

Cryptography

Modern cryptography was born in 1970's when computationally easy-to-verify but 'hard-to-solve' problems were discovered.

Cryptography is closely related to some advanced topics in computational complexity.

Synopsis

1. Computationally Secure Encryption
2. Pseudorandom Generator
3. Pseudorandom Function
4. One-Way Function
5. Zero Knowledge Proof
6. Remark

Computationally Secure Encryption

An **encryption scheme** is a pair (E, D) of algorithms such that

$$D_k(E_k(x)) = x$$

for all **key** k and **plaintext** x . Obviously E_k is **one-one** for every k .

Shannon's Perfect Secrecy

(E, D) is **perfectly secret** if for every pair $x, x' \in \{0, 1\}^m$, the distributions $E_{U_n}(x)$ and $E_{U_n}(x')$ are identical.

- ▶ n is the key length.
- ▶ U_n is the uniform distribution over $\{0, 1\}^n$.

One Time Pad Encryption Scheme, Vernan 1917

Encryption:

- ▶ Plaintext $x \in \{0, 1\}^n$.
- ▶ Generate a key $k \in_{\mathcal{R}} \{0, 1\}^n$, encrypt x by $x \oplus k$.

Decryption:

- ▶ Ciphertext $y \in \{0, 1\}^n$.
- ▶ The plaintext is recovered by $y \oplus k$.

If a key k is used twice, useful information can be derived.

One Time Pad Encryption Scheme

Fact. The one time pad encryption scheme is perfectly secure.

It is crucial that the **key** is as long as the **message**.

Shannon Theorem. Suppose (E, D) is an encryption scheme. If $n < m$, then there exist x, x' such that $E_{U_n}(x)$ and $E_{U_n}(x')$ differ.

Proof.

A proof can be read off from the proof of Lemma. □

- ▶ Perfectly secret encryption scheme is not a practical scenario.
- ▶ Modern cryptography offers a solution.

Negligible Functions

A function $\epsilon : \mathbf{N} \rightarrow [0, 1]$ is **negligible** if

$$\forall c. \exists N. \forall n \geq N. \epsilon(n) < \frac{1}{n^c}.$$

In other words ϵ is negligible if it tends to 0 faster than $\frac{1}{p(n)}$ for every polynomial $p(n)$.

- ▶ Events with negligible probability can be practically ignored.
- ▶ ϵ is **not** negligible if $\exists c. \epsilon(n) \geq \frac{1}{n^c}$ for infinitely many n .

Computationally Secure Encryption Scheme

An encryption scheme (E, D) for keys of length n and messages of length m is **computationally secure** if for every P-time PTM \mathbb{A} there is a negligible function $\epsilon : \mathbf{N} \rightarrow [0, 1]$ such that

$$\left| \Pr_{k \in_R \{0,1\}^n, x \in_R \{0,1\}^m} [\mathbb{A}(E_k(x)) = (i, b) \wedge x_i = b] - \frac{1}{2} \right| \leq \epsilon(n).$$

Is there a computationally secure encryption scheme? The answer is conditional.

Lemma. Suppose $\mathbf{P} = \mathbf{NP}$. Let (E, D) be a P-time encryption scheme with key shorter than message. A P-time algorithm \mathbb{A} exists such that for every message length m , there is a pair $x_0, x_1 \in \{0, 1\}^m$ satisfying

$$\Pr_{b \in_{\mathbf{R}} \{0,1\}, k \in_{\mathbf{R}} \{0,1\}^n} [\mathbb{A}(E_k(x_b)) = b] \geq 3/4$$

where n is the key length and $n < m$.

1. Let S be defined as follows:

$$y \in S \text{ iff } \exists k. y = E_k(x_0), \text{ where } x_0 = 0^m.$$

2. If $\mathbf{P} = \mathbf{NP}$ then S is P-time decidable by some algorithm \mathbb{A} .

► $\mathbb{A}(x) = 0$ iff $x \in S$.

3. Let $D_x = \text{distribution } E_{U_n}(x)$. Then $\Pr[\mathbb{A}(D_{x_0})=0] = 1$.

If $\Pr[D_x \in S] > \frac{1}{2}$ for all x then one would have

$$\frac{1}{2} < \Pr_x[\Pr[D_x \in S]] = \Pr_k[\Pr_x[E_k(x) \in S]] \leq \frac{1}{2},$$

where \leq holds because $|S| \leq 2^n \leq 2^{m-1}$ by the definition of S and E_k is injective.

It follows that $\Pr[D_{x_1} \in S] \leq \frac{1}{2}$ for some $x_1 \in \{0, 1\}^m$. According to the definition of \mathbb{A} , one has $\Pr[\mathbb{A}(D_{x_1})=0] \leq \frac{1}{2}$. Hence

$$\begin{aligned} \Pr_{b,k}[\mathbb{A}(E_k(x_b))=b] &= \frac{1}{2}\Pr[\mathbb{A}(D_{x_0})=0] + \frac{1}{2}\Pr[\mathbb{A}(D_{x_1})=1] \\ &= \frac{1}{2} + \frac{1}{2}\Pr[\mathbb{A}(D_{x_1})=1] \\ &\geq \frac{3}{4}. \end{aligned}$$

P \neq **NP** is necessary for modern cryptography. We do not know if it is sufficient.

