LS_Handy-CardenasL

November 18, 2020

```
[4]: import numpy as np
     # from spicy.io import loadmat
    import matplotlib.pyplot as plt
    #A = np.genfromtxt('Data_Raw.csv', delimiter=',')
     #print(A.dtype)
    Data = np.genfromtxt('Data.csv', delimiter=',')
    x_train = Data[0:1001,0:14] # features
    y_train = Data[0:1001,14] # corresponding labels
    # evaluation data
    x_eval= Data[1001:12330,0:14] # features
    y_eval = Data[1001:12330,14] # corresponding labels
    \# X = Data[0:3,0:14]
    # y = Data[:,14]
    # Classifier 1
    \#w = (X^T X)^{(-1)}X^T y
    X = x_train
    y = y_train
    w = np.linalg.inv(X.transpose()@X)@X.transpose()@y
    \#A = np.linalg.inv(X@X.T)
    print(np.round(w,2))
    [ 0.01 0.
               0.02 -0. -0. 0. 0.04 -0.09 0.01 0.01 -0.01 0.
           0. ]
     -0.
[5]: import numpy as np
     # from spicy.io import loadmat
    import matplotlib.pyplot as plt
     #A = np.genfromtxt('Data_Raw.csv', delimiter=',')
    #print(A.dtype)
    Data = np.genfromtxt('Data.csv', delimiter=',')
```

```
x_train = Data[0:1001,0:14] # features
y_train = Data[0:1001,14] # corresponding labels

# evaluation data
x_eval= Data[1001:12330,0:14] # features
y_eval = Data[1001:12330,14] # corresponding labels

# X = Data[0:3,0:14]
# y = Data[:,14]

# Classifier 1
#w = (X^T X)^(-1)X^T y

X = x_eval
y = y_eval
w = np.linalg.inv(X.transpose()@X)@X.transpose()@y
#A = np.linalg.inv(X@X.T)

print(np.round(w,2))
```

```
[]: # 2d
     x_eval = X
     y_eval = y
     # all features
     print('considering all features')
     x train = X
     w_train = w
     y_hat = np.sign(x_train@w_train)
     error_vec = [0 if i[0]==i[1] else 1 for i in np.hstack((y_hat, y_eval))]
     print('Errors: '+ str(sum(error_vec)))
     print('Percent error: '+str(100.0*sum(error_vec)/len(error_vec))+'%')
     # 3 main features
     print('considering 3 main features')
     x_train = X[:, [0, 2, 3]]
     w_train = np.linalg.inv(x_train.transpose()@x_train)@x_train.transpose()@y
     y_hat = np.sign(x_train@w_train)
     error_vec = [0 if i[0]==i[1] else 1 for i in np.hstack((y_hat, y_eval))]
     print('Errors: '+ str(sum(error_vec)))
     print('Percent error: '+str(100.0*sum(error_vec)/len(error_vec))+'%')
     # w = (X^T X)^{(-1)}X^T y
```

```
[]: # 2f
     in_data = loadmat('face_emotion_data.mat')
     X = in_data['X']
     y = in_data['y']
     x_1 = X[range(0,16), :]
     x_2 = X[range(16,32), :]
     x_3 = X[range(16,48), :]
     x_4 = X[range(48,64), :]
     x_5 = X[range(64,80), :]
     x 6 = X[range(80,96), :]
     x 7 = X[range(96,112), :]
     x_8 = X[range(112, 128), :]
     y_1 = y[range(0,16), :]
     y_2 = y[range(16,32), :]
     y_3 = y[range(16,48), :]
     y_4 = y[range(48,64), :]
     y_5 = y[range(64,80), :]
     y_6 = y[range(80,96), :]
     y_7 = y[range(96,112), :]
     y_8 = y[range(112,128), :]
     # 1 variation, hold x_8 and y_8
     print('considering 1st variation')
     x_{train} = np.vstack((x_1, x_2, x_3, x_4, x_5, x_6, x_7))
     y_{train} = np.vstack((y_1, y_2, y_3, y_4, y_5, y_6, y_7))
     w_train = np.linalg.inv(x_train.transpose()@x_train)@x_train.transpose()@y_train
     y_hat = np.sign(x_8@w_train)
     # print(w_train)
     error_vec_1 = [0 if i[0]==i[1] else 1 for i in np.hstack((y_hat, y_8))]
     print('Errors: '+ str(sum(error_vec_1)))
     print('Percent error: '+str(100.0*sum(error_vec_1)/16)+'%')
     # 2 variation, hold x_1 and y_1
     print('considering 2nd variation')
     x_{train} = np.vstack((x_2, x_3, x_4, x_5, x_6, x_7, x_8))
     y_{train} = np.vstack((y_2, y_3, y_4, y_5, y_6, y_7, y_8))
