CREAT. MATH. INFORM. Volume **27** (2018), No. 2, Pages 101 - 110 Online version at https://creative-mathematics.cunbm.utcluj.ro/ Print Edition: ISSN 1584 - 286X; Online Edition: ISSN 1843 - 441X DOI: https://doi.org/10.37193/CMI.2018.02.01

Generalized weighted statistical convergence in intuitionistic fuzzy normed linear spaces

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ABSTRACT. In this paper, we introduce a new statistical convergence type, named weighted λ -statistical convergence to generalize the concept of weighted statistical convergence with respect to the intuitionistic fuzzy norm (μ, v) . Moreover, we establish its relation to weighted statistical convergence and a new summability method, named as $(\overline{N}_{\lambda}, p)$ -summability with respect to the intuitionistic fuzzy norm (μ, v) .

1. INTRODUCTION

Fuzzy set theory defined by Zadeh [41] has been investigated for the last 40 years. It has been studied many branches of mathematics, engineering and physics such as fuzzy topology [31], metric spaces [9, 18], chaos control [11], nonlinear dynamical systems [17], population dynamics [5] and quantum physics [25].

The theory of intuitionistic fuzzy set introduced by Atanassov [3] was investigated in decision making problems and e^{∞} -theory [4, 8]. The concept of intuitionistic fuzzy metric spaces was defined by Park [30]. Moreover, the notion of the intuitionistic fuzzy normed linear space was given by Saadati and Park [33].

As an extended definition of usual convergence, the concept of statistical convergence was given by Steinhaus [40] and Fast [10]. Many years later, statistical convergence was discussed by many researchers in the theory of Fourier analysis, ergodic theory and number theory. Furthermore, statistical convergence has been investigated in the view of summability theory benefiting from the works of Schoenberg [38], Salat [34], Connor [7], Fridy [12] and Fridy and Miller [13]. As a variant of statistical convergence, the concept of λ -statistical convergence of real sequences was given by Mursaleen [29]. Applications of statistical convergence and λ -statistical convergence in the study of intuitionistic fuzzy normed linear spaces can be found in [21, 22, 28] and [20, 23, 26, 29, 36, 37] respectively.

Let $p = (p_k)_{k=0}^{\infty}$ be a sequence of nonnegatif numbers such that $p_0 > 0$ and $P_n = \sum_{k=0}^{n} p_k \to \infty$ as $n \to \infty$. Set

$$t_n = \frac{1}{P_n} \sum_{k=0}^n p_k x_k, \ n = 0, 1, 2, \dots$$

 $x = (x_k)$ is (\overline{N}, p) -summable if $\lim_{n \to \infty} t_n = L$. In this case, we write $(\overline{N}, p) - \lim x = L$.

Following the introduction of (\overline{N}, p) -summability due to Hardy [15], the concept of statistical convergence was generalized by Karakaya and Chisti [19] and called this new method as weighted statistical convergence. The concept of weighted statistical convergence was altered by Mursaleen et al. [27]. The definition was given as follows:

Received: 18.09.2017. In revised form: 11.12.2017. Accepted: 23.02.2018

²⁰¹⁰ Mathematics Subject Classification. 40D15, 40G99.

Key words and phrases. statistical convergence, λ -statistical convergence, weighted statistical convergence, strong summability, intuitionistic fuzzy normed linear space.

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Let define the weighted density of $K \subseteq \mathbb{N}$ by $\delta_{\overline{N}}(K) = \lim_{n \to \infty} \frac{1}{P_n} |K_{P_n}|$ if, the limit exists. A sequence $x = (x_k)$ is weighted statistically convergent (or $S_{\overline{N}}$ - convergent) to L if, for every $\varepsilon > 0$,

$$\delta_{\overline{N}}(\{k \in \mathbb{N} : p_k | x_k - L| \ge \varepsilon\}) = 0,$$

or equivalently

$$\lim_{n \to \infty} \frac{1}{P_n} \left| \left\{ k \le P_n : p_k \left| x_k - L \right| \ge \varepsilon \right\} \right| = 0.$$

In this case, it is denoted by $S_{\overline{N}} - \lim x = L$.

In this study, we introduce the concepts of weighted λ -statistical convergence and $(\overline{N}_{\lambda}, p)$ -summability in intuitionistic fuzzy normed linear spaces. We also investigate some inclusion relations for these new concepts in intuitionistic fuzzy normed linear spaces.

2. BASIC DEFINITIONS

In this section, we give the basic definitions satisfied to give new summability methods in the intuitionistic fuzzy normed linear spaces.

Definition 2.1. ([29]) By a sequence $\lambda = (\lambda_n)$, we mean a nondecreasing sequence of positive numbers tending to ∞ such that $\lambda_{n+1} \leq \lambda_n + 1$, $\lambda_1 = 1$. The sequence $x = (x_k)$ is λ -statistical convergent to L if for each $\varepsilon > 0$,

$$\lim_{n \to \infty} \frac{1}{\lambda_n} |\{k \in I_n : |x_k - L| \ge \varepsilon\}| = 0,$$

where $I_n = [n - \lambda_n + 1, n]$. In this case, it is denoted by $S_{\lambda} - \lim x = L$.

Definition 2.2. ([6]) Let $p = (p_k)_{k=0}^{\infty}$ be a sequence of nonnegatif numbers such that $p_0 > 0$ and $P_{\lambda_n} := \sum_{k \in I_n} p_k \to \infty$ as $n \to \infty$ and $\sigma_n := \frac{1}{P_{\lambda_n}} \sum_{k \in I_n} p_k x_k$.

- i) A sequence $x = (x_k)$ is $(\overline{N}_{\lambda}, p)$ -summable to L if $\lim_{n \to \infty} \sigma_n = L$. If $\lambda_n = n$, $x = (x_k)$ is $(\overline{N}_{\lambda}, p)$ -summability is reduced to (\overline{N}, p) -summability.
- ii) A sequence $x = (x_k)$ is strongly $(\overline{N}_{\lambda}, p)$ -summable to *L* if,

$$\lim_{n \to \infty} \frac{1}{P_{\lambda_n}} \sum_{k \in I_n} p_k |x_k - L| = 0.$$

Definition 2.3. ([6]) Let $K \subseteq \mathbb{N}$. The number

$$\delta_{\overline{N}_{\lambda}}(K) = \lim_{n \to \infty} \frac{1}{P_{\lambda_n}} \left| \{k \le P_{\lambda_n} : k \in K\} \right|$$

is said to be weighted λ -density of $K \subseteq \mathbb{N}$ if the limit exists. In case $\lambda_n = n$, weighted λ -density reduces to the weighted density defined in [27].

Definition 2.4. ([6]) A sequence $x = (x_k)$ is said to be weighted λ -statistically convergent (or $S_{\overline{N}_{\lambda}}$ - convergent) to *L* if, for every $\varepsilon > 0$,

$$\lim_{n \to \infty} \frac{1}{P_{\lambda_n}} \left| \left\{ k \le P_{\lambda_n} : p_k \left| x_k - L \right| \ge \varepsilon \right\} \right| = 0.$$

In this case, we write $S_{\overline{N}_{\lambda}} - \lim x = L$. $S_{\overline{N}_{\lambda}}$ denotes the set of all weighted λ -statistically convergent sequences.

Definition 2.5. ([39]) A binary operation $* : [0,1] \times [0,1] \rightarrow [0,1]$ is said to be a continuos *t*-norm if, it satisfies the following conditions:

- i) * is associative and commutative,
- ii) * is continuous,
- iii) a * 1 = a for all $a \in [0, 1]$,
- iv) $a * b \le c * d$ whenever $a \le c$ and $b \le d$ for each $a, b, c, d \in [0, 1]$.

Definition 2.6. ([39]) A binary operation \circ : $[0,1] \times [0,1] \rightarrow [0,1]$ is said to be a continuos *t*-conorm if, it satisfies the following conditions:

- i) o is associative and commutative,
- ii) \circ is continuous,
- iii) $a \circ 0 = a$ for all $a \in [0, 1]$,
- iv) $a \circ b \leq c \circ d$ whenever $a \leq c$ and $b \leq d$ for each $a, b, c, d \in [0, 1]$.

