## Lecture 4: Pseudorandomness

## Last Lecture

- Example of Reduction


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- Any n.u. PPT $D$ can distinguish $\left\{X_{n}\right\}$ from $\left\{Y_{n}\right\}$ with only negligible probability


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- Property 1: Closure under Efficient Operations


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- Property 1: Closure under Efficient Operations
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- Property 2: Transitivity (aka, the "Hybrid Lemma")
- If first and last hybrids are comp. distinguishable, then at least a pair of consecutive hybrids are comp. distinguishable


## Distinguishing vs Prediction

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## Definition (Prediction Advantage)

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\max _{\mathcal{A}} \operatorname{Pr}\left[b \stackrel{\$}{\leftarrow}\{0,1\}, t \leftarrow X_{n}^{b}: \mathcal{A}(t)=b\right]-\frac{1}{2}
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- Think: Comp. Indistinguishability $\Rightarrow$ Negl. Prediction Advantage?


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- Think: Comp. Indistinguishability $\Rightarrow$ Negl. Prediction Advantage?
- Think: Comp. Indistinguishability $\Leftarrow$ Negl. Prediction Advantage?


## Distinguishing vs Prediction (contd.)

## Lemma (Prediction Lemma)

Let $\left\{X_{n}^{0}\right\}$ and $\left\{X_{n}^{1}\right\}$ be ensembles of probability distributions. Let $D$ be a n.u. PPT that $\varepsilon(\cdot)$-distinguishes $\left\{X_{n}^{0}\right\}$ and $\left\{X_{n}^{1}\right\}$ for infinitely many $n \in \mathbb{N}$. Then, $\exists$ n.u. PPT $\mathcal{A}$ s.t.

$$
\operatorname{Pr}\left[b \stackrel{\Phi}{\leftarrow}\{0,1\}, t \leftarrow X_{n}^{b}: \mathcal{A}(t)=b\right]-\frac{1}{2} \geqslant \frac{\varepsilon(n)}{2}
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for infinitely many $n \in \mathbb{N}$.

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- Think: Proof?


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Want: Strings that pass all efficient tests

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## Definition (Pseudorandom Ensembles)

An ensemble $\left\{X_{n}\right\}$, where $X_{n}$ is a distribution over $\{0,1\}^{\ell(n)}$, is said to be pseudorandom if:

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\left\{X_{n}\right\} \approx\left\{U_{\ell(n)}\right\}
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- Think: How to show indistinguishability against all efficient tests?


## Next-bit Unpredictability

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## Definition (Next-bit Unpredictability)

An ensemble of distributions $\left\{X_{n}\right\}$ over $\{0,1\}^{\ell(n)}$ is next-bit unpredictable if, for all $0 \leqslant i<\ell(n)$ and n.u. PPT $\mathcal{A}, \exists$ negligible function $\nu(\cdot)$ s.t.:

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\operatorname{Pr}\left[t=t_{1} \ldots t_{\ell(n)} \leftarrow X_{n}: \mathcal{A}\left(t_{1} \ldots t_{i}\right)=t_{i+1}\right] \leqslant \frac{1}{2}+\nu(n)
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## Theorem (Completeness of Next-bit Test)

If $\left\{X_{n}\right\}$ is next-bit unpredictable then $\left\{X_{n}\right\}$ is pseudorandom.

## Next-bit Unpredictability $\Longrightarrow$ Pseudorandomness

$$
H_{n}^{(i)}:=\left\{x \leftarrow X_{n}, u \leftarrow U_{n}: x_{1} \ldots x_{i} u_{i+1} \ldots u_{\ell(n)}\right\}
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- Now, next bit unpredictability is violated


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(3) How to construct PRGs?

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The predicate $h:\{0,1\}^{n} \rightarrow\{0,1\}$ is hardcore for $f:\{0,1\}^{n} \rightarrow\{0,1\}^{m}$ if $\forall$ n.u. PPT adversary $\mathcal{A}, \exists$ negligible function $\mu(\cdot)$ such that:

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- Hardcore Predicate suffices to construct PRG


## Construction Outline: PRG from OWF

- Step 1: OWF (OWP) $\Longrightarrow$ Hardcore Predicate for OWF (OWP)


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- Step 1: OWF (OWP) $\Longrightarrow$ Hardcore Predicate for OWF (OWP)
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- Step 2: Hardcore Predicate for OWF (OWP) $\Longrightarrow$ One-bit stretch PRG
- Step3: One-bit stretch PRG $\Longrightarrow$ Poly-stretch PRG


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## Hardcore Predicate for OWP $\Longrightarrow$ One-bit stretch PRG

- Construction: $G(s)=f(s) \| h(s)$


## Hardcore Predicate for OWP $\Longrightarrow$ One-bit stretch PRG

- Construction: $G(s)=f(s) \| h(s)$
- Think: Proof?


## One-bit stretch PRG $\Longrightarrow$ Poly-stretch PRG

Intuition: Iterate the one-bit stretch PRG poly times
Construction of $G_{\text {poly }}:\{0,1\}^{n} \rightarrow\{0,1\}^{\ell(n)}$ :

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- Let $G_{1}:\{0,1\}^{n} \rightarrow\{0,1\}^{n+1}$ be a one-bit stretch PRG


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- $G_{p o l y}(s):=b_{1} \ldots b_{\ell(n)}$


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- Proof?

