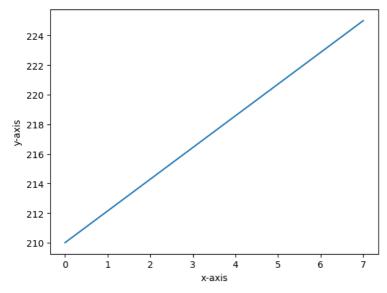
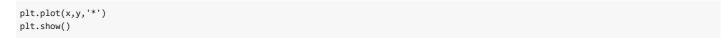
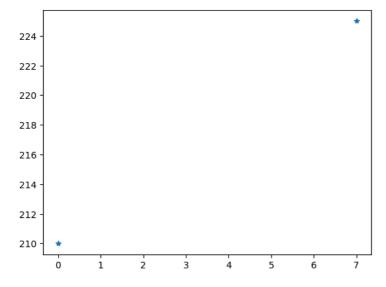
```
#draw a line in a diagram from position (0,0) to position (6,250):
import matplotlib.pyplot as plt
import numpy as np
x = np.array([0,7])
y = np.array([210,225])
plt.xlabel('x-axis')
plt.ylabel('y-axis')
plt.plot(x,y)
plt.show()
```

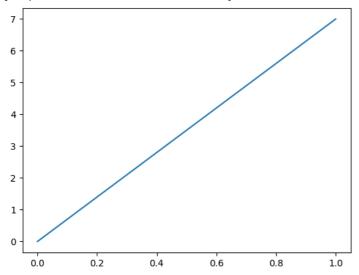






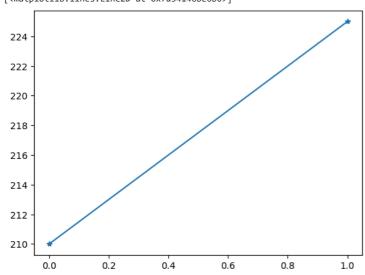
plt.plot(x)

[<matplotlib.lines.Line2D at 0x7d54151bdcf0>]



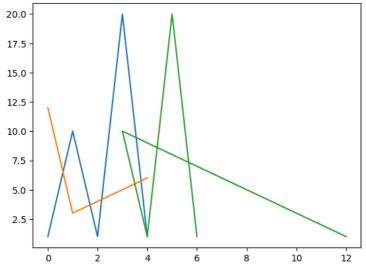
plt.plot(y,marker='*')

[<matplotlib.lines.Line2D at 0x7d54146be6b0>]



```
x = np.array([12,3,4,5,6])
y = np.array([1,10,1,20,1])
plt.plot(y)
plt.plot(x)
plt.plot(x,y)
```

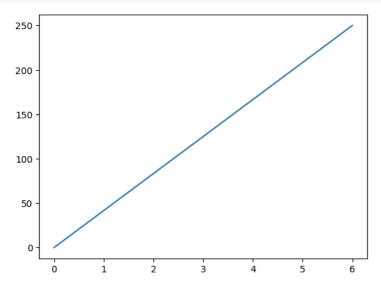
[<matplotlib.lines.Line2D at 0x7d54147d3400>]



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

xpoints = np.array([0, 6])
ypoints = np.array([0, 250])

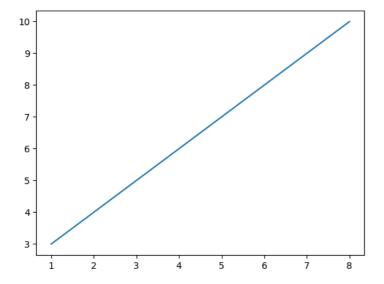
plt.plot(xpoints, ypoints)
plt.show()
```



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

xpoints = np.array([1, 8])
ypoints = np.array([3, 10])

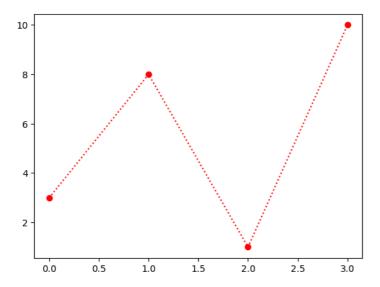
plt.plot(xpoints, ypoints)
plt.show()
```



```
import numpy as np

ypoints = np.array([3, 8, 1, 10])

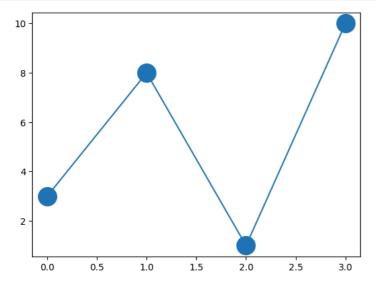
plt.plot(ypoints, 'o:r')
plt.show()
```



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

ypoints = np.array([3, 8, 1, 10])

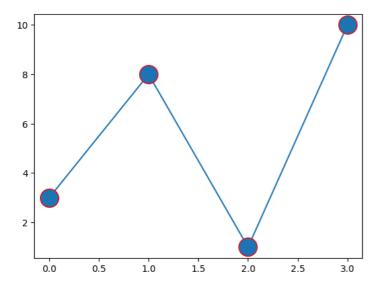
plt.plot(ypoints, marker = 'o', ms = 20)
plt.show()
```



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

ypoints = np.array([3, 8, 1, 10])

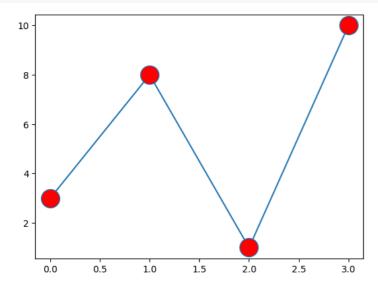
plt.plot(ypoints, marker = 'o', ms = 20, mec = 'r')
plt.show()
```



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

ypoints = np.array([3, 8, 1, 10])

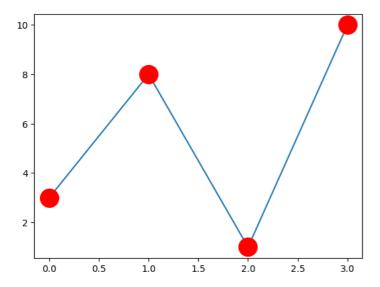
plt.plot(ypoints, marker = 'o', ms = 20, mfc = 'r')
plt.show()
```



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

ypoints = np.array([3, 8, 1, 10])

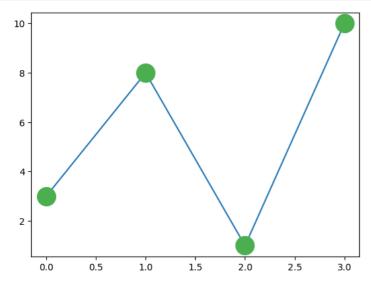
plt.plot(ypoints, marker = 'o', ms = 20, mec = 'r', mfc = 'r')
plt.show()
```



