# A Remark on Complete Convergence for Arrays of Rowwise Negatively Associated Random Variables

Tien-Chung Hu	Andrei Volodin
National Tsing Hua University	University of Regina

**ABSTRACT** We obtain complete convergence result for arrays of rowwise negatively associated random variables, which extend and generalize the results of Hu *et al.* (1998), Hu *et al.* (2003), and Sung *et al.* (2005). As applications, some well-known results on independent random variables can be easily extended to the case of negatively associated random variables.

*Keywords* Negatively associated random variables; Array of rowwise negatively associated random variables; Complete convergence.

## **1. Introduction**

The concept of complete convergence of a sequence of random variables was introduced by Hsu and Robbins (1947) as follows. A sequence  $\{U_n, n \ge 1\}$  of random variables converges completely to the constant  $\theta$  if

$$\sum_{n=1}^{\infty} P\{|U_n - \theta| > \varepsilon\} < \infty$$
 for all  $\varepsilon > 0$ 

The first results concerning the complete convergence were due to Hsu and Robbins (1947) and Erdös (1949). Since then there were many authors who devoted the study to complete convergence for sums and weighted sums of independent random variables. Hu *et al.* (1998) announced the general result on complete convergence for arrays of rowwise independent random variables, presented as Theorem A below.

Received May 2006, revised August 2006.

Tien-Chung Hu is a Professor in the Department of Mathematics at the National Tsing Hua University, Hsinchu 300, Taiwan, ROC; email: <u>tchu@math.nthu.edu.tw</u>. Andrei Volodin is an Associate Professor in the Department of Mathematics and Statistics at the University of Regina, Regina, Saskatchewan, Canada S4S 0A2; email: <u>andrei@math.uregina.ca</u>.

Jointly Published by: *The International Chinese Association of Quantitative Management* and *Chung-Hwa Data Mining Society*, Taipei, Taiwan, Republic of China.

In the following we let  $\{X_{nk}, 1 \le k \le k_n, n \ge 1\}$  be an array of random variables defined on a probability space  $(\Omega, \mathcal{F}, P), \{k_n, n \ge 1\}$  be a sequence of positive integers such that  $\lim_{n\to\infty} k_n = \infty$ , and  $\{c_n, n \ge 1\}$  be a sequence of positive constants. For an event  $A \in \mathcal{F}$  we denote  $I\{A\}$  the indicator function. We should note that all the results of this paper remain true in the case  $k_n = \infty$  for some/all  $n \ge 1$ , provided the series  $\sum_{k=1}^{\infty} X_{nk}$  converges almost surely. Certainly, we should consider sup instead of max in the case of infinite sums.

**Theorem A** Let  $\{X_{ni}, 1 \le i \le k_n, n \ge 1\}$  be array of rowwise independent random variables. Suppose that for every  $\varepsilon > 0$  and some  $\delta > 0$ : (i)  $\sum_{n=1}^{\infty} c_n \sum_{i=1}^{k_n} P\{|X_{ni}| > \varepsilon\} < \infty$ , (ii) there exists  $j \ge 2$  such that

$$\sum_{n=1}^{\infty} c_n \left( \sum_{i=1}^{k_n} E X_{ni}^2 I\{|X_{ni}| \le \delta\} \right)^j < \infty,$$

(iii)  $\sum_{i=1}^{k_n} EX_{ni}I\{|X_{ni}| \le \delta\} \to 0 \text{ as } n \to \infty.$ Then

$$\sum_{n=1}^{\infty} c_n P\left\{ \left| \sum_{k=1}^{k_n} X_{nk} \right| > \varepsilon \right\} < \infty \text{ for all } \varepsilon > 0.$$

The paper Hu *et al.* [6] unifies and extends the ideas of previously obtained results on complete convergence. In the main results of Hu *et al.* [6], no assumptions are made concerning the existence of expected values or absolute moments of the random variables. The *proof* of Hu *et al.* [6] is mistakenly based on the fact that the assumptions of their theorem imply convergence in probability of the corresponding partial sums. Counterexamples to this proof were presented in Hu and Volodin [7] and Hu *et al.* [5]. We would like to stress that the both examples are the counterexamples to the *proof* of Theorem A, but not to the result. At the same time, they mentioned that the problem whether the Theorem A is true for any positive constants  $\{c_n, n \ge 1\}$ has remained open.

