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CHARACTERIZATIONS OF TRACIAL PROPERTY VIA INEQUALITIES

TAKASHI SANO AND TAKESHI YATSU

DEPARTMENT OF MATHEMATICAL SCIENCES

FACULTY OF SCIENCE

YAMAGATA UNIVERSITY

YAMAGATA 990-8560, JAPAN.

sano@sci.kj.yamagata-u.ac.jp

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ABSTRACT. In this article, we give characterizations of a tracial property for a positive linear functional via inequalities; we have necessary and sufficient conditions for a faithful positive linear functional φ to be a positive scalar multiple of the trace by inequalities: for a non matrix monotone, increasing function f ,

$$X \leq Y \Rightarrow \varphi(f(X)) \leq \varphi(f(Y))$$

is considered. Also for a non matrix convex, convex function f ,

$$\varphi\left(f\left(\frac{X+Y}{2}\right)\right) \leq \varphi\left(\frac{f(X)+f(Y)}{2}\right)$$

is studied. We also show that suppose

$$0 \leq \varphi(p_{m,k}(X, Y))$$

for all $X, Y \geq O$, then φ should be a positive scalar multiple of the trace. Here, $p_{m,k}(X, Y)$ is the coefficient of t^k in the polynomial $(X + tY)^m$ and $1 \leq k \leq m - 1$.

Key words and phrases: Trace; Inequality; Non matrix monotone function of order 2; Non matrix convex function of order 2; Bessis-Moussa-Villani conjecture.

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1. INTRODUCTION

In operator theory, matrix monotone functions and matrix convex ones have played a significant role, for instance, see [3, 4, 1, 2]. A real-valued continuous function f on an interval I

($\subseteq \mathbb{R}$) is called matrix monotone of order n if $X \leq Y$ implies $f(X) \leq f(Y)$ for all $n \times n$ Hermitian matrices X and Y with eigenvalues in I . If f is matrix monotone of all orders, f is said to be matrix monotone or operator monotone. When f is matrix monotone of order n , for a positive linear functional φ on $n \times n$ matrices, we have

$$X \leq Y \Rightarrow \varphi(f(X)) \leq \varphi(f(Y))$$

for $n \times n$ Hermitian matrices X and Y . Also, for an increasing function f and Hermitian matrices X and Y with $X \leq Y$,

$$\text{Tr}(f(X)) \leq \text{Tr}(f(Y))$$

holds in which Tr is the standard trace on matrices (for more details, see the argument at the beginning of Section 2).

A real-valued continuous function f on an interval I ($\subseteq \mathbb{R}$) is called matrix convex of order n if the inequality

$$f\left(\frac{X+Y}{2}\right) \leq \frac{f(X)+f(Y)}{2}$$

is satisfied for all $n \times n$ Hermitian matrices X and Y with eigenvalues in I . If f is matrix convex of all orders, f is said to be matrix convex or operator convex. When f is matrix convex of order n , for a positive linear functional φ on $n \times n$ matrices,

$$\varphi\left(f\left(\frac{X+Y}{2}\right)\right) \leq \varphi\left(\frac{f(X)+f(Y)}{2}\right)$$

holds for $n \times n$ Hermitian matrices X and Y with eigenvalues in I . And for a convex function f and Hermitian matrices X and Y , we have

$$\text{Tr}\left(f\left(\frac{X+Y}{2}\right)\right) \leq \text{Tr}\left(\frac{f(X)+f(Y)}{2}\right)$$

(see basic facts on Jensen's inequalities explained before the proof of Theorem 3.3).

In this article, we give characterizations of tracial properties for positive linear functionals via inequalities; we have necessary and sufficient conditions for a faithful positive linear functional φ to be tracial by inequalities: for a non matrix monotone, increasing function f ,

$$X \leq Y \Rightarrow \varphi(f(X)) \leq \varphi(f(Y))$$

is considered. Also for a non matrix convex, convex function f ,

$$\varphi\left(f\left(\frac{X+Y}{2}\right)\right) \leq \varphi\left(\frac{f(X)+f(Y)}{2}\right)$$

is studied. We have a criterion of non matrix monotonicity of order 2 or non matrix convexity of order 2. We show a necessary and sufficient condition for the function

$$X \mapsto \{\text{Tr}(|X|^p C)\}^{\frac{1}{p}}$$

($p > 2$) to be a norm; the function is essentially the Schatten p -norm.

We also observe an inequality given by a coefficient of a certain polynomial: let $p_{m,k}(X, Y)$ be the coefficient of t^k in the polynomial $(X + tY)^m$ for $X, Y \in M_n(\mathbb{C})$, $m \in \mathbb{N}$, and $t \in \mathbb{C}$ and $1 \leq k \leq m - 1$. Suppose that

$$0 \leq \varphi(p_{m,k}(X, Y))$$

for all $X, Y \geq O$. Then φ should be a positive scalar multiple of the trace (see the remark of Proposition 3.1 about the BMV conjecture).

We remark that divided differences are useful in this article: we refer the reader to [4, 2, 6].

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2. INEQUALITIES OF NON MATRIX MONOTONE FUNCTIONS

Let $M_n(\mathbb{C})$ be the set of all complex n -square matrices and let φ be a faithful positive linear functional on $M_n(\mathbb{C})$. Let f be an increasing function on $I = (a, b)$. For Hermitian matrices $X, Y \in M_n(\mathbb{C})$ with $a1 < X \leq Y < b1$,

$$\lambda_i(X) \leq \lambda_i(Y)$$

for $i = 1, 2, \dots, n$ in which λ_i is the i -th eigenvalue with $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$. Since f is increasing,

$$\lambda_i(f(X)) = f(\lambda_i(X)) \leq f(\lambda_i(Y)) = \lambda_i(f(Y)).$$

Hence, it follows that

$$\text{Tr}(f(X)) = \sum_{i=1}^n \lambda_i(f(X)) \leq \sum_{i=1}^n \lambda_i(f(Y)) = \text{Tr}(f(Y)).$$

Let us study the following inequality for a strictly increasing, differentiable function f on $I = (a, b)$ and $\varepsilon \in [0, 1]$:

$$f'(\lambda)\alpha^2(1 + \varepsilon) - 2\alpha\sqrt{1 - \alpha^2} \frac{f(\lambda) - f(\mu)}{\lambda - \mu}(1 - \varepsilon) + f'(\mu)(1 - \alpha^2)(1 + \varepsilon) \geq 0$$

for all $\mu, \lambda \in I$ ($\mu < \lambda$) and all $\alpha \in [0, 1]$. By considering $0 < \alpha < 1$, we have the equivalent inequality

$$\frac{\alpha}{\sqrt{1 - \alpha^2}} f'(\lambda) + \frac{\sqrt{1 - \alpha^2}}{\alpha} f'(\mu) \geq 2 \frac{1 - \varepsilon}{1 + \varepsilon} \cdot \frac{f(\lambda) - f(\mu)}{\lambda - \mu}$$

for all $\mu, \lambda \in I$ ($\mu < \lambda$) and all $\alpha \in (0, 1)$. Let

$$t := \frac{\alpha}{\sqrt{1 - \alpha^2}}, \quad \delta := \frac{1 - \varepsilon}{1 + \varepsilon}.$$

Then notice that

$$0 < \alpha < 1 \Leftrightarrow 0 < t < \infty, \quad 0 \leq \varepsilon \leq 1 \Leftrightarrow 0 \leq \delta \leq 1,$$

and $\varepsilon = 1$ if and only if $\delta = 0$. The corresponding inequality is described as

$$\frac{1}{2} \left(t f'(\lambda) + \frac{1}{t} f'(\mu) \right) \geq \delta \frac{f(\lambda) - f(\mu)}{\lambda - \mu}$$

for all $0 < t < \infty$ and all $\mu, \lambda \in I$ ($\mu < \lambda$). Hence, by considering arithmetic-geometric mean inequality in the left-hand side, we have

$$\sqrt{f'(\lambda)f'(\mu)} \geq \delta \frac{f(\lambda) - f(\mu)}{\lambda - \mu}.$$

In this case, the condition $\varepsilon = 1$ or $\delta = 0$ is given by

$$\inf_{\lambda > \mu} \frac{\sqrt{f'(\lambda)f'(\mu)}}{\frac{f(\lambda) - f(\mu)}{\lambda - \mu}} = 0.$$

We summarize our observation as follows:

Proposition 2.1. *Let f be a strictly increasing, continuously differentiable function on $I = (a, b)$ and $\varepsilon \in [0, 1]$. Suppose that*

$$\inf_{\lambda > \mu} \frac{\sqrt{f'(\lambda)f'(\mu)}}{\frac{f(\lambda) - f(\mu)}{\lambda - \mu}} = 0.$$

Then the inequality

$$f'(\lambda)\alpha^2(1 + \varepsilon) - 2\alpha\sqrt{1 - \alpha^2} \frac{f(\lambda) - f(\mu)}{\lambda - \mu}(1 - \varepsilon) + f'(\mu)(1 - \alpha^2)(1 + \varepsilon) \geq 0$$

holds for all $\mu, \lambda \in I$ ($\mu < \lambda$) and all $\alpha \in [0, 1]$ if and only if $\varepsilon = 1$.

The following are examples:

$$\begin{array}{ll} x^p (p > 1) \text{ on } (0, a), & x^p (p > 1) \text{ on } (a, \infty), \\ e^x \text{ on } (a, \infty), & e^x \text{ on } (-\infty, a) \end{array}$$

for a constant a .

By direct computations, it is easy to see that each example satisfies the condition so details are left to the reader.

Theorem 2.2. *Let φ be a faithful positive linear functional on $M_n(\mathbb{C})$ and let f be a function as in Proposition 2.1. Then*

$$(2.1) \quad \varphi(f(X)) \leq \varphi(f(Y)) \quad \text{whenever} \quad aI < X \leq Y < bI$$

if and only if φ is a positive scalar multiple of the trace.

Proof. At the beginning of this section it was explained that if φ is a positive scalar multiple of the trace then the inequality (2.1) holds. We show the converse: since there is uniquely a positive definite matrix D such that $\varphi(X) = \text{Tr}(XD)$ for $X \in M_n(\mathbb{C})$, we have to prove that D is a positive scalar multiple of the identity matrix. Taking into consideration

$$\varphi(V^* \cdot V) = \text{Tr}(\cdot VDV^*)$$

for all unitary V and that VDV^* is diagonal for a unitary V , we assume that D is a diagonal matrix $\text{diag}(d_1, \dots, d_n)$. To show that $d_i = d_j$ for any pair of d_i, d_j ($i \neq j$), we consider matrices $X = (x_{kl})$ with x_{kl} zero except for $(k, l) = (i, i), (i, j), (j, i), (j, j)$. Hence, it suffices to consider the case $n = 2$ so that we suppose

$$D = \text{diag}(\varepsilon, 1)$$

for a number ε ($0 < \varepsilon \leq 1$). We show that $\varepsilon = 1$.

Let

$$U = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}, \quad A_{\lambda, \mu} = \begin{pmatrix} \lambda & 0 \\ 0 & \mu \end{pmatrix}, \quad P_\alpha = \begin{pmatrix} \alpha^2 & \alpha\sqrt{1-\alpha^2} \\ \alpha\sqrt{1-\alpha^2} & 1-\alpha^2 \end{pmatrix}$$

for λ, μ ($\lambda \neq \mu, a < \lambda, \mu < b$), α ($0 \leq \alpha \leq 1$). For all $t > 0$,

$$UA_{\lambda, \mu}U^* \leq U(A_{\lambda, \mu} + tP_\alpha)U^*,$$

and $a1 < A_{\lambda, \mu} + tP_\alpha < b1$ for small $t > 0$. Then, by assumption

$$\text{Tr}(Uf(A_{\lambda, \mu})U^*D) \leq \text{Tr}(Uf(A_{\lambda, \mu} + tP_\alpha)U^*D).$$

This implies that

$$\left. \frac{d}{dt} \text{Tr}(Uf(A_{\lambda, \mu} + tP_\alpha)U^*D) \right|_{t=0} \geq 0.$$

