

AN ESTIMATION OF QUASI-ARITHMETIC MEAN BY ARITHMETIC MEAN AND ITS APPLICATIONS

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ABSTRACT. The quasi-arithmetic mean inequality says that if f is an increasing strictly convex function on an interval I , then $f^{-1}(\langle f(A)x, x \rangle) \geq \langle Ax, x \rangle$ for all unit vectors x in a Hilbert space H and a selfadjoint operator A on H , whose spectrum is contained in I . In this paper, we consider reverse inequalities of the quasi-arithmetic mean inequality. For each $\lambda > 0$ we observe an upper bound of a difference

$$f^{-1}(\langle f(A)x, x \rangle) - \lambda \langle Ax, x \rangle.$$

We find a condition on vectors x which attain the optimal bounds.

Replacing a given function $f(t)$ by a power, the logarithmic and the exponential function, we show these reverse quasi-arithmetic mean inequalities and equality conditions, in which the obtained constants are expressed by a generalized Kantorovich constant, the Specht ratio and the logarithmic mean.

1. INTRODUCTION

Let f be a strictly increasing continuous function on an interval I . Then

$$(1.1) \quad f^{-1}\left(\frac{1}{n} \sum_{i=1}^n f(a_i)\right)$$

is called the quasi-arithmetic mean of $a = (a_1, \dots, a_n) \in I^n (\subset \mathbb{R}^n)$ by f (cf. [15]). Typical examples are arithmetic, geometric and harmonic means which correspond to functions $f(t) = t$, $\log t$ and $-\frac{1}{t}$, respectively.

Throughout this paper, an operator means a bounded linear operator on a Hilbert space H . For each unit vector $x \in H$, we consider

$$(1.2) \quad f^{-1}(\langle f(A)x, x \rangle)$$

for all selfadjoint operators A whose spectra are contained in I , as an operator version of the quasi-arithmetic mean (1.1). Incidentally, $\langle Ax, x \rangle$ is regarded as the arithmetic mean. Indeed, (1.1) is obtained by putting $A = \text{diag}(a_1, \dots, a_n)$ and

$x = \frac{1}{\sqrt{n}} \begin{pmatrix} 1 \\ \vdots \\ 1 \end{pmatrix}$ in (1.2), and obviously $\langle Ax, x \rangle = \frac{1}{n} \sum a_i$. If we choose the logarithmic function $f(t) = \log t$, then its quasi-arithmetic mean $\exp\langle (\log A)x, x \rangle$ for a fixed unit

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vector $x \in H$ is called the determinant of a positive invertible operator A in [3], [5]. It is known that if A satisfies $0 < m \leq A \leq M$, then

$$(1.3) \quad \exp\langle(\log A)x, x\rangle \leq \langle Ax, x\rangle \leq S(h) \exp\langle(\log A)x, x\rangle$$

where $h := \frac{M}{m}$ and $S(h)$ is the *Specht ratio* [22] defined by

$$(1.4) \quad S(h) := \frac{h^{\frac{1}{h-1}}}{e \log h^{\frac{1}{h-1}}} \quad \text{for } h > 0 \text{ with } h \neq 1 \quad \text{and} \quad S(1) := 1.$$

The Specht ratio $S(h)$ is an upper bound of the arithmetic mean by the geometric one (cf. [2], [5], [24]). In general, for every (strictly increasing) convex (resp. concave) f , the quasi-arithmetic mean inequality

$$(1.5) \quad f^{-1}(\langle f(A)x, x\rangle) \geq \langle Ax, x\rangle \quad (\text{resp. } f^{-1}(\langle f(A)x, x\rangle) \leq \langle Ax, x\rangle)$$

holds for all unit vectors $x \in H$ which is equivalent to the Jensen inequality

$$\langle f(A)x, x\rangle \geq f(\langle Ax, x\rangle) \quad (\text{resp. } \langle f(A)x, x\rangle \leq f(\langle Ax, x\rangle)).$$

On the other hand, the following reverse inequality of (1.5) for $f(t) = t^p$ is given by Furuta [8]: For $p > 1$ (resp. $0 < p < 1$)

$$(1.6) \quad \begin{aligned} \langle Ax, x\rangle^p &\leq \langle A^p x, x\rangle \leq K(h, p) \langle Ax, x\rangle^p \\ (\text{resp. } \langle Ax, x\rangle^p &\geq \langle A^p x, x\rangle \geq K(h, p) \langle Ax, x\rangle^p) \end{aligned}$$

where $K(h, p)$ is a *generalized Kantorovich constant* (cf. [2], [9], [10]) defined by

$$(1.7) \quad K(h, p) := \frac{1}{h-1} \frac{h^p - h}{p-1} \left(\frac{p-1}{h^p - h} \frac{h^p - 1}{p} \right)^p$$

for all $h > 0$ and $p \in \mathbb{R}$.

In this paper, we give bounds of a difference

$$(1.8) \quad f^{-1}(\langle f(A)x, x\rangle) - \lambda \langle Ax, x\rangle$$

for each $\lambda > 0$. Precisely, we determine the optimal constant $F(\lambda) = F(m, M, f, \lambda)$ (resp. $G(\lambda) = G(m, M, f, \lambda)$) such that

$$(1.9) \quad f^{-1}(\langle f(A)x, x\rangle) - \lambda \langle Ax, x\rangle \leq F(\lambda) \quad (\text{resp. } f^{-1}(\langle f(A)x, x\rangle) - \lambda \langle Ax, x\rangle \geq G(\lambda))$$

for strictly increasing convex (resp. concave) functions f on $[m, M]$ without differentiability. Next we obtain useful estimations by applying it to functions t^p , $\log t$, e^t and $t \log t$. In particular, we have bounds of the ratio $f^{-1}(\langle f(A)x, x\rangle) / \langle Ax, x\rangle$ and the difference $f^{-1}(\langle f(A)x, x\rangle) - \langle Ax, x\rangle$ which include (1.3), (1.6) and other known results. Furthermore, we investigate a specific vector x which gives the equality for the obtained inequality (1.9).

2. MAIN RESULTS

In this section, we consider an optimal estimation of a difference $f^{-1}(\langle f(A)x, x \rangle) - \lambda \langle Ax, x \rangle$ for each $\lambda > 0$ under some conditions.

For the sake of convenience, we define two constants α_f and β_f for all real valued function $f(t)$ on the interval $[m, M]$ with $m < M$ as follows:

$$\alpha_f := \frac{f(M) - f(m)}{M - m} \quad \text{and} \quad \beta_f := \frac{Mf(m) - mf(M)}{M - m}.$$

In particular, if $f(t)$ is strictly increasing, then $\alpha_f > 0$.

We first estimate the quasi-arithmetic mean $f^{-1}(\langle f(A)x, x \rangle)$ by the arithmetic mean $\langle Ax, x \rangle$ for an increasing function f . We use the Mond-Pečarić method [12].

Lemma 2.1. *Let A be a selfadjoint operator on a Hilbert space H such that $m \leq A \leq M$ for some scalars $m < M$. Let $f(t)$ be a strictly increasing continuous function on the interval $[m, M]$. If there exists $t_0 \in [m, M]$ where it satisfies an inequality $f(t) \leq \alpha_f(t - t_0) + f(t_0)$, then for each $\lambda > 0$*

$$(2.1) \quad f^{-1}(\langle f(A)x, x \rangle) - \lambda \langle Ax, x \rangle \leq \max_{m \leq t \leq M} \{f^{-1}(\alpha_f(t - t_0) + f(t_0)) - \lambda t\}$$

holds for all unit vectors $x \in H$.

Proof. The inequality $f(t) \leq \alpha t + \beta$ implies $f(A) \leq \alpha A + \beta$. Since f^{-1} is increasing and $m \leq \langle Ax, x \rangle \leq M$, it follows that

$$f^{-1}(\langle f(A)x, x \rangle) - \lambda \langle Ax, x \rangle \leq f^{-1}(\alpha \langle Ax, x \rangle + \beta) - \lambda \langle Ax, x \rangle \leq \max_{m \leq t \leq M} \{f^{-1}(\alpha t + \beta) - \lambda t\}.$$

Moreover, if we put $\alpha = \alpha_f$ and $\beta = -\alpha_f t_0 + f(t_0)$, then (2.1) holds. \square

Here we put

$$f^{(-)}(t) := \lim_{h \rightarrow -0} \frac{f(t+h) - f(t)}{h} \quad \text{and} \quad f^{(+)}(t) := \lim_{h \rightarrow +0} \frac{f(t+h) - f(t)}{h}$$

if exist, and I_f means the interval spanned by $\frac{\alpha_f}{f^{(-)}(M)}$ and $\frac{\alpha_f}{f^{(+)}(m)}$. (In particular, if $f^{(-)}(t), f^{(+)}(t) = \pm\infty$ for $t \in [m, M]$, then we put $\frac{\alpha_f}{f^{(-)}(t)}, \frac{\alpha_f}{f^{(+)}(t)} = 0$.) If $f(t)$ is convex, then $f^{(-)}(t) \leq f^{(+)}(t)$. Moreover we suppose that $f(t)$ is increasing. Then for each $\lambda \in I_f$, there exists a unique $\mu_\lambda \in [m, M]$ such that $f^{(-)}(\mu_\lambda) \leq \frac{\alpha_f}{\lambda} \leq f^{(+)}(\mu_\lambda)$.

In our previous note [20], we gave reverse inequalities related to the Jensen one under an assumption that $f(t)$ is twice differentiable. But we have the following theorem without differentiability.

Theorem 2.2. *Let A be a selfadjoint operator on a Hilbert space H such that $m \leq A \leq M$ for some scalars $m < M$. Let $f(t)$ be a real valued continuous increasing and strictly convex (resp. strictly concave) function on $[m, M]$ with $f^{(+)}(m) \neq 0$*

(resp. $f^{(-)}(M) \neq 0$). Then for each $\lambda > 0$

$$(2.2) \quad \begin{aligned} & f^{-1}(\langle f(A)x, x \rangle) - \lambda \langle Ax, x \rangle \leq F(m, M, f, \lambda) \\ & \text{(resp. } f^{-1}(\langle f(A)x, x \rangle) - \lambda \langle Ax, x \rangle \geq G(m, M, f, \lambda)) \end{aligned}$$

holds for all unit vectors $x \in H$ where

$$(2.3) \quad \begin{aligned} F(m, M, f, \lambda) &:= \begin{cases} (1 - \lambda)M & \text{if } 0 < \lambda < \frac{\alpha_f}{f^{(-)}(M)} \\ \mu_\lambda - \frac{f(\mu_\lambda) - \beta_f}{\alpha_f} \lambda & \text{if } \lambda \in I_f \\ (1 - \lambda)m & \text{if } \lambda > \frac{\alpha_f}{f^{(+)}(m)} \end{cases} \\ \left(\text{resp. } G(m, M, f, \lambda) &:= \begin{cases} (1 - \lambda)m & \text{if } 0 < \lambda < \frac{\alpha_f}{f^{(+)}(m)} \\ \mu_\lambda - \frac{f(\mu_\lambda) - \beta_f}{\alpha_f} \lambda & \text{if } \lambda \in I_f \\ (1 - \lambda)M & \text{if } \lambda > \frac{\alpha_f}{f^{(-)}(M)} \end{cases} \right). \end{aligned}$$

Furthermore the equivalent condition for which the equality of (2.2) holds is as follows:

(i) If $0 < \lambda < \frac{\alpha_f}{f^{(-)}(M)}$ (resp. $0 < \lambda < \frac{\alpha_f}{f^{(+)}(m)}$), then x is an eigenvector for M (resp. m).

(ii) If $\lambda \in I_f$, then

$$(2.4) \quad x = x_f(\lambda) := \sqrt{1 - \frac{f(\mu_\lambda) - f(m)}{f(M) - f(m)}} e_m + \sqrt{\frac{f(\mu_\lambda) - f(m)}{f(M) - f(m)}} e_M$$

where e_m and e_M are corresponding unit eigenvectors to m and M , respectively.

(iii) If $\lambda > \frac{\alpha_f}{f^{(+)}(m)}$ (resp. $\lambda > \frac{\alpha_f}{f^{(-)}(M)}$), then x is an eigenvector for m (resp. M).

Proof. We only prove the case that $f(t)$ is convex. Then we have $f(t) \leq \alpha_f(t - m) + f(m) = \alpha_f t + \beta_f$ for all $t \in [m, M]$. Since $f(t)$ is strictly increasing on $[m, M]$, $f^{-1}(t)$ can be defined on $[f(m), f(M)]$ and strictly increasing. If $t_0 = m$ in (2.1), then we have

$$f^{-1}(\langle f(A)x, x \rangle) - \lambda \langle Ax, x \rangle \leq \max_{m \leq t \leq M} \{f^{-1}(\alpha_f t + \beta_f) - \lambda t\}$$

holds for all unit vectors $x \in H$. We put

$$h_\lambda(t) := f^{-1}(\alpha_f t + \beta_f) - \lambda t \quad \text{for } t \in [m, M].$$

Then we see the strict concavity of $h_\lambda(t)$ by the strict convexity of $f(t)$ and $h_\lambda^{(\pm)}(t) = \frac{\alpha_f}{f^{(\pm)}(f^{-1}(\alpha_f t + \beta_f))} - \lambda$, respectively. We divide into three cases

$$(i) \ 0 < \lambda < \frac{\alpha_f}{f^{(-)}(M)}, \quad (ii) \ \lambda \in I_f := \left[\frac{\alpha_f}{f^{(-)}(M)}, \frac{\alpha_f}{f^{(+)}(m)} \right] \quad \text{and} \quad (iii) \ \lambda > \frac{\alpha_f}{f^{(+)}(m)}.$$

Firstly we suppose (ii), i.e., $\lambda \in I_f$. Since $h_\lambda(t)$ is strictly concave and $h_\lambda^{(-)}(M) \leq 0 \leq h_\lambda^{(+)}(m)$, there exists a unique $t_\lambda \in [m, M]$ such that $h_\lambda^{(+)}(t_\lambda) \leq 0 \leq h_\lambda^{(-)}(t_\lambda)$ and so $\max_{m \leq t \leq M} h_\lambda(t) = h_\lambda(t_\lambda)$. (If either $h_\lambda^{(+)}(m)$ or $h_\lambda^{(-)}(M) = 0$, then the maximum

is given by $t = t_\lambda = m$ or M , respectively.) Let $\mu_\lambda := f^{-1}(\alpha_f t_\lambda + \beta_f) \in [m, M]$. Then we have $t_\lambda = \frac{f(\mu_\lambda) - \beta_f}{\alpha_f}$ and

$$\max_{m \leq t \leq M} h_\lambda(t) = h_\lambda(t_\lambda) = f^{-1}(\alpha_f t_\lambda + \beta_f) - \lambda t_\lambda = \mu_\lambda - \frac{f(\mu_\lambda) - \beta_f}{\alpha_f} \lambda.$$