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

ypoints = np.array([3, 8, 1, 10])

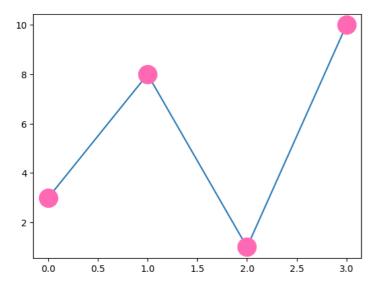
plt.plot(ypoints, marker = 'o', ms = 20, mec = '#4CAF50', mfc = '#4CAF50')
plt.show()
```



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

ypoints = np.array([3, 8, 1, 10])

plt.plot(ypoints, marker = 'o', ms = 20, mec = 'hotpink', mfc = 'hotpink')
plt.show()
```

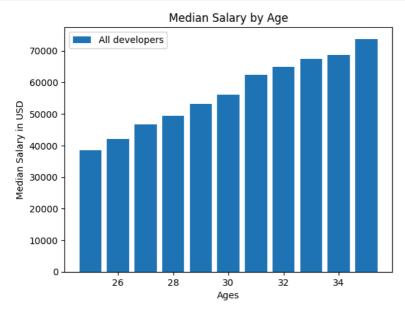


Bar_Charts

```
import pandas as pd
import matplotlib.pyplot as plt
x = [25, 26, 27, 28, 29, 30, 31, 32, 33, 34, 35]
devs_y = [38496, 42000, 46752, 49320, 53200, 56000, 62316, 64928, 67317, 68748, 73752]
```

Plotting the bar plot

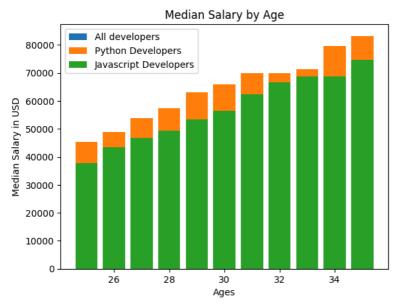
```
plt.bar(x, devs_y, label="All developers")
plt.xlabel("Ages")
plt.ylabel("Median Salary in USD")
plt.title("Median Salary by Age")
plt.legend()
plt.show()
```



2. Adding more bars to the same plot

```
py_devs_y = [45372, 48876, 53850, 57287, 63016, 65998, 70003, 70000, 71418, 79674, 83238]
js_devs_y = [37810, 43515, 46823, 49293, 53437, 56373, 62375, 66674, 68745, 68746, 74583]
```

```
plt.bar(x, devs_y, label="All developers")
plt.bar(x, py_devs_y, label="Python Developers")
plt.bar(x, js_devs_y, label="Javascript Developers")
plt.xlabel("Ages")
plt.ylabel("Median Salary in USD")
plt.title("Median Salary by Age")
plt.legend()
plt.show()
```



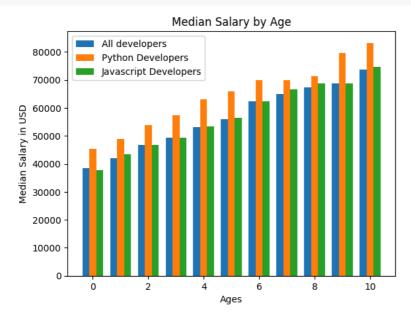
→ 3. Adjusting the width of the plot

```
import numpy as np

x_indexes = np.arange(len(x))

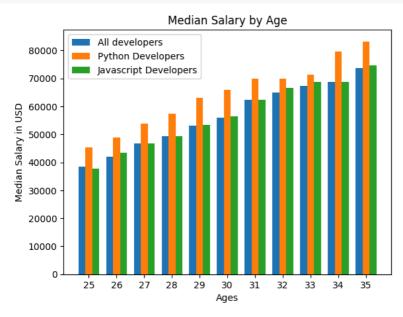
width = 0.25

plt.bar(x_indexes - width, devs_y, width=width, label="All developers")
plt.bar(x_indexes, py_devs_y, width = width, label="Python Developers")
plt.bar(x_indexes + width, js_devs_y, width=width, label="Javascript Developers")
plt.xlabel("Ages")
plt.ylabel("Median Salary in USD")
plt.title("Median Salary by Age")
plt.legend()
plt.show()
```



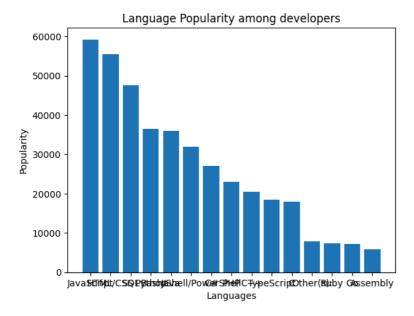
4. Changing the xlabels

```
plt.bar(x_indexes - width, devs_y, width=width, label="All developers")
plt.bar(x_indexes, py_devs_y, width = width, label="Python Developers")
plt.bar(x_indexes + width, js_devs_y, width=width, label="Javascript Developers")
plt.xlabel("Ages")
plt.ylabel("Median Salary in USD")
plt.title("Median Salary by Age")
plt.xticks(ticks=x_indexes, labels=x) #changing the xlabel
plt.legend()
plt.show()
```



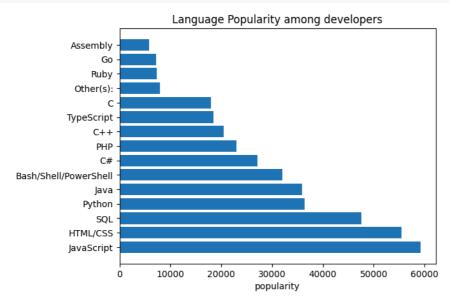
▼ 5. Plotting the bar plot from pandas dataframe

```
import pandas as pd
data = pd.read_csv('/content/data.csv')
from collections import Counter
ids = data['Responder_id']
language_responses = data['LanguagesWorkedWith']
language_counter = Counter()
for response in language_responses:
    language_counter.update(response.split(";"))
languages = []
popularity = []
for item in language_counter.most_common(15):
    languages.append(item[0])
    popularity.append(item[1])
print(languages)
print(popularity)
     ['JavaScript', 'HTML/CSS', 'SQL', 'Python', 'Java', 'Bash/Shell/PowerShell', 'C#', 'PHP', 'C++', 'TypeScript', 'C', 'Other(s):', 'Rub [59219, 55466, 47544, 36443, 35917, 31991, 27097, 23030, 20524, 18523, 18017, 7920, 7331, 7201, 5833]
plt.bar(languages, popularity)
plt.xlabel("Languages")
plt.ylabel("Popularity")
plt.title("Language Popularity among developers")
plt.show()
```



