

## THE RUIN PROBLEM FOR FINITE MARKOV CHAINS

BY THOMAS HÖGLUND

*Royal Institute of Technology, Stockholm*

We derive an asymptotic approximation of the joint distribution  $\text{prob}(N(u) - n \in A, S_{N(u)} - u \in B)$  as  $n$  and  $u \rightarrow \infty$ . Here  $N(u) = \min\{n; S_n > u\}$  denotes the first passage time for a random walk of the form  $S_n = \sum_{k=1}^n U_k(\xi_{k-1}, \xi_k)$ , where  $\xi_0, \xi_1, \dots$  is a finite Markov chain and where  $\{U_k(i, j)\}_{k=1}^\infty$  is a sequence of independent random variables. The approximation holds for all sets  $B$  and a fairly large class of sets  $A$ .

**1. Introduction and results.** Consider a Markovian random walk of the form

$$(1.1) \quad S_n = \sum_{k=1}^n U_k(\xi_{k-1}, \xi_k), \quad n > 0,$$

and the first passage time

$$(1.2) \quad N(u) = \min\{n > 0; S_n > u\}.$$

Here  $\{\xi_k\}_{k=0}^\infty$  is an irreducible and stationary Markov chain with finite state space  $\{1, 2, \dots, r\}$ , and  $U_k(i, j)$  are integer-valued random variables such that the matrices  $\{U_k(i, j)\}_{i, j=1}^r$ ,  $k = 1, 2, \dots$ , have a common distribution and are independent of each other and of the Markov chain.

Let  $A \subset \mathbb{Z}$  (the integers) and  $B \subset \mathbb{Z}_+$  (the positive integers) and put

$$(1.3) \quad Z_{ij}(n, u) = \text{prob}(N(u) \in n + A, S_{N(u)} \in u + B, \xi_{N(u)} = j | \xi_0 = i).$$

We shall in this paper determine the asymptotic behaviour of  $Z_{ij}(n, u)$  as  $n$  and  $u$  tend to infinity. The approximation is valid for all  $B \subset \mathbb{Z}_+$  and for a fairly large class of sets  $A$ .

If we let  $r = 1$ , we get the corresponding result for ordinary random walks (on  $\mathbb{Z}$ ). The other extreme is when  $U_1(i, j)$  is a deterministic function of  $(i, j)$ .

The more general problem of estimating the probability that a two-dimensional random walk  $(T_k, S_k)_{k=1}^\infty$  hits the set  $(n, u) + A \times B$  the first time  $S_k > u$  can be treated by the same method [see Höglund (1990a) for the corresponding problem for independent random variables].

Miller (1962a, b) considered the case  $A = \mathbb{Z}$ ,  $B = \mathbb{Z}_+$  and found the asymptotic behaviour of  $\text{prob}(N(u) < \infty)$  as  $u \rightarrow \infty$  using a Wiener–Hopf factorization. Presman (1969) used a similar technique to study the asymptotic behaviour of  $\text{prob}(N(u) \leq n)$  ( $A = \{\dots, -2, -1, 0\}$ ,  $B = \mathbb{Z}_+$ ). This technique was further developed by Arndt (1981, 1982). He obtained upper bounds for

---

Received September 1989; revised February 1990.

AMS 1980 subject classifications. Primary 60J10; secondary 60J15.

Key words and phrases. Boundary crossing, large deviations, Markov chains.

$\text{prob}(N(u) < \infty)$ ,  $\text{prob}(N(u) \leq n)$  and their difference and he gave conditions under which explicit calculations are possible.

Let  $(p_{ij})$  stand for the transition probability matrix and define the (matrix-valued) probability density of  $U_k$ ,  $P(u) = (P_{ij}(u))$ , by

$$(1.4) \quad \begin{aligned} P_{ij}(u) &= \text{prob}(U_k(\xi_{k-1}, \xi_k) = u, \xi_k = j | \xi_{k-1} = i) \\ &= \text{prob}(U_k(i, j) = u) p_{ij}. \end{aligned}$$

The approximation will be expressed in terms of quantities related to the matrices  $\hat{P}(\theta) = (\hat{P}_{ij}(\theta))$ ,  $\theta \in \Theta$ , where

$$(1.5) \quad \hat{P}_{ij}(\theta) = \sum_u e^{\theta u} P_{ij}(u) = p_{ij} E e^{\theta U_1(i, j)}$$

and where  $\Theta$  denotes the interior of the set of  $\theta \in \mathbf{R}$  for which this matrix is finite. The set  $\Theta$  is then an open interval which we shall assume is nonempty.

The matrix  $\hat{P}(\theta)$  is thus a positive and irreducible matrix whose coefficients are analytic in  $\theta$ ; and hence  $\hat{P}(\theta)$  has a maximal positive eigenvalue  $\lambda(\theta)$  corresponding to strictly positive left and right eigenvectors  $\sigma(\theta) = \{\sigma_i(\theta)\}$  and  $\rho(\theta) = \{\rho_i(\theta)\}$ . This eigenvalue is simple and analytic in  $\Theta$  and  $\sigma_i(\theta)$  and  $\rho_i(\theta)$  can be chosen to be analytic in  $\Theta$ . We shall use the normalization  $\sigma(\theta) \cdot \rho(\theta) = 1$ .

