

Nonuniform Berry-Esseen bound for self-normalized series

Nguyen Chi Dzung · Pham Viet Hung

Abstract In this paper, we shall obtain nonuniform Berry-Esseen bounds in the central limit theorem for self-normalized series. We establish the exponential Berry-Esseen bounds for the probability of the self-normalized series under the condition that the third moment is finite.

Keywords Nonuniform bound, Berry-Esseen inequality, Random power series, Self-normalized series

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1 Introduction

Let X_1, X_2, \dots be a sequence of independent random variables with $\mathbb{E}X_i = 0$ and $0 < \mathbb{E}X_i^2 < \infty$ for $i \geq 1$. Let $b \in (0, 1)$ be the discount factor. The random power series S_b can be defined as

$$S_b = X_0 + bX_1 + b^2X_2 + \dots$$

From the financial point of view, X_i stands for the (random) money that we will get at i -th year of a contract, for example a coupon bond and S_b is the present value of the cash flow. In the literature, S_b is also called the perpetuities (see [2], [9], [10]) for more detail.

The study of this quantity has drawn much of interest and it has a long history for more than 50 years. Let us mention some remarkable results for the simplest case that the random variables X_i 's are independent, identically distribution (i.i.d). In 1971, Gerber [8] provided a Berry-Esseen bound for the following central limit theorem as $b \rightarrow 1^-$,

$$\sqrt{1 - b^2}S_b \xrightarrow{d} \mathcal{N}(0, \sigma^2),$$

where $\sigma^2 = \mathbb{E}X_1^2$.

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In 1974, Lai [13] proved the strong law of large number

$$\frac{X_b}{1-b} \xrightarrow{a.s.} 0.$$

The law of iterated logarithm

$$\limsup_{b \rightarrow 1^-} \sqrt{\frac{1-b^2}{\log \log(1/(1-b^2))}} S_b = \sqrt{2}\sigma$$

was proved by Gaposhkin [7] in 1965.

Recently, Iksanov consider a generalization with stochastic discount rates and provide the analogue versions of the above results.

In this paper, we are interested in the self-normalized series (denote by S_b/V_b)

$$\frac{X_0 + bX_1 + b^2X_2 + \dots}{\sqrt{X_0^2 + b^2X_1^2 + b^4X_2^2 + \dots}}$$

with $V_b^2 = X_0^2 + b^2X_1^2 + b^4X_2^2 + \dots$

Self-normalized series can be seen as an extension of self-normalized sum, defined as

$$\frac{S_n}{V_n} = \frac{X_1 + \dots + X_n}{\sqrt{X_1^2 + \dots + X_n^2}},$$

where again the random variables X_i 's are i.i.d with mean zero and finite variance.

The self-normalized sum is also an attractive research direction both in Probability and Statistics, (see [15, Self-normalized limit theorem: A survey, Probability Surveys]) for more detail.

From the distribution of S_n/V_n , one can make a suitable change of variable to deduce the distribution of the classical Student t -statistics and also the studentized t -statistics. This research direction has been studied extensively with many interesting and nice results: Nonuniform Berry-Esseen bound [12, Wang and Jing], Cramér type large (moderate) deviation [11, Jing, Shao and Wang], the law of iterated logarithm [11, Jing, Shao and Wang], Donsker type functional central limit theorem [3, Csörgő, Szyszkowicz and Wang]. It is also interesting to consider some questions for the self-normalized series model. In 2006, Fu and Huang [6] confirmed the self-normalized law of iterated logarithm.

The purpose of this paper is to establish a nonuniform Berry-Esseen bound for the self-normalized series S_b/V_b . In other words, we wish to obtain a bound for

$$\delta_b(x) := |\mathbb{P}(S_b/V_b \leq x) - \Phi(x)|.$$

Our main result is the following theorem.

Theorem 1 *Let X_1, X_2, \dots be a sequence of independent, symmetric random variables with $\mathbb{E}(|X_j|^3) < \infty$ for all $j = 1, 2, \dots$. Set $B_b^2 = \sum_{j=1}^{\infty} b^{2j} \mathbb{E}X_j^2$, $L_{3b} = B_b^{-3} \sum_{j=1}^{\infty} b^{3j} \mathbb{E}|X_j|^3$.*

(i) *If $|x| \leq (5L_{3b}^{1/3})^{-1}$, we have*

$$\delta_b(x) \leq A \left((1+x^2)L_{3b} + \sum_{j=1}^{\infty} \mathbb{P}(|b^j X_j| > B_b/(6|x|)) \right) \exp\left(-\frac{x^2}{2}\right). \quad (1.1)$$

(ii) If $|x| > (5L_{3b}^{1/3})^{-1}$, we have

$$\delta_b(x) \leq \left(1 + \frac{1}{\sqrt{2\pi}|x|}\right) \exp\left(-\frac{x^2}{2}\right). \quad (1.2)$$

Under the assumption $\mathbb{E}|X_j|^3 < \infty$, by applying Markov's inequality for $|x| \leq (5L_{3b}^{1/3})^{-1}$, we obtain the following corollary.

