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# ON THE RATE OF CONVERGENCE OF THE INVARIANCE PRINCIPLE FOR STATIONARY SEQUENCES

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## ABSTRACT

In this paper, the author estimates the rate of convergence of the invariance principle for some strictly stationary sequences of random variables satisfying the  $\phi$ -mixing condition.

### 1. Introduction and results

Let  $\{X_i, i \geq 1\}$  be a strictly stationary sequence of random variables on a probability space  $(\Omega, F, P)$  and suppose  $EX_1=0$  and  $E|X_1|^{2+\delta} < \infty$  for some  $\delta > 0$ . For positive integers  $a$  and  $b$  with  $0 < a < b$ , let  $F_a^b$  denote the  $\sigma$ -field generated by random variables  $\{X_a, \dots, X_b\}$ . Suppose that the sequence  $\{X_i\}$  satisfies the  $\phi$ -mixing condition in the sense that

$$\sup_{a \geq 1} \sup_{A \in F_1^a, B \in F_{a+n}^\infty} |P(A \cap B) - P(A)P(B)| / P(A) \equiv \phi(n) \rightarrow 0$$

as  $n \rightarrow \infty$ .

It is known that if

$$\phi(n) = O(n^{-\beta}) \text{ for some } \beta > (2+\delta)/(1+\delta)$$

then the limit

$$(1.1) \quad \sigma^2 = \lim_{n \rightarrow \infty} n^{-1} E(\sum_{i=1}^n X_i)^2$$

exists. (See, e. g. [5].)

Let  $S_n = \sum_{i=1}^n X_i$  and  $S_0 = 0$ . Suppose  $\sigma^2 > 0$  and define a continuous polygonal line  $\{X_n(t), 0 \leq t \leq 1\}$  by

$$X_n(t) = (\sigma^2 n)^{-1/2} S_{[nt]} + (nt - [nt]) (\sigma^2 n)^{-1/2} X_{[nt]+1}$$

where  $[b]$  denotes an integer part of  $b$ .

Let  $C=C[0, 1]$  be the space of continuous functions on  $[0, 1]$  with the uniform metric  $d(x, y)=\sup_{0 \leq t \leq 1} |x(t) - y(t)|$  and  $\mathcal{C}$  be the smallest  $\sigma$ -field containing all open sets in  $C$ .

Let  $P_n$  be the distribution of  $\{X_n(t)\}$  and  $W$  be the Wiener measure on  $(C, \mathcal{C})$ . The Prokhorov-Lévy metric  $\rho(\cdot, \cdot)$  on the space of probability measures on  $(C, \mathcal{C})$  is defined by

$$\rho(R, Q) = \inf\{\varepsilon > 0; R(B) \leq \varepsilon + Q\{y; d(x, y) < \varepsilon, x \in B\}, \\ Q(B) \leq \varepsilon + R\{y; d(x, y) < \varepsilon, x \in B\} \text{ for all } B \in \mathcal{C}\}$$

where  $R$  and  $Q$  are probability measures on  $(C, \mathcal{C})$ .

In this paper we shall show the following results concerning the rate of convergence of  $\rho(P_n, W)$  to zero.

**Theorem 1.** *Let  $\{X_i\}$  be a strictly stationary sequence, which is  $\phi$ -mixing with coefficient  $\phi(n)$  satisfying*

$$(1.2) \quad \phi(n) = O(e^{-\gamma n})$$

for some  $\gamma > 0$  as  $n \rightarrow \infty$ . Suppose that  $EX_1 = 0$  and  $\sigma = 1$  in (1.1). If  $E|X_1|^{2+\delta} < \infty$  for some  $0 < \delta \leq 2$ , then as  $n \rightarrow \infty$

$$\rho(P_n, W) = O(n^{-\delta/2(3+2\delta)} \log n).$$

**Theorem 2.** *In Theorem 1, replace condition (1.2) by*

$$\phi(n) = O(n^{-\beta})$$

for some  $\beta > 2(2+\delta)/(1+\delta)$ . Then we have

$$\rho(P_n, W) = O(n^{-\delta/(2(3+2\delta) + 3(2+\delta)/\beta)} \log n).$$

It should be mentioned that Yoshihara [9] gave some results on the rate of convergence of  $\rho(P_n, W)$  for an absolutely regular sequence under the moment condition  $E|X_1|^{4+\varepsilon} < \infty$  for some  $\varepsilon > 0$ . Our moment condition is weaker than his, although his mixing condition is weaker than ours.

