

**IDENTITY TESTING & LOWER BOUNDS
FOR
READ- k OBLIVIOUS ABPS**

Ben Lee Volk

Joint with

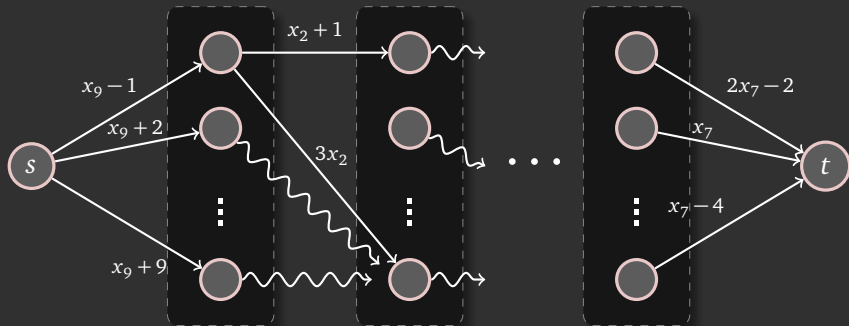
Matthew Anderson

Michael A. Forbes

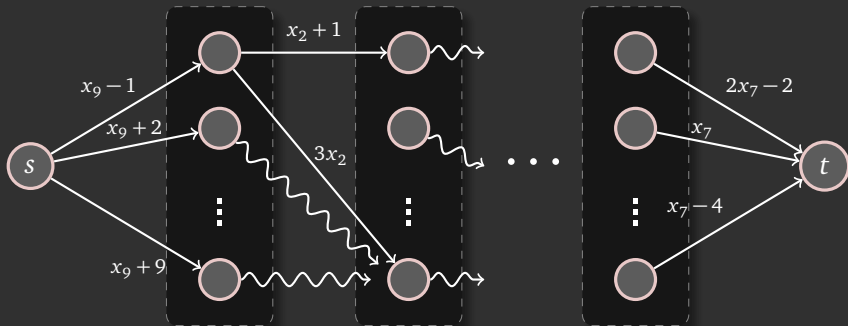
Ramprasad Saptharishi

Amir Shpilka

READ-ONCE OBLIVIOUS ABPS

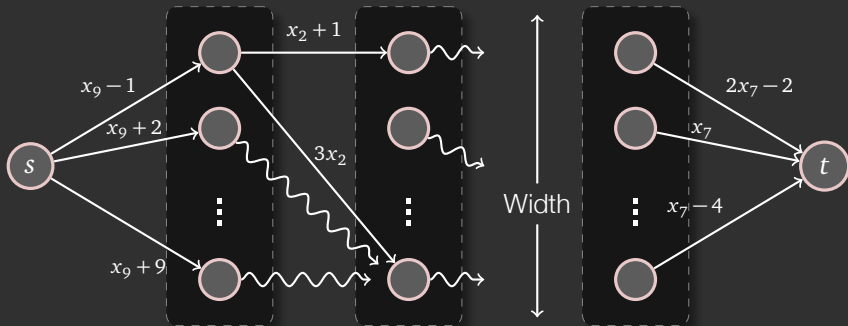


READ-ONCE OBLIVIOUS ABPS



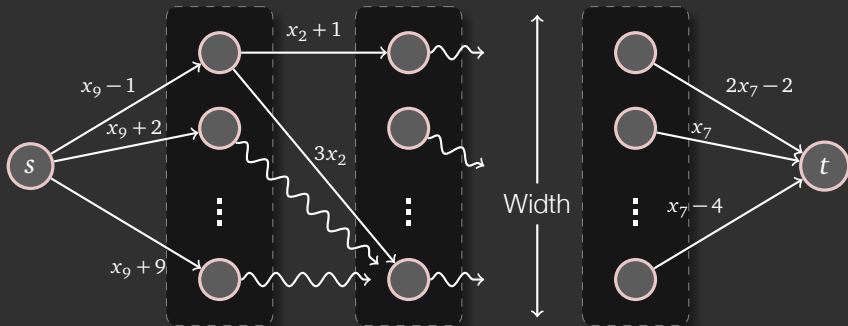
- Each $s \rightarrow t$ path computes multiplication of edge labels
- Program computes the sum of those over all $s \rightarrow t$ paths
- **Read Once**: Each var appears in one layer

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Equivalently: f is the $(1, 1)$ entry of the iterated matrix product

$$\prod_{i=1}^n M_i(x_{\pi(i)})$$

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This talk is about **read- k oblivious ABPs**.

