

# Introduction to Matrix Algebra

**Matrix and its components; square, symmetric, diagonal, transpose, identity matrix; trace; matrix arithmetic by element, matrix-vector product, matrix-matrix product, matrix inverse, projection matrix, orthogonal matrix**

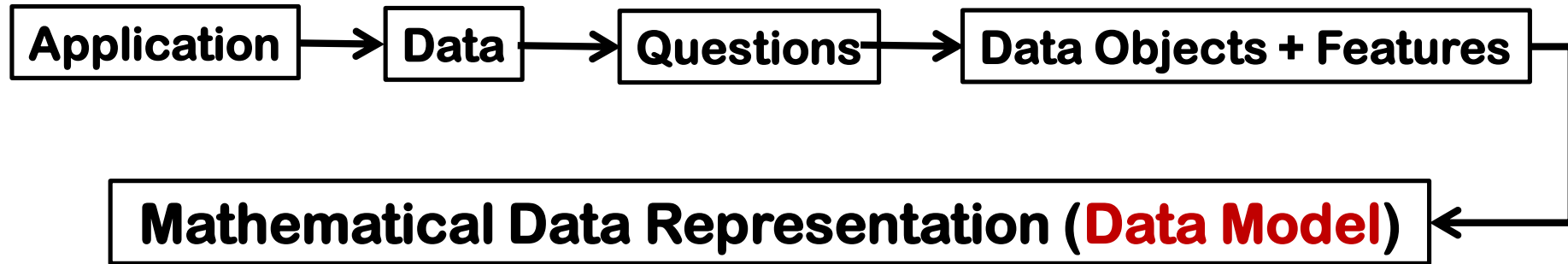
**Nagiza F. Samatova, [samatova@csc.ncsu.edu](mailto:samatova@csc.ncsu.edu)**

Professor, Department of Computer Science  
North Carolina State University

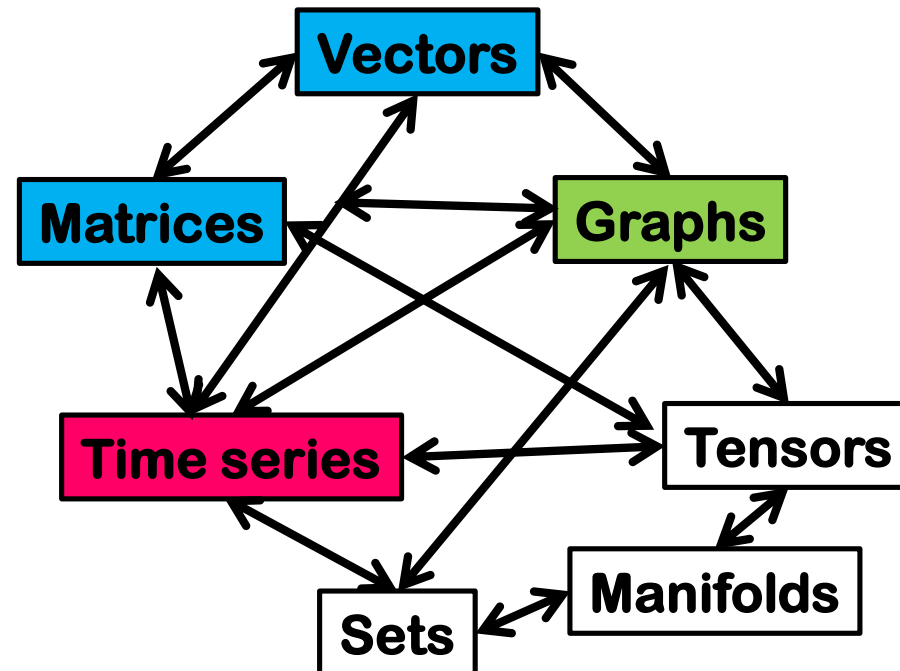
# Matrix Algebra

## **MATRIX DATA MODEL IN DATA SCIENCE**

# Data Model in the Data Science (DS) Process



Different DS methods  
require different  
*abstractions for data  
representation*



Not one hat fits all

More than one models  
is needed

Models are related but often  
in a complementary way

# Data Objects as Matrices

**Example: A collection of text documents on the Web**

## Original Documents

D1:	Child Safety at Home
D2:	Infant & Toddler First Aid
D3:	Your Baby's Health and Safety: From Infant to Toddler

## Parsed Documents

D1:	Child Safety Home
D2:	Infant Toddler
D3:	Bab Health Safety Infant Toddler

## t-d term-document matrix

	D1:	D2:	D3:
T1:	0	0	1
T2:	1	0	0
T3:	0	0	1
T4:	1	0	0
T5:	0	1	1
T6:	1	0	1
T7:	0	1	1

## Terms=Features=Dimensions

T1:	Bab
T2:	Child
T3:	Health
T4:	Home
T5:	Infant
T6:	Safety
T7:	Toddler

**Mining such data ~ studying matrices**

# Vector & Matrix Algebra

## **VECTOR AND MATRIX OPERATORS IN PYTHON**

# Vector Operators in Python ( $x$ is a vector, $A$ is a matrix)

Definition	Python Syntax	Example
To create $x$	$x = \text{xrange}(x_n)$	$x = \text{range}(12)$
To extract a diagonal from $A$ as a vector	$\text{np.diag}(A)$	$A = \text{np.reshape}(\text{range}(12), (3,4))$ $\text{np.diag}(A)$
To find the length of $x$	$\text{len}(x)$	$\text{len}(x)$ # 12
To access the $i^{\text{th}}$ element in $x$	$x[i]$	$x[3]$

