

## Research Article

# On a Matrix Trace Inequality for Hermitian Matrices

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**Abstract:** Da-Wei Chang, obtained the matrix trace inequality for Hermitian matrices  $\text{tr}((AB)^{2k}) \leq \text{tr}(A^{2k}B^{2k})$  for any integer  $k$ . In this paper, we give an equality condition for this inequality by using the weak majorization of eigenvalue.

**Keywords:** trace function, trace inequality, the weak majorization

**MSC:** 14A18

## 1. Introduction

Let  $M_n(\mathbb{C})$  be the set of all  $n \times n$  matrices over the complex number field  $\mathbb{C}$ . The eigenvalues of  $A \in M_n(\mathbb{C})$  are  $\lambda_1(A), \lambda_2(A), \dots, \lambda_n(A)$ , with  $|\lambda_1(A)| \geq |\lambda_2(A)| \geq \dots \geq |\lambda_n(A)|$ . The singular values of  $A \in M_n(\mathbb{C})$ , denoted by  $\sigma_1(A), \sigma_2(A), \dots, \sigma_n(A)$ , are the eigenvalues of the matrix  $|A| = (AA^*)^{1/2}$  arranged in such a way that  $\sigma_1(A) \geq \sigma_2(A) \geq \dots \geq \sigma_n(A)$ . Note that  $\sigma_i^2(A) = \lambda_i(AA^*) = \lambda_i(A^*A)$ , so for a positive semidefinite matrix  $A$ , we have  $\sigma_i(A) = \lambda_i(A)$  for all  $i = 1, 2, \dots, n$ .

Given two real vectors  $x = (x_1, x_2, \dots, x_n)$  and  $y = (y_1, y_2, \dots, y_n)$  in decreasing order we say that  $x$  is weakly log majorized by  $y$  and we write  $\log x \prec_w \log y$  if  $\prod_{i=1}^k x_i \leq \prod_{i=1}^k y_i$ , for every  $k = 1, 2, \dots, n$  but if equality occurs at  $k = n$  we write it as  $\log x \prec \log y$ . We say that  $x$  is weakly majorized by  $y$  and we write  $x \prec_w y$  if  $\sum_{i=1}^k x_i \leq \sum_{i=1}^k y_i$  for every  $k = 1, 2, \dots, n$ . We say that  $x$  is majorized by  $y$  and we write  $x \prec y$  if

$$x \prec_w y \text{ and } \sum_{i=1}^n x_i = \sum_{i=1}^n y_i.$$

A normal matrix  $A$  can be decomposed as  $A = UDU^*$  using the spectral decomposition, where  $D = \text{diag}\{\lambda_1(A), \lambda_2(A), \dots, \lambda_n(A)\}$  and  $U$  is a unitary matrix.

It is well know that for all positive semidefinite matrices  $A$  and  $B$ ,

$$0 \leq \text{tr}(AB) \leq \text{tr}(A) \text{tr}(B).$$

Lieb and Thirring [1] showed that

$$\operatorname{tr}((AB)^\alpha) \leq \operatorname{tr}(A^\alpha B^\alpha) \quad (1)$$

for all positive operators  $A, B$  on a separable Hilbert space and any real number  $\alpha \geq 1$ . In the case where  $\alpha = m$  is a positive integer, some upper and lower bounds for the inequality (1) were obtained by Marcus [2], Le Couteur [3] and proved again by Bushell and Trustrum [4] as

$$\sum_{i=1}^n \lambda_i^m(A) \lambda_{n-i+1}^m(B) \leq \operatorname{tr}((AB)^m) \leq \operatorname{tr}(A^m B^m) \leq \sum_{i=1}^n \lambda_i^m(A) \lambda_i^m(B). \quad (2)$$

