

BOUNDS FOR THE NORMALIZED JENSEN–MERCER FUNCTIONAL

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*Dedicated to Professor Josip Pečarić
 on the occasion of his 60th birthday*

Abstract. We introduce the normalized Jensen-Mercer functional

$$\mathcal{M}_n(f, \mathbf{x}, \mathbf{p}) = f(a) + f(b) - \sum_{i=1}^n p_i f(x_i) - f\left(a + b - \sum_{i=1}^n p_i x_i\right)$$

and establish the inequalities of type $M\mathcal{M}_n(f, \mathbf{x}, \mathbf{q}) \geq \mathcal{M}_n(f, \mathbf{x}, \mathbf{p}) \geq m\mathcal{M}_n(f, \mathbf{x}, \mathbf{q})$, where f is a convex function, $\mathbf{x} = (x_1, \dots, x_n)$ and m and M are real numbers satisfying certain conditions. We prove them for the case when \mathbf{p} and \mathbf{q} are nonnegative n -tuples and when \mathbf{p} and \mathbf{q} satisfy the conditions for the Jensen-Steffensen inequality. We also give their integral versions in both cases.

1. Introduction

In paper [4] A. McD. Mercer proved the following variant of Jensen’s inequality, to which we will refer as to the ”Jensen-Mercer inequality”.

THEOREM A. *Let $[a, b]$ be an interval in \mathbb{R} and $x_1, \dots, x_n \in [a, b]$. Let w_1, \dots, w_n be nonnegative real numbers such that $W_n = \sum_{i=1}^n w_i > 0$. If $f : [a, b] \rightarrow \mathbb{R}$ is a convex function, then*

$$f\left(a + b - \frac{1}{W_n} \sum_{i=1}^n w_i x_i\right) \leq f(a) + f(b) - \frac{1}{W_n} \sum_{i=1}^n w_i f(x_i). \quad (1.1)$$

In paper [1] is proved that (1.1) remains valid even in the case when the condition ” $\mathbf{w} = (w_1, \dots, w_n)$ is nonnegative n -tuple” is somewhat relaxed. More precisely the following is true.

THEOREM B. *Let $[a, b]$ be an interval in \mathbb{R} and $\mathbf{x} = (x_1, \dots, x_n) \in [a, b]^n$ be a monotonic n -tuple. Let $\mathbf{w} = (w_1, \dots, w_n)$ be a real n -tuple such that*

$$0 \leq W_k \leq W_n \quad (k = 1, \dots, n), \quad W_n > 0, \quad (1.2)$$

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where $W_k = \sum_{i=1}^k w_i$ ($k = 1, 2, \dots, n$). If $f : [a, b] \rightarrow \mathbb{R}$ is a convex function, then (1.1) holds.

Let \mathcal{P}_n denotes the set of all nonnegative real n -tuples (p_1, \dots, p_n) with the property $\sum_{i=1}^n p_i = 1$. For any convex function $f : [a, b] \rightarrow \mathbb{R}$ and for any choice of n -tuples $\mathbf{x} = (x_1, \dots, x_n) \in [a, b]^n$ and $\mathbf{p} = (p_1, \dots, p_n) \in \mathcal{P}_n$ we define

$$\mathcal{M}_n(f, \mathbf{x}, \mathbf{p}) := f(a) + f(b) - \sum_{i=1}^n p_i f(x_i) - f\left(a + b - \sum_{i=1}^n p_i x_i\right) \quad (1.3)$$

and we call it the normalized Jensen-Mercer functional. For a fixed function f and n -tuple \mathbf{x} , $\mathcal{M}_n(f, \mathbf{x}, \cdot)$ can be observed as a function on \mathcal{P}_n . Note that \mathcal{P}_n is obviously a convex subset in \mathbb{R}^n and because of Theorem A, $\mathcal{M}_n(f, \mathbf{x}, \mathbf{p}) \geq 0$ for all $\mathbf{p} \in \mathcal{P}_n$.

In Section 2 we establish the inequalities of type $m\mathcal{M}_n(f, \mathbf{x}, \mathbf{q}) \geq \mathcal{M}_n(f, \mathbf{x}, \mathbf{p}) \geq M\mathcal{M}_n(f, \mathbf{x}, \mathbf{q})$, where m and M are real constants satisfying certain conditions. We prove them for the case when \mathbf{p} and \mathbf{q} are nonnegative n -tuples and when \mathbf{p} and \mathbf{q} satisfy the conditions for the Jensen-Steffensen inequality. In Section 4 we give the integral versions of all results from Section 2 and 3.

2. Bounds for the normalized Jensen-Mercer functional

We assume the notations from introduction.

THEOREM 1. Let $\mathbf{p} = (p_1, \dots, p_n)$ and $\mathbf{q} = (q_1, \dots, q_n)$ be two n -tuples from \mathcal{P}_n . Let m and M be any real constants such that

$$m \geq 0, \quad p_i - mq_i \geq 0, \quad Mq_i - p_i \geq 0 \quad (i = 1, \dots, n). \quad (2.1)$$

If $f : [a, b] \rightarrow \mathbb{R}$ is a convex function and $\mathbf{x} = (x_1, \dots, x_n)$ is any n -tuple from $[a, b]^n$, then

$$M\mathcal{M}_n(f, \mathbf{x}, \mathbf{q}) \geq \mathcal{M}_n(f, \mathbf{x}, \mathbf{p}) \geq m\mathcal{M}_n(f, \mathbf{x}, \mathbf{q}). \quad (2.2)$$

Proof. Suppose that $\mathbf{p}, \mathbf{q} \in \mathcal{P}_n$ and $m, M \in \mathbb{R}$ satisfy (2.1). From $p_i - mq_i \geq 0$ ($i = 1, \dots, n$) follows that $1 - m = \sum_{i=1}^n (p_i - mq_i) \geq 0$ i.e., $m \leq 1$, and from $Mq_i - p_i \geq 0$ ($i = 1, \dots, n$) follows that $M - 1 = \sum_{i=1}^n (Mq_i - p_i) \geq 0$ i.e., $M \geq 1$. If $m = 1$ or $M = 1$, then $\mathbf{p} = \mathbf{q}$ and (2.2) obviously holds. Hence, it remains to consider the case when $m < 1$ and $M > 1$.

Applying Theorem A with $w_i := p_i - mq_i$ and using the convexity of the function f we obtain the right inequality in (2.2).

