

Article

Jensen Functional, Quasi-Arithmetic Mean and Sharp Converses of Hölder's Inequalities

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Abstract: In this article, we give sharp two-sided bounds for the generalized Jensen functional $J_n(f, g, h; \mathbf{p}, \mathbf{x})$. Assuming convexity/concavity of the generating function h , we give exact bounds for the generalized quasi-arithmetic mean $A_n(h; \mathbf{p}, \mathbf{x})$. In particular, exact bounds are determined for the generalized power means in terms from the class of Stolarsky means. As a consequence, some sharp converses of the famous Hölder's inequality are obtained.

Keywords: quasi-arithmetic means; power means; convex functions; Hölder's inequality

1. Introduction

Recall that the Jensen functional $J_n(\phi; \mathbf{p}, \mathbf{x})$ is defined on an interval $I \subseteq \mathbb{R}$ by

$$J_n(\phi; \mathbf{p}, \mathbf{x}) := \sum_1^n p_i \phi(x_i) - \phi\left(\sum_1^n p_i x_i\right),$$

where $\phi : I \rightarrow \mathbb{R}$, $\mathbf{x} = (x_1, x_2, \dots, x_n) \in I^n$ and $\mathbf{p} = \{p_i\}_1^n$, $\sum_1^n p_i = 1$, is a positive weight sequence.

If ϕ is a convex function on I , then the inequality

$$0 \leq J_n(\phi; \mathbf{p}, \mathbf{x})$$

holds for each $\mathbf{x} \in I^n$ and any positive weight sequence \mathbf{p} .

Jensen's inequality plays a fundamental role in many parts of mathematical analysis and applications. For example, well known $\mathcal{A} - \mathcal{G} - \mathcal{H}$ inequality, Hölder's inequality, Ky Fan inequality, etc., are proven by the help of Jensen's inequality (cf. [1–4]).

Assuming that $\mathbf{x} \in [a, b]^n \subset I^n$, our aim in this paper is to determine some sharp bounds for the generalized Jensen functional

$$J_n(f, g, h; \mathbf{p}, \mathbf{x}) := f\left(\sum_1^n p_i h(x_i)\right) - g\left(h\left(\sum_1^n p_i x_i\right)\right),$$

for suitably chosen functions f, g and h , such that

$$c_{f,g,h}(a, b) \leq J_n(f, g, h; \mathbf{p}, \mathbf{x}) \leq C_{f,g,h}(a, b),$$

i.e., the bounds which does not depend on \mathbf{p} or \mathbf{x} , but only on a, b and functions f, g and h .

Our global bounds will be entirely presented in terms of elementary means.

Recall that the *mean* is a map $M : \mathbb{R}_+ \times \mathbb{R}_+ \rightarrow \mathbb{R}_+$, with a property

$$\min(x, y) \leq M(x, y) \leq \max(x, y),$$

for each $x, y \in \mathbb{R}_+$.



Citation: Simić, S.; Todorčević, V. Jensen Functional, Quasi-Arithmetic Mean and Sharp Converses of Hölder's Inequalities. *Mathematics* **2021**, *9*, 3104. <https://doi.org/10.3390/math9233104>

Academic Editor: Marius Radulescu

Received: 15 November 2021

Accepted: 28 November 2021

Published: 1 December 2021

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In order to make our results condensed and applicable, we shall use in the sequel the class of so-called Stolarsky (or extended) two-parametric mean values, defined for positive values of $x, y, x \neq y$ by the following:

$$E_{r,s}(x, y) = \begin{cases} \left(\frac{r(x^s - y^s)}{s(x^r - y^r)}\right)^{1/(s-r)}, & rs(r-s) \neq 0 \\ \exp\left(\frac{-1}{s} + \frac{x^s \log x - y^s \log y}{x^s - y^s}\right), & r = s \neq 0 \\ \left(\frac{x^s - y^s}{s(\log x - \log y)}\right)^{1/s}, & s \neq 0, r = 0 \\ \sqrt{xy}, & r = s = 0, \\ x, & y = x > 0. \end{cases}$$

In this form, it was introduced by Keneth Stolarsky in [5].

Most of the classical two variable means are just special cases of the class E .

For example,

$$A(x, y) = E_{1,2}(x, y) = \frac{x + y}{2}$$

is the arithmetic mean;

$$G(x, y) = E_{0,0}(x, y) = E_{-r,r}(x, y) = \sqrt{xy}$$

is the geometric mean;

$$L(x, y) = E_{0,1}(x, y) = \frac{x - y}{\log x - \log y}$$

is the logarithmic mean;

$$I(x, y) = E_{1,1}(x, y) = (x^x / y^y)^{\frac{1}{x-y}} / e$$

is the identric mean, etc.

More generally, the r -th power mean

$$A_r(x, y) = \left(\frac{x^r + y^r}{2}\right)^{1/r}$$

is equal to $E_{r,2r}(x, y)$.

Theory of Stolarsky means is very well developed, cf. [6,7] and references therein.

Some basic properties are listed in the following:

Means $E_{r,s}(x, y)$ are

a. symmetric in both parameters, i.e., $E_{r,s}(x, y) = E_{s,r}(x, y)$;

b. symmetric in both variables, i.e., $E_{r,s}(x, y) = E_{r,s}(y, x)$;

c. homogeneous of order one, that is $E_{r,s}(tx, ty) = tE_{r,s}(x, y)$, $t > 0$;

d. monotone increasing in either r or s ;

e. monotone increasing in either x or y ; and

f. logarithmically convex in either r or s for $r, s \in \mathbb{R}_-$ and logarithmically concave for $r, s \in \mathbb{R}_+$.

