

## Linear Matrix Inequalities vs Convex Sets

Harry Dym	Weitzman Inst.
Damon Hay	Florida → Texas
Igor Klep	Everywhere in Solvenia
Scott McCullough	University of Florida
Victor Vinnikov	Ben Gurion U of the Negev
I am Bill Helton	UCSD

**Advertisement:** Try noncommutative computation

**NCA**lgebra<sup>1</sup>  
**NCS**o**ST**ools<sup>2</sup>

---

<sup>1</sup> Helton, deOliveira (UCSD), Stankus (CalPoly SanLobispo ), Miller

<sup>2</sup> Igor Klep

# Outline

Semidefinite Programming and LMIs

Linear Systems give nc poly inequalities

LMIs and Convex Sets

NC Real Algebraic Geometry: Positivstellensatz

Change of Variables to achieve Free Convexity

- nc maps

- Proper nc maps

- Pencil balls and maps

- NC Proper maps are bianalytic

Summary

## SDP and LMIs

A **linear pencil** is a matrix valued function  $L$  of the form

$$L(\mathbf{x}) := L_0 + L_1 \mathbf{x}_1 + \cdots + L_g \mathbf{x}_g,$$

where  $L_0, L_1, L_2, \dots, L_g$  are symmetric matrices and  $\mathbf{x} := \{\mathbf{x}_1, \dots, \mathbf{x}_g\}$  are  $m$  real parameters.

A **Linear Matrix Inequality (LMI)** is one of the form:

$$L(\mathbf{x}) \succeq 0.$$

The set of solutions

$$\mathcal{G} := \{(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_g) : L_0 + L_1 \mathbf{x}_1 + \cdots + L_g \mathbf{x}_g \text{ is PosSD}\}$$

is a convex set. Solutions can be found numerically for problems of modest size. This is called

**Semidefinite Programming**    **SDP**

Notation: a monic LMI is one with  $L_0 = I$ .

# SDP is Everywhere

**Main development in convex programming in the last 20 years.**

- Many problems from many areas can be converted to SDPs.
- Max -Cut, etc, etc is approximable by SDP
- Linear systems theory, classic problems convert to Linear Matrix Inequalities LMIs. Compromises opens many uses of SDP
- Computational Real Algebraic Geometry RAG (RAG is critical in this talk)
- Lyapunov functions for Nonlin Sys.      Statistics
- Types of Chemical Problems      **Many others**

See Nemirovski's Plenary ICM address 2006

"Semidefinite Programming" gets about **70,000 hits on google**

"Symmetric Matrix" gets about **300,000 hits on google**

In **linear engineering systems** a special class of SDP is fundamental: SDP with **matrix (not scalar) variables**.

OLD FASHION ORDINARY **LMI's**  
COMMUTATIVE VARIABLES  
**NOT DIM FREE**

slLMIrepVVPart

# Which Sets have LMI REPRESENTATIONS?

**QUESTION (Vague):**

**ARE CONVEX PROBLEMS ALL TREATABLE WITH LMI's?**

**DEFINITION:**

A set  $\mathcal{C} \subset \mathbb{R}^g$  has an **Linear Matrix Inequality (LMI) Representation** provided that there are symmetric matrices  $L_0, L_1, L_2, \dots, L_g$  for which the **Linear Pencil**,  $L(\mathbf{x}) := L_0 + L_1\mathbf{x}_1 + \dots + L_g\mathbf{x}_g$ , has positivity set,

$$\mathcal{G} := \{(\mathbf{x}_1, \mathbf{x}_2, \dots, \mathbf{x}_g) : L_0 + L_1\mathbf{x}_1 + \dots + L_g\mathbf{x}_g \text{ is PosSD}\}$$

equals the set  $\mathcal{C}$ ; that is,  $\mathcal{C} = \mathcal{G}$ .

**Which convex sets have an LMI representation?**

Parrilo and Sturmfels preprint 2000 (open for  $g = 2$ ).

