## EECS 545 Homework 2

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## Problem 2

Code and result is as follows:

```
import numpy as np
# # Process the data
z = np.genfromtxt('spambase.data', dtype = float, delimiter = ',')
np.random.seed(0) # Seed the random number generator
rp = np.random.permutation(z.shape[0]) # random permutation of indices
z = z[rp,:] # shuffle the rows of z
x = z[:,:-1]
y = z[:,-1]
# Quantize variables with option 2 where values equal to the median to 1
row_num,col_num = x.shape; y = list(y)
x_{train} = x[0:2000,:]
y_train = y[0:2000]
x_test = x[2000:row_num,:]
y_test= y[2000:row_num]
mid_train = np.median(x_train,axis=0)
for i in range(2000):
   for j in range(col_num):
       if x_train[i,j] > mid_train[j]:
          x_{train[i,j]} = 2
       else:
          x_{train[i,j]} = 1
# Build up Naive Bayes Classifier
## compute the probability of y=0,1 and the conditional probability of x_j=1,2 given y=0,1
n_1 = sum(y_train); n_0 = len(y_train) - n_1
pi_1 = n_1/len(y_train); pi_0 = 1-pi_1 # the pmf of y
count_x_equ_1_y_equ_0 = np.zeros(col_num)
```

```
count_x_equ_1_y_equ_1 = np.zeros(col_num)
count_x_equ_2_y_equ_0 = np.zeros(col_num)
count_x_equ_2_y_equ_1 = np.zeros(col_num)
for 1 in range(len(y_train)):
   if y_train[1] == 0:
       for m in range(col_num):
          if x train[1,m] == 1:
              count_x_equ_1_y_equ_0[m] += 1
          elif x_train[1,m] == 2:
              count_x_equ_2_y_equ_0[m] += 1
   elif y_train[l] == 1:
       for n in range(col_num):
          if x_train[1,n] == 1:
              count_x_equ_1_y_equ_1[n] += 1
          elif x_train[1,n] == 2:
              count_x_equ_2_y_equ_1[n] += 1
p_yto0_j_xto1 = np.zeros(col_num)
p_yto1_j_xto1 = np.zeros(col_num)
p_yto0_j_xto2 = np.zeros(col_num)
p_yto1_j_xto2 = np.zeros(col_num)
for i in range(col_num):
   p_yto0_j_xto1[i] = count_x_equ_1_y_equ_0[i]/n_0
   p_yto0_j_xto2[i] = count_x_equ_2_y_equ_0[i]/n_0
for i in range(col_num):
   p_yto1_j_xto1[i] = count_x_equ_1_y_equ_1[i]/(len(y_train)-n_0)
   p_yto1_j_xto2[i] = count_x_equ_2_y_equ_1[i]/(len(y_train)-n_0)
# Test data
## quantize the test data with median of training data, and compute the test result
for i in range(x_test.shape[0]):
   for j in range(x test.shape[1]):
       if x_test[i,j] > mid_train[j]:
          x \text{ test[i,j]} = 2
       else:
          x_{test[i,j]} = 1
y_test_result = np.zeros(x_test.shape[0])
for i in range(x_test.shape[0]):
   y0 = 1; y1 = 1
   for j in range(x_test.shape[1]):
       if x \text{ test[i,j]} == 1:
          y0 = y0*p_yto0_j_xto1[j]
          y1 = y1*p_yto1_j_xto1[j]
       elif x_test[i,j] == 2:
```

```
y0 = y0*p_yto0_j_xto2[j]
          y1 = y1*p_yto1_j_xto2[j]
   if (pi_0*y0) >= (pi_1*y1):
       y_test_result[i] = 0
   else:
       y_test_result[i] = 1
## Test error
error = 0
for i in range(len(y_test)):
   if y_test[i] != y_test_result[i]:
       error += 1
print("The test error of spam emails by Naive Bayes classifier is %f." %(error/len(y_test)) )
## Sanity check
major = 1-y_train.count(1)/len(y_train) # In the training data, the major class for emails
    is "not spam", so we assume to predict all emails are not spam emails.
error_sanity = y_test.count(1)/len(y_test)
print("The sanity check error is %f." %(error_sanity))
the result is:
The test error of spam emails by Naive Bayes classifier is 0.105344.
The sanity check error is 0.386774.
```

