

I.4. CONVEX LMI MODELLING

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Minors

A **minor** of a matrix F is the determinant of a submatrix of F , say with row index I and column index J

If $I = J$ this is a **principal minor**

If $I = J = 1 \dots k$ this is a **leading principal minor**

A symmetric matrix F is positive definite iff all its leading principal minors are positive

A symmetric matrix F is positive semidefinite iff all its principal minors are nonnegative

Positive semidefiniteness

If $F \in \mathbb{R}^{m \times m}$ we have $2^m - 1$ principal minors

A simpler and equivalent criterion follows from the fact that a univariate polynomial $t \mapsto f(t) = \sum_k f_{m-k} t^k = \prod_k (t - t_k)$ which has **only real roots** satisfies $t_k \leq 0$ iff $f_k \geq 0$

Apply to **characteristic polynomial**

$$t \mapsto f(t) = \det(tI_m + F) = \sum_{k=0}^m f_{m-k}(F) t^k$$

A symmetric matrix F is positive semidefinite iff $f_i(F) \geq 0$, $\forall i$

Only m polynomials to be checked, they are (signed) sums of principal minors

Geometry of LMI sets

Given symmetric matrices F_i we want to characterize the shape in \mathbb{R}^n of the LMI set

$$\mathcal{F} = \{x \in \mathbb{R}^n : F(x) = F_0 + \sum_{i=1}^n x_i F_i \succeq 0\}$$

Build characteristic polynomial

$$t \mapsto f(t, x) = \det(tI_m + F(x)) = \sum_{k=0}^m f_{m-k}(x)t^k$$

which is monic, i.e. $f_0(x) = 1$

Matrix $F(x)$ is PSD iff $f_i(x) \geq 0$ for all $i = 1, \dots, m$

Semialgebraic description

Diagonal minors are multivariate **polynomials** of the x_i

So the LMI set can be described as

$$\mathcal{F} = \{x \in \mathbb{R}^n : f_i(x) \geq 0, i = 1, 2, \dots\}$$

which is a **basic semialgebraic** set

(basic = intersection of polynomial level-sets)

Moreover, it is a **convex** set

Example of 2D LMI feasible set

$$F(x) = \begin{bmatrix} 1 - x_1 & x_1 + x_2 & x_1 \\ x_1 + x_2 & 2 - x_2 & 0 \\ x_1 & 0 & 1 + x_2 \end{bmatrix} \succeq 0$$

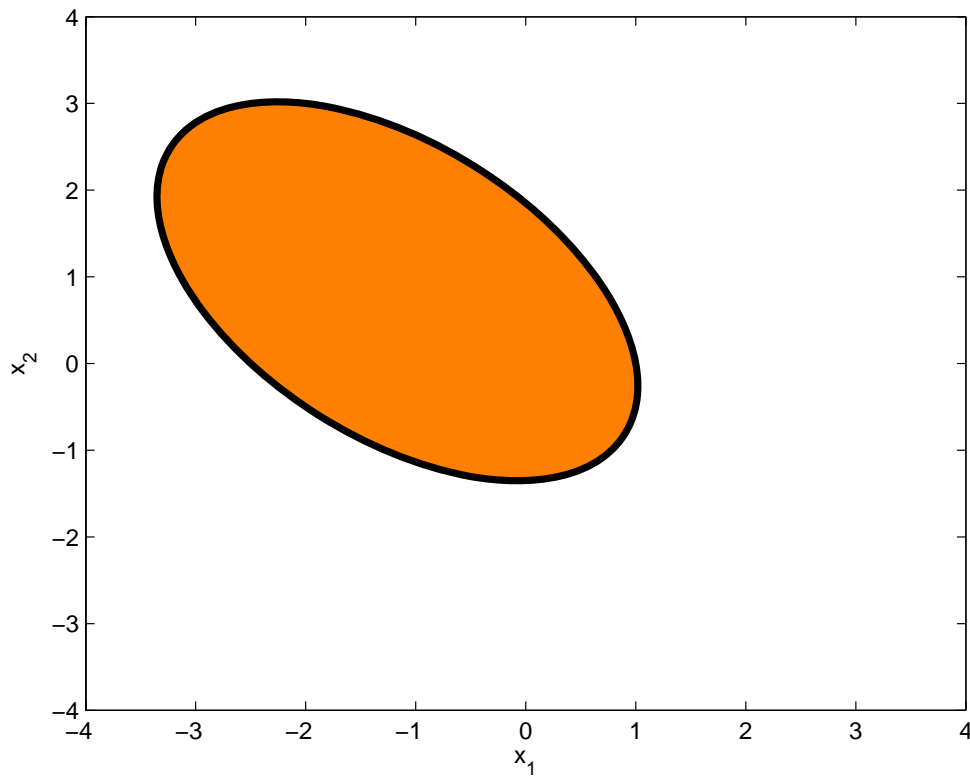
System of 3 polynomial inequalities $f_i(x) \geq 0$

1st order minors: $f_1(x) = 4 - x_1 \geq 0$

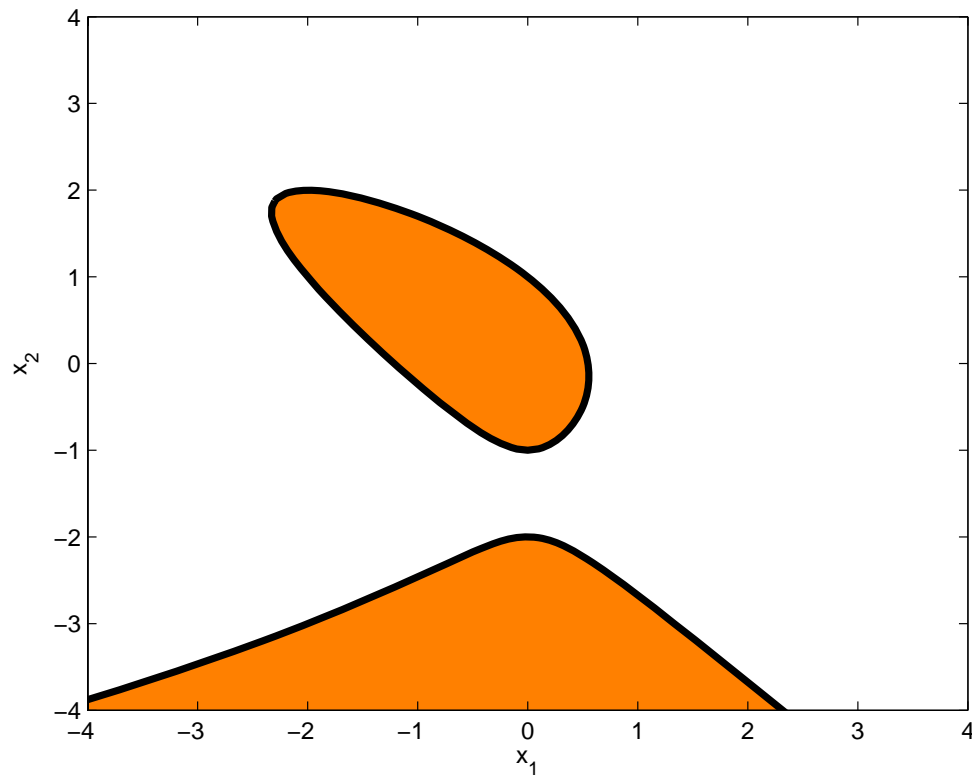
LMI set = intersection of an infinite number of halfspaces

$\{x : y^T F(x) y \geq 0\}$ for all $y \in \mathbb{R}^3$

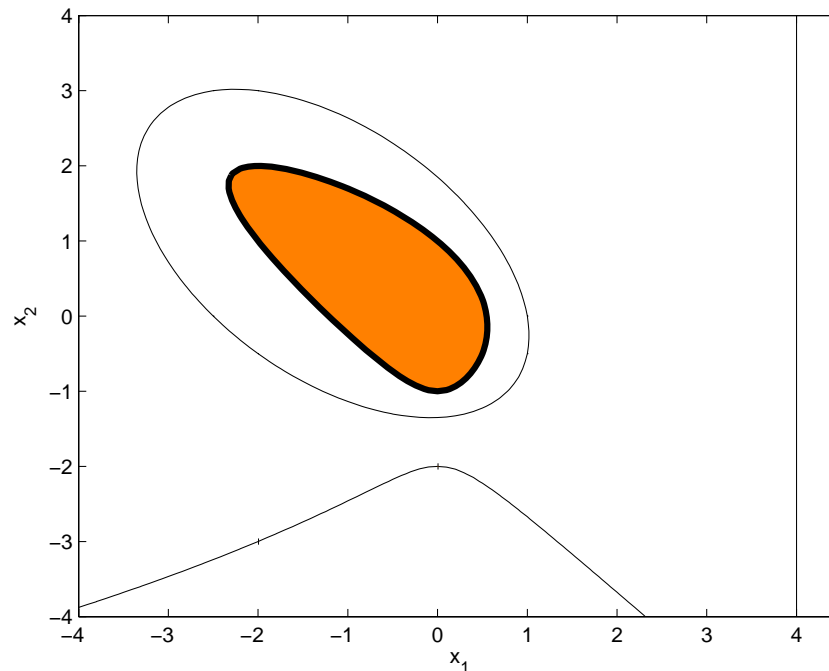
2nd order: $f_2(x) = 5 - 3x_1 + x_2 - 2x_1^2 - 2x_1x_2 - 2x_2^2 \geq 0$



3rd order: $f_3(x) = 2 - 2x_1 + x_2 - 3x_1^2 - 3x_1x_2 - 2x_2^2 - x_1x_2^2 - x_2^3 \geq 0$