A subscript  $\theta$  on probabilities and expectations indicates that the underlying probability measure is given by the initial distribution

$$(1.6) \quad \text{prob}_\theta(\xi_0 = i) = \sigma_i(\theta) \rho_i(\theta),$$

and the cylinder set probabilities

$$(1.7) \quad \begin{aligned} \text{prob}_\theta(U_k(\xi_{k-1}, \xi_k) = u_k, \xi_k = i_k, k = 1, \dots, n | \xi_0 = i_0) \\ = \prod_{k=1}^n \frac{e^{\theta u_k} P_{i_{k-1}i_k}(u_k) \rho_{i_k}(\theta)}{\lambda(\theta) \rho_{i_{k-1}}(\theta)}. \end{aligned}$$

We shall write  $m(\theta)$  for the expectation of  $U_1(\xi_0, \xi_1)$  with respect to this measure:

$$(1.8) \quad m(\theta) = E_\theta U_1(\xi_0, \xi_1) = \sigma(\theta) \cdot \hat{P}'(\theta) \rho(\theta) / \lambda(\theta).$$

LEMMA 1.1. *The function  $m$  satisfies  $m(\theta) = \lambda'(\theta) / \lambda(\theta)$  and  $m'(\theta) > 0$  for all  $\theta$  unless  $\text{prob}(U_1(i, j) = a + v(j) - v(i)) = 1$  for some integer  $a$  and some sequence  $v(1), \dots, v(r)$ .*

PROOF. Differentiating the identity

$$(1.9) \quad \hat{P}(\theta) \rho(\theta) = \lambda(\theta) \rho(\theta),$$

we get

$$(1.10) \quad \hat{P}'(\theta) \rho(\theta) + \hat{P}(\theta) \rho'(\theta) = \lambda'(\theta) \rho(\theta) + \lambda(\theta) \rho'(\theta).$$

Multiply (scalar product) by  $\sigma(\theta)$  and observe that

$$(1.11) \quad \sigma(\theta) \cdot \hat{P}(\theta)\rho'(\theta) = \lambda(\theta)\sigma(\theta) \cdot \rho'(\theta).$$

The equality of the lemma follows. The inequality is Theorem 1.2 of Keilson and Wishart (1964).  $\square$

The approximation will depend on the support of the distribution,  $L(n, u) = (L_{ij}(n, u))$ , of the first strict ascending ladder point  $(N, S_N)$ :

$$(1.12) \quad L_{ij}(n, s) = \text{prob}(N = n, S_N = s, \xi_N = j | \xi_0 = i).$$

Here and below  $N = N(0)$ . This distribution defines a certain group  $G(L) \subset \mathbb{Z}^2$ . A complication here is that  $L$  need not be irreducible even though  $P$  is. We shall avoid this complication simply by assuming irreducibility. Define with  $t = (n, u)$ ,

$$(1.13) \quad L^{k*}(t) = \sum_{t_1 + \dots + t_k = t} L(t_1) \cdots L(t_k),$$

that is,  $L^{k*}$  is the distribution of the  $k$ th strict ascending ladder point. Put

$$(1.14) \quad S_{ij}(L) = \bigcup_{k=1}^{\infty} \{(n, s); L_{ij}^{k*}(n, s) > 0\}$$

and let  $G_{ij}(L)$  denote the smallest subgroup of  $\mathbb{Z}^2$  that contains all the differences  $t_1 - t_2$ ,  $t_1 \in S_{ij}(L)$ ,  $t_2 \in S_{ij}(L)$ . It turns out that if  $L$  is irreducible [i.e., the matrix  $\sum_t L(t)$  is irreducible], then the groups  $G_{ij}(L) = G(L)$  do not depend on  $i, j$  and  $S_{ij}(L) \subset v(i) - v(j) + G(L)$  for some sequence  $v(1), \dots, v(r)$  in  $\mathbb{Z}^2$  [see Lemma 1.1 in Höglund (1990b)].

Let

$$(1.15) \quad \omega_j(b, \theta) = e^{-\theta b} \frac{\text{prob}_{\theta}^*(N < \infty, S_N \geq b, \xi_N = j)}{E_{\theta}^* S_N} \rho_j(\theta)^{-1},$$

where the star indicates that the underlying probability distribution is given by the transition probabilities (1.7), but the initial distribution (1.6) which is stationary for  $\text{prob}_{\theta}(\xi_1 = j | \xi_0 = i)$  is replaced by the initial distribution which is stationary for  $\text{prob}_{\theta}(N < \infty, \xi_N = j | \xi_0 = i)$ :

$$(1.16) \quad \text{prob}_{\theta}^*(\xi_0 = i) = \sigma_i^*(\theta)\rho_i(\theta).$$

Here  $\sigma^*(\theta)$  is the unique left positive eigenvector of the matrix  $\sum_n \sum_u e^{\theta u} L(n, u) / \lambda(\theta)^n$ , that satisfies  $\sigma^*(\theta) \cdot \rho(\theta) = 1$ .

We shall determine the asymptotic behaviour of  $Z(n, u)$  when  $(n, u)$  tends to infinity in the cone

$$(1.17) \quad \{(\tau, \tau m(\theta)); \tau > 0, \theta \in \Theta, m(\theta) > 0\}.$$

It follows from Lemma 1.1 that the equation  $m(\theta) = u/n$  has a unique solution  $\theta = \hat{\theta}(u/n)$  when  $(n, u)$  belongs to this cone. We shall also need the

set

$$(1.18) \quad \hat{A} = \left\{ \theta \in \Theta; \sum_{a \in A} \lambda(\theta)^a < \infty \right\}$$

$$= \begin{cases} \{ \theta \in \Theta; \lambda(\theta) > 1 \} & \text{if } \inf A = -\infty, \sup A < \infty, \\ \Theta & \text{if } \inf A > -\infty, \sup A < \infty, \\ \{ \theta \in \Theta; \lambda(\theta) < 1 \} & \text{if } \inf A > -\infty, \sup A = \infty. \end{cases}$$

Note that  $\hat{A}^c$  does not stand for the complement of  $\hat{A}$  but

$$\hat{A}^c = \left\{ \theta \in \Theta; \sum_{a \in A^c} \lambda(\theta)^a < \infty \right\}.$$

**THEOREM 1.2.** *Suppose that  $L$  is irreducible and that  $G(L) = \mathbb{Z}^2$ . If  $m(\theta) > 0$ , then  $\text{prob}_\theta(N < \infty) = 1$  and the expressions (1.15) are well-defined.*

*Let  $n$  and  $u \rightarrow \infty$  in the cone (1.17) in such a way that  $\hat{\theta} = \hat{\theta}(u/n)$  stays within but away from the boundary of the set  $\{ \theta \in \Theta; m(\theta) > 0, \theta \in \hat{A} \cup \hat{A}^c \}$ .*

(i) *If  $\hat{\theta}(u/n) \in \hat{A}$ , then*

$$(1.19) \quad Z_{ij}(n, u) = \lambda(\hat{\theta})^n e^{-\hat{\theta}u} (2\pi n Q(\hat{\theta}))^{-1/2} \rho_i(\hat{\theta})$$

$$\times \left( \sum_{a \in A} \lambda(\hat{\theta})^a \sum_{b \in B} \omega_j(b, \hat{\theta}) + o(1) \right),$$

*uniformly in  $\hat{\theta}(u/n)$ . Here  $0 < Q(\theta) < \infty$  is as in (3.18).*

(ii) *If  $\hat{\theta}(u/n) \in \hat{A}^c$  and if there is a  $\kappa \in \Theta$  such that  $\lambda(\kappa) = 1$  and  $m(\kappa) > 0$ , then*

$$(1.20) \quad Z_{ij}(n, u) = e^{-\kappa u} \rho_i(\kappa) \left( \sum_{b \in B} \omega_j(b, \kappa) + o(1) \right),$$

*uniformly in  $\hat{\theta}(u/n)$ .*

The theorem will be proved in the following sections.

Write  $I$  for the interval  $I = \{ m(\theta); \theta \in \Theta, m(\theta) > 0 \}$ . The function  $m(\theta)$  is strictly increasing and hence the approximation (i) holds for  $u/n \in m(\hat{A}) \cap I$  and the approximation (ii) for  $u/n \in m(\hat{A}^c) \cap I$ . Furthermore,  $\log \lambda(\theta)$  is convex and  $\lambda(0) = 1$ . So if  $m(0) > 0$ , then  $\kappa = 0$ , and if  $m(0) < 0$ , then there is a possibility that  $\lambda(\kappa) = 1$  for some  $\kappa > 0$ , and then  $m(\kappa)$  is necessarily positive.

Thus, for example, if we let  $A = \{ \dots, -2, -1, 0 \}$ ,  $B = \mathbb{Z}_+$ , we get an approximation for

$$\text{prob} \left( \max_{k \leq n} S_k > u \right).$$

The approximation (i) holds when  $m(\kappa) < u/n \in I$  and (ii) holds when  $m(\kappa) > u/n \in I$ . Here we assumed the existence of  $\kappa$ .

The condition  $G(L) = \mathbb{Z}^2$  is just a normalization; see the comment around formula (1.14) in Höglund (1990b).

Theorem 1.2(ii) is applicable to the case  $A = \mathbb{Z}$ .

**COROLLARY 1.3.** *Suppose that  $L$  is irreducible and that  $G(L) = \mathbb{Z}^2$ . If  $0 \in \Theta$  and if there is a  $\kappa \in \Theta$  such that  $\lambda(\kappa) = 1$  and  $m(\kappa) > 0$ , then  $\text{prob}_\kappa(N < \infty) = 1$ ,  $\omega_j(b, \kappa)$  are well defined and*

$$(1.21) \quad \begin{aligned} & \text{prob}(N(u) < \infty, S_{N(u)} \in u + B, \xi_{N(u)} = j | \xi_0 = i) \\ & = e^{-\kappa u} \rho_i(\kappa) \left( \sum_{b \in B} \omega_j(b, \kappa) + o(1) \right) \end{aligned}$$

as  $u \rightarrow \infty$ .

Thus in particular,

$$(1.22) \quad \text{prob}(S_{N(u)} = u + s | \xi_0 = i) \rightarrow \text{prob}_0^*(S_N \geq s) / E_0^* S_N$$

for  $s = 1, 2, \dots$  when  $m(0) > 0$ . Here we used the fact that we can take  $\rho_i(0) = 1$  for all  $i$ .

It follows from Proposition 2.3 that the corollary holds as it stands under the slightly weaker assumption  $\{u \in \mathbb{Z}; (n, u) \in G(L)\} = \mathbb{Z}$ .

In Section 2 we shall express the approximations in quantities that are only implicitly determined by  $P$  via  $L$ . In Section 3 we relate some of these quantities to the corresponding ones for  $P$ , which I think is preferable.

The above approximations are of the form  $e^{-nH} n^{-1/2} D$  or  $e^{-nH} D$ . The exponent  $H$  is directly given by  $P$  and  $u/n$ , but  $D$  only implicitly. I think that it is possible to express even this directly via an analogue, Spitzer's formula when  $B = \mathbb{Z}_+$ , but I have not done this. See formula (2.13) in Höglund (1990a) for the corresponding expressions for ordinary random walks.

**2. The implicit solution.** The idea of the proof is to show that  $Z$  is of the form (2.9) for some suitably chosen  $F$  and then simply apply a renewal theorem to the right-hand side of this identity to get an approximation for  $Z$ .

Let  $Z_{ij}^s(n, u)$ ,  $u \geq 0$ , denote the probability (1.3) in the special case when  $A = \{0\}$ ,  $B = \{s\}$  and define  $Z_{ij}^s(n, u) = 0$  for  $u < 0$ .

The event  $N(u) = n$ ,  $S_{N(u)} = u + s$ ,  $\xi_{N(u)} = j$  occurs if and only if either  $N(0) = n$ ,  $S_{N(0)} = u + s$ ,  $\xi_{N(0)} = j$  or else there is an  $m \leq n$ ,  $v \leq u$  and a  $k \in \{1, \dots, r\}$  such that  $N(0) = m$ ,  $S_{N(0)} = v$ ,  $\xi_{N(0)} = k$  and then  $N(u) = n$ ,  $S_{N(u)} = u + s$ ,  $\xi_{N(u)} = j$ .

Therefore

$$(2.1) \quad Z^s(n, u) = F^s(n, u) + L * Z^s(n, u).$$

Here

$$(2.2) \quad F_{ij}^s(n, u) = \begin{cases} \text{prob}(N = n, S_N = u + s, \xi_N = j | \xi_0 = i) & \text{for } u \geq 0, \\ 0 & \text{for } u < 0, \end{cases}$$

and the star denotes convolution

$$(2.3) \quad L * Z^s(n, u) = \sum_{m,v} L(m, v) Z^s(n - m, u - v).$$

The equation (2.1) implies

$$(2.4) \quad Z^s = \sum_{k=0}^{h-1} L^{k*} * F^s + L^{h*} * Z^s$$

for  $h = 1, 2, \dots$ . Observe that  $\text{supp}(L) \subset [1, \infty) \times [1, \infty)$  and hence that  $\text{supp}(L^{h*}) \subset [h, \infty) \times [h, \infty)$ ; use the fact that  $Z^s(n, u) = 0$  when  $n \leq 0$  or  $u < 0$  and conclude that  $L^{h*} * Z^s(n, u) = 0$  for all  $h \geq \min(n, u + 1)$ .

Therefore

$$(2.5) \quad Z^s = R * F^s,$$

where

$$(2.6) \quad R = \sum_{n=0}^{\infty} L^{n*}.$$

Let  $A, B$  and  $Z$  be as in (1.3) and note that

$$(2.7) \quad \begin{aligned} \sum_{k-n \in A} \sum_{s \in B} Z^s(k, u) &= Z(n, u), \\ \sum_{k-n \in A} \sum_{s \in B} F^s(k, u) &= \text{prob}(N - n \in A, S_N - u \in B, \xi_N = j | \xi_0 = i) \end{aligned}$$

for  $u \geq 0$ . The identity (2.9) therefore follows from (2.5) with

$$(2.8) \quad F_{ij}(n, u) = \begin{cases} \text{prob}(N - n \in A, S_N - u \in B, \xi_N = j | \xi_0 = i) & \text{for } u \geq 0, \\ 0 & \text{for } u < 0. \end{cases}$$

LEMMA 2.1. *The probabilities (1.3) satisfy*

$$(2.9) \quad Z = R * F$$

for all  $A \subset \mathbb{Z}, B \subset \mathbb{Z}_+$ .

First some notation: Let

$$(2.10) \quad \hat{L}_{ij}(\eta, \theta) = \sum_n \sum_u e^{\eta n + \theta u} L_{ij}(n, u) = E(e^{\eta N + \theta S_N} \delta_j(\xi_N) | \xi_0 = i),$$

where  $\delta_j(\xi) = 1$  if  $\xi = j$  and  $\delta_j(\xi) = 0$  if  $\xi \neq j$ . Let furthermore  $\Theta^L$  denote the interior of the set of  $\zeta = (\eta, \theta) \in \mathbb{R}^2$  for which  $\hat{L}(\zeta) < \infty$ . (We do not exclude the possibility  $\Theta^L = \emptyset$  in this section.) Assume that  $\hat{L}(\zeta)$  is irreducible and write  $\lambda(\zeta)$  for its maximal positive eigenvalue and  $\sigma(\zeta)$  and  $\rho(\zeta)$  for the corresponding left and, respectively, right positive eigenvectors, normalized in such a way that  $\sigma(\zeta) \cdot \rho(\zeta) = 1$ . The symbols  $\lambda, \rho$  and  $\sigma$  thus have another meaning in this section than that in the introduction.

Let

$$(2.11) \quad \lambda'(\zeta) = (\partial_1 \lambda(\zeta), \partial_2 \lambda(\zeta)) \quad \text{and} \quad \lambda''(\zeta) = \begin{pmatrix} \frac{\partial^2 \lambda(\zeta)}{\partial \eta^2} & \frac{\partial^2 \lambda(\zeta)}{\partial \eta \partial \theta} \\ \frac{\partial^2 \lambda(\zeta)}{\partial \theta \partial \eta} & \frac{\partial^2 \lambda(\zeta)}{\partial \theta^2} \end{pmatrix}$$

denote the gradient and the second order derivative matrix.

