

CSCE 5300

Data Visualization

Matplotlib

Why Data Visualization

“A picture is worth a thousand words”

- Data visualization is the first step of analysis work.
- It gives intuitive understanding of data.
- Helps you to see data in certain meaningful patterns.
- Visual representations enhances the human cognitive process.

Benefits of Data Visualization

- Data visualization allow users to see several different perspectives of data
- Data visualization makes it possible to interpret vast amounts of data.
- It offers ability to note expectations in data
- Exploring trends within a database through visualization by letting analysts navigate through data and visually orient themselves to the patterns in the data

Matplotlib

- Matplotlib is an amazing visualization library in Python for 2D plots of arrays
- Matplotlib is a multi-platform data visualization library built on NumPy arrays and designed to work with the broader SciPy stack.
- It was introduced by John Hunter in the year 2002.
- One of the greatest benefits of visualization is that it allows us visual access to huge amounts of data in easily digestible visuals.
- Matplotlib consists of several plots like line, bar, scatter, histogram etc

Importing matplotlib

- `from matplotlib import pyplot as plt` *or*
- `import matplotlib.pyplot as plt`

Basic plots in Matplotlib

- Matplotlib comes with a wide variety of plots.
- Plots helps to understand trends, patterns, and to make correlations.
- They're typically instruments for reasoning about quantitative information.

Line plot

- `# importing matplotlib module`
- `from matplotlib import pyplot as plt`
-
- `# x-axis values`
- `x = [5, 2, 9, 4, 7]`
-
- `# Y-axis values`
- `y = [10, 5, 8, 4, 2]`
-
- `# Function to plot`
- `plt.plot(x,y)`
-
- `# function to show the plot`
- `plt.show()`

Bar plot

- `# importing matplotlib module`
- `from matplotlib import pyplot as plt`
-
- `# x-axis values`
- `x = [5, 2, 9, 4, 7]`
-
- `# Y-axis values`
- `y = [10, 5, 8, 4, 2]`
-
- `# Function to plot the bar`
- `plt.bar(x,y)`
-
- `# function to show the plot`
- `plt.show()`

Histogram

```
#importing matplotlib module  
From matplotlib import pyplot as plt
```

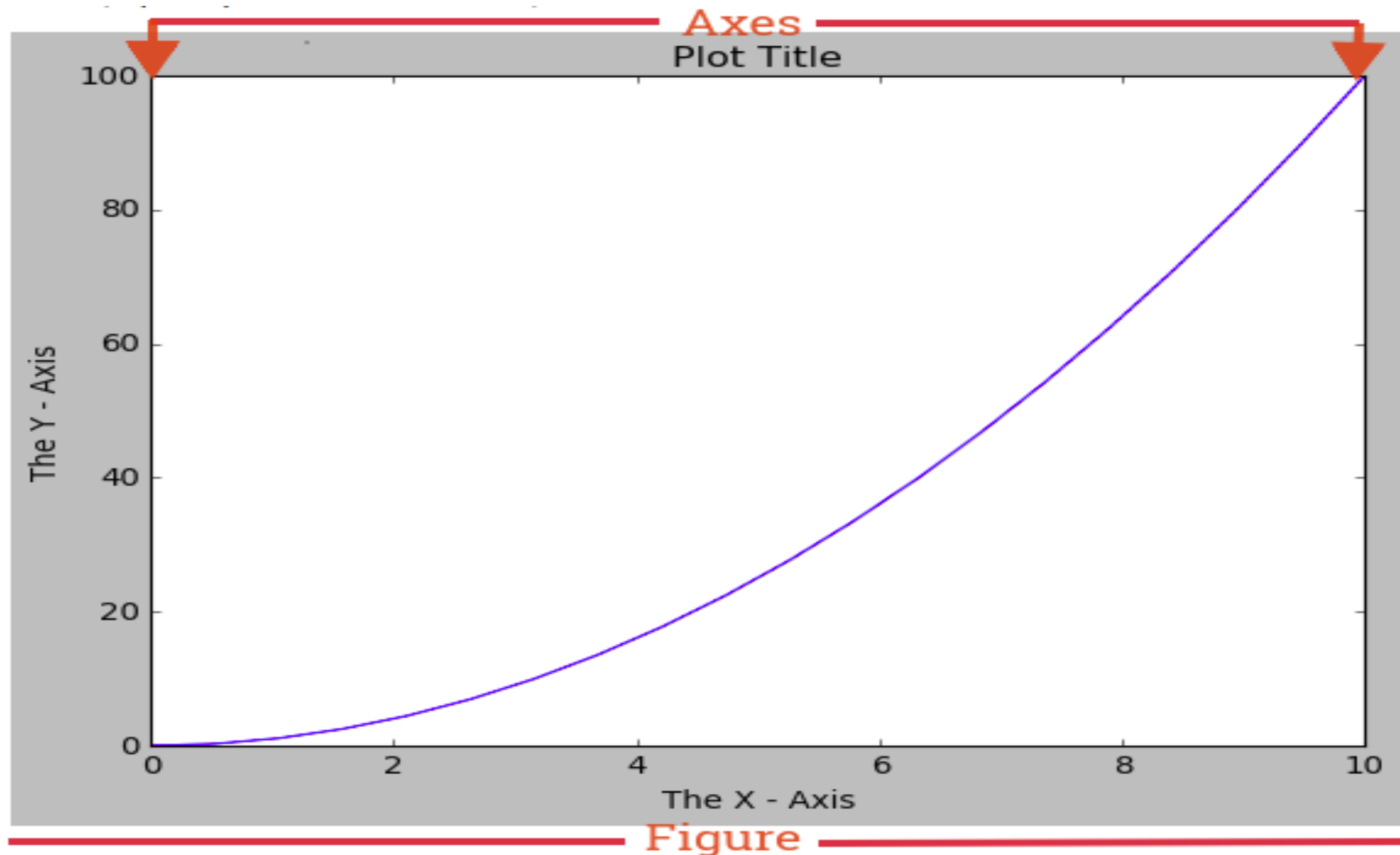
```
#Y-axis values  
y = [10, 5, 8, 4, 2]
```

```
#Function to plot histogram  
plt.hist(y)
```

```
#Function to show the plot  
plt.show()
```

Scatter Plot :

- `# importing matplotlib module`
- `from matplotlib import pyplot as plt`
-
- `# x-axis values`
- `x = [5, 2, 9, 4, 7]`
-
- `# Y-axis values`
- `y = [10, 5, 8, 4, 2]`
-
- `# Function to plot scatter`
- `plt.scatter(x, y)`
-
- `# function to show the plot`
- `plt.show()`

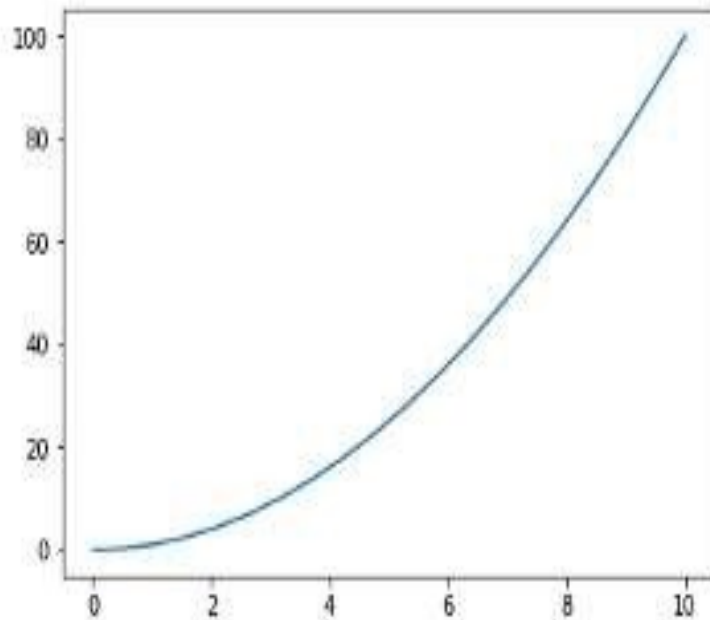


Functional Approach:

```
In [3]: import numpy as np  
x = np.linspace(0, 10, 20) #Generate 20 datapoints between 0 and 10  
y = x**2                  #Generate array 'y' from square of 'x'
```