Similarly, applying Theorem A with $w_i := Mq_i - p_i$ and using the convexity of the function f we obtain the left inequality in (2.2). \square

COROLLARY 1. Let $\mathbf{p} = (p_1, \dots, p_n)$ and $\mathbf{q} = (q_1, \dots, q_n)$ be two n -tuples from \mathcal{P}_n such that $q_i > 0$ ($i = 1, \dots, n$). Let

$$m = m(\mathbf{p}, \mathbf{q}) := \min_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\}, \quad M = M(\mathbf{p}, \mathbf{q}) := \max_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\}. \quad (2.3)$$

If $f : [a, b] \rightarrow \mathbb{R}$ is a convex function and $\mathbf{x} = (x_1, \dots, x_n)$ is any n -tuple from $[a, b]^n$, then (2.2) holds.

Proof. Obviously $m \geq 0$ and

$$\frac{p_i}{q_i} - m \geq 0, \quad M - \frac{p_i}{q_i} \geq 0 \quad (i = 1, \dots, n),$$

which implies

$$p_i - m q_i \geq 0, \quad M q_i - p_i \geq 0 \quad (i = 1, \dots, n).$$

Hence, m and M satisfy the conditions of Theorem 1. \square

3. Bounds for the normalized Jensen-Mercer functional under the Jensen-Steffensen conditions

Let $\tilde{\mathcal{P}}_n$ denotes the set of all real n -tuples $\mathbf{p} = (p_1, \dots, p_n)$ satisfying the following Jensen-Steffensen conditions

$$0 \leq P_k \leq 1 \quad (k = 1, \dots, n-1), \quad P_n = 1, \tag{3.1}$$

where $P_k := \sum_{i=1}^k p_i \quad (k = 1, \dots, n)$. Since any n -tuple \mathbf{p} from \mathcal{P}_n obviously satisfies (3.1), $\mathcal{P}_n \subseteq \tilde{\mathcal{P}}_n$. Notice that $\tilde{\mathcal{P}}_n$ is also a convex subset of \mathbb{R}^n .

Let $f : [a, b] \rightarrow \mathbb{R}$ be a convex function, $\mathbf{x} = (x_1, \dots, x_n) \in [a, b]^n$ any monotonic n -tuple and $\mathbf{p} = (p_1, \dots, p_n) \in \tilde{\mathcal{P}}_n$. Then $\sum_{i=1}^n p_i x_i \in [a, b]$ (see for example [1]) and $\mathcal{M}_n(f, \mathbf{x}, \mathbf{p})$ is well defined. Also, because of Theorem B, $\mathcal{M}_n(f, \mathbf{x}, \mathbf{p}) \geq 0$ for all $\mathbf{p} \in \tilde{\mathcal{P}}_n$.

THEOREM 2. Let $\mathbf{p} = (p_1, \dots, p_n)$ and $\mathbf{q} = (q_1, \dots, q_n)$ be two n -tuples from $\tilde{\mathcal{P}}_n$. Let m and M be any real constants such that

$$m \geq 0, \quad P_k - m Q_k \geq 0, \quad (1 - P_k) - m(1 - Q_k) \geq 0 \quad (k = 1, \dots, n-1) \tag{3.2}$$

and

$$M Q_k - P_k \geq 0, \quad M(1 - Q_k) - (1 - P_k) \geq 0 \quad (k = 1, \dots, n-1), \tag{3.3}$$

where $P_k = \sum_{i=1}^k p_i$, $Q_k = \sum_{i=1}^k q_i$. If $f : [a, b] \rightarrow \mathbb{R}$ is a convex function and $\mathbf{x} = (x_1, \dots, x_n) \in [a, b]^n$ is any monotonic n -tuple, then

$$M \mathcal{M}_n(f, \mathbf{x}, \mathbf{q}) \geq \mathcal{M}_n(f, \mathbf{x}, \mathbf{p}) \geq m \mathcal{M}_n(f, \mathbf{x}, \mathbf{q}). \tag{3.4}$$

Proof. Assume that $\mathbf{p}, \mathbf{q} \in \tilde{\mathcal{P}}_n$ and $m, M \in \mathbb{R}$ satisfy (3.2) and (3.3). From (3.2) follows that it has to be $m \leq 1$, and from (3.3) follows that it has to be $M \geq 1$. If $m = 1$ or $M = 1$, then $\mathbf{p} = \mathbf{q}$ and (3.4) obviously holds. Hence, it remains to consider the case when $m < 1$ and $M > 1$.

To prove the right inequality in (3.4) we consider the n -tuple $\mathbf{w} = (w_1, \dots, w_n)$ defined by $w_i := p_i - m q_i \quad (i = 1, \dots, n)$. From (3.2) follows that \mathbf{w} satisfies conditions

(1.2). Now, we follow our proof of Theorem 1, but instead of using Theorem A we use Theorem B.

To prove the left inequality in (3.4) we consider the n -tuple $\mathbf{w} = (w_1, \dots, w_n)$ defined by $w_i := Mq_i - p_i$, ($i = 1, \dots, n$). From (3.3) follows that \mathbf{w} satisfies conditions (1.2). Again, we follow our proof of Theorem 1, but using Theorem B instead of Theorem A. \square

COROLLARY 2. *Let $\mathbf{p} = (p_1, \dots, p_n)$ and $\mathbf{q} = (q_1, \dots, q_n)$ be two n -tuples from $\tilde{\mathcal{P}}_n$. For $k \in \{1, \dots, n\}$ denote $P_k := \sum_{i=1}^k p_i$, $Q_k := \sum_{i=1}^k q_i$. Assume that $0 < Q_k < 1$ for all $k \in \{1, \dots, n-1\}$ and define*

$$\tilde{m} = \tilde{m}(\mathbf{p}, \mathbf{q}) := \min \left\{ \frac{P_k}{Q_k}, \frac{1-P_k}{1-Q_k} : k = 1, \dots, n-1 \right\}, \quad (3.5)$$

$$\tilde{M} = \tilde{M}(\mathbf{p}, \mathbf{q}) := \max \left\{ \frac{P_k}{Q_k}, \frac{1-P_k}{1-Q_k} : k = 1, \dots, n-1 \right\}. \quad (3.6)$$

If $f : [a, b] \rightarrow \mathbb{R}$ is a convex function and if $\mathbf{x} = (x_1, \dots, x_n) \in [a, b]^n$ is any monotonic n -tuple, then

$$\tilde{M} \mathcal{M}_n(f, \mathbf{x}, \mathbf{q}) \geq \mathcal{M}_n(f, \mathbf{x}, \mathbf{p}) \geq \tilde{m} \mathcal{M}_n(f, \mathbf{x}, \mathbf{q}). \quad (3.7)$$

