

## THE GENERALIZED INVERSE OF A NONNEGATIVE MATRIX

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ABSTRACT. Necessary and sufficient conditions are given in order that a nonnegative matrix have a nonnegative Moore-Penrose generalized inverse.

1. **Introduction.** Let  $A$  be an arbitrary  $m \times n$  real matrix. Then the Moore-Penrose generalized inverse of  $A$  is the unique  $n \times m$  real matrix  $A^+$  satisfying the equations

$$A = AA^+A, \quad A^+ = A^+AA^+, \\ (AA^+)^T = AA^+, \quad \text{and} \quad (A^+A)^T = A^+A.$$

The properties and applications of  $A^+$  are described in a number of papers including Penrose [7], [8], Ben-Israel and Charnes [1], Cline [2], and Greville [6]. The main value of the generalized inverse, both conceptually and practically, is that it provides a solution to the following least squares problem: *Of all the vectors  $x$  which minimize  $\|b - Ax\|$ , which has the smallest  $\|x\|^2$ ?* The solution is  $x = A^+b$ .

If  $A$  is nonnegative (written  $A \geq 0$ ), that is, if the components of  $A$  are all nonnegative real numbers, then  $A^+$  is not necessarily nonnegative. In particular, if  $A \geq 0$  is square and nonsingular, then  $A^+ = A^{-1} \geq 0$  if and only if  $A$  is monomial, i.e.,  $A$  can be expressed as a product of a diagonal matrix and a permutation matrix, so that  $A^{-1} = DA^T$  for some diagonal matrix  $D$  with positive diagonal elements. The main purpose of this paper is to give necessary and sufficient conditions on  $A \geq 0$  in order that  $A^+ \geq 0$ . Certain properties of such nonnegative matrices are then derived.

2. **Results.** In order to simplify the discussion to follow, it will be convenient to introduce a canonical form for a nonnegative symmetric

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idempotent matrix. Flor [5] has shown that if  $E$  is any nonnegative idempotent matrix of rank  $r$ , then there exists a permutation matrix  $P$  such that

$$PEP^T = \begin{pmatrix} J & JB & 0 & 0 \\ 0 & 0 & 0 & 0 \\ AJ & AJB & 0 & 0 \\ 0 & 0 & 0 & 0 \end{pmatrix}$$

where  $A$  and  $B$  are arbitrary nonnegative matrices of appropriate sizes and

$$J = \begin{pmatrix} J_1 & & & 0 \\ & \cdot & & \\ & & \cdot & \\ 0 & & & J_r \end{pmatrix}$$

with each  $J_r$  a nonnegative idempotent matrix of rank 1. This gives the following lemma.

LEMMA 1. *Let  $E \geq 0$  be a symmetric idempotent matrix of rank  $r$  with  $q$  nonzero rows. Then there exists integers  $\lambda_1, \dots, \lambda_r$  and a permutation matrix  $P$  such that  $q = \lambda_1 + \dots + \lambda_r$  and such that  $PEP^T$  has the form*

$$(1) \quad PEP^T = \left( \begin{array}{ccc|c} J_1 & & 0 & \\ & \cdot & & 0 \\ & & \cdot & \\ 0 & & & J_r \\ \hline & & & 0 \end{array} \right)$$

where each  $J_i$  is a  $\lambda_i \times \lambda_i$  positive idempotent matrix of rank 1.

The main result is given next. The theorem characterizes  $A \geq 0$  so that  $A^+ \geq 0$ , and its proof indicates a method by which such an  $A^+$  can be constructed readily.

THEOREM 1. *Let  $A$  be an  $m \times n$  nonnegative matrix of rank  $r$ . Then the following statements are equivalent.*

- (i)  $A^+$  is nonnegative.

(ii) There exists a permutation matrix  $P$  such that  $PA$  has the form<sup>2</sup>

$$(2) \quad PA = \begin{pmatrix} B_1 \\ \cdot \\ \cdot \\ \cdot \\ B_r \\ 0 \end{pmatrix}$$

where each  $B_i$  has rank 1 and where the rows of  $B_i$  are orthogonal to the rows of  $B_j$  whenever  $i \neq j$ .

(iii)  $A^+ = DA^T$  for some diagonal matrix  $D$  with positive diagonal elements.

PROOF. Suppose (i) holds so that  $A, A^+ \geq 0$ . Since  $E = AA^+$  is a symmetric idempotent, there exists a permutation matrix  $P$  so that  $K = PEP^T$  has the form (1). Let  $B = PA$ . Then  $B^+ = A^+P^T$ ,  $BB^+ = K$ ,  $KB = B$ , and  $B^+K = B^+$ . Now  $B$  can be partitioned into the form (2), where  $r$  is the rank of  $A$  and where each  $B_i$ ,  $1 \leq i \leq r$ , is a  $\lambda_i \times n$  matrix with no zero rows, since  $A$  and  $B$  have the same number of nonzero rows. It remains to show each  $B_i$  has rank 1 and  $B_iB_j^T = 0$ , for  $1 \leq i \neq j \leq r$ . Let  $C = B^+$ . Then  $C$  can be partitioned into the form