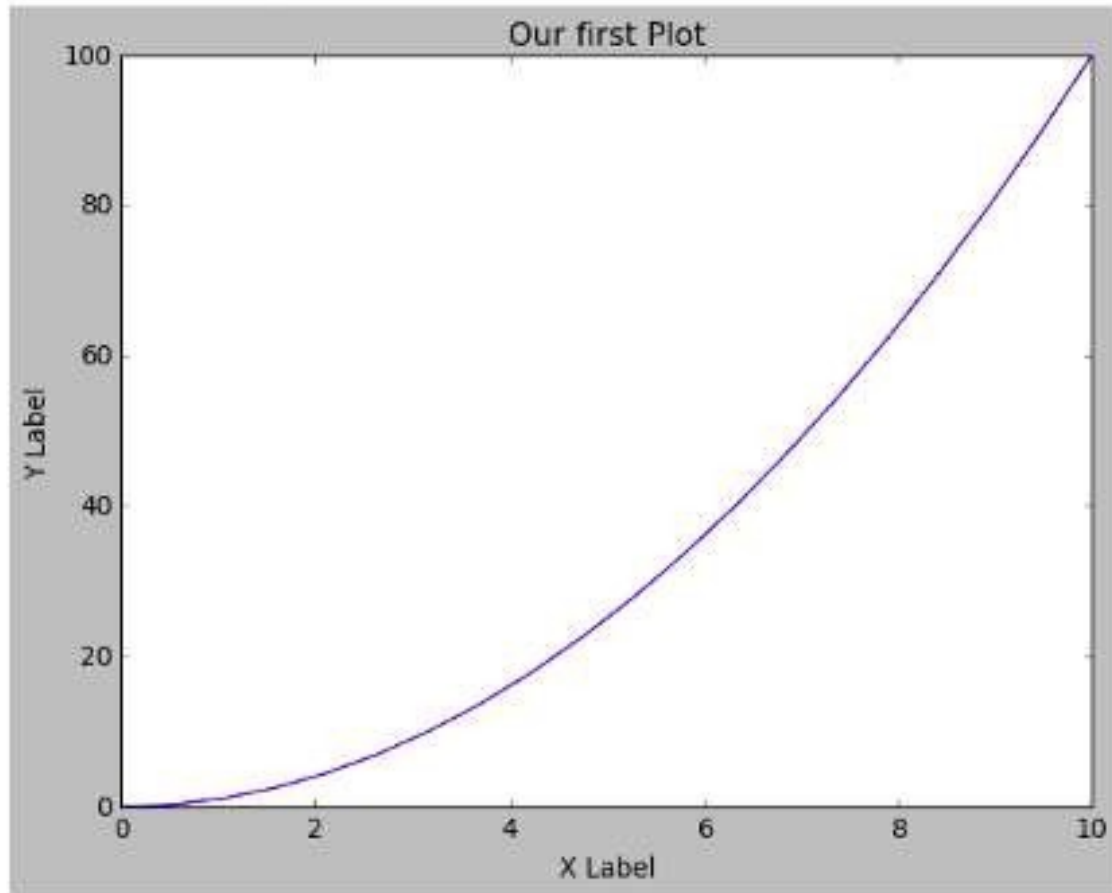
```
In [4]: plt.plot(x,y)
```

```
Out[4]: [<matplotlib.lines.Line2D at 0x7f5a339fcd30>]
```



```
In [12]: plt.plot(x,y)
plt.title('Our first Plot')
plt.xlabel('X Label')
plt.ylabel('Y Label')
```

```
Out[12]: Text(0,0.5,'Y Label')
```



Matplotlib allows us easily create multi-plots on the same figure using the `.subplot()` method. This `.subplot()` method takes in three parameters, namely:

`nrows`: the number of

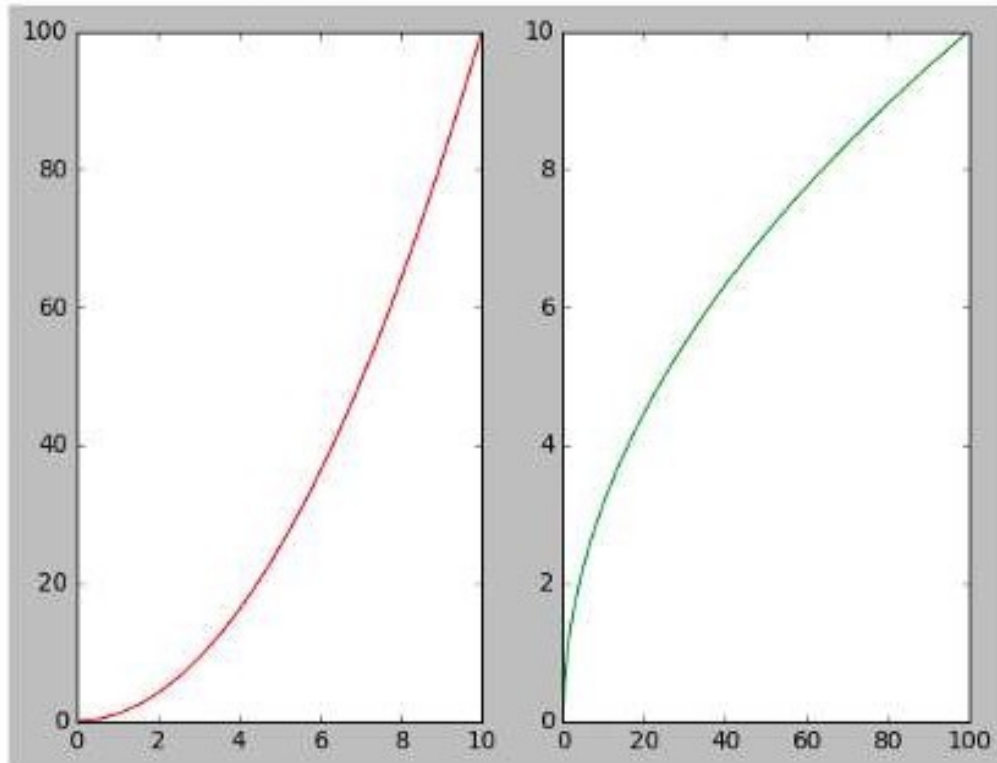
`rows ncols`: the number of
columns

`plot_number` : refers to a
specific plot in the Figure.

Using `.subplot()` we will create a two plots on the same canvas:

```
In [13]: # plt.subplot(nrows, ncols, plot_number)
plt.subplot(1, 2, 1)
plt.plot(x, y, 'red') # More on color options later

plt.subplot(1, 2, 2)
plt.plot(y, x, 'green');
```



Object oriented Interface

This is the best way to create plots.

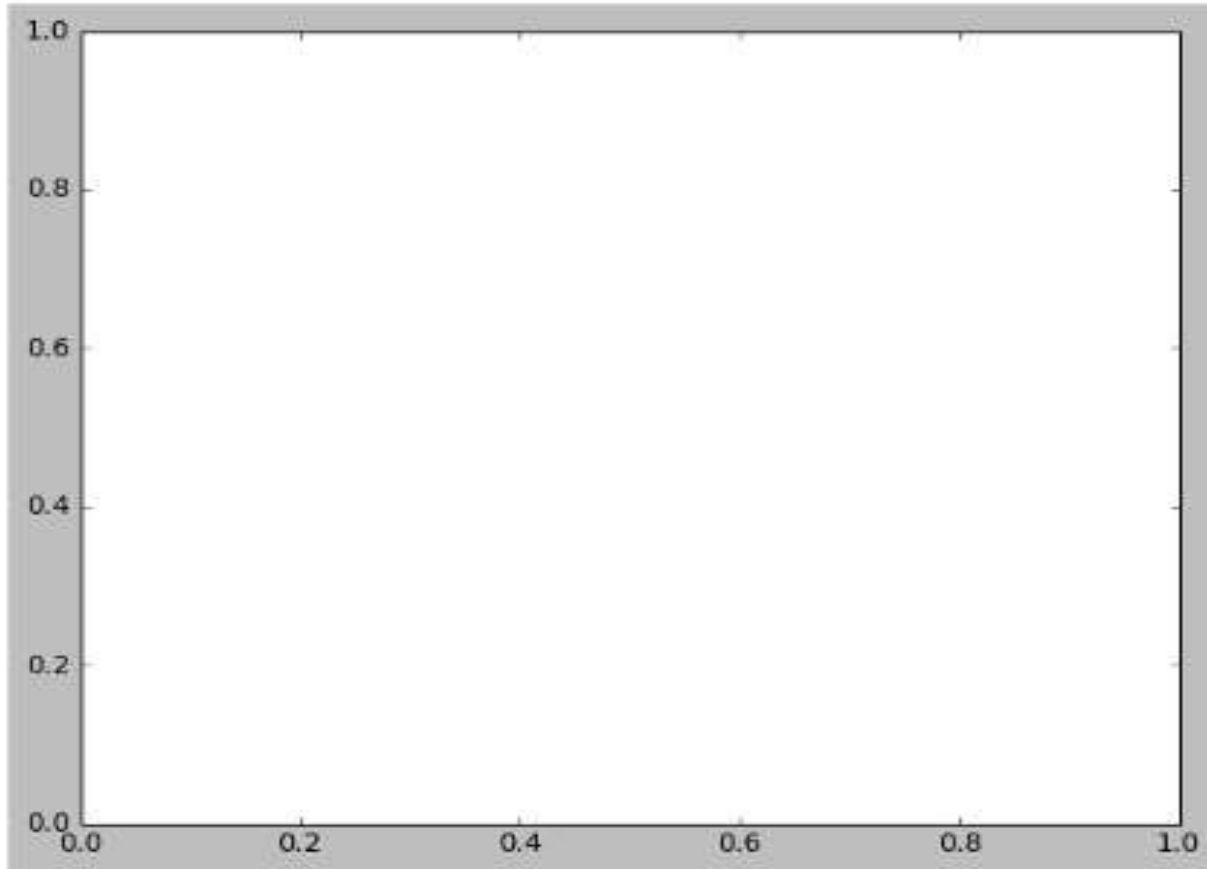
The idea here is to create Figure objects and call methods off it. Let's create a blank Figure using the `.figure()` method

```
In [14]: fig = plt.figure()  
<Figure size 640x480 with 0 Axes>
```


