# 29.13.implement\_SGD\_1.0

August 1, 2018

### 1 Implement SGD on bston dataset

Dataset from sklearn load\_boston

The GD is theta=theta-alpha*derivative*( $cost\ function$ ) equ=y=wtx+b;  $cost=(y-wtx-b)^2$   $determinant\ wrt\ w=>-2(y-wtx-b)x\ determinant\ wrt\ b=>-2(y-wtx-b)$ ; in general: loss These are in vec notation

h=np.dot(x,theta); loss=h-y; cost=np.sum(loss\*\*2)/(2m); gradient=np.dot(xt,loss)/m; theta=theta-alphagradient

## 2 Objective

Implement the model and compare the intercept and weights with sklearn

#### 3 Import data and libraries and split train and test by 80:20

```
In [139]: from sklearn.datasets import load_boston
          import numpy as np
          import matplotlib.pyplot as plt
          boston = load_boston()
          print(boston.data.shape)
          boston.data[0:1]
          y=np.array(boston.target)
          x=np.array(boston.data)
          from sklearn.preprocessing import normalize, StandardScaler
          x = normalize(x, norm='11', axis=0)
          #y = normalize(y.reshape(-1, 1), norm='l1', axis=0)
          #y=y.ravel()
          #print("shape of y",y.shape)
          x_train=x[0:450]
          x_test=x[451:505]
          y_train=y[0:450]
          y_{test=y}[451:505]
(506, 13)
```

#### 4 Create my model and cost function to calculate cost

```
In [140]: # Create all functions
          \#m=no \ of \ obs, n=no \ of \ features
          def grad(x,y,theta,alpha=.0001,iteration=1000,k=200):
            #for SGD
            m=x.shape[0]
            idx=np.random.randint(m,size=k)
            x=x[idx,:]
            y=np.random.choice(y,k)
            #print('for SGD new size',x.shape,y.shape)
            m, n=np.shape(x)
            xt=x.transpose()
            #print("x y, theta , m,n,shape before multiply",x.shape,y.shape,theta.shape,m,n)
            oldcost=0
            bold=0
            1=0
            thetaprev=theta
            prevalpha=alpha
            optimumcost=0
            ind=0
            for i in range(0,iteration):
               1=1+1
               h=np.dot(x,theta)
               #print("\nbefore loss h and y",h.shape,y.shape)
               loss=h-y
               #print('loss',loss)
               cost=np.sum(loss**2)/(2*k)
               #if (oldcost^=0 & cost-oldcost<20):
                       break:
               #if change is very slow increase alpha
               b=(cost-oldcost)/oldcost
               #print("At iteration %d Cost : %f oldcost : %f alpha %f cost compare %f" %(i,co
               #if ((b>0 & bold<0) | (b<0 & bold>0)):
               if((cost>oldcost) & (oldcost>0) & (ind==0)):
                   #set optimumcost only one time
                      ind=1
                      optimumcost=oldcost
                      alpha=alpha/2
                      print("opt cost,ind",optimumcost,ind)
                      theta=thetaprev
                      print('new alpha',alpha,'cost',cost,'old cost',oldcost)
                      continue
               elif ((b>-.05) & (ind==0)):
```

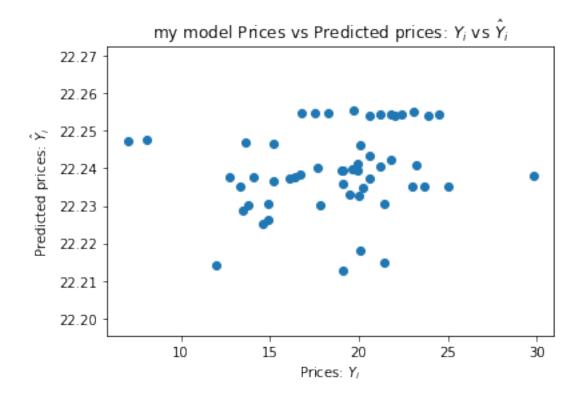
```
alpha=alpha*2
     #elif (ind==1):
             alpha=alpha*1.15
             ind=2
     gradient=np.dot(xt,loss)/k
     #print("\ngradient\n", gradient)
     theta=theta-alpha*gradient
     aa=alpha*gradient
     b=(cost-oldcost)/oldcost
     oldcost=cost
     thetaprev=theta
     prevalpha=alpha
  return theta, cost
def data(x,y,k=10):
  x=np.append(x, np.ones([x.shape[0], 1]), axis=1) #with intercept
  #without intercept
  n1=np.shape(x)[1]
  theta=np.random.randn(n1)
  print("theta passed shaped",theta.shape,theta)
  newtheta,cost=grad(x,y,theta)
  print('coefficient of mymodel',newtheta)
  return newtheta, cost
def predictcost(x,newtheta):
    x=np.append(x, np.ones([x.shape[0], 1]), axis=1) #with intercept
    #without intercept
    y=np.dot(x,newtheta)
    return y
def cost1(y_pred,y_test):
    loss=-y_pred+y_test
    cost=np.sum(loss**2)/(2*y_test.shape[0])
    return cost
```