Let $h : I \rightarrow J$ be a continuous and strictly monotone function on an interval $I \subset \mathbb{R}$. Then, its inverse function $h^{-1} : J \rightarrow I$ exists and generates so-called quasi - arithmetic mean $\mathcal{A}_h(\mathbf{p}, \mathbf{x})$, given by

$$\mathcal{A}_h(\mathbf{p}, \mathbf{x}) := h^{-1}\left(\sum_1^n p_i h(x_i)\right),$$

where $\mathbf{x} = (x_1, x_2, \dots, x_n) \in I^n$ and $\mathbf{p} = \{p_i\}_1^n$, $\sum_1^n p_i = 1$ is a positive weight sequence.

Quasi-arithmetic means are introduced in [1] and then investigated by a plenty of researchers with most interesting results (cf. [8]). In this article, we shall give tight two-sided bounds for the difference

$$\mathcal{A}_h(\mathbf{p}, \mathbf{x}) - \mathcal{A}(\mathbf{p}, \mathbf{x}).$$

An important special case is the class of generalized power means $\mathcal{B}_s(\mathbf{p}, \mathbf{x})$, generated by $h(x) = x^s, s \in \mathbb{R} \setminus \{0\}$,

$$\mathcal{B}_s(\mathbf{p}, \mathbf{x}) = \left(\sum_1^n \mathbf{p}_i x_i^s \right)^{1/s}.$$

It is well known fact that power means are monotone increasing in $s \in \mathbb{R}$ (cf. [1]). Some important particular cases are

$$\mathcal{B}_{-1}(\mathbf{p}, \mathbf{x}) = \left(\sum_1^n \mathbf{p}_i / x_i \right)^{-1} := \mathcal{H}(\mathbf{p}, \mathbf{x});$$

$$\mathcal{B}_0(\mathbf{p}, \mathbf{x}) = \lim_{s \rightarrow 0} \mathcal{B}_s(\mathbf{p}, \mathbf{x}) = \prod_1^n x_i^{\mathbf{p}_i} := \mathcal{G}(\mathbf{p}, \mathbf{x});$$

$$\mathcal{B}_1(\mathbf{p}, \mathbf{x}) = \sum_1^n \mathbf{p}_i x_i := \mathcal{A}(\mathbf{p}, \mathbf{x}),$$

that is, the generalized harmonic, geometric and arithmetic means, respectively. Therefore,

$$\mathcal{H}(\mathbf{p}, \mathbf{x}) \leq \mathcal{G}(\mathbf{p}, \mathbf{x}) \leq \mathcal{A}(\mathbf{p}, \mathbf{x}),$$

represents the celebrated $\mathcal{A} - \mathcal{G} - \mathcal{H}$ inequalities.

Some converses of these inequalities will be given in this paper.

For arbitrary positive sequences a and b and real numbers s, t with $1/s + 1/t = 1$, the celebrated Hölder’s inequalities says that

$$\sum_1^n a_i b_i \leq \left(\sum_1^n a_i^s \right)^{1/s} \left(\sum_1^n b_i^t \right)^{1/t}, \quad s > 1;$$

and

$$\sum_1^n a_i b_i \geq \left(\sum_1^n a_i^s \right)^{1/s} \left(\sum_1^n b_i^t \right)^{1/t}, \quad 0 < s < 1.$$

We shall give in the sequel precise estimations of the difference

$$\sum_1^n a_i b_i - \left(\sum_1^n a_i^s \right)^{1/s} \left(\sum_1^n b_i^t \right)^{1/t},$$

and the quotient

$$\left(\sum_1^n a_i^s \right)^{1/s} \left(\sum_1^n b_i^t \right)^{1/t} / \sum_1^n a_i b_i,$$

that is,

$$\sum_1^n a_i b_i \leq \left(\sum_1^n a_i^s \right)^{1/s} \left(\sum_1^n b_i^t \right)^{1/t} \leq \frac{E_{s+t,s}(a, b) E_{s+t,t}(a, b)}{G^2(a, b)} \sum_1^n a_i b_i,$$

for $1/s + 1/t = 1, s, t > 1; a \leq a_i^{1/t} / b_i^{1/s} \leq b, i = 1, 2, \dots, n$.

2. Results and Proofs

Our main result concerning the generalized Jensen functional $J_n(f, g, h; \mathbf{p}, \mathbf{x})$ is given by the following:

Theorem 1. Let $f : J \rightarrow \mathbb{R}$, $g : J \rightarrow \mathbb{R}$, $h : I \rightarrow J$ be continuous and eventually differentiable functions on their domains.

For $x \in [a, b]^n \subset I^n$, let h be convex on I and f be an increasing function on J .

Then,

$$c_{f,g,h}(a, b) := \min_p [(f \circ h + g \circ h)(pa + (1 - p)b)] \\ \leq J_n(f, g, h; \mathbf{p}, \mathbf{x}) \leq \\ \max_p [f(ph(a) + (1 - p)h(b)) - g(h(pa + (1 - p)b))] := C_{f,g,h}(a, b).$$

Both bounds $c_{f,g,h}(a, b)$ and $C_{f,g,h}(a, b)$ are sharp.

