

NORM INEQUALITIES RELATED TO THE MATRIX GEOMETRIC MEAN OF NEGATIVE POWER

MOHSEN KIAN* AND YUKI SEO**

Received February 12, 2018; revised March 27, 2018

ABSTRACT. In this paper, we show norm inequalities related to the matrix geometric mean of negative power for positive definite matrices: For positive definite matrices A and B ,

$$\left\| e^{(1-\beta)\log A + \beta\log B} \right\| \leq \|A \sharp_{\beta} B\| \leq \|A^{1-\beta} B^{\beta}\|$$

for every unitarily invariant norm and $-1 \leq \beta \leq -\frac{1}{2}$, where the β -quasi geometric mean $A \sharp_{\beta} B$ is defined by $A \sharp_{\beta} B = A^{\frac{1}{2}}(A^{-\frac{1}{2}}BA^{-\frac{1}{2}})^{\beta}A^{\frac{1}{2}}$. For our purposes, we show the Ando-Hiai log-majorization of negative power.

1 Introduction. Let $\mathbb{M}_n = \mathbb{M}_n(\mathbb{C})$ be the algebra of $n \times n$ complex matrices and denote the matrix absolute value of any $A \in \mathbb{M}_n$ by $|A| = (A^*A)^{\frac{1}{2}}$. For $A \in \mathbb{M}_n$, we write $A \geq 0$ if A is positive semidefinite and $A > 0$ if A is positive definite, that is, A is positive and invertible. For two Hermitian matrices A and B , we write $A \geq B$ if $A - B \geq 0$, and it is called the Löwner ordering. A norm $\|\cdot\|$ on \mathbb{M}_n is said to be unitarily invariant if $\|UXV\| = \|X\|$ for all $X \in \mathbb{M}_n$ and unitary U, V .

Let A and B be two positive definite matrices. The arithmetic-geometric mean inequality says that

$$(1.1) \quad A \sharp_{\alpha} B \leq (1 - \alpha)A + \alpha B \quad \text{for all } \alpha \in [0, 1],$$

where the α -geometric mean $A \sharp_{\alpha} B$ is defined by

$$A \sharp_{\alpha} B = A^{\frac{1}{2}} \left(A^{-\frac{1}{2}} B A^{-\frac{1}{2}} \right)^{\alpha} A^{\frac{1}{2}} \quad \text{for all } \alpha \in [0, 1],$$

also see [11]. As another matrix geometric mean, we recall that the chaotic geometric mean $A \diamond_{\alpha} B$ is defined by

$$A \diamond_{\alpha} B = e^{(1-\alpha)\log A + \alpha\log B} \quad \text{for all } \alpha \in \mathbb{R},$$

also see [5, Section 3.5]. If A and B commute, then $A \diamond_{\alpha} B = A \sharp_{\alpha} B = A^{1-\alpha} B^{\alpha}$ for $\alpha \in [0, 1]$. In [4], Bhatia and Grover showed precise norm estimations of the arithmetic-geometric mean inequality (1.1) as follows: For each $\alpha \in [0, 1]$ and any unitarily invariant norm $\|\cdot\|$

$$\begin{aligned} \|A \sharp_{\alpha} B\| &\leq \|A \diamond_{\alpha} B\| \leq \|B^{\frac{\alpha}{2}} A^{1-\alpha} B^{\frac{\alpha}{2}}\| \\ &\leq \left\| \frac{1}{2} (A^{1-\alpha} B^{\alpha} + B^{\alpha} A^{1-\alpha}) \right\| \leq \|A^{1-\alpha} B^{\alpha}\| \leq \|(1 - \alpha)A + \alpha B\| \end{aligned}$$

2000 *Mathematics Subject Classification.* Primary 15A45; Secondary 47A64.

Key words and phrases. Ando-Hiai inequality, matrix geometric mean, unitarily invariant norm, positive definite matrix.

and

$$\begin{aligned} \|A \sharp_\alpha B\| &\leq \|A \diamond_\alpha B\| \\ &\leq \left\| \left(B^{\frac{\alpha p}{2}} A^{(1-\alpha)p} B^{\frac{\alpha p}{2}} \right)^{\frac{1}{p}} \right\| \leq \left\| ((1-\alpha)A^p + \alpha B^p)^{\frac{1}{p}} \right\| \quad \text{for all } p > 0. \end{aligned}$$

For convenience in symbolic expression, we define $A \natural_\beta B$ for $\beta \in [-1, 0)$ and positive definite matrices A, B as follows:

$$(1.2) \quad A \natural_\beta B = A^{\frac{1}{2}} (A^{-\frac{1}{2}} B A^{-\frac{1}{2}})^\beta A^{\frac{1}{2}} \quad \text{for all } \beta \in [-1, 0),$$

whose formula is the same as \sharp_α . Though $A \natural_\beta B$ for $\beta \in [-1, 0)$ are not matrix means in the sense of Kubo-Ando theory [11], it is known in [7] that $A \natural_\beta B$ have matrix mean like properties for any positive definite matrices A and B . Thus we call (1.2) the β -quasi geometric mean for $\beta \in [-1, 0)$. For more detail, see [7].

