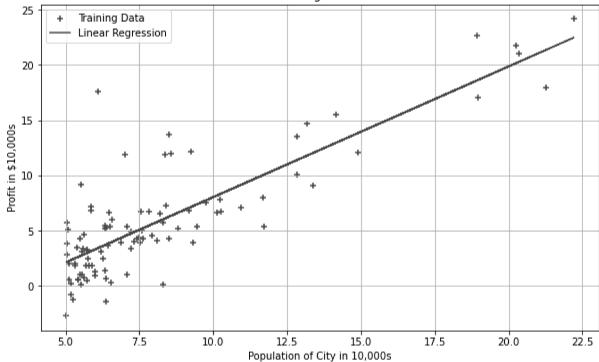
```
import numpy as np
 In [1]:
          import pandas as pd
          import matplotlib.pyplot as plt
In [18]: df = pd.read_csv('https://raw.githubusercontent.com/satishgunjal/datasets/master/univa
          #df = pd.DataFrame(pd.read_csv("UProfits.csv"))
          df.head(100) # To get first n rows from the dataset default value of n is 5
Out[18]:
             population
                          profit
           0
                 6.1101 17.59200
           1
                         9.13020
                 5.5277
           2
                 8.5186 13.66200
           3
                 7.0032 11.85400
           4
                 5.8598
                         6.82330
          92
                 5.8707
                         7.20290
          93
                 5.3054
                         1.98690
          94
                 8.2934
                         0.14454
          95
                13.3940
                         9.05510
          96
                 5.4369
                         0.61705
         97 rows × 2 columns
In [19]:
         len(df)
Out[19]:
In [20]: dataset = df.values[:,:] # get input values from first column
          print('dataset = ', dataset[:10,:])
         dataset = [[ 6.1101 17.592 ]
          [ 5.5277 9.1302]
           [ 8.5186 13.662 ]
           [ 7.0032 11.854 ]
           [ 5.8598 6.8233]
           [ 8.3829 11.886 ]
           [ 7.4764 4.3483]
           [ 8.5781 12.
           [ 6.4862 6.5987]
           [ 5.0546 3.8166]]
In [21]: X = df.values[:, 0] # get input values from first column
          y = df.values[:, 1] # get output values from second column
          len(X), len(y)
         (97, 97)
Out[21]:
In [22]: print('X = ', X[: 5]) # Show only first 5 records
```

```
print('y = ', y[: 5])
          X = [6.1101 \ 5.5277 \ 8.5186 \ 7.0032 \ 5.8598]
          y = [17.592]
                          9.1302 13.662 11.854
                                                  6.8233]
In [23]: print('X = ', X[:5]) # Show only first 5 records
          print('y = ', y[:5])
          X = [6.1101 \ 5.5277 \ 8.5186 \ 7.0032 \ 5.8598]
          y = [17.592 \quad 9.1302 \quad 13.662 \quad 11.854]
                                                   6.8233]
In [24]: plt.scatter(X,y, color='red',marker= '+')
          plt.grid()
          plt.rcParams["figure.figsize"] = (6,6)
          plt.xlabel('Population of City in 10,000s')
          plt.ylabel('Profit in $10,000s')
          plt.title('Scatter plot of training data')
          Text(0.5, 1.0, 'Scatter plot of training data')
Out[24]:
                            Scatter plot of training data
            25
            20
            15
          Profit in $10,000s
            10
             5
             0
                 5.0
                       7.5
                                    12.5
                                           15.0
                                                  17.5
                                                        20.0
                                                               22.5
                              Population of City in 10,000s
In [25]:
          # Using reshape function convert X 1D array to 2D array of dimension 97x1
          m=len(X)
          X_1 = X.reshape(m, 1)
          print('X_1 = ', X_1[:5,:]) # Show only first 5 records
          X_1 = [[6.1101]]
           [5.5277]
           [8.5186]
           [7.0032]
           [5.8598]]
In [26]: #Lets create a matrix with single column of ones
          m=len(X)
          \#X_0 = np.array((m, 1))
          X_0 = np.ones((m, 1))
          X 0[:5], len(X 0)
```

```
(array([[1.],
Out[26]:
                  [1.],
                 [1.],
                 [1.],
                  [1.]]),
          97)
In [27]: # Lets use hstack() function from numpy to stack X_0 and X_1 horizontally (i.e. column
         # This will be our final X matrix (feature matrix)
         X = np.hstack((X 0, X 1))
         X[:5]
         array([[1.
