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# RESEARCH

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# Some inequalities related to $2 \times 2$ block sector partial transpose matrices

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## Abstract

In this article, two inequalities related to  $2 \times 2$  block sector partial transpose matrices are proved, and we also present a unitarily invariant norm inequality for the Hua matrix which is sharper than an existing result.

MSC: 15A45; 15A60

Keywords: Sector partial transpose matrices; Unitarily invariant norms; Linear maps

## **1** Introduction

We denote by  $\mathbb{M}_n$  the set of  $n \times n$  complex matrices.  $\mathbb{M}_n(\mathbb{M}_k)$  is the set of  $n \times n$  block matrices with each block in  $\mathbb{M}_k$ . The  $n \times n$  identity matrix is denoted by  $I_n$ . We use  $\|\cdot\|$  for an arbitrary unitarily invariant norm. A positive semidefinite matrix A will be expressed as  $A \ge 0$ . Likewise, we write A > 0 to refer that A is a positive definite matrix. The singular values of A, denoted by  $s_1(A), s_2(A), \ldots, s_n(A)$ , are the eigenvalues of the positive semidefinite matrix  $|A| = (A^*A)^{1/2}$ , arranged in decreasing order and repeated according to multiplicity as  $s_1(A) \ge s_2(A) \ge \cdots \ge s_n(A)$ . When A is Hermitian, we enumerate eigenvalues of A in nonincreasing order  $\sigma_1(A) \ge \sigma_2(A) \ge \cdots \ge \sigma_n(A)$ . Recall that  $C \in \mathbb{M}_{m \times n}$  is (strictly) contractive if  $(I_n > C^*C) \ I_n \ge C^*C$ . The geometric mean of two positive definite matrices  $A, B \in \mathbb{M}_n$ , denoted by  $A \sharp B$ , is the positive definite solution of the Riccati equation  $XB^{-1}X = A$  and has the explicit expression  $A \ddagger B = A^{\frac{1}{2}}(A^{-\frac{1}{2}}BA^{-\frac{1}{2}})^{\frac{1}{2}}A^{\frac{1}{2}}$ . More details on the matrix geometric mean can be found in [2, Chap. 4].

The numerical range of  $A \in \mathbb{M}_n$  is defined by

$$W(A) = \{x^*Ax | x \in \mathbb{C}^n, x^*x = 1\}.$$

For basic properties of numerical range, see [5]. Also, we define a sector on the complex plane

$$S_{\alpha} = \{z \in \mathbb{C} | \Re z \ge 0, |\Im z| \le (\Re z) \tan(\alpha)\}, \quad \alpha \in \left[0, \frac{\pi}{2}\right).$$

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Actually, the class of matrices T with  $W(T) \subseteq S_{\alpha}$  and the class of T with positive definite real part (i.e. accretive matrices) are both called sector matrices. Sector matrices have been the subject of a number of recent papers [3, 8, 14].

A matrix  $H = (H_{ij})_{i,j=1}^n \in \mathbb{M}_n(\mathbb{M}_k)$  is said to be positive partial transpose (i.e. PPT) if H is positive semidefinite and its partial transpose  $H^{\tau} = (H_{ji})_{j,i=1}^n$  is also positive semidefinite. Inspired by PPT, Kuai [6] defined a new conception called sectorial partial transpose (i.e. SPT). That is, if  $W(A) \subseteq S_{\alpha}$  for  $A = (A_{ij})_{i,j=1}^n \in \mathbb{M}_n(\mathbb{M}_k)$ , then  $W(A^{\tau}) \subseteq S_{\alpha}$ . Thus, it is natural to extend the results for PPT matrices to SPT matrices.

Hiroshima [4, Theorem 1] proved the following result.

**Theorem 1.1** Let  $H = \begin{pmatrix} A & X \\ X^* & B \end{pmatrix} \in \mathbb{M}_2(\mathbb{M}_n)$  be PPT. Then

$$\|H\| \le \|A + B\|. \tag{1}$$

As the application of Theorem 1.1, Lin and Hiroshima [10, Theorem 3.3] presented a relation between the norm of diagonal blocks of the Hua matrix, e.g., [12] and the norm of its off diagonal blocks.