# Vector Operators in R ( $x$ is a vector, $A$ is a matrix)

Definition	R Syntax	Example
To create $x$	$x \leftarrow c(x_1, x_2, \dots, x_n)$	<code>x &lt;- c(1:12)</code>
To extract a diagonal from $A$ as a vector	<code>diag(A)</code>	<code>A &lt;- matrix(1:12, 3, 4)</code> <code>diag(A)</code>
To find the length of $x$	<code>length(x)</code>	<code>length(x)</code> # [1] 12
To access the $i^{th}$ element in $x$	<code>x[i]</code>	<code>x[3]</code>

# Matrix Operators in **Python** ( $A$ , $B$ are matrices)

Definition	Python Syntax	Example
To create $A$ with <code>np.reshape()</code>	<code>A = np.reshape(range(<math>x_n</math>), (rows, columns))</code>	<code>m=np.reshape(range(12),(3,4))</code>
To create $A$ from some other object	<code>A = np.array([[values], [values]])</code> <code>np.asmatrix(A)</code>	<code>m = np.array([[1, 2], [3, 4]])</code> <code>np.asmatrix(m)</code>
To create a diagonal matrix from the diagonals of $A$	<code>np.diag(np.diag(A))</code>	<code>np.diag(np.diag(m))</code>
Extract diagonal cells of $A$	<code>np.diag(A, k)</code> $k > 0$ for diagonals above main, $k < 0$ – below main diagonal	<code>np.diag(m, 1)</code>
To find the shape of $A$	<code>A.shape</code>	<code>m.shape</code> # (3, 4)
To find the number of rows in $A$	<code>A.shape[0]</code>	<code>m.shape[0]</code> # 3
To find the number of columns in $A$	<code>A.shape[1]</code>	<code>m.shape[1]</code> # 4



# Matrix Operators in Python ( $A$ , $B$ are matrices) (Cont.)

Definition	Python Syntax	Example
To access the $(i, j)^{th}$ elements in $A$	$A[i, j]$	$m[2, 3]$
To access the $i^{th}$ row in $A$	$A[i, :]$ or $A[i, ]$ or $A[i]$	$m[2]$
To access the $j^{th}$ column in $A$	$A[:, j]$	$m[:, 2]$
To access a subset of rows in $A$	$A[i_1:i_2, :]$ or $A[i_1:i_2]$	$m[1:3, :]$
To access a subset of columns in $A$	$A[:, j_1:j_2]$	$m[:, 2:4]$
To access a sub-matrix	$A[i_1:i_2, j_1:j_2]$	$m[1:3, 1:4]$

# Matrix Operators in **Python** ( $A, B$ are matrices) (Cont.)

Definition	Python Syntax
Addition $A+B$ , the same dimensions	$A + B$
Subtraction $A-B$ , the same dimensions	$A - B$
Hadamard (element-by-element) multiplication, $A \odot B$	$A * B$
Multiplication $A \times B$ , $\text{ncol}(A) = \text{nrow}(B)$	<code>np.matmul(A,B)</code>
Transpose $A^T$ or $A'$	$A.T$
Inversion $A^{-1}$	<code>np.linalg.inv(A)</code>
Determinant	<code>np.linalg.det(A)</code>
Eigen-analysis of $A$	<code>np.linalg.eig(A)</code>
Trace of $A$	<code>np.sum(np.diag(A))</code>
To join vectors into $A$ as columns	<code>np.concatenate((A,B),axis=1)</code>
To join vectors into $A$ as rows	<code>np.concatenate((A,B),axis=0)</code>
To join $A$ and $B$ side-by-side ( $\text{nrow}(A) = \text{nrow}(B)$ )	<code>np.hstack((A, B))</code>
To stack $A$ and $B$ on top of each other ( $\text{ncol}(A) = \text{ncol}(B)$ )	<code>np.vstack((A, B))</code>

# Matrix Operators in R ( $A$ , $B$ are matrices)

Definition	R Syntax	Example
To create $A$ with $m$ rows and $n$ cols	$A \leftarrow \text{matrix}(\text{data}, \text{nrow}=m, \text{ncol}=n, \text{byrow}=F)$	<code>m&lt;-matrix(1:12, 3, 4)</code>
To create a diagonal matrix from	<code>diag(c(<math>x_{11}, \dots, x_{nn}</math>)), diag(const, nrow, ncol)</code>	<code>diag(1, 3, 4)</code>
To create a diagonal matrix from $A$	<code>diag(diag(A))</code>	<code>diag(diag(m))</code>
To create $A$ from a dataframe $df$	<code>data.matrix(df)</code>	<code>data.matrix(df)</code>
To create $A$ from some other object	<code>as.matrix(object)</code>	
To find the dimensions of $A$	<code>dim(A)</code>	<code>dim(m) # [1] 3 4</code>
To find the number of rows in $A$	<code>nrow(A)</code>	<code>nrow(m) # 3</code>
To find the number of columns in $A$	<code>ncol(A)</code>	<code>ncol(m) # 4</code>
To access the $(i, j)^{th}$ elements in $A$	<code>A[i, j]</code>	<code>m[2, 3]</code>
To access the $i^{th}$ row in $A$	<code>A[i, ]</code>	<code>m[3, ]</code>
To access the $j^{th}$ column in $A$	<code>A[, j]</code>	<code>m[, 2]</code>
To access a subset of rows in $A$	<code>A[i<sub>1</sub>:i<sub>2</sub>, ]</code>	<code>m[2:3, ]</code>
To access a subset of columns in $A$	<code>A[, j<sub>1</sub>:j<sub>2</sub>]</code>	<code>m[, 3:4]</code>
To access a sub-matrix	<code>A[i<sub>1</sub>:i<sub>2</sub>, j<sub>1</sub>:j<sub>2</sub>]</code>	<code>m[2:3, 2:4]</code>

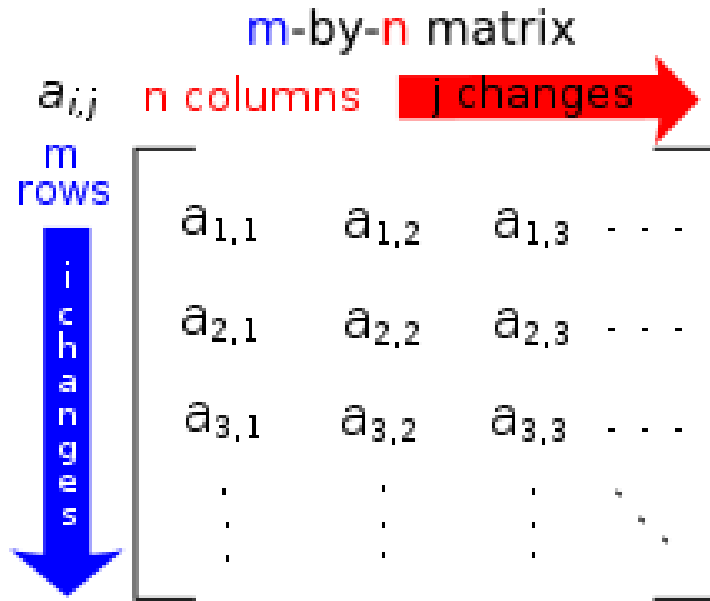
# Matrix Operators in R ( $A$ , $B$ are matrices) (Cont.)

Definition	R Syntax
Addition $A+B$ , the same dimensions	$A + B$
Subtraction $A-B$ , the same dimensions	$A - B$
Hadamard (element-by-element) multiplication, $A \odot B$	$A * B$
Multiplication $A \times B$ , $\text{ncol}(A) = \text{nrow}(B)$	$A \%*\% B$
Transpose $A^T$ or $A'$	$t(A)$
Inversion $A^{-1}$	$\text{solve}(A)$
Determinant $\det(A)$ or $ A $	$\det(A)$
Eigen-analysis of $A$	$\text{eigen}(A)$
Trace of $A$	$\text{sum}(\text{diag}(A))$
To join vectors into $A$ as columns	$\text{cbind}(vec_1, vec_2, \dots, vec_n)$
To join vectors into $A$ as rows	$\text{rbind}(vec_1, vec_2, \dots, vec_m)$
To join $A$ and $B$ side-by-side ( $\text{nrow}(A) = \text{nrow}(B)$ )	$\text{cbind}(A, B)$
To stack $A$ and $B$ on top of each other ( $\text{ncol}(A) = \text{ncol}(B)$ )	$\text{rbind}(A, B)$

# Matrix Algebra

## **MATRIX DEFINITIONS AND TYPES**

# Matrix and Its Components



An  $m \times n$  **matrix**  $A$  is a rectangular array of **scalar numbers**:

- $m$  rows and  $n$  columns, or
- $m$  by  $n$  matrix, or
- a matrix of **order**  $m \times n$ , or
- $A$  has dimensions  $m$  and  $n$

$$A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 4 & 6 \end{pmatrix} \text{ is a } 2 \times 3 \text{ matrix}$$

The numbers  $a_{ij}$  are the **components** (or **elements**) of  $A$ .

A [**column**] vector is a matrix with one column, or an  $m \times 1$  matrix

A [**row**] vector is an  $1 \times n$  matrix

A **square** matrix:  $m = n$

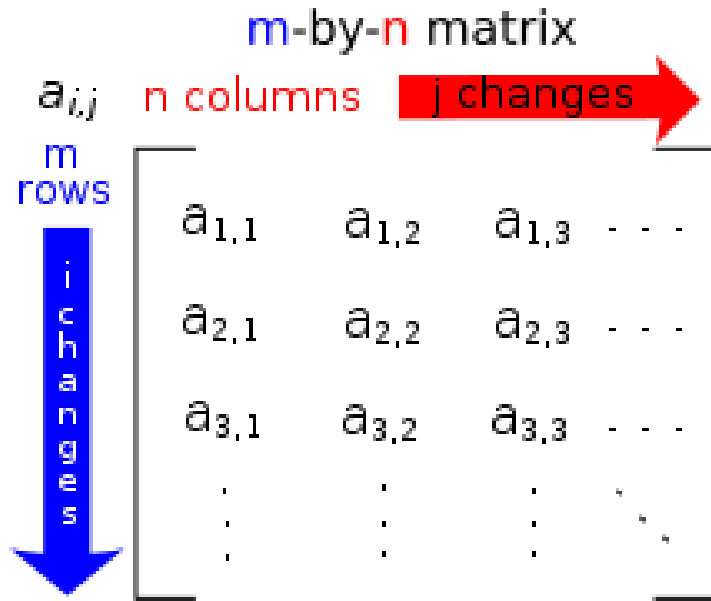
Example in **R**:

```
A <- matrix(c(1,0,2,-6,3,0), nrow=2, ncol=3)
A
```

