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# Uncertainty relations for generalized metric adjusted skew information and generalized metric adjusted correlation measure

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## Abstract

In this paper, we give a Heisenberg-type or a Schrödinger-type uncertainty relation for generalized metric adjusted skew information or generalized metric adjusted correlation measure. These results generalize the previous result of Furuichi and Yanagi (*J. Math. Anal. Appl.* 388:1147-1156, 2012).

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## Introduction

We start from the Heisenberg uncertainty relation [1]:

$$V_{\rho}(A)V_{\rho}(B) \geq \frac{1}{4}|Tr[\rho[A, B]]|^2$$

for a quantum state (density operator)  $\rho$  and two observables (self-adjoint operators)  $A$  and  $B$ . The further stronger result was given by Schrödinger in [2,3]:

$$V_{\rho}(A)V_{\rho}(B) - |Re\{Cov_{\rho}(A, B)\}|^2 \geq \frac{1}{4}|Tr[\rho[A, B]]|^2,$$

where the covariance is defined by  $Cov_{\rho}(A, B) \equiv Tr[\rho(A - Tr[\rho A]I)(B - Tr[\rho B]I)]$ .

The Wigner-Yanase skew information represents a measure for non-commutativity between a quantum state  $\rho$  and an observable  $H$ . Luo introduced the quantity  $U_{\rho}(H)$  representing a quantum uncertainty excluding the classical mixture [4]:

$$U_{\rho}(H) \equiv \sqrt{V_{\rho}(H)^2 - (V_{\rho}(H) - I_{\rho}(H))^2},$$

with the Wigner-Yanase skew information [5]:

$$I_{\rho}(H) \equiv \frac{1}{2}Tr[(i[\rho^{1/2}, H_0])^2] = Tr[\rho H_0^2] - Tr[\rho^{1/2}H_0\rho^{1/2}H_0], \quad H_0 \equiv H - Tr[\rho H]I,$$

and then he successfully showed a new Heisenberg-type uncertainty relation on  $U_{\rho}(H)$  in [4]:

$$U_{\rho}(A)U_{\rho}(B) \geq \frac{1}{4}|Tr[\rho[A, B]]|^2. \quad (1)$$

As stated in [4], the physical meaning of the quantity  $U_\rho(H)$  can be interpreted as follows. For a mixed state  $\rho$ , the variance  $V_\rho(H)$  has both classical mixture and quantum uncertainty. Also, the Wigner-Yanase skew information  $I_\rho(H)$  represents a kind of quantum uncertainty [6,7]. Thus, the difference  $V_\rho(H) - I_\rho(H)$  has a classical mixture so that we can regard that the quantity  $U_\rho(H)$  has a quantum uncertainty excluding a classical mixture. Therefore, it is meaningful and suitable to study an uncertainty relation for a mixed state by the use of the quantity  $U_\rho(H)$ .

Recently, a one-parameter extension of the inequality (1) was given in [8]:

$$U_{\rho,\alpha}(A)U_{\rho,\alpha}(B) \geq \alpha(1 - \alpha)|\text{Tr}[\rho[A, B]]|^2,$$

where

$$U_{\rho,\alpha}(H) \equiv \sqrt{V_\rho(H)^2 - (V_\rho(H) - I_{\rho,\alpha}(H))^2},$$

with the Wigner-Yanase-Dyson skew information  $I_{\rho,\alpha}(H)$  defined by

$$I_{\rho,\alpha}(H) \equiv \frac{1}{2}\text{Tr}[(i[\rho^\alpha, H_0])(i[\rho^{1-\alpha}, H_0])] = \text{Tr}[\rho H_0^2] - \text{Tr}[\rho^\alpha H_0 \rho^{1-\alpha} H_0].$$

It is notable that the convexity of  $I_{\rho,\alpha}(H)$  with respect to  $\rho$  was successfully proven by Lieb in [9]. The further generalization of the Heisenberg-type uncertainty relation on  $U_\rho(H)$  has been given in [10] using the generalized Wigner-Yanase-Dyson skew information introduced in [11]. Recently, it is shown that these skew informations are connected to special choices of quantum Fisher information in [12]. The family of all quantum Fisher informations is parametrized by a certain class of operator monotone functions  $\mathcal{F}_{op}$  which were justified in [13]. The Wigner-Yanase skew information and Wigner-Yanase-Dyson skew information are given by the following operator monotone functions:

$$f_{\text{WY}}(x) = \left(\frac{\sqrt{x} + 1}{2}\right)^2,$$

$$f_{\text{WYD}}(x) = \alpha(1 - \alpha)\frac{(x - 1)^2}{(x^\alpha - 1)(x^{1-\alpha} - 1)}, \quad \alpha \in (0, 1),$$

respectively. In particular, the operator monotonicity of the function  $f_{\text{WYD}}$  was proved in [14] (see also [15]). On the other hand, the uncertainty relation related to the Wigner-Yanase skew information was given by Luo [4], and the uncertainty relation related to the Wigner-Yanase-Dyson skew information was given by Yanagi [8]. In this paper, we generalize these uncertainty relations to the uncertainty relations related to quantum Fisher informations by using (generalized) metric adjusted skew information or correlation measure.

### Operator monotone functions

Let  $M_n(\mathbb{C})$  (respectively  $M_{n,sa}(\mathbb{C})$ ) be the set of all  $n \times n$  complex matrices (respectively all  $n \times n$  self-adjoint matrices), endowed with the Hilbert-Schmidt scalar product  $\langle A, B \rangle = \text{Tr}(A^*B)$ . Let  $M_{n,+}(\mathbb{C})$  be the set of strictly positive elements of  $M_n(\mathbb{C})$  and  $M_{n,+1}(\mathbb{C})$  be the set of strictly positive density matrices, that is  $M_{n,+1}(\mathbb{C}) = \{\rho \in M_n(\mathbb{C}) | \text{Tr}\rho = 1, \rho > 0\}$ . If it is not otherwise specified, from now on, we shall treat the case of faithful states, that is  $\rho > 0$ .