6. Plotting Horizontal bar chart

```
plt.barh(languages, popularity)
plt.xlabel("popularity")
plt.title("Language Popularity among developers")
plt.show()
```



languages.reverse()
popularity.reverse()

https://insights.stackoverflow.com/survey/2020#technology-programming-scriptingand-markup-languages

plt.figure(figsize=(8,6)) plt.barh(languages, popularity) plt.xlabel("popularity") plt.title("Language Popularity among developers") plt.tight_layout() plt.show()

Show Your Creativity

Automobile Land Speed Records (GR 5-10)

In the first recorded automobile race in 1898, Count Gaston de Chasseloup-Laubat of Paris, France, drove 1 kilometer in 57 seconds for an average speed of 39.2 miles per hour(mph) or 63.1 kilometers per hour (kph). In 1904, Henry Ford drove his Ford Arrow across frozen Lake St.

Clair, MI, at an average speed of 91.4 mph. Now, the North American Eagle is trying to break a land speed record of 800 mph. The Federation International deL'Automobile (FIA), the world's governing body for motor sport and land speed records, recorded the following land speed records.

```
import matplotlib.pyplot as plt
import pandas as pd

from google.colab import drive
data = pd.read_csv('/content/LandRecords.csv')
```

data.head()

	Speed (mph)	Driver	Car	Engine	Date
0	407.447	Craig Breedlove	Spirit of America	GE J47	8/5/1963
1	413.199	Tom Green	Wingfoot Express	WE J46	10/2/1964
2	434.220	Art Arfons	Green Monster	GE J79	10/5/1964
3	468.719	Craig Breedlove	Spirit of America	GE J79	10/13/1964
4	526.277	Craig Breedlove	Spirit of America	GE J79	10/15/1965

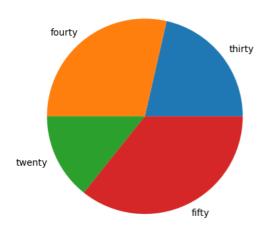
from matplotlib import pyplot as plt

→ 1. Plotting the Pie Chart

```
slices = [30, 40, 20, 50] #sum needs not be 100plt.pie(slices) plt.show()
```

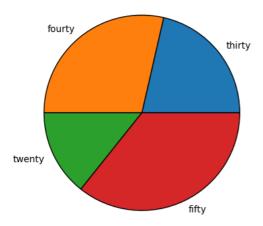
→ 2. Adding labels to the pie chart

```
labels = ['thirty','fourty', 'twenty','fifty']
plt.pie(slices, labels=labels)
plt.show()
```



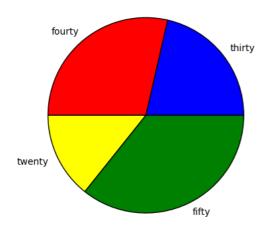
3. setting edge color

```
plt.pie(slices, labels=labels, wedgeprops={'edgecolor':'black'})
plt.show()
```



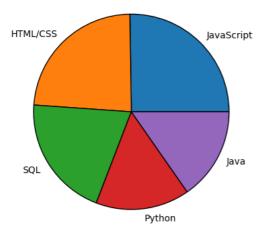

```
color = ['blue','red','yellow','green']

#hexadecimal color codes can also be used
plt.pie(slices, labels=labels, colors=color, wedgeprops={'edgecolor':'black'})
plt.show()
```



→ 5. plotting real world data

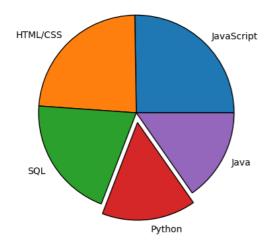
```
labels = ['JavaScript', 'HTML/CSS', 'SQL', 'Python', 'Java']
slices = [59219, 55466, 47544, 36443, 35917]
plt.pie(slices, labels=labels, wedgeprops={'edgecolor':'black'})
plt.show()
```



→ 6. Exploding the slice

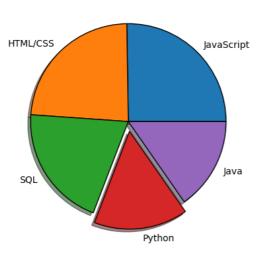
```
explode = [0, 0, 0, 0.1, 0]

plt.pie(slices, labels=labels, explode=explode, wedgeprops={'edgecolor':'black'})
plt.show()
```



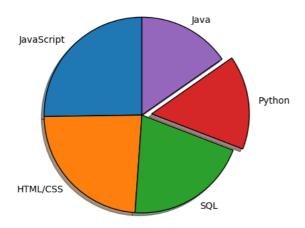
→ 7. adding shadow to the chart

plt.pie(slices, labels=labels, explode=explode, shadow=True, wedgeprops={'edgecolor':'black'})
plt.show()



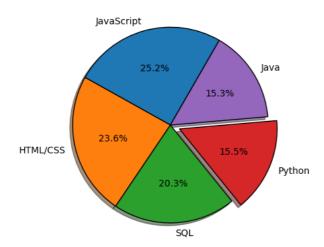
8. setting the starting angle

plt.pie(slices, labels=labels, explode=explode, shadow=True, startangle=90, wedgeprops={'edgecolor':'black'})
plt.show()



y 9. displaying percentage of each slices

plt.pie(slices, labels=labels, explode=explode, shadow=True, startangle=60, autopct="%0.1f%%", wedgeprops={'edgecolor':'black'})
plt.show()



Show Your Creativity

Covid 19 India Data as on 5th Sept 2020

import matplotlib.pyplot as plt
import pandas as pd

from google.colab import drive
data = pd.read_csv('/content/data.csv')
data.head()

Resp	onder_id	LanguagesWorkedWith				
0	1	HTML/CSS;Java;JavaScript;Python				
1	2	C++;HTML/CSS;Python				
2	3	HTML/CSS				
3	4	C;C++;C#;Python;SQL				
4	5	C++:HTML/CSS:Java:JavaScript:Pvthon:SQL:VBA				

