

Lower bounds for matrix factorization ranks via noncommutative polynomial optimization

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Four matrix factorization ranks

- ▶ For a nonnegative $m \times n$ matrix A
 - ▶ **nonnegative rank** $\text{rank}_+(A)$: smallest d for which $A = (\langle x_i, y_j \rangle)$ with $x_1, \dots, x_m, y_1, \dots, y_n \in \mathbb{R}_+^d$
 - ▶ **positive semidefinite rank** $\text{psd-rank}(A)$: smallest d for which $A = (\langle X_i, Y_j \rangle)$ with $X_1, \dots, X_m, Y_1, \dots, Y_n$ $d \times d$ Hermitian PSD
- ▶ **Symmetric ranks** for a symmetric $n \times n$ matrix A
 - ▶ **completely positive rank** $\text{cp-rank}(A)$: smallest d for which $A = (\langle x_i, x_j \rangle)$ with $x_1, \dots, x_n \in \mathbb{R}_+^d$ when A is **completely positive (CP)**
 - ▶ **completely positive semidefinite rank** $\text{cpsd-rank}(A)$: smallest d for which $A = (\langle X_i, X_j \rangle)$ with X_1, \dots, X_n $d \times d$ Hermitian PSD when A is **completely positive semidefinite (CPSD)**

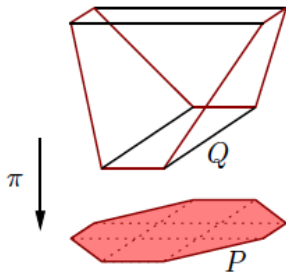
$$\text{CP}^n \subseteq \text{CPSD}^n \subseteq \text{PSD}^n$$

Common approach to lower bound these four matrix factorization ranks using (noncommutative tracial) polynomial optimization

Motivation for rank_+ and psd-rank

rank_+ and psd-rank are used in (quantum) communication complexity, linear/semidefinite extension complexity

[Yannakakis 1991, Gouveia-Parrilo-Thomas 2013]



Motivation for CP and CPSD

- ▶ **CP** is used to model discrete optimization problems
[de Klerk-Pasechnik'02, Burer'09]
- ▶ **CPSD** is used to model quantum graph parameters [L-Piovesan'15]
- ▶ **CPSD** used to model bipartite quantum correlations in $C_q(m, k)$

$$p = (p(a, b|s, t) := \langle \Psi, A_s^a \otimes B_t^b \Psi \rangle), \text{ with } d \in \mathbb{N}, \Psi \in \mathbb{C}^d \otimes \mathbb{C}^d \text{ unit vector, } A_s^a, B_t^b \text{ } d \times d \text{ Hermitian PSD, } \sum_{a=1}^k A_s^a = \sum_{b=1}^k B_t^b = I \text{ for } s, t \in [m]$$

Smallest such $d =$ **entanglement dimension** of p

- ▶ $C_q(m, k)$ is an affine slice of **CPSD**^{2mk} [Mancinska-Roberson'14]
[Sikora-Varvitsiotis'15]
- ▶ If p is **synchronous**: $p(a, b|s, s) = 0$ whenever $a \neq b$, then its **entanglement dimension** is equal to **cpsd-rank**(A_p), where $(A_p)_{(a,s),(b,t)} = p(a, b|s, t)$ [G-dL-L'17]
- ▶ $C_q(m, k)$ is not closed [Slofstra'17] [Dykema-Paulsen-Prakash'17]
 \rightsquigarrow **CPSD**ⁿ is not closed for $n \geq 1942$, for $n \geq 10$

Basic bounds

Upper bounds:

- ▶ For $A \in \mathbb{R}_+^{m \times n}$: $\text{psd-rank}(A) \leq \text{rank}_+(A) \leq \min\{m, n\}$
- ▶ For $A \in \text{CP}^n$: $\text{cp-rank}(A) \leq \binom{n+1}{2}$
- ▶ For $A \in \text{CPSD}^n$: No upper bound exists on cpsd-rank in terms of n

Lower bounds:

- ▶ $\text{rank}(A) \leq \text{rank}_+(A), \text{cp-rank}(A)$
- ▶ $\sqrt{\text{rank}(A)} \leq \text{psd-rank}(A), \text{cpsd-rank}(A)$