A function  $f : (0, +\infty) \rightarrow \mathbb{R}$  is said to be operator monotone if, for any  $n \in \mathbb{N}$  and  $A, B \in M_{n,+}(\mathbb{C})$  such that  $0 \leq A \leq B$ , the inequalities  $0 \leq f(A) \leq f(B)$  hold. An operator monotone function is said to be symmetric if  $f(x) = xf(x^{-1})$  and normalized if  $f(1) = 1$ .

**Definition 1.**  $\mathcal{F}_{op}$  is the class of functions  $f : (0, +\infty) \rightarrow (0, +\infty)$  such that

1.  $f(1) = 1$ ,
2.  $tf(t^{-1}) = f(t)$ ,
3.  $f$  is operator monotone.

**Example 1.** Examples of elements of  $\mathcal{F}_{op}$  are given by the following list:

$$f_{\text{RLD}}(x) = \frac{2x}{x+1}, \quad f_{\text{WY}}(x) = \left( \frac{\sqrt{x}+1}{2} \right)^2,$$

$$f_{\text{BKM}}(x) = \frac{x-1}{\log x}, \quad f_{\text{SLD}}(x) = \frac{x+1}{2},$$

$$f_{\text{WYD}}(x) = \alpha(1-\alpha) \frac{(x-1)^2}{(x^\alpha-1)(x^{1-\alpha}-1)}, \quad \alpha \in (0, 1).$$

*Remark 1.* Any  $f \in \mathcal{F}_{op}$  satisfies

$$\frac{2x}{x+1} \leq f(x) \leq \frac{x+1}{2}, \quad x > 0.$$

For  $f \in \mathcal{F}_{op}$ , define  $f(0) = \lim_{x \rightarrow 0} f(x)$ . We introduce the sets of regular and non-regular functions

$$\mathcal{F}_{op}^r = \{f \in \mathcal{F}_{op} \mid f(0) \neq 0\}, \quad \mathcal{F}_{op}^n = \{f \in \mathcal{F}_{op} \mid f(0) = 0\}$$

and notice that trivially  $\mathcal{F}_{op} = \mathcal{F}_{op}^r \cup \mathcal{F}_{op}^n$ .

**Definition 2.** For  $f \in \mathcal{F}_{op}^r$ , we set

$$\tilde{f}(x) = \frac{1}{2} \left[ (x+1) - (x-1)^2 \frac{f(0)}{f(x)} \right], \quad x > 0.$$

**Theorem 1.** ([12,16,17]) *The correspondence  $f \rightarrow \tilde{f}$  is a bijection between  $\mathcal{F}_{op}^r$  and  $\mathcal{F}_{op}^n$ .*

### Metric adjusted skew information and correlation measure

In the Kubo-Ando theory of matrix means, one associates a mean to each operator monotone function  $f \in \mathcal{F}_{op}$  by the formula

$$m_f(A, B) = A^{1/2} f(A^{-1/2} B A^{-1/2}) A^{1/2},$$

where  $A, B \in M_{n,+}(\mathbb{C})$ . Using the notion of matrix means, one may define the class of monotone metrics (also called quantum Fisher informations) by the following formula:

$$\langle A, B \rangle_{\rho, f} = \text{Tr}(A \cdot m_f(L_\rho, R_\rho)^{-1}(B)),$$

where  $L_\rho(A) = \rho A, R_\rho(A) = A\rho$ . In this case, one has to think of  $A, B$  as tangent vectors to the manifold  $M_{n,+1}(\mathbb{C})$  at the point  $\rho$  (see [12,13]).

**Definition 3.** For  $A, B \in M_{n,sa}(\mathbb{C})$  and  $\rho \in M_{n,+1}(\mathbb{C})$ , we define the following quantities:

$$\text{Corr}_\rho^f(A, B) = \text{Tr}[\rho AB] - \text{Tr}[A \cdot m_f(L_\rho, R_\rho)B],$$

$$\begin{aligned} \text{Corr}_\rho^{s(f)}(A, B) &= \frac{f(0)}{2} \langle i[\rho, A], i[\rho, B] \rangle_{\rho, f}, \\ I_\rho^f(A) &= \text{Corr}_\rho^f(A, A), \\ C_\rho^f(A, B) &= \text{Tr}[A \cdot m_f(L_\rho, R_\rho)B], \\ C_\rho^f(A) &= C_\rho^f(A, A), \\ U_\rho^f(A) &= \sqrt{V_\rho(A)^2 - (V_\rho(A) - I_\rho^f(A))^2}, \end{aligned}$$

The quantity  $I_\rho^f(A)$  is known as metric adjusted skew information [18], and the metric adjusted correlation measure  $\text{Corr}_\rho^f(A, B)$  was also previously defined in [18].

Then we have the following proposition.

**Proposition 1.** ([16,19]) *For  $A, B \in M_{n,sa}(\mathbb{C})$  and  $\rho \in M_{n,+1}(\mathbb{C})$ , we have the following relations, where we put  $A_0 = A - \text{Tr}[\rho A]I$  and  $B_0 = B - \text{Tr}[\rho B]I$ :*

