

# **m-almost everywhere convergence of intuitionistic fuzzy observables induced by Borel measurable function**

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**Abstract:** In paper [4] we studied the upper and the lower limits of sequence of intuitionistic fuzzy observables. We used an intuitionistic fuzzy state  $\mathbf{m}$  for a definition the notion of almost everywhere convergence. We compared two concepts of  $\mathbf{m}$ -almost everywhere convergence. The aim of this paper is to show the connection between almost everywhere convergence in classical probability space induced by Kolmogorov construction and  $\mathbf{m}$ -almost everywhere convergence in intuitionistic fuzzy space. We studied the sequence of intuitionistic fuzzy observables induced by Borel measurable function.

**Keywords:** Intuitionistic fuzzy event, Intuitionistic fuzzy observable, Intuitionistic fuzzy state, Joint intuitionistic fuzzy observable, Product, Upper limit, Lower limit,  $\mathbf{m}$ -almost everywhere convergence, Function of several intuitionistic fuzzy observables, Borel measurable function, Kolmogorov construction.

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## **1 Introduction**

In [1–3] K. T. Atanassov introduced the notion of intuitionistic fuzzy sets. Then in [7] B. Riečan defined the intuitionistic fuzzy state on the family of intuitionistic fuzzy events

$$\mathcal{F} = \{(\mu_A, \nu_A) ; \mu_A + \nu_A \leq 1_\Omega\},$$

where  $\mu_A, \nu_A$  are  $\mathcal{S}$ -measurable functions,  $\mu_A, \nu_A : \Omega \rightarrow [0, 1]$ , as a mapping  $\mathbf{m}$  from the family  $\mathcal{F}$  to the set  $R$  by the formula

$$\mathbf{m}((\mu_A, \nu_A)) = (1 - \alpha) \int_{\Omega} \mu_A dP + \alpha \left( 1 - \int_{\Omega} \nu_A dP \right),$$

where  $P : \mathcal{S} \rightarrow [0, 1]$  is a probability measure and  $\alpha \in [0, 1]$ .

In paper [4] we defined the upper and the lower limits for sequence of intuitionistic fuzzy observables. We used an intuitionistic fuzzy state  $\mathbf{m}$  for a definition the notion of almost everywhere convergence. We compared two concepts of  $\mathbf{m}$ -almost everywhere convergence.

In this paper we study the  $\mathbf{m}$ -almost everywhere convergence of sequence of intuitionistic fuzzy observables  $g_n(x_1, \dots, x_n) : \mathcal{B}(R) \rightarrow \mathcal{F}$  given by

$$g_n(x_1, \dots, x_n) = h_n \circ g_n^{-1},$$

where  $h_n : \mathcal{B}(R^n) \rightarrow \mathcal{F}$  is the joint intuitionistic fuzzy observable of intuitionistic fuzzy observables  $x_1, \dots, x_n$  and  $g_n : R^n \rightarrow R$  is a Borel measurable function. We show the connection between  $\mathbf{m}$ -almost everywhere convergence of this sequence of intuitionistic fuzzy observables and  $P$ -almost everywhere convergence of random variables in classical probability space induced by Kolmogorov construction.

**Remark.** Note that in a whole text we use a notation “IF” in short as the phrase “intuitionistic fuzzy.”

## 2 IF-events, IF-states and IF-observables

First we start with definitions of basic notions.

**Definition 2.1.** Let  $\Omega$  be a nonempty set. An IF-set  $\mathbf{A}$  on  $\Omega$  is a pair  $(\mu_A, \nu_A)$  of mappings  $\mu_A, \nu_A : \Omega \rightarrow [0, 1]$  such that  $\mu_A + \nu_A \leq 1_{\Omega}$ .

**Definition 2.2.** Start with a measurable space  $(\Omega, \mathcal{S})$ . Hence  $\mathcal{S}$  is a  $\sigma$ -algebra of subsets of  $\Omega$ . An IF-event is called an IF-set  $\mathbf{A} = (\mu_A, \nu_A)$  such that  $\mu_A, \nu_A : \Omega \rightarrow [0, 1]$  are  $\mathcal{S}$ -measurable.

The family of all IF-events on  $(\Omega, \mathcal{S})$  will be denoted by  $\mathcal{F}$ ,  $\mu_A : \Omega \rightarrow [0, 1]$  will be called the membership function,  $\nu_A : \Omega \rightarrow [0, 1]$  be called the non-membership function.

