

# Algorithms for structured approximation and interpolation

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# Context

And notations

## Goal

Design algorithms with **good complexity** bounds for fundamental operations on **polynomials and matrices**

Exact Computations

**Base Field  $\mathbb{K}$**   
 $\mathbb{Z}/p\mathbb{Z}, \mathbb{F}_{p^e}, \mathbb{Q}, \dots$

**Algebraic Complexity**  
Upper bound on the number of operations in  $\mathbb{K}$

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Notations:

- $\tilde{O}$ : asymptotic bound hiding logarithmic factors.  
 $\rightsquigarrow$  **Fast polynomial multiplication:  $\tilde{O}(d)$  using FFT-based methods.**
- $\omega$ : exponent of matrix multiplication,  $2 < \omega \leq 3$ , best known  $\omega \approx 2.371$ .

# Modular approximation

Problem and previous works

Problem 1:

$\text{ApproxMod}(M, F \in \mathbb{K}[x]^{n \times m}, \nu = \{\nu_1, \dots, \nu_m\})$   
find  $g$  such that  $F \cdot g = 0 \pmod{M}$  and  $\deg(g_j) < \nu_j$

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Fundamental problems:

- Hermite–Padé approximation when  $M = x^d$ .
- Berlekamp–Massey algorithm when  $M = x^d$  and  $m = 2$ .
- Rational interpolation when  $M = (x - a_1) \cdots (x - a_d)$  with  $a_i$ 's distinct.

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Family of polynomials  
 $\text{ApproxMultMod}(\mathfrak{M}, F, \nu)$

$$\mathfrak{M} = (M_1, \dots, M_n) \text{ with}$$

$$D = \sum_{i=1}^n \deg(M_i)$$

Polynomial matrix  
 $\text{ApproxMatrixMod}(M, F, \nu)$

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**Polynomial approach:**

- Beckermann and Labahn 1994:  $\text{ApproxMultMod}$  solved in  $\tilde{O}(m^\omega \cdot D)$ .
- Neiger 2016:  $\text{ApproxMultMod}$  solved in  $\tilde{O}(\max(n, m)^{\omega-1} \cdot D)$ .
- Neiger and Vu 2017:  $\text{ApproxMatrixMod}$  solved in  $\tilde{O}(\max(n, m)^{\omega-1} \cdot D)$ .

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**Structured approach:**

- Chowdhury et al. 2015:  $\text{ApproxMultMod}$  solved in  $\tilde{O}(\max(n, m)^{\omega-1} \cdot n)$ .
- **Not done yet for  $\text{ApproxMatrixMod}$ .**

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Reduction to a structured linear system

# Structured matrices

Displacement rank approach

Example: Toeplitz matrix

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$$\mathbb{K}^{4 \times 4}$$

$$\begin{bmatrix} a_0 & a_1 & a_2 & a_3 \\ a_{-1} & a_0 & a_1 & a_2 \\ a_{-2} & a_{-1} & a_0 & a_1 \\ a_{-3} & a_{-2} & a_{-1} & a_0 \end{bmatrix}$$

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$$A \in \mathbb{K}^{n \times m}$$

Displacement rank:

For  $P \in \mathbb{K}^{n \times n}$  and  $Q \in \mathbb{K}^{m \times m}$ , we define it as the rank of the displacement operator  $\Delta[P, Q]$  such that:

$$\Delta[P, Q](A) = A - P \cdot A \cdot Q$$

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$$\begin{array}{c}
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 \begin{bmatrix}
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 \end{array}
 -
 \begin{array}{c}
 \mathbb{Z}_{n,0} \cdot A \cdot \mathbb{Z}_{m,0}^T \\
 \begin{bmatrix}
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 0 & a_{-1} & a_0 & a_1 \\
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 \end{bmatrix}
 \end{array}
 =
 \begin{array}{c}
 \text{Displacement result} \\
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Quasi-Toeplitz Structure:  $P = \mathbb{Z}_{n,0}$ ,  $Q = \mathbb{Z}_{m,0}^T$

$$\mathbb{Z}_{n,0} = \begin{bmatrix}
 0 & & & & \\
 1 & 0 & & & \\
 & \ddots & \ddots & & \\
 & & & 1 & 0
 \end{bmatrix} \in \mathbb{K}^{n \times n}$$

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Quasi-Toeplitz Structure:  $P = \mathbb{Z}_{n,0}$ ,  $Q = \mathbb{Z}_{m,0}^T$

Generators:

$(G, H) \in \mathbb{K}^{n \times \alpha} \times \mathbb{K}^{m \times \alpha}$  are called a  **$\phi_+$ -generator of length  $\alpha$  for  $A$**  if

$$\Delta[\mathbb{Z}_{n,0}, \mathbb{Z}_{m,0}^T](A) = G \cdot H^T.$$

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 \left[ \begin{array}{cccc}
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$$\begin{array}{c}
 \mathbf{G} \\
 \left[ \begin{array}{cc}
 a_0 & 1 \\
 a_{-1} & 0 \\
 a_{-2} & 0 \\
 a_{-3} & 0
 \end{array} \right] \times \mathbf{H}^T = \text{Displacement result} \\
 \mathbf{G} \in \mathbb{K}^{4 \times 2} \quad \left[ \begin{array}{cccc}
 1 & 0 & 0 & 0 \\
 0 & a_1 & a_2 & a_3
 \end{array} \right] \mathbf{H} \in \mathbb{K}^{4 \times 2} = \left[ \begin{array}{cccc}
 a_0 & a_1 & a_2 & a_3 \\
 a_{-1} & 0 & 0 & 0 \\
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 \end{array} \right] \\
 \text{Displacement rank } \alpha = 2
 \end{array}$$

# Structured linear system

Definition and previous works

$$A \in \mathbb{K}^{n \times m}$$

Problem 2:

$$\text{LinearSystem}[\mathbb{Z}_{n,0}, \mathbb{Z}_{m,0}^T, \alpha](G, H, v)$$

Using generators of A:  $G \in \mathbb{K}^{n \times \alpha}, H \in \mathbb{K}^{m \times \alpha}$

$$\text{find } u \text{ such that } A \cdot u = v$$

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Previous works:

- Kailath, Kung, Morf 1979: solved in  $\tilde{O}(\alpha \cdot \max(n, m)^2)$ .
- Kaltofen 1994, 1995: solved in  $\tilde{O}(\alpha^2 \cdot \max(n, m))$ . Better since  $\alpha \leq \min(n, m)$
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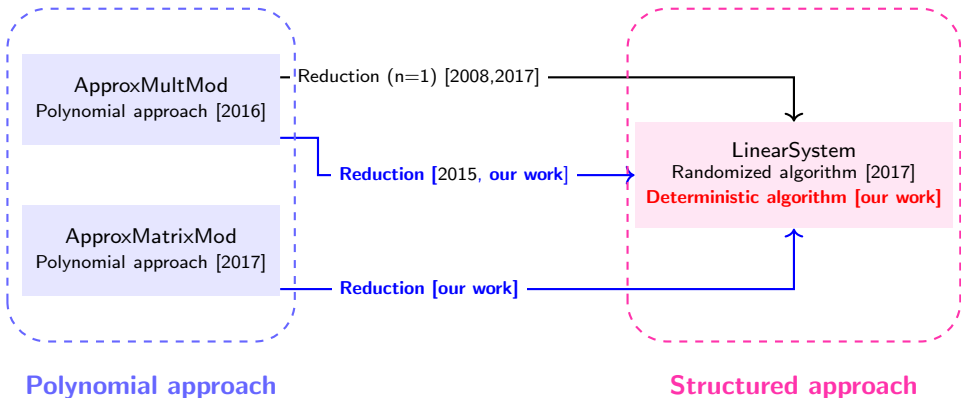
**The algorithms are probabilistic.**

The field  $\mathbb{K}$  has to be large enough for the algorithms to work with positive probability.  
 e.g.,  $15000 \times 15000$  system with  $\alpha = 30$  solved in  $< 1$  sec, but requires  $|\mathbb{K}| > 2.25 \cdot 10^8 = 15000^2$ .

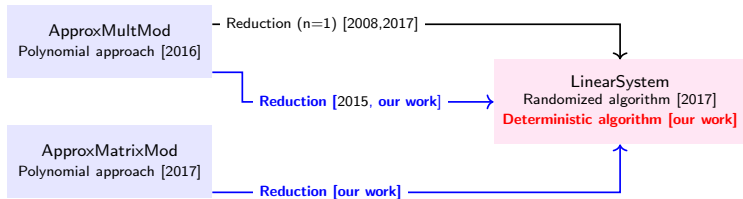
# Main results

Bridging the gap between the two approaches

Our contributions:



# Matrix modular approximation: reduction



# Matrix modular approximation

Modulus polynomial matrices

## ApproxMatrixMod

**Input:** a modulus polynomial matrix  $M \in \mathbb{K}[x]^{n \times n}$ ,  
a polynomial matrix  $F \in \mathbb{K}[x]^{n \times m}$  such that  $\text{rdeg}(F) < \text{rdeg}(M)$ ,  
a set of positive integers  $\nu = \{\nu_1, \dots, \nu_m\}$ .

**Output:** nonzero  $g \in \mathbb{K}[x]^{m \times 1}$  such that  $F \cdot g = 0 \pmod{M}$  and  $\deg(g_j) < \nu_j$ .

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M

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

$$\text{rdeg}(M) = (2, 1, 2)$$

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F

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**Output:** nonzero  $g \in \mathbb{K}[x]^{m \times 1}$  such that  $F \cdot g = 0 \text{ mod } M$  and  $\deg(g_j) < \nu_j$ .

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 F & \cdot & g & = & 0 & \text{mod} & M \\
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$$\deg(g_j) < \nu_j \text{ for } j = 1, 2, 3, 4$$

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## Theorem

The problem  $\text{ApproxMatrixMod}(M, F, \nu)$  can be solved in  $\tilde{O}(\max(m, n)^{\omega-1} \cdot D)$ .

$\text{ApproxMatrixMod}(M, F, \nu)$

Find  $g$  such that  
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Reduction

$\text{LinearSystem}(G, H, 0)$

Using generators of  $A$   
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Modulus polynomial matrices

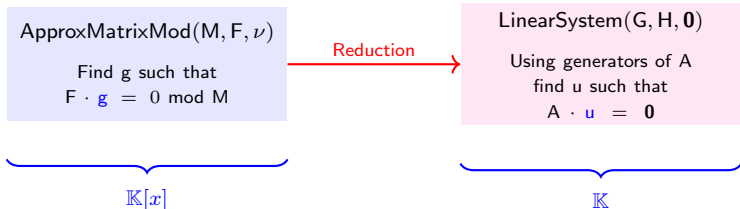
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# Multiplication matrix

From polynomials to coefficients

**Multiplication matrix:** For  $M(x) = x^d + m_{d-1}x^{d-1} + \dots + m_1x + m_0 \in \mathbb{K}[x]$

$$\mathbb{X}(M) = \begin{bmatrix} 0 & \cdots & 0 & -m_0 \\ 1 & \cdots & 0 & -m_1 \\ \vdots & \ddots & \vdots & \vdots \\ 0 & \cdots & 1 & -m_{d-1} \end{bmatrix}$$

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$$\mathbb{X}(M)^k \cdot \begin{bmatrix} p_0 \\ \vdots \\ p_{d-1} \end{bmatrix}$$

coefficient vector

$$x^k \cdot (p_0 + \dots + p_{d-1}x^{d-1}) \text{ rem } M$$

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coefficient vector

$$\rightarrow x^k \cdot (p_0 + \dots + p_{d-1}x^{d-1}) \text{ rem } M$$

Generalization to  $M$ :  $\text{rdeg}(M) = [d_1, \dots, d_n]$

Let  $D = \sum_{i=1}^n d_i$ .

$$\mathbb{X}(M) = \begin{bmatrix} \mathbb{X}(M_{11}) & C_{12} & \cdots & C_{1n} \\ C_{21} & \mathbb{X}(M_{22}) & \cdots & C_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ C_{n1} & C_{n2} & \cdots & \mathbb{X}(M_{nn}) \end{bmatrix}$$

$$C_{ij} = \begin{bmatrix} 0 & \cdots & 0 & -\text{coeff}(M_{ij}, 0) \\ 0 & \cdots & 0 & -\text{coeff}(M_{ij}, 1) \\ \vdots & \ddots & \vdots & \vdots \\ 0 & \cdots & 0 & -\text{coeff}(M_{ij}, d_i - 1) \end{bmatrix}.$$

# Definition of A

## Reduction

Let  $C = \nu_1 + \dots + \nu_m$ .

