

Primal and dual optimization problems related to matrix factorizations

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Works:

- ① R. Balan, K. Okoudjou, A. Poria, *On a Feichtinger Problem*, *Operators and Matrices* vol. 12(3), 881-891 (2018)
<http://dx.doi.org/10.7153/oam-2018-12-53>
- ② R. Balan, K.A. Okoudjou, M. Rawson, Y. Wang, R. Zhang, *Optimal l1 Rank One Matrix Decomposition*, in "Harmonic Analysis and Applications", Rassias M., Ed. Springer (2021)

Problem Formulation

Let $Sym^+(\mathbb{C}^n) = \{A \in \mathbb{C}^{n \times n}, A^* = A \geq 0\}$. For $A \in Sym^+(\mathbb{C}^n)$,

$$\gamma_+(A) := \inf_{A = \sum_{k \geq 1} x_k x_k^*} \sum_k \|x_k\|_1^2$$

The *matrix conjecture*: There is a universal constant C_0 such that, for every $n \geq 1$ and $A \in Sym^+(\mathbb{C}^n)$,

$$\gamma_+(A) \leq C_0 \|A\|_1 := C_0 \sum_{k,l=1}^n |A_{k,l}|$$

Regardless of the answer, we are interested in finding:

$$C_n = \sup_{A \geq 0} \frac{\gamma_+(A)}{\|A\|_1}$$

Motivation

A Feichtinger Problem

At a 2004 Oberwolfach meeting, Hans Feichtinger asked the following question:

(Q1) Given a positive semi-definite trace-class operator $T : L^2(\mathbb{R}^d) \rightarrow L^2(\mathbb{R}^d)$, $Tf(x) = \int K(x, y)f(y)dy$, with $K \in M^1(\mathbb{R}^d \times \mathbb{R}^d)$, and its spectral factorization, $T = \sum_k \langle \cdot, h_k \rangle h_k$, must it be $\sum_k \|h_k\|_{M^1}^2 < \infty$?

A modified version of the question is:

(Q2) Given T as before, i.e., $T = T^* \geq 0$, $K \in M^1(\mathbb{R}^d \times \mathbb{R}^d)$, is there a factorization $T = \sum_k \langle \cdot, g_k \rangle g_k$ such that $\sum_k \|g_k\|_{M^1}^2 < \infty$?

Motivation (2)

Matrix Language

Consider an infinite matrix $A = (A_{m,n})_{m,n \geq 0}$ so that

$$\|A\|_{\wedge} := \|A\|_{\mathbf{1}} := \sum_{m,n \geq 0} |A_{m,n}| < \infty.$$

This implies that A acts on $l^2(\mathbb{N})$ as a trace-class compact operator.

Assume additionally $A = A^* \geq 0$ as a quadratic form.

Let $(e_k)_{k \geq 0}$ denote an orthogonal set of eigenvectors normalized so that

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Using previous equivalence and some functional analysis arguments:

Proposition

If (Q2) is answered affirmatively, then the matrix conjecture must be true.

Current Status of the Matrix Conjecture (1)

The infimum is achieved:

$$\gamma_+(A) := \inf_{A=\sum_{k \geq 1} x_k x_k^*} \sum_k \|x_k\|_1^2 = \min_{A=\sum_{k=1}^{n^2} x_k x_k^*} \sum_k \|x_k\|_1^2.$$

Upper bounds:

$$\gamma_+(A) \leq n \operatorname{trace}(A) \leq n \|A\|_1 = n \sum_{k,j} |A_{k,j}|$$

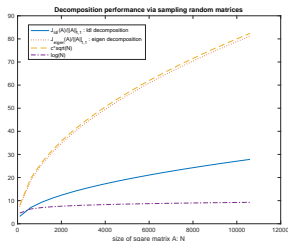
$$\gamma_+(A) \leq n \operatorname{trace}(A) \leq n^2 \|A\|_{op}$$

Lower bounds:

$$\|A\|_1 = \min_{A=\sum_{k \geq 1} x_k y_k^*} \sum_k \|x_k\|_1 \|y_k\|_1 \leq \gamma_+(A)$$

Convexity: for $A, B \in \operatorname{Sym}^+(\mathbb{C}^n)$ and $t \geq 0$,

$$\gamma_+(A+B) \leq \gamma_+(A) + \gamma_+(B) \quad , \quad \gamma_+(tA) = t\gamma_+(A)$$



Maximum of $\sum_k \|x_k\|_1^2 / \|A\|_1$ over 30 random noise realizations, where x_k 's are obtained from the eigendecomposition, or the LDL factorization.

