

Graduate Discrete Math (21-701) Notes

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Abstract

Lecture notes based on Graduate Discrete Math (21-701)

1 Graphs

Definition 1.1. *Graph is a set of objects (V, E) and $E \subseteq \binom{V}{2}$*

Definition 1.2. *Walk is a sequence of vertices*

Definition 1.3. *A path is a walk without repeated vertices*

Definition 1.4. *A proper K -coloring of a graph is a function $c : V \rightarrow [k]$ such that $\forall u, v \in V, u \sim v \implies c(u) \neq c(v)$*

Theorem 1.5. *A graph is 2 colorable if and only if there is no odd cycles in G*

Proof. (\implies) AFSOC there exist an odd cycle, C in G . Define the vertices of C as v_1, v_2, \dots, v_k where k is odd. Define $c(v) = \begin{cases} \text{red} & d(v, v_1) \text{ is even} \\ \text{blue} & d(v, v_1) \text{ is odd} \end{cases}$

Then $c(v_1)$ and $c(v_k)$ are both red so a contradiction.

(\impliedby) We can assume each component is connected. Choose v_0 and define $c(v) = \begin{cases} \text{red} & d(v, v_0) \text{ is even} \\ \text{blue} & d(v, v_0) \text{ is odd} \end{cases}$

If there exist vertices u, v with uv an edge such that $d(u, v_0) \equiv d(v, v_0) \pmod{2}$ then consider the cycle, C formed by shortest path from $v_0 \rightarrow u$ and $v_0 \rightarrow v$ with uv . Then $|C| = d(u, v_0) + d(v, v_0) + 1$ is odd and we're done. \square

2 Hypergraphs

Definition 2.1. *A collection \mathcal{H} of subsets of a vertex set V .*

Definition 2.2. *\mathcal{H} is k -uniform if $|f| = k, \forall f \in \mathcal{H}$*

Definition 2.3. *A proper k -coloring of \mathcal{H} is an assignment $c : V \rightarrow [k]$ such that $\forall f \in \mathcal{H}, |c(f)| = k$*

Definition 2.4. *A rainbow coloring of \mathcal{H} is an assignment $c : V \rightarrow [k] \forall f \in \mathcal{H}, |c(f)| = |f|$*

Example 2.5. *What is the least number of edges in a k -uniform graph that is not 2-colorable?*

Let this number be $m(k)$ then $m(1) = 0, m(2) = 3, m(3) \geq 7$

Theorem 2.6. *If \mathcal{H} is a 3-uniform hypergraph with less than 6 edges then \mathcal{H} is 2-colorable*

Proof. Using induction on $|V|$

(Base Case) For $n = 6$, consider all balanced 2-colorings of V there are $\binom{6}{3} = 20$. Each hyperedge is incompatible with 2 of those colorings (namely those where the edges are 3 blue or 3 red). Thus, at least $20 - 12 > 0$ of these colorings can be proper.

(Induction Hypothesis) Suppose $n \geq 7$

Claim 2.7. *There are 2 vertices u and v not in any common edge.*

Each edge connects $\binom{3}{2} = 3$ pairs of vertices. There are $\binom{7}{2} = 21$ pairs of vertices overall. So some pair of vertices is not connected as $21 > 18$.

Define \mathcal{H}' by merging u, v into w

Claim 2.8. *\mathcal{H}' is 3-uniform*

Because no edge contains both u and v the merging doesn't create a 2 set and every edge is still has size 3.

Additionally, $||\mathcal{H}'|| \leq ||\mathcal{H}|| \leq 6$ so by induction hypothesis \mathcal{H}' is 2-colorable. Giving both u and v the same color as w and keeping the rest of the colors the same.

If an edge of e of \mathcal{H} avoids $\{u, v\}$ then it is properly colored in \mathcal{H}' . If e contains u or v then after merging it corresponds to an edge of \mathcal{H}' containing w . If e is monochromatic in \mathcal{H} then it would be monochromatic in \mathcal{H}' . This would be a contradiction so edge is monochromatic in \mathcal{H} and thus a proper coloring. \square

Remark 2.9. *Suppose it has 7 edges and vertices. Consider the coloring 4 red and 3 blue. Then there are $\binom{7}{3} = 35$ such colorings. If \mathcal{H} is not 2 colorable then there are $\binom{3}{3} + \binom{4}{3} = 5$ excluded coloring for all distinct edges. There are 4 forbidden configurations for any configurations that are not 2 colorable \mathcal{H} with $|\mathcal{H}| = 7$ on 7 vertices*

3 Probabilistic Method

Theorem 3.1. $m(k) \geq 2^{k-1}$

Proof. Color vertices of \mathcal{H} randomly red or blue. For each edge f , define E_f to be the event that f is monochromatic then $Pr[E_f] = \frac{1}{2^{k-1}}$

$$Pr \left[\bigcup_{f \in \mathcal{H}} E_f \right] \leq \sum_{f \in \mathcal{H}} Pr [E_f] = \frac{|\mathcal{H}|}{2^{k-1}} < 1$$

So there is non-zero probability that there exist a coloring with no monochromatic edges if $|\mathcal{H}| < 2^{k-1}$ \square

Theorem 3.2. Erdős-Selfridge Theorem: *Given hypergraph \mathcal{H} , consider a game between a maker and breaker. The maker's goal is to color some edge all blue and breaker's goal is to prevent all blue edges.*

If \mathcal{H} is k -uniform and $|\mathcal{H}| < 2^{k-1}$ then the breaker has a winning strategy even as player 2.

Proof. Let $\phi(f) = \begin{cases} 0 & \text{if blocked by breaker} \\ \frac{2^{\#\text{blue}\in f}}{2^n} & \text{otherwise} \end{cases}$

be the "danger function". Define

$$\phi(\mathcal{H}) = \sum_{f \in \mathcal{H}} \phi(f)$$

Observe that if an edge is all blue, then $\phi(\mathcal{H}) \geq 1$

At start of the game $\phi(\mathcal{H}) = \frac{|\mathcal{H}|}{2^n}$. The worst case for when maker moves is increasing by $\frac{|\mathcal{H}|}{2^n}$ when the chosen vertex is in all edges. Then when breaker moves,

$$-\sum_{f \ni v_1} \phi(f).$$

When maker goes after

$$\sum_{f \ni v_2} \phi(f).$$

Notice

$$\sum_{f \ni v_1} \phi(f) > \sum_{f \ni v_2} \phi(f)$$

otherwise breaker played optimally.

So as long as $\frac{|\mathcal{H}|}{2^{n-1}} < 1$ then breaker wins. \square

Definition 3.3. Incidence matrix of a hypergraph \mathcal{H} with $|V| = n$ and $|\mathcal{H}| = m$ is defined as

$$I_{i,j} = \begin{cases} 1 & \text{if } v_n \in f_m \\ 0 & \text{otherwise} \end{cases}$$

Theorem 3.4. Hall's Theorem

If G is a bipartite on (A, B) there is a complete matching if and only if

$$\forall S \subseteq A, |\Gamma(S)| \geq |S|$$

where $\Gamma(S) = \{u \in B \mid \exists v \in S, u \sim v\}$.

Theorem 3.5. Consider complete graph $\mathcal{P}(X)$ where $|X| = n$.

$\mathcal{P}(X)$ has levels

$$\binom{X}{0}, \binom{X}{1}, \dots, \binom{X}{n}$$

$\forall k < \frac{n}{2}$, there is an injection

$$f_k : \binom{X}{k} \rightarrow \binom{X}{k+1}$$

such that $\forall S \in \binom{X}{k}, S \subseteq f_k(S)$

Proof. Consider bipartite graph $\left(\binom{X}{k}, \binom{X}{k+1} \right)$, if $f \in \binom{X}{k}, g \in \binom{X}{k+1}$ then we define $f \sim g$ if $f \subseteq g$. Then for some $S \subseteq \binom{X}{k}$ then $|\Gamma(S)| \geq \frac{|S|(n-k)}{k+1}$. \square

Definition 3.6. For a sperner system is a hypergraph \mathcal{H} that satisfy if

$$\forall f, g \in \mathcal{H}, f \not\subseteq g$$

Theorem 3.7. If \mathcal{H} is a sperner system of n -vertices then

$$|\mathcal{H}| \leq \binom{n}{\lfloor \frac{n}{2} \rfloor}$$

Theorem 3.8. LYM Inequality on a sperner family \mathcal{H} ,

$$\sum_{f \in \mathcal{H}} \frac{1}{\binom{n}{|f|}} \leq 1$$

Proof. Suppose $F = \{\emptyset, \{1\}, \dots, \{1, 2, \dots, n\}\}$

Note: Any sperner family can share at most one edge with F .

Consider a random permutation $\sigma \in S_n$ and define $\mathcal{H}_\sigma = \{\sigma(f) | f \in \mathcal{H}\}$.

For any $\sigma \in S_n$, $|\mathcal{H}_\sigma \cap F| \leq 1$

Now choose any σ , uniformly at random and define $\mathcal{X} = |\mathcal{H}_\sigma \cap F|$ is a random variable and $\mathcal{X} \leq 1$.

$$\text{Let } \mathcal{X} = I_f \text{ where } I_f = \begin{cases} 1 & \sigma(f) \in F \\ 0 & \text{otherwise} \end{cases}$$

$$\mathbb{E}[\mathcal{X}] = \sum_{f \in \mathcal{H}} \mathbb{E}[I_f] = \sum_{f \in \mathcal{H}} \frac{1}{\binom{n}{|f|}} \leq 1$$

□

Definition 3.9. Define the "shadow" of $\mathcal{H} \subseteq \binom{X}{r}$ as $\partial\mathcal{H} \subseteq \binom{X}{r-1}$

$$\partial\mathcal{H} = \left\{ S \subseteq \binom{X}{r-1} : \exists T \in \mathcal{H}, S \subseteq T \right\}$$

Theorem 3.10. Let $n = |X|$ then

$$\frac{|\mathcal{H}|}{\binom{n}{r}} \leq \frac{|\partial\mathcal{H}|}{\binom{n}{r-1}}$$

with equality only if \mathcal{H} is empty on $\binom{X}{r}$

Proof. Suppose \mathcal{H} is a sperner system, not all on one level.

Write $\mathcal{H}_i = \mathcal{H} \cap \binom{X}{i}$ then $H = \mathcal{H}_i \cup \mathcal{H}_{i+1} \cup \dots \cup \mathcal{H}_j$ where $i < j$ and \mathcal{H}_i nonempty.

We can instead of \mathcal{H}_j we can write $\partial\mathcal{H}_j$ as $\partial\mathcal{H}_j \subseteq \mathcal{H}_j$

Suppose \mathcal{H} maximizes the sum $\sum_{f \in \mathcal{H}} \frac{1}{\binom{|f|}{r}}$ among all sperner graphs.

Let $S \in \partial\mathcal{H}$ and $T \subseteq \mathcal{H}$. Define a bipartite graph from $S \rightarrow T$ and edges if $S \subseteq T$.

For $T \in \mathcal{H}$, $\deg(T) = r$ for $S \in \partial\mathcal{H}$, $\deg(S) = n - (r - 1)$.

So $|\mathcal{H}| \cdot r = b \leq |\partial\mathcal{H}| \cdot (n - r + 1)$.

$$\text{Then } |\mathcal{H}| \cdot r \binom{n}{r} \leq |\partial\mathcal{H}| \cdot (n - r + 1) \binom{n}{r} \implies \frac{|\mathcal{H}|}{\binom{n}{r}} \leq \frac{|\partial\mathcal{H}|}{\binom{n}{r-1}}$$

□

Definition 3.11. An intersecting hypergraph has any 2 hyperedges intersect.

Theorem 3.12. For an intersecting hypergraph on n -vertices and r -uniform,

If $r = \frac{n}{2}$ then we can fix 1 vertex and complete the remaining $\frac{n}{2} - 1$ vertices. So $\binom{n-1}{\frac{n}{2}-1}$

If $r > \frac{n}{2}$ then 2^n .

If $r < \frac{n}{2}$ then $\binom{n-1}{r-1}$.

We'll prove the last statement

Proof. Assume $n = lk$ for some l and for any $\sigma \in S_n$ define $\mathcal{H}_\sigma = \{\sigma(f) | f \in \mathcal{H}\}$. Define \mathcal{F} to be a k -uniform hypergraph with l non-intersecting edges.

If \mathcal{H} is intersecting then $|\mathcal{H}_\sigma \cap \mathcal{F}| \leq 1$.

Let $\mathcal{X} = |\mathcal{H}_\sigma \cap \mathcal{F}|$. Then

$$\mathbb{E}[\mathcal{X}] = \sum_f \mathbb{E}[I_f] = |\mathcal{H}| \mathbb{P}[\sigma(f) \in \mathcal{F}] = |\mathcal{H}| \frac{l}{\binom{n}{k}} = \frac{|\mathcal{H}|}{\binom{n-1}{k-1}}$$

$$\text{So } |\mathcal{H}| \leq \binom{n-1}{k-1}$$

Consider the case when n is not divisible by k . If $n \geq 2k$ then fix a cyclic ordering π of the n vertices. For that ordering consider the n cyclic k -intervals for $i = 1, 2, \dots, n$

$$I_i(\pi) = \{\pi(i), \pi(i+1), \dots, \pi(i+k-1)\}$$

indices taken modulo n .