Da-Wei Chang [5] obtained the inequality

$$\operatorname{tr}((AB)^{2k}) \leq \operatorname{tr}(A^{2k} B^{2k}), \quad (3)$$

where  $k$  is a positive integer,  $A$  and  $B$  are hermitian. Wang and Gong [6] generalized the inequality (2) in terms of majorization and proved

$$\log \lambda^{1/\alpha}(A^\alpha B^\alpha) \prec \log \lambda^{1/\beta}(A^\beta B^\beta), \quad 0 < \alpha \leq \beta, \quad (4)$$

$$\lambda^{1/\alpha}(A^\alpha B^\alpha) \prec_\omega \lambda^{1/\beta}(A^\beta B^\beta), \quad 0 < \alpha \leq \beta, \quad (5)$$

$$\lambda^{1/\beta}(A^\beta B^\beta) \prec_\omega \lambda^{1/\alpha}(A^\alpha B^\alpha), \quad \alpha \leq \beta < 0, \quad (6)$$

$$\lambda^\alpha(AB) \prec_\omega \lambda(A^\alpha B^\alpha), \quad |\alpha| \geq 1, \quad (7)$$

$$\lambda(A^\alpha B^\alpha) \prec_\omega \lambda^\alpha(AB), \quad |\alpha| \leq 1. \quad (8)$$

In 1999 Chang [5] proved  $\operatorname{tr}((AB)^{2k}) \leq \operatorname{tr}(A^{2k} B^{2k})$  for Hermitian matrices  $A, B$ . In this paper we show that equality holds if and only if  $A, B$  commute by using the weak majorization of eigenvalue.

## 2. Main results

Throughout this section, we work with square matrices. First of all we have to show that for any Hermitian matrices  $A$  and  $B$ ,  $\operatorname{tr}((AB)^2) = \operatorname{tr}(A^2 B^2)$  if and only if  $AB = BA$ .

**Lemma 1** Let  $A \in M_n(\mathbb{R})$  be a diagonal matrix and  $B \in M_n(\mathbb{C})$  a Hermitian matrix. If  $\operatorname{tr}((AB)^2) = \operatorname{tr}(A^2 B^2)$ , then  $AB = BA$ .

**Proof.** Suppose that  $\operatorname{tr}((AB)^2) = \operatorname{tr}(A^2 B^2)$ , while

$$A = \begin{bmatrix} a_1 & 0 & \cdots & 0 \\ 0 & a_2 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & a_n \end{bmatrix} \quad \text{and} \quad B = \begin{bmatrix} b_{11} & b_{12} & \cdots & b_{1n} \\ \bar{b}_{12} & b_{22} & \cdots & b_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ \bar{b}_{1n} & \bar{b}_{2n} & \cdots & b_{nn} \end{bmatrix}$$

in which  $a_1, a_2, \dots, a_n \in \mathbb{R}$  and  $b_{11}, b_{22}, \dots, b_{nn} \in \mathbb{R}$ . Then we have

$$\begin{aligned} \operatorname{tr}((AB)^2) &= (a_1 b_{11} a_1 b_{11} + a_1 b_{12} a_2 \bar{b}_{12} + \cdots + a_1 b_{1n} a_n \bar{b}_{1n}) \\ &\quad + (a_2 \bar{b}_{12} a_1 b_{12} + a_2 b_{22} a_2 b_{22} + \cdots + a_2 b_{2n} a_n \bar{b}_{2n}) \\ &\quad + \cdots + (a_n \bar{b}_{1n} a_1 b_{1n} + a_n \bar{b}_{2n} a_2 b_{2n} + \cdots + a_n b_{nn} a_n b_{nn}) \\ &= \sum_{i=1}^n \sum_{j=1}^n a_i a_j |b_{ij}|^2 \end{aligned}$$

and

$$\begin{aligned} \operatorname{tr}(A^2 B^2) &= (a_1 a_1 b_{11} b_{11} + a_1 a_1 b_{12} \bar{b}_{12} + \cdots + a_1 a_1 b_{1n} \bar{b}_{1n}) \\ &\quad + (a_2 a_2 \bar{b}_{12} b_{12} + a_2 a_2 b_{22} b_{22} + \cdots + a_2 a_2 b_{2n} \bar{b}_{2n}) \\ &\quad + \cdots + (a_n a_n \bar{b}_{1n} b_{1n} + a_n a_n \bar{b}_{2n} b_{2n} + \cdots + a_n a_n b_{nn} b_{nn}) \\ &= \sum_{i=1}^n \sum_{j=1}^n a_i^2 |b_{ij}|^2. \end{aligned}$$

