



Statistical Relative Approximation on Modular Spaces

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Abstract. In the present paper, using the concept of statistical relative convergence, we study the problem of approximation to a function by means of double sequences of positive linear operators defined on a modular space. Also, a non-trivial application is presented.

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1. Introduction and Preliminaries

Let $C(X)$ be the space of all continuous real valued functions on a compact subset X of the real numbers and (L_n) be the sequence of positive linear operators on $C(X)$ say $(L_n(f; x))$. Korovkin [15] established the sufficient conditions for the uniform convergence of (L_n) to a function f by using the test functions $1, x, x^2$. Many researchers have investigated these conditions for various operators defined on different spaces (see for instance [2, 13, 25]). Recently, Demirci and Orhan [9] introduced statistical relative uniform convergence of single sequences by using the notions of the natural density and the relative uniform convergence. Then, Yilmaz et al. [25] defined a new type of modular convergence by using the notion of the relative uniform convergence. More recently, Demirci and Kolay [10] studied statistical relative modular convergence of single sequences. In this paper, we investigate the problem of statistical relative approximation to a function f by means of double sequences of positive linear operators defined on a modular space.

Let us first remind of the concept of statistical convergence for double sequences.

A double sequence $x = \{x_{m,n}\}$ is said to be convergent in Pringsheim’s sense if, for every $\varepsilon > 0$, there exists $N = N(\varepsilon) \in \mathbb{N}$, the set of all natural numbers, such that $|x_{m,n} - L| < \varepsilon$ whenever $m, n > N$, where L is called the Pringsheim limit of x and denoted by $P - \lim_{m,n} x_{m,n} = L$ (see [23]). We shall call such an x , briefly, “ P -convergent”. A double sequence is called bounded if there exists a positive number M such that $|x_{m,n}| \leq M$ for all $(m, n) \in \mathbb{N}^2$. Note that in contrast to the case for single sequences, a convergent double sequence need not to be bounded.

Statistical convergence of single sequences was introduced by Steinhaus [24] and studied by many authors [11, 12, 14]. Recently, this concept was extended to the double sequences. If $E \subset \mathbb{N}^2$ is a two-dimensional subset of positive integers, then $E_{j,k}$ denotes the set $\{(m, n) \in E : m \leq j, n \leq k\}$ and $|E_{j,k}|$ denotes the cardinality of $E_{j,k}$. The double natural density of E [19] is given by

$$\delta_2(E) := P - \lim_{j,k} \frac{1}{jk} |E_{j,k}|,$$

if it exists. For example, let $E = \{(j^2, k^2) : j, k \in \mathbb{N}\}$ then $\delta_2(E) = 0$. The number sequence $x = \{x_{m,n}\}$ is statistically convergent to L provided that for every $\varepsilon > 0$, the set

$$E := E_{j,k}(\varepsilon) := \{m \leq j, n \leq k : |x_{m,n} - L| \geq \varepsilon\}$$

has natural density zero; in that case we write $st_2 - \lim_{m,n} x_{m,n} = L$.

Clearly, a P -convergent double sequence is statistically convergent to the same value but its converse is not always true. Also, note that a statistically convergent double sequence need not to be bounded. For example, consider the double sequence $x = \{x_{m,n}\}$ given by

$$x_{m,n} = \begin{cases} mn, & m = k^2 \text{ and } n = l^2 \\ 1, & \text{otherwise.} \end{cases}, \quad k, l = 1, 2, \dots$$

Then, clearly $st_2 - \lim_{m,n} x_{m,n} = 1$. Nevertheless, x is neither convergent nor bounded.

Also, the statistical convergence for double sequences was characterized in [19] as given below:

A double sequence $x = \{x_{m,n}\}$ is statistically convergent to L if and only if there exists a set $S \subset \mathbb{N}^2$ such that the natural density of S is 1 and

$$P - \lim_{m,n \rightarrow \infty \text{ and } (m,n) \in S} x_{m,n} = L.$$

The concepts of *statistical superior limit* and *inferior limit* for double sequences have been introduced by Çakan and Altay [5]. For any real double sequence $x = \{x_{m,n}\}$, the statistical superior limit of x is

$$st_2 - \limsup_{m,n} x_{m,n} = \begin{cases} \sup G_x, & \text{if } G_x \neq \emptyset, \\ -\infty, & \text{if } G_x = \emptyset, \end{cases}$$

where $G_x := \{C \in \mathbb{R} : \delta_2(\{(m, n) : x_{m,n} > C\}) \neq 0\}$ and \emptyset denotes the empty set. We note that, in general, by $\delta_2(K) \neq 0$ we mean either $\delta_2(K) > 0$ or K fails to have the double natural density. Similarly, the statistical inferior limit of x is

$$st_2 - \liminf_{m,n} x_{m,n} = \begin{cases} \inf F_x, & \text{if } F_x \neq \emptyset, \\ \infty, & \text{if } F_x = \emptyset, \end{cases}$$

where $F_x := \{D \in \mathbb{R} : \delta_2(\{(m, n) : x_{m,n} < D\}) \neq 0\}$. As in the ordinary superior or inferior limit, it was proved that

$$st_2 - \liminf_{m,n} x_{m,n} \leq st_2 - \limsup_{m,n} x_{m,n}$$

and also that, for any double sequence $x = \{x_{m,n}\}$ satisfying

$$\delta_2(\{(m, n) : |x_{m,n}| > M\}) = 0 \text{ for some } M > 0,$$

$$st_2 - \lim_{m,n} x_{m,n} = L \text{ iff } st_2 - \liminf_{m,n} x_{m,n} = st_2 - \limsup_{m,n} x_{m,n} = L.$$

Now, we focus on modular spaces.

Let $I = [a, b]$ be a bounded interval of the real line \mathbb{R} provided with the Lebesgue measure. Then, by $X(I^2)$ we denote the space of all real-valued measurable functions on $I^2 = [a, b] \times [a, b]$ provided with equality a.e. As usual, let $C(I^2)$ denote the space of all continuous real-valued functions, and $C^\infty(I^2)$ denote the space of all infinitely differentiable functions on I^2 . In this case, we say that a functional $\rho : X(I^2) \rightarrow [0, +\infty]$ is a *modular* on $X(I^2)$ provided that the following conditions hold:

- (i) $\rho(f) = 0$ if and only if $f = 0$ a.e. in I^2 ,
- (ii) $\rho(-f) = \rho(f)$ for every $f \in X(I^2)$,
- (iii) $\rho(\alpha f + \beta g) \leq \rho(f) + \rho(g)$ for every $f, g \in X(I^2)$ and for any $\alpha, \beta \geq 0$ with $\alpha + \beta = 1$.

