RESEARCH

Open Access

Variance of uncertain random variables

Haiying Guo and Xiaosheng Wang*

*Correspondence: xswang@hebeu.edu.cn School of Sciences, Hebei University of Engineering, Handan 056038, China

Abstract

Uncertain random variable is developed to describe the phenomenon which mixes uncertainty with randomness. The variance of uncertain random variable may provide a degree of the spread of the distribution around its expected value. This paper presents a formula to obtain the variance of uncertain random variable.

Keywords: Uncertainty theory; Uncertain random variable; Chance measure; Variance

1 Introduction

Prior to today, probability theory and fuzzy set theory are two common mathematical tools to model indeterminacy phenomena and have been widely applied in information theory, engineering, management science, and so on. We know a fundamental premise of applying probability theory is that the estimated probability is closed enough to the real frequency. However, such as 'strength of bridge', 'about one million tons', 'tall', and 'most', due to lack of observed data and the complexity of environment, when making decisions, people have to consult with domain experts. In this case, information and knowledge cannot be described well by random variables. For fuzzy set theory, it was still challenged by many scholars after it was founded. Liu [1] presented several paradoxes to show that fuzzy variable and fuzzy set are not suitable for modeling uncertain quantities and unsharp concepts, respectively.

In order to deal with non-random phenomena, an uncertainty theory was founded by Liu [2] and refined by Liu [3] and then became a branch of mathematics for modeling belief degrees. One the hand, Liu presented some basic and important theoretical works of uncertainty theory [3]. On the other hand, as an application of uncertainty theory, Liu [4] proposed a spectrum of uncertain programming which is a type of mathematical programming involving uncertain variables and applied uncertain programming to system reliability design, facility location problem, vehicle routing problem, project scheduling problem, finance, and so on. In order to well interpret expert's experimental data by uncertainty theory, Liu first proposed uncertain statistics, and Wang et al. build several uncertain statistical methods such as the method of moments [5], uncertain Delphi method [6], and uncertain hypothesis testing for expert's empirical data [7] and discussed uncertain variance of sample and its application [8]. Nowadays, uncertainty theory was well developed on both theory section and practice section. For exploring the recent developments of uncertainty theory, the readers may consult the book [1].

Inspired by Kwakernaak [9,10], Puri and Ralescu [11], Kruse and Meyer [12], and Liu and Liu [13,14], uncertain random variable was first defined by Liu [15] to describe



© 2014 Guo and Wang; licensee Springer. This is an Open Access article distributed under the terms of the Creative Commons Attribution License (http://creativecommons.org/licenses/by/2.0), which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited. complex systems in which uncertainty and randomness frequently appear together. Otherwise, in order to model uncertain random phenomena well, Liu [15] also introduce the concepts of uncertain random arithmetic, chance measure, chance distribution, expected value, variance, and so on. As everyone knows, the variance of uncertain random variable will provide a degree of the spread of the distribution around its expected value. Although the concept of variance for uncertain random variable has been defined by expected value, we still do not know how to obtain the variance from chance distribution. This paper will build the formula for the convenience of obtaining the variance from chance distribution. The remainder of this paper is organized as follows. The next section is intended to introduce some concepts in uncertainty theory as they are needed. A formula for obtaining the variance of uncertain random variable and some examples are proposed in Section 3. Finally, a conclusion is drawn in Section 4.

2 Preliminary

Uncertainty theory was founded by Liu [2] in 2007 and refined by Liu [3] in 2010 and became a branch of mathematics based on normality, duality, subadditivity, and product axioms. For exploring the recent developments of uncertainty theory, the readers can consult the book by Liu [1]. Based on uncertainty theory and probability theory, chance theory was pioneered by Liu [15] in order to model the uncertain random compound systems. In this section, some useful concepts are introduced. Let $(\Gamma, \mathcal{L}, \mathcal{M})$ be an uncertainty space, where Γ is a non-empty set, \mathcal{L} is a σ -algebra defined on Γ , and \mathcal{M} is an uncertain measure, and let $(\Omega, \mathcal{A}, Pr)$ be a probability space, where Ω is a sample space, \mathcal{A} is a σ -algebra defined on Ω , and Pr is a probability measure.

