

Positive Semidefinite Rank

João Gouveia



FCTUC FACULDADE DE CIÊNCIAS
E TECNOLOGIA
UNIVERSIDADE DE COIMBRA

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with Hamza Fawzi (MIT), Pablo Parrilo (MIT), Richard Z.
Robinson (U.Washington) and Rekha Thomas (U.Washington)

Section 1

Definition and Basic Properties

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The smallest size of a semidefinite factorization is denoted by **positive semidefinite rank** of M , $\text{rank}_{\text{psd}}(M)$

Basic Properties

Properties

The psd rank is:

- (i) invariant under transpositions or nonnegative scalings;
- (ii) subadditive;
- (iii) at least $\approx \sqrt{2\text{rank}}$;
- (iv) at most the smallest dimension of the matrix;
- (v) ...

How does the rank function look like?

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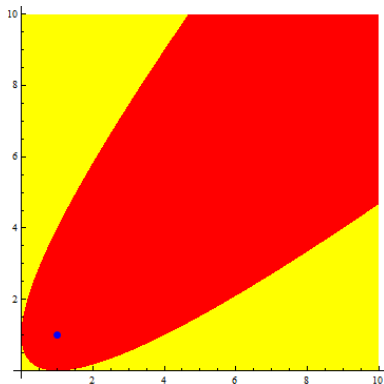
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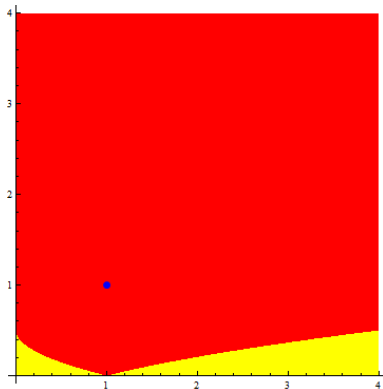
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$$\mathcal{P}_{p,q,k} := \{M \in \mathbb{R}_+^{p \times q} \mid \text{rank}_{\text{psd}}(M) \leq k\}$$

is a closed semialgebraic set inside the $\text{rank} \leq \binom{k+1}{2}$ variety.

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Even in the case $(3, 3, 2)$ the precise description is not easy.

Geometric Motivation

Given a polytope P described as a convex hull of n points and a polyhedron Q described by m inequalities with $P \subseteq Q$ we define $S_{P,Q} \subseteq \mathbb{R}_+^{n \times m}$ as the evaluation of the inequalities of Q at the points of P .

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Theorem (Semidefinite Yannakakis Theorem)

$\text{rank}_{\text{psd}}(S_{P,Q}) \leq k$ if and only if there is a convex set C with an sdp representation of size k such that $P \subseteq C \subseteq Q$.

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Lemma (Gillis-Glineur 12)

All nonnegative matrices of rank $n + 1$ can be seen as generalized slack matrices of polyhedra of dimension n .

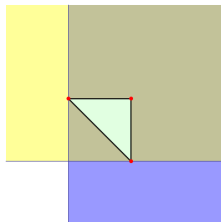
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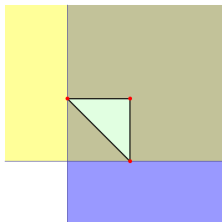
$M = S_{P,Q}$ with $\begin{cases} P = \text{conv}\{(1, 0), (0, 1), (1, 1)\} \\ Q = \{(x, y) : 1 \geq 0, x \geq 0, y \geq 0\} \end{cases} :$



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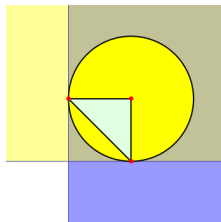


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Section 2

Computing Semidefinite Rank

Low (usual) rank cases

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Rank 3

$$\text{rank}(M) = 3 \Rightarrow \text{rank}_{\text{psd}}(M) \geq 2$$

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Can we say more for rank 3?

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Rank 3

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Can we say more for rank 3?

Let M_n be the (rank 3) slack matrix of a regular n -gon then $\text{rank}_{\text{psd}}(M_n) \rightarrow +\infty$.

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Lemma

Let $M = S_{PQ}$ with $\text{rank}(M) = 3$ then $\text{rank}_{\text{psd}}(M) = 2$ if and only if there is an ellipse E with $P \subseteq E \subseteq Q$.

