

## On complete convergence in Marcinkiewicz-Zygmund type SLLN for random variables

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**Abstract.** We consider a generalization of Baum-Katz theorem for random variables satisfying some cover conditions. Consequently, we get the results for many dependent structures, such as END,  $\varrho^*$ -mixing,  $\varrho^-$ -mixing and  $\varphi$ -mixing, etc.

### §1 Introduction

In this paper, we consider a sequence  $\{X_n, n \geq 1\}$  of random variables defined on some probability space  $(\Omega, \mathcal{F}, P)$ . Hsu and Robbins [10] introduced the following concept of complete convergence. A sequence  $\{X_n, n \geq 1\}$  is said to converge completely to a constant  $C$  if

$$\sum_{n=1}^{\infty} P(|X_n - C| > \varepsilon) < \infty,$$

for all  $\varepsilon > 0$ . For independent and identically distributed (i.i.d., in short) random variables  $\{X, X_n, n \geq 1\}$ , let  $S_n = \sum_{k=1}^n X_k, n \geq 1$  be the partial sums, Hsu and Robbins [10] proved that  $S_n/n$  converge completely to  $EX$ , provided  $DX < \infty$ . Erdős [8] proved the converse theorem. This Hsu-Robbins-Erdős's theorem was generalized in different ways. Katz [13], Baum and Katz [1], and Chow [7] formed the following generalization of Marcinkiewicz-Zygmund type.

**Theorem 1.1.** *Let  $\{X, X_n, n \geq 1\}$  be a sequence of i.i.d. random variables and let  $\alpha p \geq 1, \alpha > 1/2$ . Then the following statements are equivalent:*

- (i)  $E|X|^p < \infty$  and  $EX = 0$  if  $p \geq 1$ ;
- (ii)  $\sum_{n=1}^{\infty} n^{\alpha p - 2} P(|S_n| > \varepsilon n^\alpha) < \infty$  for all  $\varepsilon > 0$ ;
- (iii)  $\sum_{n=1}^{\infty} n^{\alpha p - 2} P(\max_{1 \leq i \leq n} |S_i| > \varepsilon n^\alpha) < \infty$  for all  $\varepsilon > 0$ ;

If  $\alpha p > 1, \alpha > 1/2$  the above are also equivalent to

- (iv)  $\sum_{n=1}^{\infty} n^{\alpha p - 2} P(\sup_{i \geq n} i^{-\alpha} |S_i| > \varepsilon) < \infty$  for all  $\varepsilon > 0$ .

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In many stochastic models, the assumption that random variables are independent is not plausible. Increases in some random variables are often related to decreases in other random variables, so an assumption of dependence is more appropriate than an assumption of independence. One of the important dependence structure is the extended negatively dependent structure, which was introduced by Liu [18] as follows.

**Definition 1.1.** Random variables  $X_k, k = 1, \dots, n$  are said to be lower extended negatively dependent (LEND) if there is some  $M > 0$  such that, for all real numbers  $x_k, k = 1, \dots, n$ ,

$$P \left\{ \bigcap_{k=1}^n (X_k \leq x_k) \right\} \leq M \prod_{k=1}^n P\{X_k \leq x_k\}; \tag{1.1}$$

they are said to be upper extended negatively dependent (UEND) if there is some  $M > 0$  such that, for all real numbers  $x_k, k = 1, \dots, n$ ,

$$P \left\{ \bigcap_{k=1}^n (X_k > x_k) \right\} \leq M \prod_{k=1}^n P\{X_k > x_k\}; \tag{1.2}$$

and they are said to be extended negatively dependent (END) if they are both LEND and UEND. A sequence of infinitely many random variables  $\{X_k, k = 1, 2, \dots\}$  is said to be LEND/UEND/END if there is some  $M > 0$  such that, for each positive integer  $n$ , the random variables  $X_1, X_2, \dots, X_n$  are LEND/UEND/END, respectively.

In the case  $M = 1$ , the formula of END random variables reduces to the notion of negatively orthant dependent (NOD, in short) random variables which was introduced by Joag-Dev and Proschan [12]. They also pointed out that negatively associated (NA, in short) random variables are NOD random variables and then END.

As pointed out in Liu [18], the END structure covers many negative dependence structures and, more interestingly, it covers certain positive dependence structures. Hence, studying the limiting behavior of END random variables is of great significance. There appeared literatures. For example, Liu [18] obtained the precise large deviations for dependent random variables with heavy tails. Shen [25] presented some probability inequalities and gave some applications. Wu et al [33] considered some complete moment convergence and mean convergence theorems for the partial sums of END random variables. Qiu et al [24], Wang et al [31, 32] and Shen et al [27] investigated some results on complete convergence of END random variables. Chen et al [6], Xu and Yan [35], Yan [36-39] considered the SLLN and complete convergence for END random variables, and so on.

Another group of dependencies is formed by mixing type structures defined by special sequences of mixing coefficients. Some of them, however defined in a way that is significantly different from the negative dependence structures, have many similar properties which allow us to use in consideration methods and tools similar to those used in END case. In further consideration we will deal with three types of mixing dependencies:  $\varrho^*$ -mixing,  $\varphi$ -mixing and  $\varrho^-$ -mixing. One can easily find the exact definitions of them in the corresponding references.

Bradley [2] and Miller [19] studied various limit properties of random fields under the assumption  $\varrho^*(k) \rightarrow 0, k \rightarrow \infty$ . We refer to the results obtained under the condition  $\varrho^*(k) < 1$

for some  $k \in \mathbb{N}$  which is important in estimating the moments of partial and maxima of partial sums, see Bryc and Smolenski [3] and Peligrad [20]. Peligrad [20], Utev and Peligrad [29] investigated the properties of the maximum of partial sums and used them to obtain an invariance principle, Peligrad and Gut [21] presented Rosenthal-type maximal inequality and convergence rate for the Marcinkiewicz-Zygmund type SLLN, Cai [4] obtained SLLN and complete convergence for random variables with different distributions.

