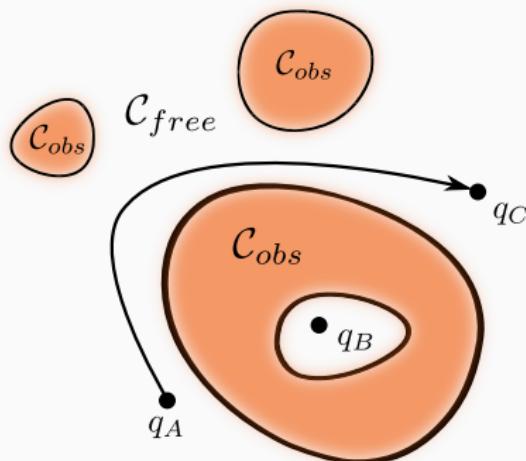


Connectivity in Real Algebraic Sets: Algorithms & Applications

17th October 2024

CFHP Seminar



Rémi PRÉBET

Joint works with M. SAFEY EL DIN, É. SCHOST

Md N. ISLAM, A. POTEAUX

D.CHABLAT, D.SALUNKHE, P. WENGER

SLIDES:

rprebet.github.io/#talks

INTERACTIVE EXAMPLE:

rprebet.github.io/cristal.html

Algorithms for polynomial systems with real variables

Semi-algebraic sets

Real solutions of **polynomial** systems of equations and inequalities

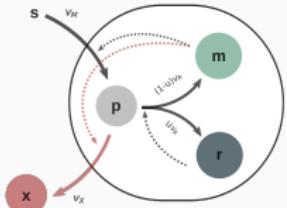
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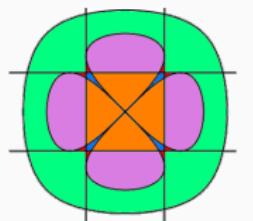
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$$(x - 1)^2 + \frac{(y - 1)^2}{9} - 1 = 0$$



Biology



Physics

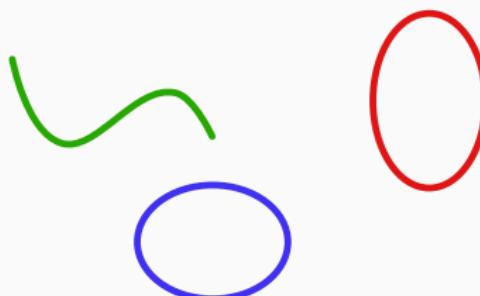


Robotics

Algorithms for polynomial systems with real variables

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Fundamental algorithmic problems

Project : what is the set of possible values?

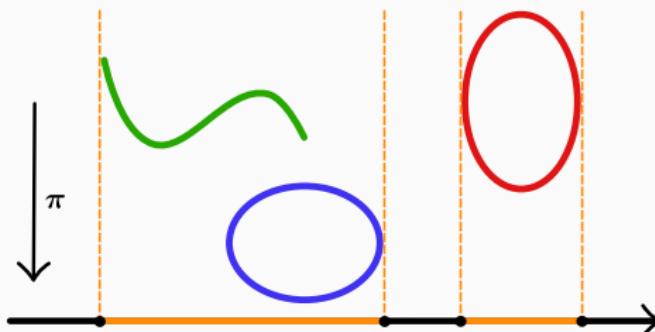
Sample : are there any solutions?

Connect : are two points connected?

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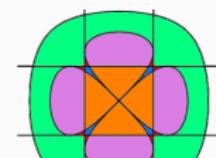


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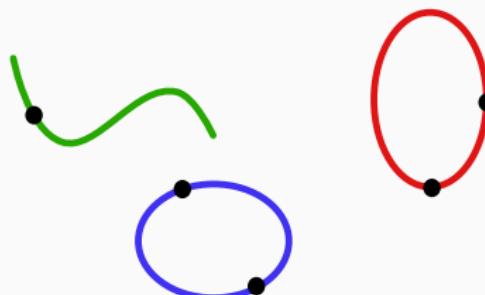
■ 2, ■ 4, ■ 6, ■ 8, ■ 10

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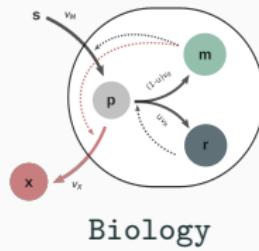


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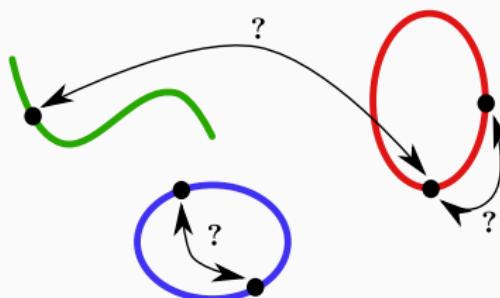


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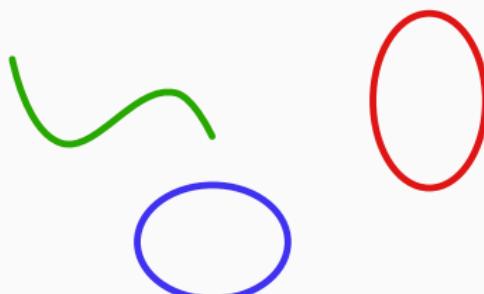


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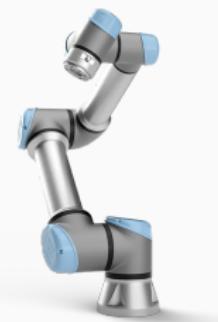


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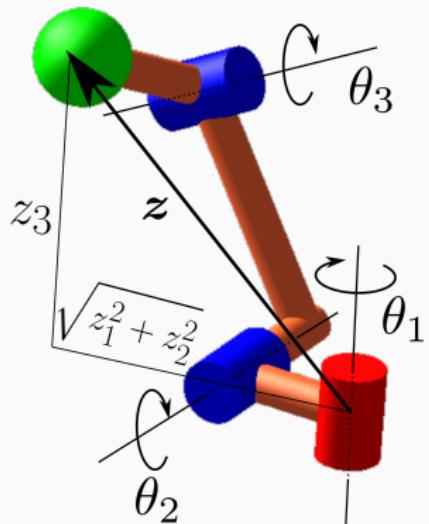


Robotics

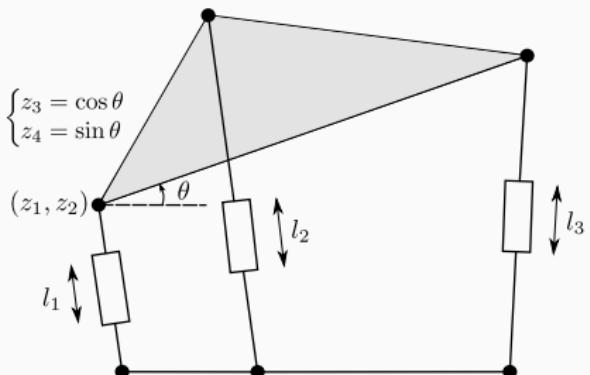
A quick look at robotics

Kinematic map of a robot

$$\begin{aligned} \mathcal{K}: \quad \mathbb{R}^d &\rightarrow \mathbb{R}^d \\ (\ell, \theta) &\mapsto z = (z_1(\ell, \theta), \dots, z_d(\ell, \theta)) \end{aligned}$$



An Orthogonal 3R Serial Robot



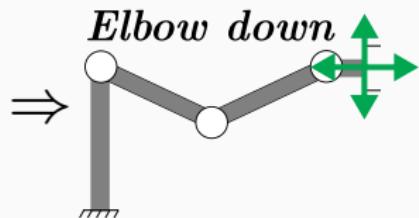
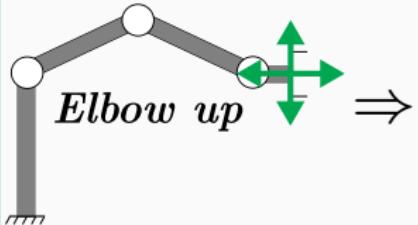
A 3-RPR Planar Parallel Robot

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Not cuspidal: controlled posture 

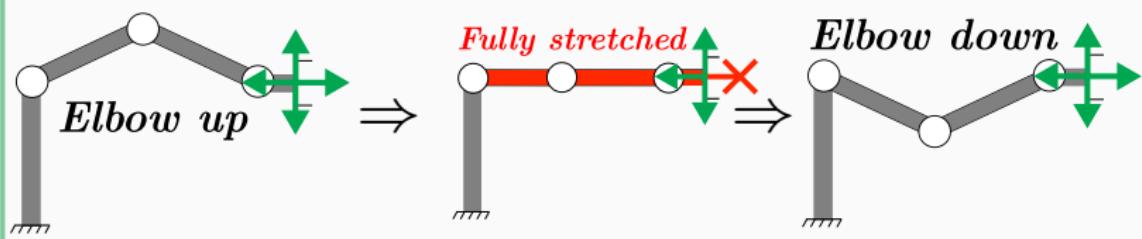


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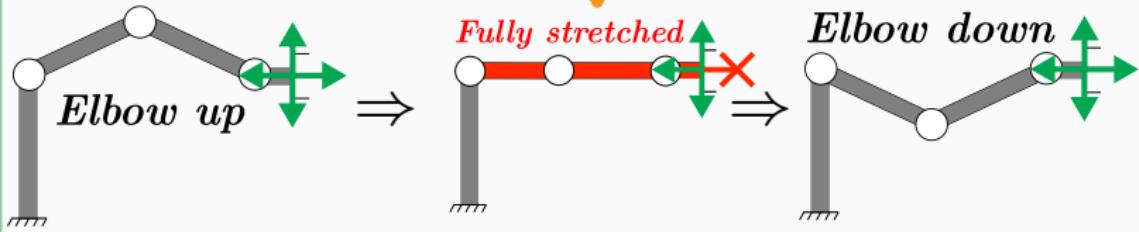
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Singular posture

Configurations (ℓ, θ) s.t.
 $\text{Jac}_{\ell, \theta}(\mathcal{K})$ is rank deficient

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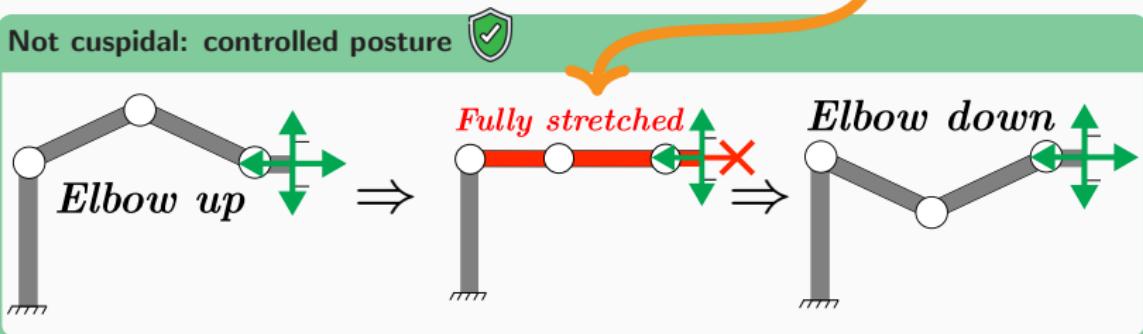
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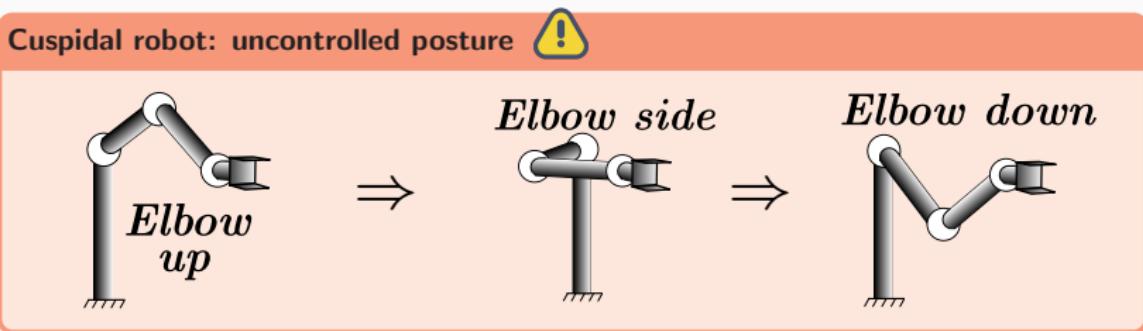
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Not cuspidal: controlled posture



Cuspidal robot: uncontrolled posture



Cuspidal robot

Theorem

[Borrel & Liégeois, 1986]

A robot **cannot** move between two associated postures,
without passing by a singular posture

Cuspidal robot

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Cuspidal robots can **induce problem**
for task planning

Open problem

Cuspidality decision for a general robot



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(Chablat, P., Safey, Salunkhe, Wenger; 2022)

There exists a **general** algorithm

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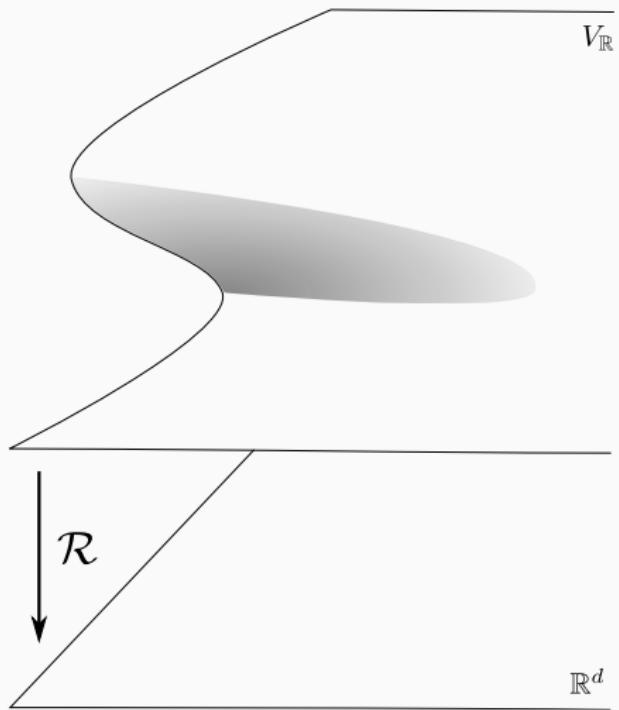
There exists a general algorithm with singly exponential complexity

The algebraic cuspidality problem

Data

Data: $\mathbf{f} = (f_1, \dots, f_s)$ and $\mathcal{R} = (r_1, \dots, r_d)$ polynomials in $\mathbb{R}[x_1, \dots, x_n]$

Assumptions: $V = V(\mathbf{f})$ is d -equidimensional and $V_{\mathbb{R}} = V \cap \mathbb{R}^n \subsetneq \text{sing}(V)$