Current Status of the Matrix Conjecture (2)

Lower bound is achieved, $\gamma_+(A) = \|A\|_1$:

- ① If $A = xx^*$ is of rank one.
- ② If $A \geq 0$ is a diagonally dominant matrix, $A_{ii} \geq \sum_{k \neq i} |A_{i,k}|$.
- ③ If $A \geq 0$ admits a Non Negative Matrix Factorization (NNMF), $A = BB^T$ with $B_{ij} \geq 0$.

Continuity, Lipschitz and linear program reformulation:

- ① $\gamma_+ : \text{Sym}^+(\mathbb{C}^n) \rightarrow \mathbb{R}$ is continuous.
- ② If $A, B \geq \delta I$ and $\text{trace}(A), \text{trace}(B) \leq 1$ then

$$|\gamma_+(A) - \gamma_+(B)| \leq \left(\frac{n}{\delta^2} + n^2 \right) \|A - B\|_{op}.$$

- ③ Let $S_1 = \{x \in \mathbb{C}^n, \|x\|_1 = 1\}$ denote the compact unit sphere with respect to the l^1 norm, and let $\mathcal{B}(S_1)$ denote the set of Borel measures over S_1 . Then:

$$\gamma_+(A) = \inf_{\mu \in \mathcal{B}(S_1): \int_{S_1} xx^* d\mu(x) = A} \mu(S_1), \quad \mu^*(x) = \sum_{k=1}^m \lambda_k \delta(x - g_k)$$

where $\gamma_+(A) = \sum_{k=1}^m \lambda_k$ and $A = \sum_{k=1}^m g_k g_k^*$ is the optimal factorization.

Primal and dual problems for γ_+

The linear program is convex optimization problem (which is great), but it is defined in an infinite dimensional space (not so great!).

Its dual program enjoys strong duality (this may not be obvious due to infinite dimensional technical issues):

Theorem (2024a)

Assume $A \geq 0$. Its associated primal (min) - dual (max) problems are:

$$\max_{T=T^*: \langle T, x \rangle \leq 1, \forall \|x\|_1 \leq 1} \text{trace}(TA) = \min_{\mu \in \mathcal{B}(S_1): \int_{S_1} xx^* d\mu(x) = A} \mu(S_1) = \gamma_+(A)$$

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Note the quantity:

$$\rho_1(T) = \max_{x: \|x\|_1 \leq 1} \langle Tx, x \rangle$$

The dual problem and C_n turn into:

$$\max_{T=T^*: \rho_1(T) \leq 1} \text{trace}(TA)$$

$$C_n = \max_{A \geq 0: \|A\|_1 \leq 1} \gamma_+(A) = \max_{A \geq 0: \|A\|_1 \leq 1} \max_{T=T^*: \rho_1(T) \leq 1} \text{trace}(TA) = \max_{A \geq 0} \max_{T=T^*: \rho_1(T) > 0} \frac{\text{trace}(TA)}{\|A\|_1 \rho_1(T)}$$

The bound ρ_1

Recall, for $T = T^*$:

$$\rho_1(T) = \max_{x: \|x\|_1 \leq 1} \langle Tx, x \rangle$$

How to compute it?

Easy cases:

- 1 If $T \leq 0$ then $\rho_1(T) = 0$
- 2 If $T \geq 0$ then

$$\rho_1(T) = \max_k T_{k,k} = \max_{i,j} |T_{i,j}| =: \|T\|_\infty$$

This resembles the *numerical radius* of a matrix, $r(T) = \max_{\|x\|_2=1} |\langle Tx, x \rangle|$, which for hermitian matrices equals the largest singular value (operator norm). Note differences: (i) $\|\cdot\|_2 \rightarrow \|\cdot\|_1$; (ii) no absolute value $|\cdot|$.

