

On Complete Moment Convergence for Randomly Weighted Sums of NSD Random Variables

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Abstract. In this paper, we investigate the complete moment convergence and complete convergence for randomly weighted sums of negatively superadditive dependent (NSD, in short) random variables. The results obtained in the paper generalize the convergence theorem for constant weighted sums to randomly weighted sums of dependent random variables. In addition, strong law of large numbers for NSD sequence is obtained.

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1 Introduction

Let $\{X_n, n \geq 1\}$ be a sequence of random variables defined on a fixed probability space (Ω, \mathcal{F}, P) . The concept of negatively superadditive dependent (NSD, in short) random variables was introduced by Hu [1] based on the class of superadditive functions. Superadditive structure functions have important reliability interpretations, which describe whether a system is more series-like or more parallel-like. The concepts of superadditive structure function and NSD random variables were introduced by Kemperman [2] and Hu [1] as follows.

Definition 1.1 ([2]). A function $\phi : R^n \rightarrow R$ is called superadditive if

$$\phi(\mathbf{x} \vee \mathbf{y}) + \phi(\mathbf{x} \wedge \mathbf{y}) \geq \phi(\mathbf{x}) + \phi(\mathbf{y}) \text{ for all } \mathbf{x}, \mathbf{y} \in R^n,$$

where \vee is for componentwise maximum and \wedge is for componentwise minimum.

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Definition 1.2 ([1]). A random vector $\mathbf{X} = (X_1, \dots, X_n)$ is said to be negatively superadditive dependent (NSD, in short) if

$$E\phi(X_1, \dots, X_n) \leq E\phi(X_1^*, \dots, X_n^*), \quad (1.1)$$

where X_1^*, \dots, X_n^* are independent such that X_i^* and X_i have the same distribution for each i and ϕ is a superadditive function such that the expectations in (1.1) exist.

Definition 1.3. A sequence of random variables $\{X_n, n \geq 1\}$ is said to be negatively superadditive dependent if for all $n \geq 1$, (X_1, \dots, X_n) is negatively superadditive dependent.

Hu [1] gave an example illustrating that NSD random variables are not necessarily negatively associated (NA, in short). Christofides and Vaggelatou [3] indicated that NA random variables are NSD. Negatively superadditive dependent structure is an extension of negatively associated structure and sometimes more useful than negatively associated structure. For example, the structure function of a monotone coherent system can be superadditive [4], so inequalities derived from NSD can give one-side or two-side bounds of the system reliability. The notion of NSD random variables has wide applications in multivariate statistical analysis and reliability theory. Eghbal *et al.* [5,6] provided some inequalities and strong law of large numbers of quadratic forms of NSD random variables under some assumptions. Shen *et al.* [7] obtained Khintchine-Kolmogorov-type convergence theorem and strong stability for NSD random variables. Shen *et al.* [8] discussed the Marcinkiewicz-type strong law of large numbers and integrability of supremum for NSD random variables. Wang *et al.* [9] and Wang *et al.* [10] obtained the complete convergence of weighted sums for an array of rowwise NSD random variables. For more details about complete convergence for dependent case, one can refer to Cabrera *et al.* [11], Yang *et al.* [12], Li *et al.* [13] and Wang *et al.* [14].

The main purpose of this paper is to study the complete moment convergence and complete convergence for randomly weighted sums of NSD random variables. As an application, a strong law of large numbers is obtained for NSD structure.

The following concept of stochastic domination will be used in the main results of the paper.

Definition 1.4 ([15]). A sequence of random variables $\{X_n, n \geq 1\}$ is said to be stochastically dominated by random variable X if there exists a positive constant C such that

$$P(|X_n| \geq x) \leq CP(|X| \geq x),$$

for all $x \geq 0$ and all $n \geq 1$.

Throughout this paper, let $I(A)$ be the indicator of the set A and $X^+ = \max\{0, X\}$. C denotes a positive constant which may be different in various places. $a_n = O(b_n)$ represents $a_n \leq Cb_n$ for all $n \geq 1$.

2 Some lemmas

The following lemmas will be needed in this paper.

