

Near-Optimal Pseudorandom Generators for Constant-Depth Read-Once Formulas

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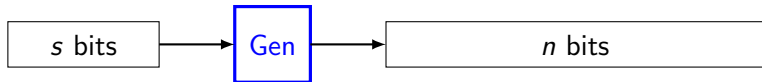
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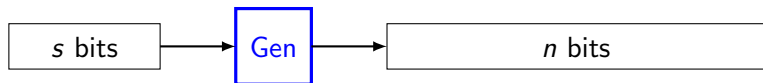
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- ▶ But randomness is **costly**
- ▶ An algorithm that uses **fewer random bits** is better

Pseudorandom generators (PRGs)



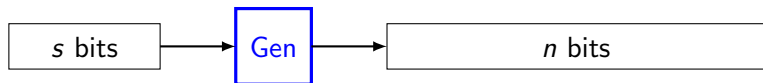
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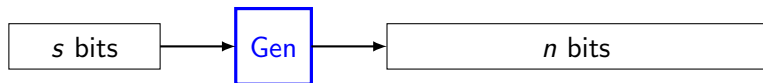


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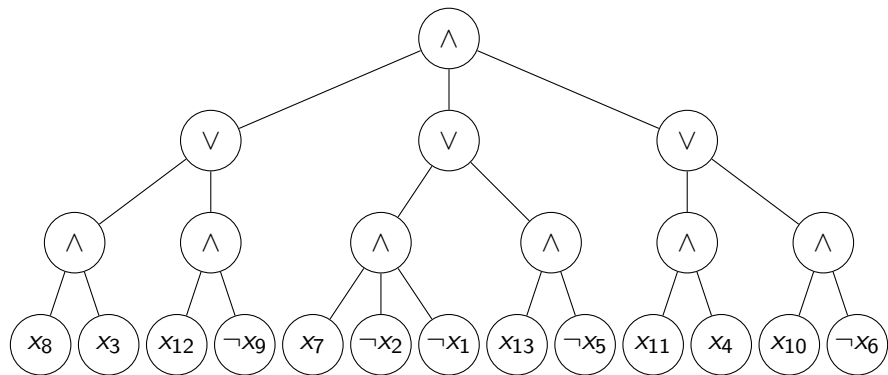


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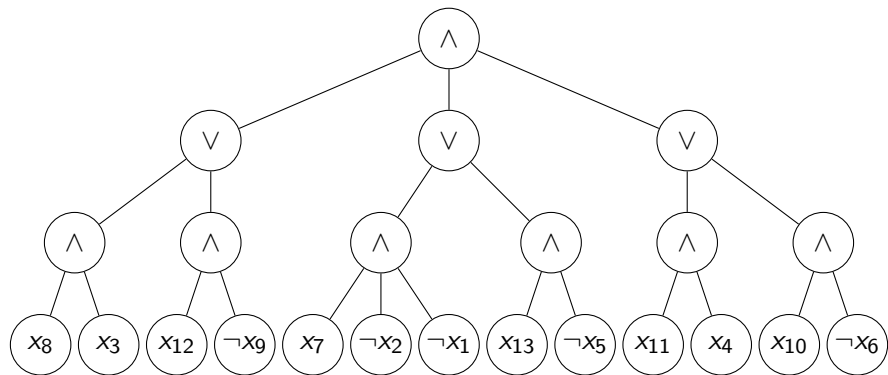
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- ▶ Minimize seed length $s = s(n, \varepsilon)$

Read-once formulas

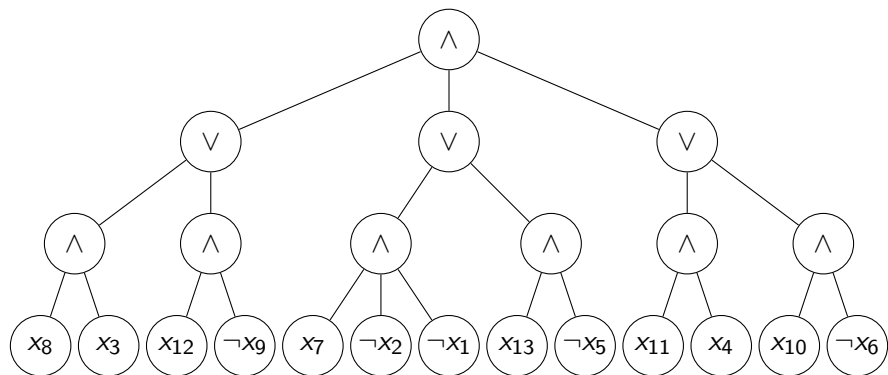


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- ▶ Read-once version of \mathbf{AC}^0

