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AN INEQUALITY IMPROVING THE SECOND HERMITE-HADAMARD INEQUALITY FOR CONVEX FUNCTIONS DEFINED ON LINEAR SPACES AND APPLICATIONS FOR SEMI-INNER PRODUCTS

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Abstract

An inequality for convex functions defined on linear spaces is obtained which contains in a particular case a refinement for the second part of the celebrated Hermite-Hadamard inequality. Applications for semi-inner products on normed linear spaces are also provided.

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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 $f: [a, b] \to \mathbb{R}$ and the attached function $g(a, b): [0, 1] \to \mathbb{R}$, $g(a, b)(t) := f[(1 - t) \, a + tb]$, $t \in [0, 1]$.

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

(ii)
$$g'_{+}(a,b)(0) = (\nabla_{+}f(a))(b-a)$$

(iii)
$$g'_{-}(a,b)(1) = (\nabla_{-}f(b))(b-a)$$

where $(\nabla_+ f(x))(y)$ are the Gâteaux lateral derivatives, we recall that

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

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

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



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

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

For other related results see the monograph on line [1].

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

(iv)
$$\langle x, y \rangle_s := (\nabla_+ f_0(y))(x) = \lim_{t \to 0+} \left[\frac{\|y + tx\|^2 - \|y\|^2}{2t} \right];$$

(v)
$$\langle x, y \rangle_i := (\nabla_- f_0(y))(x) = \lim_{s \to 0^-} \left[\frac{\|y + sx\|^2 - \|y\|^2}{2s} \right];$$

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 [2]), assuming that $p, q \in \{s, i\}$ and $p \neq q$:

(a)
$$\langle x, x \rangle_n = ||x||^2$$
 for all $x \in X$;

(aa)
$$\langle \alpha x, \beta y \rangle_n = \alpha \beta \langle x, y \rangle_n$$
 if $\alpha, \beta \ge 0$ and $x, y \in X$;

(aaa)
$$\left| \langle x, y \rangle_p \right| \le \|x\| \, \|y\|$$
 for all $x, y \in X$;

(av)
$$\langle \alpha x + y, x \rangle_n = \alpha \langle x, x \rangle_n + \langle y, x \rangle_n$$
 if $x, y \in X$ and $\alpha \in \mathbb{R}$;



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(v)
$$\langle -x, y \rangle_p = -\langle x, y \rangle_q$$
 for all $x, y \in X$;

(va)
$$\langle x+y,z\rangle_n \leq ||x|| \, ||z|| + \langle y,z\rangle_n$$
 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 $f_0(x) = \frac{1}{2} ||x||^2$, one may deduce the inequality

(1.1)
$$\left\| \frac{x+y}{2} \right\|^2 \le \int_0^1 \left\| (1-t) x + ty \right\|^2 dt \le \frac{\left\| x \right\|^2 + \left\| y \right\|^2}{2}$$

for any $x, y \in X$. The same (HH) inequality applied for $f_1(x) = ||x||$, will give the following refinement of the triangle inequality:

(1.2)
$$\left\| \frac{x+y}{2} \right\| \le \int_0^1 \|(1-t)x + ty\| \, dt \le \frac{\|x\| + \|y\|}{2}, \ x, y \in X.$$

In this paper we point out an integral inequality for convex functions which is related to the first Hermite-Hadamard inequality in (HH) and investigate its applications for semi-inner products in normed linear spaces.



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2. The Results

We start with the following lemma which is also of interest in itself.

