

## ON THE CONVEXITY OF SOME TRACE FUNCTIONS

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*Abstract.* In this paper, we study the convexity of the trace function  $A \rightarrow \text{Tr } B^* f(A) B$ . And we get an extension of the Peierls-Bogolyubov inequality and the joint convexity of the trace geometric mean.

### 1. Introduction

Throughout this paper, we write  $\mathbf{M}_n$  to denote the  $n \times n$  complex matrices,  $\mathbf{H}_n$  the subset consisting of self-adjoint (Hermitain) matrices,  $\mathbf{P}_n$  and  $\mathbf{P}_n^+$  the subset consisting of positive semidefinite matrices and positive definite matrices, respectively. We write  $A \geq 0$  if  $A$  is positive semidefinite and  $A > 0$  if  $A$  is positive definite.

It is well known that, when  $f$  is convex, the trace function  $A \rightarrow \text{Tr } f(A)$  is convex for self-adjoint matrices. Some related results can be found in [6, Introduction]. See also [7]. Several important concepts in operator theory and quantum information theory are closely related to the convexity of this type of trace functions. For example, the Schatten  $p$ -norms  $(\text{Tr } |A|^p)^{\frac{1}{p}}$ , ( $p > 1$ ), the von Neumann entropy  $-\text{Tr } A \log A$ , and the Peierls-Bogolyubov inequality [5, 1, 4].

The Peierls-Bogolyubov inequality plays a very important role in quantum information theory, especially in the calculation of the partition function. It states that

$$\log \frac{\text{Tr } \exp(A+B)}{\text{Tr } \exp A} \geq \frac{\text{Tr } \exp(A) B}{\text{Tr } \exp A}.$$

It can also state the log-convexity of the trace function  $\text{Tr } \exp(A)$ , i.e., the map

$$A \rightarrow \log \text{Tr } \exp(A)$$

is convex on  $\mathbf{H}_n$ .

In [8], Hansen et al. studied the Peierls-Bogolyubov inequality for deformed exponentials. The deformed logarithm denoted  $\log_q$  is defined by setting

$$\log_q a = \begin{cases} \frac{a^{q-1} - 1}{q - 1} & q \neq 1, \\ \log a & q = 1, \end{cases} \quad a > 0.$$

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The deformed logarithm is also denoted by the  $q$ -logarithm. The deformed exponential function or the  $q$ -exponential is defined as the inverse function to the  $q$ -logarithm. It is denoted by  $\exp_q$  and is given by the formula

$$\exp_q a = \begin{cases} (a(q-1) + 1)^{1/(q-1)}, & a > -1/(q-1), & q > 1, \\ (a(q-1) + 1)^{1/(q-1)}, & a < -1/(q-1), & q < 1, \\ \exp a, & a \in \mathbb{R}, & q = 1. \end{cases}$$

In [8], the authors studied the convexity of the function

$$f(A) = \log_r \operatorname{Tr} B^* \exp_q(A)B.$$

It has been proved that

- (i) If  $q \leq 0$  and  $r \geq q$ , then  $f$  is convex.
- (ii) If  $\frac{3}{2} \leq q \leq 2$  and  $r \geq q$ , then  $f$  is convex.
- (iii) If  $q \geq 2$  and  $r \leq q$ , then  $f$  is concave.

We find there is a gap for  $0 \leq q \leq \frac{3}{2}$  that has not been studied yet. In this paper, we tackle with this problem and complement the corresponding results. We will also study the convexity of the trace function  $\exp_q \operatorname{Tr} H^* \log_q AH$ .

Moreover, we study the joint convexity of the trace function

$$(A, B) \rightarrow \operatorname{Tr} B^{\frac{1}{2}} f(B^{-\frac{1}{2}} A B^{-\frac{1}{2}}) B^{\frac{1}{2}}.$$

It enables us to consider the joint convexity of the trace geometric mean.

## 2. Main results

### 2.1. The convexity of trace functions

**THEOREM 2.1.** *Let  $B \in \mathbf{M}_n$ . If  $f : \mathbb{R} \rightarrow \mathbb{R}$  is a convex function, then the trace function*

$$F(A) = \operatorname{Tr} B^* f(A)B \tag{2.1}$$

*is convex for  $A \in \mathbf{H}_n$ .*

*Proof.* Let  $\{P_i\}$  be a pairwise orthogonal family of minimal projections with  $\sum_{i=1}^n P_i = I_n$ . Let  $\sum_j s_j Q_j$  be the spectral decomposition of a self-adjoint matrix  $C$ .