## EXAMPLE

$$\mathcal{C} := \{(\mathbf{x}_1, \mathbf{x}_2) : 1 + 2\mathbf{x}_1 + 3\mathbf{x}_2 - (3\mathbf{x}_1 + 5\mathbf{x}_2)(3\mathbf{x}_1 + 2\mathbf{x}_2) \geq 0\}$$

has the LMI Rep

$$\mathcal{C} = \{\mathbf{x} : \mathbf{L}(\mathbf{x}) \succeq 0\} \quad \text{here } \mathbf{x} := (\mathbf{x}_1, \mathbf{x}_2)$$

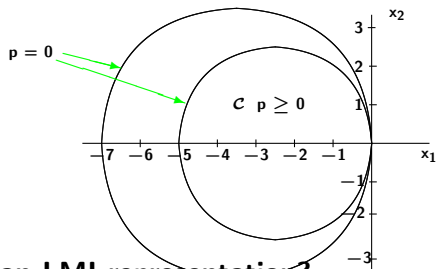
with

$$\mathbf{L}(\mathbf{x}) = \begin{pmatrix} 1 + 2\mathbf{x}_1 + 3\mathbf{x}_2 & 3\mathbf{x}_1 + 5\mathbf{x}_2 \\ 3\mathbf{x}_1 + 2\mathbf{x}_2 & 1 \end{pmatrix}$$

Pf: The determinant of  $\mathbf{L}(\mathbf{x})$  is pos iff  $\mathbf{L}(\mathbf{x})$  is PosSD.

## QUESTION 1

Does this set  $\mathcal{C}$  which is the inner component of



have an LMI representation?

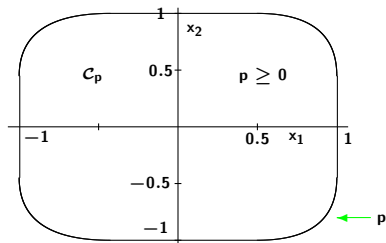
$$p(x_1, x_2) = (x_1^2 + x_2^2)(x_1^2 + x_2^2 + 12x_1 - 1) + 36x_1^2 \geq 0$$

$$\mathcal{C} := \text{inner component of } \{x \in \mathbb{R}^2 : p(x) \geq 0\}$$



## QUESTION 2

Does this set have an LMI representation?



$$p(\mathbf{x}_1, \mathbf{x}_2) = 1 - \mathbf{x}_1^4 - \mathbf{x}_2^4 \geq 0$$

$\mathcal{C}_p := \{\mathbf{x} \in \mathbb{R}^2 : p(\mathbf{x}) \geq 0\}$  has degree 4.

# Rigid Convexity

DEFINE: A convex set  $\mathcal{C}$  in  $\mathbb{R}^g$  with minimal degree (denote degree by  $d$ ) defining polynomial  $p$  to be **rigidly convex** provided

for every point  $x^0$  in  $\mathcal{C}$  and every line  $\ell$  through  $x^0$  (except for maybe a finite number of lines), the line  $\ell$  intersects the the zero set  $\{x \in \mathbb{R}^g : p(x) = 0\}$  of  $p$  in exactly  $d$  points<sup>3</sup>.

The “line test”

---

<sup>3</sup>In this counting one ignores lines which go thru  $x^0$  and hit the boundary of  $\mathcal{C}$  at  $\infty$ .

**DEF: A REAL ZERO polynomial in  $g$  variables satisfies for each  $x \in \mathbb{R}^g$ ,**

**(RZ)  $p(\mu x) = 0$  implies  $\mu$  is real**

**LEMMA: If  $\mathcal{C}$  contains  $0$  in its interior, and  $p$  is the minimal degree defining polynomial for  $\mathcal{C}$ , then  $p$  satisfies the Real Zeroes Condition if and only if  $\mathcal{C}$  is rigidly convex.**

**If  $\mathcal{C}$  does not contain  $0$  shift it.**

## IN $\mathbb{R}^2$ RIGID CONVEXITY RULES

**THM** (Vinnikov + H). **IF**  $\mathcal{C}$  is a closed convex set in  $\mathbb{R}^g$  with an LMI representation, **THEN**  $\mathcal{C}$  is rigidly convex.

**When  $g = 2$** , the converse is true, namely, a rigidly convex degree  $d$  set has a LMI representation with symmetric matrices  $L_j \in \mathbb{R}^{d \times d}$ .

The Proof of necessity is trivial. The Proof of sufficiency ( $g = 2$ ) is not at all elementary. Uses algebraic geometry methods of Vinnikov. These have been refined and extended by Vinnikov and Joe Ball.