If  $\mathbf{A} = (\mu_A, \nu_A) \in \mathcal{F}$ ,  $\mathbf{B} = (\mu_B, \nu_B) \in \mathcal{F}$ , then we define the Łukasiewicz binary operations  $\oplus, \odot$  on  $\mathcal{F}$  by

$$\begin{aligned} \mathbf{A} \oplus \mathbf{B} &= ((\mu_A + \mu_B) \wedge 1_{\Omega}, (\nu_A + \nu_B - 1_{\Omega}) \vee 0_{\Omega}), \\ \mathbf{A} \odot \mathbf{B} &= ((\mu_A + \mu_B - 1_{\Omega}) \vee 0_{\Omega}, (\nu_A + \nu_B) \wedge 1_{\Omega}) \end{aligned}$$

and the partial ordering is then given by

$$\mathbf{A} \leq \mathbf{B} \iff \mu_A \leq \mu_B, \nu_A \geq \nu_B.$$

In paper we use max-min connectives defined by

$$\mathbf{A} \vee \mathbf{B} = (\mu_A \vee \mu_B, \nu_A \wedge \nu_B),$$

$$\mathbf{A} \wedge \mathbf{B} = (\mu_A \wedge \mu_B, \nu_A \vee \nu_B)$$

and the de Morgan rules

$$(a \vee b)^* = a^* \wedge b^*,$$

$$(a \wedge b)^* = a^* \vee b^*,$$

where  $a^* = 1 - a$ .

**Example 2.3.** Fuzzy set  $f : \Omega \rightarrow [0, 1]$  can be regarded as IF-set, if we put

$$\mathbf{A} = (f, 1_\Omega - f).$$

If  $f = \chi_A$ , then the corresponding IF-set has the form

$$\mathbf{A} = (\chi_A, 1_\Omega - \chi_A) = (\chi_A, \chi_{A'}).$$

In this case  $\mathbf{A} \oplus \mathbf{B}$  corresponds to the union of sets,  $\mathbf{A} \odot \mathbf{B}$  to the intersection of sets and  $\leq$  to the set inclusion.

In the IF-probability theory [7, 9] instead of the notion of probability, we use the notion of state.

**Definition 2.4.** Let  $\mathcal{F}$  be the family of all IF-events in  $\Omega$ . A mapping  $\mathbf{m} : \mathcal{F} \rightarrow [0, 1]$  is called an IF-state, if the following conditions are satisfied:

- (i)  $\mathbf{m}((1_\Omega, 0_\Omega)) = 1$ ,  $\mathbf{m}((0_\Omega, 1_\Omega)) = 0$ ;
- (ii) if  $\mathbf{A} \odot \mathbf{B} = (0_\Omega, 1_\Omega)$  and  $\mathbf{A}, \mathbf{B} \in \mathcal{F}$ , then  $\mathbf{m}(\mathbf{A} \oplus \mathbf{B}) = \mathbf{m}(\mathbf{A}) + \mathbf{m}(\mathbf{B})$ ;
- (iii) if  $\mathbf{A}_n \nearrow \mathbf{A}$  (i.e.  $\mu_{A_n} \nearrow \mu_A$ ,  $\nu_{A_n} \searrow \nu_A$ ), then  $\mathbf{m}(\mathbf{A}_n) \nearrow \mathbf{m}(\mathbf{A})$ .

Probably the most useful result in the IF-state theory is the following representation theorem ([7]):

**Theorem 2.5.** To each IF-state  $\mathbf{m} : \mathcal{F} \rightarrow [0, 1]$  there exists exactly one probability measure  $P : \mathcal{S} \rightarrow [0, 1]$  and exactly one  $\alpha \in [0, 1]$  such that

$$\mathbf{m}(\mathbf{A}) = (1 - \alpha) \int_{\Omega} \mu_A dP + \alpha \left( 1 - \int_{\Omega} \nu_A dP \right)$$

for each  $\mathbf{A} = (\mu_A, \nu_A) \in \mathcal{F}$ .

The third basic notion in the probability theory is the notion of an observable. Let  $\mathcal{J}$  be the family of all intervals in  $R$  of the form

$$[a, b) = \{x \in R : a \leq x < b\}.$$

Then the  $\sigma$ -algebra  $\sigma(\mathcal{J})$  is denoted  $\mathcal{B}(R)$  and it is called the  $\sigma$ -algebra of Borel sets, its elements are called Borel sets.

**Definition 2.6.** By an IF-observable on  $\mathcal{F}$  we understand each mapping  $x : \mathcal{B}(R) \rightarrow \mathcal{F}$  satisfying the following conditions:

- (i)  $x(R) = (1_\Omega, 0_\Omega)$ ,  $x(\emptyset) = (0_\Omega, 1_\Omega)$ ;
- (ii) if  $A \cap B = \emptyset$ , then  $x(A) \odot x(B) = (0_\Omega, 1_\Omega)$  and  $x(A \cup B) = x(A) \oplus x(B)$ ;
- (iii) if  $A_n \nearrow A$ , then  $x(A_n) \nearrow x(A)$ .

