## CS 6815: Pseudorandomness and Combinatorial Constructions

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## 1 Introduction

Recall the following claim from the previous lecture:
Claim 1.1. Suppose $f:\{0,1\}^{n} \rightarrow\{0,1\}$ is $(S, \epsilon)$-hard. Then $G:\{0,1\}^{n} \rightarrow\{0,1\}^{n+1}$ defined as $G(x)=x \circ f(x)$ is an $\left(S^{\prime}, \epsilon^{\prime}\right)$ pseudorandom generator $(P R G)$ with $S^{\prime}=S-1$ and $\epsilon^{\prime}=\epsilon$.

This shows that the assumption of a hard function $f$ allows us to extend $n$ bits to $n+1$ bits. The following idea from Nisan and Wigderson will show the construction of a much better PRG based on the same hardness assumption.

## 2 Nisan-Wigderson

Claim 2.1. Suppose $f:\{0,1\}^{n} \rightarrow\{0,1\}$ is $(S, \epsilon)$-hard. Then there is an $\left(S^{\prime}, \epsilon^{\prime}\right) P R G, G:\{0,1\}^{r} \rightarrow$ $\{0,1\}^{m}$ where $S^{\prime}=S-O\left(m 2^{k}\right)$ and $\epsilon^{\prime}=m \epsilon$.

Before we prove this claim we will need to define ( $n, k$ ) designs which are an essential component of these PRG's. It will be the case that if we can pick better designs then we will get better PRG's.

Definition 2.2. An $(n, k)$ design is a set system $S_{1}, S_{2}, \cdots S_{m} \subseteq[r]$ such that $\left|S_{i}\right|=n$ and $\forall i, j$ where $i \neq j$ we have $\left|S_{i} \cap S_{j}\right| \leq k$ (small intersection) and the size of such a design is $m$.

The amount of sets $m$, and intersection size $k$ will depend on the choice $n$. Also, since the size of the sets is at least $n$ we must have $r \geq n$. With this definition we can now provide a construction of the PRG's claimed in 2.1.

Claim 2.3. Suppose $f:\{0,1\}^{n} \rightarrow\{0,1\}$ is $(S, \epsilon)$-hard. Let $x \in\{0,1\}^{r}$ (a seed). For any $W \subseteq[r]$ let $\left.x\right|_{W}$ denote the projection of $x$ onto coordinates $W$. Let $z \in\{0,1\}^{m}$ be $G(x)$ where $z_{i}=f\left(\left.x\right|_{S_{i}}\right)$ and $S_{1}, \cdots, S_{m}$ form an $(n, k)$ design. That is to say $G(x)=z=f\left(\left.x\right|_{S_{1}}\right) \circ f\left(\left.x\right|_{S_{2}}\right) \circ \cdots \circ f\left(\left.x\right|_{S_{m}}\right)$, the concatenation of $f$ evaluated at $x$ projected onto each of the sets $S_{i}$. Then $G$ is an $\left(S^{\prime}, \epsilon^{\prime}\right) P R G$ with parameters described in Claim 2.1.

Essentially we are using many evaluations of the hard function $f$ to construct a PRG. The runtime of $G$ will then be $m \cdot T(n)$ where $T(n)$ is the runtime of $f(f \in D T I M E(T(n)))$ plus the construction of an $(n, k)$ design. For a dream PRG we would want $m$ to be exponential in $r$ and $r$ polynomial in $n$. Is it reasonable to assume such sets exist? (We shall see a construction in the next class)

We now prove by way of contradiction that $G$ is an $\left(S^{\prime}, \epsilon^{\prime}\right)$ PRG.
Proof. Suppose $G$ is not an $\left(S^{\prime}, \epsilon^{\prime}\right)$ PRG. We will use the hybrid argument to show this is a contradiction. Consider the following sequence of hybrid distributions:

$$
H_{0}=f\left(\left.x\right|_{S_{1}}\right) \circ f\left(\left.x\right|_{S_{2}}\right) \circ \cdots \circ f\left(\left.x\right|_{S_{m}}\right)
$$

