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# Research Article Precise Rates in Log Laws for NA Sequences

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Let  $X_1, X_2, ...$  be a strictly stationary sequence of negatively associated (NA) random variables with  $EX_1 = 0$ , set  $S_n = X_1 + \cdots + X_n$ , suppose that  $\sigma^2 = EX_1^2 + 2\sum_{n=2}^{\infty} EX_1X_n > 0$  and  $EX_1^2 < \infty$ , if  $-1 < \alpha \le 1$ ;  $EX_1^2 (\log |X_1|)^{\alpha} < \infty$ , if  $\alpha > 1$ . We prove  $\lim_{\epsilon \downarrow 0} \epsilon^{2\alpha+2} \sum_{n=1}^{\infty} ((\log n)^{\alpha}/n)P(|S_n| \ge \sigma(\epsilon + \kappa_n)\sqrt{2n\log n}) = 2^{-(\alpha+1)}(\alpha+1)^{-1}E|N|^{2\alpha+2}$ , where  $\kappa_n = O(1/\log n)$  and N is the standard normal random variable.

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# 1. Introduction

A finite family of random variables,  $X_1, X_2, ..., X_n$ , is said to be NA if, for every pair of disjoint subsets  $T_1$  and  $T_2$  of  $\{1, 2, ..., n\}$ ,

$$Cov(f_1(X_i, i \in T_1), f_2(X_j, j \in T_2)) \le 0,$$
(1.1)

whenever  $f_1$  and  $f_2$  are coordinatewise increasing and the covariance exists. An infinite family is NA if every finite subfamily is NA. This definition was introduced by Alam and Saxena [1] and Joag-Dev and Proschan [2], and has found many applications in percolation theory, multivariate statistical analysis, and reliability theory (see, e.g., Barlow and Proschan [3]).

Let  $\{X_n : n \ge 1\}$  be a sequence of NA random variables on some probability space  $(\Omega, \mathcal{F}, P)$  with mean zero and finite variance. As usual, set  $S_0 = 0$ ,  $S_n = X_1 + \cdots + X_n$ ,  $n \ge 1$ , and write  $\sigma_n^2 = ES_n^2$ . Under appropriate covariance conditions, many limit theorems have been obtained. For example, the central limit theorem was proved by Newman [4].

THEOREM 1.1. Let  $\{X_n : n \ge 1\}$  be strictly stationary NA sequences with mean zero and

$$0 < \sigma^2 = EX_1^2 + 2\sum_{n=2}^{\infty} EX_1 X_n < \infty,$$
(1.2)

then

$$\frac{S_n}{(\sigma\sqrt{n})} \xrightarrow{\mathfrak{D}} N(0,1), \quad as \ n \longrightarrow \infty.$$
(1.3)

Further results are three series theorems (see, e.g., Matula [5]), probability inequalities (cf. Roussas [6], Shao [7]), weak convergence (see, e.g., Zhang [8]), the complete convergence (cf. Liang and Su [9], Liang [10]), and the law of the iterated logarithm (see, e.g., Shao and Su [11], Zhang [12]), and so forth.

Note that in the above-mentioned limit theorems, the convergence rates of logarithm are little known, the purpose of the present paper is to investigate the precise asymptotics in the law of the logarithm for NA sequences. It is well known that NA sequences can contain independent random variables as special case, many authors have given lots of beautiful results for independent variables. Let us first recall parts of those results, it is very convenient to adopt the following notations: let  $X_1, X_2, \ldots$  be independent and identically distributed (i.i.d.) nondegenerate random variables with  $EX_1 = 0$  and  $EX_1^2 = \sigma^2 < \infty$ , set  $S_n = X_1 + \cdots + X_n$ ,  $\log x = \log_e(x \lor e)$ . Chow and Lai [13] studied the following results.

THEOREM 1.2. Suppose that  $\operatorname{Var} X_1 = \sigma^2$  and  $\alpha \ge 1$ . Then the following are equivalent:

$$\sum_{n=1}^{\infty} n^{\alpha-2} P\Big( |S_n| \ge \epsilon \sqrt{2n \log n} \Big) < \infty, \quad \forall \epsilon > \sigma \sqrt{\alpha - 1};$$

$$\sum_{n=1}^{\infty} n^{\alpha-2} P\Big( \max_{1 \le k \le n} |S_k| \ge \epsilon \sqrt{2n \log n} \Big) < \infty, \quad \forall \epsilon > \sigma \sqrt{\alpha - 1};$$

$$\sum_{n=1}^{\infty} n^{\alpha-2} P\Big( |S_n| \ge \epsilon \sqrt{2n \log n} \Big) < \infty, \quad for \ some \ \epsilon > 0;$$

$$EX_1 = 0, \qquad \frac{E |X_1|^{2\alpha}}{(\log |X_1|)^{\alpha}} < \infty.$$
(1.4)

Heyde [14] presented an interesting and beautiful result. THEOREM 1.3. If  $EX_1 = 0$  and  $EX_1^2 < \infty$ , then

$$\lim_{\epsilon \to 0} \epsilon^2 \sum_{n=1}^{\infty} P(|S_n| \ge \epsilon n) = EX_1^2.$$
(1.5)

This is a precise estimate for the convergence rate of probability series as  $\epsilon \downarrow 0$ , which has been generalized and extended in several directions. For  $\alpha = 1$  in Theorem 1.2, Gut and Sp*ă*taru [15] obtained the results as follows.

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THEOREM 1.4. Suppose that  $EX_1 = 0$  and  $EX_1^2 = \sigma^2 < \infty$ . Then, for  $0 \le \delta \le 1$ ,

$$\lim_{\epsilon \downarrow 0} \epsilon^{2\delta+2} \sum_{n=1}^{\infty} \frac{(\log n)^{\delta}}{n} P\Big( |S_n| \ge \epsilon \sqrt{n \log n} \Big) = \frac{\sigma^{2\delta+2} E|N|^{2\delta+2}}{\delta+1},$$
(1.6)

where N is a standard normal random variable.

Our starting point is Theorem 1.4, the present work will give the analogue of (1.6) for NA sequences. From now on, we adopt the following notations: let  $X_1, X_2,...$  be strictly stationary NA sequences with  $EX_1 = 0$  and  $EX_1^2 < \infty$ ,  $\sigma^2 = EX_1^2 + 2\sum_{n=2}^{\infty} EX_1X_n > 0$ , and set  $S_n = X_1 + \cdots + X_n$ ,  $M_n = \max_{1 \le k \le n} |S_k|$ , write log for the natural logarithm,  $\log x = \log_e(x \lor e)$ , [*z*] denotes the largest integer which is not larger than *z*, *C* denotes positive constant, independent of  $\epsilon$ , it may take different values in each appearance. The paper is organized as follows: we first introduce our main results, after which the proofs of Theorems 2.1 and 2.4 are exposed in Sections 3 and 4, respectively. We now state the main results.

#### 2. Main results

THEOREM 2.1. Let  $\kappa_n = O(1/\log n)$ ,  $EX_1^2 < \infty$ , *if*  $-1 < \alpha \le 1$ ;  $EX_1^2(\log |X_1|)^{\alpha} < \infty$ , *if*  $\alpha > 1$ . Then

$$\lim_{\epsilon \downarrow 0} \epsilon^{2\alpha+2} \sum_{n=1}^{\infty} \frac{(\log n)^{\alpha}}{n} P\Big( |S_n| \ge \sigma(\epsilon + \kappa_n) \sqrt{2n \log n} \Big)$$
  
= 2<sup>-(\alpha+1)</sup>(\alpha + 1)^{-1} E|N|^{2\alpha+2}, (2.1)

where N is a standard normal random variable.

