## MIT 6.875

## Foundations of Cryptography Lecture 7

## Recap + Today

Define one-way functions (OWF).

Define Hardcore bits (HCB).

Show that one-way functions* $+\mathrm{HCB} \Rightarrow$ PRG

Goldreich-Levin Theorem: "every OWF has a HCB."

## Recap + Today

Define one-way functions (OWF).

Define Hardcore bits (HCB).

Show that one-way functions* $+\mathrm{HCB} \Rightarrow \mathrm{PRG}$

Goldreich-Levin Theorem: for every OWF/OWP F, there is another OWF/OWP $F^{\prime}$ which has a HCB.

## Goldreich-Levin (GL) Theorem: Version 1

Let $\left\{B_{r}:\{0,1\}^{n} \rightarrow\{0,1\}\right\}$ where

$$
B_{r}(x)=\langle r, x\rangle=\sum_{i=1}^{n} r_{i} x_{i} \bmod 2
$$

be a collection of predicates (one for each $r$ ). Then, for every one-way function $F$, a random $B_{r}$ is hardcore. That is, for every one-way function F, every PPT A, there is a negligible function $\mu$ s.t.
$\operatorname{Pr}\left[x \leftarrow\{0,1\}^{n} ; r \leftarrow\{0,1\}^{n}: A(F(x), r)=B_{r}(x)\right] \leq \frac{1}{2}+\mu(n)$

## GL Theorem: Version 2

For every one-way function
$F$, there is a related one-way function

$$
F^{\prime}(x, r)=(F(x), r)
$$

which has a deterministic hardcore predicate. In particular, the predicate $B(x, r)=\langle r, x\rangle \bmod 2$ is hardcore for $F^{\prime}$.
$\operatorname{Pr}\left[x \leftarrow\{0,1\}^{n} ; r \leftarrow\{0,1\}^{n}: A\left(F^{\prime}(x, r)\right)=\langle r, x\rangle\right] \leq \frac{1}{2}+\mu(n)$

## Key Point:

This statement is sufficient to construct PRGs from any OWP.

## If there are OWPs, then there are PRGs

## CONSTRUCTION

Let $F$ be a one-way permutation, then $G$ defined below is a PRG.

Then, define $G(x, r)=\mathrm{F}^{\prime}(\mathrm{x}, \mathrm{r})\|\langle r, x\rangle=\mathrm{F}(x)\| r \|\langle r, x\rangle$.

Theorem: $G$ is a PRG assuming $F$ is a one-way permutation.

## We proved a weaker version in L6:

Let's assume a pretty good predictor $P$

$$
\operatorname{Pr}\left[x \leftarrow\{0,1\}^{n} ; r \leftarrow\{0,1\}^{n}: P(F(x), r)=\langle r, x\rangle\right] \geq \frac{\mathbf{3}}{\mathbf{4}}+1 / p(n)
$$

Then there is a OWF inverter $A$.

$$
\operatorname{Pr}\left[x \leftarrow\{0,1\}^{n}: A(F(x)) \in F^{-1}(F(x))\right] \geq 1 / p^{\prime}(n)
$$

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First, we used an averaging argument.
Claim: For at least a $1 / 2 p(n)$ fraction of the $x$,

$$
\operatorname{Pr}\left[r \leftarrow\{0,1\}^{n}: P(F(x), r)=\langle r, x\rangle\right] \geq \frac{3}{4}+1 / 2 p(n)
$$

Call these the good $x$.
Proof: On the board.

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$$

## Key Idea: Linearity

Pick a random $r$ and ask $P$ to tells us $\langle r, x\rangle$ and $\left\langle r+e_{i}, x\right\rangle$. Subtract the two answers to get $\left\langle e_{i}, x\right\rangle=x_{i}$.

Proof: $\operatorname{Pr}\left[\right.$ we compute $x_{i}$ correctly]
$\geq \operatorname{Pr}\left[\mathrm{P}\right.$ predicts $\langle r, x\rangle$ and $\left\langle r+e_{i}, x\right\rangle$ correctly]
$=1-\operatorname{Pr}\left[\mathrm{P}\right.$ predicts $\langle r, x\rangle$ or $\left\langle r+e_{i}, x\right\rangle$ wrong $]$
$\geq 1-(\operatorname{Pr}[\mathrm{P}$ predicts $\langle r, x\rangle$ wrong] + $\operatorname{Pr}\left[\mathrm{P}\right.$ predicts $\left\langle r+e_{i}, x\right\rangle$ wrong]) (by union bound)
$\geq 1-2 \cdot\left(\frac{1}{4}-\frac{1}{2 p(n)}\right)=\frac{1}{2}+1 / p(n)$

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For at least a $1 / 2 p(n)$ fraction of the $x$,

$$
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$$

Inverter A:
Repeat for each $i \in\{1,2, \ldots, n\}$ :
Repeat $\mathrm{O}\left(\log n(p(n))^{2}\right)$ times:
Pick a random $r$ and ask $P$ to tells us $\langle r, x\rangle$ and $\left\langle r+e_{i}, x\right\rangle$. Subtract the two answers to get a guess for $x_{i}$.

