Cryptography: notes

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Chapter 1

Efficient implementations of elementary operations

1.1 Notation

- Let b be a numeric base.
- Let n be a number in N.
- Length of a number: $l_b(n)$, k. It's equal to log(n).
- (*a*, *b*) is the Maximum Common Divisor of *a*, *b*.
- Let $n \in \mathbb{N}$: $n = (d_{k-1}, d_{k-2}, \dots, d_1, d_0)^1$.
- $\varphi(n)$: the number of elements a in [1, n] such that (a, n) = 1.
- \equiv_p is the equivalence in base p. Ex.: $5 \equiv_3 = 5 \mod 3 = 2$.
- Let $\mathbb{Z}_n[X]$ be the set of polynomials in X with coefficients in \mathbb{Z}_n .

1.2 Classification of the algorithms' complexity

In order to better identify the classes of complexity of the algorithms, the following 3 classes are defined:

- Polynomial time: $O(log^{\alpha}(n))$ bit operations, where $\alpha > 0$.
- Exponential time: $O(exp(c \cdot log(n)))$ bit operations, where c > 0.

 $^{^{1}}d_{k-1}\neq 0$

• Sub-exponential time: $O(exp(c \cdot log(n))^{\alpha})$ bit operations, where c > 0, $\alpha \in]0,1[$.

•

1.3 Basic bit operations

1.3.1 Sum of 3 bits - 3-bit-sum

Given n_1, n_2 their sum produces $n_1 + n_2$ and their carry. Since $n_1, n_2 \in [0, 1]$, then this operation can be done in O(1).

1.3.2 Summation of 2 numbers

Given n_1 , n_2 their sum produces $n_1 + n_2$. Since the sum is computed bit by bit, the 3-bit-sum is performed $max\{lenght(n_1), length(n_2)\}$ times. Each time the carry on of the previous sum is added to the two digits. This operation has then complexity: $O(max\{lenght(n_1), length(n_2)\}) = O(max\{lenght(n_1), length(n_2)\})$

1.3.3 Summation of n numbers

The summation of n numbers is simply the sum of two numbers, but performed n-1 times.

Let's assume that $\forall i \in [1, n] : M \ge a_i$. The complexity of this operation is then: $O((n-1) \cdot log(M)) = O(n)$.

1.3.4 Product of 2 numbers

If we consider the classic implementation of the binary multiplication, that is just a sequence of summations.

- The number of summations to execute is equal to the length of the smallest number, O(log(n)).
- The maximum cost of a single summation is O(log(m)).

• Then, $T(m \cdot n) = O(log(m) \cdot log(n))$, but, if we consider the worst case², that becomes $O(log^2(m))$.

1.3.5 Division of 2 numbers

Let's consider the division of two numbers m, n. This operations consists in finding two numbers q, r such that $m = q \cdot n + r$.

This is achieved by performing a succession of subtractions, until the ending condition $0 \le r < n$ is reached.

- Let's consider that the number of steps of this algorithm is O(log(q)).
- Moreover, $q \le m$: #steps = O(log(m)).
- It's assumed that the cost of the single subtraction is O(log(n)).
- Then, $T(\frac{m}{n}) = O(log(n) \cdot log(m))$.

Production of *n* numbers

Let's assume that $j \in [1, s+1]$ and $M = max(m_j)$. The cost of the operation $\prod_{j=1}^{s+1} m_j$ is then $O(s^2 \cdot log^2(M))$. This will now be considered our inductive hypothesis.

Proof by induction, on s:

- (1) Base case: $T(m_1 \cdot m_2) = O(log(m_1) \cdot log(m_2)) = O(k_1 \cdot k_2) \le c \cdot k_M^2$.
- (2) Base case: $T(m_1 \cdot m_2 \cdot m_3) = T(m_1 \cdot m_2) + T((m_1 \cdot m_2) \cdot m_3)$ $\leq c \cdot k_M^2 + c \cdot k_{m_1 \cdot m_2} + k_{m_3}$ $\leq c \cdot k_M^2 + c \cdot k_{M^2} + k_M$
- Inductive step: we assume the inductive hypothesis to be true up to s. Then,

$$T(\prod_{j=1}^{s+1} m_j) = T([\prod_{j=1}^{s} m_j] \cdot m_{s+1})$$

$$\leq c \cdot \sum_{j=1}^{s} (j \cdot k_M^2)$$

$$= c \cdot k_M^2 \cdot \frac{s \cdot (s-1)}{2}$$

$$= O(k_M^2 \cdot s^2)$$

$$= O(s^2 \cdot log^2(M))$$

²two numbers that are equally large

Applications

- An analogous dimonstration can be used to prove that $T(\prod_{i=1}^{s+1} m_i \mod n) =$ $O(s \cdot log^2(M))$
- This proof can be used to show that $T(m!) = O(m \cdot log^2(m))$.

Optimizations of more complex operations

1.4.1 Powers & Modular Powers

Let's consider what follows: $a^n = a \cdot a \cdot a \cdot \cdots \cdot a$, where a is repeated n times.

Trivial implementation

The most trivial implementation would consists in computing the product $\prod_{i=1}^{n} a$. This would imply a cost of $O(n^2 \cdot log^2(a))$.

What follows is a suggestion that could improve the cost of this operation.

Square & Multiply method for scalars, modular powers

Each number in \mathbb{Z} can be represented in a binary notation.

Let's consider $n = (b_{k-1}, b_{k-2}, ..., b_0) = \sum_{i=0}^{k-1} b_i \cdot 2^i$.

It is clear that we can spare a lot of computational resources by just calculating the powers of 2 and summing the ones that have $b_i = 1$. The following algorithm explains the procedure in detail. Let's compute the complexity of this algorithm:

- All of the assignments $X \leftarrow Y$ are implemented in O(log(Y)).
- The cost of 3 is $O(log(a) \cdot log(n))$, because it ensures that $A \le n$.
- Instructions 5 and 6 can be executed in O(log(m)).
- Instrucion 8 can be executed in $O(log^2(n))$.
- The cost of 10 is $O(log^2(n))$, because it ensures that $A \le n$.
- The loop is executed log(m) times.
- The total cost of this algorithm is then $O(log(n) \cdot log(a) + log(m) \cdot (log^2(n) + log(a)))$ log(m)))

 $= O(log^2(m) + log(m) \cdot log(n)).$

This algorithm can be easily converted for the computation of non-modular powers by applying the following changes:

Algorithm 1: The Square & Multiply Method

```
1 P \leftarrow 1;

2 M \leftarrow m;

3 A \leftarrow a \mod n;

4 while M > 0 do

5 q \leftarrow \lfloor \frac{M}{2} \rfloor;

6 r \leftarrow M - s \cdot q;

7 if r = 1 then

8 P \leftarrow P \cdot A \mod n;

9 end

10 A \leftarrow A^2 \mod n;

11 M \leftarrow q;

12 end

13 return P
```

```
1 A \leftarrow a \mod n \Longrightarrow A \leftarrow a;

2 P \leftarrow P \cdot A \mod n \Longrightarrow P \leftarrow P \cdot A;

3 A \leftarrow A^2 \mod n \Longrightarrow A \leftarrow A^2;
```

Square & Multiply method for polynomials

Let's consider $\Re = \frac{\mathbb{Z}_n[x]}{x^r-1}$. The modular powers of the elements in this set can be computed by using a variation of the Square & Multiply method.

- Assume that $f, g \in \Re$.
- Let $h(x) = f(x) \cdot g(x) = \sum_{j=0}^{2r-2} h_j \cdot x^j$.
- Where $h(j) = (\sum_{i=0}^{j} f_i \cdot g_{j-i} \mod n) \mod n$.
- Then, $T(h_i) = O(j \cdot log^2(n))$.
- Then, $T(h(x)) = O(\sum_{j=0}^{\log^2(n)}) = O(r^2 \cdot \log^2(n)).$

This result will be useful in the following computations. Let's now consider $\frac{h(x)}{x^r-1}$.

• When j > r - 1, $h_j \cdot x^j$ does not take any part in the computations.