On the other hand, the following reverse arithmetic-geometric mean inequality holds:

$$(1-\beta)A + \beta B \leq A \natural_\beta B \quad \text{for all } \beta \in [-1, 0),$$

also see [8]. Though we have no relation among $A \natural_\beta B$, $A \diamond_\beta B$ and $A^{1-\beta} B^\beta$ for $\beta \in [-1, 0)$ under the Löwner ordering, it follows from a proof similar to Bhatia-Grover's one in [4] that for each $\beta \in \mathbb{R}$ and any unitarily invariant norm $\|\cdot\|$

$$(1.3) \quad \|A \diamond_\beta B\| \leq \left\| B^{\frac{\beta}{2}} A^{1-\beta} B^{\frac{\beta}{2}} \right\| \leq \left\| \frac{1}{2} (A^{1-\beta} B^\beta + B^\beta A^{1-\beta}) \right\| \leq \|A^{1-\beta} B^\beta\|.$$

Also, by the Lie-Trotter formula $\lim_{t \rightarrow 0} \left(e^{\frac{t}{2} B} e^{tA} e^{\frac{t}{2} B} \right)^{\frac{1}{t}} = e^{A+B}$ and the Araki-Cordes inequality $\|B^t A^t B^t\| \leq \|(BAB)^t\|$ for all $t \in [0, 1]$, also see [3, Exercise IX.1.5, Theorem IX.2.10], it follows that for each $\beta \in [-1, 0)$

$$(1.4) \quad \|A \diamond_\beta B\| \leq \left\| \left(B^{\frac{\beta q}{2}} A^{(1-\beta)q} B^{\frac{\beta q}{2}} \right)^{\frac{1}{q}} \right\|$$

holds for all $q > 0$ and $\left\| \left(B^{\frac{\beta q}{2}} A^{(1-\beta)q} B^{\frac{\beta q}{2}} \right)^{\frac{1}{q}} \right\|$ decreases to $\|A \diamond_\beta B\|$ as $q \downarrow 0$. It is natural to ask what is the estimate of the β -quasi geometric mean in the norm inequalities (1.3) and (1.4) for $\beta \in [-1, 0)$.

In this paper, we show norm inequalities related to the β -quasi geometric mean of negative power, the chaotic geometric mean $A \diamond_\beta B$ and $A^{1-\beta} B^\beta$ for positive definite matrices A, B . Moreover, we show precise norm estimations of the reverse arithmetic-geometric mean inequality under the assumption $A \geq B$. For our purposes, we need the Ando-Hiai log-majorization of negative power.

2 Preliminaries. In this section, we have some preliminary results on the log majorization of matrices. For Hermitian matrices H, K the weak majorization $H \prec_w K$ means that

$$\sum_{i=1}^k \lambda_i(H) \leq \sum_{i=1}^k \lambda_i(K) \quad \text{for } k = 1, 2, \dots, n,$$

where $\lambda_1(H) \geq \dots \geq \lambda_n(H)$ and $\lambda_1(K) \geq \dots \geq \lambda_n(K)$ are the eigenvalues of H and K respectively. Further, the majorization $H \prec K$ means that $H \prec_w K$ and the equality holds

for $k = n$ in the above, i.e., $\text{Tr}H = \text{Tr}K$. For $A, B \geq 0$ let us write $A \prec_{w(\log)} B$ and refer to the weak log majorization if

$$\prod_{i=1}^k \lambda_i(A) \leq \prod_{i=1}^k \lambda_i(B) \quad \text{for } k = 1, 2, \dots, n.$$

Further the log majorization $A \prec_{(\log)} B$ means that $A \prec_{w(\log)} B$ and the equality holds for $k = n$ in the above, i.e.,

$$\prod_{i=1}^n \lambda_i(A) = \prod_{i=1}^n \lambda_i(B) \quad \text{i.e.,} \quad \det A = \det B.$$

Note that when $A, B > 0$ the log majorization $A \prec_{(\log)} B$ is equivalent to $\log A \prec \log B$. It is known that for positive semidefinite $A, B \geq 0$,

$$A \prec_{w(\log)} B \implies A \prec_w B \implies \|A\| \leq \|B\|$$

for any unitarily invariant norm. See [1, 12] for theory of majorization for matrices.

For each matrix X and $k = 1, 2, \dots, n$, let $C_k(X)$ denote the k -fold antisymmetric tensor power of X . See [12] for details. Then (1)-(3) below are basic facts, (4) is easily seen from (2) and (3), and (5) follows from the Binet-Cauchy theorem.

Lemma 2.1. (1) $C_k(X^*) = C_k(X)^*$.

(2) $C_k(XY) = C_k(X)C_k(Y)$ for every pair of matrices X, Y .

(3) $C_k(X^{-1}) = C_k(X)^{-1}$ for nonsingular X .

(4) $C_k(A^p) = C_k(A)^p$ for every positive definite $A > 0$ and all $p \in \mathbb{R} \setminus \{0\}$.

(5) For every $A > 0$, $\prod_{i=1}^k \lambda_i(A) = \lambda_1(C_k(A))$ for $k = 1, 2, \dots, n$ and consequently, for $A, B > 0$, $\lambda_1(C_k(A)) \leq \lambda_1(C_k(B))$ for all $k = 1, \dots, n$ if and only if $A \prec_{w(\log)} B$.

3 Ando-Hiai Log-Majorization of negative power. For $0 \leq \alpha \leq 1$, the matrix α -geometric mean is the matrix mean corresponding to the matrix monotone function t^α . Note that $A \sharp_\alpha B = B \sharp_{1-\alpha} A$ and if $AB = BA$ then $A \sharp_\alpha B = A^{1-\alpha} B^\alpha$, and $(A, B) \mapsto A \sharp_\alpha B$ is jointly monotone, also see [5, Lemma 3.2].

On the other hand, the β -quasi geometric mean for $\beta \in [-1, 0)$ has the following properties in [7]; for any positive definite matrices A, B and C

(i) **consistency with scalars:** If A and B commute, then $A \natural_\beta B = A^{1-\beta} B^\beta$.

(ii) **homogeneity:** $(\alpha A) \natural_\beta (\alpha B) = \alpha(A \natural_\beta B)$ for all $\alpha > 0$.

(iii) **right reverse monotonicity:** $B \leq C$ implies $A \natural_\beta B \geq A \natural_\beta C$.

We recall the log-majorization theorem due to Ando-Hiai [2]: For each $\alpha \in [0, 1]$

$$A^r \sharp_\alpha B^r \prec_{(\log)} (A \sharp_\alpha B)^r \quad \text{for } r \geq 1,$$

or equivalently

$$(A^p \sharp_\alpha B^p)^{\frac{1}{p}} \prec_{(\log)} (A^q \sharp_\alpha B^q)^{\frac{1}{q}} \quad \text{for } 0 < q < p.$$

To show the main theorem related to the β -quasi geometric mean for $\beta \in [-1, 0)$, we need the following Ando-Hiai log-majorization of negative power $\beta \in [-1, 0)$:

Theorem 3.1. For every positive definite matrices $A, B > 0$ and $\beta \in [-1, 0)$,

$$(3.1) \quad A^r \sharp_{\beta} B^r \prec_{(\log)} (A \sharp_{\beta} B)^r \quad \text{for all } 0 < r \leq 1$$

or equivalently

$$(3.2) \quad (A \sharp_{\beta} B)^r \prec_{(\log)} (A^r \sharp_{\beta} B^r) \quad \text{for all } r \geq 1,$$

$$(3.3) \quad (A^q \sharp_{\beta} B^q)^{\frac{1}{q}} \prec_{(\log)} (A^p \sharp_{\beta} B^p)^{\frac{1}{p}} \quad \text{for all } 0 < q \leq p.$$

Proof. The equivalence of (3.1)-(3.3) is immediate. It is easy to see by Lemma 2.1 that for $k = 1, \dots, n$

$$C_k(A^r \sharp_{\beta} B^r) = C_k(A)^r \sharp_{\beta} C_k(B)^r$$

and

$$C_k((A \sharp_{\beta} B)^r) = (C_k(A) \sharp_{\beta} C_k(B))^r.$$

Also,

$$\det(A^r \sharp_{\beta} B^r) = (\det A)^{r(1-\beta)} (\det B)^{r\beta} = \det(A \sharp_{\beta} B)^r.$$

Hence, in order to prove (3.1), it suffices to show that

$$(3.4) \quad \lambda_1(A^r \sharp_{\beta} B^r) \leq \lambda_1(A \sharp_{\beta} B)^r \quad \text{for all } 0 < r \leq 1.$$

For this purpose we may prove that $A \sharp_{\beta} B \leq I$ implies $A^r \sharp_{\beta} B^r \leq I$, because both sides of (3.4) have the same order of homogeneity for A, B , so that we can multiply A, B by a positive constant.

First let us assume $\frac{1}{2} \leq r \leq 1$ and write $r = 1 - \varepsilon$ with $0 \leq \varepsilon \leq \frac{1}{2}$. Let $C = A^{\frac{1}{2}} B^{-1} A^{\frac{1}{2}}$. Then $B^{-1} = A^{-\frac{1}{2}} C A^{-\frac{1}{2}}$ and $A \sharp_{\beta} B = A^{\frac{1}{2}} C^{-\beta} A^{\frac{1}{2}}$. If $A \sharp_{\beta} B \leq I$, then $C^{-\beta} \leq A^{-1}$ so that $A \leq C^{\beta}$ and $A^{\varepsilon} \leq C^{\beta\varepsilon}$ for $0 \leq \varepsilon \leq \frac{1}{2}$ by Löwner-Heinz inequality. Since $-\beta \in (0, 1]$ and $1 - \varepsilon \in [\frac{1}{2}, 1]$, we now get