COROLLARY 2.2. Under the conditions in Theorem 2.1,

$$\lim_{\epsilon \downarrow 0} \epsilon^{2\alpha+2} \sum_{n=1}^{\infty} \frac{(\log n)^{\alpha}}{n} P\Big( |S_n| \ge \sigma \epsilon \sqrt{2n \log n} \Big) = \frac{E|N|^{2\alpha+2}}{2^{(\alpha+1)}(\alpha+1)}.$$
 (2.2)

COROLLARY 2.3. Suppose that  $EX_1^2 < \infty$ . Then

$$\lim_{\epsilon \downarrow 0} \epsilon^2 \sum_{n=1}^{\infty} n^{-1} P\Big( |S_n| \ge \sigma(\epsilon + \kappa_n) \sqrt{2n \log n} \Big) = \frac{EN^2}{2}.$$
(2.3)

THEOREM 2.4. Let  $\kappa_n = O(1/\log n)$ ,  $EX_1^2 < \infty$ ,  $if -1 < \alpha \le 1/2$ ;  $EX_1^2(\log |X_1|)^{\alpha} < \infty$ ,  $if \alpha > 1/2$ . Then

$$\lim_{\epsilon \downarrow 0} \epsilon^{2\alpha+2} \sum_{n=1}^{\infty} \frac{(\log n)^{\alpha}}{n} P\left(M_n \ge \sigma\left(\epsilon + \kappa_n\right) \sqrt{2n \log n}\right)$$
$$= 2^{-\alpha} (\alpha+1)^{-1} E|N|^{2\alpha+2} \sum_{n=0}^{\infty} \frac{(-1)^n}{(2n+1)^{2\alpha+2}}.$$
(2.4)

Without loss of generality, throughout the paper, we will suppose that  $\sigma^2 = 1$ . Let  $\Phi(x)$  denote the standard normal distribution function, and put  $\Psi(x) = 1 - \Phi(x) + \Phi(-x)$ ,  $x \ge 0$ .

## 3. Proof of Theorem 2.1

In order to prove this result easily, we separate the proof into two propositions, the first one can be formulated as follows.

**PROPOSITION 3.1.** Suppose that N be a nondegenerate Gaussian random variable. Then

$$\lim_{\epsilon \downarrow 0} \epsilon^{2\alpha+2} \sum_{n=1}^{\infty} \frac{(\log n)^{\alpha}}{n} P(|N| \ge (\epsilon + \kappa_n) \sqrt{2\log n})$$
$$= 2^{-(\alpha+1)} (\alpha+1)^{-1} E|N|^{2(\alpha+1)}.$$
(3.1)

*Proof.* Noting the definition of  $\kappa_n$ , we first show that

$$\lim_{\epsilon \downarrow 0} \epsilon^{2\alpha+2} \sum_{n=1}^{\infty} \frac{(\log n)^{\alpha}}{n} P\Big(|N| \ge \epsilon \sqrt{2\log n}\Big) = 2^{-(\alpha+1)} (\alpha+1)^{-1} E|N|^{2\alpha+2}.$$
(3.2)

By integral formula and transformation, it is enough to show that for any  $\alpha > -1$ ,

$$\lim_{\epsilon \to 0} \epsilon^{2\alpha+2} \sum_{n=1}^{\infty} \frac{(\log n)^{\alpha}}{n} P(|N| \ge \epsilon \sqrt{2\log n})$$

$$= \lim_{\epsilon \to 0} \epsilon^{2\alpha+2} \sum_{n=1}^{\infty} \int_{n}^{n+1} \frac{(\log x)^{\alpha}}{x} P(|N| \ge \epsilon \sqrt{2\log x}) dx$$

$$= 2^{-\alpha} \int_{0}^{\infty} y^{2\alpha+1} P(|N| \ge y) dy$$

$$= 2^{-(\alpha+1)} (\alpha+1)^{-1} E|N|^{2\alpha+2}.$$
(3.3)

Write

$$A_{n}(\epsilon) = \left| P\left( |N| \ge \epsilon \sqrt{2\log n} \right) - P\left( |N| \ge (\epsilon + \kappa_{n}) \sqrt{2\log n} \right) \right|.$$
(3.4)

The proof of (3.1) should be completed, if one could show that

$$\lim_{\epsilon \downarrow 0} \epsilon^{2\alpha+2} \sum_{n=1}^{\infty} \frac{(\log n)^{\alpha}}{n} A_n(\epsilon) = 0,$$
(3.5)

 $\Box$ 

the proof of (3.5) is similar to that of Proposition 2.2 in Huang and Zhang [16].

