

TRACE OPERATOR INEQUALITY FOR SUPERQUADRATIC FUNCTIONS

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ABSTRACT. In this work, some operator trace inequalities are proved. An extension of Klein's inequality for all Hermitian matrices is proved. A non-commutative version (or Hansen-Pedersen version) of the Jensen trace inequality is provided as well. A generalization of the result for any positive Hilbert space operators acts on a positive unital linear map is established.

1. INTRODUCTION

Let $\mathcal{B}(\mathcal{H})$ be the Banach algebra of all bounded linear operators defined on a complex Hilbert space $(\mathcal{H}; \langle \cdot, \cdot \rangle)$ with the identity operator $1_{\mathcal{H}}$ in $\mathcal{B}(\mathcal{H})$. When $\mathcal{H} = \mathbb{C}^n$, we identify $\mathcal{B}(\mathcal{H})$ with the algebra $\mathfrak{M}_{n \times n}$ of n -by- n complex matrices. Then, $\mathfrak{M}_{n \times n}^+$ is just the cone of n -by- n positive semidefinite matrices.

Let $A \in \mathfrak{M}_{n \times n}(\mathbb{C})$, the trace of a square matrix equals the sum of the eigenvalues counted with multiplicities. Moreover, the trace of a Hermitian matrix is real. If A is a linear operator represented by a square matrix with real or complex entries and if $\lambda_1, \dots, \lambda_n$ are the eigenvalues of A , then $\text{Tr}(A) = \sum_j \lambda_j$. This follows from the fact that A is always similar to its Jordan form, an upper triangular matrix having $\lambda_1, \dots, \lambda_n$ on the main diagonal.

The inner product

$$\langle A, B \rangle = \text{Tr}(A^* B)$$

which is defined on the space of all complex (or real) $m \times n$ matrices, is called the Frobenius norm, which satisfies submultiplicative property as matrix norm.

If A and B are real positive semi-definite matrices of the same size, using the Cauchy-Schwarz inequality, we have

$$0 \leq \text{Tr}^2(AB) \leq \text{Tr}(A^2) \text{Tr}(B^2) \leq \text{Tr}^2(A) \text{Tr}^2(B).$$

The concept of trace of a matrix is generalized to the trace class of compact operators on Hilbert spaces, and the analog of the Frobenius norm is called the Hilbert-Schmidt norm, which is can be defined as [19]:

$$\langle A, B \rangle = \text{Tr}(A^* B) = \sum_i \langle Ae_i, Be_i \rangle$$

over all orthonormal basis of \mathcal{H} , $\{e_i : i \in I\}$.

A Hilbert-Schmidt operator, is a bounded operator A on a Hilbert space \mathcal{H} with finite Hilbert-Schmidt norm

$$\|A\|_{\text{HS}}^2 = \text{Tr}(A^* A) = \sum_{i \in I} \|Ae_i\|^2,$$

where $\|\cdot\|$ is the norm of \mathcal{H} and Tr is the trace of a nonnegative selfadjoint operator. This definition is independent of the choice of the basis, and therefore

$$\|A\|_{\text{HS}}^2 = \sum_{i,j} |\langle e_i, Ae_j \rangle|^2 = \|A\|_2^2.$$

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A bounded linear operator A over a separable Hilbert space \mathcal{H} is said to be in the trace class if for some (and hence all) orthonormal bases $\{e_i\}_i$ of \mathcal{H} , the sum of positive terms

$$\|A\|_1 = \text{Tr } |A| := \sum_i \langle |A| e_i, e_i \rangle = \sum_i \langle (A^*A)^{1/2} e_i, e_i \rangle,$$

is finite. In this case, the trace of A , which is given by the sum

$$\text{Tr}(A) = \sum_i \langle A e_i, e_i \rangle,$$

is absolutely convergent and is independent of the choice of the orthonormal basis. When \mathcal{H} is finite-dimensional, every operator is trace class and this definition of trace of A coincides with the definition of the trace of a matrix.

A function $f : J \rightarrow \mathbb{R}$ is called convex iff

$$(1.1) \quad f(t\alpha + (1-t)\beta) \leq tf(\alpha) + (1-t)f(\beta),$$

for all points $\alpha, \beta \in J$ and all $t \in [0, 1]$. If $-f$ is convex then we say that f is concave. Moreover, if f is both convex and concave, then f is said to be affine.

