

Computing Popov and Hermite forms of rectangular polynomial matrices

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Vincent Neiger, Johan Rosenkilde, Grigory Solomatov

XLIM – University of Limoges, France
Technical University of Denmark

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- Context and contribution
- Algorithmic tools and general approach
- Overview of new algorithms

\mathbb{K} a fieldMatrices over $\mathbb{K}[X]$

$$\begin{bmatrix} 3X^3 + X^2 + 5X + 3 & 6X + 5 & 2X + 1 \\ 5 & 5X^2 + 3X + 1 & 5X + 3 \\ 3X + 4 & X^3 + 4X + 1 & 4X^2 + 3 \end{bmatrix}$$

Usual matrix operations

- matrix multiplication
- rank, determinant
- system solving, inversion

Transformations to normal forms

- triangularization \rightsquigarrow Hermite form
- row reduction \rightsquigarrow Popov form
- diagonalization \rightsquigarrow Smith form

Input matrix:
$$\begin{bmatrix} 3X^3 + X^2 + 5X + 3 & 6X + 5 & 2X + 1 \\ 5 & 5X^2 + 3X + 1 & 5X + 3 \\ 3X + 4 & X^3 + 4X + 1 & 4X^2 + 3 \end{bmatrix}$$

Transform, via elementary row operations,

$$\begin{cases} \text{row}_i \leftarrow \text{row}_i + p(X)\text{row}_j \\ \text{row}_i \leftrightarrow \text{row}_j \\ \text{row}_i \leftarrow \alpha \text{row}_i, \alpha \in \mathbb{K} \setminus \{0\} \end{cases}$$

\rightsquigarrow into Popov form [Popov '72]

$$\begin{bmatrix} X^3 + 5X^2 + 4X + 1 & 2X + 4 & 3X + 5 \\ 1 & X^2 + 2X + 3 & X + 2 \\ 3X + 2 & 4X & X^2 \end{bmatrix}$$

\rightsquigarrow into Hermite form [Hermite 1851]

$$\begin{bmatrix} X^6 + 6X^4 + X^3 + X + 4 & 0 & 0 \\ 5X^5 + 5X^4 + 6X^3 + 2X^2 + 6X + 3 & X & 0 \\ 3X^4 + 5X^3 + 4X^2 + 6X + 1 & 5 & 1 \end{bmatrix}$$

Context and contribution

Goal: fast algorithms

\mathbb{K} a field

Matrices over $\mathbb{K}[X]$

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Transformations to **normal forms**

- triangularization \rightsquigarrow **Hermite form**
[Gupta-Storjohann '11] [Labahn-Neiger-Zhou '17]
- row reduction \rightsquigarrow **Popov form**
[Gupta-Sarkar-Storjohann-Valeriotte '11 & '12]
- diagonalization \rightsquigarrow **Smith form**

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Usual **matrix operations**

- **matrix multiplication**
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cost $\mathcal{O}^{\sim}(m^{\omega} d)$

$m \times m$ matrix
with degree d

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For a matrix in $\mathbb{K}[X]^{m \times n}$ with $m \leq n$:

Popov form

deterministic algorithm with cost $\mathcal{O}^{\sim}(m^{\omega-1}nd)$

d the **degree** of the matrix
size of Popov form is $\mathcal{O}(mnd)$

previous fastest: $\mathcal{O}(r m n d^2)$ [Mulders-Storjohann '03] and $\mathcal{O}^{\sim}(m n^{\omega} d)$ [*]

Las Vegas: $\mathcal{O}^{\sim}(m^{\omega-1} n d)$ assuming full row rank [Sarkar-Storjohann '11]

[*] = based on some kernel computation [Beckermann-Labahn-Villard '06]

For a matrix in $\mathbb{K}[X]^{m \times n}$ with $m \leq n$:

Hermite form

deterministic algorithm with cost $\mathcal{O}^{\sim}(m^{\omega-1}n\delta)$

$\delta \leq md$

size of Hermite form can be $\Theta(mn\delta)$

$\delta = \min(\text{sum of row/col degrees})$

previous fastest: $\mathcal{O}^{\sim}(n^{\omega+1}\delta)$ [\star]

(speed-up factor $\geq n$)

[\star] = based on some kernel computation [Beckermann-Labahn-Villard '06]

