An Iterative Procedure for Computing the Maximal Root of a Positive Matrix

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I. Introduction. A. Kolmogoroff proved the following theorem [2].

Theorem A. Suppose A is a positive matrix of order n. Then A is similar to a positive generalized stochastic matrix with each row sum equal to the maximal positive characteristic root, $\lambda(A)$, of A.

The proof of Theorem A depends upon the fact that each positive matrix has a positive characteristic vector corresponding to the maximal positive characteristic root.

Professor A. T. Brauer developed a practical method in [1] for computing the maximal characteristic root of a positive matrix to any desired degree of accuracy. In particular, he proved

THEOREM B. Let A be a positive square matrix of order n, and let R(r) denote the largest (smallest) row sum of A. If $\epsilon > 0$, there exists a matrix, $F(\epsilon)$, which is similar to A such that $R^* - r^* < \epsilon$, where R^* (r^*) denotes the largest (smallest) row sum of $F(\epsilon)$.

It is known that $r \leq \lambda(A) \leq R$ [1, Theorem 13, p. 21], and the inequality is strict unless R = r. Thus it follows immediately from Theorem B that $\lambda(A)$ can be computed to any desired degree of accuracy.

We shall offer a new method for computing the maximal root of a positive matrix.

II. An Iterative Procedure for Computing the Maximal Root. We shall use the following notation. Suppose A is a positive matrix of order n. Let R_i denote the ith row sum of A, $R = \max_i \{R_i\}$, and $r = \min_i \{R_i\}$.

THEOREM 1. Suppose A is a positive matrix of order n, with R > r. Let $S = \text{diag } (R_1, \dots, R_n)$. Then $S^{-1}AS$ is a positive matrix with each row sum in the interval (r, R).

Proof. Assume, without loss of generality, that $R = R_1 \ge R_2 \ge \cdots \ge R_n = r$. Then $S^{-1}AS = (a_{ij}R_j/R_i)$ is a positive matrix. Moreover,

$$\frac{\sum_{j=1}^{n} a_{ij} R_{j}}{R_{i}} < \frac{\sum_{j=1}^{n} a_{ij} R_{1}}{R_{i}} = R,$$

and

$$\frac{\sum_{j=1}^{n} a_{ij} R_{j}}{R_{i}} > \frac{\sum_{j=1}^{n} a_{ij} R_{n}}{R_{i}} = r , \text{ for } i = 1, 2, \dots, n .$$

Hence each row sum of $S^{-1}AS$ lies in the interval (r, R).

If we denote $B_1 = S^{-1}AS$, then we can transform B_1 by a similarity transformation, obtaining B_2 , such that each row sum of B_2 lies in the interval determined

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by the largest and smallest row sums of B_1 . We continue in this manner and obtain a sequence $B_1, B_2, \dots, B_n, \dots$ such that the difference of the largest row sum and the smallest row sum of B_i is decreasing at each step. Theorem 2 assures us that we do obtain convergence to a positive generalized stochastic matrix.

THEOREM 2. Suppose A is a positive matrix of order n, and let $B_1, B_2, \dots, B_n, \dots$ denote the sequence of matrices obtained from A by the procedure outlined above. Assume $a_{k1}/R_k = \min_i \{a_{i1}/R_i\}$. Then

max row sum of B_n – min row sum of $B_n \leq (1 - a_{k1}/R_k)^n(R - r)$.

Proof. Note the following inequalities:

$$i\text{th row sum of } B_1 = \sum_{j=1}^n a_{ij} \frac{R_j}{R_i} \le \sum_{j=1}^{n-1} \frac{a_{ij}R}{R_i} + \frac{a_{in}r}{R_i} = \sum_{j=1}^{n-1} \frac{a_{ij}(R-r)}{R_i} + r$$

$$= \left(1 - \frac{a_{in}}{R_i}\right)(R-r) + r = \left(1 - \frac{a_{in}}{R_i}\right)R + \frac{a_{in}}{R_i}r.$$

Assume $\min_{i} \{a_{in}/R_{i}\} = a_{pn}/R_{p}$. Then

(1) ith row sum of $B_1 \le (1 - a_{pn}/R_p)R + (a_{pn}/R_p)r$ for each i . Similarly, we have

(2) ith row sum of $B_1 \ge (1 - a_{k1}/R_k)r + (a_{k1}/R_k)R$ for each i, where $a_{k1}/R_k = \min_i \{a_{i1}/R_i\}$.

Using (1) and (2), it follows that

max row sum of B_1 — min row sum of B_1 $\leq (1 - a_{pn}/R_p)R + (a_{pn}/R_p)r - [(1 - a_{k1}/R_k)r + (a_{k1}/R_k)R]$ $= (1 - [a_{pn}/R_p + a_{k1}/R_k])(R - r) < (1 - a_{k1}/R_k)(R - r)$.

Hence the result holds when n = 1; i.e.

- (3) max row sum of B_1 min row sum of $B_1 < (1 a_{k1}/R_k)(R r)$. In the same manner, it is true that
- (4) $\max \text{ row sum of } B_2 \min \text{ row sum of } B_2$ $\leq \left(1 \min_{i} \left\{\frac{a_{i1}R_1}{\sum_{j=1}^{n} a_{ij}R_j}\right\}\right) (\max \text{ row sum of } B_1 \min \text{ row sum of } B_1).$

But

(5)
$$\frac{a_{i1}R_1}{\sum_{i=1}^n a_{ii}R_i} \ge \frac{a_{i1}R_1}{\sum_{i=1}^n a_{ii}R_1} \ge \frac{a_{k1}}{R_k} \quad \text{for each } i.$$

Hence

$$\min_{i} \left\{ \frac{a_{i1}R_1}{\sum_{j=1}^{n} a_{ij}R_j} \right\} \ge \frac{a_{k1}}{R_k} .$$

Finally, with (4) and (5), we have

(6) max row sum of B_2 – min row sum of $B_2 \le (1 - a_{k1}/R_k)^2(R - r)$.

The remainder of the proof is an easy induction.

Theorem 1 yields a simple method for the computation of the maximal positive root of a positive matrix.

III. An Example. The following computations were performed at the University of North Carolina at Charlotte Computer Center, and the author gratefully acknowledges the use of an algorithm prepared by Professors D. Nixon and L. Davis. Let

$$A = \begin{pmatrix} 4 & 7 & 25 & 2\\ 1 & 12 & 5 & 18\\ 3 & 6 & 15 & 2\\ 5 & 2 & 2 & 1 \end{pmatrix}.$$

	$\min \ row \ sum$	max row sum
A:	10	38
B_1 :	21.6666	32.4
B_2 :	25.5974	26.9567
B_3 :	26.2425	26.6185
B_4 :	26.3759	26.4030
B_5 :	26.3807	26.3931
$B_{\mathfrak{6}}$:	26.3831	26.3873
B_7 :	26.3845	26.3853
B_8 :	26.3849	26.3850

All numbers are truncated at four decimal places, and the computations were performed in double-precision arithmetic.

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^{1.} Alfred Brauer, "On the characteristic roots of non-negative matrices," Recent Advances in Matrix Theory, (Proc. Advanced Seminar, Math. Res. Center, U. S. Army), Univ. of Wisconsin Press, Madison, Wis., 1964, pp. 3–38. MR 29 #5835.

2. N. DMITRIEV & E. DYNKIN, "On characteristic roots of stochastic matrices," Bull. Acad. Sci. URSS Sér. Math., v. 10, 1946, pp. 167–184. MR 8, 129.