For a given π define

$$\mathcal{X}_\pi := \#\{f \in \mathcal{H} \mid f \text{ is one of the intervals } I_i(\pi)\}$$

Any two sets counted in \mathcal{X}_π must intersect since \mathcal{H} is intersecting. Among the n cyclic k -intervals at most k of them can be pairwise intersecting since we can fix one vertex however $k+1$ intervals will force two of them to be disjoint. So for every π , $\mathcal{X}_\pi \leq k$.

So

$$\mathbb{E}[\mathcal{X}_\pi] = |\mathcal{H}| \frac{k!(n-k)!}{(n-1)!} = |\mathcal{H}| \frac{n}{\binom{n}{k}} \leq k \implies |\mathcal{H}| \leq \binom{n-1}{k-1}$$

□

We'll try constructing such a configuration.

If $|\mathcal{H}| = \binom{n-1}{k-1}$ then $|\mathcal{H}_\sigma \cap \mathcal{F}| = k$ for each $\sigma \in S_n$. Then there is an i such that $\mathcal{I} = \begin{cases} \{i-k+1, i-k+2, \dots, i\} \\ \{i-k+2, i-k+3, \dots, i+1\} \\ \vdots \\ \{i, i+1, \dots, i+k+1\} \end{cases}$

Suppose $a_1, \dots, a_{k-1} \in [n]$ with no $a_j = i-k, i-k-1, \dots, i, \dots, i-k+1$

Consider a permutation σ sending $a_1 \rightarrow i+1, a_2 \rightarrow i+2, \dots, a_{k-1} \rightarrow i+k-1$ and fixing $i-k, \dots, i$.

We know, $|\mathcal{H}_\sigma \cap \mathcal{F}|$ includes all edges of \mathcal{I}

Now let σ be any permutation such that $\mathcal{H} \cap \mathcal{F}$ includes i of \mathcal{I} . It suffices to show $\mathcal{H}_\sigma \cap \mathcal{F}$ includes all of \mathcal{I} for any transposition.

Lemma 3.13. *Adjacent transposition generates S_n .*

Proof. (Case 1) If $j, j+1 \in \{i-k+1, i-k, \dots, i+k-1\}$, neither is i so they're both on same side of i .

Letting $f_0 = \{i-k+1, \dots, i\}$ and $f_1 = \{i, \dots, i+k-1\}$ then $\tau(f_0) = f_0$ and $\tau(f_1) = f_1$

(Case 2) If $j = i+k-1$ and $j+1 = i+k$ then $\tau(f_0) = f_0$ and $\tau(\{i-k, \dots, i-1\}) = \{i-k, \dots, i-1\}$. □

Theorem 3.14. *Let $\alpha_1, \dots, \alpha_n \sim Ber(p)$, choosing numbers β_1, \dots, β_n with $\sum \beta_i = 1$ then $\mathbb{P}[\sum \beta_i \alpha_i \geq \frac{1}{2}] \geq p$*

Proof. Define \mathcal{H} on $[n]$ by $f \in \mathcal{H}$ if $\sum_{i \in f} \alpha_i \geq \frac{1}{2}$. For simplicity, assume no sum is $\frac{1}{2}$. Then

$$\mathbb{P}\left[\sum \beta_i \alpha_i \geq \frac{1}{2}\right] = \sum_{f \in \mathcal{H}} p^{|f|} (1-p)^{n-|f|}$$

Define $h_k = |\mathcal{H} \cap \binom{X}{k}|$

$$\begin{aligned}\mathbb{P}\left[\sum \beta_i \alpha_i \geq \frac{1}{2}\right] &= \sum_k h_k p^k (1-p)^{n-k} \\ &= \sum_{k \leq \frac{n}{2}} h_k p^k (1-p)^{n-k} + h_{n-k} p^{n-k} (1-p)^k \\ &= \sum_{k \leq \frac{n}{2}} (h_k + h_n) p^{n-k} (1-p)^k - h_k (p^{n-k} (1-p)^k - p^k (1-p)^{n-k})\end{aligned}$$

$$\begin{aligned}\text{Note: } h_k + h_{n-k} &\geq \binom{n}{k} \text{ since it or its complement has to be in } \mathcal{H} \\ &\geq \sum_{k \leq \frac{n}{2}} \binom{n}{k} p^{n-k} (1-p)^k - h_k (p^{n-k} (1-p)^k - p^k (1-p)^{n-k}) \\ &= \sum_{k \leq \frac{n}{2}} \binom{n}{k} p^{n-k} (1-p)^k + \binom{n-1}{k-1} p^k (1-p)^{n-k} \\ &= p \sum_{k \leq \frac{n}{2}} \binom{n-1}{k-1} p^k (1-p)^{n-k} \\ &= p \sum_{k \leq \frac{n}{2}} \binom{n-1}{k-1} p^k (1-p)^{n-k} \\ &= p \sum_{k \leq \frac{n}{2}} \binom{n-1}{k-1} p^{k-1} (1-p)^{(n-k)-(k-1)} \\ &= p\end{aligned}$$

□

Theorem 3.15. If there are 10 points in the plane then they can be covered by 10 non-intersecting unit circles.

Proof. Given any collection $X \subseteq \mathbb{R}^2$, $|x| = 10$. Consider a random translation of the hexagonal circle pattern.

Let $\mathcal{Z} = \#$ points in X covered then

$$\mathbb{E}[\mathcal{Z}] = \mathbb{E}[I_1] + \cdots + \mathbb{E}[I_{10}] = 10 \cdot \frac{\pi}{\frac{6}{\sqrt{3}}} \approx 9.07$$

So there exist a translation such that $\mathcal{Z} = 10$

□

Theorem 3.16. Given a graph, G on n vertices and $\frac{nd}{2}$ edges, $d \geq 1$. Then $\alpha(G) \geq \frac{n}{2d}$.

Proof. Let $S \subseteq V$ be a random subset defined by $\mathbb{P}[v \in S] = p$, p to be determined. Let $X = |S|$ and $Y = \mathbb{E}[G|_S]$. For each $e = \{i, j\} \in E$, let Y_e be indicator random variable for the event $i, j \in S$ so that

$$Y = \sum_{e \in E} Y_e$$

For any such e ,

$$\begin{aligned}\mathbb{E}[Y_e] &= \mathbb{P}[i, j \in S] = p^2 \\ \mathbb{E}[Y] &= \frac{nd}{2} p^2\end{aligned}$$

Clearly, $\mathbb{E}[X] = np$ so $\mathbb{E}[X - Y] = np - \frac{nd}{2} p^2$.

Setting $p = \frac{1}{d}$ then $\mathbb{E}[X - Y] = \frac{n}{2d}$.

So there exist a S such that the number of vertices minus the number of edges is at least $\frac{n}{2d}$.

Create S^* from S by deleting one vertex from each edge in S and delete it and this leaves S^* with at least $\frac{n}{2d}$ vertices. With all edges destroyed we leave S^* an independent set. □

Theorem 3.17. Erdos Chromatic Number Girth Theorem

$\forall k \in \mathbb{Z}^+, \exists$ graph of girth greater than or equal to k and chromatic number k .

Proof. Idea: Choose random graph $G \sim G(n, p)$. To show a graph satisfies both properties we need the the number of short cycles (length less than k) to be 0 and there are no independent set of size no more than $\frac{n}{k}$.

For the first statement let $X = \#\text{cycles with length } \leq k$ then

$$\mathbb{E}[X] = \sum_C \mathbb{E}[I_C] = \sum_{j=3}^k \sum_{|C|=j} \mathbb{E}[I_C] = \sum_{j=3}^k \binom{n}{j} \frac{(j-1)!}{2} p^j \leq \sum_{j=3}^k n^j p^j \leq (np)^{k+1}$$

To have $\mathbb{E}[X] = O(1)$, we need $p = O\left(\frac{1}{n}\right)$

We want no independent set of size $a \approx \frac{n}{k}$ so

$$\begin{aligned} \mathbb{P}[\alpha(G_{n,p}) \geq a] &\leq \binom{n}{a} (1-p)^{\binom{a}{2}} \\ &\leq n^a (1-p)^{a(a-1)/2} \\ &\leq n^a e^{-pa(a-1)/2} \\ &= \left(ne^{-p(a-1)/2}\right)^a \\ &= \left(e^{\ln(n)-p(a-1)/2}\right)^a \end{aligned}$$

Not possible since we need $p \geq 5\ln(n)/a$ but $p = O\left(\frac{1}{n}\right)$ from previous condition.

To fix this issue consider an alteration. If $p = \frac{n^\epsilon}{n}$ and $0 < \epsilon < \frac{1}{k}$ then we have $\mathbb{P}(\alpha(G_{n,p}) \geq \frac{n}{2k}) \rightarrow 0$ since $\frac{n^\epsilon}{n} >> \frac{5\ln(n)-2k}{n}$.

To fix the short cycle issue,

$$\mathbb{E}[X] = \sum_{j=3}^k (np)^j \leq (k-3)(np)^k \leq kn^{\epsilon k}$$

By Markov's inequality

$$\mathbb{P}\left(X \geq \frac{n}{2}\right) \leq \frac{kn^{\epsilon k}}{n/2}$$

Choose n large enough such that both probabilities are greater than $\frac{1}{2}$. Then there exists a graph on n vertices with no independent set of size $\frac{n}{2k}$ and less than $\frac{n}{2}$ short cycles. Delete an vertex from each short cycle to make a graph G' with $\frac{n}{2}$ vertices, no short cycles and no independent set of size $\frac{n}{2k}$. So

$$\chi(G') = \frac{n'}{\alpha(G')} \geq \frac{\frac{n}{2}}{\frac{n}{2k}} \geq k.$$

□

When does $G_{n,p}$ have triangles?

If $X = \#\text{triangles}$ then

$$\mathbb{E}[X] = \sum_{\text{Triangle } T \in K_n} E[I_T] = \binom{n}{3} p^3 \sim n^3 p^3 / 6 = O(1)$$

if $p = O\left(\frac{1}{n}\right)$

Is $G_{n,p}$ connected for $p = \frac{c}{n}$?

Let $X = \#\text{spanning trees}$ of $G_{n,p}$ then

$$\mathbb{E}[X] = \sum_{T \in G_{n,p}} E(I_T) = n^{n-2} p^{n-1} = n^{n-2} \frac{c^{n-1}}{n^{n-1}} = \frac{c^{n-1}}{n} \rightarrow \infty.$$

Let $Y = \#\text{isolated vertices}$ then

$$\mathbb{E}[Y] = \sum_{v \in V} \mathbb{P}(v \text{ isolated}) = n(1-p)^{n-1} \approx ne^{-p(n-1)} \approx ne^{-c}$$

For $p = O\left(\frac{1}{n}\right)$, let $X = \#\text{triangles}$. Then

$$\mathbb{E}[X] = \binom{n}{3} p^3 \rightarrow 0.$$

So $\mathbb{P}(G_{n,p} \text{ having triangles}) \rightarrow 0$.

Theorem 3.18. *Threshold of \mathcal{H} in $G_{n,p}$.*

Consider $G_{n,p}$ with $p = p(n)$ and \mathcal{H} is fixed graph with k vertices and l edges.

Define $\epsilon = \epsilon(\mathcal{H}) = \frac{l}{k}$ and $\epsilon' = \epsilon'(\mathcal{H}) = \max_{J \subseteq \mathcal{H}} \epsilon(J)$. If $p^\epsilon \cdot n \rightarrow 0$ then $\mathbb{E}[\#\mathcal{H} \text{ in } G_{n,p}] \rightarrow 0$. Now to show the other side, if $p^{\epsilon'} n \rightarrow \infty$ (if $p = \omega\left(\frac{1}{n^{1/\epsilon'}}\right)$) then $G_{n,p}$ has \mathcal{H} as a subgraph with high probability.

Proof. $\mathbb{E}[\#\mathcal{H} \in G_{n,p}] \leq \binom{n}{k} hp^l \leq C(np^\epsilon)^k$ where $h = \frac{k!}{\text{Aut}(\mathcal{H})}$

If $p^{\epsilon'} n \rightarrow 0$, there is some argument for densest subgraph J . \square

Proof. Let $X = \#\mathcal{H}$ subgraph in $G_{n,p}$ then by Chebyshev,

$$\mathbb{P}[X \leq 0] \leq \frac{\text{Var}[X]}{\mathbb{E}[X]^2}.$$

We can compute

$$\mathbb{E}[X^2] = \sum_{H_1, H_2 \in \mathcal{H}} \mathbb{P}[H_1, H_2 \subseteq G_{n,p}] = \sum_{t=0}^k \sum_{|H_1 \cap H_2|=t} \mathbb{P}(H_1, H_2 \subseteq G_{n,p})$$

For $t = 0$ we have

$$\sum_{|H_1 \cap H_2|=0} \mathbb{P}(H_1 \subseteq G_{n,p}) \mathbb{P}(H_2 \subseteq G_{n,p}) \leq \mathbb{E}[X]^2$$

So

$$\mathbb{E}[X^2] - \mathbb{E}[X]^2 \leq \sum_{t=1}^k \sum_{|H_1 \cap H_2|=t} \mathbb{P}(H_1, H_2 \subseteq G) = \sum_{t=1}^k \sum_{|H_1 \cap H_2|=t} \binom{k}{t} \binom{n-k}{k-t} h p^{e(H_1 \cup H_2)}$$

By PIE, $e(H_1 \cup H_2) \geq 2l - \epsilon't$ since $e(H_1 \cap H_2) \leq \epsilon't$ as $H_1 \cap H_2$ is a subgraph of H_1

$$\begin{aligned} \sum_{t=1}^k \sum_{|H_1 \cap H_2|=t} \binom{k}{t} \binom{n-k}{k-t} h p^{e(H_1 \cup H_2)} &\leq \sum_{t=1}^k \sum_{|H_1 \cap H_2|=t} \binom{k}{t} \binom{n-k}{k-t} h p^{2l - \epsilon't} \\ &= \sum_{t=1}^k \sum_{|H_1 \cap H_2|=t} \binom{n}{k} h \cdot h \binom{n-k}{k-t} p^{l - \epsilon't} \\ &\quad (\sum_{H \in \mathcal{H}} p^l = \mathbb{E}[X] = \binom{n}{k} \cdot h \cdot p^l) \\ &\leq \mathbb{E}[X] \sum_{t=1}^k h \binom{k}{t} \binom{n-k}{k-t} p^{l - \epsilon't} \\ &\leq \mathbb{E}[X] \sum_{t=1}^k h \cdot C \cdot n^k \cdot \frac{1}{n^t} \cdot p^{l - \epsilon't} \\ &\leq \mathbb{E}[X] \sum_{t=1}^k C' \cdot h \cdot \binom{n}{k} \cdot p^l \cdot \frac{1}{(np^{\epsilon'})^t} \\ &= \mathbb{E}[X]^2 \sum_{t=1}^k C' \left(\frac{1}{np^{\epsilon'}}\right)^t \rightarrow 0 \end{aligned}$$

\square

Theorem 3.19. Chernoff Bound

Suppose you had independent random values ζ_1, \dots, ζ_n with $\zeta_i \in \{-1, 1\} \forall i$ and $\mathbb{P}(\zeta_i = 1) = \mathbb{P}(\zeta_i = -1) = \frac{1}{2}$.

Let $X = \sum_{i=1}^n \zeta_i$

$$\begin{aligned}
 \mathbb{P}(X > a) &= \mathbb{P}(e^{tX} > e^{ta}) \\
 &\leq \frac{\mathbb{E}(e^{tX})}{e^{ta}} \\
 &= \frac{\mathbb{E}(e^{t\sum \zeta_i})}{e^{ta}} \\
 &= \frac{\prod_{i=1}^n \mathbb{E}[e^{t\zeta_i}]}{e^{ta}} \\
 &= \frac{\left(\frac{e^t + e^{-t}}{2}\right)^n}{e^{ta}} \\
 &\leq e^{nt^2/2 - ta} \\
 &= e^{a^2/2n - a^2/n} \quad (\text{for } t = \frac{a}{n}) \\
 &= e^{-\frac{a^2}{2n}}
 \end{aligned}$$

Question: How many vectors can I have in \mathbb{R}^d all at a common pairwise angle.

Equivalently, $\exists \alpha, v_i \cdot v_j = \begin{cases} \alpha & \text{if } i \neq j \\ 1 & \text{otherwise} \end{cases}$

$$\begin{aligned}
 \text{Let } V &= \begin{bmatrix} | & | & & | \\ v_1 & v_2 & \cdots & v_m \\ | & | & & | \end{bmatrix} \\
 \text{Then } V^T V &= G = \begin{bmatrix} v_1 \cdot v_1 & v_1 \cdot v_2 & \cdots & v_1 \cdot v_m \\ \vdots & v_2 \cdot v_2 & \cdots & \\ \vdots & \vdots & \ddots & \end{bmatrix}
 \end{aligned}$$

Claim: $\text{Rank}(G) \leq d$.

For each $\vec{x} \in \text{Ker}(V) \implies V\vec{x} = 0 \implies V^T V \vec{x} = 0 \implies G\vec{x} = 0$

We know $\text{Rank}(V) \leq d$ so $\text{Rank}(G) = \text{Rank}(V^T V) \leq d$.

We can write $G = \alpha J_m + (1-\alpha)I_m$ then J_m has eigenvalues m with multiplicity at least 1 and eigenvectors with multiplicity $\geq m-1$.

$$\begin{bmatrix} 1 \\ -1 \\ 0 \\ \vdots \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ -1 \\ \vdots \end{bmatrix}, \dots, \begin{bmatrix} 1 \\ 0 \\ 0 \\ \vdots \\ -1 \end{bmatrix}$$

If \vec{x} is an eigenvector of J_m with eigenvalue λ then $G\vec{x} = \alpha\lambda\vec{x} + (1-\alpha)\vec{x} = (\alpha\lambda + 1 - \alpha)\vec{x}$.

G has eigenvalues $1 - \alpha$ with multiplicity at least $m-1$.

So $\text{Rank}(G) \geq n-1$ and consequently $d \geq n-1$.

Question: How many unit vectors v_1, \dots, v_n can I have in \mathbb{R}^d with all pairs approx equal angle.

Equivalently, $\exists \alpha$ such that $(*) v_i \cdot v_j \in (\alpha - \epsilon, \alpha + \epsilon)$ if $i \neq j$ and 1 if $i = j$.

Theorem 3.20. $\forall \epsilon \in (0, 1)$, there is a $c > 0$ such that for all sufficiently large d there is a collection $v_1, \dots, v_m \in \mathbb{R}^d$ of vectors satisfying $(*)$ with $m \geq 2^{cd}$

Consider the easier problem when $\alpha = 0$. Choose m vectors $v_i \in \{1, -1\}^d$ at random with $\mathbb{P}(v_i^j = +1) = \mathbb{P}(v_i^j = -1) = \frac{1}{2}$. We know $v_i \cdot v_i = n$ then let $u_i = \frac{v_i}{\sqrt{n}}$. Then $\mathbb{E}(u_i \cdot u_v) = 0$ as $\zeta_{i,j,k} =$

$\begin{cases} 1 & \text{if } v_i^k \cdot v_j^k = 1 \\ -1 & \text{if } v_i^k \cdot v_j^k = -1 \end{cases}$ By Chernoff,

$$\mathbb{P}(v_i \cdot v_j \notin (-\epsilon n, \epsilon n)) \leq 2e^{-\epsilon^2 n/2}$$

$$\mathbb{P}(\exists i, j \text{ s.t. } u_i \cdot u_j \notin (-\epsilon n, \epsilon n)) \leq \binom{m}{2} 2e^{-\epsilon^2 n/2} \leq m^2 e^{-\epsilon^2 n/2} = m^2 \beta^{-n}$$

Theorem 3.21. If $\zeta_1, \zeta_2, \dots, \zeta_n$ are independent random variables with $\mathbb{E}(\zeta_i) = 0, |\zeta_i| \leq 1, X = \sum \zeta_i$ then

$$\mathbb{P}(X \geq a) \leq e^{-\frac{a^2}{2n}}$$

Question: I flip biased coins with head prob = $\frac{1}{3}$ n times. Bound the probability # head $\geq \frac{n}{2}$.

Define $\zeta_i = \begin{cases} 1 & \text{if head} \\ -\frac{p}{1-p} & \text{otherwise} \end{cases}$ (Subtract expected value and dividing by $1-p$) Then $\mathbb{E}[\zeta_i] = 0, |\zeta_i| \leq 1$ for $p < \frac{1}{2}$. Let $X = h + (n-h)\frac{-p}{1-p}$

Theorem 3.22. For random variables X, Y

$$\begin{aligned} \mathbb{E}[XY] &= \sum_{x,y} xy \mathbb{P}(X=x, Y=y) \\ &= \sum_{x,y} xy \mathbb{P}(Y=y) \mathbb{P}(X=x|Y=y) \\ &= \sum_y y \mathbb{P}(Y=y) \sum_x x \mathbb{P}(X=x|Y=y) \\ &= \mathbb{E}(Y \mathbb{E}(X|Y)) \end{aligned}$$

Theorem 3.23. Let X be a random variable and A an event then

$$\mathbb{E}(X|A) = \frac{1}{\mathbb{P}(A)} \sum_{w \in A} p(w) \mathcal{X}(w)$$

Theorem 3.24. Let ζ_1, \dots, ζ_n be random variable with $\mathbb{E}(\zeta_i) = 0$ and $|\zeta_i| \leq 1$.

$$\mathbb{E}(e^{t \sum \zeta_i}) \leq \left(\frac{e^t + e^{-t}}{2} \right)^n$$

Proof.

$$\mathbb{E}(e^{t \sum \zeta_i}) = \mathbb{E} \left(\prod_{i=1}^n e^{t \zeta_i} \right) = \mathbb{E} \left(\prod_{i=1}^{n-1} e^{t \zeta_i} \mathbb{E} \left(e^{t \zeta_n} \mid \prod_{i=1}^{n-1} e^{t \zeta_i} \right) \right)$$

We know want to upper bound $\mathbb{E} \left(e^{t \zeta_n} \mid \prod_{i=1}^{n-1} e^{t \zeta_i} \right)$.

By convexity, $e^{t \zeta_i} \leq h(\zeta_i) = \frac{1}{2} [(1-\zeta_i)e^{-t} + (1+\zeta_i)e^t] \implies \mathbb{E}[e^{t \zeta_i}] \leq \frac{e^t + e^{-t}}{2}$

$$\mathbb{E} \left(e^{t \zeta_n} \mid \prod_{i=1}^{n-1} e^{t \zeta_i} \right) \leq \left(\frac{e^t + e^{-t}}{2} \right) \mathbb{E} \left(\prod_{i=1}^{n-1} e^{t \zeta_i} \right) \leq \left(\frac{e^t + e^{-t}}{2} \right)^n$$

□

Definition 3.25. Random variables X_0, X_1, \dots is a martingale if it satisfies the following properties

1. $\mathbb{E}[|X_i|] < \infty$
2. $\mathbb{E}[X_{i+1}|X_1, \dots, X_i] = X_i$

Theorem 3.26. Azuma's Theorem: If X_0, X_1, \dots is martingale and $|X_{i+1} - X_i| \leq 1$ then

$$\mathbb{P}(X_n - X_0 \geq \lambda\sqrt{n}) \leq e^{-\lambda^2/2}$$

Proof. By Markov's Inequality we have

$$\mathbb{P}(X_n - X_0 \geq \lambda\sqrt{n}) \leq \frac{\mathbb{E}[e^{t(X_n - X_0)}]}{e^{t\lambda\sqrt{n}}}$$

We can telescope the numerator as

$$\begin{aligned}\mathbb{E}[e^{t(X_n - X_0)}] &= \mathbb{E}\left[\prod_{k=1}^n e^{t(X_k - X_{k-1})}\right] \\ &= \mathbb{E}\left[\mathbb{E}\left[e^{t(X_n - X_{n-1})}|X_0, X_1, X_2, \dots, X_{n-1}\right] \prod_{k=1}^{n-1} e^{t(X_k - X_{k-1})}\right]\end{aligned}$$

By Theorem 3.23 we have $\mathbb{E}[e^{t(X_n - X_{n-1})}|X_1, X_2, \dots, X_{n-1}] \leq e^{t^2/2}$ so

$$\mathbb{E}\left[\mathbb{E}\left[e^{t(X_n - X_{n-1})}|X_0, X_1, X_2, \dots, X_{n-1}\right] \prod_{k=1}^n e^{t(X_k - X_{k-1})}\right] \leq e^{nt^2/2}$$

So

$$\mathbb{P}(X_n - X_0 \geq \lambda\sqrt{n}) \leq e^{nt^2/2 - t\lambda\sqrt{n}}.$$

The RHS achieves it's max at $t = \frac{\lambda}{\sqrt{n}}$. Thus

$$\mathbb{P}(X_n - X_0 \geq \lambda\sqrt{n}) \leq e^{-\lambda^2/2}$$

□

Definition 3.27. Doob's Martingale: Let X and Y_1, Y_2, \dots be random variables with $\mathbb{E}[|Y_i|] < \infty$. Define $X_i := \mathbb{E}(X|Y_1, \dots, Y_i)$ for $i \geq 1$. with $X_0 = \mathbb{E}[X]$

Theorem 3.28. McDiarmid's Inequality: Let $f : \mathcal{Y}_1 \times \mathcal{Y}_2 \times \dots \times \mathcal{Y}_n \rightarrow \mathbb{R}$ that is 1-lipschitz. If Y_1, Y_2, \dots, Y_n are independent random variables where $Y_i \in \mathcal{Y}_i$. For $X := f(Y_1, \dots, Y_n)$ satisfies

$$\mathbb{P}(X - \mathbb{E}(X) \geq \lambda\sqrt{n}) \leq e^{-2\lambda^2}.$$

Example 3.29. For m balls into n bins, let $X = \#$ empty bins then

$$\mathbb{E}[X] = n \left(1 - \frac{1}{n}\right)^m \approx ne^{-\alpha}$$

where $m = \alpha n$. If we let $Y_i = \text{pos of ball } i$ then $f(Y_1, \dots, Y_m) = \#$ empty bins. So $\mathbb{P}(X \geq \mathbb{E}[X] + \lambda\sqrt{n}) \leq e^{-2\lambda^2}$

Theorem 3.30.

$$\mathbb{P}(A|B \cap C) = \frac{\mathbb{P}(A \cap B|C)}{\mathbb{P}(B|C)}$$

Theorem 3.31. Lovasz Local Lemma: Given a collection $\mathcal{B} = \{B_1, B_2, \dots, B_n\}$ of bad events. If B_1, \dots, B_n have a dependency graph of max degree d and $4pd \leq 1$ where $\mathbb{P}(B_i) \leq p$ then

$$\mathbb{P}\left(\bigcap \overline{B_i}\right) > 0.$$

Proof. We'll prove the statement by induction on $k = |S|$ for $S \subseteq [n]$. Assume $4pd \leq 1$ then we want to show $\forall i, \mathbb{P}(B_i | \bigcap_{j \in S} \overline{B}_j) \leq 2p$.

