

DERIVATIONS OF GENERALIZED INVERSE USING CONTOUR INTEGRATION AND INTERPOLATION POLYNOMIAL

Mohammad Maruf Ahmed
Department of Mathematics and Natural Science
BRAC University, 66 Mohakhali C/A
Dhaka – 1212, Bangladesh
Email: maruf_1978@yahoo.com

ABSTRACT

This paper deals with a representation of Generalized inverse (g-inverse) by the contour integral formula that supports the four major properties of g-inverse. Here we have used Cauchy's integral formula. These are verified numerically. This paper also includes the derivation of g-inverse by using minimal polynomial. Here we express A^+ as a Lagrange-Sylvester interpolation polynomial in powers of A, A^* . Mathematica codes are used in these examples.

Key words: contour integral, minimal polynomial, inconsistent system, eigenvalue.

1. INTRODUCTION

The inverse of a nonsingular square matrix and its various properties are available in textbooks. It is stated that if a matrix A has an inverse, the matrix must be square and its determinant must be non-zero. Let us consider a system of linear equations $Ax=b$.

If A is an $n \times n$ non-singular matrix, the solution to the system in the equation $Ax=b$ exists and is unique and is given by $x = A^{-1}b$

However, there are cases where A is not a square matrix (i.e. rectangular matrix) and also the cases where A is $n \times n$ singular matrix; i.e when the linear equations are inconsistent. In these cases there may still be solution to the system and a unified theory to treat all cases may be desirable. One such theory involves the use of **generalized inverse of matrices**. The generalized inverse is also referred to as **Pseudo-inverse, Moore-Penrose inverses or simply g-inverse** with possible subscripting of the letter g.

Moore [3] first published the work on generalized inverses. **Penrose** [4] defined uniquely determined generalized inverse matrix and investigated some of its properties.

2. DEFINITION

Generalized inverse (g-inverse)

Let A be $m \times n$ matrix of rank $R(A) = r \leq \min(m, n)$. Then a *generalized inverse (g-inverse)* of A is an $n \times m$ matrix denoted by A^- such that $x = A^-b$ is a solution of both the consistent and inconsistent set of linear equations $Ax=b$.

A matrix A^- satisfying $AA^-A = A$ obviously coincides with A^{-1} when A^{-1} exists.

3. DIFFERENT CLASSES OF G- INVERSES

Let A be an $m \times n$ matrix over the complex field \mathbb{C} . Clearly, analogous results are obtainable when the matrices are defined over a real field.

Consider the following matrix equations:

$$(i) \quad AXA = A, \tag{1.1}$$

$$(ii) \quad XAX = X, \tag{1.2}$$

$$(iii) \quad (XA)^* = XA, \tag{1.3}$$

$$(iv) \quad (AX)^* = AX, \tag{1.4}$$

where $*$ denotes the conjugate transpose.

X is a **g-inverse** if equation (1.1) is satisfied and we denote $X = A^-$.

- (a) If (1.1) and (1.2) are satisfied then X is a **reflexive g-inverse** and we denote $X=A^r$.
- (b) If (1.1), (1.2) & (1.3) are satisfied then X is a **left weak g-inverse** and we denote $X=A^w$.
- (c) If (1.1), (1.2) & (1.4) are satisfied then X is **right weak g-inverse** and we denote $X=A^n$.
- (d) If (1.1), (1.2), (1.3) & (1.4) are all satisfied then we call X is **Pseudo -inverse** or (**Moore & Penrose generalized inverse**) and we denote $X=A^+$. It is also known as **M-P g-inverse**.

4. SPECIAL REPRESENTATION OF GENERALIZED INVERSE BY CONTOUR INTEGRATION

Here we will discuss two theorems that represent the generalized inverse using contour integral formula. Numerical examples will support our proof.

Theorem 1 If A is any $m \times n$ matrix such that $(AA^*)^{-1}$ exists, then

$$A^+ = \frac{1}{2\pi i} \int_C A^* (AA^* - Iz)^{-1} \frac{1}{z} dz$$

where C is a closed contour containing non-zero eigenvalues of AA^* but not containing the zero eigenvalue of AA^* in or on C .

