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# A General Law of Precise Asymptotics for the Complete Moment Convergence\*\*\*

Yong ZHANG\* Xiaoyun YANG\* Zhishan DONG\*\*

**Abstract** The authors achieve a general law of precise asymptotics for a new kind of complete moment convergence of i.i.d. random variables, which includes complete convergence as a special case. It can describe the relations among the boundary function, weighted function, convergence rate and limit value in studies of complete convergence. This extends and generalizes the corresponding results of Liu and Lin in 2006.

**Keywords** Complete moment convergence, General law, Precise asymptotics **2000 MR Subject Classification** 60F15, 60G50

#### 1 Introduction

Throughout this paper, let  $\{X, X_n, n \geq 1\}$  be a sequence of i.i.d. random variables,  $S_n = \sum_{i=1}^n X_i$  and N be the standard normal random variable. C denotes positive constants, possibly varying from place to place, and [x] denotes the largest integer  $\leq x$ .

Since P. L. Hsu and H. Robbins [6] introduced the concept of complete convergence, there have been extensions in several directions. One of them is to discuss the precise rate and limit value of  $\sum_{n=1}^{\infty} \varphi(n)P\{|S_n| \geq \varepsilon g(n)\}$  as  $\varepsilon \downarrow a$ ,  $a \geq 0$ , where  $\varphi(x)$  and g(x) are the positive functions defined on  $[0, \infty)$ . We call  $\varphi(x)$  and g(x) weighted function and boundary function respectively. The first result in this direction was due to C. C. Heyde [5], who proved that

$$\lim_{\varepsilon \downarrow 0} \varepsilon^2 \sum_{n=1}^{\infty} P\{|S_n| \ge \varepsilon n\} = EX^2, \tag{1.1}$$

where EX = 0 and  $EX^2 < \infty$ . For analogous results in more general case, see [2, 3, 9, 11], etc. The research in this field is called the precise asymptotics. W. D. Liu and Z. Y. Lin [7] studied the precise asymptotics for a new kind of complete moment convergence of i.i.d. random variables.

W. D. Liu and Z. Y. Lin [7] achieved the following three results.

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<sup>\*</sup>School of Mathematics, Jilin University (Qianwei Campus), Changchun 130012, China. E-mail: zyong2661@sina.com

<sup>\*\*</sup>Corresponding author. School of Mathematics, Jilin University (Qianwei Campus), Changchun 130012, China. E-mail: Dongzs@mail.jlu.edu.cn

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Theorem 1.1 Suppose that

$$EX = 0$$
,  $EX^2 = \sigma^2$  and  $EX^2 \log^+ |X| < \infty$ . (1.2)

Then we have

$$\lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} \sum_{n=1}^{\infty} \frac{1}{n^2} E S_n^2 I\{|S_n| \ge \varepsilon n\} = 2\sigma^2.$$
 (1.3)

Conversely, if (1.3) is true, then (1.2) holds.

Theorem 1.2 Suppose that

$$EX = 0, \quad EX^2 = \sigma^2 < \infty. \tag{1.4}$$

Then we have

$$\lim_{\varepsilon \downarrow 0} \varepsilon^{2-p} \sum_{n=1}^{\infty} \frac{1}{n^p} E|S_n|^p I\{|S_n| \ge \varepsilon n\} = \frac{2}{2-p} \sigma^2$$
 (1.5)

for  $0 \le p < 2$ . Conversely, if (1.5) is true for some  $0 \le p < 2$ , then (1.4) holds.

Theorem 1.3 Suppose that

$$EX = 0$$
,  $EX^2 = \sigma^2$  and  $EX^2(\log^+|X|)^\delta < \infty$ , (1.6)

where  $0 < \delta \le 1$ . Then we have

$$\lim_{\varepsilon \downarrow 0} \varepsilon^{2\delta} \sum_{n=1}^{\infty} \frac{(\log n)^{\delta-1}}{n^2} ES_n^2 I\{|S_n| \ge \varepsilon \sqrt{n \log n}\} = \frac{\sigma^{2\delta+2}}{\delta} E|N|^{2\delta+2}. \tag{1.7}$$

Conversely, if (1.7) is true, then (1.6) holds.

In this paper, we will extend the scope of the weighted functions and boundary functions, and give a general law of precise asymptotics of i.i.d. random variables, which extend and generalize the direct part of [7, Theorems 1.1–1.3].

#### 2 Main Results

We will make some appropriate limitations to  $\varphi(x)$  and g(x) in the following theorems, and then get some corollaries according to the kind of g(x). From these corollaries we can conclude a series of interesting results, which contain the direct part of [7, Theorems 1.1–1.3].

**Theorem 2.1** Let g(x) be a positive and differentiable function defined on  $[n_0, \infty)$ , which is strictly increasing to  $\infty$ . g'(x) is monotone. If g'(x) is monotone nondecreasing, we assume  $\lim_{n\to\infty} \frac{g'(n+1)}{g'(n)} = 1$ . Assume that  $\varphi(x) = \frac{g'(x)}{g(x)}$  is monotone, and if  $\varphi(x)$  is monotone nondecreasing, we assume  $\lim_{n\to\infty} \frac{\varphi(n+1)}{\varphi(n)} = 1$ . And assume that the following condition is satisfied:

$$\lim_{n \to \infty} \sup n\varphi(n) < \infty. \tag{2.1}$$

Finally suppose that the following conditions of i.i.d. random variables  $\{X, X_n, n \geq 1\}$  are satisfied:

$$EX = 0$$
,  $EX^2 = \sigma^2$  and  $EX^2 \log^+ |X| < \infty$ . (2.2)

Then we have

$$\lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} \sum_{n=n_0}^{\infty} \varphi(n) E \left| \frac{S_n}{\sqrt{n}} \right|^2 I\{|S_n| \ge \varepsilon \sqrt{ng(n)}\} = 2\sigma^2.$$
 (2.3)

**Theorem 2.2** Let g(x) be a positive and differentiable function defined on  $[n_0, \infty)$ , which is strictly increasing to  $\infty$ . g'(x) is monotone. If g'(x) is monotone nondecreasing, we assume  $\lim_{n\to\infty} \frac{g'(n+1)}{g'(n)} = 1$ . Assume that  $\varphi(x) = \frac{g'(x)}{g^{ps}(x)}$  is monotone, where  $\frac{1}{s} > p \geq 0$ . If  $\varphi(x)$  is monotone nondecreasing, we assume  $\lim_{n\to\infty} \frac{\varphi(n+1)}{\varphi(n)} = 1$ . And assume that the following condition is satisfied:

$$\lim_{n \to \infty} n^{\frac{p}{2}} \varphi(n) < \infty. \tag{2.4}$$

Finally suppose that the following conditions of i.i.d. random variables  $\{X, X_n, n \geq 1\}$  are satisfied:

$$EX = 0, EX^2 = \sigma^2 < \infty \quad for \ 0 \le p < 2;$$
 (2.5)

$$EX = 0, EX^2 = \sigma^2, EX^2 \log^+ |X| < \infty \text{ for } p = 2;$$
 (2.6)

$$EX = 0, EX^2 = \sigma^2, E|X|^p < \infty \text{ for } p > 2.$$
 (2.7)

