

**SOME DISCRETE INEQUALITIES FOR CONVEX FUNCTIONS
DEFINED ON LINEAR SPACES**

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ABSTRACT. In this paper we provide some discrete inequalities related to the Hermite-Hadamard result for convex functions defined on convex subsets in a linear space. Applications for norms and univariate real functions with an example for the logarithm, are also given.

1. INTRODUCTION

Let X be a real linear space, $a, b \in X$, $a \neq b$ and let $[a, b] := \{(1 - \lambda)a + \lambda b, \lambda \in [0, 1]\}$ be the *segment* generated by a and b . We consider the function $f : [a, b] \rightarrow \mathbb{R}$ and the attached function $g(a, b) : [0, 1] \rightarrow \mathbb{R}$, $g(a, b)(t) := f[(1 - t)a + tb]$, $t \in [0, 1]$.

It is well known that f is convex on $[a, b]$ iff $g(a, b)$ is convex on $[0, 1]$, and the following lateral derivatives exist and satisfy

- (i) $g'_{\pm}(a, b)(s) = (\nabla_{\pm} f[(1 - s)a + sb])(b - a)$, $s \in [0, 1]$
- (ii) $g'_{+}(a, b)(0) = (\nabla_{+} f(a))(b - a)$
- (iii) $g'_{-}(a, b)(1) = (\nabla_{-} f(b))(b - a)$

where $(\nabla_{\pm} f(x))(y)$ are the *Gâteaux lateral derivatives*, we recall that

$$\begin{aligned} (\nabla_{+} f(x))(y) & : = \lim_{h \rightarrow 0^{+}} \left[\frac{f(x + hy) - f(x)}{h} \right], \\ (\nabla_{-} f(x))(y) & : = \lim_{k \rightarrow 0^{-}} \left[\frac{f(x + ky) - f(x)}{k} \right], \quad x, y \in X. \end{aligned}$$

The following inequality is the well-known Hermite-Hadamard integral inequality for convex functions defined on a segment $[a, b] \subset X$:

$$f\left(\frac{a+b}{2}\right) \leq \int_0^1 f[(1-t)a + tb] dt \leq \frac{f(a) + f(b)}{2}, \quad (\text{HH})$$

which easily follows by the classical Hermite-Hadamard inequality for the convex function $g(a, b) : [0, 1] \rightarrow \mathbb{R}$

$$g(a, b)\left(\frac{1}{2}\right) \leq \int_0^1 g(a, b)(t) dt \leq \frac{g(a, b)(0) + g(a, b)(1)}{2}.$$

For other related results see the monograph on line [4].

We have the following result [2] related to the first Hermite-Hadamard inequality in (HH):

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Theorem 1. Let X be a linear space, $a, b \in X$, $a \neq b$ and $f : [a, b] \subset X \rightarrow \mathbb{R}$ be a convex function on the segment $[a, b]$. Then for any $s \in (0, 1)$ one has the inequality

$$\begin{aligned} & \frac{1}{2} \left[(1-s)^2 (\nabla_+ f [(1-s)a + sb]) (b-a) - s^2 (\nabla_- f [(1-s)a + sb]) (b-a) \right] \\ & \leq \int_0^1 f [(1-t)a + tb] dt - f [(1-s)a + sb] \\ & \leq \frac{1}{2} \left[(1-s)^2 (\nabla_- f (b)) (b-a) - s^2 (\nabla_+ f (a)) (b-a) \right]. \end{aligned} \quad (1)$$

The constant $\frac{1}{2}$ is sharp in both inequalities. The second inequality also holds for $s = 0$ or $s = 1$.

If $f : [a, b] \rightarrow \mathbb{R}$ is as in Theorem 1 and Gâteaux differentiable in $c := (1-\lambda)a + \lambda b$, $\lambda \in (0, 1)$ along the direction $b-a$, then we have the inequality:

$$\left(\frac{1}{2} - \lambda \right) (\nabla f (c)) (b-a) \leq \int_0^1 f [(1-t)a + tb] dt - f (c). \quad (2)$$

If f is as in Theorem 1, then

$$\begin{aligned} 0 & \leq \frac{1}{8} \left[\nabla_+ f \left(\frac{a+b}{2} \right) (b-a) - \nabla_- f \left(\frac{a+b}{2} \right) (b-a) \right] \\ & \leq \int_0^1 f [(1-t)a + tb] dt - f \left(\frac{a+b}{2} \right) \\ & \leq \frac{1}{8} [(\nabla_- f (b)) (b-a) - (\nabla_+ f (a)) (b-a)]. \end{aligned} \quad (3)$$

The constant $\frac{1}{8}$ is sharp in both inequalities.

Also we have the following result [3] related to the second Hermite-Hadamard inequality in (HH):

Theorem 2. Let X be a linear space, $a, b \in X$, $a \neq b$ and $f : [a, b] \subset X \rightarrow \mathbb{R}$ be a convex function on the segment $[a, b]$. Then for any $s \in (0, 1)$ one has the inequality

$$\begin{aligned} & \frac{1}{2} \left[(1-s)^2 (\nabla_+ f [(1-s)a + sb]) (b-a) - s^2 (\nabla_- f [(1-s)a + sb]) (b-a) \right] \\ & \leq (1-s) f (a) + s f (b) - \int_0^1 f [(1-t)a + tb] dt \\ & \leq \frac{1}{2} \left[(1-s)^2 (\nabla_- f (b)) (b-a) - s^2 (\nabla_+ f (a)) (b-a) \right]. \end{aligned} \quad (4)$$

The constant $\frac{1}{2}$ is sharp in both inequalities. The second inequality also holds for $s = 0$ or $s = 1$.

If $f : [a, b] \rightarrow \mathbb{R}$ is as in Theorem 2 and Gâteaux differentiable in $c := (1-\lambda)a + \lambda b$, $\lambda \in (0, 1)$ along the direction $b-a$, then we have the inequality:

$$\left(\frac{1}{2} - \lambda \right) (\nabla f (c)) (b-a) \leq (1-\lambda) f (a) + \lambda f (b) - \int_0^1 f [(1-t)a + tb] dt. \quad (5)$$

If f is as in Theorem 2, then

$$\begin{aligned} 0 &\leq \frac{1}{8} \left[\nabla_+ f \left(\frac{a+b}{2} \right) (b-a) - \nabla_- f \left(\frac{a+b}{2} \right) (b-a) \right] \\ &\leq \frac{f(a) + f(b)}{2} - \int_0^1 f[(1-t)a + tb] dt \\ &\leq \frac{1}{8} [(\nabla_- f(b))(b-a) - (\nabla_+ f(a))(b-a)]. \end{aligned} \quad (6)$$

The constant $\frac{1}{8}$ is sharp in both inequalities.

2. THE RESULTS

Let $f : C \subset X \rightarrow \mathbb{R}$ be a convex function on C . We define the function $F_f : C \times C \rightarrow \mathbb{R}$ by

$$F_f(x, y) := \int_0^1 f((1-t)x + ty) dt. \quad (7)$$

Theorem 3. *Let $f : C \subset X \rightarrow \mathbb{R}$ be a convex function on C . Then the function F_f is convex on $C \times C$ and if $x_i, y_i \in C$ and $p_i \geq 0$ for $i = 1, \dots, n$ with $\sum_{i=1}^n p_i = 1$, then we have the inequalities*

$$\begin{aligned} \sum_{i=1}^n p_i \int_0^1 f((1-t)x_i + ty_i) dt &\geq \int_0^1 f \left((1-t) \sum_{i=1}^n p_i x_i + t \sum_{i=1}^n p_i y_i \right) dt \\ &\geq f \left(\sum_{i=1}^n p_i \left(\frac{x_i + y_i}{2} \right) \right), \end{aligned} \quad (8)$$

$$\begin{aligned} \sum_{i=1}^n p_i \left(\frac{f(x_i) + f(y_i)}{2} \right) &\geq \sum_{i=1}^n p_i \int_0^1 f((1-t)x_i + ty_i) dt \\ &\geq \sum_{i=1}^n p_i f \left(\frac{x_i + y_i}{2} \right) \geq f \left(\sum_{i=1}^n p_i \left(\frac{x_i + y_i}{2} \right) \right) \end{aligned} \quad (9)$$

and

$$\begin{aligned} \sum_{i=1}^n p_i \left(\frac{f(x_i) + f(y_i)}{2} \right) &\geq \frac{1}{2} \left[f \left(\sum_{i=1}^n p_i x_i \right) + f \left(\sum_{i=1}^n p_i y_i \right) \right] \\ &\geq \int_0^1 f \left((1-t) \sum_{i=1}^n p_i x_i + t \sum_{i=1}^n p_i y_i \right) dt. \end{aligned} \quad (10)$$