Before giving the second proposition, the following lemma is necessary.

LEMMA 3.2 [17]. Suppose that  $\{X_k : k \ge 1\}$  be NA sequences with  $EX_k = 0$ ,  $E|X_k|^p < \infty$ , for  $p \ge 2$ . Then, for any t > p/2, x > 0,

$$P(|S_n| \ge x) \le \sum_{k=1}^{n} P\left(|X_k| \ge \frac{x}{t}\right) + 2e^t \left(1 + \frac{x^2}{t \sum_{k=1}^{n} EX_k^2}\right)^{-t}.$$
(3.6)

PROPOSITION 3.3. Suppose that  $EX_1^2 < \infty$ , if  $-1 < \alpha \le 1$ ;  $EX_1^2(\log |X_1|)^{\alpha} < \infty$ , if  $\alpha > 1$ . Then

$$\lim_{\epsilon \downarrow 0} \epsilon^{2\alpha+2} \sum_{n=1}^{\infty} \frac{(\log n)^{\alpha}}{n} \left| P\left( \left| S_n \right| \ge (\epsilon + \kappa_n) \sqrt{2n \log n} \right) - P\left( \left| N \right| \ge (\epsilon + \kappa_n) \sqrt{2 \log n} \right) \right| = 0.$$
(3.7)

*Proof.* Set  $H(\epsilon) = [\exp(M/\epsilon^2)]$ , where  $M > 4, 0 < \epsilon < 1/4$ . It is easy to get

$$\sum_{n=1}^{\infty} \frac{(\log n)^{\alpha}}{n} \left| P\left( \left| S_{n} \right| \ge (\epsilon + \kappa_{n})\sqrt{2n\log n} \right) - P\left( \left| N \right| \ge (\epsilon + \kappa_{n})\sqrt{2\log n} \right) \right|$$

$$= \sum_{n \le H(\epsilon)} \frac{(\log n)^{\alpha}}{n} \left| P\left( \left| S_{n} \right| \ge (\epsilon + \kappa_{n})\sqrt{2n\log n} \right) - P\left( \left| N \right| \ge (\epsilon + \kappa_{n})\sqrt{2\log n} \right) \right|$$

$$+ \sum_{n > H(\epsilon)} \frac{(\log n)^{\alpha}}{n} \left| P\left( \left| S_{n} \right| \ge (\epsilon + \kappa_{n})\sqrt{2n\log n} \right) - P\left( \left| N \right| \ge (\epsilon + \kappa_{n})\sqrt{2\log n} \right) \right| \stackrel{\triangle}{=} I_{1} + I_{2}.$$
(3.8)

We first consider  $I_1$ . Let  $\Delta_n = \sup_x |P(|S_n| \ge x\sqrt{n}) - P(|N| \ge x)|$ , noting Theorem 1.1, since  $\Psi(x)$  is a continuous function, then, for any  $x \ge 0$ , we have  $\lim_{n\to\infty} \Delta_n = 0$ . It follows that

$$\epsilon^{2\alpha+2}I_{1} \leq \epsilon^{2\alpha+2} \sum_{n \leq H(\epsilon)} \frac{(\log n)^{\alpha}}{n} \Delta_{n} = \epsilon^{2\alpha+2} \sum_{n \leq H(\epsilon)} \frac{(\log n)^{\alpha}}{n} \Delta_{n}$$

$$\leq C\epsilon^{2\alpha+2} (\log n)^{\alpha+1} \frac{1}{(\log n)^{\alpha+1}} \sum_{n \leq H(\epsilon)} \frac{(\log n)^{\alpha}}{n} \Delta_{n}$$

$$\leq CM^{\alpha+1} \frac{1}{(\log n)^{\alpha+1}} \sum_{n \leq H(\epsilon)} \frac{(\log n)^{\alpha}}{n} \Delta_{n} \longrightarrow 0, \quad \text{as } \epsilon \neq 0.$$
(3.9)

