

Randomness is Linear in Space

Nisan and Zuckerman (1994)

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Main Result

Theorem (1)

Any randomized algorithm A that runs in space S and time T and uses $\text{poly}(S)$ random bits can be simulated using only $O(S)$ random bits in space S and time $T + \text{poly}(S)$. The distribution of the output of the simulation is within statistical distance of $\exp(-S^{1-\gamma})$ from the distribution of the output of A. Here $S = S(n) \geq \log(n)$, $T = T(n) \geq n$, and $\gamma > 0$ is an arbitrary constant.

For any polynomial $p(x)$, we need a psuedo-random generator $G : \{0, 1\}^{O(S)} \rightarrow \{0, 1\}^{p(S)}$ for space S with parameter $\exp(-S^{1-\gamma})$ running in time $\text{poly}(S)$ and space $O(S)$.

Agenda

- Pseudo-Random Generators for Space S
- Model of space-bounded computation, motivation for extractors
- Definition and intuition for extractors
- Some ideas of existence proofs for extractors

Pseudo-Random Generators for Space S

For any polynomial $p(x)$, we need a pseudo-random generator $G : \{0, 1\}^{O(S)} \rightarrow \{0, 1\}^{p(S)}$ for space S with parameter $\exp(-S^{1-\gamma})$ running in time $\text{poly}(S)$ and space $O(S)$.

Definition

A generator $G : \{0, 1\}^n \rightarrow \{0, 1\}^m$ is called a pseudo-random generator for space S with parameter ϵ if, for every randomized space S algorithm A and every input to it,

$$|\Pr[A(y) \text{ accepts}] - \Pr[A(G(x)) \text{ accepts}]| \leq \epsilon$$

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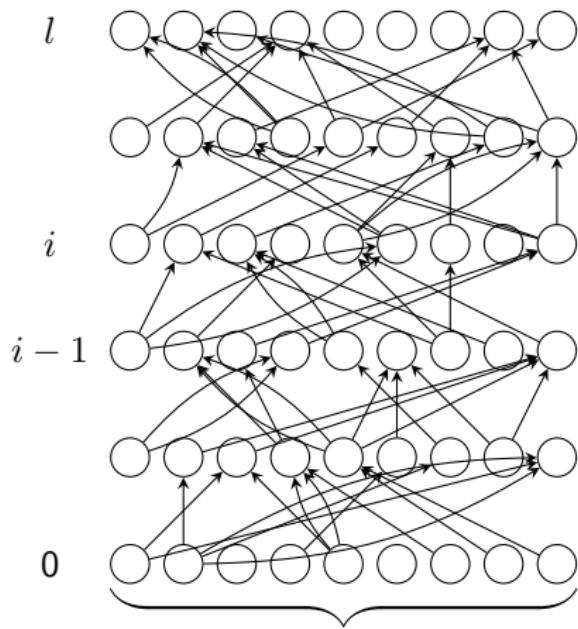
Lemma

Let $G_1 : \{0, 1\}^{R_2} \rightarrow \{0, 1\}^{R_1}$ be a generator for space S_1 with parameter ϵ_1 running in space S_2 . Let $G_2 : \{0, 1\}^{R_3} \rightarrow \{0, 1\}^{R_2}$ be a generator for space $S_1 + S_2$ with parameter ϵ_2 running in space S_3 . Then

$G_1 \circ G_2 : \{0, 1\}^{R_3} \rightarrow \{0, 1\}^{R_1}$ is a pseudo-random generator for space S_1 with parameter $\epsilon_1 + \epsilon_2$ running in space $S_2 + S_3$.

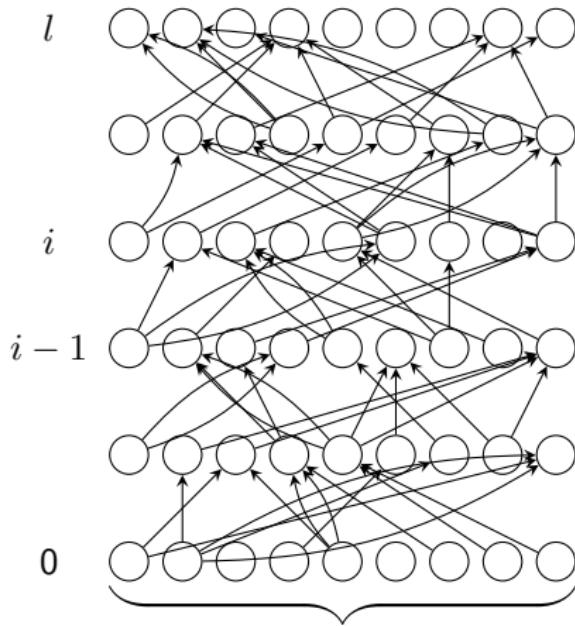
Assume $p(n) = n^c$. Thus, we only need a pseudo-random generator $G : \{0, 1\}^R \rightarrow \{0, 1\}^{\Omega(RS^\gamma)}$ with parameter $\exp(-S^{1-\gamma})$ running in time $\text{poly}(S)$ that we compose with itself $(c - 1)/\gamma$ times. Assume without loss of generality that $R \geq 4S$.

Proof



- Let M be an arbitrary space(S) machine
- Each layer represents configuration after a set of S random bits.
- Edge $((i, j), (i + 1, k))$ labeled by $r \Leftrightarrow S\text{-bit random string } r$ causes M to go from j to k .
- U_i : Distribution on layer i in response to true randomness
- D_i : Distribution on layer i after running G
- We want $\|U_l - D_l\| \leq 2^{-\Omega(S^{1-\gamma})}$. Let $\epsilon = 2^{-\Omega(S^{1-\gamma})}$.

Proof (continued)



Let $x \in \{0, 1\}^R$ be the input for G , and let $r_1, \dots, r_l \in \{0, 1\}^S$ be the outputs. Let X be the random variable for x (X is uniform).

We want $\|U_l - D_l\| \leq \epsilon$.

We prove by induction that

$$\|U_i - D_i\| \leq i\epsilon/l.$$

Let D_i^j and U_i^j be D_i and U_i conditioned on the fact that M was in configuration j at step $i - 1$.

Dividing up cases

Consider dividing into good configurations (say A) and bad configurations at step $i - 1$. Let A be the configurations j which have $D_{i-1}[j] \geq 2^{-2S}$.

- If $j \in A$, then look at $X|(i - 1, j)$.

We know that

$$\begin{aligned}\Pr[X = x | (i - 1, j)] &\leq \Pr[X = x] / \Pr[(i - 1, j)] \\ &\leq 2^{-R} 2^{2S} \leq 2^{-R/2}\end{aligned}$$

We want that $\|U_i^j - D_i^j\| \leq \epsilon'$, which would come from the distribution of r_i conditioned on $(i - 1, j)$ being quasi-random to within ϵ' .

- For $j \notin A$, $D_{i-1}[j] \leq 2^{-2S}$, and there are at most 2^S possible values of j , so

$$\sum_{j \notin A} D_{i-1}[j] \leq 2^S 2^{-2S} = 2^{-S}$$

Proof (continued)

$$\begin{aligned}\|U_i - D_i\|_1 &= \left\| \sum_j U_{i-1}[j]U_i^j - \sum_j D_{i-1}[j]D_i^j \right\|_1 \\ &\leq \left\| \sum_j U_{i-1}[j]U_i^j - \sum_j D_{i-1}[j]U_i^j \right\|_1 \\ &\quad + \left\| \sum_j D_{i-1}[j]U_i^j - \sum_j D_{i-1}[j]D_i^j \right\|_1 \\ &\leq \left(\sum_j |U_{i-1}[j] - D_{i-1}[j]| \right) \|U_i^j\|_1 \\ &\quad + \left(\sum_{j \in A} |D_{i-1}[j]| \right) \|U_i^j - D_i^j\|_1 + \left(\sum_{j \notin A} |D_{i-1}[j]| \right) \|U_i^j - D_i^j\|_1 \\ &\leq \|U_{i-1} - D_{i-1}\|_1 \cdot 1 + 1 \cdot 2\epsilon' + 2^{-S} \cdot 2.\end{aligned}$$

What we need

Distribution of X in $j \in A$ case

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Definition (δ -source)

A distribution D on $\{0, 1\}^n$ is a δ -source if for all $x \in \{0, 1\}^n$, $D(x) \leq 2^{-\delta n}$.

Note that we know that $X|(i - 1, j)$ is a δ -source.

Extractors

This motivates the need for an extractor!

Definition

Let $E : \{0, 1\}^n \times \{0, 1\}^t \rightarrow \{0, 1\}^m$. E is called a (δ, ϵ) -extractor if for every δ -source D , the distribution of $E(x, y) \circ y$ induced by choosing x from D and y uniformly in $\{0, 1\}^t$ is within statistical distance of ϵ from the uniform distribution (on $\{0, 1\}^m \times \{0, 1\}^t$.)

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Intuition for extractors

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Intuition for extractors

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Intuition for extractors

- Hash families
- Bipartite graph on $\{0, 1\}^n \times \{0, 1\}^m$ with good expansion properties.

More Extractor Intuition

Consider a set $U \subseteq \{0, 1\}^n$ with $|U| \geq 2^{\delta n}$, and suppose you have a random element from A .

How can you extract the δn bits of randomness?

An extractor gives you $\Omega(\delta^2 n)$ bits.

Existence of Extractors

Lemma

For any parameters $\delta = \delta(n)$ and $\epsilon = \epsilon(n)$ with $1/n \leq \delta \leq 1/2$ and $2^{-\delta n} \leq \epsilon \leq 1/n$, there exists an easily computable (and explicitly given) (δ, ϵ) -extractor $E : \{0, 1\}^n \times \{0, 1\}^t \rightarrow \{0, 1\}^m$, where $t = O(\log \epsilon^{-1} \log^2 n \log \delta^{-1}/\delta)$ and $m = \Omega(\delta^2 n / \log(\delta^{-1}))$.

Proof Idea: Leftover Hash Lemma and converting δ -sources to block-wise δ -sources.

Proof Ideas: Leftover Hash Lemma

Lemma (Leftover Hash Lemma)

Let $X \subset \{0, 1\}^n$, $|X| \geq 2^r$. Let $k > 0$, and H be a 2-universal family of hash functions mapping n bits to $r - 2k$. Then the distribution $(h, h(x))$ is quasi-random within $1/2^k$ (on the set $H \times \{0, 1\}^{r-2k}$), where h is chosen uniformly at random from H , and x uniformly from X .

Proof Ideas: Leftover Hash Lemma

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Corollary

Let D be a distribution on $\{0, 1\}^n$ such that for all $x \in \{0, 1\}^n$, $D(x) \leq 2^{-r}$. Let $k > 0$, and let H be a universal family of hash functions mapping n bits to $r - 2k$ bits. Then the distribution $(h, h(x))$ is quasi-random within $1/2^k$ (on the set $H \times \{0, 1\}^{r-2k}$), where h is chosen uniformly at random from H , and x according to D .

Block-wise δ -sources

Definition

A distribution D on the space $\{0, 1\}^{l_1} \times \{0, 1\}^{l_k}$ is called a block-wise δ -source if, for $1 \leq i \leq k$ and for all values $x_1 \in \{0, 1\}^{l_1}, \dots, x_i \in \{0, 1\}^{l_k}$, we have that

$$\Pr[X_i = x_i | X_1 = x_1, \dots, X_{i-1} = x_{i-1}] \leq 2^{-\delta l_i}$$