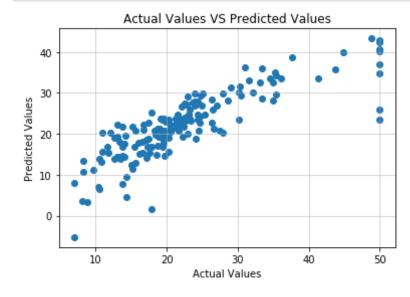
## **Assignment:Implementing SGDClassifiers**

```
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 sklearn.linear model import SGDRegressor
        from sklearn import preprocessing
        from sklearn.metrics import mean squared error
In [2]: import warnings
        warnings.filterwarnings('ignore')
In [3]: X = load boston().data
        Y = load boston().target
In [4]: from sklearn.model selection import train test split
        X tra, X tes, Y tra, Y tes = train test split(X, Y, test size = 0.33, r)
        andom state = 5)
        print(X tra.shape)
        print(X tes.shape)
        print(Y tra.shape)
        print(Y tes.shape)
        (339, 13)
        (167, 13)
```

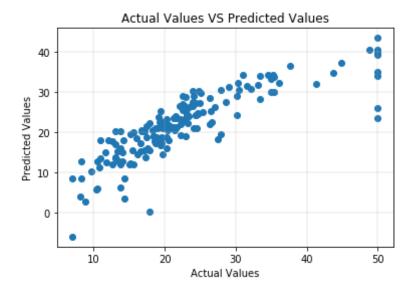
```
(339,)
        (167,)
In [5]: scaler= preprocessing.StandardScaler().fit(X tra)
        X train=scaler.transform(X tra)
        X test=scaler.transform(X tes)
In [7]: custom=pd.DataFrame(data=X train)
        custom['price']=Y tra
In [8]: def sgd(alpha, lrr var, eta=0.01, pow t=0.25,n iter=1):
            clf=SGDRegressor(alpha=alpha, learning rate=lrr var, eta0=eta, pow
        er t=pow t, n iter=n iter)
            clf.fit(X train,Y tra)
            y pred=clf.predict(X test)
            #scatter plot
            plt.scatter(Y tes,y pred)
            plt.title('Actual Values VS Predicted Values')
            plt.xlabel('Actual Values')
            plt.ylabel('Predicted Values')
            plt.grid(b=True, linewidth=0.5)
            plt.show()
            sgd error=mean squared error(Y tes,y pred)
            print('mean square error=', sqd error)
            return clf.coef , clf.intercept , sgd error
In [9]: def custom model(X, lr rate variation, alpha=0.0001, lr rate=0.01, powe
        r t=0.25, n iter=1):
            w new=np.zeros(shape=(1,13))
            b new=0
            t=1
            r=lr rate
            while(t<=n iter):</pre>
               w old=w new
```

```
b old=b new
                  w = np.zeros(shape=(1,13))
                  b = 0
                  x data=X.sample(10)
                  x=np.array(x_data.drop('price',axis=1))
                  y=np.array(x data['price'])
                  for i in range(10): # for getting the derivatives using sgd wi
          th k=10
                      y curr=np.dot(w old,x[i])+b old
                      W_+=x[i] * (y[i] - y_curr)
                      b += (y[i] - y curr)
                  w *= (-2/x.shape[0])
                  b^{*}=(-2/x.shape[0])
                  #updating the parameters
                  w \text{ new} = (w \text{ old} - r * w)
                  b new=(b old-r*b)
                  if(lr_rate_variation=='invscaling'):
                      r = lr rate / pow(t, power t)
                  t += 1
              return w new, b new
In [10]: def predict(x,w, b):
             y pred=[]
              for i in range(len(x)):
                  y=np.asscalar(np.dot(w,x[i])+b)
                  y pred.append(y)
              return np.array(y pred)
In [11]: def ploting(X_test,y_pred):
              #scatter plot
```

## **Skearn SGDRegressor**



## **Custom implementation of SGDRegressor**



```
In [21]: b=w.reshape(13,1)
```

```
In [22]: from prettytable import PrettyTable

numbering = [1,2,3,4,5,6,7,8,9,10,11,12,13]
# Initializing prettytable
ptable = PrettyTable()

# Adding columns
ptable.add_column("S.NO.",numbering)
ptable.add_column("Weights of Sklearn's SGD",w_sgd)
ptable.add_column("Weights of manual SGD",b)
```

```
# Printing the Table
print(ptable)
```

```
S.NO. | Weights of Sklearn's SGD | Weights of manual SGD
         -1.3812233430359353
                                       [-1.18969291]
  2
          0.6342440751923348
                                       [0.76183188]
  3
                                       [-0.527319]
          -0.11820806673036802
                                       [0.69989891]
           0.4512447404217142
          -1.354866106534257
                                       [-1.41940834]
           3.1880799640744666
                                       [2.47851379]
                                       [-0.30013524]
         -0.016879860138361562
  8
                                       [-2.78326128]
          -3.0270354416702983
           3.22408253504173
                                       [2.20547065]
  10
          -2.17961223091745
                                       [-1.07909456]
  11
          -2.156591952719235
                                      [-2.12206703]
  12
                                      [1.10438779]
          0.9947252014489741
  13
                                       [-3.61071477]
           -3.219698660547973
```

## **Conclusion**

Out[23]:

	Vectorizer	Mean_Sqaure_Error(MSE)
0	SGD Regressor	27
1	Custom SGD Regressor	27