

The Goldreich-Levin Theorem

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Contents

1	Some Definitions	1
2	Theorem Statement	2
3	The Proof	2
4	Connection with local list decoding of Hadamard Codes	4
5	Applications of the Goldreich-Levin Theorem	5

1 Some Definitions

We recap some basic definitions of cryptography before stating and proving the Goldreich-Levin theorem.

Definition 1 (One-Way Function). *A family of functions $f_n : \{0, 1\}^n \mapsto \{0, 1\}^{k(n)}$ are called one-way functions if they are computable in polynomial time and for every non-uniform PPT adversary \mathcal{A} ,*

$$\Pr_{x \leftarrow \{0,1\}^n} (f_n(\mathcal{A}(f_n(x))) = f_n(x)) = \text{negligible}(n)$$

where $\text{negligible}(n)$ is a function which decays super-polynomially with n .

Definition 2 (Hard-Core Predicate). *A predicate $h : \{0, 1\}^* \rightarrow \{0, 1\}$ is called a hard-core predicate for a one-way function $f : \{0, 1\}^n \mapsto \{0, 1\}^{k(n)}$ if h is computable in polynomial time and for every non-uniform PPT adversary \mathcal{A}*

$$\Pr_{x \leftarrow \{0,1\}^n} (\mathcal{A}(1^n, f(x)) = h(x)) = \frac{1}{2} + \text{negligible}(n)$$

We extensively referred to [1], [2], and [3] for our report.

2 Theorem Statement

The Goldreich-Levin theorem goes as follows:

Theorem 2.1 (Goldreich-Levin Theorem). *Let f be a one-way function with domain $\{0, 1\}^n$. Note that for any $r \in \{0, 1\}^n$, $g(x, r) := (f(x), r)$ is a one-way function too. Then $h(x, r) := \langle x, r \rangle$ is a hard-core predicate for g , where $\langle x, r \rangle$ denotes the dot product of x and r (in \mathbb{F}_2).*

3 The Proof

We proceed via contradiction: Consider a PPT adversary which can guess the hardcore bit with non-negligible probability over $\frac{1}{2}$. We shall construct a PPT adversary which can invert f with non-negligible probability. However establishing the theorem requires some lemmata, which we shall now prove.

Lemma 3.1. *Let \mathcal{A} be any PPT adversary, let $\delta > 0$. Define*

$$G_{\mathcal{A}, \delta} := \left\{ x : \Pr_{r \leftarrow \{0, 1\}^n} (\mathcal{A}(f(x), r) = \langle x, r \rangle) \geq \frac{1 + \delta}{2} \right\}$$

If $\Pr_{x, r \leftarrow \{0, 1\}^n} (\mathcal{A}(f(x), r) = \langle x, r \rangle) \geq \frac{1}{2} + \delta$, then $\Pr_{x \leftarrow \{0, 1\}^n} (x \in G_{\mathcal{A}, \delta}) \geq \frac{\delta}{2}$.

Proof. Note that

$$\begin{aligned} \Pr_{x, r \leftarrow \{0, 1\}^n} (\mathcal{A}(f(x), r) = \langle x, r \rangle) &= \Pr_{x, r \leftarrow \{0, 1\}^n} (\mathcal{A}(f(x), r) = \langle x, r \rangle | x \in G_{\mathcal{A}, \delta}) \Pr_{x \leftarrow \{0, 1\}^n} (x \in G_{\mathcal{A}, \delta}) \\ &\quad + \Pr_{x, r \leftarrow \{0, 1\}^n} (\mathcal{A}(f(x), r) = \langle x, r \rangle | x \notin G_{\mathcal{A}, \delta}) \Pr_{x \leftarrow \{0, 1\}^n} (x \notin G_{\mathcal{A}, \delta}) \\ &\leq 1 \cdot \Pr_{x \leftarrow \{0, 1\}^n} (x \in G_{\mathcal{A}, \delta}) + \frac{1 + \delta}{2} \cdot 1 \end{aligned}$$

