

Course Number: EEL 5840

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Homework 5:

Q1.

Implement the Bayes classifier, under the assumption that your likelihood model $p(x | j)$ is unimodal Gaussian and that the prior probabilities $p(j)$ are dictated by the number of samples n_j belong to R that you have for each class. This classifier is given by the following discriminant function for each class j . (the order of the matrix for the species from up to down and from left to right is 0 and 1)

1) Matrix for test data: $\begin{bmatrix} 28 & 0 \\ 0 & 32 \end{bmatrix}$

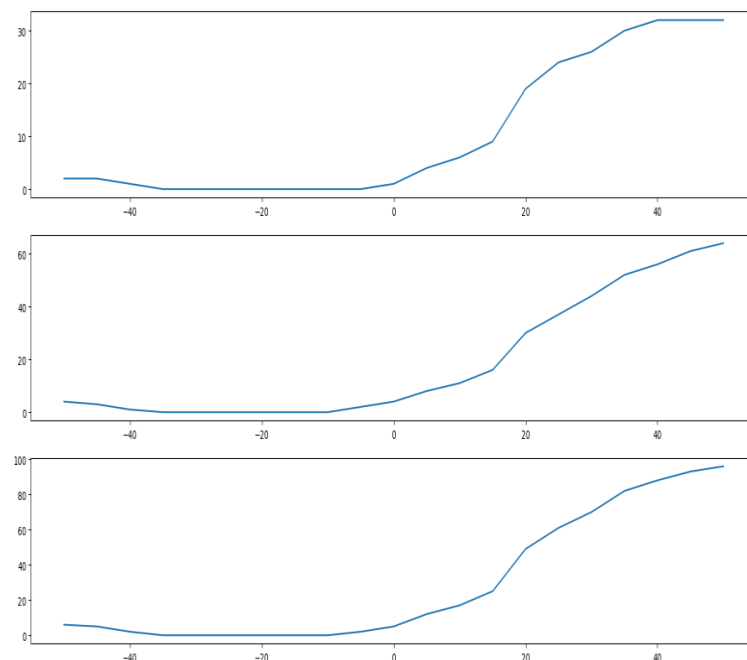
2) Matrix for training data: $\begin{bmatrix} 72 & 0 \\ 0 & 68 \end{bmatrix}$

3) Matrix for all the data: $\begin{bmatrix} 100 & 0 \\ 0 & 100 \end{bmatrix}$

4) Processing timer: 0.3445889949798584 seconds

Q2.

Implement a classifier based on linear discriminant analysis. This is a particular case of the Bayes classifier. Here, T belongs to R is a potentially non-zero threshold value that you should choose, in some fashion, on your own. The variables $\mu_0; \mu_1$ R_7 are the class means for the two classes and Σ_0, Σ_1 belongs to $R^{7 \times 7}$ are the class covariances for the two classes. (the order of the matrix for the species from up to down and from left to right is 0 and 1) To make less error I plot a picture of T and sum of error from -50 to 50 as follow and found that the threshold around -20 can cause no error.



The picture of the sum of error and threshold T

1) Matrix for test data: $\begin{bmatrix} 28 & 0 \\ 0 & 32 \end{bmatrix}$

2) Matrix for training data: $\begin{bmatrix} 72 & 0 \\ 0 & 68 \end{bmatrix}$

3) Matrix for all the data: $\begin{bmatrix} 100 & 0 \\ 0 & 100 \end{bmatrix}$

4) Processing time: 0.1093592643737793 seconds

While I did this question, I found that the right of the $g(x)$ is too big, when T is closed to zero. Afterwards, I check out the data and found that the last two dimensions are 'male' and 'female'. These two-opposite phenomena cause the covariance of the two class to be very big. So I add the regularized term λ to adjust the problem (If I drop one of the last two dimension, the problem can also be solved). I choose the λ equals to 0.01. To get the threshold, I set the different threshold T , and I got the answers above.

