

# NumPy Library

- What is NumPy?
  - NumPy stands for Numerical Python.
  - NumPy was created in 2005 by Travis Oliphant. It is an open source project and you can use it freely.
  - NumPy is a Python library used for working with arrays.
  - It also has functions for working in domain of linear algebra, fourier transform, and matrices.
- Why Use NumPy?
  - In Python we have lists that serve the purpose of arrays, but they are slow to process.
  - NumPy aims to provide an array object that is up to 50x faster than traditional Python lists.
  - The array object in NumPy is called ndarray
    - Numpy provides a lot of supporting functions that make working with ndarray very easy.

# NumPy Library

- Why is NumPy Faster Than Lists?
  - NumPy arrays are stored at one continuous place in memory unlike lists, so processes can access and manipulate them very efficiently.
  - This behavior is called locality of reference in computer science.
  - This is the main reason why NumPy is faster than lists. Also it is optimized to work with latest CPU architectures.
- Which Language is NumPy written in?
  - NumPy is a Python library and is written partially in Python, but most of the parts that require fast computation are written in C or C++.

# Installation of NumPy

```
C:\Users\Your Name>pip install numpy
```

Checking NumPy Version

```
import numpy as np  
  
print(np.__version__)
```

# Creating Array

- 0-D or Scalar Array
  - `arr = np.array(42)`
- 1-D Arrays:
  - An array that has 0-D arrays as its elements is called uni-dimensional or 1-D array.
  - `arr = np.array([1, 2, 3, 4, 5])`
- 2-D Arrays
  - An array that has 1-D arrays as its elements is called a 2-D array.
  - `arr = np.array([[1, 2, 3],  
[4, 5, 6]])`

# Creating Array

- 3-D arrays
  - An array that has 2-D arrays (matrices) as its elements is called 3-D array.

```
arr = np.array([[[1, 2, 3],  
                 [4, 5, 6]],  
               [[1, 2, 3],  
                [4, 5, 6]]])  
  
print(arr)
```

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


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

# Check Dimension of Array

```
import numpy as np

a = np.array(42)
b = np.array([1, 2, 3, 4, 5])
c = np.array([[1, 2, 3], [4, 5, 6]])
d = np.array([[[1, 2, 3], [4, 5, 6]], [[1, 2, 3], [4, 5, 6]]])

print(a.ndim)    0
print(b.ndim)    1
print(c.ndim)    2
print(d.ndim)    3
```



**Output**

# Creating Arrays

```
import numpy as np  
arr = np.array([1, 2, 3, 4, 5])  
print(type(arr))
```

```
<class 'numpy.ndarray'>
```

- To create an ndarray, we can pass a list, tuple or any array-like object into the array() method, and it will be converted into an ndarray:

```
arr = np.array((1, 2, 3, 4, 5))
```

# Indexing

## Access 1-D Array Element

```
import numpy as np
arr = np.array([1, 2, 3, 4])
print("Element at location 1 : ", arr[1])
print("Addition : ", arr[0] + arr[1])
```

```
Element at location 1 :  2
Addition :  3
```

## Access 2-D Array Element

```
import numpy as np
arr = np.array([[1, 2, 3],
                [10, 11, 12]])
print('0th Element of 1st Row: ', arr[1, 0])
```

```
0th Element of 1st Row:  10
```



# Indexing

## Access 3-D Array Element

```
import numpy as np
arr = np.array([[[1, 2, 3],
                  [4, 5, 6]],
                [[7, 8, 9],
                  [10, 11, 12]]])
print("1st 2-D Matrix, 1st Row and Col. 0 : ",arr[1, 1, 0])
```

1st 2-D Matrix, 1st Row and Col. 0 : 10

## Negative Indexing

```
import numpy as np
arr = np.array([[[1, 2, 3],
                  [7, 8, 9]])
print('Last element of 1st Row : ', arr[1, -1])
```

Last element of 1st Row : 9

# 1-D Array Slicing

```
import numpy as np
arr = np.array([0, 1, 2, 3, 4, 5, 6, 7])
print("Element from index 1 to 4 :", arr[1:5])
print("Element from index 5 onwards :", arr[5:])
print("Element from index 0 upto 3 :", arr[:4])
print("Elements using Negative indices :", arr[-4:-1])
print("Elements at even locations :", arr[0: len(arr): 2])
print("Reverse order of elements :", arr[::-1])
print("Reverse order of elements :", arr[6:2:-1])
```

```
Element from index 1 to 4 : [1 2 3 4]
Element from index 5 onwards : [5 6 7]
Element from index 0 upto 3 : [0 1 2 3]
Elements using Negative indices : [4 5 6]
Elements at even locations : [0 2 4 6]
Reverse order of elements : [7 6 5 4 3 2 1 0]
Reverse order of elements : [6 5 4 3]
```

# 2-D Array Slicing

```
import numpy as np
arr = np.array([[1, 2, 3, 4],
                [5, 6, 7, 8],
                [9, 10, 11, 12]])
print("Column 3 of Row 0 and 1 : ", arr[0:2, 3])
print("Column 0, 1, 2 of Row 1 and 2 : \n", arr[1:3, 0:3])
```

```
Column 3 of Row 0 and 1 :  [4 8]
Column 0, 1, 2 of Row 1 and 2 :
[[ 5  6  7]
 [ 9 10 11]]
```

# NumPy Array Iteration

```
import numpy as np
arr = np.array([7, 8, 9, 10])
for element in arr:
    print(element)
```

```
7
8
9
10
import numpy as np
arr = np.array([[1, 2, 3],
                [4, 5, 6]])
for element in arr:
    print(element)
```

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

```
import numpy as np
arr = np.array([[[1, 2, 3],
                  [4, 5, 6]],
                [[7, 8, 9],
                  [10, 11, 12]]])
for element in arr:
    print("Element : \n", element)
```

```
Element :
[[1 2 3]
 [4 5 6]]
Element :
[[ 7  8  9]
 [10 11 12]]
```



# NumPy Array Iteration

```
import numpy as np
arr = np.array([[1, 2],
                [3, 4]])
for row in arr:
    for element in row:
        print(element)
```

1  
2  
3  
4

  
**Iteration using nditer**

```
import numpy as np

arr = np.array([[[1, 2],
                  [3, 4]],
                [[5, 6],
                  [7, 8]]])

for x in np.nditer(arr):
    print(x)
```

1  
2  
3  
4  
5  
6  
7  
8

# NumPy Array Iteration

```
import numpy as np
arr = np.array([[1, 2, 3, 4],
                [5, 6, 7, 8]])
for x in np.nditer(arr[:, ::2]):
    print(x)
```

1  
3  
5  
7

**Iteration using ndenumerate**



```
import numpy as np
arr = np.array([[1, 2],
                [5, 6]])
for idx, x in np.ndenumerate(arr):
    print(idx, x)
```

(0, 0) 1  
(0, 1) 2  
(1, 0) 5  
(1, 1) 6

# Data Types in NumPy

- List of data types in NumPy and characters used to represent them:
  - i - integer
  - b - boolean
  - u - unsigned integer
  - f - float
  - c - complex float
  - m - timedelta
  - M - datetime
  - O - object
  - S - string
  - U - unicode string
  - V - fixed chunk of memory for other type ( void )

# Data Types in NumPy

```
import numpy as np
arr = np.array([1, 2, 3, 4])
print(arr.dtype)
```

*The NumPy array object has a property called **dtype** that returns the data type of the array*

int64

*We use the **array()** function to create arrays, this function can take an optional argument: **dtype** that allows us to define the expected data type of the array elements*

```
import numpy as np
arr = np.array([1, 2, 3, 4], dtype='f2')
print(arr)
print(arr.dtype)
```

[1. 2. 3. 4.]  
float16

*For **i**, **u**, **f**, **S** and **U** we can define size as well*



# Data Types in NumPy

```
import numpy as np
arr = np.array(['pineapple', 'banana', 'cherry'])
print(arr.dtype)
print(arr)
```

<U9

['pineapple' 'banana' 'cherry']

# Data Types in NumPy

```
import numpy as np
arr = np.array([10, 2, 3, 4], dtype='S')
print(arr)
print(arr.dtype)
```

```
[b'10' b'2' b'3' b'4']
|S2
```

```
import numpy as np
arr = np.array(['1', '2', '3'], dtype='i')
print(arr)
```

```
[1 2 3]
```