Moreover we suppose that

$$f^{-1}(\langle f(A)x, x \rangle) - \lambda \langle Ax, x \rangle = f^{-1}(\alpha_f \langle Ax, x \rangle + \beta_f) - \lambda \langle Ax, x \rangle = h_\lambda(t_\lambda),$$

for some unit vector $x = x_{f,\lambda}$. Then $\langle f(A)x, x \rangle = \alpha_f \langle Ax, x \rangle + \beta_f$ and $\langle Ax, x \rangle = t_\lambda$ by the strict concavity of h_λ . Since $f(t) < \alpha_f t + \beta_f$ for $t \in (m, M)$, the former holds if and only if x is a linear combination of eigenvectors e_m and e_M corresponding to m and M , respectively. We may write $x = \sqrt{1 - s^2}e_m + se_M$ for some $s \in [0, 1]$. So we have

$$\frac{f(\mu_\lambda) - \beta_f}{\alpha_f} = t_\lambda = \langle Ax, x \rangle = (1 - s^2)m + s^2M = (M - m)s^2 + m,$$

or $s^2 = \frac{f(\mu_\lambda) - \beta_f - m\alpha_f}{(M - m)\alpha_f} = \frac{f(\mu_\lambda) - f(m)}{f(M) - f(m)}$ and hence x is of form (2.4). On the other hand, if $x = x_f(\lambda)$ in (2.4), then we have

$$\begin{aligned} & f^{-1}(\langle f(A)x_f(\lambda), x_f(\lambda) \rangle) - \lambda \langle Ax_f(\lambda), x_f(\lambda) \rangle \\ &= f^{-1}\left(\left(f(M) - f(m)\right)\frac{f(\mu_\lambda) - f(m)}{f(M) - f(m)} + f(m)\right) - \lambda\left(\left(M - m\right)\frac{f(\mu_\lambda) - f(m)}{f(M) - f(m)} + m\right) \\ &= \mu_\lambda - \frac{f(\mu_\lambda) - f(m) + m\alpha_f}{\alpha_f} \lambda = \mu_\lambda - \frac{f(\mu_\lambda) - \beta_f}{\alpha_f} \lambda. \end{aligned}$$

We suppose (i), i.e., $0 < \lambda < \frac{\alpha_f}{f^{(-)}(M)}$. Since $h_\lambda(t)$ is strictly concave,

$$\begin{aligned} & \min\{h_\lambda^{(+)}(m), h_\lambda^{(-)}(M), h_\lambda^{(\pm)}(t) : m < t < M\} \\ &= h_\lambda^{(-)}(M) = \frac{\alpha_f}{f^{(-)}(f^{-1}(\alpha_f M + \beta_f))} - \lambda = \frac{\alpha_f}{f^{(-)}(M)} - \lambda > 0. \end{aligned}$$

Hence $h_\lambda(t)$ is strictly increasing in $[m, M]$. Then we have

$$\max_{m \leq t \leq M} h_\lambda(t) = h_\lambda(M) = f^{-1}(\alpha_f M + \beta_f) - \lambda M = f^{-1}(f(M)) - \lambda M = (1 - \lambda)M.$$

Moreover we suppose that

$$f^{-1}(\langle f(A)x, x \rangle) - \lambda \langle Ax, x \rangle = f^{-1}(\alpha_f \langle Ax, x \rangle + \beta_f) - \lambda \langle Ax, x \rangle = (1 - \lambda)M,$$

for some unit vector x . Since $h_\lambda(\langle Ax, x \rangle) = h_\lambda(M)$, we see that $\langle Ax, x \rangle = M$ and so x is an eigenvector for M .

We suppose (iii), i.e., $\lambda > \frac{\alpha_f}{f^{(+)}(m)}$. Then we see $\max\{h_\lambda^{(+)}(m), h_\lambda^{(-)}(M), h_\lambda^{(\pm)}(t) : m < t < M\} < 0$. So the maximum of $h_\lambda(t)$ attains at $t = m$, and we have

$$\max_{m \leq t \leq M} h_\lambda(t) = h_\lambda(m) = (1 - \lambda)m.$$

Moreover we suppose that

$$f^{-1}(\langle f(A)x, x \rangle) - \lambda \langle Ax, x \rangle = (1 - \lambda)m$$

if and only if x is an eigenvector for m by a similar way to (i). \square

Under the same assumptions in Theorem 2.2, we denote the secant through two points $(m, f(m))$ and $(M, f(M))$ as $y = \alpha_f t + \beta_f$. Suppose $mM \geq 0$. Since $f(t)$ is strictly convex, there exists a unique tangent line of the graph $y = f(t)$ through $(0, \beta_f)$

$$y = f(\nu_f) + f^0(\nu_f)(t - \nu_f)$$

where $(\nu_f, f(\nu_f))$ is the point of tangency and $f^0(\nu) := \frac{f(\nu) - \beta_f}{\nu}$. Since we see that $\nu_f \in [m, M]$, the equation $\beta_f = f(\nu) - f^0(\nu)\nu$ has a unique solution $\nu = \nu_f$.

By using the above fact we give explicit estimations of the ratio and difference of $f^{-1}(\langle f(A)x, x \rangle)$ by $\langle Ax, x \rangle$.

Corollary 2.3. *Suppose the hypothesis of Theorem 2.2.*

(i) *Suppose $mM \geq 0$. Then the ratio inequality*

$$(2.5) \quad f^{-1}(\langle f(A)x, x \rangle) \leq \lambda_f \langle Ax, x \rangle \quad (\text{resp. } f^{-1}(\langle f(A)x, x \rangle) \geq \lambda_f \langle Ax, x \rangle)$$

holds for all unit vectors $x \in H$ where $\lambda_f := \frac{\alpha_f}{f^0(\nu_f)} \in I_f$ is a unique solution of $F(\lambda) := F(m, M, f, \lambda) = 0$ (resp. $G(\lambda) := G(m, M, f, \lambda) = 0$). Moreover the equality of (2.5) holds if and only if both m and M are eigenvalues of A and $x = x_f(\lambda_f)$ where $x = x_f(\cdot)$ is defined in Theorem 2.2.

(ii) *The difference inequality*

$$(2.6) \quad f^{-1}(\langle f(A)x, x \rangle) - \langle Ax, x \rangle \leq \mu_1 - \frac{f(\mu_1) - \beta_f}{\alpha_f}$$

$$\left(\text{resp. } f^{-1}(\langle f(A)x, x \rangle) - \langle Ax, x \rangle \geq \mu_1 - \frac{f(\mu_1) - \beta_f}{\alpha_f} \right)$$

holds for all unit vectors $x \in H$ where $\mu = \mu_1 \in (m, M)$ is a unique solution of the equation $f^0(\mu) = \alpha_f$. Moreover the equality of (2.6) holds if and only if both m and M are eigenvalues of A and $x = x_f(1)$ where $x = x_f(\cdot)$ is defined in Theorem 2.2.