Definition 2.1. An uncertain random variable is a function ξ from a chance space $(\Gamma, \mathcal{L}, \mathcal{M}) \times (\Omega, \mathcal{A}, Pr)$ to the set of real numbers such that $\{\xi \in B\}$ is an event in $\mathcal{L} \times \mathcal{A}$ for any Borel set *B* [15].

Definition 2.2. Let ξ be an uncertain random variable on the chance space $(\Gamma, \mathcal{L}, \mathcal{M}) \times (\Omega, \mathcal{A}, Pr)$, and let *B* be a Borel set. Then, $\{\xi \in B\}$ is an uncertain random event with chance measure [15]

$$Ch\{\xi \in B\} = \int_0^1 Pr\{\omega \in \Omega \mid \mathcal{M}\{\gamma \in \Gamma \mid \xi(\gamma, \omega) \in B\} \ge x\}dx.$$
(2.1)

Note that the chance measure is in fact the expected value of the random variable $\mathcal{M}\{\xi(\cdot) \in B\}$, i.e.,

$$Ch\{\xi \in B\} = E[\mathcal{M}\{\xi(\cdot) \in B\}], \qquad (2.2)$$

where E denotes the expected value operator for random variable under probability measure Pr.

Theorem 2.1. Let ξ be an uncertain random variable. Then, the chance measure $Ch\{\xi \in B\}$ has normality, duality, and monotonicity properties, i.e., [15]

- (*i*) $Ch\{\xi \in \Re\} = 1$,
- (*ii*) $Ch\{\xi \in B\} + Ch\{\xi \in B^c\} = 1$, for any Borel set B,
- (iii) $Ch\{\xi \in B_1\} \leq Ch\{\xi \in B_2\}$, for any Borel sets $B_1 \subset B_2$.

Definition 2.3. Let ξ be an uncertain random variable. Then, its chance distribution of ξ is defined by [15]

$$\Phi(x) = Ch\{\xi \le x\}, \text{ for any } x \in \mathfrak{R}.$$

An operational law of uncertain random variables was given by Liu [16] as follows.

Theorem 2.2. Let $\eta_1, \eta_2, \dots, \eta_m$ be independent random variables with probability distributions $\Psi_1, \Psi_2, \dots, \Psi_m$, and let $\tau_1, \tau_2, \dots, \tau_n$ be independent uncertain variables with uncertainty distributions $\Upsilon_1, \Upsilon_2, \dots, \Upsilon_n$, respectively. Then, the uncertain random variable $\xi = f(\eta_1, \eta_2, \dots, \eta_m, \tau_1, \tau_2, \dots, \tau_n)$ has a chance distribution [16]

$$\Phi(x) = \int_{\mathfrak{M}^m} F(x; y_1, y_2, \cdots, y_m) d\Psi_1(y_1) d\Psi_2(y_2) \cdots d\Psi_m(y_m),$$
(2.3)

where $F(x; y_1, y_2, \dots, y_m)$ is determined by its inverse function

$$F^{-1}(\alpha; y_1, y_2, \cdots, y_m) = f(y_1, y_2, \cdots, y_m, \Upsilon_1^{-1}(\alpha), \Upsilon_2^{-1}(\alpha), \cdots, \Upsilon_n^{-1}(\alpha))$$
(2.4)

provided that $f(\eta_1, \eta_2, \dots, \eta_m, \tau_1, \tau_2, \dots, \tau_n)$ is a strictly increasing function with respect to $\tau_1, \tau_2, \dots, \tau_n$.

Note that if $\eta_1, \eta_2, \dots, \eta_m$ are non-independent random variables with joint distribution $\Psi(y_1, y_2, \dots, y_m)$, Equation 2.3 can be modified as below,

$$\Phi(x) = \int_{\Re^m} F(x; y_1, y_2, \cdots, y_m) d\Psi(y_1, y_2, \cdots, y_m).$$
(2.5)

Definition 2.4. Let ξ be an uncertain random variable. Then its expected value is defined by [15]

$$E[\xi] = \int_0^{+\infty} Ch\{\xi \ge r\} dr - \int_{-\infty}^0 Ch\{\xi \le r\} dr$$
(2.6)

provided that at least one of the two integrals is finite.

