

## CONVERGENCE RATES FOR INTEGRAL TYPE FUNCTIONALS OF WEAKLY DEPENDENT RANDOM VARIABLES

By

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### 1. Introduction and results.

Let  $\{X_i, i \in \mathbf{Z}\}$  be a strictly stationary sequence of random variables. Let  $\mathcal{F}_p^q$  denote a  $\sigma$ -field generated by random variables  $\{X_i, i=p, p+1, \dots, q\}$ . Suppose the sequence  $\{X_i\}$  satisfies the condition of the absolute regularity,

$$(1.1) \quad \beta(n) = E \left[ \sup_{A \in \mathcal{F}_n^\infty} |P(A | \mathcal{F}_{-\infty}^0) - P(A)| \right] \downarrow 0,$$

as  $n \rightarrow \infty$ .

Suppose  $EX_1=0$  and  $E|X_1|^{2+\delta} < \infty$  for some  $\delta > 0$ . Under these assumptions, if

$$\sum_{i=1}^{\infty} (\beta(i))^{2/2+\delta} < \infty,$$

then

$$(1.2) \quad \sigma^2 \equiv EX_1^2 + 2 \sum_{i=1}^{\infty} E(X_i X_{i+1}) < \infty.$$

(See e. g. [1]). Throughout this paper we assume  $\sigma > 0$ .

Let  $S_k = \sum_{i=1}^k X_i$ ,  $S_0=0$  and  $\{B(t), t \geq 0\}$  be the standard Brownian motion. Let  $f(s, x) \in C(\mathbf{R}^2)$  be a function such that  $f$  and its partial derivative of order one are of slow growth in  $x$ , i. e.  $f$  satisfies the inequality of the form,

$$(1.3) \quad |Df(s, x)| \leq K_1(1 + |x|^a),$$

where  $D$  denotes either the identity operator or first derivative, and  $a$  and  $K_1$  are some positive constants independent of  $x$  and  $s$ . Moreover assume that the probability distribution

$$F(\lambda) = P \left( \int_0^1 f(t, B(t)) dt \leq \lambda \right)$$

satisfies a first order Lipschitz condition, i. e. there exists an absolute positive constant  $K_2$  such that

$$(1.4) \quad |F(t) - F(s)| \leq K_2 |t - s|,$$

uniformly in  $t$  and  $s$ .

Under these assumptions, Yoshihara [6] proved that if  $E|X_1|^{4+\delta} < \infty$  for some  $\delta > 0$  and  $\beta(n) = O(e^{-\gamma n})$  for some  $\gamma > 0$ , then

$$(1.5) \quad \Delta_n \equiv \sup_{\lambda} \left| P\left(\frac{1}{n} \sum_{k=1}^n f\left(\frac{k}{n}, \frac{S_k}{\sigma\sqrt{n}}\right) < \lambda\right) - F(\lambda) \right| = O(n^{-1/5}(\log n)^{1+a'}),$$

where  $a' > a > 0$ .

The purpose of this paper is to estimate the left hand side of (1.5) for an absolutely regular sequence of random variables with absolute moments of all orders, by the method in the proofs given in [3] and [6], and to show the following theorem.

**Theorem 1.** *Let  $\{X_i, i \in \mathbf{Z}\}$  be a sequence of strictly stationary random variables. Suppose that the sequence  $\{X_i\}$  satisfies*

$$(1.6) \quad \beta(n) = O(e^{-\gamma n})$$

for some  $\gamma > 0$ . Suppose also that  $EX_1 = 0$  and for any  $\lambda \in \mathbf{R}^1$

$$(1.7) \quad P(|X_1| \geq \lambda) = O(\exp\{-\alpha\lambda^\delta\})$$

for some  $\alpha > 0$  and  $\delta > 0$ . Moreover assume without loss of generality that  $\sigma \equiv 1$ . Let  $f(s, t) \in C^1(\mathbf{R}^2)$  be a function satisfying the conditions (1.3) and (1.4). We then have

$$\Delta_n = O(n^{-1/4}(\log n)^\eta),$$

where  $\eta = \eta(\alpha, \delta) > 1 + a$  is some positive constant depending only on  $\alpha$  and  $\delta$ .