A concept of  $\varphi$ -mixing dependence was introduced independently by Rozanov and Volkonski [23] and Ibragimov [11]. A number of limit theorems for  $\varphi$ -mixing random variables have been established by many authors. We refer to Wang *et al* [30] (Rosenthal type maximal inequality, Hájek-Rényi type inequality, SLLN), Tuyen [28] (SLLN), and Chen *et al* [5] (complete convergence and Marcinkiewicz-Zygmund type SLLN of moving averages processes) and Kuczmaszewska [16] (complete convergence for NA,  $\varrho^*$ -mixing and  $\varphi$ -mixing sequences satisfying Petrov's condition).

Some results concerning the complete moment convergence, the complete convergence and strong law of large numbers of Marcinkiewicz-Zygmund type for moving average process generated by  $\varrho^-$ -mixing sequences one can find in Zhang [38]. We also refer to Wang and Lu [34].

In this paper, we are interested in generalizations of the Baum-Katz result. Kuczmaszewska [14] extended the result to the case of negatively associated (NA, in short) sequence. They got the following result.

**Theorem 1.2.** (Kuczmaszewska [14]). *Let  $\{X_n, n \geq 1\}$  be a sequence of NA random variables and  $X$  be a random variable possibly defined on a different space satisfying the condition*

$$\frac{1}{n} \sum_{k=1}^n P(|X_k| > x) = c \cdot P(|X| > x) \quad (1.3)$$

for all  $x > 0$ , all  $n \geq 1$  and some positive constant  $c$ . Let  $\alpha p > 1$  and  $\alpha > 1/2$ . Moreover, additionally assume that for  $p \geq 1$   $EX_n = 0$  for all  $n \geq 1$ . Then  $E|X|^p < \infty$  is equivalent to  $\sum_{n=1}^{\infty} n^{\alpha p - 2} P(\max_{1 \leq i \leq n} |S_i| > \varepsilon n^\alpha) < \infty$  for all  $\varepsilon > 0$ .

Though condition (1.3) is weak in some sense, it remains a strong condition, we even get the same result for arbitrary random variables with some rough conditions. Gut [9] introduced the following concept of weakly mean domination.

**Definition 1.2.** We say that the array  $\{X_{nk}, 1 \leq k \leq n, n \geq 1\}$  is weakly mean dominated (WMD, in short) by the random variable  $X$  if, for some  $\gamma > 0$ ,

$$\frac{1}{n} \sum_{k=1}^n P(|X_{nk}| > x) \leq \gamma P(|X| > x), \quad (1.4)$$

for all  $x > 0$  and all  $n$ .

Kuczmaszewska and Lagodowski [15] introduced another structure which can also be used to prove results for non-identically distributed random variables.

**Definition 1.3.** Random variables  $\{X_k, k \geq 1\}$  are weakly mean bounded (WMB, in short) by random variable  $X$  (possibly defined on a different probability space) iff there exist some

constant  $\gamma_1, \gamma_2 > 0$  such that for all  $x > 0$  and  $n \geq 1$

$$\gamma_1 \cdot P(|X| > x) \leq \frac{1}{n} \sum_{k=1}^n P(|X_k| > x) \leq \gamma_2 \cdot P(|X| > x). \tag{1.5}$$

Obviously, if a sequence  $\{X_k, k \geq 1\}$  and a random variable  $X$  satisfy WMB condition, they must satisfy WMD ones.

**Example 1.1.** We give an example of (1.5). Suppose that  $P(X_k = 1 - \frac{1}{k}) = P(X_k = 2 - \frac{1}{k}) = 1/2$  for  $k = 1, 2, \dots$ . Then

$$\frac{1}{n} \sum_{k=1}^n P(|X_k| > x) \rightarrow \begin{cases} 1, & 0 \leq x < 1, \\ 1/2, & 1 \leq x < 2, \\ 0, & x \geq 2. \end{cases}$$

that is, if the mean dominating random variable,  $X$ , is such that  $P(X = 1) = P(X = 2) = 1/2$ , then (1.5) is satisfied.

The aim of this paper is to consider the analogous generalization of the Baum-Katz theorem for a sequence of random variables satisfying WMD or WMB sense and some usual conditions (Marcinkiewicz-Zygmund type inequality and Rosenthal type inequality). The main results and there proofs are presented in Section 2 and Section 3 separately.

As usual, we note that  $C$  will be numerical constants whose values are without importance, and, in addition, may change between appearances.  $I(A)$  is the indicator function on the set  $A$ . Denote  $X^+ = \max(X, 0)$  and  $X^- = \max(-X, 0)$ .

## §2 Main Results

Before presenting our main results, we first give the following assumptions.

**Hypothesis.** Let  $\{X_n, n \geq 1\}$  be an arbitrary sequence of random variables satisfying for every  $\theta \geq 2$  and  $n \geq 1$

$$E \left( \left| \sum_{i=1}^n f_i(X_i) \right| \right)^\theta \leq C \left[ \sum_{i=1}^n E|f_i(X_i)|^\theta + \left( \sum_{i=1}^n E|f_i(X_i)|^2 \right)^{\theta/2} \right] \tag{2.1}$$

and

$$E \left( \max_{1 \leq k \leq n} \left| \sum_{i=1}^k f_i(X_i) \right| \right)^\theta \leq C \log^\theta n \left[ \sum_{i=1}^n E|f_i(X_i)|^\theta + \left( \sum_{i=1}^n E|f_i(X_i)|^2 \right)^{\theta/2} \right], \tag{2.2}$$

whenever  $f_1, f_2, \dots, f_n$  are all nondecreasing (or non-increasing) functions,  $E f_i(X_i) = 0$  and  $E|f_i(X_i)|^\theta < \infty$ , for all  $1 \leq i \leq n$ .