Pseudorandom Generator

Modern cryptography addresses the long key issue by studying how to generate long keys from short ones.

- ▶ An observer cannot detect efficiently any useful difference between a pseudorandom key and a truly random key.

What is a pseudorandom string? How do we characterize pseudorandom strings?

- ▶ For modern cryptography it suffices that encrypted messages are distributed in a way that looks random to all **efficient** observers.

Pseudorandom Generator

Let $G : \{0, 1\}^* \rightarrow \{0, 1\}^*$ and $\ell : \mathbf{N} \rightarrow \mathbf{N}$ be P-time computable such that $\ell(n) > n$ for all n and $|G(x)| = \ell(|x|)$ for all $x \in \{0, 1\}^*$.

G is a computationally secure **pseudorandom generator** of **stretch $\ell(n)$** if, for every P-time PTM \mathbb{A} , there exists a negligible function $\epsilon : \mathbf{N} \rightarrow [0, 1]$ such that

$$|\Pr[\mathbb{A}(G(U_n)) = 1] - \Pr[\mathbb{A}(U_{\ell(n)}) = 1]| \leq \epsilon(n).$$

1. Yao. Theory and Applications of Trapdoor Functions. FOCS 1982.



A pseudorandom generator says nothing about how it is constructed.

Unpredictability

Let $G : \{0, 1\}^* \rightarrow \{0, 1\}^*$ be P-time computable with stretch $\ell(n)$, where $\ell : \mathbf{N} \rightarrow \mathbf{N}$ is P-time computable such that $\forall n. \ell(n) > n$.

We say that G is **unpredictable** if for every P-time PTM \mathbb{B} there is a negligible function $\epsilon : \mathbf{N} \rightarrow [0, 1]$ such that

$$\left| \Pr_{x \in_R \{0,1\}^n, y = G(x), i \in_R [\ell(n)]} [\mathbb{B}(1^n, y_1, \dots, y_{i-1}) = y_i] - \frac{1}{2} \right| \leq \epsilon(n).$$

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1. M. Blum, S. Micali. How to Generate Cryptographically Strong Sequences of Pseudorandom Bits. FOCS 1982.



Unpredictability \Leftarrow Pseudorandomness

Suppose G is a pseudorandom generator. If it is not unpredictable then there is some c such that

$$\left| \Pr_{x \in_R \{0,1\}^n, y=G(x), i \in_R [\ell(n)]} [\mathbb{B}(1^n, y_1, \dots, y_{i-1}) = y_i] - \frac{1}{2} \right| \geq \frac{1}{n^c}$$

holds for a P-time PTM \mathbb{B} for infinitely many n . Some i exists such that

$$\left| \Pr_{x \in_R \{0,1\}^n, y=G(x)} [\mathbb{B}(1^n, y_1, \dots, y_{i-1}) = y_i] - \frac{1}{2} \right| \geq \frac{1}{n^c \ell(n)}$$

for infinitely many n . It follows from $\Pr[\mathbb{B}(U_{\ell(n)}) = 1] = \frac{1}{2}$ that

$$\Pr[\mathbb{B}(G(U_n)) = 1] - \Pr[\mathbb{B}(U_{\ell(n)}) = 1] \geq \frac{1}{n^c \ell(n)}$$

for infinitely many n , which is a contradiction.

Unpredictability \Rightarrow Pseudorandomness

Theorem (Yao, 1982). If G is unpredictable, then it is a pseudorandom generator.

1. Yao. Theory and Applications of Trapdoor Functions. FOCS 1982.



Unpredictability \Rightarrow Pseudorandomness

Let $\ell : \mathbf{N} \rightarrow \mathbf{N}$ be P-time computable such that $\ell(n) \geq n$.

Let $G : \{0, 1\}^* \rightarrow \{0, 1\}^*$ be P-time computable unpredictable function with stretch ℓ .

Suppose G is not a pseudorandom generator. Then there is some constant c and some P-time PTM A such that, wlog,

$$\Pr[\mathsf{A}(G(U_n)) = 1] - \Pr[\mathsf{A}(U_{\ell(n)}) = 1] \geq \frac{1}{n^c}$$

for infinitely many n .

Unpredictability \Rightarrow Pseudorandomness

For $i \leq \ell(n)$, the **hybrid** distribution \mathcal{D}_i is defined as follows:

1. choose $x \in_{\mathcal{R}} \{0, 1\}^n$ and compute $y = G(x)$;
2. output $y_1, \dots, y_i, z_{i+1}, \dots, z_{\ell(n)}$ with $z_{i+1}, \dots, z_{\ell(n)} \in_{\mathcal{R}} \{0, 1\}$.

We notice that $\mathcal{D}_0 = U_{\ell(n)}$ and $\mathcal{D}_{\ell(n)} = G(U_n)$.