# NC (free) LMI

## LMI in Matrix Variables

### OUTLINE

Semidefinite Programming and LMIs

Linear Systems give nc poly inequalities

LMIs and Convex Sets

NC Real Algebraic Geometry: Positivstellensatz

Change of Variables to achieve Free Convexity

- nc maps

- Proper nc maps

- Pencil balls and maps

- NC Proper maps are bianalytic

Summary

## Systems problems $\rightarrow$ Matrix Ineq



Many such problems Eg.  $H^\infty$  control

Freq. domain	<b>Nev-Pick</b>	1975-85
State space	<b>tricks</b>	Riccatis 1985-95
Matrix ineqs	<b>tricks</b>	LMIs 1990 -

Solve numerical LMIs, up to  $50 \times 50$  (unstructured) matrices, via semidef prog., bundle methods, etc.

$$\left. \begin{array}{l} F_1(\mathbf{X}) := \mathbf{A}\mathbf{X} + \mathbf{X}\mathbf{A}^T + \mathbf{B}\mathbf{B}^T \\ F_2(\mathbf{X}, \mathbf{Y}) \end{array} \right\} \begin{array}{l} \gamma \\ 0 \end{array} \quad \text{PosSD}$$

# MATRIX INEQUALITIES

Polynomial or Rational function of matrices are PosSDef.

Example: Get Riccati expressions like

$$\mathbf{A}\mathbf{X} + \mathbf{X}\mathbf{A}^T - \mathbf{X}\mathbf{B}\mathbf{B}^T\mathbf{X} + \mathbf{C}\mathbf{C}^T \succ 0$$

OR Linear Matrix Inequalities (LMI) like

$$\begin{pmatrix} \mathbf{A}\mathbf{X} + \mathbf{X}\mathbf{A}^T + \mathbf{C}^T\mathbf{C} & \mathbf{X}\mathbf{B} \\ \mathbf{B}^T\mathbf{X} & \mathbf{I} \end{pmatrix} \succ 0$$

which is equivalent to the Riccati inequality.

ncDimlessPartB

# $H^\infty$ Control

## ALGEBRA PROBLEM:

Given the polynomials:

$$H_{xx} = E_{11} A + A^T E_{11} + C_1^T C_1 + E_{12}^T b C_2 + C_2^T b^T E_{12}^T + E_{11} B_1 b^T E_{12}^T + E_{11} B_1 B_1^T E_{11} + E_{12} b b^T E_{12}^T + E_{12} b B_1^T E_{11}$$

$$H_{xz} = E_{21} A + \frac{a^T (E_{21} + E_{12}^T)}{2} + c^T C_1 + E_{22} b C_2 + c^T B_2^T E_{11}^T + \frac{E_{21} B_1 b^T (E_{21} + E_{12}^T)}{2} + E_{21} B_1 B_1^T E_{11}^T + \frac{E_{22} b b^T (E_{21} + E_{12}^T)}{2} + E_{22} b B_1^T E_{11}^T$$

$$H_{zx} = A^T E_{21}^T + C_1^T c + \frac{(E_{12} + E_{21}^T) a}{2} + E_{11} B_2 c + C_2^T b^T E_{22}^T + E_{11} B_1 b^T E_{22}^T + E_{11} B_1 B_1^T E_{21}^T + \frac{(E_{12} + E_{21}^T) b b^T E_{22}^T}{2} + \frac{(E_{12} + E_{21}^T) b B_1^T E_{21}^T}{2}$$

$$H_{zz} = E_{22} a + a^T E_{22}^T + c^T c + E_{21} B_2 c + c^T B_2^T E_{21}^T + E_{21} B_1 b^T E_{22}^T + E_{21} B_1 B_1^T E_{21}^T + E_{22} b b^T E_{22}^T + E_{22} b B_1^T E_{21}^T$$

(PROB)  $A, B_1, B_2, C_1, C_2$  are knowns.

Solve the inequality  $\begin{pmatrix} H_{xx} & H_{xz} \\ H_{zx} & H_{zz} \end{pmatrix} \preceq 0$  for unknowns

$a, b, c$  and for  $E_{11}, E_{12}, E_{21}$  and  $E_{22}$



**Engineering problems defined  
entirely by signal flow diagrams  
and  $L^2$  performance specs  
are equivalent to  
Polynomial Matrix Inequalities**

How and why is a long story but the correspondence <sup>4</sup> between linear systems and noncommutative algebra is on **the next slides:**

---

<sup>4</sup>Ask after the talk if you would like more detail.

# Notation

$\mathbf{x} = (x_1, \dots, x_g)$  algebraic noncommuting variables

$p(\mathbf{x})$  and nc polynomial

Eg.  $p(\mathbf{x}) = x_1x_2 + x_2x_1$

$\mathbf{X} = (X_1, \dots, X_g)$  a tuple of matrices.

