

*Research Article*

## Strong Laws of Large Numbers for Arrays of Rowwise $\rho^*$ -Mixing Random Variables

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Some strong laws of large numbers for arrays of rowwise  $\rho^*$ -mixing random variables are obtained. The result obtained not only generalizes the result of Hu and Taylor (1997) to  $\rho^*$ -mixing random variables, but also improves it.

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### 1. Introduction

Let  $\{X, X_n, n \geq 1\}$  be a sequence of independent identically distributed (i.i.d.) random variables. The Marcinkiewicz-Zygmund strong law of large numbers (SLLN) provides that

$$\begin{aligned} \frac{1}{n^{1/\alpha}} \sum_{i=1}^n (X_i - EX_i) &\longrightarrow 0 \quad \text{a.s. for } 1 \leq \alpha < 2, \\ \frac{1}{n^{1/\alpha}} \sum_{i=1}^n X_i &\longrightarrow 0 \quad \text{a.s. for } 0 < \alpha < 1 \end{aligned} \quad (1.1)$$

if and only if  $E|X|^\alpha < \infty$ . The case  $\alpha = 1$  is due to Kolmogorov. In the case of independence (but not necessarily identically distributed), Hu and Taylor [1] proved the following strong law of large numbers.

**THEOREM 1.1.** *Let  $\{X_{ni}; 1 \leq i \leq n, n \geq 1\}$  be a triangular array of rowwise independent random variables. Let  $\{a_n, n \geq 1\}$  be a sequence of positive real numbers such that  $0 < a_n \uparrow \infty$ . Let  $\psi(t)$  be a positive, even function such that  $\psi(|t|)/|t|^p$  is an increasing function of  $|t|$  and  $\psi(|t|)/|t|^{p+1}$  is a decreasing function of  $|t|$ , respectively, that is,*

$$\frac{\psi(|t|)}{|t|^p} \uparrow, \quad \frac{\psi(|t|)}{|t|^{p+1}} \downarrow, \quad \text{as } |t| \uparrow \quad (1.2)$$

for some nonnegative integer  $p$ . If  $p \geq 2$  and

$$\begin{aligned} EX_{ni} &= 0, \\ \sum_{n=1}^{\infty} \sum_{i=1}^n E \frac{\psi(|X_{ni}|)}{\psi(a_n)} &< \infty, \\ \sum_{n=1}^{\infty} \left( \sum_{i=1}^n E \left( \frac{X_{ni}}{a_n} \right)^2 \right)^{2k} &< \infty, \end{aligned} \tag{1.3}$$

where  $k$  is a positive integer, then

$$\frac{1}{a_n} \sum_{i=1}^n X_{ni} \longrightarrow 0 \quad a.s. \tag{1.4}$$

Let nonempty sets  $S, T \subset \mathcal{N}$ , and define  $\mathcal{F}_S = \sigma(X_k, k \in S)$ , and the maximal correlation coefficient  $\rho_n^* = \sup \text{corr}(f, g)$  where the supremum is taken over all  $(S, T)$  with  $\text{dist}(S, T) \geq n$  and all  $f \in L_2(\mathcal{F}_S), g \in L_2(\mathcal{F}_T)$ , and where  $\text{dist}(S, T) = \inf_{x \in S, y \in T} |x - y|$ .

A sequence of random variables  $\{X_n, n \geq 1\}$  on a probability space  $\{\Omega, \mathcal{F}, P\}$  is called  $\rho^*$ -mixing if

$$\lim_{n \rightarrow \infty} \rho_n^* < 1. \tag{1.5}$$

An array of random variables  $\{X_{ni}; i \geq 1, n \geq 1\}$  is called rowwise  $\rho^*$ -mixing random variables if for every  $n \geq 1, \{X_{ni}; i \geq 1\}$  is a  $\rho^*$ -mixing sequence of random variables.

As for  $\rho^*$ -mixing sequences of random variables, Bryc and Smoleński [2] established the moments inequality of partial sums. Peligrad [3] obtained a CLT. Peligrad [4] established an invariance principle. Peligrad and Gut [5] established the Rosenthal-type maximal inequality. Utev and Peligrad [6] obtained an invariance principle of nonstationary sequences.

The main purpose of this paper is to establish a strong law of large numbers for arrays of rowwise  $\rho^*$ -mixing random variables. The result obtained not only generalizes the result of Hu and Taylor [1] to  $\rho^*$ -mixing random variables, but also improves it.

## 2. Main results

Throughout this paper,  $C$  will represent a positive constant though its value may change from one appearance to the next, and  $a_n = O(b_n)$  will mean  $a_n \leq Cb_n$ .

Let  $\{X, X_n, n \geq 1\}$  be a sequence of independent identically distributed (i.i.d.) random variables and denote  $S_n = \sum_{i=1}^n X_i$ . The Hsu-Robbins-Erdős law of large numbers (see Hsu and Robbins [7], Erdős [8]) states that

$$\forall \varepsilon > 0, \quad \sum_{n=1}^{\infty} P(|S_n| > \varepsilon n) < \infty \tag{2.1}$$

is equivalent to  $EX = 0, EX^2 < \infty$ .