1.  $I_\rho^f(A) = I_\rho^f(A_0) = \text{Tr}[\rho A_0^2] - \text{Tr}[A_0 \cdot m_{\tilde{f}}(L_\rho, R_\rho)A_0] = V_\rho(A) - C_\rho^{\tilde{f}}(A_0),$
2.  $J_\rho^f(A) = \text{Tr}[\rho A_0^2] + \text{Tr}[A_0 \cdot m_{\tilde{f}}(L_\rho, R_\rho)A_0] = V_\rho(A) + C_\rho^{\tilde{f}}(A_0),$
3.  $0 \leq I_\rho^f(A) \leq U_\rho^f(A) \leq V_\rho(A),$
4.  $U_\rho^f(A) = \sqrt{I_\rho^f(A) \cdot J_\rho^f(A)},$
5.  $\text{Corr}_\rho^f(A, B) = \text{Corr}_\rho^f(A_0, B_0) = \text{Tr}[\rho A_0 B_0] - \text{Tr}[A_0 \cdot m_{\tilde{f}}(L_\rho, R_\rho)B_0],$
6.  $\begin{aligned} \text{Corr}_\rho^{s(f)}(A, B) &= \text{Corr}_\rho^{s(f)}(A_0, B_0) \\ &= \frac{1}{2} \text{Tr}[\rho A_0 B_0] + \frac{1}{2} \text{Tr}[\rho B_0 A_0] - \text{Tr}[A_0 \cdot m_{\tilde{f}}(L_\rho, R_\rho)B_0] \\ &= \frac{1}{2} \text{Tr}[\rho A_0 B_0] + \frac{1}{2} \text{Tr}[\rho B_0 A_0] - C_\rho^{\tilde{f}}(A_0, B_0). \end{aligned}$

Now we modify the uncertainty relation given by [20].

**Theorem 2.** *For  $f \in \mathcal{F}_{op}^r$ , it holds*

$$I_\rho^f(A) \cdot I_\rho^f(B) \geq |\text{Corr}_\rho^{s(f)}(A, B)|^2,$$

where  $A, B \in M_{n,sa}(\mathbb{C})$  and  $\rho \in M_{n,+1}(\mathbb{C})$ .

*Remark 2.* Since Theorem 2 is easily given by using the Schwarz inequality, we omit the proof. In [20] we gave the uncertainty relation

$$U_\rho^f(A) \cdot U_\rho^f(B) \geq 4f(0)|\text{Corr}_\rho^{s(f)}(A, B)|^2.$$

But since  $4f(0) \leq 1$  and  $I_\rho^f(A) \leq U_\rho^f(A)$ , it is easily given by Theorem 2.

**Theorem 3.** ([20,21]) *For  $f \in \mathcal{F}_{op}^r$  if*

$$\frac{x+1}{2} + \tilde{f}(x) \geq 2f(x), \tag{2}$$

then it holds

$$U_\rho^f(A) \cdot U_\rho^f(B) \geq f(0)|\text{Tr}(\rho[A, B])|^2, \tag{3}$$

$$U_\rho^f(A) \cdot U_\rho^f(B) \geq 4f(0)|\text{Corr}_\rho^f(A, B)|^2, \tag{4}$$

where  $A, B \in M_{n,sa}(\mathbb{C})$  and  $\rho \in M_{n,+1}(\mathbb{C})$ .

**Remark 3.** Though we cannot use the Schwarz inequality, we can get (4) in Theorem 3 by modifying the proof given by [20].

By putting

$$f_{WYD}(x) = \alpha(1 - \alpha) \frac{(x - 1)^2}{(x^\alpha - 1)(x^{1-\alpha} - 1)}, \quad \alpha \in (0, 1),$$

we obtain the following uncertainty relation.

**Corollary 1.** For  $A, B \in M_{n,sa}(\mathbb{C})$  and  $\rho \in M_{n,+1}(\mathbb{C})$ ,

$$U_\rho^{f_{WYD}}(A) \cdot U_\rho^{f_{WYD}}(B) \geq \alpha(1 - \alpha) |Tr(\rho[A, B])|^2,$$

$$U_\rho^{f_{WYD}}(A) \cdot U_\rho^{f_{WYD}}(B) \geq 4\alpha(1 - \alpha) |Corr_\rho^{f_{WYD}}(A, B)|^2,$$

where

$$Corr_\rho^{f_{WYD}}(A, B) = Tr[\rho A_0 B_0] - \frac{1}{2} Tr[\rho^\alpha A_0 \rho^{1-\alpha} B_0] - \frac{1}{2} Tr[\rho^\alpha B_0 \rho^{1-\alpha} A_0].$$

**Remark 4.** Even if (2) does not necessarily hold, then

$$U_\rho^f(A) \cdot U_\rho^f(B) \geq f(0)^2 |Tr(\rho[A, B])|^2, \tag{5}$$

$$U_\rho^f(A) \cdot U_\rho^f(B) \geq 4f(0)^2 |Corr_\rho^f(A, B)|^2, \tag{6}$$

where  $A, B \in M_{n,sa}(\mathbb{C})$  and  $\rho \in M_{n,+1}(\mathbb{C})$ . Since  $f(0) < 1$ , it is easy to show that (5) and (6) are weaker than (3) and (4), respectively.

### Generalized metric adjusted skew information and correlation measure

We give some generalizations of Heisenberg or Schrödinger uncertainty relations which include Theorem 3 as corollary.

**Definition 4.** ([22]) Let  $g, f \in \mathcal{F}_{op}^r$  satisfy

$$g(x) \geq k \frac{(x - 1)^2}{f(x)}$$

for some  $k > 0$ . We define

$$\Delta_g^f(x) = g(x) - k \frac{(x - 1)^2}{f(x)} \in \mathcal{F}_{op}. \tag{7}$$

**Definition 5.** For  $A, B \in M_{n,sa}(\mathbb{C})$  and  $\rho \in M_{n,+1}(\mathbb{C})$ , we define the following quantities:

$$Corr_\rho^{s(gf)}(A, B) = k \langle i[\rho, A], i[\rho, B] \rangle_{\rho, f},$$

$$I_\rho^{(gf)}(A) = Corr_\rho^{s(gf)}(A, A),$$

$$C_\rho^f(A, B) = Tr[A \cdot m_f(L_\rho, R_\rho)B],$$

$$C_\rho^f(A) = C_\rho^f(A, A),$$

$$U_{\rho}^{(g,f)}(A) = \sqrt{(C_{\rho}^g(A) + C_{\rho}^{\Delta_g^f}(A))(C_{\rho}^g(A) - C_{\rho}^{\Delta_g^f}(A))}.$$

The quantity  $I_{\rho}^{(g,f)}(A)$  and  $\text{Corr}_{\rho}^{s(g,f)}(A, B)$  are said to be generalized metric adjusted skew information and generalized metric adjusted correlation measure, respectively.

