

Lecture 1 — 9th January 2013

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The goal of this course is to present a formal definition of randomized algorithms and some easy applications.

1.1 Formal basis

1.1.1 Typology of problems

Given an input x , the purpose of a problem is to search for an appropriate output:

- *decision problem*: *ACCEPT* or *REJECT*
- *functional problem*: $F(x)$
- *relational problem*: y such that $x\mathcal{R}y$

1.1.2 Deterministic and randomized algorithms

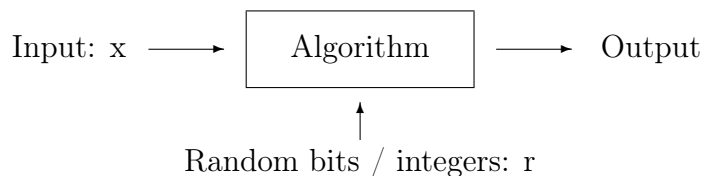
Deterministic algorithm



Goal:

- correctly solve the problem on all inputs
- efficiently: linear or polynomial time on input size

Randomized algorithm



A randomized algorithm, compared to a deterministic algorithm, has an additional input: the random variable r . We suppose that we have access to a source of uniform random bits or integers (which is basically equivalent).

Remarks:

- Behaviour depends on both x and r .
- Once r is fixed, the algorithm is deterministic.

Goal: find a randomized algo such that on all inputs:

- *Monte Carlo algorithms:* output is correct for most of random choices r , complexity is small for all random choices r
- *Las Vegas algorithms:* output is correct for all random choices r , complexity is small in average over random choices r

NB: We do not know yet how to generate random numbers with computers, we have only access to pseudo-random generators.

1.1.3 Typology of randomized algorithms

Definition 1.1. *In this course we will study 3 types of randomized algorithms, presented here for a functional problem.*

1. Algorithm A computes f without error and with average complexity T if for all inputs x :

- for all random choices r , $A(x, r) = f(x)$
- $\mathbb{E}_r[\text{complexity}(A(x, r))] \leq T$

NB: T is generally a function of size of x

Example: Quicksort with random pivot.

2. Algorithm A computes f without error, with probability $\delta < 1$ to abort and with complexity T if for all inputs x :

- for all random choices r such that $A(x, r)$ does not abort, $A(x, r) = f(x)$
- for all random choices r , $\text{complexity}(A(x, r)) \leq T$
- $\mathbb{P}_r[A(x, r) \text{ aborts}] \leq \delta$

NB: δ is often $\frac{1}{2}$. The algorithm is generally run a few times with new random bits each time, until it terminates at least once: $\delta_k = \frac{1}{2^k}$

Example: Quicksort with random pivot and finite time of execution.

3. Algorithm A computes f with bounded error $\epsilon < \frac{1}{2}$ and complexity T if for all inputs x :

- $\mathbb{P}_r[A(x, r) \neq f(x)] \leq \epsilon$
- for all random choices r , $\text{complexity}(A(x, r)) \leq T$

Theorem 1.2. Types 1 and 2 are equivalent.

Proof: The basic idea for converting an algorithm to type 2 is to stop the algorithm when T becomes too large.

- Let A be an algorithm of type 1. Let $c > 1$ be some constant.

Define $B(x, r)$: Run $A(x, r)$ and stop it after running time T . If A has terminated, output the output of A . Otherwise, abort.

If $B(x, r)$ does not abort, then its output is correct. Running time of $B \leq cT$. Let $\tau(r)$ be the running time of $A(x, r)$. $\mathbb{P}_r(B(x, r) \text{ aborts}) = \mathbb{P}_r(\tau(r) \geq cT) \leq \frac{1}{c}$ because of Markov property, since $\tau(r) \geq 0$ and $\mathbb{E}_r[\tau(r)] \leq T$. We constructed an equivalent algorithm of type 2.

- Let A be an algorithm of type 2.

Define $B(x)$: Run $A(x, r)$ with fresh random bits r until A does not abort. Output the output of A .