Proof. Let $(x, y), (u, v) \in C \times C$ and $\alpha, \beta \geq 0$ with $\alpha + \beta = 1$. Then

$$\begin{aligned} F_f(\alpha(x, y) + \beta(u, v)) &= F_f(\alpha x + \beta u, \alpha y + \beta v) \\ &= \int_0^1 f((1-t)(\alpha x + \beta u) + t(\alpha y + \beta v)) dt \\ &= \int_0^1 f(\alpha[(1-t)x + ty] + \beta[(1-t)u + tv]) dt \\ &\leq \int_0^1 [\alpha f((1-t)x + ty) + \beta f((1-t)u + tv)] dt \\ &= \alpha \int_0^1 f((1-t)x + ty) dt + \beta \int_0^1 f((1-t)u + tv) dt \\ &= \alpha F_f(x, y) + \beta F_f(u, v), \end{aligned}$$

which proves the joint convexity of the function F_f .

By Jensen's inequality for the convex function F_f we have

$$\sum_{i=1}^n p_i F_f(x_i, y_i) \geq F_f\left(\sum_{i=1}^n p_i(x_i, y_i)\right) = F_f\left(\sum_{i=1}^n p_i x_i, \sum_{i=1}^n p_i y_i\right),$$

which is equivalent to the first inequality in (8).

By Hermite-Hadamard inequality (HH) we have

$$\begin{aligned} \int_0^1 f\left((1-t)\sum_{i=1}^n p_i x_i + t\sum_{i=1}^n p_i y_i\right) dt &\geq f\left(\frac{\sum_{i=1}^n p_i x_i + \sum_{i=1}^n p_i y_i}{2}\right) \\ &= f\left(\sum_{i=1}^n p_i \left(\frac{x_i + y_i}{2}\right)\right) \end{aligned}$$

and the second part of (8) is proved.

From (HH) we also have for each $i \in \{1, \dots, n\}$ that

$$\frac{f(x_i) + f(y_i)}{2} \geq \int_0^1 f[(1-t)x_i + ty_i] dt \geq f\left(\frac{x_i + y_i}{2}\right).$$

If we multiply this inequality by $p_i \geq 0$ and sum over i from 1 to n we get the first and second inequality in (9).

The last part in (9) follows by Jensen's inequality.

Let $u := \sum_{i=1}^n p_i x_i$ and $v := \sum_{i=1}^n p_i y_i$. By Hermite-Hadamard inequality (HH) we also have

$$\frac{f(u) + f(v)}{2} \geq \int_0^1 f[(1-t)u + tv] dt,$$

which produces the second inequality in (10).

By Jensen's inequality for f we have

$$\sum_{i=1}^n p_i f(x_i) \geq f\left(\sum_{i=1}^n p_i x_i\right)$$

and

$$\sum_{i=1}^n p_i f(y_i) \geq f\left(\sum_{i=1}^n p_i y_i\right).$$

If we sum these two inequalities and divide by 2 we get the first inequality in (10). \square

The following result also holds:

Theorem 4. *With the assumptions of Theorem 3 we have*

$$\begin{aligned}
0 &\leq \frac{1}{8} \left[\sum_{i=1}^n p_i \left(\nabla_+ f \left(\frac{x_i + y_i}{2} \right) (y_i - x_i) \right) \right. \\
&\quad \left. - \sum_{i=1}^n p_i \left(\nabla_- f \left(\frac{x_i + y_i}{2} \right) (y_i - x_i) \right) \right] \\
&\leq \sum_{i=1}^n p_i \int_0^1 f[(1-t)x_i + ty_i] dt - \sum_{i=1}^n p_i f \left(\frac{x_i + y_i}{2} \right) \\
&\leq \frac{1}{8} \left[\sum_{i=1}^n p_i (\nabla_- f(y_i)) (y_i - x_i) - \sum_{i=1}^n p_i (\nabla_+ f(x_i)) (y_i - x_i) \right], \quad (11)
\end{aligned}$$

and

$$\begin{aligned}
0 &\leq \frac{1}{8} \left[\sum_{i=1}^n p_i \left(\nabla_+ f \left(\frac{x_i + y_i}{2} \right) (y_i - x_i) \right) \right. \\
&\quad \left. - \sum_{i=1}^n p_i \left(\nabla_- f \left(\frac{x_i + y_i}{2} \right) (y_i - x_i) \right) \right] \\
&\leq \sum_{i=1}^n p_i \left(\frac{f(x_i) + f(y_i)}{2} \right) - \sum_{i=1}^n p_i \int_0^1 f((1-t)x_i + ty_i) dt \\
&\leq \frac{1}{8} \left[\sum_{i=1}^n p_i (\nabla_- f(y_i)) (y_i - x_i) - \sum_{i=1}^n p_i (\nabla_+ f(x_i)) (y_i - x_i) \right]. \quad (12)
\end{aligned}$$

We also have

$$\begin{aligned}
0 &\leq \frac{1}{8} \left[\nabla_+ f \left(\sum_{i=1}^n p_i \left(\frac{x_i + y_i}{2} \right) \right) \left(\sum_{i=1}^n p_i (y_i - x_i) \right) \right. \\
&\quad \left. - \nabla_- f \left(\sum_{i=1}^n p_i \left(\frac{x_i + y_i}{2} \right) \right) \left(\sum_{i=1}^n p_i (y_i - x_i) \right) \right] \\
&\leq \int_0^1 f \left((1-t) \sum_{i=1}^n p_i x_i + t \sum_{i=1}^n p_i y_i \right) dt - f \left(\sum_{i=1}^n p_i \left(\frac{x_i + y_i}{2} \right) \right) \\
&\leq \frac{1}{8} \left[\left(\nabla_- f \left(\sum_{i=1}^n p_i y_i \right) \right) \left(\sum_{i=1}^n p_i (y_i - x_i) \right) \right. \\
&\quad \left. - \left(\nabla_+ f \left(\sum_{i=1}^n p_i x_i \right) \right) \left(\sum_{i=1}^n p_i (y_i - x_i) \right) \right] \quad (13)
\end{aligned}$$

and

$$\begin{aligned}
0 &\leq \frac{1}{8} \left[\nabla_+ f \left(\sum_{i=1}^n p_i \left(\frac{x_i + y_i}{2} \right) \right) \left(\sum_{i=1}^n p_i (y_i - x_i) \right) \right. \\
&\quad \left. - \nabla_- f \left(\sum_{i=1}^n p_i \left(\frac{x_i + y_i}{2} \right) \right) \left(\sum_{i=1}^n p_i (y_i - x_i) \right) \right] \\
&\leq \frac{1}{2} \left[f \left(\sum_{i=1}^n p_i x_i \right) + f \left(\sum_{i=1}^n p_i y_i \right) \right] - \int_0^1 f \left((1-t) \sum_{i=1}^n p_i x_i + t \sum_{i=1}^n p_i y_i \right) dt \\
&\leq \frac{1}{8} \left[\left(\nabla_- f \left(\sum_{i=1}^n p_i y_i \right) \right) \left(\sum_{i=1}^n p_i (y_i - x_i) \right) \right. \\
&\quad \left. - \left(\nabla_+ f \left(\sum_{i=1}^n p_i x_i \right) \right) \left(\sum_{i=1}^n p_i (y_i - x_i) \right) \right]. \quad (14)
\end{aligned}$$

Proof. From the inequality (3) we have for $a = x_i$ and $b = y_i$, where $i \in \{1, \dots, n\}$ that

$$\begin{aligned}
0 &\leq \frac{1}{8} \left[\nabla_+ f \left(\frac{x_i + y_i}{2} \right) (y_i - x_i) - \nabla_- f \left(\frac{x_i + y_i}{2} \right) (y_i - x_i) \right] \\
&\leq \int_0^1 f [(1-t)x_i + ty_i] dt - f \left(\frac{x_i + y_i}{2} \right) \\
&\leq \frac{1}{8} [(\nabla_- f(y_i))(y_i - x_i) - (\nabla_+ f(x_i))(y_i - x_i)],
\end{aligned}$$

for any $i \in \{1, \dots, n\}$.

If we multiply this inequality by $p_i \geq 0$ and sum over i from 1 to n , then we get

$$\begin{aligned}
0 &\leq \frac{1}{8} \sum_{i=1}^n p_i \left[\nabla_+ f \left(\frac{x_i + y_i}{2} \right) (y_i - x_i) - \nabla_- f \left(\frac{x_i + y_i}{2} \right) (y_i - x_i) \right] \\
&\leq \sum_{i=1}^n p_i \int_0^1 f [(1-t)x_i + ty_i] dt - \sum_{i=1}^n p_i f \left(\frac{x_i + y_i}{2} \right) \\
&\leq \frac{1}{8} \sum_{i=1}^n p_i [(\nabla_- f(y_i))(y_i - x_i) - (\nabla_+ f(x_i))(y_i - x_i)],
\end{aligned}$$

which is equivalent to (11).

The inequality (12) follows in a similar way by employing the inequality (6).

The inequalities (13) and (14) follow by taking $a = \sum_{i=1}^n p_i x_i$ and $b = \sum_{i=1}^n p_i y_i$ in the inequalities (3) and (6). \square

3. EXAMPLES FOR NORMS

Now, assume that $(X, \|\cdot\|)$ is a normed linear space. The function $f_0(s) = \frac{1}{2} \|x\|^2$, $x \in X$ is convex and thus the following limits exist

$$\begin{aligned}
\text{(iv)} \quad \langle x, y \rangle_s &:= (\nabla_+ f_0(y))(x) = \lim_{t \rightarrow 0^+} \left[\frac{\|y+tx\|^2 - \|y\|^2}{2t} \right]; \\
\text{(v)} \quad \langle x, y \rangle_i &:= (\nabla_- f_0(y))(x) = \lim_{s \rightarrow 0^-} \left[\frac{\|y+sx\|^2 - \|y\|^2}{2s} \right];
\end{aligned}$$

for any $x, y \in X$. They are called the *lower* and *upper semi-inner* products associated to the norm $\|\cdot\|$.

For the sake of completeness we list here some of the main properties of these mappings that will be used in the sequel (see for example [1] or [5]), assuming that $p, q \in \{s, i\}$ and $p \neq q$:

- (a) $\langle x, x \rangle_p = \|x\|^2$ for all $x \in X$;
- (aa) $\langle \alpha x, \beta y \rangle_p = \alpha\beta \langle x, y \rangle_p$ if $\alpha, \beta \geq 0$ and $x, y \in X$;
- (aaa) $|\langle x, y \rangle_p| \leq \|x\| \|y\|$ for all $x, y \in X$;
- (av) $\langle \alpha x + y, x \rangle_p = \alpha \langle x, x \rangle_p + \langle y, x \rangle_p$ if $x, y \in X$ and $\alpha \in \mathbb{R}$;
- (v) $\langle -x, y \rangle_p = -\langle x, y \rangle_q$ for all $x, y \in X$;
- (va) $\langle x + y, z \rangle_p \leq \|x\| \|z\| + \langle y, z \rangle_p$ for all $x, y, z \in X$;
- (vaa) The mapping $\langle \cdot, \cdot \rangle_p$ is continuous and subadditive (superadditive) in the first variable for $p = s$ (or $p = i$);
- (vaaa) The normed linear space $(X, \|\cdot\|)$ is smooth at the point $x_0 \in X \setminus \{0\}$ if and only if $\langle y, x_0 \rangle_s = \langle y, x_0 \rangle_i$ for all $y \in X$; in general $\langle y, x \rangle_i \leq \langle y, x \rangle_s$ for all $x, y \in X$;
- (ax) If the norm $\|\cdot\|$ is induced by an inner product $\langle \cdot, \cdot \rangle$, then $\langle y, x \rangle_i = \langle y, x \rangle = \langle y, x \rangle_s$ for all $x, y \in X$.

