# Geometry of LMIs and determinantal representations of algebraic plane curves

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# Outline

- 1. LMI optimization
- 2. Geometry of LMI sets
- 3. Rational curves
- 4. Cubics
- 5. Dixon's construction

Linear matrix inequality

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

where  $F_i$  are given symmetric real matrices and constraint  $\succeq 0$  means positive semidefinite (all eigenvalues real nonnegative)

Arise in control theory (Lyapunov 1890, Willems 1971, Boyd et al. 1994), combinatorial optimization, finance, structural mechanics, and many other areas

Key property = convex in x

## Semidefinite programming

Decision problem

 $\begin{array}{ll} \min_{\boldsymbol{x}} & \sum_{i} c_{i} \boldsymbol{x}_{i} \\ \text{s.t.} & F_{0} + \sum_{i} \boldsymbol{x}_{i} F_{i} \succeq 0 \end{array}$ 

Optimization over LMIs = semidefinite programming, versatile generalization of linear (and convex quadratic) programming to the convex cone of positive semidefinite matrices

At given accuracy can be solved in polynomial time using interiorpoint methods (Nesterov, Nemirovski 1994)

Many public-domain solvers available

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#### Geometry of LMI sets

How does an LMI set

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

look like in Euclidean space ?

Matrix F(x) is PSD iff its diagonal minors  $f_i(x)$  are nonnegative

So the LMI set can be described as

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

a convex closed basic semialgebraic set

#### Semialgebraic formulation

For an d-by-d matrix F(x) we have  $2^d - 1$  diagonal minors

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

Apply to characteristic poly  $f(t, x) = \det(tI_d + F(x)) = \sum_{k=0}^d f_{d-k}(x)t^k$ which is monic, i.e.  $f_0(x) = 1$ 

Only *d* poly ineqs  $f_k(x) \ge 0$  to be checked

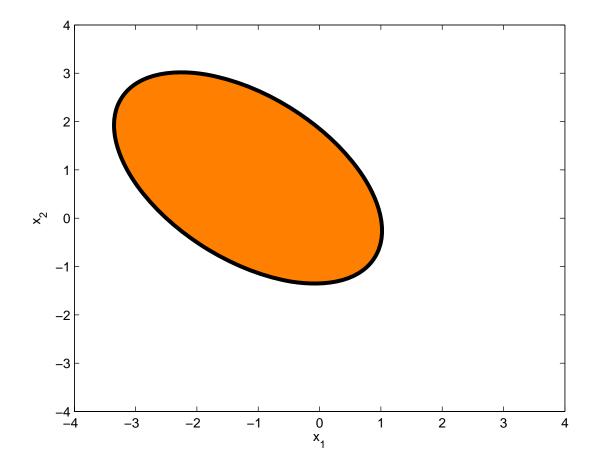
Polys  $f_k(x)$  are sums of principal minors of F(x) of order k or equivalently sums of k-term-products of eigenvalues of F(x)

#### Example of 2D LMI feasible set

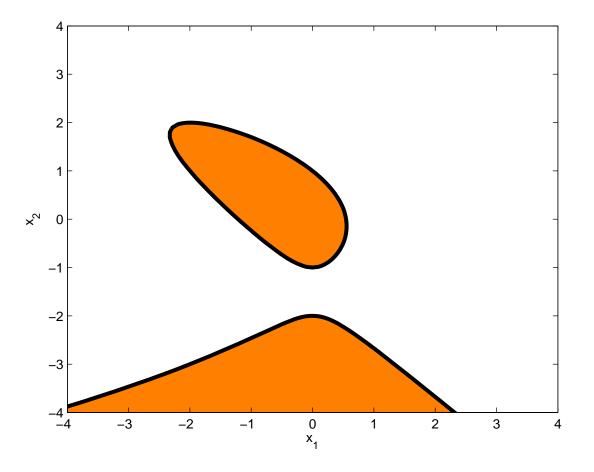
$$F(x) = \begin{bmatrix} 1 - x_1 & x_1 + x_2 & x_1 \\ x_1 + x_2 & 2 - x_2 & 0 \\ x_1 & 0 & 1 + x_2 \end{bmatrix} \succeq 0$$

System of 3 polynomial inequalities  $f_i(x) \ge 0$  $f_1(x) = \text{trace } F(x) = 4 - x_1 \ge 0$ 

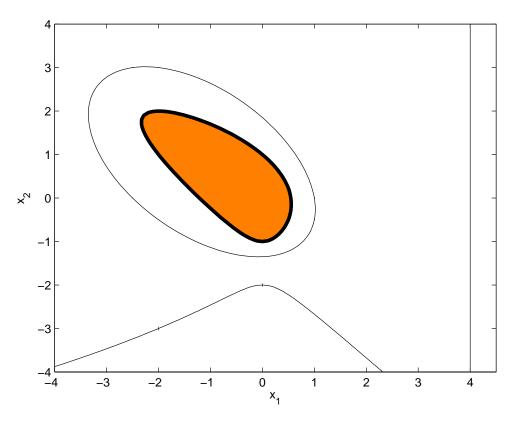
$$f_2(x) = 5 - 3x_1 + x_2 - 2x_1^2 - 2x_1x_2 - 2x_2^2 \ge 0$$



