

**Definition:** A matrix of the form

$$U = \begin{bmatrix} u_{11} & u_{12} & u_{13} & \dots & u_{1n} \\ & u_{22} & u_{23} & \dots & u_{2n} \\ & & & \ddots & \\ & & \bigcirc & & \\ & & & & u_{nn} \end{bmatrix}$$

is called an **upper triangular matrix**.

A matrix of the form

$$L = \begin{bmatrix} l_{11} & & & & \bigcirc \\ l_{21} & l_{22} & & & \\ & & & \ddots & \\ l_{n1} & l_{n2} & \dots & & l_{nn} \end{bmatrix}$$

is called a **lower triangular matrix**.

**Notation:** An upper triangular matrix is typically denoted with **U** and a lower triangular matrix is typically denoted with **L**.

properties:

$$\textcircled{1} \quad U^t = L \\ L^t = U.$$

That is,

If you transpose an upper (lower) triangular matrix, you get a lower (upper) triangular matrix.

$$\textcircled{2} \quad \begin{cases} L_1 \cdot L_2 = L \\ U_1 \cdot U_2 = U \end{cases}$$

that is, the product of two lower (upper) triangular matrices is lower (upper) triangular

\textcircled{3} A triangular matrix is **invertible** if and only if all diagonal entries are nonzero.

Example: the matrix

$$A = \begin{bmatrix} 1 & -5 & 3 & 4 \\ 0 & -2 & 1 & 0 \\ 0 & 0 & 0 & 5 \\ 0 & 0 & 0 & 1 \end{bmatrix} \quad \text{is not invertible,}$$

and the matrix

$$B = \begin{bmatrix} 4 & 0 & 0 \\ 1 & 3 & 0 \\ 0 & 2 & 1 \end{bmatrix} \quad \text{is invertible.}$$

# Diagonal Matrices:

**Definition:** A diagonal matrix is a square matrix with zero entries except possibly on the main diagonal.

• The following are examples, of diagonal matrices:

$$A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad B = \begin{bmatrix} \frac{1}{2} & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 \\ 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 4 \end{bmatrix}$$

• In general, a diagonal matrix is given by:

$$D = \begin{bmatrix} d_1 & 0 & 0 & \dots & 0 \\ 0 & d_2 & 0 & \dots & 0 \\ 0 & 0 & d_3 & \dots & 0 \\ & & & \dots & \\ & & & & d_n \end{bmatrix}$$

Notation: A lot of time, a diagonal matrix is referenced with a capital D.

## Finding the power of a diagonal matrix:

If  $D$  is a diagonal matrix, then  $D^n$  for  $n \geq 0$  is given by

$$D^n = \begin{bmatrix} d_1^n & & & \\ & d_2^n & & \\ & & \ddots & \\ & & & d_n^n \end{bmatrix}$$

• The inverse: A diagonal matrix  $D$  is invertible if and only if all the diagonal elements are nonzero.

In this case,  $D^{-1}$  is given by

$$D^{-1} = \begin{bmatrix} \frac{1}{d_1} & & & \\ & \frac{1}{d_2} & & \\ & & \ddots & \\ & & & \frac{1}{d_n} \end{bmatrix}$$

Example: For  $A = \begin{bmatrix} 2 & 0 \\ 0 & 3 \end{bmatrix}$

$$\text{Then } A^{-1} = \begin{bmatrix} \frac{1}{2} & 0 \\ 0 & \frac{1}{3} \end{bmatrix}.$$

## Identity matrix :

Definition: the identity matrix  $I_n$  of size  $n$  is the  $n \times n$  matrix in which all the elements on the main diagonal are equal to  $1$  and all other elements are equal to  $0$ .

Examples :

$$I_1 = [1] \quad , \quad I_2 = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$I_3 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} , \dots$$

$$I_n = \begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & 1 & 0 & \dots & 0 \\ \dots & \dots & \dots & \dots & \dots \\ 0 & 0 & 0 & \dots & 1 \end{bmatrix}$$

Remark:

$A \cdot I_n = I_m \cdot A$  for any  $m$ -by- $n$  matrix  $A$ . In particular,

$A \cdot I_n = I_n \cdot A = A$  for any  $n \times n$  matrix.

## Symmetric matrix

A Symmetric matrix is a square matrix that is equal to its transpose. Formally,

$$A \text{ is symmetric} \Leftrightarrow A = A^t$$

If we put,  $A = (a_{ij})$   $1 \leq i, j \leq n$

Then,

$$A \text{ is symmetric} \Leftrightarrow a_{ij} = a_{ji} \\ \text{for every } i, j.$$

**Example:** Zero matrix and identity matrix are symmetric (any diagonal matrix is symmetric).

• the matrix  $A = \begin{bmatrix} 6 & 8 & 5 \\ 8 & 2 & -1 \\ 5 & -1 & 0 \end{bmatrix}$  is symmetric.

Theorem: Let  $A$  be a square matrix. Then  $AA^t$  and  $A^tA$  are symmetric.

**Proof:** It is clear that

$$(A^tA)^t = A^t(A^t)^t = A^tA$$

i.e.  $A^tA$  is symmetric.

**Theorem:** If  $A$  and  $B$  are symmetric matrices with the same size, and if  $k$  is any scalar, then

①  $A^t$  is symmetric

②  $A+B$  and  $A-B$  are symmetric

③  $k \cdot A$  is symmetric

④ The product of two symmetric matrices is symmetric if and only if the matrices commute, i.e.  $AB = BA$ .

⑤ If  $A$  is invertible symmetric matrix, then  $A^{-1}$  is symmetric.

**Definition:** An  $n \times n$  matrix  $B$  is called skew-symmetric if  $B = -B^t$ .

Example: the matrix  $B = \begin{bmatrix} 0 & 1 & 2 \\ 1 & 0 & 1 \\ 2 & 1 & 3 \end{bmatrix}$

is symmetric.

The matrix  $B = \begin{bmatrix} 0 & 1 & 2 \\ -1 & 0 & 1 \\ -2 & -1 & 0 \end{bmatrix}$  is skew-symmetric.

## Eigenspace

**Definition:** The eigenspace associated with an eigenvalue  $h$  of an  $n \times n$  matrix is the kernel of the matrix  $A - hI_n$  and is denoted by  $E_h$ .

So  $E_h$  consists of all solutions  $v$  of the equation  $Av = hv$ . In other words,  $E_h$  consists of all eigenvectors with eigenvalue  $h$ , together with the zero vector.

**Definition:** The geometric multiplicity for a given eigenvalue is the dimension of the eigenspace  $E_h$ . That is,

$$G_m(h) = \dim E_h.$$

• The algebraic multiplicity for a given eigenvalue  $h$  is the number of times the eigenvalue is repeated. For example, if the characteristic polynomial is  $(\lambda - 1)^2 (\lambda - 2)^3$  then for  $h = 1$  the algebraic multiplicity is 2 and for  $h = 2$  the algebraic multiplicity is 3.

• The algebraic multiplicity is greater than or equal to the geometric multiplicity.