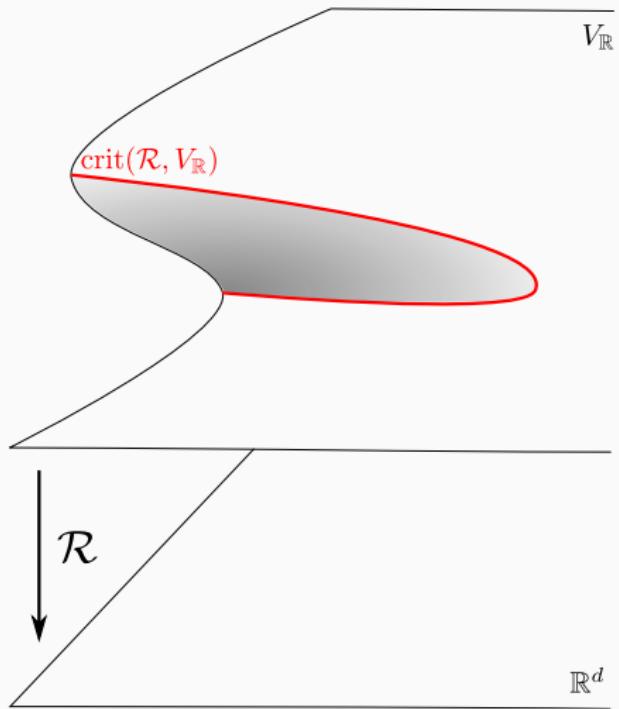


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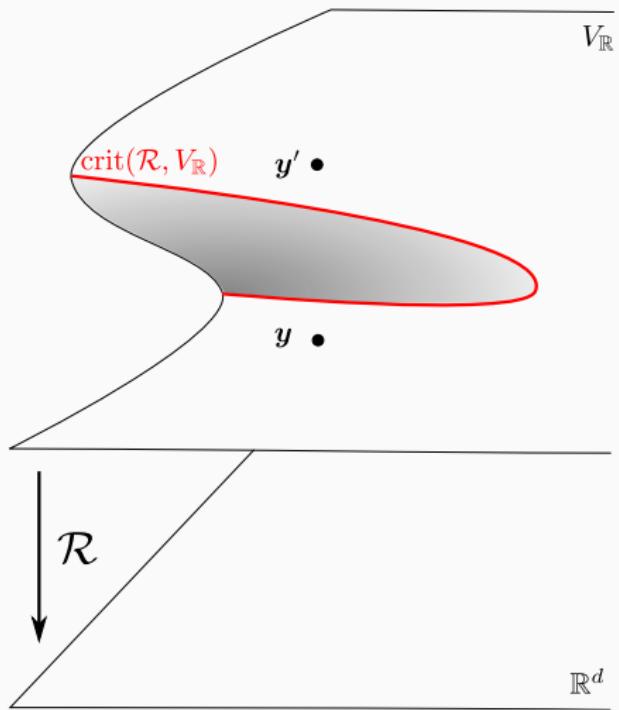
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Algebraic cuspidality problem

The restriction of \mathcal{R} to $V_{\mathbb{R}}$ is cuspidal if there is $\mathbf{y} \neq \mathbf{y}' \in V_{\mathbb{R}}$ such that



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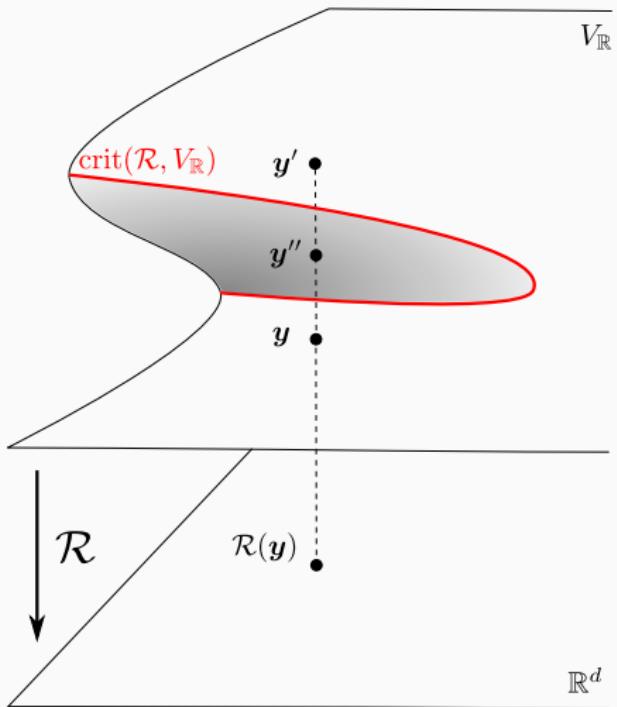
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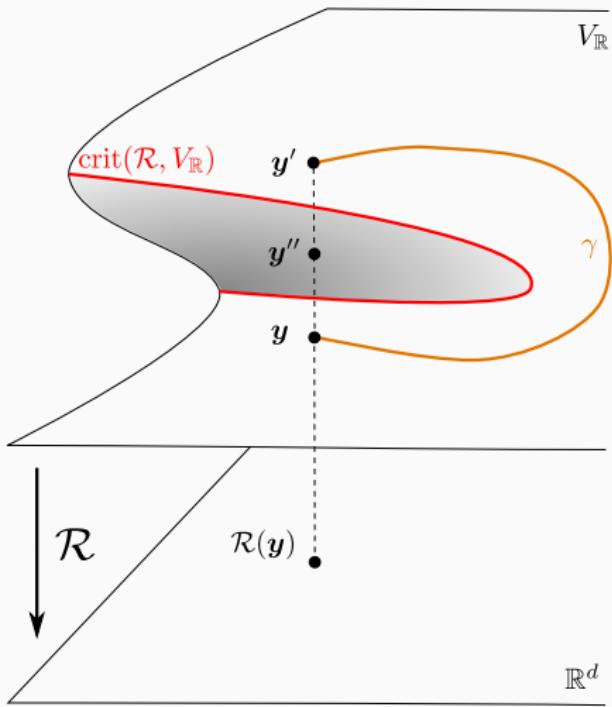
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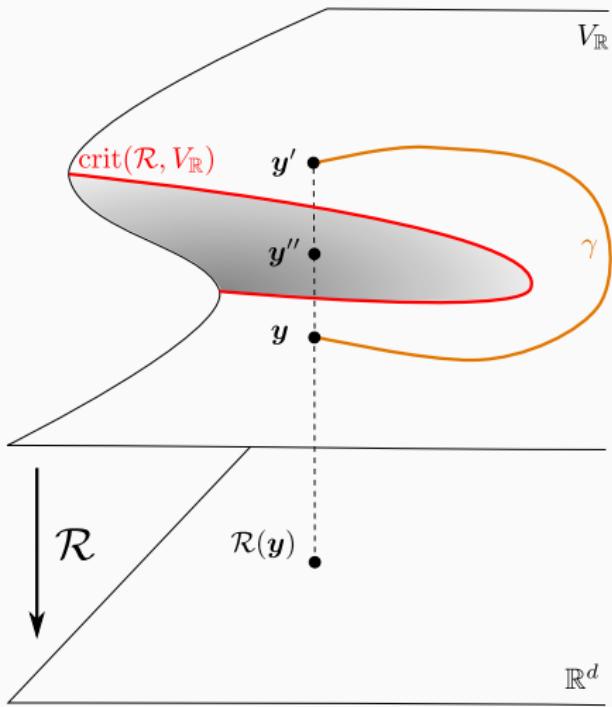
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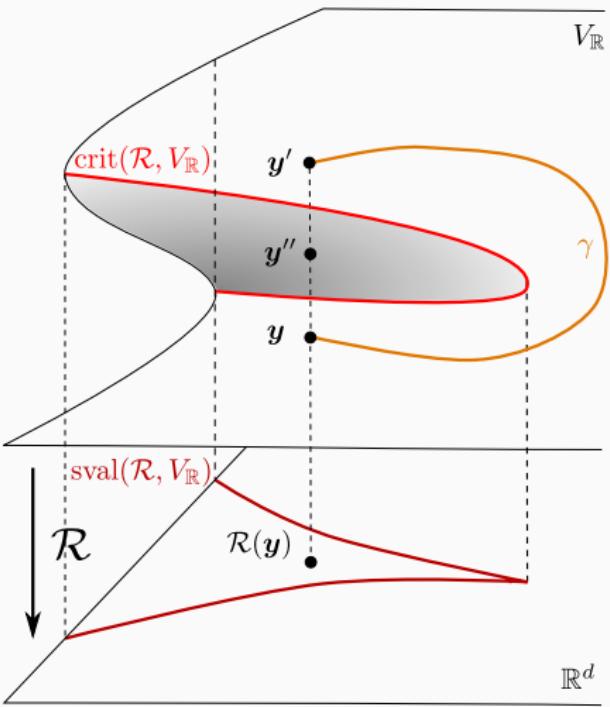
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Singular values of \mathcal{R}

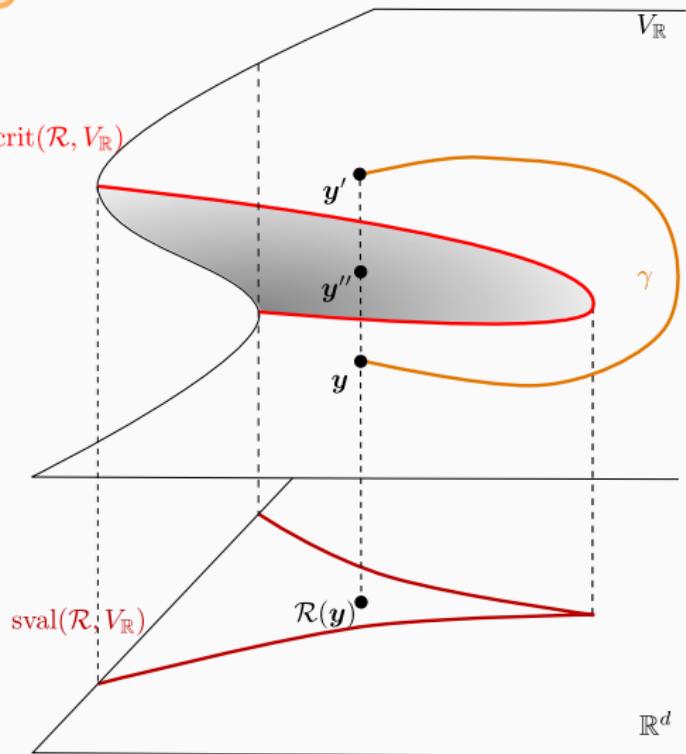
$$\text{sval}(\mathcal{R}, V) = \mathcal{R}(\text{crit}(\mathcal{R}, V))$$



The cuspidality algorithm

Thom's First Isotopy Lemma

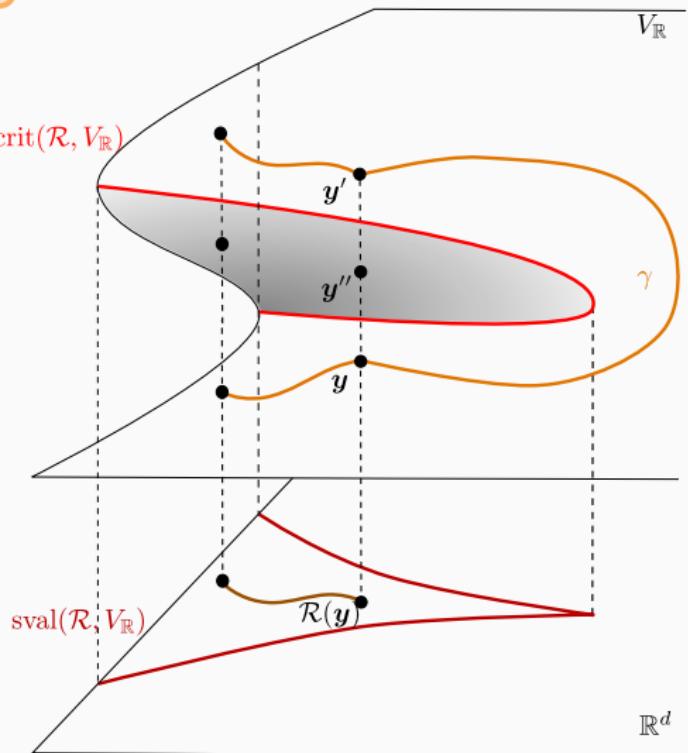
Fibers from the same connected component of $\mathbb{R}^d - \text{sval}(\mathcal{R}, V)$ have the **same type**



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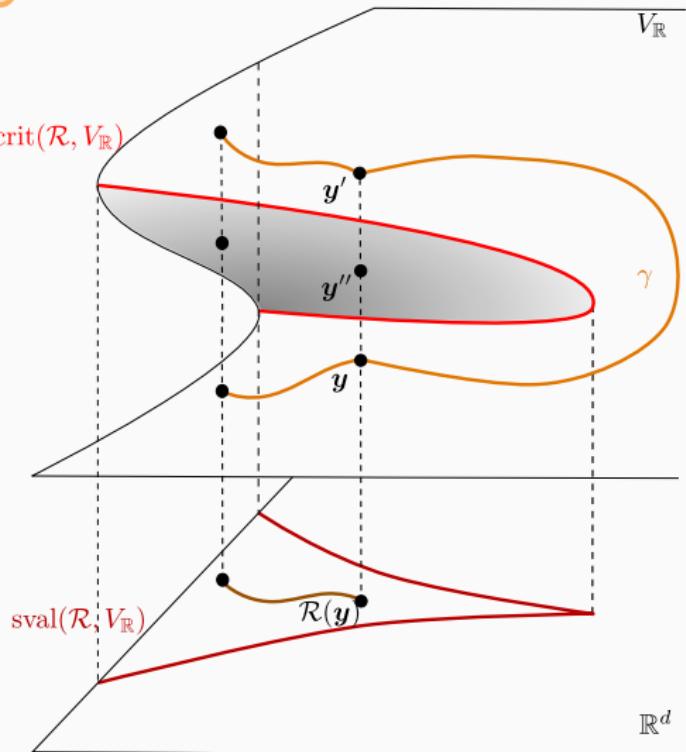
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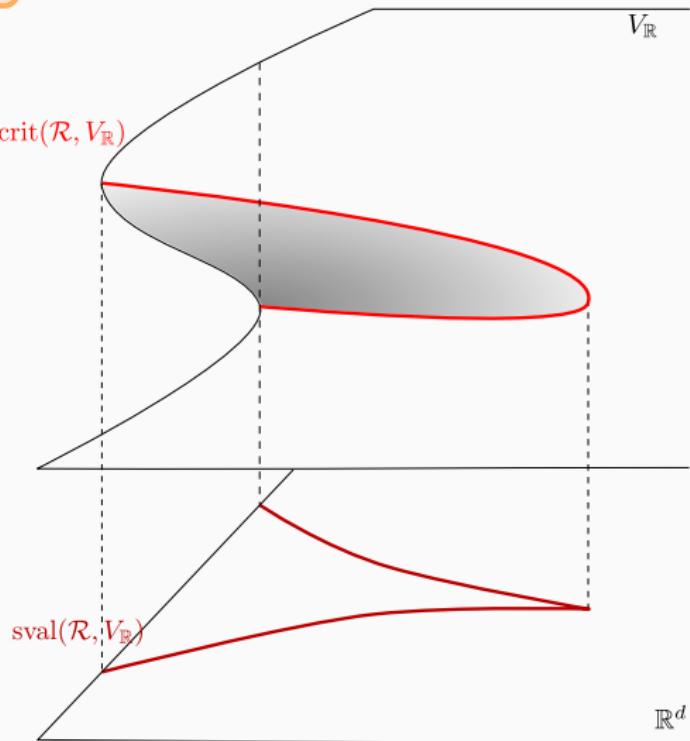
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>Main steps