Proof Let $X = \frac{1}{2\pi i} \int_C A^* (AA^* - Iz)^{-1} \frac{1}{z} dz$

Then we have to show that X satisfies the following four conditions:

- 1) $AXA = A$
- 2) $XAX = X$
- 3) $(AX)^* = AX$
- 4) $(XA)^* = XA$

$$\begin{aligned} (1) \quad AXA &= \frac{1}{2\pi i} \int_C AA^* (AA^* - Iz)^{-1} A \frac{1}{z} dz \\ &= \frac{1}{2\pi i} \int_C \{(AA^*)^{-1}\}^{-1} (AA^* - Iz)^{-1} A \frac{1}{z} dz \\ &= \frac{1}{2\pi i} \int_C \{(AA^* - Iz)(AA^*)^{-1}\}^{-1} A \frac{1}{z} dz \end{aligned}$$

$$\begin{aligned} &= \frac{1}{2\pi i} \int_C \{I - z(AA^*)^{-1}\}^{-1} A \frac{1}{z} dz \\ &= \frac{1}{2\pi i} \int_C \frac{f(z)}{(z - z_0)} dz \quad (\text{say}) \text{ where } z_0 = 0 \end{aligned}$$

$$\begin{aligned} \text{and } f(z) &= \{I - z(AA^*)^{-1}\}^{-1} A \\ &= f(z_0) \\ &= f(0) \\ &= (I - 0)^{-1} A \\ &= A \\ \therefore AXA &= A \end{aligned}$$

$$\begin{aligned} (2) \quad XAX &= \frac{1}{2\pi i} \int_C A^* (AA^* - Iz)^{-1} \frac{1}{z} dz A \frac{1}{2\pi i} \int_C A^* (AA^* - Iz)^{-1} \frac{1}{z} dz \\ &= \frac{1}{2\pi i} \int_C A^* (AA^* - Iz)^{-1} \frac{1}{z} dz I \\ &= XI \\ &= X \\ \therefore XAX &= X \end{aligned}$$

$$\begin{aligned} (3) \quad AX &= A \frac{1}{2\pi i} \int_C A^* (AA^* - Iz)^{-1} \frac{1}{z} dz = I \\ \therefore (AX)^* &= I^* = I = AX \end{aligned}$$

$$\begin{aligned} (4) \quad XA &= \frac{1}{2\pi i} \int_C A^* (AA^* - Iz)^{-1} \frac{1}{z} A dz \\ &= \frac{1}{2\pi i} \int_C \frac{f(z) dz}{(z - z_0)} \quad \text{where } z_0 = 0 \\ \text{and } f(z) &= A^* (AA^* - Iz)^{-1} A \\ &= f(z_0) \text{ (using Cauchy's integral formula)} \\ &= A^* (AA^*)^{-1} A \text{ which is hermitian.} \\ \therefore (XA)^* &= XA. \end{aligned}$$

¹ Suppose U is an open subset of the complex plane \mathbb{C} , and $f: U \rightarrow \mathbb{C}$ is a holomorphic function, and the disk $D = \{z : |z - z_0| \leq r\}$ is completely contained in U . Let C be the circle forming the boundary of D . Then we have for every a in the interior of D :

