

## Chapter 18: "Matrix Inverses"

(116)

(Throughout this chapter,  $A$  is a square  $n \times n$  matrix).

Goal: Looking for a matrix  $A^{-1}$ , so that from  $A\vec{x} = \vec{b}$  we can deduce

$$\boxed{\vec{x} = A^{-1}\vec{b}}$$

Definition: If  $A$  is an  $n \times n$  matrix and  $B$  is an  $n \times n$  matrix such that

$$AB = BA = \mathbf{I}_n,$$

then  $B$  is called an "inverse" of  $A$  and is denoted by  $B = A^{-1}$ . In this case,  $A$  is called invertible.

Recall  $\mathbf{I}_n$  is the  $n \times n$  identity matrix.

Example:

(1)  $\mathbf{I}_n^{-1} = \mathbf{I}_n$ , since  $\mathbf{I}_n \mathbf{I}_n = \mathbf{I}_n \mathbf{I}_n = (\mathbf{I}_n)^2 = \mathbf{I}_n$ .

(2)  $A = \begin{pmatrix} 1 & 1 \\ 2 & 3 \end{pmatrix}$ ,  $B = \begin{pmatrix} 3 & -1 \\ -2 & 1 \end{pmatrix}$ . Then,

$$AB = \begin{pmatrix} 1 & 1 \\ 2 & 3 \end{pmatrix} \begin{pmatrix} 3 & -1 \\ -2 & 1 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \mathbf{I}_2,$$

$$BA = \begin{pmatrix} 3 & -1 \\ -2 & 1 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 2 & 3 \end{pmatrix} = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix} = \mathbf{I}_2,$$

So,  $B = A^{-1}$  and  $A = B^{-1}$ .

\* Finding the inverse of a 2x2 matrix:

Suppose  $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$  and  $ad - bc \neq 0$ , then

$$A^{-1} = \frac{1}{ad - bc} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

If  $ad - bc = 0$ , then  $A$  is not invertible, for instance  $A = \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}$  is not invertible, since if  $B = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$  were an inverse of  $A$ , then

$$\begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix} \begin{pmatrix} a & b \\ c & d \end{pmatrix} = \begin{pmatrix} a+c & b+d \\ a+c & b+d \end{pmatrix} \neq \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

for every  $a, b, c, d \in \mathbb{R}$ . So, no inverse exists.

- Recall: If  $A$  is invertible, then  $A\vec{x} = \vec{0}$  has a unique solution.

- Lemma: Suppose  $A$  is an invertible  $n \times n$  matrix. Then any linear system  $A\vec{x} = \vec{b}$ : ( $A$  is the  $n \times n$  coefficient matrix)

(1) is consistent, and

(2) has a unique solution.

Proof:

consistency:  $A(A^{-1}\vec{b}) = (AA^{-1})\vec{b} = I_n \vec{b} = \vec{b}$ , so  $\vec{x} = A^{-1}\vec{b}$  is a solution.

uniqueness: let's  $\vec{y}$  be another solution, then  $A\vec{y} = \vec{b} \Rightarrow A^{-1}A\vec{y} = A^{-1}\vec{b}$   
 $\Rightarrow \vec{y} = A^{-1}\vec{b}$ , which is the solution, so  $\vec{x} = \vec{y}$ .

□

\* Algebraic Properties of inverses:

If  $k \neq 0$  is a scalar and  $A$  and  $C$  are invertible  $n \times n$  matrices, then

$$(1) A^{-1}, \text{ and } (A^{-1})^{-1} = A,$$

$$(2) A^k, \text{ and } (A^k)^{-1} = (A^{-1})^k,$$

$$(3) A^T, \text{ and } (A^T)^{-1} = (A^{-1})^T,$$

$$(4) kA, \text{ and } (kA)^{-1} = \frac{1}{k} A^{-1},$$

$$(5) AC, \text{ and } (AC)^{-1} = C^{-1}A^{-1}, \text{ (note the order)}$$

are all invertible matrices. Moreover, if  $AC$  is invertible, then so are  $A$  and  $C$ .

