



# **Chapter 7**

# **Randomization**

**Algorithm Theory**  
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# Randomization

## Randomized Algorithm:

- An algorithm that uses (or can use) **random coin flips** in order to make decisions

**We will see:** **randomization** can be a **powerful tool** to

- Make algorithms **faster**
- Make algorithms **simpler**
- Make the analysis simpler
  - Sometimes it's also the opposite...
- Allow to **solve problems (efficiently)** that cannot be solved (efficiently) without randomization
  - True in some computational models (e.g., for distributed algorithms)
  - Not clear in the standard sequential model

# Contention Resolution

A simple starter example (from distributed computing)

- Allows to introduce important concepts
- ... and to repeat some basic probability theory

## Setting:

- $n$  processes, 1 resource  
(e.g., communication channel, shared database, ...)
- There are time slots 1,2,3, ...
- In each time slot, only one client can access the resource
- All clients need to regularly access the resource
- If client  $i$  tries to access the resource in slot  $t$ :
  - Successful iff no other client tries to access the resource in slot  $t$

## Algorithm Ideas:

- Accessing the resource deterministically seems hard
  - need to make sure that processes access the resource at different times
  - or at least: often only a single process tries to access the resource
- **Randomized solution:**  
In each time slot, each process tries with **probability  $p$** .

## Analysis:

- How large should  $p$  be?
- How long does it take until some process  $i$  succeeds?
- How long does it take until all processes succeed?
- What are the probabilistic guarantees?

## Events:

- $\mathcal{A}_{x,t}$ : process  $x$  **tries to access** the resource in time slot  $t$ 
  - Complementary event:  $\overline{\mathcal{A}_{x,t}}$

$$\mathbb{P}(\mathcal{A}_{x,t}) = p, \quad \mathbb{P}(\overline{\mathcal{A}_{x,t}}) = 1 - p$$

- $\mathcal{S}_{x,t}$ : process  $x$  is **successful** in time slot  $t$

$$\mathcal{S}_{x,t} = \mathcal{A}_{x,t} \cap \left( \bigcap_{y \neq x} \overline{\mathcal{A}_{y,t}} \right)$$

- **Success probability** (for process  $x$ ):

# Fixing $p$

- $\mathbb{P}(\mathcal{S}_{x,t}) = p(1-p)^{n-1}$  is maximized for

$$p = \frac{1}{n} \quad \Rightarrow \quad \mathbb{P}(\mathcal{S}_{x,t}) = \frac{1}{n} \left(1 - \frac{1}{n}\right)^{n-1} .$$

- **Asymptotics:**

$$\text{For } n \geq 2: \quad \frac{1}{4} \leq \left(1 - \frac{1}{n}\right)^n < \frac{1}{e} < \left(1 - \frac{1}{n}\right)^{n-1} \leq \frac{1}{2}$$

- **Success probability:**

$$\frac{1}{en} < \mathbb{P}(\mathcal{S}_{x,t}) \leq \frac{1}{2n}$$

# Time Until First Success

## Random Variable $T_x$ :

- $T_x = t$  if proc.  $x$  is successful in slot  $t$  for the first time

- **Distribution:**

- $T_x$  is **geometrically distributed** with parameter

$$q = \mathbb{P}(\mathcal{S}_{x,t}) = \frac{1}{n} \left(1 - \frac{1}{n}\right)^{n-1} > \frac{1}{en}.$$

- **Expected time** until first success:

$$\mathbb{E}[T_x] = \frac{1}{q} < en$$

# Time Until First Success

**Failure Event  $\mathcal{F}_{x,t}$ :** Process  $x$  does not succeed in time slots  $1, \dots, t$

- The events  $\mathcal{S}_{x,t}$  are independent for different  $t$ :

$$\mathbb{P}(\mathcal{F}_{x,t}) = \mathbb{P}\left(\bigcap_{r=1}^t \overline{\mathcal{S}_{x,r}}\right) = \prod_{r=1}^t \mathbb{P}(\overline{\mathcal{S}_{x,r}}) = \left(1 - \mathbb{P}(\mathcal{S}_{x,r})\right)^t$$

