

*Lecture 8 – ECE 275A*

# **Moore-Penrose Conditions & SVD**

# Four Moore-Penrose Pseudoinverse Conditions

## MOORE-PENROSE THEOREM

Consider a linear operator  $A : \mathcal{X} \rightarrow \mathcal{Y}$ .

A linear operator  $M : \mathcal{Y} \rightarrow \mathcal{X}$  is the unique pseudoinverse of  $A$ ,  $M = A^+$ , if and only if it satisfies the

### Four Moore-Penrose Conditions:

$$\text{I. } (AM)^* = AM \quad \text{II. } (MA)^* = MA \quad \text{III. } AMA = A \quad \text{IV. } MAM = M$$



More simply we usually say that  $A^+$  is the unique p-inv of  $A$  iff

$$\text{I. } (AA^+)^* = AA^+ \quad \text{II. } (A^+A)^* = A^+A \quad \text{III. } AA^+A = A \quad \text{IV. } A^+AA^+ = A^+$$

- The theorem statement provides greater clarity because there we distinguish between a *candidate* p-inv  $M$  and the *true* p-inv  $A^+$ .
- If and only if the candidate p-inv  $M$  satisfies the four M-P conditions can we claim that indeed  $A^+ = M$ .

# ***Proof of the M-P Theorem***

First we reprise some basic facts that are consequences of the definitional properties of the pseudoinverse.

$$\text{FACT 1: } \mathcal{N}(A^+) = \mathcal{N}(A^*)$$

$$\text{FACT 2: } \mathcal{R}(A^+) = \mathcal{R}(A^*)$$

$$\text{FACT 3: } P_{\mathcal{R}(A)} = A A^+$$

$$\text{FACT 4: } P_{\mathcal{R}(A^*)} = A^+ A$$

We now proceed to prove two auxiliary theorems (Theorems A and B).

## ***Proof of the M-P Theorem – Cont.***

### **THEOREM A**

Let  $C : \mathcal{X} \rightarrow \mathcal{Y}$  and  $B : \mathcal{Y} \rightarrow \mathcal{Z}$  be linear mappings. It is readily shown that the composite mapping  $BC : \mathcal{X} \rightarrow \mathcal{Z}$  is a linear mapping where  $BC$  is defined by

$$(BC)x \triangleq B(Cx) \quad \forall x \in \mathcal{X}$$

Then

$$\mathcal{N}(B) \cap \mathcal{R}(C) = \{0\} \quad \Rightarrow \quad BC = 0 \text{ iff } C = 0$$



### **Proof**

$$\begin{aligned} BC = 0 &\Leftrightarrow (BC)x = 0 \quad \forall x && \text{(definition of zero operator)} \\ &\Leftrightarrow B(Cx) = 0 \quad \forall x && \text{(definition of composition)} \\ &\Leftrightarrow Cx = 0 \quad \forall x && \text{(because } Cx \in \mathcal{R}(C) \cap \mathcal{N}(B) = \{0\}, \forall x) \\ &\Leftrightarrow C = 0 && \text{(definition of zero operator)} \end{aligned}$$

**QED**

## ***Proof of the M-P Theorem – Cont.***

Theorem B covers the uniqueness part of the M-P Theorem.

**THEOREM B.**      **The pseudoinverse of  $A$  is unique.**      ■

**Proof.** Suppose that  $A^+$  and  $M$  are *both* p-inv's of  $A$ . Then Fact 3 gives  $P_{\mathcal{R}(A)} = AA^+ = AM$  or

$$A(A^+ - M) = 0$$

From Fact 2,  $\mathcal{R}(A^*) = \mathcal{R}(A^+) = \mathcal{R}(M)$  and as a consequence

$$\mathcal{R}(A^+ - M) \subset \mathcal{R}(A^*)$$

But  $\mathcal{R}(A^*) \perp \mathcal{N}(A)$  and therefore

$$\mathcal{R}(A^+ - M) \subset \mathcal{R}(A^*) = \mathcal{N}(A)^\perp$$

so that

$$\mathcal{N}(A) \cap \mathcal{R}(A^+ - M) = \{0\}$$

Therefore from Theorem A,

$$A^+ - M = 0 \Rightarrow A^+ = M$$

**QED**

## ***Proof of the M-P Theorem – Cont.***

**Necessity ('only if' part) of the M-P Conditions.** Assume that  $M = A^+$ .

**Necessity of M-P Conditions I & II.** Easy consequences of Facts 3 and 4.

**Necessity of M-P Condition III.** Note that Fact 4 and idempotency of a projection operator implies  $(A^+A)(A^+A) = A^+A$ , or

$$A^+ \underbrace{(AA^+A - A)}_{\triangleq C=A(A^+A-I)} = AC = 0$$

We have  $\mathcal{N}(A^+) = \mathcal{N}(A^*)$  (Fact 1) and  $\mathcal{R}(C) \subset \mathcal{R}(A) = \mathcal{N}(A^*)^\perp$ . Therefore  $\mathcal{N}(A^+) \cap \mathcal{R}(C) = \mathcal{N}(A^*) \cap \mathcal{R}(C) = \{0\}$  so that by Theorem A,  $C = 0$ .

