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# An almost sure central limit theorem of products of partial sums for $\rho^-$ -mixing sequences

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## Abstract

Let  $\{X_n, n \geq 1\}$  be a strictly stationary  $\rho^-$ -mixing sequence of positive random variables with  $EX_1 = \mu > 0$  and  $Var(X_1) = \sigma^2 < \infty$ . Denote  $S_n = \sum_{i=1}^n X_i$  and  $\gamma = \frac{\sigma}{\mu}$  the coefficient of variation. Under suitable conditions, by the central limit theorem of weighted sums and the moment inequality we show that

$$\forall x = \lim_{n \rightarrow \infty} \frac{1}{\log n} \sum_{k=1}^n \frac{1}{k} I \left\{ \left( \prod_{i=1}^k \frac{S_i}{i\mu} \right)^{\frac{1}{\gamma\sigma k}} \right\} = F(x) \text{ a.s.},$$

where  $\sigma_k^2 = Var(S_{k,k})$ ,  $S_{k,k} = \sum_{i=1}^k b_{i,k} Y_i$ ,  $b_{i,k} = \sum_{j=i}^k \frac{1}{j}$ ,  $i \leq k$  with

$b_{i,k} = 0, i > k$ ,  $Y_i = \frac{X_i - \mu}{\sigma}$ ,  $F(x)$  is the distribution function of the random variable  $e^{\sqrt{2}\mathcal{N}}$ , and  $\mathcal{N}$  is a standard normal random variable.

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## 1 Introduction and main results

For a random variable  $X$ , define  $\|X\|_p = (E|X|^p)^{1/p}$ . For two nonempty disjoint sets  $S, T \subset N$ , we define  $\text{dist}(S, T)$  to be  $\min\{|j - k|; j \in S, k \in T\}$ . Let  $\sigma(S)$  be the  $\sigma$ -field generated by  $\{X_k, k \in S\}$ , and define  $\sigma(T)$  similarly. Let  $\mathcal{C}$  be a class of functions which are coordinatewise increasing. For any real number  $x$ ,  $x^+$ , and  $x^-$  denote its positive and negative part, respectively, (except for some special definitions, for examples,  $\rho^-(s)$ ,  $\rho^-(S, T)$ , etc.). For random variables  $X, Y$ , define

$$\rho^-(X, Y) = 0 \vee \sup \frac{\text{Cov}(f(X), g(Y))}{(\text{Var}f(X))^{\frac{1}{2}} (\text{Var}g(Y))^{\frac{1}{2}}},$$

where the sup is taken over all  $f, g \in \mathcal{C}$  such that  $E(f(X))^2 < \infty$  and  $E(g(Y))^2 < \infty$ .

A sequence  $\{X_n, n \geq 1\}$  is called negatively associated (NA) if for every pair of disjoint subsets  $S, T$  of  $N$ ,

$$\text{Cov}\{f(X_i, i \in S), g(X_j, j \in T)\} \leq 0,$$

whenever  $f, g \in \mathcal{C}$ .

A sequence  $\{X_n, n \geq 1\}$  is called  $\rho^*$ -mixing if

$$\rho^*(s) = \sup \{ \rho(S, T); S, T \subset N, \text{dist}(S, T) \geq s \} \rightarrow 0 \text{ as } s \rightarrow \infty,$$

where

$$\rho(S, T) = \sup \{ |E(f - Ef)(g - Eg)| / (\|f - Ef\|_2 \cdot \|g - Eg\|_2); f \in L_2(\sigma(S)), g \in L_2(\sigma(T)) \}.$$

A sequence  $\{X_n, n \geq 1\}$  is called  $\rho^-$ -mixing, if

$$\rho^-(s) = \sup \{ \rho^-(S, T); S, T \subset N, \text{dist}(S, T) \geq s \} \rightarrow 0 \text{ as } s \rightarrow \infty.$$

where,

$$\rho^-(S, T) = 0 \vee \sup \left\{ \frac{\text{Cov} \{ f(X_i, i \in S), g(X_i, j \in T) \}}{\sqrt{\text{Var} \{ f(X_i, i \in S) \} \text{Var} \{ g(X_i, j \in T) \}}}; f, g \in C \right\}.$$

The concept of  $\rho^-$ -mixing random variables was proposed in 1999 (see [1]). Obviously,  $\rho^-$ -mixing random variables include NA and  $\rho^*$ -mixing random variables, which have a lot of applications, their limit properties have aroused wide interest recently, and a lot of results have been obtained, such as the weak convergence theorems, the central limit theorems of random fields, Rosenthal-type moment inequality, see [1-4]. Zhou [5] studied the almost sure central limit theorem of  $\rho^-$ -mixing sequences by the conditions provided by Shao: on the conditions of central limit theorem, and if  $\varepsilon_0 > 0$ ,  $\text{Var} \left( \sum_{i=1}^n \frac{1}{i} f \left( \frac{S_i}{\sigma_i} \right) \right) = O \left( \log^{2-\varepsilon_0} n \right)$ , where  $f$  is Lipschitz function. In this article, we study the almost sure central limit theorem of products of partial sums for  $\rho^-$ -mixing sequence by the central limit theorem of weighted sums and moment inequality.

