

Convergence of Weighted Sums of Random Variables ^{*}

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Abstract: In this paper we study convergence almost sure of weighted sums $\sum_{j=1}^n a_{nj}X_j$, where $\{X_n, n \geq 1\}$ is a sequence of random variables and a_{nk} is an array of real numbers under conditions on a_{nj} and sequence $\{X_n, n \geq 1\}$.

Key Words: Negative Dependent, Strong Law of Large Numbers, Weighted Sums.

1. Introduction.

Convergence theorems for weighted sums have been obtained by Chow [5], Chow and Lai [6], Hanson and Koopman [7], Pruitt [8], Stout [10], Teicher [12] and Choi and Sung [4]. Several authors extended these theorem to sum of negatively dependent random variable, see for example Bozorgnia A., Patterson, R.F. and Taylor, R.L. [3], Taylor R.L. [11] and Amini M., Azarnoosh H.A. and Bozorgnia A. [2]. In this paper, we present various condition on $\{a_{nj}\}$ and $\{X_n\}$ for which $\sum_{j=1}^n a_{nj}X_j$ converges almost sure. It is show that for stochastically bounded random variables the condition $E|X|^2 < \infty$ and $\sum_{j=1}^n a_{nj}^2 = O(n^{-\beta})$, $\beta > 1$, imply $\sum_{j=1}^n a_{nj}X_j \rightarrow 0$, almost surely and for ND random variables with $EX_n = 0$, and $\sup |X_n| \leq c$ for some constant c and various condition on $\{a_{nj}\}$, we have $\sum_{j=1}^n a_{nj}X_j \rightarrow 0$, almost surely.

Definition 1. The random variables X_1, \dots, X_n are said to be ND if we have

$$P\left[\bigcap_{j=1}^n (X_j \leq x_j)\right] \leq \prod_{j=1}^n P[X_j \leq x_j], \quad (1)$$

and

$$P\left[\bigcap_{j=1}^n (X_j > x_j)\right] \leq \prod_{j=1}^n P[X_j > x_j], \quad (2)$$

for all $x_1, \dots, x_n \in R$. An infinite sequence $\{X_n, n \geq 1\}$ is said to be ND if every finite subset X_1, \dots, X_n is ND. The conditions (1) and (2) are equivalent for $n = 2$, but these do not agree for $n \geq 3$ (see Bozorgnia, [2]).

Definition 2. A sequence $\{X_n, n \geq 1\}$ of random variable is said to be stochastically bounded by random variable X if for all n and every real number $t > 0$,

$$P(|X_n| > t) \leq P(|X| > t).$$

Lemma 1 (Serfling [8]) Let X be a r.v. with $E(X) = \mu$. If $P[a \leq X \leq b] = 1$. Then for every real number $t > 0$,

$$Ee^{t(X-\mu)} \leq e^{\frac{t^2(b-a)^2}{8}},$$

and

$$Ee^{t|X-\mu|} \leq 2e^{\frac{t^2(b-a)^2}{8}}.$$

Lemma 2 Let X be a r.v. with $E(X) = 0$. If $|X| \leq c$, W.P.1. for some constant $c < \infty$. Then for every real number t ,

$$Ee^{tX} \leq \exp\left\{\frac{c^2 t^2}{2} \exp[ct]\right\}.$$

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Proof. From the following inequality

$$e^x \leq 1 + x + \frac{1}{2}x^2 e^{|x|}, \quad \forall x \in \mathbb{R},$$

for every $t \geq 0$, we have

$$\begin{aligned} Ee^{tX} &\leq 1 + E\left\{\frac{1}{2}t^2 X^2 e^{t|X|}\right\} \leq \\ &1 + \frac{c^2 t^2}{2} \exp[ct] \leq \exp\left\{\frac{c^2 t^2}{2} \exp[ct]\right\}. \quad \square \end{aligned}$$

The next two lemmas will be needed in the proof of SLLN's, in the next section.

Lemma 3. (Bozorgnia, [2]) Let X_1, \dots, X_n be ND random variables and f_1, \dots, f_n be a sequence of Borel functions which all are monotone increasing (or all are monotone decreasing), then $f_1(X_1), \dots, f_n(X_n)$ are ND random variables.

