

# AMAZON SGDRegressor

January 23, 2019

```
In [1]: from sklearn.datasets import load_boston
        boston = load_boston()
```

```
In [2]: print(boston.data.shape)

(506, 13)
```

```
In [3]: print(boston.feature_names)
```

```
['CRIM' 'ZN' 'INDUS' 'CHAS' 'NOX' 'RM' 'AGE' 'DIS' 'RAD' 'TAX' 'PTRATIO'
 'B' 'LSTAT']
```

```
In [4]: #print(boston.target)
```

```
In [5]: #print(boston.DESCR)
```

```
In [6]: import pandas as pd
        bos = pd.DataFrame(boston.data)
        print(bos.head())
```

	0	1	2	3	4	5	6	7	8	9	10	\
0	0.00632	18.0	2.31	0.0	0.538	6.575	65.2	4.0900	1.0	296.0	15.3	
1	0.02731	0.0	7.07	0.0	0.469	6.421	78.9	4.9671	2.0	242.0	17.8	
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	
4	0.06905	0.0	2.18	0.0	0.458	7.147	54.2	6.0622	3.0	222.0	18.7	
		11	12									
0	396.90	4.98										
1	396.90	9.14										
2	392.83	4.03										
3	394.63	2.94										
4	396.90	5.33										

```
In [7]: bos['PRICE'] = boston.target
```

```
X = bos.drop('PRICE', axis = 1)
Y = bos['PRICE']
```

```
In [8]: X.duplicated().sum()
```

```
Out[8]: 0
```

```
In [10]: import numpy as np
         np.any(np.isnan(X))
```

```
Out[10]: False
```

## 1 Splitting and standardizing data

```
In [11]: from sklearn.model_selection import train_test_split
         X_train, X_test, Y_train, Y_test = train_test_split(X, Y, test_size = 0.33, random_state=42)
         print(X_train.shape)
         print(X_test.shape)
         print(Y_train.shape)
         print(Y_test.shape)
```

```
(339, 13)
```

```
(167, 13)
```

```
(339,)
```

```
(167,)
```

```
In [12]: from sklearn.preprocessing import StandardScaler
         standardization= StandardScaler()
         X_train = standardization.fit_transform(X_train)
```

## 2 Gradient Descent code

In the sixth line i concatenated ones of size equal to number of rows in train data,  $W = b_0 + w_1x_1 + w_2x_2 + w_3x_3 \dots$  is same as  $W = b_0x_0 + w_1x_1 + w_2x_2 + w_3x_3 \dots$  (where  $x_0=1$ ) so i added a feature at the end of train data with all ones which is  $x_0$ . and  $W[-1:0]$  (which is last term of weight vector is my  $b_0$ )

```
In [27]: import numpy as np
         def GD(X_train,Y_train,alpha=0.001,num_of_iter=100,fit_intercept=False,penalty = 'l2',
               lamb=0.05,change_alpha=False,tol=0.001):
             num_of_col = X_train.shape[1]
             num_of_rows = X_train.shape[0]
             if (fit_intercept==True):
                 W = np.random.normal(loc=0.1,scale=0.001,size=num_of_col+1)
                 x = np.ones((num_of_rows,1))
                 X_train = np.concatenate((X_train,x),axis=1)
             else:
                 W = np.random.normal(size=X.shape[1])

             #W = W/(sum(W**2)**0.5)
```

```

W_hat = W
W_new = W
for i in range(1,num_of_iter,1):
    W_old = W_new
    #print (W_old)
    error = Y_train - np.dot(W_old,X_train.T)
    gradient = -(2/num_of_rows)*np.dot(X_train.T,error)
    regularization = 0
    if (penalty == 'l2'):
        regularization = 2*W_old
    elif (penalty == 'l1') :
        regularization = 1
    W_new = W_old - alpha*(gradient+lamb * regularization)
    #print (W_new)
    W_hat = W_new
    if (change_alpha==True):
        alpha=alpha/2
    k = W_new - W_old
    k =abs (k)
    #print (k)
    if (sum(k)/len(k)==tol):
        #print (alpha)
        #print (i)
        W_hat = W_old
        break

return (W_hat)

```

In [28]: `W_hat = GD(X_train,Y_train,penalty='l2',alpha = 0.05,num_of_iter=5000,lamb=0.01,fit_i`

In [29]: `W_hat`

Out[29]: `array([-1.26681059, 0.807998 , -0.25736564, 0.20527258, -1.38856047,
 2.8148151 , -0.34586958, -2.67124956, 2.578236 , -1.91130961,
 -2.10081408, 1.0380248 , -3.28815899, 22.31402786])`

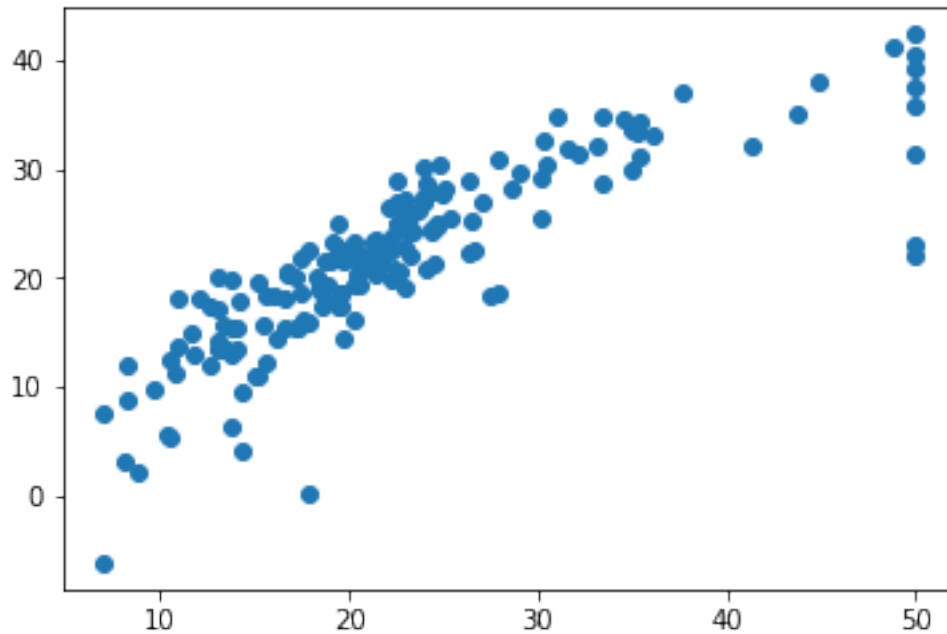
In [30]: `col = X_train.shape[1]`  
`predictions = np.dot(standardization.transform(X_test),W_hat[0:-1])+W_hat[-1:]`