Geometrically, for two point $(x, f(x))$ and $(y, f(y))$ on the graph of f are on or below the chord joining the endpoints for all $x, y \in I$, $x < y$. In symbols, we write

$$f(t) \leq \frac{f(y) - f(x)}{y - x} (t - x) + f(x)$$

for any $x \leq t \leq y$ and $x, y \in J$.

Equivalently, given a function $f : J \rightarrow \mathbb{R}$, we say that f admits a support line at $x \in J$ if there exists a $\lambda \in \mathbb{R}$ such that

$$(1.2) \quad f(t) \geq f(x) + \lambda(t - x)$$

for all $t \in J$.

The set of all such λ is called the subdifferential of f at x , and it's denoted by ∂f . Indeed, the subdifferential gives us the slopes of the supporting lines for the graph of f . So that if f is convex then $\partial f(x) \neq \emptyset$ at all interior points of its domain.

From this point of view Abramovich et al. [2] extend the above idea for what they called superquadratic functions. Namely, a function $f : [0, \infty) \rightarrow \mathbb{R}$ is called superquadratic provided that for all $x \geq 0$ there exists a constant $C_x \in \mathbb{R}$ such that

$$(1.3) \quad f(t) \geq f(x) + C_x(t - x) + f(|t - x|)$$

for all $t \geq 0$. We say that f is subquadratic if $-f$ is superquadratic. Thus, for a superquadratic function we require that f lie above its tangent line plus a translation of f itself. If f is differentiable and satisfies $f(0) = f'(0) = 0$, then one sees easily that the C_x appearing in the definition is necessarily $f'(x)$, (see [1]).

Prima facie, superquadratic function looks to be stronger than convex function itself but if f takes negative values then it may be considered as a weaker function. Therefore, if f is superquadratic and non-negative, then f is convex and increasing [2] (see also [4]).

Moreover, the following result holds for superquadratic function.

Lemma 1. [2] *Let f be superquadratic function. Then*

- (1) $f(0) \leq 0$
- (2) *If f is differentiable and $f(0) = f'(0) = 0$, then $C_x = f'(x)$ for all $x \geq 0$.*
- (3) *If $f(x) \geq 0$ for all $x \geq 0$, then f is convex and $f(0) = f'(0) = 0$.*

The next result gives a sufficient condition when convexity (concavity) implies super(sub)quadraticity.

Lemma 2. [2] *If f' is convex (concave) and $f(0) = f'(0) = 0$, then f is super(sub)quadratic. The converse of f is not true.*

Remark 1. *In general, non-negative subquadratic functions does not imply concavity. In other words, there exists a subquadratic function which is convex. For example, $f(x) = x^p$, $x \geq 0$ and $1 \leq p \leq 2$ is subquadratic and convex.*

Among others, Abramovich et al. [2] proved that the inequality

$$(1.4) \quad f\left(\int \varphi d\mu\right) \leq \int f(\varphi(s)) - f\left(\left|\varphi(s) - \int \varphi d\mu\right|\right) d\mu(s)$$

holds for all probability measures μ and all nonnegative, μ -integrable functions φ if and only if f is superquadratic. For more details, recent result and generalization the reader may refer to [4],[5] and [8]. Related operator inequalities can be found in [6], [7],[10] and [14]–[16].

2. JENSEN'S TRACE INEQUALITY

Let $f : \mathbb{R} \rightarrow \mathbb{R}$ be continuous, and let n be any integer. It's well known that if $t \mapsto f(t)$ is convex (monotone increasing), then the trace function $A \mapsto \text{Tr}(f(A))$ is convex (monotone increasing) [13] (see also [12]).

The next result shows that the trace function is real positive superquadratic function.

