

# Cryptography

## Fundamentals

Prof. Dr. Heiko Knospe

November 4, 2022

# Mathematical Fundamentals

Modern cryptography relies on mathematical structures and methods.

We briefly discuss a number of fundamental topics from discrete mathematics, elementary number theory, computational complexity and probability theory.

Algebraic structures are covered in a separate chapter.

# Sets

Sets are the most elementary mathematical structure. Finite sets play an important role in cryptography.

*Example:*  $M = \{0, 1\}^{128}$  is the set of binary strings of length 128. Elements in  $M$  can be written in the form  $b_1 b_2 \dots b_{128}$  or

$$(b_1, b_2, \dots, b_{128})$$

in vectorial notation. An element of  $M$  could, for example, represent one block of plaintext or ciphertext data. The cardinality of  $M$  is very large:

$$|M| = 2^{128} \approx 3.4 \cdot 10^{38}$$

# Small and Large Numbers

It is important to help understand the difference between small, big and inaccessible numbers in practical computations. For example, one can easily store one terabyte ( $10^{12}$  bytes, i.e., around  $2^{43}$  bits) of data. On the other hand, a large amount of resources are required to store one exabyte (one million terabytes) or  $2^{63}$  bits and more than  $2^{100}$  bits are out of reach.

The number of computing steps is also bounded: less than  $2^{40}$  steps (say CPU clocks) are easily possible,  $2^{60}$  operations require a lot of computing resources and take a significant amount of time, and more than  $2^{100}$  operations are unfeasible. It is for example impossible to test  $2^{128}$  different keys with conventional (non-quantum) computers.

# Functions

## Definition

A *function* or a *map*

$$f : X \rightarrow Y$$

consists of two sets (the *domain*  $X$  and the *codomain*  $Y$ ) and a rule which assigns an output element (an *image*)  $y = f(x) \in Y$  to each input element  $x \in X$ . The set of all  $f(x)$  is a subset of  $Y$  called the *range* or *image*  $\text{im}(f)$ . Any  $x \in X$  with  $f(x) = y$  is called a *preimage* of  $y$ . Let  $B \subset Y$ , then we say that

$$f^{-1}(B) = \{x \in X \mid f(x) \in B\}$$

is the *preimage* or *inverse image* of  $B$  under  $f$ .

# Injective, Surjective and Bijective Maps

## Definition

Let  $f : X \rightarrow Y$  be a function.

- $f$  is *injective* if different elements of the domain map to different elements of the range: for all  $x_1, x_2 \in X$  with  $x_1 \neq x_2$ , we have  $f(x_1) \neq f(x_2)$ . Equivalently,  $f$  is injective if for all  $x_1, x_2 \in X$ :

$$f(x_1) = f(x_2) \Rightarrow x_1 = x_2$$

- $f$  is *surjective* or *onto* if every element of the codomain  $Y$  is contained in the image of  $f$ , i.e., for every  $y \in Y$  there exists an  $x \in X$  with  $f(x) = y$ . In other words,  $f$  is surjective if  $\text{im}(f) = Y$ .
- $f$  is *bijective* if it is both injective and surjective. Bijective functions are invertible and possess an inverse map  $f^{-1} : Y \rightarrow X$  such that  $f^{-1} \circ f = \text{id}_X$  and  $f \circ f^{-1} = \text{id}_Y$ .

# Relations

## Definition

A relation  $R$  on  $X$  is a subset of  $X \times X$ .  $R$  is called an *equivalence relation* if it satisfies the following conditions:

- 1  $R$  is reflexive, i.e.,  $(x, x) \in R$  for all  $x \in X$ , and
- 2  $R$  is symmetric, i.e., if  $(x, y) \in R$  then  $(y, x) \in R$ , and
- 3  $R$  is transitive, i.e., if  $(x, y) \in R$  and  $(y, z) \in R$  then  $(x, z) \in R$ .

If  $(x, y) \in R$ , then  $x$  and  $y$  are called *equivalent* and we write  $x \sim y$ . For  $x \in X$ , the subset  $\bar{x} = \{y \in X \mid x \sim y\} \subset X$  is called the *equivalence class* of  $x$ . The set of equivalence classes of  $X$  gives the *quotient set*

$$X / \sim .$$

# Residue Classes modulo $n$

Let  $n \in \mathbb{N}$ ,  $n \geq 2$ . Define the following equivalence relation  $R_n$  on  $\mathbb{Z}$ :

$$R_n = \{(x, y) \in \mathbb{Z} \times \mathbb{Z} \mid x - y \in n\mathbb{Z}\}$$

Note:  $(x, y) \in R_n$  if the difference  $x - y$  is divisible by  $n$ .