Substitute a matrix for each variable

Eg.  $p(\mathbf{X}) = X_1X_2 + X_2X_1$

## Linear Systems and Algebra Synopsis

A Signal Flow Diagram with  $L^2$  based performance, eg  $H^\infty$  gives precisely a nc polynomial

$$p(\mathbf{a}, \mathbf{x}) := \begin{pmatrix} p_{11}(\mathbf{a}, \mathbf{x}) & \cdots & p_{1k}(\mathbf{a}, \mathbf{x}) \\ \vdots & \ddots & \vdots \\ p_{k1}(\mathbf{a}, \mathbf{x}) & \cdots & p_{kk}(\mathbf{a}, \mathbf{x}) \end{pmatrix}$$

The linear systems problem becomes exactly:

Given matrices  $\mathbf{A}$ .

Find matrices  $\mathbf{X}$  so that  $P(\mathbf{A}, \mathbf{X})$  is PosSemiDef.

WHY? Turn the crank using quadratic storage functions.

**BAD** Typically  $p$  is a mess, until a hundred people work on it and maybe **convert it to convex Matrix Inequalities**.

**QUESTIONS (Vague) :**

**WHICH SUCH PROBLEMS "ARE" LMI PROBLEMS.**

Clearly, such a problem must be convex and "semialgebraic".

**Which convex nc problems are NC LMIS?**

**WHICH PROBLEMS ARE TREATABLE WITH LMI's?**

This requires some kind of **change of variables theory.**

**ARE LMIs INTERESTING MATHEMATICAL OBJECTS?**

**Free Real Algebraic Geometry: Polynomials positive on NC  
Convex sets**

**These are topics of this talk**

**We shall suppress the two classes of variables  $a, x$ .**

# Outline

Semidefinite Programming and LMIs

Linear Systems give nc poly inequalities

LMIs and Convex Sets

NC Real Algebraic Geometry: Positivstellensatz

Change of Variables to achieve Free Convexity

- nc maps

- Proper nc maps

- Pencil balls and maps

- NC Proper maps are bianalytic

Summary

# NC (FREE) CONVEX SETS

## SETS

HOW DOES NC CONVEXITY compare to classical CONVEXITY?

**Q?** Can we treat many more problems with convex techniques than LMI techniques?

Scott McCullough + H

## NC SEMIALGEBRAIC SET

**EXAMPLE:**  $p$  is a symmetric NC polynomial in  $g$  variables  
 $p(0_n) = I_n$

$\mathcal{D}_p^n :=$  component of  $0$  of

$$\{\mathbf{X} \in (\mathbb{SR}^{n \times n})^g : p(\mathbf{X}) \succ 0\}$$

$\mathcal{D}_p :=$  Positivity Domain  $= \cup_n \mathcal{D}_p^n$ .

$p$  is a defining polynomial for the domain  $\mathcal{D}_p$ .

**Example:**  $\mathbf{x} = (x^1, x^2)$  and  $\mathbf{X} = (X^1, X^2)$

$$p = 1 - x_1^4 - x_2^4$$

$$\mathcal{D}_p^2 = \{\mathbf{X} \in (\mathbb{SR}^{2 \times 2})^2 : 1 - X_1^4 - X_2^4 \succ 0\}$$

## NC BASIC SEMIALGEBRAIC SET

$\mathbf{p}$  is a symmetric  $\delta \times \delta$ - matrix of NC polynomials  
 $\mathbf{p}(0_n)$  is Pos Def. Eg.

$$\mathbf{p} = \begin{pmatrix} \mathbf{p}_{11}(\mathbf{x}) & \mathbf{p}_{12}(\mathbf{x}) \\ \mathbf{p}_{21}(\mathbf{x}) & \mathbf{p}_{22}(\mathbf{x}) \end{pmatrix}$$

$\mathcal{D}_p^n :=$  component of 0 of

$$\{\mathbf{X} \in \mathbb{S}\mathbb{R}^{n \times n} : \mathbf{p}(\mathbf{X}) \succ 0\}$$

$\mathcal{D}_p :=$  Positivity Domain =  $\cup_n \mathcal{D}_p^n$ .

A **NC basic open semialgebraic set** is one of the form  $\mathcal{D}_p$ ,  
 $\mathbf{p}$  is called a defining polynomial for  $\mathcal{D}_p$ .