## 2. Preliminaries

The two lemmas are due to Ibragimov [6]. (As to Lemma 2, see also Yokoyama [8].)

**Lemma 1.** *Suppose that  $f$  is  $F_1^\alpha$ -measurable and  $g$  is  $F_{\alpha+n}^\infty$ -measurable and that  $E|f|^r < \infty$  and  $E|g|^s < \infty$  for  $r > 1$  and  $s > 1$  with  $r^{-1} + s^{-1} = 1$ . Then*

$$|E(fg) - E(f)E(g)| \leq 2(\phi(n))^{1/r} \|f\|_r \|g\|_s.$$

**Lemma 2.** *Under the assumptions in Theorem 1 or Theorem 2, for each  $\delta > 0$ ,*

as  $n \rightarrow \infty$

$$E|S_n|^{2+\delta} = o((ES_n^2)^{1+\delta/2}).$$

For some  $0 < \alpha < 1$ , let  $M = [n/[n^\alpha]] + 1$  and define  $I_j = \{(j-1)[n^\alpha] + 1, (j-1)[n^\alpha] + 2, \dots, j[n^\alpha]\}$  for  $j = 1, \dots, M-1$  and  $I_M = \{(M-1)[n^\alpha] + 1, (M-1)[n^\alpha] + 2, \dots, n\}$ . Let  $y_j = \sum_{i \in I_j} n^{-1/2} X_i$  for  $j = 1, \dots, M$ . Moreover for positive  $\theta$ , define  $U_j = \{j[n^\alpha] - [\theta(\log n)] + 1, j[n^\alpha] - [\theta(\log n)] + 2, \dots, j[n^\alpha]\}$  and  $v_j = \sum_{i \in U_j} n^{-1/2} X_i$ ,  $i = 1, \dots, M-1$  and  $v_M = 0$ .

The basic idea of the proofs of Theorems 1 and 2 is using the following approximation theorem by Berkes and Philipp [1].

**Lemma 3.** *Let  $\{X_i, i \geq 1\}$  be a sequence of random variables and  $\{L_i, i \geq 1\}$  be a sequence of  $\sigma$ -fields such that  $X_i$  is  $L_i$ -measurable for each  $i$  and for some  $n_k \geq 0$*

$$|P(A \cap B) - P(A)P(B)| \leq \eta_k P(A)$$

for all  $A \in V_{i < k} L_i$  and  $B \in L_k$ . Then without changing its distribution we can redefine the sequence  $\{X_i, i \geq 1\}$  on a probability space together with a sequence  $\{Y_i\}$  of independent random variables such that  $Y_i$  has the same distribution as  $X_i$  for each  $i$  and

$$P\{|X_i - Y_i| \geq 6\eta_i\} \leq 6\eta_i, i \geq 1.$$

The following lemma is due to Borovkov [2].

**Lemma 4.** *Let  $\{Y_i\}$  be a sequence of independent random variables with  $EY_i = 0$  and  $E|Y_i|^{2+\delta} < \infty$  for some  $\delta > 0$  for all  $i \geq 1$ . Then for each  $t > 0$ ,*

$$P\{\max_{1 \leq k \leq n} |\sum_{i=1}^k Y_i| > t\} \leq C_1 \exp\{-C_2 t^2 / \sum_{i=1}^n EY_i^2\} \\ + \sum_{i=1}^n P(|Y_i| > t/4) + C_3 (t^{-(2+\delta)} \sum_{i=1}^n E|Y_i|^{2+\delta}),$$

where  $C_1, C_2$  and  $C_3$  are positive constants depending only on  $\delta$ .

