Research Article

Moment Estimation Inequalities Based on g_{λ} **Random Variable on Sugeno Measure Space**

Jingfeng Tian,¹ Zhiming Zhang,² and Dazeng Tian³

¹ College of Science and Technology, North China Electric Power University, Baoding, Hebei Province 071051, China

² College of Mathematics and Computer Sciences, Hebei University, Baoding, Hebei Province 071002, China

³ College of Physics Science and Technology, Hebei University, Baoding, Hebei Province 071002, China

Correspondence should be addressed to Jingfeng Tian, tianjfhxm_ncepu@yahoo.cn

Received 8 October 2009; Accepted 12 December 2009

Academic Editor: Andrei Volodin

Copyright © 2010 Jingfeng Tian et al. This is an open access article distributed under the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

The definitions and properties of moment of g_{λ} random variable are provided on Sugeno measure space. Then some important moment estimation inequalities based on g_{λ} random variable are presented and proven.

1. Introduction

In 1974, the Japanese scholar Sugeno [1] presented a kind of typical nonadditive measure, Sugeno measure, which is an important generalization of probability measure [2–6]. As we all know, the definitions and properties of moment of random variable play an important role in probability theory [7–9]. Likewise, they are also very important for Sugeno measure. In this paper we present the analogous definitions and properties based on g_{λ} random variable on Sugeno measure space. Then some important moment estimation inequalities based on g_{λ} random variable are presented and proven.

2. Preliminaries

Let us recall some definitions and facts from [5].

Definition 2.1. Let X be a nonempty set, let ζ be a nonempty class of subsets of X, and let μ be a nonnegative real valued set function defined on ζ . Therefore μ satisfies the σ - λ rule (on ζ)

if and only if there exists

$$\lambda \in \left(-\frac{1}{\sup \mu}, \infty\right) \cup \{0\}$$
(2.1)

such that

$$\mu\left(\bigcup_{i=1}^{\infty} E_{i}\right) = \begin{cases} \frac{1}{\lambda} \left\{ \prod_{i=1}^{\infty} \left[1 + \lambda \cdot \mu(E_{i})\right] - 1 \right\}, & \text{as } \lambda \neq 0, \\ \sum_{i=1}^{\infty} \mu(E_{i}), & \text{as } \lambda = 0, \end{cases}$$
(2.2)

for any disjoint sequence $\{E_i\}$ of sets in ζ whose union is also in ζ .

Definition 2.2. Let \mathcal{F} be a *σ*-algebra of subsets of *X*. And μ is called Sugeno measure on \mathcal{F} if and only if it satisfies the *σ*- λ rule and $\mu(X) = 1$. Usually, Sugeno measure on \mathcal{F} is denoted by g_{λ} .

We call the triple $(X, \mathcal{F}, g_{\lambda})$ a Sugeno measure space, denoted by g_{λ} space, where $\lambda \in (-1, \infty)$. In the following, our discussion will be restricted to this space.

Theorem 2.3. For all $E, F \in \mathcal{F}, E \subset F$ imply that $g_{\lambda}(E) \leq g_{\lambda}(F)$ (monotonicity).

Theorem 2.4. Let g_{λ} be a Sugeno measure on \mathcal{F} . Then, for any $E \in \mathcal{F}$ and $F \in \mathcal{F}$,

$$g_{\lambda}(E \cup F) = \frac{g_{\lambda}(E) + g_{\lambda}(F) - g_{\lambda}(E \cap F) + \lambda g_{\lambda}(E)g_{\lambda}(F)}{1 + \lambda g_{\lambda}(E \cap F)},$$

$$g_{\lambda}(E - F) = \frac{g_{\lambda}(E) - g_{\lambda}(E \cap F)}{1 + \lambda g_{\lambda}(E \cap F)},$$

$$g_{\lambda}(E^{c}) = \frac{1 - g_{\lambda}(E)}{1 + \lambda g_{\lambda}(E)}.$$
(2.3)

In order to present the analogous definitions and properties based on g_{λ} random variable on Sugeno measure space, we recall some definitions and facts from [10].

Definition 2.5. Let ξ be a function mapping from $(X, \mathcal{F}, g_{\lambda})$ to real line \mathbb{R} . Then ξ is called a g_{λ} random variable.