$$\begin{aligned} A^r \sharp_{\beta} B^r &= A^{\frac{1-\varepsilon}{2}} (A^{\frac{\varepsilon-1}{2}} B^{1-\varepsilon} A^{\frac{\varepsilon-1}{2}})^{\beta} A^{\frac{1-\varepsilon}{2}} \\ &= A^{\frac{1-\varepsilon}{2}} (A^{\frac{1-\varepsilon}{2}} (B^{-1})^{1-\varepsilon} A^{\frac{1-\varepsilon}{2}})^{-\beta} A^{\frac{1-\varepsilon}{2}} \\ &= A^{\frac{1-\varepsilon}{2}} (A^{\frac{1-\varepsilon}{2}} (A^{-\frac{1}{2}} C A^{-\frac{1}{2}})^{1-\varepsilon} A^{\frac{1-\varepsilon}{2}})^{-\beta} A^{\frac{1-\varepsilon}{2}} \\ &= A^{\frac{1-\varepsilon}{2}} (A^{-\frac{\varepsilon}{2}} [A \sharp_{1-\varepsilon} C] A^{-\frac{\varepsilon}{2}})^{-\beta} A^{\frac{1-\varepsilon}{2}} \\ &= A^{\frac{1}{2}-\varepsilon} [A^{\varepsilon} \sharp_{-\beta} (A \sharp_{1-\varepsilon} C)] A^{\frac{1}{2}-\varepsilon} \\ &\leq A^{\frac{1}{2}-\varepsilon} [C^{\beta\varepsilon} \sharp_{-\beta} (C^{\beta} \sharp_{1-\varepsilon} C)] A^{\frac{1}{2}-\varepsilon}, \end{aligned}$$

using the joint monotonicity of matrix geometric means. Since a direct computation yields

$$C^{\beta\varepsilon} \sharp_{-\beta} (C^{\beta} \sharp_{1-\varepsilon} C) = C^{\beta(2\varepsilon-1)}$$

and by Löwner-Heinz inequality and $0 \leq 1 - 2\varepsilon \leq 1$, $C^{-\alpha} \leq A^{-1}$ implies $C^{-\beta(1-2\varepsilon)} \leq A^{-(1-2\varepsilon)}$ and thus we get

$$A^r \sharp_{\beta} B^r \leq A^{\frac{1}{2}-\varepsilon} C^{\beta(2\varepsilon-1)} A^{\frac{1}{2}-\varepsilon} \leq A^{\frac{1}{2}-\varepsilon} A^{-1+2\varepsilon} A^{\frac{1}{2}-\varepsilon} = I.$$

Therefore (3.4) is proved in the case of $\frac{1}{2} \leq r \leq 1$.

When $0 < r < \frac{1}{2}$, writing $r = 2^{-k}(1 - \varepsilon)$ with $k \in \mathbb{N}$ and $0 \leq \varepsilon \leq \frac{1}{2}$, and repeating the argument above we have

$$\begin{aligned} \lambda_1(A^r \natural_{\beta} B^r) &\leq \lambda_1(A^{2^{-(k-1)}(1-\varepsilon)} \natural_{\beta} B^{2^{-(k-1)}(1-\varepsilon)})^{\frac{1}{2}} \\ &\vdots \\ &\leq \lambda_1(A^{1-\varepsilon} \natural_{\beta} B^{1-\varepsilon})^{2^{-k}} \\ &\leq \lambda_1(A \natural_{\beta} B)^r \end{aligned}$$

and so the proof is complete. \square

By Theorem 3.1, we have the following results:

Theorem 3.2. *Let A and B be positive definite matrices and $\|\cdot\|$ any unitarily invariant norm, and $\beta \in [-1, 0)$. If f is a continuous non-decreasing function on $[0, \infty)$ such that $f(0) \geq 0$ and $f(e^t)$ is convex, then*

$$\|f(A^r \natural_{\beta} B^r)\| \leq \|f((A \natural_{\beta} B)^r)\| \quad \text{for all } 0 < r \leq 1.$$

In particular,

$$\|A^r \natural_{\beta} B^r\| \leq \|(A \natural_{\beta} B)^r\| \quad \text{for all } 0 < r \leq 1$$

or equivalently

$$\begin{aligned} \|(A \natural_{\beta} B)^r\| &\leq \|(A^r \natural_{\beta} B^r)\| \quad \text{for all } r \geq 1, \\ \|(A^q \natural_{\beta} B^q)^{\frac{1}{q}}\| &\leq \|(A^p \natural_{\beta} B^p)^{\frac{1}{p}}\| \quad \text{for all } 0 < q \leq p. \end{aligned}$$