## Problem 4

the code and the result is as follows:

```
from numpy import *
import numpy as np
import scipy.io as sio
import matplotlib.pyplot as plt

# Import the data
mnist_49_3000 = sio.loadmat('mnist_49_3000.mat')
x = mnist_49_3000['x']
y = mnist_49_3000['y']
d,n = x.shape
y += (y < 0) * 1

# Process the data, we divide the data into training part and test part
added = np.ones(n)</pre>
```

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x_original = x; y_original = y
x = mat(np.vstack((added,x))); y = mat(y)
x_train = x[:,:2000]; y_train = y[:,:2000]
x_{test} = x[:,2000:]; y_{test} = y[:,2000:]
# Initialize parameters
lamda = 10
theta = mat(np.zeros(d + 1))
dJ = mat(np.zeros(d + 1))
d2J = mat(np.zeros([d + 1,d + 1]))
# Iteration with Newton's Method:
## Notice here the stop condition for iteration is when the change of theta is less than 1%
N = 0; error0 = 10
while (error0 > 0.01):
   z = 1 / (1 + np.exp(-theta*x_train))
   dJ = x_train*(z-y_train).T+2*lamda*theta.T
   d2J = x_train * mat(diag(multiply(z, (1-z)).getA()[0])) * x_train.T +
       2*lamda*mat(np.eye(d+1,d+1))
   error0 = np.sqrt(((d2J.I * dJ).T * (d2J.I * dJ))[0,0])
   theta = theta - (d2J.I * dJ).T
   y_train_result = theta * x_train
   for 1 in range(2000):
       if y_train_result[0,1] < 0:</pre>
          y_train_result[0,1] = 0
       else:
          y_train_result[0,1] = 1
   N = N+1
print("Iteration times: ", N)
log_like = 0
for 1 in range(2000):
   z = 1 / (1 + np.exp(-theta * x_train[:,1]))
   log_like += y_train[0,1]*np.log(z)+(1-y_train[0,1])*np.log(1-z)
J = -log_like + lamda*theta*theta.T
print("Value of objective function is: ", J[0,0])
# Test data
y_test_result = theta * x_test
eta_test = 1/(1+np.exp(-y_test_result))
false = []
for 1 in range(1000):
   if y_test_result[0, 1] < 0:</pre>
       y_test_result[0, 1] = 0
   else:
```

```
y_{test_result[0, 1] = 1
error1 = 0
for m in range(1000):
   if y_test_result[0, m] != y_test[0, m]:
       error1 += 1
       false.append(m)
print("Test error is: ", error1/1000)
prob = np.zeros(1000)
x_false = x_test[1:,false]; x_false = x_false.getA()
y_false = y_test[:,false]
y_false_result = y_test_result[:,false]
for 1 in range(1000):
   if y_test_result[0,1] == 0:
       prob[1] = 1-eta_test[0,1]
       prob[1] = eta_test[0,1]
confidence = prob[false]
indx = argsort(confidence)[28:]
x_false20 = np.zeros((d,20)); y_real20 = np.zeros(20); y_pre20 = np.zeros(20)
for 1 in range(20):
   x_false20[:,1] = x_false[:,int(indx[1])]
   y_real20[1] = y_false[:,int(indx[1])]
   y_pre20[1] = y_false_result[:,int(indx[1])]
## Plot the picture of 20 most confident missclassified pictures
fig = plt.figure(num='missclassified',figsize=(8,8))
fig.suptitle("\"True\" represents real result, \"Pre\" represents predicted result\n")
for 1 in range(20):
   plt.subplot(4,5,1+1)
   if y_real20[1] == 0:
       true title = 4
      pre_title = 9
   else:
       true_title = 9
       pre_title = 4
   plt.title('True: %s, Pre: %s' %(str(true_title), str(pre_title)))
   plt.imshow(np.reshape(x_false20[:,1], (int(np.sqrt(d)), int(np.sqrt(d)))))
plt.show()
plt.close()
```

the result is:

Iteration times: 6

"True" represents real result, "Pre" represents predicted result