Left kernel basis of a matrix $A \in \mathbb{K}[X]^{m \times n}$

Matrix $K \in \mathbb{K}[X]^{k \times m}$ such that

$$\left\{ \begin{array}{l} K \text{ has full row rank} \\ KA = 0 \\ \text{rows of } K \text{ generate the kernel of } A \end{array} \right.$$

- core algorithmic tool: rank, inversion, determinant, Hermite form, ...
- kernel bases can now be computed fast in Popov form
 \rightsquigarrow combine [Zhou-Labahn-Storjohann '12] + this work
- shifted normal forms: $UA = P$
 (via shifted Popov approximant basis [Jeannerod-Neiger-Schost-Villard '16])
 \rightsquigarrow cost of this approach: unsatisfactory

Row basis of a matrix $A \in \mathbb{K}[X]^{m \times n}$

Matrix $B \in \mathbb{K}[X]^{r \times n}$ such that $\begin{cases} B \text{ has full row rank} \\ B \text{ and } A \text{ have the same row space} \end{cases}$

Fast algorithm: [Zhou-Labahn '13]

Consequence: deal with $m \geq n$ and rank-deficient matrices

Normal form of arbitrary A

Step 1: $B \leftarrow$ row basis of A // use [Zhou-Labahn '13]
Step 2: $P \leftarrow$ the normal form of B // full row rank case
Step 3: Return P

Step 1 costs $\tilde{O}(m^{\omega-1}(m+n)d)$

What do **pivots** become in the full row rank case?

nonsingular, $m \times m$

pivot of a row: rightmost entry of largest degree

$$\begin{bmatrix} [6] & [1] & [2] & [4] \\ [2] & [2] & [1] & [1] \\ [3] & [1] & [3] & [2] \\ [5] & [1] & [2] & [5] \end{bmatrix} \quad \text{Popov}$$

size $\mathcal{O}(m^2d)$

pivot of a row: rightmost nonzero entry

$$\begin{bmatrix} [12] & & & \\ [11] & [3] & & \\ [11] & [2] & [0] & \\ [11] & [2] & & [1] \end{bmatrix} \quad \text{Hermite}$$

size $\mathcal{O}(m^2d)$

nonsingular, $m \times m$

full row rank, $m \times n$

pivot of a row: rightmost entry of largest degree

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Popov

$$\begin{bmatrix} [6] & [5] & [5] & [1] & [5] & [2] & [4] \\ [2] & [2] & [2] & [2] & [1] & [1] & [1] \\ [3] & [3] & [3] & [1] & [3] & [3] & [2] \\ [5] & [5] & [5] & [1] & [5] & [2] & [5] \end{bmatrix}$$

size $\mathcal{O}(m^2d)$

size $\mathcal{O}(mnd)$

pivot of a row: rightmost nonzero entry

$$\begin{bmatrix} [12] & & & \\ [11] & [3] & & \\ [11] & [2] & [0] & \\ [11] & [2] & & [1] \end{bmatrix}$$

Hermite

size $\mathcal{O}(m^2d)$

Pivot support

From now, focus on **Popov form**

- full row rank $m \times n$ matrix \mathbf{A}
- input/output size $\mathcal{O}(mnd)$
- target cost $\mathcal{O}(m^{\omega-1}nd)$

$$\begin{bmatrix} [4] & [3] & [5] & [2] & [1] & [4] & [5] \\ [7] & [9] & [4] & [4] & [2] & [9] & [3] \\ [5] & [8] & [5] & [5] & [0] & [5] & [9] \\ [7] & [4] & [9] & [3] & [5] & [5] & [9] \end{bmatrix} = \mathbf{A}$$

unimodular

$$\begin{bmatrix} [6] & [5] & [5] & [1] & [5] & [2] & [4] \\ [2] & [2] & [2] & [2] & [1] & [1] & [1] \\ [3] & [3] & [3] & [1] & [3] & [3] & [2] \\ [5] & [5] & [5] & [1] & [5] & [2] & [5] \end{bmatrix} = \mathbf{P}$$