# Data Types in NumPy

*If a type is given in which elements can't be casted then NumPy will raise a **ValueError**.*

```
import numpy as np
arr = np.array(['a', '2', '3'], dtype='i')
print(arr)
```

-----  
**ValueError**

Traceback

<ipython-input-26-bc5f30a52b19> in <module>

```
1 import numpy as np
----> 2 arr = np.array(['a', '2', '3'], dtype='i')
      3 print(arr)
```

# Data Types in NumPy

```
import numpy as np
arr = np.array([1.1, 2.1, 3.1])
newarr = arr.astype('i')
print(newarr)
print(newarr.dtype)
```

```
[1 2 3]
int32
```

*The astype() function creates a copy of the array, and allows you to specify the data type as a parameter.*

```
import numpy as np
arr = np.array([1, 0, 3, -1])
newarr = arr.astype(float)
print(newarr)
print(newarr.dtype)
```

```
[ 1.  0.  3. -1.]
float64
```

# Shape of Array

```
import numpy as np
arr = np.array([[[1, 2, 5],
                 [3, 4, 5]],
               [[5, 6, 5],
                [7, 8, 5]]])
print(arr.shape)
```

(2, 2, 3)

The **shape** of an array is the number of elements in each dimension.

NumPy arrays have an attribute called **shape** that returns a tuple with each index having the number of corresponding elements.

```
import numpy as np
arr = np.array([1, 2, 3, 4], ndmin=2)
print(arr)
print('shape of array :', arr.shape)

arr = np.array([1, 2, 3, 4], ndmin=3)
print(arr)
print('shape of array :', arr.shape)
```

[[1 2 3 4]]

shape of array : (1, 4)

[[[1 2 3 4]]]

shape of array : (1, 1, 4)

Create an array with 2/3 dimensions using **ndmin** using a vector with values 1,2,3,4 and verify that last dimension has value 4

# Reshape Array

- Reshaping means changing the shape of an array.
- The shape of an array is the number of elements in each dimension.
- By reshaping we can add or remove dimensions or change number of elements in each dimension.

```
import numpy as np
arr = np.array([1, 2, 3, 4, 5, 6])
newarr = arr.reshape(2, 3)
print(newarr)
```

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

*Convert the following 1-D array with 12 elements into a 2-D array.*

# Joining Arrays

- We pass a sequence of arrays that we want to join to the concatenate() function, along with the axis.

```
import numpy as np
arr1 = np.array([[1, 2], [3, 4]])
arr2 = np.array([[4, 5], [6, 7]])
arr = np.concatenate((arr1, arr2),
                      axis=0)

print(arr)
```

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

*Join two 2-D arrays along  
columns (axis=0)*

```
import numpy as np
arr1 = np.array([[1, 2], [3, 4]])
arr2 = np.array([[4, 5], [6, 7]])
arr = np.concatenate((arr1, arr2),
                      axis=1)

print(arr)
```

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

*Join two 2-D arrays along rows (axis=1)*



# Joining Arrays

- We pass a sequence of arrays that we want to join to `stack()` method along with the axis.
- Stacking is same as concatenation, only difference is that stacking is done along new axis.

```
import numpy as np
arr1 = np.array([[1, 2], [3, 4]])
arr2 = np.array([[4, 5], [6, 7]])
arr = np.stack((arr1, arr2),
               axis=0)

print(arr)
```

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

 [[4 5]
  [6 7]]]
```

```
import numpy as np
arr1 = np.array([[1, 2], [3, 4]])
arr2 = np.array([[4, 5], [6, 7]])
arr = np.stack((arr1, arr2),
               axis=1)

print(arr)
```

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

 [[3 4]
  [6 7]]]
```



# Joining Arrays

- NumPy provides a helper function: `hstack()` to stack along rows.
- NumPy provides a helper function: `vstack()` to stack along columns.

```
import numpy as np
arr1 = np.array([[1, 2], [3, 4]])
arr2 = np.array([[4, 5], [6, 7]])
arr = np.hstack((arr1, arr2))
print(arr)
```

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

```
import numpy as np
arr1 = np.array([[1, 2], [3, 4]])
arr2 = np.array([[4, 5], [6, 7]])
arr = np.vstack((arr1, arr2))
print(arr)
```

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

# Sorting Arrays

- The NumPy ndarray object has a function called `sort()`, that will sort a specified array.

```
import numpy as np
arr = np.array([1, 3, 2, 6, 4])
arr = np.sort(arr)
print(arr)
```

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

```
import numpy as np
arr = np.array([[1, 6, 2], [12, 5, 8]])
arr = np.sort(arr)
print(arr)
```

```
[[ 1  2  6]
 [ 5  8 12]]
```

# Pandas

- What is Pandas?
  - Pandas is a Python library used for working with data sets.
  - It has functions for analyzing, cleaning, exploring, and manipulating data.
  - The name "Pandas" has a reference to both "Panel Data", and "Python Data Analysis" and was created by Wes McKinney in 2008.
- Why Use Pandas?
  - Pandas allows us to analyze big data and make conclusions based on statistical theories.
  - Pandas can clean messy data sets, and make them readable and relevant.
  - Relevant data is very important in data science.

# Pandas

- What Can Pandas Do?
  - Pandas gives you answers about the data. Like:
    - Is there a correlation between two or more columns?
  - What is average value?
  - Max value?
  - Min value?
  - Pandas are also able to delete rows that are not relevant, or contains wrong values, like empty or NULL values. This is called cleaning the data.

# Pandas installation

In cmd execute command : `pip install pandas`

```
Command Prompt
Microsoft Windows [Version 10.0.19043.2251]
(c) Microsoft Corporation. All rights reserved.

C:\Users\Admin>pip install pandas
Collecting pandas
  Downloading pandas-1.5.1-cp310-cp310-win_amd64.whl (10.4 MB)
    ----- 10.4/10.4 MB 355.7 kB/s eta 0:00:00
Collecting python-dateutil>=2.8.1
  Downloading python_dateutil-2.8.2-py2.py3-none-any.whl (247 kB)
    ----- 247.7/247.7 kB 389.9 kB/s eta 0:00:00
Collecting pytz>=2020.1
  Downloading pytz-2022.6-py2.py3-none-any.whl (498 kB)
    ----- 498.1/498.1 kB 385.3 kB/s eta 0:00:00
Requirement already satisfied: numpy>=1.21.0 in c:\users\admin\appdata\local\programs\python\python310\lib\site-packages
(from pandas) (1.23.4)
Collecting six>=1.5
  Downloading six-1.16.0-py2.py3-none-any.whl (11 kB)
Installing collected packages: pytz, six, python-dateutil, pandas
Successfully installed pandas-1.5.1 python-dateutil-2.8.2 pytz-2022.6 six-1.16.0

C:\Users\Admin>
```

# Pandas Series

- What is a Series?
  - A Pandas Series is like a column in a table.
  - It is a one-dimensional array holding data of any type.

```
import pandas as pd
a = [1, 7, 2]
myvar = pd.Series(a)
print(myvar)
```

```
0    1
1    7
2    2
dtype: int64
```

```
import pandas as pd
a = [1, 7, 2]
myvar = pd.Series(a, index=["x", "y", "z"])
print(myvar)
```

```
x    1
y    7
z    2
dtype: int64
```

# Access Series Elements

```
import pandas as pd
a = [1, 7, 2]
myvar = pd.Series(a, index=["x", "y", "z"])
print(myvar[0], myvar['z'])
```

1 2

**Key/Value Objects as Series**



```
import pandas as pd
calories = {"A": 420, "B": 380, "C": 390}
myvar = pd.Series(calories)
print(myvar)
```

```
A    420
B    380
C    390
dtype: int64
```

# Data Frames

- DataFrames
  - Data sets in Pandas are usually multi-dimensional tables, called DataFrames.
  - Series is like a column, a DataFrame is the whole table.

```
import pandas as pd
data = {
    "Name": ['Ajay', 'Raman', 'Umesh'],
    "Grade": [9.4, 9.2, 9.6]
}
df = pd.DataFrame(data)
print(df)
```