Definition 2.6. Let ξ be a g_{λ} random variable. Then the distribution function of ξ is defined by

$$F_{g_{\lambda}}(x) = g_{\lambda}\{\xi \le x\}, \quad \forall x \in \mathbb{R}.$$
(2.4)

Definition 2.7. Let $F_{g_{\lambda}}(x)$ be the distribution function of g_{λ} random variable ξ . Then ξ is called continuous g_{λ} random variable if there exists a nonnegative real valued function $f_{g_{\lambda}}(x)$ such that

$$F_{g_{\lambda}}(x) = \int_{-\infty}^{x} f_{g_{\lambda}}(t) dt, \quad \forall x \in \mathbb{R}$$
(2.5)

is valid. The function $f_{g_{\lambda}}(x)$ is called a density function of ξ .

In the following, our discussion will be restricted to the continuous g_{λ} random variable.

Definition 2.8. Let $F_{g_{\lambda}}(x)$ be the distribution function of g_{λ} random variable ξ . If $\int_{-\infty}^{\infty} |x| dF_{g_{\lambda}}(x) < \infty$, then we call $\int_{-\infty}^{\infty} x dF_{g_{\lambda}}(x)$ an expected value of g_{λ} random variable ξ , denoted by $E_{g_{\lambda}}(\xi)$.

Theorem 2.9. Let ξ, η be g_{λ} random variables; let *C* and *D* be constants. Then

$$E_{g_{\lambda}}(C\xi + D\eta) = CE_{g_{\lambda}}(\xi) + DE_{g_{\lambda}}(\eta).$$
(2.6)

Definition 2.10. Let ξ be a g_{λ} random variable. If $E_{g_{\lambda}}\{[\xi - E_{g_{\lambda}}(\xi)]^2\}$ exists, then $E_{g_{\lambda}}\{[\xi - E_{g_{\lambda}}(\xi)]^2\}$ is called the variance of ξ , denoted by $D_{g_{\lambda}}(\xi)$.

3. Moment Estimation Inequalities Based on g_{λ} **Random Variable**

We begin this section with a short lemma (see [11]), which will be useful in the sequel.

Lemma 3.1. Let ξ be a g_{λ} random variable whose Sugeno density function $f_{g_{\lambda}}$ exists. If the Lebesgue integral

$$\int_{0}^{+\infty} g_{\lambda}\{\xi \ge r\}dr - \int_{-\infty}^{0} g_{\lambda}\{\xi \le r\}dr + \lambda \int_{0}^{+\infty} g_{\lambda}\{\xi \ge r\} \cdot g_{\lambda}\{\xi \le r\}dr$$
(3.1)

is finite, then

$$E_{g_{\lambda}}(\xi) = \int_{0}^{+\infty} g_{\lambda}\{\xi \ge r\} dr - \int_{-\infty}^{0} g_{\lambda}\{\xi \le r\} dr + \lambda \int_{0}^{+\infty} g_{\lambda}\{\xi \ge r\} \cdot g_{\lambda}\{\xi \le r\} dr.$$
(3.2)

Theorem 3.2. Let ξ be a nonnegative g_{λ} random variable. When $\lambda \ge 0$, the inequality

$$\sum_{i=1}^{\infty} g_{\lambda}\{\xi \ge i\} \le E_{g_{\lambda}}(\xi) \le (1+\lambda) \left(1 + \sum_{i=1}^{\infty} g_{\lambda}\{\xi \ge i\}\right)$$
(3.3)

is valid; when $\lambda < 0$ *, the inequality*

$$(1+\lambda)\sum_{i=1}^{\infty} g_{\lambda}\{\xi \ge i\} \le E_{g_{\lambda}}(\xi) \le 1 + \sum_{i=1}^{\infty} g_{\lambda}\{\xi \ge i\}$$
(3.4)

holds true.

Proof. (I) When $\lambda \ge 0$, since $g_{\lambda} \{\xi \ge r\}$ is a monotone decreasing function of r, we have

$$\begin{split} E_{g_{\lambda}}(\xi) &= \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr + \lambda \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} \cdot g_{\lambda}\{\xi \leq r\} dr \\ &\geq \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr \\ &= \sum_{i=1}^{\infty} \int_{i-1}^{i} g_{\lambda}\{\xi \geq r\} dr \\ &\geq \sum_{i=1}^{\infty} \int_{i-1}^{i} g_{\lambda}\{\xi \geq i\} dr \\ &= \sum_{i=1}^{\infty} g_{\lambda}\{\xi \geq i\}, \end{split}$$

$$\begin{split} E_{g_{\lambda}}(\xi) &= \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr + \lambda \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} \cdot g_{\lambda}\{\xi \leq r\} dr \\ &\leq \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr + \lambda \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr \\ &= (1+\lambda) \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr \\ &= (1+\lambda) \sum_{i=1}^{\infty} \int_{i-1}^{i} g_{\lambda}\{\xi \geq r\} dr \\ &\leq (1+\lambda) \sum_{i=1}^{\infty} \int_{i-1}^{i} g_{\lambda}\{\xi \geq i-1\} dr \\ &= (1+\lambda) \left(1 + \sum_{i=1}^{\infty} g_{\lambda}\{\xi \geq i\}\right). \end{split}$$

