

# A NEW PROOF OF THE DENSITY HALES-JEWETT THEOREM

D. H. J. POLYMATH

ABSTRACT. The Hales–Jewett theorem asserts that for every  $r$  and every  $k$  there exists  $n$  such that every  $r$ -colouring of the  $n$ -dimensional grid  $\{1, \dots, k\}^n$  contains a combinatorial line. This result is a generalization of van der Waerden’s theorem, and it is one of the fundamental results of Ramsey theory. The theorem of van der Waerden has a famous density version, conjectured by Erdős and Turán in 1936, proved by Szemerédi in 1975, and given a different proof by Furstenberg in 1977. The Hales–Jewett theorem has a density version as well, proved by Furstenberg and Katznelson in 1991 by means of a significant extension of the ergodic techniques that had been pioneered by Furstenberg in his proof of Szemerédi’s theorem. In this paper, we give the first elementary proof of the theorem of Furstenberg and Katznelson, and the first to provide a quantitative bound on how large  $n$  needs to be. In particular, we show that a subset of  $\{1, 2, 3\}^n$  of density  $\delta$  contains a combinatorial line if  $n$  is at least as big as a tower of 2s of height  $O(1/\delta^2)$ . Our proof is surprisingly simple: indeed, it gives arguably the simplest known proof of Szemerédi’s theorem.

## 1. INTRODUCTION

**1.1. Statement of our main result.** The purpose of this paper is to give the first elementary proof of the density Hales–Jewett theorem. This theorem, first proved by Furstenberg and Katznelson [FK89, FK91], has the same relation to the Hales–Jewett theorem [HJ63] as Szemerédi’s theorem [Sze75] has to van der Waerden’s theorem [vdW27]. Before we go any further, let us state all four theorems. We shall use the notation  $[k]$  to stand for the set  $\{1, 2, \dots, k\}$ . If  $X$  is a set and  $r$  is a positive integer, then an  $r$ -colouring of  $X$  will mean a function  $\kappa: X \rightarrow [r]$ . A subset  $Y$  of  $X$  is called *monochromatic* if  $\kappa(y)$  is the same for every  $y \in Y$ .

We begin with van der Waerden’s theorem.

**Theorem 1.1.** *For every pair of positive integers  $k$  and  $r$  there exists  $N$  such that for every  $r$ -colouring of  $[N]$  there is a monochromatic arithmetic progression of length  $k$ .*

Szemerédi’s theorem is the *density version* of van der Waerden’s theorem. That is, it says that in van der Waerden’s theorem one can always find an arithmetic progression in any colour class that is used reasonably often.

**Theorem 1.2.** *For every positive integer  $k$  and every  $\delta > 0$  there exists  $N$  such that every subset  $A \subseteq [N]$  of size at least  $\delta N$  contains an arithmetic progression of length  $k$ .*

The reason it is called a density version is that we think of  $|A|/N$  as the density of  $A$  inside  $[N]$ , so the condition on  $A$  is that it has density at least  $\delta$ .

To state the Hales–Jewett theorem, we need a little more terminology. The theorem is concerned with subsets of  $[k]^n$ , elements of which we refer to as *points* (or *strings*). Instead of looking for arithmetic progressions, the Hales–Jewett theorem looks for structures known as *combinatorial lines*. There are many equivalent ways of defining these, of which one is the following. Let  $[n]$  be partitioned into sets  $X_1, \dots, X_k, W$  in such a way that  $W$  is non-empty. Then take the set of all points  $x$  such that  $x_i = j$  whenever  $j \leq k$  and  $i \in X_j$ , and  $x_i$  takes the same value for every  $i \in W$ . The only choice we have in specifying such an  $x$  is the value we assign to the coordinates  $x_i$  with  $i \in W$ , so each line contains  $k$  points.

Here is a simple example of a combinatorial line when  $k = 3$  and  $n = 8$ :

$$\{(1, 3, 1, 2, 2, 1, 1, 2), (2, 3, 2, 2, 2, 2, 1, 2), (3, 3, 3, 2, 2, 3, 1, 2)\}$$

In this case the sets  $X_1, X_2, X_3$  and  $W$  are  $\{7\}$ ,  $\{4, 5, 8\}$ ,  $\{2\}$ , and  $\{1, 3, 6\}$ , respectively.

The coordinates in  $X_1 \cup \dots \cup X_k$  are called the *fixed coordinates* of the line, and the coordinates in  $W$  are the *variable coordinates* or *wildcards*.

Another way of thinking of a line is as an element of the set  $([k] \cup \{*\})^n$ , where at least one coordinate takes the wildcard value  $*$ . To obtain the  $k$  points in the line, one lets  $j$  run from 1 to  $k$  and sets all the wildcards equal to  $j$ . For instance, in this notation the line above is

$$(*, 3, *, 2, 2, *, 1, 2).$$

With both these ways of thinking of combinatorial lines, it is clear that there is a close relationship between lines in  $[k]^n$  and points in  $[k+1]^n$ . Indeed, if one allows “degenerate lines” in which the wildcard sets are empty then there is an obvious one-to-one correspondence between the two sets. This will be very important to us later.

We are now ready to state the Hales–Jewett theorem.

**Theorem 1.3.** *For every pair of positive integers  $k$  and  $r$  there exists a positive number  $HJ(k, r)$  such that for every  $n \geq HJ(k, r)$  and every  $r$ -colouring of the set  $[k]^n$  there is a monochromatic combinatorial line.*

As with van der Waerden’s theorem, we may consider the density version of the Hales–Jewett theorem, where the density of  $A \subseteq [k]^n$  is  $|A|/k^n$ . The following theorem was first proved by Furstenberg and Katznelson [FK91].

**Theorem 1.4.** *For every positive integer  $k$  and every real number  $\delta > 0$  there exists a positive integer  $DHJ(k, \delta)$  such that if  $n \geq DHJ(k, \delta)$  and  $A$  is any subset of  $[k]^n$  of density at least  $\delta$ , then  $A$  contains a combinatorial line.*

We sometimes write “ $DHJ_k$ ” to mean the  $k$  case of this theorem. The first nontrivial case,  $DHJ_2$ , is a weak version of Sperner’s theorem [Spe28]; we discuss this further in Section 2. We also remark that the Hales–Jewett theorem easily implies van der Waerden’s theorem, and likewise for the density versions. To see this, temporarily interpret  $[m]$  as  $\{0, 1, \dots, m-1\}$  rather than  $\{1, 2, \dots, m\}$ , and identify integers in  $[N]$  with their base- $k$  representation in  $[k]^n$ . It is then easy to see that a combinatorial line in  $[k]^n$  corresponds to an arithmetic progression of length  $k$  in  $[N]$ : if the wildcard set of the line is  $S$ , then the common difference of the progression is  $\sum_{i \in S} k^{n-i}$ . However, only very few arithmetic progressions of length  $k$  in  $[N]$  arise in this way, so finding combinatorial lines is strictly harder than finding arithmetic progressions. (Further evidence for this is that several

other results are easy consequences of the Hales–Jewett theorem and its density version: in particular, it is an exercise to deduce the multidimensional Szemerédi theorem from DHJ.)

In this paper, we give a new, elementary proof of the density Hales–Jewett theorem, very different from that of Furstenberg and Katznelson (though the discovery of one part of the argument, sketched in §5.4, was in part inspired by ergodic methods). Our proof gives rise to the first known quantitative bounds for the theorem. Define the *tower function*  $T(n)$  inductively by taking  $T(1) = 2$  and  $T(n) = 2^{T(n-1)}$  (so for instance  $T(4) = 2^{2^{2^2}} = 65536$ ). More generally, define (not quite standardly) the  $k$ th function  $A_k$  in the Ackermann hierarchy by setting  $A_k(1) = 2$  and  $A_k(n) = A_{k-1}(A_k(n-1))$ , with  $A_1(n) = 2n$ . Thus, the  $k$ th function is obtained by iterating the  $(k-1)$ st function, so  $A_2(n) = 2^n$  and  $A_3(n) = T(n)$ .

**Theorem 1.5.** *In the density Hales–Jewett theorem, one may take  $DHJ_3\delta = T(O(1/\delta^2))$ . For  $k \geq 4$ , the bound  $DHJ_k\delta$  we achieve is broadly comparable to the function  $A_k(1/\delta)$ .*

By “broadly comparable” we mean something like that it is much nearer to  $A_k(1/\delta)$  than to  $A_{k+1}(1/\delta)$ . In fact, the bound we obtain is something like  $A_k(A_{k-1}(1/\delta))$ . (To give an idea, if we were to apply a composition of this kind to the function  $A_{k-1}(n) = 2^n$ , then  $A_k(n)$  would be a tower of height  $n$ , whereas  $A_k(A_{k-1}(n))$  would be a tower of height  $2^n$ .)

Another way of phrasing our result is in terms of the number  $c_{n,3}$ , the cardinality of the largest subset of  $[3]^n$  without a combinatorial line. Theorem 1.5 states that  $c_{n,3}/3^n \leq O(1/\sqrt{\log^* n})$ . The only known lower bounds appear in a parallel paper to this one that is by an overlapping set of authors [Pol09]: in that paper it is shown that  $c_{n,3} = 2, 6, 18, 52, 150, 450$  for  $n = 1, 2, 3, 4, 5, 6$ , and for large  $n$  that  $c_{n,3}/3^n \geq \exp(-O(\sqrt{\log n}))$ . Generalizing to  $DHJ_k$ , the authors show that  $c_{n,k}/k^n \geq \exp(-O(\log n)^{1/\lceil \log_2 k \rceil})$ , using ideas from recent work on the construction of Behrend [Beh46].

**1.2. The motivation for finding a new proof.** Why is it interesting to give a new proof of the density Hales–Jewett theorem? There are two main reasons. The first is connected with the history of results and techniques in this area. One of the main benefits of Furstenberg’s proof of Szemerédi’s theorem was that it introduced a technique—ergodic methods—that could be developed in many directions, which did not seem to be the case with Szemerédi’s proof. As a result, several far-reaching generalizations of Szemerédi’s theorem were proved [BL96, FK78, Fur85, FK91], and for a long time nobody could prove them in any other way than by using Furstenberg’s methods. In the last few years that has changed, and a programme has developed to find new and finitary proofs of the results that were previously known only by infinitary ergodic methods; see, e.g., [RS04, NRS06, RS06, RS07b, RS07a, Gow06, Gow07, Tao06, Tao07]. Giving a non-ergodic proof of the density Hales–Jewett theorem was seen as a key goal for this programme, especially since Furstenberg and Katznelson’s ergodic proof seemed significantly harder than the ergodic proof of Szemerédi’s theorem. Having given a purely finitary proof, we are able to obtain explicit bounds for how large  $n$  needs to be as a function of  $\delta$  and  $k$  in the density Hales–Jewett theorem. Such bounds could not be obtained via the ergodic methods even in principle, since these proofs rely on the Axiom of Choice. Admittedly, our explicit bounds are not particularly good: we start with a tower-type dependence for  $k = 3$  and go up a level of the Ackermann hierarchy each time we go from  $k$  to  $k + 1$ . However, they are in line with several other bounds in the area. For example, the best known bounds for

the multidimensional Szemerédi theorem [Gow07, NRS06] (which is an easy consequence of DHJ) are also of this type.

A second reason that a new proof of the density Hales–Jewett theorem is interesting is that it immediately implies Szemerédi’s theorem, and finding a new proof of Szemerédi’s theorem seems always to be illuminating—or at least this has been the case for the four main approaches discovered so far (combinatorial [Sze75], ergodic [Fur77, FKO82], Fourier [Gow01], hypergraph removal [Gow06, Gow07, RS04, NRS06]). Surprisingly, in view of the fact that DHJ is considerably more general than Szemerédi’s theorem and the ergodic-theory proof of DHJ is considerably more complicated than the ergodic-theory proof of Szemerédi’s theorem, the new proof we have discovered gives arguably the simplest proof yet known of Szemerédi’s theorem. It seems that by looking at a more general problem we have removed some of the difficulty. Related to this is another surprise. We started out by trying to prove the first difficult case of the theorem,  $\text{DHJ}_3$ . The experience of all four of the earlier proofs of Szemerédi’s theorem has been that interesting ideas are needed to prove results about progressions of length 3, but significant extra difficulties arise when one tries to generalize an argument from the length-3 case to the general case. Unexpectedly, it turned out that once we had proved the case  $k = 3$  of the density Hales–Jewett theorem, it was straightforward to generalize the argument to the  $k \geq 4$  cases. We do not fully understand why our proof should be different in this respect, but it is perhaps a sign that the density Hales–Jewett theorem is at a “natural level of generality”.

One might ask, if this is the case, why the proof of Furstenberg and Katznelson seems to be *more* complicated than the ergodic-theoretic proofs of Szemerédi’s theorem and its multidimensional version. An explanation for this discrepancy is that our proof appears to be genuinely different from theirs (that is, not just a translation of their proof into a more elementary language). The clearest sign of this is that they use Carlson’s theorem, a powerful result in Ramsey theory, in an essential way, whereas we have no need of any colouring results in our argument (unless you count the occasional use of the pigeonhole principle).

Before we start working towards the proof of the theorem, we would like briefly to mention that it was proved in a rather unusual “open source” way, which is why it is being published under a pseudonym. The work was carried out by several researchers, who wrote their thoughts, as they had them, in the form of blog comments at <http://gowers.wordpress.com>. Anybody who wanted to could participate, and at all stages of the process the comments were fully open to anybody who was interested. (Indeed, taking some inspiration from a few of these blog comments, Austin provided *another* new (ergodic) proof of the density Hales–Jewett theorem [Aus09].) This open process was in complete contrast to the usual way that results are proved in private and presented in a finished form. The blog comments are still available, so although this paper is a polished account of the  $\text{DHJ}_k$  argument, it is possible to read a record of the entire thought process that led to the proof. The constructions of new *lower* bounds for the  $\text{DHJ}_k$  problem, mentioned in Section 1.1, are being published by a partially overlapping set of researchers [Pol09]. The participants in the project also created a wiki, <http://michaelnielsen.org/polymath1/>, which contains sketches of the arguments, links to the blog comments, and a great deal of related material.

**1.3. Combinatorial subspaces and multidimensional DHJ.** We know from the density Hales–Jewett theorem that dense subsets of  $[k]^n$  contain combinatorial lines. It is natural to wonder whether there is a higher-dimensional version of this result, in which

one finds  $d$ -dimensional subspaces. Such a result does indeed exist, and is a straightforward consequence of DHJ, as was observed by Furstenberg and Katznelson. Since we shall need this extension, we briefly define the relevant concepts and give the proof.

A  $d$ -dimensional combinatorial subspace is just like a combinatorial line except that there are  $d$  wildcard sets instead of just one. In other words we partition the ground set  $[n]$  into  $k + d$  sets  $X_1, \dots, X_k, W_1, \dots, W_d$  such that  $W_1, \dots, W_d$  are non-empty, and the subspace consists of all sequences  $x$  such that  $x_i = j$  whenever  $i \in X_j$  and  $x$  is constant on each set  $W_r$ . There is an obvious isomorphism between  $[k]^d$  and any  $d$ -dimensional combinatorial subspace: the sequence  $z = (z_1, \dots, z_d)$  is sent to the sequence  $x$  such that  $x_i = j$  whenever  $i \in X_j$  and  $x_i = z_r$  whenever  $x \in W_r$ .

Note that there is an obvious injection from the set of all  $d$ -dimensional combinatorial subspaces of  $[k]^n$  to  $[k + d]^n$  (which becomes a bijection if one allows the subspaces to be degenerate).

The multidimensional density Hales-Jewett theorem is the following.

**Theorem 1.6.** *For every  $\delta > 0$  and every pair of integers  $k$  and  $d$  there exists a positive integer  $MDHJ(k, d, \delta)$  such that, for every  $n \geq MDHJ(k, d, \delta)$  and every subset  $A \subset [k]^n$ ,  $A$  contains a  $d$ -dimensional combinatorial subspace of  $[k]^n$ .*

We shall refer to this theorem as MDHJ, and for each  $k$  we shall refer to the result for that  $k$  as  $MDHJ_k$ .

**Proposition 1.7.** *For every  $k$ ,  $MDHJ_k$  follows from  $DHJ_k$ .*

*Proof.* We prove the result by induction on  $d$ . Suppose we know  $MDHJ_k$  for dimension  $d - 1$ , and let  $A \subseteq [k]^n$  have density at least  $\delta$ . Let  $m = MDHJ(k, d - 1, \delta/2)$ , and write a typical string  $z \in [k]^n$  as  $(x, y)$ , where  $x \in [k]^m$  and  $y \in [k]^{n-m}$ . Call a string  $y \in [k]^{n-m}$  “good if  $A_y = \{x \in [k]^m : (x, y) \in A\}$  has density at least  $\delta/2$  within  $[k]^m$ . Let  $G \subseteq [k]^{n-m}$  be the set of good  $y$ ’s. Then the density of  $G$  within  $[k]^{n-m}$  must be at least  $\delta/2$ , or  $A$  could not have density at least  $\delta$  in  $[k]^n$ .

By induction, for any good  $y$  the set  $A_y$  contains a  $(d - 1)$ -dimensional combinatorial subspace. There are at most  $M = (k + d - 1)^m$  such subspaces, because of the injection mentioned above. Therefore, there must be some subspace  $\sigma \subseteq [k]^m$  such that the set

$$G_\sigma = \{y \in [k]^{n-m} : (x, y) \in A \ \forall x \in \sigma\}$$

has density at least  $(\delta/2)/M$  within  $[k]^{n-m}$ . Provided that  $n \geq m + DHJ(k, \delta/2M)$ , we may conclude from  $DHJ_k$  that  $G_\sigma$  contains a combinatorial line,  $\lambda$ . Then  $\sigma \times \lambda$  is the desired  $d$ -dimensional subspace of  $[k]^n$  that is contained in  $A$ .  $\square$

Because we have to iterate  $DHJ_k$  with rapidly decreasing densities in order to obtain this result, the bound that we get from it is very bad indeed: it is this that causes the Ackermann-type dependence on  $k$  in our main theorem.

