

# Pseudorandom Generators from Regular One-way Functions: New Constructions with Improved Parameters

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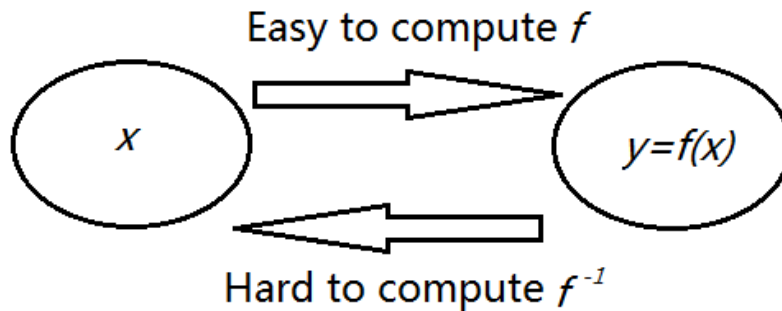
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Joint work with Xiangxue Li and Jian Weng

Asiacrypt 2013

# One-way Functions

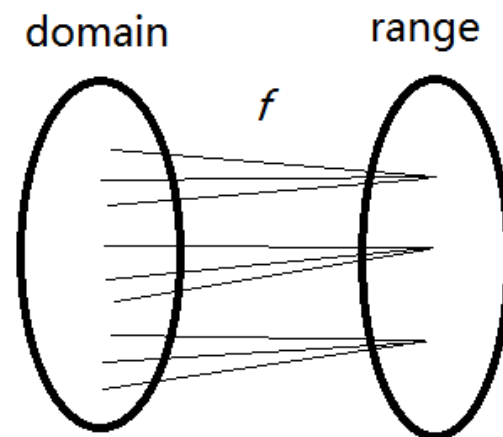
One-way functions are an ensemble of functions  $\{f_n : \{0,1\}^n \rightarrow \{0,1\}^{l(n)}\}_{n \in \mathbb{N}}$  that are



- Simplifying notation :  $f : \{0,1\}^n \rightarrow \{0,1\}^{l(n)}$
- Definition:  $f$  is a  $(t, \varepsilon)$ -one-way function (OWF) if for all adversaries  $A$  of running time  $t$ ,  $\Pr_{y \leftarrow f(U_n)} [A(y) \in f^{-1}(y)] \leq \varepsilon$
- Standard OWF:  $t \in \text{super-poly}$ ,  $\varepsilon \in \text{negl}$
- Folklore: OWFs can be assumed to be length-preserving, i.e.,  $l(n) = n$ .

# Regular Functions

- $f$  is a regular function if for any  $n$  the preimage size  $\alpha = |f^{-1}(y)|$  is fixed (independent of  $y$ ).



- Known-regular function: a regular function  $f$  whose regularity  $\alpha$  is polynomial-time computable from security parameter  $n$ .
- Unknown-regular function: a regular function  $f$  whose regularity  $\alpha$  is inefficient to approximate from security parameter  $n$ .

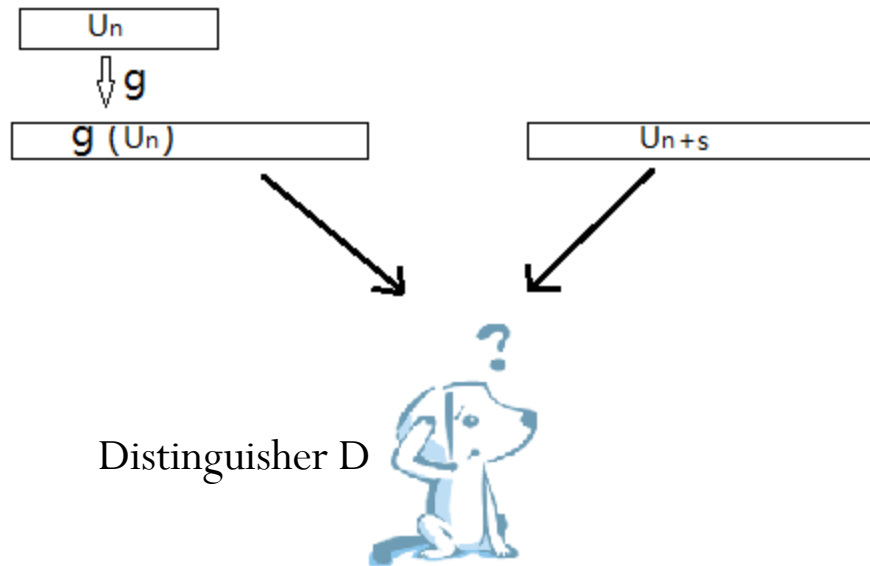
Note: one-way permutation is a special known-regular function.