1. Compute polynomials defining $\text{sval}(\mathcal{R}, V) = \mathcal{R}(\text{crit}(\mathcal{R}, V))$



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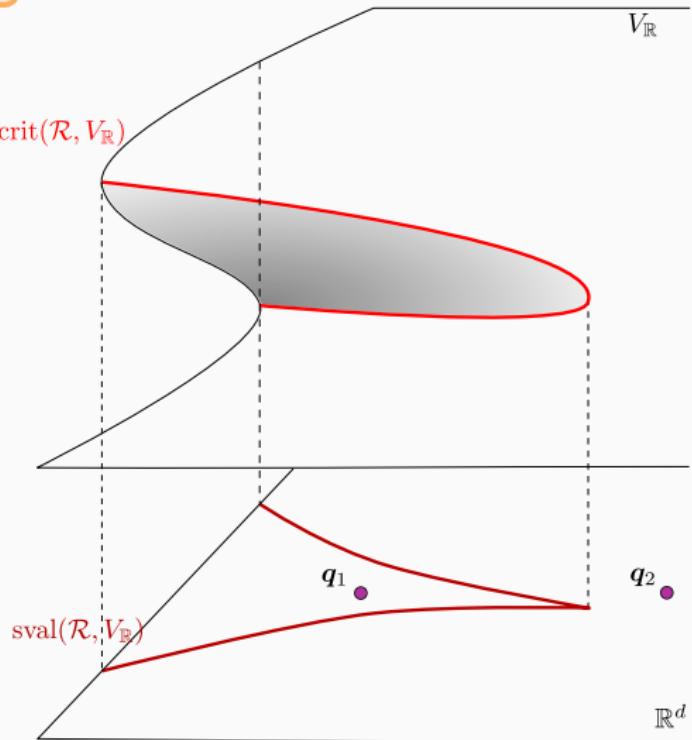
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The cuspidality algorithm

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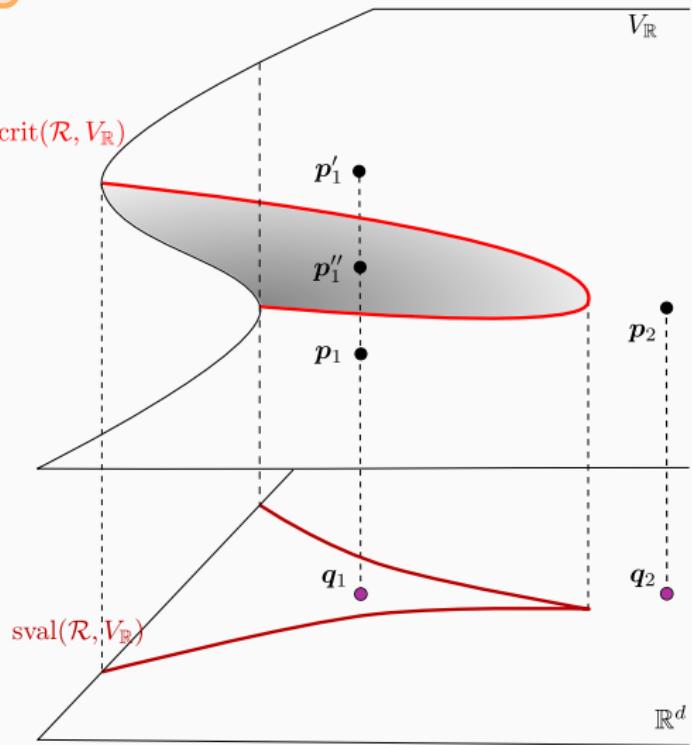
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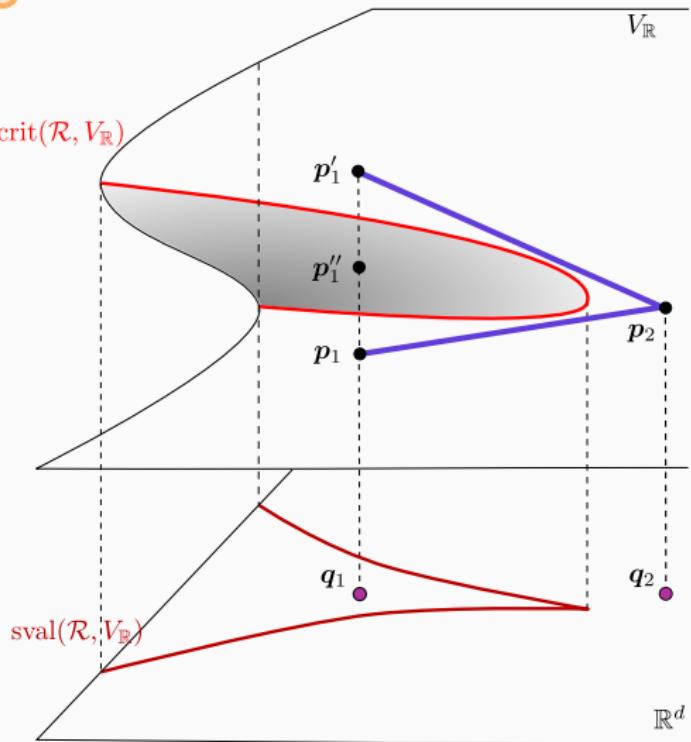
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3. Compute their preimages $\mathcal{P} = V \cap \mathcal{R}^{-1}(\mathcal{Q})$
4. Search for cuspidal pairs in \mathcal{P} by connecting points in the same connected component of $V_{\mathbb{R}} - \text{crit}(\mathcal{R}, V)$



Complexities of semi-algebraic set algorithms

Semi-algebraic set S

Defined by s polynomials (equations+inequalities) with n variables of deg $\leq D$

General algorithm [Collins; '75]

Complexity: $(sD)^{2^{O(n)}}$

$$\begin{cases} 4y + x^3 - 4x^2 - 2x - 8 = 0 \\ -2 \leq x \leq 0 \end{cases}$$
$$(x - 1)^2 + \frac{(y - 1)^2}{9} - 1 = 0$$
$$\frac{x^2}{4} + y^2 - 1 = 0$$

Fundamental algorithmic problems

Project S on t coordinates $\rightsquigarrow (sD)^{O(nt)}$ [Renegar, '92]

Optimal !

Sample points of S $\rightsquigarrow (sD)^{O(n)}$ [Basu-Pollack-Roy, '98]

Optimal !

Connect two points in S $\rightsquigarrow (sD)^{O(n^2)}$ [Canny, '88]

Not optimal !

Quantitative bounds on algebraic sets

Real algebraic sets

$$V_{\mathbb{R}} = \{f_1 = \cdots = f_p = 0\} \subset \mathbb{R}^n$$

where

$$(f_1, \dots, f_p) \subset \mathbb{R}[x_1, \dots, x_n]$$



Real trace of algebraic sets

$$V_{\mathbb{R}} = V \cap \mathbb{R}^n$$

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$$V = \{f_1 = \cdots = f_p = 0\} \subset \mathbb{C}^n$$

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$\mathcal{H}_1, \dots, \mathcal{H}_n$ generic hyperplanes

Dimension

$$\dim V = \text{smallest } d \quad \text{s.t.}$$
$$\text{card}(V \cap \mathcal{H}_1 \cap \dots \cap \mathcal{H}_d) < +\infty$$

Degree

$$\deg V = \text{card}(V \cap \mathcal{H}_1 \cap \dots \cap \mathcal{H}_d)$$

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$$V = \{p_1, \dots, p_{15}\}$$
$$\deg V = 15$$

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$$V = \{f_1 = \dots = f_p = 0\} \subset \mathbb{C}^n$$

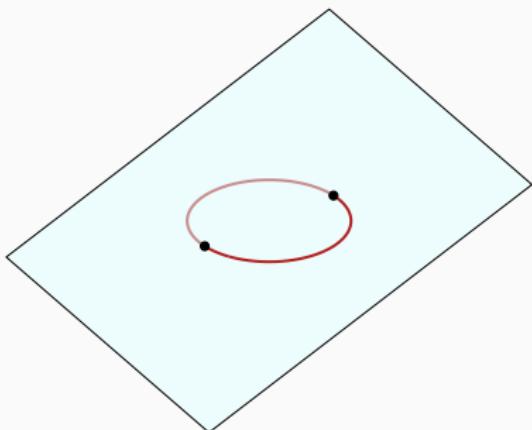
$\mathcal{H}_1, \dots, \mathcal{H}_n$ generic hyperplanes

Dimension

$$\dim V = \text{smallest } d \quad \text{s.t.}$$
$$\text{card}(V \cap \mathcal{H}_1 \cap \dots \cap \mathcal{H}_d) < +\infty$$

Degree

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$$\begin{aligned} & V(x^2 + y^2 - 1, z) \\ & \Rightarrow \deg V = 2 \end{aligned}$$

Quantitative bounds on algebraic sets

Real algebraic sets

$$V_{\mathbb{R}} = \{f_1 = \dots = f_p = 0\} \subset \mathbb{R}^n$$

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 $(f_1, \dots, f_p) \in \mathbb{R}[x_1, \dots, x_n]$

\iff

Real trace of algebraic sets

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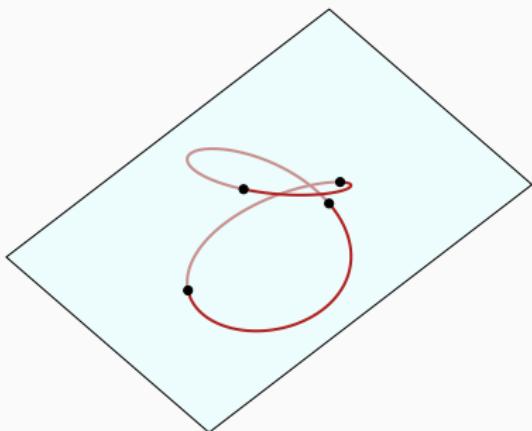
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Bézout Bound

$$\deg V \leq \prod_{j=1}^p \deg f_j$$



$$V(x^2 + y^2 - 1, 2z^2 - x - 1)$$
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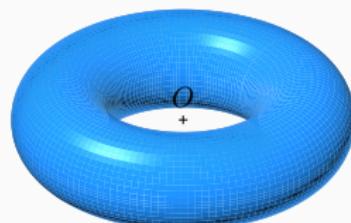
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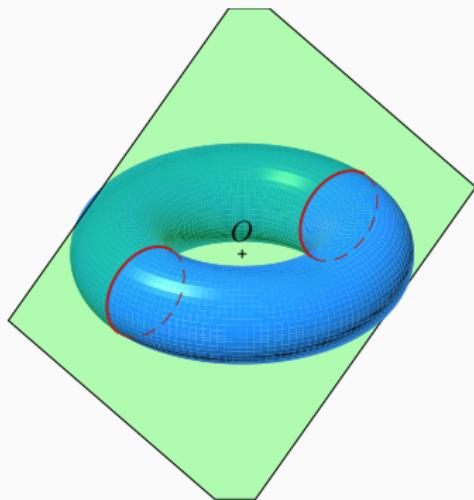
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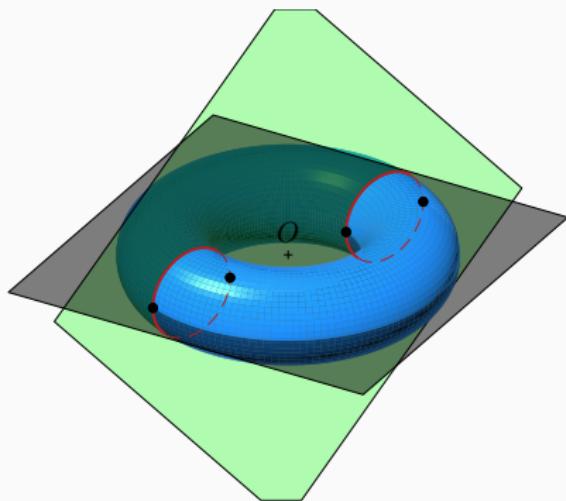
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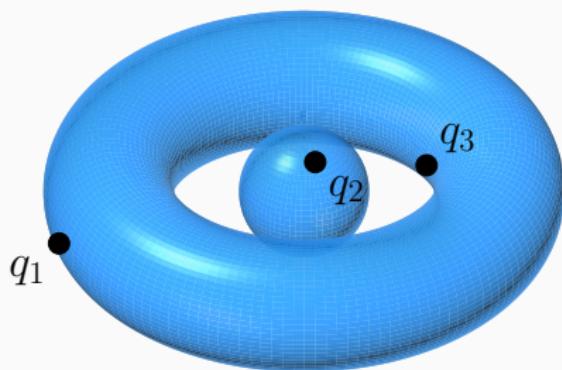
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Computing connectivity properties: Roadmaps

💡 [Canny, 1988] Compute $\mathcal{R} \subset S$ one-dimensional, sharing its connectivity

Roadmap of (S, \mathcal{P})

A semi-algebraic curve $\mathcal{R} \subset S$, containing query points (q_1, \dots, q_N) s.t.
for all connected components C of S : $C \cap \mathcal{R}$ is non-empty and connected

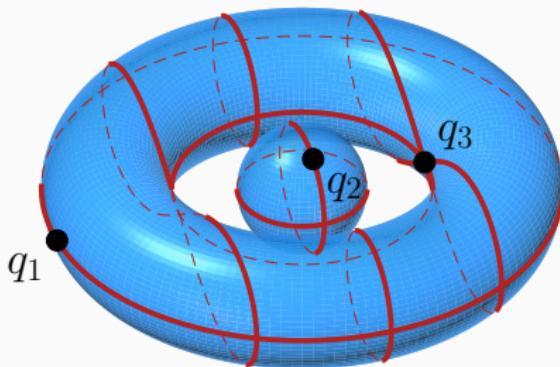


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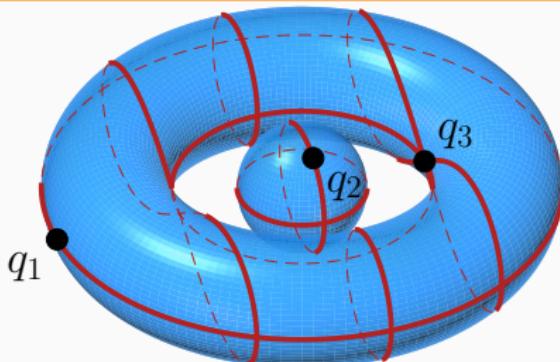
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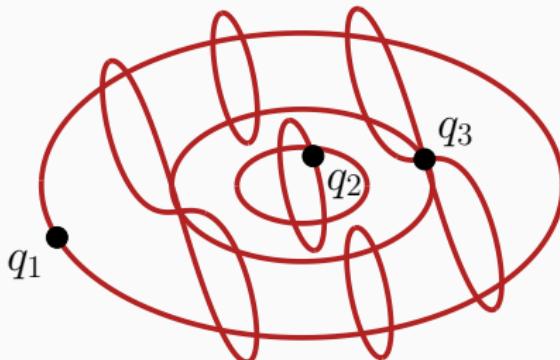
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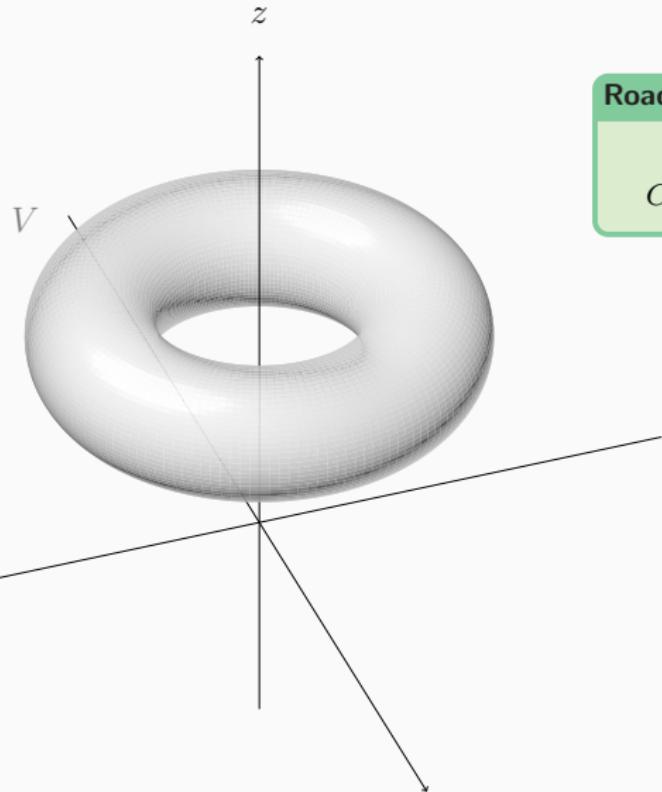
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Problem reduction