LMI feasible set = intersection of sets $\{x : f_i(x) \geq 0\}, i = 1, 2, 3$

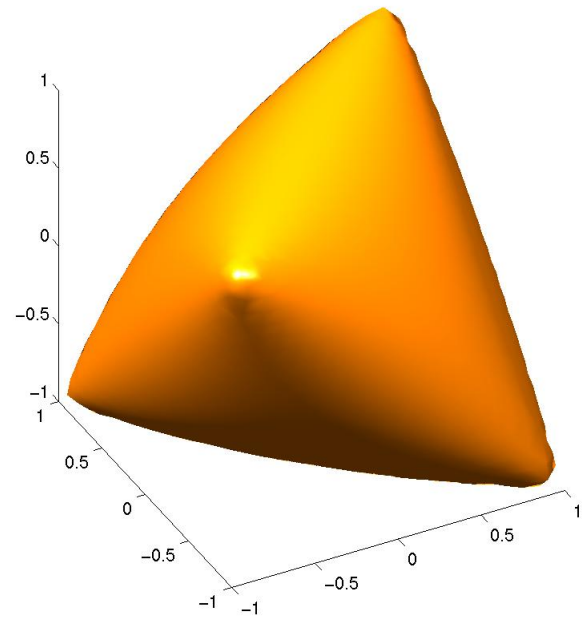


Boundary of LMI region shaped by **determinant**

Other polynomials only isolate **convex connected component**

Example of 3D LMI feasible set

$$\mathcal{F} = \{x \in \mathbb{R}^3 : \underbrace{\begin{bmatrix} 1 & x_1 & x_2 \\ x_1 & 1 & x_3 \\ x_2 & x_3 & 1 \end{bmatrix}}_{F(x)} \succeq 0\}$$

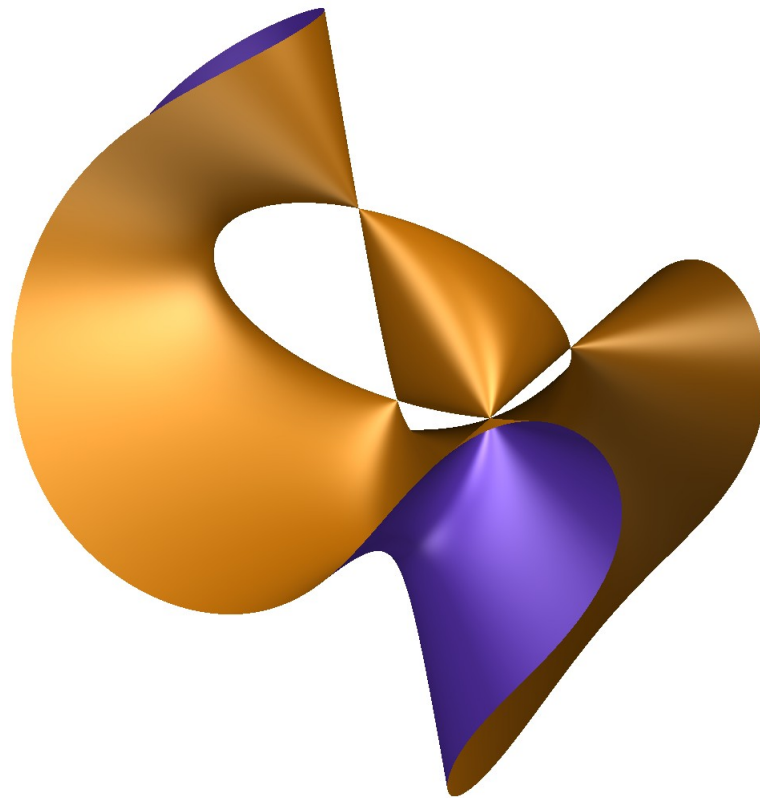


A smoothed tetrahedron..

vertices correspond to points x for which $\text{rank } F(x) = 1$

Semialgebraic formulation

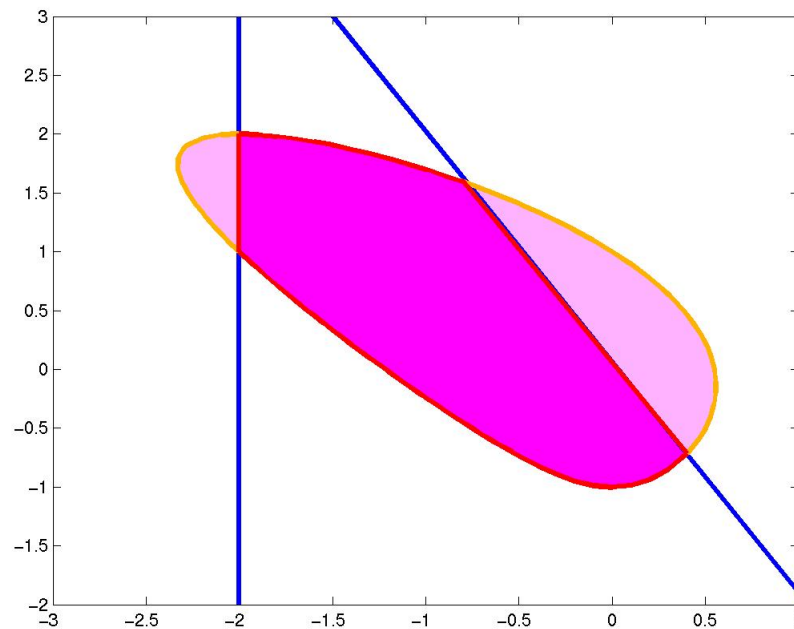
$$\mathcal{F} = \{x \in \mathbb{R}^3 : 1 + 2x_1x_2x_3 - (x_1^2 + x_2^2 + x_3^2) \geq 0, 3 - x_1^2 - x_2^2 - x_3^2 \geq 0\}$$



Intersection of LMI sets

Intersection of LMI feasible sets is also LMI

$$F(x) \succeq 0 \quad x_1 \geq -2 \quad 2x_1 + x_2 \leq 0$$



LMI sets

LMI sets are convex basic semialgebraic sets..
but are all convex basic semialgebraic sets LMI ?

Let us make a fundamental **distinction** between

- LMI representable sets
- lifted-LMI representable sets

We say that a convex set $X \subset \mathbb{R}^n$ is **LMI representable** if there exists an affine mapping $F(x)$ such that

$$x \in X \iff F(x) \succeq 0$$

LMI and lifted-LMI representability

We say that a convex set $X \subset \mathbb{R}^n$ is **lifted-LMI representable** if there exists an affine mapping $F(x, u)$ such that

$$x \in X \iff \exists u \in \mathbb{R}^m : F(x, u) \succeq 0$$

A set X is lifted-LMI representable when

$$x \in X \iff \exists u : F(x, u) \succeq 0$$

i.e. when it is the **projection** of the solution set of the LMI $F(x, u) \succeq 0$ onto the x -space and u are additional, or **lifting** variables

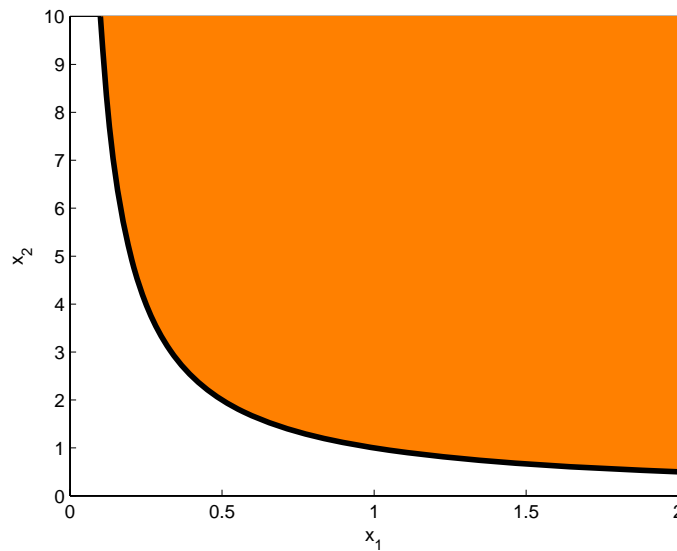
In other words, lifting variables u are **not allowed** in LMI representations

LMI and lifted-LMI functions

Similarly, a convex **function** $f : \mathbb{R}^n \rightarrow \mathbb{R}$ is LMI (or lifted-LMI) representable if its **epigraph**

$$\{x, t : f(x) \leq t\}$$

is an LMI (or lifted-LMI) representable set



Conic quadratic forms

The Lorentz, or ice-cream cone

$$\{x, t \in \mathbb{R}^n \times \mathbb{R} : \|x\|_2 \leq t\}$$

is LMI representable as

$$\left\{ x, t \in \mathbb{R}^n \times \mathbb{R} : \begin{bmatrix} tI_n & x \\ x^T & t \end{bmatrix} \succeq 0 \right\}$$

As a result, all second-order conic sets are LMI representable

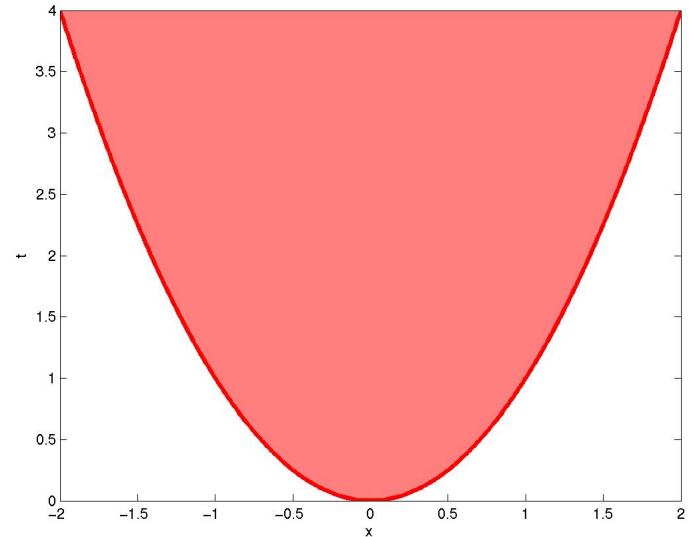
In the sequel we give a list of LMI and lifted-LMI representable sets (following Ben-Tal, Nemirovski, Nesterov)