Corollary 1 *Let X_1, X_2, \dots be a sequence of independent, symmetric random variables with $\mathbb{E}(|X_j|^3) < \infty$, for all $j = 1, 2, \dots$. Then for all $x \in \mathbb{R}$, we have*

$$\delta_b(x) \leq A \min\{(1 + |x|^3)L_{3b}, 1\} \exp\left(-\frac{x^2}{2}\right).$$

Before proving the main result, we need some the following technical lemmas.

2 Some technical lemmas

Lemma 1 *Let X_1, X_2, \dots be a sequence of independent, symmetric random variables with $\mathbb{E}(|X_n|^3) < \infty$ for all $n = 1, 2, \dots$*

(i) *For all $n \geq 1$ and $x > 0$ such that $(1 + x^3)L_{3b} \leq \frac{1}{125}$, we have*

$$\mathbb{P}(S_b > x(V_b^2 + B_b^2)/(2B_b)) = (1 - \Phi(x)) \exp(r_{1b}(x)) + \exp\left(\frac{-x^2}{2}\right) r_{2b}(x), \quad (2.1)$$

where $|r_{1b}(x)| \leq 14x^3L_{3b}$ and $|r_{2b}(x)| \leq A(1 + x^2)L_{3b} \exp(14x^3L_{3b})$.

(ii) *For $n \geq 1$ and $x \geq 1$ satisfying $x^3L_{3b} \leq \frac{1}{125}$, we have*

$$\mathbb{P}(S_b > x(V_b^2 + B_b^2)/(2B_b)) = (1 - \Phi(x))(1 + r_{3b}(x)), \quad (2.2)$$

where $|r_{3b}(x)| \leq Ax(1 + x^2)L_{3b} \exp(14x^3L_{3b})$.

Proof Set

$$h = \frac{x}{B_b}, \quad \eta_j = b^j X_j - \frac{h}{2}(b^{2j} X_j^2 - b^{2j} \sigma_j^2).$$

Then the left-hand side of (2.1) is equivalent to

$$\mathbb{P}(S_b > x(V_b^2 + B_b^2)/(2B_b)) = \mathbb{P}\left(\sum_{j=1}^{\infty} \eta_j > xB_b\right). \quad (2.3)$$

Next, we apply the conjugate method which was first introduced by Esscher [4] and improved by Feller [5]. Let ξ_1, ξ_2, \dots be independent random variables with ξ_i having distribution function defined by

$$V_j(u) = \mathbb{E}(\exp(h\eta_j)\mathbf{1}(\eta_j \leq u))/\mathbb{E}(\exp(h\eta_j)) \quad \text{for } j = 1, 2, \dots$$

We also define

$$M_b^2(h) = \sum_{j=1}^{\infty} \text{Var}(\xi_j)$$

and

$$G_b(t) = \mathbb{P} \left(\frac{\sum_{j=1}^{\infty} (\xi_j - \mathbb{E}\xi_j)}{M_b(h)} \leq t \right), \quad R_b(h) = \frac{x B_b - \sum_{j=1}^{\infty} \mathbb{E}\xi_j}{M_b(h)}.$$

By the well-known equation $\int_0^{\infty} \exp(-sx) d\Phi(x) = \exp\left(-\frac{x^2}{2}\right) (1 - \Phi(s))$ and using inverse Laplace transform, we have

$$\begin{aligned} \mathbb{P} \left(\sum_{j=1}^{\infty} \eta_j > x B_b \right) &= \left(\prod_{j=1}^{\infty} \mathbb{E} \exp(h\eta_j) \right) \int_{x B_b}^{\infty} \exp(-hu) d\mathbb{P} \left(\sum_{j=1}^{\infty} \xi_j \leq u \right) \\ &= \left(\prod_{j=1}^{\infty} \mathbb{E} \exp(h\eta_j) \right) \int_0^{\infty} \exp(-hxB_b - hM_b(h)v) dG_b(v + R_b(h)) \\ &= \left(\prod_{j=1}^{\infty} \mathbb{E} \exp(h\eta_j) \right) e^{-x^2} \left(\int_0^{\infty} \exp(-hM_b(h)v) d(G_b(v + R_b(h)) - \Phi(v)) \right. \\ &\quad \left. + \int_0^{\infty} \exp(-hM_b(h)v) d\Phi(v) \right) \\ &= I_0(h) \exp(-x^2) \left(\exp\left(\frac{x^2}{2}\right) (1 - \Phi(x)) + I_1(h) + I_2(h) + I_3(h) \right), \end{aligned} \quad (2.4)$$

where

$$\begin{aligned} I_0(h) &= \prod_{j=1}^{\infty} \mathbb{E} \exp(h\eta_j), \\ I_1(h) &= \int_0^{\infty} \exp(-hM_b(h)v) d(G_b(v + R_b(h)) - \Phi(v + R_b(h))), \\ I_2(h) &= \int_0^{\infty} \exp(-hM_b(h)v) d(\Phi(v + R_b(h)) - \Phi(v)), \\ I_3(h) &= \int_0^{\infty} \exp(-hM_b(h)v - \exp(-xv)) d\Phi(v). \end{aligned}$$

We will establish some inequalities before estimating $I_j(h)$ for $j = 1, 2, 3$.

It follows from Jensen's inequality that $\sigma_j^3 \leq \mathbb{E}|X_j|^3$.

