

# Matrix Properties

#### Linear Algebra

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# Matrix Exponential Application



□ Solve systems of linear ordinary differential equations.

$$\frac{d}{dt}y(t) = Ay(t), \qquad y(0) = y_0$$

where A is a constant matrix, is given by

$$y(t) = e^{At}y_0$$

# Matrix Exponential



Is a matrix function on square matrices (A) using Taylor series:

$$e^A = I + A + \frac{1}{2!}A^2 + \frac{1}{3!}A^3 + \frac{1}{4!}A^4 + \cdots$$

□ Special Case: When A is Diagonal:

$$A = \begin{bmatrix} \alpha & 0 \\ 0 & \beta \end{bmatrix} \Rightarrow \underline{e^A} = \begin{bmatrix} e^{\alpha} \\ 0 \end{bmatrix}$$

# Matrix Operations Complexity



 $m \times n$  matrix stored A as  $m \times n$  array of numbers (for sparse A, store only nnz(A) nonzero values)

matrix addition, scalar-matrix multiplication cost m flops

□ matrix-vector multiplication costs  $m(2n-1) \approx 2mn$  flops (for sparse A, around 2nnz(A) flops)

### Transpose



The **transpose** of a matrix results from "flipping" the rows and columns. Given a matrix  $A \in \mathbb{R}^{m \times n}$ , is the  $m \times n$  matrix whose entries are given by

$$(A^T)_{ij} = A_{ji}$$

Properties:

$$\circ$$
  $(A^T)^T = A$ 

$$\circ \quad (A+B)^T = A^T + B^T$$

$$\circ (cA)^T = c(A^T)$$

$$(AB)^{T} = B^{T}A^{T} \rightarrow (A_{1}A_{2}A_{3}\cdots A_{n})^{T} = A_{n}^{T}\cdots A_{3}^{T}A_{2}^{T}A_{1}^{T}$$

# Conjugate Transpose (Adjoint)



$$A^* = A^H = (\bar{A})^T = \overline{A^T}$$

$$A = \begin{bmatrix} 1 & -2 - i & 5 \\ 1 + i & i & 4 - 2i \end{bmatrix} \quad A^{H} = \begin{bmatrix} 1 & 1 - i \\ -2 + i & -i \\ 5 & 4 + 2i \end{bmatrix}$$

- $\Box$   $(A+B)^H=A^H+B^H$  for any two matrices A and B of the same dimensions.
- $\Box$   $(zA)^H = \bar{z}A^H$  for any complex number z and any m-by-n matrix A.
- $\Box$   $(AB)^H = B^H A^H$  for any m-by-n matrix A and any n-by-p matrix B. Note that the order of the factors is reserved.
- $\Box$   $(A^H)^H = A$  for any m-by-n matrix A

For real matrices, the conjugate transpose is just the transpose,  $A^H = A^T$ .

#### Trace



□ The **trace** of a square matrix  $A \in \mathbb{R}^{n \times n}$ , denoted trA, is the sum of diagonal elements in the matrix:

$$trA = \sum_{i=1}^{n} A_{ii},$$

$$Tr\begin{pmatrix} a_{11} & a_{12} & \dots & a_{1,n-1} & a_{1n} \\ a_{21} & a_{22} & \dots & a_{2,n-1} & a_{2n} \\ \vdots & \vdots & \ddots & \vdots & \vdots \\ a_{n-1,1} & a_{n-1,2} & \dots & a_{n-1,n-1} & a_{n-1,n} \\ a_{n1} & a_{n2} & \dots & a_{n,n-1} & a_{nn} \end{pmatrix} = a_{11} + a_{22} + \dots + a_{nn}$$

#### Trace



- □ The trace has following properties:
  - $\circ$  For  $A \in \mathbb{R}^{n \times n}$ ,  $trA = trA^T$ .
  - $\circ$  For  $A, B \in \mathbb{R}^{n \times n}$ , tr(A + B) = trA + trB.
  - o For  $A \in \mathbb{R}^{n \times n}$ ,  $t \in \mathbb{R}$ , tr(tA) = t tr A.
  - $\circ$  For A, B such that AB is square, trAB = trBA.
  - o For A, B, C such that ABC is square, trABC = trBCA = trCAB, and so on for the product of more matrices.
- Trace is a linear function on the matrix space. Why?

## Example

Show that there do not exist matrices  $A, B \in \mathcal{M}_n$  such that AB - BA = I.

