

Sparse $\sqrt{\text{FGLM}}$ using the block Wiedemann algorithm

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Main Problem

Input: $I \subset \mathbb{K}[x_1, \dots, x_n]$ zero-dimensional

- monomial basis of $Q = \mathbb{K}[x_1, \dots, x_n]/I$
- multiplication matrices $T_1, \dots, T_n \in \mathbb{K}^{D \times D}$ of x_1, \dots, x_n , with $D = \dim_{\mathbb{K}}(Q)$

Output:

- lex Gröbner basis of \sqrt{I}

Assumptions

- characteristic of \mathbb{K} larger than D
- x_n generic coordinate

Consequence: output is

$$\begin{aligned} x_1 - R_1(x_n), \\ \vdots \\ x_{n-1} - R_{n-1}(x_n), \\ R(x_n) \end{aligned}$$

Previous work

[Faugère *et al.*'93] **FGLM**

- dense matrix computations

[Rouillier'99] **RUR**

- linearly generated sequence using the trace

[Bostan *et al.*'03]

- trace \rightarrow random linear form

[Faugère, Mou'17] **Sparse FGLM**

- lex basis of I
- uses Berlekamp-Massey-Sakata in some cases

At a glance

Cons:

- computes a lex basis of \sqrt{I} (weaker output)

Pros:

- few assumptions
- simple algorithm

Key idea: use sequences of small matrices, requires less terms than scalar sequences [Coppersmith '93].

Correctness from the analysis of Coppersmith's algorithm [Villard '97], [Kaltofen, Villard '04] and generating series properties [Bostan *et al.* '03]. See also [Kaltofen '95], [Kaltofen, Yuhasz '06].

Algorithm

choose $U, V \in \mathbb{K}^{m \times D}$

$s = (UT_n^i V^t)_{0 \leq i < 2d}$, with $d = \frac{D}{m}$

$S = \text{MatrixBerlekampMassey}(s)$ and $N = S \sum_{i \geq 0} \frac{s_i}{x^{i+1}}$

$P = \text{largest invariant factor of } S$ and $R_n = \text{SquareFreePart}(P)$

$a = [0 \ \dots \ 0 P] S^{-1}$

$N^* = \text{first entry of } aN$

for $j = 1 \dots n - 1$:

$s_j = (UT_n^i T_j V^t)_{0 \leq i < d}$ and $N_j = S \sum_{i \geq 0} \frac{s_{ji}}{x^{i+1}}$

$N_j^* = \text{first entry of } aN_j$

$R_j = N_j^* / N^* \bmod R_n$

Example

Input: $I = \langle f_1^2, f_2^2, f_3 \rangle \subset \mathbb{F}_{9001}[x_1, x_2, x_3]$ of degree $D = 32$, with

$$f_1 = 4979x_1^2 + 6202x_1x_2 + \dots, \quad f_2 = 4682x_1^2 + 8290x_1x_2 + \dots, \quad f_3 = 4199x_1^2 + 2325x_1x_2 + \dots$$

Step 1 with $m = 2$

$$U = \begin{bmatrix} 1898 & 6830 & 3494 & 169 & 7991 & 3352 & \dots \\ 3161 & 8858 & 8467 & 5882 & 8037 & 3726 & \dots \end{bmatrix} \quad V = \begin{bmatrix} 7595 & 8416 & 2285 & 8351 & 550 & 7012 & \dots \\ 823 & 5686 & 6539 & 7884 & 7105 & 3427 & \dots \end{bmatrix}^t$$

Step 2 & 3 with $d = 16$

$$s = \left(\begin{bmatrix} 31 & 6977 \\ 1178 & 1695 \end{bmatrix}, \begin{bmatrix} 8212 & 1663 \\ 4811 & 4837 \end{bmatrix}, \dots \right) \xrightarrow{\text{MatrixBerlekampMassey}} \begin{aligned} S &= \begin{bmatrix} \mathbf{x}^{16} + \dots & 423\mathbf{x}^{15} + \dots \\ 6426\mathbf{x}^{15} + \dots & \mathbf{x}^{16} + \dots \end{bmatrix} \\ N &= \begin{bmatrix} 6191\mathbf{x}^{15} + \dots & 8101\mathbf{x}^{15} + \dots \\ 7116\mathbf{x}^{15} + \dots & 2129\mathbf{x}^{15} + \dots \end{bmatrix} \end{aligned}$$

Step 4: $a = [2575\mathbf{x}^7 + \dots \quad \mathbf{x}^8 + \dots]$

Step 5: $[N^*] = [2575\mathbf{x}^7 + \dots \quad \mathbf{x}^8 + \dots] \begin{bmatrix} 6191\mathbf{x}^{15} + \dots \\ 7116\mathbf{x}^{15} + \dots \end{bmatrix} = [1178\mathbf{x}^{23} + 8727\mathbf{x}^{22} + \dots]$

Step 6 for $j = 1$

$$s_1 = \left(\begin{bmatrix} 3029 & 8903 \\ 1538 & 5610 \end{bmatrix}, \begin{bmatrix} 1914 & 3734 \\ 5221 & 5431 \end{bmatrix}, \dots \right) \implies N_1 = \begin{bmatrix} 1374\mathbf{x}^{15} + \dots & 3271\mathbf{x}^{15} + \dots \\ 4027\mathbf{x}^{15} + \dots & 1575\mathbf{x}^{15} + \dots \end{bmatrix}$$

Step 7: $[N_1^*] = [2575\mathbf{x}^7 + \dots \quad \mathbf{x}^8 + \dots] \begin{bmatrix} 1374\mathbf{x}^{15} + \dots \\ 4027\mathbf{x}^{15} + \dots \end{bmatrix} = [1538\mathbf{x}^{23} + 6498\mathbf{x}^{22} + \dots]$

Parallel Computations

- Bottleneck is computing the sequence $(UT_n^i)_{0 \leq i < 2d}$
- Can parallelize by computing the sequences $(U_1T_n^i), \dots, (U_mT_n^i)$ separately, where U_i is the i^{th} row of U
- When $m = 1$, same computation as Sparse FGLM

Conclusion

References

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