The bound ρ_1 (2)

Assume $\lambda_{\max}(T) > 0$, i.e. T is NOT negative semi-definite. Then:

$$\rho_1(T) = \max_{x: \|x\|_1=1} \langle Tx, x \rangle = \max_{\substack{A \geq 0 : \\ \text{rank}(A) = 1 \\ \|A\|_1 = 1}} \text{trace}(TA)$$

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Convex relaxation:

$$\pi_+(T) := \max_{\substack{A \geq 0 : \\ \|A\|_1 = 1}} \text{trace}(TA)$$

which is a semi-definite program (SDP). Thus:

$$\rho_1(T) \leq \pi_+(T).$$

Primal and dual problems for ρ_1

The SDP enjoys strong duality:

Theorem (2024b)

Assume $T = T^*$. The primal-dual programs have strong duality:

$$\pi_+(T) = \max_{\substack{A \geq 0 : \\ \|A\|_1 = 1}} \text{trace}(TA) = \min_{Y \geq 0} \|T + Y\|_\infty$$

where $\|Z\|_\infty = \max_{i,j} |Z_{i,j}|$.

The proof of this theorem is based on the Von Neumann's min-max theorem:

$$\begin{aligned} \min_{Y \geq 0} \|T + Y\|_\infty &= \min_{Y \geq 0} \max_{A: \|A\|_1=1} \text{trace}((T + Y)A) \stackrel{vN}{=} \max_{A: \|A\|_1=1} \min_{Y \geq 0} \text{trace}((T + Y)A) = \\ &= \max_{A: \|A\|_1=1} \left(\text{trace}(TA) + \min_{Y \geq 0} \text{trace}(YA) \right) = \max_{A \geq 0: \|A\|_1=1} \left(\text{trace}(TA) + \min_{Y \geq 0} \text{trace}(YA) \right) = \\ &= \max_{A \geq 0: \|A\|_1=1} \text{trace}(TA) = \pi_+(T) \end{aligned}$$

Connexion result

The final result: the connection between $\gamma_+(A)$ and C_n on one hand, and $\rho_1(T)$ and $\pi_+(T)$ on the other hand:

Theorem (2024c)

$$C_n := \max_{\substack{A \geq 0 \\ A \neq 0}} \frac{\gamma_+(A)}{\|A\|_1} = \max_{\substack{T = T^* \\ \rho_1(T) \neq 0}} \frac{\pi_+(T)}{\rho_1(T)}$$

The proof is based on an earlier derivation:

$$\begin{aligned} C_n &= \max_{\substack{A \geq 0 \\ A \neq 0}} \max_{\substack{T = T^* \\ \rho_1(T) > 0}} \frac{\text{trace}(TA)}{\|A\|_1 \rho_1(T)} = \max_{\substack{T = T^* \\ \rho_1(T) > 0}} \max_{\substack{A \geq 0 \\ A \neq 0}} \frac{\text{trace}(TA)}{\|A\|_1 \rho_1(T)} = \\ &= \max_{\substack{T = T^* \\ \rho_1(T) > 0}} \frac{1}{\rho_1(T)} \frac{\pi_+(T)}{\rho_1(T)} \max_{\substack{A \geq 0 \\ \|A\|_1 = 1}} \text{trace}(TA) = \max_{\substack{T = T^* \\ \rho_1(T) > 0}} \frac{\pi_+(T)}{\rho_1(T)} \end{aligned}$$

Thank you!

Thank you for listening!
QUESTIONS?

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If (Q2) is answered affirmatively, then the matrix conjecture must be true.

Linear program result

Optimal Factorization from a Measure Theory Perspective

Let $S_1 = \{x \in \mathbb{C}^n, \|x\|_1 = 1\}$ denote the compact unit sphere with respect to the l^1 norm, and let $\mathcal{B}(S_1)$ denote the set of Borel measures over S_1 . For $A \in \text{Sym}(\mathbb{C}^n)^+(\mathbb{C}^n)$ consider the optimization problem:

$$(p^*, \mu^*) = \inf_{\mu \in \mathcal{B}(S_1): \int_{S_1} xx^* d\mu(x) = A} \mu(S_1) \quad (M)$$

Theorem (Optimal Measure)

For any $A \in \text{Sym}^+(\mathbb{C}^n)$ the optimization problem (M) is convex and its global optimum (minimum) is achieved by

$$p^* = \gamma_+(A) \quad , \quad \mu^*(x) = \sum_{k=1}^m \lambda_k \delta(x - g_k)$$

where $A = \sum_{k=1}^m (\sqrt{\lambda_k} g_k)(\sqrt{\lambda_k} g_k)^*$ is an optimal decomposition that achieves $\gamma_+(A) = \sum_{k=1}^m \lambda_k$.