Applying inequality (HH) for the convex function $f_r(x) = \|x\|^r$, $r \geq 1$ one may deduce the inequality

$$\left\| \frac{x+y}{2} \right\|^r \leq \int_0^1 \|(1-t)x + ty\|^r dt \leq \frac{\|x\|^r + \|y\|^r}{2} \quad (15)$$

for any $x, y \in X$.

Let $(X, \|\cdot\|)$ be a normed linear space and $x = (x_1, \dots, x_n)$, $y = (y_1, \dots, y_n)$ be n -tuples of vectors in X , then for the probability distribution $p = (p_1, \dots, p_n)$ and $r \geq 1$ we have by Theorem 3 for the convex function $f(x) = \|x\|^r$ that

$$\begin{aligned} \sum_{i=1}^n p_i \int_0^1 \|(1-t)x_i + ty_i\|^r dt &\geq \int_0^1 \left\| (1-t) \sum_{i=1}^n p_i x_i + t \sum_{i=1}^n p_i y_i \right\|^r dt \\ &\geq \left\| \sum_{i=1}^n p_i \left(\frac{x_i + y_i}{2} \right) \right\|^r, \end{aligned} \quad (16)$$

$$\begin{aligned} \sum_{i=1}^n p_i \left(\frac{\|x_i\|^r + \|y_i\|^r}{2} \right) &\geq \sum_{i=1}^n \int_0^1 p_i \|(1-t)x_i + ty_i\|^r dt \\ &\geq \sum_{i=1}^n p_i \left\| \frac{x_i + y_i}{2} \right\|^r \geq \left\| \sum_{i=1}^n p_i \left(\frac{x_i + y_i}{2} \right) \right\|^r \end{aligned} \quad (17)$$

and

$$\begin{aligned} \sum_{i=1}^n p_i \left(\frac{\|x_i\|^r + \|y_i\|^r}{2} \right) &\geq \frac{1}{2} \left[\left\| \sum_{i=1}^n p_i x_i \right\|^r + \left\| \sum_{i=1}^n p_i y_i \right\|^r \right] \\ &\geq \int_0^1 \left\| (1-t) \sum_{i=1}^n p_i x_i + t \sum_{i=1}^n p_i y_i \right\|^r dt. \end{aligned} \quad (18)$$

If we use Theorem 4 for the convex function $f(x) = \frac{1}{2} \|x\|^2$ then for $x = (x_1, \dots, x_n)$, $y = (y_1, \dots, y_n)$ n -tuples of vectors in X and for the probability distribution $p = (p_1, \dots, p_n)$

we have

$$\begin{aligned}
0 &\leq \frac{1}{4} \left[\sum_{k=1}^n p_k \left\langle y_k - x_k, \frac{x_k + y_k}{2} \right\rangle_s - \sum_{k=1}^n p_k \left\langle y_k - x_k, \frac{x_k + y_k}{2} \right\rangle_i \right] \\
&\leq \sum_{k=1}^n p_k \int_0^1 \|(1-t)x_k + ty_k\|^2 dt - \sum_{k=1}^n p_k \left\| \frac{x_k + y_k}{2} \right\|^2 \\
&\leq \frac{1}{4} \left[\sum_{k=1}^n p_k \langle y_k - x_k, y_k \rangle_i - \sum_{k=1}^n p_k \langle y_k - x_k, x_k \rangle_s \right], \quad (19)
\end{aligned}$$

and

$$\begin{aligned}
0 &\leq \frac{1}{4} \left[\sum_{k=1}^n p_k \left\langle y_k - x_k, \frac{x_k + y_k}{2} \right\rangle_s - \sum_{k=1}^n p_k \left\langle y_k - x_k, \frac{x_k + y_k}{2} \right\rangle_i \right] \\
&\leq \sum_{k=1}^n p_k \left(\frac{\|x_k\|^2 + \|y_k\|^2}{2} \right) - \sum_{k=1}^n p_k \int_0^1 \|(1-t)x_k + ty_k\|^2 dt \\
&\leq \frac{1}{4} \left[\sum_{k=1}^n p_k \langle y_k - x_k, y_k \rangle_i - \sum_{k=1}^n p_k \langle y_k - x_k, x_k \rangle_s \right]. \quad (20)
\end{aligned}$$

