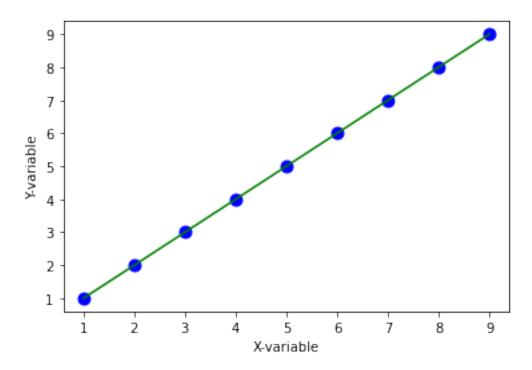
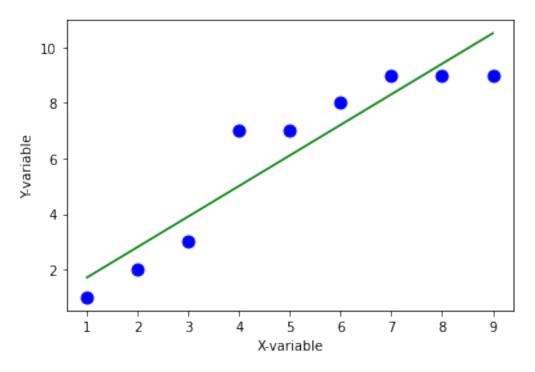
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from operator import mul
import math
def compute regcoef(x,y):
    n = len(x)
    sumxy = sum(list(map(mul,x,y)))
    num = n* sumxy - sum (x)*sum(y)
    sumxx = sum(list(map(mul,x,x)))
    denom = n*sumxx - sum(x)**2
    m = num/denom
    C = (sum(y) - m*sum(x))/n
    return(c,m)
def plot regline(x,y,b):
    plt.scatter(x,y,color='b',marker='o',s=80)
    y_pred = np.float_(x)*b[1] + b[0]
    plt.plot(x,y_pred,color='g')
    plt.xlabel('X-variable')
    plt.ylabel('Y-variable')
    plt.show()
x = [1,2,3,4,5,6,7,8,9]
y = [1,2,3,4,5,6,7,8,9]
b= compute_regcoef(x,y)
type(b)
print('intercept',b[0])
print('slope',b[1])
plot regline(x,y,b)
intercept 0.0
slope 1.0
```

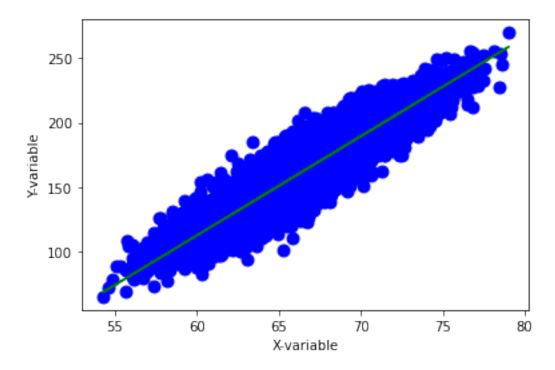


```
x= [1,2,3,4,5,6,7,8,9]
y= [1,2,3,7,7,8,9,9,9]
b= compute_regcoef(x,y)
type(b)
print('intercept',b[0])
print('slope',b[1])
plot_regline(x,y,b)
intercept 0.61111111111111103
slope 1.1
```



```
df=pd.read csv('weight-height.csv')
print(df)
      Gender
                 Height
                              Weight
0
        Male
              73.847017
                          241.893563
1
        Male
              68.781904
                          162.310473
2
        Male
              74.110105
                          212.740856
3
              71.730978
        Male
                          220.042470
4
        Male
              69.881796
                          206.349801
9995
      Female
              66.172652
                          136.777454
9996
      Female
              67.067155
                          170.867906
9997
      Female
              63.867992
                          128.475319
9998
      Female
              69.034243
                          163.852461
9999
      Female 61.944246
                          113.649103
[10000 rows x 3 columns]
df.columns
Index(['Gender', 'Height', 'Weight'], dtype='object')
df.dtypes
Gender
           object
Height
          float64
Weight
          float64
dtype: object
```

```
x=df.iloc[:,1:2].values
y=df.iloc[:,2].values
b=compute_regcoef(x,y)
type(b)
print('intercept',b[0])
print('slope',b[1])
plot_regline(x,y,b)
intercept [-350.73719181]
slope [7.71728764]
```



```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.25,rand
om_state=0)

from sklearn.linear_model import LinearRegression
regression= LinearRegression()
regression.fit(x_train,y_train)

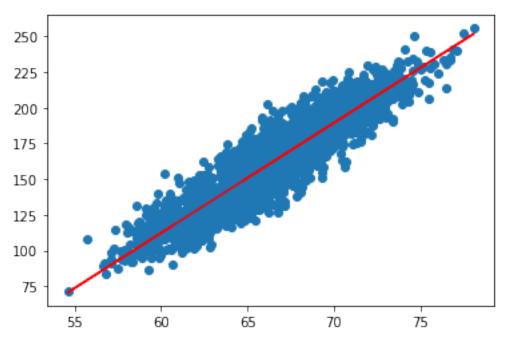
LinearRegression()

m=regression.coef_
print('Regression Coefficient / slope of regression line',m)

Regression Coefficient / slope of regression line [7.71833809]

c= regression.intercept_
print('Intercept',c)
```

```
Intercept -350.84158094506427
y pred = regression.predict(x test)
print(y_pred)
[148.71918181 168.38777621 224.33647637 ... 177.05541851 152.24692441
117.824240991
dfl = pd.DataFrame({'Actual value':y test.flatten(),'Predicted
Value':y pred.flatten()})
dfl
     Actual value
                    Predicted Value
0
        138.085796
                         148.719182
1
        187.363366
                         168.387776
2
        216.533191
                         224.336476
3
        131.761443
                         157.642640
4
        157.718438
                         149.664097
2495
        121.094631
                         149.597631
2496
        207.493691
                         203.256212
2497
        169.634399
                         177.055419
2498
        169.459631
                         152,246924
2499
        123.628360
                         117.824241
[2500 rows x 2 columns]
x test[0]*m+c
array([148.71918181])
plt.scatter(x test, y test)
plt.plot(x_test,y_pred,color='r')
plt.show()
```



```
from sklearn import metrics
print('Mean Absolute
Error',metrics.mean_absolute_error(y_test,y_pred))
print('Mean Squared Error',metrics.mean_squared_error(y_test,y_pred))
print('Root Mean Squared
Error',np.sqrt(metrics.mean_absolute_error(y_test,y_pred)))
Mean Absolute Error 9.815091837529609
Mean Squared Error 151.46996027337102
Root Mean Squared Error 3.1329046965283847
```