

# MACHINE LEARNING

## Numpy

Trainer : Sujata Mohite  
Email: [sujata.mohite@sunbeaminfo.com](mailto:sujata.mohite@sunbeaminfo.com)



# Package

- Collection of python files which you want to make it reusable libraries is called as a **package**.
- When you want to make a huge library which may not be fitting into a single file and huge logic required. You can put it into multiple python files.
- these multiple python files combined together make a package
- package = library (with multiple python files)



# Package

- Python has multiple built-in packages(eg: math,os,sys) and third party packages(eg: numpy , pandas , matplotlib , sklearn)

Eg: syntax --> `python -m pip install package-name`

`python -m pip install numpy`

`python -m pip install pandas`

`python -m pip uninstall package-name`



# Numpy

- third party package (need to be installed explicitly using pip3)
- developed for optimizing memory while creating collections
- mainly used while performing statistical calculations
- faster alternative for python collections
- free and open source
- developed in C++



# Installation

- to install the numpy package on linux/macOS

- `sudo pip3 install numpy`

- to install the numpy package on windows

- `pip install numpy`



# Usage

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- to import numpy
  - import numpy as np



# ndarray

- n-dimensional array
- collection of similar values
- the memory will get allocated contiguously
- immutable in nature

## Rules

- every value in an array must be of same type
- in multi-dimensional array, every row must be having same number of columns



# Ndarray- Data Types

- data types
  - Integer
    - int8
    - int16
    - int32
    - Int64
  - float
    - float16
    - float32
    - Float64
  - string
    - unicode characters





# Ndarray- Data Types

- to change the data type, pass dtype as second parameters

```
a1 = np.array([10, 20, 30, 40, 50])  
a2 = np.array([10, 20, 30, 40, 50], dtype=np.int32)  
a3 = np.array([10, 20, 30, 40, 50], dtype=np.int16)  
a4 = np.array([10, 20, 30, 40, 50], dtype=np.int8)
```



# Ndarray-dtype

- returns the data type of value(s) stored in the array
- e.g.

```
a1 = np.array([10, 20, 30, 40, 50])  
print(a1.dtype)  
# int64
```



# Ndarray- size

- number of values in the array
- similar to len(array)
- e.g.

```
a1 = np.array([10, 20, 30, 40, 50])  
print(a1.size)  
# 5
```



# Ndarray- itemsize

- number of bytes needed to store every value
- e.g.

```
a1 = np.array([10, 20, 30, 40, 50])  
print(a1.itemsize) # 8
```

```
a2 = np.array([10, 20, 30, 40, 50], dtype=np.int32)  
print(a2.itemsize) # 4
```

```
a3 = np.array([10, 20, 30, 40, 50], dtype=np.int16)  
print(a3.itemsize) # 2
```

```
a4 = np.array([10, 20, 30, 40, 50], dtype=np.int8)  
print(a4.itemsize) # 1
```



# Ndarray- nbytes

- return the size requirement of the array
- e.g

```
a1 = np.array([10, 20, 30, 40, 50])  
print(a1.nbytes)    # 40 => 8 bytes * 5
```



# Ndarray- ndim

- returns the dimensions of the array
- e.g

```
a1 = np.array([10, 20, 30, 40, 50])
```

```
print(a1.ndim)          # 1
```

```
▪
```

```
a2 = np.array([[10, 20], [30, 40]])
```

```
print(a2.ndim)          # 2
```



# Indexing

- Array (operations / functions)
- array used to create an array
- e.g.

```
a1 = np.array([10, 20, 30])  
print(a1) # [10 20 30]
```



# Indexing - positive indexing

- indexing using positive index values
- e.g:

```
a1 = np.array([10, 20, 30, 40, 50])  
print(f"value at index 0: {a1[0]}")           # 10  
print(f"value at index 1: {a1[1]}")           # 20  
print(f"value at index 2: {a1[2]}")           # 30  
print(f"value at index 3: {a1[3]}")           # 40  
print(f"value at index 4: {a1[4]}")           # 50
```





# Indexing - negative indexing

- indexing using negative index values
- -1 always represents the last value in an array
- -(len(array)) always represents the first value in an array
- e.g.

```
a1 = np.array([10, 20, 30, 40, 50])  
print(f"value at index -1: {a1[-1]}") # 50  
print(f"value at index -2: {a1[-2]}") # 40  
print(f"value at index -3: {a1[-3]}") # 30  
print(f"value at index -4: {a1[-4]}") # 20  
print(f"value at index -5: {a1[-5]}") # 10
```



