SIGN COMPATIBLE EXPRESSIONS FOR MINORS OF THE MATRIX I - A

BY

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ABSTRACT. Let $A = (a_{ij})$ be an $n \times n$ nonnegative matrix having row sums less than or equal to one. This paper shows that the *ij*th minor of I - A can be expressed as

where

$$(-1)^{i+j} \sum \prod r_k a_{pq}$$

$$r_k = 1 - \sum_{s=1}^n a_{ks}$$

and each $[]r_k a_{pq}$ is a product of exactly n - 1 numbers taken from r_k , a_{pq} for k, p, q = 1, ..., n. This theorem is then used to obtain perturbation results concerning the matrix I - A.

Perturbation results in matrix theory are concerned with estimating the error in matrix computations. This paper provides perturbation results for the matrix I - A where $A = (a_{ij})$ is nonnegative having row sums less than or equal to one. The method by which these perturbation results are achieved is a variant of that given by Sengupta [2] in his work on comparing stochastic eigenvectors of two irreducible stochastic matrices. The method, as we apply it, first gives expressions for the minors of I - A, in terms of the entries of A, and then uses these expressions to produce useful perturbation results for this matrix.

The theorem of the paper produces expressions for the minors of I - A.

THEOREM. Let A be an $n \times n$ nonnegative matrix having largest row sum less than or equal to one. Then

$$|(I-A)_{ij}| = (-1)^{i+j} \sum \prod r_k a_{pq}$$

where

$$r_k = 1 - \sum_{s=1}^n a_{ks}$$

and each $\prod r_k a_{pq}$ is a product of exactly n-1 numbers taken from r_k , a_{pq} for $k, p, q = 1, \ldots, n$.

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PROOF. The method of proof is induction on n. The case n = 2 can be proved by checking all choices for i and j. Thus, suppose the theorem is true for all $n \times n$ matrices A, satisfying the hypothesis, where $n < n_1$. Now let A be an $n \times n$ matrix, satisfying the hypothesis, where $n = n_1$. The argument is divided into two cases. The first case to be considered is when $i \neq j$. Here we will assume i < j as the case i > j is argued similarly.

For this case, we first define an $n \times n$ matrix $E_{pq} = (e_{rs})$ where

$$e_{rs} = \begin{cases} 1, & \text{if } r = s \text{ and } r \neq p, r \neq q, \\ 1, & \text{if } r = p \text{ and } s = q, \\ 1, & \text{if } r = q \text{ and } s = p, \\ 0, & \text{otherwise.} \end{cases}$$

Let $P = E_{12}E_{23} \cdots E_{i-1,i}$, a permutation matrix, and set P(I - A)P' = I - PAP' = I - B. Then

$$|(I - A)_{ij}| = (-1)^{i-1} |(I - B)_{1j}|.$$

Hence we need only prove the result for i = 1 and j > 1.

For this then we expand $|(I - A)_{1i}|$ about the 1st column achieving that

$$|(I - A)_{1j}| = \sum_{s>1} (-1)^{s-1+1} (-a_{s1}) |[(I - A)_{1j}]_{s1}|$$

where $[(I - A)_{1j}]_{s1}$ denotes the matrix obtained from (I - A) by deleting rows 1, s and columns j, 1.

Now noting that $[(I - A)_{1j}]_{s1} = [(I - A)_{11}]_{sj}$ and applying the induction hypothesis yields that

$$|(I - A)_{1j}| = \sum_{s>1} (-1)^{s+1} a_{s1} | [(I - A)_{11}]_{sj} |$$

= $\sum_{s>1} (-1)^{s+1} a_{s1} ((-1)^{s+j} \sum \prod (r_k + a_{k1}) a_{pq})$
= $(-1)^{1+j} \sum \prod r_k a_{pq}$

where each $\prod r_k a_{pq}$ is a product of exactly n - 1 numbers taken from r_k, a_{pq} for $k, p, q = 1, \ldots, n$.