Let $p_i = \Pr[\mathbf{A}(\mathcal{D}_i) = 1]$. By assumption for infinitely many n ,

$$p_{\ell(n)} - p_0 = (p_{\ell(n)} - p_{\ell(n)-1}) + (p_{\ell(n)-1} - p_{\ell(n)-2}) + \dots + (p_1 - p_0) \geq \frac{1}{n^c}.$$

Unpredictability \Rightarrow Pseudorandomness

Algorithm \mathbb{B} asserts that everything \mathbb{A} says is correct.

- ▶ Input 1^n , $i \in [\ell(n)]$ and y_1, \dots, y_{i-1} .
 1. randomly generate $z_i, \dots, z_{\ell(n)}$;
 2. compute $a = \mathbb{A}(y_1, \dots, y_{i-1}, z_i, \dots, z_{\ell(n)})$;
 3. output z_i if $a = 1$ and $1 - z_i$ if $a = 0$.

We are done if we can prove the following inequality

$$\Pr_{x \in_R \{0,1\}^n, y=G(x), i \in_R [\ell(n)]} [\mathbb{B}(1^n, y_1, \dots, y_{i-1}) = y_i] - \frac{1}{2} \geq \frac{1}{n^c \ell(n)},$$

which can be derived if the following holds for every $i \in [\ell(n)]$:

$$\Pr_{x \in_R \{0,1\}^n, y=G(x)} [\mathbb{B}(1^n, y_1, \dots, y_{i-1}) = y_i] = \frac{1}{2} + (p_i - p_{i-1}).$$

Unpredictability \Rightarrow Pseudorandomness

\mathbb{B} predicts y_i correctly if $a = 1 \wedge z_i = y_i$ or $a = 0 \wedge z_i = 1 - y_i$. This event happens with probability

$$\frac{1}{2} \Pr_{x,y=G(x)}[a = 1 | z_i = y_i] + \frac{1}{2} (1 - \Pr_{x,y=G(x)}[a = 1 | z_i = 1 - y_i]) .$$

Now $\Pr_{x \in_R \{0,1\}^n, y=G(x)}[a = 1 | z_i = y_i] = p_i$. On the other hand,

$$\begin{aligned} p_{i-1} &= \Pr[\textcolor{violet}{A}(\mathcal{D}_{i-1}) = 1] \\ &= \Pr[a = 1 | z_i = y_i]/2 + \Pr[a = 1 | z_i = 1 - y_i]/2 \\ &= p_i/2 + \Pr[a = 1 | z_i = 1 - y_i]/2. \end{aligned} \tag{1}$$

We get $\Pr[a = 1 | z_i = 1 - y_i] = 2p_{i-1} - p_i$ from (1).

Theorem Given a pseudorandom generator with stretch n^c , one can design a computationally secure encryption scheme (E, D) using n -length keys for n^c -length messages.

Given a random key of length n , generate a key of length n^c using the pseudorandom generator, and then apply the one-time pad encryption scheme.

Application: Derandomization

If pseudorandom generator exists, then we can construct **subexponential deterministic** algorithms for problems in **BPP**.

- ▶ This is the **derandomization** of **BPP**.

The basic idea:

- ▶ Let L be decided by an n^d -time PTM \mathbb{P} with bounded error.
- ▶ For every small ϵ let c be such that $0 < \frac{d}{c} < \epsilon < 1$.
- ▶ Apply to all strings of length $n^{\frac{d}{c}}$ the pseudorandom generator with stretch n^c and then execute \mathbb{P} by following the choices prescribed by the produced pseudorandom strings of length n^d .
- ▶ The algorithm runs in time $O(2^{n^\epsilon})$

Pseudorandom Function

Let \mathcal{F}_n denote the set of all functions of type $\{0, 1\}^n \rightarrow \{0, 1\}^n$.

1. Generally $n2^n$ bits are necessary to specify a function in \mathcal{F}_n .
2. Consequently its computation is not efficient.

We look for an efficient subset \mathcal{G}_n of \mathcal{F}_n that appears random.

1. Every element of \mathcal{G}_n is specified by n bits.
2. Every element of \mathcal{G}_n is P-time computable.
3. Yet no P-time PTM can detect noticeable difference between a random element of \mathcal{G}_n and a random element of \mathcal{F}_n .

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- ▶ There are 2^{n2^n} elements in \mathcal{F}_n .
 - ▶ There are only 2^n elements in \mathcal{G}_n .

Pseudorandom functions are pseudorandom generators with exponential stretch.

- ▶ A pseudofunction is a blackbox, a distinguisher can only ask for the values of the function at a **small** number of inputs.

Pseudorandom Function

Let $\{f_k\}_{k \in \{0,1\}^*}$ be a family of functions such that

- ▶ $f_k : \{0, 1\}^{|k|} \rightarrow \{0, 1\}^{|k|}$ for every $k \in \{0, 1\}^*$, and
- ▶ $f_k(x)$ is P-time computable from k, x .

The family $\{f_k\}_{k \in \{0,1\}^*}$ is **pseudorandom** if for every P-time probabilistic OTM \mathbb{A} there is a negligible function $\epsilon : \mathbf{N} \rightarrow [0, 1]$ such that for all n ,

$$\left| \Pr_{k \in_R \{0,1\}^n} [\mathbb{A}^{f_k}(1^n) = 1] - \Pr_{g \in_R \mathcal{F}_n} [\mathbb{A}^g(1^n) = 1] \right| \leq \epsilon(n).$$

\mathbb{A} needs no input. The string 1^n marks the input length.

