

# 1. Matrix, vector and scalar representation

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## 1.1 Matrix

Example:

$$x = \begin{bmatrix} 4.1 & 5.3 \\ -3.9 & 8.4 \\ 6.4 & -1.8 \end{bmatrix}$$

$x_{ij}$  is the element at the  $i^{th}$  row and  $j^{th}$  column. Here:  $x_{11} = 4.1, x_{32} = -1.8$ .

Dimension of matrix  $x$  is the number of rows times the number of columns.

Here  $\dim(x) = 3 \times 2$ .  $x$  is said to be a  $3 \times 2$  matrix.

The set of all  $3 \times 2$  matrices is  $\mathbb{R}^{3 \times 2}$ .

## 1.2 Vector

Example:

$$y = \begin{bmatrix} 4.1 \\ -3.9 \\ 6.4 \end{bmatrix}$$

$y_i = i^{th}$  element of  $y$ . Here:  $y_1 = 4.1, y_3 = 6.4$ .

Dimension of vector  $y$  is the number of rows.

Here  $\dim(y) = 3 \times 1$  or  $\dim(y) = 3$ .  $y$  is said to be a 3-dim vector.

The set of all 3-dim vectors is  $\mathbb{R}^3$ .

## 1.3 Scalar

Example:

$$z = 5.6$$

A scalar has no dimension.

The set of all scalars is  $\mathbb{R}$ .

Note:  $z = [5.6]$  is a 1-dim vector, not a scalar.

## Question 1: Represent the previous matrix, vector and scalar in Python

Hint: You may use numpy library, `shape()`, `type()`, `dtype`.

```
In [ ]: import numpy as np

#YOUR CODE HERE

x = np.array([[4.1, 5.3], [-3.9, 8.4], [6.4, -1.8]]) # matrix x
print(x)
print(x.shape) # size of x
print(type(x)) # type of x
print(x.dtype) # data type of x

y = x[:, 0] # vector y
print(y)
print(y.shape) # size of y

z = y[0] # scalar z
print(z)
print(z.shape) # size of z

[[ 4.1  5.3]
 [-3.9  8.4]
 [ 6.4 -1.8]]
(3, 2)
<class 'numpy.ndarray'>
float64
[ 4.1 -3.9  6.4]
(3,)
4.1
()
```

```
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```

## 2. Matrix addition and scalar-matrix multiplication

### 2.1 Matrix addition

Example:

$$\begin{bmatrix} 4.1 & 5.3 \\ -3.9 & 8.4 \\ 6.4 & -1.8 \end{bmatrix}_{3 \times 2} + \begin{bmatrix} 2.7 & 7.3 \\ 3.5 & 2.4 \\ 6.0 & -1.1 \end{bmatrix}_{3 \times 2} = \begin{bmatrix} 4.1 + 2.7 & 5.3 + 7.3 \\ -3.9 + 3.5 & 8.4 + 2.4 \\ 6.4 + 6.0 & -1.8 - 1.1 \end{bmatrix}_{3 \times 2}$$

All matrix and vector operations must satisfy dimensionality properties. For example, it is not allowed to add two matrices of different dimensionalities, such as

$$\begin{bmatrix} 4.1 & 5.3 \\ -3.9 & 8.4 \\ 6.4 & -1.8 \end{bmatrix}_{3 \times 2} + \begin{bmatrix} 2.7 & 7.3 & 5.0 \\ 3.5 & 2.4 & 2.8 \end{bmatrix}_{2 \times 3} = \text{Not allowed}$$

### 2.1 Scalar-matrix multiplication

Example:

$$\begin{array}{ccc}
 3 & \times & \begin{bmatrix} 4.1 & 5.3 \\ -3.9 & 8.4 \\ 6.4 & -1.8 \end{bmatrix} \\
 \text{No dim} & + & 3 \times 2
 \end{array}
 =
 \begin{array}{ccc}
 & & \begin{bmatrix} 3 \times 4.1 & 3 \times 5.3 \\ 3 \times -3.9 & 3 \times 8.4 \\ 3 \times 6.4 & 3 \times -1.8 \end{bmatrix} \\
 & & 3 \times 2
 \end{array}$$

## Question 2: Add the two matrices, and perform the multiplication scalar-matrix as above in Python

```

In [ ]: import numpy as np

#YOUR CODE HERE

X1 = np.array([[4.1, 5.3], [-3.9, 8.4], [6.4, -1.8]])
X2 = np.array([[2.7, 3.5], [7.3, 2.4], [5.0, 2.8]])
X = X1 + X2 # summation of X1 and X2

print(X1)
print(X2)
print(X)

Y1 = X * 4 # X multiplied by 4
Y2 = X / 3 # X divided by 3

print(X)
print(Y1)
print(Y2)