This is a fundamental theorem in probability theory and has been intensively investigated by many authors in the past decades. One of the most important results is Baum-Katz [9] law of large numbers, which states that for  $p < 2$  and  $r \geq p$ ,

$$\forall \varepsilon > 0, \sum_{n=1}^{\infty} n^{r/p-2} P(|S_n| > \varepsilon n^{1/p}) < \infty \tag{2.2}$$

if and only if  $E|X|^r < \infty$ ,  $r \geq 1$ , and  $EX = 0$ .

There are many extensions in various directions. Some of them can be found by Chow and Lai in [10, 11], where the authors propose a two-sided estimate, and by Petrov in [12].

In order to prove our main result, we need the following lemma.

LEMMA 2.1 (see Utev and Peligrad [6]). *Let  $\{X_i, i \geq 1\}$  be a  $\rho^*$ -mixing sequence of random variables,  $EX_i = 0$ ,  $E|X_i|^p < \infty$  for some  $p \geq 2$  and for every  $i \geq 1$ . Then there exists  $C = C(p)$ , such that*

$$E \max_{1 \leq k \leq n} \left| \sum_{i=1}^k X_i \right|^p \leq C \left\{ \sum_{i=1}^n E|X_i|^p + \left( \sum_{i=1}^n EX_i^2 \right)^{p/2} \right\}. \tag{2.3}$$

THEOREM 2.2. *Let  $\{X_{ni}; i \geq 1, n \geq 1\}$  be an array of rowwise  $\rho^*$ -mixing random variables. Let  $\{a_n, n \geq 1\}$  be a sequence of positive real numbers such that  $0 < a_n \uparrow \infty$ . Let  $\psi(t)$  be a positive, even function such that  $\psi(|t|)/|t|$  is an increasing function of  $|t|$  and  $\psi(|t|)/|t|^p$  is a decreasing function of  $|t|$ , respectively, that is,*

$$\frac{\psi(|t|)}{|t|} \uparrow, \quad \frac{\psi(|t|)}{|t|^p} \downarrow, \quad \text{as } |t| \uparrow \tag{2.4}$$

for some nonnegative integer  $p$ . If  $p \geq 2$  and

$$\begin{aligned} EX_{ni} &= 0, \\ \sum_{n=1}^{\infty} \sum_{i=1}^n E \frac{\psi(|X_{ni}|)}{\psi(a_n)} &< \infty, \\ \sum_{n=1}^{\infty} \left( \sum_{i=1}^n E \left( \frac{X_{ni}}{a_n} \right)^2 \right)^{v/2} &< \infty, \end{aligned} \tag{2.5}$$

where  $v$  is a positive integer,  $v \geq p$ , then

$$\forall \varepsilon > 0, \sum_{n=1}^{\infty} P \left( \max_{1 \leq k \leq n} \left| \frac{1}{a_n} \sum_{i=1}^k X_{ni} \right| > \varepsilon \right) < \infty. \tag{2.6}$$

#### 4 Discrete Dynamics in Nature and Society

*Proof of Theorem 2.2.* For all  $i \geq 1$ , define  $X_i^{(n)} = X_{ni}I(|X_{ni}| \leq a_n)$ ,  $T_j^{(n)} = (1/a_n) \sum_{i=1}^j (X_i^{(n)} - EX_i^{(n)})$ , then for all  $\varepsilon > 0$ ,

$$\begin{aligned} & P\left(\max_{1 \leq k \leq n} \left| \frac{1}{a_n} \sum_{i=1}^k X_{ni} \right| > \varepsilon\right) \\ & \leq P\left(\max_{1 \leq j \leq n} |X_{nj}| > a_n\right) + P\left(\max_{1 \leq j \leq n} |T_j^{(n)}| > \varepsilon - \max_{1 \leq j \leq n} \left| \frac{1}{a_n} \sum_{i=1}^j EX_i^{(n)} \right|\right). \end{aligned} \quad (2.7)$$

First, we show that

$$\max_{1 \leq j \leq n} \left| \frac{1}{a_n} \sum_{i=1}^j EX_i^{(n)} \right| \rightarrow 0, \quad \text{as } n \rightarrow \infty. \quad (2.8)$$

In fact, by  $EX_{ni} = 0$ ,  $\psi(|t|)/|t| \uparrow$  as  $|t| \uparrow$  and  $\sum_{n=1}^{\infty} \sum_{i=1}^n E(\psi(|X_{ni}|)/\psi(a_n)) < \infty$ , then

$$\begin{aligned} \max_{1 \leq j \leq n} \left| \frac{1}{a_n} \sum_{i=1}^j EX_i^{(n)} \right| &= \max_{1 \leq j \leq n} \left| \frac{1}{a_n} \sum_{i=1}^j EX_{ni}I(|X_{ni}| \leq a_n) \right| \\ &= \max_{1 \leq j \leq n} \left| \frac{1}{a_n} \sum_{i=1}^j EX_{ni}I(|X_{ni}| > a_n) \right| \\ &\leq \sum_{i=1}^n \frac{E|X_{ni}|I(|X_{ni}| > a_n)}{a_n} \\ &\leq \sum_{i=1}^n \frac{E\psi(|X_{ni}|)I(|X_{ni}| > a_n)}{\psi(a_n)} \\ &\leq \sum_{i=1}^n \frac{E\psi(|X_{ni}|)}{\psi(a_n)} \rightarrow 0, \quad \text{as } n \rightarrow \infty. \end{aligned} \quad (2.9)$$