Here and in the sequel, let  $b_{k,n} = \sum_{i=k}^n \frac{1}{i}$ ,  $k \leq n$  with  $b_{k,n} = 0$ ,  $k > n$ . Suppose  $\{X_n, n \geq 1\}$  be a strictly stationary  $\rho^-$ -mixing sequence of positive random variables with  $EX_1 = \mu > 0$  and  $\text{Var}(X_1) = \sigma^2 < \infty$ . Let  $\tilde{S}_n = \sum_{k=1}^n Y_k$  and  $S_{n,n} = \sum_{k=1}^n b_{k,n} Y_k$ , where  $Y_k = \frac{X_k - \mu}{\sigma}$ ,  $k \geq 1$ . Let  $\sigma_n^2 = \text{Var}(S_{n,n})$ , and  $C$  denotes a positive constant, which may take different values whenever it appears in different expressions. The following are our main results.

**Theorem 1.1** Let  $\{X_n, n \geq 1\}$  be a defined as above with  $0 < E|X_1|^r < \infty$  for a certain  $r > 2$ , denote  $S_n = \sum_{i=1}^n X_i$  and  $\gamma = \frac{\sigma}{\mu}$  the coefficient of variation. Assume that

$$(a_1) \sigma_1^2 = EX_1^2 + 2 \sum_{n=2}^{\infty} \text{Cov}(X_1, X_n) > 0,$$

$$(a_2) \sum_{n=2}^{\infty} |\text{Cov}(X_1, X_n)| < \infty,$$

$$(a_3) \rho^-(n) = O(\log^{-\delta} n), \exists \delta > 1,$$

$$(a_4) \inf_{n \in N} \frac{\sigma_n^2}{n} > 0.$$

Then

$$\forall x \quad \lim_{n \rightarrow \infty} \frac{1}{\log n} \sum_{k=1}^n \frac{1}{k} I \left\{ \frac{S_{k,k}}{\sigma_k} \leq x \right\} = \Phi(x) \text{ a.s.} \quad (1.1)$$

Here and in the sequel,  $I\{\cdot\}$  denotes indicator function and  $\Phi(\cdot)$  is the distribution function of standard normal random variable  $\mathcal{N}$ .

**Theorem 1.2** Under the conditions of Theorem 1.1, then

$$\forall x = \lim_{n \rightarrow \infty} \frac{1}{\log n} \sum_{k=1}^n \frac{1}{k} I \left\{ \left( \prod_{i=1}^k \frac{S_i}{i\mu} \right)^{\frac{1}{\gamma\sigma k}} \leq x \right\} = F(x) \text{ a.s.} \tag{1.2}$$

Here and in the sequel,  $F(\cdot)$  is the distribution function of the random variable  $e^{\sqrt{2}\mathcal{N}}$ .

## 2 Some lemmas

To prove our main results, we need the following lemmas.

**Lemma 2.1** [3] Let  $\{X_n, n \geq 1\}$  be a weakly stationary  $\rho^-$ -mixing sequence with  $EX_n = 0, 0 < EX_1^2 < \infty$ , and

- (i)  $\sigma_1^2 = EX_1^2 + 2 \sum_{n=2}^{\infty} Cov(X_1, X_n) > 0$ ,
- (ii)  $\sum_{n=2}^{\infty} |Cov(X_1, X_n)| < \infty$ ,

then

$$\frac{ES_n^2}{n} \rightarrow \sigma_1^2, \quad \frac{S_n}{\sigma_1\sqrt{n}} \xrightarrow{d} \mathcal{N}(0, 1) \text{ as } n \rightarrow \infty.$$

**Lemma 2.2** [4] For a positive real number  $q \geq 2$ , if  $\{X_n, n \geq 1\}$  is a sequence of  $\rho^-$ -mixing random variables with  $EX_i = 0, E|X_i|^q < \infty$  for every  $i \geq 1$ , then for all  $n \geq 1$ , there is a positive constant  $C = C(q, \rho^-(\cdot))$  such that

$$E \left( \max_{1 \leq j \leq n} |S_j|^q \right) \leq C \left\{ \sum_{i=1}^n E|X_i|^q + \left( \sum_{i=1}^n EX_i^2 \right)^{\frac{q}{2}} \right\}.$$

**Lemma 2.3** [6]  $\sum_{i=1}^n b_{i,n}^2 = 2n - b_{1,n}$ .

**Lemma 2.4** [[3], Theorem 3.2] Let  $\{X_{ni}, 1 \leq i \leq n, n \geq 1\}$  be an array of centered random variables with  $EX_{ni}^2 < \infty$  for each  $i = 1, 2, \dots, n$ . Assume that they are  $\rho^-$ -mixing. Let  $\{a_{ni}, 1 \leq i \leq n, n \geq 1\}$  be an array of real numbers with  $a_{ni} = \pm 1$  for  $i = 1, 2, \dots, n$ .