More lower bounds on rank_+ and cp-rank

- Fawzi-Parrilo (2016) define lower bounds $\tau_+(\cdot)$ and $\tau_{cp}(\cdot)$ based on the **atomic definition** of rank_+ and cp-rank:

$$\text{rank}_+(A) = \min d \text{ s.t. } A = u_1 v_1^T + \dots + u_d v_d^T \text{ with } u_i, v_i \in \mathbb{R}_+^n$$

$$\tau_+(A) = \min \alpha \text{ s.t. } \frac{1}{\alpha} A \in \text{conv}(R : 0 \leq R \leq A, \text{rank}(R) \leq 1)$$

$$\text{cp-rank}(A) = \min d \text{ s.t. } A = u_1 u_1^T + \dots + u_d u_d^T \text{ with } u_i \in \mathbb{R}_+^n$$

$$\tau_{cp}(A) = \min \alpha \text{ s.t. } \frac{1}{\alpha} A \in \text{conv}(R : 0 \leq R \leq A, \text{rank}(R) \leq 1, R \preceq A)$$

- Fawzi-Parrilo (2016) define SDP lower bounds $\tau_+^{\text{SOS}}(\cdot)$ and $\tau_{cp}^{\text{SOS}}(\cdot)$:

$$\tau_+^{\text{SOS}}(A) \leq \tau_+(A) \leq \text{rank}_+(A), \quad \text{rank}(A) \leq \tau_{cp}^{\text{SOS}}(A) \leq \tau_{cp}(A) \leq \text{cp-rank}(A)$$

- Link to the combinatorial 'rectangle covering' bound on rank_+ :

$$\text{rank}_+(A) \geq \chi(RG(A)) = \text{coloring number of 'rectangle graph' } RG(A)$$

$$\text{rank}_+(A) \geq \tau_+(A) \geq \chi_f(RG(A)), \quad \text{rank}_+(A) \geq \tau_+^{\text{SOS}}(A) \geq \vartheta(\overline{RG(A)})$$

New approach to bound all four factorization ranks

since no atomic definition exists for **psd-rank** and **cpsd-rank**

Commutative polynomial optimization [Lasserre, Parrilo,...]

Noncommutative eigenvalue optimization [Pironio, Navascués, Acín,...]

Noncommutative tracial optimization

[Burgdorf, Cafuta, Klep, Povh, Schweighofer,...]

$$f_*^c = \inf f(x) \text{ s.t. } x \in \mathbb{R}^n, g(x) \geq 0 (g \in S) \quad [d = 1]$$

$$f_*^{nc} = \inf \text{Tr}(f(\mathbf{X}))/d \text{ s.t. } d \in \mathbb{N}, \mathbf{X} \in (H^d)^n, g(\mathbf{X}) \succeq 0 (g \in S)$$

$$f_\infty^{nc} = \inf \tau(f(\mathbf{X})) \text{ s.t. } \mathcal{A} \text{ } C^*\text{-algebra, } \tau \text{ trace, } \mathbf{X} \in \mathcal{A}^n, g(\mathbf{X}) \succeq 0 (g \in S)$$

$$f_\infty^{nc} \leq f_*^{nc} \leq f_*^c$$

- ▶ SDP lower bounds: $\inf L(f)$ over $L \in \mathbb{R}[\mathbf{x}]_{2t}^*$, $L(1) = 1, \dots, L \in \mathbb{R}\langle \mathbf{x} \rangle_{2t}^*$
Under Archimedean condition: $f_t^c \rightarrow f_*^c$, $f_t^{nc} \rightarrow f_\infty^{nc}$ as $t \rightarrow \infty$
- ▶ Equality: $f_t^{nc} = f_*^{nc}$, $f_t^c = f_*^c$ if order t bound has flat optimal sol.

For lower bounding matrix factorization ranks: use the same framework, but now **minimize** $L(1)$ with L **not normalized** s.t. ...

Polynomial optimization approach for cpsd-rank

Assume $A = (\text{Tr}(X_i X_j))$ has a factorization by $d \times d$ Hermitian PSD matrices $\mathbf{X} = (X_1, \dots, X_n)$ and $d = \text{cpsd-rank}(A)$.