## 2. Preliminary lemmas.

In this section we give five lemmas. The basic idea of the proof of Theorem 1 is to use the following approximation theorem due to Yoshihara [5].

**Lemma 1.** *Let  $\{X_i\}$  be an absolutely regular sequence of random variables such that  $EX_1 = 0$  and  $E|X_1|^{2+\delta} < \infty$  for some  $\delta > 0$ . Let  $g(t_1, \dots, t_k)$  be any Borel function on  $\mathbf{R}^k$  with  $|g(t_1, \dots, t_k)| \leq K_g$  for some positive constant  $K_g$ . Then there exists a sequence of independent random variables  $\{Y_i\}$  such that each  $Y_i$  has the same distribution as that of  $X_i$  and*

$$|E\{g(X_{i_1}, X_{i_2}, \dots, X_{i_k})\} - E\{g(Y_{i_1}, Y_{i_2}, \dots, Y_{i_k})\}| \leq 2K_g k \beta(d)$$

for any  $1 \leq i_1 < i_2 < \dots < i_k$ , where

$$d = \max_{1 \leq j \leq k-1} (i_{j+1} - i_j).$$

The following lemma is also due to Yoshihara [7].

**Lemma 2.** *Let  $\{X_i\}$  be a strictly stationary and absolutely regular sequence of random variables with zero mean and  $E|X_1|^s < \infty$  for some  $s \geq 3$ . If the assumption (1.6) holds and  $\sigma > 0$  in (1.2), then*

$$\sup_{t > ((s-2)\log n)^{1/2}} \{t^s (\log n)^{-1} P(|S_n| \geq t(EX_1^2)^{1/2} (\log n)^{1/2} n^{1/2})\} \leq K_s n^{-(s-2)/2},$$

where  $K_s$  is a positive constant depending only on  $s$ .

We now define several sequences of random variables associated with  $\{X_i, i \in \mathbb{Z}\}$  in the following way. Let  $M = [n^{1/2}] + 1$  and define the sequences  $I_j = \{(j-1)[n^{1/2}] + 1, (j-1)[n^{1/2}] + 2, \dots, j[n^{1/2}]\}$  for  $j = 1, \dots, M-1$  and  $I_M = \{(M-1)[n^{1/2}] + 1, (M-1)[n^{1/2}] + 2, \dots, n\}$  and write

$$y_j = \sum_{i \in I_j} n^{-1/2} X_i, \quad j = 1, \dots, M.$$

Furthermore letting  $U_j = \{j[n^{1/2}] - [\theta(\log n)] + 1, j[n^{1/2}] - [\theta(\log n)] + 2, \dots, j[n^{1/2}]\}$ , define the random variables

$$v_j = \sum_{i \in U_j} n^{-1/2} X_i, \quad j = 1, \dots, M-1$$

and  $v_M = 0$ , where  $\theta$  is some positive constant. Write  $\xi_j = y_j - v_j$  for  $j = 1, \dots, M$ . Using Theorem 1 in Yokoyama [3] we have the following lemma.

**Lemma 3.** *As  $n \rightarrow \infty$ ,*

(i)  $E y_j^2 = n^{-1/2} + O(n^{-1}), \quad j = 1, \dots, M-1,$

(ii)  $E y_M^2 = (n - (M-1)[n^{1/2}])/n + O(n^{-1})$

and

(iii)  $E v_j^2 = (\theta(\log n))/n + O(n^{-1}), \quad j = 1, \dots, M-1.$

Moreover for each positive integer  $m$ ,

(iv)  $E |y_j|^{2m} \leq K_\beta K_m E |X_1|^{2m} n^{-m/2}, \quad j = 1, \dots, M,$

and

(v)  $E |v_j|^{2m} \leq K_\beta K_m E |X_1|^{2m} (\theta(\log n))^m n^{-m}, \quad j = 1, \dots, M-1,$

where  $K_m = (2m)!(2m-1)$  and  $K_\beta$  is a positive constant depending only on the coefficient  $\beta(\cdot)$ .

Take  $M$  points  $a_k, k = 1, \dots, M$  on  $[0, 1]$  such that  $a_k = k[n^{1/2}]/n$  for  $k = 1, \dots, M-1$  and  $a_M = 1$ . Let  $\{X_n(t), 0 \leq t \leq 1\}, \{\tilde{X}_n(t), 0 \leq t \leq 1\}$  and  $\{\hat{X}_n(t), 0 \leq t \leq 1\}$  be continuous polygonal lines defined by

$$X_n(t) = \begin{cases} n^{1/2}tX_1, & \text{for } t \in [0, \frac{1}{n}], \\ n^{-1/2}S_k + (nt-k)n^{-1/2}X_{k+1}, & \text{for } t \in (\frac{k}{n}, \frac{k+1}{n}], \text{ for } k=1, \dots, n-1, \end{cases}$$

$$\tilde{X}_n(t) = \begin{cases} \frac{t}{a_1}y_1, & \text{for } t \in [0, a_1], \\ \sum_{i=1}^k y_i + \frac{t-a_k}{a_{k+1}-a_k}y_{k+1}, & \text{for } t \in (a_k, a_{k+1}], \text{ } k=1, \dots, M-1 \end{cases}$$

and

$$\hat{X}_n(t) = \begin{cases} \frac{t}{a_1}\xi_1, & \text{for } t \in [0, a_1], \\ \sum_{i=1}^k \xi_i + \frac{t-a_k}{a_{k+1}-a_k}\xi_{k+1}, & \text{for } t \in (a_k, a_{k+1}], \text{ } k=1, \dots, M-1, \end{cases}$$

respectively. With these definitions, we have

**Lemma 4.** Let  $\varepsilon_n = n^{-1/4}(\log n)^\eta$ . As  $n \rightarrow \infty$ ,

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

**Proof.** Since, from the condition (1.3),

$$(2.1) \quad |f(t, x) - f(t, x')| \leq K_1(1 + \max(|x|^a, |x'|^a))|x - x'|,$$

we have for some positive  $b$  with  $b > a$  and  $b < \eta - 1$ ,

$$(2.2) \quad \begin{aligned} & P(\sup_{0 \leq t \leq 1} |f(t, X_n(t)) - f(t, \tilde{X}_n(t))| \geq \varepsilon_n) \\ & \leq P(\sup_{0 \leq t \leq 1} K_1 |X_n(t) - \tilde{X}_n(t)| (1 + \max(|X_n(t)|^a, |\tilde{X}_n(t)|^a)) \geq \varepsilon_n) \\ & \leq P(\sup_{0 \leq t \leq 1} K_1 |X_n(t) - \tilde{X}_n(t)| (1 + \max(|X_n(t)|^a, |\tilde{X}_n(t)|^a)) \geq \varepsilon_n, \\ & \quad \sup_{0 \leq t \leq 1} |X_n(t)|^a < K_1^{-1}(\log n)^b) \\ & \quad + P(\sup_{0 \leq t \leq 1} |X_n(t)|^a \geq K_1^{-1}(\log n)^b) \\ & \leq P(\sup_{0 \leq t \leq 1} |X_n(t) - \tilde{X}_n(t)| \geq \varepsilon_n (\log n)^{-b}) \\ & \quad + P\left(\max_{1 \leq k \leq n} \left| n^{-1/2} \sum_{i=1}^k X_i \right| \geq K_1^{-1}(\log n)^{b/a}\right) \\ & \equiv A_1 + A_2, \text{ say.} \end{aligned}$$

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

$$(2.3) \quad A_1 = \sum_{i=1}^M P(\sup_{a_{i-1} \leq t \leq a_i} |X_n(t) - \tilde{X}_n(t)| \geq \varepsilon_n (\log n)^{-b})$$

$$\begin{aligned} &\leq MP\left(\max_{1 \leq k \leq [n^{1/2}]} \left| \sum_{i=1}^k n^{-1/2} X_i \right| \geq \varepsilon_n (\log n)^{-b}\right) \\ &= MP\left(\max_{1 \leq k \leq [n^{1/2}]} \left| \sum_{i=1}^k n^{-1/4} X_i \right| \geq (\log n)^{\eta-b}\right). \end{aligned}$$

