

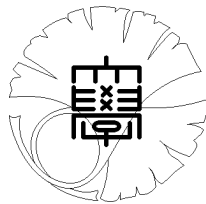
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**Asymptotic Behavior of distributions of the sum
of i.i.d. random variables with fat tail I**

by

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1 Introduction

Let (Ω, \mathcal{F}, P) be a probability space. Let X_n , $n = 1, 2, \dots$, be independent identically distributed random variables, and $F : \mathbf{R} \rightarrow [0, 1]$ and $\bar{F} : \mathbf{R} \rightarrow [0, 1]$ be given by

$$F(x) = P(X_1 \leq x), \quad \text{and} \quad \bar{F}(x) = P(X_1 > x), \quad x \in \mathbf{R}.$$

We assume the following.

(A-1) $\bar{F}(x)$ is a regular varying function of index $-\alpha$, for some $\alpha > 2$, as $x \rightarrow \infty$, i.e., if we let

$$L(x) = x^\alpha \bar{F}(x), \quad x \in \mathbf{R},$$

then $L(x) > 0$ for any $x > 0$, and for any $a > 0$

$$\frac{L(ax)}{L(x)} \rightarrow 1, \quad x \rightarrow \infty.$$

Also we assume the following.

(A2) $x^{\alpha-\delta} F(-x) \rightarrow 0$, as $x \rightarrow \infty$ for any $\delta > 0$.

Since $\alpha > 2$, we see that $E[|X_1|^2] < \infty$. We assume furthermore for simplicity that

(A3) $E[X_1] = 0$.

Our first main theorem is the following.

Theorem 1 *Assume the assumptions (A-1), (A-2) and (A-3), and let $\beta : \mathbf{N} \rightarrow (0, \infty)$ be such that*

$$\frac{\beta(n)}{(\log n)^{1/2}} \rightarrow \infty, \quad n \rightarrow \infty.$$

Then we have

$$\sup_{s \geq n^{1/2}\beta(n)} \left| \frac{P(\sum_{k=1}^n X_k > s)}{n\bar{F}(s)} - 1 \right| \rightarrow 0, \quad n \rightarrow \infty.$$

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We remark that the following has been shown essentially by Feller [1].

Theorem 2 (Feller) *Under the assumptions (A-1),(A-2) and (A-3), we have for any $N \geq 1$*

$$\left| \frac{P(\sum_{k=1}^N X_k > x)}{N\bar{F}(x)} - 1 \right| \rightarrow 0, \quad x \rightarrow \infty.$$

Note that the number of random variables is fixed in Feller's result, but is increasing in our result.

Let us assume the following furthermore.

(A-4) There is an $x_0 > 0$ such that \bar{F} is twice continuously differentiable on (x_0, ∞) and that

$$x^2 \frac{d^2}{dx^2} \log \bar{F}(x) \rightarrow \alpha, \quad x \rightarrow \infty.$$

Then we have the following.

Theorem 3 *Assume the assumptions (A-1), (A-2), (A-3) and (A-4) and let $\beta : \mathbf{N} \rightarrow (0, \infty)$ be such that*

$$\frac{\beta(n)}{(\log n)^{1/2}} \rightarrow \infty, \quad n \rightarrow \infty.$$

Let $v = E[X_1^2]$. Then we have

$$\sup_{s \geq n^{1/2}\beta(n)} \frac{s^2}{n} \left| \frac{P(\sum_{k=1}^n X_k > s)}{n\bar{F}(s)} - \left(1 + \frac{\alpha(\alpha+1)vn}{2s^2}\right) \right| \rightarrow 0, \quad n \rightarrow \infty.$$

2 Preparations

Proposition 4 *Let Y be a random variable and $\gamma \in (1, 2]$, and assume that*

$$E[|Y|^\gamma] < \infty \quad \text{and} \quad E[Y] = 0.$$

Then for any $s \in \mathbf{R} \setminus \{0\}$ and $b > 0$

$$E[\exp(sY)1_{\{|Y| \leq b\}}] \leq 1 + |s|^\gamma \left(1 + \left(\frac{1}{|s|b}\right)^{\gamma-1}\right) \exp(|s|b) E[|Y|^\gamma].$$

Proof. First, note that

$$|\exp(x) - 1| \leq 1 \vee \exp(x),$$

and

$$|\exp(x) - 1| = \left| \int_0^x e^y dy \right| \leq |x|(1 \vee \exp(x)), \quad x \in \mathbf{R}.$$

So we have

$$|\exp(x) - (1+x)| = \left| \int_0^x (e^y - 1) dy \right| \leq (|x| \wedge |x|^2)(1 \vee \exp(x))$$

for any $x \in \mathbf{R}$. Therefore we see that

$$|\exp(x) - (1+x)| \leq |x|^\gamma \exp(|x|), \quad x \in \mathbf{R}$$

for $\gamma \in (1, 2]$.

