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# RANDOM WALKS WITH RANDOM INDICES AND NEGATIVE DRIFT CONDITIONED TO STAY POSITIVE

#### BY

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Abstract. Let  $\{X_k, k \ge 1\}$  be a sequence of independent, identically distributed random variables with  $E[X_1] = \mu < 0$ , and let  $\{N_n, n \ge 0\}$ ,  $N_0 = 0$  a.s., be a sequence of positive integer-valued random variables. Form the random walk  $\{S_{N_n}, n \ge 0\}$  by setting  $S_0 = 0$ ,  $S_{N_n} = X_1 + \ldots + X_{N_n}$ ,  $n \ge 1$ .

The main result in this paper shows (under appropriate conditions on  $\{N_n, n \ge 0\}$  and  $\{X_k, k \ge 1\}$ ) that  $S_{N_n}$  conditioned on  $[S_1 > 0, ..., S_{N_n} > 0]$  converges weakly to a random variable  $S^*$  considered by Iglehart [4].

1. Introduction. We assume that  $\{X_k, k \ge 1\}$  are the coordinate functions defined on the product space

$$(\Omega, \mathscr{A}, P) = \sum_{k=1}^{\infty} (\mathbf{R}, \mathscr{B}, \pi),$$

where  $\mathbf{R} = (-\infty, \infty)$ ,  $\mathscr{B}$  is the  $\sigma$ -field of Borel sets of  $\mathbf{R}$ , and  $\pi$  is the common probability measure of the  $X_k$ 's. If  $\Lambda_n = [S_1 > 0, ..., S_n > 0]$ , then we let  $(\Lambda_n, \Lambda_n \cap \mathscr{A}, P_n)$  be the trace of  $(\Omega, \mathscr{A}, P)$  on  $\Lambda_n$ , where  $\Lambda_n \cap \mathscr{A} = \{\Lambda_n \cap A, A \in \mathscr{A}\}$  and  $P_n[A] = P[A]/P[\Lambda_n]$  for  $A \in \Lambda_n \cap \mathscr{A}$ . The expectation with respect to  $P_n$  is denoted by  $E_n\{\cdot\}$ . Let  $S_n^*$  denote the restriction of  $S_n$  to  $\Lambda_n$ , let

$$r_n = P[S_1 > 0, ..., S_n > 0]$$

and, for  $u \ge 0$ , set

$$f_n(u) = E_n \{ \exp(-uS_n^*) \} = E \{ \exp(-uS_n) | I [S_1 > 0, ..., S_n > 0] \},\$$

where  $I[\cdot]$  denotes the indicator function; in the same way, for  $\{N_n, n \ge 0\}$ , define

$$\hat{r}_n = P[S_1 > 0, \dots, S_{N_n} > 0]$$

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and, for  $u \ge 0$ ,

$$\widehat{f}_n(u) = E\{\exp(-uS_{N_n}) | I[S_1 > 0, \dots, S_{N_n} > 0]\},\$$

where  $\{N_n, n \ge 0\}$ ,  $N_0 = 0$  a.s., is a sequence of positive integer-valued random variables.

We suppose that  $\{X_k, k \ge 1\}$  is a sequence of independent, identically distibuted random variables and that the distribution of  $X_1$  satisfies the following conditions:

- (1)  $-\infty \leq E\{X_1\} = \mu < 0;$
- (2)  $\Theta(s) = E\{\exp(sX_1)\}$  converges for real  $s \in [0, a)$  for some a > 0;
- (3)  $\Theta(s)$  attains its infimum at a point  $\tau$ ,  $0 < \tau < a$ , where  $\Theta(\tau) = \gamma < 1$ and  $\Theta'(\tau) = 0$ ;
- (4) if  $X_1$  is a lattice, then  $P[X_1 = 0] > 0$ .