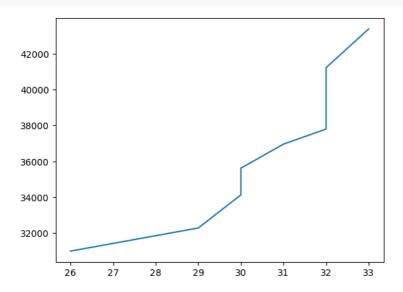
1. Creating Plots

```
# Installation: pip install matplotlib/ conda install matplotlib
import matplotlib.pyplot as plt
import random
# generating 10 random numbers between 25 to 35
ages = [random.randrange(25,35,1) for ages in range(11)]
ages = sorted(ages, reverse=False)
# generating 10 random numbers between 30k to 45k

devs = [random.randrange(30000,45000,1) for devs in range(11)]
devs = sorted(devs, reverse=False)
```

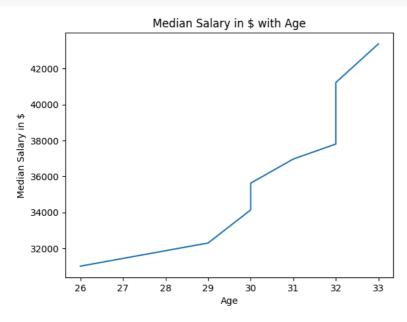
✓ 1.1. Plotting Line Plot

```
plt.plot(ages, devs)
plt.show()
```



✓ 1.2. Adding title, xlabel and ylabel

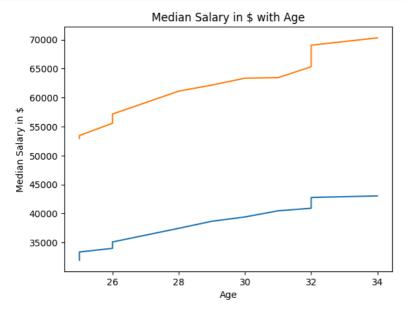
```
plt.plot(ages, devs)
plt.title("Median Salary in $ with Age") # add the title
plt.xlabel("Age") # add xlabel
plt.ylabel("Median Salary in $") #add ylabel
plt.show()
```



✓ 1.3. Adding more plot to the same graph

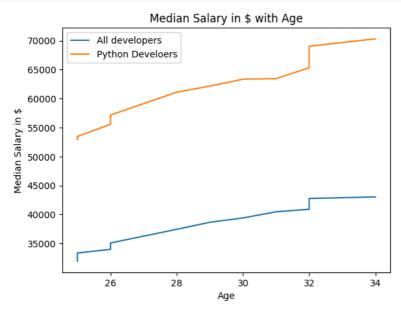
```
#creating 10 random numbers between 50k to 75k
import random
import matplotlib.pyplot as plt
ages = [random.randrange(25,35,1) for ages in range(11)]
ages = sorted(ages, reverse=False)
devs = [random.randrange(30000,45000,1) for devs in range(11)]
devs = sorted(devs, reverse=False)
py_devs = [random.randrange(50000,75000) for py_devs in range(11)]
py_devs = sorted(py_devs, reverse=False)

plt.plot(ages, devs)
plt.plot(ages, py_devs) # adding other plot to the same figure
plt.title("Median Salary in $ with Age")
plt.ylabel("Median Salary in $")
plt.ylabel("Median Salary in $")
plt.show()
```



✓ 1.4. Adding legend to the plot

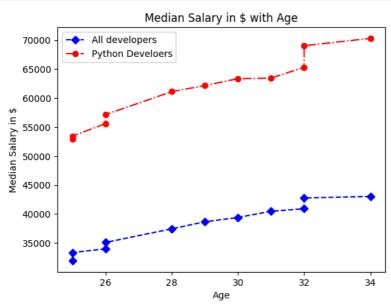
```
plt.plot(ages, devs, label = "All developers") # label
plt.plot(ages, py_devs, label = "Python Develoers")
plt.title("Median Salary in $ with Age")
plt.xlabel("Age")
plt.ylabel("Median Salary in $")
plt.legend() #plot the legend
plt.show()
```



1.5. Setting marker, linestyle and color

```
#https://matplotlib.org/api/_as_gen/matplotlib.pyplot.plot.html

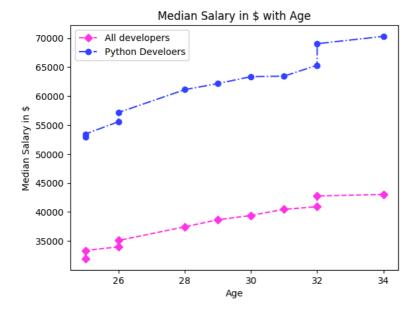
plt.plot(ages, devs, color="blue", linestyle = "--", marker = "D", label = "All developers")
plt.plot(ages, py_devs, color="red", linestyle = "-.", marker = "o", label = "Python Develoers")
plt.title("Median Salary in $ with Age")
plt.xlabel("Age")
plt.ylabel("Median Salary in $")
plt.legend()
plt.show()
```



1.6. Hexadecimal code for colors

```
#https://matplotlib.org/api/_as_gen/matplotlib.pyplot.plot.html

plt.plot(ages, devs, color="#FF33E9", linestyle = "--", marker = "D", label = "All developers")
plt.plot(ages, py_devs, color="#3344FF", linestyle = "--", marker = "o", label = "Python Develoers")
plt.title("Median Salary in $ with Age")
plt.xlabel("Age")
plt.ylabel("Median Salary in $")
plt.legend()
plt.show()
```



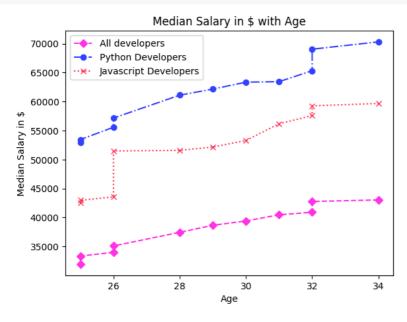
Adding other plot to the same graph

```
#creating 10 random numbers between 40k to 60k

js_devs = [random.randrange(40000,60000) for js_devs in range(11)]
js_devs = sorted(js_devs, reverse=False)

#https://matplotlib.org/api/_as_gen/matplotlib.pyplot.plot.html

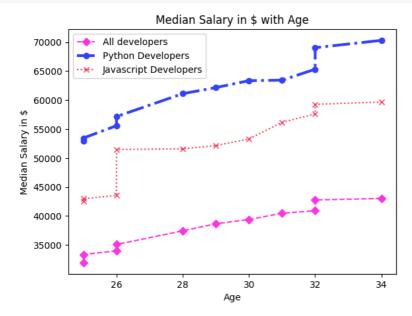
plt.plot(ages, devs, color="#FF33E9", linestyle = "--", marker = "D", label = "All developers")
plt.plot(ages, py_devs, color="#3344FF", linestyle = "--", marker = "o", label = "Python Developers")
plt.plot(ages, js_devs, color="#FF3355", linestyle = ":", marker = "x", label = "Javascript Developers")
plt.title("Median Salary in $ with Age")
plt.ylabel("Age")
plt.ylabel("Median Salary in $")
plt.legend()
plt.show()
```