Definition 2.7. ([32]) The five-tuple $(X, \mu, v, *, \circ)$ is said to be intuitionistic fuzzy normed linear space (or shortly IFNLS) where *X* is a linear space over a field *F*, * is a continuous *t*-norm, \circ is a continuous *t*-conorm, μ , v are fuzzy sets on $X \times (0, \infty)$, μ denotes the degree of membership and v denotes the degree of nonmembership of $(x, t) \in X \times (0, \infty)$ satisfying the following conditions for every $x, y \in X$ and s, t > 0:

- i) $\mu(x,t) + \upsilon(x,t) \le 1$,
- ii) $\mu(x,t) > 0$,
- iii) $\mu(x,t) = 1$ if and only if x = 0,
- iv) $\mu(\alpha x, t) = \mu\left(x, \frac{t}{|\alpha|}\right)$ if $\alpha \neq 0$,
- v) $\mu(x,t) * \mu(y,s) \le \mu(x+y,t+s),$
- vi) $\mu(x,.): (0,\infty) \rightarrow [0,1]$ is continuous,
- vii) $\lim_{t\to\infty} \mu(x,t) = 1$ and $\lim_{t\to0} \mu(x,t) = 0$,
- viii) v(x,t) < 1,

ix)
$$v(x,t) = 0$$
 if and only if $x = 0$,

x)
$$v(\alpha x, t) = v\left(x, \frac{t}{|\alpha|}\right)$$
 if $\alpha \neq 0$,

xi) $v(x,t) \circ v(y,s) \ge v(x+y,s+t)$,

- xii) $v(x,.): (0,\infty) \rightarrow [0,1]$ is continuous,
- xiii) $\lim_{t \to \infty} v(x,t) = 0$ and $\lim_{t \to 0} v(x,t) = 1$.

Definition 2.8. ([32]) Let $(X, \mu, v, *, \circ)$ be an IFNLS. A sequence $x = (x_k)$ be a sequence in X is convergent to $L \in X$ with respect to the intuitionistic fuzzy norm (μ, v) if, for every $\varepsilon > 0$ and t > 0, there exists $k_0 \in \mathbb{N}$ such that $\mu(x_k - L, t) > 1 - \varepsilon$ and $v(x_k - L, t) < \varepsilon$ for all $k \ge k_0$ where $k \in \mathbb{N}$. It is denoted by $(\mu, v) - \lim x = L$.

Theorem 2.1. ([35]) Let $(X, \mu, v, *, \circ)$ be an IFNLS. Then, a sequence $x = (x_k)$ in X is convergent to $L \in X$ with respect to the intuitionistic fuzzy norm (μ, v) if and only if $\lim_{k \to \infty} \mu (x_k - L, t) = 1$ and $\lim_{k \to \infty} v (x_k - L, t) = 0$.

3. GENERALIZED WEIGHTED STATISTICAL CONVERGENCE IN IFNLS

In this section, we generalize the concept of weighted statistical convergence with respect to the intuitionistic fuzzy norm (μ, v) following the line of Altundağ and Kamber [2]. Moreover, we investigate some connections between this definition with the concept of weighted statistical convergence and $(\overline{N}_{\lambda}, p)$ -summability with respect to the intuitionistic fuzzy norm (μ, v) . **Definition 3.9.** Let $(X, \mu, v, *, \circ)$ be an IFNLS. A sequence $x = (x_k)$ in X is said to be $(\overline{N}_{\lambda}, p)$ -summable to $L \in X$ with respect to the intuitionistic fuzzy norm (μ, v) (or $(\overline{N}_{\lambda}, p)^{(\mu, v)}$ -summable to $L \in X$) if, for every $\varepsilon > 0$ and t > 0, there exists $n_0 \in \mathbb{N}$ such that

$$\frac{1}{P_{\lambda_{n}}}\sum_{k\in I_{n}}\mu\left(p_{k}\left(x_{k}-L\right),t\right)>1-\varepsilon$$

and

$$\frac{1}{P_{\lambda_{n}}}\sum_{k\in I_{n}}\upsilon\left(p_{k}\left(x_{k}-L\right),t\right)<\varepsilon$$

for all $n \ge n_0$. In this case, we write $(\overline{N}_{\lambda}, p)^{(\mu, v)} - \lim x = L$.

