

RESEARCH

Open Access

# A note to the convergence rates in precise asymptotics

Jianjun He\*

\*Correspondence: hejj@cjl.u.edu.cn  
Department of Mathematics, China  
Jiliang University, Hangzhou,  
310018, China

## Abstract

Let  $\{X, X_n, n \geq 1\}$  be a sequence of i.i.d. random variables with zero mean. Set  $S_n = \sum_{k=1}^n X_k$ ,  $EX^2 = \sigma^2 > 0$ , and  $\lambda_{r,p}(\epsilon) = \sum_{n=1}^{\infty} n^{r/p-2} P(|S_n| \geq n^{1/p}\epsilon)$ . In this paper, the author discusses the rate of approximation of  $\frac{p}{r-p} E|N|^{2(r-p)/(2-p)}$  by  $\epsilon^{2(r-p)/(2-p)} \lambda_{r,p}(\epsilon)$  under suitable moment conditions, where  $N$  is normal with zero mean and variance  $\sigma^2 > 0$ , which improves the results of Gut and Steinebach (J. Math. Anal. Appl. 390:1-14, 2012) and extends the work He and Xie (Acta Math. Appl. Sin. 29:179-186, 2013). Specially, for the case  $r = 2$  and  $p = \frac{1}{\beta+1}$ ,  $\beta > -\frac{1}{2}$ , the author discusses the rate of approximation of  $\frac{\sigma^2}{2\beta+1}$  by  $\epsilon^2 \lambda_{2,1/(\beta+1)}(\epsilon)$  under the condition  $EX^2 I(|X| > t) = O(t^{-\delta} l(t))$  for some  $\delta > 0$ , where  $l(t)$  is a slowly varying function at infinity.

**MSC:** 60F15; 60G50

**Keywords:** convergence rate; precise asymptotics; slowly varying function

## 1 Introduction

Let  $\{X, X_n, n \geq 1\}$  be a sequence of i.i.d. random variables. Set  $S_n = \sum_{k=1}^n X_k$  and  $\lambda_{r,p}(\epsilon) = \sum_{n=1}^{\infty} n^{r/p-2} P(|S_n| \geq n^{1/p}\epsilon)$ . Heyde [1] proved that

$$\lim_{\epsilon \rightarrow 0} \epsilon^2 \lambda_{2,1}(\epsilon) = \sigma^2,$$

whenever  $EX = 0$  and  $EX^2 = \sigma^2 < \infty$ . Klesov [2] studied the rate of the approximation of  $\sigma^2$  by  $\epsilon^2 \lambda_{2,1}(\epsilon)$  under the condition  $E|X|^3 < \infty$ . He and Xie [3] improved the results of Klesov [2]. Gut and Steinebach [4] extended the results of Klesov [2] and obtained the following Theorem A. Gut and Steinebach [5] studied the general idea of proving precise asymptotics.

**Theorem A** *Let  $\{X, X_n, n \geq 1\}$  be a sequence of i.i.d. random variables with zero mean and  $0 < p < 2$ ,  $r \geq 2$ .*

(1) *If  $EX^2 = \sigma^2 > 0$  and  $E|X|^q < \infty$  for some  $r < q \leq 3$ , then*

$$\epsilon^{2(r-p)/(2-p)} \lambda_{r,p}(\epsilon) - \frac{p}{r-p} E|N|^{2(r-p)/(2-p)} = o\left(\epsilon^{\frac{p(q-2)(r-p)}{(q-p)(2-p)}}\right).$$

(2) *If  $EX^2 = \sigma^2 > 0$  and  $E|X|^q < \infty$  for some  $q \geq 3$  with  $q > \frac{2r-3p}{2-p}$ , then*

$$\epsilon^{2(r-p)/(2-p)} \lambda_{r,p}(\epsilon) - \frac{p}{r-p} E|N|^{2(r-p)/(2-p)} = o\left(\epsilon^{\frac{2p(r-p)}{(2-p)(p+2q-pq)}}\right),$$

where  $N$  is normal with mean 0 and variance  $\sigma^2 > 0$ .

The purpose of this paper is to strengthen Theorem A and extend the theorem of He and Xie [3] under suitable moment conditions. In addition, we shall discuss the rate at which  $\epsilon^{2\lambda_{2,1}/(\beta+1)}(\epsilon)$  converges to  $\frac{\sigma^2}{2\beta+1}$  under the condition  $T(t) = O(t^{-\delta}l(t))$  for some  $\delta > 0$ , where  $T(t) = EX^2I(|X| > t)$ ,  $l(t)$  is a slowly varying function at infinity. Throughout this paper,  $C$  represents a positive constant, though its value may change from one appearance to the next, and  $[x]$  denotes the integer part of  $x$ .  $\Phi(x)$  is the standard normal distribution function,  $\Phi(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^x e^{-t^2/2} dt$ ,  $\varphi(x) = \Phi'(x)$ .

## 2 Main results

From Gut and Steinebach [6], it is easy to obtain the following lemma.