Let $S = T \cup U$ where $T = \{j \in S | j \sim i\}$.

$$\begin{aligned} \mathbb{P}\left(B_i | \bigcap_{j \in T} \overline{B}_j \cap \bigcap_{j \in U} \overline{B}_j\right) &= \frac{\mathbb{P}(B_i \cap \bigcap_{j \in T} \overline{B}_j | \bigcap_{j \in U} \overline{B}_j)}{\mathbb{P}(\bigcap_{j \in T} \overline{B}_j | \bigcap_{j \in U} \overline{B}_j)} \\ &\leq \frac{p}{1 - \mathbb{P}(\bigcup_{j \in T} B_j | \bigcap_{j \in U} \overline{B}_j)} \\ &\leq \frac{p}{1 - \sum_{j \in T} \mathbb{P}(B_j | \bigcap_{k \in U} \overline{B}_k)} \\ &\leq \frac{p}{1 - 2dp} \quad (\text{Induction Hypothesis}) \\ &\leq 2p \quad (\text{Assumption}) \end{aligned}$$

We're done with the induction.

Then

$$\mathbb{P}\left(\bigcap_{i=1}^n \overline{B}_i\right) = \mathbb{P}\left(\overline{B}_n | \bigcap_{i=1}^{n-1} \overline{B}_i\right) \cdot \mathbb{P}\left(\bigcap_{i=1}^{n-1} \overline{B}_i\right) \geq \frac{(1-2p)^n}{p} > 0$$

□

Theorem 3.32. *Local Lemma (General Version)* Suppose \mathcal{B} is a collection of "bad" events with some dependency graph. Suppose we can assign real number $0 < X_A < 1$ to each $A \in \mathcal{B}$ such that

$$\mathbb{P}(A) \leq X_A \prod_{B \sim A} (1 - X_B).$$

Then

$$\mathbb{P}\left(\bigcap_{B \in \mathcal{B}} \overline{B}\right) \geq \prod_{B \in \mathcal{B}} (1 - X_B) > 0$$

Proof. We can prove it by induction on $|S|$ that if $B_1, \dots, B_t, B_{t+1}, \dots, B_S \in \mathcal{B}$ such that $A \sim B_1, \dots, B_t$ and $A \not\sim B_{t+1}, \dots, B_S$ to show

$$\mathbb{P}\left(A | \bigcap_{i=1}^S \overline{B}_i\right) \leq X_A$$

We have

$$\begin{aligned} \mathbb{P}\left(A | \bigcap_{i=1}^t \overline{B}_i \cap \bigcap_{i=t+1}^S \overline{B}_i\right) &= \frac{\mathbb{P}(A \cap \bigcap_{i=1}^t \overline{B}_i | \bigcap_{i=t+1}^S \overline{B}_i)}{\mathbb{P}(\bigcap_{i=1}^t \overline{B}_i | \bigcap_{i=t+1}^S \overline{B}_i)} \\ &\leq \frac{\mathbb{P}(A | \bigcap_{i=t+1}^S \overline{B}_i)}{\mathbb{P}(\bigcap_{i=1}^t \overline{B}_i | \bigcap_{i=t+1}^S \overline{B}_i)} \\ &\leq \frac{X_A \prod_{B \sim A} (1 - X_B)}{\prod_{i=1}^t (1 - X_B)} \\ &\leq X_A \end{aligned}$$

Thus we're done with induction. Using the statement

$$\mathbb{P}\left(\bigcap_{i=1}^S \overline{B}_i\right) = \mathbb{P}(\overline{B}_1 | \overline{B}_2) \times \dots \times \mathbb{P}\left(\overline{B}_n | \bigcap_{i=1}^S \overline{B}_i\right) \geq \prod_{i=1}^n (1 - X_{B_i}) > 0$$

□

Theorem 3.33. *Axel's Theorem:*

$\forall \epsilon > N_\epsilon$ and an infinite binary sequence such that $\forall n > N_\epsilon$, any 2 consecutive block of length n differ in $\geq (\frac{1}{2} - \epsilon) n$ places.

Proof. Let the bad events be $B_{i,n}$ where for each i , intervals $[i, \dots, i+n]$, $[i+n+1, \dots, i+2n]$ differ by less than $(\frac{1}{2} - \epsilon) n$

Let $X = \#$ places where they differ then $\mathbb{E}[X] = \frac{n}{2}$.

By Chernoff-Hoeffding's Lemma we have

$$\mathbb{P}(X - \mathbb{E}[X] \geq -\epsilon n) = \mathbb{P}(X \geq n/2 - \epsilon n) \leq e^{-2\epsilon^2 n} \leq e^{-\epsilon^2 n/10}$$

Let $X_{B_{i,n}} = e^{-\epsilon^2 n/20} \cdot \frac{1}{n^3}$ then fix B_{i_0, n_0} then we have

$$X_{B_{i_0, n_0}} = e^{-\epsilon^2 n_0/20} \cdot \frac{1}{n_0^3}$$

$$e^{-\epsilon^2 n_0/20} \cdot \frac{1}{n_0^3} \prod_{n=N_\epsilon}^T \prod_{i=i_0-2n}^{i_0+2n} \left(1 - e^{-\epsilon^2 n/20} \cdot \frac{1}{n^3}\right)^{2n+2n_0} \geq$$

□

From Homework #3

Theorem 3.34. $\forall \epsilon > 0, \exists N_\epsilon, \forall T, \exists$ a binary sequence of length T such that
(*) $\forall n > N_\epsilon$ identical blocks of length n are separated by distance $\geq (2 - \epsilon)^n$

Theorem 3.35. *Konig's Infinity Lemma:* Let G be a connected, locally finite, infinite graph then G contains an infinite path.

Theorem 3.36. $\forall \epsilon > 0, \exists N_\epsilon, \forall T, \exists$ an infinite binary sequence such that vertices of my tree are finite binary sequences with property (*).

Let \mathcal{T} be a complete tree of all binary sequences with all vertices and join $S \rightarrow S'$ if S can be obtained from S' by removing last digit of S' .

We want to show \mathcal{T} is locally finite and infinite. It is locally finite as each node has at most 2 children. It is infinite because for any string that satisfy (*), any of its prefix has to satisfy (*). By Theorem 3.34, there must be an infinite path with property (*)

4 Topology

Definition 4.1. A topology is a set X and a collection \mathcal{O} of open sets satisfying

1. $\emptyset \in \mathcal{O}, X \in \mathcal{O}$
2. \mathcal{O} is closed under finite intersection
3. \mathcal{O} is closed under arbitrary union

A collection of basic open sets are closed under finite intersections.

Definition 4.2. X is compact if every cover has a finite subcover

Definition 4.3. Product topology is

$$\prod_{\alpha \in A} \mathcal{O}_\alpha$$

where $\mathcal{O}_\alpha \subseteq X_\alpha$ is open and $\mathcal{O}_\alpha = X_\alpha$ except for finitely many.

Theorem 4.4. If X_α where $\alpha \in I$ are compact topological spaces then $\prod_{\alpha \in I} X_\alpha$ is compact.

5 Ramsey Numbers

Definition 5.1. Ramsey number $R(k, l) = \min_{n \geq 1} \{K_n \text{ contains a red } K_k \text{ or blue } K_l\}$
We can see $R(3, 3) = 6$

Theorem 5.2. $R(k, l) \leq R(k - 1, l) + R(k, l - 1)$

Proof. Let $n \geq R(k - 1, l) + R(k, l - 1)$ and consider a red/blue coloring of K_n . Fix v_0 . Since v_0 has $\geq R(k - 1, l) + R(k, l - 1) - 1$ edges,

(Case 1) If v_0 has $\geq R(k - 1, l)$ red edges then the induced subgraph of the neighbors, G' must have red K_{k-1} or blue K_l . If red K_{k-1} then $G' \cup v_0$ is a K_k , otherwise we have blue K_l .

(Case 2) If v_0 has $\geq R(k, l - 1)$ blue neighbors then same argument as case 1.

Thus we're done \square

Theorem 5.3. $R(k, k) > (1 - O(1)) \frac{k}{e\sqrt{2}} 2^{k/2}$

Proof. Flip fair coins to color a K_n red or blue. Let $X = \# \text{ monotonic } K_k$ then

$$\mathbb{E}[X] = \binom{n}{k} 2^{1 - \binom{k}{2}} \leq \left(\frac{en}{k}\right)^k \cdot 2 \cdot 2^{-k(k-1)/2}$$

If $\mathbb{E}[X] < 1$ then $R(k, k) > n$.

$$2^{1/k} \left(\frac{en}{k}\right) 2^{(k-1)/2} < 1$$

Thus

$$n < (1 - O(1)) \frac{k}{e\sqrt{2}} 2^{k/2}$$

Consequently, $R(k, k) > (1 - O(1)) \frac{k}{e\sqrt{2}} 2^{k/2}$ \square

Theorem 5.4. Alterations: Color edges of K_n randomly red or blue. Delete an vertex from each monochromatic K_k . Let $X = n - \# \text{ monochromatic cliques}$.

$$\begin{aligned} \mathbb{E}[X] &= n - \binom{n}{k} 2^{1 - \binom{k}{2}} \\ &\geq n - \left(\frac{en}{k}\right)^k \cdot 2 \cdot 2^{-k(k+1)/2} \\ &= n - 2 \left(\frac{en}{k} \cdot 2^{\frac{-k-1}{2}}\right) \end{aligned}$$

Let $n = \frac{k}{e} \cdot 2^{k/2}$ then

$$\frac{k}{e} \cdot 2^{k/2} - 2^{k/2} = (1 - o(1)) \cdot \frac{k}{e} \cdot 2^{k/2}$$

Theorem 5.5. Using Lovasz Local Lemma: Given k , fix n , randomly red/blue color edges.

Proof. Bad events: B_k for $k \in \mathcal{K}$ where \mathcal{K} is the collection of k -clique.

Then $\mathbb{P}(B_k) = 2^{1 - \binom{n}{k}}$.

If K_1, K_2 share any edges, set $B_{K_1} \sim B_{K_2}$ in dependency graph. Then

$$D \leq \binom{k}{2} \binom{n}{k-2}$$

Consequently

$$\begin{aligned} epD &\leq e \cdot 2^{1-\binom{n}{k}} \left(2 \binom{k}{2} \binom{n}{k-2} \right) < 1 \\ 4e \left(\left(\frac{en}{k-2} \right)^{k-2} \binom{k}{2} \right) &< 2^{\binom{k}{2}} \\ \left(2e \binom{k}{2} \right)^{\frac{1}{k-2}} \cdot \frac{en}{k-2} &< 2^{\binom{k}{2}-\frac{1}{k-2}} = 2^{\frac{k+1}{2}} \\ (1+o(1)) \frac{en}{k-2} &< 2^{\frac{k+1}{2}} \end{aligned}$$

So

$$n < (1-o(1)) \frac{k\sqrt{2}}{e} 2^{k/2}$$

Thus $R(k) > (1-o(1)) \frac{k\sqrt{2}}{e} 2^{k/2}$

□

Definition 5.6. Define K_k^j as the complete j uniform hypergraph on n vertices with k vertices

Definition 5.7. Define $R_j(k) = \min n$ such that any red/blue coloring of $\binom{[n]}{j}$ has a monochromatic K_k^j

Theorem 5.8. $R_r(k, l) \leq R_{r-1}(R_r(k-1, l), R_r(k, l-1))$

Proof. Let $N = R_{r-1}(R_r(k-1, l), R_r(k, l-1)) + 1$ and fix v . There are $N+1$ other vertices, Y . Each edge containing v includes an $r-1$ edge in Y . Let it inherit the color of the r edges.

(Case 1) We have $R_r(k-1, l)$ vertices in Y such that all $r-1$ subsets are red. (Case 1A) □

Let $C(k) = \min n$ such that $\forall X \subseteq \mathbb{R}^2$ such that $|X| = n$ and X has a subset S where $|S| = k$ and S is in convex position.

Then $C(1) = 1, C(2) = 2, C(3) = 3, C(4) = 5, C(5) = 9, \dots$

Theorem 5.9. $C(k) \leq R_4(5, k)$

Lemma: If $S \subseteq \mathbb{R}^2$ is k -points in general position such that any 4 of them are in convex position, then they all are. (Easy to see by triangulation)

Given a set F of four points color F red if not in convex position and blue otherwise.

Note: K_k^n is a complete k -uniform hypergraph on n vertices.

A red K_4^5 is impossible as $C(4) \leq 5$. Since we can find a blue K_4^5 and we win by lemma.

Color 3-tuples according to whether "sorted slopes" are increasing or decreasing. If $n \geq R_3(k, k)$, I can find k vertices all of whose 3 tuples are caps or all 3 tuples are cups.

Definition 5.10. Let $CC(k, l) = \min n$ such that any n pts in general position, no two have same x coordinate, have a k -cup or an l -cap.

Theorem 5.11. Erdos Szekeres: $CC(k, l) = \binom{k+l-4}{k-2} + 1$

Proof. I have k -cup or a l -cup. We'll show it by induction on $k+l$. Assume no l -cap. I do have a $(k-1)$ -cup or l cup by induction. If I delete the last point of each $(k-1)$ -cup. Then only $\binom{k+1-5}{k-3}$ points remain. So I deleted $\binom{k_l-4}{k-2} + 1 - \binom{k+l-5}{k-2} + 1$. □

Theorem 5.12. For all positive integers k and r , there exists N such that any r -coloring of the numbers $1, 2, \dots, N$ has a monochromatic k -term arithmetic progression.