✓ 1.7. Changing the line width

```
#https://matplotlib.org/api/_as_gen/matplotlib.pyplot.plot.html

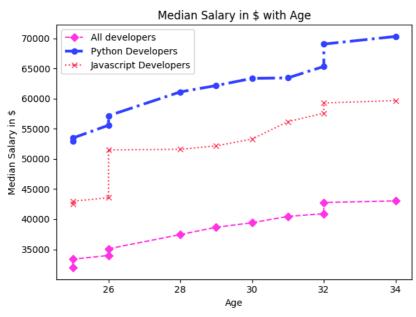
plt.plot(ages, devs, color="#FF33E9", linestyle = "--", marker = "D", label = "All developers")
plt.plot(ages, py_devs, color="#3344FF", linestyle = "--", marker = "o", linewidth=3, label = "Python Developers")
plt.plot(ages, js_devs, color="#FF3355", linestyle = ":", marker = "x", label = "Javascript Developers")
plt.title("Median Salary in $ with Age")
plt.xlabel("Age")
plt.ylabel("Median Salary in $")
plt.legend()
plt.show()
```



1.8. Add padding to the plot

```
#https://matplotlib.org/api/_as_gen/matplotlib.pyplot.plot.html

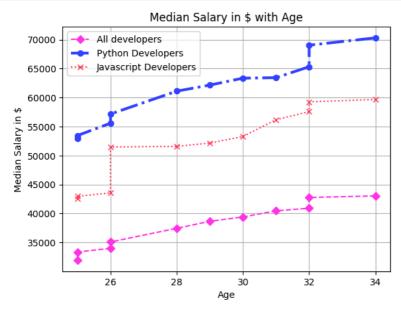
plt.plot(ages, devs, color="#FF33E9", linestyle = "--", marker = "D", label = "All developers")
plt.plot(ages, py_devs, color="#3344FF", linestyle = "--", marker = "o", linewidth=3, label = "Python Developers")
plt.plot(ages, js_devs, color="#FF3355", linestyle = ":", marker = "x", label = "Javascript Developers")
plt.title("Median Salary in $ with Age")
plt.xlabel("Age")
plt.ylabel("Median Salary in $")
plt.legend()
plt.tight_layout() #adds padding
plt.show()
```



1.9. Adding grid to the plot

```
#https://matplotlib.org/api/_as_gen/matplotlib.pyplot.plot.html

plt.plot(ages, devs, color="#FF33E9", linestyle = "--", marker = "D", label = "All developers")
plt.plot(ages, py_devs, color="#3344FF", linestyle = "--", marker = "o", linewidth=3, label = "Python Developers")
plt.plot(ages, js_devs, color="#FF3355", linestyle = ":", marker = "x", label = "Javascript Developers")
plt.title("Median Salary in $ with Age")
plt.xlabel("Age")
plt.ylabel("Median Salary in $")
plt.grid(True)
plt.legend()
plt.show()
```



1.10. Changing style of the plot

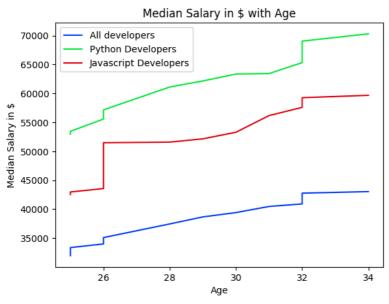
```
print(plt.style.available)

['Solarize_Light2', '_classic_test_patch', '_mpl-gallery', '_mpl-gallery-nogrid', 'bmh', 'classic', 'dark_background', 'fast', 'fiver'

#https://matplotlib.org/api/_as_gen/matplotlib.pyplot.plot.html

plt.style.use('seaborn-bright') #to change the style
plt.plot(ages, devs, label = "All developers")
plt.plot(ages, py_devs, label = "Python Developers")
plt.plot(ages, js_devs, label = "Javascript Developers")
plt.title("Median Salary in $ with Age")
plt.xlabel("Age")
plt.ylabel("Median Salary in $")
plt.legend()
plt.show()
```

<ipython-input-67-d07ee214d04f>:3: MatplotlibDeprecationWarning: The seaborn styles shipped by Matplotlib are deprecated since 3.6, a
plt.style.use('seaborn-bright') #to change the style



→ 1.11. Saving the plot

```
#https://matplotlib.org/api/_as_gen/matplotlib.pyplot.plot.html

plt.style.use('ggplot')

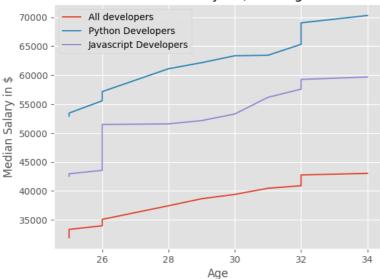
plt.plot(ages, devs, label = "All developers")
plt.plot(ages, py_devs, label = "Python Developers")
plt.plot(ages, js_devs, label = "Javascript Developers")

plt.title("Median Salary in $ with Age")
plt.xlabel("Age")
plt.ylabel("Median Salary in $")

plt.legend()
plt.savefig("plot.png")#save the plot

plt.show()
```

Median Salary in \$ with Age



for Further Reading click the below link

https://matplotlib.org/tutorials/introductory/pyplot.html

https://pythonbasics.org/matplotlib-line-chart/

```
import matplotlib.pyplot as plt
import pandas as pd
from google.colab import drive
data = pd.read_csv('/content/data_gapminder_gdp_oceania.csv',index_col='country')
print(data)

gdpPercap_1952 gdpPercap_1957 gdpPercap_1962 gdpPercap_1967 \
country
```

country	gapPercap_1952	gaprercap_1957	gapPercap_1962	gaprercap_1967	\
Australia New Zealand	10039.59564 10556.57566	10949.64959 12247.39532	12217.22686 13175.67800	14526.12465 14463.91893	
New Zealanu	10556.57566	12247.39332	131/3.0/800	14403.91893	
country	gdpPercap_1972	gdpPercap_1977	gdpPercap_1982	gdpPercap_1987	\
Australia	16788.62948	18334.19751	19477.00928	21888.88903	
New Zealand	16046.03728	16233.71770	17632.41040	19007.19129	
country	gdpPercap_1992	gdpPercap_1997	gdpPercap_2002	gdpPercap_2007	
Australia	23424.76683	26997.93657	30687.75473	34435.36744	
New Zealand	18363.32494	21050.41377	23189.80135	25185.00911	

Plot data directly from a Pandas dataframe.

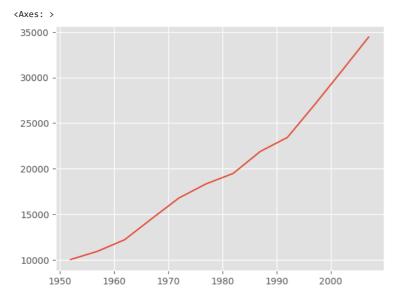
- We can also plot Pandas dataframes.
- This implicitly uses matplotlib.pyplot.
- Before plotting, we convert the column headings from a string to integer data type, since they represent numerical values

```
# Extract year from last 4 characters of each column name
# The current column names are structured as 'gdpPercap_(year)',
# so we want to keep the (year) part only for clarity when plotting GDP vs. years
# To do this we use strip(), which removes from the string the characters stated in the argument
# This method works on strings, so we call str before strip()

years = data.columns.str.strip('gdpPercap_')
# Convert year values to integers, saving results back to dataframe

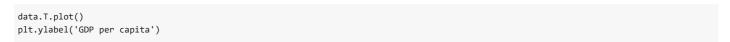
data.columns = years.astype(int)

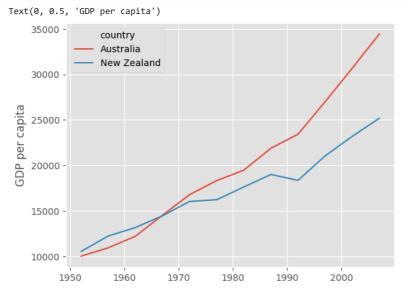
data.loc['Australia'].plot()
```



Select and transform data, then plot it.

- By default, DataFrame.plot plots with the rows as the X axis.
- We can transpose the data in order to plot multiple series.





Data can also be plotted by calling the matplotlib plot function directly.

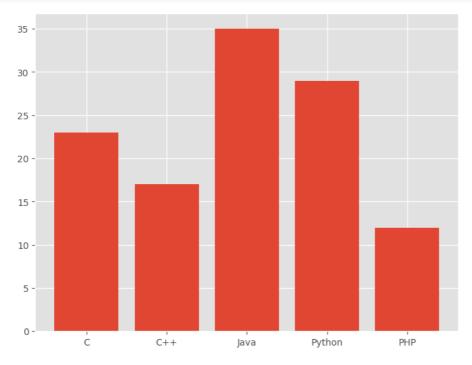
- The command is plt.plot(x, y)
- The color and format of markers can also be specified as an additional optional argument e.g., b- is a blue line, g-- is a green dashed line.

Get Australia data from dataframe

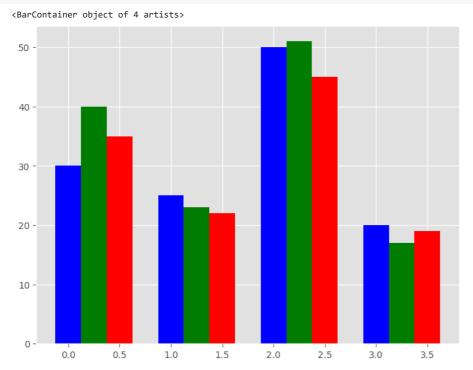
Plot with both countries

- · Select two countries' worth of data.
- Plot with differently-colored markers.
- · Create legend.

```
import matplotlib.pyplot as plt
fig = plt.figure()
ax = fig.add_axes([0,0,1,1])
langs = ['C', 'C++', 'Java', 'Python', 'PHP']
students = [23,17,35,29,12]
ax.bar(langs,students)
plt.show()
```



```
import numpy as np
import matplotlib.pyplot as plt
data = [[30, 25, 50, 20],
[40, 23, 51, 17],
[35, 22, 45, 19]]
X = np.arange(4)
fig = plt.figure()
ax = fig.add_axes([0,0,1,1])
ax.bar(X + 0.00, data[0], color = 'b', width = 0.25)
ax.bar(X + 0.25, data[1], color = 'g', width = 0.25)
ax.bar(X + 0.50, data[2], color = 'r', width = 0.25)
```



28-02-2024

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from sklearn.datasets import load breast cancer
breast = load_breast_cancer()
breast_data = breast.data
print(breast_data)
print(breast_data.shape)
    [[1.799e+01 1.038e+01 1.228e+02 ... 2.654e-01 4.601e-01 1.189e-01]
    [2.057e+01 1.777e+01 1.329e+02 ... 1.860e-01 2.750e-01 8.902e-02]
    [1.969e+01 2.125e+01 1.300e+02 ... 2.430e-01 3.613e-01 8.758e-02]
    [1.660e+01 2.808e+01 1.083e+02 ... 1.418e-01 2.218e-01 7.820e-02]
    [2.060e+01 2.933e+01 1.401e+02 ... 2.650e-01 4.087e-01 1.240e-01]
[7.760e+00 2.454e+01 4.792e+01 ... 0.000e+00 2.871e-01 7.039e-02]]
    (569, 30)
cancer = load breast cancer(as frame=True)
df=cancer.frame
df.head()
                                                               mean
                                                                               mean
        mean
              mean
                       mean
                            mean
                                      mean
                                               mean
                                                        mean
                                                                      mean
                                                                                        worst
                                                                                                 worst
                                                                                                      wors
                                                            concave
                                                                            fractal
      radius
            texture
                  perimeter
                            area
                                 smoothness
                                          compactness
                                                    concavity
                                                                   symmetry
                                                                                       texture perimeter
                                                                                                       are
                                                                           dimension
                                                             points
    0
       17.99
              10.38
                      122.80 1001.0
                                    0.11840
                                              0.27760
                                                       0.3001
                                                             0.14710
                                                                     0.2419
                                                                             0.07871
                                                                                         17.33
                                                                                                184.60 2019
       20.57
              17 77
                      132.90
                            1326.0
                                    0.08474
                                              0.07864
                                                       0.0869
                                                             0.07017
                                                                     0.1812
                                                                             0.05667
                                                                                         23.41
                                                                                                158.80
                                                                                                      1956.
    1
    2
       19.69
              21.25
                      130.00
                           1203.0
                                    0.10960
                                              0.15990
                                                       0.1974
                                                            0.12790
                                                                     0.2069
                                                                             0.05999
                                                                                         25.53
                                                                                                152.50
                                                                                                      1709.
    3
                                    0.14250
                                                                                                 98.87
       11.42
              20.38
                      77.58
                            386.1
                                              0.28390
                                                       0.2414
                                                             0.10520
                                                                     0.2597
                                                                             0.09744
                                                                                         26.50
                                                                                                       567
    4
       20.29
              14.34
                      135.10 1297.0
                                    0.10030
                                              0.13280
                                                       0.1980
                                                            0.10430
                                                                     0.1809
                                                                             0.05883
                                                                                         16.67
                                                                                                152.20 1575.
   5 rows × 31 columns
df.shape
   (569, 31)
breast_labels = breast.target
print(breast labels)
print(breast_labels.shape)
    100000000101111100100111100100111000
    101110110010000100010101011010000110011
    10111110011011001011111011110100000000
    101101011111111111111101111010111100011
    10111110110111111111111010010111111010
    1\;1\;1\;1\;1\;1\;0\;1\;0\;1\;1\;1\;1\;1\;1\;1\;0\;0\;1\;0\;1\;1\;1\;1\;1\;1\;0\;1\;1\;0\;1\;0\;1\;0\;0
    11111110000001]
    (569,)
labels = np.reshape(breast_labels,(569,1))
final_breast_data = np.concatenate([breast_data,labels],axis=1)
print(final_breast_data.shape)
    (569, 31)
breast_dataset = pd.DataFrame(final_breast_data)
print(breast_dataset.head())
                                                           8
                      1001.0 0.11840
   a
     17.99
          10.38 122.80
                                    0.27760
                                          0.3001
                                                0.14710
                                                        0.2419
                            0.08474
   1
     20.57
           17.77
                132.90
                      1326.0
                                    0.07864
                                          0.0869
                                                0.07017
                                                        0.1812
     19.69
           21.25 130.00
                      1203.0 0.10960
                                   0.15990
                                          0.1974
                                                0.12790
                                                        0.2069
           20.38
                 77.58
                       386.1 0.14250
                                   0.28390
                                          0.2414
                                                0.10520
                                                        0.2597
     11.42
```