Compute the majority of all such guesses and set the bit as $x_{i}$ Output the concatenation of all $x_{i}$ as $x$.

Analysis: Chernoff + Union Bound

## Who's the culprit here?

For at least a $1 / 2 p(n)$ fraction of the $x$,

$$
\operatorname{Pr}\left[r \leftarrow\{0,1\}^{n}: P(F(x), r)=\langle r, x\rangle\right] \geq \frac{3}{4}+1 / 2 p(n)
$$

Pick a random $r$ and ask $P$ to tells us $\langle r, x\rangle$ and $\left\langle r+e_{i}, x\right\rangle$. Subtract the two answers to get $\left\langle e_{i}, x\right\rangle=x_{i}$.

Proof: $\operatorname{Pr}\left[\right.$ we compute $x_{i}$ correctly]
$\geq \operatorname{Pr}\left[\mathrm{P}\right.$ predicts $\langle r, x\rangle$ and $\left\langle r+e_{i}, x\right\rangle$ correctly]
$=1-\operatorname{Pr}\left[\mathrm{P}\right.$ predicts $\langle r, x\rangle$ or $\left\langle r+e_{i}, x\right\rangle$ wrong $]$
$\geq 1-(\operatorname{Pr}[$ P predicts $\langle r, x\rangle$ wrong $]+$ $\operatorname{Pr}\left[\mathrm{P} \text { predicts }\left\langle r+e_{i}, x\right\rangle \text { wrong] }\right)_{(\text {by union bound) }}$
$\geq 1-2 \cdot\left(\frac{1}{4}-\frac{1}{2 p(n)}\right)=\frac{1}{2}+1 / p(n)$

## The Real Proof of the GL Theorem

 (attributed to Charlie Rackoff)Assume (after averaging) that for $\geq 1 / 2 p(n) \mathrm{f}$ $\operatorname{Pr}\left[r \leftarrow\{0,1\}^{n}: P(F(x), r)=\langle r, x\rangle\right] \geq \frac{1}{2}+1 /$

For a minute, assume we have a bit of help/ad
Pick a random $r$, ask the Oracle to tells us $\langle r, x\rangle$ and ask $P$ to tell us $\left\langle r+e_{i}, x\right\rangle$. Subtract the two answers to get $\left\langle e_{i}, x\right\rangle=x_{i}$.

Proof: $\operatorname{Pr}$ [we compute $x_{i}$ correctly]
$\geq \operatorname{Pr}\left[\mathrm{P}\right.$ predicts $\left\langle r+e_{i}, x\right\rangle$ correctly $] \geq \frac{1}{2}+1 / 2 p(n)$

## The Real Proof of the GL Theorem

Assume (after averaging) that for $\geq 1 / 2 p(n) \mathrm{f}$

$$
\operatorname{Pr}\left[r \leftarrow\{0,1\}^{n}: P(F(x), r)=\langle r, x\rangle\right] \geq \frac{1}{2}+1 /
$$

Pick a random $r$, guess $\langle r, x\rangle$ and ask $P$ to tell us $\left\langle r+e_{i}, x\right\rangle$. Subtract the two to get $\left\langle e_{i}, x\right\rangle=x_{i}$.

If our guesses are all correct, then the analysis works out just as before.

But what's the chance...?
The number of $r^{\prime} \mathrm{s}$ is $m=\mathrm{O}\left(n \log n(p(n))^{2}\right)$.

## Parsimony in Guessing

Pick random "seed vectors" $s_{1}, \ldots, s_{\log (m+1)}$, and guess $c_{j}=\left\langle s_{j}, x\right\rangle$ for all $j$.
The probability that all guesses are correct is $\frac{1}{2^{\log (m+1)}}=1 /(m+1)$ which is not bad.

From the seed vectors, generate many more $r_{i}$.
Let $T_{1}, \ldots, T_{m}$ denote all possible non-empty subsets of
$\{1,2, \ldots, \log (m+1)\}$. We will let

$$
r_{i}=\bigoplus_{j \in T_{i}} s_{j} \quad \text { and } \quad b_{i}=\bigoplus_{j \in T_{i}} c_{j}
$$

Key Observation: If the guesses $c_{1}, \ldots, c_{\log (m+1)}$ are all correct, then so are the $b_{1}, \ldots, b_{m}$.

## The OWF Inverter

Generate random $s_{1}, \ldots, s_{\log (m+1)}$ and bits $c_{1}, \ldots, c_{\log (m+1)}$.
From them, derive $r_{1}, \ldots, r_{\log (m+1)}$ and bits $b_{1}, \ldots, b_{m}$ as in the previous slide.

