

Math 270 Linear Algebra

Chapter 1 Linear Equations and Matrices

1.5 Special Types of Matrices and Partitioned Matrices

We have the following special $n \times n$ matrices:

diagonal matrix $a_{ij} = 0$ for $i \neq j$, i.e. the terms off the main diagonal are all zero

scalar matrix diagonal matrix whose diagonal elements are all equal

identity matrix diagonal matrix whose diagonal elements are all 1

Example 1

$$A = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} 3 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{bmatrix}, \quad I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

A, B, I_3 diagonal

B, I_3 scalar

I_3 identity

Note:

1. If A is $m \times n$, then $AI_n = A$ and $I_m A = A$.
2. If A is a scalar matrix, then $A = rI_n$ for some scalar r .

Proof: HW (Exercise 1)

To define the powers of a matrix:

Let p be a positive integer. Define

$$A^p = \underbrace{A \cdot A \cdot \dots \cdot A}_{p \text{ factors}}$$

If A is $n \times n$, define

$$A^0 = I_n.$$

For nonnegative integers p and q :

$$A^p A^q = A^{p+q}$$

$$(A^p)^q = A^{pq}$$

Proofs: HW (Exercise 8)

Note that $(AB)^p = A^p B^p$ does not hold for square matrices unless $AB = BA$. (HW: Exercise 9)

Still more on special $n \times n$ matrices:

upper triangular $a_{ij} = 0$ for $i > j$

lower triangular $a_{ij} = 0$ for $i < j$

Note A diagonal matrix is both upper and lower triangular.

symmetric matrix $A^T = A$

skew-symmetric matrix $A^T = -A$

Example 3

$$A = \begin{bmatrix} 1 & -1 & 0 \\ -1 & 5 & 2 \\ 0 & 2 & 4 \end{bmatrix} \text{ symmetric}$$

$$B = \begin{bmatrix} 0 & -1 & 1 \\ 1 & 0 & 2 \\ -1 & -2 & 0 \end{bmatrix} \text{ skew-symmetric}$$

Note:

1. A symmetric or skew-symmetric \Rightarrow A square.
2. A symmetric \Rightarrow the entries of A are symmetric wrt the main diagonal of A
3. A symmetric $\Leftrightarrow a_{ij} = a_{ji}$; A skew-symmetric $a_{ij} = -a_{ji}$.
4. A skew-symmetric \Rightarrow the entries on the main diagonal of A are all zero.
5. A is $n \times n \Rightarrow A = S + K$, where S is a symmetric matrix and K is a skew-symmetric matrix. Moreover, this decomposition is unique. (HW: Exercise 29)

A **submatrix** of a matrix A is obtained by crossing out some, but not all, of the rows or columns of A.

Example 4

$$A = \begin{bmatrix} 1 & 4 & 3 \\ -1 & 2 & 5 \\ 2 & 0 & 1 \\ 3 & -2 & 0 \end{bmatrix} \text{ submatrix } \begin{bmatrix} 4 & 3 \\ 2 & 5 \\ -2 & 0 \end{bmatrix}$$

partitioned matrix draw horizontal lines between rows and vertical lines between columns

Example 5

$$A = \left[\begin{array}{ccc|cc} a_{11} & a_{12} & a_{13} & a_{14} & a_{15} \\ a_{21} & a_{22} & a_{23} & a_{24} & a_{25} \\ a_{31} & a_{32} & a_{33} & a_{34} & a_{35} \end{array} \right] = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}$$

or

$$A = \left[\begin{array}{cc|c|cc} a_{11} & a_{12} & a_{13} & a_{14} & a_{15} \\ a_{21} & a_{22} & a_{23} & a_{24} & a_{25} \\ a_{31} & a_{32} & a_{33} & a_{34} & a_{35} \end{array} \right] = \begin{bmatrix} A_{11} & A_{12} & A_{13} \\ A_{21} & A_{22} & A_{23} \end{bmatrix}$$

Example 6 The augmented matrix of the system $A\vec{x} = \vec{b}$, $\left[A \mid \vec{b} \right]$, is a partitioned matrix.