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Convex Formulation

Let $P = \text{conv}(x_1, \dots, x_n)$ and $Q = \{x : Gx \leq h\}$ then $\text{rank}_{\text{psd}}(S_{PQ}) = 2$ iff there exist A, b, c such that:

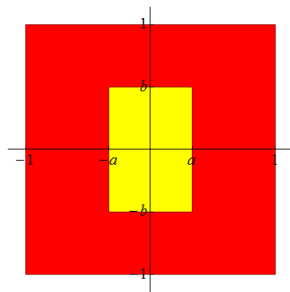
1. $A \succeq 0$, $\text{trace}(A) = 1$
2. $\begin{bmatrix} x_j \\ 1 \end{bmatrix}^T \begin{bmatrix} A & b \\ b^T & c \end{bmatrix} \begin{bmatrix} x_j \\ 1 \end{bmatrix} \leq 0 \quad \forall j$
3. $\exists \lambda_i \geq 0 : \begin{bmatrix} A & b \\ b^T & c \end{bmatrix} \succeq \lambda_i \begin{bmatrix} 0 & g_i^T/2 \\ g_i/2 & -h_i \end{bmatrix} \quad \forall i$

Example

$$M = \begin{bmatrix} 1 + a & 1 + b & 1 - a & 1 - b \\ 1 - a & 1 + b & 1 + a & 1 - b \\ 1 - a & 1 - b & 1 + a & 1 + b \\ 1 + a & 1 - b & 1 - a & 1 + b \end{bmatrix}$$

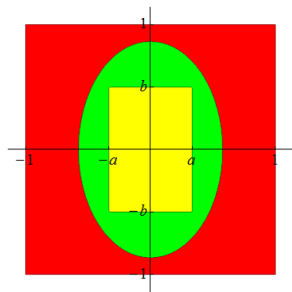
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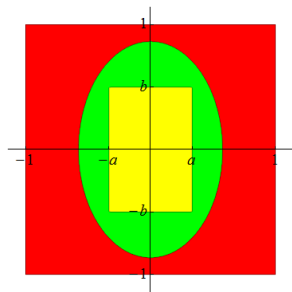
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$$\text{rank}_{\text{psd}} M = \begin{cases} 3 & \text{if } a^2 + b^2 > 1 \\ 2 & \text{if } 0 < a^2 + b^2 \leq 1 \\ 1 & \text{if } a = b = 0 \end{cases}$$

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Theorem

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- ▶ What is the complexity of computing rank_{psd} ?
- ▶ Is deciding $\text{rank}_{\text{psd}}(M) < \min\{p, q\}$ for a $p \times q$ matrix NP-hard?

Section 3

Square Root Rank

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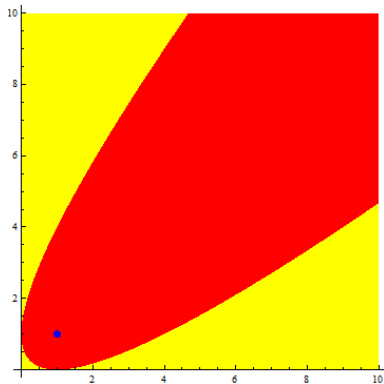
$\text{rank}_{\sqrt{\cdot}}(M)$ rank corresponds to the semidefinite rank restricted to rank one factor matrices. In particular

$$\text{rank}_{\text{psd}}(M) \leq \text{rank}_{\sqrt{\cdot}}(M).$$

How does the square root rank function look like?

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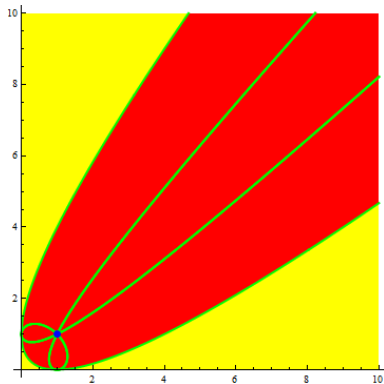
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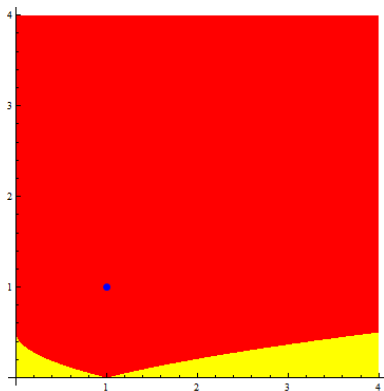


