

Research Article

A Limit Theorem for the Moment of Self-Normalized Sums

Qing-pei Zang

Department of Mathematics, Huaiyin Teachers College, Huaian 223300, China

Correspondence should be addressed to Qing-pei Zang, zqphunhu@yahoo.com.cn

Received 25 December 2008; Revised 30 March 2009; Accepted 18 June 2009

Recommended by Jewgeni Dshalalow

Let $\{X, X_n; n \geq 1\}$ be a sequence of independent and identically distributed (*i.i.d.*) random variables and X is in the domain of attraction of the normal law and $EX = 0$. For $1 \leq p < 2, b > -1$, we prove the precise asymptotics in Davis law of large numbers for $\sum_{n=1}^{\infty} ((\log n)^b/n) E\{|S_n|/V_n) - \varepsilon(2 \log n)^{(2-p)/(2p)}\} + \text{as } \varepsilon \searrow 0$.

Copyright © 2009 Qing-pei Zang. This is an open access article distributed under the Creative Commons Attribution License, which permits unrestricted use, distribution, and reproduction in any medium, provided the original work is properly cited.

1. Introduction and Main Result

Throughout this paper, we let $\{X, X_n; n \geq 1\}$ be a sequence of *i.i.d.* random variables and X is in the domain of attraction of the normal law and $EX = 0$. Put

$$S_n = \sum_{k=1}^n X_k, \quad V_n^2 = \sum_{i=1}^n X_i^2. \quad (1.1)$$

Also let $\log n = \ln(n \vee e)$. Then by the well-known Davis laws of large numbers [1],

$$\sum_{n=1}^{\infty} \frac{\log n}{n} P\left(|S_n| \geq \varepsilon \sqrt{n \log n}\right) < \infty, \quad \varepsilon > 0, \quad (1.2)$$

if and only if $EX = 0$ and $EX^2 < \infty$.

Gut and Spătaru [2] proved its precise asymptotics as follows.

Theorem A. Suppose that $EX_1 = 0$ and $EX_1^2 = \sigma^2 < \infty$. Then for $0 \leq \delta \leq 1$,

$$\lim_{\varepsilon \searrow 0} \varepsilon^{2(\delta+1)} \sum_{n=1}^{\infty} \frac{(\log n)^{\delta}}{n} P\left(|S_n| \geq \varepsilon \sqrt{n \log n}\right) = \frac{\mu^{(2\delta+2)}}{\delta+1} \sigma^{2\delta+2}, \quad (1.3)$$

where $\mu^{(2\delta+2)}$ stands for the $(2\delta+2)$ th absolute moment of the standard normal distribution.

It is well known that, for i.i.d. random variables, Chow [3] discussed the complete moment convergence, and got the following result.

Theorem B. Let $\{X, X_n; n \geq 1\}$ be a sequence of i.i.d. random variables with $EX_1 = 0$. Assume $p \geq 1$, $\alpha > 1/2$, $p\alpha > 1$, and $E(|X|^p + |X| \log(1 + |X|)) < \infty$. Then for any $\varepsilon > 0$,

$$\sum_{n=1}^{\infty} n^{p\alpha-2-\alpha} E\left\{\max_{j \leq n} |S_j| - \varepsilon n^{\alpha}\right\}_+ < \infty. \quad (1.4)$$

On the other hand, the past decade has witnessed a significant development on the limit theorems for the so-called self-normalized sum S_n/V_n , $V_n = \sqrt{\sum_{i=1}^n X_i^2}$. Bentkus and Götze [4] obtained Berry-Esseen inequalities for self-normalized sums. Wang and Jing [5] derived exponential nonuniform Berry-Esseen bound. Giné et al. [6], established asymptotic normality of self-normalized sums.

Theorem C. Let $\{X, X_n; n \geq 1\}$ be a sequence of i.i.d. random variables with $EX_1 = 0$. Then for any $x \in \mathbb{R}$,

$$\lim_{n \rightarrow \infty} P\left(\frac{S_n}{V_n} \leq x\right) = \Phi(x) \quad (1.5)$$

holds, if and only if X is in the domain of attraction of the normal law, where $\Phi(x)$ is the distribution function of the standard normal random variable.

Shao [7] showed a self-normalization large deviation result for $P(S_n/V_n \geq x\sqrt{n})$ without any moment conditions.

Theorem D. Let $\{x_n; n \geq 1\}$ be a sequence of positive numbers with $x_n \rightarrow \infty$ and $x_n = o(\sqrt{n})$ as $n \rightarrow \infty$. If $EX = 0$ and $EX^2 I(|X| \leq x)$ is slowly varying as $x \rightarrow \infty$, then

$$\lim_{n \rightarrow \infty} x_n^{-2} \ln P\left(\frac{S_n}{V_n} \geq x_n\right) = -\frac{1}{2}. \quad (1.6)$$

Since then, many subsequent developments of self-normalized sums have been obtained. For example, Csörgő et al. [8] have established Darling-Erdős theorem for self-normalized sums, and they [9] have also obtained Donsker's theorem for self-normalized partial sums processes.

Inspired by the above results, in this note we study the precise asymptotics in Davis law of large numbers for the moment of self-normalized sums. Our main result is as follows.

Theorem 1.1. *Suppose X is in the domain of attraction of the normal law and $EX = 0$. Then, for $b > -1$ and $1 \leq p < 2$, one has*

$$\begin{aligned} \lim_{\varepsilon \searrow 0} \varepsilon^{2p(b+1)/(2-p)} \sum_{n=1}^{\infty} \frac{(\log n)^b}{n} E \left\{ \frac{|S_n|}{V_n} - \varepsilon (2 \log n)^{(2-p)/(2p)} \right\}_+ \\ = \frac{2^{-b-1} (2-p)}{(b+1) (2pb+p+2)} E|N|^{(2pb+p+2)/(2-p)}, \end{aligned} \quad (1.7)$$

here and in the sequel, N is the standard normal random variable.

Remark 1.2. If $p = 1$ and $0 < \sigma^2 = EX^2 < \infty$, by the strong law of large numbers, we have $V_n^2/n \rightarrow \sigma^2$, a.s. Then, we can easily obtain the following result:

$$\lim_{\varepsilon \searrow 0} \varepsilon^{2(b+1)} \sum_{n=1}^{\infty} \frac{(\log n)^b}{n^{3/2}} E \left\{ |S_n| - \varepsilon \sigma \sqrt{2n \log n} \right\}_+ = \frac{\sigma 2^{-b-1}}{(b+1) (2b+3)} E|N|^{2b+3}. \quad (1.8)$$

Remark 1.3. As is well known, the strong approximation method is taken in order to obtain such an analogous result, however, this method is not applicable here.

2. Proof of Theorem 1.1

In this section, we set $A(\varepsilon) = \exp(M/\varepsilon^{2p/(2-p)})$, for $M > 1$ and $\varepsilon > 0$. Here and in the sequel, C will denote positive constants, possibly varying from place to place, and $[x]$ means the largest integer $\leq x$. The proof of Theorem 1.1 is based on the following propositions.

Proposition 2.1. *For $b > -1$, one has*

$$\begin{aligned} \lim_{\varepsilon \searrow 0} \varepsilon^{2p(b+1)/(2-p)} \sum_{n=1}^{\infty} \frac{(\log n)^b}{n} E \left\{ |N| - \varepsilon (2 \log n)^{(2-p)/(2p)} \right\}_+ \\ = \frac{2^{-b-1} (2-p)}{(b+1) (2pb+p+2)} E|N|^{(2pb+p+2)/(2-p)}. \end{aligned} \quad (2.1)$$

Proof. Via the change of variable $y = \varepsilon(2 \log t)^{(2-p)/(2p)}$, we have

$$\begin{aligned}
& \lim_{\varepsilon \searrow 0} \varepsilon^{2p(b+1)/(2-p)} \sum_{n=1}^{\infty} \frac{(\log n)^b}{n} E \left\{ |N| - \varepsilon(2 \log n)^{(2-p)/(2p)} \right\}_+ \\
&= \lim_{\varepsilon \searrow 0} \varepsilon^{2p(b+1)/(2-p)} \sum_{n=1}^{\infty} \frac{(\log n)^b}{n} \int_{\varepsilon(2 \log n)^{(2-p)/(2p)}}^{\infty} P(|N| \geq x) dx \\
&= \lim_{\varepsilon \searrow 0} \varepsilon^{2p(b+1)/(2-p)} \int_e^{\infty} \frac{(\log t)^b}{t} \int_{\varepsilon(2 \log t)^{(2-p)/(2p)}}^{\infty} P(|N| \geq x) dx dt \\
&= \lim_{\varepsilon \searrow 0} \frac{p2^{-b}}{2-p} \int_{\varepsilon 2^{(2-p)/(2p)}}^{\infty} y^{(2p/(2-p))(b+1)-1} \int_y^{\infty} P(|N| \geq x) dx dy \\
&= \lim_{\varepsilon \searrow 0} \frac{p2^{-b}}{2-p} \int_{\varepsilon 2^{(2-p)/(2p)}}^{\infty} P(|N| \geq x) \int_{\varepsilon 2^{(2-p)/(2p)}}^x y^{(2p/(2-p))(b+1)-1} dy dx \\
&= \lim_{\varepsilon \searrow 0} \frac{2^{-b-1}}{(b+1)} \int_{\varepsilon 2^{(2-p)/(2p)}}^{\infty} P(|N| \geq x) \left(x^{(2p/(2-p))(b+1)} - \varepsilon^{(2p/(2-p))(b+1)} \cdot 2^{b+1} \right) dx \\
&= \lim_{\varepsilon \searrow 0} \frac{2^{-b-1}}{(b+1)} \int_{\varepsilon 2^{(2-p)/(2p)}}^{\infty} x^{(2p/(2-p))(b+1)} P(|N| \geq x) dx \\
&= \frac{2^{-b-1} (2-p)}{(b+1) (2pb+p+2)} E|N|^{(2pb+p+2)/(2-p)}.
\end{aligned} \tag{2.2}$$

□

Proposition 2.2. For $b > -1$, one has

$$\lim_{\varepsilon \searrow 0} \varepsilon^{2p(b+1)/(2-p)} \sum_{n \leq A(\varepsilon)} \frac{(\log n)^b}{n} \left| E \left\{ \frac{|S_n|}{V_n} - \varepsilon(2 \log n)^{(2-p)/(2p)} \right\}_+ - E \left\{ |N| - \varepsilon(2 \log n)^{(2-p)/(2p)} \right\}_+ \right| = 0. \tag{2.3}$$

Proof. Set $\Delta_n = \sup_{x \in \mathbb{R}} |P(|S_n|/V_n \geq x) - P(|N| \geq x)|$. Then, by (1.5), it is easy to see $\Delta_n \rightarrow 0$ as $n \rightarrow \infty$. Observe that

$$\begin{aligned}
& \lim_{\varepsilon \searrow 0} \varepsilon^{2p(b+1)/(2-p)} \sum_{n \leq A(\varepsilon)} \frac{(\log n)^b}{n} \left| E \left\{ \frac{|S_n|}{V_n} - \varepsilon(2 \log n)^{(2-p)/(2p)} \right\}_+ - E \left\{ |N| - \varepsilon(2 \log n)^{(2-p)/(2p)} \right\}_+ \right| \\
&= \lim_{\varepsilon \searrow 0} \varepsilon^{2p(b+1)/(2-p)} \sum_{n \leq A(\varepsilon)} \frac{(\log n)^b}{n} \\
&\quad \times \left| \int_0^{\infty} P \left(\frac{|S_n|}{V_n} \geq x + \varepsilon(2 \log n)^{(2-p)/(2p)} \right) dx - \int_0^{\infty} P \left(|N| \geq x + \varepsilon(2 \log n)^{(2-p)/(2p)} \right) dx \right|
\end{aligned}$$

$$\begin{aligned}
&\leq \lim_{\varepsilon \searrow 0} \varepsilon^{2p(b+1)/(2-p)} \sum_{n \leq A(\varepsilon)} \frac{(\log n)^b}{n} \int_0^\infty \left| P\left(\frac{|S_n|}{V_n} \geq x + \varepsilon(2 \log n)^{(2-p)/(2p)}\right) \right. \\
&\quad \left. - \int_0^\infty P\left(|N| \geq x + \varepsilon(2 \log n)^{(2-p)/(2p)}\right) dx \right| dx \\
&\leq \lim_{\varepsilon \searrow 0} \varepsilon^{2p(b+1)/(2-p)} \sum_{n \leq A(\varepsilon)} \frac{(\log n)^b}{n} (\Delta_{n1} + \Delta_{n2} + \Delta_{n3} + \Delta_{n4}), \tag{2.4}
\end{aligned}$$

where

$$\begin{aligned}
\Delta_{n1} &= \int_0^{\min(\log n, 1/\sqrt{\Delta_n})} \left| P\left(\frac{|S_n|}{V_n} \geq x + \varepsilon(2 \log n)^{(2-p)/(2p)}\right) \right. \\
&\quad \left. - P\left(|N| \geq x + \varepsilon(2 \log n)^{(2-p)/(2p)}\right) \right| dx, \\
\Delta_{n2} &= \int_{\min(\log n, 1/\sqrt{\Delta_n})}^{n^{1/4}} \left| P\left(\frac{|S_n|}{V_n} \geq x + \varepsilon(2 \log n)^{(2-p)/(2p)}\right) \right. \\
&\quad \left. - P\left(|N| \geq x + \varepsilon(2 \log n)^{(2-p)/(2p)}\right) \right| dx, \tag{2.5} \\
\Delta_{n3} &= \int_{n^{1/4}}^{n^{1/2}} \left| P\left(\frac{|S_n|}{V_n} \geq x + \varepsilon(2 \log n)^{(2-p)/(2p)}\right) - P\left(|N| \geq x + \varepsilon(2 \log n)^{(2-p)/(2p)}\right) \right| dx, \\
\Delta_{n4} &= \int_{n^{1/2}}^\infty \left| P\left(\frac{|S_n|}{V_n} \geq x + \varepsilon(2 \log n)^{(2-p)/(2p)}\right) - P\left(|N| \geq x + \varepsilon(2 \log n)^{(2-p)/(2p)}\right) \right| dx.
\end{aligned}$$

Thus for Δ_{n1} , it is easy to see

$$\Delta_{n1} \leq \sqrt{\Delta_n} \longrightarrow 0, \quad \text{as } n \longrightarrow \infty. \tag{2.6}$$

Now we are in a position to estimate Δ_{n2} . From (1.6), and by applying $-X_i^j$ s to it, we can obtain that for large enough n and any $0 < a \leq 1/4$, there exist C and b such that $P(|S_n|/V_n > x) \leq Ce^{-(1/2-a)x^2}$ for $b < x < n^{1/2}/b$. In particular, for $b < x < n^{1/2}/b$, there exists $C > 0$ such that

$$P\left(\frac{|S_n|}{V_n} > x\right) \leq Ce^{-x^2/4}. \tag{2.7}$$

Hence, by Markov's inequality and (2.7), we have

$$\begin{aligned}\Delta_{n2} &\leq \int_{\min(\log n, 1/\sqrt{\Delta_n})}^{n^{1/4}} e^{-(x+\varepsilon(2\log n)^{(2-p)/2p})^2/4} dx + \int_{\min(\log n, 1/\sqrt{(\Delta_n)})}^{n^{1/4}} \frac{C}{\left(x + \varepsilon(2\log n)^{(2-p)/(2p)}\right)^2} dx \\ &\leq \int_{\min(\log n, 1/\sqrt{\Delta_n})}^{n^{1/4}} e^{-x^2/4} dx + \int_{\min(\log n, 1/\sqrt{(\Delta_n)})}^{n^{1/4}} \frac{C}{x^2} dx \longrightarrow 0, \quad \text{as } n \longrightarrow \infty.\end{aligned}\quad (2.8)$$

For Δ_{n3} , by Markov's inequality and (2.7), we have

$$\begin{aligned}\Delta_{n3} &\leq \int_{n^{1/4}}^{n^{1/2}} P\left(\frac{|S_n|}{V_n} \geq n^{1/4}\right) dx + \int_{n^{1/4}}^{n^{1/2}} \frac{C}{\left(x + \varepsilon(2\log n)^{(2-p)/(2p)}\right)^2} dx \\ &\leq e^{-\sqrt{n}/4} \left(n^{1/2} - n^{1/4}\right) + \int_{n^{1/4}}^{n^{1/2}} \frac{C}{x^2} dx \longrightarrow 0, \quad \text{as } n \longrightarrow \infty.\end{aligned}\quad (2.9)$$

From Cauchy inequality, it follows that

$$\frac{|S_n|}{V_n} \leq \sqrt{n}.\quad (2.10)$$

Therefore

$$\begin{aligned}\Delta_{n4} &= \int_{n^{1/2}}^{\infty} P\left(|N| \geq x + \varepsilon(2\log n)^{(2-p)/(2p)}\right) dx \\ &\leq \int_{n^{1/2}}^{\infty} \frac{C}{\left(x + \varepsilon(2\log n)^{(2-p)/(2p)}\right)^2} dx \\ &\leq \int_{n^{1/2}}^{\infty} \frac{C}{x^2} dx \longrightarrow 0, \quad \text{as } n \longrightarrow \infty.\end{aligned}\quad (2.11)$$

Denote $\Delta'_n = \Delta_{n1} + \Delta_{n2} + \Delta_{n3} + \Delta_{n4}$, then, since the weighted average of a sequence that converges to 0 also converges to 0, it follows that, for any $M > 1$,

$$\begin{aligned}&\lim_{\varepsilon \searrow 0} \varepsilon^{2p(b+1)/(2-p)} \sum_{n \leq A(\varepsilon)} \frac{(\log n)^b}{n} \left| E\left\{ \frac{|S_n|}{V_n} - \varepsilon(2\log n)^{(2-p)/(2p)} \right\}_+ - E\left\{ |N| - \varepsilon(2\log n)^{(2-p)/(2p)} \right\}_+ \right| \\ &\leq \lim_{\varepsilon \searrow 0} \varepsilon^{2p(b+1)/(2-p)} \sum_{n \leq A(\varepsilon)} \frac{(\log n)^b}{n} \Delta'_n \longrightarrow 0, \quad \text{as } \varepsilon \searrow 0.\end{aligned}\quad (2.12)$$

The proof is completed. \square

Proposition 2.3. For $b > -1$, one has

$$\lim_{M \rightarrow \infty} \lim_{\varepsilon \searrow 0} \varepsilon^{2p(b+1)/(2-p)} \sum_{n > A(\varepsilon)} \frac{(\log n)^b}{n} E \left\{ |N| - \varepsilon (2 \log n)^{(2-p)/(2p)} \right\}_+ = 0. \quad (2.13)$$

Proof. Note that

$$\begin{aligned} & \varepsilon^{2p(b+1)/(2-p)} \sum_{n > A(\varepsilon)} \frac{(\log n)^b}{n} E \left\{ |N| - \varepsilon (2 \log n)^{(2-p)/(2p)} \right\}_+ \\ & \leq \varepsilon^{2p(b+1)/(2-p)} \int_{A(\varepsilon)}^{\infty} \frac{(\log t)^b}{t} \int_{\varepsilon (2 \log t)^{(2-p)/(2p)}}^{\infty} P(|N| \geq x) dx dt \\ & \leq \int_{\sqrt{2M}}^{\infty} y^{(2p/(2-p))(b+1)-1} \int_y^{\infty} P(|N| \geq x) dx dy \\ & = \int_{\sqrt{2M}}^{\infty} P(|N| \geq x) \int_{\sqrt{2M}}^x y^{(2p/(2-p))(b+1)-1} dy dx \\ & \leq C \int_{\sqrt{2M}}^{\infty} x^{(2p/(2-p))(b+1)} P(|N| \geq x) dx \longrightarrow 0, \quad \text{as } M \longrightarrow \infty. \end{aligned} \quad (2.14)$$

So this proposition is proved now. \square

Proposition 2.4. For $b > -1$, one has

$$\lim_{M \rightarrow \infty} \lim_{\varepsilon \searrow 0} \varepsilon^{2p(b+1)/(2-p)} \sum_{n > A(\varepsilon)} \frac{(\log n)^b}{n} E \left\{ \frac{|S_n|}{V_n} - \varepsilon (2 \log n)^{(2-p)/(2p)} \right\}_+ = 0. \quad (2.15)$$

Proof. Note that

$$\begin{aligned} & \varepsilon^{2p(b+1)/(2-p)} \sum_{n > A(\varepsilon)} \frac{(\log n)^b}{n} E \left\{ \frac{|S_n|}{V_n} - \varepsilon (2 \log n)^{(2-p)/(2p)} \right\}_+ \\ & = \varepsilon^{2p(b+1)/(2-p)} \sum_{n > A(\varepsilon)} \frac{(\log n)^b}{n} \int_0^{\infty} P \left(\frac{|S_n|}{V_n} \geq x + \varepsilon (2 \log n)^{(2-p)/(2p)} \right) dx \\ & = B_1 + B_2 + B_3, \end{aligned} \quad (2.16)$$

where

$$\begin{aligned}
 B_1 &= \varepsilon^{2p(b+1)/(2-p)} \sum_{n>A(\varepsilon)} \frac{(\log n)^b}{n} \int_0^{n^{1/4}} P\left(\frac{|S_n|}{V_n} \geq x + \varepsilon(2 \log n)^{(2-p)/(2p)}\right) dx, \\
 B_2 &= \varepsilon^{2p(b+1)/(2-p)} \sum_{n>A(\varepsilon)} \frac{(\log n)^b}{n} \int_{n^{1/4}}^{n^{1/2}} P\left(\frac{|S_n|}{V_n} \geq x + \varepsilon(2 \log n)^{(2-p)/(2p)}\right) dx, \\
 B_3 &= \varepsilon^{2p(b+1)/(2-p)} \sum_{n>A(\varepsilon)} \frac{(\log n)^b}{n} \int_{n^{1/2}}^{\infty} P\left(\frac{|S_n|}{V_n} \geq x + \varepsilon(2 \log n)^{(2-p)/(2p)}\right) dx.
 \end{aligned} \tag{2.17}$$

For B_1 , by (2.7), we have

$$\begin{aligned}
 B_1 &\leq C \varepsilon^{2p(b+1)/(2-p)} \sum_{n>A(\varepsilon)} \frac{(\log n)^b}{n} \int_0^{n^{1/4}} e^{-(x+\varepsilon(2 \log n)^{(2-p)/(2p)})^2/4} dx \\
 &\leq C \varepsilon^{2p(b+1)/(2-p)} \sum_{n>A(\varepsilon)} \frac{(\log n)^b}{n} \int_0^{\infty} e^{-(x+\varepsilon(2 \log n)^{(2-p)/(2p)})^2/4} dx \\
 &= C \varepsilon^{2p(b+1)/(2-p)} \sum_{n>A(\varepsilon)} \frac{(\log n)^b}{n} \int_{\varepsilon(2 \log n)^{(2-p)/(2p)}}^{\infty} e^{-x^2/4} dx \\
 &\leq C \varepsilon^{2p(b+1)/(2-p)} \int_{A(\varepsilon)}^{\infty} \frac{(\log n)^b}{t} \int_{\varepsilon(2 \log t)^{(2-p)/(2p)}}^{\infty} e^{-x^2/4} dx dt \\
 &\leq C \int_{\sqrt{2M}}^{\infty} y^{(2p/(2-p))(b+1)-1} \int_y^{\infty} e^{-x^2/4} dx dy \\
 &= C \int_{\sqrt{2M}}^{\infty} e^{-x^2/4} \int_{\sqrt{2M}}^x y^{(2p/(2-p))(b+1)-1} dy dx \\
 &\leq C \int_{\sqrt{2M}}^{\infty} x^{(2p/(2-p))(b+1)} e^{-x^2/4} dx \longrightarrow 0, \quad \text{as } M \longrightarrow \infty.
 \end{aligned} \tag{2.18}$$

For B_2 , using (2.7) again, we have

$$\begin{aligned}
 B_2 &\leq \varepsilon^{2p(b+1)/(2-p)} \sum_{n>A(\varepsilon)} \frac{(\log n)^b}{n} \left(n^{1/2} - n^{1/4}\right) P\left(\frac{|S_n|}{V_n} \geq n^{1/4} + \varepsilon(2 \log n)^{(2-p)/(2p)}\right) \\
 &\leq C \varepsilon^{2p(b+1)/(2-p)} \sum_{n>A(\varepsilon)} \frac{(\log n)^b}{n} \left(n^{1/2} - n^{1/4}\right) e^{-(n^{1/4} + \varepsilon(2 \log n)^{(2-p)/(2p)})^2/4}
 \end{aligned}$$

$$\begin{aligned}
&\leq C\varepsilon^{2p(b+1)/(2-p)} \sum_{n>A(\varepsilon)} \frac{(\log n)^b}{n} \left(n^{1/2} - n^{1/4}\right) e^{-\sqrt{n}/4} e^{-\varepsilon^2(2\log n)^{(2-p)/p}/4} \\
&\leq C\varepsilon^{2p(b+1)/(2-p)} \sum_{n>A(\varepsilon)} \frac{(\log n)^b}{n} e^{-\varepsilon^2(2\log n)^{(2-p)/p}/4} \\
&\leq C\varepsilon^{2p(b+1)/(2-p)} \int_{A(\varepsilon)}^{\infty} \frac{(\log t)^b}{t} e^{-\varepsilon^2(2\log t)^{(2-p)/p}/4} dt \\
&\quad \left(\text{by letting } z = \frac{\varepsilon^2(2\log t)^{(2-p)/p}}{4}\right) \\
&\leq C \int_{(2M)^{(2-p)/p}/4}^{\infty} z^{(p(b+1))/(2-p)-1} e^{-z} dz \longrightarrow 0, \quad \text{as } M \longrightarrow \infty.
\end{aligned} \tag{2.19}$$

By noting that (2.10), it is easily seen that

$$B_3 = 0. \tag{2.20}$$

Combining (2.18), (2.19), and (2.20), the proposition is proved. \square

Our main result follows from the propositions using the triangle inequality.

Acknowledgments

The author thanks the referees for pointing out some errors in a previous version, as well as for several comments that have led to improvements in this work. Thanks are also due to Doctor Ke-ang Fu of Zhejiang University in China for his valuable suggestion in the preparation of this paper.

References

- [1] J. A. Davis, "Convergence rates for probabilities of moderate deviations," *Annals of Mathematical Statistics*, vol. 39, pp. 2016–2028, 1968.
- [2] A. Gut and A. Spătaru, "Precise asymptotics in the law of the iterated logarithm," *The Annals of Probability*, vol. 28, no. 4, pp. 1870–1883, 2000.
- [3] Y. S. Chow, "On the rate of moment convergence of sample sums and extremes," *Bulletin of the Institute of Mathematics. Academia Sinica*, vol. 16, no. 3, pp. 177–201, 1988.
- [4] V. Bentkus and F. Götze, "The Berry-Esseen bound for Student's statistic," *The Annals of Probability*, vol. 24, no. 1, pp. 491–503, 1996.
- [5] Q. Wang and B.-Y. Jing, "An exponential nonuniform Berry-Esseen bound for self-normalized sums," *The Annals of Probability*, vol. 27, no. 4, pp. 2068–2088, 1999.
- [6] E. Giné, F. Götze, and D. M. Mason, "When is the student t -statistic asymptotically standard normal?" *The Annals of Probability*, vol. 25, no. 3, pp. 1514–1531, 1997.
- [7] Q.-M. Shao, "Self-normalized large deviations," *The Annals of Probability*, vol. 25, no. 1, pp. 285–328, 1997.

- [8] M. Csörgő, B. Szyszkowicz, and Q. Wang, "Darling-Erdős theorem for self-normalized sums," *The Annals of Probability*, vol. 31, no. 2, pp. 676–692, 2003.
- [9] M. Csörgő, B. Szyszkowicz, and Q. Wang, "Donsker's theorem for self-normalized partial sums processes," *The Annals of Probability*, vol. 31, no. 3, pp. 1228–1240, 2003.