Proof. Since $a \leq x_i \leq b$, there exist non-negative numbers $\lambda_i, \mu_i; \lambda_i + \mu_i = 1$, such that $x_i = \lambda_i a + \mu_i b$, $i = 1, 2, \dots, n$.

Hence,

$$J_n(f, g, h; \mathbf{p}, \mathbf{x}) = f\left(\sum_1^n p_i h(x_i)\right) - g\left(h\left(\sum_1^n p_i x_i\right)\right) = f\left(\sum_1^n p_i h(\lambda_i a + \mu_i b)\right) - g\left(h\left(\sum_1^n p_i (\lambda_i a + \mu_i b)\right)\right) \\ \leq f\left(\sum_1^n p_i (\lambda_i h(a) + \mu_i h(b))\right) - g\left(h\left(a \sum_1^n p_i \lambda_i + b \sum_1^n p_i \mu_i\right)\right) \\ = f(ph(a) + (1 - p)h(b)) - g(h(pa + (1 - p)b)) \leq \max_p [f(ph(a) + (1 - p)h(b)) - g(h(pa + (1 - p)b))],$$

where we denoted $\sum_1^n p_i \lambda_i := p \in [0, 1]$.

The above estimate is valid for arbitrary sequences \mathbf{p} and \mathbf{x} . To prove its sharpness, suppose that the maxima is reached at the point $\mathbf{p} = \mathbf{p}_0$, i.e.,

$$\max_p [f(ph(a) + (1 - p)h(b)) - g(h(pa + (1 - p)b))] \\ = f(p_0 h(a) + (1 - p_0)h(b)) - g(h(p_0 a + (1 - p_0)b)) = C_{f,g,h}(a, b).$$

Then

$$J_n(f, g, h; \mathbf{p}_0, \mathbf{x}_0) = C_{f,g,h}(a, b),$$

where

$$\mathbf{p}_0 = (p_0, p_2, \dots, p_n), \mathbf{x}_0 = (a, b, \dots, b).$$

On the other hand, since h is a convex function on I , by Jensen's inequality we get

$$\sum_1^n p_i h(x_i) \geq h\left(\sum_1^n p_i x_i\right).$$

Because f is an increasing function, it follows that

$$J_n(f, g, h; \mathbf{p}, \mathbf{x}) = f\left(\sum_1^n p_i h(x_i)\right) - g\left(h\left(\sum_1^n p_i x_i\right)\right) \geq f\left(h\left(\sum_1^n p_i x_i\right)\right) - g\left(h\left(\sum_1^n p_i x_i\right)\right) \\ = f(h(pa + (1 - p)b)) - g(h(pa + (1 - p)b)) \geq \min_p [(f \circ h + g \circ h)(pa + (1 - p)b)] := c_{f,g,h}(a, b).$$

A simple analysis of the constant $c_{f,g,h}(a, b)$ reveals the next: if minima of the function $(f \circ h + g \circ h)(t)$ exists for $t = t_0 \in [a, b]$, then $c_{f,g,h}(a, b) = (f \circ h + g \circ h)(t_0)$, taken for $\mathbf{p} = \mathbf{p}_0 = (b - t_0)/(b - a)$.

Otherwise, we have that $c_{f,g,h}(a, b) = \min\{(f \circ h + g \circ h)(a), (f \circ h + g \circ h)(b)\}$.

Those results are evidently sharp, since

$$J_n(f, g, h; \mathbf{p}, \mathbf{x}_0) = c_{f,g,h}(a, b),$$

with $\mathbf{x}_0 = (t_0, \dots, t_0)$, $\mathbf{x}_0 = (a, \dots, a)$ or $\mathbf{x}_0 = (b, \dots, b)$, respectively. \square

Theorem 1 with its variants (a decreasing function f , concave function h) is the source of a plenty of interesting inequalities. Further investigations are left to the reader.

Sometimes, it is a difficult problem to evaluate exact maxima in this theorem.

For this cause, we shall give in the sequel two estimations of $J_n(f, g, h; \mathbf{p}, \mathbf{x})$ with the unique maxima, which could be easily calculated.

Theorem 2. Under the conditions of Theorem 1, assume firstly that f is a convex function on J . Then,

$$J_n(f, g, h; \mathbf{p}, \mathbf{x}) \leq \max_p [p(f \circ h)(a) + (1 - p)(f \circ h)(b) - (g \circ h)(pa + (1 - p)b)].$$

Assuming that $g \circ h$ is a concave function, we obtain

$$J_n(f, g, h; \mathbf{p}, \mathbf{x}) \leq \max_p [f(ph(a) + (1 - p)h(b)) - (p(g \circ h)(a) + (1 - p)(g \circ h)(b))].$$

Now, both maxima can be easily determined by the standard technique.

Proof. By Theorem 1, we know that there exists $p \in [0, 1]$ such that

$$J_n(f, g, h; \mathbf{p}, \mathbf{x}) \leq f(ph(a) + (1 - p)h(b)) - g(h(pa + (1 - p)b)).$$

If additionally f is convex on J , then

$$f(ph(a) + (1 - p)h(b)) \leq p(f \circ h)(a) + (1 - p)(f \circ h)(b).$$

Hence,

$$\begin{aligned} J_n(f, g, h; \mathbf{p}, \mathbf{x}) &\leq p(f \circ h)(a) + (1 - p)(f \circ h)(b) - (g \circ h)(pa + (1 - p)b) \\ &\leq \max_p [p(f \circ h)(a) + (1 - p)(f \circ h)(b) - (g \circ h)(pa + (1 - p)b)]. \end{aligned}$$

Similarly, if $g \circ h$ is a concave function on J , we have

$$g(h(pa + (1 - p)b)) = (g \circ h)(pa + (1 - p)b) \geq p(g \circ h)(a) + (1 - p)(g \circ h)(b),$$

and

$$J_n(f, g, h; \mathbf{p}, \mathbf{x}) \leq \max_p [f(ph(a) + (1 - p)h(b)) - (p(g \circ h)(a) + (1 - p)(g \circ h)(b))].$$

□

An important special case is the converse of Jensen’s inequality.

Theorem 3. Let ϕ be a convex function on $I \subset \mathbb{R}$ and, for $[\xi, \eta] \subset I$, let $\mathbf{x} \in [\xi, \eta]^n$. Then,

$$0 \leq J_n(\phi; \mathbf{p}, \mathbf{x}) \leq \max_p [p\phi(\xi) + (1 - p)\phi(\eta) - \phi(p\xi + (1 - p)\eta)] := T_\phi(\xi, \eta).$$

If ϕ is a concave function, then

$$0 \leq -J_n(\phi; \mathbf{p}, \mathbf{x}) \leq \max_p [\phi(p\xi + (1 - p)\eta) - (p\phi(\xi) + (1 - p)\phi(\eta))] = -T_\phi(\xi, \eta).$$

The constant $T_\phi(\xi, \eta)$ is sharp since there exist sequences $\mathbf{p}_0, \mathbf{x}_0$ such that

$$J_n(\phi; \mathbf{p}_0, \mathbf{x}_0) = T_\phi(\xi, \eta).$$

Proof. This is a simple consequence of Theorem 1 obtained for $f(x) = g(x) = x; h = \phi$. If ϕ is a concave function, then $-\phi$ is convex and the proof follows from the first part of this theorem. □

In this case, the bound $T_\phi(\xi, \eta)$ can be explicitly calculated.

Theorem 4. For a differentiable convex mapping ϕ , we have that

$$T_\phi(\xi, \eta) = \frac{\phi(\eta) - \phi(\xi)}{\eta - \xi} \Theta_\phi(\xi, \eta) + \frac{\eta\phi(\xi) - \xi\phi(\eta)}{\eta - \xi} - \phi(\Theta_\phi(\xi, \eta)),$$

where $\Theta_\phi(\xi, \eta)$ is the Lagrange mean value of numbers ξ and η , defined by

$$\Theta_\phi(\xi, \eta) := (\phi')^{-1}\left(\frac{\phi(\eta) - \phi(\xi)}{\eta - \xi}\right).$$

The function T_ϕ is positive and symmetric, i.e., $T_\phi(\xi, \eta) = T_\phi(\eta, \xi)$ and $\lim_{\eta \rightarrow \xi} T_\phi(\xi, \eta) = 0$.

Proof. If the maximum is taken at the point $\mathbf{p} = \mathbf{p}_0$, by the standard technique we get

$$\phi'(p_0\xi + (1 - p_0)\eta)(\xi - \eta) = \phi(\xi) - \phi(\eta),$$

that is,

$$p_0\xi + (1 - p_0)\eta = \Theta_\phi(\xi, \eta).$$

Therefore,

$$p_0 = \frac{\Theta_\phi(\xi, \eta) - \eta}{\xi - \eta}; \quad 1 - p_0 = \frac{\xi - \Theta_\phi(\xi, \eta)}{\xi - \eta},$$

and

$$\begin{aligned} \max_{\mathbf{p}} [p\phi(\xi) + (1 - p)\phi(\eta) - \phi(p\xi + (1 - p)\eta)] &= p_0\phi(\xi) + (1 - p_0)\phi(\eta) - \phi(p_0\xi + (1 - p_0)\eta) \\ &= \frac{\Theta_\phi(\xi, \eta) - \eta}{\xi - \eta} \phi(\xi) + \frac{\xi - \Theta_\phi(\xi, \eta)}{\xi - \eta} \phi(\eta) - \phi(\Theta_\phi(\xi, \eta)) = T_\phi(\xi, \eta). \end{aligned}$$

□

Now, some important inequalities concerning quasi-arithmetic mean can be easily obtained from Theorem 1 by putting $f = g = h^{-1}$. Nevertheless, in order to avoid unnecessary monotonicity issues, we turn another way.

Our main result is contained in the following:

Theorem 5. For $a \leq x_i \leq b$, $i = 1, 2, \dots, n$; $a, b \in I$, let $h : I \rightarrow J$ be continuous and strictly monotone function and assume that $h^{-1} : J \rightarrow I$ is convex. Then,

$$0 \leq \mathcal{A}(\mathbf{p}, \mathbf{x}) - \mathcal{A}_h(\mathbf{p}, \mathbf{x}) \leq T_{h^{-1}}(h(a), h(b)),$$

where the constant $T_\phi(\xi, \eta)$ is defined in Theorems 3 and 4.

If h^{-1} is a concave function, then

$$0 \leq \mathcal{A}_h(\mathbf{p}, \mathbf{x}) - \mathcal{A}(\mathbf{p}, \mathbf{x}) \leq -T_{h^{-1}}(h(a), h(b)).$$

Proof. We shall give a simple proof of this theorem.