$$f(a) = \frac{1}{2\pi i} \oint_C \frac{f(z)}{z - a} dz$$

where the contour integral is to be taken counter-clockwise

Thus X satisfies the four conditions of M-P g-inverse. Hence $X = A^+$

So, we have $A^+ = \frac{1}{2\pi i} \int_C A^* (AA^* - Iz)^{-1} \frac{1}{z} dz$

Example Consider a complex matrix

$$A = \begin{pmatrix} 0 & -i \\ 1 & -2 \end{pmatrix}$$

$$\text{Then } A^* = \begin{pmatrix} 0 & 1 \\ i & -2 \end{pmatrix} \text{ and}$$

$$AA^* = \begin{pmatrix} 1 & 2i \\ -2i & 5 \end{pmatrix}$$

Now $\det[AA^*] = 1 \neq 0$. So $(AA^*)^{-1}$ exists and

$$(AA^*)^{-1} = \begin{pmatrix} 5 & -2i \\ 2i & 1 \end{pmatrix}$$

$$A^+ = \frac{1}{2\pi i} \int_C A^* (AA^* - Iz)^{-1} \frac{1}{z} dz$$

$$= \frac{1}{2\pi i} \int_C \frac{A^* (AA^* - Iz)^{-1}}{z} dz$$

$$= \frac{1}{2\pi i} \int_C \frac{f(z) dz}{z - z_0} \text{ (say) where } z_0 = 0$$

$$\text{and } f(z) = A^* (AA^* - Iz)^{-1}$$

$$= f(0)$$

$$= A^* (AA^*)^{-1}$$

$$\& (AA^*)^{-1} = \begin{pmatrix} 5 & -2i \\ 2i & 1 \end{pmatrix}$$

$$A^* = A^* (AA^*)^{-1}$$

$$= \begin{pmatrix} 0 & 1 \\ i & -2 \end{pmatrix} \begin{pmatrix} 5 & -2i \\ 2i & 1 \end{pmatrix}$$

$$= \begin{pmatrix} 2i & 1 \\ i & 0 \end{pmatrix}$$

Now we will verify the four conditions

$$(1) AA^+ A = A$$

$$(2) A^+ AA^+ = A^+$$

$$(3) (AA^+)^* = AA^+$$

$$(4) (A^+ A)^* = A^+ A$$

$$\begin{aligned} (1) AA^+ A &= \begin{pmatrix} 0 & -i \\ 1 & -2 \end{pmatrix} \begin{pmatrix} 2i & 1 \\ i & 0 \end{pmatrix} \begin{pmatrix} 0 & -i \\ 1 & -2 \end{pmatrix} \\ &= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & -i \\ 1 & -2 \end{pmatrix} \\ &= \begin{pmatrix} 0 & -i \\ 1 & -2 \end{pmatrix} \\ &= A \end{aligned}$$

$$\begin{aligned} (2) A^+ AA^+ &= \begin{pmatrix} 2i & 1 \\ i & 0 \end{pmatrix} \begin{pmatrix} 0 & -i \\ 1 & -2 \end{pmatrix} \begin{pmatrix} 2i & 1 \\ i & 0 \end{pmatrix} \\ &= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \begin{pmatrix} 2i & 1 \\ i & 0 \end{pmatrix} \\ &= \begin{pmatrix} 2i & 1 \\ i & 0 \end{pmatrix} \\ &= A \end{aligned}$$

$$\begin{aligned} (3) AA^+ &= \begin{pmatrix} 0 & -i \\ 1 & -2 \end{pmatrix} \begin{pmatrix} 2i & 1 \\ i & 0 \end{pmatrix} \\ &= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \end{aligned}$$

$$\therefore (AA^+)^* = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = AA^+$$

$$\begin{aligned} (4) A^+ A &= \begin{pmatrix} 2i & 1 \\ i & 0 \end{pmatrix} \begin{pmatrix} 0 & -i \\ 1 & -2 \end{pmatrix} \\ &= \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} \end{aligned}$$

$$\therefore (A^+ A)^* = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = A^+ A$$

Hence all the four conditions are satisfied.

Theorem 2 The M-P g-inverse of a $m \times n$ matrix A of complex numbers is given by the formula

$$A^+ = \int_0^\infty e^{-A^* t} A^* dt$$

5. AN INTERPOLATION POLYNOMIAL FOR THE M-P INVERSE

Proof Let $X = \int_0^\infty e^{-A^*At} A^* dt$, then we have to show that X satisfies the four condition of M-P g-inverse.

$$\begin{aligned} (1) AXA &= A \int_0^\infty e^{-A^*At} A^* A dt \\ &= -A[e^{-\infty} - e^0] \\ &= -A\left[\frac{1}{e^\infty} - I\right] \\ &= A \end{aligned}$$

