

# 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 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  $O(S)$ .

# Agenda

- Psuedo-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  $\epsilon$  if, for every randomized space  $S$  algorithm  $A$  and every input to it,

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

# Pseudo-Random Generators for Space $S$

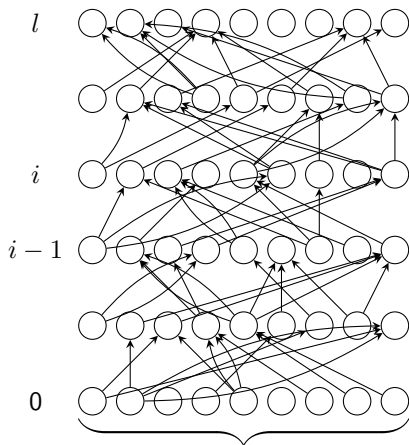
For any polynomial  $p(x)$ , we need a pseudo-random generator  $G : \{0, 1\}^{O(S)} \rightarrow \{0, 1\}^{p(S)}$  for space  $S$  with parameter  $\exp(-S^{1-\gamma})$  running in time  $\text{poly}(S)$  and space  $O(S)$ .

## Lemma

*Let  $G_1 : \{0, 1\}^{R_2} \rightarrow \{0, 1\}^{R_1}$  be a generator for space  $S_1$  with parameter  $\epsilon_1$  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  $\epsilon_2$  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  $R \geq 4S$ .

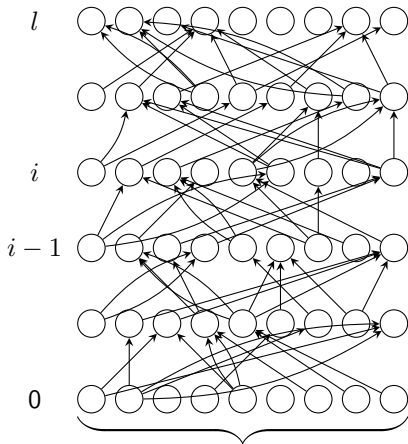
# Proof



$O(2^S)$  configurations

- 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  $r \Leftrightarrow S$ -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)



$O(2^S)$  configurations

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  $U_i$  and  $U_j$  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  $\epsilon'$ .

- 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

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

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## Definition ( $\delta$ -source)

A distribution  $D$  on  $\{0, 1\}^n$  is a  $\delta$ -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  $\delta$ -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  $(\delta, \epsilon)$ -extractor if for every  $\delta$ -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  $\epsilon$  from the uniform distribution (on  $\{0, 1\}^m \times \{0, 1\}^t$ .)

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Intuition for extractors

- Hash families

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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  $\delta n$  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)  $(\delta, \epsilon)$ -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  $\delta$ -sources to block-wise  $\delta$ -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  $X$ .*

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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 $\delta$ -sources

## Definition

A distribution  $D$  on the space  $\{0, 1\}^{l_1} \times \{0, 1\}^{l_k}$  is called a block-wise  $\delta$ -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

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