Proof. By [9, Proposition 4.4.13], if $A \prec_{w(\log)} B$ for positive definite matrices A and B and f is a continuous non-decreasing function on $[0, \infty)$ such that $f(0) \geq 0$ and $f(e^t)$ is convex, then $f(A) \prec_w f(B)$ and so $\|f(A)\| \leq \|f(B)\|$. Hence Theorem 3.2 follows from Theorem 3.1. \square

Corollary 3.3. *For every positive definite matrices $A, B > 0$ and $\beta \in [-1, 0)$,*

$$A \natural_{\beta} B \leq I \quad \text{implies} \quad A^r \natural_{\beta} B^r \leq I \quad \text{for all } 0 < r \leq 1.$$

4 Norm inequalities for quasi geometric mean. In this section, we show the main norm inequalities related to the quasi geometric mean for positive definite matrices. By [5, Lemma 5.5], we have the following quasi-geometric mean version of the Lie-Trotter formula: If A and B are positive definite matrices, then for each $\beta \in [-1, 0)$

$$(4.1) \quad A \diamond_{\beta} B = \lim_{p \rightarrow 0} (A^p \natural_{\beta} B^p)^{\frac{1}{p}}$$

and so for each $\beta \in [-1, 0)$ $\|(A^p \natural_{\beta} B^p)^{\frac{1}{p}}\|$ decreases to $\|A \diamond_{\beta} B\|$ as $p \downarrow 0$. Hence we have the following norm inequality for the quasi geometric mean of negative power:

Theorem 4.1. *Let A and B be positive definite matrices. Then for every unitarily invariant norm*

$$\|A \diamond_{\beta} B\| \leq \|A \natural_{\beta} B\| \quad \text{for all } \beta \in [-1, 0).$$

Proof. By Theorem 3.2, it follows that

$$\left\| (A^q \natural_{\beta} B^q)^{\frac{1}{q}} \right\| \leq \left\| (A^p \natural_{\beta} B^p)^{\frac{1}{p}} \right\| \quad \text{for all } 0 < q < p$$

and as $q \rightarrow 0$ and $p = 1$ we have the desired inequality by (4.1). \square

Theorem 4.2. *Let A and B be positive definite matrices. Then for every unitarily invariant norm*

$$(4.2) \quad \|A \natural_{\beta} B\| \leq \|A^{1-\beta} B^{\beta}\| \quad \text{for all } \beta \in [-1, -\frac{1}{2}].$$

Proof. For the matrix norm $\|\cdot\|$, by the Araki-Cordes inequality $\|B^t A^t B^t\| \leq \|(BAB)^t\|$ for all $t \in [0, 1]$, we have for $-1 \leq \beta \leq -\frac{1}{2}$

$$\begin{aligned} \|A \natural_{\beta} B\| &= \left\| A^{\frac{1}{2}} (A^{-\frac{1}{2}} B A^{-\frac{1}{2}})^{\beta} A^{\frac{1}{2}} \right\| \\ &= \left\| A^{\frac{1}{2}} (A^{\frac{1}{2}} B^{-1} A^{\frac{1}{2}})^{-\beta} A^{\frac{1}{2}} \right\| \\ &\leq \left\| A^{-\frac{1}{2\beta}} A^{\frac{1}{2}} B^{-1} A^{\frac{1}{2}} A^{-\frac{1}{2\beta}} \right\|^{-\beta} \quad \text{by } \frac{1}{2} \leq -\beta \leq 1 \\ &= \left\| A^{\frac{\beta-1}{2\beta}} B^{-1} A^{\frac{\beta-1}{2\beta}} \right\|^{-\beta} \\ &\leq \|A^{1-\beta} B^{2\beta} A^{1-\beta}\|^{\frac{1}{2}} \quad \text{for } \frac{1}{2} \leq -\frac{1}{2\beta} \leq 1 \\ &= \left\| (A^{1-\beta} B^{2\beta} A^{1-\beta})^{\frac{1}{2}} \right\| \end{aligned}$$

and this implies

$$\lambda_1(A \natural_{\beta} B) \leq \lambda_1((A^{1-\beta} B^{2\beta} A^{1-\beta})^{\frac{1}{2}}) = \lambda_1(|B^{\beta} A^{1-\beta}|).$$

Replacing A and B by (5) of Lemma 2.1, we obtain

$$\prod_{i=1}^k \lambda_i(A \natural_{\beta} B) \leq \prod_{i=1}^k \lambda_i(|B^{\beta} A^{1-\beta}|) \quad \text{for } k = 1, \dots, n.$$

Hence we have the weak log majorization $A \natural_{\beta} B \prec_{w(\log)} |B^{\beta} A^{1-\beta}|$ and this implies

$$\|A \natural_{\beta} B\| \leq \| |B^{\beta} A^{1-\beta}| \| = \|B^{\beta} A^{1-\beta}\| = \|A^{1-\beta} B^{\beta}\|$$

for every unitarily invariant norm and so we have the desired inequality (4.2). \square

Remark 4.3. In Theorem 4.2, the inequality $\|A \natural_{\beta} B\| \leq \|A^{1-\beta} B^{\beta}\|$ does not always hold for $-1/2 < \beta < 0$. In fact, if we put $\beta = -\frac{1}{3}$, $A = \begin{pmatrix} 2 & 1 \\ 1 & 2 \end{pmatrix}$ and $B = \begin{pmatrix} 2 & 1 \\ 1 & 1 \end{pmatrix}$, then we have the matrix norm $\|A \natural_{-\frac{1}{3}} B\| = 3.385$ and $\|A^{\frac{4}{3}} B^{-\frac{1}{3}}\| = 3.375$, and so $\|A \natural_{\beta} B\| > \|A^{1-\beta} B^{\beta}\|$.