(II) When $\lambda < 0$, owing to the monotonicity of $g_{\lambda} \{\xi \ge r\}$ we also have

$$\begin{split} E_{g_{\lambda}}(\xi) &= \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr + \lambda \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr \\ &\geq \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr + \lambda \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr \\ &= (1+\lambda) \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr \\ &= (1+\lambda) \sum_{i=1}^{\infty} \int_{i-1}^{i} g_{\lambda}\{\xi \geq r\} dr \\ &\geq (1+\lambda) \sum_{i=1}^{\infty} \int_{i-1}^{i} g_{\lambda}\{\xi \geq i\} dr \\ &= (1+\lambda) \sum_{i=1}^{\infty} g_{\lambda}\{\xi \geq i\}, \end{split}$$
(3.6)
$$E_{g_{\lambda}}(\xi) &= \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr + \lambda \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} \cdot g_{\lambda}\{\xi \leq r\} dr \\ &\leq \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr \\ &= \sum_{i=1}^{\infty} \int_{i-1}^{i} g_{\lambda}\{\xi \geq r\} dr \\ &= \sum_{i=1}^{\infty} \int_{i-1}^{i} g_{\lambda}\{\xi \geq i-1\} dr \\ &\leq \sum_{i=1}^{\infty} \int_{i-1}^{i} g_{\lambda}\{\xi \geq i-1\} dr \\ &= 1 + \sum_{i=1}^{\infty} g_{\lambda}\{\xi \geq i\}. \end{split}$$

Definition 3.3. Let ξ be a g_{λ} random variable and k a positive number. Then (1) the expected value $E_{g_{\lambda}}(\xi^k)$ is called the *k*th moment, (2) the expected value $E_{g_{\lambda}}(|\xi|^k)$ is called the *k*th absolute moment, (3) the expected value $E_{g_{\lambda}}\{[\xi - E_{g_{\lambda}}(\xi)]^k\}$ is called the *k*th central moment, and (4) the expected value $E_{g_{\lambda}}\{[\xi - E_{g_{\lambda}}(\xi)]^k\}$ is called the *k*th absolute central moment.

Theorem 3.4. *Let* ξ *be a nonnegative* g_{λ} *random variable and* k *a positive number. Then*

$$E_{g_{\lambda}}\left(\xi^{k}\right) = k \int_{0}^{+\infty} r^{k-1} g_{\lambda}\left\{\xi \ge r\right\} dr + k\lambda \int_{0}^{+\infty} r^{k-1} g_{\lambda}\left\{\xi \ge r\right\} \cdot g_{\lambda}\left\{\xi \le r\right\} dr.$$
(3.7)

Proof. From Lemma 3.1, we infer

$$E_{g_{\lambda}}\left(\xi^{k}\right) = \int_{0}^{+\infty} g_{\lambda}\left\{\xi^{k} \ge x\right\} dx + \lambda \int_{0}^{+\infty} g_{\lambda}\left\{\xi^{k} \ge x\right\} \cdot g_{\lambda}\left\{\xi^{k} \le x\right\} dx$$
$$= \int_{0}^{+\infty} g_{\lambda}\left\{\xi \ge r\right\} dr^{k} + \lambda \int_{0}^{+\infty} g_{\lambda}\left\{\xi \ge r\right\} \cdot g_{\lambda}\left\{\xi \le r\right\} dr^{k}$$
$$= k \int_{0}^{+\infty} r^{k-1} g_{\lambda}\left\{\xi \ge r\right\} dr + k\lambda \int_{0}^{+\infty} r^{k-1} g_{\lambda}\left\{\xi \ge r\right\} \cdot g_{\lambda}\left\{\xi \le r\right\} dr.$$

$$(3.8)$$

Similar to the case of credibility theory [12], we have the following: Theorems 3.5, 3.6, and 3.7.