Quadratic forms

The **Euclidean norm** $\{x, t \in \mathbb{R}^n \times \mathbb{R} : \|x\|_2 \leq t\}$
is LMI representable (see previous slide)

The **squared Euclidean norm** $\{x, t \in \mathbb{R}^n \times \mathbb{R} : x^T x \leq t\}$
is also LMI representable as

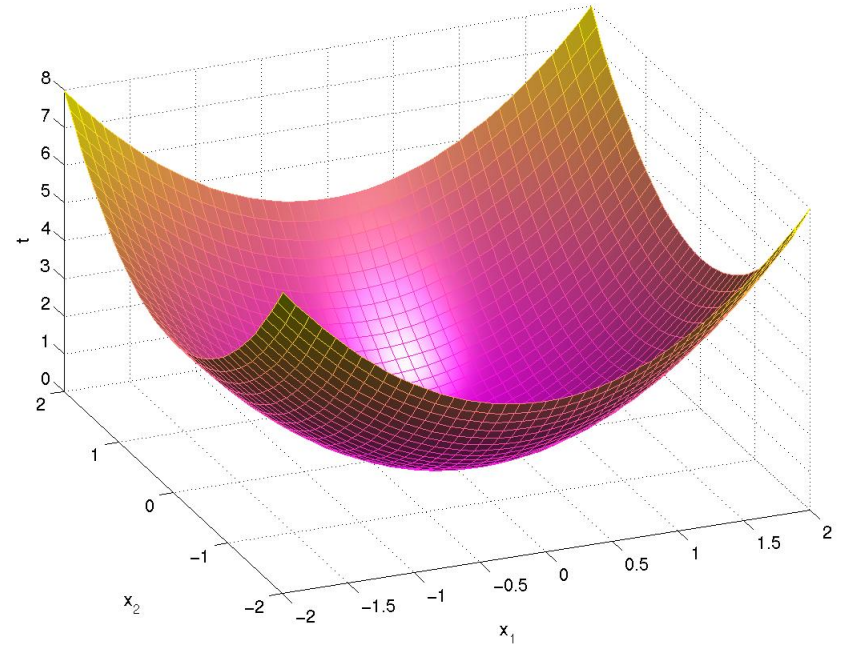
$$\begin{bmatrix} t & x^T \\ x & I_n \end{bmatrix} \succeq 0$$



Quadratic forms (2)

The **convex quadratic** set $\{x \in \mathbb{R}^n : x^T A x + b^T x + c \leq 0\}$ with $A = A^T \succeq 0$ is LMI representable as

$$\begin{bmatrix} -b^T x - c & x^T D^T \\ Dx & I_n \end{bmatrix} \succeq 0$$

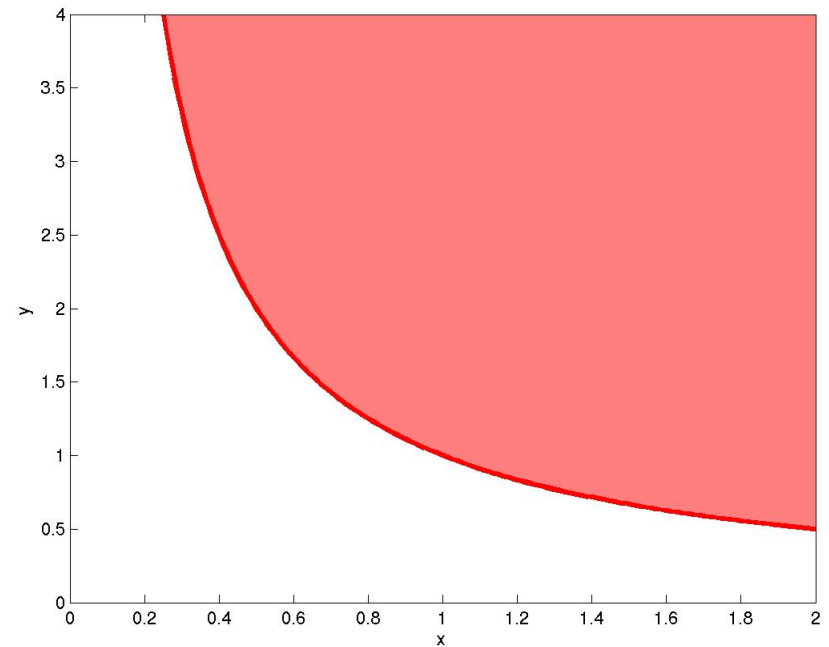


where D is the Cholesky factor of $A = D^T D$

Hyperbola

The branch of **hyperbola** $\{x, y \in \mathbb{R}^2 : x \geq 0, xy \geq 1\}$ is LMI representable as

$$\begin{bmatrix} x & 1 \\ 1 & y \end{bmatrix} \succeq 0$$

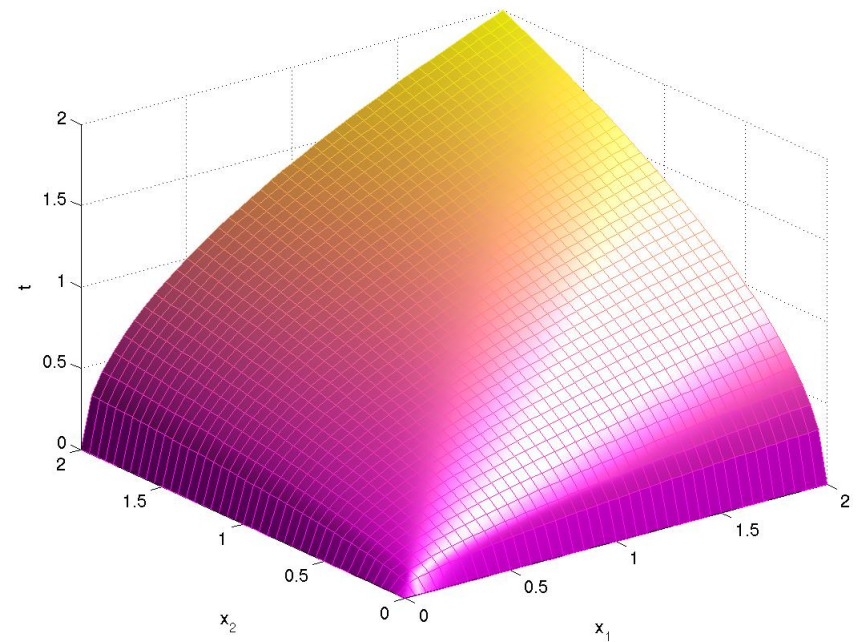


Geometric mean of two variables

The hypograph of the **geometric mean** of 2 variables

$\{x_1, x_2, t \in \mathbb{R}^3 : x_1, x_2 \geq 0, \sqrt{x_1 x_2} \geq t\}$ is LMI representable as

$$\begin{bmatrix} x_1 & t \\ t & x_2 \end{bmatrix} \succeq 0$$



Geometric mean of several variables

The hypograph of the geometric mean of 2^k variables
 $\{x_1, \dots, x_{2^k}, t \in \mathbb{R}^{2^k+1} : x_i \geq 0, (x_1 \cdots x_{2^k})^{1/2^k} \geq t\}$
is **lifted-LMI** representable

Proof: **iterate** the previous construction by introducing **lifting variables**

Example with $k = 3$:

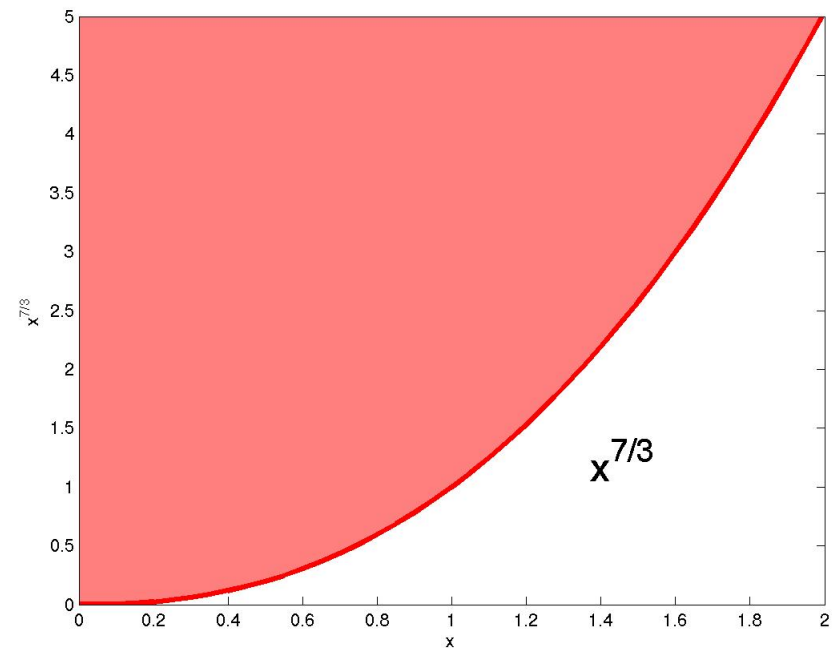
$$\begin{array}{l} (x_1 x_2 \cdots x_8)^{1/8} \geq t \\ \left. \begin{array}{l} \sqrt{x_1 x_2} \geq x_{11} \\ \sqrt{x_3 x_4} \geq x_{12} \\ \sqrt{x_5 x_6} \geq x_{13} \\ \sqrt{x_7 x_8} \geq x_{14} \end{array} \right\} \left. \begin{array}{l} \sqrt{x_{11} x_{12}} \geq x_{21} \\ \sqrt{x_{13} x_{14}} \geq x_{22} \end{array} \right\} \sqrt{x_{21} x_{22}} \geq t \end{array}$$