Recall that a modular ρ is called *N-quasi convex* if the following is satisfied:

- there exists a constant $N \geq 1$ such that

$$\rho(\alpha f + \beta g) \leq N\alpha\rho(Nf) + N\beta\rho(Ng)$$

holds for every $f, g \in X(I^2)$, $\alpha, \beta \geq 0$ with $\alpha + \beta = 1$. In particular, if $N = 1$, then ρ is called *convex*.

Furthermore, a modular ρ is called *N-quasi semiconvex* if it holds:

- there exists a constant $N \geq 1$ such that

$$\rho(af) \leq Na\rho(Nf)$$

holds for every $f \in X(I^2)$ and $a \in (0, 1]$.

It is clear that every *N-quasi semiconvex modular* is *N-quasi convex*. We should recall that the above two concepts were introduced and discussed in details by Bardaro et al. [3, 4].

We now consider some appropriate vector subspaces of $X(I^2)$ by means of a modular ρ as follows:

$$L^\rho(I^2) := \left\{ f \in X(I^2) : \lim_{\lambda \rightarrow 0^+} \rho(\lambda f) = 0 \right\}$$

and

$$E^\rho(I^2) := \{ f \in L^\rho(I^2) : \rho(\lambda f) < +\infty \text{ for all } \lambda > 0 \}.$$

Here, $L^\rho(I^2)$ is called the *modular space* generated by ρ ; and $E^\rho(I^2)$ is called the space of the finite elements of $L^\rho(I^2)$. Observe that if ρ is N -quasi semiconvex, then the space

$$\{ f \in X(I^2) : \rho(\lambda f) < +\infty \text{ for some } \lambda > 0 \}$$

coincides with $L^\rho(I^2)$. The notions about modulars are introduced in [20] and widely discussed in [4] (see also [16, 21]).

Moore [18] introduced the notion of uniform convergence of a sequence of functions relative to a scale function. Then, Chittenden [7] gave the following definition of relative uniform convergence is equivalent to the definition given by Moore:

A sequence (f_n) of functions, defined on an interval $I \equiv (a \leq x \leq b)$, converges *relatively uniformly to a limit function* f if there exists a function $\sigma(x)$, called a scale function $\sigma(x)$ such that for every $\varepsilon > 0$ there is an integer n_ε such that for every $n > n_\varepsilon$ the inequality $|f_n(x) - f(x)| < \varepsilon |\sigma(x)|$ holds uniformly in x on the interval I . The sequence (f_n) is said to converge *uniformly relatively to the scale function* σ or more simply, *relatively uniformly*. It will be observed that uniform convergence is the special case of relative uniform convergence in which the scale function is a non-zero constant (for more properties and details, see also [6–8]).

Now we introduce the notions of the *relative modular (or strong) convergence* and *statistical relative modular (or strong) convergence* for double sequences as follows:

Definition 1. Let $\{f_{m,n}\}$ be a double function sequence whose terms belong to $L^\rho(I^2)$. Then, $\{f_{m,n}\}$ is *relatively modularly convergent* to a function $f \in L^\rho(I^2)$ iff there exists a function $\sigma(x, y)$, called a scale function $\sigma \in X(I^2)$, $|\sigma(x, y)| \neq 0$ such that

$$P - \lim_{m,n} \rho \left(\lambda_0 \left(\frac{f_{m,n} - f}{\sigma} \right) \right) = 0 \text{ for some } \lambda_0 > 0. \tag{1}$$

And also, $\{f_{m,n}\}$ is *relatively F -norm convergent* (or, *relatively strongly convergent*) to f iff

$$P - \lim_{m,n} \rho \left(\lambda \left(\frac{f_{m,n} - f}{\sigma} \right) \right) = 0 \text{ for every } \lambda > 0. \tag{2}$$

It can be immediately seen that (1) and (2) are equivalent if and only if the modular ρ satisfies the Δ_2 -condition, i.e. there exists a constant $M > 0$ such that $\rho(2f) \leq M\rho(f)$ for every $f \in X(I^2)$. Indeed, relative strong convergence of the double sequence $\{f_{m,n}\}$ to f is equivalent to the condition $P - \lim_{m,n} \rho\left(2^N \lambda \left(\frac{f_{m,n}-f}{\sigma}\right)\right) = 0$, for all $N = 1, 2, \dots$ and some $\lambda > 0$. Let $\{f_{m,n}\}$ be relatively modularly convergent to f , hence there exists a $\lambda > 0$ such that $P - \lim_{m,n} \rho\left(\lambda \left(\frac{f_{m,n}-f}{\sigma}\right)\right) = 0$. Δ_2 -condition implies by induction that $\rho\left(2^N \lambda \left(\frac{f_{m,n}-f}{\sigma}\right)\right) \leq M^N \rho\left(\lambda \left(\frac{f_{m,n}-f}{\sigma}\right)\right)$, then we get

$$P - \lim_{m,n} \rho\left(2^N \lambda \left(\frac{f_{m,n}-f}{\sigma}\right)\right) = 0.$$

Definition 2. Let $\{f_{m,n}\}$ be a function sequence whose terms belong to $L^\rho(I^2)$. Then, $\{f_{m,n}\}$ is said to be *statistically relatively modularly convergent* to a function $f \in L^\rho(I^2)$ if there exists a function $\sigma(x, y)$, called a scale function $\sigma \in X(I^2)$, $|\sigma(x, y)| \neq 0$ such that

$$st_2 - \lim_{m,n} \rho\left(\lambda_0 \left(\frac{f_{m,n}-f}{\sigma}\right)\right) = 0 \text{ for some } \lambda_0 > 0.$$

Also, $\{f_{m,n}\}$ is *statistically relatively F-norm convergent* (or, *statistically relatively strongly convergent*) to f iff

$$st_2 - \lim_{m,n} \rho\left(\lambda \left(\frac{f_{m,n}-f}{\sigma}\right)\right) = 0 \text{ for every } \lambda > 0.$$

It will be observed that statistical modular convergence is the special case of statistical relative modular convergence in which the scale function is a non-zero constant (cf. [22]). Moreover, if $\sigma(x, y)$ is bounded, statistical relative modular convergence implies statistical modular convergence. However, statistical relative modular convergence does not imply statistical modular convergence, when $\sigma(x, y)$ is unbounded. This is illustrated by the following example:

Example 1. Take $I = [0, 1]$ and let $\varphi : [0, \infty) \rightarrow [0, \infty)$ be a continuous function for which the following conditions hold:

- φ is convex
- $\varphi(0) = 0, \varphi(u) > 0$ for $u > 0$ and $\lim_{u \rightarrow \infty} \varphi(u) = \infty$.

