

# Strong Laws of Large Numbers for Weighted Sums of Random Variables

Soo Hak Sung

Department of Applied Mathematics, Pai Chai University

## 확률변수의 가중합에 대한 대수의 강법칙

성수학

배재대학교 자연과학대학 자연과학부 전산정보수학 전공

Strong laws of large numbers are established for the weighted sums of i.i.d. random variables which have higher order moment condition. One of the results of Sung (2001) is extended.

높은 차수의 적률을 갖는 i.i.d. 확률 변수의 가중합에 대한 강대수 법칙을 유도한다. 또한 Sung (2001)의 결과 중 하나를 확장한다.

Key words : Strong laws of large numbers, weight sums, i.i.d. random variables, almost sure convergence

### 1. Introduction

Let  $\{X, X_n, n \geq 1\}$  be a sequence of i.i.d. random variables and  $\{a_{ni}, 1 \leq i \leq n, n \geq 1\}$  be an array of constants. The almost sure (a.s.) limiting behavior of weighted sums

$\sum_{i=1}^n a_{ni}X_i$  was studied by many authors (see, Bai and Cheng (2000), Choi and Sung (1987), Cuzick (1995), Sung (2001), Wu (1999)). Recently, Sung (2001) proved the strong law of large numbers

$$\sum_{i=1}^n a_{ni}X_i/b_n \rightarrow 0 \text{ a.s.} \tag{1}$$

when  $\{X, X_n, n \geq 1\}$  is a sequence of i.i.d. random variables with  $EX=0$  and

$$E[\exp(h|X|^\gamma)] < \infty \text{ for some } h > 0 \text{ and some } 0 < \gamma \leq 1, \quad \forall \tag{2}$$

and  $\{a_{ni}, 1 \leq i \leq n, n \geq 1\}$  is an array of constants satisfying

$$A_\alpha = \limsup_{n \rightarrow \infty} A_{\alpha, n} < \infty, \quad A_{\alpha, n} = \sum_{i=1}^n |a_{ni}|^\alpha / n \tag{3}$$

for some  $1 < \alpha \leq 2$ , where  $b_n = n^{1/\alpha}(\log n)^{1/\gamma + \beta}$  ( $\beta > 0$ ).

In this paper, we extend the result of Sung (2001).

## 2. Main Result

We state and prove our main result.

**Theorem 1.** Let  $0 < \gamma \leq 1$ . Let  $\{X, X_n, n \geq 1\}$  be i.i.d. random variables satisfying  $EX=0$  and (2). Let  $\{a_{ni}, 1 \leq i \leq n, n \geq 1\}$  be an array of constants such that

$$\max_{1 \leq i \leq n} |a_{ni}| = o\left(\frac{1}{(\log n)^{1/\gamma}}\right) \tag{4}$$

and

$$\sum_{i=1}^n a_{ni}^2 = o\left(\frac{1}{\log n}\right). \tag{5}$$

Then  $\sum_{i=1}^n a_{ni}X_i \rightarrow 0$  a.s.

**Proof.** By (5), we have

$$E\left(\sum_{i=1}^n a_{ni}X_i\right)^2 = \sum_{i=1}^n a_{ni}^2 EX^2 = o\left(\frac{1}{\log n}\right).$$

It follows that

$$\sum_{i=1}^n a_{ni} X_i \rightarrow 0 \text{ in probability.}$$

Hence, by Theorem 3.2.1 in Stout (1974), it suffices to prove that

$$\sum_{i=1}^n a_{ni} X_i^s \rightarrow 0 \text{ a.s.,}$$

where  $(X_n^s)$  is a symmetric version of  $(X_n)$ . So we need only to prove the result for  $(X_n)$  symmetric.

Define  $X_{ni}' = X_i I(|X_i| \leq (h^{-1} \log n)^{1/\gamma})$  and  $X_{ni}'' = X_i - X_{ni}'$  for  $1 \leq i \leq n$  and  $n \geq 1$ . Note that  $E[e^{h|X|^{\gamma}}] < \infty$  is equivalent to  $\sum_{n=1}^{\infty} P(|X_n| > (h^{-1} \log n)^{1/\gamma}) < \infty$ .

