The strong law of large numbers
for arrays of NA random variables

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Abstract

Let \( \{X_{ni} | 1 \leq i \leq n, n \geq 1\} \) be an array of rowwise negatively
associated random variables under some suitable conditions. Then it is
shown that for some \( \frac{1}{2} < t \leq 1, \) \( n^{-1/t} \max_{1 \leq k \leq n} \sum_{i=1}^{k} X_{ni} \to 0 \)
completely as \( n \to \infty \) if and only if \( E|X|^{2t} < \infty \) and \( E|X_{ni}| = 0 \) and
\( \frac{1}{\sqrt{n}} \max_{1 \leq j \leq k} |\sum_{i=1}^{j} X_{ni}| \to 0 \)
completely as \( n \to \infty \) implies \( E|X|^\frac{1}{t} < \infty \).

AMS Mathematics Subject Classification: Primary 60F05 ; Secondary
62E10, 45E10.
Keywords and phrases: Negatively associated random variables, Strong law
of large numbers, Complete convergence.

1 Introduction

The concept of negatively associated random variables was introduced by Joag-
Dev and Proschan ([7]) although a very special case was first introduced by
Lehmann([9]). Many authors derived several important properties about neg-
atively associated (NA) sequences and also discussed some applications in the
area of statistics, probability, reliability and multivariate analysis. Compared
to positively associated random variables, the study of NA random variables has received less attention in the literature. Readers may refer to Karlin and Rinott ([8]), Ebrahimi and Ghosh ([3]), Block et al. ([2]), Newman ([12]), Joag-Dev ([6]), Joag-Dev and Proschan ([7]), Matula ([11]) and Roussas ([13]) among others.

Recently, some authors focused on the problem of limiting behavior of partial sums of NA sequences. Su et al. ([14]) derived some moment inequalities of partial sums and a weak convergence for a strongly stationary NA sequence. Su and Qin ([15]) studied some limiting results for NA sequences. More recently, Liang and Su ([10]), and Baek, Kim and Liang ([1]) considered some complete convergence for weighted sums of NA sequences.

Let \( \{X_{nk}\} \) be an array of random variables with \( E X_{nk} = 0 \) for all \( n \) and \( k \) and let \( 1 \leq p < 2 \). Then

\[
\frac{1}{n^{1/p}} \sum_{k=1}^{n} X_{nk} \rightarrow 0 \text{ completely as } n \rightarrow \infty
\]  

and where complete convergence is defined (Hsu and Robbins ([4])) by

\[
\sum_{n=1}^{\infty} P\left(\left|\frac{1}{n^{1/p}} \sum X_{nk}\right| > \varepsilon\right) < \infty \text{ for each } \varepsilon > 0.
\]  

Hu, Mátricz and Taylor ([5]) showed that for an array of i.i.d. random variables \( \{X_{nk}\} \), (1.1) holds if and only if \( E|X_{11}|^{2p} < \infty \).

The main purpose of this paper is to extend a similar results above to row-wise NA random variables, since independent and identically random variables are a special case of NA random variables. That is, we investigate the necessary and sufficient condition for \( \frac{1}{n^{1/t}} \max_{1 \leq k \leq n} |\sum_{i=1}^{k} X_{ni}| \rightarrow 0 \) completely as \( n \rightarrow \infty \) where \( \frac{1}{2} < t \leq 1 \) and let \( \{k_n\} \) and \( \{r_n\} \) be two increasing positive sequences satisfying some conditions, then, we show that \( \frac{1}{r_n} \max_{1 \leq j \leq k_n} |\sum_{i=1}^{j} X_{ni}| \rightarrow 0 \) completely as \( n \rightarrow \infty \) implies \( E|X|^{1+1/t} < \infty \) in NA setting.

