THE ASYMPTOTIC BEHAVIOR OF A CERTAIN MARKOV CHAIN

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1. Introduction. This note is concerned with the limiting behavior of a Markov chain which arose as a model for a peculiar kind of growth process. Its precise definition is as follows: let $S_1 = 1$, and

$$(1.1) S_{N+1} = S_N + X_N (N+1-S_N), N = 1, 2, \cdots,$$

where $\{X_N\}$ is a sequence of independent random variables satisfying $0 \le X_N \le 1$. A problem that we pose, but do not solve, is to find necessary and sufficient conditions on the sequence $\{X_N\}$ to ensure that S_N/N converges a.s. to 1. A sufficient condition for this is presented, and also some related facts concerning the moments of S_N . It should be noted that the Markov chain S_N does not have temporally stationary transition probabilities and that the state space may be uncountable.

2. Convergence results.

THEOREM 1. If there exists an $\alpha > 0$ such that $E(X_i) \geq \alpha$ for all i, then $S_N/N \to 1$, a.s.

PROOF. Since $S_N = (1 - X_{N-1}) S_{N-1} + N \cdot X_{N-1}$ and $E(1 - X_N) \le 1 - \alpha < 1$, for all N,

$$E(N - S_N) = E\{(1 - X_{N-1})(N - S_{N-1})\}\$$

$$= E(1 - X_{N-1}) \cdot E(N - 1 - S_{N-1}) + E(1 - X_{N-1})\$$

$$\leq (1 - \alpha) E(N - 1 - S_{N-1}) + (1 - \alpha).$$

By induction then $E(N-S_N) \leq \sum_{1}^{N-1} (1-\alpha)^i < K < \infty$. Hence $E(N-S_N) < K$, for all N. Similarly,

$$E\{(N - S_{N})^{2}\} = E\{(1 - X_{N-1})^{2}\}E\{(N - S_{N-1})^{2}\}$$

$$= E\{(1 - X_{N-1})^{2}\}E\{(N - 1 - S_{N-1})^{2}\} + E\{(1 - X_{N-1})^{2}\}$$

$$+ 2E\{(1 - X_{N-1})^{2}\}E\{(N - 1 - S_{N-1})^{2}\}$$

$$\leq (1 - \alpha)E\{(N - 1 - S_{N-1})^{2}\} + K'$$

where K' does not depend on N. Again by induction we have $E\{(N-S_N)^2\} \le K''$, for all N. Now for $\epsilon > 0$,

$$P(|S_N/N-1| > \epsilon) = P(|S_N-N|N^{-1} > \epsilon) \le E\{(N-S_N)^2\}/N^2\epsilon^2 \le K''/N^2\epsilon^2$$
.
Thus $\sum_{1}^{\infty} P(|S_N/N-1| > \epsilon) < \infty$ which implies that $S_N/N \to 1$, a.s.

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THEOREM 2. The sequence $\{\text{Var }S_N\}$ is bounded. Moreover, if $E(X_N) = \alpha$, and $E(X_N^2) = \alpha_2$, for all N, where $\alpha > 0$ and α_2 are constants then $\lim_{N\to\infty} \text{Var } S_N = \alpha^{-2} \cdot (\alpha_2 - \alpha^2) (2\alpha - \alpha_2)^{-1}$.

PROOF. The first statement follows from the proof of the last theorem: Var $S_N = \operatorname{Var}(N - S_N) \le E\{(N - S_N)^2\} \le K''$. Now suppose $E(X_N) = \alpha$, $E(X_N)^2 = \alpha$, and Var $S_N = \sigma_N^2$, for all N. Then by squaring both sides of (1.1), taking expectations, and performing some routine algebraic reductions we derive the identity

(2.1)
$$\sigma_{N+1}^2 = \beta \sigma_N^2 + r_N + \gamma, \quad N = 1, 2, \cdots, \quad \sigma_1^2 = 0,$$

where $\beta = 1 - 2\alpha + \alpha_2$, $\gamma = \alpha_2/\alpha^2 - 1$, and $r_N \to 0$ as $N \to \infty$. Since $0 < \beta < 1$, it is clear that the solution to the difference equation (2.1) converges to a limit L satisfying

$$L = \beta L + \gamma,$$

So that $L = \gamma/1 - \beta = \alpha^{-2} \cdot (\alpha_2 - \alpha^2) (2\alpha - \alpha_2)^{-1}$. (To derive (2.1) one should first note that $E(S_N) = N + 1 - \alpha^{-1} + \alpha^{-1} (1 - \alpha)^N$.)

3. Remarks.

- (i) Theorem 1 may fail to hold if, for example, $E(X_N) \to 0$. In fact, if $E(X_N) = 1/N + 1$, then $E(S_N/N) \to \frac{1}{2}$ so that $S_N/N \to 1$.
- (ii) For the case of an arbitrary sequence $\{X_N\}$, $0 \le X_N \le 1$, it is not known whether $\lim_{N\to\infty} E(N-S_N)$ exists.
- (iii) All of these questions pertain to the stability of the solutions to a linear stochastic difference equation, for various conditions on the (random) coefficients. It would be of interest to have some general results along these lines.

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