Arbitrary dimension $\xrightarrow[\text{ROADMAP}]{} \text{Dimension 1}$



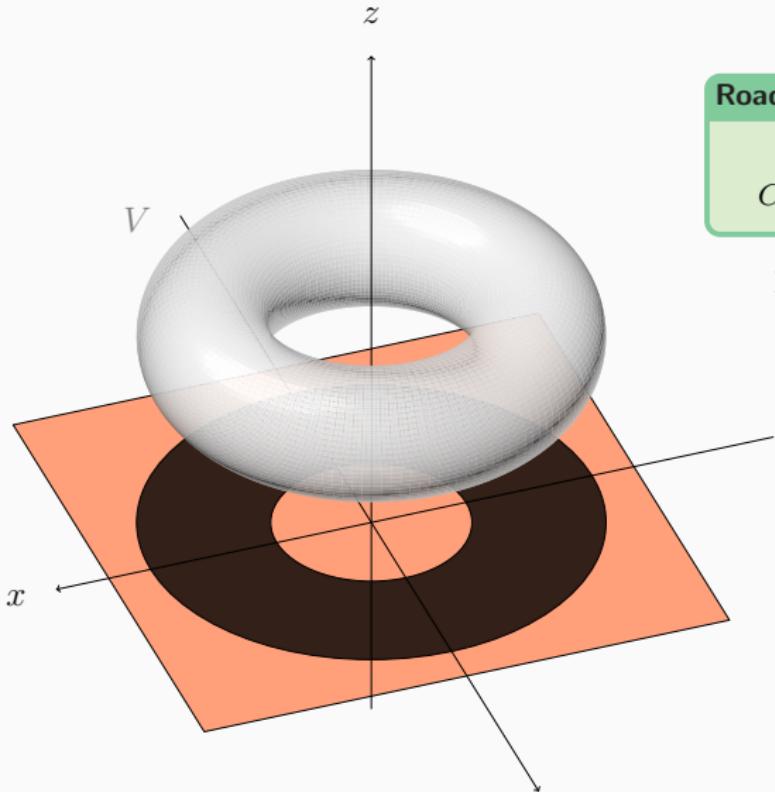
Canny's strategy



Roadmap property

$\forall C$ connected component,
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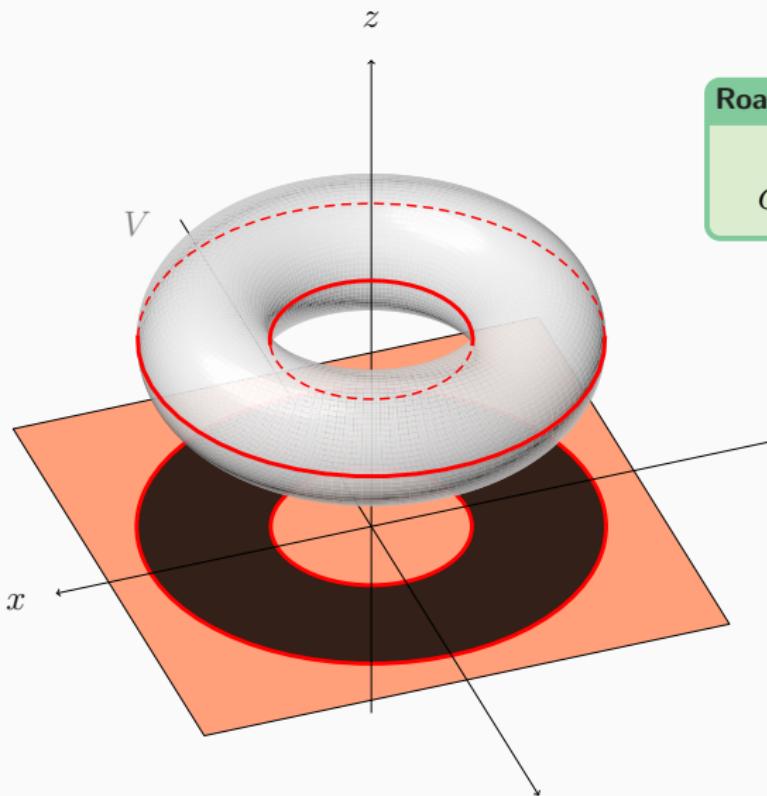
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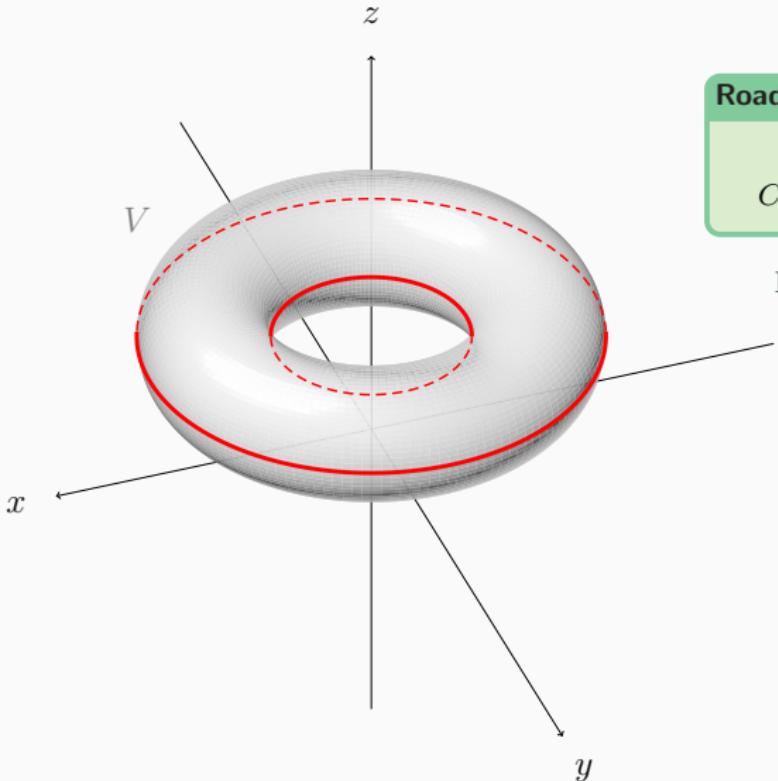
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Intersects all the
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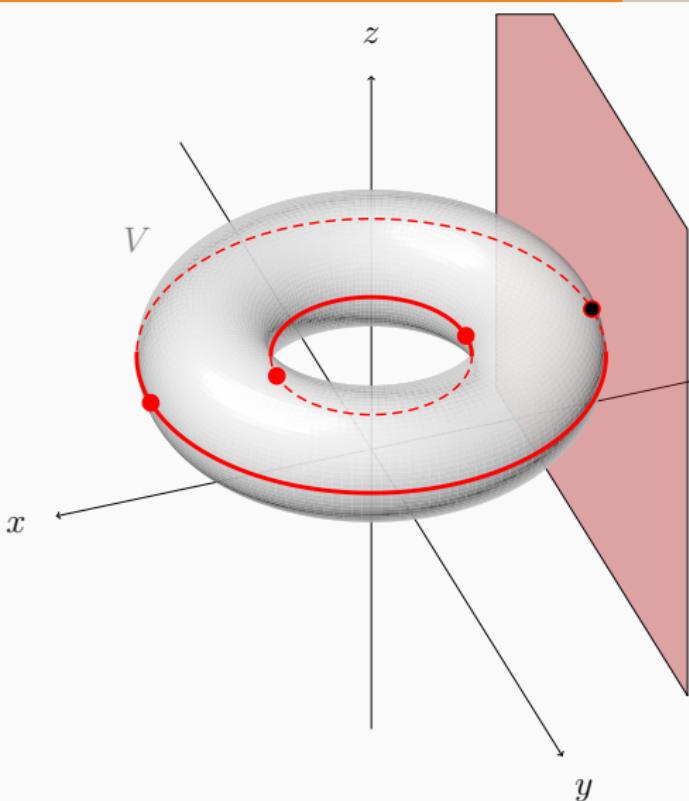
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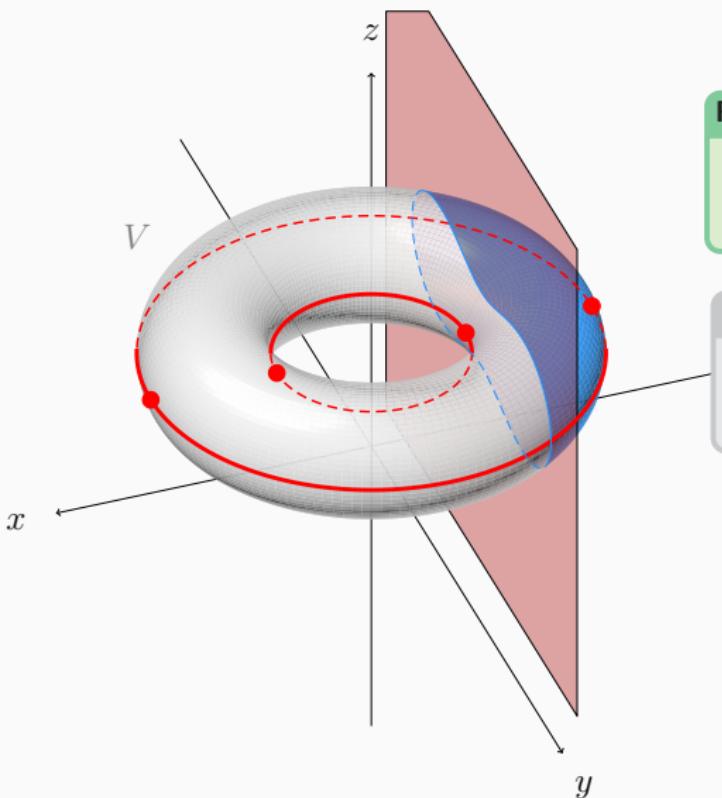
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Morse theory

“Scan” $W(\pi_2, V)$ at the critical values
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- We repair the connectivity failures with critical fibers
- We repeat the process at every critical value

