

Fast computation of shifted Popov forms of polynomial matrices via systems of linear modular equations

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Hermite and Popov forms

$\mathbf{A} \in \mathbb{K}[X]^{m \times m}$ nonsingular

\rightsquigarrow via elementary row operations,
transform \mathbf{A} into

Hermite form [Hermite, 1851]

triangular

Popov form [Popov, 1972]

row reduced

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column normalized

$$\begin{bmatrix} 4 & & & \\ 3 & 7 & & \\ 1 & 5 & 3 & \\ 3 & 6 & 1 & 2 \end{bmatrix}$$

Popov form [Popov, 1972]

row reduced
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$$\begin{bmatrix} 7 & 0 & 1 & 5 \\ 0 & 1 & & 0 \\ & & 2 & \\ 6 & 0 & 1 & 6 \end{bmatrix}$$

Invariant: $\sigma = \deg(\det(\mathbf{A})) = 4 + 7 + 3 + 2 = 7 + 1 + 2 + 6$

Hermite and Popov forms

$\mathbf{A} \in \mathbb{K}[X]^{m \times m}$ nonsingular

\rightsquigarrow via elementary row operations,
transform \mathbf{A} into

basis of $\mathcal{M} \subset \mathbb{K}[X]^{1 \times m}$ of rank m

\rightsquigarrow find the reduced Gröbner basis
of \mathcal{M} for either term order

Hermite form [Hermite, 1851]

triangular
column normalized } POT

$$\begin{bmatrix} 4 & & & \\ 3 & 7 & & \\ 1 & 5 & 3 & \\ 3 & 6 & 1 & 2 \end{bmatrix}$$

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Invariant: $\sigma = \deg(\det(\mathbf{A})) = 4 + 7 + 3 + 2 = 7 + 1 + 2 + 6$
 $=$ dimension of $\mathbb{K}[X]^{1 \times m} / \mathcal{M}$ as a \mathbb{K} -vector space

Example: constrained bivariate interpolation

As in [Guruswami-Sudan list-decoding](#) of Reed-Solomon codes

M of degree σ ; L of degree $< \sigma$

$$\mathbf{A} = \begin{bmatrix} M & & & & & \\ -L & 1 & & & & \\ -L^2 & & 1 & & & \\ \vdots & & & \ddots & & \\ -L^{m-1} & & & & & 1 \end{bmatrix}$$

Problem: find $\mathbf{p} = [p_1 \ \cdots \ p_m] \in \text{RowSpace}(\mathbf{A})$ such that

$$(\star) \quad \deg(p_j) < N_j \quad \text{for all } j$$

Approach:

- compute the Popov form \mathbf{P} of \mathbf{A} with **degree weights** on the columns
- return **row of \mathbf{P} which satisfies (\star)**

Shifted Popov form

degree shift: $\mathbf{s} = (s_1, \dots, s_m) \in \mathbb{Z}^m$ acting as additive weights

\rightsquigarrow shifted row reduced: minimizes the \mathbf{s} -degree of $\mathbf{p} = [p_1 \ \cdots \ p_m]$

$$\text{rdeg}_{\mathbf{s}}(\mathbf{p}) = \max_j (\deg(p_j) + s_j)$$

Degree constraints: $\deg(p_j) < N_j$ for all $j \Leftrightarrow \text{rdeg}_{(-N_1, \dots, -N_m)}(\mathbf{p}) < 0$

s-Popov form = **s**-row reduced + column normalized

Canonical form:

$\mathbf{U}\mathbf{A} = \mathbf{P}$ for unique unimodular \mathbf{U} and **s**-Popov form \mathbf{P}

Shifted Popov form: examples

Connects Popov and Hermite forms. Examples with $m = 4, \sigma = 16$:

$\mathbf{s} = (0, 0, 0, 0)$ Popov	$\begin{bmatrix} 4 & 3 & 3 & 3 \\ 3 & 4 & 3 & 3 \\ 3 & 3 & 4 & 3 \\ 3 & 3 & 3 & 4 \end{bmatrix}$	$\begin{bmatrix} 7 & 0 & 1 & 5 \\ 0 & 1 & & 0 \\ & & 2 & \\ 6 & 0 & 1 & 6 \end{bmatrix}$
$\mathbf{s} = (0, 2, 4, 6)$ s-Popov	$\begin{bmatrix} 7 & 4 & 2 & 0 \\ 6 & 5 & 2 & 0 \\ 6 & 4 & 3 & 0 \\ 6 & 4 & 2 & 1 \end{bmatrix}$	$\begin{bmatrix} 8 & 5 & 1 & \\ 7 & 6 & 1 & \\ & & 2 & \\ 0 & 1 & & 0 \end{bmatrix}$
$\mathbf{s} = (0, \sigma, 2\sigma, 3\sigma)$ Hermite	$\begin{bmatrix} 16 & & & \\ 15 & 0 & & \\ 15 & & 0 & \\ 15 & & & 0 \end{bmatrix}$	$\begin{bmatrix} 4 & & & \\ 3 & 7 & & \\ 1 & 5 & 3 & \\ 3 & 6 & 1 & 2 \end{bmatrix}$

Recall: $\delta_1 + \dots + \delta_m = \sigma = \deg(\det(\mathbf{A})) = \deg(\det(\mathbf{P}))$

\rightsquigarrow for \mathbf{P} , average column degree: $\sigma/m \Rightarrow$ size: $\mathcal{O}(m\sigma)$

Result

Problem

Input: $\mathbf{A} \in \mathbb{K}[X]^{m \times m}$ square nonsingular
 shift $\mathbf{s} \in \mathbb{Z}^m$

Output: the \mathbf{s} -Popov form \mathbf{P} of \mathbf{A}

Previous fastest algorithm: $\tilde{O}(m^\omega(d + \text{amp}(\mathbf{s})))$, deterministic
 [Gupta-Sarkar-Storjohann-Valeriotte, 2012] + [Sarkar-Storjohann, 2011]
 $\text{amp}(\mathbf{s}) = \max(\mathbf{s}) - \min(\mathbf{s})$ is between 0 and m^2d
 \rightsquigarrow worst-case $\tilde{O}(m^{\omega+2}d)$

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Here: $\tilde{O}(m^\omega \sigma(\mathbf{A})/m) \subseteq \tilde{O}(m^\omega d)$, probabilistic

- no dependency in \mathbf{s} (except in log factors)
- takes some degree structure into account:
 $\sigma(\mathbf{A})/m \leq \text{avg. row degree, avg. column degree}$

Reduction to $\deg(\mathbf{A}) \leq \sigma(\mathbf{A})/m$

Problem

- Input: $\mathbf{A} \in \mathbb{K}[X]^{m \times m}$ square nonsingular
 shift $\mathbf{s} \in \mathbb{Z}^m$
- Output: the \mathbf{s} -Popov form \mathbf{P} of \mathbf{A}

With **no field operation**, one can build

- $\tilde{\mathbf{A}} \in \mathbb{K}[X]^{\tilde{m} \times \tilde{m}}$
- $\mathbf{t} \in \mathbb{Z}^{\tilde{m}}$

such that

- $\tilde{m} \leq 3m$ and $\deg(\tilde{\mathbf{A}}) \leq \lceil \sigma(\mathbf{A})/m \rceil$,
- \mathbf{s} -Popov form of \mathbf{A} = principal submatrix of the \mathbf{t} -Popov form of $\tilde{\mathbf{A}}$

thanks to **partial linearization** techniques

[Gupta et al., 2012]

Iterative Popov form algorithm

[Wolovich, 1974] and [Mulders-Storjohann, 2003]