### 3. Proof of Theorem 1

Let

$$\alpha = \delta / (3 + 2\delta),$$

$$\varepsilon_n = K n^{-\delta/2(3+2\delta)} \log n, \quad n \geq 1,$$

$$\lambda_n = n^{-\delta/(3+2\delta)} \log n, \quad n \geq 1,$$

where  $K$  is larger than  $(\theta/C_2)^{1/2}$ .

Using Lemmas 1 and 2, we can easily prove the following

**Lemma 5.** *For  $j = 1, \dots, M-1$ , as  $n \rightarrow \infty$*

$$E y_j^2 = n^{\alpha-1} + o(n^{-1}),$$

$$E|y_j|^{2+\delta} = O(n^{(\alpha-1)(1+\delta/2)}),$$

$$Ev_j^2 = \theta(\log n)/n + O(n^{-1})$$

and

$$E|v_j|^{2+\delta} = O(n^{-(1+\delta/2)}(\log n)^{1+\delta/2}).$$

Also

$$Ey_M^2 = (n - (M-1)n^\alpha)/n + O(n^{-1})$$

and

$$E|y_M|^{2+\delta} = O(n^{(\alpha-1)(1+\delta/2)}).$$

By the same argument as in the proof of Theorem 1 in [3] (p. 213), if we construct the Brownian motion  $\{B(t), 0 \leq t \leq 1\}$  on  $(\Omega, F, P)$  such that

$$(3.1) \quad P\{\sup_{0 \leq t \leq 1} |X_n(t) - B(t)| \geq \epsilon_n\} = O(\epsilon_n)$$

as  $n \rightarrow \infty$ , the statement of the theorem is concluded. Hence first we have to define the Brownian motion.

Denote  $\xi_j = y_j - v_j$ ,  $j=1, \dots, M$  and denote  $M_a^b$  the  $\sigma$ -field generated by the random variables  $\{\xi_a, \dots, \xi_b\}$  for  $0 < a < b$ . For any  $A \in M_1^a$  and  $B \in M_{a+1}^b$  we have

$$|P(A \cap B) - P(A)P(B)| \leq \phi([\theta \log n])P(A) = O(n^{-\theta'})P(A)$$

uniformly with respect to  $a$  with  $1 \leq a \leq M-1$ . Hence, by applying Lemma 3, we see that there exist a sequence  $\{\tilde{\xi}_1, \dots, \tilde{\xi}_M\}$  and a sequence of independent random variables  $\{\tilde{Y}_1, \dots, \tilde{Y}_M\}$  on a probability space  $(\tilde{\Omega}, \tilde{F}, \tilde{P})$  such that the joint distribution of  $\{\tilde{\xi}_1, \dots, \tilde{\xi}_M\}$  is same as  $\{\xi_1, \dots, \xi_M\}$  and  $\tilde{\xi}_i$  has the same distribution as  $Y_i$  with

$$(3.2) \quad P\{|\tilde{\xi}_i - \tilde{Y}_i| \geq C_4 n^{-\theta'}\} \leq C_5 n^{-\theta'}$$

for each  $i=1, \dots, M$ , where  $C_4$  and  $C_5$  are absolute positive constants.

