

Modern Cryptography

Private Key Encryption Scheme

Shashank Singh

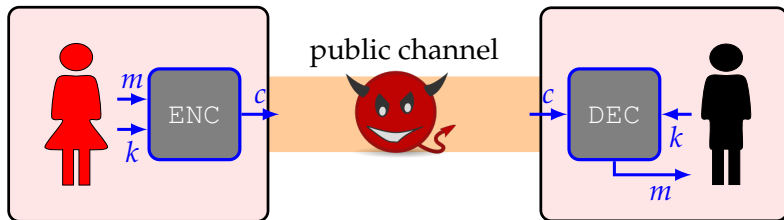
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SETTING OF PRIVATE-KEY CRYPTOGRAPHY..

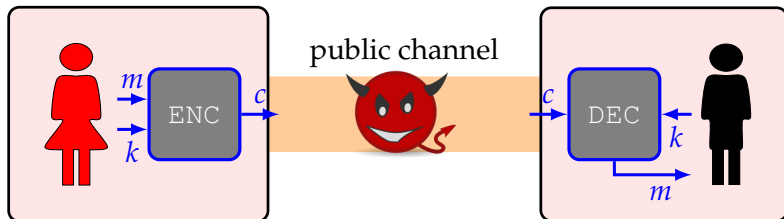
CLASSICAL CRYPTOGRAPHY



- ▶ Before sending the message (plaintext) m , Alice transforms (encrypts) it into a message c (ciphertext), using an algorithm ENC and a key k .
- ▶ Bob, on receiving c , decrypts it to get m , using a corresponding algorithm DEC and the same key k .

SETTING OF PRIVATE-KEY CRYPTOGRAPHY..

CLASSICAL CRYPTOGRAPHY





- ▶ The key k , needs to be (somehow) shared between the two communicating parties in advance and it is not known to the adversary.
- ▶ Alice and Bob could be same. Recall the **disk encryption**, where the same party encrypts the data on a disk and later decrypts it to get back the data.


ALGORITHMS

-a step by step solution to a problem.

Dictionary



**algorithm**
/ˈalgərɪð(ə)m/
noun
a process or set of rules to be followed in calculations or other problem-solving operations, especially by a computer.
"a basic algorithm for division"

Translations, word origin and more definitions

ALGORITHMS

Important features of an algorithm

1. Finiteness: An algorithm must always terminate after a finite number of steps.
2. Definiteness: Each step of an algorithm must be precisely defined.
3. Input: An algorithm has zero or more inputs: quantities that are given to it initially before the algorithm begins, or dynamically as the algorithm runs.
4. Output: An algorithm has one or more outputs: quantities that have a specified relation to the inputs.
5. Effectiveness: Its operations must all be sufficiently basic that they can in principle be done exactly and in a finite length of time by someone using pencil and paper.

(TIME) COMPLEXITY OF AN ALGORITHM

```
def is_prime(n):  
    for a in primes_upto(sqrt(n)):  
        if (a divides n):  
            return False  
    return True
```

- ▶ The time complexity deals with how fast or slow a particular algorithm performs.
- ▶ We define it as a numerical function $T(n)$, which represent the running time of the algorithm as a function of input size (in bits) n .
- ▶ But $T(n)$ depends on the implementation! A given algorithm will take different amounts of time on the same inputs depending factors as: processor speed; instruction set, disk speed, brand of compiler and etc. So we want to define $T(n)$, which do not depend on the above factors.

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- ▶ The time complexity deals with how fast or slow a particular algorithm performs.
- ▶ We define it as a numerical function $T(n)$, which represent the running time of the algorithm as a function of input size (in bits) n .
- ▶ The way around is to estimate efficiency of each algorithm **asymptotically**. We will measure time $T(n)$ as the number of elementary “steps” (in a model of computation), provided each such step takes constant time.

ASYMPTOTIC NOTATIONS

In this section we will consider functions that have \mathbb{N} as their domain and $\mathbb{R}_{\geq 0}$ as the range.