Row reduction:

$$\mathbf{A} = \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 3 & 2 & 3 & 2 \end{bmatrix} \longrightarrow \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 3 & 2 & 2 & 2 \end{bmatrix} \longrightarrow \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 2 & 2 & 2 & 2 \end{bmatrix} = \mathbf{R}$$

Column normalization:

$$\mathbf{R} = \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 2 & 2 & 2 & 2 \end{bmatrix} \longrightarrow \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 2 & 2 & 1 & 2 \end{bmatrix} \longrightarrow \begin{bmatrix} 3 & 2 & 0 & 0 \\ 2 & 5 & 1 & 1 \\ 2 & 0 & 2 & 1 \\ 2 & 2 & 1 & 2 \end{bmatrix} = \mathbf{P}$$

Cost bound: $\mathcal{O}(m^3 d^2)$

\rightsquigarrow incorporate

- fast matrix multiplication $\mathcal{O}(m^\omega)$?
- fast polynomial arithmetic $\tilde{\mathcal{O}}(d)$?

Obstacle: size of the transformation

unimodular transformation \mathbf{U} may have size beyond $\mathcal{O}(m^\omega d)$

Example:

- \mathbf{A} unimodular: $\mathbf{A}^{-1}\mathbf{A} = \mathbf{I}_m$
- $\mathbf{P} = \mathbf{I}_m$ for any \mathbf{s}
- $\mathbf{U} = \mathbf{A}^{-1}$

$$\begin{array}{ccc}
 \mathbf{A} & & \mathbf{U} = \mathbf{A}^{-1} \\
 \left[\begin{array}{cccccc} 0 & & & & & \\ d & 0 & & & & \\ & d & 0 & & & \\ & & & \ddots & \ddots & \\ & & & & d & 0 \end{array} \right] & \longrightarrow & \left[\begin{array}{cccccc} 0 & & & & & \\ d & 0 & & & & \\ 2d & d & 0 & & & \\ \vdots & \ddots & \ddots & \ddots & & \\ (m-1)d & \cdots & 2d & d & 0 & \end{array} \right] \\
 \text{degree } d & & \text{sum of degrees } \Theta(m^3 d)
 \end{array}$$

Fast Popov form

Step 1: fast row reduction

$$\tilde{O}(m^\omega d)$$

[Giorgi et al., 2003], probabilistic
[Gupta et al., 2012], deterministic

Step 2: fast column normalization

$$\tilde{O}(m^\omega d)$$

[Sarkar-Storjohann, 2011]

[Giorgi et al., 2003]:

- Expansion of \mathbf{A}^{-1} is ultimately linearly recurrent
- Find $2d + 1$ high-degree terms \mathbf{B} in expansion of \mathbf{A}^{-1}
- Reconstruct \mathbf{R} as $\mathbf{B} = \frac{*}{\mathbf{R}} \bmod X^{2d+1}$

\rightsquigarrow uses $\deg(\mathbf{R}) \leq d$, which does **not** hold for **arbitrary** shifts
(even $\deg(\mathbf{P})$ may be md)

Hermite form in $\tilde{\mathcal{O}}(m^\omega d)$

[Gupta-Storjohann, 2011], [Gupta, 2011]:

Step 1: Smith form computation: $\mathbf{UAV} = \mathbf{S}$ (probabilistic)
 \rightsquigarrow modular equations describing $\text{RowSpace}(\mathbf{A})$

Step 2: find pivot degrees $\delta = (\delta_1, \dots, \delta_m)$ by triangularization
 from a matrix involving \mathbf{V} and \mathbf{S}

Step 3: use δ to find Hermite basis of solutions to the equations

[Zhou, 2012], [Zhou-Labahn, 2016]:

Step 1: find pivot degrees δ by (partial) triangularization
 (using kernel bases and column bases, deterministic)

