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Antoine Mhanna. Norm Inequalities of Positive Semi-Definite Matrices. 2015. hal-01182244v1

HAL Id: hal-01182244 https://inria.hal.science/hal-01182244v1

Preprint submitted on 14 Aug 2015 (v1), last revised 14 Sep 2015 (v5)

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Norm Inequalities of Positive Semi-Definite Matrices

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Abstract

For positive block-matrix we write $M = \begin{pmatrix} A & X \\ X^* & B \end{pmatrix} \in \mathbb{M}_{n+m}^+$, with $A \in \mathbb{M}_n^+$, $B \in \mathbb{M}_m^+$. The main result is first to study the consequences of a decomposition lemma due to C. Bourrin and second to extend the class of these P.S.D. matrices M written by blocks that satisfies the inequality: $||M|| \leq ||A + B||$ for all symmetric norms and to give examples whenever it is necessary.

Keywords : Matrix Analysis, Hermitian matrices, symmetric norms.

1 Introduction

Let A be an $n \times n$ matrix and F an $m \times m$ matrix, (m > n) written by blocks such that A is a diagonal block and all entries other than those of A are zeros, then the two matrices have the same singular values and for all unitarily invariant norms $||A|| = ||F|| = ||A \oplus 0||$, we say then that the symmetric norm on \mathbb{M}_m induces a symmetric norm on \mathbb{M}_n , so for square matrices we may assume that our norms are defined on all spaces \mathbb{M}_n , $n \ge 1$. The spectral norm is denoted by $||.||_s$, the Frobenius norm by $||.||_{(2)}$, and the Ky Fan k-norms by $||.||_k$. Let \mathbb{M}_n^+ denote the set of positive and semi-definite part of the space of $n \times n$ complex matrices and M be any positive block-matrices; that is, $M = \begin{pmatrix} A & X \\ X^* & B \end{pmatrix} \in \mathbb{M}_{n+m}^+$, with $A \in \mathbb{M}_n^+$, $B \in \mathbb{M}_m^+$.

Decomposition of block-matrices $\mathbf{2}$

Lemma 2.1. For every matrix M in \mathbb{M}_{n+m}^+ written in blocks, we have the decomposition: $\begin{pmatrix} A & X \\ X^* & B \end{pmatrix} = U \begin{pmatrix} A & 0 \\ 0 & 0 \end{pmatrix} U^* + V \begin{pmatrix} 0 & 0 \\ 0 & B \end{pmatrix} V^*$

for some unitaries $U, V \in \mathbb{M}_{n+m}$.

Proof. Factorize the positive matrix as a square of positive matrices:

$$\begin{pmatrix} A & X \\ X^* & B \end{pmatrix} = \begin{pmatrix} C & Y \\ Y^* & D \end{pmatrix} \cdot \begin{pmatrix} C & Y \\ Y^* & D \end{pmatrix}$$

we verify that the right hand side can be written as $T^*T + S^*S$ so :

$$\begin{pmatrix} C & Y \\ Y^* & D \end{pmatrix} \cdot \begin{pmatrix} C & Y \\ Y^* & D \end{pmatrix} = \underbrace{\begin{pmatrix} C & 0 \\ Y^* & 0 \end{pmatrix}}_{T^*} \cdot \underbrace{\begin{pmatrix} C & Y \\ 0 & 0 \end{pmatrix}}_{T} + \underbrace{\begin{pmatrix} 0 & Y \\ 0 & D \end{pmatrix}}_{S^*} \cdot \underbrace{\begin{pmatrix} 0 & 0 \\ Y^* & D \end{pmatrix}}_{S} \cdot$$
Since $TT^* = \begin{pmatrix} CC + YY^* & 0 \\ 0 & 0 \end{pmatrix} = \begin{pmatrix} A & 0 \\ 0 & 0 \end{pmatrix}, SS^* = \begin{pmatrix} 0 & 0 \\ 0 & Y^*Y + DD \end{pmatrix} = \begin{pmatrix} 0 & 0 \\ 0 & B \end{pmatrix}$ and AA^* is unitarily congruent to A^*A for any square matrix A , the lemma follows. \Box

Remark 1. As a consequence of this lemma we have:

$$||M|| \le ||A|| + ||B||$$

for all symmetric norms.