Then we have

$$\lim_{\varepsilon \downarrow 0} \varepsilon^{\frac{1}{s} - p} \sum_{n = n_0}^{\infty} \varphi(n) E \left| \frac{S_n}{\sqrt{n}} \right|^p I\{ |S_n| \ge \varepsilon \sqrt{n} g^s(n) \} = \frac{\sigma^{\frac{1}{s}}}{1 - ps} E |N|^{\frac{1}{s}}. \tag{2.8}$$

**Remark 2.1** In Theorem 2.1 or Theorem 2.2,  $\lim_{n\to\infty} \frac{\varphi(n+1)}{\varphi(n)} = 1$ , (2.1) and (2.4) are all mild conditions.  $g(x) = x^{\alpha}$ ,  $(\log x)^{\beta}$ ,  $(\log \log x)^{\gamma}$  with some suitable conditions of  $\alpha > 0$ ,  $\beta > 0$ ,  $\gamma > 0$  and some others all satisfy these conditions.

**Remark 2.2** Letting g(x) = x in Theorem 2.1, we can get the direct part of [7, Theorem 1].

**Remark 2.3** If  $s = \frac{1}{2}$ ,  $0 \le p < 2$  and g(x) = x, we can get the direct part of [7, Theorem 2].

**Remark 2.4** If  $s = \frac{1}{2\delta+2}$ , p = 2 and  $g(x) = (\log x)^{\delta+1}$ , we can get the direct part of [7, Theorem 3] with  $EX^2 \log^+ |X| < \infty$  instead of  $EX^2 (\log^+ |X|)^{\delta} < \infty$ . W. D. Liu and Z. Y. Lin [7] restricts  $0 < \delta \le 1$ , but in our theorem we only restrict  $\delta > 0$ .

**Remark 2.5** If p = 0, the conditions of Theorem 2.2 still hold, then we can get [1, Theorem 1] with  $\alpha = 2$ .

**Remark 2.6** If p = 0,  $s = \frac{1}{2}$ ,  $g(x) = \log \log x$ , we can get [3, Theorem 2].

**Remark 2.7** If p = 0,  $s = \frac{2-t}{2(r-t)}$ ,  $g(x) = x^{\frac{r}{t}-1}$  for 0 < t < 2, r > t, we can get [2, Corollary 1].

#### 3 Several Lemmas

To prove our theorems, we need some lemmas as follows.

**Lemma 3.1** (see [1]) Suppose that  $0 < \alpha \le 2$ ,  $\{X, X_n, n \ge 1\}$  is a sequence of i.i.d. random variables which belongs to the normal domain of attraction of a nondegenerate stable distribution  $G_{\alpha}$  with characteristic exponent  $\alpha$  and EX = 0 when  $1 < \alpha \le 2$ . Let g(x) be a positive and differentiable function defined on  $[n_0, \infty)$ , which is strictly increasing to  $\infty$ . g'(x) is monotone. If g'(x) is monotone nondecreasing, we assume  $\lim_{n\to\infty} \frac{g'(n+1)}{g'(n)} = 1$ . Then,  $\forall s > \frac{1}{\alpha}$ , we have

$$\lim_{\varepsilon \downarrow 0} \varepsilon^{\frac{1}{s}} \sum_{n=n_0}^{\infty} g'(n) P\{|S_n - a_n| \ge \varepsilon b_n g^s(n)\} = E|Z_{\alpha}|^{\frac{1}{s}},$$

where  $Z_{\alpha}$  is a random variable having the distribution  $G_{\alpha}$ ,  $a_n$  and  $b_n$  are the centralizing and normalizing constants respectively.

**Lemma 3.2** Suppose that  $\{X, X_n, n \geq 1\}$  is a sequence of i.i.d. random variables with  $E|X|^{\beta} < \infty$ , where  $1 < \beta \leq 2$ . Let  $S_n = \sum_{i=1}^n X_i$ . Then  $\forall x, y > 0$ , we have

$$P\{|S_n| \ge x\} \le nP\{|X| \ge y\} + 2n^{\frac{x}{y}} \left(\frac{eE|X|^{\beta}}{xy^{\beta-1}}\right)^{\frac{x}{y}}.$$
 (3.1)

**Proof** To prove (3.1), it suffices to show that

$$P\{|S_n| \ge x\} \le nP\{|X| \ge y\} + 2e^{\frac{x}{y}} \left(\frac{nE|X|^{\beta}}{nE|X|^{\beta} + xy^{\beta - 1}}\right)^{\frac{x}{y}}.$$
 (3.2)

However, (3.2) is [11, Lemma 2], which, in turn, is based on [8, Theorem 1.2].

**Lemma 3.3** (see [10]) Suppose that  $\{X, X_n, n \ge 1\}$  is a sequence of i.i.d. random variables,  $S_n = \sum_{i=1}^n X_i$ . We have

$$C_1 \lambda^{-2} E X^2 I\{|X| \ge \lambda\} \le \sum_{n=1}^{\infty} P\{|S_n| \ge n\lambda\} \le C_2 \lambda^{-2} E X^2 I\{|X| \ge \lambda\}$$
 (3.3)

for any  $\lambda > 0$ , where  $C_1$  and  $C_2$  are positive absolute constants.

## 4 Proof of Theorem 2.1

Set  $b(\varepsilon) = [g^{-1}(\varepsilon^{-2})]$ , where  $g^{-1}(x)$  is the inverse function of g(x). Without loss of generality, we assume that  $\sigma^2 = 1$ .