# 5 Use my model on boston data

```
In [141]: import warnings
     warnings.filterwarnings('ignore')

import pandas as pd
     #print("train test split", x_train.shape, x_test.shape, y_train.shape, y_test.shape)
     #scale=StandardScaler(with_mean=True)
     #x_train=scale.fit_transform(x_train)
     #x_test=scale.transform(x_test)
```

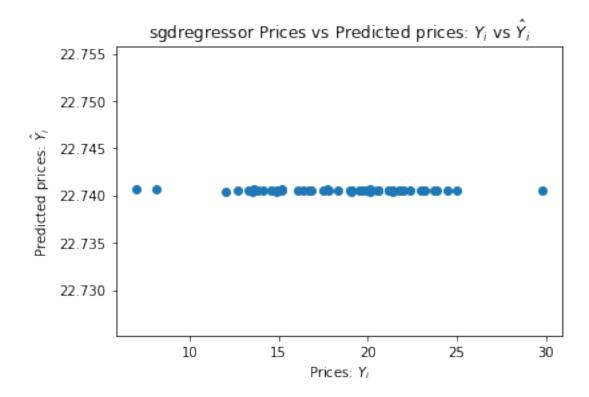
```
newtheta,cost=data(x_train,y_train)
         print("mymodel cost from the train",cost)
          #predict
         y_pred=predictcost(x_train,newtheta)
         cost=cost1(y_pred,y_train)
         y_pred=predictcost(x_test,newtheta)
         c=cost1(y_pred,y_test)
          #print(y_pred.shape,y_test.shape)
         print("my model cost from the test",c)
         aa=pd.DataFrame({'type':['mymodel'],'train_cost':[cost],'test_cost':[c],'Coeff':[new
          #print(aa)
         plt.scatter(y_test, y_pred)
         plt.xlabel("Prices: $Y_i$")
         plt.ylabel("Predicted prices: $\hat{Y}_i$")
         plt.title("my model Prices vs Predicted prices: $Y_i$ vs $\hat{Y}_i$")
         plt.show()
theta passed shaped (14,) [ 0.84181168  0.28353384 -1.65890218  0.67759257 -0.23250494  1.5515
  0.95602726 -0.60950194 1.19612826 0.36614641 0.18908912 0.16858242
  1.53765322 0.13158901]
opt cost, ind 36.6582077629 1
new alpha 1.6384 cost 36.7077999525 old cost 36.6582077629
coefficient of mymodel [ -4.30642126e+00
                                           4.82291220e+00 -2.55874326e+00 -1.06998034e+01
  -6.13320333e-01
                   1.74705492e+00 -1.84281057e-03 5.26956375e-01
 -5.46160077e-01 -2.90538283e-01
                                     2.86636557e-01 4.13640841e-01
  9.42275633e-01 2.22545481e+01]
mymodel cost from the train 36.529936717
my model cost from the test 15.2824107048
```