**1.4. Density-increment strategies.** Very briefly, our proof of  $DHJ_k$  follows a *density-increment strategy*, a technique that was pioneered by Roth [Rot53] in his proof of the  $k = 3$  case of Szemerédi’s theorem. There are now many such proofs in the literature, of which most have the following form. One would like to prove that every dense subset  $A$  of a mathematical structure  $S$  (such as an arithmetic progression or the set  $[k]^n$ ) contains a subset  $X$  of a certain type (such as a subprogression of length  $k$  or a combinatorial line). It is usually hard to show this in one step, so instead one proves that if  $A$  has density  $\delta$  in  $S$

and does *not* contain a subset of the desired kind, then  $S$  has a substructure  $S'$  such that the density of  $A$  inside  $S'$  is at least  $\delta + c$ , where  $c$  is some positive constant that depends only on  $\delta$ . This is the density increment. If  $S'$  is of a similar nature to  $S$ , then one can iterate this argument, and if  $S$  is large enough, then one can continue iterating until the density exceeds 1 and one has a contradiction, from which one deduces that  $A$  must after all contain a subset  $X$  of the desired kind.

Even getting directly from  $S$  to a density increment on a substructure  $S'$  in one step is usually too hard, so typically there is an intermediate stage. First, one finds a set  $T$  that is in some sense “simple” such that the density of  $A$  inside  $T$  is at least  $\delta + c$ . Then one proves that “simple” sets  $T$  can be partitioned into substructures  $S_1, \dots, S_N$  and uses an averaging argument to show that the density of  $A$  inside some  $S_i$  is also at least  $\delta + c$ . There are also variants of this: for instance, it is enough to find subsets  $S_1, \dots, S_N$  of  $T$  such that every element of  $T$  is in the same number of  $S_i$ , or even in approximately the same number of  $S_i$ .

A few proofs that have this basic structure are Roth’s proof itself (where the intermediate structure is a mod- $N$  arithmetic progression, which can be partitioned into genuine arithmetic progressions), Gowers’s proof of Szemerédi’s theorem [Gow01], and an argument of Shkredov [Shk06a, Shk06b] that gives strong bounds for the “corners problem”, a result that we shall discuss in detail in Section 4.

## 2. SPERNER’S THEOREM AND ITS MULTIDIMENSIONAL VERSION

The case  $k = 2$  of the density Hales-Jewett theorem is equivalent to the following statement: for every  $\delta > 0$  there exists  $n$  such that if  $\mathcal{A}$  is a collection of at least  $\delta 2^n$  subsets of  $[n]$  then there exist distinct sets  $A, B \in \mathcal{A}$  such that  $A \subset B$ . The equivalence is easily seen if one looks at the characteristic functions of the sets, in which case one sees that a pair  $(A, B)$  with  $A \subset B$  corresponds to a combinatorial line in  $\{0, 1\}^n$ .

Exact bounds are known for this theorem. The nicest proof is the following one, which will have a considerable influence on our later proofs. Recall that an *antichain* is a collection of sets such that no set in the collection is a proper subset of any other.

**Theorem 2.1.** *Let  $n$  be a positive integer and let  $\mathcal{A}$  be an antichain of subsets of  $[n]$ . Then  $|\mathcal{A}| \leq \binom{n}{\lfloor n/2 \rfloor}$ .*

*Proof.* Consider the following way of choosing a random subset of  $[n]$ . One chooses a random permutation  $\pi$  of  $[n]$  and a random integer  $m \in \{0, 1, \dots, n\}$  and takes the set  $A = \{\pi(1), \dots, \pi(m)\}$ . Since  $\mathcal{A}$  is an antichain, for each  $\pi$  there is at most one  $m$  such that the resulting set belongs to  $\mathcal{A}$ . Thus, the probability of choosing a set in  $\mathcal{A}$  is at most  $1/(n+1)$ .

Now the probability of choosing a particular set  $A$  of size  $m$  is  $(n+1)^{-1} \binom{n}{m}^{-1}$ . Therefore, if we want  $\mathcal{A}$  to be as large as possible but for the probability of choosing a set in  $\mathcal{A}$  to be at most  $(n+1)^{-1}$ , then we must choose  $\mathcal{A}$  to consist of sets of size  $m$  such that  $\binom{n}{m}$  is maximized. It follows that we cannot choose more than  $\binom{n}{\lfloor n/2 \rfloor}$  sets, as claimed.  $\square$

We shall also need a multidimensional version of Sperner’s theorem. This time we are trying to maximize the size of  $\mathcal{A}$  subject to the condition that it is not possible to find a  $d$ -dimensional combinatorial subspace, which in set-theoretic terms means a collection of disjoint non-empty sets  $A, A_1, \dots, A_d$  such that  $A \cup \bigcup_{i \in E} A_i \in \mathcal{A}$  for every  $E \subset \{1, 2, \dots, d\}$ .

The result we need was proved by Gunderson, Rödl and Sidorenko. However, for the convenience of the reader we give a proof here, which is somewhat simpler than theirs and gives a slightly better bound. (This improvement has an imperceptible effect on our bound for DHJ<sub>3</sub> though.)

We begin with an easy and standard lemma. As usual, if  $X$  is a finite set and  $Y$  is a subset of  $X$ , we write  $\mu(Y)$  for  $|Y|/|X|$ .

**Lemma 2.2.** *Let  $X$  be a finite set and let  $X_\gamma$  be a random subset of  $X$ , where  $\gamma$  is an element of a probability space  $\Gamma$ . Suppose that  $\mathbb{E}_\gamma \mu(X_\gamma) = \delta$ . Now let  $\gamma$  and  $\gamma'$  be chosen independently from  $\Gamma$ . Then  $\mathbb{E}_{\gamma, \gamma'} \mu(X_\gamma \cap X_{\gamma'}) \geq \delta^2$ .*

*Proof.* Let  $\xi_\gamma$  be the characteristic function of  $X_\gamma$ . Then

$$\begin{aligned} \delta^2 &= (\mathbb{E}_\gamma \mu(X_\gamma))^2 \\ &= (\mathbb{E}_\gamma \mathbb{E}_x \xi_\gamma(x))^2 \\ &\leq \mathbb{E}_x (\mathbb{E}_\gamma \xi_\gamma(x))^2 \\ &= \mathbb{E}_x \mathbb{E}_{\gamma, \gamma'} \xi_\gamma(x) \xi_{\gamma'}(x) \\ &= \mathbb{E}_{\gamma, \gamma'} \mu(X_\gamma \cap X_{\gamma'}). \end{aligned}$$

The inequality above is Cauchy-Schwarz. The result follows.  $\square$

**Theorem 2.3.** *Let  $\mathcal{A}$  be a collection of subsets of  $[n]$  that contains no  $d$ -dimensional combinatorial subspace. Then the density of  $\mathcal{A}$  is at most  $(25/n)^{1/2^d}$ .*

*Proof.* Let  $\delta$  be the density of  $\mathcal{A}$ . (That is,  $\mathcal{A}$  has cardinality  $\delta 2^n$ .) For  $i = 1, 2, \dots, d-1$  let  $n_i = \lfloor n/4^{d-i} \rfloor$  and let  $n_d = n - (n_1 + \dots + n_{d-1})$ . Note that  $n_d \geq (2/3)n$ .

Let us partition  $[n]$  into sets  $J_1 \cup \dots \cup J_{d-1} \cup E$  with  $|J_i| = \lfloor n/4^{d-i} \rfloor$ . Note that  $|E| \geq (2/3)n$ .

Now consider the following way of choosing a random subset  $A$  of  $[n]$ . First we choose a random permutation  $\pi$  of  $[n]$ . Then we choose a random integer  $s$  according to the binomial distribution with parameters  $n_1$  and  $1/2$ . Next, we let  $B$  be a random subset of  $\{\pi(n_1 + 1), \dots, \pi(n)\}$ . Finally, we let  $A$  be the set  $\{\pi(1), \dots, \pi(s)\} \cup B$ . The resulting distribution on  $A$  is uniform, as can be seen by conditioning on the set  $\{\pi(1), \dots, \pi(n_1)\}$ .

Let us write  $A_{\pi, s}$  for the set  $\{\pi(1), \dots, \pi(s)\}$  and  $X_{\pi, s}$  for the set of all  $B \subset \{\pi(n_1 + 1), \dots, \pi(n)\}$  such that  $A_{\pi, s} \cup B \in \mathcal{A}$ . Then the average density of  $X_{\pi, s}$  (in the set of all subsets of  $\{\pi(n_1 + 1), \dots, \pi(n)\}$ ) is  $\delta$ . Therefore, by Lemma 2.2, if we first choose  $\pi$  randomly and then choose  $s$  and  $t$  independently at random from the binomial distribution as we did for  $s$  above, then the average density of  $X_{\pi, s} \cap X_{\pi, t}$  is at least  $\delta^2$ .

We would like  $s$  and  $t$  to be distinct. The probability that  $s = t$  is equal to  $2^{-n_1} \binom{2n_1}{n_1}$  (since it is the same as the probability that  $s + t = n_1$ ), which is well known to be at most  $n_1^{-1/2}$ , which in turn is at most  $2^{d-1} n^{-1/2}$ . Therefore, the expected density of  $X_{\pi, s} \cap X_{\pi, t}$  conditional on  $s \neq t$  is at least  $\delta^2 - 2^{d-1} n^{-1/2}$ .

Let us choose  $s < t$  such that  $\mu(X_{\pi, s} \cap X_{\pi, t}) \geq \delta^2 - 2^{d-1} n^{-1/2}$ , and let us write  $A_0^{(1)}$  and  $A_1^{(1)}$  for  $A_{\pi, s}$  and  $A_{\pi, t}$ . Note that  $A_0^{(1)}$  is a proper subset of  $A_1^{(1)}$ , that both are disjoint from the set  $\{\pi(n_1 + 1), \dots, \pi(n)\}$  and that  $A_0^{(1)} \cup B$  and  $A_1^{(1)} \cup B$  both belong to  $\mathcal{A}$  for every  $B \in X_{\pi, s} \cap X_{\pi, t}$ .

Now let us run the argument again, with  $n$  replaced by  $n - n_1$ ,  $n_1$  replaced by  $n_2$  and  $\mathcal{A}$  replaced by the set  $\mathcal{A}_1 = X_{\pi,s} \cap X_{\pi,t}$ . It gives us sets  $A_0^{(2)}$  and  $A_1^{(2)}$  and a set  $\mathcal{A}_2$  of subsets of  $\{\pi(n_2 + 1), \dots, \pi(n)\}$  such that  $A_0^{(2)}$  is a proper subset of  $A_1^{(2)}$ , both  $A_0^{(2)}$  and  $A_1^{(2)}$  are disjoint from  $\{\pi(n_2 + 1), \dots, \pi(n)\}$ , both  $A_0^{(2)} \cup B$  and  $A_1^{(2)} \cup B$  belong to  $\mathcal{A}_1$  for every  $B \in \mathcal{A}_2$ , and the density of  $\mathcal{A}_2$  is at least

$$(\delta^2 - 2^{d-1}n^{-1/2})^2 - 2^{d-2}n^{-1/2} \geq \delta^4 - 2^{d-1}n^{-1/2},$$

where for the last inequality we used the fact that  $\delta < 1/2$ .

If we continue this process and have shown that  $\mathcal{A}_r$  has density at least  $\delta^{2^r} - 2^{d-r+1}n^{-1/2}$ , then at the next stage we obtain  $\mathcal{A}_{r+1}$  with density at least

$$(\delta^{2^r} - 2^{d-r+1}n^{-1/2})^2 - 2^{d-r-1}n^{-1/2} \geq \delta^{2^{r+1}} - 2^{d-r}n^{-1/2}.$$

Therefore, as long as  $\delta^{2^{d-1}} - 4n^{-1/2} \geq 1/2\sqrt{2n/3}$ , then by Sperner's theorem  $\mathcal{A}_{d-1}$  contains two sets  $A_d^{(0)}$  and  $A_d^{(1)}$ , with  $A_d^{(0)}$  a proper subset of  $A_d^{(1)}$ . This gives us the desired combinatorial subspace (which consists of all sets of the form  $A_1^{(\epsilon_1)} \cup \dots \cup A_d^{(\epsilon_d)}$  such that each  $\epsilon_i$  is either 0 or 1).

The inequality we need is true if  $n \geq 5^2/\delta^{2^d}$ , so the theorem is proved.  $\square$

### 3. EQUAL-SLICES MEASURE AND PROBABILISTIC DHJ

The proof of Sperner's theorem can be regarded as follows. First, one chooses a different measure on the power set of  $[n]$ , where to choose a set you first choose its cardinality  $m$  uniformly at random from  $\{0, 1, 2, \dots, n\}$  and you then choose a random set of size  $m$ . The set of all subsets of  $[n]$  of size  $m$  is sometimes denoted by  $[n]^{(m)}$  and called a *layer* or *slice* of the cube. We therefore call the resulting probability measure on the power set of  $[n]$ , or equivalently on  $[2]^n$ , the *equal-slices measure*.

This measure arises so naturally in the averaging argument that we used to prove Sperner's theorem that it is tempting to say that the "real" theorem is that the maximum possible equal-slices measure of an antichain is  $1/(n+1)$ . One then converts that into a slightly artificial (and weaker) statement about the uniform measure.

The advantage of equal-slices measure is not just cosmetic, however: it and its obvious generalization to  $[k]^n$  will play a crucial role in our proof. Rather than saying straight away why this should be, we shall prove a result using equal-slices measure and explain why it would be problematic to give a uniform version.

But before we do that, let us give a formal definition of the equal-slices measure on  $[k]^n$ . This time we choose, uniformly at random from all possibilities, a  $k$ -tuple  $(a_1, \dots, a_k)$  of non-negative integers that add up to  $n$ , and then we choose a sequence  $x \in [k]^n$  such that for each  $j$  the set  $X_j = \{i : x_i = j\}$  has cardinality  $a_j$ , again uniformly from all possibilities (of which there are  $\binom{n}{a_1, \dots, a_k}$ ).

The number of slices can be worked out by a "holes and pegs" argument: given any subset  $B = \{b_1, \dots, b_{k-1}\}$  of  $\{1, 2, \dots, n+k-1\}$  of size  $k-1$ , let  $a_i$  be the number of integers strictly between  $b_{i-1}$  and  $b_i$ , where we treat  $b_0$  as 0 and  $b_k$  as  $n+k$ . This gives us all possible sequences  $(a_1, \dots, a_k)$  exactly once each, so the number of slices is  $\binom{n+k-1}{k-1}$ .

For use in the proof of the next theorem, we note that if  $k=3$  then the number of slices with  $a_2=0$  is  $n+1$ , so the probability that  $a_2=0$  is  $(n+1)/\binom{n+2}{2}$ , which equals  $2/(n+2)$ .



We can easily define equal-slices measure for combinatorial lines as well. Indeed, there is a one-to-one correspondence between lines in  $[k]^n$  and points in  $[k+1]^n$ , at least if one allows the lines to be degenerate. If  $y \in [k+1]^n$ , then the corresponding line consists of all points of the form  $y^{k+1 \rightarrow j}$  with  $j \in [k]$ ; in other words, the set of  $i$  such that  $y_i = k+1$  is treated as a wildcard set.

**3.1. A probabilistic version of Sperner's theorem.** As mentioned in the introduction, our proof of DHJ uses a density-increment strategy: that is, we assume that  $A$  does not contain a line and deduce that  $A$  has increased density inside some subspace. In almost all known proofs of this kind, one can in fact get away with a weaker hypothesis. If  $A$  is a dense set inside which one wishes to find some structure, then one can find a density increment on the assumption that  $A$  has “too few” subsets of the kind one is looking for, or more generally “the wrong number” of such subsets, where “the right number” is the number you would expect if  $A$  is a random subset of density  $\delta$ . Similarly, it is also possible to find equivalent versions of the theorems that say that a set  $A$  of density  $\delta$  contains not just one subset of the desired kind, but “many” such subsets, where this means that if you choose a random such subset then with probability at least  $c = c(\delta) > 0$  it will lie in  $A$ . A statement like this is called a “probabilistic version” of the density theorem.

This is a sufficiently important feature of previously known arguments that it is initially unsettling to observe that it is false for DHJ even when  $k = 2$ . The reason is a simple one. By standard measure-concentration results, almost all points in  $[2]^n$  have roughly  $n/2$  1s and  $n/2$  2s. By the same results, almost all combinatorial lines have roughly  $n/3$  fixed 1s,  $n/3$  fixed 2s and  $n/3$  variable coordinates. (A precise statement expressing this can be found in Lemma 6.2 below.) It follows that there is a set of density almost 1 (the set of sequences with roughly equal numbers of 1s and 2s) that contains only a tiny fraction of all lines (ones with roughly  $n/2$  fixed 1s, roughly  $n/2$  fixed 2s and a very small wildcard set).

However, this does not mean that there is no probabilistic version of DHJ, which is fortunate as we shall need one later. It merely means that the uniform measure is the wrong measure in which to express it. To illustrate this point, we now prove a “probabilistic” version of DHJ<sub>2</sub>. It tells us that an equal-slices-dense subset of  $[2]^n$  must contain an equal-slices-dense set of lines.

**Theorem 3.1.** *Let  $A$  be a subset of  $[2]^n$  of equal-slices density  $\delta$ . Then the set of (possibly degenerate) combinatorial lines in  $A$  has equal-slices density at least  $\delta^2(n+1)/(n+2)$ .*

*Proof.* Let  $\pi$  be a random permutation of  $[n]$  and let  $s$  and  $t$  be elements of  $\{0, 1, 2, \dots, n\}$  chosen independently and uniformly at random. Let us write  $x_{\pi, m}$  for the sequence that takes the value 1 at  $\pi(1), \dots, \pi(m)$  and 0 everywhere else, and let  $X_\pi$  be the number of the sequences  $x_{\pi, s}$  that belong to  $A$ . Then  $\mathbb{E}X_\pi = \delta n$ , by the definition of equal-slices measure.

From this it follows that  $\mathbb{E}X_\pi^2$  is at least  $\delta^2 n^2$ . But  $X_\pi^2$  is the number of pairs  $(s, t)$  such that both  $x_{\pi, s}$  and  $x_{\pi, t}$  belong to  $A$ . Therefore, if we choose a random pair  $\{x_{\pi, s}, x_{\pi, t}\}$  then the probability that both its constituent sequences belong to  $A$  is at least  $\delta^2$ .

