

## Decoding Interleaved Gabidulin Codes using Alekhnovich's Algorithm

Sven Puchinger<sup>1</sup>, Sven Müelich<sup>1</sup>, David Mödinger<sup>2</sup>,  
Johan Rosenkilde né Nielsen<sup>3</sup>, Martin Bossert<sup>1</sup>

<sup>1</sup>Institute of Communications Engineering, Ulm University, Germany

<sup>2</sup>Institute of Distributed Systems, Ulm University, Germany

<sup>3</sup>Department of Applied Mathematics & Computer Science, Technical University of Denmark, Denmark

ACCT, June 22, 2016

## 1 Motivation

## 2 Skew Variant of Alekhnovich's Algorithm

## 3 Conclusion

## 1 Motivation

## 2 Skew Variant of Alekhnovich's Algorithm

## 3 Conclusion

$\mathbb{F}[x; \sigma]$  : Polynomials of the form

$$a = \sum_{i=0}^d a_i x^i : a_i \in \mathbb{F}_{q^m}, d \in \mathbb{N}$$

# Skew Polynomial Ring

$\mathbb{F}[x; \sigma]$  : Polynomials of the form

$$a = \sum_{i=0}^d a_i x^i : a_i \in \mathbb{F}_{q^m}, d \in \mathbb{N}$$

Addition (+)       $a + b = \sum_i (a_i + b_i) x^i$

# Skew Polynomial Ring

$\mathbb{F}[x; \sigma]$  : Polynomials of the form

$$a = \sum_{i=0}^d a_i x^i : a_i \in \mathbb{F}_{q^m}, d \in \mathbb{N}$$

Addition (+)       $a + b = \sum_i (a_i + b_i)x^i$

Multiplication ( $\cdot$ )     $a \cdot b = \sum_i \left( \sum_{j=0}^i a_j \sigma^j(b_{i-j}) \right) x^i$       (non-commutative)

$\mathbb{F}[x; \sigma]$  : Polynomials of the form

$$a = \sum_{i=0}^d a_i x^i : a_i \in \mathbb{F}_{q^m}, d \in \mathbb{N}$$

Addition (+)       $a + b = \sum_i (a_i + b_i)x^i$

Multiplication ( $\cdot$ )     $a \cdot b = \sum_i \left( \sum_{j=0}^i a_j \sigma^j(b_{i-j}) \right) x^i$       (non-commutative)

Degree                 $\deg a = \max\{i : a_i \neq 0\}$

$\mathbb{F}[x; \sigma]$  : Polynomials of the form

$$a = \sum_{i=0}^d a_i x^i : a_i \in \mathbb{F}_{q^m}, d \in \mathbb{N}$$

Addition (+)       $a + b = \sum_i (a_i + b_i)x^i$

Multiplication (·)     $a \cdot b = \sum_i \left( \sum_{j=0}^i a_j \sigma^j(b_{i-j}) \right) x^i$       (non-commutative)

Degree                 $\deg a = \max\{i : a_i \neq 0\}$

## Properties

- Linearized polynomials isomorphic to  $\mathbb{F}_{q^m}[x; \cdot^q]$  (Frobenius automorphism)
- Evaluation is linear map

## Gabidulin Codes

- Rank-metric analogues of Reed–Solomon codes
- Defined using evaluation of skew polynomials
- Interleaving of  $\ell$  codes  $\Rightarrow$  increase decoding radius to  $\frac{\ell}{\ell+1}(d - 1)$  (w.h.p.).

## Gabidulin Codes

- Rank-metric analogues of Reed–Solomon codes
- Defined using evaluation of skew polynomials
- Interleaving of  $\ell$  codes  $\Rightarrow$  increase decoding radius to  $\frac{\ell}{\ell+1}(d - 1)$  (w.h.p.).

## Key Equation

$n, k_1, \dots, k_\ell \in \mathbb{N}$  code parameters,  
 $S_1, \dots, S_\ell \in \mathbb{F}[x, \sigma]$  syndromes,  
 $\Lambda \in \mathbb{F}[x, \sigma]$  error span polynomial.

$$\Lambda S_i \equiv \Omega_i \pmod{x^{n-k_i}}$$

$$\deg \Omega_i < \deg \Lambda$$

$\Lambda$  minimal degree

## Shift Register Synthesis Problem

**Given**  $s_1, \dots, s_\ell, g_1, \dots, g_\ell \in \mathbb{F}[x, \sigma]$ ,  
**find**  $\lambda, \omega_1, \dots, \omega_\ell \in \mathbb{F}[x, \sigma]$ , s.t.  $\forall i$

$$\lambda s_i \equiv \omega_i \pmod{g_i}$$

$$\deg \omega_i < \deg \lambda$$

$\lambda$  minimal degree

### Leading Position

Rightmost position of maximal degree.

$$\boldsymbol{v} = \begin{bmatrix} x^2 & 1 & x & \boxed{x^2} & x \end{bmatrix} \Rightarrow \text{LP}(\boldsymbol{v}) = 3$$

### Leading Position

Rightmost position of maximal degree.

$$\mathbf{v} = \begin{bmatrix} x^2 & 1 & x & \boxed{x^2} & x \end{bmatrix} \Rightarrow \text{LP}(\mathbf{v}) = 3$$

### weak Popov form

Matrix with all leading positions different.

$$\begin{bmatrix} x^2 & \boxed{x^2} & 1 \\ \boxed{x^3} & x & x^2 \\ 1 & x^4 & \boxed{x^4} \end{bmatrix}$$