Lemma 3. *Given $f : [0, \infty) \rightarrow \mathbb{R}$ be a continuous function. If f is superquadratic, then the associated trace function, $A \mapsto \text{Tr}(f(A))$ is superquadratic function, for all Hermitian $n \times n$ matrices. Moreover, we have*

$$(2.1) \quad \text{Tr}\left(f\left(\frac{A+B}{2}\right)\right) + \text{Tr}\left(f\left(\frac{|A-B|}{2}\right)\right) \leq \frac{\text{Tr}(f(A)) + \text{Tr}(f(B))}{2}$$

for all Hermitian $n \times n$ matrices.

Proof. Let $\lambda_1, \dots, \lambda_n$ and μ_1, \dots, μ_n be the eigenvalues of A and B ; respectively. Then, for $0 \leq \alpha \leq 1$, we have

$$\begin{aligned} & \text{Tr}(f(\alpha A + (1-\alpha)B)) \\ &= \sum_j f(\alpha\lambda_j + (1-\alpha)\mu_j) \\ &\leq \sum_j [\alpha f(\lambda_j) + (1-\alpha)f(\mu_j) - \alpha f((1-\alpha)|\lambda_j - \mu_j|) - (1-\alpha)f(\alpha|\lambda_j - \mu_j|)] \\ &= \alpha \sum_j f(\lambda_j) + (1-\alpha) \sum_j f(\mu_j) - \alpha \sum_j f((1-\alpha)|\lambda_j - \mu_j|) - (1-\alpha) \sum_j f(\alpha|\lambda_j - \mu_j|) \\ &= \alpha \text{Tr}(f(A)) + (1-\alpha) \text{Tr}(f(B)) - \alpha \text{Tr}(f((1-\alpha)|A-B|)) - (1-\alpha) \text{Tr}(f(\alpha|A-B|)), \end{aligned}$$

which proves that $\text{Tr}(f(A))$ is superquadratic function on the set of all positive definite Hermitian $n \times n$ matrices. The inequality holds by setting $\alpha = \frac{1}{2}$. \square

Remark 2. *Another proof of Lemma 3 could be done using the Spectral Decomposition of A such that the real function f is defined on the spectrum of A .*

For all Hermitian $n \times n$ matrices A and B and all differentiable convex functions $f : \mathbb{R} \rightarrow \mathbb{R}$ with derivative f' , the well known Klein inequality reads that [18]:

$$(2.2) \quad \text{Tr}[f(A) - f(B) - (A-B)f'(B)] \geq 0,$$

The equality holds if and only if $A = B$.

For the choice $f(t) = t \log t$ ($t > 0$), we obtain [17]:

$$S(A, B) = \text{Tr} A (\log A - \log B) \geq \text{Tr}(A - B)$$

for B strictly positive and A nonnegative. The left-hand side is called relative entropy. If A and B are density matrices, i.e., $\text{Tr}(A) = \text{Tr}(B) = 1$, then $S(A, B) \geq 0$. This is a classical application of the Klein inequality (cf. [18]).

Let R, F be two Hermitian $n \times n$ matrices such that $\text{Tr}(e^R) = 1$. Define $g = \text{Tr}(Fe^R)$, then we obtain the Peierl–Bogoliubov inequality [18]:

$$\text{Tr}(e^F e^R) \geq \text{Tr}(e^{F+R}) \geq \text{Tr}(e^g).$$

The proof of this inequality follows from Klein inequality. Take $f(t) = e^x$, $A = R + F$ and $B = R + gI$.

Employing the concept of superquadraticity, one may has the following refinement of Klein's inequality (2.2).

Theorem 1. For all positive-definite Hermitian $n \times n$ matrices A and B , and all differentiable superquadratic functions $f : [0, \infty) \rightarrow \mathbb{R}$, satisfy that $f(0) = f'(0) = 0$, the inequality

$$(2.3) \quad \text{Tr} [f(A) - f(B) - (A - B)f'(B) - f(|A - B|)] \geq 0,$$

holds. The equality holds if and only if $A = B$.

Proof. Let f be superquadratic on $[0, \infty)$. Since f is differentiable and satisfies $f(0) = f'(0) = 0$, then one sees that the C_x in (1.3) is necessarily $f'(x)$ ([1]). Therefore, we have

$$f(t) \geq f(x) + f'(x)(t - x) + f(|t - x|)$$

for all $t, x \geq 0$. Therefore, if λ_j and μ_j ($1 \leq j \leq n$) are the eigenvalues of positive-definite Hermitian $n \times n$ matrices A and B , respectively; thus we have

$$f(\lambda_j) \geq f(\mu_j) + f'(\mu_j)(\lambda_j - \mu_j) + f(|\lambda_j - \mu_j|).$$

Taking the sum over j , we get

$$\sum_j f(\lambda_j) \geq \sum_j f(\mu_j) + \sum_j f'(\mu_j)(\lambda_j - \mu_j) + \sum_j f(|\lambda_j - \mu_j|),$$

which is equivalent to write

$$\text{Tr}(f(A)) \geq \text{Tr}(f(B)) + \text{Tr}((A - B)f'(B)) + \text{Tr}(f(|A - B|)),$$

and this is exactly (2.2). \square

For the choice of the superquadratic function $f(t) = t^2 \log(t)$ ($t > 0$), one could introduce a new refinement of the relative entropy defined above; which it has important applications in statistical mechanics. Similarly, one could refine the Peierl–Bogoliubov inequality using (2.3). For more details see [17], [18] and the references therein. We left the details to the interested reader.

A generalization of Lemma 3 could be stated as follows:

Theorem 2. Let $w_k \geq 0$ ($k = 1, \dots, n$) be positive scalars such that $W_n = \sum_{j=1}^n w_k$. Let f be a real-valued continuous function defined on an interval I and let m and n be natural numbers. If f is superquadratic function (in ordinary sense), then the inequality

$$(2.4) \quad \text{Tr} \left(f \left(\frac{1}{W_n} \sum_{k=1}^n w_k A_k \right) \right) \leq \frac{1}{W_n} \sum_{k=1}^n w_k \text{Tr}(f(A_k)) - \frac{1}{W_n} \sum_{k=1}^n w_k \text{Tr} \left(f \left(\left| A_k - \frac{1}{W_n} \sum_{l=1}^n w_l \text{Tr}(A_l) \right| \right) \right)$$

holds for every n -tuple (A_1, \dots, A_n) of positive $m \times m$ matrices with spectra contained in I .

In particular, we have

$$(2.5) \quad \text{Tr} \left(f \left(\frac{1}{n} \sum_{k=1}^n A_k \right) \right) \leq \frac{1}{n} \sum_{k=1}^n \text{Tr}(f(A_k)) - \frac{1}{n} \sum_{k=1}^n \text{Tr} \left(f \left(\left| A_k - \frac{1}{n} \sum_{l=1}^n \text{Tr}(A_l) \right| \right) \right)$$

Proof. Since f is superquadratic then the inequality

$$(2.6) \quad f \left(\frac{1}{W_n} \sum_{k=1}^n w_k a_k \right) \leq \sum_{k=1}^n \frac{w_k}{W_n} f(a_k) - \sum_{k=1}^n \frac{w_k}{W_n} f \left(\left| a_k - \sum_{l=1}^n \frac{w_l}{W_n} a_l \right| \right),$$

holds for every finite positive sequence of real numbers a_k ($k = 1, \dots, n$) and every positive scalars w_k such that $W_n = \sum_{j=1}^n w_k$, (see [2]).

Now, let $\lambda_1^{(j)}, \dots, \lambda_m^{(j)}$ be the eigenvalues of A_j ($1 \leq j \leq n$); respectively. Then, on utilizing the above inequality we have

$$\begin{aligned} \operatorname{Tr} \left(f \left(\frac{1}{W_n} \sum_{k=1}^n w_k A_k \right) \right) &= f \left(\frac{1}{W_n} \sum_{k=1}^n w_k \left(\sum_{j=1}^m \lambda_k^{(j)} \right) \right) \\ &\leq \sum_{k=1}^n \frac{w_k}{W_n} f \left(\sum_{j=1}^m \lambda_k^{(j)} \right) - \sum_{k=1}^n \frac{w_k}{W_n} f \left(\left| \sum_{j=1}^m \lambda_k^{(j)} - \sum_{l=1}^n \frac{w_l}{W_n} \sum_{j=1}^m \lambda_l^{(j)} \right| \right) \\ &= \operatorname{Tr} \left(\sum_{k=1}^n \frac{w_k}{W_n} f(A_k) \right) - \operatorname{Tr} \left(\sum_{k=1}^n \frac{w_k}{W_n} f \left(\left| A_k - \operatorname{Tr} \left(\sum_{l=1}^n \frac{w_l}{W_n} A_l \right) \right| \right) \right) \\ &= \frac{1}{W_n} \sum_{k=1}^n w_k \operatorname{Tr} (f(A_k)) - \frac{1}{W_n} \sum_{k=1}^n w_k \operatorname{Tr} \left(f \left(\left| A_k - \frac{1}{W_n} \sum_{l=1}^n w_l \operatorname{Tr}(A_l) \right| \right) \right) \end{aligned}$$

