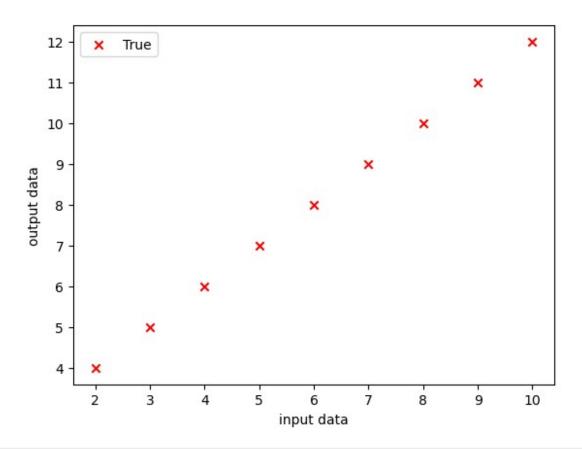
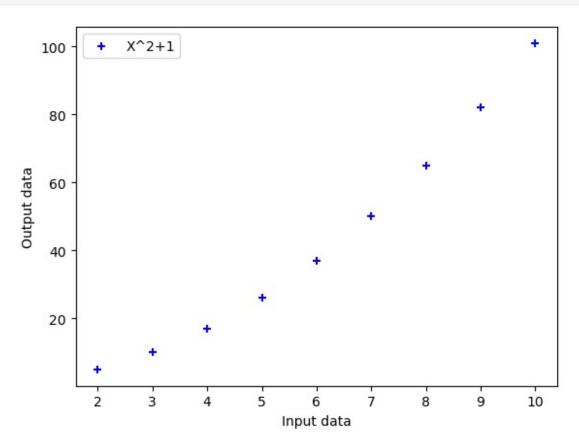
```
import numpy as np
import matplotlib.pyplot as plt
\#Y = X + 2
X = np.array([2,3,4,5,6,7,8,9,10])
Y = np.array([4,5,6,7,8,9,10,11,12])
len(X)
9
X = X.reshape(-1,1)
Y = Y.reshape(-1,1)
Χ
array([[ 2],
       [ 3],
       [4],
       [5],
       [ 6],
       [7],
       [8],
       [ 9],
       [10]])
print(f"size of input data : {len(X)}")
size of input data: 9
\#(2,4)(3,5)
for i in range(len(X)):
    print((X[i],Y[i]))
(2, 4)
(3, 5)
(4, 6)
(5, 7)
(6, 8)
(7, 9)
(8, 10)
(9, 11)
(10, 12)
plt.scatter(X,Y,marker='x',c='red',label='True')
plt.xlabel("input data")
plt.ylabel("output data")
plt.legend()
plt.show()
```



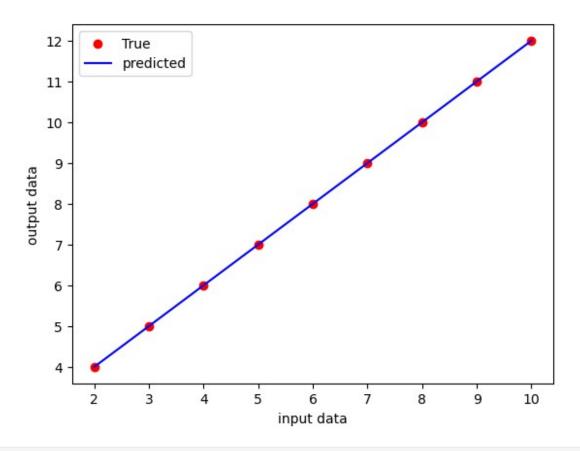
```
#y=x**2+1
x = np.array([2,3,4,5,6,7,8,9,10])
y= np.array([5,10,17,26,37,50,65,82,101])
print(f"size of x is: {len(x)}")
size of x is: 9
for i in range (len(x)):
    print((x[i],y[i]))
(2, 5)
(3, 10)
(4, 17)
(5, 26)
(6, 37)
(7, 50)
(8, 65)
(9, 82)
(10, 101)
plt.scatter(x,y,marker='+',c='b',label='X^2+1')
plt.xlabel("Input data")
plt.ylabel("Output data")
```

```
plt.legend()
plt.show()
```



