

L7

Lecture VII : §1.6 Algebraic properties of matrix operations
 §1.9 Matrix inverses

Recall : Last time we defined :

- addition, scalar multiplication of matrices
- dot product on n -dimensional column vectors ($\mathbb{R}^n = n \times 1$ matrices) & norms
- matrix multiplication : A of size $m \times n$ & B of size $n \times s$

size = # rows & # columns

(operate entry by entry)

($\mathbb{R}^n = n \times 1$ matrices) & norms

$$AB = \begin{bmatrix} a_{11} & \cdots & a_{1n} \\ a_{i1} & \cdots & a_{in} \\ a_{m1} & & a_{mn} \end{bmatrix}_{m \times n} \cdot \begin{bmatrix} b_{11} & \cdots & b_{1s} \\ b_{i1} & \cdots & b_{is} \\ b_{m1} & & b_{ms} \end{bmatrix}_{n \times s} = \begin{bmatrix} c_{11} & \cdots & c_{1s} \\ \vdots & \ddots & \vdots \\ c_{m1} & \cdots & c_{ms} \end{bmatrix}_{m \times s}$$

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$$(AB)_{ij} = a_{i1}b_{1j} + a_{i2}b_{2j} + \cdots + a_{in}b_{nj} = \sum_{k=1}^n a_{ik}b_{kj} \quad \text{for } 1 \leq i \leq m \quad 1 \leq j \leq s$$

$$(\text{jth col } AB = A \cdot \text{jth col } B) \\ (\text{in } \mathbb{R}^m) \quad \begin{matrix} m \times n \\ \text{in } \mathbb{R}^n \end{matrix} \quad \begin{matrix} \text{in } \mathbb{R}^n \\ \text{in } \mathbb{R}^s \end{matrix}$$

Example : $\begin{bmatrix} 3 & -1 \\ -3 & 5 \end{bmatrix}_{2 \times 1} \cdot \begin{bmatrix} -4 \\ 7 \end{bmatrix}_{1 \times 1} = \begin{bmatrix} 3(-4) + (-1) \cdot 7 \\ (-3)(-4) + 5 \cdot 7 \end{bmatrix} = \begin{bmatrix} -19 \\ 47 \end{bmatrix}_{2 \times 1} ; \quad \begin{bmatrix} -4 \\ 7 \end{bmatrix}_{2 \times 1} \begin{bmatrix} 3 & -1 \\ -3 & 5 \end{bmatrix}_{2 \times 2} \text{ not defined}$

Why this definition? Rules of substitution.

$$\text{Ex} : \begin{cases} x_1 = 3y_1 - y_2 \\ x_2 = -3y_1 + 5y_2 \end{cases} \quad \begin{cases} y_1 = -4z \\ y_2 = 7z \end{cases} \quad \rightsquigarrow \quad \begin{aligned} x_1 &= (3(-4) - 1 \cdot 7)z = -19z \\ x_2 &= (-3(-4) + 5 \cdot 7)z = 47z \end{aligned}$$

§1. Algebraic Properties:

Theorem 1 : A, B, C $m \times n$ matrices. Then :

- (1) $A + B = B + A$ [Commutative]
- (2) $(A + B) + C = A + (B + C)$ [Associative]
- (3) [Neutral Element] The zero matrix 0 of size $m \times n$ ($0_{ij} = 0$ for all i, j) satisfies $A + 0 = 0 + A = A$ for all $m \times n$ matrices A.
- (4) [Inverses] Given A, the matrix $P = (-A_{ij})_{i,j}$ satisfies $A + P - P + A = 0$

Why? Addition is done on each entry separately & these properties are true in \mathbb{R} .