Proof. Since $0 < Q_k < 1$ for all $k \in \{1, \dots, n-1\}$, \tilde{m} and \tilde{M} are well defined and obviously (3.2) and (3.3) are satisfied for $m = \tilde{m}$ and $M = \tilde{M}$. Therefore we can apply Theorem 2 to obtain (3.7). \square

We can consider the uniform distribution $\mathbf{u} = (\frac{1}{n}, \dots, \frac{1}{n})$ and the corresponding nonweighted Jensen-Mercer functional

$$\mathcal{M}_n(f, \mathbf{x}) := \mathcal{M}_n(f, \mathbf{x}, \mathbf{u}) = f(a) + f(b) - \frac{1}{n} \sum_{i=1}^n f(x_i) - f\left(a + b - \frac{1}{n} \sum_{i=1}^n x_i\right).$$

Then we can state the following special case of Corollary 2.

COROLLARY 3. *Let $\mathbf{p} = (p_1, \dots, p_n)$ be n -tuple from $\tilde{\mathcal{P}}_n$. For $k \in \{1, \dots, n\}$ denote $P_k := \sum_{i=1}^k p_i$ and define*

$$\tilde{m}_0 := n \cdot \min \left\{ \frac{P_k}{k}, \frac{1-P_k}{n-k} : k = 1, \dots, n-1 \right\},$$

$$\tilde{M}_0 := n \cdot \max \left\{ \frac{P_k}{k}, \frac{1-P_k}{n-k} : k = 1, \dots, n-1 \right\}.$$

If $f : [a, b] \rightarrow \mathbb{R}$ is a convex function and if $\mathbf{x} = (x_1, \dots, x_n) \in [a, b]^n$ is any monotonic n -tuple, then

$$\tilde{M}_0 \mathcal{M}_n(f, \mathbf{x}) \geq \mathcal{M}_n(f, \mathbf{x}, \mathbf{p}) \geq \tilde{m}_0 \mathcal{M}_n(f, \mathbf{x}).$$

Next, we show that Theorem 2 in some way provides an improvement of Corollary 1. Denote by Π_n the set of all permutations of $(1, 2, \dots, n)$. Suppose that $\pi = (\pi(1), \pi(2), \dots, \pi(n)) \in \Pi_n$ and denote $\mathbf{a}_\pi := (a_{\pi(1)}, a_{\pi(2)}, \dots, a_{\pi(n)})$ for any n -tuple $\mathbf{a} = (a_1, a_2, \dots, a_n)$. First we prove one simple auxiliary result.

LEMMA 1. Let $\mathbf{p} = (p_1, \dots, p_n)$ and $\mathbf{q} = (q_1, \dots, q_n)$ be two nonnegative n -tuples from \mathcal{P}_n . If $q_i > 0$ for all $i \in \{1, \dots, n\}$, then $\tilde{m}(\mathbf{p}, \mathbf{q})$ and $\tilde{M}(\mathbf{p}, \mathbf{q})$ are well defined by (3.5) and (3.6) and

$$\max_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\} \geq \tilde{M}(\mathbf{p}, \mathbf{q}), \quad \tilde{m}(\mathbf{p}, \mathbf{q}) \geq \min_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\}. \tag{3.8}$$

Proof. Since $q_i > 0$ for all $i \in \{1, \dots, n\}$, it is obvious that $0 < Q_k < 1$ for all $k \in \{1, \dots, n-1\}$, so that $\tilde{m}(\mathbf{p}, \mathbf{q})$ and $\tilde{M}(\mathbf{p}, \mathbf{q})$ are well defined by (3.5) and (3.6). Also, for any $k \in \{1, \dots, n-1\}$ we can write

$$P_k := \sum_{i=1}^k p_i = \sum_{i=1}^k \frac{p_i}{q_i} q_i, \quad 1 - P_k = \sum_{i=k+1}^n p_i = \sum_{i=k+1}^n \frac{p_i}{q_i} q_i.$$

Now,

$$\max_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\} Q_k = \max_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\} \sum_{i=1}^k q_i \geq P_k \geq \min_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\} \sum_{i=1}^k q_i = \min_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\} Q_k,$$

i.e.,

$$\max_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\} \geq \frac{P_k}{Q_k} \geq \min_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\}.$$

Similarly,

$$\max_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\} \geq \frac{1 - P_k}{1 - Q_k} \geq \min_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\} \quad \text{for all } k \in \{1, \dots, n-1\},$$

and (3.8) immediately follows. \square

REMARK 1. It is clear that inequalities stated in Lemma 1 can be strict. For example, if $n = 5$, $\mathbf{p} = (\frac{1}{6}, \frac{1}{3}, \frac{1}{9}, \frac{1}{6}, \frac{2}{9})$ and $\mathbf{q} = (\frac{1}{5}, \frac{1}{5}, \frac{1}{5}, \frac{1}{5}, \frac{1}{5})$, then

$$\max_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\} = \frac{5}{3} > \tilde{M}(\mathbf{p}, \mathbf{q}) = \frac{5}{4}, \quad \tilde{m}(\mathbf{p}, \mathbf{q}) = \frac{5}{6} > \min_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\} = \frac{5}{9}.$$

It is not hard to see that generally

$$\max_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\} = \max_{\pi \in \Pi_n} \tilde{M}(\mathbf{p}_\pi, \mathbf{q}_\pi), \quad \min_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\} = \min_{\pi \in \Pi_n} \tilde{m}(\mathbf{p}_\pi, \mathbf{q}_\pi).$$

THEOREM 3. Let $f : [a, b] \rightarrow \mathbb{R}$ be a convex function and $\mathbf{x} = (x_1, \dots, x_n) \in [a, b]^n$ be any n -tuple. Let $\pi = (\pi(1), \pi(2), \dots, \pi(n))$ be a permutation of $(1, 2, \dots, n)$ such that \mathbf{x}_π is monotonic (nondecreasing or nonincreasing). If $\mathbf{p} = (p_1, \dots, p_n)$ and $\mathbf{q} = (q_1, \dots, q_n)$ are two n -tuples from \mathcal{P}_n such that $q_i > 0$ for all $i \in \{1, \dots, n\}$, then

$$\begin{aligned} \max_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\} \mathcal{M}_n(f, \mathbf{x}, \mathbf{q}) &\geq \tilde{M}(\mathbf{p}_\pi, \mathbf{q}_\pi) \mathcal{M}_n(f, \mathbf{x}, \mathbf{q}) \geq \mathcal{M}_n(f, \mathbf{x}, \mathbf{p}) \\ &\geq \tilde{m}(\mathbf{p}_\pi, \mathbf{q}_\pi) \mathcal{M}_n(f, \mathbf{x}, \mathbf{q}) \geq \min_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\} \mathcal{M}_n(f, \mathbf{x}, \mathbf{q}), \end{aligned} \tag{3.9}$$

where $\tilde{m}(\mathbf{p}_\pi, \mathbf{q}_\pi)$ and $\tilde{M}(\mathbf{p}_\pi, \mathbf{q}_\pi)$ are defined as in (3.5) and (3.6). The first and the last inequality can be strict.