Lemma 2.1. Let $h : [\alpha, \beta] \subset \mathbb{R} \to \mathbb{R}$ be a convex function on $[\alpha, \beta]$. Then for any $\gamma \in [\alpha, \beta]$ one has the inequality

$$(2.1) \quad \frac{1}{2} \left[(\beta - \gamma)^2 h'_+(\gamma) - (\gamma - \alpha)^2 h'_-(\gamma) \right]$$

$$\leq (\gamma - \alpha) h(\alpha) + (\beta - \gamma) h(\beta) - \int_{\alpha}^{\beta} h(t) dt$$

$$\leq \frac{1}{2} \left[(\beta - \gamma)^2 h'_-(\beta) - (\gamma - \alpha)^2 h'_+(\alpha) \right].$$

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

Proof. It is easy to see that for any locally absolutely continuous function $h:(\alpha,\beta)\to\mathbb{R}$, we have the identity

(2.2)
$$\int_{\alpha}^{\beta} (t - \gamma) h'(t) dt = (\gamma - \alpha) h(\alpha) + (\beta - \gamma) h(\beta) - \int_{\alpha}^{\beta} h(t) dt$$

for any $\gamma \in (\alpha, \beta)$, where h' is the derivative of h which exists a.e. on (α, β) .

Since h is convex, then it is locally Lipschitzian and thus (2.2) holds. Moreover, for any $\gamma \in (\alpha, \beta)$, we have the inequalities

(2.3)
$$h'(t) \le h'_{-}(\gamma) \text{ for a.e. } t \in [\alpha, \gamma]$$



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and

$$(2.4) h'(t) \ge h'_{+}(\gamma) \text{ for a.e. } t \in [\gamma, \beta].$$

If we multiply (2.3) by $\gamma - t \ge 0$, $t \in [\alpha, \gamma]$ and integrate on $[\alpha, \gamma]$, we get

(2.5)
$$\int_{\alpha}^{\gamma} (\gamma - t) h'(t) dt \leq \frac{1}{2} (\gamma - \alpha)^2 h'_{-}(\gamma)$$

and if we multiply (2.4) by $t - \gamma \ge 0$, $t \in [\gamma, \beta]$, and integrate on $[\gamma, \beta]$, we also have

(2.6)
$$\int_{\gamma}^{\beta} (t - \gamma) h'(t) dt \ge \frac{1}{2} (\beta - \gamma)^2 h'_{+}(\gamma).$$

Now, if we subtract (2.5) from (2.6) and use the representation (2.2), we deduce the first inequality in (2.1).

If we assume that the first inequality (2.1) holds with a constant C > 0 instead of $\frac{1}{2}$, i.e.,

$$(2.7) \quad C\left[\left(\beta - \gamma\right)^{2} h'_{+}(\gamma) - \left(\gamma - \alpha\right)^{2} h'_{-}(\gamma)\right] \\ \leq \left(\gamma - \alpha\right) h\left(\alpha\right) + \left(\beta - \gamma\right) h\left(\beta\right) - \int_{0}^{\beta} h\left(t\right) dt$$

and take the convex function $h_0(t) := k \left| t - \frac{\alpha + \beta}{2} \right|, k > 0, t \in [\alpha, \beta]$, then

$$h'_{0+}\left(\frac{\alpha+\beta}{2}\right) = k,$$

$$h'_{0-}\left(\frac{\alpha+\beta}{2}\right) = -k,$$



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$$h_0(\alpha) = \frac{k(\beta - \alpha)}{2} = h_0(\beta),$$

$$\int_{\alpha}^{\beta} h_0(t) dt = \frac{1}{4} k(\beta - \alpha)^2,$$

and the inequality (2.7) becomes, for $\gamma = \frac{\alpha + \beta}{2}$,

$$C\left[\frac{1}{4}(\beta - \alpha)^2 k + \frac{1}{4}(\beta - \alpha)^2 k\right] \le \frac{1}{4}k(\beta - \alpha)^2,$$

giving $C \leq \frac{1}{2}$, which proves the sharpness of the constant $\frac{1}{2}$ in the first inequality in (2.1).

If either $h'_{+}(\alpha) = -\infty$ or $h'_{-}(\beta) = -\infty$, then the second inequality in (2.1) holds true.