Therefore we have

$$\begin{aligned} & |E[\exp(sY1_{\{|Y|\leq b\}})] - (1 + E[sY1_{\{|Y|\leq b\}}])| \\ & \leq |s|^\gamma \exp(|s|b)E[|Y|^\gamma]. \end{aligned}$$

Since

$$|E[sY1_{\{|Y|\leq b\}}]| = |sE[Y, |Y| > b]| \leq |s|b^{-(\gamma-1)}E[|Y|^\gamma],$$

we have our assertion. ■

Proposition 5 *Let X be a random variable and $\gamma \in (1, 2]$. We assume that*

$$E[|X|^\gamma] < \infty \text{ and } E[X] = 0.$$

Then for any $t > 0$ and $n \geq 1$

$$n \log E[\exp(\pm \frac{1}{tn^{1/\gamma}} X 1_{\{|X|\leq tn^{1/\gamma}\}})] \leq \frac{6}{t^\gamma} E[|X|^\gamma].$$

Proof. Let $Y = (1/t)X$, $s = \pm n^{-1/\gamma}$, $b = n^{1/\gamma}$, and apply Proposition 4. Since $\log(1+x) \leq x$, $x \geq 0$, we have our assertion. ■

Now let X_n , $n = 1, 2, \dots$, be independent identically distributed random variables and $\gamma \in (1, 2]$. Throughout this section we assume that

$$E[|X_1|^\gamma] < \infty \text{ and } E[X_1] = 0.$$

Proposition 6 *For any $s, t > 0$ and $\varepsilon > 0$*

$$P(|\sum_{k=1}^n X_k 1_{\{|X_k|\leq tn^{1/\gamma}\}}| \geq sn^{1/\gamma}) \leq 2 \exp(\frac{6}{t^\gamma} E[|X_1|^\gamma]) \exp(-\frac{s}{t}).$$

Proof. We see that

$$\begin{aligned} & P(\pm \sum_{k=1}^n X_k 1_{\{|X_k|\leq tn^{1/\gamma}\}} \geq sn^{1/\gamma}) \\ & \leq \exp(-\frac{s}{t}) E[\exp(\frac{\pm 1}{tn^{1/\gamma}} \sum_{k=1}^n X_k 1_{\{|X_k|\leq tn^{1/\gamma}\}})] \\ & \leq \exp(-\frac{s}{t}) E[\exp(\frac{\pm 1}{tn^{1/\gamma}} X_1 1_{\{|X_1|\leq tn^{1/\gamma}\}})]^n. \end{aligned}$$

Then by Proposition 5 we have our assertion.

Let $F : \mathbf{R} \rightarrow [0, 1]$ and $\bar{F} : \mathbf{R} \rightarrow [0, 1]$ be given by

$$F(x) = P(X_1 \leq x), \quad x \in \mathbf{R}$$

and

$$\bar{F}(x) = P(X_1 > x), \quad x \in \mathbf{R}.$$

Then we have the following.

Proposition 7 (1) For any $t, s > 0$, and $n \geq 2$,

$$P\left(\left|\sum_{k=2}^n X_k 1_{\{|X_k| \leq tn^{1/\gamma}\}}\right| > sn^{1/\gamma}\right) \leq 2 \exp\left(\frac{6}{t^\gamma} E[|X_1|^\gamma]\right) \exp\left(-\frac{s}{t}\right).$$

(2) For any $s, t > 0$, $\varepsilon \in (0, 1)$ with $t < (1 - \varepsilon)s$,

$$\begin{aligned} & \left|P\left(\sum_{k=1}^n X_k > sn^{1/\gamma}\right) - nP\left(X_1 + \sum_{k=2}^n X_k 1_{\{|X_k| \leq tn^{1/\gamma}\}} > sn^{1/\gamma}\right), \left|\sum_{k=2}^n X_k 1_{\{|X_k| \leq tn^{1/\gamma}\}}\right| \leq \varepsilon sn^{1/\gamma}\right| \\ & \leq 2n(n-1)(F(-tn^{1/\gamma}) + \bar{F}(tn^{1/\gamma}))^2 + 2 \exp\left(\frac{6}{t^\gamma} E[|X_1|^\gamma]\right) \exp\left(-\frac{s}{t}\right) \\ & \quad + 2n(F(-tn^{1/\gamma}) + \bar{F}(tn^{1/\gamma})) \exp\left(\frac{6}{t^\gamma} E[|X_1|^\gamma]\right) \exp\left(-\frac{\varepsilon s}{2t}\right) \end{aligned}$$

Proof. Note that

$$\begin{aligned} & P\left(\left|\sum_{k=2}^n X_k 1_{\{|X_k| \leq tn^{1/\gamma}\}}\right| > sn^{1/\gamma}\right) \\ & = P\left(\left|\sum_{k=1}^{n-1} X_k 1_{\{|X_k| \leq \tilde{t}(n-1)^{1/\gamma}\}}\right| > \tilde{s}(n-1)^{1/\gamma}\right), \end{aligned}$$

where

$$\tilde{t} = t\left(\frac{n}{n-1}\right)^{1/\gamma}, \quad \tilde{s} = s\left(\frac{n}{n-1}\right)^{1/\gamma}.$$

So we have the assertion (1) from Proposition 6 .

