , noting that  $d \ge \nu n/2$  and that  $p \ge cN^{-1/m(H)} \ge \phi t/n$  if c is large enough. Finally, note that in (9), each H-free  $\ell$ -graph  $I \in \mathcal{I}$  is contained in  $C(T_I)$  and  $(1 + \nu)pd \le (\pi(H) + \gamma)p\binom{N}{\ell}$ , so the probability in the statement of the theorem is bounded by

$$\exp\{-\nu^2 p d/32\} \le \exp\left\{-\gamma^3 p \binom{N}{\ell}/512\right\},\,$$

completing the proof.

Proof of Theorem 10.1. Notice that  $\pi(H) = 1 - 1/(\chi(H) - 1)$  and  $\chi(H) \ge 3$ . It is a standard exercise, either using the stability arguments of Erdős and Simonovits or using Szemerédi's regularity lemma, that there exists  $\epsilon > 0$  such that if C is a 2-graph on vertex set [N] for N sufficiently large with  $e(C) \ge (1 - \frac{1}{\chi(H) - 1} - 11\epsilon) \binom{N}{2}$  and such that C contains at most  $\epsilon \binom{N}{v(H)}$  copies of H, then there exists a subgraph  $F \subset C$  of size  $e(F) \le (\gamma/2)\binom{N}{2}$  such that C - F is  $(\chi(H) - 1)$ -partite. We may and shall assume that  $\epsilon \le 1/66$  and  $65\epsilon^2 \le \gamma^3$ .

(A word of explanation is included here for those not so familiar with such arguments. In the case that C is H-free, the assertion just made is precisely the stability theorem, as we mentioned after the statement of Theorem 10.1. The stability proof is readily adapted to the present situation where C contains few copies of H. An alternative approach uses Szemerédi's regularity lemma and the original stability theorem, though the constants involved are much larger. In this argument, the graph is partitioned by the regularity lemma, and the reduced graph, whose vertices represent the parts of the partition and whose edges represent regular pairs of positive density, must itself have density at least  $1 - \frac{1}{\chi(H)-1} - 11\epsilon$ . By a counting lemma very like Lemma 10.4 below, the reduced graph then shows the reduced graph is close to  $(\chi(H) - 1)$ -partite. Hence the same holds for C itself.)

The argument is now roughly as follows. An *H*-free subgraph of G(N, p) must lie in some container *C*. If the subgraph has size larger than  $p(1 - \frac{1}{\chi(H)-1} - \epsilon)\binom{N}{2}$  then it is unlikely that *C* has size smaller than  $(1 - \frac{1}{\chi(H)-1} - 11\epsilon)\binom{N}{2}$ . But if *C* has size larger than this then it can be made  $(\chi(H) - 1)$ -partite by removing few edges, and the same will then likely be true of the random subgraph.

Let  $\mathcal{I}$  be the set of *H*-free graphs on vertex set [N]. For  $I \in \mathcal{I}$  let  $T = T_I$ , C = C(T)and  $c' = c(H, \epsilon)$  be given by Theorem 2.3 with  $\epsilon$  as above. Let

$$\mathcal{I}_1 = \left\{ I \in \mathcal{I} : e(C(T_I)) \ge \left(1 - \frac{1}{\chi(H) - 1} - 11\epsilon\right) \binom{N}{2} \right\},\$$
  
$$\mathcal{I}_2 = \mathcal{I} - \mathcal{I}_1.$$

For  $I \in \mathcal{I}_1$  let  $F = F(T_I) \subset C(T_I)$  be as above, so that  $C(T_I) - F(T_I)$  is  $(\chi(H) - 1)$ -partite. Let X = G(N, p). Let  $E_1$  be the event that there exists  $I \subset X$ ,  $I \in \mathcal{I}_1$  such that  $|F(T_I) \cap X| \geq \gamma p {N \choose 2}$ . Let  $E_2$  be the event that there exists  $I \subset X$ ,  $I \in \mathcal{I}_2$  such that  $|C(T_I) \cap X| \geq (1 - \frac{1}{\chi(H) - 1} - \epsilon) p {N \choose 2}$ . Observe that  $E_0 \subset E_1 \cup E_2$ .