Assume that $h'_{+}\left(\alpha\right)$ and $h'_{-}\left(\beta\right)$ are finite. Since h is convex on $\left[\alpha,\beta\right]$, we have

(2.8)
$$h'(t) \ge h'_+(\alpha)$$
 for a.e. $t \in [\alpha, \gamma]$ $(\gamma \text{ may be equal to } \beta)$

and

$$(2.9) \qquad h'\left(t\right) \leq h'_{-}\left(\beta\right) \text{ for a.e. } t \in \left[\gamma,\beta\right] \ \, \left(\gamma \text{ may be equal to } \alpha\right).$$

If we multiply (2.8) by $\gamma - t \ge 0$, $t \in [\alpha, \gamma]$ and integrate on $[\alpha, \gamma]$, then we deduce

(2.10)
$$\int_{\alpha}^{\gamma} (\gamma - t) h'(t) dt \ge \frac{1}{2} (\gamma - \alpha)^2 h'_{+}(\alpha)$$



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and if we multiply (2.9) by $t - \gamma \ge 0$, $t \in [\gamma, \beta]$, and integrate on $[\gamma, \beta]$, then we also have

(2.11)
$$\int_{\gamma}^{\beta} (t - \gamma) h'(t) dt \leq \frac{1}{2} (\beta - \gamma)^2 h'_{-}(\beta).$$

Finally, if we subtract (2.10) from (2.11) and use the representation (2.2), we deduce the second part of (2.1).

Now, assume that the second inequality in (2.1) holds with a constant D > 0 instead of $\frac{1}{2}$, i.e.,

(2.12)
$$(\gamma - \alpha) f(\alpha) + (\beta - \gamma) f(\beta) - \int_{\alpha}^{\beta} f(t) dt$$

$$\geq D \left[(\beta - \gamma)^{2} f'_{-}(\beta) - (\gamma - \alpha)^{2} f'_{+}(\alpha) \right].$$

If we consider the convex function h_0 given above, then we have $h'_{-}(\beta) = k$, $h'_{+}(\alpha) = -k$ and by (2.12) applied for h_0 and $x = \frac{\alpha + \beta}{2}$ we get

$$\frac{1}{4}k(\beta - \alpha)^{2} \le D\left[\frac{1}{4}k(\beta - \alpha)^{2} + \frac{1}{4}k(\beta - \alpha)^{2}\right],$$

giving $D \geq \frac{1}{2}$, and the sharpness of the constant $\frac{1}{2}$ is proved.

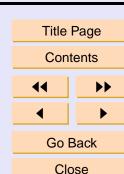
Corollary 2.2. With the assumptions of Lemma 2.1 and if $\gamma \in (\alpha, \beta)$ is a point of differentiability for h, then

(2.13)
$$\left(\frac{\alpha+\beta}{2}-\gamma\right)h'(\gamma)$$



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$$\leq \left(\frac{\gamma-\alpha}{\beta-\alpha}\right)h\left(\alpha\right) + \left(\frac{\beta-\gamma}{\beta-\alpha}\right)h\left(\beta\right) - \frac{1}{\beta-\alpha}\int_{\alpha}^{\beta}h\left(t\right)dt.$$

Now, recall that the following inequality, which is well known in the literature as the Hermite-Hadamard inequality for convex functions, holds

$$(2.14) h\left(\frac{\alpha+\beta}{2}\right) \leq \frac{1}{\beta-\alpha} \int_{\alpha}^{\beta} h\left(t\right) dt \leq \frac{h\left(\alpha\right)+h\left(\beta\right)}{2}.$$

The following corollary provides some bounds for the difference

$$\frac{h(\alpha) + h(\beta)}{2} - \frac{1}{\beta - \alpha} \int_{\alpha}^{\beta} h(t) dt.$$

Corollary 2.3. *Let* $h : [\alpha, \beta] \to \mathbb{R}$ *be a convex function on* $[\alpha, \beta]$ *. Then we have the inequality*

$$(2.15) 0 \leq \frac{1}{8} \left[h'_{+} \left(\frac{\alpha + \beta}{2} \right) - h'_{-} \left(\frac{\alpha + \beta}{2} \right) \right] (\beta - \alpha)$$

$$\leq \frac{h(\alpha) + h(\beta)}{2} - \frac{1}{\beta - \alpha} \int_{\alpha}^{\beta} h(t) dt$$

$$\leq \frac{1}{8} \left[h'_{-}(\beta) - h'_{+}(\alpha) \right] (\beta - \alpha).$$

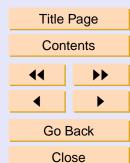
The constant $\frac{1}{8}$ is sharp in both inequalities.

