

Solving parameter-dependent semi-algebraic systems with Hermite matrices

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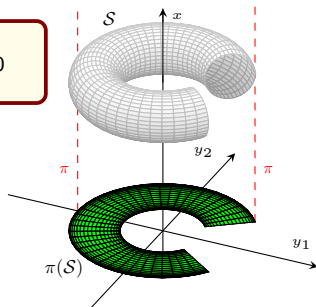
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Solving parametric polynomial systems with inequalities

$$f_1(\mathbf{y}, \mathbf{x}) = \dots = f_p(\mathbf{y}, \mathbf{x}) = 0, \quad g_1(\mathbf{y}, \mathbf{x}) > 0, \dots, g_s(\mathbf{y}, \mathbf{x}) > 0$$

- $\mathbf{y} = (y_1, \dots, y_t)$ are **parameters**
- $\mathbf{x} = (x_1, \dots, x_n)$ are **variables**
- $\pi: (\mathbf{y}, \mathbf{x}) \mapsto \mathbf{y}$ the \mathbf{y} -coordinate projection

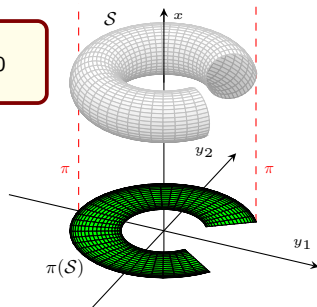


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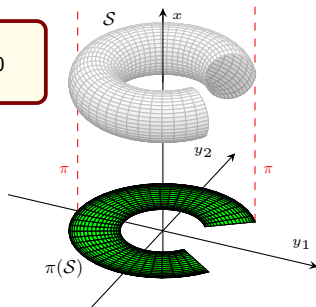


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Goals

- Classify the possible number of real roots
- Describe the regions where these numbers are achieved

→ **Applications** in Robotics, Computer Vision, Physics,...

Real Solution Classification

Given a **semi-algebraic** (s.a) set $\mathcal{S} \subseteq \mathbb{R}^{t+n}$ defined by

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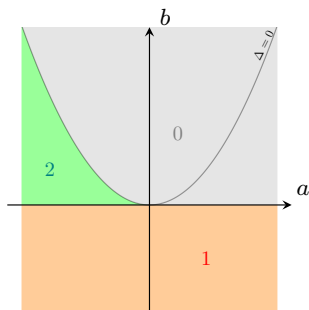
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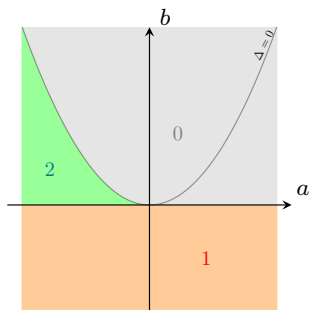
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Valid Classification:

r	η	Φ
0	(1, 1)	$\Delta < 0 \vee (a > 0 \wedge b > 0)$
1	(1, -1)	$b < 0$
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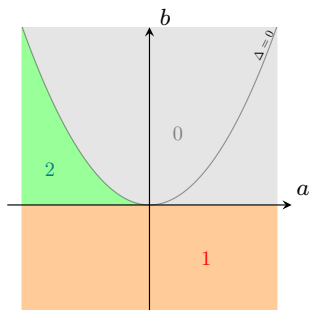
Problem

Compute $(\Phi_i, \eta_i, r_i)_{1 \leq i \leq \ell}$ with Φ_i a **s.a** formula in $\mathbb{Q}[y]$ defining the **s.a** set $\bar{\mathcal{T}}_i \subseteq \mathbb{R}^t$, $\eta_i \in \mathcal{T}_i$ and $r_i \geq 0$ st,

- for all $\eta \in \mathcal{T}_i$, $\sharp S \cap \pi^{-1}(\eta) = r_i$
- $\bigcup_{i=1}^{\ell} \mathcal{T}_i$ is dense in \mathbb{R}^t