Many attempts to solve this problem led to weaker variants of this theorem. Hu *et al.* [5] gave a first partial solution to this question. Next partial solution was given by Kuczmaszewska [10] and the question was solved completely in Sung *et al.* [18], where the result is proved as stated in Hu *et al.* [6]. The approach of Sung *et al.* [18] is different from those of Hu *et al.* [6] in the sense that it does not use the symmetrization procedure.

The next paper which, to our best knowledge, presents the most general results on complete convergence for arrays of rowwise *independent* random variables is Kruglov *et al.* [9]. As it is shown in corollaries of that paper, the main results of Hu *et al.* [6], Hu *et al.* [5], Kuczmaszewska [10], and Sung *et al.* [18] can be derived easily from the first theorem of Kruglov *et al.* [9].

The proofs of the main results of Kruglov et al. [9], as well as of Hu et al. [6], Hu et al. [5], Kuczmaszewska [10], and Sung et al. [18], make use of the well-known Hoffmann-Jørgensen's inequality (cf. Hoffmann-Jørgensen [3]). Hoffmann-Jørgensen's maximal inequality is a powerful tool which has now become a standard technique in proving limit theorem for independent random variables.

The main purpose of this investigation is to extend Theorem A for the case of arrays of rowwise *negatively associated* random variables. But for associated random variables, especially for negative associated random variables, it is still an open question whether Hoffmann-Jørgensen's maximal inequality is true or not. In order to extend Theorem A to the case of negatively associated random variables, we need to find another way. In the paper, we use the exponential inequality for negatively associated random variables, which is established by Shao [15], cf. Lemma below.

Recall that a finite family of random variables  $\{X_i, 1 \le i \le n\}$  is said to be *negatively associated* (abbreviated to NA in the following) if for any disjoint subsets A and B of  $\{1, 2, \dots, n\}$ and any real coordinate-wise nondecreasing functions f on  $\mathbf{R}^A$  and g on  $\mathbf{R}^B$ ,

$$\operatorname{Cov}\left(f(X_i, i \in A), g(X_j, j \in B)\right) \le 0$$

whenever the covariance exists. An infinite family of random variables  $\{X_i, i \ge 1\}$  is NA if every finite subfamily is NA.

This concept was introduced by Joag-Dev and Proschan [8]. They also pointed out and proved in their paper that a number of well-known multivariate distributions possess the NA property. NA random variables have wide applications in reliability theory and multivariate statistical analysis. Recently Su et al. [17] showed that NA structure plays an important role in risk management. Because of these reasons the notions of NA random variables have received more and more attention in recent years. A great number of papers for NA random variables are now in literature. We refer to Joag-Dev and Proschan [8] for fundamental properties, Newman [14] for the central limit theorem, Matula [13] for the three series theorem, Shao and Su [16] for the law of the iterated logarithm, Shao [15] for moment inequalities, Liu et al. [12] for the Hàjek-Rènyi inequality and Barbour et al. [1] for the Poison approximation.

Shao [15] showed the following important exponent inequality of Kolmogorov's type.

**Lemma** Let  $\{X_i, 1 \le i \le n\}$  be a sequence of NA mean zero random variables with  $E|X_i|^2 < 1$  $\infty$  for every  $1 \le i \le n$  and  $B_n = \sum_{i=1}^n EX_i^2$ . Then for all  $\varepsilon > 0$ , a > 0

$$P\left\{\max_{1\leq k\leq n} |\sum_{i=1}^{k} X_{i}| \geq \varepsilon\right\} \leq 2P\left\{\max_{1\leq i\leq n} |X_{i}| > a\right\} + 4\exp\left\{-\frac{\varepsilon^{2}}{4(a\varepsilon + B_{n})}\right\}$$
$$\times \left[1 + \frac{2}{3}\ln\left(1 + \frac{a\varepsilon}{B_{n}}\right)\right]\right\}.$$

Note also that if  $\{X_i, i \ge 1\}$  is a sequence of NA random variables, then the truncation in the usual way, for example the sequence  $\{X_i I\{|X_i| < \delta\}, i \ge 1\}$ , where  $\delta > 0$ , is not necessary NA any more. We should use so-called *monotone truncation* (cf. the definition of random variables  $Y_{nk}$  in the proof of Theorem B) in order to preserve this property.

#### 2. Main Results, Corollaries, and Remarks

Now we state the main results. The proofs will be detailed in next section.