Step 2: use δ to find Hermite form of \mathbf{A}

s-Popov form not triangular for arbitrary \mathbf{s}

Reduction to linear modular equations: example

$$\mathbf{I}_m \begin{bmatrix} M & & & & & \\ & -L & & & & \\ & & 1 & & & \\ & & & -L^2 & & \\ & & & & 1 & \\ & & & & & \ddots & \\ & & & & & & \ddots & \\ & & & & & & & 1 \end{bmatrix} \begin{bmatrix} 1 & & & & & \\ & L & & & & \\ & & 1 & & & \\ & & & L^2 & & \\ & & & & 1 & \\ & & & & & \ddots & \\ & & & & & & \ddots & \\ & & & & & & & L^{m-1} & & \\ & & & & & & & & & 1 \end{bmatrix} = \begin{bmatrix} M & & & & & \\ & 1 & & & & \\ & & 1 & & & \\ & & & 1 & & \\ & & & & \ddots & \\ & & & & & 1 \end{bmatrix}$$

\rightsquigarrow for $\mathbf{p} = [p_1 \ \cdots \ p_m]$,

$$\mathbf{p} \in \text{RowSpace}(\mathbf{A}) \Leftrightarrow [p_1 \ \cdots \ p_m] \begin{bmatrix} 1 \\ L \\ L^2 \\ \vdots \\ L^{m-1} \end{bmatrix} = 0 \pmod{M}$$

$$\Leftrightarrow p_1 1 + p_2 L + \cdots + p_m L^{m-1} = 0 \pmod{M}$$

Reduction to system of modular equations

Smith form of \mathbf{A} :

$$\mathbf{UAV} = \text{diag}(1, \dots, 1, m_1, \dots, m_n)$$

Consider $\mathfrak{M} = (m_1, \dots, m_n)$ and $[\mathbf{0} \mid \mathbf{F}] = \mathbf{V} \bmod (1, \dots, 1, \mathfrak{M})$

$\rightsquigarrow (\mathfrak{M}, \mathbf{F})$ computed in probabilistic $\tilde{O}(m^\omega d)$ [Gupta-Storjohann, 2011]

Then

$$\text{RowSpace}(\mathbf{A}) = \{\mathbf{p} \in \mathbb{K}[X]^{1 \times m} \mid \mathbf{pF} = 0 \bmod \mathfrak{M}\}$$

\rightsquigarrow \mathbf{s} -Popov form of $\mathbf{A} = \mathbf{s}$ -Popov basis of solutions for $(\mathfrak{M}, \mathbf{F})$

Linear systems of modular equations

Input: nonzero moduli $\mathfrak{M} = (m_1, \dots, m_n)$
 system matrix $\mathbf{F} \in \mathbb{K}[X]^{m \times n}$ with $\deg(\mathbf{F}_{*,j}) < \deg(m_j)$
 shift $\mathbf{s} \in \mathbb{Z}^m$

Output: \mathbf{P} the \mathbf{s} -Popov solution basis for $(\mathfrak{M}, \mathbf{F})$

for $\sigma = \deg(m_1) + \dots + \deg(m_n),$ $\deg(\det(\mathbf{P})) \leq \sigma$	\implies	<table border="1"> <thead> <tr> <th>I/O size</th> <th>target cost</th> </tr> </thead> <tbody> <tr> <td>$\mathcal{O}(m\sigma)$</td> <td>$\tilde{\mathcal{O}}(m^{\omega-1}\sigma)$</td> </tr> </tbody> </table>	I/O size	target cost	$\mathcal{O}(m\sigma)$	$\tilde{\mathcal{O}}(m^{\omega-1}\sigma)$
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Order bases: $m_1 = \dots = m_n = X^{\sigma/n} \rightsquigarrow \tilde{\mathcal{O}}(m^{\omega-1}\sigma)$

[Giorgi et al., 2003] [Storjohann, 2006] [Zhou-Labahn, 2012] [Jeannerod et al., 2016]

Interpolation bases: $m_j =$ product of known linear factors $\rightsquigarrow \tilde{\mathcal{O}}(m^{\omega-1}\sigma)$

[Beckermann-Labahn, 2000] [Jeannerod et al., 2015+2016]