# Pseudorandom Generators

$g : \{0,1\}^n \rightarrow \{0,1\}^{n+s}$  is a  $(t, \epsilon)$ -pseudorandom generator (PRG) with stretch  $s$  if for all distinguishers  $D$  of running time  $t$ ,

$$|\Pr[D(g(U_n)) = 1] - \Pr[D(U_{n+s}) = 1]| \leq \epsilon$$

$t \in \text{super-poly}$ ,  $\epsilon \in \text{negl}$ ,  $U_n$  is uniform distribution over  $\{0,1\}^n$



# Entropies, computational and statistical distance

collision entropy  $\mathbf{H}_2(X) \stackrel{\text{def}}{=} -\log \sum_x \Pr[X = x]^2$

min-entropy  $\mathbf{H}_\infty(X) \stackrel{\text{def}}{=} -\log (\max_x \Pr[X = x])$

conditional collision entropy  $\mathbf{H}_2(X|Z) \stackrel{\text{def}}{=} -\log ( \mathbb{E}_{z \leftarrow Z} [ \sum_x \Pr[X = x|Z = z]^2 ] )$

conditional min-entropy  $\mathbf{H}_\infty(X|Z) \stackrel{\text{def}}{=} -\log ( \mathbb{E}_{z \leftarrow Z} [ \max_x \Pr[X = x|Z = z] ] )$

*computational distance* between  $X$  and  $Y$

$X$  and  $Y$  are  $(t, \varepsilon)$ -close, denoted by  $\text{CD}_t(X, Y) \leq \varepsilon$ ,

if for every probabilistic distinguisher  $D$  of running time up to  $t$  it holds that

$$| \Pr[D(X) = 1] - \Pr[D(Y) = 1] | \leq \varepsilon$$

*statistical distance* between  $X$  and  $Y$ , denoted by  $\text{SD}(X, Y)$ , is defined by

$$\text{SD}(X, Y) \stackrel{\text{def}}{=} \frac{1}{2} \sum_x |\Pr[X = x] - \Pr[Y = x]| = \text{CD}_\infty(X, Y)$$

shorthand:  $\text{SD}(X, Y|Z) = \text{SD}((X, Z), (Y, Z))$

$\text{CD}_t(X, Y|Z) = \text{CD}_t((X, Z), (Y, Z))$

# Leftover Hash Lemma

## **leftover hash lemma**

*For any integers  $d < k \leq n$ , there exists a (polynomial-time computable) universal hash function family  $\mathcal{H} \stackrel{\text{def}}{=} \{h : \{0,1\}^n \rightarrow \{0,1\}^{k-d}\}$  such that for any joint distribution  $(X, Z)$  where  $X \in \{0,1\}^n$  and  $\mathbf{H}_2(X|Z) \geq k$ , we have*

$$\text{SD}(H(X), U_{k-d} \mid H, Z) \leq 2^{-\frac{d}{2}}$$

*where  $H$  is uniformly distributed over the members of  $\mathcal{H}$ , the description size of  $H$  is called seed length, and  $d$  is called entropy loss, i.e., the difference between the entropy of  $X$  (given  $Z$ ) and the number of bits that were extracted from  $X$ .*

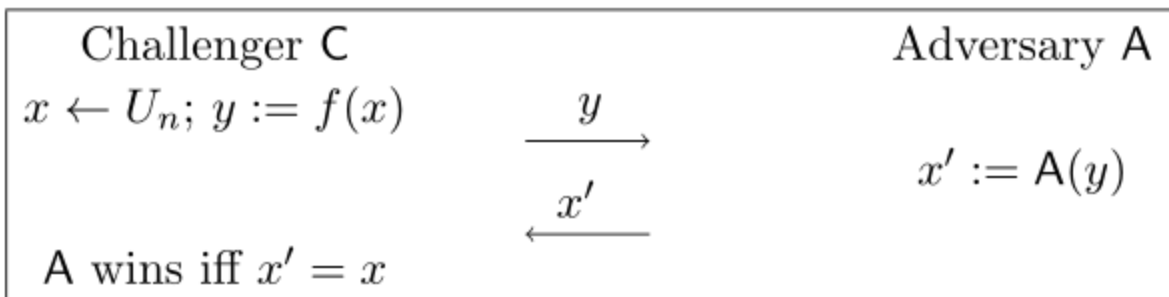
Informally: universal hash functions are good randomness extractors