Canny's strategy



Roadmap property

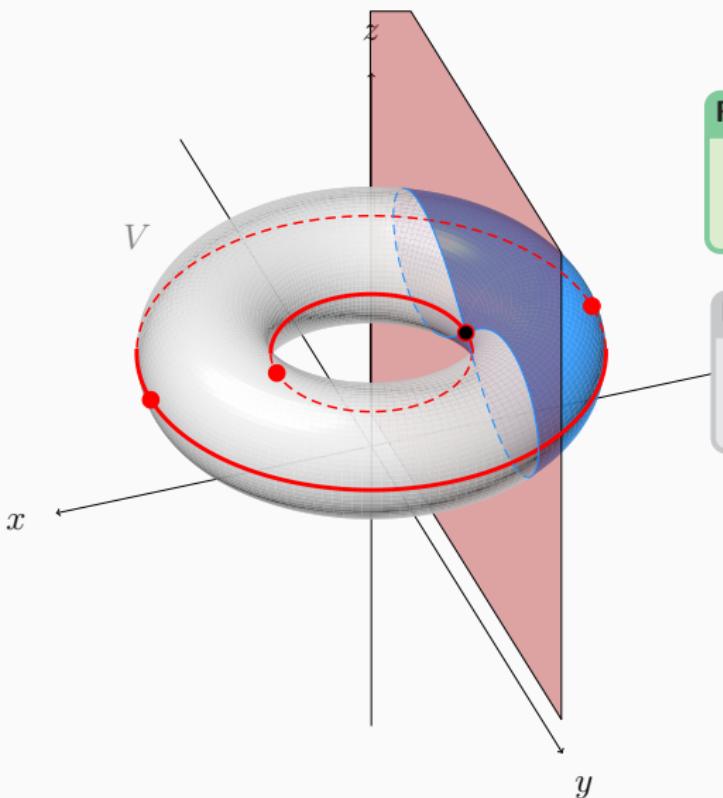
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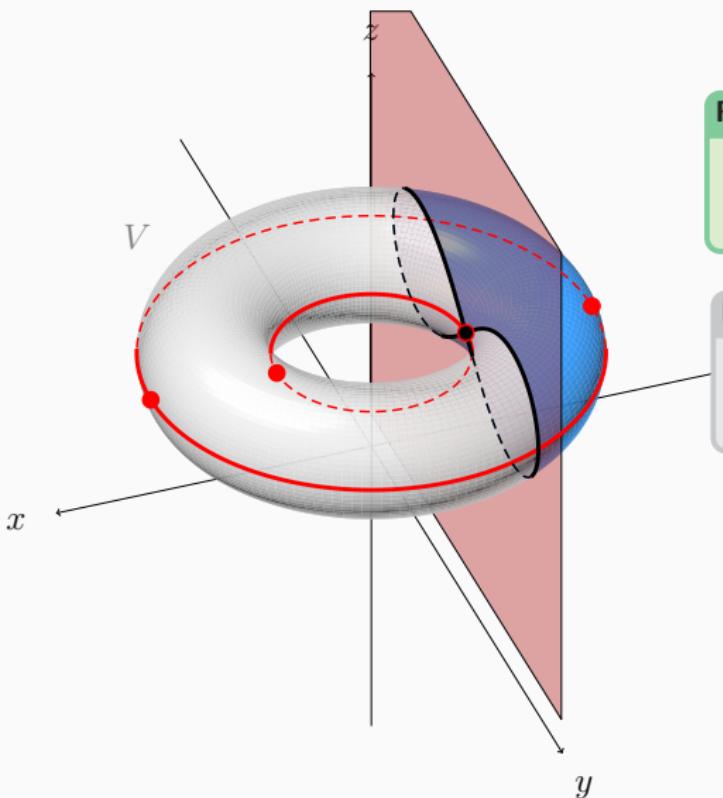
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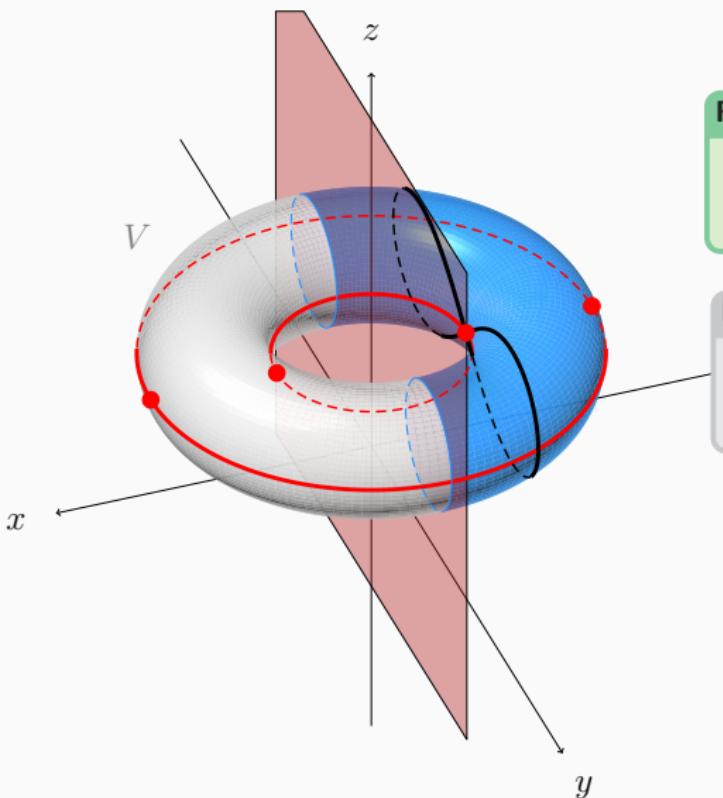
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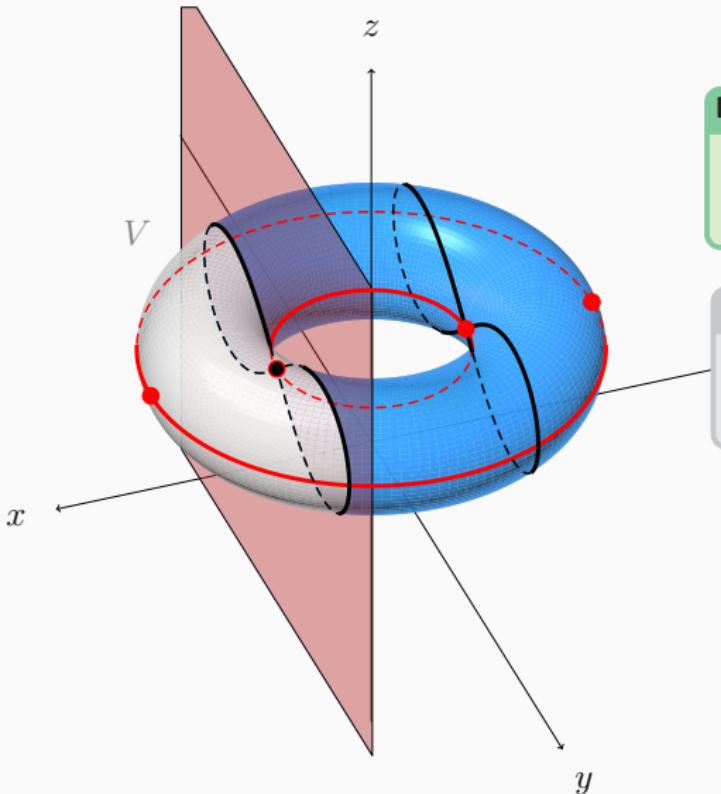
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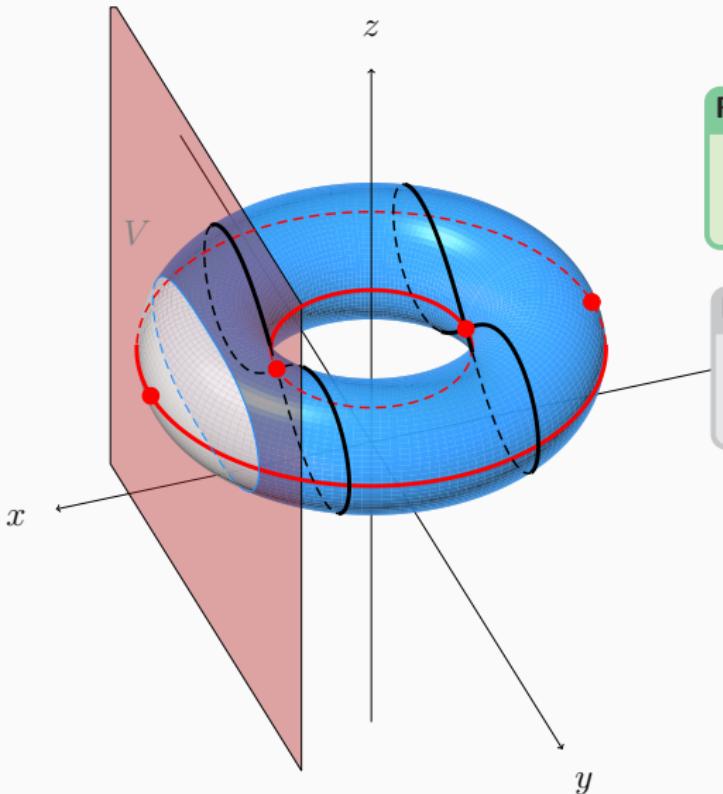
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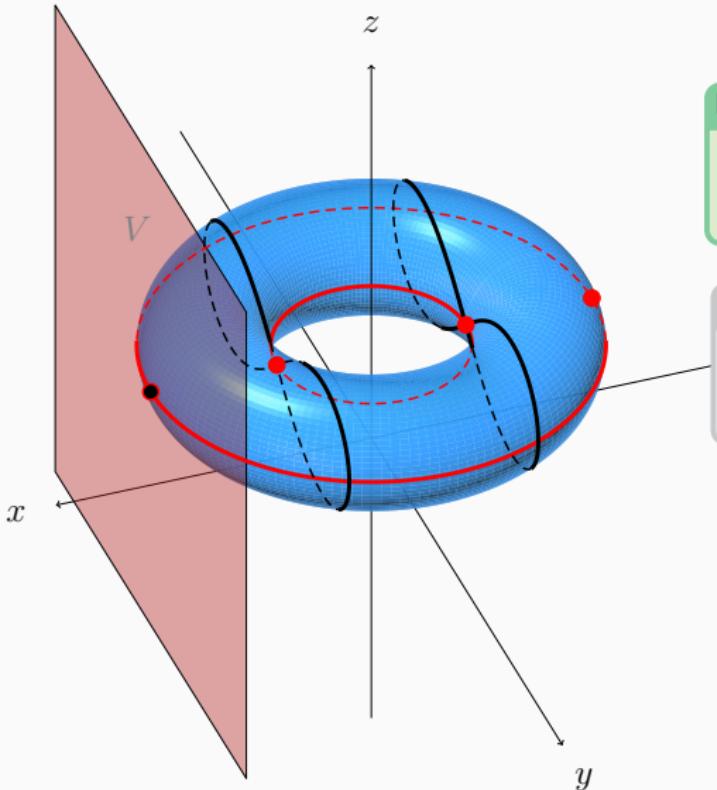
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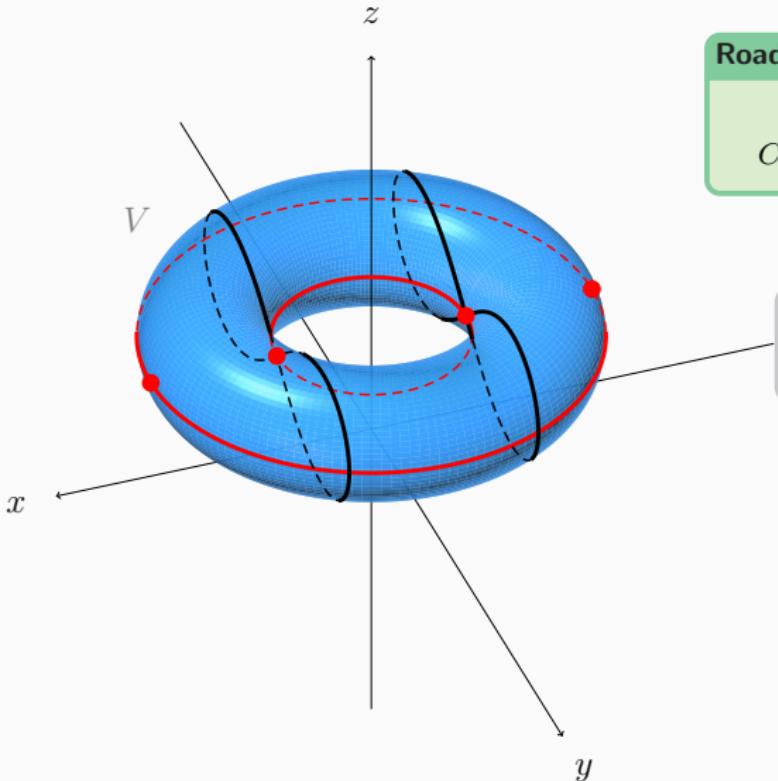
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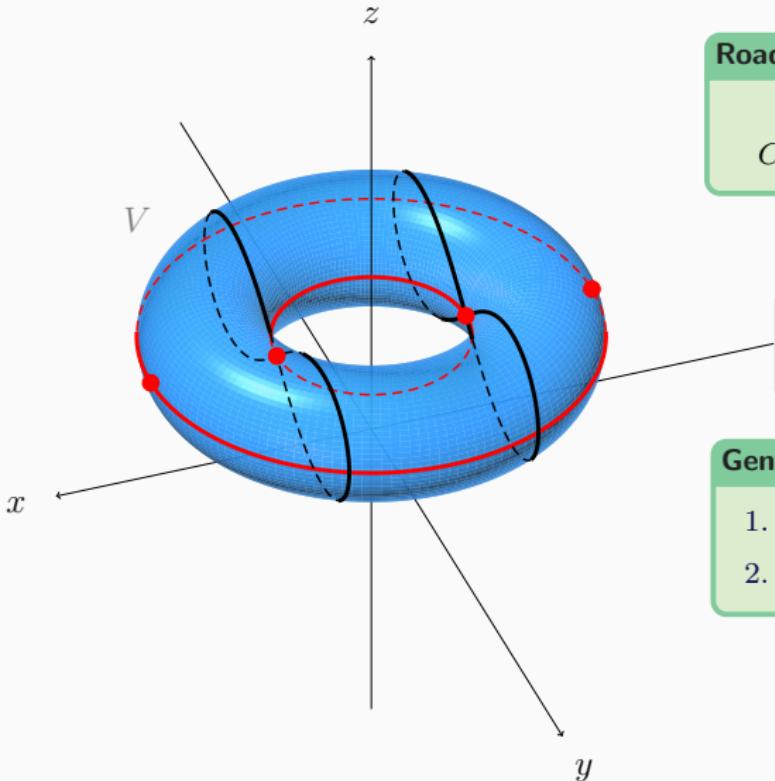


Roadmap property

$\forall C$ connected component,
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$W(\pi_2, V)$ polar variety
 F critical fibers

Canny's strategy



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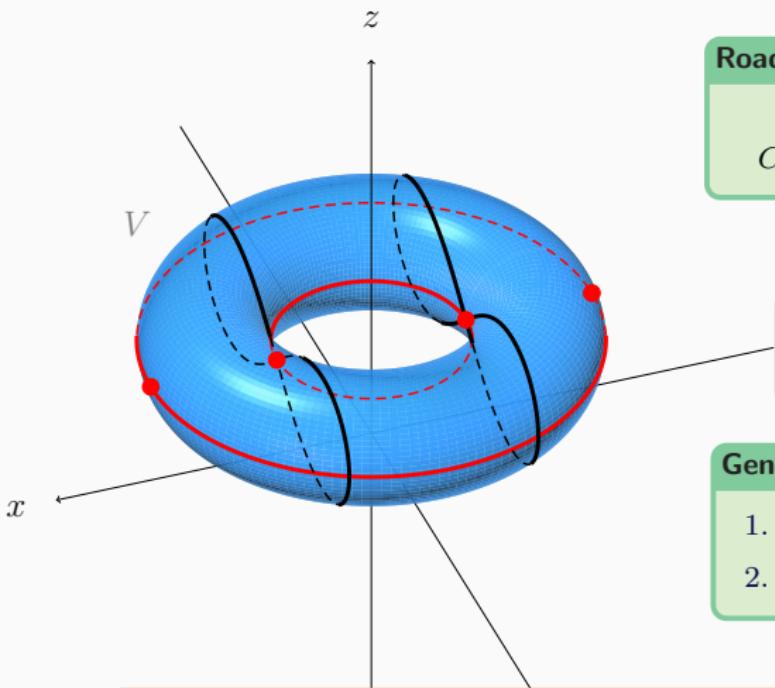
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Genericity assumptions

1. $W(\pi_2, V)$ has dimension 1
2. F has dimension $\dim(V) - 1$

Canny's strategy



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Theorem [Canny, 1988]

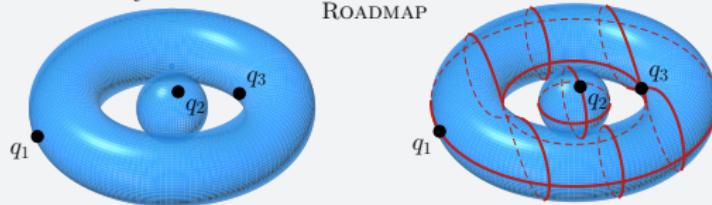
If V is bounded, $W(\pi_2, V) \cup F$ has dimension $\dim(V) - 1$
and satisfies the Roadmap property

Connectivity queries, roadmaps

[Canny, 1988] **Roadmap**: curve capturing the connectivity of the set

Reduction chain

Arbitrary dimension $\xrightarrow{\text{ROADMAP}}$ Dimension 1

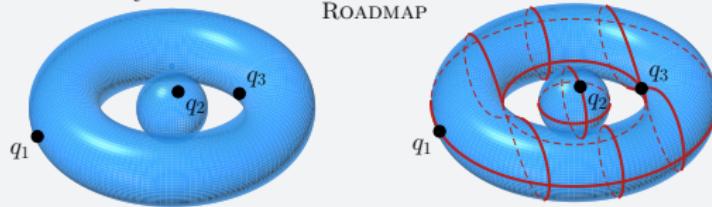


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Towards optimal

1982 $D^{2O(n)}$

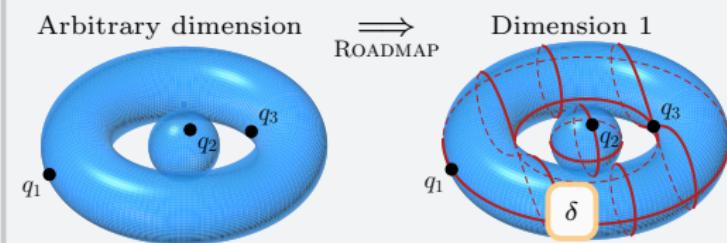
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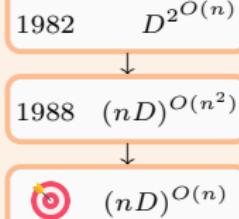
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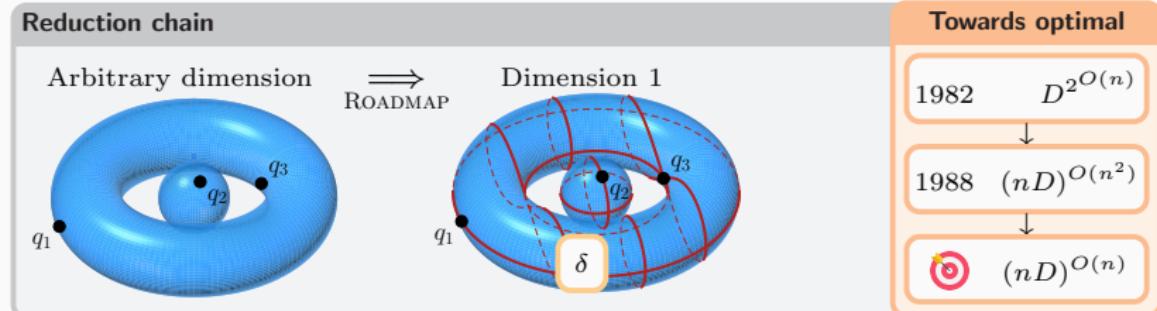
Complexity result (P., Safey El Din, Schost; 2024)

(smooth & unbounded case)

We can compute a roadmap of degree $\delta = (n^2 D)^{2n \log_2 n + O(n)}$ in $O(\delta^3)$ arithmetic ops

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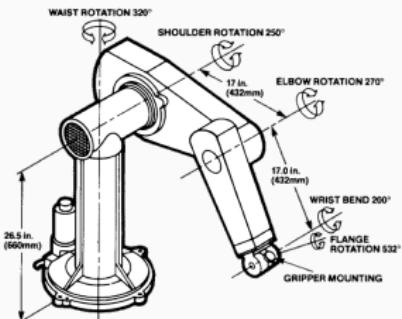
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Computation of a roadmap NEW!