We are now able to state the corresponding result for convex functions defined on linear spaces.



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Theorem 2.4. Let X be a linear space, $a, b \in X$, $a \neq b$ and $f : [a, b] \subset X \to \mathbb{R}$ be a convex function on the segment [a, b]. Then for any $s \in (0, 1)$ one has the inequality

$$(2.16) \frac{1}{2} \left[(1-s)^2 \left(\nabla_+ f \left[(1-s) \, a + sb \right] \right) (b-a) - s^2 \left(\nabla_- f \left[(1-s) \, a + sb \right] \right) (b-a) \right]$$

$$\leq (1-s) \, f \left(a \right) + s f \left(b \right) - \int_0^1 f \left[(1-t) \, a + tb \right] dt$$

$$\leq \frac{1}{2} \left[(1-s)^2 \left(\nabla_- f \left(b \right) \right) (b-a) - s^2 \left(\nabla_+ f \left(a \right) \right) (b-a) \right].$$

The constant $\frac{1}{2}$ is sharp in both inequalities.

The second inequality also holds for s = 0 or s = 1.

Proof. Follows by Lemma 2.1 applied for the convex function

$$h(t) = g(a,b)(t) = f[(1-t)a + tb], \quad t \in [0,1],$$

and for the choices $\alpha = 0$, $\beta = 1$, and $\gamma = s$.

Corollary 2.5. If $f:[a,b] \to \mathbb{R}$ is as in Theorem 2.4 and Gâteaux differentiable in $c:=(1-\lambda)$ $a+\lambda b$, $\lambda \in (0,1)$ along the direction (b-a), then we have the inequality:

$$(2.17) \quad \left(\frac{1}{2} - \lambda\right) \left(\nabla f\left(c\right)\right) \left(b - a\right)$$

$$\leq \left(1 - \lambda\right) f\left(a\right) + \lambda f\left(b\right) - \int_{0}^{1} f\left[\left(1 - t\right) a + tb\right] dt.$$



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The following result related to the second Hermite-Hadamard inequality for functions defined on linear spaces also holds.

Corollary 2.6. If f is as in Theorem 2.4, then

$$(2.18) \quad \frac{1}{8} \left[\nabla_{+} f\left(\frac{a+b}{2}\right) (b-a) - \nabla_{-} f\left(\frac{a+b}{2}\right) (b-a) \right]$$

$$\leq \frac{f(a) + f(b)}{2} - \int_{0}^{1} f\left[(1-t) a + tb \right] dt$$

$$\leq \frac{1}{8} \left[\left(\nabla_{-} f(b) \right) (b-a) - \left(\nabla_{+} f(a) \right) (b-a) \right].$$

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

Now, let $\Omega \subset \mathbb{R}^n$ be an open convex set in \mathbb{R}^n .

If $F:\Omega\to\mathbb{R}$ is a differentiable convex function on $\Omega,$ then, obviously, for any $\bar c\in\Omega$ we have

$$\nabla F(\bar{c})(\bar{y}) = \sum_{i=1}^{n} \frac{\partial F(\bar{c})}{\partial x_i} \cdot y_i, \ \bar{y} \in \mathbb{R}^n,$$

where $\frac{\partial F}{\partial x_i}$ are the partial derivatives of F with respect to the variable x_i $(i=1,\ldots,n)$.