We shall determine the asymptotic behaviour of  $Z(n, u)$  when  $(n, u)$  tends to infinity in the cone

$$(2.12) \quad \{\tau \lambda(\zeta); \tau > 0, \zeta \in \Theta^L, \lambda(\zeta) = 1\}.$$

The probability measure corresponding to that given by (1.6) and (1.7) is

$$(2.13) \quad \begin{aligned} \text{prob}_\zeta(\xi_0 = i) &= \sigma_i(\zeta) \rho_i(\zeta), \\ \text{prob}_\zeta(N = n, S_N = u, \xi_N = j | \xi_0 = i) &= \frac{e^{\eta n + \theta u} L_{ij}(n, u) \rho_j(\zeta)}{\lambda(\zeta) \rho_i(\zeta)}. \end{aligned}$$

It is seen in the same way as in the proof of Lemma 1.1 that  $\lambda'(\zeta) = (E_\zeta N, E_\zeta S_N)$ , where  $E_\zeta$  stands for expectation with respect to the previous probability measure. The probabilistic interpretation of  $\lambda'(\zeta)$  is, however, more complicated.

The set that corresponds to  $\hat{A}$  is

$$(2.14) \quad \check{A} = \left\{ \eta; \sum_{a \in A} e^{-\eta a} < \infty \right\}$$

and instead of  $\hat{\theta}(u/n)$  we shall consider  $\check{\zeta}(n, u)$  which is the unique  $\zeta = \check{\zeta}(n, u)$  for which  $\lambda(\zeta) = 1$  and  $\lambda'(\zeta)$  and  $(n, u)$  have the same direction. See Lemma 1.4 in Höglund (1990b).

PROPOSITION 2.2. *Assume that  $L$  is irreducible and that  $G(L) = \mathbb{Z}^2$ . Put*

$$(2.15) \quad c_{ij}(b, \zeta) = e^{-\theta b} \rho_i(\zeta) \text{prob}_\zeta(N < \infty, S_N \geq b, \xi_N = j) / \rho_j(\zeta).$$

Let  $u$  and  $n \rightarrow \infty$  in the cone (2.12) in such a way that  $\check{\zeta} = \check{\zeta}(n, u) = (\check{\eta}(n, u), \check{\theta}(n, u))$  stays within but away from the boundary of the set of  $\zeta = (\eta, \theta)$  determined by

$$(2.16) \quad \zeta \in \Theta^L, \quad (\eta, 0) \in \Theta^L, \quad \lambda(\zeta) \neq 0, \quad \eta \in \check{A} \cup \check{A}^c.$$

(i) *If  $\check{\eta}(n, u) \in \check{A}$ , then*

$$(2.17) \quad \begin{aligned} Z_{ij}(n, u) &= e^{-\check{\eta}n - \check{\theta}u} (2\pi T(\check{\zeta})C(\check{\zeta}))^{-1/2} \\ &\times \left( \sum_{a \in A} e^{-\check{\eta}a} \sum_{b \in B} c_{ij}(b, \check{\zeta}) + o(1) \right) \end{aligned}$$

uniformly in  $\check{\zeta}(n, u)$ . Here  $T(\check{\zeta})$  is determined by  $(n, u) = T\lambda(\check{\zeta})$  and  $C(\zeta) = \lambda(\zeta) \cdot \lambda''(\zeta)^{-1}\lambda'(\zeta)$ . The existence of  $\lambda''(\zeta)^{-1}$  is part of the conclusion.

(ii) If  $\check{\eta}(n, u) \in \check{A}^c$  and if there is a  $\kappa \geq 0$  such that  $\lambda(0, \kappa) = 1$  and  $\partial_2\lambda_L(0, \kappa) > 0$ , then

$$(2.18) \quad Z_{ij}(n, u) = e^{-\kappa u} \partial_2\lambda(0, \kappa)^{-1} \left( \sum_{b \in B} c_{ij}(b, (0, \kappa)) + o(1) \right)$$

uniformly in  $\check{\zeta}(n, u)$ .

The case  $A = \mathbb{Z}$  will be needed in the proof. In this case the functions  $Z$  and  $F$  will depend only on the variable  $u$  and we can replace  $L$  in the identities (2.6) and (2.9) by  $L_2(u) = \sum_n L(n, u)$ , the marginal distribution for  $S_N$ .

**PROPOSITION 2.3.** Assume that  $L$  is irreducible and that  $G(L_2) = \mathbb{Z}$ . If  $(0, 0) \in \Theta^L$  and if there is a  $(0, \kappa) \in \Theta^L$  such that  $\lambda(0, \kappa) = 1$  and  $\partial_2\lambda(0, \kappa) > 0$ , then

$$(2.19) \quad \begin{aligned} &\text{prob}(N(u) < \infty, S_{N(u)} \in u + B, \xi_{N(u)} = j | \xi_0 = i) \\ &= e^{-\kappa u} \partial_2\lambda(0, \kappa)^{-1} \left( \sum_{b \in B} c_{ij}(b, (0, \kappa)) + o(1) \right) \end{aligned}$$

as  $u \rightarrow \infty$ .

**PROOF OF PROPOSITION 2.3.** It follows from Lemma 2.1 and the one-dimensional version of Theorem 1.5 in Höglund (1990b) that the expression to the left in (2.19) equals

$$(2.20) \quad e^{-\kappa u} (\partial_2\lambda(0, \kappa))^{-1} \left( \sum_{\nu} \rho_i(0, \kappa) \sigma_{\nu}(0, \kappa) \hat{F}_{\nu,j}(\kappa) + o(1) \right),$$

provided the sums

$$(2.21) \quad \hat{F}_{ij}(\theta) = \sum_u e^{\theta u} F_{ij}(u) = E(e^{\theta S_N} \delta_j(\xi_N) | \xi_0 = i)$$

are convergent for  $\theta$  in a neighbourhood of  $\kappa$ . Here  $F_{ij}(u) = F_{ij}(n, u)$  as defined in (2.8) but with  $A = \mathbb{Z}$ .