Combining this and the assumption $(1 + x^3)L_{3b} \leq \frac{1}{125}$, we have

$$b^j \sigma_j h = \frac{b^j \sigma_j x}{B_b} \leq (x^3 B_b^{-3} b^{3j} \mathbb{E}|X_j|^3)^{1/3} \leq \frac{1}{5}. \quad (2.5)$$

Thus

$$h\eta_j = -\frac{1}{2}h^2(b^j X_j - h^{-1})^2 + \frac{1}{2} + \frac{1}{2}b^{2j}\sigma_j^2 h^2 \leq \frac{13}{25}. \quad (2.6)$$

From (2.5), the symmetry assumption and $\mathbb{E}|X_j|^3 < \infty$, we have

$$\begin{aligned}
|\mathbb{E}(\eta_j \mathbf{1}(|b^j X_j| \leq h^{-1}))| &= \left| \mathbb{E} \left(\left(b^j X_j - \frac{h}{2}(b^{2j} X_j^2 - b^{2j} \sigma_j^2) \right) \mathbf{1}(|b^j X_j| > h^{-1}) \right) \right| \\
&\leq \mathbb{E}(|b^j X_j| \mathbf{1}(|b^j X_j| > h^{-1})) + \frac{h}{2} \mathbb{E}((b^j X_j)^2 \mathbf{1}(|b^j X_j| > h^{-1})) \\
&\quad + \frac{h}{2} \mathbb{E}((b^j \sigma_j)^2 \mathbf{1}(|b^j X_j| > h^{-1})) \\
&\leq h^2 \mathbb{E}(|b^j X_j|^3 \mathbf{1}(|b^j X_j| > h^{-1})) + \frac{h^2}{2} \mathbb{E}(|b^j X_j|^3 \mathbf{1}(|b^j X_j| > h^{-1})) \\
&\quad + \frac{h^2}{2} \mathbb{E}(|b^j \sigma_j|^3 b^{2j} \sigma_j^2 h^2 \mathbf{1}(|b^j X_j| > h^{-1})) \\
&\leq 2h^2 \mathbb{E}(|b^j X_j|^3 \mathbf{1}(|b^j X_j| > h^{-1})).
\end{aligned} \tag{2.7}$$

Similarly, we also have

$$|\mathbb{E}(\eta_j^2 \mathbf{1}(|b^j X_j| \leq h^{-1}) - b^{2j} \sigma_j^2)| \leq \frac{3}{2} h (\mathbb{E}|b^j X_j|^3 + h b^{4j} \sigma_j^4), \tag{2.8}$$

$$\mathbb{E}(|\eta_j|^3 \mathbf{1}(|b^j X_j| \leq h^{-1})) \leq 6 \mathbb{E}(|b^j X_j|^3 \mathbf{1}(|b^j X_j| \leq h^{-1})) + 2h^3 b^{6j} \sigma_j^6. \tag{2.9}$$

We have

$$\begin{aligned}
\mathbb{E} \exp(n\eta_j) &= \mathbb{E}(\exp(h\eta_j) \mathbf{1}(|b^j X_j| \leq h^{-1})) + \mathbb{E}(\exp(h\eta_j) \mathbf{1}(|b^j X_j| > h^{-1})) \\
&= \mathbb{E} \left(\left(1 + h\eta_j + \frac{1}{2}(h\eta_j)^2 \right) \mathbf{1}(|b^j X_j| \leq h^{-1}) \right) + \mathbb{E}(\exp(h\eta_j) \mathbf{1}(|b^j X_j| > h^{-1})) \\
&\quad + \mathbb{E} \left(\left(\exp(h\eta_j) - 1 - h\eta_j - \frac{1}{2}(h\eta_j)^2 \right) \mathbf{1}(|b^j X_j| \leq h^{-1}) \right) \\
&= 1 + \frac{1}{2} h^2 b^{2j} \sigma_j^2 + l_{1j}(h) \\
&= \exp \left(\frac{1}{2} h^2 b^{2j} \sigma_j^2 + l_{2j}(h) \right),
\end{aligned} \tag{2.10}$$

$$(\mathbb{E} \exp(h\eta_j))^{-1} = 1 - \frac{1}{2} h^2 b^{2j} \sigma_j^2 + l_{3j}(h), \tag{2.11}$$

where

$$\begin{aligned}
l_{1j}(h) &= -\mathbb{P}(|b^j X_j| > h^{-1}) + h \mathbb{E}(\eta_j \mathbf{1}(|b^j X_j| \leq h^{-1})) + \frac{1}{2} h^2 \mathbb{E}(\eta_j^2 \mathbf{1}(|b^j X_j| \leq h^{-1}) - b^{2j} \sigma_j^2) \\
&\quad + \mathbb{E}(\exp(h\eta_j) \mathbf{1}(|b^j X_j| > h^{-1})) + \mathbb{E} \left(\left(\exp(h\eta_j) - 1 - h\eta_j - \frac{1}{2}(h\eta_j)^2 \right) \mathbf{1}(|b^j X_j| \leq h^{-1}) \right).
\end{aligned}$$

Applying the elementary inequality $\left|e^x - 1 - x - \frac{x^2}{2}\right| \leq \frac{|x|^3 e^{|x|}}{6}$ for all $x \in \mathbb{R}$, and noting that $\exp(xh\eta_j) \leq 2$ for $0 \leq x \leq 1$, we get

$$\begin{aligned} |l_{1j}(h)| &\leq h\mathbb{E}(\eta_j \mathbf{1}(|b^j X_j| \leq h^{-1})) + \frac{1}{2}h^2|\mathbb{E}(\eta_j^2 \mathbf{1}(|b^j X_j| \leq h^{-1}) - b^{2j}\sigma_j^2)| \\ &\quad + \frac{1}{3}h^3\mathbb{E}(|\eta_j|^3 \mathbf{1}(|b^j X_j| \leq h^{-1})) + 3\mathbb{P}(|b^j X_j| > h^{-1}) \\ &\leq 2h^3\mathbb{E}(|\eta_j|^3 \mathbf{1}(|b^j X_j| > h^{-1})) + \frac{3}{4}h^3(\mathbb{E}|b^j X_j|^3 + hb^{4j}\sigma_j^4) \\ &\quad + \frac{1}{3}h^3 6\mathbb{E}(|\eta_j|^3 \mathbf{1}(|b^j X_j| \leq h^{-1})) + 2h^3 b^{6j}\sigma_j^6 \\ &\leq 7h^3\mathbb{E}|b^j X_j|^3, \end{aligned}$$

$$\begin{aligned} |l_{2j}(h)| &\leq 2|l_{1j}(h)| \leq 14h^3\mathbb{E}|b^j X_j|^3, \\ |l_{3j}(h)| &\leq 2|l_{1j}(h)| \leq 14h^3\mathbb{E}|b^j X_j|^3. \end{aligned}$$

It is proved by Wang and Jing [12] that

$$|\mathbb{E}(\eta_j \exp(h\eta_j)) - hb^{2j}\sigma_j^2| \leq 16h^2\mathbb{E}|b^j X_j|^3, \quad (2.12)$$

$$|\mathbb{E}(\eta_j^2 \exp(h\eta_j)) - b^{2j}\sigma_j^2| \leq 30h\mathbb{E}|b^j X_j|^3, \quad (2.13)$$

$$\mathbb{E}(|\eta_j|^3 \exp(h\eta_j)) \leq 30\mathbb{E}|b^j X_j|^3. \quad (2.14)$$

It follows from (2.5)–(2.14) that

$$\mathbb{E}\xi_j = \frac{\mathbb{E}(\eta_j \exp(h\eta_j))}{\mathbb{E}(\exp(h\eta_j))} = hb^{2j}\sigma_j^2 + l_{4j}(h), \quad (2.15)$$

where

$$l_{4j}(h) = \left(\frac{1}{\mathbb{E}(\exp(h\eta_j))} - 1\right) \mathbb{E}(\eta_j \exp(h\eta_j)) + \mathbb{E}(\eta_j \exp(h\eta_j)) - hb^{2j}\sigma_j^2.$$

Thus, by (2.5), (2.11), (2.12), we get

$$\begin{aligned} |l_{4j}(h)| &\leq \left|\left(\frac{1}{\mathbb{E}(\exp(h\eta_j))} - 1\right) \mathbb{E}(\eta_j \exp(h\eta_j))\right| + |\mathbb{E}(\eta_j \exp(h\eta_j)) - hb^{2j}\sigma_j^2| \\ &\leq |l_{3j}(h) - \frac{1}{2}h^2 b^{2j}\sigma_j^2|(hb^{2j}\sigma_j^2 + 16h^2 b^{3j}\mathbb{E}|X_j|^3) + 16h^2 b^{3j}\mathbb{E}|X_j|^3 \\ &\leq 22h^4 b^{5j}\sigma_j^2 \mathbb{E}|X_j|^3 + 224h^5 b^{6j}(\mathbb{E}|X_j|^3)^2 + 16h^2 b^{3j}\mathbb{E}|X_j|^3 + \frac{1}{2}h^3 b^{4j}\sigma_j^4 \\ &\leq 20h^2 b^{3j}\mathbb{E}|X_j|^3. \end{aligned}$$

Similarly, we also have

$$\text{Var}(\xi_j) = \frac{\mathbb{E}(\eta_j^2 \exp(h\eta_j))}{(\mathbb{E} \exp(h\eta_j))^2} - (\mathbb{E}\xi_j)^2 = b^{2j}\sigma_j^2 + l_{5j}(h), \quad (2.16)$$

and

$$\mathbb{E}|\xi_j|^3 = \mathbb{E}(|\eta|^3 \exp(h\eta_j))/\mathbb{E} \exp(h\eta_j) \leq 34b^{3j}\mathbb{E}|X_j|^3, \quad (2.17)$$

where $|l_{5j}(h)| \leq 41hb^{3j}\mathbb{E}|X_j|^3$.

It follows $|l_{5j}(h)| \leq 41hb^{3j}\mathbb{E}|X_j|^3$ and the assumption $(1+x^3)L_{3b} \leq 1/125$, it is easy to obtain that

$$M_b^2(h) = B_b^2 + \sum_{j=1}^{\infty} l_{5j}(h) > \frac{2}{3}B_b^2. \quad (2.18)$$

We are now estimate $I_j(h)$, $0 \leq j \leq 3$. For $I_0(h)$, using (2.10) we have

$$I_0(h) = \exp\left(\frac{1}{2}h^2B_b^2 + \sum_{j=1}^{\infty} l_{2j}(h)\right) = \exp\left(\frac{x^2}{2}\right) \exp\left(\sum_{j=1}^{\infty} l_{2j}(h)\right). \quad (2.19)$$

By (2.15)–(2.19), the Berry-Esseen bound and Taylor expansion, we have

$$I_1(h) \leq \sup_x |G_n(v) - \Phi(v)| \leq \frac{A}{M_b^3(h)} \sum_{j=1}^{\infty} \mathbb{E}|\xi_j - \mathbb{E}\xi_j|^3 \leq AL_{3b}, \quad (2.20)$$

$$I_2(h) \leq \sup_x |\Phi(v + R_n(h)) - \Phi(v)| \leq \frac{A}{M_b(h)} \sum_{j=1}^{\infty} |l_{4j}(h)| \leq Ax^2L_{3b}. \quad (2.21)$$

By applying the mean value estimate to $I_3(h)$ [see Petrov [14], page 227], we have

$$\begin{aligned} I_3(h) &\leq \frac{1}{x} \left| \frac{M_b(h)}{B_b} - 1 \right| \max \left\{ 1, \frac{B_b^2}{M_b^2} \right\} \\ &\leq \frac{3}{2x} \left| \frac{M_b^2(h) - B_b^2}{B_b(M_b(h) + B_b)} \right| \\ &\leq AL_{3b}. \end{aligned} \quad (2.22)$$

Combining (2.3), (2.4) and (2.19)–(2.22), we get

$$\begin{aligned} \mathbb{P}(S_b > x(V_b^2 + B_b^2)/(2B_b)) &= \exp\left(-\frac{x^2}{2}\right) \exp(r_{1b}(x)) \left(\exp\left(\frac{x^2}{2}\right) (1 - \Phi(x)) + I_1(h) + I_2(h) + I_3(h) \right) \\ &= (1 - \Phi(x)) \exp(r_{1b}(x)) + \exp\left(-\frac{x^2}{2}\right) r_{2b}(x), \end{aligned}$$

where $r_{1b}(x) = \sum_{j=1}^{\infty} l_{2j}(h)$ and $r_{2b}(x) = \exp(r_{1b}(x))(I_1(h) + I_2(h) + I_3(h))$.

Thus

$$|r_{1b}(x)| = \sum_{j=1}^{\infty} |l_{2j}(h)| \leq 14x^3L_{3b},$$

and

$$|r_{2b}(x)| = \exp(|r_{1b}(x)|)(|I_1(h)| + |I_2(h)| + |I_3(h)|) \leq A(1+x^2)L_{3b} \exp(14x^3L_{3b}).$$

We thus have completed the proof of Lemma 2.1(i).

(ii) By the same proof as part (i), we also see that in the case $n \geq 1$ and $x \geq 1$ satisfying $x^3L_{3b} \leq 1/125$, (2.1) holds. Hence, the proof of (2.2) obtains by using the inequalities $e^t \leq 1 + te^t$ and $1 - \Phi(t) \leq \frac{1}{\sqrt{2\pi t}} \exp\left(-\frac{t^2}{2}\right)$ for $t > 0$.

Lemma 2 Let $\varepsilon_1, \varepsilon_2, \dots$ be i.i.d. Rademacher random variables, that is, $\mathbb{P}(\varepsilon_j = \pm 1) = 1/2$. Then for any $x \geq 1$ and any sequence a_1, a_2, \dots satisfying $|a_j| \leq B_b/(6x)$ and $\sum_{n=1}^{\infty} a_j^2 > \frac{4}{9}B_b^2$, we have

$$\mathbb{P}\left(\sum_{j=1}^{\infty} a_j \varepsilon_j > x \left(\sum_{j=1}^{\infty} a_j^2\right)^{1/2}\right) \leq (1 - \Phi(x)) (1 + Ax(1 + x^2)L_{3b}^* \exp(2x^3 L_{3b}^*)), \quad (2.23)$$

where $L_{3b}^* = B_b^{-3} \sum_{j=1}^{\infty} |a_j|^3$.

The proof of Lemma 2 follows very similar lines to those of Lemma 1, so we omit it.

Lemma 3 Let $\{a_i, i \geq 1\}$ be any sequence of real numbers. Set $A_{\infty}^2 = \sum_{i=1}^{\infty} a_i^2$ and $T_{\infty} = \sum_{i=1}^{\infty} \varepsilon_i a_i$. Then, for all $x > 0$

$$\mathbb{P}(T_{\infty} > x A_{\infty}) \leq \exp\left(-\frac{x^2}{2}\right).$$

where $\{\varepsilon_i, i \geq 1\}$ is i.i.d Rademacher random variables.

Proof For any $u > 0$, by Markov's inequality, we have

$$\begin{aligned} \mathbb{P}(T_{\infty} > x A_{\infty}) &\leq \exp(-ux A_{\infty}) \mathbb{E} \exp(u T_{\infty}) = \exp(-ux A_{\infty}) \prod_{i=1}^{\infty} \cosh(ua_i) \\ &\leq \exp(-ux A_{\infty}) \prod_{i=1}^{\infty} \exp\left(\frac{u^2 a_i^2}{2}\right) = \exp(-ux A_{\infty} + u^2 A_{\infty}^2/2). \end{aligned}$$

The proof of Lemma 3 is complete by choosing $u = -xA_{\infty}^2$.

Lemma 4 Let X_1, X_2, \dots be independent, symmetric random variables. Then for any $x \geq 0$ and $n \geq 1$, we have

$$\mathbb{P}(S_b > x V_b) \leq \exp\left(-\frac{x^2}{2}\right).$$

Proof Similarly, in [12, Lemma 43], we assume that $\{X_j, j \geq 1\}$ are defined on a probability space $(\Omega, \mathcal{F}, \mathbb{P})$ which also supports a sequence of independent Rademacher random variables $\{\varepsilon_j, j \geq 1\}$ independent of $\{X_j, j \geq 1\}$. By the symmetry of X_j and independence X_j and ε_j , we have that

$$\begin{aligned} \mathbb{P}(S_b > x V_b) &= \mathbb{P}\left(\sum_{j=1}^{\infty} X_j \varepsilon_j > x V_b\right) \\ &= \int \int \dots \mathbb{P}\left(\sum_{j=1}^{\infty} x_j \varepsilon_j > x \left(\sum_{j=1}^{\infty} x_j^2\right)^{1/2}\right) dF_1(x_1) dF_2(x_2) \dots \\ &\leq \exp\left(-\frac{x^2}{2}\right) \text{ (by Lemma 3).} \end{aligned}$$

This ends the proof of the lemma.

Lemma 5 *Let X_1, X_2, \dots be independent random variables. Then for $x \geq 1, y \geq 0$ and $k \geq 1$, we have*

$$\mathbb{P}(S_b > xV_b, |b^k X_k| > y) \leq \mathbb{P}(|b^k X_k| > y) \mathbb{P} \left(\sum_{j \neq k, j=1}^{\infty} b^j X_j > (x^2 - 1)^{1/2} \left(\sum_{j \neq k, j=1}^{\infty} b^{2j} X_j^2 \right)^{1/2} \right).$$

Proof We observe that, for any real number a , then

$$\begin{aligned} ab^k X_k - \frac{x}{2}(b^{2k} X_k^2 + a^2) &= -\frac{x}{2} \left(b^k X_k - \frac{a}{x} \right)^2 - \frac{xa^2}{2} + \frac{a^2}{2x} \\ &\leq \frac{a^2}{2} \left(\frac{1}{x} - x \right). \end{aligned}$$

Thus, we get that

$$\begin{aligned} \mathbb{P}(S_b > xV_b, |b^k X_k| > y) &= \mathbb{P} \left(S_b > \inf_{a>0} \frac{x}{2a} (V_b^2 + a^2), |b^k X_k| > y \right) \\ &= \mathbb{P} \left(\sup_{a>0} \left(\sum_{j=1}^{\infty} (ab^j X_j - \frac{x}{2} b^{2j} X_j^2) - \frac{x}{2} a^2 \right) > 0, |b^k X_k| > y \right) \\ &= \mathbb{P} \left(\sup_{a>0} \left(\sum_{j \neq k, j=1}^{\infty} (ab^j X_j - \frac{x}{2} b^{2j} X_j^2) + ab^k X_k - \frac{x}{2} b^{2k} X_k^2 - \frac{x}{2} a^2 \right) > 0, |b^k X_k| > y \right) \\ &\leq \mathbb{P} \left(\sup_{a>0} \left(\sum_{j \neq k, j=1}^{\infty} (ab^j X_j - \frac{x}{2} b^{2j} X_j^2) + \frac{a^2}{2} \left(\frac{1}{x} - x \right) \right) > 0, |b^k X_k| > y \right) \\ &= \mathbb{P} \left(\sum_{j \neq k, j=1}^{\infty} b^j X_j > \inf_{a>0} \frac{x}{2a} \left(\sum_{j \neq k, j=1}^{\infty} b^{2j} X_j^2 + a^2 \left(1 - \frac{1}{x^2} \right) \right), |b^k X_k| > y \right) \\ &= \mathbb{P} \left(\sum_{j \neq k, j=1}^{\infty} b^j X_j > (x^2 - 1)^{1/2} \left(\sum_{j=1, j \neq k}^{\infty} b^{2j} X_j^2 \right)^{1/2}, |b^k X_k| > y \right) \\ &= \mathbb{P}(|b^k X_k| > y) \mathbb{P} \left(\sum_{j \neq k, j=1}^{\infty} b^j X_j > (x^2 - 1)^{1/2} \left(\sum_{j \neq k, j=1}^{\infty} b^{2j} X_j^2 \right)^{1/2} \right). \end{aligned} \tag{2.24}$$

The lemma is proved.

We are now ready to prove Theorem 1. To bound $\delta_b(x)$, it suffices to consider $x > 0$ since we can simply apply result to $-X_j$ when $x < 0$. For $0 < x \leq 1$, (1.1) was proved by Bentkus, Bloznelis and Götze [1]. We consider the case where $1 \leq x \leq (5L_{3b}^{1/3})^{-1}$. By applying the elementary inequality $2B_b V_b \leq B_b^2 + V_b^2$ and Lemma 1(ii), we have

$$\begin{aligned} \mathbb{P}(S_b > xV_b) &\geq \mathbb{P}(2B_b S_b > x(V_b^2 + S_b^2)) \\ &\geq (1 - \Phi(x))(1 - Ax(1 + x^2)L_{3b}). \end{aligned} \tag{2.25}$$

Combining this and the inequality $1 - \Phi(x) \leq \frac{1}{\sqrt{2\pi}x} \exp\left(-\frac{x^2}{2}\right)$ for $x \geq 0$ implies that

$$\mathbb{P}(S_b \leq xV_b) - \Phi(x) \leq (1 - \Phi(x))Ax(1 + x^2)L_{3b} \leq A(1 + x^2)L_{3b} \exp\left(-\frac{x^2}{2}\right).$$

Hence, to prove (1.1), it suffices to show that

$$\mathbb{P}(S_b > xV_b) \leq (1 - \Phi(x)) + A \left((1 + x^2)L_{3b} + \sum_{j=1}^{\infty} \mathbb{P}(|b^j X_j| > B_b/(6x)) \exp\left(-\frac{x^2}{2}\right) \right). \quad (2.26)$$

Set

$$Y_j = b^j X_j \mathbf{1}(|b^j X_j| \leq B_b/(6x)), \quad S_b^* = \sum_{j=1}^{\infty} Y_j, \quad V_b^{*2} = \sum_{j=1}^{\infty} Y_j^2.$$

For $x \geq 1$, it follows from Lemma 4 and Lemma 5 that

$$\begin{aligned} \mathbb{P}(S_b > xV_b) &\leq \sum_{k=1}^{\infty} \mathbb{P}\left(S_b > xV_b, |b^k X_k| > \frac{B_b}{6x}\right) + \mathbb{P}(S_b^* > xV_b^*) \\ &\leq \sum_{k=1}^{\infty} \mathbb{P}\left(|b^k X_k| > \frac{B_b}{6x}\right) \mathbb{P}\left(\sum_{j \neq k, j=1}^{\infty} b^j X_j > (x^2 - 1)^{1/2} \left(\sum_{j \neq k, j=1}^{\infty} b^{2j} X_j^2\right)^{1/2}\right) + \mathbb{P}(S_b^* > xV_b^*) \\ &\leq \sum_{k=1}^{\infty} \mathbb{P}\left(|b^k X_k| > \frac{B_b}{6x}\right) \exp\left(-\frac{x^2 - 1}{2}\right) + \mathbb{P}(S_b^* > xV_b^*) \\ &\leq e \sum_{k=1}^{\infty} \mathbb{P}\left(|b^k X_k| > \frac{B_b}{6x}\right) \exp\left(-\frac{x^2}{2}\right) + \mathbb{P}(S_b^* > xV_b^*). \end{aligned} \quad (2.27)$$

Denote $F_j(x)$ to be distribution function of X_j for $j \geq 1$. Using the assumption $|x| \leq (5L_{3b}^{1/3})^{-1}$, the inequalities $e^t \leq 1 + te^t$ and $e^t \geq 1 + t$ for $t \geq 0$, we have for $i \geq 1$,

$$\begin{aligned} \prod_{j \neq i, j=1}^{\infty} \mathbb{E}(\exp(2x^3 B_b^{-3} |Y_j|^3)) &\leq \prod_{j \neq i, j=1}^{\infty} \left(1 + 2x^3 B_b^{-3} \mathbb{E}|Y_j|^3 e^{2/125}\right) \\ &\leq \prod_{j \neq i, j=1}^{\infty} \exp(4x^3 B_b^{-3} \mathbb{E}|Y_j|^3) \\ &\leq \exp(4x^3 L_{3b}) \\ &\leq 2. \end{aligned} \quad (2.28)$$

We will now bound $\mathbb{P}(S_b^* > xV_b^*)$. Similarly in the proof of Lemma 3, we assume that $\{Y_j, j \geq 1\}$ are defined on a probability space $(\Omega, \mathcal{F}, \mathbb{P})$ which also supports a sequence of independent Rademacher

random variables $\{\varepsilon_j, j \geq 1\}$ independent of the sequence $\{Y_j, j \geq 1\}$. By symmetry of X_j , we have that

$$\begin{aligned} \mathbb{P}(S_b^* > xV_b^*) &= \mathbb{P}\left(\sum_{j=1}^{\infty} Y_j \varepsilon_j > xV_b^*\right) \\ &\leq \mathbb{P}\left(\sum_{j=1}^{\infty} Y_j \varepsilon_j > xV_b^*, V_b^{*2} > \frac{4}{9}B_b^2\right) + \mathbb{P}\left(V_b^{*2} \leq \frac{4}{9}B_b^2\right). \end{aligned} \quad (2.29)$$

It follows from (2.28) and Lemma 2 that

$$\begin{aligned} &\mathbb{P}\left(\sum_{j=1}^{\infty} Y_j \varepsilon_j > xV_b^*, V_b^{*2} > \frac{4}{9}B_b^2\right) \\ &= \int \int \dots \int_{\sum_{j=1}^{\infty} y_j^2 > \frac{4}{9}B_b^2, |y_j| \leq B_b/(6x), j=1,2,\dots} \mathbb{P}\left(\sum_{j=1}^{\infty} y_j \varepsilon_j > x \left(\sum_{j=1}^{\infty} y_j^2\right)^{1/2}\right) dF_1(y_1) dF_2(y_2) \dots \\ &\leq (1 - \Phi(x)) \int \int \dots \int_{\dots |y_j| \leq B_b/(6x), j=1,2,\dots} \\ &\quad \times \left(1 + Ax(1+x^2)B_b^{-3} \sum_{i=1}^{\infty} |y_i|^3 \exp(2x^3 B_b^{-3} \sum_{j=1}^{\infty} |y_j|^3)\right) dF_1(y_1) dF_2(y_2) \dots \\ &\leq (1 - \Phi(x)) \left(1 + Ax(1+x^2)B_b^{-3} \sum_{i=1}^{\infty} \mathbb{E}\left(|Y_i|^3 \exp(2x^3 B_b^{-3} \sum_{j=1}^{\infty} |Y_j|^3)\right)\right) \\ &\leq (1 - \Phi(x)) \left(1 + Ax(1+x^2)B_b^{-3} \sum_{i=1}^{\infty} \left(\mathbb{E}|Y_i|^3 \left(\prod_{j \neq i, j=1}^{\infty} \mathbb{E} \exp(2x^3 B_b^{-3} \sum_{j=1}^{\infty} |Y_j|^3)\right)\right)\right) \\ &\leq (1 - \Phi(x)) + A(1+x^2)L_{3b} \exp\left(\frac{-x^2}{2}\right). \end{aligned} \quad (2.30)$$

We will next bound $\mathbb{P}\left(V_b^{*2} \leq \frac{4}{9}B_b^2\right)$. Firstly, we note that

$$\sum_{j=1}^{\infty} \mathbb{E}\left(b^{2j} X_j^2 \mathbf{1}\left(|b^j X_j| > \frac{B_b}{6x}\right)\right) \leq \frac{6x}{B_b} \sum_{j=1}^{\infty} \mathbb{E}|b^j X_j|^3 \leq \frac{6}{125} B_b^2. \quad (2.31)$$

$$\sum_{j=1}^{\infty} \mathbb{E}Y_j^2 = B_b^2 - \sum_{j=1}^{\infty} \mathbb{E}(b^{2j} X_j^2 \mathbf{1}(|b^j X_j| > B_b/(6x))). \quad (2.32)$$

Then for any $t > 0$, by using (2.31), (2.32), the inequalities $1 + |x| \leq e^{|x|}$, $e^x \leq 1 + x + 1/2x^2e^{|x|}$, $\text{Var}(Y_j^2) \leq \mathbb{E}Y_j^4 \leq b^{3j}E|X_j|^3B_b/(6x)$ and Markov's inequality, we have

$$\begin{aligned} \mathbb{P}\left(V_b^{*2} \leq \frac{4}{9}B_b^2\right) &= \mathbb{P}\left(\sum_{j=1}^{\infty}(\mathbb{E}Y_j^2 - Y_j^2) > \frac{5}{9}B_b^2 - \sum_{j=1}^{\infty}\mathbb{E}(b^{2j}X_j^2\mathbf{1}(|b^jX_j| > B_b/(6x)))\right) \\ &\leq \mathbb{P}\left(\sum_{j=1}^{\infty}(\mathbb{E}Y_j^2 - Y_j^2) > \frac{1}{2}B_b^2\right) \\ &\leq e^{-t/2} \prod_{j=1}^{\infty} \mathbb{E} \exp(tB_b^{-2}(\mathbb{E}Y_j^2 - Y_j^2)) \\ &\leq e^{-t/2} \prod_{j=1}^{\infty} \left(1 + \frac{1}{2}t^2B_b^{-4} \text{Var}(Y_j^2) \exp\left(\frac{tx^{-2}}{36}\right)\right) \\ &\leq e^{-t/2} \prod_{j=1}^{\infty} \exp\left(\frac{t^2b^{3j}\mathbb{E}|X_j|^3}{6xB_b^3} \left(\frac{tx^{-2}}{36}\right)\right) \\ &\leq e^{-t/2} \prod_{j=1}^{\infty} \exp\left(\frac{t^2}{6x}L_{3b} \left(\frac{tx^{-2}}{36}\right)\right). \end{aligned}$$

Choosing $t = 4x^2(1 + x^{-2} \log L_{3b}^{-1/2})$, we obtain

$$\mathbb{P}\left(V_b^{*2} \leq \frac{4}{9}B_b^2\right) \leq AL_{3b} \exp\left(\frac{-x^2}{2}\right). \quad (2.33)$$

Combining (2.29), (2.30) and (2.33), we get

$$\mathbb{P}(S_b^* > xV_b^*) \leq (1 - \Phi(x)) + A(1 + x^2)L_{3b} \exp\left(\frac{-x^2}{2}\right). \quad (2.34)$$

Then (2.26) implies from (2.27) and (2.34). The proof of Theorem 1.1 is thus completed.

(ii) The proof of (1.2) implies from the inequality $1 - \Phi(x) \leq \frac{1}{\sqrt{2\pi x}} \exp\left(-\frac{x^2}{2}\right)$ for $x > 0$ and Lemma 4.

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