Note that  $(1/(\log n)^{\alpha+1}) \sum_{n \le H(\epsilon)} ((\log n)^{\alpha}/n) \Delta_n \to 0, \epsilon \downarrow 0$ , then (3.9) holds. Turn to  $I_2$ , one can get

$$I_{2} \leq \sum_{n>H(\epsilon)} \frac{(\log n)^{\alpha}}{n} P\Big(|N| \geq (\epsilon + \kappa_{n})\sqrt{2\log n}\Big) \\ + \sum_{n>H(\epsilon)} \frac{(\log n)^{\alpha}}{n} P\Big(|S_{n}| \geq (\epsilon + \kappa_{n})\sqrt{2n\log n}\Big) \stackrel{\triangle}{=} I_{3} + I_{4}.$$

$$(3.10)$$

Without loss of generality, we can assume that  $0 < \epsilon < 1/4$ , M > 4, then one can get  $H(\epsilon) - 1 \ge \sqrt{H(\epsilon)}$ . Note, in particular, that the definition of  $\kappa_n$ , for *n* large enough, we have  $|\kappa_n| < \epsilon/4$ . Then, for  $I_3$ , it follows that

$$\begin{aligned} \epsilon^{2\alpha+2}I_{3} &\leq \epsilon^{2\alpha+2} \sum_{n>H(\epsilon)} \frac{(\log n)^{\alpha}}{n} P\Big(|N| \geq \frac{\epsilon}{2} \sqrt{2\log n}\Big) \\ &\leq \epsilon^{2\alpha+2} \int_{H(\epsilon)-1}^{\infty} \frac{(\log x)^{\alpha}}{x} P\Big(|N| \geq \frac{\epsilon}{2} \sqrt{2\log x}\Big) dx \\ &\leq \epsilon^{2\alpha+2} \int_{\sqrt{H(\epsilon)}}^{\infty} \frac{(\log x)^{\alpha}}{x} P\Big(|N| \geq \frac{\epsilon}{2} \sqrt{2\log x}\Big) dx \\ &\leq C \int_{\sqrt{M/4}}^{\infty} y^{2\alpha+1} P\big(|N| > y\big) dy \longrightarrow 0, \quad \text{as } M \longrightarrow \infty, \end{aligned}$$
(3.11)

uniformly with respect to  $0 < \epsilon < 1/4$ . We finally estimate  $I_4$ , by Lemma 3.2, which yields, for *n* large enough,

$$P(|S_n| \ge (\epsilon + \kappa_n)\sqrt{2n\log n}) \le P(|S_n| \ge \frac{\epsilon}{2}\sqrt{2n\log n})$$
  
$$\le nP(|X_1| \ge \frac{\epsilon}{2m}\sqrt{2n\log n}) + 2e^m \left(1 + \frac{\epsilon^2\log n}{2mEX_1^2}\right)^{-m}$$
  
$$\stackrel{\triangle}{=} I_5 + I_6,$$
(3.12)

where *m* is a positive integer to be specified later. Then, observe that  $n > H(\epsilon)$  implies  $\log n > M/\epsilon^2$ , for  $I_5$ , if  $-1 < \alpha \le 1$ , the proof is similar to that of Lemma 3.2 [15]; if  $\alpha > 1$ , applying Fubini's theorem, it turns out that

$$\sum_{n>H(\epsilon)} \frac{(\log n)^{\alpha}}{n} I_{5} = \sum_{n>H(\epsilon)} (\log n)^{\alpha} P\left(\left|X_{1}\right| \geq \frac{\epsilon \sqrt{2n \log n}}{2m}\right)$$

$$\leq C \sum_{n>H(\epsilon)} (\log n)^{\alpha} \sum_{j=n}^{\infty} P\left(\frac{\sqrt{2Mj}}{2m} \leq |X_{1}| < \frac{\sqrt{2M(j+1)}}{2m}\right)$$

$$\leq C \sum_{j>H(\epsilon)} P\left(\frac{\sqrt{2Mj}}{2m} \leq |X_{1}| < \frac{\sqrt{2M(j+1)}}{2m}\right) \sum_{n=H(\epsilon)}^{j} (\log n)^{\alpha}$$

$$\leq C \sum_{j>H(\epsilon)} j(\log j)^{\alpha} P\left(j \leq \frac{2m^{2}X_{1}^{2}}{M} < j+1\right) \leq \frac{CEX_{1}^{2}(\log |X_{1}|)^{\alpha}}{M}.$$
(3.13)