The probability of  $E_2$  is bounded by applying Lemma 10.3 to the collection  $\mathcal{I}_2$ , with  $L = [N]^{(2)}$ ,  $n = \binom{N}{2}$ ,  $D(T_I) = C(T_I)$ ,  $\nu = 10\epsilon$ ,  $d = (1 - \frac{1}{\chi(H) - 1} - 11\epsilon)\binom{N}{2} \ge \frac{1}{3}\binom{N}{2}$ , s = c' and  $t = c'N^{2-1/m(H)}$ ; provided  $p \ge cN^{-1/m(H)}$  and c, N are sufficiently large,

$$\mathbb{P}(E_2) \le \exp\left\{-\nu^2 p d/32\right\} \le \exp\left\{-25\epsilon^2 p \binom{N}{2}/24\right\}$$

The probability of  $E_1$  is bounded by applying Lemma 10.3 to the collection  $\mathcal{I}_1$ , with  $D(T_I) = F(T_I), \nu = \gamma, d = (\gamma/2) {N \choose 2}, s = c'$  and  $t = c' N^{2-1/m(H)}$ ; provided  $p \ge c N^{-1/m(H)}$  and c, N are sufficiently large,

$$\mathbb{P}(E_1) \le \exp\left\{-\nu^2 p d/32\right\} = \exp\left\{-p\gamma^3 \binom{N}{2}/64\right\} \le \exp\left\{-65p\epsilon^2 \binom{N}{2}/64\right\}.$$

Since  $\mathbb{P}(E_0) \leq \mathbb{P}(E_1) + \mathbb{P}(E_2)$  and  $pN^2$  is large, this completes the proof of Theorem 10.1.  $\Box$ 

In order to prove Theorem 10.2, we use a slight variation of a standard counting lemma. It says that containers with few canonical copies of H must contain a bipartite subgraph that has a substantial number of vertices but is nevertheless very sparse.

**Lemma 10.4.** Let H be a graph and let  $f : (0,1) \to (0,1)$ . Then there exists  $\eta > 0$  and  $\epsilon > 0$  such that the following is true. Let C be a graph of order N = nh, where h = v(H), whose vertices are partitioned into sets  $V_1, \ldots, V_h$  each of size n. Suppose  $i_H(G) < \epsilon N^h$ . Then there exists  $\gamma \ge \eta$ ,  $\{i, j\} \in E(H)$  and  $A \subset V_i$ ,  $B \subset V_j$ , of size  $|A|, |B| = \gamma n$ , such that  $e(C[A, B]) \le f(\gamma)n^2$ .

*Proof.* Let  $\delta_0 = 1$  and, for  $i = 1, 2, \ldots, h$ , define

$$\epsilon_i = \frac{1}{2h} \prod_{k < i} \delta_k$$
 and  $\delta_i = f(\epsilon_i).$ 

Let  $\eta = \epsilon_h$  and  $\epsilon = \prod_{i=1}^h \epsilon_i$ . The following process generates canonical copies of H with vertices labelled by  $v_1, \ldots, v_h$ . Let  $V_{i,1} = V_i$  for  $i \in [h]$ . For each  $i = 1, \ldots, h$ , do the following.

(1) For each j > i such that  $\{i, j\} \in E(H)$ , let

$$A_{i,j} = \{ v \in V_{i,i} : |V_{j,i} \cap N(v)| < \delta_i |V_{j,i}| \}.$$

- (2) Select  $v_i$  from  $V_{i,i} \setminus \bigcup_{j>i:\{i,j\}\in E(H)} A_{i,j}$ . (If no such vertex exists then stop.)
- (3) For each j > i, let  $V_{j,i+1} \subset V_j$  be a set of size  $\delta_i |V_{j,i}|$ , chosen arbitrarily from  $V_{j,i} \cap N(v_i)$  if  $\{i, j\} \in E(H)$ , and otherwise chosen arbitrarily from  $V_{j,i}$ .