[[ 4.1  5.3]
 [-3.9  8.4]
 [ 6.4 -1.8]]
[[2.7 3.5]
 [7.3 2.4]
 [5.  2.8]]
[[ 6.8  8.8]
 [ 3.4 10.8]
 [11.4  1. ]]
[[ 6.8  8.8]
 [ 3.4 10.8]
 [11.4  1. ]]
[[27.2 35.2]
 [13.6 43.2]
 [45.6  4. ]]
[[2.26666667 2.93333333]
 [1.13333333 3.6       ]
 [3.8        0.33333333]]

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```

## 3. Matric-vector multiplication

### 3.1 Example

Example:

$$\begin{array}{ccc} \begin{bmatrix} 4.1 & 5.3 \\ -3.9 & 8.4 \\ 6.4 & -1.8 \end{bmatrix} & \times & \begin{bmatrix} 2.7 \\ 3.5 \end{bmatrix} = \begin{bmatrix} 4.1 \times 2.7 + 5.3 \times 3.5 \\ -3.9 \times 2.7 + 8.4 \times 3.5 \\ 6.4 \times 2.7 - 1.8 \times 3.5 \end{bmatrix} \\ 3 \times 2 & 2 \times 1 & = 3 \times 1 \end{array}$$

Dimension of the matrix-vector multiplication operation is given by contraction of  $3 \times 2$  with  $2 \times 1 = 3 \times 1$ .

## 3.2 Formalization

$$\begin{array}{ccccc} [A] & \times & [x] & = & [y] \\ m \times n & & n \times 1 & = & m \times 1 \end{array}$$

Element  $y_i$  is given by multiplying the  $i^{th}$  row of  $A$  with vector  $x$ :

$$\begin{array}{ccccc} y_i & = & A_i & & x \\ 1 \times 1 & = & 1 \times n & \times & n \times 1 \end{array}$$

It is not allowed to multiply a matrix  $A$  and a vector  $x$  with different  $n$  dimensions such as

$$\begin{array}{ccccc} \begin{bmatrix} 4.1 & 5.3 \\ -3.9 & 8.4 \\ 6.4 & -1.8 \end{bmatrix} & \times & \begin{bmatrix} 2.7 \\ 3.5 \\ -7.2 \end{bmatrix} & = & ? \\ 3 \times 2 & \times & 3 \times 1 & = & \text{not allowed} \end{array}$$

## Question 3: Multiply the matrix and vector above in Python

```
In [ ]: import numpy as np

#YOUR CODE HERE

A = np.array([[4.1, 5.3], [-3.9, 8.4], [6.4, -1.8]])
x = np.array([[2.7], [3.5]])
y = np.dot(A, x)      # multiplication of A and x
print(A)
print(A.shape)        # size of A
print(x)
print(x.shape)         # size of x
print(y)
print(y.shape)         # size of y

[[ 4.1  5.3]
 [-3.9  8.4]
 [ 6.4 -1.8]]
(3, 2)
[[2.7]
 [3.5]]
(2, 1)
[[29.62]
 [18.87]
 [10.98]]
(3, 1)

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```

## 4. Matrix-matrix multiplication

### 4.1 Example

$$\begin{bmatrix} 4.1 & 5.3 \\ -3.9 & 8.4 \\ 6.4 & -1.8 \end{bmatrix}_{3 \times 2} \times \begin{bmatrix} 2.7 & 3.2 \\ 3.5 & -8.2 \end{bmatrix}_{2 \times 2} = \begin{bmatrix} 4.1 \times 2.7 + 5.3 \times 3.5 & 4.1 \times 3.2 + 5.3 \times (-8.2) \\ -3.9 \times 2.7 + 8.4 \times 3.5 & -3.9 \times 3.2 + 8.4 \times (-8.2) \\ 6.4 \times 2.7 - 1.8 \times 3.5 & 6.4 \times 3.2 - 1.8 \times (-8.2) \end{bmatrix}_{3 \times 2}$$

Dimension of the matrix-matrix multiplication operation is given by contraction of  $3 \times 2$  with  $2 \times 2 = 3 \times 2$ .

### 4.2 Formalization

$$\begin{matrix} [A] \\ m \times n \end{matrix} \times \begin{matrix} [X] \\ n \times p \end{matrix} = \begin{matrix} [Y] \\ m \times p \end{matrix}$$

Like for matrix-vector multiplication, matrix-matrix multiplication can be carried out only if  $A$  and  $X$  have the same  $n$  dimension.

### 4.3 Linear algebra operations can be parallelized/distributed

Column  $Y_i$  is given by multiplying matrix  $A$  with the  $i^{th}$  column of  $X$ :

$$\begin{matrix} Y_i \\ 1 \times 1 \end{matrix} = \begin{matrix} A \\ 1 \times n \end{matrix} \times \begin{matrix} X_i \\ n \times 1 \end{matrix}$$

Observe that all columns  $X_i$  are independent. Consequently, all columns  $Y_i$  are also independent. This allows to vectorize/parallelize linear algebra operations on (multi-core) CPUs, GPUs, clouds, and consequently to solve all linear problems (including linear regression) very efficiently, basically with one single line of code ( $Y = AX$  for millions/billions of data). With Moore's law (computers speed increases by 100x every decade), it has introduced a computational revolution in data analysis.

### Question 4: Multiply the two matrices above in Python

