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On the Law of the Iterated Logarithm for Sequences of Dependent Random Variables

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Abstract—Sufficient conditions for the applicability of the law of the iterated logarithm to sequences of dependent random variables are obtained. As a corollary, a theorem on the law of the iterated logarithm for a sequence of *m*-orthogonal random variables is proved.

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Theorem 1. Let $\{Y_n; n = 1, 2, ...\}$ be a sequence of random variables on a probability space, and let $\{a_n; n = 1, 2, ...\}$ be a sequence of positive numbers such that $a_n \to \infty$ as $n \to \infty$ and the following condition (Condition A) holds: for any $\varepsilon > 0$,

$$a_{n+r} \ge a_n(1-\varepsilon) \tag{1}$$

for all $r \ge 1$ and all sufficiently large n. Suppose that

$$\sum_{n=1}^{\infty} P\left(\max_{[c^n] \le k < [c^{n+1}]} Y_k > (1+\varepsilon)a_{[c^n]}\right) < \infty$$
⁽²⁾

for some c > 1 and any $\varepsilon > 0$. Then

$$\limsup \frac{Y_n}{a_n} \le 1 \quad almost \ surely. \tag{3}$$

Proof. Let δ be any positive number, and let c > 1. By virtue of Condition A, we have

$$P(Y_n > (1 + \delta)a_n \text{ i.o.})$$

$$\leq P\left(\max_{[c^n] \leq k < [c^{n+1}]} Y_k > (1 + \delta)a_{[c^n]}(1 - \varepsilon) \text{ i.o.}\right)$$

for any $\varepsilon > 0$. Therefore,

$$P(Y_n > (1 + \delta)a_n \text{ i.o.})$$

$$\leq P\left(\max_{[c^n] \leq k < [c^{n+1}]} Y_k > \left(1 + \frac{\delta}{2}\right)a_{[c^n]} \text{ i.o.}\right),$$

provided that the positive number ε is so small that $(1 + \delta)(1 - \varepsilon) > 1 + \delta/2$.

Taking c satisfying condition (2) and applying the Borel–Cantelli lemma, we obtain inequality (3).

Condition A is a weakening of the condition that the normalizing number sequence $\{a_n\}$ is nondecreasing; this condition makes it possible to apply Theorem 1 to sequences of random variables when the normalizing sequence is not nondecreasing but satisfies Condition A. Using Theorem 1, we can obtain a sufficient condition for the applicability of the law of the iterated logarithm to sums of *m*-dependent random variables. Recall that a sequence $\{X_n; n = 1, 2, ...\}$ of random variables is referred to as a *sequence of m-dependent random variables*, where *m* is a nonnegative integer, if the random vectors $(X_p, ..., X_q)$ and $(X_r, ..., X_s)$ are independent for any integer *p*, *q*, *r*, and *s* satisfying the conditions $1 \le p \le q < r \le s$ and r - q > m.

The notion of a sequence of *m*-dependent random variables was introduced by Hoeffding and Robbins in a classical paper [1], which also contains conditions for the applicability of the central limit theorem to sequences of *m*-dependent random variables. At present, the literature on limit theorems for sums of *m*-dependent random variables has become very extensive.

In [2], the notion of a sequence of *m*-orthogonal random variables was introduced. Given a nonnegative integer *m*, we say that a sequence $\{X_n; n = 1, 2, ...\}$ of random variables defined on some probability

space is a sequence of *m*-orthogonal random variables if $EX_n^2 < \infty$ for any *n* and $E(X_kX_j) = 0$ provided that |k-j| > m. In particular, any sequence of 0-orthogonal random variables is a sequence of orthogonal random variables. If $\{X_n\}$ is a sequence of *m*-orthogonal random variables. This remains true under the replacement of *m*-dependence by the weaker condition of pairwise *m*-dependence. Note that the verification of *m*-orthogonal random variables is of certain interest thanks to the great attention given to limit theorems for sums of orthogonal random variables and sums of *m*-dependent random variables. Below, we present a theorem on the law of the iterated logarithm for sequences of *m*-orthogonal random variables.

Theorem 2. Let $\{X_n; n = 1, 2, ...\}$ be a sequence of *m*-orthogonal random variables with zero expectations. *Put*

$$S_n = \sum_{k=1}^n X_k, \quad B_n = ES_n^2, \quad \chi(n) = (2B_n \log \log B_n)^{1/2}.$$

Suppose that

$$B_n \to \infty,$$
 (4)

$$\frac{EX_n^2}{B_n} \to 0 \tag{5}$$

as $n \to \infty$, and

$$\sum_{n=1}^{\infty} P\left(\max_{[c^n] \le k < [c^{n+1}]} S_k > (1+\varepsilon)\chi([c^n])\right) < \infty$$
(6)

for some c > 1 and any $\varepsilon > 0$. Then

$$\limsup \frac{S_n}{\chi(n)} \le 1 \quad almost \ surely.$$
⁽⁷⁾

Proof. Let us show that the sequence $\{B_n\}$ satisfies Condition A. By virtue of the Cauchy–Bunyakovskii inequality, we have

$$B_n = E(S_{n+1} - X_{n+1})^2 = B_{n+1} + EX_{n+1}^2 - 2E(S_{n+1}X_{n+1}) = B_{n+1} + EX_{n+1}^2 + 2\Theta(ES_{n+1}^2EX_{n+1}^2)^{1/2}$$

where $|\theta| \le 1$. Therefore, we can write

$$\frac{B_n}{B_{n+1}} = 1 + \frac{EX_{n+1}^2}{B_{n+1}} + 2\Theta \left(\frac{EX_{n+1}^2}{B_{n+1}}\right)^{1/2}$$

Condition (5) implies

$$\frac{B_n}{B_{n+1}} \to 1 \quad (n \to \infty). \tag{8}$$

For any integer $p \ge 1$, we have

$$B_{n+p} = B_n + E(X_{n+1} + \dots + X_{n+p})^2 + 2E[S_n(X_{n+1} + \dots + X_{n+p})] \ge B_n + 2E[S_n(X_{n+1} + \dots + X_{n+p})],$$
$$E[S_n(X_{n+1} + \dots + X_{n+p})] = E[(X_{n-m+1} + \dots + X_n)(X_{n+1} + \dots + X_{n+p})]$$

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for n > m by virtue of the *m*-orthogonality condition. Again, using this condition, we can easily show that

$$E\left|S_{n}(X_{n+1}+\ldots+X_{n+p})\right| \leq \sum_{j=0}^{m} \sum_{i=0}^{m-j} \left|E(X_{n-j}X_{n+i})\right|.$$

Hence

$$B_{n+p} \ge B_n - 2\sum_{j=0}^m \sum_{i=0}^{m-j} \left| E(X_{n-j}X_{n+i}) \right| \ge B_n - 2\sum_{j=0}^m \sum_{i=0}^{m-j} \left(EX_{n-j}^2 EX_{n+i}^2 \right)^{1/2}.$$
(9)

It follows from (5) and (8) that, for any fixed integer *p*, we have

$$\frac{EX_{n+p}^{2}}{B_{n}} = \frac{EX_{n+p}^{2}}{B_{n+p}} \frac{B_{n+p}}{B_{n+p-1}} \dots \frac{B_{n+1}}{B_{n}} \to 0 \quad (n \to \infty).$$
(10)

Relations (9) and (10) imply

$$B_{n+s} \ge B_n \left(1 - 2\sum_{j=0}^m \sum_{i=0}^{m-j} \left(\frac{EX_{n-j}^2}{B_n} \frac{EX_{n+i}^2}{B_n} \right)^{1/2} \right),$$

$$B_{n+s} \ge B_n (1-\varepsilon)$$
(11)

for any $\varepsilon > 0$ and $s \ge 1$ and all sufficiently large *n*. Thus, the sequence $\{B_n\}$ satisfies Condition A.

Inequality (11) remains valid for B_n replaced by $\chi(n) = (2B_n \log \log B_n)^{1/2}$. Thus, the sequence of $a_n = \chi(n)$ satisfies Condition A. Condition (2) of Theorem 1 also holds for $Y_n = S_n$ and $a_n = \chi(n)$ by virtue of (6). According to Theorem 1, relation (7) holds.

Theorem 2 remains true under the replacement of *m*-orthogonality by *m*-dependence (provided that the other conditions in the theorem are satisfied) or by the even weaker condition of pairwise *m*-dependence. The resulting theorem generalizes Theorem 2 of [3], in which conditions (4) and (5) are assumed to hold and, instead of (6), it is required that, for any b > 1, there exist positive constants *C* and δ such that

$$P\left(\max_{1\le k\le n} S_k \ge b\chi(n)\right) \le C(\log B_n)^{-1-\delta}$$

for all sufficiently large n. The theorem of [2] differs from this theorem in that the m-dependence condition is replaced by the m-orthogonality condition.

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