# NC LIN MATRIX INEQS vs NC CONVEXITY

## NC LINEAR MATRIX INEQUALITIES LMIs

**GIVEN** a linear pencil  $L(\mathbf{x}) := L_0 + L_1 x_1 + \cdots + L_g x_g$   
symmetric matrices

**FIND** matrices  $\mathbf{X} := \{X_1, X_2, \cdots, X_g\}$  making  $L(\mathbf{X})$   
Pos SemiDef.  $L(\mathbf{X}) := L_0 + L_1 \otimes X_1 + \cdots + L_g \otimes X_g$

## QUESTION:

Which Convex nc semialgebraic sets have an NC LMI representation?

What is the NC line test?

## LMI REPS for NC CONVEX SETS

**THM:** McCullough-H

**SUPPOSE**  $p$  is an NC symmetric polynomial,  $p(0) \succ 0$

**IF**  $\mathcal{D}_p^n := \{\mathbf{X} \in \mathbb{S}\mathbb{R}^{n \times n} : p(\mathbf{X}) \succ 0\}$  is “convex” for all  $n$ ,

**IF**  $\mathcal{D}_p$  is bounded.

**THEN** there is a monic linear pencil  $L(x)$  making

$$\mathcal{D}_p = \mathcal{D}_L.$$

**THM says:** Every bounded nc basic open semialgebraic set has an NC LMI representation.

**Proof:** Long; its for another time.

Uses mostly recent methods NC Real Algebraic Geometry.

Separating one point with an LMI is pretty easy.

**THE HARD PROBLEM:** find a finite set of separating L's. □

# LMI's GIVE NC SEPARATING HYPERPLANES

THM: (Winkler + Effros)  $\sim$

**GIVEN** A bounded closed “NC convex set”  $\mathcal{C}$ .

Consider the  $n \times n$  matrix level of this set  $\mathcal{C}_n \subset \mathbb{R}^{n \times n}$ . If

$M = (M_1, \dots, M_g)$  is in  $\partial\mathcal{C}$ , then there is a linear pencil  $L(x)$  with

$$L(M) \text{ not invertible and } L(X) \succ 0$$

for all  $X$  inside  $\mathcal{C}$ . Here  $X$  can be of any dimension.

Pf: Complete Positivity - Straight forward Arvesonism, 1970ish.

# NC Real Algebraic Geometry

## NC convexity yields a Perfect Positivstellensatz

**OUTLINE** Joint work with Klep and McCullough

Semidefinite Programming and LMIs

Linear Systems give nc poly inequalities

LMIs and Convex Sets

NC Real Algebraic Geometry: Positivstellensatz

Change of Variables to achieve Free Convexity

- nc maps

- Proper nc maps

- Pencil balls and maps

- NC Proper maps are bianalytic

Summary

# NC convex Positivstellensatz

It is perfect

**THM**    **SUPPOSE**  $L$  is a monic linear pencil;  $\mathcal{D}_L$  has interior.

$L(X) \succeq 0$  implies  $p(X) \succeq 0$     **IFF**

$$p(x) = s(x)^* s(x) + \sum_j^{\text{finite}} f_j(x)^* L(x) f_j(x),$$

where  $s, f_j$  are vectors of noncommutative polynomials each of degree no greater than  $\left\lceil \frac{\deg(p)}{2} \right\rceil_+$ .

# Compare to commutative PosSS

Commutativity has blemishes

This result contrasts sharply with the commutative setting:

**THM (COMMUTATIVE)**

**IF**  $L(X) \succeq 0$  implies  $p(X) \succ 0$ , **THEN**

$$p(x) = s(x)^* s(x) + \sum_j^{\text{finite}} f_j(x)^* L(x) f_j(x),$$

where  $s, f_j$  are vectors of polynomials each of ( **high** ) degree with bound depending on  $\frac{1}{\min_{\{x:L(x)\succeq 0\}} p(x)}$ .

Thus we call the nc convex PosSS a “ **Perfect PosSS** ”

## Duality: PosSS $\longleftrightarrow$ Moment Problems

Matrices of Moments have Hankel structure

The main ingredient of the proof is an analysis of rank preserving extensions of truncated noncommutative Hankel matrices.