# Unpredictability Pseudoentropy (UP)

**Definition 2.5 (unpredictability pseudo-entropy)** For distribution ensemble  $(X, Z)$ , we say that  $X$  has  $k$  bits of pseudo-entropy conditioned on  $Z$  for all  $t$ -time adversaries, denoted by  $\mathbf{H}_t(X|Z) \geq k$ , if for any  $n \in \mathbb{N}$  and any probabilistic adversary  $\mathbf{A}$  of running time  $t$

$$\Pr_{(x,z) \leftarrow (X,Z)} [\mathbf{A}(z) = x] \leq 2^{-k}$$

Alternatively, we say that  $X$  is  $2^{-k}$ -hard to predict given  $Z$  for all  $t$ -time adversaries.



The interactive game between  $\mathbf{A}$  and  $\mathbf{C}$  that defines unpredictability pseudo-entropy, where  $x \leftarrow U_n$  denotes sampling a random  $x \in \{0, 1\}^n$ .

# Goldreich-Levin Theorem

**Goldreich-Levin Theorem :** For  $(X, Y) \in \{0, 1\}^n \times \{0, 1\}^*$ , and for any integer  $m \leq n$ , there exists a function family  $\mathcal{H}_C \stackrel{\text{def}}{=} \{h_c : \{0, 1\}^n \rightarrow \{0, 1\}^m\}$  of description size  $\Theta(n)$ , such that

- If  $Y = f(X)$  for any  $(t, \varepsilon)$ -OWF  $f$  and  $X$  uniform over  $\{0, 1\}^n$ , then we have

$$\text{CD}_{t'}(H_C(X), U_m \mid Y, H_C) \in O(2^m \cdot \varepsilon) . \quad (1)$$

- If  $X$  is  $\varepsilon$ -hard to predict given  $Y$  for all  $t$ -time adversaries, namely,  $\mathbf{H}_t(X|Y) \geq \log(1/\varepsilon)$ , then we have

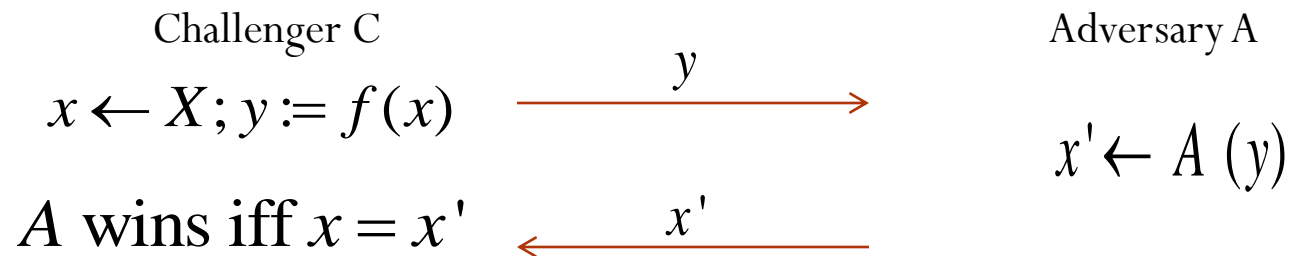
$$\text{CD}_{t'}(H_C(X), U_m \mid Y, H_C) \in O(2^m \cdot (n \cdot \varepsilon)^{\frac{1}{3}}) . \quad (2)$$

where  $t' = t \cdot (\varepsilon/n)^{O(1)}$  and function  $H_C$  is uniformly distributed over the members of  $\mathcal{H}_C$ .



## A Key Observation about Unpredictability Pseudoentropy

- Unpredictability Pseudoentropy (UP) :  $X$  has  $m$  bits of UP given  $f(X)$  for  $t$ -time adversaries if every  $A$  of running time  $t$  wins the following game with probability no greater than  $2^{-m}$

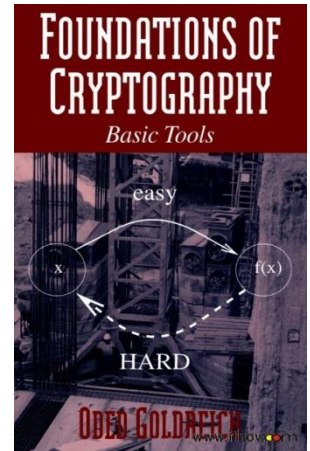


- Question: what's the UP of  $X$  given  $f(X)$  if  $f$  is a  $(t, \varepsilon)$ -regular OWF with  $|f^{-1}(y)| = 2^k$  ?
- Observation:  $X$  given  $f(X)$  has  $k + \log(1/\varepsilon)$  bits of UP.
- Rationale:  $\Pr[A(f(X)) \in f^{-1}(f(X))] \leq \varepsilon \implies \Pr[A(f(X)) = X] \leq 2^{-k} \cdot \varepsilon$