Theorem 5.13. If \mathbb{N} is partitioned into 2 sets, one contains arbitrary long arithmetic progression.

Statement 1: $\forall k, \exists N$ such that any 2 coloring of $[N]$ has a monochromatic k -term arithmetic progression. If such a statement is false for k_0 , then for all n there is a coloring of $[n]$ with no k_0 arithmetic progression. With Konig's Lemma there exists a coloring of \mathbb{N} with no k_0 -term arithmetic progression.

Statement 1 implies

Statement 2 $\forall r, \forall k, \exists W(k, r)$ such that any r -coloring of $[N]$ for $N \geq W(k, r)$ admits a k -term monochromatic A.P.

Some values of $W(k, r)$ are $W(k; 1) = k$, $W(2, r) = r + 1$, $W(2, 2) = 3$ and $W(3, 2) = 9$.

Theorem 5.14. $W(3, 2) \leq 325$

Note: The technique used here can be used for the general case.

Proof. Consider 65 blocks of 5 spots each. Within the first 33 blocks, there must be 2 blocks of the same coloring. Let the blocks be $b_1, b_2 \in [33]$. Of the first block consider the first 3 spots then if it's same color then we're done, WLOG for $a_1, a_2 \in [3]$ with $a_1 < a_2$ say $5b_1 + a_1, 5b_1 + a_2$ be red. Let $a_3 = 2a_2 - a_1 \in [5]$. If $7b_1 + a_3$ is red then we're done as $7b_1 + a_1, 7b_1 + a_2, 7b_1 + a_3$ is a mono A.P. So say $7b_1 + a_3$ is blue.

Since b_2 is the same coloring then let $b_3 = 2b_2 - b_1 \in [65]$. If $7b_3 + a_3$ is red then we have $7b_1 + a_1, 7b_2 + a_2, 7b_3 + a_3$. Otherwise if blue we have $7b_1 + a_3, 7b_2 + a_3, 7b_3 + a_3$.

Thus we're done and $W(3, 2) \geq 65 \cdot 5 = 325$ \square

Definition 5.15. $WF(k, l, r) = \text{minimum } N \text{ such that any } r\text{-coloring of } [N] \text{ admits } l \text{ color focused } k\text{-term A.P or a } k+1 \text{ term A.P.}$

Theorem 5.16.

$$WF(2, 2, r) \leq (2r^{2r+1} + 1)(2r + 1)$$

$$WF(2, 3, r) \leq (2r^{2r^{2r+1}+1} + 1)(2r^{2r+1} + 1)(2r + 1)$$

Definition 5.17. Hales-Jewett: $\forall r, \forall n, \exists d$ such that in any r -coloring of $[n]^d$ hypercube, there is a monochromatic line.

Definition 5.18. A combinatorial line is a set of points represented by a string in $([n] \cup \{x\})^d \setminus [n]^d$. The points of the line are obtained by substituting $x = 1, 2, \dots, n$.

Definition 5.19. A geometric line $([n] \cup \{x, \bar{x}\})^d \setminus [n]^d$ obtained by substituting in $x = 1, \dots, n$ and $\bar{x} = n - x + 1$.

Given an A.P-free coloring of $[N]$ want to give a line free coloring of $[N]^d$. Define $\phi : [n]^d \rightarrow (n-1)d$ by $\phi(a_0, a_1, \dots, a_{d-1}) = a_0 + a_1 + \dots + a_{d-1}$. Then we have

$$HJ(2, r) \leq d \iff 2^d < r \iff HJ(2, r) \leq \log_2 r$$

$HJ^c(2, r) = r$ as if we take any of $(0, \dots, 0), (1, 0, \dots, 0), \dots$ there are $d+1 > r \implies$ a monochromatic combinatorial line.

For $HJ(3, 2)$, take $p \in [3]^d$

Additive Combinatorics

Definition 5.20.

$$A + A = \{a + a' | a, a' \in A\}$$

$$A \cdot A = \{a \cdot a' | a, a' \in A\}$$

Theorem 5.21. $\max(|A + A|, |A \cdot A|) \geq |A|^{1+\epsilon}$

Suppose we have a set A , $X = A + A, Y = A \cdot A$. Let $\mathcal{P} = X \times Y = (A + A) \times (A \cdot A)$. Let $\mathcal{L} = \{\{y|y = a(x - a')\}|a, a' \in A\}$. Then $|\mathcal{L}| = |A|^2$.

Define $i(\mathcal{L}, \mathcal{P})$ to be the number of incidences between the points and lines in \mathcal{P} and \mathcal{L} . For any line containing $a, a' \in A$, the equation is $y = a(x - a')$. For a point $p = (a' + a'', a \cdot a'')$ we have $a' + a'' \in A + A$ and $a \cdot a'' \in A \cdot A$.

$$i(\mathcal{L}, \mathcal{P}) \geq |\mathcal{L}| \cdot |A| = |A|^3$$

Then

$$i(\mathcal{L}, \mathcal{P}) = O(|\mathcal{L}|^{2/3} |\mathcal{P}|^{2/3} + |\mathcal{L}| + |\mathcal{P}|).$$

So $|A|^3 \leq i(\mathcal{L}, \mathcal{P}) \leq C(|A|^{4/3} (|A + A| \cdot |A \cdot A|^{2/3}))$ as $|\mathcal{L}| + |\mathcal{P}| = O(|\mathcal{L}|^{2/3} |\mathcal{P}|^{2/3})$. We also have $|A|^2 \leq |\mathcal{P}| \leq |A|^4$ and $|\mathcal{P}|^{1/2} \leq |\mathcal{L}| \leq |\mathcal{P}|$. So $C|A|^{5/2} \leq |A \cdot A||A + A|$ and consequently

$$\max(|A + A|, |A \cdot A|) \geq \epsilon|A|^{5/4}$$

Planar Graphs

Theorem 5.22. *Euler's Formula for Planar Graphs:*

$$|V| - |E| + |F| = 2$$

Theorem 5.23. *Suppose G is a connected planar graph with $m \geq 3$. Then*

$$m \leq 3n - 6$$

Proof. Consider the bipartite graph of $E(G)$ and $|F(G)|$. For each edge there is at most 2 faces and each face is closed by at least 3 edges. So

$$2|E| \leq \sum \deg(e) = \sum \deg(f) \geq |F| \cdot 3$$

$$\text{So } n - m + f = 2 \implies n - m + \frac{2}{3}m \geq 2 \implies n - \frac{1}{3}m \geq 2$$

□

Definition 5.24. *Let $Cr(G)$ is the minimum number of crossing in any drawing.*

Given G , if $e(G) \geq 3n$, G is not planar $Cr(G) \geq m - 3n$ since at least $m - 3n$ edges must be removed to make G planar.

Consider G with G_p =graph where each vertex stays with probability p . Then $\mathbb{E}(np) = pn$ and $\mathbb{E}(mp) = p^2m$. Then

$$p^4Cr(G) \geq \mathbb{E}(Cr(G_p)) \geq \mathbb{E}(m_p - 3n_p) = \mathbb{E}(m_p) - 3\mathbb{E}(n_p) \geq p^2m - 3pn$$

So $Cr(G) \geq \frac{m}{p^2} - \frac{3n}{p^3}$ and is maximized when $p = \frac{4n}{m}$ only when $4n \leq m$. Then $\frac{m}{p^2} - \frac{3n}{p^3} = \frac{m^3}{64n^2}$.

Theorem 5.25. *For any collection \mathcal{L} of lines in \mathbb{R}^3 , there are at most $O(|\mathcal{L}|^{3/2})$ joints.*

We just have to show the following lemma to imply the theorem.

Lemma 5.26. *In any collection of lines with $|J|$ joints, there exist some line in $\leq 3|J|^{1/3}$ joints.*

Lemma 5.26 \implies theorem 5.25 as we define $J(L) = \text{most joints in } |L| \text{ lines}$.

$$J(L) \leq J(L - 1) + 3J^{1/3} \leq J(L - 2) + 3(J - 1)^{1/3} + 3J^{1/3} \leq \dots$$

$$\text{So } J \leq 3J^{1/3}L \iff J^{2/3} \leq 3L \iff J \leq \sqrt{27}L^{3/2}$$

Given an arbitrary field, $\text{Poly}_D(\mathbb{F}^n)$ and $S = \{a_1, \dots, a_k\}$ for $a_i \in \mathbb{F}^n$. We want to find a nonzero polynomial that vanishes at J .

Let $T : \text{Poly}_D(\mathbb{F}^n) \rightarrow \mathbb{F}^k$ defined as $T(p) = \begin{bmatrix} p(a_1) \\ \vdots \\ p(a_k) \end{bmatrix}$

By rank nullity theorem,

$$\dim(Im(T)) \leq k \implies \dim(\ker(T)) \geq \dim(\text{Poly}_D(\mathbb{F}^n)) - k$$

If $\dim(\text{Poly}_D(\mathbb{F}^n)) > k$, $\exists p \in \text{Poly}_D(\mathbb{F}^n)$ vanishes at S , $|S| = k$.

Let

$$\mathcal{D} = \{x_1^{d_1}, x_2^{d_2}, \dots, x_n^{d_n} \mid \sum d_i \leq D\}$$

This is a basis for $\text{Poly}_D(\mathbb{F}^n)$. By stars and bars we have $|\mathcal{D}| = \binom{D+n}{n} \geq \frac{D^n}{n!} > k$. We need $\frac{D^3}{3!} > J$ so $D > 3J^{1/3}$.

AFSOC each line has more than $D > 3J^{1/3}$ joints.

If $p \in \text{Poly}_D(\mathbb{F})$, $\forall a \in \mathbb{F}$, $\exists c \in \mathbb{F}$ such that $p(x) = (x - a)q(x) + c$. If a is a root, $p(x) = (x - a)q(x)$. A line is a function $\gamma(t) = a + bt$ for $a, b \in \mathbb{F}^n$ then $q(t) := p(\gamma(t))$ is a polynomial in $\text{Poly}_D(\mathbb{F})$. $\deg(q)$ has to have at most the degree of p so $\deg(q) < D$. By our assumption the line has more than D joints so $\deg(q) > D$. The only way $q(t)$ can have more than D roots is if q is the zero polynomial. So we can conclude our polynomial p must be identically 0 on the union of all lines in \mathcal{L} .

Each joint is the intersection of 3 lines and p is zero on all lines. So the direction derivative along each of the lines is 0 and as they are linearly independent we have $\nabla p = 0$ at every joint. Consider $p_1 = \frac{\partial p}{\partial x}$, $\deg(p_1) \leq D - 1$ and p_1 vanishes at every joint in J so we contradict the minimality of D so the assumption that all lines have $> D$ joints is false.

Lemma 5.27. If $P(x_1, \dots, x_n)$ is a non-zero polynomial over \mathbb{F}_q , with total degree $D \leq q - 1$ then $P(x)$ cannot be zero for all $x \in \mathbb{F}_q^n$.

Proof. We can write

$$P(x) = \sum_{k=0}^D q_k(x_1, \dots, x_{n-1}) x_n^k$$

We'll show the statement by induction on n .

(**Base Case** $n = 1$) AFSOC $P(x_1)$ is nonzero and vanishes on all of \mathbb{F}_q . Since $\deg(P) \leq q - 1$ but P has q distinct roots, P must be the zero polynomial. This is a contradiction.

(**Inductive Step**) AFSOC $P(x) = 0$ on all $x \in \mathbb{F}_q^n$. We write

$$P(x) = \sum_{k=0}^D q_k(x_1, \dots, x_{n-1}) x_n^k$$

Since $P(x)$ is a nonzero polynomial, there must exist at least one k for which $q_k(x_1, \dots, x_{n-1})$ is a non-zero polynomial.

Fix the first $n - 1$ variables. Let (a_1, \dots, a_{n-1}) be an arbitrary point in \mathbb{F}_q^{n-1} . Define

$$Q(t) := P(a_1, \dots, a_{n-1}, t)$$

We can express $Q(t) = \sum_{k=0}^D q_k(a_1, \dots, a_{n-1}) t^k$. For this fixed (a_1, \dots, a_{n-1}) , each $q_k(a_1, \dots, a_{n-1})$ is a constant in \mathbb{F}_q . This implies $\deg(Q) \leq D \leq q - 1$.

By assumption, $P(x) = 0$ everywhere, so $Q(t) = P(a_1, \dots, a_{n-1}, t) = 0$ for all $t \in \mathbb{F}_q$. From our base case, a single-variable polynomial of degree $\leq q - 1$ that has q roots must be the zero polynomial. This means all coefficients of $Q(t)$ must be zero.

Therefore $q_k(a_1, \dots, a_{n-1}) = 0$ for all k . Since (a_1, \dots, a_{n-1}) was arbitrarily chosen, this holds for all points in \mathbb{F}_q^{n-1} .

This means each q_k is a polynomial in $n - 1$ variables that vanishes on all of \mathbb{F}_q^{n-1} . The total degree of P is $D = \max_k(\deg(q_k) + k)$, which implies $\deg(q_k) \leq D \leq q - 1$. By our inductive hypothesis, a polynomial in $n - 1$ variables of degree $\leq q - 1$ that vanishes everywhere must be the zero polynomial.