Also, the standard fact, see [2, page 124] for instance, yields

$$\begin{aligned} \left. \frac{d}{dt} \right|_{t=0} f(A_{\lambda, \mu} + tP_\alpha) &= \begin{pmatrix} f'(\lambda) & \frac{f(\lambda)-f(\mu)}{\lambda-\mu} \\ \frac{f(\lambda)-f(\mu)}{\lambda-\mu} & f'(\mu) \end{pmatrix} \circ P_\alpha \\ &= \begin{pmatrix} f'(\lambda)\alpha^2 & \frac{f(\lambda)-f(\mu)}{\lambda-\mu}\alpha\sqrt{1-\alpha^2} \\ \frac{f(\lambda)-f(\mu)}{\lambda-\mu}\alpha\sqrt{1-\alpha^2} & f'(\mu)(1-\alpha^2) \end{pmatrix}, \end{aligned}$$

where \circ stands for the Hadamard (i.e., entry-wise) product. Hence, it follows that

$$\begin{aligned} 0 &\leq \left. \frac{d}{dt} \text{Tr}(f(A_{\lambda, \mu} + tP_\alpha)U^*DU) \right|_{t=0} \\ &= \text{Tr} \left(\left. \frac{d}{dt} \right|_{t=0} f(A_{\lambda, \mu} + tP_\alpha) \cdot U^*DU \right) \\ &= \text{Tr} \left(\begin{pmatrix} f'(\lambda)\alpha^2 & \frac{f(\lambda)-f(\mu)}{\lambda-\mu}\alpha\sqrt{1-\alpha^2} \\ \frac{f(\lambda)-f(\mu)}{\lambda-\mu}\alpha\sqrt{1-\alpha^2} & f'(\mu)(1-\alpha^2) \end{pmatrix} \cdot \frac{1}{2} \begin{pmatrix} 1+\varepsilon & -(1-\varepsilon) \\ -(1-\varepsilon) & 1+\varepsilon \end{pmatrix} \right) \\ &= \frac{1}{2} \left\{ f'(\lambda)\alpha^2(1+\varepsilon) - 2\alpha\sqrt{1-\alpha^2} \frac{f(\lambda)-f(\mu)}{\lambda-\mu}(1-\varepsilon) + f'(\mu)(1-\alpha^2)(1+\varepsilon) \right\} \end{aligned}$$

for all α, λ, μ . Therefore, thanks to Proposition 2.1, the proof is completed. \square

In the proof of Theorem 2.2, a criterion of non matrix monotonicity of order 2 is obtained:

Corollary 2.3. *Let f be a function as in Proposition 2.1. Then f is not matrix monotone of order 2.*

As a corollary of Theorem 2.2, we have

Theorem 2.4. *Let $C \in M_n(\mathbb{C})$ be a positive definite matrix and let $p > 2$ be given. Then the function*

$$\mu(X) := \text{Tr}(|X|^p C)^{\frac{1}{p}} \quad (X \in M_n(\mathbb{C}))$$

is a norm if and only if C is a positive scalar multiple of the identity matrix.

Proof. Suppose that μ is a norm. Then, by definition

$$\mu(UX) = \mu(X)$$

for all unitary U . For positive semidefinite matrices X, Y with $X \leq Y$, there is a contraction V such that $X^{\frac{1}{2}} = VY^{\frac{1}{2}}$ and V is a convex combination of unitary matrices: $V = \sum_{i=1}^N \lambda_i U_i$, where $\lambda_i \geq 0, U_j$ is unitary ($j = 1, 2, \dots, N$) and $\sum_{i=1}^N \lambda_i = 1$. Hence, we have

$$\mu\left(X^{\frac{1}{2}}\right) = \mu\left(\left(\sum_{i=1}^N \lambda_i U_i\right) Y^{\frac{1}{2}}\right) \leq \sum_{i=1}^N \lambda_i \mu\left(U_i Y^{\frac{1}{2}}\right) = \mu\left(Y^{\frac{1}{2}}\right)$$

since μ is a norm. Therefore, the faithful positive linear functional on $M_n(\mathbb{C})$ defined by

$$\varphi(\cdot) := \text{Tr}(\cdot C)$$

satisfies

$$0 \leq X \leq Y \implies \varphi\left(X^{\frac{p}{2}}\right) \leq \varphi\left(Y^{\frac{p}{2}}\right).$$

Since $\frac{p}{2} > 1$, it follows from Theorem 2.2 that C is a scalar multiple of the identity matrix; the proof is completed. \square

3. INEQUALITIES OF NON MATRIX CONVEX FUNCTIONS

Let us start this section with the following observation of tracial properties:

Proposition 3.1. *Let φ be a faithful positive linear functional on $M_n(\mathbb{C})$. Let $p_{m,k}(X, Y)$ be the coefficient of t^k in the polynomial $(X + tY)^m$ for $X, Y \in M_n(\mathbb{C})$, $m \in \mathbb{N}$, and $t \in \mathbb{C}$ and $1 \leq k \leq m - 1$. Suppose that*

$$\varphi(p_{m,k}(X, Y)) \geq 0$$

for all $X, Y \geq 0$. Then φ should be a positive scalar multiple of the trace.