Denote  $\sigma_n^2 = Var \left( \sum_{i=1}^n a_{ni} X_{ni} \right)$  and suppose that

$$\sup_n \frac{1}{\sigma_n^2} \sum_{i=1}^n EX_{ni}^2 < \infty,$$

and

$$\limsup_{n \rightarrow \infty} \frac{1}{\sigma_n^2} \sum_{\substack{i,j: |i-j| \geq A \\ 1 \leq i,j \leq n}} Cov(X_{ni}, X_{nj})^- \rightarrow 0 \text{ as } A \rightarrow \infty,$$

and the following Lindeberg condition is satisfied:

$$\frac{1}{\sigma_n^2} \sum_{i=1}^n EX_{ni}^2 I\{|X_{ni}| \geq \varepsilon \sigma_n\} \rightarrow 0 \text{ as } n \rightarrow \infty$$

for every  $\varepsilon > 0$ . Then

$$\frac{1}{\sigma_n} \sum_{i=1}^n a_{ni} X_{ni} \xrightarrow{d} \mathcal{N}(0, 1) \text{ as } n \rightarrow \infty.$$

**Lemma 2.5** Let  $\{X_n, n \geq 1\}$  be a strictly stationary sequence of  $\rho^-$ -mixing random variables with  $EX_n = 0$  and  $\sum_{n=2}^{\infty} |\text{Cov}(X_1, X_n)| < \infty$ ,  $\{a_{ni}, 1 \leq i \leq n, n \geq 1\}$  be an array of real numbers such that  $\sup_n \sum_{i=1}^n a_{ni}^2 < \infty$  and  $\max_{1 \leq i \leq n} |a_{ni}| \rightarrow 0$  as  $n \rightarrow \infty$ . If  $\text{Var}\left(\sum_{i=1}^n a_{ni} X_i\right) = 1$  and  $\{X_n^2\}$  is a uniformly integrable family, then

$$\sum_{i=1}^n a_{ni} X_i \xrightarrow{d} \mathcal{N}(0, 1) \text{ as } n \rightarrow \infty.$$

**Proof** Notice that

$$\sum_{i=1}^n a_{ni} X_i = \sum_{i=1}^n \frac{a_{ni}}{|a_{ni}|} |a_{ni}| X_i =: \sum_{i=1}^n b_{ni} Y_{ni},$$

where  $b_{ni} = \frac{a_{ni}}{|a_{ni}|}$  and  $Y_{ni} = |a_{ni}| X_i$ . Then  $\{Y_{ni}, 1 \leq i \leq n, n \geq 1\}$  is an array of  $\rho^-$ -mixing centered random variables with  $EY_{ni}^2 = a_{ni}^2 EX_i^2 < \infty$  and  $b_{ni} = \pm 1$  for  $i = 1, 2, \dots, n$  and  $\sigma_n^2 = \text{Var}\left(\sum_{i=1}^n b_{ni} Y_{ni}\right) = 1$ . Note that  $\{X_n^2\}$  is a uniformly integrable family, we have

$$\sup_n \frac{1}{\sigma_n^2} \sum_{i=1}^n EY_{ni}^2 = \sup_n \sum_{i=1}^n a_{ni}^2 EX_i^2 \leq \sup_n \sum_{i=1}^n a_{ni}^2 \cdot \sup_i EX_i^2 < \infty,$$

and

$$\begin{aligned}
 & \limsup_{n \rightarrow \infty} \frac{1}{\sigma_n^2} \sum_{\substack{ij: |i-j| \geq A \\ 1 \leq ij \leq n}} \text{Cov}(Y_{ni}, Y_{nj})^- \\
 &= \limsup_{n \rightarrow \infty} \sum_{\substack{ij: |i-j| \geq A \\ 1 \leq ij \leq n}} \text{Cov}(|a_{ni}| X_i, |a_{nj}| X_j)^- \\
 &\leq \limsup_{n \rightarrow \infty} \sum_{\substack{ij: |i-j| \geq A \\ 1 \leq ij \leq n}} |a_{ni}| \cdot |a_{nj}| \cdot |\text{Cov}(X_i, X_j)| \\
 &\leq C \left( \limsup_{n \rightarrow \infty} \left( \sum_{\substack{ij: |i-j| \geq A \\ 1 \leq ij \leq n}} |a_{ni}|^2 \cdot |\text{Cov}(X_i, X_j)| + \sum_{\substack{ij: |i-j| \geq A \\ 1 \leq ij \leq n}} |a_{nj}|^2 \cdot |\text{Cov}(X_i, X_j)| \right) \right) \\
 &\leq C \sup_n \sum_{i=1}^n |a_{ni}|^2 \cdot \sum_{i>A} |\text{Cov}(X_1, X_i)| \\
 &\rightarrow 0 \text{ as } A \rightarrow \infty,
 \end{aligned}$$

and  $\forall \varepsilon > 0$ , we get

$$\begin{aligned}
 & \frac{1}{\sigma_n^2} \sum_{i=1}^n E Y_{ni}^2 I\{|Y_{ni}| \geq \varepsilon \sigma_n\} \\
 &= \sum_{i=1}^n a_{ni}^2 E X_i^2 I\{|a_{ni}| \cdot |X_i| \geq \varepsilon\} \\
 &\leq \sup_n \sum_{i=1}^n a_{ni}^2 \cdot E X_1^2 I\{|a_{ni}| \cdot |X_1| \geq \varepsilon\} \\
 &\leq \sup_n \sum_{i=1}^n a_{ni}^2 \cdot E X_1^2 I\{\max_{1 \leq i \leq n} |a_{ni}| \cdot |X_1| \geq \varepsilon\} \rightarrow 0 \text{ as } n \rightarrow \infty,
 \end{aligned}$$

thus the conclusion is proved by Lemma 2.4.

