

Kronecker products and iterated matrix multiplication

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- 1 Motivation
- 2 Iterated matrix multiplication
- 3 Generalized Cayley–Hamilton
- 4 Tensor products and Kronecker products
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Motivation

- Valiant (1979): determinant vs permanent question: Given m , what is the smallest n such that $\exists y_{i,j} \in \mathbb{F} \cup \{x_{i,j}\}$ with

$$\det \begin{pmatrix} y_{1,1} & \cdots & y_{1,n} \\ \vdots & \ddots & \vdots \\ y_{n,1} & \cdots & y_{n,n} \end{pmatrix} = \sum_{\pi \in \mathfrak{S}_m} \underbrace{\prod_{i=1}^m x_{i,\pi(i)}}_{=: \text{per}_m}$$

$$\det \begin{pmatrix} 0 & x_{11} & x_{12} & x_{13} & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & x_{22} & x_{23} & 0 \\ 0 & 0 & 1 & 0 & x_{21} & 0 & x_{23} \\ 0 & 0 & 0 & 1 & 0 & x_{21} & x_{22} \\ x_{33} & 0 & 0 & 0 & 1 & 0 & 0 \\ x_{32} & 0 & 0 & 0 & 0 & 1 & 0 \\ x_{31} & 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix} = \text{per}_3.$$

- Conj.: n grows superpolynomially in m .
- Known to be false in characteristic 2 because $\text{per}_m = \det_m$. Hence, cannot be proved via field-independent methods
- Replacements per_m that work over every field come from combinatorics (Hamiltonian cycles), not from algebra
- Mulmuley–Sohoni (2001): Geometric complexity theory (GCT) to resolve det vs per
- Kadish–Landsberg (2012), Bürgisser–I–Panova (2016): Some GCT conjectures are false, since $\deg(\det_n) \neq \deg(\text{per}_m)$
- Approaches via the width of polynomials have some issues (Gesmundo 2015 + my AGATES Warsaw talk 2022)

We fix this. We abandon determinantal complexity.

- Works over all commutative semirings.
 - Works for polynomials, tensors, and exterior algebra.
- (this is exactly the level of generality in `mathlib`)

Nice corollaries of our work (unintended):

- Generalization of the Cayley–Hamilton theorem, new record ABP construction for the determinant (Mahajan–Vinay 1997), answer a question from Mahajan–Vinay (1999) about combinatorial algorithms for the determinant
- Gurvits' (2004, 2005) VNP-hardness results for the mixed discriminant and Cayley's first hyperdeterminant hold over every commutative ring (Gurvits had fields of $\text{char} \neq 2$). Also higher order permanents and higher order hyperpfaffians.
- Christandl–Fawzi–Ta–Zuiddam (2022) asymptotic spectrum works for polynomials in arbitrary characteristic

Let

$$X_m^{(k)} := \begin{pmatrix} x_{1,1}^{(k)} & \cdots & x_{1,m}^{(k)} \\ \vdots & \ddots & \vdots \\ x_{m,1}^{(k)} & \cdots & x_{m,m}^{(k)} \end{pmatrix}.$$

For fixed m and d , determine the smallest n such that

$$(X_m^{(1)} \cdots X_m^{(d)})^{\boxtimes 2} \leq X_n^{(1)} \cdots X_n^{(d)}$$

For fixed d , the smallest n grows polynomially in m , i.e., $n = O(m^{\xi_d + \varepsilon})$.

Conjecture: The sequence $(\xi_d)_{d \in \mathbb{N}}$ is unbounded.

Implies Valiant's conjecture, and is implied by Bläser–Engels' conjecture.

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Let f be a homogeneous degree d polynomial.