*Shift Register Synthesis Problem*

**Given**  $s_1, \dots, s_\ell, g_i, \dots, g_\ell \in \mathbb{F}[x, \sigma]$ , **find**  $\lambda, \omega_1, \dots, \omega_\ell \in \mathbb{F}[x, \sigma]$ , **s.t.**  $\forall i$

$$\lambda s_i \equiv \omega_i \pmod{g_i} \quad (1)$$

$$\deg \omega_i < \deg \lambda \quad (2)$$

$$\lambda \text{ minimal degree} \quad (3)$$

## Shift Register Synthesis Problem

**Given**  $s_1, \dots, s_\ell, g_i, \dots, g_\ell \in \mathbb{F}[x, \sigma]$ , **find**  $\lambda, \omega_1, \dots, \omega_\ell \in \mathbb{F}[x, \sigma]$ , **s.t.**  $\forall i$

$$\lambda s_i \equiv \omega_i \pmod{g_i} \quad (1)$$

$$\deg \omega_i < \deg \lambda \quad (2)$$

$$\lambda \text{ minimal degree} \quad (3)$$

Solution of (1) in row span of

$$\mathbf{B} = \begin{bmatrix} 1 & s_1 & s_2 & \dots & s_\ell \\ & g_1 & & & \\ & & g_2 & & \\ & & & \ddots & \\ & & & & g_\ell \end{bmatrix}$$

## Shift Register Synthesis Problem

**Given**  $s_1, \dots, s_\ell, g_1, \dots, g_\ell \in \mathbb{F}[x, \sigma]$ , **find**  $\lambda, \omega_1, \dots, \omega_\ell \in \mathbb{F}[x, \sigma]$ , **s.t.**  $\forall i$

$$\lambda s_i \equiv \omega_i \pmod{g_i} \quad (1)$$

$$\deg \omega_i < \deg \lambda \quad (2)$$

$$\lambda \text{ minimal degree} \quad (3)$$

Solution of (1) in row span of

$$\mathbf{B} = \left[ \begin{array}{ccccc} 1 & s_1 & s_2 & \dots & s_\ell \\ & g_1 & & & \\ & & g_2 & & \\ & & & \ddots & \\ & & & & g_\ell \end{array} \right]$$

## Shift Register Synthesis Problem

**Given**  $s_1, \dots, s_\ell, g_1, \dots, g_\ell \in \mathbb{F}[x, \sigma]$ , **find**  $\lambda, \omega_1, \dots, \omega_\ell \in \mathbb{F}[x, \sigma]$ , **s.t.**  $\forall i$

$$\lambda s_i \equiv \omega_i \pmod{g_i} \quad (1)$$

$$\deg \omega_i < \deg \lambda \quad (2)$$

$$\lambda \text{ minimal degree} \quad (3)$$

Solution of (1) in row span of

$$\mathbf{B} = \left[ \begin{array}{cccccc} 1 & s_1 & s_2 & \dots & s_\ell \\ & \boxed{g_1} & & & \\ & & \boxed{g_2} & & \\ & & & \ddots & \\ & & & & \boxed{g_\ell} \end{array} \right] \xrightarrow{\substack{\text{bring in wPf} \\ \text{row operations}}} \left[ \begin{array}{cccccc} & & \square & & & & \\ & & & \square & & & \\ & & & & \square & & \\ & & & & & \square & \\ & & & & & & \square \end{array} \right]$$

## Shift Register Synthesis Problem

**Given**  $s_1, \dots, s_\ell, g_1, \dots, g_\ell \in \mathbb{F}[x, \sigma]$ , **find**  $\lambda, \omega_1, \dots, \omega_\ell \in \mathbb{F}[x, \sigma]$ , **s.t.**  $\forall i$

$$\lambda s_i \equiv \omega_i \pmod{g_i} \quad (1)$$

$$\deg \omega_i < \deg \lambda \quad (2)$$

$$\lambda \text{ minimal degree} \quad (3)$$

Solution of (1) in row span of

$$B = \left[ \begin{array}{cccccc} 1 & s_1 & s_2 & \dots & s_\ell \\ & g_1 & & & & \\ & & g_2 & & & \\ & & & \ddots & & \\ & & & & g_\ell & \end{array} \right] \xrightarrow{\text{bring in wPf}} \left[ \begin{array}{cccccc} & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ \lambda & & \omega_1 & \omega_2 & \dots & \omega_\ell \end{array} \right]$$

The diagram shows the matrix  $B$  on the left and its row echelon form on the right. The matrix  $B$  has columns for the constant term 1 and shift register states  $s_1, s_2, \dots, s_\ell$ . Below each state column, there is a box containing a generator polynomial  $g_1, g_2, \dots, g_\ell$ . An arrow labeled "bring in wPf" points from the left matrix to the right matrix, indicating the row reduction process. The right matrix is in row echelon form, with the first column containing the value  $\lambda$  and the subsequent columns containing the values  $\omega_1, \omega_2, \dots, \omega_\ell$ . The other entries in the right matrix are represented by empty boxes.

## Shift Register Synthesis Problem

**Given**  $s_1, \dots, s_\ell, g_1, \dots, g_\ell \in \mathbb{F}[x, \sigma]$ , **find**  $\lambda, \omega_1, \dots, \omega_\ell \in \mathbb{F}[x, \sigma]$ , **s.t.**  $\forall i$

$$\lambda s_i \equiv \omega_i \pmod{g_i} \quad (1)$$

$$\deg \omega_i < \deg \lambda \quad (2)$$

$$\lambda \text{ minimal degree} \quad (3)$$

Solution of (1) in row span of

$$\mathbf{B} = \left[ \begin{array}{cccccc} 1 & s_1 & s_2 & \dots & s_\ell \\ & g_1 & & & & \\ & & g_2 & & & \\ & & & \ddots & & \\ & & & & g_\ell & \end{array} \right] \xrightarrow{\substack{\text{bring in wPf} \\ \text{row operations}}} \left[ \begin{array}{cccccc} & & & & & \\ & & & & & \\ & & & & & \\ & & & & & \\ \lambda & & \omega_1 & \omega_2 & \dots & \omega_\ell \\ & & & & & \\ & & & & & \end{array} \right]$$

Question: How fast can we do that?