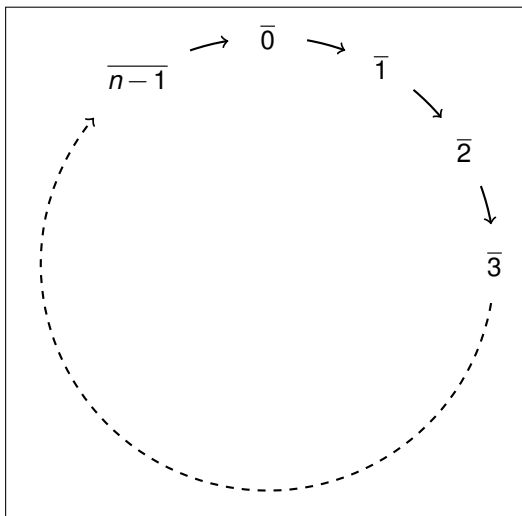
The equivalence class of  $x \in \mathbb{Z}$  is the set

$$\bar{x} = \{\dots, x - 2n, x - n, x, x + n, x + 2n, \dots\}.$$

Now we have  $n$  *different* equivalence classes and the quotient set  $\mathbb{Z}/\sim$  has  $n$  elements. We call this set the *residue classes modulo  $n$*  or *integers modulo  $n$*  and denote it by  $\mathbb{Z}_n$  or  $\mathbb{Z}/(n)$ . Each residue class has a *standard representative* in the set  $\{0, 1, \dots, n-1\}$  and elements in the same residue class are called *congruent modulo  $n$* .



# Residue Classes modulo $n$



## Example: $\mathbb{Z}_2$

$\mathbb{Z}_2 = \{\bar{0}, \bar{1}\}$  has only two elements. One has  $\overline{-1} = \bar{1} = \bar{3}$  and  $\overline{-2} = \bar{0} = \bar{2}$ . The difference of two elements which are in the same class is divisible by 2 (i.e., their difference is even).

The standard representatives are 0, 1 and we have

$$\bar{0} = \{\dots, -4, -2, 0, 2, 4, \dots\},$$

$$\bar{1} = \{\dots, -3, -1, 1, 3, 5, \dots\}.$$

We may simply write 0 and 1 for these two classes.

# XOR, AND, OR

Elements of  $\mathbb{Z}_2$  can be added modulo 2, and addition is the same as the XOR operation on bits. The multiplication modulo 2 corresponds to the AND operation.

$\oplus$	0	1
0	0	1
1	1	0

$\cdot$	0	1
0	0	0
1	0	1

**Table:** XOR and AND operations.

Furthermore, there is an OR operation on bits and  $x \text{ OR } y = x \oplus y \oplus x \cdot y$ .

OR	0	1
0	0	1
1	1	1

**Table:** OR operation

# Example: $\mathbb{Z}_{26}$

$\mathbb{Z}_{26} = \{\overline{0}, \overline{1}, \dots, \overline{25}\}$  has 26 elements. For example, one has  $\overline{-14} = \overline{38}$ , since  $-14 - 38 = -52$  is a multiple of 26. The integers  $-14$  and  $-38$  are congruent modulo 26 and we write

$$-14 \equiv 38 \pmod{26}.$$

The standard representative of this residue class is 12 and

$$\overline{12} = \{\dots, -14, 12, 38, 64, \dots\}.$$

# Computations with Residue Classes

Residue classes can be added, subtracted and multiplied. An arbitrary integer representative can be used, and it is reasonable to choose a small representative.

*Examples:* a)  $79 - 180 \pmod{26} \equiv 1 - 24 \equiv 1 + 2 = 3 \pmod{26}$ .

b)  $234577 \cdot 2328374 \cdot 2837289374 \pmod{3} \equiv 1 \cdot 2 \cdot 2 \equiv 1 \pmod{3}$ .

However, division is more tricky since rational numbers  $\frac{b}{a}$  are not representatives of residue classes. We say that  $a$  is invertible modulo  $n$  if there exists  $x \in \mathbb{Z}$  such that

$$ax \equiv 1 \pmod{n}.$$

Then  $x \equiv (a \pmod{n})^{-1}$ .