**Theorem 1.2** If the Hua matrix is given by

$$H := \begin{pmatrix} (I - A^*A)^{-1} & (I - B^*A)^{-1} \\ (I - A^*B)^{-1} & (I - B^*B)^{-1} \end{pmatrix},$$

where  $A, B \in \mathbb{M}_{m \times n}$  are strictly contractive, then

$$2\|(I - A^*B)^{-1}\| \le \|(I - A^*A)^{-1} + (I - B^*B)^{-1}\|$$
(2)

for any unitarily invariant norm.

Actually, it was only recently observed that *H* is PPT; see [1].

Lin [7] obtained a singular value inequality for PPT matrices related to a linear map.

**Theorem 1.3** Let  $A, B, X \in \mathbb{M}_n$ . If

$$M = \begin{pmatrix} A & X \\ X^* & B \end{pmatrix}$$

is PPT, then for the linear map  $\Phi: C \to C + \text{Tr}(C)I$ ,

$$s_j(\Phi(X)) \leq s_j(\Phi(A \sharp B)), \quad j = 1, \dots, n.$$

In this paper, we extend Theorem 1.1 and Theorem 1.3 to SPT matrices and show a stronger inequality than (2).

### 2 Main results

We start with some lemmas. The first three lemmas are quite standard in matrix analysis.

**Lemma 2.1** ([13, p. 63]) *If*  $H \in \mathbb{M}_n$ , *then* 

$$\sigma_j(\operatorname{Re} H) \le s_j(H), \quad j = 1, \dots, n. \tag{3}$$

**Lemma 2.2** ([11, Theorem 1]) Let  $H = \begin{pmatrix} A & X \\ X^* & B \end{pmatrix} \in \mathbb{M}_{m+n}$  be positive semidefinite with  $A \in \mathbb{M}_m, B \in \mathbb{M}_n$ . Then

$$2s_j(X) \le s_j(H), \quad j = 1, \dots, \min\{m, n\}.$$
 (4)

**Lemma 2.3** ([2, p.106]) Let  $A, B \in \mathbb{M}_n$  be positive definite matrices. Then, for all  $X \in M_n$ ,

$$X^*(A \sharp B)X \le (X^*AX) \sharp (X^*BX).$$
<sup>(5)</sup>

The next lemma is due to Zhang [14, Lemma 3.1].

**Lemma 2.4** Let  $A \in \mathbb{M}_n$  have  $W(A) \subseteq S_\alpha$  for some  $\alpha \in [0, \frac{\pi}{2})$ . Then

$$\|A\| \le \sec(\alpha) \|\operatorname{Re}A\| \tag{6}$$

for any unitarily invariant norm.

The following result about geometric mean has been proved by Lin and Sun [9].

**Lemma 2.5** Let  $A, B \in \mathbb{M}_n$  be matrices with positive semidefinite real part. Then

$$(\operatorname{Re} A)\sharp(\operatorname{Re} B) \le \operatorname{Re}(A\sharp B). \tag{7}$$

Now we are ready to present our results. The first theorem is an extension of Theorem 1.1.

**Theorem 2.6** Let  $H_{11}, H_{12}, H_{21}, H_{22} \in \mathbb{M}_n$ . If  $H = \begin{pmatrix} H_{11} & H_{12} \\ H_{21} & H_{22} \end{pmatrix}$  is SPT, then

 $||H|| \le \sec(\alpha) ||H_{11} + H_{22}||$ 

for any unitarily invariant norm.

*Proof* Since *H* is a sector partial transpose matrix, then we know that

$$\operatorname{Re} H = \begin{pmatrix} \frac{H_{11} + H_{11}^*}{2} & \frac{H_{12} + H_{21}^*}{2} \\ \frac{H_{21} + H_{12}^*}{2} & \frac{H_{22} + H_{22}^*}{2} \end{pmatrix}$$

is PPT.

