

수포자도 도전해 볼 만한

Mathematics in DeepLearning

Lecture1. Elementary Linear Algebra

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Numbers and arrays

Scalar : a

Vector : \boldsymbol{a}

Matrix : \mathbf{A}

Tensor : \mathbf{A}

Definition A vector \mathbf{a} is an array of numbers

$$\mathbf{a} = (a_1, a_2, \dots, a_n)$$

a_1 : the first element of \mathbf{a}

a_i : the i -th element of \mathbf{a}

Quiz) what is the A and B? $A \leq i \leq B$

row vector $\mathbf{a} = (a_1, a_2, \dots, a_n)$

column vector $\mathbf{a} = \begin{pmatrix} a_1 \\ \vdots \\ a_n \end{pmatrix}$

$(\text{row vector})^T = \text{column vector}$

Operations of vectors : addition

$$\begin{aligned}\mathbf{a} + \mathbf{b} &= (a_1, a_2, \cdots, a_n) + (b_1, b_2, \cdots, b_n) \\ &= (a_1 + b_1, a_2 + b_2, \cdots, a_n + b_n)\end{aligned}$$

$$\begin{aligned}\mathbf{a} - \mathbf{b} &= (a_1, a_2, \cdots, a_n) - (b_1, b_2, \cdots, b_n) \\ &= (a_1 - b_1, a_2 - b_2, \cdots, a_n - b_n)\end{aligned}$$

Operations of vectors : multiplications

$$c\mathbf{a} = (ca_1, ca_2, \cdots, ca_n)$$

$$\begin{aligned}\mathbf{a} \odot \mathbf{b} &= (a_1, a_2, \cdots, a_n) \odot (b_1, b_2, \cdots, b_n) \\ &= (a_1b_1, a_2b_2, \cdots, a_nb_n)\end{aligned}$$

$$\begin{aligned}\mathbf{a} \cdot \mathbf{b} &= (a_1, a_2, \cdots, a_n) \cdot (b_1, b_2, \cdots, b_n) \\ &= (a_1b_1 + a_2b_2 + \cdots + a_nb_n)\end{aligned}$$

Operations of vectors : multiplications

$$\begin{aligned}\mathbf{a} \times \mathbf{b} &= (a_1, a_2, a_3) \times (b_1, b_2, b_3) \\ &= (a_2b_3 - a_3b_2, a_3b_1 - a_1b_3, a_1b_2 - a_2b_1)\end{aligned}$$

Definition A matrix **A** is a 2 D array of numbers

$$\mathbf{A} = \begin{pmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{pmatrix} = (a_{ij})_{m \times n}$$

$$= \begin{pmatrix} a_{11} & a_{1j} & a_{1n} \\ \vdots & a_{ij} & \vdots \\ a_{m1} & a_{mj} & a_{mn} \end{pmatrix} = (\mathbf{a}_1 \mathbf{a}_j \mathbf{a}_n) \\ = \begin{pmatrix} \mathbf{a}_1 \\ \mathbf{a}_i \\ \mathbf{a}_m \end{pmatrix}$$

Operations of matrices : addition, transpose

$$\begin{aligned}\mathbf{A} + \mathbf{B} &= (a_{ij})_{m \times n} + (b_{ij})_{m \times n} \\ &= (a_{ij} + b_{ij})_{m \times n}\end{aligned}$$

$$\mathbf{A} = (a_{ij})_{m \times n} \Rightarrow \mathbf{A}^T = (a_{ji})_{n \times m}$$

Operations of matrices : multiplications

$$\mathbf{cA} = \mathbf{c} \begin{pmatrix} a_{11} & \cdots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \cdots & a_{mn} \end{pmatrix} = \begin{pmatrix} ca_{11} & \cdots & ca_{1n} \\ \vdots & \ddots & \vdots \\ ca_{m1} & \cdots & ca_{mn} \end{pmatrix}$$

$$\begin{aligned} \mathbf{A} \odot \mathbf{B} &= (a_{ij})_{m \times n} \odot (b_{ij})_{m \times n} \\ &= \begin{pmatrix} a_{11}b_{11} & \cdots & a_{1n}b_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1}b_{m1} & \cdots & a_{mn}b_{mn} \end{pmatrix} \end{aligned}$$

Operations of matrices : multiplications

$$\mathbf{AB} = (a_{ij})_{m \times p} (b_{ij})_{p \times n} = (c_{ij})_{m \times n}$$

$$= \begin{pmatrix} a_{11} & \dots & a_{1p} \\ \vdots & \ddots & \vdots \\ a_{m1} & \dots & a_{mp} \end{pmatrix} \begin{pmatrix} b_{11} & \dots & b_{1n} \\ \vdots & \ddots & \vdots \\ b_{p1} & \dots & b_{pn} \end{pmatrix} = \begin{pmatrix} c_{11} & \dots & c_{1n} \\ \vdots & c_{ij} & \vdots \\ c_{m1} & \dots & c_{mn} \end{pmatrix}$$

