# Advanced Algorithms (I)

Shanghai Jiao Tong University

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March 2nd, 2020

#### Information

Instructor: 张驰豪 (chihao@sjtu.edu.cn)

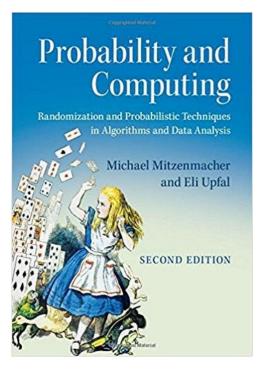
TA: 杨凤麟 (yangfl@sjtu.edu.cn)

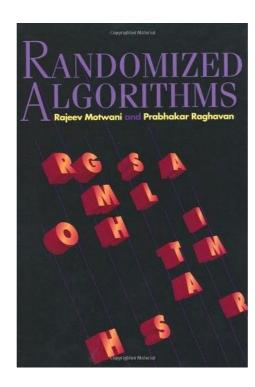
Every Monday, 10:00 am - 11:40 am

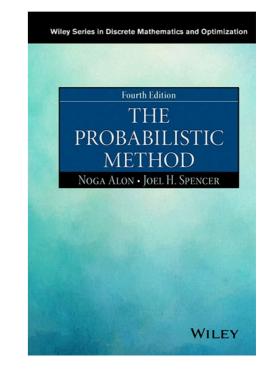
Zoom @ 123363659

Office Hour: via Canvas or WeChat Group

### References







Probability and Computing

M. Mitzenmacher & E. Upfal

Randomized Algorithms

R. Motwani & P. Raghavan

The Probabilistic Method

N. Alon & J. Spencer

# **Polynomial Identity Testing**

Given two polynomials

$$F(x) = \prod_{i=1}^{d} (x - a_i)$$
 and  $G(x) = \sum_{i=0}^{d} b_i x^i$ 

Is  $F(x) \equiv G(x)$ ?

Example.  $F(x) = (x - 1)(x - 2)(x + 3); G(x) = x^3 - 7x + 6$ 

One can expand F(x) and compare the coefficients...

It takes  $O(d^2)$  arithmetic operations.

Can be improved to  $O(d \log d)$  using FFT.

If random coins are allowed...

- The problem can be solved much faster
- at the cost of making error.

#### Choosing a uniform number $s \in \{1, 2, ..., 100d\}$

Test whether F(s) = G(s)

- If  $F(x) \equiv G(x)$ , then it always holds that F(s) = G(s)
- If  $F(x) \not\equiv G(x)$ , how likely is that  $F(s) \neq G(s)$ ?

**Theorem**. (Fundamental Theorem of Algebra) A polynomial of degree d has at most d roots in  $\mathbb{C}$  Our algorithm outputs wrong answer only when

- $F(x) \not\equiv G(x)$ ; and
- *s* is a root of F(x) G(x)

This happens with probability at most  $\frac{1}{100}$ .

It only costs O(d) operations to compute F(s) and G(s).

One can repeat the algorithm *t* times:

- error reduces to  $100^{-t}$ ;
- cost increases to O(td).

## Multi-variable Polynomials

The idea applies to a more general setting.

Let  $F, G \in \mathbb{F}(x_1, ..., x_n)$  for some field  $\mathbb{F}$ ,

$$| \text{Is } F(x_1, ..., x_n) \equiv G(x_1, ..., x_n)?$$

#### Theorem. (Schwartz-Zippel Theorem)

Let  $Q \in \mathbb{F}[x_1, ..., x_n]$  be a non-zero multivariate polynomial of degree at most d. For any set  $U \subseteq \mathbb{F}$ , it

holds that

$$\Pr_{r_1,...,r_n \in_R U} [Q(r_1,...,r_n) = 0] \le \frac{d}{|U|}$$

# Proof of Schwartz-Zippel

Induction on *n*, case n = 1 is Fundamental Theorem of Algebra.

Assuming it holds for smaller *n*...

$$Q(x_1, ..., x_n) = \sum_{i=0}^k x_1^i \cdot Q_i(x_2, ..., x_n)$$

$$\Pr\left[Q=0\right] \le \Pr\left[Q_k=0\right] + \Pr\left[Q=0 \,|\, Q_k \neq 0\right] \le \frac{d-k}{|\, U\,|} + \frac{k}{|\, U\,|}$$

Our randomized algorithm generalizes to the multivariable setting by Schwartz-Zippel theorem.