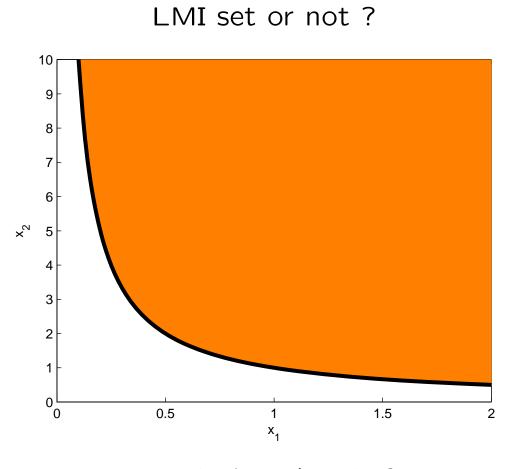
 $f_3(x) = \det F(x) = 2 - 2x_1 + x_2 - 3x_1^2 - 3x_1x_2 - 2x_2^2 - x_1x_2^2 - x_2^3 \ge 0$ 



LMI set = intersection of level-sets  $f_k(x) \ge 0$ , k = 1, 2, 3



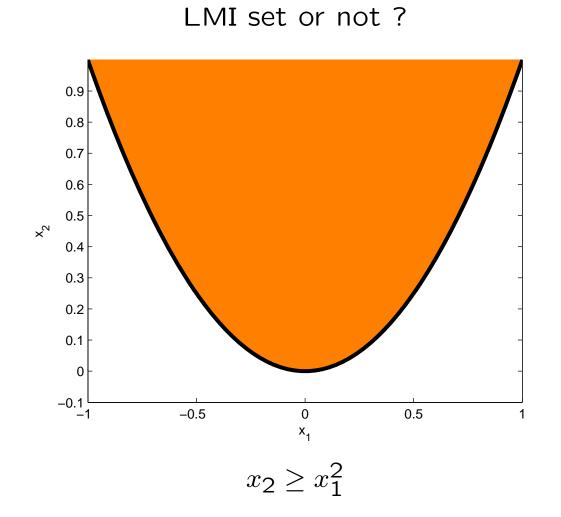
Boundary of LMI region shaped by determinant Other polys only isolate convex connected component



 $x_1x_2 \ge 1$  and  $x_1 \ge 0$ 

 $x_1x_2 \ge 1$  and  $x_1 \ge 0$ 

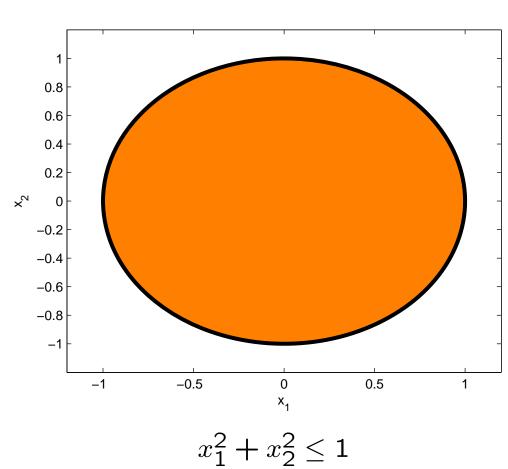
 $\iff \begin{bmatrix} x_1 & 1 \\ 1 & x_2 \end{bmatrix} \succeq 0$ 



 $x_2 \ge x_1^2$ 

 $\iff$ 

$$\left[\begin{array}{cc} 1 & x_1 \\ x_1 & x_2 \end{array}\right] \succeq 0$$

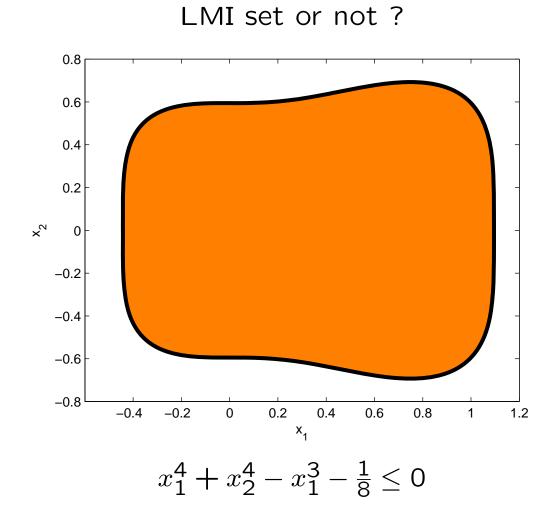


LMI set or not ?