Finally, in order to prove the strong law of large numbers for array of NA random variables, we give an important definition and some lemmas which will be used in obtaining the strong law of large numbers in the next section. **Definition 1.1** ([7]). Random variables \( X_1, \ldots, X_n \) are said to be negatively associated (NA) if for any two disjoint nonempty subsets \( A_1 \) and \( A_2 \) of \( \{1, \cdots, n\} \) and \( f_1 \) and \( f_2 \) are any two coordinatewise nondecreasing functions,

\[
\text{Cov}\left(f_1(X_i, \ i \in A_1), \ f_2(X_j, \ j \in A_2)\right) \leq 0,
\]

whenever the covariance is finite. An infinite family of random variables is NA if every finite subfamily is NA.
The strong law of large numbers for arrays of NA random variables

Lemma 1.2([10]). Let \( \{X_i | i \geq 1\} \) be a sequence of NA random variables and \( \{a_{ni} | 1 \leq i \leq n, \ n \geq 1\} \) be an array of real numbers. If \( P(\max_{1 \leq j \leq n} |a_{nj}X_j| > \varepsilon) < \delta \) for \( \delta \) small enough and \( n \) large enough, then

\[
\sum_{j=1}^{n} P(|a_{nj}X_j| > \varepsilon) = O(1)P(\max_{1 \leq j \leq n} |a_{nj}X_j| > \varepsilon)
\]

for sufficient large \( n \).

Lemma 1.3([5]). For any \( r \geq 1 \), \( E|X|^r < \infty \) if and only if

\[
\sum_{n=1}^{\infty} n^{r-1}P(|X| > n) < \infty.
\]

More precisely,

\[
r^{-r} \sum_{n=1}^{\infty} n^{r-1}P(|X| > n) \leq E|X|^r \leq 1 + r^{2r} \sum_{n=1}^{\infty} n^{r-1}P(|X| > n).
\]

Lemma 1.4([5]). If \( r \geq 1 \) and \( t > 0 \), then

\[
E|X|^rI(|X| \leq n^{1/t}) \leq r \int_{0}^{n^{1/t}} t^{r-1}P(|X| > t)dt
\]

and

\[
E|X|^rI(|X| > n^{1/t}) = n^{1/t}P(|X| > n^{1/t}) + \int_{n^{1/t}}^{\infty} P(|X| > t)dt.
\]

2 Main results

Theorem 2.1. Let \( \frac{1}{2} < t \leq 1 \) and let \( \{X_{ni} | 1 \leq i \leq n, \ n \geq 1\} \) be an array of rowwise NA random variables such that \( EX_{ni} = 0 \) and \( P(|X_{ni}| > x) = O(1)P(|X| > x) \) for all \( x \geq 0 \). If \( E|X|^{2t} < \infty \), then

\[
\frac{1}{n^{1/t}} \max_{1 \leq k \leq n} \left| \sum_{i=1}^{k} X_{ni} \right| \to 0 \ \text{completely as} \ n \to \infty.
\]

Proof. We define that for \( 1 \leq i \leq n, \ n \geq 1 \) and \( \frac{1}{2} < t \leq 1 \)

\[
Y_{ni} = X_{ni}I(|X_{ni}| \leq n^{1/t}) + n^{1/t}I(X_{ni} > n^{1/t}) - n^{1/t}I(X_{ni} < -n^{1/t}).
\]
To prove Theorem 2.1, it suffices to show that
\[
\sum_{n=1}^{\infty} P \left( \max_{1 \leq k \leq n} \left| \sum_{i=1}^{k} X_{ni} - \sum_{i=1}^{k} Y_{ni} \right| \geq \varepsilon n^{1/t} \right) < \infty \quad \text{for all } \varepsilon > 0, \tag{2.1}
\]
\[
\frac{1}{n^{1/t}} \max_{1 \leq k \leq n} \left| \sum_{i=1}^{k} EY_{ni} \right| \rightarrow 0, \tag{2.2}
\]
\[
\sum_{n=1}^{\infty} P \left( \max_{1 \leq k \leq n} \left| \sum_{i=1}^{k} Y_{ni} \right| \geq \varepsilon n^{1/t} \right) < \infty, \quad \text{for all } \varepsilon > 0. \tag{2.3}
\]

The proofs of (2.1) – (2.3) can be found in the following Lemmas 2.1 - 2.3.

**Lemma 2.1.** If \( E|X|^{2t} < \infty \), then (2.1) holds.

**Proof.**

\[
\sum_{n=1}^{\infty} P \left( \max_{1 \leq k \leq n} \left| \sum_{i=1}^{k} X_{ni} - \sum_{i=1}^{k} Y_{ni} \right| \geq \varepsilon n^{1/t} \right) \\
\leq \sum_{n=1}^{\infty} P \left( \bigcup_{i=1}^{n} X_{ni} \neq Y_{ni} \right) \\
\leq \sum_{n=1}^{\infty} \sum_{i=1}^{n} P(|X_{ni}| > n^{1/t}) \\
= \sum_{n=1}^{\infty} O(1) n P(|X| > n^{1/t}) \\
\leq O(1) E|X|^{2t} < \infty,
\]

when \( \frac{1}{2} < t \leq 1 \) since \( E|X|^{2t} < \infty \).