The **Waring rank** $\text{WR}(f)$ is defined as the smallest r such that $\exists$ linear forms with

$$\begin{aligned} f &= (\ell_1 \ \ell_2 \ \cdots \ \ell_r) \begin{pmatrix} \ell_1 & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \ell_r \end{pmatrix} \begin{pmatrix} \ell_1 & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \ell_r \end{pmatrix} \cdots \begin{pmatrix} \ell_1 & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \ell_r \end{pmatrix} \begin{pmatrix} \ell_1 \\ \vdots \\ \ell_r \end{pmatrix} \\ &= \text{tr} \left(\begin{pmatrix} \ell_1 & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \ell_r \end{pmatrix} \cdots \begin{pmatrix} \ell_1 & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \ell_r \end{pmatrix} \right) \end{aligned}$$

The **Chow rank** $\text{CR}(f)$ is defined as the smallest r such that $\exists$ linear forms with

$$\begin{aligned} f &= (\ell_1^{(1)} \ \ell_2^{(1)} \ \cdots \ \ell_r^{(1)}) \begin{pmatrix} \ell_1^{(2)} & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \ell_r^{(2)} \end{pmatrix} \begin{pmatrix} \ell_1^{(3)} & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \ell_r^{(3)} \end{pmatrix} \cdots \begin{pmatrix} \ell_1^{(d-1)} & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \ell_r^{(d-1)} \end{pmatrix} \begin{pmatrix} \ell_1^{(d)} \\ \vdots \\ \ell_r^{(d)} \end{pmatrix} \\ &= \text{tr} \left(\begin{pmatrix} \ell_1^{(1)} & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \ell_r^{(1)} \end{pmatrix} \cdots \begin{pmatrix} \ell_1^{(d)} & 0 & 0 \\ 0 & \ddots & 0 \\ 0 & 0 & \ell_r^{(d)} \end{pmatrix} \right) \end{aligned}$$

Mateusz Michałek on Tuesday: “There is nothing stopping us from dropping the word *diagonal*.”
So we now fill **all** matrix entries with linear forms!

Two variants:

- The smallest r is called the **width**:

$$f = (\ell_{1,1}^{(1)} \ell_{1,2}^{(1)} \cdots \ell_{1,r}^{(1)}) \begin{pmatrix} \ell_{1,1}^{(2)} & \cdots & \ell_{1,r}^{(2)} \\ \vdots & \ddots & \vdots \\ \ell_{r,1}^{(2)} & \cdots & \ell_{r,r}^{(2)} \end{pmatrix} \begin{pmatrix} \ell_{1,1}^{(3)} & \cdots & \ell_{1,r}^{(3)} \\ \vdots & \ddots & \vdots \\ \ell_{r,1}^{(3)} & \cdots & \ell_{r,r}^{(3)} \end{pmatrix} \cdots \begin{pmatrix} \ell_{1,1}^{(d-1)} & \cdots & \ell_{1,r}^{(d-1)} \\ \vdots & \ddots & \vdots \\ \ell_{r,1}^{(d-1)} & \cdots & \ell_{r,r}^{(d-1)} \end{pmatrix} \begin{pmatrix} \ell_{1,1}^{(d)} \\ \vdots \\ \ell_{1,r}^{(d)} \end{pmatrix}$$

- The smallest r is the width's little cousin ($\leq w(f)$):

$$f = \text{tr} \left(\begin{pmatrix} \ell_{1,1}^{(1)} & \cdots & \ell_{1,r}^{(1)} \\ \vdots & \ddots & \vdots \\ \ell_{r,1}^{(1)} & \cdots & \ell_{r,r}^{(1)} \end{pmatrix} \cdots \begin{pmatrix} \ell_{1,1}^{(d)} & \cdots & \ell_{1,r}^{(d)} \\ \vdots & \ddots & \vdots \\ \ell_{r,1}^{(d)} & \cdots & \ell_{r,r}^{(d)} \end{pmatrix} \right)$$

- Taking traces **looks** nice, but:
the noncommutative variant is not Zariski-closed, and **CAVEAT** when working field-independently!