Recall that  $\eta - b > 1$ . Using Lemma 2 with  $s=10$ , we obtain that

$$P\left(\left| \sum_{i=1}^k n^{-1/4} X_i \right| \geq (\log n)^{\eta-b}\right) = o(n^{-2})$$

uniformly in  $k$  with  $1 \leq k \leq M$ . Thus we have from (2.3) that

$$(2.4) \quad A_1 \leq M \sum_{k=1}^{[n^{1/2}]} P\left(\left| \sum_{i=1}^k n^{-1/4} X_i \right| \geq (\log n)^{\eta-b}\right) = o(\varepsilon_n).$$

Furthermore it is easily proved that  $A_2 = o(\varepsilon_n)$ , too. Hence the lemma is shown by (2.3) and (2.4).

**Lemma 5.** As  $n \rightarrow \infty$ ,

$$P\left(\sup_{0 \leq t \leq 1} |f(t, \tilde{X}_n(t)) - f(t, \hat{X}_n(t))| \geq \varepsilon_n\right) = o(\varepsilon_n).$$

**Proof.** By the inequality (2.1) we have

$$\begin{aligned} (2.5) \quad &P\left(\sup_{0 \leq t \leq 1} |f(t, \tilde{X}_n(t)) - f(t, \hat{X}_n(t))| \geq \varepsilon_n\right) \\ &\leq P\left(\sup_{0 \leq t \leq 1} K_1 |\tilde{X}_n(t) - \hat{X}_n(t)| (1 + \max(|\tilde{X}_n(t)|^a, |\hat{X}_n(t)|^a)) \geq \varepsilon_n\right) \\ &\leq P\left(\sup_{0 \leq t \leq 1} |\tilde{X}_n(t) - \hat{X}_n(t)| \geq \varepsilon_n (\log n)^{-b}\right) \\ &\quad + P\left(\sup_{0 \leq t \leq 1} \max(|\tilde{X}_n(t)|^a, |\hat{X}_n(t)|^a) \geq K_1^{-1} (\log n)^b\right) \\ &\equiv B_1 + B_2, \quad \text{say.} \end{aligned}$$

Since  $\eta - b > 1$ , we apply Lemma 2 to  $B_1$  with  $s=4$  to see that

$$\begin{aligned} (2.6) \quad &B_1 = P\left(\max_{1 \leq k \leq M-1} \left| \sum_{i=1}^k v_i \right| \geq \varepsilon_n (\log n)^{-b}\right) \\ &\leq \sum_{k=1}^{M-1} P\left(\left| \sum_{i=1}^k v_i \right| \geq \theta^{-1/2} (E v_i^2)^{1/2} (\log n)^{\eta-b-1/2} n^{1/4}\right) \\ &= o(n^{-1/2}) \\ &= o(\varepsilon_n) \end{aligned}$$

We next estimate  $B_2$  as follows. From the definitions of  $\tilde{X}_n(t)$  and  $\hat{X}_n(t)$  we have

$$(2.7) \quad B_2 \leq P\left(\sup_{0 \leq t \leq 1} |\tilde{X}_n(t)|^a \geq K_1^{-1} (\log n)^b\right) + P\left(\sup_{0 \leq t \leq 1} |\hat{X}_n(t)|^a \geq K_1^{-1} (\log n)^b\right)$$

$$\begin{aligned}
&= P\left(\max_{1 \leq k \leq M} \left| \sum_{i=1}^k y_i \right| \geq K_1^{-1/a} (\log n)^{b/a}\right) + P\left(\max_{1 \leq k \leq M} \left| \sum_{i=1}^k \xi_i \right| \geq K_1^{-1/a} (\log n)^{b/a}\right) \\
&= B_{21} + B_{22}, \quad \text{say.}
\end{aligned}$$