B is always correct. A run of A aborts with probability $\delta < 1$.

$$\mathbb{P}(A(x, r) \text{ aborts } k \text{ times}) \leq \delta^k$$

$$\mathbb{E}[\text{running time of } B] \leq \sum_{k=0}^{\infty} (\delta^k T) = \frac{T}{1 - \delta} = 2T \text{ for } \delta = \frac{1}{2}$$

We constructed an equivalent algorithm of type 1. □

Definition 1.3. A randomized algorithm applied to a decision problem can have several types of error:

1. Algorithm A has a **one-sided error** ϵ if

- if the appropriate output for x is *ACCEPT*, then $\mathbb{P}_r[A(x, r) \text{ accepts}] = 1$
- if the appropriate output for x is *REJECT*, then $\mathbb{P}_r[A(x, r) \text{ accepts}] \leq \epsilon$

NB: In this case the algorithm is run a few times and x is accepted if it has been accepted by every execution. For $\epsilon = \frac{1}{2}$, $\epsilon_k = \frac{1}{2^k}$

2. Algorithm A has a **two-sided error** ϵ if

- if the appropriate output for x is *ACCEPT*, then $\mathbb{P}_r[A(x, r) \text{ accepts}] \geq 1 - \epsilon$
- if the appropriate output for x is *REJECT*, then $\mathbb{P}_r[A(x, r) \text{ accepts}] \leq \epsilon$

NB: In this case the algorithm is run a few times and x is accepted if it has been accepted by most executions. Generally, $\epsilon = \frac{1}{3}$.

1.1.4 Complexity classes

Our interest lies in two complexity classes:

1. **ZPP complexity class**, with *zero-error* algorithms:
algorithms of type 1 or 2 with T polynomial on input size and $\delta = \frac{1}{2}$ for type 2
2. **BPP complexity class**, with *bounded-error* algorithms:
algorithms of type 3 with T polynomial on input size and $\epsilon = \frac{1}{3}$

1.2 Applications

1.2.1 Matrix multiplication

Decision problem:

- input: A, B and C , $n \times n$ matrices over an arbitrary ring
- output: decide if $A \times B = C$

Freivald's test:

- Choose $r \in \{0, 1\}^n$
- Evaluate $u = Cr$, $v = Br$ and $w = Av$
- Return *ACCEPT* if $u = w$, else *REJECT*

Theorem 1.4. *Freivald's algorithm has a one-sided error:*

- If $AB = C$, $\mathbb{P}[\text{algorithm accepts}] = 1$
- If $AB \neq C$, $\mathbb{P}[\text{algorithm accepts}] \leq \frac{1}{2}$

Since its running time is at most $3n^2$, it belongs to BPP complexity class.

Proof: Assume there are two indices i and j such that $(AB)_{ij} \neq C_{ij}$. Let $D = C - AB$. Then $D_{ij} \neq 0$, $D \neq 0$. We want to prove $\mathbb{P}_{r \in \{0,1\}^n} [Dr = 0] \leq \frac{1}{2}$.

$$(Dr)_i = \sum_k D_{ik}r_k = D_{ij}r_j + f((r_k)_{k \neq j})$$

$$\mathbb{P}[Dr = 0] \leq \mathbb{P}[(Dr)_i = 0]$$

Fix r_1, \dots, r_n excepts r_j . Then $v = f((r_k)_{k \neq j})$.

- If $v = -D_{ij}$: if $r_j = 0$ then $(Dr)_i \neq 0$, if $r_j = 1$ then $(Dr)_i = D_{ij} - D_{ij} = 0$.
Conditional probability of $(Dr)_i = 0$ is $\frac{1}{2}$.

- If $v = 0$: if $r_j = 0$ then $(Dr)_i = 0$, if $r_j = 1$ then $(Dr)_i = D_{ij} \neq 0$. Conditional probability of $(Dr)_i = 0$ is $\frac{1}{2}$.
- Otherwise: for $r_j = 0, 1$ $(Dr)_i \neq 0$.

$$\mathbb{P}[(Dr)_i = 0] \leq \frac{1}{2}$$

□

1.2.2 Finding prime numbers

Primality testing

Decision problem:

- input: an integer $N \geq 2$
- output: decide if N is prime

The sieve of Eratosthenes gives a result in \sqrt{N} steps which is too long.