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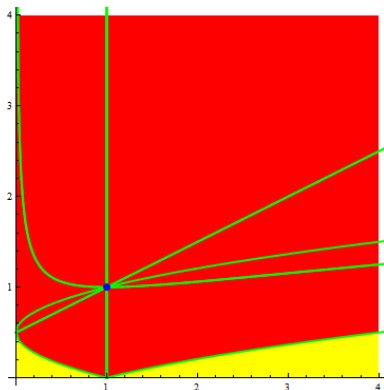
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$$\text{However } \text{rank}_{\sqrt{-}}(Q^k) = k.$$

Further bad news on the square-root rank

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$$\text{rank}_{\sqrt{\cdot}} \begin{bmatrix} 1 & 0 & \dots & 0 & a_1^2 \\ 0 & 1 & \ddots & 0 & a_2^2 \\ \vdots & \ddots & \ddots & \vdots & \vdots \\ 0 & 0 & \dots & 1 & a_n^2 \\ 1 & 1 & \dots & 1 & 0 \end{bmatrix} = n \text{ iff } \{a_1, \dots, a_n\} \text{ can be partitioned}$$

What works for square root rank

0/1 matrices

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Psd minimal polytopes

P a d -dimensional polytope, then $\text{rank}_{\text{psd}}(S_P) \geq d + 1$ with equality if and only if $\text{rank}_{\sqrt{\cdot}}(S_P) = d + 1$.

PSD minimal polytopes

\mathbb{R}^2 characterization

A 2-dimensional polytope is sdp-minimal iff it is a **triangle** or a **quadrilateral**.

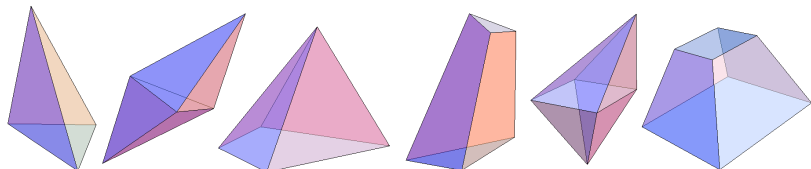
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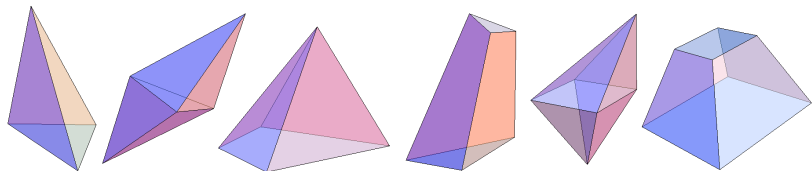
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Kostya's talk for more news on that.

Section 4

Dependency on the field

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Can the gap be 2?

Rational psd rank

We could also define $\text{rank}_{\text{psd}}^{\mathbb{Q}}$ by restricting to rational factors.

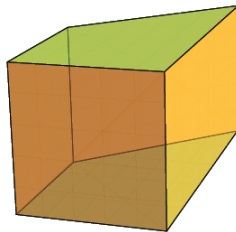
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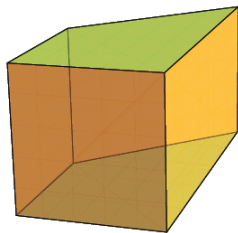
$$M = \begin{pmatrix} 0 & 0 & 2 & 1 & 0 & 1 \\ 1 & 0 & 0 & 2 & 0 & 1 \\ 0 & 1 & 2 & 0 & 0 & 1 \\ 1 & 2 & 0 & 0 & 0 & 1 \\ 0 & 0 & 2 & 1 & 1 & 0 \\ 1 & 0 & 0 & 2 & 1 & 0 \\ 0 & 1 & 2 & 0 & 1 & 0 \\ 1 & 2 & 0 & 0 & 1 & 0 \end{pmatrix}$$



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psd minimal \Rightarrow rank one factors $\Rightarrow \begin{bmatrix} 1 & 1 \\ 2 & 1 \end{bmatrix}$ cannot appear.

Section 5

Space of factorizations

Space of factorizations

Given M with $\text{rank}_{\text{psd}}(M) = k$ consider $\mathcal{SF}(M)$ the set of its $tgk \times k$ psd factorizations:

$$\mathcal{SF}(M) = \{(A_1, \dots, A_n, B_1, \dots, B_m) \in \text{PSD}_k^{m+n} : M_{ij} = \langle A_i, B_j \rangle, \forall i, j\}.$$

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For any $L \in \text{GL}(k)$ it is easy to see

$$(A_1, \dots, B_m) \in \mathcal{SF}(M) \Leftrightarrow (L^T A_1 L, \dots, L^{-1} B_m L^{-T}) \in \mathcal{SF}(M)$$

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so $\text{GL}(k)$ acts on this set.