We also have

$$\begin{aligned}
0 &\leq \frac{1}{4} \left[\left\langle \sum_{k=1}^n p_k (y_k - x_k), \sum_{k=1}^n p_k \left(\frac{x_k + y_k}{2} \right) \right\rangle_s \right. \\
&\quad \left. - \left\langle \sum_{k=1}^n p_k (y_k - x_k), \sum_{k=1}^n p_k \left(\frac{x_k + y_k}{2} \right) \right\rangle_i \right] \\
&\leq \int_0^1 \left\| (1-t) \sum_{k=1}^n p_k x_k + t \sum_{k=1}^n p_k y_k \right\|^2 dt - \left\| \sum_{k=1}^n p_k \left(\frac{x_k + y_k}{2} \right) \right\|^2 \\
&\leq \frac{1}{4} \left[\left\langle \sum_{k=1}^n p_k (y_k - x_k), \sum_{k=1}^n p_k y_k \right\rangle_i - \left\langle \sum_{k=1}^n p_k (y_k - x_k), \sum_{k=1}^n p_k x_k \right\rangle_s \right] \quad (21)
\end{aligned}$$

and

$$\begin{aligned}
0 &\leq \frac{1}{4} \left[\left\langle \sum_{k=1}^n p_k (y_k - x_k), \sum_{k=1}^n p_k \left(\frac{x_k + y_k}{2} \right) \right\rangle_s \right. \\
&\quad \left. - \left\langle \sum_{k=1}^n p_k (y_k - x_k), \sum_{k=1}^n p_k \left(\frac{x_k + y_k}{2} \right) \right\rangle_i \right] \\
&\leq \frac{1}{2} \left[\left\| \sum_{k=1}^n p_k x_k \right\|^2 + \left\| \sum_{k=1}^n p_k y_k \right\|^2 \right] - \int_0^1 \left\| (1-t) \sum_{k=1}^n p_k x_k + t \sum_{k=1}^n p_k y_k \right\|^2 dt \\
&\leq \frac{1}{4} \left[\left\langle \sum_{k=1}^n p_k (y_k - x_k), \sum_{k=1}^n p_k y_k \right\rangle_i - \left\langle \sum_{k=1}^n p_k (y_k - x_k), \sum_{k=1}^n p_k x_k \right\rangle_s \right]. \quad (22)
\end{aligned}$$

4. EXAMPLES FOR FUNCTIONS OF A REAL VARIABLE

If $f : I \rightarrow \mathbb{R}$ is convex on the interval I and $p_i \geq 0$, $i \in \{1, \dots, n\}$ with $\sum_{i=1}^n p_i = 1$, then

$$\sum_{i=1}^n p_i \int_0^1 f((1-t)x_i + ty_i) dt \geq \int_0^1 f\left((1-t) \sum_{i=1}^n p_i x_i + t \sum_{i=1}^n p_i y_i\right) dt \quad (23)$$

$$\sum_{i=1}^n p_i \left(\frac{f(x_i) + f(y_i)}{2}\right) \geq \sum_{i=1}^n p_i \int_0^1 f((1-t)x_i + ty_i) dt. \quad (24)$$

If $f : I \rightarrow \mathbb{R}$ is convex and differentiable on the interior of I then for all $x_i \in I$ and $p_i \geq 0$, $i \in \{1, \dots, n\}$ with $\sum_{i=1}^n p_i = 1$, then by (11) and (12) we get

$$\begin{aligned} 0 &\leq \sum_{i=1}^n p_i \int_0^1 f[(1-t)x_i + ty_i] dt - \sum_{i=1}^n p_i f\left(\frac{x_i + y_i}{2}\right) \\ &\leq \frac{1}{8} \sum_{i=1}^n p_i [f'(y_i) - f'(x_i)](y_i - x_i), \end{aligned} \quad (25)$$

and

$$\begin{aligned} 0 &\leq \sum_{i=1}^n p_i \left(\frac{f(x_i) + f(y_i)}{2}\right) - \sum_{i=1}^n p_i \int_0^1 f((1-t)x_i + ty_i) dt \\ &\leq \frac{1}{8} \sum_{i=1}^n p_i [f'(y_i) - f'(x_i)](y_i - x_i). \end{aligned} \quad (26)$$

