

Stopping Rules for S_n/n , and the Class $L \log L$

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1. Introduction

If $(f_1, f_2, \dots) = (f_n)$ is a stochastic process a stopping rule for (f_n) will be defined to be a non-defective positive integer valued random variable t such that for every positive integer k $\{t=k\} \in \sigma(f_1, \dots, f_k)$, the σ -field generated by f_1, \dots, f_k . Here we will be concerned with stopping rules for the sequence (S_n/n) where $S_n = X_1 + \dots + X_n$, X_1, X_2, \dots independent identically distributed random variables with finite expectation, investigating the possibility of the existence of stopping rules t for (S_n/n) such that $E(S_t/t) = \infty$. It is proved (Theorem 1) that such stopping rules exist if and only if $E(X_1 \log^+ X_1) = \infty$, where $\log^+ a = \log(a)$ if $a > 1$, 0 if $a \leq 1$. This immediately implies the result of Burkholder in [1] that if $E(X_1 \log^+ X_1) = \infty$ then $E(\sup(S_n/n)) = \infty$. Similar results are proved if (X_n/n) is looked at in place of (S_n/n) .

Since $E|S_n/n| \leq E|X_1|$ for all n , recent results (see, for example, [4]) on stopping a stochastic process (f_n) to get $E|f_t| = \infty$ are not applicable here, since they require that $\sup E|f_n| = \infty$. The work by Chow, Robbins and others on stopping S_n/n to maximize $E(S_t/t)$ has dealt with essentially different (although obviously related) questions than this paper.

From now on “ t is a stopping rule” will mean that t is a stopping rule for the particular process under discussion. Note that since $\sigma(S_1, S_2/2, \dots, S_n/n) = \sigma(X_1, X_2/2, \dots, X_n/n) = \sigma(X_1, \dots, X_n)$, the stopping rules for the processes (S_n/n) , (X_n/n) and (X_n) are the same.

2. Some Inequalities Concerning S_n/n

Let $f(n) = \sum_{k=1}^{n-1} (n-k)/k$. By approximating $f(n)$ with an integral it is seen that $\lim_{n \rightarrow \infty} f(n)/((n+1) \log(n+1)) = 1$. Thus there is a positive number which will be called ε such that

$$f(n) > \varepsilon(n+1) \log(n+1), \quad n = 2, 3, \dots$$

In this section X will be a random variable with finite expectation and X_1, X_2, \dots will be independent random variables each with the same distribution as X . B will stand for $\max(1, E|X|)$, and if Y is a random variable Y^+ will designate $\max(Y, 0)$.

By a well known inequality, if $\lambda > 0$ then $P(\sup |S_n/n| > \lambda) \leq E|X_1|/\lambda$. For a proof see the introduction to [3]. Thus

$$P(\sup_n |S_n/n| \geq 2B) \leq \frac{1}{2}. \tag{1}$$

Let $A_n = ((X_n/n) - 4B)^+ I(\sup_{k < n} |S_k/k| \leq 2B)$. Then, since $X_n/n \leq S_n/n + |S_{n-1}/(n-1)|$,

$$A_n \leq S_n/n \text{ on } \{A_n > 0\} \tag{2}$$

Also, since if X_n/n exceeds $4B$ then at least one of $|S_n/n|, |S_{n-1}/(n-1)|$ must exceed $2B$,

$$P(A_i > 0, A_j > 0) = 0 \text{ if } i \neq j. \tag{3}$$

Next a lower bound is found for $\sum EA_k$. Since X_n and $\{\sup_{k < n} |S_k/k| \leq 2B\}$ are independent, (1) implies $EA_n \geq E((X_n/n) - 4B)^+/2$. Let $p_i = P(4Bi \leq X < 4B(i+1))$. Then

$$\begin{aligned} \sum_1^\infty EA_n &\geq (\frac{1}{2}) \sum_1^\infty E(((X_n/n) - 4B)^+) \\ &= (\frac{1}{2}) \sum_{n=1}^\infty (1/n) E((X - 4Bn)^+) \\ &\geq (\frac{1}{2}) \sum_{n=1}^\infty (4B/n) \sum_{k=n+1}^\infty p_k(k-n) \\ &= 2B \sum_{k=2}^\infty p_k f(k) \geq 2B\epsilon \sum_{k=2}^\infty p_k(k+1) \log(k+1) \\ &\geq 2B\epsilon [E((X^+/4B) \log^+(X^+/4B)) - p_1 2 \log 2] \\ &\geq 2B\epsilon [(1/4B) E(X^+ (\log^+ X^+ - \log 4 - \log B)) - 2 \log 2]. \end{aligned}$$

Since $B \leq 1 + E|X|$, this completes the proof of the following lemma, noting that $\log^+ X^+ = \log^+ X$.

Lemma. *There are positive constants, K, C such that if X is an integrable random variable and X_1, X_2, \dots are independent random variables each having the distribution of X then $\sum EA_n \geq CEX \log^+ X - KE|X| \log^+ E|X| - K$.*

3. Construction of Stopping Times

Suppose the conditions of the lemma are satisfied and that $E(X \log^+ X) = \infty$. Then $\sum EA_n = \infty$, and thus if $\{t = n\} = \{A_n > 0\}$, $E((S_t/t) I(t < \infty)) = \infty$, using (2) and (3). Unfortunately t is not a stopping rule since $P(t < \infty) < 1$. Most of the work in proving Theorem 1 below involves changing t slightly to remedy this defect.