It has been proved by Bahadur and Rao [1] Theorem 1 that conditions (1)-(4) imply

(5) 
$$P[S_n > 0] \sim (2\pi n)^{-1/2} \gamma^n (\alpha \tau)^{-1}, \quad n \to \infty$$

where  $\alpha = \Theta''(\tau)/\gamma$ ,  $0 < \alpha < \infty$ . In the same way we have

In the same way we have

(6) 
$$E \{ \exp(-uS_n) : S_n > 0 \} \sim (2\pi\alpha)^{-1/2} \gamma^n (\alpha(\tau+u))^{-1}, \quad n \to \infty.$$

Put now

$$M = \left[ (2\pi)^{1/2} \alpha \tau \right]^{-1} \exp \left\{ \sum_{n=1}^{\infty} \left( \gamma^{-n} / n^{3/2} \right) P \left[ S_n > 0 \right] \right\}$$

which is finite by (5). Under assumptions (1)-(4) Iglehart [4] has proved that

(7)  $r_n \sim (\gamma^n/n^{3/2}) M$ 

and, for  $u \ge 0$ 

(8) 
$$\lim_{n \to \infty} f_n(u) = [\tau/(\tau+u)] \exp \{ \sum_{n=1}^{\infty} [\gamma^{-n}/n^{3/2}] [E \{ \exp(-uS_n^+) \} - 1] \} \equiv f(u).$$

2. Results. Proofs of theorems 1-3 are given in Section 3.

THEOREM 1. Suppose that conditions (1)-(4) are satisfied. If  $\{N_n, n \ge 0\}$ ,  $N_0 = 0$  a.s., is a sequence of positive integer-valued random variables independent of  $\{X_k, k \ge 1\}$  and  $\{\alpha_n, n \ge 1\}$  is a sequence of positive real numbers such that for any given  $\varepsilon > 0$ 

(9) 
$$P[|N_n/\alpha_n - \lambda| \ge \varepsilon] = o(E(\gamma^{N_n}/N_n^{3/2}))$$

with  $\alpha_n \to \infty$ ,  $n \to \infty$ , and  $\lambda$  is a random variable such that

(10) 
$$P[\lambda \ge a] = 1$$
 for a constant  $a > 0$ ,

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then for 
$$u \ge 0$$

(11)  $\lim_{n\to\infty}\hat{f}_n(u)=f(u).$ 

Remarks. Note that if  $\lambda$  is a degenerate random variable at a > 0, then (10) is trivially satisfied. Moreover, we note that in general (9) cannot be replaced by the weaker condition  $N_n/\alpha_n \xrightarrow{P} a$ ,  $n \to \infty$  (P. – in probability) which is used in the random central limit theorem. This fact is established by the following

Example 1. Let  $\{X_k, k \ge 1\}$  be a sequence of random variables which satisfies (1)-(4) with  $\gamma = 1/2$  and independent of  $\{N_n, n \ge 1\}$ , where  $N_n$  is as follows:

$$P[N_n = 1] = 1/2^n n^{3/2}, \quad P[N_n = n] = 1 - 1/2^n n^{3/2}.$$

Then for any given  $\varepsilon > 0$ 

$$P[|N_n/n-1| \ge \varepsilon] = 1/2^n n^{3/2} \to 0, \quad n \to \infty,$$

i.e.  $N_n/n \xrightarrow{P} 1, n \to \infty$ .

In this case we have

$$\hat{r}_n = r_1 P[N_n = 1] + r_n P[N_n = n]$$
  
~  $P[X_1 > 0]/2^{n/2} + M(1 - 1/2^n n^{3/2})/2^n n^{3/2}, \quad n \to \infty.$ 

Hence, for  $u \ge 0$  we have

$$\hat{f}_n(u) = \frac{f_1(u)r_1/2^n n^{3/2} + f_n(u) M (1 - 1/2^n n^{3/2})/2^n n^{3/2}}{r_1/2^n n^{3/2} + M (1 - 1/2^n n^{3/2})/2^n n^{3/2}} \rightarrow \frac{f_1(u)r_1 + Mf(u)}{r_1 + M} \neq f(u), \quad n \to \infty.$$

Furthermore, we shall see that, in case where  $\lambda$  is nondegenerated random variable, condition (10) cannot be replaced by  $P[\lambda > 0] = 1$  without changing (9).