Lemma 2.1 ([1]). *If (X_1, \dots, X_n) is NSD and f_1, \dots, f_n are all nondecreasing functions, then $(f_1(X_1), \dots, f_n(X_n))$ is NSD.*

Lemma 2.2 ([9]). *Let $p > 1$ and $\{X_n, n \geq 1\}$ be a sequence of NSD random variables with $E|X_n|^p < \infty$ for each $n \geq 1$. Then for all $n \geq 1$,*

$$E \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k X_i \right|^p \right) \leq 2^{3-p} \sum_{i=1}^n E|X_i|^p, \quad \text{for } 1 < p \leq 2,$$

$$E \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k X_i \right|^p \right) \leq 2 \left(\frac{15p}{\ln p} \right)^p \left[\sum_{i=1}^n E|X_i|^p + \left(\sum_{i=1}^n EX_i^2 \right)^{\frac{p}{2}} \right], \quad \text{for } p > 2.$$

Lemma 2.3 ([16]). *Let $\{Y_n, n \geq 1\}$ and $\{Z_n, n \geq 1\}$ be sequences of random variables. Then for any $q > 1, \varepsilon > 0$, and $a > 0$,*

$$E \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k (Y_i + Z_i) \right| - \varepsilon a \right)^+ \\ \leq \left(\frac{1}{\varepsilon^q} + \frac{1}{q-1} \right) \frac{1}{a^{q-1}} \left(E \max_{1 \leq k \leq n} \left| \sum_{i=1}^k Y_i \right|^q \right) + E \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k Z_i \right| \right).$$

Lemma 2.4 ([17]). *Suppose that $\{X_n, n \geq 1\}$ is a sequence of random variables which is stochastically dominated by a random variable X . Then for any $q > 0$ and $x > 0$,*

$$E|X_n|^q I(|X_n| \leq x) \leq C[E|X|^q I(|X| \leq x) + x^q P(|X| > x)],$$

$$E|X_n|^q I(|X_n| > x) \leq CE|X|^q I(|X| > x).$$

3 Main results

Theorem 3.1. *Assume $\alpha > \frac{1}{2}, p \geq 2$. Let $\{X_n, n \geq 1\}$ be a sequence of NSD random variables which has mean zero and is stochastically dominated by a random variable X . Assume that $\{A_{ni}, 1 \leq i \leq n, n \geq 1\}$ is an array of independent random variables satisfying that $\{A_{ni}, 1 \leq i \leq n, n \geq 1\}$ is independent of $\{X_n, n \geq 1\}$. Let $\beta \geq 1$ such that*

$$\sum_{i=1}^n |A_{ni}|^q = O(n^\beta), \tag{3.1}$$

for some $q > \frac{2(\alpha p - 1)}{2\alpha - 1}$. If $E|X|^p < \infty$, then for every $\varepsilon > 0$,

$$\sum_{n=1}^{\infty} n^{\alpha p - \alpha - \beta - 1} E \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k A_{ni} X_i \right| - \varepsilon n^\alpha \right)^+ < \infty. \tag{3.2}$$

Further

$$\sum_{n=1}^{\infty} n^{\alpha p - \beta - 1} P\left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k A_{ni} X_i \right| > \varepsilon n^{\alpha}\right) < \infty. \tag{3.3}$$

Proof. For $n \geq 1$ and $1 \leq i \leq n$, denote

$$\begin{aligned} X_{ni} &= -n^{\alpha} I(X_i < -n^{\alpha}) + X_i I(|X_i| \leq n^{\alpha}) + n^{\alpha} I(X_i > n^{\alpha}), \\ X_{ni}^* &= n^{\alpha} I(X_i < -n^{\alpha}) - n^{\alpha} I(X_i > n^{\alpha}) + X_i I(|X_i| > n^{\alpha}), \end{aligned}$$

and $\tilde{X}_{ni} = X_{ni} - EX_{ni}$. Thus $X_i = \tilde{X}_{ni} + EX_{ni} + X_{ni}^*$ for $1 \leq i \leq n$.

By Lemma 2.3 with $a = n^{\alpha}$, we obtain that

$$\begin{aligned} & \sum_{n=1}^{\infty} n^{\alpha p - \alpha - \beta - 1} E\left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k A_{ni} X_i \right| - \varepsilon n^{\alpha}\right)^+ \\ & \leq C \sum_{n=1}^{\infty} n^{\alpha p - \alpha q - \beta - 1} E\left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k A_{ni} \tilde{X}_{ni} \right|^q\right) + \sum_{n=1}^{\infty} n^{\alpha p - \alpha - \beta - 1} E\left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k A_{ni} X_{ni}^* \right|\right) \\ & \quad + \sum_{n=1}^{\infty} n^{\alpha p - \alpha - \beta - 1} E\left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k A_{ni} E(X_{ni}) \right|\right) \\ & \doteq H_1 + H_2 + H_3. \end{aligned} \tag{3.4}$$