# The FIRST CONSTRUCTION (from known-regular OWF)

- $g(X, h_1, h_2, h_c) = (h_1(f(X_1)), h_2(X_1), h_c(X_1), h_1, h_2, h_c)$

A complicated proof by Goldreich in Section 3.5.2 of



# PRGs from Known-Regular OWFs by three extractions (a three-line proof)

- Assumption:  $f$  is  $(t, \varepsilon)$ -one-way and  $2^k$ -regular, i.e.  $|f^{-1}(y)| = 2^k$
- Construction and Proof.

1.  $H_\infty(f(X)) = n - k$   extract  $(n - k)$  bits using  $h_1$

2.  $H_\infty(X | f(X)) = k$   extract  $k$  bits using  $h_2$

3. chain rule:

$H_{\text{up}}^t(X | f(X)) = k + \log(1/\varepsilon)$    $H_{\text{up}}^t(X | f(X), h_2(X)) \geq \log(1/\varepsilon)$

extract  $O(\log(1/\varepsilon))$  bits using hard-core function  $h_c$

- This completes the proof for the folklore construction, i.e.  $g(X, h_1, h_2, h_c) = (h_1(f(X_1)), h_2(X_1), h_c(X_1), h_1, h_2, h_c)$  is a PRG.
- Parameters: seed length linear in  $n$ , and a single call to  $f$ .

# Tightening the security bounds

- $g(x, h_1, h_2, h_c) = (h_1(f(x)), h_2(x), h_c(x), h_1, h_2, h_c)$

The proof for 3<sup>rd</sup> extraction: consider  $f'(x, h_2) = (f(x), h_2(x), h_2)$

$\because x$  is  $\varepsilon$ -hard to predict given  $f'(x, h_2)$ , i.e.  $H_{up}^t(X | f'(X, H_2)) \geq \log(1/\varepsilon)$

$\therefore$  by Goldreich-Levin Thm,  $h_c(x)$  is  $2^m (n \cdot \varepsilon)^{1/3}$ -close to  $U_m$  given  $f'(x, h_2)$

- A tighter approach (use the tight version of Goldreich-Levin)?

if  $f'$  is an  $\varepsilon'$ -hard OWF, then  $h_c(x)$  is  $(2^m \cdot \varepsilon')$ -close to  $U_m$  given  $f'(x, h_2)$

1. Goldreich show  $\varepsilon' = O(\varepsilon^{1/5})$  in [Gol01, vol-1]

2. We show  $\varepsilon' = 3\sqrt{\varepsilon}$  against  $t$ -time adversaries

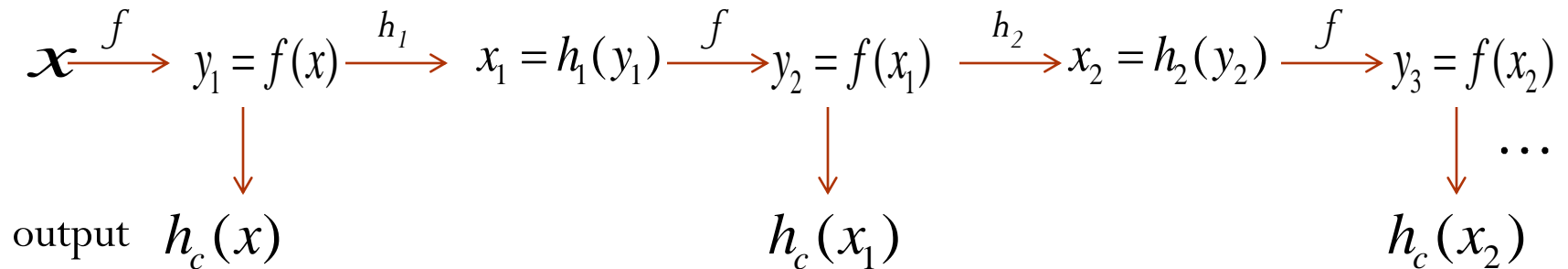
the idea: show  $f'$  is almost 1-to-1, i.e.  $H_2(f'(X, H_2) | H_2) \geq n - 1$

