

```
import pandas as pd
import math
import numpy as np
import matplotlib.pyplot as plt
```

```
from google.colab import drive, files
uploaded = files.upload()
```

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Saving data n2 n3.xlsx to data n2 n3.xlsx

```
# function for normalising data
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```
def norm(data):
    mean = np.mean(data, axis = 0)
    std = np.std(data, axis = 0)
    norm_data = (data-mean)/std
    return norm_data
```

```
# defining the cost function
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```
def costfunction(X, y, wt):
    hyp = np.dot(X, wt.T)
    J = (0.5/len(y))*np.sum((hyp - y)**2)
    return J
```

```
#function for regularisation
```

```
def wt_regularisation(lamb, wt, reg==.2):
    ..wt_reg==np.zeros(wt.shape)
    ..if reg==.1:
    ....costreg==(lamb/2)*np.sum(np.abs(wt))
    ....wt_reg==(lamb/2)*np.sign(wt)
    ..elif reg==.2:
    ....costreg==(lamb/2)*np.sum((wt)**2)
    ....wt_reg==lamb*wt
    ....#.print(wt_reg)
    ..return wt_reg
```

```
#function for regularisation
```

```
def J_regularisation(lamb, wt, reg==.2):
    ..costreg==0
    ..if reg==.1:
    ....costreg==(lamb/2)*np.sum(np.abs(wt))
    ..elif reg==.2:
    ....costreg==(lamb/2)*np.sum((wt)**2)
    ..return costreg
```

```
def bgd(alpha, lamb, iters, X, y):
```

```
    ..w==np.random.rand(X.shape[1])
    ..for i in range(iters):
    ....hyp==np.dot(X, w.T)
    ....#.w==w+(-alpha/len(y))*(np.dot(hyp-y,X)+wt_regularisation(lamb, w))
```

```

...#w=w-(alpha/len(y))*(np.dot(hyp-y,X)-wt_regularisation(lamb,w))
...w=w*(1-(alpha/len(y))*lamb)-(alpha/len(y))*np.dot(hyp-y,X)
..return w

def mbgd(alpha, lamb, iters, batch_size, X, y):
..w=np.random.rand(X.shape[1])
..for i in range(iters):
...rand_ind=np.random.randint(len(y))
...X_batch=X[rand_ind:rand_ind+batch_size]
...y_batch=y[rand_ind:rand_ind+batch_size]
...hyp=np.dot(X_batch,w.T)
...w=w-(alpha/len(y))*(np.dot(hyp-y_batch,X_batch)-wt_regularisation(lamb,w))
..return w

def sgd(alpha, lamb, iters, X, y):
..w=np.random.rand(X.shape[1])
..for i in range(iters):
...rand_ind=np.random.randint(len(y))
...X_ind=X[rand_ind:rand_ind+1]
...y_ind=y[rand_ind:rand_ind+1]
...hyp=np.dot(X_ind,w.T)
...#print(hyp.shape)
...#print(y_ind.shape)
...w=w-(alpha/len(y))*(np.dot(hyp-y_ind,X_ind)-wt_regularisation(lamb,w))
..return w

def error(y_testing, y_predicted):
..k=len(y_testing)
..#print(k)
..mae=(0.5/k)*(np.sum(abs(y_testing-y_predicted)))
..mse=(0.5/k)*np.sum((y_testing-y_predicted)**2)
..y_testingdiff=y_testing-np.mean(y_testing)
..y_predicteddiff=y_predicted-np.mean(y_predicted)
..#cc=np.sum(np.multiply(y_testingdiff,y_predicteddiff))/(np.sqrt(np.sum((y_testingdif
..cc=np.corrcoef(y_predicted,y_testing).mean()
..return print('MAE:',mae,'\nMSE:',mse,'\nCC:',cc)

#extracting the data and separating
data=pd.read_excel("data_q2_q3.xlsx")
data=np.asarray(data)
np.random.shuffle(data)
data=np.norm(data)

#splitting the data into training, testing and validation
rows=data.shape[0]
training_data=data[0:int(rows*0.7)]
testing_data=data[int(rows*0.7)::int(rows*0.9)]
validation_data=data[int(rows*0.9)::rows+1]

print(rows,len(training_data),len(testing_data),len(validation_data))

