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ON A QUASI-LINEAR EQUATION

Richard Bellman

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SUMMARY


 The purpose of this paper is to establish some limit theorems for the solutions of  $x_1(n+1) = \max_{j=1, \dots, q} \sum_{j=1}^q a_{1j}(q)x_j(n)$ . *non-linear recurrence relations.*

Recurrence relations of this kind occur in various dynamic programming problems. ( ) 

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ON A QUASI-LINEAR EQUATION

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§1. Introduction

The purpose of this note is to establish some limit theorems for ~~the~~ non-linear recurrence relations

$$(1) \quad x_1(n+1) = \text{Max}_q \sum_{j=1}^n a_{1j}(q)x_j(n), \quad j = 1, 2, \dots, n, \quad n \geq 0,$$

under certain assumptions concerning the initial values  $c_1 = x_1(0)$ , and the coefficient matrices  $A(q) = (a_{1j}(q))$ .

Equations of this type occur in various parts of the theory of dynamic programming, as we shall indicate below, and are, in addition, of interest in furnishing a link between the theory of linear and non-linear operations, as we have discussed elsewhere, cf [1], [2], [3], [4].

§2. The Homogeneous Equation

Let us consider the equation

$$(1) \quad \lambda y_1 = \text{Max}_q \sum_{j=1}^n a_{1j}(q)y_j, \quad i = 1, 2, \dots, n,$$

where we impose the following conditions.

- (2) (a)  $q = (q_1, q_2, \dots, q_n)$  runs over some set of values,  $S$ , with the property that the maximum is attained in (1),
- (b)  $\infty > m \geq a_{1j}(q) > 0$ ,  $i, j = 1, 2, \dots, n$  for  $q \in S$ ,
- (c) for any  $q$ , let  $\phi(q)$  denote the characteristic root of  $A(q) = (a_{1j}(q))$  of largest absolute value, the Perron

root, known to be positive. We assume that there exists at least one value of  $q$  for which  $\phi(q)$  assumes its maximum for  $q \in S$ .

We shall now prove

Theorem 1. Under these conditions, there exists a unique positive  $\lambda$  with the property that (1) has a positive solution,  $y_i > 0$ ,  $i=1,2,\dots,n$ . This solution is unique up to a multiplicative constant, and

$$(3) \quad \lambda = \text{Max}_{q \in S} \phi(q).$$

Proof. We begin by showing the existence of a positive  $\lambda$  and a positive set of solutions  $\{y_i\}$ . Consider the region defined by  $y_i \geq 0$ ,  $\sum_{i=1}^n y_i = 1$ . The normalized transformation

$$(4) \quad y_i' = \left[ \text{Max}_q \sum_{j=1}^n a_{ij}(q)y_j \right] / \left[ \sum_{i=1}^n \text{Max}_q \sum_{j=1}^n a_{ij}(q)y_j \right],$$

is a continuous mapping of this region into itself. Hence there exists a fixed point,  $\{y_i\}$ . This fixed point is a solution of (1), with  $\lambda$  the denominator in (4). Each component  $y_i$  is positive because of the positivity of  $a_{ij}(q)$ .

To show that this solution is unique up to a multiplicative constant, let  $[\mu, z]$  be another solution of (1) with  $\mu > 0$  and  $z$  a positive vector. Let  $\{q\}$  be the set of values for which the maximum is attained in (1), and  $\{\bar{q}\}$  the similar set associated with  $z$ . Observe that we may have different sets for each  $i$ . We have then

$$(5) \quad \lambda y_1 = \sum_j a_{1j}(q)y_j \geq \sum a_{1j}(\bar{q})y_j, \quad 1 = 1, 2, \dots, n,$$

$$\mu z_1 = \sum_j a_{1j}(\bar{q})z_j.$$

Let us now assume, without loss of generality that  $\lambda < \mu$ . Let  $\epsilon$  be a positive constant chosen so that one, at least, of the components  $y_1 - \epsilon z_1$  is zero, one at least is positive, and the others are non-negative. This can always be accomplished if  $y$  and  $z$  are not proportional. If  $1$  is an index for which  $y_1 - \epsilon z_1$  is zero, we have

$$(6) \quad 0 = \mu(y_1 - \epsilon z_1) > \lambda y_1 - \epsilon \mu z_1 \geq \sum_{j=1}^n a_{1j}(\bar{q})(y_j - \epsilon z_j) > 0,$$

Since  $a_{1j}(\bar{q}) > 0$ , a contradiction. Hence  $\lambda = \mu$  and  $y$  and  $z$  are proportional.