Theorem 1.5. *Fermat's little theorem:* p prime number $\Rightarrow \forall a \in [1, p-1], a^{p-1} = 1 [p]$

Two random primality tests are based on the above theorem:

- Miller-Rabin test: $O((\log N)^2)$ running time
- Solovay-Strassen test: $O((\log N)^2)$ running time

Tentative algorithm

Lemma 1.6. *Assume there is $1 \leq a < N$ such that $a \wedge N = 1$ and $a^{N-1} \neq 1 [N]$. Then*

$$\mathbb{P}_{1 \leq a < N} [a^{N-1} = 1 [N] | a \wedge N = 1] \leq \frac{1}{2}$$

Primality test algorithm:

- Input: $N \geq 2$
- Select a random $a \in [1, N-1]$
- If $a \wedge N \neq 1$ then reject (in this case N is not prime, because $(a \wedge N) | N$)
- Compute a^{N-1} with rapid exponentiation: $a^{2r} = (a^r)^2$, $a^{2r+1} = a(a^r)^2$
- Accept if $a^{N-1} = 1 [N]$, otherwise reject

Remarks:

- Running time is $O(\log N)$.
- If N is prime then the algorithm accepts N with probability 1.

Corollary 1.7. Assume there is $1 \leq a < N$ such that $a \wedge N = 1$ and $a^{N-1} \neq 1 [N]$. Then $\mathbb{P}_a(\text{algorithm accepts } N) \leq \frac{1}{2}$

Proof: Take N non prime such that there is $1 \leq a < N$ such that $a \wedge N = 1$ and $a^{N-1} \neq 1 [N]$

$$\begin{aligned} \mathbb{P}_a(\text{algorithm accepts } N) &= \mathbb{P}_a(a \wedge N = 1 \text{ and } a^{N-1} = 1 [N]) \\ &= \underbrace{\mathbb{P}_a(a^{N-1} = 1 [N] | a \wedge N = 1)}_{\leq \frac{1}{2}} \times \underbrace{\mathbb{P}_a(a \wedge N = 1)}_{\leq 1} \\ &\leq \frac{1}{2} \end{aligned}$$

□

Definition 1.8. An integer N is a Carmichael number if there is $1 \leq a < N$ such that $a \wedge N = 1$ and $a^{N-1} \neq 1 [N]$ and N is not prime.

The smallest Carmichael number is $561 = 3 \times 11 \times 17$

Miller-Rabin test

Lemma 1.9. If p is prime then the only solution of $x^2 = 1 [p]$ are $\pm 1 \pmod{p}$.

Algorithm:

- Input: $N \geq 2$
- If $N = 2$, ACCEPT. Otherwise if $2|N$, REJECT.
- Take $a \in [2, N - 1]$ uniformly at random.
- If $a \wedge N \neq 1$, REJECT
- Let $N - 1 = 2^t u$ ($t \geq 1$ since N is odd). Compute $b = a^u$. Let $i \leq t$ be the smallest integer such that $b^{2^i} = 1$.
- If i does not exist, REJECT (since $b^{2^t} \neq 1 [N]$, Fermat's test fails)
- If $i = 0$ or $b^{2^{i-1}} = -1$, ACCEPT
- Otherwise, REJECT

Remark: Running time is $O(\log N)$.

Prime finding algorithm

Relational problem:

- input: integer N
- output: prime $p \in [N, 2N]$

Theorem 1.10. Let $\pi(x)$ be the number of prime numbers lower than x . Then $\pi(x) \sim \frac{x}{\ln x}$ while $x \rightarrow +\infty$

Algorithm:

- Take a random $p \in [N, 2N]$
- Check if p is prime
- If p is prime, output p
- If not, start again

Analysis: The number of primes between N and $2N$ is $\pi(2N) - \pi(N-1) \sim \frac{2N}{\ln 2N} - \frac{N}{\ln N} \sim \frac{N}{\ln N}$. Therefore p is prime with probability $\sim \frac{1}{\ln N}$.