**Voilà**

## nc Hankel Matrices

$$\begin{bmatrix} 1 & x_1 & x_2 & x_1^2 & x_1x_2 & x_2^2 \\ x_1 & x_1^2 & x_1x_2 & x_1^3 & x_1^2x_2 & x_1x_2^2 \\ x_2 & x_1x_2 & x_2^2 & x_1^2x_2 & x_1x_2^2 & x_2^3 \\ x_1^2 & x_1^3 & x_1^2x_1 & x_1^4 & x_1^3x_2 & x_1^2x_2^2 \\ x_1x_2 & x_1^2x_2 & x_1x_2^2 & x_1^3x_2 & x_1^2x_2^2 & x_1x_2^3 \\ x_2^2 & x_1x_2^2 & x_2^3 & x_1^2x_2^2 & x_1x_2^3 & x_2^4 \end{bmatrix} \\ = \text{first column} \times (\text{first column})^T$$

**Hankel:** Replace each monomial  $x^\alpha$  by a number  $y_\alpha = \ell(x^\alpha)$ ,  
we get:

$$\text{Hank} := \begin{bmatrix} 1 & y_{10} & y_{01} & y_{20} & y_{11} & y_{02} \\ y_{10} & y_{20} & y_{11} & y_{30} & y_{21} & y_{12} \\ y_{01} & y_{11} & y_{02} & y_{21} & y_{12} & y_{03} \\ y_{20} & y_{30} & y_{21} & y_{40} & y_{31} & y_{22} \\ y_{11} & y_{21} & y_{12} & y_{31} & y_{22} & y_{13} \\ y_{02} & y_{12} & y_{03} & y_{22} & y_{13} & y_{04} \end{bmatrix}$$

**True confession-** the Hankels above are commutative.  
**nc Hankels** need lots of laptop battery on the plane



## The nc world is flat

**THM** Every positive definite noncommutative **Hankel** matrix has a “1-step flat” extension.

meaning

Given  $y_{ij}$  up to a certain degree, find the  $y_{ij}$  making

$$\text{Hank} := \begin{bmatrix} 1 & y_{10} & y_{01} & y_{20} & y_{11} & y_{02} \\ y_{10} & y_{20} & y_{11} & y_{30} & y_{21} & y_{12} \\ y_{01} & y_{11} & y_{02} & y_{21} & y_{12} & y_{03} \\ y_{20} & y_{30} & y_{21} & y_{40} & y_{31} & y_{22} \\ y_{11} & y_{21} & y_{12} & y_{31} & y_{22} & y_{13} \\ y_{02} & y_{12} & y_{03} & y_{22} & y_{13} & y_{04} \end{bmatrix}$$

**Hank** is PosSemiDef,  
and  
rank **Hankel** = rank **Hank**.

## Proof of free convex PosSS

**Patch flat extension for nc Hankel theorem together with the usual Hahn Banach separating linear functional plus GNS (Putinar) proof.**

**Long story.**

# Changing variables to achieve NC convexity

Semidefinite Programming and LMIs

Linear Systems give nc poly inequalities

LMIs and Convex Sets

NC Real Algebraic Geometry: Positivstellensatz

Change of Variables to achieve Free Convexity

- nc maps

- Proper nc maps

- Pencil balls and maps

- NC Proper maps are bianalytic

Summary

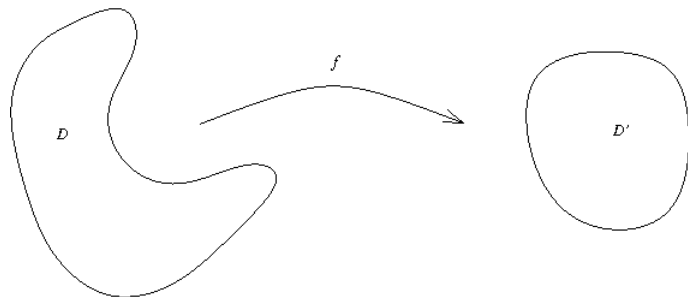
# Change of variables

## Main issues

- ▶ Which sets can be transformed into nc convex sets?

**Equivalently**

- ▶ Which sets can be transformed into nc LMI domains?



Based on joint work with Igor Klep and Scott McCullough.

# NC (free) analytic maps

Change of variables

(1) **Analytic nc polynomials have no  $x^*$ .**

# NC (free) analytic maps

Change of variables

- (1) Analytic nc polynomials have no  $\mathbf{x}^*$ .
- (2)  $\mathbf{f}(\mathbf{x}_1) = \mathbf{x}_1^*$  is **not** an nc analytic map.

# NC (free) analytic maps

Change of variables

- (1) Analytic nc polynomials have no  $x^*$ .
- (2)  $f(x_1) = x_1^*$  is **not** an nc analytic map.
- (3) Convergent **power series**

$$f = \sum_{w \in \langle x \rangle} c_w w, \quad c_w \in \mathbb{C}$$

are **nc analytic maps**. Here  $w$  are words in  $x_j$ .