Now each such pair forms a combinatorial line. If  $s \leq t$ , then this line consists of all sequences  $x$  such that  $x_i = 1$  if  $i \in \{\pi(1), \dots, \pi(s)\}$ ,  $x_i = 0$  if  $i \in \{\pi(t+1), \dots, \pi(n)\}$ , and  $x$  is constant on the set  $\{\pi(s+1), \dots, \pi(t)\}$ . (Thus, the set  $\{\pi(s+1), \dots, \pi(t)\}$  is the wildcard set.) If  $t \leq s$  then we simply interchange the roles of  $s$  and  $t$  in the above.

(If  $t = s$  then we have a degenerate line and interchanging the roles of  $s$  and  $t$  makes no difference.)

There is one technical detail that we need to address, which is that the probability  $p(\ell)$  that we choose a particular combinatorial line  $\ell$  is not quite the equal-slices probability  $q(\ell)$ . In particular, the probability that the line is degenerate is  $(n+1)^{-1}$  instead of  $2(n+2)^{-1}$ . However, if we condition on the event that  $s \neq t$ , then we are choosing a random subset of  $\{0, 1, 2, \dots, n\}$  of size 2, and such pairs are in one-to-one correspondence with triples  $(a_1, a_2, a_3)$  such that  $a_1 + a_2 + a_3 = n$  and  $a_2 \neq 0$ . Thus,  $p(\ell) = (n+2)q(\ell)/2(n+1)$  if  $\ell$  is degenerate, and  $p(\ell) = (1 - (n+1)^{-1})q(\ell)/(1 - 2(n+2)^{-1}) = (n+2)q(\ell)/(n+1)$  if  $\ell$  is non-degenerate.

From the above calculation it follows that the set of lines in  $A$  has equal-slices density at least  $\delta^2(n+1)/(n+2)$ , as claimed.  $\square$

The equal-slices density of the set of degenerate lines is  $O(n^{-1})$ , so this result implies that there is a dense set of non-degenerate combinatorial lines in  $A$  as well.

**3.2. Non-degenerate equal-slices measure.** For technical reasons, it is sometimes convenient, when talking about equal-slices measure, to condition on the event that every  $j \in [k]$  is equal to  $x_i$  for some  $i$ . Indeed, we have already seen in the proof of Theorem 3.1 that degenerate slices—that is, slices for which this condition does not hold—can be slightly problematic. It turns out that if we condition on the slices not being degenerate, then we can prove a useful lemma that would hold only approximately, and after tedious consideration of the degenerate cases, if we used the equal-slices measure itself.

Let us therefore define the *non-degenerate equal-slices measure* on  $[k]^n$  as follows. One first chooses a random  $k$ -tuple of positive (rather than non-negative) integers  $(a_1, \dots, a_k)$  that add up to  $n$  and then a random sequence  $x \in [k]^n$  such that  $|X_j| = a_j$  for each  $j$ , where as before  $X_j$  is the set  $\{i \in [n] : x_i = j\}$ .

A helpful equivalent way of defining this measure is as follows. To select a random point  $x \in [k]^n$ , one places  $n$  points  $q_1, \dots, q_n$  around a circle in a random order. That creates  $n$  gaps between consecutive points. One chooses a random set of  $k$  of these gaps and places further points  $r_1, \dots, r_k$  into the gaps, again in a random order. Finally, one sets  $x_i$  to be  $j$  if and only if  $r_j$  is the first point out of  $r_1, \dots, r_k$  that you come to if you go round clockwise starting at  $q_i$ .

Note that since the  $q_i$  are in a random order, precisely the same distribution will arise if the  $r_j$  are placed in some fixed order rather than their order too being randomized. However, it is more convenient to randomize everything. Note also that since we do not allow two different  $r_j$  to occupy the same gap, for each  $j$  there exists  $i$  such that  $x_i = j$ . Finally, note that apart from this constraint, all slices are equally likely. Therefore, we really do have the equal-slices measure conditioned on the event that the slices are non-degenerate.

To see the effect that this conditioning has, let us give an upper bound for the probability is that a slice  $i$ s degenerate.

**Lemma 3.2.** *Let  $x$  be an equal-slices random point of  $[k]^n$ . Then the probability that no coordinate of  $x$  is equal to  $k$  is  $\frac{k-1}{n+k-1}$ . In particular, it is at most  $k/n$ .*

*Proof.* To choose  $k$  non-negative integers  $a_1, \dots, a_k$  that add up to  $n$ , and to do so uniformly from all possibilities, one can choose a random subset  $P = \{p_1 < \dots < p_{k-1}\} \subset$

$\{1, 2, \dots, n+k-1\}$  of size  $k-1$  (of “pegs”) and let  $a_i$  be the number of integers strictly between  $p_{i-1}$  and  $p_i$ , where we set  $p_0 = 0$  and  $p_k = n+k$ . The probability that no coordinate of  $x$  is equal to  $k$  is the probability that  $a_k = 0$ , which is the probability that  $n+k-1 \in P$ , which is  $\frac{k-1}{n+k-1}$ , as claimed.  $\square$

**Corollary 3.3.** *Let  $\nu$  and  $\tilde{\nu}$  be the equal-slices and non-degenerate equal-slices measures on  $[k]^n$ , respectively. Then for any set  $A \subset [k]^n$  we have  $|\nu(A) - \tilde{\nu}(A)| \leq k^2/n$*

*Proof.* It follows from Lemma 3.2 that the probability that a slice is degenerate is at most  $k^2/n$ . Therefore, if  $A$  is a set that consists only of non-degenerate sequences, then its non-degenerate equal-slices measure is  $(1-c)^{-1}$  times its equal-slices measure, for some  $c < k^2/n$ . Therefore, for such a set,  $0 \leq \tilde{\nu}(A) - \nu(A) = c\tilde{\nu}(A) \leq k^2/n$ . If  $A$  consists only of degenerate sequences, then  $0 \leq \nu(A) - \tilde{\nu}(A) = \nu(A) \leq k^2/n$ . The result follows, since if one takes a union of sets of the two different kinds, then the differences cancel out rather than reinforcing each other.  $\square$

For later use, we slightly generalize Lemma 3.2.

**Lemma 3.4.** *Let  $x$  be chosen randomly from  $[k]^n$  using the equal-slices distribution. Then the probability that fewer than  $m$  coordinates of  $x$  are equal to  $k$  is at most  $mk/n$ .*

*Proof.* Let  $P$  be as in the proof of Lemma 3.2. This time we are interested in the probability that  $p_{k-1} \geq n+k-m$ . The number with  $p_{k-1} = n+k-s$  is  $\binom{n+k-s-1}{k-2}$ , which is at most  $\binom{n+k-2}{k-2}$ , which as we noted in the proof of Lemma 3.2 is at most  $\frac{k}{n} \binom{n+k-1}{k-1}$ . The result follows.  $\square$

**Corollary 3.5.** *Let  $x$  be chosen randomly from  $[k]^n$  using the equal-slices distribution. Then the probability that there exists  $j \in [k]$  such that fewer than  $m$  coordinates of  $x$  are equal to  $j$  is at most  $mk^2/n$ .*

*Proof.* This follows immediately from Lemma 3.4.  $\square$

Now let us return to our discussion of the non-degenerate equal-slices measure. The next result tells us that it has a beautiful property. Let us use the expression  $\tilde{\nu}$ -random to mean “random and chosen according to the non-degenerate equal-slices measure”. Then the property is that a  $\tilde{\nu}$ -random point in a  $\tilde{\nu}$ -random subspace with no fixed coordinates is a  $\tilde{\nu}$ -random point. This result will enable us to carry out clean averaging arguments when we are using equal-slices measure.

We have not said what we mean by a  $\tilde{\nu}$ -random subspace with no fixed coordinates, but the definition is a straightforward modification of our earlier definition of the equal-slices density of a set of combinatorial lines. First, a  $d$ -dimensional subspace with no fixed coordinates is simply a subspace obtained by partitioning  $[n]$  into  $d$  non-empty sets  $X_1, \dots, X_d$  and taking the set of all sequences  $x \in [k]^n$  that are constant on each  $X_i$ . For brevity, let us call these *special* subspaces.

As we mentioned earlier, just as a combinatorial line in  $[k]^n$  can be associated with a point in  $[k+1]^n$ , so a  $d$ -dimensional combinatorial subspace in  $[k]^n$  can be associated with a point in  $[k+d]^n$ . If the subspace is special, then it will in fact be associated with a point in  $[d]^n$ .

In the reverse direction, if  $x \in [k+d]^n$ , then the corresponding  $d$ -dimensional subspace is the set of all points  $y$  such that  $y_i = j$  whenever  $j \in [k]$  and  $x_i = j$ , and  $y$  is constant on all sets of the form  $X_j = \{i : x_i = j\}$  when  $j > k$ . Thus, the wildcard sets are the  $d$  sets

$X_{k+1}, \dots, X_{k+d}$ . In the case of special subspaces, we take instead  $x$  to belong to  $[d]^n$  and the wildcard sets are  $X_1, \dots, X_d$ .

Therefore, when we talk about the equal-slices measure or non-degenerate equal-slices measure of a set of special  $d$ -dimensional subspaces, we are associating with each subspace a point in  $[d]^n$  and taking the corresponding measure there. (A small detail is that for this to work we need the wildcard sets in the combinatorial subspace to form a sequence rather than just a set. In other words, if we permute the ‘‘basis’’ then we are considering the result as a different subspace, even though it consists of the same points. Alternatively, one could regard the correspondence as being  $d!$ -to-one.)

**Lemma 3.6.** *Let  $n, k$  and  $d$  be positive integers with  $n \geq k + d$ . Suppose that a point  $x \in [k]^n$  is chosen randomly by first choosing a  $\tilde{\nu}$ -random special  $d$ -dimensional subspace  $V$  of  $[k]^n$  and then choosing a  $\tilde{\nu}$ -random point in  $V$ . Then the resulting distribution is the non-degenerate equal-slices measure on  $[k]^n$ .*

*Proof.* To prove this we use the second method of defining the non-degenerate equal-slices measure. That is, we choose a random subspace as follows. First, we place  $n$  points  $q_1, \dots, q_n$  in a random order around a circle. Next, we choose  $d$  points  $r_1, \dots, r_d$  and place them in random gaps between the  $q_i$ , with no two of the  $r_h$  occupying the same gap. Then the wildcard set  $X_h$  will consist of all  $h$  such that  $r_h$  is the first of the points  $r_1, \dots, r_d$  if you go clockwise round the circle from  $q_i$ . Let us call the set of points  $q_i$  with this property, together with  $r_h$ , the  $h$ th *block*.

How do we then choose a random point  $x$  in this subspace? We can think of it as follows. We take the  $d$  blocks and randomly permute them. We then randomly place  $k$  points  $s_1, \dots, s_k$  in gaps between blocks (with no two  $s_j$  in the same gap). Then  $x_i = j$  if  $s_j$  is the first of the points  $s_1, \dots, s_k$  if you go clockwise round from  $q_i$  (after the blocks have been permuted).

Now consider a second way of choosing a random point in  $[k]^n$ . We proceed exactly as above, except that this time we do not bother to permute the blocks. We claim that this gives rise to exactly the same distribution.

To see this, let us call two valid arrangements of the points  $q_1, \dots, q_n$  and  $r_1, \dots, r_d$  *equivalent* if one is obtained from the other by a permutation of the blocks. Then all the equivalence classes have size  $d!$ , so randomly choosing an arrangement is the same as randomly choosing an arrangement and then randomly changing it to an equivalent arrangement.

Now the second way of choosing a random sequence amounts to choosing the random points  $q_1, \dots, q_n$  and  $r_1, \dots, r_d$ , randomly choosing  $k$  of the points  $r_1, \dots, r_d$  and calling them  $s_1, \dots, s_k$  (in a random order) and finally using the points  $q_1, \dots, q_n, s_1, \dots, s_k$  to define a point in  $[k]^n$  in the usual way. But this is precisely the non-degenerate equal-slices measure on  $[k]^n$ .  $\square$

**3.3. A probabilistic version of the density Hales-Jewett theorem.** With the help of Corollary 3.3 and Lemma 3.6, it is straightforward to prove that a probabilistic version of  $\text{DHJ}_k$  follows from an ‘‘equal-slices version’’. Let us begin by stating the equal-slices version.

**Theorem 3.7.** *For every  $\delta > 0$  and every positive integer  $k$  there exists  $n$  such that every set  $A \subset [k]^n$  of equal-slices density at least  $\delta$  contains a combinatorial line.*

We shall show later that Theorem 3.7 follows from  $\text{DHJ}_k$  itself. For now let us assume it and deduce a probabilistic version. We shall write  $\text{EDHJ}(k, \delta)$  for the smallest integer  $m$  such that every subset  $A \subset [k]^m$  of equal-slices density at least  $\delta$  contains a combinatorial line.

**Theorem 3.8.** *Let  $\delta > 0$  and let  $k$  be an integer greater than or equal to 2. Then there exists  $\theta = \text{PDHJ}(k, \delta)$  such that for every  $n \geq \max\{m, 4k^2/\delta\}$  and every  $A \subset [k]^n$  of equal-slices density at least  $\delta$  the equal-slices density of the set of combinatorial lines in  $A$  is at least  $\theta$ . Moreover, if  $m = \text{EDHJ}(k, \delta/4)$  then we can take  $\theta = (\delta/9)(k+1)^{-m}$ .*

*Proof.* (Assuming Theorem 3.7.) By Corollary 3.3 the non-degenerate equal-slices density  $\tilde{\nu}(A)$  of  $A$  is at least  $\delta - k^2/n$ . Since  $n \geq 4k^2/\delta$ , this is at least  $3\delta/4$ .

Let  $V$  be a random  $m$ -dimensional special subspace of  $[k]^n$ , chosen according to the non-degenerate equal-slices measure. Then Lemma 3.6 implies that the expected non-degenerate equal-slices density of  $A$  inside  $V$  is also at least  $3\delta/4$ , from which it follows that with probability at least  $\delta/4$  this density is at least  $\delta/2$ .

Let  $V$  be a subspace inside which  $A$  has non-degenerate equal-slices density at least  $\delta/2$ . Remove from  $A \cap V$  all degenerate strings. The resulting set  $A' \cap V$  still has density at least  $\delta/2$ . By Corollary 3.3 again, this implies that the equal-slices density of  $A'$  inside  $V$  is at least  $\delta/4$ .

But by our choice of  $m$  this means that with probability at least  $\delta/4$  the set  $A' \cap V$  contains a combinatorial line. Moreover, since  $A' \cap V$  contains no degenerate strings, this line must have fixed coordinates of every single value.

The number of such lines is at most  $(k+1)^m$ . Therefore, if you choose a random special subspace and inside it you choose a line according to the non-degenerate equal-slices measure, then with probability at least  $(\delta/4)(k+1)^{-m}$  it will be a line in  $A$ .

But by Lemma 3.6 the way we have just chosen this line was according to the non-degenerate equal-slices measure. By the proof of Corollary 3.3, the equal-slices probability is at least  $(\delta/4)(k+1)^{-m}(1 - (k+1)^2/n)$ . By our assumption that  $n \geq 4k^2/\delta$  (and that  $k \geq 2$ ), this is at least  $(\delta/9)(k+1)^{-m}$ .  $\square$

#### 4. A MODIFICATION OF AN ARGUMENT OF AJTAI AND SZEMERÉDI

After Szemerédi proved his theorem on arithmetic progressions, it was natural to try to prove the multidimensional version, which states that for every finite subset  $H$  of  $\mathbb{Z}^d$  and every  $\delta > 0$  there exists  $N$  such that every subset  $A$  of  $[N]^d$  of size at least  $\delta N^d$  contains a subset of the form  $aH + b$  with  $a > 0$ . A full proof of this result had to wait for the ergodic approach of Furstenberg: the result is due to Furstenberg and Katznelson [FK78]. However, Ajtai and Szemerédi managed to prove the first genuinely multidimensional case of the theorem, where  $H$  is the set  $\{(0,0), (1,0), (0,1)\}$ , by means of a clever deduction from Szemerédi's theorem itself. Their argument is based on a density-increment strategy, but it is not organized in quite the way that was described in §1.4. However, it is possible to reorganize the steps so that it follows that general outline very closely: in this section we briefly sketch this slight modification of their argument because it provides a template for our proof of the density Hales-Jewett theorem.

Let  $\delta > 0$ , let  $N$  be a large integer, and let  $A$  be a subset of  $[N]^2$  of density at least  $\delta$ . Our aim is to show that  $A$  contains a triple of the form  $\{(x, y), (x+d, y), (x, y+d)\}$  with

$d > 0$ . We shall call such configurations *corners*. The theorem of Ajtai and Szemerédi is the following.

**Theorem 4.1.** *For every  $\delta > 0$  there exists  $N$  such that every subset  $A \subset [N]^2$  of density at least  $\delta$  contains a triple  $\{(x, y), (x + d, y), (x, y + d)\}$  with  $d > 0$ .*

Before we sketch the proof, we make the general remark that there are three privileged directions, horizontal, vertical and parallel to the line  $x + y = 0$ , which correspond to the three lines that are defined by pairs of points from the set  $\{(0, 0), (1, 0), (0, 1)\}$ . Indeed, one could argue that the formulation of the problem is an unnatural one, and that instead of the grid  $[N]^2$  one should consider a triangular portion of a triangular lattice, so that there is a symmetry between the three directions. We shall not do this, but when we come to relate the argument of this section to the proof of DHJ, it will help to bear this point in mind.

We shall regard certain subsets of  $[N]^2$  as “simple” or “somewhat structured”. We define a *1-set* to be a subset of the form  $X \times [N]$ . We call such sets 1-sets because whether or not a point  $(x, y)$  belongs to  $X \times [N]$  depends only on its first coordinate  $x$ . A more symmetrical, and therefore preferable, explanation is this. We represent our points not by pairs  $(x, y)$  with  $x, y \in [N]$  but as *triples*  $(x, y, z)$  such that  $x, y \in [N]$  and  $x + y + z = 2N + 1$ . (We have chosen  $2N$  so that  $z$  lies between 1 and  $2N - 1$ , but all we care about is that  $x + y + z$  should be constant.) It is still true that whether or not the point represented by a triple  $(x, y, z)$  belongs to  $X \times [N]$  depends only on  $x$ . In other words, if  $(x, y, z)$  belongs to a 1-set, then so does  $(x, y + u, z - u)$  for every  $u$ . Another way of looking at this, which turns out to correspond more closely to what we shall do when we prove DHJ, is think of a 1-set as a *23-insensitive set*, meaning that membership of the set is unaffected by changes to the second and third coordinates.