- $O(\ell n^2)$  [Li, Nielsen, Puchinger, Sidorenko, WCC'15]
- $O(\ell^3 n^{1.69})$  [this paper]  $\leftarrow \mathbb{F}[x, \sigma]$ -variant of [Alekhnovich, 2005]

## 1 Motivation

## 2 Skew Variant of Alekhnovich's Algorithm

## 3 Conclusion

## Degree of a Vector/Matrix

- $\deg \mathbf{m} = \max_i \{\deg m_i\}$
- $\deg \mathbf{M} = \sum_i \deg \mathbf{m}_i$

## Degree of a Vector/Matrix

- $\deg \mathbf{m} = \max_i \{\deg m_i\}$
- $\deg \mathbf{M} = \sum_i \deg \mathbf{m}_i$

$$\deg \begin{bmatrix} x^2 & x \\ x^3 & x^3 \end{bmatrix} = 2 + 3 = 5$$

## Degree of a Vector/Matrix

- $\deg \mathbf{m} = \max_i \{\deg m_i\}$
- $\deg \mathbf{M} = \sum_i \deg \mathbf{m}_i$

$$\deg \begin{bmatrix} x^2 & x \\ x^3 & x^3 \end{bmatrix} = 2 + 3 = 5$$

## Accuracy Approximation/Length

$$a = \boxed{\phantom{0}}^{x^d}_{x^0}$$

## Degree of a Vector/Matrix

- $\deg \mathbf{m} = \max_i \{\deg m_i\}$
- $\deg \mathbf{M} = \sum_i \deg \mathbf{m}_i$

$$\deg \begin{bmatrix} x^2 & x \\ x^3 & x^3 \end{bmatrix} = 2 + 3 = 5$$

## Accuracy Approximation/Length

$$a = \begin{bmatrix} \square & x^d \\ & x^0 \end{bmatrix} \longrightarrow a|_t = \begin{bmatrix} \square & x^d \\ \dots & x^{d-t+1} \\ \dots & x^0 \end{bmatrix}$$

## Degree of a Vector/Matrix

- $\deg \mathbf{m} = \max_i \{\deg m_i\}$
- $\deg \mathbf{M} = \sum_i \deg \mathbf{m}_i$

$$\deg \begin{bmatrix} x^2 & x \\ x^3 & x^3 \end{bmatrix} = 2 + 3 = 5$$

## Accuracy Approximation/Length

$$a = \begin{array}{c|c} x^d & \\ \hline & \end{array} \longrightarrow a|_t = \begin{array}{c|c} x^d & \\ \hline x^0 & \end{array} \dots \begin{array}{c|c} x^d & \\ \hline x^0 & \end{array} \quad \text{len}(a) = t$$

## Degree of a Vector/Matrix

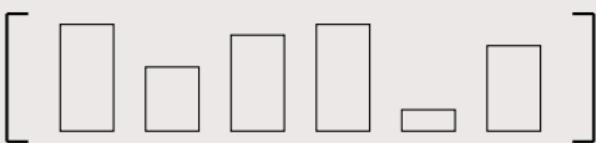
- $\deg \mathbf{m} = \max_i \{\deg m_i\}$
- $\deg \mathbf{M} = \sum_i \deg \mathbf{m}_i$

$$\deg \begin{bmatrix} x^2 & x \\ x^3 & x^3 \end{bmatrix} = 2 + 3 = 5$$

## Accuracy Approximation/Length

$$a = \begin{bmatrix} \square & x^d \\ & \longrightarrow a|_t = \begin{bmatrix} \square & x^d \\ x^0 & \dots & x^d \\ & \dots & x^{d-t+1} \\ & \dots & x^0 \end{bmatrix} \end{bmatrix}$$

$\text{len}(a) = t$



## Degree of a Vector/Matrix

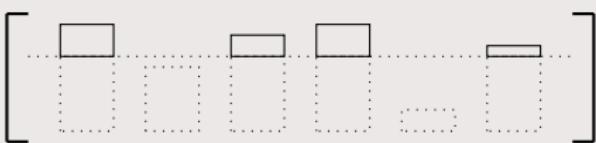
- $\deg \mathbf{m} = \max_i \{\deg m_i\}$
- $\deg \mathbf{M} = \sum_i \deg \mathbf{m}_i$

$$\deg \begin{bmatrix} x^2 & x \\ x^3 & x^3 \end{bmatrix} = 2 + 3 = 5$$

## Accuracy Approximation/Length

$$a = \begin{bmatrix} \square & x^d \\ & \longrightarrow a|_t = \square & x^d \\ x^0 & & x^{d-t+1} \\ & \diagup & \uparrow \\ & & \square \\ & & x^0 \end{bmatrix}$$

$\text{len}(a) = t$



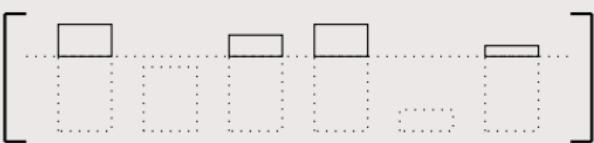
## Degree of a Vector/Matrix

- $\deg \mathbf{m} = \max_i \{\deg m_i\}$
- $\deg \mathbf{M} = \sum_i \deg \mathbf{m}_i$

$$\deg \begin{bmatrix} x^2 & x \\ x^3 & x^3 \end{bmatrix} = 2 + 3 = 5$$

## Accuracy Approximation/Length

$$a = \begin{bmatrix} \square & x^d \\ & \longrightarrow a|_t = \square & x^d \\ x^0 & & x^{d-t+1} \\ & \uparrow & \\ & \text{len}(a) = t & \end{bmatrix}$$