# The Second Construction

(NEW, improving the Randomized Iterate)

# The Randomized Iterate

- Goldreich, Krawczyk and Luby (SICOMP 93) :  
PRGs from known regular OWFs with seed length  $O(n^3)$
- Haitner, Harnik and Reingold (CRYPTO 2006):  
PRGs from unknown regular OWFs with seed length  $O(n \cdot \log n)$



$h_1, h_2, \dots$  are random pairwise independent hash,  $h_c$  is hard-core function

# Lower bounds by Holenstein and Sinha (FOCS12)

- Asymptotic setting: Any black-box construction of PRG must make  $\Omega(n/\log n)$  calls to an arbitrary (including unknown regular) OWF.
- Concrete setting : Any black-box construction of PRG must make  $\Omega(n/\log(1/\varepsilon))$  calls to an arbitrary (including unknown regular)  $(\varepsilon^{-1}, \varepsilon)$ -secure OWF.

# PRGs from unknown-regular OWFs: a new construction

- Assumption:  $f$  is  $(t, \varepsilon)$ -one-way and  $2^k$ -regular (  $k$  is unknown).
- The goal: a PRG construction oblivious of  $k$ .
- The idea: transform  $f$  into a known-regular OWF  $\bar{f}$

$$f : \{0,1\}^n \rightarrow \mathcal{Y}, \text{ where } \mathcal{Y} \subseteq \{0,1\}^n$$

$$\text{define } \bar{f} : \mathcal{Y} \times \{0,1\}^n \rightarrow \mathcal{Y}$$

$$\bar{f}(y, r) = f(y \oplus r)$$

where  $\oplus$ : "bitwise XOR",  $y \leftarrow f(U_n)$ ,  $r \leftarrow U'_n$

1.  $\bar{f}$  is also a  $(t, \varepsilon)$ -one-way function
2.  $\bar{f}$  is a  $2^n$ -regular function, i.e.  $|\bar{f}^{-1}(y, r)| = 2^n$  regardless of  $k$



# PRGs from unknown-regular OWFs: a new construction (cont'd)

- Given a one-way function with known pre-image size  $2^n$   

$$\overline{f} : \mathcal{Y} \times \{0,1\}^n \rightarrow \mathcal{Y}$$
- Similarly,  $(Y, R)$  has  $n + \log(1/\varepsilon)$  bits of UP given  $\overline{f}(Y, R)$ .
- We get a special PRG  $\overline{g} : \mathcal{Y} \times \{0,1\}^n \rightarrow \mathcal{Y} \times \{0,1\}^{n+\Theta(\log(1/\varepsilon))}$
- Done?

No,  $n$  bits needed to sample from  $\mathcal{Y}$  (i.e.  $f(U_n)$ )

stretch :  $-n + \Theta(\log(1/\varepsilon)) + \Theta(\log(1/\varepsilon)) + \Theta(\log(1/\varepsilon)) \dots$

To make it positive: iterate  $\overline{g}$

- In summary: a PRG from unknown regular OWF with linear seed length (hybrid argument) and  $\Theta(n/\log(1/\varepsilon))$  OWF calls.
- Tight (Holenstein and Sinha, FOCS 2012): BB construction of PRG requires  $\Omega(n/\log(1/\varepsilon))$  OWF calls, and  $\Omega(n/\log n)$  calls in general.

# Summary

- PRG from any known-regular  $\epsilon$ -hard OWF:  
seed length  $\tilde{O}(m)$  and a  $\tilde{O}(m)$  calls to the underlying OWF
- PRG from any unknown-regular  $\epsilon$ -hard OWF :  
seed length  $\tilde{O}(n)$  and  $\tilde{O}(m/\log(1/\epsilon))$  OWF calls

Question: remove the dependency on  $\epsilon$  ?

Yes, by paying a factor  $\omega(1)$  in seed length and number of calls.

Why? Due to the entropy loss of the Leftover Hash Lemma.

Given 1-to-1 OWF  $f : \{0,1\}^n \rightarrow \{0,1\}^{n+1}$  (without knowing  $\epsilon$  )

Run  $q = \omega(1)$  copies of  $f$ , extracting  $2 \log n$  hardcore bits per copy,  
followed by a single extraction with entropy loss set to  $q \cdot \log n$  .

# More details

Full version at eprint <http://eprint.iacr.org/2013/270>

Thank you!