Let  $D = |\text{rdeg}(M)| = d_1 + \dots + d_n$ .

We define  $A = [A_1 \quad A_2 \quad \dots \quad A_m] \in \mathbb{K}^{D \times C}$  with:

$$A_j = \begin{bmatrix} \bar{F}_{*j} & \mathbb{X}(M) \cdot \bar{F}_{*j} & \dots & \mathbb{X}(M)^{\nu_j-1} \cdot \bar{F}_{*j} \end{bmatrix} \in \mathbb{K}^{D \times \nu_j}.$$

$\downarrow$   
 $F_{*j}$

$\downarrow$   
 $x \cdot F_{*j} \text{ rem } M$

$\downarrow$   
 $x^{\nu_j-1} \cdot F_{*j} \text{ rem } M$

Here  $\bar{F}_{*j}$  is the vector of coefficients of  $F_{*j}$  the  $j$ -th column of  $F$ .

# Definition of A

## Reduction

Let  $C = \nu_1 + \dots + \nu_m$ . **Number of unknowns**

Let  $D = |\text{rdeg}(M)| = d_1 + \dots + d_n$ . **Number of equations**

We define  $A = [A_1 \quad A_2 \quad \dots \quad A_m] \in \mathbb{K}^{D \times C}$  with:

$$A_j = \begin{bmatrix} \bar{F}_{*j} & \mathbb{X}(M) \cdot \bar{F}_{*j} & \dots & \mathbb{X}(M)^{\nu_j-1} \cdot \bar{F}_{*j} \end{bmatrix} \in \mathbb{K}^{D \times \nu_j}.$$

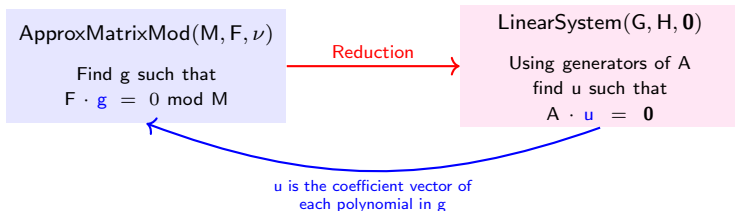
$\downarrow$   
 $F_{*j}$

$\downarrow$   
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Reduction:



# Generators of A

Reduction cost

**Goal:** Find  $(G, H)$  generators of length  $O(\max(n, m))$  for A.

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We proceed by splitting the displacement operator:

$$A - Z_{D,0} \cdot A \cdot Z_{C,0}^T = A - X(M) \cdot A \cdot Z_{C,0}^T + Y \cdot A \cdot Z_{C,0}^T - \delta \cdot A \cdot Z_{C,0}^T - N \cdot E \cdot A \cdot Z_{C,0}^T$$

$$G = \begin{bmatrix} Y & R & T & N \end{bmatrix} \in \mathbb{K}^{D \times (m+3n)} \text{ and } H = \begin{bmatrix} Z & U & U & Q \end{bmatrix} \in \mathbb{K}^{C \times (m+3n)}$$

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LinearSystem(G, H, 0)

Using generators of length  $O(\max(n, m))$  for A  
find u such that  $A \cdot u = 0$

$$\tilde{O}(\max(n, m)^{\omega-1} \cdot D)$$

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The total cost of the reduction:

$$\tilde{O}(n^{\omega-1} \cdot D + m \cdot n^{\omega-2} \cdot D + n^{\omega-1} \cdot C) \leq \tilde{O}(\max(n, m)^{\omega-1} \cdot D).$$

# High degree coefficients

## Problem overview

### LastCoeffModMat

**Input:**  $M \in \mathbb{K}[x]^{n \times n}$ ,  $F \in \mathbb{K}[x]^{n \times m}$ ,  $\nu = \{\nu_1, \dots, \nu_m\}$   
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**Output:** For  $j = 1, \dots, m$  and  $k_j = 0, \dots, \nu_j - 1$ ,  
 $\mathbf{a}_{k_j, j} = [\text{coeff}([x^{k_j} \cdot F \text{ rem } M]_{ij}, d_i - 1)]_{i=1, \dots, n}$

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$$\overline{Q_{0, j}} = 0,$$

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$$\vdots$$

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$$\rightsquigarrow \overline{Q_\nu} = \overline{M}^{-1} \cdot \overline{F} \text{ rrem } \mathbb{X}^\nu$$

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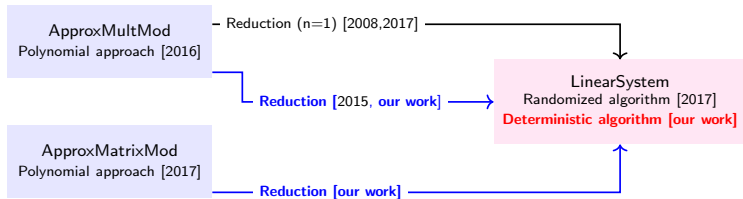
Truncated inverse product lemma:  $\tilde{O}(n^{\omega-1} \cdot C)$

- State of the art: the problem is solved **under the assumption that**  $\max(\nu) \leq 2 \cdot \frac{C}{m}$ .