**Lemma 2.6** [2] Suppose that  $f_1(x)$  and  $f_2(y)$  are real, bounded, absolutely continuous functions on  $R$  with  $|f_1'(x)| \leq C_1$  and  $|f_2'(y)| \leq C_2$ . Then for any random variables  $X$  and  $Y$ ,

$$|\text{Cov}(f_1(X), f_2(Y))| \leq C_1 C_2 \{-\text{Cov}(X, Y) + 8\rho^-(X, Y)\|X\|_{2,1}\|Y\|_{2,1}\},$$

where  $\|X\|_{2,1} = \int_0^\infty P^{\frac{1}{2}}(|X| > x) dx$ .

**Lemma 2.7** Let  $\{X_n, n \geq 1\}$  be a strictly stationary sequence of  $\rho^-$ -mixing random variables with  $E X_1 = 0$ ,  $0 < E X_1^2 < \infty$  and

$$\begin{aligned}
 0 < \sigma_1^2 &= E X_1^2 + 2 \sum_{n=2}^{\infty} \text{Cov}(X_1, X_n) < \infty, \\
 \sum_{n=2}^{\infty} |\text{Cov}(X_1, X_n)| &< \infty,
 \end{aligned}$$

then for  $0 < p < 2$ , we have

$$n^{-\frac{1}{p}} S_n \rightarrow 0 \text{ a.s. as } n \rightarrow \infty.$$

**Proof** By Lemma 2.1, we have

$$\lim_{n \rightarrow \infty} \frac{ES_n^2}{n} = \sigma_1^2. \tag{2.1}$$

Let  $n_k = k^\alpha$ , where  $\alpha > \max \left\{ 1, \frac{p}{2-p} \right\}$ . By (2.1), we get

$$\sum_{k=1}^{\infty} P \left\{ |S_{n_k}| \geq \varepsilon n_k^{\frac{1}{p}} \right\} \leq \sum_{k=1}^{\infty} \frac{ES_{n_k}^2}{\varepsilon^2 n_k^{\frac{2}{p}}} \leq \sum_{k=1}^{\infty} \frac{C}{\varepsilon^2 k^{\alpha(\frac{2}{p}-1)}} < \infty.$$

From Borel-Cantelli lemma, it follows that

$$n_k^{-\frac{1}{p}} S_{n_k} \rightarrow 0 \text{ a.s. as } k \rightarrow \infty. \tag{2.2}$$

And by Lemma 2.2, it follows that

$$\begin{aligned} & \sum_{k=1}^{\infty} P \left\{ \max_{n_k \leq n < n_{k+1}} \frac{|S_n - S_{n_k}|}{n^{\frac{1}{p}}} \geq \varepsilon \right\} \leq \sum_{k=1}^{\infty} \frac{E \max_{n_k \leq n < n_{k+1}} |S_n - S_{n_k}|^2}{\varepsilon^2 n_k^{\frac{2}{p}}} \\ &= \sum_{k=1}^{\infty} \frac{E \max_{n_k \leq n < n_{k+1}} \left| \sum_{i=n_{k+1}}^n X_i \right|^2}{\varepsilon^2 n_k^{\frac{2}{p}}} \leq C \sum_{k=1}^{\infty} \frac{(n_{k+1} - n_k)}{\varepsilon^2 n_k^{\frac{2}{p}}} \leq C \sum_{k=1}^{\infty} \frac{1}{k^{\alpha(\frac{2}{p}-1)}} < \infty. \end{aligned}$$

By Borel-Cantelli lemma, we conclude that

$$\max_{n_k \leq n < n_{k+1}} \frac{|S_n - S_{n_k}|}{n^{\frac{1}{p}}} \rightarrow 0 \text{ a.s. as } n \rightarrow \infty. \tag{2.3}$$

For every  $n$ , there exist  $n_k$  and  $n_{k+1}$  such that  $n_k \leq n < n_{k+1}$ , by (2.2) and (2.3), we have

$$\begin{aligned} \frac{|S_n|}{n^{\frac{1}{p}}} &= \frac{|S_n - S_{n_k} + S_{n_k}|}{n^{\frac{1}{p}}} \\ &\leq \frac{|S_{n_k}|}{n_k^{\frac{1}{p}}} + \max_{n_k \leq n < n_{k+1}} \frac{|S_n - S_{n_k}|}{n^{\frac{1}{p}}} \rightarrow 0 \text{ a.s. as } n \rightarrow \infty. \end{aligned}$$

The proof is now completed.