## Multiplication of Length- $t$ Matrices

Multiply polynomials of length  $\leq t$  ( $\omega$  matrix mult. exponent):

$$\mathcal{M}(t) \in O(t^{\frac{\omega+1}{2}}) \subseteq O(t^{1.69})$$

$(\ell + 1) \times (\ell + 1)$ -matrices:  $O(\ell^3 \mathcal{M}(t))$

## Simple Transformation

Find two rows  $\mathbf{m}_i$  and  $\mathbf{m}_j$  with

$$\text{LP}(\mathbf{m}_i) = \text{LP}(\mathbf{m}_j) \text{ and } \deg \mathbf{m}_i \geq \deg \mathbf{m}_j.$$

Replace  $\mathbf{m}_i \leftarrow \mathbf{m}_i - \alpha x^\delta \mathbf{m}_j$ , s.t. leading monomial at  $\text{LP}(\mathbf{m}_i)$  cancels.

Remember row operation by matrix  $\mathbf{U}$ .

$$\mathbf{M} = \begin{bmatrix} x^2 & \boxed{x^2 + x} \\ 1 & \boxed{x} \end{bmatrix}$$

## Simple Transformation

Find two rows  $\mathbf{m}_i$  and  $\mathbf{m}_j$  with

$$\text{LP}(\mathbf{m}_i) = \text{LP}(\mathbf{m}_j) \text{ and } \deg \mathbf{m}_i \geq \deg \mathbf{m}_j.$$

Replace  $\mathbf{m}_i \leftarrow \mathbf{m}_i - \alpha x^\delta \mathbf{m}_j$ , s.t. leading monomial at  $\text{LP}(\mathbf{m}_i)$  cancels.

Remember row operation by matrix  $\mathbf{U}$ .

$$\mathbf{M} = \begin{bmatrix} x^2 & \boxed{x^2 + x} \\ 1 & \boxed{x} \end{bmatrix} \xrightarrow{\mathbf{m}_1 \leftarrow \mathbf{m}_1 - x\mathbf{m}_2}$$

## Simple Transformation

Find two rows  $\mathbf{m}_i$  and  $\mathbf{m}_j$  with

$$\text{LP}(\mathbf{m}_i) = \text{LP}(\mathbf{m}_j) \text{ and } \deg \mathbf{m}_i \geq \deg \mathbf{m}_j.$$

Replace  $\mathbf{m}_i \leftarrow \mathbf{m}_i - \alpha x^\delta \mathbf{m}_j$ , s.t. leading monomial at  $\text{LP}(\mathbf{m}_i)$  cancels.

Remember row operation by matrix  $\mathbf{U}$ .

$$\mathbf{M} = \begin{bmatrix} x^2 & \boxed{x^2 + x} \\ 1 & \boxed{x} \end{bmatrix} \xrightarrow{\mathbf{m}_1 \leftarrow \mathbf{m}_1 - x\mathbf{m}_2} \begin{bmatrix} x^2 - x & x \\ 1 & \boxed{x} \end{bmatrix}$$

## Simple Transformation

Find two rows  $\mathbf{m}_i$  and  $\mathbf{m}_j$  with

$$\text{LP}(\mathbf{m}_i) = \text{LP}(\mathbf{m}_j) \text{ and } \deg \mathbf{m}_i \geq \deg \mathbf{m}_j.$$

Replace  $\mathbf{m}_i \leftarrow \mathbf{m}_i - \alpha x^\delta \mathbf{m}_j$ , s.t. leading monomial at  $\text{LP}(\mathbf{m}_i)$  cancels.

Remember row operation by matrix  $\mathbf{U}$ .

$$\mathbf{M} = \begin{bmatrix} x^2 & \boxed{x^2 + x} \\ 1 & \boxed{x} \end{bmatrix} \xrightarrow{\mathbf{m}_1 \leftarrow \mathbf{m}_1 - x\mathbf{m}_2} \begin{bmatrix} x^2 - x & x \\ 1 & \boxed{x} \end{bmatrix} = \begin{bmatrix} 1 & -x \\ 0 & 1 \end{bmatrix} \cdot \mathbf{M}$$

## Simple Transformation

Find two rows  $\mathbf{m}_i$  and  $\mathbf{m}_j$  with

$$\text{LP}(\mathbf{m}_i) = \text{LP}(\mathbf{m}_j) \text{ and } \deg \mathbf{m}_i \geq \deg \mathbf{m}_j.$$

Replace  $\mathbf{m}_i \leftarrow \mathbf{m}_i - \alpha x^\delta \mathbf{m}_j$ , s.t. leading monomial at  $\text{LP}(\mathbf{m}_i)$  cancels.

Remember row operation by matrix  $\mathbf{U}$ .

$$\mathbf{M} = \begin{bmatrix} x^2 & \boxed{x^2 + x} \\ 1 & \boxed{x} \end{bmatrix} \xrightarrow{\mathbf{m}_1 \leftarrow \mathbf{m}_1 - x\mathbf{m}_2} \begin{bmatrix} x^2 - x & x \\ 1 & \boxed{x} \end{bmatrix} = \begin{bmatrix} 1 & -x \\ 0 & 1 \end{bmatrix} \cdot \mathbf{M}$$

$$\mathbf{U} = R(\mathbf{M})$$

Apply simple transformations until  $\mathbf{U} \cdot \mathbf{M}$  in wPf or  $\deg(\mathbf{U} \cdot \mathbf{M}) \leq \deg \mathbf{M} - 1$ .