Definition 3.10. Let $(X, \mu, v, *, \circ)$ be an IFNLS. A sequence $x = (x_k)$ in X is said to be weighted λ -statistically convergent to $L \in X$ with respect to the intuitionistic fuzzy norm (μ, v) (or $S_{\overline{N_{\lambda}}}^{(\mu,v)}$ - convergent to $L \in X$) if, for every $\varepsilon > 0$ and t > 0,

$$\delta_{\overline{N}_{\lambda}}\left(\left\{k \in \mathbb{N} : \mu\left(p_{k}\left(x_{k}-L\right), t\right) \leq 1-\varepsilon \text{ or } \upsilon\left(p_{k}\left(x_{k}-L\right), t\right) \geq \varepsilon\right\}\right) = 0, \quad (3.1)$$

or equivalently

$$\lim_{n \to \infty} \frac{1}{P_{\lambda_n}} \left| \left\{ k \le P_{\lambda_n} : \mu \left(p_k \left(x_k - L \right), t \right) \le 1 - \varepsilon \text{ or } \upsilon \left(p_k \left(x_k - L \right), t \right) \ge \varepsilon \right\} \right| = 0.$$

In this case, we write $S_{\overline{N}}^{(\mu,v)} - \lim x = L$.

- **Remark 3.1.** (i) If $\lambda_n = n$ for all $n \in \mathbb{N}$ in Definition 3.9, then $(\overline{N}_{\lambda}, p)^{(\mu, v)}$ summability is reduced to $(\overline{N}, p)^{(\mu, v)}$ summability introduced in [2].
 - (ii) If $p_k = 1$ for all $k \in \mathbb{N}$ in Definition 3.9, then $(\overline{N}_{\lambda}, p)^{(\mu, v)}$ summability is reduced to $(V, \lambda)^{(\mu, v)}$ summability introduced in [16].
 - (iii) If $p_k = 1$ for all $k \in \mathbb{N}$ in Definition 3.10, then the concept of $S_{\overline{N}_{\lambda}}^{(\mu,v)}$ convergence is reduced to the concept of $S_{\lambda}^{(\mu,v)}$ convergence introduced in [26].
 - (iv) If $\lambda_n = n$ for all $n \in \mathbb{N}$ in Definition 3.10, then the concept of $S_{\overline{N_{\lambda}}}^{(\mu,\upsilon)}$ convergence is reduced to the concept of $S_{\overline{N}}^{(\mu,\upsilon)}$ convergence introduced in [2].

It is easily seen that if $S_{\overline{N_{\lambda}}}^{(\mu,\upsilon)} - \lim x = L$, then $S_{\overline{N}}^{(\mu,\upsilon)} - \lim x = L$, since $\frac{\lambda_n}{n}$ is bounded by 1 and so $\frac{P_{\lambda_n}}{P_n}$ is bounded by 1. We prove the following theorem for the reverse situation.

Theorem 3.2. Let $(X, \mu, v, *, \circ)$ be an IFNLS and $x = (x_k)$ be a sequence in X. If $\liminf_{n \to \infty} \frac{P_{\lambda_n}}{P_n} > 0$ and $S_{\overline{N}}^{(\mu,v)} - \lim_{n \to \infty} x = L$, then $S_{\overline{N}_{\lambda}}^{(\mu,v)} - \lim_{n \to \infty} x = L$.

Proof. Suppose that $\liminf_{n\to\infty} \frac{P_{\lambda_n}}{P_n} > 0$. Then, there exists $\delta > 0$ such that $\delta \leq \frac{P_{\lambda_n}}{P_n} \leq 1$. Then for every $\varepsilon > 0$ and t > 0, we have

$$\frac{1}{P_n} \left| \left\{ k \le P_n : \mu \left(p_k \left(x_k - L \right), t \right) \le 1 - \varepsilon \text{ or } v \left(p_k \left(x_k - L \right), t \right) \ge \varepsilon \right\} \right|$$

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$$\geq \frac{\delta}{P_{\lambda_n}} \left| \left\{ k \leq P_{\lambda_n} : \mu \left(p_k(x_k - L), t \right) \leq 1 - \varepsilon \text{ or } v \left(p_k(x_k - L), t \right) \geq \varepsilon \right\} \right|.$$

By the assumption, it follows that $S_{\overline{N}_{\lambda}}^{(\mu,\upsilon)} - \lim x = L$.

In the following theorems, we establish the relation between $S_{\overline{N}_{\lambda}}^{(\mu,\nu)}$ - convergence and $(\overline{N}_{\lambda}, p)^{(\mu,\nu)}$ - summability in intuitionistic fuzzy normed linear spaces.

