

A Near-Optimal Depth-Hierarchy Theorem for Small-Depth Multilinear Circuits

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Joint work with

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More Resources

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More power?

Power of resources

Turing machines

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Non-explicit separations.

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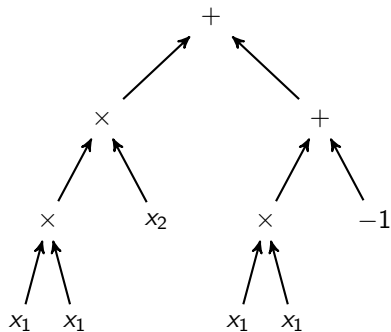
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Today we will focus on arithmetic formulas as a model of computation.

A model of computation for polynomials

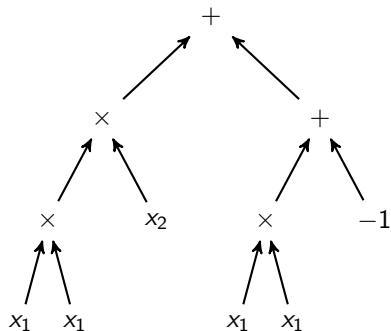
Arithmetic formula

Definition: An arithmetic formula



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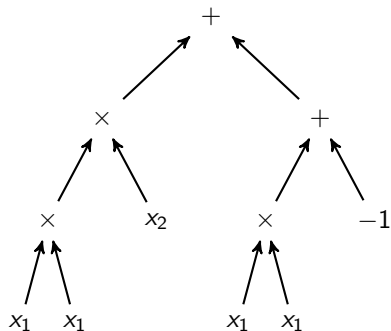
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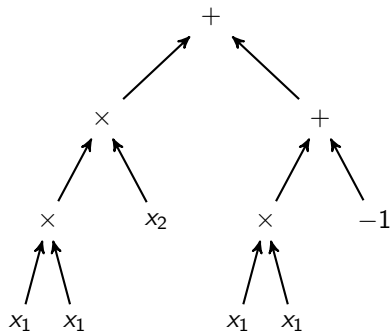
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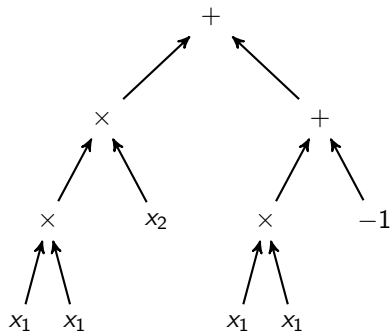
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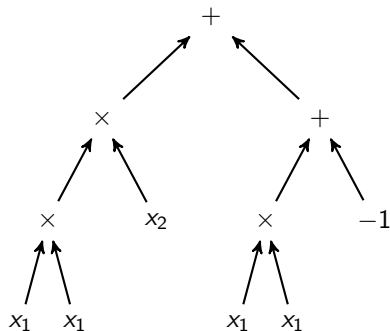
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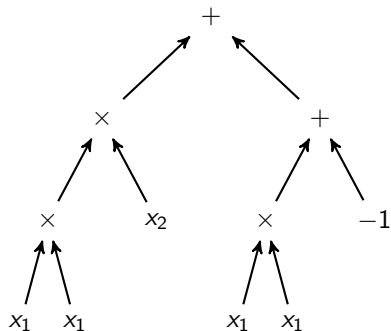
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Size and product-depth

The number of nodes in the tree is the size of the formula.

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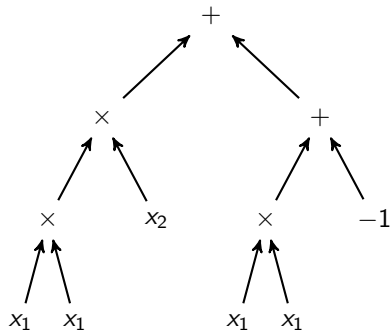
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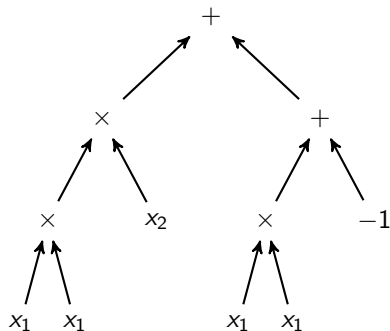
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It may not always be a succinct representation.

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This is a $\prod \sum$ formula for the same polynomial.

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Extended and appended by a line of work.

[[Raz06](#),[RSY07](#),[RY09](#),[DMPY12](#),[KV17](#)].

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This $\sum \Pi \Sigma$ realization is non-multilinear!

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[Kayal, Nair, Saha, 15] show that this is not possible.

