## INEQUALITIES ON $2 \times 2$ BLOCK ACCRETIVE MATRICES

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(Communicated by F. Kittaneh)

Abstract. A 2×2 block matrix  $\begin{pmatrix} A & X \\ Y^* & B \end{pmatrix}$  is accretive partial transpose (APT) if both  $\begin{pmatrix} A & X \\ Y^* & B \end{pmatrix}$  and  $\begin{pmatrix} A & Y^* \\ X & B \end{pmatrix}$  are accretive. This article presents some inequalities related to this class of matrices. One of our results refines a recent inequality in [Oper. Matrices, 15 (2021) 581–587].

## 1. Introduction

Let  $\mathbb{M}_n$  be the set of all  $n \times n$  complex matrices. If  $A \in \mathbb{M}_n$  is positive semidefinite (definite), then we write  $A \ge 0$  (A > 0). For two Hermitian matrices A, B of the same size,  $A \ge B$  (A > B) means that  $A - B \ge 0$  (A - B > 0). We say that  $A \in \mathbb{M}_n$  is accretive if its real part  $\operatorname{Re} A := \frac{A + A^*}{2}$  is positive definite, where  $A^*$  means the conjugate transpose of A. It is known that for every  $A \ge 0$ , there exists a unique  $B \ge 0$  such that  $B^2 = A$  [5, Theorem 7.2.6] and we denote  $A^{1/2} = B$ . If all eigenvalues of A are real, then they are arranged nonincreasingly  $\lambda_1(A) \ge \ldots \ge \lambda_n(A)$ ; the singular values of A are the eigenvalues of |A|, where  $|A| = (A^*A)^{\frac{1}{2}}$ , i.e.,  $s_j(A) = \lambda_j(|A|)$ ,  $j = 1, \ldots, n$ . The geometric mean of two positive definite matrices  $A, C \in \mathbb{M}_n$  is defined by

$$A \sharp C := A^{\frac{1}{2}} \left( A^{-\frac{1}{2}} C A^{-\frac{1}{2}} \right)^{\frac{1}{2}} A^{\frac{1}{2}}.$$
 (1)

It is known that the notion of geometric mean could be extended to cover all positive semidefinite matrices; see [2, p. 107]. Recently, Drury [3] defined the geometric mean of two accretive matrices via the following formula

$$A \sharp C = \left(\frac{2}{\pi} \int_0^\infty (tA + t^{-1}C)^{-1} \frac{dt}{t}\right)^{-1},$$

and proved the relationship (1) is also valid for two accretive matrices  $A, C \in \mathbb{M}_n$ . The readers can consult [3] for more properties.

Mathematics subject classification (2020): 15A45, 15A42, 47A30.

Keywords and phrases: Geometric mean, positive semidefinite matrices, singular value inequalities.



For the  $2 \times 2$  block matrix

$$M = \begin{pmatrix} A & B \\ B^* & C \end{pmatrix} \in \mathbb{M}_{2n}$$

with each block in  $\mathbb{M}_n$ , its partial transpose is defined by

$$M^{\tau} := \begin{pmatrix} A & B^* \\ B & C \end{pmatrix}.$$

A matrix M is called partial positive transpose (PPT) if M and  $M^{\tau}$  are positive semidefinite. We extend the notion to accretive matrices. If

$$M = \begin{pmatrix} A & X \\ Y^* & C \end{pmatrix} \in \mathbb{M}_{2n}$$

and

$$M^{\tau} := \begin{pmatrix} A & Y^* \\ X & C \end{pmatrix}$$

are both accretive, then we say that M is APT (i.e., accretive partial transpose). Clearly, the class of APT matrices includes the class of PPT matrices. Lee [6] obtained a matrix inequality involving the off-diagonal block of a PPT matrix and the geometric mean of its diagonal blocks.

THEOREM 1.1. [6, Theorem 2.1] Let  $\begin{pmatrix} A & B \\ B^* & C \end{pmatrix} \in \mathbb{M}_{2n}$  be PPT. Then, for some unitary matrix  $V \in \mathbb{M}_n$ ,

$$|B| \leqslant \frac{A \sharp C + V^* (A \sharp C) V}{2}$$

Recently, Fu et al.[4] presented a stronger result.

THEOREM 1.2. [4, Theorem 2.3] Let  $\begin{pmatrix} A & B \\ B^* & C \end{pmatrix} \in \mathbb{M}_{2n}$  be PPT. Then

$$|B| \leqslant (A \sharp C) \sharp (V^* (A \sharp C) V), \ |B^*| \leqslant (A \sharp C) \sharp (V (A \sharp C) V^*),$$

where  $V \in \mathbb{M}_n$  is any unitary matrix such that B = V|B|.

When  $\begin{pmatrix} A & B \\ B^* & C \end{pmatrix} \in \mathbb{M}_{2n}$  is positive semidefinite, Fu et al.[4, Theorem 2.2] also obtained that

$$|B| \leqslant (V^*AV) \sharp C, \ |B| \leqslant A \sharp (VCV^*).$$

$$\tag{2}$$

Liu et al. [8] extended Theorem 1.1 to the case of APT matrices.