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► **Main result:** PRG for read-once \mathbf{AC}^0 with seed length

$$\log(n/\varepsilon) \cdot O(d \log \log(n/\varepsilon))^{2d+2}.$$

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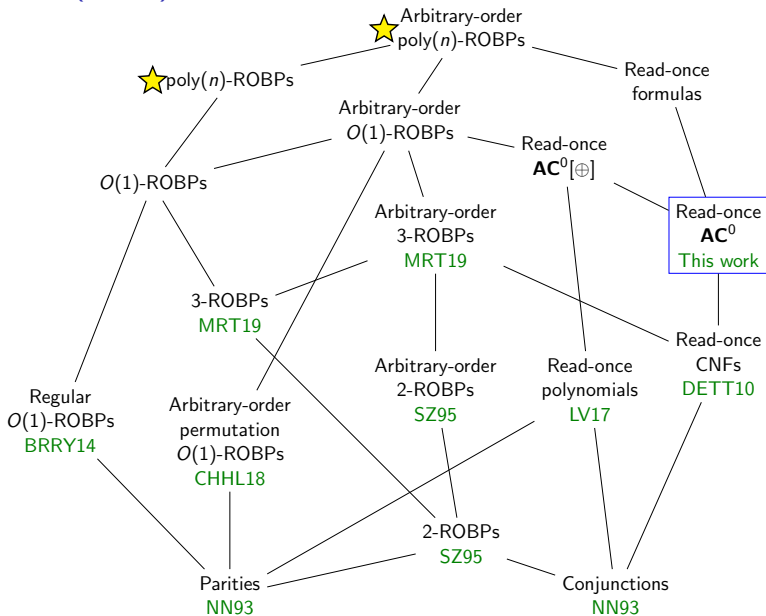
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- ▶ Read-once **AC**⁰ is one of the **frontiers** of this progress

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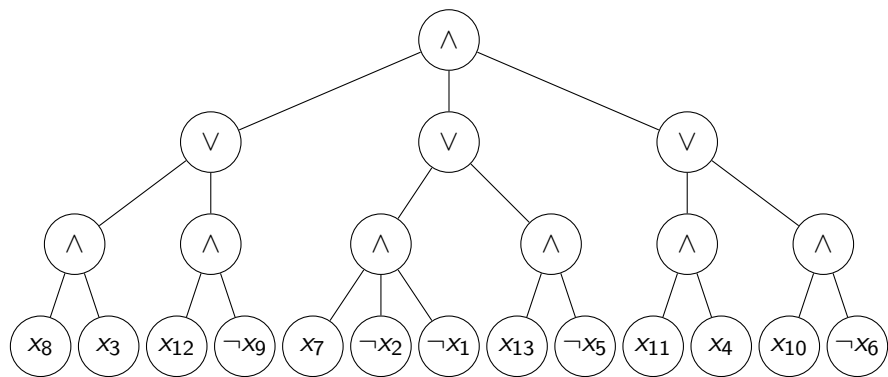
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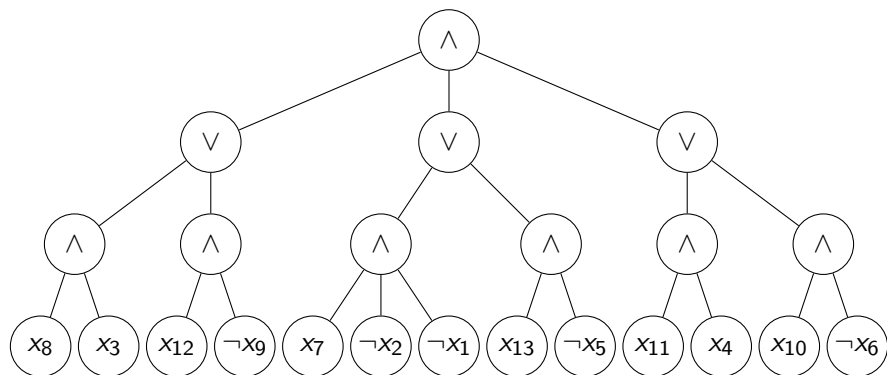
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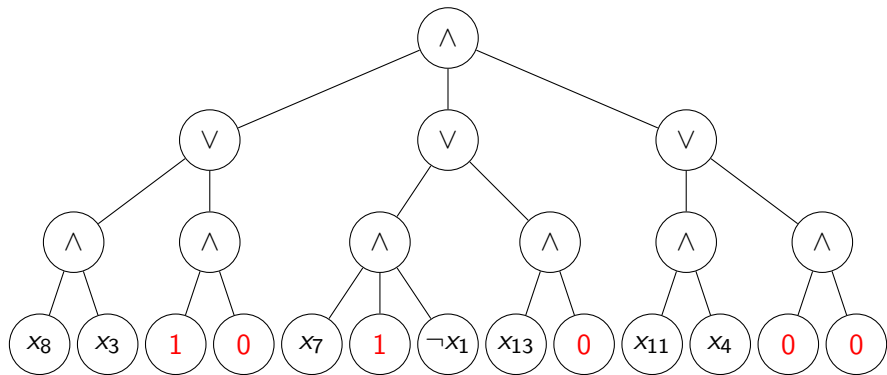
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Restriction notation

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- ▶ (Proof involves clever Fourier analysis, building on [RSV13, HLV18, CHRT18])