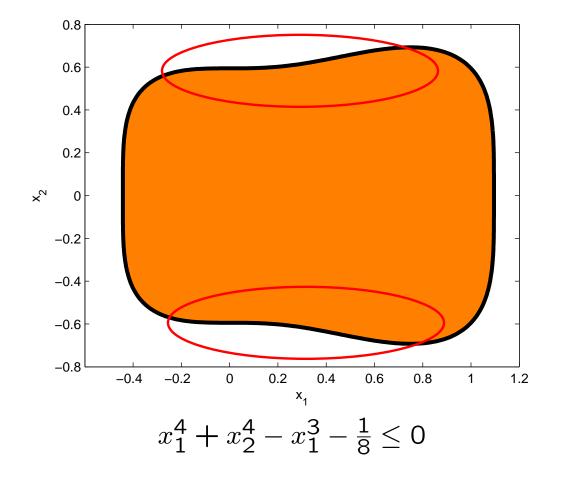
 $x_1^2 + x_2^2 \le 1$ 

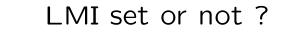
$$\iff$$

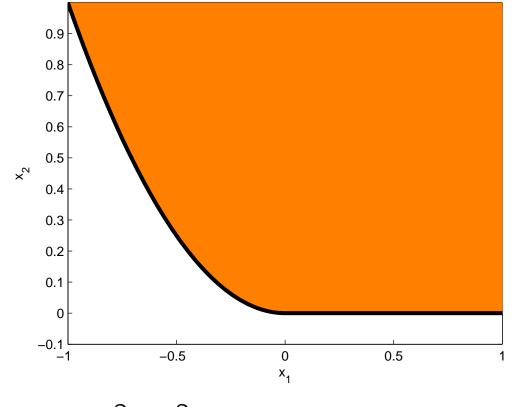
$$\begin{bmatrix} 1+x_1 & x_2 \\ x_2 & 1-x_1 \end{bmatrix} \succeq 0$$



