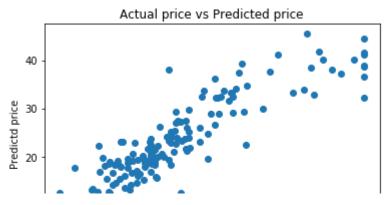
```
import warnings
warnings.filterwarnings("ignore")
from sklearn.datasets import load boston
from random import seed
from random import randrange
from csv import reader
from math import sqrt
from sklearn import preprocessing
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
from prettytable import PrettyTable
from sklearn.linear_model import SGDRegressor
from sklearn import preprocessing
from sklearn.metrics import mean squared error
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()
 С→
                0
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                                                                   8
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                                                                               10
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                                                                                              12
                                                                                                   13
                         2.31
                                    0.538
                                           6.575
                                                   65.2
                                                         4.0900
                                                                      296.0
         0.00632
                   18.0
                               0.0
                                                                 1.0
                                                                              15.3
                                                                                    396.90
                                                                                            4.98
                                                                                                  3.0
         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
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                                                                      242.0
                         7.07
                                    0.469
                                           7.185
                                                   61.1
                                                         4.9671
                                                                 2.0
                                                                             17.8
      2
         0.02729
                    0.0
                                                                                    392.83
                                                                                            4.03
                                                                                                  4.0
                         2.18
         0.03237
                    0.0
                               0.0
                                    0.458
                                           6.998
                                                   45.8
                                                         6.0622
                                                                 3.0
                                                                      222.0
                                                                              18.7
                                                                                    394.63
                                                                                            2.94
                                                                                                  6.0
         0.06905
                         2.18
                               0.0
                                    0.458
                                           7.147
                                                   54.2
                                                        6.0622
                                                                 3.0
                                                                      222.0
                                                                                    396.90
                                                                                            5.33
                    0.0
                                                                             18.7
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#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)
data train=pd.DataFrame(X train)
data train['price']=y train
data train.head()
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7
                 0
                            1
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                                                             4
                                                                        5
                                                                                   6
         -0.425469
                    -0.470768
                               -0.954686
                                          -0.231455
                                                     -0.919581
                                                                0.215100
                                                                           -0.747410
                                                                                      0.454022
                                                                                                 -0.7
         -0.426323
                    2.992576 -1.330157 -0.231455
                                                    -1.227311
                                                                -0.883652
                                                                          -1.691588
                                                                                      3.163428
                                                                                                 -0.6
#SGD implementation for linear regression
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=data_train.sample(25) #sampling random k=batch size=20 data
    x=np.array(data.drop('price',axis=1))
    y=np.array(data['price'])
    for i in range(k):
        temp_vectors+=(-2)*x[i]*(y[i]-(np.dot(w,x[i])+b))#partial differentiation wrt w dl/dw=1/k(-2)
        temp intercept+=(-2)*(y[i]-(np.dot(w,x[i])+b))#partial differentiation wrt b dl/db=1/k(-2)*(
    W=(w-lr_rate*(temp_vectors)/k)
    B=(b-lr_rate*(temp_intercept)/k)
    iteration-=1
print(W)
print(B)
     [[-1.07154866 0.6824004 -0.73812727 0.94985895 -1.53072618
                                                                         1.83740877
        0.12717258 -3.04013226 1.95119164 -1.05883575 -2.31322222
                                                                         0.84870653
       -1.83316004 2.73721903]]
     [22.13209163]
#prediction on x test
y predic linear regression = [ ]
for i in range(len(X test)):
    val=np.dot(W,X test[i])+B #val= wTx+b
    y predic linear regression.append(np.asscalar(val))
#Scatter plot of actual price vs predicted price
plt.scatter(y_test, y_predic_linear_regression)
plt.xlabel('Actual price')
plt.ylabel('Predictd price')
plt.title('Actual price vs Predicted price')
plt.show()
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```



MSE_linear_regression = mean_squared_error(y_test,y_predic_linear_regression)
print('mean squared error =',MSE_linear_regression)

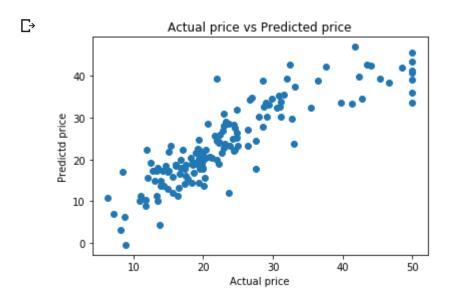
mean squared error = 23.05958235093564

```
#SGD regression sklearn implementation
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```
model = SGDRegressor(learning_rate ='constant',eta0 = 0.01, penalty = None, n_iter_no_change = 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()
```



MSE_sgd=mean_squared_error(y_test,y_pred_sgd)
print('mean squared error =',MSE_sgd)

mean squared error = 25.215752448802053

#comparison between MSE of own implementation and SGD sklearn implementation
print('MSE of manual implementation = ',MSE_linear_regression)
print('='*50)

print('MSE of SGD sklearn implementation = ',MSE_sgd)