For the case i = j, we assume without loss of generality that i = j = 1. Write

$$(I-A)_{11} = \begin{vmatrix} r_2 + \sum_{s \neq 2} a_{2s} & -a_{23} & \cdots & -a_{2n} \\ -a_{32} & r_3 + \sum_{s \neq 3} a_{3s} & \cdots & -a_{3n} \\ \cdots & \cdots & \cdots & \cdots & \cdots & \cdots \\ -a_{n2} & -a_{n3} & \cdots & r_n + \sum_{s \neq n} a_{ns} \end{vmatrix}.$$

Adding columns two through n - 1 to column one yields

$$B = \begin{cases} r_2 + a_{21} & -a_{23} & \cdots & -a_{2n} \\ r_3 + a_{31} & r_3 + \sum_{s \neq 3} a_{3s} & \cdots & -a_{3n} \\ \vdots & \vdots & \vdots & \vdots \\ r_n + a_{n1} & -a_{n3} & \cdots & r_n + \sum_{s \neq n} a_{ns} \end{cases}$$

Expanding the determinant about the first column of B and noting that $B_{s1} = [(I - A)_{11}]_{s+1,2}$ yields, by the induction hypothesis, that

$$|(I - A)_{11}| = \sum_{s=1}^{n-1} (-1)^{s+1} (r_{s+1} + a_{s+1,1}) |[(I - A)_{11}]_{s+1,2}|$$

= $\sum_{s=1}^{n-1} (-1)^{s+1} (r_{s+1} + a_{s+1,1}) ((-1)^{s+1} (\sum \prod (r_k + a_{k1}) a_{pq}))$
= $\sum \prod r_k a_{pq}$

where each $\prod r_k a_{pq}$ is a product of exactly n-1 numbers taken from r_k , a_{pq} for k, p, q = 1, ..., n.

This theorem is now applied to yield our first perturbation result. This result estimates the error in computing $(I - A)^{-1}$, when it exists.

COROLLARY 1. Let A and \hat{A} be $n \times n$ nonnegative matrices with row sums less than or equal to one and having spectral radius less than 1. Set $B = (I - A)^{-1}$ and $\hat{B} = (I - \hat{A})^{-1}$. If

(1) $a_{ij} \leq \hat{\theta} \hat{a}_{ij}, r_k \leq \hat{\theta} \hat{r}_k$ and (2) $\hat{a}_{ij} \leq \theta a_{ij}, \hat{r}_k \leq \theta r_k$ then, for all $b_{ij} \neq 0$, $(\hat{b}_{ij} - b_{ij})/b_{ij} \leq \theta^{n-1}\hat{\theta}^n - 1$.

PROOF. Note that

$$b_{ij} = \frac{(-1)^{i+j} |(I-A)_{ji}|}{|(I-A)|}$$
 and $\hat{b}_{ij} = \frac{(-1)^{i+j} |(I-\hat{A})_{ji}|}{|(I-\hat{A})|}$.

Thus, if $b_{ij} \neq 0$, application of the theorem yields that

$$\frac{\hat{b}_{ij}}{b_{ij}} = \frac{|(I-\hat{A})_{ji}| |I-A|}{|(I-A)_{ji}| |I-\hat{A}|} < \frac{\theta^{n-1}|(I-A)_{ji}|\hat{\theta}^n|I-\hat{A}|}{|(I-A)_{ji}| |I-\hat{A}|} < \theta^{n-1}\hat{\theta}^n.$$

Hence,

$$\frac{\hat{b}_{ij}-b_{ij}}{b_{ij}} \leq \theta^{n-1}\hat{\theta}^n - 1. \quad \Box$$

A second perturbation result estimates the error in solving Leontif's open economic model.

COROLLARY 2. Let A and \hat{A} be $n \times n$ nonnegative matrices having largest row sums less than or equal to one and having spectral radii less than one. Let b and \hat{b} be $1 \times n$ nonnegative vectors with x(I - A) = b and $\hat{x}(I - \hat{A}) = \hat{b}$. If

(1) $a_{ij} \leq \hat{\theta} \hat{a}_{ij}, r_k(A) \leq \hat{\theta} r_k(\hat{A}), b_i \leq \hat{\theta} \hat{b}_i \text{ and}$ (2) $\hat{a}_{ij} < \theta a_{ij}, r_k(\hat{A}) \leq \theta r_k(A), \hat{b}_i \leq \theta b_i$ then, for all $x_i \neq 0, (\hat{x}_i - x_i)/x_i \leq (\theta \hat{\theta})^n - 1$.