Example in **Python**:

```
A = np.reshape([1,0,2,-6,3,0], (2,3))
A
```

# Matrix Transpose, $A^T$ or $A'$



$$(A^T)_{i,j} = A_{j,i}$$

$$\begin{aligned}(A^T)^T &= A \\ (A')' &= A\end{aligned}$$

$$\begin{aligned}\xrightarrow{\text{blue}} & \begin{bmatrix} 1 & 2 & 3 \\ 0 & -6 & 0 \end{bmatrix}^T \\ \xrightarrow{\text{brown}} & = \begin{bmatrix} 1 & 0 \\ 2 & -6 \\ 3 & 0 \end{bmatrix}\end{aligned}$$

## Example in R:

```
A <- matrix(c(1,0,2,-6,3,0), nrow=2, ncol=3)
A
B <- t(A)
B
t(B)
help(t)
```

## Example in Python:

```
A = np.reshape([1,0,2,-6,3,0], (2,3))
B = A.T
B
B.T
help(np.transpose)
```

# Ex #7: Transpose of the Matrix

Consider the following 3-by-2 matrix:

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

a. What is the value of  $A[2, 1] =$  ?

b. Show its transpose matrix and validate with Python or R code:  $A^T =$

c. What is the value of  $A^T[1, 2] =$  ?

d. If the matrix had  $m$  rows and  $n$  columns, then how many rows the transpose matrix will have?



# Square Symmetric Matrix

A square matrix  $A$  is **symmetric** if:

- $A^T = A$  or
- $A' = A$  or
- $a_{ij} = a_{ji}$  for every element of  $A$

Which matrix is symmetric?

1)  $A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}$

2)  $A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 4 & 6 \end{pmatrix}$

3)  $A = \begin{pmatrix} 1 & 2 \\ 2 & 4 \end{pmatrix}$

# Python: Diagonal Matrix

A square matrix  $D_n$  with all elements NOT on the diagonal equal to zero (0) is **diagonal** if:

- $d_{ij} = d_{ji} = 0$  for every  $(i, j)$  element of  $D_n$ , with  $i \neq j$ , AND
- $d_{ii} \neq 0$  for at least one value of  $i$

Which matrix is diagonal?

$$1) A = \begin{pmatrix} 1 & 0 \\ 0 & 4 \end{pmatrix}$$

$$2) A = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 6 \end{pmatrix}$$

$$3) A = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

Example in Python:

```
1 A = np.reshape(np.array([1,0,0,2]),(2,2))
2 print(A)
3 B = np.diag(np.array([1,0,0,2]))
4 print(B)
5 np.fill_diagonal(A,5)
6 print(A)
7 C = np.reshape(np.array([1,3,4,2]),(2,2))
8 print(C)
9 np.diag(np.diag(C))
```

# R: Diagonal Matrix

A square matrix  $D_n$  with all elements NOT on the diagonal equal to zero (0) is **diagonal** if:

- $d_{ij} = d_{ji} = 0$  for every  $(i, j)$  element of  $D_n$ , with  $i \neq j$ , AND
- $d_{ii} \neq 0$  for at least one value of  $i$

Which matrix is diagonal?

$$1) A = \begin{pmatrix} 1 & 0 \\ 0 & 4 \end{pmatrix}$$

$$2) A = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 6 \end{pmatrix}$$

$$3) A = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

Example in R:

```
A <- matrix (c (1,0,0,2), nrow=2, ncol=2)
A
B <- diag ( c (1,0,0,2) )
B
help (diag)
diag(A) <- 5
A
C <- matrix (c (1,3,4,2), nrow=2, ncol=2)
C
diag (diag (C) )
```

# Identity Matrix

A diagonal matrix  $I_n$  with all the diagonal equal to one (1) is the **identity** matrix:

- $a_{ij} = a_{ji} = 0$  for every  $(i, j)$  element of  $I_n$ , with  $i \neq j$ , AND
- $a_{ii} = 1$  for **all** values of  $i$

Which matrix is identical?

$$1) A = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$2) A = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 0 & 6 \end{pmatrix}$$

$$3) A = \begin{pmatrix} 0 & 0 \\ 0 & 0 \end{pmatrix}$$

Example in **Python**:

```
I = np.diag(np.array([1,1,1]))  
I
```

Example in **R**:

```
I <- diag (c(1, 1, 1))  
I
```

# Trace of a Square Matrix

The **trace** of a square matrix  $A$  is the sum of all the diagonal elements of the matrix:

- $\text{trace}(A) = \text{tr}(A) = \sum_{i=1}^n a_{ii}$

Example in **Python**:

What is the trace of the matrix?