Theorem 4.4. *Let A and B be positive definite matrices. Then for every unitarily invariant norm*

$$(4.3) \quad \|A \diamond_{\beta} B\| \leq \left\| \left(B^{\frac{\beta q}{2}} A^{(1-\beta)q} B^{\frac{\beta q}{2}} \right)^{\frac{1}{q}} \right\| \leq \|A \natural_{\beta} B\| \quad \text{for } 0 < q \leq \frac{1}{2} \text{ and } \beta \in [-1, 0)$$

and

$$(4.4) \quad \|A \natural_{\beta} B\| \leq \|A^{1-\beta} B^{\beta}\| \leq \left\| \left(B^{\frac{\beta p}{2}} A^{(1-\beta)p} B^{\frac{\beta p}{2}} \right)^{\frac{1}{p}} \right\| \quad \text{for } p \geq 2 \text{ and } \beta \in [-1, -\frac{1}{2}].$$

Proof. Since the first inequality in (4.3) follows from the Lie-Trotter formula (4.1), we show the second inequality in (4.3). By Theorem 3.2, we have $\left\| (A^r \natural_{\beta} B^r)^{\frac{1}{r}} \right\| \leq \|A \natural_{\beta} B\|$ for all $0 < r \leq 1$. For the matrix norm $\|\cdot\|$, by the Araki-Cordes inequality $\|(BAB)^t\| \leq \|B^t A^t B^t\|$ for all $t \geq 1$, we have for $0 < r \leq 1$

$$\begin{aligned} \|A \natural_{\beta} B\| &\geq \|A^r \natural_{\beta} B^r\|^{\frac{1}{r}} \\ &= \|B^r \natural_{1-\beta} A^r\|^{\frac{1}{r}} \\ &= \left\| B^{\frac{r}{2}} (B^{-\frac{r}{2}} A^r B^{-\frac{r}{2}})^{1-\beta} B^{\frac{r}{2}} \right\|^{\frac{1}{r}} \\ &\geq \left\| B^{\frac{\beta r}{2(1-\beta)}} A^r B^{\frac{\beta r}{2(1-\beta)}} \right\|^{\frac{1-\beta}{r}} \quad \text{by } 0 < \frac{1}{1-\beta} < 1 \\ &\geq \left\| B^{\frac{\beta r}{4}} A^{\frac{(1-\beta)r}{2}} B^{\frac{\beta r}{4}} \right\|^{\frac{2}{r}} \quad \text{by } \frac{1}{2} < \frac{1-\beta}{2} \leq 1. \end{aligned}$$

If we put $q = \frac{r}{2}$, then we have $\left\| \left(B^{\frac{\beta q}{2}} A^{(1-\beta)q} B^{\frac{\beta q}{2}} \right)^{\frac{1}{q}} \right\| \leq \|A \natural_{\beta} B\|$ for $0 < q \leq \frac{1}{2}$ and this implies

$$\lambda_1 \left(\left(B^{\frac{\beta q}{2}} A^{(1-\beta)q} B^{\frac{\beta q}{2}} \right)^{\frac{1}{q}} \right) \leq \lambda_1(A \natural_{\beta} B).$$

Replacing A and B by (5) of Lemma 2.1, we obtain

$$\prod_{i=1}^k \lambda_1 \left(\left(B^{\frac{\beta q}{2}} A^{(1-\beta)q} B^{\frac{\beta q}{2}} \right)^{\frac{1}{q}} \right) \leq \prod_{i=1}^k \lambda_1(A \natural_{\beta} B) \quad \text{for } k = 1, \dots, n,$$

which gives the second inequality in (4.3).