Iterated matrix multiplication

For $f \in \text{Sym}^d V \cup \bigwedge^d V \cup \bigotimes^d V$, the **width** $w(f)$ is defined as the smallest n such that $\exists$ homog. lin. $\ell_{i,j}^{(k)}$ with

$$f = (\ell_{1,1}^{(1)} \cdots \ell_{1,n}^{(1)}) \begin{pmatrix} \ell_{1,1}^{(2)} & \cdots & \ell_{1,n}^{(2)} \\ \vdots & \ddots & \vdots \\ \ell_{n,1}^{(2)} & \cdots & \ell_{n,n}^{(2)} \end{pmatrix} \begin{pmatrix} \ell_{1,1}^{(3)} & \cdots & \ell_{1,n}^{(3)} \\ \vdots & \ddots & \vdots \\ \ell_{n,1}^{(3)} & \cdots & \ell_{n,n}^{(3)} \end{pmatrix} \cdots \begin{pmatrix} \ell_{1,1}^{(d-1)} & \cdots & \ell_{1,n}^{(d-1)} \\ \vdots & \ddots & \vdots \\ \ell_{n,1}^{(d-1)} & \cdots & \ell_{n,n}^{(d-1)} \end{pmatrix} \begin{pmatrix} \ell_{1,1}^{(d)} \\ \ell_{2,1}^{(d)} \\ \vdots \\ \ell_{n,1}^{(d)} \end{pmatrix}$$

Nisan's 1990 result for $f \in \bigotimes^d V$

$$w(f) = \max\{\text{rk}(\mathcal{F}_j(f)) \mid 1 \leq j \leq d-1\}$$

where $\mathcal{F}_j(f)$ is f interpreted as a matrix in $(\bigotimes^j V) \otimes (\bigotimes^{d-j} V)$.

Forbes' observation 2016: For tensors we have width = border-width.

This does not hold when working with the trace instead.

This definition works for arbitrary commutative semirings.

Bläser–Dörfler–I (2020), “De-bordering”: Over $\mathbb{C}$, for a homogeneous polynomial f , we have $\underline{\text{WR}}(f) \geq w(f)$.

Bürgisser–Cucker–Lairez (2020): width to describe structured systems of polynomial eqns, and solve Smale's 17th problem.

Restrictions

- $w(f)$ is defined as the smallest n such that $\exists$ homog. lin. $\ell_{i,j}^{(k)}$ such that

$$f = (\ell_{1,1}^{(1)} \cdots \ell_{1,n}^{(1)}) \begin{pmatrix} \ell_{1,1}^{(2)} & \cdots & \ell_{1,n}^{(2)} \\ \vdots & \ddots & \vdots \\ \ell_{n,1}^{(2)} & \cdots & \ell_{n,n}^{(2)} \end{pmatrix} \begin{pmatrix} \ell_{1,1}^{(3)} & \cdots & \ell_{1,n}^{(3)} \\ \vdots & \ddots & \vdots \\ \ell_{n,1}^{(3)} & \cdots & \ell_{n,n}^{(3)} \end{pmatrix} \cdots \begin{pmatrix} \ell_{1,1}^{(d-1)} & \cdots & \ell_{1,n}^{(d-1)} \\ \vdots & \ddots & \vdots \\ \ell_{n,1}^{(d-1)} & \cdots & \ell_{n,n}^{(d-1)} \end{pmatrix} \begin{pmatrix} \ell_{1,1}^{(d)} \\ \ell_{2,1}^{(d)} \\ \vdots \\ \ell_{n,1}^{(d)} \end{pmatrix}$$

- We write $f \leq h$ if $f = h(\ell_1, \dots, \ell_N)$ for homogeneous linear ℓ_i .