We shall use a superscript to indicate the dependence on the set  $B$ . A change of the order of summation yields

$$(2.22) \quad \begin{aligned} \hat{F}_{ij}^B(\theta) &= \sum_{b \in B} e^{-\theta b} \sum_{v \geq b} e^{\theta v} \sum_m L_{ij}(m, v) \\ &\leq \hat{F}_{ij}^{\mathbb{Z}^+}(\theta) = \begin{cases} \sum_m \sum_v (e^{\theta v} - 1) L_{ij}(m, v) / (e^{\theta} - 1) & \text{if } \theta \neq 0, \\ \sum_m \sum_v v L_{ij}(m, v) & \text{if } \theta = 0, \end{cases} \end{aligned}$$

and these sums are convergent for all  $(0, \theta)$  in  $\Theta^L$  since this set is open and since  $(0, 0) \in \Theta^L$ .



That the sum in (2.20) equals the sum in (2.19) follows from the second identity in (2.13) with  $\zeta = (0, \kappa)$ .  $\square$

PROOF OF PROPOSITION 2.2. The proof of part (i) is similar to the proof of Proposition 2.3. We use the two-dimensional version of the above mentioned theorem and have to verify that the sums

$$(2.23) \quad \hat{F}_{ij}^B(\zeta) = \sum_n \sum_u e^{\eta n + \theta u} F_{ij}(n, u) = E(e^{\eta N + \theta S_N} \delta_j(\xi_N) | \xi_0 = i)$$

are convergent for  $\zeta$  in the set determined by (2.16).

In this case,

$$(2.24) \quad \begin{aligned} \hat{F}_{ij}^B(\zeta) &= \sum_{a \in A} e^{-\eta a} \sum_{b \in B} e^{-\theta b} \sum_m \sum_{v \geq b} e^{\eta m + \theta v} L_{ij}(m, v) \\ &\leq \hat{F}_{ij}^{Z^+}(\zeta) \\ &= \sum_{a \in A} e^{-\eta a} \times \begin{cases} \sum_m \sum_v e^{\eta m} (e^{\theta v} - 1) L_{ij}(m, v) / (e^\theta - 1) & \text{if } \theta \neq 0, \\ \sum_m \sum_v e^{\eta m} v L_{ij}(m, v) & \text{if } \theta = 0, \end{cases} \end{aligned}$$

and these sums are convergent for all  $\zeta$  in the set (2.16) since this set is open.

The remainder of part (i) follows from the second identity in (2.13).

In order to prove part (ii), note that

$$(2.25) \quad Z^A(n, u) + Z^{A^c}(n, u) = Z^Z(n, u).$$

Here we used a superscript to indicate the dependence on  $A$ . We can apply part (i) to  $Z^{A^c}(n, u)$  and Proposition 2.3 to  $Z^Z(n, u)$ . Furthermore,

$$(2.26) \quad \check{\eta}(n, u) n + \check{\theta}(n, u) u = \max_{\zeta} \{ \eta n + \theta u; \lambda(\zeta) = 1 \} \geq \kappa u$$

and hence  $Z^A(n, u) = Z^Z(n, u)(1 + O(n^{-1/2}))$ .

**3. Proof of the explicit version.** We shall need an auxiliary function  $G(n, u) = (G_{ij}(n, u))$ , where for  $n \geq 0$ ,

$$(3.1) \quad G_{ij}(n, u) = \text{prob}(N > n, S_n = u, \xi_n = j | \xi_0 = i).$$

Here  $S_0 = 0$  and hence  $G(0, u) = \delta(0, u)$ , where  $\delta(n, u) \neq 0$  only when  $(n, u) = (0, 0)$  and equals  $I$ , the identity matrix, in that case. Let us agree that both  $G(n, u)$  and  $L(n, u)$  equal 0 when  $n < 0$ .

By an obvious modification of the argument [page 599 in Feller (1971)]:

$$(3.2) \quad \begin{aligned} &\{N = n, S_n = u, \xi_n = j\} \\ &= \bigcup_{k=1}^r \bigcup_{t \leq 0} (\{N > n - 1, S_{n-1} = t, \xi_{n-1} = k\} \\ &\quad \cap \{U(k, \xi_n) = u - t, \xi_n = j\}) \end{aligned}$$

for  $u > 0$  and hence

$$(3.3) \quad L(n, u) = \sum_t G(n - 1, t)P(u - t), \quad u > 0.$$

In a similar way we obtain

$$(3.4) \quad G(n, u) = \delta(n, u) + \sum_t G(n - 1, t)P(u - t), \quad u \leq 0,$$

and hence

$$(3.5) \quad L(n, u) + G(n, u) = \delta(n, u) + \sum_t G(n - 1, t)P(u - t)$$

for all  $u$ .

Define  $\hat{L}(\eta, \theta)$  as in (2.10) and

$$(3.6) \quad \hat{G}(\eta, \theta) = \sum_{n=0}^{\infty} e^{\eta n} \sum_u e^{\theta u} G(n, u).$$

LEMMA 3.1. *If  $\theta \in \Theta$ ,  $m(\theta) > 0$  and  $e^{\eta\lambda(\theta)} = 1$ , then  $\hat{G}$  converges in a neighbourhood of the point  $(\eta, \theta)$ .*

*Let  $l = \inf_{\theta \in \Theta} \lambda(\theta)$ , then  $(-\infty, \log(1/l)) \times \Theta \subset \Theta^L$ .*

PROOF. Let  $\theta' < \theta$ , then

$$(3.7) \quad \begin{aligned} \hat{G}(\eta, \theta) &\leq \sum_n \sum_{u \leq 0} e^{\eta n} e^{\theta u} P^{n*}(u) \leq \sum_n \sum_u e^{\eta n} e^{\theta' u} P^{n*}(u) \\ &= \sum_n e^{\eta n} \hat{P}(\theta')^n. \end{aligned}$$

The latter sum converges if and only if  $\sum_n e^{\eta n} \lambda(\theta')^n < \infty$ , that is, if  $e^{\eta\lambda(\theta')} < 1$ . But  $\log \lambda$  has positive derivative  $m(\theta)$  at the point  $\theta$  and hence  $\lambda(\theta') < \lambda(\theta)$  when  $\theta'$  is less than but sufficiently close to  $\theta$ .

