

# Matrices

## Lecture-1

In a basis  $\hat{e}_i$ , both vectors and linear operators can be described in terms of their components with respect to the basis. These components may be displayed as an array of numbers called a matrix.

If a linear operator  $A$  transforms vectors from an  $N$ -dim vector space, for which we choose a basis  $\hat{e}_j, j=1, \dots, N$ , into vectors belonging to an  $M$ -dim vector space, with basis  $f_i, i=1, \dots, M$ , then we may represent the operator  $A$  by the matrix

$$A = \begin{pmatrix} A_{11} & A_{12} & \dots & A_{1N} \\ A_{21} & A_{22} & \dots & A_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ A_{M1} & A_{M2} & \dots & A_{MN} \end{pmatrix}$$

Component  $A_{ij}$  is also denoted by  $(A)_{ij}$

$$\sum_j (A+B)_{ij} x_j = \sum_j A_{ij} x_j + \sum_j B_{ij} x_j$$

$$\sum_j (\lambda A)_{ij} x_j = \lambda \sum_j A_{ij} x_j$$

$$\sum_j (AB)_{ij} x_j = \sum_k A_{ik} (Bx)_k = \sum_j \sum_k A_{ik} B_{kj} x_j.$$

$$(A+B)_{ij} = A_{ij} + B_{ij}$$

$$(\lambda A)_{ij} = \lambda A_{ij}$$

$$(AB)_{ij} = \sum_k A_{ik} B_{kj}$$

①

Matrix addition and multiplication by a scalar

$$S_{ij} = A_{ij} + B_{ij}$$

$$A + B = B + A$$

matrix addition is commutative and associative.

Multiplication by a scalar is distributive and associative.

$$\vec{y} = A \vec{x}$$

$$y_i = \sum_{j=1}^N A_{ij} x_j \quad \text{for } i=1, \dots, M$$

$$\vec{y} = A \vec{x}$$

$$\begin{pmatrix} y_1 \\ y_2 \\ \vdots \\ y_M \end{pmatrix} = \begin{pmatrix} A_{11} & A_{12} & \dots & A_{1N} \\ A_{21} & A_{22} & \dots & A_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ A_{M1} & A_{M2} & \dots & A_{MN} \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_N \end{pmatrix}$$

$$y_2 = A_{21} x_1 + A_{22} x_2 + \dots + A_{2N} x_N$$

If instead we operate with  $A$  on a basis  $e_j$  having all components zero except for the  $j$ th, which equals unity

$$A e_j = \begin{pmatrix} A_{11} & A_{12} & \dots & A_{1N} \\ A_{21} & A_{22} & \dots & A_{2N} \\ \vdots & \vdots & \ddots & \vdots \\ A_{M1} & A_{M2} & \dots & A_{MN} \end{pmatrix} \begin{pmatrix} 0 \\ 0 \\ \vdots \\ 1 \\ \vdots \\ 0 \end{pmatrix} = \begin{pmatrix} A_{1j} \\ A_{2j} \\ \vdots \\ A_{Mj} \end{pmatrix}$$

$$P_{ij} = \sum_{k=1}^N A_{ik} B_{kj} \quad \text{for } i=1, \dots, M \\ j=1, \dots, R$$

(2)

$$AB \neq BA$$

multiplication of matrix is not in general commutative.

$$(A+B)C = AC + BC$$

$$C(A+B) = CA + CB$$

The null and identity matrices

$$AO = O = OA$$

$$A+O = O+A = A$$

The identity matrix  $I$  has the property

$$AI = IA = A$$

identity matrix must be square

$$I_N = \begin{pmatrix} 1 & & 0 \\ & \dots & \\ 0 & & 1 \end{pmatrix}$$

Transpose of a matrix

$$(AB)^T = B^T A^T$$

$$(AB)^T_{ij} = (AB)_{ji} = \sum_k A_{jk} B_{ki}$$
$$(ABC \dots G)^T = G^T \dots C^T B^T A^T$$

$$= \sum_k (A^T)_{kj} (B^T)_{ik}$$

$$= \sum_k (B^T)_{ik} (A^T)_{kj}$$

$$= (B^T A^T)_{ij}$$

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The Complex and Hermitian conjugates of a matrix