- When j = r, then, $\frac{h_r \cdot x^r}{x^r 1} = h_r + \frac{h_r}{x^r 1}$ Or, in other words: $h_r \cdot x^r \equiv_{x^r - 1} h_r$.
- In the other cases: $h_{r+i} \cdot x^{r+i} \equiv_{x^r-1} h_{r-i} \cdot x^{r-i}$ for $1 \le i \le r-2$.

Then,
$$h(x) \equiv_{(n,x^r-1)} f(x) \cdot g(x) \equiv [\sum_{j=0}^{r-2} ((h_j + h_{r+j}) \mod n) \cdot x^j] + h_{r-1} \cdot x^{r-1}.$$

Therefore, $T(h(x) \mod (n, x^r - 1)) = O(r^2 \cdot log^2(n))$.

Finally, we can analyze the complexity of the computation of the modular power h(x) elevated to n.

In order to optimize the use of the computational resources, we can use a variation of the Square & Multiply method (See 1); although, this time, the computation of the partial products will be conducted by using the previously explained procedure (See 3).

The cost of this method would then be $T(\#Loops \cdot (h(x) \mod (n, x^r - 1))) = O(log(n) \cdot r^2 \cdot log(n)) = O(r^2 \cdot log^3(n)).$

1.4.2 Finding the *b*-expansion of $n(n_b)$

Let's consider the cost in bit operations of the conversion of a number n to a new base h

The algorithm used will be the classical: a succession of divisions by b.

- Let's consider $r_i \in \{0, 1, ..., b-1\}$.
- Let $n_b = (r_{k+1}, r_k, \dots, r_1, r_0)$.
- Then:

-
$$n = q_0 \cdot b + r_0$$

- $q_0 = q_1 \cdot b + r_1$
- ...
- $q_k = 0 \cdot b + q_k$

- Consider then that $q_k = r_{k+1}$
- And that $b^{k+2} > n > b^{k+1} \to (k+2) \cdot log(b) \le log(n) \le (k+1) \cdot log(b)$.
- $\therefore k = O(\frac{log(n)}{log(b)}).$

We can now proceed with the computation of the cost of this operation:

$$T(n_b) = T(\#Divisions \cdot q_i \bmod b) = O(\frac{log(n)}{log(b)} \cdot log(n) \cdot log(b)) = O(log^2(n))$$

1.4.3 How to use Bezout formula to compute modular inverses

An efficient way of computing the modular inverse of a given number a with in the group $\mathbb{Z}m^*$ uses the corollary of the *Bezout identity* and the *Extended Euclidean Algorithm*.

That is, given $a \cdot x \equiv_m 1$, we want to compute x.

Extended Euclidean Algorithm, given a, m computes the gcd(a, m) and also returns the coefficients x, y for which ax + my = 1.

At this point, the modular inverse of a in $\mathbb{Z}m^*$ is x:

- Let's consider that $ax + my \equiv_m 1$;
- Since $my \equiv_m 0$, then $ax \equiv_m 1$;
- $\therefore x \mod n$ is the modular inverse of a in $\mathbb{Z}m^*$ for its definition.

The complexity of this operation is then O(log(x)log(n)), because we have to compute the remainder of the division between x and n (this does not take into account the execution of the *Extended Euclidean Algorithm*).

1.4.4 Computing the order of an element in a cyclic group

The order of an element a in $\mathbb{Z}p^*$ (m = order(a)) is the minimum m such that $a^m \equiv_p 1$.

This problem is computationally hard, because the most efficient way to compute order(a) is to brute force its value.

The only optimization available is that we don't have to compute the modulars powers of a from scratch each time, but we can save the results at each iteration. Therefore, at each step we can only compute the modular product $(a^{p-1} \cdot a) \mod p$, that has a cost $O(log^2(p))$. The cost of this algorithm is then $O(order(a) \cdot log^2(p))$, because we have to compute $a^i \mod p$ for each attempt to find order(a).

1.4.5 Extended Euclidean Algorithm

The Extended Euclidean Algorithm is a variation of the classic Euclidean Algorithm, that computes the GCD between two numbers *a*, *b*.