Useful idea in other LMI representability problems

Rational power functions

Following the same ideas, the increasing rational power functions

$$f(x) = x^{p/q}, \quad x \geq 0$$

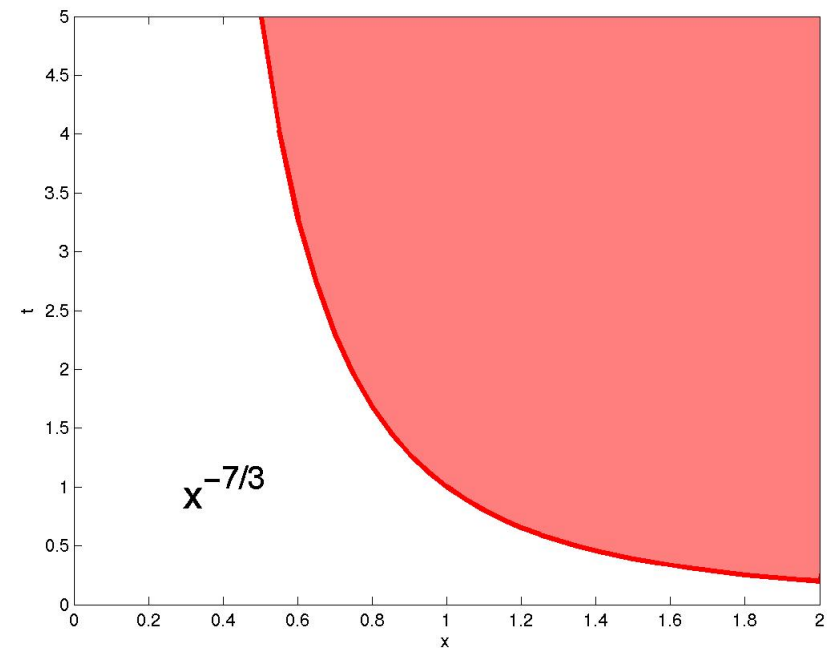


with rational $p/q \geq 1$, are lifted-LMI representable

Rational power functions

Similarly, the decreasing rational power functions

$$g(x) = x^{-p/q}, \quad x \geq 0$$



with rational $p/q \geq 0$, are lifted-LMI representable

Rational power functions

Example: $\{x, t : x \geq 0, x^{7/3} \leq t\}$

Start from lifted-LMI representable $\hat{t} \leq (\hat{x}_1 \cdots \hat{x}_8)^{1/8}$
and replace

$$\begin{aligned}\hat{t} &= \hat{x}_1 = x \geq 0 \\ \hat{x}_2 &= \hat{x}_3 = \hat{x}_4 = t \geq 0 \\ \hat{x}_5 &= \hat{x}_6 = \hat{x}_7 = \hat{x}_8 = 1\end{aligned}$$

to get

$$\begin{aligned}x &\leq x^{1/8}t^{3/8} \\ x^{7/8} &\leq t^{3/8} \\ x^{7/3} &\leq t\end{aligned}$$

Same idea works for any rational $p/q \geq 1$

- **lift** = use additional variables, and
- **project** in the space of original variables

Even power monomial

The epigraph of **even power monomial** $\mathcal{F} = \{x, t : x^{2p} \leq t\}$ where p is a positive integer is lifted-LMI representable

Indeed $\{x, t : x^{2p} \leq t\} \iff \{x, y, t : x^2 \leq y\}$ and $\{x, y, t : y \geq 0, y^p \leq t\}$, both lifted-LMI representable

Use **lifting** y and **project** back onto x, t

Similarly, **even power polynomials** are lifted-LMI representable (several monomials)

Quartic level set

Model quartic level set

$$\mathcal{F} = \{x, t : x^4 \leq t\}$$

as

$$\mathcal{F} = \{x, t : \exists y : y \geq x^2, t \geq y^2, y \geq 0\}$$

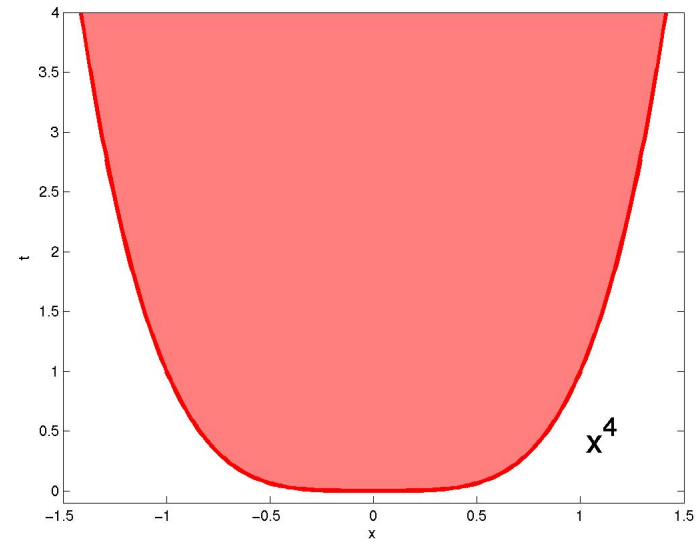
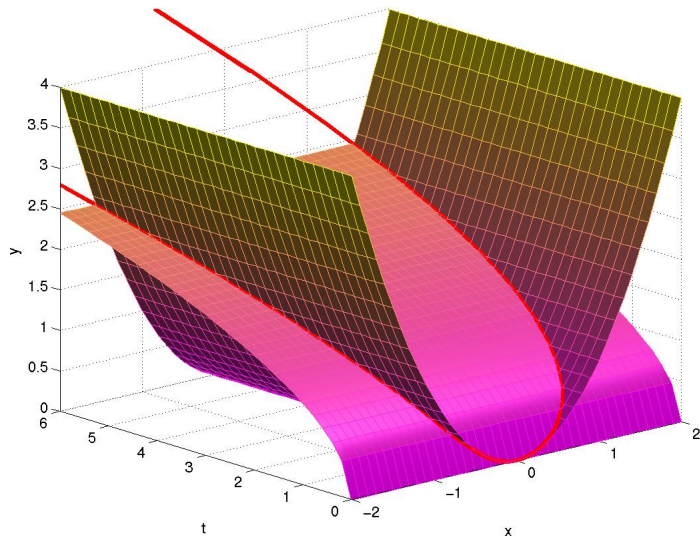
LMI in x, t and y

$$\begin{bmatrix} 1 & x \\ x & y \end{bmatrix} \succeq 0 \quad \begin{bmatrix} 1 & y \\ y & t \end{bmatrix} \succeq 0$$

It can be shown that it is impossible to remove the lifting variable y while keeping a (finite-dimensional) LMI formulation

Quartic level set: from 3D to 2D

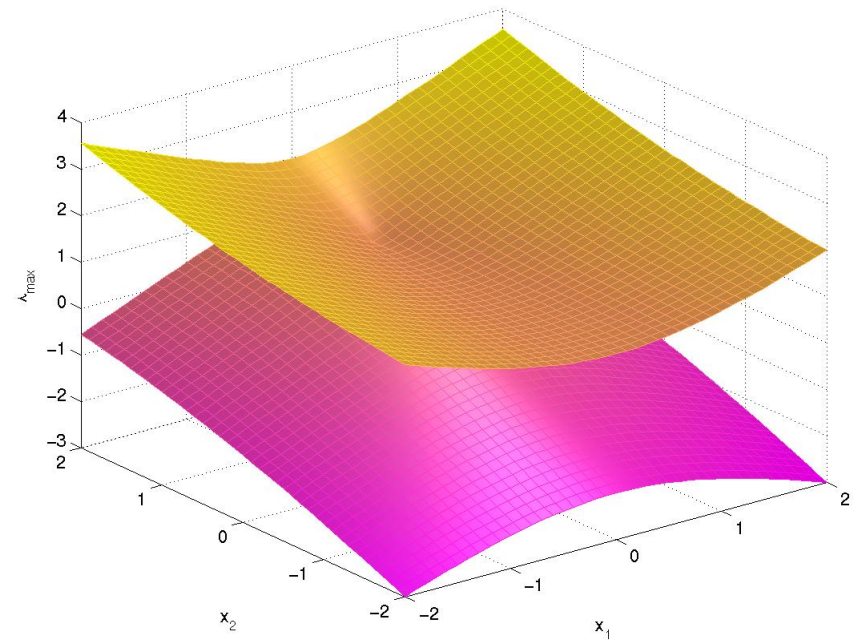
$$\mathcal{F} = \{x, t : x^4 \leq t\}$$



Largest eigenvalue

Function **largest eigenvalue** of a symmetric matrix $\{X = X^T \in \mathbb{R}^{n \times n}, t \in \mathbb{R} : \lambda_{\max}(X) \leq t\}$ is LMI representable as

$$X \preceq tI_n$$



Eigenvalues of matrix $\begin{bmatrix} 1 & x_1 \\ x_1 & x_2 \end{bmatrix}$

Sums of largest eigenvalues

Let

$$S_k(X) = \sum_{i=1}^k \lambda_i(X), \quad k = 1, \dots, n$$

denote the **sum of the k largest eigenvalues** of an n -by- n symmetric matrix X