Namely, since h^{-1} is a convex function, applying the first part of Theorem 3 with $\phi = h^{-1}$, we obtain

$$0 \leq \sum_1^n p_i h^{-1}(x_i) - h^{-1}\left(\sum_1^n p_i x_i\right) \leq T_{h^{-1}}(a, b).$$

Now, by changing variables $x_i \rightarrow h(x_i)$ $i = 1, 2, \dots, n$, we get $h^{-1}(x_i) \rightarrow h^{-1} \circ h(x_i) = x_i$ and $a \rightarrow h(a)$, $b \rightarrow h(b)$.

Hence, $h(a) \leq h(x_i) \leq h(b)$ or $h(b) \leq h(x_i) \leq h(a)$ depending on the monotonicity of h . However, since $T_\phi(\xi, \eta)$ is symmetric in variables, we finally get

$$0 \leq \sum_1^n p_i x_i - h^{-1} \left(\sum_1^n p_i h(x_i) \right) \leq T_{h^{-1}}(h(a), h(b)).$$

The second part of this theorem can be proved along the same lines. \square

The most striking example of quasi-arithmetic means is the class of generalized power means $\mathcal{B}_s(\mathbf{p}, \mathbf{x})$, generated by $h(x) = x^s$, $h^{-1}(x) = x^{1/s}$, $s \in \mathbb{R} \setminus \{0\}$, i.e.,

$$\mathcal{B}_s(\mathbf{p}, \mathbf{x}) = \left(\sum_1^n p_i x_i^s \right)^{1/s}.$$

As an application of Theorem 5, we shall estimate the difference $\mathcal{B}_s(\mathbf{p}, \mathbf{x}) - \mathcal{A}(\mathbf{p}, \mathbf{x})$.

Theorem 6. Let $a \leq x_i \leq b$, $i = 1, 2, \dots, n$; $0 < a < b$.

Then,

$$0 \leq \mathcal{B}_s(\mathbf{p}, \mathbf{x}) - \mathcal{A}(\mathbf{p}, \mathbf{x}) \leq \frac{s-1}{s} \left(E_{s,1}(a, b) - \frac{G^2(a, b)}{E_{s,s-1}(a, b)} \right), \quad s > 1;$$

$$0 \leq \mathcal{A}(\mathbf{p}, \mathbf{x}) - \mathcal{B}_s(\mathbf{p}, \mathbf{x}) \leq \frac{1-s}{s} (E_{1,s}(a, b) - E_{1-s,-s}(a, b)), \quad 0 < s < 1;$$

$$0 \leq \mathcal{A}(\mathbf{p}, \mathbf{x}) - \mathcal{B}_s(\mathbf{p}, \mathbf{x}) \leq \frac{s-1}{s} (E_{1-s,-s}(a, b) - E_{1,s}(a, b)), \quad s < 0.$$

Proof. Let $h(x) = x^s$, $h^{-1}(x) = x^{1/s}$, $s \in \mathbb{R} \setminus \{0\}$.

If $s > 1$ then h^{-1} is a concave function on \mathbb{R}^+ . Hence, by the second part of Theorem 5, we get

$$0 \leq \mathcal{B}_s(\mathbf{p}, \mathbf{x}) - \mathcal{A}(\mathbf{p}, \mathbf{x}) \leq -T_{x^{1/s}}(a^s, b^s).$$

Applying the result from Theorem 4, a simple calculation gives

$$\Theta_{x^{1/s}}(a^s, b^s) = \left(\frac{b^s - a^s}{s(b-a)} \right)^{s/(s-1)} = E_{s,1}^s(a, b) = \frac{b^s - a^s}{s(b-a)} E_{s,1}(a, b).$$

Hence,

$$-T_{x^{1/s}}(a^s, b^s) = (\Theta(\cdot))^{1/s} - \frac{b-a}{b^s - a^s} \Theta(\cdot) - \frac{ab^s - ba^s}{b^s - a^s}$$

$$= E_{s,1}(a, b) - \frac{1}{s} E_{s,1}(a, b) - \frac{ab(b^{s-1} - a^{s-1})}{b^s - a^s} = \frac{s-1}{s} \left(E_{s,1}(a, b) - \frac{G^2(a, b)}{E_{s,s-1}(a, b)} \right).$$

In cases $0 < s < 1$ or $s < 0$, one should apply the first part of Theorem 5, since $h^{-1} = x^{1/s}$ is convex on \mathbb{R}^+ . Proceeding as above, the result follows. \square

As a consequence, we obtain some converses of the $\mathcal{A}(\mathbf{p}, \mathbf{x}) - \mathcal{G}(\mathbf{p}, \mathbf{x}) - \mathcal{H}(\mathbf{p}, \mathbf{x})$ inequality.

Corollary 1. Let $a \leq x_i \leq b$, $i = 1, 2, \dots, n$; $0 < a < b$.

Then,

$$0 \leq \mathcal{A}(\mathbf{p}, \mathbf{x}) - \mathcal{H}(\mathbf{p}, \mathbf{x}) \leq 2(A(a, b) - G(a, b)).$$

Proof. Putting $s = -1$, we get

$$\begin{aligned} 0 &\leq \mathcal{A}(\mathbf{p}, \mathbf{x}) - \mathcal{B}_{-1}(\mathbf{p}, \mathbf{x}) = \mathcal{A}(\mathbf{p}, \mathbf{x}) - \mathcal{H}(\mathbf{p}, \mathbf{x}) \\ &\leq 2(E_{2,1}(a, b) - E_{1,-1}(a, b)) = 2(A(a, b) - G(a, b)). \end{aligned}$$

□

Corollary 2. Let $a \leq x_i \leq b, i = 1, 2, \dots, n; 0 < a < b$.