If we denote  $x(A) = (x^\flat(A), 1_\Omega - x^\sharp(A))$  for each  $A \in \mathcal{B}(R)$ , then  $x^\flat, x^\sharp : \mathcal{B}(R) \rightarrow \mathcal{T}$  are observables, where  $\mathcal{T} = \{f : \Omega \rightarrow [0, 1]; f \text{ is } \mathcal{S} - \text{measurable}\}$ .

**Remark 2.7.** Sometimes we need to work with  $n$ -dimensional IF-observable  $x : \mathcal{B}(R^n) \rightarrow \mathcal{F}$  defined as a mapping with the following conditions:

- (i)  $x(R^n) = (1_\Omega, 0_\Omega)$ ,  $x(\emptyset) = (0_\Omega, 1_\Omega)$ ;
- (ii) if  $A \cap B = \emptyset$ ,  $A, B \in \mathcal{B}(R^n)$ , then  $x(A) \odot x(B) = (0_\Omega, 1_\Omega)$  and  $x(A \cup B) = x(A) \oplus x(B)$ ;
- (iii) if  $A_n \nearrow A$ , then  $x(A_n) \nearrow x(A)$  for each  $A, A_n \in \mathcal{B}(R^n)$ .

If  $n = 1$  we simply say that  $x$  is an IF-observable.

Similarly to the classical case, the following theorem can be proved ([9]).

**Theorem 2.8.** Let  $x : \mathcal{B}(R) \rightarrow \mathcal{F}$  be an IF-observable,  $\mathbf{m} : \mathcal{F} \rightarrow [0, 1]$  be an IF-state. Define the mapping  $\mathbf{m}_x : \mathcal{B}(R) \rightarrow [0, 1]$  by the formula

$$\mathbf{m}_x(C) = \mathbf{m}(x(C)).$$

Then  $\mathbf{m}_x : \mathcal{B}(R) \rightarrow [0, 1]$  is a probability measure.

### 3 Product operation, joint IF-observable and function of several IF-observables

In [5] we introduced the notion of product operation on the family of IF-events  $\mathcal{F}$  and showed an example of this operation.

**Definition 3.1.** We say that a binary operation  $\cdot$  on  $\mathcal{F}$  is product if it satisfying the following conditions:

- (i)  $(1_\Omega, 0_\Omega) \cdot (a_1, a_2) = (a_1, a_2)$  for each  $(a_1, a_2) \in \mathcal{F}$ ;
- (ii) the operation  $\cdot$  is commutative and associative;

(iii) if  $(a_1, a_2) \odot (b_1, b_2) = (0_\Omega, 1_\Omega)$  and  $(a_1, a_2), (b_1, b_2) \in \mathcal{F}$ , then

$$(c_1, c_2) \cdot ((a_1, a_2) \oplus (b_1, b_2)) = ((c_1, c_2) \cdot (a_1, a_2)) \oplus ((c_1, c_2) \cdot (b_1, b_2))$$

and

$$((c_1, c_2) \cdot (a_1, a_2)) \odot ((c_1, c_2) \cdot (b_1, b_2)) = (0_\Omega, 1_\Omega)$$

for each  $(c_1, c_2) \in \mathcal{F}$ ;

(iv) if  $(a_{1n}, a_{2n}) \searrow (0_\Omega, 1_\Omega)$ ,  $(b_{1n}, b_{2n}) \searrow (0_\Omega, 1_\Omega)$  and  $(a_{1n}, a_{2n}), (b_{1n}, b_{2n}) \in \mathcal{F}$ , then  $(a_{1n}, a_{2n}) \cdot (b_{1n}, b_{2n}) \searrow (0_\Omega, 1_\Omega)$ .

The following theorem provides an example of product operation for IF-events.

**Theorem 3.2** ([5, Theorem 1]). *The operation  $\cdot$  defined by*

$$(x_1, y_1) \cdot (x_2, y_2) = (x_1 \cdot x_2, y_1 + y_2 - y_1 \cdot y_2)$$

*for each  $(x_1, y_1), (x_2, y_2) \in \mathcal{F}$  is product operation on  $\mathcal{F}$ .*

In [8] B. Riečan defined the notion of a joint IF-observable and proved its existence.

**Definition 3.3.** *Let  $x, y : \mathcal{B}(R) \rightarrow \mathcal{F}$  be two IF-observables. The joint IF-observable of the IF-observables  $x, y$  is a mapping  $h : \mathcal{B}(R^2) \rightarrow \mathcal{F}$  satisfying the following conditions:*

(i)  $h(R^2) = (1_\Omega, 0_\Omega)$ ,  $h(\emptyset) = (0_\Omega, 1_\Omega)$ ;

(ii) if  $A, B \in \mathcal{B}(R^2)$  and  $A \cap B = \emptyset$ , then

$$h(A \cup B) = h(A) \oplus h(B) \text{ and } h(A) \odot h(B) = (0_\Omega, 1_\Omega);$$

(iii) if  $A, A_1, \dots \in \mathcal{B}(R^2)$  and  $A_n \nearrow A$ , then  $h(A_n) \nearrow h(A)$ ;

(iv)  $h(C \times D) = x(C) \cdot y(D)$  for each  $C, D \in \mathcal{B}(R)$ .