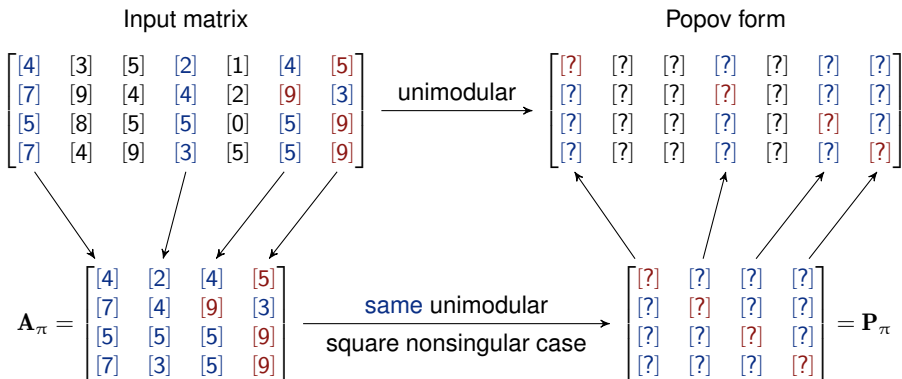
pivot support of \mathbf{A} : indices of the pivots in \mathbf{P}

\rightsquigarrow any multiple \mathbf{UA} has its pivots in the pivot support

Algorithmic approach

- Find the pivot support of \mathbf{A}
- Use the pivot support to compute \mathbf{P}

When the pivot support is known



Algorithm in $\mathcal{O}^{\sim}(m^{\omega-1}nd)$

Step 1: $\mathbf{A}_\pi \leftarrow$ submatrix of \mathbf{A} with columns in pivot support

Step 2: $\mathbf{P}_\pi \leftarrow \mathbf{U}\mathbf{A}_\pi =$ Popov form of \mathbf{A}_π // nonsingular case

Step 3: $\mathbf{P}_{\neq\pi} \leftarrow \mathbf{U}\mathbf{A}_{\neq\pi} = (\mathbf{P}_\pi\mathbf{A}_\pi^{-1})\mathbf{A}_{\neq\pi} \bmod X^{d+1}$

Saturation of $\mathbf{A} = \mathbb{K}[X]^{1 \times n} \cap \mathbb{K}(X)^{1 \times m} \mathbf{A}$
 = left kernel of any right kernel basis of \mathbf{A}

same rank as \mathbf{A} + contains the row space of $\mathbf{A} \Rightarrow$ same pivot support

Pivot Support in $\tilde{O}(n^\omega d)$

Step 1: $\mathbf{K} \leftarrow$ right kernel basis of \mathbf{A} // use [Zhou-Labahn-Storjohann '12]

Step 2: $\mathbf{B} \leftarrow$ pivot support revealing left kernel basis of \mathbf{K}

Step 3: return the indices of pivots in \mathbf{B}

Ensuring efficiency in Step 2

- \mathbf{K} has large average row degree (not suitable for [Zhou-Labahn-Storjohann '12])
- \mathbf{K} has low average column degree, and we have $\deg(\mathbf{B}) \leq d$ since $\mathbf{AK} = \mathbf{0}$

\rightsquigarrow obtain \mathbf{B} via fast Popov approximant basis [Jeanerod-Neiger-Schost-Villard '16]

$$\mathbf{BK} = \mathbf{0} \bmod \text{diag}(X^{\delta_1}, \dots, X^{\delta_{n-r}})$$

Goal: when $n \gg m$, lower the cost from $\mathcal{O}^{\sim}(n^{\omega} d)$ to $\mathcal{O}^{\sim}(m^{\omega-1} n d)$

$$\begin{bmatrix} [4] & [3] & [5] & [2] & [1] & [4] & [5] & [4] & [2] & [9] & [0] & [3] \\ [7] & [9] & [4] & [4] & [2] & [9] & [3] & [6] & [3] & [6] & [7] & [4] \\ [5] & [8] & [5] & [5] & [0] & [5] & [9] & [6] & [4] & [6] & [2] & [1] \end{bmatrix}$$

Strategy

apply the previous algorithm iteratively to

- n/m overlapping submatrices of A
- each of dimensions $m \times 2m$

Goal: when $n \gg m$, lower the cost from $\mathcal{O}^{\sim}(n^{\omega} d)$ to $\mathcal{O}^{\sim}(m^{\omega-1} n d)$