THEOREM 1.3. [8, Theorem 3.4] Let  $\begin{pmatrix} A & X \\ Y^* & C \end{pmatrix} \in \mathbb{M}_{2n}$  be APT. Then, for some unitary matrix  $V \in \mathbb{M}_n$ ,

$$|X+Y| \leq \operatorname{Re}\left(A \sharp C + V^*(A \sharp C)V\right).$$

The main objective of this paper is to offer a refined result of Theorem 1.3.

In Section 2, we first present an inequality on  $2 \times 2$  block accretive matrices. It will then be applied to obtain a refinement of Theorem 1.3. As a consequence, a singular values inequality is given. At last, we will give an alternative proof of the inequality  $A \# A^* \ge \text{Re}A$  when  $A \in \mathbb{M}_n$  is an accretive matrix.

## 2. Main results

We first summarize some properties of the geometric mean of positive semidefinite matrices; see [2, Chapter 4].

PROPOSITION 2.1. Let  $A, C \ge 0$ . Then

- (i)  $A \not = A^{1/2} U C^{1/2}$  for some unitary matrix U.
- (*ii*)  $(A \not = C)^{-1} = A^{-1} \not = C^{-1}$  when A, C > 0.
- (iii)  $X^*AX \ddagger X^*CX \ge X^*(A \ddagger C)X$  with equality holds if X is nonsingular.

(iv) 
$$A \sharp C = \max \left\{ X : X = X^*, \ \begin{pmatrix} A & X \\ X & C \end{pmatrix} \ge 0 \right\}.$$

For a general  $2 \times 2$  block accretive matrix, we give the following two inequalities on its off-diagonal block and the geometric mean of its diagonal blocks.

THEOREM 2.2. Let 
$$\begin{pmatrix} A & X \\ Y^* & C \end{pmatrix} \in \mathbb{M}_{2n}$$
 be accretive. Then  
 $\left| \frac{X+Y}{2} \right| \leq (U^*(\operatorname{Re} A)U) \sharp \operatorname{Re} C$  and  $\left| \frac{X^*+Y^*}{2} \right| \leq \operatorname{Re} A \sharp (U(\operatorname{Re} C)U^*),$ 

where  $U \in \mathbb{M}_n$  is any unitary matrix such that  $\frac{X+Y}{2} = U \left| \frac{X+Y}{2} \right|$ .

*Proof.* Since 
$$\begin{pmatrix} A & X \\ Y^* & C \end{pmatrix}$$
 is accretive,  $\operatorname{Re}\begin{pmatrix} A & X \\ Y^* & C \end{pmatrix} = \begin{pmatrix} \operatorname{Re}A & \frac{X+Y}{2} \\ \frac{X^*+Y^*}{2} & \operatorname{Re}C \end{pmatrix}$  is posi-

tive definite. Hence by (2), we have

$$\left|\frac{X+Y}{2}\right| \leqslant (U^*(\operatorname{Re} A)U) \sharp \operatorname{Re} C,$$

and

$$\left|\frac{X^* + Y^*}{2}\right| \leqslant \operatorname{Re} A \sharp (U(\operatorname{Re} C)U^*). \quad \Box$$

It is clear that U in Theorem 2.2 is the unitary matrix in the polar decomposition of  $\frac{X+Y}{2}$ .

Theorem 2.2 leads us to an improvement of Theorem 1.3.

THEOREM 2.3. Let 
$$\begin{pmatrix} A & X \\ Y^* & C \end{pmatrix} \in \mathbb{M}_{2n}$$
 be APT. Then  
 $\left| \frac{X+Y}{2} \right| \leq (\operatorname{Re}A \sharp \operatorname{Re} C) \sharp (U^*(\operatorname{Re}A \sharp \operatorname{Re} C)U),$ 

and

$$\left|\frac{X^* + Y^*}{2}\right| \leq (\operatorname{Re}A \sharp \operatorname{Re}C) \sharp (U(\operatorname{Re}A \sharp \operatorname{Re}C)U^*),$$

where  $U \in \mathbb{M}_n$  is any unitary matrix such that  $\frac{X+Y}{2} = U \left| \frac{X+Y}{2} \right|$ .

*Proof.* Since 
$$\begin{pmatrix} A & X \\ Y^* & C \end{pmatrix}$$
 and  $\begin{pmatrix} A & Y^* \\ X & C \end{pmatrix}$  are accretive,  
 $\operatorname{Re}\begin{pmatrix} A & X \\ Y^* & C \end{pmatrix} = \begin{pmatrix} \operatorname{Re}A & \frac{X+Y}{2} \\ \frac{X^*+Y^*}{2} & \operatorname{Re}C \end{pmatrix}$  and  $\operatorname{Re}\begin{pmatrix} A & Y^* \\ X & C \end{pmatrix} = \begin{pmatrix} \operatorname{Re}A & \frac{X^*+Y^*}{2} \\ \frac{X+Y}{2} & \operatorname{Re}C \end{pmatrix}$ 

are positive definite. This means that  $\operatorname{Re}\begin{pmatrix} A & X \\ Y^* & C \end{pmatrix}$  is PPT.