Remarks

1. If A and B are both $m \times n$ matrices that are partitioned in the same way, then $A + B$ is obtained by simply adding the corresponding submatrices of A and B .
2. Similarly, if A is a partitioned matrix, then the scalar multiple cA is obtained by forming the scalar multiple of each submatrix.

Example 7 block multiplication

$$A = \left[\begin{array}{cc|cc} 0 & 1 & 2 & 0 \\ 3 & 2 & -1 & 2 \\ \hline 1 & 0 & 1 & 0 \\ 0 & 2 & -3 & 1 \end{array} \right] = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}$$

and

$$\text{Let } B = \left[\begin{array}{ccc|cc} 1 & 2 & -1 & 1 & 0 \\ 2 & 0 & 3 & 0 & -1 \\ \hline -1 & 2 & 0 & 1 & 3 \\ 1 & 1 & 2 & 0 & -1 \end{array} \right] = \begin{bmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{bmatrix}$$

Then

$$AB = \begin{bmatrix} A_{11}B_{11} + A_{12}B_{21} & A_{11}B_{12} + A_{12}B_{22} \\ A_{21}B_{11} + A_{22}B_{21} & A_{21}B_{12} + A_{22}B_{22} \end{bmatrix} = \dots$$

Example 8 adjoining matrices – reverse of partitioning

$$\text{Let } A = \begin{bmatrix} 1 \\ 2 \end{bmatrix}, B = [0 \quad -1 \quad 5], C = \begin{bmatrix} 0 & 2 & 1 \\ 1 & -1 & 3 \end{bmatrix}$$

Then

$$[A \quad C] = \left[\begin{array}{c|cc} 1 & 0 & 2 & 1 \\ 2 & 1 & -1 & 3 \end{array} \right]$$

$$[C \quad B] = \left[\begin{array}{ccc|c} 0 & 2 & 1 & \\ \hline 1 & -1 & 3 & \\ 0 & -1 & 5 & \end{array} \right]$$

$$[C^T \quad B^T] = \left[\begin{array}{cc|c} 0 & 1 & 0 \\ 2 & -1 & -1 \\ \hline 1 & 3 & 5 \end{array} \right]$$

Definition An $n \times n$ matrix A is called **nonsingular**, or **invertible**, if there exists an $n \times n$ matrix B s.t. $AB = BA = I_n$; B is called an **inverse** of A . Otherwise, A is called **singular** or **noninvertible**.

Remark It will be shown later in 2.2 that if $AB = I_n$ then $BA = I_n$. Thus, we need only verify that $AB = I_n$ to show that B is an inverse of A .

Example 9

$$A = \begin{bmatrix} 1 & 2 \\ 3 & -1 \end{bmatrix}, B = \begin{bmatrix} \frac{1}{7} & \frac{2}{7} \\ \frac{3}{7} & -\frac{1}{7} \end{bmatrix}$$

$$AB = BA = I_2$$

Theorem The inverse of a matrix, if it exists, is unique.

Proof:

Let B and C be inverses of A . Then

$$AB = BA = I_n \text{ and}$$

$$AC = CA = I_n$$

$$B = BI_n = B(AC) = (BA)C = I_n C = C$$

Thus, the inverse of A , if it exists, is unique.

Notation A^{-1} inverse of A ; $AA^{-1} = A^{-1}A = I$

Finding A^{-1}

Example 10

$$\text{Let } A = \begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix}, A^{-1} = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

$$\text{Then } AA^{-1} = \begin{bmatrix} a+3c & b+3d \\ 2a+4c & 2b+4d \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$a+3c=1 \quad (1)$$

$$b+3d=0 \quad (2)$$

$$\Rightarrow 2a+4c=0 \quad (3)$$

$$2b+4d=1 \quad (4)$$

$$\Rightarrow a = -2, b = 3/2, c = 1, d = -1/2$$

$$\text{Thus, } A^{-1} = \begin{bmatrix} -2 & 3/2 \\ 1 & -1/2 \end{bmatrix}$$

Example 11

$$A = \begin{bmatrix} 1 & -2 \\ -2 & 4 \end{bmatrix}, A^{-1} = \begin{bmatrix} a & b \\ c & d \end{bmatrix}$$

$$AA^{-1} = \begin{bmatrix} a-2c & b-2d \\ -2a+4c & -2b+4d \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$\Rightarrow \emptyset \Rightarrow A$ is singular