So by (6) we have

$$\|H\| \le \sec(\alpha) \|\operatorname{Re} H\|$$
  
$$\le \sec(\alpha) \|\operatorname{Re} H_{11} + \operatorname{Re} H_{22}\| \quad (by (1))$$
  
$$\le \sec(\alpha) \|H_{11} + H_{22}\|.$$

*Remark* 2.7 When  $H_{12} = H_{21}^*$  and  $\alpha = 0$ , then *H* is PPT in Theorem 2.6. Thus, our result is Hiroshima's inequality (1).

Next will give a stronger inequality than Theorem 1.2.

Theorem 2.8 Let the Hua matrix be given by

$$H := \begin{pmatrix} (I - A^*A)^{-1} & (I - B^*A)^{-1} \\ (I - A^*B)^{-1} & (I - B^*B)^{-1} \end{pmatrix},$$

where  $A, B \in \mathbb{M}_{m \times n}$  are strictly contractive. Then

$$\|(I - A^*B)^{-1}\| \le \|(I - A^*A)^{-1} \sharp (I - B^*B)^{-1}\|$$

for any unitarily invariant norm.

*Proof* Since *H* is PPT, then

$$\begin{pmatrix} (I-A^*A)^{-1} & (I-B^*A)^{-1} \\ (I-A^*B)^{-1} & (I-B^*B)^{-1} \end{pmatrix}, \qquad \begin{pmatrix} (I-A^*A)^{-1} & (I-A^*B)^{-1} \\ (I-B^*A)^{-1} & (I-B^*B)^{-1} \end{pmatrix}$$

are both positive semidefinite matrices.

Hence,

$$(I - B^*B)^{-1} \ge (I - A^*B)^{-1}(I - A^*A)(I - B^*A)^{-1}$$

and

$$(I - B^*B)^{-1} \ge (I - B^*A)^{-1} (I - A^*A) (I - A^*B)^{-1}.$$
(8)

Clearly, by unitary similarity transformation,

$$\begin{pmatrix} (I - B^*B)^{-1} & (I - A^*B)^{-1} \\ (I - B^*A)^{-1} & (I - A^*A)^{-1} \end{pmatrix}$$

is also positive semidefinite.

Therefore,

$$(I - A^*A)^{-1} \ge (I - B^*A)^{-1} (I - B^*B) (I - A^*B)^{-1}.$$
(9)

Thus,

$$(I - B^*B)^{-1} \sharp (I - A^*A)^{-1}$$
  
-  $(I - B^*A)^{-1} ((I - A^*A)^{-1} \sharp (I - B^*B)^{-1})^{-1} (I - A^*B)^{-1}$   
 $\geq (I - B^*B)^{-1} \sharp (I - A^*A)^{-1}$   
-  $((I - B^*A)^{-1} (I - A^*A) (I - A^*B)^{-1}) \sharp ((I - B^*A)^{-1} (I - B^*B) (I - A^*B)^{-1})$ 

(by (5) and monotonicity)

$$\geq (I - B^*B)^{-1} \sharp (I - A^*A)^{-1} - (I - B^*B)^{-1} \sharp (I - A^*A)^{-1} \quad (by (8) \text{ and } (9))$$
  
= 0.

In a similar way, we can prove

$$(I - B^*B)^{-1} \sharp (I - A^*A)^{-1} - (I - A^*B)^{-1} ((I - A^*A)^{-1} \sharp (I - B^*B)^{-1})^{-1} (I - B^*A)^{-1} \ge 0.$$

So

$$K := \begin{pmatrix} (I - A^*A)^{-1} \sharp (I - B^*B)^{-1} & (I - A^*B)^{-1} \\ (I - B^*A)^{-1} & (I - B^*B)^{-1} \sharp (I - A^*A)^{-1} \end{pmatrix}$$

is PPT.

Therefore,

$$2\|(I - A^*B)^{-1}\| \le \|K\| \quad (by (4))$$
  
$$\le \|((I - A^*A)^{-1}\sharp(I - B^*B)^{-1}) + ((I - B^*B)^{-1}\sharp(I - A^*A)^{-1})\|$$
  
$$(by (1))$$
  
$$= 2\|(I - B^*B)^{-1}\sharp(I - A^*A)^{-1}\|.$$

*Remark* 2.9 Obviously, our result is sharper than (2).

