

OPTIMIZING A PARAMETRIC LINEAR FUNCTION OVER A NON-COMPACT REAL VARIETY

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Problem Statements

Let h_1, \dots, h_p be polynomials in $\mathbb{R}[X]$ which define the algebraic variety

$$\mathcal{V} = \{x \in \mathbb{C}^n \mid h_1(x) = \dots = h_p(x) = 0\}.$$

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- ▶ How to compute a polynomial $\Phi \in \mathbb{R}[c_0, \mathbf{c}]$ s.t. c_0^* can be obtained by solving $\Phi(c_0, \gamma) = 0$ for a generic $\gamma \in \mathbb{R}^n$?

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- ▶ Can we compute a polynomial family $\{\Phi_i\} \in \mathbb{R}[c_0, \mathbf{c}]$, s.t. $\forall \gamma \in \mathbb{R}^n$, there exists k

$$\Phi_k(c_0, \gamma) \neq 0, \quad \Phi_k(c_0^*, \gamma) = 0?$$

State of the Art

Previous work on the problem:

- ▶ CAD can be used to describe the optimal value function by a sequence of polynomials of degree **doubly exponential** in n . [Brown, Collins, Hong, McCallum among many others]

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- ▶ For the specialized optimization problem, the algorithm based on modified **polar varieties** [Greuet, Guo, Safey El Din, Zhi], [Greuet, Safey El Din] allows us to compute a polynomial of degree **singly exponential** in n whose roots contain the maximal value. It works for **noncompact** cases.

Our Contributions

We generalize the results of Rostalski and Sturmfels and have the following conclusions:

- ▶ When \mathcal{V} is nonsmooth and compact in \mathbb{R}^n , dual varieties of regular locus and singular locus give such a polynomial Φ for generic parameter's value γ .

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We compute finitely many pairs of polynomials (Φ_i, Z_i) where $\Phi_i \in \mathbb{Q}[c_0, \mathbf{c}]$ and $Z_i \in \mathbb{Q}[\mathbf{c}]$ such that

- ▶ for each γ , there exists k such that $\gamma \notin V(Z_k)$ and $\Phi_k(c_0, \gamma) \neq 0$;
- ▶ if c_0^* is finite for γ , $\Phi_k(c_0^*, \gamma) = 0$.

Main Results for Smooth and Noncompact Case

Let $\mathcal{V}^* \subset \mathbb{P}^n$ be the dual variety to the projective closure of \mathcal{V} and $C_{\mathbf{h}}$ the closure of the convex hull of $\mathcal{V} \cap \mathbb{R}^n$.

We extend the result of [Rostalski, Sturmfels] and have the following conclusions:

Theorem

Suppose that \mathcal{V} is **equidimensional** and **smooth**, then

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Theorem

When \mathcal{V} is **irreducible**, **smooth** and $C_{\mathbf{h}}$ contains no lines, we have

- ▶ \mathcal{V}^* is an irreducible hypersurface,
- ▶ its defining polynomial is $\Phi(-c_0, c_1, \dots, c_n)$, where $\Phi(c_0^*, \gamma) = 0$ for each $\gamma \in \mathbb{R}^n$.

Bad Parameters' Values Cases

$$\Phi = \Phi_0(c_1, \dots, c_n)c_0^m + \Phi_1(c_1, \dots, c_n)c_0^{m-1} + \dots + \Phi_m(c_1, \dots, c_n)$$

Example

We consider the optimization problem:

$$\sup c_1x_1 + c_2x_2 + c_3x_3 + c_4x_4 \quad \text{s.t. } x \in \mathcal{V} \cap \mathbb{R}^4,$$

where $\mathcal{V} = \mathbf{V}(x_4 - (x_1 + x_1^2x_2^2 + x_1^4x_2x_3)^2)$. The dual variety \mathcal{V}^* is defined by

$$\begin{aligned} \Phi := & 1073741824c_0^{12}c_2^4c_4^2 + 268435456c_0^{11}c_1^2c_2^4c_4 - 134217728c_0^{11}c_2^2c_3^2c_4^3 \\ & - 33554432c_0^{10}c_1^2c_2^2c_3^2c_4^2 + \dots + 520093696c_0^9c_1c_2^3c_3^2c_4^3. \end{aligned}$$

- ▶ When $\gamma \in V(c_2c_4, c_3c_4, c_3c_2c_1)$, $\Phi(c_0, \gamma) \equiv 0$.
- ▶ $\Phi(c_0, 0, 0, 0, -1) \equiv 0$ which gives no info on c_0^* .