	Name	Grade
0	Ajay	9.4
1	Raman	9.2
2	Umesh	9.6



# Access Elements of DataFrame

```
import pandas as pd
data = {
    "Name": ['Ajay', 'Raman', 'Umesh'],
    "Grade": [9.4, 9.2, 9.6]
}
df = pd.DataFrame(data)
print(df.loc[0])
```

```
Name      Ajay
Grade      9.4
Name: 0, dtype: object
```

```
import pandas as p
data = {
    "Name" : ['Taru', 'Siraj', 'Aneeqa'],
    "grades" : [8.9, 9.6, 8]
}
df = p.DataFrame(data)
print(df)

print(df.loc[0:1])
```

```
      Name  grades
0   Taru    8.9
1  Siraj    9.6
2 Aneeqa    8.0
      Name  grades
0   Taru    8.9
1  Siraj    9.6
```

*Pandas use the **loc** attribute to return one or more specified row(s)*

# Create Index for DataFrame

```
import pandas as pd
data = {
    "Name": ['Ajay', 'Raman', 'Umesh'],
    "Grade": [9.4, 9.2, 9.6]
}
df = pd.DataFrame(data,
                   index=[101, 102, 103])
print(df)
```

	Name	Grade
101	Ajay	9.4
102	Raman	9.2
103	Umesh	9.6

```
import pandas as pd
data = {
    "Name": ['Ajay', 'Raman', 'Umesh'],
    "Grade": [9.4, 9.2, 9.6]
}
df = pd.DataFrame(data,
                   index=[101, 102, 103])
print(df.loc[101])
```

Name	Ajay
Grade	9.4

Name: 101, dtype: object

# Manipulating DataFrame

```
import pandas as pd
data = {
    "r_no": [1, 2, 3, 4, 5, 6, 7],
    "Grade": [9.4, 9.2, 9.6,
              8.2, 8.8, 9.5, 8.4]
}
df = pd.DataFrame(data)
print(df)
```

	r_no	Grade
0	1	9.4
1	2	9.2
2	3	9.6
3	4	8.2
4	5	8.8
5	6	9.5
6	7	8.4

```
import pandas as pd
data = {
    "r_no": [1, 2, 3, 4, 5, 6, 7],
    "Grade": [9.4, 9.2, 9.6,
              8.2, 8.8, 9.5, 8.4]
}
df = pd.DataFrame(data)
df = df.rename(columns={'r_no': 'Roll_no'})
print(df)
```

	Roll_no	Grade
0	1	9.4
1	2	9.2
2	3	9.6
3	4	8.2
4	5	8.8
5	6	9.5
6	7	8.4

# Manipulating DataFrame

```
import pandas as pd
data = {
    "r_no": [1, 2, 3, 4, 5, 6, 7],
    "Grade": [9.4, 9.2, 9.6,
              8.2, 8.8, 9.5, 8.4]
}
df = pd.DataFrame(data)
df = df.sort_values(by=['Grade'],
                    ascending=False)
print(df)
```

	r_no	Grade
2	3	9.6
5	6	9.5
0	1	9.4
1	2	9.2
4	5	8.8
6	7	8.4
3	4	8.2

```
import pandas as pd
data = {
    "r_no": [1, 2, 3, 4, 5, 6, 7],
    "Grade": [9.4, 9.3, 9.6,
              8.2, 8.8, 9.4, 8.4]
}
df = pd.DataFrame(data)
df = df.sort_values(by=['Grade', 'r_no'],
                    ascending=[True, False])
print(df)
```

	r_no	Grade
3	4	8.2
6	7	8.4
4	5	8.8
1	2	9.3
5	6	9.4
0	1	9.4
2	3	9.6



# Manipulating DataFrame

```
import pandas as pd
data = {
    "r_no": [1, 2, 3, 4, 5, 6, 7],
    "Grade": [9.4, 9.3, 9.6,
              8.2, 8.8, 9.4, 8.4]
}
df = pd.DataFrame(data)

def Calculate_Marks(a):
    return a*10

df['Marks'] = df['Grade'].apply(Calculate_Marks)
print(df)
```

	r_no	Grade	Marks
0	1	9.4	94.0
1	2	9.3	93.0
2	3	9.6	96.0
3	4	8.2	82.0
4	5	8.8	88.0
5	6	9.4	94.0
6	7	8.4	84.0

# Manipulating DataFrame

```
import pandas as pd
data = {
    "r_no": [1, 2, 3, 4, 5, 6, 7],
    "Grade": [9.4, 9.3, 9.6,
              8.2, 8.8, 9.4, 8.4]
}
df = pd.DataFrame(data)

def Increase_Grades(a):
    return a + 0.2

df['Grade'] = df['Grade'].apply(Increase_Grades)
print(df)
```

	r_no	Grade
0	1	9.6
1	2	9.5
2	3	9.8
3	4	8.4
4	5	9.0
5	6	9.6
6	7	8.6

# DataFrame: Apply Filter

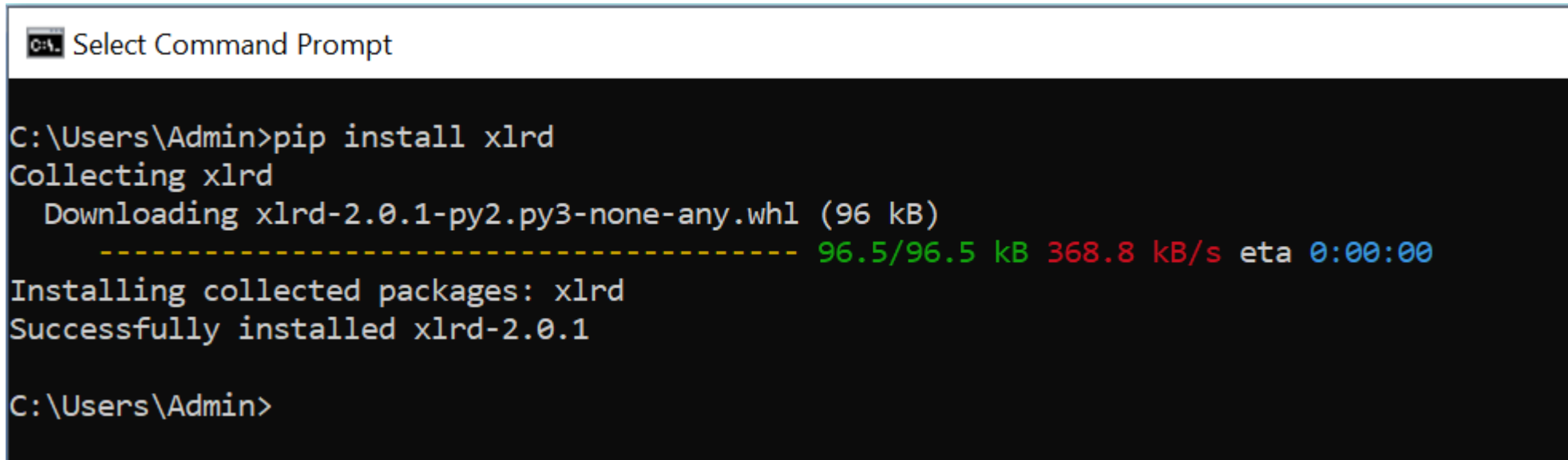
- Select Students with Grade above 9.0

```
import pandas as pd
data = {
    "r_no": [1, 2, 3, 4, 5, 6, 7],
    "Grade": [9.4, 9.3, 9.6,
              8.2, 8.8, 9.4, 8.4]
}
df = pd.DataFrame(data)
print(df[df.Grade > 9.0])
```

	r_no	Grade
0	1	9.4
1	2	9.3
2	3	9.6
5	6	9.4

# Excel file data read module installation

In cmd execute command : `pip install xlrd`



```
C:\Users\Admin>pip install xlrd
Collecting xlrd
  Downloading xlrd-2.0.1-py2.py3-none-any.whl (96 kB)
    ----- 96.5/96.5 kB 368.8 kB/s eta 0:00:00
Installing collected packages: xlrd
Successfully installed xlrd-2.0.1

C:\Users\Admin>
```