```
4 20.29 14.34 135.10 1297.0 0.10030 0.13280 0.1980 0.10430 0.1809

9 ... 21 22 23 24 25 26 27 \
0 0.07871 ... 17.33 184.60 2019.0 0.1622 0.6656 0.7119 0.2654
1 0.05667 ... 23.41 158.80 1956.0 0.1238 0.1866 0.2416 0.1860
2 0.05999 ... 25.53 152.50 1709.0 0.1444 0.4245 0.4504 0.2430
3 0.09744 ... 26.50 98.87 567.7 0.2098 0.8663 0.6869 0.2575
4 0.05883 ... 16.67 152.20 1575.0 0.1374 0.2050 0.4000 0.1625
```

28 29 30 0 0.4601 0.11890 0.0 1 0.2750 0.08902 0.0 2 0.3613 0.08758 0.0 3 0.6638 0.17300 0.0 4 0.2364 0.07678 0.0

[5 rows x 31 columns]

```
features = breast.feature_names
print(features)
```

```
['mean radius' 'mean texture' 'mean perimeter' 'mean area' 'mean smoothness' 'mean compactness' 'mean concavity' 'mean concave points' 'mean symmetry' 'mean fractal dimension' 'radius error' 'texture error' 'perimeter error' 'area error' 'smoothness error' 'compactness error' 'concavity error' 'concave points error' 'symmetry error' 'fractal dimension error' 'worst radius' 'worst texture' 'worst perimeter' 'worst area' 'worst smoothness' 'worst compactness' 'worst concavity' 'worst concave points' 'worst symmetry' 'worst fractal dimension']
```

features_labels = np.append(features, 'label')

breast_dataset.columns = features_labels
breast_dataset.head()

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	mean fractal dimension	 worst texture	worst perimeter	wors are
0	17.99	10.38	122.80	1001.0	0.11840	0.27760	0.3001	0.14710	0.2419	0.07871	 17.33	184.60	2019.
1	20.57	17.77	132.90	1326.0	0.08474	0.07864	0.0869	0.07017	0.1812	0.05667	 23.41	158.80	1956.
2	19.69	21.25	130.00	1203.0	0.10960	0.15990	0.1974	0.12790	0.2069	0.05999	 25.53	152.50	1709.
3	11.42	20.38	77.58	386.1	0.14250	0.28390	0.2414	0.10520	0.2597	0.09744	 26.50	98.87	567.
4	20.29	14.34	135.10	1297.0	0.10030	0.13280	0.1980	0.10430	0.1809	0.05883	 16.67	152.20	1575.
5 rows x 31 columns													

```
breast_dataset['label'].replace(0,'benign',inplace = True)
breast_dataset['label'].replace(1,'malignant',inplace = True)
breast_dataset.tail()
```

	mean radius	mean texture	mean perimeter	mean area	mean smoothness	mean compactness	mean concavity	mean concave points	mean symmetry	mean fractal dimension	•••	worst texture	worst perimeter	wo a
564	21.56	22.39	142.00	1479.0	0.11100	0.11590	0.24390	0.13890	0.1726	0.05623		26.40	166.10	202
565	20.13	28.25	131.20	1261.0	0.09780	0.10340	0.14400	0.09791	0.1752	0.05533		38.25	155.00	173
566	16.60	28.08	108.30	858.1	0.08455	0.10230	0.09251	0.05302	0.1590	0.05648		34.12	126.70	112
567	20.60	29.33	140.10	1265.0	0.11780	0.27700	0.35140	0.15200	0.2397	0.07016		39.42	184.60	182
568	7.76	24.54	47.92	181.0	0.05263	0.04362	0.00000	0.00000	0.1587	0.05884		30.37	59.16	26
_														

5 rows × 31 columns

```
from sklearn.preprocessing import StandardScaler
x = breast_dataset.loc[:, features].values
x = StandardScaler().fit_transform(x) # normalizing the features
print(x.shape)
```

(569, 30)

```
np.mean(x), np.std(x)
```

(-6.118909323768877e-16, 1.0)

