

The Multivariate Resultant is NP-hard in any Characteristic

Bruno Grenet, Pascal Koiran and Natacha Portier



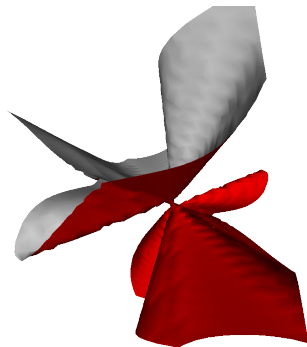
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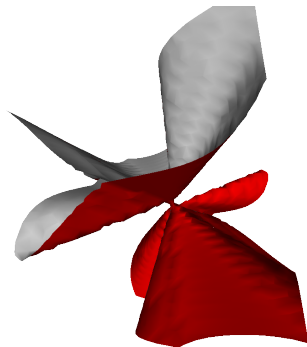
Motivation

- General framework: Resolution of polynomial systems



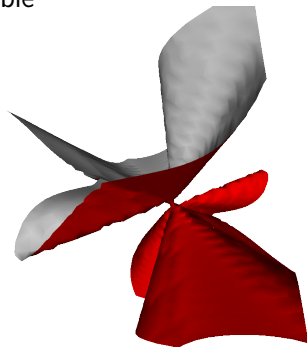
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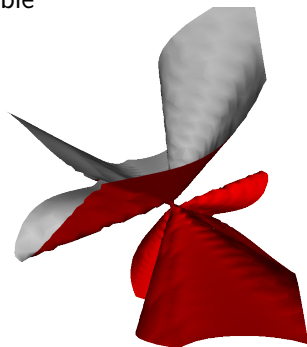
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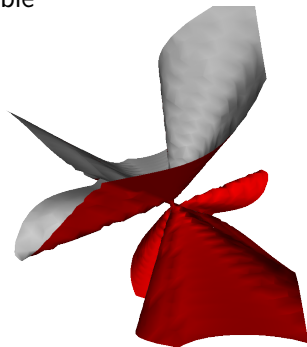
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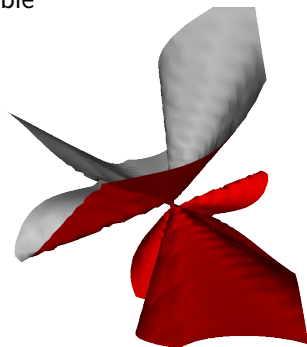
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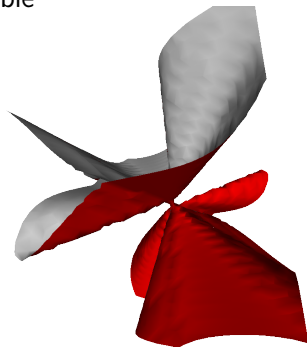
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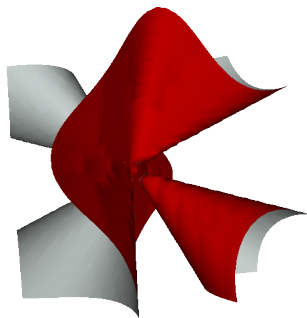
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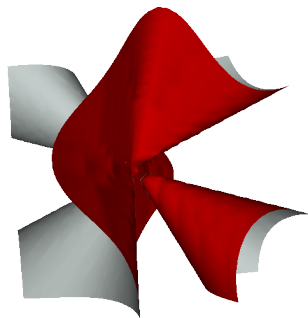
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- A few words about Elimination Theory



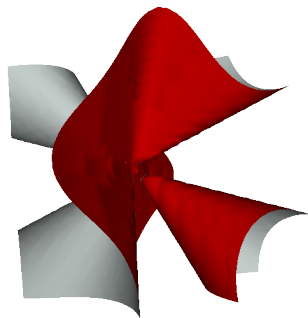
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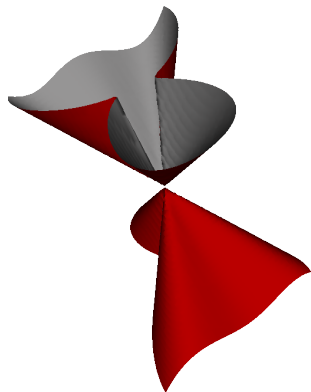
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- A few words about Elimination Theory
- First (simple) results about polynomial system solving
- Two ideas to prove NP-hardness



Outline

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



General form

$$f_1, \dots, f_s \in \mathbb{K}[x_1, \dots, x_n]$$

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For which $\gamma_{i,\alpha}$ is there a root?