Next step

- Now we need to add a set of axes
`.add_axes()`
(left, bottom, width, and height)

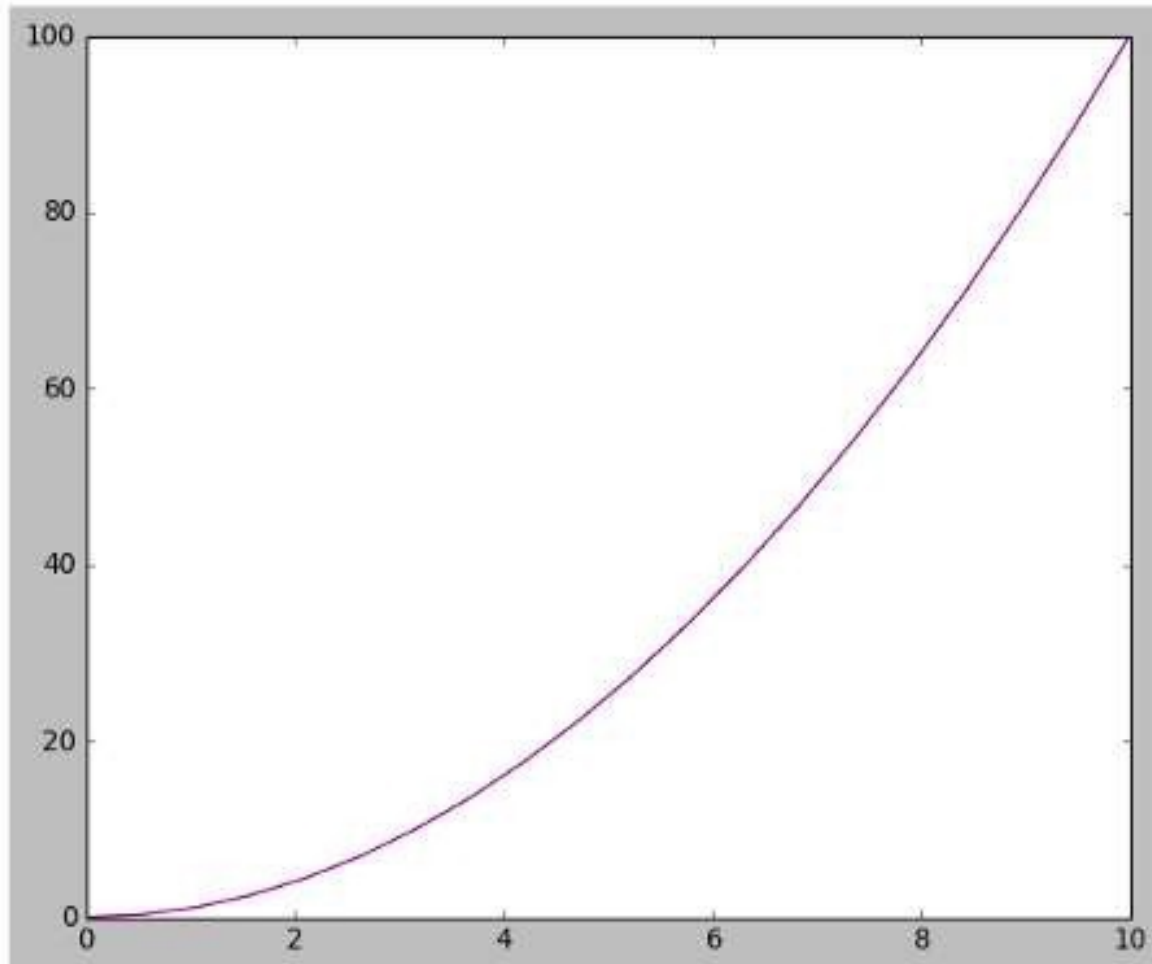
```
In [17]: fig = plt.figure()  
ax = fig.add_axes([0.1, 0.2, 0.8, 0.9])
```



```
In [18]: fig = plt.figure()
ax = fig.add_axes([0.1, 0.2, 0.8, 0.9])

ax.plot(x,y, 'purple')
```

```
Out[18]: [<matplotlib.lines.Line2D at 0x7f00630b6208>]
```

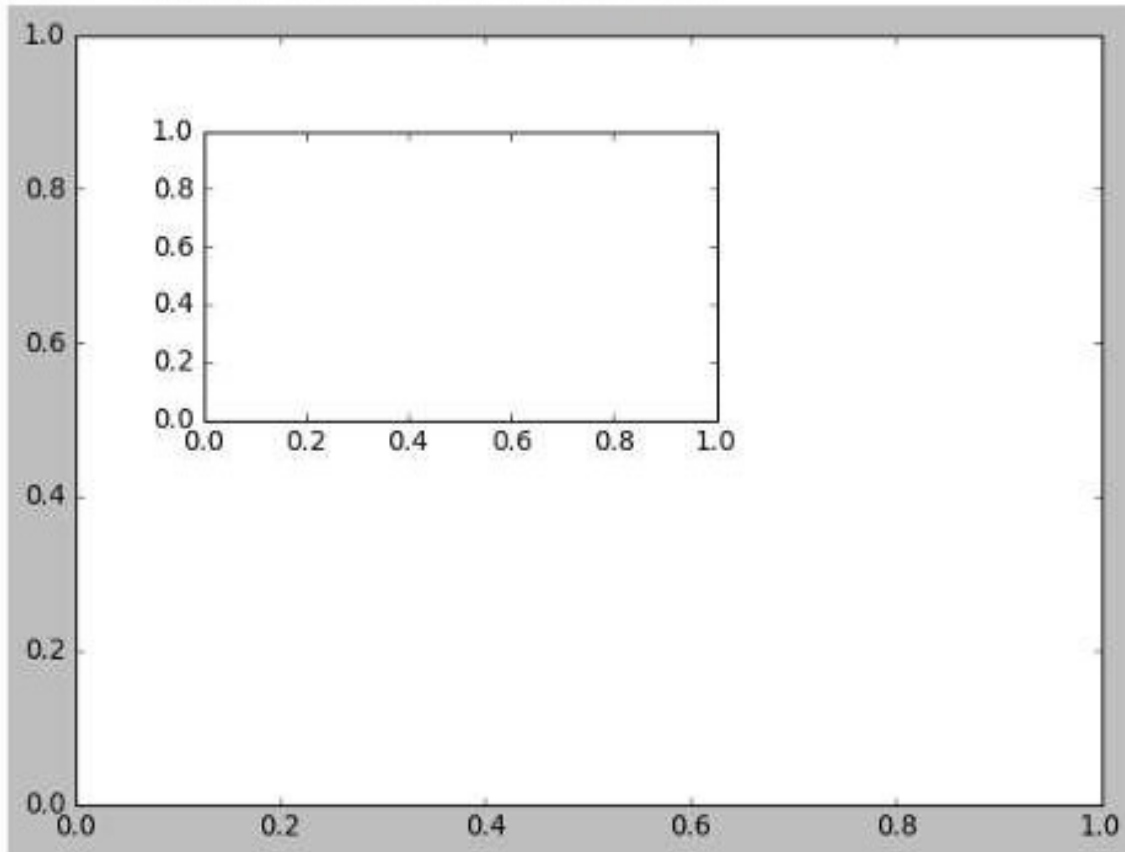


```
In [22]: fig = plt.figure()
ax = fig.add_axes([0.1, 0.2, 0.8, 0.9])

ax.plot(x,y, 'purple')
ax.set_xlabel('X Label')
ax.set_ylabel('Y Label')
ax.set_title('Our First Plot using Object Oriented Approach')
```

Something interesting ,figure in figure

```
In [24]: fig = plt.figure()  
axes1 = fig.add_axes([0.1, 0.1, 0.8, 0.8])  
axes2 = fig.add_axes([0.2, 0.5, 0.4, 0.3])
```