Since  $b/a > 1$ , we also apply Lemma 2 to  $B_{21}$  and  $B_{22}$  with  $s=4$  and we obtain that

$$\begin{aligned}
(2.8) \quad B_{21} &\leq \sum_{k=1}^M P\left(\left| \sum_{i=1}^k y_i \right| \geq K_1^{-1/a} (E y_1^2)^{1/2} (\log n)^{b/a} n^{1/4}\right) \\
&= o(n^{-1/2}) \\
&= o(\varepsilon_n)
\end{aligned}$$

and

$$(2.9) \quad B_{22} = o(\varepsilon_n).$$

Thus we obtain the lemma from (2.5)-(2.9).

### 3. Skorokhod embedding.

Let  $\mathcal{M}_p^q$  be the  $\sigma$ -field generated by the random variables  $\{\xi_p, \xi_{p+1}, \dots, \xi_q\}$  for all  $p$  and  $q$  with  $1 \leq p \leq q \leq M$ . From the condition (1.1) we have

$$\max_{1 \leq k \leq M-1} E \left\{ \sup_{A \in \mathcal{M}_{k+1}^M} |P(A | \mathcal{M}_k^k) - P(A)| \right\} = O(n^{-\theta r}).$$

Thus, from Lemma 1, there exists a sequence of independent random variables  $\{Y_i, i=1, \dots, M\}$  such that each  $Y_i$  has the same distribution as that of  $\xi_i$  and

$$(3.1) \quad \left| P\left(\max_{1 \leq k \leq M} \left| \sum_{i=1}^k \xi_i \right| \geq t\right) - P\left(\max_{1 \leq k \leq M} \left| \sum_{i=1}^k Y_i \right| \geq t\right) \right| = O(n^{-\theta r + 1/2})$$

for any  $t \geq 0$ .

On the other hand, applying the Skorokhod embedding theorem (see e. g. [2]) to  $\{Y_i\}$ , we can construct a Brownian motion  $\{B(t), t \geq 0\}$  and a sequence of independent and positive random variables  $\{T_i, i=1, \dots, M\}$  such that the joint distributions of  $\left\{B(T_1), B(T_1+T_2) - B(T_1), \dots, B\left(\sum_{i=1}^M T_i\right) - B\left(\sum_{i=1}^{M-1} T_i\right)\right\}$  are the same as those of  $\{Y_1, Y_2, \dots, Y_M\}$ , and moreover

$$(3.2) \quad ET_i = EY_i^2 \quad \text{and} \quad E|T_i|^m \leq 2m! E|Y_i|^{2m}, \quad i=1, \dots, M,$$

for each positive integer  $m$ .

Using this argument, we have the following lemma.

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

$$P\left(\sum_{k=1}^M (a_k - a_{k-1}) f\left(a_k, B\left(\sum_{i=1}^k T_i\right)\right) \leq \lambda - 2\varepsilon_n\right) = o(\varepsilon_n)$$

$$\begin{aligned} &\leq P\left(\int_0^1 f(t, \hat{X}_n(t))dt \leq \lambda\right) \\ &\leq P\left(\sum_{k=1}^M (a_k - a_{k-1})f\left(a_k, B\left(\sum_{i=1}^k T_i\right)\right) \leq \lambda + 2\varepsilon_n\right) + o(\varepsilon_n). \end{aligned}$$

The detail of the proof is just the same as that of Lemma 3.3 in Yoshihara [6] and is omitted.

