

▼ 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 3×2 matrix.

The set of all 3×2 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([ 4.1, -3.9,  6.4])
12 print(y)
13 print(y.shape)    # size of y
14
15 z = np.array(5.6)
16 print(z)
17 print(z.shape)    # size of z
18
```



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

▼ 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:

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

▼ 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. ,  2.8]])
7
8 X = X1 + X2
9
10 print(X1)
11 print(X2)
12 print(X)
13
14 Y1 = np.dot(4, X)
15 Y2 = X / 3
16
17 print(X)
18 print(Y1)
19 print(Y2)
20
```



```
[[ 4.1  5.3]
 [-3.9  8.4]
 [ 6.4 -1.8]]
[[2.7 3.5]
 [7.3 2.4]
 [5.  2.8]]
[[ 0.  0.  0.  0.]
```

3.1 Example

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

3.2 Formalization

$$\begin{array}{ccc} \left[\mathbf{A} \right] & \times & \left[\mathbf{x} \right] = \left[\mathbf{y} \right] \\ m \times n & n \times 1 & = m \times 1 \end{array}$$

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

<https://colab.research.google.com/drive/1tQHbPznD97pi6oStDZL5XI5uE3rbXcX> #printMode=true

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

$3 \times 2 \quad \times \quad 3 \times 1 \quad = \quad \text{not allowed}$

▼ Question 3: Multiply the matrix and vector 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], [ 3.5]])
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

```

```

[[ 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

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

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

4.2 Formalization

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

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 :

$$\begin{array}{ccccc} Y_i & = & A & \times & X_i \\ 1 \times 1 & = & 1 \times n & \times & n \times 1 \end{array}$$

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
9 print(A)
10 print(A.shape)    # size of A
11 print(X)
12 print(X.shape)    # size of X
13 print(Y)
14 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)

```

▼ 5. Some linear algebra properties

5.1 Matrix multiplication is *not* commutative

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

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

5.2 Scalar multiplication is associative

$$\alpha \times B = B \times \alpha$$

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

5.3 Transpose matrix

$$X_{ij}^T = X_{ji}$$

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

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:

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

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

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.

```

1 import numpy as np
2 import numpy.linalg as lin
3 #YOUR CODE HERE
4
5 A = np.array([[ 2.7, -8.2], [ 3.5,  5.4], [ 3.2, -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 = lin.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
17 print(Ainv)
18 print(Ainv.shape) # size of Ainv

```

```
19 print(AAinv)
20 print(AAinv.shape)    # size of AAinv
```

```
[[ 2.7  3.5  3.2]
 [-8.2  5.4 -1.7]]
(3, 2)
(2, 3)
[[ 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)
```