

RATES OF CONVERGENCE FOR WEIGHTED SUMS OF RANDOM VARIABLES¹

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For $N = 1, 2, \dots$ let $S_N = \sum_k a_{N,k} X_k$ where $a_{N,k}$ is a real number for $N, k = 1, 2, \dots$ and $\{X_k\}$ is a sequence of not necessarily independent random variables. For the case $0 < t < 1$, with assumptions closely related to $E|X_k|^t < \infty$ it is shown that the rate of convergence of $P(|S_N| > \epsilon)$ to zero is related to $\sum_k |a_{N,k}|^t$. The theorems presented here extend some of the results in the literature to not necessarily independent sequences $\{X_k\}$.

1. Introduction and summary. Let X_k for $k = 1, 2, \dots$ be a sequence of random variables (not necessarily independent), let $a_{N,k}$ for $N, k = 1, 2, \dots$ be real numbers, let $0 < t < 1$ and ρ_N be a sequence of positive numbers such that $\sum_k |a_{N,k}|^t \leq \rho_N$, and let $S_{N,M} = \sum_{k=1}^M a_{N,k} X_k$ for $N, M = 1, 2, \dots$. In Section 2 with assumptions closely related to $E|X_k|^t < \infty$, we show that for each N , $S_{N,M}$ has an almost sure limit S_N as $M \rightarrow \infty$ and that the rate at which $P(|S_N| > \epsilon)$ converges to zero is related to ρ_N . We conclude with some remarks about the case $t = 1$.

The results of this paper are similar to those of [2], [3], [4], and [7]. In the references cited above the random variables were assumed to be independent. However in [4] it was observed that Theorems 1a and 2a of that paper were valid if the assumption of independence was omitted. Since Theorems 1a and 2a of [4] were generalizations of Theorems 1 and 2 of [2], the question is raised as to whether Theorems 3 and 4 of [2] can be generalized to include dependent sequences $\{X_k\}$ for $0 < t < 1$. In [7] Theorem 4 of [2] was generalized to the case $0 < t < 1$ but the sequence $\{X_k\}$ was still assumed to be independent. The above question is answered in the affirmative by Theorems 3 and 4 of this paper.

2. Results. Using the notation of Section 1, define for $y \geq 0$,

$$F_k(y) = P(|X_k| \geq y) \quad \text{and} \quad F(y) = \sup_k F_k(y).$$

Throughout this paper C will denote various positive constants whose exact values do not matter. Where appropriate, summations will be taken over those values of k for which $a_{N,k} \neq 0$ and integrals will be Lebesgue-Stieltjes integrals.

We now prove the following

LEMMA. *If $y^t F(y) \leq B < \infty$ for all $y > 0$, then for each N as $M \rightarrow \infty$ $S_{N,M}$ has an a.s. limit which we will denote by S_N .*

PROOF. We define $Y_{N,k} = X_k I_{[|a_{N,k} X_k| < 1]}$ and observe that for each N and for

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each $\epsilon > 0$

$$(1) \quad P(\sup_{j \geq 1} |S_{N, M+j} - S_{N, M}| > \epsilon) \leq \sum_{k=M+1}^{\infty} P(|a_{N, k} X_k| \geq 1) + P(\sup_{j \geq 1} |\sum_{k=M+1}^{M+j} a_{N, k} Y_{N, k}| > \epsilon).$$

Since $y^t F(y)$ is bounded for all $y > 0$, we see that the second expression in (1) is bounded by $C \sum_{k=M+1}^{\infty} |a_{N, k}|^t$. Using the inequality of Theorem 1 of [5] with $c_i \equiv 1$ and $r = 1$, the third expression in (1) is bounded by

$$C \sum_{k=M+1}^{\infty} |a_{N, k}| \int_{[0, |a_{N, k}|^{-1})} x |dF_k(x)|$$

(where the last integral is taken with respect to the Lebesgue-Stieltjes measure corresponding to $-F_k$). Integrating by parts we see the last expression is bounded by

$$\begin{aligned} & C \sum_{k=M+1}^{\infty} |a_{N, k}| \int_0^{|a_{N, k}|^{-1}} F(x) dx \\ & \leq C \sum_{k=M+1}^{\infty} |a_{N, k}| \int_0^{|a_{N, k}|^{-1}} x^{-t} dt \\ & = C \sum_{k=M+1}^{\infty} |a_{N, k}|^t. \end{aligned}$$

So expression (1) tends to zero as $M \rightarrow \infty$ and hence $S_{N, M}$ has an almost sure limit (cf. page 115 of [6]).

The following theorems give rates of convergence for $P(|S_N| > \epsilon)$.

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

$$P(|S_N| > \epsilon) = O(\rho_N).$$

THEOREM 2. *If $y^t F(y) \rightarrow 0$ as $y \rightarrow \infty$ and if $\max_k |a_{N, k}| \rightarrow 0$ as $N \rightarrow \infty$, then for every $\epsilon > 0$*

$$P(|S_N| > \epsilon) = o(\rho_N).$$

For Theorems 3 and 4 we assume that ρ_N is of the form $CN^{-\rho}$ and hence there exists a constant β such that

$$(2) \quad \max_k |a_{N, k}| \leq CN^{-\beta}.$$

For Theorem 3 let s be a constant such that $0 < s < t$ and let α be a constant such that $\sum_k |a_{N, k}|^s \leq CN^\alpha$. As in [4] it can be shown that we may assume $\beta \geq \rho/t$, $\beta \geq -\alpha/s$, and $\rho \geq \beta(t-s) - \alpha$.

THEOREM 3. *If $\beta > 0$ and if F satisfies*

$$(3) \quad \lim_{y \rightarrow \infty} F(y) = 0 \quad \text{and} \quad \int_0^\infty y^t |dF(y)| < \infty,$$

then for every $\epsilon > 0$

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

THEOREM 4. *If $\beta > 0$ and if there exists a non-increasing real valued function $G(x)$ satisfying (3) and such that $G(x) \geq F(x)$ for all $x > 0$ and*

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

then for every $\varepsilon > 0$

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

Note. Theorem 1 was proved in [4] and has been included here for completeness. Theorem 2 was proved in [4] under the assumption that $\rho_N \rightarrow 0$ as $N \rightarrow \infty$; however, examining that proof we see that the weaker assumption $\max_k |a_{N,k}| \rightarrow 0$ as $N \rightarrow \infty$ would suffice. Rohatgi in [7] extended Theorem 4 of [2] to the case $0 < t < 1$, but did not give an extension of Theorem 3 of [2]. Theorem 4 of this paper extends the above work to dependent sequences $\{X_k\}$ and removes assumption (6) of [7]. Theorem 3 of this paper extends Theorem 3 of [2] to the case $0 < t < 1$ and to not necessarily independent sequences $\{X_k\}$.

We now prove Theorems 3 and 4.

PROOFS. First we observe that

$$(5) \quad \begin{aligned} P(|S_N| > \varepsilon) &\leq \sum_k F(|a_{N,k}|^{-1}) + P(|\sum_k a_{N,k} Y_{N,k}| > \varepsilon) \end{aligned}$$

where $Y_{N,k}$ is defined as in the proof of the lemma. The proofs of Theorems 3 and 4 are completed by showing that the last two expressions in (5) behave as specified in the theorems.

To show that the second expression in (5) behaves as specified in Theorem 3 one only needs to mimic the proof given for Theorem 3 of [3] found on pages 446 and 447. For Theorem 4 the argument on pages 351 and 352 of [2] suffices. It should be noted that the two arguments cited do not require $\beta(t - s) - \alpha > 0$ or $\rho > 0$ but only that $\beta > 0$.

In considering the last expression in (5), we define $\delta_{N,M} = \text{card. } \{k: M^{-1} \leq |a_{N,k}|\}$ for $N, M = 1, 2, \dots$. Using the Markov Inequality we see that

$$(6) \quad \begin{aligned} \sum_N N^{\beta(t-s)-\alpha-1} P(|\sum_k a_{N,k} Y_{N,k}| > \varepsilon) &\leq C \sum_N N^{\beta(t-s)-\alpha-1} E |\sum_k a_{N,k} Y_{N,k}| \\ &\leq C \sum_N N^{\beta(t-s)-\alpha-1} \sum_k |a_{N,k}| \int_{[0, |a_{N,k}|^{-1})} x |dF_k(x)| \\ &\leq C \sum_N N^{\beta(t-s)-\alpha-1} \sum_k |a_{N,k}| \int_0^{|a_{N,k}|^{-1}} F(x) dx \end{aligned}$$

$$(7) \quad \leq C \sum_N N^{\beta(t-s)-\alpha-1} \sum' |a_{N,k}| \int_0^{|a_{N,k}|^{-1}} F(x) dx$$

$$(8) \quad + C \sum_N N^{\beta(t-s)-\alpha-1} \sum_{M=2}^\infty (\delta_{N,M} - \delta_{N,M-1})(M - 1)^{-1} \int_0^M F(x) dx$$

where the prime on the summation in expression (7) indicates it is to be taken over those values of k for which $|a_{N,k}| \geq 1$. Since $\beta > 0$ expression (7) is finite. Expression (8) is bounded by

$$(9) \quad \begin{aligned} C \sum_N N^{\beta(t-s)-\alpha-1} \sum_{M=2}^\infty (\delta_{N,M} - \delta_{N,M-1}) M^{-1} \sum_{j=1}^M F(j - 1) &\leq C \sum_{j=1}^\infty F(j - 1) \sum_{M=j}^\infty M^{-2} \sum_{N=1}^\infty N^{\beta(t-s)-\alpha-1} \delta_{N,M} . \end{aligned}$$

We now obtain estimates for $\delta_{N,M}$. Since $\sum_k |a_{N,k}|^s \leq CN^\alpha$ and $\max_k |a_{N,k}| \leq CN^{-\beta}$, we see that $\delta_{N,M} = 0$ unless $N \leq CM^{1/\beta}$ and $\delta_{N,M} \leq CN^\alpha M^s$. Therefore

(9) is bounded by

$$\begin{aligned} & C \sum_{j=1}^{\infty} F(j-1) \sum_{M=j}^{\infty} M^{-(2-s)} \sum_{N=1}^{\lfloor CM^{1/\beta} \rfloor} N^{\beta(t-s)-1} \\ & \leq C \sum_{j=1}^{\infty} j^{t-1} F(j-1) \leq C + C \sum_{j=1}^{\infty} j^{t-1} F(j) \\ & \leq C + C \int_0^{\infty} x^t |dF(x)| < \infty . \end{aligned}$$

For Theorem 4 an argument similar to the one beginning at (6) shows that it is sufficient to consider

$$(10) \quad \sum_N N^{\rho-1} \sum_k |a_{N,k}| \int_0^{|a_{N,k}|^{-1}} F(x) dx .$$

From (2) we see that there exists a positive constant A such that $|a_{N,k}|^{-1} \geq AN^\beta$ for $N, k = 1, 2, \dots$. Expression (10) is equal to

$$(11) \quad \sum_N N^{\rho-1} \sum_k |a_{N,k}| \int_0^{AN^\beta} F(x) dx + \sum_N N^{\rho-1} \sum_k |a_{N,k}| \int_{AN^\beta}^{|a_{N,k}|^{-1}} F(x) dx .$$

The first expression in (11) is bounded by

$$(12) \quad C \sum_N N^{-1-\beta(1-t)} \sum_{k=1}^N F(A(k-1)^\beta) [k^\beta - (k-1)^\beta] .$$

Applying the Mean Value Theorem to the function $1 - (1-x)^\beta$, one can show that there exists a constant C depending only on β such that $k^\beta - (k-1)^\beta \leq CK^{\beta-1}$. Hence expression (12) is bounded by

$$\begin{aligned} & C \sum_{k=1}^{\infty} k^{\beta-1} F(A(k-1)^\beta) \sum_{N=k}^{\infty} N^{-1-\beta(1-t)} \leq C \sum_{k=1}^{\infty} k^{\beta t-1} F(A(k-1)^\beta) \\ & \leq C + C \sum_{k=1}^{\infty} k^{\beta t} [F(Ak^\beta) - F(A(k+1)^\beta)] \leq C + C \int_0^{\infty} x^t |dF(x)| < \infty . \end{aligned}$$

Choose N_0 so that $AN_0^\beta \geq 1$. Using (4) we see that

$$\begin{aligned} & \sum_{N=N_0}^{\infty} N^{\rho-1} \sum_k |a_{N,k}| \int_{AN^\beta}^{|a_{N,k}|^{-1}} F(x) dx \\ & \leq C \sum_{N=N_0}^{\infty} N^{\rho+\beta t-1} G(AN^\beta) \sum_k |a_{N,k}| \int_{AN^\beta}^{|a_{N,k}|^{-1}} x^{-t} dx \\ & \leq C \sum_{N=1}^{\infty} N^{\beta t-1} G(AN^\beta) \leq C \int_0^{\infty} x^t |dG(x)| < \infty . \end{aligned}$$

We have shown that the second expression in (11) is finite and the proofs are completed.

In [3] (see Theorem 6), it was shown that $\int_0^{\infty} x^t \log^+ x |dF(x)|$ finite implies the existence of the hypothesized G of Theorem 4. In [3] and [4] the sharpness of these theorems has been investigated for sequences of independent random variables.

For the case $t = 1$ it was shown in [4] that Theorems 1 and 2 are not valid for independent random variables even if it is assumed that $\rho_N \rightarrow 0$ as $N \rightarrow \infty$. However for $t = 1$ it was shown that with the additional hypotheses that $\rho_N \rightarrow 0$ as $N \rightarrow \infty$ and $\limsup_{T \rightarrow \infty} \sup_k |\int_{[-T, T]} x dP(X_k \leq x)| < \infty$ the conclusions of Theorems 1 and 2 hold for independent variables. The following example shows that this is not the case for dependent variables.

EXAMPLE. Let Z, Y_1, Y_2, \dots be independent random variables such that $P(Z = -1) = P(Z = 1) = \frac{1}{2}$, Y_1, Y_2, \dots are identically distributed, $P(Y_1 \geq 0) = 1$, $yP(Y_1 \geq y) \rightarrow 0$ as $y \rightarrow \infty$, and $EY_1 = \infty$. Set $X_k = ZY_k$ for $k = 1, 2, \dots$. Clearly $yF(y) \rightarrow 0$ as $y \rightarrow \infty$ and $\int_{[-T, T]} x dP(X_k \leq x) = 0$ for all T and $k =$

1, 2, \dots . Now $N^{-1} \sum_{k=1}^N Y_k \rightarrow_{\text{a.s.}} \infty$ since $EY_1 = \infty$ and so there exists a sequence of positive numbers $\delta_N \rightarrow \infty$ such that

$$P(|N^{-1} \sum_{k=1}^N Y_k| > \delta_N) \geq \frac{1}{2} \quad \text{for } N = 1, 2, \dots$$

Let $a_{N,k}$ be $(N\delta_N)^{-1}$ for $1 \leq k \leq N$ and zero for $k > N$. For this example $\rho_N = \delta_N^{-1}$ and $\rho_N^{-1}P(|S_N| > 1) \rightarrow \infty$.

In [1] an example of a stationary ergodic sequence X_k was given for which $EX_1 = 0$, $|X_1| = 1$, and

$$\sum_N N^{-1}P(|N^{-1} \sum_{k=1}^N X_k| > \epsilon) = \infty.$$

Hence Theorems 3 and 4 do not hold for $t = 1$ if $\rho = 0$. It would be of interest to know if they hold in the case $t = 1$ for $\beta(t - s) - \alpha > 0$ or $\rho > 0$, respectively.

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