

**SOME RESULTS GIVING RATES OF CONVERGENCE IN  
THE LAW OF LARGE NUMBERS FOR WEIGHTED  
SUMS OF INDEPENDENT RANDOM VARIABLES**

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Let  $\{X_N\}$  for  $N=1, 2, \dots$  be an independent sequence of random variables, and let  $S_N = X_1 + \dots + X_N$ . Probabilists have expended considerable effort investigating the convergence of  $\{(S_N - b_N)/a_N\}$  where  $\{b_N\}$  and  $\{a_N\}$  are sequences of centering and weighting constants respectively. Recently Baum and Katz in [1], [2], and [3] have investigated the rate of convergence to zero of appropriately normalized sums, obtaining (along with some other results) results on the convergence of series of the form  $\sum N^\gamma P\{|S_N - N\mu| > N^\beta\}$  where the  $X$ 's are assumed to be identically distributed with common mean  $\mu$ . Pruitt in [4] obtained a sufficient condition for sums of the form  $S_N = \sum_k a_{N,k} X_k$  to converge to  $\mu$ .

Now let  $\{X_k\}$  be an independent sequence of random variables having finite first moments and define

$$F(y) = \sup_k P\{|X_k - EX_k| > y\}.$$

Let  $\{a_{N,k}\}$  for  $N, k=1, 2, \dots$  be real numbers such that

$$(1) \quad \max_k |a_{N,k}| = CN^{-\beta},$$

$$(2) \quad \sum_k |a_{N,k}| \leq CN^\alpha,$$

$$(3) \quad \sum_k |a_{N,k}|^t \leq CN^{-\rho}.$$

Define

$$S_N = \sum_k a_{N,k}(X_k - EX_k).$$

We have obtained the following five theorems.

**THEOREM 1.** *If  $\rho > 0, \beta > 0, \alpha < \beta, t > 1$ , and  $y^t F(y) \leq M < \infty$  for all  $y > 0$ , then for every  $\epsilon > 0$*

$$P\{|S_N| > \epsilon\} \leq O(N^{-\rho}).$$

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THEOREM 2. If  $\rho > 0, \beta > 0, \alpha < \beta, t > 1$ , and  $y^t F(y) \rightarrow 0$  as  $y \rightarrow \infty$ , then for every  $\epsilon > 0$

$$P\{|S_N| > \epsilon\} = o(N^{-\rho}).$$

THEOREM 3. If  $\beta(t-1) - \alpha > 0, \beta > 0, \alpha < \beta, t > 1$ , and  $F$  satisfies

$$\lim_{y \rightarrow \infty} F(y) = 0 \quad \text{and} \quad \int y^t |dF(y)| < \infty,$$

then for every  $\epsilon > 0$

$$\sum_N N^{\beta(t-1) - \alpha - 1} P\{|S_N| > \epsilon\} < \infty.$$

THEOREM 4. If  $\rho > 0, \beta > 0, \alpha < \beta, t \geq 1$ , and there exists a nonnegative and nonincreasing real-valued function  $G(x) \geq F(x)$  satisfying

$$\lim_{y \rightarrow \infty} G(y) = 0 \quad \text{and} \quad \int_0^\infty y^t |dG(y)| < \infty$$

such that

$$(4) \quad \sup_{x \geq 1} \sup_{y \geq x} \frac{y^t F(y)}{x^t G(x)} = \gamma < \infty,$$

then for every  $\epsilon > 0$

$$(5) \quad \sum_N N^{\rho-1} P\{|S_N| > \epsilon\} < \infty.$$

THEOREM 5. If  $\rho > 0, \beta > 0, \alpha < \beta, t \geq 1$ , and  $F$  satisfies

$$\lim_{y \rightarrow \infty} F(y) = 0 \quad \text{and} \quad \int_0^\infty y^t \log^+ y |dF(y)| < \infty,$$

then (5) holds for every  $\epsilon > 0$ .

One should immediately notice that for  $t \geq 1$  we have

$$\begin{aligned} \sum_k |a_{N,k}|^t &\leq \left(\max_k |a_{N,k}|\right)^{t-1} \sum_k |a_{N,k}| \\ &\leq C^2 N^{\alpha - \beta(t-1)} \end{aligned}$$

so that  $\rho$  can be assumed to be at least as large as  $\beta(t-1) - \alpha$ . Note that Theorem 1 implies  $\sum_N N^{\rho-1-\delta} P\{|S_N| > \epsilon\} < \infty$  for every  $\delta > 0$  so that the additional assumptions used in Theorems 4 and 5 do not give very much more than that already obtained in Theorem 1. The

assumption (4) can readily be violated but most "reasonable" distributions will satisfy it.

Though Theorems 4 and 5 are considerably stronger than Theorem 3, two known results can be obtained as corollaries of Theorem 3 by specializing the constant  $t$  and the constants  $\alpha$  and  $\beta$  from (1) and (2). Theorem 2 of [4] is obtained by setting  $t = 1 + 1/\gamma$ ,  $\alpha = 0$ , and  $\beta = \gamma$ ; a part of Theorem 3 of [2] and [3] is obtained by leaving  $t$  as is and setting  $\alpha = 1 - r/t$  and  $\beta = r/t$  with  $\frac{1}{2} < r/t \leq 1$ .

The motivation for this work was as follows. The average  $\frac{1}{2}X_1 + \frac{1}{4}X_2 + \frac{1}{4}X_3$  is at least as "fine" an average as is  $\frac{1}{2}X_1 + \frac{1}{2}X_2$  and in some sense the first average is "finer" than the second. It has always seemed reasonable to the authors that a finer average than the standard average  $(1/N)\{X_1 + \cdots + X_N\}$  should not hurt convergence any and might actually improve the rate of convergence if one could use the right quantitative measure of the improvement in averaging. The exponent  $\rho$  used in (3) seems to be the correct measure of averaging to use.

The methods used in the proof of these theorems apparently originated with Erdős [5]. The method was modified and improved by Katz [1] and modified still more by Pruitt [4]. Detailed proofs will appear elsewhere.

#### REFERENCES

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