Proof. We only prove the case that $f(t)$ is convex. It follows from $0 < f^{(+)}(m) \leq \alpha_f \leq f^{(-)}(M)$ that we may consider the case $\lambda \in I_f$ to estimate $f^{-1}(\langle f(A)x, x \rangle)$ by $\langle Ax, x \rangle$ in (2.3). Indeed, we see $\lambda \neq 1$ and $F(\lambda) \neq 0$ in other two cases.

We put $\lambda_f := \frac{\alpha_f}{f^0(\nu_f)} \in I_f$. It implies that there exists a unique solution $\mu = \mu_{\lambda_f} \in [m, M]$ such that $f^{(-)}(\mu) \leq \frac{\alpha_f}{\lambda_f} \leq f^{(+)}(\mu)$. Moreover the inequality $f^{(-)}(\mu) \leq f^0(\nu_f) \leq f^{(+)}(\mu)$ has a unique solution $\mu = \nu_f$. So it implies that $\mu_{\lambda_f} = \nu_f$ and hence

$$F(m, M, f, \lambda_f) = \mu_{\lambda_f} - \frac{f(\mu_{\lambda_f}) - \beta_f}{\alpha_f} \lambda_f = \nu_f - \frac{f(\nu_f) - \beta_f}{f^0(\nu_f)} = 0.$$

Hence (2.5) is ensured by (2.2). Next the difference inequality (2.6) is obtained by $\lambda = 1$ in (2.2), in which we remark $1 \in I_f$ and $\mu_1 \in (m, M)$ by the mean value theorem. Furthermore, equality conditions of inequalities (2.5) and (2.6) are obtained by (2.4). \square

3. APPLICATIONS FOR THE POWER FUNCTION

We consider applications of Theorem 2.2 to the power function $f(t) = t^p$ ($t > 0$) for $p > 0$. Here we cite the Hölder-McCarthy inequality [18]: For $p \geq 1$ (resp. $0 \leq p \leq 1$)

$$(3.1) \quad \langle A^p x, x \rangle \geq \langle Ax, x \rangle^p \quad (\text{resp. } \langle A^p x, x \rangle \leq \langle Ax, x \rangle^p)$$

(cf. [6], [8], [17]). We give a complementary inequality of (3.1), that is, the following estimation of a difference $\langle A^p x, x \rangle^{1/p} - \lambda \langle Ax, x \rangle$ for each $\lambda > 0$. As a matter of fact, we concentrate the case of $\lambda \in I_p = I_{t^p}$ spanned by $\frac{h^p-1}{ph^{p-1}(h-1)}$ and $\frac{h^p-1}{p(h-1)}$.

Theorem 3.1. *Let A be a positive operator on a Hilbert space H such that $0 < m \leq A \leq M$ for some scalars $m < M$ and $h := \frac{M}{m} (> 1)$. If $p > 1$ (resp. $0 < p < 1$), then for every $\lambda \in I_p$*

$$(3.2) \quad \begin{aligned} \langle A^p x, x \rangle^{\frac{1}{p}} - \lambda \langle Ax, x \rangle &\leq m \frac{h^p - h}{h^p - 1} \left\{ \left(\frac{K(h, p)}{\lambda} \right)^{\frac{1}{p-1}} - \lambda \right\} \\ \left(\text{resp. } \langle A^p x, x \rangle^{\frac{1}{p}} - \lambda \langle Ax, x \rangle \right) &\geq m \frac{h^p - h}{h^p - 1} \left\{ \left(\frac{K(h, p)}{\lambda} \right)^{\frac{1}{p-1}} - \lambda \right\} \end{aligned}$$

holds for all unit vectors $x \in H$ where $K(h, p)$ is defined by (1.7).

Moreover the equality of (3.2) holds if and only if both m and M are eigenvalues of A and

$$(3.3)$$

$$\begin{aligned} x &= x_p(\lambda) \\ &:= \sqrt{\frac{1}{h^p - 1} \left\{ h^p - \left(\frac{h^p - 1}{p\lambda(h-1)} \right)^{\frac{p}{p-1}} \right\}} e_m + \sqrt{\frac{1}{h^p - 1} \left\{ \left(\frac{h^p - 1}{p\lambda(h-1)} \right)^{\frac{p}{p-1}} - 1 \right\}} e_M \end{aligned}$$

where e_m and e_M are corresponding unit eigenvectors to m and M , respectively.

Proof. We suppose $p > 1$ and replace $f(t)$ by t^p for all $t > 0$ in Theorem 2.2. Then we have $\alpha_f = \frac{M^p - m^p}{M - m} (= m^{p-1} \frac{h^p - 1}{h - 1})$, $\beta_f = \frac{Mm^p - mM^p}{M - m} (= m^p \frac{h - h^p}{h - 1})$ and $I_p := I_f = [\frac{h^p - 1}{ph^{p-1}(h-1)}, \frac{h^p - 1}{p(h-1)}]$. For each $\lambda \in I_p$ it follows that the equation $(\mu^p)' = m^{p-1} \frac{h^p - 1}{\lambda(h-1)}$

has a unique solution $\mu = \mu_\lambda := m \left(\frac{h^p - 1}{p\lambda(h-1)} \right)^{\frac{1}{p-1}} \in [m, M]$, and

$$\begin{aligned} \mu_\lambda - \frac{f(\mu_\lambda) - \beta_f}{\alpha_f} \lambda &= m \left(\frac{h^p - 1}{p\lambda(h-1)} \right)^{\frac{1}{p-1}} - \lambda m \frac{h-1}{h^p-1} \left(\frac{h^p - 1}{p\lambda(h-1)} \right)^{\frac{p}{p-1}} + m \frac{h-h^p}{h^p-1} \lambda \\ &= \frac{p-1}{p} m \left(\frac{h^p - 1}{p\lambda(h-1)} \right)^{\frac{1}{p-1}} + m \frac{h-h^p}{h^p-1} \lambda \\ &= m \frac{h^p - h}{h^p - 1} \left[\left\{ \left(\frac{h^p - 1}{h^p - h} \frac{p-1}{p} \right)^p \left(\frac{h^p - h}{\lambda(h-1)(p-1)} \right) \right\}^{\frac{1}{p-1}} - \lambda \right] \\ &= m \frac{h^p - h}{h^p - 1} \left\{ \left(\frac{K(h, p)}{\lambda} \right)^{\frac{1}{p-1}} - \lambda \right\}, \end{aligned}$$

which gives (3.2).

Moreover, since we have

$$\frac{f(\mu_\lambda) - f(m)}{f(M) - f(m)} = \frac{m^p \left(\frac{h^p - 1}{p\lambda(h-1)} \right)^{\frac{p}{p-1}} - m^p}{M^p - m^p} = \frac{1}{h^p - 1} \left\{ \left(\frac{h^p - 1}{p\lambda(h-1)} \right)^{\frac{p}{p-1}} - 1 \right\},$$

the equality condition is given in (3.3) by (2.4).

We can apply a similar discussion to the above in the case of $0 < p < 1$. \square

Comparing with our previous result [23, Corollary 2], the equivalent condition (3.3) that the equality of (3.2) holds, is added. The following ratio inequality and difference one were shown by Furuta [8] and Izumino [16], respectively. They are obtained as a corollary of Theorem 3.1 by the facts that $\lambda = K(h, p)^{\frac{1}{p}} \in I_p$ is a unique solution of the equation $\left(\frac{K(h, p)}{\lambda} \right)^{\frac{1}{p-1}} - \lambda = 0$ and $1 \in I_p$.