Here: $\tilde{\mathcal{O}}(m^{\omega-1}\sigma)$ for arbitrary moduli, $n \in \mathcal{O}(m)$

Overview of the algorithm

Similarly to [Jeannerod et al., 2016] for interpolation bases,
 divide-and-conquer on n (number of equations):

- recursive calls give $\mathbf{P}^{(1)}$ and $\mathbf{P}^{(2)} \rightsquigarrow \mathbf{P} = \text{ColumnNormalize}(\mathbf{P}^{(2)}\mathbf{P}^{(1)})$
- deduce \mathbf{s} -pivot degrees δ of \mathbf{P}
- compute \mathbf{P} when knowing δ [Gupta-Storjohann, 2011]

\rightsquigarrow base case: one equation

Difficulty: no recurrence relations like in order/interpolation bases

\rightsquigarrow compute a shifted Popov kernel basis with arbitrary shift:

$$\mathbf{p}\mathbf{F} = 0 \bmod \mathbf{m} \quad \Leftrightarrow \quad \text{for some } q, \quad [\mathbf{p} \quad q] \begin{bmatrix} \mathbf{F} \\ \mathbf{m} \end{bmatrix} = 0$$

New divide-and-conquer approach on the shift

New divide-and-conquer approach on the shift

Recall $\deg(\mathbf{F}) < \deg(\mathbf{m}) = \sigma$

Reduction to kernel basis:

$$\begin{bmatrix} \mathbf{P} & \mathbf{q} \end{bmatrix} = (\mathbf{s}, \min(\mathbf{s}))\text{-Popov kernel basis of } \begin{bmatrix} \mathbf{F} \\ \mathbf{m} \end{bmatrix}$$

Reduction to order basis:

$$\begin{bmatrix} \mathbf{P} & \mathbf{q} \\ * & * \end{bmatrix} = (\mathbf{s}, \min(\mathbf{s}))\text{-Popov order basis for } \begin{bmatrix} \mathbf{F} \\ \mathbf{m} \end{bmatrix} \text{ and } \text{amp}(\mathbf{s}) + 2\sigma$$

\rightsquigarrow Base case: $\text{amp}(\mathbf{s}) \in \mathcal{O}(\sigma)$, cost $\tilde{\mathcal{O}}(m^{\omega-1}\sigma)$ [Jeannerod et al., 2016]

Divide-and-conquer on $\text{amp}(\mathbf{s})$:

$$\mathbf{s} = (\mathbf{s}^{(1)}, \mathbf{s}^{(2)}), \quad \mathbf{F} = \begin{bmatrix} \mathbf{F}^{(1)} \\ \mathbf{F}^{(2)} \end{bmatrix} \quad \text{with} \quad \text{amp}(\mathbf{s}^{(i)}) \approx \text{amp}(\mathbf{s})/2$$

New divide-and-conquer approach on the shift

- ① recursive call to find **splitting index** and $\delta^{(1)}$:

$$\begin{bmatrix} \mathbf{P}^{(1)} & \mathbf{0} \\ * & * \end{bmatrix} = \mathbf{s}^{(1)}\text{-Popov sol. basis for } (\mathbf{F}^{(1)}, \mathbf{m}) \rightsquigarrow \text{UpdateSplit}(\mathbf{s}, \mathbf{F})$$

- ② residual computation thanks to **known** $\delta^{(1)}$:

$$\mathbf{A} = \begin{bmatrix} \mathbf{P}^{(1)} & \mathbf{0} & \mathbf{q}^{(1)} \\ * & \mathbf{P}^{(0)} & * \\ * & \mathbf{0} & q \end{bmatrix} = \mathbf{u}\text{-order basis for } \begin{bmatrix} \mathbf{F}^{(1)} \\ \mathbf{F}^{(2)} \\ \mathbf{m} \end{bmatrix} \rightsquigarrow \begin{bmatrix} \mathbf{0} \\ \mathbf{G} \\ \mathbf{n} \end{bmatrix} = \mathbf{A} \begin{bmatrix} \mathbf{F}^{(1)} \\ \mathbf{F}^{(2)} \\ \mathbf{m} \end{bmatrix}$$