(def: same as before except that now every variable appears in at most k layers)

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- PRG with seed length \sqrt{s} for size- s programs [**Impagliazzo-Meka-Zuckerman**]

READ- k OBLIVIOUS ABPS

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Lower Bound: There is a polynomial $f \in \text{VP}$ that requires read- k oblivious ABPs of width $\exp(n/k^k)$.

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PIT: There is a white-box* PIT algorithm for read- k oblivious ABPs, of running time $\exp(n^{1-1/2^{k-1}})$.

*only the order in which the variables appear is important

EVALUATION DIMENSION

Reminder: $f \in \mathbb{F}[x_1, \dots, x_n]$, $S \subseteq [n]$.

$$\text{eval-dim}_{S, \bar{S}}(f) = \dim \text{span} \{f|_{\mathbf{x}_S = \alpha} \mid \alpha \in \mathbb{F}^{|S|}\}.$$

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Characterizes ROABP complexity:

Theorem [Nisan]: f has ROABP of width w in variable order x_1, x_2, \dots, x_n iff for every $i \in [n]$,

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(same as rank of partial derivative matrix)

WARM-UP: 2-PASS ABP

Same as ROABP but with two “passes”:

x_1	x_2	\dots	x_{n-1}	x_n	x_1	x_2	\dots	x_{n-1}	x_n
-------	-------	---------	-----------	-------	-------	-------	---------	-----------	-------

$$f = \left(M_1^1(x_1) M_2^1(x_2) \cdots M_n^1(x_n) \cdot M_1^2(x_1) M_2^2(x_2) \cdots M_n^2(x_n) \right)_{(1,1)}$$

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Fixing $x_1 = \alpha_1, x_2 = \alpha_2$:

$$\left(N^1(\alpha_1, \alpha_2) M_3^1(x_3) \cdots M_n^1(x_n) \cdot N^2(\alpha_1, \alpha_2) M_3^2(x_3) \cdots M_n^2(x_n) \right)_{(1,1)}$$

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Fixing x_1, x_2, \dots, x_i :

$$f|_{x_{[i]}=\alpha} = \left(N^1(\alpha_1, \dots, \alpha_i) M_{i+1}^1(x_{i+1}) \cdots M_n^1(x_n) \right. \\ \left. N^2(\alpha_1, \dots, \alpha_i) M_{i+1}^2(x_{i+1}) \cdots M_n^2(x_n) \right)_{(1,1)}$$

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GENERALIZE: k -PASS ABP

Theorem: If f is computed by a width- w k -pass ABP in variable order x_1, x_2, \dots, x_n , then for every $i \in [n]$, $\text{eval-dim}_{[i], \overline{[i]}}(f) \leq w^{2k}$.

