

# Fingerprinting and primality

RA-MIRI QT Curs 2020-2021

# Fingerprinting technique

Freivalds algorithm is an example of the algorithmic **fingerprinting** technique, we do not want to compute, but just to check.

We want to compare two items,  $A_1$  and  $A_2$ , instead of comparing them directly, we compute **random fingerprints**  $\phi(A_1)$  and  $\phi(A_2)$  and compare these.

We seek a fingerprint function  $\phi()$  with the following properties:

- ▶ If  $A_1 = A_2$  then whp  $\Pr[\phi(A_1) = \phi(A_2)] = 1$ .
- ▶ If  $A_1 \neq A_2$  then  $\Pr[\phi(A_1) = \phi(A_2)] = 0$ .
- ▶ It is a lot more efficient to compute and compare  $\phi(A_1)$  and  $\phi(A_2)$ , than computing and comparing  $A_1$  and  $A_2$ .

Notice that for Freivalds' algorithm, if  $A$  is  $n \times n$  matrix, then  $\phi(A) = Ar$ , for a random  $n$ -dimensional Boolean vector  $r$ .

# Database consistency

## From MR 7.4

Alice and Bob are in different continents. Each has a copy of a huge database with  $N$  bits. Alice maintain its large  $N$ -bit database  $X = \{x_{N-1}, \dots, x_0\}$  of information, while Bob maintains a second copy  $Y = \{y_{N-1}, \dots, y_0\}$  of the same database.

Periodically they want to check consistency of their copies, i.e., to check that both are the same.

Alice could send  $X$  to Bob, and he could compare it to  $Y$ . But this requires transmission of  $N$  bits, which is costly and error-prone.

Instead, suppose Alice first computes a much smaller fingerprint  $\phi(X)$  and sends this to Bob. He then computes  $\phi(Y)$  and compares it with  $\phi(X)$ . If the fingerprints are equal, he announces that the copies are identical.

# Database consistency

## From MR 7.4

Alice and Bob are in different continents. Each has a copy of a huge database with  $N$  bits. Alice maintain its large  $N$ -bit database  $X = \{x_{N-1}, \dots, x_0\}$  of information, while Bob maintains a second copy  $Y = \{y_{N-1}, \dots, y_0\}$  of the same database.

Periodically they want to check consistency of their copies, i.e., to check that both are the same.

Alice could send  $X$  to Bob, and he could compare it to  $Y$ . But this requires transmission of  $N$  bits, which is costly and error-prone.

Instead, suppose Alice first computes a much smaller fingerprint  $\phi(X)$  and sends this to Bob. He then computes  $\phi(Y)$  and compares it with  $\phi(X)$ . If the fingerprints are equal, he announces that the copies are identical.

What kind of fingerprint function should we use here?

# Database consistency

## From MR 7.4

Alice and Bob are in different continents. Each has a copy of a huge database with  $N$  bits. Alice maintain its large  $N$ -bit database  $X = \{x_{N-1}, \dots, x_0\}$  of information, while Bob maintains a second copy  $Y = \{y_{N-1}, \dots, y_0\}$  of the same database.

Periodically they want to check consistency of their copies, i.e., to check that both are the same.

Alice could send  $X$  to Bob, and he could compare it to  $Y$ . But this requires transmission of  $N$  bits, which is costly and error-prone.

Instead, suppose Alice first computes a much smaller fingerprint  $\phi(X)$  and sends this to Bob. He then computes  $\phi(Y)$  and compares it with  $\phi(X)$ . If the fingerprints are equal, he announces that the copies are identical.

What kind of fingerprint function should we use here?

How many bits do we need to send?

# Database consistency

## From MR 7.4

Alice and Bob are in different continents. Each has a copy of a huge database with  $N$  bits. Alice maintain its large  $N$ -bit database  $X = \{x_{N-1}, \dots, x_0\}$  of information, while Bob maintains a second copy  $Y = \{y_{N-1}, \dots, y_0\}$  of the same database.

Periodically they want to check consistency of their copies, i.e., to check that both are the same.

Alice could send  $X$  to Bob, and he could compare it to  $Y$ . But this requires transmission of  $N$  bits, which is costly and error-prone.

Instead, suppose Alice first computes a much smaller fingerprint  $\phi(X)$  and sends this to Bob. He then computes  $\phi(Y)$  and compares it with  $\phi(X)$ . If the fingerprints are equal, he announces that the copies are identical.

