### **Objective- Implement SGD to Linear Regression**

### **Importing libraries**

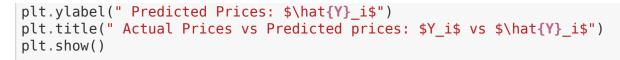
```
In [1]: import warnings
        warnings.filterwarnings("ignore")
        from sklearn.datasets import load boston
        from random import seed
        from random import randrange
        from csv import reader
        from math import sqrt
        from sklearn import preprocessing
        import 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]: # Loading of Boston dataset
        from sklearn.datasets import load boston
        boston=load boston()
In [3]: # Shape of data
        print(boston.data.shape)
        (506, 13)
In [4]: # Number of rows
        print(boston.data.shape[0])
```

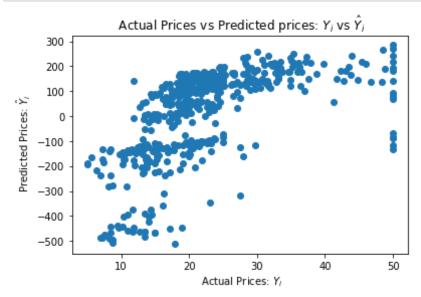
```
In [5]: # Number of columns
         print(boston.data.shape[1])
         13
 In [6]: # Columns Name of Boston dataset
         print(boston.feature names)
         ['CRIM' 'ZN' 'INDUS' 'CHAS' 'NOX' 'RM' 'AGE' 'DIS' 'RAD' 'TAX' 'PTRATI
         0'
          'B' 'LSTAT']
         Importing Data
 In [7]: # Dividing dataset into data(input variables) and target(output variable)
         X = pd.DataFrame(boston.data)
         Y = pd.DataFrame(boston.target)
 In [8]: # Preprocessing of data
         scaler = preprocessing.StandardScaler()
         scaler.fit(X)
         X data=pd.DataFrame(scaler.transform(X))
 In [9]: #Concatenating Input and output variables into a single DataFrame
         data = pd.concat([X data,Y], axis=1,ignore index=True)
In [10]: #Printing Top 5 rows
         data.head(5)
Out[10]:
                   0
                                             3
                                                              5
            -0.417713 0.284830
                             |-1.287909|-0.272599|-0.144217|0.413672|-0.120013|0.140214|-0
```

	0	1	2	3	4	5	6	7	
1	-0.415269	-0.487722	-0.593381	-0.272599	-0.740262	0.194274	0.367166	0.557160	-0
2	-0.415272	-0.487722	-0.593381	-0.272599	-0.740262	1.282714	-0.265812	0.557160	-0
3	-0.414680	-0.487722	-1.306878	-0.272599	-0.835284	1.016303	-0.809889	1.077737	-0
4	-0.410409	-0.487722	-1.306878	-0.272599	-0.835284	1.228577	-0.511180	1.077737	-0

## **SGD** regressor of Sklearn

```
In [11]: import warnings
         warnings.filterwarnings("ignore")
         #Applying SGDRegressor() and fitting the model on Data
         clf = SGDRegressor()
         clf.fit(X data, Y)
         #Printing the mean square error (Actual Y- Predicted Y)
         print(mean squared error(Y, clf.predict(X data)))
         22.782371903483956
In [12]: # Finding the intercept value
         clf.intercept
Out[12]: array([22.35309641])
In [13]: # individual weights and storing as list in Dataframe
         SGD wght=pd.DataFrame(list(clf.coef ))
In [14]: #Plotting the Scatter plot of Predicted vs Actual price of Boston datas
         plt.scatter(Y, clf.predict(X))
         plt.xlabel(" Actual Prices: $Y i$")
```





## **Manual SGD implementation**

In [27]: #Taking random weight vector 'w' into Dataframe
w=pd.DataFrame(X\_data.loc[0,:])

w=w.T #Changing into row vector as default type is column vector
w #Printing the weight vector 'w'

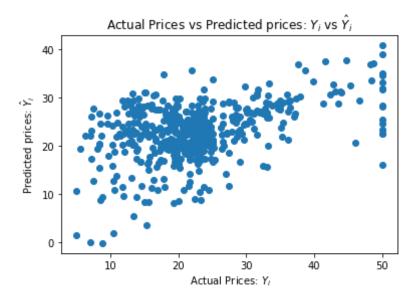
Out[27]:

		0	1	2	3	4	5	6	7	
(	)	-0.417713	0.28483	-1.287909	-0.272599	-0.144217	0.413672	-0.120013	0.140214	-0.98

In [28]: b=0 #Choosing Default value of intercept(scalar)

```
r=1 #Learning rate(r)
In [29]: # Reference: https://stackoverflow.com/questions/50328545/stochastic-gr
         adient-descent-for-linear-regression-on-partial-derivatives
         while(True):
             random data=data.sample(100) #Taking the random data of 100 sample
         S
             random data=random data.reset index(drop=True)
             X=random data.loc[:,0:12] #Taking all the input variables needed t
         o predict the class label
             Y=random data.loc[:,13] #Taking the output variable(class label)
             w optimized=0.0
             b optimized=0.0
             for k in range(100):
                 \# -2x * (v-(mx +b))
                 w_optimized= w_optimized+(-2*X.loc[k]*(Y.loc[k]-np.dot(X.loc[k
         1,w.T)-b))
                  \# -2(v - (mx + b))
                 b optimized=b optimized+(float(-2*(Y.loc[k]-np.dot(X.loc[k],w.T
         )-b)))
             #updating the parameters
             w0=w-r^*(w \text{ optimized}/100)
             b0=b-r*(b optimized/100)
             r=r/2 #reducing the learning rate to half in each iteration
             #Comparing whether the obtained w is same as previous w
             if(np.array(w) == np.array(w0)).all():
                 break:
             else:
                 w=w0 #updating the value of w in each iteration
```

```
b=b0 #updating the value of b in each iteration
In [30]: # Obtained value of weight vector
Out[30]:
                                              3
                                                               5
          0 -3.239091 | 0.973096 | -0.278184 | -2.432222 | -0.827658 | 4.525984 | 2.053458 | -0.699801 | 1.4
In [311: b0
Out[31]: 22.87076428299188
In [32]: y pred=[]
         X=data.loc[:,0:12]
         v=data.loc[:,13]
         for i in range (506):
              # Determining the hyperplane using the formula, y=mx+c
              # where m=w0.T(vector), x=X[i] i=0,1,2...506 , c=b0(scalar)
              y pred.append(np.dot(X.loc[i],w0.T)+b0)
         y pred=np.asarray(y pred)
In [33]: #Plotting the Scatter plot of Predicted vs Actual price of Boston datas
         import matplotlib.pyplot as plt
         plt.scatter(y, y pred)
         plt.xlabel(" Actual Prices: $Y i$")
         plt.ylabel("Predicted prices: $\hat{Y} i$")
         plt.title("Actual Prices vs Predicted prices: $Y i$ vs $\hat{Y} i$")
         plt.show()
```



```
In [36]: #Printing the mean square error (Actual Y- Predicted Y)
    print(mean_squared_error(y,y_pred))
```

24.14143389935003

# **Summary**

- The mean square error obtained from Manual SGD is almost similar to the mean square error obtained from the Sklearn SGD with small difference.
- The intercept term is also similar in both the cases.