On the other hand, using the Skorokhod embedding theorem (Theorem 7 in [7]) we construct the Brownian motion  $\{B^*(t), 0 \leq t \leq 1\}$  and a sequence of independent and positive random variables  $\{T_i^*, i=1, \dots, M\}$  on another probability space  $(\Omega^*, F^*, P^*)$  such that

$$\{(B^*(T_1^*), B^*(T_1^* + T_2^*) - B^*(T_1^*), \dots, B^*(\sum_{i=1}^M T_i^*) - B^*(\sum_{i=1}^{M-1} T_i^*)\} \stackrel{d}{=} (Y_1, Y_2, \dots, Y_M),$$

and

$$(3.3) \quad ET_i^* = E\tilde{Y}_i^2 \quad \text{and} \quad E|T_i^*|^{1+\delta/2} \leq C_6 E|\tilde{Y}_i|^{2+\delta}$$

for all  $1 \leq i \leq M$ , where " $\stackrel{d}{=}$ " means the equality of joint distributions and  $C_6$  is a positive constant depending only on  $\delta$ .

Using Lemma A1 in [1] we can redefine  $(\{\xi_i\}, \{Y_i\}, \{T_i\}, \{B(t)\})$  on a common probability space  $(\Omega, F, P)$  such that the distribution of  $(\{\tilde{\xi}_i\}, \{\tilde{Y}_i\})$  and  $(\{B^*(\sum_{k=1}^i T_k^*) - B^*(\sum_{k=1}^{i-1} T_k^*)\}, \{T_i^*\}, \{B^*(t)\})$  are same as those of  $(\{\xi_i\}, \{Y_i\})$  and  $(\{Y_i\}, \{T_i\}, \{B(t)\})$ , respectively.

Define  $\{a_0, a_1, \dots, a_M\}$  by  $a_i = i[n^\alpha]/n$  for  $i=0, 1, \dots, M-1$  and  $a_M = 1$ . Let  $\{\tilde{X}_n(t), 0 \leq t \leq 1\}$  and  $\{\tilde{B}_n(t), 0 \leq t \leq 1\}$  be continuous polygonal lines defined by

$$\tilde{X}_n(t) = X_n(a_k) + (X_n(a_{k+1}) - X_n(a_k))(t - a_k)/(a_{k+1} - a_k),$$

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for  $t \in [a_k, a_{k+1}]$ ,  $k=0, 1, \dots, M-1$ , and

$$\tilde{B}_n(t) = B(a_k) + (B(a_{k+1}) - B(a_k))(t - a_k)/(a_{k+1} - a_k),$$

for  $t \in [a_k, a_{k+1}]$ ,  $k=0, 1, M-1$ , respectively.

**Lemma 6.** As  $n \rightarrow \infty$

$$P\{\sup_{0 \leq t \leq 1} |\tilde{X}_n(t) - \tilde{B}_n(t)| \geq \varepsilon_n\} = o(\varepsilon_n).$$

*Proof.* We have

$$\begin{aligned} & P\{\sup_{0 \leq t \leq 1} |\tilde{X}_n(t) - \tilde{B}_n(t)| \geq \varepsilon_n\} \\ &= P\{\max_{1 \leq k \leq M} |\sum_{i=1}^k y_i - B(a_k)| \geq \varepsilon_n\} \\ &\leq P\{\max_{1 \leq k \leq M} |\sum_{i=1}^k (y_i - v_i) - B(a_k)| \geq \varepsilon_n/2\} \\ &\quad + P\{\max_{1 \leq k \leq M} |\sum_{i=1}^k v_i| \geq \varepsilon_n/2\} \\ &\equiv A_1 + A_2, \end{aligned}$$

say. Applying Lemma 3 to  $A_2$ , we see that there exist independent and identically distributed random variables  $\{V_i\}$  such that  $V_i$  has the same distribution as  $v_i$  and

$$P\{|v_i - V_i| \geq C_7 e^{-\gamma n^{\alpha/2}}\} \leq C_8 e^{-\gamma n^{\alpha/2}},$$

where  $C_7$  and  $C_8$  are absolute positive constants. Thus by Lemmas 4 and 5 we have