Another special kind of set is one of the form  $X \times Y$ . This is the intersection of the 1-set  $X \times [N]$  and the 2-set  $[N] \times Y$ . In this section we shall call it a 12-set (which is not to be confused with a 12-insensitive set, which we are calling a 3-set).

Now let us sketch the argument that gives us corners. The basic idea is a *density increment strategy*, which has been used to prove many density theorems. (A few examples can be found in [Rot53], [Sze75], [Gow01], [Shk06b], [Shk06a], and [LM08], but this is by no means an exhaustive list.) We shall show that if  $A$  does not contain a corner, then there is some subset of  $[N]^2$  that looks like  $[m]^2$ , and inside that subset  $A$  has an increased density. We can iterate this argument until eventually we reach a contradiction when the relative density of  $A$  inside some subset becomes greater than 1.

**4.1. Finding a dense diagonal.** The first step is to find a set of the form  $\{(x, y) : x + y = t\}$  that contains a reasonable number of points of  $A$ . Since there are  $2N - 1$  such sets and  $A$  has size at least  $\delta N^2$ , at least one such set contains at least  $\delta N/2$  points of  $A$ .

**4.2. A dense 12-set that is disjoint from  $A$ .** Suppose that we have found  $t$  such that the number of points of  $A$  in the diagonal  $\{(x, y) : x + y = t\}$  is at least  $\delta N/2$ . Let us write these points as  $(x_1, y_1), \dots, (x_{2m}, y_{2m})$  with  $x_1 < \dots < x_{2m}$ . If the number of points of  $A$  on the diagonal is odd, we just omit one of them. Let  $X = \{x_1, \dots, x_m\}$  and let  $Y = \{y_{m+1}, \dots, y_{2m}\}$ . Then no point of  $X \times Y$  can belong to  $A$ , since if  $(x_i, y_j) \in A$  then the three points  $(x_i, y_j), (x_j, y_j)$  and  $(x_i, y_i)$  all belong to  $A$ , and they form a corner since  $x_j - x_i = y_i - y_j > 0$ . The size of  $X \times Y$  is  $m^2$ , and  $m \geq \lfloor \delta N/4 \rfloor$ , so (ignoring the integer part)  $X \times Y$  has density at least  $\delta^2/16$  or so.

**4.3. A dense 12-set that correlates with  $A$ .** If  $A$  is disjoint from a dense 12-set  $X \times Y$  then it must make up for this with an increased density in the complement of  $X \times Y$ . However, the complement of  $X \times Y$  splits up into the three 12-sets  $X \times Y^c$ ,  $X^c \times Y$  and  $X^c \times Y^c$ . A simple averaging argument shows that in at least one of these three 12-sets the relative density of  $A$  is at least  $\delta + \delta^3/48$ . Thus, we have sets  $U$  and  $V$  such that the density of  $A$  inside the 12-set  $U \times V$  is at least  $\delta + \delta^3/48$ . Moreover, a very crude argument shows that the  $U \times V$  must have density at least  $\delta^3/48$  inside  $[N]^2$ .

**4.4. A dense 1-set can be almost entirely partitioned into large grids.** As mentioned earlier, our eventual aim is to find a subset of  $[N]^2$  of a similar type, inside which  $A$  has increased density. The subsets that will interest us are *grids*, which are sets of the form  $P \times Q$ , where  $P$  is an arithmetic progression and  $Q$  is a translate of  $P$ .

Given a dense 1-set  $X \times [N]$ , we can partition almost all of it into grids as follows. Suppose that the density of  $X$  is  $\theta$  and let  $\epsilon$  be some positive constant that is much smaller than  $\theta$  (but independent of  $N$ ). Since  $X$  has density at least  $\epsilon$ , by Szemerédi's theorem it contains an arithmetic progression  $P_1$  of length at least  $m$ , where  $m$  tends to infinity with  $N$ . If the set  $X \setminus P_1$  still has density at least  $\epsilon$ , then it contains an arithmetic progression of length  $m$ . Indeed, we can partition  $X$  into sets  $P_0, P_1, \dots, P_r$ , where  $P_1, \dots, P_r$  are arithmetic progressions of length at least  $m$  and  $P_0$  is a residual set of density less than  $\epsilon$ .

For each  $i$ , we can then straightforwardly partition almost all of  $P_i \times [N]$  into sets of the form  $P_i \times Q_{ij}$ , where each  $Q_{ij}$  is a translate of  $P_i$ . (It helps if each  $P_i$  has diameter at most  $\epsilon N$ , but it is easy to ensure that this is the case.) We can therefore partition all but an arbitrarily small proportion of  $X \times [N]$  into grids of size tending to infinity with  $N$ .

**4.5. A dense 12-set can be almost entirely partitioned into large grids.** It is easy to deduce from the previous step a similar statement about 12-sets. Indeed, let  $X$  and  $Y$  be dense sets, and begin by partitioning almost all of  $X \times [N]$  into large grids  $P_i \times Q_i$ . (We have changed the indexing of these grids.) The intersection of  $X \times Y$  with any of these grids  $P_i \times Q_i$  is  $P_i \times (Y \cap Q_i)$ , since  $P_i \subset X$ . Therefore, if  $Y \cap Q_i$  has positive density inside  $Q_i$ , we can use the previous step to partition almost all of  $P_i \times (Y \cap Q_i)$  into subgrids, still with size tending to infinity. By a simple averaging argument, the proportion of points in  $X \times Y$  that are contained in grids  $P_i \times Q_i$  inside which  $Y$  is sparse is small. So by this means we have partitioned almost all of  $X \times Y$  into grids with sizes that tend to infinity.

**4.6. A density increment on a large grid.** By Step 3, we have a dense 12-set  $X \times Y$  inside which the density of  $A$  is at least  $\delta + \delta^3/48$ . By Step 5 we can partition almost all of  $X \times Y$  into large grids. If we choose "almost" appropriately, we can ensure that the density of that part of  $A$  that lies in these large grids is at least  $\delta + \delta^3/100$ . But then by averaging we can find a large grid  $P \times Q$  such that the density of  $A$  inside  $P \times Q$  is at least  $\delta + \delta^3/100$ . This is exactly what we need for our density-increment strategy, so the proof is complete.

## 5. A DETAILED SKETCH OF A PROOF OF DHJ<sub>3</sub>

In this section, we shall explain in some detail how our proof works in the case  $k = 3$ . As mentioned in the previous section, the structure of our proof is closely modelled on the structure of the argument of Ajtai and Szemerédi (in the slightly modified form in which we have presented it). However, to make that clear, we need to explain what the

counterparts are of concepts such as “grid”, “12-set” and the like. So let us begin by discussing a dictionary that will guide us in our proof.

Everything flows from the following simple thought: whereas a typical point in  $[N]^2$  can be thought of as a triple  $(x, y, z)$  such that  $x + y + z = 2N + 1$ , a typical point in  $[3]^n$  can be thought of as a triple of disjoint sets  $(X, Y, Z)$  such that  $X \cup Y \cup Z = [n]$ : to turn such a triple into a sequence  $(x_1, \dots, x_n)$  let  $x_i = 1$  if  $i \in X$ , 2 if  $i \in Y$  and 3 if  $i \in Z$ .

A corner in  $[N]^2$  can be defined symmetrically as a triple of points of the form  $\{(x + u, y, z), (x, y + u, z), (x, y, z + u)\}$  such that  $x + y + z + u = 2N + 1$  and  $u \neq 0$ . This translates very nicely: a combinatorial line is a triple of points of the form  $\{(X \cup U, Y, Z), (X, Y \cup U, Z), (X, Y, Z \cup U)\}$  such that  $X, Y, Z$  and  $U$  partition  $[n]$  and  $U \neq \emptyset$ .

A diagonal in  $[N]^2$  is a set of the form  $D_t = \{(x, y, z) : x + y = t\}$ . It therefore makes sense to define a “diagonal” in  $[3]^n$  to be a set of the form  $\{(X, Y, Z) : X \cup Y = T\}$  for some subset  $T \subset [n]$ . In other words, it is the collection of all triples  $(X, Y, Z)$  that partition  $[n]$ , but now  $Z$  is a fixed set (equal to the complement of  $T$  above).

Recall that a 1-set in  $[N]^2$  is a set of the form  $X \times [N]$ , or in symmetric notation a set of the form  $\{(x, y, z) : x \in X\}$ . The obvious generalization of this notion to  $[3]^n$  is a set of the form  $\{(X, Y, Z) : X \in \mathcal{X}\}$  for some collection  $\mathcal{X}$  of subsets of  $[n]$ . A subset  $S$  of  $[3]^n$  is a 1-set if and only if it is *23-insensitive* in the following sense: if  $(X, Y, Z) \in S$ , then  $(X, Y', Z') \in S$  whenever  $Y' \cup Z' = Y \cup Z$ . Equivalently, if a sequence  $x \in [3]^n$  belongs to  $S$ , then so do all sequences that can be formed from  $x$  by changing some 2s to 3s and/or some 3s to 2s.

The natural definition of a 12-set is now clear: as in the case of subsets of  $[N]^2$ , it should be the intersection of a 1-set with a 2-set.

We should also mention that the notion of Cartesian product has an analogue. The Cartesian product of  $X$  and  $Y$  is the intersection of the 1-set  $X \times [N]$  with the 2-set  $[N] \times Y$ . So if we are given two collections  $\mathcal{X}$  and  $\mathcal{Y}$  of subsets of  $[n]$ , then the analogue of their Cartesian product ought to be the 12-set  $\{(X, Y, Z) : X \in \mathcal{X}, Y \in \mathcal{Y}, X \cap Y = \emptyset\}$ . Since  $X$  and  $Y$  determine  $Z$ , we can think of this as a set of pairs, and then the resemblance with a true Cartesian product is that much closer: it is (equivalent to) the set of all pairs  $(X, Y)$  such that  $X \in \mathcal{X}$ ,  $Y \in \mathcal{Y}$  and  $X$  and  $Y$  are disjoint. We shall call this the *disjoint product* of  $\mathcal{X}$  and  $\mathcal{Y}$  and write it as  $\mathcal{X} \boxtimes \mathcal{Y}$ .

There is one concept that has a non-obvious (though still natural) translation from the  $[N]^2$  world to the  $[3]^n$  world, namely that of a grid. At first sight, it might seem extremely unlikely that the Ajtai-Szemerédi can be generalized to give a proof of  $\text{DHJ}_3$ . After all, their proof could be regarded as the beginnings of a sort of induction: they deduce the first non-trivial case of the two-dimensional theorem from the full one-dimensional theorem (namely Szemerédi’s theorem). If one is attempting to prove  $\text{DHJ}_3$ , the obvious candidate for a statement “one level down” is  $\text{DHJ}_2$ , but that is a much less deep statement than Szemerédi’s theorem. So it seems that our only hope will be if Ajtai and Szemerédi did not after all need a tool as powerful as Szemerédi’s theorem.

One of the key ideas of our proof is that this is indeed the case, though the result we need is not  $\text{DHJ}_2$  but its multidimensional version  $\text{MDHJ}_2$  proved in the last section. The appropriate replacement of the notion of a long arithmetic progression in  $[N]$  is a combinatorial subspace of  $[2]^n$ . We then have to decide what the analogue of a grid is. Given the concepts so far, it should be something like the disjoint product of two “parallel” combinatorial subspaces of  $[2]^n$ , and we would like that to give us a combinatorial subspace



of  $[3]^n$  (since we want the analogue of a grid to be a structure that resembles  $[3]^n$ ). All this can be done. A  $d$ -dimensional combinatorial subspace of  $[2]^n$  is defined by taking disjoint sets  $X_0, X_1, \dots, X_d$  and defining  $U$  to be the set of all unions  $X_0 \cup \bigcup_{i \in A} X_i$  such that  $A \subset [d]$ . It is natural to define two such subspaces to be parallel if they are defined by sequences of sets  $(X_0, X_1, \dots, X_d)$  and  $(Y_0, Y_1, \dots, Y_d)$  such that  $X_i = Y_i$  for every  $i \geq 1$ , and also, since we want to take a disjoint product, to add the condition that  $X_0$  and  $Y_0$  should be disjoint. If we do that, then a typical point in the disjoint product is a pair  $(X, Y)$  such that  $X = X_0 \cup \bigcup_{i \in A} X_i$  and  $Y = Y_0 \cup \bigcup_{i \in B} X_i$  such that  $A \cap B = \emptyset$ . If we set  $Z = [n] \setminus (X \cup Y)$ , we see easily that this is precisely a  $d$ -dimensional combinatorial subspace of  $[3]^n$ :  $X_0$  and  $Y_0$  are the sets where the fixed coordinates are 1 and 2, respectively, and the wildcard sets are  $X_1, \dots, X_d$ .

With these concepts in mind, let us now give an overview of the proof of DHJ<sub>3</sub>. (To generalize this discussion to DHJ<sub>k</sub> is straightforward: the Ajtai-Szemerédi argument can be used to deduce a “ $k$ -dimensional corners” theorem from the  $(k-1)$ -dimensional Szemerédi theorem, and that provides a template for our deduction of DHJ<sub>k+1</sub> from MDHJ<sub>k</sub>, which itself can be deduced from PDHJ<sub>k</sub>, which follows from DHJ<sub>k</sub>.)

**5.1. Finding a dense diagonal.** Recall that we are defining a diagonal in  $[3]^n$  to be a set of the form  $\{(X, Y, Z) : X \cup Y = T\}$ . Equivalently, one fixes a set  $Z$  and defines the associated diagonal to be the set of all sequences in  $[3]^n$  that take the value 3 in  $Z$  and 1 or 2 everywhere else.

Obviously the diagonals form a partition of  $[3]^n$ , so if  $A \subset [3]^n$  is a set of density  $\delta > 0$  then by averaging we can find a diagonal inside which  $A$  still has density  $\delta$ . We can also ensure that this diagonal is not too small by throwing away the very small fraction of  $[3]^n$  that is contained in small diagonals.

It is not completely obvious at this stage what probability measure we want to take on  $[3]^n$ , but note that the argument so far is general enough to apply to any measure.

**5.2. A dense 12-set that is disjoint from  $A$ .** What should we do next? In the equivalent stage of the corners argument we were assuming that  $A$  contained no corners. Then every pair of points of  $A$  in our dense diagonal implied that a third point (the bottom of the corner of which those two points formed the diagonal) did not belong to  $A$ . Moreover, the set of points that we showed did not belong to  $A$  formed a dense 12-set. So now we would like to do something similar.

At first, the situation looks very promising, since if  $(X, Y, Z)$  and  $(X', Y', Z)$  are two points with  $X \subset X'$ , both belonging to the diagonal determined by the set  $Z$ , then we can set  $U = X' \setminus X$  and write these two points as  $(X, Y \cup U, Z)$  and  $(X \cup U, Y, Z)$ . Then the point  $(X, Y, Z \cup U)$  cannot lie in  $A$ , since otherwise the three points would form a combinatorial line in  $A$ .

So what can we say about the set of all forbidden points? These are all points of the form  $(X, Y, Z \cup U)$  such that both  $(X \cup U, Y, Z)$  and  $(X, Y \cup U, Z)$  belong to  $A$ . Now  $Z$  is a fixed set (that defines the particular diagonal we are talking about), so if we are presented with a point  $(X, Y, Z \cup U)$  then we can work out what  $U$  is. Let  $\mathcal{X}$  be the set of all  $X \subset [n] \setminus Z$  such that  $(X, [n] \setminus (X \cup Z), Z) \in A$ . Then the set of all  $(X, Y, Z \cup U)$  such that  $(X, Y \cup U, Z) \in A$  is precisely the set of all  $(X, Y, Z \cup U)$  such that  $X \in \mathcal{X}$ . This would be a 1-set if we were not insisting that every point took the value 3 in the set  $Z$ . However, the set of all such points forms a subspace of  $[3]^n$  (of dimension  $n - |Z|$ ),

and inside that subspace we have a 1-set. Similarly, the set of all  $(X, Y, Z \cup U)$  such that  $(X \cup U, Y, Z) \in A$  is a 2-set inside the same subspace: this time we define  $\mathcal{Y}$  to be the set of all  $Y$  such that  $([n] \setminus (Y \cup Z), Y, Z) \in A$  and take the set of all points  $(X, Y, Z \cup U)$  such that  $Y \in \mathcal{Y}$ .

Thus, the good news is that we have found a 12-set that is disjoint from  $A$ , but the bad news is that this 12-set is in a subspace of  $[3]^n$  rather than in the whole space.

**5.3. A dense 12-set that correlates with  $A$ .** In the proof of the result about corners, we used a simple averaging argument at this stage: if there is a dense 12-set that is disjoint from  $A$  then one of three other 12-sets must have an unexpectedly large intersection with  $A$ . However, we cannot argue as straightforwardly here, since the 12-set we have found is *not* dense.

There are in fact two problems here. The first is the obvious one that we have restricted to a subspace, the density of which will be very small. To see this, note that for almost all points  $(X, Y, Z)$  in  $[3]^n$  the sets  $X, Y$  and  $Z$  have size very close to  $n/3$ . Therefore, it may well be that  $A$  consists solely of such points, in which case when we pass to the subspace that takes the value 3 on some fixed  $Z$  we will lose approximately  $n/3$  dimensions.

The second problem is that even when we do restrict to such a subspace we find that  $A$  may well have tiny density, since almost all triples in such a subspace will be of the form  $(X, Y, Z \cup U)$  with  $X, Y$  and  $U$  all of approximately the same size, and it may well be that no such triples belong to  $A$ , since then  $X, Y$  and  $Z \cup U$  do *not* all have approximately the same size.