# Excel file data read module installation

In cmd execute command : pip install openpyxl

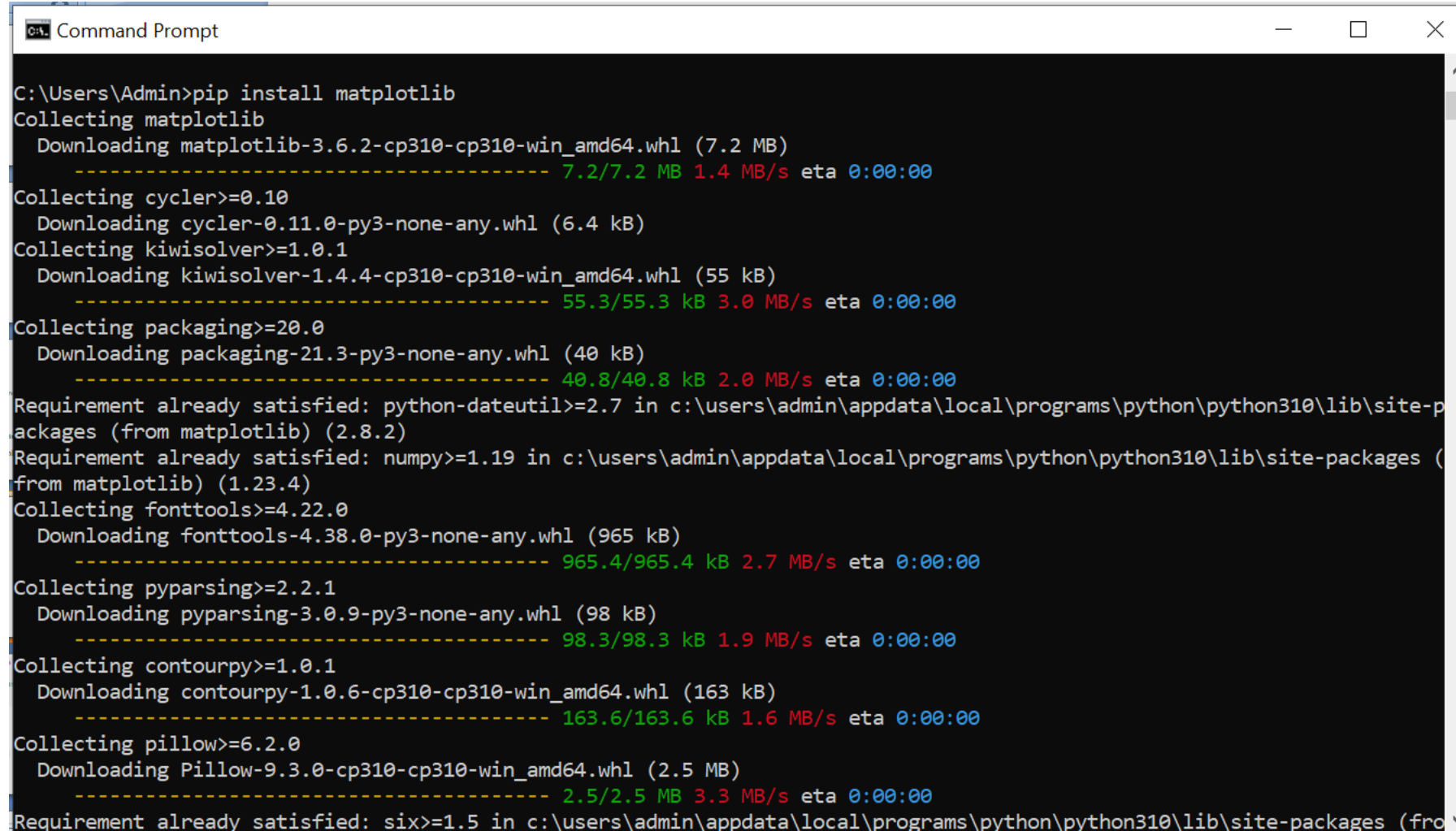
```
y C:\Users\Admin>pip install openpyxl
5> Collecting openpyxl
  Downloading openpyxl-3.0.10-py2.py3-none-any.whl (242 kB)
  ----- 242.1/242.1 kB 42.3 kB/s eta 0:00:00
Collecting et_xmlfile
  Downloading et_xmlfile-1.1.0-py3-none-any.whl (4.7 kB)
Installing collected packages: et_xmlfile, openpyxl
Successfully installed et_xmlfile-1.1.0 openpyxl-3.0.10
```

# Matplotlib

- What is Matplotlib?
  - Matplotlib is a low level graph plotting library in python that serves as a visualization utility.
  - Matplotlib was created by John D. Hunter.
  - Matplotlib is open source and we can use it freely.
  - Matplotlib is mostly written in python,
    - a few segments are written in C, Objective-C and Javascript for Platform compatibility.
- Most of the Matplotlib utilities lies under the pyplot submodule, and are usually imported under the plt alias:
  - `import matplotlib.pyplot as plt`

# Matplotlib installation

In cmd execute command : `pip install matplotlib`



```
C:\Users\Admin>pip install matplotlib
Collecting matplotlib
  Downloading matplotlib-3.6.2-cp310-cp310-win_amd64.whl (7.2 MB)
    ----- 7.2/7.2 MB 1.4 MB/s eta 0:00:00
Collecting cyclor>=0.10
  Downloading cyclor-0.11.0-py3-none-any.whl (6.4 kB)
Collecting kiwisolver>=1.0.1
  Downloading kiwisolver-1.4.4-cp310-cp310-win_amd64.whl (55 kB)
    ----- 55.3/55.3 kB 3.0 MB/s eta 0:00:00
Collecting packaging>=20.0
  Downloading packaging-21.3-py3-none-any.whl (40 kB)
    ----- 40.8/40.8 kB 2.0 MB/s eta 0:00:00
Requirement already satisfied: python-dateutil>=2.7 in c:\users\admin\appdata\local\programs\python\python310\lib\site-packages (from matplotlib) (2.8.2)
Requirement already satisfied: numpy>=1.19 in c:\users\admin\appdata\local\programs\python\python310\lib\site-packages (from matplotlib) (1.23.4)
Collecting fonttools>=4.22.0
  Downloading fonttools-4.38.0-py3-none-any.whl (965 kB)
    ----- 965.4/965.4 kB 2.7 MB/s eta 0:00:00
Collecting pyparsing>=2.2.1
  Downloading pyparsing-3.0.9-py3-none-any.whl (98 kB)
    ----- 98.3/98.3 kB 1.9 MB/s eta 0:00:00
Collecting contourpy>=1.0.1
  Downloading contourpy-1.0.6-cp310-cp310-win_amd64.whl (163 kB)
    ----- 163.6/163.6 kB 1.6 MB/s eta 0:00:00
Collecting pillow>=6.2.0
  Downloading Pillow-9.3.0-cp310-cp310-win_amd64.whl (2.5 MB)
    ----- 2.5/2.5 MB 3.3 MB/s eta 0:00:00
Requirement already satisfied: six>=1.5 in c:\users\admin\appdata\local\programs\python\python310\lib\site-packages (from
```