Algorithm for Non-parametric Optimization

Construct a one dimensional subvariety $\mathcal{C} \subseteq \mathcal{V}$ such that

$$\sup_{x \in \mathcal{V} \cap \mathbb{R}^n} f(x) = \sup_{x \in \mathcal{C} \cap \mathbb{R}^n} f(x).$$

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Asymptotic Values of f on \mathcal{V}

$\{z \in \mathbb{R} \mid \exists y_k \in \mathcal{V}, k = 1, 2, \dots \text{ such that } \|y_k\| \rightarrow \infty, f(y_k) \rightarrow z\}$.

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Output: $\Phi(c_0)$ s.t. $f(S_1 \cup S_2 \cup S_3) \subseteq \mathbf{V}(\Phi)$

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Polar Varieties

Suppose \mathcal{V} is an equidimensional and smooth variety. Let \mathbf{H} be $\{h_1, \dots, h_p\}$ and $\langle h_1, \dots, h_p \rangle$ is radical. $\text{MaxMinors}(\text{jac}(\mathbf{H}, \mathbf{X}_{\geq i+1}))$ is denoted to be the $(n-i) \times (n-i)$ minors of the Jacobian of \mathbf{H} respect to x_{i+1}, \dots, x_n .

Polar Varieties [Bank, Giusti, Heintz, Mbakop, Pardo, Safey El Din, Schost]

Polar varieties are defined to be a sequence of varieties $\{W_i\}$, where W_{n-i+1} is the **critical locus** of

$$\pi_i : (X_1, \dots, X_n) \longrightarrow (X_1, \dots, X_i)$$

restricted to \mathcal{V} .

W_{n-i+1} is the variety of \mathbf{H} and $\text{MaxMinors}(\text{jac}(\mathbf{H}, \mathbf{X}_{\geq i+1}))$.

Modified Polar Varieties

Modified Polar Varieties

Let W_{n-i+1} be the variety of \mathbf{H} , $\text{MaxMinors}(\text{Jac}([f, \mathbf{H}], \mathbf{X}_{\geq i+1}))$ and

$$W = \cup_{i=1}^d M_i, \text{ where } M_i = W_{n-i+1} \cap V(X_1, \dots, X_{i-1}), 1 \leq i \leq d.$$

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After a generic linear change of coordinates,

- ▶ $f(V \cap \mathbb{R}^n) = f(W \cap \mathbb{R}^n)$;
- ▶ the set of asymptotic value of f on W is finite.
- ▶ $\dim(\mathcal{C}) = 1$ for $\mathcal{C} := \overline{W \setminus \text{Crit}(f, \mathcal{V})}$.

Algorithm for Parametric Optimization

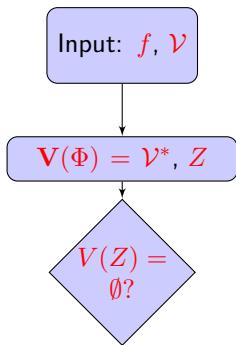
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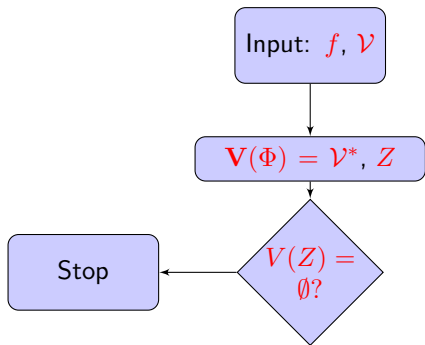
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$\mathbf{V}(\Phi) = \mathcal{V}^*, Z$