Equations involving unitary matrices are called unitary orbits representations. Recall that if $A \in \mathbb{M}_n$, $R(A) = \frac{A + A^*}{2}$ and $I(A) = \frac{A - A^*}{2i}$.

Corollary 2.1. For every matrix in \mathbb{M}_{2n}^+ written in blocks of the same size, we have the decomposition:

$$\begin{pmatrix} A & X \\ X^* & B \end{pmatrix} = U \begin{pmatrix} \frac{A+B}{2} - R(X) & 0 \\ 0 & 0 \end{pmatrix} U^* + V \begin{pmatrix} 0 & 0 \\ 0 & \frac{A+B}{2} + R(X) \end{pmatrix} V^*$$

for some unitaries $U, V \in \mathbb{M}_{2n}$.

Proof. Let $J = \frac{1}{\sqrt{2}} \begin{pmatrix} I & -I \\ I & I \end{pmatrix}$ where I is the identity of \mathbb{M}_n , J is a unitary matrix, and we have:

$$J\begin{pmatrix} A & X\\ X^* & B \end{pmatrix} J^* = \underbrace{\begin{pmatrix} \frac{A+B}{2} - R(X) & \frac{A-B}{2} + \frac{X^*-X}{2} \\ \frac{A-B}{2} - \frac{X-X^*}{2} & \frac{A+B}{2} + R(X) \end{pmatrix}}_{N}$$

Now we factorize N as a square of positive matrices:

$$\begin{pmatrix} A & X \\ X^* & B \end{pmatrix} = J^* \begin{pmatrix} L & M \\ M^* & F \end{pmatrix} \cdot \begin{pmatrix} L & M \\ M^* & F \end{pmatrix} J$$

and let:

$$\delta = J^* \begin{pmatrix} L & M \\ M^* & F \end{pmatrix} = \frac{1}{\sqrt{2}} \begin{pmatrix} L+M^* & M+F \\ M^*-L & F-M \end{pmatrix}$$
$$\psi = \begin{pmatrix} L & M \\ M^* & F \end{pmatrix} J = \frac{1}{\sqrt{2}} \begin{pmatrix} L+M & M-L \\ F+M^* & F-M^* \end{pmatrix}$$

A direct computation shows that:

$$\delta.\psi = \frac{1}{2} \begin{pmatrix} (L+M^*)(L+M) + (M+F)(F+M^*) & (L+M^*)(M-L) + (M+F)(F-M^*) \\ \\ (M^*-L)(L+M) + (F-M)(F+M^*) & (M^*-L)(M-L) + (F-M)(F-M^*) \end{pmatrix}$$

$$= \Gamma^*\Gamma + \Phi^*\Phi$$
(1)

where: $\Gamma = \frac{1}{\sqrt{2}} \begin{pmatrix} L+M & M-L \\ 0 & 0 \end{pmatrix}$, and $\Phi = \frac{1}{\sqrt{2}} \begin{pmatrix} 0 & 0 \\ F+M^* & F-M^* \end{pmatrix}$ to finish notice that for any square matrix A, A^*A is unitarily congruent to AA^* and, $\Gamma\Gamma^*$, $\Phi\Phi^*$ have the required form.

The previous corollary implies that $\frac{A+B}{2} \ge R(X)$ and $\frac{A+B}{2} \ge -R(X)$.

Corollary 2.2. For every matrix in \mathbb{M}_{2n}^+ written in blocks of the same size, we have the decomposition:

$$\begin{pmatrix} A & X \\ X^* & B \end{pmatrix} = U \begin{pmatrix} \frac{A+B}{2} + I(X) & 0 \\ 0 & 0 \end{pmatrix} U^* + V \begin{pmatrix} 0 & 0 \\ 0 & \frac{A+B}{2} - I(X) \end{pmatrix} V^*$$

for some unitaries $U, V \in \mathbb{M}_{n+m}$.