# Slicing

- getting a slice using range of values
- e.g.

```
a1 = np.array([10, 20, 30, 40, 50])
```

```
print(a1[2:4])           # [30 40]  
print(a1[2:])            # [30 40 50]  
print(a1[:4])            # [10 20 30 40]  
print(a1[:])             # [10 20 30 40 50]
```



# Slicing – array indexing

- getting a slice (portion) of an array by using array of index values
- e.g.

```
a1 = np.array([10, 20, 30, 40, 50])  
print(a1[[2, 3]])           # [30 40]
```



# Slicing-Boolean indexing

- getting a slice (portion) of an array by using array of boolean values representing required positions
- e.g

```
a1 = np.array([10, 20, 30, 40, 50])  
print(a1[[False, False, True, True, False]])      # [30 40]  
print(a1[[False, True, False, True, True]])        # [20 40 50]
```



# Filtering

- getting the required values based on certain conditions
- e.g.

```
salaries = np.array([10000, 15000, 12000, 30000, 45000, 50000, 55000,  
120000, 150000])
```

```
print(salaries[salaries > 100000])          # [120000 150000]
```



# Ndarray- shape

- returns number of values in the respective dimension
- e.g

```
a1 = np.array([10, 20, 30, 40, 50])  
print(a1.shape) # (5,)
```

```
a2 = np.array([[10, 20], [30, 40]])  
print(a2.shape) # (2, 2)
```

```
a3 = np.array([[10, 20, 30], [40, 50, 60]])  
print(a3.shape) # (2, 3)
```



# Ndarray- reshape

- used to convert shape of an array
- e.g.

```
a1 = np.array([[10, 20, 30], [40, 50, 60]])
print(f"shape of a1 = {a1.shape}") # (2, 3)
# [
#  [10 20 30]
#  [40 50 60]
# ]
```

```
a2 = a1.reshape([3, 2])
print(f"shape of a2 = {a2.shape}") # (3, 2)
# [
#  [10 20]
#  [30 40]
#  [50 60]
# ]
```

```
a3 = a1.reshape([1, 6])
print(f"shape of a3 = {a3.shape}") # (1, 6)
# [
#  [10 20 30 40 50 60]
# ]
```

```
a4 = a1.reshape([6, 1])
print(f"shape of a4 = {a4.shape}") # (6, 1)
# [
#  [10]
#  [20]
#  [30]
#  [40]
#  [50]
#  [60]
# ]
```



# Ndarray- arange

- used to create an array with consecutive numbers starting from first to the last value
- note: the upper bound will never be included in the array
- e.g.

```
a1 = np.arange(1, 5)
print(a1)           # [1 2 3 4]
```

```
a1 = np.arange(1, 5, 2)
print(a1)           # [1 3]
```





# Ndarray- ones

- used to create an array with all values set to one
- e.g.

```
a1 = np.ones(5, dtype=int8)
print(a1)
```

```
# [1 1 1 1 1]
```

```
a2 = np.ones([2, 2], dtype=int8)
print(a2)
```

```
# [
#  [1 1]
#  [1 1]
# ]
```



# Ndarray - zeros

- used to create an array with all values set to zero
- e.g.

```
a1 = np.zeros(5)
print(a1)
```

```
# [0 0 0 0 0]
```

```
a2 = np.zeros([2, 2], dtype=int8)
print(a2)
```

```
# [
#  [0 0]
#  [0 0]
# ]
```



# Ndarray - random

- used to create a random array
- e.g.

```
a1 = np.random.random(5)
print(a1)      # [0.344 0.1234 0.42 0.5223 0.7777]
```

```
a2 = np.random.randint(1, 10, 5)
      # any five random numbers from 1 to 10
```

```
print(a2)      # [1 5 2 7 4]
```



# Mathematical Operations

- all mathematical operations are supported on numpy array
- e.g.

```
a1 = np.array([1, 2, 3])  
a2 = np.array([1, 2, 3])  
  
print(a1 + a2) # [2 4 6]  
print(a1 - a2) # [0 0 0]  
print(a1 / a2) # [1. 1. 1.]  
print(a1 // a2) # [1 1 1]  
print(a1 * a2) # [1 4 9]  
print(a1 % a2) # [0 0 0]  
print(a1 ** 3) # [1 8 27]
```



# Broadcast operations

- performing an operation on every member of an array
- e.g.

```
a1 = np.array([1, 2, 3])
```

```
print(a1 + 10)    # [11 12 13]
print(a1 - 10)    # [-9 -8 -7]
print(a1 / 10)    # [0.1 0.2 0.3]
print(a1 // 10)   # [0 0 0]
print(a1 * 10)    # [10 20 30]
print(a1 ** 2)    # [1 4 9]
print(a1 % 10)    # [1 2 3]
```



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**Thank You!!**