Theorem 3.3. Let $(X, \mu, v, *, \circ)$ be an IFNLS, $x = (x_k)$ be a sequence in X and $\frac{P_{\lambda_n}}{\lambda_n} \ge 1$ for all $n \in \mathbb{N}$. If $x = (x_k)$ in X is $(\overline{N}_{\lambda}, p)^{(\mu, v)}$ - summable to $L \in X$, then $x = (x_k)$ is $S_{\overline{N}_{\lambda}}^{(\mu, v)}$ - convergent to $L \in X$.

Proof. Suppose that $(\overline{N}_{\lambda}, p)^{(\mu, v)} - \lim x = L$. For every $\varepsilon > 0$ and t > 0, let $K_{D} = \{k \leq P_{\lambda} : \mu(p, (x, -L), t) \leq 1 - \varepsilon \text{ or } \mu(p, (x, -L), t) > \epsilon\}$

$$\mathbf{A}_{P_{\lambda_n}}(\varepsilon) = \{k \le P_{\lambda_n} : \mu(p_k(x_k - L), t) \le 1 - \varepsilon \text{ or } v(p_k(x_k - L), t) \ge \varepsilon\}$$

and

$$K_{P_{\lambda_n}}^c\left(\varepsilon\right) = \left\{k \le P_{\lambda_n} : \mu\left(p_k\left(x_k - L\right), t\right) > 1 - \varepsilon \text{ and } \upsilon\left(p_k\left(x_k - L\right), t\right) < \varepsilon\right\}.$$

Then,

$$\frac{1}{P_{\lambda_n}} \sum_{k \in I_n} \mu\left(p_k\left(x_k - L\right), t\right) = \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \mu\left(p_k\left(x_k - L\right), t\right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}^c(\varepsilon)}} \mu\left(p_k\left(x_k - L\right), t\right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}^c(\varepsilon)}} \mu\left(p_k\left(x_k - L\right), t\right) + \frac{1}{P_n} \left| K_{P_{\lambda_n}}^c(\varepsilon) \right| \left(1 - \varepsilon\right).$$

$$(3.2)$$

By inequality (3.2.), we have $\lim_{n\to\infty} \frac{1}{P_{\lambda_n}} \left| K_{P_{\lambda_n}}^c(\varepsilon) \right| = 1$. On the other hand, for every $\varepsilon > 0$ and t > 0,

$$\frac{1}{P_{\lambda_n}} \sum_{k \in I_n} \upsilon \left(p_k \left(x_k - L \right), t \right) = \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}^c(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_$$

By inequality (3.3.), we have $\lim_{n \to \infty} \frac{1}{P_{\lambda_n}} |K_{P_{\lambda_n}}(\varepsilon)| = 0$. Hence, we obtain $S_{\overline{N}_{\lambda}}^{(\mu,\nu)} - \lim x = L$.

The following example shows that the converse of Theorem 3.3. is not valid in general.

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Example 3.1. Let $(\mathbb{R}, |.|)$ denote the space of real numbers with the usual norm and let a * b = ab and $a \circ b = \min \{a + b, 1\}$ for all $a, b \in [0, 1]$. For all $x \in \mathbb{R}$ and every t > 0, consider $\mu(x, t) := \frac{t}{t+|x|}$ and $v(x, t) := \frac{|x|}{t+|x|}$. In this case, $(\mathbb{R}, \mu, v, *, \circ)$ is an IFNLS. Consider $\lambda_n = n$ for all $n \in \mathbb{N}$ and $p_k = \frac{1}{k+1}$ for all $k \in \mathbb{N}$ and define a sequence $x = (x_k)$ whose terms are given by

$$x_k = \begin{cases} \sqrt{k}, & \text{if } k = m^8 (m \in \mathbb{N}), \\ 0, & \text{otherwise.} \end{cases}$$

Then, for every $0 < \varepsilon < 1$ and for every t > 0, let

$$K_{P_n}(\varepsilon) = \{k \le P_n : \mu(p_k x_k, t) \le 1 - \varepsilon \text{ or } v(p_k x_k, t) \ge \varepsilon\}.$$