Hence, consider the functional ρ^φ on $X(I^2)$ defined by

$$\rho^\varphi(f) := \int_0^1 \int_0^1 \varphi(|f(x, y)|) dx dy \quad \text{for } f \in X(I^2).$$

In this case, ρ^φ is a convex modular on $X(I^2)$, which satisfies all assumptions listed in this section. Consider the Orlicz space generated by φ as

follows:

$$L_\varphi^\rho(I^2) := \{f \in X(I^2) : \rho^\varphi(\lambda f) < +\infty \text{ for some } \lambda > 0\}.$$

For each $m, n \in N$, define $g_{m,n} : [0, 1] \times [0, 1] \rightarrow \mathbb{R}$ by

$$g_{m,n}(x, y) = \begin{cases} 1, & m = k^2 \text{ and } n = l^2 \\ & (x, y) \in (0, \frac{1}{m}) \times (0, \frac{1}{n}); \\ m^2n(1 - mnxy), & m \neq k^2 \text{ and } n \neq l^2 \\ 0, & (x, y) = (0, 0) \text{ or } (x, y) \in [\frac{1}{m}, 1] \times [\frac{1}{n}, 1]; \\ & m \neq k^2 \text{ and } n \neq l^2 \end{cases},$$

$k, l = 1, 2, \dots$ (3)

If $\varphi(x) = x^p$ for $1 \leq p < \infty, x \geq 0$, then $L_\varphi^\rho(I^2) = L_p(I^2)$. Moreover we have for any function $f \in L_\varphi^\rho(I^2)$

$$\rho^\varphi(f) = \|f\|_{L_p}^p.$$

It is clear that $\{g_{m,n}\}$ does not converge statistically modularly but converges to $g = 0$ statistically modularly relative to a scale function.

$\sigma(x, y) = \begin{cases} 1, & (x, y) = (0, 0) \\ \frac{1}{x^2y}, & (x, y) \in (0, 1] \times (0, 1] \end{cases}$ on $L_1([0, 1] \times [0, 1])$. Indeed, for some $\lambda_0 > 0$, with the choice of $p = 1$ we have $\rho^\varphi(g) = \|g\|_{L_1}$,

$$\begin{aligned} \rho(\lambda_0(g_{m,n} - g)) &= \|\lambda_0(g_{m,n} - g)\|_{L_1} \\ &= \begin{cases} 1, & m = k^2 \text{ and } n = l^2 \\ \frac{3m}{4}, & m \neq k^2 \text{ and } n \neq l^2 \end{cases}, k, l = 1, 2, \dots, \end{aligned} \tag{4}$$

then we have

$$st_2 - \lim_{m,n} \|\lambda_0(g_{m,n} - g)\|_{L_1} \neq 0.$$

Using the scale function σ ,

$$\rho\left(\lambda_0\left(\frac{g_{m,n} - g}{\sigma}\right)\right) = \begin{cases} \frac{1}{6}, & m = k^2 \text{ and } n = l^2 \\ \frac{1}{12mn}, & m \neq k^2 \text{ and } n \neq l^2 \end{cases}, k, l = 1, 2, \dots,$$

we get

$$st_2 - \lim_{m,n} \left\| \lambda_0\left(\frac{g_{m,n} - g}{\sigma}\right) \right\|_{L_1} = 0.$$

On the other hand, we can easily see that $\{g_{m,n}\}$ does not converge to $g = 0$ modularly relative to a scale function $\sigma(x, y) = \begin{cases} 1, & (x, y) = (0, 0) \\ \frac{1}{x^2y}, & (x, y) \in (0, 1] \times (0, 1] \end{cases}$ on $L_1([0, 1] \times [0, 1])$. Indeed from (4), we observed that the sequence

$\left\{ \rho\left(\lambda_0\left(\frac{g_{m,n} - g}{\sigma}\right)\right) \right\}$ has two subsequences with different limit points so $\{g_{m,n}\}$ is not relatively modularly convergent to $g = 0$.

In this paper, we will need the following assumptions on a modular ρ :

- ρ is *monotone* if $\rho(f) \leq \rho(g)$ for $|f| \leq |g|$,
- ρ is *finite* if $\chi_A \in L^\rho(I^2)$ whenever A is measurable subset of I^2 such that $\mu(A) < \infty$,
- ρ is *absolutely finite* if ρ is finite and, for every $\varepsilon > 0, \lambda > 0$, there exists a $\delta > 0$ such that $\rho(\lambda\chi_B) < \varepsilon$ for any measurable subset $B \subset I^2$ with $\mu(B) < \delta$,
- ρ is *strongly finite* if $\chi_{I^2} \in E^\rho(I^2)$,
- ρ is *absolutely continuous* provided that there exists an $\alpha > 0$ such that, for every $f \in X(I^2)$ with $\rho(f) < +\infty$, the following condition holds: for every $\varepsilon > 0$ there is $\delta > 0$ such that $\rho(\alpha f\chi_B) < \varepsilon$ whenever B is any measurable subset of I^2 with $\mu(B) < \delta$.

Observe now that (see [2,3]) if a modular ρ is monotone and finite, then we have $C(I^2) \subset L^\rho(I^2)$. In a similar manner, if ρ is monotone and strongly finite, then $C(I^2) \subset E^\rho(I^2)$. Also, if ρ is monotone, absolutely finite and absolutely continuous, then $\overline{C^\infty(I^2)} = L^\rho(I^2)$. Some important relations between the above properties may be found in [1,3,4,17,21].

2. Statistical Relative Korovkin Theorems in Modular Spaces

In this section, we apply the notion of statistical relative modular convergence of a double sequences of positive linear operators defined on a modular space to prove a Korovkin type approximation theorem.

Let ρ be a monotone and finite modular on $X(I^2)$. Assume that D is a set satisfying $C^\infty(I^2) \subset D \subset L^\rho(I^2)$. We can construct such a subset D when ρ is monotone and finite (see [2]). Assume further that $\mathbb{T} := \{T_{m,n}\}$ is a sequence of positive linear operators from D into $X(I^2)$ for which there exists a subset $X_{\mathbb{T}} \subset D$ containing $C^\infty(I^2)$ and $\sigma \in X(I^2)$ is an unbounded function satisfying $\sigma(x, y) \neq 0$ such that

$$st_2 - \limsup_{m,n} \rho \left(\lambda \left(\frac{T_{m,n}h}{\sigma} \right) \right) \leq R\rho(\lambda h) \tag{5}$$

holds for every $h \in X_{\mathbb{T}}, \lambda > 0$ and for an absolute positive constant R .

