NOTES

AN EXAMPLE IN DENUMERABLE DECISION PROCESSES

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- 1. Introduction and summary. This note uses the terminology and notation of Ross [3]. The example presented here is a denumerable Markovian decision process which has an optimal nonstationary rule which is better than every stationary rule. This answers a question of Derman [1].
 - 2. The example.

$$I = \{0, 1, 1', 2, 2', 3, 3', \cdots\}, \qquad K_0 = K_{i'} = 1, \qquad K_i = 2,$$

$$P(0, i:1) = P(0, i':1) = \frac{3}{2}(\frac{1}{4})^i, \qquad i > 0,$$

$$P(i, 0:1) = (\frac{1}{2})^i \qquad = 1 - P(i, i':1),$$

$$P(i, 0:2) = \frac{1}{2} \qquad = 1 - P(i, i + 1:2),$$

$$P(i', 0:1) = (\frac{1}{2})^i \qquad = 1 - P(i', i':1),$$

$$C(0, \cdot) = 1, \qquad \text{all other costs are zero, i.e.,}$$

$$C(i, \cdot) = 0 = C(i', \cdot), \qquad i > 0.$$

Let R_n be the stationary deterministic rule which takes action 2 at states 0 < i < n and action 1 elsewhere

$$M_{00}(R_n) \, = \, 1 \, + \, \sum_{j=1}^{\infty} \tfrac{3}{2} (\tfrac{1}{4})^j M_{j0}(R_n) \, + \, \sum_{j=1}^{\infty} \tfrac{3}{2} (\tfrac{1}{4})^j 2^j.$$

Now $j \ge n \Rightarrow M_{j0}(R_n) = 2^j$, whereas

$$j < n \Rightarrow M_{j0}(R_n) = \frac{1}{2} + 2(\frac{1}{2})^2 + \dots + (n-j)(\frac{1}{2})^{n-j} + (\frac{1}{2})^{n-j}[n-j+2^n]$$
$$= 2 + 2^j - (\frac{1}{2})^{n-j-1}.$$

Therefore

$$\begin{split} M_{00}(R_n) &= \frac{5}{2} + \sum_{j=1}^{\infty} \frac{3}{2} (\frac{1}{4})^j (2 + 2^j) - \sum_{j=1}^{n-1} \frac{3}{2} (\frac{1}{4})^j (\frac{1}{2})^{n-j-1} \\ &- 2 \sum_{j=n}^{\infty} \frac{3}{2} (\frac{1}{4})^j \\ &= 5 - \sum_{j=1}^{n-1} \frac{3}{2} (\frac{1}{2})^j (\frac{1}{2})^{n-j-1} - 3 \sum_{j=n}^{\infty} (\frac{1}{4})^j. \end{split}$$

Hence $M_{00}(R_n) < 5$ for all n, and $M_{00}(R_n) \rightarrow 5$ as $n \rightarrow \infty$.

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Now let R be any stationary rule, let P_i be the probability that R takes action 1 when in state i. Now

$$M_{00}(R) = 1 + \sum_{i=1}^{\infty} \frac{3}{2} (\frac{1}{4})^{i} M_{i0}(R) + \sum_{j=1}^{\infty} \frac{3}{2} (\frac{1}{4})^{j} 2^{j}$$

but

$$M_{j0}(R) = \sum_{n=j}^{\infty} \left[P_n \prod_{k=j}^{n-1} (1 - P_k) \right] M_{j0}(R_n) + 2 \prod_{k=j}^{\infty} (1 - P_k)$$

$$< (2 + 2^j) \left[\sum_{n=j}^{\infty} P_n \prod_{k=j}^{n-1} (1 - P_k) + \prod_{k=j}^{\infty} (1 - P_k) \right] = 2 + 2^j.$$

Consequently

$$M_{00}(R) < 5$$
 for all stationary rules R , and $\varphi(i, R) > \frac{1}{5}$ for all stationary rules R , for all i .

However if we consider the nonstationary rule R^* which uses

$$R_1$$
 for $t = 1, 2, \dots, N_1$,
 R_2 for $t = N_1 + 1, \dots, N_1 + N_2$,
 \vdots
 R_n for $t = \sum_{i=1}^{n-1} N_i + 1, \dots, \sum_{i=1}^{n} N_i$,
 \vdots

it can be shown (as in Theorem 4.3 of Ross [3]) that there exists N_i 's such that $\varphi(i, R^*) = \lim_n \varphi(i, R_n) = \frac{1}{5}$. It also follows from Theorem 3.1 of Ross [3] that R^* is optimal.

Since every stationary rule gives rise to a recurrent Markov chain the results of Fisher [2] may also be used to show this result.

REFERENCES

- Derman, C. (1966). Denumerable state Markovian decision processes—average cost criterion. Ann. Math. Statist. 37 1545-1553.
- [2] Fisher, L. (1968). On recurrent denumerable decision processes. Ann. Math. Statist. 39 424-432.
- 3] Ross, S. M. (1968). Non-discounted denumerable Markovian decision models. Ann. Math. Statist. 39 412-423.