**Remark 2.1.** A lot of dependent structures, for example such as  $\rho^*$ -mixing,  $\varphi$ -mixing, NA, ND, END, etc., satisfy (2.1) and (2.2) in Hypothesis. We list some of them in the following Lemma 2.1, Lemma 2.2 and Lemma 2.3 .

**Lemma 2.1.** (cf.Liu [18]) Let  $\{X_n, n \geq 1\}$  be a sequence of END random variables. For each  $n \geq 1$ , if  $f_1, f_2, \dots, f_n$  are all nondecreasing ( or nonincreasing ) functions, then random variables  $f_1(X_1), f_2(X_2), \dots, f_n(X_n)$  are also END.

**Lemma 2.2.** (cf. Wang et al.[31]) Let  $p \geq 2$  and  $\{X_n, n \geq 1\}$  be a sequence of END random variables with  $EX_n = 0$  and  $E|X_n|^p < \infty$  for each  $n \geq 1$ . Then there exists a positive constant  $C_p$  depending only on  $p$  such that (2.1) and (2.2) hold for  $\theta = p$  and  $f_i(X_i) = X_i$ ,  $i \geq 1$ .

**Lemma 2.3.** (cf. Utev and Peligrad [29], Wang and Lu [34], Wang et al.[30]) Let  $p \geq 2$  and  $\{X_n, n \geq 1\}$  be a sequence of  $\varrho^*$ -mixing,  $\varrho^-$ -mixing or  $\varphi$ -mixing random variables with  $EX_n = 0$  and  $E|X_n|^p < \infty$  for each  $n \geq 1$ . Moreover, if  $X_n, n \geq 1$  are  $\varphi$ -mixing we assume that  $\sum_{n=1}^{\infty} \varphi^{\frac{1}{2}}(n) < \infty$ . Then there exists a positive constant  $C_p$  depending only on  $p$  such that (2.2) holds for  $\theta = p$  and  $f_i(X_i) = X_i$ ,  $i \geq 1$ .

**Theorem 2.1.** Suppose  $\alpha p > 1$ ,  $\alpha > 1/2$ . Let  $\{X_n, n \geq 1\}$  be an arbitrary sequence of random variables with  $EX_n = 0$  for all  $n \geq 1$  if  $p > 1$ , and  $X$  be a random variable possibly defined on a different probability space satisfying (1.4) for all  $x > 0$ , all  $n \geq 1$  and some positive constant  $\gamma$ . Assume that  $\{X_n, n \geq 1\}$  satisfies the conditions of Hypothesis. Then  $E|X|^p < \infty$  implies that for all  $\varepsilon > 0$

$$\sum_{n=1}^{\infty} n^{\alpha p - 2} P\left(\max_{1 \leq k \leq n} |S_k| > \varepsilon n^{\alpha}\right) < \infty, \quad (2.3)$$

and

$$\sum_{n=1}^{\infty} n^{\alpha p - 2} P\left(\sup_{i \geq n} i^{-\alpha} |S_i| > \varepsilon\right) < \infty, \quad (2.4)$$

where  $S_k = \sum_{i=1}^k X_i$ ,  $1 \leq k \leq n$ .

The next theorem presents the necessary condition for (2.3) under assumption that random variables  $\{X_n, n \geq 1\}$  and  $X$  satisfy WMB condition (1.5).

**Theorem 2.2.** Suppose  $\alpha p > 1$ ,  $\alpha > 1/2$ . Let  $\{X_n, n \geq 1\}$  be a sequence of random variables satisfying (2.1) for  $\theta = 2$  and  $X$  be a random variable possibly defined on a different probability space satisfying (1.5) for all  $x > 0$ , all  $n \geq 1$  and some positive constants  $\gamma_1$  and  $\gamma_2$ . Then (2.3) implies  $E|X|^p < \infty$ .

As a consequence of Theorem 2.1 and Theorem 2.2 we get the following result.

**Corollary 2.1.** Suppose  $\alpha p > 1$ ,  $\alpha > 1/2$ . Let  $\{X_n, n \geq 1\}$  be a sequence of END,  $\varrho^*$ -mixing,  $\varrho^-$ -mixing or  $\varphi$ -mixing random variables and  $X$  be a random variable possibly defined on a different probability space satisfying (1.5) for all  $x > 0$ , all  $n \geq 1$  and some positive constants  $\gamma_1$  and  $\gamma_2$ . Moreover, we assume  $EX_n = 0$  for all  $n \geq 1$  if  $p > 1$  and  $\sum_{n=1}^{\infty} \varphi^{\frac{1}{2}}(n) < \infty$  in case of  $\varphi$ -mixing sequence. Then the following statements are equivalent:

- (i)  $E|X|^p < \infty$ ;
- (ii)  $\sum_{n=1}^{\infty} n^{\alpha p - 2} P(\max_{1 \leq i \leq n} |S_i| > \varepsilon n^{\alpha}) < \infty$  for all  $\varepsilon > 0$ .

**Remark 2.2.** Since END and  $\varrho^-$ -mixing random variables include NA random variables, our result also holds for NA case.

### §3 Proofs of the Main Results

In proving the main results, the following fundamental lemma will be used many times.

