# 1. Matrix, vector and scalar representation

#### 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:

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

 $x_i=i^{th}$  element of x. Here:  $x_1=4.1, x_3=6.4$ .

Dimension of vector x is the number of rows.

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

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

#### 1.3 Scalar

Example:

$$x = 5.6$$

A scalar has no dimension.

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

Note: x = [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.

```
1 import numpy as np
2
3 #YOUR CODE HERE
4
5 x = np.array([[4.1,5.3],[-3.9,8.4],[6.4,-1.8]])
6 print(x)
7 print(x.shape)  # size of x
8 print(type(x))  # type of x
9 print(x.dtype)  # data type of x
10
11 y = np.array(x[:,0])
```

```
12 print(y)
13 print(y.shape)  # size of y
14
15 z = 5.6
16 print(z)
17 print(np.shape(z))  # size of z
18

Discrepance [[ 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,)
       5.6
       ()
```

# 2. Matrix addition and scalar-matrix multiplication

#### 2.1 Matrix addition

Example:

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

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

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

$$3 \times 2 + 2 \times 3 = \text{Not allowed}$$

# 2.1 Scalar-matrix multiplication

Example:

$$3 imes \begin{bmatrix} 4.1 & 5.3 \\ -3.9 & 8.4 \\ 6.4 & -1.8 \end{bmatrix} = \begin{bmatrix} 3 imes 4.1 & 3 imes 5.3 \\ 3 imes -3.9 & 3 imes 8.4 \\ 3 imes 6.4 & 3 imes -1.8 \end{bmatrix}$$
To dim  $+ 3 imes 2 = 3 imes 2$ 

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

```
1 import numpy as np
2
3 #YOUR CODE HERE
4
5 X1 = np.array([[4.1,5.3],[-3.9,8.4],[6.4,-1.8]])
6 X2 = np.array([[2.7,3.5],[7.3,2.4],[5.0,2.8]])
```

```
7 X = X1 + X2 # summation of X1 and X2
 9 print(X1)
10 print(X2)
11 print(X)
12
13 Y1 = 4*X # X multiplied by 4
14 \ Y2 = X/3 \# X \ divided \ by \ 3
15
16 print(X)
17 print(Y1)
18 print(Y2)
19
[-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.]]
    ΓΓ27.2 35.27
     [13.6 43.2]
     [45.6 4.]]
    [[2.26666667 2.93333333]
    [1.13333333 3.6
         0.3333333377
```

# 3. Matric-vector multiplication

### 3.1 Example

Example:

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

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

#### 3.2 Formalization

$$egin{bmatrix} m{A} & \times & m{x} & = & m{y} \ m imes n & n imes 1 & = & m imes 1 \end{bmatrix}$$

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

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

$$\begin{bmatrix} 4.1 & 5.3 \\ -3.9 & 8.4 \\ 6.4 & -1.8 \end{bmatrix} \times \begin{bmatrix} 2.7 \\ 3.5 \\ -7.2 \end{bmatrix} = ?$$

$$3 \times 2 \times 3 \times 1 = \text{not allowe}$$

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

```
1 import numpy as np
 2
 3 #YOUR CODE HERE
 5 A = np.array([[4.1,5.3],[-3.9,8.4],[6.4,-1.8]])
 6 \times = \text{np.array}([[2.7],[3.5]])
 7 y = np.dot(A,x) # multiplication of A and x
 8 print(A)
 9 print(A.shape) # size of A
10 print(x)
11 print(x.shape) # size of x
12 print(y)
13 print(y.shape) # size of y
14
[ 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)
```

# 4. Matrix-matrix multiplication

### 4.1 Example

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

$$3 \times 2 \times 2 \times 2 = 3 \times 2$$

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

$$egin{bmatrix} m{A} & \times & m{X} & = & m{Y} \ m imes n & m imes p & = & m imes p \end{bmatrix}$$

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:

$$Y_i = A \times X_i$$
  
 $1 \times 1 = 1 \times n \times n \times 1$ 

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

```
1 import numpy as np
 2
 3 #YOUR CODE HERE
 4
 5 A = np.array([[4.1,5.3],[-3.9,8.4],[6.4,-1.8]])
 6 X = np.array([[2.7,3.2],[3.5,-8.2]])
7 Y = np.dot(A,X) # matrix multiplication of A and X
8 print(A)
9 print(A.shape) # size of A
10 print(X)
11 print(X.shape) # size of X
12 print(Y)
13 print(Y.shape) # size of Y
14
[ 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 -81.367
    Γ 10.98 35.24]]
   (3, 2)
```

# 5. Some linear algebra properties

### 5.1 Matrix multiplication is *not* commutative

# 5.2 Scalar multiplication is associative

### 5.3 Transpose matrix

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

### 5.4 Identity matrix

$$I = I_n = Diag([1,1,\ldots,1])$$

such that

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

Examples:

$$I_2 = egin{bmatrix} 1 & 0 \ 0 & 1 \end{bmatrix}$$
  $I_3 = egin{bmatrix} 1 & 0 & 0 \ 0 & 1 & 0 \ 0 & 0 & 1 \end{bmatrix}$ 

#### 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:

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

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

```
1 import numpy as np
2
3 #YOUR CODE HERE
4
5 A = np.array([[2.7,3.5,3.2],[-8.2,5.4,-1.7]])
6 AT = A.T  # transpose of A
7
8 print(AT)
9 print(A.shape)  # size of A
10 print(AT.shape)  # size of AT
11
12 A = np.array([[2.7,3.5],[3.2,-8.2]])
13 Ainv = np.linalg.inv(A)  # inverse of A
14 AAinv = np.dot(A,Ainv)  # multiplication of A and A inverse
15 print(A)
16 print(A.shape)  # size of A
```

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