

# Convex Algebraic Geometry



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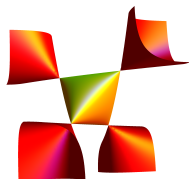
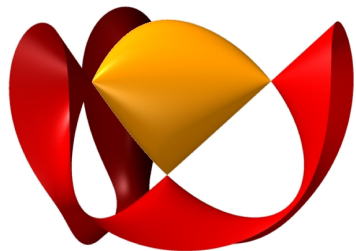
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Algebraic techniques can help answer questions about these convex sets. Convexity provides additional tools and challenges.

# Motivational Example: The elliptope

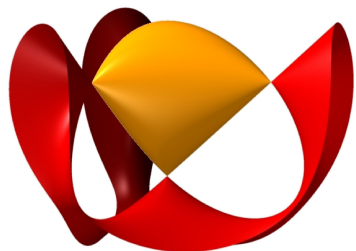
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$$\left\{ (x, y, z) \in \mathbb{R}^3 : \begin{pmatrix} 1 & x & y \\ x & 1 & z \\ y & z & 1 \end{pmatrix} \succeq 0 \right\}$$



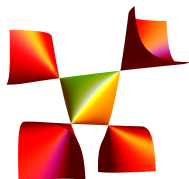
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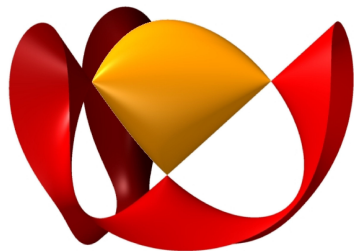
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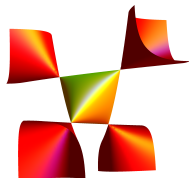
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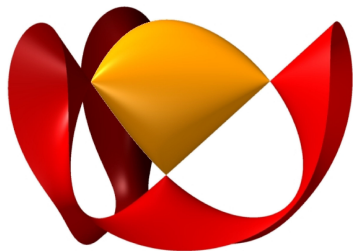
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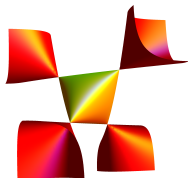


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appears in ...

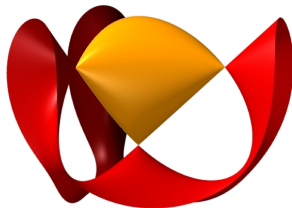
- ▶ statistics as set of correlation matrices
- ▶ combinatorial optimization





# The elliptope: convex algebraic structure

Convex structure



# The ellipsope: convex algebraic structure

## Convex structure

- ▶  $\infty$ -many zero-dim'l faces
- ▶ 6 one-dim'l faces
- ▶ 4 vertices



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## Algebraic structure

Bounded by a cubic hypersurface,

$$\{(x, y, z) \in \mathbb{R}^3 : f = 2xyz - x^2 - y^2 - z^2 + 1 = 0\}$$

that has 4 nodes and contains 6 lines.

# Low-rank matrices

The elliptope exhibits *general* behavior for its size and dimension.

For general  $A_0, A_1, A_2, A_3 \in \mathbb{R}_{sym}^{3 \times 3}$  the set of matrices

$$\{A_0 + xA_1 + yA_2 + zA_3 : (x, y, z) \in \mathbb{R}^3 \text{ or } \mathbb{C}^3\}$$

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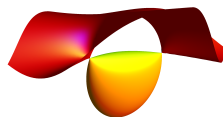
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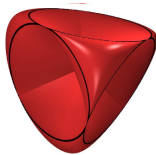
**Why?** The set of matrices of rank  $\leq 1$  is variety of codimension 3 and degree 4 in  $\mathbb{R}_{sym}^{3 \times 3} \cong \mathbb{R}^6$ .

If  $A_0 \succ 0$ , there will always be 2 or 4 matrices of rank-one over  $\mathbb{R}$ .



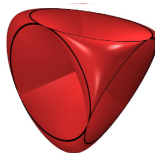
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The dual of the elliptope is bounded by the union of a **quartic surface** and four planes. Writing down the solution of a random linear optimization problem over the elliptope requires solving a **degree four** polynomial.

# Moments and Sums of Squares

The cone of **nonnegative polynomials**

$$\text{NN}_{n,2d} = \{p \in \mathbb{R}[x_1, \dots, x_n]_{\leq 2d} : p(x) \geq 0 \text{ for all } \mathbb{R}^n\}$$

is **convex**, **semialgebraic**, and contains the cone of **sums of squares**

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$$NN_{n,2d}^\circ = \text{conv}\{\lambda(1, x_1, \dots, x_n, x_1^2, x_1x_2, \dots, x_n^{2d}) : \lambda \in \mathbb{R}, x \in \mathbb{R}^n\}.$$

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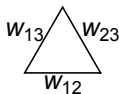
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Moreover  $SOS_{n,2d}^\circ$  is a **spectrahedron**!

# Sums of squares and the Goemans-Williamson relaxation

MAXCUT: Given weights  $w_e \in \mathbb{R}$  to the edges of a graph  $G = (V, E)$ , find a cut  $V \rightarrow \{\pm 1\}$  maximizing the summed weight of mixed edges.

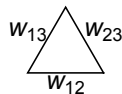
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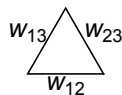
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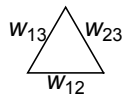
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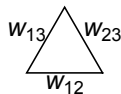
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