

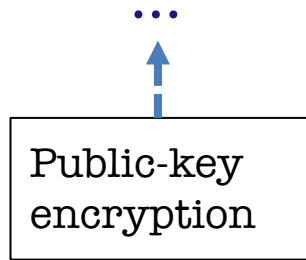
MIT 6.875 & Berkeley CS276

Foundations of Cryptography
Lecture 3

Roadmap of the Course: Worlds in Crypto

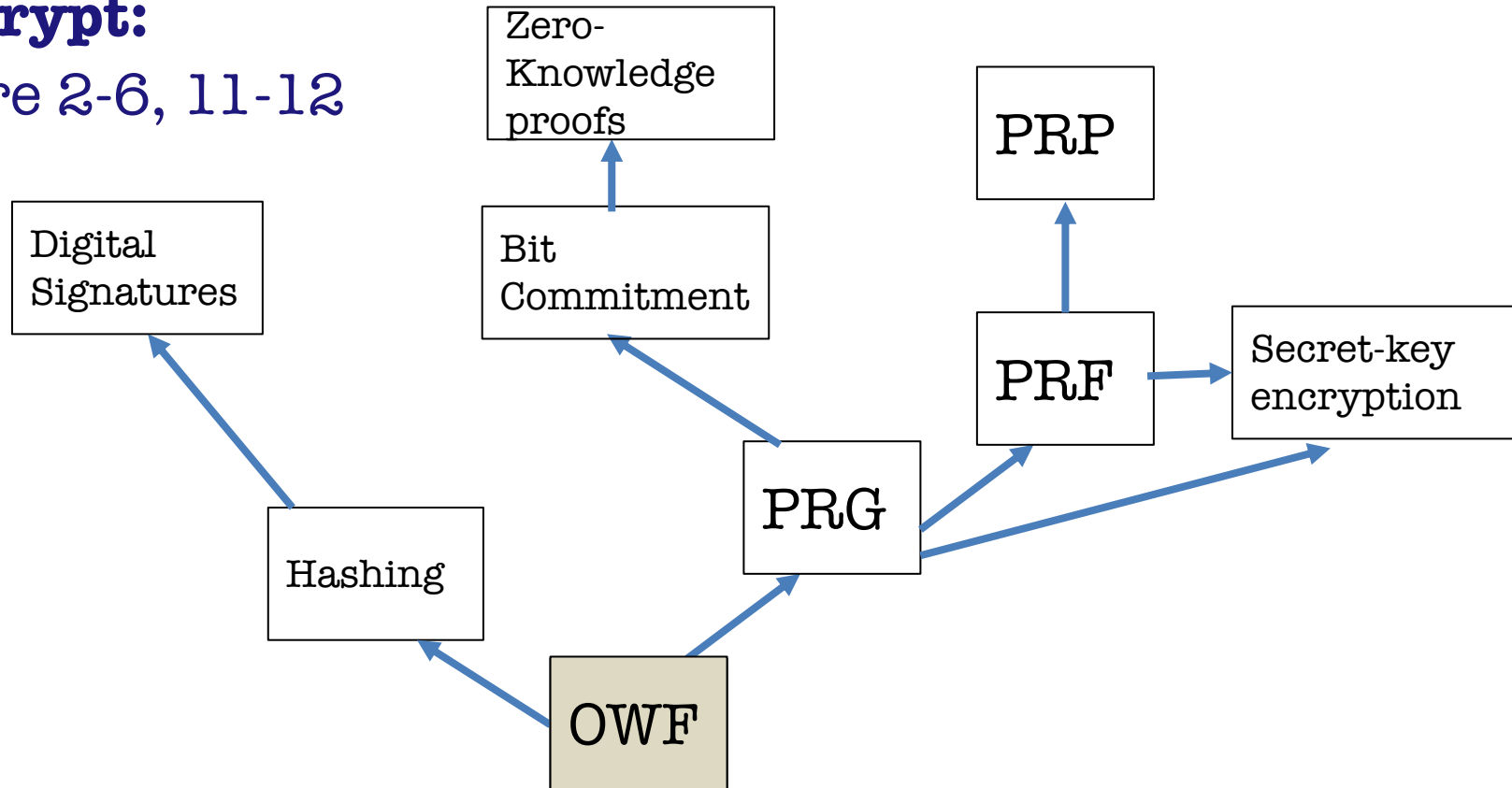
Cryptomania:

Lecture 7-10,...