Then we have the following proposition.

**Proposition 2.** For  $A, B \in M_{n,sa}(\mathbb{C})$  and  $\rho \in M_{n,+1}(\mathbb{C})$ , we have the following relations, where we put  $A_0 = A - \text{Tr}[\rho A]I$  and  $B_0 = B - \text{Tr}[\rho B]I$ :

1.  $I_{\rho}^{(g,f)}(A) = I_{\rho}^{(g,f)}(A_0) = C_{\rho}^g(A_0) - C_{\rho}^{\Delta_g^f}(A_0),$
2.  $J_{\rho}^{(g,f)}(A) = C_{\rho}^g(A_0) + C_{\rho}^{\Delta_g^f}(A_0),$
3.  $U_{\rho}^{(g,f)}(A) = \sqrt{I_{\rho}^{(g,f)}(A) \cdot J_{\rho}^{(g,f)}(A)},$
4.  $\text{Corr}_{\rho}^{s(g,f)}(A, B) = \text{Corr}_{\rho}^{s(g,f)}(A_0, B_0).$

**Theorem 4.** For  $f \in \mathcal{F}_{op}^r$ , it holds

$$I_{\rho}^{(g,f)}(A) \cdot I_{\rho}^{(g,f)}(B) \geq |\text{Corr}_{\rho}^{s(g,f)}(A, B)|^2,$$

where  $A, B \in M_{n,sa}(\mathbb{C})$  and  $\rho \in M_{n,+1}(\mathbb{C})$ .

*Proof of Theorem 4.* We define for  $X, Y \in M_n(\mathbb{C})$

$$\text{Corr}_{\rho}^{s(g,f)}(X, Y) = k(i[\rho, X], i[\rho, Y])_{\rho f}.$$

Since

$$\begin{aligned} \text{Corr}_{\rho}^{s(g,f)}(X, Y) &= k\text{Tr}((i[\rho, X])^* m_f(L_{\rho}, R_{\rho})^{-1} i[\rho, Y]) \\ &= k\text{Tr}((i(L_{\rho} - R_{\rho})X)^* m_f(L_{\rho}, R_{\rho})^{-1} i(L_{\rho} - R_{\rho})Y) \\ &= \text{Tr}(X^* m_g(L_{\rho}, R_{\rho})Y) - \text{Tr}(X^* m_{\Delta_g^f}(L_{\rho}, R_{\rho})Y), \end{aligned}$$

it is easy to show that  $\text{Corr}_{\rho}^{s(g,f)}(X, Y)$  is an inner product in  $M_n(\mathbb{C})$ . Then we can get the result by using the Schwarz inequality. □

**Theorem 5.** For  $f \in \mathcal{F}_{op}^r$ , if

$$g(x) + \Delta_g^f(x) \geq \ell f(x) \tag{8}$$

for some  $\ell > 0$ , then it holds

$$U_{\rho}^{(g,f)}(A) \cdot U_{\rho}^{(g,f)}(B) \geq k\ell |\text{Tr}[\rho[A, B]]|^2, \tag{9}$$

where  $A, B \in M_{n,sa}(\mathbb{C})$  and  $\rho \in M_{n,+1}(\mathbb{C})$ .

In order to prove Theorem 5, we need the following lemmas.

**Lemma 1.** If (7) and (8) are satisfied, then we have the following inequality:

$$m_g(x, y)^2 - m_{\Delta_g^f}(x, y)^2 \geq k\ell(x - y)^2.$$

*Proof of Lemma 1.* By (7) and (8), we have

$$m_{\Delta_g^f}(x, y) = m_g(x, y) - k \frac{(x - y)^2}{m_f(x, y)}, \tag{10}$$

$$m_g(x, y) + m_{\Delta_g^f}(x, y) \geq \ell m_f(x, y). \tag{11}$$

Therefore, by (10) and (11),

$$\begin{aligned} & m_g(x, y)^2 - m_{\Delta_g^f}(x, y)^2 \\ &= \left\{ m_g(x, y) - m_{\Delta_g^f}(x, y) \right\} \left\{ m_g(x, y) + m_{\Delta_g^f}(x, y) \right\} \\ &\geq k \frac{(x - y)^2}{m_f(x, y)} \ell m_f(x, y) \\ &= k \ell (x - y)^2. \end{aligned}$$

We have the following expressions for the quantities  $I_\rho^{(g,f)}(A)$ ,  $J_\rho^{(g,f)}(A)$ ,  $U_\rho^{(g,f)}(A)$ , and  $\text{Corr}_\rho^{s(g,f)}(A, B)$  by using Proposition 2 and a mean  $m_{\Delta_g^f}$ .