Hence, by the Borel-Cantelli lemma,  $\sum_{i=1}^n |X_{ni}''|$  is bounded a.s. It follows by (4) that

$$\left| \sum_{i=1}^n a_{ni} X_{ni}'' \right| \leq \max_{1 \leq i \leq n} |a_{ni}| \sum_{i=1}^n |X_{ni}''| = o(1) \frac{\sum_{i=1}^n |X_{ni}''|}{(\log n)^{1/\gamma}} \rightarrow 0$$

a.s. as  $n \rightarrow \infty$ . Hence, to prove the theorem, it remains to prove that

$$\sum_{i=1}^n a_{ni} X_{ni}' \rightarrow 0 \text{ a.s.} \tag{6}$$

From an inequality  $e^x \leq 1 + x + \frac{x^2}{2} e^{|x|}$  for all  $x \in \mathbb{R}$ , we have

$$E[e^{t a_{ni} X_{ni}'}] \leq 1 + E\left[\frac{1}{2} t^2 a_{ni}^2 |X_{ni}'|^2 e^{4 a_{ni} |X_{ni}'|}\right]$$

for any  $t > 0$ . Let  $\varepsilon > 0$  be given. By putting  $t = 2 \log n / \varepsilon$ , we obtain

$$\begin{aligned} E[e^{t a_{ni} X_{ni}'}] &\leq 1 + \left(\frac{1}{2}\right) \left(\frac{2}{\varepsilon}\right)^2 a_{ni}^2 \log^2 n E[|X_{ni}'|^2 e^{\frac{2}{\varepsilon} |a_{ni} X_{ni}'| \log n}] \\ &= 1 + \left(\frac{1}{2}\right) \left(\frac{2}{\varepsilon}\right)^2 a_{ni}^2 \log^2 n E[e^{\log |X_{ni}'|^2} e^{\frac{2}{\varepsilon} |a_{ni} X_{ni}'| \log n}]. \end{aligned}$$

Since  $|X_{ni}'| = |X_i| I(|X_i| \leq (h^{-1} \log n)^{1/\gamma})$ , we have that

$$\begin{aligned}
 & E[e^{\log|X_{ni}|^2} e^{\frac{2}{\varepsilon}|a_{ni}|X_{ni}|\log n}] \\
 &= \sum_{j=2}^n E[e^{\log|X|^2} I(|X| \leq (h^{-1} \log n)^{1/\gamma}) e^{\frac{2}{\varepsilon}|a_{ni}|X} I(|X| \leq (h^{-1} \log n)^{1/\gamma}) \log n \times \\
 &\quad I((h^{-1} \log(j-1))^{1/\gamma} < |X| \leq (h^{-1} \log j)^{1/\gamma})] \\
 &\leq \sum_{j=2}^n e^{2 \log(h^{-1} \log j)^{1/\gamma}} e^{\frac{2}{\varepsilon}|a_{ni}|(h^{-1} \log j)^{1/\gamma} \log n} P((h^{-1} \log(j-1))^{1/\gamma} < |X| \leq (h^{-1} \log j)^{1/\gamma}).
 \end{aligned} \tag{7}$$

Observe that

$$\begin{aligned}
 & 2 \log(h^{-1} \log j)^{1/\gamma} + \frac{2}{\varepsilon} |a_{ni}| (h^{-1} \log j)^{1/\gamma} \log n \\
 &= 2 \log(h^{-1} \log j)^{1/\gamma} + \frac{2}{\varepsilon} \frac{o(1)}{(\log n)^{1/\gamma}} (h^{-1} \log j)^{1/\gamma} \log n \\
 &= 2 \log(h^{-1} \log j)^{1/\gamma} + o(1) \frac{\log n}{(\log n)^{1/\gamma}} (\log j)^{1/\gamma} \\
 &\leq \log(j-1)
 \end{aligned}$$

for all sufficiently large  $j$  ( $j_0 < j \leq n$ ). Thus we have by (7) that

$$\begin{aligned}
 & E[e^{\log|X_{ni}|^2} e^{\frac{2}{\varepsilon}|a_{ni}|X_{ni}|\log n}] \\
 &\leq \sum_{j=j_0}^n e^{2 \log(h^{-1} \log j)^{1/\gamma}} e^{\frac{2}{\varepsilon}|a_{ni}|(h^{-1} \log j)^{1/\gamma} \log n} P((h^{-1} \log(j-1))^{1/\gamma} < |X| \leq (h^{-1} \log j)^{1/\gamma}) \\
 &\quad + \sum_{j=j_0+1}^n e^{\log(j-1)} P((h^{-1} \log(j-1))^{1/\gamma} < |X| \leq (h^{-1} \log j)^{1/\gamma}) \\
 &\leq C + E[e^{\#|X|^{\gamma}}]
 \end{aligned}$$

for some constant  $C > 0$ . It follows that

$$\begin{aligned}
 E[e^{t a_{ni} X_{ni}}] &\leq 1 + \left(\frac{1}{2}\right) \left(\frac{2}{\varepsilon}\right)^2 a_{ni}^2 \log^2 n (C + E[e^{\#|X|^{\gamma}}]) \\
 &\leq \exp\left\{\left(\frac{1}{2}\right) \left(\frac{2}{\varepsilon}\right)^2 a_{ni}^2 \log^2 n (C + E[e^{\#|X|^{\gamma}}])\right\}.
 \end{aligned}$$