**Theorem B** Let  $\{X_{nk}, 1 \le k \le k_n, n \ge 1\}$  be an array of rowwise NA random variables such that conditions (i) and (ii) are satisfied. Then

$$\sum_{n=1}^{\infty} c_n P\left\{ \max_{1 \le m \le k_n} \left| \sum_{k=1}^{m} [X_{nk} - E(X_{nk}I\{|X_{nk}| \le \delta\})] \right| > \varepsilon \right\} < \infty$$

for any  $\varepsilon > 0$ .

The following two corollaries of Theorem B are immediate.

**Corollary 1** Let  $\{X_{nk}, 1 \le k \le k_n, n \ge 1\}$  be an array of rowwise NA random variables such that conditions (i), (ii), and (iii) are satisfied. Then

$$\sum_{n=1}^{\infty} c_n P\left\{ \left| \sum_{k=1}^{k_n} X_{nk} \right| > \varepsilon \right\} < \infty$$

for any  $\varepsilon > 0$ .

**Corollary 2.** Let  $\{X_{nk}, 1 \le k \le k_n, n \ge 1\}$  be an array of rowwise NA random variables such that conditions (i), (ii), and

(*iii*)'  $\max_{1 \le m \le k_n} |\sum_{i=1}^m EX_{ni}I\{|X_{ni}| \le \delta\}| \to 0 \text{ as } n \to \infty$ are satisfied. Then

$$\sum_{n=1}^{\infty} c_n P\left\{ \max_{1 \le m \le k_n} \left| \sum_{k=1}^m X_{nk} \right| > \varepsilon \right\} < \infty$$

for any  $\varepsilon > 0$ .

Similar argument as in Remark 2 of Hu et al. [6], we have

**Theorem C** Let  $\{X_{ni}, 1 \le i \le k_n, n \ge 1\}$  be an array of rowwise NA random variables with  $EX_{nk} = 0$  for all  $n \ge 1$  and  $1 \le k \le k_n$ . Let  $\phi(x)$  be a real function such that for some  $\delta > 0$ :

$$\sup_{x>\delta}\frac{x}{\phi(x)}<\infty \text{ and } \sup_{0\leq x\leq\delta}\frac{x^2}{\phi(x)}<\infty.$$

Suppose that for all  $\varepsilon > 0$ , conditions (i) and

(ii)' there exist  $j \ge 2$  such that

$$\sum_{n=1}^{\infty} c_n \left( \sum_{k=1}^{k_n} E\phi(|X_{nk}|) \right)^j < \infty \text{ and } \sum_{k=1}^{k_n} E\phi(|X_{nk}|) \to 0 \text{ as } n \to \infty$$

are satisfied, then

$$\sum_{n=1}^{\infty} c_n P\left\{ \max_{1 \le m \le k_n} \left| \sum_{k=1}^m X_{nk} \right| > \varepsilon \right\} < \infty$$

for any  $\varepsilon > 0$ .

**Remark 1.** Suppose  $\liminf_{n\to\infty} c_n > 0$ , then condition (i) in Corollary 2 is also necessary. In fact,

$$\sum_{n=1}^{\infty} c_n P\left\{ \max_{1 \le m \le k_n} \left| \sum_{k=1}^m X_{nk} \right| > \varepsilon \right\} < \infty$$

implies that

$$\sum_{n=1}^{\infty} c_n P\left\{ \max_{1 \le i \le k_n} |X_{ni}| > \varepsilon \right\} < \infty, \text{ for all } \varepsilon > 0,$$

hence we have  $P \{ \max_{1 \le i \le k_n} |X_{ni}| > \varepsilon \} \to 0$  as  $n \to \infty$ . By Lemma 2 of Liang [11], we have for sufficiently large *n*,

$$\sum_{i=1}^{k_n} P\{|X_{ni}| > \varepsilon\} \le CP\left\{\max_{1 \le i \le k_n} |X_{ni}| > \varepsilon\right\}$$

for some C > 0, therefore condition (i) holds.

**Remark 2.** By Theorems B and C we have that Corollaries 1 and 2 of Hu *et al.* [6] are also true if we assume negative association instead of independence and replace weighted sums by the maxima of weighted sums.

#### 3. Proofs

In the following, C always stands for a positive constant which may differ from one place to another.