**Lemma 3.1.** (cf. Gut [9]) *Let  $\{X_n, n \geq 1\}$  be a sequence of random variables satisfying a weak dominating condition with mean dominating random variable  $X$ . Let  $r > 0$  and for some  $A > 0$*

$$\begin{aligned} X'_i &= X_i I(|X_i| \leq A), & X''_i &= X_i I(|X_i| > A), \\ X^*_i &= X_i I(|X_i| \leq A) - AI(X_i < -A) + AI(X_i > A), \end{aligned}$$

and

$$\begin{aligned} X' &= XI(|X| \leq A), & X'' &= XI(|X| > A), \\ X^* &= XI(|X| \leq A) - AI(X < -A) + AI(X > A). \end{aligned}$$

Then

- (i) *if  $E|X|^r < \infty$ , then  $\frac{1}{n} \sum_{i=1}^n E|X_i|^r \leq CE|X|^r$ ;*
- (ii)  *$\frac{1}{n} \sum_{i=1}^n E|X'_i|^r \leq C(E|X'|^r + A^r P(|X| > A))$  for any  $A > 0$ ;*
- (iii)  *$\frac{1}{n} \sum_{i=1}^n E|X''_i|^r \leq CE|X''|^r$  for any  $A > 0$ ;*
- (iv)  *$\frac{1}{n} \sum_{i=1}^n E|X^*_i|^r \leq CE|X^*|^r$  for any  $A > 0$ .*

Now, we present the proofs of the main results step by step.

**Proof of Theorem 2.1.** We first take

$$0 < p' < p, \quad \frac{1}{\alpha p} < q < 1$$

such that

$$\alpha(p - p') > \alpha(p - p')q > 1, \quad \text{and} \quad p - p' > 1 \text{ if } p > 1.$$

For all  $1 \leq i \leq n, n \geq 1$ , denote that

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

Then, for every  $1 \leq i \leq n, n \geq 1$

$$X_i = X_{ni}^{(1)} + X_{ni}^{(2)} - X_{ni}^{(3)} \quad \text{and} \quad X_{ni}^{(2)} \geq 0, \quad X_{ni}^{(3)} \geq 0.$$

Thus,

$$\begin{aligned} & \sum_{n=1}^{\infty} n^{\alpha p - 2} P \left( \max_{1 \leq k \leq n} |S_k| > \varepsilon n^\alpha \right) \leq \sum_{j=1}^3 \sum_{n=1}^{\infty} n^{\alpha p - 2} P \left( \max_{1 \leq k \leq n} \left| \sum_{i=1}^k X_{ni}^{(j)} \right| > \varepsilon n^\alpha / 3 \right) \\ &= \sum_{n=1}^{\infty} n^{\alpha p - 2} P \left( \max_{1 \leq k \leq n} \left| \sum_{i=1}^k X_{ni}^{(1)} \right| > \varepsilon n^\alpha / 3 \right) + \sum_{n=1}^{\infty} n^{\alpha p - 2} P \left( \sum_{i=1}^n X_{ni}^{(2)} > \varepsilon n^\alpha / 3 \right) \\ & \quad + \sum_{n=1}^{\infty} n^{\alpha p - 2} P \left( \sum_{i=1}^n X_{ni}^{(3)} > \varepsilon n^\alpha / 3 \right) \\ &\triangleq I_1 + I_2 + I_3. \end{aligned}$$

To prove (2.3), it suffices to show that  $I_k < \infty, k = 1, 2, 3$ .

For  $I_1$ , we first prove that

$$n^{-\alpha} \max_{1 \leq k \leq n} \left| \sum_{i=1}^k EX_{ni}^{(1)} \right| \rightarrow 0, \quad n \rightarrow \infty. \tag{3.1}$$

We will do it in three cases.

**Case I.** Let  $\alpha \leq 1$ . Then  $\alpha p > 1$  implies  $p > 1$  and, according to the assumption,  $EX_n = 0, n \geq 1$ . It follows from Lemma 3.1 that

$$\begin{aligned} n^{-\alpha} \max_{1 \leq k \leq n} \left| \sum_{i=1}^k EX_{ni}^{(1)} \right| &\leq n^{-\alpha} \max_{1 \leq k \leq n} \left| \sum_{i=1}^k [EX_i I(|X_i| > n^{\alpha q}) + n^{\alpha q} P(|X_i| > n^{\alpha q})] \right| \\ &\leq 2n^{-\alpha} \sum_{i=1}^n E|X_i| I(|X_i| > n^{\alpha q}) \leq Cn^{-[\alpha q(p-p')-1]-\alpha(1-q)} \rightarrow 0, \quad n \rightarrow \infty. \end{aligned}$$

**Case II.** Let  $\alpha > 1$  and  $p > 1$ . We have

$$\begin{aligned} n^{-\alpha} \max_{1 \leq k \leq n} \left| \sum_{i=1}^k EX_{ni}^{(1)} \right| &\leq n^{-\alpha} \sum_{i=1}^n [E|X_i| I(|X_i| \leq n^{\alpha q}) + n^{\alpha q} P(|X_i| > n^{\alpha q})] \\ &\leq Cn^{1-\alpha} [E|X| I(|X| \leq n^{\alpha q}) + n^{\alpha q} P(|X| > n^{\alpha q})] \leq Cn^{1-\alpha} \rightarrow 0, \quad n \rightarrow \infty. \end{aligned}$$

