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Complete convergence for Sung's type weighted sums of END random variables

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Abstract

In this paper, the author studies the complete convergence results for Sung's type weighted sums of sequences of END random variables and obtains some new results. These results extend and improve the corresponding theorems of Sung (*Discrete Dyn. Nat. Soc.* 2010:630608, 2010, doi:10.1155/2010/630608).

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1 Introduction and main results

The concept of complete convergence was introduced by Hsu and Robbins [1] as follows. A sequence of random variables $\{X_n, n \geq 1\}$ is said to converge completely to a constant c if

$$\sum_{n=1}^{\infty} P(|X_n - c| > \varepsilon) < \infty, \quad \forall \varepsilon > 0.$$

From then on, many authors have devoted their study to complete convergence; see [2–6], and so on.

Recently, Sung [5] obtained a complete convergence result for weighted sums of identically distributed ρ^* -mixing random variables (we call these Sung's type weighted sums).

Theorem A *Let $p > 1/\alpha$ and $1/2 < \alpha \leq 1$. Let $\{X, X_n, n \geq 1\}$ be a sequence of identically distributed ρ^* -mixing random variables with $EX = 0$ and $E|X|^p < \infty$. Assume that $\{a_{ni}, 1 \leq i \leq n, n \geq 1\}$ is an array of real numbers satisfying*

$$\sum_{i=1}^n |a_{ni}|^q = O(n) \tag{1.1}$$

for some $q > p$. Then

$$\sum_{n=1}^{\infty} n^{p\alpha-2} P\left(\max_{1 \leq j \leq n} \left| \sum_{i=1}^j a_{ni} X_i \right| > \varepsilon n^\alpha\right) < \infty, \quad \forall \varepsilon > 0. \tag{1.2}$$

Conversely, (1.2) implies $EX = 0$ and $E|X|^p < \infty$ if (1.2) holds for any array $\{a_{ni}, 1 \leq i \leq n, n \geq 1\}$ with (1.1) for some $q > p$.

In this paper, we will extend Theorem A under the END setup. We firstly introduce the concept of END random variables.

Definition 1.1 Random variables Y_1, Y_2, \dots are said to be *extended negatively dependent* (END) if there exists a constant $M > 0$ such that, for each $n \geq 2$,

$$P(Y_1 \leq y_1, \dots, Y_n \leq y_n) \leq M \prod_{i=1}^n P(Y_i \leq y_i)$$

and

$$P(Y_1 > y_1, \dots, Y_n > y_n) \leq M \prod_{i=1}^n P(Y_i > y_i)$$

hold for every sequence $\{y_1, \dots, y_n\}$ of real numbers.

The concept was introduced by Liu [7]. When $M = 1$, the notion of END random variables reduces to the well-known notion of so-called *negatively dependent* (ND) random variables, which was firstly introduced by Embraimi and Ghosh [8]; some properties and limit results can be found in Alam and Saxena [9], Block *et al.* [10], Joag-Dev and Proschan [11], and Wu and Zhu [12]. As is mentioned in Liu [7], the END structure is substantially more comprehensive than the ND structure in that it can reflect not only a negative dependence structure but also a positive one, to some extent. Liu [7] pointed out that the END random variables can be taken as negatively or positively dependent and provided some interesting examples to support this idea. Joag-Dev and Proschan [11] also pointed out that negatively associated (NA) random variables must be ND and ND is not necessarily NA, thus NA random variables are END. A great number of articles for NA random variables have appeared in the literature. But very few papers are written for END random variables. For example, for END random variables with heavy tails Liu [7] obtained the precise large deviations and Liu [13] studied sufficient and necessary conditions for moderate deviations, and Qiu *et al.* [3] and Wu and Guan [14] studied complete convergence for weighted sums and arrays of rowwise END, and so on.

Now we state the main results; some lemmas and the proofs will be detailed in the next section.

Theorem 1.1 Let $p > 1/\alpha$ and $1/2 < \alpha \leq 1$. Let $\{X, X_n, n \geq 1\}$ be a sequence of identically distributed END random variables with $EX = 0$ and $E|X|^p < \infty$. Assume that $\{a_{ni}, 1 \leq i \leq n, n \geq 1\}$ is an array of real numbers satisfying (1.1) for some $q > p$. Then (1.2) holds. Conversely, (1.2) implies $EX = 0$ and $E|X|^p < \infty$ if (1.2) holds for any array $\{a_{ni}, 1 \leq i \leq n, n \geq 1\}$ with (1.1) for some $q > p$.