#### Kronecker sum



 $\Box$  A and B are square matrices, the Kronecker sum is: ( $I_b$  identity matrix with size b\*b)

$$A \oplus B = A \otimes I_b + I_a \otimes B$$

Properties:

$$\exp(A) \otimes \exp(B) = \exp(A \oplus B)$$

## Example

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \oplus \begin{bmatrix} b_{11} & b_{12} \\ b_{21} & b_{22} \end{bmatrix} = \begin{bmatrix} a_{11} + b_{11} & b_{12} & a_{12} & 0 \\ b_{21} & a_{11} + b_{22} & 0 & a_{12} \\ a_{21} & 0 & a_{22} + b_{11} & b_{12} \\ 0 & a_{21} & b_{21} & a_{22} + b_{22} \end{bmatrix}.$$

# Elementary Matrices



□ An elementary matrix is one that is obtained by performing a single elementary row operation on an identity matrix.

#### Note

If an elementary row operation is performed on an m x m matrix A, the resulting matrix can be written as EA, where m x m matrix E is created by performing the same row operation on  $I_m$ .

## Example

$$E_{1} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -4 & 0 & 1 \end{bmatrix}, E_{2} = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}, E_{3} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 5 \end{bmatrix}, A = \begin{bmatrix} a & b & c \\ d & e & f \\ g & h & i \end{bmatrix}$$



- An  $m \times n$  matrix is
  - $\circ$  Tall m > n
  - Wide n > m
  - $\circ$  Square m=n
- Main diagonal of matrix

$$A_{n\times n} = \begin{bmatrix} & & \\ & & \\ & & \\ & & \end{bmatrix} \quad a_{11}, a_{22}, \dots, a_{nn}$$

$$a_{11}, a_{22}, \dots, a_{nn}$$

Anti diagonal of matrix

$$A_{n \times n} = \begin{bmatrix} & & & \\ & & &$$

$$a_{1,n}$$
,  $a_{2,n-1}$ , ...,  $a_{n,1}$ 



☐ Identity matrix

 $I \in \mathbb{R}^{n \times n}$ , is a square matrix with ones on the diagonal and zeros everywhere else. That is,  $I_{ij} = \begin{cases} 1 & i=j \\ 0 & i \neq j \end{cases}$ .

It has the property that for all  $A \in \mathbb{R}^{m \times n}$ , AI = A = IA.

$$I = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \rightarrow I_n = \begin{bmatrix} e_1 & e_2 & e_3 \end{bmatrix}$$

Diagonal matrix a matrix where all non-diagonal elements are 0.  $D = diag(d_1, ..., d_n)$ ,

with 
$$D_{ij} = \begin{cases} d_{ij} & i = j \\ 0 & i \neq j \end{cases}$$
 
$$A = \operatorname{diag}(a_1, \dots, a_m) = \begin{bmatrix} a_1 & \dots & 0 \\ \vdots & a_i & \vdots \\ 0 & \dots & a_m \end{bmatrix}$$
 Clearly,  $I = \operatorname{diag}(1, 1, \dots, 1)$ .

Scalar matrix A special kind of diagonal matrix in which all diagonal elements are the same

$$\begin{bmatrix} 3 & 0 & 0 \\ 0 & 3 & 0 \\ 0 & 0 & 3 \end{bmatrix}$$



- $\Box$  A square matrix A over R is called:
  - $\circ$  symmetric if  $A^T = A$
  - o skew-symmetric if  $A^T = -A$  (Good Property??)
  - o  $A^TA$  must be symmetric (<u>A with any size, it is not necessary for A to be a square matrix</u>)

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 2 & 4 & -1 \\ 3 & -1 & 4 \end{bmatrix} \text{ and } B = \begin{bmatrix} 0 & 1 & 2 \\ -1 & 0 & -3 \\ -2 & 3 & 0 \end{bmatrix}$$

 $\Box$  A is orthogonal if  $AA^T = A^TA = I$ 

$$A = \begin{bmatrix} \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{2}} & \frac{-1}{\sqrt{2}} & 0 \\ \frac{1}{\sqrt{6}} & \frac{1}{\sqrt{6}} & \frac{-2}{\sqrt{6}} \end{bmatrix}$$

#### Example

The matrix exponential of a skew-symmetric matrix is an orthogonal matrix?

#### Hermitian Matrix



□ Hermitian matrix (or self-adjoint matrix) is a complex square matrix that is equal to its own conjugate transpose

$$A \ Hermitian \iff A = A^H$$

□ conjugate transpose

$$A^H = A^* = (\overline{A})^T$$

## Unitary matrix



$$U^*U = UU^* = UU^{-1} = I$$

#### Note

If U is a square, complex matrix, then the following conditions are equivalent:

- 1. U is unitary.
- 2.  $U^*$  is unitary.
- 3. U is invertible with  $U^{-1} = U^*$ .
- 4. The columns of U form an orthonormal basis of  $\mathbb{C}^n$  with respect to usual inner product. In other words,  $U^*U = 1$ .
- 5. The rows of U form an orthonormal basis of  $\mathbb{C}^n$  with respect to usual inner product. In other words,  $UU^* = 1$ .