Proof. The proof is similar to Corollary 2.1, we have: $J_1 \begin{pmatrix} A & X \\ X^* & B \end{pmatrix} J_1^* = \begin{pmatrix} A & iX \\ -iX^* & B \end{pmatrix}$ where $J_1 = \begin{pmatrix} I & 0 \\ 0 & -iI \end{pmatrix}$, and $K = JJ_1 \begin{pmatrix} A & X \\ X^* & B \end{pmatrix} J_1^* J^* = \begin{pmatrix} \frac{A+B}{2} + I(X) & * \\ * & \frac{A+B}{2} - I(X) \end{pmatrix}$

here (*) means an unspecified entry, the proof is similar to that in Corollary 2.1 but for reader's convenience we give the main headlines: first factorize K as a square of positive matrices; that is, $M = \begin{pmatrix} A & X \\ X^* & B \end{pmatrix} = J_1^* J^* L^2 J J_1$ next decompose L^2 as in Lemma 2.1 to obtain

$$M = J_1^* J^* (T^*T + S^*S) J J_1 = J_1^* J^* (T^*T) J J_1 + J_1^* J^* (S^*S) J J_1$$

where $TT^* = \begin{pmatrix} \frac{A+B}{2} + I(X) & 0\\ 0 & 0 \end{pmatrix}$ and $SS^* = \begin{pmatrix} 0 & 0\\ 0 & \frac{A+B}{2} - I(X) \end{pmatrix}$ finally the congruence property completes the proof.

The existence of unitaries U and V in the decomposition process need not to be unique as one can take the special case; that is, M any diagonal matrix with diagonal entries equals a nonnegative number k, explicitly $M = kI = U\left(\frac{k}{2}I\right)U^* + V\left(\frac{k}{2}I\right)V^*$ for any U and V unitaries.

Remark 2. Notice that from the Courant-Fischer theorem if $A, B \in \mathbb{M}_n^+$, then the eigenvalues of each matrix are the same as the singular values and $A \leq B \Longrightarrow ||A||_k \leq ||B||_k$, for all $k = 1, \dots, n$, also $A < B \Longrightarrow ||A||_k < ||B||_k$, for all $k = 1, \dots, n$.

Corollary 2.3. For every matrix in \mathbb{M}_{2n}^+ written in blocks of the same size, we have:

$$\begin{pmatrix} A & X \\ X^* & B \end{pmatrix} \leq \frac{1}{2} \left\{ U \begin{pmatrix} A+B+|X-X^*| & 0 \\ 0 & 0 \end{pmatrix} U^* + V \begin{pmatrix} 0 & 0 \\ 0 & A+B+|X-X^*| \end{pmatrix} V^* \right\}$$

for some unitaries $U, V \in \mathbb{M}_{n+m}$.

Proof. This a consequence of the fact that $I(X) \leq |I(X)|$.

3 Symmetric Norms and Inequalities

In [1] they found that if X is hermitian then

$$\|M\| \le \|A + B\| \tag{2}$$

for all symmetric norms. It has been given counter-examples showing that this does not necessarily holds if X is a normal but not Hermitian matrix, the main idea of this section is to give examples and counter-examples in a general way and to extend the previous inequality to a larger class of P.S.D. matrices written by blocks satisfying (2).

Theorem 3.1. If A and B are positive definite matrices of same size. Then

$$\begin{pmatrix} A & X \\ X^* & B \end{pmatrix} > 0 \Longleftrightarrow A \ge XB^{-1}X^*$$

Proof. Write $\begin{pmatrix} A & X \\ X^* & B \end{pmatrix} = \begin{pmatrix} I & XB^{-1} \\ 0 & I \end{pmatrix} \begin{pmatrix} A - XB^{-1}X^* & 0 \\ 0 & B \end{pmatrix} \begin{pmatrix} I & 0 \\ XB^{-1} & I \end{pmatrix}$ where I is the identity matrix, and that complete the proof since for any matrix A,

$$A \ge 0 \Longleftrightarrow X^*AX \ge 0, \ \forall X$$

Theorem 3.2. Let $M = \begin{pmatrix} A & B \\ C & D \end{pmatrix}$ be any square matrix written by blocks of same size, if AC = CA then $\det(M) = \det(AD - CB)$