$$\begin{aligned} A_2 &\leq P\{\max_{1 \leq k \leq M-1} |\sum_{i=1}^k (v_i - V_i)| \geq \varepsilon_n/4\} \\ &\quad + P\{\max_{1 \leq k \leq M-1} |\sum_{i=1}^k V_i| > \varepsilon_n/4\} \\ &\leq (M-1)P\{|v_i - V_i| \geq \varepsilon_n/4(M-1)\} + C_1 \exp\{-C_2 \varepsilon_n^2 / \sum_{i=1}^{M-1} E V_i^2\} \\ &\quad + \sum_{i=1}^{M-1} P\{|V_i| \geq \varepsilon_n/16\} + C_3 (\sum_{i=1}^{M-1} E |V_i|^{2+\delta} \varepsilon_n^{-(2+\delta)})^2 \\ &= o(\varepsilon_n), \end{aligned}$$

as  $n \rightarrow \infty$ .

We next estimate  $A_1$ . Denote  $Z_k = \sum_{i=1}^k (T_i - ET_i)$ . Recall  $y_i - v_i = \xi_i$ . Since  $\sum_{i=1}^k Y_i = B(\sum_{i=1}^k T_i)$  by the construction of  $\{Y_i\}$ , we have

$$\begin{aligned} A_1 &\leq P\{\max_{1 \leq k \leq M} |\sum_{i=1}^k (\xi_i - Y_i)| \geq \varepsilon_n/4\} \\ &\quad + P\{\max_{1 \leq k \leq M} |\sum_{i=1}^k Y_i - B(a_k)| \geq \varepsilon_n/4\} \\ &\leq P\{\sum_{i=1}^M |\xi_i - Y_i| \geq \varepsilon_n/4\} \\ &\quad + P\{\max_{1 \leq k \leq M} |B(Z_k + \sum_{i=1}^k ET_i) - B(a_k)| \geq \varepsilon_n/4\} \\ &\leq \sum_{i=1}^M P\{|\xi_i - Y_i| \geq \varepsilon_n/4M\} \\ &\quad + P\{\max_{1 \leq k \leq M} |B(Z_k + \sum_{i=1}^k ET_i) - B(a_k)| \geq \varepsilon_n/4, \\ &\quad \quad \quad \max_{1 \leq k \leq M} |Z_k| \leq \varepsilon_n\} \\ &\quad + P\{\max_{1 \leq k \leq M} |Z_k| > \varepsilon_n\} \\ &\equiv L_1 + L_2 + L_3, \end{aligned}$$

say. By (3.2) we have for any sufficiently large  $\theta$

$$L_1 = O(n^{1-\alpha-\theta r}) = o(\varepsilon_n).$$

Since  $ET_i = EY_i^2$  and  $Y_i$  has the same distribution as  $\xi_i = y_i - v_i$ , we have  $ET_i = E(y_i - v_i)^2$ , thus

$$\begin{aligned} & \max_{1 \leq k \leq M} \left| \sum_{i=1}^k ET_i - a_k \right| \\ & \leq \max_{1 \leq k \leq M} \left| \sum_{i=1}^k Ey_i^2 - a_k \right| + 2 \sum_{i=1}^M |E(y_i v_i)| + \sum_{i=1}^M Ev_i^2. \end{aligned}$$

By Lemmas 1 and 5 we have

$$\begin{aligned} |E(y_i v_i)| & \leq |E\{(\sum_{i=1}^{\lfloor n^\alpha \rfloor - \lfloor \theta(\log n) \rfloor} n^{-1/2} X_i) v_i\}| + Ev_i^2 \\ & \leq \sum_{i=1}^{\lfloor n^\alpha \rfloor - \lfloor \theta(\log n) \rfloor} 2(\phi(\lfloor n^\alpha \rfloor - \lfloor \theta(\log n) \rfloor + 1 - i))^{(1+\delta)/(2+\delta)} \\ & \quad \times \|n^{-1/2} X_i\|_{(2+\delta)/(1+\delta)} \|v_i\|_{2+\delta} + Ev_i^2 \\ & = O(n^{-1}(\log n)), \end{aligned}$$

thus

$$\sum_{i=1}^M |E(y_i v_i)| = O(\lambda_n).$$

And also we have from Lemma 5

$$\max_{1 \leq k \leq M} \left| \sum_{i=1}^k Ey_i^2 - a_k \right| = o(\lambda_n)$$

and

$$\sum_{i=1}^M Ev_i^2 = o(\lambda_n).$$

Hence there exists a positive constant  $C_9$  such that

$$\max_{1 \leq k \leq M} \left| \sum_{i=1}^k ET_i - a_k \right| \leq C_9 \lambda_n.$$

