

1 Construction of PRG from OWF

Today, we continue and finish the proof of the following Theorem by Goldreich and Levin. Let's recall and continue.

Theorem 1. Construction of Hard-Core Predicate for Permutation

Suppose f is an one way permutation, and we sample $x, r \leftarrow \{0, 1\}^n$, and we construct:

$$f'(x, r) := f(x) \| r$$

$$h(x, r) := x \odot r = \sum_{i=1}^n x_i \cdot r_i \pmod{2}$$

then f' is an one way permutation, while h is a hard-core predicate with respect to f' .

Proof. We first check f' . It is a permutation because $(f')^{-1}(a, b) = (f^{-1}(a), b)$. Then f' is OWF, it is straightforward that it is easy to compute. For the analysis of hardness to invert, it is similar to the proof of Theorem ??, note that we substitute the sampling of a uniformly random bit $b \xleftarrow{\$} \{0, 1\}$ to $b' \xleftarrow{\$} \{0, 1\}^n$ and everything follows similarly. It remains to check that h is a hard-core predicate, we proceed by contradiction. Suppose h is not a hard core predicate, then by definition there exists an NUPPT A and a polynomial p such that for infinitely many $n \in \mathbb{N}$:

$$\Pr_{x,r}[A(f'(x, r)) = h(x, r)] \geq \frac{1}{2} + \frac{1}{p(n)}$$

We want to construct an adversary B that can invert f . Firstly we consider the perfect case, which is:

$$\Pr_{x,r}[A(f'(x, r)) = h(x, r)] = 1$$

In this case we observe that it means $\forall x, r$, we have $A(f(x) \| r) = x \odot r$. For example we have:

$$A(f(x) \| 10 \cdots 0) = x_1 \cdots x_n \odot 10 \cdots 0 = x_1$$

thus in general, this applies to the case when for the strings e_1, \dots, e_n , for $e_i = 0, \dots, 1, \dots, 0$, it means i^{th} bit is 1, while all other bits are zero. Then for B , we can just simply run $A(y \| e_i) = x_i$ for n times, and we can see that B will always successfully invert f , which is a contradiction. Then we relax the probability of a little bit, say:

$$\Pr_{x,r}[A(f'(x, r)) = h(x, r)] = \frac{3}{4} + \frac{1}{p(n)}$$

Then $B(y)$ will be $\forall i \in \{1, \dots, n\}$, we will get $z_i \leftarrow A(y \| e_i)$, and we repeat this procedure n times, we get $z_1 \cdots z_n$ and hopefully this will be x . However, the overall successful probability for this is low, thus this will not work as we wished. Now we replace:

$$z_i \leftarrow A(y \| r) \oplus A(y \| r \oplus e_i)$$

for uniformly random r , this is well-defined because one can show:

$$(x \odot r) \oplus (x \odot (r \oplus e_i)) = x \odot e_i = x_i$$

and we will succeed. Note that we will have the following probability:

$$\Pr_{x,r}[z_i = x_i] \geq \frac{1}{2} + 2\frac{1}{p(n)}$$

following from the constructions above and by union bound. However, we note that repeat this procedure for another n round is not satisfactory enough.

Now, we want to ask this question: can this procedure work for all pairs x and y ? The answer is No, because this procedure depends on fixed y , we cannot randomly sample another y , making this probability highly dependent. We make a new claim: there are many “good” x : we define the set:

$$G := \{x \in \{0, 1\}^n : \Pr_r[A(f(x)||r) = x \odot r] \geq \frac{3}{4} + \frac{1}{2p(n)}\}$$

and we claim that the cardinality of this set is $|G| \geq \frac{1}{2p(n)} \cdot 2^n$. We modify our procedure: For $j = 1, \dots, m$:

1. Randomly sample $r \leftarrow \{0, 1\}^n$.
2. We get $z_{ij} \leftarrow A(y||r) \oplus A(y||r \oplus e_i)$. Note that:

$$(x \odot r) \oplus (x \odot (r \oplus e_i)) = x \odot e_i = x_i$$

and

$$\Pr_{x,r}[z_{ij} = x_i] \geq \frac{1}{2} + \frac{1}{p(n)}$$

3. Note that z_{ij} is independent from $j \in [m]$, and we just take z_i to be the majority:

$$z_i \leftarrow \text{maj}(z_{i1}, \dots, z_{im})$$

We consider the event $\mathbb{X}_j :=$ the event $z_{ij} = x_i$, and the expectation is:

$$\mu = \mathbb{E}[\mathbb{X}_j = 1] = \frac{1}{2} + \frac{1}{p(n)}$$

and we have

$$\Pr\left[\sum_{j \in [m]} \leq \frac{m}{2}\right] = (1 - \delta) \cdot m \cdot \mu \leq e^{-\Omega(m \cdot \alpha^2)}$$

by Chernoff bound. Which indicates if we take $m = n \cdot \alpha^2$, everything will follow. Observe that in this reduction, we have two failing cases:

1. $x \notin G$.
2. x_i fails in the above procedure.