# Types of Charts

- Pie Chart
  - With Pyplot, you can use the `pie()` function to draw pie charts

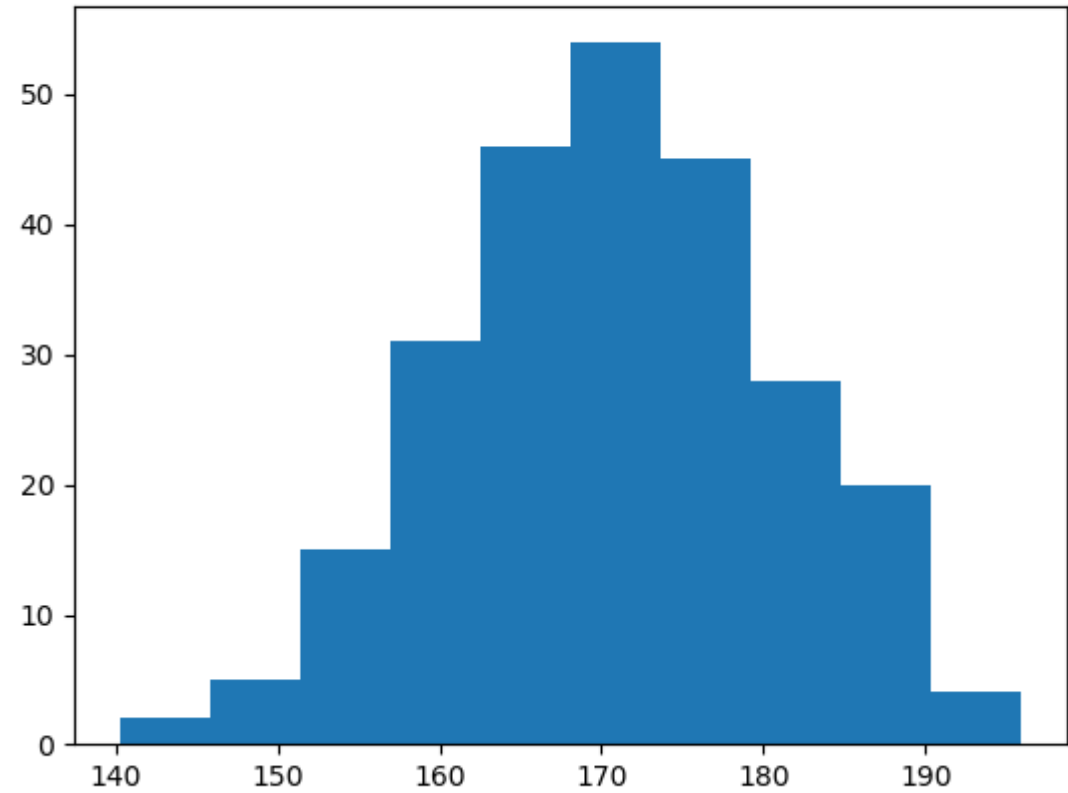
```
import matplotlib.pyplot as plt  
import numpy as np
```

```
y = np.array([35, 25, 25, 15])  
plt.pie(y)  
plt.show()
```



# Types of Charts

- Histogram
  - A histogram is a graph showing *frequency* distributions.
  - It is a graph showing the number of observations within each given interval.
  - Example: Say you ask for the height of 250 people, you might end up with a histogram like this:



*You can read from the histogram that there are approximately:*

*2 people from 140 to 145cm*

*5 people from 145 to 150cm*

*15 people from 151 to 156cm*

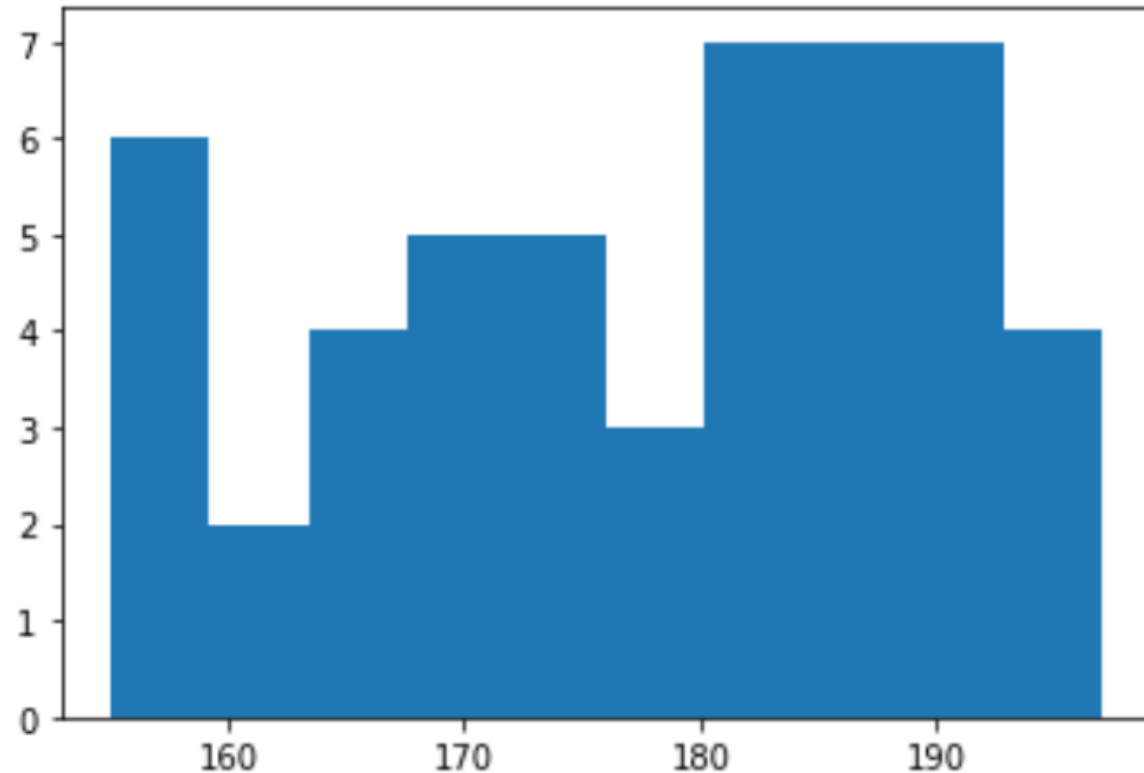
# Types of Charts

## *Plotting Histogram using hist()*

The `hist()` function will use an array of numbers to create a histogram, the array is sent into the function as an argument.

```
import matplotlib.pyplot as plt
import numpy as np

x = np.random.randint(150, 200, 50)
plt.hist(x)
plt.show()
```