Let us denote

$$\tilde{F}(x) = P(|X_1| > x) \leq F(-x) + \bar{F}(x), \quad x > 0.$$

Note that

$$P\left(\sum_{k=1}^n X_k > sn^{1/\gamma}\right) = \sum_{m=0}^n I_m,$$

where

$$I_m = P\left(\sum_{k=1}^n X_k > sn^{1/\gamma}, \sum_{k=1}^n 1_{\{|X_k| > tn^{1/\gamma}\}} = m\right), \quad m = 0, 1, \dots, n.$$

Then we have

$$I_m = \binom{n}{m} P\left(\sum_{k=1}^n X_k > sn^{1/\gamma}, |X_i| > tn^{1/\gamma}, i = 1, \dots, m, |X_j| \leq tn^{1/\gamma}, j = m+1, \dots, n\right),$$

for $m = 0, 1, \dots, n$. So we see that

$$\sum_{m=2}^n I_m \leq \sum_{m=2}^n \frac{n(n-1)}{m(m-1)} \binom{n-2}{m-2} \tilde{F}(tn^{1/\gamma})^m (1 - \tilde{F}(tn^{1/\gamma}))^{n-m} \leq \frac{n(n-1)}{2} \tilde{F}(tn^{1/\gamma})^2. \quad (1)$$

Also, by Proposition 6, we have

$$I_0 \leq 2 \exp\left(-\frac{s}{t}\right) \exp\left(\frac{6}{t^\gamma} E[|X_1|^\gamma]\right). \quad (2)$$

Let

$$A_1 = \{|X_1| > tn^{1/\gamma}\}, \quad A_2 = \{|X_k| \leq tn^{1/\gamma}, k = 2, 3, \dots, n\},$$

$$B_1 = \{X_1 + \sum_{k=2}^n X_k 1_{\{|X_k| \leq tn^{1/\gamma}\}} > sn^{1/\gamma}\},$$

and

$$B_2 = \{|\sum_{k=2}^n X_k 1_{\{|X_k| \leq tn^{1/\gamma}\}}| \leq \varepsilon sn^{1/\gamma}\}.$$

Note that $B_1 \cap B_2 \subset A_1$, since $t < (1 - \varepsilon)s$. So we see that

$$\begin{aligned} & |P(B_1 \cap A_1 \cap A_2) - P(B_1 \cap B_2)| \\ & \leq P(B_1 \cap B_2^c \cap A_1 \cap A_2) + P(B_1 \cap B_2 \cap A_1 \cap A_2^c) \\ & \leq P(A_1)P(B_2^c) + P(A_1)P(A_2^c). \end{aligned} \quad (3)$$

Note that

$$P(A_2^c) \leq \sum_{k=2}^n P(|X_k| > tn^{1/\gamma}) = (n-1)\tilde{F}(tn^{1/\gamma}).$$

Also, by the assertion (1) we have

$$P(B_2^c) \leq 2 \exp\left(\frac{6}{t^\gamma} E[|X_1|^\gamma]\right) \exp\left(-\frac{\varepsilon s}{2t}\right).$$

Since $I_1 = nP(B_1 \cap A_1 \cap A_2)$, we have the assertion from Equations (1), (2) and (3).

This completes the proof. \blacksquare

3 Proof of Theorem 1

Now let us prove Theorem 1. Let $\beta : \mathbf{N} \rightarrow (0, \infty)$ be such that

$$\frac{\beta(n)}{(\log n)^{1/2}} \rightarrow \infty, \quad n \rightarrow \infty.$$

Assume that Theorem 1 is not valid. Then there is a sequence of positive numbers $\{s'_n\}_{n=1}^\infty$ such that $s'_n \geq n^{1/2}\beta(n)$, $n = 1, 2, \dots$, and

$$\overline{\lim}_{n \rightarrow \infty} \left| \frac{P(\sum_{k=1}^n X_k > s'_n)}{n\tilde{F}(s'_n)} - 1 \right| > 0.$$