Convergence works manageably, but we do not discuss convergence in this talk. **Think polynomials!** See Dan Voiculescu, Dima KV and Victor V, or see our HKMc paper for a light version

$$f = \begin{pmatrix} f_1 \\ \vdots \\ f_{\tilde{g}} \end{pmatrix}$$

# Change of variables

Proper nc maps

$$M(\mathbb{C})^g := \bigcup_n (\mathbb{C}^{n \times n})^g$$

Let  $\mathcal{U} \subset M(\mathbb{C})^g$  and  $\mathcal{V} \subset M(\mathbb{C})^{\tilde{g}}$  be given nc basic semialgebraic domains.

- ▶ An nc map  $f : \mathcal{U} \rightarrow \mathcal{V}$  is **proper** if each  $f[n] : \mathcal{U}(n) \rightarrow \mathcal{V}(n)$  is proper. That is, if  $K \subset \mathcal{V}(n)$  is compact, then  $f^{-1}(K)$  is compact.



# Change of variables

Proper nc maps

$$M(\mathbb{C})^g := \bigcup_n (\mathbb{C}^{n \times n})^g$$

Let  $\mathcal{U} \subset M(\mathbb{C})^g$  and  $\mathcal{V} \subset M(\mathbb{C})^g$  be given nc basic semialgebraic domains.

- ▶ An nc map  $f : \mathcal{U} \rightarrow \mathcal{V}$  is **proper** if each  $f[n] : \mathcal{U}(n) \rightarrow \mathcal{V}(n)$  is proper. That is, if  $K \subset \mathcal{V}(n)$  is compact, then  $f^{-1}(K)$  is compact.
- ▶ In other words: for all  $n$ , if  $(z_j)$  is a sequence in  $\mathcal{U}(n)$  and  $z_j \rightarrow \partial\mathcal{U}(n)$ , then  $f(z_j) \rightarrow \partial\mathcal{V}(n)$ .

# Change of variables

Proper nc maps

$$M(\mathbb{C})^g := \bigcup_n (\mathbb{C}^{n \times n})^g$$

Let  $\mathcal{U} \subset M(\mathbb{C})^g$  and  $\mathcal{V} \subset M(\mathbb{C})^{\tilde{g}}$  be given nc basic semialgebraic domains.

- ▶ An nc map  $f : \mathcal{U} \rightarrow \mathcal{V}$  is **proper** if each  $f[n] : \mathcal{U}(n) \rightarrow \mathcal{V}(n)$  is proper. That is, if  $K \subset \mathcal{V}(n)$  is compact, then  $f^{-1}(K)$  is compact.
- ▶ In other words: for all  $n$ , if  $(z_j)$  is a sequence in  $\mathcal{U}(n)$  and  $z_j \rightarrow \partial\mathcal{U}(n)$ , then  $f(z_j) \rightarrow \partial\mathcal{V}(n)$ .
- ▶ In the case  $g = h$  and both  $f$  and  $f^{-1}$  are (proper) nc maps, we say  $f$  is a **bianalytic** nc map.

# Change of variables

## Pencil balls

- ▶ Let  $\mathbb{B}^g := \{\mathbf{X} \in M(\mathbb{C})^g : \|\mathbf{X}\| < 1\}$  be the **operator ball**.
- ▶ To a linear pencil

$$\Lambda(\mathbf{x}) := L_1 \mathbf{x}_1 + \dots + L_g \mathbf{x}_g,$$

$L_j$  not necessarily symmetric, we associate its **pencil ball**

$$\mathcal{B}_\Lambda := \{\mathbf{X} \mid \|\Lambda(\mathbf{X})\| < 1\} = \mathcal{D} \begin{pmatrix} I & \Lambda \\ \Lambda^* & I \end{pmatrix}.$$

# Change of variables

## Pencil balls

- ▶ Let  $\mathbb{B}^g := \{X \in M(\mathbb{C})^g : \|X\| < 1\}$  be the **operator ball**.
- ▶ To a linear pencil

$$\Lambda(x) := L_1 x_1 + \dots + L_g x_g,$$

$L_j$  not nec symmetric, we associate its **pencil ball**

$$\mathcal{B}_\Lambda := \{X \mid \|\Lambda(X)\| < 1\} = \mathcal{D} \begin{pmatrix} I & \Lambda \\ \Lambda^* & I \end{pmatrix}.$$

## THM

**SUPPOSE**  $\Lambda$  is a **minimal dimensional defining pencil** for  $\mathcal{B}_\Lambda$ . **IF**  $f : \mathcal{B}_\Lambda \rightarrow \mathbb{B}^{\tilde{g}}$  is a proper analytic nc map (with matrix coefficients) with  $f(0) = 0$ , **THEN** there is a contraction-valued analytic  $J : \mathcal{B}_\Lambda \rightarrow \mathbb{B}$  such that

$$f(x) = U \begin{pmatrix} \Lambda(x) & 0 \\ 0 & J(x) \end{pmatrix} V^*$$

for some unitaries  $U, V$ .