Example 2. Let  $(\langle 0, 1 \rangle, \mathscr{B}(\langle 0, 1 \rangle), P)$  be a probability space, where P is the Lebesgue measure and  $\mathscr{B}(\langle 0, 1 \rangle)$  is the  $\sigma$ -field of Borel subsets of  $\langle 0, 1 \rangle$ . Assume that  $\{X_k, k \ge 1\}$  is a sequence of random variables defined on  $(\langle 0, 1 \rangle, \mathscr{B}(\langle 0, 1 \rangle))$  and satisfying (1)-(4). Let  $\{N_n, n \ge 1\}$  be a sequence of random variables independent of  $X_k, k \ge 1$ , defined as follows:

$$N_n(\omega) = \begin{cases} 1, & \omega \in \langle 0, 1/n \rangle, \\ k, & \omega \in \langle (k-1)/n^4, k/n^4 \rangle, k = n^3 + 1, \dots, n^4 \end{cases}$$

We see that, for any given  $\varepsilon > 0$ ,

 $P[|N_n/n^4 - \lambda] \ge \varepsilon] = 0$ 

for *n* sufficiently large, where  $\lambda$  is the random variable, uniformly distributed on  $\langle 0, 1 \rangle$ .

Iglehart [4] has proved that, for  $u \ge 0$ ,

$$r_n f_n(u) \sim (\gamma^n/n^{3/2}) M_1(u),$$

where

$$M_1(u) = [(2\pi)^{1/2} \alpha (\tau + u)]^{-1} \exp \left\{ \sum_{n=1}^{\infty} (\gamma^{-n}/n) E\left\{ \exp(-uS_n): S_n > 0 \right\} \right\} < \infty.$$

In this case we have, for  $u \ge 0$ ,

$$\hat{f}_{n}(u) = \left(f_{1}(u)r_{1}/n + \sum_{k=n^{3}+1}^{n^{4}} f_{k}(u)r_{k}/n^{4}\right)/\hat{r}_{n}$$

$$= \frac{f_{1}(u)r_{1} + \sum_{k=n^{3}+1}^{n^{4}} f_{k}(u)r_{k}/n^{3}}{r_{1} + \sum_{k=n^{3}+1}^{n^{4}} r_{k}/n^{3}} \rightarrow f_{1}(u) \neq f(u), \quad n \to \infty,$$

since

$$\sum_{k=n^3+1}^{n^4} r_k f_k(u)/n^3 \sim (1/n^3) \sum_{k=n^3+1}^{n^4} (\gamma^k/k^{3/2}) M_1(u) \to 0,$$

and

$$\sum_{n=n^3+1}^{n^4} (r_k/n^3) \sim (1/n^3) \sum_{k=n^3+1}^{n^4} (\gamma^k/k^{3/2}) M \to 0, \ n \to \infty.$$

In the case where  $\lambda$  is a nondegenerated random variable which satisfies only  $P[\lambda > 0] = 1$ , we have

THEOREM 2. Suppose that conditions (1)-(4) are satisfied. If  $\{N_n, n \ge 0\}$ ,  $N_0 = 0$  a.s., is a sequence of positive integer-valued random variables independent of  $\{X_k, k \ge 1\}$  and  $\{\alpha_n, n \ge 1\}$  is a sequence of positive real numbers such that  $\lim_{n \to \infty} \alpha_n = \infty$ , and

(12) 
$$P[|N_n/\alpha_n - \lambda| \ge \varepsilon_n] = o(E(\gamma^{N_n}/N_n^{3/2})),$$

(13) 
$$P[\lambda < 2\varepsilon_n] = o(E(\gamma^{N_n}/N_n^{3/2})),$$

then (11) holds, where  $\lambda$  is a positive random variable and  $\{\varepsilon_n, n \ge 1\}$  is a sequence of positive numbers such that  $0 < \varepsilon_n \to 0$ ,  $\alpha_n \varepsilon_n \to \infty$ ,  $n \to \infty$ .

We now establish (11) without the assumption of independence  $\{X_k, k \ge 1\}$  and  $\{N_n, n \ge 0\}$ . First we shall give an example which shows that in this case assumptions of type (9) and (12) are not sufficient for (11).

Example 3. Let  $\{X_k, k \ge 1\}$  be a sequence of independent, identically distributed random variables such that  $X_1$  is uniformly distributed on  $\langle -2, 1 \rangle$ . It can be verified that  $X_1$  satisfies conditions (1)-(4). Assume that

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 $\{N_n, n \ge 1\}$  is a sequence of positive integer-valued random variables such that

$$[N_n = n] = [X_{n+1} \in \langle -2, 0\rangle], \quad [N_n = n+1] = [X_{n+1} \in \langle 0, 1\rangle].$$

Note that for any given  $\varepsilon > 0$ 

$$P[|N_n/n-1| \ge \varepsilon] = 0,$$

whenever  $n > n_0 = \lfloor 1/\varepsilon \rfloor$ .