Next, for $s \geq 1$, it follows from Theorem 3.2 that $\|A \natural_{\beta} B\| \leq \left\| (A^s \natural_{\beta} B^s)^{\frac{1}{s}} \right\|$. For the matrix norm $\|\cdot\|$, we have

$$\begin{aligned} \|A \natural_{\beta} B\| &\leq \|A^s \natural_{\beta} B^s\|^{\frac{1}{s}} \\ &= \left\| A^{\frac{s}{2}} (A^{\frac{s}{2}} B^{-s} A^{\frac{s}{2}})^{-\beta} A^{\frac{s}{2}} \right\|^{\frac{1}{s}} \\ &\leq \left\| A^{-\frac{(1-\beta)s}{2\beta}} B^{-s} A^{-\frac{(1-\beta)s}{2\beta}} \right\|^{-\frac{\beta}{s}} \quad \text{by } \frac{1}{2} \leq -\beta \leq 1 \\ &\leq \left\| A^{(1-\beta)s} B^{2\beta s} A^{(1-\beta)s} \right\|^{\frac{1}{2s}} \quad \text{by } \frac{1}{2} \leq -\frac{1}{2\beta} \leq 1. \end{aligned}$$

If we put $p = 2s$, then we have

$$\begin{aligned} \|A \natural_{\beta} B\| &\leq \left\| A^{\frac{(1-\beta)p}{2}} B^{\beta p} A^{\frac{(1-\beta)p}{2}} \right\|^{\frac{1}{p}} \\ &= \text{spr} \left(A^{\frac{(1-\beta)p}{2}} B^{\beta p} A^{\frac{(1-\beta)p}{2}} \right)^{\frac{1}{p}} \\ &= \text{spr} \left(B^{\frac{\beta p}{2}} A^{(1-\beta)p} B^{\frac{\beta p}{2}} \right)^{\frac{1}{p}} \\ &= \left\| B^{\frac{\beta p}{2}} A^{(1-\beta)p} B^{\frac{\beta p}{2}} \right\|^{\frac{1}{p}} \\ &= \left\| \left(B^{\frac{\beta p}{2}} A^{(1-\beta)p} B^{\frac{\beta p}{2}} \right)^{\frac{1}{p}} \right\| \end{aligned}$$

for $p \geq 2$, where $\text{spr}(X)$ is the spectral radius of X . By the argument similar to above, we have the inequality (4.4). \square

Let A and B be positive definite matrices in \mathbb{M}_n and $\beta \in [-1, 0)$. Since there is the case that $(1-\beta)A + \beta B$ is not positive semidefinite, we have no relation between $\|(1-\beta)A + \beta B\|$ and $\|A \natural_{\beta} B\|$ though $(1-\beta)A + \beta B \leq A \natural_{\beta} B$. Suppose that $A \geq B$. Then $(1-\beta)A^p + \beta B^p$ is positive definite for all $p \in (0, 1]$. In particular $0 < (1-\beta)A + \beta B \leq A \natural_{\beta} B$ and so $\|(1-\beta)A + \beta B\| \leq \|A \natural_{\beta} B\|$ for every unitarily invariant norm. Thus under the assumption $A \geq B$, we consider the refinement of this norm inequality. For this, we need the following result due to J. I. Fujii [6]: A real valued continuous function f on an interval J is matrix concave if and only if

$$(4.5) \quad f((1-\beta)H + \beta K) \leq (1-\beta)f(H) + \beta f(K)$$

for all Hermitian matrices H and K with $\sigma(H), \sigma(K)$ and $\sigma((1-\beta)H + \beta K) \subset J$ and $\beta \in [-1, 0)$.

Let $0 < q < p \leq 1$. Then the function $f(t) = t^{\frac{q}{p}}$ on $[0, \infty)$ is matrix concave and by (4.5)

$$(4.6) \quad ((1-\beta)A^p + \beta B^p)^{\frac{q}{p}} \leq (1-\beta)A^q + \beta B^q.$$

Note that $(1-\beta)A^p + \beta B^p > 0$ for all $p \in (0, 1]$ since $A \geq B$. This implies that

$$\lambda_i((1-\beta)A^p + \beta B^p)^{\frac{q}{p}} \leq \lambda_i((1-\beta)A^q + \beta B^q) \quad \text{for all } i = 1, \dots, n.$$

Taking q -th roots of both sides, we obtain

$$\lambda_i((1-\beta)A^p + \beta B^p)^{\frac{1}{p}} \leq \lambda_i((1-\beta)A^q + \beta B^q)^{\frac{1}{q}} \quad \text{for all } i = 1, \dots, n$$

and so $\left\|((1-\beta)A^p + \beta B^p)^{\frac{1}{p}}\right\|$ is a decreasing function of p .

On the other hand, taking the logarithm of both sides in (4.6) and by (4.5), we obtain

$$\begin{aligned} \log((1-\beta)A^p + \beta B^p)^{\frac{1}{p}} &\leq \frac{1}{q} \log((1-\beta)A^q + \beta B^q) \\ &\leq (1-\beta) \log A + \beta \log B \end{aligned}$$

and this implies

$$\lambda_i\left(\log((1-\beta)A^p + \beta B^p)^{\frac{1}{p}}\right) \leq \lambda_i((1-\beta) \log A + \beta \log B) \quad \text{for all } i = 1, \dots, n.$$

Taking the exponent of both sides, we obtain

$$\lambda_i((1-\beta)A^p + \beta B^p)^{\frac{1}{p}} \leq \lambda_i\left(e^{(1-\beta) \log A + \beta \log B}\right) \quad \text{for all } i = 1, \dots, n$$

and so

$$\left\|((1-\beta)A^p + \beta B^p)^{\frac{1}{p}}\right\| \leq \left\|e^{(1-\beta) \log A + \beta \log B}\right\|$$

for all $p \in (0, 1]$. Summing up, we obtain the following result:

Theorem 4.5. *Let A and B be positive definite matrices in \mathbb{M}_n such that $A \geq B$ and $\beta \in [-1, 0)$. Then for every unitarily invariant norm*

$$\|(1-\beta)A + \beta B\| \leq \left\|((1-\beta)A^p + \beta B^p)^{\frac{1}{p}}\right\| \leq \|A \diamond_{\beta} B\| \leq \|A \natural_{\beta} B\|$$

for all $p \in (0, 1]$.