Since $A_{ni} X_i = A_{ni}^+ X_i - A_{ni}^- X_i$, without loss of generality, we can assume that $A_{ni} \geq 0, 1 \leq i \leq n$. In view of Lemma 2.1, it is easy to see that $\{X_{ni}, 1 \leq i \leq n, n \geq 1\}$ is a sequence of NSD random variables. So we have $\{A_{ni} \tilde{X}_{ni}, 1 \leq i \leq n, n \geq 1\}$ is a sequence of NSD random variables with mean zero. By $q > \frac{2(\alpha p - 1)}{2\alpha - 1}$ and $p \geq 2$, it is easy to see that $q \geq 2$. Thus by Lemma 2.2 we can get that

$$\begin{aligned} H_1 & \leq C \sum_{n=1}^{\infty} n^{\alpha p - \alpha q - \beta - 1} \left[\sum_{i=1}^n E|A_{ni} \tilde{X}_{ni}|^q + \left(\sum_{i=1}^n E(A_{ni} \tilde{X}_{ni})^2 \right)^{\frac{q}{2}} \right] \\ & = C \sum_{n=1}^{\infty} n^{\alpha p - \alpha q - \beta - 1} \sum_{i=1}^n E|A_{ni} \tilde{X}_{ni}|^q + C \sum_{n=1}^{\infty} n^{\alpha p - \alpha q - \beta - 1} \left(\sum_{i=1}^n E(A_{ni} \tilde{X}_{ni})^2 \right)^{\frac{q}{2}} \\ & \doteq H_{11} + H_{12}. \end{aligned} \tag{3.5}$$

By (3.1) and Lemma 2.4, it follows that

$$\begin{aligned}
 H_{11} &\leq C \sum_{n=1}^{\infty} n^{\alpha p - \alpha q - \beta - 1} \sum_{i=1}^n E(|A_{ni}|^q |X_{ni}|^q) \\
 &= C \sum_{n=1}^{\infty} n^{\alpha p - \alpha q - \beta - 1} \sum_{i=1}^n E|A_{ni}|^q [E|X_i|^q I(|X_i| \leq n^\alpha) + n^{\alpha q} E I(|X_i| > n^\alpha)] \\
 &\leq C \sum_{n=1}^{\infty} n^{\alpha p - \alpha q - 1} E|X|^q I(|X| \leq n^\alpha) + C \sum_{n=1}^{\infty} n^{\alpha p - 1} E I(|X| > n^\alpha) \\
 &\doteq H_{11}^* + H_{11}^{**}.
 \end{aligned} \tag{3.6}$$

Observe that $q > \frac{2(\alpha p - 1)}{2\alpha - 1}$ and $p \geq 2$, then we have $q > p$. Therefore it has

$$\begin{aligned}
 H_{11}^* &= C \sum_{n=1}^{\infty} n^{\alpha p - \alpha q - 1} \sum_{m=1}^n E|X|^q I((m-1)^\alpha < |X| \leq m^\alpha) \\
 &= C \sum_{m=1}^{\infty} E|X|^q I((m-1)^\alpha < |X| \leq m^\alpha) \sum_{n=m}^{\infty} n^{\alpha p - \alpha q - 1} \leq CE|X|^p < \infty,
 \end{aligned} \tag{3.7}$$

$$\begin{aligned}
 H_{11}^{**} &= C \sum_{n=1}^{\infty} n^{\alpha p - 1} \sum_{m=n}^{\infty} E I(m^\alpha < |X| \leq (m+1)^\alpha) \\
 &= C \sum_{m=1}^{\infty} E I(m^\alpha < |X| \leq (m+1)^\alpha) \sum_{n=1}^m n^{\alpha p - 1} \leq CE|X|^p < \infty.
 \end{aligned} \tag{3.8}$$

Combining (3.1) with Hölder’s inequality and $q > p \geq 2$, one has

$$\begin{aligned}
 \sum_{i=1}^n EA_{ni}^2 &\leq \left(\sum_{i=1}^n E|A_{ni}|^q \right)^{\frac{2}{q}} \left(\sum_{i=1}^n 1 \right)^{1 - \frac{2}{q}} \leq n^{\beta \frac{2}{q} + 1 - \frac{2}{q}}, \\
 EX_{ni}^2 &= E[|X_i|^2 I(|X_i| \leq n^\alpha) + n^{2\alpha} I(|X_i| > n^\alpha)] \\
 &\leq CE[X^2 I(|X| \leq n^\alpha) + X^2 I(|X| > n^\alpha)] \\
 &= CEX^2.
 \end{aligned}$$

Thus, by the fact that $q > \frac{2(\alpha p - 1)}{2\alpha - 1}$ it follows that

$$\begin{aligned}
 H_{12} &\leq C \sum_{n=1}^{\infty} n^{\alpha p - \alpha q - \beta - 1} \left[\sum_{i=1}^n (EA_{ni}^2 EX_{ni}^2) \right]^{\frac{q}{2}} \\
 &\leq C \sum_{n=1}^{\infty} n^{\alpha p - \alpha q - \beta - 1} n^{(\beta \frac{2}{q} + 1 - \frac{2}{q}) \frac{q}{2}} \\
 &\leq C \sum_{n=1}^{\infty} n^{\alpha p - \alpha q + \frac{q}{2} - 2} < \infty.
 \end{aligned} \tag{3.9}$$