Thus

$$\sum_{i=1}^n \sum_{j=1}^n a_i a_j |b_{ij}|^2 = \sum_{i=1}^n \sum_{j=1}^n a_i^2 |b_{ij}|^2.$$

Now

$$AB - BA = \begin{bmatrix} 0 & (a_1 - a_2)b_{12} & \cdots & (a_1 - a_n)b_{1n} \\ (a_2 - a_1)\bar{b}_{12} & 0 & \cdots & (a_2 - a_n)b_{2n} \\ \vdots & \vdots & \ddots & \vdots \\ (a_n - a_1)\bar{b}_{1n} & (a_n - a_2)\bar{b}_{2n} & \cdots & 0 \end{bmatrix}$$

But since  $|b_{ij}| = |b_{ji}|$ ,

$$\begin{aligned} \sum_{i=1}^n \sum_{j=1}^n (a_i - a_j)^2 |b_{ij}|^2 &= \sum_{i=1}^n \sum_{j=1}^n a_i^2 |b_{ij}|^2 + \sum_{i=1}^n \sum_{j=1}^n a_j^2 |b_{ij}|^2 - 2 \sum_{i=1}^n \sum_{j=1}^n a_i a_j |b_{ij}|^2 \\ &= 2 \left\{ \sum_{i=1}^n \sum_{j=1}^n a_i^2 |b_{ij}|^2 - \sum_{i=1}^n \sum_{j=1}^n a_i a_j |b_{ij}|^2 \right\}. \end{aligned}$$

Thus  $AB - BA = 0$ . □

**Theorem 2** Let  $A$  and  $B$  be Hermitian matrices. Then  $AB = BA$  if and only if  $\text{tr}((AB)^2) = \text{tr}(A^2 B^2)$ .

**Proof.** If  $AB = BA$ , then it is easy to see that  $\text{tr}((AB)^2) = \text{tr}(A^2 B^2)$ . Next suppose that  $\text{tr}((AB)^2) = \text{tr}(A^2 B^2)$ . Since  $A$  is Hermitian,  $A = UDU^*$  where  $U$  is a unitary matrix and  $D$  is a diagonal matrix whose diagonal entries are eigenvalues of  $A$ . Thus

$$\begin{aligned} \text{tr}((AB)^2) &= \text{tr}[(UDU^*)B(UDU^*)B] = \text{tr}[D(U^*BU)D(U^*BU)] \\ &= \text{tr}([D(U^*BU)]^2) \end{aligned} \tag{9}$$

and

$$\begin{aligned} \text{tr}(A^2 B^2) &= \text{tr}[(UDU^*)(UDU^*)B^2] = \text{tr}[(UD^2U^*)B^2] \\ &= \text{tr}[D^2(U^*B^2U)]. \end{aligned} \tag{10}$$

From equality of (9) and (10), Lemma 1 implies that  $AB = BA$ . □

Next we will give an equality condition for the inequality (3) in case  $k \geq 1$ .

**Theorem 3** Let  $A$  and  $B$  be Hermitian matrices. Then for any natural number  $k \geq 1$ ,  $AB = BA$  if and only if  $\text{tr}((AB)^{2k}) = \text{tr}(A^{2k} B^{2k})$ .

**Proof.** If  $A$  and  $B$  commute the results is trivial. To prove sufficiency, we will show that if  $\text{tr}((AB)^{2k}) = \text{tr}(A^{2k} B^{2k})$ ,  $k \geq 1$  then  $\text{tr}((AB)^2) = \text{tr}(A^2 B^2)$  which implies  $AB = BA$ . Suppose that  $\text{tr}((AB)^2) \neq \text{tr}(A^2 B^2)$ , so from the inequality (3) we have that

$$\sum_{i=1}^n \lambda_i^2(AB) < \sum_{i=1}^n \lambda_i(A^2 B^2). \tag{11}$$

Thus

$$\text{tr}((AB)^{2k}) = \sum_{i=1}^n \lambda_i^{2k-2}((AB)) < \sum_{i=1}^n \lambda_i^{2k-2}(A^2 B^2). \tag{12}$$