Then,

$$0 \leq \mathcal{A}(\mathbf{p}, \mathbf{x}) - \mathcal{G}(\mathbf{p}, \mathbf{x}) \leq L(a, b) \log \frac{L(a, b)I(a, b)}{G^2(a, b)}.$$

Proof. We have,

$$\begin{aligned} \mathcal{A}(\mathbf{p}, \mathbf{x}) - \mathcal{G}(\mathbf{p}, \mathbf{x}) &= \lim_{s \rightarrow 0} (\mathcal{A}(\mathbf{p}, \mathbf{x}) - \mathcal{B}_s(\mathbf{p}, \mathbf{x})) \\ &\leq \lim_{s \rightarrow 0} \left(\frac{1-s}{s} (E_{1,s}(a, b) - E_{1-s,-s}(a, b)) \right) \\ &= L(a, b) \log \frac{L(a, b)I(a, b)}{G^2(a, b)}. \end{aligned}$$

□

The sequences \mathbf{p} and \mathbf{x} in Theorem 6 are arbitrary. Specializing a little bit, we obtain sharp converses of slightly generalized Hölder’s inequalities.

Theorem 7. Let $\{t_i\}_1^n, \{u_i\}_1^n, \{v_i\}_1^n$ be any sequences of positive real numbers with $a \leq u_i v_i^{1-t} \leq b$ for some constants $0 < a < b$ and $1/s + 1/t = 1$ for some $s, t \in \mathbb{R}$.

Then,

$$0 \leq \left(\sum_1^n t_i u_i^s \right)^{1/s} \left(\sum_1^n t_i v_i^t \right)^{1/t} - \sum_1^n t_i u_i v_i \leq C_s(a, b) \sum_1^n t_i v_i^t,$$

with

$$C_s(a, b) = \frac{s-1}{s} \left(E_{s,1}(a, b) - \frac{G^2(a, b)}{E_{s,s-1}(a, b)} \right),$$

and $s > 1$;

$$0 \leq \sum_1^n t_i u_i v_i - \left(\sum_1^n t_i u_i^s \right)^{1/s} \left(\sum_1^n t_i v_i^t \right)^{1/t} \leq D_s(a, b) \sum_1^n t_i v_i^t,$$

where

$$D_s(a, b) = \frac{1-s}{s} \left(E_{s,1}(a, b) - E_{1-s,-s}(a, b) \right),$$

and $0 < s < 1$.

Proof. Changing variables

$$p_i = t_i v_i^t / \sum_1^n t_i v_i^t; \quad x_i = u_i v_i^{1-t}, \quad i = 1, 2, \dots, n,$$

yields

$$p_i x_i = t_i u_i v_i / \sum_1^n t_i v_i^t;$$

$$p_i x_i^s = t_i v_i^t (u_i v_i^{1-t})^s / \sum_1^n t_i v_i^t = t_i u_i^s v_i^{s+t-st} / \sum_1^n t_i v_i^t = t_i u_i^s / \sum_1^n t_i v_i^t.$$

Now, applying Theorem 6 for $s > 1$, we get

$$\begin{aligned} 0 \leq \mathcal{B}_s(\mathbf{p}, \mathbf{x}) - \mathcal{A}(\mathbf{p}, \mathbf{x}) &= \left(\sum_1^n \mathbf{p}_i \mathbf{x}_i^s\right)^{1/s} - \sum_1^n \mathbf{p}_i \mathbf{x}_i \\ &= \frac{\left(\sum_1^n t_i u_i^t\right)^{1/s}}{\left(\sum_1^n t_i v_i^t\right)^{1/s}} - \frac{\sum_1^n t_i u_i v_i}{\sum_1^n t_i v_i^t} \leq \frac{s-1}{s} \left(E_{s,1}(a, b) - \frac{G^2(a, b)}{E_{s,s-1}(a, b)}\right), \end{aligned}$$

and the result clearly follows by multiplying both sides with $\sum_1^n t_i v_i^t$.

Applying the same procedure in the case $0 < s < 1$, we obtain the second part of this theorem. \square

Finally, we prove another sharp converses of Hölder’s inequalities. For this cause, we shall estimate firstly the expression

$$\mathcal{F}_{s,t}(\mathbf{p}, \mathbf{x}) := \left(\sum_1^n \mathbf{p}_i \mathbf{x}_i^s\right)^{1/s} \left(\sum_1^n \mathbf{p}_i \mathbf{x}_i^{-t}\right)^{1/t}, \quad 1/s + 1/t = 1, \quad s, t \in \mathbb{R}.$$

Lemma 1. Let $a \leq x_i \leq b, i = 1, 2, \dots, n$ for some $0 < a < b$.

If $s > 1$, we have

$$1 \leq \mathcal{F}_{s,t}(\mathbf{p}, \mathbf{x}) \leq \frac{E_{s,s+t}(a, b)E_{t,s+t}(a, b)}{G^2(a, b)},$$

and

$$\frac{E_{s,s+t}(a, b)E_{t,s+t}(a, b)}{G^2(a, b)} \leq \mathcal{F}_{s,t}(\mathbf{p}, \mathbf{x}) \leq 1,$$

for $0 < s < 1$.