Example:

~~•~~ Simplify  $(A^T B)^{-1} A^T$ :

$$(A^T B)^{-1} A^T = B^{-1} \underbrace{(A^T)^{-1} A^T}_I = B^{-1}.$$

~~•~~ Matrix Inversion Algorithm:

Recall that if  $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix}$ , then  $A$  is invertible  $\Leftrightarrow \det(A) \neq 0$ ,

and

$$A^{-1} = \frac{1}{\det(A)} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}, \quad \det(A) = ad - bc.$$

Goal: Solve  $AB = I_n$ , for the unknown matrix  $B$ . (If  $A$  is invertible)

(1) write B and I<sub>n</sub> as collections of column vectors:

$$B = [\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n], \quad I_n = [\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n],$$

$\vec{e}_i$  is the *i*th standard basis of  $\mathbb{R}^n$ .

(2) Multiply A & B:

$$AB = [A\vec{v}_1, A\vec{v}_2, \dots, A\vec{v}_n] = [\vec{e}_1, \vec{e}_2, \dots, \vec{e}_n] = I_n,$$

which gives rise to a number of matrix equations to solve,

$$A\vec{v}_1 = \vec{e}_1, \quad A\vec{v}_2 = \vec{e}_2, \quad \dots, \quad A\vec{v}_n = \vec{e}_n$$

(If A is invertible, then each of these equations will be consistent and will have a unique solution, which are the columns of  $A^{-1}$  (or B)).

(3) Row reduce each of the augmented matrices:

$$[A | \vec{e}_1], [A | \vec{e}_2], \dots, [A | \vec{e}_n].$$

and assuming A is invertible, the RREFs will be as

$$[I | \vec{v}_1], [I | \vec{v}_2], \dots, [I | \vec{v}_n],$$

which is exactly the row reduction of  $[A | \vec{e}_1 \vec{e}_2 \dots \vec{e}_n] = [A | I]$ ,

whose RREF will be  $[I | B] = [I | A^{-1}]$ .

Example:  $A = \begin{pmatrix} 2 & 7 & 1 \\ 1 & 4 & -1 \\ 1 & 3 & 0 \end{pmatrix}$ . Find  $A^{-1}$ ?

Apply the elementary row operations to the super-augmented matrix

$$[A | I] = \left( \begin{array}{ccc|ccc} 2 & 7 & 1 & 1 & 0 & 0 \\ 1 & 4 & -1 & 0 & 1 & 0 \\ 1 & 3 & 0 & 0 & 0 & 1 \end{array} \right) \quad \begin{array}{l} \sim \\ R_1 \leftrightarrow R_2 \end{array}$$

$$\left( \begin{array}{ccc|ccc} 1 & 4 & -1 & 1 & 0 & 0 \\ 2 & 7 & 1 & 0 & 1 & 0 \\ 1 & 3 & 0 & 0 & 0 & 1 \end{array} \right) \quad \begin{array}{l} \sim \\ -2R_1 + R_2 \rightarrow R_2 \\ -R_1 + R_3 \rightarrow R_3 \end{array}$$

$$\left( \begin{array}{ccc|ccc} 1 & 4 & -1 & 1 & 0 & 0 \\ 0 & -1 & 3 & 0 & 1 & -2 \\ 0 & -1 & 1 & 0 & -1 & 1 \end{array} \right) \quad \sim \dots \sim$$

$$\left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 1 & -\frac{3}{2} & -\frac{3}{2} & \frac{11}{2} \\ 0 & 1 & 0 & 0 & \frac{1}{2} & \frac{1}{2} & -\frac{3}{2} \\ 0 & 0 & 1 & 0 & -\frac{1}{2} & \frac{1}{2} & -\frac{1}{2} \end{array} \right),$$

Hence,  $A^{-1} = \frac{1}{2} \begin{pmatrix} -3 & -3 & 11 \\ 1 & 1 & -3 \\ 1 & -1 & -1 \end{pmatrix}$ . (check  $AB = BA = I_3$ .)

\* Lemma: If  $A$  and  $B$  are  $n \times n$  matrices such that  $AB = I_n$ , then  $BA = I_n$  as well.

\* Note: In the matrix inversion algorithm if  $A$  does not row reduce to the identity matrix in RREF, or if  $\text{rank}(A) < n$ , then  $A\vec{x} = \vec{b}$  won't have a unique solution, or in particular,  $A$  can't be invertible.

