## Lecture 14 Bounds by Counting Dimensions

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## 1 Multivariate Polynomials

The following lemma generalizes the fact that there is a bijection between univariate polynomials of degree $d$ and their evaluations on $d+1$ points, the case of multivariate polynomials.

Lemma 1. If $S_{1}, S_{2}, \ldots S_{n} \subset \mathbb{F}$ are all sets of size $d+1$, and $P_{d}$ is the set of polynomials whose degree in any variable is at most $d$, then there is a bijection between polynomials $p \in P_{d}$, their evaluations $\mathbb{F}^{(d+1)^{n}}$, where the coordinates of the evaluations are indexed by elements of $S_{1} \times \cdots \times S_{n}$, and the coordinate that corresponds to $x$ is $p(x)$.

Proof We shall show that the map from polynomials to evaluations is a full rank linear transformation. Given any point $\left(\alpha_{1}, \alpha_{2}, \ldots, \alpha_{n}\right) \in S_{1} \times \cdots \times S_{n}$, we define the polys $f_{i}\left(X_{i}\right)=$ $\prod_{\alpha \in S_{i}, \alpha \neq \alpha_{i}} \frac{X_{i}-\alpha}{\alpha_{i}-\alpha}$. And define $f_{\alpha}=\prod_{i} f_{i}\left(X_{i}\right)$. This is a polynomial whose degree in each variable is at most $d$, and for $\beta \in S_{1} \times \cdots \times S_{n}$,

$$
f_{\alpha}(\beta)= \begin{cases}1 & \text { if } \alpha=\beta \\ 0 & \text { else }\end{cases}
$$

Thus the rank of the evaluation map is full.
Last time, we used polynomials to prove the following theorem:
Theorem 2. If $\mathcal{F}$ is a family of sets such that for all $A \neq B \in \mathcal{F},|A \cap B| \in L$, then $|\mathcal{F}| \leq \sum_{i=0}^{|L|}\binom{n}{i}$.
Without much additional work, you will prove the following in homework:
Theorem 3. If $\mathcal{F}$ is a family of sets and $p$ is a prime such that for all $A \neq B \in \mathcal{F},|A \cap B| \in L$ $\bmod p$ and $|A| \notin L \bmod p$, then $|\mathcal{F}| \leq \sum_{i=0}^{|L|}\binom{n}{i}$.

## 2 Ramsey Graphs

An undirected graph on $n$ vertices is called $k$-Ramsey if it does not have an independent set or clique of size $k$. Note that if a graph is $k$-Ramsey, then it is also $(k+1)$-Ramsey. How small can $k$ be as a function of the number of vertices $n$ ?

In one of the first illustrations of the probabilistic method, Erdős showed the following theorem:
Theorem 4. A random graph on $n$ vertices is $2 \log n$-Ramsey with high probability.

Proof The probability that any fixed set of size $k$ is a clique or independent set is exactly $2^{-\binom{k}{2}+1}$. Thus, by the union bound, the probability that there is any clique or independent set of size $k$ is at most

$$
2^{-\binom{k}{2}+1}\binom{n}{k} \leq 2^{-k^{2} / 2+k / 2+1}(e n / k)^{k}=2^{-k^{2} / 2+k / 2+1+k \log n-k \log k+k \log e}
$$

When $k=2 \log n$, this is $2^{-2 \log n \log \log n+\log n+1+2 \log (e n)}$, which tends to 0 as $n$ tends to infinity.
Can we construct such a graph deterministically with a fast algorithm? Frankl and Wilson gave the following construction based on Fisher's inequalities: For a prime $p>2$, the vertices of the graph are $\binom{\left[p^{3}\right]}{p^{2}-1}$ (namely subsets of $\left[p^{3}\right]$ ) of size $p^{2}-1$. There is an edge between $A, B$ if and only if $|A \cap B|=-1 \bmod p$.
Theorem 5. The graph constructed above is $k$-Ramsey for $k=\sum_{i=0}^{p-1}\binom{p^{3}}{i}$.
Proof If there is an independent set $S$, for $A \neq B \in S,|A \cap B| \in\{0,1,2, \ldots, p-2\} \bmod p$, and further, $|A|=-1 \bmod p$. Thus, by Theorem $3,|S| \leq k$.

On the other hand, if $S$ is a clique, then for $A \neq B \in S,|A \cap B| \in\left\{p-1,2 p-1, \ldots, p^{2}-p-1\right\}$, which is a set of size $p-1$. So by Theorem $2,|S| \leq k$.

In the above theorem, $k \leq p^{O(p)}=2^{O(p \log p)}$ and $n=p^{\Omega\left(p^{2}\right)}=2^{\Omega\left(p^{2} \log p\right)}$. Thus we obtain a $k$-Ramsey graph with $k \approx 2^{\sqrt{\log n}}$, which is much larger than the $k \approx 2^{\log \log n}$ promised by the probabilistic method. A few years ago, Barak, Rao, Shaltiel and Wigderson improved this to give an algorithm with better performance.

Theorem 6 (BRSW). There is a polynomial time algorithm that computes the adjacency matrix of a size $n$ graph that is $2^{(\log n)^{o(1)}}$-Ramsey.

Unfortunately, this algorithm is much more complicated than the construction of Frankl and Wilson.

## 3 Bounds on Besicovitch Sets

Let $\mathbb{F}$ be a finite field of size $q$. A set $S \subset \mathbb{F}^{n}$ is called a Besicovitch set if it contains a line in every direction, i.e. for every $x \in \mathbb{F}^{n}$, there exists a $y \in \mathbb{F}^{n}$ such that the line $\{t x+y: t \in \mathbb{F}\} \subseteq S$. How small can such a set be? We give a proof by Dvir.
Theorem 7. $|S| \geq\binom{ q-1+n}{n}$.
Proof Fix any such set $S$ with $|S|<\binom{q-1+n}{n}$. Then consider the space of degree $q-1$ polynomials. This is a space of dimension $\binom{q-1+n}{n}$. We claim that there must be a non-zero polynomial $g$ in this space such that $g$ maps every point of $S$ to 0 . Indeed, the constraint that $g(x)=0$ is a linear constraint on the coefficients of $g$. Since the number of coefficients is more than the number of constraints, there must be some non-zero setting of coefficients that satisfies all constraints ${ }^{1}$.

[^0]Thus, there is a non-zero polynomial $g$ of total degree $d \leq q-1$ such that for fixed $x$ there exists $y$ for which $L=\{t \cdot x+y: t \in \mathbb{F}\}, g(L)=0$. Since $g(t \cdot x+y)$ is a univariate polynomial of degree at most $d$, and this polynomial has $q$ roots, it must be the 0 polynomial.

Now consider the degree $d$ coefficient of $g\left(t \cdot x+y\right.$ ) (as a univariate polynomial in $t$ ). Let $g_{d}$ be the degree $d$ homogenous part of $g$. Then the degree $d$ coefficient of $g(t \cdot x+y)$ is exactly $g_{d}(x)$. Thus $g_{d}(x)=0$ for every $x \in \mathbb{F}^{n}$, so by Lemma $1, g_{d}$ must be the 0 polynomial, which is a contradiction.


[^0]:    ${ }^{1}$ If this argument is not clear, consider the matrix $M$ whose rows correspond to points of $S$, columns corresponds to coefficients of a polynomial from the space, and the $M_{i, j}$ is the evaluation of the $j$ 'th monomial on the $i$ 'th element of $S$. Thus, for a column vector $A$, the product $M A$ denotes the evaluation of the polynomial that corresponds to the coefficients $A$ at the points of $S$. Since the number of rows of $M$ is less than the number of columns, the columns are not linearly independent, and so there must be a non-zero $A$ such that $M A=0$.