Using (2.16), we may state that

$$(2.19)\left(\frac{1}{2} - \lambda\right) \sum_{i=1}^{n} \frac{\partial F\left(\lambda \bar{a} + (1 - \lambda)\bar{b}\right)}{\partial x_{i}} \cdot (b_{i} - a_{i})$$



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$$\leq (1 - \lambda) F(\bar{a}) + \lambda F(\bar{b}) - \int_0^1 F[(1 - t) \bar{a} + t \bar{b}] dt$$

$$\leq \frac{1}{2} \left[(1 - \lambda)^2 \sum_{i=1}^n \frac{\partial F(\bar{b})}{\partial x_i} \cdot (b_i - a_i) - \lambda^2 \sum_{i=1}^n \frac{\partial F(\bar{a})}{\partial x_i} \cdot (b_i - a_i) \right]$$

for any $\bar{a}, \bar{b} \in \Omega$ and $\lambda \in (0, 1)$. In particular, for $\lambda = \frac{1}{2}$, we get

$$(2.20) 0 \leq \frac{F(\bar{a}) + F(b)}{2} - \int_{0}^{1} F[(1-t)\bar{a} + t\bar{b}] dt$$
$$\leq \frac{1}{8} \sum_{i=1}^{n} \left(\frac{\partial F(\bar{b})}{\partial x_{i}} - \frac{\partial F(\bar{a})}{\partial x_{i}} \right) \cdot (b_{i} - a_{i}).$$

In (2.20) the constant $\frac{1}{8}$ is sharp.



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3. Applications for Semi-Inner Products

Let $(X, \|\cdot\|)$ be a real normed linear space. We may state the following results for the semi-inner products $\langle \cdot, \cdot \rangle_i$ and $\langle \cdot, \cdot \rangle_s$.

Proposition 3.1. For any $x, y \in X$ and $\sigma \in (0, 1)$ we have the inequalities:

$$(3.1) \qquad (1 - \sigma)^{2} \langle y - x, (1 - \sigma) x + \sigma y \rangle_{s} - \sigma^{2} \langle y - x, (1 - \sigma) x + \sigma y \rangle_{i}$$

$$\leq (1 - \sigma) \|x\|^{2} + \sigma \|y\|^{2} - \int_{0}^{1} \|(1 - t) x + ty\|^{2} dt$$

$$\leq (1 - \sigma)^{2} \langle y - x, y \rangle_{i} - \sigma^{2} \langle y - x, y \rangle_{s}.$$

The second inequality in (3.1) also holds for $\sigma = 0$ or $\sigma = 1$.

The proof is obvious by Theorem 2.4 applied for the convex function $f(x) = \frac{1}{2} ||x||^2$, $x \in X$.

If the space is *smooth*, then we may put $[x,y] = \langle x,y \rangle_i = \langle x,y \rangle_s$ for each $x,y \in X$ and the first inequality in (3.1) becomes

(3.2)
$$(1-2\sigma)[y-x,(1-\sigma)x+\sigma y]$$

$$\leq (1-\sigma)\|x\|^2 + \sigma\|y\|^2 - \int_0^1 \|(1-t)x+ty\|^2 dt.$$

An interesting particular case one can get from (3.1) is the one for $\sigma = \frac{1}{2}$,

(3.3)
$$0 \le \frac{1}{8} \left[\langle y - x, y + x \rangle_s - \langle y - x, y + x \rangle_i \right] \\ \le \frac{\|x\|^2 + \|y\|^2}{2} - \int_0^1 \|(1 - t)x + ty\|^2 dt$$



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$$\leq \frac{1}{4} \left[\langle y - x, y \rangle_i - \langle y - x, x \rangle_s \right].$$

The inequality (3.3) provides a refinement and a counterpart for the second inequality in (1.1).

If we consider now two linearly independent vectors $x, y \in X$ and apply Theorem 2.4 for $f(x) = ||x||, x \in X$, then we get

Proposition 3.2. For any linearly independent vectors $x, y \in X$ and $\sigma \in (0, 1)$, one has the inequalities:

$$(3.4) \quad \frac{1}{2} \left[(1 - \sigma)^2 \frac{\langle y - x, (1 - \sigma) x + \sigma y \rangle_s}{\|(1 - \sigma) x + \sigma y\|} - \sigma^2 \frac{\langle y - x, (1 - \sigma) x + \sigma y \rangle_i}{\|(1 - \sigma) x + \sigma y\|} \right]$$

$$\leq (1 - \sigma) \|x\| + \sigma \|y\| - \int_0^1 \|(1 - t) x + ty\| dt$$

$$\leq \frac{1}{2} \left[(1 - \sigma)^2 \frac{\langle y - x, y \rangle_i}{\|y\|} - \sigma^2 \frac{\langle y - x, x \rangle_s}{\|x\|} \right].$$

The second inequality also holds for $\sigma = 0$ or $\sigma = 1$.