Super-resolution and Convex Optimizations

$$\gamma_+(A) = \min_{x_1, \dots, x_m : A = \sum_k x_k x_k^*} \sum_{k=1}^m \|x_k\|_1^2, \quad m = n^2 \quad (P)$$

$$p^* = \inf_{\mu \in \mathcal{B}(S_1) : A = \int_{S_1} x x^* d\mu(x)} \int_{S_1} d\mu(x) \quad (M)$$

Remarks

- 1 *The optimization problem (P) is non-convex, but finite-dimensional. The optimization problem (M) is convex, but infinite-dimensional.*
- 2 *If $g_1, \dots, g_m \in S_1$ in the support of μ^* are known so that $\mu^* = \sum_{k=1}^m \lambda_k \delta(x - g_k)$, then the optimal $\lambda_1, \dots, \lambda_m \geq 0$ are determined by a linear program. More general, (M) is an infinite-dimensional linear program.*
- 3 *Finding the support of μ^* is an example of a super-resolution problem. One possible approach is to choose a redundant dictionary (frame) that includes the support of μ^* , and then solve the induced linear program.*

Proof of the Optimal Measure Result

Recall: we want to show the following problems admit same solution:

$$\gamma_+(A) = \min_{x_1, \dots, x_m : A = \sum_k x_k x_k^*} \sum_{k=1}^m \|x_k\|_1^2, \quad m = n^2 \quad (P)$$

$$p^* = \inf_{\mu \in \mathcal{B}(S_1) : A = \int_{S_1} x x^* d\mu(x)} \int_{S_1} d\mu(x) \quad (M)$$

a. Assume $A = \sum_{k=1}^m x_k x_k^*$ is a global minimum for (P). Then

$\mu(x) = \sum_{k=1}^m \|x_k\|_1^2 \delta(x - \frac{x_k}{\|x_k\|_1})$ is a feasible solution for (M). This shows

$$p^* \leq \gamma_+(A).$$

b. For reverse: Let μ^* be an optimal measure in (M). Fix $\varepsilon > 0$. Construct a disjoint partition $(U_l)_{1 \leq l \leq L}$ of S_1 so that each U_l is included in some ball $B_\varepsilon(z_l)$ of radius ε with $\|z_l\|_1 = 1$. Thus $U_l \subset B_\varepsilon(z_l) \cap S_1$.

For each l , compute $x_l = \frac{1}{\mu^*(U_l)} \int_{U_l} x d\mu^*(x) \in B_\varepsilon(z_l)$. Let $g_l = \sqrt{\mu^*(U_l)} x_l$.

Proof: The Optimal Measure Result (cont)

Key inequality:

$$0 \leq R_l := \int_{U_l} (x - x_l)(x - x_l)^* d\mu^*(x) = \int_{U_l} xx^* d\mu^*(x) - \mu^*(U_l)x_l x_l^*$$

Sum over l and with $R = \sum_{l=1}^L R_l$ get

$$A = \int_{S_1} xx^* d\mu^*(x) \leq \sum_{l=1}^L g_l g_l^* + R$$

By sub-additivity and homogeneity:

$$\gamma_+(A) \leq \sum_{l=1}^L \|g_l\|_1^2 + \gamma_+(R) \leq \sum_{l=1}^L \mu^*(U_l) \|x_l\|_1^2 + n \operatorname{trace}(R)$$

But $\|x_l - z_l\|_1 \leq \varepsilon$ and $\|x - x_l\|_1 \leq 2\varepsilon$ for every $x \in U_l$. Hence $\|x_l\|_1 \leq 1 + \varepsilon$ and $\operatorname{trace}(R_l) \leq 4\mu^*(U_l)\varepsilon^2$.

Proof: The Optimal Measure Result (end)

Thus:

$$\gamma_+(A) \leq \mu^*(S_1) + (2\varepsilon + \varepsilon^2 + 4n\varepsilon^2)\mu^*(S_1)$$

Since $\varepsilon > 0$ is arbitrary, it follows

$$\gamma_+(A) \leq \mu^*(S_1) = p^*$$

This ends the proof of the measure result. \square

Second New Result: The Continuity Property

Theorem (The Continuity Property)

The map $\gamma_+ : (\text{Sym}^+(\mathbb{C}^n), \|\cdot\|) \rightarrow \mathbb{R}$ is continuous.