Note that

$$|V_{j,i}| = \delta_{i-1}|V_{j,i-1}| = \delta_{i-1}\delta_{i-2}|V_{j,i-2}| = \dots = \left(\prod_{k < i} \delta_k\right)n = 2h\epsilon_i n = 2\epsilon_i N.$$

Now if  $|A_{i,j}| \leq |V_{i,i}|/2h$  for every  $\{i, j\} \in E(H)$ , then the number of choices for  $v_i$  is at least

$$\left| V_{i,i} \setminus \bigcup_{j > i: \{i,j\} \in E(H)} A_{i,j} \right| \ge |V_{i,i}|/2 \ge \epsilon_i N ,$$

giving at least  $\epsilon N^h$  canonical copies of H, a contradiction. Thus there exists  $\{i, j\} \in E(H)$ with  $|A_{i,j}| \geq |V_{i,i}|/2h = \epsilon_i n$ . Then putting  $\gamma = \epsilon_i \geq \eta$  and taking  $A \subset A_{i,j}$  and  $B \subset V_{j,i}$ of size  $|A|, |B| = \gamma n$  gives a pair of sets with  $e(C[A, B]) \leq |A|\delta_i|V_{j,i}| \leq \delta_i n^2 = f(\gamma)n^2$  as required.

Proof of Theorem 10.2. The argument is broadly this. Each G(n, M, H) with relatively few canonical copies of H must lie in a container C which itself has few copies. Lemma 10.4 states that C has a very sparse bipartite subgraph, from which G(n, M, H) is very unlikely to pick up many edges. But if G(n, M, H) fails to pick up such edges it will fail to be regular. It is a crucial feature of the argument that, in order that  $\alpha$  can be as small as we like, the bipartite subgraph can be made as sparse as we like whilst not being too small. This accounts for the appearance of Lemma 10.4.

Here then are the details. We shall take c to be sufficiently large as necessary. Let N = hn, let  $p = M/n^2$  and let X = G(n, M, H). For ease of notation we shall often identify graphs with their edge sets. Define  $f: (0,1) \to (0,1)$  by  $2f(\gamma) = (\alpha/4)^{2/\gamma^2}$ . Let  $\epsilon, \eta > 0$  be given by Lemma 10.4. Let  $\tilde{G} \subset G(N, H)$  be the set of canonical copies of H (that is, the  $n^h = N^h/h^h$  copies whose vertices are aligned to the partition  $V_1 \cup \cdots \cup V_h$  of [N].) Reduce  $\epsilon$  if necessary so that  $e(\tilde{G}) \ge \epsilon N^h$ . Choose  $\tilde{c} \ge 2$  larger than  $c(H, \epsilon)$  as given by Theorem 9.2, and large enough so that inequality (11) below holds. Set  $q = (\eta^2/h^2\tilde{c}^3)p$ . Certainly  $q \le 1/\tilde{c}$ , and moreover  $q \ge N^{-1/m(H)}$  holds if c is large enough because  $M \ge cn^{2-1/m(H)}$ . Hence we may apply Theorem 9.2 with  $H, \epsilon, q$  and  $\tilde{G}$ . Let  $\mathcal{C}$  be given by the theorem.

Consider the tuples  $T = (T_1, \ldots, T_s)$  described in Theorem 9.2. For each such T, let  $J(T) = T_1 \cup \cdots \cup T_s$ , and define the following probabilistic events:

 $E_T: J(T) \subset X \subset C(T)$  and X is  $(H, \eta, p)$ -regular,  $F_T: J(T) \subset X$ ,  $G_T: X \subset C(T)$  and X is  $(H, \eta, p)$ -regular.

Theorem 9.2 states that if  $i_H(X) < q^{e(H)}N^h$  then there exists  $T = (T_1, \ldots, T_s)$  with  $J(T) \subset X \subset C(T)$ ; thus if in addition X is  $(H, \eta, p)$ -regular then  $E_T$  holds. We may assume that  $1/c \leq \eta$  and also that  $n^h p^{e(H)}/c < q^{e(H)}N^h$ . Therefore, by the union bound, to complete the proof it is enough to show that  $\sum_T \mathbb{P}(E_T) \leq \alpha^M$ . Note that  $E_T = F_T \cap G_T$  and so  $\mathbb{P}(E_T) = \mathbb{P}(F_T)\mathbb{P}(G_T|F_T)$ .