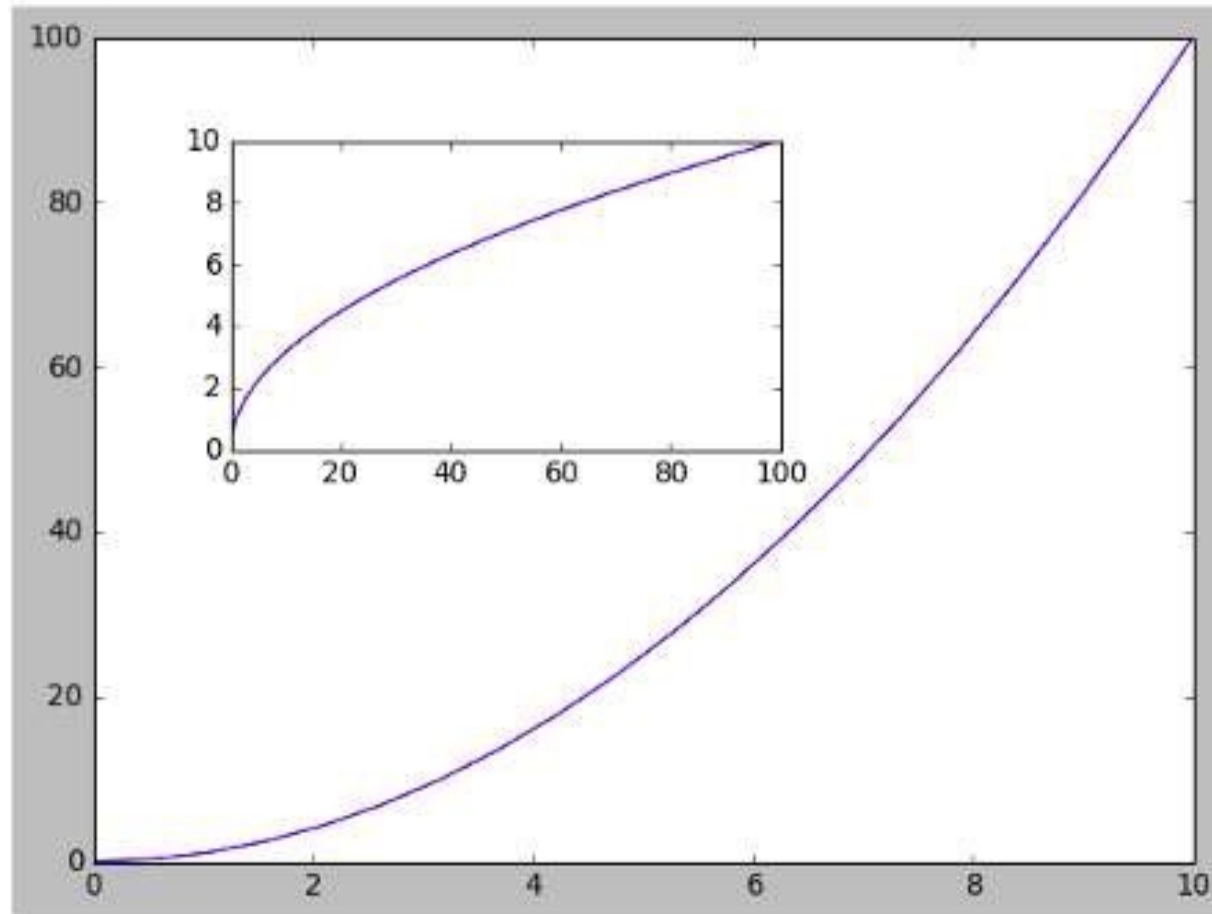


```
In [26]: fig = plt.figure()

axes1 = fig.add_axes([0.1, 0.1, 0.8, 0.8])
axes2 = fig.add_axes([0.2, 0.5, 0.4, 0.3])

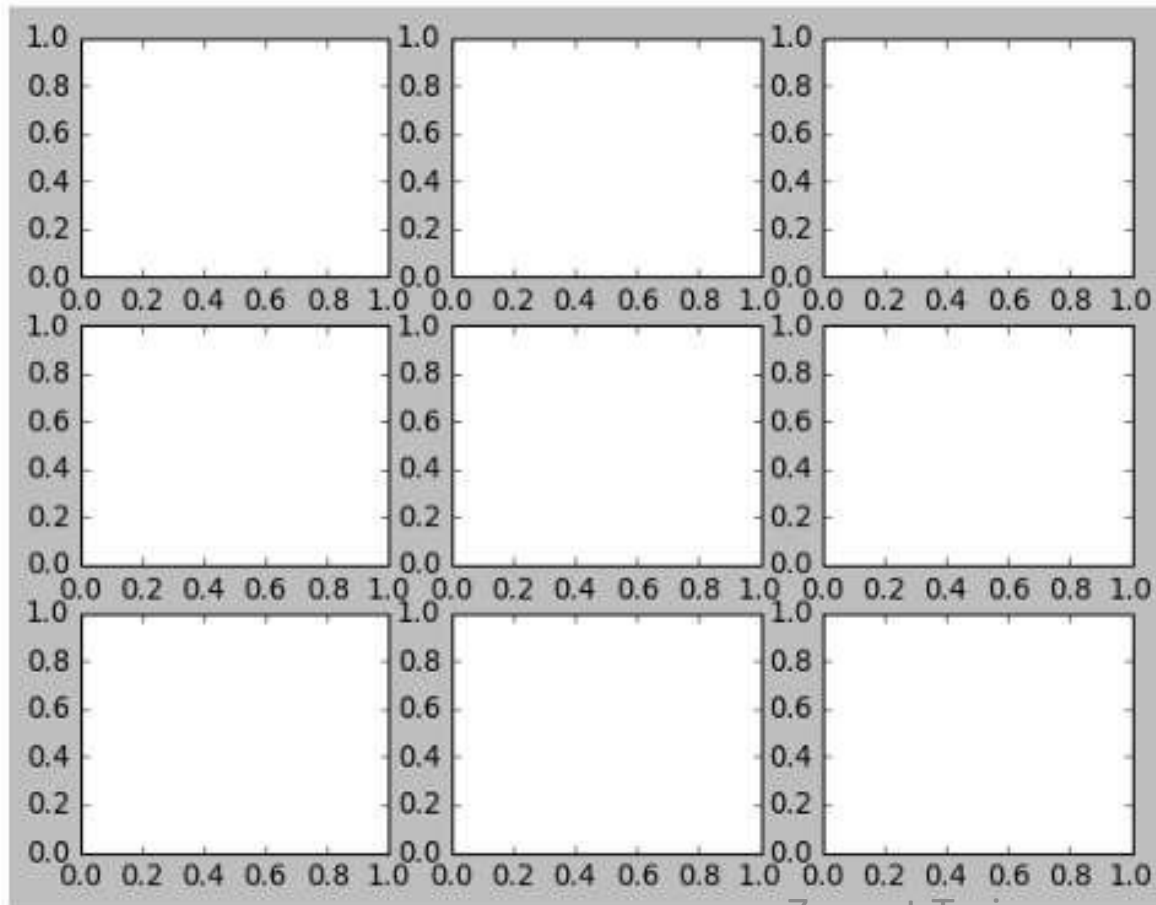
axes1.plot(x,y)
axes2.plot(y,x)
```

```
Out[26]: [<matplotlib.lines.Line2D at 0x7f00630991d0>]
```



We can create a matrix of subplot for example 3×3

```
In [30]: # Empty canvas of 3 by 3 subplots  
fig, axes = plt.subplots(nrows=3, ncols=3)
```



Zeenat Tariq

Add, `plt.tight_layout`

The only difference between `plt.figure()` and `plt.subplots()` is that `plt.subplots()` automatically does what the `.add_axes()` method of `.figure()` will do for you based off the number of rows and columns you specify.


```
In [34]: # Empty canvas of 3 by 3 subplots
fig, ax = plt.subplots(nrows=3, ncols=3)

ax[0,1].plot(x,y)
ax[1,2].plot(y,x)

plt.tight_layout()
```