Proof. Following the method from the proof of Theorem 1, we get

$$\mathbf{x}_i^s = \lambda_i a^s + \mu_i b^s, \quad \lambda_i + \mu_i = 1, \quad i = 1, 2, \dots, n.$$

If $s > 1$, then also $t > 1$, hence the function $\mathbf{x}^{-t/s}$ is convex.

Therefore,

$$\mathbf{x}_i^{-t} = (\mathbf{x}_i^s)^{-t/s} = (\lambda_i a^s + \mu_i b^s)^{-t/s} \leq \lambda_i (a^s)^{-t/s} + \mu_i (b^s)^{-t/s} = \lambda_i a^{-t} + \mu_i b^{-t},$$

and

$$\begin{aligned} \mathcal{F}_{s,t}(\mathbf{p}, \mathbf{x}) &= \left(\sum_1^n \mathbf{p}_i \mathbf{x}_i^s\right)^{1/s} \left(\sum_1^n \mathbf{p}_i \mathbf{x}_i^{-t}\right)^{1/t} \\ &\leq (a^s \sum_1^n \mathbf{p}_i \lambda_i + b^s \sum_1^n \mathbf{p}_i \mu_i)^{1/s} (a^{-t} \sum_1^n \mathbf{p}_i \lambda_i + b^{-t} \sum_1^n \mathbf{p}_i \mu_i)^{1/t} \\ &= (pa^s + qb^s)^{1/s} (pa^{-t} + qb^{-t})^{1/t}, \end{aligned}$$

where we put

$$\sum_1^n \mathbf{p}_i \lambda_i := p, \quad \sum_1^n \mathbf{p}_i \mu_i := q; \quad p + q = 1.$$

Therefore, it follows that

$$\begin{aligned} \mathcal{F}_{s,t}(\mathbf{p}, \mathbf{x}) &\leq \max_p [(pa^s + qb^s)^{1/s} (pa^{-t} + qb^{-t})^{1/t}] \\ &= (p_0 a^s + q_0 b^s)^{1/s} (p_0 a^{-t} + q_0 b^{-t})^{1/t}. \end{aligned}$$

By the standard technique we obtain that this maxima satisfy the equation

$$\frac{s(p_0a^s + q_0b^s)}{a^s - b^s} = \frac{-t(p_0a^{-t} + q_0b^{-t})}{a^{-t} - b^{-t}},$$

that is,

$$p_0 = \frac{1}{s+t} \left(\frac{sb^s}{b^s - a^s} - \frac{ta^t}{b^t - a^t} \right); \quad q_0 = \frac{1}{s+t} \left(\frac{tb^t}{b^t - a^t} - \frac{sa^s}{b^s - a^s} \right).$$

Henceforth,

$$p_0a^{-t} + q_0b^{-t} = \frac{s}{s+t} \frac{a^{-t}b^s - a^sb^{-t}}{b^s - a^s} = \frac{s}{s+t} \frac{(ab)^{-t}(b^{s+t} - a^{s+t})}{b^s - a^s},$$

and

$$p_0a^s + q_0b^s = \frac{t}{s+t} \frac{b^{s+t} - a^{s+t}}{b^t - a^t}.$$

Therefore,

$$(p_0a^{-t} + q_0b^{-t})^{1/t} = E_{s+t,s}(a, b) / G^2(a, b);$$

$$(p_0a^s + q_0b^s)^{1/s} = E_{s+t,t}(a, b),$$

and we finally obtain

$$\begin{aligned} \max_p [(pa^s + qb^s)^{1/s} (pa^{-t} + qb^{-t})^{1/t}] &= (p_0a^s + q_0b^s)^{1/s} (p_0a^{-t} + q_0b^{-t})^{1/t} \\ &= \frac{E_{s+t,s}(a, b) E_{s+t,t}(a, b)}{G^2(a, b)}. \end{aligned}$$

On the other hand, by the monotonicity in s of \mathcal{B}_s , we get

$$1 \leq \frac{\mathcal{B}_s(\mathbf{p}, \mathbf{x})}{\mathcal{B}_{-t}(\mathbf{p}, \mathbf{x})} = \mathcal{F}_{s,t}(\mathbf{p}, \mathbf{x}),$$

since $s > -t$.

In the case $0 < s < 1$, we have that $t < 0$. Therefore,

$$0 > st = s + t,$$

and

$$\mathcal{F}_{s,t}(\mathbf{p}, \mathbf{x}) \leq 1,$$

since $s < -t$.

Additionally, $-t/s > 1$, hence $x^{-t/s}$ is a convex function.

Therefore,

$$\mathbf{x}_i^{-t} = (\mathbf{x}_i^s)^{-t/s} = (\lambda_i a^s + \mu_i b^s)^{-t/s} \leq \lambda_i (a^s)^{-t/s} + \mu_i (b^s)^{-t/s} = \lambda_i a^{-t} + \mu_i b^{-t}.$$

However, because the exponent $1/t$ is negative in this case, we obtain

$$\begin{aligned} \mathcal{F}_{s,t}(\mathbf{p}, \mathbf{x}) &= \left(\sum_1^n p_i \mathbf{x}_i^s \right)^{1/s} \left(\sum_1^n p_i \mathbf{x}_i^{-t} \right)^{1/t} \\ &\geq (a^s \sum_1^n p_i \lambda_i + b^s \sum_1^n p_i \mu_i)^{1/s} (a^{-t} \sum_1^n p_i \lambda_i + b^{-t} \sum_1^n p_i \mu_i)^{1/t} \\ &= (pa^s + qb^s)^{1/s} (pa^{-t} + qb^{-t})^{1/t}. \end{aligned}$$

Therefore, we get

$$\mathcal{F}_{s,t}(\mathbf{p}, \mathbf{x}) \geq \min_p [(pa^s + qb^s)^{1/s} (pa^{-t} + qb^{-t})^{1/t}],$$

and, proceeding as above, the second part of this theorem follows. \square

We are now able to formulate our main result.