Minicrypt:

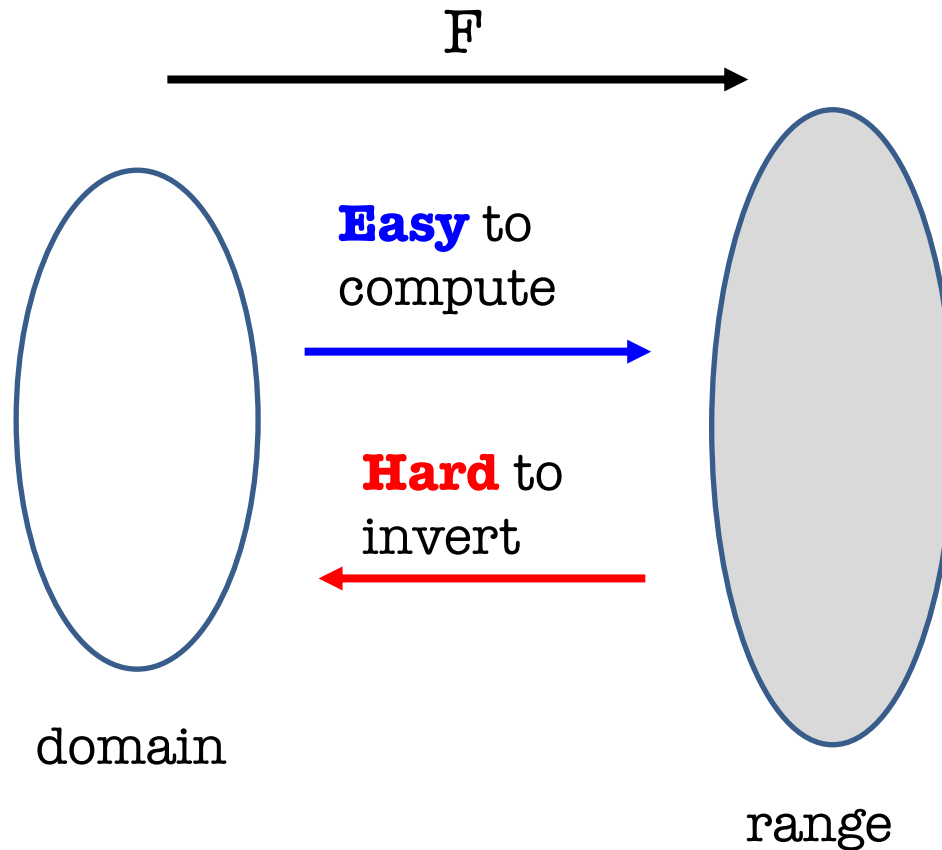
Lecture 2-6, 11-12



Today

1. Define one-way functions (OWF).
2. Define Hardcore bits (HCB).
3. Show that one-way functions* + HCB \Rightarrow PRG
4. Goldreich-Levin Theorem: every OWF has a HCB.

One-way Functions (Informally)



One-way Functions (Take 1)

A function (family) $\{F_n\}_{n \in \mathbb{N}}$ where $F_n: \{0,1\}^n \rightarrow \{0,1\}^{m(n)}$ is one-way if for every p.p.t. adversary A , there is a negligible function μ s.t.

$$\Pr[x \leftarrow \{0,1\}^n; y = F_n(x): A(1^n, y) = x] \leq \mu(n)$$

Consider $F_n(x) = \mathbf{0}$ for all x .

This is one-way according to the above definition.

In fact, impossible to find *the* inverse even if A has unbounded time.

Conclusion: not a useful/meaningful definition.