Since  $A$  and  $B$  are Hermitian,  $A^2$  and  $B^2$  are positive semidefinite. Therefore by (7) we have that

$$\sum_{i=1}^n \lambda_i^{2k-1} (A^2 B^2) \leq \sum_{i=1}^n \lambda_i (A^{2k} B^{2k}). \quad (13)$$

We can therefore conclude that if  $\text{tr}((AB)^2) < \text{tr}(A^2 B^2)$  then  $\text{tr}((AB)^{2k}) < \text{tr}(A^{2k} B^{2k})$ . Hence by the inequality (3) we have, if  $\text{tr}((AB)^{2k}) = \text{tr}(A^{2k} B^{2k})$  then  $\text{tr}((AB)^2) = \text{tr}(A^2 B^2)$  which implies  $AB = BA$ .  $\square$

Next we will give an equality condition for the inequality (3) in case  $k < 0$ . First we will consider  $k = -1$ , after which any  $k \leq 1$  can be considered. To be able to find the square root of any matrix, we need to consider positive semidefinite Hermitian matrices.

**Lemma 4** Let  $A$  be a positive semidefinite diagonal matrix and  $B \in M_n(\mathbb{C})$  a positive semidefinite matrix. If  $\text{tr}((AB)^{1/2}) = \text{tr}(A^{1/2} B^{1/2})$ , then  $AB = BA$ .

**Proof.** Since  $B$  is positive semidefinite,  $B = UDU^*$  where  $U$  is a unitary matrix and  $D$  is a diagonal matrix whose diagonal entries are non-negative eigenvalues of  $B$ . Then

$$\text{tr}((AB)^{1/2}) = \text{tr}(A^{1/2} B^{1/2}) = \text{tr}(A' B'), \quad (14)$$

where  $A' = U^* A^{1/2} U$  and  $B' = D^{1/2}$ . We also have that

$$\begin{aligned} \text{tr}((A'^2 B'^2)^{1/2}) &= \sum_{i=1}^n \lambda_i^{1/2} (A'^2 B'^2) \\ &= \sum_{i=1}^n \lambda_i^{1/2} (U^* A U D) \\ &= \sum_{i=1}^n \lambda_i^{1/2} (A U D U^*) \\ &= \text{tr}((AB)^{1/2}). \end{aligned} \quad (15)$$

From equation (14) and (15) we have that  $\text{tr}(A' B') = \text{tr}((A'^2 B'^2)^{1/2})$ . But

$$\text{tr}((A'^2 B'^2)^{1/2}) = \sum_{i=1}^n \lambda_i^{1/2} (A'^2 B'^2) = \sum_{i=1}^n \lambda_i^{1/2} ((A' B') (A' B')^*) = \sum_{i=1}^n \sigma_i(A' B').$$

Thus  $\sum_{i=1}^n \lambda_i(A' B') = \text{tr}(A' B') = \text{tr}((A'^2 B'^2)^{1/2}) = \sum_{i=1}^n \sigma_i(A' B')$ .

We know that

$$\left| \sum_{i=1}^n \lambda_i(A' B') \right| \leq \sum_{i=1}^n \left| \lambda_i(A' B') \right| \leq \sum_{i=1}^n \lambda_i(|A' B'|) = \sum_{i=1}^n \sigma_i(A' B')$$

so  $\sum_{i=1}^n \lambda_i(A'B') = \sum_{i=1}^n |\lambda_i(A'B')| = \sum_{i=1}^n \sigma_i(A'B')$ .