We denote the value of $T_{m,n}f$ at a point $(x, y) \in I^2$ by $T_{m,n}(f(u, v); x, y)$ or briefly, $T_{m,n}(f; x, y)$.

Theorem 1. *Let ρ be a monotone, strongly finite, absolutely continuous and N -quasi semiconvex modular on $X(I^2)$. Let $\mathbb{T} := \{T_{m,n}\}$ be a double sequence of positive linear operators from D into $X(I^2)$ satisfying (5) and suppose that $\sigma_i(x, y)$ is an unbounded function satisfying $|\sigma_i(x, y)| \geq a_i > 0 (i = 0, 1, 2, 3)$. Assume that*

$$st_2 - \lim_{m,n} \rho \left(\lambda \left(\frac{T_{m,n}(e_i) - e_i}{\sigma_i} \right) \right) = 0 \quad \text{for every } \lambda > 0 \text{ and } i = 0, 1, 2, 3 \tag{6}$$

where $e_0(x, y) = 1, e_1(x, y) = x, e_2(x, y) = y$ and $e_3(x, y) = x^2 + y^2$. Now let f be any function belonging to $L^\rho(I^2)$ such that $f - g \in X_{\mathbb{T}}$ for every $g \in C^\infty(I^2)$. Then, we have

$$st_2 - \lim_{m,n} \rho \left(\lambda_0 \left(\frac{T_{m,n}f - f}{\sigma} \right) \right) = 0 \text{ for some } \lambda_0 > 0. \tag{7}$$

where $\sigma(x, y) = \max \{|\sigma_i(x, y)|; i = 0, 1, 2, 3\}$.

Proof. We first claim that

$$st_2 - \lim_{m,n} \rho \left(\eta \left(\frac{T_{m,n}g - g}{\sigma} \right) \right) = 0 \text{ for every } g \in C(I^2) \cap D \text{ and every } \eta > 0. \tag{8}$$

To see this assume that g belongs to $C(I^2) \cap D$. By the continuity of g on I^2 , given $\varepsilon > 0$, there exists a number $\delta > 0$ such that for all $(u, v), (x, y) \in I^2$ satisfying $|u - x| < \delta$ and $|v - y| < \delta$ we have

$$|g(u, v) - g(x, y)| < \varepsilon. \tag{9}$$

Also we get for all $(u, v), (x, y) \in I^2$ satisfying $|u - x| > \delta$ and $|v - y| > \delta$ that

$$|g(u, v) - g(x, y)| \leq \frac{2M}{\delta^2} \left\{ (u - x)^2 + (v - y)^2 \right\} \tag{10}$$

where $M := \sup_{(x,y) \in I^2} |g(x, y)|$. Combining (9) and (10) we have for

$(u, v), (x, y) \in I^2$ that

$$|g(u, v) - g(x, y)| < \varepsilon + \frac{2M}{\delta^2} \left\{ (u - x)^2 + (v - y)^2 \right\}.$$

namely,

$$\begin{aligned} -\varepsilon - \frac{2M}{\delta^2} \left\{ (u - x)^2 + (v - y)^2 \right\} \\ < g(u, v) - g(x, y) < \varepsilon + \frac{2M}{\delta^2} \left\{ (u - x)^2 + (v - y)^2 \right\}. \end{aligned} \tag{11}$$

Since $T_{m,n}$ is linear and positive, by applying $T_{m,n}$ to (11) for every $m, n \in \mathbb{N}$ we get

$$\begin{aligned} -\varepsilon T_{m,n}(e_0; x, y) - \frac{2M}{\delta^2} T_{m,n} \left((u - x)^2 + (v - y)^2; x, y \right) \\ < T_{m,n}(g; x, y) - g(x, y) T_{m,n}(e_0; x, y) \\ < \varepsilon T_{m,n}(e_0; x, y) + \frac{2M}{\delta^2} T_{m,n} \left((u - x)^2 + (v - y)^2; x, y \right) \end{aligned}$$

and hence,

$$\begin{aligned} |T_{m,n}(g; x, y) - g(x, y)| &\leq |T_{m,n}(g; x, y) - g(x, y) T_{m,n}(e_0; x, y)| \\ &\quad + |g(x, y) T_{m,n}(e_0; x, y) - g(x, y)| \end{aligned}$$

$$\begin{aligned} &\leq \varepsilon T_{m,n}(e_0; x, y) + M |T_{m,n}(e_0; x, y) - e_0(x, y)| \\ &\quad + \frac{2M}{\delta^2} T_{m,n}\left((u-x)^2 + (v-y)^2; x, y\right) \end{aligned}$$

holds for every $x, y \in I$ and $m, n \in \mathbb{N}$. The above inequality implies that

$$\begin{aligned} |T_{m,n}(g; x, y) - g(x, y)| &\leq \varepsilon + \left\{ \varepsilon + M + \frac{4M}{\delta^2} E^2 \right\} |T_{m,n}(e_0; x, y) - e_0(x, y)| \\ &\quad + \frac{4M}{\delta^2} E |T_{m,n}(e_1; x, y) - e_1(x, y)| \\ &\quad + \frac{4M}{\delta^2} E |T_{m,n}(e_2; x, y) - e_2(x, y)| \\ &\quad + \frac{2M}{\delta^2} |T_{m,n}(e_3; x, y) - e_3(x, y)|. \end{aligned}$$

where $E := \max\{|x|, |y|\}$. Now we multiply the both-sides of the above inequality by $\frac{1}{|\sigma(x, y)|}$ and for any $\eta > 0$, the last inequality gives that

$$\begin{aligned} \eta \left| \frac{T_{m,n}(g; x, y) - g(x, y)}{\sigma(x, y)} \right| &\leq \frac{\eta\varepsilon}{|\sigma(x, y)|} + K\eta \left\{ \left| \frac{T_{m,n}(e_0; x, y) - e_0(x, y)}{\sigma(x, y)} \right| \right. \\ &\quad + \left| \frac{T_{m,n}(e_1; x, y) - e_1(x, y)}{\sigma(x, y)} \right| \\ &\quad + \left| \frac{T_{m,n}(e_2; x, y) - e_2(x, y)}{\sigma(x, y)} \right| \\ &\quad \left. + \left| \frac{T_{m,n}(e_3; x, y) - e_3(x, y)}{\sigma(x, y)} \right| \right\}, \end{aligned}$$

