

Signature-based Gröbner basis computations for approximate input

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Signature-based Gröbner basis computation algorithms: application to solving polynomial systems with inexact input data in \mathbb{C} .

1 Definitions

2 Problems of application

3 Known methods

4 Inexact zero

- Approximate arithmetic instead of floats
- Non-invertible approximate elements classification
- Examples

5 New ideas

- Strict modular method for approximate Gröbner basis
- Strict modular method for approximate linear system
- No-restart TSV method for signature-based algorithms
- “Strict modular” and “no-restart TSV” methods: system solving

- Compute Gröbner basis G for ideal generated by given polynomials $I = (f_1, \dots, f_n)$ in polynomial ring P
- Are one of the most effective algorithms for Gröbner basis computation
- Track *signature* of every polynomial p , corresponding to “highest” element of input-representation

$$p = \sum_{i=1}^n q_i f_i$$

- Using signature-based rules maintain an ordered queue of polynomials to process

- Arises from different areas (statistics, video analytics, etc.)
- Formulated as a set of polynomial equations $f_i = 0$ with *approximate number* coefficients

Definitions

Approximate (complex) number is a pair (c, ε) , $c \in \mathbb{C}$, $\varepsilon \in \mathbb{R}$

Specialization of complex number a is $\hat{a} \in \mathbb{C}$, $|\hat{a} - c| < \varepsilon$

Requirements for approximate Gröbner basis used to solve polynomial system: a set G of polynomials with approximate coefficients satisfying following:

The approximate Gröbner basis $GB(f_i)$ should have specialization $\widehat{GB(f_i)}$ equal to exact Gröbner basis $GB(\hat{f}_i)$ for any specialization \hat{f}_i of input data f_i .

- Algorithm branching points: is $a = 0$ in $f = ax^2 + by^2$?
 - formal determine $\text{HM}(f)$: is it x^2 or y^2 ?
 - practical can we reduce polynomial $x^3 + xy$ by f ?
- Branching points in signature-based algorithms are unavoidable:
 - need test coefficients of non-reducible highest monomials (HC)
 - specific polynomial selection is required for correctness
- Approximated numbers having zero specialization
 - may be input data coefficients
 - may appear as a result of inexact computations
 - don't allow comparison with zero with boolean result

- Comprehensive Gröbner basis computation
 - treat coefficients as symbols and use symbolic computation
 - slow arithmetic with huge symbolic expressions
- Exact Gröbner basis computation in \mathbb{Q}
 - selects some \mathbb{Q} -specialization of approximate number
 - slow arithmetic with huge denominators
 - solves task only for single specialization
- Approximate Gröbner basis computation using numeric or modular methods for zero-comparison
 - result correctness is not guaranteed even for single specialization
- Approximate Gröbner basis computation changing monomial order to skip HC zero-comparison (TSV, etc.)
 - requires restart from scratch of signature-based algorithms

Addition $(c_1, \varepsilon_1) + (c_2, \varepsilon_2) = (c_1 + c_2, \varepsilon_1 + \varepsilon_2)$

Multiplication $(c_1, \varepsilon_1) \times (c_2, \varepsilon_2) = (c_1 c_2, \varepsilon_1 |c_2| + \varepsilon_2 |c_1| + \varepsilon_1 \varepsilon_2)$

Subtraction based on addition and multiplication by -1:

$$(c_1, \varepsilon_1) - (c_2, \varepsilon_2) = (c_1 - c_2, \varepsilon_1 + \varepsilon_2)$$

Inversion defined only for numbers not having zero specialization, or equivalently satisfying $|c| - \varepsilon > 0$:

$$\frac{1}{(c, \varepsilon)} = \left(\frac{1}{c}, \frac{\varepsilon}{|c| (|c| - \varepsilon)} \right)$$

- Track all possible specializations
- Slower than floating point arithmetic only by a constant factor