**Proof of Theorem B.** The conclusion of the theorem is obvious if  $\sum_{n=1}^{\infty} c_n < \infty$ . Let  $\{Y_{nk}, 1 \le k \le k_n, n \ge 1\}$  be a monotone truncation of  $\{X_{nk}, 1 \le k \le k_n, n \ge 1\}$ , that is

$$Y_{nk} = X_{nk}I\{|X_{nk}| \le \delta\} + \delta I\{X_{nk} > \delta\} - \delta I\{X_{nk} < -\delta\}$$

where  $\delta > 0$  and  $1 \le k \le k_n, n \ge 1$ . Note that by Property 6 of Joag-Dev and Proschan [8] (applied twice) we can conclude that  $\{Y_{nk}, 1 \le k \le k_n, n \ge 1\}$  is an array of rowwise NA random variables.

For  $n \ge 1$  and  $1 \le m \le k_n$  let

$$T_m = \sum_{k=1}^m Y_{nk}, \ S_m = \sum_{k=1}^m X_{nk}, \ S'_m = \sum_{k=1}^m X_{nk} I\{|X_{nk}| \le \delta\},$$
$$A = \bigcap_{k=1}^{k_n} \{X_{nk} = X_{nk} I\{|X_{nk}| \le \delta\}\}, \ A^c = \bigcup_{k=1}^{k_n} \{X_{nk} \ne X_{nk} I\{|X_{nk}| \le \delta\}\}.$$

Note that for any  $n \ge 1$ 

$$\begin{split} &P\{\max_{1 \le m \le k_n} |S_m - ES'_m| > \varepsilon\} \\ &= P\left\{\{\max_{1 \le m \le k_n} |S_m - ES'_m| > \varepsilon\} \cap A^c\right\} + P\left\{\{\max_{1 \le m \le k_n} |S'_m - ES'_m| > \varepsilon\} \cap A\right\} \\ &\leq \sum_{k=1}^{k_n} P\{|X_{nk}| > \delta\} + P\{\max_{1 \le m \le k_n} |S'_m - ET_m| > \varepsilon\} \\ &\leq \sum_{k=1}^{k_n} P\{|X_{nk}| > \delta\} + P\{\max_{1 \le m \le k_n} |T_m - ET_m| > \varepsilon/2\} \\ &+ P\{\max_{1 \le m \le k_n} |(S'_m - T_m) - E(S'_m - T_m)| > \varepsilon/2\} \\ &\leq \sum_{k=1}^{k_n} P\{|X_{nk}| > \delta\} + P\{\max_{1 \le m \le k_n} |T_m - ET_m| > \varepsilon/2\} \\ &+ P\{\max_{1 \le m \le k_n} \delta\} \sum_{k=1}^{m} [I\{X_{nk} > \delta\} - I\{X_{nk} < -\delta\} \\ &- E(I\{X_{nk} > \delta\} - I\{X_{nk} < -\delta\})]] > \varepsilon/2\} \\ &\leq \sum_{k=1}^{k_n} P\{|X_{nk}| > \delta\} + P\{\max_{1 \le m \le k_n} |T_m - ET_m| > \varepsilon/2\} \\ &+ CE\max_{1 \le m \le k_n} |\sum_{k=1}^{m} [I\{X_{nk} > \delta\} - I\{X_{nk} < -\delta\} \\ &- E(I\{X_{nk} > \delta\} - I\{X_{nk} < -\delta\})]] (by Markov's inequality) \\ &\leq \sum_{k=1}^{k_n} P\{|X_{nk}| > \delta\} + P\{\max_{1 \le m \le k_n} |T_m - ET_m| > \varepsilon/2\} \\ &+ CE\max_{1 \le m \le k_n} \sum_{k=1}^{m} [I\{X_{nk} > \delta\} + I\{X_{nk} < -\delta\} + P\{X_{nk} > \delta\} + P\{X_{nk} < -\delta\}] \\ &\leq \sum_{k=1}^{k_n} P\{|X_{nk}| > \delta\} + P\{\max_{1 \le m \le k_n} |T_m - ET_m| > \varepsilon/2\} \\ &+ CE\max_{1 \le m \le k_n} \sum_{k=1}^{m} [I\{X_{nk} > \delta\} + I\{X_{nk} < -\delta\} + P\{X_{nk} > \delta\} + P\{X_{nk} < -\delta\}] \\ &\leq \sum_{k=1}^{k_n} P\{|X_{nk}| > \delta\} + P\{\max_{1 \le m \le k_n} |T_m - ET_m| > \varepsilon/2\} \\ &+ CE\sum_{k=1}^{k_n} [I\{X_{nk} > \delta\} + I\{X_{nk} < -\delta\} + P\{X_{nk} > \delta\} + P\{X_{nk} < -\delta\}] \end{aligned}$$