$\therefore AXA = A$

$$\begin{aligned} (2) XAX &= \int_0^\infty e^{-A^*At} A^* dt A \int_0^\infty e^{-A^*At} A^* dt \\ &= -[e^{-\infty} - e^0] \int_0^\infty e^{-A^*At} A^* dt \\ &= \int_0^\infty e^{-A^*At} A^* dt \\ &= X \\ \therefore XAX &= X \end{aligned}$$

$$\begin{aligned} (3) AX &= A \int_0^\infty e^{-A^*At} A^* dt \\ &= -[e^{-\infty} - e^0] \\ &= I \\ \therefore (AX)^* &= AX \end{aligned}$$

$$\begin{aligned} (4) XA &= \int_0^\infty e^{-A^*At} A^* dt \\ &= -[e^{-\infty} - e^0] \\ &= I \\ \therefore (XA)^* &= XA \end{aligned}$$

Hence X satisfies the four conditions of M-P g-inverse

So, $A^+ = \int_0^\infty e^{-A^*At} A^* dt$

Here we express A^+ as a Lagrange-Sylvester interpolation polynomial in powers of A, A^* . For any complex square matrix A let $\sigma(A)$ denote the spectrum of A and $\psi(A)$ its *minimal polynomial* written as $\psi(\lambda) = \prod_{\mu \in \sigma(A)} (\lambda - \mu)^{v(\mu)}$, where the root $\mu \in \sigma(A)$ is *simple* if $v(\mu) = 1$ and *multiple* otherwise.

We intend to construct A^+ as the matrix function $f(A)$ corresponding to the scalar function $f(\lambda) = \lambda^+$ and consider only the case where $\lambda = 0 \in \sigma(A)$ as otherwise A is nonsingular.

Corollary If $\lambda = 0$ is a simple root, this effort to construct A^+ this way lead only to the satisfaction of (1.1) and (1.2).

We use therefore $A^+ = (A^*A)^+ A^*$ to construct A^+ by using the Lagrange-Sylvester interpolation polynomial to give explicit expressions associated with (A^*A) , as all the roots in $\sigma(A^*A)$ are simple.

$$A^*A = \sum_{\lambda \in \sigma(A^*A)} \lambda \frac{\prod_{\lambda \neq \mu \in \sigma(A^*A)} (A^*A - \mu I)}{\prod_{\lambda \neq \mu \in \sigma(A^*A)} (\lambda - \mu)} \quad (1.5)$$

so that $A^+ = \sum_{\lambda \in \sigma(A^*A)} \lambda^+ \left(\frac{\prod_{\lambda \neq \mu \in \sigma(A^*A)} (A^*A - \mu I)}{\prod_{\lambda \neq \mu \in \sigma(A^*A)} (\lambda - \mu)} \right) A^*$

We call (1.5) the *Lagrange-Sylvester interpolation polynomial* for A^+ .

Example Let $A = \begin{pmatrix} 1 & -1 \\ 0 & 0 \end{pmatrix}$

Then $A^*A = \begin{pmatrix} 1 & -1 \\ -1 & 1 \end{pmatrix}$

And $\psi(A^*A) = (A^*A)^2 - 2(A^*A)$ is the minimal polynomial.

Writing $\psi(\lambda) = \lambda(\lambda - 2)$

we have $(A^*A)^+ = \frac{1}{2} \frac{(A^*A)}{2} = \begin{pmatrix} 1/4 & -1/4 \\ -1/4 & 1/4 \end{pmatrix}$

and

$$A^+ = (A^*A)^+ A^* = \begin{pmatrix} 1/4 & -1/4 \\ -1/4 & 1/4 \end{pmatrix} \begin{pmatrix} 1 & 0 \\ -1 & 0 \end{pmatrix} = \begin{pmatrix} 1/2 & 0 \\ -1/2 & 0 \end{pmatrix}.$$

Example Let $A = \begin{pmatrix} 1 & 0 & 0 & -1 \\ -1 & 1 & 0 & 0 \\ 0 & -1 & 1 & 0 \\ 0 & 0 & -1 & 1 \end{pmatrix}$

$$A^*A = \begin{pmatrix} 2 & -1 & 0 & -1 \\ -1 & 2 & -1 & 0 \\ 0 & -1 & 2 & -1 \\ -1 & 0 & -1 & 2 \end{pmatrix}$$