To $\mathcal{F}_k(M) := \mathcal{SF}(M)/\text{GL}(k)$ we call the **space of factorizations** of M .

Example

Recall that $M = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 1 & 1 & 1 \end{bmatrix}$ has $\text{rank}_{\text{psd}}(M) = 2$.

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$$\left\{ \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix}, \begin{bmatrix} 1 & a \\ a & 1 \end{bmatrix}, \begin{bmatrix} 1 & \frac{-1}{2a} \\ \frac{-1}{2a} & 1 \end{bmatrix}, \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix}, \begin{bmatrix} 0 & 0 \\ 0 & 1 \end{bmatrix} \right\}$$

with $a \in [1/2, 1]$.

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Given $P \subseteq Q$, let $\mathcal{C}_k(P, Q)$ be the set of PSD_k -representable sets C such that $P \subseteq C \subseteq Q$.

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In fact, it is invariant with respect to the $\text{GL}(k)$ action so we have a map

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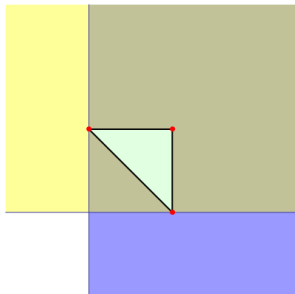
from the space of factorizations to that of “sandwiched” sets.

Example revisited

Lets look again at matrix $M = \begin{bmatrix} 1 & 1 & 0 \\ 1 & 0 & 1 \\ 1 & 1 & 1 \end{bmatrix}$.

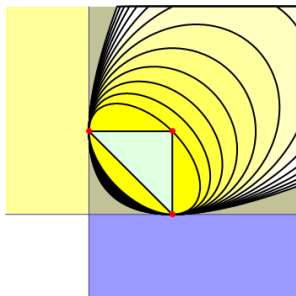
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Here we actually have a one to one correspondence

Extremal case

Theorem

If $M = S_{P,Q}$ has $\text{rank}_{\text{psd}}(M) = k$ and $\text{rank}(M) = \binom{k+1}{2}$ then $\mathcal{F}_k(M)$ and $\mathcal{C}_k(P, Q)$ are homeomorphic.

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- ▶ $\mathcal{F}_4(M)$ has at least 2 points.

Connectedness

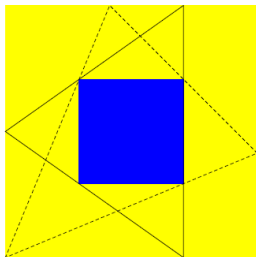
Proposition

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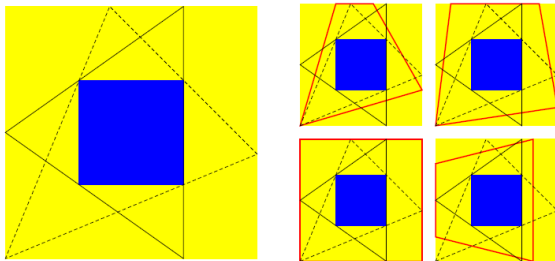


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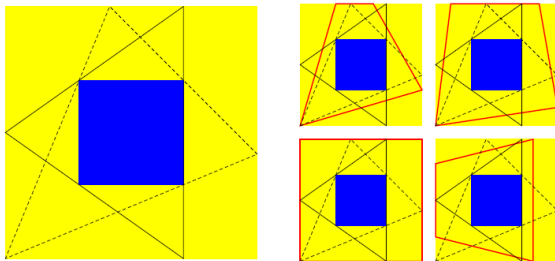


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Question

When is $\mathcal{F}_k(M)$ connected?

The end



H. Fawzi, J. Gouveia, P. Parrilo, R. Z. Robinson and R.R. Thomas.

Positive semidefinite rank.

arXiv preprint arXiv:1407.4095, 2014.



H. Fawzi, J. Gouveia, and R. Z. Robinson.

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Polytopes of minimum positive semidefinite rank.

Discrete & Computational Geometry, 50(3):679–699, 2013.

Thank you