STRONG LAW OF LARGE NUMBERS FOR LEVEL-WISE INDEPENDENT FUZZY RANDOM VARIABLES

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ABSTRACT. In this paper, we obtain a strong law of large numbers for sums of level-wise independent and level-wise identically distributed fuzzy random variables.

1. Introduction

Laws of large numbers for sums of independent random sets have been studied by Artstein and Hart [1], Artstein and Vitale [2], Puri and Ralescu [16], Taylor and Inoue [18], Uemura [19], etc. These results have been generalized to the case of fuzzy random variables by several people. A SLLN for sums of independent and identically distributed fuzzy random variables was obtained by Kruse [14], and a SLLN for sums of independent fuzzy random variables by Miyakoshi and Shimbo [15]. Also, Klement, Puri and Ralescu [12] proved some limit theorems which includes a SLLN, and Inoue [10] obtained a SLLN for sums of independent tight fuzzy random sets. Recently, Hong and Kim [9] generalized Marcinkiewicz law of large numbers to fuzzy random variables.

In this paper, we obtain a SLLN for sums of level-wise independent and level-wise identically distributed fuzzy random variables by using a metric which is stronger than one in works mentioned previously. The representation theorem of fuzzy numbers by Goetschel and Voxman [7] will be used.

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2. Preliminaries

In this section, we describe some basic concepts of fuzzy numbers. Let R denote the real line. A fuzzy number is a fuzzy set $\tilde{u}: R \to [0, 1]$ with the following properties;

- (1) \tilde{u} is normal, i.e., there exists $x \in R$ such that $\tilde{u}(x) = 1$.
- (2) \tilde{u} is upper semicontinuous.
- (3) supp $\tilde{u} = cl\{x \in R : \tilde{u}(x) > 0\}$ is compact.
- (4) \tilde{u} is a convex fuzzy set, i.e., $\tilde{u}(\lambda x + (1 \lambda)y) \ge \min(\tilde{u}(x), \tilde{u}(y))$ for $x, y \in R$ and $\lambda \in [0, 1]$.

For a fuzzy set \tilde{u} , we define

$$L_{lpha} ilde{u} = \left\{egin{array}{ll} \{x: ilde{u}(x) \geq lpha\}, & 0 < lpha \leq 1 \ supp ilde{u}, & lpha = 0 \end{array}
ight.$$

Then, it follows that \tilde{u} is a fuzzy number if and only if $L_1\tilde{u} \neq \phi$ and $L_{\alpha}\tilde{u}$ is a closed bounded interval for each $\alpha \in [0,1]$. From this characterization of fuzzy numbers, a fuzzy number \tilde{u} is completely determined by the end points of the intervals $L_{\alpha}\tilde{u} = [u_{\alpha}^-, u_{\alpha}^+]$. We denote the family of all fuzzy numbers by F(R).

THEOREM 2.1 ([7]). For $\tilde{u} \in F(R)$, we denote $u^{-}(\alpha) = u_{\alpha}^{-}$ and $u^{+}(\alpha) = u_{\alpha}^{+}$. Then the followings hold;

- (1) $u^{-}(\alpha)$ is a bounded increasing function on [0,1].
- (2) $u^+(\alpha)$ is a bounded decreasing function on [0,1].
- (3) $u^-(1) \le u^+(1)$.
- (4) $u^{-}(\alpha)$ and $u^{+}(\alpha)$ are left continuous on (0,1) and right continuous at 0.
- (5) If $v^-(\alpha)$ and $v^+(\alpha)$ satisfy above (1)-(4), then there exists unique $\tilde{v} \in F(R)$ such that $v_{\alpha}^- = v^-(\alpha)$, $v_{\alpha}^+ = v^+(\alpha)$.