To get round these problems, we do two things. First, we do not use the uniform measure on  $[3]^n$  but instead the equal-slices measure. This deals with the second problem, since for an equal-slices random triple  $(X, Y, Z)$  it is no longer the case that the sets  $X, Y$  and  $Z$  almost always have approximately the same size. Second, we argue that we may assume that the restriction of  $A$  to almost all subspaces has density at least  $\delta - \eta$  for some very small  $\eta$ . This observation is standard in proofs of density theorems: roughly speaking, if  $A$  often has smaller density than this, then somewhere it must have substantially larger density (by averaging), and then we have completed the iteration step in a particularly simple way. But if  $A$  almost always has density at least  $\delta - \eta$ , then when we use an averaging argument to find a diagonal that contains many points of  $A$ , we can also ask for  $A$  to have density at least  $\delta - \eta$  inside the subspace we are forced to drop down to.

Once all these arguments have been made precise, the conclusion is that there is a subspace  $V$  of  $[3]^n$  of reasonably large dimension such that the density of  $A$  inside  $V$  is at least  $\delta - \eta$ , and a dense 12-set inside that subspace that is disjoint from  $A$ . Then a simple averaging argument similar to the one in the corners proof gives us a dense 12-set in that subspace inside which the relative density of  $A$  is at least  $\delta + c(\delta)$ . (For this we must make sure we choose  $\eta$  sufficiently small for the small density decrease to be more than compensated for by the subsequent density increase.)

Thus, although the statement and proof of this step are directly modelled on the corresponding step for the corners proof, there are some important differences: we show that  $A$  correlates *locally* (that is, in some subspace of density that tends to zero) with a 12-set, whereas in the corners proof a global correlation is found. We do not know whether a dense subset of  $[3]^n$  that contains no combinatorial line must correlate globally with a 12-set. (Strictly speaking, we *do* know, since we have proved that every dense subset of  $[3]^n$  contains a combinatorial line. However, one can obtain a better formulation of the

question by replacing the assumption that the set contains no lines by the assumption that it contains few lines.) A second difference is that although we start with a set  $A$  that is equal-slices dense, the local correlation that the proof ends up giving is with respect to the uniform measure. (There is a general principle operating here, which is that equal-slices measure does not behave well when you restrict to combinatorial subspaces.)

**5.4. A dense 1-set can be almost entirely partitioned into large combinatorial subspaces.** Bearing in mind our dictionary, the next stage of the proof should be to partition almost all of a dense 1-set into combinatorial subspaces of dimension tending to infinity.

Let us recall what a 1-set, or a 23-insensitive set, is. It is a set  $A \subset [3]^n$  with the property that if  $x \in A$ ,  $y \in [3]^n$  and  $\{i : x_i = 1\} = \{i : y_i = 1\}$ , then  $y \in A$ . Equivalently, using set-theoretic notation, it is a set of triples of the form  $\{(X, Y, Z) : X \in \mathcal{X}\}$  for some collection  $\mathcal{X}$  of subsets of  $[n]$ .

At this stage of the corners proof, one starts with a 1-set  $X \times [N]$ , applies Szemerédi's theorem over and over again to remove arithmetic progressions  $P_i$  from  $X$  until it is no longer dense, and then partitions the sets  $P_i \times [N]$  into sets of the form  $P_i \times Q_{ij}$ , where the  $Q_{ij}$  are translates of  $P_i$ .

If we follow the proof of the corners theorem, then we should expect an argument along the following lines. We start with the 1-set  $\{(X, Y, Z) : X \in \mathcal{X}\}$ . We then partition almost all of  $\mathcal{X}$ , which can be thought of as a subset of  $[2]^n$ , into large combinatorial subspaces using repeated applications of MDHJ<sub>2</sub>. For each one of these subspaces  $U$ , we then partition the disjoint product  $U \boxtimes [3]^n$  into combinatorial subspaces.

Unfortunately, this last step does not work, which leads us to the second point where our argument is more complicated than that of Ajtai and Szemerédi, and the second place where we use localization to get us out of trouble. The difficulty is this. If  $U$  is the  $d$ -dimensional subspace defined by the sets  $(X_0, X_1, \dots, X_d)$ , then  $U \boxtimes [3]^n$  consists of all triples  $(X, Y, Z)$  of disjoint sets such that  $X$  is a union of  $X_0$  with some of the sets  $X_i$ . A combinatorial subspace inside this set must have wildcard sets that are unions of the  $X_i$  with  $i \geq 1$ , which means that it cannot contain any point  $(X, Y, Z)$  such that  $Y \cap X_i$  and  $Z \cap X_i$  are non-empty for every  $i$ .

This is a genuine difficulty, but we can get round it. The way we do so may at first look a little dangerous, but it turns out to work. The argument proceeds in five steps as follows.

- Let  $B$  be a 23-insensitive set of density  $\eta$ . Let  $m$  be a positive integer to be chosen later (for now it is sufficient to think of it as a number that tends to infinity but is much much smaller than  $n$ ), and choose a random element of  $[3]^n$  by randomly permuting the ground set  $[n]$  and then taking a pair  $(x, y)$ , where  $x$  is chosen uniformly from  $[2]^m$  and  $y$  is chosen uniformly from  $[3]^{n-m}$ . (Here we are regarding  $x$  as supported on the first  $m$  elements of the randomly permuted ground set and  $y$  as supported on the last  $n - m$  elements.) For sufficiently small  $m$ , the distribution of  $(x, y)$  is approximately uniform, so if for each  $y$  we let  $E_y = \{x : (x, y) \in B\}$ , then  $E_y$  has density at least  $\eta/3$  in  $[2]^m$  for a set of  $y$  of density at least  $\eta/3$ . (This is not the main reason that we need  $m$  to be small, so this step will be true with a great deal of room to spare.)
- For each such  $y$  use MDHJ<sub>2</sub> to find a  $d$ -dimensional combinatorial subspace  $U$  of  $[2]^m$  that lives inside  $E_y$ , and hence has the property that  $(x, y) \in B$  for every  $x \in U$ . (Here,  $d$  depends on  $m$  and  $\eta$ .)

- By the pigeonhole principle, we can find a subset  $T$  of  $[3]^{n-m}$  of density  $\theta = \theta(m, d, \eta)$  and a combinatorial subspace  $U \subset [2]^m$  such that  $U \times T \subset B$ . Let us choose  $T$  to be maximal: that is,  $T$  is the set of all  $y \in [3]^{n-m}$  such that  $U \times \{y\} \subset B$ . Since  $B$  is a 23-insensitive set, it follows that if we allow the wild-card sets of  $U$  to take the value 3 as well, then all the resulting points will still belong to  $B$ . That is, we have the same statement as above but now  $U$  is a combinatorial subspace of  $[3]^m$ . This is the point of our argument “where the induction happens”.
- $U \times T$  is a union of combinatorial subspaces, and there are quite a lot of them. It is tempting at this stage to remove them from  $B$  and start again. But unfortunately there is no reason to suppose that  $B \setminus (U \times T)$  will be 23-insensitive. (We give an example to illustrate this just after this proof outline.) However, this turns out not to be too serious a problem, because for every  $x \in X$  the set  $(B \setminus (U \times T)) \cap (\{x\} \times [3]^{n-m})$  is a 23-insensitive subset of  $\{x\} \times [3]^{n-m}$ . In other words, we can partition  $B \setminus (U \times T)$  into locally 23-insensitive sets and run the argument again.
- Using this basic idea, we develop an iterative proof. Whenever we are faced with a set of small density we regard it as part of our “error set” and leave it alone. And from any set of large density we remove a disjoint union of combinatorial subspaces and partition the rest into locally 23-insensitive sets. If we are careful, we can choose  $m$  in such a way that the combinatorial subspaces have dimension that tends to infinity with  $n$ , but the number of iterations before there are no dense sets left is smaller than  $n/m$ , so we never “run out of dimensions”. In this way we prove that a 23-insensitive set can almost all be partitioned into combinatorial subspaces.

Here, as promised, is an example of a 23-insensitive set  $B$  such that removing  $U \times T$  leaves us with a set that is no longer 23-insensitive. Let  $m = 2$  and  $n = 3$  and let  $B$  be the 23-insensitive set  $\{11, 22, 23, 32, 33\} \times \{2, 3\}$ . Then  $B$  contains the set  $\{11, 22, 33\} \times \{1, 2, 3\}$ , which is of the form  $U \times T$  with  $U$  a subspace and  $T$  23-insensitive (and it is the only non-empty subset of this form). If we remove this from  $B$ , we end up with the set  $\{23, 32\} \times \{2, 3\}$ , which is no longer 23-insensitive. It is, however, 23-insensitive in the third coordinate.

**5.5. A dense 12-set can be almost entirely partitioned into large combinatorial subspaces.** This stage of the argument is very similar to the corresponding stage of the corners argument and needs little comment. One simply checks that the intersection of a 13-insensitive set with a combinatorial subspace is 13-insensitive inside that subspace (which is almost trivial). Then, given an intersection of a 23-insensitive set and a 13-insensitive set, one applies the result of the previous section to the 23-insensitive set, partitioning almost all of it into subspaces, and then applies the same argument to the 13-insensitive set inside each subspace.

**5.6. A density increment on a large combinatorial subspace.** Again, this stage of the argument is very similar to the corresponding stage of the corners argument. If  $A$  has increased density on a (locally) 23-insensitive set, and if that set can be almost entirely partitioned into combinatorial subspaces of dimension tending to infinity, then by averaging we must be able to find one of these combinatorial subspaces inside which  $A$  has increased density.

We are not quite in a position to iterate at this point, because we started out with a set of equal-slices measure  $\delta$  and ended up finding a combinatorial subspace on which the *uniform* density had gone up. However, it turns out to be quite easy to pass from that to a further subspace inside which  $A$  has an equal-slices density increment, at which point we are done.

## 6. MEASURE FOR MEASURE

As we have already mentioned, there are some arguments that work better when we use product measures, and others when we use equal-slices measures. This appears to be an unavoidable situation, so we need a few results that will tell us that if we can prove a statement in terms of one measure then we can deduce a statement in terms of another. In this section, we shall collect together a number of such results, so that later on in the paper we can simply apply them when the need arises. The results we prove are just technical calculations, so the reader may prefer to take them on trust. The statements we shall need later are Corollary 6.4, Corollary 6.5 and Lemma 6.6.

We begin with a standard definition that will tell us when we regard two probability measures as being close.

**Definition.** Let  $\mu$  and  $\nu$  be two probability measures on a finite set  $X$ . The total variation distance  $d(\mu, \nu)$  is defined to be  $\max_{A \subset X} |\mu(A) - \nu(A)|$ .

In order to prove that we can switch from one probability measure to another, we shall make use of the following very simple general principle.

**Lemma 6.1.** Let  $\mu$  and  $\nu_1, \dots, \nu_m$  be probability measures, let  $a_1, \dots, a_m$  be positive real numbers that add up to 1, and suppose that  $d(\mu, \sum_{i=1}^m a_i \nu_i) \leq \eta$ . Then for every  $\alpha \in [0, 1]$  and every set  $A$  such that  $\mu(A) \geq \alpha$  there exists  $i$  such that  $\nu_i(A) \geq \alpha - \eta$ .

*Proof.* From our assumptions it follows that  $\sum_{i=1}^m a_i \nu_i(A) \geq \alpha - \eta$ , so by averaging it follows that there exists  $i$  such that  $\nu_i(A) \geq \alpha - \eta$ .  $\square$

**6.1. From uniform measure to equal-slices measure.** Before we apply Lemma 6.1, let us prove a simple but useful technical lemma.

**Lemma 6.2.** Let  $x$  be an element of  $[k]^n$  chosen uniformly at random, and for each  $j \in [k]$  let  $X_j = \{i : x_i = j\}$ . Then with probability at least  $1 - 2k \exp(-2n^{1/3})$  the sets  $X_j$  all have size between  $n/k - n^{2/3}$  and  $n/k + n^{2/3}$ .

*Proof.* The size of  $X_j$  is binomial with parameters  $n$  and  $1/k$ . Standard bounds for the tail of the binomial distribution therefore tell us that the probability that  $|X_j|$  differs from  $n/k$  by at least  $r$  is at most  $2 \exp(-2r^2/n)$ . (This particular bound follows from Azuma's inequality.) The result follows.  $\square$

As a first application of Lemma 6.1 we shall prove that a set of uniform density  $\delta$  has equal-slices density almost as great on some combinatorial subspace. The actual result we shall prove is, however, slightly more general. To set it up, we shall need a little notation.

Let  $m < n$ , let  $\sigma$  be an injection from  $[m]$  to  $[n]$ , let  $J = \sigma([m])$  and let  $\bar{J}$  be the complement of  $J$ . Then we can write each element of  $[k]^n$  as a pair  $(x, y)$  with  $x \in [k]^J$  and  $y \in [k]^{\bar{J}}$ . An element of  $[k]^J$  is a function from  $J$  to  $[k]$ . Given an element  $x = (x_1, \dots, x_m)$  of  $[k]^m$ , let  $\phi_\sigma(x)$  be the element of  $[k]^J$  that takes  $j \in J$  to  $x_{\sigma^{-1}(j)}$ . In other words,  $\phi_\sigma$  takes

an element of  $[k]^m$  and uses  $\sigma$  to turn it into an element of  $[k]^J$  in the obvious way. Given  $y \in [k]^{\bar{J}}$ , we also define a map  $\phi_{\sigma,y} : [k]^m \rightarrow [k]^n$  by taking  $\phi_{\sigma,y}(x)$  to be  $(\phi_\sigma(x), y)$ . Thus,  $\phi_{\sigma,y}$  is a bijection between  $[k]^m$  and the combinatorial subspace  $S_{J,y} = \{(x, y) : x \in [k]^J\}$  (in which the wildcard sets are all singletons  $\{i\}$  such that  $i \in J$ ).

Now let  $\nu$  be a probability measure on  $[k]^m$ . For each pair  $(\sigma, y)$  as above, we can define a probability measure  $\nu_{\sigma,y}$  on  $[k]^n$  by “copying”  $\nu$  in the obvious way. That is, given a subset  $A \subset [k]^n$  we let  $\nu_{\sigma,y}(A) = \nu(\phi_{\sigma,y}^{-1}(A))$ .

We now show that if  $m$  is sufficiently small, then the average of all the measures  $\nu_{\sigma,y}$  is close to the uniform measure on  $[k]^n$ .

**Lemma 6.3.** *Let  $\eta > 0$ , let  $k \geq 2$  be a positive integer, let  $n \geq (16k/\eta)^{12}$ , let  $m \leq n^{1/4}$ , let  $\nu$  be a probability measure on  $[k]^m$  and let  $\mu$  be the uniform measure on  $[k]^n$ . Then  $d(\mu, \mathbb{E}_{\sigma,y} \nu_{\sigma,y}) \leq \eta$ , where the average is over all pairs  $(\sigma, y)$  as defined above.*

*Proof.* We shall prove the result in the case where all of  $\nu$  is concentrated at a single point. Since all other probability measures are convex combinations of these “delta measures” (and their copies are the same convex combinations of the copies of the delta measures), the result will follow.

Let  $u$ , then, be an element of  $[k]^m$  and for each  $C \subset [k]^m$  let  $\nu(C) = 1$  if  $u \in C$  and 0 otherwise. For each injection  $\sigma : [m] \rightarrow [n]$  and each  $y \in [k]^{\bar{J}}$  (where  $\bar{J}$  is again the complement of  $\sigma([m])$ ), the measure  $\nu_{\sigma,y}$  is the delta measure at  $\phi_{\sigma,y}(u)$ . That is,  $\nu_{\sigma,y}(A) = 1$  if  $\phi_{\sigma,y}(u) \in A$  and  $\nu_{\sigma,y}(A) = 0$  otherwise.

What, then, is  $\mathbb{E}_{\sigma,y} \nu_{\sigma,y}(A)$ ? To answer this, let us see what happens when  $A$  is a singleton  $\{z\}$ . Then  $\nu_{\sigma,y}(A) = 1$  if and only if the restriction of  $z$  to  $J$  is  $\phi_\sigma(u)$  and the restriction of  $z$  to  $\bar{J}$  is  $y$ . So  $\mathbb{E}_{\sigma,y} \nu_{\sigma,y}(A)$  is the probability, for a randomly chosen pair  $(\sigma, y)$ , that  $z_{\sigma(i)} = u_i$  for every  $i \in [m]$  and the restriction of  $z$  to  $\bar{J}$  is  $y$ .

For every  $\sigma$ , the probability of the second event given  $\sigma$  is  $k^{m-n}$ , so it remains to calculate the probability that  $z_{\sigma(i)} = u_i$  for every  $i$ . For each  $j \in [k]$ , let  $X_j = \{i : z_i = j\}$  and let  $n_j$  be the cardinality of  $X_j$ . Now let us choose the values  $\sigma(1), \sigma(2), \dots, \sigma(m)$  one at a time and estimate the conditional probability that  $\sigma(i) \in X_{u_i}$  given that  $\sigma(h) \in X_{u_h}$  for every  $h < i$ . If we set  $p = \min_j n_j$  and  $q = \max_j n_j$ , then each conditional probability of this kind will be at most  $q/(n-m)$  and at least  $(p-m)/(n-m)$ .

Lemma 6.2 tells us that with probability at least  $1 - 2k \exp(-2n^{1/3})$  we have the bounds  $n/k - n^{2/3} \leq p$  and  $q \leq n/k + n^{2/3}$ . If those bounds hold, then the probability that  $\sigma(i) \in X_{u_i}$  for every  $i \in [m]$  lies between  $(1/k - 2n^{-1/3})^m$  and  $(1/k + 2n^{-1/3})^m$ . (Here we are using the inequality that  $(n/k + n^{2/3})/(n - n^{1/4}) \leq 1/k + 2n^{-1/3}$ , which holds if  $k \geq 2$  and  $n \geq 8$ .) Therefore, it lies between  $k^{-m}(1 - \eta/4)$  and  $k^{-m}(1 + \eta/4)$ . (This inequality is valid if  $n \geq (16k/\eta)^{12}$ , as we are assuming.)

We have just shown that the value of the measure  $\mathbb{E}_{\sigma,y} \nu_{\sigma,y}$  on a singleton  $\{z\}$  is approximately equal to the value taken by the uniform measure, provided that the singleton has roughly the same number of coordinates of each value.