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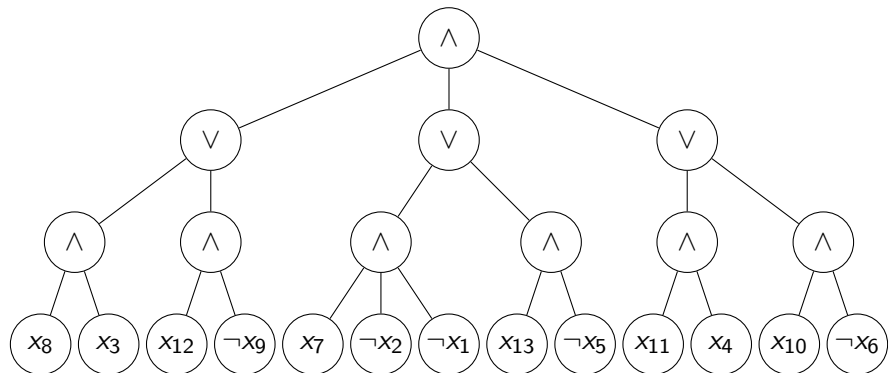
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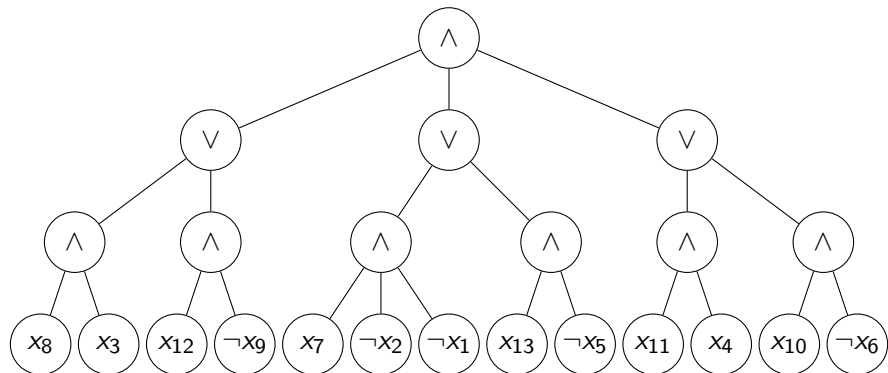
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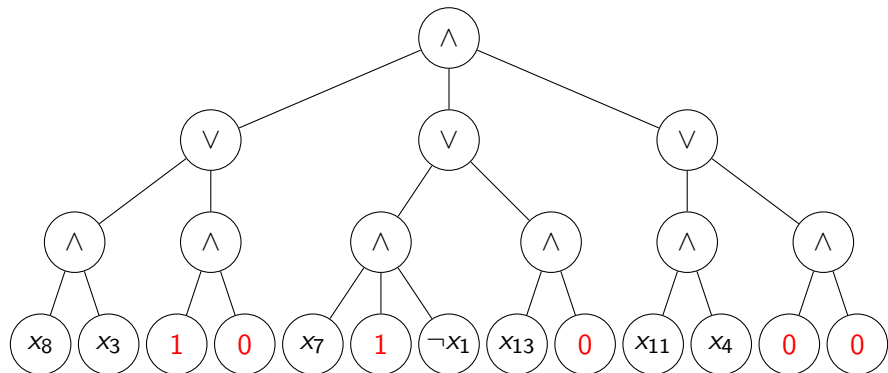
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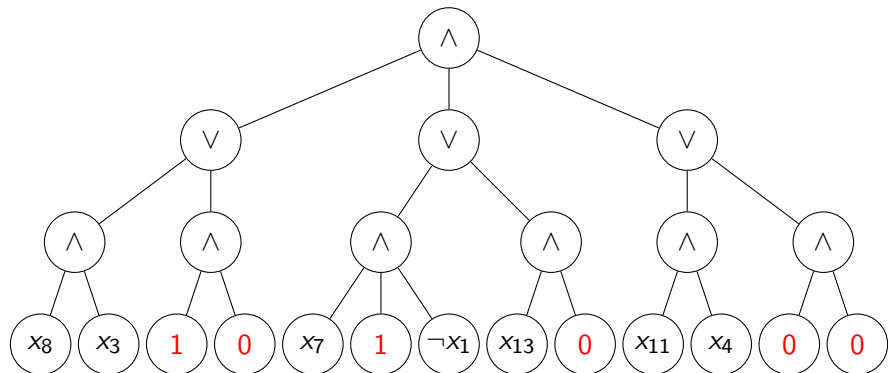
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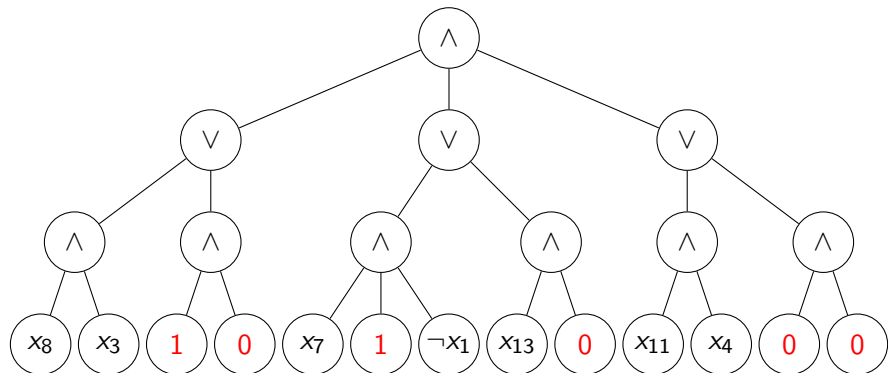
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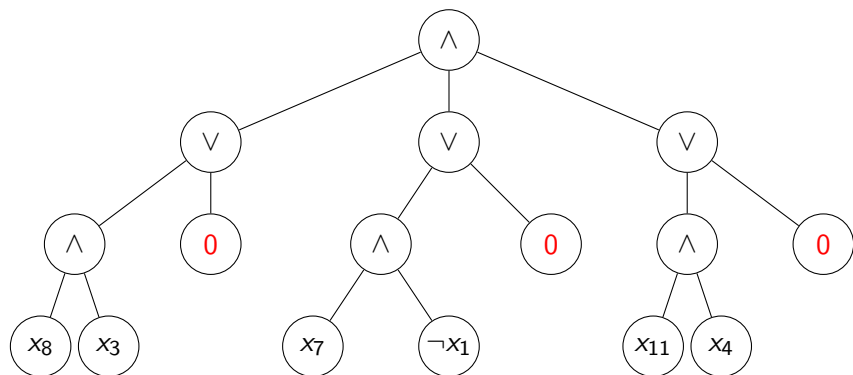
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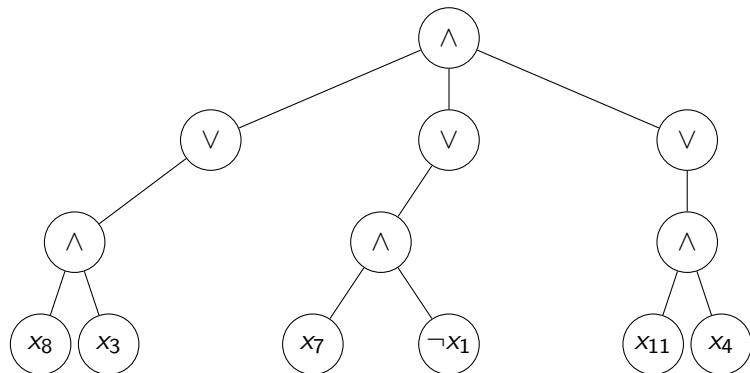
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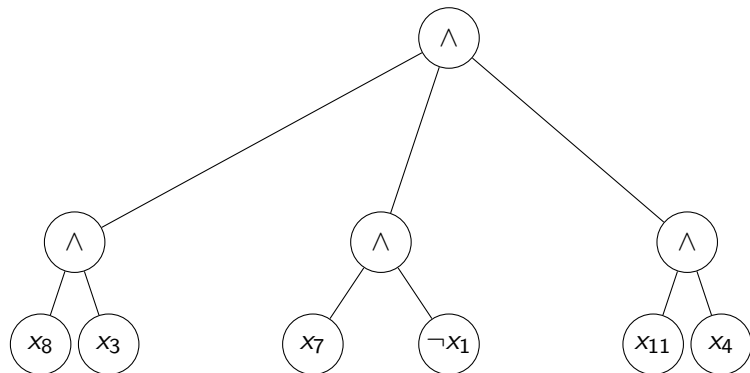
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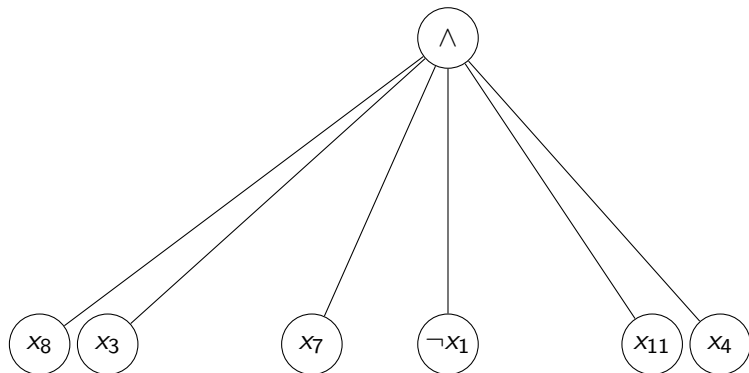
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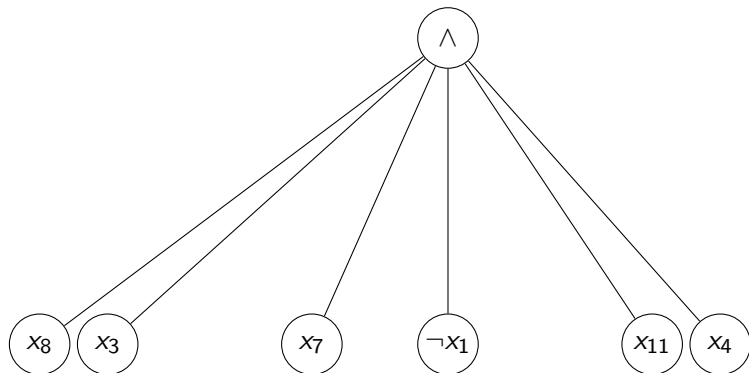
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- ▶ Step 2: Fool restricted formula, **taking advantage of simplicity**