**Lemma 2.** Let  $\{|\phi_1\rangle, |\phi_2\rangle, \dots, |\phi_n\rangle\}$  be a basis of eigenvectors of  $\rho$ , corresponding to the eigenvalues  $\{\lambda_1, \lambda_2, \dots, \lambda_n\}$ . We put  $a_{jk} = \langle \phi_j | A_0 | \phi_k \rangle$ ,  $b_{jk} = \langle \phi_j | B_0 | \phi_k \rangle$ , where  $A_0 \equiv A - \text{Tr}[\rho A] I$  and  $B_0 \equiv B - \text{Tr}[\rho B] I$  for  $A, B \in M_{n,sa}(\mathbb{C})$  and  $\rho \in M_{n,+1}(\mathbb{C})$ . Then we have

$$\begin{aligned} I_\rho^{(g,f)}(A) &= \sum_{j,k} m_g(\lambda_j, \lambda_k) a_{jk} a_{kj} - \sum_{j,k} m_{\Delta_g^f}(\lambda_j, \lambda_k) a_{jk} a_{kj} \\ &= 2 \sum_{j < k} \left\{ m_g(\lambda_j, \lambda_k) - m_{\Delta_g^f}(\lambda_j, \lambda_k) \right\} |a_{jk}|^2, \end{aligned}$$

$$\begin{aligned} J_\rho^{(g,f)}(A) &= \sum_{j,k} m_g(\lambda_j, \lambda_k) a_{jk} a_{kj} + \sum_{j,k} m_{\Delta_g^f}(\lambda_j, \lambda_k) a_{jk} a_{kj} \\ &\geq 2 \sum_{j < k} \left\{ m_g(\lambda_j, \lambda_k) + m_{\Delta_g^f}(\lambda_j, \lambda_k) \right\} |a_{jk}|^2, \end{aligned}$$

$$U_\rho^{(g,f)}(A)^2 = \left( \sum_{j,k} m_g(\lambda_j, \lambda_k) |a_{jk}|^2 \right)^2 - \left( \sum_{j,k} m_{\Delta_g^f}(\lambda_j, \lambda_k) |a_{jk}|^2 \right)^2,$$

and

$$\begin{aligned} & \text{Corr}_\rho^{s(g,f)}(A, B) \\ &= \sum_{j,k} m_g(\lambda_j, \lambda_k) a_{jk} b_{kj} - \sum_{j,k} m_{\Delta_g^f}(\lambda_j, \lambda_k) a_{jk} b_{kj} \\ &= \sum_{j < k} \left( m_g(\lambda_j, \lambda_k) - m_{\Delta_g^f}(\lambda_j, \lambda_k) \right) a_{jk} b_{kj} + \sum_{j < k} \left( m_g(\lambda_k, \lambda_j) - m_{\Delta_g^f}(\lambda_k, \lambda_j) \right) a_{kj} b_{jk}. \end{aligned}$$

We are now in a position to prove Theorem 5.

*Proof of Theorem 5.* At first we prove (9). Since

$$\text{Tr}(\rho[A, B]) = \sum_{j,k} (\lambda_j - \lambda_k) a_{jk} b_{kj},$$

$$|T\mathcal{H}(\rho[A, B])| \leq \sum_{j,k} |\lambda_j - \lambda_k| |a_{jk}| |b_{kj}|.$$

Then by Lemma 1, we have

$$\begin{aligned} & k\ell |T\mathcal{H}(\rho[A, B])|^2 \\ & \leq \left\{ \sum_{j,k} \sqrt{k\ell} |\lambda_j - \lambda_k| |a_{jk}| |b_{kj}| \right\}^2 \\ & \leq \left\{ \sum_{j,k} \left( m_g(\lambda_j, \lambda_k)^2 - m_{\Delta_g^f}(\lambda_j, \lambda_k)^2 \right)^{1/2} |a_{jk}| |b_{kj}| \right\}^2 \\ & \leq \left\{ \sum_{j,k} \left( m_g(\lambda_j, \lambda_k) - m_{\Delta_g^f}(\lambda_j, \lambda_k) \right) |a_{jk}|^2 \right\} \left\{ \sum_{j,k} \left( m_g(\lambda_j, \lambda_k) + m_{\Delta_g^f}(\lambda_j, \lambda_k) \right) |b_{kj}|^2 \right\} \\ & = I_\rho^{(g,f)}(A) J_\rho^{(g,f)}(B). \end{aligned}$$

By a similar way, we also have

$$I_\rho^{(g,f)}(B) J_\rho^{(g,f)}(A) \geq k\ell |T\mathcal{H}(\rho[A, B])|^2.$$

Hence, we have the desired inequality (9). □

We give some examples satisfying the condition (8).

**Example 2.** Let

$$\begin{aligned} g(x) &= \frac{x+1}{2}, \\ f(x) &= \alpha(1-\alpha) \frac{(x-1)^2}{(x^\alpha-1)(x^{1-\alpha}-1)}, \quad \alpha \in (0, 1), \\ k &= \frac{f(0)}{2} = \frac{\alpha(1-\alpha)}{2}, \quad \ell = 2. \end{aligned}$$

Then

$$g(x) + \Delta_g^f(x) \geq 2f(x).$$

*Proof of Example 2.* In [10,21] we give

$$(x^{2\alpha} - 1)(x^{2(1-\alpha)} - 1) \geq 4\alpha(1-\alpha)(x-1)^2$$

for  $x > 0$  and  $0 \leq \alpha \leq 1$ . Then we have

$$g(x) + \Delta_g^f(x) \geq 2f(x). \quad \square$$

**Example 3.** Let

$$\begin{aligned} g(x) &= \left( \frac{\sqrt{x}+1}{2} \right)^2, \\ f(x) &= \alpha(1-\alpha) \frac{(x-1)^2}{(x^\alpha-1)(x^{1-\alpha}-1)}, \quad \alpha \in (0, 1), \end{aligned}$$



$$k = \frac{f(0)}{8} = \frac{\alpha(1-\alpha)}{8}, \quad \ell = \frac{3}{2}.$$

Then

$$g(x) + \Delta_g^f(x) \geq \frac{3}{2}f(x)$$

holds for  $0 < \alpha < 1$ .