### 3 Proof of the theorems

**Proof of Theorem 1.1** By the property of  $\rho^-$ -mixing sequence, it is easy to see that  $\{Y_n\}$  is a strictly stationary  $\rho^-$ -mixing sequence with  $EY_1 = 0$  and  $EY_1^2 = 1$ . We first prove

$$\frac{S_{n,n}}{\sigma_n} \xrightarrow{d} \mathcal{N}(0, 1) \text{ as } n \rightarrow \infty. \tag{3.1}$$

Let  $a_{ni} = \frac{b_{i,n}}{\sigma_n}$ ,  $1 \leq i \leq n, n \geq 1$ . Obviously,

$$\text{Var} \left( \sum_{i=1}^n a_{ni} Y_i \right) = 1.$$

From condition  $(a_4)$  in Theorem 1.1 and Lemma 2.3, we have

$$\sup_n \sum_{i=1}^n a_{ni}^2 = \sup_n \sum_{i=1}^n \frac{b_{i,n}^2}{\sigma_n^2} = \sup_n \frac{2n - b_{1,n}}{\sigma_n^2} \leq C \sup_n \frac{2n - b_{1,n}}{n} < \infty,$$

and

$$\max_{1 \leq i \leq n} |a_{ni}| = \max_{1 \leq i \leq n} \frac{b_{i,n}}{\sigma_n} \leq \frac{b_{1,n}}{\sigma_n} \leq \frac{C \log n}{\sqrt{n}} \rightarrow 0 \text{ as } n \rightarrow \infty.$$

By stationarity of  $\{Y_n, n \geq 1\}$  and  $E|X_1|^2 < \infty$ , we know that  $\{Y_n^2\}$  is uniformly integrable, and from condition  $(a_2)$  in Theorem 1.1, we get  $\sum_{n=2}^{\infty} |\text{Cov}(Y_1, Y_n)| < \infty$ , so applying Lemma 2.5, we have

$$\sum_{i=1}^n a_{ni} Y_i \xrightarrow{d} \mathcal{N}(0, 1).$$

Notice that

$$\sum_{i=1}^n a_{ni} Y_i = \sum_{i=1}^n \frac{b_{i,n} Y_i}{\sigma_n} = \frac{S_{n,n}}{\sigma_n},$$

so (3.1) is valid. Let  $f(x)$  be a bounded Lipschitz function and have a Radon-Nikodym derivative  $h(x)$  bounded by  $\Gamma$ . From (3.1), we have

$$Ef \left( \frac{S_{n,n}}{\sigma_n} \right) \rightarrow Ef(\mathcal{N}(0, 1)) \text{ as } n \rightarrow \infty,$$

thus

$$\frac{1}{\log n} \sum_{k=1}^n \frac{1}{k} Ef \left( \frac{S_{k,k}}{\sigma_k} \right) - Ef(\mathcal{N}(0, 1)) \rightarrow 0 \text{ as } n \rightarrow \infty. \tag{3.2}$$

On the other hand, note that (1.1) is equivalent to

$$\lim_{n \rightarrow \infty} \frac{1}{\log n} \sum_{k=1}^n \frac{1}{k} f \left( \frac{S_{k,k}}{\sigma_k} \right) = \int_{-\infty}^{\infty} f(x) d\Phi(x) = Ef(\mathcal{N}(0, 1)) \text{ a.s.} \tag{3.3}$$

from Section 2 of Peligrad and Shao [7] and Theorem 7.1 on  $P_{42}$  from Billingsley [8]. Hence, to prove (3.3), it suffices to show that

$$T_n = \frac{1}{\log n} \sum_{k=1}^n \frac{1}{k} \left[ f \left( \frac{S_{k,k}}{\sigma_k} \right) - Ef \left( \frac{S_{k,k}}{\sigma_k} \right) \right] \rightarrow 0 \text{ a.s. } n \rightarrow \infty \tag{3.4}$$

by (3.2). Let  $\xi_k = f\left(\frac{S_{k,k}}{\sigma_k}\right) - Ef\left(\frac{S_{k,k}}{\sigma_k}\right)$ ,  $1 \leq k \leq nm$  we have

$$\begin{aligned}
 ET_n^2 &= \frac{1}{\log^2 n} E\left(\sum_{k=1}^n \frac{\xi_k}{k}\right)^2 \\
 &\leq \frac{1}{\log^2 n} \sum_{1 \leq k \leq l \leq n, 2k \geq l} \frac{|E\xi_k \xi_l|}{kl} + \frac{1}{\log^2 n} \sum_{1 \leq k \leq l \leq n, 2k \geq l} \frac{|E\xi_k \xi_l|}{kl} \\
 &:= I_1 + I_2.
 \end{aligned} \tag{3.5}$$

By the fact that  $f$  is bounded, we have

$$I_1 \leq \frac{C}{\log^2 n} \sum_{k=1}^n \sum_{l=k}^{2k} \frac{1}{kl} = \frac{C}{\log^2 n} \sum_{k=1}^n \frac{1}{k} \sum_{l=k}^{2k} \frac{1}{l} \leq C(\log^{-1} n). \tag{3.6}$$