Corollary 3.2. *Assume that the conditions of Theorem 3.1 hold and $p > 1$ (resp. $0 < p < 1$).*

(i) *The ratio inequality*

$$(3.4) \quad \langle A^p x, x \rangle^{\frac{1}{p}} \leq K(h, p)^{\frac{1}{p}} \langle Ax, x \rangle \quad \left(\text{resp. } \langle A^p x, x \rangle^{\frac{1}{p}} \geq K(h, p)^{\frac{1}{p}} \langle Ax, x \rangle \right)$$

holds for all unit vectors $x \in H$ where $K(h, p)$ is defined by (1.7). Moreover the equality of (3.4) holds if and only if both m and M are eigenvalues of A and

$$\begin{aligned} x &= x_p(K(h, p)^{\frac{1}{p}}) \\ &= \sqrt{\frac{1}{h^p - 1} \left(h^p - \frac{1}{h-1} \frac{h^p - h}{p-1} \right)} e_m + \sqrt{\frac{1}{h^p - 1} \left(\frac{1}{h-1} \frac{h^p - h}{p-1} - 1 \right)} e_M. \end{aligned}$$

(ii) *The difference inequality*

$$(3.5) \quad \begin{aligned} & \langle A^p x, x \rangle^{\frac{1}{p}} - \langle Ax, x \rangle \leq m \frac{h^p - h}{h^p - 1} \left(K(h, p)^{\frac{1}{p-1}} - 1 \right) \\ & \left(\text{resp. } \langle A^p x, x \rangle^{\frac{1}{p}} - \langle Ax, x \rangle \geq m \frac{h^p - h}{h^p - 1} \left(K(h, p)^{\frac{1}{p-1}} - 1 \right) \right) \end{aligned}$$

holds for all unit vectors $x \in H$ where $K(h, p)$ is defined by (1.7). Moreover the equality of (3.5) holds if and only if both m and M are eigenvalues of A and

$$\begin{aligned} & x = x_p(1) \\ & = \sqrt{\frac{1}{h^p - 1} \left\{ h^p - \left(\frac{h^p - 1}{p(h-1)} \right)^{\frac{p}{p-1}} \right\}} e_m + \sqrt{\frac{1}{h^p - 1} \left\{ \left(\frac{h^p - 1}{p(h-1)} \right)^{\frac{p}{p-1}} - 1 \right\}} e_M. \end{aligned}$$

We have the following result which is given in [25] as a modification of [20, Corollaries 7 and 15]:

Corollary 3.3. (Yamazaki [25]) *Let A be a positive operator on H such that $0 < m \leq A \leq M$ for some scalars $m < M$ and $h = \frac{M}{m} (> 1)$. Then*

(i) *The ratio inequality*

$$(3.6) \quad \langle A^p x, x \rangle^{\frac{1}{p}} \leq K\left(h^r, \frac{p}{r}\right)^{\frac{1}{p}} \langle A^r x, x \rangle^{\frac{1}{r}}$$

holds for $p > r > 0$ and all unit vectors $x \in H$.

(ii) *The difference inequality*

$$(3.7) \quad \langle A^p x, x \rangle - \langle Ax, x \rangle^p \leq m^p \frac{h^p - h}{h - 1} \left\{ K(h, p)^{\frac{1}{p-1}} - 1 \right\}$$

holds for $p > 1$ and all unit vectors $x \in H$.

Proof. In (3.4) we replace A and p by A^r and $\frac{p}{r}$, respectively. Then we have (3.6).

In (3.5) we replace A and p by A^p and $\frac{1}{p}$, respectively. Then it follows from the inversion formula [2, Lemma 4] that

$$\begin{aligned} \langle A^{p \cdot \frac{1}{p}} x, x \rangle^{\frac{1}{1/p}} - \langle A^p x, x \rangle & \geq m^p \frac{h^{p \cdot \frac{1}{p}} - h^p}{h^{p \cdot \frac{1}{p}} - 1} \left(K\left(h^p, \frac{1}{p}\right)^{\frac{1}{\frac{1}{p}-1}} - 1 \right) \\ & = -m^p \frac{h^p - h}{h - 1} \left(K\left(h^p, \frac{1}{p}\right)^{\frac{p}{1-p}} - 1 \right) \\ & = -m^p \frac{h^p - h}{h - 1} \left(K(h, p)^{\frac{1}{p-1}} - 1 \right). \end{aligned}$$

Therefore we have (3.7). □

4. APPLICATIONS FOR LOGARITHMIC AND EXPONENTIAL FUNCTIONS

We consider applications to the logarithmic function $f(t) = \log t$ in Theorem 2.2. Here we note that for a unit vector $x \in H$, the *determinant* $\exp\langle(\log A)x, x\rangle$ for all positive invertible operators A is defined by J.I. Fujii and Y. Seo [3]. The determinant is considered as a continuous (weighted) geometric mean (with the weighted x). The following inequality is a variational expression of the arithmetic - geometric mean inequality:

$$(4.1) \quad \exp\langle(\log A)x, x\rangle \leq \langle Ax, x\rangle.$$

We give a complementary inequality of (4.1), that is, a lower bound of a difference $\exp\langle(\log A)x, x\rangle - \lambda\langle Ax, x\rangle$ for each $\lambda > 0$. For the sake of convenient, we use the *logarithmic mean*

$$(4.2) \quad L(m, M) := \frac{M - m}{\log M - \log m} \quad \text{and} \quad L(m, m) := m$$

for all $0 < m, M$.

Theorem 4.1. *Let A be a positive invertible operator on a Hilbert space H such that $0 < m \leq A \leq M$ for some scalars $m < M$ and $h := \frac{M}{m} (> 1)$. Then for each $\lambda \in I_{\log} := \left[\frac{m}{L(m, M)}, \frac{M}{L(m, M)} \right]$*

$$(4.3) \quad \exp\langle(\log A)x, x\rangle - \lambda\langle Ax, x\rangle \geq -\lambda L(m, M) \log(\lambda S(h))$$

holds for all unit vectors $x \in H$ where $L(m, M)$ and $S(h)$ are defined by (4.2) and (1.4), respectively.

Moreover the equality of (4.3) holds if and only if both m and M are eigenvalues of A and

$$(4.4) \quad x = x_{\log(\lambda)} := \sqrt{1 - \frac{\log(\lambda L(1, h))}{\log h}} e_m + \sqrt{\frac{\log(\lambda L(1, h))}{\log h}} e_M$$

where e_m and e_M are corresponding unit eigenvectors to m and M , respectively.

Proof. We replace $f(t)$ by $\log t$ in Theorem 2.2. Then we have $\alpha_f = L(m, M)^{-1}$, $\beta_f = \frac{M \log m - m \log M}{M - m}$ and $I_f = I_{\log}$. For every $\lambda \in I_{\log}$ the equation $(\log \mu)' = (\lambda L(m, M))^{-1}$ has a unique solution $\mu = \mu_\lambda := \lambda L(m, M) (\in [m, M])$. Hence we have

$$\begin{aligned} \mu_\lambda - \frac{f(\mu_\lambda) - \beta_f}{\alpha_f} \lambda &= \lambda L(m, M) - \lambda L(m, M) \left(\log(\lambda L(m, M)) - \frac{M \log m - m \log M}{M - m} \right) \\ &= \lambda L(m, M) \log \frac{e(\log M - \log m) m^{\frac{M}{M-m}}}{\lambda(M - m) M^{\frac{m}{M-m}}} \\ &= \lambda L(m, M) \log \frac{e \log h^{\frac{1}{h-1}}}{\lambda h^{\frac{1}{h-1}}} = -\lambda L(m, M) \log(\lambda S(h)) \end{aligned}$$

which gives (4.3).

