

On the rate of convergence in the global central limit theorem for random sums of uniformly strong mixing random variables

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Abstract. We present upper bounds of the integral $\int_{-\infty}^{\infty} |x|^l |\mathbf{P}\{Z_N < x\} - \Phi(x)| dx$ for $0 \leq l \leq 1 + \delta$, where $0 < \delta \leq 1$, $\Phi(x)$ is a standard normal distribution function, and $Z_N = S_N / \sqrt{\mathbf{E}S_N^2}$ is the normalized random sum with $\mathbf{E}S_N^2 > 0$ ($S_N = X_1 + \dots + X_N$) of centered random variables X_1, X_2, \dots satisfying the uniformly strong mixing condition. The number of summands N is a nonnegative integer-valued random variable independent of X_1, X_2, \dots .

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1 Introduction and main results

Let X_1, X_2, \dots be a sequence of real centered random variables (r.v.s). For $a \leq b$, we denote by \mathcal{F}_a^b the σ -algebra of events generated by r.v.s X_a, X_{a+1}, \dots, X_b . As usual, \mathbb{R} is the real line, $\mathbb{N} = \{1, 2, \dots\}$, $\mathbb{N}_0 = \{0, 1, 2, \dots\}$, and $\mathbf{1}_A$ is the indicator of an event A .

We consider the weak dependence condition defined between the “past” and “future” in terms of the uniformly strong mixing coefficient $\varphi(\tau)$ introduced by Ibragimov (1959): We say that a sequence of r.v.s X_1, X_2, \dots satisfies the uniformly strong mixing (u.s.m.) condition (or the φ -mixing condition) with the u.s.m. coefficient $\varphi(\tau)$ if

$$\varphi(\tau) = \sup_{t \in \mathbb{N}} \sup_{\substack{A \in \mathcal{F}_1^t, B \in \mathcal{F}_{t+\tau}^\infty \\ \mathbf{P}(A) > 0}} \frac{|\mathbf{P}(AB) - \mathbf{P}(A)\mathbf{P}(B)|}{\mathbf{P}(A)} \xrightarrow{\tau \rightarrow \infty} 0 \quad (1.1)$$

(see [4] or [5]).

In what follows, $\Phi(x)$ is the standard normal distribution function. By $C(\cdot)$ with an index or without it we denote a positive finite factor depending only on the quantities indicated in the parentheses (not necessarily the same at different occurrences).

Recall the following result for sums with a fixed number n of random summands satisfying the u.s.m. condition (1.1), which will be used to prove the corresponding results for random sums.

Theorem A. (See [15, Cor. 3].) *Let a sequence of r.v.s X_1, X_2, \dots with $\mathbf{E}X_i = 0$ and $\mathbf{E}|X_i|^{2+\delta} < \infty$, where $0 < \delta \leq 1$, $i = 1, \dots, n$, satisfy the u.s.m. condition (1.1) with coefficient $\varphi(\tau) \leq Ke^{-\mu\tau}$, where $0 < K < \infty$ and $\mu > 0$ are constants. Denote*

$$Z_n = \frac{S_n}{\sqrt{\mathbf{E}S_n^2}}, \quad S_n = \sum_{i=1}^n X_i, \quad \mathcal{I}_{l,n} = \int_{-\infty}^{\infty} |x|^l |\mathbf{P}\{Z_n < x\} - \Phi(x)| dx,$$

$$\lambda_{l,n} = |\mathbf{E}|Z_n|^l - \mathbf{E}|Y|^l|, \quad L_{r,n} = \frac{1}{(\mathbf{E}S_n^2)^{r/2}} \sum_{i=1}^n \mathbf{E}|X_i|^r,$$

where $\mathbf{E}S_n^2 > 0$, and Y is a standard normal r.v. Then

$$\mathcal{I}_{l,n} \leq C_0 L_{2+\delta,n} \ln^{1+\delta}(1+n) \tag{1.2}$$

if (i) $0 \leq l \leq 1$ or (ii) $1 < l \leq 1 + \delta$ and $L_{2,n} \leq C_*$; and

$$\lambda_{l,n} \leq C_0 L_{2+\delta,n} \ln^{1+\delta}(1+n)$$

if (i) $1 \leq l \leq 2$ or (ii) $2 < l \leq 2 + \delta$ and $L_{2,n} \leq C_*$.

Here $C_0 = C(K, \mu, l)$ in cases (i), and $C_0 = C(K, \mu, l, C_*)$ in cases (ii).

Recall that to prove Theorem A, we used the powerful and general direct Stein method introduced in [13] for estimating the rate of convergence of sums of weakly dependent r.v.s to the normal distribution.

In this paper, we are interested in estimates of the quantities

$$\mathcal{I}_l = \int_{-\infty}^{\infty} |x|^l |\mathbf{P}\{Z_N < x\} - \Phi(x)| dx, \quad \lambda_l = |\mathbf{E}|Z_N|^l - \mathbf{E}|Y|^l|,$$

where

$$Z_N = \frac{S_N}{\sqrt{\mathbf{E}S_N^2}}, \quad S_N = \sum_{i=1}^N X_i, \quad S_0 = 0,$$

assuming that $\mathbf{E}S_N^2 > 0$, the number of summands N is a nonnegative integer-valued r.v. independent of X_1, X_2, \dots , the centered summands X_1, X_2, \dots satisfy the u.s.m. condition (1.1), and Y is a standard normal r.v.

There are not many results on the convergence rate in the central limit theorem for random sums with weakly dependent summands. Strictly stationary sequences satisfying the u.s.m. condition (1.1), assuming that the number of summands and summands are dependent, were considered in [9]. Similar results for strictly stationary sequences of martingales have been obtained in [8]. A stationary sequence of m -dependent r.v.s, assuming that the number of summands and summands are independent, was investigated in the recent paper [10]. Without the convergence rate, the asymptotic normality of random sums of stationary m -dependent r.v.s was investigated in [12], in the recent paper [6], and that of martingales in [11].

For a wide range of various well-known and less common methods for estimation of the accuracy of probabilistic approximations, we refer to [2] and references therein. For the results on the convergence rate in the central limit theorem for weakly dependent random variables, we refer to [1, 7, 14] and references therein.

However, the author has not found any published results on the upper bounds of the quantities \mathcal{I}_l and λ_l , $l \geq 0$, for random sums with summands satisfying the u.s.m. condition (1.1). Note that for independent summands X_1, X_2, \dots , the corresponding upper estimates of \mathcal{I}_l for $0 \leq l \leq 1 + \delta$, where $0 < \delta \leq 1$ (and λ_l for $1 \leq l \leq 2 + \delta$) have been obtained in the recent paper [16].

To investigate the asymptotic normality and the convergence rate for random sums of independent (as well as dependent) summands, we use, as usual, the additional r.v.s

$$A_N = \sum_{i=1}^N \mathbf{E}X_i, \quad B_N^2 = \sum_{i=1}^N \mathbf{V}X_i, \quad l_{r,N} = \sum_{i=1}^N \mathbf{E}|X_i|^r.$$

In [17], seemingly for the first time, we introduced the additional r.v.s

$$\kappa_N^2 = \sum_{i=1}^N \sum_{j=1}^N \text{cov}(X_i, X_j) = B_N^2 + 2 \sum_{1 \leq i < j \leq N} \text{cov}(X_i, X_j),$$

which are very useful for investigating the asymptotics of the normality and the convergence rate for random sums of dependent (including weakly dependent) summands. Here $\text{cov}(\xi, \eta) = \mathbf{E}\xi\eta - \mathbf{E}\xi\mathbf{E}\eta$ is the covariance of real r.v.s ξ and η . Moreover, we assume that $\sum_{i=1}^0 (\cdot) = 0$.

The main results of this paper are Theorems 1–3.

Theorem 1. *Let a sequence of real r.v.s X_1, X_2, \dots with $\mathbf{E}X_i = 0$ and $\mathbf{E}|X_i|^{2+\delta} < \infty$, where $0 < \delta \leq 1$, $i = 1, 2, \dots$, satisfy the u.s.m. condition (1.1) with coefficient $\varphi(\tau) \leq Ke^{-\mu\tau}$, where $0 < K < \infty$ and $\mu > 0$ are constants. Let N be a nonnegative integer-valued r.v. independent of X_1, X_2, \dots . Then*

$$\mathcal{I}_l \leq C_1 \frac{\mathbf{E}l_{2+\delta,N} \ln^{1+\delta}(1+N)}{(\mathbf{E}\kappa_N^2)^{(2+\delta)/2}} + C_2 \frac{\mathbf{E}|\kappa_N^2 - \mathbf{E}\kappa_N^2|}{\mathbf{E}\kappa_N^2}, \tag{1.3}$$

for $0 \leq l \leq 1$; if, in addition, $L_{2,k} \leq C_*$ for $k = 1, 2, \dots$, then

$$\mathcal{I}_l \leq C_3 \frac{\mathbf{E}l_{l+1,N}}{(\mathbf{E}\kappa_N^2)^{(l+1)/2}} + C_4 \frac{\mathbf{E}l_{2+\delta,N} \ln^{1+\delta}(1+N)}{(\mathbf{E}\kappa_N^2)^{(2+\delta)/2}} + C_5 \frac{\mathbf{E}|\kappa_N^{l+1} - (\mathbf{E}\kappa_N^2)^{(l+1)/2}|}{(\mathbf{E}\kappa_N^2)^{(l+1)/2}} \tag{1.4}$$

for $1 < l \leq 1 + \delta$. Here the factors $C_1 = C_1(l, \delta, C_0)$, $C_2 = C_2(l)$, $C_3 = C(K, \mu, l, C_*)$, $C_4 = C(l, \delta, C_0)$, and $C_5 = C(K, \mu, l, C_*)$, where C_0 is taken from Theorem A.

In particular, if the summands are identically distributed with zero mixed moments, then we have the following result.

Corollary 1. *Let X, X_1, X_2, \dots be real identically distributed r.v.s with $\mathbf{E}X = 0$, $0 < \sigma^2 = \mathbf{E}X^2$, $\beta_{2+\delta} = \mathbf{E}|X|^{2+\delta} < \infty$, where $0 < \delta \leq 1$, and $\mathbf{E}X_i X_j = 0$, $1 \leq i \neq j \leq \infty$, satisfy the u.s.m. condition (1.1) with coefficient $\varphi(\tau) \leq Ke^{-\mu\tau}$, where $0 < K < \infty$ and $\mu > 0$ are constants. Let N be a nonnegative integer-valued r.v. with $\mathbf{E}N > 0$, independent of X_1, X_2, \dots . Then $Z_N = S_N/(\sigma\sqrt{\mathbf{E}N})$,*

$$\mathcal{I}_l \leq C_1 \frac{\beta_{2+\delta} \mathbf{E}N \ln^{1+\delta}(1+N)}{\sigma^{2+\delta} (\mathbf{E}N)^{(2+\delta)/2}} + C_2 \frac{\mathbf{E}|N - \mathbf{E}N|}{\mathbf{E}N} \tag{1.5}$$

for $0 \leq l \leq 1$ with $\mathbf{E}N \ln^{1+\delta}(1+N) < \infty$, and

$$\mathcal{I}_l \leq C_3 \frac{\beta_{l+1}}{\sigma^{l+1}} \frac{1}{(\mathbf{E}N)^{(l-1)/2}} + C_4 \frac{\beta_{2+\delta} \mathbf{E}N \ln^{1+\delta}(1+N)}{\sigma^{2+\delta} (\mathbf{E}N)^{(2+\delta)/2}} + C_5 \frac{\mathbf{E}|N^{(l+1)/2} - (\mathbf{E}N)^{(l+1)/2}|}{(\mathbf{E}N)^{(l+1)/2}}$$

for $1 < l \leq 1 + \delta$ with $\mathbf{E}N^{(l+1)/2} < \infty$.

By λ_l we denote the absolute value of the difference between the absolute moments of the random sum Z_N and the standard normal r.v. Y ,

$$\lambda_l = |\mathbf{E}|Z_N|^l - \mathbf{E}|Y|^l|.$$

The estimates of λ_l follow from the estimates of \mathcal{I}_l of Theorem 1. Namely, we have the following result.

Theorem 2. *Let the conditions of Theorem 1 hold. Then*

$$\lambda_l \leq C_6 \frac{\mathbf{E}l_{2+\delta,N} \ln^{1+\delta}(1+N)}{(\mathbf{E}\kappa_N^2)^{(2+\delta)/2}} + C_7 \frac{\mathbf{E}|\kappa_N^2 - \mathbf{E}\kappa_N^2|}{\mathbf{E}\kappa_N^2}$$

for $1 \leq l \leq 2$; if, in addition, $L_{2,k} \leq C_*$ for $k = 1, 2, \dots$, then

$$\lambda_l \leq C_8 \frac{\mathbf{E}l_{l,N}}{(\mathbf{E}\kappa_N^2)^{l/2}} + C_9 \frac{\mathbf{E}l_{2+\delta,N} \ln^{1+\delta}(1+N)}{(\mathbf{E}\kappa_N^2)^{(2+\delta)/2}} + C_{10} \frac{\mathbf{E}|\kappa_N^l - (\mathbf{E}\kappa_N^2)^{l/2}|}{(\mathbf{E}\kappa_N^2)^{l/2}}$$

for $2 < l \leq 2 + \delta$.

Here the factors $C_i = lC_{i-5}$, $i = 6, 7, 8, 9, 10$, where C_1, C_2, C_3, C_4 , and C_5 are taken from Theorem 1.

