

# ON THE RATE OF CONVERGENCE FOR THE LAW OF LARGE NUMBERS<sup>1</sup>

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**Introduction.** Let  $X_1, X_2, \dots$  be a sequence of independent random variables with a common expectation  $\mu$ . If  $t > 1$  and  $E|X_i|^t \leq C$  for all  $i$  then it is known  $\sum_{i=1}^n X_i/n \rightarrow \mu$  with probability 1. The main result of this paper is that

$$(1) \quad q_n(\epsilon) = P[\sup_{k \geq n} |\sum_{i=1}^k X_i - k\mu|/k \geq \epsilon] = O(1/n^{t-1})$$

for each  $\epsilon > 0$ . Assuming that  $X_1, X_2, \dots$  are also identically distributed, Baum and Katz [1] showed, among other things, that  $EX_1 = \mu$  and  $E|X_1|^t < \infty$  if, and only if,

$$(2) \quad \sum_{i=1}^{\infty} n^{t-2} q_n(\epsilon) < \infty \quad \text{for each } \epsilon > 0.$$

We note that (2) implies that  $q_n(\epsilon) = o(1/n^{t-1})$  (e.g., see the lemma on page 113 of [1]).

**THEOREM.** Let  $X_1, X_2, \dots$  be a sequence of independent random variables with a common expectation  $\mu$ . If  $t > 1$  and  $E|X_i|^t \leq C$  for all  $i$  then, for each  $\epsilon > 0$ ,

$$P[\sup_{k \geq n} |\sum_{i=1}^k X_i - k\mu|/k \geq \epsilon] = O(1/n^{t-1}).$$

The first step in the proof is a generalization of the Hájék-Rényi inequality [2].

**LEMMA.** Suppose  $r \geq 1$ . Let  $Y_1, Y_2, \dots$  be a sequence of independent random variables with  $EY_i = 0$  and  $E|Y_i|^r < \infty$  for all  $i$ . Put  $S_k = Y_1 + \dots + Y_k$  for all  $k$ . Then for each nonincreasing sequence  $\{c_k\}$  of positive constants, for each choice of integers  $0 < n < m \leq \infty$  and for each  $\epsilon > 0$

$$(3) \quad \epsilon^r P[\sup_{n \leq k \leq m} c_k |S_k| \geq \epsilon] \leq c_n^r E|S_n|^r + \sum_{n+1}^m c_k^r [E|S_k|^r - E|S_{k-1}|^r].$$

The usual Hájék-Rényi inequality is obtained from (3) when  $r = 2$ . The proof of (3) follows the proof for  $r = 2$  given in [2] after noting that if  $j \leq k$  then  $E|S_k|^r I_A \geq E|S_j|^r I_A$  where  $A$  is any event defined by  $Y_1, \dots, Y_j$ . (See Loève [3], problem 2, page 263).

**PROOF OF THE THEOREM.** We may take  $\mu = 0$ ,  $\epsilon < 1$ , and  $C > 1$  without loss of generality.

$$(4) \quad \begin{aligned} &P\{\bigcup_{k=n}^{\infty} [|\sum_{i=1}^k X_i| \geq k\epsilon]\} \\ &= P\{\bigcup_{k=n}^{\infty} [|\sum_{i=1}^k X_i| \geq k\epsilon]; (\bigcup_{i=1}^n [|X_i| > n-1]) \cup (\bigcup_{n+1}^{\infty} [|X_k| > k-1])\} \\ &\quad + P\{\bigcup_{k=n}^{\infty} [|\sum_{i=1}^k X_i| \geq k\epsilon]; \bigcap_{i=1}^n [|X_i| \leq n-1]; \bigcap_{n+1}^{\infty} [|X_k| \leq k-1]\}. \end{aligned}$$

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The first term of the right-hand side of (4) is bounded by

$$P\{(\mathbf{U}_1^n [|X_k| > n - 1]) \cup (\mathbf{U}_{n+1}^\infty [|X_k| > k - 1])\} \\ \leq nC/(n - 1)^t + \sum_{n+1}^\infty C/(k - 1)^t = O(1/n^{t-1}).$$

For the second term of the right-hand side of (4) let

$$Y_j = X_j I_{[|X_j| \leq n-1]} - EX_j I_{[|X_j| \leq n-1]}, \quad 1 \leq j \leq n \\ = X_j I_{[|X_j| \leq j-1]} - EX_j I_{[|X_j| \leq j-1]}, \quad j \geq n + 1$$

so that  $Y_1, Y_2, \dots$  is a sequence of independent random variables where, for all  $i$ ,  $EY_i = 0$ ,  $E|Y_i|^t \leq 2^t C$  and  $E|Y_i|^r < \infty$  for all  $r \geq 1$ . Choose  $N$  such that

$$|EX_j I_{[|X_j| \leq k-1]}| = |EX_j I_{[|X_j| > k-1]}| \leq C/(k - 1)^{t-1} < \frac{1}{2}\epsilon$$

for all  $j$  and for all  $k \geq N$ . Thus for  $n \geq N$

$$\{\mathbf{U}_{k=n}^\infty [|\sum_1^k X_i| \geq k\epsilon]; \mathbf{U}_1^n [|X_k| \leq n - 1]; \mathbf{U}_{n+1}^\infty [|X_k| \leq k - 1]\} \\ \subseteq \mathbf{U}_n^\infty [|S_k| \geq k\frac{1}{2}\epsilon]$$

where  $S_k = Y_1 + \dots + Y_k$  for all  $k$ . We now use the lemma to show that  $P\{\mathbf{U}_n^\infty [|S_k| \geq k\epsilon]\} = O(1/n^{t-1})$  for each  $\epsilon > 0$ .