*Example:*  $(3 \pmod{10})^{-1} \equiv 7$ , since  $3 \cdot 7 \equiv 1 \pmod{10}$ .

# Invertible Residue Classes

## Proposition

*An integer  $a$  is invertible modulo  $n$  if and only if  $\gcd(a, n) = 1$ , i.e., if the greatest common divisor of  $a$  and  $n$  is 1.*

*Example:* 3 is invertible modulo 10, but 2 is not invertible modulo 10.

## Definition

The invertible integers modulo  $n$  are called units mod  $n$ . The subset of units of  $\mathbb{Z}_n$  is denoted by  $\mathbb{Z}_n^*$ .

*Example:*  $\mathbb{Z}_{10}^* = \{\overline{1}, \overline{3}, \overline{7}, \overline{9}\}$ .

# Prime Numbers

## Definition

An integer  $p \geq 2$  is called a prime number if  $p$  is only divisible by  $\pm 1$  and  $\pm p$ .

If  $p$  is prime, then

$$\mathbb{Z}_p^* = \{\overline{1}, \dots, \overline{p-1}\}.$$

Prime numbers play an important role in public-key cryptography.

The *Prime Number Theorem* states that the density of primes in the first  $N$  integers is approximately

$$\frac{1}{\ln(N)}.$$

# Extended Euclidean Algorithm

One of the key algorithms in elementary number theory is the *Extended Euclidean Algorithm*. The algorithm takes two nonzero integers  $a, b$  as input and computes  $\gcd(a, b)$  as well as two integers  $x, y \in \mathbb{Z}$  such that

$$\gcd(a, b) = ax + by.$$

The Extended Euclidean Algorithm is very efficient and can be used to compute the multiplicative inverse of  $a \bmod n$ . If  $\gcd(a, n) = 1$  then the algorithm outputs  $x, y \in \mathbb{Z}$  such that

$$1 = ax + ny.$$

Then

$$1 \equiv ax \pmod{n}$$

and thus  $x \equiv (a \bmod n)^{-1}$ .



# Extended Euclidean Algorithm

---

**Input:**  $a, b \in \mathbb{N}$

**Output:**  $\gcd(a, b)$ ,  $x, y \in \mathbb{Z}$  such that  $\gcd(a, b) = ax + by$

**Initialisation:**  $x_0 = 1, x_1 = 0, y_0 = 0, y_1 = 1, sign = 1$

```
1: while  $b \neq 0$  do
2:    $r = a \bmod b$  // remainder of the integer division  $a : b$ 
3:    $q = a/b$  // integer quotient
4:    $a = b$ 
5:    $b = r$ 
6:    $xx = x_1$ 
7:    $yy = y_1$ 
8:    $x_1 = q \cdot x_1 + x_0$ 
9:    $y_1 = q \cdot y_1 + y_0$ 
10:   $x_0 = xx$ 
11:   $y_0 = yy$ 
12:   $sign = -sign$ 
13: end while
14:  $x = sign \cdot x_0$ 
15:  $y = -sign \cdot y_0$ 
16:  $\gcd = a$ 
17: return  $\gcd, x, y$ 
```

# Modular Exponentiation I

Modular exponentiation with a large basis, exponent and modulus plays an important role in cryptography. How can we efficiently compute

$$x^a \mod n ?$$

If  $a = 2^k$  then  $k$ -fold squaring modulo  $n$  gives the result:

$$x^a \mod n = (((x^2 \mod n)^2 \mod n)^2 \mod n \dots)^2 \mod n$$

For example,  $x^{256} \mod n$  can be computed with only 8 squaring operations. After each squaring, reduce mod  $n$  in order to reduce the size of the result.

# Modular Exponentiation II

If the exponent is not a power of 2, then it can still be written as a *sum of powers* of 2. This gives a product of factors of type  $x^{(2^k)} \bmod n$ , and each factor can be computed by  $k$  modular squarings. We call this the *Fast Exponentiation Algorithm*.