Moreover, since we have

$$\frac{f(\mu_\lambda) - f(m)}{f(M) - f(m)} = \frac{\log(\lambda L(m, M)) - \log m}{\log M - \log m} = \frac{1}{\log h} \log \left(\lambda \frac{h-1}{\log h} \right) = \frac{1}{\log h} \log(\lambda L(1, h)),$$

the equality condition is given in (4.4) by (2.4). \square

The following ratio inequality and difference one with equality conditions were shown in [5] and [3], respectively. They are obtained as a corollary of Theorem 4.1 by the facts that $\lambda = S(h)^{-1} (\in I_{\log})$ is a unique solution of the equation $-L(m, M) \log(\lambda S(h)) = 0$ and $1 \in I_{\log}$. The constant $L(m, M) \log S(h)$ which is represented in the difference inequality is called the *Mond-Shisha difference* (cf. [3], [4], [12], [21]).

Corollary 4.2. *Assume that the conditions of Theorem 4.1 hold.*

(i) *The ratio inequality*

$$(4.5) \quad \exp\langle (\log A)x, x \rangle \geq S(h)^{-1} \langle Ax, x \rangle$$

holds for all unit vectors $x \in H$. Moreover the equality of (4.5) holds if and only if both m and M are eigenvalues of A and

$$x = x_{\log}(S(h)^{-1}) = \sqrt{\frac{h}{h-1} - \frac{1}{\log h}} e_m + \sqrt{\frac{1}{\log h} - \frac{1}{h-1}} e_M.$$

(ii) *The difference inequality*

$$(4.6) \quad \exp\langle (\log A)x, x \rangle - \langle Ax, x \rangle \geq -L(m, M) \log S(h)$$

holds for all unit vectors $x \in H$. Moreover the equality of (4.6) holds if and only if both m and M are eigenvalues of A and

$$x = x_{\log}(1) = \sqrt{1 - \frac{\log L(1, h)}{\log h}} e_m + \sqrt{\frac{\log L(1, h)}{\log h}} e_M.$$

We give an upper bound of a difference $\log\langle e^A x, x \rangle - \lambda \langle Ax, x \rangle$ for each $\lambda > 0$. Ando, Hiai and Petz used this type inequality in [1] and [14]. We need a 2-variable extension of the Specht ratio defined by

$$(4.7) \quad S(h, \lambda) := \frac{h^{\frac{\lambda}{h-1}}}{e \log h^{\frac{\lambda}{h-1}}} \quad \text{for } h > 0, \lambda > 0 \quad \text{and} \quad S(1, \lambda) := \frac{e^\lambda}{\lambda e}.$$

Theorem 4.3. *Let A be a selfadjoint operator on a Hilbert space H such that $m \leq A \leq M$ for some scalars $m < M$. Then for each $\lambda \in I_{\exp} := \left[\frac{e^M - e^m}{e^{M(M-m)}}, \frac{e^M - e^m}{e^{m(M-m)}} \right]$.*

$$(4.8) \quad \log\langle e^A x, x \rangle - \lambda \langle Ax, x \rangle \leq \log\{e^{(1-\lambda)m} S(e^{M-m}, \lambda)\}$$

holds for all unit vectors $x \in H$ where $S(h, \lambda)$ is defined by (4.7).

Moreover the equality of (4.8) holds if and only if both m and M are eigenvalues of A and

$$(4.9) \quad x = x_{\exp}(\lambda) := \sqrt{\frac{e^{M-m}}{e^{M-m}-1} - \frac{1}{\lambda(M-m)}} e_m + \sqrt{\frac{1}{\lambda(M-m)} - \frac{1}{e^{M-m}-1}} e_M$$

where e_m and e_M are corresponding unit eigenvectors to m and M , respectively.

Proof. If we replace $f(t)$ by e^t in Theorem 2.2, then we have $\alpha_f = \frac{e^M - e^m}{M-m}$, $\beta_f = \frac{Me^m - me^M}{M-m}$ and $I_f = I_{\exp}$. For every $\lambda \in I_{\exp}$ the equation $(e^\mu)' = \frac{e^M - e^m}{\lambda(M-m)}$ has a unique solution $\mu = \mu_\lambda := \log \frac{e^M - e^m}{\lambda(M-m)}$ ($\in [m, M]$). Hence putting $h := e^{M-m}$ we have

$$\begin{aligned} \mu_\lambda - \frac{f(\mu_\lambda) - \beta_f}{\alpha_f} \lambda &= \log \frac{e^M - e^m}{\lambda(M-m)} - 1 + \frac{Me^m - me^M}{e^M - e^m} \lambda \\ &= \log \frac{(h-1)e^m}{\lambda e \log h} + \left(\frac{M-m}{h-1} - m \right) \lambda = \log \frac{e^{m(1-\lambda)} h^{\frac{\lambda}{h-1}}}{e \log h h^{\frac{\lambda}{h-1}}} = \log \{ e^{(1-\lambda)m} S(h, \lambda) \}, \end{aligned}$$

which gives (4.8).

Moreover, since we have

$$\frac{f(\mu_\lambda) - f(m)}{f(M) - f(m)} = \frac{\frac{e^M - e^m}{\lambda(M-m)} - e^m}{e^M - e^m} = \frac{1}{\lambda(M-m)} - \frac{1}{e^{M-m} - 1},$$

the equality condition is given in (4.9) by (2.4). \square

Next we see that for $mM \geq 0$ the equation $\log \{ e^{(1-\lambda)m} S(e^{M-m}, \lambda) \} = 0$ has a unique solution $\lambda = \lambda_{\exp}$ ($\in I_{\exp}$), and moreover $1 \in I_{\exp}$. So we have the following corollary:

Corollary 4.4. *Assume that the conditions of Theorem 4.3 hold.*

(i) *Let $mM \geq 0$. Then the equation $e^{(1-\lambda)m} S(e^{M-m}, \lambda) = 1$ has a unique solution $\lambda = \lambda_{\exp}$ ($\in I_{\exp}$), and the ratio inequality*

$$(4.10) \quad \log \langle e^A x, x \rangle \leq \lambda_{\exp} \langle Ax, x \rangle$$

holds for all unit vectors $x \in H$. Moreover the equality of (4.10) holds if and only if both m and M are eigenvalues of A and

$$x = x_{\exp}(\lambda_{\exp}) = \sqrt{\frac{e^{M-m}}{e^{M-m}-1} - \frac{1}{\lambda_{\exp}(M-m)}} e_m + \sqrt{\frac{1}{\lambda_{\exp}(M-m)} - \frac{1}{e^{M-m}-1}} e_M.$$

(ii) *The difference inequality*

$$(4.11) \quad \log \langle e^A x, x \rangle - \langle Ax, x \rangle \leq \log S(e^{M-m})$$

holds for all unit vectors $x \in H$. Moreover the equality of (4.11) holds if and only if both m and M are eigenvalues of A and

$$x = x_{\exp}(1) = \sqrt{\frac{e^{M-m}}{e^{M-m}-1} - \frac{1}{(M-m)}} e_m + \sqrt{\frac{1}{(M-m)} - \frac{1}{e^{M-m}-1}} e_M.$$