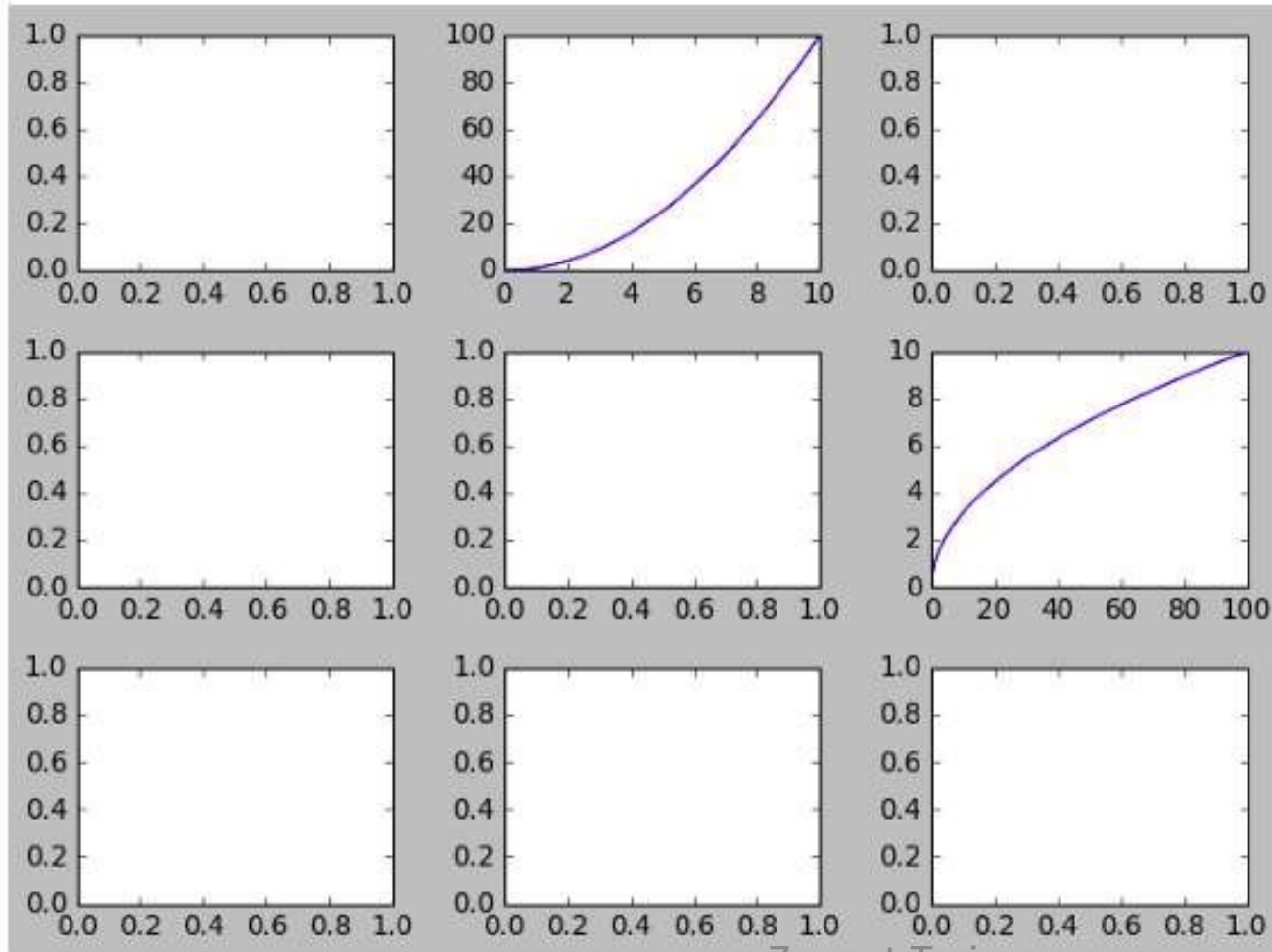
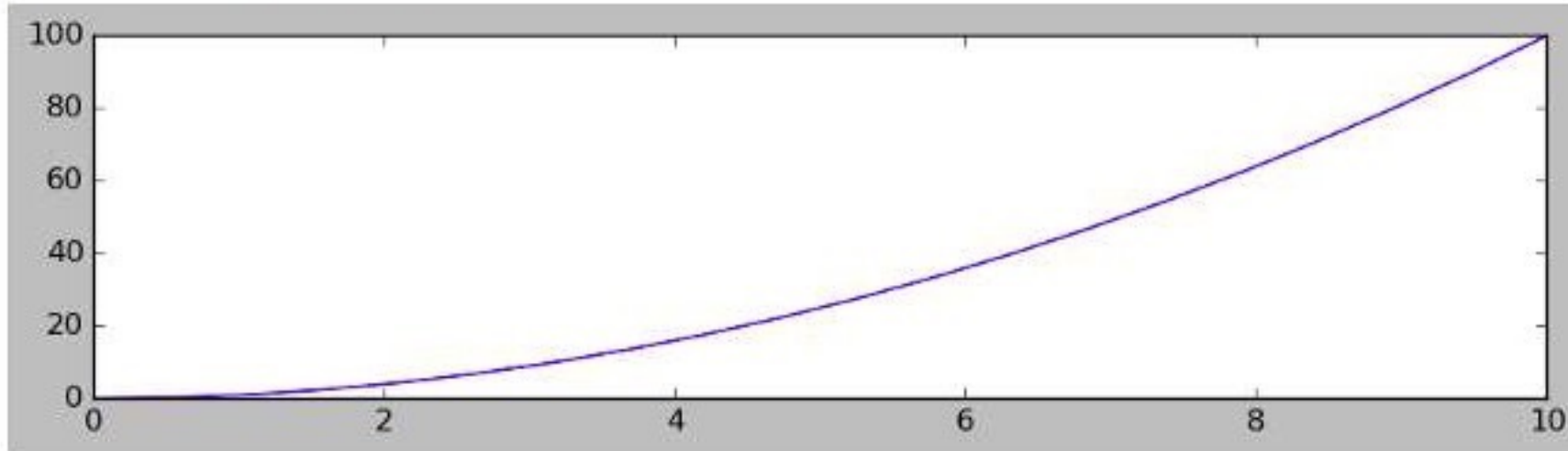


Figure size, aspect ratio, and DPI IN FIGURE

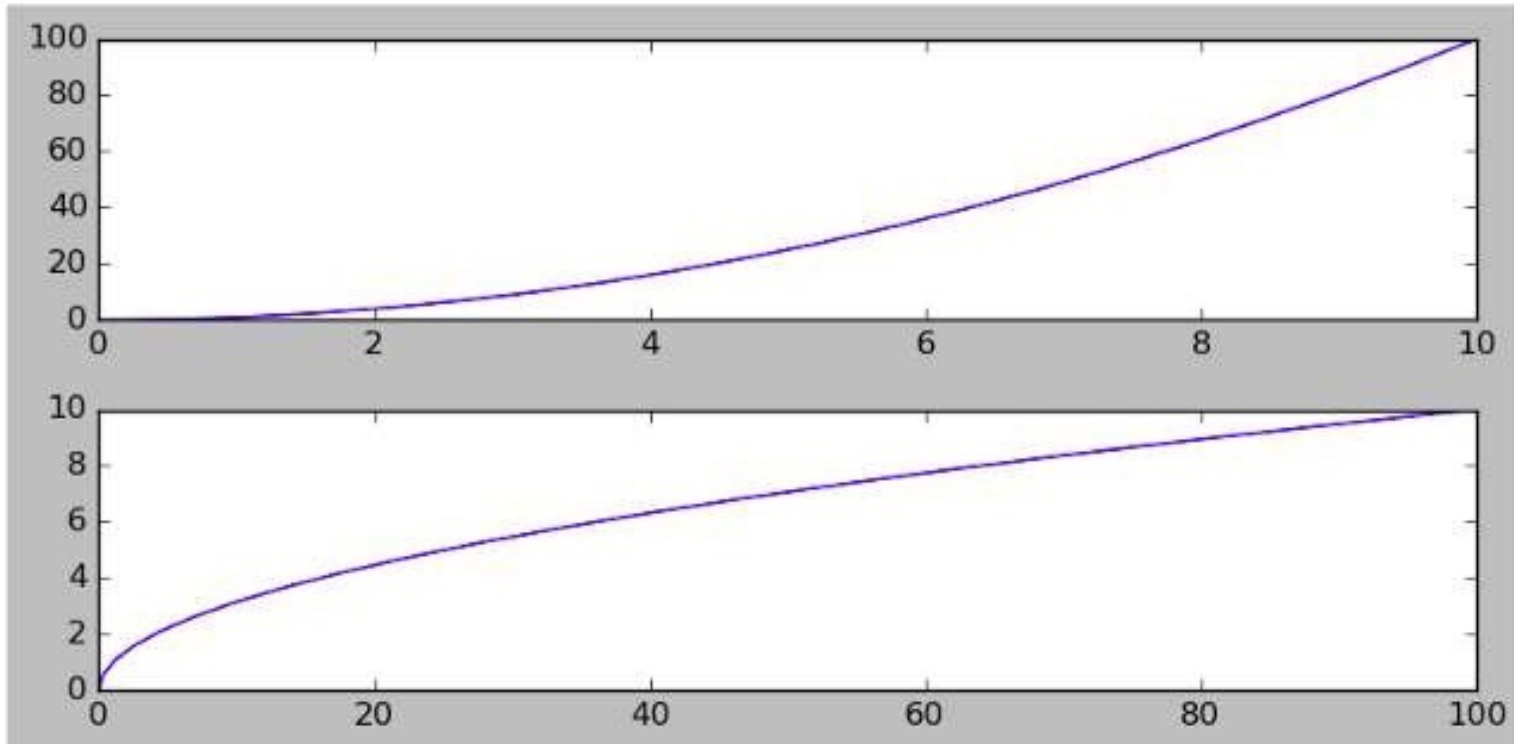
```
In [39]: fig = plt.figure(figsize=(8,2), dpi = 100)  
  
         ax = fig.add_axes([0,0,1,1])  
         ax.plot(x,y)
```

```
Out[39]: [<matplotlib.lines.Line2D at 0x7f00624b9a20>]
```