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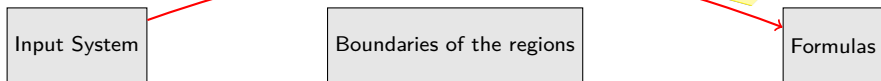
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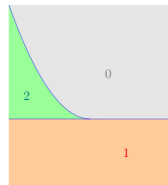
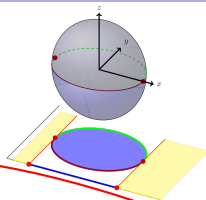
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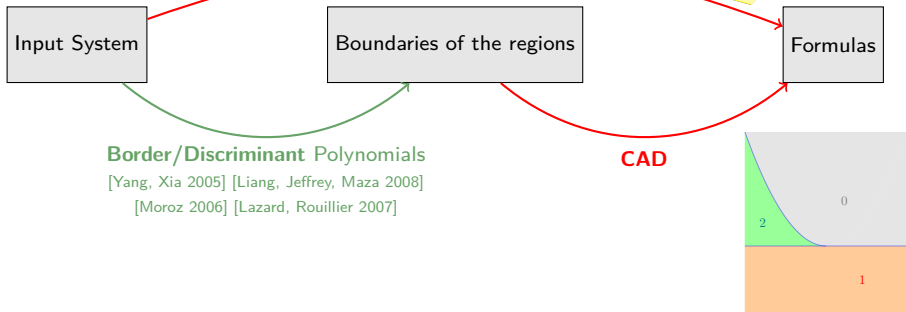
Border/Discriminant Polynomials

[Yang, Xia 2005] [Liang, Jeffrey, Maza 2008]

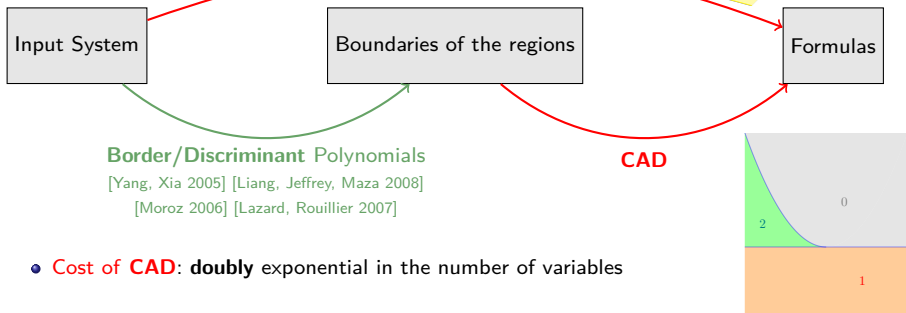
[Moroz 2006] [Lazard, Rouillier 2007]



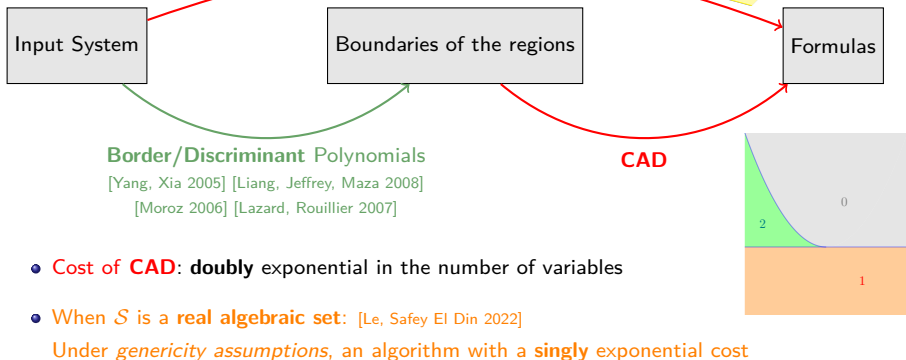
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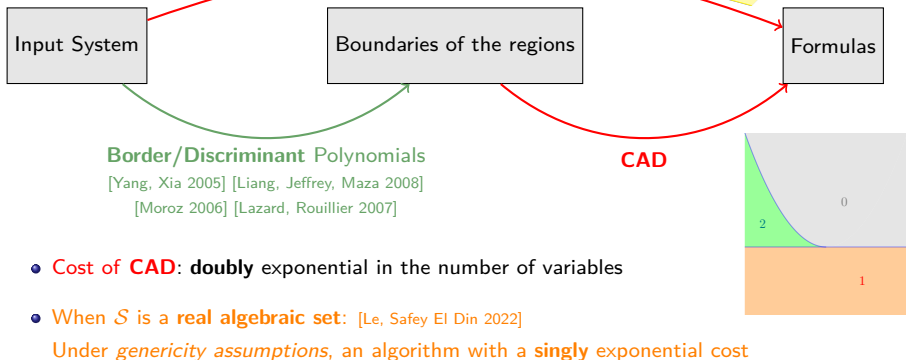
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Question: Can we achieve a **singly** exponential complexity for **semi-algebraic sets**?