A Remark on Complete Convergence for Arrays of Rowwise .....

$$\leq P\{\max_{1\leq m\leq k_n} |T_m - ET_m| > \varepsilon/2\} + (1+2C)\sum_{k=1}^{k_n} P\{|X_{nk}| > \delta\}.$$

By (i) it is enough to prove that for all  $\varepsilon > 0$ 

$$\sum_{n=1}^{\infty} c_n P\{\max_{1 \le m \le k_n} |T_m - ET_m| > \varepsilon\} < \infty.$$

Let  $B_n = \sum_{k=1}^{k_n} \operatorname{Var}(Y_{nk})$ . For any  $\varepsilon > 0$  and a > 0 set

$$\begin{split} \mathbf{N}_1 &= \{n : B_n > a\varepsilon\},\\ \mathbf{N}_2 &= \left\{n : \sum_{k=1}^{k_n} P\{|X_{nk}| > \delta\} > \min\{1, a\varepsilon/(2\delta^2), a/(4\delta)\}\right\},\\ \mathbf{N}_3 &= \left\{n : \sum_{k=1}^{k_n} EX_{nk}^2 I\{|X_{nk}| \le \delta\} > \min\{a\varepsilon/2, a^2/16\}\right\},\\ \mathbf{N}_4 &= \mathbf{N} - (\mathbf{N}_2 \cup \mathbf{N}_3). \end{split}$$

Since

$$B_n = \sum_{k=1}^{k_n} (EY_{nk}^2 - (EY_{nk})^2) \le \sum_{k=1}^{k_n} EY_{nk}^2$$
  
=  $\delta^2 \sum_{k=1}^{k_n} P\{|X_{nk}| > \delta\} + \sum_{k=1}^{k_n} EX_{nk}^2 I\{|X_{nk}| \le \delta\},$ 

we have  $\mathbf{N}_1 \subseteq \mathbf{N}_2 \cup \mathbf{N}_3$ . Note that

$$\sum_{n \in \mathbf{N}_2 \cup \mathbf{N}_3} c_n P\{\max_{1 \le m \le k_n} |T_m - ET_m| > \varepsilon\} \le \sum_{n \in \mathbf{N}_2 \cup \mathbf{N}_3} c_n$$

$$\le \left(\min\{1/2, a\varepsilon/(4\delta^2), a/(8\delta)\}\right)^{-1} \sum_{n=1}^{\infty} c_n \sum_{k=1}^{k_n} P\{|X_{nk}| > \delta\}$$

$$+ \left(\min\{a\varepsilon/4, a^2/32\}\right)^{-j} \sum_{n=1}^{\infty} c_n \left(\sum_{k=1}^{k_n} EX_{nk}^2 I\{|X_{nk}| \le \delta\}\right)^j < \infty.$$

Hence it is sufficient to prove that

$$\sum_{n\in\mathbb{N}_4}c_n P\left\{\max_{1\leq m\leq k_n}|\sum_{k=1}^m(Y_{nk}-EY_{nk})|>\varepsilon\right\}<\infty.$$

By Lemma we have that

$$\sum_{n \in \mathbf{N}_4} c_n P \left\{ \max_{1 \le m \le k_n} \left| \sum_{k=1}^m (Y_{nk} - EY_{nk}) \right| > \varepsilon \right\}$$

$$\leq \sum_{n \in \mathbf{N}_4} c_n \left[ 2P \left\{ \max_{1 \leq k \leq k_n} |Y_{nk} - EY_{nk}| > a \right\} + 4 \exp \left\{ -\frac{\varepsilon^2}{4(a\varepsilon + B_n)} \left[ 1 + \frac{2}{3} \ln \left( 1 + \frac{a\varepsilon}{B_n} \right) \right] \right\} \right]$$

