## Demystifying the border of depth-3 algebraic circuits

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## Algebraic Complexity Theory

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- [P ? NP, Aronson 2011] calls GCT "The String Theory of Computer Science".


## Algebraic circuits



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$\square$ E.g. $\operatorname{dc}\left(x_{1} \cdots x_{n}\right)=n$, since

$$
x_{1} \cdots x_{n}=\operatorname{det}\left(\begin{array}{cccc}
x_{1} & 0 & \ldots & 0 \\
0 & x_{2} & \ldots & 0 \\
\vdots & \vdots & \vdots & \vdots \\
0 & 0 & \ldots & x_{n}
\end{array}\right)
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Connections: Linear algebra, Volume, counting planar matchings.

## ‘Hard' polynomials?

$\square$ Are there hard polynomial families $\left(f_{n}\right)_{n}$ such that it cannot be computed by an $n^{c}$-size circuit, for every constant $c$ ? i.e. $\operatorname{size}\left(f_{n}\right)=n^{\omega(1)}$ ?

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## VNP = "hard to compute?" [Valiant 1979]

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## Valiant's Conjecture [Valiant 1979]

$\operatorname{VBP} \neq \mathrm{VNP} \& \mathrm{VP} \neq \mathrm{VNP}$. Equivalently, dc $($ perm $n)$ and size $($ perm $n)$ are both $n^{\omega(1)}$.

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> Assuming GRH (Generalized Riemann hypothesis), the results hold over $\mathbb{C}$ as well.

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- Can there be 'algebraic natural proofs' to prove VP $\neq \mathrm{VNP}$ ? Some answers: [Chatterjee-Kumar-Ramya-Saptharishi-Tengse 2020, Kumar-Ramya-Saptharishi-Tengse 2020].


## Border Complexity and GCT

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DR $(h) \leq r$ is denoted as $h \in \Sigma^{[r]} \wedge \Sigma$ (homogeneous depth-3 diagonal circuits).

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- Such $f=b^{2}-4 a c$ is sometimes called a 'polynomial obstruction' or a 'separating polynomial'.
$\square X_{1}$ is a closed set. If there are three sequences $\left(a_{n}, b_{n}, c_{n}\right)$ such that $a_{n} \rightarrow a, b_{n} \rightarrow b, c_{n} \rightarrow c$, i.e. limits exist, such that $\left(a_{n}, b_{n}, c_{n}\right) \in X_{1}$, then $(a, b, c) \in X_{1}$.


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=x^{2} y+\epsilon x y^{2}+\frac{\epsilon^{2}}{3} y^{3} \xrightarrow{\epsilon \rightarrow 0} x^{2} y=: h \quad \text { (coefficient-wise). }
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=x^{2} y+\epsilon x y^{2}+\frac{\epsilon^{2}}{3} y^{3} \xrightarrow{\epsilon \rightarrow 0} x^{2} y=: h \quad \text { (coefficient-wise). }
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- Note: $\operatorname{WR}\left(h_{\epsilon}\right) \leq 2$, for any fixed non-zero $\epsilon$. $\operatorname{But} \operatorname{WR}(h)=3$ !


## Approximation helps

- Example: $\operatorname{WR}\left(x^{2} y\right) \leq 3$, because

$$
x^{2} y=\frac{1}{6} \cdot(x+y)^{3}-\frac{1}{6} \cdot(x-y)^{3}-\frac{1}{3} \cdot y^{3} .
$$

- Prove: $\operatorname{WR}\left(x^{2} y\right)=3$.

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The subtlety is gone: $X_{n}:=\{h \mid \overline{\mathrm{WR}}(h) \leq n\}$, is now a closed set.
On to proving lower bounds: To show $\overline{\mathrm{WR}}(p)>n$, for some $p$, it suffices to show that $p \notin X_{n}$, i.e. find a continuous function $f$ that vanishes on $X_{n}$ but not on $p$.

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- Yes! Via 'approximative circuits'.


## Approximative circuits (continued)



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$>g(\boldsymbol{x}, \epsilon) \in \mathbb{F}\left[x_{1}, \ldots, x_{n}, \epsilon\right]$, i.e. it is a polynomial of the form

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Summary: $g_{0}$ is really something non-trivial and being 'approximated' by the circuit since $\lim _{\epsilon \rightarrow 0} g(\boldsymbol{x}, \epsilon)=g_{0}$.

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- Symmetry-characterization avoids the Razborov-Rudich barrier: Very few functions are symmetry-characterized, so symmetry-characterization violates the largeness criterion!


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$\square$ Upper bounds and lower bounds are dual to each other.
Further potential applications in identity testing and understanding its 'robustness'.

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They compute polynomials of the form $\sum_{i=1}^{k} \prod_{j=1}^{d} \ell_{i j}$, where $\ell_{i j}$ are linear polynomials (i.e. $a_{0}+a_{1} x_{1}+\ldots+a_{n} x_{n}$, for $a_{i} \in \mathbb{F}$ ).

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$\square$ No. E.g. the Inner Product polynomial $\langle\boldsymbol{x}, \boldsymbol{y}\rangle=x_{1} y_{1}+\ldots+x_{k+1} y_{k+1}$ cannot be written as a $\Sigma^{[k]} \Pi^{[d]} \Sigma$ circuit, regardless of the product fan-in $d$ !

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

Let $P$ be any homogeneous $n$-variate degree $d$ polynomial. Then, $P \in \overline{\Sigma^{[2]} \Pi^{[D]} \Sigma}$, where $D:=\exp (n, d)$.

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

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Infinitely many factorizations may give infinitely many limits.

## Proof sketch for $k=2$

- $T_{1}+T_{2}=f(\boldsymbol{x})+\epsilon \cdot S(\boldsymbol{x}, \epsilon)$, where $T_{i} \in \Pi \Sigma \in \mathbb{F}(\epsilon)[\boldsymbol{x}]$. Assume $\operatorname{deg}(f)=d$.


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$\square$ We devise a technique called DiDIL - Divide, Derive, Interpolate with Limit.

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f+\epsilon \cdot S & =T_{1}+T_{2} \\
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\Longrightarrow \Phi(f) / \tilde{T}_{2}+\epsilon \cdot \Phi(S) / \tilde{T}_{2} & =\epsilon^{a_{2}}+\Phi\left(T_{1}\right) / \tilde{T}_{2} \\
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\partial_{z}\left(\Phi\left(T_{1}\right) / \tilde{T}_{2}\right) & =\Phi\left(T_{1}\right) / \tilde{T}_{2} \cdot \operatorname{dlog}\left(\Phi\left(T_{1}\right) / \tilde{T}_{2}\right) \\
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$\square$ Here $\Sigma$ means just a linear polynomial $\ell$.

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Derandomizing border depth-3 circuits

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D Derandomizing PIT, for restricted cases, has many algorithmic applications:
$>$ Graph Theory [Lovasz' 79], [Fenner-Gurjar-Theirauf' 19]
$>$ Primality Testing [Agrawal-Kayal-Saxena'04].

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## Theorem 2 (Derandomizing polynomal-sized depth-3 top-fanin- $k$ circuits) [Dutta-Dwivedi-Saxena 2021]

There exists an explicit quasipolynomial-time $\left(s^{O(\log \log s)}\right)$ hitting set for size-s $\Sigma^{[k]} \Pi \Sigma$ circuits, for any constant $k$.

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Thank you \& stay safe!