**Lemma 2.1** *Let  $\{X, X_n, n \geq 1\}$  be a sequence of i.i.d. normal distribution random variables with zero mean and variance  $\sigma^2 > 0$ . Set  $0 < p < 2$  and  $r \geq 2$ , then*

$$\begin{aligned} & \epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{\infty} n^{(r/p)-2} P(|S_n| \geq n^{1/p}\epsilon) - \frac{p}{r-p} E|N|^{2(r-p)/(2-p)} \\ &= \begin{cases} O(\epsilon^{2(r-p)/(2-p)}), & r < 3p, \\ O(\epsilon^{4p/(2-p)}), & r \geq 3p. \end{cases} \end{aligned} \tag{2.1}$$

**Lemma 2.2** (Bingham *et al.* [7]) *Let  $l(t)$  be a slowly varying function. We have*

(1) *for any  $\eta > 0$ ,*

$$\lim_{t \rightarrow \infty} t^\eta l(t) = \infty, \quad \lim_{t \rightarrow \infty} t^{-\eta} l(t) = 0;$$

(2) *if  $0 < \delta < 1$ , then*

$$\int_a^t s^{-\delta} l(s) ds \sim \frac{1}{1-\delta} t^{1-\delta} l(t), \quad t \rightarrow \infty;$$

(3) *if  $\delta > 1$ , then*

$$\int_t^\infty s^{-\delta} l(s) ds \sim -\frac{1}{1-\delta} t^{1-\delta} l(t), \quad t \rightarrow \infty;$$

(4) *if  $\delta = 1$ , then  $L(t) = \int_t^\infty \frac{l(s)}{s} ds$ ,  $m(t) = \int_a^t \frac{l(s)}{s} ds$  are slowly varying functions; and*

$$\lim_{t \rightarrow \infty} \frac{l(t)}{L(t)} = 0, \quad \lim_{t \rightarrow \infty} \frac{l(t)}{m(t)} = 0.$$

**Theorem 2.1** *Let  $\{X, X_n, n \geq 1\}$  be a sequence of i.i.d. random variables with zero mean and  $0 < p < 2$ ,  $r \geq 2$ .*

(1) *If  $EX^2 = \sigma^2 > 0$  and  $E|X|^3 < \infty$  for some  $r < 3$ , then*

$$\epsilon^{2(r-p)/(2-p)} \lambda_{r,p}(\epsilon) - \frac{p}{r-p} E|N|^{2(r-p)/(2-p)} = \begin{cases} O(\epsilon^{2(r-p)/(2-p)}), & 2 \leq r < \frac{3p}{2}, \\ O(\epsilon^{p/(2-p)} \log \frac{1}{\epsilon}), & r = \frac{3p}{2}, \\ O(\epsilon^{p/(2-p)}), & \frac{3p}{2} < r < 3. \end{cases} \tag{2.2}$$

(2) If  $EX^2 = \sigma^2 > 0$  and  $E|X|^{2+\delta} < \infty$  for some  $0 < \delta < 1$ ,  $r < 2 + \delta$ , then

$$\begin{aligned} & \epsilon^{2(r-p)/(2-p)} \lambda_{r,p}(\epsilon) - \frac{P}{r-p} E|N|^{2(r-p)/(2-p)} \\ &= \begin{cases} O(\epsilon^{2(r-p)/(2-p)}), & 2 \leq r < (1 + \delta/2)p, \\ O(\epsilon^{p\delta/(2-p)} \log \frac{1}{\epsilon}), & r = (1 + \delta/2)p, \\ O(\epsilon^{p\delta/(2-p)}), & (1 + \delta/2)p < r < 2 + \delta. \end{cases} \end{aligned} \tag{2.3}$$

(3) If  $EX^2 = \sigma^2 > 0$  and  $E|X|^q < \infty$  for some  $q \geq 3$  with  $q > \frac{2r-3p}{2-p}$ , then

$$\epsilon^{2(r-p)/(2-p)} \lambda_{r,p}(\epsilon) - \frac{P}{r-p} E|N|^{2(r-p)/(2-p)} = \begin{cases} O(\epsilon^{2(r-p)/(2-p)}), & 2 \leq r < \frac{3p}{2}, \\ O(\epsilon^{p/(2-p)} \log \frac{1}{\epsilon}), & r = 3p/2, \\ O(\epsilon^{p/(2-p)}), & r > 3p/2, \end{cases} \tag{2.4}$$

where  $N$  is normal with mean 0 and variance  $\sigma^2 > 0$ .

**Remark 2.1** Clearly, Theorem 1 and Theorem 2 in He and Xie [3] are special cases of Theorem 2.1, by taking  $r = 2$  and  $p = 1$ .

**Remark 2.2** If  $0 < p < 2$ ,  $r \geq 2$ , we have  $\min(\frac{2(r-p)}{2-p}, \frac{p\delta}{2-p}) > \frac{p\delta(r-p)}{(2+\delta-p)(2-p)}$  for  $r < 2 + \delta = q \leq 3$  and  $\min(\frac{2(r-p)}{2-p}, \frac{p}{2-p}) > \frac{2(r-p)p}{(2-p)(p+2q-pq)}$  for some  $q \geq 3$  with  $q > \frac{2r-3p}{2-p}$ . So, the results of Theorem 2.1 are stronger than those of Theorem A.

**Theorem 2.2** Let  $\{X, X_n; n \geq 1\}$  be a sequence of i.i.d random variables with zero mean, and let  $T(t) = O(t^{-\delta}l(t))$  for some  $\delta > 0$ , where  $l(t)$  is a slowly varying function at infinity. Set  $EX^2 = \sigma^2 > 0$  and  $\beta > -\frac{1}{2}$ .

(1) If  $\delta > 1$ , then

$$\epsilon^2 \lambda_{2,1/(\beta+1)}(\epsilon) - \frac{\sigma^2}{2\beta+1} = \begin{cases} O(\epsilon^2), & -\frac{1}{2} < \beta < -\frac{1}{4}, \\ O(\epsilon^2 \log \frac{1}{\epsilon}), & \beta = -\frac{1}{4}, \\ O(\epsilon^{1/(2\beta+1)}), & \beta > -\frac{1}{4}. \end{cases} \tag{2.5}$$

(2) If  $0 < \delta < 1$ , then

$$\epsilon^2 \lambda_{2,1/(\beta+1)}(\epsilon) - \frac{\sigma^2}{2\beta+1} = \begin{cases} O(\epsilon^2), & -\frac{1}{2} < \beta < -\frac{1}{2} + \frac{\delta}{4}, \\ O(\epsilon^{\delta/(2\beta+1)} l(\epsilon^{-1/(2\beta+1)})), & \beta \geq -\frac{1}{2} + \frac{\delta}{4}. \end{cases} \tag{2.6}$$

(3) If  $\delta = 1$ , then

$$\epsilon^2 \lambda_{2,1/(\beta+1)}(\epsilon) - \frac{\sigma^2}{2\beta+1} = \begin{cases} O(\epsilon^2 + \epsilon^2 \int_1^{\epsilon^{-1/(2\beta+1)}} \frac{l(t)}{t} dt), & -\frac{1}{2} < \beta < -\frac{1}{4}, \\ O(\epsilon^2 (1 + \int_1^{\epsilon^{-5}} \frac{l(t)}{t} dt) \log \frac{1}{\epsilon}), & \beta = -\frac{1}{4}, \\ O(\epsilon^{1/(2\beta+1)} (1 + \int_1^{\epsilon^{-(2\beta+3)/(2\beta+1)}} \frac{l(t)}{t} dt)), & \beta > -\frac{1}{4}. \end{cases} \tag{2.7}$$

**Remark 2.3** For  $r = 2$ ,  $p = \frac{1}{\beta+1}$ . If  $l(t) = 1$ , then the result of Theorem 2.2 is weaker than that of Theorem 2.1 for  $0 < \delta < 1$ ,  $\beta \geq -\frac{1}{2} + \frac{\delta}{4}$ , and weaker than that of Theorem 2.1 for

$\delta = 1$ . But the condition  $T(t) = O(t^{-\delta})$  is weaker than the condition  $E|X|^{2+\delta} < \infty$ . If  $l(t) \rightarrow 0$  as  $t \rightarrow \infty$ , then the result of Theorem 2.2 is the same as that of Theorem 2.1 for  $0 < \delta < 1$ .

**Remark 2.4** For  $\delta > 0$ , the condition  $E|X|^{2+\delta} < \infty$  is neither sufficient nor necessary for the condition  $T(t) = O(t^{-\delta}l(t))$ . Here are some suitable examples.

**Example 1** Let  $X$  be a random variable with density  $f(x) = \frac{C(1+\delta \ln|x|)}{|x|^{3+\delta} \ln^2|x|} I(|x| > e)$ , where  $C$  is a normalizing constant, and  $0 < \delta < 1$ , then  $EX = 0$  and  $T(t) = \frac{C}{t^\delta \ln t} I(t > e)$ ,  $l(t) = \frac{1}{\ln t}$  is a slowly varying function at infinity. But  $E|X|^{2+\delta} = C \int_{|x|>e} \frac{1+\delta \ln|x|}{|x| \ln^2|x|} dx = \infty$ .

**Example 2** Let  $X$  be a random variable with density  $f(x) = \frac{C(\delta \ln^2|x| + |x|(\ln|x| - 1))}{|x|^{\delta+3} \ln^2|x| e^{|x|/\ln|x|}} I(|x| > e)$ , where  $0 < \delta < 1$ , then  $EX = 0$  and  $T(t) = \frac{C}{t^\delta e^{t/\ln t}} I(t > e)$ ,  $h(t) = \frac{1}{e^{t/\ln t}}$ ,  $E|X|^{2+\delta} < \infty$ . But  $h(t) = \frac{1}{e^{t/\ln t}}$  is not a slowly varying function at infinity.

In fact, we have the following result.

**Theorem 2.3** Suppose  $X$  is a real random variable and  $\delta > 0$ . Then  $E|X|^{2+\delta} < \infty$  if and only if  $t^\delta T(t) \rightarrow 0$  and  $\int_t^\infty s^{\delta-1} T(s) ds \rightarrow 0$  as  $t \rightarrow \infty$ .

**Remark 2.5** If  $t^\delta T(t)$  is bounded as  $t \rightarrow \infty$  for some  $\delta > 0$ , then we have  $E|X|^{2+\alpha} < \infty$  for every  $\alpha \in (0, \delta)$  from Theorem 2.3.

**Remark 2.6** Let  $X$  be a random variable with zero mean. If there exist positive constants  $C_1$  and  $C_2$  such that  $C_1 l(t) \leq t^\delta T(t) \leq C_2 l(t)$  for sufficiently large  $t$  and some  $\delta > 0$ , where  $l(t)$  is a slowly varying function at infinity, then from Lemma 2.2(4) and Theorem 2.3, we have

$$E|X|^{2+\delta} < \infty \iff \int_t^\infty \frac{l(s)}{s} ds \rightarrow 0 \text{ as } t \rightarrow \infty.$$

### 3 Proofs of the main results

*Proof of Theorem 2.1* Without loss of generality, we suppose that  $\sigma^2 = 1$ ,  $0 < \epsilon < 1$ . Since

$$P(|S_n| \geq n^{1/p} \epsilon) = 2(1 - \Phi(n^{(2-p)/2p} \epsilon)) + R_n, \tag{3.1}$$

where

$$R_n = P(S_n \leq -n^{1/p} \epsilon) - \Phi(-n^{1/p-1/2} \epsilon) + \Phi(n^{1/p-1/2} \epsilon) - P(S_n \leq n^{1/p} \epsilon).$$

From (3.1), we have

$$\begin{aligned} & \epsilon^{2(r-p)/(2-p)} \lambda_{r,p}(\epsilon) - \frac{p}{r-p} E|N|^{2(r-p)/(2-p)} \\ &= 2\epsilon^{2(r-p)/(2-p)} \sum_{n=1}^\infty n^{r/p-2} (1 - \Phi(n^{(2-p)/2p} \epsilon)) - \frac{p}{r-p} E|N|^{2(r-p)/(2-p)} \\ & \quad + \epsilon^{2(r-p)/(2-p)} \sum_{n=1}^\infty n^{r/p-2} R_n. \end{aligned} \tag{3.2}$$

By Lemma 2.1, in order to prove Theorem 2.1, we only need to estimate  $\epsilon^{2(r-p)/(2-p)} \times \sum_{n=1}^{\infty} n^{r/p-2} R_n$ .

(1) On account of a non-uniform estimate of the central limit theorem by Nagaev [8], for every  $x \in R$ ,

$$\left| P\left(\frac{S_n}{\sqrt{n}} < x\right) - \Phi(x) \right| \leq \frac{CE|X|^3}{\sqrt{n}(1+|x|)^3}. \tag{3.3}$$

By (3.3),  $|R_n| \leq \frac{CE|X|^3}{\sqrt{n}(1+\epsilon n^{(2-p)/2p})^3}$ .

(a) If  $r < 3p/2$ , then

$$\epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{\infty} n^{r/p-2} R_n \leq C\epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{\infty} n^{r/p-5/2} = O(\epsilon^{2(r-p)/(2-p)}). \tag{3.4}$$

(b) If  $3p/2 < r < 3$ , then

$$\begin{aligned} & \epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{\infty} n^{r/p-2} R_n \\ & \leq C\epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{\infty} \frac{n^{r/p-2}}{\sqrt{n}(1+\epsilon n^{(2-p)/2p})^3} \\ & \leq C\epsilon^{2(r-p)/(2-p)} \left( \sum_{n=1}^{[\epsilon^{-2p/(2-p)}]} \frac{n^{r/p-2}}{\sqrt{n}} + \epsilon^{-3} \sum_{n=[\epsilon^{-2p/(2-p)}]+1}^{\infty} n^{r/p-5/2-(6-3p)/2p} \right) \\ & = O(\epsilon^{p/(2-p)}). \end{aligned} \tag{3.5}$$

(c) If  $r = 3p/2$ , then

$$\begin{aligned} \epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{\infty} n^{r/p-2} R_n & \leq C\epsilon^{p/(2-p)} \left( \sum_{n=1}^{[\epsilon^{-2p/(2-p)}]} \frac{1}{n} + \epsilon^{-3} \sum_{n=[\epsilon^{-2p/(2-p)}]+1}^{\infty} n^{-1-(6-3p)/2p} \right) \\ & = O\left(\epsilon^{p/(2-p)} \log \frac{1}{\epsilon}\right). \end{aligned} \tag{3.6}$$

From (2.1), (3.2), (3.4), (3.5) and (3.6), we obtain (2.2). This completes the proof of part (1).

(2) By the inequality in Osipov and Petrov [9], there exists a bounded and decreasing function  $\psi(u)$  on the interval  $(0, \infty)$  such that  $\lim_{u \rightarrow \infty} \psi(u) = 0$  and

$$\left| P\left(\frac{1}{\sqrt{n}\sigma} S_n < x\right) - \Phi(x) \right| \leq \frac{\psi(\sqrt{n}(1+|x|))}{n^{\delta/2}(1+|x|)^{2+\delta}}.$$

Let  $x = n^{(2-p)/2p}\epsilon$ , we have  $|R_n| \leq \frac{2\psi(\sqrt{n}(1+n^{(2-p)/2p}\epsilon))}{n^{\delta/2}(1+n^{(2-p)/2p}\epsilon)^{2+\delta}}$ , so that:

(a) If  $2 < r < (1 + \delta/2)p$ , then

$$\epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{\infty} n^{r/p-2} R_n \leq \epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{\infty} n^{r/p-2-\delta/2} = O(\epsilon^{2(r-p)/(2-p)}). \tag{3.7}$$

(b) If  $(1 + \delta/2)p < r < 2 + \delta$ , then by noticing that  $\lim_{u \rightarrow \infty} \psi(u) = 0$  for any  $\eta > 0$ , there exists a natural number  $N_0$  such that  $\psi(\sqrt{n}) < \eta$  whenever  $n > N_0$ . We conclude that

$$\begin{aligned} & \epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{\infty} n^{r/p-2} R_n \\ & \leq C \epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{\infty} \frac{2n^{r/p-2} \psi(\sqrt{n}(1 + n^{(2-p)/2p}\epsilon))}{n^{\delta/2}(1 + \epsilon n^{(2-p)/2p})^{2+\delta}} \\ & \leq C \epsilon^{2(r-p)/(2-p)} \left( \sum_{n=1}^{N_0} n^{r/p-2-\delta/2} \psi(\sqrt{n}) + \eta \sum_{n=N_0+1}^{[\epsilon^{-2p/(2-p)}]} n^{r/p-2-\delta/2} \right) \\ & \quad + C \epsilon^{2(r-p)/(2-p)-2-\delta} \psi(\epsilon^{-p/(2-p)}) \sum_{n=[\epsilon^{-2p/(2-p)}]+1}^{\infty} n^{r/p-2-\delta/2-(1/p-1/2)(2+\delta)} \\ & \leq \epsilon^{2(r-p)/(2-p)} N_0^{r/p-1-\delta/2} + C \eta \epsilon^{p\delta/(2-p)} + C \psi(\epsilon^{-p/(2-p)}) \epsilon^{p\delta/(2-p)} \\ & = o(\epsilon^{p\delta/(2-p)}). \end{aligned} \tag{3.8}$$

(c) If  $r = (1 + \delta/2)p$ , then

$$\epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{\infty} n^{r/p-2} R_n = O\left(\epsilon^{p\delta/(2-p)} \log \frac{1}{\epsilon}\right). \tag{3.9}$$

By (2.1) and combining with (3.2), (3.7), (3.8) and (3.9), we obtain (2.3), which completes the proof of part (2).

(3) We make use of the following large deviation estimate in Petrov [10]:

$$\left| P\left(\frac{1}{\sqrt{n}\sigma} S_n < x\right) - \Phi(x) \right| \leq \frac{C}{\sqrt{n}(1 + |x|)^q}, \quad x > 0.$$

So,  $|R_n| \leq \frac{C}{\sqrt{n}(1 + \epsilon n^{(2-p)/2p})^q}$ . Hence we have the following.

(a) If  $r < 3p/2$ , then

$$\epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{\infty} n^{r/p-2} R_n \leq \epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{\infty} n^{r/p-5/2} = O(\epsilon^{2(r-p)/(2-p)}). \tag{3.10}$$

(b) If  $r > 3p/2$ , then  $\frac{r}{p} - \frac{5}{2} - \frac{2q-pq}{2p} < -1$ . By noting that  $q > \frac{2r-3p}{2-p}$ , we obtain

$$\begin{aligned} & \epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{\infty} n^{r/p-2} R_n \leq C \epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{\infty} \frac{n^{r/p-2}}{(1 + \epsilon n^{(2-p)/2p})^q \sqrt{n}} \\ & \leq C \epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{[\epsilon^{-2p/(2-p)}]} n^{r/p-2-1/2} \\ & \quad + C \epsilon^{2(r-p)/(2-p)-q} \sum_{n=[\epsilon^{-2p/(2-p)}]+1}^{\infty} n^{r/p-2-1/2-(2-p)q/2p} \\ & = O(\epsilon^{p/(2-p)}). \end{aligned} \tag{3.11}$$

(c) If  $r = 3p/2$ , then

$$\begin{aligned} \epsilon^{2(r-p)/(2-p)} \sum_{n=1}^{\infty} n^{r/p-2} R_n &\leq C\epsilon^{p/(2-p)} \sum_{n=1}^{[\epsilon^{-2p/(2-p)}]} \frac{1}{n} + C\epsilon^{p/(2-p)-q} \sum_{n=[\epsilon^{-2p/(2-p)}]+1}^{\infty} n^{-1-(2-p)q/2p} \\ &= O\left(\epsilon^{p/(2-p)} \log \frac{1}{\epsilon}\right). \end{aligned} \tag{3.12}$$

By (2.1), from (3.2), (3.10), (3.11) and (3.12), we have (2.4), which completes the proof of part (3).  $\square$

*Proof of Theorem 2.2* We write

$$\begin{aligned} \epsilon^2 \lambda_{2,1/(\beta+1)}(\epsilon) &= \frac{1}{2\beta+1} \\ &= \left( \frac{2\epsilon^2}{\sqrt{2\pi}} \sum_{n=1}^{\infty} n^{2\beta} \int_{\epsilon n^{\beta+1/2}}^{\infty} e^{-t^2/2} dt - \frac{1}{2\beta+1} \right) \\ &\quad + \epsilon^2 \left( \sum_{n=1}^{[\epsilon^{-4/(2\beta+1)}]} + \sum_{n=[\epsilon^{-4/(2\beta+1)}]+1}^{\infty} \right) n^{2\beta} \left( P(|S_n| \geq \epsilon n^{\beta+1}) - \frac{2}{\sqrt{2\pi}} \int_{\epsilon n^{\beta+1/2}}^{\infty} e^{-t^2/2} dt \right) \\ &=: I_1 + I_2 + I_3. \end{aligned} \tag{3.13}$$

First, according to Lemma 2.1, we have

$$I_1 = \begin{cases} O(\epsilon^2), & -\frac{1}{2} < \beta < \frac{1}{2}, \\ O(\epsilon^{4/(2\beta+1)}), & \beta \geq \frac{1}{2}. \end{cases} \tag{3.14}$$

For  $I_3$ , applying Lemma 2.3 of Xie and He [11], and letting  $x = 2y = n^{\beta+1}\epsilon$ , we obtain

$$P(|S_n| \geq n^{\beta+1}\epsilon) \leq nP\left(|X| \geq \frac{1}{2}n^{\beta+1}\epsilon\right) + 8e^2\epsilon^{-4}n^{-4\beta-2}. \tag{3.15}$$

Observing the following fact

$$\frac{2}{\sqrt{2\pi}} \int_{\epsilon n^{\beta+1/2}}^{\infty} e^{-t^2/2} dt = 2(1 - \Phi(n^{\beta+1/2}\epsilon)) \leq \frac{2\varphi(n^{\beta+1/2}\epsilon)}{n^{\beta+1/2}\epsilon} = O(\epsilon^{-5}n^{-5\beta-5/2}), \tag{3.16}$$

from (3.15) and (3.16), we have

$$\begin{aligned} I_3 &\leq \epsilon^2 \sum_{n=[\epsilon^{-4/(2\beta+1)}]+1}^{\infty} n^{2\beta} P(|S_n| \geq \epsilon n^{\beta+1}) + \epsilon^2 \sum_{n=[\epsilon^{-4/(2\beta+1)}]+1}^{\infty} \frac{2n^{2\beta}}{\sqrt{2\pi}} \int_{\epsilon n^{\beta+1/2}}^{\infty} e^{-t^2/2} dt \\ &\leq \epsilon^2 \sum_{n=[\epsilon^{-4/(2\beta+1)}]+1}^{\infty} n^{2\beta+1} P\left(|X| > \frac{\epsilon n^{\beta+1}}{2}\right) + C\epsilon^{-2} \sum_{n=[\epsilon^{-4/(2\beta+1)}]+1}^{\infty} n^{-2\beta-2} \\ &\quad + C\epsilon^{-3} \sum_{n=[\epsilon^{-4/(2\beta+1)}]+1}^{\infty} n^{-3\beta-5/2} \end{aligned}$$

$$\begin{aligned}
 &\leq \epsilon^2 \sum_{n=[\epsilon^{-4/(2\beta+1)}]_+}^{\infty} n^{2\beta+1} \int_{|x| \geq \frac{1}{2} n^{\beta+1} \epsilon} dF(x) + O(\epsilon^2) + O(\epsilon^3) \\
 &\leq \epsilon^2 \sum_{n=[\epsilon^{-4/(2\beta+1)}]_+}^{\infty} n^{2\beta+1} \sum_{k=n}^{\infty} \int_{\frac{1}{2} k^{\beta+1} \epsilon \leq x < \frac{1}{2} (k+1)^{\beta+1} \epsilon} dF(x) + O(\epsilon^2) \\
 &\leq \epsilon^2 \sum_{k=[\epsilon^{-4/(2\beta+1)}]_+}^{\infty} \sum_{n=1}^k n^{2\beta+1} \int_{\frac{1}{2} k^{\beta+1} \epsilon \leq x < \frac{1}{2} (k+1)^{\beta+1} \epsilon} dF(x) + O(\epsilon^2) \\
 &\leq C \epsilon^2 \sum_{k=[\epsilon^{-4/(2\beta+1)}]_+}^{\infty} k^{2\beta+2} \int_{\frac{1}{2} k^{\beta+1} \epsilon \leq x < \frac{1}{2} (k+1)^{\beta+1} \epsilon} dF(x) + O(\epsilon^2) \\
 &\leq C \sum_{k=[\epsilon^{-4/(2\beta+1)}]_+}^{\infty} \int_{\frac{1}{2} k^{\beta+1} \epsilon \leq x < \frac{1}{2} (k+1)^{\beta+1} \epsilon} x^2 dF(x) + O(\epsilon^2) \\
 &\leq C \int_{x \geq \frac{1}{2} \epsilon^{-(2\beta+3)/(2\beta+1)}} x^2 dF(x) + O(\epsilon^2) \\
 &= CT(\epsilon^{-(2\beta+3)/(2\beta+1)}) + O(\epsilon^2).
 \end{aligned}$$

Using the assumption on  $T(t)$  and Lemma 2.2(1), we can obtain

$$I_3 = \begin{cases} O(\epsilon^2), & -\frac{1}{2} < \beta \leq \frac{\min(\delta, 1)}{4} - \frac{1}{2}, \\ O(\epsilon^{1/(2\beta+1)}), & \beta \geq -\frac{1}{4}, \delta \geq 1, \\ O(\epsilon^{\delta/(2\beta+1)} l(\epsilon^{-1/(2\beta+1)})), & \beta \geq -\frac{1}{2} + \frac{\delta}{4}, 0 < \delta < 1. \end{cases} \tag{3.17}$$

For  $I_2$ , by Bikelis’s inequality (see [12]), we have

$$\begin{aligned}
 I_2 &\leq \epsilon^2 \sum_{n=1}^{[\epsilon^{-4/(2\beta+1)}]} \frac{Cn^{2\beta}}{(1 + \epsilon n^{\beta+1/2})^3 \sqrt{n}} \int_0^{(1+\epsilon n^{\beta+1/2})\sqrt{n}} T(v) dv \\
 &\leq \epsilon^2 \sum_{n=1}^{[\epsilon^{-2/(2\beta+1)}]} n^{2\beta-1/2} \int_0^{(1+\epsilon n^{\beta+1/2})\sqrt{n}} T(v) dv \\
 &\quad + \epsilon^{-1} \sum_{n=[\epsilon^{-2/(2\beta+1)}]_+}^{[\epsilon^{-4/(2\beta+1)}]} n^{-\beta-2} \int_0^{(1+\epsilon n^{\beta+1/2})\sqrt{n}} T(v) dv.
 \end{aligned}$$

Now, the proof of Theorem 2.2 will be divided into the following cases.

Case 1 of  $\delta > 1$ .

Noting that  $T(t) \leq EX^2 = 1$ , let  $\delta_1$  be a real number such that  $1 < \delta_1 < \delta$ , by Lemma 2.2(1),  $\lim_{t \rightarrow \infty} t^{\delta_1 - \delta} l(t) = 0$ . Therefore, there is a real number  $T_0 > 0$  such that  $|\frac{l(t)}{t^{\delta - \delta_1}}| < 1$  whenever  $t > T_0$ . Then

$$\int_0^{\infty} T(t) dt \leq \int_0^1 T(t) dt + \int_1^{\infty} T(t) dt \leq C + \int_{T_0}^{\infty} \frac{1}{t^{\delta_1}} dt < \infty.$$



We have

$$\begin{aligned}
 I_2 &\leq C\epsilon^2 \sum_{n=1}^{[\epsilon^{-2/(2\beta+1)}]} n^{2\beta-1/2} + C\epsilon^{-1} \sum_{n=[\epsilon^{-2/(2\beta+1)}]+1}^{[\epsilon^{-4/(2\beta+1)}]+1} n^{-\beta-2} \\
 &= \begin{cases} O(\epsilon^2), & -\frac{1}{2} < \beta \leq -\frac{1}{4}, \\ O(\epsilon^2 \log \frac{1}{\epsilon}), & \beta = -\frac{1}{4}, \\ O(\epsilon^{1/(2\beta+1)}), & \beta > -\frac{1}{4}. \end{cases} \tag{3.18}
 \end{aligned}$$

From (3.13), (3.14), (3.17) and (3.18), we obtain (2.5).

Case 2 of  $0 < \delta < 1$ .

(a) If  $-\frac{1}{2} < \beta < -\frac{1}{2} + \frac{\delta}{4}$ , then  $\sum_{n=1}^{\infty} n^{2\beta-1/2} < \infty$  and  $\int_1^{\infty} t^{4\beta+1-\delta} l(t) dt < \infty$ . Making use of Lemma 2.2(2)-(3), we have

$$\begin{aligned}
 I_2 &\leq C\epsilon^2 \sum_{n=1}^{[\epsilon^{-2/(2\beta+1)}]} n^{2\beta-1/2} \left( 1 + \int_1^{2\sqrt{n}} T(t) dt \right) \\
 &\quad + C\epsilon^{-1} \sum_{n=[\epsilon^{-2/(2\beta+1)}]+1}^{[\epsilon^{-4/(2\beta+1)}]} n^{-\beta-2} \left( 1 + \int_1^{2\epsilon n^{\beta+1}} T(t) dt \right) \\
 &\leq C\epsilon^2 \sum_{n=1}^{[\epsilon^{-2/(2\beta+1)}]} n^{2\beta-1/2} (\sqrt{n})^{1-\delta} l(\sqrt{n}) + O(\epsilon^2) \\
 &\quad + C\epsilon^{1/(2\beta+1)} + C\epsilon^{-1} \sum_{n=[\epsilon^{-2/(2\beta+1)}]+1}^{[\epsilon^{-4/(2\beta+1)}]} n^{-\beta-2} (2n^{\beta+1}\epsilon)^{1-\delta} l(2n^{\beta+1}\epsilon) \\
 &\leq C\epsilon^2 \int_1^{\epsilon^{-2/(2\beta+1)}} x^{2\beta-1/2} (\sqrt{x})^{1-\delta} l(\sqrt{x}) dx \\
 &\quad + C\epsilon^{-\delta} \int_{\epsilon^{-2/(2\beta+1)}}^{\infty} x^{-\beta-2} l(2x^{\beta+1}\epsilon) x^{(\beta+1)(1-\delta)} dx + O(\epsilon^2) \\
 &\leq C\epsilon^2 \int_1^{\epsilon^{-1/(2\beta+1)}} t^{4\beta+1-\delta} l(t) dt + C \int_{\epsilon^{-1/(2\beta+1)}}^{\infty} \frac{l(t)}{t^{1+\delta}} dt + O(\epsilon^2) \\
 &\leq C\epsilon^2 + C\epsilon^{\delta/(2\beta+1)} l(\epsilon^{-1/(2\beta+1)}) \\
 &= O(\epsilon^2). \tag{3.19}
 \end{aligned}$$

(b) If  $\beta \geq -\frac{1}{2} + \frac{\delta}{4}$ , then we have

$$\begin{aligned}
 I_2 &\leq C\epsilon^2 \epsilon^{-(4\beta+1)/(2\beta+1)} \left( 1 + \int_1^{2\epsilon^{-1/(2\beta+1)}} T(t) dt \right) + C\epsilon^{1/(2\beta+1)} + C\epsilon^{\delta/(2\beta+1)} l(\epsilon^{-1/(2\beta+1)}) \\
 &\leq C\epsilon^{1/(2\beta+1)} (1 + (2\epsilon^{-(1-\delta)/(2\beta+1)} l(\epsilon^{-1/(2\beta+1)}))) + C\epsilon^{\delta/(2\beta+1)} l(\epsilon^{-1/(2\beta+1)}) \\
 &\leq C\epsilon^{\delta/(2\beta+1)} l(\epsilon^{-1/(2\beta+1)}). \tag{3.20}
 \end{aligned}$$

Therefore

$$I_2 = \begin{cases} O(\epsilon^2), & -\frac{1}{2} < \beta \leq -\frac{1}{2} + \frac{\delta}{4}, \\ O(\epsilon^{\delta/(2\beta+1)}l(\epsilon^{1/(2\beta+1)})), & \beta \geq -\frac{1}{2} + \frac{\delta}{4}. \end{cases} \quad (3.21)$$

Combining the estimate with (3.11) and (3.14), by (3.10), this implies that (2.6) follows.

Case 3 of  $\delta = 1$ .

(a) If  $-\frac{1}{2} < \beta < -\frac{1}{4}$ , then  $\sum_{n=1}^{\infty} n^{2\beta-\frac{1}{2}} < \infty$ . We have

$$\begin{aligned} I_2 &\leq C\epsilon^2 \left( 1 + \int_1^{\epsilon^{-1/(2\beta+1)}} T(t) dt \right) \sum_{n=1}^{[\epsilon^{-2/(2\beta+1)}]} n^{2\beta-1/2} \\ &\quad + C\epsilon^{-1} \left( 1 + \int_1^{\epsilon^{-\frac{2\beta+3}{2\beta+1}}} T(t) dt \right) \sum_{n=[\epsilon^{-2/(2\beta+1)}]+1}^{[\epsilon^{-4/(2\beta+1)}]} n^{-\beta-2} \\ &\leq C\epsilon^2 \left( 1 + \int_1^{\epsilon^{-1/(2\beta+1)}} \frac{l(t)}{t} dt \right) + \epsilon^{1/(2\beta+1)} \left( 1 + \int_1^{\epsilon^{-\frac{2\beta+3}{2\beta+1}}} \frac{l(t)}{t} dt \right) \\ &\leq C\epsilon^2 \left( 1 + \int_1^{\epsilon^{-1/(2\beta+1)}} \frac{l(t)}{t} dt \right). \end{aligned} \quad (3.22)$$

(b) If  $\beta > -\frac{1}{4}$ , then we have

$$\begin{aligned} I_2 &\leq C\epsilon^2 \epsilon^{\frac{-2}{2\beta+1}(2\beta+1/2)} \left( 1 + \int_1^{\epsilon^{-1/(2\beta+1)}} \frac{l(t)}{t} dt \right) + C\epsilon^{1/(2\beta+1)} \left( 1 + \int_1^{\epsilon^{-(2\beta+3)/(2\beta+1)}} \frac{l(t)}{t} dt \right) \\ &\leq C\epsilon^{1/(2\beta+1)} \left( 1 + \int_1^{\epsilon^{-1/(2\beta+1)}} \frac{l(t)}{t} dt \right) + \epsilon^{1/(2\beta+1)} \left( 1 + \int_1^{\epsilon^{-(2\beta+3)/(2\beta+1)}} \frac{l(t)}{t} dt \right) \\ &\leq C\epsilon^{1/(2\beta+1)} \left( 1 + \int_1^{\epsilon^{-(2\beta+3)/(2\beta+1)}} \frac{l(t)}{t} dt \right). \end{aligned} \quad (3.23)$$

(c) If  $\beta = -\frac{1}{4}$ , then we have

$$\begin{aligned} I_2 &\leq C\epsilon^2 \log \frac{1}{\epsilon} \left( 1 + \int_1^{\epsilon^{-2}} \frac{l(t)}{t} dt \right) + C\epsilon^2 \left( 1 + \int_1^{\epsilon^{-5}} \frac{l(t)}{t} dt \right) \\ &\leq C\epsilon^2 \log \frac{1}{\epsilon} \left( 1 + \int_1^{\epsilon^{-5}} \frac{l(t)}{t} dt \right) \end{aligned}$$

so that

$$I_2 = \begin{cases} O(\epsilon^2 (1 + \int_1^{\epsilon^{-1/(2\beta+1)}} \frac{l(t)}{t} dt)), & -\frac{1}{2} < \beta \leq -\frac{1}{4}, \\ O(\epsilon^2 (1 + \int_1^{\epsilon^{-5}} \frac{l(t)}{t} dt) \log \frac{1}{\epsilon}), & \beta = -\frac{1}{4}, \\ O(\epsilon^{1/(2\beta+1)} (1 + \int_1^{\epsilon^{-(2\beta+3)/(2\beta+1)}} \frac{l(t)}{t} dt)), & \beta > -\frac{1}{4}. \end{cases} \quad (3.24)$$

Combining the estimate with (3.14) and (3.17), by (3.13), this implies that (2.7) follows, and hence Theorem 2.2 is proved.  $\square$

*Proof of Theorem 2.3* Set  $T_1(t) = E|X|^{2+\delta}I(|X| > t)$ . First, note that

$$\begin{aligned} E|X|^{2+\delta}I(|X| > t) &= \int_{|x|>t} |x|^{2+\delta} dF(x) \\ &= \int_{|x|>t} x^2 \left( \int_t^{|x|} \delta y^{\delta-1} dy \right) dF(x) + t^\delta \int_{|x|>t} x^2 dF(x) \\ &= \int_t^\infty \delta y^{\delta-1} \left( \int_{|x|>y} x^2 dF(x) \right) dy + t^\delta T(t) \\ &= \delta \int_t^\infty s^{\delta-1} T(s) ds + t^\delta T(t). \end{aligned}$$

We have

$$T_1(t) = \delta \int_t^\infty s^{\delta-1} T(s) ds + t^\delta T(t).$$

Since  $\int_t^\infty s^{\delta-1} T(s) ds \geq 0$ ,  $t^\delta T(t) \geq 0$ , we have

$$T_1(t) \rightarrow 0 \iff t^\delta T(t) \rightarrow 0 \text{ and } \int_t^\infty s^{\delta-1} T(s) ds \rightarrow 0 \text{ as } t \rightarrow \infty.$$

Next, it is easy to get

$$E|X|^{2+\delta} < \infty \iff T_1(t) \rightarrow 0 \text{ as } t \rightarrow \infty.$$

From the above facts, the proof of Theorem 2.3 is complete. □

**Competing interests**

The author declares that they have no competing interests.

**Acknowledgements**

The author would like to thank the referee for helpful comments.

Received: 3 February 2013 Accepted: 18 July 2013 Published: 11 August 2013

**References**

1. Heyde, CC: A supplement to the strong law of large numbers. *J. Appl. Probab.* **12**, 903-907 (1975)
2. Klesov, OI: On the convergence rate in a theorem of Heyde. *Theory Probab. Math. Stat.* **49**, 83-87 (1994)
3. He, JJ, Xie, TF: Asymptotic property for some series of probability. *Acta Math. Appl. Sin.* **29**, 179-186 (2013)
4. Gut, A, Steinebach, J: Convergence rates in precise asymptotics. *J. Math. Anal. Appl.* **390**, 1-14 (2012)
5. Gut, A, Steinebach, J: Precise asymptotics-a general approach. *Acta Math. Hung.* **138**, 365-385 (2013)
6. Gut, A, Steinebach, J: Correction to 'Convergence rates in precise asymptotics'. Preprint (<http://www2.math.uu.se/~allan/90correction.pdf>) (2012)
7. Bingham, NH, Goldie, CM, Teugels, JL: *Regular Variation*. Cambridge University Press, Cambridge (1987)
8. Nagaev, SV: Some limit theorems for large deviation. *Theory Probab. Appl.* **10**, 214-235 (1965)
9. Osipov, LV, Petrov, VV: On an estimate of remainder term in the central limit theorem. *Theory Probab. Appl.* **12**, 281-286 (1967)
10. Petrov, VV: *Limit Theorems of Probability Theory*. Oxford University Press, Oxford (1995)
11. Xie, TF, He, JJ: Rate of convergence in a theorem of Heyde. *Stat. Probab. Lett.* **82**, 1576-1582 (2012)
12. Bikelis, A: Estimates of the remainder in the central limit theorem. *Liet. Mat. Rink.* **6**, 323-346 (1966)

doi:10.1186/1029-242X-2013-378

**Cite this article as:** He: A note to the convergence rates in precise asymptotics. *Journal of Inequalities and Applications* 2013 **2013**:378.