Remark that the non-negativity of $\text{Tr}(p_{m,k}(X, Y))$ for all positive matrices X, Y is (equivalent to) the Bessis-Moussa-Villani conjecture (see [9, 7] and also [6]); it is known that it is the case if one of the following is satisfied:

- (1) $k \leq 2$ (or $m \leq 5$),
- (2) $n = 2$ (see Fact 5 in [8]),
- (3) $n = 3, k = 6$ (see [7]).

Proof. Due to the same argument for $\varphi = \text{Tr}(\cdot D)$ as in the proof of Theorem 2.2, it suffices to consider the case $n = 2$ so that we suppose

$$D = \text{diag}(\varepsilon, 1)$$

for a number ε ($0 < \varepsilon \leq 1$). We show that $\varepsilon = 1$.

At first consider the case $k \geq 2$; let

$$U = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}, \quad A = \begin{pmatrix} \alpha^2 & \alpha\sqrt{1-\alpha^2} \\ \alpha\sqrt{1-\alpha^2} & 1-\alpha^2 \end{pmatrix},$$

$$B = \begin{pmatrix} 1 & 0 \\ 0 & \alpha \end{pmatrix}$$

for a number α ($0 < \alpha < 1$). Since $A^2 = A$, $p_{m,k}(A, B)$ is described as

$$AB^k + B^k A + \text{the terms including } BAB + \text{the terms } AB^k A.$$

Notice that for a number λ , $A \begin{pmatrix} 1 & 0 \\ 0 & \lambda \end{pmatrix} A$ is

$$\begin{pmatrix} \alpha^4 + \alpha^2(1-\alpha^2)\lambda & \alpha^3\sqrt{1-\alpha^2} + \alpha(1-\alpha^2)\sqrt{1-\alpha^2}\lambda \\ \alpha^3\sqrt{1-\alpha^2} + \alpha(1-\alpha^2)\sqrt{1-\alpha^2}\lambda & \alpha^2(1-\alpha^2) + (1-\alpha^2)^2\lambda \end{pmatrix},$$

$$\begin{pmatrix} 1 & 0 \\ 0 & \lambda \end{pmatrix} A \begin{pmatrix} 1 & 0 \\ 0 & \lambda \end{pmatrix} = \begin{pmatrix} \alpha^2 & \alpha\lambda\sqrt{1-\alpha^2} \\ \alpha\lambda\sqrt{1-\alpha^2} & \lambda^2(1-\alpha^2) \end{pmatrix},$$

and

$$AB^k = \begin{pmatrix} \alpha^2 & \alpha^{k+1}\sqrt{1-\alpha^2} \\ \alpha\sqrt{1-\alpha^2} & \alpha^k(1-\alpha^2) \end{pmatrix}.$$

Thus, we have for $l \geq 2$

$$AB^l A = o(\alpha) \quad (\alpha \rightarrow 0), \quad BAB = o(\alpha) \quad (\alpha \rightarrow 0),$$

where o is Landau's small o , and

$$\begin{aligned} 0 &\leq \varphi(p_{m,k}(UAU^*, UBU^*)) \\ &= \varphi(U p_{m,k}(A, B) U^*) \\ &= \varphi(U \{AB^k + B^k A + o(\alpha)\} U^*) \\ &= \alpha \varphi \left(U \left\{ \frac{1}{\alpha} (AB^k + B^k A) + o(1) \right\} U^* \right) \quad (\alpha \rightarrow 0). \end{aligned}$$

Dividing this inequality by $\alpha > 0$ and taking α as $\alpha \rightarrow 0$, we have

$$\begin{aligned} 0 &\leq \varphi \left(U \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} U^* \right) \\ &= \varphi \left(\begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \right) \\ &= \text{Tr} \left(\begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix} \begin{pmatrix} \varepsilon & 0 \\ 0 & 1 \end{pmatrix} \right) = \varepsilon - 1, \end{aligned}$$

since

$$\frac{1}{\alpha} AB^k \rightarrow \begin{pmatrix} 0 & 0 \\ 1 & 0 \end{pmatrix} (\alpha \rightarrow 0).$$

Hence, $\varepsilon \geq 1$ or $\varepsilon = 1$.