which gives the required result. The particular case follows by setting $w_k = 1$ for all $k = 1, \dots, n$ so that $W_n = n$. \square

Corollary 1. Let $w_k \geq 0$ ($k = 1, \dots, n$) be positive scalars such that $W_n = \sum_{j=1}^n w_k$. For every n -tuple (A_1, \dots, A_n) of positive $m \times m$ matrices with spectra contained in $[0, \infty)$. Then the inequality

$$(2.7) \quad \operatorname{Tr} \left(\left(\frac{1}{W_n} \sum_{k=1}^n w_k A_k \right)^p \right) \leq \frac{1}{W_n} \sum_{k=1}^n w_k \operatorname{Tr} (A_k^p) - \frac{1}{W_n} \sum_{k=1}^n w_k \operatorname{Tr} \left(\left| A_k - \frac{1}{W_n} \sum_{l=1}^n w_l \operatorname{Tr}(A_l) \right|^p \right)$$

holds for all $p \geq 2$. In particular, we have

$$(2.8) \quad \operatorname{Tr} \left(\left(\frac{1}{n} \sum_{k=1}^n A_k \right)^p \right) \leq \frac{1}{n} \sum_{k=1}^n \operatorname{Tr} (A_k^p) - \frac{1}{n} \sum_{k=1}^n \operatorname{Tr} \left(\left| A_k - \frac{1}{n} \sum_{l=1}^n \operatorname{Tr}(A_l) \right|^p \right)$$

In 2003, Hansen & Pedersen [11], proved the following non-commutative trace version of Jensen's inequality

$$\operatorname{Tr} \left(f \left(\sum_{k=1}^n C_k^* A_k C_k \right) \right) \leq \operatorname{Tr} \left(\sum_{k=1}^n C_k^* f(A_k) C_k \right)$$

for every n -tuple (A_1, \dots, A_n) of positive $m \times m$ matrices with spectra contained in I and every n -tuple (C_1, \dots, C_n) of $m \times m$ matrices with $\sum_{k=1}^n C_k^* C_k = 1$, where f is assumed to be convex on I .

Using the concept of superquadratic functions, one could give the following refinement of Hansen–Pedersen trace inequality.

Theorem 3. Let f be a real-valued continuous function defined on an interval I and let m and n be natural numbers. If f is superquadratic function (in ordinary sense), then the inequality

$$(2.9) \quad \operatorname{Tr} \left(f \left(\sum_{k=1}^n C_k^* A_k C_k \right) \right) \leq \operatorname{Tr} \left(\sum_{k=1}^n C_k^* f(A_k) C_k \right) - \operatorname{Tr} \left(\sum_{k=1}^n C_k^* f \left(\left| A_k - \operatorname{Tr} \left(\sum_{j=1}^n C_j^* A_j C_j \right) \right| \right) C_k \right)$$

holds for every n -tuple (A_1, \dots, A_n) of positive $m \times m$ matrices with spectra contained in I and every n -tuple (C_1, \dots, C_n) of $m \times m$ matrices with $\sum_{k=1}^n C_k^* C_k = 1$. Conversely, if the inequality (2.9) is satisfied for some n and m , where $n > 1$, then f is superquadratic function. If f is subquadratic, then the inequality (2.9) is reversed.

