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Asymptotics for sums of a function of normalized independent sums

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Abstract

We derive a central limit theorem for sums of a function of independent sums of independent and identically distributed random variables. In particular we show that previously known result from Rempała and Wesolowski (Statist. Probab. Lett. 74, 129), which can be obtained by applying the logarithm as the function, holds true under weaker assumptions.

Key words: Products of sums of iid rv's, Limit distribution, Central limit theorem, Lognormal distribution

2000 MSC: G0F05

1. Introduction

Throughout this paper let $(X_{k,n})_{k=1,\dots,n}$; $n = 1, 2, \dots$ be a triangular array of independent and identically distributed (iid) random variables (rv's) with the same distribution as X . Let us define the (mutually independent) partial sums $S_{n,n} = \sum_{k=1}^n X_{k,n}$. We will work with real functions f defined at least on an interval I such that $\mathbb{P}(X \in I) = 1$. We will also write $\log^+ x$ for $\log(x \vee 1)$.

The asymptotic behavior of the product of partial sums of a sequence of iid rv's has been studied in several papers (see e.g. Lu and Qi (2004) for a brief review). In particular it was shown in Rempała and Wesolowski (2002) that if (X_n) is a sequence of iid positive square integrable rv's with $\mathbb{E}X_1 = \mu$, $\text{Var}(X_1) = \sigma^2 > 0$, then setting $S_n = \sum_{k=1}^n X_k$ and $\gamma = \sigma/\mu$ we have as

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$n \rightarrow \infty$

$$\left(\prod_{k=1}^n \frac{S_k}{k\mu} \right)^{\frac{1}{\gamma\sqrt{n}}} \xrightarrow{d} e^{\sqrt{2}\mathcal{N}},$$

where $\xrightarrow{d}$ stands for convergence in distribution and $\mathcal{N}$ is a standard normal random variable. This result was extended in Qi (2003) and Lu and Qi (2004) to a general limit theorem covering the case when the underlying distribution is integrable and belongs to the domain of attraction of a stable law with index from the interval $[1, 2]$.

This study brought an interest to the array case, where we no longer consider a sequence (X_n) but a triangular array $(X_{k,n})$. In Rempała and Wesolowski (2005) the analogous result was obtained, namely

$$\left(n^{\frac{\gamma^2}{2}} \prod_{k=1}^n \frac{S_{k,k}}{k\mu} \right)^{\frac{1}{\gamma\sqrt{\log n}}} \xrightarrow{d} e^{\mathcal{N}}, \quad (1)$$

under the assumption $\mathbb{E}|X|^p < \infty$ for some $p > 2$.

The purpose of this paper is to show that the above result holds true under the assumption $\mathbb{E}X^2(\log^+ |X|)^{1/2} < \infty$. However, it is no longer true in general when only $\mathbb{E}X^2 < \infty$ is assumed. We will show that under this assumption, different normalisation is needed. Furthermore, we will set our discussion in a more general setting. It is straightforward that (1) is a simple corollary from

$$\frac{\sum_{k=1}^n f(S_{k,k}/k) - b_n}{a_n} \xrightarrow{d} \mathcal{N},$$

if one sets $f(x) = \log x$ and chooses the sequences a_n, b_n properly.

2. Main result

Theorem 1. *Suppose that $\mathbb{E}|X|^2 < \infty$ and denote $\mu = \mathbb{E}X$, $\sigma^2 = \text{Var}(X)$. Let f be a real function with bounded third derivative on some neighbourhood of μ . Then as $n \rightarrow \infty$*

$$\frac{\sum_{k=1}^n f(S_{k,k}/k) - b_n}{a_n} \xrightarrow{d} \sigma f'(\mu)\mathcal{N},$$

where

$$a_n = \sqrt{\log n}, \quad b_n = nf(\mu) + \frac{f''(\mu)}{2} \sum_{k=1}^n \frac{1}{k} \mathbb{E}|X - \mu|^2 1_{\{|X - \mu| \leq \sigma k\}}.$$