Repeat for each $i \in\{1,2, \ldots, n\}$ :
Repeat $100 \boldsymbol{n}(\boldsymbol{p}(\boldsymbol{n}))^{2}$ times:
Ask $P$ to tells us $\left\langle r_{i}+e_{i}, x\right\rangle$. XOR P's reply with $b_{i}$ to get a guess for $x_{i}$.

Compute the majority of all such guesses and set the bit as $x_{i}$ Output the concatenation of all $x_{i}$ as $x$.

## Analysis of the Inverter

Let's condition on the guesses $c_{1}, \ldots, c_{\log (m+1)}$ being all correct.

The main issue: The $r_{i}$ are not independent (can't do Chernoff)

Key Observation: The $r_{i}$ are pairwise independent.
Therefore, can apply Chebyshev!

We have that
$p:=\operatorname{Pr}[$ Inverter succeeds $\mid$ all guesses correct, $\operatorname{good} \mathrm{x}] \geq 0.99$.
(Pf. on the board, also in the next two slides)

## Putting it all together

$\operatorname{Pr}[$ Inverter succeeds]
$\geq \operatorname{Pr}[$ Inverter succeeds | all guesses correct, good $x]$. $\operatorname{Pr}[$ all guesses correct $] \cdot \operatorname{Pr}[\operatorname{good} x]$
$=\frac{1}{m+1} \cdot \frac{1}{2 p(n)} \cdot p$
$=\frac{1}{2 n^{2} p(n)^{3}} \cdot p$
So, it suffices to show that $p$ is large.

By our calculation (on the board), $p \geq 0.99$, so we are done.

Can also make the success probability $\approx 1 / p(n)$ by enumerating over all the "guesses". Each guess results in a supposed inverse, but we can check which of them is the actual inverse!

## The Coding-Theoretic View of GL

$x \rightarrow(\langle x, r\rangle)_{r \in\{0,1\}^{n}}$ can be viewed as a highly redundant, exponentially long encoding of $x=$ the Hadamard code.
$P(F(x), r)$ can be thought of as providing access to a noisy codeword.

What we proved:

- unique decoding algorithm for Hadamard code with error rate $\frac{1}{4}-1 / p(n)$.
- list-decoding algorithm for Hadamard code with error rate $\frac{1}{2}-1 / p(n)$.


## Hardcore Predicates from any

## List-Decodable Code

(due to Impagliazzo and Sudan)
$x \rightarrow C(x)$ is the encoding.
Given a $C(x)$ that is incorrect at $\frac{1}{2}-\varepsilon$ fraction of the locations, a list-decoder outputs a list $\left\{x_{1}, \ldots, x_{m}\right\}$ of possibilities for $x$.

The hardcore predicate is

$$
B_{i}(x)=C(x)_{i}
$$

A hardcore-bit predictor gives us access to a corrupted codeword. Running the list-decoder on it gives us the list of possible inverses. The fact that the OWF is easy to compute means that we can filter out the bogus (non-)inverses.

## Recap

l. Defined one-way functions (OWF).
2. Defined Hardcore bits (HCB).
3. Goldreich-Levin Theorem: every OWF has a HCB. (showed proof for an important special case)
4. Show that one-way permutations (OWP) $\Rightarrow$ PRG
(in fact, one-way functions $\Rightarrow$ PRG, but that's a much harder theorem)

## Universal Hardcore Predicate Conjecture 1

For every one-way function $F$,
there exists a circuit $B_{F}$ s.t.
for every PPT Circuit/Turing Machine A, there is a negligible function $\mu$ s.t.

$$
\operatorname{Pr}\left[x \leftarrow\{0,1\}^{n}: A(F(x))=B_{F}(x)\right] \leq \frac{1}{2}+\mu(n)
$$

In fact: I conjecture that for every one-way function $F$, there exists an $r_{F}$ for which the predicate $B_{r_{F}}(x)=\left\langle r_{F}, x\right\rangle$ that is hardcore.


## Universal Hardcore Predicate Conjecture 2

For every one-way function $F$, there is an efficiently generatable circuit $B_{F}$ s.t. for every PPT Circuit/Turing Machine A, there is a negligible function $\mu$ s.t.

$$
\operatorname{Pr}\left[x \leftarrow\{0,1\}^{n}: A(F(x))=B_{F}(x)\right] \leq \frac{1}{2}+\mu(n)
$$



## Other Topics (Time permitting)

1. $\mathrm{OWF} \Rightarrow \mathrm{PRG}$ ?
2. Pseudorandom Permutations from Pseudorandom Functions (the Luby-Rackoff construction)

## Minicrypt:



Candidate Constructions: from number theory, geometry, combinatorics,...