One-way Functions (Take 1)

A function (family) $\{F_n\}_{n \in \mathbb{N}}$ where $F_n: \{0,1\}^n \rightarrow \{0,1\}^{m(n)}$ is one-way if for every p.p.t. adversary A , there is a negligible function μ s.t.

$$\Pr[x \leftarrow \{0,1\}^n; y = F_n(x): A(1^n, y) = x] \leq \mu(n)$$

The Right Definition: Impossible to find *an* inverse in p.p.t.

One-way Functions: The Definition

A function (family) $\{F_n\}_{n \in \mathbb{N}}$ where $F_n: \{0,1\}^n \rightarrow \{0,1\}^{m(n)}$ is one-way if for every p.p.t. adversary A , there is a negligible function μ s.t.

$$\Pr[x \leftarrow \{0,1\}^n; y = F_n(x); A(1^n, y) = x' : y = F_n(x')] \leq \mu(n)$$

- Can always find *an* inverse with unbounded time
- ... but should be hard with probabilistic polynomial time

One-way Permutations:

One-to-one one-way functions with $m(n) = n$.

One-way Functions: Candidates

Subset sum:

$$G(a_1, \dots, a_n, x_1, \dots, x_n) = (a_1, \dots, a_n, \sum_{i=1}^n x_i a_i \bmod 2^{n+1})$$

where a_i are random n -bit numbers, and x_i are random bits.

One-way functions candidates are abundant in nature.

We will see many other candidates from number theory, coding theory, combinatorics later in class.

Today

1. Define one-way functions (OWF).
2. Define Hardcore bits (HCB).
3. Show that one-way *permutations* (OWP) \Rightarrow PRG
4. Goldreich-Levin Theorem: every OWF has a HCB.

Hardcore Bits

If F is a one-way function, we know it's hard to compute a pre-image of $F(x)$ for a randomly chosen x .

How about computing partial information about an inverse?

Exercise: There are one-way functions for which it is easy to compute the first half of the bits of the inverse.

Hardcore Bits

If F is a one-way function, we know it's hard to compute a pre-image of $F(x)$ for a randomly chosen x .

HARDCORE BIT (Take 1)

Nevertheless, there has to be a hardcore set of hard to invert inputs. Concretely: Does there exist some bit b of $F(x)$ that can't be guessed with probability non-negligibly better than $1/2$?

- Any bit can be guessed correctly w.p. $1/2$
- So, “hard to compute” \rightarrow “hard to guess with probability non-negligibly better than $1/2$ ”

Hardcore Bits

If F is a one-way function, we know it's hard to compute a pre-image of $F(x)$ for a randomly chosen x .

HARDCORE BIT (Take 1)

For any function (family) $F: \{0,1\}^n \rightarrow \{0,1\}^m$, a bit $i = i(n)$ is hardcore if for every p.p.t. adversary A , there is a negligible function μ s.t.

$$\Pr[x \leftarrow \{0,1\}^n; y = F(x): A(y) = x_i] \leq \frac{1}{2} + \mu(n)$$

Does every one-way function have a hardcore bit?

(Hard) Exercise: There are functions that are one-way, yet *every* bit is somewhat easy to predict (say, with probability $\frac{1}{2} + 1/n$).

So, we will generalize the notion of a hardcore “bit”.