$$\Gamma_{n,d} := (x_{1,1}^{(1)} \cdots x_{1,n}^{(1)}) \begin{pmatrix} x_{1,1}^{(2)} & \cdots & x_{1,n}^{(2)} \\ \vdots & \ddots & \vdots \\ x_{n,1}^{(2)} & \cdots & x_{n,n}^{(2)} \end{pmatrix} \begin{pmatrix} x_{1,1}^{(3)} & \cdots & x_{1,n}^{(3)} \\ \vdots & \ddots & \vdots \\ x_{n,1}^{(3)} & \cdots & x_{n,n}^{(3)} \end{pmatrix} \cdots \begin{pmatrix} x_{1,1}^{(d-1)} & \cdots & x_{1,n}^{(d-1)} \\ \vdots & \ddots & \vdots \\ x_{n,1}^{(d-1)} & \cdots & x_{n,n}^{(d-1)} \end{pmatrix} \begin{pmatrix} x_{1,1}^{(d)} \\ x_{2,1}^{(d)} \\ \vdots \\ x_{n,1}^{(d)} \end{pmatrix}$$

$$w(f) \leq n \text{ iff } f \leq \Gamma_{n,d}$$

For $f \in \text{Sym}^d \mathbb{C}^N$ let

$$\text{stab}(f) = \{g \in \text{GL}_N \mid gf = f\}.$$

These all have reductive stabilizers: $x_1 x_2 \cdots x_d$, $x_1^d + x_2^d + \cdots + x_d^d$, $\det_n$, per_m

BUT $\text{stab}(\Gamma_{n,d})$ is not reductive!

$$\Gamma_{n,d} = (x_{1,1}^{(1)} \cdots x_{1,r}^{(1)}) \begin{pmatrix} x_{1,1}^{(2)} & \cdots & x_{1,r}^{(2)} \\ \vdots & \ddots & \vdots \\ x_{r,1}^{(2)} & \cdots & x_{r,r}^{(2)} \end{pmatrix} \begin{pmatrix} x_{1,1}^{(3)} & \cdots & x_{1,r}^{(3)} \\ \vdots & \ddots & \vdots \\ x_{r,1}^{(3)} & \cdots & x_{r,r}^{(3)} \end{pmatrix} \cdots \begin{pmatrix} x_{1,1}^{(d-1)} & \cdots & x_{1,r}^{(d-1)} \\ \vdots & \ddots & \vdots \\ x_{r,1}^{(d-1)} & \cdots & x_{r,r}^{(d-1)} \end{pmatrix} \begin{pmatrix} x_{1,1}^{(d)} \\ x_{2,1}^{(d)} \\ \vdots \\ x_{r,1}^{(d)} \end{pmatrix}$$

$$\Xi_{n,d} := \begin{pmatrix} x_{1,1}^{(1)} & \cdots & x_{1,n}^{(1)} \\ \vdots & \ddots & \vdots \\ x_{n,1}^{(1)} & \cdots & x_{n,n}^{(1)} \end{pmatrix} \begin{pmatrix} x_{1,1}^{(2)} & \cdots & x_{1,n}^{(2)} \\ \vdots & \ddots & \vdots \\ x_{n,1}^{(2)} & \cdots & x_{n,n}^{(2)} \end{pmatrix} \cdots \begin{pmatrix} x_{1,1}^{(d)} & \cdots & x_{1,n}^{(d)} \\ \vdots & \ddots & \vdots \\ x_{n,1}^{(d)} & \cdots & x_{n,n}^{(d)} \end{pmatrix}$$

	Zariski-closed End-orbit	reductive stabilizer
$\Gamma_{n,d}$	✓	☹
$\text{tr}(\Xi_{n,d})$	☹	✓
$\Xi_{n,d}$	✓	✓

$$\Xi_{n,d} \in \mathbb{C}^n \otimes \text{Sym}^d \mathbb{C}^{n^2} \otimes \mathbb{C}^{n*}$$

$$\text{stab}_{\text{GL}_n \times \text{GL}_{n^2} \times \text{GL}_n}(\Xi_{n,d}) = \text{GL}_n^{d+1}$$