Theorem (18.1) ;

suppose  $A$  is an  $n \times n$  matrix and  $\text{rank}(A) = n$ . Then  $A$  is invertible and  $A^{-1}$  can be computed by the matrix inversion algorithm, or by row reducing

$$[A | I] \sim \dots \sim [I | A^{-1}]$$

If  $\text{rank}(A) < n$ , then  $A$  is not invertible.

Examples : Find  $A^{-1}$ , if

(1)  $A = \begin{pmatrix} 1 & 2 & 3 \\ 0 & 0 & 1 \\ 0 & 0 & 4 \end{pmatrix}$  :  $\text{rank}(A) < 3 \Rightarrow A$  is not invertible.  
(no leading one in the 2<sup>nd</sup> column)

(2)  $A = \begin{pmatrix} 1 & 2 & 3 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{pmatrix}$  :  $\text{rank}(A) = 3 \Rightarrow$

$$[A | I] = \left( \begin{array}{ccc|ccc} 1 & 2 & 3 & 1 & 0 & 0 \\ 0 & 1 & 1 & 0 & 1 & 0 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right) \sim \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 1 & -2 & -3 \\ 0 & 1 & 0 & 0 & 1 & -1 \\ 0 & 0 & 1 & 0 & 0 & 1 \end{array} \right)$$

$$\Rightarrow A^{-1} = \begin{pmatrix} 1 & -2 & -3 \\ 0 & 1 & -1 \\ 0 & 0 & 1 \end{pmatrix}$$

(3)  $A = \begin{pmatrix} 2 & 2 & 0 \\ 1 & 1 & 2 \\ 1 & -1 & 3 \end{pmatrix}$  :  $[A | I] = \left( \begin{array}{ccc|ccc} 2 & 2 & 0 & 1 & 0 & 0 \\ 1 & 1 & 2 & 0 & 1 & 0 \\ 1 & -1 & 3 & 0 & 0 & 1 \end{array} \right)$

$$\sim \left( \begin{array}{ccc|ccc} 1 & 1 & 2 & 0 & 1 & 0 \\ 2 & 2 & 0 & 1 & 0 & 0 \\ 1 & -1 & 3 & 0 & 0 & 1 \end{array} \right) \sim \dots \sim \left( \begin{array}{ccc|ccc} 1 & 0 & 0 & 5/8 & -3/4 & 1/2 \\ 0 & 1 & 0 & -1/8 & 3/4 & -1/2 \\ 0 & 0 & 1 & -1/4 & 1/2 & 0 \end{array} \right)$$

$$\Rightarrow A^{-1} = \frac{1}{8} \begin{pmatrix} 5 & -6 & 4 \\ -1 & 6 & -4 \\ -2 & 4 & 0 \end{pmatrix}$$

## Theorem (18.2) : Invertible Matrix Theorem

Let  $A$  be a  $n \times n$  matrix, then the followings are equivalent:

- (1)  $\text{rank}(A) = n$ ,
- (2)  $A\vec{x} = \vec{0} \Rightarrow \vec{x} = \vec{0}$  (trivial solution is the only solution),
- (3)  $A\vec{x} = \vec{b}$  is consistent for all  $\vec{b} \in \mathbb{R}^n$ ,
- (4) Every linear system  $A\vec{x} = \vec{b}$  has a unique solution,
- (5) The RREF of  $A$  is  $I_n$ ,
- (6)  $\text{Null}(A) = \{0\}$ ,
- (7)  $\text{Col}(A) = \mathbb{R}^n$ ,
- (8)  $\text{Row}(A) = \mathbb{R}^n$ ,
- (9)  $\text{rank}(A^T) = n$ ,
- (10) Columns of  $A$  are LI,
- (11) Rows of  $A$  are LI,
- (12) Columns of  $A$  span  $\mathbb{R}^n$ ,
- (13) Rows of  $A$  span  $\mathbb{R}^n$ ,
- (14) Columns of  $A$  form a basis for  $\mathbb{R}^n$ ,
- (15) Rows of  $A$  form a basis for  $\mathbb{R}^n$ ,
- (16)  $A$  is invertible,
- (17)  $A^T$  is invertible.

Examples: Check all the statements for  $A_1 = \begin{pmatrix} 2 & 2 & 0 \\ 1 & 1 & 2 \\ 1 & -1 & 3 \end{pmatrix}$ ,  $A_2 = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}$ .