Definition (O -notation)

We say that a function $f(n)$ is *big-oh* of $g(n)$, written as $f(x) = O(g(n))$, if there exists positive constants c and n_0 such that

$$0 \leq f(n) \leq c \cdot g(n) \text{ for all } n \geq n_0.$$

In other words $O(g(n))$ denotes a set of functions that satisfy the above.

Remark

If $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)}$ exists and $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} \neq \infty$ then $f(n) = O(g(n))$.

ASYMPTOTIC NOTATIONS..

Example

- ▶ Show that $f(x) = x^2 + 2x + 1$ is $O(x^2)$.
- ▶ Show that n^2 is not $O(n)$.

ASYMPTOTIC NOTATIONS..

Definition (o -notation)

We say that a function $f(n)$ is *small-oh* or *little-oh* of $g(n)$, written as $f(x) = o(g(n))$, if for any positive non zero constant c , there exist a positive constant n_0 such that

$$0 \leq f(n) < c \cdot g(n) \text{ for all } n \geq n_0.$$

In other words $o(g(n))$ denotes a set of functions that satisfy the above.

Remark

If $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)}$ exists and $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = 0$ then $f(n) = o(g(n))$.

Example

Let $f(n) = n^2$, then $f(n) \neq o(n^2)$ but $f(n) = o(n^2 \log n)$.

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ASYMPTOTIC NOTATIONS..

Definition (Ω -notation)

We say that a function $f(n)$ is *big-omega* of $g(n)$, written as $f(x) = \Omega(g(n))$, if there exists positive constants c and n_0 such that

$$0 \leq c \cdot g(n) \leq f(n) \text{ for all } n \geq n_0.$$

In other words $\Omega(g(n))$ denotes a set of functions that satisfy the above.

Remark

If $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)}$ exists and $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} \neq 0$ then $f(n) = \Omega(g(n))$.

ASYMPTOTIC NOTATIONS..

Definition (ω -notation)

We say that a function $f(n)$ is *little-omega* of $g(n)$, written as $f(x) = \omega(g(n))$, if for any positive non zero constant c , there exist a positive nonzero constant n_0 such that

$$0 \leq c \cdot g(n) < f(n) \text{ for all } n \geq n_0.$$

In other words $\Omega(g(n))$ denotes a set of functions that satisfy the above.

Remark

If $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)}$ exists and $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = \infty$ then $f(n) = \omega(g(n))$.

Θ-NOTATION

Definition (Θ-Notation)

We say that a function $f(n)$ is **theta** of $g(n)$, written as $f(n) = \Theta(g(n))$, if there exists positive constants c_1, c_2 and n_0 such that

$$0 \leq c_2 \cdot g(n) \leq f(n) \leq c_1 \cdot g(n) \text{ for all } n \geq n_0.$$

Remark

If $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)}$ exists and $\lim_{n \rightarrow \infty} \frac{f(n)}{g(n)} = c$, where c is a non-zero positive constant, then $f(n) = \Theta(g(n))$.

TERMINOLOGY FOR COMPLEXITY OF ALGORITHMS

Complexity (input size is n)	Terminology
$\Theta(1)$	Constant complexity
$\Theta(\log n)$	Logarithmic complexity
$\Theta(n)$	Linear complexity
$\Theta(n \log n)$	Linearithmic complexity
$\Theta(n^b)$, for positive integer b	Polynomial complexity
$\Theta(b^n)$, where $b > 1$	Exponential complexity
$\Theta(n!)$	Factorial Complexity

