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Research Article

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

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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

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

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

$$F_{g_1}(x) = g_{\lambda}\{\xi \le x\}, \quad \forall x \in \mathbb{R}.$$
 (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

$$E_{g_1}(C\xi + D\eta) = CE_{g_1}(\xi) + DE_{g_1}(\eta). \tag{2.6}$$

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

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

is finite, then

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

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

$$E_{g_{\lambda}}(\xi) = \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr + \lambda \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} \cdot g_{\lambda}\{\xi \leq r\} dr$$

$$\geq \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr$$

$$= \sum_{i=1}^{\infty} \int_{i-1}^{i} g_{\lambda}\{\xi \geq r\} dr$$

$$\geq \sum_{i=1}^{\infty} \int_{i-1}^{i} g_{\lambda}\{\xi \geq i\} dr$$

$$= \sum_{i=1}^{\infty} g_{\lambda}\{\xi \geq i\},$$

$$E_{g_{\lambda}}(\xi) = \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr + \lambda \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} \cdot g_{\lambda}\{\xi \leq r\} dr$$

$$\leq \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr + \lambda \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr$$

$$= (1 + \lambda) \int_{0}^{+\infty} g_{\lambda}\{\xi \geq r\} dr$$

$$= (1 + \lambda) \sum_{i=1}^{\infty} \int_{i-1}^{i} g_{\lambda}\{\xi \geq i\} dr$$

$$\leq (1 + \lambda) \sum_{i=1}^{\infty} \int_{i-1}^{i} g_{\lambda}\{\xi \geq i\} dr$$

$$= (1 + \lambda) \left(1 + \sum_{i=1}^{\infty} g_{\lambda}\{\xi \geq i\}\right).$$

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

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

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

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 kth moment, (2) the expected value $E_{g_{\lambda}}(|\xi|^k)$ is called the kth absolute moment, (3) the expected value $E_{g_{\lambda}}\{[\xi - E_{g_{\lambda}}(\xi)]^k\}$ is called the kth central moment, and (4) the expected value $E_{g_{\lambda}}\{[|\xi - E_{g_{\lambda}}(\xi)|]^k\}$ is called the kth absolute central moment.

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

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

Proof. From Lemma 3.1, we infer

$$E_{g_{\lambda}}(\xi^{k}) = \int_{0}^{+\infty} g_{\lambda} \{\xi^{k} \geq x\} dx + \lambda \int_{0}^{+\infty} g_{\lambda} \{\xi^{k} \geq x\} \cdot g_{\lambda} \{\xi^{k} \leq x\} dx$$

$$= \int_{0}^{+\infty} g_{\lambda} \{\xi \geq r\} dr^{k} + \lambda \int_{0}^{+\infty} g_{\lambda} \{\xi \geq r\} \cdot g_{\lambda} \{\xi \leq r\} dr^{k}$$

$$= k \int_{0}^{+\infty} r^{k-1} g_{\lambda} \{\xi \geq r\} dr + k\lambda \int_{0}^{+\infty} r^{k-1} g_{\lambda} \{\xi \geq r\} \cdot g_{\lambda} \{\xi \leq r\} dr.$$

$$\square$$

$$(3.8)$$

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

$$E_{g_{\lambda}}[f(\xi)] \le \frac{n - E_{g_{\lambda}}(\xi)}{n - m} f(m) + \frac{E_{g_{\lambda}}(\xi) - m}{n - m} f(n). \tag{3.9}$$

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

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

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 $\lambda \geq 0$, we have

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

$$\ge \int_{0}^{+\infty} g_{\lambda}\{|\xi|^{t} \ge r\} dr. \tag{3.12}$$

Since $E_{g_{\lambda}}(|\xi|^t) < \infty$, we obtain $\int_0^{+\infty} g_{\lambda}\{|\xi|^t \ge r\} dr < \infty$. Consequently,

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

Since

$$\int_{x^{t}/2}^{\infty} g_{\lambda} \left\{ |\xi|^{t} \ge r \right\} dr \ge \int_{x^{t}/2}^{x^{t}} g_{\lambda} \left\{ |\xi|^{t} \ge r \right\} dr \ge \frac{1}{2} x^{t} g_{\lambda} \{ |\xi| \ge x \}, \tag{3.14}$$

we have

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

(2) When λ < 0, we have

$$E_{g_{\lambda}}(|\xi|^{t}) = \int_{0}^{+\infty} g_{\lambda} \{|\xi|^{t} \geq r\} dr + \lambda \int_{0}^{+\infty} g_{\lambda} \{|\xi|^{t} \geq r\} \cdot g_{\lambda} \{|\xi|^{t} \leq r\} dr$$

$$\geq \int_{0}^{+\infty} g_{\lambda} \{|\xi|^{t} \geq r\} dr + \lambda \int_{0}^{+\infty} g_{\lambda} \{|\xi|^{t} \geq r\} dr$$

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

$$(3.16)$$

Since

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

we obtain

$$(1+\lambda)\int_0^{+\infty} g_\lambda \left\{ |\xi|^t \ge r \right\} dr < \infty. \tag{3.18}$$

Consequently,

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

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Since

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

we have

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

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

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

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

$$= (1 + \lambda) \int_{0}^{+\infty} g_{\lambda}\{|\xi|^{s} \geq r\} dr$$

$$= (1 + \lambda) \left(\int_{0}^{l} g_{\lambda}\{|\xi|^{s} \geq r\} dr + \int_{l}^{+\infty} g_{\lambda}\{|\xi|^{s} \geq r\} dr \right)$$

$$= (1 + \lambda) \left(\int_{0}^{l} g_{\lambda}\{|\xi|^{s} \geq r\} dr + \int_{l}^{+\infty} sr^{s-1} g_{\lambda}\{|\xi| \geq r\} dr \right)$$

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

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

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

$$E_{g_{\lambda}}(|\xi|^{s}) \le (1+\lambda) \left(\int_{0}^{l} g_{\lambda}\{|\xi|^{s} \ge r\} dr + s \int_{0}^{+\infty} r^{s-t-1} dr \right) < \infty.$$
 (3.23)

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

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

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

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

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

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

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

$$(3.24)$$

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

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

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