Theorem 3.5. Let ξ be a g_{λ} random variable that takes values in [m, n] and has expected value $E_{g_{\lambda}}(\xi)$, and let f(x) be a convex function on [m, n]. Then

$$E_{g_{\lambda}}\left[f(\xi)\right] \le \frac{n - E_{g_{\lambda}}(\xi)}{n - m} f(m) + \frac{E_{g_{\lambda}}(\xi) - m}{n - m} f(n).$$

$$(3.9)$$

Theorem 3.6. Let ξ be a g_{λ} random variable that takes values in [m, n] and has expected value $E_{g_{\lambda}}(\xi)$. Then

$$D_{g_{\lambda}}(\xi) \le \left[E_{g_{\lambda}}(\xi) - m \right] \left[n - E_{g_{\lambda}}(\xi) \right].$$
(3.10)

Theorem 3.7. Let ξ be a g_{λ} random variable that takes values in [m, n] and has expected value μ . Then, for any positive integer k,

$$E_{g_{\lambda}}(|\xi|^{k}) \leq \frac{n-\mu}{n-m}|m|^{k} + \frac{\mu-m}{n-m}|n|^{k},$$

$$E_{g_{\lambda}}(|\xi-\mu|^{k}) \leq \frac{n-\mu}{n-m}|\mu-m|^{k} + \frac{\mu-m}{n-m}|n-\mu|^{k}.$$
(3.11)

Theorem 3.8. Let ξ be a g_{λ} random variable and t > 0. Then $E_{g_{\lambda}}(|\xi|^{t}) < \infty$ if and only if $\sum_{i=1}^{\infty} g_{\lambda}\{|\xi| > i^{1/t}\} < \infty$.

Proof. From $g_{\lambda}\{|\xi|^t \ge i\} = g_{\lambda}\{|\xi| \ge i^{1/t}\}$ and Theorem 3.2, the conclusion is valid.

Theorem 3.9. Let ξ be a g_{λ} random variable and t > 0. If $E_{g_{\lambda}}(|\xi|^{t}) < \infty$, then $\lim_{x \to \infty} x^{t} g_{\lambda}\{|\xi| \ge x\} = 0$. Conversely, if there exists one positive number t such that $\lim_{x \to \infty} x^{t} g_{\lambda}\{|\xi| \ge x\} = 0$, then $E_{g_{\lambda}}(|\xi|^{s}) < \infty$ for any s, where $0 \le s < t$.

Proof. (1) When $\lambda \ge 0$, we have

$$E_{g_{\lambda}}\left(\left|\boldsymbol{\xi}\right|^{t}\right) = \int_{0}^{+\infty} g_{\lambda}\left\{\left|\boldsymbol{\xi}\right|^{t} \ge r\right\} dr + \lambda \int_{0}^{+\infty} g_{\lambda}\left\{\left|\boldsymbol{\xi}\right|^{t} \ge r\right\} \cdot g_{\lambda}\left\{\left|\boldsymbol{\xi}\right|^{t} \le r\right\} dr$$

$$\ge \int_{0}^{+\infty} g_{\lambda}\left\{\left|\boldsymbol{\xi}\right|^{t} \ge r\right\} dr.$$
(3.12)

Since $E_{g_{\lambda}}(|\xi|^{t}) < \infty$, we obtain $\int_{0}^{+\infty} g_{\lambda}\{|\xi|^{t} \ge r\} dr < \infty$. Consequently,

$$\lim_{x \to \infty} \int_{x^t/2}^{\infty} g_{\lambda} \Big\{ |\xi|^t \ge r \Big\} dr = 0.$$
(3.13)

Since

$$\int_{x^{t}/2}^{\infty} g_{\lambda} \Big\{ |\xi|^{t} \ge r \Big\} dr \ge \int_{x^{t}/2}^{x^{t}} g_{\lambda} \Big\{ |\xi|^{t} \ge r \Big\} dr \ge \frac{1}{2} x^{t} g_{\lambda} \{ |\xi| \ge x \},$$
(3.14)