Hardcore Bits

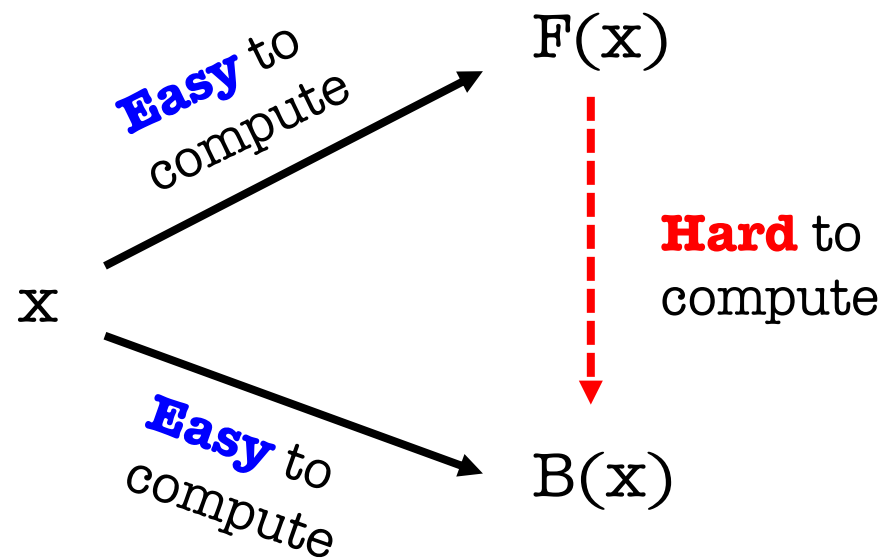
HARDCORE PREDICATE (Definition)

For any function (family) $F: \{0,1\}^n \rightarrow \{0,1\}^m$, a function $B: \{0,1\}^n \rightarrow \{0,1\}$ is a hardcore **predicate** if for every p.p.t. adversary A , there is a negligible function μ s.t.

$$\Pr[x \leftarrow \{0,1\}^n; y = F(x): A(y) = B(x)] \leq \frac{1}{2} + \mu(n)$$

For us, henceforth, a hardcore bit will mean a hardcore predicate.

Hardcore Predicate (in pictures)



Discussion on the Definition

HARDCORE PREDICATE (Definition)

For any function (family) $F: \{0,1\}^n \rightarrow \{0,1\}^m$, a bit $B: \{0,1\}^n \rightarrow \{0,1\}$ is a hardcore **predicate** (HCP) if for every p.p.t. adversary A , there is a negligible function μ s.t.

$$\Pr[x \leftarrow \{0,1\}^n; y = F(x): A(y) = B(x)] \leq \frac{1}{2} + \mu(n)$$

1. Definition of HCP makes sense for *any* function family, not just one-way functions.
2. Some functions can have information-theoretically hard to guess predicates (e.g., compressing functions)
3. We'll be interested in settings where x is uniquely determined given $F(x)$, yet $B(x)$ is hard to predict given $F(x)$

Today

1. Define one-way functions (OWF).
2. Define Hardcore bits (HCB).
3. Show that one-way *permutations* (OWP) \Rightarrow PRG
4. Goldreich-Levin Theorem: every OWF has a HCB.

OWP \Rightarrow PRG

CONSTRUCTION

Let F be a one-way permutation, and B an associated hardcore predicate for F .

Then, define $G(x) = F(x) \parallel B(x)$.

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

(Note that G stretches by one bit. Shafi will tell you how to extend the stretch of G to any poly number of bits.)

OWP \Rightarrow PRG

CONSTRUCTION

Let F be a one-way permutation, and B an associated hardcore predicate for F .

Then, define $G(x) = F(x) \parallel B(x)$.

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

Proof (next slide): From Distinguishing to Predicting.

OWP \Rightarrow PRG

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

Proof: Assume for contradiction that G is not a PRG. Therefore, there is a p.p.t. distinguisher D and a polynomial function p such that

$$\Pr[x \leftarrow \{0,1\}^n; y = G(x): D(y) = 1] - \Pr[y \leftarrow \{0,1\}^{n+1} : D(y) = 1] \geq 1/p(n)$$

Think: D outputs “1” = D thinks its input is pseudorandom

OWP \Rightarrow PRG

Theorem: G is a PRG assuming F is a one-way permutation and B is its hardcore predicate .

Proof: Assume for contradiction that G is not a PRG. Therefore, there is a p.p.t. distinguisher D and a polynomial function p such that

$$\Pr[x \leftarrow \{0,1\}^n; y = G(x): D(y) = 1] - \Pr[y \leftarrow \{0,1\}^{n+1} : D(y) = 1] \geq 1/p(n)$$

We will construct a hardcore predictor A and show:

$$\Pr[x \leftarrow \{0,1\}^n: A(F(x)) = B(x)] \geq \frac{1}{2} + 1/p'(n)$$

OWP \Rightarrow PRG

Theorem: G is a PRG assuming F is a one-way permutation and B is its hardcore predicate .