Thus, each q_k is the zero polynomial. This implies $P(x)$ is the zero polynomial, which contradicts our initial assumption that $P(x)$ is a non-zero polynomial. \square

Theorem 5.28. If $N \subseteq \mathbb{F}_q^n$ is a set with the property that for all $x \in \mathbb{F}_q^n$, there is a line L_x such that $L_x \setminus \{x\} \subseteq N$, then $|N| \geq \epsilon_n q^n$ where $\epsilon_n > 0$ depends only on n . (The proof shows $\epsilon_n = (10n)^{-n}$).

Proof. Assume for the sake of contradiction that $|N| < \left(\frac{q}{10n}\right)^n$.

We know from the polynomial method that there exists a non-zero polynomial $p \in \text{Poly}_D(\mathbb{F}_q^n)$ that vanishes on N , with degree $D \leq 2n|N|^{1/n}$.

Using our AFSOC, we can bound this degree D :

$$D \leq 2n|N|^{1/n} < 2n \left(\frac{q}{10n}\right)^n = \frac{q}{5}$$

So, we have found a non-zero polynomial p with total degree $D < q/5$.

Now, consider any arbitrary $x \in \mathbb{F}_q^n$. By the theorem's premise, there is a line L_x through x such that $L_x \setminus \{x\} \subseteq N$. We can parametrize this line as $\gamma(t) = x + d \cdot t$ for $t \in \mathbb{F}_q$, where $d \in \mathbb{F}_q^n \setminus \{0\}$ is a direction vector. Note that $\gamma(0) = x$, and $L_x \setminus \{x\} = \{\gamma(t) \mid t \in \mathbb{F}_q \setminus \{0\}\}$.

Define a new, single-variable polynomial $R(t) := p(\gamma(t))$. The degree of $R(t)$ is at most the total degree of p , so $\deg(R) \leq D < q/5$.

Since p vanishes on N , p must vanish on $L_x \setminus \{x\}$. This means $R(t) = p(\gamma(t)) = 0$ for all $t \in \mathbb{F}_q \setminus \{0\}$. The set $\mathbb{F}_q \setminus \{0\}$ has $q - 1$ elements, so $R(t)$ has $q - 1$ distinct roots.

We have a polynomial $R(t)$ with $\deg(R) \leq D < q/5$. For any $q \geq 3$, we have $q/5 \leq q - 2$ (since $10 \leq 4q$).

Thus, $R(t)$ is a polynomial with degree strictly less than $q - 1$, but it has $q - 1$ roots. A non-zero polynomial cannot have more roots than its degree. Therefore, $R(t)$ must be the zero polynomial.

If $R(t)$ is the zero polynomial, it must be zero for all t , including $t = 0$.

$$R(0) = p(\gamma(0)) = p(x) = 0$$

Since $x \in \mathbb{F}_q^n$ was arbitrary, we have shown $p(x) = 0$ for all $x \in \mathbb{F}_q^n$. We also know $\deg(p) = D < q/5$, which implies $\deg(p) \leq q - 1$.

By Lemma 5.27, any polynomial with degree $\leq q - 1$ that vanishes on all of \mathbb{F}_q^n must be the zero polynomial. This contradicts our choice of p as a non-zero polynomial.

Therefore, our initial assumption was false, and we must have $|N| \geq (\frac{q}{10n})^n$. \square

Lemma 5.29. If $p \in \text{poly}_{a-1}(\mathbb{F}_q^n)$ is nonzero, $|\text{zero}(P)| < q^n$.

Maximized when $x_1^{q-1} - 1$

Theorem 5.30. Schatz-Zippel: If nonzero $p \in \text{poly}_D(\mathbb{F}^n)$ and $S \subseteq \mathbb{F}$ a finite subset. For random $s_1, \dots, s_n \in S$

$$\mathbb{P}_{s_1, \dots, s_n}(p(s_1, \dots, s_n) = 0) \leq \frac{D}{|S|}$$

In other words, $|\text{zero}(p) \cap S^n| \leq D|S|^{n-1}$

Proof. Let $p \in \text{poly}_D(\mathbb{F}^n)$ be nonzero. We're done if $n = 1$. Do induction on n .

$$p(x_1, \dots, x_n) = \sum_{k=0}^n q_k(x_1, \dots, x_{n-1}) x_n^k.$$

Choose k_0 to be largest such that $q_{k_0} \neq 0$. By induction

$$\mathbb{P}_{s_1, \dots, s_{n-1}}(q_{k_0}(s_1, \dots, s_{n-1}) = 0) \leq \frac{D - k_0}{|S|}$$

$$\mathbb{P}_{s_1, \dots, s_n}(p(s_1, \dots, s_n) = 0) \leq \mathbb{P}_{s_1, \dots, s_n}(q_{k_0}(s_1, \dots, s_{n-1}) = 0) + \mathbb{P}_{s_1, \dots, s_n}(p(s_1, \dots, s_n) = 0 | q_{k_0}(s_1, \dots, s_{n-1}) \neq 0)$$

Note: This is just $\mathbb{P}(B) \leq \mathbb{P}(C) + \mathbb{P}(B \mid \neg C) \cdot \mathbb{P}(\neg C)$

$q_{k_0}(s_1, \dots, s_{n-1}) \neq 0 \implies p(s_1, \dots, s_{n-1}, x_n)$ has degree k_0 .

$$\frac{\sum_{s_1, \dots, s_{n-1}} \prod_{q(s_1, \dots, s_{n-1}) \neq 0} \frac{1}{|S|^{n-1}} \mathbb{P}(p(s_1, \dots, s_n) = 0 | q_{k_0}(s_1, \dots, s_{n-1}) \neq 0)}{\sum_{s_1, \dots, s_{n-1}} \prod_{q(s_1, \dots, s_{n-1}) \neq 0} \frac{1}{|S|^{n-1}}}$$

$$\mathbb{P}_{s_1, \dots, s_n}(q_{k_0}(s_1, \dots, s_{n-1}) + \mathbb{P}_{s_1, \dots, s_n}(p(s_1, \dots, s_n) = 0 | q_{k_0}(s_1, \dots, s_{n-1}) \neq 0) \leq \frac{D - k_0}{|S|} + \frac{k_0}{|S|} = \frac{D}{|S|}$$

\square

Theorem 5.31. Extremal Schwartz-Zippel. If p nonzero of degree d , $p = 0$ on S^n then $|S| \leq d$.

Example 5.32.

$$p = \prod_{s \in S} (x_i - s)$$

is 0 for all $x \in S^n$. This holds for $S \times \{1\} \times \{1\} \times \dots$

Example 5.33.

$$q = \prod_{a_1 \in S_1} (x_1 - a_1) \prod_{a_2 \in S_2} (x_2 - a_2) \prod_{a_3 \in S_3} (x_3 - a_3)$$

Say S_1, S_2, S_3 has size 4, 3, 2, respectively. If in a $5 \times 4 \times 3$ box then q is definitely not zero polynomial.

Theorem 5.34. *Combinatorial Nullstellensatz:* Suppose p is a nonzero polynomial in $\text{Poly}_d(\mathbb{F}^n)$ of degree d and the monomial $x_1^{j_1}x_2^{j_2}\cdots x_n^{j_n}$ for $j_1 + \cdots + j_n = d$ has nonzero coefficients then $\forall S_1, \dots, S_n \subseteq \mathbb{F}$ with $|S_i| \geq j_i + 1$ for all i , $p(x) \neq 0$ for some $x \in S_1 \times S_2 \times \cdots \times S_n$.

Proof. We'll show it by induction on d .

Suppose $f \equiv 0$ on $S_1 \times S_2 \times \cdots \times S_n$ where $S_i \subseteq \mathbb{F}$.

WLOG $j_i \geq 1$, $\forall s \in S_1$ we have $f(x) = (x_1 - s)q_s(x) + r(x)$ then $\deg(q_s) = d - 1$, moreover coefficients of $x_1^{j_1-1}x_2^{j_2}\cdots x_n^{j_n}$. For any $s \in S_1, s_2 \in S_2, \dots$ we have $f(s, s_2, \dots, s_n) = 0 \implies r_s(s, s_2, \dots, s_n) = 0 \forall s_2, \dots, s_n \implies r_s(s_1, \dots, s_n) = 0 \forall s_1 \in S_1, s_2 \in S_2, \dots$

By our assumption we have $0 = f(s_1, \dots, s_n) = (s_1 - s)q_s(s_1, \dots, s_n)$ for all $s, s_1 \in S_1, \dots, s_n \in S_n$. For $s \neq s_1$ we learn $q_s(s_1, \dots, s_n) = 0$. Since q_s is zero on $S_1 \setminus \{s\} \times S_2 \times \cdots \times S_n \implies$ we must have $|S_1 \setminus \{s\}| \leq j_1 - 1$ or for some i , $|S_i| \leq j_i$. by induction hypothesis. \square

Example 5.35. $\chi'(G)$ list chromatic number of G . We want to find $\chi'(C_n)$.

Consider an assignment c_i to each vertex i , $c_i \in \mathbb{N}$. $f(c_1, \dots, c_n) = (c_2 - c_1)(c_3 - c_2)\cdots(c_n - c_{n-1})(c_1 - c_n)$. The leading term of f is $2c_1c_2\cdots c_n$. For all sets $S_1, \dots, S_n \subseteq \mathbb{N}$, chromatic number implies $\exists c_1 \in S_1, c_2 \in S_2, \dots, c_n \in S_n$ such that $f(c_1, \dots, c_n) \neq 0$.

Example 5.36. Cauchy Davenport: Let p prime, $A, B \subseteq \mathbb{Z}_p$

$$|A + B| \geq \min(p, |A| + |B| - 1)$$

Proof. Case 1: If $|A| + |B| - 1 \geq p$

Consider $x \in \mathbb{Z}_p$ then $|x - A| = |A|$ so $|A - x| + |B| \geq p + 1$ and $\exists y \in A_x \cap B \implies \exists a \in A, b \in B$ such that $y = b, y = x - a$ so $x = a + b$.

Case 2: If $|A| + |B| - 1 < p$

Consider any set $C \subseteq \mathbb{Z}$ of size $|C| = |A| + |B| - 2$. We want to show $\exists x \in A + B, x \notin C$.

Define

$$f(a, b) = \prod_{c \in C} (a + b - c)$$

We have $\deg(f) = |C| = |A| + |B| - 2$, consider the monomial $a^{|A|-1}b^{|B|-1}$ then the coefficient is $\binom{|A|+|B|-2}{|A|-1} \neq 0$ in \mathbb{Z}_p . So for $|A|, |B|$ we have a choice of a, b such that $a + b \notin C$. \square

Definition 5.37. Finite Kakeya: In $\mathbb{F}_a^n, \forall a, \exists b$ such that $\{at + b | t \in \mathbb{F}_a\} \subseteq K$. Then K is a kakeya set.

Theorem 5.38. Chevalley-Warning theorem:

Let $a = p^l$ for $f_1, \dots, f_k \in \mathbb{F}_a[x_1, \dots, x_n]$.

If $\sum_i \deg(f_i) < n$ then the number of common zeros is a multiple of p . In particular: if there's 1 common zero then there is more.

Example 5.39. Given any n numbers a_1, \dots, a_n there is a nonempty subset that sums to 0 (mod n).

Proof. Let $S_0 = \{\}, S_1 = \{a_1\}, \dots, S_n = \{a_1, \dots, a_n\}$ then there exists i, j such that $S_i = S_j$ so \square

Theorem 5.40. Erdos-Ginzburg-Ziv Theorem:

How large a collection of numbers do I require to ensure that some n -subset sum to a multiple of n ? $2n - 1$ is enough

Proof. (Main Case) $n = p$ is a prime

Given numbers a_1, \dots, a_{2p-1} , we'll give two polynomials in $2p - 1$ variables x_1, \dots, x_{p-1}

We want a polynomial such that x_i behaves like indicators for $a_i \in S$. So $x_i^{p-1} \equiv 1 \pmod{p}$ by FLT.

$$f(x_1, \dots, x_{2p-1}) = \sum_{x_i} x_i^{p-1} = \#\{i | x_i \neq 0\}$$

$$g(x_1, \dots, x_{2p-1}) = \sum_{x_i} a_i x_i^{p-1} = \sum_{x_i \neq 0} a_i$$

We have $2p-2 < 2p-1$ and the trivial solution exist so a non-trivial solution exist by Chevalley-Warning (General Case) Induction on n , a_1, \dots, a_{2n-1}

If n not prime, let p be a prime factor of n , $m = \frac{n}{p}$. Find a set I_i , $|I_i| = p$ such that $\sum_{j \in I_i} a_j \equiv 0 \pmod{p}$ for $i \in [2m-1]$

Say $\sum_{i \in I_j} a_i = b_i \equiv 0 \pmod{p}$

Let $c_i = \frac{b_i}{p}$, we can find $c_{i_1}, c_{i_2}, \dots, c_{i_m}$ such that

$$\sum_{j=1}^m c_{i_j} = \sum_{j=1}^m \sum_{t \in I_{i_j}} t = \left(\sum_{j=1}^m c_{i_j} \right) p \equiv 0$$

□

Theorem 5.41. *There exist an order of at least d^2 2-distance set in \mathbb{R}^d ?*

Proof. Suppose $S = \{p_1, \dots, p_m\}$ has just 2 distances α and β . Consider the polynomial, $f \in \mathbb{R}[x_1, \dots, x_d]$ defined as

$$f_i(X) = (||X - p_i||^2 - \alpha^2)(||X - p_i||^2 - \beta^2)$$

$$f_i(X) = \begin{cases} \alpha^2 \beta^2 & \text{if } X = p_i \\ 0 & \text{otherwise} \end{cases}$$

Claim: f_i 's are independent.