What kind of fingerprint function should we use here?

How many bits do we need to send?

Which is the error in the fingerprint test?

# Review of Algebra 1

Given  $a, b, n \in \mathbb{Z}$ ,  $a$  **congruent** with  $b$  modulo  $n$  ( $a \equiv b \pmod{n}$ ) if  $n \mid (a - b)$ .

1.  $a \pmod{n} = b \Rightarrow a \equiv b \pmod{n}$ .
2.  $(a + b) \pmod{n} \equiv ((a \pmod{n}) + (b \pmod{n})) \pmod{n}$ .
3.  $(a \cdot b) \pmod{n} \equiv ((a \pmod{n}) \cdot (b \pmod{n})) \pmod{n}$ .
4.  $a + (b + c) \equiv (a + b) + c \pmod{n}$  (associativity)
5.  $ab \equiv ba \pmod{n}$  (commutativity)
6.  $a(b + c) \equiv ab + ac \pmod{n}$  (distributivity)

$n$  partitions  $\mathbb{Z}$  in  $n$  equivalence classes:  $\mathbb{Z}_n = \{0, 1, \dots, n - 1\}$ .

For any  $m \in \mathbb{Z}$ ,  $m \pmod{n} \in \mathbb{Z}_n$ .

Define  $\mathbb{Z}_n^+ = \{1, \dots, n - 1\}$ .  $(\mathbb{Z}_n, +_n, \cdot_n)$  form a commutative ring,

# Review of Algebra 2

## Theorem (Prime number Theorem)

Let  $n \in \mathbb{Z}$  and let  $\pi(n)$  be the number of primes  $\leq n$ , then

$$\pi(n) \sim \frac{n}{\ln n}, \text{ as } n \rightarrow \infty.$$

The frequency of primes slowly decay as the integers increase in length.

For ex. if  $n = 10^4$ ,  $\pi(n) = 1929$  and  $\frac{n}{\ln n} = 1086$ ,  
while, if  $n = 10^7$ ,  $\pi(n) = 664579$  and  $\frac{n}{\ln n} = 620420$ .



## Review of Algebra 3

Lemma: If  $n \in \mathbb{Z}$  has  $N$ -bits, then  $n \leq 2^N$ , and at most  $N$  different primes can divide  $n$ .

As prime numbers are  $\geq 2$ , the # of distinct primes that divide  $n$  is  $\leq N$ , because if we multiply together more than  $N$  numbers that are at least 2, then we get a number greater than  $2^N$

For ex. if  $n = 33$ , ( $33_2 = 100001$ ), so  $N = 6$  and  $2^6 = 64$ . Besides,  $\pi(33) = 11$  of which only 2 of them divide 33 ( $2 < 6$ )

Corollary: Let  $p_i$  be the  $i$ -th. prime number, then the value of  $p_i \sim i \ln i$

For ex. if  $i = 1000$ , then  $p_i \sim 1000 \ln(1000) = 6907$  and the exact value is  $p_{1000} = 7919$

# Solution to the database consistency problem

If Alice (A) has  $X$  and Bob (B) has  $Y$ , they use the following algorithm to check they are the same:

- ▶ See the data as  $N$ -bit integers:  $\mathbf{x} = \sum_{i=0}^{N-1} x_i 2^i$  and  $\mathbf{y} = \sum_{i=0}^{N-1} y_i 2^i$ .
- ▶ A chooses u.a.r. a prime  $p \in [2, 3, 5, \dots, m]$ , for suitable  $m = cN \ln N$ . (The number of primes in  $2^N$  is  $N$ )
- ▶ A computes  $\phi(\mathbf{x}) = \mathbf{x} \bmod p$  and sends the result together with the value  $p$  to B.
- ▶ B computes  $\phi(\mathbf{y}) = \mathbf{y} \bmod p$  and compares with the quantity he got from A.
- ▶ If  $\phi(\mathbf{x}) \neq \phi(\mathbf{y})$  for sure  $X \neq Y$ , but it is possible  $\phi(\mathbf{x}) = \phi(\mathbf{y})$  and  $X \neq Y$ . (This happens if  $\mathbf{x} \bmod p = \mathbf{y} \bmod p$ , with  $\mathbf{x} \neq \mathbf{y}$ ).

# Bounding the probability of error

By the Prime Number Theorem  $\pi(m) \sim \frac{m}{\ln m}$ , so as we see below, we need to take  $m = cN \ln N$ , for constant  $c > 1$ .