*Example:* Compute  $6^{41} \bmod 59$ . We have  $41 = 2^5 + 2^3 + 2^0$  and first compute the following sequence of squares:

$$6^2 \equiv 36 \bmod 59$$

$$6^4 \equiv 36^2 \equiv 57 \bmod 59$$

$$6^8 \equiv 57^2 \equiv 4 \bmod 59$$

$$6^{16} \equiv 4^2 \equiv 16 \bmod 59$$

$$6^{32} \equiv 16^2 \equiv 20 \bmod 59$$

$$\text{Then } 6^{41} = 6^{32} \cdot 6^8 \cdot 6 \equiv 20 \cdot 4 \cdot 6 \equiv 8 \bmod 59.$$

# Cardinality

## Proposition

*Let  $X$  and  $Y$  be finite sets of cardinality  $|X|$  and  $|Y|$ , respectively. Then:*

- 1**  $|X \times Y| = |X| \cdot |Y|$  and  $|X^k| = |X|^k$  for  $k \in \mathbb{N}$ .
- 2** Suppose  $|X| = n$  and  $k \leq n$ . Then the number of subsets of  $X$  of cardinality  $k$  is given by the binomial coefficient  $\binom{n}{k} = \frac{n!}{k!(n-k)!}$ .

*Example:* There are  $\binom{128}{2} = \frac{128 \cdot 127}{2} = 8128$  different binary words of length 128 with exactly two ones and 126 zeros.

# Euler's $\phi$ -Function

## Definition

Let  $n \in \mathbb{N}$ . Then Euler's  $\phi$ -function is defined by the cardinality of the units mod  $n$ , i.e.,

$$\phi(n) = |\mathbb{Z}_n^*|$$

*Examples:* a)  $\phi(10) = 4$ .

b) If  $p$  is a prime number, then  $\phi(p) = p - 1$ .

c) If  $p$  and  $q$  are different prime numbers, then  $\phi(pq) = (p - 1)(q - 1)$  (Why?).

# Permutations

## Definition

Let  $S$  be a finite set. A *permutation* of  $S$  is a bijective map  $\sigma : S \rightarrow S$ .

## Proposition

Let  $S$  be a finite set and  $|S| = n$ . Then there are  $n!$  permutations of  $S$ .

Note: the factorial increases very fast, for example

$$50! \approx 3.04 \cdot 10^{64}.$$

# Permutations in Cryptography

Cryptographic operations often use permutations. A randomly chosen family of permutations of a set like  $M = \{0, 1\}^{128}$  would constitute an ideal block cipher. However, it is impossible to write down or store a general permutation since  $M$  has  $2^{128}$  elements. Much simpler (and much less secure) are *bit permutations*, which permute only the *position* of the bits.

*Example:*  $(5\ 7\ 1\ 2\ 8\ 6\ 3\ 4)$  defines a permutation on  $X = \{0, 1\}^8$ : a byte  $(b_1, b_2, \dots, b_8)$  is mapped to  $(b_5, b_7, b_1, b_2, b_8, b_6, b_3, b_4)$ . There are  $8!$  bit permutations of  $X$  (a small number), but  $(2^8)!$  general permutations (a very large number).

# Big-O Notation

We often need to analyze the computational complexity of algorithms, i.e., the resources (running time and space) as a function of the input size.

## Definition

Let  $f, g : \mathbb{N} \rightarrow \mathbb{R}$  be two functions on  $\mathbb{N}$ . Then we say that  $g$  is an *asymptotic upper bound* for  $f$ , if there exists a real number  $C \in \mathbb{R}$  and an integer  $n_0 \in \mathbb{N}$  such that

$$|f(n)| \leq C |g(n)| \text{ for all } n \geq n_0.$$

One writes  $f = O(g)$  or  $f \in O(g)$ .



# Asymptotic Complexity: Examples

- 1  $f(n) = 2n^3 + n^2 + 7n + 2$ . Since  $n^2 \leq n^3$ ,  $n \leq n^3$  and  $1 \leq n^3$  for  $n \geq 1$ , one has  $f(n) \leq (2 + 1 + 7 + 2)n^3$ . Set  $C = 12$  and  $n_0 = 1$ . Thus  $f = O(n^3)$  and so  $f$  has cubic growth in  $n$ .
- 2  $f(n) = 100 + \frac{20}{n+1}$ . Set  $C = 101$  and  $n_0 = 19$ . Since  $\frac{20}{n+1} \leq 1$  for  $n \geq 19$ , we have  $f = O(1)$ . Hence  $f$  is asymptotically bounded by a constant.
- 3  $f(n) = 5\sqrt{2^{n+3} + n^2 - 2n}$ . Then  $f = O(2^{n/2})$ , and so  $f$  grows exponentially in  $n$ .