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1. Recursively sample $G_d, G'_d \in \{0, 1\}^n$
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 3. $X = \text{Res}(G_d \oplus D, G'_d \oplus D')$

Preserving expectation

- ▶ **Claim:** For any depth- $(d + 1)$ read-once \mathbf{AC}^0 formula f ,

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- ▶ **Proof:** Read-once \mathbf{AC}^0 can be simulated by constant-width ROBPs [CSV15]
- ▶ So we can simply apply Forbes-Kelley result:

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where $t = O((\log \log n)^2)$

- ▶ Actually we only prove this statement “up to sandwiching”

$\Delta \mapsto \text{polylog } n$: Proof outline

- ▶ Chen, Steinke, Vadhan '15: Read-once **AC**⁰ simplifies under truly random restrictions

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$\Delta \mapsto \text{polylog } n$: Proof outline

- ▶ Chen, Steinke, Vadhan '15: Read-once \mathbf{AC}^0 simplifies under **truly** random restrictions
- ▶ Testing for simplification is **another read-once \mathbf{AC}^0** problem
- ▶ So we can derandomize the [CSV15] analysis:

$$X = \text{Res}(G_d \oplus D, G'_d \oplus D')$$

Collapse under truly random restrictions

- ▶ Assume f is a **biased** read-once **AC⁰** formula:

$$\mathbb{E}[f] \leq \rho \text{ or } \mathbb{E}[f] \geq 1 - \rho$$

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$$\Pr_{R^{os}}[f|_{R^{os}} \text{ nonconstant}] \leq \rho + \frac{1}{n^{100}},$$

where $s = O(\log \log n)$

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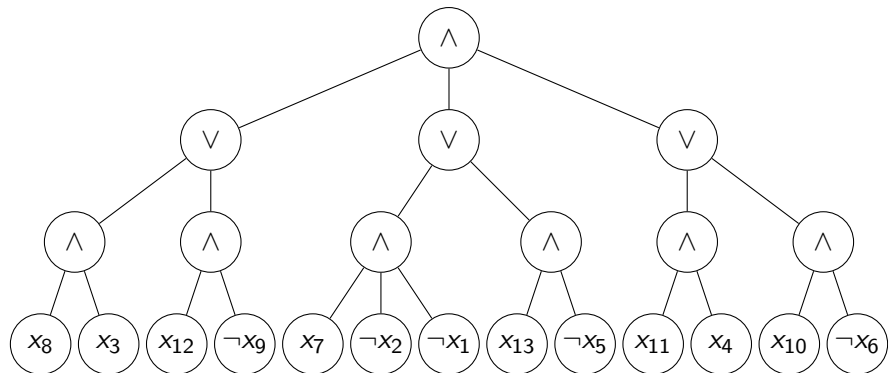
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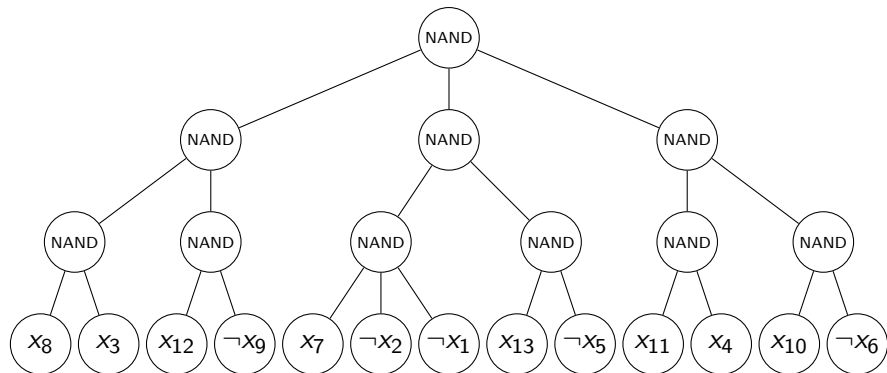
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- ▶ (Proof uses Fourier analysis)

NAND formulas



NAND formulas



Collapse under truly random restrictions (continued)

► **Corollary:** If $\mathbb{E}[f] \geq 1 - \rho$, then

$$\Pr_{R^{\text{os}}}[f|_{R^{\text{os}}} \neq 1] \leq 2\rho + \frac{1}{n^{100}}$$

Collapse under truly random restrictions (continued)

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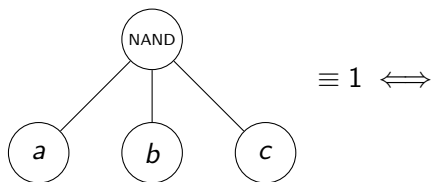
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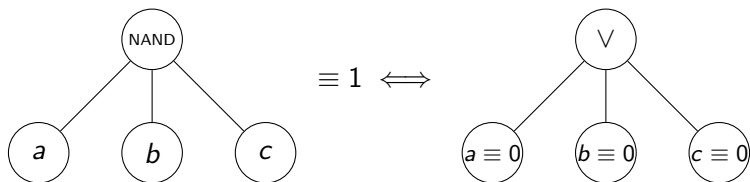
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- ▶ **Lemma:** Can be decided in **depth- d read-once \mathbf{AC}^0**

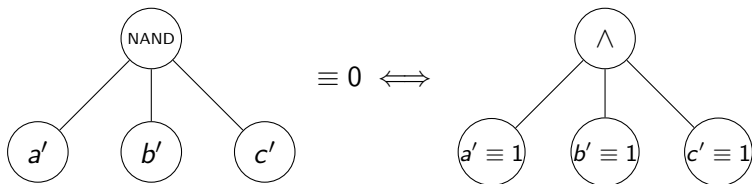
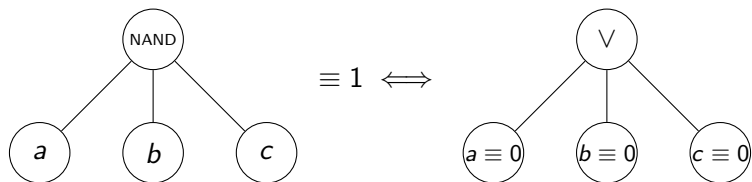
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Deciding whether $\exists f \in \mathcal{F}, f|_{\text{Res}(y,z)} \equiv 1$ (continued)