Since

$$K_{P_n}(\varepsilon) = \left\{ k \le P_n : \frac{t}{t + \frac{|x_k|}{k+1}} \le 1 - \varepsilon \text{ or } \frac{\frac{|x_k|}{k+1}}{t + \frac{|x_k|}{k+1}} \ge \varepsilon \right\}$$
$$= \left\{ k \le P_n : |x_k| \ge \frac{(k+1)\varepsilon t}{(1-\varepsilon)} \right\} \subseteq \left\{ k \le P_n : x_k = \sqrt{k} \right\},$$

we have

$$\lim_{n \to \infty} \frac{1}{P_n} \left| K_{P_n} \left(\varepsilon \right) \right| = 0.$$

So, by definition 3.10. we obtain $S_{\overline{N}_{\lambda}}^{(\mu,v)} - \lim x = 0$. However, since as $n \to \infty$,

$$\frac{1}{P_n}\sum_{k=1}^n \mu\left(p_k x_k, t\right) \to \infty$$

and

$$\frac{1}{P_n}\sum_{k=1}^n \upsilon\left(p_k x_k, t\right) \to \infty,$$

 $x = (x_k)$ is not $(\overline{N}_{\lambda}, p)$ -summable to 0 with respect to intuitionistic fuzzy norm (μ, v) .

Theorem 3.4. Let $(X, \mu, \upsilon, *, \circ)$ be an IFNLS, $x = (x_k)$ in X and $\frac{P_{\lambda_n}}{\lambda_n} \ge 1$ for all $n \in \mathbb{N}$. If $S_{\overline{N_\lambda}}^{(\mu,\upsilon)} - \lim x = L, \mu(p_k(x_k - L), t) \ge 1 - M$ and $\upsilon(p_k(x_k - L), t) \le M$ for all $k \in \mathbb{N}$, then $(\overline{N_\lambda}, p)^{(\mu,\upsilon)} - \lim x = L$.

Proof. Suppose that $S_{\overline{N}_{\lambda}}^{(\mu,\upsilon)} - \lim x = L$. Then, for every $\varepsilon > 0$ and t > 0,

$$K_{P_{\lambda_n}}\left(\varepsilon\right) = \left\{k \le P_{\lambda_n} : \mu\left(p_k\left(x_k - L\right), t\right) \le 1 - \varepsilon \quad \text{or } \upsilon\left(p_k\left(x_k - L\right), t\right) \ge \varepsilon\right\}$$

and

$$K_{P_{\lambda_n}}^c\left(\varepsilon\right) = \left\{k \le P_{\lambda_n} : \mu\left(p_k\left(x_k - L\right), t\right) > 1 - \varepsilon \text{ and } \upsilon\left(p_k\left(x_k - L\right), t\right) < \varepsilon\right\}.$$

Hence,

Generalized weighted ...

$$\frac{1}{P_{\lambda_n}} \sum_{k \in I_n} \mu\left(p_k\left(x_k - L\right), t\right) = \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \mu\left(p_k\left(x_k - L\right), t\right) + \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}^c(\varepsilon)}} \mu\left(p_k\left(x_k - L\right), t\right) = S_1\left(n\right) + S_2\left(n\right)$$

where

$$S_{1}(n) = \frac{1}{P_{\lambda_{n}}} \sum_{\substack{k \in I_{n} \\ k \in K_{P_{\lambda_{n}}}(\varepsilon)}} \mu\left(p_{k}\left(x_{k}-L\right), t\right)$$
(3.4)

and

$$S_{2}(n) = \frac{1}{P_{\lambda_{n}}} \sum_{\substack{k \in I_{n} \\ k \in K_{P_{\lambda_{n}}}^{c}(\varepsilon)}} \mu\left(p_{k}\left(x_{k}-L\right), t\right).$$
(3.5)

If $k \in K_{P_{\lambda_n}}(\varepsilon)$, then

$$S_{1}(n) = \frac{1}{P_{\lambda_{n}}} \sum_{\substack{k \in I_{n} \\ k \in K_{P_{\lambda_{n}}}(\varepsilon)}} \mu\left(p_{k}\left(x_{k}-L\right), t\right) \ge \frac{\left|K_{P_{\lambda_{n}}}\left(\varepsilon\right)\right|}{P_{\lambda_{n}}}\left(1-M\right).$$
(3.6)