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Theorem [Basu, Pollack, Roy 2005] : $\rho \leq s^t d^{O(n+t)}$

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- **Implementation** solving instances that were **out of reach**

Hermite's quadratic forms

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For $\mathbf{f} = (f_1, \dots, f_p) \in \mathbb{K}[\mathbf{x}]$ s.t. $\langle \mathbf{f} \rangle_{\mathbb{K}} \subseteq \mathbb{K}[\mathbf{x}]$ **zero-dim** and $g \in \mathbb{K}[\mathbf{x}]$

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Tarski-queries

When $\mathbb{K} = \mathbb{R}$ or \mathbb{Q}

$\text{TaQ}(g, \mathbf{f}) :=$

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Remark:

$$\begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & -1 \\ 0 & 1 & 1 \end{bmatrix} \cdot \begin{bmatrix} c(g=0) \\ c(g>0) \\ c(g<0) \end{bmatrix} = \begin{bmatrix} \text{TaQ}(1, \mathbf{f}) \\ \text{TaQ}(g, \mathbf{f}) \\ \text{TaQ}(g^2, \mathbf{f}) \end{bmatrix} \quad \text{where } c(g \diamond 0) := \#\{x \mid \mathbf{f}(x) = 0 \wedge g(x) \diamond 0\}$$

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Choice of basis (Le, Safey El Din 2022)

- G a Gröbner basis of $\langle \mathbf{f} \rangle \subseteq \mathbb{Q}[\mathbf{x}, \mathbf{y}]$ wrt the elimination ordering $\text{grevlex}(\mathbf{x}) \succ \text{grevlex}(\mathbf{y})$ (with $x_1 \succ \dots \succ x_n$ and $y_1 \succ \dots \succ y_t$)

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- G a Gröbner basis of $\langle f \rangle \subseteq \mathbb{Q}[\mathbf{x}, \mathbf{y}]$ wrt the elimination ordering $\text{grevlex}(\mathbf{x}) \succ \text{grevlex}(\mathbf{y})$ (with $x_1 \succ \dots \succ x_n$ and $y_1 \succ \dots \succ y_t$)
- G is also a Gröbner basis of $\langle f \rangle_{\mathbb{K}} \subseteq \mathbb{K}[\mathbf{x}]$ wrt $\text{grevlex}(\mathbf{x})$
- The set B of all monomials in \mathbf{x} that are not reducible by $\text{lm}(G)$ is a **basis** of $A_{\mathbb{K}}$

Classification for $s = 1$

$$f_1 = \cdots = f_p = 0, \quad g > 0$$

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Recall: For η in a Zariski dense subset of \mathbb{C}^t

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Key idea

- Over a **connected component** of the s-a set defined by the **non-vanishing locus** of the *leading principal minors* of $\mathcal{H}_1, \mathcal{H}_g, \mathcal{H}_{g^2}$, c_η is invariant
- Sample one point in every **connected component** using [Le, Safey El Din 2022]
- Deduce formulas for the **classification** from the **sign patterns** of these minors

Algorithm for $s = 1$

Algorithm 1: Real Solution Classification for 1 inequality

Algorithm for $s = 1$

Input : $f_1 = \dots = f_p = 0, \quad g > 0$ s.t. for generic $\eta \in \mathbb{C}^t$, $\mathbf{f}(\eta, \cdot) = 0$ is zero-dim

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- 5 **for** $\eta \in L$ **do**
 - 6 $T_\eta \leftarrow (\text{Sign}(\mathcal{H}_1(\eta)), \text{Sign}(\mathcal{H}_g(\eta)), \text{Sign}(\mathcal{H}_{g^2}(\eta)))^t$
 - 7 Solve $M \cdot c_\eta = T_\eta$ to compute $r_\eta := c(g(\eta, \cdot) > 0)$
 - 8 $\Phi_\eta \leftarrow$ **sign pattern of Minors evaluated in η**
- 9 **end**
- 10 **return** $(\Phi_\eta, \eta, r_\eta)_{\eta \in L}$

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Generalize the identity for $s = 1$ to a **tensor identity** [Ben-Or, Kozen, Reif 1984] [Basu, Pollack, Roy 2006]

$$M = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & -1 \\ 0 & 1 & 1 \end{bmatrix}$$

$$M \cdot \begin{bmatrix} c(0, \mathcal{Z}) \\ c(1, \mathcal{Z}) \\ c(-1, \mathcal{Z}) \end{bmatrix} = \begin{bmatrix} \text{TaQ}(1, \mathcal{Z}) \\ \text{TaQ}(Q_1, \mathcal{Z}) \\ \text{TaQ}(Q_1^2, \mathcal{Z}) \end{bmatrix}$$

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Problem

Need to compute 3^s Hermite matrices \rightarrow **exceed our target complexity** (polynomial in s)

How to avoid unrealizable sign conditions

$$d := \max(\deg f_i, \deg g_j)$$

$$f_1 = \dots = f_p = 0, \quad g_1 > 0, \dots, g_s > 0, \quad \mathbf{x} = (x_1, \dots, x_n), \quad \mathbf{y} = (y_1, \dots, y_t)$$

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The *number of sign conditions* realized by \mathbf{g} on the roots of \mathbf{f} is bounded by

$$\rho := \binom{s}{t} 4^{t+1} d(2d-1)^{n+t-1} = d^{O(n+t)} s^t.$$

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- Control the needed number of **Hermite matrices**.