**Lemma 2.2.** If \( E|X|^{2t} < \infty \) and \( EX_{ni} = 0 \), then
\[
\frac{1}{n^{1/t}} \max_{1 \leq k \leq n} \left| \sum_{i=1}^{k} EY_{ni} \right| \rightarrow 0.
\]

**Proof.** To prove \( \frac{1}{n^{1/t}} \max_{1 \leq k \leq n} |\sum_{i=1}^{k} EY_{ni}| \rightarrow 0 \), it suffices to show that \( \sum_{n=1}^{\infty} \frac{1}{n^{1/t}} \max_{1 \leq k \leq n} |\sum_{i=1}^{k} EY_{ni}| < \infty \). Note that by \( EX_{ni} = 0 \), we have
\[
\sum_{n=1}^{\infty} \frac{1}{n^{1/t}} \max_{1 \leq k \leq n} |\sum_{i=1}^{k} EY_{ni}| \\
\leq \sum_{n=1}^{\infty} \frac{1}{n^{1/t}} \sum_{i=1}^{n} E|X_{ni}| I(|X_{ni}| > n^{1/t}) + \sum_{n=1}^{\infty} \frac{1}{n^{1/t}} \sum_{i=1}^{n} n^{1/t} P(|X_{ni}| > n^{1/t}) \\
=: I_1 + I_2 \text{ (say)}.
\]
First, to estimate $I_1$, by using Lemma 1.4,

$$I_1 = \sum_{n=1}^{\infty} \frac{1}{n^{1/t}} \sum_{i=1}^{n} \left[ n^{1/t} P(|X_{ni}| > n^{1/t}) + \int_{n^{1/t}}^{\infty} P(|X_{ni}| > x)dx \right]$$

$$= O(1) \sum_{n=1}^{\infty} n P(|X| > n^{1/t}) + O(1) \sum_{n=1}^{\infty} \frac{n}{n^{1/t}} \int_{n^{1/t}}^{\infty} P(|X| > x)dx =: I_1' \text{(say)}.$$

Letting $x = n^{1/t} s$ and applying Lemma 1.3, we have

$$I_1' = O(1) \sum_{n=1}^{\infty} n P(|X| > n^{1/t}) + O(1) \sum_{n=1}^{\infty} n \int_{1}^{\infty} P(|X| > n^{1/t} s)ds$$

$$\leq O(1) E|X|^{2t} + O(1) \int_{1}^{\infty} \sum_{n=1}^{\infty} n P(|s^{-1} X|^t > n)ds$$

$$\leq O(1) E|X|^{2t} + O(1) E|X|^{2t} \int_{1}^{\infty} s^{-2t} ds$$

$$= O(1) E|X|^{2t} < \infty.$$

As to $I_2$, we have

$$I_2 = \sum_{n=1}^{\infty} \frac{1}{n^{1/t}} \sum_{i=1}^{n} n^{1/t} P(|X_{ni}| > n^{1/t})$$

$$= \sum_{n=1}^{\infty} \sum_{i=1}^{n} P(|X_{ni}| > n^{1/t})$$

$$= O(1) \sum_{n=1}^{\infty} n P(|X| > n^{1/t})$$

$$\leq O(1) E|X|^{2t} < \infty.$$

**Lemma 2.3.** If $E|X|^{2t} < \infty$, then $\sum_{n=1}^{\infty} P\left( \max_{1 \leq k \leq n} \left| \sum_{i=1}^{k} Y_{ni} \right| \geq \varepsilon n^{1/t} \right) < \infty$ for all $\varepsilon > 0$.