Since $S_{\overline{N}_{\lambda}}^{(\mu,\upsilon)} - \lim x = L$, we have

$$\lim_{n \to \infty} S_1(n) \ge 0. \tag{3.7}$$

If $k\in K^{c}_{_{P_{\lambda_{n}}}}(\varepsilon)$, then we have

$$S_{2}(n) = \frac{1}{P_{\lambda_{n}}} \sum_{\substack{k \in I_{n} \\ k \in K_{P_{\lambda_{n}}}^{c}(\varepsilon)}} \mu\left(p_{k}\left(x_{k}-L\right), t\right) > \frac{\left|K_{P_{\lambda_{n}}}^{c}\left(\varepsilon\right)\right|}{P_{\lambda_{n}}}\left(1-\varepsilon\right)$$
(3.8)

which yields that

$$\lim_{n \to \infty} S_2(n) > (1 - \varepsilon).$$
(3.9)

Using equalities (3.4.)-(3.5.) and inequalities (3.6.)-(3.9.), we get

$$\lim_{n \to \infty} \frac{1}{P_{\lambda_n}} \sum_{k \in I_n} \mu\left(p_k\left(x_k - L\right), t\right) = 1.$$
(3.10)

Similarly,

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$$\frac{1}{P_{\lambda_n}} \sum_{k \in I_n} \upsilon \left(p_k \left(x_k - L \right), t \right) = \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right)$$
$$+ \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}(\varepsilon)}} \upsilon \left(p_k \left(x_k - L \right), t \right)$$
$$= S_3 \left(n \right) + S_4 \left(n \right)$$

where

$$S_{3}(n) = \frac{1}{P_{\lambda_{n}}} \sum_{\substack{k \in I_{n} \\ k \in K_{P_{\lambda_{n}}}(\varepsilon)}} v\left(p_{k}\left(x_{k} - L\right), t\right)$$

and

$$S_4(n) = \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}^c(\varepsilon)}} \upsilon(p_k(x_k - L), t).$$

If $k \in K_{P_{\lambda_{n}}}\left(\varepsilon\right)$, then as $n \to \infty$

$$S_{3}(n) = \frac{1}{P_{\lambda_{n}}} \sum_{\substack{k \in I_{n} \\ k \in K_{P_{\lambda_{n}}}(\varepsilon)}} \upsilon \left(p_{k} \left(x_{k} - L \right), t \right)$$

$$\leq \frac{1}{P_{\lambda_{n}}} \left| K_{P_{\lambda_{n}}}(\varepsilon) \right| M \to 0$$
(3.11)

since
$$\lim_{n \to \infty} \frac{1}{P_{\lambda_n}} \left| K_{P_{\lambda_n}} \left(\varepsilon \right) \right| = 0.$$

If $k \in K_{P_{\lambda_n}}^c \left(\varepsilon \right)$,

$$S_4 \left(n \right) = \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}^c \left(\varepsilon \right)}} \upsilon \left(p_k \left(x_k - L \right), t \right)$$

$$< \frac{1}{P_{\lambda_n}} \sum_{\substack{k \in I_n \\ k \in K_{P_{\lambda_n}}^c \left(\varepsilon \right)}} \varepsilon = \frac{1}{P_{\lambda_n}} \left| K_{P_{\lambda_n}}^c \left(\varepsilon \right) \right| \varepsilon$$
(3.12)

which yields that

$$\lim_{n \to \infty} S_4(n) < \varepsilon. \tag{3.13}$$

By inequalities (3.11.)-(3.13.), we get

$$\lim_{n \to \infty} \frac{1}{P_{\lambda_n}} \sum_{k \in I_n} \upsilon \left(p_k \left(x_k - L \right), t \right) = 0.$$
(3.14)

Using equalities (3.10.) and (3.14.), we have

$$\left(\overline{N}_{\lambda}, p\right)^{(\mu, v)} - \lim x = L.$$
(3.15)

4. CONCLUSION

In this paper, we defined a new variant of summability method, called as $(\overline{N}_{\lambda}, p)$ -summability in IFNLS and used this summability method to introduce a new variant of statistical convergence, called as weighted λ -statistical convergence in IFNLS. Moreover, we investigate some relations between these concepts. In establishing most of the proofs, we have benefited from the properties of intuitionistic fuzzy norm. The obtained results and relations in this paper are more general than the corresponding results for normed linear spaces.

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