- ▶ At bottom, we get one additional layer:

$$(\text{Res}(y, z)_i \equiv b) \iff (y_i = 0 \wedge z_i = b)$$

$$(\neg \text{Res}(y, z)_i \equiv b) \iff (y_i = 0 \wedge z_i = 1 - b)$$

Deciding whether $\exists f \in \mathcal{F}, f|_{\text{Res}(y,z)} \equiv 1$ (continued)

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- ▶ At top: “ $\exists f \in \mathcal{F}$ ” is one more \vee gate (merge with top \vee gates)

Collapse under pseudorandom restrictions

- ▶ Let \mathcal{F} be a set of depth- $(d - 1)$ formulas on disjoint variables
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- ▶ Hybrid argument:

$$\Pr_{X^{os}}[\forall f \in \mathcal{F}, f|_{X^{os}} \neq 1] \approx \Pr_{R^{os}}[\forall f \in \mathcal{F}, f|_{R^{os}} \neq 1]$$

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- ▶ Assume that for every gate g in f , $\mathbb{E}[\neg g] \geq 1/\text{poly}(n)$

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- ▶ **Lemma:** With high probability over X^{os} ,

$$\Delta(f|_{X^{\text{os}}}) \leq \sqrt{\Delta(f)} \cdot \text{polylog } n$$

Illustration: $\Delta \mapsto \sqrt{\Delta}$ polylog n

Total depth $d + 1$

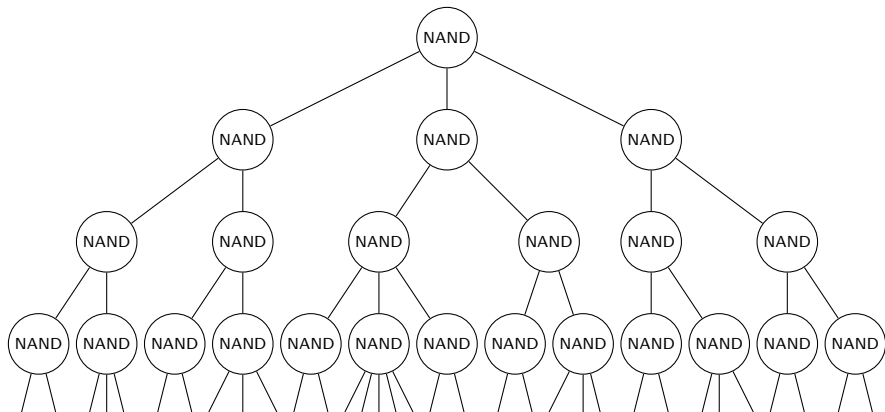
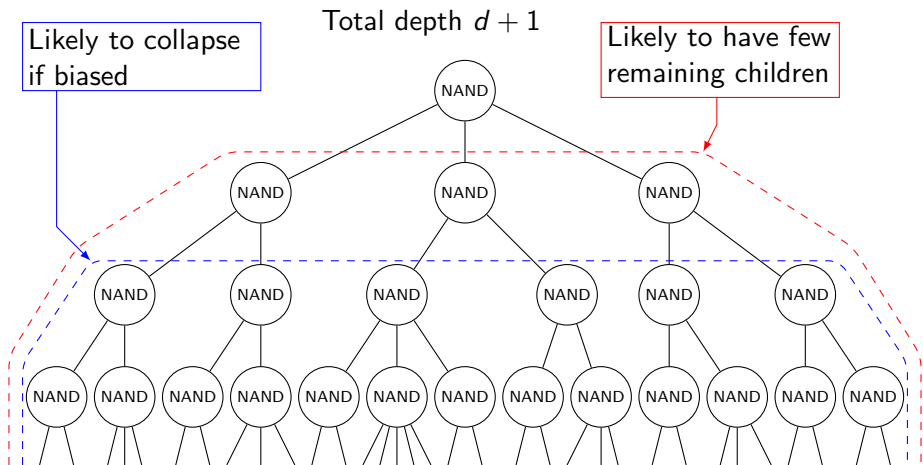