Then by the same argument as in the proof of Lemma 5 in [7], we have

$$\begin{aligned} L_2 & \leq \sum_{k=1}^M P\{\sup_{|t| \leq (1+C_9)\lambda_n} |B(t+a_k) - B(a_k)| \geq \varepsilon/4\} \\ & \leq 2M P\{\sup_{0 \leq t \leq (1+C_9)\lambda_n} |B(t)| \geq \varepsilon_n/4\} \\ & \leq 8M P\{|B(1)| \geq \varepsilon_n/4((1+C_9)\lambda_n)^{1/2}\} \\ & = o(\varepsilon_n). \end{aligned}$$

On the other hand by the Kolmogorov inequality and the Marcinkiewicz-Zygmund inequality we have

$$L_3 \leq \lambda_n^{-(1+\delta/2)} E|Z_M|^{1+\delta/2} \leq C_{10} \lambda_n^{-(1+\delta/2)} \sum_{i=1}^M E|T_i - ET_i|^{1+\delta/2},$$

where  $C_{10}$  is a positive constant depending only on  $\delta$ . Since  $Y_i$  has the same distribution as  $\xi_i$ , using (3.3) and Lemma 5, we have

$$L_3 \leq C_{11} \lambda_n^{-(1+\delta/2)} \sum_{i=1}^M E|\xi_i|^{2+\delta} = o(\varepsilon_n),$$

where  $C_{11}$  is an absolute constant. Thus we have  $A_1 = o(\varepsilon_n)$ , which concludes the lemma.

The following lemma is due to Borovkov (Lemma 2 in [3]).

**Lemma 7.** As  $n \rightarrow \infty$

$$P\{\sup_{0 \leq t \leq 1} |\tilde{B}_n(t) - B(t)| \geq \varepsilon_n\} = o(\varepsilon_n).$$

Finally we shall prove

**Lemma 8.** As  $n \rightarrow \infty$

$$P\{\sup_{0 \leq t \leq 1} |X_n(t) - \tilde{X}_n(t)| \geq \varepsilon_n\} = o(\varepsilon_n).$$

*Proof.* Let  $q$  be a real number such that  $\phi(q) < 1/10$  and define

$$\chi(j, k) = \begin{cases} 0 & \text{if } [k/q]q + j > k, \\ 1 & \text{if } [k/q]q + j \leq k. \end{cases}$$

By the definition of  $\{X_n(t)\}$  and  $\{\tilde{X}_n(t)\}$  we have

$$\begin{aligned} & P\{\sup_{0 \leq t \leq 1} |X_n(t) - \tilde{X}_n(t)| \geq \varepsilon_n\} \\ & \leq \sum_{i=1}^M P\{\sup_{a_{i-1} \leq t \leq a_i} |X_n(t) - \tilde{X}_n(t)| \geq \varepsilon_n\} \\ & \leq M P\{\max_{1 \leq k \leq [n^\alpha]} |\sum_{i=1}^k n^{-1/2} X_i - (k/[n^\alpha]) \sum_{i=1}^{[n^\alpha]} n^{-1/2} X_i| \geq \varepsilon_n\} \\ & \leq M P\{\max_{1 \leq k \leq [n^\alpha]} |\sum_{i=1}^k n^{-1/2} X_i| \geq \varepsilon_n/2\} \\ & \leq M P\{\max_{1 \leq k \leq [n^\alpha]} |\sum_{j=1}^q \sum_{i=1}^{[k/q] + \chi(j, k)} n^{-1/2} X_{(i-1)q+j}| \geq \varepsilon_n/2\} \\ & \leq (Mq) P\{\max_{1 \leq k \leq [(n^\alpha)/q]+1} |\sum_{i=1}^k n^{-1/2} X_{(i-1)q+1}| \geq \varepsilon_n/2q\} \end{aligned}$$