**Theorem 3.4** ([8, Theorem 3.3]). *For each two IF-observables  $x, y : \mathcal{B}(R) \rightarrow \mathcal{F}$  there exists their joint IF-observable.*

**Remark 3.5.** *The joint IF-observable of IF-observables  $x, y$  from Definition 3.3 is a two-dimensional IF-observable.*

If we have several IF-observables and a Borel measurable function, we can define the IF-observable, which is the function of several IF-observables. Regarding this we provide the following definition.

**Definition 3.6.** *Let  $x_1, \dots, x_n : \mathcal{B}(R) \rightarrow \mathcal{F}$  be IF-observables,  $h_n$  their joint IF-observable and  $g_n : R^n \rightarrow R$  a Borel measurable function. Then we define the IF-observable  $g_n(x_1, \dots, x_n) : \mathcal{B}(R) \rightarrow \mathcal{F}$  by the formula*

$$g_n(x_1, \dots, x_n)(A) = h_n(g_n^{-1}(A)).$$

*for each  $A \in \mathcal{B}(R)$ .*

## 4 Lower and upper limits, m-almost everywhere convergence

In [4] we defined the notions of lower and upper limits for a sequence of IF-observables and showed the connection between two kinds of m-almost everywhere convergence.

**Definition 4.1.** We shall say that a sequence  $(x_n)_n$  of IF-observables has  $\limsup_{n \rightarrow \infty}$ , if there exists an IF-observable  $\bar{x} : \mathcal{B}(R) \rightarrow \mathcal{F}$  such that

$$\bar{x}((-\infty, t)) = \bigvee_{p=1}^{\infty} \bigvee_{k=1}^{\infty} \bigwedge_{n=k}^{\infty} x_n \left( \left( -\infty, t - \frac{1}{p} \right) \right)$$

for every  $t \in R$ . We write  $\bar{x} = \limsup_{n \rightarrow \infty} x_n$ .

Note that if another IF-observable  $y$  satisfies the above condition, then  $\mathbf{m} \circ y = \mathbf{m} \circ \bar{x}$ .

**Definition 4.2.** A sequence  $(x_n)_n$  of IF-observables has  $\liminf_{n \rightarrow \infty}$ , if there exists an IF-observable  $\underline{x}$  such that

$$\underline{x}((-\infty, t)) = \bigvee_{p=1}^{\infty} \bigwedge_{k=1}^{\infty} \bigvee_{n=k}^{\infty} x_n \left( \left( -\infty, t - \frac{1}{p} \right) \right)$$

for all  $t \in R$ . Notation:  $\underline{x} = \liminf_{n \rightarrow \infty} x_n$ .

**Theorem 4.3** ([4, Theorem 3.3]). The IF-observables  $\bar{x}, \underline{x}$  from Definition 4.1 and Definition 4.2 can be expressed in the following form

$$\bar{x}(A) = \left( \overline{x^b}(A), 1_{\Omega} - \overline{x^{\sharp}}(A) \right),$$

$$\underline{x}(A) = \left( \underline{x^b}(A), 1_{\Omega} - \underline{x^{\sharp}}(A) \right),$$

for each  $A \in \mathcal{B}(R)$ . Here  $\overline{x^b}, \underline{x^b}$  are upper and lower limits of sequence  $(x_n^b)_1^{\infty}$  of observables in tribe  $\mathcal{T}$  and  $\overline{x^{\sharp}}, \underline{x^{\sharp}}$  are upper and lower limits of sequence  $(x_n^{\sharp})_1^{\infty}$  of observables in tribe  $\mathcal{T}$  (see [6]).

**Proposition 4.1** ([4, Proposition 3.1]). If a sequence of IF-observables  $(x_n)_n$  has  $\bar{x} = \limsup_{n \rightarrow \infty} x_n$  and  $\underline{x} = \liminf_{n \rightarrow \infty} x_n$ , then

$$\bar{x}((-\infty, t)) \leq \underline{x}((-\infty, t)),$$

for every  $t \in R$ .

**Proposition 4.2** ([4, Proposition 4.1]). A sequence  $(x_n)_n$  of IF-observables converges m-almost everywhere to 0 if and only if

$$\begin{aligned} \mathbf{m} \left( \bigvee_{p=1}^{\infty} \bigvee_{k=1}^{\infty} \bigwedge_{n=k}^{\infty} x_n \left( \left( -\infty, t - \frac{1}{p} \right) \right) \right) &= \mathbf{m} \left( \bigvee_{p=1}^{\infty} \bigwedge_{k=1}^{\infty} \bigvee_{n=k}^{\infty} x_n \left( \left( -\infty, t - \frac{1}{p} \right) \right) \right) = \\ &= \mathbf{m}(0_{\mathcal{F}}((-\infty, t))), \end{aligned}$$

for every  $t \in R$ .