Illustration: $\Delta \mapsto \sqrt{\Delta}$ polylog n



Proof that $\Delta \mapsto \sqrt{\Delta} \text{polylog } n$

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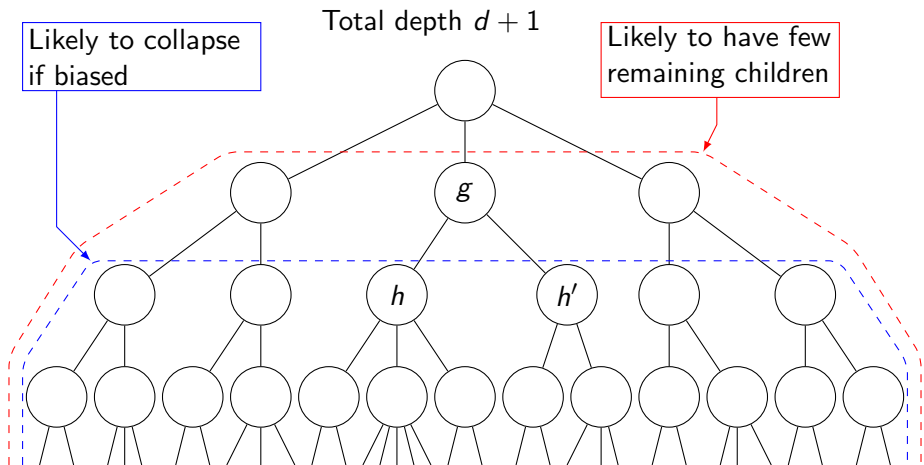
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- ▶ Consider one bucket $\mathcal{B} = \{h : \mathbb{E}[h] \approx 1 - \rho\}$

Illustration: $\Delta \mapsto \sqrt{\Delta}$ polylog n (continued)



Proof that $\Delta \mapsto \sqrt{\Delta} \text{polylog } n$ (continued)

► $\mathcal{B} = \{h : \mathbb{E}[h] \approx 1 - \rho\}$

Proof that $\Delta \mapsto \sqrt{\Delta} \text{polylog } n$ (continued)

- ▶ $\mathcal{B} = \{h : \mathbb{E}[h] \approx 1 - \rho\}$
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$$1/\text{poly}(n) \leq \mathbb{E}[\neg g]$$

Proof that $\Delta \mapsto \sqrt{\Delta}$ polylog n (continued)

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$$1/\text{poly}(n) \leq \mathbb{E}[\neg g] \leq \prod_{h \in \mathcal{B}} \mathbb{E}[h]$$

Proof that $\Delta \mapsto \sqrt{\Delta}$ polylog n (continued)

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▶ We also trivially have $|\mathcal{B}| \leq \Delta$

Proof that $\Delta \mapsto \sqrt{\Delta} \text{polylog } n$ (continued)

- ▶ Let $L = \#\{h \in \mathcal{B} : h|_{X^{os}} \neq 1\}$ (number of **living** children)

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$$\leq \frac{1}{\binom{M}{k}} \cdot \binom{|\mathcal{B}|}{k} \cdot \left(O(\rho)^k + \frac{1}{n^{200}} \right)$$

Proof that $\Delta \mapsto \sqrt{\Delta} \text{polylog } n$ (continued)

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$$\begin{aligned} \text{► } \Pr[L \geq M] &= \Pr \left[\binom{L}{k} \geq \binom{M}{k} \right] && \text{Pascal} \\ &\leq \frac{1}{\binom{M}{k}} \cdot \mathbb{E} \left[\binom{L}{k} \right] && \text{Markov} \\ &\leq \frac{1}{\binom{M}{k}} \cdot \binom{|\mathcal{B}|}{k} \cdot \left(O(\rho)^k + \frac{1}{n^{200}} \right) \\ &\leq \left(\frac{|\mathcal{B}|e}{M} \right)^k \cdot \left(O(\rho)^k + \frac{1}{n^{200}} \right) && \text{Stirling} \end{aligned}$$

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$\leq \frac{1}{\binom{M}{k}} \cdot \mathbb{E}\left[\binom{L}{k}\right]$ Markov

$\leq \frac{1}{\binom{M}{k}} \cdot \binom{|\mathcal{B}|}{k} \cdot \left(O(\rho)^k + \frac{1}{n^{200}}\right)$

$\leq \left(\frac{|\mathcal{B}|e}{M}\right)^k \cdot \left(O(\rho)^k + \frac{1}{n^{200}}\right)$ Stirling

$\leq \left(\frac{1}{\sqrt{\Delta}}\right)^k + \frac{1}{n^{200}} \cdot (\sqrt{\Delta})^k$

$\leq \frac{2}{n^{100}}$

□

Finishing proof of main lemma

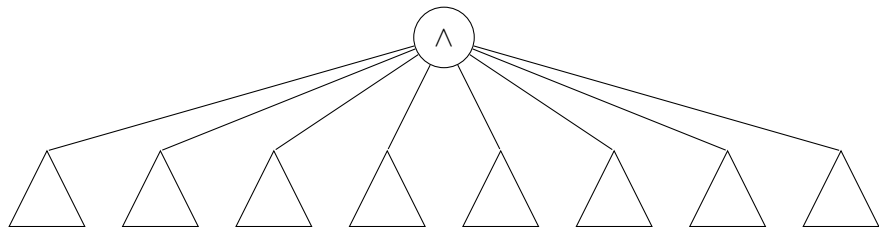
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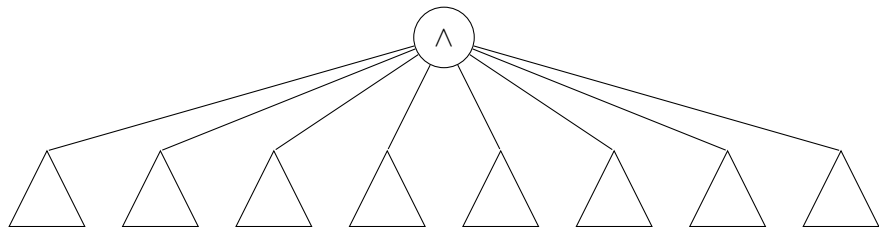
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- ▶ Therefore, after $t = O((\log \log n)^2)$ restrictions, $\Delta = \text{polylog } n$
- ▶ Total cost so far: $\tilde{O}(\log n)$ truly random bits