Furthermore, one can easily get  $\limsup_{\epsilon \downarrow 0} e^{2\alpha+2} \sum_{n>H(\epsilon)} (\log n)^{\alpha} I_5/n = 0$ . We finally estimate  $I_6$ , by the arbitrarity of m(>1), one can obviously choose an appropriate positive

integer *m*, such that  $m > \alpha + 1$ . Then we have

$$\sum_{n>H(\epsilon)} \frac{(\log n)^{\alpha}}{n} I_{6} \leq C \sum_{n>H(\epsilon)} \frac{(\log n)^{\alpha}}{n} \left(\frac{\epsilon^{2} \log n}{2mEX_{1}^{2}}\right)^{-m}$$

$$\leq C \sum_{n>H(\epsilon)} \frac{(\log n)^{\alpha}}{n} (\epsilon^{2} \log n)^{-m}$$

$$\leq C\epsilon^{-2m} \int_{H(\epsilon)-1}^{\infty} \left(\frac{(\log x)^{\alpha-m}}{x}\right) dx$$

$$\leq C\epsilon^{-2m} (\log (H(\epsilon)))^{\alpha+1-m}$$

$$\leq C\epsilon^{-2\alpha-2} M^{\alpha+1-m},$$
(3.14)

it is easy to get  $\lim_{M\to\infty} \epsilon^{2\alpha+2} \sum_{n>H(\epsilon)} (\log n)^{\alpha} I_6/n = 0$ , uniformly with respect to  $0 < \epsilon < 1/4$ . Thus the proof of Proposition 3.3 is completed.

*Proof of Theorem 2.1.* Combining Propositions 3.1 and 3.3, one can complete the proof of this theorem immediately.

## 4. Proof of Theorem 2.4

The following propositions will simplify the proof of Theorem 2.4, which are stated as follows.

**PROPOSITION 4.1.** Suppose that  $\{W(t) : t \ge 0\}$  be a standard Wiener process (Brownian motion). Then

$$\lim_{\epsilon \downarrow 0} \epsilon^{2\alpha+2} \sum_{n=1}^{\infty} \frac{(\log n)^{\alpha}}{n} P\left(\sup_{0 \le s \le 1} |W(s)| \ge (\epsilon + \kappa_n) \sqrt{2\log n}\right)$$

$$= 2^{-\alpha} (\alpha+1)^{-1} E|N|^{2\alpha+2} \sum_{n=0}^{\infty} \frac{(-1)^n}{(2n+1)^{2\alpha+2}}.$$
(4.1)

Proof. Noting the result of Billingsley [18],

$$P\left(\sup_{0\leq s\leq 1} |W(s)| \geq x\right) = 1 - \sum_{k=-\infty}^{\infty} (-1)^{k} P\left((2k-1)x \leq N \leq (2k+1)x\right)$$
$$= 4 \sum_{k=0}^{\infty} (-1)^{k} P\left(N \geq (2k+1)x\right) = 2 \sum_{k=0}^{\infty} (-1)^{k} P\left(|N| \geq (2k+1)x\right),$$
(4.2)

where N is the standard normal random variable. Then, according to Proposition 3.1, one can complete the proof easily.

LEMMA 4.2 [7]. Suppose that  $\{X_n : n \ge 1\}$  be strictly stationary NA sequences,  $EX_1 = \mu$ ,  $0 < \text{Var} X_1 = \sigma^2 < \infty$ , and  $B^2 = EX_1^2 + 2\sum_{n=2}^{\infty} EX_1 X_n > 0$ , set  $S_m = \sum_{k=1}^m X_k$ , write

$$W_n(t) = \frac{1}{B\sqrt{n}} \left( S_m + (nt - m)X_{m+1} - nt\mu \right), \quad m \le n < m+1, \ 0 \le t \le T.$$
(4.3)

Then

$$W_n(t) \xrightarrow{\mathfrak{G}} W(t) \quad in \ C[0,T],$$

$$(4.4)$$

where W(t) is the standard Wiener process and C[0,T] is the usual C space on [0,T].