In accordance to Proposition 4.2 we can extend the notion of  $\mathbf{m}$ -almost everywhere convergence in the following way.

**Definition 4.4.** A sequence  $(x_n)_n$  of an IF-observables converges  $\mathbf{m}$ -almost everywhere to an IF-observable  $x$ , if

$$\begin{aligned} \mathbf{m}\left(\bigvee_{p=1}^{\infty} \bigvee_{k=1}^{\infty} \bigwedge_{n=k}^{\infty} x_n\left(\left(-\infty, t - \frac{1}{p}\right)\right)\right) &= \mathbf{m}\left(\bigvee_{p=1}^{\infty} \bigwedge_{k=1}^{\infty} \bigvee_{n=k}^{\infty} x_n\left(\left(-\infty, t - \frac{1}{p}\right)\right)\right) = \\ &= \mathbf{m}(x((-\infty, t))), \end{aligned}$$

for every  $t \in R$ .

## 5 $P$ -almost everywhere convergence and $\mathbf{m}$ -almost everywhere convergence

The main result of this section is given in Theorem 5.1. The main step is presented in the following proposition.

Recall, that the corresponding probability space is  $(R^N, \sigma(\mathcal{C}), P)$ , where  $\mathcal{C}$  is the family of all sets of the form

$$\{(t_i)_{i=1}^{\infty} : t_1 \in A_1, \dots, t_n \in A_n\},$$

and  $P$  is the probability measure determined by the equality

$$P(\{(t_i)_{i=1}^{\infty} : t_1 \in A_1, \dots, t_n \in A_n\}) = \mathbf{m}(x_1(A_1) \cdot \dots \cdot x_n(A_n)).$$

The corresponding projections  $\xi_n : R^N \rightarrow R$  are defined by the equality

$$\xi_n((t_i)_{i=1}^{\infty}) = t_n.$$

**Proposition 5.1.** Let  $(x_n)_n$  be a sequence of IF-observables,  $(\xi_n)_n$  the sequence of corresponding projections,  $g_n : R^n \rightarrow R$  be a Borel measurable functions ( $n = 1, 2, \dots$ ). Then

$$\begin{aligned} P\left(\{u \in R^N : \limsup_{n \rightarrow \infty} g_n(\xi_1(u), \dots, \xi_n(u)) < t\}\right) &\leq \\ &\leq \mathbf{m}\left(\bigvee_{p=1}^{\infty} \bigvee_{k=1}^{\infty} \bigwedge_{n=k}^{\infty} g_n(x_1, \dots, x_n)\left(\left(-\infty, t - \frac{1}{p}\right)\right)\right), \\ P\left(\{u \in R^N : \liminf_{n \rightarrow \infty} g_n(\xi_1(u), \dots, \xi_n(u)) < t\}\right) &\geq \\ &\geq \mathbf{m}\left(\bigvee_{p=1}^{\infty} \bigwedge_{k=1}^{\infty} \bigvee_{n=k}^{\infty} g_n(x_1, \dots, x_n)\left(\left(-\infty, t - \frac{1}{p}\right)\right)\right). \end{aligned}$$

*Proof.* We have

$$\begin{aligned}
P\left(\{u \in R^N : \limsup_{n \rightarrow \infty} g_n(\xi_1(u), \dots, \xi_n(u)) < t\}\right) &= \\
&= P\left(\bigcup_{p=1}^{\infty} \bigcup_{k=1}^{\infty} \bigcap_{n=k}^{\infty} \left\{u \in R^N : g_n(u_1, \dots, u_n) < t - \frac{1}{p}\right\}\right) = \\
&= \lim_{p \rightarrow \infty} \lim_{k \rightarrow \infty} \lim_{i \rightarrow \infty} P\left(\bigcap_{n=k}^{k+i} \left(\pi_n^{-1}\left(g_n^{-1}\left(\left(-\infty, t - \frac{1}{p}\right)\right)\right)\right)\right) = \\
&= \lim_{p \rightarrow \infty} \lim_{k \rightarrow \infty} \lim_{i \rightarrow \infty} P\left(\pi_{k+i}^{-1}\left(\bigcap_{n=k}^{k+i} g_n^{-1}\left(\left(-\infty, t - \frac{1}{p}\right)\right)\right)\right) = \\
&= \lim_{p \rightarrow \infty} \lim_{k \rightarrow \infty} \lim_{i \rightarrow \infty} \mathbf{m}\left(h_{k+i}\left(\bigcap_{n=k}^{k+i} g_n^{-1}\left(\left(-\infty, t - \frac{1}{p}\right)\right)\right)\right) \leq \\
&\leq \lim_{p \rightarrow \infty} \lim_{k \rightarrow \infty} \lim_{i \rightarrow \infty} \mathbf{m}\left(\bigwedge_{n=k}^{k+i} (h_{k+i} \circ g_n^{-1})\left(\left(-\infty, t - \frac{1}{p}\right)\right)\right) = \\
&= \lim_{p \rightarrow \infty} \lim_{k \rightarrow \infty} \lim_{i \rightarrow \infty} \mathbf{m}\left(\bigwedge_{n=k}^{k+i} g_n(x_1, \dots, x_n)\left(\left(-\infty, t - \frac{1}{p}\right)\right)\right) = \\
&= \mathbf{m}\left(\bigvee_{p=1}^{\infty} \bigvee_{k=1}^{\infty} \bigwedge_{n=k}^{\infty} g_n(x_1, \dots, x_n)\left(\left(-\infty, t - \frac{1}{p}\right)\right)\right).
\end{aligned}$$