1)  $A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix}$

2)  $A = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 4 & 6 \end{pmatrix}$

3)  $A = \begin{pmatrix} 2 & 2 \\ 2 & 4 \end{pmatrix}$

```
C = np.reshape(np.array([1,3,4,2]), (2,2))
C
trace = np.sum(np.diag(C))
trace
```

Example in **R**:

```
C <- matrix(c(1,3,4,2), nrow=2, ncol=2)
C
trace <- sum(diag(C))
trace
```

# Trace of the Transposed Matrix

$$\textit{trace}(A) = \textit{trace}(A^T)$$

# Matrix Algebra

## **MATRIX ARITHMETIC: ELEMENT-BY-ELEMENT**

# Matrix Operations: **Element-by-Element**

$$\dim(A) = \dim(B) = m \times n$$


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

**Scalar  
Multiplications**

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

**Addition or  
Subtraction**

$$(A \odot B)_{ij} = A_{ij} \cdot B_{ij}$$

**Hadamard  
Multiplication**

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

**Division**



# Matrix Operations: **Element-by-Element**

## Example in **Python**:

```
A = np.reshape([54,49,49,41,26,43,49,50,58,71],(5,2))
B = np.reshape(range(1,11),(5,2))
print(A)
print(B)
```

## Example in **R**:

```
A=matrix(c(54,49,49,41,26,43,49,50,58,71),nrow=5,ncol=2)
B=matrix(c(1:10), nrow=5, ncol=2)
A
B
```

```
2*A+3
A+B
A-B
A*B
A/B
```

To perform **element-by-element** operations

# Examples: Element-by-Element Matrix Operations

$$A = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \quad B = \begin{pmatrix} 5 & 6 \\ 7 & 8 \end{pmatrix} \quad \lambda = 2$$

$$A + B = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} + \begin{pmatrix} 5 & 6 \\ 7 & 8 \end{pmatrix} = \begin{pmatrix} 6 & 8 \\ 10 & 12 \end{pmatrix} = 2 \cdot \begin{pmatrix} 3 & 4 \\ 5 & 6 \end{pmatrix}$$

$$A - B = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} + \begin{pmatrix} 5 & 6 \\ 7 & 8 \end{pmatrix} = \begin{pmatrix} -4 & -4 \\ -4 & -4 \end{pmatrix}$$

$$A \odot B = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} + \begin{pmatrix} 5 & 6 \\ 7 & 8 \end{pmatrix} = \begin{pmatrix} 5 & 12 \\ 21 & 32 \end{pmatrix}$$

# Matrix Algebra

## **MATRIX-VECTOR AND MATRIX-MATRIX PRODUCTS**

# The Inner Product of Two Vectors

$$a' = (a_1, a_2, \dots, a_n) \text{ and } x = \begin{pmatrix} x_1 \\ \vdots \\ x_n \end{pmatrix}$$

## Inner Product

$$a'x = a_1x_1 + a_2x_2 + \dots + a_nx_n$$

### Problem:

The company purchased the supplies for three products at a unit cost of \$3, \$1, and \$10, respectively. The total order size was 50, 100, and 30 units, resp.

What is the total cost for the order to the company?

### Solution as a Vector Product:

$$a' = (3, 1, 10) \text{ and } x = \begin{pmatrix} 50 \\ 100 \\ 30 \end{pmatrix}$$

$$\begin{aligned} a'x &= (3, 1, 10) \begin{pmatrix} 50 \\ 100 \\ 30 \end{pmatrix} \\ &= 3(50) + 1(100) + 10(30) \\ &= 550 \end{aligned}$$

# Matrix-Vector Multiplication

$A = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \end{pmatrix}$  is a matrix and  $x = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$  is a column vector

## Matrix-vector multiplication:

$$Ax = \begin{pmatrix} a_{11}x_1 + a_{12}x_2 + a_{13}x_3 \\ a_{21}x_1 + a_{22}x_2 + a_{23}x_3 \end{pmatrix} \leftarrow \begin{array}{l} \text{by rows of } A: \\ \text{inner product of} \\ \text{a row of } A \text{ with } x \end{array}$$

### Problem:

The company purchased the supplies for three products at a unit cost of \$3, \$1, and \$10, respectively, in the first quarter and at a unit cost of \$2, \$2, and \$8, in the second quarter. The total order size was 50, 100, and 30 units, resp., in each quarter

What is the total cost for the order to the company for both quarters?

### Solution as a Vector Product:

$$A = \begin{pmatrix} 3 & 1 & 10 \\ 2 & 2 & 8 \end{pmatrix} \text{ and } x = \begin{pmatrix} 50 \\ 100 \\ 30 \end{pmatrix}$$

$$\begin{aligned} Ax &= \begin{pmatrix} 3 & 1 & 10 \\ 2 & 2 & 8 \end{pmatrix} \begin{pmatrix} 50 \\ 100 \\ 30 \end{pmatrix} \\ &= \begin{pmatrix} 3(50) + 1(100) + 10(30) \\ 2(50) + 2(100) + 8(30) \end{pmatrix} \\ &= \begin{pmatrix} 550 \\ 540 \end{pmatrix} = y \end{aligned}$$

