

## Lecture 1 — 23th September 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 An introducing example : Freival's Algorithm

### Decision problem:

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

*Remarks:* since 2011 with an improvement from Virginia Williams, an explicit matrix multiplication has an asymptotic complexity of  $O(n^{2.3727})$ .

### Freivald's test:

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

This algorithm uses  $O(n^2)$  additions and multiplications on the coefficients.

**Theorem 1.1.** *Freivald's algorithm has a one-sided error:*

- If  $AB = C$ ,  $\mathbb{P}(\text{algorithm accepts}) = 1$
- If  $AB \neq C$ ,  $\mathbb{P}(\text{algorithm rejects}) \geq \frac{1}{2}$

*Remarks:* If  $AB \neq C$ , since this algorithm has an one-sided error, by running  $k$  independent executions we have  $\mathbb{P}(\text{algorithm accepts after } k \text{ independent executions}) \leq 1/2^k$ . In practice  $k = 100$  is acceptable. For comparison, cosmic rays induce errors on computer with larger probability. In 1996, a studies by IBM revealed that they induced one error per 256 megabytes of RAM per month, which means a probability of  $1.4 \times 10^{-15}$  per byte per second, which is greater than  $2^{-49}$ . Another comparison on large number, is that  $2^{100}$  is far greater than the age of the universe in second, which is less than  $2^{60}$  (for now...).

### Proof:

- If  $AB = C$  then  $u = Cr = (AB)r = A(Br) = Av = w$  thus  $\mathbb{P}(\text{algorithm accepts}) = 1$ .

- If  $AB \neq C$ : Particular case on  $\mathbb{Z}_2$

Let  $F = \{r \in \{0, 1\}^n : (A * B)r = Cr\} \subseteq \{0, 1\}^n$ .  $F \neq \{0, 1\}^n$  and  $F$  is a subspace of vector space  $\{0, 1\}^n$ . Thus using Lagrange's Theorem we have  $|F| \leq \frac{1}{2}|\{0, 1\}^n|$

Hence  $\mathbb{P}(\text{algorithm accepts}) = \mathbb{P}(r \in F) \leq \frac{|F|}{|\{0, 1\}^n|} \leq \frac{1}{2}$

- If  $AB \neq C$ : general case

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 Formal basis

### 1.2.1 Deterministic and randomized algorithms

#### Deterministic algorithm



*Goal:*

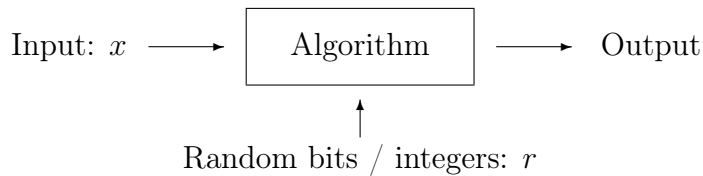
- correctly solve the problem on all inputs
- efficiently (wished): linear or polynomial time on input size

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



*Goal:* find a randomized algo such that on all inputs  $x$  and given a time  $T$ :

- *Monte Carlo algorithms:*
  - $\mathbb{P}[A(x, r) \text{ is correct}] \geq \frac{2}{3}$
  - $\forall r$  execution time of  $A(x, r) \leq T$
- *Las Vegas algorithms:*
  - $\forall r, A(x, r)$  is correct
  - $\mathbb{E}_r(\text{execution time of } A(x, r)) \leq T$

### 1.2.2 Monte-Carlo algorithms

#### One-sided error

**Definition 1.2.** A Monte-Carlo algorithm is said to have an one-sided error if it verify one of the following:

- It is always correct when returning *ACCEPT* (true-biased)
- It is always correct when returning *REJECT* (false-biased)

**Example:** matrix product (true-biased one-sided error algorithm)

- If  $AB = C$  then  $\mathbb{P}(\text{Algorithm on } (A, B, C) \text{ accepts}) = 1$
- Else  $\mathbb{P}(\text{Algorithm on } (A, B, C) \text{ reject}) \geq \frac{1}{2}$

## Success probability amplification

Since  $\mathbb{P}[A(x, r) \text{ is correct}] \geq \frac{2}{3}$ , you can amplify the probability of a correct answer by doing  $k$  independent executions of the algorithm and returning the most answered output.

