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

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#### Abstract

A $2 \times 2$ block matrix $\left(\begin{array}{cc}A & X \\ Y^{*} & B\end{array}\right)$ is accretive partial transpose (APT) if both $\left(\begin{array}{cc}A & X \\ Y^{*} & B\end{array}\right)$ and $\left(\begin{array}{ll}A & Y^{*} \\ X & B\end{array}\right)$ 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 $A \geqslant 0(A>0)$. For two Hermitian matrices $A, B$ of the same size, $A \geqslant B(A>B)$ means that $A-B \geqslant 0(A-B>0)$. 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 $A \geqslant 0$, there exists a unique $B \geqslant 0$ 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 $\lambda_{1}(A) \geqslant \ldots \geqslant \lambda_{n}(A)$; the singular values of $A \in \mathbb{M}_{n}$, denoted by $s_{j}(X)$, are similarly arranged. Note that the singular values of $A$ are the eigenvalues of $|A|$, where $|A|=\left(A^{*} A\right)^{\frac{1}{2}}$, 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

$$
\begin{equation*}
A \sharp C:=A^{\frac{1}{2}}\left(A^{-\frac{1}{2}} C A^{-\frac{1}{2}}\right)^{\frac{1}{2}} A^{\frac{1}{2}} . \tag{1}
\end{equation*}
$$

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

$$
A \sharp C=\left(\frac{2}{\pi} \int_{0}^{\infty}\left(t A+t^{-1} C\right)^{-1} \frac{d t}{t}\right)^{-1},
$$

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.

[^0]For the $2 \times 2$ block matrix

$$
M=\left(\begin{array}{cc}
A & B \\
B^{*} & C
\end{array}\right) \in \mathbb{M}_{2 n}
$$

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

$$
M^{\tau}:=\left(\begin{array}{ll}
A & B^{*} \\
B & C
\end{array}\right)
$$

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

$$
M=\left(\begin{array}{cc}
A & X \\
Y^{*} & C
\end{array}\right) \in \mathbb{M}_{2 n}
$$

and

$$
M^{\tau}:=\left(\begin{array}{ll}
A & Y^{*} \\
X & C
\end{array}\right)
$$

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 $\left(\begin{array}{cc}A & B \\ B^{*} & C\end{array}\right) \in \mathbb{M}_{2 n}$ 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 $\left(\begin{array}{cc}A & B \\ B^{*} & C\end{array}\right) \in \mathbb{M}_{2 n}$ be PPT. Then

$$
|B| \leqslant(A \sharp C) \sharp\left(V^{*}(A \sharp C) V\right),\left|B^{*}\right| \leqslant(A \sharp C) \sharp\left(V(A \sharp C) V^{*}\right),
$$

where $V \in \mathbb{M}_{n}$ is any unitary matrix such that $B=V|B|$.
When $\left(\begin{array}{cc}A & B \\ B^{*} & C\end{array}\right) \in \mathbb{M}_{2 n}$ is positive semidefinite, Fu et al.[4, Theorem 2.2] also obtained that

$$
\begin{equation*}
|B| \leqslant\left(V^{*} A V\right) \sharp C,|B| \leqslant A \sharp\left(V C V^{*}\right) . \tag{2}
\end{equation*}
$$

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

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

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

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 $A \sharp A^{*} \geqslant \operatorname{Re} A$ 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 $A, C \geqslant 0$. Then
(i) $A \sharp C=A^{1 / 2} U C^{1 / 2}$ for some unitary matrix $U$.
(ii) $(A \sharp C)^{-1}=A^{-1} \sharp C^{-1}$ when $A, C>0$.
(iii) $X^{*} A X \sharp X^{*} C X \geqslant X^{*}(A \sharp C) X$ with equality holds if $X$ is nonsingular.
(iv) $A \sharp C=\max \left\{X: X=X^{*},\left(\begin{array}{ll}A & X \\ X & C\end{array}\right) \geqslant 0\right\}$.

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 $\left(\begin{array}{cc}A & X \\ Y^{*} & C\end{array}\right) \in \mathbb{M}_{2 n}$ be accretive. Then

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

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 $\left(\begin{array}{cc}A & X \\ Y^{*} & C\end{array}\right)$ is accretive, $\operatorname{Re}\left(\begin{array}{cc}A & X \\ Y^{*} & C\end{array}\right)=\left(\begin{array}{cc}\operatorname{Re} A & \frac{X+Y}{2} \\ \frac{X^{*}+Y^{*}}{2} & \operatorname{Re} C\end{array}\right)$ is positive definite. Hence by (2), we have

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

and

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

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 $\left(\begin{array}{cc}A & X \\ Y^{*} & C\end{array}\right) \in \mathbb{M}_{2 n}$ be APT. Then

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

and

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

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 $\left(\begin{array}{cc}A & X \\ Y^{*} & C\end{array}\right)$ and $\left(\begin{array}{cc}A & Y^{*} \\ X & C\end{array}\right)$ are accretive,

$$
\operatorname{Re}\left(\begin{array}{cc}
A & X \\
Y^{*} & C
\end{array}\right)=\left(\begin{array}{cc}
\operatorname{Re} A & \frac{X+Y}{2} \\
\frac{X^{*}+Y^{*}}{2} & \operatorname{Re} C
\end{array}\right) \quad \text { and } \quad \operatorname{Re}\left(\begin{array}{ll}
A & Y^{*} \\
X & C
\end{array}\right)=\left(\begin{array}{cc}
\operatorname{Re} A & \frac{X^{*}+Y^{*}}{2} \\
\frac{X+Y}{2} & \operatorname{Re} C
\end{array}\right)
$$

are positive definite. This means that $\operatorname{Re}\left(\begin{array}{cc}A & X \\ Y^{*} & C\end{array}\right)$ is PPT.
By Theorem 1.2, we have

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

and

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

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

Corollary 2.4. Let $\left(\begin{array}{cc}A & X \\ Y^{*} & C\end{array}\right) \in \mathbb{M}_{2 n}$ be APT. Then

$$
\prod_{j=1}^{k} s_{j}\left(\frac{X+Y}{2}\right) \leqslant \prod_{j=1}^{k} s_{j}(A \sharp C), \quad k=1, \ldots, n .
$$

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

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

where $W$ is any unitary matrix such that

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

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

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

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

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 $A \sharp A^{*} \geqslant \operatorname{Re} A$.

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

Using Proposition 2.1 (ii) and (iii),

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

So $M=\left(\begin{array}{cc}A \sharp A^{*} & A \\ A^{*} & A \sharp A^{*}\end{array}\right)$ is positive semidefinite. Similarly, $M^{\tau}=\left(\begin{array}{cc}A \sharp A^{*} & A^{*} \\ A & A \sharp A^{*}\end{array}\right)$ is also positive semidefinite. This means that $M$ is PPT. Therefore,

$$
\frac{M+M^{\tau}}{2}=\left(\begin{array}{cc}
A \sharp A^{*} & \operatorname{Re} A \\
\operatorname{Re} A & A \sharp A^{*}
\end{array}\right) \geqslant 0 .
$$

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

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.

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