In the same way we see that  $\hat{L}(\eta, \theta)$  converges if  $e^{\eta\lambda(\theta')} < 1$  for some  $\theta' \geq \theta$ . It follows from (3.8) that  $\hat{L}(\eta, \theta)$  converges if  $\hat{G}(\eta, \theta)$  does. Therefore  $\hat{L}(\eta, \theta)$  converges if  $e^{\eta\lambda(\theta')} < 1$  for some  $\theta'$ .  $\square$

We get from (3.5),

$$(3.8) \quad \hat{L}(\eta, \theta) + \hat{G}(\eta, \theta) = I + e^{\eta} \hat{G}(\eta, \theta) \hat{P}(\theta)$$

or

$$(3.9) \quad \hat{G}(\eta, \theta)(I - e^{\eta} \hat{P}(\theta)) = I - \hat{L}(\eta, \theta).$$

We shall compare the two chains defined by (1.7) and (1.6), respectively, with (2.13). In order to keep clear the difference between the two chains we shall add a subscript or superscript  $P$  and, respectively,  $L$ .

Let  $\zeta = (\eta, \theta)$ . The mean-value vector  $m^L(\zeta) = (E_\zeta^L N, E_\zeta^L S_N)$  is defined by

$$(3.10) \quad \begin{aligned} E_\zeta^L N &= \sigma^L(\zeta) \cdot \frac{\partial \hat{L}(\zeta)}{\partial \eta} \rho^L(\zeta) / \lambda_L(\zeta), \\ E_\zeta^L S_N &= \sigma^L(\zeta) \cdot \frac{\partial \hat{L}(\zeta)}{\partial \theta} \rho^L(\zeta) / \lambda_L(\zeta). \end{aligned}$$

**THEOREM 3.2.** *Assume that  $m^P(\theta) > 0$ .*

(i) *The variables  $N$  and  $S_N$  are proper with respect to  $\text{prob}_\theta$  and  $E_\theta(t^N | \xi_0 = i) < \infty$  for some  $t > 1$  and all  $i \in X$ .*

(ii)  *$\lambda_L(\eta, \theta) = 1$  if and only if  $e^\eta \lambda_P(\theta) = 1$ .*

(iii) *If  $e^\eta \lambda_P(\theta) = 1$ , then  $E_\zeta^L N < \infty$  and*

$$(3.11) \quad \begin{aligned} E_\zeta^L S_N &= m^P(\theta) E_\zeta^L N, \\ \rho^L(\zeta) &= \rho^P(\theta), \\ \sigma^L(\zeta) \hat{G}(\zeta) &= E_\zeta^L N \sigma^P(\theta). \end{aligned}$$

(iv)

$$(3.12) \quad E \left( \lambda_P(\theta)^{-N} e^{\theta S_N} \frac{\rho_{\xi_N}^P(\theta)}{\rho_{\xi_0}^P(\theta)} \middle| \xi_0 = i \right) = 1.$$

The first of the identities (3.11), which is a Markov version of Wald's identity, is certainly related to the generalized Wald's identity of Franken and Lisek (1982). Note that it implies that  $m^L(\zeta)$  has the same direction as the vector  $(1, m^P(\theta))$  which is what we shall need. The identity (3.12) is a special case of Folgerung 1 of KÜCHLER and SEMJONOV (1979). We shall not use it but it serves as an intermediate step in the proof.

**PROOF OF THEOREM 3.2.** Put  $\zeta_0 = (\eta(\theta), \theta)$ , where  $\eta(\theta)$  is defined by  $e^{\eta(\theta)} \lambda(\theta) = 1$ .

The identity (3.9) yields

$$(3.13) \quad (I - \hat{L}(\zeta_0)) \rho^P(\theta) = 0,$$

that is, (3.12) holds and hence  $N$  and  $S_N$  are proper. The remainder of (i) follows from Lemma 3.1 and the observation that  $\hat{L}(\zeta)$  converges whenever  $\hat{G}(\zeta) < \infty$  [let  $t = e^\eta \lambda_P(\theta)$ ].

Another consequence of (3.9) is that  $\rho^L(\zeta_0) = \rho^P(\theta)$  and  $\lambda_L(\zeta_0) = 1$  [recall that  $\lambda_L(\zeta)$  is the only eigenvalue that has positive eigenvectors].  $\hat{L}(\zeta)$  increases strictly with  $\eta$  and the same must be true for  $\lambda_L(\zeta)$  because

$$(3.14) \quad \lambda_L(\zeta) = \max_{f \geq 0} \min_{i \in X} \frac{\hat{L}(\zeta) f(i)}{f(i)}$$

[see Wielandt (1950)]. Therefore  $\text{sign}(\lambda_L(\zeta) - 1) = \text{sign}(e^{\eta\lambda_P(\theta)} - 1)$  and hence (ii) follows.