Pseudorandom Generator \Leftarrow Pseudorandom Function

Suppose $\{f_k\}_{k \in \{0,1\}^*}$ is a pseudorandom family of functions. It follows from definition

- ▶ that for every polynomial $\ell(n)$, the map G defined by

$$k \in \{0,1\}^n \mapsto f_k(1), \dots, f_k(\ell(n)) \in \{0,1\}^{n\ell(n)}$$

is a pseudorandom generator.

Goldreich-Goldwasser-Micali Theorem.

Suppose that there exists a pseudorandom generator G with stretch $\ell(n) = 2n$. Then there exists a pseudorandom function family.

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1. O. Goldreich, S. Goldwasser, S. Micali. How to Construct Random Functions. FOCS 1984.



Pseudorandom Generator \Rightarrow Pseudorandom Function

Let G be a pseudorandom generator with stretch $2n$.

- ▶ $G_0(x)$ is the first n bits of $G(x)$;
- ▶ $G_1(x)$ is the last n bits of $G(x)$.

For each seed $k \in \{0, 1\}^n$ the function f_k is defined by

$$f_k(x) = G_{x_n}(G_{x_{n-1}}(\dots G_{x_1}(k)\dots)).$$

We will prove that $\{f_k\}_{k \in \{0,1\}^*}$ is a pseudorandom function family.

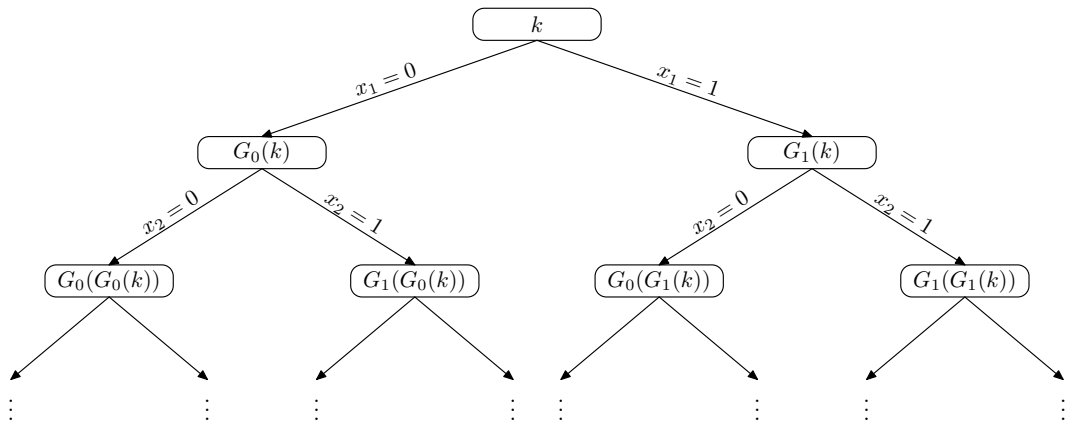


Figure: The Algorithm that Calculates $f_k(x)$.

Pseudorandom Generator \Rightarrow Pseudorandom Function

Let \mathbb{A} be a T -time PTM that distinguishes $\{f_k\}_{k \in \{0,1\}^n}$ and \mathcal{F}_n . I.e. some c exists such that the following inequality holds for infinitely many n ,

$$\Pr_{g \in_R \mathcal{F}_n}[\mathbb{A}^g(1^n) = 1] - \Pr_{k \in_R \{0,1\}^n}[\mathbb{A}^{f_k}(1^n) = 1] \geq \frac{1}{n^c}.$$

We construct a P-time PTM \mathbb{B} that distinguishes U_{2n} and $G(U_n)$ with $\frac{1}{nT} \cdot \frac{1}{n^c}$ bias.

- ▶ Define a **random** implementation \mathcal{O} of the oracles f_{U_n} in terms of G .
 1. generate a seed $k \in_R \{0,1\}^n$ randomly;
 2. run the algorithm that calculates f_k on **all** queries.
- ▶ We then modify \mathcal{O} to get $\{\mathcal{O}_i\}_{i \leq nT}$ using hybrid approach.

Pseudorandom Generator \Rightarrow Pseudorandom Function

For $i \leq nT$ the random oracle \mathcal{O}_i is defined as follows:

1. In the first i -th steps generate children randomly.
2. After the first i -steps generate children pseudo-randomly using G .
3. The random answers must be **consistent**!

Clearly \mathcal{O}_0 is \mathcal{O} , and \mathcal{O}_{nT} is a random function.

Let $p_i = \Pr[\mathbb{A}^{\mathcal{O}_i}(1^n) = 1]$. Observe that

$$\blacktriangleright p_0 = \Pr_{k \in_R \{0,1\}^n}[\mathbb{A}^{f_k}(1^n) = 1] \text{ and } p_{nT} = \Pr_{g \in_R \mathcal{F}_n}[\mathbb{A}^g(1^n) = 1].$$

By assumption $p_{nT} - p_0 \geq \frac{1}{n^c}$.