- $x_1^d + x_2^d = \prod_{k=0}^{d-1} (x_1 - \zeta_k x_2)$, where $\zeta_k = e^{(2k+1)\pi i/d}$
- $w(x_1^d + \cdots + x_n^d) \leq \lceil \frac{n}{2} \rceil$
- Kumar 2019: $w(x_1^d + \cdots + x_n^d) = \lceil \frac{n}{2} \rceil$
- Easy to see:

$$w \left(\begin{pmatrix} x_1^d & 0 & \cdots & 0 \\ 0 & x_2^d & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ 0 & \cdots & 0 & x_n^d \end{pmatrix} \right) = n.$$

Gradient and Hessian

Giorgio Ottaviani mentioned yesterday:

- $\nabla(h \boxtimes f) = \nabla(h) \boxtimes \nabla(f)$
- $\text{Hess}(h \boxtimes f) = \text{Hess}(h) \boxtimes \text{Hess}(f)$

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Computing Determinants

$$X_n := \begin{pmatrix} x_{1,1} & \cdots & x_{1,n} \\ \vdots & & \vdots \\ x_{n,1} & \cdots & x_{n,n} \end{pmatrix}$$

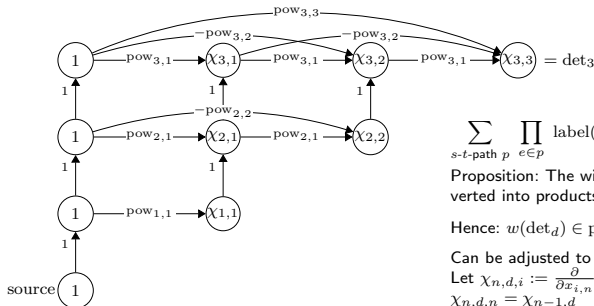
$$\det(X_n + \text{diag}(t, \dots, t)) = \sum_{d \geq 0} \chi_{n,d} \cdot t^{n-d}$$

Strengthening of the Cayley–Hamilton theorem

For all $n, d \in \mathbb{N}$ we have $(\nabla \chi_{n,d+1})^T = \sum_{i=0}^d (-1)^i \chi_{n,d-i} (X_n)^i$.

Take the bottom-right entry: $\text{pow}_{n,i} := [(X_n)^i]_{n,n}$ $\chi_{n-1,d} = \sum_{i=0}^d (-1)^i \chi_{n,d-i} \text{pow}_{n,i}$. Rearrange:

$$\chi_{n,d} = \chi_{n-1,d} + \sum_{i=1}^d (-1)^{i+1} \chi_{n,d-i} \text{pow}_{n,i}.$$



$$\sum_{s-t\text{-path } p} \prod_{e \in p} \text{label}(e) = \det_d$$

Proposition: The width of such graph constructions can be converted into products of matrices of the same width.

Hence: $w(\det_d) \in \text{poly}(d)$

Can be adjusted to beat [Mahajan–Vinay 1997]:

Let $\chi_{n,d,i} := \frac{\partial}{\partial x_{i,n}} \chi_{n,d+1}$.

$$\chi_{n,d,n} = \chi_{n-1,d}$$

$$\chi_{n,d,i} = - \sum_{j=1}^n \chi_{n,d-1,j} \cdot x_{j,i} \text{ for } i < n.$$

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- Tensor product $\otimes$ increases the tensor order:

▶ $\mathbb{F}^n \otimes \mathbb{F}^m = \mathbb{F}^{n \times m}$ matrices.

▶ $\mathbb{F}^n \otimes \mathbb{F}^m \otimes \mathbb{F}^\ell = \mathbb{F}^{n \times m \times \ell}$ order 3 tensors.

- Kronecker product $\boxtimes$ preserves the tensor order:

▶ $\mathbb{F}^n \boxtimes \mathbb{F}^m = \mathbb{F}^{nm}$.