```

```

#splitting into input and output
X_tr==training_data[:,0:-1]..#input
y_tr==training_data[:, -1]....#output

X_ts==testing_data[:,0:-1]..#input
y_ts==testing_data[:, -1]....#output

X_val==validation_data[:,0:-1]..#input
y_val==validation_data[:, -1]....#output

print(X_tr.shape, X_ts.shape, X_val.shape)
print(y_tr.shape, y_ts.shape, y_val.shape)


(80, 4) (23, 4) (12, 4)
(80,) (23,) (12,)


#defining X for regression model
m==X_tr.shape[0]
one_tr==np.ones([m,1])
X_tr==np.append(one_tr, X_tr, axis==1)

m==X_val.shape[0]
one_val==np.ones([m,1])
X_val==np.append(one_val, X_val, axis==1)

m==X_ts.shape[0]
one_ts==np.ones([m,1])
X_ts==np.append(one_ts, X_ts, axis==1)

print(X_tr.shape, X_ts.shape, X_val.shape)


(80, 5) (23, 5) (12, 5)


wt_bgd==np.random.rand(X_tr.shape[1])
alpha_vals==np.linspace(0.001,0.1,num=50)
l_vals==np.linspace(0.0001,0.1,num=50)
err_bgd==1000000
n==0
for a in alpha_vals:
    for l in l_vals:
        wt==bgd(a, l, 1000, X_tr, y_tr)
        #print(wt)
        temp_err==(1/len(y_val))*np.sum((y_val--np.dot(X_val, wt.T)**2))
        #print(wt)
        #print(mse_err)
        if temp_err<err_bgd:
            wt_bgd==wt
            err_bgd==temp_err
            #print(wt_bgd, '\n\n')
            n+=1

print(wt_bgd)

```

```

y_pred_bgd=np.dot(X_ts,wt_bgd.T)
#print(y_pred_bgd,'\n\n',y_ts)
error(y_ts,y_pred_bgd)###mae,mse,cc

[0.29408527 0.86488077 0.35132413 0.07686659 0.09459979]
MAE: 0.3725170453070251
MSE: 0.37423793462132715
CC: 0.8194593541569737

```

```

wt_sgd=np.random.rand(X_tr.shape[1])
alpha_vals=np.logspace(-4,-2,num=50)
l_vals=np.logspace(-3,0,num=50)
err_sgd=1000000
n=0
for a in alpha_vals:
    for l in l_vals:
        wt=sgd(a,l,1000,X_tr,y_tr)
        #print(wt)
        temp_err=(1/len(y_val))*np.sum((y_val-np.dot(X_val,wt.T)**2))
        #print(wt)
        #print(mse_err)
        if temp_err<err_sgd:
            wt_sgd=wt
            err_sgd=temp_err
            #print(wt_bgd,'\n\n')
            n+=1

#print(wt_sgd)
y_pred_sgd=np.dot(X_ts,wt_sgd.T)
#print(y_pred_sgd,'\n\n',y_ts)
error(y_ts,y_pred_sgd)###mae,mse,cc

```

```

☞ MAE: 0.5449139398367401
MSE: 0.9917652028026064
CC: 0.790726564715108

```

```

wt_mbgd=np.random.rand(X_tr.shape[1])
alpha_vals=np.logspace(0.001,0.1,num=50)
l_vals=np.logspace(-3,0,num=50)
err_mbgd=1000000
n=0
for a in alpha_vals:
    for l in l_vals:
        wt=mbgd(a,l,1000,30,X_tr,y_tr)
        #print(wt)
        temp_err=(1/len(y_val))*np.sum((y_val-np.dot(X_val,wt.T)**2))
        #print(wt)
        #print(mse_err)
        if temp_err<err_mbgd:
            wt_mbgd=wt
            err_mbgd=temp_err
            #print(wt_bgd,'\n\n')
            n+=1

```

```
#.print(wt_mbgd)
y_pred_mbgd=np.dot(X_ts,wt_mbgd.T)
#.print(y_pred_mbgd,'\n\n',y_ts)
error(y_ts,y_pred_mbgd)###mae,mse,cc
```

```
MAE: 0.2727196082684354
MSE: 0.22302457636961326
CC: 0.8470644207784837
```

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