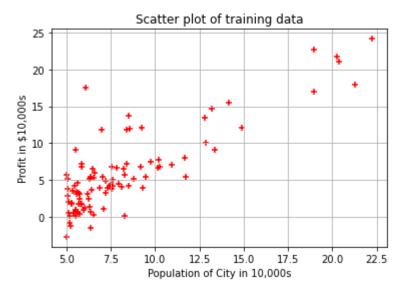
```
In [3]:
         import numpy as np
         import pandas as pd
         import matplotlib.pyplot as plt
In [6]:
         df = pd.read csv('https://raw.githubusercontent.com/satishgunjal/datasets/master/univar
         df.head() # To get first n rows from the dataset default value of n is 5
         M=len(df)
         Μ
Out[6]: 97
In [7]:
         X = df.values[:, 0] # get input values from first column
         y = df.values[:, 1] # get output values from second column
         m = len(y) # Number of training examples
         print('X = ', X[: 5]) # Show only first 5 records
         print('y = ', y[: 5])
         print('m = ', m)
        X = [6.1101 \ 5.5277 \ 8.5186 \ 7.0032 \ 5.8598]
                     9.1302 13.662 11.854
            [17.592
                                               6.82331
        m = 97
In [8]:
         X = df.values[:, 0] # get input values from first column
         y = df.values[:, 1] # get output values from second column
         m = len(y) # Number of training examples
         print('X = ', X[: 97]) # Show only first 5 records
         print('y = ', y[: 97])
         print('m = ', m)
        X = \begin{bmatrix} 6.1101 & 5.5277 & 8.5186 & 7.0032 & 5.8598 & 8.3829 & 7.4764 & 8.5781 & 6.4862 \end{bmatrix}
          5.0546 5.7107 14.164
                                  5.734
                                          8.4084 5.6407 5.3794 6.3654 5.1301
          6.4296 7.0708 6.1891 20.27
                                          5.4901
                                                  6.3261
                                                         5.5649 18.945 12.828
         10.957 13.176 22.203
                                  5.2524 6.5894
                                                 9.2482 5.8918 8.2111 7.9334
          8.0959 5.6063 12.836
                                  6.3534 5.4069
                                                 6.8825 11.708
                                                                  5.7737 7.8247
                                          5.5416
                                                 7.5402 5.3077
          7.0931 5.0702 5.8014 11.7
                                                                 7.4239 7.6031
          6.3328 6.3589 6.2742 5.6397 9.3102
                                                  9.4536 8.8254
                                                                5.1793 21.279
         14.908 18.959
                          7.2182 8.2951 10.236
                                                  5.4994 20.341 10.136
                                                                          7.3345
          6.0062 7.2259 5.0269 6.5479
                                         7.5386 5.0365 10.274
                                                                  5.1077
                                                                         5.7292
          5.1884 6.3557 9.7687
                                  6.5159
                                         8.5172 9.1802 6.002
                                                                  5.5204 5.0594
          5.7077 7.6366 5.8707 5.3054 8.2934 13.394
                                                          5.43691
        y = [17.592]
                        9.1302 13.662
                                         11.854
                                                   6.8233 11.886
                                                                     4.3483 12.
          6.5987
                            3.2522 15.505
                                              3.1551
                                                      7.2258
                                                                0.71618 3.5129
                   3.8166
          5.3048
                   0.56077 3.6518
                                    5.3893
                                              3.1386
                                                     21.767
                                                                4.263
                                                                         5.1875
                           13.501
                                     7.0467 14.692
                                                      24.147
                                                               -1.22
                                                                         5.9966
          3.0825 22.638
         12.134
                                             4.1164
                                                       3.3928 10.117
                   1.8495
                            6.5426
                                    4.5623
                                                                         5.4974
          0.55657 3.9115
                            5.3854
                                     2.4406
                                              6.7318
                                                       1.0463
                                                                5.1337
                                                                         1.844
          8.0043
                   1.0179
                            6.7504
                                     1.8396
                                              4.2885
                                                       4.9981
                                                                1.4233
                                                                        -1.4211
                                                     -0.74279 17.929
          2.4756
                   4.6042
                            3.9624
                                     5.4141
                                              5.1694
                                                                        12.054
         17.054
                   4.8852
                           5.7442
                                     7.7754
                                              1.0173 20.992
                                                                6.6799
                                                                         4.0259
          1.2784
                   3.3411 -2.6807
                                     0.29678 3.8845
                                                     5.7014
                                                                6.7526
                                                                         2.0576
          0.47953 0.20421 0.67861 7.5435
                                              5.3436
                                                       4.2415
                                                                6.7981
                                                                         0.92695
                                                       1.9869
          0.152
                   2.8214
                            1.8451
                                     4.2959
                                              7.2029
                                                                0.14454 9.0551
          0.61705]
        m = 97
In [9]:
         plt.scatter(X,y, color='red',marker= '+')
```

```
plt.grid()
plt.rcParams["figure.figsize"] = (10,6)
plt.xlabel('Population of City in 10,000s')
plt.ylabel('Profit in $10,000s')
plt.title('Scatter plot of training data')
```