**Case III.** Let  $\alpha > 1$  and  $p \leq 1$ . We have

$$\begin{aligned} n^{-\alpha} \max_{1 \leq k \leq n} \left| \sum_{i=1}^k EX_{ni}^{(1)} \right| &\leq n^{-\alpha} \sum_{i=1}^n [E|X_i| I(|X_i| \leq n^{\alpha q}) + n^{\alpha q} P(|X_i| > n^{\alpha q})] \\ &\leq Cn^{1-\alpha} [E|X| I(|X| \leq n^{\alpha q}) + n^{\alpha q} P(|X| > n^{\alpha q})] \leq Cn^{-[\alpha q(p-p')-1]-\alpha(1-q)} \rightarrow 0, \quad n \rightarrow \infty. \end{aligned}$$

By (3.1), to prove  $I_1 < \infty$ , we prove only that

$$I_1^* \triangleq \sum_{n=1}^{\infty} n^{\alpha p-2} P \left( \max_{1 \leq k \leq n} \left| \sum_{i=1}^k (X_{ni}^{(1)} - EX_{ni}^{(1)}) \right| > \varepsilon n^\alpha / 6 \right) < \infty. \tag{3.2}$$

From Hypothesis, for each  $n \geq 1, \{X_{ni}^{(1)} - EX_{ni}^{(1)}, 1 \leq i \leq n\}$  remain satisfy the inequalities in Hypothesis. By  $\alpha q(p - p') > 1$  and  $0 < q < 1$ , we have for  $p \leq 2$

$$\alpha - \frac{1}{2} - \alpha q \left( 1 - \frac{p-p'}{2} \right) > \alpha - \frac{1}{2} - \alpha \left( 1 - \frac{p-p'}{2} \right) = \frac{\alpha(p-p')-1}{2} > 0.$$

By taking

$$\tau > \max \left\{ 2, p, \frac{p - (p-p')q}{1-q}, \frac{\alpha p - 1}{\alpha - \frac{1}{2}}, \frac{\alpha p - 1}{\alpha - \frac{1}{2} - \alpha q \left( 1 - \frac{p-p'}{2} \right)} \right\},$$

Chebyshev's inequality and Hypothesis we get

$$\begin{aligned} I_1^* &\leq C \sum_{n=1}^{\infty} n^{\alpha p-2-\alpha \tau} E \left( \max_{1 \leq k \leq n} \left| \sum_{i=1}^k (X_{ni}^{(1)} - EX_{ni}^{(1)}) \right| \right)^\tau \\ &\leq C \sum_{n=1}^{\infty} n^{\alpha p-2-\alpha \tau} \log^\tau n \sum_{i=1}^n E |X_{ni}^{(1)}|^\tau + C \sum_{n=1}^{\infty} n^{\alpha p-2-\alpha \tau} \log^\tau n \left( \sum_{i=1}^n E |X_{ni}^{(1)}|^2 \right)^{\tau/2} \\ &\triangleq I_{11}^* + I_{12}^*. \end{aligned}$$

Again by Hypothesis

$$I_{11}^* \leq C \sum_{n=1}^{\infty} n^{-\alpha(1-q)\left(\tau - \frac{p-q(p-p')}{1-q}\right)-1} \log^{\tau} n < \infty,$$

and

$$I_{12}^* \leq C \sum_{n=1}^{\infty} n^{\alpha p-2-\alpha\tau} \log^{\tau} n \left\{ \sum_{i=1}^n [E|X_i|^2 I(|X_i| \leq n^{\alpha q}) + n^{2\alpha q} P(|X_i| > n^{\alpha q})] \right\}^{\tau/2}.$$

Now, we prove  $I_{12}^* < \infty$  in two cases:  $p > 2$  and  $0 < p \leq 2$ .

If  $p > 2$ , then

$$I_{12}^* \leq C \sum_{n=1}^{\infty} n^{\alpha p-2-\alpha\tau} \log^{\tau} n \cdot n^{\tau/2} (EX^2)^{\tau/2} \leq C \sum_{n=1}^{\infty} n^{-(\alpha-\frac{1}{2})\left(\tau - \frac{\alpha p-1}{\alpha-\frac{1}{2}}\right)-1} \log^{\tau} n < \infty.$$

For  $0 < p \leq 2$  we have

$$\begin{aligned} I_{12}^* &\leq C \sum_{n=1}^{\infty} n^{\alpha p-2-\alpha\tau} \log^{\tau} n \cdot \left(n^{\alpha q(2-(p-p'))} \cdot n\right)^{\tau/2} \left(E|X|^{p-p'}\right)^{\tau/2} \\ &\leq C \sum_{n=1}^{\infty} n^{\alpha p-2-\tau\left[\alpha-\frac{1}{2}-\alpha q\left(1-\frac{p-p'}{2}\right)\right]} \log^{\tau} n < \infty. \end{aligned}$$

This ends the proof of  $I_1 < \infty$ . Next for  $I_2 < \infty$  and each  $1 \leq i \leq n, n \geq 1$ , let

$$Y_{ni}^{(2)} = (X_i - n^{\alpha q})I(n^{\alpha q} < X_i \leq n^{\alpha q} + n^{\alpha}) + n^{\alpha}I(X_i > n^{\alpha q} + n^{\alpha}).$$

Then  $Y_{ni}^{(2)} \geq 0$  and

$$X_{ni}^{(2)} = Y_{ni}^{(2)} + (X_i - n^{\alpha q} - n^{\alpha})I(X_i > n^{\alpha q} + n^{\alpha}).$$

Thus,

$$I_2 \leq \sum_{n=1}^{\infty} n^{\alpha p-2} P\left(\sum_{i=1}^n Y_{ni}^{(2)} > \frac{\varepsilon n^{\alpha}}{6}\right) + \sum_{n=1}^{\infty} n^{\alpha p-2} \sum_{i=1}^n P(X_i > n^{\alpha q} + n^{\alpha}) \triangleq I_{21} + I_{22}.$$