Remark 1.1 The tool is the maximal Rosenthal's moment inequality in the proof of Theorem A. But we do not know whether the maximal Rosenthal's moment inequality holds or not for an END sequence. So the proof of Theorem 1.1 is different from that of Theorem A.

Remark 1.2 Theorem 1.1 does not discuss the very interesting case: $p\alpha = 1$. In fact, it is still an open problem whether (1.2) holds or not even in the partial sums of an END sequence when $p\alpha = 1$. But we have the following partial result.

Theorem 1.2 Let $\{X, X_n, n \geq 1\}$ be a sequence of identically distributed END random variables with $EX = 0$. Assume that $\{a_{ni}, 1 \leq i \leq n, n \geq 1\}$ is an array of real numbers satisfying (1.1) for some $q > 1$. Then

$$\sum_{n=1}^{\infty} n^{-1} P\left(\max_{1 \leq j \leq n} \left| \sum_{i=1}^j a_{ni} X_i \right| > \varepsilon n\right) < \infty, \quad \forall \varepsilon > 0. \tag{1.3}$$

Conversely, (1.3) implies $EX = 0$ if (1.3) holds for any array $\{a_{ni}, 1 \leq i \leq n, n \geq 1\}$ with (1.1) for some $q > 1$.

Throughout this paper, C always stands for a positive constant which may differ from one place to another.

2 Lemmas and proofs of main results

To prove the main result, we need the following lemmas.

Lemma 2.1 ([7]) Let X_1, X_2, \dots, X_n be END random variables. Assume that f_1, f_2, \dots, f_n are Borel functions all of which are monotone increasing (or all are monotone decreasing). Then $f_1(X_1), f_2(X_2), \dots, f_n(X_n)$ are END random variables.

The following lemma is due to Chen *et al.* [15] when $1 < r < 2$ and Shen [16] when $r \geq 2$.

Lemma 2.2 For any $r > 1$, there is a positive constant C_r depending only on r such that if $\{X_n, n \geq 1\}$ is a sequence of END random variables with $EX_n = 0$ for every $n \geq 1$, then, for all $n \geq 1$,

$$E \left| \sum_{i=1}^n X_i \right|^r \leq C_r \sum_{i=1}^n E|X_i|^r$$

holds when $1 < r < 2$ and

$$E \left| \sum_{i=1}^n X_i \right|^r \leq C_r \left\{ \sum_{i=1}^n E|X_i|^r + \left(\sum_{i=1}^n E|X_i|^2 \right)^{r/2} \right\}$$

holds when $r \geq 2$.

By Lemma 2.2 and the same argument as Theorem 2.3.1 in Stout [17], the following lemma holds.

Lemma 2.3 For any $r > 1$, there is a positive constant C_r depending only on r such that if $\{X_n, n \geq 1\}$ is a sequence of END random variables with $EX_n = 0$ for every $n \geq 1$, then, for all $n \geq 1$,

$$E \max_{1 \leq k \leq n} \left| \sum_{i=1}^k X_i \right|^r \leq C_r (\log n)^r \sum_{i=1}^n E|X_i|^r$$

holds when $1 < r < 2$ and

$$E \max_{1 \leq k \leq n} \left| \sum_{i=1}^k X_i \right|^r \leq C_r (\log n)^r \left\{ \sum_{i=1}^n E|X_i|^r + \left(\sum_{i=1}^n E|X_i|^2 \right)^{r/2} \right\}$$

holds when $r \geq 2$.

Lemma 2.4 *Let $p > 1/\alpha$ and $1/2 < \alpha \leq 1$. Let $\{X, X_n, n \geq 1\}$ be a sequence of identically distributed END random variables with $EX = 0$ and $E|X|^p < \infty$. Assume that $\{a_{ni}, 1 \leq i \leq n, n \geq 1\}$ is an array of real numbers satisfying $|a_{ni}| \leq 1$ for $1 \leq i \leq n$ and $n \geq 1$. Then (1.2) holds.*

Proof Without loss of generality, we can assume that

$$a_{ni} \geq 0, \quad \forall 1 \leq i \leq n, n \geq 1, \tag{2.1}$$

from which it follows that

$$\sum_{i=1}^n a_{ni}^\tau \leq n, \quad \forall \tau \geq 1. \tag{2.2}$$

Since $p > 1/\alpha$ and $1/2 < \alpha \leq 1$, we have $0 \leq (1 - \alpha)/(p\alpha - \alpha) < 1$. We take t as given such that $(1 - \alpha)/(p\alpha - \alpha) < t < 1$.

