#### Lecture 11 Hardness vs Randomness

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CS 860 - Graduate Complexity Theory Fall 2022



#### • Nisan-Wigderson (NW) Generators

#### **Pseudorandom Generators**

#### Definition 1 (Pseudorandom Distributions)

A distribution R over  $\{0,1\}^m$  is  $(s,\varepsilon)\text{-pseudorandom}$  if for every circuit C such that  $S(C)\leq s$ 

$$\left|\Pr[C(R)=1] - \Pr[C(U_m)=1]\right| < \varepsilon$$

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• We say that  $G : \{0,1\}^{\ell} \to \{0,1\}^m$  is  $(s,\varepsilon)$ -pseudorandom if the distribution  $G(U_{\ell})$  is  $(s,\varepsilon)$ -pseudorandom.

# Constructing PRGs

- ▶ It seems to be very hard to construct PRGs unconditionally
- Today: one can use hard boolean functions to construct PRGs
  Idea:
  - 1. unpredictability equivalent to pseudorandomness ([Yao 1982])
  - 2. a hard function should be hard to predict

## Nisan-Wigderson PRG

Definition 2 (Average-Case Hardness)

Given  $f: \{0,1\}^n \to \{0,1\}$ , its average-case hardness, denoted by H(f), is the smallest  $s \in \mathbb{N}$  such that

$$\forall C \text{ circuit s.t. } S(C) \leq s \Rightarrow \Pr_x[C(x) = f(x)] \leq 1/2 + 1/s$$

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#### Theorem 3 (Special case of [NW 1994])

If there is  $L \in E$  and  $\delta > 0$  such that for all sufficiently large n,  $H(L_n) \ge 2^{\delta n}$ , then there is constant c > 0 and family of PRGs  $G_m : \{0,1\}^{c \log m} \to \{0,1\}^m$  which are computable in poly(m)time and are (2m, 1/8)-pseudorandom.

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• In particular, the above implies P = BPP.

## Combinatorial designs

#### Definition 4 (Combinatorial designs)

Given integers  $t>\ell>d>0,$  the family  $\{S_1,\ldots,S_m\}$  of subsets of [t] is a  $(t,\ell,d)\text{-design}$  if

- 1.  $|S_i| = \ell$  for all  $i \in [m]$
- 2.  $|S_i \cap S_j| \le d$  for all  $i \ne j$

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#### Proposition 5

For every  $\ell \in \mathbb{N}^*$  and  $\gamma \in (0,1)$ , there is  $t = O(\gamma^{-1}\ell)$  such that a  $(t, \ell, \gamma \ell)$ -design  $\{S_1, \ldots, S_m\}$ , where  $m := 2^{\gamma \ell}$ , can be constructed in  $O(2^t \cdot tm^2)$  time.

#### NW generators: construction

Notation: if  $x \in \{0,1\}^t$  and  $S \subseteq [t]$ , let  $x_S \in \{0,1\}^{|S|}$  be the string obtained by selecting the bits of S (in order) from x

#### Definition 6 (NW generators)

For a boolean function  $f: \{0,1\}^\ell \to \{0,1\}$  and a design  $S := \{S_1, \ldots, S_m\}$  over [t], the NW-generator is given by

$$NW_{f,\mathcal{S}}(x) := f_1(x) \circ \cdots \circ f_m(x)$$

where  $f_i(x) := f(x_{S_i})$ 

#### Proof of Pseudorandomness

Follows from the following lemma:

#### Lemma 7

Let  $t, \ell, \gamma$  as in Proposition 5 and  $m := 2^{\gamma \ell}$ . If  $f : \{0, 1\}^{\ell} \to \{0, 1\}$ and  $S := \{S_1, \ldots, S_m\}$  be a  $(t, \ell, \log m)$ -design over [t]. If  $D : \{0, 1\}^m \to \{0, 1\}$  is s.t.

$$|\Pr_{r}[D(r) = 1] - \Pr_{x}[D(NW_{f,\mathcal{S}}(x)) = 1]| > \varepsilon$$

then there is a circuit C with  $S(C) = O(m^2)$  s.t.

$$|\Pr_x[D(C(x)) = f(x)] - 1/2| > \varepsilon/m.$$

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Above lemma shows that a distinguisher for the generator yields a distinguisher for f

Main idea: if D distinguishes  $NW_{f,S}$  from uniform, then can find a bit of output of  $NW_{f,S}$  where we can notice this difference. From this bit, we can non-trivially predict f.

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Define distributions  $H_0, \ldots, H_m$  over  $\{0, 1\}^m$  as follows:

- Sample  $u \sim U_m$  and  $v \sim NW_{f,\mathcal{S}}(U_t)$
- $H_i$  given by  $v_{[i]} \circ u_{[m] \setminus [i]}$

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$$H_0 = U_m$$
 and  $H_m = NW_{f,\mathcal{S}}(U_t)$ 

By hypothesis of lemma, there is  $b_0 \in \{0,1\}$  s.t.

$$\Pr_{x}[D'(NW_{f,\mathcal{S}}(x)) = 1] - \Pr_{r}[D'(r) = 1] > \varepsilon$$

where  $D'(x) = b_0 \oplus D(x)$ .

Note

$$\varepsilon < \Pr_{x}[D'(NW_{f,S}(x)) = 1] - \Pr_{r}[D'(r) = 1]$$
  
=  $\Pr[D'(H_m) = 1] - \Pr[D'(H_0) = 1]$   
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• Assume  $S_i = [\ell]$  and let  $\{0,1\}^t = \{0,1\}^\ell \times \{0,1\}^{t-\ell}$  s.t. x = (y,z)

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• Above inequality  $\Rightarrow$  good distinguisher for  $f: \{0,1\}^{\ell} \rightarrow \{0,1\}$ 

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2. compute  $f_1(x), \ldots, f_{i-1}(x)$   $(x = (y, z), f_i(x) := f(x_{S_i}))$ 

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- 3. If  $D'(f_1(x), \ldots, f_{i-1}(x), r_i, \ldots, r_m) = 1$ , output  $r_i$ . Else, output  $1 - r_i$ .

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**Claim:**  $\Pr_{y,z,r}[A(y) = f(y)] > 1/2 + \varepsilon/m$ 

Same proof as last lecture's.

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By averaging, there are fixed z,r such that A when given z,r approximates f well.

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- ▶ By design property,  $i \neq j \Rightarrow |S_i \cap S_j| \le \log m$ .  $f_j(y, z) = f_j(x) = f(x_{S_j})$  depends on  $\le \log m$  bits of y!

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- $\blacktriangleright \ \ \, {\rm Since we have fixed } z,r$

 $f_j$  computed by circuit of size O(m)

So all m bits can be computed by a  $O(m^2)$  sized circuit!

## Proof of Theorem 3

• Let  $f_{\ell}: \{0,1\}^{\ell} \to \{0,1\}$  be given by  $f(x) := L_{\ell}(x)$ .

• Let  $G_m := NW_{f,S}$  with the parameters  $\ell, \gamma, t$  and  $m = 2^{\gamma \ell}$  from Proposition 5 (design)

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- ▶ By Definition 1,  $G_m$  is not (2m, 1/8)-pseudorandom  $\Rightarrow$  exists cicuit D with  $S(D) \le 2m$  s.t.

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▶ By Lemma 7, there is circuit  $\Phi$  of size  $O(m^2)$  such that

$$\Pr_x[\Phi(x) = f(x)] > 1/2 + 1/8m = 1/2 + 2^{-\gamma\ell - 3}$$

which contradicts  $H(L_{\ell}) \geq 2^{\delta \ell}$  when  $\gamma < \delta/3$ 

- ▶ Take p to be a prime number and consider  $\mathbb{F}_p$  finite field with p elements. Let  $t = p^2$ .
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- There are  $p^{d+1}$  polynomials of degree  $\leq d$

## References I

Arora, Sanjeev and Barak, Boaz (2009) Computational Complexity, A Modern Approach Cambridge University Press

Chapter 20

Papadimitriou, C (1994) Computational Complexity Addison-Wesley



Trevisan, Luca (2002)

Lecture notes See webpage



#### Goldreich, Oded (2006)

Computational complexity: a conceptual perspective. Chapter 6 https://www.wisdom.weizmann.ac.il/~oded/cc-drafts.html

Lectures 23, 24

## References II

Babai, L and Fortnow, L and Nisan, N and Wigderson, A (1993)

BPP has subsexponential time simulations unless EXPTIME has publishable proofs

Computational Complexity



Impagliazzo, Russell (1995)

Hard-core distributions for somewhat hard problems  $\overline{\text{FOCS}}$ 



Impagliazzo, Russell and Wigderson, Avi (1997) P = BPP unless E has subexponential circuits STOC

Impagliazzo, Russell and Wigderson, Avi (1998) Randomness vs Time: Derandomization under a uniform assumption FOCS

## References III

#### Nisan, Noam and Wigderson, Avi (1994)

Hardness vs Randomness

Journal of Computer and System Sciences

#### Yao, Andrew C. (1982)

Theory and applications of trapdoor functions FOCS