To show that  $\lambda = \text{Max}_q \phi(q)$ , we proceed as follows. Let  $\mu = \text{Max}_q \phi(q)$ . It is clear that  $\lambda$ , as the characteristic root of some  $A^q$ , satisfies the inequality  $\lambda \leq \mu$ . Assume that actually  $\mu > \lambda$ . Let  $z = (z_1, z_2, \dots, z_n)$  be a positive characteristic vector associated with  $\mu$  and  $\bar{q}$  a set of  $q$ -values which yield  $\mu = \phi(\bar{q})$ . Then we have

$$(7) \quad \mu z_1 = \sum_{j=1}^n a_{1j}(\bar{q})z_j \leq \text{Max}_q \sum_{j=1}^n a_{1j}(q)z_j$$

Since  $y_1$  is positive, we can find a positive constant  $m$  such that  $z_1 \leq m y_1$  for  $1 = 1, 2, \dots, n$ . Hence, (7) yields

$$(8) \quad \mu z_1 \leq m \text{Max}_q \sum_{j=1}^n a_{1j}(q)y_j = m \lambda y_1$$

Thus  $z_1 \leq my_1 \lambda/\mu$ . Iterating this we obtain  $z_1 \leq my_1 (\lambda/\mu)^k$ , for arbitrary  $k$ . Since  $\lambda/\mu < 1$ , by assumption, this yields  $z_1 = 0$ , a contradiction. Hence  $\lambda = \mu$ .

### §3. The Recurrence Relation

Let us now return to the recurrence relation of (1.1) and prove

Theorem 2. If, in addition to the conditions of (2.2), we assume that there is a unique  $q$  for which the maximum value of  $\phi(q)$  is attained and that  $c_1 > 0$ , then

$$(1) \quad x_1(n) \sim ay_1 \lambda^n,$$

as  $n \rightarrow \infty$ , where  $a$  is a constant dependent upon the initial values  $c_1$ .

Proof. Let us take  $c_1 > 0$  without loss of generality. There are then two positive constants  $k$  and  $K$  such that  $ky_1 \leq c_1 \leq Ky_1, i=1,2,\dots,n$ .

Let us show inductively that

$$(2) \quad ky_1 \lambda^n \leq x_1(n) \leq Ky_1 \lambda^n.$$

Assume that we have the result for  $n$ , then

$$(3) \quad x_1(n+1) \leq K\lambda^n \text{Max}_q \sum_{j=1}^n a_{1j}(q)y_j = K\lambda^{n+1}y_1 \\ \geq k\lambda^n \text{Max}_q \sum_{j=1}^n a_{1j}(q)y_j = k\lambda^{n+1}y_1.$$

To establish the asymptotic behavior we show that for  $n$  sufficiently large the set of  $q$ 's which furnish the maximum in (1.1) is precisely the set which yields  $\lambda = \text{Max } \phi(q)$ .

Assume the contrary. This means that infinitely often we employ a set  $\{\bar{q}\}$  which is not identical with the  $q$  which furnishes the maximum in  $\phi(q)$ .

We then have

$$(4) \quad x_1(n+1) = \sum_{j=1}^n a_{1j}(\bar{q})x_j(n), \quad i=1,2,\dots,n$$

$$\leq \left(\sum_{j=1}^n a_{1j}(\bar{q})y_j\right) K\lambda^n.$$

For some index  $i$  we must have

$$(5) \quad \sum_{j=1}^n a_{1j}(\bar{q})y_j < \lambda y_1,$$

with strict inequality. For if  $\sum_{j=1}^n a_{1j}(\bar{q})y_j \geq \lambda y_1$  for all  $i$ , the characteristic root of  $A(\bar{q}) = (a_{ij}(\bar{q}))$  of largest absolute value,  $\phi(\bar{q})$ , would at least equal  $\lambda = \underset{q}{\text{Max}} \phi(q)$ , which would contradict

the assumption concerning the uniqueness of the maximum of  $\phi(q)$ .

