

# Generalized Inverse of a Matrix And Its Applications.

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# Generalized Inverse of a Matrix and Its Application



## Definition

Type(1)-Minimum norm

Type(2)-Least Squared

Type(3)-Minimum norm  
least squares Solution

## Inverse Matrix

A is  $m$  by  $m$  nonsingular matrix, Then there exists an inverse with the property

$$AA^{-1} = A^{-1}A = I$$

## Left(or Right) Inverse Matrix

A is  $m$  by  $n$  rectangular matrix, Then

Case i) rank  $n \leq m$ , there exists a left inverse

$$A_L^{-1} = (A^T A)^{-1} A^T$$

Case ii) rank  $m \leq n$ , there exists a right inverse

$$A_R^{-1} = A^T (A A^T)^{-1}$$

When such inverses **do not exist**,

can we represent a solution of the consistent equation  $Ax = y$  in the form  $x = Gy$ ?

**If such a  $G$  exists, we call it a generalized inverse.**

# Generalized Inverse of a Matrix and Its Application



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## Generalized Inverse

Three equivalent definitions of a G inverse

### *Definition 1.*

An  $n$  by  $m$  matrix  $G$  is said to be a  $g$  inverse of an  $m$  by  $n$  matrix  $A$  if  $x = Gy$  is a solution to the equation  $Ax = y$  For any  $y$  such that the equation  $Ax = y$  is consistent.

### *Definition 2.*

$G$  is a  $g$  inverse of  $A$  if  $AGA = A$

### *Definition 3.*

$G$  is a  $g$  inverse of  $A$  if  $AG$  is idempotent\* and  $R(AG) =$

# Generalized Inverse of a Matrix and Its Application



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## Generalized Inverse

A matrix  $G$  satisfying any one of these definitions is called a  $g$  inverse.

### *Theorem 1.*

Let  $Ax = y$  be a consistent equation and  $A^-$  be a  $g$  inverse of  $A$ .

- i) Then  $x = A^-y$  is a solution.
- ii) The class of all solutions is provided by  
$$x = A^-y + (I - A^-A)z, z \text{ arbitrary}$$

# Generalized Inverse of a Matrix and Its Application



Definition

Type(1)-Minimum norm

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## The g inverse for minimum norm solution

From Theorem 1-ii, the general solution is  $x = Gy + (I - GA)z$ .

So, There exists a choice of  $G$  independently of  $y$  such that the solution  $Gy$  has a minimum norm.

**Proof.** If such a  $G$  exists,

$$\|Gy\| \leq \|Gy + (I - GA)z\|, \text{ for all } z \text{ and } y$$

That is,

$$\|GAx\| \leq \|GAx + (I - GA)z\|$$

This implies  $(GAx, (I - GA)z) = 0$  for all  $z$  and  $x$ ,

$$\Rightarrow (GA)^T(I - GA) = 0 \text{ or } (GA)^T = (GA)$$

# Generalized Inverse of a Matrix and Its Application



Definition

Type(1)-Minimum norm

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least squares Solution

## The g inverse for minimum norm solution

*Theorem 2.*

Let  $Ax = y$  be a consistent equation and  $G$  be a g inverse of  $A$  such that  $Gy$  is a minimum norm solution.

i) Then,  $AGA = A$ ,  $(GA)^T = (GA)$

ii)  $AGA = A$ ,  $(GA)^T N = N(GA)$  if  $(y, x)_n = x^T N y$   
->Minimum N norm g inverse.

Especially *Theorem 2-ii*,

$G = N^{-1} A^T (A N^{-1} A^T)^{-1}$  satisfies the conditions of the theorem for any choice of the g inverse.

# Generalized Inverse of a Matrix and Its Application



Definition

Type(1)-Minimum norm

Type(2)-Least Squared

Type(3)-Minimum norm  
least squares Solution

## The g inverse for a least square solution

From Theorem 1-ii, the general solution is  $x = Gy + (I - GA)z$ .

So, There exists a least squares solution by minimizing  $\|Ax - y\|$ .

**Proof.** If such a  $G$  exists,

$$\|AGy - y\| \leq \|Ax - y\| \text{ for all } x, y$$

This implies  $(Aw, (AG - I)y) = 0$  for all  $y$  and  $x$ , and  $w = x - Gy$  implies  $A^T(AG - I) = 0$ .

$$\Rightarrow AG = (AG)^T, AGA = A$$

# Generalized Inverse of a Matrix and Its Application



Definition

Type(1)-Minimum norm

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Type(3)-Minimum norm  
least squares Solution

## The g inverse for minimum norm solution

***Theorem 3.***

Let  $Ax = y$  be a consistent equation and  $G$  be a g inverse of  $A$  such that  $Gy$  is a least squares solution

i) Then,  $AGA = A$ ,  $(AG)^T = (AG)$

ii)  $AGA = A$ ,  $(AG)^T M = M(GA)$  if  $(y, x)_m = x^T M y$   
->  $M$  least squares g inverse.

Especially ***Theorem 3-ii***,

$G = (A^T M A)^{-1} A^T M$  satisfies the conditions of the theorem for any choice of the g inverse.



# Generalized Inverse of a Matrix and Its Application



Definition

Type(1)-Minimum norm

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Type(3)-Minimum norm  
least squares Solution

## The g inverse for minimum norm least square solution

From Theorem 1-ii, the general solution is  $x = Gy + (I - GA)z$ .

So, There exists matrix  $G$  such that  $Gy$  has minimum norm in the class of least squares solutions.

**Proof.** If such a  $G$  exists,

$$\|Gy\|_n \leq \|\zeta\|_n, \{\zeta: \|A\zeta - y\|_m \leq \|Ax - y\|_m\} \text{ for all } x, y \\ \Rightarrow \|Gy\| \leq \|\zeta\|, \{\zeta: A^T A \zeta \leq A^T y\} \text{ for all } y$$

$$\text{This implies } A^T(I - AG) = 0 \text{ and } G^T(I - GA) = 0 \\ \Rightarrow AGA = A, (AG)^T = AG, GAG = G, (GA)^T = GA$$

# Generalized Inverse of a Matrix and Its Application



Definition

Type(1)-Minimum norm

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least squares Solution

## The g inverse for minimum norm solution

### *Theorem 4.*

Let  $Ax = y$  be a consistent equation and  $G$  be a  $g$  inverse of  $A$  such that  $Gy$  is a minimum norm solution.

i) Then,  $AGA = A$ ,  $(AG) = (AG)^T$ ,  $GAG = G$ ,  $(GA) = (GA)^T$

ii)  $AGA = A$ ,  $GAG = G$ ,  $(AG)^T M = MAG$ ,  $(GA)^T N = NG$

if  $(y, x)_m = x^T M y$ ,  $(y, x)_n = x^T N y$

->Minimum N norm M least squares  $g$  inverse.

Especially *Theorem 4-i*,

$G = A^T A (A^T A A^T A)^{-1} A^T$  satisfies the conditions of the theorem for any choice of the  $g$  inverse.