*Proof of Example 3.* Since

$$\begin{aligned} & \frac{1}{2} \left( \frac{1 + \sqrt{x}}{2} \right)^2 - \frac{1}{8} (x^\alpha - 1)(x^{1-\alpha} - 1) \\ &= \frac{1}{8} (x + 2\sqrt{x} + 1 - x - 1 + x^\alpha + x^{1-\alpha}) \\ &= \frac{1}{8} (2\sqrt{x} + x^\alpha + x^{1-\alpha}) \\ &= \frac{1}{8} (x^{\alpha/2} + x^{(1-\alpha)/2})^2 \geq 0, \end{aligned}$$

we have

$$2 \left( \frac{1 + \sqrt{x}}{2} \right)^2 \geq \frac{1}{8} (x^\alpha - 1)(x^{1-\alpha} - 1) + \frac{3}{2} \left( \frac{1 + \sqrt{x}}{2} \right)^2.$$

Since

$$\alpha(1-\alpha) \frac{(x-1)^2}{(x^\alpha - 1)(x^{1-\alpha} - 1)} \leq \left( \frac{1 + \sqrt{x}}{2} \right)^2,$$

we have

$$2 \left( \frac{1 + \sqrt{x}}{2} \right)^2 \geq \frac{1}{8} (x^\alpha - 1)(x^{1-\alpha} - 1) + \frac{3}{2} \alpha(1-\alpha) \frac{(x-1)^2}{(x^\alpha - 1)(x^{1-\alpha} - 1)}.$$

Then we have

$$g(x) + \Delta_g^f(x) \geq \frac{3}{2}f(x)$$

□

**Example 4.** Let

$$g(x) = \left( \frac{x^\gamma + 1}{2} \right)^{1/\gamma} \quad \left( \frac{3}{4} \leq \gamma \leq 1 \right),$$

$$f(x) = \left( \frac{\sqrt{x} + 1}{2} \right)^2,$$

$$k = \frac{f(0)}{4} = \frac{1}{16}, \quad \ell = 2.$$

Then  $g(x) + \Delta_g^f(x) \geq 2f(x)$ .

In order to prove Example 4, we need the following lemma.

**Lemma 3.** For  $x > 0$ , we set the function of  $y$  as

$$F(y) \equiv \left( \frac{1 + x^y}{2} \right)^{1/y}.$$

Then  $F(y)$  has following properties:

1.  $F(y)$  is monotone increasing for  $y \in \mathbb{R}$ .

2.  $F(y)$  is convex for  $y < 0$ .
3.  $F(y)$  is concave for  $y \geq 1/2$ .

We give the proof of Lemma 3 in the Appendix.

*Proof of Example 4.* By Lemma 3,

$$2 \left( \frac{1 + x^{3/4}}{2} \right)^{4/3} \geq \frac{1+x}{2} + \left( \frac{1 + \sqrt{x}}{2} \right)^2.$$

It follows from the monotonicity that

$$\left( \frac{1 + x^y}{2} \right)^{1/y} \geq \left( \frac{1 + x^{3/4}}{2} \right)^{4/3}$$

for  $y \in [3/4, 1]$ . Then

$$2 \left( \frac{1 + x^y}{2} \right)^{1/y} \geq \frac{1+x}{2} + \left( \frac{1 + \sqrt{x}}{2} \right)^2$$

for  $y \in [3/4, 1]$ . Therefore, we have

$$2 \left( \frac{1 + x^y}{2} \right)^{1/y} - \left( \frac{\sqrt{x} - 1}{2} \right)^2 \geq 2 \left( \frac{\sqrt{x} + 1}{2} \right)^2.$$

Hence, we have

$$g(x) + \Delta_g^f(x) \geq 2f(x).$$

□

## Appendix

*Proof of Lemma 3.*

- (i) Since  $F(y) > 0$  for  $x > 0$  and  $t \in \mathbb{R}$ , it is sufficient to prove  $\frac{d}{dy} \log F(y) > 0$  for the proof of  $F'(y) > 0$ . We have

$$\frac{d}{dy} \log F(y) = \frac{1}{y^2} \left( \log 2 + \frac{x^y \log x^y}{1 + x^y} - \log(1 + x^y) \right).$$

Then we put

$$G(r) \equiv (r + 1) \log 2 + r \log r - (r + 1) \log(r + 1), \quad (r > 0),$$

where we put  $x^y \equiv r > 0$ . From elementary calculations, we have  $G(r) \geq G(1) = 0$  which implies  $\frac{d}{dy} \log F(y) > 0$ .

- (ii) We firstly set  $f(y) \equiv \log F(y)$ . Since  $F(y) > 0$ , we have only to prove  $f''(y) > 0$  for the proof of  $F''(y) > 0$ . We set again  $g(y) \equiv \frac{1+x^y}{2}$ , ( $x > 0, y < 0$ ). Then we have  $\frac{d^2}{dy^2} \log g(y) \equiv \frac{x^y (\log x)^2}{(1+x^y)^2} > 0$ . In addition, by  $f(y) = \frac{1}{y} \log g(y)$ , we have

$$f'(y) = \frac{1}{y} \frac{g'(y)}{g(y)} - \frac{1}{y^2} \log g(y) > 0.$$

By  $\frac{d^2}{dy^2} \log g(y) = \frac{g(y)g''(y) - \{g'(y)\}^2}{g(y)^2}$ , we have

$$f''(y) = \frac{1}{y} \frac{g(y)g''(y) - \{g'(y)\}^2}{g(y)^2} - \frac{2}{y^2} \frac{g'(y)}{g(y)} + \frac{2}{y^3} \log g(y) = \frac{1}{y} \frac{d^2}{dy^2} \log g(y) - \frac{2}{y} f'(y).$$

We prove  $f''(y) > 0$  for  $y < 0$ . We calculate

$$\begin{aligned} f''(y) &= \frac{1}{y} \frac{x^y (\log x)^2}{(1+x^y)^2} - \frac{2}{y^2} \frac{1}{y} \left( \log 2 + \frac{x^y \log x^y}{1+x^y} - \log(1+x^y) \right) \\ &= \frac{1}{y^3 (1+x^y)^2} \left\{ -2x^y (1+x^y) \log x^y + x^y (\log x^y)^2 + 2(1+x^y)^2 \log \frac{1+x^y}{2} \right\}. \end{aligned}$$