For the case $k = 1$, let

$$C = \begin{pmatrix} 1 & 0 \\ 0 & \alpha^2 \end{pmatrix} (= B^2).$$

Then $p_{m,1}(A, C)$ is $AC + CA +$ the terms ACA . Notice that $ACA = AB^2A = o(\alpha)$ ($\alpha \rightarrow 0$) and the preceding argument for $k \geq 2$ works similarly. Therefore, the proof is completed. \square

Remark 3.2. It follows from the proof that the inequality assumption for $0 \leq X, Y \leq 1$ is sufficient for the assertion.

Theorem 3.3. Let φ be a faithful positive linear functional on $M_n(\mathbb{C})$ and let f be a twice continuously differentiable convex function on $[0, \infty)$ with

$$f^{[2]}(0, 0, 0) = 0, \quad f^{[2]}(1, 0, 0) > 0,$$

where $f^{[2]}$ is the second divided difference of f . Then

$$(3.1) \quad \varphi \left(\frac{f(X) + f(Y)}{2} \right) \geq \varphi \left(f \left(\frac{X+Y}{2} \right) \right)$$

holds for all $X, Y \geq 0$ if and only if φ is a positive scalar multiple of the trace.

Also, $f(t) = t^p$ ($p > 2$) on $[0, \infty)$ is such an example.

Before giving a proof, let us recall basic facts on matrix convex continuous functions: let f be a matrix convex continuous function of order n on an interval I . Then by definition,

$$\frac{f(X) + f(Y)}{2} \geq f \left(\frac{X+Y}{2} \right)$$

holds for Hermitian matrices $X, Y \in M_n(\mathbb{C})$ with eigenvalues in I . This yields

$$\varphi \left(\frac{f(X) + f(Y)}{2} \right) \geq \varphi \left(f \left(\frac{X+Y}{2} \right) \right)$$

for a positive linear functional φ on $M_n(\mathbb{C})$.

We also recall basic facts on Jensen's inequalities: for a convex continuous function f on I and Hermitian matrices $X, Y \in M_n(\mathbb{C})$ with eigenvalues in I ,

$$(3.2) \quad \text{Tr} \left(\frac{f(X) + f(Y)}{2} \right) \geq \text{Tr} \left(f \left(\frac{X+Y}{2} \right) \right)$$

is satisfied; this inequality is well-known, for instance, see [10, Proposition 3.1]: von Neumann observes the convexity $x \mapsto \text{Tr}(f(x))$. E. H. Lieb gives a description of $\text{Tr}(f(x))$ and B. Simon has further arguments. F. Hansen and G.K. Pedersen study generalizations; Jensen's operator inequality and Jensen's trace inequality. There are also many articles on these kinds of inequalities. See the introduction and references in [5] about Jensen's inequalities.

Proof. We have a proof of Jensen's inequality (3.2) for the reader's convenience: Ky Fan's maximum principle (for instance, see [2, page 35]) means that

$$\sum_{i=1}^k \lambda_i(X) + \sum_{i=1}^k \lambda_i(Y) \geq \sum_{i=1}^k \lambda_i(X + Y)$$

for $k = 1, 2, \dots, n - 1$, and

$$\sum_{i=1}^n \lambda_i(X) + \sum_{i=1}^n \lambda_i(Y) = \sum_{i=1}^n \lambda_i(X + Y),$$

where λ_i is the i -th eigenvalue with $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$. In other words, $\{\lambda_i(\frac{X+Y}{2})\}$ is majorized by $\{\frac{\lambda_i(X)+\lambda_i(Y)}{2}\}$. Then the majorization theory ([2, page 40]) says that

$$\sum_{i=1}^n f\left(\frac{\lambda_i(X) + \lambda_i(Y)}{2}\right) \geq \sum_{i=1}^n f\left(\lambda_i\left(\frac{X + Y}{2}\right)\right).$$

Hence, from the convexity of f in the left-hand side, the above inequality (3.2) for Tr and f follows. Therefore, if φ is a positive scalar multiple of the trace, we have the inequality (3.1).