If, in addition, the summands are identically distributed with zero mixed moments, then from Theorem 2 we obtain the following result.

Corollary 2. *Let X, X_1, X_2, \dots be real identically distributed r.v.s with $\mathbf{E}X = 0$, $0 < \sigma^2 = \mathbf{E}X^2$, $\beta_{2+\delta} = \mathbf{E}|X|^{2+\delta} < \infty$, where $0 < \delta \leq 1$, and $\mathbf{E}X_i X_j = 0$, $1 \leq i \neq j \leq \infty$, satisfy the u.s.m. condition (1.1) with coefficient $\varphi(\tau) \leq K e^{-\mu\tau}$, where $0 < K < \infty$ and $\mu > 0$ are constants. Let N be a nonnegative integer-valued r.v. with $\mathbf{E}N > 0$, independent of X_1, X_2, \dots . Then $Z_N = S_N / (\sigma\sqrt{\mathbf{E}N})$,*

$$\lambda_l \leq C_6 \frac{\beta_{2+\delta} \mathbf{E}N \ln^{1+\delta}(1+N)}{\sigma^{2+\delta} (\mathbf{E}N)^{(2+\delta)/2}} + C_7 \frac{\mathbf{E}|N - \mathbf{E}N|}{\mathbf{E}N} \tag{1.6}$$

for $1 \leq l \leq 2$ with $\mathbf{E}N \ln^{1+\delta}(1+N) < \infty$, and

$$\lambda_l \leq C_8 \frac{\beta_l}{\sigma^l} \frac{1}{(\mathbf{E}N)^{(l-2)/2}} + C_9 \frac{\beta_{2+\delta} \mathbf{E}N \ln^{1+\delta}(1+N)}{\sigma^{2+\delta} (\mathbf{E}N)^{(2+\delta)/2}} + C_{10} \frac{\mathbf{E}|N^{l/2} - (\mathbf{E}N)^{l/2}|}{(\mathbf{E}N)^{l/2}}$$

for $2 < l \leq 2 + \delta$ with $\mathbf{E}N^{l/2} < \infty$.

To present the results for three concrete random indices N , we recall the definition of the τ -shifted \mathcal{L} distribution (τ -shifted Poisson distribution, τ -shifted binomial distribution, τ -shifted negative binomial distribution, and so on), which was first introduced in [16]. We write $\xi \sim \mathcal{L}$ if the distribution of a r.v. ξ is \mathcal{L} .

DEFINITION 1. We say that a discrete r.v. N is distributed by the τ -shifted \mathcal{L} distribution ($\tau \geq 0$) (for short, $N - \tau \sim \mathcal{L}$) or that N is a τ -shifted r.v. if for any discrete r.v. $\xi \sim \mathcal{L}$ taking values x_k with probabilities p_k ,

$$\mathbf{P}\{N = x_k + \tau\} = \mathbf{P}\{\xi = x_k\} = p_k.$$

In particular, the 0-shifted \mathcal{L} distribution coincides with the \mathcal{L} distribution.

DEFINITION 2. We say that a r.v. N is distributed by the τ -shifted Poisson distribution with parameters $\tau \in \mathbb{N}_0$ and $\lambda > 0$ (for short, $N - \tau \sim \mathcal{P}(\lambda)$) if

$$\mathbf{P}\{N = k + \tau\} = \frac{\lambda^k}{k!} e^{-\lambda}, \quad k = 0, 1, 2, \dots$$

DEFINITION 3. We say that a r.v. N is distributed by the τ -shifted binomial distribution with parameters $\tau \in \mathbb{N}_0$, $n \in \mathbb{N}$, and $0 < p < 1$ (for short, $N - \tau \sim \mathcal{B}(n, p)$) if

$$\mathbf{P}\{N = k + \tau\} = \binom{n}{k} p^k (1 - p)^{n-k}, \quad k = 0, 1, \dots, n.$$

DEFINITION 4. We say that a r.v. N is distributed by the τ -shifted negative binomial distribution with parameters $\tau \in \mathbb{N}_0$, $r \in \mathbb{N}$, and $0 < p < 1$ (for short, $N - \tau \sim \mathcal{NB}(r, p)$) if

$$\mathbf{P}\{N = k + \tau\} = \binom{k-1}{r-1} p^r (1-p)^{k-r}, \quad k = r, r+1, \dots$$

Now, we present the following statement for three presented τ -shifted \mathcal{L} distributions.

Theorem 3. Let X, X_1, X_2, \dots be real identically distributed r.v.s with $\mathbf{E}X = 0$, $0 < \sigma^2 = \mathbf{E}X^2$, $\beta_{2+\delta} = \mathbf{E}|X|^{2+\delta} < \infty$, where $0 < \delta \leq 1$, and $\mathbf{E}X_i X_j = 0$, $1 \leq i \neq j \leq \infty$, satisfy the u.s.m. condition (1.1) with coefficient $\varphi(\tau) \leq K e^{-\mu\tau}$, where $0 < K < \infty$ and $\mu > 0$ are constants. Let N be a nonnegative integer-valued r.v. independent of X_1, X_2, \dots . Then $Z_N = S_N / (\sigma\sqrt{\mathbf{E}N})$, and:

1. If $N - \tau \sim \mathcal{P}(\lambda)$ with $\tau \in \mathbb{N}_0$ and $\lambda > 0$, then

$$\mathcal{I}_l \leq C_{11} \frac{\ln^{1+\delta}(1 + \tau + \lambda)}{(\tau + \lambda)^{\delta/2}} \tag{1.7}$$

for $0 \leq l \leq 1$, and

$$\lambda_l \leq C_{12} \frac{\ln^{1+\delta}(1 + \tau + \lambda)}{(\tau + \lambda)^{\delta/2}} \tag{1.8}$$

for $1 \leq l \leq 2$.

2. If $N - \tau \sim \mathcal{B}(n, p)$ with $\tau \in \mathbb{N}_0$, $n \in \mathbb{N}$, and $0 < p < 1$, then

$$\mathcal{I}_l \leq C_{13} \frac{\ln^{1+\delta}(1 + \tau + np)}{(\tau + np)^{\delta/2}} \tag{1.9}$$

for $0 \leq l \leq 1$, and

$$\lambda_l \leq C_{14} \frac{\ln^{1+\delta}(1 + \tau + np)}{(\tau + np)^{\delta/2}} \tag{1.10}$$

for $1 \leq l \leq 2$.

3. If $N - \tau \sim \mathcal{NB}(r, p)$ with $\tau \in \mathbb{N}_0$, $r \in \mathbb{N}$, and $0 < p < 1$, then

$$\mathcal{I}_l \leq C_{15} \frac{\ln^{1+\delta}(1 + \tau + r/p)}{(\tau p + r)^{\delta/2}} \tag{1.11}$$

for $0 \leq l \leq 1$, and

$$\lambda_l \leq C_{16} \frac{\ln^{1+\delta}(1 + \tau + r/p)}{(\tau p + r)^{\delta/2}} \tag{1.12}$$

for $1 \leq l \leq 2$.