#### 6 Use SGDREGRESSOR model on boston data

```
In [142]: import warnings
          warnings.filterwarnings('ignore')
          # Use sklearn
          import pandas as pd
          from sklearn.linear_model import SGDRegressor
          x_train=pd.DataFrame(x_train)
          x_test=pd.DataFrame(x_test)
          y_train=pd.DataFrame(y_train)
          y_test=pd.DataFrame(y_test)
          #print(x_train)
          #print(y_train)
          print(y_train.shape,y_test.shape,x_train.shape,x_test.shape)
          lm = SGDRegressor(fit_intercept=True)
          lm.fit(x_train, y_train)
          #print(lm.coef_)
          print(lm)
          #y_test.reshape(-1,1)
          #np.reshape(y_test, 54)
```

```
#print(y_test.shape)
         y_pred = lm.predict(x_train)
         print('coefficient of sgdregressor model',lm.intercept_,lm.coef_)
         #c=cost1(y_pred,y_train)
         y_predtrain = lm.predict(x_train)
         print("sklearn Cost from train")
         print(lm.score(x_train,y_train))
         \#c1 = cost1(y\_pred, y\_test)
         y_predtest = lm.predict(x_test)
         print("shape",y_predtrain.shape,y_train.shape)
         print(type(cost1(y_predtrain.reshape(450,1),y_train)))
         print("train cost :",cost1(y_predtrain.reshape(450,1),y_train)[0])
         print("test cost",cost1(y_predtest.reshape(54,1),y_test)[0])
         #print(y_predtrain)
         #print(y_train)
         #print("sklearn Cost from test", lm.score(y_predtest, y_test))
         bb=pd.DataFrame({'type':['sgdregressor'], 'train_cost':[cost1(y_predtrain.reshape(450
         aa=aa.append(bb)
         plt.scatter(y_test, y_predtest)
         plt.xlabel("Prices: $Y_i$")
         plt.ylabel("Predicted prices: $\hat{Y}_i$")
         plt.title("sgdregressor Prices vs Predicted prices: $Y_i$ vs $\hat{Y}_i$")
         plt.show()
         print(aa)
(450, 1) (54, 1) (450, 13) (54, 13)
SGDRegressor(alpha=0.0001, average=False, epsilon=0.1, eta0=0.01,
      fit_intercept=True, l1_ratio=0.15, learning_rate='invscaling',
      loss='squared_loss', max_iter=None, n_iter=None, penalty='12',
      power_t=0.25, random_state=None, shuffle=True, tol=None, verbose=0,
      warm_start=False)
coefficient of sgdregressor model [ 22.73990766] [-0.03697114  0.10995223  0.01883134  0.10206
  0.03142505 0.0571567 0.01386032 0.0279774 0.03979366 0.05169953
  0.00950983]
sklearn Cost from train
-0.000943380226666
shape (450,) (450, 1)
<class 'pandas.core.series.Series'>
train cost: 45.1897757515
test cost 17.2183729505
```

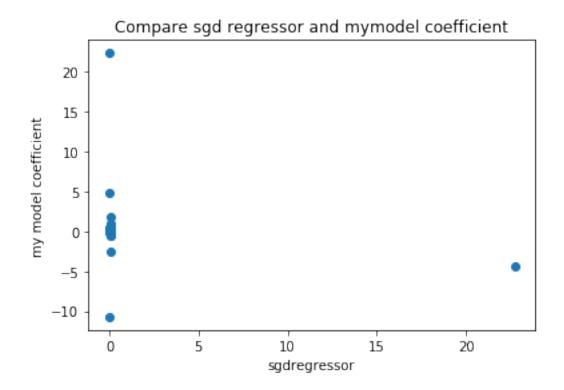


```
Coeff test_cost train_cost \
0 [-4.30642126306, 4.82291219613, -2.55874326256... 15.282411 45.445280
0 [-0.0369711351004, 0.109952233902, 0.018831339... 17.218373 45.189776

type
0 mymodel
0 sgdregressor
```

### 7 Score comparision of mymodel and SGDregressor

```
sklearn coeff [ 22.73990766] [-0.03697114  0.10995223  0.01883134  0.10206679
                                                                               0.03709152
  0.03142505 0.0571567
                          0.01386032 0.0279774
                                                  0.03979366
                                                              0.05169953
  0.00950983]
mymodel theta including intercept -4.30642126306 [
                                                    4.82291220e+00 -2.55874326e+00
                                                                                     -1.069980
                  -1.84281057e-03
   1.74705492e+00
                                     5.26956375e-01
                                                     -5.46160077e-01
  -2.90538283e-01
                    2.86636557e-01
                                                      9.42275633e-01
                                     4.13640841e-01
  2.22545481e+01]
```



#### 8 Observation

- 1. While trying learning rate of different value sometimes cost is going towards very hogh value, so tried with higher value
- 2. After trying more iteration cost is going down with a very small alpha value
- 3. sklearn model is performing better with cost close to 0 but my model cost is close to 30
- 4. Best cost is achieved by different trial cost function
- 5. theta are not changing much from initial random value (when alpha like .000005)
- 6. when alpha taken like .05 , gradient is becoming too high. cost function is becoming inf, theta are becoming very high
- 7. sklearn model is always getting global minimum but my model doesnot