Applying the algorithm on a selected set of columns of  $\overline{F}$ .

$$\rightsquigarrow \log_2(m) \text{ steps since } C = \sum_{j=1}^m \nu_j.$$

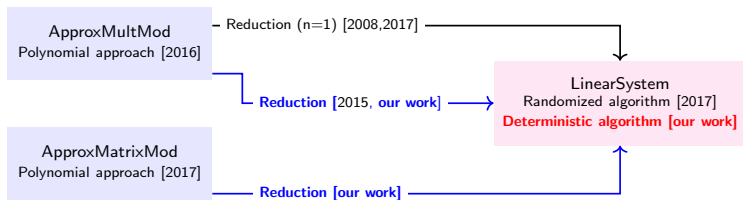
## Structured linear system: deterministic algorithm



# Quasi-Toeplitz linear system

Deterministic algorithm

We recall:



$\text{LinearSystem}[\mathbb{Z}_{n,0}, \mathbb{Z}_{m,0}^T, \alpha]$

**Input:**  $G \in \mathbb{K}^{n \times \alpha}, H \in \mathbb{K}^{m \times \alpha}$  are  $\phi_+$ -generators of length  $\alpha$  for  $A$ ,  $v \in \mathbb{K}^{n \times 1}$ .

**Output:** nonzero  $u \in \mathbb{K}^{m \times 1}$  such that  $A \cdot u = v$  if it exists, otherwise return  $\emptyset$ .

## Theorem

We can solve  $\text{LinearSystem}[\mathbb{Z}_{n,0}, \mathbb{Z}_{m,0}^T, \alpha]$  in  $\tilde{O}(\alpha^{\omega-1} \cdot \max(n, m))$  **deterministically**.

$\rightsquigarrow$  no constraint on the size of the field  $\mathbb{K}$

# Polynomial formulation

First step

$\text{LinearSystem}[\mathbb{Z}_{n,0}, \mathbb{Z}_{m,0}^T, \alpha](G, H, v)$

Using generators of  $A$ , find  $u$  such that  $A \cdot u = v$

# Polynomial formulation

First step

LinearSystem $[\mathbb{Z}_{n,0}, \mathbb{Z}_{m,0}^T, \alpha](G, H, v)$

Using generators of A, find  $u$  such that  $A \cdot u = v$

$\Sigma$ -LU decomposition of A:

$$\phi_+(A) = G \cdot H^T$$

$$\begin{aligned}
 A \cdot u = v &\Leftrightarrow \sum_{i=1}^{\alpha} \mathbb{L}(G_{*i}) \cdot \mathbb{U}(H_{*i}) \cdot u = v \\
 &\Leftrightarrow \sum_{i=1}^{\alpha} \begin{bmatrix} G_{1i} & 0 & \cdots & 0 \\ G_{2i} & G_{1i} & \ddots & \vdots \\ \vdots & \ddots & \ddots & 0 \\ G_{mi} & \cdots & G_{2i} & G_{1i} \end{bmatrix} \cdot \begin{bmatrix} H_{1i} & H_{2i} & \cdots & H_{mi} \\ 0 & H_{1i} & \ddots & \vdots \\ \vdots & \ddots & \ddots & H_{2i} \\ 0 & \cdots & 0 & H_{1i} \end{bmatrix} \cdot u = v
 \end{aligned}$$

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$$\Leftrightarrow \sum_{i=1}^{\alpha} f_i(x) \cdot (\text{rev}_{n-1}(g_i) \cdot u(x) \text{ quo } x^{n-1}) \bmod x^n = v(x)$$

# Polynomial formulation

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Let  $F = [f_1(x) \quad f_2(x) \quad \cdots \quad f_{\alpha}(x)]$ .

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LinearSystem( $G, H, v$ )

Find  $u$  such that

$$A \cdot u = v$$

Polynomial formulation

Find the basis of solutions

$$\{p \in \mathbb{K}[x]^{\alpha \times 1} \mid F \cdot p = v(x) \bmod x^n\}$$

# Approximant basis

Second step

Find the basis of solutions

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- $\mathbf{P} \in \mathbb{K}[x]^{\alpha \times \alpha}$  the basis of solutions  $\{\mathbf{q} \in \mathbb{K}[x]^{\alpha \times 1} \mid \mathbf{F} \cdot \mathbf{q} = 0 \bmod x^n\}$
- $\mathbf{s} \in \mathbb{K}[x]^{\alpha \times 1}$  such that  $\mathbf{F} \cdot \mathbf{s} = v(x) \bmod x^n$

$$\mathbf{p} = \mathbf{P} \cdot \boldsymbol{\lambda} + \mathbf{s}$$

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Approximant basis

Find  $u(x)$  and  $\lambda$  such that

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# Simultaneous extended Hermite-Padé approximation

Third step

Find  $u(x)$  and  $\lambda$  such that

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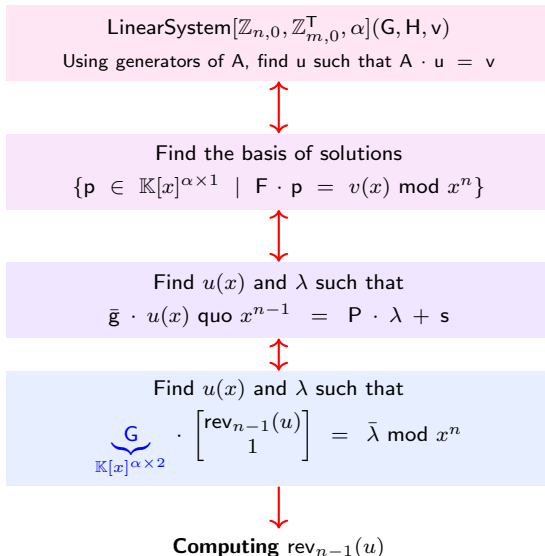
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Simultaneous extended  
Hermite-Padé approximation

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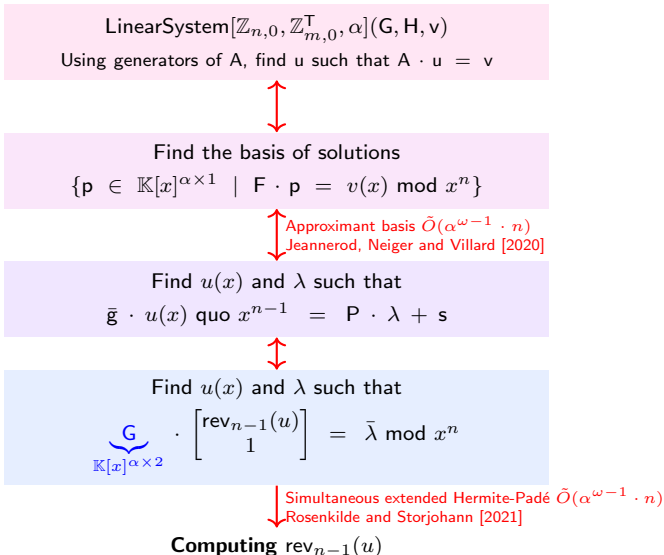
# Overview of the algorithm