We want to bound the probability that  $\mathbf{x} \neq \mathbf{y}$  but  $\phi(\mathbf{x}) = \phi(\mathbf{y})$ , i.e.,

$$\begin{aligned}\Pr[\mathbf{x} \bmod p = \mathbf{y} \bmod p | \mathbf{x} \neq \mathbf{y}] &= \Pr[p \text{ divides } |\mathbf{x} - \mathbf{y}|] \\ &= \frac{\# \text{ of primes dividing } |\mathbf{x} - \mathbf{y}|}{\# \text{ primes } \leq m} \\ &\leq \frac{N}{m / \ln m} = \frac{N \ln m}{cN \ln N} = \frac{\ln m}{c \ln N} \\ &= \frac{\ln(cN \ln N)}{c \ln N} = \frac{\ln N + \ln(c \ln N)}{c \ln N} \\ &= \frac{1}{c} + \frac{\ln(c \ln N)}{c \ln N} = \frac{1}{c} + o(1)\end{aligned}$$

**Lemma:** Taking  $c = 1/\epsilon$  for a chosen  $0 < \epsilon < 1$ , the algorithm achieves an error probability of  $\leq \epsilon$ .

Choosing a large  $m \Rightarrow$ , i.e. a large  $c$ , we have a larger selection for  $p$ , so it is less likely that  $p$  divides  $|\mathbf{x} - \mathbf{y}|$ .

# Communication bits

**Lemma:** The fingerprint algorithm to check the consistency of two databases with  $N$  bits uses  $O(\lg N)$  bits of communication.

**Proof:** A sends to B  $p$  and  $\mathbf{x} \bmod p$ , both are  $\leq m$ .

Since  $m = cN \ln N$ , then  $m$  requires

$\lg(cN \ln N) = \lg N + \lg(c \ln N) \sim O(\lg N)$  bits, so the number of transmitted bits is  $O(\lg n)$ . □

We proved that by using a more efficient representation of the data (modular), the randomized fingerprinting algorithm gives an exponential decrease in the amount of communication at a small cost in correctness.

# How to pick a random prime number

Problem: Given an integer  $N$  we want to pick a random prime  $p \in [2, \dots, 2^N - 1]$ .

Recall: if  $n$  has  $N$  bits  $\Rightarrow n \leq 2^N - 1$  and  $N \geq \lg n$ .

Assume we have an efficient algorithm **Prime?** which tell us if an integer is a prime, or not.

Define the set  $P = \{p \mid 1 < p \leq 2^N - 1, \text{ and } p \text{ is prime}\}$ .

We want to pick u.a.r.  $p \in P$  (i.e., with probability  $\frac{1}{|P|}$ )

```
Pickprime( $p$ )  
for  $i = 0$  to  $t$  do  
   $p = \text{Rand}(2^N - 1)$   
  if Prime?( $p$ ) =  $T$  then  
    return  $p$ 
```

$t$  will be fixed later

First analyze one iteration of the algorithm

After we analyze the probability of error after amplifying  $t$  times.

## Analysis of the algorithm

Let  $A$  be the event that a random generated  $N$ -bit integer is a prime in  $P$ :

$$\Pr[A] = \frac{|P|}{2^N} = \frac{(2^N / \ln 2^N)}{2^N} = \frac{1}{N \ln 2} = \frac{1.442}{N}.$$

If  $N = 2000$  then  $\Pr[A] = 0.000721$ , therefore the probability of failing is  $\Pr[\bar{A}] = 0.999271$ . Quite high !

Taking into consideration the  $t$ -amplification,

$$\Pr[\text{Failure after } t \text{ repetitions}] = \left(1 - \frac{1.442}{N}\right)^t \leq e^{-\frac{1.442t}{N}},$$

so taking  $t = 10N$  suffices to make small the probability of failure.

## Analysis of the algorithm: Numerical example

If  $N = 2000$  taking  $t = 10N = 20000$  yields

$\Pr[\text{Failure}] = 0.00004539$  and  $\Pr[\text{Success}] = 0.999955$ . If

$t = N = 2000$ ,  $\Pr[\text{Success}] = 0.76425$ .

In practice, most of the algorithms to generate a large prime, follows the previous scheme (see for ex.

<https://asecuritysite.com/encryption/random3>)

# The Primality problem

From Cormen et al., 31.8 (3rd edition)

INPUT:  $n \in \mathbb{N}$ . QUESTION: Is  $n$  prime?