Now we estimate  $I_2$ , if  $l > 2k$ , we have

$$\begin{aligned}
 S_{l,l} - S_{2k,2k} &= (b_{1,l}Y_1 + b_{2,l}Y_2 + \dots + b_{l,l}Y_l) - (b_{1,2k}Y_1 + b_{2,2k}Y_2 + \dots + b_{2k,2k}Y_{2k}) \\
 &= (b_{2k+1,l}Y_{2k+1} + \dots + b_{l,l}Y_l) + b_{2k+1,l}\tilde{S}_{2k},
 \end{aligned}$$

and

$$\begin{aligned}
 |E\xi_k \xi_l| &= \left| \text{Cov}\left(f\left(\frac{S_{k,k}}{\sigma_k}\right), f\left(\frac{S_{l,l}}{\sigma_l}\right)\right) \right| \\
 &\leq \left| \text{Cov}\left(f\left(\frac{S_{k,k}}{\sigma_k}\right), f\left(\frac{S_{l,l}}{\sigma_l}\right) - f\left(\frac{S_{l,l} - S_{2k,2k} - b_{2k+1,l}\tilde{S}_{2k}}{\sigma_l}\right)\right) \right| \\
 &\quad + \left| \text{Cov}\left(f\left(\frac{S_{k,k}}{\sigma_k}\right), f\left(\frac{S_{l,l} - S_{2k,2k} - b_{2k+1,l}\tilde{S}_{2k}}{\sigma_l}\right)\right) \right|.
 \end{aligned}$$

By Lemma 2.3 and condition  $(a_2)$  in Theorem 1.1, we have

$$\begin{aligned}
 \text{Var}(S_{k,k}) &= \sum_{i=1}^k b_{i,k}^2 EY_i^2 + 2 \sum_{j=1}^{k-1} \sum_{i=j+1}^k b_{i,k} b_{j,k} \text{Cov}(Y_i, Y_j) \\
 &\leq \sum_{i=1}^k b_{i,k}^2 + 2 \sum_{j=1}^k b_{j,k}^2 \sum_{i=j+1}^k |\text{Cov}(Y_i, Y_j)| \\
 &\leq Ck,
 \end{aligned} \tag{3.7}$$

and

$$\begin{aligned}
 &\text{Var}\left(S_{l,l} - S_{2k,2k} - b_{2k+1,l}\tilde{S}_{2k}\right) \\
 &= \sum_{i=2k+1}^l b_{i,l}^2 EY_i^2 + 2 \sum_{j=2k+1}^{l-1} \sum_{i=j+1}^l b_{i,l} b_{j,l} \text{Cov}(Y_i, Y_j) \\
 &\leq \sum_{i=2k+1}^l b_{i,l}^2 + 2 \sum_{j=1}^l b_{i,l}^2 \sum_{i=j+1}^l |\text{Cov}(Y_i, Y_j)| \\
 &\leq Cl.
 \end{aligned}$$



By Lemma 2.6, the definition of  $\rho^-$ -mixing sequence and condition  $(a_4)$ , we have

$$\begin{aligned} & \left| \text{Cov} \left( f \left( \frac{S_{k,k}}{\sigma_k} \right), f \left( \frac{S_{l,l} - S_{2k,2k} - b_{2k+1,l} \tilde{S}_{2k}}{\sigma_l} \right) \right) \right| \\ & \leq C \left\{ -\text{Cov} \left( \frac{S_{k,k}}{\sigma_k}, \frac{S_{l,l} - S_{2k,2k} - b_{2k+1,l} \tilde{S}_{2k}}{\sigma_l} \right) \right. \\ & \quad \left. + 8\rho^- \left( \frac{S_{k,k}}{\sigma_k}, \frac{S_{l,l} - S_{2k,2k} - b_{2k+1,l} \tilde{S}_{2k}}{\sigma_l} \right) \cdot \left\| \frac{S_{k,k}}{\sigma_k} \right\|_{2,1} \cdot \left\| \frac{S_{l,l} - S_{2k,2k} - b_{2k+1,l} \tilde{S}_{2k}}{\sigma_l} \right\|_{2,1} \right\} \\ & \leq C\rho^-(k) \left( \text{Var} \left( \frac{S_{k,k}}{\sigma_k} \right) \right)^{\frac{1}{2}} \left( \text{Var} \left( \frac{S_{l,l} - S_{2k,2k} - b_{2k+1,l} \tilde{S}_{2k}}{\sigma_l} \right) \right)^{\frac{1}{2}} \\ & \quad + C\rho^-(k) \left\| \frac{S_{k,k}}{\sigma_k} \right\|_{2,1} \cdot \left\| \frac{S_{l,l} - S_{2k,2k} - b_{2k+1,l} \tilde{S}_{2k}}{\sigma_l} \right\|_{2,1} \\ & \leq C\rho^-(k) + C\rho^-(k) \left\| \frac{S_{k,k}}{\sigma_k} \right\|_{2,1} \cdot \left\| \frac{S_{l,l} - S_{2k,2k} - b_{2k+1,l} \tilde{S}_{2k}}{\sigma_l} \right\|_{2,1}. \end{aligned}$$

By the inequality  $\|X\|_{2,1} \leq \frac{r}{r-2} \|X\|_r$  ( $r > 2$ ) (cf. Zhang [[2], p. 254] or Ledoux and Talagrand [[9], p. 251]), we get

$$\left\| \frac{S_{k,k}}{\sigma_k} \right\|_{2,1} \leq \frac{r}{r-2} \left\| \frac{S_{k,k}}{\sigma_k} \right\|_r = \frac{r}{r-2} \frac{1}{\sigma_k} (E|S_{k,k}|^r)^{\frac{1}{r}},$$

and

$$\begin{aligned} E|S_{k,k}|^r &= E \left| \sum_{j=1}^k b_{j,k} Y_j \right|^r \\ &\leq C \left\{ \sum_{j=1}^k b_{j,k}^r E|X_j|^r + \left( \sum_{j=1}^k b_{j,k}^2 EX_j^2 \right)^{\frac{r}{2}} \right\} \\ &\leq C \left\{ k (\log^r k) + k^{\frac{r}{2}} \right\}, \end{aligned}$$

thus

$$\left\| \frac{S_{k,k}}{\sigma_k} \right\|_{2,1} \leq C \left( \frac{r}{r-2} \cdot \frac{\log k}{k^{\frac{1}{2}-\frac{1}{r}}} + \frac{r}{r-2} \right) < C,$$

similarly,

$$\left\| \frac{S_{l,l} - S_{2k,2k} - b_{2k+1,l} \tilde{S}_{2k}}{\sigma_l} \right\|_{2,1} < C,$$

hence

$$\left| \text{Cov} \left( f \left( \frac{S_{k,k}}{\sigma_k} \right), f \left( \frac{S_{l,l} - S_{2k,2k} - b_{2k+1,l} \tilde{S}_{2k}}{\sigma_l} \right) \right) \right| \leq C\rho^-(k).$$

Similarly to (3.7), we have

$$\begin{aligned} \text{Var}(S_{2k,2k}) &= \sum_{i=1}^{2k} b_{i,2k}^2 EY_i^2 + 2 \sum_{j=1}^{2k-1} \sum_{i=j+1}^{2k} b_{i,2k} b_{j,2k} \text{Cov}(Y_i, Y_j) \\ &\leq \sum_{i=1}^{2k} b_{i,2k}^2 + 2 \sum_{j=1}^{2k-1} b_{i,2k}^2 \sum_{i=j+1}^{2k} |\text{Cov}(Y_i, Y_j)| \\ &\leq Ck, \end{aligned}$$

and

$$\begin{aligned} \text{Var}(\tilde{S}_{2k}) &= \text{Var}\left(\sum_{i=1}^{2k} Y_i\right) \\ &= \sum_{i=1}^{2k} EY_i^2 + 2 \sum_{i=1}^{2k-1} \sum_{j=i+1}^{2k} |\text{Cov}(Y_i, Y_j)| \\ &= 2k + 2 \sum_{i=1}^{2k-1} \sum_{j=2}^{2k-i+1} \text{Cov}(Y_i, Y_j) \\ &\leq Ck. \end{aligned}$$

Since  $f$  is a bounded Lipschitz function, we have

$$\begin{aligned} &\left| \text{Cov}\left(f\left(\frac{S_{k,k}}{\sigma_k}\right), f\left(\frac{S_{l,l}}{\sigma_l}\right) - f\left(\frac{S_{l,l} - S_{2k,2k} - b_{2k+1,l}\tilde{S}_{2k}}{\sigma_l}\right)\right) \right| \\ &\leq CE \frac{|S_{2k,2k} + b_{2k+1,l}\tilde{S}_{2k}|}{\sigma_l} \\ &\leq C \frac{(\text{Var}(S_{2k,2k}))^{\frac{1}{2}}}{\sigma_l} + C \frac{b_{2k+1,l}(\text{Var}(\tilde{S}_{2k}))^{\frac{1}{2}}}{\sigma_l} \\ &\leq C\left(\frac{k}{l}\right)^{\frac{1}{2}} + C\left(\frac{k}{l}\right)^{\frac{1}{2}} \log \frac{1}{2k} \\ &\leq C\left(\frac{k}{l}\right)^\varepsilon, \end{aligned}$$

where  $0 < \varepsilon < \frac{1}{2}$ . Hence if  $l > 2k$ , we have

$$|E\xi_k \xi_l| \leq C \left[ \rho^-(k) + \left(\frac{k}{l}\right)^\varepsilon \right].$$

Thus

$$\begin{aligned}
 I_2 &\leq \frac{C}{\log^2 n} \sum_{l=2}^n \sum_{k=1}^{l-1} \frac{1}{k^{1-\varepsilon} l^{1+\varepsilon}} + \frac{C}{\log^2 n} \sum_{l=2}^n \frac{1}{l} \sum_{k=1}^{l-1} \frac{\rho^-(k)}{k} \\
 &\leq \frac{C}{\log^2 n} \sum_{l=2}^n \frac{1}{l^{1+\varepsilon}} \frac{(l-1)^\varepsilon}{\varepsilon} + \frac{C}{\log^2 n} \sum_{l=2}^n \frac{1}{l} \sum_{k=1}^n \frac{\log^{-\delta} k}{k} \\
 &\leq \frac{C}{\log^2 n} \sum_{l=2}^n \frac{1}{l} + \frac{C}{\log^2 n} \sum_{l=2}^n \frac{1}{l} \sum_{k=1}^n \frac{\log^{-\delta} k}{k} \\
 &\leq C \log^{-1} n.
 \end{aligned} \tag{3.8}$$