Proof: Assume for contradiction that G is not a PRG. Therefore, there is a p.p.t. distinguisher D and a polynomial function p such that

$$\Pr[x \leftarrow \{0,1\}^n; y = G(x): D(y) = 1] - \Pr[y \leftarrow \{0,1\}^{n+1} : D(y) = 1] \geq 1/p(n)$$

We will construct a hardcore predictor A and show:

$$\Pr[x \leftarrow \{0,1\}^n: A(F(x)) = B(x)] \geq \frac{1}{2} + 1/p'(n)$$

OWP \Rightarrow PRG

Let's look closely at D.

$$\Pr[x \leftarrow \{0,1\}^n; y = G(x): D(y) = 1] - \Pr[y \leftarrow \{0,1\}^{n+1} : D(y) = 1] \geq 1/p(n)$$

By definition:

$$\Pr[x \leftarrow \{0,1\}^n; y = F(x)|B(x): D(y) = 1] - \Pr[y \leftarrow \{0,1\}^{n+1} : D(y) = 1] \geq 1/p(n)$$

OWP \Rightarrow PRG

Let's look closely at D.

$$\frac{\Pr[x \leftarrow \{0,1\}^n; y = G(x): D(y) = 1]}{\Pr[y \leftarrow \{0,1\}^{n+1} : D(y) = 1]} \geq 1/p(n)$$

A syntactic change:

$$\frac{\Pr[x \leftarrow \{0,1\}^n; y = F(x)|B(x): D(y) = 1]}{\Pr[y_0 \leftarrow \{0,1\}^n, y_1 \leftarrow \{0,1\}, y = y_0|y_1: D(y) = 1]} \geq 1/p(n)$$

OWP \Rightarrow PRG

Let's look closely at D.

$$\Pr[x \leftarrow \{0,1\}^n; y = G(x): D(y) = 1] - \Pr[y \leftarrow \{0,1\}^{n+1} : D(y) = 1] \geq 1/p(n)$$

Rewriting the second term:

$$\Pr[x \leftarrow \{0,1\}^n; y = F(x)|B(x): D(y) = 1] - \Pr[x \leftarrow \{0,1\}^n, y_1 \leftarrow \{0,1\}, y = F(x)|y_1: D(y) = 1] \geq 1/p(n)$$

\equiv

$$\Pr[x \leftarrow \{0,1\}^n, y = F(x)|\mathbf{0}: D(y) = 1] + \Pr[x \leftarrow \{0,1\}^n, y = F(x)|\mathbf{1}: D(y) = 1]$$

OWP \Rightarrow PRG

Let's look closely at D.

$$\Pr[x \leftarrow \{0,1\}^n; y = G(x): D(y) = 1] - \Pr[y \leftarrow \{0,1\}^{n+1} : D(y) = 1] \geq 1/p(n)$$

Rewriting the second term (again):

$$\Pr[x \leftarrow \{0,1\}^n; y = F(x)|B(x): D(y) = 1] - \Pr[x \leftarrow \{0,1\}^n, y_1 \leftarrow \{0,1\}, y = F(x)|y_1: D(y) = 1] \geq 1/p(n)$$

\equiv

$$\frac{\Pr[x \leftarrow \{0,1\}^n, y = F(x)|\mathbf{B}(x): D(y) = 1] + \Pr[x \leftarrow \{0,1\}^n, y = F(x)|\overline{\mathbf{B}(x)}: D(y) = 1]}{2}$$

OWP \Rightarrow PRG

Let's look closely at D.

$$\Pr[x \leftarrow \{0,1\}^n; y = G(x): D(y) = 1] - \Pr[y \leftarrow \{0,1\}^{n+1} : D(y) = 1] \geq 1/p(n)$$

Putting things together:

$$\frac{1}{2} (\Pr[x \leftarrow \{0,1\}^n; y = F(x) | \mathbf{B}(x): D(y) = 1] - \Pr[x \leftarrow \{0,1\}^n, y = F(x) | \overline{\mathbf{B}(x)}: D(y) = 1]) \geq 1/p(n)$$

In English: D says “1” more often when fed with the “right bit” than the “wrong bit”.

