

# The SDP-exact region in quadratic optimization

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# Quadratic programming

Consider a quadratically constrained quadratic program:

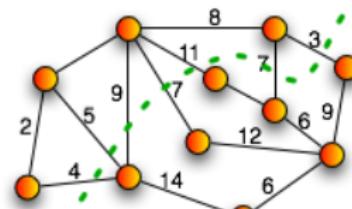
$$\begin{aligned} \min_{x \in \mathbb{R}^n} \quad & g(x) := x^T C x + 2c^T x \\ & f_i(x) := x^T A x + 2a_i^T x + \alpha_i = 0, \quad i = 1, \dots, m \end{aligned} \tag{QP}$$

QP's are nonconvex and NP-hard. We consider convex relaxations based on *semidefinite programming* (SDP).

## Example (Max-Cut)

Given a graph  $G = (V, E)$  with weights  $c_{ij}$ ,

$$\begin{aligned} \min_x \quad & \sum_{ij \in E} c_{ij} x_i x_j \\ \text{s.t.} \quad & x_i^2 = 1 \text{ for } i \in V \end{aligned}$$



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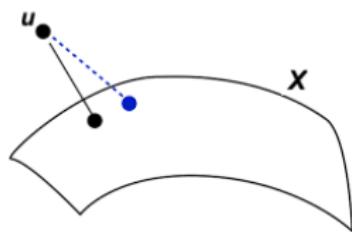
$$\begin{aligned} \min_{x \in \mathbb{R}^n} \quad & g(x) := x^T Cx + 2c^T x \\ & f_i(x) := x^T Ax + 2a_i^T x + \alpha_i = 0, \quad i = 1, \dots, m \end{aligned} \tag{QP}$$

QP's are nonconvex and NP-hard. We consider convex relaxations based on *semidefinite programming* (SDP).

## Example (Nearest point to quadratic variety)

Given a variety  $X \subset \mathbb{R}^n$  defined by quadrics, and a point  $u \in \mathbb{R}^n$ ,

$$\begin{aligned} \min_x \quad & \|x - u\|^2 \\ \text{s.t.} \quad & x \in X \end{aligned}$$

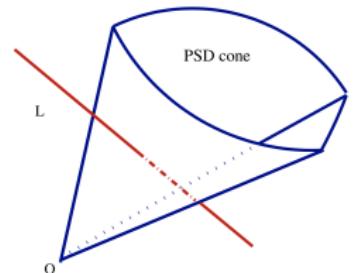


# Semidefinite Programming

A semidefinite program (SDP) is

$$\begin{aligned} \min_{X \in \mathbb{S}^d} \quad & \mathcal{C} \bullet X \\ \text{s.t.} \quad & \mathcal{A}_i \bullet X = b_i \text{ for } i \in [m] \\ & X \succeq 0 \end{aligned}$$

where  $\mathcal{C}, \mathcal{A}_i \in \mathbb{S}^d$ ,  $b_i \in \mathbb{R}$  are given.



## SDP relaxation

Consider the quadratic program

$$\begin{aligned} \min_{x \in \mathbb{R}^n} \quad & g(x) := x^T C x + 2c^T x \\ & f_i(x) := x^T A_i x + 2a_i^T x + \alpha_i = 0, \text{ for } i \in [m] \end{aligned} \tag{QP}$$

Let the following matrices in  $\mathbb{S}^{n+1}$

$$\mathcal{C} := \begin{bmatrix} 0 & c^T \\ c & C \end{bmatrix}, \quad \mathcal{A}_0 := \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \quad \mathcal{A}_i := \begin{bmatrix} \alpha_i & a_i^T \\ a_i & A_i \end{bmatrix}$$

The QP can be equivalently written as

$$\begin{aligned} \min_{x \in \mathbb{R}^n, X \in \mathbb{S}^{n+1}} \quad & \mathcal{C} \bullet X \\ & \mathcal{A}_0 \bullet X = 1 \\ & \mathcal{A}_i \bullet X = 0 \text{ for } i \in [m] \end{aligned}$$

$$X = \begin{bmatrix} 1 & x^T \end{bmatrix} \begin{bmatrix} 1 \\ x \end{bmatrix}$$