**Proof.** From the definition of NA random variables, we know that $\{Y_{ni} \mid 1 \leq i \leq k, n \geq 1\}$ is still an array of rowwise NA random variables. Thus, we obtain that

$$\sum_{n=1}^{\infty} P\left( \max_{1 \leq k \leq n} \left| \sum_{i=1}^{k} Y_{ni} \right| \geq \varepsilon n^{1/t} \right)$$

$$\leq O(1) \sum_{n=1}^{\infty} \frac{1}{n} E\left( \max_{1 \leq k \leq n} \left| \sum_{i=1}^{k} Y_{ni} \right| \right)^t$$

$$\leq O(1) \sum_{n=1}^{\infty} \frac{1}{n} \sum_{i=1}^{n} E|Y_{ni}|^t$$
\[
\leq O(1) \left[ \sum_{n=1}^{\infty} \frac{1}{n} \sum_{i=1}^{n} E|X_{ni}|^t I(|X_{ni}| \leq n^{1/t}) + \sum_{n=1}^{\infty} \frac{1}{n} \sum_{i=1}^{n} P(|X_{ni}| > n^{1/t}) \right] =: I_3 + I_4 \text{ (say)}.
\]

First, we prove that \( I_3 < \infty \). Let \( G_{ni}(x) = P(|X_{ni}| \leq x) \), then we have

\[
I_3 = O(1) \sum_{n=1}^{\infty} \frac{1}{n} \sum_{i=1}^{n} E|X_{ni}|^t I(|X_{ni}| \leq n^{1/t})
\leq O(1) \sum_{n=1}^{\infty} \sum_{i=1}^{n} \int_{0}^{n^{1/t}} \left( \frac{x}{n^{1/t}} \right)^t dG_{ni}(x)
= O(1) \sum_{n=1}^{\infty} \sum_{i=1}^{n} \int_{0}^{1} \int_{(ns)^{1/t}}^{n^{1/t}} dG_{ni}(x) ds
= O(1) \sum_{n=1}^{\infty} \sum_{i=1}^{n} \int_{1}^{\infty} P((ns)^{1/t} < |X_{ni}| < n^{1/t}) ds
\leq O(1) \int_{0}^{1} \sum_{n=1}^{\infty} nP(|X| > (ns)^{1/t}) ds
\leq O(1) E|X|^{2t} < \infty.
\]

Also, the proof of \( I_4 \) is similar to that of Lemma 2.3.

Corollary 1 below is a corresponding result for a sequence of rowwise NA random variables.

**Corollary 1.** Let \( \frac{1}{2} < t \leq 1 \) and let \( \{X_i | i \geq 1\} \) be a sequence of NA random variables such that \( EX_i = 0 \) for all \( i \) and \( P(|X_i| > x) = O(1) P(|X| > x) \) for all \( x \geq 0 \). If \( E|X|^{2t} < \infty \), then

\[
\frac{1}{n^{1/t}} \max_{1 \leq k \leq n} \sum_{i=1}^{k} X_i \rightarrow 0 \ \text{completely as} \ \ n \rightarrow \infty.
\]

**Theorem 2.2.** Let \( \frac{1}{2} < t \leq 1 \) and let \( \{X_{ni} | 1 \leq i \leq n, n \geq 1\} \) be an array of rowwise NA random variables such that \( P(|X| > x) = O(1) P(|X_{ni}| > x) \) for all \( x \geq 0 \). Assume that \( \frac{1}{n^{1/t}} \max_{1 \leq k \leq n} \sum_{i=0}^{k} X_{ni} \rightarrow 0 \) completely as \( n \rightarrow \infty \), then \( E|X|^{2t} < \infty \) and \( EX_{ni} = 0 \).

**Proof.** From the assumptions, for any \( \varepsilon > 0 \),

\[
\sum_{n=1}^{\infty} P\left( \max_{1 \leq k \leq n} \sum_{i=1}^{k} X_{ni} \geq \varepsilon n^{1/t} \right) < \infty, \quad (2.4)
\]

By Lemma 1.2, we obtain that

\[
\sum_{i=1}^{n} P(|X_{ni}| \geq \varepsilon n^{1/t}) = O(1) P\left( \max_{1 \leq k \leq n} \sum_{i=1}^{k} X_{ni} \geq \varepsilon n^{1/t} \right),
\]

which, together with (2.4) and assumptions, we have

\[ \sum_{n=1}^{\infty} n P(|X| \geq \epsilon n^{1/t}) < \infty \]

which is equivalent to \( E|X|^{2t} < \infty \), by Lemma 1.3.