Remarks

- 1 This statement extends the continuity result from $\text{Sym}^{++}(\mathbb{C}^n) = \{A = A^* > 0\}$ to $\text{Sym}^+(\mathbb{C}^n) = \{A = A^* \geq 0\}$.
- 2 Proof is based on a (new?) comparison result between non-negative operators.
- 3 Global Lipschitz is still open.

The Continuity Property

The proof is based on the following two lemmas:

Lemma (L1)

Let $A \in \text{Sym}^+(\mathbb{C}^n)$ of rank $r > 0$. Let $\lambda_r > 0$ denote the r^{th} eigenvalue of A , and let $P_{A,r}$ denote the orthogonal projection onto the range of A . For any $0 < \varepsilon < 1$ and $B \in \text{Sym}^+(\mathbb{C}^n)$ such that $\|A - B\|_{\text{Op}} \leq \frac{\varepsilon \lambda_r}{1 - \varepsilon}$, the following holds true:

$$A - (1 - \varepsilon)P_{A,r}BP_{A,r} \geq 0 \quad (1)$$

Lemma (L2)

Let $A \in \text{Sym}^+(\mathbb{C}^n)$ of rank $r > 0$. Let $\lambda_r > 0$ denote the r^{th} eigenvalue of A . For any $0 < \varepsilon < \frac{1}{2}$ and $B \in \text{Sym}^+(\mathbb{C}^n)$ such that $\|A - B\|_{\text{Op}} \leq \varepsilon \lambda_r$, the following holds true:

$$B - (1 - \varepsilon)P_{B,r}AP_{B,r} \geq 0 \quad (2)$$

where $P_{B,r}$ denotes the orthogonal projection onto the top r eigenspace of B .

Proof of Continuity of γ_+

Fix $A \in \text{Sym}^+(\mathbb{C}^n)$. Let $(B_j)_{j \geq 1}$, $B_j \in \text{Sym}^+(\mathbb{C}^n)$, be a convergent sequence to A . We need to show $\gamma_+(B_j) \rightarrow \gamma_+(A)$.

Let $A = \sum_{k=1}^{n^2} x_k x_k^*$ be the optimal decomposition of A such that

$$\gamma_+(A) = \sum_{k=1}^{n^2} \|x_k\|_1^2.$$

If $A = 0$ then $\gamma_+(A) = 0$ and

$$0 \leq \gamma_+(B_j) \leq n \text{trace}(B_j) \leq n^2 \|B_j\|_{op}.$$

Hence $\lim_j \gamma_+(B_j) = 0$.

Assume $\text{rank}(A) = r > 0$ and let $\lambda_r > 0$ denote the smallest strictly positive eigenvalue of A . Let $\varepsilon \in (0, \frac{1}{2})$ be arbitrary. Let $J = J(\varepsilon)$ be so that

$\|A - B_j\|_{op} < \varepsilon \lambda_r$ for all $j > J$. Let $B_j = \sum_{k=1}^{n^2} y_{j,k} y_{j,k}^*$ be the optimal decomposition of B_j such that $\gamma_+(B_j) = \sum_{k=1}^{n^2} \|y_{j,k}\|_1^2$.

Let $\Delta_j = A - (1 - \varepsilon) P_{A,r} B_j P_{A,r}$. By Lemma L1, for any $j > J$,

$$\gamma_+(A) \leq (1 - \varepsilon) \gamma_+(P_{A,r} B_j P_{A,r}) + \gamma_+(\Delta_j) \leq (1 - \varepsilon) \sum_{k=1}^{n^2} \|P_{A,r} y_{j,k}\|_1^2 + n \text{trace}(\Delta_j)$$

Proof of Continuity of γ_+ (cont)

Pass to a subsequence j' of j so that $y_{j',k} \rightarrow y_k$, for every $k \in [n^2]$, and $\gamma_+(B_{j'}) \rightarrow \liminf_j \gamma_+(B_j)$. Then $\lim_{j'} P_{A,r} y_{j',k} = P_{A,r} y_k = y_k$ and

$$\lim_{j'} \sum_{k=1}^{n^2} \|P_{A,r} y_{j',k}\|_1^2 = \lim_{j'} \sum_{k=1}^{n^2} \|y_{j',k}\|_1^2 = \lim_j \inf \gamma_+(B_j)$$

On the other hand, $\lim_j \text{trace}(\Delta_j) = \varepsilon \text{trace}(A)$. Hence:

$$\gamma_+(A) \leq (1 - \varepsilon) \lim_j \inf \gamma_+(B_j) + \varepsilon \text{trace}(A)$$

Since $\varepsilon > 0$ is arbitrary, it follows $\gamma_+(A) \leq \liminf_j \gamma_+(B_j)$.