## 1.3 Reminder on probabilities

### 1.3.1 Definitions

- **Discrete random variable**  $X$  (finite) from  $\Omega$  (finite)

*Example:* random bit  $B$  on  $\Omega = \{0, 1\}$

- **Stochastic process:**  $(X_t)_{t \in T}$  with  $T \in \mathbb{N}$

- **Halting time**  $\tau$  such as  $\tau = t$  depends only from  $X_1, \dots, X_t$

*Example:*  $\tau$ : time to get a 0 from a random bit stream,  $\mathbb{E}(\tau) = 2$

- $\mathbb{P}[\tau = 1] = 1/2$
- $\mathbb{P}[\tau = 2] = 1/4$
- $\mathbb{P}[\tau = k] = 1/2^k$
- $\mathbb{E}(\tau) = \sum_k k \mathbb{P}[\tau = k] = 2$

### 1.3.2 Bernoulli

**Theorem 1.3.** *If  $\mathbb{P}[B_t = 0] = p$ , then  $\mathbb{E}(\tau) = 1/p$*

#### Application

Let  $\Omega = \{1, 2, \dots, n\}$ ,  $X$  a discrete random value from  $\omega$ ,  $X_1, \dots, X_t$  a stochastic process  
Let  $\tau$  be the smallest  $t$  such as  $\{X_1, \dots, X_t\} = \Omega$

Then  $\mathbb{E}(\tau) \approx n \log(n)$

#### Proof

$\tau = \sum_{i=1}^n \tau_i$  with  $\tau_i$  time to get a new value knowing we already have  $i - 1$  different values.  
Then  $\mathbb{P}(\tau_i) = \frac{n-i+1}{n}$  and using the theorem we have  $\mathbb{E}(\tau_i) = \frac{n}{n-i+1}$   
Hence  $\mathbb{E}(\tau) = \sum_{i=1}^n \mathbb{E}(\tau_i) = \sum_{i=1}^n \frac{n}{n-i+1} \approx n \log(n)$

### 1.3.3 Markov inequality

**Theorem 1.4.**  *$X \geq 0$  a discrete random variable,  $\mu = \mathbb{E}(X)$ . Then  $\forall a > 0, \mathbb{P}(X > a\mu) \leq \frac{1}{a}$*

### 1.3.4 Chernoff bound

**Theorem 1.5.**  $X_1, \dots, X_n$  independent random variables from  $\{0, 1\}$  such as  $\forall i, \mathbb{P}[X_i = 1] = \mu_i = \mathbb{E}(X_i)$ . Let  $X = \frac{1}{n} \sum X_i$  and  $\mu = \frac{1}{n} \sum \mu_i = \mathbb{E}(X)$   
Then  $\forall \delta > 0, \mathbb{P}[|X - \mu| \geq \delta\mu] \leq 2^{-\mu\delta^2n/3}$

## 1.4 Application: Primality testing

*Decision problem:*

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

$N$  is  $n = \log_2(N)$  long. The sieve of Eratosthenes gives a result in  $\sqrt{N}$  steps which is too long ( $O(2^{n/2})$  operations).

### 1.4.1 Fermat's little theorem approach

**Fermat's little theorem**

**Theorem 1.6.**  $p \geq 2$  prime number  $\Rightarrow \forall a \in [1, p - 1], a^{p-1} = 1[p]$

**Tentative algorithm**

*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.

**Algorithm's proof**

**Lemma 1.7.** Assume there is  $1 \leq a < N$  such that  $a \wedge N = 1$  and  $a^{N-1} \neq 1 [N]$ . Then  $\mathbb{P}_{1 \leq a < N} [a^{N-1} = 1 [N] | a \wedge N = 1] \leq \frac{1}{2}$

**Proof:** Let  $G = \{b \in \{1, \dots, N-1\} | GCD(b, N) = 1\}$ .  $G$  is an abelian group for the operation  $(X \bmod N)$ . Let  $F = \{b \in G | b^{N-1} = 1 [N]\}$ .

$F \neq G$  and  $F$  is a subgroup hence  $|F| \leq 1/2|G|$  (Lagrange's Theorem)  $\square$

**Corollary 1.8.** 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  $a^{N-1} \neq 1 [N]$

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

$\square$

**Carmichael number**

**Definition 1.9.** An non-prime integer  $N$  is a Carmichael number if all  $1 \leq a < N$  such that  $a \wedge N = 1$  satisfy  $a^{N-1} \equiv 1 [N]$ .

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

There are 255 Carmichael number  $\leq 100000000$

**1.4.2 Miller-Rabin test**

**Lemma 1.10.** 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  $O(\log N)$ .