

NumPy is the foundation of the Python machine learning stack.

NumPy allows for efficient operations on the data structures often used in machine learning.

The data structures are vectors, matrices, arrays.

In [20]:

```
1  #You need to create a vector.
2  #Use NumPy to create a one-dimensional array
3  import numpy as np
4
5  vector_row = np.array([[1,2,3]])
6  print(vector_row)
7  vector_col = np.array([[1],[2],[3]])
8  print(vector_col)
9  vector_col
10
11
```

```
[[1 2 3]]
[[1]
 [2]
 [3]]
```

Out[20]: array([[1],
 [2],
 [3]])

```
In [10]: 1  #Creating a two dimensional Matrix
          2  import numpy as np
          3
          4  matrix = np.array([[1,2],[3,4],[5,6]])
          5  print(matrix)
          6
          7  # other way with mat() method
          8  matrix = np.mat([[1,2],[3,4],[5,6]])
          9  matrix
```

```
[[1 2]
 [3 4]
 [5 6]]
```

```
Out[10]: matrix([[1, 2],
                  [3, 4],
                  [5, 6]])
```

In [38]:

```
1  #Selecting elements
2  import numpy as np
3
4  vector_row = np.array([1,2,3])
5  #select specific element
6  print(vector_row[2])
7
8  #select all elements upto index 2
9  print(vector_row[:2])
10
11 #select all elements
12 print(vector_row[:])
13
14 matrix = np.array([[1,2,3],
15                    [4,5,6],
16                    [7,8,9]])
17 #select element of third row and second column
18 print(matrix[2,1])
19 #Select first two rows and all columns
20 print(matrix[:2,:])
21 #Select all rows and the second column
22 print(matrix[:,1:2])
23
24
```

3

[1 2]

[1 2 3]

8

[[1 2 3]

[4 5 6]]

[[2]

[5]

[8]]

```
In [42]: 1  #Describing a Matrix in terms of shape,size and dimensions
2
3  import numpy as np
4  #create matrix
5  matrix = np.array([[1,2,3],
6                     [4,5,6],
7                     [7,8,9]])
8
9  # View number of rows and columns
10 print(matrix.shape)
11
12 # View number of elements (rows * columns)
13 print(matrix.size)
14
15 # View number of dimensions
16 print(matrix.ndim)
17
```

(3, 3)

9

2

NumPy's vectorize class converts a function into a function that can apply to all elements in an array or slice of an array

In [6]:

```
1  #You want to apply some function to multiple elements in an array.
2  #Use NumPy's vectorize
3
4  # Load library
5  import numpy as np
6
7  # Create matrix
8  matrix = np.array([[1, 2, 3],
9                     [4, 5, 6],
10                    [7, 8, 9]])
11
12 # Create function that adds 100 to something
13 add_100 = lambda i: i + 100
14
15 # Create vectorized function
16 vectorized_add_100 = np.vectorize(add_100)
17
18 # Apply function to all elements in matrix
19 print(vectorized_add_100(matrix))
20 print()
21
22 #One more simple way
23 new_matrix = matrix+100
24 print(new_matrix)
```

```
[[101 102 103]
 [104 105 106]
 [107 108 109]]
```

```
[[101 102 103]
 [104 105 106]
 [107 108 109]]
```

In [10]:

```
1  # Finding the Maximum and Minimum Values
2
3  # Load library
4  import numpy as np
5  # Create matrix
6  matrix = np.array([[1, 2, 3],
7                     [4, 5, 6],
8                     [7, 8, 9]])
9
10 # Return maximum element
11 print(np.max(matrix))
12
13 # Return minimum element
14 print(np.min(matrix))
15
16 # Find maximum element in each row axis = 0 is for row
17 print(np.max(matrix, axis=0))
18
19 # Find maximum element in each column axis = 1 is for column
20 print(np.max(matrix, axis=1))
```

9

1

[7 8 9]

[3 6 9]

In [15]:

```
1  # Calculating the Average, Variance, and Standard Deviation
2
3  # Load library
4  import numpy as np
5
6  # Create matrix
7  matrix = np.array([[1, 2, 3],
8                     [4, 5, 6],
9                     [7, 8, 9]])
10
11
12 #Return Average
13 print("Average:",np.average(matrix))
14
15 # Return mean
16 print("Mean:",np.mean(matrix))
17
18 # Return variance
19 print("Variance:",np.var(matrix))
20
21 # Return standard deviation
22 print("Standard Deviation:",np.std(matrix))
23
```

