

# Matrix Algebra

Ch.8

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## 1 Matrix Algebra

### Matrix

**Definition 1** (Matrix). *Matrix is a rectangular array of numbers (scalars)*

Let  $a_{ij} \in \mathbb{R}$  or  $A_{ij} \in \mathbb{R}$  be the  $i$ th row and  $j$ th column element of matrix  $A$

**Definition 2** (Equal).

$$A = B \iff \begin{cases} \text{same size} \\ a_{ij} = b_{ij} \quad \forall i, j \end{cases}$$

### Addition, Subtraction

Let  $A, B$  be  $n \times k$  matrices and  $r \in \mathbb{R}$

**Definition 3** (Addition).

$$(A + B)_{ij} := a_{ij} + b_{ij} \quad \forall i, j$$

Important note: the first  $+$  and the second  $+$  are not same operators

**Definition 4** (Subtraction).

$$(A - B)_{ij} := a_{ij} - b_{ij} \quad \forall i, j$$

### Multiplications of Matrices

**Definition 5** (Scalar Multiplication).

$$(rA)_{ij} := rA_{ij} \quad \forall i, j$$

Let  $A$  be  $n \times k$  matrix and  $B$  be  $k \times m$  matrix. Then  $AB$  is  $n \times m$  matrix.

**Definition 6** (Matrix Multiplication).

$$(AB)_{ij} := A_{i1}B_{1j} + A_{i2}B_{2j} + \cdots + A_{ik}B_{kj} = \sum_{r=1}^k A_{ir}B_{rj}$$

For  $n \times n$  matrices, identity matrix  $I_n$  is a multiplicative identity.

$$AI = IA = A$$

## Laws of Matrix Algebra

### Laws of Matrix Algebra

$$(A + B) + C = A + (B + C) \quad (\text{Associative Law for Addition})$$

$$(AB)C = A(BC) \quad (\text{Associative Law for Multiplication})$$

$$A + B = B + A \quad (\text{Commutative Law for Addition})$$

$$A(B + C) = AB + AC \quad (\text{Distributive Law})$$

$$(A + B)C = AC + BC \quad (\text{Distributive Law})$$

Important Note:  $AB \neq BA$

## Transpose

**Definition 7** (Transpose).  $A^\top$  ( $n \times m$ ) is a transpose of  $A$  ( $m \times n$ ) if:

$$(A^\top)_{ij} := A_{ji} \quad \forall i, j$$

$$(A \pm B)^\top = A^\top \pm B^\top$$

$$(A^\top)^\top = A$$

$$(rA)^\top = rA^\top$$

$$(AB)^\top = B^\top A^\top \quad (\text{Theorem 8.1})$$

## 2 Special Kinds of Matrices

### Special Kinds of Matrices (1)

Suppose  $A$  is  $k \times n$  matrix. Then,

**Definition 8** (Special Kinds of Matrices (1)). •  $A$  is a square matrix if  $k = n$

- $A$  is a column matrix if  $n = 1$
- $A$  is a row matrix if  $k = 1$
- $A$  is a diagonal matrix if  $k = n$  and  $a_{ij} = 0 \quad \forall i \neq j$
- $A$  is a scalar matrix if  $A = tI_n$
- $A$  is an upper-triangular matrix if  $a_{ij} = 0 \quad \forall i > j$
- $A$  is a lower-triangular matrix if  $a_{ij} = 0 \quad \forall i < j$

### Special Kinds of Matrices (2)

**Definition 9** (Special Kinds of Matrices (2)). •  $A$  is a symmetric matrix if  $A$  is square matrix and  $a_{ij} = a_{ji} \quad \forall i, j$ . Or,  $A^T = A$

- $A$  is an Idempotent matrix if  $AA = A$
- $A$  is a permutation matrix if  $A$  is the result of  $I_n$  with  $ERO_1$  (row exchange)
- $A$  is a nonsingular matrix if  $\text{rank} A = \# \text{row} = \# \text{column}$

If a coefficient matrix of a system of linear equations is nonsingular, this system has only one solution  $\mathbf{x} = A^{-1}\mathbf{b}$

## 3 Elementary Matrices

### Elementary Matrix

Let  $E$  be an elementary matrix of some  $ERO$ s. Then,

**Theorem 1** (8.3).  $ERO$  with a matrix  $A$  is equivalent to  $EA$

**Theorem 2** (8.2). • Let  $E1_{ij}$  be the permutation matrix with interchanging  $R_i$  and  $R_j$  of  $I_n$ , then  $E1_{ij}$  is equivalent to  $ERO_1(i, j)$

- Let  $E2_{k,j,i}$  be the result of  $ERO_2(k, j, i)$  from  $I_n$ , then  $E2_{k,j,i}$  is equivalent to  $ERO_2(k, j, i)$
- Let  $E3_{k,i}$  be the result of  $ERO_3(k, i)$  from  $I_n$ , then  $E3_{k,i}$  is equivalent to  $ERO_3(k, i)$