Remark 4.5. Taking $m = 0$ in (4.10), we have $\lambda = \lambda_{\exp} := \frac{e^M-1}{M}$ as the solution of the equation $\log\{e^{(1-\lambda)m} S(e^{M-m}, \lambda)\} = 0$. Hence the following ratio inequality related to (4.10) holds:

$$\log\langle e^A x, x \rangle \leq \frac{e^M - 1}{M} \langle Ax, x \rangle.$$

Replacing A and λ with $\log A^p$ and $\frac{q}{p}$, respectively in Theorem 4.3, we have the following result by Furuta and Giga [11]:

Corollary 4.6. (Furuta and Giga [11, Theorem 5.1]) *Let A be a positive invertible operator on H such that $0 < m \leq A \leq M$ for some scalars $m < M$ and $h = \frac{M}{m} (> 1)$. Then*

$$(4.12) \quad S_h(p, q) \exp\langle (\log A^q)x, x \rangle \geq \langle A^p x, x \rangle (\geq \exp\langle (\log A^q)x, x \rangle)$$

hold for all unit vectors x and all $p, q > 0$ with $q \leq \frac{h^p-1}{\log h} \leq qh^p$ where

$$(4.13) \quad S_h(p, q) := m^{p-q} \frac{h^{\frac{q}{h^p-1}}}{e \log h^{\frac{q}{h^p-1}}}.$$

The constant $S_h(p, q)$ ($= m^{p-q} S(h^p, \frac{q}{p})$) has been already appeared in [19].

Concluding this section, we give reverse inequalities of the inequality

$$(4.14) \quad \log\langle Ax, x \rangle \geq \langle (\log A)x, x \rangle$$

for all unit vectors x and a positive invertible operator A .

Corollary 4.7. *Let A be a positive invertible operator on H such that $0 < m \leq A \leq M$ for some scalars $m < M$ and $h = \frac{M}{m} (> 1)$. Then for each $\lambda \in I_{\log} := \left[\frac{L(m, M)}{M}, \frac{L(m, M)}{m} \right]$.*

$$(4.15) \quad \log\langle Ax, x \rangle - \lambda\langle (\log A)x, x \rangle \leq \log\{m^{1-\lambda} S(h, \lambda)\}$$

holds for all unit vectors $x \in H$, where $S(h, \lambda)$ is defined in (4.7). Moreover the equality of (4.15) holds if and only if both m and M are eigenvalues of A and

$$(4.16) \quad x = \sqrt{\frac{h}{h-1} - \frac{1}{\lambda \log h}} e_m + \sqrt{\frac{1}{\lambda \log h} - \frac{1}{h-1}} e_M$$

where e_m and e_M are corresponding unit eigenvectors to m and M , respectively.

In particular, the following inequalities hold:

(i) *The ratio inequality*

$$(4.17) \quad \log \langle Ax, x \rangle \leq \lambda_{\log} \langle (\log A)x, x \rangle$$

holds for $m \geq 1$ or $0 < M \leq 1$ and all unit vectors $x \in H$ where $\lambda = \lambda_{\log} (\in I_{\log})$ is a unique solution of the equation $m^{1-\lambda} S(h, \lambda) = 1$. Moreover the equality of (4.17) holds if and only if both m and M are eigenvalues of A and

$$(4.18) \quad x = \sqrt{\frac{h}{h-1} - \frac{1}{\lambda_{\log} \log h}} e_m + \sqrt{\frac{1}{\lambda_{\log} \log h} - \frac{1}{h-1}} e_M$$

where e_m and e_M are corresponding unit eigenvectors to m and M , respectively.

(ii) *The difference inequality*

$$(4.19) \quad \log \langle Ax, x \rangle - \langle (\log A)x, x \rangle \leq \log S(h).$$

holds for all unit vectors $x \in H$. Moreover the equality of (4.19) holds if and only if both m and M are eigenvalues of A and

$$(4.20) \quad x = \sqrt{\frac{h}{h-1} - \frac{1}{\log h}} e_m + \sqrt{\frac{1}{\log h} - \frac{1}{h-1}} e_M$$

where e_m and e_M are corresponding unit eigenvectors to m and M , respectively.

Remark 4.8. We obtain (4.5) by taking exponential in both sides of (4.19). Incidentally (4.19) has been given in [7].

5. FURUTA-PEČARIĆ INEQUALITY

From the convexity of $t \log t$, the Jensen inequality implies

$$\langle (A \log A)x, x \rangle \geq \langle Ax, x \rangle \log \langle Ax, x \rangle$$

for all unit vectors x and positive invertible operators A . In [13], Furuta and Pečarić showed the following difference inequality as a reverse inequality of it:

Theorem FP. (Furuta and Pečarić [13]) *Let A be a positive invertible operator on a Hilbert space H such that $0 < m \leq A \leq M$ for some scalars $m < M$ and $h := \frac{M}{m} (> 1)$. Then the difference inequality*

$$(5.1) \quad \langle Ax, x \rangle \log \langle Ax, x \rangle - \langle (A \log A)x, x \rangle \geq -\frac{M}{L(1, h)} (S(h) - 1)$$

holds for all unit vectors $x \in H$ where $S(h)$ is defined by (1.4).

From the viewpoint of the quasi-arithmetic mean inequality (1.5), we give an estimation of a difference $\langle Ax, x \rangle \log \langle Ax, x \rangle - \lambda \langle (A \log A)x, x \rangle$ which is an extension of Theorem FP. Moreover we discuss the case the equality holds for the obtained inequality. For the sake of convenience, we recall (4.7); $S(h, \lambda) := \frac{h^{\frac{\lambda}{h-1}}}{e \log h^{\frac{\lambda}{h-1}}}$ ($h > 0, \lambda > 0$) and $S(h, 1) = S(h)$ in particular.

Theorem 5.1. *Let A be a positive invertible operator on a Hilbert space H such that $0 < m \leq A \leq M$ for some scalars $m < M$, $h = \frac{M}{m} (> 1)$ and $I_0 := \left[\frac{L(1,h)}{h}, \frac{L(1,h)}{h} \log(eh) \right]$. Then for each $\lambda \in I_0$*

$$(5.2) \quad \begin{aligned} & \langle Ax, x \rangle \log \langle Ax, x \rangle - \lambda \langle (A \log A)x, x \rangle \\ & \geq (1 - \lambda)(\log m) \langle Ax, x \rangle - \frac{\lambda M}{L(1, h)} (h^{\lambda-1} S(h, \lambda) - 1) \end{aligned}$$

holds for all unit vectors $x \in H$.

Moreover the equality of (5.2) holds if and only if both m and M are eigenvalues of A and

$$(5.3) \quad x = x_0(\lambda) := \sqrt{1 - \frac{h^{\frac{\lambda h}{h-1}}/e - m}{h-1}} e_m + \sqrt{\frac{h^{\frac{\lambda h}{h-1}}/e - m}{h-1}} e_M$$

where e_m and e_M are corresponding unit eigenvectors to m and M , respectively.