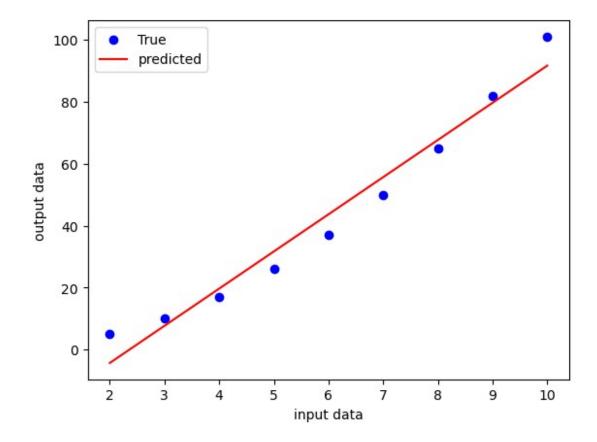
```
pip install scikit-learn
Requirement already satisfied: scikit-learn in
/opt/anaconda3/lib/python3.12/site-packages (1.5.1)
Requirement already satisfied: numpy>=1.19.5 in
/opt/anaconda3/lib/python3.12/site-packages (from scikit-learn)
(1.26.4)
Requirement already satisfied: scipy>=1.6.0 in
/opt/anaconda3/lib/python3.12/site-packages (from scikit-learn)
(1.13.1)
Requirement already satisfied: joblib>=1.2.0 in
/opt/anaconda3/lib/python3.12/site-packages (from scikit-learn)
(1.4.2)
Requirement already satisfied: threadpoolctl>=3.1.0 in
/opt/anaconda3/lib/python3.12/site-packages (from scikit-learn)
(3.5.0)
Note: you may need to restart the kernel to use updated packages.
from sklearn.linear model import LinearRegression
model LR =LinearRegression()
```

```
model_LR.fit(X,Y)
LinearRegression()
model_LR.predict([[11]])
array([[13.]])
test_x = np.array([2,3,4,5,6,7,8,9,10]).reshape(-1,1)
predict_y = model_LR.predict(test_x)
predict y
array([[ 4.],
       [5.],
       [ 6.],
       [7.],
       [8.],
       [ 9.],
       [10.],
       [11.],
       [12.]])
plt.scatter(X,Y,marker='o',c='red',label='True')
plt.plot(test_x,predict_y,c='blue',label='predicted')
plt.xlabel("input data")
plt.ylabel("output data")
plt.legend()
plt.show()
```



```
#y=x**2+1
X = np.array([2,3,4,5,6,7,8,9,10])
Y= np.array([5,10,17,26,37,50,65,82,101])
print(f"size of x is: {len(x)}")
size of x is: 9
X = X.reshape(-1,1)
Y = Y.reshape(-1,1)
Χ
array([[ 2],
         3],
         4],
         5],
         6],
        7],
        8],
       [ 9],
       [10]])
model_LR =LinearRegression()
model_LR.fit(X,Y)
```

```
LinearRegression()
model LR.predict([[11]])
array([[103.6666667]])
test_x = np.array([2,3,4,5,6,7,8,9,10]).reshape(-1,1)
predict_y = model_LR.predict(test_x)
predict_y
array([[-4.33333333],
        [ 7.66666667],
        [19.66666667],
        [31.66666667],
        [43.66666667],
        [55.66666667],
        [67.66666667],
        [79.66666667],
       [91.66666667]])
plt.scatter(X,Y,marker='o',c='blue',label='True')
plt.plot(test_x,predict_y,c='red',label='predicted')
plt.xlabel("input data")
plt.ylabel("output data")
plt.legend()
plt.show()
```



```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.linear model import LinearRegression
df = pd.read csv("Housing.csv")
df
        price
                                              stories mainroad guestroom
                       bedrooms
                                  bathrooms
                area
basement
0
     13300000
                7420
                                           2
                                                    3
                                                            yes
                                                                        no
no
1
     12250000
                8960
                                                    4
                                                            yes
                                                                        no
no
     12250000
                9960
                              3
                                           2
2
                                                    2
                                                            yes
                                                                        no
yes
     12215000
3
                7500
                                           2
                                                    2
                                                            yes
                                                                        no
yes
     11410000
                7420
4
                                                    2
                                                            yes
                                                                       yes
yes
. .
                                                                        . . .
. . .
      1820000
                3000
                              2
540
                                                    1
                                                            yes
                                                                        no
yes
541
      1767150
                2400
                              3
                                                    1
                                                             no
                                                                        no
no
      1750000
                3620
                              2
542
                                                    1
                                                            yes
                                                                        no
no
543
      1750000
                2910
                              3
                                                     1
                                                             no
                                                                        no
no
544
      1750000
               3850
                              3
                                                    2
                                                            yes
                                                                        no
no
    hotwaterheating airconditioning
                                        parking prefarea furnishingstatus
0
                  no
                                   yes
                                               2
                                                       yes
                                                                   furnished
                                               3
                                                                   furnished
1
                  no
                                                        no
                                   yes
2
                                               2
                                                       yes
                                                             semi-furnished
                  no
                                    no
3
                                                                   furnished
                                               3
                  no
                                   yes
                                                       yes
                                               2
                                                                   furnished
                  no
                                   yes
                                                        no
540
                                               2
                                                        no
                                                                 unfurnished
                  no
                                    no
```

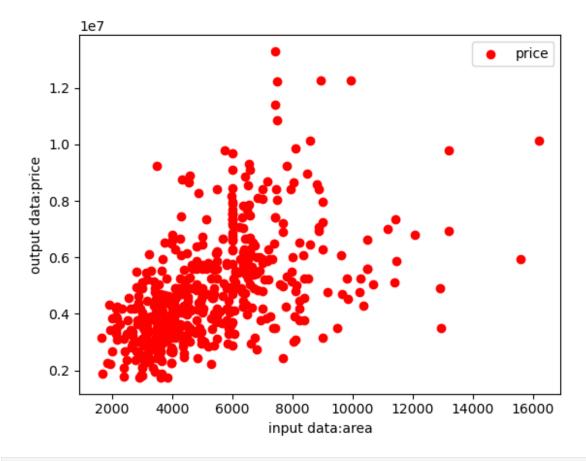
```
541
                                              0
                                                            semi-furnished
                                   no
                                                       no
                  no
542
                                                                unfurnished
                  no
                                   no
                                                       no
543
                                                                  furnished
                  no
                                   no
                                                       no
544
                                                                unfurnished
                  no
                                              0
                                                       no
                                   no
X=df['area']
Χ
0
       7420
       8960
1
2
       9960
3
       7500
4
       7420
540
       3000
541
       2400
       3620
542
543
       2910
544
       3850
Name: area, Length: 545, dtype: int64
Y=df['price']
Υ
0
       13300000
1
       12250000
2
       12250000
3
       12215000
4
       11410000
540
        1820000
541
        1767150
542
        1750000
543
        1750000
544
        1750000
Name: price, Length: 545, dtype: int64
X=np.array(X).reshape(-1,1)
Y=np.array(Y).reshape(-1,1)
Χ
array([[ 7420],
        [ 8960],
        [ 9960],
```

```
[ 2910],
    [ 3850]])

print(f"size of input data : {len(X)}")

size of input data : 545

plt.scatter(X,Y,marker='o',c='red',label='price')
plt.xlabel("input data:area")
plt.ylabel("output data:price")
plt.legend()
plt.show()
```