Complex conjugate  $A^*$   
Hermitian  $A^{\dagger}$

The Complex conjugate of a matrix  $A$  is the matrix obtained by taking the Complex conjugate of each of the elements of  $A$ , i.e.

$$(A^*_{ij}) = (A_{ij})^*$$

if a matrix is real (i.e. it contains only real elements) then  $A^* = A$

The Hermitian conjugate (or adjoint) of a matrix  $A$  is simply the transpose of its complex conjugate, or eqv. the complex conjugate of its transpose, i.e.

$$A^{\dagger} = (A^*)^T = (A^T)^*$$

$$(A B \dots G)^{\dagger} = G^{\dagger} \dots B^{\dagger} A^{\dagger}$$

if  $A$  is real (so that  $A^* = A$ ) then  $A^{\dagger} = A^T$ , and so the Hermitian conjugate may be considered as a generalization of the transpose of to complex matrices.

inner product of two vectors. Suppose that in a given orthonormal basis the vectors  $\vec{a}$  and  $\vec{b}$  may be represented by the column matrices

$$a = \begin{pmatrix} a_1 \\ \vdots \\ a_N \end{pmatrix} \quad \text{and} \quad b = \begin{pmatrix} b_1 \\ \vdots \\ b_N \end{pmatrix}$$

taking the Hermitian conjugate of  $a$ , to give a row matrix, and multiplying by  $b$  we obtain

$$a^\dagger b = \begin{pmatrix} a_1^* & a_2^* & \dots & a_N^* \end{pmatrix} \begin{pmatrix} b_1 \\ \vdots \\ b_N \end{pmatrix} = \sum_{i=1}^N a_i^* b_i$$

inner product  $\langle a | b \rangle$

for real vectors  $a^\dagger b = \sum_{i=1}^N a_i b_i$

if the basis  $\hat{e}_i$  is not orthonormal, so that in gen:

$$\langle \hat{e}_i | \hat{e}_j \rangle = G_{ij} \neq \delta_{ij}$$

$$\begin{aligned} \langle \vec{a} | \vec{b} \rangle &= \langle a_1 \hat{e}_1 + \dots + a_N \hat{e}_N | b_1 \hat{e}_1 + \dots + b_N \hat{e}_N \rangle \\ &= \sum_{i=1}^N a_i^* b_i \langle \hat{e}_i | \hat{e}_i \rangle + \sum_i \sum_{j \neq i} a_i^* b_j \langle \hat{e}_i | \hat{e}_j \rangle \\ &= \sum_{i=1}^N a_i^* b_i \end{aligned}$$

Then if  $\vec{a} = \sum_{i=1}^N a_i \hat{e}_i$  and  $\vec{b} = \sum_{i=1}^N b_i \hat{e}_i$

$$\begin{aligned} \langle \vec{a} | \vec{b} \rangle &= \left\langle \sum_{i=1}^N a_i \hat{e}_i \left| \sum_{j=1}^N b_j \hat{e}_j \right. \right\rangle \\ &= \sum_{i=1}^N \sum_{j=1}^N a_i^* b_j \langle \hat{e}_i | \hat{e}_j \rangle \\ &= \sum_{i=1}^N \sum_{j=1}^N a_i^* G_{ij} b_j \end{aligned}$$

in gen where the base vectors  $\hat{e}_1, \hat{e}_2, \dots, \hat{e}_N$  are not orthonormal (or orthogonal)

$$G_{ij} = \langle \hat{e}_i | \hat{e}_j \rangle$$

$$G_{ij} = G_{ji}^* \quad \|\vec{a}\| = \langle \vec{a} | \vec{a} \rangle \text{ is real, symmetric}$$

$$\langle \vec{a} | \vec{a} \rangle^* = \sum_{i=1}^N \sum_{j=1}^N a_i G_{ij}^* a_j^* = \sum_{j=1}^N \sum_{i=1}^N a_j^* G_{ji} a_i = \langle \vec{a} | \vec{a} \rangle$$

the scalar product of  $\vec{a}$  and  $\vec{b}$  in terms of their components with respect to this basis

$$\langle \vec{a} | \vec{b} \rangle = \sum_{i=1}^N \sum_{j=1}^N a_i^* G_{ij} b_j = \mathbf{a}^\dagger \mathbf{G} \mathbf{b}$$

where  $\mathbf{G}$  is the  $N \times N$  matrix with elements  $G_{ij}$