Proof. Since π is chosen so that \mathbf{x}_π is monotonic we can apply Corollary 2 and Lemma 1 to the n -tuples \mathbf{p}_π and \mathbf{q}_π to get

$$\begin{aligned} \max_{1 \leq i \leq n} \left\{ \frac{p_{\pi(i)}}{q_{\pi(i)}} \right\} \mathcal{M}_n(f, \mathbf{x}_\pi, \mathbf{q}_\pi) &\geq \tilde{M}(\mathbf{p}_\pi, \mathbf{q}_\pi) \cdot \mathcal{M}_n(f, \mathbf{x}_\pi, \mathbf{q}_\pi) \geq \mathcal{M}_n(f, \mathbf{x}_\pi, \mathbf{p}_\pi) \\ &\geq \tilde{m}(\mathbf{p}_\pi, \mathbf{q}_\pi) \cdot \mathcal{M}_n(f, \mathbf{x}_\pi, \mathbf{q}_\pi) \\ &\geq \min_{1 \leq i \leq n} \left\{ \frac{p_{\pi(i)}}{q_{\pi(i)}} \right\} \cdot \mathcal{M}_n(f, \mathbf{x}_\pi, \mathbf{q}_\pi). \end{aligned}$$

Since $\mathcal{M}_n(f, \mathbf{x}, \mathbf{p})$ doesn't change if we simultaneously permute the components of \mathbf{x} and \mathbf{p} , we have $\mathcal{M}_n(f, \mathbf{x}_\pi, \mathbf{p}_\pi) = \mathcal{M}_n(f, \mathbf{x}, \mathbf{p})$ and $\mathcal{M}_n(f, \mathbf{x}_\pi, \mathbf{q}_\pi) = \mathcal{M}_n(f, \mathbf{x}, \mathbf{q})$. Also it is obvious that $\max_{1 \leq i \leq n} \left\{ \frac{p_{\pi(i)}}{q_{\pi(i)}} \right\} = \max_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\}$ and $\min_{1 \leq i \leq n} \left\{ \frac{p_{\pi(i)}}{q_{\pi(i)}} \right\} = \min_{1 \leq i \leq n} \left\{ \frac{p_i}{q_i} \right\}$. Therefore, sequence of inequalities (3.9) holds. By Remark 1 the first and the last inequality in the sequence can be strict. \square

4. Integral versions

In [3] the following theorem is proved.

THEOREM C. *Let $(\Omega, \mathcal{A}, \mu)$ be a probability measure space, and let $x : \Omega \rightarrow [a, b]$ ($-\infty < a < b < \infty$) be a measurable function. Then for any continuous convex function $f : [a, b] \rightarrow \mathbb{R}$,*

$$f \left(a + b - \int_{\Omega} x d\mu \right) \leq f(a) + f(b) - \int_{\Omega} f(x) d\mu$$

holds.

It can analogously be proved that for a measure space $(\Omega, \mathcal{A}, \mu)$ with $0 < \mu(\Omega) < \infty$ the integral version of the Jensen-Mercer inequality

$$f \left(a + b - \frac{1}{\mu(\Omega)} \int_{\Omega} x d\mu \right) \leq f(a) + f(b) - \frac{1}{\mu(\Omega)} \int_{\Omega} f(x) d\mu \quad (4.1)$$

holds. In a special case when $\Omega = [\alpha, \beta]$, where $-\infty < \alpha < \beta < \infty$ and $\lambda : [\alpha, \beta] \rightarrow \mathbb{R}$ is any nondecreasing function such that $\lambda(\beta) \neq \lambda(\alpha)$ inequality (4.1) becomes

$$f \left(a + b - \frac{1}{\lambda(\beta) - \lambda(\alpha)} \int_{\alpha}^{\beta} x(t) d\lambda(t) \right) \leq f(a) + f(b) - \frac{1}{\lambda(\beta) - \lambda(\alpha)} \int_{\alpha}^{\beta} f(x(t)) d\lambda(t). \quad (4.2)$$

Also, we can prove that (4.2) remains valid even in the case when the condition " λ is nondecreasing function " is somewhat relaxed. We use the following integral variant of the Jensen-Steffensen inequality given by R. P. Boas [2] (see also [5, p. 59]):

THEOREM D. Let $x : [\alpha, \beta] \rightarrow (a, b)$ be a continuous and monotonic function (either nondecreasing or nonincreasing), where $-\infty < \alpha < \beta < \infty$ and $-\infty \leq a < b \leq \infty$, and let function $\lambda : [\alpha, \beta] \rightarrow \mathbb{R}$ be either continuous or of bounded variation satisfying

$$\lambda(\alpha) \leq \lambda(t) \leq \lambda(\beta) \text{ for all } t \in [\alpha, \beta], \quad \lambda(\beta) - \lambda(\alpha) > 0. \tag{4.3}$$

If $f : (a, b) \rightarrow \mathbb{R}$ is a convex function, then

$$f\left(\frac{1}{\lambda(\beta) - \lambda(\alpha)} \int_{\alpha}^{\beta} x(t) d\lambda(t)\right) \leq \frac{1}{\lambda(\beta) - \lambda(\alpha)} \int_{\alpha}^{\beta} f(x(t)) d\lambda(t). \tag{4.4}$$

holds.

THEOREM 4. Let $x : [\alpha, \beta] \rightarrow [a, b]$ be a continuous and monotonic function, where $-\infty < \alpha < \beta < +\infty$ and $-\infty < a < b < +\infty$. Let function $\lambda : [\alpha, \beta] \rightarrow \mathbb{R}$ be either continuous or of bounded variation satisfying (4.3). Then for any continuous convex function $f : [a, b] \rightarrow \mathbb{R}$ inequality (4.2) holds.