**4. Proof of Theorem 1.**

To prove the theorem, it is enough to show that

$$(4.1) \quad P\left(\left|\sum_{i=1}^M (a_k - a_{k-1})f\left(a_k, B\left(\sum_{i=1}^k T_i\right)\right) - \int_0^1 f(t, B(t))dt\right| \geq \varepsilon_n\right) = o(\varepsilon_n).$$

This is seen in the following way. Once (4.1) holds, we have

$$\begin{aligned} &P\left(\int_0^1 f(t, B(t))dt \leq \lambda - \varepsilon_n\right) - o(\varepsilon_n) \\ &\leq P\left(\sum_{i=1}^M (a_k - a_{k-1})f\left(a_k, B\left(\sum_{i=1}^k T_i\right)\right) \leq \lambda\right) \\ &\leq P\left(\int_0^1 f(t, B(t))dt \leq \lambda + \varepsilon_n\right) + o(\varepsilon_n), \end{aligned}$$

it holds in view of (1.4) that

$$\begin{aligned} &\left|P\left(\sum_{k=1}^M (a_k - a_{k-1})f\left(a_k, B\left(\sum_{i=1}^k T_i\right)\right) \leq \lambda\right) - P\left(\int_0^1 f(t, B(t))dt \leq \lambda\right)\right| \\ &\leq P\left(\lambda - \varepsilon_n < \int_0^1 f(t, B(t))dt < \lambda + \varepsilon_n\right) + o(\varepsilon_n) \\ &\leq 2K_2\varepsilon_n + o(\varepsilon_n). \end{aligned}$$

From this we can prove the theorem from Lemmas 4-6.

Now we shall show (4.1) by the adaptation of Sawyer [3]. Define  $f^A(s, x) \in C^1(\mathbf{R}^2)$  for any  $A > 0$  to be such that

$$f^A(s, x) = f(s, x) \quad \text{for } |x| \leq A,$$

and

$$Df^A(s, k) \leq 2K_1(1 + A^a) \quad \text{for } |x| > A,$$

uniformly in  $s$  and  $x$ , where  $D$  is the operator in (1.3). We then have that

$$(4.2) \quad \begin{aligned} &P\left(\left|\sum_{k=1}^M (a_k - a_{k-1})f\left(a_k, B\left(\sum_{i=1}^k T_i\right)\right) - \int_0^1 f(t, B(t))dt\right| \geq \varepsilon_n\right) \\ &\leq P\left(\left|\sum_{k=1}^M (a_k - a_{k-1})f^A\left(a_k, B\left(\sum_{i=1}^k T_i\right)\right) - \int_0^1 f^A(t, B(t))dt\right| \geq \varepsilon_n\right) \end{aligned}$$

$$+P\left(\sup_{0 \leq t \leq \frac{1}{2}} |B(t)| \geq A\right) + P\left(\sum_{i=1}^M T_i > 2\right) \\ \equiv C_1 + C_2 + C_3, \quad \text{say.}$$

From the assumption (1.7), we have that as  $m \rightarrow \infty$ ,

$$E|X_1|^{2m} \leq 4mK_1 \int_0^1 x^{2m-1} \exp(-\alpha x^\delta) dx \leq \frac{4mK_1}{\alpha^{(2m/\delta)-1}} \Gamma\left(\frac{2m}{\delta}\right) \leq \left(\frac{2m}{\delta}\right)^{2m/\delta}.$$

Thus, from Lemma 2 and (3.2), there exists a positive constant  $K_3$  depending only on  $\delta$  such that

$$(4.3) \quad ET_i = E(y_i - v_i)^2 = a_i - a_{i-1} + O(n^{-1} \log n)$$

and

$$(4.4) \quad ET_i^m \leq 2m! E|y_i - v_i|^{2m} \leq (K_3 m)^{K_3 m} n^{-m/2},$$

for each  $i=1, \dots, M$ . Therefore it follows from Tchebyshev's inequality and Burkholder's inequality that

$$(4.5) \quad C_3 = P\left(\sum_{i=1}^M (T_i - ET_i) > 2 - \sum_{i=1}^M ET_i\right) \\ = P\left(\sum_{i=1}^M (T_i - ET_i) > 1 + o(1)\right) \\ \leq 2E \left| \sum_{i=1}^M (T_i - ET_i) \right|^2 \\ \leq K_4 \sum_{i=1}^M E|y_i - v_i|^4 \\ \leq K_5 n^{-1/2} \\ = o(\varepsilon_n),$$

where  $K_4$  and  $K_5$  are absolute positive constants. Furthermore, setting  $A = \log n$ , we also obtain that

$$(4.6) \quad C_2 \leq 2P(|B(1)| \geq \sqrt{2}A) \leq \sqrt{2}A^{-1} \exp(-A^2) = o(\varepsilon_n).$$