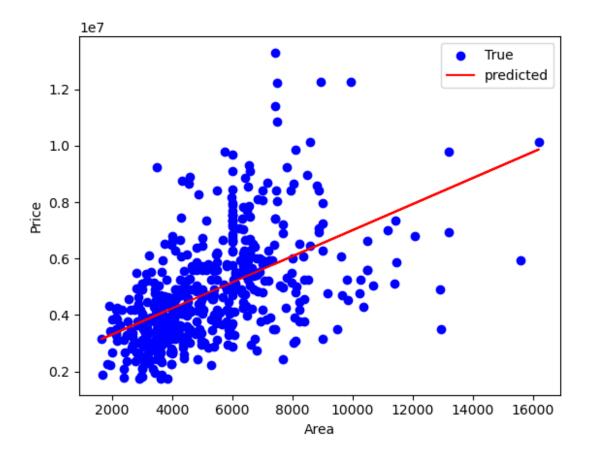


```
#LINEAR REGRESSION MODEL
model_LR = LinearRegression()

model_LR.fit(X,Y) #providing X & Y to the model
LinearRegression()
model_LR.predict([[2500]]) #predicting
array([[3542245.71807839]])
```

```
predict_y = model_LR.predict(X)

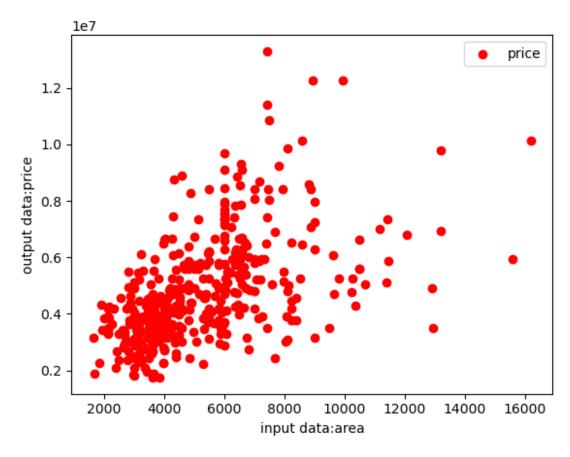
plt.scatter(X,Y,c="blue",label="True")
plt.plot(X,predict_y,c="red",label="predicted")
plt.xlabel("Area")
plt.ylabel("Price")
plt.legend()
plt.show()
```



```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.linear_model import LinearRegression
from sklearn.model selection import train test split
df = pd.read csv("Housing.csv")
df
        price area
                       bedrooms
                                  bathrooms
                                             stories mainroad guestroom
basement \
     13300000
                7420
                                          2
0
                              4
                                                    3
                                                            yes
                                                                        no
no
     12250000
                8960
1
                                                                        no
                                                            yes
no
2
     12250000
                9960
                              3
                                          2
                                                    2
                                                            yes
                                                                        no
yes
3
     12215000
                7500
                              4
                                          2
                                                    2
                                                            yes
                                                                        no
yes
4
     11410000
                7420
                                                    2
                                                            yes
                                                                       yes
yes
. .
. . .
540
      1820000
                3000
                                                    1
                                                            yes
                                                                        no
yes
541
      1767150
                2400
                              3
                                                    1
                                                             no
                                                                        no
no
                              2
542
      1750000
                3620
                                                    1
                                                            yes
                                                                        no
no
543
      1750000
                2910
                              3
                                                             no
                                                                        no
no
      1750000
                3850
544
                              3
                                                    2
                                                            yes
                                                                        no
no
    hotwaterheating airconditioning
                                        parking prefarea furnishingstatus
0
                                                                   furnished
                                               2
                  no
                                   yes
                                                       yes
1
                                                                   furnished
                  no
                                   yes
                                               3
                                                        no
2
                                               2
                                                             semi-furnished
                  no
                                    no
                                                      yes
3
                                               3
                                                                   furnished
                  no
                                   yes
                                                       yes
4
                                               2
                                                                   furnished
                  no
                                   yes
                                                        no
                                                                unfurnished
540
                                               2
                                                        no
                  no
                                    no
```

```
540
                  no
                                              2
                                                       no
                                                                unfurnished
                                   no
541
                                                            semi-furnished
                                              0
                  no
                                   no
                                                       no
542
                                                                unfurnished
                  no
                                   no
                                                       no
                                                                  furnished
543
                  no
                                   no
                                                       no
544
                                              0
                                                                unfurnished
                  no
                                   no
                                                       no
X=df['area']
Χ
0
       7420
1
       8960
2
       9960
3
       7500
4
       7420
540
       3000
541
       2400
542
       3620
543
       2910
       3850
544
Name: area, Length: 545, dtype: int64
Y=df['price']
Υ
0
       13300000
1
       12250000
2
       12250000
3
       12215000
4
       11410000
540
        1820000
        1767150
541
542
        1750000
543
        1750000
544
        1750000
Name: price, Length: 545, dtype: int64
X=np.array(X).reshape(-1,1)
Y=np.array(Y).reshape(-1,1)
#using sklearn split dataset
X train, X test, Y train, Y test = train test split(X, Y, test size = 0.2)
```

```
1.1.1
x train = X[0:435]
y_{train} = Y[0:435]
x test = X[435:]
y_{test} = Y[435:]
#manually
X \text{ train} = X[0:435]
Y \text{ train} = Y[0:435]
X \text{ test} = X[435:]
Y \text{ test} = Y[435:]
print(f"size of training data of X : {len(X train)}")
print(f"size of testing data of X : {len(X_test)}")
print(f"size of training data of Y : {len(Y train)}")
print(f"size of testing data of Y : {len(Y test)}")
size of training data of X : 436
size of testing data of X : 109
size of training data of Y: 436
size of testing data of Y : 109
plt.scatter(X train,Y train,marker='o',c='red',label='price')
plt.xlabel("input data:area")
plt.ylabel("output data:price")
plt.legend()
plt.show()
```



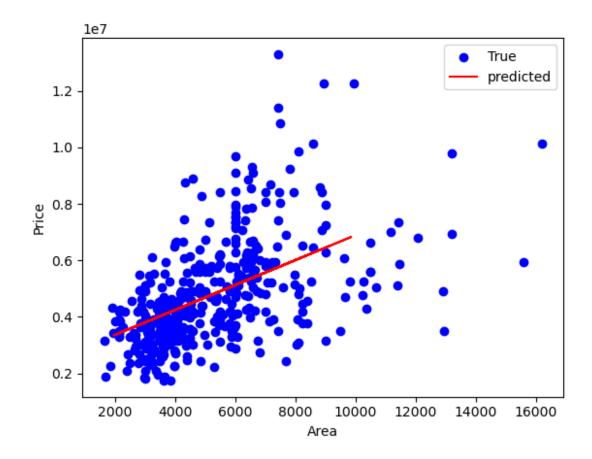
```
#LINEAR REGRESSION MODEL
model_LR = LinearRegression()

model_LR.fit(X_train,Y_train) #providing X & Y to the model
LinearRegression()

model_LR.predict([[2500]]) #predicting
array([[3585124.1100557]])

predict_y = model_LR.predict(X_test)

plt.scatter(X_train,Y_train,c="blue",label="True")
plt.plot(X_test,predict_y,c="red",label="predicted")
plt.xlabel("Area")
plt.ylabel("Price")
plt.legend()
plt.show()
```



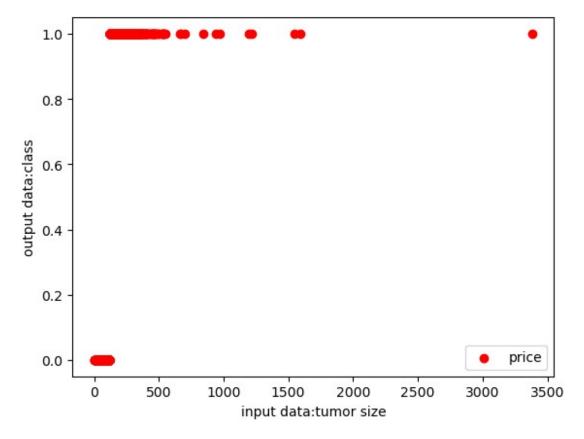
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from sklearn.linear_model import LogisticRegression
from sklearn.model selection import train test split
df = pd.read csv("Brain Tumor.csv")
df
      Class
             Tumor Size Unnamed: 2
0
               98.613971
                                 NaN
          0
1
          0
                                 NaN
               63.858816
2
          0
              81.867206
                                   S
3
          1
             151.229741
                                 NaN
4
          1
            174.988756
                                 NaN
. . .
                                 . . .
         . . .
            158.437600
          1
3743
                                 NaN
3744
          1
             161.158675
                                 NaN
3745
          1
             167.130118
                                 NaN
3746
          1
             223.812932
                                 NaN
          1 239.251388
3747
                                 NaN
[3748 rows x 3 columns]
df.head()
   Class
          Tumor Size Unnamed: 2
0
           98.613971
       0
                             NaN
1
       0
           63.858816
                             NaN
2
       0
           81.867206
                               S
3
          151.229741
       1
                             NaN
4
       1
          174.988756
                             NaN
df.tail()
      Class
            Tumor Size Unnamed: 2
            158.437600
3743
          1
                                 NaN
3744
          1
             161.158675
                                 NaN
          1
             167.130118
                                 NaN
3745
3746
          1
             223.812932
                                 NaN
3747
          1 239.251388
                                 NaN
X=df['Tumor_Size']
Χ
0
         98.613971
1
         63.858816
2
         81.867206
3
        151.229741
```

```
4
        174.988756
3743
        158.437600
3744
        161.158675
3745
        167.130118
3746
        223.812932
        239.251388
3747
Name: Tumor Size, Length: 3748, dtype: float64
Y=df['Class']
Υ
0
        0
        0
1
2
        0
3
        1
4
        1
3743
       1
3744
        1
3745
        1
3746
        1
3747
        1
Name: Class, Length: 3748, dtype: int64
X=np.array(X).reshape(-1,1)
Y=np.array(Y).reshape(-1,1)
#using sklearn split dataset
X train, X test, Y train, Y test = train test split(X, Y, test size = 0.2)
x_{train} = X[0:435]
y train = Y[0:435]
x test = X[435:]
y_{test} = Y[435:]
#manually
X train = X[0:435]
Y train = Y[0:435]
X \ test = X[435:]
Y \ test = Y[435:]
\n train = X[0:435]\ny train = Y[0:435]\nx test = X[435:]\ny test =
Y[435:]\n\mmanually\nX train = X[0:435]\nY train = Y[0:435]\nX test =
X[435:]\nY_{test} = Y[435:]\n'
print(f"size of training data of X : {len(X_train)}")
print(f"size of testing data of X : {len(X test)}")
```