The above theorem implies that we can identify a fuzzy number \tilde{u} with the parametrized representation $\{(u_{\alpha}^-, u_{\alpha}^+) | 0 \le \alpha \le 1\}$. Suppose now that \tilde{u}, \tilde{v} are fuzzy numbers represented by $\{(u_{\alpha}^-, u_{\alpha}^+) | 0 \le \alpha \le 1\}$ and $\{(v_{\alpha}^-, v_{\alpha}^+) | 0 \le \alpha \le 1\}$, respectively. If we define

$$(\tilde{u} + \tilde{v})(z) = \sup_{x+y=z} \min(\tilde{u}(x), \tilde{v}(y)),$$

$$(\lambda \tilde{u})(z) = \begin{cases} \tilde{u}(z/\lambda), & \lambda \neq 0 \\ \tilde{0}, & \lambda = 0 \end{cases}$$

where $\tilde{0} = I_{\{0\}}$ is the indicator function of $\{0\}$, then

$$\begin{split} \tilde{u} + \tilde{v} &= \{(u_{\alpha}^- + v_{\alpha}^-, v_{\alpha}^+ + v_{\alpha}^+) | \ 0 \leq \alpha \leq 1\}, \\ \lambda \tilde{u} &= \left\{ \begin{array}{ll} \{(\lambda u_{\alpha}^-, \lambda u_{\alpha}^+) | \ 0 \leq \alpha \leq 1\}, & \lambda \geq 0 \\ \{(\lambda u_{\alpha}^+, \lambda u_{\alpha}^-) | \ 0 \leq \alpha \leq 1\}, & \lambda < 0. \end{array} \right. \end{split}$$

Now, we define two metrics d, d^* on F(R) by

(2.1)
$$d(\tilde{u}, \tilde{v}) = \sup_{0 \le \alpha \le 1} d_H(L_\alpha \tilde{u}, L_\alpha \tilde{v})$$

(2.2)
$$d^*(\tilde{u}, \tilde{v}) = \int_0^1 d_H(L_\alpha \tilde{u}, L_\alpha \tilde{v}) d\alpha$$

where d_H is the Hausdorff metric defined as

$$d_H(L_{\alpha}\tilde{u},L_{\alpha}\tilde{v})=\max(|u_{\alpha}^--v_{\alpha}^-|,|u_{\alpha}^+-v_{\alpha}^+|).$$

Also, the norm $\|\tilde{u}\|$ of fuzzy number \tilde{u} will be defined as

$$\|\tilde{u}\| = d(\tilde{u}, \tilde{0}) = \max(|u_0^-|, |u_0^+|).$$

3. Fuzzy random variables

Throughout this paper, (Ω, \mathcal{A}, P) denotes a complete probability space. If $\tilde{X}: \Omega \to F(R)$ is a fuzzy number valued function and B is a subset of R, then $\tilde{X}^{-1}(B)$ denotes the fuzzy subset of Ω defined by

$$\tilde{X}^{-1}(B)(\omega) = \sup_{x \in B} \tilde{X}(\omega)(x)$$

for every $\omega \in \Omega$. The function $\tilde{X}: \Omega \to F(R)$ is called a fuzzy random variable if for every closed subset B of R, the fuzzy set $\tilde{X}^{-1}(B)$ is measurable when considered as a function from Ω to [0,1]. If we denote $\tilde{X}(\omega) = \{(X_{\alpha}^{-}(\omega), X_{\alpha}^{+}(\omega)|\ 0 \leq \alpha \leq 1\}$, then it is well-known that \tilde{X} is a fuzzy random variable if and only if for each $\alpha \in [0,1]$, X_{α}^{-} and X_{α}^{+} are random variables in the usual sense (See Kim and Ghil [11]). Hence, if $\sigma(\tilde{X})$ is the smallest σ -field which makes \tilde{X} a fuzzy random variable, then $\sigma(\tilde{X})$ is consistent with $\sigma(\{X_{\alpha}^{-}, X_{\alpha}^{+}|\ 0 \leq \alpha \leq 1\})$. This enables us to define the concept of independence for fuzzy random variables as in the case of classical random variables.

