

On the Variance of a Fuzzy Random Variable **

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In this paper, we shall introduce the concept of the variance of a fuzzy random variable based on [3, 4], and give formulas to calculus of the expected value and the variance of a fuzzy random variable taking in $\mathcal{F}(R^1)$.

Keywords: Fuzzy random variable, The expectation and variance of a fuzzy random variable

L.P.Madeh [3] introduced the concept of a fuzzy random variable and studied the expectation of it by using random variables set theory [1, 2, 5]. Li [4] further discussed the properties of fuzzy random variables. We shall investigate the variance of a fuzzy random variable based on [3, 4] as follows.

Let (Ω, \mathcal{A}, P) be a probability space where P is a nonatomic probability measure, R^n n -dimensional Euclidean space, \mathcal{B} the set of all Borel subsets of R^n and $K_0(R^n)$ the set of all nonempty compact subsets of R^n .

Definition 1 [4] A fuzzy subset $u: R^1 \rightarrow [0, 1]$ is called a fuzzy number, if

- (1) $\forall \alpha \in (0, 1], u_\alpha \triangleq \{x \in R^1: u(x) = \alpha\}$ is a closed interval;
- (2) $u_1 \triangleq \{x \in R^1: u(x) = 1\} \neq \emptyset$.

A Fuzzy subset $u: R^n \rightarrow [0, 1]$ is called an extended fuzzy number,

** This term is supported by NNSFC.

if

(1) $\forall \alpha \in (0, 1]$, $u_\alpha \triangleq \{x \in \mathbb{R}^n: u(x) \geq \alpha\}$ is compact;

(2) $u_1 \triangleq \{x \in \mathbb{R}^n: u(x) = 1\} \neq \emptyset$.

$\mathcal{F}_0(\mathbb{R}^n)$ denotes the set of all extended fuzzy numbers, and $\mathcal{F}(\mathbb{R}^1)$ the set of all fuzzy numbers.

Definition 2 [3] A fuzzy random variable is a function $X: \Omega \rightarrow \mathcal{F}_0(\mathbb{R}^n)$ such that

$$\{(\omega, x): x \in X_\alpha(\omega)\} \in \mathcal{A} \times \mathcal{B}, \quad \text{for every } \alpha \in (0, 1]$$

where $X_\alpha(\omega) = \{x \in \mathbb{R}^n: X(\omega)(x) \geq \alpha\}$.

The equivalent definition of a fuzzy random variable was given in [4].

Property 1 [4] X is a fuzzy random variable if and only if for every $\alpha \in [0, 1]$, for every $A \in \mathcal{B}_\sigma(K_0)$,

$$\{\omega \in \Omega: X_\alpha(\omega) \cap A \neq \emptyset\} \in \mathcal{A}$$

where $\mathcal{B}(K_0)$ denotes the set of all Borel σ -algebras of subsets of $K_0(\mathbb{R}^n)$.

Property 2 [4] Let X_i be a fuzzy random variable, and $a_i \in \mathbb{R}^1$, $i = 1, \dots, m$. Then the both $\sum_{i=1}^m a_i X_i$ and $\prod_{i=1}^m X_i$ are fuzzy random variables, and for every $\alpha \in [0, 1]$

$$\left(\sum_{i=1}^m a_i X_i\right)_\alpha = \sum_{i=1}^m a_i (X_i)_\alpha, \quad \left(\prod_{i=1}^m X_i\right)_\alpha = \prod_{i=1}^m (X_i)_\alpha.$$

Definition 3 [3] The expected value of the fuzzy random variable of X , denoted by $E(X)$, is the extended number $v \in \mathcal{F}_0(\mathbb{R}^n)$ such that

$$v_\alpha = \{E(f): f \text{ is a P-integral section of } X_\alpha\} \triangleq E(X_\alpha),$$

for every $\alpha \in [0, 1]$.

Now we define the variance of X as follows.

Definition 4 Let X^2 be integrably bounded [3], $E((X-E(X))^2)$ is called the variance of X , denoted by $D^2(X)$.

The existent and uniqueness of $D^2(X)$ can be obtained by property 2 and theorem 3.1 in [3].

In general case, the computations of the expected value and variance of X are complex just as L.P.Madeh pointed out. The following we limit a fuzzy random variable to taking in $\mathcal{F}(R^1)$.

Theorem: Let X^2 be integrably bounded, then we have

$$D^2(X) = E(X^2) - (E(X))^2 .$$

Proof: We first prove the property:

$$E(X_1+X_2) = E(X_1) + E(X_2)$$

where X_i is any fuzzy random variable, $i = 1, 2$.

Let $(X_i)_\alpha(\omega) = [a_i(\omega, \alpha), b_i(\omega, \alpha)]$, for every $\alpha \in (0, 1]$ and for every $\omega \in \Omega$, $i = 1, 2$. Then for some ω , $a_i(\omega, \alpha)$ is an increasing function of α and $b_i(\omega, \alpha)$ is a decreasing function of α .