For H_2 and H_3 , similar to the proof of (3.7) and (3.8), we have that

$$\begin{aligned}
 H_2 &\leq \sum_{n=1}^{\infty} n^{\alpha p - \alpha - \beta - 1} \sum_{i=1}^n E|A_{ni}|E|X_{ni}^*| \\
 &\leq C \sum_{n=1}^{\infty} n^{\alpha p - \alpha - 1} E|X|I(|X| > n^\alpha) \\
 &= C \sum_{n=1}^{\infty} n^{\alpha p - \alpha - 1} \sum_{m=n}^{\infty} E|X|I(m^\alpha < |X| \leq (m+1)^\alpha) \\
 &\leq C \sum_{m=1}^{\infty} E[|X|I(m^\alpha < |X| \leq (m+1)^\alpha)]m^{\alpha p - \alpha} \\
 &\leq CE|X|^p < \infty,
 \end{aligned} \tag{3.10}$$

$$\begin{aligned}
 H_3 &\leq \sum_{n=1}^{\infty} n^{\alpha p - \alpha - \beta - 1} \sum_{i=1}^n |E(A_{ni}X_{ni})| \\
 &\leq C \sum_{n=1}^{\infty} n^{\alpha p - \alpha - 1} E|X|I(|X| > n^\alpha) \\
 &\leq CE|X|^p < \infty.
 \end{aligned} \tag{3.11}$$

Combining (3.4)-(3.11), we can obtain (3.2) immediately. By (3.2) and [16], it follows that

$$\begin{aligned}
 &\infty > \sum_{n=1}^{\infty} n^{\alpha p - \alpha - \beta - 1} E \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k A_{ni}X_i \right| - \varepsilon n^\alpha \right)^+ \\
 &\geq \sum_{n=1}^{\infty} n^{\alpha p - \alpha - \beta - 1} \int_0^{\varepsilon n^\alpha} P \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k A_{ni}X_i \right| - \varepsilon n^\alpha > t \right) dt \\
 &\geq \varepsilon \sum_{n=1}^{\infty} n^{\alpha p - \beta - 1} P \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k A_{ni}X_i \right| > 2\varepsilon n^\alpha \right),
 \end{aligned} \tag{3.12}$$

which implies (3.3). □

If taking $\alpha p = 1 + \beta$, $\frac{1}{2} < \alpha < \frac{1+\beta}{2}$, then we get the following Corollary.

Corollary 3.1. Let $\alpha p = 1 + \beta$, $\frac{1}{2} < \alpha < \frac{1+\beta}{2}$. Let $\{X_n, n \geq 1\}$ be a sequence of NSD random variables which has mean zero and is stochastically dominated by a random variable X . Assume that $\{A_{ni}, 1 \leq i \leq n, n \geq 1\}$ is an array of independent random variables satisfying that $\{A_{ni}, 1 \leq i \leq n, n \geq 1\}$ is independent of $\{X_n, n \geq 1\}$. Let $\beta \geq 1$ such that

$$\sum_{i=1}^n |A_{ni}|^q = O(n^\beta),$$

for some $q > \frac{2\beta}{2\alpha-1}$. If $E|X|^p < \infty$, then for every $\varepsilon > 0$,

$$\sum_{n=1}^{\infty} n^{-\alpha} E \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k A_{ni} X_i \right| - \varepsilon n^{\alpha} \right)^+ < \infty. \quad (3.13)$$

Further

$$\sum_{n=1}^{\infty} P \left(\max_{1 \leq k \leq n} \left| \sum_{i=1}^k A_{ni} X_i \right| > \varepsilon n^{\alpha} \right) < \infty. \quad (3.14)$$

By Borel-Cantelli lemma and (3.14), the strong law of large numbers can be obtained as follows.

Corollary 3.2. Under the conditions of Corollary 3.1, we have

$$\lim_{n \rightarrow \infty} \frac{1}{n^{\alpha}} \max_{1 \leq k \leq n} \left| \sum_{i=1}^k A_{ni} X_i \right| = 0, \quad a.s. \quad (3.15)$$

Remark 3.1. Theorem 3.1 and Corollary 3.1 present complete moment convergence and complete convergence for randomly weighted sums of NSD random variables, which extend the results of Shen *et al.* ([8]) and Wang *et al.* ([9]). Note that in [8, 9] the complete moment convergence for constant weighted sums of NSD random variables were obtained.

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