Algorithm \mathbb{B} .

1. Input $k \in \{0, 1\}^{2n}$.
2. Generate $i \in_{\mathbb{R}} [nT]$.
3. Run $\mathbb{A}^{\mathcal{O}_i}(1^n)$, with the modification of \mathcal{O}_i that in the i -th invocation the two children are the first respectively the last n bits of k .

The following can be easily verified.

- ▶ If $k \in_{\mathbb{R}} U_{2n}$, then \mathbb{B} 's output is distributed as $\mathbb{A}^{\mathcal{O}_i}(1^n)$.
- ▶ If $k \in_{\mathbb{R}} G(U_n)$, then \mathbb{B} 's output is distributed as $\mathbb{A}^{\mathcal{O}_{i-1}}(1^n)$.

Using hybrid argument, $\Pr[\mathbb{B}(U_{2n}) = 1] - \Pr[\mathbb{B}(G(U_n)) = 1]$ is

$$\sum_{i \in [nT]} \frac{\Pr[\mathbb{A}^{\mathcal{O}_i}(1^n) = 1]}{nT} - \sum_{i \in [nT]} \frac{\Pr[\mathbb{A}^{\mathcal{O}_{i-1}}(1^n) = 1]}{nT} = \frac{p_{nT}}{nT} - \frac{p_0}{nT} \geq \frac{1}{nT} \cdot \frac{1}{n^c}.$$

Application: One Key for Many Messages

By Goldreich-Goldwasser-Micali Theorem and Yao's Theorem, the string

$$f_k(r_1)f_k(r_2)f_k(r_3)\dots f_k(r_{\ell(k)})$$

is unpredictable.

1. **Alice** encrypts a message $x \in \{0, 1\}^n$ by choosing $r \in_R \{0, 1\}^n$ and sends $(r, f_k(r) \oplus x)$ to **Bob**, where $k \in \{0, 1\}^n$ is the key.
2. **Bob** receives (r, y) and calculates $f_k(r) \oplus y$ to recover x .

Application: Message Authentication Code

For the same reason the following protocol is secure.

1. Alice sends x to Bob.
2. Bob sends $(x, f_k(x))$ to Alice.
3. Alice receives (x, y) and checks if $y = f_k(x)$ to verify that the message has not been corrupted.

Application: Lower Bound for Machine Learning

In machine learning the goal is to learn a function f from a sequence of examples $(r_1, f(r_1)), \dots, (r_k, f(r_k))$.

- ▶ The existence of pseudorandom function implies that even if f is P-time computable, there is no way to learn it in P-time.

One-Way Function

Suppose $G : \{0, 1\}^* \rightarrow \{0, 1\}^*$ is a pseudorandom generator.

For every P-time PTM \mathbb{A} there must be a negligible function $\epsilon : \mathbf{N} \rightarrow [0, 1]$ such that the following holds for every n ,

$$\Pr_{x \in_R \{0,1\}^n} [\mathbb{A}(1^n, G(x)) = x' \wedge G(x') = G(x)] \leq \epsilon(n).$$

One-Way Function

A P-time function $f : \{0, 1\}^* \rightarrow \{0, 1\}^*$ is a **one-way function** if for every P-time PTM \mathbb{A} there is a negligible function $\epsilon : \mathbf{N} \rightarrow [0, 1]$ such that for every n ,

$$\Pr_{x \in_R \{0,1\}^n, y=f(x)}[\mathbb{A}(1^n, y) = x' \wedge f(x') = y] \leq \epsilon(n).$$

Let $f : \{0, 1\}^* \rightarrow \{0, 1\}^*$ be a P-time computable function such that $\forall x. |x| \leq |f(x)|$.

- ▶ If $\mathbf{P} = \mathbf{NP}$ then $\{(l, u, y) \mid \exists x. f(x) = y \wedge l \leq x \leq u\} \in \mathbf{P}$.
- ▶ By divide-and-conquer one can compute f^{-1} in P-time.

The existence of one way function implies $\mathbf{P} \neq \mathbf{NP}$.

Integer multiplication is believed to be one-way.

Theorem. If one-way **permutations** exist, then for every $c \in \mathbf{N}$, there exists a pseudorandom generator with stretch $S(n) = n^c$.

1. Q. Yao. Theory and Applications of Trapdoor Functions. FOCS 1982.

Theorem. If one-way **functions** exist, then for every $c \in \mathbf{N}$, there exists a pseudorandom generator with stretch $S(n) = n^c$.

1. J. Håstad, R. Impagliazzo, L. Levin and M. Luby. A Pseudorandom Generator from any One-way Function. SIAM Journal on Computing, 28:1364-1396, 1999.

The crucial step is to obtain a pseudorandom generator that extends input by one bit.

- ▶ If f is a one-way permutation, then $G(x, r) = f(x), r, x \odot r$ is a pseudorandom generator. Intuitively r is random, $f(x)$ is pseudorandom, and the $(2n+1)$ -th bit cannot be predicted with probability noticeably larger than $1/2$.

We shall prove **Theorem** using Goldreich-Levin Theorem.