Therefore, we can conclude that  $\lambda_i(A'B')$  is non-negative for all  $i = 1, 2, \dots, n$ . Since the arrangement of eigenvalues and singular values,  $\lambda_i(A'B') = \sigma_i(A'B')$  for all  $i = 1, 2, \dots, n$  which implies  $A'B'$  is a positive semidefinite matrix. Then we have,

$$\begin{aligned} \sum_{i=1}^n \lambda_i((A'B')^2) &= \sum_{i=1}^n \lambda_i((A'B')(A'B')^*) \\ &= \sum_{i=1}^n \lambda_i(A'B'B'A') \\ &= \sum_{i=1}^n \lambda_i(A^2B^2). \end{aligned}$$

Thus by Lemma 1,  $A'B' = B'A'$ . Hence  $A'$  and  $B'$  are simultaneously unitarily diagonalizable and commute. Therefore we can conclude that  $AB = BA$ .  $\square$

**Theorem 1** Let  $A$  and  $B$  be positive semidefinite matrices. Then  $AB = BA$  if and only if  $\text{tr}((AB)^{1/2}) = \text{tr}(A^{1/2}B^{1/2})$ .

**Proof.** Since  $A$  and  $B$  are positive semidefinite matrices,  $A$  and  $B$  are simultaneously unitary diagonalizable. If  $AB = BA$ , then  $(AB)^{1/2} = A^{1/2}B^{1/2}$ . Thus  $\text{tr}((AB)^{1/2}) = \text{tr}(A^{1/2}B^{1/2})$ . Next assume that  $\text{tr}((AB)^{1/2}) = \text{tr}(A^{1/2}B^{1/2})$ . Assume that  $A = UDU^*$  is the spectral decomposition of  $A$ . Thus

$$\text{tr}(A^{1/2}B^{1/2}) = \text{tr}(UD^{1/2}U^*B^{1/2}) = \text{tr}(D^{1/2}U^*B^{1/2}U). \quad (16)$$

By the unitary invariance of eigenvalues we have that

$$\text{tr}((AB)^{1/2}) = \text{tr}((UDU^*B)^{1/2}) = \text{tr}((DU^*B)^{1/2}). \quad (17)$$

By Lemma 2 and the equations (16) and (17), we have that  $DU^*BU = U^*BUD$ , which implies that  $AB = BA$ .  $\square$

**Theorem 6** Let  $A$  and  $B$  be positive semidefinite matrices and  $k$  a positive integer. Then  $AB = BA$  if and only if  $\text{tr}((AB)^{1/2^k}) = \text{tr}(A^{1/2^k}B^{1/2^k})$ .

**Proof.** Necessity is obvious because  $A$  and  $B$  are normal and commute, so they are simultaneously unitary diagonalizable. Next we will prove the sufficiency case. In [6] Wang and Gong proved the weak majorization  $\lambda(A^\alpha B^\alpha) \prec_w \lambda^\alpha(AB)$  for  $|\alpha| \leq 1$ , so we have that

$$\text{tr}(A^{1/2^k}B^{1/2^k}) \leq \text{tr}(AB)^{1/2^k}.$$

We will show that  $\text{tr}(A^{1/2^k}B^{1/2^k}) = \text{tr}(AB)^{1/2^k}$  implies  $\text{tr}(A^{1/2}B^{1/2}) = \text{tr}(AB)^{1/2}$ . If  $\text{tr}(A^{1/2}B^{1/2}) \neq \text{tr}(AB)^{1/2}$ , then  $\sum_{i=1}^n \lambda_i(A^{1/2}B^{1/2}) < \sum_{i=1}^n \lambda_i^{1/2}(AB)$ . We apply (8) and get

$$\begin{aligned}
\operatorname{tr}(A^{1/2^k} B^{1/2^k}) &= \sum_{i=1}^n \lambda_i(A^{1/2^k} B^{1/2^k}) \\
&\leq \sum_{i=1}^n \lambda_i^{1/2^k}(A^{1/2} B^{1/2}) \\
&< \sum_{i=1}^n \lambda_i^{1/2^k}(AB) \\
&= \operatorname{tr}(AB)^{1/2^k}
\end{aligned}$$

a contradiction. Thus  $\lambda(A^{1/2} B^{1/2}) = \lambda^{1/2}(AB)$ , by Theorem 3 we have that  $AB = BA$ . □

## Conflict of interest

Author declares there is no conflict of interest at any point with reference to research findings.

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