Theorem 8. Let $\{t_i\}_1^n, \{u_i\}_1^n, \{v_i\}_1^n$ be arbitrary sequences of positive numbers with $a \leq u_i^{1/t}/v_i^{1/s} \leq b$ for some constants $0 < a < b$ and $1/s + 1/t = 1$; $s, t \in \mathbb{R}$.

For $s > 1$, we have

$$\sum_1^n t_i u_i v_i \leq \left(\sum_1^n t_i u_i^s\right)^{1/s} \left(\sum_1^n t_i v_i^t\right)^{1/t} \leq \frac{E_{s,s+t}(a,b)E_{t,s+t}(a,b)}{G^2(a,b)} \sum_1^n t_i u_i v_i,$$

and

$$\frac{E_{s,s+t}(a,b)E_{t,s+t}(a,b)}{G^2(a,b)} \sum_1^n t_i u_i v_i \leq \left(\sum_1^n t_i u_i^s\right)^{1/s} \left(\sum_1^n t_i v_i^t\right)^{1/t} \leq \sum_1^n t_i u_i v_i,$$

for $0 < s < 1$.

Proof. Changing variables

$$p_i = t_i u_i v_i / \sum_1^n t_i u_i v_i; \quad x_i = u_i^{1/t} v_i^{-1/s}, \quad i = 1, 2, \dots, n,$$

we get

$$p_i x_i^s = t_i u_i v_i (u_i^{1/t} v_i^{-1/s})^s / \sum_1^n t_i u_i v_i = t_i u_i^{1+s/t} / \sum_1^n t_i u_i v_i = t_i u_i^s / \sum_1^n t_i u_i v_i;$$

$$p_i x_i^{-t} = t_i u_i v_i (u_i^{1/t} v_i^{-1/s})^{-t} / \sum_1^n t_i u_i v_i = t_i v_i^{1+t/s} / \sum_1^n t_i u_i v_i = t_i v_i^t / \sum_1^n t_i u_i v_i,$$

and

$$\mathcal{F}_{s,t}(t, u, v) = \left(\frac{\sum_1^n t_i u_i^s}{\sum_1^n t_i u_i v_i}\right)^{1/s} \left(\frac{\sum_1^n t_i v_i^t}{\sum_1^n t_i u_i v_i}\right)^{1/t} = \frac{(\sum_1^n t_i u_i^s)^{1/s} (\sum_1^n t_i v_i^t)^{1/t}}{\sum_1^n t_i u_i v_i}.$$

Now, an application of Lemma 1 gives the result. \square

3. Conclusions

In this article, we give further development of our results from [3]. Sharp two-sided bounds are explicitly determined for the generalized Jensen functional $J_n(f, g, h; \mathbf{p}, \mathbf{x})$ and, consequently, for Jensen’s inequality and quasi-arithmetic means. Exact converses of $\mathcal{A} - \mathcal{G} - \mathcal{H}$ inequalities and some forms of Hölder’s inequalities are also given. Since Theorem 1 achieved its definite form with very mild conditions posed on the generating functions f, g and h , there remains a lot of work to apply its results in different areas of mathematics.

Author Contributions: Theoretical part, S.S.; numerical part with examples, V.T. All authors have read and agreed to the published version of the manuscript.

Funding: Vesna Todorčević is supported by Researchers Supporting Project number 11143, Faculty of Organizational Sciences, University of Belgrade, Serbia.

Institutional Review Board Statement: Not applicable.

Informed Consent Statement: Not applicable.

Acknowledgments: The authors are grateful to the referees for their valuable comments.

Conflicts of Interest: The authors declare no conflict of interests.

References

1. Hardy, G.H.; Littlewood, J.E.; Polya, G. *Inequalities*; Cambridge University Press: Cambridge, UK, 1978.
2. Dragomir, S.S. Some reverses of the Jensen inequality for functions of self-adjoint operators in Hilbert spaces. *J. Inequal. Appl.* **2010**, *496821*, 15.
3. Simic, S. Sharp global bounds for Jensen's inequality. *Rocky Mt. J. Math.* **2011**, *41*, 2021–2031. [[CrossRef](#)]
4. Simic, S. Some generalizations of Jensen's inequality. *arXiv* **2020**, arXiv:2011.10746.
5. Stolarsky, K.B. Generalizations of the logarithmic mean. *Math. Mag.* **1975**, *48*, 87–92. [[CrossRef](#)]
6. Qi, F. Logarithmic convexity of extended mean values. *Proc. Am. Math. Soc.* **2001**, *130*, 1787–1796. [[CrossRef](#)]
7. Neuman, E.; Páles, Z. On comparison of Stolarsky and Gini means. *J. Math. Anal. Appl.* **2003**, *278*, 274–284. [[CrossRef](#)]
8. Matkowski, J.; Páles, Z. Characterization of generalized quasi-arithmetic means. *Acta Sci. Math.* **2015**, *81*, 34. [[CrossRef](#)]