```
In [1]: import warnings
        warnings.filterwarnings("ignore")
        from sklearn.datasets import load_boston
         from random import seed
        from random import randrange
         from csv import reader
         from math import sqrt
         from sklearn import preprocessing
         import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         from prettytable import PrettyTable
         from sklearn.linear model import SGDRegressor
         from sklearn import preprocessing
         from sklearn.metrics import mean squared error
In [2]: X = load boston().data
        Y = load boston().target
         df=pd.DataFrame(X)
         #some intuition
         df[13]=df[10]//df[12] #here we set a column 13 such that df[13]=Boston
         data['Medv']//Boston data['B']
        X=df.as matrix()
        df.head()
Out[2]:
                0
                                                            9 10
                                                                          12 13
         0 0.00632 18.0 2.31 0.0 0.538 6.575 65.2 4.0900 1.0 296.0 15.3 396.90 4.98 3.0
         1 0.02731
                   0.0 7.07 0.0 0.469 6.421 78.9 4.9671 2.0 242.0 17.8 396.90 9.14 1.0
         2 0.02729
                   0.0 7.07 0.0 0.469 7.185
                                          61.1 4.9671 2.0 242.0
                                                             17.8
         3 0.03237
                   0.0 2.18 0.0 0.458 6.998 45.8 6.0622 3.0 222.0
                                                             18.7 394.63 2.94 6.0
         4 0.06905
                   0.0 2.18 0.0 0.458 7.147 54.2 6.0622 3.0 222.0 18.7 396.90 5.33 3.0
```

```
In [3]: #Splitting whole data into train and test
         from sklearn.model selection import train test split
         X train, X test, y train, y test=train test split(X, Y, test size=0.3,
         random state=4)
         # applying column standardization on train and test data
         scaler = preprocessing.StandardScaler()
         X train = scaler.fit transform(X train)
         X test=scaler.transform(X test)
         df train=pd.DataFrame(X train)
         df train['price']=y train
         df train.head()
Out[3]:
                           1
                                                                               7
          0 -0.425469 -0.470768 -0.954686 -0.231455 -0.919581 0.215100 -0.747410 0.454022 -0.76446
          1 -0.426323 2.992576 -1.330157 -0.231455 -1.227311 -0.883652 -1.691588 3.163428 -0.65156
          2 -0.385190 -0.470768 -0.705828 4.320494 -0.423795 -0.125423
                                                                0.818985 -0.353904 -0.19996
          3 -0.249268 -0.470768 -0.423497 -0.231455 -0.158805 -0.228336
                                                                1.021567 -0.021755 -0.65156
          4 -0.365945 0.395068 -1.030363 -0.231455 0.157472 3.102729 -0.060078 -0.646202 -0.53866
In [4]: #SGD implementation for linear regression
         #function having parameter X train, y train, no of iteration, learning rat
         #intialising no of iteration=100, learning rate =0.01
         #batch size=20
         W,B,iteration,lr rate,k=np.zeros(shape=(1,14)),0,750,0.01,25 #intialise
          W and B to zero
         while iteration>=0:
```

```
w,b,temp vectors,temp intercept=W,B,np.zeros(shape=(1,14)),0
            data=df train.sample(25) #sampling random k=batch size=20 data
            x=np.array(data.drop('price',axis=1))
            v=np.array(data['price'])
            for i in range(k):
                temp\_vectors+=(-2)*x[i]*(y[i]-(np.dot(w,x[i])+b))#partial diffe
        rentiation wrt w dl/dw=1/k(-2x)*(y-wTx-b)
                temp intercept+=(-2)*(y[i]-(np.dot(w,x[i])+b))#partial differen
        tiation wrt b dl/db=1/k(-2)*(y-wTx-b)
            W=(w-lr rate*(temp vectors)/k)
            B=(b-lr rate*(temp intercept)/k)
            iteration-=1
        print(W)
        print(B)
        [[-1.27290222  0.53695967  -0.66141371  0.92096371  -1.4356316
                                                                        1.636325
        89
           0.26658427 -2.81929856 2.14914024 -0.92267911 -2.14815104 0.999576
        79
          -1.97622219 2.66317572]]
        [22.27101495]
In [5]: #prediction on x test
        #https://www.geeksforgeeks.org/numpy-asscalar-in-python/
        y predic lr=[]
        for i in range(len(X test)):
            val=np.dot(W,X test[i])+B \#val= wTx+b
            y predic lr.append(np.asscalar(val))
In [6]: #Scatter plot of actual price vs predicted price
        plt.scatter(y test,y predic lr)
        plt.xlabel('Actual price')
        plt.ylabel('Predictd price')
```

```
plt.title('Actual price vs Predicted price')
         plt.show()
                        Actual price vs Predicted price
            40
            30
          Predictd price
            20
            10
                   10
                            20
                                              40
                                Actual price
 In [7]: MSE_lr=mean_squared_error(y_test,y_predic_lr)
         print('mean squared error =',MSE lr)
         mean squared error = 24.258404283487806
In [10]: #SGD regression sklearn implementation
         #intialising no of iteration=100,eta0=1
         #taking t=2 and power t=1 such that for each iteration eta0=eta0/pow(2,
         1) ,it means half each times
         model=SGDRegressor(learning rate='constant',eta0=0.01,penalty=None,max
         iter=100)
         model.fit(X train,y train)
         y pred sgd=model.predict(X test)
         #Scatter plot of actual price vs predicted price
```

plt.scatter(y_test,y_pred_sgd)

```
plt.xlabel('Actual price')
         plt.ylabel('Predictd price')
         plt.title('Actual price vs Predicted price')
          plt.show()
                        Actual price vs Predicted price
            40
          Predictd price
            20
            10
                            20
                   10
                                     30
                                              40
                                                       50
                                Actual price
In [11]: MSE sgd=mean squared error(y test,y pred sgd)
          print('mean squared error =',MSE sgd)
         mean squared error = 24.834732966390735
In [12]: #Comparison between weights obtained from own implementation and weight
         s obtained from sqd implementation
         from prettytable import PrettyTable
         x = PrettyTable()
         x.field names=['Weight vector manual','Weight vector SGD sklearn']
         weight sqd=model.coef
         for i in range(13):
              x.add row([W[0][i],weight sgd[i]])
         print(x)
```

| Weight vector manual | Weight vector SGD sklearn |

```
-1.2729022153064808
                                      -1.7383713107960062
            0.5369596698100564
                                      0.9455387554214926
                                      -0.41331803536355016
           -0.6614137122022188
            0.9209637069264854
                                      0.9017778913579981
           -1.4356315991605868
                                      -1.6015520405581498
            1.6363258899143136
                                       1.048928088760244
            0.2665842712625248
                                      0.08704812372570098
            -2.819298559451317
                                      -3.0231576999056533
            2.1491402371476007
                                        2.61590860443553
           -0.9226791124626965
                                      -1.6275359256507602
            -2.148151040199819
                                      -2.1822133327621853
            0.9995767929920171
                                      0.7908475151392433
           -1.9762221899180858
                                      -1.8902500986675943
In [13]: #comparison between MSE of own implementation and SGD sklearn implement
         ation
         print('MSE of manual implementation = ',MSE lr)
         print('-'*50)
         print('MSE of SGD sklearn implementation = ',MSE sgd)
         MSE of manual implementation = 24.258404283487806
         MSE of SGD sklearn implementation = 24.834732966390735
In [ ]:
```