**Total Cost:  $\text{sum}(y) = 550 + 540$**

# Matrix-Matrix Multiplication

$$\dim(A) = m \times n$$

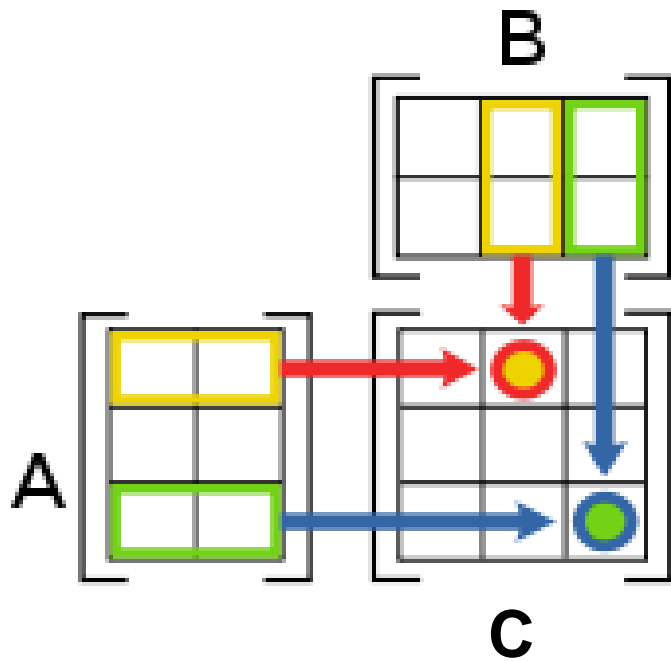
$$C = A * B$$

$$\dim(B) = n \times k$$

$$\dim(C) = m \times k$$

$$(A * B)_{i,j} = A_{i,1}B_{1,j} + A_{i,2}B_{2,j} + \dots + A_{i,n}B_{n,j}$$

$$= \sum_{r=1}^n A_{i,r}B_{r,j}$$



$$1 * 1 + 0 * 1 + 2 * 0 = 1$$

$$\begin{bmatrix} \underline{1} & \underline{0} & \underline{2} \\ -1 & 3 & 1 \end{bmatrix} \times \begin{bmatrix} 3 & \underline{1} \\ 2 & \underline{1} \\ 1 & \underline{0} \end{bmatrix} = \begin{bmatrix} 5 & \underline{1} \\ 4 & 2 \end{bmatrix}$$

$\underline{2 \times 3} \quad \underline{2 \times 3} \quad \underline{3 \times 2}$

In Python:

$$C = \text{np.matmul}(A, B)$$

In R:

$$C = A \%*\% B$$

# Example: Matrix-Matrix Multiplication

## Problem:

The company purchased the supplies for three products at a unit cost of \$3, \$1, and \$10, respectively, in the first quarter and at a unit cost of \$2, \$2, and \$8, in the second quarter.

The total order size for each quarter was 50, 100, and 30 units, resp., during the first year and 60, 80, and 40 units during the second year?

What is the total cost for the order to the company for two years?

## Solution as a Matrix Product:

$$A = \begin{pmatrix} 3 & 1 & 10 \\ 2 & 2 & 8 \end{pmatrix} \text{ and } B = \begin{pmatrix} 50 & 60 \\ 100 & 80 \\ 30 & 40 \end{pmatrix}$$

$$\begin{aligned} AB &= \begin{pmatrix} 3 & 1 & 10 \\ 2 & 2 & 8 \end{pmatrix} \begin{pmatrix} 50 & 60 \\ 100 & 80 \\ 30 & 40 \end{pmatrix} \\ &= \begin{pmatrix} 550 & 660 \\ 540 & 600 \end{pmatrix} \\ &= C \end{aligned}$$

$$\text{Total Cost: } \text{sum}(C) = 550 + 540 + 660 + 600$$

# Matrix Multiplication: **Non-Commutative**

$$A_{2 \times 2} = \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \quad B_{2 \times 2} = \begin{pmatrix} 5 & 6 \\ 7 & 8 \end{pmatrix} \quad C_{2 \times 2} = A \times B$$

$$\begin{aligned} A \times B &= \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \times \begin{pmatrix} 5 & 6 \\ 7 & 8 \end{pmatrix} \\ &= \begin{pmatrix} 1 \times 5 + 2 \times 7 & 1 \times 6 + 2 \times 8 \\ 3 \times 5 + 4 \times 7 & 3 \times 6 + 4 \times 8 \end{pmatrix} \\ &= \begin{pmatrix} 5 + 14 & 6 + 16 \\ 15 + 28 & 18 + 32 \end{pmatrix} \\ &= \begin{pmatrix} 19 & 22 \\ 43 & 50 \end{pmatrix} \end{aligned}$$

$$\begin{aligned} B \times A &= \begin{pmatrix} 5 & 6 \\ 7 & 8 \end{pmatrix} \times \begin{pmatrix} 1 & 2 \\ 3 & 4 \end{pmatrix} \\ &= \begin{pmatrix} 5 + 18 & 10 + 24 \\ 7 + 24 & 14 + 32 \end{pmatrix} \\ &= \begin{pmatrix} 23 & 34 \\ 31 & 46 \end{pmatrix} \end{aligned}$$


$$A \times B \neq B \times A$$




# Example: Multiplication of Rectangular Matrices

$$A_{2 \times 3} = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix} \quad B_{3 \times 2} = \begin{pmatrix} 1 & 2 \\ 3 & 4 \\ 5 & 6 \end{pmatrix}$$

$$A_{2 \times 3} \times B_{3 \times 2} = C_{2 \times 2} = \begin{pmatrix} 22 & 28 \\ 49 & 64 \end{pmatrix}$$

  
match

$$B_{3 \times 2} \times A_{2 \times 3} = W_{3 \times 3} = \begin{pmatrix} 9 & 12 & 15 \\ 19 & 26 & 33 \\ 29 & 40 & 51 \end{pmatrix}$$

  
match

$$A \times B \neq B \times A$$

# Multiplication with Identity Matrix: Commutative

Let  $I_m$  be the identity matrix of size  $m \times m$  and  $I_n$  be the identity matrix of size  $n \times n$

$$A_{n \times m} \times I_m = I_n \times A_{n \times m} = A_{n \times m}$$

$$A_{2 \times 3} = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{pmatrix}$$

$$I_2 = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

$$I_3 = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix}$$

- **Verify that  $A \times I = I \times A = A$**
- **Note: subscripts are not included if the context is clear**