```
feat_cols = ['features'+str(1) for i in range (x.shape[1])]
normalised_breast = pd.DataFrame(x,columns = feat_cols)
print(normalised_breast)
         features1 features1 features1 features1 features1 \
                   -2.073335 1.269934 0.984375
-0.353632 1.685955 1.908708
    0
          1.097064
                                                  1.568466
                                                             3.283515
          1.829821 -0.353632
                                         1.908708 -0.826962 -0.487072
    1
                    0.456187 1.566503 1.558884
                                                    0.942210 1.052926
          1.579888
     3
         -0.768909
                    0.253732 -0.592687 -0.764464
                                                    3.283553
                                                              3,402909
                  -1.151816 1.776573 1.826229
                                                    0.280372 0.539340
          1.750297
                    0.721473 2.060786
     564
         2.110995
                                         2.343856
                                                  1.041842 0.219060
                               1.615931
     565
          1.704854
                    2.085134
                                         1.723842
                                                    0.102458 -0.017833
    566
          0.702284
                    2.045574 0.672676
                                         0.577953 -0.840484 -0.038680
          1.838341
                              1.982524
                                                  1.525767
                                                              3.272144
    567
                    2.336457
                                         1.735218
                    1.221792 -1.814389 -1.347789 -3.112085 -1.150752
         -1.808401
    568
         features1 features1 features1 ...
                                                        features1 features1
                                         2.255747 ...
    0
          2.652874
                    2,532475
                               2.217515
                                                         1.886690 -1.359293
                               0.001392 -0.868652 ...
    1
         -0.023846
                    0.548144
                                                         1.805927 -0.369203
                             0.93500.
2.867383
                               0.939685 -0.398008 ...
    2
          1.363478
                    2.037231
                                                        1.511870 -0.023974
                                                        -0.281464 0.133984
    3
          1.915897
                    1.451707
                                        4.910919
                                                  . . .
                  1.428493 -0.009560 -0.562450 ...
                                                        1.298575 -1.466770
     4
          1.371011
                                                   ...
                    2.320965 -0.312589 -0.931027 ...
     564
         1.947285
                                                         1.901185
                                                                  0.117700
     565
          0.693043
                    1.263669
                             -0.217664
                                        -1.058611
                                                         1.536720
                                                                   2.047399
                                                   . . .
                                                                   1.374854
     566
          0.046588
                    0.105777
                             -0.809117 -0.895587
                                                         0.561361
                                                   . . .
     567
          3.296944
                    2.658866
                               2.137194
                                         1.043695
                                                        1.961239
                                                                   2.237926
                                                  . . .
    568
         -1.114873 -1.261820 -0.820070 -0.561032 ...
                                                        -1.410893
                                                                  0.764190
         features1 features1 features1 features1 \
    0
                    2,001237
                               1.307686
                                                   2,109526
                                                              2,296076
          2.303601
                                         2,616665
                    1.890489 -0.375612 -0.430444 -0.146749
          1.535126
                                                              1.087084
    1
                    1.456285 0.527407
                                                              1.955000
    2
          1.347475
                                         1.082932
                                                   0.854974
    3
         -0.249939
                  -0.550021 3.394275
                                         3.893397
                                                   1.989588
                                                              2.175786
     4
          1.338539
                  1.220724 0.220556 -0.313395 0.613179 0.729259
     564
          1.752563
                    2.015301
                              0.378365 -0.273318
                                                    0.664512
                                                              1.629151
     565
          1.421940
                    1.494959
                              -0.691230
                                        -0.394820
                                                    0.236573
                                                              0.733827
          0.579001
                    0.427906 -0.809587
                                         0.350735
                                                              0.414069
     566
                                                    0.326767
          2.303601
                    1.653171
                               1.430427
                                         3.904848
                                                    3.197605
                                                              2.289985
                  -1.075813 -1.859019 -1.207552 -1.305831 -1.745063
    568
         -1.432735
         features1 features1
    0
          2.750622
                    1.937015
                    0.281190
         -0.243890
    1
    2
          1.152255
                    0.201391
    3
          6.046041
                    4.935010
    4
         -0.868353 -0.397100
         -1.360158 -0.709091
    564
    565
         -0.531855
                   -0.973978
    566 -1.104549
                   -0.318409
    567
          1.919083
                    2.219635
    568 -0.048138 -0.751207
    [569 rows x 30 columns]
```

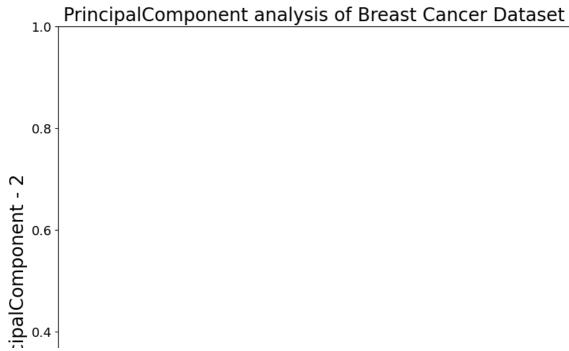
normalised_breast.head()

```
features1 features1 features1 features1 features1 features1 features1
               -2.073335
                           1.269934
                                                                       2.652874
                                                                                  2.532
    1.097064
                                      0.984375
                                                 1.568466
                                                            3.283515
1
    1.829821
               -0.353632
                           1.685955
                                     1.908708
                                               -0.826962
                                                           -0.487072
                                                                      -0.023846
                                                                                  0.548
    1.579888
               0.456187
                           1.566503
                                     1.558884
                                                0.942210
                                                            1.052926
                                                                       1.363478
                                                                                  2.037
    -0 768909
3
               0.253732
                          -0.592687
                                     -0.764464
                                                3.283553
                                                            3.402909
                                                                       1.915897
                                                                                  1.451
    1.750297
               -1.151816
                           1.776573
                                     1.826229
                                                 0.280372
                                                            0.539340
                                                                       1.371011
                                                                                  1 428
5 rows × 30 columns
```

```
from sklearn.decomposition import PCA
pca_breast = PCA(n_components=2)
PrincipalComponents_breast = pca_breast.fit_transform(x)
Principal_breast_Df = pd.DataFrame(data = PrincipalComponents_breast,columns = ['principal component 1','principal component 2'])
Principal_breast_Df.tail()
```

	principal component 1	principal component 2
564	6.439315	-3.576817
565	3.793382	-3.584048
566	1.256179	-1.902297
567	10.374794	1.672010
568	-5.475243	-0.670637

```
Traceback (most recent call last)
/usr/local/lib/python3.10/dist-packages/pandas/core/indexes/base.py in get_loc(self, key, method, tolerance)
-> 3802
                       return self._engine.get_loc(casted_key)
                    except KeyError as err:
                              — 💲 10 frames -
pandas/_libs/hashtable_class_helper.pxi in pandas._libs.hashtable.PyObjectHashTable.get_item()
pandas/_libs/hashtable_class_helper.pxi in pandas._libs.hashtable.PyObjectHashTable.get_item()
KeyError: 'Principal component 1'
The above exception was the direct cause of the following exception:
                                          Traceback (most recent call last)
/usr/local/lib/python3.10/dist-packages/pandas/core/indexes/base.py in get_loc(self, key, method, tolerance)
                        return self._engine.get_loc(casted_key)
                    except KeyError as err:
-> 3804
                        raise KeyError(key) from err
  3805
                    except TypeError:
                        # If we have a listlike key, _check_indexing_error will raise
KeyError: 'Principal component 1'
<Figure size 640x480 with 0 Axes>
```



2.3 correlation Regression