#### Proposition 4.1

$$\lim_{\varepsilon\downarrow 0}\frac{1}{-\log\varepsilon}\sum_{n=n_0}^\infty\varphi(n)\int_{\varepsilon\sqrt{g(n)}}^\infty 2xP\{|N|\ge x\}\mathrm{d}x=2.$$

**Proof** At first, we discuss the relations between the integral and the series. If  $\varphi(y)$  is nonincreasing, then  $\varphi(y) \int_{\varepsilon\sqrt{g(y)}}^{\infty} 2x P\{|N| \ge x\} dx$  is nonincreasing. Hence we have

$$\int_{n_0+1}^{\infty} \varphi(y) \int_{\varepsilon\sqrt{g(y)}}^{\infty} 2x P\{|N| \ge x\} dx dy \le \sum_{n=n_0+1}^{\infty} \varphi(n) \int_{\varepsilon\sqrt{g(n)}}^{\infty} 2x P\{|N| \ge x\} dx dy \le \int_{n_0}^{\infty} \varphi(y) \int_{\varepsilon\sqrt{g(y)}}^{\infty} 2x P\{|N| \ge x\} dx dy.$$

Then we have

$$\lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} \sum_{n=n_0}^{\infty} \varphi(n) \int_{\varepsilon \sqrt{g(n)}}^{\infty} 2x P\{|N| \ge x\} dx$$

$$= \lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} \int_{n_0}^{\infty} \frac{g'(y)}{g(y)} \int_{\varepsilon \sqrt{g(y)}}^{\infty} 2x P\{|N| \ge x\} dx dy$$

$$= \lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} \int_{g(n_0)}^{\infty} \frac{1}{t} \int_{\varepsilon \sqrt{t}}^{\infty} 2x P\{|N| \ge x\} dx dt$$

$$= \lim_{\varepsilon \downarrow 0} \frac{2}{-\log \varepsilon} \int_{\varepsilon \sqrt{g(n_0)}}^{\infty} \frac{1}{y} \int_{y}^{\infty} 2x P\{|N| \ge x\} dx dy$$

$$= \lim_{\varepsilon \downarrow 0} 2 \int_{\varepsilon \sqrt{g(n_0)}}^{\infty} 2x P\{|N| \ge x\} dx$$

$$= 2. \tag{4.1}$$

If  $\varphi(y)$  is nondecreasing, then by  $\lim_{n\to\infty} \frac{\varphi(n+1)}{\varphi(n)} = 1$  for any  $0 < \delta < 1$ , there exists  $n_1 = n_1(\delta)$ , such that  $\frac{\varphi(n+1)}{\varphi(n)} < 1 + \delta$  and  $\frac{\varphi(n)}{\varphi(n+1)} > 1 - \delta$  for  $n \ge n_1$ . Thus we have

$$(1+\delta)^{-1} \int_{n_1+1}^{\infty} \varphi(y) \int_{\varepsilon\sqrt{g(y)}}^{\infty} 2x P\{|N| \ge x\} dx dy$$

$$\le \sum_{n=n_1+1}^{\infty} \varphi(n) \int_{\varepsilon\sqrt{g(n)}}^{\infty} 2x P\{|N| \ge x\} dx$$

$$\le (1-\delta)^{-1} \int_{n_1}^{\infty} \varphi(y) \int_{\varepsilon\sqrt{g(y)}}^{\infty} 2x P\{|N| \ge x\} dx dy.$$

Therefore

$$\lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} (1+\delta)^{-1} \int_{n_1+1}^{\infty} \varphi(y) \int_{\varepsilon \sqrt{g(y)}}^{\infty} 2x P\{|N| \ge x\} dx dy$$

$$\leq \lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} \sum_{n=n_0}^{\infty} \varphi(n) \int_{\varepsilon \sqrt{g(n)}}^{\infty} 2x P\{|N| \ge x\} dx$$

$$\leq \lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} (1-\delta)^{-1} \int_{n_1}^{\infty} \varphi(y) \int_{\varepsilon \sqrt{g(y)}}^{\infty} 2x P\{|N| \ge x\} dx dy.$$

And then by (4.1), we know

$$2(1+\delta)^{-1} \leq \lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} \sum_{n=n_0}^{\infty} \varphi(n) \int_{\varepsilon \sqrt{g(n)}}^{\infty} 2x P\{|N| \geq x\} \mathrm{d}x \leq 2(1-\delta)^{-1}.$$

Let  $\delta \downarrow 0$ . Then we can conclude

$$\lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} \sum_{n=n_0}^{\infty} \varphi(n) \int_{\varepsilon \sqrt{g(n)}}^{\infty} 2x P\{|N| \ge x\} \mathrm{d}x = 2.$$

**Remark 4.1** In the following, for simplicity, we will omit the discuss of  $\varphi(x)$ , but the process is similar to the discussion of Proposition 4.1.

#### Proposition 4.2

$$\lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} \sum_{n=n_0}^{b(\varepsilon)} \varphi(n) \Big| \int_{\varepsilon \sqrt{g(n)}}^{\infty} 2x P\{|S_n| \ge \sqrt{n}x\} dx - \int_{\varepsilon \sqrt{g(n)}}^{\infty} 2x P\{|N| \ge x\} dx \Big| = 0. \quad (4.2)$$

Proof Obviously,

$$\sum_{n=n_0}^{b(\varepsilon)} \varphi(n) \Big| \int_{\varepsilon\sqrt{g(n)}}^{\infty} 2x P\{|S_n| \ge \sqrt{n}x\} dx - \int_{\varepsilon\sqrt{g(n)}}^{\infty} 2x P\{|N| \ge x\} dx \Big|$$

$$= \sum_{n=n_0}^{b(\varepsilon)} \frac{g'(n)}{g(n)} \Big| \int_0^{\infty} 2(x + \varepsilon\sqrt{g(n)}) P\{|S_n| \ge \sqrt{n}(x + \varepsilon\sqrt{g(n)})\} dx$$

$$- \int_0^{\infty} 2(x + \varepsilon\sqrt{g(n)}) P\{|N| \ge (x + \varepsilon\sqrt{g(n)})\} dx \Big|$$

$$\leq \sum_{n=n_0}^{b(\varepsilon)} \frac{g'(n)}{g(n)} (\Delta_{n1} + \Delta_{n2} + \Delta_{n3}),$$

where

$$\Delta_{n} = \sup_{x} |P\{|S_{n}| \geq \sqrt{n}x\} - P\{|N| \geq x\}|,$$

$$\Delta_{n1} = \int_{0}^{\Delta_{n}^{-\frac{1}{4}}} 2(x + \varepsilon \sqrt{g(n)})|P\{|S_{n}| \geq \sqrt{n}(x + \varepsilon \sqrt{g(n)})\} - P\{|N| \geq (x + \varepsilon \sqrt{g(n)})\}|dx,$$

$$\Delta_{n2} = \int_{\Delta_{n}^{-\frac{1}{4}}}^{\infty} 2(x + \varepsilon \sqrt{g(n)})P\{|S_{n}| \geq \sqrt{n}(x + \varepsilon \sqrt{g(n)})\}dx,$$

$$\Delta_{n3} = \int_{\Delta_{n}^{-\frac{1}{4}}}^{\infty} 2(x + \varepsilon \sqrt{g(n)})P\{|N| \geq (x + \varepsilon \sqrt{g(n)})\}dx.$$

