

## AN IMBEDDING METHOD FOR COMPUTING THE GENERALIZED INVERSES\*

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### Abstract

This paper deals with a system of ordinary differential equations with known conditions associated with a given matrix. By using analytical and computational methods, the generalized inverses of the given matrix can be determined. Among these are the weighted Moore-Penrose inverse, the Moore-Penrose inverse, the Drazin inverse and the group inverse. In particular, a new insight is provided into the finite algorithms for computing the generalized inverse and the inverse.

### §1. Introduction

In [1, 2], the imbedding method for nonlinear matrix eigenvalue problems and for computational linear algebra are presented.

In many engineering problems we must find the generalized inverses of a given matrix.

Let  $A \in C^{m \times n}$ . Throughout this paper, let  $M$  and  $N$  be positive definite matrices of order  $m$  and  $n$  respectively. Then, there is a unique matrix  $X \in C^{n \times m}$  satisfying

$$AXA = A, \quad XAX = X, \quad (MAX)^* = MAX, \quad (NXA)^* = NXA. \quad (1.1)$$

This  $X$  is called the weighted Moore-Penrose inverse of  $A$ , and is denoted by  $X = A_{MN}^+$ . In particular, when  $M = I_m, N = I_n$ , the matrix  $X$  that satisfies (1.1) is called the Moore-Penrose inverse of  $A$ , and is denoted by  $X = A^+$ , i.e.,  $A^+ = A_{I_m I_n}^+$ .

Let  $A \in C^{n \times n}$ . The smallest nonnegative integer  $k$  such that

$$\text{rank}(A^k) = \text{rank}(A^{k+1}) \quad (1.2)$$

is called the index of  $A$ , and is denoted by  $\text{Ind}(A)$ .

Let  $A \in C^{n \times n}$ . With  $\text{Ind}(A) = k$  and if  $X \in C^{n \times n}$  is such that

$$A^{k+1}X = A^k, \quad XAX = X, \quad AX = XA \quad (1.3)$$

then  $X$  is called the Drazin inverse of  $A$ , and is denoted by  $X = A_d$ . In particular, when  $\text{Ind}(A) = 1$ , the matrix  $X$  that satisfies (1.3) is called the group inverse of  $A$ , and is denoted by  $X = A^\#$ .

An imbedding method for the Moore-Penrose inverse is given in [3]. In this paper, the imbedding methods for the weighted Moore-Penrose inverse Moore-Penrose inverse, the Drazin inverse and the group inverse are presented, and these methods have a uniform formula.

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First, we show the generalized inverses can be characterized in terms of a limiting process. These expressions involve the inverse of the matrix  $B_t(z)$ , where  $B_t(z)$  is a matrix of  $z$ . Secondly, we show how this problem may be reduced to integrating a system of ordinary differential equations subject to initial conditions. In particular, a new insight is provided into a series of finite algorithms for computing the generalized inverses and the inverse in [4-6, 9].

### §2. Generalized Inverses as a Limit

In this section, we will show how the generalized inverses  $A^+, A_{MN}^+, A_d$  and  $A^\#$  can be characterized in terms of a limiting process respectively.

**Theorem 2.1.** *Let  $A \in C^{m \times n}$ ,  $\text{rank} A = r$ . Then*

$$A_{MN}^+ = \lim_{z \rightarrow 0} (N^{-1}A^*MA - zI)^{-1}N^{-1}A^*M \tag{2.1}$$

where  $z$  tends to zero through negative values.

*Proof.* From the  $(M, N)$ -singular value decomposition theorem<sup>[7]</sup>, there exists an  $M$ -unitary matrix  $U \in C^{m \times m}$  and an  $N^{-1}$ -unitary matrix  $V \in C^{n \times n}$  such that

$$A = U \begin{pmatrix} D & 0 \\ 0 & 0 \end{pmatrix} V^* \tag{2.2}$$

where

$$U^*MU = I_m, \quad V^*N^{-1}V = I_n, \tag{2.3}$$

$$D = \text{diag}(d_1, d_2, \dots, d_r), \quad d_i > 0, \quad i = 1, 2, \dots, r \tag{2.4}$$

and

$$A_{MN}^+ = N^{-1}V \begin{pmatrix} D^{-1} & 0 \\ 0 & 0 \end{pmatrix} U^*M. \tag{2.5}$$

Let

$$N^{-1/2}V = \tilde{V} = (v_1, v_2, \dots, v_n), \tag{2.6}$$

$$M^{1/2}U = \tilde{U} = (u_1, u_2, \dots, u_m), \tag{2.7}$$

then

$$\tilde{V}^* = \tilde{V}^{-1}, \quad \tilde{U}^* = \tilde{U}^{-1} \tag{2.8}$$

and

$$A_{MN}^+ = N^{-1/2} \left( \sum_{i=1}^r d_i^{-1} v_i u_i^* \right) M^{1/2}. \tag{2.9}$$

Since

$$N^{-1}A^*MA = N^{-1/2} \left( \sum_{i=1}^r d_i^2 v_i v_i^* \right) N^{1/2} \tag{2.10}$$