By Lemma 3.1,

$$I_{22} \leq \sum_{n=1}^{\infty} n^{\alpha p-2} \sum_{i=1}^n P(|X_i| > n^{\alpha}) \leq C \sum_{n=1}^{\infty} n^{\alpha p-1} P(|X| > n^{\alpha}) \leq CE|X|^p < \infty. \tag{3.3}$$

Next we prove only that  $I_{21} < \infty$ . Similarly, we can get

$$n^{-\alpha} \sum_{i=1}^n EY_{ni}^{(2)} \rightarrow 0, \quad n \rightarrow \infty. \tag{3.4}$$

By (3.4), to prove  $I_{21} < \infty$ , it is sufficient to show that

$$I_{21}^* = \sum_{n=1}^{\infty} n^{\alpha p-2} P\left(\left|\sum_{i=1}^n \left(Y_{ni}^{(2)} - EY_{ni}^{(2)}\right)\right| > \frac{\varepsilon n^{\alpha}}{12}\right) < \infty.$$

We will prove it in two cases:  $0 < p < 2$  and  $p \geq 2$ .

For  $0 < p < 2$ , by Chebyshev's inequality, Hypothesis, Lemma 3.1 and (3.3) we have

$$I_{21}^* \leq C \sum_{n=1}^{\infty} n^{\alpha p-2-2\alpha} \sum_{i=1}^n E|X_i|^2 I(|X_i| \leq 2n^{\alpha}) + C \sum_{n=1}^{\infty} n^{\alpha p-2} \sum_{i=1}^n P(|X_i| > n^{\alpha}) < \infty.$$

Let  $p \geq 2$ . By taking

$$\kappa > \max\left\{p, \frac{\alpha p - 1}{\alpha - \frac{1}{2}}, \frac{2(\alpha p - 1)}{\alpha(p - p') - 1}\right\},$$



Chebyshev's inequality and Hypothesis we have

$$I_{21}^* \leq C \sum_{n=1}^{\infty} n^{\alpha p-2-\alpha \kappa} \sum_{i=1}^n E \left| Y_{ni}^{(2)} \right|^{\kappa} + C \sum_{n=1}^{\infty} n^{\alpha p-2-\alpha \kappa} \left( \sum_{i=1}^n E \left| Y_{ni}^{(2)} \right|^2 \right)^{\kappa/2}$$

$$\triangleq I_{211}^* + I_{212}^*.$$

By the similar method, we can get  $I_{211}^* < \infty, I_{212}^* < \infty$ , so omit them. To prove the second thesis of Theorem 2.1, it is enough to show that (2.3) implies (2.4). For  $0 < \varepsilon < 1$  and  $\alpha p > 1$  we have

$$\begin{aligned} & \sum_{n=1}^{\infty} n^{\alpha p-2} P \left( \sup_{k \geq n} k^{-\alpha} |S_k| > \varepsilon \right) = \sum_{j=1}^{\infty} \sum_{n=2^{j-1}}^{2^j-1} n^{\alpha p-2} P \left( \sup_{k \geq n} k^{-\alpha} |S_k| > \varepsilon \right) \\ & \leq C \sum_{j=1}^{\infty} 2^{j(\alpha p-1)} \sum_{i=j}^{\infty} P \left( \max_{2^{i-1} \leq k < 2^i} k^{-\alpha} |S_k| > \varepsilon \right) \leq C \sum_{i=1}^{\infty} 2^{i(\alpha p-1)} P \left( \max_{2^{i-1} \leq k < 2^i} k^{-\alpha} |S_k| > \varepsilon \right) \\ & \leq C \sum_{i=1}^{\infty} 2^{i(\alpha p-1)} P \left( \max_{1 \leq k \leq 2^i} |S_k| > \varepsilon 2^{-\alpha} 2^{i\alpha} \right) \leq C \sum_{i=1}^{\infty} n^{\alpha p-2} P \left( \max_{1 \leq k \leq n} |S_k| > \varepsilon 2^{-2\alpha} n^{\alpha} \right) < \infty. \end{aligned}$$

This ends the proof of Theorem 2.1. □

**Proof of Theorem 2.2.** In order to prove the result, it suffices to show that

$$\sum_{n=1}^{\infty} n^{\alpha p-1} P(|X| > n^{\alpha}) < \infty. \tag{3.5}$$

For simplicity, we denote  $A = \{\max_{1 \leq i \leq n} |X_i| > n^{\alpha}\}$  and  $S_0 = 0$ . By (2.3), we first note that for  $0 < \varepsilon < 1$

$$\begin{aligned} \sum_{n=1}^{\infty} n^{\alpha p-2} P(A) & \leq \sum_{n=1}^{\infty} n^{\alpha p-2} P \left( \max_{1 \leq i \leq n} |S_i - S_{i-1}| > \varepsilon n^{\alpha} \right) \\ & \leq 2 \sum_{n=1}^{\infty} n^{\alpha p-2} P \left( \max_{1 \leq i \leq n} |S_i| > \varepsilon n^{\alpha} / 2 \right) < \infty. \end{aligned} \tag{3.6}$$

Therefore,

$$P(A) = P \left( \max_{1 \leq i \leq n} |X_i| > n^{\alpha} \right) \rightarrow 0, \quad \text{as } n \rightarrow \infty. \tag{3.7}$$