Since $\Pr_{x, r \leftarrow \{0, 1\}^n} (\mathcal{A}(f(x), r) = \langle x, r \rangle) \geq \frac{1}{2} + \delta$, we get our desired result. \square

Lemma 3.2. *Let $X_1, X_2, \dots, X_{m'}$ be pairwise independent Bernoulli random variables with parameter p . Define $X := \sum_{i=1}^{m'} X_i$. Then*

$$\Pr(|X - \mathbb{E}[X]| \geq m'\delta) \leq \frac{1}{4m'\delta^2}$$

Proof. Denote by μ the value of $\mathbb{E}[X]$.

Note that

$$\begin{aligned} \text{Var}(X) &= \mathbb{E}[(X - \mu)^2] = \mathbb{E}[X^2 - 2\mu X + \mu^2] \\ &= \mathbb{E}[X^2] - 2\mu\mathbb{E}[X] + \mu^2 = \mathbb{E} \left[\sum_{i=1}^{m'} X_i^2 + 2 \sum_{1 \leq i < j \leq m'} X_i X_j \right] - 2\mu\mathbb{E}[X] + \mu^2 \end{aligned}$$

$$= \sum_{i=1}^{m'} \mathbb{E}[X_i^2] + 2 \sum_{1 \leq i < j \leq m'} \mathbb{E}[X_i X_j] - 2\mu^2 + \mu^2$$

Since X_i, X_j are pairwise independent for $i \neq j$, $\mathbb{E}[X_i X_j] = \mathbb{E}[X_i] \mathbb{E}[X_j] = p^2$. Moreover, $\mathbb{E}[X_i^2] = p$. Consequently

$$\text{Var}(X) = \sum_{i=1}^{m'} p + 2 \sum_{1 \leq i < j \leq m'} p^2 - \mu^2 = m'p(1 - p)$$

where the last equality follows since $\mu = m'p$.

The desired result then follows by invoking Chebyshev's inequality and noting that $p(1 - p) \leq \frac{1}{4}$ for every $p \in [0, 1]$. \square

Now, let $\delta = \text{non-negligible}(n) > 0$ be the advantage of our adversary \mathcal{A} in calculating the hardcore bit, ie:- $\Pr_{x, r \leftarrow \{0,1\}^n}(\mathcal{A}(f(x), r) = \langle x, r \rangle) \geq \frac{1}{2} + \delta$. Set $m := \lceil \frac{2n}{\delta^2} \rceil, k := 1 + \lceil \log_2(m) \rceil$. Uniformly choose k random vectors t_1, t_2, \dots, t_k from $\{0, 1\}^k$. Now, let $S \subseteq \{1, 2, \dots, k\} =: [k]$ be any non-empty set. Then we define r_S as $r_S := \sum_{i \in S} t_i$. This way we can generate $2^k - 1 = m' \geq m$ random vectors. Note that all the vectors r_S are themselves distributed uniformly in $\{0, 1\}^n$ since a linear combination of uniform random vectors from $\{0, 1\}^n$ is itself a uniform random vector ¹.

Note that for any two sets $S_1 \neq S_2$, r_{S_1}, r_{S_2} are independent. Consequently, all our m' random vectors are pairwise independent.

Now assume we already know the correct values of $\langle x, t_i \rangle$ for every $i \in [k]$. Then we know the values $\langle x, r_S \rangle$ for every $S \subseteq [k]$, since $\langle x, r_S \rangle = \langle x, \sum_{i \in S} t_i \rangle = \sum_{i \in S} \langle x, t_i \rangle$.

Let e_i be the i^{th} unit vector of $\{0, 1\}^n$. For any $S \subseteq [k]$, since r_S are uniformly random, we get that $r_S \oplus e_i$ is uniformly random too. Moreover, note that $\langle x, r_S \oplus e_i \rangle - \langle x, r_S \rangle = \langle x, e_i \rangle = x_i$.