And final complexity



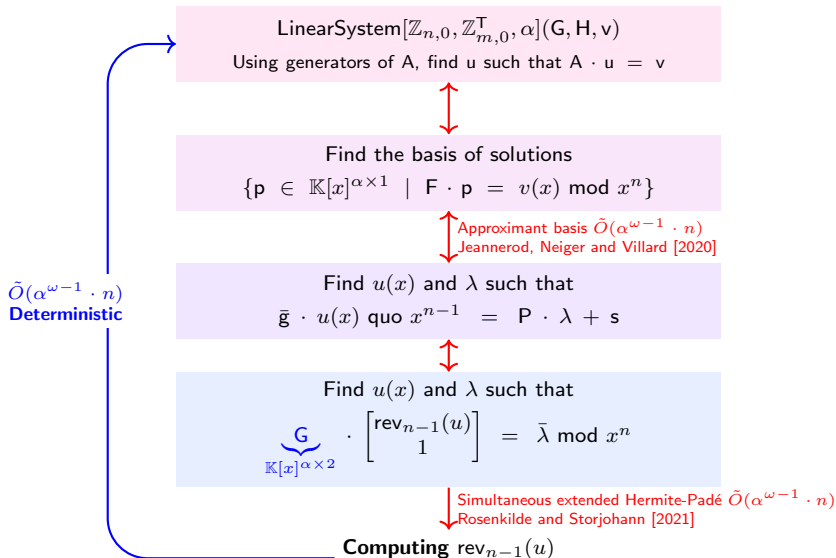
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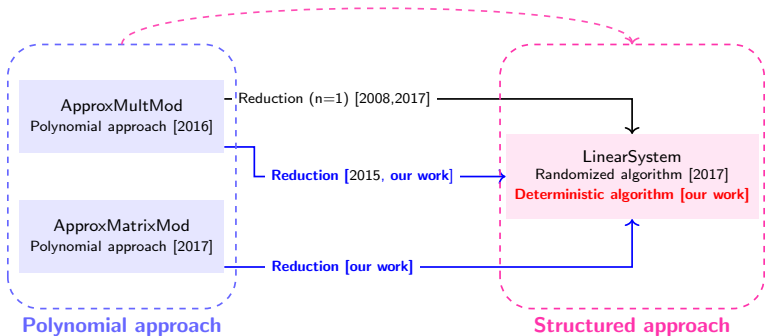
And final complexity



## Conclusion

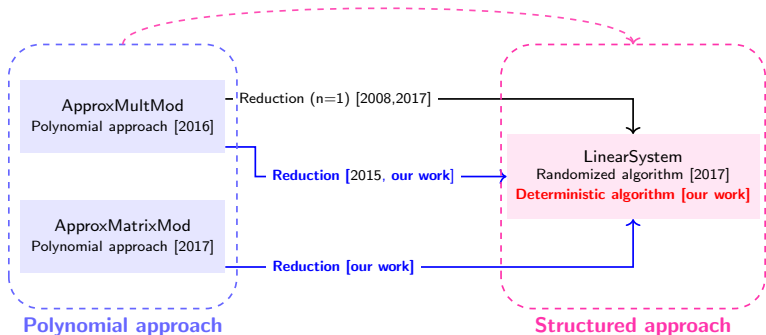
# Conclusion

Summary and future work



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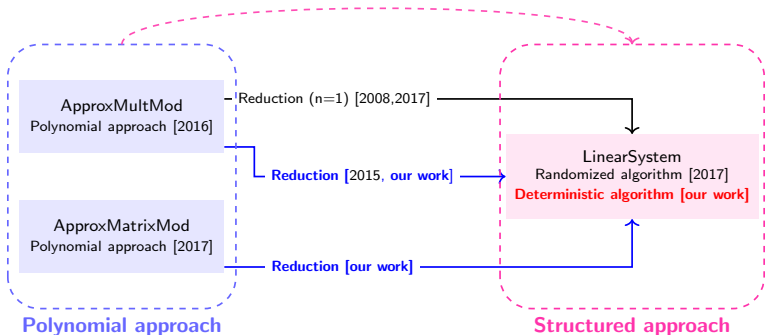


Perspectives:

- Extend our last result to other structured linear systems.
- Lift the  $\tilde{O}$  notation.

# Conclusion

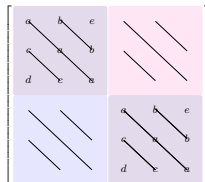
Summary and future work



## Perspectives:

- Extend our last result to other structured linear systems.
- Lift the  $\tilde{O}$  notation.
- Compute a basis of solutions for structured linear systems.
- Study cases with multiple nested structures.

e.g.,  $F = [1, p, p^2, \dots, p^{m-1}]$  for some  $p \in \mathbb{K}[x]$



**BTTB**

## Acknowledgment

# Thank you!

# Appendix

# Polynomial matrices

Modulus polynomial and row reduced

$$M \in \mathbb{K}[x]^{n \times n}$$

Modulus polynomial:

$M_{ii}$  monic,  $\deg(M_{ii}) = d_i$

$$\begin{bmatrix} M_{11} & M_{12} & \cdots & M_{1n} \\ M_{21} & M_{22} & \cdots & M_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ M_{n1} & M_{n2} & \cdots & M_{nn} \end{bmatrix}$$

$\deg(M_{ij}) < d_i$  for  $j \neq i$

Row reduced:

A row reduced matrix is a matrix whose **row leading coefficients matrix** is **invertible**.

e.g.  $\begin{bmatrix} x & x+1 \\ x & x^2 \end{bmatrix}$        $\text{rlm} = \begin{bmatrix} 1 & 1 \\ 0 & 1 \end{bmatrix}$

## Lemma

$M$  modulus polynomial  $\implies M$  is row reduced.