The epigraph $\{X = X^T \in \mathbb{R}^{n \times n}, t \in \mathbb{R} : S_k(X) \leq t\}$ is lifted-LMI representable as

$$\begin{aligned} t - ks - \text{trace } Z &\preceq 0 \\ Z &\preceq 0 \\ Z - X + sI_n &\preceq 0 \end{aligned}$$

where Z and s are **liftings**

Determinant of a PSD matrix

The **determinant**

$$\det(X) = \prod_{i=1}^n \lambda_i(X)$$

is not a convex function of X , but the function

$$f_q(X) = -\det^q(X), \quad X = X^T \succeq 0$$

is convex when $q \in [0, 1/n]$ is rational

The epigraph $\{X = X^T \in \mathbb{R}^{n \times n}, t \in \mathbb{R} : f_q(X) \leq t\}$ is lifted-LMI representable

$$\begin{bmatrix} X & \Delta \\ \Delta^T & \text{diag } \Delta \\ t \leq (\delta_1 \cdots \delta_n)^q \end{bmatrix} \succeq 0$$

since we know that the latter constraint (hypograph of a concave monomial) is lifted-LMI representable

Here Δ is a lower triangular matrix of **liftings** with diagonal entries δ_i

Application: extremal ellipsoids

A little excursion in the world of ellipsoids and polytopes..

Various representations of an ellipsoid in \mathbb{R}^n

$$\begin{aligned} E &= \{x \in \mathbb{R}^n : x^T P x + 2x^T q + r \leq 0\} \\ &= \{x \in \mathbb{R}^n : (x - x_c)^T P (x - x_c) \leq 1\} \\ &= \{x = Qy + x_c \in \mathbb{R}^n : y^T y \leq 1\} \\ &= \{x \in \mathbb{R}^n : \|Rx - x_c\| \leq 1\} \end{aligned}$$

where

$$Q = R^{-1} = P^{-1/2} \succ 0$$

Ellipsoid volume

Volume of ellipsoid $E = \{Qy + x_c : y^T y \leq 1\}$

$$\text{vol } E = k_n \det Q$$

where k_n is volume of n -dimensional unit ball

$$k_n = \begin{cases} \frac{2^{(n+1)/2} \pi^{(n-1)/2}}{n(n-2)!!} & \text{for } n \text{ odd} \\ \frac{2\pi^{n/2}}{n(n/2-1)!} & \text{for } n \text{ even} \end{cases}$$

n	1	2	3	4	5	6	7	8
k_n	2.00	3.14	4.19	4.93	5.26	5.17	4.72	4.06

Unit ball has maximum volume for $n = 5$

Outer and inner ellipsoidal approximations

Let $S \subset \mathbb{R}^n$ be a **solid** = a closed bounded convex set with nonempty interior

- the largest volume ellipsoid E_{in} contained in S is unique and satisfies $E_{\text{in}} \subset S \subset nE_{\text{in}}$
- the smallest volume ellipsoid E_{out} containing S is unique and satisfies $E_{\text{out}}/n \subset S \subset E_{\text{out}}$

These are **Löwner-John** ellipsoids

Factor n reduces to \sqrt{n} if S is symmetric

How can these ellipsoids be computed ?

Ellipsoid in polytope

Let the intersection of hyperplanes

$$S = \{x \in \mathbb{R}^n : a_i^T x \leq b_i, i = 1, \dots, m\}$$

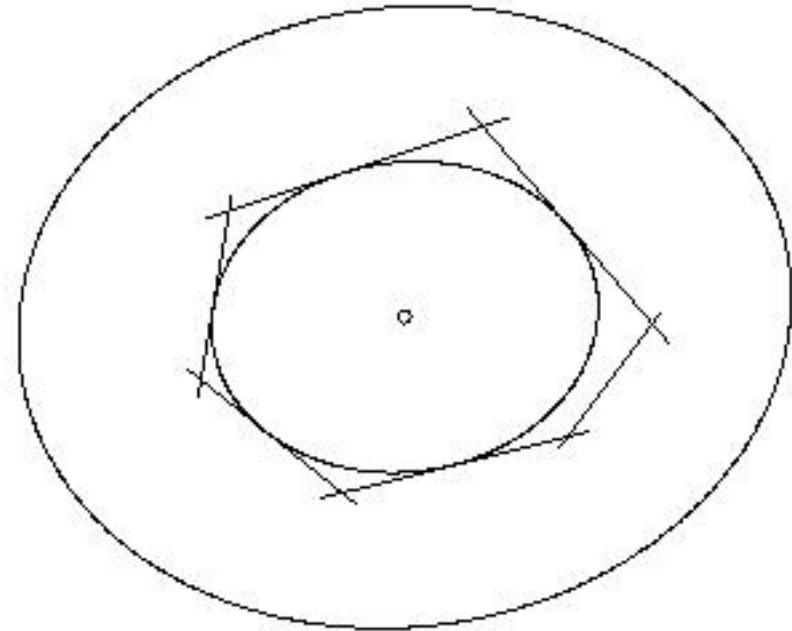
describe a **polytope** = bounded nonempty polyhedron

The **largest volume ellipsoid contained** in S is

$$E = \{Qy + x_c : y^T y \leq 1\}$$

where Q, x_c are optimal solutions of the LMI problem

$$\begin{aligned} \max \quad & \det^{1/n} Q \\ & Q \succeq 0 \\ & \|Qa_i\|_2 \leq b_i - a_i^T x_c \end{aligned}$$



Polytope in ellipsoid

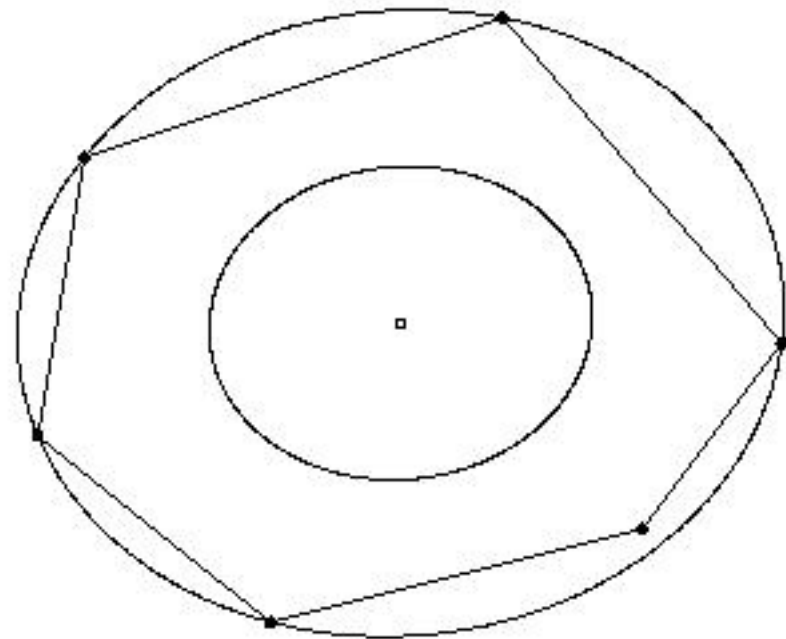
Let the convex hull of vertices $S = \text{conv} \{x_1, \dots, x_m\}$ describe a **polytope**

The **smallest volume ellipsoid containing** S is

$$E = \{x : (x - x_c)^T P (x - x_c) \leq 1\}$$

where $P, x_c = -P^{-1}q$ are optimal solutions of the LMI problem

$$\begin{aligned} \max \quad & t \\ & t \leq \det^{1/n} P \\ & \begin{bmatrix} P & q \\ q^T & r \end{bmatrix} \succeq 0 \\ & x_i^T P x_i + 2x_i^T q + r \leq 1 \end{aligned}$$



Sums of largest singular values

Let

$$\Sigma_k(X) = \sum_{i=1}^k \sigma_i(X), \quad k = 1, \dots, n$$

denote the **sum of the k largest singular values** of an n -by- m matrix X

Then the epigraph $\{X \in \mathbb{R}^{n \times m}, t \in \mathbb{R} : \Sigma_k(X) \leq t\}$ is lifted-LMI representable since

$$\sigma_i(X) = \lambda_i \left(\begin{bmatrix} 0 & X^T \\ X & 0 \end{bmatrix} \right)$$

and the sum of largest eigenvalues of a symmetric matrix is lifted-LMI representable

Positive polynomials

The set of univariate polynomials that are positive on the real axis is **lifted-LMI representable** in the coefficient space

Can be proved with cone duality (Nesterov) or with theory of moments (Lasserre) - more on that later

The even polynomial

$$p(s) = p_0 + p_1s + \cdots + p_{2n}s^{2n}$$

satisfies $p(s) \geq 0$ for all $s \in \mathbb{R}$ if and only if

$$\begin{aligned} p_k &= \sum_{i+j=k} X_{ij}, & k &= 0, 1, \dots, 2n \\ &= \text{trace } H_k X \end{aligned}$$

for some **lifting** matrix $X = X^T \succeq 0$

Sum-of-squares decomposition

The expression of p_k with Hankel matrices H_k comes from

$$p(s) = [1 \quad s \quad \cdots \quad s^n] X [1 \quad s \quad \cdots \quad s^n]^*$$

hence $X \succeq 0$ naturally implies $p(s) \geq 0$

Conversely, existence of X for any polynomial $p(s) \geq 0$ follows from the existence of a **sum-of-squares** (SOS) decomposition (with at most two elements) of

$$p(s) = \sum_k q_k^2(s) \geq 0$$

Matrix X has entries $X_{ij} = \sum_k q_{k_i} q_{k_j}$

Seeking the lifting matrix amounts to seeking an SOS decomposition

Primal and dual formulations

Global minimization of polynomial $p(s) = \sum_{k=0}^n p_k s^k$

Global optimum p^* : maximum value of \hat{p} such that $p(s) - \hat{p} \geq 0$

Primal LMI problem

$$\begin{aligned} \max \quad & \hat{p} = p_0 - \text{trace } H_0 X \\ \text{s.t.} \quad & \text{trace } H_k X = p_k, \quad k = 1, \dots, n \\ & X \succeq 0 \end{aligned}$$