#### Normal Matrix



 $\square$  A matrix  $A \in \mathcal{M}_n(\mathbb{C})$  is called **normal** if  $A^*A = AA^*$ 

□ A normal and upper triangle matrix is a diagonal matrix.



□ Submatrix of matrix: A matrix obtained by deleting some of the rows and/or columns of a matrix is said to be a submatrix of the given matrix.

$$A = \begin{bmatrix} 1 & 4 & 5 \\ 0 & 1 & 2 \end{bmatrix}$$

$$\begin{bmatrix} 1 \end{bmatrix}$$
,  $\begin{bmatrix} 2 \end{bmatrix}$ ,  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$ ,  $\begin{bmatrix} 1 \\ 5 \end{bmatrix}$ ,  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$ ,  $\begin{bmatrix} 1 \\ 2 \end{bmatrix}$ ,  $\begin{bmatrix} 1 \\ 1 \end{bmatrix}$ ,  $\begin{bmatrix} 1 \\ 0 \end{bmatrix}$ ,



#### Zero or null Matrix

If  $A \in M_{m \times m}$ , and c is a scalar,

then (1) 
$$A + 0_{m \times m} = A$$

X So,  $\mathbf{0}_{m \times n}$  is also called the additive identity for the set of all  $m \times n$  matrices

(2) 
$$A + (-A) = 0_{m \times m}$$

X Thus, -A is called the additive inverse of A

(3) 
$$cA = 0_{m \times m} \Rightarrow c = 0 \text{ or } A = 0_{m \times m}$$

All above properties are very similar to the counterpart properties for the real number 0



□ Block Matrix whose entries are matrices, such as

$$A = \begin{bmatrix} B & C \\ D & E \end{bmatrix} \rightarrow submatrix or block of A$$

$$B = \begin{bmatrix} 0 & 2 & 3 \end{bmatrix}, C = \begin{bmatrix} -1 \end{bmatrix}, D = \begin{bmatrix} 2 & 2 & 1 \\ 1 & 3 & 5 \end{bmatrix}, E = \begin{bmatrix} 4 \\ 4 \end{bmatrix}$$

then

$$\begin{bmatrix} B & C \\ D & E \end{bmatrix} = \begin{bmatrix} 0 & 2 & 3 & -1 \\ 2 & 2 & 1 & 4 \\ 1 & 3 & 5 & 4 \end{bmatrix}$$

- Matrices in each block row must have same height (row dimension)
- Matrices in each block column must have same width (column dimension)
- Note: A is not a square matrix but it is a block square matrix



#### Block Matrix

- o Transpose of block matrix  $\begin{bmatrix} A & B \\ C & D \end{bmatrix}^T = \begin{bmatrix} A^T & C^T \\ B^T & D^T \end{bmatrix}$
- Multiplication

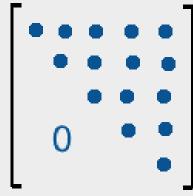
$$A = \begin{bmatrix} 2 & -3 & 1 & 0 & -4 \\ 1 & 5 & -2 & 3 & -1 \\ 0 & -4 & -2 & 7 & -1 \end{bmatrix} = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix}, \qquad B = \begin{bmatrix} 6 & 4 \\ -2 & 1 \\ -3 & 7 \\ -1 & 3 \\ 5 & 2 \end{bmatrix} = \begin{bmatrix} B_1 \\ B_2 \end{bmatrix}$$

$$AB = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \begin{bmatrix} B_1 \\ B_2 \end{bmatrix} = \begin{bmatrix} A_{11}B_1 + A_{12}B_2 \\ A_{21}B_1 + A_{22}B_2 \end{bmatrix} = \begin{bmatrix} -5 & 4 \\ -6 & 2 \\ 2 & 1 \end{bmatrix}$$

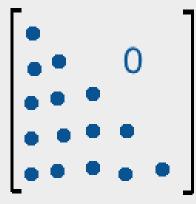


#### □ Triangular matrix

- $\circ$  Upper triangular  $a_{ij} = 0$ , i > j
- $\circ$  Lower triangular  $a_{ij} = 0$ , i < j



Upper Triangular Matrix



Lower Triangular Matrix



#### ■ Sparse matrix

- $_{\circ}$  Density of matrix  $A_{m imes n}$
- Density of identity matrix?
- Sparse matrix has low density

$$1 \ge \frac{nnz(A)}{mn}$$



- Let  $A = [a_{ij}]$  be an  $n \times n$  matrix with  $a_{ij} = \begin{cases} 1 & if \ i = j + 1 \\ 0 & other \end{cases}$ . Then  $A^n = 0$  and  $A^k \neq 0$  for 1 < k < n 1
- $\square$  Nilpotent: A for which a positive integer p exists such that  $A^p = 0$ .
- $\Box$  Order of nilpotency (degree, index): Least positive integer p for which  $A^p = 0$  is called the.