we have

$$\lim_{x \to \infty} x^t g_{\lambda}\{|\xi| \ge x\} = 0. \tag{3.15}$$

(2) When $\lambda < 0$, we have

$$E_{g_{\lambda}}\left(\left|\boldsymbol{\xi}\right|^{t}\right) = \int_{0}^{+\infty} g_{\lambda}\left\{\left|\boldsymbol{\xi}\right|^{t} \ge r\right\} dr + \lambda \int_{0}^{+\infty} g_{\lambda}\left\{\left|\boldsymbol{\xi}\right|^{t} \ge r\right\} \cdot g_{\lambda}\left\{\left|\boldsymbol{\xi}\right|^{t} \le r\right\} dr$$
$$\ge \int_{0}^{+\infty} g_{\lambda}\left\{\left|\boldsymbol{\xi}\right|^{t} \ge r\right\} dr + \lambda \int_{0}^{+\infty} g_{\lambda}\left\{\left|\boldsymbol{\xi}\right|^{t} \ge r\right\} dr$$
$$= (1+\lambda) \int_{0}^{+\infty} g_{\lambda}\left\{\left|\boldsymbol{\xi}\right|^{t} \ge r\right\} dr.$$
(3.16)

Since

$$E_{g_{\lambda}}\left(\left|\boldsymbol{\xi}\right|^{t}\right) < \infty, \tag{3.17}$$

we obtain

$$(1+\lambda)\int_{0}^{+\infty}g_{\lambda}\left\{|\xi|^{t}\geq r\right\}dr<\infty.$$
(3.18)

Consequently,

$$\lim_{x \to \infty} (1+\lambda) \int_{x^t/2}^{\infty} g_{\lambda} \Big\{ |\xi|^t \ge r \Big\} dr = 0.$$
(3.19)

Since

$$(1+\lambda) \int_{x^{t}/2}^{\infty} g_{\lambda} \Big\{ |\xi|^{t} \ge r \Big\} dr \ge (1+\lambda) \int_{x^{t}/2}^{x^{t}} g_{\lambda} \Big\{ |\xi|^{t} \ge r \Big\} dr \ge \frac{1}{2} (1+\lambda) x^{t} g_{\lambda} \{ |\xi| \ge x \},$$
(3.20)

we have

$$\lim_{x \to \infty} x^t g_{\lambda}\{|\xi| \ge x\} = 0.$$
(3.21)

Conversely, if $\lim_{x\to\infty} x^t g_{\lambda}\{|\xi| \ge x\} = 0$, then there exists one number l such that $x^t g_{\lambda}\{|\xi| \ge x\} \le 1$, for all $x \ge l$.

(3) When $\lambda \ge 0$, for any *s*, where $0 \le s < t$, we have

$$\begin{split} E_{g_{\lambda}}(|\xi|^{s}) &= \int_{0}^{+\infty} g_{\lambda}\{|\xi|^{s} \ge r\}dr + \lambda \int_{0}^{+\infty} g_{\lambda}\{|\xi|^{s} \ge r\} \cdot g_{\lambda}\{|\xi|^{s} \le r\}dr \\ &\leq \int_{0}^{+\infty} g_{\lambda}\{|\xi|^{s} \ge r\}dr + \lambda \int_{0}^{+\infty} g_{\lambda}\{|\xi|^{s} \ge r\}dr \\ &= (1+\lambda) \int_{0}^{+\infty} g_{\lambda}\{|\xi|^{s} \ge r\}dr \\ &= (1+\lambda) \left(\int_{0}^{l} g_{\lambda}\{|\xi|^{s} \ge r\}dr + \int_{l}^{+\infty} g_{\lambda}\{|\xi|^{s} \ge r\}dr \right) \\ &= (1+\lambda) \left(\int_{0}^{l} g_{\lambda}\{|\xi|^{s} \ge r\}dr + \int_{l}^{+\infty} sr^{s-1}g_{\lambda}\{|\xi| \ge r\}dr \right) \\ &\leq (1+\lambda) \left(\int_{0}^{l} g_{\lambda}\{|\xi|^{s} \ge r\}dr + s \int_{l}^{+\infty} r^{s-t-1}dr \right) \\ &\leq (1+\lambda) \left(\int_{0}^{l} g_{\lambda}\{|\xi|^{s} \ge r\}dr + s \int_{0}^{+\infty} r^{s-t-1}dr \right). \end{split}$$

