

Pseudorandomness - Lecture 14.

Agenda: - Intro to pseudorandom generators
- Hybrid argument.

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Lecture #: 14.

What is a pseudorandom object/distribution?

An object that exhibits some property that makes an observer think it was picked randomly.

The stupid algo for max-cut — pairwise ~~of~~ independence.

$\bigoplus_{i \in S} x_i$

— ϵ -biased distributions

Empirical avg of samples

— expander walk.

Any randomized algo
running in time $O(n^2)$

— ??

Computational indistinguishability:

Two RVs X, Y taking values in $\{0,1\}^m$ is ϵ -c.i.
for a class $C = \{T: \{0,1\}^m \rightarrow [0,1]\}$ of test functions

iff

$$|E[T(X)] - E[T(Y)]| \leq \epsilon \text{ for all } T \in C.$$

That is, as far as "tests" from C are concerned, they behave roughly similarly whether they are fed X or Y .

PRG for a class \mathcal{C} : $(u_1, u_2, \dots, u_t) \mapsto (u_{t+1}, u_2, u_3, \dots, u_{t+1})$

A map $G: \{0,1\}^d \rightarrow \{0,1\}^m$ is an ϵ -PRG for \mathcal{C} if the RVs U_m and $G(U_d)$ are ϵ -comp. ind for \mathcal{C} . i.e.,

$$\left| \mathbb{E}_{x \sim U_m} [T(x)] - \mathbb{E}_{y \sim U_d} [T(G(y))] \right| \leq \epsilon$$

for all $T \in \mathcal{C}$.

(Often \mathcal{C} corresponds to size m^2 circuits or subclasses of "efficient computation")

Again, we often care for families: $\{G_m: \{0,1\}^{d(m)} \rightarrow \{0,1\}^m\}$.

Desire: - Want d as small as possible
 - Want $G_d(y)$ to be efficiently computable.

Defn: $\{G_m: \{0,1\}^{d(m)} \rightarrow \{0,1\}^m\}$ is $t(m)$ -computable

if there is an algorithm M s.t.

$$M(1^m, x) = G_m(x) \quad \text{and } M \text{ runs in time } t$$

$$\& M(1^m) = d(m)$$

Thm: Suppose for all m , there is $t(m)$ -comp. $1/8$ -PRG $\{G_d: \{0,1\}^{d(m)} \rightarrow \{0,1\}^m\}$ for $\{C_m\}$ where C_m are boolean fns computed by circuits of size $\leq m$.

Then, $BPP \subseteq \bigcup_{c>0} DTIME(2^{d(n^c)} \cdot (n^c + t(n^c)))$

Pf: A is a rand. algo running in time $\leq n^c = m$.

$\Rightarrow A$ uses $\leq m$ random bits.

$x \in L \Rightarrow \Pr[A(x, r) = 1] \geq 2/3$

$x \notin L \Rightarrow \Pr[A(x, r) = 1] \leq 1/3$

Algo B (input x):

\triangleright Build a circuit $T: \{0,1\}^m \rightarrow \{0,1\}$

$$T(r) = A(x, r)$$

\triangleright Run over all $y \in \{0,1\}^{d(m)}$,

compute $z = G(y)$.

count # y : $A(z) = 1$.

\triangleright Acc of this # $> \frac{1}{2} \cdot 2^{d(m)}$.

PRG guarantee $\Rightarrow B$ is correct.

□

Define $\{G_m: \{0,1\}^{d(m)} \rightarrow \{0,1\}^m\}$ is

- \triangleright mildly explicit if it is $\text{poly}(m, 2^{d(m)})$ -computable
- \triangleright fully explicit if it is $\text{poly}(m)$ computable.

output the whole truth table in $\text{poly}(L)$

Computing G on a specific seed.

Do PRGs exist at all?

Always check if a "random object works".

