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# On almost sure limiting behavior of a dependent random sequence

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Republic of China**Abstract**

We study some sufficient conditions for the almost certain convergence of averages of arbitrarily dependent random variables by certain summability methods. As corollaries, we generalized some known results.

**MSC:** 60F15**Keywords:** dependent random variable; summability; dominated random sequence

## 1 Introduction

In reference [1], Chow and Teicher gave a limit theorem of almost certain summability of i.i.d. random variables as follows.

**Theorem** (Chow *et al.*, 1971) *Let  $a(x)$ ,  $x > 0$  be a positive non-increasing function and  $a_n = a(n)$ ,  $A_n = \sum_{k=1}^n a_k$ ,  $b_n = A_n/a_n$ , where*

- (1)  $A_n \rightarrow \infty$ ;
- (2)  $0 < \liminf_n \frac{b_n}{n} a(\log b_n) \leq \limsup_n \frac{b_n}{n} a(\log b_n) < \infty$ ;
- (3)  $xa(\log^+ x)$  is non-decreasing for  $x > 0$ , then i.i.d.  $\{X, X_n\}$  are  $a_n$  summable, i.e.,

$$T_n = A_n^{-1} \sum_{k=1}^n a_k X_k - C_n \rightarrow 0 \quad a.c.$$

for some choice of centering constants  $C_n$ , if and only if

$$E|X|a(\log^+ |X|) < \infty.$$

Motivated by Chow and Teicher's idea, in this paper we consider the problem of arbitrarily dependent random variables and their limiting behavior from a new perspective.

Throughout this paper, let  $\mathbb{N}$  denote the set of positive integers,  $\{X, X_n, \mathcal{F}_n, n \in \mathbb{N}\}$  be a stochastic sequence defined on the probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ , i.e., the sequence of  $\sigma$ -fields  $\{\mathcal{F}_n, n \in \mathbb{N}\}$  in  $\mathcal{F}$  is increasing in  $n$ , and  $\{\mathcal{F}_n\}$  are adapted to random variables  $\{X_n\}$ ,  $\mathcal{F}_0$  denotes the trivial  $\sigma$  field  $\{\Phi, \Omega\}$  and  $\mathbf{I}_{[\cdot]}$  the indicator function.

We begin by introducing some terminology and lemmas.

**Definition 1** (Adler *et al.*, 1987 [2]) Let  $\{X_n, n \in \mathbb{N}\}$  be a sequence of random variables, and it is said to be stochastically dominated by a random variable  $X$  (we write  $\{X_n, n \in \mathbb{N}\} < X$ )

if there exists a constant  $C > 0$ , for almost every  $\omega \in \Omega$ , such that

$$\sup_{n \in \mathbb{N}} \mathbb{P}\{|X_n| > t\} \leq C\mathbb{P}\{|X| > t\} \quad \text{for all } t > 0.$$

**Lemma 1** (Chow *et al.*, 1978 [3]) *Let  $\{X_n, \mathcal{F}_n, n \in \mathbb{N}\}$  be an  $L_p$  ( $1 \leq p \leq 2$ ) martingale difference sequence, if  $\sum_{n=1}^{\infty} E(|X_n|^p | \mathcal{F}_{n-1}) < \infty$ , then  $\sum_{n=1}^{\infty} X_n$  a.c. converges.*

**Lemma 2** *Let  $\{X, X_n, n \in \mathbb{N}\}$  be a sequence of random variables. If  $\{X_n\} \prec X$ , then for all  $t > 0$ ,*

$$\mathbb{E}|X_n|^2 \mathbf{1}_{\{|X_n| \leq t\}} \leq C[t^2 \mathbb{P}(|X| > t) + \mathbb{E}X^2 \mathbf{1}_{\{|X| \leq t\}}].$$

