

## REPRESENTATIONS FOR THE WEIGHTED MOORE-PENROSE INVERSE OF A PARTITIONED MATRIX\*

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### Abstract

The weighted Moore-Penrose inverse of a partitioned matrix  $A = (U V)$  is discussed. Representations for the weighted Moore-Penrose inverse of the matrix  $A$  are derived, which extend some earlier results.

### §1. Introduction

Various expressions for the generalized inverse have been developed by a number of authors. Greville [4] has developed a representation for the generalized inverse of a partitioned matrix  $A_k = (A_{k-1}, a_k)$  and presented a famous recursive method for computing the M-P inverse of  $A$ .

Wang and Chen<sup>[5]</sup> extended Greville's result to compute the weighted M-P inverse of  $A_k = (A_{k-1}, a_k)$ . The result is as follows: Let  $A_k$  be the submatrix of  $A$  consisting of the first  $k$  columns and  $A_k$  be partitioned as  $A_k = (A_{k-1}, a_k)$ . The matrix  $N_k \in C^{k \times k}$  is the leading principal submatrix of  $N$ , and  $N_k$  is partitioned as  $N_k = \begin{pmatrix} N_{k-1} & l_k \\ l_k^* & n_{kk} \end{pmatrix}$ . Let  $X_{k-1} = (A_{k-1})_{MN_{k-1}}^+$ ,  $X_k = (A_k)_{MN_k}^+$ ,  $d_k = X_{k-1}a_k$ , and  $c_k = a_k - A_{k-1}d_k = (I - A_{k-1}X_{k-1})a_k$ . Then

$$X_k = \begin{pmatrix} X_{k-1} - d_k b_k^* - (I - X_{k-1}A_{k-1})N_{k-1}^{-1}l_k b_k^* \\ b_k^* \end{pmatrix}, \quad (1.1)$$

where

$$b_k^* = \begin{cases} (c_k^* M c_k)^{-1} c_k^* M, & \text{if } c_k \neq 0, \\ \delta_k^{-1} (d_k^* N_{k-1} - l_k^*) X_{k-1}, & \text{if } c_k = 0, \end{cases} \quad (1.2)$$

and

$$\delta_k = n_{kk} + d_k^* N_{k-1} d_k - (d_k^* l_k + l_k^* d_k) - l_k^* (I - X_{k-1}A_{k-1}) N_{k-1}^{-1} l_k \quad (1.3)$$

is a positive real scalar.

Cline<sup>[3]</sup> discussed the M-P inverse of any matrix  $A$  partitioned as  $A = (U V)$ , in which  $U$  and  $V$  are submatrices, and presented an expression for the M-P inverse of  $A$  under some conditions (see [3], Theorem 1).

It is the purpose of this paper to develop representations for the weighted M-P inverse of a partitioned matrix  $A = (U V)$ . A more general result is given without any conditions.

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**Lemma 1.** *Let the columns of the matrices  $R$  and  $S$  consist of a basis for  $N(A^*)$  and  $N(A)$  respectively, then*

$$I - AA_{MN}^+ = M^{-1}R(R^*M^{-1}R)^{-1}R^*, \quad I - A_{MN}^+A = S(S^*NS)^{-1}S^*N.$$

**§2. Main Results**

We begin by combining expressions in (1.2) into a single expression. Since  $c_k$  is a single column vector,  $c_k \neq 0$  implies  $b_k^* = (c_k^*Mc_k)^{-1}c_k^*M = (c_k)_{M,\delta_k}^+$ , and thus  $(c_k)_{M,\delta_k}^+c_k = I$ . Further,  $c_k = 0$  implies  $(c_k)_{M,\delta_k}^+ = 0$ . Then we can rewrite  $b_k^*$  as

$$b_k^* = (c_k)_{M,\delta_k}^+ + [1 - (c_k)_{M,\delta_k}^+c_k]\delta_k^{-1}(d_k^*N_{k-1} - 1_k^*)X_{k-1}. \tag{2.1}$$

Now consider an arbitrary matrix  $A = (U \ V)$ , where  $U \in C^{m \times n_1}$  and  $V \in C^{m \times (n-n_1)}$ , the hermitian positive definite matrix  $N$  is partitioned as

$$N = \begin{pmatrix} N_1 & L \\ L^* & N_2 \end{pmatrix}, \tag{2.2}$$

where  $N_1 \in C^{n_1 \times n_1}$ . Corresponding to  $d_k, c_k$  and  $\delta_k$ , let

$$D = U_{MN_1}^+V, \tag{2.3}$$

$$C = (I - UU_{MN_1}^+)V, \tag{2.4}$$

$$K = N_2 + D^*N_1D - (D^*L + L^*D) - L^*(I - U_{MN_1}^+U)N_1^{-1}L. \tag{2.5}$$

We shall prove that  $K$  is hermitian positive definite.

Let the columns of  $W_1$  be consist of a basis for  $N(U)$ , and  $W_2 = \begin{pmatrix} W_1 & -D \\ 0 & I_{n_2} \end{pmatrix}$ , then  $W_2$  has full column rank, thus the matrix  $W_2^*NW_2$  is hermitian positive definite. Then the second diagonal block of  $(W_2^*NW_2)^{-1}$  is also hermitian positive definite, we can show that it is equal to  $K^{-1}$  by using Lemma 1. Hence  $K$  is hermitian positive definite. Then we have our main theorem.

**Theorem 1.** *Let  $A \in C^{m \times n}, A = (U \ V)$ , where  $U \in C^{m \times n_1}, n_1 < n, N$  be partitioned as (2.2),  $D, C$  and  $K$  be defined as before, then*

$$A_{MN}^+ = \begin{pmatrix} U_{MN_1}^+ - DH - (I - U_{MN_1}^+U)N_1^{-1}LH \\ H \end{pmatrix}, \tag{2.6}$$

where

$$H = C_{MK}^+ + (I - C_{MK}^+C)K^{-1}(D^*N_1 - L^*)U_{MN_1}^+. \tag{2.7}$$

*Proof.* Let the right hand side of (2.6) be  $X$ , then we can prove Theorem 1 by verifying

$$AXA = A, \quad XAX = X, \quad (MAX)^* = MAX, \quad (NXA)^* = NXA.$$

Omitted.