Definition 3.1. Let \tilde{X}, \tilde{Y} be two fuzzy random variables whose representations are $\{(X_{\alpha}^-, X_{\alpha}^+)|\ 0 \leq \alpha \leq 1\}$ and $\{(Y_{\alpha}^-, Y_{\alpha}^+)|\ 0 \leq \alpha \leq 1\}$, respectively.

- (1) \tilde{X} and \tilde{Y} are called independent if the σ -fields $\sigma(\tilde{X})$ and $\sigma(\tilde{Y})$ are independent.
- (2) \tilde{X} and \tilde{Y} are called level-wise independent if for each $\alpha \in [0,1]$ the σ -fields $\sigma(X_{\alpha}^{-}, X_{\alpha}^{+})$ and $\sigma(Y_{\alpha}^{-}, Y_{\alpha}^{+})$ are independent.
- (3) \tilde{X} and \tilde{Y} are called level-wise identically distributed if for each $\alpha \in [0,1], (X_{\alpha}^{-}, X_{\alpha}^{+})$ and $(Y_{\alpha}^{-}, Y_{\alpha}^{+})$ are identically distributed random vectors.

Note that the definitions (2) and (3) is firstly introduced in this paper.

DEFINITION 3.2. A fuzzy random variable $\tilde{X}=\{(X_{\alpha}^-,X_{\alpha}^+)|\ 0\leq\alpha\leq 1\}$ is called integrable if for each $\alpha\in[0,1],\ X_{\alpha}^-$ and X_{α}^+ are integrable, equivalently, $\int \|\tilde{X}\|dP<\infty$. In this case, the expectation of \tilde{X} is defined by

$$E ilde{X} = \int ilde{X}dP = \{(\int X_{lpha}^- dP, \int X_{lpha}^+ dP) | \, 0 \leq lpha \leq 1\}$$

4. Main Result

In this section, a SLLN with respect to the metric d defined as in (2.1) will be obtained. In earlier works, the metric d^* defined as in (2.2) have been used (see [9],[10],[12]). Note that d is stronger than d^* . First, we need a subspace $F_C(R)$ of F(R). Let $F_C(R) = \{\tilde{u} \in F(R) | u_{\alpha}^- \text{ and } u_{\alpha}^+ \text{ are continuous when considered as functions of } \alpha\}$. Then it is known that $\tilde{u} \in F_C(R)$ if and only if for any $\beta \in (0,1)$, there exist at most two different x_1, x_2 such that $\tilde{u}(x_1) = \tilde{u}(x_2) = \beta$ (See [4] Theorem 5.1). Note that if \tilde{X} is $F_C(R)$ -valued, then $E\tilde{X} \in F_C(R)$. Before we state the main result, we recall the following lemma which is well-known in the classical Analysis.

LEMMA 4.1. Let (f_n) be a sequence of monotonic functions on [0,1]. If $f_n(x)$ converges pointwise to a continuous function f(x) on [0,1], then $f_n(x)$ converges to f(x) uniformly.

We now state the SLLN for sums of level-wise independent fuzzy random variables.

THEOREM 4.2. Let $\{\tilde{X}_n\}$ be a sequence of level-wise independent and level-wise identically distributed fuzzy random varibles with $E\|\tilde{X}_1\| < \infty$. If $E\tilde{X}_1 \in F_C(R)$, then

$$d(\frac{1}{n}\sum_{i=1}^{n}\tilde{X}_{i},E\tilde{X}_{1})\longrightarrow 0$$
 a.s.