## NOT LMI: not convex

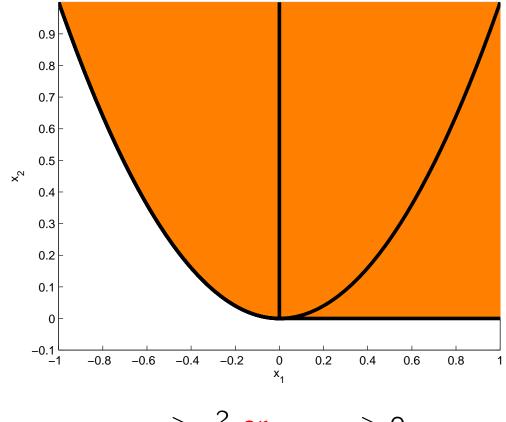




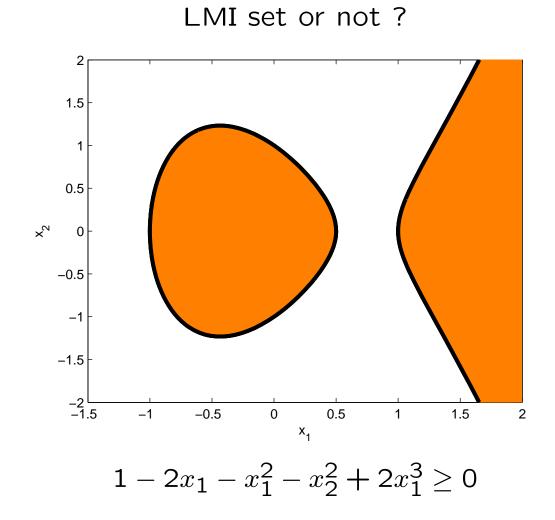


 $\{x \in \mathbb{R}^2 : t^2 + 2x_1t + x_2 \ge 0, \forall t \in \mathbb{R}\}$ 

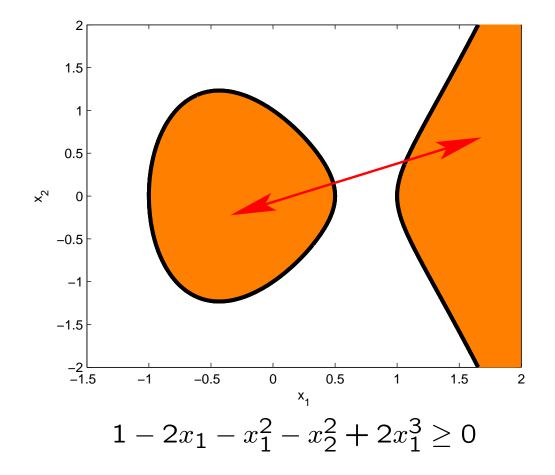
# NOT LMI: not basic semialgebraic

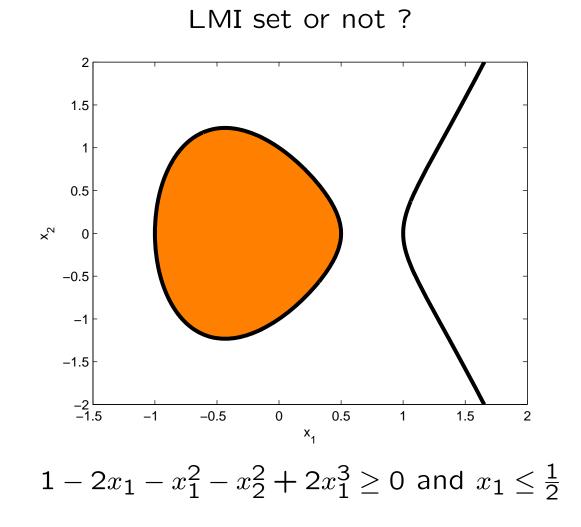


 $x_2 \ge x_1^2 \text{ or } x_1, x_2 \ge 0$ 



#### NOT LMI: not connected

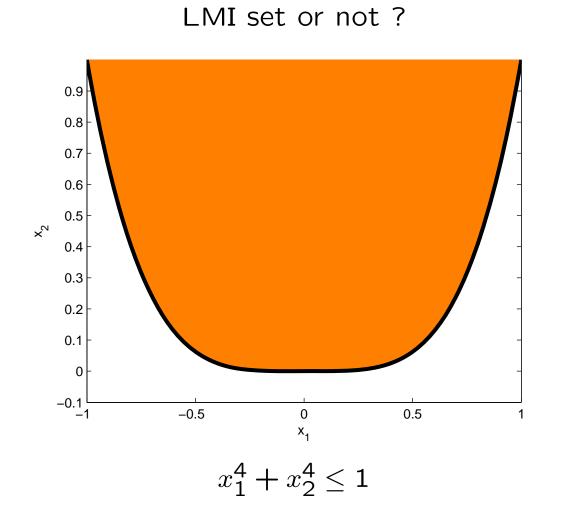




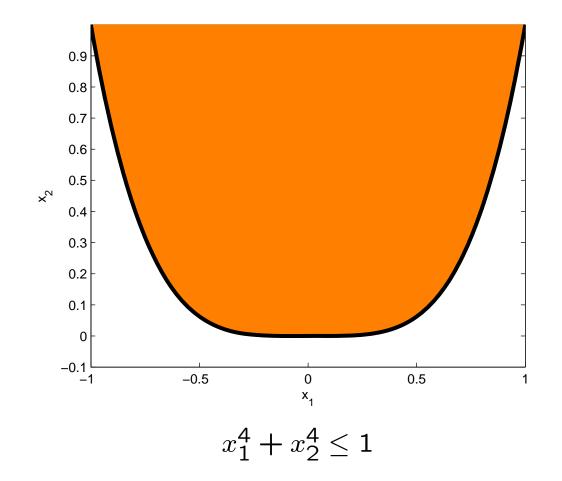
$$1 - 2x_1 - x_1^2 - x_2^2 + 2x_1^3 \ge 0$$
 and  $x_1 \le \frac{1}{2}$ 

 $\iff$ 

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



# NOT LMI: not rigidly convex



#### Determinantal representation

Consider the non-empty semialgebraic set

$$\mathcal{F} = \{ x \in \mathbb{R}^n : f(x) \ge 0 \}$$

where f(x) is a given polynomial of degree d

Without loss of generality, assume that we are given a point e (typically the origin) satisfying f(e) = 1

Since the boundary of an LMI set is shaped by a determinant, can we find symmetric real matrices  $F_i$  such that

$$F(x) = F_0 + \sum_{i=1}^n x_i F_i$$
, det  $F(x) = f(x)$ 

So we would like to find a linear symmetric determinantal representation for polynomial f(x)

#### Definite determinantal representation = LMI

Once we have det F(x) = f(x), we would like to know whether

$$\mathcal{F} = \operatorname{closure} \left\{ x \in \mathbb{R}^n : \det F(x) > 0 \right\} \ni e$$
$$= \left\{ x \in \mathbb{R}^n : F(x) \ge 0 \right\}$$

Since f(e) = 1, it holds  $e \in int \mathcal{F}$  and  $F(e) \succ 0$  so the representation must be definite for  $\mathcal{F}$  to be expressed as an LMI

Under which conditions on f can we find such a definite representation ?

Define the algebraic curve

$$\mathcal{C} = \{ x \in \mathbb{R}^n : f(x) = 0 \}$$

containing the boundary of  ${\mathcal F}$ 

# Rigid convexity

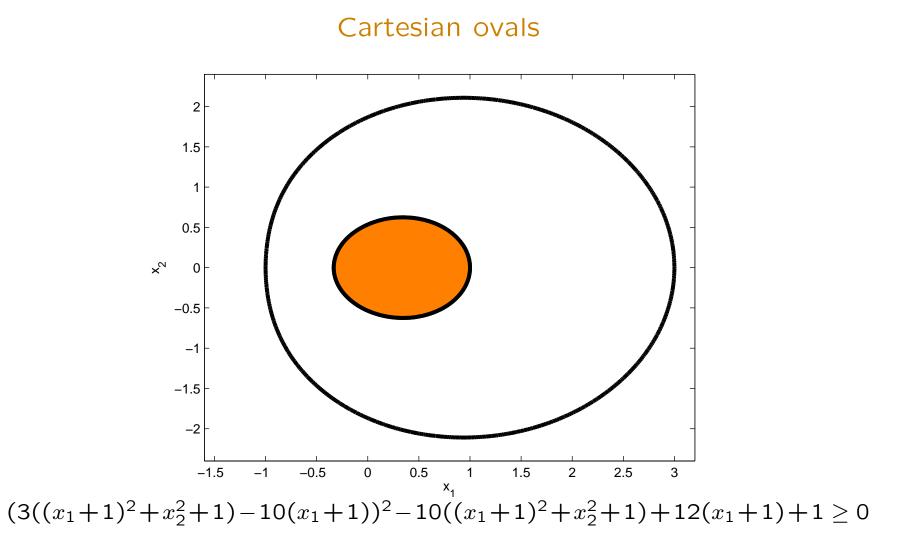
Necessary condition for  $\mathcal{F}$  to have a definite symmetric linear determinantal, or LMI representation:

Any line passing through an interior point of  $\mathcal{F}$  must intersect  $\mathcal{C}$  exactly d times (counting multiplicities and points at infinity)

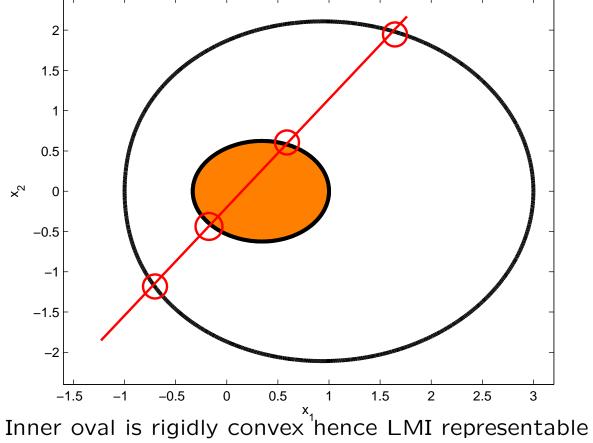
Rigid convexity implies convexity

Strong result by Helton and Vinnikov (2002): the condition is also sufficient in the plane, i.e. for n = 2

Also sufficient for n > 2 ?



## Cartesian ovals



#### Constructive methods

Checking rigid convexity amounts to checking positive semidefiniteness of the Hermite matrix of polynomial p(x) for all x

Given f(x) and e, once we know that the set

$$\mathcal{F} = \{x \in \mathbb{R}^n : f(x) \ge 0\} \ni e$$

is rigidly convex, how can we systematically build symmetric matrices  $F_i$  such that

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

and so  $f(x) = \det F(x)$ ? When/how can we enforce  $F_0 = I$ ?

In the sequel we focus exclusively on the plane case (n = 2)

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#### Rational curves

An algebraic plane curve of genus zero, that is, with a maximal number of singularities

$$\{x \in \mathbb{R}^2 : f(x) = 0\}$$

admits a rational parametrization

$$x_1(t) = \frac{f_1(t)}{f_0(t)}, \quad x_2(t) = \frac{f_2(t)}{f_0(t)}$$

with  $f_i(t)$  real polys of real indeterminate t

Degrees of  $f_i$  do not exceed degree of f

Coeffs of  $f_i$  chosen in (typically small) algebraic extension of the coeff field of f

#### Bezoutian

Determinantal representation follows from the resultant of the two polys

$$g_1(t, x_1) = f_0(t) - x_1(t)f_1(t)$$
  

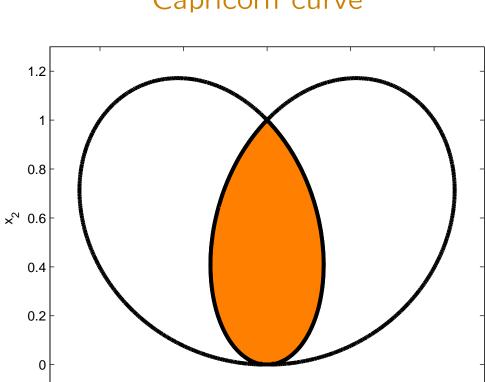
$$g_2(t, x_2) = f_0(t) - x_2(t)f_2(t)$$

with respect to t (variable to be eliminated)

Bezout matrix  $B_t(g_1, g_2)$  is symmetric and linear in x such that

$$\det B_t(g_1, g_2) = f(x)$$

hence  $F(x) = B_t(g_1, g_2)$  is a valid symmetric linear determinantal representation of f



# Capricorn curve

 $f(x) = x_1^2(x_1^2 + x_2^2) - 2(x_1^2 + x_2^2 - x_2)^2$ 

0 x<sub>1</sub>

0.5

1

-0.5

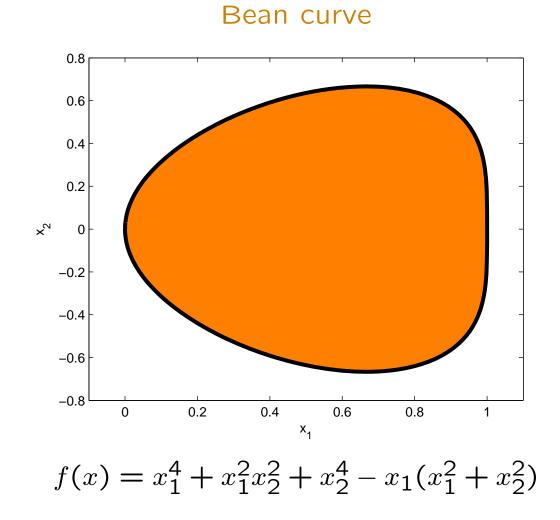
-1

#### Capricorn LMI

$$F(x) = \begin{bmatrix} 1960 - 868x - 1924y & -952 - 940x + 740y \\ -952 - 940x + 740y & 776 + 540x + 476y \\ -168 + 180x + 180y & -8 - 36x - 84y \\ 56 - 4x - 52y & -72 + 20x + 52y \end{bmatrix}$$

$$\begin{bmatrix} -168 + 180x + 180y & 56 - 4x - 52y \\ -8 - 36x - 84y & -72 + 20x + 52y \\ 40 + 60x + 92y & 8 + 20x - 28y \\ 8 + 20x - 28y & 8 - 4x - 4y \end{bmatrix}$$

Definite around e = (0, 1/2) hence LMI representable



#### Bean determinant

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

Indefinite around e = (1/2, 0)

Not rigidly convex hence not LMI representable

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#### Cubics

When deg f(x) = 3 the genus of f(x) can be 0 (rational, or singular cubic) or 1 (elliptic, or smooth cubic)

Homogeneize  $f(x_0, x_1, x_2) = x_0^3 f(\frac{x_1}{x_0}, \frac{x_2}{x_0})$ , define Hessian matrix

$$H(f(x)) = \left[\frac{\partial^2 f(x)}{\partial x_i \partial x_j}\right]_{i,j=0,1,2}$$

which is real symmetric linear, and Hessian

$$h(x) = \det H(f(x))$$

which is cubic

An elliptic curve has 9 flexes  $x^*$  (3 of which are real) satisfying

$$f(x^*) = h(x^*) = 0$$

#### Parametrized Hessian

f(x) and h(x) share the same flexes and we know a symmetric linear determinantal representation for h(x), so use linear homotopy to find one for f(x) (thanks to Frédéric Han)

For  $t \in \mathbb{R}$  define parametrized Hessian

$$g(x,t) = \det H(tf(x) + h(x))$$

and find  $t^*$  satisfying

$$g(x^*, t^*) = f(x^*)$$

at a real flex  $x^*$  by solving equation of degree 3

Three distinct determinantal representations not equivalent by congruence transformation, one of which is definite hence LMI

#### Elliptic curve

Find a symmetric linear determinantal representation for

$$f(x) = x_1^3 - x_2^2 - x_1$$

First build Hessian

$$h(x) = \det H(f(x)) = 8(x_0^3 + 3x_0x_1^2 - 3x_1x_2^2)$$

Parametrized Hessian

 $g(t,x) = \det H(tf(x) + h(x)) = 24t^3x_0x_1^2 - 576t^2x_0^2x_1 + \dots + 110592x_1^3$ matches f(x) at flex  $x_0^* = 0$  for

$$t^* \in \{0, 24, -24\}$$

yielding the following three representations...

## Elliptic curve

$$F^{1}(x) = \begin{bmatrix} 1 & -x_{2} & x_{1} \\ -x_{2} & -x_{1} & 0 \\ x_{1} & 0 & 1 \end{bmatrix}$$

$$F^{2}(x) = 4^{-\frac{1}{3}} \begin{bmatrix} 1+3x_{1} & -x_{2} & -1+x_{1} \\ -x_{2} & -1-x_{1} & -x_{2} \\ -1+x_{1} & -x_{2} & 1-x_{1} \end{bmatrix}$$

$$F^{3}(x) = 4^{-\frac{1}{3}} \begin{bmatrix} 1 - 3x_{1} & -x_{2} & 1 + x_{1} \\ -x_{2} & 1 - x_{1} & x_{2} \\ 1 + x_{1} & x_{2} & 1 + x_{1} \end{bmatrix}$$

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#### Dixon's construction

A. C. Dixon described in 1902 an explicit construction of a symmetric linear determinantal representation of a plane algebraic curve based on the knowledge of a contact curve

Given a curve f(x) = 0 of degree d, a contact curve g(x) = 0 is a curve of degree d - 1 touching f(x) = 0 at  $\frac{1}{2}d(d - 1)$  points

Once g(x) is given, a determinantal representation for f(x) follows by simple linear algebra

Better illustrated with an example..

#### Adjoint matrix

Consider the elliptic curve

$$f(x) = 1 - 2x_1 - x_1^2 - x_2^2 + 2x_1^3 = 0$$

with (known) determinantal representation

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

Consider its adjoint (or adjugate) matrix

$$V(x) = F^{-1}(x)f(x) = \begin{bmatrix} 1 - 2x_1 - x_2^2 & x_1(-1 + 2x_1) & x_1x_2 \\ x_1(-1 + 2x_1) & 1 - 2x_1 & -x_2 \\ x_1x_2 & -x_2 & 1 - x_1^2 \end{bmatrix}$$

which contains quadratic cofactors

#### Quotient ring

Since F(x)V(x) = f(x)I, the cofactors in a column of V(x) generate a basis for the quotient ring  $\mathbb{R}[x]/f(x)$ 

$$\begin{bmatrix} 1 & x_1 & 0 \\ x_1 & 1 & x_2 \\ 0 & x_2 & 1 - 2x_1 \end{bmatrix} \begin{bmatrix} 1 - 2x_1 - x_2^2 & x_1(-1+2x_1) & x_1x_2 \\ x_1(-1+2x_1) & 1 - 2x_1 & -x_2 \\ x_1x_2 & -x_2 & 1 - x_1^2 \end{bmatrix} = \begin{bmatrix} f(x) & 0 & 0 \\ 0 & f(x) & 0 \\ 0 & 0 & f(x) \end{bmatrix}$$

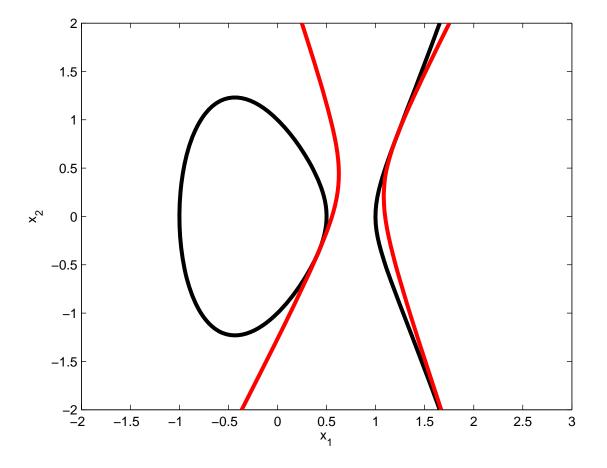
#### Net of contact curves

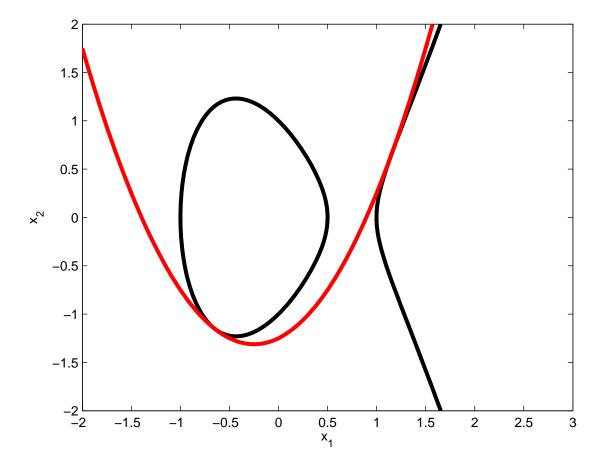
For all  $v \in \mathbb{R}^3$  the conic