Theorem 2.3. Let $\eta_1, \eta_2, \dots, \eta_m$ be independent random variables with probability distributions $\Psi_1, \Psi_2, \dots, \Psi_m$, respectively, and let $\tau_1, \tau_2, \dots, \tau_n$ be uncertain variables, then the uncertain random variable [16]

 $\xi = f(\eta_1, \cdots, \eta_m, \tau_1, \cdots, \tau_n)$

has an expected value

$$E[\xi] = \int_{\mathbb{R}^m} E[f(y_1, \cdots, y_m, \tau_1, \cdots, \tau_n)] d\Psi_1(y_1) \cdots d\Psi_m(y_m)$$

where $E[f(y_1, \dots, y_m, \tau_1, \dots, \tau_n)]$ is the expected value of the uncertain variable $f(\eta_1, \dots, \eta_m, \tau_1, \dots, \tau_n)$ for any real numbers y_1, \dots, y_m .

As a special case, Liu [16] proved the following theorem.

Theorem 2.4. Let η be a random variable and let τ be an uncertain variable [16]. Then

 $E[\eta + \tau] = E[\eta] + E[\tau] \tag{2.7}$

and

$$E[\eta\tau] = E[\eta] E[\tau].$$
(2.8)

3 Variance of uncertain random variables

A formal definition of variance for uncertain random variable was presented by Liu [15] as follows.

Definition 3.1. Let ξ be an uncertain random variable with expected value *e*. Then, its variance is defined by [15]

$$V[\xi] = E[(\xi - e)^2].$$
(3.9)

The variance of uncertain random variable provides a degree of spread of the distribution around its expected value. The following theorem presents a formula to obtain the variance for a class of uncertain random variable.

Theorem 3.1. Let $\eta_1, \eta_2, \dots, \eta_m$ be independent random variables with probability distributions $\Psi_1, \Psi_2, \dots, \Psi_m$, respectively, and let $\tau_1, \tau_2, \dots, \tau_n$ be uncertain variables; then, the uncertain random variable

 $\xi = f(\eta_1, \cdots, \eta_m, \tau_1, \cdots, \tau_n)$

has variance

$$V[\xi] = 2 \int_{\mathbb{R}^m} \int_0^\infty x[1 - F(e + x; y_1, \cdots, y_m) + F(e - x; y_1, \cdots, y_m)] \, dx d\Psi_1(y_1) \cdots d\Psi_m(y_m)$$
(3.10)

where

$$F(x; y_1, \cdots, y_m) = \mathcal{M}\{f(y_1, \cdots, y_m, \tau_1, \cdots, \tau_n) \le x\}$$

and

$$e = \int_{\mathfrak{R}^m} E[f(y_1, \cdots, y_m, \tau_1, \cdots, \tau_n)] d\Psi_1(y_1) \cdots d\Psi_m(y_m)$$

are the uncertainty distribution and expected value of uncertain random variable ξ , respectively.

Proof. Suppose that the uncertain random variable $\xi = f(\eta_1, \dots, \eta_m, \tau_1, \dots, \tau_n)$ has the expected value *e*, by the definitions of variance and chance measure, we have

$$V[\xi] = \int_{0}^{+\infty} Ch\{(\xi - e)^{2} \ge x\} dx$$

= $\int_{0}^{+\infty} E_{Pr}[\mathcal{M}\{\xi - e)^{2} \ge x\}] dx$
= $\int_{0}^{+\infty} E_{Pr}[\mathcal{M}\{(f(\eta_{1}, \dots, \eta_{m}, \tau_{1}, \dots, \tau_{n}) - e)^{2} \ge x\}] dx$
= $\int_{0}^{+\infty} \int_{\mathbb{R}^{m}} \mathcal{M}\{(f(y_{1}, \dots, y_{m}; \tau_{1}, \dots, \tau_{n}) - e)^{2} \ge x\} d\Psi_{1}(y_{1}) \dots d\Psi_{m}(y_{m}) dx$
= $\int_{\mathbb{R}^{m}} \int_{0}^{+\infty} \mathcal{M}\{(f(y_{1}, \dots, y_{m}; \tau_{1}, \dots, \tau_{n}) - e)^{2} \ge x\} dx d\Psi_{1}(y_{1}) \dots d\Psi_{m}(y_{m}).$
(3.11)