The identity

$$(3.15) \quad \sigma^L(\zeta_0)\hat{G}(\zeta_0)(I - \lambda_P(\theta)^{-1}\hat{P}(\theta)) = 0$$

follows from (3.9) and hence there is a positive number  $c$  such that

$$(3.16) \quad \sigma^L(\zeta_0)\hat{G}(\zeta_0) = c\sigma^P(\theta).$$

Differentiate (3.9) with respect to  $\eta$ . Multiply by  $\sigma^L(\zeta_0)$  and  $\rho^L(\zeta_0)$  from the left and right and conclude that

$$(3.17) \quad c = E_{\zeta_0}^L N.$$

The remaining identity follows in the same way after differentiation of (3.9) with respect to  $\theta$ .  $\square$

We shall finally use the results of this section to show how Theorem 1.2 follows from Proposition 2.2.

PROOF OF THEOREM 1.2. Assume  $m(\theta) > 0$ . Define  $\eta(\theta)$  by  $e^{\eta(\theta)\lambda(\theta)} = 1$ . It follows from Theorem 3.2(ii) that  $\lambda^L(\eta, \theta) = 1$  if and only if  $\eta = \eta(\theta)$ .

Lemma 3.1 implies that  $(\eta(\theta), \theta) \in \Theta^L$  and  $(\eta(\theta), 0) \in \Theta^L$ , for all  $\theta$  in the open set  $\{\theta \in \Theta; m(\theta) > 0\}$ . Thus in particular  $E_{\zeta} S_N < \infty$ .

Theorem 3.2(iii) implies that  $\lambda'_L(\zeta) \neq 0$ ,  $\check{\theta}(n, u) = \hat{\theta}(u/n)$ ,  $T(\check{\zeta}) = n/E_{\theta}^P N$  and  $\rho_i^L(\check{\zeta}) = \rho_i^P(\hat{\theta})$ .

It is clear that  $\eta(\theta) \in \check{A}$  if and only if  $\theta \in \hat{A}$ .

The condition on  $L$  implies that  $\dim G(L) = 2$  and hence that  $\lambda''_L(\zeta)$  is strictly positive definite [see Lemma 1.2 in Höglund (1990b)]. The constant  $Q$  is thus given by

$$(3.18) \quad Q(\theta) = \frac{\lambda'_L(\zeta) \cdot \lambda''_L(\zeta)^{-1} \lambda'_L(\zeta)}{E_{\zeta} N (E_{\zeta} S_N)^2},$$

where  $\zeta = (\eta(\theta), \theta)$ .

The identity

$$(3.19) \quad c_{ij}(b, \zeta)/E_{\theta}^* S_N = \rho_i^P(\theta)\omega_j(b, \theta)$$

follows from the fact that  $\rho_i^L(\zeta) = \rho_i^P(\theta)$  and  $\sigma_i^L(\zeta) = \sigma_i^*(\theta)$ , when  $e^{\eta\lambda(\theta)} = 1$ .  $\square$

### REFERENCES

ARNDT, K. (1981). Properties of limit functionals of a random walk on finite Markov chains. *Math. Operationsforsch. Statist. Ser. Statist.* **12** 85–100. (In Russian.)  
 ARNDT, K. (1982). Finding in explicit form the distribution of the supremum of a random walk on a Markov chain. *Predel'nye Teoremy Teorii Veroyatnoste i Smezhnye Voprosy, Tr. Inst. Mat.* **1** 139–146. (In Russian.)

- FELLER, W. (1971). *An Introduction to Probability Theory and its Applications 2*, 2nd ed. Wiley, New York.
- FRANKEN, P. and LISEK, B. (1982). On Wald's identity for dependent variables. *Z. Wahrsch. Verw. Gebiete* **60** 143–150.
- HÖGLUND, T. (1990a). An asymptotic expression for the probability of ruin within finite time. *Ann. Probab.* **18** 378–389.
- HÖGLUND, T. (1990b). A multidimensional renewal theorem for finite Markov chains. *Ark. Mat.* **28** 273–287.
- KATO, T. (1966). *Permutation Theory for Linear Operators*. Springer, New York.
- KEILSON, J. and WISHART, D. M. G. (1964). A central limit theorem for processes defined on a finite Markov chain. *Proc. Cambridge Philos. Soc.* **60** 547–567.
- KÜCHLER, I. and SEMJONOV, A. (1979). Die Waldsche Fundamentalidentität und ein sequentieller Quotientest für eine zufällige Irrfahrt über einen homogenen irreduziblen Markovschen Kette mit endlichem Zustandsraum. *Math. Operationsforsch. Statist. Ser. Statist.* **10** 319–331.
- MILLER, H. D. (1961). A convexity property in the theory of random variables defined on a finite Markov chain. *Ann. Math. Statist.* **32** 1261–1270.
- MILLER, H. D. (1962a). A matrix factorization problem in the theory of random variables defined on a finite Markov chain. *Proc. Cambridge Philos. Soc.* **58** 268–285.
- MILLER, H. D. (1962b). Absorption probabilities for sums of random variables defined on a finite Markov chain. *Proc. Cambridge Philos. Soc.* **58** 286–298.
- PRESMAN, E. L. (1969). Factorization methods and boundary problems for sums of random variables given on Markov chains. *Izv. Akad. Nauk SSSR Ser. Mat.* **33** 815–852.
- WIELANDT, H. (1950). Unzerlegbare nicht-negative Matrizen. *Math. Z.* **52** 642–648.

DEPARTMENT OF MATHEMATICS  
ROYAL INSTITUTE OF TECHNOLOGY  
S-100 44 STOCKHOLM  
SWEDEN