Let  $B$  be the set of all “balanced” sequences  $z$ . That is,  $B$  is the set of  $z$  such that the assumptions of the above argument are satisfied. Then  $\mathbb{E}_{\sigma,y} \nu_{\sigma,y}(B) \geq (1 - 2k \exp(-2n^{1/3}))(1 - \eta/4) \geq 1 - \eta/2$ , from which it follows that  $\mathbb{E}_{\sigma,y} \nu_{\sigma,y}(B^c) \leq \eta/2$ . Therefore, if  $A$  is any subset of  $[k]^n$ , we have that

$$\mathbb{E}_{\sigma,y} \nu_{\sigma,y}(A) \leq \mu(A)(1 + \eta/4) + \eta/2$$

and

$$\mathbb{E}_{\sigma,y} \nu_{\sigma,y}(A) \geq \mu(A)(1 - \eta/4) - \eta/2.$$

Since  $\mu(A) \leq 1$ , it follows that  $|\mu(A) - \mathbb{E}_{\sigma,y} \nu_{\sigma,y}(A)| \leq \eta$ .

As commented at the beginning of the proof, the result for arbitrary  $\nu$  follows from this result, since we can write it as a convex combination of delta measures and apply the triangle inequality.  $\square$

Armed with this result, we now prove two statements that will be helpful to us later on.

**Corollary 6.4.** *Let  $A$  be a subset of  $[k]^n$  of uniform density  $\delta$ , let  $\eta > 0$ , let  $m \leq n^{1/4}$  and suppose that  $n \geq (16k/\eta)^{12}$ . Let  $J$  be a random subset of  $[n]$  of size  $m$  and let  $y$  be a random element of  $[k]^{\bar{J}}$ . Then the expected equal-slices density of  $A$  inside the combinatorial subspace  $S_{J,y}$  is at least  $\delta - \eta$ . In particular, there exist  $J$  and  $y$  such that the equal-slices density of  $A$  inside  $S_{J,y}$  is at least  $\delta - \eta$ .*

*Proof.* Let  $\nu$  be the equal-slices measure on  $[k]^m$  and apply Lemma 6.3. It implies that  $\mathbb{E}_{\sigma,y} \nu_{\sigma,y}(A) \geq \delta - \eta$ , from which it follows that there exists a pair  $(\sigma, y)$  such that  $\nu_{\sigma,y}(A) \geq \delta - \eta$ . But  $\nu_{\sigma,y}$  is the equal-slices measure on the combinatorial subspace  $S_{J,y}$ , where  $J = \sigma([m])$ , which is  $m$ -dimensional.  $\square$

For the next lemma we need some notation. Given a subset  $J \subset [n]$  of size  $m$  and a sequence  $y \in [k]^{\bar{J}}$ , let us write  $S'_{J,y}$  for the set of all sequences in  $S_{J,y}$  that never take the value  $k$  in  $J$ . Thus,  $S'_{J,y}$  is a copy of  $[k-1]^m$ . By the equal-slices density on  $S'_{J,y}$  we mean the image of the equal-slices density on  $[k-1]^m$  (where this is considered as a set in itself and not as a subset of  $[k]^m$ ).

**Corollary 6.5.** *Let  $A$  be a subset of  $[k]^n$  of uniform density  $\delta$ , let  $\eta > 0$ , let  $m \leq n^{1/4}$  and suppose that  $n \geq (16k/\eta)^{12}$ . Let  $J$  be a random subset of  $[n]$  of size  $m$  and let  $y$  be a random element of  $[k]^{\bar{J}}$ . Then the expected equal-slices density of  $A$  inside the set  $S'_{J,y}$  is at least  $\delta - \eta$ . In particular, there exist  $J$  and  $y$  such that the equal-slices density of  $A$  inside  $S'_{J,y}$  is at least  $\delta - \eta$ .*

*Proof.* Let  $\nu'$  be the measure on  $[k]^m$  defined by taking  $\nu'(A)$  to be the equal-slices measure of  $A \cap [k-1]^m$  (considered as a subset of  $[k-1]^m$ ). In other words,  $\nu'(A)$  is the probability that  $x \in A$  if you choose a random  $(k-1)$ -tuple  $(r_1, \dots, r_{k-1})$  of positive integers that add up to  $m$  and then let  $x$  be a random element of  $[k-1]^m$  with  $r_j$   $j$ s for each  $j$ .

Applying Lemma 6.3, we find that  $\mathbb{E}_{\sigma,y} \nu'_{\sigma,y}(A) \geq \delta - \eta$ , from which it follows that there exists a pair  $(\sigma, y)$  such that  $\nu'_{\sigma,y}(A) \geq \delta - \eta$ . But  $\nu'_{\sigma,y}$  is the equal-slices measure on the set  $S'_{J,y}$ , where  $J = \sigma([m])$ .  $\square$

**6.2. From equal-slices measure to uniform measure.** We would now like to go in the other direction, passing from a set of equal-slices density  $\delta$  to a subspace inside which the uniform density is at least  $\delta - \eta$  for some small  $\eta$ . As before, we need to use a result that says that a typical sequence  $x$  is not too imbalanced. Since we are choosing  $x$  from the equal-slices measure, the conclusion we can hope for is much weaker than the conclusion of Lemma 6.2: the result we use is Lemma 3.5, which tells us that with high probability every value will be taken a reasonable number of times.

The result we prove in this subsection states that if  $A$  has equal-slices density  $\delta$ , then there is a distribution on the  $m$ -dimensional subspaces of  $[k]^n$  such that if you choose one at random then the expected uniform density of  $A$  in that subspace is at least  $\delta - \beta$ .

**Lemma 6.6.** *Let  $\delta, \beta > 0$ , let  $m, n$  and  $k$  be positive integers, and suppose that  $m \leq \min\{\beta n/8k, \beta n/2k^2\}$ . Let  $A$  be a subset of  $[k]^n$  of equal-slices density  $\delta$ . Let  $J$  be a random subset of  $[n]$  of size  $[m]$ , let  $x$  be chosen uniformly at random from  $[k]^J$  and let  $y$  be chosen randomly, according to equal-slices measure, from  $[k]^{\overline{J}}$  (with this choice made independently of  $x$ ). Then the probability that  $(x, y) \in A$  is between  $\delta - \beta$  and  $\delta + \beta$ .*

*Proof.* Let  $z$  be an element of  $[k]^n$ . We shall estimate the probability that  $(x, y) = z$ , when  $x$  and  $y$  are chosen as in the statement of the theorem, and compare that with the equal-slices probability of the singleton  $\{z\}$ . To do this, let us define  $u_j$ , for each  $j \in [k]$ , to be the number of  $i$  such that  $x_i = j$ . Let us also assume that  $u_j \geq 1$  for every  $j$ .

We start by considering the case  $m = 1$ . In other words, we first pick a random  $i$  and randomly choose some  $j \in [k]$ . Then we randomly choose  $y$  from equal-slices measure on  $[k]^{[n] \setminus \{i\}}$ . And then we would like to know the probability that  $j = z_i$  and  $y_h = z_h$  for every  $h \neq i$ .

The probability that  $j = z_i$  is  $1/k$ , since we chose  $j$  uniformly. Now let us suppose that  $z_i$  is in fact equal to 1. Then the probability that  $y_h = z_h$  for every  $h \neq i$  is the equal-slices measure of a singleton that consists of a sequence in  $[k-1]^{n-1}$  with  $u_1 - 1$  1s and  $u_j$   $j$ s for every  $j > 1$ . That measure is equal to

$$\binom{n+k-2}{k-1}^{-1} \binom{n-1}{u_1-1, u_2, \dots, u_k}^{-1}$$

(It is here that we are assuming that  $u_1 \neq 0$ .) For comparison, the equal-slices measure of  $\{z\}$  in  $[k]^n$  is

$$\binom{n+k-1}{k-1}^{-1} \binom{n}{u_1, u_2, \dots, u_k}^{-1}.$$

The first measure divided by the second equals  $(n+k-1)/u_1$ .

It follows that the probability that  $(x, y) = z$  given that  $z_i = 1$  is  $(n+k-1)/ku_1$ . Therefore, by the law of total probability, the probability that  $(x, y) = z$  is

$$\sum_{j=1}^k \frac{1}{k} \frac{u_j}{n} \frac{n+k-1}{u_j} = \frac{n+k-1}{n}$$

times the equal-slices probability of  $z$ .

Now let us consider the more general case where  $|J| = m$ . Again we shall look at the probability that  $(x, y) = z$ , but this time we shall assume that  $u_j \geq m$  for every  $j$ . We claim that the probability of getting  $z$  is

$$r_{n,k,m} = \frac{(n+k-1)(n+k-2) \dots (n+k-m)}{n(n-1) \dots (n-m+1)}$$

times the equal-slices measure of  $\{z\}$ . This follows easily from what we have just done and induction. Indeed, by induction we know that if we choose a random set  $J'$  of size  $m-1$  and choose  $x$  uniformly from  $[k]^{J'}$  and  $y$  using equal-slices from  $[k]^{\overline{J'}}$ , then the probability that  $(x, y) = z$  is  $r_{n,k,m-1}$  times the equal-slices measure of  $\{z\}$ . If we now change the way we choose  $y$  by uniformly picking one coordinate and using equal-slices to pick the rest, then by the case  $m=1$  we multiply this probability by a further  $(n+k-m)/(n-m+1)$ , which gives us  $r_{n,k,m}$  times the equal-slices measure of  $\{z\}$ .



Now  $r_{n,k,m}$  is at least 1 and at most  $(1 + k/(n - m))^m$ . Given our assumption about  $m$ , this is at most  $1 + \beta/2$ . Thus, for every  $z$  with at least  $m$  coordinates of each value, the probability that  $(x, y) = z$  lies between  $\nu(\{z\})$  and  $(1 + \beta/2)\nu(\{z\})$ , where  $\nu$  is equal-slices measure. By Lemma 3.5, the equal-slices probability that  $z$  does not have at least  $m$  coordinates of each value is at most  $mk^2/n$ , which by assumption is at most  $\beta/2$ .

Now let  $A$  be any subset of  $[k]^n$  of density  $\delta$  and let  $B$  be the set of all sequences such that for some  $j$  there are fewer than  $m$  coordinates equal to  $j$ . Then if we choose  $(x, y)$  randomly in the manner stated, the probability that it belongs to  $A$  is at most  $(1 + \beta/2)\nu(A) + \beta/2$ , since the probability that it belongs to  $B$  is at most the equal-slices measure of  $B$  (as we see by looking at  $B^c$ ). The probability is also at least  $\nu(A) - \beta/2$ , for similar reasons. This proves the lemma.  $\square$

We now show that DHJ implies the equal-slices version of DHJ (which we stated earlier as Theorem 3.7).

**Corollary 6.7.** *Let  $k$  be a positive integer and suppose that  $DHJ_k$  is true. Let  $\delta > 0$  and let  $n \geq (16k^2/\delta)DHJ(k, \delta/2)$ . Then every subset of  $[k]^n$  of equal-slices density at least  $\delta$  contains a combinatorial line.*

*Proof.* By Lemma 6.6 there exists a combinatorial subspace  $V$  of dimension not less than  $DHJ(k, \delta/2)$  such that the uniform density of  $A$  in  $V$  is at least  $\delta/2$ . The result follows.  $\square$

**6.3. From uniform measure on  $[k]^n$  to uniform measure on  $[k - 1]^m$ .** We need one more result of a similar kind. This time it says that if we choose a random set  $J \subset [n]$  of size  $m$  and choose  $y$  uniformly at random from  $[k]^J$  and  $x$  uniformly at random from  $[k - 1]^J$ , then the distribution of  $(x, y)$  is approximately uniform. This can be proved as another almost immediate corollary of Lemma 6.3. However, we shall give a direct proof instead, since this case is an easy one and the proof is short.

**Lemma 6.8.** *Let  $\eta > 0$  and let  $m$  and  $n$  be positive integers with  $m \leq n^{1/4}$  and  $n \geq (12/\eta)^{12}$ . Let  $J$  be a random subset of  $[n]$  of size  $m$ , let  $y$  be a random element of  $[k]^J$  and let  $x$  be a random element of  $[k - 1]^J$  (in both cases chosen uniformly). Then the total variation distance between the resulting distribution on  $(x, y)$  and the uniform distribution on  $[k]^n$  is at most  $\eta$ .*

*Proof.* Let  $z$  be an element of  $[k]^n$ , let  $X$  be the set of coordinates  $i$  such that  $z_i = k$  and let  $r$  be the cardinality of  $X$ . By the proof of Lemma 6.2, the probability that  $r$  lies between  $n/k - n^{2/3}$  and  $n/k + n^{2/3}$  is at least  $1 - 2\exp(-2n^{1/3})$ , which is at least  $1 - \eta/3$ . Let us assume that  $z$  has this property. Now choose  $J$  and let us calculate the probability that  $(x, y) = z$  conditional on this choice of  $J$ .

If  $J \cap X \neq \emptyset$ , then the probability is zero. If, however, if  $J \cap X = \emptyset$  then it is  $(k - 1)^{-m}k^{-(m-n)}$ . The probability that  $J \cap X = \emptyset$  is  $\binom{n-r}{m}$ , which lies between  $(1 - 1/k - n^{-1/3} - m/n)^m$  and  $(1 - 1/k + n^{-1/3})^m$ . A simple calculation shows that it therefore lies between  $(1 - 1/k)^m(1 - 4n^{-1/12})$  and  $(1 - 1/k)^m(1 + 4n^{-1/12})$ . Therefore, the probability that  $(x, y) = z$  lies between  $k^{-n}(1 \pm \eta/3)$ .

Let  $B$  be the set of all  $z$  such that  $r$  does *not* lie between  $n/k - n^{2/3}$  and  $n/k + n^{2/3}$ . Then the probability that  $(x, y) \in B$  is at most  $1 - (1 - \eta/3)^2 \leq 2\eta/3$ . Therefore, if  $A$  is any subset of  $[k]^n$  and  $\delta$  is the density of  $A$ , the probability that  $(x, y) \in A$  lies between  $(\delta - \eta/3)(1 - \eta/3)$  and  $\delta(1 + \eta/3) + 2\eta/3$ , which proves the lemma.  $\square$

7. A DENSE SET WITH NO COMBINATORIAL LINE CORRELATES LOCALLY WITH AN INTERSECTION OF INSENSITIVE SETS

In this section we shall carry out the first three stages of the proof of  $\text{DHJ}_k$  (corresponding to the first three stages of the sketch proofs given earlier of the corners theorem and  $\text{DHJ}_3$ ).

**7.1. Finding a dense diagonal.** Let  $A$  be a subset of  $[k]^n$  of density  $\delta$ . The aim of this subsection is to find a combinatorial subspace  $V$  of  $[k]^n$  with two properties. First, the density of  $A$  inside  $V$  is not much smaller than  $\delta$ , and second, there are many points of  $A$  in  $V$  for which the variable coordinates take values in  $[k-1]$ . The densities in both cases are with respect to equal-slices measure. The second statement corresponds to the title of this subsection: this step is analogous to finding a dense diagonal in the corners proof. However, that proof gave us a dense structured set that was disjoint from  $A$ . Here, what we get is a structured set that is dense in a subspace. This will not help us at all unless  $A$  still has density almost  $\delta$  (or better) in that subspace. Thus, there is slightly more to this step than there was in the corners proof.

**Lemma 7.1.** *Let  $A \subset [k]^n$  be a set of uniform density  $\delta$ , let  $0 < \eta \leq \delta/4$ , let  $m \leq n^{1/4}$  and suppose that  $n \geq (16k/\eta)^{12}$ . Then there exists a pair  $(J, y)$ , where  $J$  is a subset of  $[n]$  of size  $m$  and  $y \in [k]^J$ , such that one of the following two possibilities holds:*

- (i) *the equal-slices density of  $A$  in the subspace  $S_{J,y}$  is at least  $\delta + \eta$ ;*
- (ii) *the equal-slices density of  $A$  in  $S_{J,y}$  is at least  $\delta - 4\eta\delta^{-1}$  and the equal-slices density of  $A$  in  $S'_{J,y}$  is at least  $\delta/4$ .*

*Proof.* By Corollary 6.4, if we choose  $J$  and  $y$  randomly then the expected equal-slices density of  $A$  in  $S_{J,y}$  is at least  $\delta - \eta$ . If the density is never more than  $\delta + \eta$ , then the probability that it is less than  $\delta - 4\eta\delta^{-1}$  is less than  $\delta/2$ , since otherwise the average would be at most

$$(1 - \delta/2)(\delta + \eta) + (\delta/2)(\delta - 4\eta\delta^{-1}) = \delta + (1 - \delta/2)\eta - 2\eta < \delta - \eta,$$

a contradiction.

By Corollary 6.5 the average density of  $A$  in a random set  $S'_{J,y}$  is at least  $\delta - \eta$ . Therefore, the probability that  $A$  has density less than  $\delta/4$  in  $S'_{J,y}$  is at most  $1 - \delta/2$ , since otherwise the average would be at most

$$\delta/2 + (1 - \delta/2)(\delta/4) < 3\delta/4 \leq \delta - \eta,$$

another contradiction.

It follows that if (i) does not hold then with positive probability (ii) holds.  $\square$

What Lemma 7.1 tells us is that either we can pass to a subspace and get a density increment of  $\eta$ , in which case we can move to the next stage of the iteration (after passing to a further subspace to convert this density increment into a uniform density increment), or we find a “dense diagonal” in a subspace in which  $A$  has not lost a significant amount of density.

**7.2. A “simple” locally dense set that is almost disjoint from  $A$ .** Let us suppose that the second possible conclusion of Lemma 7.1 holds (for an  $\eta$  that we are free to choose later). Then we have a combinatorial subspace  $V$  of  $m$  dimensions and  $A$  contains many points in  $V$  for which the variable coordinates are all in  $[k-1]$ . For simplicity, and without loss of generality, let us assume that  $V = [k]^m$ , and let us write  $A$  for  $A \cap V$ . So we are

given that  $A$  has equal-slices density at least  $\delta - \gamma$  (where  $\gamma = 4\eta\delta^{-1}$ ) and inside  $[k - 1]^m$  has equal-slices density at least  $\delta/4$ . Let us write  $B$  for  $A \cap [k - 1]^m$ . Finally, if  $x \in [k]^m$  and  $i, j \in [k]$ , let us write  $x^{i \rightarrow j}$  for the sequence that turns all the  $i$ s of  $x$  into  $j$ s.