Dual LMI problem

$$\begin{aligned} \min \quad & p_0 + \sum_{k=1}^n p_k y_k \\ \text{s.t.} \quad & H_0 + \sum_{k=1}^n H_k y_k \succeq 0 \end{aligned}$$

with Hankel structure (**moment matrix**)

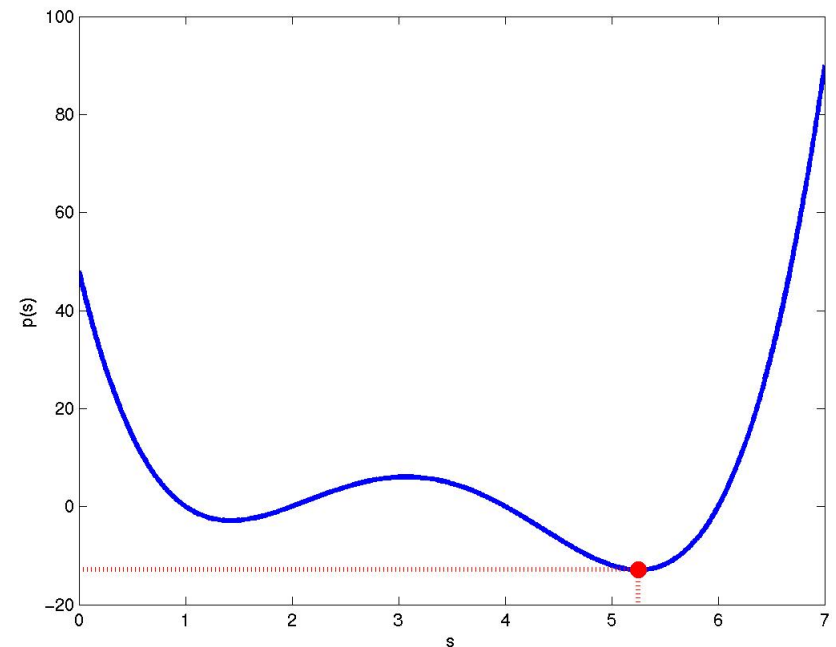
Positive polynomials and LMIs

Example: **Global minimization** of the polynomial

$$p(s) = 48 - 92s + 56s^2 - 13s^3 + s^4$$

Solving the **dual** LMI problem yields $p^* = p(5.25) = -12.89$

$$\begin{array}{ll} \min & 48 - 92y_1 + 56y_2 - 13y_3 + y_4 \\ \text{s.t.} & \begin{bmatrix} 1 & y_1 & y_2 \\ y_1 & y_2 & y_3 \\ y_2 & y_3 & y_4 \end{bmatrix} \succeq 0 \end{array}$$



Complex LMIs

The **complex** valued LMI

$$F(\boldsymbol{x}) = A(\boldsymbol{x}) + jB(\boldsymbol{x}) \succeq 0$$

is equivalent to the real valued LMI

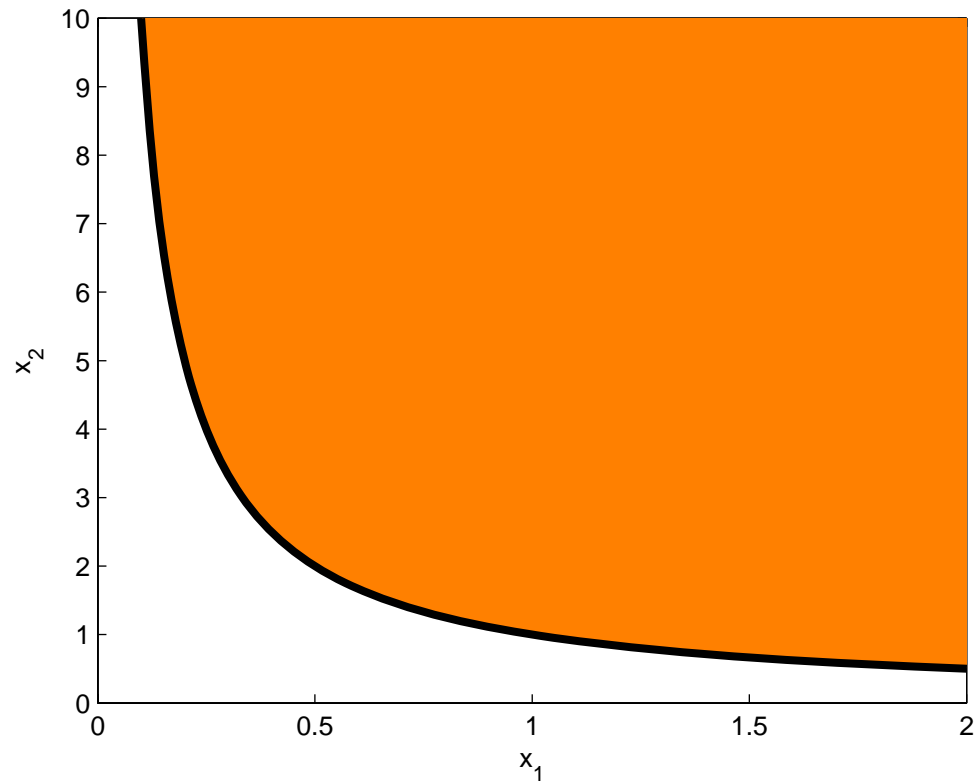
$$\begin{bmatrix} A(\boldsymbol{x}) & B(\boldsymbol{x}) \\ -B(\boldsymbol{x}) & A(\boldsymbol{x}) \end{bmatrix} \succeq 0$$

If there is a complex solution to the LMI
then there is a **real** solution to the same LMI

Note that matrix $A(\boldsymbol{x}) = A^T(\boldsymbol{x})$ is symmetric
whereas $B(\boldsymbol{x}) = -B^T(\boldsymbol{x})$ is skew-symmetric

INTERLUDE

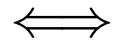
LMI set or not ?



$$x_1 x_2 \geq 1 \text{ and } x_1 \geq 0$$

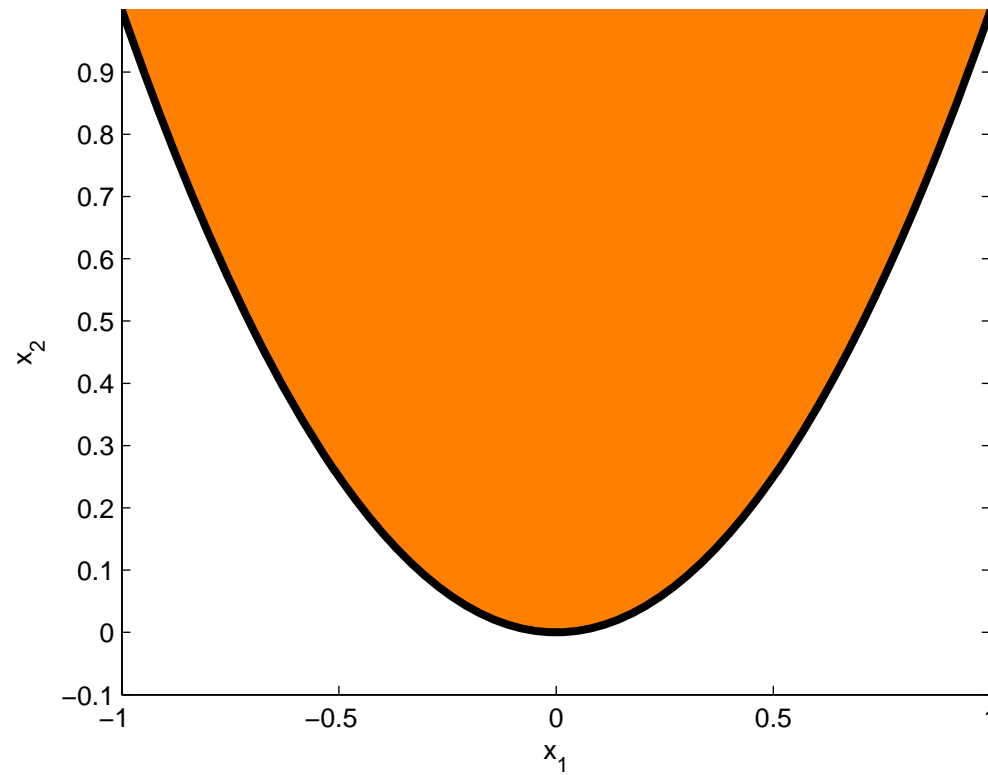
LMI

$$x_1 x_2 \geq 1 \text{ and } x_1 \geq 0$$



$$\begin{bmatrix} x_1 & 1 \\ 1 & x_2 \end{bmatrix} \succeq 0$$

LMI set or not ?