Output: polynomials of max deg **200**
1 polynomial $\approx 20\,000$ coeffs
1 coeff ≈ 3000 digits

Time: **4h10**



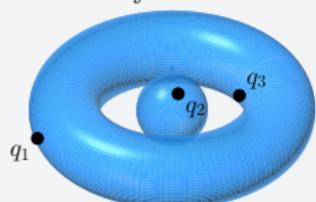
PUMA 560 [Unimation, 1984]

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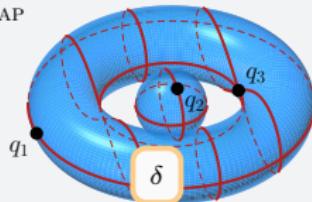
Reduction chain

Arbitrary dimension



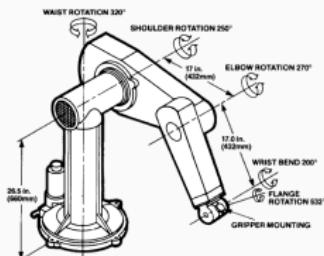
ROADMAP

Dimension 1



ANALYSIS

Graph



PUMA 560 [Unimation, 1984]

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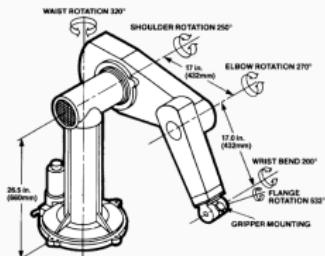
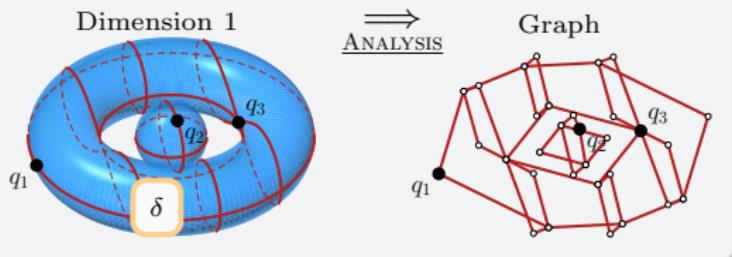


Connectivity queries, roadmaps

[Canny, 1988] **Roadmap**: curve capturing the connectivity of the set

State-of-the-art : geometry

Space	Complexity
\mathbb{R}^2	$O^\sim(\delta^6)$
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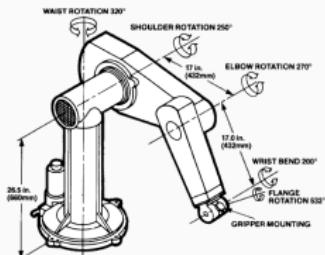
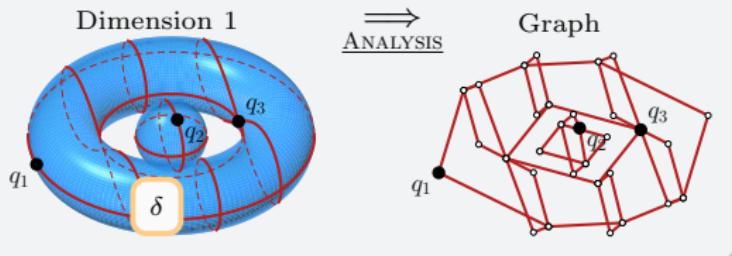


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Avoid computing the complete geometry (Islam, Poteaux, P.; 2023)

We can compute a connectivity graph for a curve of degree δ in $O^\sim(\delta^6)$ bit operations

Data representation and quantitative estimate

Theorem

In a *generic* system of coordinates,
 V is *birational* to a hypersurface of \mathbb{C}^{d+1} through:
 $\pi_{d+1}: (\mathbf{x}_1, \dots, \mathbf{x}_n) \mapsto (\mathbf{x}_1, \dots, \mathbf{x}_{d+1})$



V equidimensional
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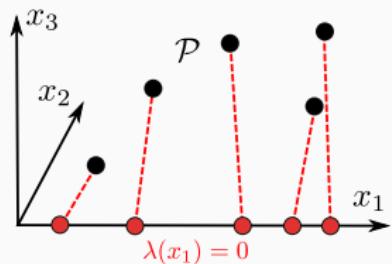


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Zero-dimensional parametrization of $\mathcal{P} \subset \mathbb{C}^n$ finite

$(\lambda, \vartheta_2, \dots, \vartheta_n) \subset \mathbb{Z}[x_1]$ s.t.

$$\mathcal{P} = \left\{ \left(\mathbf{x}_1, \frac{\vartheta_2(\mathbf{x}_1)}{\lambda'(\mathbf{x}_1)}, \dots, \frac{\vartheta_n(\mathbf{x}_1)}{\lambda'(\mathbf{x}_1)} \right) \text{ s.t. } \lambda(\mathbf{x}_1) = 0 \right\}$$



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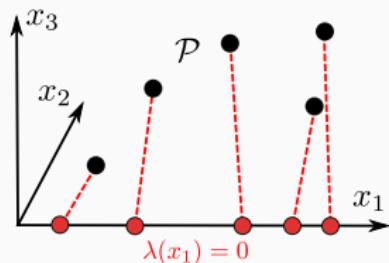


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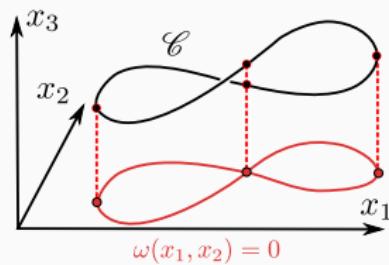
$$\mathcal{P} = \left\{ \left(\mathbf{x}_1, \frac{\vartheta_2(\mathbf{x}_1)}{\lambda'(\mathbf{x}_1)}, \dots, \frac{\vartheta_n(\mathbf{x}_1)}{\lambda'(\mathbf{x}_1)} \right) \text{ s.t. } \lambda(\mathbf{x}_1) = 0 \right\}$$



One-dimensional parametrization of $\mathcal{C} \subset \mathbb{C}^n$ algebraic curve

$(\omega, \rho_3, \dots, \rho_n) \subset \mathbb{Z}[\mathbf{x}_1, \mathbf{x}_2]$ s.t.

$$\mathcal{C} = \overline{\left\{ \left(\mathbf{x}_1, \mathbf{x}_2, \frac{\rho_3(\mathbf{x}_1, \mathbf{x}_2)}{\partial_{x_2} \omega(\mathbf{x}_1, \mathbf{x}_2)}, \dots, \frac{\rho_n(\mathbf{x}_1, \mathbf{x}_2)}{\partial_{x_2} \omega(\mathbf{x}_1, \mathbf{x}_2)} \right) \text{ s.t. } \omega(\mathbf{x}_1, \mathbf{x}_2) = 0 \text{ and } \partial_{x_2} \omega(\mathbf{x}_1, \mathbf{x}_2) \neq 0 \right\}}^Z$$



Data representation and quantitative estimate

Theorem

In a *generic* system of coordinates,
 V is *birational* to a hypersurface of \mathbb{C}^{d+1} through:
 $\pi_{d+1}: (\mathbf{x}_1, \dots, \mathbf{x}_n) \mapsto (\mathbf{x}_1, \dots, \mathbf{x}_{d+1})$

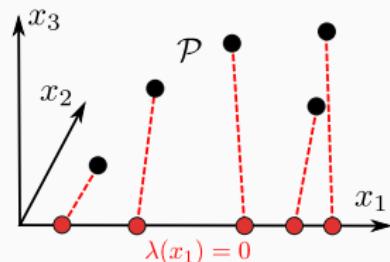


V equidimensional
of dimension d

Zero-dimensional parametrization of $\mathcal{P} \subset \mathbb{C}^n$ finite

$(\lambda, \vartheta_2, \dots, \vartheta_n) \subset \mathbb{Z}[x_1]$ s.t.

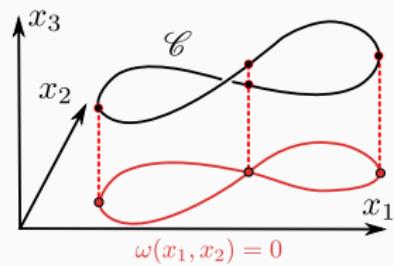
$$\mathcal{P} = \left\{ \left(\mathbf{x}_1, \frac{\vartheta_2(\mathbf{x}_1)}{\lambda'(\mathbf{x}_1)}, \dots, \frac{\vartheta_n(\mathbf{x}_1)}{\lambda'(\mathbf{x}_1)} \right) \text{ s.t. } \lambda(\mathbf{x}_1) = 0 \right\}$$



One-dimensional parametrization of $\mathcal{C} \subset \mathbb{C}^n$ algebraic curve

$(\omega, \rho_3, \dots, \rho_n) \subset \mathbb{Z}[x_1, x_2]$ s.t.

$$\mathcal{C} = \left\{ \left(\mathbf{x}_1, \mathbf{x}_2, \frac{\rho_3(\mathbf{x}_1, \mathbf{x}_2)}{\partial_{x_2} \omega(\mathbf{x}_1, \mathbf{x}_2)}, \dots, \frac{\rho_n(\mathbf{x}_1, \mathbf{x}_2)}{\partial_{x_2} \omega(\mathbf{x}_1, \mathbf{x}_2)} \right) \middle| \begin{array}{l} \text{s.t. } \omega(\mathbf{x}_1, \mathbf{x}_2) = 0 \quad \text{and} \quad \partial_{x_2} \omega(\mathbf{x}_1, \mathbf{x}_2) \neq 0 \end{array} \right\}^Z$$



Magnitude of a polynomial

$f \in \mathbb{Z}[x_1, \dots, x_n]$ has *magnitude* (δ, τ) if
 $\deg(f) \leq \delta \quad \text{and} \quad |\text{coeffs}(f)| \leq 2^\tau$

Soft-O notation

$$\tilde{O}(N) = O(N \log(N)^\alpha)$$

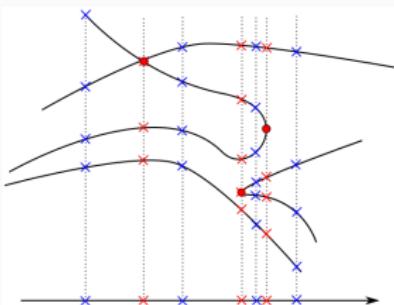
Results

Data

- $\mathcal{R} \subset \mathbb{Z}[x_1, x_2]$ of magnitude (δ, τ) , encoding an algebraic curve $\mathcal{C} \subset \mathbb{C}^n$;
- $\mathcal{P} \subset \mathbb{Z}[x_1]$ of magnitude (δ, τ) , encoding a finite $\mathcal{P} \subset \mathcal{C}$;

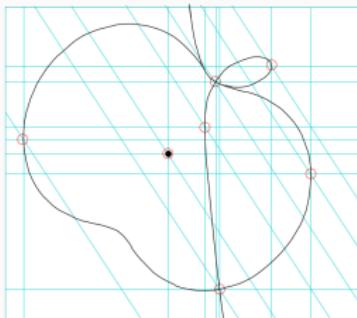
Computing topology

Ambient dimension	Bit complexity	Reference
$n = 2$	$\tilde{O}(\delta^5(\delta + \tau))$	[Kobel, Sagraloff; '15] [D.Diatta, S.Diatta, Rouiller, Roy, Sagraloff; '22]



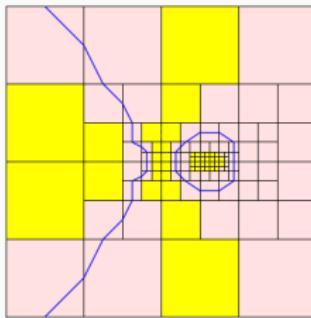
Cylindrical Algebraic Decomposition

[Collins, '75] [Kerber, Sagraloff; '12]



Multiple projections

[Seidel, Wolpert; '05]



Subdivision

[Burr, Choi, Galehouse, Yap; '05]

Results

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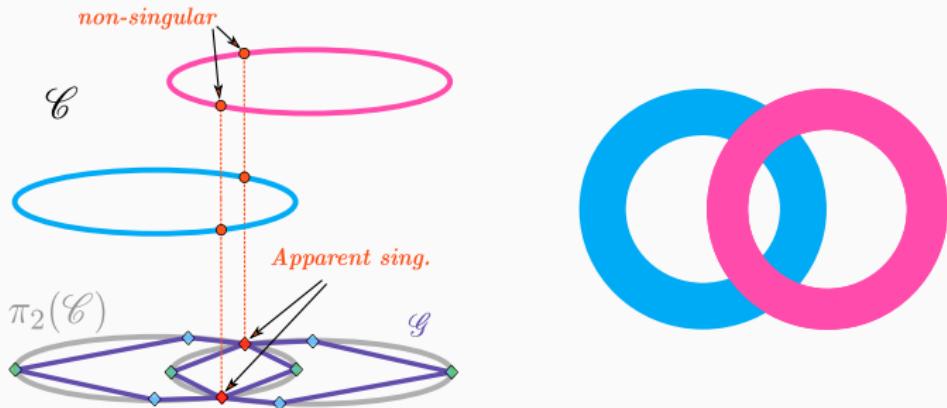
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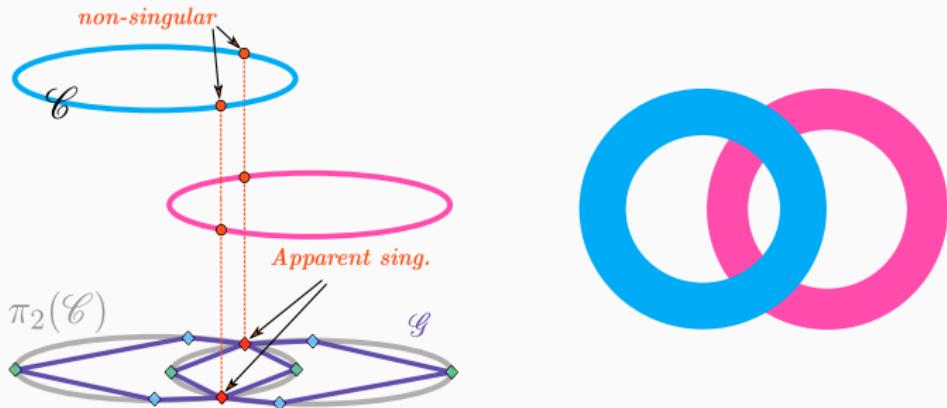
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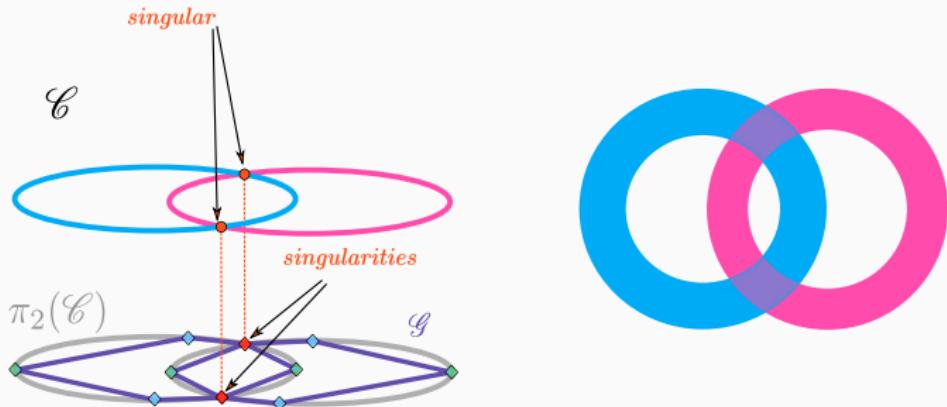
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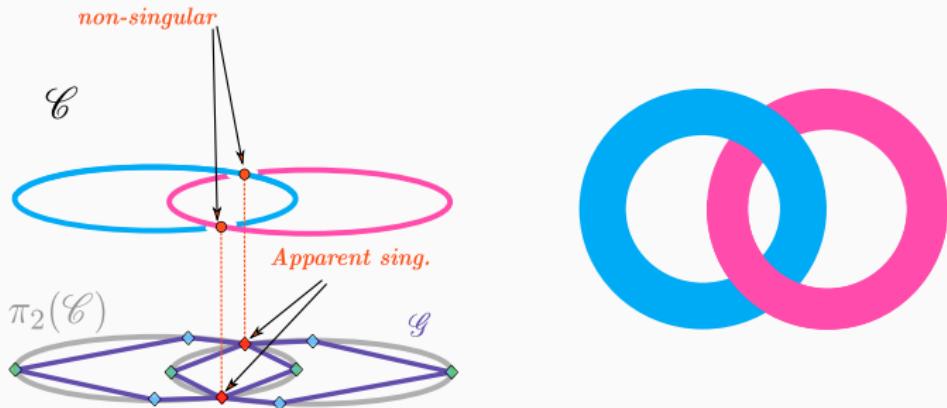
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Computing connectivity - Main Result NEW!