It also provides the coefficients λ , μ such that $\lambda \cdot a + \mu \cdot b = GCD(a, b)$. This algorithm has a cost $O(log^3(max\{a,b\}))$.

1.4.6 Computation of square and m-th root of n

The following algorithm can be used to compute efficiently $\lfloor \sqrt[m]{n} \rfloor$. It is assumed that the length of the result is known and is l. Let's consider the cost of this algorithm:

Algorithm 2: The Extended Euclidean Algorithm

```
Data: a, b
   Result: (\lambda, \mu, GCD(a, b))
 1 old_r \leftarrow a;
 2 r \leftarrow b;
 s old_s ← 1;
 4 s \leftarrow 0;
 5 old_t \leftarrow 0;
 6 t \leftarrow 1;
 7 while r \neq 0 do
        quotient \leftarrow floor(old\_r/r);
        old_r \leftarrow r;
        old\_s \leftarrow s;
10
        old\_t \leftarrow t;
11
        r \leftarrow old\_r - quotient \cdot r;
        s \leftarrow old\_s - quotient \cdot s;
        t \leftarrow old\_t - quotient \cdot t;
14
15 end
16 return (s,t,old_r)
```

Algorithm 3: The Efficient m-th root of n

```
Data: n, m

Result: \lfloor \sqrt[m]{n} \rfloor

1 x_0 \leftarrow 2^{l-1};

2 for i \leftarrow 1 to l-1 do

3 | x_i \leftarrow x_{i-1} + 2^{l-i-1};

4 | \text{if } x_i^m > n \text{ then} \rangle

5 | x_i \leftarrow x_{i-1};

6 | \text{end} \rangle

7 end

8 return x_{l-1}
```

- Computing x_i^m has cost $O(log^2(n))$
- Comparing x_i^m and n has cost O(log(n)).
- The length of the loop is O(log(n)) iterations.
- The total cost is therefore $O(log^3(n))$.

1.4.7 Compute n, m given n^m

We can extract the base and the exponent of an integer by making different attempts. Let's consider the cost of this operation:

- We have to make at most *m* attempts by brute force;
- At each attempt we have to compute $\lfloor sqrt[m_i]n^m \rfloor$;
- This operation has total cost of $\sum_{m=3}^{log(n)} O(\frac{log(n)}{m} \cdot log^2(m) \cdot log(m)) + O(log^3(n))^3$;
- That is equal to $O(log^3(n))\sum_{m=3}^{log(n)}O(\frac{log^3(m)}{m})+O(log^3(n));$
- $\sum_{m=3}^{log(n)} O(\frac{log^3(m)}{m})$ can be approximated by calculating the correspondant integral, to O(loglog(n)).
- The final cost is therefore $O(log^3(n) \cdot (loglog(n))^2)$ = $O(log^{3+\epsilon})$, with $\epsilon \in (0,1)$

 $[\]frac{3 \log(n)}{m}$ is the length of the loop

Chapter 2

Elements of Number Theory

2.1 Definitions of Number Theory

2.1.1 The cyclic group \mathbb{Z}_n^*

The cyclic group \mathbb{Z}_n^* is defined as $\{a \in \mathbb{Z}_n : (a, n) = 1\}$.

The **generator** of \mathbb{Z}_n^* is a number in \mathbb{Z}_n^* such that $\forall a \in \mathbb{Z}_n^* \exists i : a \equiv_m g^i$. For this reason, \mathbb{Z}_n^* is also referred as $\langle g \rangle$.

An interesting property is that only for $[1,2,4,\phi,\phi^{\alpha},2\phi^{\alpha}]$, \mathbb{Z}_{n}^{*} is cyclic, where ϕ is a prime number and ϕ^{α} is a power of a prime number.

2.1.2 Pseudoprime number

For an integer a > 1, if a composite integer x divides ax - 1 - 1, then x is called a **Fermat pseudoprime** to base a.

In other words, a composite integer is a Fermat pseudoprime to base a if it successfully passes the Fermat primality test for the base *a*.

2.2 Reminders of Modular Arithmetic

2.2.1 Little Fermat's Theorem

Theorem 1 (Little Fermat's Theorem). Consider what follows:

- Let p be a prime number and $p \nmid a$.
- *Then,* $a^{p-1} \equiv_p 1$.