We note that if the space is smooth, then we have

(3.5)
$$\left(\frac{1}{2} - \sigma\right) \cdot \frac{[y - x, (1 - \sigma)x + \sigma y]}{\|(1 - \sigma)x + \sigma y\|}$$

$$\leq (1 - \sigma)\|x\| + \sigma\|y\| - \int_0^1 \|(1 - t)x + ty\| dt$$



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and for $\sigma = \frac{1}{2}$, (3.4) will give the simple inequality

$$(3.6) 0 \leq \frac{1}{8} \left[\left\langle y - x, \frac{\frac{x+y}{2}}{\left\| \frac{x+y}{2} \right\|} \right\rangle_{s} - \left\langle y - x, \frac{\frac{x+y}{2}}{\left\| \frac{x+y}{2} \right\|} \right\rangle_{i} \right]$$

$$\leq \frac{\|x\| + \|y\|}{2} - \int_{0}^{1} \|(1-t)x + ty\| dt$$

$$\leq \frac{1}{8} \left[\left\langle y - x, \frac{y}{\|y\|} \right\rangle_{i} - \left\langle y - x, \frac{x}{\|x\|} \right\rangle_{s} \right].$$

The inequality (3.6) provides a refinement and a conterpart of the second inequality in (1.2).

Moreover, if we assume that $(H, \langle \cdot, \cdot \rangle)$ is an inner product space, then by (3.6) we get for any $x, y \in H$ with ||x|| = ||y|| = 1 that

(3.7)
$$0 \le 1 - \int_0^1 \|(1-t)x + ty\| dt \le \frac{1}{8} \|y - x\|^2.$$

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

Indeed, if we choose $H = \mathbb{R}$, $\langle a, b \rangle = a \cdot b$, x = -1, y = 1, then we get equality in (3.7).

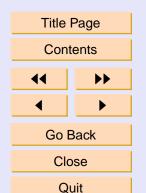
We give now some examples.

1. Let $\ell^2(\mathbb{K})$, $\mathbb{K} = \mathbb{C}$, \mathbb{R} ; be the Hilbert space of sequences $x = (x_i)_{i \in \mathbb{N}}$ with



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 $\sum_{i=0}^{\infty} |x_i|^2 < \infty$. Then, by (3.7), we have the inequalities

(3.8)
$$0 \leq 1 - \int_0^1 \left(\sum_{i=0}^\infty |(1-t)x_i + ty_i|^2 \right)^{\frac{1}{2}} dt$$
$$\leq \frac{1}{8} \cdot \sum_{i=0}^\infty |y_i - x_i|^2,$$

for any $x, y \in \ell^2(\mathbb{K})$ provided $\sum_{i=0}^{\infty} |x_i|^2 = \sum_{i=0}^{\infty} |y_i|^2 = 1$.

2. Let μ be a positive measure, $L_2(\Omega)$ the Hilbert space of μ -measurable functions on Ω with complex values that are 2-integrable on Ω , i.e., $f \in L_2(\Omega)$ iff $\int_{\Omega} |f(t)|^2 d\mu(t) < \infty$. Then, by (3.7), we have the inequalities

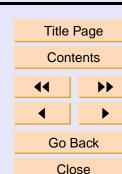
(3.9)
$$0 \leq 1 - \int_{0}^{1} \left(\int_{\Omega} |(1 - \lambda) f(t) + \lambda g(t)|^{2} d\mu(t) \right)^{\frac{1}{2}} d\lambda$$
$$\leq \frac{1}{8} \cdot \int_{\Omega} |f(t) - g(t)|^{2} d\mu(t)$$

for any $f, g \in L_2(\Omega)$ provided $\int_{\Omega} |f(t)|^2 d\mu(t) = \int_{\Omega} |g(t)|^2 d\mu(t) = 1$.



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