- ③ recursive call to find $\delta^{(2)} \rightsquigarrow \mathbf{s}$ -pivot degree $\delta = (\delta^{(1)}, \delta^{(0)} + \delta^{(2)})$

$$\mathbf{P}^{(2)} = (\mathbf{s}^{(2)} + \delta^{(0)})\text{-Popov sol. basis for } (\mathbf{G}, \mathbf{n})$$

- ④ compute \mathbf{P} from δ via order basis at order $\mathcal{O}(\sigma)$

Conclusion

Linear systems of modular equations

- $\tilde{\mathcal{O}}(m^{\omega-1}\sigma)$
- return **basis** of solutions
- in **s-Popov form**
- deterministic

Shifted row reduction of polynomial matrices

- $\tilde{\mathcal{O}}(m^{\omega-1}\sigma(\mathbf{A}))$
- return **s-Popov form**
- probabilistic

Example: constrained bivariate interpolation

As in Guruswami-Sudan list-decoding of Reed-Solomon codes:

$$\mathcal{M} = \left\{ Q = \sum_{0 \leq j < m} Q_j(X) Y^j \in \mathbb{K}[X, Y] \mid Q(x_i, y_i) = 0 \text{ for } 1 \leq i \leq \sigma \right\}$$

Define $M = (X - x_1) \cdots (X - x_\sigma)$ and $L \in \mathbb{K}[X]$ s.t. $L(x_i) = y_i$

$$\rightsquigarrow \text{basis of } \mathcal{M}: \begin{pmatrix} M \\ Y - L \\ Y^2 - L^2 \\ \vdots \\ Y^{m-1} - L^{m-1} \end{pmatrix} \longleftrightarrow \mathbf{A} = \begin{bmatrix} M & & & & \\ -L & 1 & & & \\ -L^2 & & 1 & & \\ \vdots & & & \ddots & \\ -L^{m-1} & & & & 1 \end{bmatrix}$$

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Problem: find $Q \in \mathcal{M}$ satisfying $\deg(Q_j) < N_j$ for $0 \leq j < m$

Approach:

- compute the Popov form \mathbf{P} of \mathbf{A} with **degree weights** on the columns
- return **row of \mathbf{P} which satisfies (iii)**

Reduction to linear modular equations: example

$$\mathbf{I}_m \begin{bmatrix} M & & & & & \\ -L & 1 & & & & \\ -L^2 & & 1 & & & \\ \vdots & & & \ddots & & \\ -L^{m-1} & & & & 1 & \end{bmatrix} \begin{bmatrix} 1 & & & & & \\ L & 1 & & & & \\ L^2 & & 1 & & & \\ \vdots & & & \ddots & & \\ L^{m-1} & & & & 1 & \end{bmatrix} = \begin{bmatrix} M & & & & & \\ & 1 & & & & \\ & & 1 & & & \\ & & & \ddots & & \\ & & & & 1 & \\ & & & & & 1 \end{bmatrix}$$

In other words, for $Q = \sum_{j < m} Q_j(X) Y^j$,

$$Q(x_i, y_i) = 0 \text{ for all } i \Leftrightarrow [Q_0 \quad \cdots \quad Q_{m-1}] \begin{bmatrix} 1 \\ L \\ L^2 \\ \vdots \\ L^{m-1} \end{bmatrix} = 0 \pmod{M}$$