Goldreich-Levin Theorem. Suppose $f : \{0, 1\}^* \rightarrow \{0, 1\}^*$ is a one-way permutation. Then for every P-time PTM \mathbb{A} there is a negligible function $\epsilon : \mathbf{N} \rightarrow [0, 1]$ such that

$$\left| \Pr_{x, r \in_{\mathbf{R}} \{0, 1\}^n} [\mathbb{A}(f(x), r) = x \odot r] - \frac{1}{2} \right| \leq \epsilon(n),$$

where $x \odot r = \sum_{i=1}^n x_i r_i \pmod{2}$.

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1. O. Goldreich, L. Levin. A Hard-Core Predicate for All One-Way Functions. STOC 1989.



We call $x \odot r$ the **hard core bit** of the function $xr \mapsto f(x)r$.

Scenario:

- ▶ We know $f(x)$ and that $\mathbb{A}(f(x), r)$ approximates $x \odot r$ to some extent.
- ▶ We hope to recover x .

1. If $\mathbb{A}(f(x), r) = x \odot r$ for all r , then it is easy to recover x by the following algorithm:

- ▶ Run $\mathbb{A}(f(x), e^1), \dots, \mathbb{A}(f(x), e^n)$.
- ▶ Paste the resulting n bits to get x .

2. Suppose $\Pr_{r \in_R \{0,1\}^n}[\mathbb{A}(f(x), r) = x \odot r] \geq 0.9$.

Now $x \odot r$ is uniformly distributed. So by union bound

$$\Pr_{r \in_R \{0,1\}^n}[(\mathbb{A}(f(x), r) \neq x \odot r) \vee (\mathbb{A}(f(x), r \oplus e^i) \neq x \odot (r \oplus e^i))] \leq 0.2.$$

The equality $(x \odot r) \oplus (x \odot (r \oplus e^i)) = x \odot (r \oplus r \oplus e^i) = x \odot e^i = x_i$ implies that

$$\Pr_{r \in_R \{0,1\}^n}[\mathbb{A}(f(x), r) \oplus \mathbb{A}(f(x), r \oplus e^i) = x_i] \geq 0.8, \quad (2)$$

which can be amplified to $1 - 1/10n$ by majority vote.

- If we decrease 0.9 to 0.75, then 0.8 goes down to 0.5, rendering the lower bound in (2) utterly useless.

Algorithm **B**: The input is $y = f(x)$.

1. $m := 200n$.
2. Choose $r^1, \dots, r^m \in_{\mathbb{R}} \{0, 1\}^n$.
3. For i from 1 to n do
 - 3.1 $z_1 := \mathbb{A}(f(x), r^1)$, $z'_1 := \mathbb{A}(f(x), r^1 \oplus e^i)$,
 \dots ,
 $z_m := \mathbb{A}(f(x), r^m)$, $z'_m := \mathbb{A}(f(x), r^m \oplus e^i)$.
 - 3.2 guess that x_i is the majority value of $\{z_j \oplus z'_j\}_{j \in [m]}$.

Analysis of \mathbb{B} :

1. Let random variable Z_j be defined by

$$Z_j(r^j) = \begin{cases} 1, & \text{if } \mathbb{A}(f(x), r^j) = x \odot r^j \text{ and } \mathbb{A}(y, r^j \oplus e^i) = x \odot (r^j \oplus e^i), \\ 0, & \text{otherwise.} \end{cases}$$

2. Clearly Z_1, \dots, Z_m are independent. Let $Z = Z_1 + \dots + Z_m$.
3. $E[Z_j] \geq 0.8$ and $E[Z] \geq 0.8m$.
4. $\Pr[|Z - E[Z]| \geq 0.3m] \leq 1/(0.3\sqrt{m})^2$ by Chebychev inequality.
5. It follows from $m = 200n$ that $\Pr[Z \leq 0.5m] \leq 1/10n$.

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- ▶ **Chebychev inequality:** $\Pr[|Z - E[Z]| \geq k\sqrt{\text{Var}(Z)}] \leq 1/k^2$.
 - ▶ $\text{Var}(Z) = \sum_{j=1}^m \text{Var}(Z_j) \leq m$ since $\text{Var}(Z_j) \leq 1$ for all j .

3. Suppose there are constant $c \in \mathbf{N}$ and P-time PTM \mathbb{A} such that

$$\Pr_{x,r \in_R \{0,1\}^n} [\mathbb{A}(f(x), r) = x \odot r] - \frac{1}{2} \geq \frac{1}{n^c}$$

for infinitely many n . There is at least a $\frac{1}{2n^c}$ fragment of x 's, the **good** x 's, such that

$$\Pr_{r \in_R \{0,1\}^n} [\mathbb{A}(f(x), r) = x \odot r] - \frac{1}{2} \geq \frac{1}{2n^c}$$

for infinitely many n 's.

Lemma. Suppose $a_1, a_2, \dots, a_n \in [0, 1]$ and $\rho = (\sum_{i \in [n]} a_i)/n$. There is at least $\frac{\rho}{2}$ fraction of a_i 's such that $a_i \geq \frac{\rho}{2}$.