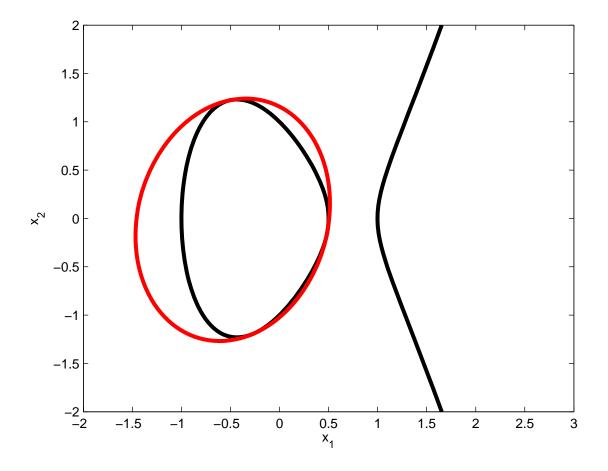
$$g(x,v) = v^T V(x)v = 0$$

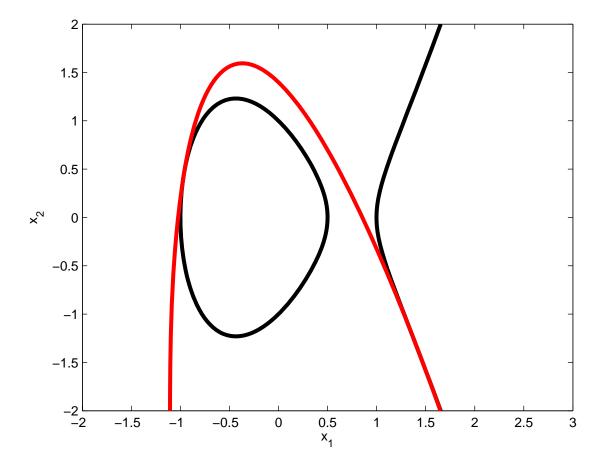
is a contact curve touching f(x) = 0 at 3 points

Here are some random contact conics for our cubic..

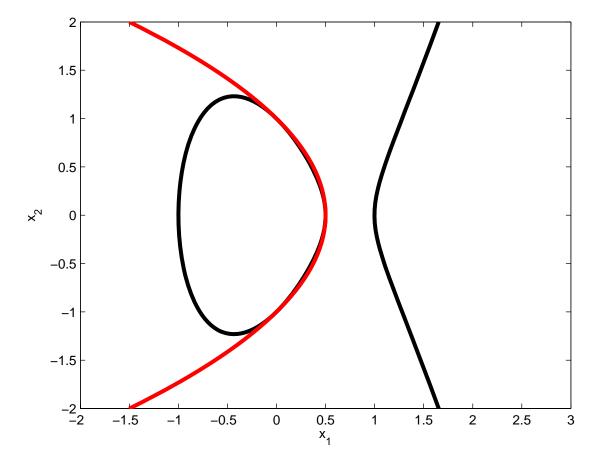








## Contact parabola



Let (1,1) be the contact parabola

$$V(x) = \begin{bmatrix} 1 - 2x_1 - x_2^2 & ? & ? \\ x_1(-1 + 2x_1) & ? & ? \\ x_1x_2 & ? & ? \end{bmatrix}$$

Build basis for quotient ring  $\mathbb{R}[x]/f(x)$ 

$$V(x) = \begin{bmatrix} 1 - 2x_1 - x_2^2 & x_1(-1 + 2x_1) & x_1x_2 \\ x_1(-1 + 2x_1) & ? & ? \\ x_1x_2 & ? & ? \end{bmatrix}$$

All these contact conics share common touching points

$$V(x) = \begin{bmatrix} 1 - 2x_1 - x_2^2 & x_1(-1 + 2x_1) & x_1x_2 \\ x_1(-1 + 2x_1) & 1 - 2x_1 & ? \\ x_1x_2 & ? & ? \end{bmatrix}$$
(2.2) shares contact points with (1.1) and (2.1)

(2,2) shares contact points with (1,1) and (2,1)

$$V(x) = \begin{bmatrix} 1 - 2x_1 - x_2^2 & x_1(-1 + 2x_1) & x_1x_2 \\ x_1(-1 + 2x_1) & 1 - 2x_1 & ? \\ x_1x_2 & ? & 1 - x_1^2 \end{bmatrix}$$
  
(3,3) shares contact points with (1,1) and (3,1)

$$V(x) = \begin{bmatrix} 1 - 2x_1 - x_2^2 & x_1(-1 + 2x_1) & x_1x_2 \\ x_1(-1 + 2x_1) & 1 - 2x_1 & -x_2 \\ x_1x_2 & -x_2 & 1 - x_1^2 \end{bmatrix}$$
  
(3,2) shares contact points with (2,2) and (3,3)

### Inverting the adjoint

$$f(x)V^{-1}(x) = \begin{bmatrix} 1 & x_1 & 0 \\ x_1 & 1 & x_2 \\ 0 & x_2 & 1 - 2x_1 \end{bmatrix}$$

Invert adjoint matrix to recover linear representation

### Dixon's construction

Given f(x) of degree d, and a contact curve g(x) of degree d-1:

- let g(x) be a diagonal entry in V(x)
- derive the whole row and column in V(x) by generating a basis of the quotient ring  $\mathbb{R}[x]/f(x)$
- derive the remaining entries in V(x) by solving linear equations in the quotient ring

• compute 
$$F(x) = f^{d-2}(x)V(x)^{-1}$$

A key issue remains: how can we algebraically and systematically find a contact curve g(x) ?

Impact of the choice of the contact points ?

Definiteness of the representation ?

### Open problems

For genus zero curves:

- how many distinct parametrizations ?
- how can we detect/enforce definiteness ?

For positive genus curves:

- how can we find a contact curve ?
- how can we detect/enforce definiteness ?
- for quartics, should we use the bitangents (Edge 1935) ?
- can we singularize the curve (inverse quadratic mappings of Hilbert/Hurwitz to solve singularities) and use the Bezoutian ?

# Example of 3D LMI set

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