$$\Leftrightarrow Q(X, L) = 0 \pmod{M}$$

Previous algorithms

Here, \star = probabilistic algorithm, $d = \deg(\mathbf{A})$

Algorithm	Problem	Cost bound
[Hafner-McCurley, 1991]	Hermite form	$\tilde{O}(m^4 d)$
[Storjohann-Labahn, 1996]	Hermite form	$\tilde{O}(m^{\omega+1} d)$
[Villard, 1996]	Popov & Hermite forms	$\tilde{O}(m^{\omega+1} d + (md)^\omega)$
[Alekhovich, 2002]	weak Popov form	$\tilde{O}(m^{\omega+1} d)$
[Mulders-Storjohann, 2003]	Popov & Hermite forms	$O(m^3 d^2)$
[Giorgi et al., 2003]	$\mathbf{0}$ -reduction	$\tilde{O}(m^\omega d)$ \star
[1] = [Sarkar-Storjohann, 2011]	Popov form of $\mathbf{0}$ -reduced	$\tilde{O}(m^\omega d)$
[Gupta-Storjohann, 2011]	Hermite form	$\tilde{O}(m^\omega d)$ \star
[2] = [Gupta et al., 2012]	$\mathbf{0}$ -reduction	$\tilde{O}(m^\omega d)$
[Zhou-Labahn, 2012/2016]	Hermite form	$\tilde{O}(m^\omega d)$
[1] + [2]	s-Popov form for any s	$\tilde{O}(m^\omega (d + \text{amp}(\mathbf{s})))$

Iterative Popov form algorithm

[Wolovich, 1974] and [Mulders-Storjohann, 2003]

Row reduction:

$$\mathbf{A} = \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 3 & 2 & 3 & 2 \end{bmatrix} \longrightarrow \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 3 & 2 & 2 & 2 \end{bmatrix}$$

Column normalization:

Cost bound: $\mathcal{O}(m^3 d^2)$

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Column normalization:

$$\mathbf{R} = \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 2 & 2 & 2 & 2 \end{bmatrix} \longrightarrow \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 2 & 2 & 1 & 2 \end{bmatrix}$$

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$$\mathbf{A} = \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 3 & 2 & 3 & 2 \end{bmatrix} \longrightarrow \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 3 & 2 & 2 & 2 \end{bmatrix} \longrightarrow \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 2 & 2 & 2 & 2 \end{bmatrix} = \mathbf{R}$$

Column normalization:

$$\mathbf{R} = \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 2 & 2 & 2 & 2 \end{bmatrix} \longrightarrow \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 2 & 2 & 1 & 2 \end{bmatrix} \longrightarrow \begin{bmatrix} 3 & 2 & 0 & 0 \\ 2 & 5 & 1 & 1 \\ 2 & 0 & 2 & 1 \\ 2 & 2 & 1 & 2 \end{bmatrix} = \mathbf{P}$$

Cost bound: $\mathcal{O}(m^3 d^2)$

Iterative Popov form algorithm

[Wolovich, 1974] and [Mulders-Storjohann, 2003]

Row reduction:

$$\mathbf{A} = \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 3 & 2 & 3 & 2 \end{bmatrix} \longrightarrow \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 3 & 2 & 2 & 2 \end{bmatrix} \longrightarrow \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 2 & 2 & 2 & 2 \end{bmatrix} = \mathbf{R}$$

Column normalization:

$$\mathbf{R} = \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 2 & 2 & 2 & 2 \end{bmatrix} \longrightarrow \begin{bmatrix} 3 & 2 & 0 & 0 \\ 4 & 5 & 0 & 0 \\ 2 & 0 & 2 & 1 \\ 2 & 2 & 1 & 2 \end{bmatrix} \longrightarrow \begin{bmatrix} 3 & 2 & 0 & 0 \\ 2 & 5 & 1 & 1 \\ 2 & 0 & 2 & 1 \\ 2 & 2 & 1 & 2 \end{bmatrix} = \mathbf{P}$$

Cost bound: $\mathcal{O}(m^3 d^2)$

\rightsquigarrow incorporate

- fast matrix multiplication $\mathcal{O}(m^\omega)$?
- fast polynomial arithmetic $\tilde{\mathcal{O}}(d)$?