**Necessity of M-P Condition IV.** Note that Fact 3 and idempotency of a projection operator implies  $(AA^+)(AA^+) = AA^+$ , or

$$A \underbrace{(A^+AA^+ - A^+)}_{\triangleq C=A^+(AA^+-I)} = AC = 0$$

With  $\mathcal{R}(A^+) = \mathcal{R}(A^*)$  (Fact 2) we have  $\mathcal{R}(C) \subset \mathcal{R}(A^+) = \mathcal{R}(A^*) = \mathcal{N}(A)^\perp$ . Therefore  $\mathcal{N}(A) \cap \mathcal{R}(C) = \{0\}$  so that by Theorem A,  $C = 0$ .

## ***Proof of the M-P Theorem – Cont.***

### **Sufficiency('if' part) of the M-P Conditions.**

Here we assume that  $M$  satisfies all four of the M-P conditions and then show as a consequence that  $M = A^+$ .

We do this using the following steps.

- (1) First prove that  $P_{\mathcal{R}(A)} = AM$  (proving that  $AM = AA^+$  via uniqueness of projection operators).
- (2) Prove that  $P_{\mathcal{R}(A^*)} = MA$  (proving that  $MA = A^+A$ ).
- (3) Finally, prove that as a consequence of (1) and (2),  $M = A^+$ .

# Proof of the M-P Theorem – Cont.

## Sufficiency – Cont.

### Step (1):

From M-P conditions 1 & 3,  $(AM)^* = AM$  and  $AM = (AMA)M = (AM)(AM)$ , showing that  $AM$  is an orthogonal projection operator. But onto what? Obviously onto a subspace of  $\mathcal{R}(A)$  as  $\mathcal{R}(AM) \subset \mathcal{R}(A)$ . However

$$\mathcal{R}(A) = A(\mathcal{X}) = AMA(\mathcal{X}) = AM(A(\mathcal{X})) = AM(\mathcal{R}(A)) \subset AM(\mathcal{Y}) = \mathcal{R}(AM) \subset \mathcal{R}(A)$$

yields the stronger statement that  $\mathcal{R}(AM) = \mathcal{R}(A)$ . Thus  $AM$  is the orthogonal projector onto the range of  $A$ ,  $P_{\mathcal{R}(A)} = AM = AA^+$ .

### Step (2):

From M-P conditions 2 & 3,  $(MA)^* = MA$  and  $MA = M(AMA) = (MA)(MA)$ , showing that  $MA$  is an orthogonal projection operator. Note that M-P conditions 3 and 2 imply  $A^* = (AMA)^* = A^*M^*A^*$  and  $MA = (MA)^* = A^*M^*$ . We have

$$\mathcal{R}(A^*) = A^*(\mathcal{Y}) = (A^*M^*A^*)(\mathcal{Y}) = (A^*M^*)(\mathcal{R}(A^*)) \subset \underbrace{(A^*M^*)(\mathcal{X})}_{=\mathcal{R}(A^*M^*)=\mathcal{R}(MA)} \subset \mathcal{R}(A^*)$$

showing that  $\mathcal{R}(MA) = \mathcal{R}(A^*)$ . Thus  $P_{\mathcal{R}(A^*)} = MA = A^+A$ .



## *Proof of the M-P Theorem – Cont.*

### **Sufficiency – Cont.**

#### **Step (3):**

Note that we have yet to use M-P condition 4,  $MAM = M$ . From M-P condition 4 and the result of Step (2) we have

$$MAM = P_{\mathcal{R}(A^*)}M = M$$

Obviously, then  $\mathcal{R}(M) \subset \mathcal{R}(A^*)$ , as can be rigorously shown via the subspace chain

$$\mathcal{R}(M) = M(\mathcal{Y}) = P_{\mathcal{R}(A^*)}M(\mathcal{Y}) = P_{\mathcal{R}(A^*)}(\mathcal{R}(M)) \subset P_{\mathcal{R}(A^*)}(\mathcal{X}) = \mathcal{R}(A^*)$$

Recalling that  $\mathcal{R}(A^+) = \mathcal{R}(A^*)$  (Fact 2), it therefore must be the case that

$$\mathcal{R}(M - A^+) \subset \mathcal{R}(A^*) = \mathcal{N}(A)^\perp$$

Using the result of Step (1),  $P_{\mathcal{R}(A)} = AM = AA^+$ , we have

$$A(M - A^+) = 0$$

with  $\mathcal{N}(A) \cap \mathcal{R}(M - A^+) = \{0\}$ . Therefore Theorem A yields  $M - A^+ = 0$ .