Final step: MRT PRG

- ▶ **Theorem** (Meka, Reingold, Tal '19): There is an explicit PRG with seed length $\tilde{O}(\log(n/\varepsilon))$ that fools functions of the form

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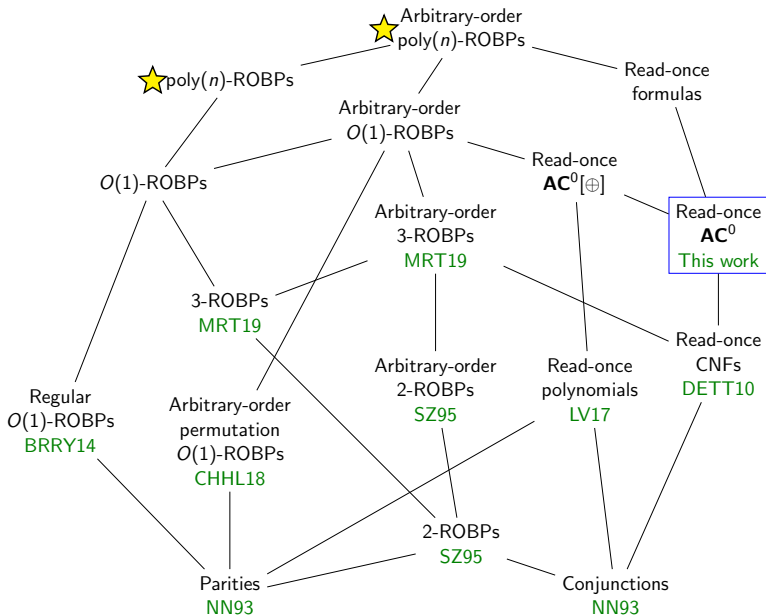
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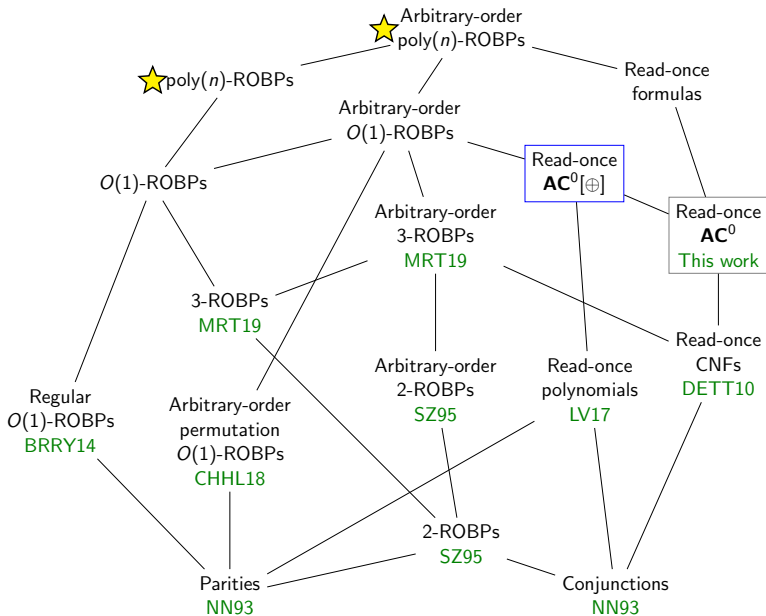
$$f = \bigwedge_{i=1}^m f_i = \sum_{S \subseteq [m]} \frac{(-1)^{|S|}}{2^m} \prod_{i \in S} (-1)^{f_i}$$

Directions for further research

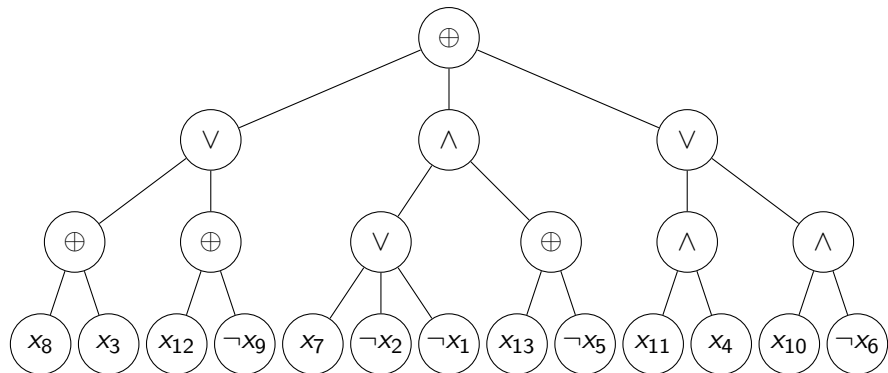
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Read-once $AC^0[\oplus]$



Fooling read-once $\mathbf{AC}^0[\oplus]$

- ▶ Natural next step toward derandomizing **BPL**

Fooling read-once $\mathbf{AC}^0[\oplus]$

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- ▶ Thanks! Questions?