# Ex #8: Matrix-Matrix Multiplication

1. Generate in Python a 3-by-3 matrix A filled with one's everywhere but the diagonal; and with 3's on the diagonal.
2. Generate in Python a 3-by-2 matrix B filled with two's.

a. Multiply matrix A by matrix B and print the resulting matrix C:

$$C = AB =$$

b. What is the size of this new matrix C, i.e. number of rows and cols?

c. Can you multiply matrix B by A to get matrix D (why or why not):

$$D = BA =$$

d. Show with manual calculations how you will get the value of:

$$C[2, 2] =$$

# Matrix Algebra

## **PROJECTION AND ORTHOGONAL MATRIX**

# Ex #9: Projection Matrix

1. Generate in Python a 4-by-2 matrix  $A$  and plot its 4 rows as points in 2-dim.
2. Generate a 2-by-2 matrix with diagonal elements as one's and off-diagonal matrix as zero's (aka *identity* matrix,  $I$ ).

a. Multiply in Python matrix  $A$  by matrix  $I$  and print the resulting matrix  $C$ :

$$C = A \times I =$$

b. Set the  $I[2, 2]$  element to zero and assign the new matrix to  $P$ .

c. Multiply matrix  $A$  by the modified matrix  $I$  to get matrix  $D$ :

$$D = A \times P =$$

d. Add the rows of matrix  $D$  as four points in 2-dim. to your original plot of  $A$ .

Do you observe that these new points are projections of the original points?

What axis are the points of  $A$  projected to in  $D$ ?

e. How will you modify the identity matrix  $I$  to project the points of  $A$  on the other axis?

# Orthogonal Matrix

A square  $m \times m$  matrix  $A$  is **orthogonal** if:

$$A \times A^T = A^T \times A = I_m - \text{the identity matrix}$$

**Example:** Verify (manually or using Python) that the following matrices are orthogonal:

$$A_{2 \times 2} = \frac{1}{\sqrt{2}} \begin{pmatrix} 1 & -1 \\ 1 & 1 \end{pmatrix} \quad B_{2 \times 2} = \begin{pmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{pmatrix}$$

**You can choose any  $\theta$ , e.g.,  $\theta = \frac{\pi}{2}, \frac{\pi}{4}, \frac{\pi}{3}$**

**Side Note:** An orthogonal matrix whose elements are all  $+1$  or  $-1$  is known as Hadamard matrix; it has a role in statistical experimental design.

# Matrix Algebra

## **INVERSES**

# Inverse of a Square Matrix, $\text{nrow}(A)=\text{ncol}(A)=n$

Identity Matrix,  $I_n$

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

Matrix Inverse,  $A^{-1}$

$$A \times A^{-1} = A^{-1} \times A = I_n$$

Examples: Verify (manually) the inverse of the matrix

$$A = \begin{pmatrix} 2 & 3 \\ 3 & 4 \end{pmatrix}$$

$$A^{-1} = \begin{pmatrix} -4 & 3 \\ 3 & -2 \end{pmatrix}$$

$$A = \begin{pmatrix} 3 & 4 \\ 2 & 3 \end{pmatrix}$$

$$A^{-1} = \begin{pmatrix} 3 & -2 \\ -4 & 3 \end{pmatrix}$$



# Inverse of a Square Matrix, $\text{nrow}(A)=\text{ncol}(A)=n$

Identity Matrix,  $I_n$

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

Matrix Inverse,  $A^{-1}$

$$A \times A^{-1} = A^{-1} \times A = I_n$$

Example in R:

```
A <- matrix(rnorm(16), nrow=4, ncol=4)
options(digits=3)
A
IA <- solve(A)
IA
A %*% IA
IA %*% A
help(solve)
```

Example in Python:

```
1 A = np.random.rand(4,4)
2 np.set_printoptions(precision=3)
3 print(A)
4 IA = np.linalg.inv(A)
5 print(IA)
6 print(np.matmul(A, IA))
7 print(np.matmul(IA, A))
```

# Inverse of Inverse

$$(A^{-1})^{-1} = A$$

Example in **Python**:

```
A = np.reshape(np.array([2,3,3,4]), (2,2))
print(A)
IA = np.linalg.inv(A)
print(IA)
```

Example in **R**:

```
A <- matrix ( c(2, 3, 3, 4), nrow=2, ncol=2, byrow=F)
A
IA <- solve (A)
IA
solve (IA)
```

# Python: Inverse of Matrix Product

$$(A \times B)^{-1} = B^{-1} \times A^{-1}$$

## Example in Python:

```
1 A = np.reshape([2,3,3,4],(2,2),order='F')
2 print(A)
3 B = np.reshape([3,4,2,3],(2,2),order='F')
4 print(B)
5 C = np.matmul(A,B)
6 print(C)
7 IC = np.linalg.inv(C)
8 print(IC)
9 np.matmul(np.linalg.inv(B),np.linalg.inv(A))
```

```
[[2 3]
 [3 4]]
[[3 2]
 [4 3]]
[[18 13]
 [25 18]]
[[-18.  13.]
 [ 25. -18.]]

array([[ -18.,   13.],
       [  25.,  -18.]])
```