Lemma 4. (Bozorgnia, [2]) Let X_1, \dots, X_n be ND nonnegative random variables, then

$$E\left[\prod_{j=1}^n X_j\right] \leq \prod_{j=1}^n E[X_j].$$

2. Weighted sums of stochastically bounded r.v.'s

In this section, we prove two theorems for weighted sums of stochastically bounded random variables under the condition on a_{nj} .

Theorem 1. Let $\{X_n, n \geq 1\}$ be a sequence of stochastically bounded by random variable X , with $EX^2 < \infty$ and let $\{a_{nj}, 1 \leq j \leq n\}$, for each $n \geq 1$ be a triangular array of real numbers with $\sum_{j=1}^n a_{nj}^2 = O(n^{-\beta})$, $\beta > 1$. Then

$$\sum_{j=1}^n a_{nj} X_j \longrightarrow 0 \quad a.s.$$

Proof. From Cauchy Schwartz inequality we have

$$\left|\sum_{j=1}^n a_{nj} X_j\right|^2 \leq \left(\sum_{j=1}^n a_{nj}^2\right) \left(\sum_{j=1}^n X_j^2\right) \leq cn^{-\beta} \sum_{j=1}^n X_j^2.$$

For fixed $n \geq 1$, choose an integer k such that $2^{k-1} \leq n < 2^k$. Then

$$\limsup \left|\sum_{j=1}^n a_{nj} X_j\right|^2 \leq c \limsup \frac{1}{(2^{k-1})^\beta} \sum_{j=1}^{2^k} X_j^2, \quad (3)$$

we have

$$\begin{aligned} \sum_{k=1}^{\infty} P\left((2^{k-1})^{-\beta} \sum_{j=1}^{2^k} X_j^2 > \varepsilon\right) &\leq \frac{1}{\varepsilon} \sum_{k=1}^{\infty} (2^{k-1})^{-\beta} \sum_{j=1}^{2^k} EX_j^2 \\ &= \frac{1}{\varepsilon} \sum_{j=1}^{\infty} EX_j^2 \sum_{\{k: 2^k \geq j\}} 2^{-k\beta} = O(1) \sum_{j=1}^{\infty} \frac{EX_j^2}{j^\beta} < \\ &O(1) \sum_{j=1}^{\infty} \frac{EX_j^2}{j^\beta} < \infty. \end{aligned}$$

Thus by Borel Cantelli lemma we have

$$\limsup_k 2^{-\beta(k-1)} \sum_{j=1}^{2^k} X_j^2 = 0. \quad W.P.1$$

and by inequality (3)

$$\limsup_n \left|\sum_{j=1}^n a_{nj} X_j\right| = 0. \quad W.P.1$$

this complete the proof. \square

Theorem 2. Let $\{X_n, n \geq 1\}$ be a sequence of stochastically bounded independent random variables with $EX_n = 0$ and $EX^2 < \infty$. Let $\{a_{nj}, 1 \leq j \leq n, n \geq 1\}$ be a triangular array of real numbers with

$$\sum_{j=1}^n |a_{nj} - a_{n(j+1)}| = O(n^{-\beta}), \quad \beta > 1,$$

where $a_{n(n+1)} = 0$. Then

$$\sum_{j=1}^n a_{nj} X_j \longrightarrow 0, \quad a.s.$$

Proof. By the Following inequality

$$\begin{aligned} \left|\sum_{j=1}^n a_{nj} X_j\right| &\leq \max_{1 \leq i \leq n} \left|\sum_{j=1}^i X_j\right| \left(\sum_{j=1}^n |a_{nj} - a_{n(j+1)}|\right) \\ &\leq cn^{-\beta} \max_{1 \leq i \leq n} \left|\sum_{j=1}^i X_j\right|, \quad (4) \end{aligned}$$