$$
\begin{gathered}
H_{1}=b_{1} \circ f\left(\left.x\right|_{S_{2}}\right) \circ \cdots \circ f\left(\left.x\right|_{S_{m}}\right) \\
H_{2}=b_{1} \circ b_{2} \circ \cdots \circ f\left(\left.x\right|_{S_{m}}\right) \\
H_{i}=b_{1} \circ b_{2} \circ \cdots \circ b_{i} \circ f\left(\left.x\right|_{S_{i+1}}\right) \circ \cdots \circ f\left(\left.x\right|_{S_{m}}\right) \\
\vdots \\
H_{m}=b_{1} \circ b_{2} \circ \cdots \circ b_{m}
\end{gathered}
$$

where $x \sim U_{r}$ and $b_{1}, b_{2}, \cdots b_{m} \sim U_{1}$. For any $i$ the distribution replaces the first $i$ function values with uniform random bits. The hybrids go from from $H_{0}$ which is just the distribution of $G(x)$ until $H_{m}$ which is the uniform distribution on $m$ random bits. Due to our assumption that $G$ is not an $\left(S^{\prime}, \epsilon^{\prime}\right)$ PRG, there must exist some distinguisher $D$ that distinguishes between the output of $G$ and the uniform distribution. Rephrasing this in terms of the hybrids, we have $\left|\operatorname{Pr}\left[D\left(H_{m}\right)=1\right]-\operatorname{Pr}\left[D\left(H_{0}\right)=1\right]\right| \geq \epsilon^{\prime}$. We can manipulate this probability into the following telescoping sum

$$
\left|\operatorname{Pr}\left[D\left(H_{m}\right)=1\right]-\operatorname{Pr}\left[D\left(H_{0}\right)=1\right]\right|=\left|\sum_{i=0}^{m-1} \operatorname{Pr}\left[D\left(H_{i+1}\right)=1\right]-\operatorname{Pr}\left[D\left(H_{i}\right)=1\right]\right|
$$

By the triangle inequality we have

$$
\left|\sum_{i=0}^{m-1} \operatorname{Pr}\left[D\left(H_{i+1}\right)=1\right]-\operatorname{Pr}\left[D\left(H_{i}\right)=1\right]\right| \leq \sum_{i=0}^{m-1}\left|\operatorname{Pr}\left[D\left(H_{i+1}\right)=1\right]-\operatorname{Pr}\left[D\left(H_{i}\right)=1\right]\right|
$$

and thus $\epsilon^{\prime} \leq \sum_{i=0}^{m-1}\left|\operatorname{Pr}\left[D\left(H_{i+1}\right)=1\right]-\operatorname{Pr}\left[D\left(H_{i}\right)=1\right]\right|$. Notice that it cannot be the case that all $m$ terms in the summation are less than $\frac{\epsilon^{\prime}}{m}$ or they would not sum to at least $\epsilon^{\prime}$. Therefore, there must be some $i$ for which $\left|\operatorname{Pr}\left[D\left(H_{i+1}\right)=1\right]-\operatorname{Pr}\left[D\left(H_{i}\right)=1\right]\right| \geq \frac{\epsilon^{\prime}}{m}$. Therefore, by assuming that there was a distinguisher for $H_{0}$ and $H_{m}$ with $\epsilon^{\prime}$, we necessarily have a distinguisher between at least one $H_{i}$ and $H_{i+1}$ with $\frac{\epsilon^{\prime}}{m}$.

We will now consider a randomized algorithm $\mathcal{A}$ (which can eventually into a circuit). We will show that under conditions on $\frac{\epsilon^{\prime}}{m}$, we can construct an $\epsilon$ distinguisher from $\mathcal{A}$ which will contradict the fact that $f$ is $(S, \epsilon)$-hard.