## Simple Transformation

Find two rows  $\mathbf{m}_i$  and  $\mathbf{m}_j$  with

$$\text{LP}(\mathbf{m}_i) = \text{LP}(\mathbf{m}_j) \text{ and } \deg \mathbf{m}_i \geq \deg \mathbf{m}_j.$$

Replace  $\mathbf{m}_i \leftarrow \mathbf{m}_i - \alpha x^\delta \mathbf{m}_j$ , s.t. leading monomial at  $\text{LP}(\mathbf{m}_i)$  cancels.

Remember row operation by matrix  $\mathbf{U}$ .

$$\mathbf{M} = \begin{bmatrix} x^2 & \boxed{x^2 + x} \\ 1 & \boxed{x} \end{bmatrix} \xrightarrow{\mathbf{m}_1 \leftarrow \mathbf{m}_1 - x\mathbf{m}_2} \begin{bmatrix} x^2 - x & x \\ 1 & \boxed{x} \end{bmatrix} = \begin{bmatrix} 1 & -x \\ 0 & 1 \end{bmatrix} \cdot \mathbf{M}$$

$\mathbf{U} = R(\mathbf{M})$

Apply simple transformations until  $\mathbf{U} \cdot \mathbf{M}$  in wPf or  $\deg(\mathbf{U} \cdot \mathbf{M}) \leq \deg \mathbf{M} - 1$ .

$\mathbf{U} = R(\mathbf{M}, t)$

Apply  $R(\mathbf{M})$   $t$  times.  $\Rightarrow \deg(\mathbf{U} \cdot \mathbf{M}) \leq \deg \mathbf{M} - t$

## Simple Transformation

Find two rows  $\mathbf{m}_i$  and  $\mathbf{m}_j$  with

$$\text{LP}(\mathbf{m}_i) = \text{LP}(\mathbf{m}_j) \text{ and } \deg \mathbf{m}_i \geq \deg \mathbf{m}_j.$$

Replace  $\mathbf{m}_i \leftarrow \mathbf{m}_i - \alpha x^\delta \mathbf{m}_j$ , s.t. leading monomial at  $\text{LP}(\mathbf{m}_i)$  cancels.

Remember row operation by matrix  $\mathbf{U}$ .

$$\mathbf{M} = \begin{bmatrix} x^2 & \boxed{x^2 + x} \\ 1 & \boxed{x} \end{bmatrix} \xrightarrow{\mathbf{m}_1 \leftarrow \mathbf{m}_1 - x\mathbf{m}_2} \begin{bmatrix} x^2 - x & x \\ 1 & \boxed{x} \end{bmatrix} = \begin{bmatrix} 1 & -x \\ 0 & 1 \end{bmatrix} \cdot \mathbf{M}$$

$\mathbf{U} = R(\mathbf{M})$

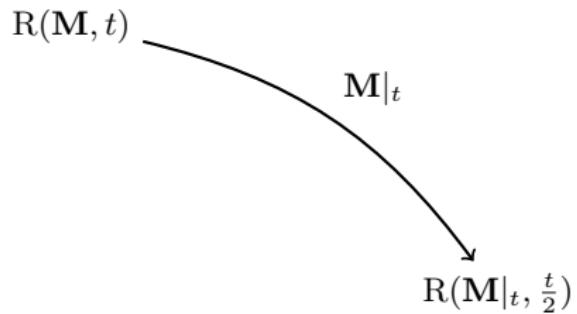
Apply simple transformations until  $\mathbf{U} \cdot \mathbf{M}$  in wPf or  $\deg(\mathbf{U} \cdot \mathbf{M}) \leq \deg \mathbf{M} - 1$ .

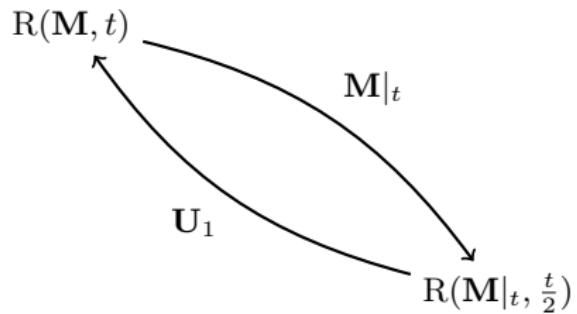
$\mathbf{U} = R(\mathbf{M}, t)$

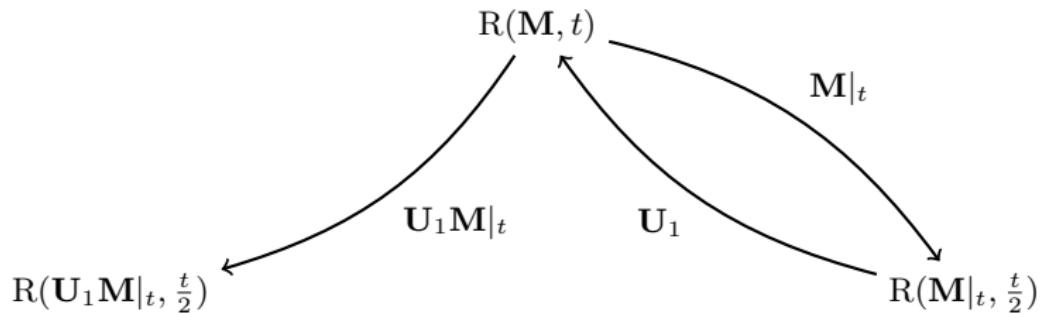
Apply  $R(\mathbf{M})$   $t$  times.  $\Rightarrow \deg(\mathbf{U} \cdot \mathbf{M}) \leq \deg \mathbf{M} - t$