PROOF. Let $\tilde{X}_n = \{(X_{n\alpha}^-, X_{n\alpha}^+) | 0 \le \alpha \le 1\}$. Then for each $\alpha \in [0, 1]$, $\{(X_{n\alpha}^-, X_{n\alpha}^+)\}$ is a sequence of independent and identically distributed random vectors with $E|X_{n\alpha}^-| < \infty$ and $E|X_{n\alpha}^+| < \infty$ in the classical sense. By Kolmogorov's strong law of large numbers,

$$\frac{1}{n}\sum_{i=1}^{n}X_{i\alpha}^{-}\longrightarrow EX_{1\alpha}^{-} \ a.s.$$

and

$$\frac{1}{n}\sum_{i=1}^{n}X_{i\alpha}^{+}\longrightarrow EX_{1\alpha}^{+} \ a.s.$$

Now, let $\{r_k\}$ be a countable dense subset of [0,1] with $r_0=0, r_1=1$. Then there exist $B_k \in \mathcal{A}$ with $P(B_k)=0$ such that for each $\omega \notin B_k$

(4.1)
$$\frac{1}{n} \sum_{i=1}^{n} X_{ir_k}^{-}(\omega) \longrightarrow EX_{1r_k}^{-}$$

(4.2)
$$\frac{1}{n} \sum_{i=1}^{n} X_{ir_k}^+(\omega) \longrightarrow EX_{1r_k}^+$$

If we define $B = \bigcup_{k=0}^{\infty} B_k$, then P(B) = 0 and for each $\omega \notin B$, (4.1) and (4.2) hold for all r_k . Now, we will show that for each $\omega \notin B$

$$\frac{1}{n}\sum_{i=1}^{n}X_{i\alpha}^{-}(\omega)\longrightarrow EX_{1\alpha}^{-} \text{ uniformly in } \alpha\in[0,1].$$

By Lemma 4.1, it suffices to show that for each $\omega \notin B$, and each α ,

$$\frac{1}{n}\sum_{i=1}^{n}X_{i\alpha}^{-}(\omega)\longrightarrow EX_{1\alpha}^{-}.$$

Let $\omega \notin B$ and $\epsilon > 0$ be fixed. Then by the continuity of $EX_{1\alpha}^-$ as a function of α , there exists $\delta > 0$ such that

$$|\alpha - \beta| < \delta$$
 implies $|EX_{1\alpha}^- - EX_{1\beta}^-| < \epsilon$

If we take r_l, r_m so that $\alpha - \delta < r_l < \alpha < r_m < \alpha + \delta$, then

$$EX_{1r_m}^- - \epsilon < EX_{1\alpha}^- < EX_{1r_l}^- + \epsilon.$$

Hence, by the monotonicity of $X_{i\alpha}^{-}(\omega)$ with respect to α ,

$$\frac{1}{n} \sum_{i=1}^{n} X_{ir_{l}}^{-}(\omega) - EX_{1r_{l}}^{-} - \epsilon < \frac{1}{n} \sum_{i=1}^{n} X_{i\alpha}^{-}(\omega) - EX_{1\alpha}^{-}
< \frac{1}{n} \sum_{i=1}^{n} X_{ir_{m}}^{-}(\omega) - EX_{1r_{m}}^{-} + \epsilon$$

which implies

$$\frac{1}{n}\sum_{i=1}^{n}X_{i\alpha}^{-}(\omega)\longrightarrow EX_{1\alpha}^{-}.$$

Similarly, it can be proved that for each $\omega \notin B$

$$\frac{1}{n} \sum_{i=1}^{n} X_{i\alpha}^{+}(\omega) \longrightarrow EX_{1\alpha}^{+} \text{ uniformly in } \alpha \in [0,1].$$

Therefore, for each $\omega \notin B$.

$$d\left(\frac{1}{n}\sum_{i=1}^{n}\tilde{X}_{i}(\omega),E\tilde{X}_{1}\right)\longrightarrow0.$$

 \Box

COROLLARY 4.3. Let $\{\tilde{X}_n\}$ be a sequence of level-wise independent and level-wise identically distributed $F_C(R)$ -valued fuzzy random variables. There exists $\tilde{b} \in F_C(R)$ such that

$$(4.3) d\left(\frac{1}{n}\sum_{i=1}^{n}\tilde{X}_{i}, \tilde{b}\right) \longrightarrow 0 \quad a.s.$$

if and only if $E\|\tilde{X}_1\| < \infty$. Furthermore, if (4.3) holds, then $\tilde{b} = E\tilde{X}_1$.