The second inequality can be proved similarly.  $\square$

**Theorem 5.1.** Let  $(x_n)_n$  be a sequence of IF-observables,  $(\xi_n)_n$  be the sequence of corresponding projections,  $(g_n)_n$  be a sequence of Borel measurable functions  $g_n : R^n \rightarrow R$ . If the sequence  $(g_n(\xi_1, \dots, \xi_n))_n$  converges  $P$ -almost everywhere, then the sequence  $(g_n(x_1, \dots, x_n))_n$  converges  $\mathbf{m}$ -almost everywhere and

$$\mathbf{m}\left(\limsup_{n \rightarrow \infty} g_n(x_1, \dots, x_n)((-\infty, t))\right) = \mathbf{m}\left(\liminf_{n \rightarrow \infty} g_n(x_1, \dots, x_n)((-\infty, t))\right)$$

for each  $t \in R$ . Moreover

$$P\left(\{u \in R^N : \limsup_{n \rightarrow \infty} g_n(\xi_1(u), \dots, \xi_n(u)) < t\}\right) = \mathbf{m}\left(\limsup_{n \rightarrow \infty} g_n(x_1, \dots, x_n)((-\infty, t))\right)$$

for each  $t \in R$ .

*Proof.* Let the sequence  $(g_n(\xi_1, \dots, \xi_n))_n$  converges  $P$ -almost everywhere, then

$$\begin{aligned}
P\left(\{u \in R^N : \limsup_{n \rightarrow \infty} g_n(\xi_1(u), \dots, \xi_n(u)) < t\}\right) &= \\
&= P\left(\{u \in R^N : \liminf_{n \rightarrow \infty} g_n(\xi_1(u), \dots, \xi_n(u)) < t\}\right).
\end{aligned} \tag{1}$$

Put

$$\begin{aligned}
\varphi(t) &= \bigvee_{p=1}^{\infty} \bigvee_{k=1}^{\infty} \bigwedge_{n=k}^{\infty} g_n(x_1, \dots, x_n)\left(\left(-\infty, t - \frac{1}{p}\right)\right), \\
\psi(t) &= \bigvee_{p=1}^{\infty} \bigwedge_{k=1}^{\infty} \bigvee_{n=k}^{\infty} g_n(x_1, \dots, x_n)\left(\left(-\infty, t - \frac{1}{p}\right)\right),
\end{aligned}$$



and

$$\eta_n(u) = g_n(\xi_1(u), \dots, \xi_n(u)).$$

Since  $\varphi(t) \leq \psi(t)$ , then

$$\mathbf{m}(\varphi(t)) \leq \mathbf{m}(\psi(t)).$$

By Proposition 5.1 and (1) we obtain

$$\begin{aligned} \mathbf{m}(\psi(t)) &\leq P\left(\{u \in R^N : \liminf_{n \rightarrow \infty} \eta_n(u) < t\}\right) = P\left(\{u \in R^N : \limsup_{n \rightarrow \infty} \eta_n(u) < t\}\right) \\ &\leq \mathbf{m}(\varphi(t)). \end{aligned}$$

Hence

$$\mathbf{m}(\varphi(t)) = \mathbf{m}(\psi(t)),$$

and moreover

$$P\left(\{u \in R^N : \limsup_{n \rightarrow \infty} \eta_n(u) < t\}\right) = \mathbf{m}(\varphi(t)) = \mathbf{m}(\psi(t)).$$

Denote the common value by

$$F(t) = P\left(\{u \in R^N : \limsup_{n \rightarrow \infty} \eta_n(u) < t\}\right) = \mathbf{m}(\varphi(t)) = \mathbf{m}(\psi(t)).$$