For $1 \leq i \leq n, n \geq 1$, set

$$\begin{aligned} X_{ni}^{(1)} &= -n^{t\alpha} I(X_i < -n^{t\alpha}) + X_i I(|X_i| \leq n^{t\alpha}) + n^{t\alpha} I(X_i > n^{t\alpha}), \\ X_{ni}^{(2)} &= (X_i - n^{t\alpha}) I(n^{t\alpha} < X_i \leq n^\alpha) + n^\alpha I(X_i > n^\alpha), \\ X_{ni}^{(3)} &= (X_i - n^{t\alpha} - n^\alpha) I(X_i > n^\alpha), \\ X_{ni}^{(4)} &= (X_i + n^{t\alpha}) I(-n^\alpha \leq X_i < -n^{t\alpha}) - n^\alpha I(X_i < -n^\alpha), \\ X_{ni}^{(5)} &= (X_i + n^{t\alpha} + n^\alpha) I(X_i < -n^\alpha). \end{aligned}$$

Therefore

$$\begin{aligned} & \sum_{n=1}^{\infty} n^{p\alpha-2} P \left(\max_{1 \leq j \leq n} \left| \sum_{i=1}^j a_{ni} X_i \right| > \varepsilon n^\alpha \right) \\ & \leq \sum_{l=1}^5 \sum_{n=1}^{\infty} n^{p\alpha-2} P \left(\max_{1 \leq j \leq n} \left| \sum_{i=1}^j a_{ni} X_{ni}^{(l)} \right| > \varepsilon n^\alpha / 5 \right) := \sum_{l=1}^5 I_l. \end{aligned}$$

For I_3 ,

$$\begin{aligned} I_3 & \leq \sum_{n=1}^{\infty} n^{p\alpha-2} P \left(\bigcup_{i=1}^n (X_{ni}^{(3)} \neq 0) \right) \leq \sum_{n=1}^{\infty} n^{p\alpha-2} \sum_{i=1}^n P(|X_i| > n^\alpha) \\ & = \sum_{n=1}^{\infty} n^{p\alpha-1} P(|X| > n^\alpha) \leq CE|X|^p < \infty. \end{aligned} \tag{2.3}$$

By the same argument as (2.3), we also have $I_5 < \infty$.

For I_1 , by $EX = 0$, Markov's inequality, (2.1), (2.2), and $(1 - \alpha)/(p\alpha - \alpha) < t < 1$,

$$\begin{aligned} n^{-\alpha} \max_{1 \leq j \leq n} \left| \sum_{i=1}^j a_{ni} EX_{ni}^{(1)} \right| &\leq 2n^{-\alpha} \sum_{i=1}^n a_{ni} E|X_i| I(|X_i| > n^{t\alpha}) \\ &\leq 2n^{-\alpha} E|X| I(|X| > n^{t\alpha}) \sum_{i=1}^n a_{ni} \\ &\leq 2n^{1-\alpha-(p\alpha-\alpha)t} E|X|^p I(|X| > n^{t\alpha}) \rightarrow 0, \quad n \rightarrow \infty. \end{aligned} \tag{2.4}$$

Hence, to prove $I_1 < \infty$, it is enough to show that

$$I_1^* = \sum_{n=1}^{\infty} n^{p\alpha-2} P\left(\max_{1 \leq j \leq n} \left| \sum_{i=1}^j a_{ni} (X_{ni}^{(1)} - EX_{ni}^{(1)}) \right| > \varepsilon n^\alpha / 10 \right) < \infty.$$

By the Markov inequality, Lemma 2.1, Lemma 2.3, the C_r -inequality, (2.1), and (2.2), for any $r \geq 2$,

$$\begin{aligned} I_1^* &\leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2} E \max_{1 \leq j \leq n} \left| \sum_{i=1}^j a_{ni} (X_{ni}^{(1)} - EX_{ni}^{(1)}) \right|^r \\ &\leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2} (\log n)^r \left\{ \sum_{i=1}^n a_{ni}^r E|X_{ni}^{(1)}|^r + \left(\sum_{i=1}^n a_{ni}^2 E|X_{ni}^{(1)}|^2 \right)^{r/2} \right\} \\ &\leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-1} (\log n)^r E|X_{n1}^{(1)}|^r + C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2+r/2} (\log n)^r (E|X_{n1}^{(1)}|^2)^{r/2} \\ &:= CI_{11}^* + CI_{12}^*. \end{aligned} \tag{2.5}$$