Since  $n < b(\varepsilon)$  implies  $\varepsilon \sqrt{q(n)} < 1$ , we have

$$\Delta_{n1} \le \int_0^{\Delta_n^{-\frac{1}{4}}} 2(x + \varepsilon \sqrt{g(n)}) \Delta_n dx \le \Delta_n (\Delta_n^{-\frac{1}{4}} + \varepsilon \sqrt{g(n)})^2 \le (\Delta_n^{\frac{1}{4}} + \Delta_n^{\frac{1}{2}})^2.$$
 (4.3)

For  $\Delta_{n3}$ , by Markov inequality, we have

$$\Delta_{n3} \le C \int_{\Delta_n^{-\frac{1}{4}}}^{\infty} \frac{1}{(x + \varepsilon \sqrt{g(n)})^3} \mathrm{d}x \le C \Delta_n^{\frac{1}{2}}.$$
(4.4)

Now we estimate  $\Delta_{n2}$ . By Lemma 3.2, we have

$$\Delta_{n2} \leq \int_{\Delta_n^{-\frac{1}{4}}}^{\infty} 2(x + \varepsilon \sqrt{g(n)}) nP \Big\{ |X| \geq \frac{\sqrt{n}}{2} (x + \varepsilon \sqrt{g(n)}) \Big\} dx$$
$$+ C \int_{\Delta_n^{-\frac{1}{4}}}^{\infty} 2(x + \varepsilon \sqrt{g(n)}) n^2 \frac{1}{n^2 (x + \varepsilon \sqrt{g(n)})^4} dx$$
$$\stackrel{\triangle}{=} I_1 + I_2.$$

By Fubini Theorem and the fact  $0 \le \Delta_n \le 1$ , we have

$$I_{1} = E \int_{\Delta_{n}^{-\frac{1}{4}}}^{\infty} 2(x + \varepsilon \sqrt{g(n)}) nI \Big\{ |X| \ge \frac{\sqrt{n}}{2} (x + \varepsilon \sqrt{g(n)}) \Big\} dx$$

$$\le EI \{4|X| \ge \sqrt{n}\} \int_{0}^{\infty} 2(x + \varepsilon \sqrt{g(n)}) nI \Big\{ |X| \ge \frac{\sqrt{n}}{2} (x + \varepsilon \sqrt{g(n)}) \Big\} dx$$

$$\le EI \{4|X| \ge \sqrt{n}\} \int_{0}^{\frac{2|X|}{\sqrt{n}} - \varepsilon \sqrt{g(n)}} 2n(x + \varepsilon \sqrt{g(n)}) dx$$

$$\le 4E|X|^{2} I \{4|X| \ge \sqrt{n}\}$$

$$(4.5)$$

and

$$I_2 \le C \int_{\Delta_n^{-\frac{1}{4}}}^{\infty} \frac{1}{(x + \varepsilon \sqrt{g(n)})^3} \mathrm{d}x \le C \Delta_n^{\frac{1}{2}}.$$
(4.6)

From (4.3)–(4.6) and the fact  $\Delta_n \to 0$  as  $n \to \infty$ , we can get

$$\Delta_{n1} + \Delta_{n2} + \Delta_{n3} \to 0$$
, as  $n \to \infty$ . (4.7)

Then by (4.7), the monotonicity of  $\varphi(x)$  and Toeplitz Lemma (see [4]), we get (4.2). The proposition is now proved.

#### Proposition 4.3

$$\lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} \sum_{n=h(\varepsilon)+1}^{\infty} \varphi(n) \int_{\varepsilon \sqrt{g(n)}}^{\infty} 2x P\{|N| \ge x\} dx = 0.$$
 (4.8)

**Proof** By the monotonicity of  $\varphi(x)$  and Markov inequality, we have

$$\lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} \sum_{n=b(\varepsilon)+1}^{\infty} \varphi(n) \int_{\varepsilon \sqrt{g(n)}}^{\infty} 2x P\{|N| \ge x\} dx$$

$$\leq \lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} \sum_{n=b(\varepsilon)+1}^{\infty} \varphi(n) \int_{\varepsilon \sqrt{g(n)}}^{\infty} 2x \frac{E|N|^4}{x^4} dx$$

$$\leq \lim_{\varepsilon \downarrow 0} \frac{C}{-\varepsilon^2 \log \varepsilon} \sum_{n=b(\varepsilon)+1}^{\infty} \frac{g'(n)}{g^2(n)}$$

$$\leq \lim_{\varepsilon \downarrow 0} \frac{C}{-\varepsilon^2 \log \varepsilon} \int_{-\varepsilon^2}^{\infty} \frac{1}{y^2} dy \leq \lim_{\varepsilon \downarrow 0} \frac{C}{-\log \varepsilon} = 0.$$

#### Proposition 4.4

$$\lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} \sum_{n=b(\varepsilon)+1}^{\infty} \varphi(n) \int_{\varepsilon\sqrt{g(n)}}^{\infty} 2x P\{|S_n| \ge \sqrt{n}x\} dx = 0.$$
 (4.9)

**Proof** Set

$$X'_{i} = X_{i}I\{|X_{i}| \le \lambda\}, \quad X''_{i} = X_{i}I\{|X_{i}| > \lambda\},$$
  
 $S'_{n} = \sum_{i=1}^{n} (X'_{i} - EX'_{i}), \quad S''_{n} = \sum_{i=1}^{n} (X''_{i} - EX''_{i}),$ 

where  $\lambda > 1$ . Clearly,  $S_n = S'_n + S''_n$ . Therefore,

$$\sum_{n=b(\varepsilon)+1}^{\infty} \varphi(n) \int_{\varepsilon\sqrt{g(n)}}^{\infty} 2x P\{|S_n| \ge \sqrt{n}x\} dx$$

$$\leq \sum_{n=b(\varepsilon)+1}^{\infty} \varphi(n) \int_{\varepsilon\sqrt{g(n)}}^{\infty} 2x P\{|S_n'| \ge \frac{\sqrt{n}x}{2}\} dx + \sum_{n=b(\varepsilon)+1}^{\infty} \varphi(n) \int_{\varepsilon\sqrt{g(n)}}^{\infty} 2x P\{|S_n''| \ge \frac{\sqrt{n}x}{2}\} dx$$

$$\stackrel{\triangle}{=} I_3 + I_4.$$

For I<sub>3</sub>, by the monotonicity of  $\varphi(x)$ , Markov inequality and  $E|S'_n|^4 \leq Cn^2\lambda^4$ , we have

$$\lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} I_{3} \leq \lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} \sum_{n=b(\varepsilon)+1}^{\infty} \varphi(n) \int_{\varepsilon \sqrt{g(n)}}^{\infty} 2x \frac{4E|S'_{n}|^{4}}{n^{2}x^{4}} dx$$

$$\leq \lim_{\varepsilon \downarrow 0} \frac{C}{-\varepsilon^{2} \log \varepsilon} \sum_{n=b(\varepsilon)+1}^{\infty} \frac{g'(n)}{g^{2}(n)}$$

$$\leq \lim_{\varepsilon \downarrow 0} \frac{C}{-\varepsilon^{2} \log \varepsilon} \int_{\varepsilon^{-2}}^{\infty} \frac{1}{y^{2}} dy \leq \lim_{\varepsilon \downarrow 0} \frac{C}{-\log \varepsilon} = 0. \tag{4.10}$$