Proof. • Let's consider $A = \{na \mod p : n \in [1, p-1]\}$

- A has all distinct elements due to the Lemma 1.
- Then, $A = \mathbb{Z}_n^*$.
- Then, $\Pi_{n \in A} n \equiv_p \Pi_{n=1}^{p-1} na \Rightarrow (p-1)! \equiv_p (p-1)! a^{p-1}$
- Therefore, $a^{p-1} \equiv_p 1$ for the Wilson's Theorem 2.2.2

2.2.2 Wilson's Theorem

Theorem 2 (Wilson's Theorem). Consider what follows:

- Let $n \in \mathbb{N} \setminus \{0, 1\}$
- *Then,* $(n-1)! \equiv_n -1$.

Proof. The proof is by induction on *n*:

- Base case: n = 2Trivially, $1! \equiv_2 -1$
- Inductive step:

The theorem is assumed to be true up until n-1. Let's consider the case of n:

- Consider the polynomial g(x) = (x-1)(x-2)...(x-(n-1)).
- g has degree p-1 and costant term (p-1)!. It's roots are [1, p-1].
- Consider $h(x) = x^{p-1} 1$. h has also degree p-1 and leading term x^{p-1} .
- Let f(x) = g(x) h(x).
- Then, f has degree at most p-2 (since the leading terms cancel), and modulo p also has the n-1 roots 1,2,...,n-1.
- But Lagrange's theorem says it cannot have more than p-2 roots.
- $: f \equiv_n 0.$
- Its costant term, $(n-1)! + 1 \equiv_n 0 \iff (n-1)! \equiv_n -1$

2.2.3 Euler-Fermat's Theorem

Theorem 3 (Euler-Fermat's Theorem). *Consider what follows:*

- Let $n \in \mathbb{Z}$ be a number.
- Then, $a^{\varphi(n)} \equiv_n 1$.

2.2.4 Bézout's identity

Theorem 4 (Bezout's identity). *Consider what follows:*

- Let a and b be integers with greatest common divisor d.
- Then there exist integers x and y such that ax + by = d.
- Moreover, the integers of the form az + bt are exactly the multiples of d.

2.2.5 Chinese Reminder's Theorem

Theorem 5 (Chinese Reminder's Theorem). *Given a system of congruences:*

- $x \equiv_{m_1} a_1$
- $x \equiv_{m_2} a_2$
- ...
- $x \equiv_{m_r} a_r$

In which each module is prime with each others, that is $\forall i \neq j : (m_i, m_j) = 1$, then there exists a simultaneous solution x to all of the congruences, and any two solutions are congruent to one another modulo $M = m_1 m_2 \dots m_r$.

Proof. Consider what follows:

- Suppose that x' and x'' are two solutions.
- Let x = x' x''.
- Then x must be congruent to 0 modulo each m_i and hence modulo M.
- Let $M_i = M/m_i$, to be the product of all of the moduli except for the i-th.
- Then $GCD(m_i, M_i) = 1$ and therefore $\exists N_i : M_i N_i \equiv_{m_i} 1$.
- Set $x = \sum_i a_i M_i N_i$;
- Then, for each i we see that the terms in the sum other than the i-th term are all divisible by m_i , because $m_i|M_j$ when $i \neq j$.
- Thus, for each *i* we have: $x \equiv_{m_i} a_i M_i N_i \equiv_{m_i} a_i$,

2.2.6 Bijection of a modular function Lemma

Lemma 1 (Bijection of a modular function Lemma). *Consider* $a \in \mathbb{Z}_n^*$ *and* $f : \mathbb{Z}_n \to \mathbb{Z}_n$. *Then,* f *is a bijection.*

Proof. Part 1: *f* is injective.

- Assume $f(x_1) = f(x_2) \in \mathbb{Z}_n \iff ax_1 \equiv_n ax_2$.
- Then $\exists k \in \mathbb{Z} : ax_1 ax_2 = kn$.
- Therefore, $a(x_1 x_2) = kn \iff a^{-1}a(x_1 x_2) = a^{-1}kn \iff (x_1 x_2) = a^{-1}kn \iff x_1 x_2 \equiv_m 0.$
- $\therefore x_1 \equiv_m x_2$, so f is an injection.