Then for positive definite matrix  $K(=BB^*)$ , we have  $\text{Tr} KP_i > 0$  for  $i = 1, \dots, n$ . Thus

$$\begin{aligned} \text{Tr} Kf(C) &= \sum_j f(s_j) \text{Tr} KQ_j = \sum_i \left( \sum_j f(s_j) \text{Tr} KQ_j P_i \right) \\ &= \sum_i \left[ \frac{\sum_j f(s_j) \text{Tr} KQ_j P_i}{\text{Tr} KP_i} \text{Tr} KP_i \right] \\ &\geq \sum_i (\text{Tr} KP_i) f \left( \sum_j s_j \frac{\text{Tr} KQ_j P_i}{\text{Tr} KP_i} \right) \\ &= \sum_i (\text{Tr} KP_i) f \left( \frac{\text{Tr} KCP_i}{\text{Tr} KP_i} \right). \end{aligned} \tag{2.2}$$

Now, assume  $\sum_{i=1}^n \mu_i P_i$  is the spectral decomposition of the convex combination

$$A = \lambda C_1 + (1 - \lambda) C_2.$$

It follows that

$$\begin{aligned} \text{Tr} Kf(A) &= \sum_i f(\mu_i) \text{Tr} KP_i \\ &= \sum_i (\text{Tr} KP_i) f \left( \frac{\text{Tr} KAP_i}{\text{Tr} KP_i} \right) \\ &= \sum_i (\text{Tr} KP_i) f \left( \frac{\text{Tr} K(\lambda C_1 + (1 - \lambda) C_2) P_i}{\text{Tr} KP_i} \right) \\ &\leq \sum_i (\text{Tr} KP_i) \left[ \lambda f \left( \frac{\text{Tr} KC_1 P_i}{\text{Tr} KP_i} \right) + (1 - \lambda) f \left( \frac{\text{Tr} KC_2 P_i}{\text{Tr} KP_i} \right) \right] \\ &\leq \lambda \text{Tr} Kf(C_1) + (1 - \lambda) \text{Tr} Kf(C_2), \end{aligned}$$

where the first inequality follows from the convexity of  $f$ , and the second inequality follows from applying the inequality (2.2) twice. Hence we have that

$$\text{Tr} Kf(\lambda C_1 + (1 - \lambda) C_2) \leq \lambda \text{Tr} Kf(C_1) + (1 - \lambda) \text{Tr} Kf(C_2), \tag{2.3}$$

with  $K$  positive definite.

For arbitrary matrix  $B \in \mathbf{M}_n$ ,  $K = BB^*$  is positive semidefinite. We set  $K_\varepsilon = K + \varepsilon I$  for an arbitrary  $\varepsilon \in (0, 1)$ . Then  $K_\varepsilon$  is positive definite and satisfies

$$\text{Tr} K_\varepsilon f(\lambda C_1 + (1 - \lambda) C_2) \leq \lambda \text{Tr} K_\varepsilon f(C_1) + (1 - \lambda) \text{Tr} K_\varepsilon f(C_2).$$

Letting  $\varepsilon \rightarrow 0$ , we have the inequality (2.3) holds for positive semidefinite matrix  $K$ . Hence, we have  $A \rightarrow \text{Tr} B^* f(A) B$  is convex.  $\square$

Now we consider the Jensen inequalities related to the trace function  $\text{Tr} B^* f(A) B$ . In this part, we denote the set of  $n \times m$  matrices by  $\mathbf{M}_{nm}$ .

**THEOREM 2.2.** *Let  $f : \mathbb{R} \rightarrow \mathbb{R}$  be convex. Then the following statements hold and they are equivalent:*

(i) *For  $B \in \mathbf{M}_n$ ,  $F(A) = \text{Tr} B^* f(A) B$  is convex on  $\mathbf{H}_n$ .*

(ii) *For  $A_i \in \mathbf{H}_n$ ,  $B \in \mathbf{M}_m$ , and  $H_i \in \mathbf{M}_{nm}$ , with  $\sum_i H_i^* H_i = I_m$ ,*

$$\text{Tr} B^* f\left(\sum_i H_i^* A_i H_i\right) B \leq \sum_i \text{Tr} B^* H_i^* f(A_i) H_i B.$$

(iii) *For  $A \in \mathbf{H}_n$ , and  $V \in \mathbf{M}_{nm}$ ,  $B \in \mathbf{M}_m$  with  $V^* V = I_m$ ,*

$$\text{Tr} B^* f(V^* A V) B \leq \text{Tr} B^* V^* f(A) V B.$$

*Proof.* According to Theorem 2.1, we should only prove the equivalence.