Proof. Suppose first that A is invertible, let us write M as

$$M = \begin{pmatrix} Z & 0 \\ V & I \end{pmatrix} \begin{pmatrix} I & E \\ 0 & F \end{pmatrix}$$
(3)

upon calculation we find that: Z = A, V = C, $E = A^{-1}B$, $F = D - CA^{-1}B$ taking the determinant on each side of (3) we get:

$$\det(M) = \det(A(D - CA^{-1}B)) = \det(AD - CB)$$

the result follows by a continuity argument since the Determinant function is a continuous function. $\hfill \square$

Given the matrix $M = \begin{pmatrix} A & X \\ X^* & 0 \end{pmatrix}$ a matrix in \mathbb{M}_{2n}^+ written by blocks of same size, we know that it M is not P.S.D., to see this notice that all the 2 × 2 extracted principle submatrices of M are P.S.D if and only if X = 0 and A is positive semi-definite. Even if a proof of this exists and would take two lines, it is quite nice to see a different constructive proof, a direct consequence of Lemma 2.1.

Theorem 3.3. Given
$$\begin{pmatrix} A & X \\ X^* & B \end{pmatrix}$$
 a matrix in \mathbb{M}_{2n}^+ written in blocks of same size:
1. If $\begin{pmatrix} A & X \\ X^* & 0 \end{pmatrix}$ is positive semi-definite, $I(X) > 0$ or $I(X) < 0$, then there exist a matrix Y such that $M = \begin{pmatrix} A & Y \\ Y^* & 0 \end{pmatrix}$ is positive semi-definite and:
 $\left\| \begin{pmatrix} A & Y \\ Y^* & 0 \end{pmatrix} \right\| > \|A\|$
(4)

for all symmetric norms.

r

2. If $\begin{pmatrix} 0 & X \\ X^* & B \end{pmatrix}$ is positive semi-definite, I(X) > 0 or I(X) < 0 then there exist a matrix Y such that $M = \begin{pmatrix} 0 & Y \\ Y^* & B \end{pmatrix}$ is positive semi-definite and:

$$\left\| \begin{pmatrix} 0 & Y \\ Y^* & B \end{pmatrix} \right\| > \|B\| \tag{5}$$

The same result holds if we replaced I(X) by R(X) because $\begin{pmatrix} A & iX \\ -iX^* & B \end{pmatrix}$ is unitarily congruent to $\begin{pmatrix} A & X \\ X^* & B \end{pmatrix}$.

Proof. Without loss of generality we can consider I(X) > 0 cause $\begin{pmatrix} A & X \\ X^* & B \end{pmatrix}$ and $\begin{pmatrix} A & -X \\ -X^* & B \end{pmatrix}$ are unitarily congruent, we will show the first statement as the second one has a similar proof, from Corollary 2.2 we have:

$$\begin{pmatrix} A & X \\ X^* & 0 \end{pmatrix} \ge U \begin{pmatrix} \frac{A}{2} & 0 \\ 0 & 0 \end{pmatrix} U^* + U \begin{pmatrix} I(X) & 0 \\ 0 & 0 \end{pmatrix} U^* + V \begin{pmatrix} 0 & 0 \\ 0 & \frac{A}{2} \end{pmatrix} V^*$$

Since $\begin{pmatrix} A & X \\ X^* & 0 \end{pmatrix}$ is congruent to $L = \begin{pmatrix} A & lX \\ lX^* & 0 \end{pmatrix}$ for any $l \in \mathbb{C}$, L is P.S.D. A is a fixed matrix, we have $\left\| U \begin{pmatrix} \frac{A}{2} & 0 \\ 0 & 0 \end{pmatrix} U^* + V \begin{pmatrix} 0 & 0 \\ 0 & \frac{A}{2} \end{pmatrix} V^* \right\|_k = \beta \|A\|_k$ for some $\beta \leq 1$ finally we set Y = lX where $l \in \mathbb{R}$ is large enough to have $\|M\|_k > \|A\|_k$, $\forall k$ thus $\|M\| > \|A\|$ for all symmetric norms.