**Lemma 7.2.** *Let  $A$  be a subset of  $[k]^m$  that contains no combinatorial line, and let  $B = A \cap [k - 1]^m$ . For each  $j \leq k - 1$  let  $C_j$  be the set  $\{x \in [k]^m : x^{k \rightarrow j} \in B\}$ . Then  $C_j$  is a  $jk$ -insensitive set, and  $A \cap C_1 \cap \dots \cap C_{k-1} \subset [k - 1]^m$ .*

*Proof.* Since the condition for belonging to  $C_j$  depends only on  $x^{k \rightarrow j}$ , it is trivial that  $C_j$  is  $jk$ -insensitive.

Suppose now that  $x \in C_1 \cap \dots \cap C_{k-1}$  and that at least one coordinate of  $x$  takes the value  $k$ . Let  $X$  be the set of coordinates where  $x = k$ . Then if you change all the coordinates in  $X$  to  $j$ , you end up with a point that belongs to  $A$ , since  $x \in C_j$ . Therefore, since  $A$  contains no combinatorial line, it follows that  $x$  itself does not belong to  $A$ .  $\square$

**Lemma 7.3.** *Let  $A, B$  and  $C_1, \dots, C_{k-1}$  be the subsets of  $[k]^m$  defined in Lemma 7.2 and let  $C = C_1 \cap \dots \cap C_{k-1}$ . Then for every  $\delta > 0$  there exists  $\theta > 0$  such that if  $B$  has equal-slices density at least  $\delta/4$  in  $[k - 1]^m$ , then  $C \setminus [k - 1]^m$  has equal-slices density at least  $\theta$  in  $[k]^m$ .*

*Proof.* There is a one-to-one correspondence between combinatorial lines in  $B$  and points in  $C \setminus [k - 1]^m$ . Moreover, this one-to-one correspondence preserves equal-slices measure (for the trivial reason that we defined the equal-slices measure on the set of combinatorial lines in  $[k - 1]^m$  by treating them as points in  $[k]^m$ ). By the probabilistic version of  $\text{DHJ}_{k-1}$  there exists  $\theta = \text{PDHJ}(k - 1, \delta/4) > 0$  such that the equal-slices density of combinatorial lines in  $B$  is at least  $\theta$ .  $\square$

From this lemma and Lemma 3.2 we see that  $\nu(A \cap C) \leq (k/\theta m)\nu(C)$ . (Recall that  $\nu$  is the equal-slices measure.) If  $m$  is large enough, that will be significantly less than  $\delta$ . This is the sense in which  $A$  is ‘‘almost disjoint’’ from  $C$ .

### 7.3. A ‘‘simple’’ locally dense set that correlates with $A$ .

**Lemma 7.4.** *Let  $A, B$  and  $C_1, \dots, C_{k-1}$  be the subsets of  $[k]^m$  defined in Lemma 7.2, let  $C = C_1 \cap \dots \cap C_{k-1}$ , and suppose that  $C$  has density  $\theta$ . Let  $0 < \gamma \leq \delta/4$  and suppose also that  $\nu(A) \geq \delta - \gamma$  and that  $\nu(A \cap C) \leq (\delta/2)\nu(C)$ . Then there exist sets  $D_1, \dots, D_{k-1}$  such that  $D_i$  is  $ik$ -insensitive for each  $i$  and such that  $\nu(A \cap D) \geq (\delta - \gamma)\nu(D) + \delta\theta/4k$ , where  $D = D_1 \cap \dots \cap D_{k-1}$ .*

*Proof.* We begin with the observation that

$$[k]^m = \bigcup_{i=1}^k C_1 \cap \dots \cap C_{i-1} \cap C_i^c \cap \dots \cap C_{k-1}^c,$$

and that this union is in fact a partition of  $[k]^m$ . For each  $i$  let us write  $D^{(i)}$  for the set  $C_1 \cap \dots \cap C_{i-1} \cap C_i^c \cap \dots \cap C_{k-1}^c$ . Then  $D^{(k)} = C$ . From our assumptions, we know that

$$\begin{aligned} \nu(A \cap (D^{(1)} \cup \dots \cup D^{(k-1)})) &\geq \delta - \gamma - (\delta/2)\nu(D^{(k)}) \\ &= (\delta - \gamma)(1 - \nu(D^{(k)})) + (\delta/2 - \gamma)\nu(D^{(k)}) \\ &\geq (\delta - \gamma)(1 - \nu(D^{(k)})) + \delta\theta/4. \end{aligned}$$

Since  $1 - \nu(D^{(k)}) = \nu(D^{(1)} \cup \dots \cup D^{(k-1)})$ , it follows by averaging that there exists  $i$  such that  $\nu(A \cap D^{(i)}) \geq \delta - \gamma + \delta\theta/4(k-1)$ . Since both  $C_i$  and  $C_i^c$  are  $ik$ -insensitive, this proves the lemma.  $\square$

Now for the next part of our argument we need to use the uniform measure. In order to do this, we must use our measure-transfer results again. Basically, all we do is randomly restrict to a small subspace  $V$  with the uniform measure on it and apply Lemma 6.6, but that is not quite the whole story since we want *two* things to happen: that the relative density of  $A \cap D \cap V$  inside  $D \cap V$  is still bigger than  $\delta$ , and also that the relative density of  $D \cap V$  inside  $V$  is not too small.

**Lemma 7.5.** *Let  $\beta > 0$  and let  $k, r$  and  $m$  be positive integers such that  $r \leq \min\{\beta m/8k, \beta m/2k^2\}$ . Let  $A$  and  $D$  be subsets of  $[k]^m$  such that  $\nu(A \cap D) \geq (\delta - \gamma)\nu(D) + 3\beta$ . Then there exists a combinatorial subspace  $V$  of  $[k]^m$  of dimension  $r$  such that  $\mu_V(D \cap V) \geq \gamma\mu(V)$  and  $\mu_V(A \cap D \cap V) \geq (\delta - \gamma)\mu_V(D \cap V) + \beta$ , where  $\mu_V$  is the uniform probability measure on  $V$ .*

*Proof.* Let us choose  $V$  by randomly choosing a set  $J \subset [m]$  of size  $r$ , randomly choosing  $y \in [k]^J$  using equal-slices measure, and taking the subspace  $S_{J,y}$ . By Lemma 6.6, the expectation of  $\mu_V(A \cap D \cap V) - (\delta - \gamma)\mu_V(D \cap V)$  is at least  $\nu(A \cap D) - \beta - (\delta - \gamma)\nu(D) - \beta$ , which is at least  $\beta$  by our assumed lower bound for  $\nu(A \cap D)$ .  $\square$

Note that the conclusion of the lemma implies that  $\mu_V(D \cap V)$  is at least  $\eta$ .

Let us now put together the results of this section.

**Lemma 7.6.** *Let  $\delta > 0$ , let  $k$  be a positive integer, let  $\theta = \text{PDHJ}(k-1, \delta/4)$ , let  $\eta = \delta^2\theta/96k$ , let  $\beta = \delta\theta/12k$  and let  $\gamma = 4\delta^{-1}\eta = \delta\theta/24k = \beta/2$ . Let  $n$  be a positive integer, let  $m = \lfloor n^{1/4} \rfloor$ , let  $r = \lfloor \beta m/8k^2 \rfloor$  and suppose that  $n \geq (16k/\eta)^{12}$ . Let  $A$  be a subset of  $[k]^n$  of uniform density  $\delta$ . Then either  $A$  contains a combinatorial line or there is an  $r$ -dimensional combinatorial subspace  $W$  of  $[k]^n$  and sets  $D_1, \dots, D_{k-1} \subset W$  such that  $D_j$  is  $jk$ -insensitive for each  $j$ , and such that if we set  $D$  to be  $D_1 \cap \dots \cap D_{k-1}$ , then  $\mu_W(D) \geq \gamma$  and  $\mu_W(A \cap D) \geq (\delta + \gamma)\mu_W(D)$ .*

*Proof.* Let  $m = \lfloor n^{1/4} \rfloor$ . Then, by Lemma 7.1, either there is an  $m$ -dimensional subspace  $V$  such that  $\mu_V(A) \geq \delta + \eta$ , in which case we are done (since we can pass to a random  $r$ -dimensional subspace of  $V$  and on average we will have the same density increment) or there is an  $m$ -dimensional subspace  $V$  such that the equal-slices density of  $A$  in  $V$  is at least  $\delta - 4\eta\delta^{-1}$  and the equal-slices density of  $A$  in  $V'$  is at least  $\delta/4$ , where  $V'$  is the set of points in  $V$  with no variable coordinate equal to  $k$ .

Let  $B = A \cap V'$ . Then Lemma 7.3 gives us a  $\theta > 0$  and sets  $C_1, \dots, C_{k-1}$  such that  $C_i$  is  $ik$ -insensitive, the intersection  $C = C_1 \cap \dots \cap C_{k-1}$  is such that  $C \setminus V'$  has equal-slices density at least  $\theta$ , and  $C \setminus V'$  is disjoint from  $A$ . The value of  $\theta$  can be taken to be  $\text{PDHJ}(k-1, \delta/4)$ .

Let  $\gamma = 4\eta\delta^{-1} = \beta/2$ . It is easily checked that  $k/\theta m \leq \delta/2$  and that  $\delta\theta/4k \geq 2\gamma$ . Therefore, Lemma 7.4 tells us that we can find sets  $D_1, \dots, D_{k-1}$  such that  $D_i$  is  $ik$ -insensitive, and such that if  $D = D_1 \cap \dots \cap D_{k-1}$ , then  $\nu(A \cap D) \geq (\delta - \gamma)\nu(D) + \delta\theta/4k$ .

Finally, Lemma 7.5 with  $\beta = \delta\theta/12k$  gives us an  $r$ -dimensional subspace  $W$  of  $V$  such that  $\mu_W(A \cap D \cap W) \geq (\delta - \gamma)\mu_W(D \cap W) + \beta$ . This implies that  $\mu_W(A \cap D \cap W) \geq (\delta + \gamma)\mu_W(D \cap W)$  and that  $\mu_W(D \cap W) \geq \gamma$ , as claimed.  $\square$

8. ALMOST PARTITIONING LOW-COMPLEXITY SETS INTO SUBSPACES

We have completed one of the two main stages of the proof, which corresponds to the first three steps of the proof we sketched of the corners theorem (and also to the first three steps of our sketch proof of DHJ<sub>3</sub>). In this section we shall carry out a task that corresponds to the next two steps. So far, we have obtained a density increment on a dense subset  $D$  of a subspace  $W$ . This helps us, because  $D$  is an intersection of  $ik$ -insensitive sets, and therefore has low complexity, in a certain useful sense. Our job now is to show that low-complexity sets can be almost completely partitioned into combinatorial subspaces with dimension tending to infinity. To prove this, we shall follow the scheme of argument presented in Section 5.4. (That argument was presented for the case  $k = 3$ , but it can be straightforwardly generalized.)

8.1. A  $1k$ -insensitive set can be almost entirely partitioned into large subspaces.

We begin by proving the result for  $1k$ -insensitive sets, and hence for  $jk$ -insensitive sets whenever  $j < k$ . It will then be straightforward to deduce the result for intersections of such sets.

**Lemma 8.1.** *Let  $\eta > 0$ , and let  $d, m$  and  $n$  be positive integers with  $m \geq MDHJ_{k-1}(d, \eta)$  and  $n \geq \eta^{-1}m(k+d)^m$ . Let  $D$  be a  $1k$ -insensitive subset of  $[k]^n$ . Then there are disjoint combinatorial subspaces  $V_1, \dots, V_N$ , each of which has dimension  $d$  and is a subset of  $D$ , such that  $\mu(V_1 \cup \dots \cup V_N) \geq \mu(D) - 3\eta$ .*

*Proof.* Let us write a typical element of  $[k]^n$  as  $(x, y)$ , where  $x \in [k]^m$  and  $y \in [k]^{n-m}$ . For each  $y$  let us write  $D_y$  for the set  $\{x \in [k]^m : (x, y) \in D\}$  and  $E_y$  for the set  $\{x \in [k-1]^m : (x, y) \in D\} = E_y \cap [k-1]^m$ . Then by Lemma 6.8 the average density of the sets  $E_y$  is at least  $\gamma - \eta \geq 2\eta$ . It follows that the density of  $y$  such that  $E_y$  has density at least  $\eta$  (in  $[k-1]^m$ ) is at least  $\eta$ .

If  $E_y$  has density at least  $\eta$ , then by our assumption about  $m$  it follows that it contains a  $d$ -dimensional combinatorial subspace  $U'_y$  (where this means a subspace of  $[k-1]^m$ ). Since  $D$  is  $1k$ -insensitive, and therefore so is  $D_y$ , it follows that  $D_y$  contains a  $d$ -dimensional combinatorial subspace  $U_y$  (where this means a subspace of  $[k]^m$ ).

The number of possible  $d$ -dimensional subspaces of  $[k]^m$  is at most  $(k+d)^m$  (since we have to decide for each coordinate  $i \in [m]$  whether to give it a fixed value in  $[k]$  or to put it into one of the  $d$  wildcard sets), so by the pigeonhole principle there must exist a subspace  $U \subset [k]^m$  such that the set  $T = \{y \in [k]^{n-m} : U \times \{y\} \subset D\}$  has density at least  $\eta(k+d)^{-m}$ . Since  $D$  is  $1k$ -insensitive, it follows that  $T$  is also  $1k$ -insensitive.

The set  $U \times T$  is a subset of  $D$  of density at least  $\eta(k+d)^{-m}$ , and it is a union of the  $d$ -dimensional subspaces  $U \times \{y\}$  with  $y \in T$ . We now remove  $U \times T$  from  $D$ .

The resulting set  $D_1 = D \setminus (U \times T)$  is not necessarily  $1k$ -insensitive, but for every  $x \in [k]^m$  the set  $\{y : (x, y) \in D_1\}$  is  $1k$ -insensitive: this follows immediately from the fact that both  $D$  and  $T$  are  $1k$ -insensitive. Thus, we can at least partition  $[k]^n$  into subspaces inside each of which  $D_1$  is  $1k$ -insensitive.

This gives us the basis for an inductive argument. The inductive hypothesis is that  $D_r$  is a set of density at least  $2\eta$  such that for every  $x \in [k]^{rm}$  the set  $\{y \in [k]^{n-rm} : (x, y) \in D_r\}$  is  $1k$ -insensitive, and that  $D \setminus D_r$  is a union of  $d$ -dimensional subspaces of density at least  $\eta(k+d)^{-m}$ . We have essentially just given the proof of the inductive step, but we need to generalize the argument very slightly.

To do this, let us write a typical element of  $D_r$  as  $(x, y, z)$  with  $x \in [k]^{rm}$ ,  $y \in [k]^m$  and  $z \in [k]^{n-(r+1)m}$ . For each  $x \in [k]^{rm}$  let  $(D_r)_x$  be  $\{(y, z) \in [k]^{n-rm} : (x, y, z) \in D_r\}$  and for each pair  $(x, z)$ , let  $(E_r)_{x,z}$  be the set  $\{y \in [k-1]^m : (x, y, z) \in D_r\}$ . Then the average density of the sets  $(D_r)_x$  is the density of  $D_r$ , which is at least  $3\eta$ . It follows from Lemma 6.8 that the average density of the sets  $(E_r)_{x,z}$  is at least  $2\eta$ , provided that  $n - rm \geq (12/\eta)^{12}$ . Therefore, the density of pairs  $(x, z)$  such that  $(E_r)_{x,z}$  has density at least  $\eta$  is at least  $\eta$ .

If  $(E_r)_{x,z}$  has density at least  $\eta$ , then it contains a  $d$ -dimensional combinatorial subspace  $U'_{x,z}$ , where this is a subspace of  $[k-1]^m$ . Since  $(D_r)_x$  is  $1k$ -insensitive, it follows that it also contains a  $d$ -dimensional combinatorial subspace  $U_{x,z}$ , where this time we mean a subspace of  $[k]^m$ . By the pigeonhole principle there is a  $d$ -dimensional subspace  $U \subset [k]^m$  such that the set  $T = \{(x, z) \in [k]^{rm} \times [k]^{n-(r+1)m} : \{x\} \times U \times \{z\} \subset D_r\}$  has density at least  $\eta(k+d)^{-m}$ .

Let  $D_{r+1} = D_r \setminus T \times U$  (where we interpret  $T \times U$  to mean  $\{(x, y, z) : (x, z) \in T, y \in U\}$ ). Then  $T \times U$  is a union of  $d$ -dimensional subspaces of density at least  $\eta(k+d)^{-m}$ , and for every  $(x, y)$  the set  $\{z \in [k]^{n-(r+1)m} : (x, y, z) \in D_{r+1}\}$  is  $1k$ -insensitive.

Clearly we cannot iterate this process more than  $\eta^{-1}(k+d)^m$  times. Therefore, since  $n \geq \eta^{-1}m(k+d)^m$ , it follows that we can write  $D$  as a disjoint union of  $d$ -dimensional combinatorial subspaces and a residual set of density at most  $3\eta$ , as claimed.  $\square$

**8.2. An intersection of  $jk$ -insensitive sets can be almost entirely partitioned into large subspaces.** The main result of this subsection is a very straightforward consequence of Lemma 8.1. Let  $F$  be the function that bounds  $n$  in terms of  $d$  in that lemma (and also  $\eta$  and  $k$ , which we shall regard as fixed): that is,  $F(d) = \lceil \eta^{-1}m(k+d)^m \rceil$ , where  $m = \text{MDHJ}_{k-1}(d, \eta)$ . Let  $F^{(k-1)}(d)$  denote the result of applying  $F$  to  $d$   $k-1$  times.