$$C = (C_1, \dots, C_r, 0)$$

where, for  $1 \leq i \leq r$ ,  $C_i$  is an  $n \times \lambda_i$  matrix with no zero columns. Moreover, since  $CB$  is symmetric, a column of  $B$  is nonzero if and only if the corresponding row of  $C$  is nonzero. Now  $KB = B$  implies that  $J_iB_i = B_i$ , so that  $B_i$  has rank 1, for  $1 \leq i \leq r$ . It remains to show that the rows of  $B_i$  are orthogonal to the rows of  $B_j$  for  $i \neq j$ . Since  $BC = K$  has the form (1),

$$\begin{aligned} B_iC_j &= J_i, \quad \text{if } i = j, \quad \text{and} \\ &= 0, \quad \text{if } i \neq j, \end{aligned}$$

for  $1 \leq i, j \leq r$ . Suppose the  $l$ th column of  $B_i$  is nonzero. Then  $B_iC_k = 0$  for  $k \neq i$  implies that the  $l$ th row of  $C_k$  is zero. However, since the  $l$ th row of  $C$  is nonzero, the  $l$ th row of  $C_i$  is nonzero. In this case, the  $l$ th column of  $B_k$  is zero for all  $k \neq i$ , since  $B_kC_i = 0$ . Thus

$$B_iB_j^T = 0 \quad \text{for all } 1 \leq i \neq j \leq r,$$

and (ii) is established.

<sup>2</sup> Note that the zero block may not be present.

Now assuming (ii) holds, let  $B = PA$  have the form (2). Then for  $1 \leq i \leq r$ , there exist column vectors  $x_i, y_i$  such that  $B_i = x_i y_i^T$ . Furthermore,  $B_i^+$  is the nonnegative matrix

$$B_i^+ = (\|x_i\|^2 \|y_i\|^2)^{-1} B_i^T$$

and moreover  $B^+ = (B_1^+, \dots, B_r^+, 0)$ , since  $B_i B_j^T = 0$  for  $i \neq j$ . In particular then,  $B^+ = DB^T$  where  $D$  is a diagonal matrix with positive diagonal elements and thus  $A^+ = DA^T$ , yielding (iii).

Clearly (iii) implies (i) so the proof is complete.

The next theorem considers doubly stochastic matrices, that is, square matrices  $B \geq 0$  whose row sums and column sums are 1. The matrix  $A \geq 0$  is said to be diagonally equivalent to a doubly stochastic matrix if there exist diagonal matrices  $D_1$  and  $D_2$  such that  $D_1 A D_2$  is doubly stochastic. Classes of nonnegative matrices with this property have been the subject of several recent papers (for example, see Djoković [4]). Part of the following theorem identifies another such class.

**THEOREM 2.** *Let  $A \geq 0$  be square with no zero rows or columns. If  $A^+ \geq 0$  then  $A$  is diagonally equivalent to a doubly stochastic matrix. Moreover, if  $A$  is doubly stochastic then  $A^+$  is doubly stochastic if and only if the equation  $A = AXA$  has a doubly stochastic solution, in which case  $A^+ = A^T$ .*

**PROOF.** The first statement follows since there exist permutation matrices  $P$  and  $Q$  such that

$$PAQ = \begin{pmatrix} B_1 & & & 0 \\ & \cdot & & \\ & & \cdot & \\ 0 & & & B_r \end{pmatrix}$$

where each  $B_i$  is a positive square matrix.

For the second statement note that a doubly stochastic idempotent matrix  $E$  is necessarily symmetric; for in particular, there exists a permutation matrix  $P$  such that  $PEP^T$  has the form (1), where each row and column is nonzero and where each  $J_i$  is a positive, idempotent doubly stochastic matrix of rank 1. Then each entry of  $J_i$  is  $1/\lambda_i$  so that  $PEP^T$  and, accordingly,  $E$  are symmetric matrices. This means that  $A^+$  is the only possible doubly stochastic solution to the equations  $A = AYA$  and  $Y = YAY$ , since  $AY$  and  $YA$  are symmetric and  $A^+$  is unique. Thus  $A^+$  is doubly stochastic if and only if  $A = AXA$  has a doubly stochastic solution, in which case  $A^+ = XAX$ , and so  $A^+ = A^T$  by Theorem 1.

The final result determines the singular values of  $A$  (i.e., the positive square roots of the nonzero eigenvalues of  $A^T A$ ) whenever  $A^+ \geq 0$ .

**THEOREM 3.** *Let  $A \geq 0$  be an  $m \times n$  real matrix with  $A^+ \geq 0$  and let  $PA$  have the form (2). Let  $\{x_i, y_i\}_{i=1}^r$  be column vectors so that  $B_i = x_i y_i^T$  for  $1 \leq i \leq r$ . Then the singular values of  $A$  are the numbers  $\|x_i\| \cdot \|y_i\|$ .*

**PROOF.** The eigenvalues of  $AA^T$  are the eigenvalues of  $BB^T$ . But these are the eigenvalues of the matrices  $B_i B_i^T$  for  $1 \leq i \leq r$ , that is, the numbers  $\|x_i\|^2 \cdot \|y_i\|^2$ .

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