Part 2: *f* is surjective.

- Let $b \in \mathbb{Z}_n$.
- Let $\overline{x} = a^{-1}b \in \mathbb{Z}_n$.
- Then, $f(\overline{x}) \equiv_m a \cdot \overline{x}$ $\equiv_m a \cdot a^{-1} \cdot b \equiv_m b$.
- Therefore, *f* is surjective.

Since f is injective $\land f$ is surjective, then f is bijective.

2.2.7 Euler's φ function Lemmas

Lemma 2 (Sum of the prime divisors' φ -function). $\forall n \in \mathbb{N} \setminus \{0\} : \sum_{\frac{n}{d}} \varphi(d) = n$, where $\frac{n}{d}$ is the set of prime divisors of n.

Proof. • Let *B* be $\{\frac{h}{n}: h \in \mathbb{Z}_n \land n \in \mathbb{N}\}.$

- Therefore, $B = \bigcup_{\frac{n}{d}} \{a \in \{1, \dots, n\} \land (a, d) = 1\}.$
- Then, $n = |B| = \sum_{\substack{d \ n}} \varphi(d)$.
- Consider $(a_1, d_1) = 1 \land (a_2, d_2) = 1$, where $d_1 | n \land d_2 | n$.

- Then, $\frac{a_1}{d_1} = \frac{a_2}{d_2} \iff a_1 d_2 = a_2 d_1$.
- Then, $d_1|a_1d_2 \Rightarrow d_1|d_2 \wedge$ $d_2|a_2d_1 \Rightarrow d_2|d_1$.
- $\therefore d_1 = d_2$. This is clearly a contradiction, because in *B* each divisor is counted

Lemma 3 (Number of divisors of a prime number's power). Let $p \in \mathbb{N}$ be a prime num*ber, and* $\alpha \in \mathbb{N}$.

Then, $\varphi(p^{\alpha}) = p^{\alpha-1}(p-1)$.

Proof. The proof is by induction on α .

Case base: $\alpha = 1$

Then,
$$p = \sum_{\frac{d}{p}} \varphi(d) = \varphi(1) + \varphi(p) = 1 + \varphi(p)$$

 $\Rightarrow \varphi(p) = p - 1$.

Case base: $\alpha = 2$

Then,
$$p^2 = \sum_{\frac{d}{p^2}} \varphi(d) = \varphi(1) + \varphi(p) + \varphi(p^2)$$

$$= 1 + p - 1 + \varphi(p^2) \Rightarrow \varphi(p^2) = p^2 - p - 1 + 1$$

 $= p \cdot (p-1) = p^{\alpha-1}(p-1)$

Inductive Step: we can now assume that $\varphi(p^{\alpha}) = p^{\alpha-1}(p-1)$ up until $\alpha - 1$. We'll proceed now to demonstrate that this is also valid for each α .

$$p^{\alpha} = \sum_{\frac{d}{p^{\alpha}}} \varphi(d)$$

$$= \sum_{i=0}^{\alpha-1} \varphi(p^i) + \varphi(p^\alpha)$$

$$= \sum_{i=0}^{\alpha-1} \varphi(p^i) + \varphi(p^{\alpha})$$

$$\Rightarrow \varphi(p^{\alpha}) = p^{\alpha} - p^{\alpha-1} = p^{\alpha-1}(p-1).$$

Multiplicativity of the Euler's φ -function

Theorem 6 (Multiplicativity of the Euler's φ -function). Let's consider the Euler's func $tion \varphi(n) = |\mathbb{Z}_n^*$.

Let's consider that $n = \prod_{i=1}^r p_i^{\alpha_i}$, where p_i is a prime number, and $\alpha_i \in \mathbb{N}$.

Then, =

Proof.

Useful Facts

• The GMP library is a free library for arbitrary precision arithmetic. It implements all the basic arithmetic operations with the maximum efficency possible.

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