The case  $1 < t \leq 2$  follows easily from the lemma by taking  $r = 2$ ,  $\{c_k\} = \{1/k\}$  and using the argument given below for  $t > 2$ . Thus assume  $t > 2$  and put  $\{c_k\} = \{1/k\}$  and  $r = 2b$  where  $b$  is the smallest integer  $\geq t$ . We see from (3) that

$$(\epsilon)^{2b} P\{\mathbf{U}_n^\infty [|S_k| \geq k\epsilon]\} \leq ES_n^{2b}/n^{2b} + \sum_{n+1}^\infty [ES_k^{2b} - ES_{k-1}^{2b}]/k^{2b}.$$

The proof will be completed if we can show that there exist constants  $A$  and  $B$  such that (a)  $ES_n^{2b} \leq An^{2b-t+1}$  and (b)  $ES_k^{2b} - ES_{k-1}^{2b} \leq Bk^{2b-t}$  for all  $k \geq n + 1$ .

Suppose  $k \geq n \geq N$ . Then

$$ES_k^{2b} = \sum_{\tau=1}^b \sum_{\delta_1, \dots, \delta_\tau \geq 2; \delta_1 + \dots + \delta_\tau = 2b} C_{\delta_1, \dots, \delta_\tau}^{2b} \sum_{1 \leq i_1 < \dots < i_\tau \leq k} EY_{i_1}^{\delta_1} \dots EY_{i_\tau}^{\delta_\tau}$$

For  $i \leq k$ ,  $E|Y_i|^\delta \leq 2^t C$  (or  $2^t C k^{\delta-t}$ ) if  $2 \leq \delta \leq t$  (or  $\delta > t$ ) so that

$$|\sum_{1 \leq i_1 < \dots < i_\tau \leq k} EY_{i_1}^{\delta_1} \dots EY_{i_\tau}^{\delta_\tau}| \leq k^\tau (2^t C)^r k^\alpha = (2^t C)^r k^{\alpha+\tau},$$

where  $\alpha = \sum_{i: \delta_i > t} (\delta_i - t)$ . It is easy to see that  $\alpha + \tau \leq 2b - t + 1$  so that

$$(5) \quad ES_k^{2b} \leq \{\sum_{\tau=1}^b \sum_{\delta_1, \dots, \delta_\tau \geq 2; \delta_1 + \dots + \delta_\tau = 2b} C_{\delta_1, \dots, \delta_\tau}^{2b} (2^t C)^\tau\} k^{2b-t+1}.$$

The proof of (a) now follows from (5) by taking  $k = n$ .

To prove (b) note that  $ES_k^{2b} - ES_{k-1}^{2b} = \sum_{j=2}^{2b} C_j^{2b} EY_k^j ES_{k-1}^{2b-j}$ . Because  $E|Y_k^j| \leq 2^t C$  for  $2 \leq j < b$  (or  $\leq 2^t C k^{j-t}$  if  $j \geq b$ ) and  $E|S_{k-1}^{2b-j}| \leq 2^t C (k-1)^{2b-j}$  for  $b < j \leq 2b$  we conclude that  $|EY_k^j ES_{k-1}^{2b-j}|$  is

$$(6a) \quad \leq (2^t C)^2 E|S_{k-1}^{2b-j}| \leq (2^t C)^2 ES_{k-1}^{2b-2}, \quad 2 \leq j < b;$$

$$(6b) \quad \leq (2^t C) k^{b-t} E|S_{k-1}^b|, \quad j = b;$$

$$(6c) \quad \leq (2^t C)^2 (k-1)^{2b-j} \leq (2^t C)^2 k^{2b-t}, \quad b < j \leq 2b.$$

Using (5) there is a constant  $D$  such that  $ES_{k-1}^{2b-2} \leq ES_k^{2b-2} \leq (ES_k^{2b})^{(1-1/b)} \leq Dk^{2b-t}$ . Similarly,  $E|S_{k-1}^b| \leq E|S_k^b| \leq (ES_k^{2b})^{\frac{1}{2}} \leq D'k^b$  for some constant  $D'$ . Finally then, we see from (6) that there is a constant  $H$  such that

$$EY_k^j ES_{k-1}^{2b-j} \leq Hk^{2b-t}, \quad 2 \leq j \leq 2b,$$

and (b) is proven.

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