where $K := \max\left\{\varepsilon + M + \frac{4M}{\delta^2} E^2, \frac{4M}{\delta^2} E, \frac{2M}{\delta^2}\right\}$. Now, applying the modular ρ to both-sides of the above inequality, since ρ is monotone and $\sigma(x, y) = \max\{|\sigma_i(x, y)|; i = 0, 1, 2, 3\}$, we have

$$\begin{aligned} \rho\left(\eta \left(\frac{T_{m,n}g - g}{\sigma}\right)\right) &\leq \rho\left(\eta \frac{\varepsilon}{|\sigma|} + \eta K \left| \frac{T_{m,n}e_0 - e_0}{\sigma_0} \right| + \eta K \left| \frac{T_{m,n}e_1 - e_1}{\sigma_1} \right| \right. \\ &\quad \left. + \eta K \left| \frac{T_{m,n}e_2 - e_2}{\sigma_2} \right| + \eta K \left| \frac{T_{m,n}e_3 - e_3}{\sigma_3} \right| \right). \end{aligned}$$

Hence, we may write that

$$\begin{aligned} \rho\left(\eta \left(\frac{T_{m,n}g - g}{\sigma}\right)\right) &\leq \rho\left(\frac{5\eta\varepsilon}{\sigma}\right) + \rho\left(5\eta K \left(\frac{T_{m,n}e_0 - e_0}{\sigma_0}\right)\right) \\ &\quad + \rho\left(5\eta K \left(\frac{T_{m,n}e_1 - e_1}{\sigma_1}\right)\right) \\ &\quad + \rho\left(5\eta K \left(\frac{T_{m,n}e_2 - e_2}{\sigma_2}\right)\right) \end{aligned}$$

$$+ \rho \left(5\eta K \left(\frac{T_{m,n}e_3 - e_3}{\sigma_3} \right) \right)$$

Since ρ is N -quasi semiconvex and strongly finite, we have, assuming $0 < \varepsilon \leq 1$,

$$\begin{aligned} \rho \left(\eta \left(\frac{T_{m,n}g - g}{\sigma} \right) \right) &\leq N\varepsilon\rho \left(\frac{5\eta N}{\sigma} \right) + \rho \left(5\eta K \left(\frac{T_{m,n}e_0 - e_0}{\sigma_0} \right) \right) \\ &+ \rho \left(5\eta K \left(\frac{T_{m,n}e_1 - e_1}{\sigma_1} \right) \right) \\ &+ \rho \left(5\eta K \left(\frac{T_{m,n}e_2 - e_2}{\sigma_2} \right) \right) \\ &+ \rho \left(5\eta K \left(\frac{T_{m,n}e_3 - e_3}{\sigma_3} \right) \right). \end{aligned}$$

For a given $r > 0$, choose an $\varepsilon \in (0, 1]$ such that $N\varepsilon\rho \left(\frac{5\eta N}{\sigma} \right) < r$. Now define the following sets:

$$\begin{aligned} G_\eta &:= \left\{ (m, n) : \rho \left(\eta \left(\frac{T_{m,n}g - g}{\sigma} \right) \right) \geq r \right\} \\ G_{\eta,i} &:= \left\{ (m, n) : \rho \left(5\eta K \left(\frac{T_{m,n}e_i - e_i}{\sigma_i} \right) \right) \geq \frac{r - N\varepsilon\rho \left(\frac{5\eta N}{\sigma} \right)}{4} \right\}, \end{aligned}$$

where $i = 0, 1, 2, 3$. Then, it is easy to see that $G_\eta \subseteq \bigcup_{i=0}^3 G_{\eta,i}$. So, we can write that

$$\delta_2(G_\eta) \leq \sum_{i=0}^3 \delta_2(G_{\eta,i}).$$

Using the hypothesis (6), we get

$$\delta_2(G_\eta) = 0,$$

which proves our claim (8). Obviously (8) also holds for every $g \in C^\infty(I^2)$. Now let $f \in L^\rho(I^2)$ satisfying $f - g \in X_{\mathbb{T}}$ for every $g \in C^\infty(I^2)$. Since $\mu(I^2) < \infty$ and ρ is strongly finite and absolutely continuous, we can see that ρ is also absolutely finite on $X(I^2)$. Using these properties of the modular ρ , it is known from [4, 17] that the space $C^\infty(I^2)$ is modularly dense in $L^\rho(I^2)$, i.e., there exists a sequence $\{g_{k,j}\} \subset C^\infty(I^2)$ such that

$$P - \lim_{k,j} \rho(3\lambda_0^*(g_{k,j} - f)) = 0 \quad \text{for some } \lambda_0^* > 0.$$

This means that, for every $\varepsilon > 0$, there is a positive number $k_0 = k_0(\varepsilon)$ so that

$$\rho(3\lambda_0^*(g_{k,j} - f)) < \varepsilon \quad \text{for every } k, j \geq k_0. \tag{12}$$

On the other hand, by the linearity and positivity of the operators $T_{m,n}$, we may write that

$$\begin{aligned} \lambda_0^* |T_{m,n}(f; x, y) - f(x, y)| &\leq \lambda_0^* |T_{m,n}(f - g_{k_0, k_0}; x, y)| \\ &\quad + \lambda_0^* |T_{m,n}(g_{k_0, k_0}; x, y) - g_{k_0, k_0}(x, y)| \\ &\quad + \lambda_0^* |g_{k_0, k_0}(x, y) - f(x, y)| \end{aligned}$$

holds for every $x, y \in I$ and $m, n \in \mathbb{N}$. Applying the modular ρ in the last inequality and using the monotonicity of ρ and moreover multiplying the both-sides of the above inequality by $\frac{1}{|\sigma(x,y)|}$, we have

$$\begin{aligned} \rho \left(\lambda_0^* \left(\frac{T_{m,n}f - f}{\sigma} \right) \right) &\leq \rho \left(3\lambda_0^* \frac{T_{m,n}(f - g_{k_0, k_0})}{\sigma} \right) \\ &\quad + \rho \left(3\lambda_0^* \left(\frac{T_{m,n}g_{k_0, k_0} - g_{k_0, k_0}}{\sigma} \right) \right) \\ &\quad + \rho \left(3\lambda_0^* \left(\frac{g_{k_0, k_0} - f}{\sigma} \right) \right). \end{aligned}$$

