# Probabilistic Method and Random Graphs Lecture 8. The Method of Counting\&Expectation 

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${ }^{1}$ The slides are mainly based on Chapter 6 of Probability and Computing.

## Comments, questions, or suggestions?

## Recap of Lecture 6

- Random graph models
$-\mathscr{G}_{n, p}, \mathscr{G}_{n, m}$
- Decoupling: $\mathscr{G}_{n, m} \sim\left(\mathscr{G}_{n, p} \mid\right.$ there are $m$ edges $)$

- Other random graphs: with specified degree distribution, scale-free, small world


## Recap of Lecture 6

- The existence of (sharp) threshold functions
- Mostly on $\mathscr{G}_{n, p}$
- Sharp threshold functions of connectivity: $\frac{\ln n}{n}$
- That of major component existence: $\frac{1}{n}$
- A Hamiltonian cycle can be found efficiently
w.h.p. in $\mathscr{G}_{n, p}$ with $p \geq 40 \frac{\ln n}{n}$


## Probabilistic Method <br> -Elegance from graph theory

- A warm-up example:
- n players against each other
- "Top-k" players get prize

- But, are you sure no controversy exists?
- Controversy: prize-winners all defeated by a loser
- Unfortunately, when players are too many, controversy does exist w.h.p.


## Proof (non-constructive)

- $S$ : a $k$-subset of players
- $A_{S}$ : no player beats all member of $S$
- Consider a random tournament
- $\operatorname{Pr}\left(A_{S}\right)=\left(1-2^{-k}\right)^{n-k}$
- $\operatorname{Pr}($ no controversy $)=\operatorname{Pr}\left(\cup A_{S}\right)$

$$
\begin{aligned}
& \leq \sum \operatorname{Pr}\left(A_{S}\right)=\binom{n}{k}\left(1-2^{-k}\right)^{n-k} \\
& =o\left(\frac{1}{n}\right)
\end{aligned}
$$

- Find a controversial one? Just sampling


## Cool?

A piece of cake in probabilistic method!

## What is the Probabilistic Method?

- Proving the existence of an object satisfying a certain property without constructing it
- Underlying principle

- Pioneered by Erdős in 1940's
- Construct by flipping coins
- Naturally lead to (randomized) algorithms


## Main Probabilistic Methods

- Counting argument
- First-moment method
- Second-moment method
- Higher-moment methd
- Lovasz local lemma



## Counting Argument

- Construct a probability space and calculate the probability
- Algorithm design: sampling
- Application
- Tournament
- Ramsey number: an observation by S. Szalai, 1950's


## Ramsey Number

- Given a complete graph $K(n)$, 2-color its edges
- Ramsey number $R(k, l)$
- the smallest number $n$ such that for every complete graph $G$ with at least $n$ vertices, any 2-coloring of $G$ would either has a red $K(k)$ or a blue $K(l)$

There is a coloring without
red $K(k)$ and blue $K(l) \quad R(k, l)$

Every coloring has either a red $K(k)$ or a blue $K(l)$

Number of vertices of the complete graph

## Ramsey Number is Well defined

- Ramsey Theorem: $R(k, l)$ is finite for any $k, l$
- Proved by F. Ramsey in 1930 and determined $R(3,3)$
- Origin of Ramsey theory
- The existence of rather large good substructure in a big structure
- How much is $R(k, l)$ ?
- Upper bound: $R(k, l) \leq R(k-1, l)+R(k, l-1)$
- Proved by P. Erdős and G. Szekeres in 1935
- The $2^{\text {nd }}$ cornerstone of Ramsey theory
- By $R(k, 2)=R(2, k)=k, R(k, l) \leq\binom{ k+l-2}{k-1}$
- It implies $R(k, k) \leq 4^{k}$
- Best: $k^{-c \frac{\ln k}{\ln \ln k} 4^{k}}$ by Conlon in 2009