Proof.  $\text{rlm}$  of  $M$  is the identity matrix.

# Polynomial matrices

## Euclidean division

For any row reduced  $M \in \mathbb{K}[x]^{n \times n}$  with  
 $\text{rdeg}(M) = [d_1, \dots, d_n]$

and any  $F \in \mathbb{K}[x]^{n \times m}$  with  
 $\text{rdeg}(F) < [d_1 + k, \dots, d_n + k]$

There exists **unique** matrices  $Q, R \in \mathbb{K}[x]^{n \times m}$  such that:

$$F = M \cdot Q + R$$

with  $\text{rdeg}(R) < \text{rdeg}(M)$  and  $\text{deg}(Q) < k$ .

### Notation

$M \in \mathbb{K}[x]^{n \times m}$ :

- $\text{rdeg}(M) = [d_1, \dots, d_n]$  such that  
 $d_i = \max_{j=1, \dots, m} (\text{deg}(M_{ij}))$
- $\text{deg}(M) =$   
 $\max_{i=1, \dots, n, j=1, \dots, m} (\text{deg}(M_{ij}))$

# Applications

## Family of polynomials approximation:

- Block Wiedemann algorithm
- Factorization of integer numbers
- Polynomial multivariate systems solving

## Polynomial matrix approximation:

- Acceleration of the computation of the characteristic polynomial
- Computations for univariate matrices

## Structured linear systems:

- Error correcting codes: Reed-Solomon

# Algorithms

## Part 2

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**Algorithm 1:** Coefficient  $d - 1$  of each of the polynomials  $x^k P \text{ rem } M$  for  $k = 0, \dots, \nu$

---

**Input:**  $M = m_0 + m_1x + \dots + m_dx^d$ ,  
 $P = p_0 + p_1x + \dots + p_{d-1}x^{d-1}$ ,  $\nu$

**Output:** [coeff( $P \text{ rem } M$ ,  $d - 1$ ), coeff( $xP \text{ rem } M$ ,  $d - 1$ ), ..., coeff( $x^{\nu-1}P \text{ rem } M$ ,  $d - 1$ )]

$g = \text{rev}_d(M) \text{ rem } x^\nu$ ;

$h = \text{rev}_{d-1}(P) \text{ rem } x^\nu$ ;

$f = g^{-1} \cdot h \text{ rem } x^\nu$ ;

**for**  $i = 0$  **to**  $\nu - 1$  **do**

$a_i = \text{coeff}(f, i)$ ;

**return** [ $a_0, \dots, a_{\nu-1}$ ];

---



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### Algorithm 2: Truncated Inverse Product

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**Input:**  $G \in \mathbb{K}[x]^{n \times n}$  such that  $G$  is row reduced,  $P \in \mathbb{K}[x]^{n \times m}$  and  
 $\nu = \{\nu_1, \dots, \nu_m\}$ .

**Output:**  $G^{-1} \cdot P \text{ rem } \mathbb{X}^\nu$ , where  
 $\mathbb{X}^\nu = \text{diag}(x^{\nu_1}, x^{\nu_2}, \dots, x^{\nu_m})$ .

**Initialization:**  $P' = P$ ,  $\nu' = \nu$ ,  
 $S = \{1, \dots, m\}$ ,  $\ell = 1$ ,  $C = \sum_{j=1}^m \nu_j$ ,

$R \in \mathbb{K}[x]^{n \times m}$  with  $R = [R_{*j}]_{j=1, \dots, m}$ .  
 $S_\ell = \{j \in S \mid \nu_j \leq 2 \cdot \frac{C}{m}\}$ ; //  $S_1$

**for**  $\ell = 1$  **to**  $\lceil \log_2(C/m) \rceil$  **do**

$P' \in \mathbb{K}[x]^{n \times |S_\ell|}$ , where  $P' = [P_{*j}]_{j \in S_\ell}$ ;

$\nu' = (\nu_j)_{j \in S_\ell}$ ;

$\mathbb{X}^{\nu'} \in \mathbb{K}[x]^{|S_\ell| \times |S_\ell|}$ , where

$\mathbb{X}^{\nu'} = \text{diag}(x^{\nu_j})_{j \in S_\ell}$ ;

  Computing  $B = G^{-1} \cdot P' \text{ rem } \mathbb{X}^{\nu'}$ ;

$i = 1$ ;

**for**  $j \in S_\ell$  **do**

$R_{*j} = B_{*i}$ ;

$i = i + 1$ ;

$S_{\ell+1} = \{j \in S \mid 2^\ell \cdot \frac{C}{m} < \nu_j \leq 2^{\ell+1} \cdot \frac{C}{m}\}$ ;

**return**  $R$ ;

---

# Algorithms

## Part 2

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**Algorithm 3:** Solving a structured linear system

---

**Input:**  $G \in \mathbb{K}^{n \times \alpha}$ ,  $H \in \mathbb{K}^{m \times \alpha}$ ,  $v \in \mathbb{K}^{n \times 1}$

**Output:** LinearSystem( $G, H, v$ )

// Transform to polynomial matrices

**for**  $j = 0$  **to**  $\alpha - 1$  **do**

$f_j(x) = \sum_{i=0}^{n-1} G_{ij} x^i$ ;

$F = [f_0(x), f_1(x), \dots, f_{\alpha-1}(x)]$ ;

**for**  $j = 0$  **to**  $\alpha - 1$  **do**

$g_j(x) = \sum_{i=0}^{m-1} H_{ij} x^i$ ;

$v(x) = \sum_{i=0}^{n-1} v_i x^i$ ;

// Approximant basis computation

$\begin{bmatrix} P & S \\ 0 & \mu \end{bmatrix} = \text{ApproximantBasis}([F \quad -v], n)$ ;

**if**  $\deg(\mu) > 0$  **then**

**return**  $\emptyset$ ;

**else**

$\deg(\mu) = 0$

$\mu(x) = 1$ ;

