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## Rate of Convergence in the Central Limit Theorem for iid Pareto Variables

Claas Becker

*Hochschule RheinMain, 65022 Wiesbaden, Germany, [claas.becker@hs-rm.de](mailto:claas.becker@hs-rm.de)*

Manuel Bohnet

*Hochschule RheinMain, 65022 Wiesbaden, Germany, [manuel\\_bohnet@web.de](mailto:manuel_bohnet@web.de)*

Sarah Kummert

*Hochschule RheinMain, 65022 Wiesbaden, Germany, [sarah.kummert@gmx.de](mailto:sarah.kummert@gmx.de)*

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## RATE OF CONVERGENCE IN THE CENTRAL LIMIT THEOREM FOR IID PARETO VARIABLES

CLAAS BECKER\*, MANUEL BOHNET, AND SARAH KUMMERT

**ABSTRACT.** We estimate the rate of convergence in the central limit theorem for a sequence of iid Pareto variables  $X_k$  with shape parameter  $r$ . If  $r \leq 4$ ,  $E(|X_1|^3) = \infty$  and the Berry-Esseen theorem cannot be applied. In these cases the rate of convergence is very slow and can be expressed as a function of  $r$ .

### 1. Introduction

Let  $(X_k)_{k \in \mathbb{N}}$  be a sequence of iid random variables with mean  $\mu$  and variance  $\sigma^2$ . By the central limit theorem, the standardized random sums

$$Z_n = \frac{1}{\sqrt{n\sigma^2}} \sum_{k=1}^n (X_k - \mu)$$

converge weakly to the standard normal distribution. The rate of convergence can be measured in terms of

$$\|F_n - \Phi\|_\infty = \sup_{x \in \mathbb{R}} \{|F_n(x) - \Phi(x)|\}$$

where  $F_n$  is the cumulative distribution function of  $Z_n$  and  $\Phi$  is the cumulative distribution function of the standardized normal distribution. It is well known that without further assumptions on the distribution of the  $X_k$ , the rate of convergence cannot be faster than  $1/\sqrt{n}$ , see [3], p.448.

If  $E(|X_1|^3) < \infty$ , then, by the Berry-Esseen theorem,

$$\|F_n - \Phi\|_\infty \leq \frac{CE(|X_1|^3)}{\sigma^3 \sqrt{n}}.$$

In recent years, significant progress has been made to obtain sharp estimates for the constant  $C$ , see [2], [4], [5] and the references therein.

In this paper, we focus on iid random variables with a Pareto distribution. Their probability density function is given by

$$f(x) = \begin{cases} (r-1) x_{\min}^{r-1} x^{-r}, & x \geq x_{\min} > 0 \\ 0, & \text{otherwise} \end{cases} \quad (1.1)$$

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\* Corresponding author.

with shape parameter  $r > 1$  and location parameter  $x_{\min}$ . We are interested in tail behaviour and therefore, for simplicity, put  $x_{\min} = 1$ .

If  $r \leq 4$ ,  $E(|X_1|^3) = \infty$  and the Berry-Esseen theorem cannot be applied. For the same reason, an Edgeworth expansion does not work. Instead, we rely on the technique of truncated moments.

## 2. The Truncation Method

For each  $X_k$ , we take any  $\hat{\tau}_k, \tau_k > 0$  and define the truncated random variables

$$\bar{X}_k = X_k \cdot 1_{]-\hat{\tau}_k, \tau_k[} \quad \text{and} \quad X'_k = X_k - \bar{X}_k .$$

The following theorem is due to Feller [1].

**Theorem 2.1.** *Let  $(X_k)_{k \in \mathbb{N}}$  be a sequence of mutually independent random variables. We assume  $E(X_k) = 0$  and  $E(X_k^2) = \sigma_k^2 < \infty$  for all  $k \in \mathbb{N}$ . Put*

$$\begin{aligned} \beta'_k &= E(X_k'^2) & b' &= \beta'_1 + \dots + \beta'_n \\ \gamma_k &= E(|\bar{X}_k|^3) & c &= \gamma_1 + \dots + \gamma_n \\ s^2 &= \sigma_1^2 + \dots + \sigma_n^2 . \end{aligned}$$

Then

$$\|F_n - \Phi\|_\infty \leq 6 \left( \frac{c}{s^3} + \frac{b'}{s^2} \right) \quad (2.1)$$

where  $F_n$  is the cumulative distribution function of the standardized sum

$$\frac{1}{\sqrt{s^2}} \sum_{k=1}^n X_k .$$

Note that the dependency on  $n$  is hidden in the constants  $b'$ ,  $c$  and  $s^2$ .

## 3. Rate of Convergence

**Theorem 3.1.** *Assume that  $(Y_k)_{k \in \mathbb{N}}$  are iid Pareto with shape parameter  $r > 3$  and location parameter  $x_{\min} = 1$ . We put  $\mu = E(Y_n)$ ,  $X_n = Y_n - \mu$  and  $\sigma^2 = E(X_n^2)$ . Let  $F_n$  denote the cumulative distribution function of*

$$Z_n = \frac{1}{\sqrt{n\sigma^2}} \sum_{k=1}^n X_k .$$

Then there exist constants  $c_1, c_2, c_3 > 0$  such that the asymptotic behaviour<sup>1</sup> of the bound in inequality (2.1) is given by

$$\|F_n - \Phi\|_\infty \leq 6 \left( \frac{c}{s^3} + \frac{b'}{s^2} \right) \sim \begin{cases} c_1 n^{\frac{1}{2}(3-r)}, & 3 < r < 4 \\ c_2 n^{-\frac{1}{2}} \ln n, & r = 4 \\ c_3 n^{-\frac{1}{2}}, & r > 4. \end{cases}$$

<sup>1</sup> $a_n \sim b_n$  iff  $\lim_{n \rightarrow \infty} \frac{a_n}{b_n} = 1$

*Proof.* We have

$$\mu = E(Y_k) = \frac{r-1}{r-2} \quad \text{and} \quad E(Y_k^2) = \frac{r-1}{r-3}.$$

The probability density function of  $X_k$  is

$$f(x) = \begin{cases} (r-1)(x+\mu)^{-r}, & x \geq 1-\mu \\ 0, & \text{otherwise.} \end{cases}$$

Since the shifted Pareto distributions have support in the interval  $[1-\mu, \infty[$ , we can use symmetric truncation  $\bar{X}_k = X_k \cdot 1_{] -\tau_k, \tau_k[}$ . We try the ansatz

$$\tau_k = k^\alpha \quad \text{with} \quad \alpha > 0.$$

The main challenge is to choose  $\alpha$  in such a way that both terms  $\frac{c}{s^3}$  and  $\frac{b'}{s^2}$  in inequality (2.1) decay sufficiently fast.

(1) *The case*  $3 < r < 4$ :

We compute  $\beta'_k$  and  $\gamma_k$ . A simple computation shows that

$$\begin{aligned} \beta'_k = E(X_k'^2) &= (r-1) \int_{\tau_k}^{\infty} x^2 (x+\mu)^{-r} dx \\ &= (r-1) \int_{\tau_k+\mu}^{\infty} (x-\mu)^2 x^{-r} dx \\ &= -(r-1) \left( \frac{\tilde{\tau}_k^{3-r}}{3-r} - 2\mu \frac{\tilde{\tau}_k^{2-r}}{2-r} + \mu^2 \frac{\tilde{\tau}_k^{1-r}}{1-r} \right) \end{aligned} \tag{3.1}$$

with the abbreviation  $\tilde{\tau}_k = \tau_k + \mu$ . Therefore

$$b' = \sum_{k=1}^n \beta'_k \leq -(r-1) \sum_{k=1}^n \left( \frac{\tilde{\tau}_k^{3-r}}{3-r} + \mu^2 \frac{\tilde{\tau}_k^{1-r}}{1-r} \right).$$

$$\begin{aligned} \gamma_k &= E(|\bar{X}_k|^3) \\ &= -2(r-1) \left( \frac{\mu^{4-r}}{4-r} - 3\mu \frac{\mu^{3-r}}{3-r} + 3\mu^2 \frac{\mu^{2-r}}{2-r} - \mu^3 \frac{\mu^{1-r}}{1-r} \right) \\ &\quad + (r-1) \left( \frac{\tilde{\tau}_k^{4-r} + 1}{4-r} - 3\mu \frac{\tilde{\tau}_k^{3-r} + 1}{3-r} + 3\mu^2 \frac{\tilde{\tau}_k^{2-r} + 1}{2-r} - \mu^3 \frac{\tilde{\tau}_k^{1-r} + 1}{1-r} \right) \end{aligned} \tag{3.2}$$

Therefore the following inequality holds

$$c = \sum_{k=1}^n \gamma_k \leq (r-1) \sum_{k=1}^n \left( \frac{\tilde{\tau}_k^{4-r}}{4-r} - 3\mu \frac{\tilde{\tau}_k^{3-r}}{3-r} - \mu^3 \frac{\tilde{\tau}_k^{1-r}}{1-r} \right) + n(r-1)(\theta_1 - 2\theta_2)$$

with constants

$$\theta_1 = \frac{1}{4-r} - 3\mu \frac{1}{3-r} - \mu^3 \frac{1}{1-r} \quad \text{and} \quad \theta_2 = 3\mu^2 \frac{\mu^{2-r}}{2-r}.$$

Using  $\tilde{\tau}_k = \tau_k + \mu = k^\alpha + \mu$ , we arrive at the inequalities

$$\tilde{\tau}_k^{4-r} \leq 3k^{\alpha(4-r)} \quad \tilde{\tau}_k^{3-r} \leq k^{\alpha(3-r)} \quad \tilde{\tau}_k^{1-r} \leq k^{\alpha(1-r)} .$$

Therefore

$$\begin{aligned} c + sb' &\leq (r-1) \left( \frac{3}{4-r} \sum_{k=1}^n k^{\alpha(4-r)} - \frac{3\mu}{3-r} \sum_{k=1}^n k^{\alpha(3-r)} - \frac{\mu^3}{1-r} \sum_{k=1}^n k^{\alpha(1-r)} \right) \\ &\quad + n(r-1)(\theta_1 - 2\theta_2) \\ &\quad - s(r-1) \left( \frac{1}{3-r} \sum_{k=1}^n k^{\alpha(3-r)} + \frac{\mu^2}{1-r} \sum_{k=1}^n k^{\alpha(1-r)} \right) . \end{aligned}$$

Note that  $s$  is not a constant since  $s^2 = \sum_{k=1}^n \sigma^2 = n\sigma^2$ . Approximating integrals by Riemann sums, we have

$$\sum_{k=1}^{n-1} k^{\alpha(4-r)} \leq \int_1^n x^{\alpha(4-r)} dx = \frac{n^{\alpha(4-r)+1} - 1}{\alpha(4-r) + 1} .$$

If  $\alpha \neq -\frac{1}{3-r}$ , we have

$$\sum_{k=2}^n k^{\alpha(3-r)} \leq \int_1^n x^{\alpha(3-r)} dx = \frac{n^{\alpha(3-r)+1} - 1}{\alpha(3-r) + 1} .$$

Analogously, if  $\alpha \neq -\frac{1}{1-r}$ , we obtain

$$\sum_{k=2}^n k^{\alpha(1-r)} \leq \int_1^n x^{\alpha(1-r)} dx = \frac{n^{\alpha(1-r)+1} - 1}{\alpha(1-r) + 1} .$$

Combining these inequalities, we arrive at

$$\begin{aligned} 6 \frac{c + sb'}{s^3} &\leq \frac{6\beta_1}{\sigma^3} \left( 3 \frac{n^{\alpha(4-r)-\frac{1}{2}} - n^{-\frac{3}{2}}}{\beta_2} + 3 \frac{n^{\alpha(4-r)-\frac{3}{2}}}{4-r} - 3\mu \frac{n^{\alpha(3-r)-\frac{1}{2}} - n^{-\frac{3}{2}}}{\beta_3} \right. \\ &\quad \left. - n^{-\frac{3}{2}} \frac{3\mu}{3-r} - \mu^3 \frac{n^{\alpha(1-r)-\frac{1}{2}} - n^{-\frac{3}{2}}}{\beta_4} - n^{-\frac{3}{2}} \frac{\mu^3}{1-r} + n^{-\frac{1}{2}}(\theta_1 - 2\theta_2) \right. \\ &\quad \left. - \sigma \frac{n^{\alpha(3-r)} - n^{-1}}{\beta_3} - \sigma \frac{n^{-1}}{3-r} - \sigma \mu^2 \frac{n^{\alpha(1-r)} - n^{-1}}{\beta_4} - \sigma \frac{n^{-1}\mu^2}{1-r} \right) \end{aligned}$$

with constants

$$\begin{aligned} \beta_1 &= r-1 & \beta_2 &= \alpha(4-r)^2 + 4-r \\ \beta_3 &= \alpha(3-r)^2 + 3-r & \beta_4 &= \alpha(1-r)^2 + 1-r . \end{aligned}$$

It turns out that, as  $n \rightarrow \infty$ , only the two terms containing  $n^{\alpha(4-r)-\frac{1}{2}}$  and  $n^{\alpha(3-r)}$  ultimately determine the rate of convergence. For fast convergence of the first term, it would be best to put  $\alpha = 0$ . But for  $\alpha = 0$  the second term does not converge at all. Comparing the exponents in both terms, we find that  $\alpha = \frac{1}{2}$  is optimal. This choice for  $\alpha$  is consistent with the restrictions  $\alpha \neq -\frac{1}{3-r}$  and  $\alpha \neq -\frac{1}{1-r}$  which we made before.

A review of our previous computations in the cases  $\alpha = -\frac{1}{3-r}$  and  $\alpha = -\frac{1}{1-r}$  shows that the term containing  $n^{\alpha(4-r)-\frac{1}{2}}$  does not vanish. Therefore these choices for  $\alpha$  do not improve the asymptotic estimate.

(2) *The case  $r > 4$ :*

Equations (3.1) and (3.2) also hold in this case. We arrive at the inequalities

$$b' = \sum_{k=1}^n \beta'_k \leq -(r-1) \sum_{k=1}^n \left( \frac{\tilde{\tau}_k^{3-r}}{3-r} + \mu^2 \frac{\tilde{\tau}_k^{1-r}}{1-r} \right)$$

$$c = \sum_{k=1}^n \gamma_k \leq (r-1) \sum_{k=1}^n \left( -3\mu \frac{\tilde{\tau}_k^{3-r}}{3-r} - \mu^3 \frac{\tilde{\tau}_k^{1-r}}{1-r} \right) + n(r-1)(\bar{\theta}_1 - 2\bar{\theta}_2)$$

with constants

$$\bar{\theta}_1 = \frac{3\mu}{r-3} + \frac{\mu^3}{r-1}$$

$$\bar{\theta}_2 = \frac{\mu^{4-r}}{4-r} + 3\mu^2 \frac{\mu^{2-r}}{2-r}.$$

With the inequalities

$$\tilde{\tau}_k^{3-r} \leq k^{\alpha(3-r)} \quad \text{and} \quad \tilde{\tau}_k^{1-r} \leq k^{\alpha(1-r)}$$

we arrive at

$$c + sb' \leq (r-1) \left( -\frac{3\mu}{3-r} \sum_{k=1}^n k^{\alpha(3-r)} - \frac{\mu^3}{1-r} \sum_{k=1}^n k^{\alpha(1-r)} \right)$$

$$+ n(r-1)(\bar{\theta}_1 - 2\bar{\theta}_2)$$

$$- s(r-1) \left( \frac{1}{3-r} \sum_{k=1}^n k^{\alpha(3-r)} + \frac{\mu^2}{1-r} \sum_{k=1}^n k^{\alpha(1-r)} \right).$$

Again, approximating integrals by Riemann sums, we obtain, if  $\alpha \neq -\frac{1}{3-r}$  and  $\alpha \neq -\frac{1}{1-r}$

$$\sum_{k=2}^n k^{\alpha(3-r)} \leq \int_1^n x^{\alpha(3-r)} dx = \frac{n^{\alpha(3-r)+1} - 1}{\alpha(3-r) + 1}$$

$$\sum_{k=2}^n k^{\alpha(1-r)} \leq \int_1^n x^{\alpha(1-r)} dx = \frac{n^{\alpha(1-r)+1} - 1}{\alpha(1-r) + 1}.$$

This leads to the inequality

$$6 \frac{c + sb'}{s^3} \leq \frac{6\beta_1}{\sigma^3} \left( -3\mu \frac{n^{\alpha(3-r)-\frac{1}{2}} - n^{-\frac{3}{2}}}{\beta_3} - \frac{3\mu n^{-\frac{3}{2}}}{3-r} - \mu^3 \frac{n^{\alpha(1-r)-\frac{1}{2}} - n^{-\frac{3}{2}}}{\beta_4} \right. \\ \left. - \frac{\mu^3 n^{-\frac{3}{2}}}{1-r} + n^{-\frac{1}{2}}(\bar{\theta}_1 - 2\bar{\theta}_2) \right. \\ \left. - \sigma \frac{n^{\alpha(3-r)} - n^{-1}}{\beta_3} - \sigma \frac{n^{-1}}{3-r} - \sigma \mu^2 \frac{n^{\alpha(1-r)} - n^{-1}}{\beta_4} - \sigma \frac{n^{-1}\mu^2}{1-r} \right)$$

with constants  $\beta_1, \beta_3$ , and  $\beta_4$  as before.

It is the term containing  $n^{-\frac{1}{2}}$  that determines the rate of convergence. We can, for instance, put  $\alpha = 1$  which is in line with the conditions  $\alpha \neq -\frac{1}{3-r}$  and  $\alpha \neq -\frac{1}{1-r}$ .

A review of our previous computations in the cases  $\alpha = -\frac{1}{3-r}$  and  $\alpha = -\frac{1}{1-r}$  shows that the term containing  $n^{-\frac{1}{2}}$  does not vanish. Therefore these choices for  $\alpha$  do not improve the asymptotic estimate.

(3) *The case  $r = 4$ :*

In this case we have

$$\beta'_k = 3 \left( \tilde{\tau}_k^{-1} - \mu \tilde{\tau}_k^{-2} + \frac{\mu^2}{3} \tilde{\tau}_k^{-3} \right) \leq 3 \left( \tilde{\tau}_k^{-1} + \frac{\mu^2}{3} \tilde{\tau}_k^{-3} \right)$$

with  $\tilde{\tau}_k = \tau_k + \mu$  as before.

$$\gamma_k = \int_{-\infty}^{\infty} |y|^3 3 (y + \mu)^{-4} 1_{[1-\mu, \tau_k](y)} dy \leq 9\mu + \mu^3 + 3 \left( \ln \tilde{\tau}_k + 3\mu \tilde{\tau}_k^{-1} + \frac{\mu^3}{3} \tilde{\tau}_k^{-3} \right)$$

where no attempt at sharp estimates for constants has been made. With the inequalities

$$\ln \tilde{\tau}_k \leq \alpha \ln k + \ln 3 \quad \tilde{\tau}_k^{-1} \leq k^{-\alpha} \quad \tilde{\tau}_k^{-3} \leq k^{-3\alpha}$$

we obtain

$$c + sb' \leq 3\alpha \sum_{k=1}^n \ln k + 9\mu \sum_{k=1}^n k^{-\alpha} + \mu^3 \sum_{k=1}^n k^{-3\alpha} + n(3 \ln 3 + \theta) \\ + 3s \sum_{k=1}^n k^{-\alpha} + s\mu^2 \sum_{k=1}^n k^{-3\alpha}$$

with the constant  $\theta = 9\mu + \mu^3$ . It is the sum containing  $\ln k$  that makes the case  $r = 4$  different. Approximating integrals by Riemann sums, we have, if  $\alpha \neq 1$  and

$\alpha \neq \frac{1}{3}$ ,

$$\begin{aligned} \sum_{k=1}^{n-1} \ln k &\leq \int_1^n \ln x \, dx = n \ln n - n + 1 \\ \sum_{k=2}^n k^{-\alpha} &\leq \int_1^n x^{-\alpha} \, dx = \frac{n^{1-\alpha} - 1}{1-\alpha} \\ \sum_{k=2}^n k^{-3\alpha} &\leq \int_1^n x^{-3\alpha} \, dx = \frac{n^{1-3\alpha} - 1}{1-3\alpha}. \end{aligned}$$

Using the equation  $s^2 = \sum_{k=1}^n \sigma^2 = n\sigma^2$ , we obtain the inequality

$$\begin{aligned} &6 \frac{c + sb'}{s^3} \\ &\leq \frac{6}{\sigma^3} \left( 3\alpha (\ln n (n^{-\frac{1}{2}} + n^{-\frac{3}{2}})) - n^{-\frac{1}{2}} + n^{-\frac{3}{2}} \right) + 9\mu \left( \frac{n^{-\frac{1}{2}-\alpha} - n^{-\frac{3}{2}}}{1-\alpha} + n^{-\frac{3}{2}} \right) \\ &\quad + \mu^3 \left( \frac{n^{-\frac{1}{2}-3\alpha} - n^{-\frac{3}{2}}}{1-3\alpha} + n^{-\frac{3}{2}} \right) + n^{-\frac{1}{2}} (3 \ln 3 + \theta) \\ &\quad + 3\sigma \left( \frac{n^{-\alpha} - n^{-1}}{1-\alpha} + n^{-1} \right) + \sigma\mu^2 \left( \frac{n^{-3\alpha} - n^{-1}}{1-3\alpha} + n^{-1} \right). \end{aligned}$$

If we want the term containing  $\ln n n^{-\frac{1}{2}}$  to vanish, we have to put  $\alpha = 0$ . However, in this case the term  $n^{-\alpha}$  does not converge any more. Therefore the choice  $\alpha = \frac{1}{2}$  is optimal.

A review of our computations shows that in the cases  $\alpha = 1$  and  $\alpha = \frac{1}{3}$  the term containing  $\ln n n^{-\frac{1}{2}}$  does not vanish so that the estimate is not improved.  $\square$

#### 4. Concluding Remarks

If  $r \leq 3$ , the variance of the Pareto distribution is infinite and the formulation of the central limit theorem in terms of standardized sums is not possible.

In the case  $3 < r < 4$ ,  $E(|X_1|^3) = \infty$  and convergence is slower than in the Berry-Esseen theorem. Note that the bound in theorem 3.1 is a monotonously decreasing function of  $r$ . As  $r$  approaches 3 from above, the rate of convergence can be arbitrarily slow. For instance, in the case  $r = 3.5$ , the rate of convergence is proportional to  $n^{-\frac{1}{4}}$ . Of course, using this methodology, we have just obtained upper bounds. However, we have done extensive computer simulations that indicate that the rate of convergence in this case is indeed proportional to  $n^{-\frac{1}{4}}$ . Though we have taken great care of numerical inaccuracies, computer simulations should always be taken with caution.

The case  $r = 4$  is borderline because for  $r \leq 4$   $E(|X_1|^3) = \infty$ , but for  $r > 4$   $E(|X_1|^3) < \infty$ . For  $r > 4$  but close to 4, the rate of convergence is faster than for  $r = 4$ . Conversely, for  $r < 4$  but close to 4, the rate of convergence is slower than



for  $r = 4$ . This is true because for large  $n$

$$n^{-\frac{1}{2}} \ln n \leq n^{-\frac{1}{2}+\epsilon}$$

since  $\lim_{n \rightarrow \infty} \frac{\ln n}{n^\epsilon} = 0$ , which can be seen by applying l'Hôpital's rule.

If  $r > 4$ ,  $E(|X_1|^3) < \infty$ , and our estimate in theorem 3.1 gives the same rate of convergence as the Berry-Esseen theorem.

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CLAAS BECKER: HOCHSCHULE RHEINMAIN, 65022 WIESBADEN, GERMANY  
*Email address*: `claas.becker@hs-rm.de`

MANUEL BOHNET: HOCHSCHULE RHEINMAIN, 65022 WIESBADEN, GERMANY  
*Email address*: `manuel_bohnet@web.de`

SARAH KUMMERT: HOCHSCHULE RHEINMAIN, 65022 WIESBADEN, GERMANY  
*Email address*: `sarah.kummert@gmx.de`