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Algorithm 1
Require: \(x, b^{\prime}\)
    \(z \leftarrow 0^{r}\) (an \(r\)-length string of 0's)
    \(j \leftarrow 1\)
    for \(j \leq r\) do
        if \(j \in S_{i+1}\) then \(z_{j}=x_{j}\)
        else \(z_{j} \sim U_{1}\)
        end if
    end for
    Sample \(b_{1}, b_{2}, \cdots, b_{i} \sim U_{1}\)
    Output \(D\left(b_{1} \circ b_{2} \circ \cdots \circ b_{i} \circ b^{\prime} \circ f\left(\left.z\right|_{S_{i+2}}\right) \circ f\left(\left.z\right|_{S_{i+3}}\right) \circ \cdots, f\left(\left.z\right|_{S_{m}}\right)\right)\)
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For this algorithm $\mathcal{A}$, we are given a seed $x$ and a bit $b^{\prime}$ and $\mathcal{A}$ has knowledge of both $i$ (the value for which $\left.\left|\operatorname{Pr}\left[D\left(H_{i+1}\right)=1\right]-\operatorname{Pr}\left[D\left(H_{i}\right)=1\right]\right| \geq \frac{\epsilon^{\prime}}{m}\right)$ and an ( $n, k$ ) design. Importantly this bit $b^{\prime}$ is either chosen uniformly at random or $b^{\prime}=f(x)$. If $b^{\prime}$ is a random bit then over all choices
of random bits and values of $x$ we see that $b_{1} \circ b_{2} \circ \cdots \circ b_{i} \circ b^{\prime} \circ f\left(\left.z\right|_{S_{i+2}}\right) \circ f\left(\left.z\right|_{S_{i+3}}\right) \circ \cdots, f\left(\left.z\right|_{S_{m}}\right)$ is the distribution $H_{i+1}$. Similarly, over all choices of random bits and values of $x$, we see that if $b^{\prime}=f(x)$ then the distribution is $H_{i}$. Since $D$ is an $\frac{\epsilon^{\prime}}{m}$ distinguisher this means that

$$
\underset{b_{1}, \cdots, b_{i},\left.z\right|_{\overline{S_{i+1}}}}{\operatorname{Pr}} \operatorname{Pr}_{x}[\mathcal{A}(x, f(x))=1]-\underset{b_{1}, \cdots, b_{i},\left.z\right|_{\overline{S_{i+1}}}}{\operatorname{Pr}} \operatorname{Pr}\left[\mathcal{A}\left(x, b^{\prime}\right)=1\right] \geq \frac{\epsilon^{\prime}}{m}
$$

Where $\left.z\right|_{\overline{S_{i+1}}}$ is the set of uniformly sampled bits (the indices which are not in $S_{i+1}$ ). Using similar reasoning to the previous result with sums, we know that if the total probability of this is greater than $\frac{\epsilon^{\prime}}{m}$, then there must exist some values $b_{1}, \cdots, b_{i}, Z_{\bar{S}_{i}}$, for which $\operatorname{Pr}_{x}[\mathcal{A}(x, f(x))=$ $1]-\operatorname{Pr}_{x}\left[\mathcal{A}\left(x, b^{\prime}\right)=1\right] \geq \frac{\epsilon^{\prime}}{m}$. Therefore, this algorithm constructs an $\frac{\epsilon^{\prime}}{m}$ distinguisher for $f$ and if we set $\epsilon \geq \frac{\epsilon^{\prime}}{m}$ then this an $\epsilon$ distinguisher. It remains to find $S$ which is the size of a circuit made by derandomizing algorithm $\mathcal{A}$. Notice that the we must calculate $D$ during the algorithm which means we need a circuit of size $S^{\prime}$. We also need $r-n$ random bits as inputs to fully construct $z$ and depending on $i$ we might need at most another $m$ bits for $b_{1}, b_{2}, \cdots$. Knowledge of $i$ can also be counted to take $\log (m)$ bits. Finally, we make at most $m$ function calls to $f$ so this adds size $m \cdot c(f)$ size to the circuit where $c(f)$ is the circuit size needed to compute $f$. However, by the property of the $(n, k)$ design, each set $S_{j}$ has an intersection of size at most $k$ with $S_{i+1}$. Therefore, computing $f\left(\left.z\right|_{S_{j}}\right)$ depends only on at most $k$ bits. Therefore, we can include a lookup table in our circuit for these calculations which has size $2^{k}$. Thus the total size of the circuit is $S^{\prime}+O\left(m 2^{k}\right)$. If we let $S^{\prime}+O\left(m 2^{k}\right)<S$ then we have constructed a size less than $S$ circuit which $\epsilon$ approximates $f$. This contradicts the hardness assumption on $f$. Hence, $G$ must be an $\left(S-O\left(m 2^{k}\right), m \epsilon\right)$ PRG.

We also include the following lemma which relates PRG's to hard languages.
Lemma 2.4. Suppose $L \subset\{0,1\}^{*}$ is in DTIME (T(n)) and is $(S, \epsilon)$-hard. Then there exists a $\operatorname{PRG}\left\{G_{n}\right\}_{n \geq 1}$ where $G_{n}:\{0,1\}^{r(n)} \rightarrow\{0,1\}^{m(n)}$ that is $\left(S-O\left(m(n) \cdot 2^{k(n)}\right), m(n) \cdot \epsilon\right)$ hard.