Def : The identity matrix of size $n \times n$, written I_n , has 1's in the diagonal, and 0's elsewhere. $I_n = \begin{bmatrix} 1 & 0 & \cdots & 0 \\ 0 & 1 & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ 0 & 0 & \cdots & 1 \end{bmatrix}$ diagonal

$$\text{Ex} : I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \dots$$

Theorem 2: A of size $m \times n$, B of size $n \times s$, C of size $s \times q$.

$$(1) (\underbrace{AB})_{ij} = \underbrace{A(\underbrace{BC})}_{\substack{m \times s \\ s \times q}} \quad [\text{Associative}] \quad \begin{matrix} m \times n \\ n \times s \\ n \times q \end{matrix} \quad \begin{matrix} m \times q \text{ matrix} \end{matrix}$$

$$(2) \alpha, \beta \text{ scalars in } \mathbb{R}: \alpha(\beta A) = \alpha \beta A$$

$$(3) \alpha(AB) = (\alpha A)B = A(\alpha B)$$

$$(4) [\text{Neutral elements}]: A = I_m A = A I_n$$

$$\begin{matrix} m \times m & m \times n & m \times n & n \times n \end{matrix}$$

Proof: Compute expressions on each side & check whether matrices of the same size with the same entries.

$$(1) (AB)_{ij} = \sum_{k=1}^n A_{ik} B_{kj}$$

$$\begin{aligned} ((AB)C)_{il} &= \sum_{j=1}^s (AB)_{ij} C_{jl} = \sum_{j=1}^s \left(\sum_{k=1}^n A_{ik} B_{kj} \right) C_{je} \\ &= \sum_{j=1}^s \sum_{k=1}^n A_{ik} B_{kj} C_{je} \xrightarrow{\text{distribute}} \sum_{k=1}^n \sum_{j=1}^s A_{ik} B_{kj} C_{je} = \sum_{k=1}^n A_{ik} \underbrace{\sum_{j=1}^s B_{kj} C_{je}}_{(BC)_{ke}} = (A(BC))_{il} \end{aligned}$$

$$(2) (\alpha(\beta A))_{ij} = \alpha(\beta A)_{ij} = \alpha \beta A_{ij} = ((\alpha \beta) A)_{ij}$$

defin.
scalar prod.

$$(3) (\alpha(AB))_{ij} = \alpha(AB)_{ij} = \alpha \sum_{k=1}^n A_{ik} B_{kj} = \sum_{k=1}^n A_{ik} (\alpha B_{kj}) = (A \cdot (\alpha B))_{ij}$$

αB " "

$$\text{& also } = \sum_{k=1}^n (\underbrace{\alpha A_{ik}}_{(\alpha A)_{ik}}) B_{kj} = ((\alpha A)B)_{ij}$$

$$(4) (I_m A)_{ij} = \sum_{k=1}^m (I_m)_{ik} A_{kj} = \underset{k=1}{\overset{m}{\downarrow}} 1 \cdot A_{ij} = A_{ij}$$

$$(A \cdot I_n)_{ij} = \sum_{k=1}^n A_{ik} (I_n)_{kj} = \underset{\substack{\text{only } k=i \text{ survives} \\ \text{only } k=j \text{ survives}}}{\downarrow} A_{ij} \cdot 1 = A_{ij}$$

All matrices have the same sizes & same entries, so the equalities in all 4 statements hold. \square

• Next: Relate addition, multiplication & scalar multiplication.

- Theorem 3: (1) $A \in \mathbb{R}^{m \times n}$ & C of size $n \times p$, then
 $(A+B)C = AC + BC$ (m \times p) [Distribution 1]
- (2) A of size $m \times n$, $B \in \mathbb{R}^{n \times p}$, then
 $A(B+C) = AB + AC$ (m \times p) [Distribution 2]
- (3) α, β scalars, A of size $m \times n$, then $(\alpha+\beta)A = \alpha A + \beta A$.
- (4) α scalar, $A \in \mathbb{R}^{m \times n}$, then $\alpha(A+B) = \alpha A + \alpha B$.

Proof: Same idea as with Theorem 2. Check matrices on each side have the same size & the same entries.

$$\begin{aligned} (1) ((A+B)C)_{ij} &= \sum_{k=1}^n (A+B)_{ik} C_{kj} = \sum_{k=1}^n (A_{ik} + B_{ik}) C_{kj} = \\ &= \sum_{k=1}^n (A_{ik} C_{kj}) + (B_{ik} C_{kj}) = \underbrace{\sum_{k=1}^n (A_{ik} C_{kj})}_{\text{order sum}} + \underbrace{\sum_{k=1}^n B_{ik} C_{kj}}_{(BC)_{ij}} \end{aligned}$$