Hence, observing $|\sigma| \geq a > 0 (a = \max \{a_i : i = 0, 1, 2, 3\})$ we may write that

$$\begin{aligned} \rho \left(\lambda_0^* \left(\frac{T_{m,n}f - f}{\sigma} \right) \right) &\leq \rho \left(3\lambda_0^* \frac{T_{m,n}(f - g_{k_0, k_0})}{\sigma} \right) \\ &\quad + \rho \left(3\lambda_0^* \left(\frac{T_{m,n}g_{k_0, k_0} - g_{k_0, k_0}}{\sigma} \right) \right) \\ &\quad + \rho \left(\frac{3\lambda_0^*}{a} (g_{k_0, k_0} - f) \right). \end{aligned} \tag{13}$$

Then, it follows from (12) and (13) that

$$\begin{aligned} \rho \left(\lambda_0^* \left(\frac{T_{m,n}f - f}{\sigma} \right) \right) &\leq \varepsilon + \rho \left(3\lambda_0^* \frac{T_{m,n}(f - g_{k_0, k_0})}{\sigma} \right) \\ &\quad + \rho \left(3\lambda_0^* \left(\frac{T_{m,n}g_{k_0, k_0} - g_{k_0, k_0}}{\sigma} \right) \right). \end{aligned} \tag{14}$$

So, taking statistical limit superior as $m, n \rightarrow \infty$ in the both-sides of (14) and also using the facts that $g_{k_0, k_0} \in C^\infty(I^2)$ and $f - g_{k_0, k_0} \in X_{\mathbb{T}}$, we obtained from (5) that

$$\begin{aligned} st_2 - \limsup_{m,n} \rho \left(\lambda_0^* \left(\frac{T_{m,n}f - f}{\sigma} \right) \right) &\leq \varepsilon + R\rho(3\lambda_0^*(f - g_{k_0, k_0})) \\ &\quad + st_2 - \limsup_{m,n} \rho \left(3\lambda_0^* \left(\frac{T_{m,n}g_{k_0, k_0} - g_{k_0, k_0}}{\sigma} \right) \right), \end{aligned}$$

which gives

$$\begin{aligned}
 st_2 - \limsup_{m,n} \rho \left(\lambda_0^* \left(\frac{T_{m,n}f - f}{\sigma} \right) \right) \\
 \leq \varepsilon(R + 1) + st_2 - \limsup_{m,n} \rho \left(3\lambda_0^* \left(\frac{T_{m,n}g_{k_0,k_0} - g_{k_0,k_0}}{\sigma} \right) \right). \tag{15}
 \end{aligned}$$

By (8), since

$$st_2 - \lim_{m,n} \rho \left(3\lambda_0^* \left(\frac{T_{m,n}g_{k_0,k_0} - g_{k_0,k_0}}{\sigma} \right) \right) = 0,$$

we get

$$st_2 - \limsup_{m,n} \rho \left(3\lambda_0^* \left(\frac{T_{m,n}g_{k_0,k_0} - g_{k_0,k_0}}{\sigma} \right) \right) = 0. \tag{16}$$

Combining (15) with (16), we conclude that

$$st_2 - \limsup_{m,n} \rho \left(\lambda_0^* \left(\frac{T_{m,n}f - f}{\sigma} \right) \right) \leq \varepsilon(R + 1).$$

Since $\varepsilon > 0$ was arbitrary, we find

$$st_2 - \limsup_{m,n} \rho \left(\lambda_0^* \left(\frac{T_{m,n}f - f}{\sigma} \right) \right) = 0.$$

Furthermore, since $\rho \left(\lambda_0^* \left(\frac{T_{m,n}f - f}{\sigma} \right) \right)$ is non-negative for all $m, n \in \mathbb{N}$, we can easily show that

$$st_2 - \lim_{m,n} \rho \left(\lambda_0^* \left(\frac{T_{m,n}f - f}{\sigma} \right) \right) = 0,$$

which completes the proof. □

If the modular ρ satisfies the Δ_2 -condition, then one can get immediately the following result from Theorem 1.

Theorem 2. *Let $\mathbb{T} := \{T_{m,n}\}$, ρ and σ be the same as in Theorem 1. If ρ satisfies the Δ_2 -condition, then the following statements are equivalent:*

- (a) $st_2 - \lim_{m,n} \rho \left(\lambda \left(\frac{T_{m,n}e_i - e_i}{\sigma_i} \right) \right) = 0$ for every $\lambda > 0$ and $i = 0, 1, 2, 3$,
- (b) $st_2 - \lim_{m,n} \rho \left(\lambda \left(\frac{T_{m,n}f - f}{\sigma} \right) \right) = 0$ for every $\lambda > 0$ provided that f is any function belonging to $L^\rho(I^2)$ such that $f - g \in X_{\mathbb{T}}$ for every $g \in C^\infty(I^2)$.

If one replaces the scale function by a nonzero constant, then the condition (5) reduces to

$$st_2 - \limsup_{m,n} \rho (\lambda (T_{m,n}h)) \leq R\rho (\lambda h) \tag{17}$$

for every $h \in X_{\mathbb{T}}$, $\lambda > 0$ and for an absolute positive constant R . In this case, the next results which were obtained in [22] follows from our main theorems, Theorems 1 and 2.

Corollary 1 [22]. *Let ρ be a monotone, strongly finite, absolutely continuous and N -quasi semiconvex modular on $X(I^2)$. Let $\mathbb{T} := \{T_{m,n}\}$ be a double sequence of positive linear operators from D into $X(I^2)$ satisfying (17). If $\{T_{m,n}e_i\}$ is statistically strongly convergent to e_i for each $i = 0, 1, 2, 3$, then $\{T_{m,n}f\}$ is statistically modularly convergent to f provided that f is any function belonging to $L^\rho(I^2)$ such that $f - g \in X_{\mathbb{T}}$ for every $g \in C^\infty(I^2)$.*

Corollary 2 [22]. $\mathbb{T} := \{T_{m,n}\}$ and ρ be the same as in Corollary 1. If ρ satisfies the Δ_2 -condition, then the following statements are equivalent:

- (a) $\{T_{m,n}e_i\}$ is statistically strongly convergent to e_i for each $i = 0, 1, 2, 3$,
- (b) $\{T_{m,n}f\}$ is statistically strongly convergent to f provided that f is any function belonging to $L^\rho(I^2)$ such that $f - g \in X_{\mathbb{T}}$ for every $g \in C^\infty(I^2)$.