```
print(f"size of training data of Y : {len(Y_train)}")
print(f"size of testing data of Y : {len(Y_test)}")

size of training data of X : 2998
size of testing data of Y : 2998
size of training data of Y : 2998
size of testing data of Y : 750

plt.scatter(X_train,Y_train,marker='o',c='red',label='price')
plt.xlabel("input data:tumor size")
plt.ylabel("output data:class")
plt.legend()
plt.show()
```

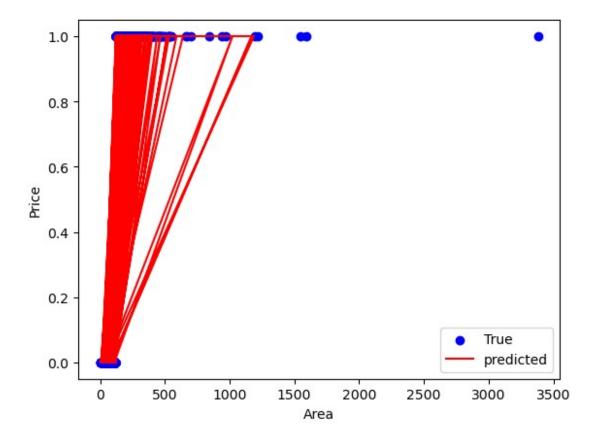


```
#LINEAR REGRESSION MODEL
model_LR = LogisticRegression()

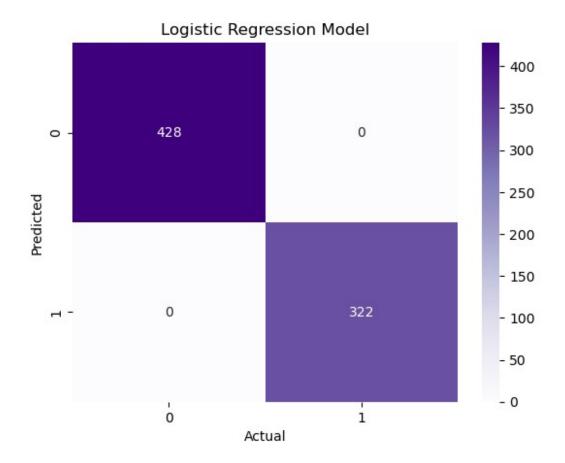
model_LR.fit(X_train,Y_train) #providing X & Y to the model

/opt/anaconda3/lib/python3.12/site-packages/sklearn/utils/
validation.py:1339: DataConversionWarning: A column-vector y was
passed when a 1d array was expected. Please change the shape of y to
(n_samples, ), for example using ravel().
    y = column_or_1d(y, warn=True)
```

```
LogisticRegression()
model_LR.predict([[2900]]) #predicting
array([1])
predict_y = model_LR.predict(X_test)
plt.scatter(X_train,Y_train,c="blue",label="True")
plt.plot(X_test,predict_y,c="red",label="predicted")
plt.xlabel("Area")
plt.ylabel("Price")
plt.legend()
plt.show()
```