Theorem 1. *Let X be an integrable variable and X_1, X_2, \dots be independent random variables each with the same distribution as X . The following statements are equivalent:*

- (i) $E(X \log^+ X) < \infty$.
- (ii) $E(S_t/t) < \infty$ for every stopping rule t .
- (iii) $E(X_t/t) < \infty$ for every stopping rule t .

Proof of Theorem 1. (i) \Rightarrow (ii), (i) \Rightarrow (iii) are immediate consequences of the fact that if $E(X \log^+ X) < \infty$ then $E \sup(S_n/n) < \infty$ and $E \sup(X_n/n) < \infty$. See [1], p. 891.

(ii) \Rightarrow (i) Assuming $E(X \log^+ X) = \infty$ a stopping rule t will be constructed so that $E(S_t/t) = \infty$.

Let a be a real number such that $0 < P(X > a) < 1$. By the lemma,

$\sum EA_n = \infty$. Pick integers $1 = N_1 < N_2 < \dots$ and sets $\phi = D(N_1), D(N_2), \dots$ to satisfy, if $R_k = \widehat{D}(N_1) \widehat{D}(N_2) \dots \widehat{D}(N_k)$,

(a) $D(N_j)$ is either $\{X_{N_j} > a\}$ or $\{X_{N_j} \leq a\}$, $j = 2, 3, \dots$

(b) $\sum_{k=N_j+1}^{N_{j+1}} E(A_k I(R_j)) \geq 1$.

(c) $\sum_{k=N_j+1}^{\infty} E(A_k I(R_j)) = \infty$.

This is possible, because having chosen $N_1, N_2, \dots, N_u, D(N_1), \dots, D(N_u)$, since (c) holds, N_{u+1} can be picked so that

$$\sum_{k=N_u+1}^{N_{u+1}} E(A_k I(R_u)) \geq 1.$$

Since

$$\sum_{k=N_{u+1}+1}^{\infty} E(A_k I(R_u)) = \infty,$$

either

$$\sum_{k=N_{u+1}+1}^{\infty} E(A_k I(R_u, \{X_{N_{u+1}} > a\})) = \infty$$

or

$$\sum_{k=N_{u+1}+1}^{\infty} E(A_k I(R_u, \{X_{N_{u+1}} \leq a\})) = \infty.$$

Pick $D(N_{u+1}) = \{X_{N_{u+1}} \leq a\}$ if the first holds and $\{X_{N_{u+1}} > a\}$ if the second holds but the first doesn't.

Let $\tau = \inf\{n: A_n > 0\}$, $v = \inf\{n: n \text{ is some } N_i \text{ and } I(D_n) > 0\}$, and let $\eta = \min(\tau, v)$.

$P(\eta < \infty) \geq P(v < \infty) = 1$, since $\sum P(D(N_i)) = \infty$ and the sets $D(N_1), D(N_2), \dots$ are independent. Using (2) and (3),

$$\begin{aligned} E(S_\eta/\eta)^+ &\geq \sum_{n=1}^{\infty} E(A_n I(\eta = n)) \\ &= \sum_{j=1}^{\infty} \sum_{k=N_j+1}^{N_{j+1}} E(A_k I(\eta = k)) \\ &= \sum_{j=1}^{\infty} \sum_{k=N_j+1}^{N_{j+1}} E(A_k I(R_j)) \geq 1 + 1 + \dots = \infty. \end{aligned}$$

Now let t be the first time n such that $n \geq \eta$ and $S_n/n \geq -2B$. Then t is the required stopping rule, since $S_t/t \geq S_\eta/\eta$, so $E(S_t/t)^+ = \infty$, while $S_t/t \geq -2B$ so $E(S_t/t)^- < \infty$.

(iii) \Rightarrow (i) Again suppose that $E(X \log^+ X) = \infty$ and let the notation be as just above. Since $X_n/n > A_n$ on $\{t = n\}$, the same proof shows that $E(X_\eta/\eta)^+ = \infty$. If t^* is the first time k after η such that $X_k > 0$, then $X_{t^*}/t^* \geq X_\eta/\eta$ and $X_{t^*}/t^* \geq 0$ so $E(X_{t^*}/t^*) \geq E(X_\eta/\eta)^+ = \infty$.

Theorem 2. *There exist positive constants F, G, H such that if X is an integrable random variable, X_1, X_2, \dots have the same distribution as X , and T is the class of all stopping rules then both $\sup_{t \in T} E|S_t/t|$ and $\sup_{t \in T} E|X_t/t|$ are bounded below by $FE(|X| \log^+ |X|) - GE|X| \log^+ E|X| - G$ and bounded above by $H + HE(|X| \log^+ |X|)$.*

Proof. $\sup |X_n/n| \leq 2 \sup |S_n/n|$, and since

$$E(\sup |S_n/n|) \leq e/(e-1) + e/(e-1) E(|X| \log^+ |X|),$$

due to Doob (see [1], p. 891), H can be taken as $2e/(e-1)$.

It is no loss of generality to assume $E(X^+ \log^+ X) \geq (E|X| \log^+ |X|)/2$. Thus, if T_N is the first time that $A_k > 0$ or N , whichever comes first, using (2) and (3)

$$E|S_{T_N}/T_N| \geq \sum_1^N EA_i, \quad E|X_{T_N}/T_N| \geq \sum_1^N EA_i.$$

From an application of the lemma it follows that G may be taken to be K and F to be $C/2$.

References

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