Code 1:

```
import numpy as np
import matplotlib.pyplot as plt
import math
import time
start_time = time.time()

#initial parameters and load data
data = np.loadtxt("dataset.txt")
alldata = data[:,1:8]
training = data[0:140,1:8]
test = data[140:200,1:8]
validate_training = data[0:140,0]
validate_test = data[140:200,0]
validate_data = data[:,0]

count = 0
for i in range(140):
    if validate_training[i] == 1:
        count = count + 1
p_prior1 = count/140
p_prior0 = 1 - (count/140)
```

```
tr_1 = np.zeros([count,7])
```

```
tr_0 = np.zeros([140-count,7])
```

```
ind1 = 0
```

```
ind0 = 0
```

```
for i in range(140):
```

```
    if validate_training[i] == 1:
```

```
        tr_1[ind1] = training[i,:]
```

```
        ind1 = ind1 + 1
```

```
    if validate_training[i] == 0:
```

```
        tr_0[ind0] = training[i,:]
```

```
        ind0 = ind0 + 1
```

```
mean_1 = np.zeros(7)
```

```
mean_0 = np.zeros(7)
```

```
for i in range(7):
```

```
    mean_1[i] = sum(tr_1[:,i])/count
```

```
    M1 = mean_1[i]*np.ones([count,1])
```

```
    mean_0[i] = sum(tr_0[:,i])/(140-count)
```

```
M0 = mean_0[i]*np.ones([(140-count),1])
```

```
sig_1 = 0
```

```
sig_0 = 0
```

```
for i in range(count):
```

```
    sig_1 = sig_1 + (tr_1[i,:]-np.asmatrix(mean_1)).T@(tr_1[i,:]-  
np.asmatrix(mean_1))/count
```

```
for i in range(140-count):
```

```
    sig_0 = sig_0 + (tr_0[i,:]-np.asmatrix(mean_0)).T@(tr_0[i,:]-  
np.asmatrix(mean_0))/(140-count)
```

```
#sig_1 = np.cov(tr_1.T)
```

```
#sig_0 = np.cov(tr_0.T)
```

```
count1 = 0
```

```
count0 = 0
```

```
for i in range(60):
```

```
    g1 = -0.5*((np.asmatrix(test[i,:])  
np.asmatrix(mean_1))@np.linalg.inv(sig_1)@(np.asmatrix(test[  
i,:]) - np.asmatrix(mean_1)).T) - 3.5*math.log1p(2*math.pi) -  
0.5*math.log1p(np.linalg.det(sig_1)) + math.log1p(p_prior1)
```

```
    g0 = -0.5*((np.asmatrix(test[i,:])
```

```
np.asmatrix(mean_0))@np.linalg.inv(sig_0)@(np.asmatrix(test[
i,:]) - np.asmatrix(mean_0)).T) - 3.5*math.log1p(2*math.pi) -
0.5*math.log1p(np.linalg.det(sig_0)) + math.log1p(p_prior0)
```

```
if g1 >= g0:
```

```
    if validate_test[i] == 1:
```

```
        count1 = count1 + 1
```

```
    else:
```

```
        if g0 > g1:
```

```
            if validate_test[i] == 0:
```

```
                count0 = count0 + 1
```

```
c = 0
```

```
for i in range(60):
```

```
    if validate_test[i] == 1:
```

```
        c = c + 1
```

```
error1 = c - count1
```

```
error0 = 60 - c - count0
```

```
Matrix1 = np.ones([2,2])
```

```
Matrix1[0,0] = count0
```

```
Matrix1[1,0] = error0
```



```
Matrix1[0,1] = error1
```

```
Matrix1[1,1] = count1
```

```
count1 = 0
```

```
count0 = 0
```

```
for i in range(140):
```

```
    g1 = -0.5*((np.asmatrix(training[i,:]) -  
np.asmatrix(mean_1))@np.linalg.inv(sig_1)@(np.asmatrix(train  
ing[i,:]) - np.asmatrix(mean_1)).T) - 3.5*math.log1p(2*math.pi)  
- 0.5*math.log1p(np.linalg.det(sig_1)) + math.log1p(p_prior1)
```

```
    g0 = -0.5*((np.asmatrix(training[i,:]) -  
np.asmatrix(mean_0))@np.linalg.inv(sig_0)@(np.asmatrix(train  
ing[i,:]) - np.asmatrix(mean_0)).T) - 3.5*math.log1p(2*math.pi)  
- 0.5*math.log1p(np.linalg.det(sig_0)) + math.log1p(p_prior0)
```

```
    if g1 >= g0:
```

```
        if validate_training[i] == 1:
```

```
            count1 = count1 + 1
```

```
    else:
```

```
        if g0 > g1:
```

```
            if validate_training[i] == 0:
```

```
count0 = count0 + 1
```

```
c = 0
```

```
for i in range(140):
```

```
    if validate_training[i] == 1:
```

```
        c = c + 1
```

```
error1 = c - count1
```

```
error0 = 140 - c - count0
```

```
Matrix2 = np.ones([2,2])
```

```
Matrix2[0,0] = count0
```

```
Matrix2[1,0] = error0
```

```
Matrix2[0,1] = error1
```

```
Matrix2[1,1] = count1
```

```
count1 = 0
```

```
count0 = 0
```

```
for i in range(200):
```

```
    g1 = -0.5*((np.asmatrix(alldata[i,:]) -  
np.asmatrix(mean_1))@np.linalg.inv(sig_1)@(np.asmatrix(allda
```

```

ta[i,:]) - np.asmatrix(mean_1)).T) - 3.5*math.log1p(2*math.pi) -
0.5*math.log1p(np.linalg.det(sig_1)) + math.log1p(p_prior1)

    g0          =          -0.5*((np.asmatrix(alldata[i,:])          -
np.asmatrix(mean_0))@np.linalg.inv(sig_0)@(np.asmatrix(allda
ta[i,:]) - np.asmatrix(mean_0)).T) - 3.5*math.log1p(2*math.pi) -
0.5*math.log1p(np.linalg.det(sig_0)) + math.log1p(p_prior0)

    if g1 >= g0:
        if validate_data[i] == 1:
            count1 = count1 + 1
        else:
            if g0 > g1:
                if validate_data[i] == 0:
                    count0 = count0 + 1

c = 0
for i in range(200):
    if validate_data[i] == 1:
        c = c + 1

error1 = c - count1
error0 = 200 - c - count0

```

```
Matrix3 = np.ones([2,2])
```

```
Matrix3[0,0] = count0
```

```
Matrix3[1,0] = error0
```

```
Matrix3[0,1] = error1
```

```
Matrix3[1,1] = count1
```

```
print(Matrix1)#test
```

```
print(Matrix2)#training
```

```
print(Matrix3)#all
```

```
print("--- %s seconds ---" % (time.time() -  
start_time))#processing time
```

Code 2:

```
import numpy as np
import matplotlib.pyplot as plt
import math
import time
start_time = time.time()

#initial parameters and load data
data = np.loadtxt("dataset.txt")
alldata = data[:,1:8]
training = data[0:140,1:8]
test = data[140:200,1:8]
validate_training = data[0:140,0]
validate_test = data[140:200,0]
validate_data = data[:,0]

count = 0
for i in range(140):
    if validate_training[i] == 1:
        count = count + 1
p_prior1 = count/140
p_prior0 = 1 - (count/140)
```

```
tr_1 = np.zeros([count,7])
```

```
tr_0 = np.zeros([140-count,7])
```

```
ind1 = 0
```

```
ind0 = 0
```

```
for i in range(140):
```

```
    if validate_training[i] == 1:
```

```
        tr_1[ind1] = training[i,:]
```

```
        ind1 = ind1 + 1
```

```
    if validate_training[i] == 0:
```

```
        tr_0[ind0] = training[i,:]
```

```
        ind0 = ind0 + 1
```

```
mean_1 = np.zeros(7)
```

```
mean_0 = np.zeros(7)
```

```
for i in range(7):
```

```
    mean_1[i] = sum(tr_1[:,i])/count
```

```
    M1 = mean_1[i]*np.ones([count,1])
```

```
    mean_0[i] = sum(tr_0[:,i])/(140-count)
```

```
M0 = mean_0[i]*np.ones([(140-count),1])
```

```
sig_1 = 0
```

```
sig_0 = 0
```

```
for i in range(count):
```

```
    sig_1 = sig_1 + (tr_1[i,:]-np.asmatrix(mean_1)).T@(tr_1[i,:]-  
np.asmatrix(mean_1))/count
```

```
for i in range(140-count):
```

```
    sig_0 = sig_0 + (tr_0[i,:]-np.asmatrix(mean_0)).T@(tr_0[i,:]-  
np.asmatrix(mean_0))/(140-count)
```

```
#sig_1 = np.cov(tr_1.T)
```

```
#sig_0 = np.cov(tr_0.T)
```

```
lambda_0 = 0.01
```

```
lambda_1 = 0.01
```

```
sig_0 = sig_0 + lambda_0*np.eye(7)
```

```
sig_1 = sig_1 + lambda_1*np.eye(7)
```

```
count1 = 0
```

```
count0 = 0
```

```

T = -5
for i in range(60):
    g1 =
np.asmatrix(test[i,:])@np.linalg.inv(0.5*(sig_1+sig_0))@(np.as
matrix(mean_1)-np.asmatrix(mean_0)).T
    g0 =
0.25*(np.asmatrix(mean_0)@np.linalg.inv(sig_0)@np.asmatrix(
mean_0).T -
np.asmatrix(mean_1)@np.linalg.inv(sig_1)@np.asmatrix(mean
_1).T + T)
    if g1 >= g0:
        if validate_test[i] == 1:
            count1 = count1 + 1
        else:
            if g1 < g0:
                if validate_test[i] == 0:
                    count0 = count0 + 1
print(g0)

c = 0
for i in range(60):
    if validate_test[i] == 1:

```



```
c = c + 1
```

```
error1 = c - count1
```

```
error0 = 60 - c - count0
```

```
Matrix1 = np.ones([2,2])
```

```
Matrix1[0,0] = count0
```

```
Matrix1[1,0] = error0
```

```
Matrix1[0,1] = error1
```

```
Matrix1[1,1] = count1
```

```
count1 = 0
```

```
count0 = 0
```

```
T = -10
```

```
for i in range(140):
```

```
    g1 =
```

```
    np.asmatrix(training[i,:])@np.linalg.inv(0.5*(sig_1+sig_0))@(np
```

```
    .asmatrix(mean_1)-np.asmatrix(mean_0)).T
```

```

g0 =
0.25*(np.asmatrix(mean_0)@np.linalg.inv(sig_0)@np.asmatrix(
mean_0).T
-
np.asmatrix(mean_1)@np.linalg.inv(sig_1)@np.asmatrix(mean
_1).T + T)

```

```

if g1 >= g0:

```

```

    if validate_training[i] == 1:

```

```

        count1 = count1 + 1

```

```

    else:

```

```

        if g1 < g0:

```

```

            if validate_training[i] == 0:

```

```

                count0 = count0 + 1

```

```

c = 0

```

```

for i in range(140):

```

```

    if validate_training[i] == 1:

```

```

        c = c + 1

```

```

error1 = c - count1

```

```

error0 = 140 - c - count0

```

```

Matrix2 = np.ones([2,2])

```

Matrix2[0,0] = count0

Matrix2[1,0] = error0

Matrix2[0,1] = error1

Matrix2[1,1] = count1

count1 = 0

count0 = 0

T = -10

for i in range(200):

```
    g1 =
    np.asmatrix(alldata[i,:])@np.linalg.inv(0.5*(sig_1+sig_0))@(np.
    asmatrix(mean_1)-np.asmatrix(mean_0)).T
    g0 =
    0.25*(np.asmatrix(mean_0)@np.linalg.inv(sig_0)@np.asmatrix(
    mean_0).T -
    np.asmatrix(mean_1)@np.linalg.inv(sig_1)@np.asmatrix(mean
    _1).T + T)
    if g1 >= g0:
```

```
        if validate_data[i] == 1:
            count1 = count1 + 1
    else:
        if g1 < g0:
            if validate_data[i] == 0:
                count0 = count0 + 1
print(g0)
```

```
c = 0
for i in range(200):
    if validate_data[i] == 1:
        c = c + 1
```

```
error1 = c - count1
error0 = 200 - c - count0
```

```
Matrix3 = np.ones([2,2])
Matrix3[0,0] = count0
Matrix3[1,0] = error0
Matrix3[0,1] = error1
Matrix3[1,1] = count1
```

```
print(Matrix1)#test
print(Matrix2)#training
print(Matrix3)#all
print("--- %s seconds ---" % (time.time() -
start_time))#processing time
```