$$x_2 \geq x_1^2$$

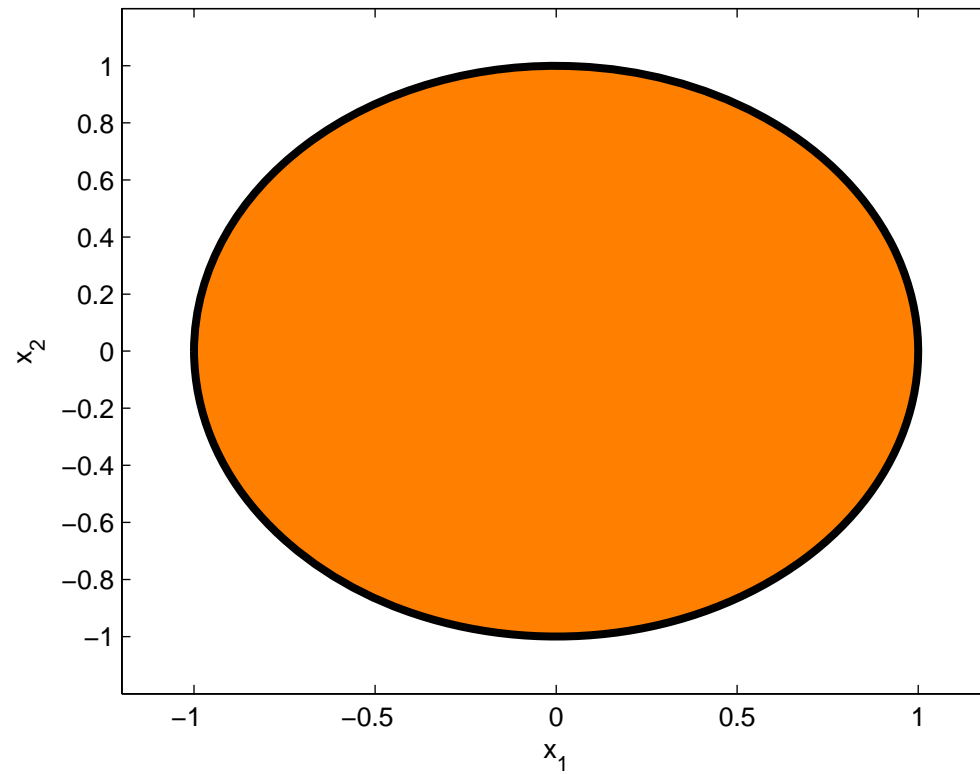
LMI

$$x_2 \geq x_1^2$$

\iff

$$\begin{bmatrix} 1 & x_1 \\ x_1 & x_2 \end{bmatrix} \succeq 0$$

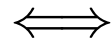
LMI set or not ?



$$x_1^2 + x_2^2 \leq 1$$

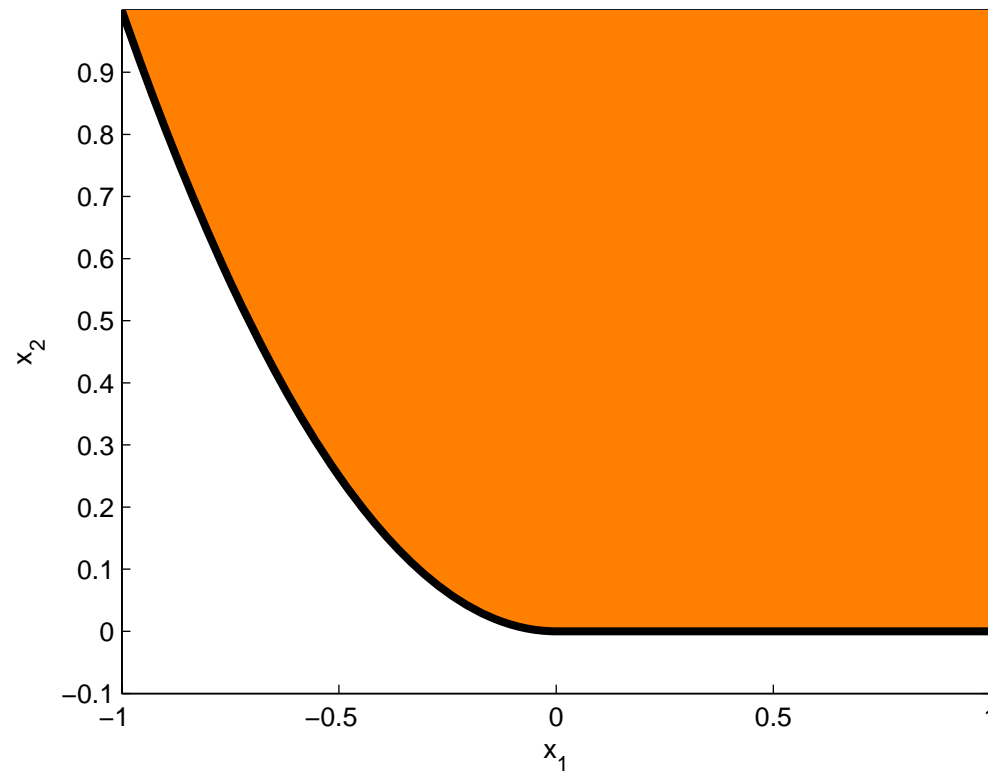
LMI

$$x_1^2 + x_2^2 \leq 1$$



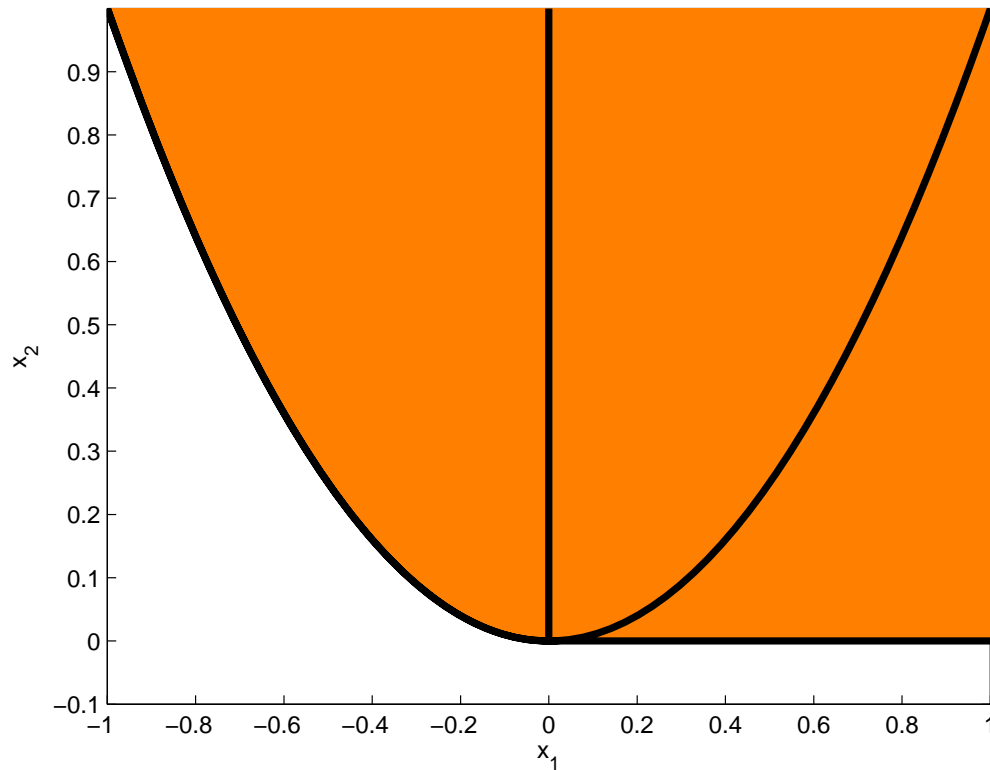
$$\begin{bmatrix} 1 + x_1 & x_2 \\ x_2 & 1 - x_1 \end{bmatrix} \succcurlyeq 0$$

LMI set or not ?



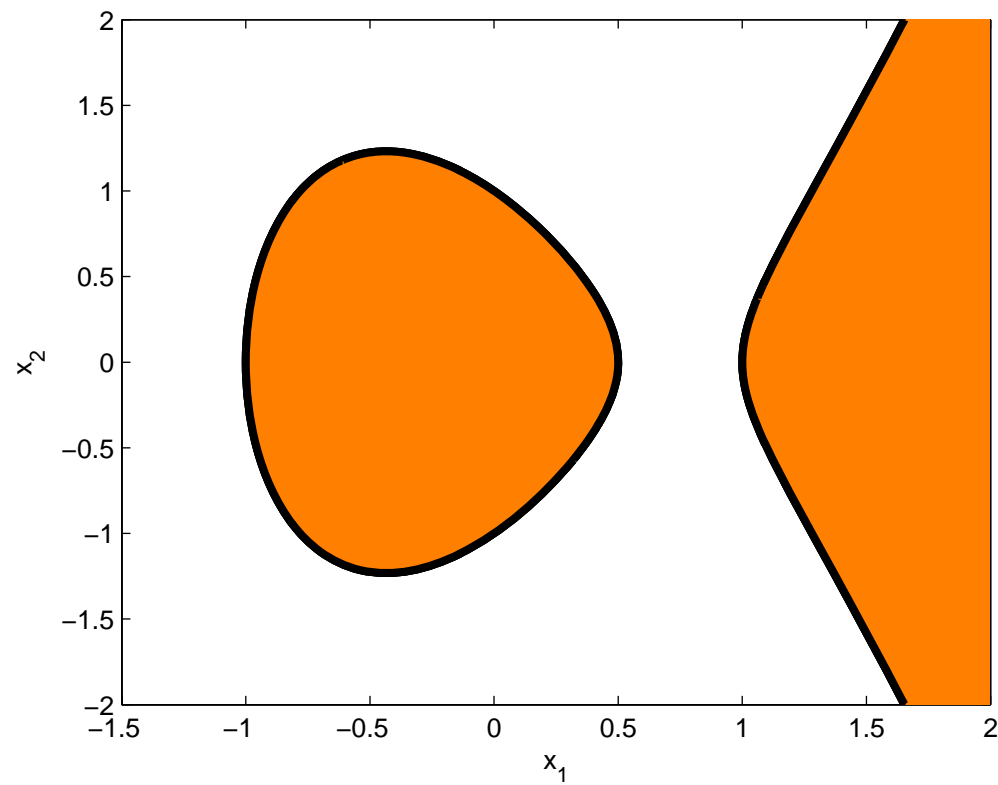
$$\{x \in \mathbb{R}^2 : t^4 + 2x_1t^2 + x_2 \geq 0, \forall t \in \mathbb{R}\}$$

NOT LMI: not basic semialgebraic



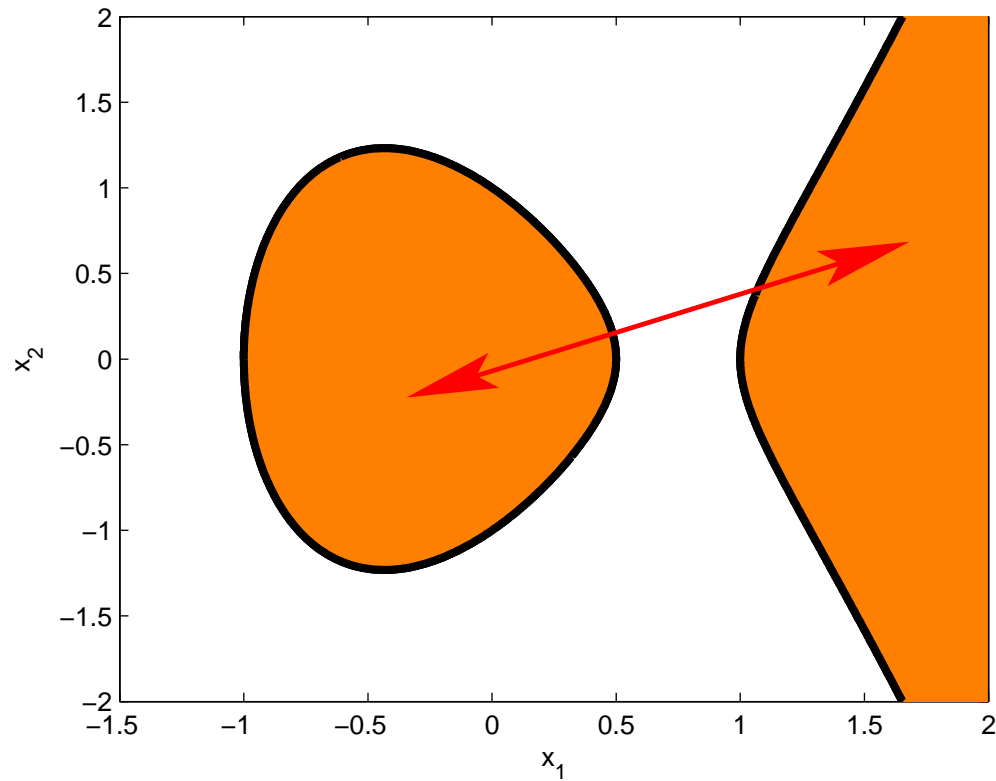
$$x_2 \geq x_1^2 \text{ or } x_1, x_2 \geq 0$$

LMI set or not ?



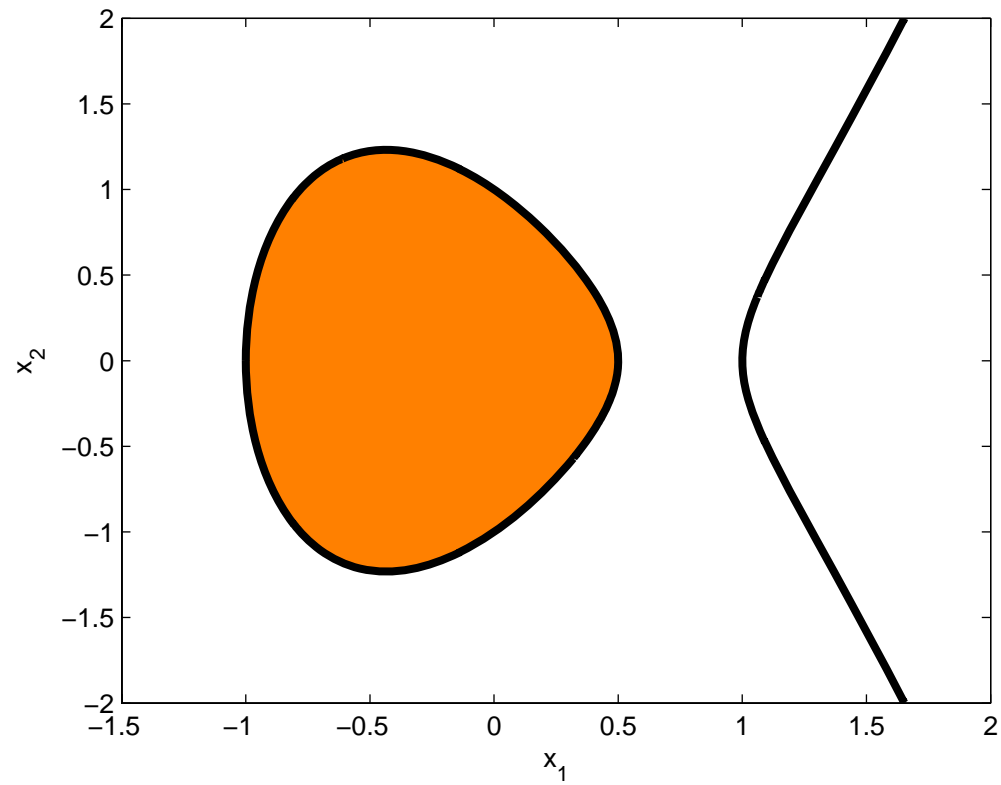
$$1 - 2x_1 - x_1^2 - x_2^2 + 2x_1^3 \geq 0$$

NOT LMI: not connected



$$1 - 2x_1 - x_1^2 - x_2^2 + 2x_1^3 \geq 0$$

LMI set or not ?



$$1 - 2x_1 - x_1^2 - x_2^2 + 2x_1^3 \geq 0 \text{ and } x_1 \leq \frac{1}{2}$$

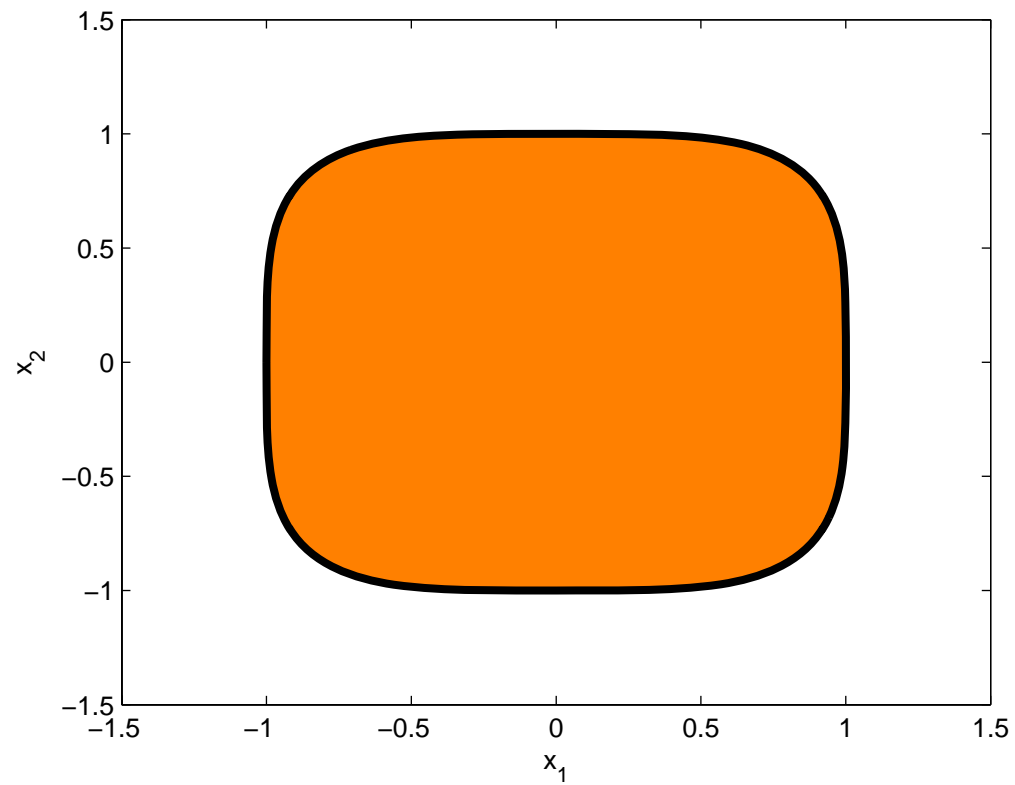
LMI

$$1 - 2x_1 - x_1^2 - x_2^2 + 2x_1^3 \geq 0 \text{ and } x_1 \leq \frac{1}{2}$$

\Leftrightarrow

$$\begin{bmatrix} 1 & x_1 & 0 \\ x_1 & 1 & x_2 \\ 0 & x_2 & 1 - 2x_1 \end{bmatrix} \succeq 0$$

LMI set or not ?



$$x_1^4 + x_2^4 \leq 1$$