Naive algorithm:

```
Is  $n \in \mathbb{N}$  prime?  
for  $a = 2, 3, \dots, \sqrt{n}$  do  
    if  $a | n$  then  
        return composite  
return prime
```

Recall that in arithmetic complexity, for large  $n$  ( $n = 2^{2024}$ ), the input size is the number of bits  $N$  to express  $n$  i.e.,  $n = 2^N$  and  $N = \lg n$

Complexity of the algorithm:  $T(N) = O(2^{N/2} N^2)$  Too slow!



# Randomized algorithms for Primality Testing

Theorem (Fermat's Little Th., XVII)

*If  $n$  is prime, then for all  $a \in \mathbb{Z}_n^+$ ,  $a^{n-1} \equiv 1 \pmod{n}$ .*

Fermat only works in one direction:

**BUT**  $\exists n \in \mathbb{Z}$  s.t. for all  $a$ ,  $a^{n-1} \equiv 1 \pmod{n}$  with  $n$  **NOT** prime.

The **Carmichael numbers**:  $n \in \mathbb{Z}$  is a **Carmichael number** if, for each  $a \in \mathbb{Z}_n^*$ ,  $a^{n-1} \equiv 1 \pmod{n}$  and  $n$  is not prime.

Carmichael numbers **are very rare** (255 with value  $< 100000000$ )

561, 1105, 1729,  $\dots$

For example  $561 = 3 \times 11 \times 17$

# Test of pseudo-primality

(Assuming the non-existence of Carmichael numbers)

For any  $n \in \mathbb{Z}$ ,  $n$  is a **pseudo-prime** if  $n$  is composite and  $\forall a \in \mathbb{Z}_n^+$ ,  
 $a^{n-1} \equiv 1 \pmod n$ .

```
Is  $n \in \mathbb{N}$  prime?  
 $a = \mathbf{rand} (1, n - 1)$   
if  $a^{n-1} \equiv 1 \pmod n$  then  
    return pseudo-prime  
else  
    return composite
```

Complexity:  $O(N^3)$ .

## Test of pseudo-primality: Error probability

If the algorithm says *composite*  $n$  is composite

If the algorithm says *pseudo-prime* if  $n$  is prime, the answer is correct, but if  $n$  is composite it errs. This happens with probability  $\leq 1/2$ .

The previous algorithm has one-side error, therefore amplifying  $t$  times the algorithm, the probability of error goes down to  $\leq 1/2^t$ .

**Repeated-Fermat**  $n, t$

```
for  $i = 1$  to  $t$  do
   $a = \text{rand}(1, n - 1)$ 
  if  $a^{n-1} \not\equiv 1 \pmod n$  then
    return non-prime
  else
    return prime
```

# Taking into consideration the Carmichel numbers

Sketch of a Monte-Carlo algorithm for deciding if a given  $n$  is a prime: G. Miller (1976), M. Rabin (1980)

- ▶ If equation  $x^2 \equiv 1 \pmod{n}$  has exactly solutions  $x = \pm 1$  that implies  $n$  is prime.
- ▶ If there is another solution different than  $\pm 1$ , then  $n$  can not be prime.
- ▶ To see if  $n$  is prime: Randomly choose an integer  $a < n$ , if  $a^2 \equiv 1 \pmod{n}$ , then  $a$  is a non-trivial root of 1 mod  $n$ , so  $n$  is not prime. Such an  $a$  is denoted a **witness** to the compositeness of  $n$ . Otherwise,  $n$  may be a prime.

The error of the resulting Monte-Carlo algorithm is  $1/2^t$  and the complexity is  $O(tN^3)$ .

## Deciding primality

- ▶ For a long time it was open to prove that  $\text{primality} \in \text{P}$ . In 2002, Agrawal, Kayal, Saxena, (AKS) gave a deterministic polynomial time algorithm for Primality.
- ▶ If  $n \leq 2^N$  the best implementation for the AKS is  $\tilde{O}(N^6) = O(N^6 \lg N)$ .
- ▶ AKS has terrible running time, and it is not clear that it can be improved in the near future.
- ▶ From the computational point the Miller-Rabin's algorithm is the basis for existing efficient algorithms.
- ▶ However, the Fermat pseudo-primality test can also work fairly nicely, (if we are dealing with  $N = 9$ , the probability of hitting a Carmichel number is 0.000000255).