# Demystifying the border of depth-3

Joint works with Pranjal Dutta & Prateek Dwivedi. [CCC'21, FOCS'21, FOCS'22]

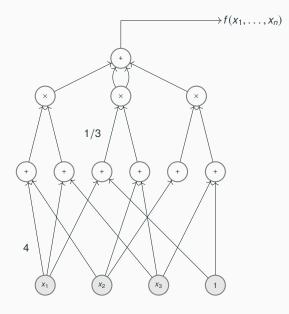
Nitin Saxena CSE, IIT Kanpur

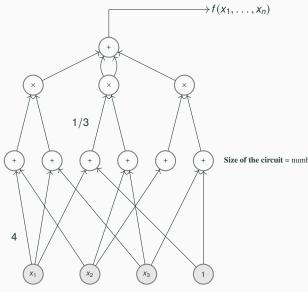
August, 2023 WAC @ Göteborg, Sweden

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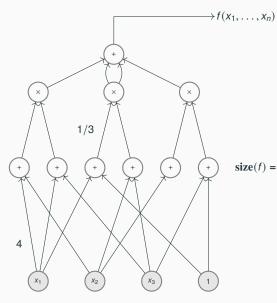
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**Basic Definitions and Terminologies** 

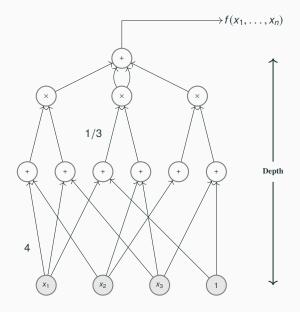


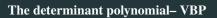


Size of the circuit = number of nodes + edges



size(f) = min size of the circuit computing f





### The determinant polynomial-VBP

□ Let  $X_s = [x_{i,j}]_{1 \le i,j \le s}$  be an  $s \times s$  matrix of distinct variables  $x_{i,j}$ . Let  $S_s := \{\pi \mid \pi : \{1, ..., s\} \longrightarrow \{1, ..., s\}$  such that  $\pi$  is bijective  $\}$ . Define

$$det_{\mathcal{S}} := det(X_{\mathcal{S}}) = \sum_{\pi \in \mathcal{S}_{\mathcal{S}}} sgn(\pi) \prod_{i=1}^{\mathcal{S}} x_{i,\pi(i)}.$$

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- □ VBP: The class VBP is defined as the set of all sequences of polynomials  $(f_n)_n$  with polynomially bounded determinantal-complexity  $dc(f_n)$ .
- □ Relates tightly to Algebraic Branching Programs ABP, or IMM: Iterated Matrix Multiplication.

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☐ The minimum dimension of the matrix  $X_s$  to compute f, is called the **permanental complexity** pc(f).

## Valiant's Conjecture- VNP

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#### Valiant's Conjecture [Valiant 1979]

 $VBP \neq VNP \& VP \neq VNP$ .

Equivalently,  $dc(perm_n)$  and  $size(perm_n)$  are both  $n^{\omega(1)}$ .

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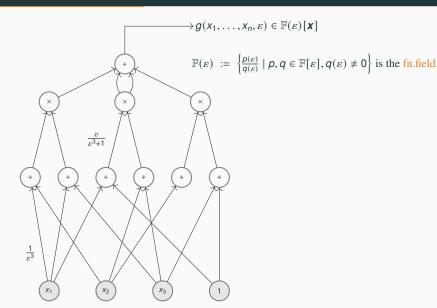
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☐ This motivates a new model: 'approximative circuit'.

# **Approximative circuits**



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, i.e. it is a polynomial of the form

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- □ Summary:  $g_0$  is **non-trivially** 'approximated' by the circuit, since  $\lim_{\varepsilon \to 0} g(\mathbf{x}, \varepsilon) = g_0$ .

# Algebraic approximation— $\overline{\text{VP}}$

### Algebraic Approximation [Bürgisser 2004]

A polynomial  $h \in \mathbb{F}[x]$  has approximative complexity s, if there is a  $g \in \mathbb{F}[\varepsilon][x]$ , of size s, and an error polynomial  $S(x, \varepsilon) \in \mathbb{F}[\varepsilon][x]$  such that

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- $\square$   $\overline{\text{size}}(h) \leq \text{size}(h) \leq \exp(\overline{\text{size}}(h)).$
- ☐ Curious eg.: Complexity of degree-**s** factor of a size-**s** polynomial? VP? VNP?

**Border Depth-3 Circuits** 

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- □ Impossibility result: The *Inner Product* polynomial  $\langle \boldsymbol{x}, \boldsymbol{y} \rangle := x_1 y_1 + x_2 y_2 + x_3 y_3$  cannot be written as a  $\Sigma^{[2]}\Pi\Sigma$  circuit,

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- $\square$  How about  $\overline{\Sigma^{[2]}\Pi\Sigma}$ ?