Notice that there exist a permutation matrix P such that $P\begin{pmatrix} A & X \\ X^* & 0 \end{pmatrix} P^{-1} = \begin{pmatrix} 0 & X^* \\ X & A \end{pmatrix}$ and since I(X) > 0 if and only if $I(X^*) < 0$, the two assertions of Theorem 3.3 are equivalent up to a permutation similarity.

Corollary 3.1. If $M = \begin{pmatrix} A & X \\ X^* & 0 \end{pmatrix}$, A a positive semi-definite matrix, and we have one of the following conditions:

- 1. R(X) > 0
- 2. R(X) < 0
- 3. I(X) > 0

4. I(X) < 0

Then M can't be positive semi-definite.

Proof. By Remark 1 any positive matrix M written in blocks must satisfy $||M|| \le ||A|| + ||B||$ for all symmetric norms which is not the case of the matrix M constructed in Theorem 3.3.

Finally we get:

Theorem 3.4. If $X \neq 0$ and B = 0, $A \ge 0$, the matrix $M = \begin{pmatrix} A & X \\ X^* & 0 \end{pmatrix}$ cannot be positive semi-definite.

Proof. Suppose the converse, so $M = \begin{pmatrix} A & X \\ X^* & 0 \end{pmatrix}$ is positive semi-definite, without loss of generality the only case we need to discuss is when R(X) has positive and negative eigenvalues, by Corollary 2.1 we can write:

$$M = U \begin{pmatrix} \frac{A}{2} - R(X) & 0\\ 0 & 0 \end{pmatrix} U^* + V \begin{pmatrix} 0 & 0\\ 0 & \frac{A}{2} + R(X) \end{pmatrix} V^*$$

for some unitaries $U, V \in \mathbb{M}_{2n}$.

Now if R(X) has $-\alpha$ the smallest negative eigenvalue $R(X) + (\alpha + \epsilon)I > 0$ consequently the matrix

$$H = U \begin{pmatrix} \frac{A}{2} - R(X) & 0\\ 0 & 0 \end{pmatrix} U^* + V \begin{pmatrix} 0 & 0\\ 0 & \frac{A}{2} + (\alpha + \epsilon)I + R(X) + (\alpha + \epsilon)I \end{pmatrix} V^*$$
(6)

$$= \begin{pmatrix} A + 2(\alpha + \epsilon)I & X + (\alpha + \epsilon)I \\ (X + (\alpha + \epsilon)I)^* & 0 \end{pmatrix}$$
(7)

is positive semi-definite with R(Y) > 0, where $Y = X + (\alpha + \epsilon)I$, by Corollary 3.1 this is a contradiction.

A natural question would be how many are the nontrivial P.S.D.matrices written by blocks ? The following lemma will show us how to construct some of them.

Lemma 3.1. Let A and B be any $n \times n$ positive definite matrices, then there exist an integer $t \ge 1$ such that the matrix $F_t = \begin{pmatrix} tA & X \\ X^* & tB \end{pmatrix}$ is positive definite.

Proof. Recall from Theorem 3.1 that F_1 is positive definite if and only if $A > XB^{-1}X^*$, which is equivalent to $x^*Ax > x^*XB^{-1}X^*x$ for all $x \in \mathbb{C}^n$. Set $f(x) := x^*Ax$ and $g(x) := x^*XB^{-1}X^*x$ and let us suppose, to the contrary, that there exist a vector z such that $f(z) \leq g(z)$ since f(x) and g(x) are homogeneous functions of degre d = 2 over \mathbb{R} if $f(x) \geq g(x)$ for all x such that $||x||_s = 1$ then $f(x) \geq g(x)$ for any $x \in \mathbb{C}^n$. So let us set $K = \max_{\|x\|_s = 1} g(x)$, and $L = \min_{\|x\|_s = 1} f(x)$ since g(x) and f(x) are continuous functions and $\{x; \|x\|_s = 1\}$ is compact, there exist a vector w respectively v such that K = g(w), respectively L = f(v). Now choose $t \geq 1$ such that $tf(v) > \frac{g(w)}{t}$, to obtain

$$x^*(tA)x \geq v^*(tA)v > w^*X(tB)^{-1}X^*w \geq x^*X(tB)^{-1}X^*x$$

for all x such that $||x||_s = 1$, thus $x^*(tA)x > x^*X(tB)^{-1}X^*x$ for any $x \in \mathbb{C}^n$ which completes the proof.

