

Lecture 10: PROBABILISTIC METHOD

A method to show the existence of a prescribed kind of mathematical object/structure. It works by showing that if one randomly chooses objects from a specified ensemble, the probability that the result is of prescribed kind is strictly greater than 0.

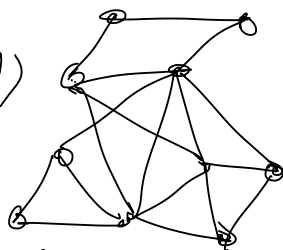
So such an object must exist.

- Even though we use probability, we conclude with certainty that an object with prescribed property exists.
- Often a non-constructive method.
- Used all over the place in mathematics & computer science
- Alon & Spencer, The Probabilistic Method, is a classic textbook

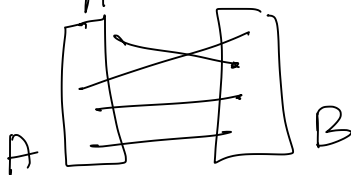
Theorem. Given any graph $G = (V, E)$, it has a bipartite subgraph with at least half the edges.

$$|E'| \geq \frac{|E|}{2}$$

$$H = (V, E')$$



Pf. (You might have seen this as an approx algo for Maximum Cut problem)



$$V = A \cup B$$

disjoint union
 $\geq \frac{1}{2}$ edges go across the cut $[A: B]$

For each $v \in V$, place $v \in A$ uniformly
or $v \in B$ at random

$$\text{For } e=(u,v) \quad \Pr[e \text{ is cut}] = \frac{2}{4} = \frac{1}{2}.$$

$\mathbb{1}_e$ = Indicator random variable for
event that ~~edge~~ e is cut.

$$\overset{\text{Expectation}}{\mathbb{E}[\mathbb{1}_e]} = \Pr[e \text{ is cut}] = \frac{1}{2}$$

$$N = \# \text{ edges cut by the process} = \sum_{e \in E} \mathbb{1}_e$$

$$\mathbb{E}[N] = \mathbb{E}\left[\sum_{e \in E} \mathbb{1}_e\right]$$

$$= \sum_{e \in E} \mathbb{E}[\mathbb{1}_e] = \sum_{e \in E} \frac{1}{2} = \frac{|E|}{2}.$$

Random process gives a cut that in expectation
(aka on average) cuts $\frac{1}{2}$ the edges

↳ Not everyone can be ^{strictly} below the average!

$\Rightarrow \exists$ cut $[A:B]$ s.t. $\geq \frac{|E|}{2}$ edges
cross it.

This "gives" the desired bipartite subgraph.
with $\geq \frac{|E|}{2}$ edges.

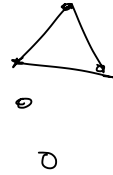
Above is a randomized Factor $\frac{1}{2}$
approximation algo for Max-Cut

For longest time, this was the best known approx. algorithm!

- In 1994, Goemans-Williamson gave an algo, based on semidefinite programming, that guaranteed a 87.8% approximation.
- This is believed to be best possible, thanks to some PCP theory.

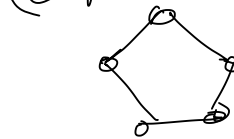
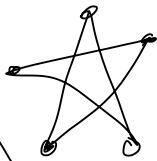
Ramsey graphs

In any graph on n (big enough) vertices, there must be a triangle (K_3) or an indep. set of size 3.

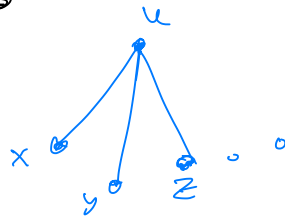


In fact $n=6$ suffices.

OTOH, for $n=5$, here's a graph with no K_3 or $\overline{K}_3$ (no clique or ind. set of 3 vertices)



Pf idea:



Let's say u has ≥ 3 nbors.

Either x, y, z is an indep. set
or u and 2 of them form a triangle.

Definition : Let $k \in \mathbb{N}$.

The k^{th} Ramsey number, denoted $R(k)$
is the minimum n such that every
 n -vertex graph has either a k -clique
or an indep. set of size k .

(Above: $R(3) = 6$)

$R(k)$ exists for every k , in fact

$$R(k) \leq \binom{2k-2}{k-1} \leq O\left(\frac{4^k}{\sqrt{k}}\right)$$

- $R(4) = 18$. (known) $\rightarrow$ (proof by induction)
- $R(5)$ is not known!

$$R(5) \in [43, 48]$$

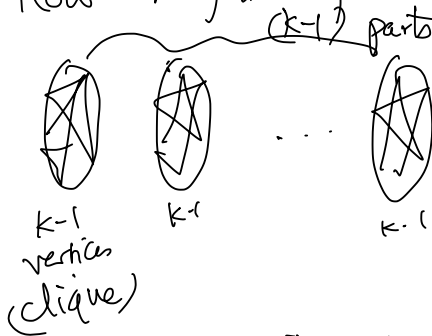
$$(R(10) \in [798, 23556])$$

$\hookrightarrow$ huge gap)

Lower bounds on $R(k)$

$R(k) > m \iff$ Existence of a graph
on m vertices with
no k -clique or k -indep set.