Code 3:

```
import numpy as np
import matplotlib.pyplot as plt
import math
import time

start_time = time.time()

fig = plt.figure(figsize=(30,10))

#initial parameters and load data
data = np.loadtxt("dataset.txt")
alldata = data[:,1:8]
training = data[0:140,1:8]
test = data[140:200,1:8]
validate_training = data[0:140,0]
validate_test = data[140:200,0]
validate_data = data[:,0]

count = 0

for i in range(140):
    if validate_training[i] == 1:
        count = count + 1

p_prior1 = count/140
```

```
p_prior0 = 1 - (count/140)
```

```
tr_1 = np.zeros([count,7])
```

```
tr_0 = np.zeros([140-count,7])
```

```
ind1 = 0
```

```
ind0 = 0
```

```
for i in range(140):
```

```
    if validate_training[i] == 1:
```

```
        tr_1[ind1] = training[i,:]
```

```
        ind1 = ind1 + 1
```

```
    if validate_training[i] == 0:
```

```
        tr_0[ind0] = training[i,:]
```

```
        ind0 = ind0 + 1
```

```
mean_1 = np.zeros(7)
```

```
mean_0 = np.zeros(7)
```

```
for i in range(7):
```

```
    mean_1[i] = sum(tr_1[:,i])/count
```

```
    M1 = mean_1[i]*np.ones([count,1])
```

```
mean_0[i] = sum(tr_0[:,i])/(140-count)
```

```
M0 = mean_0[i]*np.ones([(140-count),1])
```

```
sig_1 = 0
```

```
sig_0 = 0
```

```
for i in range(count):
```

```
    sig_1 = sig_1 + (tr_1[i,:]-np.asmatrix(mean_1)).T@(tr_1[i,:]-  
np.asmatrix(mean_1))/count
```

```
for i in range(140-count):
```

```
    sig_0 = sig_0 + (tr_0[i,:]-np.asmatrix(mean_0)).T@(tr_0[i,:]-  
np.asmatrix(mean_0))/(140-count)
```

```
#sig_1 = np.cov(tr_1.T)
```

```
#sig_0 = np.cov(tr_0.T)
```

```
lambda_0 = 0.01
```

```
lambda_1 = 0.01
```

```
sig_0 = sig_0 + lambda_0*np.eye(7)
```

```
sig_1 = sig_1 + lambda_1*np.eye(7)
```

```
count1 = 0
```

```
count0 = 0
```



```

error = np.zeros(21)

d = np.arange(-50,55,5)

for t in range(0,21,1):
    T = t*5 - 50
    for i in range(60):
        g1 =
np.asmatrix(test[i,:])@np.linalg.inv(0.5*(sig_1+sig_0))@(np.as
matrix(mean_1)-np.asmatrix(mean_0)).T
        g0 =
0.25*(np.asmatrix(mean_0)@np.linalg.inv(sig_0)@np.asmatrix(
mean_0).T -
np.asmatrix(mean_1)@np.linalg.inv(sig_1)@np.asmatrix(mean
_1).T + T)
        if g1 >= g0:
            if validate_test[i] == 1:
                count1 = count1 + 1
            else:
                if g1 < g0:
                    if validate_test[i] == 0:
                        count0 = count0 + 1

c = 0

```

```

for i in range(60):
    if validate_test[i] == 1:
        c = c + 1
    error1 = c - count1
    error0 = 60 - c - count0
    error[t] = error1 + error0
    count1 = 0
    count0 = 0
p1 = fig.add_subplot(*[3,1,1])
p1.plot(d,error)

for t in range(0,21,1):
    T = t*5 - 50
    for i in range(140):
        g1 =
np.asmatrix(training[i,:])@np.linalg.inv(0.5*(sig_1+sig_0))@(np
.assmatrix(mean_1)-np.assmatrix(mean_0)).T
        g0 =
0.25*(np.assmatrix(mean_0)@np.linalg.inv(sig_0)@np.assmatrix(
mean_0).T
-
np.assmatrix(mean_1)@np.linalg.inv(sig_1)@np.assmatrix(mean
_1).T + T)

```

```

    if g1 >= g0:
        if validate_training[i] == 1:
            count1 = count1 + 1
        else:
            if g1 < g0:
                if validate_training[i] == 0:
                    count0 = count0 + 1

c = 0
for i in range(140):
    if validate_training[i] == 1:
        c = c + 1

error1 = c - count1
error0 = 140 - c - count0
error[t] = error1 + error0
count1 = 0
count0 = 0

p2 = fig.add_subplot(*[3,1,2])
p2.plot(d,error)

for t in range(0,21,1):
    T = t*5 - 50
    for i in range(200):

```

```

g1 =
np.asmatrix(alldata[i,:])@np.linalg.inv(0.5*(sig_1+sig_0))@(np.
asmatrix(mean_1)-np.asmatrix(mean_0)).T

g0 =
0.25*(np.asmatrix(mean_0)@np.linalg.inv(sig_0)@np.asmatrix(
mean_0).T
-
np.asmatrix(mean_1)@np.linalg.inv(sig_1)@np.asmatrix(mean
_1).T + T)

if g1 >= g0:
    if validate_data[i] == 1:
        count1 = count1 + 1
    else:
        if g1 < g0:
            if validate_data[i] == 0:
                count0 = count0 + 1

c = 0
for i in range(200):
    if validate_data[i] == 1:
        c = c + 1

error1 = c - count1
error0 = 200 - c - count0
error[t] = error1 + error0

```

```
count1 = 0
```

```
count0 = 0
```

```
p3 = fig.add_subplot(*[3,1,3])
```

```
p3.plot(d,error)
```