Theorem 3.5. Let $A = diag(\lambda_1, \dots, \lambda_n)$, $B = diag(\nu_1, \dots, \nu_n)$ and $M = \begin{pmatrix} A & X \\ X^* & B \end{pmatrix}$ a given positive semi-definite matrix. If X^* commute with A and X^*X equals a diagonal matrix, then

$$\|M\| \le \|A + B\|$$

for all symmetric norms. The same inequality holds if X commute with B and XX^* is diagonal.

Proof. From Corollary ?? it suffices to prove the inequality for the Ky Fan k-norms $k = 1, \dots, n$. Let $P = \begin{pmatrix} 0 & I_n \\ I_n & 0 \end{pmatrix}$ where I_n is the identity matrix of order n, since $\begin{pmatrix} B & X^* \\ X & A \end{pmatrix} = P \begin{pmatrix} A & X \\ X^* & B \end{pmatrix} P^{-1}$ and $\begin{pmatrix} A & X \\ X^* & B \end{pmatrix}$ have same singular values, we will discuss only the first case; that is, when X^* commute with A and X^*X is diagonal, as the second case will follows. Let $D := X^*X = \begin{pmatrix} d_1 & 0 & \dots & 0 \\ 0 & d_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & d_n \end{pmatrix}$, as X^* commute with A, from Theorem 3.2 we conclude that the eigenvalues of $\begin{pmatrix} A & X \\ X^* & B \end{pmatrix}$ are the roots of

$$\det((A - \mu I_n)(B - \mu I_n) - D) = 0$$

Equivalently the eigenvalues are all the solutions of the n equations:

$$\begin{aligned} &(\lambda_1 - \mu)(\nu_1 - \mu) - d_1 &= 0\\ &(\lambda_2 - \mu)(\nu_2 - \mu) - d_2 &= 0\\ &(\lambda_3 - \mu)(\nu_3 - \mu) - d_3 &= 0\\ &\vdots\\ &(\lambda_n - \mu)(\nu_n - \mu) - d_n &= 0\end{aligned}$$

Each equation is of 2^{nd} degree, if we denote by a_i and b_i the two solutions of the i^{th} equation we deduce that:

$$a_1 + b_1 = \lambda_1 + \nu_1$$

$$a_2 + b_2 = \lambda_2 + \nu_2$$

$$\vdots$$

$$a_n + b_n = \lambda_n + \nu_n$$

But

$$A + B = \begin{pmatrix} \lambda_1 + \nu_1 & 0 & \cdots & 0 \\ 0 & \lambda_2 + \nu_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & \lambda_n + \nu_n \end{pmatrix}$$

and each diagonal entry of A + B is equal the sum of two nonegative eigenvalues of M, thus we have necessarily: $||M||_k \leq ||A + B||_k$ for all $k = 1, \dots, n$ which completes the proof.

Example 3.1. Let

$$M_x = \begin{pmatrix} x & 0 & \frac{i}{2} & 0\\ 0 & \frac{99}{100} & 0 & -\frac{i}{2}\\ -\frac{i}{2} & 0 & \frac{99}{100} & 0\\ 0 & \frac{i}{2} & 0 & \frac{1}{2} \end{pmatrix}$$

If $\frac{3}{10} \le x \le \frac{1}{2}$, M_x is positive definite and we have:

$$\|M_x\| \le \|A + B\| \tag{8}$$

for all symmetric norms, where $A = \begin{pmatrix} x & 0 \\ 0 & \frac{99}{100} \end{pmatrix}$ and $B = \begin{pmatrix} \frac{99}{100} & 0 \\ 0 & \frac{1}{2} \end{pmatrix}$. From Corollary ?? if M_x is positive definite for $x = \frac{3}{10}$ then M_x is P.D. for all $x > \frac{3}{10}$. The eigenvalues of

 $M_{\frac{3}{10}}$ which are the same as the singular values of $M_{\frac{3}{10}}$ are:

$$\lambda_1 = \frac{149}{200} + \frac{\sqrt{12401}}{200} \approx 1.301 \tag{9}$$

$$\lambda_2 = \frac{129}{200} + \frac{\sqrt{14761}}{200} \approx 1.25 \tag{10}$$

$$\lambda_3 = \frac{149}{200} - \frac{\sqrt{12401}}{200} \approx 0.188 \tag{11}$$

$$\lambda_4 = \frac{129}{200} - \frac{\sqrt{14761}}{200} \approx 0.0375 \tag{12}$$

And the (8) inequality follows from Theorem 3.5.