# Change of variables

Pencil balls. An important special case

► **Let**

$$\Lambda = \sum_{i,j} \mathbf{E}_{ij} x_{ij},$$

where  $\mathbf{E}_{ij}$  are the matrix units.

# Change of variables

Pencil balls. An important special case

▶ Let

$$\Lambda = \sum_{i,j} E_{ij} x_{ij},$$

where  $E_{ij}$  are the matrix units.

▶ Then

$$\mathcal{B}_\Lambda = \{X = (X_{ij})_{i,j} \mid \|X\| < 1\}.$$

▶ The nc automorphism group of  $\mathcal{B}_\Lambda$  is transitive and contains maps of the form

$$\mathcal{F}_v(u) := v - (I - vv^*)^{1/2} u (I - v^* u)^{-1} (I - v^* v)^{1/2},$$

where  $v$  is a scalar matrix of norm  $< 1$ .

Generalizes Muhly -Soledad finite dimensional path algebras result.

# Proper nc maps tend to be bianalytic

## Theorem

Let  $\mathcal{U}$  be an nc domain in  $\mathbf{g}$  variables,  
let  $\mathcal{V}$  be an nc domain in  $\tilde{\mathbf{g}}$  variables,  
and suppose  $\mathbf{f} : \mathcal{U} \rightarrow \mathcal{V}$  is an nc analytic map.

- (1) If  $\mathbf{f}$  is proper, then it is one-to-one, and  $\mathbf{f}^{-1} : \mathbf{f}(\mathcal{U}) \rightarrow \mathcal{U}$  is an nc map.
- (2) If  $\mathbf{g} = \tilde{\mathbf{g}}$  and  $\mathbf{f} : \mathcal{U} \rightarrow \mathcal{V}$  is proper and continuous, then  $\mathbf{f}$  is bianalytic.

## Change of variables $\mathbf{g} = \tilde{\mathbf{g}}$ is rigid

Circular domains:

- ▶ A subset  $S$  of a complex vector space is **circular** if  $\exp(it)s \in S$  whenever  $s \in S$  and  $t \in \mathbb{R}$ .
- ▶ An nc domain  $\mathcal{U}$  is **circular** if each  $\mathcal{U}(n)$  is circular.



## Change of variables $\mathfrak{g} = \tilde{\mathfrak{g}}$ is rigid

Proper nc self-maps tend to be linear

- ▶ A subset  $S$  of a complex vector space is **circular** if  $\exp(it)s \in S$  whenever  $s \in S$  and  $t \in \mathbb{R}$ .
- ▶ An nc domain  $\mathcal{U}$  is **circular** if each  $\mathcal{U}(n)$  is circular.

### Theorem

Let  $\mathcal{U}, \mathcal{V}$  be nc domains in  $\mathfrak{g}$  variables containing  $\mathbf{0}$ , and suppose  $\mathbf{f} : \mathcal{U} \rightarrow \mathcal{V}$  is a proper analytic nc map with  $\mathbf{f}(\mathbf{0}) = \mathbf{0}$ .

If  $\mathcal{U}$  and  $\mathcal{V}$  are *circular*, then  $\mathbf{f}$  is *linear*.

**Generalizes:**

**Gelu Popescu map result for Reinhardt Domains**

**Muhly- Solel path algebras.**

## Summary

### Highlights to remember:

- (1) **A free convex bounded basic open semialgebraic set is the set of solutions to some LMI.**
- (2) **Much of classical real algebraic geometry goes thru to free cases. For free convex situations there is a "perfect" PosSS.**
- (3) **A proper free analytic map is one-to-one. (Partly in progress.)  
A proper free analytic map between domains of the same dimension is bianalytic. If the domains are circular and  $f(0) = 0$ , it is linear.**
- (4) **A proper free analytic map  $f$  on pencil balls with  $f(0) = 0$  is the direct sum of a linear pencil ball isometry plus junk.**
- (5)
- (6)