LEMMA 4.3 [11]. Let  $\{X_n : n \ge 1\}$  be a sequence of NA random variable with mean zero and finite second moments. Set  $S_n = X_1 + \cdots + X_n$  and  $B_n^2 = \sum_{k=1}^n EX_k^2$ . Then for all x > 0, a > 0, and  $0 < \beta < 1$ ,

$$P\left(\max_{1\le k\le n} |S_k| \ge x\right) \le 2P\left(\max_{1\le k\le n} |X_k| \ge a\right) + \frac{2}{1-\beta} \exp\left(-\frac{\beta x^2}{2(ax+B_n^2)}\right).$$
(4.5)

PROPOSITION 4.4. Suppose that  $EX_1^2 < \infty$ , if  $-1 < \alpha \le 1/2$ ;  $EX_1^2(\log |X_1|)^{\alpha} < \infty$ , if  $\alpha > 1/2$ . Then

$$\lim_{\epsilon \downarrow 0} \epsilon^{2\alpha+2} \sum_{n \ge 1} \frac{(\log n)^{\alpha}}{n} \left| P\left( M_n \ge (\epsilon + \kappa_n) \sqrt{2n \log n} \right) - P\left( \sup_{0 \le s \le 1} |W(s)| \ge (\epsilon + \kappa_n) \sqrt{2 \log n} \right) \right| = 0.$$

$$(4.6)$$

*Proof.* Let  $H(\epsilon)$  be as above, it follows that

$$\sum_{n=1}^{\infty} \frac{(\log n)^{\alpha}}{n} \left| P\left(M_n \ge (\epsilon + \kappa_n)\sqrt{2n\log n}\right) - P\left(\sup_{0\le s\le 1} |W(s)| \ge (\epsilon + \kappa_n)\sqrt{2\log n}\right) \right|$$

$$= \sum_{n\le H(\epsilon)} \frac{(\log n)^{\alpha}}{n} \left| P\left(M_n \ge (\epsilon + \kappa_n)\sqrt{2n\log n}\right) - P\left(\sup_{0\le s\le 1} |W(s)| \ge (\epsilon + \kappa_n)\sqrt{2\log n}\right) \right|$$

$$+ \sum_{n>H(\epsilon)} \frac{(\log n)^{\alpha}}{n} \left| P\left(M_n \ge (\epsilon + \kappa_n)\sqrt{2n\log n}\right) - P\left(\sup_{0\le s\le 1} |W(s)| \ge (\epsilon + \kappa_n)\sqrt{2\log n}\right) \right|$$

$$\stackrel{\triangle}{=} I'_1 + I'_2. \tag{4.7}$$

Noting Lemma 4.2, we have  $M_n/\sqrt{n} \xrightarrow{\mathfrak{D}} \sup_{0 \le t \le 1} |W(t)|$ , as  $n \to \infty$ . Similar to Theorem 2.1, one can get  $\lim_{\epsilon \downarrow 0} \epsilon^{2\alpha+2} I'_1 = 0$ . We now estimate  $I'_2$ , it turns out that

$$I_{2}^{\prime} \leq \sum_{n > H(\epsilon)} \frac{(\log n)^{\alpha}}{n} P\left(\sup_{0 \leq s \leq 1} |W(s)| \geq (\epsilon + \kappa_{n})\sqrt{2\log n}\right) + \sum_{n > H(\epsilon)} \frac{(\log n)^{\alpha}}{n} P\left(\max_{1 \leq k \leq n} |S_{k}| \geq (\epsilon + \kappa_{n})\sqrt{2n\log n}\right) \stackrel{\triangle}{=} I_{3}^{\prime} + I_{4}^{\prime}.$$

$$(4.8)$$

Observe that  $P(\sup_{0 \le s \le 1} |W(s)| \ge x) \le 2P(|N| \ge x)$ , see [18]. Similar to Theorem 2.1, we have  $\lim_{\epsilon \ge 0} \epsilon^{2\alpha+2} I'_3 = 0$ . We then consider  $I'_4$ , as a matter of fact, by Lemma 4.3, take  $x = \epsilon \sqrt{2n \log n}/2$ ,  $a = (2n\epsilon \sqrt{\log n})^{1/2}$ . For *n* large enough, one could get