## SDP relaxation

Consider the quadratic program

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The **SDP relaxation** of QP is

$$\begin{aligned} \min_{X \in \mathbb{S}^n} \quad & C \bullet X \\ & \mathcal{A}_0 \bullet X = 1 \\ & \mathcal{A}_i \bullet X = 0 \text{ for } i \in [m] \\ & X \succeq 0 \end{aligned} \tag{SDP}$$

The relaxation is **exact** if the optimal solution is rank-one.

# The SDP-exact region

$$\min_{x \in \mathbb{R}^n} \quad g(x) \quad \text{s.t.} \quad f_i(x) = 0, \text{ for } i \in [m] \quad (QP)$$

Let  $\mathbf{f} = (f_1, \dots, f_m)$  be fixed quadrics. The **SDP-exact region** is

$$\mathcal{R}_{\mathbf{f}} = \{ g \in \mathbb{R}[x]_{\leq 2} : \text{the QP is solved exactly by its SDP relaxation} \}.$$

The SDP-exact region is a semialgebraic set in  $\mathbb{R}[x]_{\leq 2} \cong \mathbb{S}^{n+1}$ .

## Problem

Given  $\mathbf{f}$ , compute/characterize the SDP-exact region. If the computation is intractable, we would like to understand its degree.

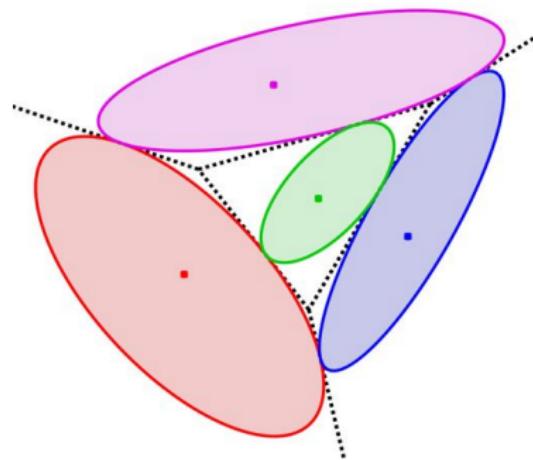
## Nearest point to four points

For  $u \in \mathbb{R}^n$ , consider

$$\min_x \|x - u\|^2 \quad s.t. \quad x \in X$$

$X := \{ \text{zero set of two polynomials in two variables} \}$

The SDP-exact region is the union of four conics, each contained in a *Voronoi cell*.



## Max-Cut

Given a graph  $G = (V, E)$  with  $V = [n]$  and weights  $c_{ij}$

$$\begin{aligned} \min_x \quad & \sum_{ij \in E} c_{ij} x_i x_j \\ \text{s.t.} \quad & x_i^2 = 1 \text{ for } i \in [n] \end{aligned}$$

Let  $S := \{C : \mathcal{L}(C) \succeq 0\}$  be the spectrahedron of the *Laplacian* matrix

$$\mathcal{L}(C) = \begin{pmatrix} -\sum_{j \neq 1} c_{1j} & c_{12} & c_{13} & \cdots & c_{1n} \\ c_{12} & -\sum_{j \neq 2} c_{2j} & c_{23} & \cdots & c_{2n} \\ c_{13} & c_{23} & -\sum_{j \neq 3} c_{3j} & \cdots & c_{3n} \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ c_{1n} & c_{2n} & c_{3n} & \cdots & -\sum_{j \neq n} c_{jn} \end{pmatrix}.$$

The SDP-exact region consists of  $2^{n-1}$  copies of  $S$ .