IN SUBPLOTS

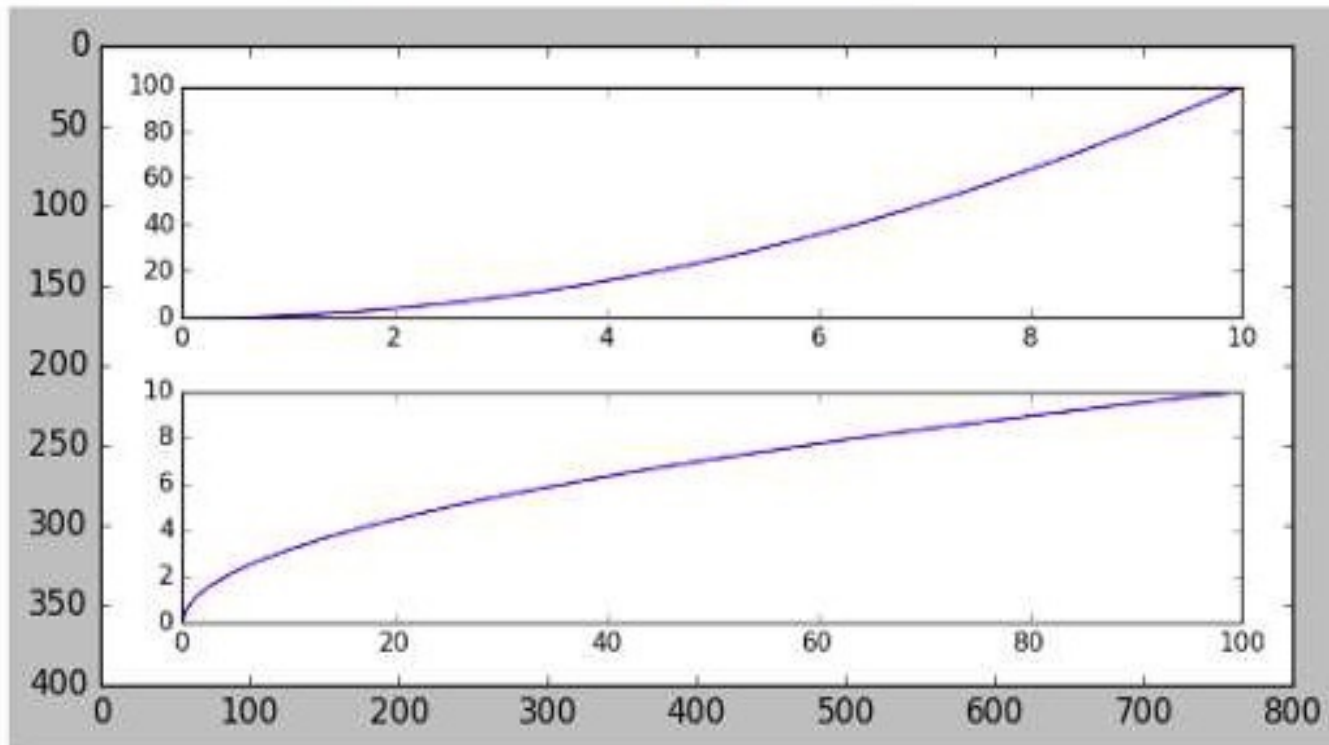
```
In [44]: fig, ax = plt.subplots(nrows=2, ncols=1, figsize=(8,4), dpi = 100)
ax[0].plot(x,y)
ax[1].plot(y,x)
plt.tight_layout()
```



SAVE PICTURE ALSO VIA
`FIG.SAVEFIG('NAME.PNG')`

```
In [52]: import matplotlib.image as mpimg  
plt.imshow(mpimg.imread('my_figure.png'))
```

```
Out[52]: <matplotlib.image.AxesImage at 0x7f00629afc88>
```

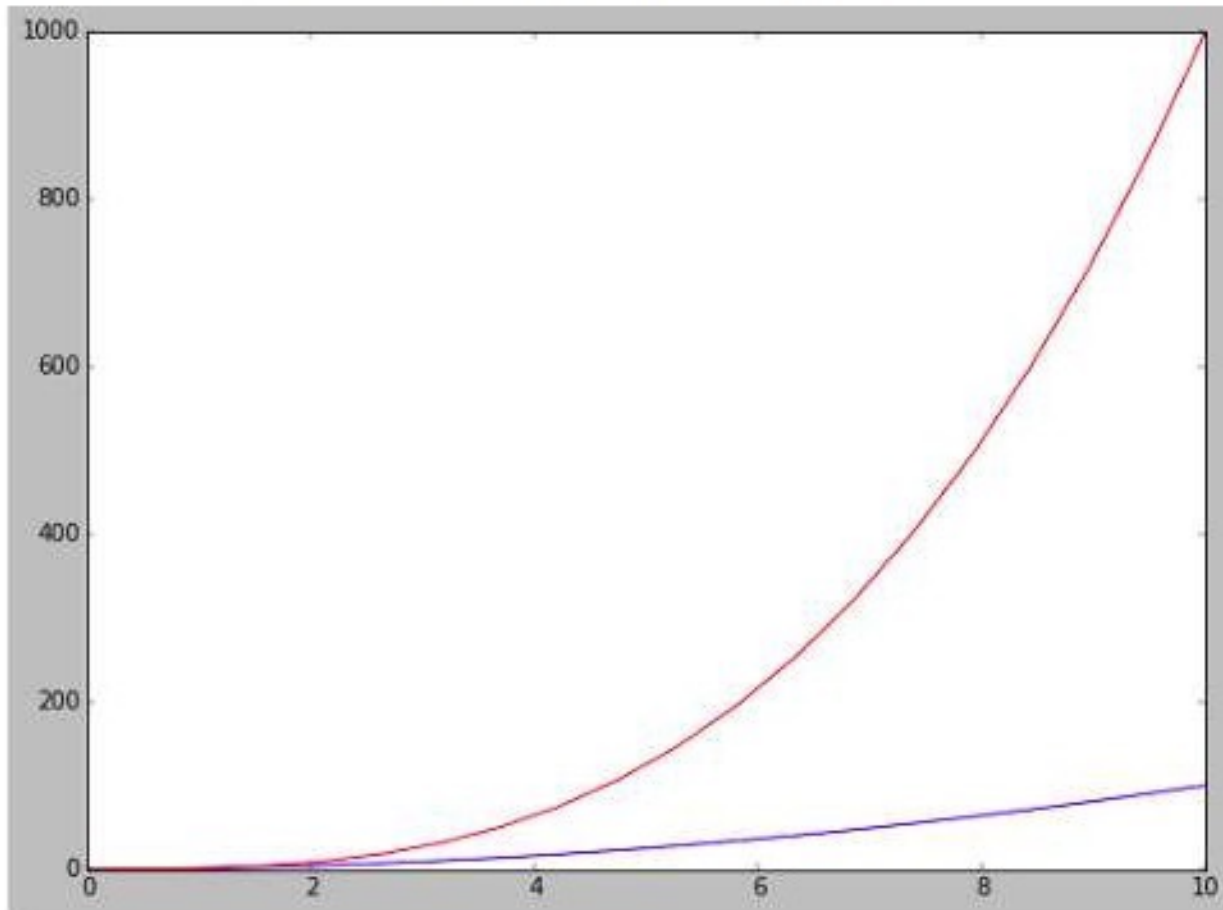


Legends

- Legends allows us to distinguish between plots. With Legends, you can use label texts to identify or differentiate one plot from another. For example, say we have a figure having two plots like below

```
In [74]: fig = plt.figure(figsize=(8,6), dpi = 60)
ax = fig.add_axes([0,0,1,1])
ax.plot(x,x**2)
ax.plot(x,x**3, 'red')
```