Let $s_n = n^{-1/2}s'_n \geq \beta(n)$. Let us take an $r \in ((\alpha + 2)/(2\alpha), 1)$ and fix it. Let t_n , $n = 1, 2, \dots$, be a sequence of positive numbers given by

$$t_n = (\log n)^{-1/2} + (\log n)^{-1}s_n^{(1+r)/2}, \quad n \geq 2.$$

Then we have the following.

$$t_n s_n \geq \frac{\beta(n)}{(\log n)^{1/2}} \rightarrow \infty, \quad n \rightarrow \infty, \quad (4)$$

$$\frac{2s_n}{t_n} \geq ((\log n)^{1/2} s_n) \wedge ((\log n) s_n^{(1-r)/2}), \quad n \geq 2, \quad (5)$$

$$\frac{2s_n}{(\log n)t_n} \geq \left(\frac{\beta(n)}{(\log n)^{1/2}}\right) \wedge (s_n^{(1-r)/2}) \rightarrow \infty, \quad n \rightarrow \infty, \quad (6)$$

and

$$\frac{(t_n n^{1/2})^2}{(s'_n)^{1+r}} \geq \frac{(\log n)^{-2} s_n^{1+r} n}{s_n^{1+r} n^{(1+r)/2}} \rightarrow \infty, \quad n \rightarrow \infty. \quad (7)$$

Therefore by Equation (7), we have

$$\frac{(F(-t_n n^{1/2}) + \bar{F}(t_n n^{1/2}))^2}{\bar{F}(s'_n)^{2r}} \rightarrow 0, \quad n \rightarrow \infty.$$

Since $2r - 1 > 2/\alpha$, we have

$$(s'_n)^2 \frac{(F(-t_n n^{1/2}) + \bar{F}(t_n n^{1/2}))^2}{\bar{F}(s'_n)} \rightarrow 0, \quad n \rightarrow \infty. \quad (8)$$

Also, by Equations (4), (5) and (6) we see that for any $m \geq 1$

$$\begin{aligned} & (ns'_n)^m \exp\left(\frac{m}{t_n^2} - \frac{1}{m} \frac{s_n}{t_n}\right) \\ &= \exp\left(\frac{m}{t_n^2} \left(1 - \frac{1}{3m^2} t_n s_n\right)\right) n^{3m/2} \exp\left(-(\log n) \frac{1}{3m} \frac{s_n}{(\log n)t_n}\right) \\ & \quad \times (s_n)^m \exp\left(-\frac{1}{3m} \frac{s_n}{t_n}\right) \rightarrow 0, \quad n \rightarrow \infty. \end{aligned}$$

Let us take an $\varepsilon \in (0, 1)$ and fix it. For $n \geq 1$, let

$$c_n(\varepsilon) = P\left(X_1 + \sum_{k=2}^n X_k 1_{\{|X_k| \leq t_n n^{1/2}\}} > s_n n^{1/2}, \left| \sum_{k=2}^n X_k 1_{\{|X_k| \leq t_n n^{1/2}\}} \right| \leq \varepsilon s_n n^{1/2}\right).$$

Note that $t_n < (1 - \varepsilon)s_n$ for sufficiently large n . Taking $\gamma = 2$ in Proposition 7 (2) we see that by Equation (8)

$$\frac{(s'_n)^2}{n} \frac{1}{n \bar{F}(s'_n)} \left| P\left(\sum_{k=1}^n X_k > s'_n\right) - nc_n(\varepsilon) \right| \rightarrow 0, \quad n \rightarrow \infty. \quad (9)$$

Note that

$$\bar{F}(s'_n(1 + \varepsilon)) - P\left(\left| \sum_{k=2}^n X_k 1_{\{|X_k| \leq t_n n^{1/2}\}} \right| > \varepsilon s_n n^{1/2}\right) \leq c_n(\varepsilon) \leq \bar{F}(s'_n(1 - \varepsilon)).$$

By Proposition 7 (1), we have

$$\frac{1}{\bar{F}(s'_n)} P\left(\left| \sum_{k=2}^n X_k 1_{\{|X_k| \leq t_n n^{1/2}\}} \right| > \varepsilon s_n n^{1/2}\right) \rightarrow 0, \quad n \rightarrow \infty.$$

Thus we have

$$(1 + \varepsilon)^{-\alpha} = \liminf_{n \rightarrow \infty} \frac{\bar{F}(s'_n(1 + \varepsilon))}{\bar{F}(s'_n)} \leq \liminf_{n \rightarrow \infty} \frac{c_n(\varepsilon)}{\bar{F}(s'_n)} \leq \liminf_{n \rightarrow \infty} \frac{P(\sum_{k=1}^n X_k > s'_n)}{n\bar{F}(s'_n)}$$

and

$$\liminf_{n \rightarrow \infty} \frac{P(\sum_{k=1}^n X_k > s'_n)}{n\bar{F}(s'_n)} \leq \liminf_{n \rightarrow \infty} \frac{c_n(\varepsilon)}{\bar{F}(s'_n)} \leq \liminf_{n \rightarrow \infty} \frac{\bar{F}(s'_n(1 - \varepsilon))}{\bar{F}(s'_n)} = (1 - \varepsilon)^{-\alpha}$$