Moreover, by Lemma 3.1 and (1.5)

$$\begin{aligned} \gamma_1 \cdot n P(|X| > n^{\alpha}) & \leq \sum_{k=1}^n P(|X_k| > n^{\alpha}) \\ & = \sum_{k=1}^n P \left( |X_k| > n^{\alpha}, \max_{1 \leq i < k} |X_i| > n^{\alpha} \right) + \sum_{k=1}^n P \left( |X_k| > n^{\alpha}, \max_{1 \leq i < k} |X_i| \leq n^{\alpha} \right) \\ & = \sum_{k=1}^n P \left( X_k^+ > n^{\alpha}, \max_{1 \leq i < k} |X_i| > n^{\alpha} \right) + \sum_{k=1}^n P \left( X_k^- > n^{\alpha}, \max_{1 \leq i < k} |X_i| > n^{\alpha} \right) + P(A), \end{aligned} \tag{3.8}$$

since the sets  $\{|X_k| > n^{\alpha}, \max_{1 \leq i < k} |X_i| \leq n^{\alpha}\}, 1 \leq k \leq n$  are disjoint. By Hypothesis,  $\{X_n^+, n \geq 1\}$  still satisfy the inequality in Hypothesis. It follows from Cauchy-Schwarz in-

equality and Hypothesis that

$$\begin{aligned}
 & \sum_{k=1}^n P\left(X_k^+ > n^\alpha, \max_{1 \leq i < k} |X_i| > n^\alpha\right) = E \left[ \sum_{k=1}^n I\left(X_k^+ > n^\alpha, \max_{1 \leq i < k} |X_i| > n^\alpha\right) \right] \\
 & \leq E \left[ \sum_{k=1}^n \{I(X_k^+ > n^\alpha) - P(X_k^+ > n^\alpha)\} I(A) \right] + E \left[ \sum_{k=1}^n P(X_k^+ > n^\alpha) I(A) \right] \\
 & \leq \sqrt{C_2 \sum_{k=1}^n E \{I(X_k^+ > n^\alpha) - P(X_k^+ > n^\alpha)\}^2 P(A) + P(A) \sum_{k=1}^n P(X_k^+ > n^\alpha)} \\
 & \leq \sqrt{C_2 \sum_{k=1}^n P(X_k^+ > n^\alpha) P(A) + P(A) \sum_{k=1}^n P(X_k^+ > n^\alpha)} \\
 & \leq \sqrt{C_2 \gamma_2 n P(|X| > n^\alpha) P(A) + \gamma_2 P(A) \cdot n P(|X| > n^\alpha)} \\
 & \leq \frac{\gamma_1 n}{4} P(|X| > n^\alpha) + \frac{\gamma_2 C_2}{\gamma_1} P(A) + \gamma_2 n P(A) P(|X| > n^\alpha), \tag{3.9}
 \end{aligned}$$

where we used the following inequality

$$\sqrt{ab} \leq \frac{a\gamma_1}{4\gamma_2 C_2} + \frac{\gamma_2 C_2}{\gamma_1} b, \quad a \geq 0, b \geq 0.$$

Similarly, we can have  $\sum_{k=1}^n P(X_k^- > n^\alpha, \max_{1 \leq i < k} |X_i| > n^\alpha) \leq \frac{\gamma_1 n}{4} P(|X| > n^\alpha) + \frac{\gamma_2 C_2 P(A)}{\gamma_1} + \gamma_2 n P(A) P(|X| > n^\alpha)$ . Taking into account with (3.8) and (3.9), we have

$$\frac{\gamma_1 n}{2} P(|X| > n^\alpha) \leq \frac{2\gamma_2 C_2 + \gamma_1}{\gamma_1} P(A) + 2\gamma_2 n P(A) P(|X| > n^\alpha).$$

By (3.7), for sufficiently large  $n$  we have

$$P(A) = P\left(\max_{1 \leq i \leq n} |X_i| > n^\alpha\right) < \frac{\gamma_1}{8\gamma_2},$$

and consequently

$$nP(|X| > n^\alpha) \leq \frac{4(2\gamma_2 C_2 + \gamma_1)}{\gamma_1} P\left(\max_{1 \leq i \leq n} |X_i| > n^\alpha\right). \tag{3.10}$$

Relations (3.6) and (3.10) give (3.5) and in conclusion we get the desired condition  $E|X|^p < \infty$ . This ends the proof of Theorem 2.2. □

### References

- [1] L E Baum, M Katz. *Convergence rates in the law of large numbers*, Trans Amer Math Soc, 1965, 120: 108-123.
- [2] R C Bradley. *On the spectral density and asymptotic normality of weakly dependent random fields*, J Theor Probab, 1992, 5(2): 355-373.
- [3] W Bryc, W Smolenski. *Moment conditions for almost sure convergence of weakly correlated random variables*, Proc of Amer Math Soc, 1993, 119(2): 629-635.
- [4] G H Cai. *Strong law of large numbers for  $\varrho^*$ -mixing sequences with different distributions*, Disc Dynamics in Nat Soc, 2006, 2006: 1-7.