## Rank-one region in SDP

Consider the SDP

$$\min_{X \in \mathbb{S}^d} \quad \mathcal{C} \bullet X \quad \text{s.t.} \quad \mathcal{A}_i \bullet X = b_i \text{ for } i \in [m], \quad X \succeq 0 \quad (\text{P})$$

Let  $\mathcal{A}, b$  be fixed. We define the **rank-one region** as

$$\mathcal{R}_{\mathcal{A}, b} = \{ C \in \mathbb{S}^d : \text{ there is an optimal solution of rank one } \}.$$

This is a generalization of the SDP-exact region to arbitrary SDP's.

## Rank-one region: primal-dual characterization

Consider the primal-dual pair

$$\min_{X \in \mathbb{S}^d} \quad \mathcal{C} \bullet X \quad \text{s.t.} \quad \mathcal{A}_i \bullet X = b_i \text{ for } i \in [m], \quad X \succeq 0 \quad (\text{P})$$

$$\max_{\lambda \in \mathbb{R}^\ell, Y \in \mathbb{S}^d} \quad b^T \lambda \quad \text{s.t.} \quad Y = \mathcal{C} - \sum_i \lambda_i \mathcal{A}_i \quad Y \succeq 0 \quad (\text{D})$$

The rank-one region is defined by the *critical equations*

$$\mathcal{R}_{\mathcal{A}, b} = \left\{ C \in \mathbb{S}^d : \begin{array}{l} \mathcal{A}_i \bullet X = b_i, \quad Y = \mathcal{C} - \sum_i \lambda_i \mathcal{A}_i, \quad X \cdot Y = 0, \\ X \succeq 0, \quad Y \succeq 0, \quad \text{rank} X = 1, \quad \text{rank} Y = d-1 \end{array} \right\}$$

## Rank-one region: primal-dual characterization

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**Computation:** For generic  $\mathcal{A}, b$ , the boundary hypersurface  $\partial_{\text{alg}} \mathcal{R}_{\mathcal{A}, b}$  can be computed by eliminating  $X, Y, \lambda$  in the equations.

$$\mathcal{A}_i \bullet X = b_i, \quad Y = \mathcal{C} - \sum_i \lambda_i \mathcal{A}_i, \quad X \cdot Y = 0, \quad \text{minors}_2(X) = 0$$

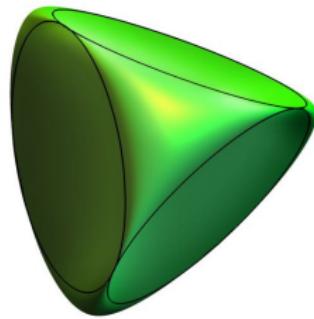
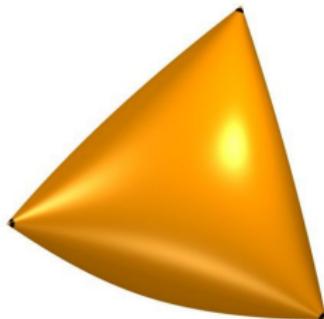
## Rank-one region in MaxCut SDP

$$\min_{X \in \mathbb{S}^d} \quad \mathcal{C} \bullet X \quad \text{s.t.} \quad X_{ii} = 1 \text{ for } i \in [n], \quad X \succeq 0 \quad (\text{P})$$

$$\max_{\lambda \in \mathbb{R}^\ell} \quad \sum_i \lambda_i \quad \text{s.t.} \quad Y = \mathcal{C} - \text{Diag}(\lambda) \quad Y \succeq 0 \quad (\text{D})$$

The rank-one region has  $2^{n-1}$  pieces, one for each  $X = xx^T$ ,  $x \in \{\pm 1\}^n$ .

$$\mathcal{R}_{\mathcal{A}, b} = \bigcup_{X \text{ vertex}} \left\{ C \in \mathbb{S}^d : X \cdot Y = 0, \quad Y = \mathcal{C} - \text{Diag}(\lambda), \quad Y \succeq 0, \right\}$$



# Generic degree of the rank-one region

## Theorem

Consider an SDP with generic  $\mathcal{A}$  and  $b$ . The boundary degree  $\partial_{\text{alg}} \mathcal{R}_{\mathcal{A}, b}$  is

$$2^{\ell-1}(d-1)\binom{d}{\ell} - 2^\ell \binom{d}{\ell+1} \quad \text{for } 3 \leq \ell \leq d$$

If  $\ell = 2$  then  $\mathcal{R}_{\mathcal{A}, b}$  is dense and the degree is  $\binom{d+1}{3}$ .