Thus, if we put

$$h(y) \equiv -2x^y (1+x^y) \log x^y + x^y (\log x^y)^2 + 2(1+x^y)^2 \log \frac{1+x^y}{2},$$

then we have only to prove  $h(y) < 0$  for  $y < 0$ . Since we have  $h(0) = 0$ , we have only to prove  $h'(y) > 0$  for  $y < 0$ . Here we have

$$h'(y) = -x^y \log x \left\{ 4x^y \log x^y - (\log x^y)^2 - 4(1+x^y) \log \frac{1+x^y}{2} \right\}.$$

If we set again

$$l(t) \equiv 4t \log t - (\log t)^2 - 4(t+1) \log \frac{t+1}{2},$$

where we put  $x^y \equiv t > 0$ , then we prove the following cases:

- (a) If  $x < 1$  (i.e.,  $t > 1$ ), then  $l(t) > 0$ .
- (b) If  $x > 1$  (i.e.,  $0 < t < 1$ ), then  $l(t) < 0$ .

For case (a), we calculate

$$l'(t) = \frac{1}{t} (4t \log 2 + (4t-2) \log t - 4t \log(t+1))$$

and

$$l''(t) = \frac{2 \{ (t+1) \log t + t - 1 \}}{t^2 (t+1)} > 0, (t > 1).$$

Thus, we have  $l'(t) \geq l'(1) = 0$ , and then we have  $l(t) \geq l(1) = 0$ . For case (b), we easily find that

$$l''(t) = \frac{2 \{ (t+1) \log t + t - 1 \}}{t^2 (t+1)} < 0, (0 < t < 1).$$

Thus, we have  $l'(t) \geq l'(1) = 0$ , and then we have  $l(t) \leq l(1) = 0$ .

(iii) We calculate

$$\frac{d^2}{dy^2} F(y) = \frac{1}{y^4} \left( \frac{1+x^y}{2} \right)^{1/y} h(x, y),$$

where

$$\begin{aligned} h(x, y) &= (\log 2 - 2y) \log 2 + \frac{2 \log 2}{1+x^y} \{ x^y \log x^y - (1+x^y) \log(1+x^y) \} \\ &\quad + \frac{1}{(1+x^y)^2} \{ x^y y^2 (x^y + y) (\log x)^2 \} \\ &\quad - \frac{1}{(1+x^y)^2} \{ 2x^y (1+x^y) (y + \log(1+x^y)) \log x^y \} \\ &\quad + \{ 2y + \log(1+x^y) \} \log(1+x^y). \end{aligned}$$

We prove  $h(x, y) \leq 0$  for  $x > 0$  and  $y \geq 1/2$ . Then we have

$$\frac{dh(x, y)}{dx} = -\frac{x^{-1+y} y^2 \log x}{(1+x^y)^3} \left\{ (x^y (y-2) - y) \log x^y + 2(1+x^y) \log \left( \frac{1+x^y}{2} \right) \right\}.$$

Here we note that  $\frac{dh(1,y)}{dx} = 0$ . We also put

$$g(x, y) = \{x^y(-2 + y) - y\} \log x^y + 2(1 + x^y) \log \left( \frac{1 + x^y}{2} \right).$$

If we have  $g(x, y) \geq 0$  for  $x > 0$  and  $y \geq 1/2$ , then we have  $\frac{dh(x,y)}{dx} \geq 0$  for  $0 < x \leq 1$  and  $\frac{dh(x,y)}{dx} \leq 0$  for  $x \geq 1$ . Thus, we then obtain  $h(x, y) \leq h(1, y) = 0$  for  $y \geq 1/2$ , due to  $\frac{dh(1,y)}{dx} = 0$ . Therefore, we have only to prove  $g(x, y) \geq 0$  for  $x > 0$  and  $y \geq 1/2$ .

(a) For the case  $0 < x \leq 1$ , we have

$$\frac{dg(x, y)}{dx} = \frac{y}{x} \left\{ y(x^y - 1) + (y - 2)x^y \log x^y + 2x^y \log \left( \frac{x^y + 1}{2} \right) \right\}.$$

Since  $g(1, y) = 0$ , if we prove  $\frac{dg(x,y)}{dx} \leq 0$ , then we can prove  $g(x, y) \geq g(1, y) = 0$  for  $y \geq 1/2$  and  $0 < x \leq 1$ . Since we have the relations

$$\frac{x - 1}{\sqrt{x}} \leq \log x \leq \frac{2(x - 1)}{x + 1} \leq 0$$

for  $0 < x \leq 1$ , we calculate

$$\begin{aligned} & y(x^y - 1) + (y - 2)x^y \log x^y + 2x^y \log \left( \frac{x^y + 1}{2} \right) \\ & \leq y(x^y - 1) + (y - 2)x^y \frac{(x^y - 1)}{x^{y/2}} + 2x^y \frac{2 \left( \frac{x^y + 1}{2} - 1 \right)}{\frac{x^y + 1}{2} + 1} \\ & = \frac{x^y - 1}{x^y + 3} \left\{ 3(y - 2)x^{y/2} + (y - 2)x^{3y/2} + 3y + (y + 4)x^y \right\}. \end{aligned}$$

Thus, we have only to prove

$$k(y) \equiv 3(y - 2)x^{y/2} + (y - 2)x^{3y/2} + 3y + (y + 4)x^y \geq 0$$

for  $0 < x \leq 1$  and  $y \geq 1/2$ . Since it is trivial  $k(y) \geq 0$  for  $y \geq 2$ , we assume  $1/2 \leq y < 2$  from here. To this end, we prove that  $k_1(y) \equiv 3(y - 2)x^{y/2} + (y - 2)x^{3y/2}$  is monotone increasing for  $1/2 \leq y < 2$  and  $k_2(y) \equiv 3y + (y + 4)x^y$  is also monotone increasing for  $1/2 \leq y < 2$ . We easily find that

$$\frac{dk_1(y)}{dy} = \frac{1}{2}x^{y/2} \{2(x^y + 3) + 3(x^y + 1)(y - 2) \log x\} > 0,$$

for  $0 < x \leq 1$  and  $1/2 \leq y < 2$ .