Ambient dimension	Bit complexity	Reference
$n \geq 2$	$\tilde{O}(\delta^5(\delta + \tau))$	[Islam, Poteaux, P.; 2023]

Avoid computation of the complete topology!

Algorithm

Input

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- $\mathcal{P} \subset \mathbb{Z}[x_1]$ of magnitude (δ, τ) , encoding a finite $\mathcal{P} \subset \mathcal{C}$;
- [some assumptions]

Output

A connectivity graph of $\mathcal{C} \cap \mathbb{R}^n$ capturing the points in $\mathcal{P} \cap \mathbb{R}^n$.

1. $\mathcal{D}, \mathcal{Q} \leftarrow \text{Proj2D}(\mathcal{R}), \text{Proj2D}(\mathcal{P})$;
2. $\mathcal{G} \leftarrow \text{Topo2D}(\mathcal{D}, \mathcal{Q})$;
3. $\mathcal{Q}_{\text{app}} \leftarrow \text{ApparentSingularities}(\mathcal{R})$;
4. $\mathcal{G}' \leftarrow \text{NodeResolution}(\mathcal{G}, \mathcal{Q}_{\text{app}})$;
5. return $\text{ConnectGraph}(\mathcal{Q}, \mathcal{G}')$;

\mathcal{C}



Algorithm

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\mathcal{C}



$\pi_2(\mathcal{C})$



Algorithm

Input

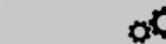
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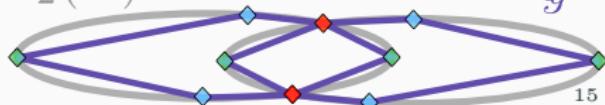


- ↔ resultants
- ↔ \mathbb{R} -root isolation
- ↳ univariate
- ↳ bivariate triangular

Planar topology

$\tilde{O}(\delta^5(\delta + \tau))$

$\pi_2(\mathcal{C})$



Algorithm

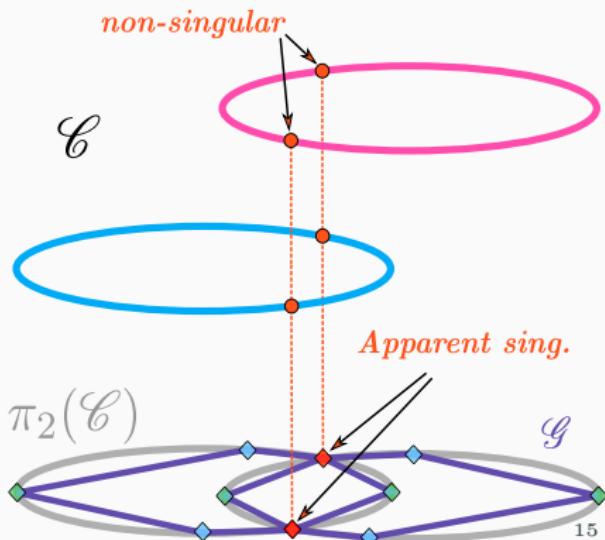
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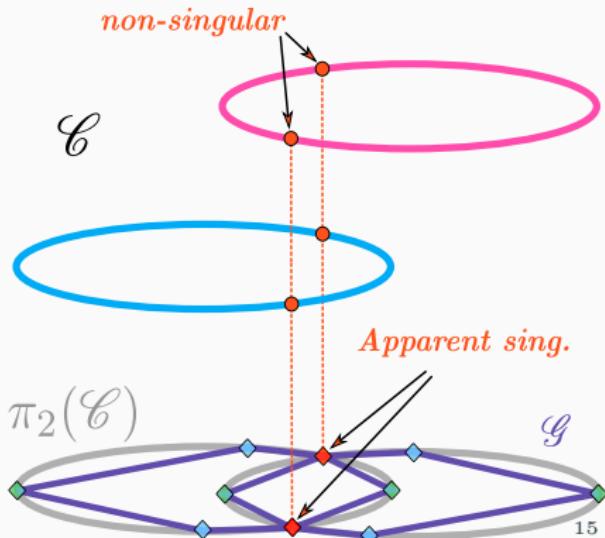
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Apparent sing.
identification

$\tilde{O}(\delta^5(\delta + \tau))$



Algorithm

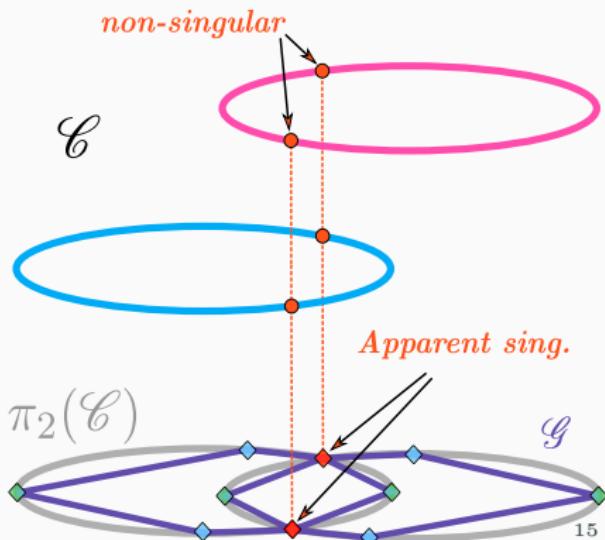
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Algorithm

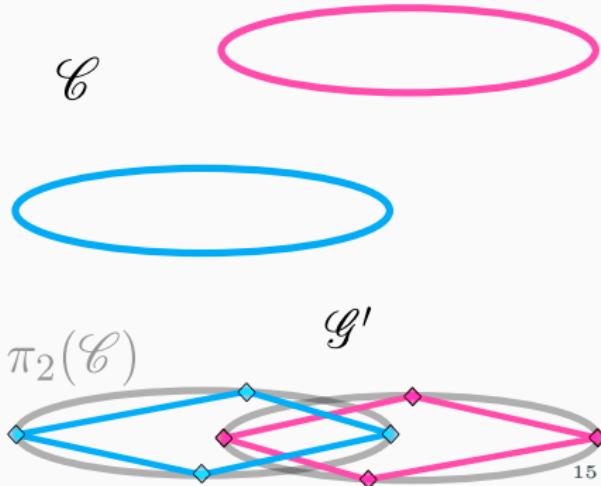
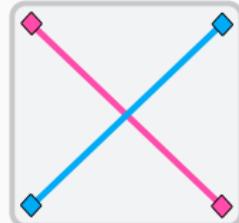
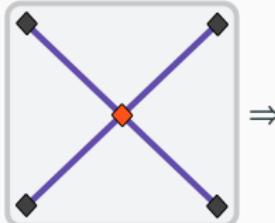
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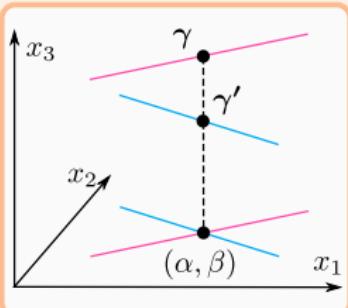
Apparent singularities: key idea

- $\mathcal{R} = (\omega, \rho_3, \dots, \rho_n) \subset \mathbb{Z}[x_1, x_2]$ encoding $\mathcal{C} \subset \mathbb{C}^n$ in generic coordinates;
- $\mathcal{A}(x_1, x_2) = \partial_{x_2}^2 \omega \cdot \partial_{x_1} \rho_3 - \partial_{x_1 x_2}^2 \omega \cdot \partial_{x_2} \rho_3 \in \mathbb{Z}[x_1, x_2]$

Generic apparent singularities

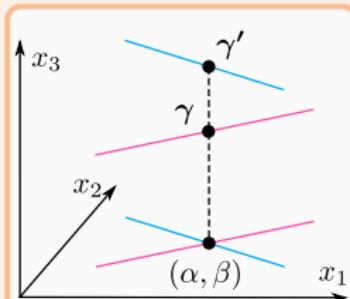
NEW!

Generic projection only introduce finitely many nodes:



Below

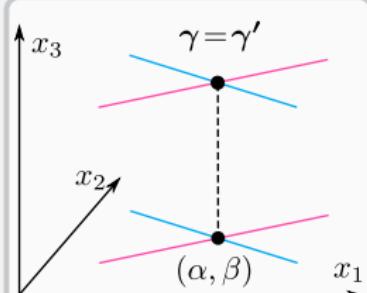
\neq



Above

Space singularities

Spatial nodes project as:



Same

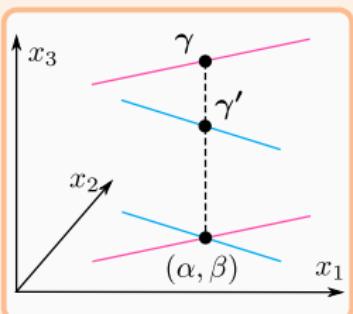
Apparent singularities: key idea

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Generic apparent singularities

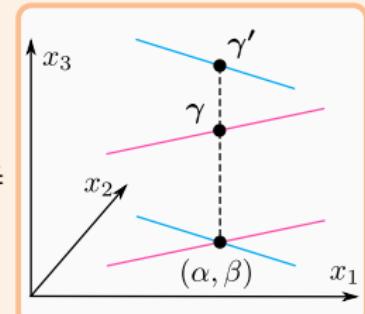
NEW!

Generic projection only introduce finitely many nodes:



Below

$$\mathcal{A}(\alpha, \beta) > 0$$

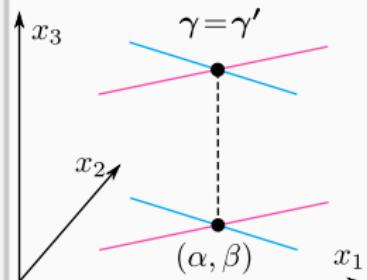


Above

$$\mathcal{A}(\alpha, \beta) < 0$$

Space singularities

Spatial nodes project as:



Same

$$\mathcal{A}(\alpha, \beta) = 0$$

Proposition - Generalization of [El Kahoui; '08]

If (α, β) is a node then $\mathcal{A}(\alpha, \beta) = \gamma - \gamma'$

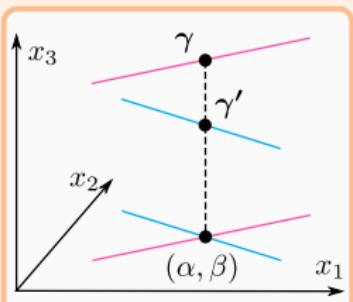
Apparent singularities: key idea

- $\mathcal{R} = (\omega, \textcolor{brown}{\rho_3}, \dots, \rho_n) \subset \mathbb{Z}[x_1, x_2]$ encoding $\mathcal{C} \subset \mathbb{C}^n$ in generic coordinates;
- $\mathcal{A}(x_1, x_2) = \partial_{x_2}^2 \omega \cdot \partial_{x_1} \textcolor{brown}{\rho_3} - \partial_{x_1 x_2}^2 \omega \cdot \partial_{x_2} \textcolor{brown}{\rho_3} \in \mathbb{Z}[x_1, x_2]$

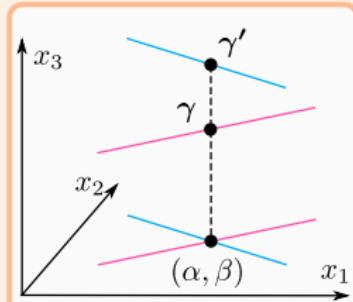
Generic apparent singularities

NEW!

Generic projection only introduce finitely many nodes:



Apparent singularity

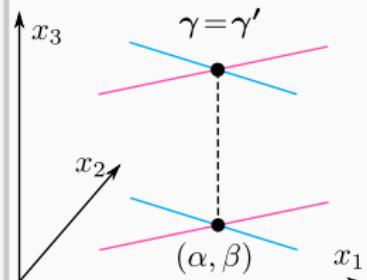


Apparent singularity

$$\mathcal{A}(\alpha, \beta) \neq 0$$

Space singularities

Spatial nodes project as:



Actual singularity

$$\mathcal{A}(\alpha, \beta) = 0$$

Proposition - Generalization of [El Kahoui; '08]

If (α, β) is a node then $\mathcal{A}(\alpha, \beta) = \gamma - \gamma'$

Algorithm

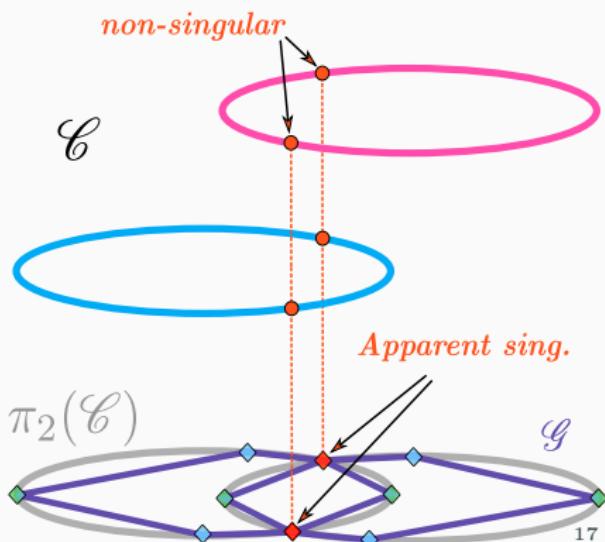
Input

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- \mathcal{C} satisfies genericity assumptions w.r.t. \mathcal{P}

Output

A connectivity graph of $\mathcal{C} \cap \mathbb{R}^n$ capturing the points in $\mathcal{P} \cap \mathbb{R}^n$.