PROOF. The sufficiency follows immediately from theorem 4.2. To prove the converse, if (4.3) holds, then for any $\alpha \in [0, 1]$,

$$\frac{1}{n} \sum_{i=1}^{n} X_{i\alpha}^{-} \longrightarrow b_{\alpha}^{-} \ a.s.$$

and

$$\frac{1}{n} \sum_{i=1}^{n} X_{i\alpha}^{+} \longrightarrow b_{\alpha}^{+} \ a.s.$$

By the converse of Kolmogorov's strong law of large numbers,

$$E|X_{1\alpha}^-| < \infty, E|X_{1\alpha}^+| < \infty$$
 for each $\alpha \in [0, 1]$

which implies $E\|\tilde{X}_1\| < \infty$ and $\tilde{b} = E\tilde{X}_1$.

EXAMPLE. Let $\tilde{u} \in F_c(R)$ be fixed and let $\{Y_n\}$ be i.i.d. with $E|Y_1| < \infty$ in the usual sense. Define $\tilde{X}_n(w)(x) = \tilde{u}(x - Y_n(w))$ i.e., $\tilde{X}_n(w)$ is the translation of \tilde{u} by $Y_n(w)$ in x-axis. Then

$$X_{n,\alpha}^-(w)=u_{\alpha}^-+Y_n(w)$$
 and $X_{n,\alpha}^+(w)=u_{\alpha}^++Y_n(w)$

Hence the above theorem implies that

$$d\left(\frac{1}{n}\sum_{i=1}^{n}\tilde{X}_{i},\,E\tilde{X}_{1}\right)\longrightarrow0\quad a.s.$$

where $(E\tilde{X}_1)(x) = \tilde{u}(x - EY_1)$.

As a final result, we give a generalization of Chung's SLLN to the case of fuzzy random variables.

THEOREM 4.4. Let $\{\tilde{X}_n\}$ be a sequence of fuzzy random variables. If $\{\|\tilde{X}_n\|\}$ are independent random variables in classical sense and

$$(4.4) \sum_{n=1}^{\infty} \frac{1}{n} E \|\tilde{X}_n\| < \infty,$$

then

$$d\left(\frac{1}{n}\sum_{i=1}^{n}\tilde{X}_{i},\frac{1}{n}\sum_{i=1}^{n}E\tilde{X}_{i}\right)\longrightarrow 0 \ a.s.$$

PROOF. First we note that

$$d(\frac{1}{n}\sum_{i=1}^{n}\tilde{X}_{i}, \frac{1}{n}\sum_{i=1}^{n}E\tilde{X}_{i}) \leq \frac{1}{n}\sum_{i=1}^{n}d(\tilde{X}_{i}, E\tilde{X}_{i})$$

$$\leq \frac{1}{n}\sum_{i=1}^{n}(\|\tilde{X}_{i}\| + E\|\tilde{X}_{i}\|).$$

Since $\{\|\tilde{X}_n\|\}$ is a sequence of independent random variables, (4.4) and Chung's law of large numbers yields

(4.5)
$$\lim_{n \to \infty} \frac{1}{n} \sum_{i=1}^{n} (\|\tilde{X}_i\| - E\|\tilde{X}_i\|) = 0 \quad a.s.$$

Now, applying the Kronecker lemma to (4.4), we obtain

$$\lim_{n \to \infty} \frac{1}{n} \sum_{i=1}^{n} E \|\tilde{X}_i\| = 0$$

which implies, together with (4.5),

$$\lim_{n \to \infty} \frac{1}{n} \sum_{i=1}^{n} (\|\tilde{X}_i\| = 0 \ a.s.$$

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This gives the desired result.

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