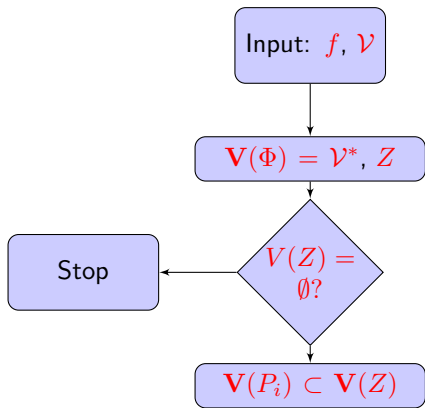
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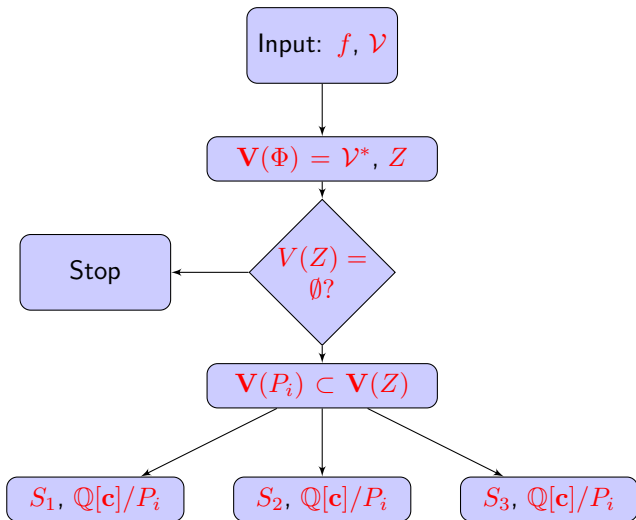
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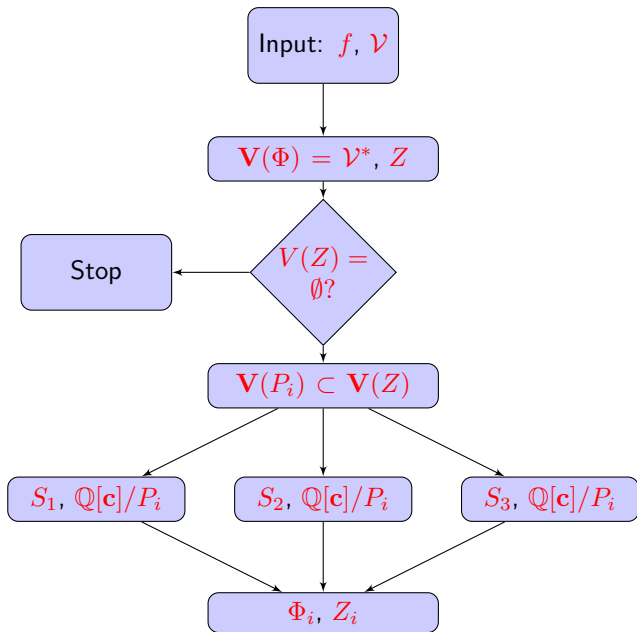
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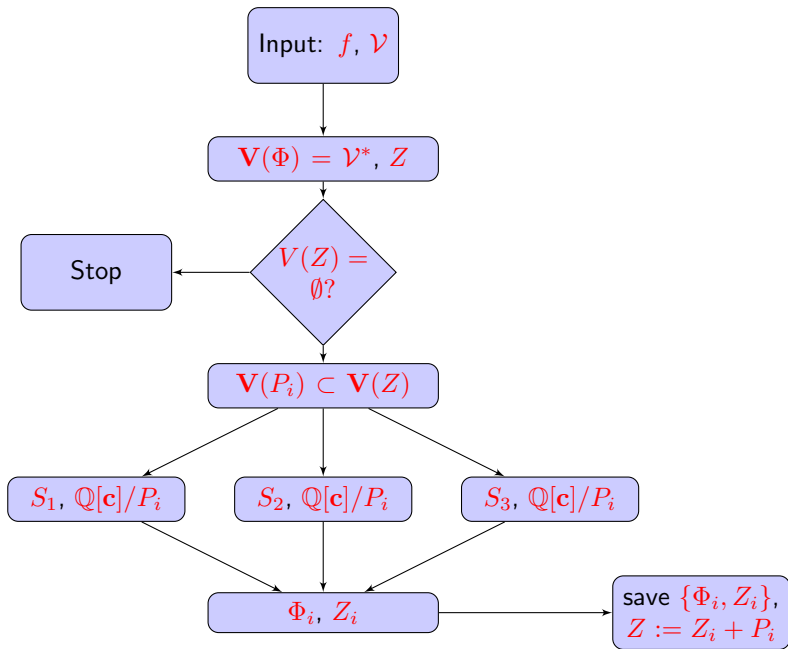
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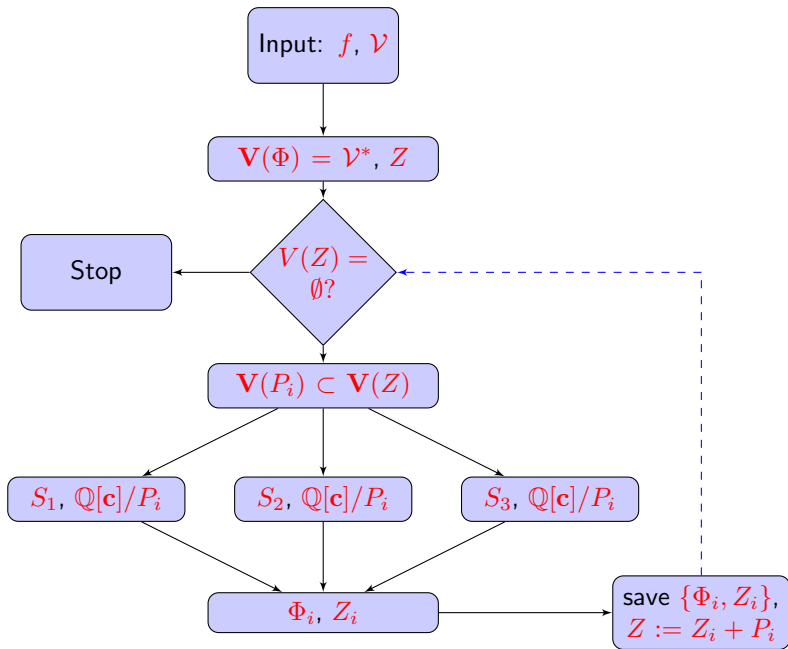
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Identify Bad Parameters' Values