OWP \Rightarrow PRG

Let's look closely at D.

$$\Pr[x \leftarrow \{0,1\}^n; y = G(x): D(y) = 1] - \Pr[y \leftarrow \{0,1\}^{n+1} : D(y) = 1] \geq 1/p(n)$$

Putting things together:

$$\left(* \right) \frac{1}{2} \left(\Pr[x \leftarrow \{0,1\}^n; y = F(x) | \mathbf{B}(x): D(y) = 1] - \Pr[x \leftarrow \{0,1\}^n, y = F(x) | \overline{\mathbf{B}(x)}: D(y) = 1] \right) \geq 1/p(n)$$

Now, let's use D to *predict* the right bit.

OWP \Rightarrow PRG

The Predictor A works as follows:

Get as input $z = F(x)$; Pick a random bit b ; and feed D with input $z|b$.

If D says “1”, output b as the prediction for the hardcore bit and if D says “0”, output \bar{b} .

Analysis of the Predictor A

$$\begin{aligned} & \Pr[x \leftarrow \{0,1\}^n: A(F(x)) = B(x)] \\ &= \Pr[x \leftarrow \{0,1\}^n: D(F(x)|b) = 1 \mid b = B(x)] \Pr[b = B(x)] + \\ & \quad \Pr[x \leftarrow \{0,1\}^n: D(F(x)|b) = 0 \mid b \neq B(x)] \Pr[b \neq B(x)] \\ &= \frac{1}{2} (\Pr[x \leftarrow \{0,1\}^n: D(F(x)|b) = 1 \mid b = B(x)] + \\ & \quad \Pr[x \leftarrow \{0,1\}^n: D(F(x)|b) = 0 \mid b \neq B(x)]) \\ &= \frac{1}{2} (\Pr[x \leftarrow \{0,1\}^n: D(F(x)|B(x)) = 1] + \\ & \quad \Pr[x \leftarrow \{0,1\}^n: D(F(x)|\overline{B(x)}) = 0]) \\ &= \frac{1}{2} (\Pr[x \leftarrow \{0,1\}^n: D(F(x)|B(x)) = 1] + \\ & \quad 1 - \Pr[x \leftarrow \{0,1\}^n: D(F(x)|\overline{B(x)}) = 1]) \\ &= \frac{1}{2} (1 + (*)) \geq \frac{1}{2} + 1/p(n) \end{aligned}$$



Today

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A Hardcore Predicate for all OWF

Let's shoot for a *universal* hardcore predicate.

i.e., a single predicate B where it is hard to guess $B(x)$ given $F(x)$

Is this possible?

Turns out the answer is “no”. Pick your favorite amazing B . I claim that you can construct a one-way function F for which B is not hard-core. I will leave it to you as an exercise.

So, what is one to do?

Goldreich-Levin (GL) Theorem

Let $\{B_r: \{0,1\}^n \rightarrow \{0,1\}\}$ 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, a **random** B_r is hardcore for **every** one-way function F . That is, for every one-way function F , every PPT A , there is a negligible function μ s.t.

$$\Pr[x \leftarrow \{0,1\}^n; r \leftarrow \{0,1\}^n: A(F(x), r) = B_r(x)] \leq \frac{1}{2} + \mu(n)$$

Alternative Interpretation 1: For every one-way function F , there is a related one-way function $F'(x, r) = (F(x), r)$ which has a *deterministic* hardcore predicate.