The proof follows the strategy by [Nie-Ranestad-Sturmfels] to compute the algebraic degree of SDP.

Algebraic degree of SDP					
$\ell \backslash d$	3	4	5	6	7
2	6	12	20	30	42
3	4	16	40	80	140
4		8	40	120	280
5			16	96	336
6				32	224

Rank-one boundary degrees					
$\ell \backslash d$	3	4	5	6	7
2	4	10	20	35	66
3	8	40	120	280	560
4		24	144	504	1344
5			64	448	1792
6				160	1280

## Generic degree of the SDP-exact region

$$\begin{aligned} \min_{x \in \mathbb{R}^n} \quad & g(x) := x^T C x + 2c^T x \\ & f_i(x) := x^T A_i x + 2a_i^T x + \alpha_i = 0, \quad i = 1, \dots, m \end{aligned} \tag{QP}$$

Consider the Hessian of the Lagrangian

$$H(\lambda) = \nabla^2 L(x, \lambda) = C - \sum_i \lambda_i A_i \in \mathbb{S}^n$$

The SDP-exact region is

$$\mathcal{R}_f = \bigcup_{x \in V_f} \left\{ (c, C) : c - \sum_i \lambda_i a_i + H(\lambda)x = 0, \quad H(\lambda) \succeq 0 \right\}$$

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## Theorem

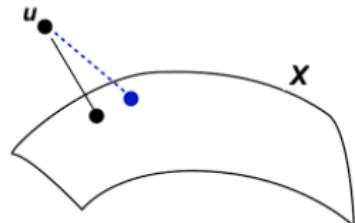
Consider a QP with generic equations  $f$ . The boundary degree  $\partial_{\text{alg}} \mathcal{R}_f$  is

$$2^m \left( n \binom{n}{m} - \binom{n}{m+1} \right).$$

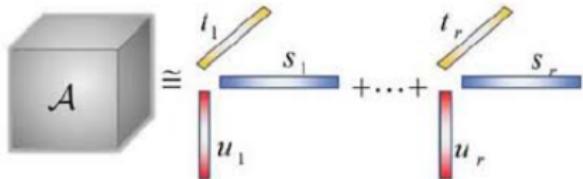
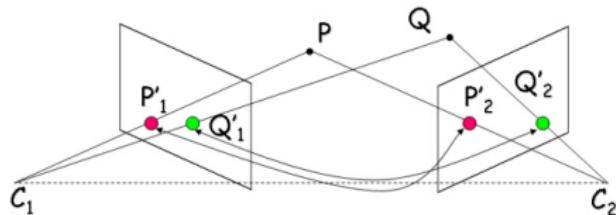
# Nearest point problems

Let  $X \subset \mathbb{R}^n$  quadratic and  $u \in \mathbb{R}^n$ .

$$\begin{aligned} \min_x \quad & \|x - u\|^2 \\ \text{s.t.} \quad & x \in X \end{aligned}$$



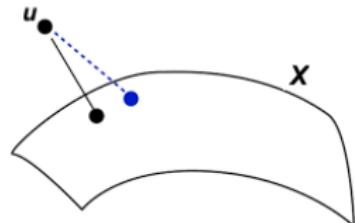
Many different applications



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$$\begin{aligned} \min_x \quad & \|x - u\|^2 \\ \text{s.t.} \quad & x \in X \end{aligned}$$



The SDP-exact region for the Euclidean Distance (ED) problem is

$$\mathcal{R}_f^{ed} = \{ u \in \mathbb{R}^n : \text{problem is solved exactly by its SDP relaxation} \}.$$

This is an affine slice of  $\mathcal{R}_f$ .