If one replaces the statistical limit by the Pringsheim limit, then the condition (5) reduces to

$$P - \limsup_{m,n} \rho \left(\lambda \left(\frac{T_{m,n}h}{\sigma} \right) \right) \leq R\rho(\lambda h) \tag{18}$$

for every $h \in X_{\mathbb{T}}, \lambda > 0$ and for an absolute positive constant R . In this case, the following results immediately follows from our Theorems 1 and 2.

Corollary 3. *Let ρ be a monotone, strongly finite, absolutely continuous and N -quasi semiconvex modular on $X(I^2)$. Let $\mathbb{T} := \{T_{m,n}\}$ be a double sequence of positive linear operators from D into $X(I^2)$ satisfying (18). Moreover suppose that $\sigma_i(x, y)$ is an unbounded function satisfying $|\sigma_i(x, y)| \geq a_i > 0 (i = 0, 1, 2, 3)$. If $\{T_{m,n}e_i\}$ is relatively strongly convergent to e_i for each $i = 0, 1, 2, 3$, then $\{T_{m,n}f\}$ is relatively modularly convergent to f provided that f is any function belonging to $L^\rho(I^2)$ such that $f - g \in X_{\mathbb{T}}$ for every $g \in C^\infty(I^2)$.*

Corollary 4. $\mathbb{T} := \{T_{m,n}\}, \rho$ and $\sigma_i (i = 0, 1, 2, 3)$ be the same as in Corollary 3. If ρ satisfies the Δ_2 -condition, then the following statements are equivalent:

- (a) $\{T_{m,n}e_i\}$ is relatively strongly convergent to e_i for each $i = 0, 1, 2, 3$,
- (b) $\{T_{m,n}f\}$ is relatively strongly convergent to f provided that f is any function belonging to $L^\rho(I^2)$ such that $f - g \in X_{\mathbb{T}}$ for every $g \in C^\infty(I^2)$.

3. Concluding Remarks

In this section, we display an example such that our Korovkin-type statistical approximation results in modular spaces are stronger than the Corollaries 1 and 3.

Example 2. Take $I = [0, 1]$ and φ, ρ^φ and $L_\varphi^\rho(I^2)$ be the same as in Example 1. Then consider the following bivariate Bernstein-Kantorovich operator $\mathbb{U} := \{U_{m,n}\}$ on the space $L_\varphi^\rho(I^2)$ which is defined by:

$$\begin{aligned}
 &U_{m,n}(f; x, y) \\
 &= \sum_{i=0}^m \sum_{k=0}^n p_{i,k}^{(m,n)}(x, y) (m+1)(n+1) \times \int_{i/(m+1)}^{(i+1)/(m+1)} \int_{k/(n+1)}^{(k+1)/(n+1)} f(t, s) ds dt
 \end{aligned} \tag{19}$$

for $x, y \in I$, where $p_{i,k}^{(m,n)}(x, y)$ defined by

$$p_{i,k}^{(m,n)}(x, y) = \binom{m}{i} \binom{n}{k} x^i y^k (1-x)^{m-i} (1-y)^{n-k}.$$

Also, it is clear that,

$$\sum_{i=0}^m \sum_{k=0}^n p_{i,k}^{(m,n)}(x, y) = 1. \tag{20}$$

Observe that the operator $U_{m,n}$ maps $L_\varphi^\rho(I^2)$ into itself. Because of (20), as in the proof of [2] Lemma 5.1 and similar to Example 1 [22], we can use the Jensen inequality and obtain that for every $f \in L_\varphi^\rho(I^2)$ and $m, n \in \mathbb{N}$ there is an absolute constant $M > 0$ such that

$$\rho^\varphi\left(\frac{U_{m,n}f}{\sigma}\right) \leq M\rho^\varphi(f).$$

Moreover, the property (18) is satisfied with the choice of $X_U := L_\varphi^\rho(I^2)$. Then, by Corollary 3, we know that, for any function $f \in L_\varphi^\rho(I^2)$ such that $f - g \in X_U$ for every $g \in C^\infty(I^2)$, $\{U_{m,n}f\}$ is relatively modularly convergent to f .

If $\varphi(x) = x^p$ for $1 \leq p < \infty, x \geq 0$, then $L_\varphi^\rho(I^2) = L_p(I^2)$. Moreover we have $\rho^\varphi(f) = \|f\|_{L_p}^p$. For $p = 1$, we have $\rho^\varphi(f) = \|f\|_{L_1}$. Then, using the operators $U_{m,n}$, we define the sequence of positive linear operators $V := \{V_{m,n}\}$ on $L_1(I^2)$ as follows:

$$\begin{aligned}
 &V_{m,n}(f; x, y) = (1 + g_{m,n}(x, y))U_{m,n}(f; x, y) \\
 &\text{for } f \in L_1(I^2), (x, y) \in I^2 \text{ and } m, n \in \mathbb{N}
 \end{aligned} \tag{21}$$

where $\{g_{m,n}\}$ is the same as in (3) and we choose $\sigma_i(x, y) = \sigma(x, y)$ ($i = 0, 1, 2, 3$), where $\sigma(x, y) = \begin{cases} 1, & (x, y) = (0, 0) \\ \frac{1}{x^2 y}, & (x, y) \in (0, 1] \times (0, 1] \end{cases}$. As in the proof of Lemma 5.1 [2] and similar to Example 1 [22], we get, for every $f \in L_1(I^2)$, $\lambda > 0$ and for positive constant C , that

$$st_2 - \limsup_{m,n} \left\| \lambda \left(\frac{V_{m,n}f}{\sigma} \right) \right\|_{L_1} \leq C \|\lambda f\|_{L_1}. \tag{22}$$

We now claim that

$$st_2 - \lim_{m,n} \left\| \lambda \left(\frac{V_{m,n}e_i - e_i}{\sigma} \right) \right\|_{L_1} = 0, \quad i = 0, 1, 2, 3. \tag{23}$$

Indeed, first observe that

$$\begin{aligned}
 V_{m,n}(e_0; x, y) &= 1 + g_{m,n}(x, y), \\
 V_{m,n}(e_1; x, y) &= (1 + g_{m,n}(x, y)) \left(\frac{mx}{m+1} + \frac{1}{2(m+1)} \right), \\
 V_{m,n}(e_2; x, y) &= (1 + g_{m,n}(x, y)) \left(\frac{ny}{n+1} + \frac{1}{2(n+1)} \right), \\
 V_{m,n}(e_3; x, y) &= (1 + g_{m,n}(x, y)) \left(\frac{m(m-1)x^2}{(m+1)^2} + \frac{2mx}{(m+1)^2} + \frac{1}{3(m+1)^2} \right. \\
 &\quad \left. + \frac{n(n-1)y^2}{(n+1)^2} + \frac{2ny}{(n+1)^2} + \frac{1}{3(n+1)^2} \right).
 \end{aligned}$$