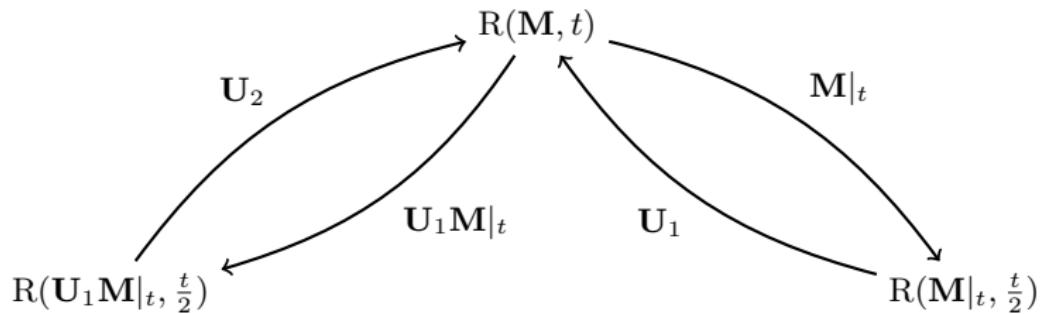
Transform  $\mathbf{B}$  (from shift register) in wPf: **Only  $t \in O(n)$  necessary!**

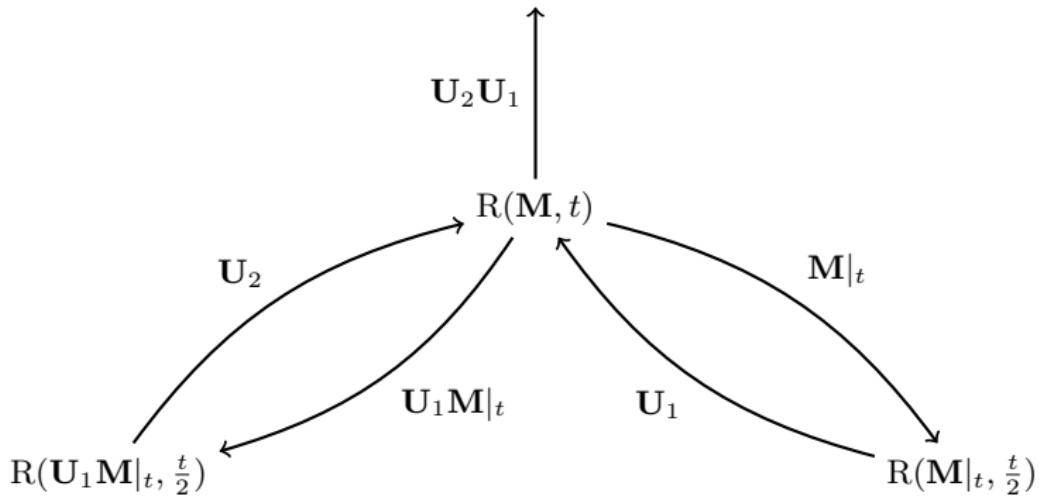
$$\mathbf{R}(\mathbf{M}, t)$$