For some $\alpha \in (0, 1]$, let

$$\begin{aligned} \min (X_i)_\alpha : \Omega \rightarrow R^1 \\ \omega \mapsto \min(X_i)_\alpha(\omega) = a_i(\omega, \alpha) . \end{aligned}$$

For every $c \in R^1$,

$$\begin{aligned} \{ \omega \in \Omega : \min (X_i)_\alpha < c \} = & \left(\bigcup_{\substack{q \in \mathbb{Q} \\ q > c}} \{ \omega \in \Omega : (X_i)_\alpha(\omega) \cap \{q\} \neq \emptyset \} \right) \\ & \cap \{ \omega \in \Omega : (X_i)_\alpha(\omega) \cap \{c\} \neq \emptyset \} \end{aligned}$$

where \mathbb{Q} is the set of all rational numbers.

We have $\{ \omega : \min(X_i)_\alpha < c \} \in \mathcal{A}$ according to the equivalent definition of a fuzzy random variable, a.e. the function $\min(X_i)_\alpha$ measurable about \mathcal{B} . Then $\min(X_i)_\alpha$ is a P -integrable section of

$(X_i)_\alpha$ because of $a_i(\omega, \alpha) \in [a_i(\omega, \alpha), b_i(\omega, \alpha)]$.

Similarly, the function

$$\begin{aligned} \max (X_i)_\alpha : \Omega &\longrightarrow \mathbb{R}^1 \\ \omega &\longmapsto \max (X_i)_\alpha(\omega) = b_i(\omega, \alpha) \end{aligned}$$

is also P-integral section of $(X_i)_\alpha$.

It is obvious that for every P-integral section f of $(X_i)_\alpha$, we have $E(\min(X_i)_\alpha) \leq E(f) \leq E(\max(X_i)_\alpha)$. So we can obtain:

$$\begin{aligned} E((X_i)_\alpha) &= [E(\min(X_i)_\alpha), E(\max(X_i)_\alpha)] \\ &= [E(a_i(\omega, \alpha)), E(b_i(\omega, \alpha))] \end{aligned}$$

then

$$\begin{aligned} (E(X_1) + E(X_2))_\alpha &= (E(X_1))_\alpha + (E(X_2))_\alpha \\ &= [E(a_1(\omega, \alpha)), E(b_1(\omega, \alpha))] \\ &\quad + [E(a_2(\omega, \alpha)), E(b_2(\omega, \alpha))] \\ &= [E(a_1(\omega, \alpha)) + E(a_2(\omega, \alpha)), E(b_1(\omega, \alpha)) \\ &\quad \quad \quad + E(b_2(\omega, \alpha))] \\ &= [E(a_1(\omega, \alpha) + a_2(\omega, \alpha)), E(b_1(\omega, \alpha) + b_2(\omega, \alpha))] \\ &= (E(X_1 + X_2))_\alpha \end{aligned}$$

Similarly, we have $E(\alpha X) = \alpha E(X)$, for every $\alpha \in \mathbb{R}^1$.

Then

$$\begin{aligned} D(X^2) &= E((X - E(X))^2) \\ &= E(X^2 - 2E(X)X + (E(X))^2) \\ &= E(X^2) - 2E(X)E(X) + (E(X))^2 \\ &= E(X^2) - (E(X))^2 \end{aligned}$$

By the proof of the theorem, we know, if $X_\alpha(w) = [a(w, \alpha), b(w, \alpha)]$ then

$$E(X) = \bigcup_{\alpha \in (0, 1]} \alpha \cdot [E(a(w, \alpha)), E(b(w, \alpha))]$$

and

$$\begin{aligned} (E(X))^2 &= \bigcup_{\alpha \in (0, 1]} \alpha \cdot [E(a(w, \alpha)), E(b(w, \alpha))]^2 \\ &= \bigcup_{\alpha \in (0, 1]} \alpha \cdot [a'(\alpha), b'(\alpha)] \end{aligned}$$

according to L.A.Zadeh's extension principle. Where

$$\begin{aligned} a'(\alpha) &= \min \{ E^2(a(w, \alpha)), E(a(w, \alpha)) \cdot E(b(w, \alpha)), E^2(b(w, \alpha)) \} \\ &= \max \{ E^2(a(w, \alpha)), E(a(w, \alpha)) \cdot E(b(w, \alpha)), E^2(b(w, \alpha)) \} \end{aligned}$$

In the similar way, we have

$$(X^2(w))_\alpha = (X_\alpha(w))^2 = [\tilde{a}(w, \alpha), \tilde{b}(w, \alpha)]$$

where

$$\tilde{a}(w, \alpha) = \min \{ a^2(w, \alpha), a(w, \alpha) \cdot b(w, \alpha), b^2(w, \alpha) \}$$

$$\tilde{b}(w, \alpha) = \max \{ a^2(w, \alpha), a(w, \alpha) \cdot b(w, \alpha), b^2(w, \alpha) \}$$

then

$$\begin{aligned} E((X_\alpha)^2) &= E((X^2)_\alpha) \\ &= [E(\tilde{a}(w, \alpha)), E(\tilde{b}(w, \alpha))] \end{aligned}$$

So we have

$$D^2(X) = \bigcup_{\alpha \in (0, 1]} \alpha \cdot [E(\tilde{a}(w, \alpha)) - b'(\alpha), E(\tilde{b}(w, \alpha)) - a'(\alpha)]$$

Note: In general case, the above-proved theorem is incorrect.

References

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