(i)  $\rightarrow$  (iii). Set  $Z = \begin{pmatrix} A & 0 \\ 0 & M \end{pmatrix}$ ,  $C = \begin{pmatrix} B & 0 \\ 0 & 0 \end{pmatrix}$ ,  $U = \begin{pmatrix} V & P \\ 0 & -V^* \end{pmatrix}$ ,  $W = \begin{pmatrix} V & -P \\ 0 & V^* \end{pmatrix}$ , where  $M \in \mathbf{H}_m$  and  $P = I - VV^*$ . Note that  $V$  is isometry, and  $V^* P = 0$ ,  $PV = 0$ , then  $U$  and  $W$  are unitary. By calculation,

$$U^* Z U = \begin{pmatrix} V^* A V & V^* A P \\ P A V & P A P + V M V^* \end{pmatrix}, \quad W^* Z W = \begin{pmatrix} V^* A V & -V^* A P \\ -P A V & P A P + V M V^* \end{pmatrix},$$

$$\frac{1}{2} (U^* Z U + W^* Z W) = \begin{pmatrix} V^* A V & 0 \\ 0 & P A P + V M V^* \end{pmatrix}.$$

Thus

$$\begin{aligned} & \text{Tr} B^* f(V^* A V) B \\ &= \text{Tr} \begin{pmatrix} B^* & 0 \\ 0 & 0 \end{pmatrix} f \begin{pmatrix} V^* A V & 0 \\ 0 & P A P + V M V^* \end{pmatrix} \begin{pmatrix} B & 0 \\ 0 & 0 \end{pmatrix} \\ &= \text{Tr} C^* f \left( \frac{1}{2} (U^* Z U + W^* Z W) \right) C \\ &\leq \frac{1}{2} \text{Tr} C^* f(U^* Z U) C + \frac{1}{2} \text{Tr} C^* f(W^* Z W) C \\ &= \text{Tr} C^* \left( \frac{1}{2} U^* f(Z) U + \frac{1}{2} W^* f(Z) W \right) C \\ &= \text{Tr} \begin{pmatrix} B^* & 0 \\ 0 & 0 \end{pmatrix} \begin{pmatrix} V^* f(A) V & 0 \\ 0 & P f(A) P + V f(M) V^* \end{pmatrix} \begin{pmatrix} B & 0 \\ 0 & 0 \end{pmatrix} \\ &= \text{Tr} B^* V^* f(A) V B. \end{aligned}$$

(iii)  $\rightarrow$  (ii). Set

$$A = \begin{bmatrix} A_1 & & & \\ & A_2 & & \\ & & \ddots & \\ & & & A_n \end{bmatrix}, \quad V = \begin{bmatrix} H_1 & 0 & \cdots & 0 \\ H_2 & 0 & \cdots & 0 \\ \vdots & 0 & \cdots & 0 \\ H_n & 0 & \cdots & 0 \end{bmatrix}, \quad D = \begin{bmatrix} B & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \end{bmatrix}.$$

Then

$$V^*AV = \begin{bmatrix} \sum_i H_i^* A_i H_i & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \\ \vdots & 0 & \cdots & 0 \\ 0 & 0 & \cdots & 0 \end{bmatrix}.$$

It follows that

$$\begin{aligned} \text{Tr } B^* f\left(\sum_i H_i^* A_i H_i\right) B &= \text{Tr } D^* f(V^*AV) D \\ &\leq \text{Tr } D^* V^* f(A) V D \\ &= \text{Tr } B^* \left(\sum_i H_i^* f(A_i) H_i\right) B \\ &= \sum_i \text{Tr } B^* H_i^* f(A_i) H_i B. \end{aligned}$$

(ii)  $\rightarrow$  (i) is easy to prove. We omit it.  $\square$

**2.2. An extension of Peierls-Bogolyubov’s inequality**

By Theorem 2.1, we know that the trace function  $A \rightarrow \text{Tr } B^* A^p B$  is concave for  $0 < p \leq 1$ , and convex for  $p < 0$ , and  $p > 1$ . In [8], the authors obtained the following proposition.