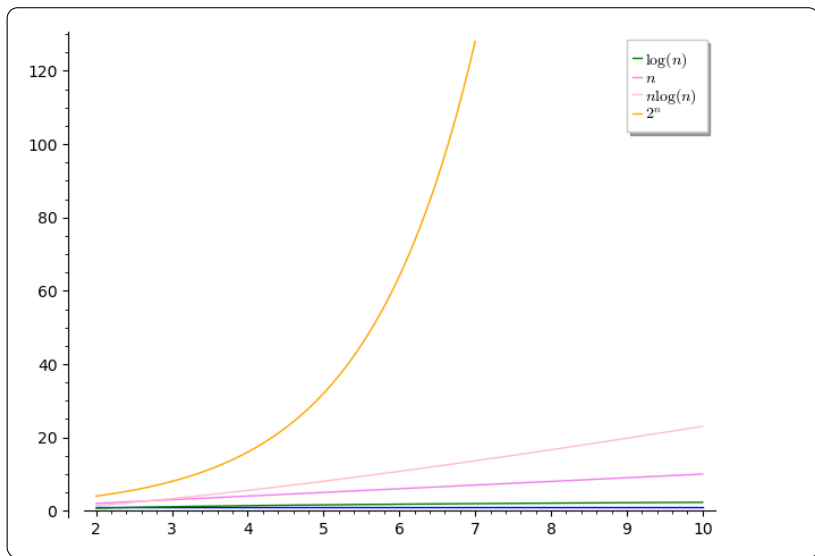
Efficient

- Algorithms with time complexity $\Theta(n^b)$, where n is input size and b is a non-zero positive integer, are called **polynomial time algorithms**.

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PPT- PROBABILISTIC POLYNOMIAL TIME ALGORITHM

- ▶ The polynomial time algorithm is an algorithms with time complexity $O(n^b)$, where n is input size and b is a fixed non zero positive integer.
- ▶ A **probabilistic algorithm** is one that has the capability of “tossing coins”, i.e. the algorithm has access to a random source of randomness that yields unbiased random bits that are independently equal to 1 with $\frac{1}{2}$ probability and to 0 with $\frac{1}{2}$ probability.
- ▶ A **probabilistic polynomial-time** algorithm is a probabilistic algorithm that may only perform a polynomial amount of operations including at most a polynomial number of coin-flips.

PRIVATE KEY ENCRYPTION

Let \mathcal{M} , \mathcal{K} and \mathcal{C} represent the set of possible messages (plaintexts), the set of possible keys and the set of possible ciphertexts respectively.



PRIVATE KEY ENCRYPTION..

A private key encryption algorithm is basically a set of three algorithms (**we will be more specific later**) (GEN, ENC, DEC), which have the following functionalities:

- GEN It is a probabilistic algorithm, called **key generation algorithm**. It outputs a key $k \in \mathcal{K}$ chosen according to some distribution.
- ENC It is called **encryption algorithm**. It takes as input a key $k \in \mathcal{K}$ and a message $m \in \mathcal{M}$ and outputs a **ciphertext** $c \in \mathcal{C}$.
- DEC It is known as **decryption algorithm**. It takes as input a key k and a ciphertext c and outputs a plaintext m .

Furthermore, it must satisfy the following correctness requirements:

$$\text{DEC}_k(\text{ENC}_k(m)) = m \forall m \in \mathcal{M}, \forall k \in \mathcal{K}. \quad (1)$$

KERCKHOFFS' PRINCIPLE



The cipher method must not be required to be secret, and it must be able to fall into the hands of the enemy without inconvenience.

- ▶ The security of a cryptographic scheme relies solely on the secrecy of the key, not on the secrecy of the underlying algorithms.

CAESER'S CIPHER (SHIFT CIPHER)

<i>A</i>	<i>B</i>	<i>C</i>	<i>D</i>	<i>E</i>	<i>F</i>	<i>G</i>	<i>H</i>	<i>I</i>	<i>J</i>	<i>K</i>	<i>L</i>	<i>M</i>
0	1	2	3	4	5	6	7	8	9	10	11	12
<i>N</i>	<i>O</i>	<i>P</i>	<i>Q</i>	<i>R</i>	<i>S</i>	<i>T</i>	<i>U</i>	<i>V</i>	<i>W</i>	<i>X</i>	<i>Y</i>	<i>Z</i>
13	14	15	16	17	18	19	20	21	22	23	24	25

Let $\mathcal{M} = \mathcal{C} = \mathcal{K} = \mathbb{Z}_{26} := \{0, 1, \dots, 25\}$.

For $0 \leq k \leq 25$, define

$$\text{ENC}_k(m) = (m + k) \mod 26$$

and

$$\text{DEC}_k(c) = (c - k) \mod 26$$

- ▶ Caesar's Cipher is the oldest recorded cipher, which is a Shift Cipher with the key $k = 3$.