▶ $\mathbb{F}^{n_1 \times n_2} \boxtimes \mathbb{F}^{m_1 \times m_2} = \mathbb{F}^{(n_1 m_1) \times (n_2 m_2)}$:

$$A = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \quad B = \begin{pmatrix} e & f \\ g & h \end{pmatrix}$$

$$A \boxtimes B = \begin{pmatrix} aB & bB \\ cB & dB \end{pmatrix} = \begin{pmatrix} ae & af & be & bf \\ ag & ah & bg & bh \\ ce & cf & de & df \\ cg & ch & dg & dh \end{pmatrix}$$

- ▶ $\boxtimes$ works for every fixed degree d .
- ▶ $\boxtimes$ = “ $\otimes$ and then flatten”

Kronecker product and the matrix multiplication exponent

- The matrix multiplication tensor

$$M_n = \sum_{i,j,k \in \{1, \dots, n\}} (e_i \boxtimes e_j) \otimes (e_j \boxtimes e_k) \otimes (e_k \boxtimes e_i) \in \mathbb{F}^{n^2} \otimes \mathbb{F}^{n^2} \otimes \mathbb{F}^{n^2}$$

- $M_n \boxtimes M_m = M_{nm}$
- $M_n \boxtimes M_m = M_{nm}$ was first used by Strassen (1969) for subcubic matrix multiplication algorithms.
- Used today for the Coppersmith–Winograd tensor to obtain the fastest matrix multiplication algorithms.
- Many tasks in computational linear algebra run in roughly the same time:
 - ▶ matrix multiplication
 - ▶ solving systems of linear equations
 - ▶ determinant
 - ▶ matrix inverse
 - ▶ LU decomposition
 - ▶ etc.

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$$\Xi_{n,d} := \begin{pmatrix} x_{1,1}^{(1)} & \cdots & x_{1,n}^{(1)} \\ \vdots & \ddots & \vdots \\ x_{n,1}^{(1)} & \cdots & x_{n,n}^{(1)} \end{pmatrix} \begin{pmatrix} x_{1,1}^{(2)} & \cdots & x_{1,n}^{(2)} \\ \vdots & \ddots & \vdots \\ x_{n,1}^{(2)} & \cdots & x_{n,n}^{(2)} \end{pmatrix} \cdots \begin{pmatrix} x_{1,1}^{(d)} & \cdots & x_{1,n}^{(d)} \\ \vdots & \ddots & \vdots \\ x_{n,1}^{(d)} & \cdots & x_{n,n}^{(d)} \end{pmatrix}$$

This section: Define $\Xi_{n,d} \boxtimes \Xi_{n,d}$

Kronecker product of polynomials in characteristic 0

- Polarization $\text{polar} : \mathbb{C}[x_1, \dots, x_N]_d \rightarrow (\bigotimes^d \mathbb{C}^N)^{\mathfrak{S}_d}$
- Restitution $\text{resti} : \bigotimes^d \mathbb{C}^N \rightarrow \mathbb{C}[x_1, \dots, x_N]_d$

Definition in char 0

Let $h, f \in \mathbb{C}[x_1, \dots, x_N]_d$.

$$h \boxtimes f := \frac{1}{d!} \text{resti}(\text{polar}(h) \boxtimes \text{polar}(f))$$

$\text{mon}_d = x_1 x_2 \cdots x_d$.

- The permanent polynomial: $\text{mon}_d \boxtimes \text{mon}_d$.
- The hyperpermanent: $\text{mon}_d^{\boxtimes m}$.
- The rectangular permanent: $\text{mon}_d \boxtimes e_{n,d}$.
- The mixed discriminant: $\det_d \boxtimes \text{mon}_d$.
- Cayley's first hyperdeterminant: $\det_d^{\boxtimes m}$.