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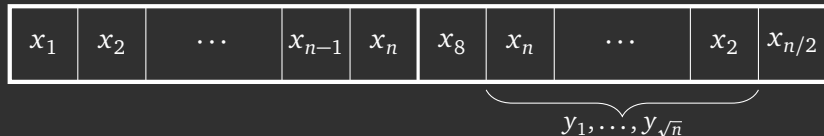
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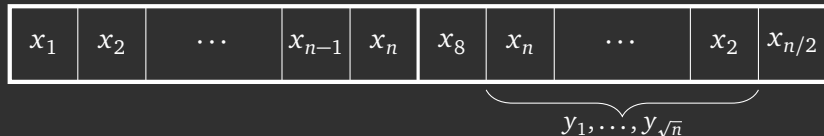
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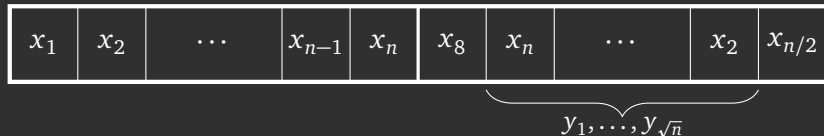
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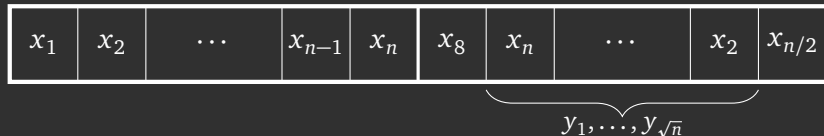


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What you get is a 2-pass ABP over \mathbf{y} vars. In other words, ignoring $\bar{\mathbf{y}}$, for every $i \in [\sqrt{n}]$, $\text{eval-dim}_{[i], [\bar{i}]}(f) \leq w^4$.

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Naturally generalizes to k passes with different orders.

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This is still not a general read- k oblivious ABP!

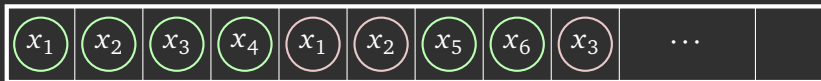
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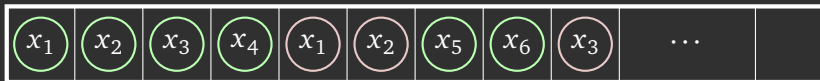
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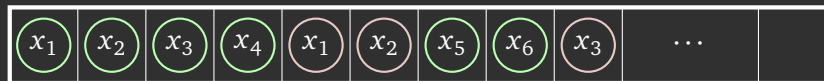


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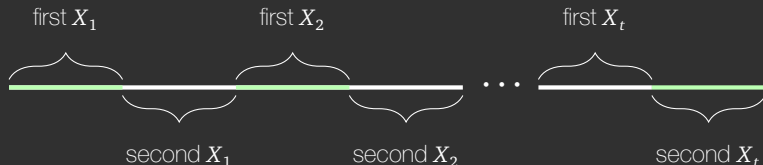
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Generalizes to read- k : apply Erdős-Szekeres to every sequence and make every pair regularly-interleaving.

Wrap-up: PIT algorithm with running time $\exp(n^{1-1/2^{k-1}})$ for read- k oblivious ABPs.

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- This is very close to being true

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Claim: We can fix $n/10$ variables and partition the remaining to subsets S, T with $|S|, |T| \geq n/k^k$ and $\text{eval-dim}_{S,T}(f) \leq w^{2k}$

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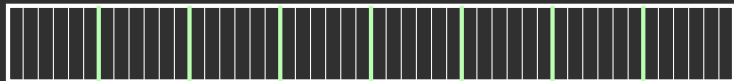
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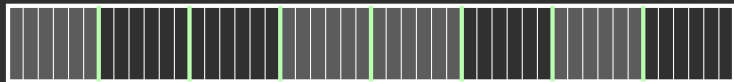


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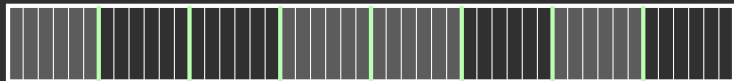


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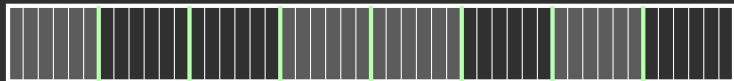
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what's left is to find a polynomial such that $\text{eval-dim}_{S,T} \geq 2^{\min\{|S|, |T|\}}$

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PIT: A white-box PIT algorithm for read- k oblivious ABPs, with running time $\exp(n^{1-1/2^{k-1}})$.

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THANK YOU