Here $C_i = C_i(K, \mu, \sigma, \beta_{2+\delta}, \delta)$, $i = 11, \dots, 16$.

Since the 0-shifted \mathcal{L} distribution coincides with the \mathcal{L} distribution, taking $\tau = 0$ in Theorem 3, we obtain the corresponding estimates of \mathcal{I}_l and λ_l for Poisson, binomial, and negative binomial random sums.

2 Auxiliary results

To prove the main results, we use some lemmas, which were also useful for estimating \mathcal{I}_l and λ_l for $l \geq 0$ for random sums of independent summands (see [16]) and for estimating the uniform metrics $\sup_{x \in \mathbb{R}} |\mathbf{P}\{Z_N < x\} - \Phi(x)|$ for random sums of summands satisfying the u.s.m. condition (1.1) (see [17]). Here we present them without proofs.

In the following lemma, we give a relationship between the first moments of the random sum S_N and the corresponding moment characteristics of r.v.s A_N and κ_N^2 .

Lemma 1. (See [17].) *Let X_1, X_2, \dots be (arbitrarily dependent, not necessarily identically distributed) r.v.s with $\mathbf{E}X_i^2 < \infty$, and let N be a nonnegative integer-valued r.v. independent of X_1, X_2, \dots . Denote*

$$S_N = \sum_{i=1}^N X_i, \quad A_N = \sum_{i=1}^N \mathbf{E}X_i, \quad B_N^2 = \sum_{i=1}^N \mathbf{V}X_i, \quad \kappa_N^2 = \sum_{i=1}^N \sum_{j=1}^N \text{cov}(X_i, X_j).$$

Then

$$\begin{aligned} \mathbf{E}S_N &= \mathbf{E}A_N, & \mathbf{E}S_N^2 &= \mathbf{E}\kappa_N^2 + \mathbf{E}A_N^2, \\ \mathbf{V}S_N &= \mathbf{E}\kappa_N^2 + \mathbf{V}A_N. \end{aligned} \tag{2.1}$$

If r.v.s X_1, X_2, \dots are independent, then

$$\mathbf{E}S_N^2 = \mathbf{E}B_N^2 + \mathbf{E}A_N^2, \quad \mathbf{V}S_N = \mathbf{E}B_N^2 + \mathbf{V}A_N.$$

In Lemma 2, we present the upper estimates of the second moment $\mathbf{E}S_N^2$ and variance $\mathbf{V}S_N$ of the random sum S_N with summands satisfying the u.s.m. condition.

Lemma 2. (See [17].) *Let a sequence of real r.v.s X_1, X_2, \dots satisfy the u.s.m. condition (1.1) with $\sum_{\tau=1}^{\infty} \varphi^{1/2}(\tau) < \infty$, and let N be a nonnegative integer-valued r.v. independent of X_1, X_2, \dots . Let the notation of Lemma 1 hold, and let*

$$b_N^2 = \sum_{i=1}^N \mathbf{E}X_i^2, \quad C_{\infty}(\varphi^{1/2}) = 1 + 4 \sum_{\tau=1}^{\infty} \varphi^{1/2}(\tau).$$

Then

$$\begin{aligned} \mathbf{E}S_N^2 &\leq C_{\infty}(\varphi^{1/2}) \mathbf{E}b_N^2 + \mathbf{E}A_N^2, \\ \mathbf{V}S_N &\leq C_{\infty}(\varphi^{1/2}) \mathbf{E}b_N^2 + \mathbf{V}A_N. \end{aligned} \tag{2.2}$$

We recall that to obtain inequality (2.2) while estimating $\mathbf{E}S_k^2$ with a fixed number $k = 1, 2, \dots$ of summands, it suffices to use inequality (1.3) in [5, p. 363, Lemma 1.1] or the inequality in [3, p. 278, Thm. A.6].

To transfer estimate (1.2) for random sums, we need Lemmas 3, 4, and 5.

Lemma 3. (See [16, Lemma 1].) *Let ξ and η be real r.v.s with $\mathbf{E}|\xi|^{l+1} < \infty$ and $\mathbf{E}|\eta|^{l+1} < \infty$ for some $l \geq 0$, respectively. Then*

$$\int_{-\infty}^{\infty} |x|^l |\mathbf{P}\{\xi < x\} - \mathbf{P}\{\eta < x\}| dx \leq \frac{\mathbf{E}|\xi|^{l+1} + \mathbf{E}|\eta|^{l+1}}{l + 1}.$$

In particular, if $\eta = Y$ is a standard normal r.v., then, for $l \geq 0$,

$$\int_{-\infty}^{\infty} |x|^l |\mathbf{P}\{\xi < x\} - \Phi(x)| dx \leq \begin{cases} \frac{1}{l+1}(\mathbf{E}|\xi|^{l+1} + \frac{2^{(l+1)/2}\Gamma((l+2)/2)}{\sqrt{\pi}}) & \text{if } l \geq 0, \\ \frac{1}{l+1}(\mathbf{E}|\xi|^{l+1} + 1) & \text{if } 0 \leq l \leq 1, \\ \frac{1}{l+1}(\mathbf{E}|\xi|^{l+1} + \frac{2\sqrt{2}}{\sqrt{\pi}}) & \text{if } 1 < l \leq 2, \end{cases} \quad (2.3)$$

where Γ is the gamma function. Moreover, if $\mathbf{E}\xi^2 = 1$, then, for all $0 \leq l \leq 1$,

$$\int_{-\infty}^{\infty} |x|^l |\mathbf{P}\{\xi < x\} - \Phi(x)| dx \leq \frac{2}{l + 1}. \quad (2.4)$$

Furthermore, to estimate $\mathcal{I}_{l,n}$ for $1 < l \leq 1 + \delta$, we use an estimate, which follows from estimate (58) in [15] under the condition of exponentially decreasing u.s.m. coefficient $\varphi(\tau)$ and truncation level $t = 1$.

Lemma 4. *Let X_1, X_2, \dots be a sequence of r.v.s with $\mathbf{E}X_i = 0$ and $\mathbf{E}|X_i|^\kappa < \infty$ for $i = 1, \dots, n$, where $\kappa \geq 2$, satisfying the u.s.m. condition (1.1) with $\varphi(\tau) \leq Ke^{-\mu\tau}$, where $0 < K < \infty$ and $\mu > 0$ are constants. Let $L_{2,n} = \sum_{i=1}^n \mathbf{E}X_i^2 / \mathbf{E}S_n^2 \leq C_*$. Denote*

$$\mathcal{I}_{l,n} = \int_{-\infty}^{\infty} |x|^l |\mathbf{P}\{Z_n < x\} - \Phi(x)| dx, \quad Z_n = \frac{S_n}{\sqrt{\mathbf{E}S_n^2}}, \quad S_n = \sum_{i=1}^n X_i,$$

where $\mathbf{E}S_n^2 > 0$. Then, for all $1 < l \leq \kappa - 1$,

$$\mathcal{I}_{l,n} \leq C_0 \left(1 + \frac{1}{(\mathbf{E}S_n^2)^{(l+1)/2}} \sum_{i=1}^n \mathbf{E}|X_i|^{l+1} \mathbf{1}_{\{|X_i| > \sqrt{\mathbf{E}S_n^2}\}} \right), \quad (2.5)$$

where $C_0 = C(K, \mu, l, C_*)$.

Lemma 5. (See [16].) *For all $a > 0$ and $l \geq 0$, we have*

$$\int_{-\infty}^{\infty} |x|^l |\Phi(xa) - \Phi(x)| dx = \frac{2^{(l+2)/2}\Gamma((l+2)/2)}{\sqrt{2\pi}} |a - 1| \int_0^1 \frac{dt}{(\gamma(t))^{(l+2)/2}} \quad (2.6)$$

$$\leq \frac{2^{(l+2)/2}\Gamma((l+2)/2)}{\sqrt{2\pi}(l+1)} \cdot \left| 1 - \frac{1}{a^{l+1}} \right|, \quad (2.7)$$

where $\gamma(t) = [1 + t(a - 1)]^2 > 0$.

In Lemmas 6, 7, and 8, we present some useful estimates of $(\mathbf{E}|N - \mathbf{E}N|)/\mathbf{E}N$, where the random number N is a τ -shifted Poisson r.v., a τ -shifted binomial r.v., and a τ -shifted negative binomial r.v., respectively.

Lemma 6. (See [17].) *If $N - \tau \sim \mathcal{P}(\lambda)$ with $\tau \in \mathbb{N}_0$ and $\lambda > 0$, then*

$$\frac{\mathbf{E}|N - \mathbf{E}N|}{\mathbf{E}N} \leq \frac{1}{\sqrt{\tau + \lambda}}. \tag{2.8}$$

Lemma 7. (See [17].) *If $N - \tau \sim \mathcal{B}(n, p)$ with $\tau \in \mathbb{N}_0$, $n \in \mathbb{N}$, and $0 < p < 1$, then*

$$\frac{\mathbf{E}|N - \mathbf{E}N|}{\mathbf{E}N} < \frac{1}{\sqrt{\tau + np}}. \tag{2.9}$$

Lemma 8. (See [17].) *If $N - \tau \sim \mathcal{NB}(r, p)$ with $\tau \in \mathbb{N}_0$, $r \in \mathbb{N}$, and $0 < p < 1$, then*

$$\frac{\mathbf{E}|N - \mathbf{E}N|}{\mathbf{E}N} < \frac{1}{\sqrt{\tau p + r}}. \tag{2.10}$$

3 Basic inequality for \mathcal{I}_l with $\mathbf{E}X_i = 0, i = 1, 2, \dots$

Denote

$$\begin{aligned} \mathcal{I}_l &= \int_{-\infty}^{\infty} |x|^l |\Delta(x)| dx, & \Delta(x) &= \mathbf{P}\{S_N < x\sqrt{\mathbf{E}S_N^2}\} - \Phi(x), & S_N &= \sum_{i=1}^N X_i, \\ \xi_k &= \frac{S_k}{\sqrt{\mathbf{E}S_k^2}}, & a_k &= \frac{\sqrt{\mathbf{E}S_N^2}}{\sqrt{\mathbf{E}S_k^2}}, & S_k &= \sum_{i=1}^k X_i, \quad k = 1, 2, \dots, \end{aligned}$$

where X_1, X_2, \dots are (arbitrarily dependent) r.v.s, $\mathbf{E}S_N^2 > 0$, and $S_0 = 0$. It is clear that if N is a nonnegative integer-valued r.v. with $p_k = \mathbf{P}\{N = k\}, k = 0, 1, 2, \dots$, independent of X_1, X_2, \dots , then, for all $x \in \mathbb{R}$,

$$\Delta(x) = \sum_{k=0}^{\infty} [\mathbf{P}\{S_k < x\sqrt{\mathbf{E}S_N^2}\} - \Phi(x)] p_k.$$

Let

$$K(\alpha) = \{k \in \mathbb{N} : |\mathbf{E}S_k^2 - \mathbf{E}S_N^2| \leq (1 - \alpha)\mathbf{E}S_N^2\}$$

and

$$\bar{K}(\alpha) = \{k \in \mathbb{N} : |\mathbf{E}S_k^2 - \mathbf{E}S_N^2| > (1 - \alpha)\mathbf{E}S_N^2\}$$

for $\alpha \in (0, 1)$.

First, we observe that $\mathbf{E}S_k^2 \geq \alpha\mathbf{E}S_N^2 > 0$ for $k \in K(\alpha)$. Since

$$[\mathbf{P}\{S_0 < x\sqrt{\mathbf{E}S_N^2}\} - \Phi(x)] p_0 = [\mathbf{1}_{\{x>0\}} - \Phi(x)] p_0$$

and

$$\begin{aligned} & \sum_{k=1}^{\infty} [\mathbf{P}\{S_k < x\sqrt{\mathbf{E}S_N^2}\} - \Phi(x)]p_k \\ &= \sum_{k \in K(\alpha)} [\mathbf{P}\{\xi_k < xa_k\} - \Phi(xa_k)]p_k + \sum_{k \in K(\alpha)} [\Phi(xa_k) - \Phi(x)]p_k \\ &+ \sum_{k \in \bar{K}(\alpha)} [\mathbf{P}\{\xi_k < xa_k\} - \Phi(x)]p_k, \end{aligned}$$

we can state the following:

Basic inequality for \mathcal{I}_l . Let X_1, X_2, \dots be (arbitrarily dependent, not necessarily identically distributed) r.v.s with $\mathbf{E}X_i = 0$ for all $i = 1, 2, \dots$, and let N be a nonnegative integer-valued r.v. with $p_k = \mathbf{P}\{N = k\}$, $k = 0, 1, 2, \dots$, independent of X_1, X_2, \dots . Then, for all $l \geq 0$,

$$\mathcal{I}_l \leq \sum_1 + \sum_2 + \sum_3 + \sum_4, \tag{3.1}$$

where

$$\begin{aligned} \sum_1 &= \sum_{k \in K(\alpha)} \int_{-\infty}^{\infty} |x|^l |\mathbf{P}\{\xi_k < xa_k\} - \Phi(xa_k)| dx p_k, \\ \sum_2 &= \sum_{k \in K(\alpha)} \int_{-\infty}^{\infty} |x|^l |\Phi(xa_k) - \Phi(x)| dx p_k, \\ \sum_3 &= \int_{-\infty}^{\infty} |x|^l |\mathbf{1}_{\{x>0\}} - \Phi(x)| dx p_0, \\ \sum_4 &= \sum_{k \in \bar{K}(\alpha)} \int_{-\infty}^{\infty} |x|^l |\mathbf{P}\{S_k < x\sqrt{\mathbf{E}S_N^2}\} - \Phi(x)| dx p_k, \end{aligned}$$

where $K(\alpha) = \{k \in \mathbb{N}: |\mathbf{E}S_k^2 - \mathbf{E}S_N^2| \leq (1 - \alpha)\mathbf{E}S_N^2\}$ and $\bar{K}(\alpha) = \{k \in \mathbb{N}: |\mathbf{E}S_k^2 - \mathbf{E}S_N^2| > (1 - \alpha)\mathbf{E}S_N^2\}$ for $\alpha \in (0, 1)$.

4 Proofs of Theorems 1–3 and Corollaries 1, 2

In what follows, the sequence of r.v.s X_1, X_2, \dots satisfies the u.s.m. condition (1.1).

First of all, note that under the condition $\mathbf{E}X_i = 0$ for all $i = 1, 2, \dots$,

$$\kappa_N^2 = \sum_{i=1}^N \sum_{j=1}^N \mathbf{E}X_i X_j = b_N^2 + 2 \sum_{1 \leq i < j \leq N} \mathbf{E}X_i X_j,$$

where $b_N^2 = \sum_{i=1}^N \mathbf{E}X_i^2$.

Proof of Theorem 1. For the sum ξ_k with fixed number k of summands, we use the notation

$$\mathcal{I}_{l,k} = \int_{-\infty}^{\infty} |x|^l |\mathbf{P}\{\xi_k < x\} - \Phi(x)| dx, \quad k = 1, 2, \dots$$

Estimation of \sum_1 . First, we observe that

$$\sum_1 = \sum_{k \in K(\alpha)} \frac{1}{a_k^{l+1}} \mathcal{I}_{l,k} p_k. \tag{4.1}$$

Since $a_k \geq 1/\sqrt{2-\alpha}$ and $\mathbf{E}S_k^2 \geq \alpha \mathbf{E}S_N^2$ for $k \in K(\alpha)$, using (1.2) of Theorem A, from (4.1) we obtain that, for all $0 \leq l \leq 1 + \delta$,

$$\begin{aligned} \sum_1 &\leq (2-\alpha)^{(l+1)/2} C_0 \sum_{k \in K(\alpha)} \frac{1}{(\mathbf{E}S_k^2)^{(2+\delta)/2}} \sum_{i=1}^k \mathbf{E}|X_i|^{2+\delta} \ln^{1+\delta}(1+k) p_k \\ &\leq \frac{(2-\alpha)^{(l+1)/2} C_0 \mathbf{E}l_{2+\delta,N} \ln^{1+\delta}(1+N) \mathbf{1}_{\{N \in K(\alpha)\}}}{\alpha^{(2+\delta)/2} (\mathbf{E}S_N^2)^{(2+\delta)/2}}, \end{aligned} \tag{4.2}$$

where $l_{2+\delta,k} = \sum_{i=1}^k \mathbf{E}|X_i|^{2+\delta}$ and C_0 is taken from (1.2) of Theorem A.

Estimation of \sum_2 . To estimate \sum_2 , using (2.6) of Lemma 5, we get that, for all $l \geq 0$,

$$\int_{-\infty}^{\infty} |x|^l |\Phi(x a_k) - \Phi(x)| dx = \frac{2^{(l+2)/2} \Gamma((l+2)/2)}{\sqrt{2\pi}} |a_k - 1| \int_0^1 \frac{dt}{(\gamma_k(t))^{(l+2)/2}}, \tag{4.3}$$

where $\gamma_k(t) = [1 + t(a_k - 1)]^2$.

Since

$$\frac{1}{\sqrt{2-\alpha}} \leq a_k \leq \frac{1}{\sqrt{\alpha}}$$

for $k \in K(\alpha)$, we easily obtain that

$$\gamma_k(t) \geq \frac{1}{2-\alpha} \tag{4.4}$$

for $0 \leq t \leq 1$, $\alpha \in (0, 1)$, and $k \in K(\alpha)$ (see [16, p. 255]).

The upper bound of $|a_k - 1|$ for $k \in K(\alpha)$ easily follows:

$$|a_k - 1| = \frac{|\mathbf{E}S_k^2 - \mathbf{E}S_N^2|}{\sqrt{\mathbf{E}S_k^2}(\sqrt{\mathbf{E}S_k^2} + \sqrt{\mathbf{E}S_N^2})} \leq \frac{1}{\alpha + \sqrt{\alpha}} \frac{|\mathbf{E}S_k^2 - \mathbf{E}S_N^2|}{\mathbf{E}S_N^2}. \tag{4.5}$$

Substituting (4.5) and (4.4) into (4.3), we obtain that, for all $l \geq 0$ and $k \in K(\alpha)$,

$$\int_{-\infty}^{\infty} |x|^l |\Phi(x a_k) - \Phi(x)| dx \leq \frac{[2(2-\alpha)]^{(l+2)/2} \Gamma((l+2)/2)}{\sqrt{2\pi}(\alpha + \sqrt{\alpha})} \frac{|\kappa_k^2 - \mathbf{E}S_N^2|}{\mathbf{E}S_N^2}. \tag{4.6}$$

Here we used the fact that $\mathbf{E}S_k^2 = \kappa_k^2$ for $k = 1, 2, \dots$ if $\mathbf{E}X_i = 0$ for $i = 1, 2, \dots$. It only remains to substitute (4.6) into the expression of \sum_2 . We obtain that, for all $l \geq 0$,

$$\sum_2 \leq \frac{[2(2 - \alpha)]^{(l+2)/2} \Gamma((l+2)/2) \mathbf{E}|\kappa_N^2 - \mathbf{E}S_N^2| \mathbf{1}_{\{N \in K(\alpha)\}}}{\sqrt{2\pi}(\alpha + \sqrt{\alpha}) \mathbf{E}S_N^2}. \tag{4.7}$$

Estimation of $\sum_3 + \sum_4$. Taking into account that $S_0 = 0$, from (2.3) of Lemma 3 the estimate of \sum_3 for $0 \leq l \leq 2$ follows:

$$\sum_3 \leq \frac{p_0}{l+1} \cdot \begin{cases} 1 & \text{if } 0 \leq l \leq 1, \\ \frac{2\sqrt{2}}{\sqrt{\pi}} & \text{if } 1 < l \leq 2. \end{cases} \tag{4.8}$$

We easily see that

$$\sum_4 \leq \sum_{41} + \sum_{42} + \sum_{43}, \tag{4.9}$$

where

$$\begin{aligned} \sum_{41} &= \sum_{k \in \overline{K}^-(\alpha)} \frac{1}{a_k^{l+1}} \mathcal{I}_{l,k} p_k, & \sum_{42} &= \sum_{k \in \overline{K}^+(\alpha)} \frac{1}{a_k^{l+1}} \mathcal{I}_{l,k} p_k, \\ \sum_{43} &= \sum_{k \in \overline{K}(\alpha) - \infty}^{\infty} \int |x|^l |\Phi(xa_k) - \Phi(x)| dx p_k, \end{aligned}$$

and $\overline{K}(\alpha) = \overline{K}^-(\alpha) \cup \overline{K}^+(\alpha)$ is rewritten as the union of $\overline{K}^-(\alpha) = \{k \in \mathbb{N} : \mathbf{E}S_k^2 < \alpha \mathbf{E}S_N^2\}$ and $\overline{K}^+(\alpha) = \{k \in \mathbb{N} : \mathbf{E}S_k^2 > (2 - \alpha) \mathbf{E}S_N^2\}$.

Since $1/a_k \leq \sqrt{\alpha}$ for $k \in \overline{K}^-(\alpha)$ and $\mathcal{I}_{l,k} \leq 2/(l+1)$ for $0 \leq l \leq 1$ by (2.4) of Lemma 3, using (4.8), we obtain that, for $0 \leq l \leq 1$,

$$\begin{aligned} \sum_3 + \sum_{41} &\leq \frac{1}{l+1} p_0 + \frac{2\alpha^{(l+1)/2}}{l+1} \sum_{k \in \overline{K}^-(\alpha)} p_k \\ &\leq \max\{1, 2\alpha^{(l+1)/2}\} \frac{1}{l+1} \sum_{k \geq 0 : |\kappa_k^2 - \mathbf{E}S_N^2| > (1-\alpha)\mathbf{E}S_N^2} p_k \\ &\leq \max\{1, 2\alpha^{(l+1)/2}\} \frac{1}{(l+1)(1-\alpha)} \frac{\mathbf{E}|\kappa_N^2 - \mathbf{E}S_N^2| \mathbf{1}_{\{N \in \overline{K}(\alpha) \cup \{0\}\}}}{\mathbf{E}S_N^2}. \end{aligned} \tag{4.10}$$

Now let $1 < l \leq 1 + \delta$, $0 < \delta \leq 1$. In this case, instead of (2.4) of Lemma 3, we use (2.5) of Lemma 4, whereby for any fixed $k = 1, 2, \dots$,

$$\mathcal{I}_{l,k} \leq C_0 \left(1 + \frac{1}{(\mathbf{E}S_k^2)^{(l+1)/2}} \sum_{i=1}^k \mathbf{E}|X_i|^{l+1} \mathbf{1}_{\{|X_i| > \sqrt{\mathbf{E}S_k^2}\}} \right),$$

and therefore

$$\frac{1}{a_k^{l+1}} \mathcal{I}_{l,k} \leq C_0 \frac{1}{a_k^{l+1}} + C_0 \frac{1}{(\mathbf{E}S_N^2)^{(l+1)/2}} \sum_{i=1}^k \mathbf{E}|X_i|^{l+1} \mathbf{1}_{\{|X_i| > \sqrt{\mathbf{E}S_k^2}\}}. \tag{4.11}$$

Thus, using (4.8) and (4.11) and recalling that $1/a_k \leq \sqrt{\alpha}$ for $k \in \overline{K}^-(\alpha)$, we obtain that, for $1 < l \leq 1 + \delta$, $0 < \delta \leq 1$,

$$\begin{aligned} \sum_3 + \sum_{41} &\leq \frac{2\sqrt{2}p_0}{\sqrt{\pi}(l+1)} + C_0\alpha^{(l+1)/2} \sum_{k \in \overline{K}^-(\alpha)} p_k \\ &\quad + C_0 \frac{1}{(\mathbf{E}S_N^2)^{(l+1)/2}} \sum_{k \in \overline{K}^-(\alpha)} \sum_{i=1}^k \mathbf{E}|X_i|^{l+1} \mathbf{1}_{\{|X_i| > \sqrt{\mathbf{E}S_k^2}\}} p_k \\ &\leq \max \left\{ \frac{2\sqrt{2}}{\sqrt{\pi}(l+1)}, C_0\alpha^{(l+1)/2} \right\} \sum_{k \in \overline{K}^-(\alpha) \cup \{0\}} p_k + C_0 \frac{\mathbf{E}l_{l+1,N} \mathbf{1}_{\{N \in \overline{K}^-(\alpha)\}}}{(\mathbf{E}S_N^2)^{(l+1)/2}} \\ &\leq \max \left\{ \frac{2\sqrt{2}}{\sqrt{\pi}(l+1)}, C_0\alpha^{(l+1)/2} \right\} \frac{1}{1-\alpha} \frac{\mathbf{E}|\kappa_N^2 - \mathbf{E}S_N^2| \mathbf{1}_{\{N \in \overline{K}^-(\alpha) \cup \{0\}\}}}{\mathbf{E}S_N^2} \\ &\quad + C_0 \frac{\mathbf{E}l_{l+1,N} \mathbf{1}_{\{N \in \overline{K}^-(\alpha)\}}}{(\mathbf{E}S_N^2)^{(l+1)/2}}. \end{aligned} \tag{4.12}$$

To estimate \sum_{42} , we use (1.2) of Theorem A. Since $\mathbf{E}S_k^2 > (2 - \alpha)\mathbf{E}S_N^2$ for $k \in \overline{K}^+(\alpha)$, we obtain that, for all $0 \leq l \leq 1 + \delta$, $0 < \delta \leq 1$,

$$\begin{aligned} \sum_{42} &\leq C_0 \sum_{k \in \overline{K}^+(\alpha)} \frac{1}{a_k^{l+1}} \frac{1}{(\mathbf{E}S_k^2)^{(2+\delta)/2}} \sum_{i=1}^k \mathbf{E}|X_i|^{2+\delta} \ln^{1+\delta}(1+k)p_k \\ &= C_0 \sum_{k \in \overline{K}^+(\alpha)} \frac{1}{(\mathbf{E}S_N^2)^{(l+1)/2}} \frac{1}{(\mathbf{E}S_k^2)^{(1+\delta-l)/2}} \sum_{i=1}^k \mathbf{E}|X_i|^{2+\delta} \ln^{1+\delta}(1+k)p_k \\ &\leq C_0 \sum_{k \in \overline{K}^+(\alpha)} \frac{1}{(\mathbf{E}S_N^2)^{(l+1)/2}} \frac{\sum_{i=1}^k \mathbf{E}|X_i|^{2+\delta} \ln^{1+\delta}(1+k)p_k}{((2-\alpha)\mathbf{E}S_N^2)^{(1+\delta-l)/2}} \\ &= \frac{C_0}{(2-\alpha)^{(1+\delta-l)/2}} \frac{\mathbf{E}l_{2+\delta,N} \ln^{1+\delta}(1+N) \mathbf{1}_{\{N \in \overline{K}^+(\alpha)\}}}{(\mathbf{E}S_N^2)^{(2+\delta)/2}}. \end{aligned} \tag{4.13}$$

To estimate \sum_{43} , we use (2.7) of Lemma 5 and obtain that, for all $l \geq 0$,

$$\begin{aligned} \sum_{43} &\leq \frac{2^{(l+2)/2} \Gamma((l+2)/2)}{\sqrt{2\pi}(l+1)} \sum_{k \in \overline{K}(\alpha)} \left| 1 - \frac{1}{a_k^{l+1}} \right| p_k \\ &\leq \frac{2^{(l+2)/2} \Gamma((l+2)/2)}{\sqrt{2\pi}(l+1)} \frac{\mathbf{E}|\kappa_N^{l+1} - (\mathbf{E}S_N^2)^{(l+1)/2}| \mathbf{1}_{\{N \in \overline{K}(\alpha)\}}}{(\mathbf{E}S_N^2)^{(l+1)/2}}. \end{aligned} \tag{4.14}$$

Substituting (4.10) in the case $0 \leq l \leq 1$ ((4.12) in the case $1 < l \leq 1 + \delta$), (4.13), and (4.14) into (4.9) and observing that the function $f(l) = |1 - 1/a^{l+1}|$, where $0 < a < \infty$, is nondecreasing for $l \in [-1, \infty)$, we

obtain that

$$\begin{aligned} \sum_3 + \sum_4 &\leq \left(\max\{1, 2\alpha^{(l+1)/2}\} \frac{1}{1-\alpha} + \frac{2^{(l+2)/2}\Gamma((l+2)/2)}{\sqrt{2\pi}} \right) \\ &\quad \times \frac{1}{l+1} \frac{\mathbf{E}|\kappa_N^2 - \mathbf{E}S_N^2| \mathbf{1}_{\{N \in \overline{K}(\alpha) \cup \{0\}\}}}{\mathbf{E}S_N^2} \\ &\quad + \frac{C_0}{(2-\alpha)^{(1+\delta-l)/2}} \frac{\mathbf{E}l_{2+\delta,N} \ln^{1+\delta}(1+N) \mathbf{1}_{\{N \in \overline{K}^+(\alpha)\}}}{(\mathbf{E}S_N^2)^{(2+\delta)/2}} \end{aligned} \tag{4.15}$$

for $0 \leq l \leq 1$ and

$$\begin{aligned} \sum_3 + \sum_4 &\leq \left(\max\left\{ \frac{2\sqrt{2}}{\sqrt{\pi}}, C_0 \alpha^{(l+1)/2} \right\} \frac{1}{1-\alpha} + \frac{2^{(l+2)/2}\Gamma((l+2)/2)}{\sqrt{2\pi}} \right) \\ &\quad \times \frac{1}{l+1} \frac{\mathbf{E}|\kappa_N^{l+1} - (\mathbf{E}S_N^2)^{(l+1)/2}| \mathbf{1}_{\{N \in \overline{K}(\alpha) \cup \{0\}\}}}{(\mathbf{E}S_N^2)^{(l+1)/2}} + C_0 \frac{\mathbf{E}l_{l+1,N} \mathbf{1}_{\{N \in \overline{K}^-(\alpha)\}}}{(\mathbf{E}S_N^2)^{(l+1)/2}} \\ &\quad + \frac{C_0}{(2-\alpha)^{(1+\delta-l)/2}} \frac{\mathbf{E}l_{2+\delta,N} \ln^{1+\delta}(1+N) \mathbf{1}_{\{N \in \overline{K}^+(\alpha)\}}}{(\mathbf{E}S_N^2)^{(2+\delta)/2}} \end{aligned} \tag{4.16}$$

for $1 < l \leq 1 + \delta, 0 < \delta \leq 1$.

Substituting (4.2), (4.7), and (4.15) for $0 \leq l \leq 1$ ((4.16) for $1 < l \leq 1 + \delta$) into (3.1), taking into account that $\mathbf{E}S_N^2 = \mathbf{E}\kappa_N^2$ in the case of $\mathbf{E}X_i = 0$ for $i = 1, 2, \dots$ and $\sum_{i=1}^0(\cdot) = 0$ (see (2.1) of Lemma 1), and taking a concrete $\alpha \in (0, 1)$, for example, $\alpha = 1/2$, we obtain estimates (1.3) and (1.4) of Theorem 1.

Theorem 1 is proved. \square

Proof of Corollary 1. The proof immediately follows from Theorem 1. \square

Proof of Theorem 2. The proof immediately follows from Theorem 1 since for all $l \geq 1$,

$$\lambda_l \leq l\mathcal{I}_{l-1}. \quad \square \tag{4.17}$$

Proof of Corollary 2. The proof immediately follows from Theorem 2. \square

Proof of Theorem 3. Since $\mathcal{I}_l \leq 2/(l+1)$ for $0 \leq l \leq 1$ and $\lambda_l \leq 2$ for $1 \leq l \leq 2$ (see (2.4) and (4.17)), we assume, without loss of generality, that $\mathbf{E}N$ is sufficiently large. To estimate the first terms in (1.5) of Corollary 1 and in (1.6) of Corollary 2, we use the estimate

$$\mathbf{E}N \ln^{1+\delta}(1+N) \leq \sqrt{\mathbf{V}N} \mathbf{E}^{1/2} \ln^{2(1+\delta)}(1+N) + \mathbf{E}N \mathbf{E} \ln^{1+\delta}(1+N). \tag{4.18}$$

Now observing that the functions $f_1(x) = \ln^{2(1+\delta)}(e^{1+2\delta} + 1 + x)$ and $f_2(x) = \ln^{1+\delta}(e^\delta + 1 + x)$, where $0 < \delta \leq 1$, are strictly concave for all $x \in (-1, \infty)$, we obtain by Jensen's inequality that

$$\mathbf{E} \ln^{2(1+\delta)}(1+N) < \ln^{2(1+\delta)}(e^{1+2\delta} + 1 + \mathbf{E}N), \tag{4.19}$$

$$\mathbf{E} \ln^{1+\delta}(1+N) < \ln^{1+\delta}(e^\delta + 1 + \mathbf{E}N). \tag{4.20}$$

Now substituting (4.19) and (4.20) into (4.18), substituting the obtained inequality into (1.5) and (1.6) of Corollaries 1 and 2, respectively, and estimating, in the corresponding cases of the number N of summands, the second terms in (1.5) and (1.6) by (2.8) of Lemma 6, by (2.9) of Lemma 7, and by (2.10) of Lemma 8, we obtain (1.7)–(1.12) of Theorem 3.

Theorem 3 is proved. \square

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