If $p \geq 2$, then $(p\alpha - 1)/(\alpha - 1/2) \geq p$. Taking r such that $r > (p\alpha - 1)/(\alpha - 1/2)$,

$$I_{12}^* \leq \sum_{n=1}^{\infty} n^{(p-r)\alpha-2+r/2} (\log n)^r (E|X|^2)^{r/2} < \infty.$$

Since $r > p$ and $t < 1$, by the C_r -inequality, we get

$$\begin{aligned} I_{11}^* &\leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-1} (\log n)^r \{ E|X|^r I(|X| \leq n^{t\alpha}) + n^{tr\alpha} P(|X| > n^{t\alpha}) \} \\ &\leq C \sum_{n=1}^{\infty} n^{(p-r)(1-t)\alpha-1} (\log n)^r E|X|^p < \infty. \end{aligned} \tag{2.6}$$

If $p < 2$, then we can take $r = 2$, in this case $I_{11}^* = I_{12}^*$ in (2.5). Since $r > p$ and $t < 1$, (2.6) still holds. Therefore $I_1 < \infty$.

For I_2 , note that $I_2 = \sum_{n=1}^{\infty} n^{p\alpha-2} P(\sum_{i=1}^n a_{ni} X_{ni}^{(2)} > \varepsilon n^\alpha / 5)$, by (2.1) and (2.2),

$$\begin{aligned} 0 &\leq n^{-\alpha} \sum_{i=1}^n E(a_{ni} X_{ni}^{(2)}) \leq n^{1-\alpha} \{ EX I(n^{t\alpha} < X \leq n^\alpha) + n^\alpha P(X > n^\alpha) \} \\ &\leq n^{1-\alpha} E|X| I(|X| > n^{t\alpha}) + nP(|X| > n^\alpha) \end{aligned}$$

$$\begin{aligned} &\leq n^{1-\alpha-(p\alpha-\alpha)t} E|X|^p I(|X| > n^{t\alpha}) + n^{1-p\alpha} E|X|^p I(|X| > n^\alpha) \\ &\rightarrow 0, \quad n \rightarrow \infty. \end{aligned} \tag{2.7}$$

Hence, in order to prove $I_2 < \infty$, it is enough to show that

$$I_2^* = \sum_{n=1}^{\infty} n^{p\alpha-2} P\left(\sum_{i=1}^n a_{ni}(X_{ni}^{(2)} - EX_{ni}^{(2)}) > \varepsilon n^\alpha / 10\right) < \infty.$$

By the Markov inequality, Lemma 2.1, Lemma 2.2, the C_r -inequality, (2.1), and (2.2), we have, for any $r \geq 2$,

$$\begin{aligned} I_2^* &\leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2} \left\{ \sum_{i=1}^n a_{ni}^r E|X_{ni}^{(2)}|^r + \left(\sum_{i=1}^n a_{ni}^2 E|X_{ni}^{(2)}|^2 \right)^{r/2} \right\} \\ &\leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-1} E|X_{n1}^{(2)}|^r + C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2+r/2} (E|X_{n1}^{(2)}|^2)^{r/2} \\ &:= CI_{21}^* + CI_{22}^*. \end{aligned} \tag{2.8}$$

If $p \geq 2$, we take r such that $r > (p\alpha - 1)/(\alpha - 1/2)$. It follows that

$$I_{22}^* \leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2+r/2} (E|X|^2)^{r/2} < \infty.$$

Since $r > p$, we get by (2.9) of Sung [4]

$$I_{21}^* \leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-1} E|X|^r I(|X| \leq n^\alpha) + C \sum_{n=1}^{\infty} n^{p\alpha-1} P(|X| > n^\alpha) \leq CE|X|^p < \infty. \tag{2.9}$$

If $p < 2$, then we take $r = 2$, in this case $I_{21}^* = I_{22}^*$. Since $r > p$, (2.9) still holds. Therefore, $I_2 < \infty$. Similar to the proof of $I_2 < \infty$, we also have $I_4 < \infty$. Thus, (1.2) holds. \square