Let $L \in \mathbb{R}\langle x_1, \dots, x_n \rangle^*$ be the real part of the trace evaluation $L_{\mathbf{X}}$ at \mathbf{X} :

$$L_{\mathbf{X}}(p) = \text{Tr}(p(\mathbf{X})), \quad L(p) = \text{Re}(\text{Tr}(p(\mathbf{X}))) \quad \text{for } p \in \mathbb{R}\langle x_1, \dots, x_n \rangle$$

- (0) $L(1) = d$
- (1) $L(x_i x_j) = A_{ij}$ for all $i, j \in [n]$
- (2) L is symmetric ($L(p^*) = L(p)$), tracial ($L(pq) = L(qp)$)
- (3) L is positive ($L(p^* p) \geq 0$)
- (4) L positive on localizing polynomials: $L(p^*(\sqrt{A_{ii}}x_i - x_i^2)p) \geq 0 \forall i$

$$L \geq 0 \text{ on } \underbrace{\text{cone}\{p^* g p : g \in \{1\} \cup \{\sqrt{A_{ii}}x_i - x_i^2 : i \in [n]\}, p \in \mathbb{R}\langle \mathbf{x} \rangle\}}_{\mathcal{M}(S_A^{\text{cpsd}})}$$

Get lower bounds by minimizing $L(1)$ over $L \in \mathbb{R}\langle \mathbf{x} \rangle_{2t}^*$ satisfying (1)-(4).

Lower bounds for cpsd-rank

For an integer $t \in \mathbb{N} \cup \{\infty\}$

$$\xi_t^{\text{cpsd}}(A) = \min L(1) \text{ s.t. } L \in \mathbb{R}\langle \mathbf{x} \rangle_{2t}^* \text{ symmetric, tracial, } A = (L(x_i x_j)) \\ L \geq 0 \text{ on } \mathcal{M}_{2t}(S_A^{\text{cpsd}})$$

$$\xi_*^{\text{cpsd}}(A) \text{ is } \xi_\infty^{\text{cpsd}}(A) \text{ with extra constraint } \text{rank}(M(L) = (L(u^* v))) < \infty$$

$$\xi_1^{\text{cpsd}}(A) \leq \dots \leq \xi_t^{\text{cpsd}}(A) \leq \dots \leq \xi_\infty^{\text{cpsd}}(A) \leq \xi_*^{\text{cpsd}}(A) \leq \text{cpsd-rank}(A)$$

- ▶ Asymptotic convergence: $\xi_t^{\text{cpsd}}(A) \rightarrow \xi_\infty^{\text{cpsd}}(A)$ as $t \rightarrow \infty$
 $\xi_\infty^{\text{cpsd}}(A) = \min \alpha$ s.t. $\frac{1}{\alpha}A = (\tau(X_i X_j))$, where \mathcal{A} C^* -algebra with trace τ , $\mathbf{X} \in \mathcal{A}^n$ s.t. $\sqrt{A_{ii}}X_i - X_i^2 \succeq 0$ for $i \in [n]$
- ▶ $\xi_*^{\text{cpsd}}(A) = \min \alpha$ s.t. ... \mathcal{A} finite dimensional ...
 $= \min L(1)$ s.t. L conic combination of trace evaluations ...
- ▶ $\xi_t^{\text{cpsd}}(A) = \xi_*^{\text{cpsd}}(A)$ if $\xi_t^{\text{cpsd}}(A)$ has a flat optimal solution

Strengthening and extending the bounds

One can strengthen the basic bounds by adding constraints on L :

1. $L(p^*(v^T Av - (\sum_i v_i x_i)^2)p) \geq 0$ for all $v \in \mathbb{R}^n$ [v-constraints]
2. $L(p^* g p g') \geq 0$ for g, g' localizing for A [Berta et al.'16]
3. $L(p x_i x_j) = 0$ if $A_{ij} = 0$ [zeros propagate]
4. $L(p(\sum_i v_i x_i)) = 0$ for all $v \in \ker A$ [kernel vectors propagate]

One can extend the bounds:

- ▶ Asymmetric setting (for rank_+ and psd-rank): use two sets of variables $x_1, \dots, x_m, y_1, \dots, y_n$
- ▶ Commutative setting (for rank_+ and cp-rank): use polynomials in commutative variables, after viewing nonnegative vectors as diagonal PSD matrices

Small example

Consider $A = \begin{pmatrix} 1 & 1/2 & 0 & 0 & 1/2 \\ 1/2 & 1 & 1/2 & 0 & 0 \\ 0 & 1/2 & 1 & 1/2 & 0 \\ 0 & 0 & 1/2 & 1 & 1/2 \\ 1/2 & 0 & 0 & 1/2 & 1 \end{pmatrix}$

- ▶ $\text{cpsd-rank}(A) \leq 5$

because if $\mathbf{X} = \text{Diag}(1, 1, 0, 0, 0)$ and its cyclic shifts then $\mathbf{X}/\sqrt{2}$ is a factorization of A

- ▶ $L = \frac{1}{2}L_{\mathbf{X}}$ is feasible for $\xi_*^{\text{cpsd}}(A)$, with value $L(1) = 5/2$

Hence $\xi_*^{\text{cpsd}}(A) \leq 5/2$, in fact $\xi_1^{\text{cpsd}}(A) = \xi_*^{\text{cpsd}}(A) = 5/2$

- ▶ $\xi_{2,\nu}^{\text{cpsd}}(A) = 5 = \text{cpsd-rank}(A)$

with the ν -constraints for $\nu = (1, -1, 1, -1, 1)$ and its cyclic shifts

Lower bounds for cp-rank

$$\xi_t^{cp}(A) = \min L(1) \quad \text{s.t.} \quad L \in \mathbb{R}[\mathbf{x}]_{2t}^*, \quad A = (L(x_i x_j)), \quad L \geq 0 \text{ on } \mathcal{M}_{2t}(S_A^{cp})$$

$$\text{where } S_A^{cp} = \{\sqrt{A_{ii}}x_i - x_i^2 : i \in [n]\} \cup \{A_{ij} - x_i x_j : i, j \in [n]\}$$

$\xi_{t,\dagger}^{cp}(A)$ has the additional constraints:

(P) $L(ug) \geq 0$ for $g \in \{1\} \cup S_A^{cp}$ and monomials u with $\deg(ug) \leq 2t$

(T) $A^{\otimes l} - (L(u^*v))_{u,v \in \langle \mathbf{x} \rangle_{=l}} \succeq 0$ for $2 \leq l \leq t$

Comparison to the bounds τ_{cp}^{sos} and τ_{cp} of Fawzi-Parrilo (2016):

- ▶ $\xi_t^{cp}(A) \leq \xi_\infty^{cp}(A) = \xi_*^{cp}(A) \leq \tau_{cp}(A)$
- ▶ $\tau_{cp}^{sos}(A) \leq \xi_{2,\dagger}^{cp}(A) \leq \xi_{\infty,\dagger}^{cp}(A) \leq \xi_{*,\dagger}^{cp}(A) = \tau_{cp}(A)$
- ▶ $\tau_{cp}(A)$ is also reached as asymptotic limit when using the v -constraints for a dense subset of \mathbb{S}^{n-1} instead of (P)-(T)

Example: $A_{a,b} = \begin{pmatrix} (q+a)I_p & J \\ J & (p+b)I_q \end{pmatrix} \in \mathcal{S}^{p+q}$ for $a, b \in [0, 1]^2$

- ▶ $\xi_{2,\dagger}^{cp}(A_{a,b}) \geq pq$
- ▶ $\xi_{2,\dagger}^{cp}(A_{a,b}) = 6 = \text{cp-rank}(A_{a,b})$ is tight for $(p, q) = (2, 3)$
 $5 \leq \tau_{cp}^{sos} < 6$ for all nonzero $(a, b) \in [0, 1]^2$, equal to 5 on subregion

Lower bounds for rank_+ and psd-rank

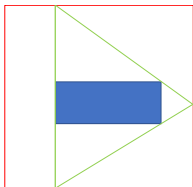
Same approach: as **no a priori bound** on the eigenvalues of the factors
... **rescale** the factors to get such bounds and thus **localizing constraints**