Finally we shall estimate  $C_1$  as follows. According to the proof in [3], we see easily that

$$\left| \sum_{k=1}^M (a_k - a_{k-1}) f^A\left(a_k, B\left(\sum_{i=1}^k T_i\right)\right) - \int_0^1 f(t, B(t)) dt \right| \\ = \sum_{k=0}^{M-1} \int_{\zeta_k}^{\zeta_{k+1}} \frac{\partial}{\partial s} f^A\left(a_k + \bar{\rho}_{kn}(s), B\left(\sum_{i=1}^k T_i\right) + \rho_{kn}(s)\right) (s - a_k) ds$$

$$\begin{aligned}
 & + \sum_{k=0}^{M-1} \int_{\zeta_k}^{\zeta_{k+1}} \frac{\partial}{\partial x} f^A(a_k + \bar{\rho}_{kn}(s), B(\sum_{i=1}^k T_i) + \rho_{kn}(s))(B(s) - B(\sum_{i=1}^k T_i)) ds \\
 & + \sum_{k=0}^{M-1} f^A(a_k, B(\sum_{i=1}^k T_i))(T_{k+1} - ET_{k+1}) + f^A(0, 0)a_1 \\
 & - f^A(1, B(\sum_{i=1}^M T_i))(a_M - a_{M-1}) + f^A(a_{M-1}, B(\sum_{i=1}^M T_i))(a_M - a_{M-1}) \\
 & - \sum_{k=0}^{M-1} f^A(a_k, B(\sum_{i=1}^k T_i))([\theta \log n]/n + O(n^{-1})) \\
 & - \int_1^{\zeta_M} f^A(t, B(t)) dt \\
 & \equiv \sum_{i=1}^8 G_i, \text{ say.}
 \end{aligned}$$

where  $|\bar{\rho}_{kn}(s)| \leq 1$ ,  $|\rho_{kn}(s)| \leq 1$  and

$$\zeta_k = \sum_{i=1}^k T_i \quad \text{for } k=1, \dots, M \text{ and } \zeta_0=0.$$

By Tchebyshev's inequality, we have that

$$\begin{aligned}
 (4.7) \quad C_1 &= P\left(\left|\sum_{i=1}^8 G_i\right| \geq \epsilon_n\right) \leq \sum_{i=1}^8 P(|G_i| \geq \epsilon_n/8) \\
 &\leq \sum_{i=1}^8 (\epsilon_n/8)^{-m} E|G_i|^m.
 \end{aligned}$$

Now we have from the definition of  $f^A(\cdot, \cdot)$  that

$$E|G_3|^m \leq (2K_1)^m (\log n)^{m a} E \left| \sum_{k=1}^M (T_k - ET_k) \right|^m.$$

Thus, using Burkholder's inequality and the inequality (4.4), we have that

$$\begin{aligned}
 (\epsilon_n/8)^{-m} E|G_3|^m &\leq (K_1/4)^m (K_3 m)^{K_3 m} (\log n)^{(a-\eta)m} \\
 &\leq (K_6 m)^{K_6 m} (\log n)^{(a-\eta)m},
 \end{aligned}$$

where  $K_6$  is a positive constant depending only on  $\delta$ . Therefore, if we set  $m = \lceil (\log n)/K_6 \rceil$  then

$$(\epsilon_n/8)^{-m} E|G_3|^m \leq n^{(1+(a-\eta)/K_6) \log \log n} = o(\epsilon_n),$$

for some  $\eta > a + K_6$ .

Now we can also estimate other  $E|G_i|^m$  in (4.7) to have

$$(\epsilon_n/8)^{-m} E|G_i|^m = o(\epsilon_n)$$

for each  $i$ . We therefore have that  $C_1 = o(\epsilon_n)$  and finish the proof of Theorem 1

from (4.2), (4.5) and (4.6).

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