# Complexity of Algorithms

## Definition

If the running time of an algorithm is  $f(n)$ , where  $f$  is a *polynomial* and  $n$  is the input *size*, then the algorithm has *polynomial running time* and belongs to the complexity class **P**.

Polynomial-time algorithms are usually regarded as *efficient*.

In computer science, one is usually interested in the *worst-case* complexity of algorithms. However, when looking at the complexity of attacks against cryptographic schemes, their *average-case* complexity is much more important.

# Complexity of Algorithms: Examples

- The functions in the above examples (1) and (2) are polynomial.
- The running time of the Extended Euclidean Algorithm on input  $a, b \in \mathbb{N}$  is  $O(\text{size}(a) \text{size}(b))$ , so the algorithm is polynomial on the maximal input size.
- The running time of multiplying two numbers modulo  $n$  is  $O(\text{size}(n)^2)$ , which is polynomial.
- The running time of fast exponentiation modulo  $n$  is  $O(\text{size}(n)^3)$ , which is also polynomial.
- An algorithm which loops through  $N = 2^n$  items has exponential running time in  $n$ .

# Negligible Functions

We need the notion of a *negligible* function in the context of the probability of successful attacks.

## Definition

Let  $f : \mathbb{N} \rightarrow \mathbb{R}$  be a function. We say that  $f$  is *negligible* in  $n$ , if  $f = O(\frac{1}{q(n)})$  for all polynomials  $q$ , or equivalently, if  $f = O(\frac{1}{n^c})$  for all  $c > 0$ .

Negligible functions are eventually smaller than any inverse polynomial. This means that  $f(n)$  approaches zero faster than any of the functions  $\frac{1}{n}, \frac{1}{n^2}, \frac{1}{n^3}, \dots$

*Example:*  $f(n) = 10e^{-n}$  and  $2^{-\sqrt{n}}$  are negligible in  $n$ .

$f(n) = \frac{1}{n^2+3n}$  is not negligible, since  $f(n) = O(\frac{1}{n^2})$ , but  $f \neq O(\frac{1}{n^3})$ .

# Probability

We refer to textbooks on probability theory. We only consider *discrete probability spaces* and need the following notions:

- Probability space  $(\Omega, \mathcal{S}, Pr)$ , where  $\Omega$  is a sample space,  $\mathcal{S} = \mathcal{P}(\Omega)$  the set of events and  $Pr : \mathcal{S} \rightarrow [0, 1]$  a probability distribution.
- Independent events  $A, B$ , i.e.,  $P(A \cap B) = P(A) \cdot P(B)$ , and mutually independent events  $A_1, \dots, A_n$ .
- The conditional probability  $P[A|B] = \frac{P(A \cap B)}{P(B)}$  of events  $A, B$ .
- Random variables  $X : \Omega \rightarrow \mathbb{R}$ , their expectation  $E[X]$  and variance  $V[X]$ .
- Probability mass function (pmf)  $p_X(x) = Pr[X = x]$  of a random variable  $X$ .

# Uniform Distribution and Random Bits

## Definition

$Pr$  has a uniform distribution if all elementary events have equal probability:  $Pr[\{\omega\}] = \frac{1}{|\Omega|}$  for all  $\omega \in \Omega$ .

Random bits (or random numbers) are quite important in cryptography (but difficult to generate).

## Definition

A random bit generator (RBG) outputs a sequence of bits such that the corresponding random variables  $X_1, X_2, X_3, \dots$  satisfy

- 1  $Pr[X_n = 0] = Pr[X_n = 1] = \frac{1}{2}$  for all  $n \in \mathbb{N}$  (uniform distribution),  
and
- 2  $X_1, X_2, \dots, X_n$  are mutually independent for all  $n \in \mathbb{N}$ .

# Birthday Paradox

Let  $x_1, x_2, \dots, x_n$  be a sequence in a sample space  $\Omega$ . We say that there is a *collision* if at least two elements in the sequence are identical.

## Proposition

*Let  $Pr$  be a uniform distribution on a set  $\Omega$  of cardinality  $n$ . If we draw  $k = \left\lceil \sqrt{2 \ln(2)n} \right\rceil \approx 1.2 \sqrt{n}$  independent samples from  $\Omega$ , then the probability of a collision is around 50%.*

This fact is called *birthday paradox*: only  $k = 23$  random birthdays ( $n = 365$ ) are on average sufficient for a collision.

For  $|\Omega| = 2^n$ , around  $2^{n/2}$  independent samples probably give a collision.