```
cr = classification_report(predict_y,Y_test)
print (cr)
              precision
                            recall f1-score
                                               support
                              1.00
           0
                   1.00
                                        1.00
                                                    428
           1
                   1.00
                              1.00
                                        1.00
                                                    322
                                        1.00
                                                    750
    accuracy
                   1.00
                              1.00
                                        1.00
                                                    750
   macro avg
weighted avg
                                                    750
                    1.00
                              1.00
                                        1.00
acc = accuracy_score(predict_y,Y_test)
acc
1.0
import seaborn as sns
sns.heatmap(cm,cmap="Purples",annot=True,fmt=
".0f",xticklabels=[0,1],yticklabels=[0,1])
plt.xlabel("Actual")
plt.ylabel("Predicted")
plt.title("Logistic Regression Model")
plt.show()
```



sns.heatmap(cm,annot=True,cmap="Reds",fmt=".0f",xticklabels=[0,1],yticklabels=[0,1])

<Axes: >

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

x = np.array([185,170,168,179,182,188,180,180,183,180,180,177]);

y = np.array([72,56,60,68,72,77,71,70,84,88,67,76])

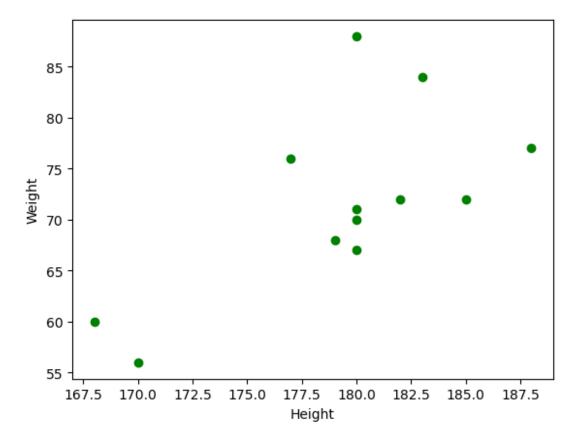
print(len(x))

12

print(len(y))

12

plt.scatter(x,y,marker='o',c='green')
plt.xlabel("Height")
plt.ylabel("Weight")
plt.show()
```



```
from sklearn.cluster import KMeans
d = np.vstack((x,y)).T
kmean = KMeans(n_clusters=3, random_state=0)
```

```
kmean.fit(d)

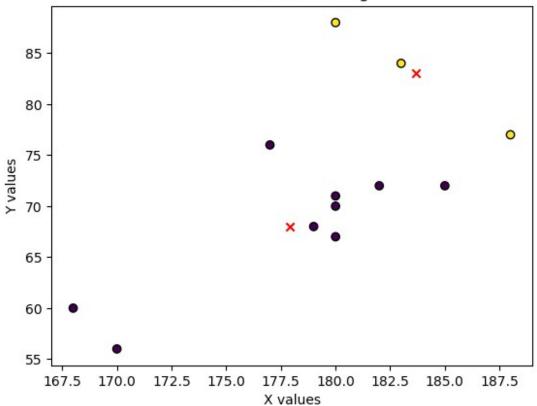
KMeans(n_clusters=3, random_state=0)

labels = kmeans.fit_predict(d)

cen = kmeans.cluster_centers_

plt.scatter(d[:, 0], d[:, 1], c=labels, cmap='viridis', marker='o', edgecolor='k')
plt.scatter(cen[:, 0], cen[:, 1], c='red', marker='x')
plt.state('K-Means Clustering')
plt.xlabel('X values')
plt.ylabel('Y values')
plt.show()
```

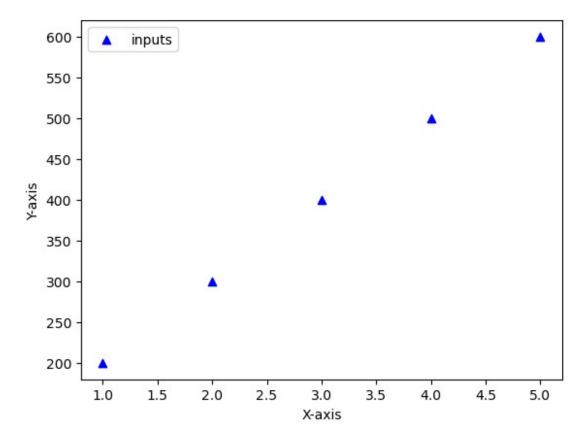
## K-Means Clustering



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

X=np.array([1.0,2.0,3.0,4.0,5.0])
Y=np.array([200,300,400,500,600])

#create scatter plot
plt.scatter(X,Y,c='blue',marker='^',label='inputs')
plt.xlabel("X-axis")
plt.ylabel("Y-axis")
plt.legend()
plt.show()
```



```
m=len(X)

#y^=f(x)=wx+b

def linear_regression_model(X,w,b):
    f_x = np.zeros(m)
    for i in range(m):
        f_x[i] = w*X[i]+b
    return f_x

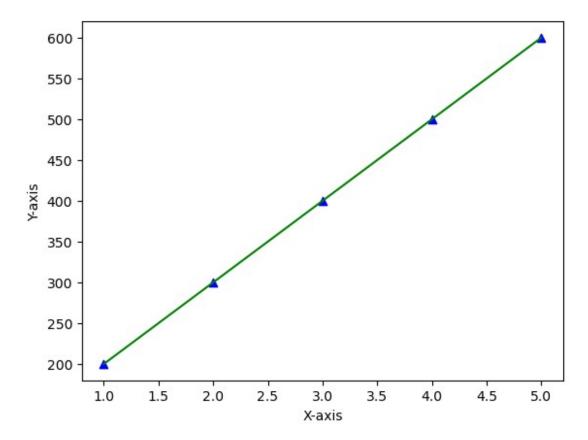
w=100
b=100
```

```
predict_y = linear_regression_model(X,w,b)

print(predict_y)