Proof. Suppose that x is continuous and nondecreasing i.e., $x(t_1) \leq x(t_2)$ for $t_1 < t_2 \in [\alpha, \beta]$. Let $\tilde{\alpha}, \tilde{\beta}$ be any real numbers such that $\tilde{\alpha} < \alpha$ and $\beta < \tilde{\beta}$. We define function $\tilde{x} : [\tilde{\alpha}, \tilde{\beta}] \rightarrow [a, b]$ by

$$\tilde{x}(t) = \begin{cases} a + \frac{x(\alpha) - a}{\alpha - \tilde{\alpha}}(t - \tilde{\alpha}), & t \in [\tilde{\alpha}, \alpha]; \\ x(t), & t \in [\alpha, \beta]; \\ b + \frac{b - x(\beta)}{\tilde{\beta} - \beta}(t - \tilde{\beta}), & t \in [\beta, \tilde{\beta}]. \end{cases}$$

\tilde{x} is also continuous and nondecreasing. Furthermore, $\tilde{x}([\tilde{\alpha}, \alpha]) = [a, x(\alpha)]$, $\tilde{x}([\beta, \tilde{\beta}]) = [x(\beta), b]$ and $\tilde{x}(t) = x(t)$ for all $t \in [\alpha, \beta]$.

Next, we define two functions $\tilde{\lambda}_s : [\tilde{\alpha}, \tilde{\beta}] \rightarrow \mathbb{R}$ and $\tilde{\lambda}_c : [\tilde{\alpha}, \tilde{\beta}] \rightarrow \mathbb{R}$ by

$$\tilde{\lambda}_s(t) = \begin{cases} 1, & t = \tilde{\alpha}; \\ 0, & t \in (\tilde{\alpha}, \tilde{\beta}); \\ -1, & t = \tilde{\beta}, \end{cases} \text{ and } \tilde{\lambda}_c(t) = \begin{cases} 1, & t \in [\tilde{\alpha}, \alpha]; \\ \frac{\lambda(\beta) - \lambda(t)}{\lambda(\beta) - \lambda(\alpha)}, & t \in [\alpha, \beta]; \\ 0, & t \in [\beta, \tilde{\beta}]. \end{cases}$$

Notice that $\tilde{\lambda}_s$ is a step function with only jumps at end points of the interval $[\tilde{\alpha}, \tilde{\beta}]$. Therefore, the integral $\int_{\tilde{\alpha}}^{\tilde{\beta}} y(t) d\tilde{\lambda}_s(t)$ is well defined for any function $y : [\tilde{\alpha}, \tilde{\beta}] \rightarrow \mathbb{R}$ continuous at the points $\tilde{\alpha}$ and $\tilde{\beta}$, and we have

$$\begin{aligned} \int_{\tilde{\alpha}}^{\tilde{\beta}} y(t) d\tilde{\lambda}_s(t) &= y(\tilde{\alpha}) [\tilde{\lambda}_s(\tilde{\alpha} + 0) - \tilde{\lambda}_s(\tilde{\alpha})] + y(\tilde{\beta}) [\tilde{\lambda}_s(\tilde{\beta}) - \tilde{\lambda}_s(\tilde{\beta} - 0)] \\ &= -y(\tilde{\alpha}) - y(\tilde{\beta}). \end{aligned} \tag{4.5}$$

Also, if λ is continuous on $[\alpha, \beta]$ then $\tilde{\lambda}_c$ is continuous on $[\tilde{\alpha}, \tilde{\beta}]$, and if λ is of bounded variation on $[\alpha, \beta]$ then $\tilde{\lambda}_c$ is of bounded variation on $[\tilde{\alpha}, \tilde{\beta}]$. Therefore the integral $\int_{\tilde{\alpha}}^{\tilde{\beta}} y(t) d\tilde{\lambda}_c(t)$ is well defined for any continuous and piecewise monotonic function $y : [\tilde{\alpha}, \tilde{\beta}] \rightarrow \mathbb{R}$, and we have

$$\begin{aligned} \int_{\tilde{\alpha}}^{\tilde{\beta}} y(t) d\tilde{\lambda}_c(t) &= \int_{\tilde{\alpha}}^{\alpha} y(t) d\tilde{\lambda}_c(t) + \int_{\alpha}^{\beta} y(t) d\tilde{\lambda}_c(t) + \int_{\beta}^{\tilde{\beta}} y(t) d\tilde{\lambda}_c(t) \\ &= \int_{\alpha}^{\beta} y(t) d\tilde{\lambda}_c(t) = \int_{\alpha}^{\beta} y(t) d \left[\frac{\lambda(\beta) - \lambda(t)}{\lambda(\beta) - \lambda(\alpha)} \right] \\ &= -\frac{1}{\lambda(\beta) - \lambda(\alpha)} \int_{\alpha}^{\beta} y(t) d\lambda(t). \end{aligned} \quad (4.6)$$

Now we define $\tilde{\lambda} : [\tilde{\alpha}, \tilde{\beta}] \rightarrow \mathbb{R}$ by $\tilde{\lambda}(t) = \tilde{\lambda}_c(t) - \tilde{\lambda}_s(t)$, $t \in [\tilde{\alpha}, \tilde{\beta}]$. Notice that

$$\tilde{\lambda}(\tilde{\beta}) - \tilde{\lambda}(\tilde{\alpha}) = \tilde{\lambda}_c(\tilde{\beta}) - \tilde{\lambda}_c(\tilde{\alpha}) - \tilde{\lambda}_s(\tilde{\beta}) + \tilde{\lambda}_s(\tilde{\alpha}) = 0 - 1 + 1 + 1 = 1. \quad (4.7)$$

From previous observations we conclude that the integral $\int_{\tilde{\alpha}}^{\tilde{\beta}} y(t) d\tilde{\lambda}(t)$ is well defined for any continuous and piecewise monotonic function $y : [\tilde{\alpha}, \tilde{\beta}] \rightarrow \mathbb{R}$, and from (4.5) and (4.6) we have

$$\begin{aligned} \int_{\tilde{\alpha}}^{\tilde{\beta}} y(t) d\tilde{\lambda}(t) &= \int_{\tilde{\alpha}}^{\tilde{\beta}} y(t) d\tilde{\lambda}_c(t) - \int_{\tilde{\alpha}}^{\tilde{\beta}} y(t) d\tilde{\lambda}_s(t) \\ &= -\frac{1}{\lambda(\beta) - \lambda(\alpha)} \int_{\alpha}^{\beta} y(t) d\lambda(t) + y(\tilde{\alpha}) + y(\tilde{\beta}). \end{aligned} \quad (4.8)$$