# Nearest point problems

## Theorem

The SDP-exact region satisfies

$$\mathcal{R}_f^{ed} = \bigcup_{x \in V_f} \left( x - \frac{1}{2} \text{Jac}_f(x) \cdot S_f^{ed} \right)$$

where

$$S_f^{ed} = \left\{ \lambda \in \mathbb{R}^m : \sum_i \lambda_i A_i \prec I_n \right\}, \quad \text{Jac}_f(x) : \mathbb{R}^m \rightarrow \text{NormalSpace}(x)$$

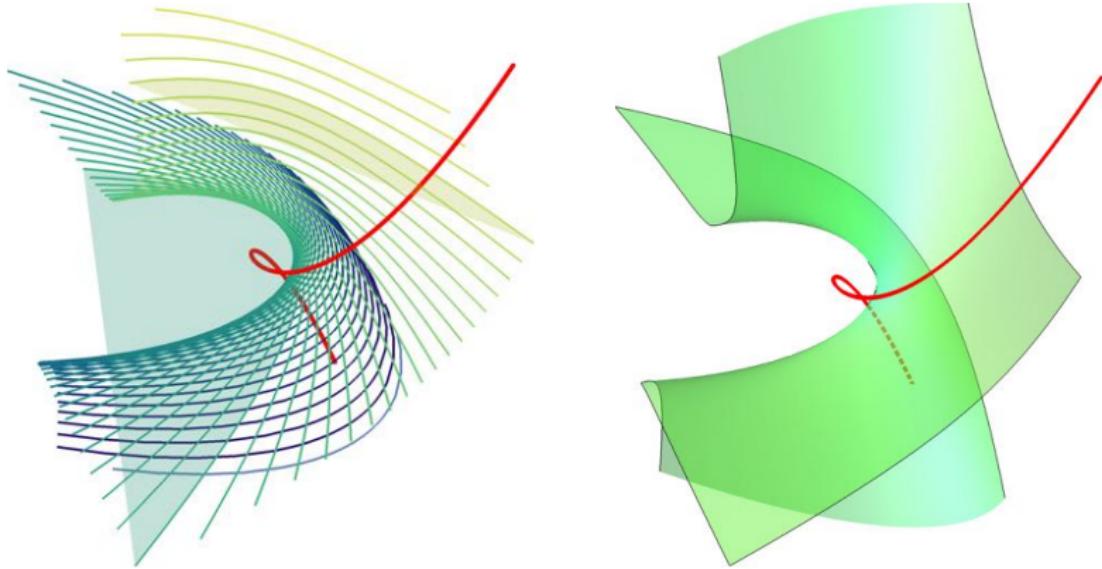
This is a bundle of *spectrahedral shadows*.

## Corollary (C.-Agarwal-Parrilo-Thomas)

If  $x$  be a smooth point, then  $\mathcal{R}_f^{ed}$  contains an open ball around  $x$ .