For  $I_4$ , by condition (2.1), Fubini Theorem and Lemma 3.3, we get

$$I_{4} = \sum_{n=b(\varepsilon)+1}^{\infty} \varphi(n) \int_{\varepsilon\sqrt{g(n)}}^{\infty} 2xP\left\{|S_{n}''| \ge \frac{\sqrt{n}x}{2}\right\} dx$$

$$= \sum_{n=b(\varepsilon)+1}^{\infty} n\varphi(n) \int_{0}^{\infty} 2(t+\varepsilon)P\left\{|S_{n}''| \ge \frac{n(t+\varepsilon)}{2}\right\} dt$$

$$\le C \int_{0}^{\infty} (t+\varepsilon) \sum_{n=1}^{\infty} P\left\{|S_{n}''| \ge \frac{n(t+\varepsilon)}{2}\right\} dt$$

$$\le C \int_{0}^{\infty} (t+\varepsilon) \frac{E|Y|^{2}I\{|Y| \ge \frac{t+\varepsilon}{2}\}}{(t+\varepsilon)^{2}} dt$$

$$= CE\left(|Y|^{2} \int_{0}^{\infty} \frac{I\{|Y| \ge \frac{t+\varepsilon}{2}\}}{(t+\varepsilon)} dt\right)$$

$$\le CE|Y|^{2} \log^{+} 2|Y| - C(\log \varepsilon)E|Y|^{2},$$

where Y denotes  $X_1'' - EX_1''$ . Then by (2.2) we have

$$\lim_{\lambda \to \infty} \limsup_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} I_4 = 0. \tag{4.11}$$

From (4.10) and (4.11), we complete the proof of Proposition 4.4.

Proof of Theorem 2.1 Note that

$$\sum_{n=n_0}^{\infty} \varphi(n) E \left| \frac{S_n}{\sqrt{n}} \right|^2 I\{ |S_n| \ge \varepsilon \sqrt{ng(n)} \}$$

$$= \varepsilon^2 \sum_{n=n_0}^{\infty} \varphi(n) g(n) P\{ |S_n| \ge \varepsilon \sqrt{ng(n)} \} + \sum_{n=n_0}^{\infty} \varphi(n) \int_{\varepsilon \sqrt{g(n)}}^{\infty} 2x P\{ |S_n| \ge \sqrt{nx} \} dx.$$

In order to prove (2.3), we just need to show that

$$\lim_{\varepsilon \downarrow 0} \frac{\varepsilon^2}{-\log \varepsilon} \sum_{n=n_0}^{\infty} g'(n) P\{|S_n| \ge \varepsilon \sqrt{ng(n)}\} = 0$$
 (4.12)

and

$$\lim_{\varepsilon \downarrow 0} \frac{1}{-\log \varepsilon} \sum_{n=n_0}^{\infty} \varphi(n) \int_{\varepsilon \sqrt{g(n)}}^{\infty} 2x P\{|S_n| \ge \sqrt{n}x\} dx = 2.$$
 (4.13)

By the same argument of Lemma 3.1, we know that for any  $\frac{1}{s} > p$ ,

$$\lim_{\varepsilon \downarrow 0} \varepsilon^{\frac{1}{s}} \sum_{n=n_0}^{\infty} g'(n) P\{|S_n| \ge \varepsilon \sqrt{n} g^s(n)\} = E|N|^{\frac{1}{s}}. \tag{4.14}$$

Let  $s = \frac{1}{2}$  in (4.14). Then we can obtain (4.12). (4.13) can be proved by Propositions 4.1–4.4 and the triangular inequality.

#### 5 Proof of Theorem 2.2

Set  $d(\varepsilon) = [g^{-1}(M\varepsilon^{-\frac{1}{s}})]$ , where  $g^{-1}(x)$  is the inverse function of g(x),  $M \ge 1$ . Without loss of generality, we assume that  $\sigma^2 = 1$ .

**Proposition 5.1** For p > 0, we have

$$\lim_{\varepsilon \downarrow 0} \varepsilon^{\frac{1}{s} - p} \sum_{n = n_0}^{\infty} \varphi(n) \int_{\varepsilon g^s(n)}^{\infty} p x^{p-1} P\{|N| \ge x\} \mathrm{d}x = \frac{ps}{1 - ps} E|N|^{\frac{1}{s}}.$$
 (5.1)

**Proof** At first we discuss the relations between the integral and the series. If  $\varphi(y)$  is nonincreasing, then  $\varphi(y) \int_{\varepsilon g^s(y)}^{\infty} px^{p-1} P\{|N| \ge x\} dx$  is nonincreasing. Hence we have

$$\int_{n_0+1}^{\infty} \varphi(y) \int_{\varepsilon g^s(y)}^{\infty} px^{p-1} P\{|N| \ge x\} dx dy \le \sum_{n=n_0+1}^{\infty} \varphi(n) \int_{\varepsilon g^s(n)}^{\infty} px^{p-1} P\{|N| \ge x\} dx dy.$$

$$\le \int_{n_0}^{\infty} \varphi(y) \int_{\varepsilon g^s(y)}^{\infty} px^{p-1} P\{|N| \ge x\} dx dy.$$

