

RESULTS ON PROBABILITIES OF MODERATE DEVIATIONS¹

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The convergence rate problem for probabilities of moderate deviations is completely solved by giving a necessary and sufficient condition for the existence of the absolute moment of order $c^2 + 2$, $c > 0$, in terms of probabilities of moderate deviations. Furthermore, a result on the rate of convergence of probabilities of moderate deviations is given under a weaker moment condition than in Rubin and Sethuraman (1965).

1. Introduction. The purpose of this paper is to improve a theorem of Rubin and Sethuraman [5], where the asymptotic behavior of probabilities of moderate deviations, i.e. of probabilities of the form

$$P^n\{\mathbf{x} \in X^n : |\sum_{i=1}^n f(x_i)| > c(n \log n)^{\frac{1}{2}}\}$$

is considered under appropriate moment conditions on f .

Furthermore, we shall prove a theorem on the convergence of

$$\sum_{n=1}^{\infty} n^{(c_0^2/2)-1} (\log n)^{(c_0^2/2)+1} P^n\{\mathbf{x} \in X^n : |\sum_{i=1}^n f(x_i)| > c(n \log n)^{\frac{1}{2}}\}$$

for $c > c_0 > 0$, which improves Theorem 3 of Davis [3]. Our Theorem 2 completely solves the convergence rate problem for probabilities of moderate deviations.

2. Results on moderate deviations. Let (X, \mathcal{A}, P) be a probability space and $f: X \rightarrow \mathbb{R}$ an \mathcal{A} -measurable function.

$P^n | \mathcal{A}^n$ denotes the independent product of n identical components $P | \mathcal{A}$.

Let

$$\Phi(t) = (2\pi)^{-\frac{1}{2}} \int_{-\infty}^t \exp[-\frac{1}{2}r^2] dr$$

$$N_{(0, \sigma^2)}(-\infty, t) = \Phi(t\sigma^{-1}).$$

For any measurable g , $P(g) = \int g dP$.

In the following we shall use the inequality

$$(2.1) \quad |c(2\pi \log n)^{\frac{1}{2}} n^{c^2/2} \Phi(-c(\log n)^{\frac{1}{2}}) - 1| \leq c^{-2} (\log n)^{-1},$$

which is standard and follows from Feller (Lemma 2, page 175).

THEOREM 1. *If $P(f) = 0$, $P(f^2) = 1$, and $P(|f|^{c_0^2+2}) < \infty$ for some $c_0 > 0$, then*

$$\sum_{n=1}^{\infty} n^{(c_0^2/2)-1} (\log n)^{(c_0^2/2)+1} |P^n\{\mathbf{x} \in X^n : |\sum_{i=1}^n f(x_i)| > c(n \log n)^{\frac{1}{2}}\} - 2\Phi(-c(\log n)^{\frac{1}{2}})|$$

converges for all $c \geq c_0$.

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By (2.1),

$$\sum_{n=1}^{\infty} n^{(c_0^{2/2})-1} (\log n)^{(c_0^{2/2})+1} \Phi(-c(\log n)^{\frac{1}{2}})$$

converges for all $c > c_0$ and diverges for $c = c_0$. This combined with Theorem 1 implies that the series in Theorem 2 below converges for all $c > c_0$ and diverges for $c = c_0$, if $P(f) = 0$, $P(f^2) = 1$, and $P(|f|^{c_0^{2/2}}) < \infty$.

Hence, our Theorem 1 and Theorem 2 of Davis ([3] page 2020) lead to a complete solution of the convergence rate problem for the probability of moderate deviations:

THEOREM 2. *Assume that $P(f^2) = 1$. Let $c_0 > 0$ be given. Then*

(i) *$P(f) = 0$ and $P(|f|^{c_0^{2/2}}) < \infty$ if and only if*

$$\sum_{n=1}^{\infty} n^{(c_0^{2/2})-1} (\log n)^{(c_0^{2/2})+1} P^n\{\mathbf{x} \in X^n : |\sum_{i=1}^n f(x_i)| > c(n \log n)^{\frac{1}{2}}\}$$

converges for all $c > c_0$.

(ii) *$P(f) = 0$ and $P(|f|^{c_0^{2/2}}) < \infty$ imply the divergence of the series in (i) for $c = c_0$.*

Using the result of Theorem 2(i) together with Proposition 1 of Davis ([3] page 2022) we obtain the following generalization of Theorem 3 of Davis, page 2023 (where $c_0^2 \geq 1$ is assumed to be an integer, and where the exponent of $(\log n)$ is $(c_0^2 - 1)/2$):

THEOREM 3. *If $P(f) = 0$, $P(f^2) = 1$, and $P(|f|^{c_0^{2/2}}) < \infty$ for some $c_0 > 0$, then*

$$\sum_{n=1}^{\infty} n^{(c_0^{2/2})-1} (\log n)^{(c_0^{2/2})+1} P^n\{\mathbf{x} \in X^n : \sup_{k \geq n} |(k \log k)^{-\frac{1}{2}} \sum_{i=1}^k f(x_i)| > c\}$$

converges for all $c > c_0$.

When the moment generating function exists Cramér has shown that

$$P^n\{\mathbf{x} \in X^n : |\sum_{i=1}^n f(x_i)| > n\alpha_n\} \sim \frac{2}{\alpha_n(2\pi n)^{\frac{1}{2}}} \exp[-\frac{1}{2}n\alpha_n^2]$$

if $n\alpha_n^3 \rightarrow 0$, $n\alpha_n^2 \rightarrow \infty$. For moderate deviations, i.e. for the case

$$\alpha_n = c \left(\frac{\log n}{n}\right)^{\frac{1}{2}}$$

Rubin and Sethuraman ([5] page 332, Theorem 4) have established the above result under the much less restrictive moment condition

$$P(|f|^q) < \infty \quad \text{for some } q > c^2 + 2.$$

We obtain a result on the rate of convergence under the condition

$$P(|f|^{c^2+2}) < \infty,$$

namely,

THEOREM 4. *If $P(f) = 0$, $P(f^2) = 1$, and $P(|f|^{c^2+2}) < \infty$ for some $c > 0$, then*

there exists a constant $d > 0$ such that for all $n \in \mathbb{N}$,

$$\left| \frac{P^n\{\mathbf{x} \in X^n : |\sum_{i=1}^n f(x_i)| > c(n \log n)^{\frac{1}{2}}\}}{(2/c(2\pi)^{\frac{1}{2}})n^{-c^2/2}(\log n)^{-\frac{1}{2}}} - 1 \right| \leq d(\log n)^{-\beta/2-\frac{1}{2}},$$

where $\beta = \min(1, c^2)$.

The proof of Theorem 4 follows easily from (2.2) together with Chebychev’s inequality, (2.11), and (2.1).

The bound given in Theorem 4 cannot be improved in general concerning the rate of convergence. To see this, assume that f is distributed as $N_{(0,1)}$. Then,

$$P^n\{\mathbf{x} \in X^n : |\sum_{i=1}^n f(x_i)| > c(n \log n)^{\frac{1}{2}}\} = 2\Phi(-c(\log n)^{\frac{1}{2}}).$$

Furthermore, by Feller (Problem 1, page 193),

$$|c(2\pi \log n)^{\frac{1}{2}}n^{c^2/2}\Phi(-c(\log n)^{\frac{1}{2}}) - 1| \geq c^{-2}(\log n)^{-1}(1 - 3c^{-2}(\log n)^{-1}).$$

PROOF OF THEOREM 1. In the proof we shall use ideas related to those of Rubin and Sethuraman.

To simplify our notations we shall use in the following $d > 0$ and $\alpha > 0$ as generic constants not depending on $n \in \mathbb{N}$.

(i) Let f_n denote f truncated at $r(n \log n)^{\frac{1}{2}}$, where $r = (1/4c) \min(1, c_0^2)$.

We remark that, with this truncation,

$$\begin{aligned} \exp\left[c\left(\frac{\log n}{n}\right)^{\frac{1}{2}}f_n\right] &\leq n^{c_0^2/4} && c_0^2 < 1 \\ &\leq n^{\frac{1}{2}} && c_0^2 \geq 1. \end{aligned}$$

(This will be used in the proof of (2.4)).

We have

$$(2.2) \quad \begin{aligned} &P^n\{\mathbf{x} \in X^n : |\sum_{i=1}^n f(x_i)| > c(n \log n)^{\frac{1}{2}}\} \\ &- P^n\{\mathbf{x} \in X^n : |\sum_{i=1}^n f_n(x_i)| > c(n \log n)^{\frac{1}{2}}\} \\ &\leq nP\{x \in X : |f(x)| > r(n \log n)^{\frac{1}{2}}\}. \end{aligned}$$

As

$$\sum_{n=1}^{\infty} n^{c_0^2/2}(\log n)^{(c_0^2/2)+1}P\{x \in X : |f(x)| > r(n \log n)^{\frac{1}{2}}\}$$

converges by Lemma 1 of Davis ([3] page 2017), it suffices to prove the assertion for the truncated variables f_n .

To this aim, let

$$g_n = \exp\left[c\left(\frac{\log n}{n}\right)^{\frac{1}{2}}f_n\right] \quad \text{and} \quad \beta_n = P(g_n).$$

Define $\tilde{Q}_n | \mathscr{A}$ by $\tilde{Q}_n(A) = \beta_n^{-1}P(1_A g_n)$, $A \in \mathscr{A}$, and let $Q_n | \mathscr{A}^n$ be the independent product of n identical components $\tilde{Q}_n | \mathscr{A}$.

With $\mu_n = \tilde{Q}_n(f_n)$ and $c_n = c(\log n)^{\frac{1}{2}} - n^{\frac{1}{2}}\mu_n$ we have

$$(2.3) \quad \begin{aligned} &P^n\{\mathbf{x} \in X^n : \sum_{i=1}^n f_n(x_i) > c(n \log n)^{\frac{1}{2}}\} \\ &= \beta_n^n \exp[-c(n \log n)^{\frac{1}{2}}\mu_n] \int_{c_n}^{\infty} \exp[-c(\log n)^{\frac{1}{2}}t]F_n(dt) \end{aligned}$$

where $F_n | \mathcal{B}$ is the measure induced by $Q_n | \mathcal{A}^n$ and

$$\mathbf{x} \rightarrow n^{-\frac{1}{2}} \sum_{i=1}^n (f_n(x_i) - \mu_n) .$$

The reduction in (2.3) is standard and has been used by Cramér [2] and Bahadur and Rao [1].

(ii) Obviously,

$$(2.4) \quad \begin{aligned} |P(f_n)| &\leq dn^{-\frac{1}{2}-\alpha} \\ 0 \leq 1 - P(f_n^2) &\leq dn^{-\alpha} \\ P(|f_n|^3 g_n) &\leq dn^{\frac{1}{2}-\alpha} . \end{aligned}$$

Hence, by a Taylor expansion of $s \rightarrow \exp[s]$ about $s = 0$,

$$(2.5) \quad \begin{aligned} \left| \beta_n - 1 - c^2 \frac{\log n}{2n} \right| &\leq dn^{-1-\alpha} \\ \left| \mu_n - c \left(\frac{\log n}{n} \right)^{\frac{1}{2}} \right| &\leq dn^{-\frac{1}{2}-\alpha} \\ |\sigma_n^2 - 1| &\leq dn^{-\alpha} , \end{aligned}$$

where $\sigma_n^2 = \tilde{Q}_n(f_n^2) - \mu_n^2$.

From (2.5) we obtain by standard arguments

$$(2.6) \quad |n^{c^2/2} \beta_n^n \exp[-c(n \log n)^{\frac{1}{2}} \mu_n] - 1| \leq dn^{-\alpha}$$

and

$$(2.7) \quad |c_n| = |c(\log n)^{\frac{1}{2}} - n^{\frac{1}{2}} \mu_n| \leq dn^{-\alpha} .$$

Furthermore, by (2.4) and (2.5),

$$\tilde{Q}(|f_n - \mu_n|^3) \sigma_n^{-3} \leq 8P(|f_n|^3 g_n) \beta_n^{-1} \sigma_n^{-3} \leq dn^{-\alpha+\frac{1}{2}} .$$

Hence, by the Berry-Esséen theorem,

$$(2.8) \quad \sup_{t \in \mathbb{R}} |F_n(-\infty, t) - N_{(0, \sigma_n^2)}(-\infty, t)| \leq dn^{-\alpha} .$$

This implies together with (2.7)

$$(2.9) \quad \left| \int_{c_n}^{\infty} \exp[-c(\log n)^{\frac{1}{2}} t] (F_n - N_{(0, \sigma_n^2)})(dt) \right| \leq dn^{-\alpha} .$$

As

$$\int_{c_n}^{\infty} \exp[-c(\log n)^{\frac{1}{2}} t] N_{(0, \sigma_n^2)}(dt) = \exp[\frac{1}{2} \sigma_n^2 c^2 \log n] \Phi(-c_n \sigma_n^{-1} - c \sigma_n (\log n)^{\frac{1}{2}}) ,$$

it is straightforward to show that (2.1), (2.5), and (2.7) imply,

$$(2.10) \quad \left| \int_{c_n}^{\infty} \exp[-c(\log n)^{\frac{1}{2}} t] N_{(0, \sigma_n^2)}(dt) - n^{c^2/2} \Phi(-c(\log n)^{\frac{1}{2}}) \right| \leq dn^{-\alpha} .$$

From (2.3), (2.6), and (2.8)–(2.10) we finally obtain

$$(2.11) \quad n^{c^2/2} |P^n\{\mathbf{x} \in X^n : \sum_{i=1}^n f_n(x_i) > c(n \log n)^{\frac{1}{2}}\} - \Phi(-c(\log n)^{\frac{1}{2}})| \leq dn^{-\alpha} .$$

As

$$\sum_{n=1}^{\infty} n^{(c_0^2/2)-1} (\log n)^{(c_0^2/2)+1} n^{-\alpha-(c^2/2)}$$

converges for $c \geq c_0$ (recall that $\alpha > 0$), the proof of Theorem 1 is completed.

REFERENCES

- [1] BAHADUR, R. R. and RANGA RAO, R. (1960). On deviations of the sample mean. *Ann. Math. Statist.* **31** 1015–1027.
- [2] CRAMÉR, H. (1938). Sur un nouveau théorème limite de la probabilités. *Actualités Sci. Indust.* No. 736.
- [3] DAVIS, J. A. (1968). Convergence rates for probabilities of moderate deviations. *Ann. Math. Statist.* **39** 2016–2028.
- [4] FELLER, W. (1970). *An Introduction to Probability Theory and its Applications* 1 3rd ed. Wiley, New York.
- [5] RUBIN, H. and SETHURAMAN, J. (1965). Probabilities of moderate deviations. *Sankhyā Ser. A* **27** 325–346.

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