**Lemma 8.2.** *Let  $\eta > 0$ , and let  $d$  and  $n$  be positive integers such that  $n \geq F^{(k-1)}(d)$ . For each  $j \in [k-1]$  let  $D_j$  be a  $jk$ -insensitive subset of  $[k]^n$  and let  $D = D_1 \cap \dots \cap D_{k-1}$ . Then there are disjoint combinatorial subspaces  $V_1, \dots, V_N$ , each of which has dimension  $d$  and is a subset of  $D$ , such that  $\mu(V_1 \cup \dots \cup V_N) \geq \mu(D) - 3(k-1)\eta$ .*

*Proof.* We prove the result by induction on the number of insensitive sets in the intersection (which is not quite the same as proving it by induction on  $k$ ). That is, we prove by induction that if  $n \geq F^{(j)}(d)$  then the conclusion of the lemma holds for  $D^{(j)} = D_1 \cap \dots \cap D_j$  and with an error of at most  $3j\eta$  instead of  $3(k-1)\eta$ .

Lemma 8.1 does the case  $j = 1$ . In general, if we have the result for  $j-1$ , then let  $n \geq F^{(j)}(d) = F(F^{(j-1)}(d))$ . Then by Lemma 8.1 we can partition  $D_j$  into combinatorial subspaces  $V_1, \dots, V_N$  of dimension  $F^{(j-1)}(d)$  together with a residual set of density at most  $3\eta$ . The intersection of any  $D_h$  with any  $V_i$  is  $hk$ -insensitive, and  $V_i \subset D_j$ , so

$$D^{(j)} \cap V_i = D^{(j-1)} \cap V_i = (D_1 \cap V_i) \cap \dots \cap (D_{j-1} \cap V_i)$$

is an intersection of insensitive sets to which we can apply the inductive hypothesis.

That allows us to partition each  $V_i$  into combinatorial subspaces  $V_{is}$  of dimension  $d$  together with a residual set of relative density (in  $V_i$ ) at most  $3(j-1)\eta$ . The union of these new residual sets has density at most  $3(j-1)\eta$  in  $[k]^n$  (since the subspaces  $V_i$  are disjoint), so we have partitioned  $D^{(j)}$  into a union of  $d$ -dimensional combinatorial subspaces together with a residual set of density at most  $3j\eta$ . This completes the inductive step.  $\square$

9. COMPLETING THE PROOF

At this stage our argument is essentially finished. In this section we shall spell out why our lemmas show that  $\text{DHJ}_k$  follows from  $\text{DHJ}_{k-1}$ . We shall begin with a qualitative argument. After that, we shall informally discuss how the bounds we obtain for  $\text{DHJ}_k$  depend on those that we obtain for  $\text{DHJ}_{k-1}$ . Finally, we shall exploit the fact that we have good bounds when  $k = 2$  to give a more careful analysis of the bounds we obtain for  $\text{DHJ}_3$ , which turn out to be of tower type.

**9.1. Proof that  $\text{DHJ}_{k-1}$  implies  $\text{DHJ}_k$ .** Let  $A \subset [k]^n$  be a set of density  $\delta$ . Our aim will be to find a combinatorial subspace  $V$  of dimension tending to infinity with  $n$  such that the relative density of  $A \cap V$  in  $V$  is at least  $\delta + c$ , where  $c$  depends only on  $\delta$  and  $k$ . If we can do that, then we will be able to apply a simple iterative argument to complete the proof.

Lemma 7.6 says that either  $A$  contains a combinatorial line or we can find an  $r$ -dimensional subspace  $W$  and subsets  $D_1, \dots, D_{k-1}$  of  $W$  such that if  $D = D_1 \cap \dots \cap D_{k-1}$  then  $\mu_W(D)$  (the density of  $D$  inside  $W$ ) is at least  $\gamma$  and  $\mu_W(A \cap D) \geq (\delta + \gamma)\mu(D)$ . Here,  $r$  tends to infinity with  $n$  for given  $\delta$  and  $k$  (and increases as  $\delta$  increases), and  $\gamma$  is a parameter that depends on  $\delta$  and  $k$  only. To be precise, if we let  $\theta = \text{PDHJ}(k-1, \delta/4)$ , then we can take  $\gamma = \delta\theta/24k$  and  $r = \lfloor \delta\theta \lfloor n^{1/4} \rfloor / 96k^3 \rfloor$ . Thus, this step depends on the fact that  $\text{DHJ}_{k-1}$  implies  $\text{PDHJ}_{k-1}$ .

Now apply Lemma 8.2 with  $[k]^n$  replaced by the  $r$ -dimensional subspace  $W$  and with  $\eta = \gamma^2/6(k-1)$ . Then we can find disjoint combinatorial subspaces  $V_1, \dots, V_N$  of  $W$  such that each has dimension equal to the largest  $d$  for which  $r \geq F^{(k-1)}(d)$ , each is a subset of  $D$ , and  $\mu_W(V_1 \cup \dots \cup V_N) \geq \mu_W(D) - \gamma^2/2$ . Here  $d$  depends on  $\eta$  and  $k$  as well as  $r$  (the dependence was suppressed in our notation for the function  $F$ ) and tends to infinity as  $r$  tends to infinity. The function  $F$  is defined in terms of the function  $\text{MDHJ}_{k-1}$ , so this step depends on the fact that  $\text{DHJ}_{k-1}$  implies  $\text{MDHJ}_{k-1}$ .

It follows that

$$\begin{aligned} \mu_W(A \cap (V_1 \cup \dots \cup V_M)) &\geq (\delta + \gamma)\mu(D) - \gamma^2/2 \\ &\geq (\delta + \gamma/2)\mu(D) \\ &\geq (\delta + \gamma/2)\mu_W(V_1 \cup \dots \cup V_M). \end{aligned}$$

Therefore, by averaging there must be some  $i$  such that  $\mu_W(A \cap V_i) \geq (\delta + \gamma/2)\mu(V_i)$ .

Since  $d$ , the dimension of  $W_i$  tends to infinity with  $r$  and  $r$  tends to infinity with  $n$ , and since  $\gamma$  depends on  $\delta$  and  $k$  only, we have found our desired density increment on a subspace. We may now repeat the argument. Either  $A \cap V_i$  contains a combinatorial line, or we can pass to a further subspace (with dimension tending to infinity with  $d$  and hence with  $n$ ) inside which the relative density is at least  $\delta + \gamma$ . (In fact, we can do slightly better, since we have now replaced  $\delta$  by  $\delta + \gamma/2$  so the density increment at this second stage will be better than  $\gamma/2$ .) Since the density of  $A$  inside any subspace is always at most 1, there can be at most  $2/\gamma$  iterations of this procedure before we eventually find a combinatorial line. Since this number of iterations depends only on  $\delta$  and  $k$ , if the original  $n$  is large enough,  $A$  must have contained a combinatorial line.

Since  $\text{DHJ}_1$  is trivial and  $\text{DHJ}_2$  follows from Sperner's theorem, the proof of the general case of  $\text{DHJ}$  is complete.

**9.2. What bound comes out of the above argument?** Let us briefly consider how the bound that we obtain for  $\text{DHJ}_k$  relates to the bound that we obtain for  $\text{DHJ}_{k-1}$ .

We note first that  $\text{EDHJ}(k-1, \delta)$  is bounded above by  $(16k^2/\delta)\text{DHJ}(k, \delta/2)$ , by Corollary 6.7 (but all we really care about for the purposes of this discussion is that the two functions are of broadly similar type). Next, recall from Theorem 3.7 that if  $A \subset [k-1]^n$  has equal-slices density at least  $\delta$ , then the equal-slices density of the set of combinatorial lines in  $A$  is at least  $(\delta/9)k^{-m}$ , where  $m = \text{EDHJ}(k-1, \delta/4)$ . That is,  $\text{PDHJ}(k-1, \delta)$  is exponentially small as a function of  $\text{EDHJ}(k-1, \delta)$ , and hence as a function of  $\text{DHJ}(k-1, \delta)$ . In particular, if  $\text{DHJ}(k-1, \delta)$  is already a tower-type function, then  $\text{PDHJ}(k-1, \delta)$  behaves broadly like the reciprocal of  $\text{DHJ}(k-1, \delta)$ . It follows that the subspace we pass to in Lemma 7.6 has dimension broadly comparable to  $n/\text{DHJ}(k-1, \delta)$ . Equivalently, if we want to pass to an  $r$ -dimensional subspace then we need  $n$  to be at least  $r\text{DHJ}(k-1, \delta)$  or so.

The next step depends on  $\text{MDHJ}_{k-1}$ , and this is where things get very expensive. The proof we gave of  $\text{MDHJ}_{k-1}$  yields a bound that is obtained as follows. Define  $G_{k-1}(x)$  to be  $\exp(\text{DHJ}(k-1, 1/x))$ . Then  $\text{MDHJ}(k-1, d, \delta)$  is bounded above by  $G_{k-1}^{(d)}(1/\delta)$ , where  $G_{k-1}^{(d)}$  is the  $d$ -fold iteration of  $G_{k-1}$ . The function  $F$  that comes into Lemma 8.2 is broadly comparable to  $\text{MDHJ}(k-1, d, \delta)$  (again, assuming that  $\text{MDHJ}(k-1, d, \delta)$  is at least of tower type), so  $F^{(k-1)}$  is something like  $G_{k-1}^{(d(k-1))}$ .

This function is so much bigger than the function  $r \mapsto \text{DHJ}(k-1, \delta)$  that we can more or less ignore the former. Therefore, if we want to end up with a  $d$ -dimensional subspace after one round of the main iteration, we need to start with  $n$  being something like the  $d(k-1)$ -fold iteration of a function that has similar behaviour to the function  $G_{k-1}$  defined above, which is pretty similar to the function  $d \mapsto \text{DHJ}(k-1, 1/d)$ . We then have to run the whole iteration  $2/\gamma$  times, where  $\gamma$  is broadly comparable to  $\text{DHJ}(k-1, \delta)^{-1}$ . So eventually we need  $n$  to be larger than  $(k-1)d\text{DHJ}(k-1, \delta)$  iterations of the function  $d \mapsto \text{DHJ}(k-1, 1/d)$ , which is roughly  $\text{DHJ}(k-1, \delta)$  iterations.

To rephrase slightly, if we let  $\text{RDHJ}_{k-1}(s) = \text{DHJ}(k-1, 1/s)$  (the ‘‘R’’ stands for ‘‘reciprocal’’ here), then  $\text{RDHJ}_k(s)$  is obtained by iterating the function  $\text{RDHJ}_{k-1}$  roughly  $\text{RDHJ}_{k-1}(s)$  times.

This means that as  $k$  increases by 1, the function  $\text{RDHJ}_k$  goes up by one level in the Ackermann hierarchy. (It is bigger than the corresponding level of the Ackermann function, but not in an interesting way.)

**9.3. Bounds for  $\text{DHJ}_3$ .** When  $k = 3$ , we can obtain much better bounds because in this case we have reasonable bounds for  $\text{MDHJ}_{k-1}$ . Let us therefore do the analysis a little more carefully.

First, note that Theorems 3.1 and 2.3 tell us that we can take  $\text{PDHJ}(2, \delta)$  to be  $\delta^2/2$  and  $\text{MDHJ}(2, d, \delta)$  to be  $25\delta^{-2^d}$ . Therefore, returning to the argument given in §9.1 and setting  $k = 3$ , we can take  $\theta$  to be  $\delta^2/32$ ,  $\gamma = \delta^3/2304$ , and  $r = \lfloor \delta^3 \lfloor n^{1/4} \rfloor / 41472 \rfloor$ .

We apply Lemma 8.2 with  $\eta = \gamma^2/6(k-1) = \delta^6/12(2304)^2$ , which is at least  $\delta^6/2^{27}$ . Therefore,  $\text{MDHJ}(2, d, \eta)$  is at most  $25(2^{27}\delta^{-6})^{2^d}$ , and if  $d \geq 10$ , say, then  $F(d)$  can be bounded above by  $2 \uparrow \delta^{-1} \uparrow 2 \uparrow 2d$ , where the symbol  $\uparrow$  denotes exponentiation and  $x \uparrow y \uparrow z$  means  $x \uparrow (y \uparrow z)$ . It follows that  $F^{(2)}(d)$  is at most  $2 \uparrow \delta^{-1} \uparrow 2 \uparrow 2 \uparrow \delta^{-1} \uparrow 2 \uparrow 3d$ . (The final 3 instead of 2 is to (over)compensate for losing a factor of 2 earlier on in the tower.)



We may therefore take  $d$  to be  $\delta \log^{(6)} r$ , where  $\log^{(6)}$  is the six-fold iterated logarithm. In fact, the factor of  $\delta$  is unduly generous, so, bearing in mind our bound for  $r$  in terms of  $n$ , it is safe to take  $d$  to be  $(\delta/2) \log^{(6)} n$ . (Strictly speaking, we need to assume that  $n$  is sufficiently large, but if we are generous later then this requirement will be met by a huge margin.)

The number of iterations we need is certainly no more than  $2304/\delta^3$ , but we can in fact do slightly better. It takes at most  $2304/\delta^2$  iterations for the density to increase from  $\delta$  to  $2\delta$ . Therefore, the total number of iterations is at most  $2304\delta^{-2}(1+1/4+1/16+\dots) = 3072\delta^{-2}$ . It follows that DHJ(3,  $\delta$ ) is bounded above by a tower of 2s of height  $20000\delta^{-2}$ . (Since  $20000 > 6 \times 3072$ , the dimension of the space will still be vast when the iterations come to an end.) This proves the estimate claimed in Theorem 1.5.

## REFERENCES

- [Aus09] Tim Austin. Deducing the Density Hales–Jewett Theorem from an infinitary removal lemma. <http://arxiv.org/abs/0903.1633>, 2009.
- [Beh46] Felix A. Behrend. On sets of integers which contain no three terms in arithmetical progression. *Proceedings of the National Academy of Sciences*, 32(12):331–332, 1946.
- [BL96] Vitaly Bergelson and Alexander Leibman. Polynomial extensions of van der Waerden’s and Szemerédi’s theorems. *Journal of the American Mathematical Society*, 9(3):725–753, 1996.
- [FK78] Hillel Furstenberg and Yitzhak Katznelson. An ergodic Szemerédi theorem for commuting transformations. *Journal d’Analyse Mathématique*, 34(1):275–291, 1978.
- [FK89] Hillel Furstenberg and Yitzhak Katznelson. A density version of the Hales–Jewett Theorem for  $k = 3$ . *Discrete Mathematics*, 75:227–241, 1989.
- [FK91] Hillel Furstenberg and Yitzhak Katznelson. A density version of the Hales–Jewett Theorem. *Journal d’Analyse Mathématique*, 57:64–119, 1991.
- [FKO82] Hillel Furstenberg, Yitzhak Katznelson, and Donald Ornstein. The ergodic theoretical proof of szemerédi’s theorem. *Bulletin of the American Mathematical Society*, 7(3):527–552, 1982.
- [Fur77] Hillel Furstenberg. Ergodic behavior of diagonal measures and a theorem of Szemerédi on arithmetic progressions. *Journal d’Analyse Mathématique*, 31(1):204–256, 1977.
- [Fur85] Hillel Furstenberg. An ergodic Szemerédi theorem for IP-systems and combinatorial theory. *Journal d’Analyse Mathématique*, 45(1):117–168, 1985.
- [Gow01] W. Timothy Gowers. A new proof of Szemerédi’s theorem. *Geometric and Functional Analysis*, 11(3):465–588, 2001.
- [Gow06] W. Timothy Gowers. Quasirandomness, counting and regularity for 3-uniform hypergraphs. *Combinatorics, Probability and Computing*, 15(1–2):143–184, 2006.
- [Gow07] W. Timothy Gowers. Hypergraph regularity and the multidimensional Szemerédi theorem. *Annals of Mathematics*, 166(3):897–946, 2007.
- [HJ63] Alfred W. Hales and Robert I. Jewett. Regularity and positional games. *Transactions of the American Mathematical Society*, 106(2):222–229, 1963.
- [LM08] Michael T. Lacey and William McClain. Three dimensional corners: A Box Norm proof. <http://arxiv.org/abs/0804.3019>, 2008.
- [NRS06] Brendan Nagle, Vojtěch Rödl, and Mathias Schacht. The counting lemma for regular  $k$ -uniform hypergraphs. *Random Structures and Algorithms*, 28(2):113–179, 2006.
- [Pol09] D. H. J. Polymath. Density Hales–Jewett and Moser numbers in low dimensions. Unpublished, <http://michaelnielsen.org/polymath1/>, 2009.
- [Rot53] Klaus F. Roth. On certain sets of integers. *Journal of the London Mathematical Society*, 28(1):104–109, 1953.
- [RS04] Vojtěch Rödl and Jozef Skokan. Regularity lemma for  $k$ -uniform hypergraphs. *Random Structures and Algorithms*, 25(1):1–42, 2004.
- [RS06] Vojtěch Rödl and Jozef Skokan. Applications of the regularity lemma for uniform hypergraphs. *Random Structures and Algorithms*, 28(2):180–194, 2006.

- [RS07a] Vojtěch Rödl and Mathias Schacht. Regular partitions of hypergraphs: Counting lemmas. *Combinatorics, Probability and Computing*, 16(6):887–901, 2007.
- [RS07b] Vojtěch Rödl and Mathias Schacht. Regular partitions of hypergraphs: regularity lemmas. *Combinatorics, Probability and Computing*, 16(6):833–885, 2007.
- [Shk06a] Ilya D. Shkredov. On a generalization of Szemerédi’s theorem. *Proceedings of the London Mathematical Society*, 93(3):723–760, 2006.
- [Shk06b] Ilya D. Shkredov. On a problem of Gowers. *Izvestiya: Mathematics*, 70(2):385–425, 2006.
- [Spe28] Emanuel Sperner. Ein Satz über Untermengen einer endlichen Menge. *Mathematische Zeitschrift*, 27(1):544–548, 1928.
- [Sze75] Endre Szemerédi. On sets of integers containing no  $k$  elements in arithmetic progression. *Acta Arithmetica*, 27:199–245, 1975.
- [Tao06] Terence Tao. A quantitative ergodic theory proof of Szemerédi’s theorem. *The Electronic Journal of Combinatorics*, 13(1), 2006.
- [Tao07] Terence Tao. A correspondence principle between (hyper)graph theory and probability theory, and the (hyper)graph removal lemma. *Journal d’Analyse Mathématique*, 103(1):1–45, 2007.
- [vdW27] Bartel L. van der Waerden. Beweis einer Baudetschen Vermutung. *Nieuw Archief voor Wiskunde*, 15:212–216, 1927.