Finally, as an application, we show a refinement of the generalized Golden-Thompson inequality in terms of the quasi geometric means. Let H and K be Hermitian matrices. The Golden-Thompson trace inequality is

$$\text{Tr}[e^{H+K}] \leq \text{Tr}[e^H e^K].$$

Hiai-Petz [10] proved the complemented Golden-Thompson inequality:

$$\| \| e^H \#_{\alpha} e^K \| \| \leq \| \| e^{(1-\alpha)H+\alpha K} \| \| \quad \text{for all } \alpha \in [0, 1]$$

for every unitarily invariant norm. By Theorem 4.1 and Theorem 4.2, we have a refinement of the Golden-Thompson inequality in terms of the quasi geometric means:

$$\| \| e^{(1-\beta)H+\beta K} \| \| \leq \| \| e^H \natural_{\beta} e^K \| \| \leq \| \| e^{(1-\beta)H} e^{\beta K} \| \|$$

for all $\beta \in [-1, -\frac{1}{2}]$ and so

$$\text{Tr}[e^{H+K}] \leq \text{Tr}[e^{\frac{1}{1-\beta}H} \natural_{\beta} e^{\frac{1}{\beta}K}] \leq \text{Tr}[e^H e^K].$$

In particular, if we put $\alpha = \frac{1}{2}$ and $\beta = -\frac{1}{2}$, then we have

$$\text{Tr}[e^{2H} \#_{\frac{1}{2}} e^{2K}] \leq \text{Tr}[e^{H+K}] \leq \text{Tr}[e^{\frac{2}{3}H} \natural_{-\frac{1}{2}} e^{-2K}] \leq \text{Tr}[e^H e^K].$$

Acknowledgement. The second author is partially supported by JSPS KAKENHI Grant Number JP 16K05253.

REFERENCES

- [1] T. Ando, *Majorization, doubly stochastic matrices, and comparison of eigenvalues*, Linear Algebra Appl., **118** (1989), 163–248.
- [2] T. Ando and F. Hiai, *Log-majorization and complementary Golden-Thompson type inequalities*, Linear Algebra Appl., **197,198** (1994), 113–131.
- [3] R. Bhatia, *Matrix Analysis*, Springer, New York, 1997.
- [4] R. Bhatia and P. Grover, *Norm inequalities related to the matrix geometric mean*, Linear Algebra Appl., **437** (2012), 726–733.
- [5] M. Fujii, J. Mičić Hot, J. Pečarić and Y. Seo, *Recent Developments of Mond-Pečarić Method in Operator Inequalities*, Monographs in Inequalities 4, Element, Zagreb, 2012.
- [6] J. I. Fujii, *An external version of the Jensen operator inequality*, Sci. Math. Japon., **73**, No.2&3 (2011), 125–128.
- [7] J.I. Fujii and Y. Seo, *Tsallis relative operator entropy with negative parameters*, Adv. Oper. Theory, **1** (2016), No.2, 219–236.
- [8] T. Furuta, *Invitation to linear operators*, Taylor&Francis, London, 2001.
- [9] F. Hiai, *Matrix Analysis: Matrix Monotone Functions, Matrix Means, and Majorization*, Inter. Information Sci., **16**, No. 2 (2010), 139–248.
- [10] F. Hiai and D. Petz, *The Golden-Thompson trace inequality is complemented*, Linear Algebra Appl., **181** (1993), 153–185.
- [11] F. Kubo and T. Ando, *Means of positive linear operators*, Math. Ann., **246**(1980), 205–224.
- [12] A.W. Marshall and I. Olkin, *Inequalities: Theory of majorization and its applications*, Mathematics in Science and Engineering, Vol. **143**, Academic Press, 1979.

Communicated by *Masatoshi Fujii*

* DEPARTMENT OF MATHEMATICS, FACULTY OF BASIC SCIENCES, UNIVERSITY OF BOJNORD,
P. O. BOX 1339, BOJNORD 94531, IRAN
E-mail address : `kian@member.ams.org`

** DEPARTMENT OF MATHEMATICS, OSAKA KYOIKU UNIVERSITY, ASAHIGAOKA, KASHIWARA,
OSAKA 582-8582, JAPAN
E-mail address : `yukis@cc.osaka-kyoiku.ac.jp`