

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'13].

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 $N = S \sum_{i \geq 0} \frac{s_i}{x_i^{i+1}}$
- $P =$ largest invariant factor of S and $R_n = \text{SquareFreePart}(P)$
- $a = [0 \ \dots \ 0P]S^{-1}$
- $N^* =$ first entry of aN
- for $j = 1 \dots n - 1$:
 - $s_j = (UT_n^i T_j V^t)_{0 \leq i < d}$ and $N_j = S \sum_{i \geq 0} \frac{s_{ji}}{x_i^{i+1}}$
 - $N_j^* =$ first entry of aN_j
 - $R_j = N_j^*/N^* \bmod R_n$

Example

Input: $I = \langle f_1^2, f_2^2, f_3 \rangle \subset \mathbb{F}_{9001}[x_1, x_2, x_3]$, non radical 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 & 5: $a = [2575\mathbf{x}^7 + \dots \quad \mathbf{x}^8 + \dots]$ and $R_3 = \mathbf{x}^8 + 6990x^7 + \dots$

Step 6: $N^* = [2575\mathbf{x}^7 + \dots \quad \mathbf{x}^8 + \dots] \begin{bmatrix} 6191\mathbf{x}^{15} + \dots \\ 7116\mathbf{x}^{15} + \dots \end{bmatrix} = [1178\mathbf{x}^{23} + 8727\mathbf{x}^{22} + \dots]$

Step 7 for $j = 1$

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

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

$$R_1 = \frac{1538\mathbf{x}^{23} + 6498\mathbf{x}^{22} + \dots}{1178\mathbf{x}^{23} + 8727\mathbf{x}^{22} + \dots} \bmod \mathbf{x}^8 + 6990x^7 + \dots = 7964\mathbf{x}^7 + 4071x^6 + \dots$$

Parallelization

Bottleneck: computing $(UT_n^i)_{0 \leq i < 2d}$

Bottleneck: easy to parallelize!

experiments in **LinBox** modulo $p = 65537$

name	n	D	ratio (8 cores)	sparsity
katsura8	9	256	1136/213= 5.33	0.63
katsura9	10	512	8903/1651= 5.39	0.64
cyclic7	7	924	6585/1235= 5.33	0.08
katsura10	11	1024	70211/13019= 5.39	0.63
gametwo7	7	1854	35051/64673= 5.41	0.54
schwarz11	11	2048	21699/4129= 5.25	0.27
eco12	12	1024	61605/11296= 5.45	0.55

(parallel speed-up for the sequence computation)

References

- [1] A. Bostan, B. Salvy, and É. Schost. Fast algorithms for zero-dimensional polynomial systems using duality. *Applicable Algebra in Engineering, Communication and Computing*, 14:239–272, 2003.
- [2] D. Coppersmith. Solving linear equations over GF(2): block Lanczos algorithm. *Linear Algebra and its Applications*, 192:33–60, 1993.
- [3] J.-C. Faugère, P. Gianni, D. Lazard, and T. Mora. Efficient computation of zero-dimensional Gröbner bases by change of ordering. *Journal of Symbolic Computation*, 16(4):329–344, 1993.
- [4] J.-C. Faugère and C. Mou. Sparse FGLM algorithms. *Journal of Symbolic Computation*, 80(3):538–569, 2017.
- [5] E. Kaltofen and G. Villard. On the complexity of computing determinants. *Comput. Complexity*, 13(3-4):91–130, 2004.
- [6] Erich Kaltofen and George Yuhasz. On the matrix Berlekamp-Massey algorithm. *ACM Transactions on Algorithms*, 9(4):33:1–33:24, 2013.
- [7] F. Rouillier. Solving zero-dimensional systems through the rational univariate representation. *Applicable Algebra in Engineering, Communication and Computing*, 9(5):433–461, 1999.
- [8] G. Villard. Further analysis of Coppersmith's block Wiedemann algorithm for the solution of sparse linear systems (extended abstract). In *ISSAC'97*, pages 32–39. ACM, 1997.