```
In [ ]: import numpy as np

#YOUR CODE HERE

A = np.array([[4.1, 5.3], [-3.9, 8.4], [6.4, -1.8]])
X = np.array([[2.7, 3.2], [3.5, -8.2]])
Y = np.dot(A, X)      # matrix multiplication of A and X
print(A)
print(A.shape)        # size of A
print(X)
print(X.shape)        # size of X
print(Y)
print(Y.shape)        # size of Y
```

```
[[ 4.1  5.3]
 [-3.9  8.4]
 [ 6.4 -1.8]]
(3, 2)
[[ 2.7  3.2]
 [ 3.5 -8.2]]
(2, 2)
[[ 29.62 -30.34]
 [ 18.87 -81.36]
 [ 10.98  35.24]]
(3, 2)
```

```
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## 5. Some linear algebra properties

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### 5.1 Matrix multiplication is *not* commutative

$$\begin{matrix} A & \times & B & \neq & B & \times & A \\ \begin{bmatrix} 4.1 & 5.3 \\ -3.9 & 8.4 \\ 6.4 & -1.8 \end{bmatrix} & \times & \begin{bmatrix} 2.7 & 3.2 \\ 3.5 & -8.2 \end{bmatrix} & \neq & \begin{bmatrix} 2.7 & 3.2 \\ 3.5 & -8.2 \end{bmatrix} & \times & \begin{bmatrix} 4.1 & 5.3 \\ -3.9 & 8.4 \\ 6.4 & -1.8 \end{bmatrix} \end{matrix}$$

### 5.2 Scalar multiplication is associative

$$\begin{matrix} \alpha & \times & B & = & B & \times & \alpha \\ 4.1 & \times & \begin{bmatrix} 2.7 & 3.2 \\ 3.5 & -8.2 \end{bmatrix} & = & \begin{bmatrix} 2.7 & 3.2 \\ 3.5 & -8.2 \end{bmatrix} & \times & 4.1 \end{matrix}$$

### 5.3 Transpose matrix

$$\begin{matrix} X_{ij}^T & = & X_{ji} \\ \begin{bmatrix} 2.7 & 3.2 & 5.4 \\ 3.5 & -8.2 & -1.7 \end{bmatrix}^T & = & \begin{bmatrix} 2.7 & 3.5 \\ 3.2 & -8.2 \\ 5.4 & -1.7 \end{bmatrix} \end{matrix}$$

### 5.4 Identity matrix

$$I = I_n = \text{Diag}([1, 1, \dots, 1])$$

such that

$$I \times A = A \times I$$

Examples:

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

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

## 5.5 Matrix inverse

For any square  $n \times n$  matrix  $A$ , the matrix inverse  $A^{-1}$  is defined as

$$AA^{-1} = A^{-1}A = I$$

Example:

$$\begin{bmatrix} 2.7 & 3.5 \\ 3.2 & -8.2 \end{bmatrix} \times \begin{bmatrix} 0.245 & 0.104 \\ 0.095 & -0.080 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$

$$A \quad \times \quad A^{-1} = I$$

Some matrices do not hold an inverse such as zero matrices. They are called degenerate or singular.

**Question 5: Compute the matrix transpose as above in Python. Determine also the matrix inverse in Python.**

```
In [ ]: import numpy as np

#YOUR CODE HERE

A = np.array([[ 2.7,  3.5,  3.2], [-8.2,  5.4, -1.7]])
AT = np.transpose(A)      # transpose of A

print(AT)
print(A.shape)      # size of A
print(AT.shape)     # size of AT

A = np.array([ [2.7,3.5], [3.2,-8.2] ])
Ainv = np.linalg.inv(A)  # inverse of A
AAinv = A.dot(Ainv)      # multiplication of A and A inverse
print(A)
print(A.shape)          # size of A
print(Ainv)
print(Ainv.shape)       # size of Ainv
print(AAinv)
print(AAinv.shape)      # size of AAinv

[[ 2.7 -8.2]
 [ 3.5  5.4]
 [ 3.2 -1.7]]
(2, 3)
(3, 2)
[[ 2.7  3.5]
 [ 3.2 -8.2]]
(2, 2)
[[ 0.24595081  0.104979 ]
 [ 0.0959808  -0.0809838 ]]
(2, 2)
[[ 1.00000000e+00  9.02056208e-17]
 [-3.96603366e-18  1.00000000e+00]]
(2, 2)

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```