Average: 5.0

Mean: 5.0

Variance: 6.666666666666667

Standard Deviation: 2.581988897471611

In [22]:

```
1  #Transposing a Vector or Matrix by using T method
2
3  # Load Library
4  import numpy as np
5
6  # Create matrix
7  matrix = np.array([[1, 2, 3],
8                     [4, 5, 6],
9                     [7, 8, 9]])
10 #Create Vector
11 vector = np.array([[1,2,3,4,5]])
12
13 # Transpose matrix
14 print(matrix.T)
15 print("\n")
16 #Transpose Vector
17 print(vector.T)
18
```

```
[[1 4 7]
 [2 5 8]
 [3 6 9]]
```

```
[[1]
 [2]
 [3]
 [4]
 [5]]
```



```
In [23]: 1  #Flattening a Matrix means transform a matrix into a one-dimensional array by using flatten().
2
3  # Load library
4  import numpy as np
5  # Create matrix
6  matrix = np.array([[1, 2, 3],
7                     [4, 5, 6],
8                     [7, 8, 9]])
9  # Flatten matrix
10 matrix.flatten()
11
```

```
Out[23]: array([1, 2, 3, 4, 5, 6, 7, 8, 9])
```

```
In [29]: 1  #Finding the Rank of a Matrix.
2  #Use NumPy's linear algebra method matrix_rank.
3
4
5  # Load library
6  import numpy as np
7  # Create matrix
8  matrix = np.array([[1, 2, 3],
9                     [4, 5, 6],
10                    [7, 8, 9]])
11
12 #Rank of a matrix
13 print("Rank of Matrix:",np.linalg.matrix_rank(matrix))
14
15
```

```
Rank of Matrix: 2
```

In [31]:

```
1  # Getting the Diagonal elements of a Matrix using diagonal method
2  #Getting trace of the matrix. The trace of a matrix is the sum of the diagonal elements
3
4  # Load library
5  import numpy as np
6
7  # Create matrix
8  matrix = np.array([[1, 2, 3],
9                     [2, 4, 6],
10                    [3, 8, 9]])
11
12 # Return diagonal elements
13 print(matrix.diagonal())
14
15 #Return the trace
16 print(matrix.trace())
```

[1 4 9]

14

In [32]:

```
1  #operations of two matrices. Addition, Subtraction, Multiplication.
2
3  # Load library
4  import numpy as np
5
6  # Create matrix
7  matrix_a = np.array([[1, 1, 1],
8                        [1, 1, 1],
9                        [1, 1, 2]])
10
11 # Create matrix
12 matrix_b = np.array([[1, 3, 1],
13                      [1, 3, 1],
14                      [1, 3, 8]])
15
16 # Add two matrices
17 print(np.add(matrix_a, matrix_b))
18
19 # Subtract two matrices
20 print(np.subtract(matrix_a, matrix_b))
21
22 # Multiply two matrices
23 print(np.dot(matrix_a, matrix_b))
24
```

```
[[ 2  4  2]
 [ 2  4  2]
 [ 2  4 10]]
[[ 0 -2  0]
 [ 0 -2  0]
 [ 0 -2 -6]]
[[ 3  9 10]
 [ 3  9 10]
 [ 4 12 18]]
```

In [33]:

```
1  # Calculating inverse of a matrix using NumPy's linear algebra inv method.
2
3  # Load library
4  import numpy as np
5
6  # Create matrix
7  matrix = np.array([[1, 4],
8                     [2, 5]])
9
10 # Calculate inverse of matrix
11 print(np.linalg.inv(matrix))
12
13
```

```
[[-1.66666667  1.33333333]
 [ 0.66666667 -0.33333333]]
```

In [38]:

```
1  #Generating Random Values
2
3  # Load library
4  import numpy as np
5
6  # Generate three random floats between 0.0 and 1.0
7  print(np.random.random(10))
8
9  # Generate three random integers between 1 and 10
10 print(np.random.randint(0, 11, 3))
11
12 # Draw three numbers greater than or equal to 1.0 and less than 2.0
13 print(np.random.uniform(1.0, 2.0, 3))
```

```
[0.71998631 0.74220332 0.91882157 0.99343941 0.02097137 0.42885785
 0.71807372 0.82008064 0.78739958 0.3929567 ]
[10  8  2]
[1.02039526 1.74755115 1.30519769]
```