We show the converse: at first we have explicit calculations for the case $f(t) = t^m$ ($m \in \mathbb{N}, m \geq 3$) although we have a general treatment below: by assumption,

$$\varphi\left(\frac{X^m + (X + tY)^m}{2} - \left(\frac{X + (X + tY)}{2}\right)^m\right) \geq 0$$

for $t > 0$ and $X, Y \geq 0$. Since

$$\begin{aligned} \frac{X^m + (X + tY)^m}{2} - \left(X + \frac{1}{2}tY\right)^m &= \frac{1}{4}p_{m,2}(X, Y)t^2 + o(t^2) \quad (t \rightarrow 0), \\ 0 &\leq t^2\{\varphi(p_{m,2}(X, Y)) + o(1)\} \quad (t \rightarrow 0). \end{aligned}$$

Thus, dividing this inequality by $t^2 > 0$ and taking t as $t \rightarrow 0$, we have

$$\varphi(p_{m,2}(X, Y)) \geq 0.$$

Hence, in this case the assertion follows from Proposition 3.1.

Let us consider the general case: since

$$\varphi\left(\frac{f(X) + f(X + tY)}{2} - f\left(\frac{X + (X + tY)}{2}\right)\right) \geq 0$$

for $t > 0$ and $X, Y \geq 0$, the preceding argument yields similarly

$$\varphi\left(\frac{d^2}{dt^2}\bigg|_{t=0} f(X + tY)\right) \geq 0.$$

For the same matrices A, B, U as in the proof of Proposition 3.1,

$$\frac{1}{2} \frac{d^2}{dt^2} \Big|_{t=0} f(A + tB)$$

is of the form

$$\begin{aligned} & f^{[2]}(1, 1, 1)ABABA + f^{[2]}(1, 1, 0)ABAB(1 - A) \\ & + f^{[2]}(1, 0, 1)AB(1 - A)BA + f^{[2]}(0, 1, 1)(1 - A)BABA \\ & + f^{[2]}(1, 0, 0)AB(1 - A)B(1 - A) + f^{[2]}(0, 0, 1)(1 - A)B(1 - A)BA \\ & + f^{[2]}(0, 1, 0)(1 - A)BAB(1 - A) + f^{[2]}(0, 0, 0)(1 - A)B(1 - A)B(1 - A) \end{aligned}$$

(see the formula of the second divided difference in [2, page 129] and remark that $f^{[2]}$ is symmetric and A is an orthogonal projection: $A = 1 \cdot A + 0 \cdot (1 - A)$). The order estimation for BAB, AB^2A and the assumption $f^{[2]}(0, 0, 0) = 0$ mean

$$\varphi(f^{[2]}(1, 0, 0)(AB^2 + B^2A) + o(\alpha)) \geq 0 \quad (\alpha \rightarrow 0).$$

Hence, replacing A, B with UAU^*, UBU^* , we get

$$\varphi(U\{f^{[2]}(1, 0, 0)(AB^2 + B^2A) + o(\alpha)\}U^*) \geq 0 \quad (\alpha \rightarrow 0).$$

Therefore, as in the proof of Proposition 3.1, the proof is completed. \square

In the proof of Theorem 3.3, a criterion of non matrix convexity of order 2 is obtained:

Corollary 3.4. *Let f be a function as in Theorem 3.3. Then f is not matrix convex of order 2.*

The same argument works for the following theorem whose proof is left to the reader:

Theorem 3.5. *Let φ be a faithful positive linear functional on $M_n(\mathbb{C})$ and let f be a continuously differentiable increasing function on $[0, \infty)$ with*

$$f^{[1]}(0, 0) = 0, \quad f^{[1]}(1, 0) > 0,$$

where $f^{[1]}$ is the first divided difference of f . Then

$$\varphi(f(X)) \leq \varphi(f(Y)) \quad \text{whenever } 0 \leq X \leq Y$$

if and only if φ is a positive scalar multiple of the trace.

$f(t) = t^p$ ($p > 1$) on $[0, \infty)$ is such an example.

We remark that a proof can be obtained by the formula of the first divided difference in [1, p. 12] for

$$\frac{d}{dt} \Big|_{t=0} f(A + tB)$$

as in the proof of Theorem 3.3.

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