- We know that  $\mathbb{P}(\mathcal{S}_{x,r}) > 1/en$ :

$$\mathbb{P}(\mathcal{F}_{x,t}) < \left(1 - \frac{1}{en}\right)^t < e^{-t/en}$$



# Time Until First Success

No success by time  $t$ :  $\mathbb{P}(\mathcal{F}_{x,t}) < e^{-t/en}$

$t = \lceil en \rceil$ :  $\mathbb{P}(\mathcal{F}_{x,t}) < 1/e$

- Generally if  $t = \Theta(n)$ : **constant success probability**

$t \geq en \cdot c \cdot \ln n$ :  $\mathbb{P}(\mathcal{F}_{x,t}) < 1/e^{c \cdot \ln n} = 1/n^c$

- For **success probability**  $1 - 1/n^c$ , we need  $t = \Theta(n \log n)$ .
- We say that  $x$  succeeds **with high probability** in  $O(n \log n)$  time.

# Time Until All Processes Succeed

**Event  $\mathcal{F}_t$ :** some process has not succeeded by time  $t$

$$\mathcal{F}_t = \bigcup_{x=1}^n \mathcal{F}_{x,t}$$

**Union Bound:** For events  $\mathcal{E}_1, \dots, \mathcal{E}_k$ ,

$$\mathbb{P}\left(\bigcup_x \mathcal{E}_x\right) \leq \sum_x \mathbb{P}(\mathcal{E}_x)$$

Probability that not all processes have succeeded by time  $t$ :

$$\mathbb{P}(\mathcal{F}_t) = \mathbb{P}\left(\bigcup_{x=1}^n \mathcal{F}_{x,t}\right) \leq \sum_{x=1}^n \mathbb{P}(\mathcal{F}_{x,t}) < n \cdot e^{-t/en}.$$

# Time Until All Processes Succeed

**Claim:** With high probability, all processes succeed in the first  $O(n \log n)$  time slots.

Proof:

- $\mathbb{P}(\mathcal{F}_t) < n \cdot e^{-t/en}$
- Set  $t = \lceil en \cdot (c + 1) \ln n \rceil$

Remark:  $\Theta(n \log n)$  time slots are necessary for all processes to succeed with reasonable probability

# Primality Testing

**Problem:** Given a natural number  $n \geq 2$ , is  $n$  a prime number?

## Simple primality test:

1. **if**  $n$  is even **then**
2.     **return** ( $n = 2$ )
3. **for**  $i := 1$  **to**  $\lfloor \sqrt{n}/2 \rfloor$  **do**
4.     **if**  $2i + 1$  divides  $n$  **then**
5.         **return false**
6. **return true**

- **Running time:**  $O(\sqrt{n})$

# A Better Algorithm?

- How can we test primality efficiently?
- We need a little bit of basic number theory...

**Square Roots of Unity:** In  $\mathbb{Z}_p^*$ , where  $p$  is a prime, the only solutions of the equation  $x^2 \equiv 1 \pmod{p}$  are  $x \equiv \pm 1 \pmod{p}$

- If we find an  $x \not\equiv \pm 1 \pmod{n}$  such that  $x^2 \equiv 1 \pmod{n}$ , we can conclude that  $n$  is not a prime.

# Algorithm Idea

**Claim:** Let  $p > 2$  be a prime number such that  $p - 1 = 2^s d$  for an integer  $s \geq 1$  and some odd integer  $d \geq 3$ . Then for all  $a \in \mathbb{Z}_p^*$ ,

$$a^d \equiv 1 \pmod{p} \text{ or } a^{2^r d} \equiv -1 \pmod{p} \text{ for some } 0 \leq r < s.$$

**Proof:**

- **Fermat's Little Theorem:** Given a prime number  $p$ ,

$$\forall a \in \mathbb{Z}_p^*: a^{p-1} \equiv 1 \pmod{p}$$

# Primality Test

**We have:** If  $n$  is an odd prime and  $n - 1 = 2^s d$  for an integer  $s \geq 1$  and an odd integer  $d \geq 3$ . Then for all  $a \in \{1, \dots, n - 1\}$ ,

$$a^d \equiv 1 \pmod{n} \text{ or } a^{2^r d} \equiv -1 \pmod{n} \text{ for some } 0 \leq r < s.$$

**Idea:** If we find an  $a \in \{1, \dots, n - 1\}$  such that

$$a^d \not\equiv 1 \pmod{n} \text{ and } a^{2^r d} \not\equiv -1 \pmod{n} \text{ for all } 0 \leq r < s,$$

we can conclude that  $n$  is not a prime.