If $f(t) = \frac{1}{t}$ with $t > 0$, then for $y_i \neq x_i$, $i \in \{1, \dots, n\}$ we have

$$\int_0^1 f((1-t)x_i + ty_i) dt = \int_0^1 \frac{1}{(1-t)x_i + ty_i} dt = \frac{\ln y_i - \ln x_i}{y_i - x_i}$$

and

$$\int_0^1 \left((1-t) \sum_{i=1}^n p_i x_i + t \sum_{i=1}^n p_i y_i \right)^{-1} dt = \frac{\ln(\sum_{i=1}^n p_i x_i) - \ln(\sum_{i=1}^n p_i y_i)}{\sum_{i=1}^n p_i x_i - \sum_{i=1}^n p_i y_i},$$

provided $\sum_{i=1}^n p_i x_i \neq \sum_{i=1}^n p_i y_i$.

From (23) we get

$$\sum_{i=1}^n p_i \frac{\ln y_i - \ln x_i}{y_i - x_i} \geq \frac{\ln(\sum_{i=1}^n p_i x_i) - \ln(\sum_{i=1}^n p_i y_i)}{\sum_{i=1}^n p_i x_i - \sum_{i=1}^n p_i y_i}$$

that is equivalent to

$$\ln \left(\prod_{i=1}^n \left(\frac{y_i}{x_i} \right)^{\frac{p_i}{y_i - x_i}} \right) \geq \ln \left[\left(\frac{\sum_{i=1}^n p_i x_i}{\sum_{i=1}^n p_i y_i} \right)^{\frac{1}{\sum_{i=1}^n p_i x_i - \sum_{i=1}^n p_i y_i}} \right]$$

and to

$$\prod_{i=1}^n \left(\frac{y_i}{x_i} \right)^{\frac{p_i}{y_i - x_i}} \geq \left(\frac{\sum_{i=1}^n p_i x_i}{\sum_{i=1}^n p_i y_i} \right)^{\frac{1}{\sum_{i=1}^n p_i x_i - \sum_{i=1}^n p_i y_i}}. \quad (27)$$

From (24) we get in a similar way that

$$\exp \left[\sum_{i=1}^n p_i \left(\frac{x_i + y_i}{2x_i y_i} \right) \right] \geq \prod_{i=1}^n \left(\frac{y_i}{x_i} \right)^{\frac{p_i}{y_i - x_i}}, \quad (28)$$

from (25) we get

$$1 \leq \frac{\prod_{i=1}^n \left(\frac{y_i}{x_i}\right)^{\frac{p_i}{y_i-x_i}}}{\exp\left[\sum_{i=1}^n p_i \left(\frac{2}{x_i+y_i}\right)\right]} \leq \exp\left(\frac{1}{8} \sum_{i=1}^n p_i \frac{(x_i+y_i)(y_i-x_i)^2}{x_i^2 y_i^2}\right) \quad (29)$$

and from (26) we get

$$1 \leq \frac{\exp\left[\sum_{i=1}^n p_i \left(\frac{x_i+y_i}{2x_i y_i}\right)\right]}{\prod_{i=1}^n \left(\frac{y_i}{x_i}\right)^{\frac{p_i}{y_i-x_i}}} \leq \exp\left(\frac{1}{8} \sum_{i=1}^n p_i \frac{(x_i+y_i)(y_i-x_i)^2}{x_i^2 y_i^2}\right). \quad (30)$$

The interested reader may apply some of the above inequalities for other instances of convex functions such as $f(t) = -\ln t$, $t \ln t$, $\exp t$ etc... and we omit the details.

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