and the Kolmogorov maximal inequality for any $\varepsilon > 0$, we obtain

$$\begin{aligned}
\sum_{n=1}^{\infty} P(n^{-\beta} \max_{1 \leq i \leq n} |S_i| > \varepsilon) &\leq \frac{1}{\varepsilon^2} \sum_{n=1}^{\infty} n^{-2\beta} \sum_{j=1}^n EX_j^2 \\
&= \frac{1}{\varepsilon^2} \sum_{j=1}^{\infty} EX_j^2 \sum_{n=j}^{\infty} n^{-2\beta} = O(1) \sum_{j=1}^{\infty} \frac{EX_j^2}{j^{2\beta-1}} \\
&< O(1) \sum_{j=1}^{\infty} \frac{EX_j^2}{j^{2\beta-1}} < \infty.
\end{aligned}$$

Thus by (4) we have

$$\begin{aligned}
\sum_{n=1}^{\infty} P\left(\left|\sum_{j=1}^n a_{nj} X_j\right| > \varepsilon\right) &\leq \\
\sum_{n=1}^{\infty} P(n^{-\beta} \max_{1 \leq i \leq n} |S_n| > \varepsilon) &< \infty,
\end{aligned}$$

hence by Borel Cantelli lemma we have

$$\sum_{j=1}^n a_{nj} X_j \longrightarrow 0, \quad a.s. \quad \square$$

3. Weighted sums negatively dependent r.v.'s

In this section, we obtain some strong limit theorems for weighted sums $\sum_{j=1}^n a_{nj} X_j$ where $\{X_n, n \geq 1\}$ is a sequence of negative dependent random variables and a_{nk} is a triangular array of real numbers under the conditions on X_n and a_{nj} .

Theorem 3. Let $\{X_n, n \geq 1\}$ be a sequence of ND random variables with $EX_j = 0$ and $\sup |X_n| \leq c$, for some finite constant c . Let $\{a_{nj}, 1 \leq j \leq n, n \geq 1\}$ be a triangular array of real number with $\max |a_{nj}| = O(n^{-1})$. Then

$$\sum_{j=1}^n a_{nj} X_j \longrightarrow 0, \quad a.s.$$

Proof. By Lemmas 2, 3 and 4 we have

$$\begin{aligned}
E \exp\left\{t \sum_{j=1}^n a_{nj} X_j\right\} &\leq \prod_{j=1}^n E \exp\{t a_{nj} X_j\} \\
&\leq \exp\left\{\frac{c^2 t^2}{n} \exp\left[\frac{ct}{n}\right]\right\}.
\end{aligned}$$

By Markove inequality and for $\varepsilon > 0$, we have

$$\begin{aligned}
\sum_{n=1}^{\infty} P\left(\left|\sum_{j=1}^n a_{nj} X_j\right| > \varepsilon\right) &\leq \\
\sum_{n=1}^{\infty} e^{-t\varepsilon} E \exp\left\{t \sum_{j=1}^n a_{nj} X_j\right\} &\leq \\
\sum_{n=1}^{\infty} \exp\left\{-t\varepsilon + \frac{c^2 t^2}{n} \exp\left[\frac{ct}{n}\right]\right\} &
\end{aligned}$$

Thus by putting $t = \frac{2}{\varepsilon} \ln(n)$, we have

$$\begin{aligned}
\sum_{n=1}^{\infty} P\left(\left|\sum_{j=1}^n a_{nj} X_j\right| > \varepsilon\right) &= \\
O(1) \sum_{n=1}^{\infty} \exp\{-2 \ln(n)\} &< \infty,
\end{aligned}$$

and Borel Cantelli lemma complete the proof. \square

Corollary 1. (Choi and Sung, [3]) Let $\{X_n, n \geq 1\}$ be a sequence of independent random variables with $EX_j = 0$ and $\sup |X_n| \leq c$, for some finite constant c .