Bad parameters' values

γ is a bad parameter's value of Φ_i if

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- ▶ Some parameters' values are not in **generic position** by one random linear change.
- ▶ The computation of Gröbner basis may not be **commutative** for the **specialization** operation.

Example (continued)

$I = \langle c_2c_4, c_3c_4, c_1c_2c_3 \rangle$: $\mathbf{V}(I)$ contains the bad parameters' values.

The primary decomposition of I :

$$I = \langle c_1, c_4 \rangle \cap \langle c_2, c_4 \rangle \cap \langle c_3, c_4 \rangle \cap \langle c_2, c_3 \rangle.$$

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Let us consider $\mathbf{P} = \langle c_2, c_3 \rangle$ and solve the following optimization problem:

$$\sup c_1x_1 + c_4x_4$$

$$s.t. \quad x_4 - (x_1 + x_1^2x_2^2 + x_1^4x_2x_3)^2 = 0.$$

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- ▶ Since $V(P + Z) \neq \emptyset$, $\sqrt{\mathbf{P} + \langle Z \rangle} = \langle c_1, c_2, c_3 \rangle \cap \langle c_2, c_3, c_4 \rangle$, consider $\mathbf{P}' = \langle c_1, c_2, c_3 \rangle$.

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We get $\Phi = c_0$ and $V(Z) = \emptyset$.

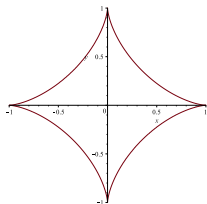
Main Results for Singular Case

Let \mathcal{V} be defined by

$$(X_1^2 + X_2^2 - 1)^3 + 27X_1^2X_2^2.$$

The defining polynomial of \mathcal{V}^* is

$$\Phi_1 := -c_1^2c_2^2 + c_0^2c_1^2 + c_2^2c_0^2.$$



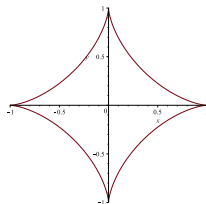
Main Results for Singular Case

Let \mathcal{V} be defined by

$$(X_1^2 + X_2^2 - 1)^3 + 27X_1^2X_2^2.$$

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Let $k = 1$ and $\mathcal{V}_k = \mathcal{V}$.

Step 1 Compute radical and equidimensional decomposition

$$\mathcal{V}_k = \cup_i \mathcal{V}_{k,i}.$$

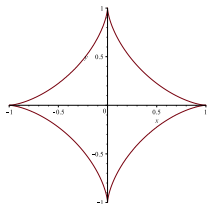
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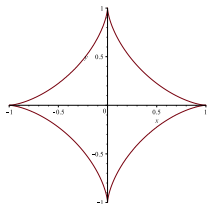
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Step 3 Compute the singular locus $\tilde{\mathcal{V}}_{k,i}$ of each $\mathcal{V}_{k,i}$. Let

$$\mathcal{V}_{k+1} = \cup_i \tilde{\mathcal{V}}_{k,i} \text{ and } k = k + 1. \text{ Go to Step 1.}$$

Main Results for Singular Case

Theorem

The algorithm terminates in a finite number k steps and we have

$$(-c_0^* : \gamma_1 : \cdots : \gamma_n) \subseteq \bigcup_{i=1}^k (\mathcal{V}^{(k)})^*.$$

for every γ such that c_0^ is finite.*

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In this example, $\text{Sing}(\mathcal{V})$ is equidimensional. Its dual variety $(\mathcal{V}^{(2)})^*$ is defined by

$$\Phi_2 = (c_0 - c_2)(c_0 + c_2)(c_0 - c_1)(c_0 + c_1)(c_0^2 + c_1^2 - 2c_1c_2 + c_2^2) \\ (c_0^2 + c_1^2 + 2c_1c_2 + c_2^2).$$

Then we have $(-c_0^* : \gamma_1 : \gamma_2) \in \mathbf{V}(\Phi_1\Phi_2)$.

Conclusions and Future Work

- ▶ How to compute a polynomial Φ when the feasible set is a real variety which is **noncompact** and **nonsmooth**?
- ▶ How to compute a polynomial Φ when the feasible set is a **semialgebraic set**?