- For every odd composite  $n > 2$ , at least  $3/4$  of all possible  $a$  satisfy the above condition
- How can we find such a *witness*  $a$  efficiently?

# Miller-Rabin Primality Test

- Given a natural number  $n \geq 2$ , is  $n$  a prime number?

## Miller-Rabin Test:

1. **if**  $n$  is even **then return** ( $n = 2$ )
2. compute  $s, d$  such that  $n - 1 = 2^s d$ ;
3. choose  $a \in \{2, \dots, n - 2\}$  uniformly at random;
4.  $x := a^d \bmod n$ ;
5. **if**  $x = 1$  **or**  $x = n - 1$  **then return probably prime**;
6. **for**  $r := 1$  **to**  $s - 1$  **do**
7.      $x := x^2 \bmod n$ ;
8.     **if**  $x = n - 1$  **then return probably prime**;
9. **return composite**;



# Analysis

## Theorem:

- If  $n$  is prime, the Miller-Rabin test always returns **true**.
- If  $n$  is composite, the Miller-Rabin test returns **false** with probability at least  $3/4$ .

## Proof:

- If  $n$  is prime, the test works for all values of  $a$
- If  $n$  is composite, we need to pick a good witness  $a$

**Corollary:** If the Miller-Rabin test is repeated  $k$  times, it fails to detect a composite number  $n$  with probability at most  $4^{-k}$ .

# Running Time

## Cost of Modular Arithmetic:

- Representation of a number  $x \in \mathbb{Z}_n$ :  $O(\log n)$  bits
- Cost of adding two numbers  $x + y \bmod n$ :
- Cost of multiplying two numbers  $x \cdot y \bmod n$ :
  - It's like multiplying degree  $O(\log n)$  polynomials  
→ use FFT to compute  $z = x \cdot y$

# Running Time

Cost of exponentiation  $x^d \bmod n$ :

- Can be done using  $O(\log d)$  multiplications
- Base-2 representation of  $d$ :  $d = \sum_{i=0}^{\lfloor \log d \rfloor} d_i 2^i$
- **Fast exponentiation:**
  1.  $y := 1$ ;
  2. **for**  $i := \lfloor \log d \rfloor$  **to** 0 **do**
  3.      $y := y^2 \bmod n$ ;
  4.     **if**  $d_i = 1$  **then**  $y := y \cdot x \bmod n$ ;
  5. **return**  $y$ ;
- **Example:**  $d = 22 = 10110_2$

# Running Time

**Theorem:** One iteration of the Miller-Rabin test can be implemented with running time  $O(\log^2 n \cdot \log \log n \cdot \log \log \log n)$ .

1. **if**  $n$  is even **then return** ( $n = 2$ )
2. compute  $s, d$  such that  $n - 1 = 2^s d$ ;
3. choose  $a \in \{2, \dots, n - 2\}$  uniformly at random;
4.  $x := a^d \bmod n$ ;
5. **if**  $x = 1$  **or**  $x = n - 1$  **then return probably prime**;
6. **for**  $r := 1$  **to**  $s - 1$  **do**
7.      $x := x^2 \bmod n$ ;
8.     **if**  $x = n - 1$  **then return probably prime**;
9. **return composite**;

# Deterministic Primality Test

- If a conjecture called the generalized Riemann hypothesis (GRH) is true, the Miller-Rabin test can be turned into a polynomial-time, deterministic algorithm
  - It is then sufficient to try all  $a \in \{1, \dots, O(\log^2 n)\}$
- It has long not been proven whether a deterministic, polynomial-time algorithm exists
- In 2002, Agrawal, Kayal, and Saxena gave an  $\tilde{O}(\log^{12} n)$ -time deterministic algorithm
  - Has been improved to  $\tilde{O}(\log^6 n)$
- In practice, the randomized Miller-Rabin test is still the fastest algorithm