# R: Inverse of Matrix Product

$$(A \times B)^{-1} = B^{-1} \times A^{-1}$$

## Example in R:

```
A <- matrix ( c(2, 3, 3, 4), nrow=2, ncol=2, byrow=F)
A
B <- matrix ( c(3, 4, 2, 3), nrow=2, ncol=2, byrow=F)
B
C <- A %*% B
C
IC <- solve (C)
IC
solve (B) %*% solve(A)
```

```
> C <- A %*% B
> C
      [,1] [,2]
[1,]   18  13
[2,]   25  18
>
> IC <- solve (C)
> IC
      [,1] [,2]
[1,]  -18  13
[2,]   25 -18
>
> solve (B) %*% solve(A)
      [,1] [,2]
[1,]  -18  13
[2,]   25 -18
```

# Python: Inverse of Transpose

$$(A^T)^{-1} = (A^{-1})^T$$

## Example in Python:

```
1 A = np.reshape(np.array([2,3,3,4]),(2,2),order='F')
2 T = A.T
3 print(A)
4 print(T)
5 IT = np.linalg.inv(T)
6 print(IT)
7 IA = np.linalg.inv(A)
8 print(IA.T)
```

```
[[2 3]
 [3 4]]
[[2 3]
 [3 4]]
[[-4.  3.]
 [ 3. -2.]]
[[-4.  3.]
 [ 3. -2.]]
```

# R: Inverse of Transpose

$$(A^T)^{-1} = (A^{-1})^T$$

## Example in R:

```
A <- matrix ( c(2, 3, 3, 4), nrow=2, ncol=2, byrow=F)
T <- t(A)
A
T
IT <- solve (T)
IT
IA <- solve (A)
t (IA)
help(t)
```

```
> A
      [,1] [,2]
[1,]     2     3
[2,]     3     4
> T
      [,1] [,2]
[1,]     2     3
[2,]     3     4
> IT <- solve (T)
> IT
      [,1] [,2]
[1,]    -4     3
[2,]     3    -2
> IA <- solve (A)
> t (IA)
      [,1] [,2]
[1,]    -4     3
[2,]     3    -2
```

# Existence of the Inverse Matrix, $\det(A) \neq 0$

## Matrix Inverse, $A^{-1}$

$$A \times A^{-1} = A^{-1} \times A = I_n$$

$$A^{-1} \text{ exists if and only if } \det(A) \neq 0$$

Determinant,  $\det(A)$ ,  $|A|$

scalar:  $\det \begin{pmatrix} a & b \\ c & d \end{pmatrix} = ad - bc$

## 2-by-2 Matrix Inverse

$$A^{-1} = \frac{1}{\det(A)} \begin{pmatrix} d & -b \\ -c & a \end{pmatrix}$$

## $n$ -by- $n$ Matrix Inverse Methods:

- LU-factorization
- Gaussian elimination
- Gauss-Jordan elimination

# Existence of the Inverse Matrix, $\det(A) \neq 0$

## Matrix Inverse, $A^{-1}$

$$A \times A^{-1} = A^{-1} \times A = I_n$$

$A^{-1}$  **exists** if and only if  $\det(A) \neq 0$

### Example in Python:

```
A = np.random.rand(4,4)
print(A)
np.linalg.det(A)
```

### Example in R:

```
A <- matrix(rnorm(16), nrow=4, ncol=4)
options(digits=3)
A
det(A)
help(det)
```



# Matrix Algebra

## **TRANSPOSE AND TRACE OF MATRIX PRODUCT**

# Python: Transpose of the Matrix Product

$$(A \times B)^T = B^T \times A^T$$

## Example in Python:

```
1 A = np.reshape(np.array([1,2,3,4,5,6]),(2,3))
2 B = np.reshape(np.array([1,2,3,4,5,6]),(3,2))
3 U = np.matmul(A,B)
4 print(U.T)
5 V = np.matmul(B.T, A.T)
6 print(V)
7 W = np.matmul(A.T, B.T)
8 print(W)
```

```
[[22 49]
 [28 64]]
[[22 49]
 [28 64]]
[[ 9 19 29]
 [12 26 40]
 [15 33 51]]
```

# R: Transpose of the Matrix Product

$$(A \times B)^T = B^T \times A^T$$

Example in R:

```
A <- matrix(c(1, 2, 3, 4, 5, 6), nrow=2, ncol=3, byrow=T)
B <- matrix(c(1, 2, 3, 4, 5, 6), nrow=3, ncol=2, byrow=T)

U <- A %*% B
t(U)
V <- t(B) %*% t(A)
V
W <- t(A) %*% t(B)
W
```

```
> U <- A %*% B
> t(U)
      [,1] [,2]
[1,]    22    49
[2,]    28    64
> V <- t(B) %*% t(A)
> V
      [,1] [,2]
[1,]    22    49
[2,]    28    64
>
> W <- t(A) %*% t(B)
> W
      [,1] [,2] [,3]
[1,]     9    19    29
[2,]    12    26    40
[3,]    15    33    51
```

# Trace of the Matrix Product

$$\textit{trace}(A \times B) = \textit{trace} (B \times A)$$