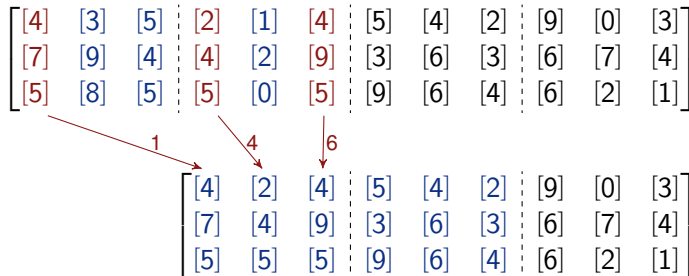
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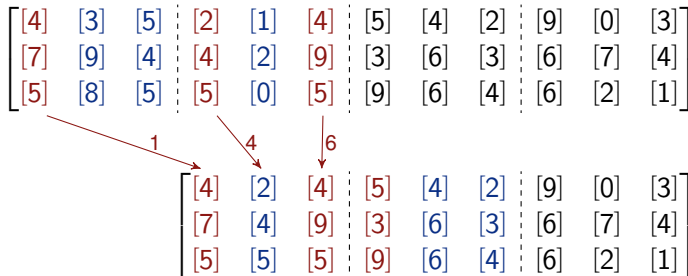


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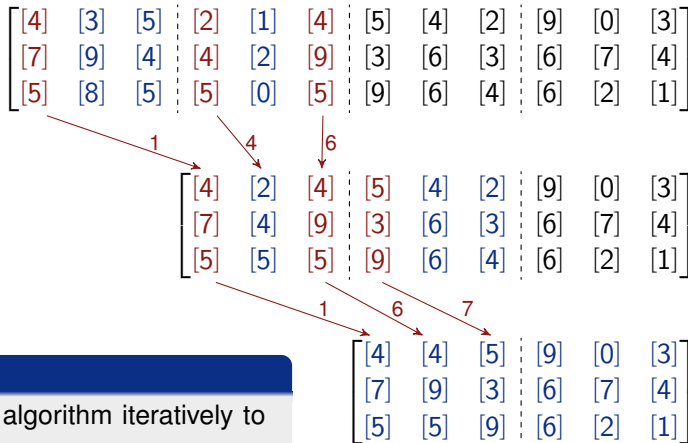
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Find the pivot support of a wide matrix

Goal: when $n \gg m$, lower the cost from $\mathcal{O}^{\sim}(n^{\omega} d)$ to $\mathcal{O}^{\sim}(m^{\omega-1} n d)$

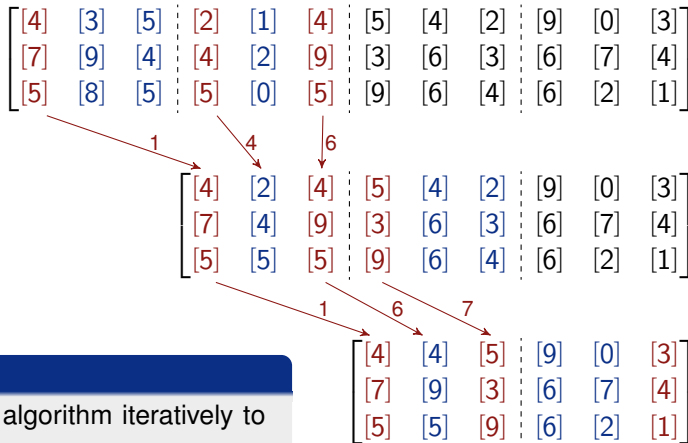


Strategy

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- n/m overlapping submatrices of A
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Goal: when $n \gg m$, lower the cost from $\mathcal{O}(n^\omega d)$ to $\mathcal{O}(m^{\omega-1}nd)$



Strategy

apply the previous algorithm iteratively to

- n/m overlapping submatrices of A
- each of dimensions $m \times 2m$

Conclusion:

- fast algorithm to **find pivot support** for Popov form
- fast Popov form **when pivot support known**

Also in the paper:

- algorithms based on **completing** the matrix into a **square** one
- second item works for **all shifted normal forms** (including **Hermite**)

Perspectives:

- normal forms for **arbitrary shifts**
- kernel bases for **arbitrary shifts**
- cost bound **sensitive** to **average** row/column degrees