Associated with (3.5), (3.6), and (3.8), we have

$$ET_n^2 \leq C \log^{-1} n. \tag{3.9}$$

To prove (3.4), let  $n_k = e^{k^\tau}$ , where  $\tau > 1$ . From (3.9), we have

$$\sum_{k=1}^\infty ET_{n_k}^2 \leq C \sum_{k=1}^\infty \log^{-1} n_k = C \sum_{k=1}^\infty \frac{1}{k^\tau} < \infty.$$

Thus  $\forall \varepsilon > 0$ , we have

$$\sum_{k=1}^\infty P\{|T_{n_k}| \geq \varepsilon\} \leq \sum_{k=1}^\infty \frac{ET_{n_k}^2}{\varepsilon^2} < \infty.$$

By Borel-Cantelli lemma, we have

$$T_{n_k} \rightarrow 0 \text{ a.s. as } k \rightarrow \infty.$$

Note that

$$\frac{\log n_{k+1}}{\log n_k} = \frac{(k+1)^\tau}{k^\tau} \rightarrow 1 \text{ as } k \rightarrow \infty.$$

For every  $n$ , there exist  $n_k$  and  $n_{k+1}$  satisfying  $n_k < n \leq n_{k+1}$ , we have

$$\begin{aligned}
 |T_n| &\leq \frac{1}{\log n_k} \left| \sum_{i=1}^{n_k} \frac{\xi_i}{i} \right| + \frac{1}{\log n_k} \sum_{i=n_k}^{n_{k+1}} \frac{|\xi_i|}{i} \\
 &\leq |T_{n_k}| + C \left( \frac{\log n_{k+1}}{\log n_k} - 1 \right) \rightarrow 0 \text{ a.s. as } n \rightarrow \infty,
 \end{aligned}$$

(3.4) is completed, so the proof of Theorem 1.1 is completed.

**Proof of Theorem 1.2** Let  $C_i = \frac{S_i}{\mu_i}$ , we have

$$\frac{1}{\gamma \sigma_k} \sum_{i=1}^k (C_i - 1) = \frac{1}{\gamma \sigma_k} \sum_{i=1}^k \left( \frac{S_i}{\mu_i} - 1 \right) = \frac{1}{\sigma_k} \sum_{i=1}^k b_{i,k} Y_i = \frac{S_{k,k}}{\sigma_k}.$$

Hence (1.1) is equivalent to

$$\forall x \quad \lim_{n \rightarrow \infty} \frac{1}{\log n} \sum_{k=1}^n \frac{1}{k} I \left\{ \frac{1}{\gamma \sigma_k} \sum_{i=1}^k (C_i - 1) \leq x \right\} = \Phi(x) \text{ a.s.} \tag{3.10}$$

On the other hand, to prove (1.2), it suffices to show that

$$\forall x \quad \lim_{n \rightarrow \infty} \frac{1}{\log n} \sum_{k=1}^n \frac{1}{k} I \left\{ \frac{1}{\gamma \sigma_k} \sum_{i=1}^k \log C_i \leq x \right\} = \Phi(x) \quad a.s. \quad (3.11)$$

By Lemma 2.7, for enough large  $i$ , for some  $\frac{4}{3} < p < 2$  we have

$$|C_i - 1| = \left| \frac{S_i}{\mu i} - 1 \right| \leq C i^{\frac{1}{p}-1} \quad a.s.$$

It is easy to know that  $\log(1+x) = x + O(x^2)$  for  $|x| < \frac{1}{2}$ , thus

$$\left| \sum_{k=1}^n \log C_k - \sum_{k=1}^n (C_k - 1) \right| \leq C \sum_{k=1}^n (C_k - 1)^2 \leq C \sum_{k=1}^n k^{\frac{2}{p}-2} \leq C n^{\frac{2}{p}-1} \quad a.s.,$$

and

$$\sum_{k=1}^n (C_k - 1) - C n^{\frac{2}{p}-1} \leq \sum_{k=1}^n \log C_k \leq \sum_{k=1}^n (C_k - 1) + C n^{\frac{2}{p}-1} \quad a.s.$$

Hence for arbitrary small  $\varepsilon > 0$ , there is  $n_0 = n_0(\omega, \varepsilon)$ , such that for every  $n > n_0$  and arbitrary  $x$ ,

$$I \left\{ \frac{1}{\gamma \sigma_k} \sum_{i=1}^k (C_i - 1) \leq x - \varepsilon \right\} \leq I \left\{ \frac{1}{\gamma \sigma_k} \sum_{i=1}^k \log C_i \leq x \right\} \leq I \left\{ \frac{1}{\gamma \sigma_k} \sum_{i=1}^k (C_i - 1) \leq x + \varepsilon \right\},$$

so by (3.10), we know that (3.11) is true, and (3.11) is equivalent to (1.2), thus the proof of Theorem 1.2 is complete.

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#### Authors' contributions

XT and YZ carried out the design of the study and performed the analysis. YZ participated in its design and coordination. All authors read and approved the final manuscript.

#### Competing interests

The authors declare that they have no competing interests.

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