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AFDELING MATHEMATISCHE BESLIJKUNDE

BN 18/73

JULY

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A NOTE ON HOWARD'S VALUE DETERMINATION STEP

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BIBLIOTHEEK MATHEMATISCH CENTRUM
AMSTERDAM

Printed at the Mathematical Centre, 49, 2e Boerhaavestraat, Amsterdam.

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AMS (MOS) subject classification scheme (1970): 90C40

1. INTRODUCTION

Let P be an $N \times N$ Markov matrix whose (i,j) element is p_{ij} ($i,j=1,\dots,N$), i.e., $p_{ij} \geq 0$ and $\sum_j p_{ij} = 1$. Let T be an N component column vector whose i th element is T_i where $T_i > 0$ for $i=1,\dots,N$, and let q be an N component column vector whose i th element is q_i ($i=1,\dots,N$). The triple (P,T,q) can be thought of as a semi-Markov reward process with transition probabilities p_{ij} , expected transition times T_i and one-transition rewards q_i . It is assumed that the Markov matrix P has a single recurrent chain. Let state N be a recurrent state of the Markov matrix P .

In each iteration of Howard's [2] well known policy-iteration algorithm a set of linear simultaneous equations must be solved. For the single chain case this set of equations is of the following type:

$$gT + v = q + Pv, \quad (1)$$

where g is an unknown scalar and v is an unknown N component column vector whose i th element is v_i ($i=1,\dots,N$). It is important to have an efficient method for solving (1). For the case where P is an aperiodic Markov matrix Morton [4] has given a simple iterative scheme to solve (1).

The purpose of this note is to demonstrate that a solution of (1) can be found by solving two sets of linear simultaneous equations which are more easy to tackle than (1). In our approach we need not require that P is aperiodic. Despite the fact that our approach is implied in the paper of Derman and Veinott[1], the theorem below seems to have passed unnoticed.

2. RESULTS

We first introduce some notation. Let T^* be the $N-1$ component column vector whose i th element is T_i , let q^* be the $N-1$ component column vector whose i th element is q_i , and let R be the $N-1$ component row vector whose i th element is p_{Ni} ($i=1, \dots, N-1$). Denote by Q the $(N-1) \times (N-1)$ matrix whose (i,j) element is p_{ij} ($i,j=1, \dots, N-1$). Observe that $Q^n \rightarrow 0$ as $n \rightarrow \infty$, since N is a recurrent state of the Markov matrix P .

We have the following theorem (cf. Derman and Veinott [1] and Theorem 1 of Morton [4])

THEOREM. *Let the column vector $x=(x_1, \dots, x_{N-1})$ be the unique solution to*

$$x = q^* + Qx, \quad (2)$$

and let the column vector $y=(y_1, \dots, y_{N-1})$ be the unique solution to

$$y = T^* + Qy. \quad (3)$$

Define the scalar g by

$$g = (q_N + Rx) / (T_N + Ry), \quad (4)$$

and define the N component column vector $v=(v_1, \dots, v_N)$ by

$$v_i = x_i - gy_i \quad \text{for } i=1, \dots, N-1, \quad v_N = 0. \quad (5)$$

Then g, v satisfy equation (1).

Proof. Let us first observe that both (2) and (3) have a unique solution, since $Q^n \rightarrow 0$ as $n \rightarrow \infty$. Denote by v^* the $N-1$ component column vector whose

ith element is v_i ($i=1, \dots, N-1$). From (2), (3) and (5),

$$gT^* + v^* = gT^* + q^* + Qx - g(T^* + Qy) = q^* + Q(x - gy) = q^* + Qv^* ,$$

while from (4) and (5) it follows that

$$gT_N^* + v_N^* = q_N^* + Rx - gRy = q_N^* + R(x - gy) = q_N^* + Rv^* .$$

Using $v_N = 0$ the theorem now follows.

Observe that g in (4) can be interpreted as the ratio of the expected return earned during a cycle and the expected length of a cycle, where a cycle is defined as the time interval between two successive visits to the recurrent state N . It is well-known that this ratio equals the long-run average return.

Remark. Suppose that $p_{iN} = 1 - \alpha_i > 0$ for $i=1, \dots, N-1$. Let z_0 be an arbitrary $N-1$ component column vector, and for $n \geq 1$ define z_n by $z_n = b + Qz_{n-1}$, where b is a given $N-1$ component column vector. Let z be the unique solution to $z = b + Qz$. Define for any $n \geq 1$,

$$u_n'(i) = z_n(i) + (1 - \alpha_i)^{-1} \min_j \{z_n(j) - z_{n-1}(j)\} \quad \text{for } i=1, \dots, N-1 ,$$

and

$$u_n''(i) = z_n(i) + (1 - \alpha_i)^{-1} \max_j \{z_n(j) - z_{n-1}(j)\} \quad \text{for } i=1, \dots, N-1 .$$

Then, for any $n \geq 1$, $u_n'(i) \leq z(i) \leq u_n''(i)$ for $i=1, \dots, N-1$, where $u_n'(i)$ is nondecreasing in n to $z(i)$ and $u_n''(i)$ is nonincreasing in n to $z(i)$ for all i . The proof of this assertion is a slight modification of proofs given by Macqueen [3] and is based on the following fact: If $Tu \leq Tw$ then $u \leq w$, where the transformation T is defined by $Tu = u - (b + Qu)$ for any $N-1$ component column vector u .

Remark. It is straightforward to extend the analysis above to the case of a general Markov matrix P ; in this case the set of simultaneous equations $g=Pg$ and $g^T+v=q+Pv$ has to be solved where g and v are unknown N component column vectors.

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