Get now $\tau_+(A) = \xi_\infty^+(A)$ directly as asymptotic limit of the SDP bounds

Example for rank_+ : [Fawzi-Parrilo'16]

$$S_{a,b} = \begin{pmatrix} 1-a & 1+a & 1+a & 1-a \\ 1+a & 1-a & 1-a & 1+a \\ 1-b & 1-b & 1+b & 1+b \\ 1+b & 1+b & 1-b & 1-b \end{pmatrix} \quad \text{for } a, b \in [0, 1]$$

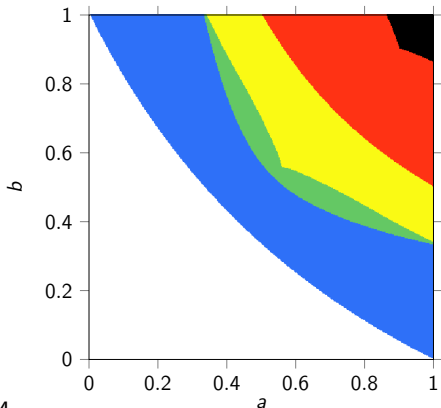
slack matrix of nested rectangles: $R = [-a, a] \times [-b, b] \subseteq P = [-1, 1]^2$



\exists triangle T s.t. $R \subseteq T \subseteq P \iff \text{rank}_+(S_{a,b}) = 3$

Extension complexity: Nested rectangle problem

White region: $\text{rank}_+(S_{a,b}) = 3 \iff (1+a)(1+b) \leq 2$



Colored: $\text{rank}_+ = 4$

Top right: $\xi_{1,\dagger}^+ > 3$

Two top right regions: $\tau_+^{\text{SOS}} > 3$

Three top right regions: $\xi_{2,\dagger}^+ > 3$

Four top right regions: $\xi_{3,\dagger}^+ > 3$

$$\begin{pmatrix} 1-a & 1+a & 1+a & 1-a \\ 1+a & 1-a & 1-a & 1+a \\ 1-b & 1-b & 1+b & 1+b \\ 1+b & 1+b & 1-b & 1-b \end{pmatrix}$$

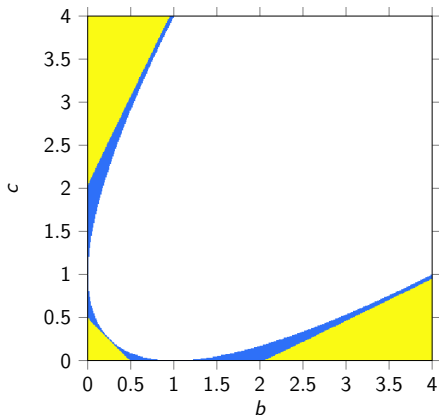
Small example for psd-rank

[Fawzi et al.'15] For $M_{b,c} = \begin{pmatrix} 1 & b & c \\ c & 1 & b \\ b & c & 1 \end{pmatrix}$

White region: $\text{psd-rank}_{\mathbb{R}} \leq 2 \iff b^2 + c^2 + 1 \leq 2(b + c + bc)$

Colored region: $\text{psd-rank}_{\mathbb{R}} = 3$

Yellow region: $\xi_2^{\text{psd}} > 2$



Concluding remarks

- ▶ Bounds via (tracial nc) polynomial optimization: [arXiv:1708.01573](https://arxiv.org/abs/1708.01573)

commutative	tracial noncommutative
completely positive cone CP^n	completely positive semidefinite cone $CPSD^n$
cp-rank, rank ₊	cpsd-rank, psd-rank

- ▶ 'Minimizing $L(1)$ ' was used by [Tang-Sha'15, Nie'16] to get bounds converging to the tensor nuclear norm (commutative setting)
- ▶ The approach extends to the nonnegative tensor rank, also considered by Fawzi-Parrilo (2016) (commutative setting)
- ▶ The bounds apply to the **complex** ranks (using Hermitian factors). How to tailor the bounds for **real** ranks?
- ▶ Extension to lower bound the entanglement dimension of a (non-synchronous) quantum correlation in [arXiv:1708.09696](https://arxiv.org/abs/1708.09696)