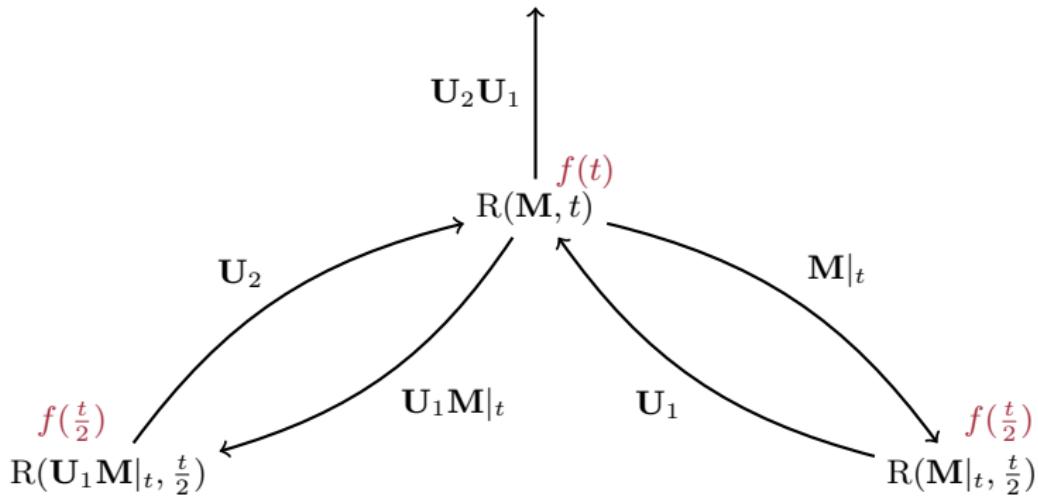




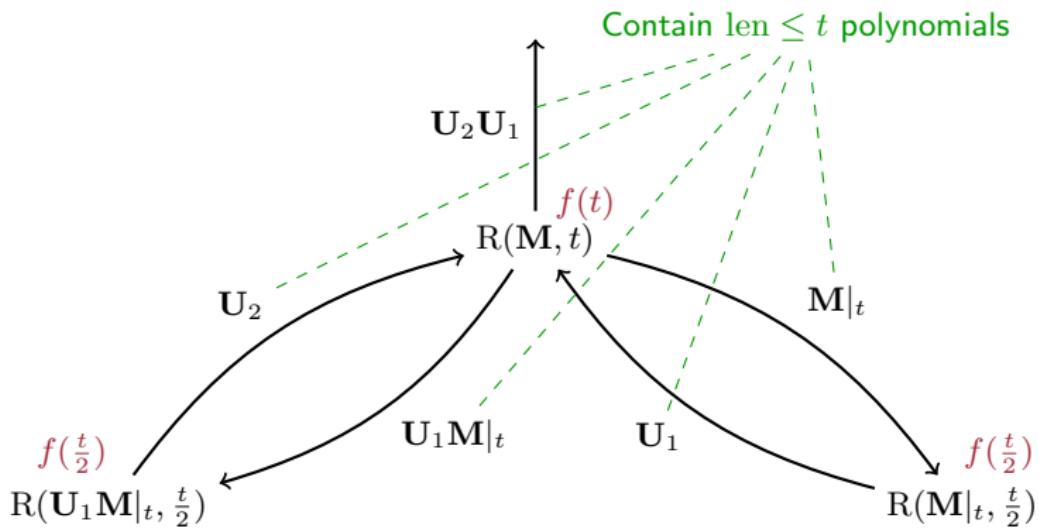


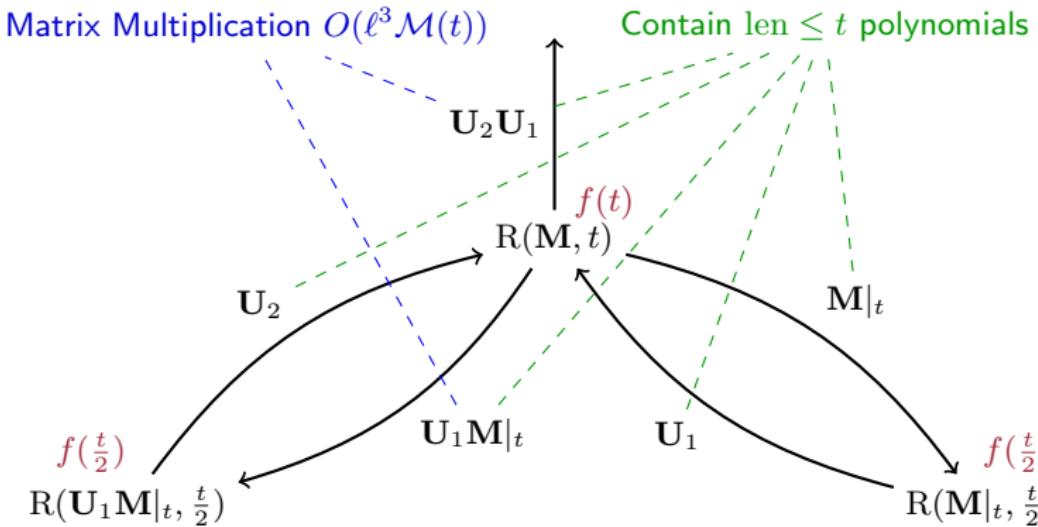






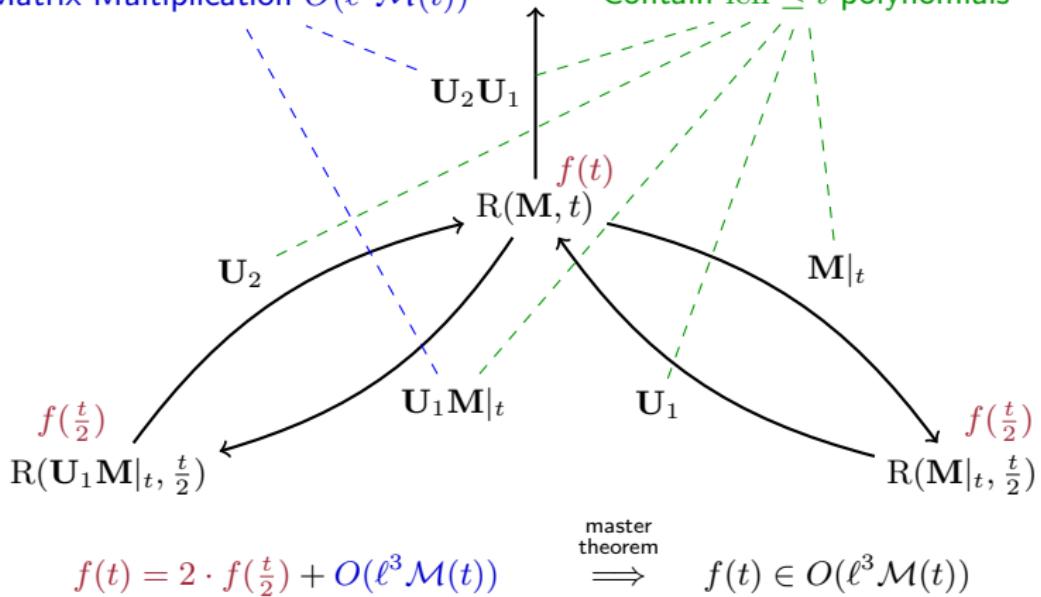
$$f(t) = 2 \cdot f\left(\frac{t}{2}\right) +$$