## Known bounding ranges

| $r s$ | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 |
| 2 |  | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 |
| 3 |  |  | 6 | 9 | 14 | 18 | 23 | 28 | 36 | 40-42 |
| 4 |  |  |  | 18 | $25^{[5]}$ | 36-41 | 49-61 | $59{ }^{[10]}-84$ | 73-115 | 92-149 |
| 5 |  |  |  |  | $43-48$ | 58-87 | 80-143 | 101-216 | 133-316 | $149{ }^{[10]}-442$ |
| 6 |  |  |  |  |  | 102-165 | $115{ }^{[10]}-298$ | 134 ${ }^{[10]}-495$ | 183-780 | 204-1171 |
| 7 |  |  |  |  |  |  | 205-540 | 217-1031 | 252-1713 | 292-2826 |
| 8 |  |  |  |  |  |  |  | 282-1870 | 329-3583 | 343-6090 |
| 9 |  |  |  |  |  |  |  |  | 565-6588 | 581-12677 |
| 10 |  |  |  |  |  |  |  |  |  | 798-23556 |

Vigleik Angeltveit; Brendan McKay (2017). " $R(5,5) \leq 48$ ". arXiv

## Proof of the upper bound

- 2-color the complete graph on $R(k-1, l)+$ $R(k, l-1)$ vertices
- Pick a vertex $u$
- Define subgraphs $G_{r}$ and $G_{b}: \forall v$,
$-v \in \begin{cases}G_{r} & \text { if }(u, v) \text { is red } \\ G_{b} & \text { if }(u, v) \text { is blue }\end{cases}$
- Either $\left|G_{r}\right| \geq R(k-1, l)$ or $\left|G_{b}\right| \geq R(k, l-1)$
- Do case-by-case analysis


## Example: Ramsey Number $R(3,3)$

$$
R(3,3) \leq 6
$$

Actually, $R(3,3)>5$


## What is $R(k, k)$ ?

- $R(k, k)>2^{k / 2}=\sqrt{2}^{k}$ (Erdős, 1947)
- Best: $[1+o(1)] \frac{k}{e} \sqrt{2}^{1+k}$ by Spencer in 1975
- For any complete graph with at most $2^{k / 2}$ vertices, there is a 2 -coloring without red $K(k)$ and blue $K(k)$
- Prove by the probabilistic method


## Proof

- Consider a graph $G$ on $n$ vertices, and 2 -color each edge randomly
- Uniform distribution on all 2-coloring
- $A_{S}$ : the subgraph on $S$ is monochromatic
- $S$ is a $k$-subset of the vertices
- $\operatorname{Pr}\left(A_{S}\right)=2^{1-\binom{k}{2}}$
- $\operatorname{Pr}\left(\mathrm{U}_{S} A_{S}\right) \leq\binom{ n}{k} 2^{1-\binom{k}{2}}<\frac{2^{1+\frac{k}{2}}}{k!} \frac{n^{k}}{\frac{k}{}^{\frac{k^{2}}{2}}}<1 \quad$ (If $\left.n=\left\lfloor 2^{k / 2}\right\rfloor\right)$
- There is a 2 -coloring of the edges such that $K\left(\left[2^{k / 2}\right]\right)$ has no monochromatic $k$ subgraph
- Together with the definition of $R(k, k), R(k, k)>2^{k / 2}$


## Randomized Algorithms

- But how to find a good coloring? By sampling!
- General approach

- Prerequisites
- Efficient sampling
- Small probability of failure
- Efficient verification (Las Vegas only)


## First-Moment method

- Use the expectation in probabilistic reasoning
- Two types of first-moment method
- Expectation argument

$$
\operatorname{Pr}(X \geq \mathbb{E}[X])>0, \operatorname{Pr}(X \leq \mathbb{E}[X])>0
$$

- Markov's inequality for non-negative $X$
- $\operatorname{Pr}(X \geq a) \leq \frac{\mathbb{E}[X]}{a}$
- When $X$ is integer-valued,

$$
\operatorname{Pr}(X \neq 0)=\operatorname{Pr}(X>0)=\operatorname{Pr}(X \geq 1) \leq \mathbb{E}[X]
$$

## First-Moment argument

- 3-CNF Boolean formula

$$
-\left(x_{1} \vee x_{2} \vee x_{3}\right) \wedge \ldots \wedge\left(\overline{x_{1}} \vee \overline{x_{3}} \vee x_{4}\right)
$$

- For such a formula, at most how many clauses can be satisfied simultaneously?
- MAX-3SAT is NP-hard
- Theorem: there is a truth assignment which satisfies $\geq \frac{7}{8}$-fraction of the clauses


## Proof

- Randomly assign truth values to each variable
- Define r.v. $X_{i}$ indicating whether clause $i$ is true
- $\mathbb{E}\left[X_{i}\right]=\frac{7}{8} \Rightarrow \mathbb{E}\left[\sum_{i=1}^{n} X_{i}\right]=\frac{7}{8} n$
- Remark: probability of sampling a good truth assignment $\geq \frac{1}{n+1}$, leading to an efficient alg.
- Optimum, since impossible to get a $\left(\frac{7}{8}+\varepsilon\right)$-approx.
- J. Hastad. Some optimal inapproximability results. STOC 1997


## Proof of $\operatorname{Pr}\left(\sum X_{i} \geq \frac{7}{8} n\right) \geq \frac{1}{n+1}$

- Let $X=\sum X_{i}$ and $p=\operatorname{Pr}\left(X \geq \frac{7}{8} n\right)$
- $\frac{7}{8} n=\mathbb{E}[X]$

$$
\begin{aligned}
& =\sum_{i<\frac{7}{8} n} i * \operatorname{Pr}(X=i)+\sum_{i \geq \frac{7}{8} n} i * \operatorname{Pr}(X=i) \\
& \leq\left(\frac{7}{8} n-\frac{1}{8}\right)(1-p)+n p \\
& =\frac{7}{8} n-\frac{1}{8}+\frac{n+1}{8} p
\end{aligned}
$$

## Expectation argument

- Turán Theorem
- Any graph $G=(V, E)$ contains an independent set of size at least $\frac{|V|}{D+1}$, where $D=\frac{2|E|}{|V|}$
- Proof: Consider the following random process for constructing an independent set $S$ :
- Initialize $S$ to be the empty set
- For each vertex $u$ in $V$ in random order, if no neighbors of $u$ are in $S$, add $u$ to $S$
- Return $S$


## Proof (Continued)

- $S$ is an independent set
- Vertex $u$ is selected with probability $\geq \frac{1}{d(u)+1}$
- See the next slide
- So, $\mathbb{E}[|S|] \geq \sum \frac{1}{d(u)+1} \geq \frac{|V|}{D+1}$ due to convexity
- Remark: probability of sampling a good independent set is $\geq \frac{1}{2 D|V|^{2}}$


## Proof: $\operatorname{Pr}(u$ is selected $) \geq \frac{1}{d(u)+1}$

- $u$ is selected if and only if $A$ occurs
- $A$ : when sampling first occurs in the neighborhood of $u, u$ rather than its neighbors is sampled
- Neighborhood: $u$ and its then-valid neighbors
- Denote the neighborhood by $N$, and the number of then-valid neighbors by $x$. Note that $x \leq d(u)$
- $\operatorname{Pr}(A)=\operatorname{Pr}(u$ is chosen $\mid$ sampling occurs in $N)$

$$
=\frac{1}{x+1} \geq \frac{1}{d(u)+1}
$$

## References

- http://www.cse.buffalo.edu/~hungngo/classe s/2011/Spring-694/lectures/sm.pdf
- http://www.cse.cuhk.edu.hk/~chi/csc51602007/notes/Probabilistic.ppt
- Erdős. Graph theory and probability I. 1959
- Erdős. Graph theory and probability II. 1961
- Alon\&Krivelevich. Extremal and Probabilistic Combinatorics. 2006