                       , 6.1101],
Out[27]:
                     , 5.5277],
                [1.
                [1.
                      , 8.5186],
                [1.
                       , 7.0032],
                [1.
                       , 5.8598]])
In [28]:
         theta = np.zeros((2,1))
         theta
         array([[0.],
Out[28]:
                [0.]])
         def compute_loss(X, y, theta):
In [29]:
           Compute loss for linear regression.
           Input Parameters
           X : 2D array where each row represent the training example and each column represent
               m= number of training examples
               n= number of features (including X_0 column of ones)
           y: 1D array of labels/target value for each traing example. dimension(m)
           theta: 2D array of fitting parameters or weights. Dimension (n,1)
           Output Parameters
            J : Scalar value : Loss
            predictions = X.dot(theta) #prediction = h
           errors = np.subtract(predictions, y)
            sqrErrors = np.square(errors)
            J = 1 / (2 * m) * np.sum(sqrErrors)
            return J
In [30]: # Lets compute the cost for theta values
         cost = compute_loss(X, y, theta)
         print('The cost for given values of theta 0 and theta 1 = ', cost)
         The cost for given values of theta_0 and theta_1 = 3111.0551861132
In [33]:
         def gradient_descent(X, y, theta, alpha, iterations):
           Compute cost for linear regression.
           Input Parameters
```

```
X : 2D, Dimension(m x n)
               m= number of training data point
               n= number of features (including X_0 column of ones)
           y : 1D array of labels/target value for each traing data point. dimension(m)
            theta: 2D array of fitting parameters or weights. Dimension (n,1)
            alpha: Learning rate. Scalar value
            iterations: Number of iterations. Scalar value.
            Output Parameters
            theta : Final Value. 2D array of fitting parameters or weights. Dimension (n,1)
            Loss-history: Conatins value of cost for each iteration. 1D array. Dimansion(m)
            loss_history = np.zeros(iterations)
            for i in range(iterations):
              predictions = X.dot(theta) #prediction (M,1)
              errors = np.subtract(predictions, y) #Error (M,1) = temp
              sum delta = (alpha / m) * X.transpose().dot(errors); #sum delta (n,1)
              theta = theta - sum delta; \#theta (n,1)
              loss_history[i] = compute_loss(X, y, theta)
            return theta, loss_history
In [51]: | theta = [0., 0.]
          iterations = 2000;
          alpha = 0.01;
In [52]: theta, loss_history = gradient_descent(X, y, theta, alpha, iterations)
          print('Final value of theta =', theta)
          print('cost_history =', loss_history)
         Final value of theta = [-3.78806857 1.18221277]
         cost history = [6.73719046 5.93159357 5.90115471 ... 4.47803526 4.47803143 4.4780276
In [53]: # Since X is list of list (feature matrix) lets take values of column of index 1 only
          plt.scatter(X[:,1], y, color='red', marker= '+', label= 'Training Data')
          plt.plot(X[:,1],X.dot(theta), color='green', label='Linear Regression')
          plt.rcParams["figure.figsize"] = (10,6)
          plt.grid()
          plt.xlabel('Population of City in 10,000s')
          plt.ylabel('Profit in $10,000s')
          plt.title('Linear Regression Fit')
          plt.legend()
Out[53]: <matplotlib.legend.Legend at 0x200dd3e2490>
```





```
In [54]: plt.plot(range(1, iterations + 1),loss_history, color='blue')
    plt.rcParams["figure.figsize"] = (10,6)
    plt.grid()
    plt.xlabel('Number of iterations')
    plt.ylabel('Cost (J)')
    plt.title('Convergence of gradient descent')
```

Out[54]: Text(0.5, 1.0, 'Convergence of gradient descent')