Hence, by (5)

$$\begin{aligned}
 P\left(\sum_{i=1}^n a_{ni} X_{ni} > \varepsilon\right) &= P\left(e^{t \sum_{i=1}^n a_{ni} X_{ni}} > e^{t\varepsilon}\right) \\
 &\leq e^{-t\varepsilon} E\left[e^{t \sum_{i=1}^n a_{ni} X_{ni}}\right] \\
 &= e^{-t\varepsilon} \prod_{i=1}^n E\left[e^{t a_{ni} X_{ni}}\right] \\
 &= e^{-t\varepsilon} \prod_{i=1}^n \exp\left\{\left(\frac{1}{2}\right) \left(\frac{2}{\varepsilon}\right)^2 a_{ni}^2 \log^2 n (C + E[e^{\#|X|^{\gamma}}])\right\} \\
 &= e^{-t\varepsilon} \exp\left\{\left(\frac{1}{2}\right) \left(\frac{2}{\varepsilon}\right)^2 \log^2 n (C + E[e^{\#|X|^{\gamma}}]) \sum_{i=1}^n a_{ni}^2\right\} \\
 &= e^{-2 \log n} \exp\{o(1) \log n\} \\
 &= \exp\{(-2 + o(1)) \log n\},
 \end{aligned}$$

which implies that

$$\sum_{n=1}^{\infty} P\left(\sum_{i=1}^n a_{ni} X_{ni}' > \varepsilon\right) < \infty.$$

By the Borel-Cantelli lemma, we have

$$\limsup_{n \rightarrow \infty} \sum_{i=1}^n a_{ni} X_{ni}' \leq 0 \text{ a.s.}$$

By replacing  $X_{ni}'$  by  $-X_{ni}'$  from the above statement, we obtain

$$\liminf_{n \rightarrow \infty} \sum_{i=1}^n a_{ni} X_{ni}' \geq 0 \text{ a.s.}$$

Hence (6) is satisfied.

The following corollary is proved by Sung (2001).

**Corollary 2.** Let  $0 < \gamma \leq 1$ . Let  $\{X, X_n, n \geq 1\}$  be i.i.d. random variables satisfying  $EX = 0$  and (2). Let  $\{a_{ni}, 1 \leq i \leq n, n \geq 1\}$  be an array of constants satisfying (3) for some  $1 < \alpha \leq 2$ . Then for  $b_n = n^{1/\alpha} (\log n)^{1/\gamma + \beta}$  ( $\beta > 0$ )

$$\sum_{i=1}^n a_{ni} X_i / b_n \rightarrow 0 \text{ a.s.}$$

**Proof.** Note that

$$\left| \frac{a_{ni}}{b_n} \right| \leq \frac{\left(\sum_{i=1}^n |a_{ni}|^\alpha\right)^{1/\alpha}}{b_n} = \frac{(nA_{a,n}^\alpha)^{1/\alpha}}{n^{1/\alpha} (\log n)^{1/\gamma + \beta}} = o\left(\frac{1}{(\log n)^{1/\gamma}}\right),$$

and

$$\sum_{i=1}^n \left| \frac{a_{ni}}{b_n} \right|^2 \leq \frac{\left(\sum_{i=1}^n |a_{ni}|^\alpha\right)^{2/\alpha}}{n^{2/\alpha} ((\log n)^{1/\gamma + \beta})^2} = \frac{(nA_{a,n}^\alpha)^{2/\alpha}}{n^{2/\alpha} ((\log n)^{1/\gamma + \beta})^2} = o\left(\frac{1}{\log n}\right)$$

since  $2(1/\gamma + \beta) > 1$ . Thus it follows by Theorem 1 that  $\sum_{i=1}^n a_{ni} X_i / b_n \rightarrow 0$  a.s.

## Acknowledgment

This study was financially supported by a Central Research Fund for the year of 2002 from Pai Chai University.

## References

- Bai, Z.D., and Cheng, P.E., 2000. Marcinkiewicz strong laws for linear statistics. *Statist. Probab. Lett.*, 46: 105-112.
- Choi, B.D., Sung, S.H., 1987. Almost sure convergence theorems of weighted sums of random variables. *Stochastic Anal. Appl.*, 5: 365-377.
- Cuzick, J., 1995. A strong law for weighted sums of i.i.d. random variables. *J. Theoret. Probab.*, 8: 625-641.
- Hsu, P.L., Robbins, H., 1947. Complete convergence and the law of large numbers. *Proc. Nat. Acad. Sci. U.S.A.*, 33: 25-31.
- Rosalsky, A., Sreehari, M., 1998. On the limiting behavior of randomly weighted partial sums. *Statist. Probab. Lett.*, 40: 403-410.
- Stout, W.F., 1974. *Almost Sure Convergence*. Academic Press, New York.
- Sung, S.H., 2001. Strong laws for weighted sums of i.i.d. random variables. *Statist. Probab. Lett.*, 52: 413-419.
- Wu, W.B., 1999. On the strong convergence of a weighted sum. *Statist. Probab. Lett.*, 44: 19-22.