For the above equation, we have

$$\int_{0}^{+\infty} \mathcal{M}\{(f(y_{1}, \cdots, y_{m}; \tau_{1}, \cdots, \tau_{n}) - e)^{2} \ge x\} dx$$

$$= \int_{0}^{+\infty} \mathcal{M}\{f(y_{1}, \cdots, y_{m}; \tau_{1}, \cdots, \tau_{n}) \ge e + \sqrt{x} \cup f(y_{1}, \cdots, y_{m}; \tau_{1}, \cdots, \tau_{n}) \le e - \sqrt{x}\} dx$$

$$\leq \int_{0}^{+\infty} \mathcal{M}\{f(y_{1}, \cdots, y_{m}; \tau_{1}, \cdots, \tau_{n}) \ge e + \sqrt{x}\} dx$$

$$+ \int_{0}^{+\infty} \mathcal{M}\{f(y_{1}, \cdots, y_{m}; \tau_{1}, \cdots, \tau_{n}) \le e - \sqrt{x}\} dx$$

$$= \int_{0}^{+\infty} [1 - F(e + \sqrt{x}; y_{1}, \cdots, y_{m}) + F(e - \sqrt{x}; y_{1}, \cdots, y_{m})] dx$$

$$= 2\int_{0}^{+\infty} x[1 - F(e + x; y_{1}, \cdots, y_{m}) + F(e - x; y_{1}, \cdots, y_{m})] dx, \qquad (3.12)$$

where $F(x; y_1, \dots, y_m)$ is the uncertainty distribution of uncertain variable $f(y_1, \dots, y_m; \tau_1, \dots, \tau_n)$ for any real numbers y_1, y_2, \dots, y_m .

By Equations 3.11 and 3.12, this theorem is proved.

Example 1. Let η be a random variable with probability distribution Ψ , and let τ be an uncertain variable with uncertainty distribution Υ . Then, the sum

$$\xi = \eta + \tau$$

has the variance

$$V[\xi] = 2 \int_{-\infty}^{+\infty} \int_{0}^{+\infty} x[1 - \Upsilon(e + x - y) + \Upsilon(e - x - y)] dx d\Psi(y)$$

where *e* is the expected value of uncertain random variable ξ and the above integral is finite.

Especially, let η be a random variable having probability density function

$$\phi(x) = \begin{cases} \frac{1}{d-c}, & \text{if } c \le x \le d \\ 0, & \text{otherwise,} \end{cases}$$
(3.13)

and let τ be an linear uncertain variable with uncertainty distribution

$$\Upsilon(x) = \begin{cases} 0, & \text{if } x < a \\ \frac{x-a}{b-a}, & \text{if } a \le x \le b \\ 1, & \text{otherwise.} \end{cases}$$
(3.14)

By Theorem 2.3, the sum

$$\xi = \eta + \tau$$

has the variance

$$V[\xi] = \frac{(d-c)^2}{12} + \frac{(b-a)^2}{12} = V[\eta] + V[\tau].$$

Example 2. Let η be a random variable with probability distribution Ψ , and let τ be an uncertain variable with uncertainty distribution Υ . Then, the product

$$\xi = \eta \eta$$

has the variance

$$V[\xi] = 2 \int_{-\infty}^{+\infty} \int_{0}^{+\infty} x \left[1 - \Upsilon \left(\frac{e+x}{y} \right) + \Upsilon \left(\frac{e-x}{y} \right) \right] dx d\Psi(y),$$

where *e* is the expected value of uncertain random variable ξ and the above integral is finite.

Especially, let η be a random variable having probability density function

$$\phi(x) = \begin{cases} \frac{1}{d-c}, & \text{if } 0 \le c \le x \le d\\ 0, & \text{otherwise} \end{cases}$$
(3.15)

and let τ be an linear uncertain variable with uncertainty distribution

$$\Upsilon(x) = \begin{cases} 0, & \text{if } x < a \\ \frac{x-a}{b-a}, & \text{if } 0 \le a \le x \le b \\ 1, & \text{otherwise.} \end{cases}$$
(3.16)

