

The multivariate resultant is NP-hard in any characteristic

Bruno Grenet, Pascal Koiran and Natacha Portier



MC2 – LIP, ÉNS Lyon
Theory Group – DCS, Univ. of Toronto

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Motivation

- General framework: Resolution of polynomial systems

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- First (simple) results about polynomial system solving
- Two ideas to prove NP-hardness

Outline

- 1 Elimination Theory
- 2 First results
- 3 NP-hardness in any characteristic
 - First idea \rightsquigarrow randomized reduction
 - Second idea \rightsquigarrow deterministic reduction

A modern theory!

MODERN
ALGEBRA

By B. L. VAN DER WAERDEN, Ph.D.
PROFESSOR OF MATHEMATICS
AT THE UNIVERSITY OF AMSTERDAM

In part a development from lectures
By E. ARTIN and E. NOETHER

VOLUME II

Translated from the second revised German edition
By THEODORE J. BENAC, Ph.D.
ASSOCIATE PROFESSOR, DEPARTMENT OF MATHEMATICS
U.S. NAVAL ACADEMY

CHAPTER XI

ELIMINATION THEORY

In elimination theory we study systems of algebraic equations in several unknowns in order to set up conditions for their solvability as well as formulas for calculating their solutions in various cases. In this chapter the corresponding theory for linear equations, i.e., the theory of determinants, is assumed as known. Furthermore, it shall also be assumed that one equation in a single unknown of degree higher than one can be solved, or more precisely, if such an equation can not be resolved in a given field, an extension field may be constructed in which it is decomposable, and in fact one in which it may be completely decomposed (Chapter 5). In the following when we refer to the "solutions of an equation" or the "zeros of a polynomial," we shall always assume that the solutions are in a suitably chosen extension field of the fixed commutative field K .

77. THE RESULTANT SYSTEM OF SEVERAL POLYNOMIALS
IN A SINGLE VARIABLE

THEOREM. Let f_1, \dots, f_r be r polynomials in a single variable of given degrees with indeterminate coefficients. Then there exists a system D_1, \dots, D_r of integral polynomials in these coefficients with the property that if these coefficients are assigned values from the field K the conditions $D_1 = 0, \dots, D_r = 0$ are necessary and sufficient in order that either the equations $f_1 = 0, \dots, f_r = 0$ have a solution in a suitable extension field or that the formal leading coefficients of all polynomials f_1, \dots, f_r vanish.

The proof is based on Kronecker's method of elimination.

First, we transform the polynomials f_1, \dots, f_r into polynomials of the same degree by multiplying every polynomial f_i by $(x-1)^{n-n_i}$ and $(x-1)^{n-n_i}$ provided that n_i is smaller than n where n is the greatest (formal) degree of the given polynomials. In this way two polynomials of formal degree n arise from f_i such that for any specialization of the coefficients their common zeros and their leading coefficients are the same as those of f_i . This system of polynomials of the same degree may contain more polynomials than are in the system f_1, \dots, f_r , but it has the same common zeros. We designate these polynomials by g_1, \dots, g_s .

1

Most recent book: 1950

General form

- Some polynomials: $f_1, \dots, f_s \in \mathbb{K}[X_1, \dots, X_n]$,

$$f_i = \sum_{|\alpha|_1 \leq d_i} \gamma_{i,\alpha} \bar{X}^\alpha$$

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- Condition on the $\gamma_{i,\alpha}$ for the system to have a root?

There exist $R_1, \dots, R_h \in \mathbb{K}[\bar{\gamma}]$ s.t.

$$R_1(\bar{\gamma}) = \dots = R_h(\bar{\gamma}) = 0 \implies \exists \bar{X}, f_1(\bar{X}) = \dots = f_s(\bar{X}) = 0$$

Special cases

- If $s = 2$ & $n = 1$ then $h = 1$:

$$R(\bar{\gamma}) = 0 \implies \exists \bar{X}, f_1(\bar{X}) = f_2(\bar{X}) = 0$$

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Definitions

Hilbert's Nullstellensatz over \mathbb{K} : $\text{HN}(\mathbb{K})$

Input: $f_1, \dots, f_s \in \mathbb{K}[x_0, \dots, x_n]$

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- $\text{H}_2\text{N}^\square(\mathbb{K})$: $s = n + 1$ homogeneous polynomials

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For all \mathbb{K} , $H_2 N^\square(\mathbb{K}) \leq_m^p \text{HN}(\mathbb{K})$

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Remark. Best known upper bound for $\mathbb{K} = \mathbb{F}_p$ is PSPACE.

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- Boolean variables

X_1, \dots, X_n

- Equations

- ▶ $X_i = \text{True}$

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$$\rightsquigarrow \left\{ \begin{array}{rcl} x_1^2 - x_0^2 & = & 0 \\ & \vdots & \\ x_n^2 - x_0^2 & = & 0 \\ u_1 x_1 + \dots + u_n x_n & = & 0 \end{array} \right.$$

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- Over \mathbb{Z} , $HN \in NP \implies BPP = NP$

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- I is a negative instance $\implies \mathbb{P}[A(I) \text{ is a positive instance}] \leq 1/3$

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- Random α_{ij} : equivalence with high probability (quantifier elimination + Schwartz-Zippel Lemma)

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Thank you!