[200. 300. 400. 500. 600.]

plt.scatter(X,Y,c='blue',marker='^',label='inputs')
plt.plot(X,predict_y,c='green')
plt.xlabel("X-axis")
plt.ylabel("Y-axis")
plt.show()
```



```
def cost_function_linear_regression(X, Y, w, b):
    m = len(X)
    sum_cost = 0

for i in range(m):
    f_x = w * X[i] + b
    sum_cost += (f_x - Y[i]) ** 2

final_cost = (1 / (2 * m)) * sum_cost
    return final_cost

j_cost_w_b = cost_function_linear_regression(X,Y,w,b)
```

```
print(f"minimum cost:{j cost w b}")
minimum cost:0.0
def gradient descent(X,Y,w,b):
    j_dw, j_db=0, 0
    m = len(X)
    for i in range(m):
          f x = w * X[i] + b
          j dw += (f x-Y[i])*X[i]
          j_db += (f_x-Y[i])
    j dw=j dw/m
    j_db=j_db/m
    return j_dw,j_db
j dw,j db = gradient descent(X,Y,w,b)
print(f"j dw:{j dw} j db:{j dw}")
j_dw:0.0 j_db:0.0
gradient_descent_algo(X,Y,tempw,tempb,iteration,learningrate,gradient_
descent):
    w,b=tempw,tempb
    for i in range(iteration):
        j dw,j db=gradient descent(X,Y,w,b)
        w = w - learningrate* j_dw
        b = b - learningrate* j db
    return w,b
tempw=0
tempb=0
learningrate=0.001
iteration = 1000
w,b =
gradient descent algo(X,Y,tempw,tempb,iteration,learningrate,gradient
descent)
for i in range(iteration):
print(f"w:{w} b:{b}")
w:115.70560348553238 b:43.29466213948348
w:115.70560348553238 b:43.29466213948348
w:115.70560348553238 b:43.29466213948348
w:115.70560348553238 b:43.29466213948348
w:115.70560348553238 b:43.29466213948348
```

```
w:115.70560348553238 b:43.29466213948348
print(f"w:{w} b:{b}")
w:115.70560348553238 b:43.29466213948348
```

```
In [3]:
        import pandas as pd
        import numpy as np
        import matplotlib.pyplot as plt
In [4]: df = pd.read_csv("Brain_Tumor.csv")
        df.head()
Out[4]:
            Class Tumor_Size Unnamed: 2
         0
               0
                   98.613971
                                     NaN
         1
               0
                   63.858816
                                     NaN
         2
               0
                   81.867206
                                       S
         3
                  151.229741
                                     NaN
         4
               1
                  174.988756
                                     NaN
In [5]: X=df["Tumor_Size"]
        Y=df["Class"]
        plt.scatter(X,Y,marker='o',c='red')
In [6]:
        plt.xlabel("Tumor_Size")
        plt.ylabel("Class")
        plt.show()
          1.0
          0.8
          0.6
          0.4
          0.2
          0.0
                          500
                                   1000
                                            1500
                                                      2000
                                                                2500
                                                                         3000
                                                                                   3500
                                             Tumor_Size
        def sigmoid_func(x,w,b):
In [7]:
            f_x = np.dot(x,w)+b
            d = np.exp(-f_x) + 1
            return 1/d;
```

```
In [11]: def cost_function(x,y,w,b):
              m = 1
              f_x = np.dot(x,w)+b
              pred = sigmoid_func(x,w,b)
              cost = (-1/m)*np.sum(y * np.log(pred)+(1-y)*np.log(1-pred))
              return cost
In [15]: X = np.array(X).reshape(-1,1)
         Y = np.array(Y).reshape(-1,1)
In [17]: X_{train} = X[0:2623]
         X_{\text{test}} = X[2623:]
         Y_train = Y[0:2623]
         Y_{test} = Y[2623:]
In [19]:
         w=0
          b=3
          x=X_test
          y=Y_test
           print(sigmoid_func(x,w,b))
           print(f"cost :{cost_function(x,y,w,b)}")
        [[0.95257413]
         [0.95257413]
         [0.95257413]
         . . .
         [0.95257413]
         [0.95257413]
         [0.95257413]]
        cost :1824.6607705204615
 In [ ]:
```