In order to bound  $\mathbb{P}(G_T|F_T)$ , let T be fixed, let J = J(T) and let  $C = C(T) \in \mathcal{C}$ . Theorem 9.2 guarantees that C contains at most  $\epsilon N^{v(H)}$  canonical copies of H. Thus by Lemma 10.4, there exists  $\{i, j\} \in E(H), A \subset V_i, B \subset V_j$ , with  $|A|, |B| = \gamma n$  and  $e(C[A, B]) \leq f(\gamma)n^2$ , where  $\gamma \geq \eta$ . If  $G_T$  holds, then X is  $(H, \eta, p)$ -regular and so

$$|(A \times B) \cap (X - J)| \ge (1 - \eta)p|A||B| - |J| \ge \gamma^2 M/2$$
(10)

(here we assumed, as we may, that  $\eta \leq 1/4$ , and noted that  $|J| \leq \tilde{c}qN^2 = \eta^2 pn^2/\tilde{c}^2 \leq \gamma^2 M/4$ ). However  $|(A \times B) \cap C| \leq f(\gamma)n^2$  and if  $G_T$  holds then  $X \subset C$ , so the probability of (10) is small. Specifically, in generating the random graph  $(X - J) \cap (V_i \times V_j)$  when conditioned on  $J \subset X$ , we are selecting a set of  $M - |J \cap (V_i \times V_j)| \leq M$  edges uniformly from at least  $n^2 - |J \cap (V_i \times V_j)| \geq n^2/2$  possible edges, and for (10) to hold, we must select at least  $\gamma^2 M/2$  edges from a set of at most  $f(\gamma)n^2$  possibilities. This probability is at most

$$\mathbb{P}(G_T|F_T) \le \binom{M}{\gamma^2 M/2} \left(\frac{f(\gamma)n^2}{n^2/2}\right)^{\gamma^2 M/2} \le 2^M (2f(\gamma))^{\gamma^2 M/2} \le (\alpha/2)^M,$$

by the definition of f.

Thus  $\sum_T \mathbb{P}(E_T) \leq \sum_T \mathbb{P}(F_T) \mathbb{P}(G_T | F_T) \leq (\alpha/2)^M \sum_T \mathbb{P}(F_T)$ , and to finish the proof it is enough to show that  $\sum_T \mathbb{P}(F_T) \leq 2^M$ . Now

$$\sum_{T} \mathbb{P}(F_T) = \sum_{T} \mathbb{P}(J(T) \subset X) = \sum_{T} \mathbb{E}\mathbf{1}_{J(T) \subset X} = \mathbb{E}\left|\left\{T : J(T) \subset X\right\}\right|.$$

Since  $T = (T_1, \ldots, T_s)$  where  $s \leq \tilde{c}$ , Lemma 6.1 tells us that  $|\{T : J(T) \subset X\}| \leq \exp\{\tilde{c}\theta|X|(1+\log(1/\theta))\}$  where  $\theta|X|$  is the average size of the  $T_i$ . Now  $|X| = Me(H) = pn^2e(H)$  and  $\theta|X| \leq \tilde{c}qN^2$ . Thus

$$\tilde{c}\theta|X|(1+\log(1/\theta)) \leq \tilde{c}(\tilde{c}qN^2)\left(1+\log\frac{pn^2e(H)}{\tilde{c}qN^2}\right)$$
$$= M\left(\eta^2/\tilde{c}\right)\left(1+\log(e(H)\tilde{c}^2/\eta^2)\right)$$
$$< M\log 2$$
(11)

as required.

#### 11. Optimality

We finish with the proof of Theorem 3.8 from §3.6. The ideas behind the proof, which is a kind of converse to the proof of Theorem 2.12, have already been sketched out but here are the details.

Proof of Theorem 3.8. We may assume that c < 1/4. We may assume that tn is large, since we can choose  $\gamma$  so that  $\gamma tn < 1/2$  for small values of tn, in which case the theorem is immediate. We prove (i) in a way that can readily be adapted for (ii). Put  $\epsilon = \min\{1, (rc/9)^{1/(r-1)}\}$ . Select a subset  $X \subset [n]$  by choosing vertices independently with probability p, where  $p = \epsilon t = \epsilon d^{-1/(r-1)}$ . As mentioned in §3.6, X is likely to be close to independent. To be precise, by standard estimates for the binomial distribution (such as [30, Corollary 2.3]) we have  $\mathbb{P}(|X| \leq (1 - c/3)pn) \leq 2e^{-c^2pn/40} < 1/3$  since  $pn = \epsilon tn$  is large. The expected value of e(G[X]) is  $p^r nd/r$  so  $\mathbb{P}(e(G[X]) > 3p^r nd/r) \leq 1/3$ . Hence, by removing a vertex from each edge of G[X], we see that, with probability at least 1/3, X contains an independent subset I with  $|I| \ge (1 - c/3)pn - 3p^r nd/r \ge (1 - 2c/3)pn$ , where the last inequality holds because  $\epsilon^{r-1} \le rc/9$ . In summary

$$\mathbb{P}(\text{there exists independent } I \subset X \text{ with } |I| \ge (1 - 2c/3)pn) \ge \frac{1}{3}.$$
(12)