The inequality $\limsup_j \gamma_+(B_j) \leq \gamma_+(A)$ follows from Lemma L2 similarly: with

$\Delta_j = B_j - (1 - \varepsilon)P_{B_j,r}AP_{B_j,r}$ and $A = \sum_{k=1}^{n^2} x_k x_k^*$ optimal,

$$\gamma_+(B_j) \leq (1 - \varepsilon)\gamma_+(P_{B_j,r}AP_{B_j,r}) + n \text{trace}(\Delta_j) = (1 - \varepsilon) \sum_{k=1}^{n^2} \|P_{B_j,r} x_k\|_1^2 + n \text{trace}(\Delta_j).$$

Next take limsup of lhs by noticing $P_{B_j,r} \rightarrow P_{A,r}$ and $\limsup_j \|\Delta_j\|_{O_p} = \varepsilon \|A\|_{O_p}$:

$\limsup_j \gamma_+(B_j) \leq (1 - \varepsilon)\gamma_+(A) + n^2 \varepsilon \|A\|_{O_p}$. Take $\varepsilon \rightarrow 0$ and result follows. 

Proof of Lemmas

Proof of Lemma L1

Let $P = P_{A,r}$. and $\Delta = A - (1 - \varepsilon)P_{A,r}BP_{A,r}$. For any $x \in \mathbb{C}^n$:

$$\begin{aligned} \langle \Delta x, x \rangle &= \langle APx, Px \rangle - (1 - \varepsilon)\langle BPx, Px \rangle = \langle (A - (1 - \varepsilon)B)Px, Px \rangle = \\ &= \varepsilon\langle APx, Px \rangle + (1 - \varepsilon)\langle (A - B)Px, Px \rangle \geq \varepsilon\lambda_r\|Px\|^2 - (1 - \varepsilon)\|A - B\|_{Op}\|Px\|^2 \geq 0 \end{aligned}$$

because $\|A - B\|_{Op} \leq \frac{\varepsilon\lambda_r}{1 - \varepsilon}$.

Proof of Lemma L2

Let $P = P_{B,r}$ and $\Delta = B - (1 - \varepsilon)P_{B,r}AP_{B,r}$. Let $C = B - P_{B,r}BP_{B,r} \geq 0$. Let μ_r be the r^{th} eigenvalue of B . Note $|\mu_r - \lambda_r| \leq \|A - B\|_{Op} \leq \varepsilon\lambda_r$. Thus $\mu_r \geq (1 - \varepsilon)\lambda_r$. For any $x \in \mathbb{C}^n$:

$$\begin{aligned} \langle \Delta x, x \rangle &= \langle Cx, x \rangle + \langle BPx, Px \rangle - (1 - \varepsilon)\langle APx, Px \rangle = \langle Cx, x \rangle + \varepsilon\langle BPx, Px \rangle + \\ &+ (1 - \varepsilon)\langle (B - A)Px, Px \rangle \geq \langle Cx, x \rangle + (\varepsilon\mu_r - (1 - \varepsilon)\|A - B\|_{Op})\|Px\|^2 \geq 0 \end{aligned}$$

because $\|A - B\|_{Op} \leq \varepsilon\lambda_r \leq \frac{\varepsilon\mu_r}{1 - \varepsilon}$.

Notations

Recall the setup.

Take $A \in \text{Sym}^+(\mathbb{C}^n) := \{A \in \mathbb{C}^{n \times n}, A^* = A \geq 0\}$.

We are interested in this quantity:

$$\gamma_+(A) := \inf_{A = \sum_{k \geq 1} x_k x_k^*} \sum_k \|x_k\|_1^2$$

Recall definitions of norms:

$$\|A\|_1 = \sum_{k,l=1}^n |A_{k,l}|, \quad \|A\|_{Op} = \max_{\|x\|_2=1} \|Ax\|_2 = s_{\max}(A)$$

The *matrix conjecture*: There is a universal constant C_0 such that, for every $n \geq 1$ and $A \in \text{Sym}^+(\mathbb{C}^n)$,

$$\gamma_+(A) \leq C_0 \|A\|_1$$