Now, under \( E|X|^{2t} < \infty \), we obtain from Theorem 2.1 that

\[ \sum_{n=1}^{\infty} P\left( \max_{1 \leq k \leq n} |\sum_{i=1}^{k} (X_{ni} - EX_{ni})| \geq \epsilon n^{1/t} \right) < \infty \text{ for any } \epsilon > 0 \quad (2.5) \]

(2.4) and (2.5) yield \( EX_{ni} = 0 \)

**Theorem 2.3.** Let \( \{X_{ni} | 1 \leq i \leq k_n, \ n \geq 1\} \) be an array of rowwise NA random variables with \( C_1 P(|X| > x) \leq C_1 \inf_{n,i} P(|X_{ni}| > x) \leq C_2 \sup_{n,i} P(|X_{ni}| > x) \) for all \( x \geq 0 \). Assume that \( \{k_n\} \) and \( \{r_n\} \) are two sequences satisfying \( r_n \geq b_1 n^r, k_n \leq b_2 n^k \), for some \( b_1, b_2, r, k > 0 \). Let

\[ \frac{1}{r_n} \max_{1 \leq j \leq k_n} |\sum_{i=1}^{j} X_{ni}| \rightarrow 0 \text{ completely as } n \rightarrow \infty. \]

If \( k + 1 < r \), then \( E|X|^{\frac{k+1}{r}} < \infty \).

**Proof.** Note that

\[ \frac{1}{r_n} \max_{1 \leq j \leq k_n} |\sum_{i=1}^{j} X_{ni}| \rightarrow 0 \text{ completely as } n \rightarrow \infty. \]

i.e.

\[ \sum_{n=1}^{\infty} P\left( \max_{1 \leq j \leq k_n} |\sum_{i=1}^{j} X_{ni}| \geq \epsilon r_n \right) < \infty, \quad \text{for all } \epsilon > 0. \quad (2.6) \]

Since \( \max_{1 \leq j \leq k_n} |X_{nj}| \leq 2 \max_{1 \leq j \leq k_n} |\sum_{i=1}^{j} X_{ni}| \), (2.6) implies

\[ \sum_{n=1}^{\infty} P\left( \max_{1 \leq j \leq n} |X_{nj}| \geq n \right) < \infty, \quad (2.7) \]

and

\[ P\left( \max_{1 \leq j \leq k_n} |X_{nj}| \geq r_n \right) \rightarrow 0 \text{ as } n \rightarrow \infty. \quad (2.8) \]

By (2.7) and (2.8), and using Lemma 1.2, we obtain that

\[ \sum_{i=1}^{k_n} P(|X_{ni}| > r_n) = O(1) P\left( \max_{1 \leq j \leq k_n} |X_{nj}| \geq r_n \right), \]
which, together with (2.7), it follows that
\[ \sum_{n=1}^{\infty} \sum_{i=1}^{k_n} P(|X_{ni}| > r_n) < \infty. \]

Thus, using the assumptions of Theorem 2.3, we have
\[ \sum_{n=1}^{\infty} k_n P(|X| > b_1 n^r) < \infty, \]
which is equivalent to \( E|X|^{k+1} < \infty \).

**Corollary 2.** Let \( \{X_{ni} | 1 \leq i \leq k_n, n \geq 1\} \) be an array of rowwise identically distributed \( NA \) random variables. Assume that \( \{k_n\} \) and \( \{r_n\} \) are two sequences satisfying \( r_n \sim n^r, k_n \sim n^k \), for some \( r, k > 0 \) where \( a_n \sim b_n \) means that \( C_1 a_n \leq b_n \leq C_2 a_n \) for large enough \( n \). If
(1) \( k + 1 < r \) or
(2) \( r \leq k + 1 < tr \) for some \( 0 < t \leq \frac{1}{2} \) and \( EX_{ni} = 0 \), then
\[ \frac{1}{r_n} \max_{1 \leq j \leq k_n} |\sum_{i=1}^{j} X_{ni}| \rightarrow 0 \quad \text{completely as } n \rightarrow \infty \] if and only if \( E|X|^{k+1} < \infty \).

**ACKNOWLEDGEMENTS.** This paper was supported by Wonkwang University Research in 2005.

**References**


Received: August 30, 2005