Properties of A^{-1}

Theorem If A and B are both nonsingular $n \times n$ matrices, then AB is nonsingular and $(AB)^{-1} = B^{-1}A^{-1}$.

Proof:

$$\begin{aligned}(AB)(B^{-1}A^{-1}) &= A(BB^{-1})A^{-1} \\ &= (AI_n)A^{-1} \\ &= AA^{-1} \\ &= I_n\end{aligned}$$

Similarly, $(B^{-1}A^{-1})(AB) = I_n$.

Thus, AB is nonsingular and by the uniqueness of the inverse of a matrix, $(AB)^{-1} = B^{-1}A^{-1}$.

Corollary If A_1, A_2, \dots, A_r are $n \times n$ nonsingular matrices, then $A_1 \cdot A_2 \cdot \dots \cdot A_r$ is nonsingular and $(A_1 \cdot A_2 \cdot \dots \cdot A_r)^{-1} = A_r^{-1}A_{r-1}^{-1} \dots A_1^{-1}$.

Proof: HW (Exercise 44)

Theorem If A is nonsingular, then A^{-1} is nonsingular and $(A^{-1})^{-1} = A$.

Proof: HW (Exercise 45)

Theorem If A is nonsingular, then A^T is nonsingular and $(A^T)^{-1} = (A^{-1})^T$.

Proof: $AA^{-1} = I_n$

$$\begin{aligned}(AA^{-1})^T &= I_n^T \\ (A^{-1})^T A^T &= I_n\end{aligned}$$

Similarly, $A^T (A^{-1})^T = I_n$.

Thus, $(A^T)^{-1} = (A^{-1})^T$.

Remarks

- If A is nonsingular, then
 - $AB = AC \Rightarrow B = C$ (HW: Exercise 50)
 - $AB = O_n \Rightarrow B = O_n$ (HW: Exercise 51)
- If A is a symmetric nonsingular matrix then A^{-1} is symmetric. (HW: Exercise 54)

Linear Systems and Inverses

A is $n \times n$
 $Ax = b$ (*) system of n linear equations in n unknowns

Suppose A is nonsingular. Then A^{-1} exists and

$$A^{-1}(Ax) = A^{-1}b$$

$$(A^{-1}A)x = A^{-1}b$$

$$I_n x = A^{-1}b$$

$x = A^{-1}b$ is a solution to (*) and the solution is unique

Thus, A is nonsingular $\Rightarrow x = A^{-1}b$ is the unique solution to (*).

Example 12 Use matrix A from Example 10

$$Ax = b$$

$$\begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix} x = \begin{bmatrix} 4 \\ 3 \end{bmatrix}$$

$$\begin{aligned} \text{Then } x &= \begin{bmatrix} 1 & 3 \\ 2 & 4 \end{bmatrix}^{-1} \begin{bmatrix} 4 \\ 3 \end{bmatrix} \\ &= \begin{bmatrix} -2 & 3/2 \\ 1 & -1/2 \end{bmatrix} \begin{bmatrix} 4 \\ 3 \end{bmatrix} \\ &= \begin{bmatrix} -7/2 \\ 5/2 \end{bmatrix} \end{aligned}$$

Example 13 If we change the RHS b to $b = \begin{bmatrix} 1 \\ 2 \end{bmatrix}$, then $x = \begin{bmatrix} -2 & 3/2 \\ 1 & -1/2 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$.