If we apply Theorem D on the functions \tilde{x} , f and $\tilde{\lambda}$ (we can do that even if the function $\tilde{\lambda}$ is neither continuous nor of bounded variation since all the integrals are well defined) we obtain

$$f \left(\frac{1}{\tilde{\lambda}(\tilde{\beta}) - \tilde{\lambda}(\tilde{\alpha})} \int_{\tilde{\alpha}}^{\tilde{\beta}} \tilde{x}(t) d\tilde{\lambda}(t) \right) \leq \frac{1}{\tilde{\lambda}(\tilde{\beta}) - \tilde{\lambda}(\tilde{\alpha})} \int_{\tilde{\alpha}}^{\tilde{\beta}} f(\tilde{x}(t)) d\tilde{\lambda}(t). \quad (4.9)$$

Using (4.7), (4.8) with $y = \tilde{x}$ and (4.8) with $y = f(\tilde{x})$, from (4.9) we have

$$f \left(a + b - \frac{1}{\lambda(\beta) - \lambda(\alpha)} \int_{\alpha}^{\beta} x(t) d\lambda(t) \right) \leq f(a) + f(b) - \frac{1}{\lambda(\beta) - \lambda(\alpha)} \int_{\alpha}^{\beta} f(x(t)) d\lambda(t). \quad (4.10)$$

In case when x is nonincreasing we define function $\tilde{x} : [\tilde{\alpha}, \tilde{\beta}] \rightarrow [a, b]$ by

$$\tilde{x}(t) = \begin{cases} b + \frac{x(\alpha)-b}{\alpha-\tilde{\alpha}}(t - \tilde{\alpha}), & t \in [\tilde{\alpha}, \alpha]; \\ x(t), & t \in [\alpha, \beta]; \\ a + \frac{a-x(\beta)}{\beta-\tilde{\beta}}(t - \tilde{\beta}), & t \in [\beta, \tilde{\beta}] \end{cases}$$

and obtain (4.10) in the same way. This completes the proof. \square

There is no loss in generality if we assume $\lambda(\beta) - \lambda(\alpha) = 1$ and consider the normalized Jensen-Mercer functional

$$\mathcal{M}(f, x, \lambda) := f(a) + f(b) - \int_{\alpha}^{\beta} f(x(t))d\lambda(t) - f\left(a + b - \int_{\alpha}^{\beta} x(t)d\lambda(t)\right). \quad (4.11)$$

Under the appropriate assumptions on f , x and λ , we always have $\mathcal{M}(f, x, \lambda) \geq 0$.

For $-\infty < \alpha < \beta < \infty$ let $\Lambda_{[\alpha, \beta]}$ denotes the class of all functions $\lambda : [\alpha, \beta] \rightarrow \mathbb{R}$ which are either continuous or of bounded variation and satisfy the condition

$$\lambda(\alpha) \leq \lambda(t) \leq \lambda(\beta) \text{ for all } t \in [\alpha, \beta], \quad \lambda(\beta) - \lambda(\alpha) = 1. \quad (4.12)$$

Notice that any nondecreasing function $\lambda : [\alpha, \beta] \rightarrow \mathbb{R}$ with $\lambda(\beta) - \lambda(\alpha) = 1$ belongs to $\Lambda_{[\alpha, \beta]}$.

Now we can state the integral analogues of the results from the previous section. First, we give the integral version of Theorem 1.

THEOREM 5. *Let λ and μ be two functions from $\Lambda_{[\alpha, \beta]}$, $-\infty < \alpha < \beta < \infty$. Let $x : [\alpha, \beta] \rightarrow [a, b]$, $-\infty < a < b < \infty$ be a continuous function and let $f : [a, b] \rightarrow \mathbb{R}$ be a convex function.*

a) *If μ is nondecreasing and if $m \geq 0$ is a constant such that the function $\rho : [\alpha, \beta] \rightarrow \mathbb{R}$ defined by*

$$\rho(t) := \lambda(t) - m\mu(t), \quad t \in [\alpha, \beta] \quad (4.13)$$

is also nondecreasing, then

$$\mathcal{M}(f, x, \lambda) \geq m\mathcal{M}(f, x, \mu). \quad (4.14)$$

b) *If λ is nondecreasing and if $M > 0$ is a constant such that the function $\sigma : [\alpha, \beta] \rightarrow \mathbb{R}$ defined by*

$$\sigma(t) := M\mu(t) - \lambda(t), \quad t \in [\alpha, \beta] \quad (4.15)$$

is also nondecreasing, then

$$M\mathcal{M}(f, x, \mu) \geq \mathcal{M}(f, x, \lambda). \quad (4.16)$$

Proof. a) Since μ and $\rho = \lambda - m\mu$ are assumed to be nondecreasing and $m \geq 0$, the function $\lambda = \rho + m\mu$ is nondecreasing too. Hence, $\mathcal{M}(f, x, \lambda)$ and $\mathcal{M}(f, x, \mu)$ are well defined. Since ρ is nondecreasing it has to be $m \leq 1$. If $m = 1$ then (4.14) holds with equality sign. Hence, it remains to consider the case when $m < 1$. In this case $\rho(\beta) - \rho(\alpha) = 1 - m > 0$ and ρ is nondecreasing by our assumption so that (4.2) can be applied to ρ . Now, using (4.2) and then the convexity of f we obtain (4.14).

b) Since λ and $\sigma = M\mu - \lambda$ are assumed to be nondecreasing and $M > 0$, the function $\mu = \frac{1}{M}(\sigma + \lambda)$ is nondecreasing too. Hence, $\mathcal{M}(f, x, \lambda)$ and $\mathcal{M}(f, x, \mu)$ are well defined. Since ρ is nondecreasing it has to be $M \geq 1$. If $M = 1$ then (4.16) holds with equality sign. Hence, it remains to consider the case when $M > 1$. In this case $\sigma(\beta) - \sigma(\alpha) = M - 1 > 0$ and σ is nondecreasing by our assumption so that (4.2) can be applied to σ . Now, using (4.2) and then the convexity of f we obtain (4.16). This completes the proof. \square