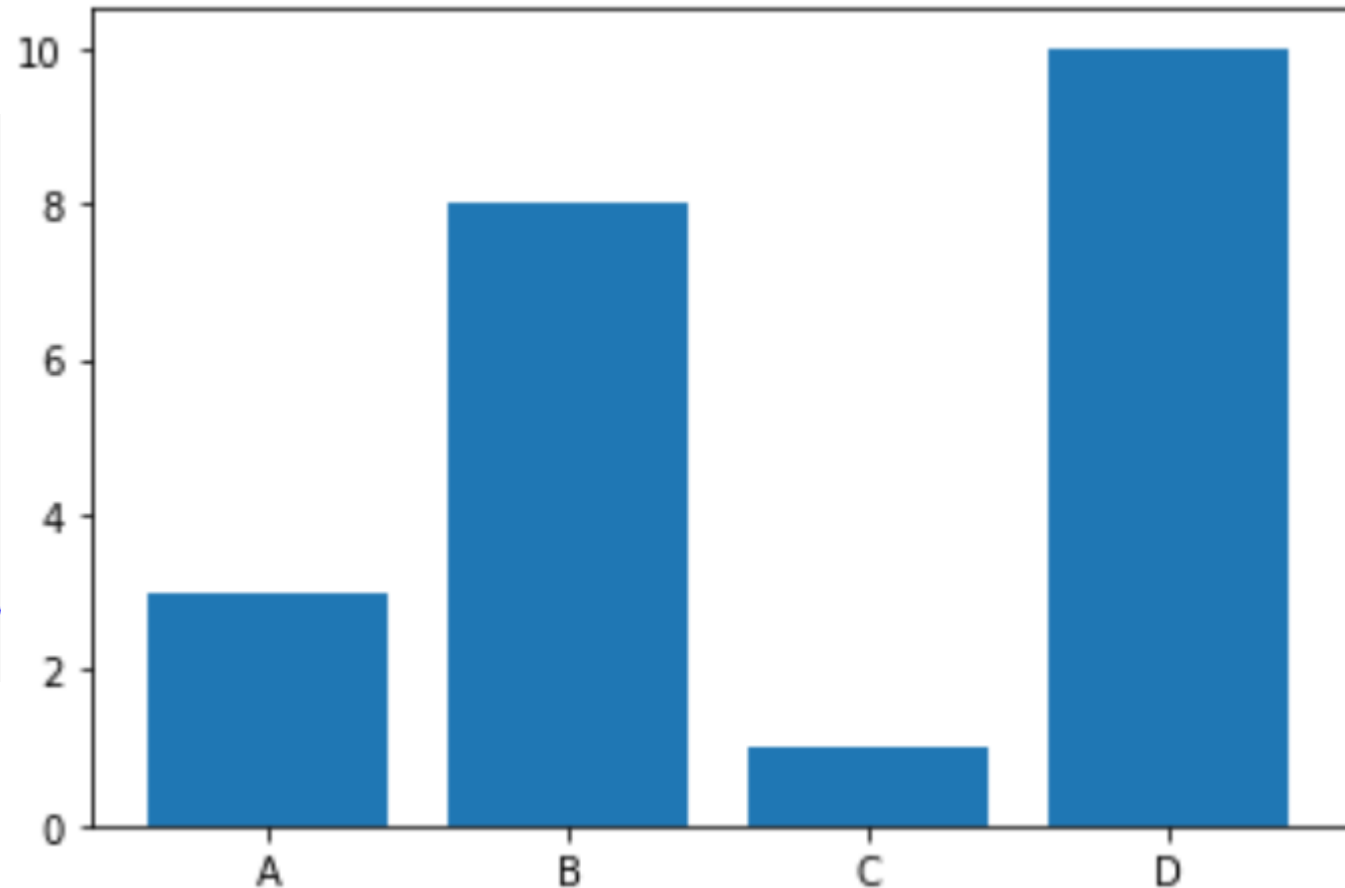


# Types of Charts

- Bar Graph
  - you can use the `bar()` function to draw bar graphs

```
import matplotlib.pyplot as plt
import numpy as np

x = np.array(["A", "B", "C", "D"])
y = np.array([3, 8, 1, 10])
plt.bar(x,y)
plt.show()
```



# Types of Charts

- Scatter Plot
  - With Pyplot, you can use the `scatter()` function to draw a scatter plot.
  - The `scatter()` function plots one dot for each observation.
  - It needs two arrays of the same length, one for the values of the x-axis, and one for values on the y-axis:

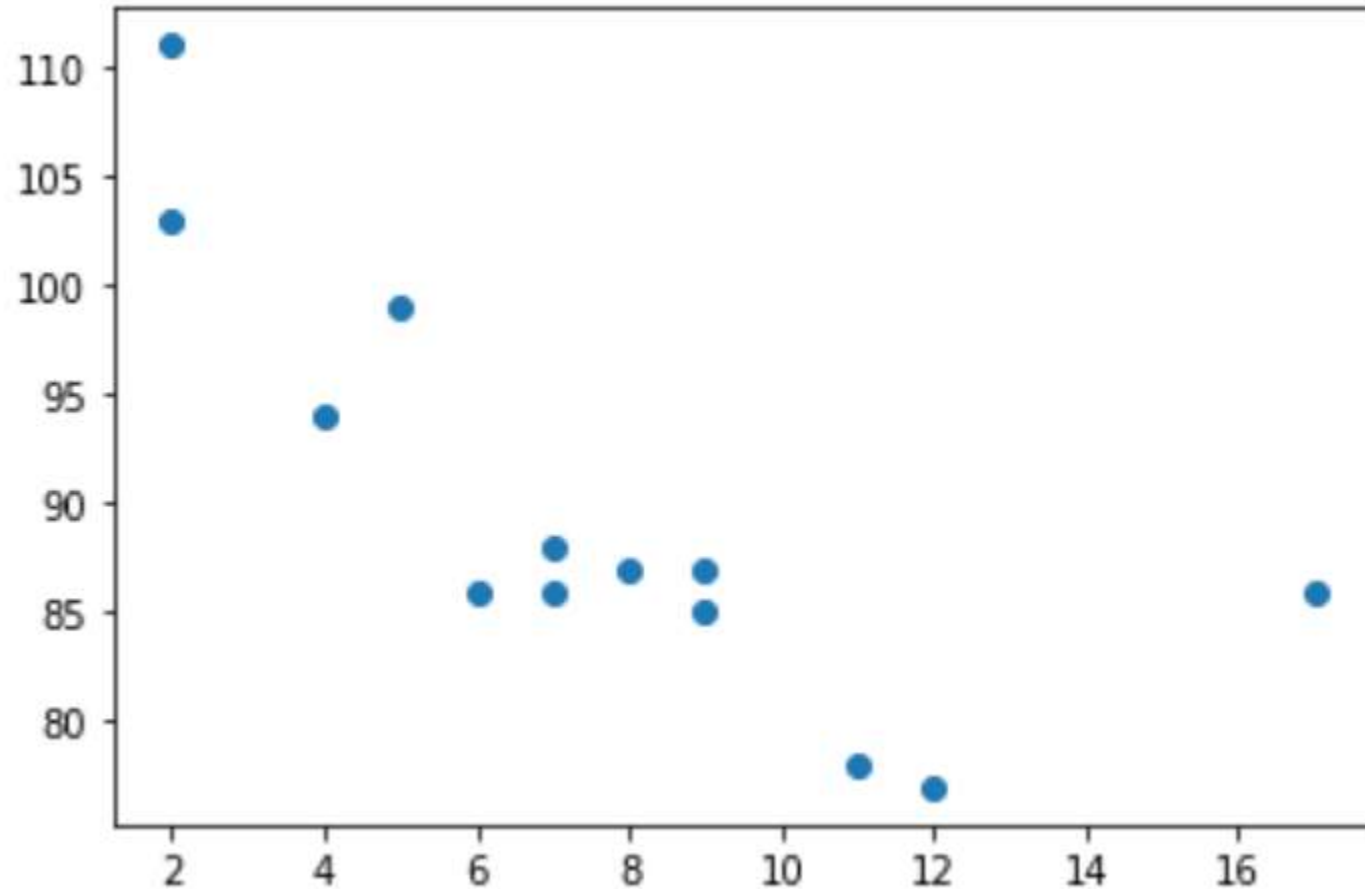


# Types of Charts

```
import matplotlib.pyplot as plt
import numpy as np

x = np.array([5, 7, 8, 7, 2, 17, 2, 9,
              4, 11, 12, 9, 6])
y = np.array([99, 86, 87, 88, 111, 86,
              103, 87, 94, 78, 77, 85, 86])

plt.scatter(x, y)
plt.show()
```



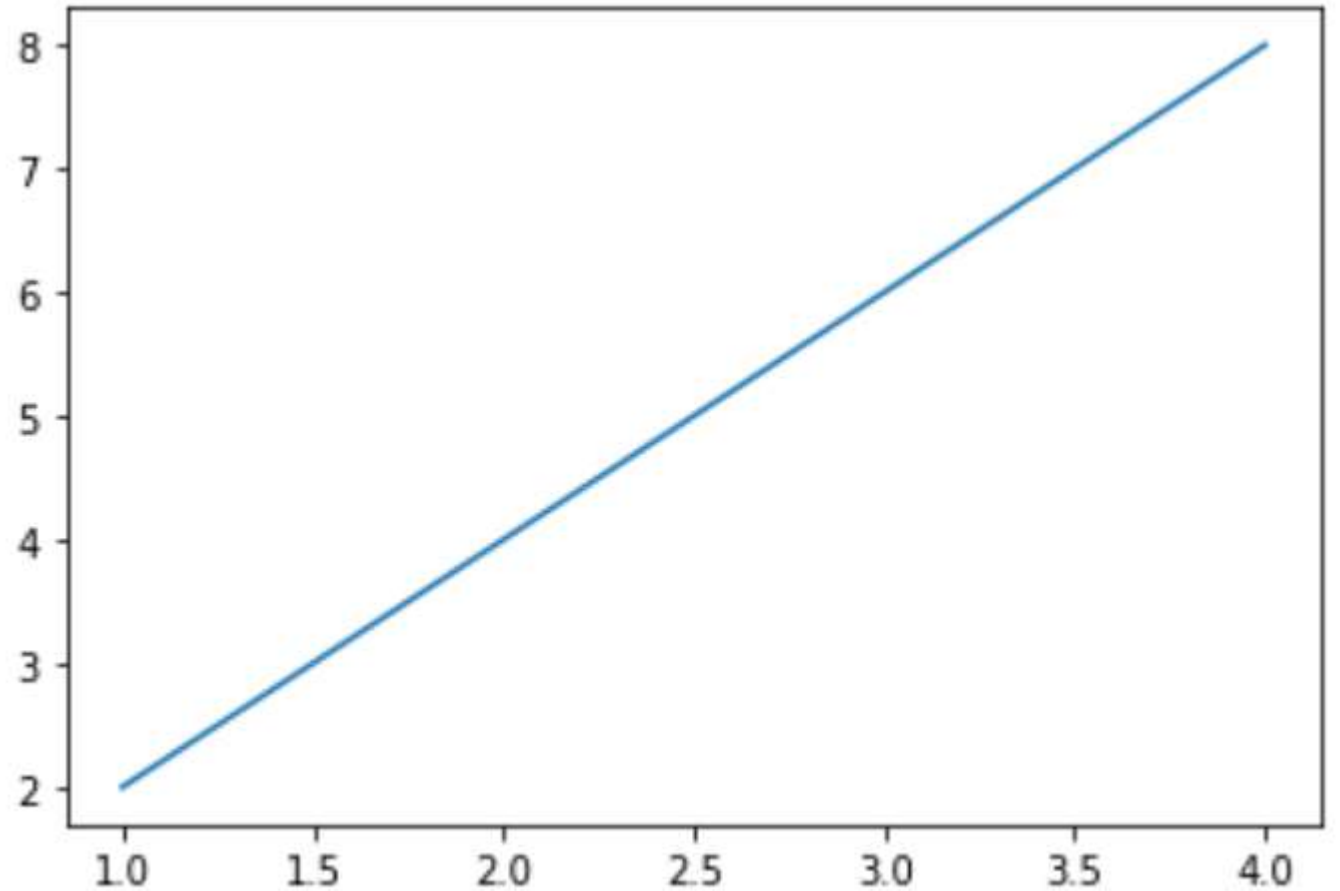
- The observation in the example above is the result of 13 cars passing by.
  - The X-axis shows how old the car is.
  - The Y-axis shows the speed of the car when it passes.
- Are there any relationships between the observations?
  - It seems that the newer the car, the faster it drives, but that could be a coincidence, after all we only registered 13 cars.