$$A = \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} \quad C = \begin{bmatrix} 5 & -3 & 2 \\ 15 & -9 & 6 \\ 10 & -6 & 4 \end{bmatrix}$$



 $\square$  Idempotent: satisfy the condition that  $A^2 = A$ 

#### Example

2 x 2:

$$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} 3 & -6 \\ 1 & -2 \end{bmatrix}$$

3 x 3:

$$\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 2 & -2 & -4 \\ -1 & 3 & 4 \\ 1 & -2 & -3 \end{bmatrix}$$

#### Note

If a matrix  $\begin{pmatrix} a & b \\ c & d \end{pmatrix}$  is idempotent, then

- $a = a^2 + bc$ .
- b = ab + bd, implying b(1 a d) = 0 so d = 1 a,
- c = ca + cd, implying c(1 a d) = 0 so d = 1 a,
- $d = d^2 + bc$



- ☐ Toeplitz: diagonal-constant matrix: values on diagonals are equal
- □ A Toeplitz matrix is not necessarily square.

$$A = \begin{bmatrix} a_0 & a_{-1} & a_{-2} & \cdots & \cdots & a_{-(n-1)} \\ a_1 & a_0 & a_{-1} & \ddots & & \vdots \\ a_2 & a_1 & \ddots & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \ddots & a_{-1} & a_{-2} \\ \vdots & & \ddots & a_1 & a_0 & a_{-1} \\ a_{n-1} & \cdots & \cdots & a_2 & a_1 & a_0 \end{bmatrix} \quad A_{i,j} = A_{i+1,j+1} = a_{i-j}$$

$$T(b) = \begin{bmatrix} b_1 & 0 & 0 & 0 \\ b_2 & b_1 & 0 & 0 \\ b_3 & b_2 & b_1 & 0 \\ 0 & b_3 & b_2 & b_1 \\ 0 & 0 & b_3 & b_2 \\ 0 & 0 & 0 & b_3 \end{bmatrix}$$

#### Permutation Matrix



- f A square n imes n matrix (P) obtained by rearranging the rows of  $I_n$
- $\square$  Permutation matrix is orthogonal ( $PP^T = I$ )

$$P = \begin{bmatrix} 0 & 1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

- o How many possible permutation matrix?
- A product of permutation matrices is again a permutation matrix
- Some power of a permutation matrix is identity. Why?  $(e. g: p^3 = I)$
- o The inverse of a permutation matrix is again a permutation matrix

# Permutation Matrix Application



$$B = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 3 & 4 \\ 5 & 6 & 7 \end{bmatrix} \quad P = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$$

 $\Box$  Interchange the columns of matrix B:  $P_{ij} = 1$  column i is moved to column j

$$BP = \begin{bmatrix} 1 & 2 & 0 \\ 0 & 3 & 4 \\ 5 & 6 & 7 \end{bmatrix} \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 0 & 2 & 1 \\ 4 & 3 & 0 \\ 7 & 6 & 5 \end{bmatrix}$$

 $\Box$  Interchange the rows of matrix B:  $P_{ij} = 1$  row j is moved to row i

$$PB = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix} \begin{bmatrix} 1 & 2 & 0 \\ 0 & 3 & 4 \\ 5 & 6 & 7 \end{bmatrix} = \begin{bmatrix} 5 & 6 & 7 \\ 0 & 3 & 4 \\ 1 & 2 & 0 \end{bmatrix}$$

# Vec Operator



The vec-operator applied on a matrix A stacks the columns into a vector

$$A = \begin{bmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{bmatrix} \qquad vec(A) = \begin{bmatrix} A_{11} \\ A_{21} \\ A_{12} \\ A_{22} \end{bmatrix}$$

Properties:

$$vec(AXB) = (B^T \otimes A)vec(X)$$
  
 $Tr(A^TB) = vec(A)^Tvec(B)$   
 $vec(A + B) = vec(A) + vec(B)$   
 $vec(\alpha A) = \alpha \cdot vec(A)$   
 $a^TXBX^Tc = vec(X)^T(B \otimes ca^T)vec(X)$ 

# Conclusion



Real Case	Complex Case
$u.v = u^T v = v^T u$	$u.v = v^*u$
Transpose () <sup>T</sup>	Conjugate transpose ()*
Orthogonal matrix $AA^T = I$	Unitary matrix $UU^* = I$
Symmetric matrix $A = A^T$	Hermitian matrix $H = H^*$