$\boxtimes$ over commutative semirings R

- Let $\text{Sym}_d(V) := (\bigotimes^d V)^{\mathfrak{S}_d}$ (symmetric tensors).
Let $\text{Sym}^d(V) := (\bigotimes^d V) / \mathfrak{S}_d$ (coinvariants = polynomials)
- R -bilinear map $\boxtimes : \bigotimes^d V \times \bigotimes^d W \rightarrow \bigotimes^d (V \boxtimes W)$.
- Restrict to $\psi : \text{Sym}_d(V) \times \bigotimes^d W \rightarrow \bigotimes^d (V \boxtimes W)$.
- Fix $s \in \text{Sym}_d(V)$.
- $\mathfrak{S}_d$ -equivariant R -linear map $\phi_s : \bigotimes^d W \rightarrow \bigotimes^d (V \boxtimes W)$, $f \mapsto \psi(s, f)$
- Apply coinvariant functor: $\psi_s : \text{Sym}^d(W) \rightarrow \text{Sym}^d(V \boxtimes W)$, $[f] \mapsto [\psi(s, f)]$.
- R -linear map $\kappa : \text{Sym}_d(V) \rightarrow \text{Hom}(\text{Sym}^d(W), \text{Sym}^d(V \boxtimes W))$, $s \mapsto \psi_s$
- Altogether: R -bilinear map

$$\boxtimes : \text{Sym}_d(V) \times \text{Sym}^d(W) \rightarrow \text{Sym}^d(V \boxtimes W), \quad (s, [f]) \mapsto \psi_s([f]) = [\psi(s, f)]$$

Definition (Kronecker product of polynomials)

Let $h, f \in R[x_1, \dots, x_N]_d$. $h \boxtimes f := \text{resti}(\text{polar}(h) \boxtimes f)$. $(h \boxtimes f = f \boxtimes h \text{ up to renaming variables})$

On monomials: $(x_1 \cdots x_d) \boxtimes (y_1 \cdots y_d) = \sum_{\pi \in \mathfrak{S}_d} (x_1 \boxtimes y_{\pi(1)}) \cdots (x_d \boxtimes y_{\pi(d)})$.

The hypercomputant $h\Xi_{n,d}$

$$h\Xi_{n,d} := \Xi_{n,d} \boxtimes \Xi_{n,d}$$

an $n^2 \times n^2$ matrix of homog. degree d polynomials. Its bottom-right entry: $h\Gamma_{n,d} = \Gamma_{n,d} \boxtimes \Gamma_{n,d}$

Hamiltonian cycle polynomial:

$$HC_d := \sum_{\phi \in \mathfrak{S}_{d-1}} x_{d,\phi(1)}^{(1)} \cdot x_{\phi(1),\phi(2)}^{(2)} \cdots x_{\phi(d-1),d}^{(d)}$$

Theorem: $HC_d \leq h\Gamma_{d,d}$.

Proof

$$\Gamma_{d,d} = \sum_{\phi: [d-1] \rightarrow [d]} x_{d,\phi(1)}^{(1)} \cdot x_{\phi(1),\phi(2)}^{(2)} \cdot x_{\phi(2),\phi(3)}^{(3)} \cdots x_{\phi(d-1),d}^{(d)}.$$

$$\Gamma_{d,d} \boxtimes \Gamma_{d,d} \geq \text{mon}_d \boxtimes \Gamma_{d,d} = \sum_{\substack{\phi: [d-1] \rightarrow [d] \\ \pi \in \mathfrak{S}_d}} x_{\pi(1),d,\phi(1)}^{(1)} \cdot x_{\pi(2),\phi(1),\phi(2)}^{(2)} \cdots x_{\pi(d),\phi(d-1),d}^{(d)}.$$