Goldreich-Levin (GL) Theorem

Let $\{B_r: \{0,1\}^n \rightarrow \{0,1\}\}$ where

$$B_r(x) = \langle r, x \rangle = \sum_{i=1}^n r_i x_i \pmod{2}$$

be a collection of predicates (one for each r). Then, a **random** B_r is hardcore for **every** one-way function F . That is, for every one-way function F , every PPT A , there is a negligible function μ s.t.

$$\Pr[x \leftarrow \{0,1\}^n; r \leftarrow \{0,1\}^n: A(F(x), r) = B_r(x)] \leq \frac{1}{2} + \mu(n)$$

Alternative Interpretation 2: For every one-way function F , there **exists** (non-uniformly) a (possibly different) hardcore predicate $\langle r_F, x \rangle$. **(Cool open problem: remove the non-uniformity)**

Proof of GL Theorem

Let's make our lives easier: assume a perfect predictor P

~~Assume for contradiction there is a predictor P~~

$$\Pr[x \leftarrow \{0,1\}^n : \Pr[\{0,1\}^n : A(F(x)) = x' : F(x') = F(x)] \geq \frac{1}{2}] \geq 1/p(n)$$

We will need to show an inverter A for F

$$\Pr[x \leftarrow \{0,1\}^n : A(F(x)) = x' : F(x') = F(x)] \geq 1/p'(n)$$

Proof of GL Theorem

Let's make our lives easier: assume a perfect predictor P

~~Assume for contradiction there is a predictor P~~

$$\Pr[x \leftarrow \{0,1\}^n; r \leftarrow \{0,1\}^n: P(F(x), r) = \langle r, x \rangle] = 1$$

The inverter A works as follows:

On input $y = F(x)$, A runs the predictor P n times, on inputs $(y, e_1), (y, e_2), \dots$, and (y, e_n) where $e_1 = 100\dots 0, e_2 = 010 \dots 0, \dots$ are the unit vectors.

Since A is perfect, it returns $\langle e_i, x \rangle = x_i$, the i^{th} bit of x on the i^{th} invocation.

Proof of GL Theorem

OK, now let's assume less: assume a pretty good predictor P

~~Assume for contradiction there is a predictor P~~

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

First, we need an **averaging argument**.

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

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

Proof: Exercise in counting.

Call these the good x .

Proof of GL Theorem

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

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

Key Idea: Linearity

Pick a random r and ask P to tell us $\langle r, x \rangle$ and $\langle r + e_i, x \rangle$.
Subtract the two answers to get $\langle e_i, x \rangle = x_i$.

Proof: $\Pr[\text{we compute } x_i \text{ correctly}]$

$$\begin{aligned} &\geq \Pr[P \text{ predicts } \langle r, x \rangle \text{ and } \langle r + e_i, x \rangle \text{ correctly}] \\ &= 1 - \Pr[P \text{ predicts } \langle r, x \rangle \text{ or } \langle r + e_i, x \rangle \text{ wrong}] \\ &\geq 1 - (\Pr[P \text{ predicts } \langle r, x \rangle \text{ wrong}] + \\ &\quad \Pr[P \text{ predicts } \langle r + e_i, x \rangle \text{ wrong}]) \quad (\text{by union bound}) \\ &\geq 1 - 2 \cdot \left(\frac{1}{4} - \frac{1}{2p(n)} \right) = \frac{1}{2} + 1/p(n) \end{aligned}$$

Proof of GL Theorem

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

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

Inverter A:

Repeat for each $i \in \{1, 2, \dots, n\}$:

Repeat $\log n / p(n)$ times:

Pick a random r and ask P to tell us $\langle r, x \rangle$ and $\langle r + e_i, x \rangle$.
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

Real Proof (will not do in class)

Assume (after averaging) that for $\geq 1/2p(n)$ fraction of the x ,

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

Key Idea: Pairwise independence

Reference: Goldreich Book Part 1, Section 2.5.2.

<http://www.wisdom.weizmann.ac.il/~oded/PSBookFrag/part2N.ps>

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 $\frac{1}{4} - 1/p(n)$.

The real proof = **list-decoding algorithm** for Hadamard code with error rate $\frac{1}{2} - 1/p(n)$.

Recap

1. 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)

Next Lecture: Back to PRGs