There must be some  $C \in \mathcal{C}$  with  $I \subset C$ , so  $\mathbb{P}(\text{there exists } C \in \mathcal{C} \text{ with } |X \cap C| \geq (1 - 2c/3)pn) \geq 1/3$ . This can happen only if  $|\mathcal{C}|$  is large, since for an individual container  $C \in \mathcal{C}$  the event  $|X \cap C| \geq (1 - 2c/3)pn$  is unlikely, because  $|C| \leq (1 - c)n$ . Indeed, choosing C' containing C with |C'| = (1 - c)n, and again using standard estimates (e.g. [30, Corollary 2.3]), we have  $\mathbb{P}(|X \cap C| \geq (1 - 2c/3)pn) \leq \mathbb{P}(|X \cap C'| \geq (1 - 2c/3)pn) \leq 2e^{-c^2pn/40}$ . Therefore  $|\mathcal{C}|2e^{-c^2pn/40} \geq 1/3$  or  $|\mathcal{C}| \geq (1/6)e^{c^2pn/40}$ , which proves (i).

We can sharpen this calculation if C is internally generated and  $C = \{C(T) : T \in \mathcal{T}\}$ . If  $I \subset X$  is an independent set then there is some  $T \in \mathcal{T}$  with  $T \subset I \subset C(T)$ . So inequality (12) implies

$$\mathbb{P}\left(|C(T) \cap X| \ge (1 - 2c/3)pn \text{ for some } T \in \mathcal{T}, T \subset X\right) \ge \frac{1}{3}.$$
(13)

We next show that small sets T cannot make much contribution to (13). Let  $\mathcal{T}^* = \{T \in \mathcal{T} : |T| \leq \gamma tn\}$ . Apply Lemma 10.3 with L = [n], s = 1, t in the lemma equal to  $\gamma tn$  here, D(T) = C(T) and d = (1 - c)n. Let  $\nu = c/3$ , so  $(1 + \nu)pd < (1 - 2c/3)pn$ . Choose  $\gamma$  so that  $\phi\gamma \leq \epsilon$ . If we take  $\mathcal{I}$  in the lemma to be those independent sets for which there exists some  $T \in \mathcal{T}^*$  with  $T \subset I \subset C(T)$ , then the conditions of the lemma are satisfied, and so

$$\mathbb{P}\left(|C(T) \cap X| \ge (1 - 2c/3)pn \text{ for some } T \in \mathcal{T}^*, T \subset X\right) \le \exp\{-\nu^2 pd/32\} \le \frac{1}{6}$$

because tn is large.

It follows now from inequality (13) that

$$\mathbb{P}\left(|C(T) \cap X| \ge (1 - 2c/3)pn \text{ for some } T \in \mathcal{T} \setminus \mathcal{T}^*, T \subset X\right) \ge \frac{1}{6}.$$

In particular  $\mathbb{P}(T \subset X \text{ for some } T \in \mathcal{T} \setminus \mathcal{T}^*) \geq 1/6$ . But  $\mathbb{P}(T \subset X) = p^{|T|}$  and so

$$\mathbb{P}(T \subset X \text{ for some } T \in \mathcal{T} \setminus \mathcal{T}^*) \leq \sum_{T \notin \mathcal{T}^*} p^{|T|} \leq |\mathcal{T} \setminus \mathcal{T}^*| p^{\gamma tr}$$

by definition of  $\mathcal{T}^*$ . Hence  $|\mathcal{T} \setminus \mathcal{T}^*| p^{\gamma tn} \ge 1/6$ . Therefore  $\mathcal{T} \setminus \mathcal{T}^* \neq \emptyset$ , which proves the first part of (ii), and moreover  $|\mathcal{T}| \ge (1/6)(1/p)^{\gamma tn} \ge e^{\gamma' nt \log(1/t)}$  for some constant  $\gamma'$ , implying the second part of (ii).