*Proof* By the integral equality

$$2 \int_0^t s \mathbb{P}(|X_n| > s) ds = t^2 \mathbb{P}(|X_n| > t) + \mathbb{E}|X_n|^2 \mathbf{1}_{\{|X_n| \leq t\}},$$

it follows that

$$\begin{aligned} \mathbb{E}|X_n|^2 \mathbf{1}_{\{|X_n| \leq t\}} &\leq 2 \int_0^t s \mathbb{P}(|X_n| > s) ds \\ &\leq 2C \int_0^t s \mathbb{P}(|X| > s) ds = C[t^2 \mathbb{P}(|X| > t) + \mathbb{E}X^2 \mathbf{1}_{\{|X| \leq t\}}]. \end{aligned} \quad \square$$

## 2 Strong law of large numbers

In this section, we always assume that  $a(x)$ ,  $x > 0$  is a positive non-increasing function and  $a_n = a(n)$ ,  $A_n = \sum_{k=1}^n a_k$ ,  $b_n = A_n/a_n$ , where

- (1)  $A_n \rightarrow \infty$ ;
- (2)  $0 < \liminf_n \frac{b_n}{n} a(\log b_n) \leq \limsup_n \frac{b_n}{n} a(\log b_n) < \infty$ ;
- (3)  $xa(\log^+ x)$  is non-decreasing for  $x > 0$ .

**Theorem 1** *Let  $\{X, X_n\}$  be a sequence of random variables with  $\{X_n\} \prec X$ . If  $E|X|a(\log^+ |X|) < \infty$ , then*

$$\lim_n \frac{1}{A_n} \sum_{k=1}^n a_k [X_k - E(X_k \mathbf{1}_{\{|X_k| \leq b_k\}} | \mathcal{F}_{k-1})] = 0, \quad \text{a.c.} \tag{2.1}$$

*Proof* To prove (2.1) by applying the Kronecker lemma, it suffices to show that

$$\text{the series } \sum_{n=1}^{\infty} \frac{X_n - E(X_n \mathbf{1}_{\{|X_n| \leq b_n\}} | \mathcal{F}_{n-1})}{b_n} \text{ converges a.c.}$$

Since  $0 < a(x) \downarrow$ , (1) guarantees that  $b_n \uparrow \infty$ . Choose  $m_0$  such that  $n \geq m_0$  implies

$$\alpha n \leq b_n a(\log b_n) \leq \beta n \tag{2.2}$$

whence  $b_n \geq \alpha n [a(\log b_m)]^{-1}$  for  $n \geq m \geq m_0$  entailing

$$\sum_{k=m}^{\infty} b_k^{-2} \leq \frac{a^2(\log b_m)}{\alpha^2 m}. \tag{2.3}$$

Put  $Y_n = X_n \mathbf{1}_{\{|X_n| \leq b_n\}}$ ,  $Z_n = X_n \mathbf{1}_{\{|X_n| > b_n\}}$ , obviously,  $X_n = Y_n + Z_n, n \in \mathbb{N}$ . Note that  $\{X_n\} \prec X$  and the condition  $E|X|a(\log^+ |X|) < \infty$ , we have

$$\begin{aligned} & \sum_{n=1}^{\infty} \mathbb{P}(|X_n| > b_n) \\ &= \sum_{n=1}^{m_0-1} \mathbb{P}(|X_n| > b_n) + \sum_{n=m_0}^{\infty} \mathbb{P}(|X_n| > b_n) \\ &\leq m_0 - 1 + C \sum_{n=m_0}^{\infty} \mathbb{P}(|X| > b_n) \\ &\leq m_0 - 1 + C \sum_{n=m_0}^{\infty} \mathbb{P}(|X|a(\log |X|) \geq b_n a(\log b_n)) \\ &\leq m_0 - 1 + C \sum_{n=m_0}^{\infty} \mathbb{P}(|X|a(\log |X|) \geq \alpha n) < \infty, \end{aligned} \tag{2.4}$$

which shows

$$\mathbb{P}(X_n \neq Z_n, \text{i.o.}) = 0. \tag{2.5}$$

Let  $W_n = \frac{Y_n}{b_n} - E(\frac{Y_n}{b_n} | \mathcal{F}_{n-1})$ , then  $(W_n, \mathcal{F}_n, n \in \mathbb{N})$  is a martingale difference sequence. Since