Remark 5.2. *Since $1 \in I_0$ in Theorem 5.1, we obtain Theorem FP by putting $\lambda = 1$ in (5.2), and that the equality of (5.1) holds if and only if both m and M are eigenvalues of A and*

$$(5.4) \quad x = \sqrt{1 - \frac{h^{\frac{h}{h-1}}/e - m}{h-1}} e_m + \sqrt{\frac{h^{\frac{h}{h-1}}/e - m}{h-1}} e_M$$

where e_m and e_M are corresponding unit eigenvectors to m and M , respectively.

For the proof of Theorem 5.1, we show the following lemma which represents a lower bound of a difference $\langle Ax, x \rangle \log \langle Ax, x \rangle - \lambda \langle (A \log A)x, x \rangle$ under $1 \leq m \leq A \leq M$:

Lemma 5.3. *Let A be a positive invertible operator on H such that $1 \leq m \leq A \leq M$, $h = \frac{M}{m} (> 1)$ and $I_1 := \left[\frac{\log(me)}{\log(Mh^{\frac{1}{h-1}})}, \frac{\log(Me)}{\log(Mh^{\frac{1}{h-1}})} \right]$. Then for each $\lambda \in I_1$*

$$(5.5) \quad \langle Ax, x \rangle \log \langle Ax, x \rangle - \lambda \langle (A \log A)x, x \rangle \geq -\frac{\lambda M}{L(1, h)} (M^{\lambda-1} S(h, \lambda) - 1)$$

holds for all unit vectors $x \in H$.

Moreover the equality of (5.5) holds if and only if both m and M are eigenvalues of A and

$$(5.6) \quad x = x_1(\lambda) := \sqrt{1 - \frac{(Mh^{\frac{1}{h-1}})^{\lambda}/e - m}{M - m}} e_m + \sqrt{\frac{(Mh^{\frac{1}{h-1}})^{\lambda}/e - m}{M - m}} e_M$$

where e_m and e_M are corresponding unit eigenvectors to m and M , respectively.

Proof. We put $f_1(t) := t \log t$ for all $t \geq 1$. Moreover in Theorem 2.2, we replace $f(t)$ and A by $f_1^{-1}(t)$ and $A \log A$, respectively. Then we see that $(f_1^{-1})'(f_1(t)) = (f_1'(t))^{-1}$

and $f_1'(t) = \log te$. So we have $\alpha_f = \{\log(Mh^{\frac{1}{k-1}})\}^{-1}$, $\beta_f = \frac{M \log h}{\log \frac{Mh}{m}}$ and $I_f = I_1$. For every $\lambda \in I_1$ the inequality (2.2) is represented as follows:

$$\langle Ax, x \rangle \log \langle Ax, x \rangle - \lambda \langle A \log Ax, x \rangle \geq \mu_\lambda \log \mu_\lambda - \frac{\mu_\lambda - \beta_f \lambda}{\alpha_f}$$

where $\mu = \mu_\lambda \in [m, M]$ is a unique solution of the equation $(f_1^{-1})'(\mu \log \mu) = \frac{\alpha_f}{\lambda}$. More precisely, we investigate the right side of the above inequality. The equation $(f_1^{-1})'(\mu \log \mu) = \frac{\alpha_f}{\lambda}$ ensures $\log \mu e = \frac{\lambda}{\alpha_f}$ and so

$$\mu = \mu_\lambda := e^{\frac{\lambda}{\alpha_f} - 1} = \frac{(Mh^{\frac{1}{k-1}})^\lambda}{e} \quad (\in [m, M]).$$

Hence we have

$$\begin{aligned} \mu_\lambda \log \mu_\lambda - \frac{\mu_\lambda - \beta_f \lambda}{\alpha_f} \lambda &= e^{\frac{\lambda}{\alpha_f} - 1} \left(\frac{\lambda}{\alpha_f} - 1 \right) - \frac{e^{\frac{\lambda}{\alpha_f} - 1} - \beta_f \lambda}{\alpha_f} \lambda = -e^{\frac{\lambda}{\alpha_f} - 1} + \frac{\beta_f \lambda}{\alpha_f} \\ &= -\frac{(Mh^{\frac{1}{k-1}})^\lambda}{e} + \frac{\log(Mh^{\frac{1}{k-1}}) \cdot M \log h}{\log \frac{Mh}{m}} \lambda \\ &= -\frac{\lambda M}{L(1, h)} \left(\frac{(h-1)M^\lambda h^{\frac{\lambda}{k-1}}}{\lambda M e \log h} - 1 \right) \\ &= -\frac{\lambda M}{L(1, h)} (M^{\lambda-1} S(h, \lambda) - 1), \end{aligned}$$

which gives (5.5).

The equality condition is given in (5.6) by (2.4). □

Proof of Theorem 5.1. In Lemma 5.3, we replace A by $\frac{A}{m}$, that is, $1 \leq A \leq h$, and the interval I_1 is corresponding to I_0 . So by (5.5) we have for $\lambda \in I_0$

$$\begin{aligned} &\left\langle \frac{A}{m} x, x \right\rangle \log \left\langle \frac{A}{m} x, x \right\rangle - \lambda \left\langle \left(\frac{A}{m} \log \frac{A}{m} \right) x, x \right\rangle \\ &= \frac{1}{m} \{ \langle Ax, x \rangle (\log \langle Ax, x \rangle - \log m) - \lambda (\langle (A \log A) x, x \rangle - (\log m) \langle Ax, x \rangle) \} \\ &= \frac{1}{m} \{ \langle Ax, x \rangle \log \langle Ax, x \rangle - \lambda \langle (A \log A) x, x \rangle - (1 - \lambda) (\log m) \langle Ax, x \rangle \} \\ &\geq -\frac{\lambda h}{L(1, h)} (h^{\lambda-1} S(h, \lambda) - 1). \end{aligned}$$

Hence we have the desired inequality (5.2). □

Next we note that the equation $-\frac{\lambda M}{L(1, h)} (M^{\lambda-1} S(h, \lambda) - 1) = 0$ has a unique solution $\lambda = \lambda_1 \in I_1$. So we have the following corollary:

Corollary 5.4. *Under the same assumption of Lemma 5.3, the ratio inequality*

$$(5.7) \quad \langle Ax, x \rangle \log \langle Ax, x \rangle \geq \lambda_1 \langle (A \log A)x, x \rangle$$

holds for all unit vectors $x \in H$ where $\lambda = \lambda_1 (\in I_1)$ is a unique solution of the equation $-\frac{\lambda M}{L(1, h)}(M^{\lambda-1} S(h, \lambda) - 1) = 0$. Moreover the equality of (5.7) holds if and only if both m and M are eigenvalues of A and

$$(5.8) \quad x = x_1(\lambda_1) := \sqrt{1 - \frac{(Mh^{\frac{1}{h-1}})^{\lambda_1}/e - m}{M - m}} e_m + \sqrt{\frac{(Mh^{\frac{1}{h-1}})^{\lambda_1}/e - m}{M - m}} e_M$$

where e_m and e_M are corresponding unit eigenvectors to m and M , respectively.

If $m = 1$ in Corollary 5.4, then we have $\lambda_1 = \frac{L(1, M)}{M}$. So the following reverse of the Jensen inequality is obtained:

$$(5.9) \quad \langle Ax, x \rangle \log \langle Ax, x \rangle \geq \frac{L(1, M)}{M} \langle (A \log A)x, x \rangle \quad \text{for all unit vectors } x \in H.$$

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