Thm: For any $m \in \mathbb{N}$ and $\epsilon > 0$, there are ^{lots of} PRGs (not explicit)
 $G: \{0,1\}^d \rightarrow \{0,1\}^m$ for size m circuits with seed length
 $d = O(\log m + \log \frac{1}{\epsilon})$.

Pf: Pick $G: \{0,1\}^d \rightarrow \{0,1\}^m$ uniformly at random.

Fix a circuit T of size m ,

$$\mathbb{E}_{y \sim \mathcal{U}_d} [T(G(y))] = \frac{1}{2^d} \sum_{i=1}^{2^d} T(z_i) \quad z_i \sim \mathcal{U}_m.$$

$\mu = \mathbb{E}[T(\mathcal{U})]$. Chernoff says you are within ϵ w.p. $\geq 1 - \exp(-\epsilon^2 \cdot 2^d)$

If G is not a PRG, then

$$\Pr_G \left[\exists T: \left| \mathbb{E}[T(\mathcal{U})] - \frac{1}{2^d} \sum_{i=1}^d T(z_i) \right| > \epsilon \right]$$

$$\leq \binom{\# \text{ circuits of size } \leq m}{\# \text{ size } \leq m} \cdot \exp(-\epsilon^2 \cdot 2^d)$$

$$\therefore \text{If } 2^d \cdot \epsilon^2 > 100 \cdot m \log m$$

$$\Rightarrow d = O(\log m + \log \frac{1}{\epsilon}) \text{ is suff.} \quad \square$$

"Can we find hay in a haystack?"

Aspirational goal: Find an explicit PRG $G: \{0,1\}^d \rightarrow \{0,1\}^m$ for size m circuits with $O(\log m + \log \frac{1}{\epsilon})$ -seed length.

(we will take anything! $d = o(m)$)

A "simpler" requirement from a PRG

Defn: (Next bit unpredictable) X r.v on $\{0,1\}^m$ is (t, ϵ) -NBU if there is no circuit P of size $\leq t$ and no $i \in [m]$ with

$$\Pr_x [P(X_1, \dots, X_{i-1}) = X_i] \geq \frac{1}{2} + \epsilon.$$

Given a prefix, guessing the next bit is hard.

Lemma: If $X \sim \{0,1\}^m$ is (t, ϵ) -pseudorandom, then X is $(t - O(1), \epsilon)$ -NBU.
 Conversely, X is (t, ϵ) -NBU, then X is $(t, \epsilon m)$ -pseudorandom.

Pf: (\Rightarrow): X was pseudorandom but next bit predictable
 \Rightarrow There is a circuit P and an index i s.t.

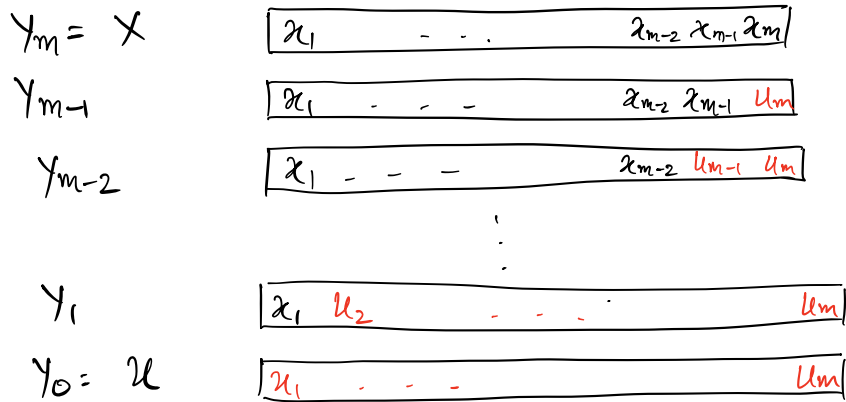
$$\Pr_x [P(X_1, \dots, X_{i-1}) = X_i] \geq \frac{1}{2} + \epsilon$$

Q { Algo on input z_1, \dots, z_m .
 Accept if $z_i = P(z_1, \dots, z_{i-1})$

$$\mathbb{E}[Q(u_m)] = \frac{1}{2}. \quad \mathbb{E}[Q(x)] \geq \frac{1}{2} + \epsilon$$

diff. by ϵ

(\Leftarrow): Given that X is (t, ϵ) -NBU
 Want to show that $X \stackrel{c.i.}{\approx}_{\epsilon m} u_m$.
 Hybrid _{walk} argument!



Assume $X = Y_m \stackrel{c.i.}{\approx}_{\epsilon} Y_{m-1} \stackrel{c.i.}{\approx}_{\epsilon} Y_{m-2} \dots \stackrel{c.i.}{\approx}_{\epsilon} Y_0$
 $\Rightarrow X \stackrel{c.i.}{\approx}_{m\epsilon} \mathcal{U}$

Suppose $Y_0 \not\stackrel{c.i.}{\approx}_{m\epsilon} Y_m \Rightarrow$ there is a P

s.t $\mathbb{E}[P(Y_0)] - \mathbb{E}[P(Y_m)] \geq m\epsilon.$

$\Rightarrow \sum_{i=1}^m \mathbb{E}[P(Y_{i-1})] - \mathbb{E}[P(Y_i)] \geq m\epsilon.$

$\Rightarrow \exists i : \mathbb{E}[P(Y_{i-1})] - \mathbb{E}[P(Y_i)] \geq \epsilon.$

(by replacing P by $\neg P$ if necc, no abs value)

$Y_{i-1} = [x_1 \dots x_{i-1} \quad u_i \quad u_{i+1} \dots u_m]$

$Y_i = [x_1 \dots x_{i-1} \quad x_i \quad u_{i+1} \dots u_m]$

P is more likely to acc Y_{i-1} than Y_i .

Define a circuit \tilde{P} which gets input x_1, \dots, x_{i-1} :

Pick z_1, \dots, z_m at random.

$b = P(x_1, \dots, x_{i-1}, z_1, \dots, z_m)$

If $b=1$, return \bar{z}_i . else return z_i .

What is the prob that \tilde{P} is right?

$$\alpha = P_0 [P(X_{1,0} \rightarrow X_{i-1}, X_i, U_{i+1,0} \rightarrow U_m) = 1]$$

$$\alpha' = P_0 [P(X_{1,0} \rightarrow X_{i-1}, \bar{X}_i, U_{i+1,0} \rightarrow U_m) = 1]$$

$$P_0 [P(X_{1,0}, X_{i-1}, U_i, U_{i+1,0} \rightarrow U_m) = 1] = \frac{1}{2} (\alpha + \alpha') \geq \alpha + \epsilon$$

$$\Rightarrow \alpha' \geq \alpha + 2\epsilon.$$

$$P_0 [\tilde{P} \text{ is correct }] \quad \begin{array}{l} \text{if } z_i = x_i \quad \& \quad b = 0 \\ \text{or } z_i = \bar{x}_i \quad \& \quad b = 1 \end{array}$$

||

$$\frac{1}{2} \cdot (1 - \alpha) + \frac{1}{2} \cdot \alpha' = \frac{1}{2} + \frac{1}{2} (\alpha' - \alpha) \geq \frac{1}{2} + \epsilon.$$

□.

How do we use this to build PRGs?

Toy case: stretch of 1.

$$G: \{0,1\}^d \rightarrow \{0,1\}^{d+1}$$

[Blum-Micali] $G(x) = x \oplus b$

$b = \text{Hard Function}(x)$.
"hard to guess".

Later in the course:

Suppose we have access to "really hard" functions,
then we can use that to build PRGs.

"If you can find hay in one haystack, you can find one in another"

[Impagliazzo-Wigderson] If $E = \text{DTIME}(2^{o(n)})$ has a language that requires circuits of size $2^{\Omega(n)}$, then $P = \text{BPP}$.