COROLLARY 4. Let λ and μ be two functions from $\Lambda_{[\alpha, \beta]}$, $-\infty < \alpha < \beta < \infty$, $x : [\alpha, \beta] \rightarrow [a, b]$, $-\infty < a < b < \infty$ be a continuous function and $f : [a, b] \rightarrow \mathbb{R}$ be a convex function.

a) Let μ is strictly increasing and define

$$\tilde{m} = \tilde{m}(\lambda, \mu) := \inf_{\alpha < t < \beta} \left\{ \inf \left\{ \frac{\lambda(t) - \lambda(s)}{\mu(t) - \mu(s)} : \alpha \leq s \leq \beta, s \neq t \right\} \right\}.$$

If $\tilde{m} \geq 0$, then

$$\mathcal{M}(f, x, \lambda) \geq \tilde{m} \mathcal{M}(f, x, \mu). \tag{4.17}$$

b) Let λ is nondecreasing, μ is strictly increasing and define

$$\tilde{M} = \tilde{M}(\lambda, \mu) := \sup_{\alpha < t < \beta} \left\{ \sup \left\{ \frac{\lambda(t) - \lambda(s)}{\mu(t) - \mu(s)} : \alpha \leq s \leq \beta, s \neq t \right\} \right\}.$$

If $\tilde{M} < \infty$, then

$$\tilde{M} \mathcal{M}(f, x, \mu) \geq \mathcal{M}(f, x, \lambda). \tag{4.18}$$

Proof. a) Since μ is strictly increasing it is injective and \tilde{m} is well defined real number in $[-\infty, \infty)$. If $\tilde{m} \geq 0$, then $\rho = \lambda - \tilde{m}\mu$ is well defined nondecreasing function on $[\alpha, \beta]$ and we can apply Theorem 5 a) with $m = \tilde{m}$ to obtain (4.17).

b) Since λ is nondecreasing and μ is strictly increasing \tilde{M} is well defined real number in $(0, \infty]$. If $\tilde{M} < \infty$ then $\sigma = \tilde{M}\mu - \lambda$ is well defined nondecreasing function on $[\alpha, \beta]$ and we can apply Theorem 5 b) with $M = \tilde{M}$ to obtain (4.18). \square

REMARK 2. If $\tilde{m} = 0$, then the inequality (4.17) is trivially fulfilled. Similarly, if $\tilde{M} = \infty$, then the inequality (4.18) is trivially fulfilled. Two simple examples which illustrate such cases are as follows.

For $[\alpha, \beta] = [0, 1]$ let $\lambda(t) = t^3$, $\mu(t) = t$, $t \in [0, 1]$. Then $\tilde{m} = 0$, $\tilde{M} = 3$.

For $[\alpha, \beta] = [0, 1]$ let $\lambda(t) = t$, $\mu(t) = t^3$, $t \in [0, 1]$. Then $\tilde{m} = \frac{1}{3}$, $\tilde{M} = \infty$.

As in the discrete case we can consider the uniform distribution i.e., the function $v \in \Lambda_{[\alpha, \beta]}$ defined by $v(t) := \frac{1}{\beta - \alpha}t$, $t \in [\alpha, \beta]$ and the corresponding nonweighted integral Jensen-Mercer functional

$$\mathcal{M}(f, x) := \mathcal{M}(f, x, v) = f(a) + f(b) - \frac{1}{\beta - \alpha} \int_{\alpha}^{\beta} f(x(t)) dt - f\left(a + b - \frac{1}{\beta - \alpha} \int_{\alpha}^{\beta} x(t) dt\right).$$

Then we can state the following special case of Corollary 4.

COROLLARY 5. *Let λ be a function from $\Lambda_{[\alpha, \beta]}$, $-\infty < \alpha < \beta < \infty$, $x: [\alpha, \beta] \rightarrow [a, b]$, $-\infty < a < b < \infty$ be a continuous function and $f: [a, b] \rightarrow \mathbb{R}$ be a convex function.*

a) *If $\tilde{m}_0 := (\beta - \alpha) \cdot \inf_{\alpha < t < \beta} \left\{ \inf \left\{ \frac{\lambda(t) - \lambda(s)}{t - s} : \alpha \leq s \leq \beta, s \neq t \right\} \right\} \geq 0$, then $\mathcal{M}(f, x, \lambda) \geq \tilde{m}_0 \mathcal{M}(f, x)$.*

b) *If λ is nondecreasing and $\tilde{M}_0 := (\beta - \alpha) \cdot \sup_{\alpha < t < \beta} \left\{ \sup \left\{ \frac{\lambda(t) - \lambda(s)}{t - s} : \alpha \leq s \leq \beta, s \neq t \right\} \right\} < \infty$, then $\tilde{M}_0 \mathcal{M}(f, x) \geq \mathcal{M}(f, x, \lambda)$.*

Next, we give the integral version of Theorem 2.

THEOREM 6. *Let λ and μ be two functions from $\Lambda_{[\alpha, \beta]}$, $-\infty < \alpha < \beta < \infty$, either both continuous or both of bounded variation. Let $x: [\alpha, \beta] \rightarrow [a, b]$, $-\infty < a < b < \infty$ be a monotonic function (either nondecreasing or nonincreasing) and $f: [a, b] \rightarrow \mathbb{R}$ be a convex function.*

a) *If $m \geq 0$ is a constant such that for all $\alpha < t < \beta$*

$$\lambda(t) - \lambda(\alpha) - m(\mu(t) - \mu(\alpha)) \geq 0, \quad \lambda(\beta) - \lambda(t) - m(\mu(\beta) - \mu(t)) \geq 0, \quad (4.19)$$

then

$$\mathcal{M}(f, x, \lambda) \geq m \mathcal{M}(f, x, \mu). \quad (4.20)$$

b) *If $M > 0$ is a constant such that for all $\alpha < t < \beta$*

$$M(\mu(t) - \mu(\alpha)) - (\lambda(t) - \lambda(\alpha)) \geq 0, \quad M(\mu(\beta) - \mu(t)) - (\lambda(\beta) - \lambda(t)) \geq 0, \quad (4.21)$$

then

$$M \mathcal{M}(f, x, \mu) \geq \mathcal{M}(f, x, \lambda). \quad (4.22)$$

Proof. First note that under the given assumptions on λ , μ , x and f , $\mathcal{M}(f, x, \lambda)$ and $\mathcal{M}(f, x, \mu)$ are well defined and nonnegative.