Thus our total failure probability will be:

$$\Pr[\text{failure}] \leq \Pr[x \notin G] + n \cdot e^{-\Omega(n)} = \left(1 - \frac{1}{2p(n)}\right) + \text{negl}$$

so our success probability should be

$$\Pr[B \text{ successfully invert}] \geq \frac{1}{2p(n)} - \text{negl} \geq \frac{1}{\text{poly}(n)}$$

which conclude this case.

Now, we face the next challenge: how can we push the probability from $\frac{3}{4}$ to our desired $\frac{1}{2} + \frac{1}{p(n)}$?

This is not easy, if we take $\frac{3}{5}$ instead of $\frac{3}{4}$ in our procedure, we will have the probability

$$\Pr_{x,r}[z_{ij} = x_i] \geq \frac{1}{5} + \frac{1}{p(n)}$$

which indicates that the union bound fails the majority selection. Before pushing the probability, let's prove the claim regarding

$$G := \{x \in \{0, 1\}^n : \Pr_r[A(f(x)||r) = x \odot r] \geq \frac{3}{4} + \frac{1}{2p(n)}\}$$

We proceed by contradiction, assume $|G| < \frac{1}{2p(n)} \cdot 2^n$, and we denote $A(f(x)||r) = x \odot r$ by event T . Then we have

$$\begin{aligned} \Pr_{x,r}[T] &= \Pr_{x,r}[x \in G] + \Pr_{x,r}[T \mid x \notin G] \cdot \Pr_x[x \notin G] \\ &\leq \frac{|G|}{2^n} + \left(\frac{3}{4} + \frac{1}{2p(n)}\right) \cdot 1 \\ &< \frac{1}{2p(n)} + \frac{3}{4} + \frac{1}{2p(n)} \\ &= \frac{3}{4} + \frac{1}{p(n)} \end{aligned}$$

But we start by $\Pr_{x,r}[T] \geq \frac{3}{4} + \frac{1}{p(n)}$, which is a contradiction. Hence the claim is proved.

Now, we back to the main theorem, we hope the term $A(y||r)$ in $z_{ij} \leftarrow A(y||r) \oplus A(y||r \oplus e_i)$ to be $x \odot r_j$, but we cannot guarantee this anymore. One can check that it is equivalent to just guessing random g_j , we hope $g_j = x \odot r_j$. From now on, we won't call $A(y||r_i)$ anymore.

Now, we are ready to do the final and formal reduction $B(y)$:

1. We sample $u_1, \dots, u_\ell \leftarrow \{0, 1\}^n$ independently, and we get pairwise independent r_1, \dots, r_m by performing bitwise XOR on subsets of (u_1, \dots, u_ℓ) , with $\ell = \log(m)$. We also sample $b_1, \dots, b_\ell \leftarrow \{0, 1\}$ independently, and again we perform bitwise XOR to obtain pairwise independent g_1, \dots, g_m . We hope this will hold:

$$b_k = x \odot u_k \quad \forall k \in [\ell]$$

If this hold, we will have the following probability:

$$\Pr[b_k = x \odot u_k \quad \forall k \in [\ell]] = \frac{1}{2^\ell} = \frac{1}{m}$$

and then we will have $g_j = x \odot r_j$ for all $j \in [m]$, which indicates:

$$\bigoplus_{k \in \text{Set } j} b_k = x \odot \bigoplus_{k \in \text{Set } j} u_k$$

then the following probability hold:

$$\Pr[g_j = x \odot r_j \forall j \in [m]] = \frac{1}{m}$$

2. Then for each i , we compute the following:

$$z_{ij} \leftarrow A(y \| r_j \oplus e_i) \oplus g_j \forall j \in [m]$$

3. We pick the major: $z_i = \text{maj}(z_{i1}, \dots, z_{im})$

4. Output z_1, \dots, z_n .

By choosing $m = \text{poly}(n, \alpha)$, we claim that we can achieve the desired result. We skip the detailed probability analysis, we just provide the sketch here: Suppose $f^{-1}(y) \in G$, and suppose $b_k = x \odot u_k$ for all $k \in [\ell]$, then by the Chebyshev's inequality (Theorem 2 below), we will have

$$\Pr[z_i \neq x_i] \leq O\left(\frac{1}{m}\right)$$

and by the union bound, we will have:

$$\Pr[\exists i \text{ s.t. } z_i \neq x_i] \leq O\left(\frac{n}{m}\right)$$

From these calculations, we see that the event $f^{-1}(y) \in G$ fails with probability $1 - \frac{1}{2p(n)}$, while the event $b_k = x \odot u_k$ for all $k \in [\ell]$ fails with probability $1 - \frac{1}{m}$. Combining everything together, we conclude the proof, the result follows. \square

Theorem 2. Chebyshev's Inequality

For pairwise independent variables $X_1, X_2, \dots, X_m \in [0, 1]$, we have

$$\Pr\left[\left|\sum_{j \in [m]} x_j - m \cdot \mu\right| \geq \delta m\right] \leq \frac{1 - \mu}{m - \delta^2}$$

where $\mu = \mathbb{E}[x_j]$ for all j .

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References