Suppose $\alpha_1 f_1 + \dots + \alpha_m f_m = 0, \forall i$, plug in p_i gives $\alpha_i \alpha^2 \beta^2 = 0 \implies \alpha_i = 0$

Claim: $x_i^{d_i} x_j^{d_j}$ with $d_1 + d_2 \leq 4$ covers all possible terms. So there is at most $O(d^2)$ possible choices. □

Example 5.42. Eventown where each club has even size and even intersection, $\geq 2^{\lceil n+\frac{1}{2} \rceil}$

Example 5.43. Oddtown where each club has odd size and even intersection

$$\text{Let } v_i = \text{indidence vector of club } i \text{ in } \mathbb{F}_2. v_i \cdot v_j = \begin{cases} 0 & \text{if } i \neq j \\ 1 & \text{otherwise} \end{cases}$$

Suppose $\alpha_1 \vec{v}_1 + \dots + \alpha_n \vec{v}_n = \vec{0}$

\vec{v}_i on both sides so $\alpha_i = 0$

Theorem 5.44. Every Polygon has a triangulation

Proof. Choose a convex vertex of the polygon (a vertex that is a vertex of the convex hull) with neighbors q, r . If $\overline{qr} \subset P^\circ$ then we're done. Otherwise we can move the point. □

Definition 5.45. We say polygon $P \sim Q$ by scissor congruency if we can cut up P and reassembled to be Q .

Lemma 5.46. (Any rectangle) \sim (Any Unit Size Rectangle)

(Any triangle) \sim (Unit Side Rectangle)

(Triangle) \sim (2 Right Triangle)

(Right Triangle) \sim (Rectangle)

Remark 5.47. From this lemma we can conclude any polygon is congruent to a rectangle of $1 \times d$

Theorem 5.48. Are equal-area polyhedra necessarily plane-dissection equivalent? This is not true.

Example 5.49. Unit cube and volume 1 reg-tetrahedron are not dissection equivalent.

Proof. Dihedral angle is an irrational multiple of π

A list of vectors \vec{v}_1, \dots is independent if for every finite sum,

$$\sum_{j=1}^k \alpha_{i_j} v_{i_j} = 0 \implies \alpha_{i_j} = 0 \forall j$$

$Span(\mathcal{L})$ is the set of vectors representable as finite linear combinations. \mathcal{L} is a basis for V if \mathcal{L} is independent and $Span(\mathcal{L}) = V$

Lemma 5.50. Zorn's Lemma: P is a poset in which every chain has an upper bound then P has a maximal element.

Doset is a set of independent sequences, ordered by inclusion.

$$\mathcal{L}_1 \subseteq \mathcal{L}_2 \dots \subseteq$$

Then $\cup \mathcal{L}_i$ is still independent. By Zorn's lemma and Doset we have every vector space has a (possibly infinite) basis.

Define α to be dihedral angle of tetrahedron, we'll use that $\frac{\alpha}{\pi}$ is irrational.

In general, we can define a linear transformation $f : \mathbb{R} \rightarrow \mathbb{R}$ such that $f(\pi) = 0$ and $f(\alpha) = 1$.

Since α and π are independent over \mathbb{Q} extended to a basis $\alpha, \pi, v_3, v_4, \dots$

Define f using this basis by defining $f(\alpha) = 1, f(\pi) = 0$

If a plane goes through an angle then $l_e = l_{e_1} = l_{e_2}$ and $\theta_e = \theta_{e_1} + \theta_{e_2}$.

If a plane goes through an edge then $l_{e_1} + l_{e_2} = l_e$ and $\theta_e = \theta_{e_1} = \theta_{e_2}$

If a plane goes through another plane and creates a new edge then $l_{e_1} = l_{e_2}$ and $\theta_{e_1} + \theta_{e_2} = \pi$

We can assign a real number to each polytope by

$$\sum_{e \in P} l_e \cdot f(\theta_e)$$

In the first case we would have

$$R(P) = \sum_{e \in P} l_e f(\theta_e)$$

If P is a cube then $R(P) = 0$ as each dihedral angle is 90° so it's a rational multiple of π . However if P is a tetrahedron has irrational multiple of π so $R(P) \neq 0$. \square

Linear Algebra

Definition 5.51. Adjacency Matrices: On the vertex set $[n]$

$$A = \begin{bmatrix} & \\ & \end{bmatrix}$$

$$A_{i,j} = \begin{cases} 1 & \text{if } i \sim j \\ 0 & \text{otherwise} \end{cases}$$

If $f : V \rightarrow \mathbb{R}$ then $Af = g$ and if $Af = \lambda f$ then it's an eigen function.

Remark 5.52. We'll denote the $n \times n$ matrix of all ones as J_n .

Theorem 5.53. J_n 's eigenvalues are $\lambda_1 = n$ with multiplicity 1 and $\lambda_2 = 0$ with multiplicity $n - 1$.

Proof. To show $\lambda = 0$ has multiplicity $n - 1$ the associated eigenvectors are

$$\begin{bmatrix} 1 \\ -1 \\ 0 \\ \vdots \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ -1 \\ \vdots \end{bmatrix}, \dots,$$

□

Theorem 5.54. $K_n = J_n - I_n$ has eigenvalue $n - 1$ with multiplicity 1 and -1 with multiplicity $n - 1$.

Remark 5.55. For any regular graph with degree d , d is an eigenvalue value with multiplicity 1.

Lemma 5.56. For any adjacency matrix A , A^2 has the property that $A_{i,j}^2 = \#\text{walks of length 2 from } i \rightarrow j$. This generalizes easily to the general case for A^k

Proof. This is easily shown from the matrix multiplication

$$A_{i,j} = \sum_{k \in [n]} A(i,k)A(k,j)$$

□

Lemma 5.57. For a d -regular graph with diameter 2 graph then the maximum number of vertices is $1 + d + d(d - 1) = d^2 + 1$ vertices.

To achieve this bound, I require $\text{girth}(G) \geq 5$

Note: Girth is the length of a shortest cycle.

Another way to achieve this bound is the Peterson Graph

Lemma 5.58. Any Moore graph (regular graph whose girth is at least twice its diameter) has $A^2 + A - (d - 1)I = J$.

Proof. We know $\lambda = d$ an eigenvalue for $f \equiv 1$. By spectral theorem, A has an orthogonal basis of real eigenvectors. Since $f_j \perp f_1$ for $j \neq 1$ then $J \cdot f_j = \vec{0}$ as J is the matrix of all ones. So

$$A^2 f_j + A f_j - (d - 1)I f_j = 0 \iff \lambda_j^2 + \lambda_j - (d - 1) = 0$$

We can conclude $\lambda = \frac{-1 \pm \sqrt{4d - 3}}{2}$

We need $1 + m_2 + m_3 = n$ and $d + m_2\lambda_2 + m_3\lambda_3 = 0$

So $2d - (m_2 + m_3) + (m_2 - m_3)\sqrt{4d - 3} = 0$

(Case 1) If $\sqrt{4d + 3} \notin \mathbb{Q}$ then $m_2 = m_3$ and $2d = d^2 \implies d = 2$

(Case 2) If $\sqrt{4d - 3} = s \in \mathbb{Z}$ then $d = \frac{s^2 + 3}{4}$ then

$$2d - d^2 + (m_2 - m_3)s = 0 \iff 8\left(\frac{s^2 + 3}{4}\right) - (s^2 + 3)^2 + 16(m_2 - m_3)s = 0$$

Expanding out we have $as^4 + bs^3 + cs^2 + ds + 15 = 0$ then $s|15 \implies s = 1, 3, 5, 15 \implies d = 1, 3, 7, 57$. □

We'll be covering graph where for any 2 vertices u, v there is exactly one common neighbor of u, v

Theorem 5.59. "There is a politician": $\exists v_0$ such that $\forall u, v_0 \sim u$

Note: This doesn't hold for infinite vertices by $H_0 = 5\text{-cycle}$ and H_{i+1} is H_i with independent path of length 2 added between parts that don't have a common neighbor in H_i .

Step 1 A counterexample must be regular

Step 1A $u \not\sim v \implies \deg(v) \geq \deg(u)$. By symmetry $\deg(v) = \deg(u)$. This is from w_1 being the common neighbor of u, v and w_2 being the common neighbor of w_1, u and z_1 being common neighbor of w_1 and v .

Step 1B Let $\deg(u) = d, \forall v \neq w_i$ we get $\deg = d$ for all w_2, \dots, w_d , we get $\deg(w_i) = \deg(v) = d$. All but w_1 are known to be degree d . Since w_i not a politician then w_1 must be the politician.

Going back to the graph with diameter 2 graph then $n = 1 + d(d-1) = d^2 - d + 1$. There are exactly 1 path of length 2 between u, v $\deg(u) = d \implies A^2$ is d along diagonal and 1 everywhere else. $A^2 = J + (d-1)I$. J has e.v. n with multiplicity 1 and 0 with multiplicity $n-1$. So A^2 has e.v. $n+d-1$ with multiplicity 1 and $d-1$ with multiplicity $n-1$.

A has eigenvalues d with multiplicity 1, $\sqrt{d-1}$ with multiplicity s and $-\sqrt{d-1}$ with multiplicity t . Also $s+t = n-1$. The trace of A is 0 so $d + \sqrt{d-1}s - \sqrt{d-1}t = 0 \implies d + (s-t)\sqrt{d-1} = 0$. So $\sqrt{d-1} \in \mathbb{Q} \implies h := \sqrt{d-1} \in \mathbb{N}$.

We have $d = \sqrt{d-1}^2 + 1 = h^2 + 1$. So $d + h(s-t) = 0 \implies h^2 + 1 = h(t-s) \implies h = 1 \implies d = 2$

Theorem 5.60. Oddtown: Clubs have odd size and intersections are even. The clubs are less than number of people.

Lemma 5.61 (Fisher's Inequality). Let $\mathcal{F} = \{A_1, \dots, A_m\}$ be a family of m distinct subsets of a universe X where $|X| = n$. Suppose there exists a constant k such that $|A_i \cap A_j| = k$ for all $i \neq j$. Furthermore, assume that $|A_i| > k$ for all i . Then:

$$m \leq n$$

Proof. Let $v_1, \dots, v_m \in \mathbb{R}^n$ be the incidence vectors of the sets A_1, \dots, A_m . That is, the x -th component of vector v_i is 1 if $x \in A_i$ and 0 otherwise.

We aim to show that these vectors are linearly independent. Consider a linear combination of these vectors equal to the zero vector, with coefficients $\alpha_1, \dots, \alpha_m \in \mathbb{R}$:

$$\sum_{i=1}^m \alpha_i v_i = 0 \tag{1}$$

We take the squared Euclidean norm (the dot product with itself) of both sides:

$$\left\langle \sum_{i=1}^m \alpha_i v_i, \sum_{j=1}^m \alpha_j v_j \right\rangle = 0$$

Expanding using the linearity of the inner product:

$$\sum_{i=1}^m \alpha_i^2 \langle v_i, v_i \rangle + \sum_{i \neq j} \alpha_i \alpha_j \langle v_i, v_j \rangle = 0$$

We observe the following properties of the incidence vectors:

- $\langle v_i, v_i \rangle = |A_i|$
- $\langle v_i, v_j \rangle = |A_i \cap A_j| = k$ (for $i \neq j$)

Substituting these values into the equation:

$$\sum_{i=1}^m \alpha_i^2 |A_i| + k \sum_{i \neq j} \alpha_i \alpha_j = 0$$

To simplify the second term, we use the identity $(\sum \alpha_i)^2 = \sum \alpha_i^2 + \sum_{i \neq j} \alpha_i \alpha_j$. Rearranging this gives $\sum_{i \neq j} \alpha_i \alpha_j = (\sum \alpha_i)^2 - \sum \alpha_i^2$. We substitute this back into our equation:

$$\begin{aligned} \sum_{i=1}^m \alpha_i^2 |A_i| + k \left[\left(\sum_{i=1}^m \alpha_i \right)^2 - \sum_{i=1}^m \alpha_i^2 \right] &= 0 \\ \sum_{i=1}^m \alpha_i^2 (|A_i| - k) + k \left(\sum_{i=1}^m \alpha_i \right)^2 &= 0 \end{aligned}$$

Since we assumed $|A_i| > k$, we have $|A_i| - k > 0$. Also, squares of real numbers are non-negative (assuming $k > 0$ and observing $(\sum \alpha_i)^2 \geq 0$). Therefore, we have a sum of non-negative terms equaling zero. This implies that every individual term must be zero. Specifically:

$$\alpha_i^2 (|A_i| - k) = 0 \quad \forall i$$

Since $|A_i| - k \neq 0$, it must be that $\alpha_i = 0$ for all i .

Thus, the vectors v_1, \dots, v_m are linearly independent. Since they exist in \mathbb{R}^n , the dimension of the subspace they span cannot exceed n , implying $m \leq n$. \square

Theorem 5.62. $R(k+1) \geq \binom{k}{3} + 1$. We can group every 3 vertices then color it red

A quadratic form/homogeneous polynomial say $q(x, y) = x^2 + 2xy + 3y^2 = [x \ y] \begin{bmatrix} 1 & 1 \\ 1 & 3 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$. There is an bijection between quadratic form and symmetric matrices.