**QED**

## ***Proof of the M-P Theorem – Cont.***

Note the similarity of the latter developments in Step 3 to the proof of Theorem B. In fact, some thought should convince yourself that the latter part of Step 3 provides justification for the claim that the pseudoinverse is unique, so that Theorem B can be viewed as redundant to the proof of the M-P Theorem.

Theorem B was stated to introduce the student to the use of Theorem A (which played a key role in the proof of the M-P Theorem) and to present the uniqueness of the pseudoinverse as a key result in its own right.

# Singular Value Decomposition (SVD)

Henceforth, let us consider only Cartesian Hilbert spaces (i.e., spaces with identity metric matrices) and consider all finite dimensional operators to be represented as complex  $m \times n$  matrices,

$$A_{m \times n} : \mathcal{X} = \mathbb{C}^n \rightarrow \mathcal{Y} = \mathbb{C}^m$$

Note that  $A$  in general is non-square and therefore does not have a spectral representation (because eigenvalues and eigenvectors are not defined).

Even if  $A$  is square, it will in general have complex valued eigenvalues and non-orthogonal eigenvectors. Even worse, a general  $n \times n$  matrix can be *defective* and not have a full set of  $n$  eigenvectors, in which case  $A$  is not diagonalizable. In the latter case, one must use generalized eigenvectors to understand the spectral properties of the matrix (which is equivalent to placing the matrix in *Jordan Canonical Form*).

It is well known that if a square,  $n \times n$  complex matrix is self-adjoint (Hermitian),  $A = A^H$ , then its eigenvalues are all real and it has a full complement of  $n$  eigenvectors that can all be chosen to be orthonormal. In this case for eigenpairs  $(\lambda_i, x_i)$ ,  $i = 1, \dots, n$ ,  $A$  has a simple spectral representation given by an orthogonal transformation,

$$A = \lambda_1 x_1 x_1^H + \dots + \lambda_n x_n x_n^H = X \Lambda X^H$$

with  $\Lambda = \text{diag}(\lambda_1 \dots \lambda_n)$ , and  $X$  is unitary,  $X^H X = X X^H = I$ , where the columns of  $X$  are comprised of the orthonormal eigenvectors  $x_i$ . If in addition, a hermitian matrix  $A$  is positive-semidefinite, denoted as  $A \geq 0$ , then the eigenvalues are all non-negative, and all strictly positive if the matrix  $A$  is invertible (positive-definite,  $A > 0$ ).

## Singular Value Decomposition (SVD) – Cont.

Given an arbitrary (nonsquare) complex matrix operator  $A \in \mathbb{C}^{m \times n}$  we can ‘regularized’ its structural properties by ‘squaring’ it to produce a hermitian, positive-semidefinite matrix, and thereby exploit the very nice properties of hermitian, positive-semidefinite matrices mentioned above.

Because matrix multiplication is noncommutative, there are two ways to ‘square’  $A$  to form a hermitian, positive-semidefinite matrix, viz

$$AA^H \quad \text{and} \quad A^H A$$

It is an easy exercise to prove that both of these forms are hermitian, positive-semidefinite, recalling that a matrix  $M$  is defined to be positive-semidefinite,  $M \geq 0$ , if and only if the associated quadratic form  $\langle x, Mx \rangle = x^H Mx$  is real and positive-semidefinite

$$\langle x, Mx \rangle = x^H Mx \geq 0 \quad \forall x$$

Note that a sufficient condition for the quadratic form to be real is that  $M$  be hermitian,  $M = M^H$ . For the future, recall that a positive-semidefinite matrix  $M$  is positive-definite,  $M > 0$ , if in addition to the non-negativity property of the associated quadratic form we also have

$$\langle x, Mx \rangle = x^H Mx = 0 \quad \text{if and only if} \quad x = 0$$

## ***Singular Value Decomposition (SVD) – Cont.***

In Lecture 9 we will show that the eigenstructures of the well-behaved hermitian, positive-semidefinite ‘squares’  $A^H A$  and  $AA^H$  are captured in the Singular value Decomposition (SVD) introduced in Example 5 of Lecture 7. As noted in that example, knowledge of the SVD enables us to compute the pseudoinverse of  $A$  in the rank deficient case.

The SVD will also allow us to compute a variety of important quantities, including the rank of  $A$ , orthonormal bases for all four fundamental subspaces of  $A$ , orthogonal projection operators onto all four fundamental subspaces of the matrix operator  $A$ , the spectral norm of  $A$ , the Frobenius norm of  $A$ , and the condition number of  $A$ .

The SVD will also provide a simple geometrically intuitive understanding of the nature of  $A$  as an operator based on the action of  $A$  as mapping hyperspheres in  $\mathcal{R}(A^*)$  to hyperellipsoids in  $\mathcal{R}(A)$  in addition to the fact that  $A$  maps  $\mathcal{N}(A)$  to 0.