# Types of Charts

- Line Plot

```
import matplotlib.pyplot as plt
import numpy as np
x = np.array([1, 2, 3, 4])
y = x*2

plt.plot(x, y)
plt.show()
```

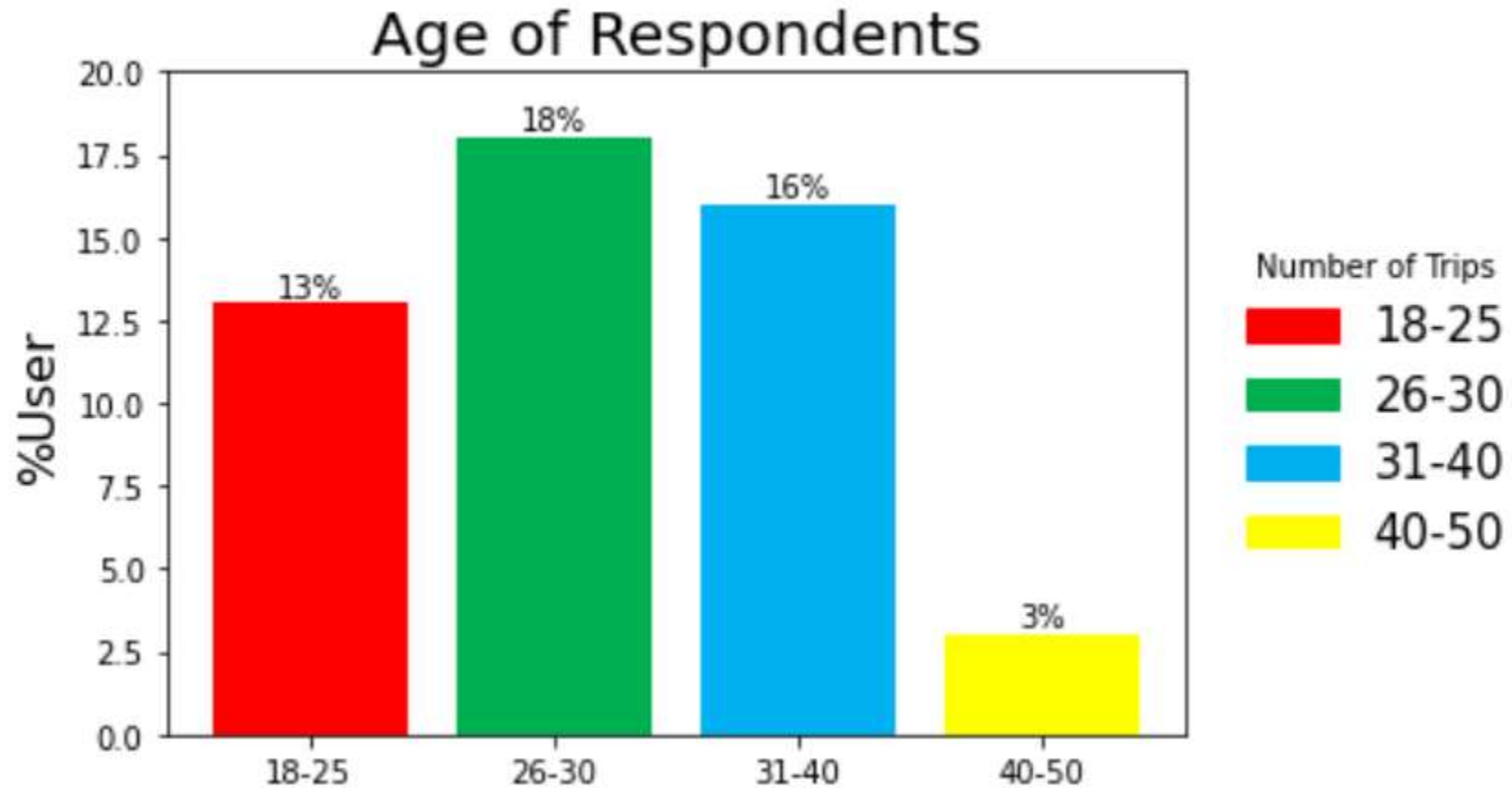


# Legends, Annotation, Style

```
import numpy as np
import matplotlib.pyplot as plt
from matplotlib.patches import Patch
color = ('red', '#00b050', '#00b0f0', 'yellow')
objects = ('18-25', '26-30', '31-40', '40-50')
y_pos = np.arange(len(objects))
performance = [13, 18, 16, 3]
width = 0.35 # the width of the bars
plt.bar(y_pos, performance, align='center', color=color)
plt.xticks(y_pos, objects)
plt.ylim(0, 20)
plt.ylabel('%User', fontsize=16)
plt.title('Age of Respondents', fontsize=20)

# map names to colors
cmap = dict(zip(performance, color))
# create the rectangles for the legend
patches = [Patch(color=v, label=k) for k, v in cmap.items()]
# add the legend
plt.legend(title='Number of Trips', labels=objects, handles=patches,
          bbox_to_anchor=(1.04, 0.5), loc='center left',
          borderaxespad=0, fontsize=15, frameon=False)
# add the annotations
for y, x in zip(performance, y_pos):
    plt.annotate(f'{y}%\n', xy=(x, y), ha='center', va='center')
```