     w_train = np.linalg.inv(x_train.transpose()@x_train)@x_train.transpose()@y_train
     y_hat = np.sign(x_1@w_train)
     # print(w_train)
     error_vec_2 = [0 if i[0]==i[1] else 1 for i in np.hstack((y_hat, y_1))]
```

```
print('Errors: '+ str(sum(error_vec_2)))
print('Percent error: '+str(100.0*sum(error_vec_2)/16)+'%')
# 3 variation, , hold x_2 and y_2
print('considering 3rd variation')
x_{train} = np.vstack((x_1, x_3, x_4, x_5, x_6, x_7, x_8))
y_{train} = np.vstack((y_1, y_3, y_4, y_5, y_6, y_7, y_8))
w_train = np.linalg.inv(x_train.transpose()@x_train)@x_train.transpose()@y_train
y hat = np.sign(x 2@w train)
# print(w train)
error vec 3 = [0 \text{ if } i[0] == i[1] \text{ else } 1 \text{ for } i \text{ in np.hstack}((y \text{ hat}, y 2))]
print('Errors: '+ str(sum(error_vec_3)))
print('Percent error: '+str(100.0*sum(error_vec_3)/16)+'%')
# 4 variation, hold x_3 and y_3
print('considering 4th variation')
x_{train} = np.vstack((x_1, x_2, x_4, x_5, x_6, x_7, x_8))
y_{train} = np.vstack((y_1, y_2, y_4, y_5, y_6, y_7, y_8))
w_train = np.linalg.inv(x_train.transpose()@x_train)@x_train.transpose()@y_train
y_hat = np.sign(x_3@w_train)
# print(w_train)
error_vec_4 = [0 \text{ if } i[0]==i[1] \text{ else } 1 \text{ for } i \text{ in } np.hstack((y_hat, y_3))]
print('Errors: '+ str(sum(error_vec_4)))
print('Percent error: '+str(100.0*sum(error vec 4)/16)+'%')
# 5 variation, hold x 4 and y 4
print('considering 5th variation')
x_{train} = np.vstack((x_1, x_2, x_3, x_5, x_6, x_7, x_8))
y_train = np.vstack((y_1, y_2, y_3, y_5, y_6, y_7, y_8))
w_train = np.linalg.inv(x_train.transpose()@x_train)@x_train.transpose()@y_train
y_hat = np.sign(x_4@w_train)
# print(w_train)
error_vec_5 = [0 \text{ if } i[0]==i[1] \text{ else } 1 \text{ for } i \text{ in } np.hstack((y_hat, y_4))]
print('Errors: '+ str(sum(error_vec_5)))
print('Percent error: '+str(100.0*sum(error_vec_5)/16)+'%')
# 6 variation, hold x 5 and y 5
print('considering 6th variation')
x_{train} = np.vstack((x_1, x_2, x_3, x_4, x_6, x_7, x_8))
y_train = np.vstack((y_1, y_2, y_3, y_4, y_6, y_7, y_8))
w_train = np.linalg.inv(x_train.transpose()@x_train)@x_train.transpose()@y_train
y_hat = np.sign(x_5@w_train)
# print(w train)
error_vec_6 = [0 \text{ if } i[0]==i[1] \text{ else } 1 \text{ for } i \text{ in } np.hstack((y_hat, y_5))]
print('Errors: '+ str(sum(error_vec_6)))
print('Percent error: '+str(100.0*sum(error_vec_6)/16)+'%')
```

```
# 7 variation, hold x_6 and y_6
print('considering 7th variation')
x_{train} = np.vstack((x_1, x_2, x_3, x_4, x_5, x_7, x_8))
y_{train} = np.vstack((y_1, y_2, y_3, y_4, y_5, y_7, y_8))
w_train = np.linalg.inv(x_train.transpose()@x_train)@x_train.transpose()@y_train
y_hat = np.sign(x_6@w_train)
# print(w_train)
error_vec_7 = [0 if i[0]==i[1] else 1 for i in np.hstack((y_hat, y_6))]
print('Errors: '+ str(sum(error_vec_7)))
print('Percent error: '+str(100.0*sum(error_vec_7)/16)+'%')
# 8 variation, hold x_7 and y_7
print('considering 8th variation')
x_{train} = np.vstack((x_1, x_2, x_3, x_4, x_5, x_6, x_8))
y_{train} = np.vstack((y_1, y_2, y_3, y_4, y_5, y_6, y_8))
w_train = np.linalg.inv(x_train.transpose()@x_train)@x_train.transpose()@y_train
y_hat = np.sign(x_7@w_train)
# print(w_train)
error_vec_8 = [0 if i[0]==i[1] else 1 for i in np.hstack((y_hat, y_7))]
print('Errors: '+ str(sum(error_vec_8)))
print('Percent error: '+str(100.0*sum(error_vec_8)/16)+'%')
print('Final performance estimate:')
→average_error_rate=(error_vec_1+error_vec_2+error_vec_3+error_vec_4+error_vec_5+error_vec_6
∽8
hola = ((1/
→16)*(sum(error_vec_1)+sum(error_vec_2)+sum(error_vec_3)+sum(error_vec_4)+sum(error_vec_5)+s
→8)*100
print(str(hola)+'%')
```

[]:

[]: