

## Lecture 26

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This lecture discusses the relationship between pseudorandomness and hard functions. The main result, accompanied by proof, is Theorem 3.

First, let's review the definition of a useful construction from the previous lecture:

**Definition 1** A collection of sets  $S_1, \dots, S_m \subseteq [d] = \{1, \dots, d\}$  is an  $(l, a)$ -**design** if

- $\forall i, |S_i| = l$
- $\forall i \neq j, |S_i \cap S_j| \leq a$ .

Note that if  $a = 0$ , then the sets  $S_1, \dots, S_l$  are all disjoint as they each have  $l$  elements,  $d$  is forced to be at least  $m \cdot l$ . For the purposes of this lecture, it is useful to think of  $m$  as being big and  $a$  relatively small.

We will use the following theorem, which we don't prove here:

**Theorem 1** For any constant  $\gamma$ , there exists an  $(l, a)$ -design with  $a = \gamma \log m$ , constructible in time  $2^{O(d)}$  and such that  $d = O(l^2/a)$ .

We now introduce another definition

**Definition 2**  $f : \{0, 1\}^l \rightarrow \{0, 1\}$  is  $(t, \alpha)$ -**average case hard** if for any nonuniform (circuit with advice) algorithm  $A$  running in time  $t(l)$  the following inequality holds for large  $l$ :

$$\Pr_{x,A} [A(x) = f(x)] < 1 - \alpha(l)$$

Note that  $x$  is of size  $l$ . We will use  $\alpha(l) = 1 - \epsilon(l)$  for  $\epsilon(l) \leq \frac{1}{t(l)}$ , hence  $1 - \alpha(l) \leq \frac{1}{2} + \epsilon(l) \leq \frac{1}{2} + \frac{1}{t(l)}$ .

The following theorem allows us to extend by 1-bit:

**Theorem 2** If  $f$  is  $(t, 1 - \epsilon)$ -average case hard, then  $G(y) := y \circ f(y)$  is a  $(t, \epsilon)$ -PRG.

We want to stretch this. Our approach is to use the *Nisan-Wigderson generator*, which we present here.

**Definition 3 (Nisan-Wigderson generator)** Given  $(l, a)$ -design  $S_1, \dots, S_m \subseteq [d]$ , define  $G : \{0, 1\}^d \rightarrow \{0, 1\}^m$  to be

$$G(x) := f(x|_{S_1}) \circ f(x|_{S_2}) \circ \dots \circ f(x|_{S_m}),$$

where  $x|_{S_i}$  is the string of length  $l = |S_i|$  obtained by selecting the bits of  $x$  indexed by  $S_i$ . For convenience, use the notation  $f_i(x) := f(x|_{S_i})$ . Note that the domain of each  $f_i$  is  $\{0, 1\}^l$ .

The intuition behind this construction is that if the sets  $S_i$  were completely disjoint, then the strings  $x|_{S_i}$  would be completely independent, since they would have no common bits, making  $G$  hard to predict. However, in this case, as we saw,  $d \geq ml$ .

What we hope is that by trading independence of the strings  $x|_{S_i}$ , by allowing a bit of overlap (bounded above by  $|S_i \cap S_j| \leq a$ ), we can still achieve satisfactory unpredictability. The following theorem quantifies these ideas:

**Theorem 3 (NW)** Assume that the following two conditions hold (to be used in the Nisan-Wigderson generator):

- there exists  $f : \{0, 1\}^l \rightarrow \{0, 1\}$  such that  $f \in E := \text{DTIME}(2^{O(l)})$  and

$$f \text{ is } \left( t, \frac{1}{2} - \frac{1}{\epsilon(l)} \right) \text{-averagecasehard}$$

- there exists an  $(l, a)$ -design  $S_1, \dots, S_m \subseteq [d]$  such that

$$m = t(l)^{1/3} \quad \text{and} \quad a = \frac{1}{3} \log t(l)$$

Then the Nisan-Wigderson generator  $G$  is a  $\frac{1}{m}$ -PRG against non-uniform time  $m$ .

Before we move on to the proof of theorem 3, we mention two interesting corollaries.

**Corollary 4** If  $f \in E = \text{DTIME}(2^{O(l)})$  such that  $f$  is  $(t, \frac{1}{2} - t)$ -average case hard for

$$\begin{aligned} t = 2^{\Omega(l)} &\implies P = \text{BPP} \\ t = 2^{l^{\Omega(1)}} &\implies \tilde{P} = \text{BPP} \\ t = l^{\omega(1)} &\implies \text{BPP} \subseteq \text{SUBEXP} \end{aligned}$$

**Corollary 5** There exists  $(m, 1/m)$  PRG for depth  $d$  circuits of size  $m$  such that the PRG is computable in polynomial time.

Now we present the proof of theorem 3:

**Proof**

Suppose the result is not true. Then there exists a next-bit predictor  $P$  such that

$$\Pr_{i,x} \left[ P(f_1(x) \circ f_2(x) \circ \dots \circ f_{i-1}(x)) = f_i(x) \right] \geq \frac{1}{2} + \frac{\epsilon}{m}. \quad (1)$$

Note that the circuit size of  $P$  is the sum of the runtime of the PRG, which is  $m$  and the size of the advice we gave  $P$  in the proof, which is  $O(m)$ , hence  $\text{size}(P) = O(m)$ .

Using a standard argument (seen before in other lectures), there exists  $i^*$  that achieves the expectation, in other words

$$\Pr_{\substack{\text{bits of } x \text{ in } S_{i^*}, \\ \text{bits of } x \text{ not in } S_{i^*}}} \left[ P(f_1(x) \circ f_2(x) \circ \dots \circ f_{i^*-1}(x)) = f_{i^*}(x) \right] \geq \frac{1}{2} + \frac{\epsilon}{m}. \quad (2)$$

Note that this is just inequality (1) as before, rewritten for  $i^*$  and with the probability split over two sets.

Now using an averaging process, we see that there must exist a setting  $Z$  of the bits of  $x$  not in  $S_i$  which achieves (2). We change notation and use the variable  $y$  to denote the  $x$ 's that has its bits not in  $S_i$  set according to the setting  $Z$ . Then (2) becomes

$$\Pr_y \left[ P(f_1(y) \circ f_2(y) \circ \dots \circ f_{i^*-1}(y)) = f_{i^*}(y) \right] \geq \frac{1}{2} + \frac{\epsilon}{m} \quad (3)$$

Note that in (3), in  $f_{i^*}(y)$ , the unset variables are those indexed by  $S_{i^*}$  and  $f_{i^*}$  depends on all these. However, on the left hand side of the equality inside the probability in (3), each  $f_j$ ,  $1 \leq j \leq i^* - 1$  depends only on the unset variables index by  $S_j \cap S_{i^*}$ , for the other variable of  $y$  have been fixed according to the setting  $Z$  chose above.

Hence, each  $f_j$  depends on  $|S_i \cap S_j| \leq a$  variables. The  $2^a$  values can be encoded as advice, giving a total advice size of  $m \cdot 2^a$ . This relatively small size of the advice (for special  $m$  and  $a$ ) is crucial in what follows.

Define  $A(y) = P(f_1(y) \circ \dots \circ f_{i^*-1}(y))$ .

- predicts  $f(y)$  with advantage at least  $\frac{\epsilon}{m} \approx \frac{1}{m^2}$

- has circuit size  $m \cdot 2^a + \text{size of}(P)$ . The latter we saw to be  $O(m)$ . Since we picked  $a, m$  to satisfy  $a = \frac{1}{3} \log t(l)$  and  $m = t(l)^{\frac{1}{3}}$ , we have that

$$\text{size}(A(y)) = m \cdot 2^a + O(m) = t(l)^{\frac{1}{3}} \cdot t(l)^{\frac{1}{3}} + O(t(l)^{\frac{1}{3}}) \ll t(l),$$

contradicting the first assumption of theorem 3. (the average case hardness assumption)

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