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

$$\begin{cases} R_1(\bar{\gamma}) = 0 \\ \vdots \\ R_h(\bar{\gamma}) = 0 \end{cases} \implies \exists \bar{a}, \begin{cases} f_1(\bar{a}) = 0 \\ \vdots \\ f_s(\bar{a}) = 0 \end{cases}$$

Two Univariate Polynomials

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- Non trivial root?

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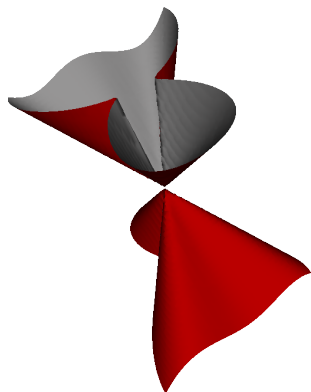
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- Resultant computable in **polynomial space**

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Definitions

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

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

Question: $\exists? \bar{a} \in \bar{\mathbb{K}}^{n+1}, \begin{cases} f_1(\bar{a}) = 0 \\ \vdots \\ f_s(\bar{a}) = 0 \end{cases}$

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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} ,

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

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- Equations

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□

Summary so far

Upper bounds

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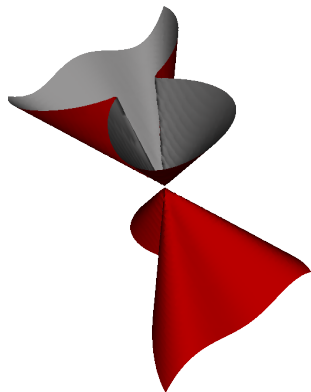
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- NP = AM “under plausible complexity conjectures”

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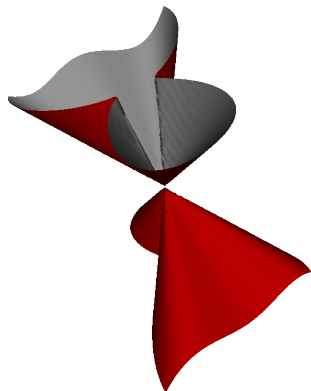
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 - ▶ Second idea: increase the number of variables

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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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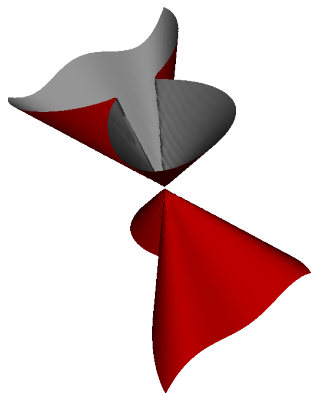
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if random α_{ij} : with high probability

(Quantifier Elimination + Schwartz-Zippel Lemma)

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- Variables over \mathbb{K} x_0 and x_1, \dots, x_n
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 - ▶ $x_0 \cdot (x_i + x_0)$
 - ▶ $x_0 \cdot (x_i + x_j)$
 - ▶ $(x_i + x_0)^2 - (x_j + x_0) \cdot (x_k + x_0)$

Introduction

- Randomized reduction: fewer polynomials
 \neq Deterministic reduction: **more variables**

H₂N

- Variables over \mathbb{K} x_0 and x_1, \dots, x_n
 - Polynomials $x_0^2 - x_i^2$ for every i $\rightarrow f_1, \dots, f_n$
- | | | |
|------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|---|-----------------------------------|
| <ul style="list-style-type: none"> ▶ $x_0 \cdot (x_i + x_0)$ ▶ $x_0 \cdot (x_i + x_j)$ ▶ $(x_i + x_0)^2 - (x_j + x_0) \cdot (x_k + x_0)$ | } | $\rightarrow f_{n+1}, \dots, f_s$ |
|------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|---|-----------------------------------|

Reduction

- New variables: y_1, \dots, y_{s-n-1}

New system

$$g(\bar{x}, \bar{y}) = \left(\begin{array}{c} \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \\ \end{array} \right)$$

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Reduction

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Equivalence?

(\bar{a}, \bar{b}) : a non trivial root

$$\begin{pmatrix} f_1(\bar{a}) \\ \vdots \\ f_n(\bar{a}) \\ f_{n+1}(\bar{a}) + \lambda b_1^2 \\ f_{n+2}(\bar{a}) - b_1^2 + \lambda b_2^2 \\ \vdots \\ f_{s-1}(\bar{a}) - b_{s-n-2}^2 + \lambda b_{s-n-1}^2 \\ f_s(\bar{a}) - b_{s-n-1}^2 \end{pmatrix}$$

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Last step

$$\det = \pm \left(\epsilon_1 + \epsilon_2 \lambda + \cdots + \epsilon_N \lambda^{N-1} \right)$$

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[Shoup 1990]
Polynomial
time algorithm

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$H_2N^\square(\mathbb{K})$ is NP-hard

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Thank you!