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

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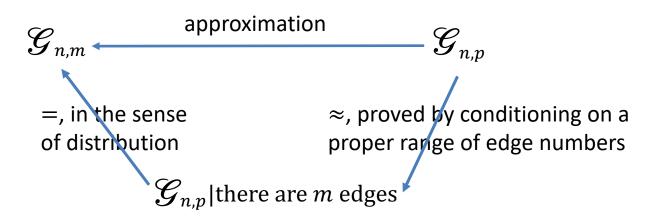
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¹The slides are mainly based on Chapter 6 of Probability and Computing.

Comments, questions, or suggestions?

Recap of Lecture 6

- Random graph models
 - $-\mathcal{G}_{n,p}, \mathcal{G}_{n,m}$
 - Decoupling: $\mathcal{G}_{n,m} \sim (\mathcal{G}_{n,p} | \text{there are } m \text{ edges})$



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

Recap of Lecture 6

- The existence of (sharp) threshold functions
 - Mostly on $\mathcal{G}_{n,p}$

– 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 $p \ge 40 \frac{\ln n}{n}$

Probabilistic Method -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

•
$$\Pr(A_S) = (1 - 2^{-k})^{n-k}$$

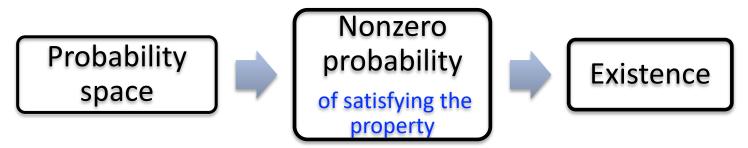
- Pr(no controversy) = Pr($\bigcup A_S$) $\leq \sum \Pr(A_S) = \binom{n}{k} (1 - 2^{-k})^{n-k}$ $= o\left(\frac{1}{n}\right)$
- Find a controversial one? Just sampling

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

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Counting Argument

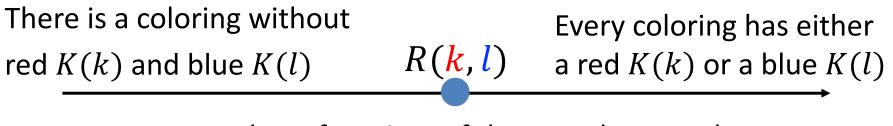
 Construct a probability space and calculate the probability

• 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 Gwould either has 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: $R(k, l) \le R(k 1, l) + R(k, l 1)$
 - Proved by P. Erdős and G. Szekeres in 1935
 - The 2nd cornerstone of Ramsey theory

$$- \operatorname{By} R(k,2) = R(2,k) = k, R(k,l) \le \binom{k+l-2}{k-1}$$

- 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 r	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)≤48". <u>arXiv</u>

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 : $\forall v$,

$$-v \in \begin{cases} G_r & \text{if } (u, v) \text{ is red} \\ G_b & \text{if } (u, v) \text{ is blue} \end{cases}$$

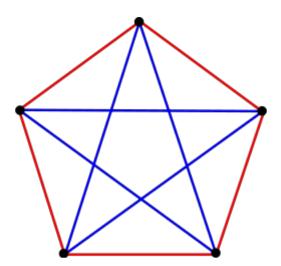
• Either $|G_r| \ge R(k-1, l)$ or $|G_b| \ge R(k, l-1)$

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
- $\Pr(A_S) = 2^{1 \binom{k}{2}}$

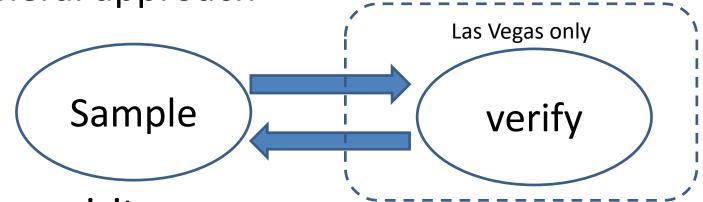
•
$$\Pr(\bigcup_{S} A_{S}) \le {\binom{n}{k}} 2^{1 - {\binom{k}{2}}} < \frac{2^{1 + \frac{k}{2}}}{k!} \frac{n^{k}}{2^{\frac{k^{2}}{2}}} < 1 \quad (\text{If } n = \lfloor 2^{k/2} \rfloor)$$

• There is a 2-coloring of the edges such that $K(\lfloor 2^{k/2} \rfloor)$ 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 $Pr(X \ge \mathbb{E}[X]) > 0, Pr(X \le \mathbb{E}[X]) > 0$
 - Markov's inequality for non-negative X

•
$$\Pr(X \ge a) \le \frac{\mathbb{E}[X]}{a}$$

• When X is integer-valued, $Pr(X \neq 0) = Pr(X > 0) = Pr(X \ge 1) \le \mathbb{E}[X]$

First-Moment argument

• 3-CNF Boolean formula

 $-(x_1 \lor x_2 \lor x_3) \land \dots \land (\overline{x_1} \lor \overline{x_3} \lor x_4)$

 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

•
$$\mathbb{E}[X_i] = \frac{7}{8} \Rightarrow \mathbb{E}[\sum_{i=1}^n X_i] = \frac{7}{8}n$$

• 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 $\Pr\left(\sum X_i \ge \frac{7}{8}n\right) \ge \frac{1}{n+1}$

• Let $X = \sum X_i$ and $p = \Pr\left(X \ge \frac{7}{8}n\right)$

•
$$\frac{7}{8}n = \mathbb{E}[X]$$

$$= \sum_{i < \frac{7}{8}n} i * \Pr(X = i) + \sum_{i \ge \frac{7}{8}n} i * \Pr(X = i)$$

$$\leq \left(\frac{7}{8}n - \frac{1}{8}\right) (1 - p) + np$$

$$= \frac{7}{8}n - \frac{1}{8} + \frac{n+1}{8}p$$

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,
$$\mathbb{E}[|S|] \ge \sum \frac{1}{d(u)+1} \ge \frac{|V|}{D+1}$$
 due to convexity

• **Remark:** probability of sampling a good independent set is $\geq \frac{1}{2D|V|^2}$

Proof: $\Pr(u \text{ is selected}) \ge \frac{1}{d(u)+1}$

- *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 $x \leq d(u)$
- Pr(A) = Pr(u is chosen | sampling occurs in N)

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

References

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