i-th row

j-th column

$$= (c_{ij})_{m \times n} = (\mathbf{a}_i^T \cdot \mathbf{b}_j)_{m \times n}$$

Tensor

An array with more than two axes

$$\mathbf{A} = (a_{ijk})_{m \times n \times p}$$

$$\begin{aligned}\mathbf{A} \otimes \mathbf{B} &= \begin{pmatrix} a_{11} & \dots & a_{1n} \\ \vdots & \ddots & \vdots \\ a_{m1} & \dots & a_{mn} \end{pmatrix} \otimes \begin{pmatrix} b_{11} & \dots & b_{1q} \\ \vdots & \ddots & \vdots \\ b_{p1} & \dots & b_{pq} \end{pmatrix} \\ &= \begin{pmatrix} a_{11}\mathbf{B} & \dots & a_{1n}\mathbf{B} \\ \vdots & \ddots & \vdots \\ a_{m1}\mathbf{B} & \dots & a_{mn}\mathbf{B} \end{pmatrix}\end{aligned}$$

$$\begin{aligned}
\mathbf{A} \otimes \mathbf{B} &= \begin{pmatrix} a_{11}\mathbf{B} & \cdots & a_{1n}\mathbf{B} \\ \vdots & \ddots & \vdots \\ a_{m1}\mathbf{B} & \cdots & a_{mn}\mathbf{B} \end{pmatrix} \\
&= \begin{pmatrix} a_{11} \\ a_{m1} \\ a_{1n} \\ a_{mn} \end{pmatrix} \otimes \begin{pmatrix} b_{11} & \cdots & b_{1q} \\ \vdots & \ddots & \vdots \\ b_{p1} & \cdots & b_{pq} \end{pmatrix} \\
&= (a_{ijk})_{mn \times p \times q}
\end{aligned}$$

Note.

$$\mathbf{A}(\mathbf{B} + \mathbf{C}) = \mathbf{AB} + \mathbf{AC}$$

$$\mathbf{A}(\mathbf{BC}) = (\mathbf{AB})\mathbf{C}$$

$$\mathbf{a}^T \mathbf{b} = \mathbf{b}^T \mathbf{a}$$

$$(\mathbf{A}^T)^T = \mathbf{A}$$

$$(\mathbf{AB})^T = \mathbf{B}^T \mathbf{A}^T$$

$$(\mathbf{A} + \mathbf{B})^T = \mathbf{A}^T + \mathbf{B}^T$$

Identity & Inverse Matrices

Definition $\mathbf{I}_n \in R^{n \times n}$ is the identity matrix

$$\mathbf{I}_n = \begin{pmatrix} 1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & 1 \end{pmatrix} \text{ identity matrix of } n$$

The matrix inverse of \mathbf{A} is denoted as \mathbf{A}^{-1} and is defined as the matrix such as

$$\mathbf{AB} = \mathbf{I}_n = \mathbf{BA}. \text{ Then } \mathbf{B} = \mathbf{A}^{-1}$$

A linear system $\mathbf{Ax} = \mathbf{b}$ is defined as

$$\sum_{j=1}^n a_{ij}x_j = b_i, \quad i \in \{1, \dots, n\}$$

If **there exist** \mathbf{A}^{-1} , $\mathbf{Ax} = \mathbf{b}$ can be solved

$$\mathbf{x} = \mathbf{A}^{-1}\mathbf{b}$$

For **there exist** \mathbf{A}^{-1} ,

we can check the determinant of the matrix.

$$\det(\mathbf{A}) \neq 0 \Rightarrow \exists \mathbf{A}^{-1}$$

$$\det(\mathbf{A}) = 0 \Rightarrow \nexists \mathbf{A}^{-1}$$

Find \mathbf{C}^{-1} using $\det(\mathbf{C})$

$$\mathbf{C} = \begin{pmatrix} -1 & 3 & 2 \\ 0 & -2 & 1 \\ 1 & 0 & -2 \end{pmatrix}$$

Norm

Definition Norm is a function. We usually measure the size of vectors using the function

$$\|a\|_p = \left(\sum_i |x_i|^p \right)^{1/p} \quad p \geq 1$$

A norm is any function f such that

- $f(x) = 0 \implies x = 0$
- $f(x + y) \leq f(x) + f(y)$
- $\forall c \in \mathbb{R}, f(ax) = |a|f(x)$

Definition Norm is a function. We usually measure the size of matrix using the function

$$\|\mathbf{A}\|_p = \left(\sum_i |x_i|^p \right)^{1/p} \quad p \geq 1$$

A norm is any function f such that

- $f(x) = 0 \implies x = 0$
- $f(x + y) \leq f(x) + f(y)$
- $\forall c \in \mathbb{R}, f(ax) = |a|f(x)$

Note.

$\mathbf{D} = (\mathbf{d}_{ii})_{n \times n}$: diagonal matrix

$\mathbf{A}^T = \mathbf{A}$: symmetric matrix

$(\mathbf{A}\mathbf{A}^T) = \mathbf{I}_n$ *or* $\mathbf{A}^T = \mathbf{A}^{-1}$: Orthogonal matrix

$\text{tr}\mathbf{A} = \sum_{i=1} a_{ii}$

Matrix decomposition

LU, QR, ..., decomposition

Spectral decomposition

Single Value Decomposition (SVD)

Q & A