The Probabilistic Method - Probabilistic Techniques

Lecture 1: "Introduction - The Method of Positive Probability"

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A powerful tool used in many applications in different topics:

I) Study of random graph models ($G_{n,p}$, $G_{n,R}$, $G_{n,k}$ etc) which are:

- typical instances for average case analysis of graph algorithms and
- abstract models of modern networks (sensor networks, social networks etc.)

II) Design and analysis of randomized algorithms:

- evolution based on random choices
- solutions provided a) either are always correct but their running time is a random variable (Las Vegas algorithms) b) or may be erroneous but are correct w.h.p. (Monte Carlo algorithms)
- trade-off performance (faster, simpler) with very small, controlled error probability.

The core of the method

The Probabilistic Method

- uses simple techniques
 - the Basic Method
 - Linearity of Expectation
- as well as complex ones
 - the Local Lemma
 - Martingales
 - Markov Chains

but there is a common, underlying concept:

The core of the method

Non-constructive (μη-κατασκευαστική) proof of existence of combinatorial structures that have certain desired properties.

The Basic Method (method of "positive probability")

- Construct (by using abstract random experiments) an appropriate probability sample space of combinatorial structures (thus, the sample points correspond to the combinatorial structures whose existence we try to prove).
- Prove that the probability of the desired property in this space is positive (i.e. non-zero).

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There is at least one point in the space with the desired property.

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There is at least one combinatorial structure with the desired property.

- comprehensible, pretty short proofs
- simple (basic knowledge of Probabilistic Theory, Graph Theory, Combinatorics suffices)
- elegant
 - qualitative ideas, subtle notions
 - not lengthy, mechanical operations
- still very powerful (use to resolve extremely difficult problems)

Examples in this lecture

- (i) Monochromatic arithmetic progressions (Van der Waerden property)
- (ii) Ramsey Numbers

Definition 1

W(k) is the smallest natural number n, such that for any two-coloring of the numbers 1, 2, ..., n there is a monochromatic arithmetic progression of k terms.

Theorem 1

$$W(k) > 2^{\frac{k}{2}}$$

Proof of Theorem 1 (1/3)

- We construct a probability space by two-coloring the numbers 1, 2, ..., *n* at random, equiprobably for the two colors and independently for every number. Clearly, the sample points of this space are random two-colorings of the *n* numbers.
- \blacksquare Let S be any fixed arithmetic progression of k terms.
- Define the event $M_S := \{S \text{ is monochromatic}\}.$
 - i.e, all terms of *S* must have the same color.
- Compute the probability $\Pr[M_S]$.
 - every term is colored red (or blue) with probability 1/2
 - all *k* terms are red-colored (or blue-colored) with probability $(\frac{1}{2})^k$

$$\Pr[M_S] = \left(\frac{1}{2}\right)^k + \left(\frac{1}{2}\right)^k = 2^{1-k}$$

Proof of Theorem 1 (2/3)

- Define the event M := {∃ at least one monochromatic arithmetic progression of k terms } ⇒ M = ⋃_{|S|=k} M_S.
- An arithmetic progression of *k* terms is defined uniquely by its two first terms ⇒ There are at most $\binom{n}{2}$ arithmetic progressions ⇒ $\#(S : |S| = k) \le \binom{n}{2}$
- Using Boole's inequality we can compute Pr[*M*]

$$\Pr[M] = \Pr\left\{\bigcup_{|S|=k} M_S\right\} \le \sum_{|S|=k} \Pr[M_S] \le \binom{n}{2} 2^{1-k}$$

• We easily get:

$$\Pr[M] < \frac{n^2}{2} 2^{1-k} = \frac{n^2}{2^k}$$

- If $n < 2^{\frac{k}{2}}$ then $\Pr[M] < 1 \Rightarrow \Pr[\overline{M}] > 0$.
- Hence, there is a two-coloring without a monochromatic arithmetic progression of *k* terms when $n < 2^{\frac{k}{2}}$.
- Thus, $W(k) > 2^{\frac{k}{2}}$.

(II) Ramsey Numbers

Definition 2

The Ramsey number R(k, l) is the smallest integer n such that in any two-coloring of the edges of the complete graph on n vertices K_n by red and blue colors, either there is a red K_k or there is a blue K_l .

Difficulty of computation:

- Ramsey (1930) proved that *R*(*k*, *l*) is finite
- Greenwood and Gleason (1955) computed R(3,3) = 6 and R(4,4) = 18
- since then there is no notable progress *R*(4, 5) is still unknown
- Erdös suggested that R(6,6) is too difficult to be computed

R(k,k): diagonal Ramsey number (a monochromatic K_k is required).

Theorem 2 (Erdös, 1947)

$$lf\binom{n}{k}2^{1-\binom{k}{2}} < 1$$
 then $R(k,k) > n$.

Proof of Theorem 2 (1/3)

- Construct a probability sample space by two-coloring at random, equiprobably (for the two colors) and independently (for the edges) every edge of K_n.
- Let *S* be any fixed set of *k* vertices and consider the edges induced.
- Define the event $M_S := \{S \text{ is monochromatic}\}.$
 - i.e. all $\binom{k}{2}$ edges in *S* have the same color.
- Compute the probability $\Pr[M_S]$.
 - every edge is colored red (or blue) with 1/2 probability

$$\Pr[M_S] = \left(\frac{1}{2}\right)^{\binom{k}{2}} + \left(\frac{1}{2}\right)^{\binom{k}{2}} = 2^{1 - \binom{k}{2}}$$

■ Define the event M := {∃ at least one monochromatic set of k vertices}.

• Hence,
$$M = \bigcup_{|S|=k} M_S$$
.

• Using Boole's inequality we can compute the Pr[M]

$$\Pr[M] \le \sum_{|S|=k} \Pr[M_S] = \binom{n}{k} 2^{1-\binom{k}{2}}$$

- If $\Pr[M] < 1 \Rightarrow \Pr[\overline{M}] > 0$ $\Rightarrow \text{ if } \binom{n}{k} 2^{1 - \binom{k}{2}} < 1 \text{ then there is a point in the sample space}$ without $M \Rightarrow$ there is a monochromatic K_k .
- Hence, it must be R(k,k) > n.

Lower Bound of Ramsey Numbers

- We proved that if $\binom{n}{k} 2^{1 \binom{k}{2}} < 1$ then R(k, k) > n
- If ⁽ⁿ⁾_k 2^{1−(^k₂)} ~ 1 then we can find the best possible lower bound for R(k, k) (with this derivation).
- By using Stirling's formula and binomial approximation we obtain:

$$\frac{n^{\underline{k}}}{k!} \cdot 2^{1-\binom{k}{2}} \sim \frac{n^{k}}{\sqrt{2\pi k} \left(\frac{k}{e}\right)^{k}} \cdot 2^{-\frac{k^{2}}{2}} \sim 1$$
$$\Rightarrow n^{k} \sim \sqrt{2\pi k} \cdot \left(\frac{k}{e}\right)^{k} \cdot 2^{\frac{k^{2}}{2}}$$
$$\Rightarrow R(k,k) > n \sim \frac{k}{e\sqrt{2}} 2^{\frac{k}{2}}$$

Bibliography

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Relevant postgraduate course

ETY course

"Randomized Algorithms"

www.ceid.upatras.gr/webpages/courses/randalgs/index.html