Define $T(x_{a,b,c}^{(k)}) = x_{b,c}^{(k)}$ if $a = c$, and $T(x_{a,b,c}^{(k)}) = 0$ otherwise. We calculate

$$T(\text{mon}_d \boxtimes \Gamma_{d,d}) = \sum_{\substack{\phi \in \mathfrak{S}_d \\ \phi(d)=d}} x_{d,\phi(1)}^{(1)} \cdot x_{\phi(1),\phi(2)}^{(2)} \cdots x_{\phi(d-1),d}^{(d)} = HC_d. \quad \square$$

With a small argument, this lifts two hardness results of Gurvits (2004, 2005) to all commutative rings:

$$HC_d \leq \text{mon}_d \boxtimes \Gamma_{d,d} \leq_p \text{mon}_d \boxtimes \det_d \leq \det_d \boxtimes \det_d$$

Exponents

$$\xi_d := \inf\{e \mid w(\mathbf{h}\Xi_{n,d}) \in O(n^e)\},$$

d	2	3	4	5	6	7			
ξ_d for tensors over $\mathbb{C}$	4	4	6	6	8	8	$\dots$	$\Theta(d)$	(Nisan)
ξ_d for polynomials over $\mathbb{R}_{\geq 0}$	4	4	6	6	6	8	$\dots$	$\Theta(d)$	(pathwidth)

Main open question

Is the set of exponents $\{\xi_d \mid d \in \mathbb{N}\}$ unbounded for polynomials over $\mathbb{C}$?

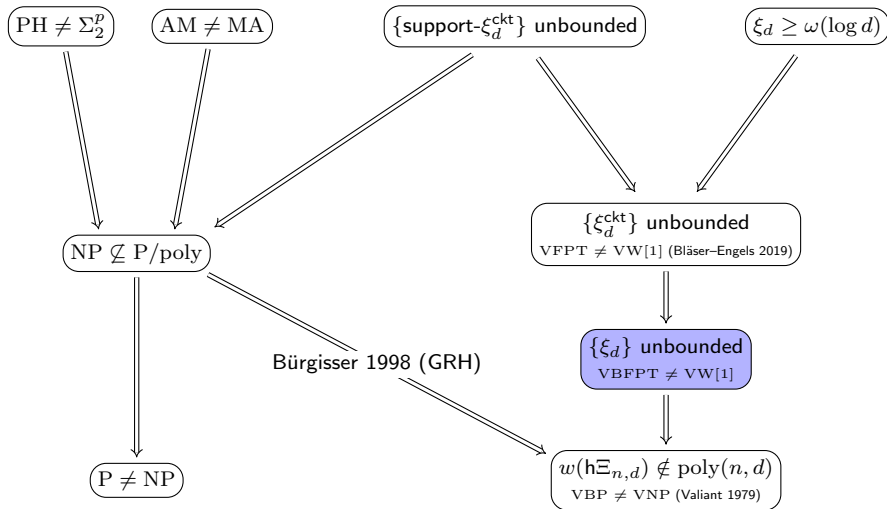
$\xi_d \geq 2$. Best known algorithm in the general case is **monotone**.

Theorem (Chatterjee–Kumar–Volk 2024): Determinantal complexity gives the **same** exponent, over every field.

Theorem (Ahmed Shaaban 2026 using GPT 5.6):

There is a commutative ring where determinantal complexity gives a different exponent.

For polynomials over $\mathbb{C}$:



Summary

- Guided by fixing some rough edges in the definitions of geometric complexity theory (GCT).
- Instead of polynomials, study **double-index** (n and d) **matrices** of polynomials.
- This gives a short existence proof for polysized ABPs for $\det_d$, gives a new record ABP for $\det_d$, and answers a question by Mahajan–Vinay.
- Both properties at the same time: The stabilizer of $\Xi_{n,d}$ is reductive and the tensor variant is Zariski-closed.
- The definition of $\boxtimes$ for polynomials solves issues with the characteristic.
- The equivariance of $\boxtimes$ gives a short way for lifting two hardness results of Gurvits to all commutative rings.
- And possibly more ...

Thank you for your attention!