Lemma 2.5 *Let $p > 1/\alpha$ and $1/2 < \alpha \leq 1$. Let $\{X_n, n \geq 1\}$ be a sequence of identically distributed END random variables with $EX = 0$. Assume that $\{a_{ni}, 1 \leq i \leq n, n \geq 1\}$ is an array of real numbers satisfying (1.1) for some $q > p$ and $a_{ni} = 0$ or $|a_{ni}| > 1$. Then (1.2) holds.*

Proof Let t be as in Lemma 2.4. Without loss of generality, we may assume that $a_{ni} \geq 0$ and $\sum_{i=1}^n a_{ni}^q \leq n$ for some $q > p$, thus, we have

$$\sum_{i=1}^n a_{ni}^\tau \leq n, \quad \forall 0 < \tau \leq q. \tag{2.10}$$

Similar to the proof of Lemma 2.4 of Sung [5], we may assume that (2.10) holds for some $p < q < 2$ when $p < 2$. For $1 \leq i \leq n, n \geq 1$, set

$$\begin{aligned} X_{ni}^{(1)} &= -n^{t\alpha} I(a_{ni}X_i < -n^{t\alpha}) + a_{ni}X_i I(|a_{ni}X_i| \leq n^{t\alpha}) + n^{t\alpha} I(a_{ni}X_i > n^{t\alpha}), \\ X_{ni}^{(2)} &= (a_{ni}X_i - n^{t\alpha}) I(n^{t\alpha} < a_{ni}X_i \leq n^\alpha) + n^\alpha I(a_{ni}X_i > n^\alpha), \end{aligned}$$

$$\begin{aligned} X_{ni}^{(3)} &= (a_{ni}X_i - n^{t\alpha} - n^\alpha)I(a_{ni}X_i > n^\alpha), \\ X_{ni}^{(4)} &= (a_{ni}X_i + n^{t\alpha})I(-n^\alpha \leq a_{ni}X_i < -n^{t\alpha}) - n^\alpha I(a_{ni}X_i < -n^\alpha), \\ X_{ni}^{(5)} &= (a_{ni}X_i + n^{t\alpha} + n^\alpha)I(a_{ni}X_i < -n^\alpha). \end{aligned}$$

Therefore

$$\begin{aligned} &\sum_{n=1}^{\infty} n^{p\alpha-2} P\left(\max_{1 \leq j \leq n} \left| \sum_{i=1}^j a_{ni}X_i \right| > \varepsilon n^\alpha\right) \\ &\leq \sum_{l=1}^5 \sum_{n=1}^{\infty} n^{p\alpha-2} P\left(\max_{1 \leq j \leq n} \left| \sum_{i=1}^j X_{ni}^{(l)} \right| > \varepsilon n^\alpha / 5\right) := \sum_{l=1}^5 I_l. \end{aligned}$$

By the proof of Lemma 2.4 in Sung [5], we have

$$\begin{aligned} I_3 &\leq \sum_{n=1}^{\infty} n^{p\alpha-2} P\left(\bigcup_{i=1}^n (X_{ni}^{(3)} \neq 0)\right) \\ &\leq \sum_{n=1}^{\infty} n^{p\alpha-2} \sum_{i=1}^n P(|a_{ni}X_i| > n^\alpha) \leq CE|X|^p < \infty. \end{aligned} \tag{2.11}$$

Similarly, we have $I_5 < \infty$.

For I_1 , since $EX_i = 0$, $p > 1/\alpha$, $1/2 < \alpha \leq 1$, $(1 - \alpha)/(p\alpha - \alpha) < t < 1$ and (2.10), we get

$$\begin{aligned} n^{-\alpha} \max_{1 \leq j \leq n} \left| \sum_{i=1}^j X_{ni}^{(1)} \right| &\leq 2n^{-\alpha} \sum_{i=1}^n E|a_{ni}X_i| I(|a_{ni}X_i| > n^{t\alpha}) \\ &\leq 2n^{-\alpha-(p-1)t\alpha} \sum_{i=1}^n a_{ni}^p E|X_i|^p I(|a_{ni}X_i| > n^{t\alpha}) \\ &\leq Cn^{-\alpha-(p-1)t\alpha} \sum_{i=1}^n a_{ni}^p \leq Cn^{1-\alpha-(p-1)t\alpha} \rightarrow 0, \quad n \rightarrow \infty. \end{aligned} \tag{2.12}$$