How might you construct such a graph?



Disjoint union of $(k-1)$ cliques of size $(k-1)$

Proves $R(k) \geq \Omega(k^2)$

For a while it was conjectured that $R(k) \leq O(k^2)$

Erdős (1947) $R(k) \geq \Omega(k \cdot 2^{k/2})$
 ($\exists$ graph on n vertices with no clique or indep set of $\approx 2 \log n$ vertices)
 How to construct such a graph?

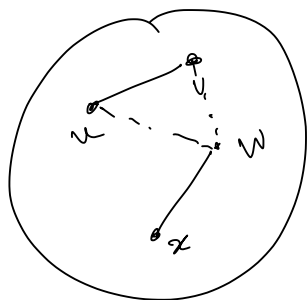
Answer: No one knows!

But Erdős showed they exist, in fact in abundance.

Proof: (Birth of the probabilistic method)
 $\rightarrow$ (alongside [Shannon 1948] proved existence of good error correction schemes)

Let's prove $R(k) > 2^{k/2}$ (for simplicity)
 (define $n = 2^{k/2}$)

PICK A RANDOM GRAPH ON n vertices.



n vertices

For each pair $\{u, v\}$,
include an edge between them
with probability $\frac{1}{2}$.
(independently for each pair)

Each possible (labeled) graph is sampled
with probability $\frac{1}{2^{\binom{n}{2}}}$

Let's Count the number of subsets that
induce a k -clique or k -ind. set. ("bad")

$G = (V, E)$ - realization of random gh.

Fix $S \subseteq V$, $|S| = k$

$X_S = X_S(G) =$ indicator r.v. that S is "bad"
i.e. S induces a k -clique or k -ind. set.

$$N = N(G) := \sum_{\substack{S \subseteq V \\ |S| = k}} X_S$$

$N = \#$ bad subsets
that violate the desired
Ramsey property.
(would like $N = 0$)

If we prove

$\mathbb{E}[N] < 1$, then that

would mean, $\exists G$ s.t. $N(G) > 0$
 ("Not everybody can be strictly above average")
 → Such a G would be desired Ramsey graph.

$$E[N] = E\left[\sum_S X_S\right]$$

(Linearity of
Expectation) =

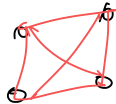
$$\sum_{\substack{S \subseteq V \\ |S|=k}} E[X_S]$$

$$= \sum_{\substack{S \subseteq V \\ |S|=k}} \Pr[S \text{ induces a } k\text{-clique or a } k\text{-indep-set}]$$

↓ ??

$$\Pr[S \text{ induces a } k\text{-clique}] + \Pr[S \text{ induces a } k\text{-ind set}]$$

$k=4$



$$\frac{1}{2^{\binom{k}{2}}}$$

+

$$\frac{1}{2^{\binom{k}{2}}}$$

$$= 2^{1 - \binom{k}{2}}$$

$$E[N] = \binom{n}{k} 2^{1 - \binom{k}{2}}$$

Recall
 $n = 2^{k/2}$

$$\leq \frac{n^k}{k!} 2^{1 - \frac{k(k-1)}{2}}$$

$$= \frac{2^{\frac{k}{2}}}{k!} \cdot 2^{1 - \frac{k}{2} + \frac{k}{2}}$$

$$= \frac{2^{\frac{k}{2} + 1}}{k!} < 1 \quad (\text{for } k \geq 4)$$

(in fact ≈ 0 for large k)

Want ~~obj~~ graph with $N=0$

$$\Pr[N > 0] = \Pr[N \geq 1] \leq \mathbb{E}[N]$$

(Markov's inequality)
 $\approx 0 \ll 1$

$$\Pr[N=0] > 0 \quad (\text{and in fact } \approx 1)$$

- Lots of recent work on explicit constructions of Ramsey graphs
- Randomness extraction.

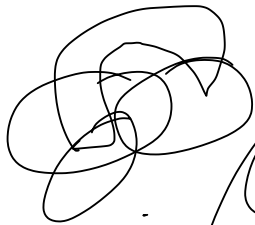
Prob. method useful to not only show existence of objects, but also to rule out certain objects (e.g. in extremal combinatorics)

Erdős-Ko-Rado Theorem

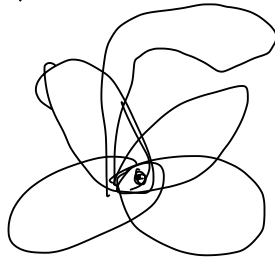
Let $\mathcal{F}$ be a family of k -element subsets of a universe of size n .

($n \geq 2k$). Suppose every pair of sets $A, B \in \mathcal{F}$ have non-empty intersection.

Then $|\mathcal{F}| \leq \binom{n-1}{k-1}$



(There's a very slick proof of this using prob. method, via random orderings).



$\binom{n-1}{k-1} = \#$ subsets of size k that contain a specific element.