We also have

$$\frac{dk_2(y)}{dy} = x^y + 3 + (y + 4)x^y \log x.$$

Here we prove  $\frac{dk_2(y)}{dy} \geq 0$  for  $0 < x \leq 1$  and  $1/2 \leq y < 2$ . We put again

$$k_3(x) \equiv x^y + 3 + (y + 4)x^y \log x,$$

then we have

$$\frac{dk_3(x)}{dx} = x^{-1+y} \{2(y + 2) + y(y + 4) \log x\}.$$

Thus, we have

$$\frac{dk_3(x)}{dx} = 0 \Leftrightarrow x = e^{-\frac{2(y+2)}{y(y+4)}} \equiv \alpha_y.$$

Since  $\frac{dk_3(x)}{dx} < 0$  for  $0 < x < \alpha_y$ , and  $\frac{dk_3(x)}{dx} > 0$  for  $\alpha_y < x \leq 1$ , we have

$$k_3(x) \geq k_3(\alpha_y) = 3 - \frac{(y+4)e^{-\frac{2(y+2)}{y+4}}}{y} \equiv k_4(y).$$

Since we have  $\frac{dk_4(y)}{dy} = \frac{8(y+2)e^{-\frac{2(y+2)}{y+4}}}{y^2(y+4)} > 0$ , the function  $k_4(y)$  is monotone increasing for  $y$ . Thus, we have

$$k_3(x) \geq k_3(\alpha_y) = 3 - \frac{(y+4)e^{-\frac{2(y+2)}{y+4}}}{y} \equiv k_4(y) \geq k_4(1/2) = 3 - \frac{9}{e^{10/9}} > 0$$

since  $e^{10/9} \simeq 3.03773$ . Therefore,  $k_2(y)$  is also a monotone increasing function of  $y$  for  $0 < x \leq 1$  and  $1/2 \leq y < 2$ . Thus,  $k(y)$  is monotone increasing for  $y \geq 1/2$ , and then we have

$$k(y) \geq k(1/2) = -\frac{3}{2}(x^{1/4} - 1)^3 \geq 0.$$

(b) For the case  $x \geq 1$ , we firstly calculate

$$\begin{aligned} \frac{dg(x,y)}{dy} &= (x^y - 1) \log x^y \\ &+ \left\{ y(x^y - 1) + (y-2)x^y \log x^y + 2x^y \log \left( \frac{1+x^y}{2} \right) \right\} \log x. \end{aligned}$$

We put

$$p(x,y) \equiv (x^y - 1)y + x^y(y-2) \log x^y + 2x^y \log \left( \frac{1+x^y}{2} \right).$$

Then we calculate

$$\begin{aligned} \frac{dp(x,y)}{dx} &= \frac{y}{x+x^{1-y}} \left\{ (1+x^y)(y-2) \log x^y \right. \\ &\left. + 2 \left( y(1+x^y) - 1 + (1+x^y) \log \left( \frac{1+x^y}{2} \right) \right) \right\}. \end{aligned}$$

Then we put

$$q(x,y) = (y-2) \log x^y + 2 \log \left( \frac{1+x^y}{2} \right) + 2y - \frac{2}{1+x^y}.$$

We have

$$\frac{dq(x,y)}{dy} = \frac{((1+x^y)^2 y - 2) \log x + (1+x^y)^2 (\log x^y + 2)}{(1+x^y)^2} > 0$$

and then

$$q(x,y) \geq q(x, 1/2) = 1 - \frac{2}{\sqrt{x}+1} + 2 \log \left( \frac{1+\sqrt{x}}{2} \right) - \frac{3}{4} \log x.$$

Since we find

$$\frac{dq(x, 1/2)}{dx} = \frac{(\sqrt{x}+3)(\sqrt{x}-1)}{4x(\sqrt{x}+1)^2} \geq 0$$

for  $x \geq 1$ , we have  $q(x,y) \geq q(x, 1/2) \geq q(1, 1/2) = 0$ . Therefore, we have  $\frac{dp(x,y)}{dx} \geq 0$ , which implies  $p(x,y) \geq p(1,y) = 0$ . Thus, we have  $\frac{dg(x,y)}{dy} \geq 0$ , and then we have  $g(x,y) \geq g(x, 1/2)$ , where

$$g(x, 1/2) = -\frac{1}{2}(3x^{1/2} + 1) \log x^{1/2} + 2(x^{1/2} + 1) \log \left( \frac{x^{1/2} + 1}{2} \right).$$

To prove  $g(x, 1/2) \geq 0$  for  $x \geq 1$  and  $y \geq 1/2$ , we put  $x^{1/2} \equiv z \geq 1$  and

$$r(z) \equiv -\frac{1}{2}(3z+1)\log z + 2(z+1)\log\left(\frac{z+1}{2}\right).$$

Since we have  $r''(z) = \frac{(z-1)^2}{2z^2(z+1)} \geq 0$  and

$$r'(z) = \frac{1}{2z} \left\{ z - 1 - 3z \log z + 4z \log\left(\frac{z+1}{2}\right) \right\},$$

we have  $r'(1) = 0$  and then we have  $r'(z) \geq 0$  for  $z \geq 1$ . Thus, we have

$r(z) \geq 0$  for  $z \geq 1$  by  $r(1) = 0$ . Finally, we have  $g(x, y) \geq g(x, 1/2) \geq 0$ , for  $x \geq 1$  and  $y \geq 1/2$ .

□

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