Definitions

- *symbolic zero* – corresponding computations in \mathbb{C} for any input data specialization give exact zero.
- *input-inspired zero* – \mathbb{C} -computations for some but not all input data specializations give exact zero
- *computation-introduced zero* – \mathbb{C} -computations for input data specializations never give exact zero

Example

$$\begin{aligned}f_1 &= y^2z + a \\f_2 &= y^2z^2 + xz + 1 \\f_3 &= y^3z + xy + 1 \\f_4 &= f_2 - zf_1 = xz - az + 1 \\f_5 &= f_3 - yf_1 = xy - ay + 1 \\f_6 &= zf_5 - yf_4 = (a - a)yz + z - y\end{aligned}$$

$HC(f_6)$ is symbolic zero.

Example

$$\begin{aligned}f_1 &= y^2z + z^2 + az \\f_2 &= xyz \\f_3 &= xy^2 + bx + 1 \\f_4 &= -(yf_2 - xf_1) = xz^2 + axz \\f_5 &= (zf_3 - xf_1) + f_4 = ((b - a) + a)xz + z\end{aligned}$$

$HC(f_5)$ may be input-inspired or computation-introduced zero.
Concrete zero type depends on a and b values.

Main idea: compute in finite field and treat all zeros as symbolic.

- Initially selected prime module
 - known methods: prime is assumed to be “big enough”
 - strict method: initially selected prime used only to compute a hypothesis of the linear system form; rechecked later
- Result approximate coefficients come from linear system solution
 - avoid computation-introduced zeros arose from $a - a$ expressions
 - detect all symbolic zeros using modular computation (like known method)
 - allow to estimate correctness probability based on system properties and prime number
 - separately select prime module for every system to ensure required correctness probability

Zero-testings are needed during Gaussian elimination. Modular computations allows to detect all symbolic zeros but may detect superfluous ones leading to error. To solve a system with error probability $\leq 2\alpha$ we should randomly take a prime number from a set of at least N_p prime numbers greater than P_0 .

- $P_0 = \sqrt{\frac{R2^RV}{\alpha}}, N_p = \frac{R2^R \log_{P_0} 2Z_0}{\alpha}$

R number of rows in a system

V number of approximate coefficients in a system

Z_0 maximal denominator of exact \mathbb{Q} coefficient

Fact

*Required number of bits is linear in the number of system rows.
Method fails on input-inspired zeros.*

TSV main idea: add monomial y and input polynomial $x_1^{j_1} \cdots x_m^{j_m} - y$ to avoid zero testing of $x_1^{j_1} \cdots x_m^{j_m}$ coefficient.
Result is GB of extended ideal $I^e = (I, x_1^{j_1} \cdots x_m^{j_m} - y, \dots)$

- new polynomial
 - original TSV method: added with small signature and restart algorithm from scratch
 - no-restart TSV method: added with “just-before-current” signature and continue processing
- real-weighted order
 - real and monomial parameter for every input polynomial
 - added polynomial signatures can be configured by parameters
 - allows addition of polynomial with signature just before given
 - applies only for signature order independent algorithms

- ① Compute a Gröbner basis of extended ideal I^e with a signature-based algorithm using real-weighted order
 - ① Compute approximate result by strict modular method
 - ② If it fails on non-invertible element, proceed by no-restart TSV
 - ② Using GB (I^e) and without using GB (I) determine one of:
 - ① P/I is vector space; find it's basis B , normal form operator ϕ
 - ② I is not zero-dimensional; then the algorithm is inapplicable
 - ③ Solve the system by “action matrix method” reformulated in terms of B and ϕ instead of GB (I).
- Problems of known methods are avoided
 - arithmetic operations has predictable complexion
 - no restarts from scratch
 - all input data specializations are considered
 - the probability of error strictly estimated
 - Open questions in complexity estimation are introduced
 - huge prime modules may appear
 - influence of real-weighted order on speed is unknown