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- **Formulas** given by sign patterns of minors of remaining Hermite matrices

Practical Results

$$f_1(\mathbf{y}, \mathbf{x}) = \cdots = f_n(\mathbf{y}, \mathbf{x}) = 0, \quad g_1(\mathbf{y}, \mathbf{x}) > 0, \dots, g_s(\mathbf{y}, \mathbf{x}) > 0$$
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				Hermite	RF	RRC
n	t	s	d			
2	2	2	2			
2	2	3	2			
3	2	1	2			
3	2	2	2			
2	3	2	2			
3	3	1	2			
2	2	1	3			
2	2	2	3			

Table: Generic dense system

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n	t	s	d	Hermite		RF	RRC
				hm	det	dv	bp
2	2	2	2	0.15 s	0.1 s	0.14 s	0.11 s
2	2	3	2	0.7 s	0.1 s	0.9 s	1 s
3	2	1	2	0.5 s	0.4 s	10 mn	7 mn
3	2	2	2	3 s	0.4 s	10 mn	14 mn
2	3	2	2	0.3 s	0.1 s	0.7 s	0.2 s
3	3	1	2	1 s	6 s	>50 h	>50 h
2	2	1	3	0.9 s	0.8 s	52 mn	47 s
2	2	2	3	5 s	1 s	57 mn	2 mn

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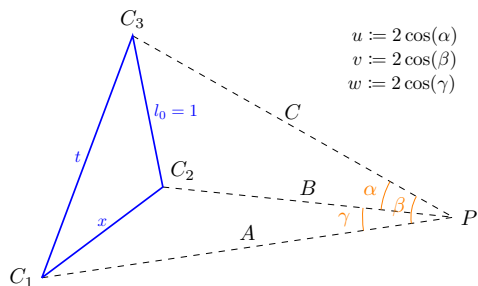
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3	2	1	2	0.5 s	0.4 s	9 s	33 s	10 mn	11 mn	7 mn
3	2	2	2	3 s	0.4 s	1 mn	57 s	10 mn	13 mn	14 mn
2	3	2	2	0.3 s	0.1 s	4 s	18mn	0.7 s	>50 h	0.2 s
3	3	1	2	1 s	6 s	4 mn	>50 h	>50 h	>50 h	>50 h
2	2	1	3	0.9 s	0.8 s	30 s	3mn	52 mn	57 mn	47 s
2	2	2	3	5 s	1 s	5 mn	6 mn	57 mn	1h 16 mn	2 mn

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Perspective-3-Point Problem



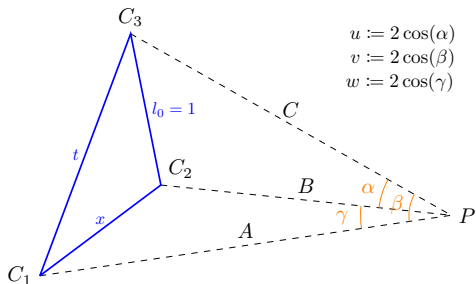
Perspective-3-Point Problem

$$\begin{cases} 1 &= A^2 + B^2 - ABu \\ t &= B^2 + C^2 - BCv \\ x &= A^2 + C^2 - ACw \end{cases}, \quad A, B, C > 0$$

with the constraints:

$$x, t > 0, \quad -2 < u, v, w < 2$$

- 3 variables : A, B, C
- 5 parameters : x, t, u, v, w



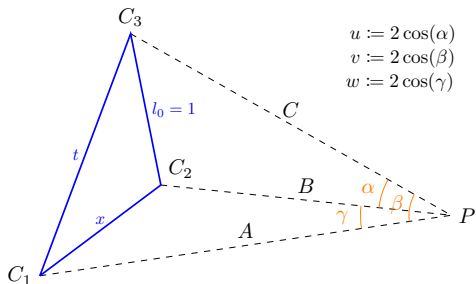
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Results

- A **complete classification** in less than **one hour** in the isosceles case ($t = 1$)
 - In the general case: able to compute the Hermite matrices and derive the **semi-algebraic conditions** from their minors.
- **Next step**: compute all the possible number of solutions and determine which conditions are feasible using the sample points routine

Conclusion

New Algorithm for **One-block Quantifier Elimination**?

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Thank you for your attention!