From (2.7) and (2.8), it follows that for  $n$  large enough,

$$P\left(\max_{1 \leq k \leq n} \left| \frac{1}{a_n} \sum_{i=1}^k X_{ni} \right| > \varepsilon\right) \leq \sum_{j=1}^n P(|X_{nj}| > a_n) + P\left(\max_{1 \leq j \leq n} |T_j^{(n)}| > \frac{\varepsilon}{2}\right). \quad (2.10)$$

Hence, we need only to prove that

$$\begin{aligned} I &=: \sum_{n=1}^{\infty} \sum_{j=1}^n P(|X_{nj}| > a_n) < \infty, \\ II &=: \sum_{n=1}^{\infty} P\left(\max_{1 \leq j \leq n} |T_j^{(n)}| > \frac{\varepsilon}{2}\right) < \infty. \end{aligned} \quad (2.11)$$

From the fact that  $\sum_{n=1}^{\infty} \sum_{i=1}^n E(\psi(|X_{ni}|)/\psi(a_n)) < \infty$ , it follows easily that

$$I = \sum_{n=1}^{\infty} \sum_{j=1}^n P(|X_{nj}| > a_n) \leq \sum_{n=1}^{\infty} \sum_{j=1}^n E \frac{\psi(|X_{nj}|)}{\psi(a_n)} < \infty. \quad (2.12)$$

By  $\nu \geq p$  and  $\psi(|t|)/|t|^p \downarrow$  as  $|t| \uparrow$ , then  $\psi(|t|)/|t|^\nu \downarrow$  as  $|t| \uparrow$ .

By Markov inequality, Lemma 2.1, and  $\sum_{n=1}^{\infty} (\sum_{i=1}^n E(X_{ni}/a_n)^2)^{\nu/2} < \infty$ , we have

$$\begin{aligned} II &= \sum_{n=1}^{\infty} P\left(\max_{1 \leq j \leq n} |T_j^{(n)}| > \frac{\varepsilon}{2}\right) \\ &\leq \sum_{n=1}^{\infty} \left(\frac{\varepsilon}{2}\right)^{-\nu} E \max_{1 \leq j \leq n} |T_j^{(n)}|^\nu \\ &\leq C \sum_{n=1}^{\infty} \left(\frac{\varepsilon}{2}\right)^{-\nu} \frac{1}{a_n^\nu} \left[ \left( \sum_{j=1}^n E|X_j^{(n)}|^2 \right)^{\nu/2} + \sum_{j=1}^n E|X_j^{(n)}|^\nu \right] \\ &\leq C \sum_{n=1}^{\infty} \frac{1}{a_n^\nu} \sum_{j=1}^n E|X_j^{(n)}|^\nu + C \sum_{n=1}^{\infty} \frac{1}{a_n^\nu} \left( \sum_{j=1}^n E|X_j^{(n)}|^2 \right)^{\nu/2} \\ &= C \sum_{n=1}^{\infty} \frac{1}{a_n^\nu} \sum_{j=1}^n E|X_{nj}|^\nu I(|X_{nj}| \leq a_n) + C \sum_{n=1}^{\infty} \frac{1}{a_n^\nu} \left( \sum_{j=1}^n E|X_j^{(n)}|^2 \right)^{\nu/2} \\ &\leq C \sum_{n=1}^{\infty} \sum_{i=1}^n E \frac{\psi(|X_{ni}|)}{\psi(a_n)} + C \sum_{n=1}^{\infty} \frac{1}{a_n^\nu} \left[ \sum_{j=1}^n E|X_j^{(n)}|^2 \right]^{\nu/2} \\ &\leq C \sum_{n=1}^{\infty} \sum_{i=1}^n E \frac{\psi(|X_{ni}|)}{\psi(a_n)} + C \sum_{n=1}^{\infty} \left( \sum_{i=1}^n E \left( \frac{X_{ni}}{a_n} \right)^2 \right)^{\nu/2} < \infty. \end{aligned} \quad (2.13)$$

Now we complete the proof of Theorem 2.2. □

**COROLLARY 2.3.** *Under the conditions of Theorem 2.2, then*

$$\frac{1}{a_n} \sum_{i=1}^n X_{ni} \longrightarrow 0 \text{ a.s.} \quad (2.14)$$

*Proof of Corollary 2.3.* By Theorem 2.2, the Proof of Corollary 2.3 is obvious. □

**Remark 2.4.** Corollary 2.3 not only generalizes the result of Hu and Taylor [1] to  $\rho^*$ -mixing random variables, but also improves it.

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