Moreover, we see that, for  $u \ge 0$ ,

$$\begin{split} \hat{f}_n(u) &= \{ E \exp(-uS_n) I [S_1 > 0, \dots, S_n > 0] I [X_{n+1} \in \langle -2, 0 \rangle] + \\ &+ E \exp(-uS_n) I [S_1 > 0, \dots, S_n > 0] \exp(-uX_{n+1}) \times \\ &\times I [X_{n+1} \in \langle 0, 1 \rangle] \} / \{ P [S_1 > 0, \dots, S_n > 0, X_{n+1} \in \langle -2, 0 \rangle] + \\ &+ P [S_1 > 0, \dots, S_{n+1} > 0, X_{n+1} \in \langle 0, 1 \rangle] \} \\ &= E (\exp(-uS_n) | I [S_1 > 0, \dots, S_n > 0]) (2/3 + (1 - e^{-u})/(3u)) \\ &= f_n(u) (2/3 + (1 - e^{-u})/(3u)) \rightarrow (2/3 + (1 - e^{-u})/(3u)) f(u), \quad n \to \infty, \end{split}$$

which proves that assumptions of type (9) and (10) are not sufficient for (11).

When  $\lambda$  is a degenerated random variable, we can prove in the considerated case the following theorem which is in some sense the strongest:

THEOREM 3. Suppose that conditions (1)-(4) hold and that  $\{N_n, n \ge 0\}$ ,  $N_0 = 0$  a.s., is a sequence of positive integer-valued random variables and  $\{\alpha_n, n \ge 1\}$  is a sequence of positive integer numbers such that  $\lim_{n \to \infty} \alpha_n = \infty$ . If

(14) 
$$P[N_n \neq \alpha_n] = o(\gamma^{\alpha n} / \alpha_n^{3/2}),$$

then (11) holds.

## 3. Proofs of the results.

Proof of Theorem 1. Let  $\varepsilon$ ,  $0 < \varepsilon < a$ , be fixed and put  $a_n = [(a-\varepsilon)\alpha_n]$ . By (9), (10) and the assumption  $\alpha_n \to \infty$ ,  $n \to \infty$ , we can choose *n* sufficiently large such that

$$0 \leq \sum_{k=1}^{a_n} P[S_1 > 0, \dots, S_k > 0] P[N_n = k] \leq \sum_{k=1}^{a_n} P[N_n = k]$$
$$\leq P[|N_n/\alpha_n - \lambda| \geq \varepsilon] = o(E(\gamma^{N_n}/N_n^{3/2})),$$

and at the same time, by (7),

$$\sum_{k=a_{n}+1}^{\infty} P[S_{1} > 0, ..., S_{k} > 0] P[N_{n} = k] \approx \sum_{k=a_{n}+1}^{\infty} (M\gamma^{k}/k^{3/2}) P[N_{n} = k]$$
$$= M \cdot E(\gamma^{N_{n}}/N_{n}^{3/2}) - M \sum_{k=1}^{a_{n}} P[N_{n} = k].$$

But

$$0 \leq M \sum_{k=1}^{a_n} (\gamma^k / k^{3/2}) P[N_n = k] \leq M \sum_{k=1}^{a_n} P[N_n = k] = o(E(\gamma^{N_n} / N_n^{3/2})).$$

Hence

(15) 
$$\hat{r}_n \sim M \cdot E(\gamma^{N_n}/N_n^{3/2}), \quad n \to \infty.$$

Put now

(16) 
$$C_{n,k} = r_k P[N_n = k]/\hat{r}_n, \quad k \ge 1, \ n \ge 1.$$

We see that  $\sum_{k=1}^{\infty} C_{n,k} = 1$  and, for fixed k, by (9) and (15)

$$0 \leq C_{n,k} \leq \frac{\sum\limits_{k=1}^{a_n} P[N_n = k]}{\hat{r}_n} = \frac{o\left(E\left(\gamma^{N_n}/N_n^{3/2}\right)\right)}{M \cdot E\left(\gamma^{N_n}/N_n^{3/2}\right)} \to 0, \quad n \to \infty,$$

which proves that  $[C_{n,k}]_{k=1,\ldots;n=1,\ldots}$  is a Toeplitz matrix. Therefore, by [5], p. 472, for  $u \ge 0$  we have

$$\widehat{f}_n(u) = \sum_{k=1}^{\infty} f_k(u) C_{n,k} \to f(u), \quad n \to \infty,$$

which completes the proof of Theorem 1.