Hence, for some component, say the first, we have

$$(6) \quad x_1(n+1) \leq \theta K\lambda^{n+1}y_1, \quad 0 < \theta < 1.$$

Since  $a_{1j}(\bar{q}^*) > 0$  for  $i, j$ , where  $\bar{q}^*$  is the value of  $q$  for which  $\lambda = \phi(\bar{q}^*)$ , we see that, for  $i = 1, 2, \dots, n$ ,

$$(7) \quad x_1(n+2) \leq K\lambda^{n+1} \left[ \sum_{j=2}^n a_{1j}(\bar{q}^*)y_1 + \theta a_{11}(\bar{q}^*)y_1 \right]$$

$$\leq \theta_1 K\lambda^{n+2}y_1,$$

where  $\theta < 1$ .

If therefore a set of  $q$ 's distinct from  $q^*$  are used  $R$  times, we obtain

$$(8) \quad x_1(n) \leq \theta_1^R K \lambda^n y_1,$$

for  $n$  sufficiently large. Since  $0 < \theta_1 < 1$ , if  $R$  is too large we eventually contradict the lower bound for  $x_1(n)$ .

Hence for  $n \geq n_0 = n_0(c_1)$ , we have

$$(9) \quad x(n+1) = A(q^*) x(n),$$

whence the asymptotic statement of (1) follows.

#### §4. A Dynamic Programming Problem

Suppose that we are engaged in a multi-stage decision process of the following type. At each stage we have our choice of various operations, which we number  $i=1,2,\dots,K$ . The  $i^{\text{th}}$  operation has a probability distribution attached with the following properties:

- (1) a. There is a probability  $p_{1k}$  that we receive  $k$  units and the process continues,  $k=1,2,\dots,R$ ,
- b. There is a probability  $p_{10}$  that we receive nothing and the process terminates.

How do we proceed so as to maximize the probability that we receive at least  $n$  units before the process terminates?

Let us define the sequence

- (2)  $u(n)$  = the probability of attaining at least  $n$  units before the termination of the process using an optimal procedure.



Then using the intuitive "principle of optimality", cf. [1], [2], [3], [4], we see that  $u(n)$  satisfies the recurrence relation

$$(3) \quad u(n) = \underset{1}{\text{Max}} \left[ \sum_{k=1}^R p_{1k} u(n-k) \right], \quad n > 0$$

$$= 1, \quad n \leq 0.$$

Using methods similar to those above, we see that for large  $n$ ,

$$(4) \quad u(n) \sim c \rho^n,$$

where  $\rho$  is the root of largest absolute value, necessarily positive, of

$$(5) \quad 1 = \sum_{k=1}^R p_{1k} \rho^{-k},$$

for the value of  $1$  which maximizes  $\rho$ .

### §5. An Analogue of a Result of Markoff

Markoff showed that if

$$(1) \quad x_i(n+1) = \sum_{j=1}^n a_{ij} x_j(n), \quad n = 0, 1, \dots,$$

$x_i(0) > 0$ , with the conditions

$$(2) \quad a_{ij} > 0, \quad \sum_j a_{ij} = 1, \quad i = 1, 2, \dots, n,$$

then

$$(3) \quad \lim_{n \rightarrow \infty} x_i(n) = c, \quad i = 1, 2, \dots, n,$$

where  $c$  depends on the initial values.

The same proof shows that the same result holds for the sequence defined by

$$(4) \quad x_1(u+1) = \text{Max}_q \sum_{j=1}^n a_{1j}(q)x_j(n),$$

provided that the conditions in (2) hold uniformly in  $q$ . The constant will, of course, in general, be different from that above.

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