Hence, in order to prove $I_1 < \infty$, it is enough to show that

$$I_1^* = \sum_{n=1}^{\infty} n^{p\alpha-2} P\left(\max_{1 \leq j \leq n} \left| \sum_{i=1}^j (X_{ni}^{(1)} - EX_{ni}^{(1)}) \right| > \varepsilon n^\alpha / 10\right) < \infty.$$

Similar to the proof of (2.5), we have, for any $r \geq 2$,

$$\begin{aligned} I_1^* &\leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2} (\log n)^r \sum_{i=1}^n E|X_{ni}^{(1)}|^r + C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2} (\log n)^r \left(\sum_{i=1}^n E|X_{ni}^{(1)}|^2\right)^{r/2} \\ &:= CI_{11}^* + CI_{12}^*. \end{aligned} \tag{2.13}$$

If $p \geq 2$, we take r such that $r > (p\alpha - 1)/(\alpha - 1/2)$. By (2.10)

$$I_{12}^* \leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2} (\log n)^r \left(\sum_{i=1}^n a_{ni}^2 E|X_i|^2\right)^{r/2} \leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2+r/2} (\log n)^r < \infty.$$

Since $r > p$ and $0 < t < 1$, we get by (2.10)

$$\begin{aligned}
 I_{11}^* &\leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2} (\log n)^r \left\{ \sum_{i=1}^n (E|a_{ni}X_i|^r I(|a_{ni}X_i| \leq n^{t\alpha}) + n^{tr\alpha} P(|a_{ni}X_i| > n^{t\alpha})) \right\} \\
 &\leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2} (\log n)^r \left\{ \sum_{i=1}^n n^{(r-p)t\alpha} a_{ni}^p E|X_i|^p \right\} \\
 &\leq C \sum_{n=1}^{\infty} n^{(p-r)(1-t)\alpha-1} (\log n)^r E|X|^p < \infty.
 \end{aligned} \tag{2.14}$$

If $p < 2$, then we take $r = 2$, in this case $I_{11}^* = I_{12}^*$ in (2.13). Since $r > p$ and $t < 1$, (2.14) still holds. Therefore $I_1 < \infty$.

For I_2 , since $(1 - \alpha)/(p\alpha - \alpha) < t < 1$, we have by (2.10)

$$\begin{aligned}
 0 &\leq n^{-\alpha} \sum_{i=1}^n E(X_{ni}^{(2)}) \leq n^{-\alpha} \sum_{i=1}^n \{Ea_{ni}X_i I(n^{t\alpha} < a_{ni}X_i \leq n^\alpha) + n^\alpha P(a_{ni}X_i > n^\alpha)\} \\
 &\leq \sum_{i=1}^n \{n^{-\alpha} Ea_{ni}X_i I(a_{ni}X_i > n^{t\alpha}) + P(a_{ni}X_i > n^\alpha)\} \\
 &\leq \sum_{i=1}^n \{n^{-(p-1)t\alpha-\alpha} E|a_{ni}X_i|^p I(|a_{ni}X_i| > n^{t\alpha}) + n^{-p\alpha} E|a_{ni}X_i|^p I(|a_{ni}X_i| > n^\alpha)\} \\
 &\leq C \sum_{i=1}^n a_{ni}^p (n^{-(p-1)t\alpha-\alpha} + n^{-p\alpha}) \\
 &\leq Cn^{1-\alpha-(p-1)t\alpha} + Cn^{1-p\alpha} \rightarrow 0, \quad n \rightarrow \infty.
 \end{aligned} \tag{2.15}$$

Hence, in order to prove $I_2 < \infty$, it is enough to show that

$$I_2^* = \sum_{n=1}^{\infty} n^{p\alpha-2} P\left(\sum_{i=1}^n (X_{ni}^{(2)} - EX_{ni}^{(2)}) > \varepsilon n^\alpha / 10\right) < \infty.$$

Similar to the proof of (2.8), we have for any $r \geq 2$

$$\begin{aligned}
 I_2^* &\leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2} \sum_{i=1}^n E|X_{ni}^{(2)}|^r + C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2} \left(\sum_{i=1}^n E|X_{ni}^{(2)}|^2\right)^{r/2} \\
 &:= CI_{21}^* + CI_{22}^*.
 \end{aligned} \tag{2.16}$$