**Applications:** Triangulation (vision), Rank-1 tensor approximation.

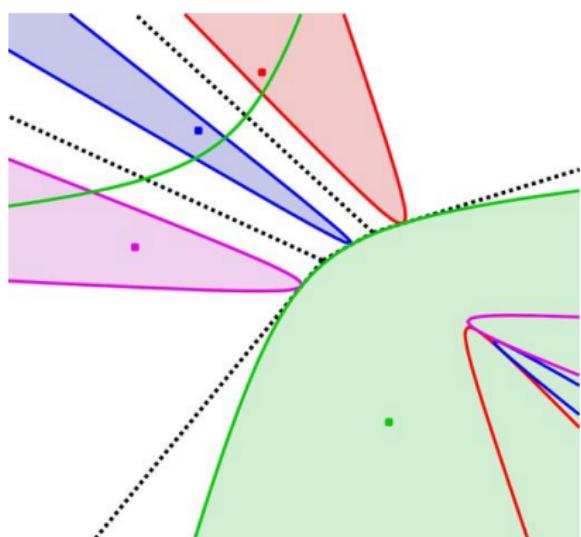
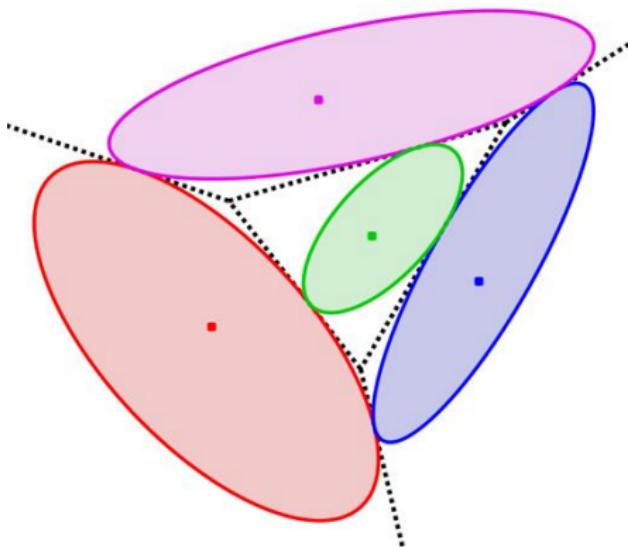
# Nearest point to the twisted cubic



$$X := \{(x_1, x_2, x_3) : x_2 = x_1^2, x_3 = x_1 x_2\}$$

## Nearest point to $2^n$ points

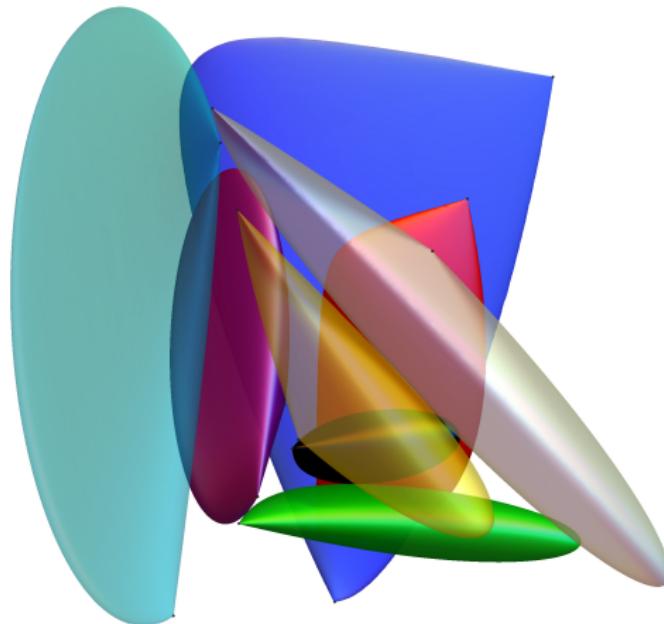
**Theorem:** Let  $m = n$  and  $\mathbf{f}$  generic, so  $X = V_{\mathbf{f}}$  is finite. The SDP-exact region consists of  $2^n$  spectrahedra. The boundaries are pairwise tangent.



$$X = (\text{four points in the plane})$$

## Nearest point to $2^n$ points

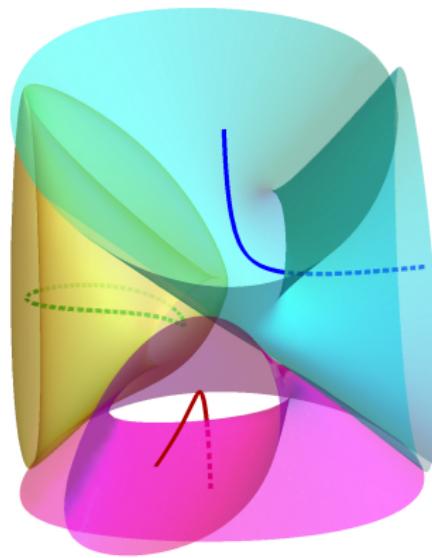
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$X =$  (eight points in 3D space)

# Nearest point to a complete intersection

**Theorem:** Let  $m \leq n$ ,  $\mathbf{f}$  generic. The degree of  $\partial_{\text{alg}} \mathcal{R}_{\mathbf{f}}^{\text{ed}}$  equals  $2^m n \binom{n-2}{m-2}$ .



