

Research Article

Precise Asymptotics in the Law of the Iterated Logarithm under Sublinear Expectations

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By an inequality of partial sum and uniform convergence of the central limit theorem under sublinear expectations, we establish precise asymptotics in the law of the iterated logarithm for independent and identically distributed random variables under sublinear expectations.

1. Introduction

Motivated by the work of g -expectation of Peng [1], Peng [2, 3] initiated the concept of the sublinear expectation space, which is a powerful tool to model the uncertainty of probability and distribution. We could consider sublinear expectation as an extension of the classical linear expectation. Peng [2, 3] constructed the basic framework, investigated basic properties, and proved the law of large number and central limit theorem under sublinear expectations. Motivated by the seminal work of Peng [2, 3], more and more limit theorems under sublinear expectation space have been established, which generalize the corresponding fundamental, important limit theorems in probability and statistics. Zhang [4–6] proved the exponential inequalities and Rosenthal's inequalities and obtained an extension of the central limit theorem and Donsker's invariance principle under sublinear expectations. Wu [7] established precise asymptotics for complete integral convergence under sublinear expectations. Yu and Wu [8] studied Marcinkiewicz-type complete convergence for weighted sums under sublinear expectations. Wu and Jiang [9] obtained a strong law of large numbers and Chover's law of the iterated logarithm under sublinear expectations. Ma and Wu [10] studied the limiting behavior of weighted sums of extended negatively dependent random variables under

sublinear expectations. Xu and Zhang [11, 12] studied three series theorem for independent random variables and the law of logarithm for arrays of random variables under sublinear expectations. Chen [13] proved strong laws of large numbers for sublinear expectations. For more results about limit theorems under sublinear expectations, the interested reader could refer to the studies of Hu et al. [14], Fang et al. [15], Kuczmaszewska [16], Wang and Wu [17], Hu and Yang [18], Zhang [19], and references therein.

Precise asymptotics in the law of the iterated logarithm is one of the fundamental problems in probability theory. Many related results have been derived in the probabilistic space. Their results can be found in the work of Gut and Spätaru [20]; Zhang [21]; Xiao et al. [22]; Huang et al. [23]; Jiang and Yang [24]; Wu and Wen [25]; Xu et al. [26]; Xu [27, 28]; and Xu [29]. However, in sublinear expectations, due to the uncertainty of sublinear expectation and related capacity, the precise asymptotics in the law of the iterated logarithm under sublinear expectations have not been reported. Motivated by the work of Wu [7], Xiao et al. [22], Xu et al. [26], and Xu [29], we try to investigate precise asymptotics in the law of the iterated logarithm under sublinear expectations. The aim of this paper is to prove the precise asymptotics in the law of the iterated logarithm for independent, identically distributed random variables under sublinear expectations. The main contribution of this paper

is that we prove an useful inequality under sublinear expectations in Lemma 1, and we extend the results of Xiao et al. [22], Xu et al. [26], and Xu [29] to those of the sublinear expectation spaces. Our results may have the potential applications in finance or engineering fields (cf. Wu [7], Peng [3], Zhang [19], and references therein). Our basic idea in this paper comes from that of Wu [7], Xiao et al. [22], Xu et al. [26], Xu [29], Spătaru [30], and Fuk and Nagaev [31]. In conclusion, our results combined with the work of Wu [7] imply heuristically that many results about precise asymptotics in the law of the iterated logarithm in probability spaces may still hold under sublinear expectations.

The rest of this paper is organized as follows: in Section 2, we summarize necessary basic notions, concepts, and relevant properties and give necessary lemmas under sublinear expectations. In Section 3, we give our main results, Theorems 1 and 2, whose proofs are presented in Sections 4 and 5, respectively.

2. Preliminaries

We use notations similar to those of Peng [3]. Let (Ω, \mathcal{F}) be a given measurable space. Let \mathcal{H} be a subset of all random variables on (Ω, \mathcal{F}) such that $I_A \in \mathcal{H}$, where $A \in \mathcal{F}$, and if $X_1, \dots, X_n \in \mathcal{H}$, then $\varphi(X_1, \dots, X_n) \in \mathcal{H}$ for each $\varphi \in C_{l,\text{Lip}}(\mathbb{R}^n)$, where $C_{l,\text{Lip}}(\mathbb{R}^n)$ denotes the linear space of (local Lipschitz) function φ satisfying

$$|\varphi(\mathbf{x}) - \varphi(\mathbf{y})| \leq C(1 + |\mathbf{x}|^m + |\mathbf{y}|^m)(|\mathbf{x} - \mathbf{y}|), \quad \forall \mathbf{x}, \mathbf{y} \in \mathbb{R}^n, \quad (1)$$

for some $C > 0$, $m \in \mathbb{N}$, depending on φ . We regard \mathcal{H} as the space of random variables.

Definition 1. A sublinear expectation \mathbb{E} on \mathcal{H} is a functional $\mathbb{E}: \mathbb{H} \rightarrow \overline{\mathbb{R}} := [-\infty, \infty]$ satisfying the following properties: for all $X, Y \in \mathcal{H}$, we have the following:

- (a) Monotonicity: if $X \geq Y$, then $\mathbb{E}[X] \geq \mathbb{E}[Y]$
- (b) Constant preserving: $\mathbb{E}[c] = c$, $\forall c \in \mathbb{R}$
- (c) Positive homogeneity: $\mathbb{E}[\lambda X] = \lambda \mathbb{E}[X]$, $\forall \lambda \geq 0$
- (d) Subadditivity: $\mathbb{E}[X + Y] \geq \mathbb{E}[X] + \mathbb{E}[Y]$ whenever $\mathbb{E}[X] + \mathbb{E}[Y]$ is not of the form $\infty - \infty$ or $-\infty + \infty$

A set function $V: \mathcal{F} \rightarrow [0, 1]$ is called a capacity if it satisfies the following:

- (a) $V(\emptyset) = 0$, $V(\Omega_1) = 1$
- (b) $V(A) \leq V(B)$, $A \subset B$, $A, B \in \mathcal{F}$

A capacity V is said to be subadditive if it satisfies $V(A + B) \leq V(A) + V(B)$, $A, B \in \mathcal{F}$.

In this paper, given a sublinear expectation space $(\Omega, \mathcal{H}, \mathbb{E})$, we define a capacity: $\mathbb{V}(A) := \inf\{\mathbb{E}[\xi]: I_A \leq \xi, \xi \in \mathcal{H}\}$, $\forall A \in \mathcal{F}$ (see Zhang [4]). Clearly, \mathbb{V} is a subadditive capacity. We also define the Choquet expectations C_V by

$$C_V(X) := \int_0^\infty \mathbb{V}(X > x) dx + \int_{-\infty}^0 (\mathbb{V}(X > x) - 1) dx. \quad (2)$$

A sublinear expectation $\mathbb{E}: \mathcal{H} \rightarrow \mathbb{R}$ is said to be continuous if it satisfies the following:

- (a) Lower continuity: $\mathbb{E}[X_n] \uparrow \mathbb{E}[X]$, if $0 \leq X_n \uparrow X$, where $X_n, X \in \mathcal{H}$
- (b) Upper continuity: $\mathbb{E}[X_n] \downarrow \mathbb{E}[X]$, if $0 \leq X_n \downarrow X$, where $X_n, X \in \mathcal{H}$

A capacity $V: \mathcal{F} \rightarrow [0, 1]$ is said to be continuous capacity if it satisfies the following:

- (1) Lower continuity: $V(A_n) \uparrow V(A)$, if $A_n \uparrow A$, where $A_n, A \in \mathcal{F}$
- (2) Upper continuity: $V(A_n) \downarrow V(A)$, if $A_n \downarrow A$, where $A_n, A \in \mathcal{F}$

Assume that $\mathbf{X} = (X_1, \dots, X_m)$, $X_i \in \mathcal{H}$, and $\mathbf{Y} = (Y_1, \dots, Y_n)$, $Y_i \in \mathcal{H}$, are two random variables on $(\Omega_1, \mathcal{H}, \mathbb{E})$. \mathbf{Y} is said to be independent of \mathbf{X} if for each $\varphi \in C_{l,\text{Lip}}(\mathbb{R}^m \times \mathbb{R}^n)$, we have $\mathbb{E}[\varphi(\mathbf{X}, \mathbf{Y})] = \mathbb{E}[\varphi(\mathbf{x}, \mathbf{Y})]_{\mathbf{x}=\mathbf{X}}$ whenever $\overline{\varphi}(\mathbf{x}) := \mathbb{E}[|\varphi(\mathbf{x}, \mathbf{Y})|] < \infty$ for each \mathbf{x} and $\mathbb{E}[|\overline{\varphi}(\mathbf{X})|] < \infty$. $\{X_n\}_{n=1}^\infty$ is said to be a sequence of independent random variables, if X_{n+1} is independent of (X_1, \dots, X_n) for each $n \geq 1$.

Suppose that \mathbf{X}_1 and \mathbf{X}_2 are two n -dimensional random vectors defined, respectively, in sublinear expectation spaces $(\Omega_1, \mathcal{H}_1, \mathbb{E}_1)$ and $(\Omega_2, \mathcal{H}_2, \mathbb{E}_2)$. They are said to be identically distributed if

$$\mathbb{E}_1[\varphi(\mathbf{X}_1)] = \mathbb{E}_2[\varphi(\mathbf{X}_2)], \quad \forall \varphi \in C_{l,\text{Lip}}(\mathbb{R}^n), \quad (3)$$

whenever the sublinear expectations are finite. $\{X_n\}_{n=1}^\infty$ is said to be identically distributed if for each $i \geq 1$, X_i and X_1 are identically distributed.

For $0 \leq \underline{\sigma}^2 \leq \overline{\sigma}^2 < \infty$, a random variable ξ under a sublinear expectation space $(\Omega, \mathcal{H}, \mathbb{E})$ is called a G-normal $\mathcal{N}(0, [\underline{\sigma}^2, \overline{\sigma}^2])$ distributed random variable, if for any $\varphi \in C_{l,\text{Lip}}(\mathbb{R}^n)$, $u(x, t) := \mathbb{E}[\varphi(x + \sqrt{t}\xi)]$ ($x \in \mathbb{R}, t \geq 0$) is the unique viscosity solution of the following heat equation:

$$\begin{aligned} \partial_t u - G(\partial_{xx}^2 u) &= 0, \\ u(0, x) &= \varphi(x), \end{aligned} \quad (4)$$

where $G(\alpha) = (\overline{\sigma}^2 \alpha^+ - \underline{\sigma}^2 \alpha^-)/2$.

In the rest of this paper, let $\{X, X_n, n \geq 1\}$ be a sequence of i.i.d. random variables under sublinear expectation space $(\Omega, \mathcal{H}, \mathbb{E})$ with $\mathbb{E}(X) = \mathbb{E}(-X) = 0$, $\mathbb{E}(X^2) = \overline{\sigma}^2 < \infty$, and $-\mathbb{E}(-X^2) = \underline{\sigma}^2$, $\lim_{c \rightarrow \infty} \mathbb{E}(X^2 - c)^+ = 0$, $C_V(X^2) < \infty$. Set $S_n = \sum_{i=1}^n X_i$. Assume that \mathbb{E} is continuous. Let ξ be a G-normal-distributed random variable with $\mathbb{E}(\xi) = \mathbb{E}(-\xi) = 0$, $\mathbb{E}(\xi^2) = \overline{\sigma}^2$, and $-\mathbb{E}(-\xi^2) = \underline{\sigma}^2$. We denote by C a positive constant which may vary from line to line.

To prove our results, we need the following lemmas.

Lemma 1. Suppose $\mathbb{E}|X|^\alpha < \infty$, $1 < \alpha \leq 2$. Then, for $x, y > 0$,

$$\mathbb{V}\{|S_n| \geq x\} \leq 2n\mathbb{V}\{|X| > y\} + 2n^{x/y} \left(\frac{e\mathbb{E}|X|^\alpha}{n\mathbb{E}|X|^\alpha + xy^{\alpha-1}} \right)^{x/y}. \quad (5)$$

Proof. We borrow the proofs from those of Theorem 2 by Fuk and Nagaev [31], and Lemma 2 by Spătaru [30]. Let

$$\begin{aligned} \tilde{X}_i &= \begin{cases} X_i & \text{for } |X_i| \leq y, \\ 0 & \text{for } |X_i| > y, \end{cases} \quad i = 1, \dots, n, \\ \tilde{S}_n &= \sum_{i=1}^n \tilde{X}_i. \end{aligned} \quad (6)$$

Therefore, by the subadditivity property of $\mathbb{V}(\cdot)$,

$$\mathbb{V}\{S_n \geq x\} \leq \mathbb{V}\{\tilde{S}_n \neq S_n\} + \mathbb{V}\{\tilde{S}_n \geq x\}. \quad (7)$$

By Markov's inequality under sublinear expectations, for any positive h ,

$$\mathbb{V}\{\tilde{S}_n \geq x\} \leq e^{-hx} \mathbb{E}\left(e^{h\tilde{S}_n}\right). \quad (8)$$

From this and (7), it follows that

$$\begin{aligned} \mathbb{V}\{S_n \geq x\} &\leq \sum_{i=1}^n \mathbb{V}\{|X_i| \geq y\} + e^{-hx} \mathbb{E}\left(e^{h\tilde{S}_n}\right) \\ &= n\mathbb{V}\{X \geq y\} + e^{-hx} \mathbb{E}\left(e^{h\tilde{S}_n}\right). \end{aligned} \quad (9)$$

Application of the monotonicity of $u^{-2}(e^{hu} - 1 - hu)$ for $u \leq y$ and $u^{-\alpha}(e^{hu} - 1 - hu)$ for $u > 0$ and the subadditivity property of sublinear expectations yields

$$\begin{aligned} \mathbb{E}e^{h\tilde{X}_i} &\leq 1 + \mathbb{E}\left(hX_i I_{|X_i| \leq y}\right) + \mathbb{E}\left(\frac{e^{hX_i} - 1 - hX_i}{X_i^2} X_i^2 I_{|X_i| \leq y}\right) \\ &\leq 1 + h\mathbb{E}\left(X_i I_{|X_i| \leq y}\right) + \frac{e^{hy} - 1 - hy}{y^2} \mathbb{E}\left(X_i^2 I_{|X_i| \leq y}\right) \\ &\leq 1 + h\mathbb{E}\left(X_i I_{|X_i| \leq y}\right) + \frac{e^{hy} - 1 - hy}{y^\alpha} \mathbb{E}\left(|X_i|^\alpha I_{|X_i| \leq y}\right) \\ &\leq 1 + h\mathbb{E}\left(X_i I_{|X_i| \leq y}\right) + \frac{e^{hy} - 1 - hy}{y^\alpha} \mathbb{E}\left(|X_i|^\alpha\right). \end{aligned} \quad (10)$$

Hence, by Lemma 1.1 in the study of Gao and Xu [32],

$$\begin{aligned} e^{-hx} \mathbb{E}\left(e^{h\tilde{S}_n}\right) &\leq \exp\left\{\left(e^{hy} - 1 - hy\right)y^{-\alpha} n\mathbb{E}\left(|X|^\alpha\right) \right. \\ &\quad \left. - hx + hn\mathbb{E}\left(XI_{|X| \leq y}\right)\right\}. \end{aligned} \quad (11)$$

Setting

$$h = \frac{1}{y} \log\left(\frac{xy^{\alpha-1}}{n\mathbb{E}\left(|X|^\alpha\right)} + 1\right), \quad (12)$$

in the right-hand side of (11), we see that

$$e^{-hx} \mathbb{E}\left(e^{h\tilde{S}_n}\right) \leq \exp\left\{\frac{x}{y} - \left(\frac{x - n\mathbb{E}\left(XI_{|X| \leq y}\right)}{y} + \frac{n\mathbb{E}\left(|X|^\alpha\right)}{y^\alpha}\right) \log\left(\frac{xy^{\alpha-1}}{n\mathbb{E}\left(|X|^\alpha\right)} + 1\right)\right\}. \quad (13)$$

Since $\mathbb{E}(X) = \mathbb{E}(-X) = 0$, by Proposition 3.6 in the study of Peng [3] and Definition 1, we see that

$$\begin{aligned} \mathbb{E}\left(XI_{|X| \leq y}\right) &= \left|\mathbb{E}\left(-XI_{|X| \geq y}\right)\right| \leq \mathbb{E}\left(|X|I_{|X| \geq y}\right) \\ &\leq \frac{1}{y^{\alpha-1}} \mathbb{E}\left(|X|^\alpha I_{|X| \geq y}\right) \leq \frac{1}{y^{\alpha-1}} \mathbb{E}\left[|X|^\alpha\right]. \end{aligned} \quad (14)$$

Therefore,

$$\frac{1}{y}\left(x - n\mathbb{E}\left(XI_{|X| \leq y}\right)\right) + \frac{n\mathbb{E}\left(|X|^\alpha\right)}{y^\alpha} \geq \frac{x}{y}. \quad (15)$$

Combining this with (13) and (9), we conclude that

$$\mathbb{V}\{S_n \geq x\} \leq n\mathbb{V}\{|X| > y\} + n^{x/y} \left(\frac{e\mathbb{E}|X|^\alpha}{n\mathbb{E}|X|^\alpha + xy^{\alpha-1}}\right)^{x/y}. \quad (16)$$

Combining (16) with the inequality derived from it with $-X$ and $-X_k$ in place of X and X_k , respectively, leads to (5). \square

Remark 1. (see Lemma 2 in [7]). For any $X \in \mathcal{H}$, we have

$$C_{\mathbb{V}}(X^2) < \infty \Leftrightarrow \int_1^\infty x\mathbb{V}(|X| > x)dx < \infty. \quad (17)$$

Lemma 2 (see Lemma 5 in [7]). Assume that $\{X_n; n \geq 1\}$ is a sequence of independent and identically distributed random variables with $\mathbb{E}[X_1] = \mathbb{E}[-X_1] = 0$ and $\lim_{c \rightarrow \infty} \mathbb{E}(X^2 - c)^+ = 0$. Write $\bar{\sigma}^2 = \mathbb{E}[X_1^2]$ and $\underline{\sigma}^2 = -\mathbb{E}[-X_1^2]$. Suppose that \mathbb{E} is continuous and set $\Delta_n(x) = \mathbb{V}(|S_n|/\sqrt{n} \geq x) - \mathbb{V}(|\xi| \geq x)$, $\xi \sim \mathcal{N}(0, [\underline{\sigma}^2, \bar{\sigma}^2])$ under \mathbb{E} . Then,

$$\Delta_n := \sup_{x \geq 0} |\Delta_n(x)| \longrightarrow 0, \text{ as } n \longrightarrow \infty. \quad (18)$$

3. Main Results

The following are our main results.

Theorem 1. For $b, d > 0$, we have

$$\lim_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n=3}^\infty \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|S_n| \geq \varepsilon \sqrt{n} (\log \log n)^d\} = \frac{C_{\mathbb{V}}(|\xi|^{b/d})}{b}. \quad (19)$$

Theorem 2. For $d > 0$, we have

$$\lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \sum_{n=3}^{\infty} \frac{1}{n \log n \log \log n} \mathbb{V}\{|S_n| \geq \varepsilon \sigma \sqrt{n} (\log \log n)^d\} = \frac{1}{d}. \quad (20)$$

In the following two sections, for $M \geq 3$ and $0 < \varepsilon < 1$, set $b(\varepsilon) = \lfloor \exp\{\exp\{M\varepsilon^{-1/d}\}\} \rfloor$.

4. Proof of Theorem 1

Proposition 1. For $b, d > 0$, we have

$$\lim_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n=3}^{\infty} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} = \frac{C_{\mathbb{V}}(|\xi|^{b/d})}{b}. \quad (21)$$

Proof.

$$\begin{aligned} & \lim_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n=3}^{\infty} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} \\ &= \lim_{\varepsilon \searrow 0} \varepsilon^{b/d} \int_{e^e}^{\infty} \frac{(\log \log y)^{b-1}}{y \log y} \mathbb{V}\{|\xi| \geq \varepsilon (\log \log y)^d\} dy \\ &= \lim_{\varepsilon \searrow 0} \varepsilon^{b/d} \int_{\varepsilon}^{\infty} (y/\varepsilon)^{(b-1)/d} (1/d) (y/\varepsilon)^{1/d-1} \frac{1}{\varepsilon} \mathbb{V}\{|\xi| \geq y\} dy \\ &= \lim_{\varepsilon \searrow 0} \int_{\varepsilon}^{\infty} (1/d) y^{b/d-1} \mathbb{V}\{|\xi| \geq y\} dy \\ &= \lim_{\varepsilon \searrow 0} \int_{\varepsilon^{b/d}}^{\infty} (1/b) \mathbb{V}\{|\xi|^{b/d} \geq t\} dt \\ &= \int_0^{\infty} (1/b) \mathbb{V}\{|\xi|^{b/d} \geq t\} dt \\ &= \frac{C_{\mathbb{V}}(|\xi|^{b/d})}{b}. \end{aligned} \quad (22)$$

Thus, this completes the proof of Proposition 1. \square

Remark 2. By the proof of (24) and (25) in the study by Wu [7], $C_{\mathbb{V}}(|\xi|^{b/d})$ is finite for any $b, d > 0$.

Proposition 2. For $b, d > 0$, we have

$$\begin{aligned} & \lim_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n \leq b(\varepsilon)} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|S_n| \geq \varepsilon \sqrt{n} (\log \log n)^d\} \\ & \quad - \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} = 0. \end{aligned} \quad (23)$$

Proof. By Lemma 2 and Toeplitz's lemma,

$$\begin{aligned} & \lim_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n \leq b(\varepsilon)} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|S_n| \geq \varepsilon \sqrt{n} (\log \log n)^d\} \\ & \quad - \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} \\ & \leq \lim_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n \leq b(\varepsilon)} \frac{(\log \log n)^{b-1}}{n \log n} \Delta_n \\ & = \lim_{\varepsilon \searrow 0} \frac{CM^b}{(\log \log(b(\varepsilon)))^b} \sum_{n \leq b(\varepsilon)} \frac{(\log \log n)^{b-1}}{n \log n} \Delta_n = 0. \end{aligned} \quad (24)$$

The proof is complete. \square

Proposition 3. For $b, d > 0$, we have

$$\lim_{M \rightarrow \infty} \limsup_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n > b(\varepsilon)} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} = 0. \quad (25)$$

Proof. We could obtain that

$$\begin{aligned} & \limsup_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n > b(\varepsilon)} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} \\ & \leq C \limsup_{\varepsilon \searrow 0} \varepsilon^{b/d} \int_{b(\varepsilon)}^{\infty} \frac{(\log \log y)^{b-1}}{y \log y} \mathbb{V}\{|\xi| \geq \varepsilon (\log \log y)^d\} dy \\ & \leq C \int_{M^d}^{\infty} t^{b/d-1} \mathbb{V}\{|\xi| \geq t\} dt \\ & = C \int_{M^b}^{\infty} \mathbb{V}\{|\xi|^{b/d} \geq t\} dt. \end{aligned} \quad (26)$$

Note that $\int_{M^b}^{\infty} \mathbb{V}\{|\xi|^{b/d} \geq t\} dt$ is integrable:

$$\int_{M^b}^{\infty} \mathbb{V}\{|\xi|^{b/d} \geq t\} dt \longrightarrow 0, \quad (27)$$

as $M \longrightarrow \infty$. Proposition 3 is established. \square

Proposition 4. For $b, d > 0$, we have

$$\lim_{M \rightarrow \infty} \limsup_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n > b(\varepsilon)} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|S_n| \geq \varepsilon \sqrt{n} (\log \log n)^d\} = 0. \quad (28)$$

Proof. When $0 < b < 2d$, by Markov's inequality under sublinear expectations, we have

$$\begin{aligned} & \limsup_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n > b(\varepsilon)} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|S_n| \geq \varepsilon \sqrt{n} (\log \log n)^d\} \\ & \leq C \limsup_{\varepsilon \searrow 0} \varepsilon^{b/d-2} \sum_{n > b(\varepsilon)} \frac{(\log \log n)^{b-1-2d}}{n^2 \log n} \mathbb{E}[S_n^2] \\ & = C \limsup_{\varepsilon \searrow 0} \varepsilon^{b/d-2} \sum_{n > b(\varepsilon)} \frac{(\log \log n)^{b-1-2d}}{n \log n} \\ & \leq C \limsup_{\varepsilon \searrow 0} \varepsilon^{b/d-2} (\log \log(b(\varepsilon)))^{b-2d} \\ & \leq CM^{b-2d} \rightarrow 0, \text{ as } M \rightarrow \infty. \end{aligned} \quad (29)$$

For $b \geq 2d$, by Lemma 1, we see that

$$\begin{aligned} & \sum_{n > b(\varepsilon)} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|S_n| \geq \varepsilon \sqrt{n} (\log \log n)^d\} \\ & \leq \sum_{n > b(\varepsilon)} \frac{(\log \log n)^{b-1}}{\log n} \mathbb{V}\{|X| > \varepsilon \sqrt{n} (\log \log n)^d / T\} \\ & + C \sum_{n > b(\varepsilon)} \frac{(\log \log n)^{b-1}}{n \log n} \frac{1}{(\log \log n)^{2dT} \varepsilon^{2T}} =: L_1 + L_2, \end{aligned} \quad (30)$$

where T is a positive constant to be specified later. On the one hand, we obtain that

$$\begin{aligned} \limsup_{\varepsilon \searrow 0} \varepsilon^{b/d} L_2 & \leq \limsup_{\varepsilon \searrow 0} \varepsilon^{b/d-2T} \sum_{n > b(\varepsilon)} \frac{(\log \log n)^{b-1-2dT}}{n \log n} \\ & \leq \limsup_{\varepsilon \searrow 0} C \varepsilon^{b/d-2T} (\log \log b(\varepsilon))^{b-2dT} \\ & \leq CM^{b-2dT} \rightarrow 0, \text{ as } M \rightarrow \infty, \end{aligned} \quad (31)$$

for any $T > b/(2d)$. On the other hand, for L_1 , without loss of generality, set $T = 1$. By the countable subadditivity property of sublinear expectations and the fact that $((\log \log x)^{b-1} / \log x) \rightarrow 0$, as $x > b(\varepsilon) \rightarrow \infty$, we obtain that

$$\begin{aligned} L_1 & = \sum_{n > b(\varepsilon)} \frac{(\log \log n)^{b-1}}{\log n} \mathbb{V}\{|X| > \varepsilon \sqrt{n} (\log \log n)^d\} \\ & \leq C \int_{x > b(\varepsilon)} \frac{(\log \log x)^{b-1}}{\log x} \mathbb{V}\{|X| > \varepsilon \sqrt{x} (\log \log x)^d\} dx \\ & \leq C \int_{x > b(\varepsilon)} \mathbb{V}\{|X|^2 > \varepsilon^2 x (\log \log x)^{2d}\} dx \\ & \leq C \varepsilon^{-2} \int_{x > b(\varepsilon) M^{2d}} \mathbb{V}\{|X|^2 > y\} dy. \end{aligned} \quad (32)$$

Hence, for $b \geq 2d$, we have

$$\limsup_{\varepsilon \searrow 0} \varepsilon^{b/d} L_1 \leq C \varepsilon^{b/d-2} \int_{x > b(\varepsilon) M^{2d}} \mathbb{V}\{|X|^2 > y\} dy = 0. \quad (33)$$

Thus, (28) holds for each $b, d > 0$.

Now, by Proposition 1-4 and the triangle inequality, $\forall \beta > 0, \exists M > 0$, which is sufficiently large, such that

$$\begin{aligned}
& \lim_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n=3}^{\infty} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|S_n| \geq \varepsilon \sqrt{n} (\log \log n)^d\} \\
& \leq \lim_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n=3}^{\infty} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} \\
& \quad + \lim_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n \leq b(\varepsilon)} \frac{(\log \log n)^{b-1}}{n \log n} \left| \mathbb{V}\{|S_n| \geq \varepsilon \sqrt{n} (\log \log n)^d\} - \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} \right| \\
& \quad + \limsup_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n > b(\varepsilon)} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} \\
& \quad + \limsup_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n > b(\varepsilon)} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|S_n| \geq \varepsilon \sqrt{n} (\log \log n)^d\} \\
& = \frac{C_{\mathbb{V}}(|\xi|^{b/d})}{b} + \beta,
\end{aligned} \tag{34}$$

$$\begin{aligned}
& \lim_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n=3}^{\infty} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|S_n| \geq \varepsilon \sqrt{n} (\log \log n)^d\} \\
& \geq \lim_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n=3}^{\infty} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} \\
& \quad - \lim_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n \leq b(\varepsilon)} \frac{(\log \log n)^{b-1}}{n \log n} \left| \mathbb{V}\{|S_n| \geq \varepsilon \sqrt{n} (\log \log n)^d\} - \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} \right| \\
& \quad - \limsup_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n > b(\varepsilon)} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} \\
& \quad - \limsup_{\varepsilon \searrow 0} \varepsilon^{b/d} \sum_{n > b(\varepsilon)} \frac{(\log \log n)^{b-1}}{n \log n} \mathbb{V}\{|S_n| \geq \varepsilon \sqrt{n} (\log \log n)^d\} \\
& \geq \frac{C_{\mathbb{V}}(|\xi|^{b/d})}{b} - \beta.
\end{aligned}$$

We derive Theorem 1 from the arbitrariness of $\beta > 0$. \square

5. Proof of Theorem 2

Proposition 5. For $d > 0$, we have

$$\lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \sum_{n=3}^{\infty} \frac{1}{n \log n \log \log n} \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} = \frac{1}{d}. \tag{35}$$

Proof. We claim that

$$\begin{aligned}
& \lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \sum_{n=3}^{\infty} \frac{1}{n \log n \log \log n} \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} \\
& = \lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \int_{e^\varepsilon}^{\infty} \frac{1}{y \log y \log \log y} \mathbb{V}\{|\xi| \geq \varepsilon (\log \log y)^d\} dy \\
& = \frac{1}{d} \lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \int_{\varepsilon}^{\infty} \frac{1}{t} \mathbb{V}\{|\xi| \geq t\} dt \\
& = \frac{1}{d}.
\end{aligned} \tag{36}$$

Indeed, by Lemma 4 in the study by Wu [7], $\forall \alpha > 0$, $\exists \delta > 0$, such that $\forall t < \delta < 1$, $\mathbb{V}\{|\xi| > t\} > 1 - \alpha d$. Therefore,

$$\begin{aligned}
 & \frac{1}{d} \left| \lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \int_{\varepsilon}^{\infty} \frac{1}{t} \mathbb{V}\{|\xi| \geq t\} dt - 1 \right| \\
 & \leq \frac{1}{d} \left| \lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \int_{\varepsilon}^{\delta} \frac{1}{t} \mathbb{V}\{|\xi| \geq t\} dt - 1 \right| + \lim_{\varepsilon \searrow 0} \left| \frac{1}{-\log \varepsilon} \int_{\delta}^{\infty} \frac{1}{t} \mathbb{V}\{|\xi| \geq t\} dt \right| \\
 & < \frac{1}{d} \lim_{\varepsilon \searrow 0} \left| \frac{1}{-\log \varepsilon} \int_{\varepsilon}^{\delta} \frac{1}{t} (1 - \alpha d) dt - 1 \right| + \lim_{\varepsilon \searrow 0} \left| \frac{1}{-\log \varepsilon} \int_{\delta}^{\infty} \frac{1}{t^3} \mathbb{E}[|\xi|^2] dt \right| \\
 & \leq \alpha.
 \end{aligned} \tag{37}$$

This establishes (35). \square

Proposition 6. For $d > 0$, we have

$$\lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \sum_{n \leq b(\varepsilon)} \frac{1}{n \log n \log \log n} \left| \mathbb{V}\{|S_n| \geq \varepsilon \sqrt{n} (\log \log n)^d\} - \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} \right| = 0. \tag{38}$$

Proof. By Lemma 2 and Toeplitz's lemma,

$$\begin{aligned}
 & \lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \sum_{n \leq b(\varepsilon)} \frac{1}{n \log n \log \log n} \left| \mathbb{V}\{|S_n| \geq \varepsilon \sqrt{n} (\log \log n)^d\} - \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} \right| \\
 & = \lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \sum_{n \leq b(\varepsilon)} \frac{1}{n \log n \log \log n} \Delta_n \\
 & = \lim_{\varepsilon \searrow 0} \frac{\log(M) - \log(\varepsilon)/d}{-\log \varepsilon} \frac{1}{\log \log b(\varepsilon)} \sum_{n \leq b(\varepsilon)} \frac{1}{n \log n \log \log n} \Delta_n = 0.
 \end{aligned} \tag{39}$$

The proof is complete. \square

Proof. By Markov inequality under sublinear expectations, we see that

Proposition 7. For $d > 0$, we have

$$\lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \sum_{n > b(\varepsilon)} \frac{1}{n \log n \log \log n} \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} = 0. \tag{40}$$

$$\begin{aligned}
 & \lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \sum_{n > b(\varepsilon)} \frac{1}{n \log n \log \log n} \mathbb{V}\{|\xi| \geq \varepsilon (\log \log n)^d\} \\
 & \leq C \lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \int_{b(\varepsilon)}^{\infty} \frac{1}{y \log y \log \log y} \mathbb{V}\{|\xi| \geq \varepsilon (\log \log y)^d\} dy \\
 & \leq C \lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \int_{M^d}^{\infty} \frac{1}{t} \mathbb{V}\{|\xi| > t\} dt \\
 & \leq C \lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \int_{M^d}^{\infty} \frac{1}{t^3} \mathbb{E}[\xi^2] dt = 0.
 \end{aligned} \tag{41}$$

Thus this proves Proposition 7. \square

Proposition 8. For $d > 0$, we have

$$\lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \sum_{n > b(\varepsilon)} \frac{1}{n \log n \log \log n} \mathbb{V}\{|S_n| \geq \varepsilon \sqrt{n} (\log \log n)^d\} = 0. \quad (42)$$

Proof. By Markov inequality under sublinear expectations, we deduce that

$$\begin{aligned} & \lim_{\varepsilon \searrow 0} \frac{1}{-\log \varepsilon} \sum_{n > b(\varepsilon)} \frac{1}{n \log n \log \log n} \mathbb{V}\{|S_n| \geq \varepsilon \sqrt{n} (\log \log n)^d\} \\ & \leq \lim_{\varepsilon \searrow 0} \frac{1}{-\varepsilon^2 \log \varepsilon} \sum_{n > b(\varepsilon)} \frac{1}{n^2 \log n (\log \log n)^{1+2d}} \mathbb{E}[S_n^2] \\ & \leq C \lim_{\varepsilon \searrow 0} \frac{1}{-\varepsilon^2 \log \varepsilon} \sum_{n > b(\varepsilon)} \frac{1}{n \log n (\log \log n)^{1+2d}} \\ & \leq C \lim_{\varepsilon \searrow 0} \frac{(\log \log(b(\varepsilon)))^{-2d}}{-\varepsilon^2 \log \varepsilon} \\ & \leq C \lim_{\varepsilon \searrow 0} \frac{M^{-2d}}{-\log \varepsilon} = 0. \end{aligned} \quad (43)$$

The proof is complete.

Finally, similar to the proof of Theorem 1, by the triangle inequality and Propositions 5–8, we finish the proof of Theorem 2. \square

Data Availability

No data were used to support this study.

Conflicts of Interest

The authors declare that they have no conflicts of interest.

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