Consequently, for every $S \subseteq [k]$, calculate the value of $\mathcal{A}(f(x), r_S \oplus e_i) - \langle x, r_S \rangle$, where \mathcal{A} is the adversary calculating the hardcore bit, obtain m' votes for the value of x_i , and take the *majority vote* of these values ².

Let ξ_S be the Bernoulli random variable denoting the probability distribution of $\mathcal{A}(f(x), r_S \oplus e_i)$ correctly calculating $\langle x, r_S \oplus e_i \rangle$. If $x \in G_{\mathcal{A}, \delta}$, then the parameter of ξ_S is at least $\frac{1+\delta}{2}$, by the definition given in [Lemma 3.1](#).

Consequently, the expected number of correct answers in the m' votes for the value of x_i is at least $\frac{m'(1+\delta)}{2}$, and thus if the majority vote turns up the wrong answer, that implies a deviation from the mean of more than $\frac{m'\delta}{2}$. By [Lemma 3.2](#), the probability of this happening is at most $\frac{1}{m'\delta^2} \leq \frac{1}{m\delta^2} \leq \frac{1}{2n}$.

Consequently, the probability that any bit is calculated wrongly is at most $\frac{1}{2n}$, which implies, by the union bound, that the probability that x is determined wrongly is at most $\frac{1}{2n} \cdot n = \frac{1}{2}$. Note that x is simply determined by

¹this can be seen through induction

²since $m' = 2^k - 1$ is an odd number, a tie is not possible

a concatenation of the bits x_i for $i \in [n]$.

Consequently, we managed to invert $f(x)$ with probability $\geq \frac{1}{2} \cdot \Pr(x \in G_{\mathcal{A}, \delta}) \geq \frac{\delta}{4}$. However since δ is not negligible, neither is $\frac{\delta}{4}$, which implies that with non-negligible probability we can invert $f(x)$, violating the assumption that it was a one-way function.

We still have to deal with one small catch: We assumed that we know $\langle x, t_i \rangle$ for every $i \in [n]$. But obviously, that is not true *a priori*. We deal with this as follows: We run the aforementioned algorithm for all $2^k = m' + 1 = \text{poly}(n)$ possible values of $(\langle x, t_i \rangle)_{i \in [k]}$. Every time, we end up with a possible value of x , whose correctness we test for by checking if applying $f(x)$ is the correct answer. Since we know that for the correct values of $(\langle x, t_i \rangle)_{i \in [k]}$, we obtain the correct value of x with probability at least $\frac{1}{2}$, we can consequently conclude that we will get the correct answer with probability at least $\frac{1}{2}$ by the end of all the 2^k iterations.

The above step blows up our runtime by 2^k , but since 2^k is polynomial in n , our algorithm remains polynomial time, and thus our overall construction of a PPT adversary continues to hold.

4 Connection with local list decoding of Hadamard Codes

The construction used to generate $2^k - 1$ pairwise independent random vectors is very similar to the concept of local list decoding for Hadamard codes: For any $x \in \{0, 1\}^n$, the Hadamard encoding of x , denoted $\text{Had}(x)$ is defined as $\text{Had}(x) := (\langle x, y \rangle)_{y \in \{0, 1\}^n} \in \{0, 1\}^{2^n}$.

In the context of the Goldreich-Levin theorem, the reason why the Hadamard code is so important is because it is $(q, \delta, \varepsilon) = (2, \frac{1}{4}, 0)$ -locally decodable: What this means is that if y is a noisy/corrupted version of $\text{Had}(x)$ such that $\|y - \text{Had}(x)\|_1 \leq \delta \cdot n = \frac{n}{4}$, then sampling just 2-bits of y allows us to recover any bit of x with probability at least $\frac{1}{2} + \varepsilon = \frac{1}{2}$.