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[Kayal, Nair, Saha, 15] show that this exponential blow-up is essential while going from $\Delta = 2$ to $\Delta = 1$.

Larger product depth $\Delta + 1 \longrightarrow \Delta$

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Open up the multiplication of summands as a sum of multiplications.

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Is the blow-up essential?

Depth hierarchy theorem

More Resources

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More power?

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More power?

More Product-depth

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Arithmetic circuit complexity world

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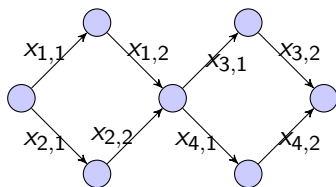


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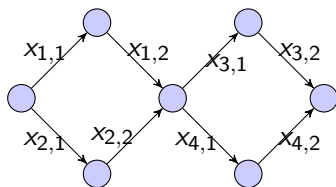


Figure: Definition of $G^{(0)}$

$$P^{(0)} = x_{1,1}x_{1,2}x_{3,1}x_{3,2} + x_{1,1}x_{1,2}x_{4,1}x_{4,2} + x_{2,1}x_{2,2}x_{3,1}x_{3,2} + x_{2,1}x_{2,2}x_{4,1}x_{4,2}.$$

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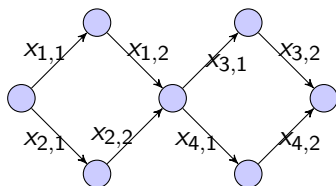


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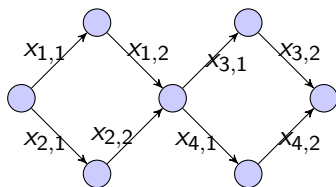


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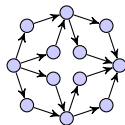


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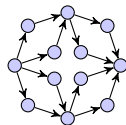
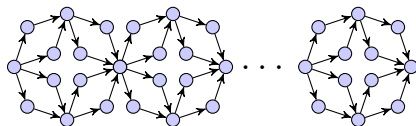


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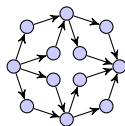
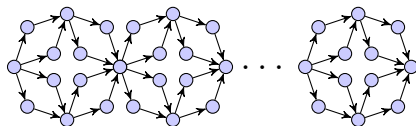


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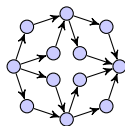
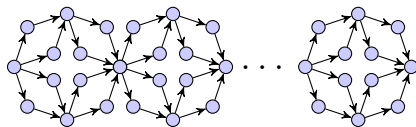


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$P^{(1)}$ is the sum of weights of all the source to sink paths in $G^{(1)}$.

$P^{(1)}$ is a polynomial over $n_1 := 8(2m)$ many variables and has product-depth 2, size $O(m) = O(n_1)$ formula.

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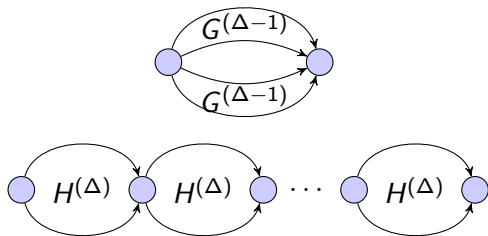


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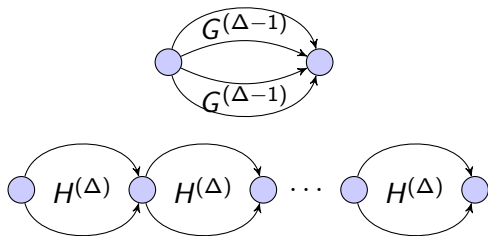


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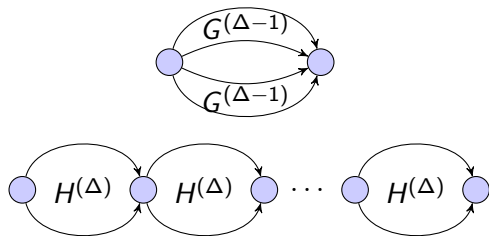


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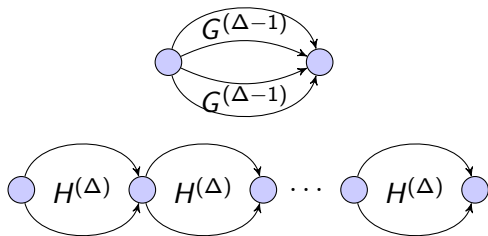


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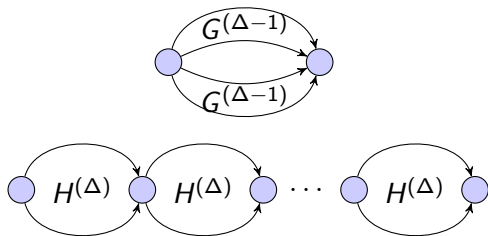


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Do similar techniques yield a non-commutative formula depth-hierarchy theorem?