PROPOSITION 2.3. ([8, Proposition 1]) *Let  $f$  be a real positive function defined on  $\mathbf{P}_n^+$  and assume  $f$  is homogeneous of degree  $p \neq 0$ .*

- (i) *If  $f$  is convex and  $p > 0$ , then  $f^{1/p}$  is convex.*
- (ii) *If  $f$  is convex and  $p < 0$ , then  $f^{1/p}$  is concave.*
- (iii) *If  $f$  is convex and  $p < 0$  and  $r > 0$ , then  $f^r$  is convex.*
- (iv) *If  $f$  is concave and  $p > 0$ , then  $f^{1/p}$  is concave.*
- (v) *If  $f$  is concave and  $p < 0$ , then  $f^{1/p}$  is convex.*
- (vi) *If  $f$  is concave and  $p > 0$  and  $r < 0$ , then  $f^r$  is convex.*

According to Proposition 2.3, we have

PROPOSITION 2.4. *Let  $B \in \mathbf{M}_n$  be an arbitrary operator and consider the function*

$$F(A) = (\text{Tr } B^* A^p B)^{1/r}$$

*defined on positive definite matrices  $\mathbf{P}_n^+$ . Then*

- (i)  *$F$  is concave for  $r \leq p < 0$ ;*

- (ii)  $F$  is convex for  $p < 0$  and  $r > 0$ ;
- (iii)  $F$  is concave for  $0 < p \leq 1$  and  $r \geq p$ ;
- (iv)  $F$  is convex for  $0 < p \leq 1$  and  $r < 0$ ;
- (v)  $F$  is convex for  $p \geq 1$  and  $0 < r \leq p$ .

Note that (iii) and (iv) and the cases of  $-1 \leq p < 0$  and  $1 \leq p \leq 2$  in (i), (ii), (v) have already been discussed in [8]. And the cases of  $p < -1$  in (i), (ii), and the case of  $p > 2$  in (v) are new.

Consider the function

$$F(A) = \log_r \operatorname{Tr} B^* \exp_q(A)B$$

$$= \frac{1}{r-1} \left( \left( \operatorname{Tr} (B^*(A(q-1) + 1)^{\frac{1}{q-1}}B) \right)^{r-1} - 1 \right).$$

Let

$$\frac{1}{q-1} = p, \quad \frac{1}{r-1} = r.$$

Then by Proposition 2.4, we have

**THEOREM 2.5.** *Let  $B$  be an arbitrary matrix and consider the function*

$$F(A) = \log_r \operatorname{Tr} B^* \exp_q(A)B$$

*defined on self-adjoint  $A > -(q-1)^{-1}$ .*

- (i) *If  $q < 1$  and  $r \geq q$ , then  $F$  is convex.*
- (ii) *If  $1 < q \leq 2$  and  $r \geq q$ , then  $F$  is convex.*
- (iii) *If  $q \geq 2$  and  $r \leq q$ , then  $F$  is concave.*

When  $q = r$ , and letting  $q \rightarrow 1$ , we have

**THEOREM 2.6.** *Let  $B$  be an arbitrary matrix, then the function*

$$F(A) = \log \operatorname{Tr} B^* \exp(A)B$$

*is convex.*

That is, we have an extension of the Peierls-Bogoliubov inequality as

**COROLLARY 2.7.** *Let  $A, B \in \mathbf{H}_n$  be self-adjoint operators, and  $C \in \mathbf{M}_n$ . Then we have the following inequality*

$$\log \frac{\operatorname{Tr} C^* \exp(A+B)C}{\operatorname{Tr} C^* \exp(A)C} \geq \frac{\operatorname{Tr} C^* (d \exp(A)B)C}{\operatorname{Tr} C^* \exp(A)C}, \tag{2.4}$$

where  $d \exp(A)$  is the Fréchet derivative of the function  $\exp(A)$ .

*Proof.* Take self-adjoint matrices  $A, B \in \mathbf{H}_n$  and define the function

$$g(t) = \log \operatorname{Tr} C^* \exp(A + tB)C.$$

Since  $g(t)$  is convex, we obtain the inequality

$$g(1) - g(0) \geq g'(0).$$

Hence the inequality (2.4) follows.  $\square$

Now we consider the concavity of another trace function related to exponential and logarithmic.