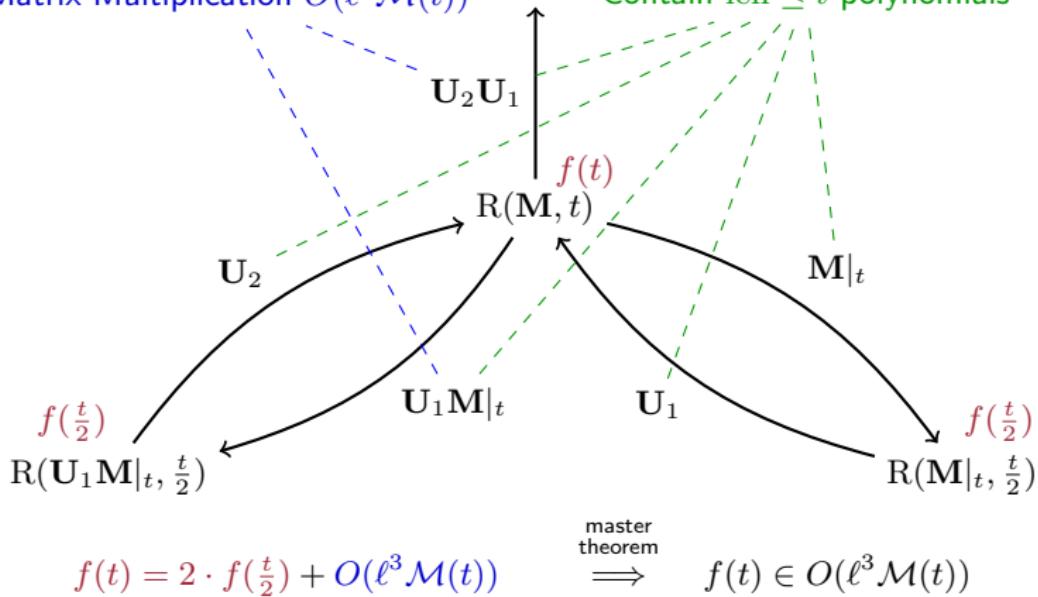


$$f(t) = 2 \cdot f\left(\frac{t}{2}\right) + O(\ell^3 \mathcal{M}(t))$$

Matrix Multiplication  $O(\ell^3 \mathcal{M}(t))$



Matrix Multiplication  $O(\ell^3 \mathcal{M}(t))$



Overall Complexity ( $\ell$  interleaving degree,  $n$  codelength)

$$t \in O(n) \implies O(\ell^3 \mathcal{M}(n))$$

## 1 Motivation

## 2 Skew Variant of Alekhnovich's Algorithm

## 3 Conclusion

Decoding Algorithm	Complexity
Skew Berlekamp–Massey [SJB11]	$O(\ell n^2)$
Skew Berlekamp–Massey (D&C) [SB14]	$O(\ell^3 \mathcal{M}(n) \log(n))$
Skew Demand-Driven [LNPS15]	$O(\ell n^2)$
Skew Alekhnovich [this paper]	$O(\ell^3 \mathcal{M}(n)) \subseteq O(\ell^3 n^{1.69})^*$

\* If  $\ell^2 \in o(\log(n))$ , additional  $\log(n)$  ( $\ell$  divisions of  $O(n^{1.69} \log(n))$ ) [PW16]

[SJB11] Sidorenko, Jiang, Bossert, "Skew-Feedback Shift-Register Synthesis and Decoding Interleaved Gabidulin Codes," *IEEE Trans. Inf. Theory*, 2011.

[SB14] Sidorenko, Bossert, "Fast Skew-Feedback Shift-Register Synthesis," *Designs, Codes & Cryptography*, 2014.

[LNPS15] Li, Nielsen, Puchinger, and Sidorenko, "Solving Shift Register Problems over Skew Polynomial Rings using Module Minimisation," *WCC*, 2015.

[PW16] Puchinger, Wachter-Zeh, "Sub-Quadratic Decoding of Gabidulin Codes," *ISIT*, 2016.

Decoding Algorithm	Complexity
Skew Berlekamp–Massey [SJB11]	$O(\ell n^2)$
Skew Berlekamp–Massey (D&C) [SB14]	$O(\ell^3 \mathcal{M}(n) \log(n))$
Skew Demand-Driven [LNPS15]	$O(\ell n^2)$
Skew Alekhnovich [this paper]	$O(\ell^3 \mathcal{M}(n)) \subseteq O(\ell^3 n^{1.69})^*$

Usually  $\ell \ll n$

\* If  $\ell^2 \in o(\log(n))$ , additional  $\log(n)$  ( $\ell$  divisions of  $O(n^{1.69} \log(n))$ ) [PW16]

[SJB11] Sidorenko, Jiang, Bossert, "Skew-Feedback Shift-Register Synthesis and Decoding Interleaved Gabidulin Codes," *IEEE Trans. Inf. Theory*, 2011.

[SB14] Sidorenko, Bossert, "Fast Skew-Feedback Shift-Register Synthesis," *Designs, Codes & Cryptography*, 2014.

[LNPS15] Li, Nielsen, Puchinger, and Sidorenko, "Solving Shift Register Problems over Skew Polynomial Rings using Module Minimisation," *WCC*, 2015.

[PW16] Puchinger, Wachter-Zeh, "Sub-Quadratic Decoding of Gabidulin Codes," *ISIT*, 2016.

Decoding Algorithm	Complexity
Skew Berlekamp–Massey [SJB11]	$O(\ell n^2)$
Skew Berlekamp–Massey (D&C) [SB14]	$O(\ell^3 \mathcal{M}(n) \log(n))$
Skew Demand-Driven [LNPS15]	$O(\ell n^2)$
Skew Alekhnovich [this paper]	$O(\ell^3 \mathcal{M}(n)) \subseteq O(\ell^3 n^{1.69})^*$

Usually  $\ell \ll n \implies$  BM D&C and Alekhnovich are fastest

\* If  $\ell^2 \in o(\log(n))$ , additional  $\log(n)$  ( $\ell$  divisions of  $O(n^{1.69} \log(n))$ ) [PW16]

[SJB11] Sidorenko, Jiang, Bossert, "Skew-Feedback Shift-Register Synthesis and Decoding Interleaved Gabidulin Codes," *IEEE Trans. Inf. Theory*, 2011.

[SB14] Sidorenko, Bossert, "Fast Skew-Feedback Shift-Register Synthesis," *Designs, Codes & Cryptography*, 2014.

[LNPS15] Li, Nielsen, Puchinger, and Sidorenko, "Solving Shift Register Problems over Skew Polynomial Rings using Module Minimisation," *WCC*, 2015.

[PW16] Puchinger, Wachter-Zeh, "Sub-Quadratic Decoding of Gabidulin Codes," *ISIT*, 2016.