The point is that we cannot afford to apply \mathbb{A} twice for probabilistic reason.

Instead of calculating $\mathbb{A}(f(x), r^1), \dots, \mathbb{A}(f(x), r^m)$, we **guess** $x \odot r_1, \dots, x \odot r_m$.

But because we are guessing the values of these expressions, we do not need to know x .

- ▶ Choose randomly distinct seeds $s^1, \dots, s^k \in_{\mathbb{R}} \{0, 1\}^n$.
 - ▶ $\{\bigoplus R\}_{R \subseteq \{s^1, \dots, s^k\}}$ are random and **pairwise** independent.
 - ▶ $\{x \odot \bigoplus R\}_{R \subseteq \{s^1, \dots, s^k\}}$ are determined by $x \odot s^1, \dots, x \odot s^k$.
-

We can afford the **exhaustive guessing** if $k = \log m$.

Algorithm **C**: The input is $y = f(x)$.

1. Input $y \in \{0, 1\}^n$. Think of y as $f(x)$ for some x .
2. $m := 10n^{2c+1}$;
3. $k := \log(m)$;
4. Generate $s^1, \dots, s^k \in_{\mathbb{R}} \{0, 1\}^n$;
5. Let R^1, \dots, R^m be subsets of $\{s^1, \dots, s^k\}$ in a canonical way;
6. For each guess $w \in \{0, 1\}^k$ do
 - 6.1 for each $i \in [n]$ do
 - 6.1.1 $x \odot s^1 := w_1, \dots, x \odot s^k := w_k$;
 $z_1 := \bigoplus_{t \in R^1} (x \odot s^t), \dots, z_m := \bigoplus_{t \in R^m} (x \odot s^t)$;
 $z'_1 := \mathbb{A}(y, \bigoplus R^1 \oplus e^i), \dots, z'_m := \mathbb{A}(y, \bigoplus R^m \oplus e^i)$;
 - 6.1.2 guess that x_i is the majority value of $\{z_j \oplus z'_j\}_{j \in [m]}$.
 - 6.2 $x := x_1 \dots x_n$;
 - 6.3 if $f(x) = y$, output x and halt.

Analysis of \mathbb{C} :

1. Let the random variable Z_j be defined by

$$Z_j(r^j) = \begin{cases} 1, & \text{if } \mathbb{A}(y, r^j \oplus e^i) = x \odot (r^j \oplus e^i), \\ 0, & \text{otherwise.} \end{cases}$$

2. Z_1, \dots, Z_m are pairwise independent and $E[Z_j] \geq 1/2 + 1/n^c$.
3. Hence $E[Z] \geq m/2 + m/n^c$, where $Z = Z_1 + \dots + Z_m$.
4. Using $\text{Var}(Z) = \sum_{j=1}^m \text{Var}(Z_j) \leq m$, we derive

$$\begin{aligned} \Pr[|Z - E[Z]| \geq m/n^c] &\leq \Pr[|Z - E[Z]| \geq \frac{\sqrt{m}}{n^c} \sqrt{\text{Var}(Z)}] \\ &\leq \frac{n^{2c}}{m} = \frac{n^{2c}}{10n^{2c+1}} = \frac{1}{10n}. \end{aligned}$$

5. Now $\Pr[Z \leq m/2] \leq \frac{1}{10n}$ follows from 3 and 4.

Theorem. Let f be a one-way **permutation**. The function mapping $x, r \in \{0, 1\}^n$ onto

$$r, f^{n^c}(x) \odot r, f^{n^c-1}(x) \odot r, \dots, f^1(x) \odot r$$

is a pseudorandom generator of stretch $n + n^c$ for every $c \in \mathbf{N}$.

Let \mathbb{A} be a P-time PTM such that for $x, r \in_{\mathbf{R}} \{0, 1\}^n$ and $i \in_{\mathbf{R}} [n^c]$,

$$\Pr[\mathbb{A}(r, f^{n^c}(x) \odot r, f^{n^c-1}(x) \odot r, \dots, f^{i+1}(x) \odot r) = f^i(x) \odot r] - \frac{1}{2} \geq \frac{1}{n^d}$$

for some $d \in \mathbf{N}$ and infinitely many n .

continued on the next slide.

The PTM $\mathbb{B}(y, r)$, where $y, r \in \{0, 1\}^n$, is designed as follows:

1. Generate $i \in_{\mathbb{R}} [n^c]$;
2. Output $\mathbb{A}(r, f^{n^c-i}(y) \odot r, \dots, f^1(y) \odot r, y \odot r)$.

The probability that $\mathbb{B}(f(x), r)$ outputs $x \odot r$ is the same as

$$\Pr[\mathbb{A}(r, f^{n^c}(x) \odot r, f^{n^c-1}(x) \odot r, \dots, f^{i+1}(x) \odot r) = f^i(x) \odot r].$$

Hence

$$\Pr_{x, r \in_{\mathbb{R}} \{0, 1\}^n}[\mathbb{B}(f(x), r) = x \odot r] - \frac{1}{2} \geq \frac{1}{n^d},$$

contradicting to Goldreich-Levin Theorem.