We can write $A = P^T B P$ where B is a diagonal matrix with eigenvalues

$$B = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}$$

Definition 5.63. For a symmetric matrix A ,

1. Positive definite if $\lambda_i > 0$ for all i
2. Negative definite if $\lambda_i < 0$ for all i

Theorem 5.64. Given symmetric A , $q_A(X) = X^T A X$. Let A be real, symmetric $n \times n$. Let $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$ be the eigen values of A . We can conclude

$$\lambda_k = \max_{U \subseteq V, \dim(U)=k} \min_{X \in U} \frac{X^T A X}{X^T X}$$

Remark 5.65. This statement doesn't care of the magnitude of X and only the direction. We're looking for the direction of greatest change.

Proof. We'll first show $\lambda_k \leq \max_U$. For this direction suffices to exhibit one good u . For v_1, \dots, v_n orthonormal eigen basis, v_i eigenvalues for λ_i . Let $U_k = \{v_1, \dots, v_k\}$ and let $X \in \text{Span}(U_k)$ so $X = \sum_{i=1}^k \alpha_i v_i$. WLOG $|X| = 1$ since the magnitude does not change the result. This implies $\sum \alpha_i^2 = 1$.

$$X^T A X = \left(\sum \alpha_j v_j \right)^T A \left(\sum \alpha_j v_j \right) \tag{2}$$

$$= \left(\sum \alpha_j v_j \right)^T \left(\sum \alpha_j \lambda_j v_j \right) \tag{3}$$

$$= \sum \lambda_j \alpha_j^2 \tag{4}$$

This is a weighted average of $\lambda_1, \dots, \lambda_k$ so this is at least the $\min(\lambda_1, \dots, \lambda_k) = \lambda_k$.

For the other direction $\lambda_k \geq \max_U$. Given any U_k , we want to show $\exists X \in U_K$ such that $\frac{X^T AX}{X^T X} \leq \lambda_k$. Let $W = \text{Span}(v_k, \dots, v_n)$ then $W = n - k + 1$. So there exist a vector $X \neq 0, X \in W \cap U_k$. WLOG, take $|X| = 1$. Since $X \in W, X = \sum_{j=k}^n \alpha_j v_j$ and

$$X^T AX = \sum_{j=k}^n \lambda_j \alpha_j^2 \leq \max(\lambda_k, \dots, \lambda_n) = \lambda_k$$

□

Example 5.66. Given a d -regular graph G with adjacency matrix A with $\lambda_1 \geq \dots \geq \lambda_n \geq -d$. Suppose I had a negative eigenvalue that is less than $-d$ then any vertex when applied the adjacency matrix will be the sum of the neighboring vertices, but we can't have $|\sum| > d^2$.

Given an independent S of size α . Define a vector $v = nI_S - \alpha I = (n - \alpha)I_S - \alpha I_{\bar{S}}$. Then $v \cdot I = 0$. We know

$$\min_{X \subseteq \mathbb{R}^n} \frac{X^T AX}{X^T X} = \lambda_k$$

So we know $\lambda_k \leq \frac{v^T Av}{v^T v}$.

$$v^T v = (nI_S - \alpha I)(nI_S - \alpha I) = \alpha n^2 - 2\alpha^2 n + \alpha^2 n = \alpha n(n - \alpha) \quad (5)$$

$$v^T Av = (nI_S - \alpha I)A(nI_S - \alpha I) \quad (6)$$

$$= nI_S A nI_S - 2\alpha^2 I_S A I + \alpha^2 I A I \quad (7)$$

$$= 0 - 2n\alpha I_S \begin{bmatrix} d \\ d \\ \vdots \\ d \end{bmatrix} + \alpha^2 nd \quad (8)$$

$$= -n\alpha^2 d \quad (9)$$

So we have $\frac{-n\alpha^2 d}{\alpha n(n - \alpha)} = \frac{-\alpha d}{(n - \alpha)} = \frac{d}{1 - \frac{n}{\alpha}} \geq \lambda_n$. When we solve for α

Definition 5.67. Expander Graphs

Definition 5.68. Let $G = (V, E)$ and $|V| = n$
Cheeger Constant

$$h(G) = \min_{S \subseteq V, |S| \leq \frac{n}{2}} \frac{e(S, \bar{S})}{|S|}$$

Remark 5.69. $h(G) \leq d$ and $h(G) = 0 \iff G$ is disconnected.

Definition 5.70. G bipartite on (L, R) with $|L| = |R| = n$ is a (d, α) expander if

1. every degree in L is d
2. every set S of size $\leq \frac{n}{d}$ in L has $\alpha|S|$ neighbors (in R)

Theorem 5.71. Let $d \equiv 4$, choose d edges from each vertex in L independent and 1 and only.

Claim: With constant probability, result is a $(d, \frac{d}{10})$ bipartite expander.

Proof. Let the bad events be for sets $S \subseteq L, T \subseteq R, |S| \leq \frac{n}{d}, |T| < \alpha|S|, E_{S,T} = \{N(S) \subseteq T\}$. Then

$$\mathbb{P}(E_{S,T}) = \left(\frac{|T|}{n}\right)^{d|S|}$$

$$\mathbb{P}(\exists S, T, |S| \leq \frac{n}{d}, |T| = \alpha|S|, E_{S,T}) \leq \sum_{s=1}^{n/d} \binom{n}{s} \binom{n}{\alpha s} \left(\frac{\alpha s}{n}\right)^d s \quad (10)$$

$$\leq \sum_{s=1}^{n/d} \binom{n}{\alpha s}^2 \left(\frac{\alpha s}{n}\right)^d s \quad (11)$$

$$\leq \sum_{s=1}^{\infty} \left(\frac{en}{\alpha s}\right)^{2\alpha s} \left(\frac{\alpha s}{n}\right)^{ds} \quad (12)$$

$$= \sum_{s=1}^{\infty} \left(\frac{n}{d}\right)^{(2\alpha-d)s} e^{2\alpha s} s^{(d-2\alpha)s} \quad (13)$$

$$= \sum_{s=1}^{\infty} \left(\frac{n}{\alpha s}\right)^{2\alpha-d}s e^{2\alpha s} \quad (14)$$

$$\leq \sum_{s=1}^{\infty} 10^{(2\alpha-d)s} e^{2\alpha s} \quad (\alpha \leq d/10) \quad (15)$$

□

Lemma 5.72. If p has 1 fermat witnesses, half of the a 's relatively prime to p are fermat witnesses.

Example 5.73. Prime Algorithm

1. Randomly choose a
2. Compute $a^p \pmod p$
If $\not\equiv a$, report not prime else report maybe prime

From the previous lemma if not prime, at least half the a will show it.

We want a deterministic expander graph and on the L

Theorem 5.74. Similar setup to therorem 5.64,

$$\lambda_k = \min_{\dim(U)=k-1} \max_{X \perp U} \frac{X^T A X}{X^T X}$$

Theorem 5.75. We want to relate the spectral gap to the Creeger Constant

Proof. $\vec{v} = nI_S - sI = (n-s)I_S - sI_{\bar{S}}$ with $s = |S|$
G is a d -regular graph, A is an adjacency matrix, we have

$$\lambda_2 = \max_{x \perp I} \frac{x^T A x}{x^T x} \geq \frac{\vec{v}^T A \vec{v}}{\vec{v}^T v}$$

Note: $\vec{v} \cdot I = ns - ns = 0$ and $\vec{v} \cdot \vec{v} = (n-s)^2 s + s^2(n-s) = s(n-s)n$.

Since the graph is d -regular we have $ds = 2e(S) + e(S, \bar{S})$ and $d(n - s) = 2e(\bar{S}) + e(S, \bar{S})$.

$$\begin{aligned}\vec{v}^T A \vec{v} &= \sum_{(i,j) \in E(G)} v_i v_j = 2e(S)(n-s)^2 - 2e(S, \bar{S})s(n-s) + 2e(\bar{S})s^2 \\ &= (ds - e(S, \bar{S}))(n-s)^2 - 2e(S, \bar{S})s(n-s) + (d(n-s)e(S, \bar{S}))s^2 \\ &= ds(n-s)n - e(S, \bar{S})[(n-s)^2 + 2s(n-s) + s^2] \\ &= ds(n-s)n - e(S, \bar{S})n^2\end{aligned}$$

Then substituting back we have

$$\lambda_2 \geq \frac{ds(n-s)n - e(S, \bar{S})n^2}{s(n-s)n} = d - \frac{e(S, \bar{S})n}{s(n-s)}$$

$\forall S \subseteq V, |S| \leq \frac{n}{2}$ we have

$$d - \lambda_2 \leq \frac{2e(S, \bar{S})}{|S|} \implies \lambda_1 - \lambda_2 \leq 2h(G)$$

□

Example 5.76. Consider $A_{K_n} = J - I$ has eigenvalues $n - 1$ with multiplicity 1 and -1 with multiplicity $n - 1$.

Lemma 5.77. Lower bound on λ_2 with G is d -regular, $A = A_G$, A^2 has eigen values $\lambda_1^2, \dots, \lambda_n^2$
 $\text{Trace}(A^2) = nd = \lambda_1^2 + \dots + \lambda_n^2$

$$nd - d^2 = \sum_{k=2}^n \lambda_k^2 \leq (n-1)\lambda_*^2 \quad (\lambda_* = \max_{i \neq 1} |\lambda_i|)$$

So we have $d - o(1) = \frac{nd - d^2}{n-1} \leq \lambda_*^2 \implies \lambda_* \geq \sqrt{d} - o(1)$

Theorem 5.78. Consider simple random walk on a graph G , G has adjacency matrix A and transition matrix $P = \begin{bmatrix} \frac{1}{\deg(v_1)} & \cdots \\ \vdots & \ddots \\ & & \frac{1}{\deg(v_n)} \end{bmatrix} A$

Remark 5.79. $P(i, j) = \mathbb{P}(\text{next at } j | \text{now at } i)$

If v_1, \dots, v_n are orthonormal eigenbasis to eigenvalues for A , also true for P . Let corresponding eigenvalues be $\lambda_1 \geq \dots \geq \lambda_n$

Consider now a stochastic vector x with $\sum_{i=1}^n x_i = 1, x_i \geq 0$, the product $xP = y$ where y is a stochastic vector with the new distribution given we start at distribution x .

$$x^T = \sum \alpha_i \vec{v}_i$$

Since P is symmetric

$$xP^t = P^t(\sum \alpha_i v_i) \quad (16)$$

$$= \sum_{i=1}^n (\alpha_i \lambda_i^t v_i) i \quad (17)$$

$$= \alpha_1 v_1 + \sum_{i=2}^n \lambda_i^t \alpha_i v_i \quad (18)$$

$$\leq \lambda_*^t \left(\sum \alpha_i v_i \right) \quad (19)$$

So we can conclude $xP^t \rightarrow \frac{1}{n} I$

Theorem 5.80. A knight makes random knight moves: let τ be the time to return to the bottom left corner. What is $\mathbb{E}[X]?$

Definition 5.81. $\rho = \text{probability of returning to origin for a random walk on } \mathbb{R}$ then a random walk is recurrent if $\rho = 1$ and transient otherwise.

Remark 5.82. A random walk is recurrent on Γ iff $\mathbb{E}[\text{visits to origin}]$ is ∞

$$\mathbb{P}(\text{I visit exactly } k \text{ times}) = p^{k-1}(1-p)$$

$$\mathbb{E}(\text{visits}) = \sum_{k=1}^{\infty} kp^{k-1}(1-p) = \frac{1}{1-p}$$

For our random walk on \mathbb{R} ,

$$\mathbb{E}[\text{visit}] = \sum_{n=0}^{\infty} \mathbb{E}[I_n] = \sum_{n=0}^{\infty} \frac{\binom{2n}{n}}{2^{2n}} = \sum_{n=0}^{\infty} a_n \geq \sum_{n \geq N} \frac{1}{\sqrt{\pi n}}$$

Where $a_n \sim \frac{1}{\pi n}$ by Stirling's Formula

The expected time to return to origin is not finite.

Theorem 5.83. The probability starting from j we reach n before reaching 0 is $p_j = \frac{j}{n}$. $p_j = \frac{1}{2}p_{j-1} + \frac{1}{2}p_{j+1}$

For a random walk on \mathbb{R}^2 let $I_n = 1$ if at 0 at $2n$.

$$\Pr(I_n) = \sum_n \frac{1}{4^{2n}} \sum_k \frac{(2n)!}{n!n!(n-k)!(n-k)!} = \sum_n \frac{1}{4^{2n}} \binom{2n}{n} \sum_k \binom{n}{k}^2 = \sum_n \frac{1}{4^{2n}} \binom{2n}{n}^2$$

For a random walk on \mathbb{R}^3 let $I_n = 1$ if I return after $2n$ steps.

$$\mathbb{E}[\text{visits}] = \sum_n \frac{1}{6^{2n}} \sum_{j,k} \frac{(2n)!}{j!^2 k!^2 (n-j-k)!^2} \leq \sum_n \frac{\binom{2n}{n} \binom{n}{n/3, n/3, n/3}}{2^{2n} 3^n} \sum_n \frac{\binom{n}{j, k, n-j-k}}{3^n}$$

By Stirling's formula, $\frac{n!}{(\frac{n}{3})!^3} \approx \frac{(n/e)^n}{(\frac{n}{3e})^3}$

For a random walk on a directed graph, let

$$\pi(y) = \mathbb{E}(\text{visits a random direction from } z \text{ and makes to } y \text{ before visiting } z \text{ again})$$