$X = (\text{curve defined by two quadrics})$

# Beyond complete intersections

We developed symbolic algorithms to compute  $\partial_{\text{alg}} \mathcal{R}_f^{\text{ed}}$  for arbitrary  $f$ , even if  $m > n$ . We have a conjecture for the degree.

## Example (Twisted cubic)

We now use three defining equations:

$$X := \{(x_1, x_2, x_3) : x_2 = x_1^2, x_3 = x_1 x_2, x_1 x_3 = x_2^2\}$$

The region  $\mathcal{R}_f^{\text{ed}}$  is dense, and its bounded by

$$\begin{aligned} & 5832u_2^3u_3^6 + 27648u_2^6u_3^2 - 62208u_1u_2^4u_3^3 - 2916u_1^2u_2^2u_3^4 + 15552u_2^4u_3^4 - 5832u_1^3u_3^5 + 8748u_1^2u_3^6 - 5832u_2^2u_3^6 - 4374u_1u_3^7 \\ & + 729u_3^8 - 41472u_1^2u_2^5 + 86400u_1^3u_2^3u_3 + 27648u_1u_2^5u_3 + 60750u_1^4u_2u_3^2 - 41472u_1^2u_2^3u_3^2 - 62208u_2^5u_3^2 - 106920u_1^3u_2u_3^3 \\ & + 85536u_1u_2^3u_3^3 + 71442u_1^2u_2u_3^4 - 19656u_2^3u_3^4 - 19440u_1u_2u_3^5 + 3888u_2u_3^6 - 84375u_1^6 - 54000u_1^4u_2^2 + 72576u_1^2u_2^4 \\ & + 202500u_1^5u_3 - 19440u_1^3u_2^2u_3 - 48384u_1u_2^4u_3 - 220725u_1^4u_3^2 + 6912u_1^2u_2^2u_3^2 + 58032u_2^4u_3^2 + 140454u_1^3u_3^3 - 35424u_1u_2^2u_3^3 \\ & - 54027u_1^2u_3^4 + 8424u_2^2u_3^4 + 11178u_1u_3^5 - 1161u_3^6 + 40050u_1^4u_2 - 50760u_1^2u_2^3 - 21132u_1^3u_2u_3 + 33840u_1u_2^3u_3 \\ & + 11880u_1^2u_2u_3^2 - 28744u_2^3u_3^2 + 3708u_1u_2u_3^3 - 1314u_2u_3^4 - 7431u_1^4 + 17736u_1^2u_2^2 + 6112u_1^3u_3 - 11824u_1u_2^2u_3 - 3246u_1^2u_3^2 \\ & + 7976u_2^2u_3^2 + 312u_1u_3^3 + 37u_3^4 - 3096u_1^2u_2 + 2064u_1u_2u_3 - 1176u_2u_3^2 + 216u_1^2 - 144u_1u_3 + 72u_3^2. \end{aligned}$$

# Summary

- We characterized the rank-one region (for SDPs) and the SDP-exact region (for QPs).
- We computed the degree for generic instances.
- The SDP-exact region for distance problems has even more structure (a bundle of spectrahedral shadows).

# Summary

- We characterized the rank-one region (for SDPs) and the SDP-exact region (for QPs).
- We computed the degree for generic instances.
- The SDP-exact region for distance problems has even more structure (a bundle of spectrahedral shadows).

If you want to know more:

- D. Cifuentes, C. Harris, B. Sturmfels, *The geometry of SDP-exactness in quadratic optimization*, Math. Prog. (2019). arXiv:1804.01796.
- D. Cifuentes, S. Agarwal, P. Parrilo, R. Thomas, *On the local stability of semidefinite relaxations*, arXiv:1710.04287.

**Thanks for your attention!**

## More SDP-exact regions