PROOF. Note first that $x = (I - A)^{-1}b$ and $\hat{x} = (I - \hat{A})^{-1}\hat{b}$. Then, if $x_i \neq 0$, application of the theorem yields that

$$\frac{\hat{x}_i}{x_i} = \frac{\sum_r (-1)^{i+r} |(I - \hat{A})_{ri}|\hat{b}_r|I - A|}{|I - \hat{A}|\sum_r (-1)^{i+r}|(I - A)_{ri}|b_r} \le (\theta\hat{\theta})^n.$$

Hence

$$\frac{\hat{x}_i - x_i}{x_i} < (\theta \hat{\theta})^n - 1. \quad \Box$$

The last perturbation result estimates the error in computing stochastic eigenvectors for stochastic matrices.

COROLLARY 3. Let A and \hat{A} be $n \times n$ irreducible stochastic matrices. Suppose α and $\hat{\alpha}$ are stochastic eigenvectors, belonging to one, for A and \hat{A} respectively. If (1) $a_{ii} \leq \hat{\theta} \hat{a}_{ii}$ and (2) $\hat{a}_{ii} \leq \theta a_{ii}$ then $(\hat{\alpha}_i - \alpha_i)/\alpha_i < (\theta \hat{\theta})^{n-1} - 1$.

PROOF. First note that the Perron-Frobenius theory [1] gives that if α and $\hat{\alpha}$ are stochastic eigenvectors, belonging to one, for A and \hat{A} respectively, then α and $\hat{\alpha}$ are the unique solutions to

$$\alpha(I-A) = 0$$
 with $\sum \alpha_i = 1$ and $\hat{\alpha}(I-\hat{A}) = 0$ with $\sum \hat{\alpha}_i = 1$

Further, by the Perron-Frobenius theory, rank $(I - A) = \operatorname{rank}(I - \hat{A}) = n$ - 1 with the first n - 1 columns of both I - A and $I - \hat{A}$ being linearly independent. Hence, the above equations are equivalent to

$$\alpha[(I-A)_n e] = e_n \text{ and } \hat{\alpha}[(I-\hat{A})_n e] = e_n$$

where $(I - A)_n$ and $(I - \hat{A})_n$ are obtained by deleting the *n*th column of (I - A) and $(I - \hat{A})$ respectively. Further e_i is the (0, 1)-vector having its only nonzero entry in the *i*th position and $e = e_1 + \cdots + e_n$. Now, by Cramer's rule

$$\alpha_i = (-1)^{i+n} \det([(I-A)_n e]_{in})/\det[(I-A)_n e]$$

and

$$\hat{\alpha}_i = (-1)^{i+n} \det\left(\left[(I-\hat{A})_n e\right]_{in}\right) / \det\left[(I-\hat{A})_n e\right].$$

Thus

$$\frac{\hat{\alpha}_i}{\alpha_i} = \frac{\det([(I-\hat{A})_n e]_{in})\det[(I-A)_n e]}{\det([(I-A)_n e]_{in})\det[(I-\hat{A})_n e]}$$

Noting that $\det[(I - \hat{A})_n e]_{in} = \det[(I - \hat{A})_{in}]$ and $\det[(I - A)_n e]_{in} = \det[(I - A)_{in}]$ and expanding $\det[(I - A)_n e]$ and $\det[(I - \hat{A})_n e]$ about the last column yields, by applying the theorem, that

$$\frac{\hat{\alpha}_i}{\alpha_i} < \frac{\left(\theta^{n-1} \sum \prod r_k a_{pq}\right) \left(\hat{\theta}^{n-1} \sum \prod \hat{r}_k \hat{a}_{pq}\right)}{\left(\sum \prod r_k a_{pq}\right) \left(\sum \prod \hat{r}_k \hat{a}_{pq}\right)} < (\theta\hat{\theta})^{n-1}.$$

Hence,

$$\frac{\hat{\alpha}_i - \alpha_i}{\alpha_i} \le (\theta\hat{\theta})^{n-1} - 1. \quad \Box$$

References

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