# Legends, Annotation, Style



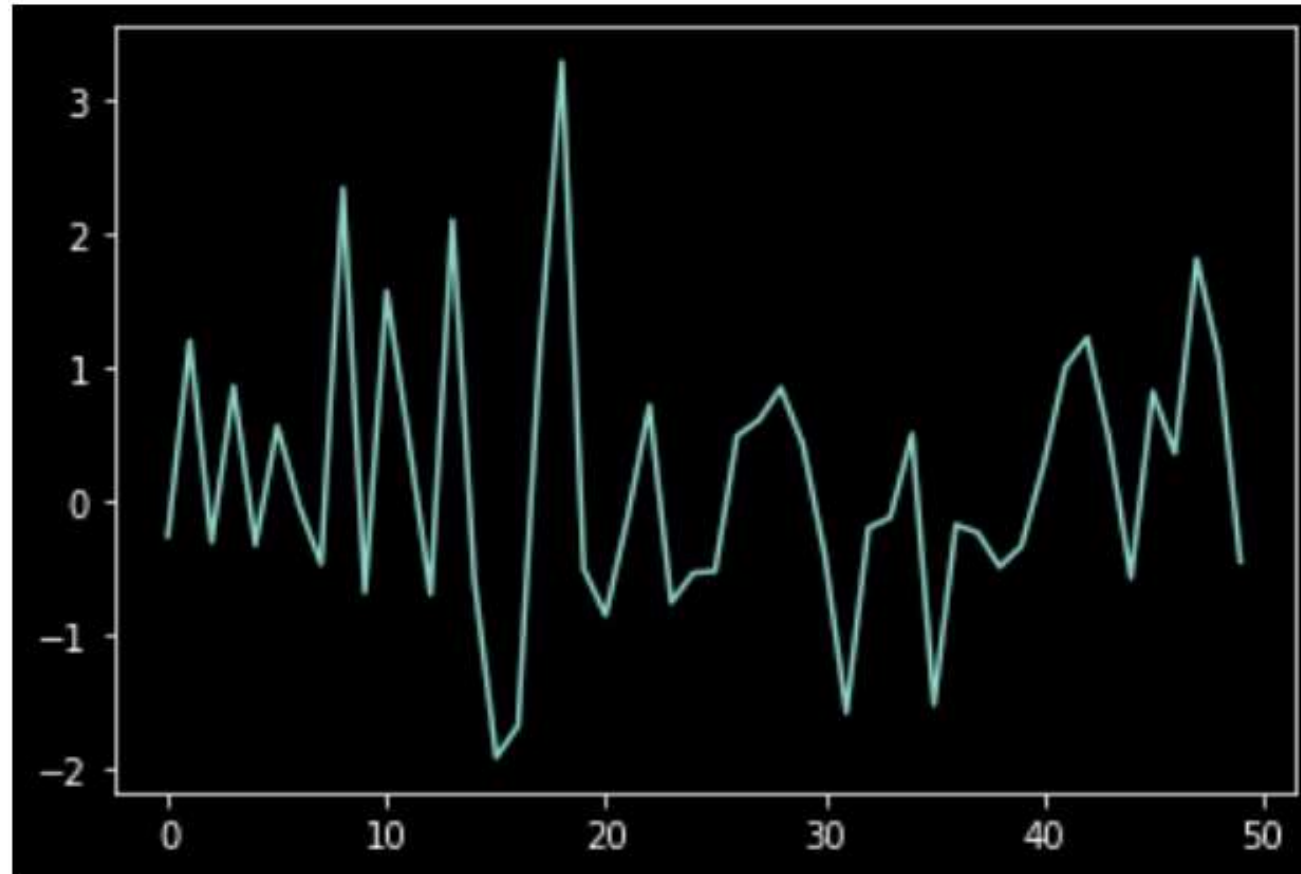
# Legends, Annotation, Style

```
import matplotlib.pyplot as plt  
print(plt.style.available)
```

['Solarize\_Light2', '\_classic\_test\_patch', 'bmh', 'classic',  
'dark\_background', 'fast', 'fivethirtyeight', 'ggplot', 'grayscale',  
'seaborn', 'seaborn-bright', 'seaborn-colorblind', 'seaborn-dark',  
'seaborn-dark-palette', 'seaborn-darkgrid', 'seaborn-deep', 'seaborn-  
muted', 'seaborn-notebook', 'seaborn-paper', 'seaborn-pastel',  
'seaborn-poster', 'seaborn-talk', 'seaborn-ticks', 'seaborn-white',  
'seaborn-whitegrid', 'tableau-colorblind10']

# Legends, Annotation, Style

```
import numpy as np
import matplotlib.pyplot as plt
from matplotlib import style
data = np.random.randn(50)
plt.style.use('dark_background')
plt.plot(data)
plt.show()
```



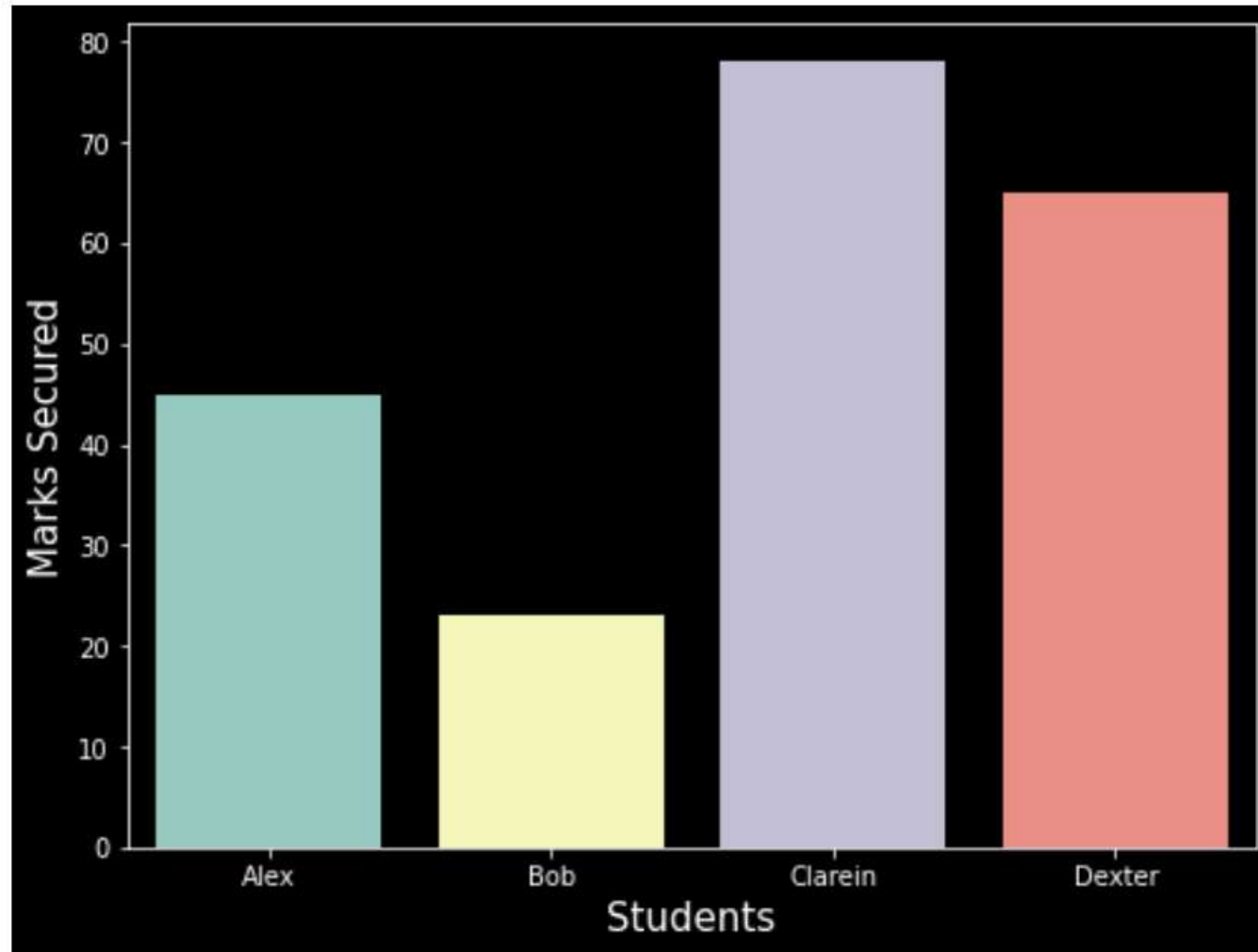
# Plotting directly from Pandas Data Frame

```
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
data = {"Name": ["Alex", "Bob", "Clarein", "Dexter"],
        "Marks": [45, 23, 78, 65]}
df = pd.DataFrame(data, columns=['Name', 'Marks'])

plt.figure(figsize=(8, 6))
plots = sns.barplot(x="Name", y="Marks", data=df)
plt.xlabel("Students", size=15)
plt.ylabel("Marks Secured", size=15)
plt.show()
```



# Plotting directly from Pandas Data Frame



# Plotting directly from NumPy Array

```
import numpy as np
import matplotlib.pyplot as plt

x = np.arange(1, 11)
y = np.array([100, 10, 300, 20, 500,
              60, 700, 80, 900, 100])

plt.title("Line graph")
plt.xlabel("X axis")
plt.ylabel("Y axis")
plt.plot(x, y, color="green")
plt.show()
```