The minimal polynomial of (A^*A) is $\psi(\lambda) = \lambda(\lambda - 2)(\lambda - 4)$

Therefore,

$$\begin{aligned} (A^*A)^+ &= \frac{1}{2} \left(\frac{A^*A(A^*A - 4I)}{2(2-4)} \right) + \frac{1}{4} \left(\frac{A^*A(A^*A - 2I)}{4(4-2)} \right) \\ &= \frac{14}{32} (A^*A) - \frac{3}{32} (A^*A)^2 \\ &= \frac{1}{16} \begin{pmatrix} 5 & -1 & -3 & -1 \\ -1 & 5 & -1 & -3 \\ -3 & -1 & 5 & -1 \\ -1 & -3 & -1 & 5 \end{pmatrix}. \end{aligned}$$

Hence

$$A^+ = \frac{1}{16} \begin{pmatrix} 5 & -1 & -3 & -1 \\ -1 & 5 & -1 & -3 \\ -3 & -1 & 5 & -1 \\ -1 & -3 & -1 & 5 \end{pmatrix} \begin{pmatrix} 1 & -1 & 0 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & 0 & 1 & -1 \\ -1 & 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} \frac{3}{8} & -\frac{3}{8} & -\frac{1}{8} & \frac{1}{8} \\ \frac{1}{8} & \frac{3}{8} & -\frac{3}{8} & -\frac{1}{8} \\ -\frac{1}{8} & \frac{1}{8} & \frac{3}{8} & -\frac{3}{8} \\ \frac{3}{8} & -\frac{1}{8} & -\frac{1}{8} & \frac{3}{8} \end{pmatrix}.$$

Mathematica code

a = {{1,0,0,-1},{-1,1,0,0},{0,-1,1,0},{0,0,-1,1}}
 {{1,0,0,-1},{-1,1,0,0},{0,-1,1,0},{0,0,-1,1}}

b = Transpose[a]
 {{1,-1,0,0},{0,1,-1,0},{0,0,1,-1},{-1,0,0,1}}

c = b.a
 {{2,-1,0,-1},{-1,2,-1,0},{0,-1,2,-1},{-1,0,-1,2}}

gi = PseudoInverse[c]
 {{5/16,-1/16,-3/16,-1/16},{-1/16,5/16,-1/16,-3/16},
 {-3/16,-1/16,5/16,-1/16},{-1/16,-3/16,-1/16,5/16}}

ginverse = gi.b
 {{3/8,-3/8,-1/8,1/8},{1/8,3/8,-3/8,-1/8},{-1/8,1/8,3/8,-3/8},
 {-3/8,-1/8,3/8,3/8}}

% // MatrixForm

$$\begin{bmatrix} \frac{3}{8} & -\frac{3}{8} & -\frac{1}{8} & \frac{1}{8} \\ \frac{1}{8} & \frac{3}{8} & -\frac{3}{8} & -\frac{1}{8} \\ -\frac{1}{8} & \frac{1}{8} & \frac{3}{8} & -\frac{3}{8} \\ -\frac{3}{8} & -\frac{1}{8} & -\frac{1}{8} & \frac{3}{8} \end{bmatrix}$$

6. REFERENCES

- [1] C. R. Rao and S. K. Mitra, Generalized inverse of matrices and its applications, Wiley, New York, 1971.
- [2] Graybill F. A., Matrices with applications in Statistics. The Wadsworth Statistics/Probability Series, Wadsworth International Group Belmont, California, 1969.
- [3] Moore E. H., 1920 on the reciprocal of the general Algebraic matrix, Abst. Bull. Amer. Math. Soc. pp.304-305.
- [4] Penrose R., 1955 A generalized inverses for matrices, Proc. Camb. Philo. Soc. 51. pp. 406-413.
- [5] Rohde C.A., Some results on generalized inverses, SIAM Review vol. 8. No. 2.