a) From (4.19) follows that it has to be $m \leq 1$. If $m = 1$ then (4.20) obviously holds with equality sign. Hence, it remains to consider the case $0 \leq m < 1$. We define the function $\rho: [\alpha, \beta] \rightarrow \mathbb{R}$ by $\rho(t) := \lambda(t) - m\mu(t)$, $t \in [\alpha, \beta]$. If λ and μ are both continuous (both of bounded variation), then ρ is continuous (of bounded variation) too. Further, from (4.19) follows that for all $\alpha < t < \beta$

$$\begin{aligned} \rho(t) - \rho(\alpha) &= \lambda(t) - \lambda(\alpha) - m(\mu(t) - \mu(\alpha)) \geq 0, \\ \rho(\beta) - \rho(t) &= \lambda(\beta) - \lambda(t) - m(\mu(\beta) - \mu(t)) \geq 0. \end{aligned}$$

Also, since $\lambda, \mu \in \Lambda_{[\alpha, \beta]}$

$$\rho(\beta) - \rho(\alpha) = \lambda(\beta) - \lambda(\alpha) - m(\mu(\beta) - \mu(\alpha)) = 1 - m > 0.$$

We conclude that the normalized function $\frac{1}{1-m}\rho$ belongs to the class $\Lambda_{[\alpha, \beta]}$. Now, we follow our proof of Theorem 5 a), but instead of using (4.2) we apply Theorem 4. Thus we obtain (4.20).

b) From (4.21) follows that it has to be $M \geq 1$. If $M = 1$ then (4.22) holds with equality sign. Hence, it remains to consider the case $M > 1$. We define the function $\sigma : [\alpha, \beta] \rightarrow \mathbb{R}$ by $\sigma(t) := M\mu(t) - \lambda(t)$, $t \in [\alpha, \beta]$. If λ and μ are both continuous (both of bounded variation), then σ is continuous (of bounded variation) too. Further, from (4.21) follows that for all $\alpha < t < \beta$

$$\begin{aligned}\sigma(t) - \sigma(\alpha) &= M(\mu(t) - \mu(\alpha)) - (\lambda(t) - \lambda(\alpha)) \geq 0, \\ \sigma(\beta) - \sigma(t) &= M(\mu(\beta) - \mu(t)) - (\lambda(\beta) - \lambda(t)) \geq 0.\end{aligned}$$

Also, since $\lambda, \mu \in \Lambda_{[\alpha, \beta]}$

$$\sigma(\beta) - \sigma(\alpha) = M(\mu(\beta) - \mu(\alpha)) - (\lambda(\beta) - \lambda(\alpha)) = M - 1 > 0.$$

We conclude that the normalized function $\frac{1}{M-1}\sigma$ belongs to the class $\Lambda_{[\alpha, \beta]}$. Now we follow our proof of Theorem 5 b) but instead of using (4.2) we apply Theorem 4. Thus we obtain (4.22). \square

COROLLARY 6. *Let λ and μ be two functions from $\Lambda_{[\alpha, \beta]}$, $-\infty < \alpha < \beta < \infty$, either both continuous or both of bounded variation. Let $x : [\alpha, \beta] \rightarrow [a, b]$, $-\infty < a < b < \infty$ be a monotonic function (either nondecreasing or nonincreasing) and $f : [a, b] \rightarrow \mathbb{R}$ be a convex function. Assume that $\mu(\alpha) < \mu(t) < \mu(\beta)$ for all $\alpha < t < \beta$, and define*

$$\tilde{m} = \tilde{m}(\lambda, \mu) := \inf \left\{ \frac{\lambda(t) - \lambda(\alpha)}{\mu(t) - \mu(\alpha)}, \frac{\lambda(\beta) - \lambda(t)}{\mu(\beta) - \mu(t)} : \alpha < t < \beta \right\}, \quad (4.23)$$

$$\tilde{M} = \tilde{M}(\lambda, \mu) := \sup \left\{ \frac{\lambda(t) - \lambda(\alpha)}{\mu(t) - \mu(\alpha)}, \frac{\lambda(\beta) - \lambda(t)}{\mu(\beta) - \mu(t)} : \alpha < t < \beta \right\}. \quad (4.24)$$

Then

$$\tilde{M} \mathcal{M}(f, x, \mu) \geq \mathcal{M}(f, x, \lambda) \geq \tilde{m} \mathcal{M}(f, x, \mu). \quad (4.25)$$

Proof. Since $\mu(\alpha) < \mu(t) < \mu(\beta)$ for all $\alpha < t < \beta$, \tilde{m} and \tilde{M} are well defined, and obviously $\tilde{m} \in [0, \infty)$ and $\tilde{M} \in (0, \infty]$. Therefore, the right inequality in (4.25) follows from Theorem 6 a) with $m = \tilde{m}$. If $\tilde{M} = \infty$, then the left inequality in (4.25) holds trivially, while for $\tilde{M} < \infty$ it follows from Theorem 6 b) with $M = \tilde{M}$. \square

As in the previous case we can consider the uniform distribution ν and the corresponding nonweighted integral Jensen-Mercer functional $\mathcal{M}(f, x)$ and state the following special case of Corollary 6:

COROLLARY 7. Let λ be a function from $\Lambda_{[\alpha,\beta]}$, $-\infty < \alpha < \beta < \infty$. Let $x : [\alpha, \beta] \rightarrow [a, b]$, $-\infty < a < b < \infty$ be a monotonic function (either nondecreasing or nonincreasing) and $f : [a, b] \rightarrow \mathbb{R}$ be a convex function. If \tilde{m}_0 and \tilde{M}_0 are defined by

$$\tilde{m}_0 := (\beta - \alpha) \cdot \inf \left\{ \frac{\lambda(t) - \lambda(\alpha)}{t - \alpha}, \frac{\lambda(\beta) - \lambda(t)}{\beta - t} : \alpha < t < \beta \right\},$$

$$\tilde{M}_0 := (\beta - \alpha) \cdot \sup \left\{ \frac{\lambda(t) - \lambda(\alpha)}{t - \alpha}, \frac{\lambda(\beta) - \lambda(t)}{\beta - t} : \alpha < t < \beta \right\},$$

then $\tilde{M}_0 \mathcal{M}(f, x) \geq \mathcal{M}(f, x, \lambda) \geq \tilde{m}_0 \mathcal{M}(f, x)$.

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