Let us study the commutation condition in Theorem 3.5. First notice that any square matrix $X = (x_{ij}) \in \mathbb{M}_n$ will commute with $A = \text{diag}(a_1, \dots, a_n)$ if and only if :

$$\begin{pmatrix} x_{1,1}a_1 & x_{1,2}a_2 & \cdots & x_{1,n}a_n \\ x_{2,1}a_1 & x_{2,2}a_2 & \cdots & x_{2,n}a_n \\ \vdots & \vdots & \ddots & \vdots \\ x_{n,1}a_1 & x_{n,2}a_2 & \cdots & x_{n,n}a_n \end{pmatrix} = \begin{pmatrix} x_{1,1}a_1 & x_{1,2}a_1 & \cdots & x_{1,n}a_1 \\ x_{2,1}a_2 & x_{2,2}a_2 & \cdots & x_{2,n}a_2 \\ \vdots & \vdots & \ddots & \vdots \\ x_{n,1}a_n & x_{n,2}a_n & \cdots & x_{n,n}a_n \end{pmatrix}$$

Corollary 3.2. Let $A = diag(\lambda_1, \dots, \lambda_n)$, $B = diag(\nu_1, \dots, \nu_n)$ and $M = \begin{pmatrix} A & X \\ X^* & B \end{pmatrix}$ a given positive semi-definite matrix. If X^* commute with A, or X commute with B, then $\|M\| \le \|A + B\|$ for all symmetric norms.

Proof. As in Theorem 3.5, we will assume without loss of generality that X^* commute with A, as the other case is similar. From the commutation condition we have.

Corollary 3.3. Let $M = \begin{pmatrix} A & X \\ X^* & B \end{pmatrix}$ be a positive semi-definite matrix written by blocks. There exist a unitary V and a unitary U such that

$$\left\| \begin{pmatrix} A & X \\ X^* & B \end{pmatrix} \right\| \le \left\| UAU^* + VBV^* \right\| := \left\| A \right\| + \left\| B \right\|$$

for all symmetric norms.

Proof. Let U and V be two unitary matrix such that $UAU^* = D_o$ and $VBV^* = G_o$ where D_o and G_o are two diagonal matrices having the same ordering o, of eigenvalues with respect to their indexes i.e., if $\lambda_n \leq \cdots \leq \lambda_1$ are the diagonal entries of D_o , and $\nu_n \leq \cdots \leq \nu_1$ are those of G_o , then if λ_i is in the (j, j) position then ν_i will be also. Consequently $||UAU^* + VBV^*|| = ||D_o + G_o|| = ||D_o|| + ||G_o|| = ||A|| + ||B||$, for all the Ky-Fan k-norms and thus for all symmetric norms. To complete the proof notice that if $T = UXV^*$ and Q is the unitary matrix $\begin{pmatrix} U & 0 \\ 0 & V \end{pmatrix}$, by Remark 1

$$\left\| \begin{pmatrix} A & X \\ X^* & B \end{pmatrix} \right\| = \left\| Q \begin{pmatrix} A & X \\ X^* & B \end{pmatrix} Q^* \right\| = \left\| \begin{pmatrix} D_o & T \\ T^* & G_o \end{pmatrix} \right\| \le \|D_o\| + \|G_o\|$$
(13)
metric norms.

for all symmetric norms.