Note that for any  $n \in \mathbf{N}_4$ 

$$\max_{1 \le k \le k_n} |EY_{nk}| \le \max_{1 \le k \le k_n} E|Y_{nk}|$$

$$\le \max_{1 \le k \le k_n} (\delta P\{|X_{nk}| > \delta\} + E|X_{nk}|I\{|X_{nk}| \le \delta\})$$

$$\le \delta \sum_{k=1}^{k_n} P\{|X_{nk}| > \delta\} + \left(\sum_{k=1}^{k_n} EX_{nk}^2 I\{|X_{nk}| \le \delta\}\right)^{1/2}$$

$$\le \delta \min\{1, a\varepsilon/(2\delta^2), a/(4\delta)\} + \left(\min\{a\varepsilon/2, a^2/16\}\right)^{1/2}$$

$$(\text{since } n \notin \mathbf{N}_2 \text{ and } n \notin \mathbf{N}_3)$$

$$\le a/4 + a/4 = a/2.$$

This implies that for any  $n \in \mathbf{N}_4$ ,  $\max_{1 \le k \le k_n} |EY_{nk}| \le a/2$  and

$$\sum_{n \in \mathbf{N}_4} c_n P\left\{\max_{1 \le k \le k_n} |Y_{nk} - EY_{nk}| > a\right\}$$
  
$$\leq \sum_{n=1}^{\infty} c_n P\left\{\max_{1 \le k \le k_n} |Y_{nk}| > a/2\right\}$$
  
$$\leq \sum_{n=1}^{\infty} c_n \sum_{k=1}^{k_n} P\left\{|X_{nk}| > \min\{\delta, a/2\}\right\} < \infty \text{ (by (i)).}$$

Therefore it is sufficient to prove that

$$\sum_{n\in\mathbb{N}_4}c_n\exp\left\{-\frac{\varepsilon^2}{4(a\varepsilon+B_n)}\left[1+\frac{2}{3}\ln\left(1+\frac{a\varepsilon}{B_n}\right)\right]\right\}<\infty.$$

When  $n \in \mathbf{N}_4$ ,  $B_n \le a\varepsilon$  and  $\sum_{k=1}^{k_n} P\{|X_{nk}| > \delta\} \le 1$ . Let  $a = \varepsilon/(12j)$ .

$$\sum_{n \in \mathbf{N}_{4}} c_{n} \exp\left\{-\frac{\varepsilon^{2}}{4(a\varepsilon + B_{n})}\left[1 + \frac{2}{3}\ln\left(1 + \frac{a\varepsilon}{B_{n}}\right)\right]\right\}$$

$$\leq \sum_{n \in \mathbf{N}_{4}} c_{n} \exp\left\{-\frac{\varepsilon^{2}}{8a\varepsilon}\left[1 + \frac{2}{3}\ln\left(1 + \frac{a\varepsilon}{B_{n}}\right)\right]\right\}$$

$$\leq \exp\{-\frac{3}{2}j\}\sum_{n \in \mathbf{N}_{4}} c_{n} \exp\left\{-j\ln\left(\frac{B_{n} + a\varepsilon}{B_{n}}\right)\right\}$$

$$= C\sum_{n \in \mathbf{N}_{4}} c_{n} \left(\frac{B_{n}}{B_{n} + a\varepsilon}\right)^{j} \leq C\sum_{n \in \mathbf{N}_{4}} c_{n} \left(\frac{B_{n}}{a\varepsilon}\right)^{j} = C\sum_{n \in \mathbf{N}_{4}} c_{n} (B_{n})^{j}$$

.

$$\leq C \sum_{n \in \mathbf{N}_{4}} c_{n} \left[ \delta^{2} \sum_{k=1}^{k_{n}} P\{|X_{nk}| > \delta\} + \sum_{k=1}^{k_{n}} EX_{nk}^{2}I\{|X_{nk}| \le \delta\} \right]^{J}$$

$$\leq C \sum_{n \in \mathbf{N}_{4}} c_{n} \left\{ \delta^{2j} \left[ \sum_{k=1}^{k_{n}} P\{|X_{nk}| > \delta\} \right]^{J} + \left[ \sum_{k=1}^{k_{n}} EX_{nk}^{2}I\{|X_{nk}| \le \delta\} \right]^{J} \right\}$$

$$\leq C \sum_{n \in \mathbf{N}_{4}} c_{n} \left\{ \delta^{2j} \sum_{k=1}^{k_{n}} P\{|X_{nk}| > \delta\} + \left[ \sum_{k=1}^{k_{n}} EX_{nk}^{2}I\{|X_{nk}| \le \delta\} \right]^{J} \right\}$$

$$(\text{since } \sum_{k=1}^{k_{n}} P\{|X_{nk}| > \delta\} \le 1)$$

$$\leq C \sum_{n=1}^{\infty} c_{n} \left\{ \delta^{2j} \sum_{k=1}^{k_{n}} P\{|X_{nk}| > \delta\} + \left[ \sum_{k=1}^{k_{n}} EX_{nk}^{2}I\{|X_{nk}| \le \delta\} \right]^{J} \right\} < \infty$$

by the assumptions. The proof is completed.  $\Box$ 

*Proof of Theorem C.* In view of Corollary 2 it is enough to show that conditions (ii) and (iii)' are satisfied.