$$\begin{aligned} & \sum_{k=1}^{\infty} E \frac{Y_k^2}{b_k^2} = \sum_{k=1}^{\infty} \frac{EX_k^2 \mathbf{1}_{\{|X_k| \leq b_k\}}}{b_k^2} \\ &\leq C \sum_{k=1}^{\infty} \left[ E \mathbf{1}_{\{|X| > b_k\}} + \frac{EX^2 \mathbf{1}_{\{|X| \leq b_k\}}}{b_k^2} \right] \quad (\text{by Lemma 2}) \\ &= C \sum_{k=1}^{\infty} E \mathbf{1}_{\{|X| > b_k\}} + C \left( \sum_{k=1}^{m_0-1} + \sum_{k=m_0}^{\infty} \right) \frac{EX^2 \mathbf{1}_{\{|X| \leq b_k\}}}{b_k^2} \\ &\leq C \sum_{k=1}^{\infty} \mathbb{P}(|X| > b_k) + C(m_0 - 1) + C \sum_{k=m_0}^{\infty} b_k^{-2} \left( \int_{\{|X_k| \leq b_{m_0-1}\}} X^2 + \sum_{i=m_0}^k \int_{[b_{i-1} < |X| \leq b_i]} X^2 \right) \\ &\leq \mathcal{O}(1) + C \sum_{i=m_0}^{\infty} \sum_{k=i}^{\infty} b_k^{-2} \int_{[b_{i-1} < |X| \leq b_i]} X^2 \quad (\text{by (2.4)}) \\ &\leq \mathcal{O}(1) + \alpha^{-2} C \sum_{i=m_0}^{\infty} i^{-1} a^2(\log b_i) \int_{[b_{i-1} < |X| \leq b_i]} X^2 \quad (\text{by (2.3)}) \\ &\leq \mathcal{O}(1) + \alpha^{-2} \beta C \sum_{i=m_0}^{\infty} a(\log b_i) \int_{[b_{i-1} < |X| \leq b_i]} |X| \quad (\text{by (2.2)}) \\ &\leq \mathcal{O}(1) + \alpha^{-2} \beta C \sum_{i=m_0}^{\infty} \int_{[b_{i-1} < |X| \leq b_i]} |X| a(\log |X|) < \infty. \end{aligned} \tag{2.6}$$

Note that

$$\begin{aligned}
 E\left[\sum_{n=1}^{\infty} E(W_n^2 | \mathcal{F}_{n-1})\right] &\leq E\left[\sum_{n=1}^{\infty} E\left(\frac{Y_n^2}{b_n^2} \middle| \mathcal{F}_{n-1}\right)\right] \\
 &= \sum_{n=1}^{\infty} E\frac{Y_n^2}{b_n^2} < \infty,
 \end{aligned}
 \tag{2.7}$$

which implies that  $\sum_{n=1}^{\infty} E(W_n^2 | \mathcal{F}_{n-1}) < \infty$  a.c. Hence, by Lemma 1, we have  $\sum_{n=1}^{\infty} W_n$  a.c. convergence.

Theorem 1 follows from (2.5) and (2.7). □

Theorem 1 also includes some particular cases of means, we can establish the following.

**Corollary 1** *Let  $\{X, X_n, n \in \mathbb{N}\}$  be a sequence of random variables with  $\{X_n\} \prec X$ . If for some  $\varepsilon > 0$ ,  $E\frac{|X|}{\log|X|\mathbf{1}_{[|X|>\varepsilon]}} < \infty$ , then*

$$\lim_n \frac{1}{\log n} \sum_{k=1}^n \left[ \frac{X_k - E(X_k \mathbf{1}_{[|X_k| \leq k \log k]} | \mathcal{F}_{k-1})}{k} \right] = 0, \quad a.c.$$

**Corollary 2** *Let  $\{X, X_n, n \in \mathbb{N}\}$  be a sequence of random variables with  $\{X_n\} \prec X$  and for some  $k \geq 2$ ,*

$$a_n = [n(\log n) \cdots (\log_{k-1} n)]^{-1},$$

where  $\log_1 n = \log n$ ,  $\log_k n = \log(\log_{k-1} n)$ ,  $k \geq 2$ , if for all large  $C > 0$ ,

$$E\frac{|X|\mathbf{1}_{[|X|>C]}}{(\log|X|) \cdots (\log_k |X|)} < \infty,$$

then

$$\lim_n \frac{1}{A_n} \sum_{k=1}^n a_k [X_k - E(X_k \mathbf{1}_{[|X_k| \leq b_k]} | \mathcal{F}_{k-1})] = 0, \quad a.c.$$

**Corollary 3** *Let  $\{X, X_n, n \in \mathbb{N}\}$  be a sequence of random variables with  $\{X_n\} \prec X$ . Further, let  $\mathcal{F}_n = \sigma(X_1, \dots, X_n)$  and  $\mathcal{F}_{-n} = \{\phi, \Omega\}$ ,  $n \geq 0$ . If  $E|X|a(\log^+ |X|) < \infty$ , then for any  $m \geq 1$ ,*

$$\lim_n \frac{1}{A_n} \sum_{k=1}^n a_k [X_k - E(X_k \mathbf{1}_{[|X_k| \leq b_k]} | \mathcal{F}_{k-m})] = 0, \quad a.c.
 \tag{2.8}$$

*Proof* Since  $\{X_{nm+l}, \mathcal{F}_{nm+l}, n \geq 1\}$  is an adapted stochastic sequence and  $\{X_{nm+l}\} \prec X$ , by Theorem 1, we have for  $l = 0, 1, \dots, m - 1$  that

$$\sum_{n=1}^{\infty} \frac{X_{nm+l} - E(X_{nm+l} \mathbf{1}_{[|X_{nm+l}| \leq b_{nm+l}]} | \mathcal{F}_{(n-1)m+l})}{b_{nm+l}} \text{ converges a.c.}$$

Therefore, we have

$$\begin{aligned} & \sum_{n=m}^{\infty} \frac{X_n - E(X_n \mathbf{1}_{\{|X_n| \leq b_n\}} | \mathcal{F}_{n-m})}{b_n} \\ &= \sum_{l=0}^{m-1} \sum_{n=1}^{\infty} \frac{X_{nm+l} - E(X_{nm+l} \mathbf{1}_{\{|X_{nm+l}| \leq b_{nm+l}\}} | \mathcal{F}_{(n-1)m+l})}{b_{nm+l}} \end{aligned} \text{ converges a.c.}$$

□

**Corollary 4** Let  $\{X_n, n \in \mathbb{N}\}$  be a sequence of  $m$ -dependent random variables. Further, let  $\mathcal{F}_n = \sigma(X_1, \dots, X_n)$  and  $\mathcal{F}_{-n} = \{\phi, \Omega\}, n \geq 0$ . If there exists a random variable  $X$  such that  $\{X_n\} < X$  and  $E|X|a(\log^+ |X|) < \infty$ , then

$$\lim_n \frac{1}{A_n} \sum_{k=1}^n a_k [X_k - E(X_k \mathbf{1}_{\{|X_k| \leq b_k\}})] = 0, \quad \text{a.c.}$$

*Proof* Note that  $\{X_n, n \in \mathbb{N}\}$  is a sequence of  $m$ -dependent random variables, then  $E(X_n | \mathcal{F}_{n-m}) = EX_n$ , Corollary 4 follows directly from Corollary 3. □

**Definition 2** (Stout, 1974) Let  $\{X_n, n \in \mathbb{N}\}$  be a sequence of random variables, and let  $\mathcal{F}_n^m = \sigma(X_n, \dots, X_m)$ . We say that the sequence  $\{X_n, n \in \mathbb{N}\}$  is  $*$ -mixing if there exists a positive integer  $M$  and a non-decreasing function  $\varphi(n)$  defined on integers  $n \geq M$  with  $\lim_n \varphi(n) = 0$  such that for  $n > M, A \in \mathcal{F}_0^m$  and  $B \in \mathcal{F}_{m+n}^\infty$ , the relation

$$|\mathbb{P}(A \cap B) - \mathbb{P}(A)\mathbb{P}(B)| \leq \varphi(n)\mathbb{P}(A)\mathbb{P}(B)$$

holds for any integer  $m \geq 1$ .

It has been proved (cf. [4]) that the  $*$ -mixing condition is equivalent to the condition

$$|\mathbb{P}(B | \mathcal{F}_0^m) - \mathbb{P}(B)| \leq \varphi(n)\mathbb{P}(B), \quad \text{a.c.}$$

for  $B \in \mathcal{F}_{m+n}^\infty$  and  $m \geq 1$  implies

$$|E(X_{n+m} | \mathcal{F}_0^m) - EX_{n+m}| \leq \varphi(n)E|X_{n+m}|, \quad \text{a.c.} \tag{2.9}$$

**Theorem 2** Let  $\{X, X_n, n \in \mathbb{N}\}$  be a sequence of  $*$ -mixing random variables with  $\{X_n\} < X$ . Further, let  $\mathcal{F}_n = \sigma(X_1, \dots, X_n)$  and  $\mathcal{F}_{-n} = \{\phi, \Omega\}, n \geq 0$ . If  $\max\{E|X|, E|X|a(\log^+ |X|)\} < \infty$ , then

$$\lim_n \frac{1}{A_n} \sum_{k=1}^n a_k [X_k - EX_k \mathbf{1}_{\{|X_k| \leq b_k\}}] = 0, \quad \text{a.c.}$$

*Proof* By Corollary 3, we have, for each  $m \geq 1$ ,

$$\lim_n \frac{1}{A_n} \sum_{k=1}^n a_k [X_k - E(X_k \mathbf{1}_{\{|X_k| \leq b_k\}} | \mathcal{F}_{k-m})] = 0, \quad \text{a.c.}$$

Since  $\{X_n, n \in \mathbb{N}\}$  is  $*$ -mixing, by (2.8) and (2.9), we obtain

$$\begin{aligned} & \left| \frac{1}{A_n} \sum_{k=1}^n a_k [X_k - EX_k \mathbf{1}_{\{|X_k| \leq b_k\}}] \right| \\ & \leq \left| \frac{1}{A_n} \sum_{k=1}^n a_k [X_k - E(X_k \mathbf{1}_{\{|X_k| \leq b_k\}} | \mathcal{F}_{k-m})] \right| \\ & \quad + \frac{1}{A_n} \sum_{k=1}^n a_k | [E(X_k \mathbf{1}_{\{|X_k| \leq b_k\}} | \mathcal{F}_{k-m}) - EX_k \mathbf{1}_{\{|X_k| \leq b_k\}}] | \\ & \leq \left| \frac{1}{A_n} \sum_{k=1}^n a_k [X_k - E(X_k \mathbf{1}_{\{|X_k| \leq b_k\}} | \mathcal{F}_{k-m})] \right| + \frac{\varphi(m)}{A_n} \sum_{k=1}^n a_k E|X_k| \mathbf{1}_{\{|X_k| \leq b_k\}} \\ & \leq \left| \frac{1}{A_n} \sum_{k=1}^n a_k [X_k - E(X_k \mathbf{1}_{\{|X_k| \leq b_k\}} | \mathcal{F}_{k-m})] \right| + \varphi(m) E|X| \rightarrow 0 \quad \text{a.c. (as } n \rightarrow \infty). \end{aligned}$$

Thus, using the Kroneker lemma, Theorem 2 follows.  $\square$

#### Competing interests

The authors declare that they have no competing interests.

#### Authors' contributions

WZ and FA carried out the design of the study and performed the analysis, WZ drafted the manuscript. All authors read and approved the final manuscript.

#### Acknowledgements

Foundation item: The National Nature Science Foundation of China (No. 11071104), Foundation of Anhui Educational Committee (KJ2012B117) and Graduate Innovation Fund of Anhui University of Technology (D2011025).

Received: 28 August 2012 Accepted: 3 January 2013 Published: 18 January 2013

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doi:10.1186/1029-242X-2013-25

**Cite this article as:** Fan and Wang: On almost sure limiting behavior of a dependent random sequence. *Journal of Inequalities and Applications* 2013 **2013**:25.

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