Since $\int_0^{+\infty} r^p dr < \infty$ for any p < -1, we have

$$E_{g_{\lambda}}(|\xi|^{s}) \leq (1+\lambda) \left(\int_{0}^{l} g_{\lambda}\{|\xi|^{s} \geq r\} dr + s \int_{0}^{+\infty} r^{s-t-1} dr \right) < \infty.$$

$$(3.23)$$

(4) When $\lambda < 0$, for any *s*, where $0 \le s < t$, we have

$$E_{g_{\lambda}}(|\xi|^{s}) = \int_{0}^{+\infty} g_{\lambda}\{|\xi|^{s} \ge r\}dr + \lambda \int_{0}^{+\infty} g_{\lambda}\{|\xi|^{s} \ge r\} \cdot g_{\lambda}\{|\xi|^{s} \le r\}dr$$

$$\leq \int_{0}^{+\infty} g_{\lambda}\{|\xi|^{s} \ge r\}dr$$

$$= \int_{0}^{l} g_{\lambda}\{|\xi|^{s} \ge r\}dr + \int_{l}^{+\infty} g_{\lambda}\{|\xi|^{s} \ge r\}dr$$

$$= \int_{0}^{l} g_{\lambda}\{|\xi|^{s} \ge r\}dr + \int_{l}^{+\infty} sr^{s-1}g_{\lambda}\{|\xi| \ge r\}dr$$

$$\leq \int_{0}^{l} g_{\lambda}\{|\xi|^{s} \ge r\}dr + s\int_{l}^{+\infty} r^{s-t-1}dr$$

$$\leq \int_{0}^{l} g_{\lambda}\{|\xi|^{s} \ge r\}dr + s\int_{0}^{+\infty} r^{s-t-1}dr.$$
(3.24)

Since $\int_{0}^{+\infty} r^{p} dr < \infty$ for any p < -1, we have

$$E_{g_{\lambda}}(|\xi|^{s}) \leq \int_{0}^{l} g_{\lambda}\{|\xi|^{s} \geq r\} dr + s \int_{0}^{+\infty} r^{s-t-1} dr < \infty.$$
(3.25)

Acknowledgment

This work was supported by the NNSF of China (no. 60773062), the NSF of Hebei Province of China (no. 2008000633), the foundation of North China Electric Power University (no. 200911033), the KSRP of Department of Education of Hebei Province of China (no. 2005001D), and the KSTRP of Ministry of Education of China (no. 206012).

References

- M. Sugeno, *Theory of fuzzy integrals and its applications*, Ph.D. dissertation, Tokyo Institute of Technology, 1974.
- [2] A. Basile, "Sequential compactness for sets of Sugeno fuzzy measures," *Fuzzy Sets and Systems*, vol. 21, no. 2, pp. 243–247, 1987.
- [3] M. Berres, "λ-additive measures on measure spaces," Fuzzy Sets and Systems, vol. 27, no. 2, pp. 159– 169, 1988.
- [4] T.-Y. Chen, J.-C. Wang, and G.-H. Tzeng, "Identification of general fuzzy measures by genetic algorithms based on partial information," *IEEE Transactions on Systems, Man, and Cybernetics, Part B*, vol. 30, no. 4, pp. 517–528, 2000.
- [5] Z. Y. Wang and G. J. Klir, Fuzzy Measure Theory, Plenum Press, New York, NY, USA, 1992.
- [6] S. T. Wierzchon, "An algorithm for identification of fuzzy measure," *Fuzzy Sets and Systems*, vol. 9, no. 1, pp. 69–78, 1983.
- [7] S. E. Graversen and G. Peškir, "Maximal inequalities for Bessel processes," *Journal of Inequalities and Applications*, vol. 2, no. 2, pp. 99–119, 1998.

- [8] S. H. Sung, "Moment inequalities and complete moment convergence," *Journal of Inequalities and Applications*, vol. 2009, Article ID 271265, 14 pages, 2009.
- [9] Q.-P. Zang, "A limit theorem for the moment of self-normalized sums," *Journal of Inequalities and Applications*, vol. 2009, Article ID 957056, 10 pages, 2009.
- [10] M. Ha, Y. Li, J. Li, and D. Tian, "The key theorem and the bounds on the rate of uniform convergence of learning theory on Sugeno measure space," *Science in China. Series F*, vol. 49, no. 3, pp. 372–385, 2006.
- [11] M. Ha, H. Zhang, W. Pedrycz, and H. Xing, "The expected value models on Sugeno measure space," International Journal of Approximate Reasoning, vol. 50, no. 7, pp. 1022–1035, 2009.
- [12] B. Liu, Uncertainty Theory. An Introduction to Its Axiomatic Foundation, vol. 154 of Studies in Fuzziness and Soft Computing, Springer; Plenum, Berlin, Germany, 2004.