Finally, suppose that G is vertex and edge transitive and that  $\delta(G, \tau) \geq 1$ . Recalling Definition 3.2, the vertex transitivity of G means that there are numbers  $D_j$ ,  $2 \leq j \leq r$ , such that  $d^{(j)}(v) = D_j$  for all  $v \in [n]$ . Let  $F_j = \{\sigma \subset [n] : |\sigma| = j, d(\sigma) = D_j\}$  and let  $G_j$ be the *j*-graph with edge set  $F_j$ . By the edge transitivity of G, every edge of G includes a member of  $F_j$ . In particular, every subset  $I \subset [n]$  that is independent in  $G_j$  is independent in G, and so C is a collection of containers for  $G_j$  too.

Let  $d_j$  be the average degree of  $G_j$ . Then  $D_j n d_j / j = D_j e(G_j) = D_j |F_j| = \sum_{\sigma \in F_j} d(\sigma) \leq {r \choose j} e(G) = {r \choose j} n d/r$ . Thus  $d_j \leq {r-1 \choose j-1} (d/D_j)$ . Recalling again Definition 3.2, we have  $\delta_j \tau^{j-1} n d = n D_j$ . Now  $\delta(G, \tau) \geq 1$ , so for some j we have  $2^{\binom{r}{2}-1-\binom{j-1}{2}} \delta_j > 1/r$ . This certainly implies  $r2^{\binom{r}{2}} \delta_j > 1$ , and so  $r2^{\binom{r}{2}} D_j > \tau^{j-1} d$ . Hence, for this value of j, we have

 $d_j \leq {\binom{r-1}{j-1}} (d/D_j) \leq {\binom{r-1}{j-1}} r 2^{\binom{r}{2}} \tau^{1-j}$ . In particular  $d_j^{-1/(j-1)} \geq \gamma' \tau$  for some positive number  $\gamma'$  depending on r.

We noted previously that C is a collection of containers for  $G_j$ . Hence, for every j, properties (i) and (ii) apply with  $t = d_j^{-1/(j-1)}$ . But there is some j with  $d_j^{-1/(j-1)} \ge \gamma' \tau$ , and so (i) and (ii) apply with  $t = \tau$ , once the value of  $\gamma$  has been suitably adjusted.  $\Box$ 

## 12. Postscript

We are extremely grateful to the referees of this paper, who expended a great deal of care and thought on their work, and made many valuable suggestions. In particular, their conscientious reading showed up a subtle error in the original version, relating to the online property. The error arose due to the use of the condition  $d_s(\sigma) \geq 2^s \tau d_{s+1}(\sigma)$  to define entry of non-singleton sets into  $\Gamma_s$ . The problem with this definition is that it is a condition relative to what happens in  $P_{s+1}$ , unlike the absolute condition  $d_s(v) \geq \tau^{r-s} d(v)$ used for the entry of vertices. This relativity breaks the online property, which is why in the present paper the online property is claimed only for simple graphs (for which non-singleton sets in  $\Gamma_s$  are irrelevant).

As a consequence of this, we were prompted to revisit our earlier ideas for entry conditions. It turns out to be possible to specify an absolute condition for entry of  $\sigma$  into  $\Gamma_s$ , which nevertheless implies the inequality  $d_s(\sigma) \geq 2^s \tau d_{s+1}(\sigma)$ . This gives rise to a slightly different algorithm, but one which still yields all the theorems of the present paper.

However the modified algorithm has many advantages. It needs only one pass through the vertex set, constructing the hypergraphs  $P_s$  simultaneously, rather than the r-1consecutive passes of the present method. The operation of the algorithm is thus more transparent. Moreover only one set T is produced in prune mode, rather than the tuple  $(T_{r-1}, \ldots, T_0)$  described here. The use of an absolute entry condition makes much clearer how the co-degree function  $\delta(G, \tau)$  arises, and how slightly different functions could be used at the expense of somewhat larger containers. Finally, the single pass approach yields the online property immediately for all hypergraphs, not just simple ones; in particular Theorem 3.7 holds for all r-graphs, giving more general colouring results. We hope to describe the modified algorithm elsewhere [57].

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DEPARTMENT OF PURE MATHEMATICS AND MATHEMATICAL STATISTICS, CENTRE FOR MATHEMATICAL SCIENCES, WILBERFORCE ROAD, CAMBRIDGE CB3 0WB, UK

E-mail address: d.saxton@dpmms.cam.ac.uk

E-mail address: a.g.thomason@dpmms.cam.ac.uk