Proof details

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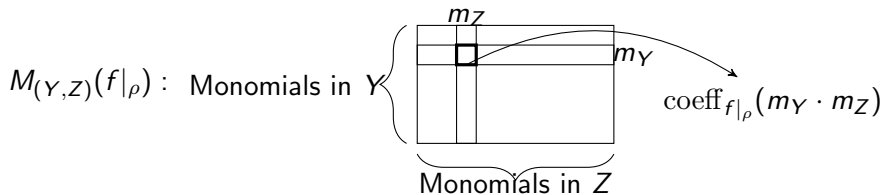
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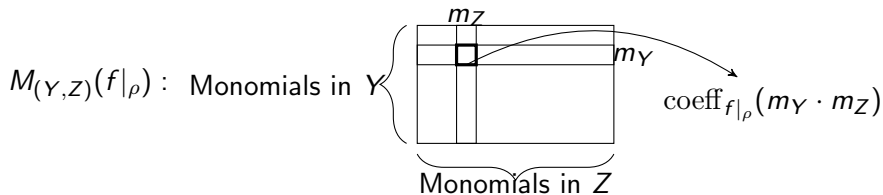
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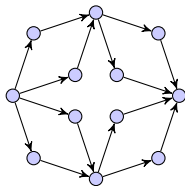
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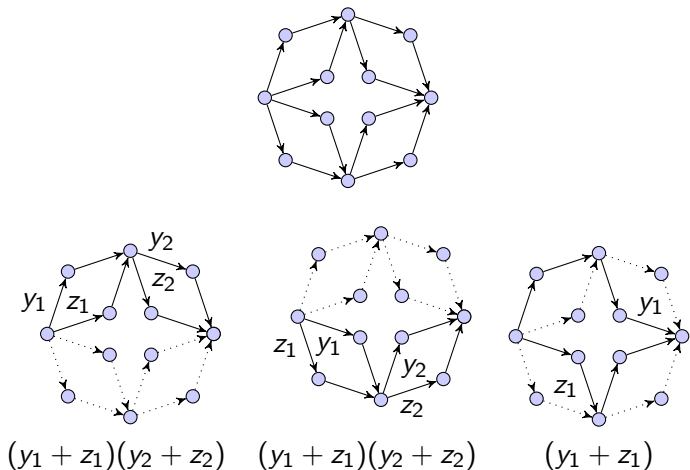
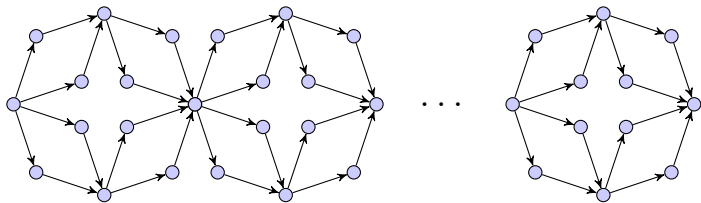


Figure: Map ρ applied to each copy of $H^{(1)}$. Edges that are not labelled have their variables set to 1. Dotted edges have their variables set to 0.

Random map ρ

Recall that $G^{(1)}$ is m copies of $H^{(1)}$.

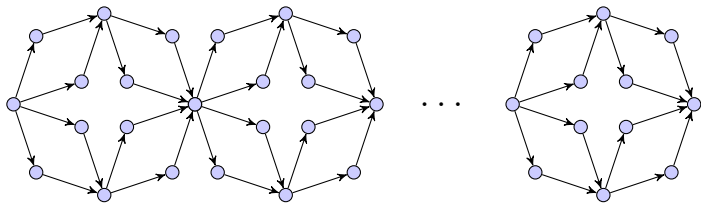


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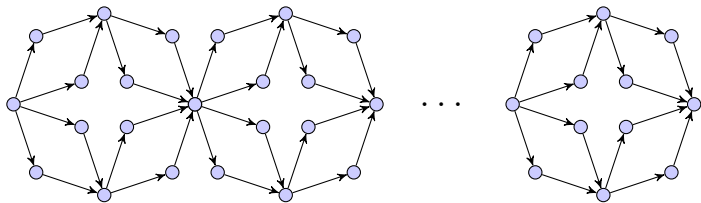


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Do similar techniques yield a non-commutative formula depth-hierarchy theorem?

Thank You!