Since  $\limsup_{n \rightarrow \infty} \eta_n$  is a random variable, then  $F : R \rightarrow [0, 1]$  is a distribution function. Evidently

$$\begin{aligned} \mathbf{m}\left(\bigvee_{n=1}^{\infty} \varphi(n)\right) &= \lim_{n \rightarrow \infty} \mathbf{m}(\varphi(n)) = \lim_{n \rightarrow \infty} F(n) = 1, \\ \mathbf{m}\left(\bigwedge_{n=1}^{\infty} \varphi(-n)\right) &= \lim_{n \rightarrow \infty} \mathbf{m}(\varphi(-n)) = \lim_{n \rightarrow \infty} F(-n) = 0. \end{aligned}$$

Since  $\mathbf{m}$  is faithful, we obtain

$$\bigvee_{n=1}^{\infty} \varphi(n) = (1_{\Omega}, 0_{\Omega}), \quad \bigwedge_{n=1}^{\infty} \varphi(-n) = (0_{\Omega}, 1_{\Omega}). \quad (2)$$

Let  $t_n \nearrow t$ . Evidently  $\varphi(t_n) \leq \varphi(t)$ , hence

$$\bigvee_{n=1}^{\infty} \varphi(n) \leq \varphi(t).$$

On the other hand to each  $p \in N$  there exist  $j, q \in N$  such that  $(-\infty, t - \frac{1}{p}) \subset (-\infty, t_j - \frac{1}{q})$ , hence

$$g_n(x_1, \dots, x_n)\left(\left(-\infty, t - \frac{1}{p}\right)\right) \leq g_n(x_1, \dots, x_n)\left(\left(-\infty, t_j - \frac{1}{q}\right)\right)$$

and therefore

$$\bigwedge_{n=k}^{\infty} g_n(x_1, \dots, x_n)\left(\left(-\infty, t - \frac{1}{p}\right)\right) \leq \bigwedge_{n=k}^{\infty} g_n(x_1, \dots, x_n)\left(\left(-\infty, t_j - \frac{1}{q}\right)\right) \leq \varphi(t_j) \leq \bigvee_{j=1}^{\infty} \varphi(t_j).$$

Since the relation holds for every  $k$  and  $p$ , we obtain

$$\varphi(t) = \bigvee_{p=1}^{\infty} \bigvee_{k=1}^{\infty} \bigwedge_{n=k}^{\infty} g_n(x_1, \dots, x_n) \left( \left( -\infty, t - \frac{1}{p} \right) \right) \leq \bigvee_{j=1}^{\infty} \varphi(t_j).$$

Hence

$$\varphi(t) = \bigvee_{j=1}^{\infty} \varphi(t_j). \quad (3)$$

Let us summarize:  $\varphi$  is non-decreasing,  $\bigvee_{n=1}^{\infty} \varphi(n) = (1_{\Omega}, 0_{\Omega})$ ,  $\bigwedge_{n=1}^{\infty} \varphi(-n) = (0_{\Omega}, 1_{\Omega})$ ,  $t_n \nearrow t$  implies  $\varphi(t_n) \nearrow \varphi(t)$ .

Put  $y_n = g_n(x_1, \dots, x_n)$ . Using max-min connectives  $\vee, \wedge$ , the De Morgan rules and equality  $y_n(A) = (y_n^b(A), 1_{\Omega} - y_n^{\sharp}(A))$  we obtain

$$\bigvee_{p=1}^{\infty} \bigvee_{k=1}^{\infty} \bigwedge_{n=k}^{\infty} y_n(A) = \left( \bigvee_{p=1}^{\infty} \bigvee_{k=1}^{\infty} \bigwedge_{n=k}^{\infty} y_n^b(A), 1_{\Omega} - \bigvee_{p=1}^{\infty} \bigvee_{k=1}^{\infty} \bigwedge_{n=k}^{\infty} y_n^{\sharp}(A) \right)$$

for each  $A \in \mathcal{B}(R)$ , where  $y_n^b, y_n^{\sharp} : \mathcal{B}(R) \rightarrow \mathcal{T}$  are observables. Put

$$\begin{aligned} \varphi^b(t) &= \bigvee_{p=1}^{\infty} \bigvee_{k=1}^{\infty} \bigwedge_{n=k}^{\infty} y_n^b \left( \left( -\infty, t - \frac{1}{p} \right) \right), \\ \varphi^{\sharp}(t) &= \bigvee_{p=1}^{\infty} \bigvee_{k=1}^{\infty} \bigwedge_{n=k}^{\infty} y_n^{\sharp} \left( \left( -\infty, t - \frac{1}{p} \right) \right). \end{aligned}$$