Then, the product

 $\xi = \eta \tau$

has the variance

$$V[\xi] = -\frac{1}{16}(a+b)^2(c+d)^2 + \frac{1}{9}(a^2+b^2+ab)(c^2+d^2+cd)$$

Remarks 3.1. The variance of uncertain random variable ξ provides a degree of spread of the distribution around its expected value *e*. If we only know its chance distribution Φ , then the variance

$$V[\xi] = \int_{0}^{+\infty} Ch\{(\xi - e)^{2} \ge x\} dx$$

= $\int_{0}^{+\infty} Ch\{(\xi \ge e + \sqrt{x}) \cup (\xi \le e - \sqrt{x})\} dx$
 $\le \int_{0}^{+\infty} (Ch\{\xi \ge e + \sqrt{x}\} + Ch\{\xi \le e - \sqrt{x}\}) dx$
= $\int_{0}^{+\infty} (1 - \Phi(e + \sqrt{x}) + \Phi(e - \sqrt{x})) dx.$

Then, we have the following stipulation

$$V[\xi] = \int_0^{+\infty} (1 - \Phi(e + \sqrt{x}) + \Phi(e - \sqrt{x})) dx.$$

4 Conclusions

Uncertain random variable and chance theory were introduced to model the uncertain random compound systems. As the same as random variable and uncertain variable, expected value is the average of uncertain random variable in the sense of chance measure and the variance of uncertain random variable provides a degree of spread of the distribution around its expected value. In this paper, a series of formulas were built to obtain the variance of uncertain random variable.

Acknowledgements

This work was supported by the National Natural Science Foundation of China (No. 61073121) and Hebei Natural Science Foundations (No. G2013402063, F2012402037).

Received: 9 February 2014 Accepted: 26 February 2014 Published: 25 March 2014

References

- 1. Liu, B: Uncertainty theory, 4th ed. http://orsc.edu.cn/liu/ut.pdf. (2014)
- 2. Liu, B: Uncertainty Theory. 2nd ed. Springer-Verlag, Berlin (2007)
- 3. Liu, B: Uncertainty Theory: A Branch of Mathematics for Modeling Human Uncertainty. Springer-Verlag, Berlin (2010)

- 4. Liu, B: Theory and Practice of Uncertain Programming. 2nd ed. Springer-Verlag, Berlin (2009)
- 5. Wang, X, Peng, Z: Method of moments for estimating uncertainty distribution. J. Uncertainty Anal. Applications. 2(5), 1–12 (2014)
- Wang, X, Gao, Z, Guo, H: Delphi method for estimating uncertainty distributions. Inform. An Int. Interdiscip. J. 15(2), 449–460 (2012)
- Wang, X, Gao, Z, Guo, H: Uncertain hypothesis testing for two experts empirical data. Math. Comput. Model. 55, 1478–1482 (2012)
- 8. Wang, X, Guo, H: Uncertain variance of sample and its application. Inform.: An Int. Interdiscip. J. **14**(1), 79–88 (2011)
- 9. Kwakernaak, H: Fuzzy random variables-1: definitions and theorems. Inform. Sci. 15, 1–29 (1978)
- Kwakernaak, H: Fuzzy random variables-2: algorithms and examples for the discrete case. Inform. Sci. 17, 713–720 (1979)
- 11. Puri, M, Ralescu, D: Fuzzy random variables. J. Math. Appl. **114**, 409–422 (1986)
- 12. Kruse, R, Meyer, K: Statistics with Vague Data. Reidel Publishing Company, Dordrecht (1987)
- Liu, YK, Liu, B: Fuzzy random variables: a scalar expected value operator. Fuzzy Optimization Decis. Making. 2(2), 143–160 (2003)
- 14. Liu, YK, Liu, B: Fuzzy random programming with equilibrium chance constraints. Inform. Sci. 170, 363–395 (2005)
- 15. Liu, YH: Uncertain random variables: a mixture of uncertainty and randomness. Soft Comput. 17(4), 625–634 (2013)
- 16. Liu, YH: Uncertain random programming with applications. Fuzzy Optimization Decis. Making. 12(2), 153–169 (2013)

doi:10.1186/2195-5468-2-6

Cite this article as: Guo and Wang: Variance of uncertain random variables. Journal of Uncertainty Analysis and Applications 2014 2:6.

Submit your manuscript to a SpringerOpen[®] journal and benefit from:

- ► Convenient online submission
- ► Rigorous peer review
- Immediate publication on acceptance
- Open access: articles freely available online
- ► High visibility within the field
- ► Retaining the copyright to your article

Submit your next manuscript at > springeropen.com