Lemma 3.2. Let

$$N = \begin{pmatrix} \begin{pmatrix} a_1 & 0 & \cdots & 0 \\ 0 & a_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & a_n \end{pmatrix} \qquad D \\ D^* \qquad \begin{pmatrix} b_1 & 0 & \cdots & 0 \\ 0 & b_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & b_n \end{pmatrix} \end{pmatrix}$$

where a_1, \dots, a_n respectively b_1, \dots, b_n are nonnegative respectively negative real numbers, $bers, A = \begin{pmatrix} a_1 & 0 & \dots & 0 \\ 0 & a_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & a_n \end{pmatrix}, B = \begin{pmatrix} b_1 & 0 & \dots & 0 \\ 0 & b_2 & \dots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \dots & b_n \end{pmatrix}$ and D is any diagonal matrix, then nor N neither -N is positive semi-definite. If $a_i + b_i \ge 0$ for all $i \le n$ then $\|N\|_s > \|A + B\|_s$.

Proof. The diagonal of N has negative and positive numbers, thus nor N neither -Nis positive semi-definite, now any two diagonal matrices will commute, in particular D^* and A, by applying Theorem 3.2 we get that the eigenvalues of N are the roots of

$$\det((A - \mu I_n)(B - \mu I_n) - D^*D) = 0$$

Equivalently the eigenvalues are all the solutions of the following equations:

$$(a_1 - \mu)(b_1 - \mu) - d_1 = 0$$

$$(a_2 - \mu)(b_2 - \mu) - d_2 = 0$$

$$(a_3 - \mu)(b_3 - \mu) - d_3 = 0$$

$$\vdots$$

$$(a_n - \mu)(b_n - \mu) - d_n = 0$$

where (d_1, \dots, d_n) are the diagonal entries of D^*D . Let us denote by x_i and y_i the two solutions of the i^{th} equation then:

$$\begin{array}{rcrcrcrcrc} x_1 + y_1 & = & a_1 + b_1 & \geq & 0 \\ x_2 + y_2 & = & a_2 + b_2 & \geq & 0 \\ & & \vdots & & \\ x_n + y_n & = & a_n + b_n & \geq & 0 \end{array}$$

It seems easy to construct examples of non P.S.D matrices N, such that $||N||_s > ||A + B||_s$, let us have a look of such inequality for P.S.D. matrices.

Example 3.2. Let

$$C = \begin{pmatrix} \frac{4}{3} & 0 & 1 & -1\\ 0 & 1 & 0 & \frac{1}{5}\\ 1 & 0 & \frac{3}{2} & 0\\ -1 & \frac{1}{5} & 0 & 2 \end{pmatrix} = \begin{pmatrix} A & X\\ X^* & B \end{pmatrix}$$

where $A = \begin{pmatrix} \frac{4}{3} & 0\\ 0 & 1 \end{pmatrix}$, $B = \begin{pmatrix} \frac{3}{2} & 0\\ 0 & 2 \end{pmatrix}$. Since the eigenvalues of C are all positive with $\lambda_1 \approx 3.008$, $\lambda_2 \approx 1.7$, $\lambda_3 \approx 0.9$, $\lambda_4 \approx 0.089$, C is positive definite and we verify that

$$3.008 \approx \|C\|_s > \|A + B\|_s = 3$$

Example 3.3. Let

$$N_y = \begin{pmatrix} 2 & 0 & 0 & 2\\ 0 & y & 0 & 0\\ 0 & 0 & 1 & 0\\ 2 & 0 & 0 & 2 \end{pmatrix} = \begin{pmatrix} A & X\\ X^* & B \end{pmatrix}$$

where $A = \begin{pmatrix} 2 & 0 \\ 0 & 1 \end{pmatrix}$ and $B = \begin{pmatrix} 1 & 0 \\ 0 & 2 \end{pmatrix}$. The eigenvalues of N_y are the numbers: $\lambda_1 = 4$, $\lambda_2 = 1$, $\lambda_3 = y$, $\lambda_4 = 0$, thus if $y \ge 0$, N_y is positive semi-definite and for all y such that $0 \le y < 1$ we have

1.
$$4 = ||N_y||_s > ||A + B||_s = 3$$

2.
$$16 + y^2 + 1 = ||N||_{(2)}^2 > ||A + B||_{(2)}^2 = 4(3 + y) + y^2 + 1$$

References

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