Then we have

$$\lim_{\varepsilon \downarrow 0} \varepsilon^{\frac{1}{s} - p} \sum_{n = n_0}^{\infty} \varphi(n) \int_{\varepsilon g^s(n)}^{\infty} px^{p-1} P\{|N| \ge x\} dx$$

$$= \lim_{\varepsilon \downarrow 0} \varepsilon^{\frac{1}{s} - p} \int_{n_0}^{\infty} \varphi(y) \int_{\varepsilon g^s(y)}^{\infty} px^{p-1} P\{|N| \ge x\} dx dy$$

$$= \lim_{\varepsilon \downarrow 0} \varepsilon^{\frac{1}{s} - p} \int_{g(n_0)}^{\infty} \frac{1}{t^{ps}} \int_{\varepsilon t^s}^{\infty} px^{p-1} P\{|N| \ge x\} dx dt$$

$$\begin{split} &= \lim_{\varepsilon \downarrow 0} \frac{p}{s} \int_{\varepsilon g^{s}(n_{0})}^{\infty} \frac{1}{t^{p+1-\frac{1}{s}}} \int_{t}^{\infty} x^{p-1} P\{|N| \geq x\} \mathrm{d}x \mathrm{d}t \\ &= \frac{p}{s} \int_{0}^{\infty} \frac{1}{t^{p+1-\frac{1}{s}}} \int_{t}^{\infty} x^{p-1} P\{|N| \geq x\} \mathrm{d}x \mathrm{d}t \\ &= \frac{p}{s} \int_{0}^{\infty} x^{p-1} P\{|N| \geq x\} \int_{0}^{x} \frac{1}{t^{p+1-\frac{1}{s}}} \mathrm{d}t \mathrm{d}x \\ &= \frac{ps}{1-ps} \int_{0}^{\infty} \frac{1}{s} x^{\frac{1}{s}-1} P\{|N| \geq x\} \mathrm{d}x \\ &= \frac{ps}{1-ps} E|X|^{\frac{1}{s}}. \end{split}$$

If  $\varphi(y)$  is nondecreasing, then by  $\lim_{n\to\infty} \frac{\varphi(n+1)}{\varphi(n)} = 1$ , the proof is similar to that of Proposition 4.1. Thus we can get Proposition 5.1 by the above steps.

**Remark 5.1** In the following, for simplicity, we will omit the discussion of  $\varphi(x)$ , but the process is similar to the discussion of Proposition 5.1.

**Proposition 5.2** For p > 0, we have

$$\lim_{\varepsilon \downarrow 0} \varepsilon^{\frac{1}{s} - p} \sum_{n=n_0}^{d(\varepsilon)} \varphi(n) \Big| \int_{\varepsilon g^s(n)}^{\infty} px^{p-1} P\{|S_n| \ge \sqrt{n}x\} dx - \int_{\varepsilon g^s(n)}^{\infty} px^{p-1} P\{|N| \ge x\} dx \Big| = 0. \quad (5.2)$$

**Proof** It is easy to see that

$$\varepsilon^{\frac{1}{s}-p} \sum_{n=n_0}^{d(\varepsilon)} \varphi(n) \Big| \int_{\varepsilon g^s(n)}^{\infty} p x^{p-1} P\{|S_n| \ge \sqrt{n}x\} dx - \int_{\varepsilon g^s(n)}^{\infty} p x^{p-1} P\{|N| \ge x\} dx \Big|$$

$$\leq \sum_{n=n_0}^{d(\varepsilon)} \varphi(n) \int_0^{\infty} p (x + \varepsilon g^s(n))^{p-1} |P\{|S_n| \ge \sqrt{n} (x + \varepsilon g^s(n))\} - P\{|N| \ge (x + \varepsilon g^s(n))\} |dx$$

$$\leq \sum_{n=n_0}^{d(\varepsilon)} \varphi(n) (\Delta'_{n1} + \Delta'_{n2} + \Delta'_{n3}),$$

where

$$\Delta'_{n1} = \int_{0}^{\Delta_{n}^{-\frac{1}{2p}}} p(x + \varepsilon g^{s}(n))^{p-1} |P\{|S_{n}| \ge \sqrt{n}(x + \varepsilon g^{s}(n))\} - P\{|N| \ge (x + \varepsilon g^{s}(n))\} |dx,$$

$$\Delta'_{n2} = \int_{\Delta_{n}^{-\frac{1}{2p}}}^{\infty} p(x + \varepsilon g^{s}(n))^{p-1} P\{|S_{n}| \ge \sqrt{n}(x + \varepsilon g^{s}(n))\} dx,$$

$$\Delta'_{n3} = \int_{\Delta_{n}^{-\frac{1}{2p}}}^{\infty} p(x + \varepsilon g^{s}(n))^{p-1} P\{|N| \ge (x + \varepsilon g^{s}(n))\} dx.$$

Since  $n \leq d(\varepsilon)$  implies  $\varepsilon g^s(n) \leq M^s$ , we have

$$\Delta'_{n1} \le \int_{0}^{\Delta_{n}^{-\frac{1}{2p}}} p(x + \varepsilon g^{s}(n))^{p-1} \Delta_{n} dx \le \Delta_{n} (\Delta_{n}^{-\frac{1}{2p}} + \varepsilon g^{s}(n))^{p} \le (\Delta_{n}^{\frac{1}{2p}} + M^{s} \Delta_{n}^{\frac{1}{p}})^{p}.$$
 (5.3)

For  $\Delta'_{n3}$ , by Markov inequality, we have

$$\Delta'_{n3} \le C \int_{\Delta_n^{-\frac{1}{2p}}}^{\infty} \frac{1}{(x + \varepsilon g^s(n))^3} \mathrm{d}x \le C \Delta_n^{\frac{1}{p}}.$$
 (5.4)

Now we estimate  $\Delta'_{n2}$ . First we consider 0 . By Markov inequality, we have

$$\Delta'_{n2} \le C \int_{\Delta_n^{-\frac{1}{2p}}}^{\infty} \frac{1}{(x + \varepsilon g^s(n))^{3-p}} dx \le C \Delta_n^{\frac{1}{p} - \frac{1}{2}}.$$
 (5.5)

For  $p \geq 2$ , by Lemma 3.2, choosing  $\beta = 2$  and x = py, we know

$$\Delta'_{n2} \leq \int_{\Delta_n^{-\frac{1}{2p}}}^{\infty} p(x + \varepsilon g^s(n))^{p-1} nP \Big\{ |X| \geq \frac{\sqrt{n}}{p} (x + \varepsilon g^s(n)) \Big\} dx$$
$$+ C \int_{\Delta_n^{-\frac{1}{2p}}}^{\infty} p(x + \varepsilon g^s(n))^{p-1} n^p \frac{1}{n^p (x + \varepsilon g^s(n))^{2p}} dx$$
$$\stackrel{\triangle}{=} J_1 + J_2.$$

By Fubini Theorem and the fact  $0 \le \Delta_n \le 1$ , we have

$$J_{1} = E \int_{\Delta_{n}^{-\frac{1}{2p}}}^{\infty} p(x + \varepsilon g^{s}(n))^{p-1} nI \Big\{ |X| \ge \frac{\sqrt{n}}{p} (x + \varepsilon g^{s}(n)) \Big\} dx$$

$$\le EI \{ 2p|X| \ge \sqrt{n} \} \int_{0}^{\infty} p(x + \varepsilon g^{s}(n))^{p-1} nI \Big\{ |X| \ge \frac{\sqrt{n}}{p} (x + \varepsilon g^{s}(n)) \Big\} dx$$

$$\le EI \{ 2p|X| \ge \sqrt{n} \} \int_{0}^{\frac{p|X|}{\sqrt{n}} - \varepsilon g^{s}(n)} p(x + \varepsilon g^{s}(n))^{p-1} ndx$$

$$\le CEI \{ 2p|X| \ge \sqrt{n} \} n \frac{|X|^{p}}{n^{\frac{p}{2}}}$$

$$\le CE|X|^{p} I \{ 2p|X| \ge \sqrt{n} \}$$

$$(5.6)$$

and

$$J_2 \le C \int_{\Delta_n^{-\frac{1}{2p}}}^{\infty} \frac{1}{(x + \varepsilon g^s(n))^{p+1}} dx \le C \Delta_n^{\frac{1}{2}}.$$
 (5.7)

From (5.3)–(5.7) and the fact  $\Delta_n \to 0$  as  $n \to \infty$ , we can get

$$\Delta'_{n1} + \Delta'_{n2} + \Delta'_{n3} \to 0 \quad \text{as } n \to \infty.$$
 (5.8)

Then by (5.8), the monotonicity of  $\varphi(x)$  and Toeplitz Lemma (see [4]), we get (5.2). The proposition is now proved.

**Proposition 5.3** For p > 0, we have

$$\lim_{M \to \infty} \limsup_{\varepsilon \downarrow 0} \varepsilon^{\frac{1}{s} - p} \sum_{n = d(\varepsilon) + 1}^{\infty} \varphi(n) \int_{\varepsilon g^{s}(n)}^{\infty} px^{p-1} P\{|N| \ge x\} \mathrm{d}x = 0. \tag{5.9}$$

**Proof** By the monotonicity of  $\varphi(x)$  and Markov inequality, we have

$$\lim_{M \to \infty} \limsup_{\varepsilon \downarrow 0} \varepsilon^{\frac{1}{s} - p} \sum_{n = d(\varepsilon) + 1}^{\infty} \varphi(n) \int_{\varepsilon g^{s}(n)}^{\infty} px^{p - 1} P\{|N| \ge x\} \mathrm{d}x$$

$$\leq \lim_{M \to \infty} \limsup_{\varepsilon \downarrow 0} C \varepsilon^{\frac{1}{s} - p} \sum_{n = d(\varepsilon) + 1}^{\infty} \frac{g'(n)}{g^{ps}(n)} \int_{\varepsilon g^{s}(n)}^{\infty} px^{p - 1} \frac{E|N|^{\frac{2}{s}}}{x^{\frac{2}{s}}} \mathrm{d}x$$

$$\leq \lim_{M \to \infty} \limsup_{\varepsilon \downarrow 0} C \varepsilon^{-\frac{1}{s}} \sum_{n=d(\varepsilon)+1}^{\infty} \frac{g'(n)}{g^2(n)}$$

$$\leq \lim_{M \to \infty} \limsup_{\varepsilon \downarrow 0} C \varepsilon^{-\frac{1}{s}} \int_{M \varepsilon^{-\frac{1}{s}}}^{\infty} \frac{1}{y^2} \mathrm{d}y$$

$$\leq \lim_{M \to \infty} \limsup_{\varepsilon \downarrow 0} C \frac{1}{M} = 0.$$

**Proposition 5.4** For p > 0, we have

$$\lim_{M \to \infty} \lim_{\varepsilon \downarrow 0} \varepsilon^{\frac{1}{s} - p} \sum_{n = d(\varepsilon) + 1}^{\infty} \varphi(n) \int_{\varepsilon g^{s}(n)}^{\infty} px^{p-1} P\{|S_{n}| \ge \sqrt{n}x\} dx = 0.$$
 (5.10)

**Proof** Define  $X_i', X_i'', S_n'$  and  $S_n''$  as in the proof of Proposition 4.4. Therefore

$$\sum_{n=d(\varepsilon)+1}^{\infty} \varphi(n) \int_{\varepsilon g^{s}(n)}^{\infty} px^{p-1} P\{|S_{n}| \geq \sqrt{n}x\} dx$$

$$\leq \sum_{n=d(\varepsilon)+1}^{\infty} \varphi(n) \Big[ \int_{\varepsilon g^{s}(n)}^{\infty} px^{p-1} P\{|S'_{n}| \geq \frac{\sqrt{n}x}{2}\} dx + \int_{\varepsilon g^{s}(n)}^{\infty} px^{p-1} P\{|S''_{n}| \geq \frac{\sqrt{n}x}{2}\} dx \Big]$$

$$\stackrel{\triangle}{=} J_{3} + J_{4}.$$

For  $J_3$ , by the monotonicity of  $\varphi(x)$ , Markov inequality and  $E|S'_n|^{\frac{2}{s}} \leq Cn^{\frac{1}{s}}\lambda^{\frac{2}{s}}$ , we have

$$\lim_{M \to \infty} \limsup_{\varepsilon \downarrow 0} \varepsilon^{\frac{1}{s} - p} J_{3} \leq \lim_{M \to \infty} \limsup_{\varepsilon \downarrow 0} C \varepsilon^{\frac{1}{s} - p} \sum_{n = d(\varepsilon) + 1}^{\infty} \varphi(n) \int_{\varepsilon g^{s}(n)}^{\infty} p x^{p - 1} \frac{E|S'_{n}|^{\frac{2}{s}}}{n^{\frac{1}{s}} x^{\frac{2}{s}}} dx$$

$$\leq \lim_{M \to \infty} \limsup_{\varepsilon \downarrow 0} C \varepsilon^{\frac{1}{s} - p} \sum_{n = d(\varepsilon) + 1}^{\infty} \varphi(n) \int_{\varepsilon g^{s}(n)}^{\infty} \frac{1}{x^{\frac{2}{s} - p + 1}} dx$$

$$\leq \lim_{M \to \infty} \limsup_{\varepsilon \downarrow 0} C \varepsilon^{-\frac{1}{s}} \sum_{n = d(\varepsilon) + 1}^{\infty} \frac{g'(n)}{g^{2}(n)}$$

$$\leq \lim_{M \to \infty} \limsup_{\varepsilon \downarrow 0} C \varepsilon^{-\frac{1}{s}} \int_{M \varepsilon^{-\frac{1}{s}}}^{\infty} \frac{1}{y^{2}} dy$$

$$\leq \lim_{M \to \infty} \limsup_{\varepsilon \downarrow 0} C \varepsilon^{-\frac{1}{s}} \int_{M \varepsilon^{-\frac{1}{s}}}^{\infty} \frac{1}{y^{2}} dy$$

$$\leq \lim_{M \to \infty} \limsup_{\varepsilon \downarrow 0} C \varepsilon^{-\frac{1}{s}} \int_{M \varepsilon^{-\frac{1}{s}}}^{\infty} \frac{1}{y^{2}} dy$$

$$\leq \lim_{M \to \infty} \limsup_{\varepsilon \downarrow 0} C \varepsilon^{-\frac{1}{s}} \int_{M \varepsilon^{-\frac{1}{s}}}^{\infty} \frac{1}{y^{2}} dy$$

For  $J_4$ , by condition (2.4), Fubini Theorem and Lemma 3.3, we get

$$\begin{aligned} \mathbf{J}_4 &= \sum_{n=d(\varepsilon)+1}^{\infty} \varphi(n) \int_{\varepsilon g^s(n)}^{\infty} p x^{p-1} P\Big\{ |S_n''| \geq \frac{\sqrt{n}x}{2} \Big\} \mathrm{d}x \\ &= \sum_{n=d(\varepsilon)+1}^{\infty} n^{\frac{p}{2}} \varphi(n) \int_0^{\infty} p (x+\varepsilon)^{p-1} P\Big\{ |S_n''| \geq \frac{n(x+\varepsilon)}{2} \Big\} \mathrm{d}x \\ &\leq C \int_0^{\infty} p (x+\varepsilon)^{p-1} \sum_{n=1}^{\infty} P\Big\{ |S_n''| \geq \frac{n(x+\varepsilon)}{2} \Big\} \mathrm{d}x \end{aligned}$$

$$\leq C \int_0^\infty p(x+\varepsilon)^{p-1} \frac{E|Y|^2 I\{|Y| \geq \frac{x+\varepsilon}{2}\}}{(x+\varepsilon)^2} dx$$
$$= CpEY^2 \int_0^\infty \frac{I\{|Y| \geq \frac{x+\varepsilon}{2}\}}{(x+\varepsilon)^{3-p}} dx,$$

where Y denotes  $X_1'' - EX_1''$ .

Note that

$$J_4 < CEY^2 \log^+ |Y| - C(\log \varepsilon)EY^2$$

for p=2, and

$$J_4 \le CE|Y|^P - C(\varepsilon^{p-2})EY^2$$

for  $p \neq 2$ , and note that  $\frac{1}{s} > p$ , so we know

$$\lim_{\lambda \to \infty} \limsup_{\varepsilon \downarrow 0} \varepsilon^{\frac{1}{s} - p} J_4 = 0.$$

From the above steps, we complete the proof.

Proof of Theorem 2.2 Since

$$E\left|\frac{S_n}{\sqrt{n}}\right|^p I\{|S_n| \ge \varepsilon \sqrt{n} \, g^s(n)\} = P\{|S_n| \ge \varepsilon \sqrt{n} \, g^s(n)\},$$

where p=0, by (4.14) we can get Theorem 2.2. Therefore we just need to discuss the case  $\frac{1}{s} > p > 0$ . Note that

$$\sum_{n=n_0}^{\infty} \varphi(n) E \left| \frac{S_n}{\sqrt{n}} \right|^p I\{|S_n| \ge \varepsilon \sqrt{n} g^s(n)\}$$

$$= \varepsilon^p \sum_{n=n_0}^{\infty} \varphi(n) g^{ps}(n) P\{|S_n| \ge \varepsilon \sqrt{n} g^s(n)\} + \sum_{n=n_0}^{\infty} \varphi(n) \int_{\varepsilon g^s(n)}^{\infty} p x^{p-1} P\{|S_n| \ge \sqrt{n} x\} dx.$$

In order to prove (2.8), we just need to show that

$$\lim_{\varepsilon \downarrow 0} \varepsilon^{\frac{1}{s}} \sum_{n=n_0}^{\infty} g'(n) P\{|S_n| \ge \varepsilon \sqrt{n} g^s(n)\} = E|N|^{\frac{1}{s}}, \tag{5.11}$$

$$\lim_{\varepsilon \downarrow 0} \varepsilon^{\frac{1}{s} - p} \sum_{n=n_0}^{\infty} \varphi(n) \int_{\varepsilon g^s(n)}^{\infty} px^{p-1} P\{|S_n| \ge \sqrt{n}x\} dx = \frac{ps}{1 - ps} E|N|^{\frac{1}{s}}.$$
 (5.12)

By (4.14), we can get (5.11). (5.12) can be proved by Propositions 5.1-5.4 and the triangular inequality.

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## References

 Cheng, F. Y. and Wang, Y. B., Precise asymptotics of partial sums for IID and NA sequences (in Chinese), Acta. Math. Sin., Ser. A, 45(5), 2004, 965–972.

- [2] Gut, A. and Spătaru, A., Precise asymptotics in the Baum-Katz and Davis law of large numbers, J. Math. Anal. Appl., 248, 2000, 233–246.
- [3] Gut, A. and Spătaru, A., Precise asymptotics in the law of the iterated logarithm, Ann. Prob., 28, 2000, 1870–1883.
- [4] Hall, P. and Heyde, C. C., Martingale Limit Theory and Its Application, Academic Press, New York, 1980.
- [5] Heyde, C. C., A supplement to the strong law of large numbers, J. Appl. Prob., 12, 1975, 173–175.
- [6] Hsu, P. L. and Robbins, H., Complete convergence and the strong law of large numbers, Proc. Nat. Acad. Sci. USA, 33, 1947, 25–31.
- [7] Liu, W. D. and Lin, Z. Y., Precise asymptotics for a new kind of complete moment convergence, Stat. Prob. Lett, 76, 2006, 1787–1799.
- [8] Nagaev, S. V., Large deviations of sums of independent random variables, Ann. Prob., 7(5), 1979, 745–789.
- [9] Pang, T. X. and Wang, J. F., Precise asymptotics in Chung's law of the iterated logarithm via self-normalization (in Chinese), Chin. Ann. Math., 28A(4), 2007, 507–518.
- [10] Pruss, A. R., A two-sided estimate in the Hsu-Robbins-Erdos law of large numbers, Stoch. Proc. Appl., 70, 1997, 173–180.
- $[11] \ \ \text{Spătaru, A., Precise asymptotics in Spitzer's law of large numbers, \textit{J. Theor. Prob., } \textbf{12}, 1999, 811-819.$