If $p \geq 2$, we take r such that $r > \{(p\alpha - 1)/(\alpha - 1/2), q\}$. By (2.10), we have

$$\begin{aligned}
 I_{22}^* &\leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2} \left(\sum_{i=1}^n E\{|a_{ni}X_i|^2 I(|a_{ni}X_i| \leq n^\alpha) + n^{2\alpha} P(|a_{ni}X_i| > n^\alpha)\}\right)^{r/2} \\
 &\leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2} \left(\sum_{i=1}^n a_{ni}^2 E|X_i|^2\right)^{r/2} \leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2+r/2} < \infty,
 \end{aligned}$$

and we get by the C_r -inequality and (2.21)-(2.23) of Sung [5] and (2.16)

$$\begin{aligned} I_{21}^* &\leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2} \sum_{i=1}^n E\{(a_{ni}X_i)^r I(n^{t\alpha} < a_{ni}X_i \leq n^\alpha) + n^{r\alpha} I(a_{ni}X_i > n^\alpha)\} \\ &\leq C \sum_{n=1}^{\infty} n^{(p-r)\alpha-2} \sum_{i=1}^n E|a_{ni}X_i|^r I(|a_{ni}X_i| \leq n^\alpha) + C \sum_{n=1}^{\infty} n^{p\alpha-2} \sum_{i=1}^n P(|a_{ni}X_i| > n^\alpha) \\ &\leq CE|X|^p < \infty. \end{aligned} \tag{2.17}$$

If $p < 2$, then we take $r = 2$, in this case $I_{21}^* = I_{22}^*$ in (2.16). Similar to the proof of Lemma 2.4 of Sung [5], (2.17) still holds. Therefore, $I_2 < \infty$. Similar to the proof of $I_2 < \infty$, we have $I_4 < \infty$. Thus, (1.2) holds. \square

Proof of Theorem 1.1 By Lemmas 2.4 and 2.5, the proof is similar to that in Sung [4], so we omit the details. \square

Proof of Theorem 1.2 Sufficiency. Without loss of generality, we can assume that $a_{ni} \geq 0$ and (1.1) holds for $1 < q \leq 2$ by the Hölder inequality. We firstly prove that

$$\sum_{n=1}^{\infty} n^{-1} P\left(\left|\sum_{i=1}^n a_{ni}X_i\right| > \varepsilon n\right) < \infty, \quad \forall \varepsilon > 0. \tag{2.18}$$

For $1 \leq i \leq n, n \geq 1$, set

$$X_{ni}^{(1)} = -nI(X_i < -n) + X_i I(|X_i| \leq n) + nI(X_i > n), \quad X_{ni}^{(2)} = X_i - X_{ni}^{(1)}.$$

Note that $EX = 0$, by the Hölder inequality,

$$n^{-1} \left| E \sum_{i=1}^n a_{ni} X_{ni}^{(1)} \right| \leq CE|X| I(|X| > n) \rightarrow 0.$$

Hence to prove (2.18), it is enough to show that for any $\varepsilon > 0$

$$I_1 = \sum_{n=1}^{\infty} n^{-1} P\left(\left|\sum_{i=1}^n a_{ni} (X_{ni}^{(1)} - EX_{ni}^{(1)})\right| > \varepsilon n\right) < \infty$$

and

$$I_2 = \sum_{n=1}^{\infty} n^{-1} P\left(\left|\sum_{i=1}^n a_{ni} X_{ni}^{(2)}\right| > \varepsilon n\right) < \infty.$$

By the Markov inequality, Lemma 2.3, the C_r -inequality, (1.1), and a standard computation

$$\begin{aligned} I_1 &\leq C \sum_{n=1}^{\infty} n^{-1-q} E \left| \sum_{i=1}^n a_{ni} (X_{ni}^{(1)} - EX_{ni}^{(1)}) \right|^q \\ &\leq C \sum_{n=1}^{\infty} n^{-1-q} \left(\sum_{i=1}^n |a_{ni}|^q \right) \{E|X|^q I(|X| \leq n) + n^q P(|X| > n)\} \end{aligned}$$

$$\begin{aligned} &\leq C \sum_{n=1}^{\infty} n^{-q} E|X|^q I(|X| \leq n) + C \sum_{n=1}^{\infty} P(|X| > n) \\ &\leq CE|X| < \infty. \end{aligned}$$

