

# Almost sure functional limit theorem for the product of partial sums

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ABSTRACT. We prove an almost sure functional limit theorem for the product of partial sums of i.i.d. positive random variables with finite second moment.

## 1. Introduction and Main Result

Limiting distributions of the product of partial sums of positive random variables have been widely studied in recent years. Arnold and Villaseñor [1] proved the limit theorem for the partial sum of a sequence of exponential random variables. Rempała and Wesolowski [4] proved it for any independent and identically distributed (i.i.d.) random variables with finite variance. Later, Qi [5] considered a sequence of random variables with  $\alpha$ -stable distribution and established the limit distribution of the product of the partial sums when  $1 < \alpha \leq 2$ .

Recently, Zhang and Huang [6] proved the following invariance principle of the product of partial sums of i.i.d. positive random variables with mean  $\mu > 0$  and variance  $\sigma^2$ :

$$\left( \prod_{k=1}^{\lfloor nt \rfloor} \frac{S_k}{\mu k} \right)^{\frac{\mu}{\sigma\sqrt{n}}} \xrightarrow{\mathcal{D}} \exp \left( \int_0^t \frac{W(s)}{s} ds \right) \text{ as } n \rightarrow \infty. \quad (1.1)$$

The goal of this paper is to obtain an almost sure version of the above invariance principle which can also be a functional version of the almost sure limit theorem obtained by Gonchigdanzan and Rempała [3]. Here is the result:

**Theorem 1.** *Let  $(X_k)_{k \geq 1}$  be a sequence of i.i.d. positive random variables with mean  $\mu > 0$  and variance  $\sigma^2$  and let  $S_n = X_1 + \cdots + X_n$ . Define a process  $\{\xi_n(t) : 0 \leq t \leq 1\}$  by*

$$\xi_n(t) := \left( \prod_{k=1}^{\lfloor nt \rfloor} \frac{S_k}{\mu k} \right)^{\frac{\mu}{\sigma\sqrt{n}}}.$$

*Let  $F_t$  be the distribution function of the random variable on the right-hand side of (1.1). Then*

$$\frac{1}{\log n} \sum_{k=1}^n \frac{1}{k} \mathbf{I}(\xi_k(t) \leq x) \xrightarrow{a.s.} F_t(x) \text{ as } n \rightarrow \infty \quad (1.2)$$

*if and only if*

$$\frac{1}{\log n} \sum_{k=1}^n \frac{1}{k} \mathbf{P}(\xi_k(t) \leq x) \longrightarrow F_t(x) \text{ as } n \rightarrow \infty. \quad (1.3)$$

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**Corollary.** Let  $(X_k)_{k \geq 1}$  be a sequence of i.i.d. positive random variables with mean  $\mu > 0$  and variance  $\sigma^2$ . Then we have

$$\frac{1}{\log n} \sum_{k=1}^n \frac{1}{k} \mathbf{I} \left( \left( \prod_{j=1}^{\lfloor kt \rfloor} \frac{S_j}{\mu j} \right)^{\frac{\mu}{\sigma \sqrt{k}}} \leq x \right) \xrightarrow{a.s.} F_t(x) \text{ as } n \rightarrow \infty.$$

## 2. Auxiliary Results and Proofs

Throughout the paper  $\log x$  and  $\log \log x$  stand for  $\ln(e \vee x)$  and  $\ln \ln(n \vee e^e)$  respectively, and " $\ll$ " is meant for the big "O" notation.

### 2.1 Auxiliary Results

**Lemma 1.** Let  $(Y_n)_{n \geq 1}$  be a sequence of random variables. Set  $S_n = Y_1 + \dots + Y_n$ . Then we have

$$E \left( \max_{1 \leq k \leq n} \left| \sum_{j=1}^k \log \left( \frac{n+1}{j} \right) Y_j \right| \right) \leq 3 \log(n+1) E \left( \max_{1 \leq k \leq n} |S_k| \right).$$

**Proof.** Observe that

$$\left| \sum_{j=1}^k \log \left( \frac{n+1}{j} \right) Y_j \right| \leq \left| \sum_{j=1}^k \log(n+1) Y_j \right| + \left| \sum_{j=2}^k \log j Y_j \right| = T_1 + T_2$$

Obviously,  $T_1 \leq \log(n+1) |S_k|$  which implies

$$\max_{1 \leq k \leq n} T_1 \leq \log(n+1) \max_{1 \leq k \leq n} |S_k|.$$

For the second term  $T_2$  we have

$$\begin{aligned} \max_{2 \leq k \leq n} T_2 &= \max_{2 \leq k \leq n} \left| \sum_{j=2}^k \log j Y_j \right| = \max_{2 \leq k \leq n} \left| \sum_{j=2}^k (Y_j + Y_{j+1} + \dots + Y_k) (\log j - \log(j-1)) \right| \\ &\leq \max_{2 \leq k \leq n} \sum_{j=2}^k |Y_j + Y_{j+1} + \dots + Y_k| (\log j - \log(j-1)) \\ &\leq 2 \max_{2 \leq k \leq n} |Y_1 + Y_2 + \dots + Y_k| \sum_{j=2}^n (\log j - \log(j-1)) \leq 2 \log(n+1) \max_{1 \leq k \leq n} |S_k|. \quad \square \end{aligned}$$

**Lemma 2.** Let  $(X_k)_{k \geq 1}$  be a sequence of i.i.d. positive random variables with mean  $\mu$  and variance  $\sigma^2$ . Then setting  $S_n = X_1 + \dots + X_n$  we have

$$\max_{0 \leq t \leq 1} \left| \frac{\mu}{\sigma \sqrt{n}} \sum_{k=1}^{\lfloor nt \rfloor} \log \frac{S_k}{\mu k} - \frac{\mu}{\sigma \sqrt{n}} \sum_{k=1}^{\lfloor nt \rfloor} \left( \frac{S_k}{\mu k} - 1 \right) \right| \xrightarrow{a.s.} 0 \text{ as } n \rightarrow \infty.$$

**Proof.** Note that  $\log(x+1) = x - r(x)$  where  $r(x)/x^2 \rightarrow \frac{1}{2}$  as  $x \rightarrow 0$ . By the strong law of large numbers we have  $S_k/k - \mu \xrightarrow{a.s.} 0$  as  $k \rightarrow \infty$ .

Thus by the law of iterated logarithm we get

$$\begin{aligned} & \left| \frac{\mu}{\sigma\sqrt{n}} \sum_{k=1}^{[nt]} \log \frac{S_k}{\mu k} - \frac{\mu}{\sigma\sqrt{n}} \sum_{k=1}^{[nt]} \left( \frac{S_k}{\mu k} - 1 \right) \right| \stackrel{a.s.}{\ll} \frac{1}{\sigma\sqrt{n}} \sum_{k=1}^{[nt]} \left( \frac{S_k}{k} - \mu \right)^2 \\ & \stackrel{a.s.}{\ll} \max_{0 \leq t \leq 1} \frac{1}{\sigma\sqrt{n}} \sum_{k=1}^{[nt]} \frac{1}{k} \log \log k \ll \frac{1}{\sigma\sqrt{n}} \log n \log \log n \rightarrow 0. \quad \square \end{aligned}$$

## 2.2 Proof of Theorem 1.

We use Berkes and Dehling's [2] technique to prove our theorem. Observe that

$$\sum_{k=1}^n \left( \frac{S_k}{k} - \mu \right) = \sum_{k=1}^n b_{k,n}(X_k - \mu)$$

where  $b_{k,n} = \sum_{j=k}^n 1/j$ . Hence by Lemma 2 it suffices to show that for any Borel-subset  $A$  of  $D[0, 1]$

$$\frac{1}{\log n} \sum_{k=2}^n \frac{1}{k} \mathbf{I} \left( \left( \frac{\hat{s}_k}{\sigma\sqrt{k}} \in A \right) - \mathbf{P} \left( \frac{\hat{s}_k}{\sigma\sqrt{k}} \in A \right) \right) \xrightarrow{a.s.} 0 \text{ as } n \rightarrow \infty. \quad (2.1)$$

where  $\hat{s}_n = \sum_{i=1}^{[nt]} b_{i,[nt]}(X_i - \mu)$ . From Berkes and Dehling [2] (p.1647), to prove (2.1) it suffices to show that for any bounded Lipschitz function  $f$  on  $D[0, 1]$  we have

$$\frac{1}{\log n} \sum_{k=1}^n \frac{1}{k} \zeta_k \xrightarrow{a.s.} 0 \text{ as } n \rightarrow \infty \quad (2.2)$$

where  $\zeta_k = f(\hat{s}_k/\sigma\sqrt{k}) - \mathbf{E}f(\hat{s}_k/\sigma\sqrt{k})$ . In fact, the following implies (2.2) (see p.1648, Berkes and Dehling (1993) for the proof):

$$\mathbf{E} \left( \sum_{k=1}^n \frac{1}{k} \zeta_k \right)^2 \ll \log^2 n (\log \log n)^{-\varepsilon} \text{ for some } \varepsilon > 0. \quad (2.3)$$

Therefore, showing (2.3) would be sufficient for the proof of Theorem 1. Observing

$$\hat{s}_l - \hat{s}_k = b_{[kt]+1, [lt]}(S_{[kt]} - [kt]\mu) + (b_{[kt]+1, [lt]}(X_{[kt]+1} - \mu) + \cdots + b_{[lt], [lt]}(X_{[lt]} - \mu))$$

for  $l \geq k$  we find that  $\hat{s}_l - \hat{s}_k - b_{[kt]+1, [lt]}(S_{[kt]} - \mu[kt])$  is independent of  $\hat{s}_{[kt]}$  which implies that

$$\text{Cov} \left( f \left( \frac{\hat{s}_k}{\sigma\sqrt{k}} \right), f \left( \frac{\hat{s}_l - \hat{s}_k - b_{[kt]+1, [lt]}(S_{[kt]} - \mu[kt])}{\sigma\sqrt{l}} \right) \right) = 0 \text{ for } l \geq k.$$

By the Lipschitz property of  $f$

$$\begin{aligned} |\mathbf{E}(\zeta_k \zeta_l)| & \ll \left| \text{Cov} \left( f \left( \frac{\hat{s}_k}{\sigma\sqrt{k}} \right), f \left( \frac{\hat{s}_l}{\sigma\sqrt{l}} \right) - f \left( \frac{|\hat{s}_l - \hat{s}_k - b_{[kt]+1, [lt]}(S_{[kt]} - \mu[kt])|}{\sigma\sqrt{l}} \right) \right) \right| \\ & \ll \mathbf{E} \left( \max_{0 \leq t \leq 1} \frac{|\hat{s}_k + b_{[kt]+1, [lt]}(S_{[kt]} - \mu[kt])|}{\sigma\sqrt{l}} \right) \\ & \ll \mathbf{E} \left( \max_{0 \leq t \leq 1} \frac{|\hat{s}_k|}{\sigma\sqrt{l}} \right) + \mathbf{E} \left( \max_{0 \leq t \leq 1} \frac{|b_{[kt]+1, [lt]}(S_{[kt]} - \mu[kt])|}{\sigma\sqrt{l}} \right) \\ & \ll \left( \frac{k}{l} \right)^{1/2} \mathbf{E} \left( \max_{0 \leq t \leq 1} \frac{|\hat{s}_k|}{\sigma\sqrt{k}} \right) + \left( \frac{k}{l} \right)^{1/2} \mathbf{E} \left( \max_{0 \leq t \leq 1} b_{[kt]+1, [lt]} \frac{|S_{[kt]} - \mu[kt]|}{\sigma\sqrt{k}} \right). \end{aligned}$$

Since  $\max_{0 \leq t \leq 1} b_{[kt]+1, [lt]} = \log(l/k) \ll (l/k)^\gamma$  (we choose  $0 < \gamma < 1/2$ )

$$\begin{aligned} |\mathbf{E}(\zeta_k \zeta_l)| &\ll \left(\frac{k}{l}\right)^{1/2} \mathbf{E}\left(\max_{0 \leq t \leq 1} \frac{1}{\sigma\sqrt{k}} \left| \sum_{i=1}^{[kt]} b_{i,k}(X_i - \mu) \right| \right) + \left(\frac{k}{l}\right)^{1/2-\gamma} \mathbf{E}\left(\max_{0 \leq t \leq 1} \frac{|S_{[kt]} - \mu[kt]|}{\sigma\sqrt{k}}\right) \\ &= \left(\frac{k}{l}\right)^{1/2} \mathbf{E}\left(\max_{0 \leq j \leq k} \frac{1}{\sigma\sqrt{k}} \left| \sum_{i=1}^j b_{i,k}(X_i - \mu) \right| \right) + \left(\frac{k}{l}\right)^{1/2-\gamma} \mathbf{E}\left(\max_{0 \leq j \leq k} \frac{|S_j - \mu j|}{\sigma\sqrt{k}}\right) \\ &= M_1 + M_2. \end{aligned}$$

Now applying Lemma 1 to  $M_1$  we obtain

$$\begin{aligned} |\mathbf{E}(\zeta_k \zeta_l)| &\ll \left(\frac{k}{l}\right)^{1/2} \log k \mathbf{E}\left(\max_{1 \leq j \leq k} \frac{1}{\sigma\sqrt{k}} \left| \sum_{i=1}^j (X_i - \mu) \right| \right) + \left(\frac{k}{l}\right)^{1/2-\gamma} \mathbf{E}\left(\max_{1 \leq j \leq k} \frac{|S_j - \mu j|}{\sigma\sqrt{k}}\right) \\ &\ll |\mathbf{E}(\zeta_k \zeta_l)| \ll \log k \left(\frac{k}{l}\right)^{1/2-\gamma} \mathbf{E}\left(\max_{1 \leq j \leq k} \frac{|S_j - \mu j|}{\sigma\sqrt{k}}\right) \ll \log k \left(\frac{k}{l}\right)^{\gamma'} \end{aligned}$$

where  $0 < \gamma' < 1/2 - \gamma$ .

On the other hand  $\mathbf{E}(\zeta_k \zeta_l) \ll 1$  because  $\zeta_k$  is bounded. Hence we have the following estimate for  $\mathbf{E}(\zeta_k \zeta_l)$ :

$$\mathbf{E}(\zeta_k \zeta_l) \ll \begin{cases} 1, & \text{if } l/k \leq \exp((\log n)^{1-\varepsilon}) \\ (k/l)^{\gamma'} \log k, & \text{if } l/k \geq \exp((\log n)^{1-\varepsilon}) \end{cases}$$

where  $\varepsilon$  is any positive number. Hence,

$$\sum_{\substack{1 \leq k \leq l \leq n \\ l/k \leq \exp((\log n)^{1-\varepsilon})}} \frac{\mathbf{E}(\zeta_k \zeta_l)}{kl} \leq \sum_{1 \leq k \leq n} \frac{1}{k} \sum_{k \leq l \leq k \exp((\log n)^{1-\varepsilon})} \frac{1}{l} \ll \sum_{k=1}^n \frac{1}{k} \log^{1-\varepsilon} n \ll \log^{2-\varepsilon} n \quad (2.4)$$

and

$$\sum_{\substack{1 \leq k \leq l \leq n \\ l/k \geq \exp((\log n)^{1-\varepsilon})}} \frac{\mathbf{E}(\zeta_k \zeta_l)}{kl} \leq e^{-\gamma'(\log n)^{1-\varepsilon}} \log n \sum_{1 \leq k \leq l \leq n} \frac{1}{kl} \ll e^{-\gamma'(\log n)^{1-\varepsilon}} \log^3 n \ll \log^{2-\varepsilon} n. \quad (2.5)$$

Thus (2.4) and (2.5) immediately imply (2.3) which completes the proof of Theorem 1.  $\square$

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