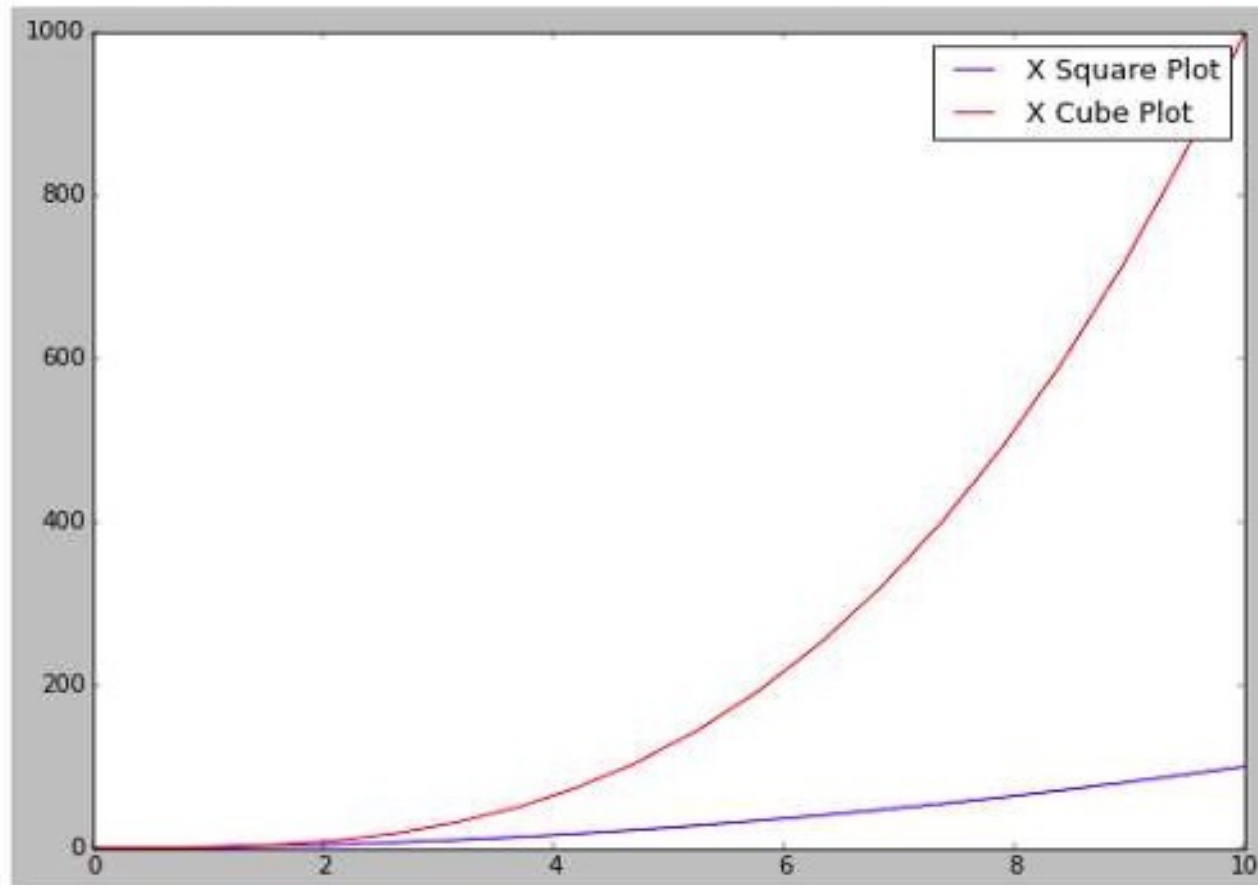
Out[74]: [<matplotlib.lines.Line2D at 0x7f0062338f60>]



```
In [75]: fig = plt.figure(figsize=(8,6), dpi = 60)
ax = fig.add_axes([0,0,1,1])
ax.plot(x,x**2, label="X Square Plot")
ax.plot(x,x**3, 'red', label='X Cube Plot')

ax.legend()
```

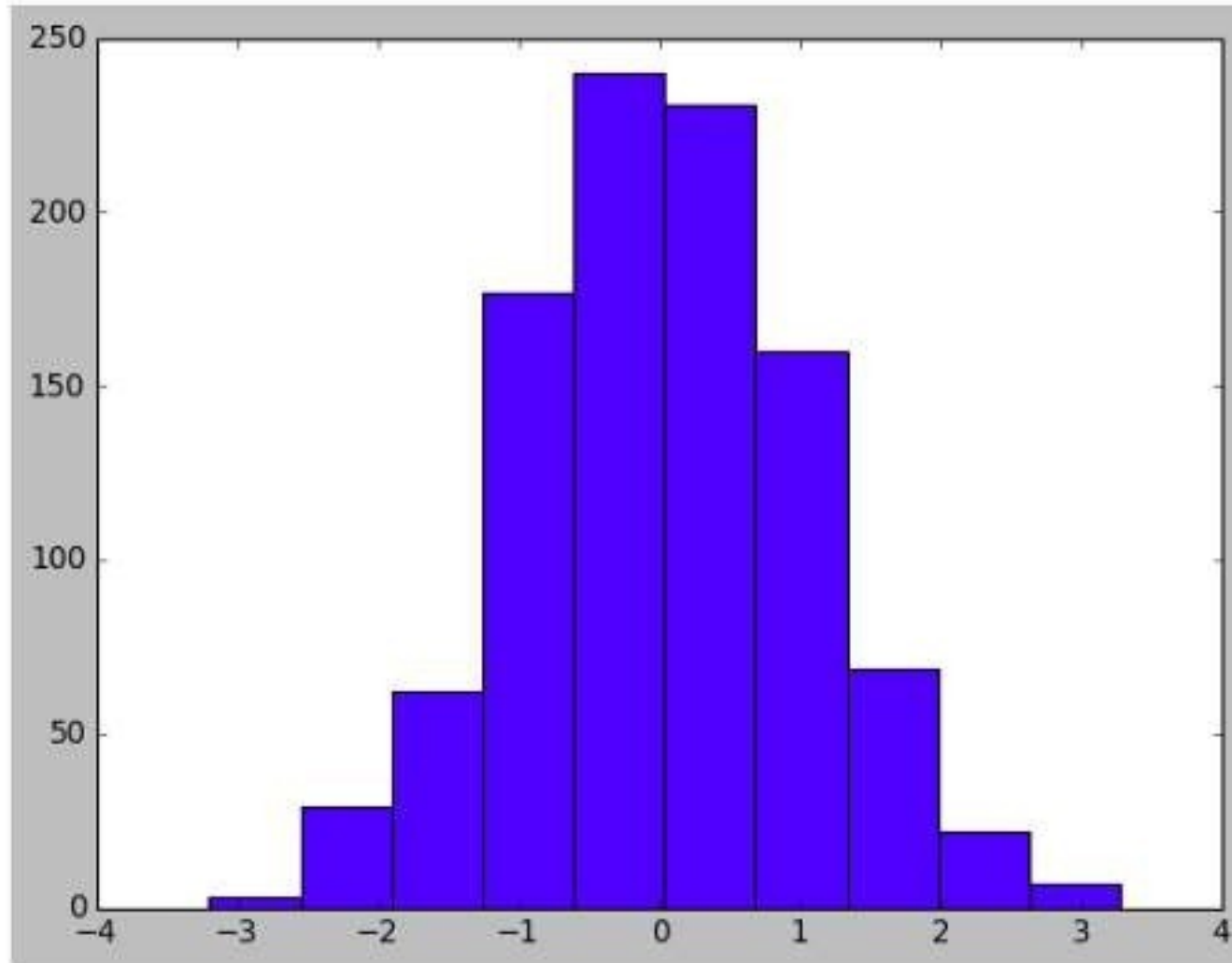
Out[75]: <matplotlib.legend.Legend at 0x7f0061e0ee80>



Plot Types

- Histogram
- Helps us understand the distribution of numeric value in a way that you can not do with mean, median and mode.


```
In [90]: x = np.random.randn(1000)  
plt.hist(x);
```



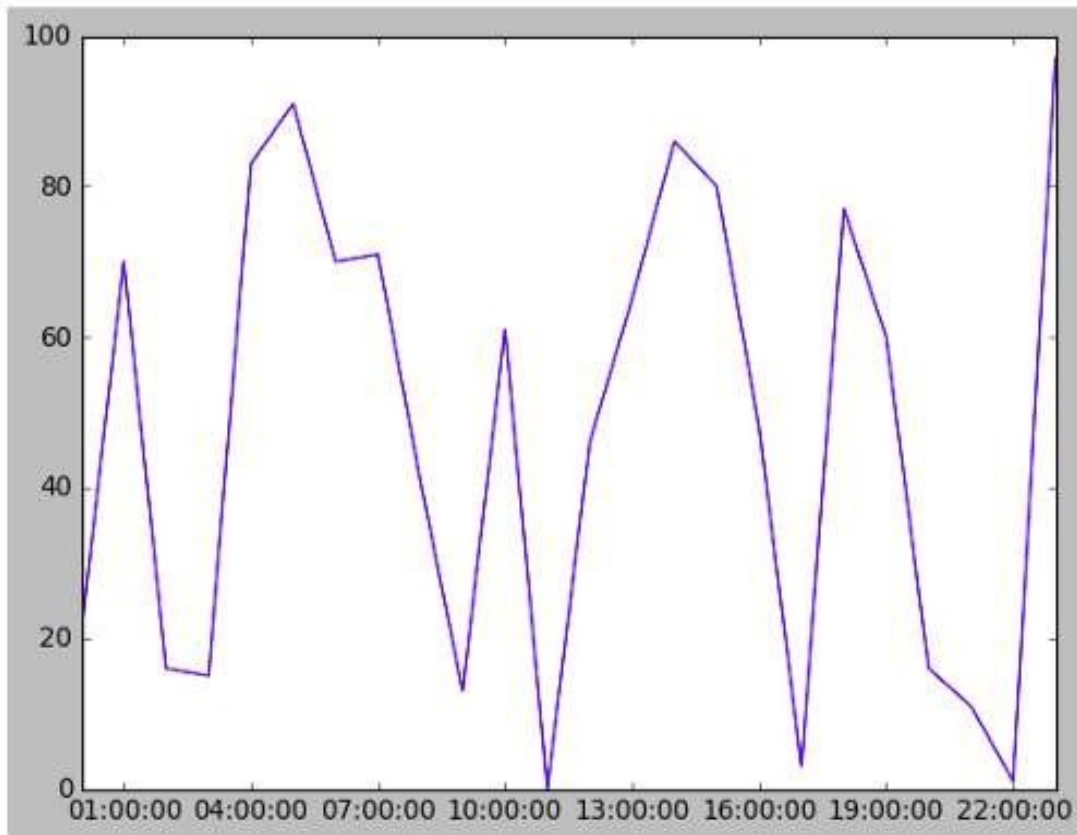
Time series (LinePlot)