Then

$$\varphi(t) = (\varphi^b(t), 1_{\Omega} - \varphi^{\sharp}(t)).$$

Since  $\varphi$  is non-decreasing, therefore  $\varphi^b, \varphi^{\sharp}$  are non-decreasing. More by (2) we have

$$\begin{aligned} (1_{\Omega}, 0_{\Omega}) &= \bigvee_{n=1}^{\infty} \varphi(n) = \bigvee_{n=1}^{\infty} (\varphi^b(n), 1_{\Omega} - \varphi^{\sharp}(n)) = \left( \bigvee_{n=1}^{\infty} \varphi^b(n), 1_{\Omega} - \bigvee_{n=1}^{\infty} \varphi^{\sharp}(n) \right), \\ (0_{\Omega}, 1_{\Omega}) &= \bigwedge_{n=1}^{\infty} \varphi(-n) = \bigwedge_{n=1}^{\infty} (\varphi^b(-n), 1_{\Omega} - \varphi^{\sharp}(-n)) = \left( \bigwedge_{n=1}^{\infty} \varphi^b(-n), 1_{\Omega} - \bigwedge_{n=1}^{\infty} \varphi^{\sharp}(-n) \right). \end{aligned}$$

Hence

$$\bigvee_{n=1}^{\infty} \varphi^b(n) = 1_{\Omega}, \quad \bigwedge_{n=1}^{\infty} \varphi^b(-n) = 0_{\Omega}, \quad (4)$$

$$\bigvee_{n=1}^{\infty} \varphi^{\sharp}(n) = 1_{\Omega}, \quad \bigwedge_{n=1}^{\infty} \varphi^{\sharp}(-n) = 0_{\Omega}. \quad (5)$$

Since  $t_n \nearrow t$  implies

$$(\varphi^b(t_n), 1_{\Omega} - \varphi^{\sharp}(t_n)) = \varphi(t_n) \nearrow \varphi(t) = (\varphi^b(t), 1_{\Omega} - \varphi^{\sharp}(t)),$$

then

$$\varphi^b(t_n) \nearrow \varphi^b(t), \quad (6)$$

$$1_\Omega - \varphi^\sharp(t_n) \searrow 1_\Omega - \varphi^\sharp(t) \iff \varphi^\sharp(t_n) \nearrow \varphi^\sharp(t). \quad (7)$$

For fixed  $\omega \in \Omega$  and arbitrary  $t \in R$  put

$$F_\omega^b(t) = \varphi^b(t)(\omega), \quad F_\omega^\sharp(t) = \varphi^\sharp(t)(\omega).$$

Evidently  $F_\omega^b, F_\omega^\sharp : R \rightarrow [0, 1]$  are the non-decreasing functions and by (4), (6) and by (5), (7) we obtain that  $F_\omega^b, F_\omega^\sharp$  are the distribution functions. Denote by  $\lambda_\omega^b, \lambda_\omega^\sharp$  the corresponding Stieltjes probability measures and define  $\bar{y}^b, \bar{y}^\sharp : \mathcal{B}(R) \rightarrow \mathcal{T}$  by the equalities

$$\bar{y}^b(A)(\omega) = \lambda_\omega^b(A), \quad \bar{y}^\sharp(A)(\omega) = \lambda_\omega^\sharp(A).$$

Then  $\bar{y}^b, \bar{y}^\sharp$  are the observables and

$$\bar{y}^b((-\infty, t))(\omega) = \lambda_\omega^b((-\infty, t)) = F_\omega^b(t) = \varphi^b(t)(\omega), \quad (8)$$

$$\bar{y}^\sharp((-\infty, t))(\omega) = \lambda_\omega^\sharp((-\infty, t)) = F_\omega^\sharp(t) = \varphi^\sharp(t)(\omega), \quad (9)$$

for each  $\omega \in \Omega$ . Using Theorem 4.3 and (8), (9) we have that there exists IF-observable  $\bar{y} = \limsup_{n \rightarrow \infty} y_n$  given by

$$\bar{y}((-\infty, t)) = \left( \bar{y}^b((-\infty, t)), 1_\Omega - \bar{y}^\sharp((-\infty, t)) \right) = (\varphi^b(t), 1_\Omega - \varphi^\sharp(t)) = \varphi(t).$$

The existence of IF-observable  $\underline{y} = \liminf_{n \rightarrow \infty} y_n = \psi$  can be proved similarly. Since  $\mathbf{m}(\varphi(t)) = \mathbf{m}(\psi(t))$ , then

$$\mathbf{m}(\bar{y}((-\infty, t))) = \mathbf{m}(\underline{y}((-\infty, t)))$$

for each  $t \in R$  and by Definition 4.4 we have that  $(y_n)_n = (g_n(x_1, \dots, x_n))_n$  converges  $\mathbf{m}$ -almost everywhere. Moreover

$$P\left(\left\{u \in R^N : \limsup_{n \rightarrow \infty} \eta_n(u) < t\right\}\right) = \mathbf{m}(\bar{y}((-\infty, t)))$$

for each  $t \in R$ . □

## 6 Conclusion

The Theorem 5.1 is important for the proof of the Individual ergodic theorem in intuitionistic fuzzy case, where we work with the sequence of several IF-observables induced by the Borel function.

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