Proof of Theorem 2. By (12) and (13) we have, for sufficiently large n, (17)  $\hat{r}_n = P[S_1 > 0, ..., S_{N_n} > 0]$ 

$$= \sum_{k=1}^{[\alpha_n e_n]} r_k P[N_n = k] + \sum_{k=[\alpha_n e_n]}^{\infty} r_k P[N_n = k] \sim M \cdot E(\gamma^{N_n}/N_n^{3/2}).$$

One can see that

$$C_{n,j} = P[S_1 > 0, ..., S_j > 0] P[N_n = j]/\hat{r}_n, \quad n \ge 1, j \ge 1$$

is a Toeplitz matrix. Indeed, we have  $C_{n,j} \ge 0$ ,  $\sum_{j=1}^{\infty} C_{n,j} = 1$ , and, by (18), (19) and (17), we get

$$0 \leq C_{n,j} \leq \frac{\sum_{k=n}^{\left[\alpha_{n}\varepsilon_{n}\right]} P\left[N_{n}=k\right]}{M \cdot E\left(\gamma^{N_{n}}/N_{n}^{3/2}\right)} \sim \frac{P\left[|N_{n}/\alpha_{n}-\lambda| \geq \varepsilon_{n}\right] + P\left[\lambda \leq 2\varepsilon_{n}\right]}{M \cdot E\left(\gamma^{N_{n}}/N_{n}^{3/2}\right)} \to 0,$$

 $n \to \infty$ , as  $j \le [\varepsilon_n \alpha_n]$  for sufficiently large *n*. Following the considerations of the proof of Theorem 1 we obtain (11).

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Proof of Theorem 3. From (14) we have

$$\hat{r}_n = P[S_1 > 0, \dots, S_{N_n} > 0] = \sum_{k=1}^{\infty} P[S_1 > 0, \dots, S_k > 0, N_n = k]$$
  
=  $P[S_1 > 0, \dots, S_{\alpha_n} > 0, N_n = \alpha_n] + P[S_1 > 0, \dots, S_{N_n} > 0, N_n \neq \alpha_n]$ 

Hence, by (7), we get

(18) 
$$M(\gamma^{\alpha_n}/\alpha_n^{3/2}) - o(\gamma^{\alpha_n}/\alpha_n^{3/2}) \leq \hat{r}_n \leq M(\gamma^{\alpha_n}/\alpha_n^{3/2}) + o(\gamma^{\alpha_n}/\alpha_n^{3/2}).$$

Taking into account that, for  $u \ge 0$ ,

$$\hat{r}_{n} \hat{f}_{n}(u) = E \{ \exp(-uS_{\alpha_{n}}) I [S_{1} > 0, ..., S_{\alpha_{n}} > 0] I [N_{n} = \alpha_{n}] \} + E \{ \exp(-uS_{N_{n}}) I [S_{1} > 0, ..., S_{N_{n}} > 0] I [N_{n} \neq \alpha_{n}] \}$$

$$= E \{ \exp(-uS_{\alpha_{n}}) I [S_{1} > 0, ..., S_{\alpha_{n}} > 0] \} - E \{ \exp(-uS_{\alpha_{n}}) I [S_{1} > 0, ..., S_{\alpha_{n}} > 0] I [N_{n} \neq \alpha_{n}] \} + E \{ \exp(-uS_{\alpha_{n}}) I [S_{1} > 0, ..., S_{N_{n}} > 0] I [N_{n} \neq \alpha_{n}] \} \}$$

and (14), we have

$$r_{\alpha_n}f_{\alpha_n}(u)-o\left(\gamma^{\alpha_n}/\alpha_n^{3/2}\right)\leqslant \hat{r}_n\,\hat{f}_n(u)\leqslant r_{\alpha_n}f_{\alpha_n}(u)+o\left(\gamma^{\alpha_n}/\alpha_n^{3/2}\right).$$

Therefore, by (18), we get (11).

Note. The problem here considered, in the case where  $E\{X_1\} = \mu = 0$ , was treated in [7].

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