Proof. Our proof is motivated by [11]. Let $A_k = \sum_{\text{sp}(A_k)} \lambda E_k(\lambda)$ denote the spectral resolution of A_k for $1 \leq k \leq n$. Thus, $E_k(\lambda)$ is the spectral projection of A_k on the eigenspace corresponding to λ ; if λ is an eigenvalue for A_k , otherwise $E_k(\lambda) = 0$. For each unit vector ξ in \mathbb{C}^m . Define the probability measure

$$\mu_\xi(S) = \left\langle \sum_{k=1}^n C_k^* E_k(S) C_k \xi, \xi \right\rangle = \sum_{k=1}^n \langle E_k(S) C_k \xi, C_k \xi \rangle$$

for any (Borel) set S in \mathbb{R} . Note that if $y = \sum_{k=1}^n C_k^* A_k C_k$, then

$$\begin{aligned} \langle y\xi, \xi \rangle &= \left\langle \sum_{k=1}^n C_k^* A_k C_k \xi, \xi \right\rangle \\ &= \left\langle \sum_{k=1}^n \sum_{\text{sp}(x_k)} \lambda E_k(\lambda) C_k \xi, C_k \xi \right\rangle \\ &= \int \lambda d\mu_\xi(\lambda). \end{aligned}$$

If a unit vector ξ is an eigenvector for y , then the corresponding eigenvalue is $\langle y\xi, \xi \rangle$, and ξ is also an eigenvector for $f(y)$ with corresponding eigenvalue $\langle f(y)\xi, \xi \rangle = f(\langle y\xi, \xi \rangle)$. In this case we have

$$\begin{aligned} \left\langle f\left(\sum_{k=1}^n C_k^* A_k C_k\right) \xi, \xi \right\rangle &= \langle f(y)\xi, \xi \rangle \\ &= f(\langle y\xi, \xi \rangle) \\ &= f\left(\int \lambda d\mu_\xi(\lambda)\right) \\ &\leq \int \left[f(\lambda) - f\left(\left|\lambda - \int \lambda d\mu_\xi(\lambda)\right|\right) \right] d\mu_\xi(\lambda) \quad (\text{by 1.4}) \\ &= \sum_{k=1}^n \left\langle \sum_{\text{sp}(x_k)} f(\lambda) - f\left(\left|\lambda - \int \lambda d\mu_\xi(\lambda)\right|\right) E_k(\lambda) C_k \xi, C_k \xi \right\rangle \\ &= \sum_{k=1}^n \langle [C_k^* f(A_k) C_k - C_k^* f(|A_k - \langle y\xi, \xi \rangle)|] C_k \xi, \xi \rangle \\ &= \sum_{k=1}^n \left\langle \left[C_k^* f(A_k) C_k - C_k^* f\left(\left|A_k - \left\langle \sum_{j=1}^n C_j^* A_j C_j \xi, \xi \right\rangle\right|\right) \right] C_k \right\rangle \xi, \xi \right\rangle. \end{aligned}$$

Summing over an orthonormal basis of eigenvectors for y we get the desired result in (2.9). \square

Corollary 2. Let f be a real-valued continuous function defined on an interval I and let m and n be natural numbers. If f is superquadratic function (in the ordinary sense), then the inequality

$$(2.10) \quad \text{Tr}(f(C^*AC)) \leq \text{Tr}(C^*f(A)C) - \text{Tr}(C^*f(|A - \text{Tr}(C^*AC)|)C),$$

holds for every positive $m \times m$ matrix A with spectrum contained in I and every $m \times m$ matrix C with $C^*C = 1$. If f is subquadratic, then the inequality (2.10) is reversed. Furthermore, we have

$$\text{Tr}((C^*AC)^p) \leq \text{Tr}(C^*A^pC) - \text{Tr}(C^*|A - \text{Tr}(C^*AC)|^pC)$$

for every $p \geq 2$, and

$$\text{Tr}((C^*AC)^p) \geq \text{Tr}(C^*A^pC) - \text{Tr}(C^*|A - \text{Tr}(C^*AC)|^pC)$$

for every $p \in (0, 2]$.