## Out[9]: Text(0.5, 1.0, 'Scatter plot of training data')



```
In [10]:
#Lets create a matrix with single column of ones
X_0 = np.ones((m, 1))
X_0[:5]
```

```
In [11]:
# Using reshape function convert X 1D array to 2D array of dimension 97x1
X_1 = X.reshape(m, 1)
X_1[:10]
```

```
In [12]:
# 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]
```

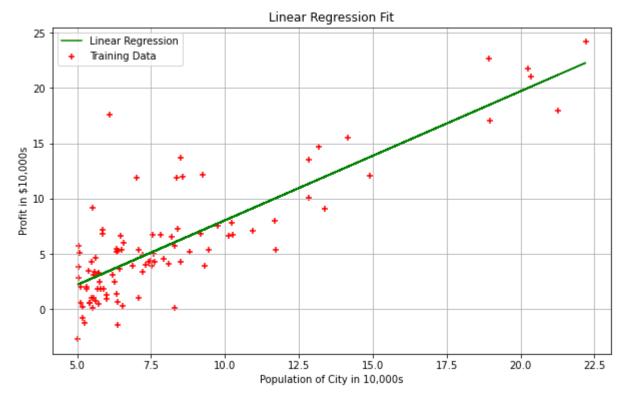
```
Out[12]: array([[1. , 6.1101], [1. , 5.5277],
```

```
[1.
                       , 8.5186],
                 [1.
                       , 7.0032],
                 [1.
                        , 5.8598]])
In [13]:
          theta = np.zeros(2)
          theta
Out[13]: array([0., 0.])
In [14]:
          def compute_cost(X, y, theta):
            Compute cost 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(1 x m)
            theta: 1D array of fitting parameters or weights. Dimension (1 x n)
            Output Parameters
            J : Scalar value.
            predictions = X.dot(theta)
            errors = np.subtract(predictions, y)
            sqrErrors = np.square(errors)
            J = 1 / (2 * m) * np.sum(sqrErrors)
            return J
In [15]:
          # Lets compute the cost for theta values
          cost = compute_cost(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 = 32.072733877455676
In [16]:
          def gradient_descent(X, y, theta, alpha, iterations):
            Compute cost 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 x 1)
            theta: 1D array of fitting parameters or weights. Dimension (1 \times n)
            alpha: Learning rate. Scalar value
            iterations: No of iterations. Scalar value.
            Output Parameters
```

theta: Final Value. 1D array of fitting parameters or weights. Dimension  $(1 \times n)$  cost\_history: Conatins value of cost for each iteration. 1D array. Dimansion $(m \times 1)$ 

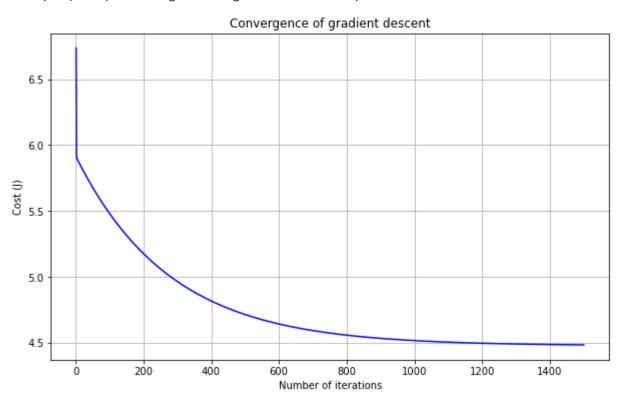
```
cost history = np.zeros(iterations)
            for i in range(iterations):
              predictions = X.dot(theta)
              errors = np.subtract(predictions, y)
              sum delta = (alpha / m) * X.transpose().dot(errors);
              theta = theta - sum delta;
              cost_history[i] = compute_cost(X, y, theta)
            return theta, cost history
In [17]:
          theta = [0., 0.]
          iterations = 1500;
          alpha = 0.01;
In [18]:
          theta, cost_history = gradient_descent(X, y, theta, alpha, iterations)
          print('Final value of theta =', theta)
          print('cost_history =', cost_history)
         Final value of theta = [-3.63029144 \ 1.16636235]
         cost history = [6.73719046 5.93159357 5.90115471 ... 4.48343473 4.48341145 4.48338826]
In [19]:
          # 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[19]: <matplotlib.legend.Legend at 0x28719cd9fa0>



```
In [20]:
    plt.plot(range(1, iterations + 1),cost_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[20]: Text(0.5, 1.0, 'Convergence of gradient descent')



In [ ]: