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# An LIL Via Self-Normalization for Sequences of Sign-Invariant Random Variables

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#### ABSTRACT

Some extensions of the law of the iterated logarithm via self-normalization are obtained for sequences of sign-invariant random variables.

KEYWORDS: LIL, Self-normalization, Sign-invariant random variables.

#### 1. INTRODUCTION

Let  $X_1, X_2, \ldots$  be a sequence of random variables and let

$$S_n = X_1 + \dots + X_n \qquad (n \ge 1),$$

$$V_n^2 = X_1^2 + \dots + X_n^2$$
  $(n \ge 1).$ 

The first result on the law of the iterated logarithm (LIL) via self-normalization was obtained by Marcinkiewicz(1937) who observed that for any symmetric distribution

$$\limsup_{n \to \infty} \frac{S_n}{(2V_n^2 L_2 V_n^2)^{1/2}} \le 1 \quad \text{a.s.}, \tag{1.1}$$

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where  $Lx = \max(1, \log_e x)$  and  $L_2x = L(Lx)$ . For this result the random variables need not be identically distributed, just independent and  $V_n^2 \to \infty$  with probability 1. Later, results were obtained with a refinement of (1.1) by Griffin and Kuelbs(1991). They also treated identically distributed case. The purpose of this paper is to generalize Griffin and Kuelbs' LIL via self-normalizations to sequences of "sign-invariant" random variables. We also consider LIL via self-normalization for a sequence of exchangeable and sign-invariant random variables.

## 2. EXTENSTIONS TO SIGN-INVARIANT SEQUENCES

Let  $(X_n)$  be a sequence of random variables on a probability space  $(\Omega, \mathcal{F}, P)$  such that every finite dimensional distribution function(d.f.) of the sequence is invariant under any changes in the sign of  $(X_n)$ . Such random variables were called "sign-invariant" (Berman,1962,1965). It is obvious that a sequence of independent random variables with distribution functions  $F_n(x)$  of  $X_n$  is sign-invariant if and only if  $F_n(x)$  is symmetric, i.e., every one-dimensional d.f. is invariant under changes in signs.

Lemma 1. Let  $X_1, X_2, ...$  be sign-invariant random variables on a probability space  $(\Omega, \mathcal{F}, P)$ . Then there exists a regular conditional distribution, say  $P^{\omega}$ , for  $X = (X_1, X_2, ...)$  given  $\sigma(|X_n|, n \geq 1)$  such that for each  $\omega \in \Omega$  the coordinate random variables  $\{\xi_n, n \geq 1\}$  of probability space  $(\mathcal{R}^{\infty}, \mathcal{B}^{\infty}, P^{\omega})$  are independent and  $P^{\omega}(\xi_n = X_n(\omega)) = \frac{1}{2} = P^{\omega}(\xi_n = -X_n(\omega))$  for all n.

**Proof.** Let 
$$B = \{\xi_1 > t_1, \dots, \xi_n > t_n\}, t_i \ge 0, i = 1, 2, \dots, n$$
. Note that 
$$P^{\omega}(B) = \begin{cases} \frac{1}{2^n}, & \text{if } |X_i(\omega)| > t_i, & i = 1, \dots, n \\ 0, & \text{if not} \end{cases},$$

and

$$P(|X_1| > t_1, \dots, |X_n| > t_n) = 2^n P(X_1 > t_1, \dots, X_n > t_n).$$

Then

$$\int P^{\omega} (\xi_1 > t_1, \dots, \xi_n > t_n) dP(\omega)$$

$$= 2^{-n} P(|X_1| > t_1, \dots, |X_n| > t_n)$$

$$= 2^{-n} 2^n P(X_1 > t_1, \dots, X_n > t_n)$$

$$= P(X_1 > t_1, \dots, X_n > t_n).$$

This proves the Lemma.

The sequences  $(X_n)$  is "exchangeable" if the joint d.f. of  $(X_1, X_2, \ldots, X_n)$  say  $G_n(x_1, \ldots, x_n)$ , is a symmetric function for each n. According to the fundamental theorem of de Finetti, there exists a sub- $\sigma$ -field of the  $\sigma$ -field  $\mathcal{F}$  and a conditional d.f.  $G^{\omega}(x)$  such that the  $(X_n)$  are conditionally independent given  $\mathcal{G}$  with the common conditional d.f.  $G^{\omega}(x)$ . More specifically, one may write,

$$G_n(x_1, \dots, x_n) = \int_{\Omega} G^{\omega}(x_1) \cdots G^{\omega}(x_n) dP(\omega), \qquad (2.1)$$

where  $G^{\omega}(x)$  is a d.f. for each  $\omega \in \Omega$ , and an  $\mathcal{F}$ -measurable function for each x. In general, for any set  $H \in \mathcal{F}$ ,

$$P(H) = \int_{\Omega} P^{\omega}(H) dP(\omega), \qquad (2.2)$$

where  $P^{\omega}(H)$  is the conditional probability of H computed under the assumption that the  $X_n$  are mutually independent with the common conditional d.f.  $G_{\omega}(x)$ .

The following lemma, due to Berman(1962), is used.

**Lemma 2.** If  $(X_n)$  are exchangeable and sign-invariant random variables, then in the representation (2.1), for almost all  $\omega$ 

$$G^{\omega}(x) = 1 - G^{\omega}(-x),$$

for all x, in the continuity set of  $G^{\omega}(x)$ .

**Theorem 1.** Let  $(X_n)$  be exchangeable and sign-invariant with  $E(X_1^2) < \infty$  and let  $\Delta = \{\omega : G_{\omega}(\cdot) \text{ is degenerate at } 0\}$ . If  $\phi$  is eventually nondecreasing and positive, then

$$P(S_n > V_n \phi(n) \text{ i.o.}) = 0 (= 1 - P(\Delta))$$

according as

$$J(\phi) < \infty (= \infty),$$

where

$$J(\phi) = \sum_{n=1}^{\infty} \frac{\phi(n)}{n} e^{-\phi^2(n)/2}.$$
 (2.3)

**Proof.** According to (2.2),

$$P(S_n > V_n \phi(n) \text{ i.o.}) = \left( \int_{\Delta} + \int_{\Omega - \Delta} \right) P^{\omega} \left( S_n > V_n \phi(n) \text{ i.o.} \right) dP(\omega).$$

For  $\omega \in \Delta$ ,  $S_n = V_n = 0$  a.s. for all n, hence  $P^{\omega}(S_n > V_n \phi(n))$  i.o.) = 0; on the other hand, by Theorem 2(Griffin and Kuelbs, 1991) and Lemma 2, for almost all  $\omega \notin \Delta$ ,

$$P^{\omega}(S_n > V_n \phi(n) \text{ i.o.}) = 0 (= 1)$$

according as

$$J(\phi) < \infty (= \infty).$$

Hence the theorem follows.

**Theorem 2.** Let  $X_1, X_2, \ldots$  be sign-invariant random variables with  $V_n^2 \to \infty$  w.p. 1. If  $\phi$  is nondecreasing and positive eventually and  $J(\phi) < \infty$  where  $J(\phi)$  is as in (2.3), then

$$P(S_n > V_n \phi(V_n^2) \text{ i.o.}) = 0.$$

**Proof.** We set  $T_n = \xi_1 + \cdots + \xi_n$ ,  $W_n^2 = \xi_1^2 + \cdots + \xi_n^2$ . By Lemma 1  $\xi_1, \xi_2, \ldots$  is independent and symmetric with respect to  $P^{\omega}$  for all  $\omega$  and by

the assumption  $V_n^2 \to \infty$  w.p.1,  $P^{\omega}(W_n^2 \to \infty) = 1$  w.p. 1. Hence it follows from Theorem 1(Griffin and Kuelbs, 1991) that

$$P(P^{\omega}(T_n > W_n \phi(W_n^2) \text{ i.o}) = 0) = 1.$$

Therefore we have

$$P(S_n > V_n \phi(V_n^2) \quad \text{i.o.})$$

$$= \int P(S_n > V_n \phi(V_n^2) \quad \text{i.o.} \quad ||X_1|, |X_2|, \dots) dP$$

$$= \int P^{\omega}(T_n > W_n \phi(W_n^2) \quad \text{i.o.}) dP = 0.$$

This completes proof.

**Theorem 3.** Let  $X, X_1, X_2, ...$  be stationary, ergodic and sign-invariant with  $0 < E(X^2) < \infty$ . If  $\phi$  is eventually nondecreasing and positive, then

$$P(S_n > V_n \phi(n) \quad \text{i.o}) = 0 (= 1) \tag{2.4}$$

according as

$$J(\phi) < \infty (= \infty),$$

where  $J(\phi)$  is as in (2.3).

To prove this we need the following lemmas.

**Lemma 3.** Let  $\{X_n, n \geq 1\}$  be a sequence of independent random variables with  $P\{X_n = \pm x_n\} = 1/2$ , and let  $|x_n| \leq n^{1/2}$  for all large n and  $(x_1^2 + \cdots + x_n^2)/n$  converges to a positive constant. If

$$\sum_{n=1}^{\infty} X_{n+1}^2 I(|X_{n+1}| \le C) \frac{\phi(n)}{n} e^{-\phi^2(n)/2}$$

diverges with probabilty 1 for some C > 0 and  $\phi(n) \uparrow \infty$ , then

$$P(S_n > (x_1^2 + \dots + x_n^2)\phi(n) \text{ i.o}) = 1.$$

**Proof.** Just follow the step 2 and step 3 in the proof Theorem 2 (Griffin and Kuelbs, 1991).

**Lemma 4.** Let  $\phi$  be eventually nondecreasing and positive such that  $L_2n \leq \phi^2(n) \leq 3L_2n$  for large n, and let  $\sum_{n=1}^{\infty} f(n) = \infty$ , where  $f(n) = \frac{\phi(n)}{n}e^{-\phi^2(n)/2}$ . Then if  $\lim_{n\to\infty} \frac{1}{n}\sum_{i=1}^n a_i = \alpha(0 < \alpha < 1)$  where  $a_i = 0$  or 1, we have

$$\sum_{n=1}^{\infty} a_n f(n) = \infty.$$

**Proof.** Let  $\{a_{n_k}\}$  be the subsequence of  $\{a_n\}$  such that  $a_{n_k} = 1$  for all k, then  $\lim_{n\to\infty} \frac{n_k}{k} = 1/\alpha$  and hence for large k,  $n_k \leq (\left[\frac{1}{\alpha}\right] + 1)k$  where [x] stands for the integer part of x. Let  $g(n) = \frac{(L_2 n)^{1/2}}{n} e^{-\phi^2(n)/2}$  and let  $\left[\frac{1}{\alpha}\right] + 1 = \beta$ , then g is eventually nonincreasing, and hence  $g(n_k) \geq g(\beta k)$  for large k. Now we note that  $3^{-1/2}f(n) \leq g(n)$  and hence  $\sum_{n=1}^{\infty} g(\beta k) = \infty$ , since  $\sum_{n=1}^{\infty} f(n) = \infty$ . Therefore we have

$$\sum_{n=1}^{\infty} a_n f(n) = \sum_{k=1}^{\infty} f(n_k) \ge \sum_{k=1}^{\infty} g(n_k) \ge c + \sum_{k=1}^{\infty} g(\beta k)$$

for some c, which proves the lemma.

**Proof of theorem.** As in the proof Theorem 2(Griffin and Kuelbs, 1991) it is standard argument to show that if Theorem 3 holds for  $\phi(n)$  nondecreasing, positive and such that

$$L_2 n \le \phi^2(n) \le 3L_2 n \tag{2.5}$$

for large n, then Theorem 3 holds without the restriction (2.5). Hence we will assume that (2.5) holds. The proof that  $J(\phi) < \infty$  implies the probability in (2.4) is zero just follows the proof of Theorem 2 (Griffin and Kuelbs, 1991) using Theorem 2 and  $V_n^2 \sim nEX^2$  which comes from the Brikhoff's ergodic theorem. Now we show that  $J(\phi) = \infty$  implies  $P(S_n > V_n \phi(n) \text{ i.o.}) = 1$ . Since  $E(X^2I|X| \le C) > 0$ , for some C > 0, we can choose B > 0 such that

$$P(B \le |X| \le C) > 0. \text{ Now}$$

$$\sum_{n=1}^{\infty} X_{n+1}^{2} I(|X_{n+1}| \le C) \frac{\phi(n)}{n} e^{-\phi^{2}(n)/2}$$

$$\ge \sum_{n=1}^{\infty} B^{2} I(B \le |X_{n+1}| \le C) \frac{\phi(n)}{n} e^{-\phi^{2}(n)/2}.$$

By the ergodic theorem,

$$\sum_{k=1}^{n} I(B \le |X_{k+1}| \le C) \sim nP(B \le |X| \le C)$$

hence by Lemma 4,

$$P\bigg(E\bigg(\sum_{n=1}^{\infty}X_{n+1}^{2}I\Big(|X_{n+1}|\leq C\Big)\frac{\phi(n)}{n}e^{-\phi^{2}(n)/2}\Big||X_{1}|,|X_{2}|,\ldots\bigg)=\infty\bigg)=1.$$

Also we can easily check by Borel-Cantelli lemma and ergodic theorem that  $P(|X_j| > j^{1/2} \text{ i.o.} | |X_1|, |X_2|, \dots) = 0$  and  $P(V_n^2/n \to EX^2 | |X_1|, |X_2|, \dots) = 1$ . Hence by Lemma 3,  $P(S_n > V_n \phi(n) \text{ i.o.} | |X_1|, |X_2|, \dots) = 1$  a.s. Therefore we have

$$P(S_n > V_n \phi(n) \text{ i.o.})$$

$$= \int P(S_n > V_n \phi(n) \text{ i.o.} ||X_1|, |X_2|, \dots) dP$$

$$= 1,$$

which completes the proof.

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