Since $\varepsilon \in (0, 1)$ is arbitrary, we see that

$$\lim_{n \rightarrow \infty} \frac{P(\sum_{k=1}^n X_k > s'_n)}{n\bar{F}(s'_n)} = 1.$$

This contradicts our assumption.

This completes the proof of Theorem 1.

4 Some estimates

In this section, we assume that (A-1) and (A-4).

Let $g : (x_0, \infty) \rightarrow \mathbf{R}$ and $H : [-1/2, 1/2] \times (2x_0, \infty) \rightarrow (0, \infty)$ be given by

$$g(x) = x^2 \frac{d^2}{dx^2} (\log \bar{F})(x) - \alpha, \quad x > x_0,$$

and

$$H(y; x) = \frac{\bar{F}(x(1 + y))}{\bar{F}(x)}, \quad y \in [-1/2, 1/2], \quad x > 2x_0.$$

We prove the following in this section.

Proposition 8 *There are functions $a : (2x_0, \infty) \rightarrow \mathbf{R}$, $c : (2x_0, \infty) \rightarrow [0, \infty)$ and a constant $C > 0$ such that $a(x) \rightarrow 0$ and $c(x) \rightarrow 0$, as $x \rightarrow \infty$, and that*

$$|H(y; x) - \{1 + a(x)y - \alpha y + \frac{\alpha(\alpha + 1)y^2}{2}\}| \leq C(c(x)y^2 + |y|^3), \quad y \in [-1/2, 1/2], \quad x > 2x_0.$$

First we prove the following.

Proposition 9 (1) *For any $x > x_0$,*

$$\frac{d}{dx} \log(x^\alpha \bar{F}(x)) = - \int_x^\infty \frac{g(z)}{z^2} dz.$$

(2) *For any $y \in [-1/2, 1/2]$ and $x > 2x_0$,*

$$H(y; x) = (1 + y)^{-\alpha} \exp\left(- \int_0^y dy' \int_{1+y'}^\infty \frac{g(xz)}{z^2} dz\right).$$

Proof. Note that

$$g(x) = x^2 \frac{d^2}{dx^2} (\log(x^\alpha \bar{F}(x)))$$

and $g(x) \rightarrow 0$ as $x \rightarrow \infty$. Then we see that

$$\frac{d}{dy} (\log(y^\alpha \bar{F}(y))) - \frac{d}{dx} (\log(x^\alpha \bar{F}(x))) = \int_x^y \frac{g(z)}{z^2} dz, \quad (10)$$

and so we see that

$$c_0 = \lim_{y \rightarrow \infty} \frac{d}{dy} (\log(y^\alpha \bar{F}(y)))$$

exists. Note that

$$\exp\left(\int_x^{2x} \frac{d}{dy} (\log(y^\alpha \bar{F}(y))) dy\right) = \frac{L(2x)}{L(x)} \rightarrow 1, \quad x \rightarrow \infty.$$

So we see that $c_0 = 0$. Therefore letting $y \rightarrow \infty$ in Equation (10) we have the assertion (1).

By the assertion (1), we have

$$\frac{d}{dy} \log((1+y)^\alpha H(y; x)) = -x \int_{x(1+y)}^\infty \frac{g(z)}{z^2} dz = - \int_{1+y}^\infty \frac{g(xz)}{z^2} dz.$$

Since $H(0; x) = 1$, we have the assertion (2). ■

Proposition 10 *Let $\tilde{a} : (2x_0, \infty) \rightarrow \mathbf{R}$ and $\tilde{c} : (2x_0, \infty) \rightarrow \mathbf{R}$ be given by*

$$\tilde{a}(x) = \frac{d}{dy} ((1+y)^\alpha H(y, x))|_{y=0},$$

and

$$\tilde{c}(x) = \sup_{y \in [-1/2, 1/2]} \left| \frac{d^2}{dy^2} ((1+y)^\alpha H(y, x)) \right|.$$

Then $\tilde{a}(x) \rightarrow 0$ and $\tilde{c}(x) \rightarrow 0$, as $x \rightarrow \infty$, and that

$$|H(y; x) - (1+y)^{-\alpha} - \tilde{a}(x)y(1+y)^{-\alpha}| \leq 2^\alpha \tilde{c}(x)y^2, \quad y \in [-1/2, 1/2], \quad x > 2x_0.$$

Proof. By Proposition 9 We have

$$\frac{d}{dy} ((1+y)^\alpha H(y; x)) = -(1+y)^\alpha H(y; x) \int_{1+y}^\infty \frac{g(xz)}{z^2} dz$$

and so

$$\tilde{a}(x) = - \int_1^\infty \frac{g(xz)}{z^2} dz$$

Similarly, we have

$$\frac{d^2}{dy^2} ((1+y)^\alpha H(y; x)) = (1+y)^\alpha H(y; x) \left\{ \left(\int_{1+y}^\infty \frac{g(xz)}{z^2} dz \right)^2 - (1+y)^{-2} g(x(1+y)) \right\}$$

Therefore we have

$$\tilde{c}(x) \leq 2^\alpha \left\{ \left(\int_{1/2}^{\infty} \frac{|g(xz)|}{z^2} dz \right)^2 + 4 \sup\{|g(x(1+y))|; y \in [-1/2, 1/2]\} \right\} \exp\left(\int_{1/2}^{\infty} \frac{|g(xz)|}{z^2} dz \right)$$

These imply that $\tilde{a}(x) \rightarrow 0$, $\tilde{c}(x) \rightarrow 0$, as $x \rightarrow \infty$. Also we have

$$|(1+y)^\alpha H(y; x) - (1 + \tilde{a}(x))| \leq \tilde{c}(x)y^2, \quad x \geq 2x_0, y \in [-1/2, 1/2].$$

This implies our assertion. ■

Now Proposition 8 is an easy corollary to Proposition 10.

5 Proof of Theorem 3.

In this section, we assume that X_n , $n = 1, 2, \dots$, are i.i.d. random variables, $\alpha > 2$ and (A-1) - (A-4) are satisfied. Let $p = (\alpha + 2)/2$ and $\beta = (\alpha + p)/2$. Then we see that $E[|X_1|^p] < \infty$ and there is a $C_0 > 1$ such that

$$F(-x) + \bar{F}(x) \leq C_0 x^{-\beta}, \quad x \geq 1.$$

Proposition 11 *Let $b(x) = E[X_1, |X_1| \leq x] = -E[X_1, |X_1| > x]$, $x > 0$. Then we have the following.*

$$(1) |b(x)| \leq E[|X_1|^p]^{1/p} (F(-x) + \bar{F}(x))^{1-1/p} \leq C_0 x^{-\beta(p-1)/p} E[|X_1|^p]^{1/p}, \quad x \geq 1.$$

(2) *There is a constant $C_1 > 1$ only dependent on p such that*

$$\begin{aligned} E\left[\left| \sum_{k=1}^n X_k 1_{\{|X_k| \leq x\}} \right|^p \right]^{1/p} &\leq C_1 n^{1/2} (E[|X_1|^p]^{1/p} + |b(x)|) + n|b(x)| \\ &\leq C_1 E[|X_1|^p]^{1/p} (1 + C_0) (n^{1/2} + nx^{-\beta(p-1)/p}) \end{aligned}$$

for any $n = 1, 2, \dots$, and $x \geq 1$.

Proof. The assertion (1) is an easy consequence of Hölder's inequality. So we prove the assertion (2). Since $E[X_k 1_{\{|X_k| \leq x\}} - b(x)] = 0$, $k = 1, 2, \dots$, we see by Burkholder-Davis-Gundy's theorem that there is a constant $C_1 > 0$ depending on p only such that

$$E\left[\left| \sum_{k=1}^n (X_k 1_{\{|X_k| \leq x\}} - b(x)) \right|^p \right]^{1/p} \leq C_1 E\left[\left| \sum_{k=1}^n (X_k 1_{\{|X_k| \leq x\}} - b(x))^2 \right|^{p/2} \right]^{1/p}$$

Then by Hölder's inequality, we have

$$\begin{aligned} E\left[\left| \sum_{k=1}^n (X_k 1_{\{|X_k| \leq x\}} - b(x)) \right|^p \right]^{1/p} &\leq C_1 E[n^{p/2-1} \sum_{k=1}^n |X_k 1_{\{|X_k| \leq x\}} - b(x)|^p]^{1/p} \\ &= C_1 n^{1/2} E[|X_1 1_{\{|X_1| \leq x\}} - b(x)|^p]^{1/p} \leq C_1 n^{1/2} (E[|X_1 1_{\{|X_1| \leq x\}}|^p]^{1/p} + |b(x)|) \end{aligned}$$

This implies our assertion. ■

Remind Proposition 8 and let

$$R(y; x) = H(y; x) - \left\{1 + a(x)y - \alpha y + \frac{\alpha(\alpha + 1)y^2}{2}\right\}, \quad y \in [-1/2, 1/2], \quad x > 2x_0.$$

Then there is a $C_2 > 0$ such that

$$|R(y; x)| \leq C_2(c(x)y^2 + y^3), \quad y \in [-1/2, 1/2], \quad x > 2x_0.$$

Let

$$Y_n(t) = \sum_{k=2}^n X_k 1_{\{|X_k| \leq tn^{1/2}\}}, \quad n \geq 2, \quad t > 0.$$

Proposition 12 *Let $r \in ((\alpha + 2)/(2\alpha), 1)$. Then for any $\varepsilon \in (0, 1/2)$*

$$\begin{aligned} \overline{\lim}_{n \rightarrow \infty} \sup \{s^2 E[|H(-\frac{1}{sn^{1/2}}Y_n(t), sn^{1/2}), |Y_n(t)| \leq \varepsilon sn^{1/2}] - (1 + \frac{\alpha(\alpha + 1)}{2s^2n}v)|]; \\ s \geq (\log n)^{1/2}, \quad t \geq (\log n)^{-1}s^{(1+r)/2}\} = 0. \end{aligned}$$

Proof. Let $s \geq (\log n)^{1/2}$, $t \geq (\log n)^{-1}s^{(1+r)/2}$, and $n \geq 3$. Then $tn^{1/2} \geq 1$. We see that

$$\begin{aligned} & E[H(-\frac{1}{sn^{1/2}}Y_n(t), sn^{1/2}), |Y_n(t)| \leq \varepsilon sn^{1/2}] - (1 + \frac{\alpha(\alpha + 1)}{2s^2}v) \\ &= \frac{a(sn^{1/2}) - \alpha}{sn^{1/2}} E[Y_n(t)] + \frac{\alpha(\alpha + 1)}{2s^2n} (E[Y_n(t)^2] - nv) \\ &- E[1 + \frac{a(sn^{1/2}) - \alpha}{sn^{1/2}} Y_n(t) + \frac{\alpha(\alpha + 1)}{2s^2n} Y_n(t)^2, |(sn^{1/2})^{-1}Y_n(t)| > \varepsilon] \\ &+ E[R(-\frac{1}{sn^{1/2}}Y_n(t), sn^{1/2}), |(sn^{1/2})^{-1}Y_n(t)| \leq \varepsilon]. \end{aligned}$$

Note that

$$\begin{aligned} s|E[Y_n(t)]| &= ns|b(tn^{1/2})| \leq C_0 E[|X_1|^p]^{1/p} s (tn^{1/2})^{-\beta(p-1)/p} \\ &\leq C_0 E[|X_1|^p]^{1/p} (n^{1/2}(\log n)^{-1})^{-\beta(p-1)/p} s^{1-(1+r)\beta(p-1)/2p}, \\ E[Y_k(t)^2] - nv &= n(E[(X_1 1_{|X_1| \leq tn^{1/2}})^2] - b(tn^{1/2})^2) + E[Y_n(t)^2] - nv \\ &= -nE[X_1^2, |X_1| > tn^{1/2}] + n(n-1)b(tn^{1/2})^2, \end{aligned}$$

and

$$\begin{aligned} & n^{-p/2} E[|Y_n(t)|^p] \\ &\leq C_1^p (1 + C_0)^p E[|X_1|^p] (1 + t^{-\beta(p-1)/p}) n^{1/2(1-\beta(p-1)/p)p}. \end{aligned}$$

Also, note that

$$r\beta(p-1)/p > 1 + \frac{3(\alpha-2)}{8} > 1.$$

So we see that

$$\frac{1}{n} |E[Y_k(t)^2] - nv| \leq E[X_1^2, |X_1| > tn^{1/2}] + C_0(n-1)n^{-\beta(p-1)/p} (\log n)^{2\beta(p-1)/p} E[|X_1|^p]^{2/p},$$

$$\begin{aligned} & s^2(sn^{1/2})^{-k} E[|Y_n(t)|^k, |(sn^{1/2})^{-1}Y_n(t)| > \varepsilon] \\ & \leq s^{2-p}\varepsilon^{-p+k}n^{-p/2} E[|Y_n(t)|^p], \quad k = 0, 1, 2, \end{aligned}$$

and

$$\begin{aligned} & s^2 E[|R(-\frac{1}{sn^{1/2}}Y_n(t), sn^{1/2})|, |(sn^{1/2})^{-1}Y_n(t)| \leq \varepsilon], \\ & \leq C_2(|c(sn^{1/2})|n^{-1} E[Y_n(t)^2] + \varepsilon^{-p}s^{2-p}n^{-p/2} E[|Y_n(t)|^p]). \end{aligned}$$

Combining them, we have our assertion. ■

Now we prove Theorem 3. We use the same idea in Section 3. Let $\beta : \mathbf{N} \rightarrow (0, \infty)$ be such that

$$\frac{\beta(n)}{(\log n)^{1/2}} \rightarrow \infty, \quad n \rightarrow \infty.$$

Assume that Theorem 3 is not valid. Then there is a sequence of positive numbers $\{s'_n\}_{n=1}^\infty$ such that $s'_n \geq n^{1/2}\beta(n)$, $n = 1, 2, \dots$, and

$$\overline{\lim}_{n \rightarrow \infty} \frac{(s'_n)^2}{n} \left| \frac{P(\sum_{k=1}^n X_k > s'_n)}{n\bar{F}(s'_n)} - \left(1 + \frac{\alpha(\alpha+1)vn}{(s'_n)^2}\right) \right| > 0.$$

Let $s_n = n^{-1/2}s'_n \geq \beta(n)$. Let us take an $r \in ((\alpha+2)/(2\alpha), 1)$ and fix it. Let t_n , $n = 1, 2, \dots$, be a sequence of positive numbers given by

$$t_n = (\log n)^{-1/2} + (\log n)^{-1}s_n^{(1+r)/2}, \quad n \geq 2.$$

Then from Proposition 12, we see that

$$\begin{aligned} & s_n^2 \left| \frac{nP(X_1 + Y_n(t) > sn^{1/2}, |Y_n(t)| \leq \varepsilon sn^{1/2})}{n\bar{F}(s_n n^{1/2})} - \left(1 + \frac{\alpha(\alpha+1)vn}{s_n^2 n}\right) \right| \\ & = s_n^2 \left| E[H(-\frac{1}{sn^{1/2}}Y_n(t), sn^{1/2}), |Y_n(t)| \leq \varepsilon sn^{1/2}] - \left(1 + \frac{\alpha(\alpha+1)vn}{s_n^2 n}\right) \right| \rightarrow 0 \end{aligned}$$

as $n \rightarrow \infty$. Then Theorem 3 follows from this, Proposition 7(2), Equations (4), (5), (6), (7) and (8).

This completes the proof of Theorem 3.

6 Remarks

Let X_n , $n = 1, 2, \dots$, be independent identically distributed random variables. and $F : \mathbf{R} \rightarrow [0, 1]$ and $\bar{F} : \mathbf{R} \rightarrow [0, 1]$ be given by

$$F(x) = P(X_1 \leq x), \quad x \in \mathbf{R}$$

and

$$\bar{F}(x) = P(X_1 > x), \quad x \in \mathbf{R}.$$

Let us assume the following.

(B-1) $\bar{F}(x)$ is a regular varying function of index $-\alpha$, $\alpha \in (0, 2]$, as $x \rightarrow \infty$, i.e., if we let

$$L(x) = x^\alpha \bar{F}(x), \quad x \in \mathbf{R},$$

$L(x) > 0$ for any $x > 0$, and for any $a > 0$

$$\frac{L(ax)}{L(x)} \rightarrow 1, \quad x \rightarrow \infty.$$

(B-2) $x^{\alpha-\delta}F(-x) \rightarrow 0$, $x \rightarrow \infty$ for any $\delta > 0$.

In the case $\alpha > 1$, we see that $E[|X_1|] < \infty$. Let us assume furthermore that

(B-3) $E[X_1] = 0$ if $\alpha > 1$.

Then we have the following.

Theorem 13 *Assume the assumptions (B-1), (B-2) and (B-3). Then for any $\gamma \in (0, \alpha)$*

$$\sup_{s \geq n^{1/\gamma}} \left| \frac{P(\sum_{k=1}^n X_k > s)}{n\bar{F}(s)} - 1 \right| \rightarrow 0, \quad n \rightarrow \infty.$$

The proof is quite similar to that of Theorem 1.

The case that $\alpha \in (0, 1]$, we need the following propositions instead of Propositions 4 and 5.

Proposition 14 *Let Y be a random variable and $\gamma \in (0, 1]$. Assume that*

$$E[|Y|^\gamma] < \infty$$

Then for any $s \in \mathbf{R} \setminus \{0\}$ and $b > 0$

$$E[\exp(sY1_{\{|Y| \leq b\}})] \leq 1 + |s|^\gamma \exp(|s|b)E[|Y|^\gamma].$$

Proposition 15 *Let X be a random variable and $\gamma \in (0, 1]$. We assume that*

$$E[|X|^\gamma] < \infty.$$

Then for any $t > 0$

$$n \log E[\exp(\pm \frac{1}{tn^{1/\gamma}} X 1_{\{|X| \leq tn^{1/\gamma}\}})] \leq \frac{6}{t^\gamma} E[|X|^\gamma].$$

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