$$P\left(\max_{1\leq k\leq n} |S_{k}| \geq (\epsilon+\kappa_{n})\sqrt{2n\log n}\right) \leq P\left(\max_{1\leq k\leq n} |S_{k}| \geq \frac{\epsilon}{2}\sqrt{2n\log n}\right)$$
$$\leq 2nP\left(|X_{1}| \geq \left(2n\epsilon\sqrt{\log n}\right)^{1/2}\right) + \frac{2}{1-\beta}\exp\left(-\frac{\beta\epsilon^{2}\log n}{8\left(\left(\epsilon\sqrt{\log n}\right)^{3/2}+1\right)}\right) \triangleq I_{5}'+I_{6}'.$$

$$(4.9)$$

Without loss of generality, we can assume  $0 < \epsilon < 1/4$ , M > 16, Notice that  $n > H(\epsilon)$  if and only if  $\log n > M/\epsilon^2$ , then for  $I'_6$ , we have

$$\epsilon^{2\alpha+2} \sum_{n>H(\epsilon)} \frac{(\log n)^{\alpha} I_{6}'}{n} \leq C\epsilon^{2\alpha+2} \sum_{n>H(\epsilon)} (\log n)^{\alpha} \exp\left(\frac{-\beta\left(\epsilon\sqrt{\log n}\right)^{1/2}/9\right)}{n}$$

$$\leq C\epsilon^{2\alpha+2} \int_{H(\epsilon)-1}^{\infty} (\log x)^{\alpha} \exp\left(\frac{-\beta\left(\epsilon\sqrt{\log x}\right)^{1/2}}{9}\right) \frac{dx}{x}$$

$$\leq C\epsilon^{2\alpha+2} \int_{\sqrt{H(\epsilon)}}^{\infty} (\log x)^{\alpha} \exp\left(\frac{-\beta\left(\epsilon\sqrt{\log x}\right)^{1/2}}{9}\right) \frac{dx}{x}$$

$$\leq C\epsilon^{2\alpha+2} \int_{\beta\sqrt[4]{M/9}}^{\infty} \epsilon^{-2\alpha-2} y^{4\alpha+3} \exp(-y) dy$$

$$\leq C \int_{\beta\sqrt[4]{M/9}}^{\infty} y^{4\alpha+3} \exp(-y) dy \longrightarrow 0, \quad \text{as } M \longrightarrow \infty.$$
(4.10)

r

We then estimate  $I'_5$ . If  $-1 < \alpha \le 1/2$ , the following result is obvious according to  $EX_1^2 < \infty$ ; if  $\alpha > 1/2$ , by Fubini's theorem, it follows that

$$\sum_{n>H(\epsilon)} \frac{(\log n)^{\alpha} I_{5}'}{n} = \sum_{n>H(\epsilon)} (\log n)^{\alpha} P\left(|X_{1}| \ge \left(2n\epsilon\sqrt{\log n}\right)^{1/2}\right)$$

$$\leq C \sum_{n>H(\epsilon)} (\log n)^{\alpha} \sum_{k=n}^{\infty} P\left(\sqrt{k} \le \frac{|X_{1}|}{\sqrt[4]{4M}} < \sqrt{k+1}\right)$$

$$\leq C \sum_{k>H(\epsilon)} P\left(\sqrt{k} \le \frac{|X_{1}|}{\sqrt[4]{4M}} < \sqrt{k+1}\right) \sum_{n=H(\epsilon)}^{k} (\log n)^{\alpha} \qquad (4.11)$$

$$\leq C \sum_{k>H(\epsilon)} k (\log k)^{\alpha} P\left(\sqrt{k} \le \frac{|X_{1}|}{\sqrt[4]{4M}} < \sqrt{k+1}\right)$$

$$\leq C E X_{1}^{2} \frac{(\log |X_{1}|)^{\alpha}}{\sqrt{4M}} < \infty.$$

 $\square$ 

Proof of Theorem 2.4. The proof follows from Propositions 4.1 and 4.4.

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