// Retrieve the solution

$B = [\bar{P}^{-1} \cdot g \quad -\bar{P}^{-1} \cdot \text{rev}_{n-1}(s)]$ ;

**return**  $[\text{coeff}(u, \alpha - 1), \dots, \text{coeff}(u, 0)]^T$ ;

---



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**Algorithm 4:** Family of polynomials approximation

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**Input:**  $\mathfrak{M} \in \mathbb{K}[x]^n$ ,  $F \in \mathbb{K}[x]^{n \times m}$ ,  
 $\nu = \{\nu_1, \dots, \nu_m\}$

**Output:** ApproxMultMod( $\mathfrak{M}, F, \nu$ )

$A \in \mathbb{K}^{D \times C}$ ;

**for**  $i = 1$  **to**  $n$  **do**

**for**  $j = 1$  **to**  $m$  **do**

$f_{ij} \in \mathbb{K}^{d_i}$  is the vector of coefficients;

$A_{ij} =$

$\begin{bmatrix} f_{ij} & \mathbb{X}(M_i) f_{ij} & \dots & \mathbb{X}(M_i)^{\nu_j - 1} f_{ij} \end{bmatrix}$ ;

Compute  $G, H \in \mathbb{K}^{* \times O(\max(m, n))}$ ;

**return**

  LinearSystem $[\mathbb{Z}_{D,0}, \mathbb{Z}_{C,0}^T, O(\max(m, n))](G, H)$ ;

---

# High degree coefficients

Problem definition

## Theorem

The problem  $\text{LastCoeffModMat}(M, F, \nu)$  is solved in  $\tilde{O}(n^{\omega-1} \cdot C)$ .

**Proof.** For  $M, F$  and  $k$ , there exist unique  $Q_k$  and  $R_k$  such that  $x^k \cdot F = M \cdot Q_k + R_k$

$$\bar{M} = \mathbb{X}^d \cdot M(x^{-1}),$$

$$\bar{F} = \mathbb{X}^{d-1} \cdot F(x^{-1}),$$

$$\bar{Q}_k = x^{k-1} \cdot Q_k(x^{-1}),$$

$$\bar{R}_k = \mathbb{X}^{d-1} \cdot R_k(x^{-1}).$$

$$\bar{F} = \bar{M} \cdot \bar{Q}_k + x^k \cdot \bar{R}_k$$

We rewrite the identity with  $k = k_j$  for each column  $j$  as:

$$\bar{F}_{*j} = \bar{M} \cdot \bar{Q}_{k_j,j} + x^{k_j} \cdot \bar{R}_{k_j,j},$$

We get  $\bar{Q}_{k_j,j} = \bar{M}^{-1} \cdot \bar{F}_{*j} \text{ rem } x^{k_j}$ .

# High degree coefficients

Correctness

$$\overline{Q_{k_j,j}} = \overline{M}^{-1} \cdot \overline{F_{*j}} \text{ rem } x^{k_j}$$

## Lemma

For each  $j$ ,  $\overline{Q_{\nu_j,j}}$  (the quotient with  $k_j = \nu_j$ ) contains the coefficients of degree  $d_i - 1$  of  $[x^{k_j} \cdot F \text{ rem } M]_{ij}$  for each  $i$ .

**Proof.** For one column  $j$

We recall the euclidean division  $x^{k_j} \cdot F_{*j} = M \cdot Q_{k_j,j} + R_{k_j,j}$

We multiply by  $x$  and get  $x^{k_j+1} \cdot F_{*j} = M \cdot (x \cdot Q_{k_j,j}) + (x \cdot R_{k_j,j})$

Note that  $x^{k_j+1} \cdot F_{*j} =$

$$M \cdot (x \cdot Q_{k_j,j}) + \left( \begin{bmatrix} \text{coeff}([x^{k_j} \cdot F \text{ rem } M]_{1j}, d_1 - 1) \\ \vdots \\ \text{coeff}([x^{k_j} \cdot F \text{ rem } M]_{nj}, d_n - 1) \end{bmatrix} \cdot \begin{bmatrix} x^{d_1} & 0 & \dots & 0 \\ 0 & x^{d_2} & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & x^{d_n} \end{bmatrix} + R'_{k_j,j} \right)$$

For ease of notation, we denote  $x^{k_j+1} \cdot F_{*j} = M \cdot (x \cdot Q_{k_j,j}) + (\mathbf{a}_{k_j,j} \cdot \mathbb{X}^d + R'_{k_j,j})$

Since  $M$  is modulus polynomial, we can write  $M = \mathbb{X}^d - M'$  and get:

$$x^{k_j+1} \cdot F_{*j} = M \cdot (x \cdot Q_{k_j,j} + \mathbf{a}_{k_j,j}) + (\mathbf{a}_{k_j,j} \cdot M' + x \cdot R'_{k_j,j})$$

# High degree coefficients

Correctness

$$\overline{Q_{k_j,j}} = \overline{M}^{-1} \cdot \overline{F_{*j}} \text{ rem } x^{k_j}$$

## Lemma

For each  $j$ ,  $\overline{Q_{\nu_j,j}}$  (the quotient with  $k_j = \nu_j$ ) contains the coefficients of degree  $d_i - 1$  of  $[x^{k_j} \cdot F \text{ rem } M]_{ij}$  for each  $i$ .

**Proof.** For one column  $j$

$$x^{k_j+1} \cdot F_{*j} = M \cdot (x \cdot Q_{k_j,j} + \mathbf{a}_{k_j,j}) + (\mathbf{a}_{k_j,j} \cdot M' + x \cdot \mathbb{R}'_{k_j,j})$$

By uniqueness of the quotient and remainder, we have:

- $Q_{k_j+1,j} = x \cdot Q_{k_j,j} + \mathbf{a}_{k_j,j}$ ,
- $R_{k_j+1,j} = \mathbf{a}_{k_j,j} \cdot M' + x \cdot \mathbb{R}'_{k_j,j}$ .