$$\begin{aligned} (2) (A(B+C))_{ij} &= \sum_{k=1}^n A_{ik} (B+C)_{kj} = \sum_{k=1}^n A_{ik} (B_{kj} + C_{kj}) \\ &= \sum_{k=1}^n (A_{ik} B_{kj}) + (A_{ik} C_{kj}) = \underbrace{\sum_{k=1}^n A_{ik} B_{kj}}_{(AB)_{ij}} + \underbrace{\sum_{k=1}^n A_{ik} C_{kj}}_{(AC)_{ij}} \end{aligned}$$

$$(3) ((\alpha+\beta)A)_{ij} = (\alpha+\beta)A_{ij} = \alpha A_{ij} + \beta A_{ij} = (\alpha A)_{ij} + (\beta A)_{ij}$$

$$\begin{aligned} (4) (\alpha(A+B))_{ij} &= \alpha(A+B)_{ij} = \alpha(A_{ij} + B_{ij}) = \underbrace{\alpha A_{ij}}_{(\alpha A)_{ij}} + \underbrace{\alpha B_{ij}}_{(\alpha B)_{ij}} \\ &= (\alpha A)_{ij} + (\alpha B)_{ij} = (\alpha A + \alpha B)_{ij} \end{aligned}$$

§2 The transpose of a matrix:

Idea: Swap rows & columns.

Def: Given A of size $m \times n$, the transpose of A is a matrix A^T of size $n \times m$ with $(A^T)_{ij} = A_{ji}$ for $1 \leq i \leq n$ $1 \leq j \leq m$

Example: $A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 0 \end{bmatrix} \Rightarrow A^T = \begin{bmatrix} 1 & 4 \\ 2 & 5 \\ 3 & 0 \end{bmatrix}$

Def: A is symmetric if $A^T = A$ (in particular $m=n$ so A is a square matrix)

Theorem 4: A, B of size $m \times n$, C of size $n \times p$:

$$(1) (A+B)^T = A^T + B^T \quad (n \times m)$$

$$(2) (AC)^T = C^T A^T \quad (p \times m)$$

$$(3) (A^T)^T = A \quad (m \times n)$$

Proof: As usual, matrices on both sides have the same size, so we only need to check they have the same entries.

$$(1) (A+B)^T_{ij} = (A+B)_{ji} = A_{ji} + B_{ji} + (A^T)_{ij} + (B^T)_{ij} = (A^T + B^T)_{ij}$$

$$(2) (AC)^T_{ij} = (AC)_{ji} = \sum_{k=1}^n A_{ik} C_{kj} = \sum_{k=1}^n C_{kj} A_{ik} = \sum_{k=1}^n (C^T)_{jk} (A^T)_{ki} = (C^T A^T)_{ji}$$

$$(3) (A^T)^T_{ij} = (A^T)_{ji} = A_{ij} \quad \square$$

Property: Assume A is an $n \times n$ matrix. Then AA^T is symmetric, of size $n \times n$.

Proof: $(AA^T)^T = (A^T)^T A^T = AA^T$ so it's symmetric. \square

Thm 4

Note: v in $\mathbb{R}^n \Rightarrow v^T$ is a row vector = matrix of size $1 \times n$

$$\text{Then } \|v\| = \sqrt{v \cdot v} = \sqrt{\sum_{i=1}^n v_i^2} = \sqrt{v^T v}$$

$$\text{Ex: } v = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \quad v^T = [1 \ 2] \text{ and } \|v\| = \sqrt{v^T \cdot v} = \sqrt{1^2 + 2^2} = \sqrt{5}$$

$$v^T = [v_1, \dots, v_n]$$

$$v = \begin{bmatrix} v_1 \\ \vdots \\ v_n \end{bmatrix}.$$

§3 Inverses of matrices:

Def: A matrix A of size $n \times n$ is invertible if we can find a matrix B

of size $n \times n$ satisfying:

$$AB = BA = I_n$$

Example: I_n is invertible ($B = I_n$ works)

Properties: If B exists, it is unique. Call it A^{-1} .

Why? Imagine B & B' both satisfy $AB = BA = I_n$
 $AB' = B'A = I_n$

Then $B = B(AB') = (BA)B' = B'$.

Why? If $A \begin{bmatrix} x_1 \\ \vdots \\ x_n \end{bmatrix} = \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$ is invertible, then $x = A^{-1} \begin{bmatrix} b_1 \\ \vdots \\ b_n \end{bmatrix}$ is the unique solution to Ax = b.