Let  $m = [(n^\alpha)/q] + 1$ . Using Theorem 1.2 in [4] and lemma 3 we have as  $n \rightarrow \infty$

$$\begin{aligned} (3.4) \quad & (Mq) P\{\max_{1 \leq k \leq m} |\sum_{i=1}^k n^{-1/2} X_{(i-1)q+1}| \geq \varepsilon_n/2q\} \\ & (Mq)(\varepsilon_n/4q)^{-(2+\delta)} E|\sum_{i=1}^m n^{-1/2} X_{(i-1)q+1}|^{2+\delta} \\ & \leq \frac{\quad}{1 - \phi(q) - (\varepsilon_n/4q)^{-(2+\delta)} \max_{1 \leq k \leq m-1} E|\sum_{i=k}^m n^{-1/2} X_{(i-1)q+1}|^{2+\delta}}. \end{aligned}$$

Similarly as in the proof of Lemma 5, we have by Lemmas 1 and 2

$$E|\sum_{i=1}^m n^{-1/2} X_{(i-1)q+1}|^{2+\delta} = O(n^{(\alpha-1)(1+\delta/2)}),$$

and also we have from the stationarity of  $\{X_i\}$

$$\begin{aligned} & \max_{1 \leq k \leq m-1} E|\sum_{i=k}^m n^{-1/2} X_{(i-1)q+1}|^{2+\delta} \\ & = \max_{2 \leq k \leq m} E|\sum_{i=1}^k n^{-1/2} X_{(i-1)q+1}|^{2+\delta}, \end{aligned}$$

which is  $O(n^{(\alpha-1)(1+\delta/2)})$  because

$$E|\sum_{i=1}^k n^{-1/2} X_{(i-1)q+1}|^{2+\delta} = O(n^{(\alpha-1)(1+\delta/2)})$$



uniformly with respect to  $k$ ,  $2 \leq k \leq m$ . Thus the right hand side of (3.4) is the order of  $o(\varepsilon_n)$  as  $n \rightarrow \infty$ . Hence the lemma is proved.

By Lemmas 6, 7 and 8 we obtain (3.1) and thus conclude the proof of Theorem 1.

#### 4. Proof of Theorem 2

It is sufficient to modify some definitions in the proof of Theorem 1. We first define  $U_j$  by  $U_j = \{j[n^\alpha] - [n^{(2-\alpha)/(1+2\beta)}] + 1, j[n^\alpha] - [n^{(2-\alpha)/(1+2\beta)}] + 2, \dots, j[n^\alpha]\}$ ,  $j=1, \dots, M-1$  and put

$$\alpha = \{2\delta\beta + 3(2+\delta)\} / \{2(3+2\delta)\beta + 3(2+\delta)\},$$

$$\varepsilon_n = K n^{-\delta / \{2(3+2\delta) + 3(2+\delta)/\beta\}} (\log n),$$

$$\lambda_n = n^{-2\delta / \{2(3+2\delta) + 3(2+\delta)/\beta\}} (\log n).$$

We note that for  $\beta > 2(2+\delta)/(1+\delta)$

$$E y_i^2 = n^{-1 + (2-\alpha)/(1+2\beta)} + O(n^{-1})$$

and

$$\max_{1 \leq k \leq M} |\sum_{i=1}^k ET_i - a_k| = O(n^{-\alpha + (2-\alpha)/(1+2\beta)}) = o(\lambda_n),$$

as  $n \rightarrow \infty$ . The rest of the proof is the same as that of Theorem 1 and so is omitted.

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