Proof. The result follows by setting $n = 1$ in Theorem 3. The second inequality follows by applying the superquadratic function $f(t) = t^p$, $p \geq 2$. Similarly, the last inequality follows by applying the subquadratic function $f(t) = t^p$, $p \in (0, 2]$. \square

The inequality (2.10) could be extended for general positive Hilbert space operators mapped under positive unital linear map, as follows:

Theorem 4. Let f be a real-valued continuous function defined on $[0, \infty)$. Let $A \in \mathcal{B}(\mathcal{H})$ be a positive operator and $\Phi : \mathcal{B}(\mathcal{H}) \rightarrow \mathcal{B}(\mathcal{K})$ be a positive unital linear map. If f is superquadratic, then we have

$$(2.11) \quad \text{Tr}(\Phi(f(A))) \geq \text{Tr}(f(\Phi(A))) + \text{Tr}(\Phi(f(|A - \text{Tr}(\Phi(A))|))).$$

In particular case, for the choice $\Phi(A) = C^*AC$, where $C \in \mathcal{B}(\mathcal{H})$ such that $C^*C = 1_{\mathcal{H}}$, we get the inequality (2.10).

Proof. Let $A \in \mathcal{B}(\mathcal{H})$ be positive. Assume that \mathcal{A} is the C^* -subalgebra of $\mathcal{B}(\mathcal{H})$ generated by A and $1_{\mathcal{H}}$. Without loss of generality, we may assume that Φ is defined on \mathcal{A} . Since every unital positive linear map on a commutative C^* -algebra is completely positive. It follows that Φ is completely positive. So there exists (by Stinespring's theorem [20]), some isometry $V : \mathcal{H} \rightarrow \mathcal{K}$; and a unital $*$ -homomorphism ρ from \mathcal{A} into the C^* -algebra $\mathcal{B}(\mathcal{H})$ such that $\Phi(A) = V^*\rho(A)V$. Clearly, $f(\rho(A)) = \rho(f(A))$, for all continuous function f .

Now, let $\{e_i : i \in I\}$ be a set of an orthonormal basis of a Hilbert space \mathcal{H} . On utilizing the continuous functional calculus for the operator $A \geq 0$ Thus,

$$\begin{aligned} \text{Tr}(f(\Phi(A))) &= \text{Tr}(f(V^*\rho(A)V)) \\ &\leq \text{Tr}(V^*f(\rho(A))V) - \text{Tr}(V^*f(|\rho(A) - V^*\rho(A)V|)V) \quad (\text{by (2.10)}) \\ &= \text{Tr}(V^*\rho(f(A))V) - \text{Tr}(V^*\rho(f(|A - \Phi(A)|))V) \\ &= \text{Tr}(\Phi(f(A))) - \text{Tr}(\Phi(f(|A - \Phi(A)|))). \end{aligned}$$

which proves the required inequality. \square

Corollary 3. Let $A \in \mathcal{B}(\mathcal{H})$ be a positive operator and $\Phi : \mathcal{B}(\mathcal{H}) \rightarrow \mathcal{B}(\mathcal{K})$ be a positive unital linear map. Then we have

$$\text{Tr}(\Phi(A^p)) \geq \text{Tr}(\Phi^p(A)) + \text{Tr}(\Phi(|A - \text{Tr}(\Phi(A))|^p)).$$

for all $p \geq 2$. The inequality is reversed for $p \in (0, 2]$.

One can easily generalized (2.11) by using Theorem 4, as follows:

Corollary 4. Let f be a real-valued continuous function defined on $[0, \infty)$. Let $A_j \in \mathcal{B}(\mathcal{H})$ ($j = 1, \dots, n$) be positive operators. Let $\Phi_j : \mathcal{B}(\mathcal{H}) \rightarrow \mathcal{B}(\mathcal{K})$ ($j = 1, \dots, n$) be a positive linear map, such that $\sum_{j=1}^n \Phi_j(1_{\mathcal{H}}) = 1_{\mathcal{K}}$. If f is superquadratic function, then

$$\text{Tr} \left(\sum_{j=1}^n \Phi_j(f(A_j)) \right) \geq \text{Tr} \left(f \left(\sum_{j=1}^n \Phi_j(A_j) \right) \right) + \text{Tr} \left(\sum_{j=1}^n \Phi_j \left(f \left(\left| A_j - \text{Tr} \left(\sum_{j=1}^n \Phi_j(A_j) \right) \right| \right) \right) \right).$$

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