By Theorem 1.2, we have

$$\left|\frac{X+Y}{2}\right| \leqslant (\operatorname{Re} A \sharp \operatorname{Re} C) \sharp (U^* (\operatorname{Re} A \sharp \operatorname{Re} C) U),$$

and

$$\left|\frac{X^* + Y^*}{2}\right| \leqslant (\operatorname{Re}A \sharp \operatorname{Re}C) \sharp (U(\operatorname{Re}A \sharp \operatorname{Re}C)U^*). \quad \Box$$

REMARK 1. It is apparent that if  $\begin{pmatrix} A & X \\ Y^* & C \end{pmatrix}$  is PPT (i.e., X = Y), Theorem 2.3 becomes Theorem 1.2.

COROLLARY 2.4. Let 
$$\begin{pmatrix} A & X \\ Y^* & C \end{pmatrix} \in \mathbb{M}_{2n}$$
 be APT. Then  
$$\prod_{j=1}^k s_j \left( \frac{X+Y}{2} \right) \leqslant \prod_{j=1}^k s_j (A \sharp C), \ k = 1, \dots, n$$

*Proof.* By Theorem 2.3 and Proposition 2.1 (i), it is easy to obtain that

$$\begin{split} \prod_{j=1}^k s_j \left( \frac{X+Y}{2} \right) &\leqslant \prod_{j=1}^k s_j ((\operatorname{Re}A \sharp \operatorname{Re}C) \sharp (U^* (\operatorname{Re}A \sharp \operatorname{Re}C) U)) \\ &\leqslant \prod_{j=1}^k s_j ((\operatorname{Re}A \sharp \operatorname{Re}C)^{\frac{1}{2}} W (U^* (\operatorname{Re}A \sharp \operatorname{Re}C) U)^{\frac{1}{2}}), \end{split}$$

where W is any unitary matrix such that

 $(\operatorname{Re}A\sharp\operatorname{Re}C)\sharp(U^*(\operatorname{Re}A\sharp\operatorname{Re}C)U) = (\operatorname{Re}A\sharp\operatorname{Re}C)^{\frac{1}{2}}W(U^*(\operatorname{Re}A\sharp\operatorname{Re}C)U)^{\frac{1}{2}}.$ 

Applying Horn inequality [9, p. 80] here, we have

$$\prod_{j=1}^{k} s_j\left(\frac{X+Y}{2}\right) \leqslant \prod_{j=1}^{k} s_j((\operatorname{Re}A \sharp \operatorname{Re}C)^{\frac{1}{2}}) s_j((\operatorname{Re}A \sharp \operatorname{Re}C)^{\frac{1}{2}})$$
$$= \prod_{j=1}^{k} s_j((\operatorname{Re}A \sharp \operatorname{Re}C)).$$

The result follows from inequality  $\operatorname{Re} A \sharp \operatorname{Re} C \leq \operatorname{Re} (A \sharp C)$  [7, Theorem 1.1] and the Fan-Hoffman inequality [1, p. 73].  $\Box$ 

Note that Corollary 2.4 is first given by Liu et al. [8, Theorem 2.1]. Next, we give an alternative proof of the inequality due to Liu et al. [8].

THEOREM 2.5. [8, Proposition 4.1] If  $A \in \mathbb{M}_n$  is accretive, then  $A \not \models A^* \ge \operatorname{Re} A$ .

*Proof.* It is clear that  $A \# A^*$  is Hermitian and accretive. Thus,  $A \# A^*$  is positive definite.

Using Proposition 2.1 (ii) and (iii),

$$A \sharp A^* - A^* (A \sharp A^*)^{-1} A = A \sharp A^* - (A^* A^{-1} A) \sharp (A^* (A^*)^{-1} A) = 0.$$

So  $M = \begin{pmatrix} A \# A^* & A \\ A^* & A \# A^* \end{pmatrix}$  is positive semidefinite. Similarly,  $M^{\tau} = \begin{pmatrix} A \# A^* & A^* \\ A & A \# A^* \end{pmatrix}$  is also positive semidefinite. This means that M is PPT. Therefore,

$$\frac{M+M^{\tau}}{2} = \begin{pmatrix} A \sharp A^* & \operatorname{Re} A \\ \operatorname{Re} A & A \sharp A^* \end{pmatrix} \ge 0.$$

By Proposition 2.1 (iv),  $\operatorname{Re} A \leq A \sharp A^*$ .  $\Box$ 

REMARK 2. We give a concise proof of Theorem 2.5 without using inner product, which is different from that in [8]. It is more concise.

*Acknowledgement.* The work of Junjian Yang is supported by the Key Laboratory of Computational Science and Application of Hainan Province, Hainan Provincial Natural Science Foundation of China (grant no. 120MS032), the National Natural Science Foundation (grant no. 12161020), the Ministry of Education of Hainan (grant no. Hnky2019ZD-13) and Hainan Provincial Natural Science Foundation for High-level Talents (grant no. 2019RC171).

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(Received September 22, 2021)

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