

# ON THE RATE OF CONVERGENCE OF PROBABILITIES OF MODERATE DEVIATIONS

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## 1. Introduction

Let  $\{X_n : n \geq 1\}$  be a sequence of independent random variables and write  $S_n = \sum_{k=1}^n X_k$ . Let

$$(1) \quad EX_i = 0, \quad EX_i^2 = \sigma_i^2$$

and let

$$(2) \quad s_n^2 = \frac{1}{n} \sum_{i=1}^n \sigma_i^2, \quad 0 < a \leq s_n \leq A < \infty.$$

Suppose that  $n^{-\frac{1}{2}}s_n^{-1}S_n$  converges in law to the standard normal distribution (see [5, 280] for necessary and sufficient conditions). Let  $\{x_n\}$  be a monotonic sequence of positive real numbers such that  $x_n \rightarrow \infty$  as  $n \rightarrow \infty$ . Then  $x_n^{-1}n^{-\frac{1}{2}}s_n^{-1}S_n \rightarrow 0$  in probability. In particular, choose  $x_n = \sqrt{\log n}$ . Then

$$(3) \quad \Pr \left\{ \left| \frac{S_n}{n} \right| > \varepsilon s_n \sqrt{\frac{\log n}{n}} \right\} \rightarrow 0$$

as  $n \rightarrow \infty$  for all  $\varepsilon > 0$ . In [6] Rubin and Sethuraman call probabilities of the form  $\Pr \{|S_n| > \varepsilon s_n \sqrt{n \log n}\}$  probabilities of moderate deviations and obtain asymptotic forms for such probabilities under appropriate moment conditions.

In this note we study the convergence rate problem for the sequences  $\Pr \{|S_n - a_n| > \varepsilon s_n \sqrt{n \log n}\}$ ,

$$\Pr \left\{ \max_{1 \leq k \leq n} \left| \frac{S_k}{s_n \sqrt{n \log n}} - b_k \right| > \varepsilon \right\} \quad \text{and} \quad \Pr \left\{ \sup_{k \geq n} \left| \frac{S_k}{s_k \sqrt{k \log k}} - c_k \right| > \varepsilon \right\}$$

where  $a_k, b_k, c_k$  are appropriate centering constants. The corresponding problem for the special case of identically distributed summands has

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recently been treated by Davis in [2] where he considers only the first and the third of above sequences.

In Theorems A and B in section 2 we assume that (1) and (2) hold and that the sequence of normed sums  $n^{-\frac{1}{2}}s_n^{-1}S_n$  converges in law to the normal distribution so that, in particular, (3) holds.  $L(\cdot)$  is a nonnegative, nondecreasing and continuous function of slow variation [3].

### 2. Results

**THEOREM A.** For  $t \geq 0$  the following statements are equivalent:

- (a)  $n^t L(n) \Pr \{|S_n| > \varepsilon s_n \sqrt{n \log n}\} \rightarrow 0$  for all  $\varepsilon > 0$ .
- (b)  $n^t L(n) \Pr \{\max_{1 \leq k \leq n} |S_k| > \varepsilon s_n \sqrt{n \log n}\} \rightarrow 0$  for all  $\varepsilon > 0$ .

If  $t > 0$ , the above statements are equivalent to

- (c)  $n^t L(n) \Pr \left\{ \sup_{k \geq n} \left| \frac{S_k}{s_k \sqrt{k \log k}} \right| > \varepsilon \right\} \rightarrow 0$  for all  $\varepsilon > 0$ .

**THEOREM B.**

- (a) For  $t \geq 0$ ,  $\sum_{n=1}^{\infty} n^{t-1} L(n) \Pr \{|S_n| > \varepsilon s_n \sqrt{n \log n}\} < \infty$  for all  $\varepsilon > 0$  if, and only if

$$\sum_{n=1}^{\infty} n^{t-1} L(n) \Pr \left\{ \max_{1 \leq k \leq n} \left| \frac{S_k}{s_n \sqrt{n \log n}} - \text{med} \left( \frac{S_k - S_n}{s_n \sqrt{n \log n}} \right) \right| > \varepsilon \right\} < \infty$$

for all  $\varepsilon > 0$ .

- (b) For  $t > 0$ ,

$$\sum_{n=1}^{\infty} n^{t-1} L(n) \Pr \left\{ \left| \frac{S_n}{s_n \sqrt{n \log n}} - \text{med} \left( \frac{S_n}{s_n \sqrt{n \log n}} \right) \right| > \varepsilon \right\} < \infty$$

for all  $\varepsilon > 0$  if, and only if

$$\sum_{n=1}^{\infty} n^{t-1} L(n) \Pr \left\{ \sup_{k \geq n} \left| \frac{S_k}{s_k \sqrt{k \log k}} - \text{med} \left( \frac{S_k}{s_k \sqrt{k \log k}} \right) \right| > \varepsilon \right\} < \infty$$

for all  $\varepsilon > 0$ .

- (c) For  $t \geq 1$  the following statements are equivalent.

$$(c_1) \sum_{n=1}^{\infty} n^{t-1} L(n) \Pr \{|S_n| > \varepsilon s_n \sqrt{n \log n}\} < \infty \quad \text{for all } \varepsilon > 0.$$

$$(c_2) \sum_{n=1}^{\infty} n^{t-1} L(n) \Pr \left\{ \max_{1 \leq k \leq n} |S_k| > s_n \sqrt{n \log n} \right\} < \infty \quad \text{for all } \varepsilon > 0.$$

$$(c_3) \sum_{n=1}^{\infty} n^{t-1} L(n) \Pr \left\{ \sup_{k \geq n} \left| \frac{S_k}{s_k \sqrt{k \log k}} \right| > \varepsilon \right\} < \infty \quad \text{for all } \varepsilon > 0.$$

THEOREM C. For  $t \geq 1$ ,

$$\sum_{n=1}^{\infty} n^{t-1} (\log n)^t \Pr \{ |S_n| > \varepsilon \sqrt{n \log n} \} < \infty$$

for all  $\varepsilon > 0$  implies  $E|X_k|^{2t} < \infty$  for all  $k$ .

REMARK 1. The  $L(n) = \log n$  case of part (b) of Theorem B has been obtained by Davis [2] in the special case of identically distributed summands.

PROOFS. The (a), (b) equivalence part of Theorem A and part (a) of Theorem B follows from the inequalities

$$\begin{aligned} (4) \quad \Pr \{ |S_n| > \varepsilon s_n \sqrt{n \log n} \} \\ \leq \Pr \left\{ \max_{i \leq k \leq n} \left| \frac{S_k}{s_n \sqrt{n \log n}} - \text{med} \left( \frac{S_k - S_n}{s_n \sqrt{n \log n}} \right) \right| > \varepsilon \right\} \\ \leq 2 \Pr \{ |S_n| > \varepsilon s_n \sqrt{n \log n} \}. \end{aligned}$$

The first of these inequalities is trivial while the second follows from Lévy's inequality [5, 247].

The  $(c_1)$ ,  $(c_2)$  equivalence part of Theorem B follows since

$$\sum_{n=1}^{\infty} n^{t-1} L(n) \Pr \{ |S_n| > \varepsilon s_n \sqrt{n \log n} \} < \infty$$

for all  $\varepsilon > 0$  implies

$$\text{med} \left( \frac{S_k - S_n}{s_n \sqrt{n \log n}} \right) \rightarrow 0 \text{ as } n \rightarrow \infty.$$

For (a), (c) equivalence part of Theorem A and part (b) of Theorem B the proof can be constructed on the lines of [4] and we do not intend to repeat the computations.

The  $(c_1)$ ,  $(c_3)$  equivalence in Theorem B follows similarly using once again the fact that for  $t \geq 1$

$$\sum_{n=1}^{\infty} n^{t-1} L(n) \Pr \{ |S_n| > \varepsilon s_n \sqrt{n \log n} \} < \infty$$

for all  $\varepsilon > 0$  implies

$$\text{med} \left( \frac{S_k - S_n}{s_n \sqrt{n \log n}} \right) \rightarrow 0 \text{ as } n \rightarrow \infty.$$

In the case of Theorem C we use the methods of Baum, Katz and Read [1] and Lemma 1 of Davis [2]. We omit the details.

REMARK 2. In Theorems A and B we may replace  $L(n)$  by an arbitrary non-negative, non-decreasing function of  $n$ .

REMARK 3. The result of Theorem C cannot be improved. This follows trivially by considering the sequences for which  $X_k = 0$ ,  $k = 2, 3, \dots$  and  $E|X_1|^{2t} < \infty$  but  $E|X_1|^{2t+\delta} = \infty$  for  $\delta > 0$ .

### References

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