The recovery technique of the above local decoding is exactly same as how we obtained the i^{th} bit of x in the proof of the Goldreich-Levin theorem: For a uniformly random $r \in \{0, 1\}^n$, sample the bit of y corresponding to $\langle x, r \rangle =: y_r$. Then x_i can be computed as $y_r \oplus y_{r \oplus e_i}$, where e_i is the standard i^{th} basis vector, and moreover this calculation is correct with probability $\geq \frac{1}{2} + \frac{1}{2} - 2\delta = \frac{1}{2}$, as claimed.

Thus the proof of Goldreich-Levin theorem is quite commonly referred to in literature as being equivalent to the list decoding of the Hadamard code.

5 Applications of the Goldreich-Levin Theorem

One of the most immediate and useful applications of this theorem is to construct *pseudo-random generators* (PRGs): Indeed, let $f : \{0, 1\}^n \mapsto \{0, 1\}^n$ be a one-way permutation. Then $g(x, r) = f(x) \parallel r \parallel \langle x, r \rangle$ is a pseudo-random generator³. Indeed, through this construction, the Goldreich-Levin theorem lays the foundation for constructing a large class of PRGs.

This construction can be easily extended to a stretch of polynomial length. Indeed,

Theorem 5.1. *If f is a one-way permutation, then*

$$g_N(x, r) := r \parallel \langle f^N(x), r \rangle \parallel \langle f^{(N-1)}(x), r \rangle \parallel \dots \parallel \langle f(x), r \rangle \parallel \langle x, r \rangle$$

is a PRG for any $N \sim \text{poly}(n)$, and f^k denotes the k -fold composition of f .

Proof. We know that pseudorandomness is equivalent to a next-bit prediction by Yao's theorem.

Now assume for the sake of contradiction that g is not a PRG: Then there would exist $i \in [N]$ and a PPT adversary \mathcal{A} such that

$$\Pr(\mathcal{A}(r \parallel \langle f^N(x), r \rangle \parallel \langle f^{N-1}(x), r \rangle \parallel \dots \parallel \langle f^{i+1}(x), r \rangle)) = \langle f^i(x), r \rangle) = \frac{1}{2} + \varepsilon$$

We describe a PPT adversary \mathcal{B} such that given $(f(z), r)$, \mathcal{B} tells us the value of $\langle z, r \rangle$ with non-negligible probability, thus violating the Goldreich-Levin theorem.

\mathcal{B} chooses an $i \in [N]$ randomly. Consider $x \in \{0, 1\}^n$ such that $f^i(x) = z$ ⁴. Note that for $\ell \geq 1$, \mathcal{B} can efficiently calculate $f^{i+\ell}(x) = f^{\ell-1}(f^i(x))$. Consequently, \mathcal{B} can, in polynomial time, generate the string $r \parallel \langle f^N(x), r \rangle \parallel \dots \parallel \langle f^{i+1}(x), r \rangle$ on its own and feed it to \mathcal{A} as an input, which would then return to \mathcal{B} the value of $\langle z, r \rangle$ with non-negligible probability, allowing \mathcal{B} to violate the Goldreich-Levin theorem. \square

References

- [1] Arora and Barak. *Computational Complexity: A Modern Approach*. 2007. URL: <https://theory.cs.princeton.edu/complexity/book.pdf>.
- [2] Omkant Pandey. *Hard Core Predicates*. 2017. URL: <https://www3.cs.stonybrook.edu/~omkant/L05-short.pdf>.
- [3] Omkant Pandey. *Proof of GL Theorem*. 2017. URL: <https://www3.cs.stonybrook.edu/~omkant/L06.pdf>.

³this can be proved through the equivalence of the definitions of pseudo-randomness and next-bit unpredictability

⁴Such a x must necessarily exist since the composition of two permutations is also a permutation, and consequently every element in our co-domain has a (unique) pre-image