- A chart that shows a trend over a period of time
- It allows you to test various hypotheses under certain conditions, like what happens different days of the week or between different times of the day

```
In [227]: import matplotlib.pyplot as plt
import datetime
import numpy as np

x = np.array([datetime.datetime(2018, 9, 28, i, 0) for i in range(24)])
y = np.random.randint(100, size=x.shape)

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



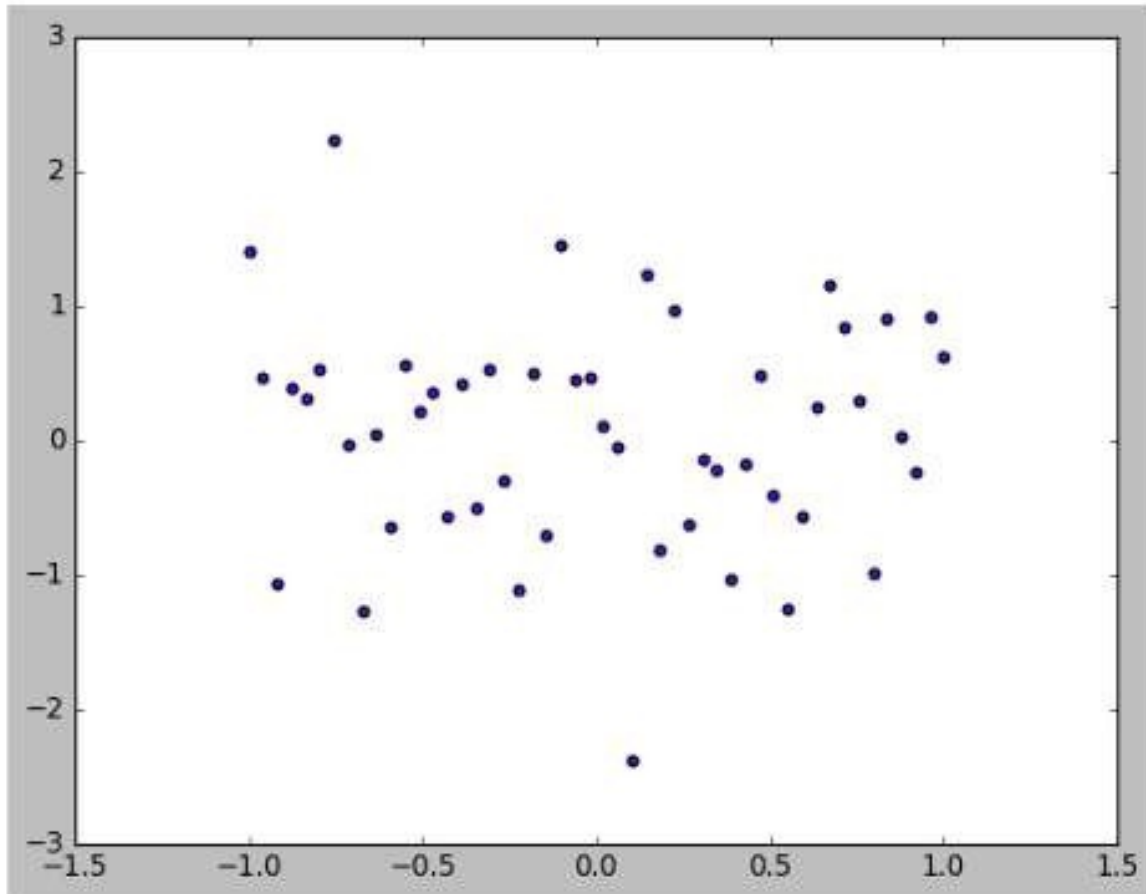
Scatter plots

- offer a convenient way to visualize how two numeric values are related in your data.
- It helps in understanding relationships between multiple variables.
- Using `.scatter()` method, we can create a scatter plot:

EXAMPLE

```
In [214]: fig, ax = plt.subplots()
x = np.linspace(-1, 1, 50)
y = np.random.randn(50)
ax.scatter(x, y)
```

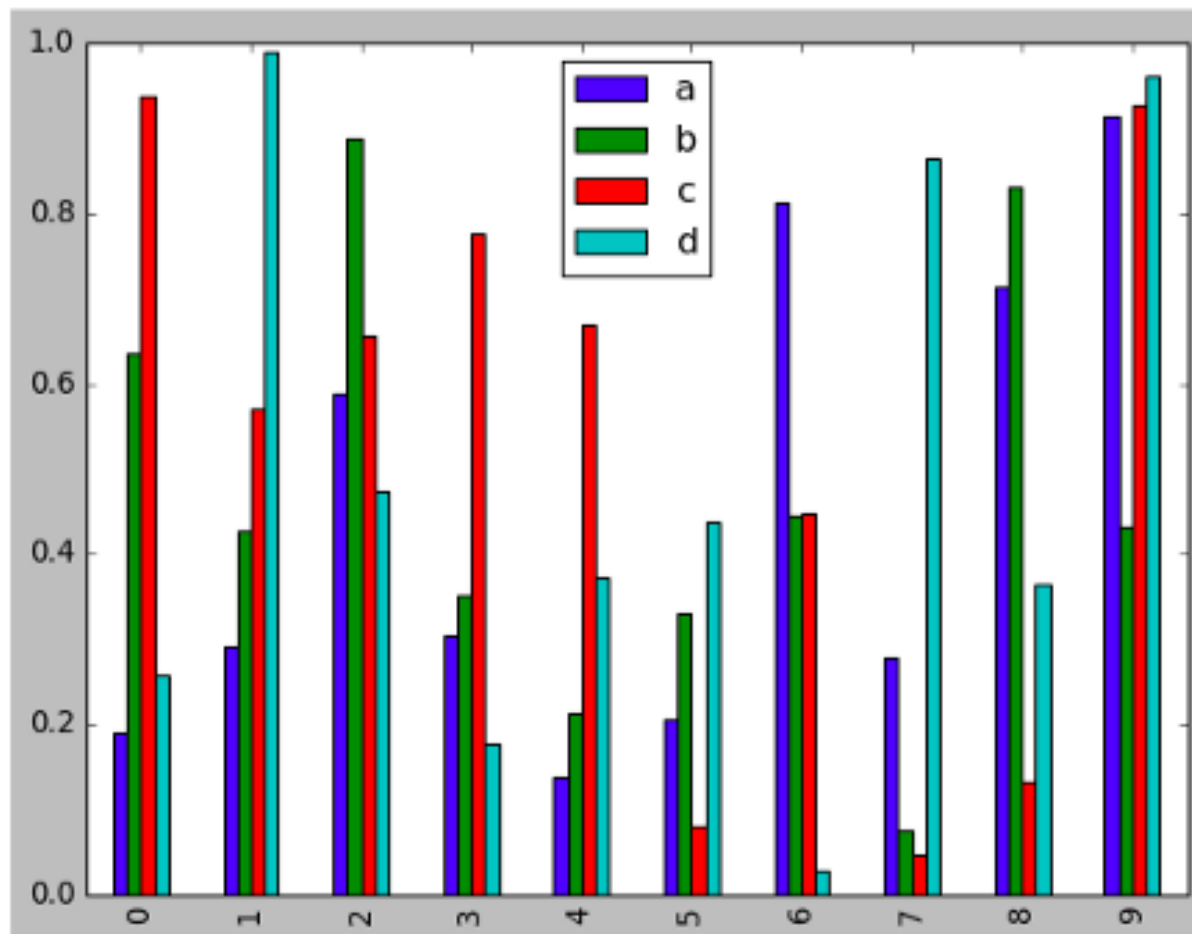
```
Out[214]: <matplotlib.collections.PathCollection at 0x7f004eaa30f0>
```



Bar graphs

- are convenient for comparing numeric values of several groups.
Using `.bar()` method, we can create a bar graph:

```
In [216]: my_df = pd.DataFrame(np.random.rand(10, 4), columns=['a', 'b', 'c', 'd'])  
my_df.plot.bar();
```



References

- <https://www.simplifiedpython.net/data-visualization-python-tutorial/>
- <https://matplotlib.org/>
- <https://mode.com/blog/python-data-visualization-libraries/>
- <https://towardsdatascience.com/top-6-python-libraries-for-visualization-which-one-to-use-fe43381cd658>