- [5] P Chen, T C Hu, A Volodin. *Limit behaviour of moving average processes under  $\phi$ -mixing assumption*, Statist Probab Lett, 2009, 79: 105-111.
- [6] Y Chen, A Chen, K Ng. *The strong law of large numbers for extended negatively dependent random variables*, J Appl Probab, 2010, 47(4): 908-922.
- [7] Y S Chow. *Delayed sums and Borel summability of independent, identically distributed random variables*, Bull Inst Math Acad Sinica, 1973, 1: 207-220.
- [8] P Erdős. *On a theorem of Hsu and Robbins*, Ann Math Stat, 1949, 20: 286-291.
- [9] A Gut. *Complete convergence for arrays*, Periodica Math Hungar, 1992, 25: 51-75.
- [10] P L Hsu, H Robbins. *Complete convergence and the law of large numbers*, Proc Nat Acad Sci USA, 1947, 33: 25-31.
- [11] I A Ibragimov. *Some limit theorems for stationary processes*, Theory Probab Appl, 1962, 7: 349-382.
- [12] K Joag-Dev, F Proschan. *Negative association of random variables with applications*, The Ann Stat, 1983, 11: 286-295.
- [13] M Katz. *The probability in the tail of a distribution*, Ann Math Stat, 1963, 34: 312-318.
- [14] A Kuczmaszewska. *On complete convergence in Marcinkiewicz-Zygmund type SLLN for negatively associated random variables*, Acta Math Hungar, 2010, 128(1-2): 116-130.
- [15] A Kuczmaszewska, Z A Lagodowski. *Convergence rates in the SLLN for some classes of dependent random fields*, J Math Anal Appl, 2011, 380: 571-584.
- [16] A Kuczmaszewska. *Convergence rate in the Petrov SLLN for dependent random variables*, Acta Math Hungar, 2016, 148(1): 56-72.
- [17] Z A Lagodowski. *An approach to complete convergence theorems for dependent random fields via application of Fuk-Nagaev inequality*, J Math Anal Appl, 2016, 437: 380-395.
- [18] L Liu. *Precise large deviations for dependent random variables with heavy tails*, Stat Probab Lett, 2009, 79: 1290-1298.
- [19] C Miller. *Three theorems on  $\rho^*$ -mixing random fields*, J Theor Probab, 1994, 7: 867-882.
- [20] M Peligrad. *Maximum of partial sums and an invariance principle for a class of weak dependent random variables*, Proc Amer Math Soc, 1998, 126(4): 1181-1189.
- [21] M Peligrad, A Gut. *Almost-sure results for a class of dependent random variables*, J Theor Probab, 1999, 12(1): 87-104.
- [22] A R Pruss. *Randomly sampled Riemann sums and complete convergence in the law of large numbers for a case without identical distribution*, Proc Amer Math Soc, 1996, 124: 919-929.
- [23] Y A Rozanov, V A Volkonski. *Some limit theorems for random function*, Theory Probab, 1959, 4: 186-207.
- [24] D H Qiu, P Y Chen, R G Antonini, A Volodin. *On the complete convergence for arrays of rowwise extended negatively dependent random variables*, J Korean Math Soc, 2013, 50(2): 379-392.
- [25] A Shen. *Probability inequalities for END sequence and their applications*, J Inequal Appl, 2011, 98: 1-12.

- [26] A Shen. *Complete convergence for weighted sums of END random variables and its applications to nonparametric regression models*, J Nonparametric Stat, 2016, 28(4): 702-715.
- [27] A Shen, M Xue, W Wang. *Complete convergence for weighted sums of extended negatively dependent random variables*, Communications in Stat-Theory and Methods, 2017, 46(3): 1433-1444.
- [28] D Q Tuyen. *A strong law for mixing random variables*, Periodica Math Hung, 1999, 38: 131-136.
- [29] S Utev, M Peligrad. *Maximal inequalities and an invariance principle for a class of weakly dependent random variables*, J Theor Probab, 2003, 16(1): 101-115.
- [30] X Wang, S Hu, W Yang, Y Shen. *On complete convergence for weighted sums of  $\varphi$ -mixing random variables*, J Ineq Appl, 2010, 2010, Article ID: 372390, 13 pages, doi: 10.1155/2010/372390.
- [31] X J Wang, X Q Li, S H Hu, X H Wang. *On complete convergence for an extended negatively dependent sequence*, Communications in Stat-Theory and Methods, 2014, 43: 2923-2937.
- [32] X J Wang, L L Zheng, C Xu, S H Hu. *Complete consistency for the estimator of nonparameter regression models based on extended negatively dependent errors*, Statistics: A Journal of Theoretical and Applied Stat, 2015, 49: 396-407.
- [33] J F Wang, F B Lu. *Inequalities of maximum of partial sums and weak convergence for a class of weak dependent random variables*, Acta Math Sin, 2006, 22: 693-700.
- [34] Y F Wu, M Z Song, C H Wang. *Complete moment convergence and mean convergence for arrays of rowwise extended negatively dependent random variables*, The Scientific World Journal, 2014, 2014, Artical ID: 478612, 7 pages, <http://dx.doi.org/10.1155/2014/478612>.
- [35] X Xu, J G Yan. *Complete moment convergence for randomly weighted sums of END sequences and its applications*, Communications in Stat-Theory and Methods, 2021, 50(12): 2877-2899.
- [36] J G Yan. *Strong Stability of a type of Jamison Weighted Sums for END Random Variables*, J Korean Math Soc, 2017, 54(3): 897-907.
- [37] J G Yan. *Almost sure convergence for weighted sums of WNOD random variables and its applications in nonparametric regression models*, Communications in Stat-Theory and Methods, 2018, 47(16): 3893-3909.
- [38] J G Yan. *Complete Convergence and Complete Moment Convergence for Maximal Weighted Sums of Extended Negatively Dependent Random Variables*, Acta Math Sinica, English Series, 2018, 34(10): 1501-1516.
- [39] J G Yan. *On Complete Convergence in Marcinkiewicz-Zygmund Type SLLN for END Random Variables and Its Applications*, Communications in Stat-Theory and Methods, 2019, 48(20): 5074-5098.
- [40] Y Zhang. *Complete moment convergence for moving average process generated by  $\rho^-$ -mixing random variables*, J Ineq Appl, 2015, 2015: 245, DOI: 10.1186/s13660-015-0766-5.

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