Lemma 1.11. Let X_1, X_2, \dots, X_n be a sequence of random variables in $0,1$ such that $\forall i, \mathbb{P}(X_i = 1) \geq p$ and T be the first i such that $X_i = 1$. Then $\mathbb{E}[T] \leq \frac{1}{p}$

The expected number of iterations before finding a prime is $\sim \ln N$. The expected time complexity of the algorithm is $O(\log N) \times (\text{primality test complexity})$.

1.2.3 Polynomial identity testing

Problem

- input: two polynomials $P(X_1, X_2, \dots, X_n), Q(X_1, X_2, \dots, X_n)$ of degree $\leq d$
- output: decide if $P = Q$

Representation of P and Q : P and Q are represented as a black box, such that they can be evaluated efficiently, given a_1, a_2, \dots, a_n . The complexity of an algorithm is the number of evaluations of P and Q .

Remark: Checking if $P = Q$ can be done by expanding them but it will cost an exponential time in their representation size.

Lemma 1.12. *Schwartz-Zippel:* Let F be a field and $S \subset F$. Let $P(X_1, \dots, X_n)$ be a non-zero polynomial of degree $\leq d$. Then $\mathbb{P}_{a_1, \dots, a_n \in S}(P(a_1, \dots, a_n) = 0) \leq \frac{d}{|S|}$

Proof: By induction on n . $n = 1$ is easy since P has at most d roots. □

Algorithm 1

- $S = \{1, 2, 3, \dots, 2d\}$
- Select $a_1, \dots, a_n \in S$ at random
- Accept if $P(a_1, \dots, a_n) = Q(a_1, \dots, a_n)$
- Reject otherwise

Analysis:

- Complexity: 2 evaluations.
- If $P = Q$ then the algorithm accepts with probability 1
- If $P \neq Q$ then $\mathbb{P}(\text{algorithm accepts}) \leq \frac{d}{|S|} \leq \frac{1}{2}$ with Schwartz-Zippel's lemma

Algorithm 2

Assume the greatest coefficient of P and Q is lower than M .

Issue: Find p such that $P = Q \Leftrightarrow P = Q \pmod p$

Then take $p \geq 2M$. In order to adapt the previous algorithm, we also need $p \geq 2d$.

First step: Find a prime between N and $2N$ where N is the maximum of $2d$ and $2M$.

Then it is same algorithm than the first one but we accept if $P(a_1, \dots, a_n) = Q(a_1, \dots, a_n) \pmod p$

1.2.4 Fingerprints**Problem**

- There are 2 players A and B.
- A's input: u , sequence of n bits
- B's input: v , sequence of n bits
- output: decide if $u = v$
- complexity = number of bits exchanged between A and B

A naive solution would be: A sends u to B. But it costs n bits.

Hashing

We define two polynomials P_u and P_v :

$$\begin{aligned}u &= u_0, u_1, \dots, u_{n-1} \\v &= v_0, v_1, \dots, v_{n-1} \\P_u &= u_0 + u_1X + \dots + u_{n-1}X^{n-1} \\P_v &= v_0 + v_1X + \dots + v_{n-1}X^{n-1}\end{aligned}$$

Remark: $u = v \Leftrightarrow P_u = P_v$

Random hash value:

- Take a prime p between n^2 and $2n^2$.
- Select a random a between 0 and $p - 1$.
- $P_u(a) \bmod p$ is the fingerprint of u in $a \bmod p$.

Protocol

Algorithm:

1. A selects p and a as described above
2. A sends $P_u(a) \bmod p$ to B
3. B checks if $P_u(a) = P_v(a) [p]$. If yes, B accepts. Else, B rejects.

Remarks:

- Number of exchanged bits: $6 \log n$
- If $u = v$, B accepts with probability 1. Otherwise, B accepts with probability $\leq \frac{1}{n}$.