Finally, we present an extension of Theorem 1.3.

**Theorem 2.10** Let  $A, B, X, Y \in \mathbb{M}_n$ . If  $M = \begin{pmatrix} A & X \\ Y^* & B \end{pmatrix}$  is SPT, then

$$s_j\left(\Phi\left(\frac{X+Y}{2}\right)\right) \le s_j\left(\Phi(A \sharp B)\right),$$
 (10)

where  $\Phi: C \to C + \operatorname{Tr}(C)I$ .

*Proof* Since *M* is SPT, then

$$\operatorname{Re} M = \begin{pmatrix} \operatorname{Re} A & (X+Y)/2\\ (X+Y)^*/2 & \operatorname{Re} B \end{pmatrix}$$

and

$$\operatorname{Re}(M^{\tau}) = \begin{pmatrix} \operatorname{Re}A & (X+Y)^*/2\\ (X+Y)/2 & \operatorname{Re}B \end{pmatrix} = (\operatorname{Re}M)^{\tau}$$

are both positive semidefinite matrices. Thus,  $\operatorname{Re} M$  is PPT.

By Theorem 1.3, we have

$$s_j\left(\Phi\left(\frac{X+Y}{2}\right)\right) \leq s_j\left(\Phi\left((\operatorname{Re} A)\sharp(\operatorname{Re} B)\right)\right).$$

Compute

$$s_{j}\left(\Phi\left(\frac{X+Y}{2}\right)\right) \leq s_{j}\left(\Phi\left(\operatorname{Re} A \sharp \operatorname{Re} B\right)\right)$$
$$\leq s_{j}\left(\Phi\left(\operatorname{Re}(A \sharp B)\right)\right) \quad (by (7))$$
$$= s_{j}\left(\operatorname{Re}\left(\Phi(A \sharp B)\right)\right)$$
$$\leq s_{j}\left(\Phi(A \sharp B)\right) \quad (by (3)).$$

*Remark* 2.11 If *M* is PPT, then (10) becomes Lin's result in Theorem 1.3.

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#### **Competing interests**

The authors declare that they have no competing interests.

#### Authors' contributions

All authors contributed almost the same amount of work to the manuscript. All authors read and approved the final manuscript.

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#### References

- 1. Ando, T.: Positivity of operator-matrices of Hua-type. Banach J. Math. Anal. 2, 1–8 (2008)
- 2. Bhatia, R.: Positive Definite Matrices, Princeton University Press, Princeton (2007)
- 3. Drury, S.W., Lin, M.: Singular values inequalities for matrices with numerical ranges in a sector. Oper. Matrices 4, 1143–1148 (2014)
- 4. Hiroshima, T.: Majorization criterion for distillability of a bipartite quantum state. Phys. Rev. Lett. 91(5), 057902 (2003)
- 5. Horn, R.A., Johnson, C.R.: Topics in Matrix Analysis. Cambridge University Press, New York (1991)
- 6. Kuai, L.: An extension of the Fiedler–Markham determinant inequality. Linear Multilinear Algebra 66, 547–553 (2018)
- 7. Lin, M.: Inequalities related to 2 × 2 block PPT matrices. Oper. Matrices 94, 917–924 (2015)
- 8. Lin, M.: Some inequalities for sector matrices. Oper. Matrices 10, 915–921 (2016)
- 9. Lin, M., Sun, F.: A property of the geometric mean of accretive operator. Linear Multilinear Algebra 65, 433–437 (2017)
- 10. Lin, M., Wolkowicz, H.: Hiroshima's theorem and matrix norm inequality. Acta Sci. Math. 81, 45–53 (2015)
- 11. Tao, Y.: More results on singular value inequalities of matrices. Linear Algebra Appl. 416, 724–729 (2006)
- 12. Xu, G., Xu, C., Zhang, F.: Contractive matrices of Hua type. Linear Multilinear Algebra 59, 159–172 (2011)
- 13. Zhan, X.: Matrix Theory. American Mathematical Society, Providence (2013)
- 14. Zhang, F.: A matrix decomposition and its applications. Linear Multilinear Algebra 63, 2033–2042 (2015)