**THEOREM 2.8.** *Let  $\operatorname{Tr} H^*H = 1$ . Consider the trace function*

$$G(A) = \exp_q \operatorname{Tr} H^* \log_q AH.$$

*We have*

- (i) *If  $q < 1$ , then  $G$  is concave.*
- (ii) *If  $1 < q \leq 2$ , then  $G$  is concave.*
- (iii) *If  $q \geq 2$ , then  $G$  is convex.*

*Proof.* Since

$$\begin{aligned} \exp_q \operatorname{Tr} H^* \log_q AH &= \left[ \left( \operatorname{Tr} H^* \frac{A^{q-1} - 1}{q-1} H \right) (q-1) + 1 \right]^{\frac{1}{q-1}} \\ &= \left[ \frac{\operatorname{Tr} H^* A^{q-1} H - \operatorname{Tr} H^* H}{q-1} (q-1) + 1 \right]^{\frac{1}{q-1}} \\ &= (\operatorname{Tr} H^* A^{q-1} H)^{\frac{1}{q-1}}, \end{aligned}$$

by Proposition 2.4, we obtain the conclusions.  $\square$

Letting  $q \rightarrow 1$ , we have

**COROLLARY 2.9.** *Let  $\operatorname{Tr} H^*H = 1$ . The trace function*

$$G(A) = \exp \operatorname{Tr} H^* \log AH$$

*is concave on self-adjoint matrices.*

### 2.3. Trace geometric mean

Now we study the joint convexity of some trace functions.

**THEOREM 2.10.** *Let  $f$  be a convex function. Then the trace function*

$$(A, B) \rightarrow \text{Tr} B^{\frac{1}{2}} f(B^{-\frac{1}{2}} A B^{-\frac{1}{2}}) B^{\frac{1}{2}}$$

*is jointly convex.*

*Proof.* Let  $\lambda \in (0, 1)$ ,  $A = \lambda A_1 + (1 - \lambda) A_2$ ,  $B = \lambda B_1 + (1 - \lambda) B_2$ . And set  $(\lambda B_1)^{\frac{1}{2}} B^{-\frac{1}{2}} = K_1$ ,  $[(1 - \lambda) B_2]^{\frac{1}{2}} B^{-\frac{1}{2}} = K_2$ . Then  $K_1, K_2$  satisfy  $K_1^* K_1 + K_2^* K_2 = I$ . Since  $f$  is convex, it follows from Theorem 2.2,

$$\begin{aligned} & \text{Tr} B^{\frac{1}{2}} f(B^{-\frac{1}{2}} A B^{-\frac{1}{2}}) B^{\frac{1}{2}} \\ &= \text{Tr} B^{\frac{1}{2}} f(K_1^* B_1^{-\frac{1}{2}} A_1 B_1^{-\frac{1}{2}} K_1 + K_2^* B_2^{-\frac{1}{2}} A_2 B_2^{-\frac{1}{2}} K_2) B^{\frac{1}{2}} \\ &\leq \text{Tr} B^{\frac{1}{2}} K_1^* f(B_1^{-\frac{1}{2}} A_1 B_1^{-\frac{1}{2}}) K_1 B^{\frac{1}{2}} + \text{Tr} B^{\frac{1}{2}} K_2^* f(B_2^{-\frac{1}{2}} A_2 B_2^{-\frac{1}{2}}) K_2 B^{\frac{1}{2}} \\ &= \lambda \text{Tr} B_1^{\frac{1}{2}} f(B_1^{-\frac{1}{2}} A_1 B_1^{-\frac{1}{2}}) B_1^{\frac{1}{2}} + (1 - \lambda) \text{Tr} B_2^{\frac{1}{2}} f(B_2^{-\frac{1}{2}} A_2 B_2^{-\frac{1}{2}}) B_2^{\frac{1}{2}}. \quad \square \end{aligned}$$

Now we consider the jointly convexity of the trace geometric mean

$$\hat{Q}_\alpha(A, B) = \text{Tr} B^{\frac{1}{2}} (B^{-\frac{1}{2}} A B^{-\frac{1}{2}})^\alpha B^{\frac{1}{2}}. \tag{2.5}$$

From Theorem 2.10, we have the following corollary.

**COROLLARY 2.11.** *The trace geometric mean  $\hat{Q}_\alpha(A, B)$  is jointly convex for  $\alpha \geq 1$  and  $\alpha < 0$ , and is jointly concave for  $\alpha \in (0, 1)$ .*

Note that the operator geometric mean  $G(A, B) = B^{\frac{1}{2}} (B^{-\frac{1}{2}} A B^{-\frac{1}{2}})^\alpha B^{\frac{1}{2}}$  is jointly concave for  $\alpha \in (0, 1)$ , and jointly convex for  $1 \leq \alpha \leq 2$  and  $-1 \leq \alpha < 0$ . See [3].

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