1. $\mathcal{D}, \mathcal{Q} \leftarrow \text{Proj2D}(\mathcal{R}), \text{Proj2D}(\mathcal{P})$;
2. $\mathcal{G} \leftarrow \text{Topo2D}(\mathcal{D}, \mathcal{Q})$;
3. $\mathcal{Q}_{\text{app}} \leftarrow \text{ApparentSingularities}(\mathcal{R})$;
4. $\mathcal{G}' \leftarrow \text{NodeResolution}(\mathcal{G}, \mathcal{Q}_{\text{app}})$;
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Algorithm

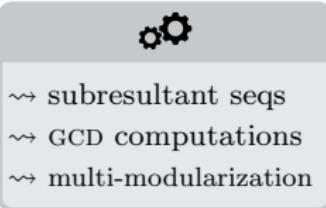
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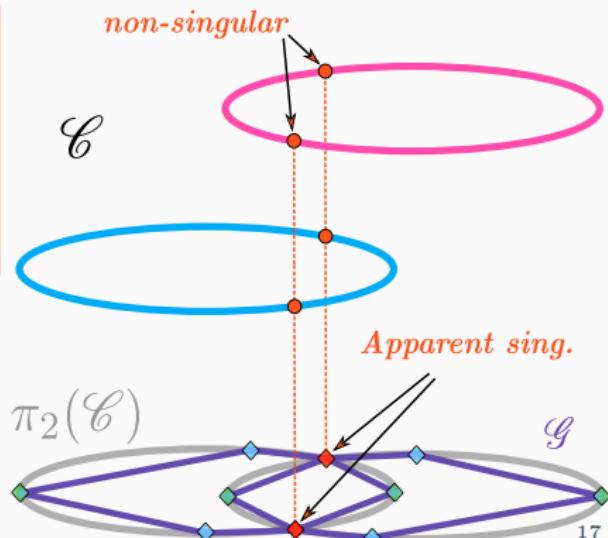
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Apparent sing.
identification
 $\tilde{O}(\delta^5(\delta + \tau))$



Algorithm

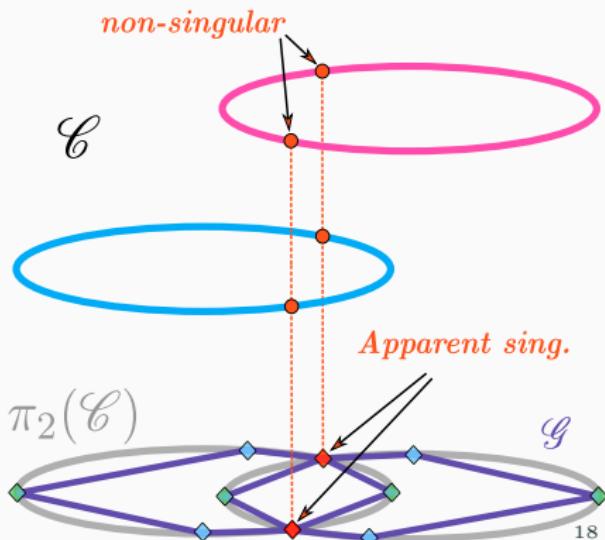
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Algorithm

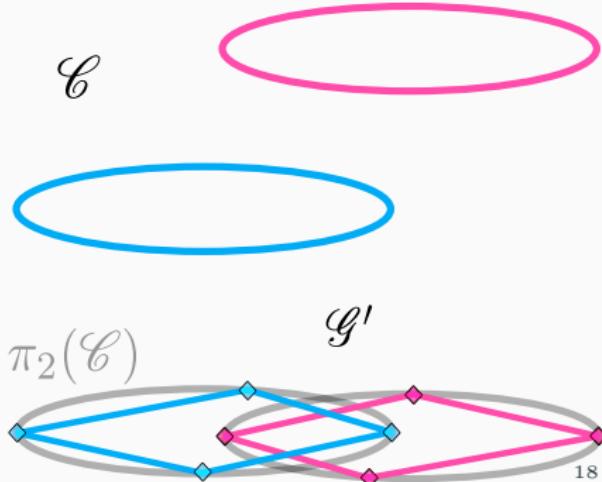
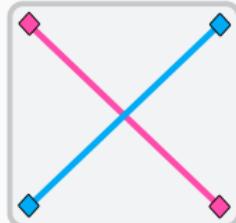
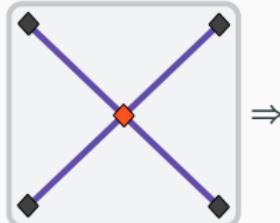
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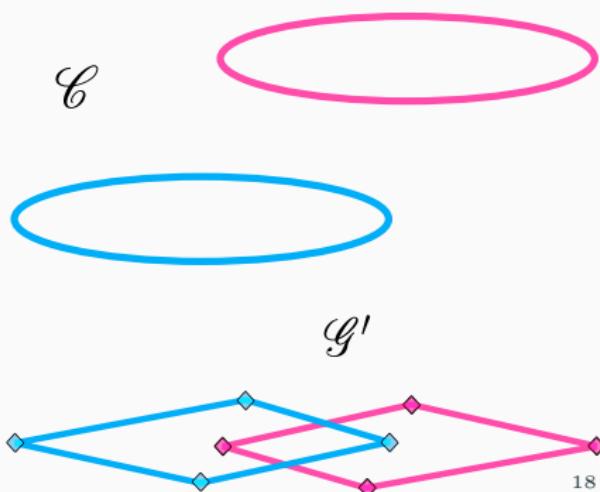
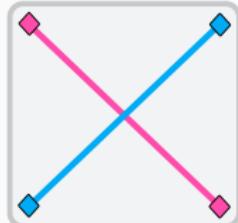
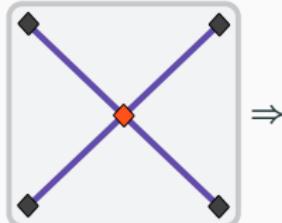
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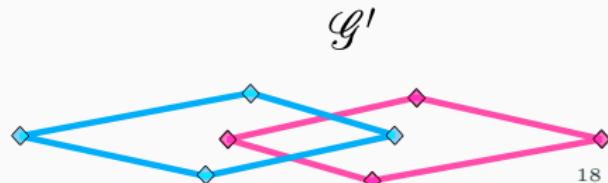
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\mathcal{C}



Overall Complexity

$$\tilde{O}(\delta^5(\delta + \tau))$$

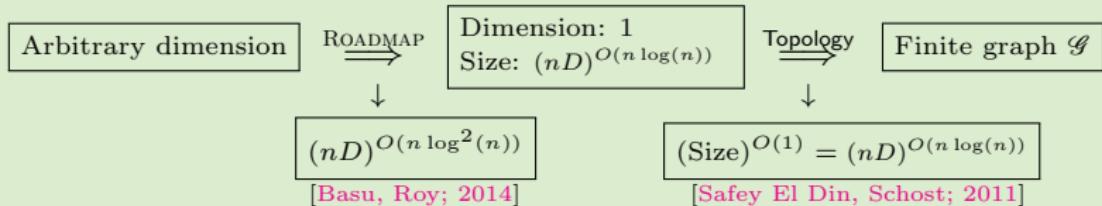


Summary

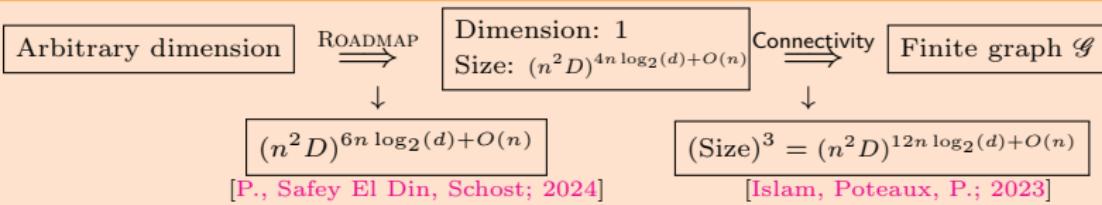
Input

Polynomials in $\mathbb{Q}[x_1, \dots, x_n]$ of max degree D defining a smooth algebraic set of dim. d

Connectivity reduction process - before



Connectivity reduction process - now



📄 Computing roadmaps in unbounded smooth real algebraic sets I: connectivity results, 2024
with M. Safey El Din and É. Schost

📄 Computing roadmaps in unbounded smooth real algebraic sets II: algorithm and complexity, 2024
with M. Safey El Din and É. Schost

📄 Algorithm for connectivity queries on real algebraic curves, 2023
with Md N. Islam and A. Poteaux

Perspectives

Algorithms

Roadmap algorithms:

- | Adapt the algorithms to structured systems: quadratic case (J.A.K.Elliott, M.Safey El Din, É.Schost)
- | Reduce the size of the roadmap by taking fewer fibers (M.Safey El Din, É.Schost)
- | Generalize the connectivity result to semi-algebraic sets
- ↓ Design optimal roadmap algorithms with complexity exponential in $O(n)$

Connectivity of s.a. curves:

- | Obtain a deterministic version of the algorithm (F.Bréhard, A.Poteaux)
- | Adapt to algebraic curves given as union (A.Poteaux)
- | Generalize to semi-algebraic curves
- ↓ Investigate the connectivity of plane curves

Applications

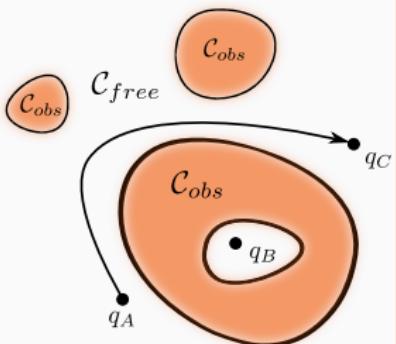
- | Analyze challenging class of robots (D.Salunkhe, P.Wenger)
- | Algorithms for rigidity and program verification problems (E.Bayarmagnai, F.Mohammadi)
- ↓ Obtain practical version of modern roadmap algorithms

Software

- | Connectivity of curves: subresultant/GCD computations deg ~ 100 (now) $\rightarrow \sim 200$ (target)
- | Build a Julia library for computational real algebraic geometry (C.Eder, R.Mohr)
- ↓ Implement a ready-to-use toolbox for roboticians

Motion planning with parameters and obstacles

Obstacles/Self-intersections

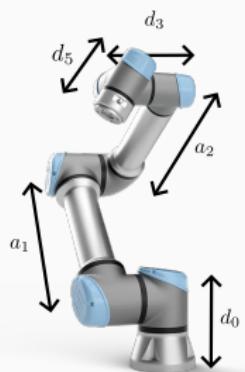


Inequalities \Rightarrow semi-algebraic

Cuspidality

Parametric classification
⇒ design guidelines

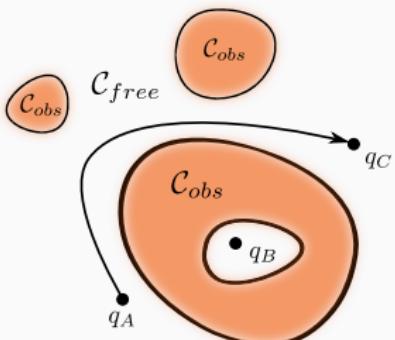
Family of robots



Parametric systems

Motion planning with parameters and obstacles

Obstacles/Self-intersections



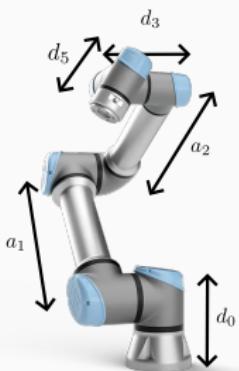
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Subroutines
efficient in **practice**

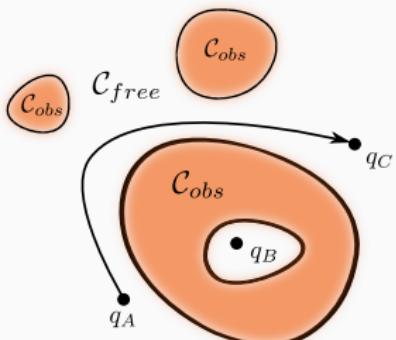
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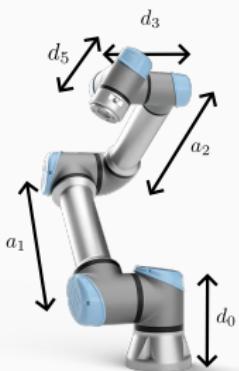


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Family of robots



Parametric systems

Roadmaps : structured semi-algebraic case

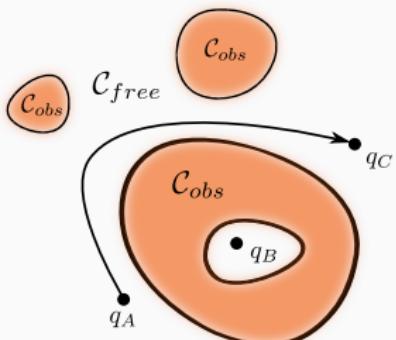
$$f_1 = \dots = f_p = 0, \quad g_1 > 0, \dots, g_s > 0$$

Structures: quadratic, multi/quasi-homogeneous

Current: $s^n(nD)^{O(n^2)}$ [Basu, Pollack, Roy ; 2000]
Goal: $s^{\textcolor{red}{n}}(n^2 D)^{4n \log_2 n + O(n)}$

Motion planning with parameters and obstacles

Obstacles/Self-intersections



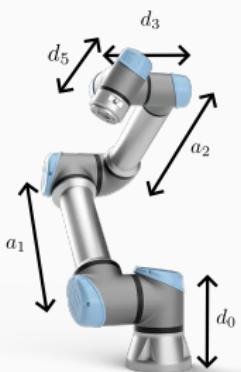
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Parametric systems

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Implementation

AlgebraicSolving.jl
 \swarrow msolve \searrow
R Isolation elimination

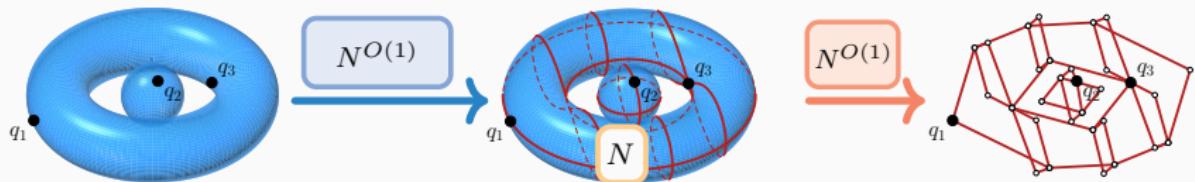


Geometry of curves

New situation

Analysing the road map becomes the most costly step

Dimension reduction

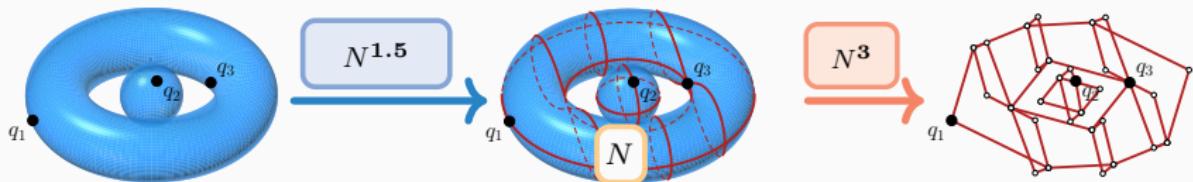


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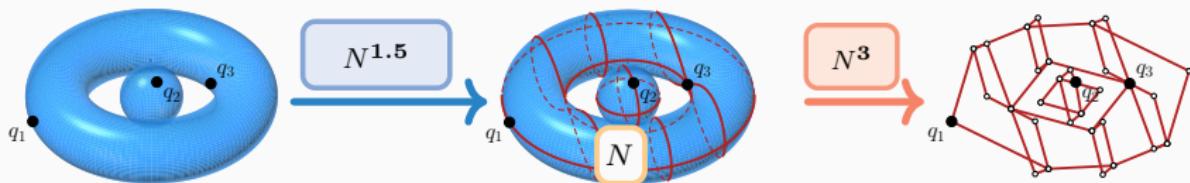


Geometry of curves

New situation

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Bottleneck

Algebraic elimination
(resultant)



with G. VILLARD

Structures des cartes routières

Algebraic

Avoid change of variables :



with F.BRÉHARD, A.POTEAUX



Geometric