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### Border depth-3 fan-in 2 circuits are 'universal' [Kumar 2020]

Let P be any n-variate degree d polynomial. Then,  $P \in \Sigma^{[2]}\Pi\Sigma$ , where the first product has fanin  $\exp(n, d)$  and the second is merely constant!

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2. Consider 
$$A(\boldsymbol{x}):=\prod_{i=1}^m(1+\ell_i^d)=\prod_{i=1}^m\prod_{j=1}^d(\alpha_j+\ell_i),$$
 for  $\alpha_j\in\mathbb{C}.$ 

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 [m can be as large as  $\exp(n, d)$ ].

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4. Divide by  $\varepsilon^d$  and rearrange to get

$$P + \varepsilon^d \cdot R(\boldsymbol{x}, \varepsilon) = -\varepsilon^{-d} + \varepsilon^{-d} \cdot \prod_{i=1}^m \prod_{j=1}^d (\alpha_j + \varepsilon \cdot \ell_i) \ \in \ \Sigma^{[2]} \Pi^{[md]} \Sigma \ .$$

Proving Upper Bounds

# De-bordering $\overline{\Sigma^{[2]}\Pi\Sigma}$ circuits

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**Remark.** The result holds if one replaces the top-fanin-2 by arbitrary constant k.

$$\square$$
  $T_1 + T_2 = f(\mathbf{x}) + \varepsilon \cdot S(\mathbf{x}, \varepsilon)$ , where  $T_i \in \Pi\Sigma \in \mathbb{F}(\varepsilon)[\mathbf{x}]$ . Assume  $\deg(f) = d$ .

Grand Idea: Reduce to k = 1!

$$\Box \ T_1 + T_2 = f(\boldsymbol{x}) + \varepsilon \cdot S(\boldsymbol{x}, \varepsilon), \text{ where } T_i \in \Pi\Sigma \in \mathbb{F}(\varepsilon)[\boldsymbol{x}]. \text{ Assume deg}(f) = d.$$

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- ☐ We devise a technique called DiDIL Divide, Derive, Induct with Limit.

### k = 2 proof continued: *Di*vide and *D*erive

 $\square$  val<sub>z</sub>(·) denotes the highest power of z dividing it (= least one across monomials). E.g.,  $h = \varepsilon z + \varepsilon^{-1} z^2 x_1 = (\varepsilon z) \cdot (1 + \varepsilon^{-2} z x_1)$ . Then, val<sub>z</sub>(h) = 1.

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$$\begin{split} \Phi(f) + \varepsilon \cdot \Phi(S) &= \Phi(T_1) + \Phi(T_2) \\ \Longrightarrow \Phi(f/T_2) + \varepsilon \cdot \Phi(S/T_2) &= \Phi(T_1/T_2) + 1 \\ \Longrightarrow \partial_Z \Phi(f/T_2) + \varepsilon \cdot \partial_Z \Phi(S/T_2) &= \partial_Z \Phi(T_1/T_2) =: g_1 \; . \end{split} \tag{1}$$

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$$\begin{split} & \lim_{\varepsilon \to 0} g_1 \equiv \lim_{\varepsilon \to 0} \Pi \Sigma / \Pi \Sigma \cdot \left( \sum \mathsf{dlog}(\Sigma) \right) \mod z^d \\ & \equiv \lim_{\varepsilon \to 0} \left( \Pi \Sigma / \Pi \Sigma \right) \cdot \left( \Sigma \wedge \Sigma \right) \mod z^d \\ & \in \overline{\left( \Pi \Sigma / \Pi \Sigma \right) \cdot \left( \Sigma \wedge \Sigma \right)} \mod z^d \;. \end{split}$$

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  $\overline{C \cdot \mathcal{D}} \subseteq \overline{C} \cdot \overline{\mathcal{D}}$ . Therefore,

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- Note: Definite integration requires setting z = 0 in  $\Phi(T_1/T_2) + 1$ ; that's why we need power-series in z.

Proving Lower Bounds

→ skip the section

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- ☐ What does work (if at all!)?

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Fix any constant  $k \ge 1$ . There is an explicit n-variate and < n degree polynomial f such that f can be computed by a  $\sum_{n=0}^{\infty} |I| = \sum_{n=0}^{\infty} I_n$  circuit of size O(n);

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- ☐ Classical is about *impossibility*. While, border is about *optimality*.

Conclusion

 $\hfill \square$  ROABP core gives us many PIT results (see our two papers).

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Thank you! Questions?