Remark 1. If we strengthen the assumption of the square integrability of random variable X to $\mathbb{E}X^2(\log^+ |X|)^{1/2} < \infty$, then we can take the sequence $\tilde{b}_n = nf(\mu) + \frac{f''(\mu)\sigma^2}{2} \log n$ instead of b_n . To see this we should show $\tilde{b}_n - b_n = o(\sqrt{\log n})$, which since $\log n - \sum_{k=1}^n \frac{1}{k} = O(1)$ is equivalent to

$$Q_n := \sum_{k=1}^n \frac{1}{k} (\sigma^2 - \mathbb{E}|X - \mu|^2 1_{\{|X - \mu| \leq \sigma k\}}) = o(\sqrt{\log n}).$$

Observe that Q_n is positive and

$$\begin{aligned} Q_n &= \sum_{k=1}^n \frac{1}{k} \mathbb{E}|X - \mu|^2 1_{\{|X - \mu| > \sigma k\}} = \mathbb{E}|X - \mu|^2 \sum_{\sigma k < |X - \mu|, k \leq n} \frac{1}{k} \\ &\sim \mathbb{E}|X - \mu|^2 \log^+ (n \wedge (|X - \mu|/\sigma)) =: \tilde{Q}_n. \end{aligned}$$

Therefore, if $\mathbb{E}X^2(\log^+ |X|)^{1/2} < \infty$, we can use the Dominated Convergence Theorem and infer that $\tilde{b}_n - b_n = o(\sqrt{\log n})$.

Remark 2. On the other hand, if for some $\varepsilon > 0$ we define a random variable X by setting $\mathbb{P}(X = \pm k_n) = C/(2k_n^2 n^2)$ and $\mathbb{P}(X = 0) = 1 - \sum_n \mathbb{P}(|X| = k_n)$, where $k_n = e^{n^{2+\varepsilon}}$ and $C = 6/\pi^2$. Then we simply have $\mu = 0$, $\sigma^2 = 1$ and $\mathbb{E}X^2(\log^+ |X|)^{1/2} = \infty$. Although, one can check that $\limsup_n Q_n/\sqrt{\log n} = \limsup_n \tilde{Q}_n/\sqrt{\log n} = \infty$, which means that we cannot use $\tilde{b}_n$ in general.

Now let us take any positive (i.e. $I \subset (0, \infty)$), nondegenerate random variable X with $\mathbb{E}X^2(\log^+ |X|)^{1/2} < \infty$ and $f(x) = \mu \log(x/\mu)$. Theorem 1 yields (1), that is the result from Rempala and Wesolowski (2005) however under weaker assumptions. Our argument shows that their result holds true even under the assumption of square integrability, although the normalizing sequences should be different. Namely, instead of the term $n^{\frac{\gamma^2}{2}}$ in (1) we should have $\exp(\frac{1}{2\mu^2} \sum_{k=1}^n \frac{1}{k} \mathbb{E}|X - \mu|^2 1_{\{|X - \mu| \leq \sigma k\}})$.

The proof of Theorem 1 relies on the Taylor's expansion of a function f in the neighbourhood of μ . Linear term in this expansion obeys a version of the classical Central Limit Theorem. Other terms are negligible mainly due to the Strong Law of Large Numbers (SLLN). This assertion will be made valid by a series of lemmas.

Lemma 2. *Under the assumptions of Theorem 1 with $\sigma > 0$*

$$\frac{1}{\sigma \sqrt{\log n}} \sum_{k=1}^n \left(\frac{S_{k,k}}{k} - \mu \right) \xrightarrow{d} \mathcal{N} \quad \text{as } n \rightarrow \infty.$$

PROOF. We may assume $\mathbb{E}X = 0$ and $\mathbb{E}X^2 = 1$ by a simple normalization argument. Since

$$\text{Var} \left(\sum_{k=1}^n \frac{S_{k,k}}{k} \right) = \sum_{k=1}^n \frac{\text{Var}(S_{k,k})}{k^2} = \sum_{k=1}^n \frac{1}{k} \sim \log n,$$

then to complete the proof it is sufficient to show the Lindeberg condition for the array $(\frac{S_{k,k}}{k\sqrt{\log n}})_{k \leq n}$, that is

$$\forall_{r>0} \quad \frac{1}{\log n} \sum_{k=1}^n \mathbb{E} \left(\frac{S_{k,k}}{k} \right)^2 1_{\{|S_{k,k}/k| > r\sqrt{\log n}\}} = o(1). \quad (2)$$

Since $\{(S_{k,k}/\sqrt{k})^2\}$ is uniformly integrable,

$$\sup_{k \in \mathbb{N}} \mathbb{E} \left(\frac{S_{k,k}}{\sqrt{k}} \right)^2 1_{\{|S_{k,k}/\sqrt{k}| > r\sqrt{\log n}\}} \rightarrow 0 \quad \text{as } n \rightarrow \infty,$$

proving (2). □

To establish the SLLN we will refer to Hsu-Robbins (1947) Law of Large Numbers (cf. Li et al. (1995) for partial bibliographies and brief discussions).

Lemma 3 (Hsu-Robbins LLN). *For a sequence (X_n) of iid rv's with $\mathbb{E}X_1 = 0$ and $\mathbb{E}X_1^2 < \infty$ the series*

$$\sum_{n=1}^{\infty} \mathbb{P}(|S_n/n| > t) \quad (3)$$

converges for every $t > 0$.

The condition (3) implies $S_n/n \rightarrow 0$ a.s. under the Borel-Cantelli lemma. Moreover if X_1 in Lemma 3 has the same distribution as X in Theorem 1, then $\mathbb{P}(|S_n/n| > t) = \mathbb{P}(|S_{n,n}/n| > t)$, i.e $S_{n,n}/n \rightarrow 0$ a.s. as well.

Before we proceed, we need some technical results derived from the elementary fact about the moments of sums of iid rv's (e.g. Hall and Heyde, 1980, p. 23).

Lemma 4 (Rosenthal's inequality). *If (X_n) is a sequence of iid rv's with the zero mean, then for any $p \geq 2$*

$$\mathbb{E}|S_n|^p \leq C_p \left(n\mathbb{E}|X_1|^p + n^{p/2}(\mathbb{E}X_1^2)^{p/2} \right),$$

where C_p is a constant depending only on p .

We will use this version of the Rosenthal's inequality to prove the following lemma, which on the other hand will simplify a number of steps in the next lemma. The later will play a crucial role in the proof of the main theorem.

Lemma 5. *Under the assumptions of Theorem 1, for every $p > 2$*

$$\mathbb{E} \sum_{k=1}^{\infty} \left(\frac{|\tilde{T}_k|}{k} \right)^p < \infty,$$

where $\tilde{T}_k = \sum_{i=1}^k (X_{i,k} 1_{\{|X_{i,k}| \leq k\}} - \mathbb{E} X_{i,k} 1_{\{|X_{i,k}| \leq k\}})$.

PROOF. Let $Y_k \stackrel{d}{=} X_{i,k} 1_{\{|X_{i,k}| \leq k\}} - \mathbb{E} X_{i,k} 1_{\{|X_{i,k}| \leq k\}}$, then by Lemma 4

$$\begin{aligned} \mathbb{E} \sum_{k=1}^{\infty} \left(\frac{|\tilde{T}_k|}{k} \right)^p &\leq C_p \sum_{k=1}^{\infty} \frac{1}{k^p} (k \mathbb{E} |Y_k|^p + k^{p/2} (\mathbb{E} Y_k^2)^{p/2}) \\ &\leq C_p 2^p \left(\sum_{k=1}^{\infty} \frac{1}{k^{p-1}} \mathbb{E} |X|^p 1_{\{|X| \leq k\}} + \sum_{k=1}^{\infty} \frac{1}{k^{p/2}} (\mathbb{E} X^2)^{p/2} \right) \\ &= C_p 2^p \left(\mathbb{E} |X|^p \sum_{k \geq |X|}^{\infty} k^{1-p} + (\mathbb{E} X^2)^{p/2} \sum_{k=1}^{\infty} \frac{1}{k^{p/2}} \right) \\ &\leq C_p 2^p \left(C \mathbb{E} |X|^2 + (\mathbb{E} X^2)^{p/2} \sum_{k=1}^{\infty} \frac{1}{k^{p/2}} \right) < \infty, \end{aligned}$$

for some positive constant C . □

Lemma 6. *Under the assumptions of Theorem 1 we have*

$$\sum_{k=1}^n \left[\left(\frac{S_{k,k} - k\mu}{k} \right)^2 - \frac{\mathbb{E} |X - \mu|^2 1_{\{|X - \mu| \leq \sigma k\}}}{k} \right] = O_{\mathbb{P}}(1), \quad (4)$$

$$\sum_{k=1}^n \left| \frac{S_{k,k} - k\mu}{k} \right|^3 = O_{\mathbb{P}}(1). \quad (5)$$

PROOF. To simplify the notation we will write S_k for $S_{k,k}$. First note that to show (4) it is sufficient to prove that

$$\sum_{k=1}^n \left(\frac{S_k^2}{k^2} - \frac{\mathbb{E} |X|^2 1_{\{|X| \leq k\}}}{k} \right) = O_{\mathbb{P}}(1),$$

for a normalized random variable X . Take any $\varepsilon > 0$ and let $T_k := \sum_{i=1}^k X_{i,k} 1_{\{|X_{i,k}| \leq k\}}$. Then $\sum_{k=1}^{\infty} \mathbb{P}(S_k \neq T_k) \leq \sum_{k=1}^{\infty} k \mathbb{P}(|X| > k) < \infty$, because $\mathbb{E}X^2 < \infty$. Hence we can take R big enough that $\sum_{k=R}^{\infty} \mathbb{P}(S_k \neq T_k) < \varepsilon/3$ and M big enough that

$$\mathbb{P}\left(\left|\sum_{k=1}^{R-1} \left(\frac{S_k^2}{k^2} - \frac{\mathbb{E}|X|^2 1_{\{|X| \leq k\}}}{k}\right)\right| > M/2\right) < \varepsilon/3.$$

So all we need to show is

$$\mathbb{P}\left(\left|\sum_{k=R}^n \left(\frac{T_k^2}{k^2} - \frac{\mathbb{E}|X|^2 1_{\{|X| \leq k\}}}{k}\right)\right| > M/2\right) < \varepsilon/3,$$

which is implied by

$$\sum_{k=1}^n \frac{T_k^2 - b_k}{k^2} = O_{\mathbb{P}}(1) \quad (6)$$

with $b_k := k \mathbb{E}|X|^2 1_{\{|X| \leq k\}}$. Observe that

$$T_k = \sum_{i=1}^k (X_{i,k} 1_{\{|X_{i,k}| \leq k\}} - \mathbb{E}X_{i,k} 1_{\{|X_{i,k}| \leq k\}}) + c_k := \tilde{T}_k + c_k,$$

where $c_k = k \mathbb{E}X 1_{\{|X| \leq k\}}$ and

$$\sum_{k=1}^n \frac{T_k^2 - b_k}{k^2} = \sum_{k=1}^n \frac{\tilde{T}_k^2 - b_k}{k^2} + \sum_{k=1}^n \frac{c_k^2}{k^2} + 2 \sum_{k=1}^n \frac{c_k \tilde{T}_k}{k^2} := I_1 + I_2 + I_3.$$

Recall that $\mathbb{E}X = 0$ so that

$$|c_k| = |k \mathbb{E}X 1_{\{|X| \leq k\}}| = |k \mathbb{E}X 1_{\{|X| > k\}}| \leq \mathbb{E}|X|^2 = 1,$$

thus $I_2 = O(1)$. We also have $I_3 = O_{\mathbb{P}}(1)$ because I_3 is bounded in L^2

$$\begin{aligned} \mathbb{E}\left(\sum_{k=1}^n \frac{c_k \tilde{T}_k}{k^2}\right)^2 &= \sum_{k=1}^n \mathbb{E}\left(\frac{c_k \tilde{T}_k}{k^2}\right)^2 \leq \sum_{k=1}^n \frac{1}{k^4} k \operatorname{Var}(X 1_{\{|X| \leq k\}}) \\ &\leq \mathbb{E}X^2 \sum_{k=1}^n \frac{1}{k^3} = O(1). \end{aligned}$$

I_1 can be rewritten as

$$I_1 = \sum_{k=1}^n \frac{\tilde{T}_k^2 - \mathbb{E}\tilde{T}_k^2}{k^2} - \sum_{k=1}^n \frac{c_k^2}{k^3} := I_{11} - I_{12}.$$

But $I_{12} = O(1)$ since $|c_k| \leq 1$. So in order to show (6) it is enough to show that

$$K_n := \sum_{k=1}^n \frac{\tilde{T}_k^2 - \mathbb{E}\tilde{T}_k^2}{k^2} = O_{\mathbb{P}}(1),$$

where $\tilde{T}_k$ is a sum of independent, mean zero rv's with the same distribution as $X1_{\{|X| \leq k\}} - \mathbb{E}X1_{\{|X| \leq k\}}$. This however follows from Lemma 5 with $p = 4$. Indeed

$$\mathbb{E}K_n^2 = \text{Var}(K_n) = \sum_{k=1}^n \frac{1}{k^4} \text{Var}(\tilde{T}_k^2) \leq \mathbb{E} \sum_{k=1}^{\infty} \left(\frac{\tilde{T}_k}{k} \right)^4 < \infty,$$

so the proof of (4) is complete.

To prove (5) it suffices to show that $\sum_{k=1}^n \left| \frac{S_k}{k} \right|^3 = O_{\mathbb{P}}(1)$ for normalized X , which by the same arguments as above is implied by

$$\sum_{k=1}^n \left| \frac{T_k}{k} \right|^3 = O_{\mathbb{P}}(1).$$

Using the same notation we have $|T_k|^3 = |\tilde{T}_k + c_k|^3 \leq 4|\tilde{T}_k|^3 + 4$. Thus

$$\sum_{k=1}^n \left| \frac{T_k}{k} \right|^3 \leq 4 \left(\sum_{k=1}^n \left| \frac{\tilde{T}_k}{k} \right|^3 + \sum_{k=1}^n \frac{1}{k^3} \right) := I_4 + I_5.$$

We obviously have $I_5 = O(1)$ and by Lemma 5 with $p = 3$ we get boundedness of I_4 in L^1 , which completes the proof. $\square$

Now we are in the position to prove the main theorem.

PROOF OF THEOREM 1. Take a_n and b_n as in the claim and denote $c_k = \mathbb{E}|X - \mu|^2 1_{\{|X - \mu| \leq \sigma k\}}$ so $b_n = \sum_{k=1}^n (f(\mu) + f''(\mu) \frac{c_k}{2k})$. By Taylor's expansion,

$$f\left(\frac{S_{k,k}}{k}\right) = f(\mu) + f'(\mu) \left(\frac{S_{k,k}}{k} - \mu\right) + \frac{f''(\mu)}{2} \left(\frac{S_{k,k}}{k} - \mu\right)^2 + O\left(\left|\frac{S_{k,k}}{k} - \mu\right|^3\right) \text{ a.s.,}$$

as a consequence of the SLLN and the assumption of boundedness of $f^{(3)}$ around μ . Using Lemma 6 we have

$$\begin{aligned} \frac{\sum_{k=1}^n f(S_{k,k}/k) - b_n}{a_n} &= \frac{f'(\mu)}{a_n} \sum_{k=1}^n \left(\frac{S_{k,k}}{k} - \mu \right) + \frac{f''(\mu)}{2a_n} \sum_{k=1}^n \left[\left(\frac{S_{k,k}}{k} - \mu \right)^2 - \frac{c_k}{k} \right] \\ &\quad + O \left(\frac{1}{a_n} \sum_{k=1}^n \left| \frac{S_{k,k}}{k} - \mu \right|^3 \right) \text{ a.s.} \\ &= \frac{f'(\mu)}{a_n} \sum_{k=1}^n \left(\frac{S_{k,k}}{k} - \mu \right) + o_{\mathbb{P}}(1). \end{aligned}$$

By Lemma 2

$$\frac{f'(\mu)}{a_n} \sum_{k=1}^n \left(\frac{S_{k,k}}{k} - \mu \right) \xrightarrow{d} \sigma f'(\mu) \mathcal{N},$$

completing the proof. $\square$

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