Obviously,

$$I_2 \leq \sum_{n=1}^{\infty} P(|X| > n) \leq CE|X| < \infty.$$

To prove (1.3), it is enough to prove that

$$\sum_{n=1}^{\infty} n^{-1} P\left(\max_{1 \leq j \leq n} \left| \sum_{i=1}^j a_{ni} (X_i^+ - EX_i^+) \right| > \varepsilon n\right) < \infty, \quad \forall \varepsilon > 0 \tag{2.19}$$

and

$$\sum_{n=1}^{\infty} n^{-1} P\left(\max_{1 \leq j \leq n} \left| \sum_{i=1}^j a_{ni} (X_i^- - EX_i^-) \right| > \varepsilon n\right) < \infty, \quad \forall \varepsilon > 0, \tag{2.20}$$

where $x^+ = \max\{0, x\}$ and $x^- = (-x)^+$.

Let $\varepsilon > 0$ be given. By $EX^+ \leq E|X| < \infty$, (1.1), and the Hölder inequality, there exists a constant $x = x(\varepsilon) > 0$ such that

$$n^{-1} \sum_{i=1}^n a_{ni} E(X_i^+ - x) I(X_i^+ > x) = n^{-1} \left(\sum_{i=1}^n a_{ni} \right) E\{(X^+ - x) I(X^+ > x)\} \leq \varepsilon/6. \tag{2.21}$$

Set

$$X_{i,x}^{(1)} = X_i^+ I(X_i^+ \leq x) + x I(X_i^+ > x), \quad X_{i,x}^{(2)} = X_i^+ - X_{i,x}^{(1)}.$$

Note that by (2.21)

$$\begin{aligned} &\max_{1 \leq j \leq n} n^{-1} \left| \sum_{i=1}^j a_{ni} (X_i^+ - EX_i^+) \right| \\ &\leq \max_{1 \leq j \leq n} n^{-1} \left| \sum_{i=1}^j a_{ni} (X_{i,x}^{(1)} - EX_{i,x}^{(1)}) \right| + \max_{1 \leq j \leq n} n^{-1} \left| \sum_{i=1}^j a_{ni} (X_{i,x}^{(2)} - EX_{i,x}^{(2)}) \right| \\ &\leq \max_{1 \leq j \leq n} n^{-1} \left| \sum_{i=1}^j a_{ni} (X_{i,x}^{(1)} - EX_{i,x}^{(1)}) \right| + n^{-1} \left| \sum_{i=1}^n a_{ni} (X_{i,x}^{(2)} - EX_{i,x}^{(2)}) \right| + \varepsilon/3. \end{aligned}$$

Therefore, to prove (2.19), it is enough to prove that

$$I_3 = \sum_{n=1}^{\infty} n^{-1} P\left(\max_{1 \leq j \leq n} \left| \sum_{i=1}^j a_{ni} (X_{i,x}^{(1)} - EX_{i,x}^{(1)}) \right| > \varepsilon n/3\right) < \infty$$

and

$$I_4 = \sum_{n=1}^{\infty} n^{-1} P \left(\left| \sum_{i=1}^n a_{ni} (X_{i,x}^{(2)} - EX_{i,x}^{(2)}) \right| > \varepsilon n/3 \right) < \infty.$$

By the Markov inequality, Lemma 2.4, and (1.1)

$$I_3 \leq C \sum_{n=1}^{\infty} n^{-1-q} E \max_{1 \leq j \leq n} \left| \sum_{i=1}^j a_{ni} (X_{i,x}^{(1)} - EX_{i,x}^{(1)}) \right|^q \leq \sum_{n=1}^{\infty} n^{-q} (\log n)^q < \infty.$$

By Lemma 2.1, $\{X_{i,x}^{(2)} - EX_{i,x}^{(2)}, i \geq 1\}$ is a sequence of identically distributed END with zero mean. Then $I_4 < \infty$ by taking $\{X_{i,x}^{(2)} - EX_{i,x}^{(2)}, i \geq 1\}$ instead of $\{X_i, i \geq 1\}$ in (2.18). Hence (2.19) holds.

The proof of (2.20) is the same as that of (2.19).

Necessity. It is similar to the proof of Theorem 2.2 in Sung [5]. Here we omit the details. So we complete the proof. \square

Competing interests

The author declares to have no competing interests.

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