So, we can see, for any $\lambda > 0$, that

$$\begin{aligned}
 &\left\| \lambda \left(\frac{V_{m,n}(e_0; x, y) - e_0(x, y)}{\sigma(x, y)} \right) \right\|_{L_1} \\
 &= \lambda \begin{cases} \int_0^1 \int_0^1 x^2 y dx dy, & m = k^2 \text{ and } n = l^2 \\ \frac{1}{n} \frac{1}{m} \int_0^1 \int_0^1 m^2 n (x^2 y - m n x^3 y^2) dx dy, & m \neq k^2 \text{ and } n \neq l^2 \end{cases}, \quad k, l = 1, 2, \dots, \\
 &= \lambda \begin{cases} \frac{1}{6} & m = k^2 \text{ and } n = l^2 \\ \frac{1}{12mn} & m \neq k^2 \text{ and } n \neq l^2 \end{cases}, \quad k, l = 1, 2, \dots, \tag{24}
 \end{aligned}$$

Now, since $P - \lim_{m,n \rightarrow \infty} \frac{1}{12mn} = 0$, we get

$$st_2 - \lim_{m,n} \left\| \lambda \left(\frac{V_{m,n}e_0 - e_0}{\sigma} \right) \right\|_{L_1} = 0.$$

which guarantees that (23) holds true for $i = 0$. Also, we have

$$\begin{aligned}
 &\left\| \lambda \left(\frac{V_{m,n}(e_1; x, y) - e_1(x, y)}{\sigma(x, y)} \right) \right\|_{L_1} \\
 &\leq \left\| \lambda \frac{g_{m,n}(x, y)}{\sigma(x, y)} \left(\frac{mx}{m+1} + \frac{1}{2(m+1)} \right) \right\|_{L_1} + \left\| \lambda \frac{x^2 y - 2x^3 y}{2(m+1)} \right\|_{L_1} \\
 &< \left\| \lambda \frac{g_{m,n}(x, y)}{\sigma(x, y)} \right\|_{L_1} + \frac{\lambda}{24(m+1)},
 \end{aligned}$$

because of $st_2 - \lim_{m,n} \left\| \lambda \frac{g_{m,n}(x, y)}{\sigma(x, y)} \right\|_{L_1} = 0$ and $P - \lim_{m,n} \frac{\lambda}{24(m+1)} = 0$, we get

$$st_2 - \lim_{m,n} \left\| \lambda \left(\frac{V_{m,n}(e_1; x, y) - e_1(x, y)}{\sigma(x, y)} \right) \right\|_{L_1} = 0.$$

Hence (23) is valid for $i = 1$. Similarly, we have

$$st_2 - \lim_{m,n} \left\| \lambda \left(\frac{V_{m,n}(e_2; x, y) - e_2(x, y)}{\sigma(x, y)} \right) \right\|_{L_1} = 0.$$

Finally, since

$$\begin{aligned} & \left\| \lambda \left(\frac{V_{m,n}(e_3; x, y) - e_3(x, y)}{\sigma(x, y)} \right) \right\|_{L_1} \\ & \leq \left\| \lambda \frac{g_{m,n}(x, y)}{\sigma(x, y)} \left(\frac{m(m-1)x^2}{(m+1)^2} + \frac{2mx}{(m+1)^2} \right. \right. \\ & \quad \left. \left. + \frac{1}{3(m+1)^2} \frac{n(n-1)y^2}{(n+1)^2} + \frac{2ny}{(n+1)^2} + \frac{1}{3(n+1)^2} \right) \right\|_{L_1} \\ & \quad + \left\| \lambda \left(\frac{(3m+1)x^4y}{(m+1)^2} + \frac{(3n+1)x^2y^3}{(n+1)^2} + \frac{2mx^3y}{(m+1)^2} + \frac{2nx^2y^2}{(n+1)^2} \right. \right. \\ & \quad \left. \left. + x^2y \left(\frac{1}{3(m+1)^2} + \frac{1}{3(n+1)^2} \right) \right) \right\|_{L_1} \\ & < 6 \left\| \lambda \frac{g_{m,n}(x, y)}{\sigma(x, y)} \right\|_{L_1} + \frac{\lambda(3m+1)}{10(m+1)^2} + \frac{\lambda(3n+1)}{12(n+1)^2} + \frac{\lambda m}{4(m+1)^2} \\ & \quad + \frac{2\lambda n}{9(n+1)^2} + \frac{\lambda}{6} \left(\frac{1}{3(m+1)^2} + \frac{1}{3(n+1)^2} \right), \end{aligned}$$

then we have,

$$st_2 - \lim_{m,n} \left\| \lambda \left(\frac{V_{m,n}(e_3; x, y) - e_3(x, y)}{\sigma(x, y)} \right) \right\|_{L_1} = 0.$$

So, our claim (23) holds true for each $i = 0, 1, 2, 3$ and for any $\lambda > 0$. Now, from (22) and (23), we can say that our sequence $V := \{V_{m,n}\}$ defined by (21) satisfy all assumptions of Theorem 1. Therefore, we conclude that

$$st_2 - \lim_{m,n} \left\| \lambda \left(\frac{V_{m,n}(f; x, y) - f(x, y)}{\sigma(x, y)} \right) \right\|_{L_1} = 0 \quad \text{for some } \lambda_0 > 0$$

holds for any $f \in L_1(I^2)$ such that $f - g \in X_V = L_1(I^2)$ for every $g \in C^\infty(I^2)$.

However, from (24) it can be seen that the sequence $\left\| \lambda \left(\frac{V_{m,n}(e_0; x, y) - e_0(x, y)}{\sigma(x, y)} \right) \right\|_{L_1}$ has two subsequences with different limit points. Since $P - \lim_{m,n} \left\| \lambda \left(\frac{V_{m,n}e_0 - e_0}{\sigma} \right) \right\|_{L_1} \neq 0$, Corollary 3 does not work for the sequence $V := \{V_{m,n}\}$. Also, since $st_2 - \lim_{m,n} \left\| \lambda (V_{m,n}e_0 - e_0) \right\|_{L_1} \neq 0$, Corollary 1 does not work for the sequence $V := \{V_{m,n}\}$ as well.

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