# High degree coefficients

Correctness

$$\overline{Q_{k_j,j}} = \overline{M}^{-1} \cdot \overline{F_{*j}} \text{ rem } x^{k_j}$$

## Lemma

For each  $j$ ,  $\overline{Q_{\nu_j,j}}$  (the quotient with  $k_j = \nu_j$ ) contains the coefficients of degree  $d_i - 1$  of  $[x^{k_j} \cdot F \text{ rem } M]_{i,j}$  for each  $i$ .

**Proof.** For one column  $j$   
We focus on the quotient:

$$Q_{k_j+1,j} = x \cdot Q_{k_j,j} + a_{k_j,j}$$

and enroll the recurrence:

$$\overline{Q_{0,j}} = 0,$$

$$\overline{Q_{1,j}} = x^0 \cdot a_{0,j} + 0 = a_{0,j},$$

$$\overline{Q_{2,j}} = x^1 \cdot a_{0,j} + a_{1,j} = x^1 \cdot a_{0,j} + a_{1,j},$$

$$\vdots$$

$$\overline{Q_{\nu_j,j}} = x^{\nu_j-1} \cdot a_{0,j} + x^{\nu_j-2} \cdot a_{1,j} + \dots + x^0 \cdot a_{\nu_j-1,j} .$$

# High degree coefficients

Complexity

$$\overline{Q_{\nu_j, j}} = \overline{M}^{-1} \cdot \overline{F_{*j}} \text{ rem } x^{\nu_j}$$

## Lemma

We can compute  $\overline{Q_{\nu}}$  in  $\tilde{O}(n^{\omega-1} \cdot C)$ .

### Proof.

We define  $\overline{Q_{\nu}} \in \mathbb{K}[x]^{n \times m}$  as  $\begin{bmatrix} \overline{Q_{\nu_1, 1}} & \overline{Q_{\nu_2, 2}} & \cdots & \overline{Q_{\nu_m, m}} \end{bmatrix}$

Since we can verify that

$$\begin{aligned} \overline{M}^{-1} \cdot \overline{F} \text{ rrem } \mathbb{X}^{\nu} &= \begin{bmatrix} \overline{M}^{-1} \cdot \overline{F_{*1}} \text{ rem } x^{\nu_1} & \overline{M}^{-1} \cdot \overline{F_{*2}} \text{ rem } x^{\nu_2} & \cdots & \overline{M}^{-1} \cdot \overline{F_{*m}} \text{ rem } x^{\nu_m} \end{bmatrix} \\ &= \begin{bmatrix} \overline{Q_{\nu_1, 1}} & \overline{Q_{\nu_2, 2}} & \cdots & \overline{Q_{\nu_m, m}} \end{bmatrix}, \end{aligned}$$

it is equivalent to compute  $\overline{Q_{\nu}} = \overline{M}^{-1} \cdot \overline{F} \text{ rem } \mathbb{X}^{\nu}$ , where

$$\begin{bmatrix} x^{\nu_1} & 0 & \cdots & 0 \\ 0 & x^{\nu_2} & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & x^{\nu_m} \end{bmatrix}$$

# High degree coefficients

Complexity

$$\overline{Q_{\nu_j, j}} = \overline{M}^{-1} \cdot \overline{F_{*j}} \text{ rem } x^{\nu_j}$$

## Lemma

We can compute  $\overline{Q_{\nu}}$  in  $\tilde{O}(n^{\omega-1} \cdot C)$ .

**Proof.**

We compute  $\overline{Q_{\nu}} = \overline{M}^{-1} \cdot \overline{F} \text{ rem } \mathbb{X}^{\nu}$ .

- State of the art: the problem is solved **under the assumption that  $\max(\nu) \leq 2 \cdot \frac{C}{m}$** .

Applying the algorithm on a selected set of columns of  $\overline{F}$ :

i.e. For a given step  $\ell$  of the algorithm, we have:

- $L_{\ell} = \{j \in \{1, \dots, m\} \mid \nu_j \leq 2^{\ell} \cdot \frac{C}{m}\}$
- $F' = [P_{*j}]_{j \in L_{\ell}} \in \mathbb{K}[x]^{n \times |L_{\ell}|}$ .
- $\nu' = (\nu_j)_{j \in L_{\ell}}$ , we get  $\mathbb{X}^{\nu'} \in \mathbb{K}[x]^{\frac{m}{2^{\ell}} \times \frac{m}{2^{\ell}}}$ .

**Number of steps:**  $\lceil \log_2(m) \rceil$ .

**Cost of each step:**  $\tilde{O}(n^{\omega-1} \cdot C)$ ,  
(since  $|L_{\ell}| \geq \frac{m}{2^{\ell}}$ ).

**Total cost:**

$\tilde{O}(n^{\omega-1} \cdot C \cdot \log_2(m)) = \tilde{O}(n^{\omega-1} \cdot C)$ .

□

# High degree coefficients

## Summary

$$\overline{Q_{\nu,j,j}} = \overline{M}^{-1} \cdot \overline{F_{*j}} \text{ rem } x^{\nu j}$$

Correctness:

### Lemma

For each  $j$ ,  $\overline{Q_{\nu,j,j}}$  contains the coefficients of degree  $d_i - 1$  of  $[x^{k_j} \cdot F \text{ rem } M]_{ij}$  for each  $i$ .

Complexity:

### Lemma

We can compute  $\overline{Q_{\nu}}$  in  $\tilde{O}(n^{\omega-1} \cdot C)$ .

Final result:

### Theorem

The problem  $\text{LastCoeffModMat}(M, F, \nu)$  is solved in  $\tilde{O}(n^{\omega-1} \cdot C)$ .

# Notes

## Berlekamp-Massey:

Finds the minimal recurrence (of order  $d$ ) from the first  $2d$  terms of a linearly recurrent sequence, and this is equivalent to compute a basis of relations for  $F = [S, -1]$  modulo  $x^{2d}$ , where the input  $S$  is the polynomial defined from the  $2d$  first terms of the sequence

- Dornstetter 1987 and Sugiyama et al. 1975: quadratic
- Brent-Gustavson-Yun 1980: quasi-linear

## Randomization in current structured linear system algorithms:

The pivots in a Gaussian elimination-type of algorithm must be on the diagonal, and authors ensure this property by multiplying the input matrix by a random choice of matrices (with a special form for efficiency).