If the polynomials are given in product form, one scan of F and G is sufficient to evaluate them.

Linear time algorithm with at most 1 % error!

It is a wide open problem in the complexity theory that whether this can be done in deterministic polynomial time.

# Some Complexity Theory

Problems solvable in deterministic polynomial-time:  ${f P}$ 

Problems solvable in randomized polynomial-time: BPP

Is 
$$\mathbf{BPP} = \mathbf{P}$$
?

## Min-Cut in a Graph

A *cut* in a graph G = (V, E) is a set of edges  $C \subseteq E$  whose removal disconnects G.

How to find the minimum cut?

It can be solved using max-flow techniques

With the fastest max-flow algorithm, it takes  $O(n \times mn)$  time.

### Karger's Min-Cut Algorithm

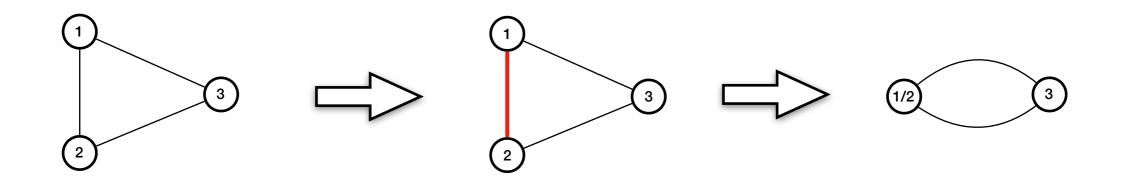


Using random bits, Karger found a much simpler algorithm

The only operation required is edge contraction

David Karger

### **Edge Contraction**



- no self-loop
- parallel edges may exist

## The Algorithm

#### Karger's Min-cut Algorithm

- Randomly choose an edge and contract it until only two vertices remains.
- 2. Output remaining edges.

## Analysis

The algorithm contracts n - 2 pair of vertices in total.

Fix an minimum cut C, we bound the probability that it survives.

Assume the removal of *C* separates  $S \subseteq V$  and  $\overline{S} = V \setminus S$ .

All contractions happen within S or  $\overline{S}$ .

For i = 1, ..., n - 2, let  $A_i$  be the event that "*i*-th contraction avoids *C*"

We need to bound

$$\Pr\left[\bigcap_{i=1}^{n-2} A_i\right] = \prod_{i=1}^{n-2} \Pr\left[A_i \middle| \bigcap_{j=1}^{i-1} A_j\right]$$

We assume |C| = k

In *i*-th contraction,

- the graph contains n i + 1 vertices;
- each vertex is of degree at least *k*.

Therefore, conditional on that C still survives,

$$\Pr\left[A_i \middle| \bigcap_{j=1}^{i-1} A_j \right] \ge 1 - \frac{2k}{k(n-i+1)} = \frac{n-i-1}{n-i+1}$$

#### Therefore

$$\Pr\left[\bigcap_{i=1}^{n-2} A_i\right] = \prod_{i=1}^{n-2} \Pr\left[A_i \middle| \bigcap_{j=1}^{i-1} A_j\right]$$
$$\geq \prod_{i=1}^{n-2} \frac{n-i-1}{n-i+1}$$
$$= \frac{n-2}{n} \cdot \frac{n-3}{n-1} \cdot \frac{1}{3}$$
$$= \frac{2}{n(n-1)}$$

So if we repeat the algorithm  $50n^2$  times, the minimum cut survives with probability at least

$$1 - \left(1 - \frac{2}{n(n-1)}\right)^{50n^2} \ge 1 - e^{-100}$$

How about the time cost?

If we store the graph in a adjacency matrix, one needs O(n) to contract an edge...

Recall Pr 
$$\begin{bmatrix} n-2\\ \bigcap_{i=1}^{n-2} A_i \end{bmatrix} = \frac{n-2}{n} \cdot \frac{n-3}{n-1} \cdots \frac{1}{3}$$

The more we contracts, the easier C gets hit

Idea: Make a copy before it becomes too bad!

The success probability can be improved to  $\Omega\left(\frac{1}{\log n}\right)$