In [31]: `predictions.shape`

Out[31]: `(167,)`

In [32]: `error = predictions - Y_test`

In [33]: `import matplotlib.pyplot as plt`  
`plt.scatter(Y_test,predictions)`  
`plt.show()`



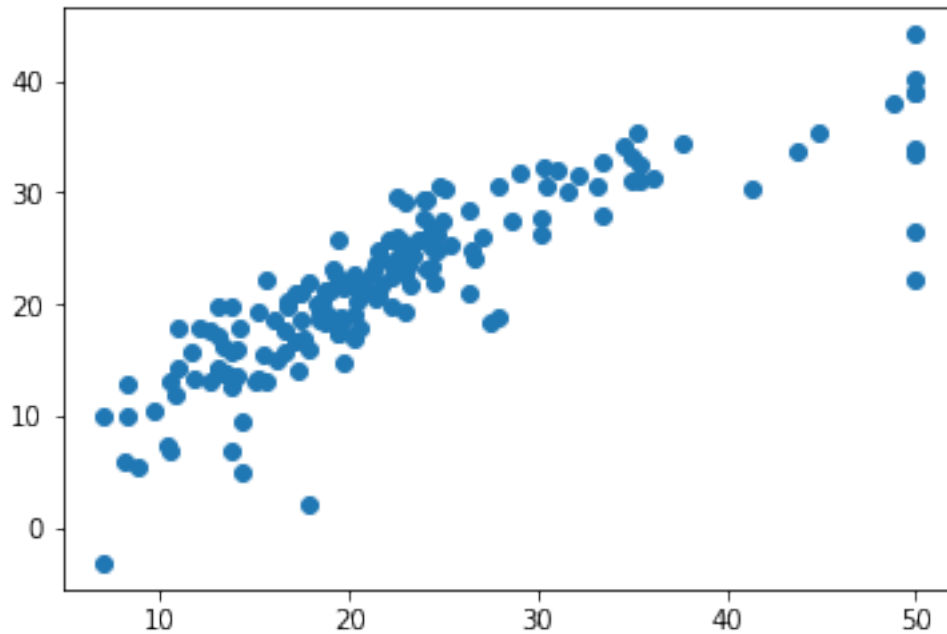
### 3 SGDRegressor

```
In [34]: from sklearn.linear_model import SGDRegressor
         clf=SGDRegressor(alpha=0.05,max_iter=5000,penalty='l2',learning_rate='constant',tol=0
         clf.fit(X_train,Y_train)
```

```
Out[34]: SGDRegressor(alpha=0.05, average=False, epsilon=0.1, eta0=0.01,
                        fit_intercept=True, l1_ratio=0.15, learning_rate='constant',
                        loss='squared_loss', max_iter=5000, n_iter=None, penalty='l2',
                        power_t=0.25, random_state=None, shuffle=True, tol=0.001, verbose=0,
                        warm_start=False)
```

```
In [35]: predictions1 = clf.predict(standardization.transform(X_test))
```

```
In [36]: plt.scatter(Y_test,predictions1)
         plt.show()
```



## 4 Weights Comparison

```
In [37]: print ('\tgradient descent\t\t\tSGD regressor')
print ('-'*100)
for i in range(13):
    print ('\t{0}\t\t\t\t\t{1}'.format(W_hat[i:i+1],clf.coef_[i:i+1]))
print ('\tintercept')
print ('\t{0}\t\t\t\t\t{1}'.format(W_hat[13:14],clf.intercept_))
```

gradient descent	SGD regressor
[-1.26681059]	[-0.95598885]
[0.807998]	[0.91266346]
[-0.25736564]	[-0.37134864]
[0.20527258]	[1.05669396]
[-1.38856047]	[-1.15997815]
[2.8148151]	[2.37457691]
[-0.34586958]	[-0.50317638]
[-2.67124956]	[-2.09729828]
[2.578236]	[1.59884316]
[-1.91130961]	[-0.75138357]
[-2.10081408]	[-1.71556694]
[1.0380248]	[0.83954643]
[-3.28815899]	[-3.23034314]
intercept	

[22.31402786]

[22.34519262]

## 5 Mean Square Error

```
In [38]: #import sklearn.metrics.mean_squared_error as MSE
         from sklearn.metrics import mean_squared_error
         mse = mean_squared_error(Y_test,predictions)
         print ('mean squared error of gradient descent = {}'.format(mse))
```

mean squared error of gradient descent = 28.890517247630147

```
In [39]: #import sklearn.metrics.mean_squared_error as MSE
         from sklearn.metrics import mean_squared_error
         mse = mean_squared_error(Y_test,predictions1)
         print ('mean squared error of SGD regressor = {}'.format(mse))
```

mean squared error of SGD regressor = 27.395627127847074