Let $\{a_{nj}, 1 \leq j \leq n, n \geq 1\}$ be a triangular array of real number with $\max |a_{nj}| = O(n^{-1})$. Then

$$\sum_{j=1}^n a_{nj} X_j \longrightarrow 0, \quad a.s.$$

Theorem 4 Let $\{X_n, n \geq 1\}$ be a sequence of ND random variables with $EX_j = 0$, and $|X_n| \leq c$, W.P.1. for some finite constants. Let $\{a_{nj}, 1 \leq j \leq n, n \geq 1\}$ be a triangular array of real number with $\max |a_{nj}| = O(n^{-\beta})$ for every $\beta > (1/2)$. Then

$$\sum_{j=1}^n a_{nj} X_j \longrightarrow 0, \quad a.s.$$

Proof. By Markove inequality, and Lemmas 1,3 and 4 for every $\varepsilon > 0$ and $t > 0$, we have

$$\begin{aligned}
P\left(\left|\sum_{j=1}^n a_{nj} X_j\right| > \varepsilon\right) &\leq e^{-t\varepsilon} E \exp\left\{t \sum_{j=1}^n a_{nj} X_j\right\} \\
&\leq 2 \exp\left\{-t\varepsilon + \frac{t^2 c^2}{8} \sum_{j=1}^n a_{nj}^2\right\}.
\end{aligned}$$

By putting $t = \frac{4\varepsilon}{c^2 \sum_{j=1}^n a_{nj}^2}$, for some $0 < K < \infty$,

$$P(|\sum_{j=1}^n a_{nj} X_j| > \varepsilon) \leq 2 \exp\left\{-\frac{2\varepsilon^2}{c^2 \sum_{j=1}^n a_{nj}^2}\right\} \leq 2 \exp\{-K^2 n^{2\beta-1}\},$$

hence

$$\sum_{n=1}^{\infty} P(|\sum_{j=1}^n a_{nj} X_j| > \varepsilon) \leq 2 \sum_{n=1}^{\infty} \exp\{-K^2 n^{2\beta-1}\} < \infty.$$

Now Borel Cantelli Lemma complete the proof. \square

Theorem 5. Let $\{X_n, n \geq 1\}$ be a sequence of ND random variables with $EX_j = 0$ and $\sigma_j^2 = EX_j^2 < \infty$. Let

$$c_n = \max\left\{\text{esssup}\left(\frac{|X_j|}{B_n}\right), 1 \leq j \leq n\right\}, \quad B_n^2 = \sum_{j=1}^n \sigma_j^2.$$

If $\sum_{n=1}^{\infty} \exp\left\{-\frac{2\varepsilon^2}{c_n^2 B_n^2 \sum_{j=1}^n a_{nj}^2}\right\} < \infty$, for every $\varepsilon > 0$, then

$$\sum_{j=1}^n a_{nj} X_j \longrightarrow 0, \quad a.s.$$

Proof. By Markov inequality, Lemma 1, 3 and 4 for every $\varepsilon > 0$, and $t > 0$,

$$\begin{aligned} P(|\sum_{j=1}^n a_{nj} X_j| > \varepsilon) &= P\left(\frac{|\sum_{j=1}^n a_{nj} X_j|}{B_n} > \frac{\varepsilon}{B_n}\right) \\ &\leq \exp\left\{\frac{-t\varepsilon}{B_n}\right\} E \exp\left\{t \frac{|\sum_{j=1}^n a_{nj} X_j|}{B_n}\right\} \leq \\ &2 \exp\left\{-\frac{t\varepsilon}{B_n} + \frac{t^2 c_n^2}{8} \sum_{j=1}^n a_{nj}^2\right\}, \end{aligned}$$

for $t = \frac{4\varepsilon}{c_n^2 \sum_{j=1}^n a_{nj}^2}$, we have

$$\begin{aligned} \sum_{n=1}^{\infty} P(|\sum_{j=1}^n a_{nj} X_j| > \varepsilon) &\leq \\ \sum_{n=1}^{\infty} 2 \exp\left\{-\frac{2\varepsilon^2}{c_n^2 B_n^2 \sum_{j=1}^n a_{nj}^2}\right\} &< \infty, \end{aligned}$$

now Borel Cantelli Lemma complete the proof. \square

Corollary 2. If $\max |a_{nj}| = O(n^{-\beta} (c_n B_n)^{-1})$ for some $\beta > (1/2)$, then

$$\sum_{j=1}^n a_{nj} X_j \longrightarrow 0, \quad a.s. \quad \square$$

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