Since f is a permutation $r, f^{n^c-i}(x) \odot r, \dots, f^1(x) \odot r, x \odot r$ is the same distribution as $r, f^{n^c}(x) \odot r, \dots, f^{i+1}(x) \odot r, f^i(x) \odot r$.

one-way function \Leftrightarrow pseudorandom generator \Leftrightarrow unpredictability

Application: Tossing Coin Over Phone

Suppose A and B want to toss a coin over phone. We can apply the following protocol.

1. A chooses $x, r \in_R \{0, 1\}^n$ and sends $(f_n(x), r)$ to B , where f_n is a one-way permutation known to both parties.
 2. B chooses $b \in_R \{0, 1\}$ and sends it to A .
 3. A sends x to B .
 4. A and B agree to use $b \oplus (x \odot r)$.
-

A cannot manipulate the result because it cannot change x .

B cannot manipulate the result because it did not know x .

A can make sure that the result is random as long as x is.

B can make sure that the result is random as long as b is.

Zero Knowledge Proof

In an interactive proof for an **NP** problem, say GI, a certificate usually reveals more than the mere fact that a statement is true.

It turns out that it is possible to design an interactive proof system such that a verifier does not learn anything from interaction apart from the fact that a statement is true.

- ▶ In the following definition, Perfect Zero Knowledge requires that no matter what a verifier learns after participating a proof for a statement x , it could have derived the same thing by itself without participating in any interaction.

1. S. Goldwasser, S. Micali and C. Rackoff The Knowledge Complexity of Interactive Proof Systems. STOC, 186-208, 1985.



Zero Knowledge Proof of **NP** Language

Suppose $L \in \mathbf{NP}$ and \mathbb{M} is a P-time TM such that $x \in L$ if and only if

$$\exists u \in \{0, 1\}^{p(|x|)}. \mathbb{M}(x, u) = 1$$

for some polynomial p .

Zero Knowledge Proof of **NP** Language

A pair \mathbb{P}, \mathbb{V} of interactive P-time PTM's is called a **zero knowledge proof** for L if they enjoy the following properties.

- ▶ **Completeness**. If $\mathbb{M}(x, u) = 1$, then $\Pr[\text{out}_{\mathbb{V}}(\mathbb{P}(x, u), \mathbb{V}(x))] \geq \frac{2}{3}$.
- ▶ **Soundness**. If $x \notin L$, then $\Pr[\text{out}_{\mathbb{V}}(\mathbb{P}^*(x, u), \mathbb{V}(x))] \leq \frac{1}{3}$ for all \mathbb{P}^* and u .
- ▶ **Perfect Zero Knowledge**. For every P-time interactive PTM \mathbb{V}^* there is an **expected** P-time PTM \mathbb{S}^* , called a **simulator**, such that for every $x \in L$ and every certificate u of x , the following holds:

$$\text{out}_{\mathbb{V}^*}(\mathbb{P}(x, u), \mathbb{V}^*(x)) \equiv \mathbb{S}^*(x),$$

meaning that the two random variables are identical even though \mathbb{S}^* does not have any access to u .

The idea of **simulation** to demonstrate security is central to many aspects of cryptography.

Zero Knowledge Proof for Graph Isomorphism

Public Input: G_0, G_1 with n vertices.

P knows: A permutation $\pi \in [n] \rightarrow [n]$ such that $G_1 = \pi(G_0)$.

P sends $H = \pi'(G_1)$ with a random permutation $\pi' \in_R [n] \rightarrow [n]$.

V sends a random bit $b \in_R \{0, 1\}$.

P sends $\pi'' = \begin{cases} \pi', & \text{if } b = 1, \\ \pi'\pi, & \text{if } b = 0. \end{cases}$

V checks if $H = \pi''(G_b)$.

If $G_0 \simeq G_1$, **V** accepts with probability one.

If $G_0 \not\simeq G_1$, **V** rejects with probability $\frac{1}{2}$.

Zero Knowledge Proof for Graph Isomorphism

Let \mathbb{V}^* be some verifier's strategy.

- ▶ If $G_0 \simeq G_1$, then \mathbb{P} 's first message has the same distribution as the message sent by the following simulator \mathbb{S}^* :
 - ▶ Generate $b' \in_{\mathbb{R}} \{0, 1\}$ and $\pi' \in_{\mathbb{R}} [n] \rightarrow [n]$;
 - ▶ Send $H = \pi'(G_{b'})$ to \mathbb{V}^* ;
 - ▶ Get some b from \mathbb{V}^* ;
 - ▶ If $b = b'$ then send π' to \mathbb{V}^* and output whatever \mathbb{V}^* outputs, otherwise restart \mathbb{S}^* .
 - ▶ The key point is that H reveals nothing about b' if $G_0 \simeq G_1$.
-

If \mathbb{V}^* runs in P-time, then \mathbb{S}^* runs in expected P-time.

Secure Multiparty Computation

Ten people working in a firm want to calculate their average salaries without revealing the salary of any of them.

Remark

Cryptography on weaker assumption (say $\mathbf{P} \neq \mathbf{NP}$) ?

Modern cryptography is founded on something not provable.