For (ii) note that

$$\sum_{k=1}^{k_n} E X_{nk}^2 I\{|X_{nk}| \le \delta\} \le \sup_{0 \le x \le \delta} \frac{x^2}{\phi(x)} \sum_{k=1}^{k_n} E \phi(|X_{nk}|).$$

For (iii)' since  $EX_{nk} = 0$  it follows that

$$\max_{1 \le m \le k_n} |\sum_{k=1}^m EX_{nk} I\{|X_{nk}| \le \delta\}| = \max_{1 \le m \le k_n} |\sum_{k=1}^m EX_{nk} I\{|X_{nk}| > \delta\}|$$
  
$$\le \sup_{x > \delta} \frac{x}{\phi(x)} \sum_{k=1}^{k_n} E\phi(|X_{nk}|) \to 0 \text{ as } n \to \infty.$$

### References

- [1] Barbour, A. D., Holst, L., and Janson, S. (1992). Poisson approximation, Oxford University Press.
- [2] Erdös, P. (1949). On a theorem of Hsu and Robbins, Ann. Math. Statist., 20, 286-291.
- [3] Hoffmann-Jørgensen, J. (1974). Sums of independent Banach space valued random variables, *Studia Math.*, **52**, 159-186.
- [4] Hsu, P. L. and Robbins, H. (1947). Complete convergence and the law of large numbers, *Proc. Nat. Acad. Sci.*, **33**, 25-31.

- [5] Hu, T.-C., Ordóñez Cabrera, M., Sung, S. H., and Volodin, A. (2003). Complete convergence for arrays of rowwise independent random variables, *Commun. Korean Math. Soc.*, 18, 375-383.
- [6] Hu, T.-C., Szynal, D., and Volodin, A. (1998). A note on complete convergence for arrays, *Statist. Probab. Lett.*, **38**, 27-31.
- [7] Hu, T.-C. and Volodin, A. (2000). Addendum to "A note on complete convergence for arrays", *Statist. Probab. Lett.*, **47**, 209-211.
- [8] Joag-Dev, K. and Proschan F. (1983). Negative association of random variables with applications, *Ann. Statist.*, **11**, 286-295.
- [9] Kruglov, V., Volodin, A., and Hu, T.-C. (2006). On complete convergence for arrays, *Statist. Probab. Lett.*, *76*, 1631-1640.
- [10] Kuczmaszewska, A. (2004). On some conditions for complete convergence for arrays, *Statist. Probab. Lett*, **66**, 399-405.
- [11] Liang, H. Y. (2000). Complete convergence for weighted sums of negatively associated random variables, *Statist. Probab. Lett.*, **45**, 85-95.
- [12] Liu, J. J., Gan, S. X., and Chen, P. Y. (1999). The Hàjek–Rènyi inequality for NA random variables and its application, *Statist. Probab. Lett.*, **43**, 99-105.
- [13] Matula, P. (1992). A note on the almost sure convergence of sums of negatively dependent random variables, *Statist. Probab. Lett.*, **15**, 209-213.
- [14] Newman, C. M. (1984). Asymptotic independence and limit theorems for positive and negatively dependent random variables, in: Y. L. Tong (ed.), Inequalities and Probability, Institute of Mathematical Statistics, Hayward, CA., 127-140.
- [15] Shao, Q. M. (2000). A comparison theorem on moment inequalities between negatively associated and independent random variables, *J. of Theoretical Probability*, **13**, 343-356.
- [16] Shao, Q. M. and Su, C. (1999). The law of the iterated logarithm for negatively associated random variables, *Stochastic Processes and Their Applications*, **83**, 139-148.
- [17] Su, C., Tiang, T., Tang, Q. H., and Liang, H. Y. (2002). On the safety of NA dependence structure, *Chinese Journal of Applied Probability and Statistics*, **16**, 400-404.
- [18] Sung, S. H., Hu, T.-C., and Volodin, A. (2005). More on complete convergence for arrays, *Statist. Probab. Lett.*, **71**, 303-311.