## Elementary Matrix

**Definition 10** (Elementary Matrix).  $E_1, E_2, E_3$  are elementary matrices corresponding to their EROs

**Theorem 3** (8.4). Let  $A \in M_n$  (set of  $n \times n$  matrices),  $E_i \in EM$  (set of elementary matrices), and  $(R)REFM$  be the set of  $(R)REF$  matrices. Then:

$$\exists E_i \quad i = 1, 2, \dots, m \quad \text{s.t.} \quad \prod_{i=m}^1 E_i A \in (R)REFM$$

or

$$E_m E_{m-1} \cdots E_2 E_1 A \in (R)REFM$$

## 4 Algebra of Square Matrices

### Inverse of Matrices

Suppose  $A, B \in M_n$

**Definition 11** (Inverse, Invertible).  $B$  is (left, or right) inverse for  $A$  if:

$$\underbrace{AB}_{B: \text{Right inverse}} = \underbrace{BA}_{B: \text{Left inverse}} = I$$

$A$  is invertible if  $\exists B$

Notation:  $B = A^{-1}$

**Theorem 4** (8.5: Uniqueness of Inverse).  $A \in M_n$  can have at most one inverse. (left inverse = right inverse)

### Inverse Matrices and the Solution of Linear Systems

**Theorem 5** (8.6). For  $A \in M_n$ ,

$$\exists A^{-1} \Rightarrow \begin{cases} A \text{ is nonsingular} \\ \text{Unique solution of } A\mathbf{x} = \mathbf{b} \Rightarrow \mathbf{x} = A^{-1}\mathbf{b} \end{cases}$$

Proof: easy

**Theorem 6** (8.7: inverse of Th8.6).

$$A \in M_n \text{ is nonsingular} \Rightarrow \exists A^{-1}$$

Proof: difficult

## Calculation of Inverse Matrix

## Calculation of Inverse Matrix

$$[A|I] \xrightarrow{EROs} [I|A^{-1}]$$

If RREF is not  $I_n$ ,  $\nexists A^{-1}$

**Theorem 7** (8.8). Let  $A = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \in M_2$ .  $A$  is nonsingular iff  $ad - bc \neq 0$

For general case ( $A \in M_n$ ), see Ch.9

## Equivalent statements

**Theorem 8** (8.9). For  $A \in M_n$ , the following statements are equivalent

1.  $\exists A^{-1}$
2.  $A$  has right inverse
3.  $A$  has left inverse
4.  $A\mathbf{x} = \mathbf{b}$  has at least one solution for every  $\mathbf{b}$
5.  $A\mathbf{x} = \mathbf{b}$  has at most one solution for every  $\mathbf{b}$
6.  $A$  is nonsingular
7.  $\text{rank} A = n$

## Properties of Inverse Matrices and Their Exponentials

**Theorem 9** (8.10). If  $A, B \in M_n$  and  $\exists A^{-1}, B^{-1}$ ,

1.  $(A^{-1})^{-1} = A$
2.  $(A^T)^{-1} = (A^{-1})^T$
3.  $\exists (AB)^{-1} \wedge (AB)^{-1} = B^{-1}A^{-1}$

**Definition 12** (Matrix Exponential).

$$A^m := \prod_{i=1}^m A$$

$$A^{-m} := (A^{-1})^m$$

## Exponential Properties of Invertible Matrices

**Theorem 10** (8.11).

$$\exists A^{-1} \Rightarrow \begin{cases} \exists A^{-m} & \forall m \in \mathbb{N} \\ A^r A^s = A^{r+s} & \forall r, s \in \mathbb{N} \\ \forall r \in \mathbb{R} - \{0\}, \quad \exists (rA)^{-1} \wedge (rA)^{-1} = \frac{1}{r} A^{-1} \end{cases}$$

Important Note:  $(AB)^k \neq A^k B^k$

## 5 Partitioned Matrices

### Partitioned Matrices

Sometimes, matrix of matrices can be more convenient.

**Definition 13** (Submatrix, Partitioned matrix). • A submatrix of matrix  $A$  is a matrix obtained by deleting some  $R_i$  or  $C_j$

- A partitioned matrix is a matrix partitioned into submatrices by horizontal and/or vertical lines which extended along entire rows or columns of a matrix  $A$

### Partitioned Matrices

**Theorem 11** (8.15). Let  $A$  be a square matrix partitioned as

$$A = \begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix}$$

and  $A_{11}, A_{22} \in M_n$ . Then,

$$\exists A_{22}^{-1} \wedge \exists D^{-1} \wedge D = A_{11} - A_{12} A_{22}^{-1} A_{21}$$

$$\Rightarrow A^{-1} = \begin{pmatrix} D^{-1} & -D^{-1} A_{12} A_{22}^{-1} \\ -A_{22}^{-1} A_{21} D^{-1} & A_{22}^{-1} (I + A_{21} D^{-1} A_{12} A_{22}^{-1}) \end{pmatrix}$$