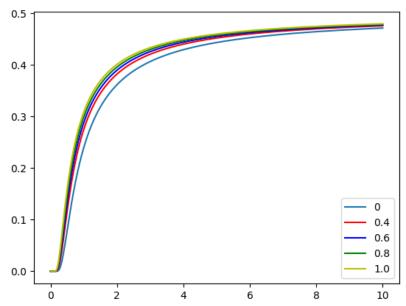
Materials Informatics – Fall 2017 Computer Project 1 – Solutions Due on: Oct 10 2017 11:59pm Jianfeng Song Jsong26@tamu.edu

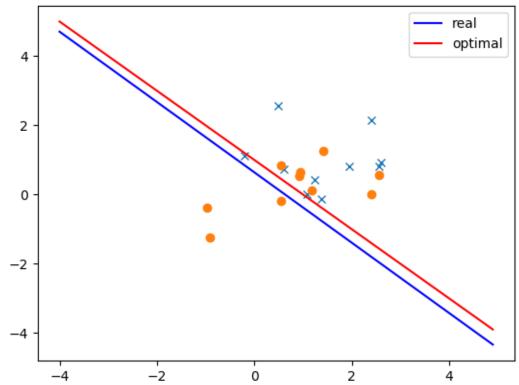
# Assignment 1:

a.  $\begin{cases}
e = \phi(\frac{1}{2}\sigma)
\end{cases}$   $\begin{cases}
e = \phi(\frac{1}{2}\sigma)
\end{cases}$   $\begin{cases}
c = (u_1 + u_2) = (u_1 - u_2)
\end{cases}$   $\begin{cases}
u_1 = (1, 1) \quad u_2 = (0, 0)
\end{cases}$   $\begin{cases}
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\end{cases}$   $e = (u_1 + u_2)$   $e = (u_1 + u_2)
\end{cases}$   $e = (u_1 + u_2)$   $e = (u_1 + u$ 



When we fixed p, as we increase variance the optimal error is also increased, when we fixed variance, the optimal error will increase as we increase p.





LDA uses the Gaussian assumption to estimate the optimal classifier, using the sample means and sample covariance matrices.

$$\hat{\mu}_{0} = \frac{1}{n_{0}} \sum_{i=1}^{n} X_{i} I_{Y_{i}=0} \qquad \hat{\mu}_{1} = \frac{1}{n_{1}} \sum_{i=1}^{n} X_{i} I_{Y_{i}=1}$$

$$\hat{\Sigma} = \frac{1}{n-2} \sum_{i=1}^{n} \left[ (X_{i} - \hat{\mu}_{0})(X_{i} - \hat{\mu}_{0})^{T} I_{Y_{i}=0} + (X_{i} - \hat{\mu}_{1})(X_{i} - \hat{\mu}_{1})^{T} I_{Y_{i}=1} \right]$$

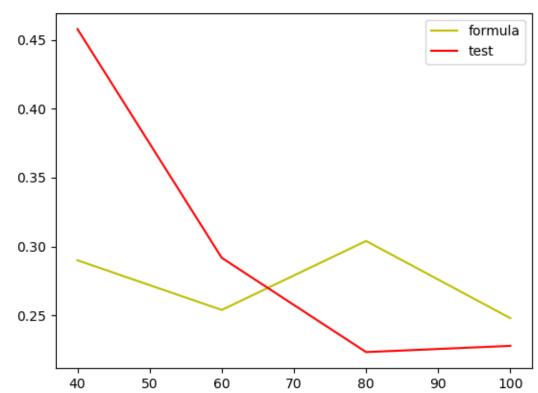
$$a_{n} = \hat{\Sigma}^{-1} (\hat{\mu}_{1} - \hat{\mu}_{0})$$

$$b_{n} = -\frac{1}{2} (\hat{\mu}_{1} - \hat{\mu}_{0})^{T} \hat{\Sigma}^{-1} (\hat{\mu}_{1} + \hat{\mu}_{0})$$

$$g_{n}(x) = a_{n}^{T} x + b_{n} = 0$$

By develop the optimal classifier, we find that the optimal classifier has better describe about the data. Our designed LDA classifier should be better as we increase the size of our data, so the mean and covariance will be close to optimal classifier.





As we increase size of sample data, my test set error and formula error both became smaller. When the size of sample data is small, the test set error is bigger than formula error. In the figure above, when size of sample data is more than 65, then the test set error is better than formula error.

#### Assignment 2:

a. This is my training data after pre-processing

```
[[ 6.0000000e-02
                     7.46000000e+00
                                       6.47610000e+01
                                                        8.96000000e+00
   1.82000000e+01
                     3.30000000e+01]
   6.00000000e-02
                     1.33000000e+01
                                       6.72400000e+01
                                                        4.00000000e-01
   1.73000000e+01
                     6.80000000e+011
   1.11800000e+01
                                       1.00000000e-02
                     7.06500000e+01
                                                        1.00000000e-02
   1.81300000e+01
                     1.93000000e+01]
 [ 1.40000000e+01
                     6.96700000e+01
                                       2.00000000e-02
                                                        8.00000000e-02
    1.62000000e+01
                     2.30000000e+01]
   2.60000000e-02
                     1.12000000e+01
                                       6.82300000e+01
                                                        1.18000000e+00
   1.89000000e+01
                     3.47000000e+01]
 [ 7.0000000e-02
                     1.61300000e+01
                                       5.48180000e+01
                                                        9.64000000e+00
                     6.50000000e+01]]
   1.84800000e+01
```

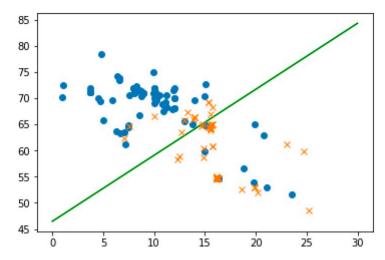
### This is my test data after pre-processing

```
[[ 7.00000000e-03
                     2.30000000e+01
                                       6.11270000e+01
                                                        4.00000000e-02
    1.58000000e+01
                     6.44000000e+01]
  4.00000000e-03
                     1.56000000e+01
                                       6.43170000e+01
                                                        3.00000000e-02
   1.75000000e+01
                     5.06000000e+01]
 [ 2.00000000e-02
                     7.56000000e+00
                                       7.05000000e+01
                                                        7.82000000e+00
                     6.39000000e+00]
    7.88000000e+00
 [ 3.00000000e-02
                     1.63000000e+01
                                       5.46980000e+01
                                                        9.64000000e+00
    1.84800000e+01
                     7.30000000e+01]
  2.60000000e-02
                     1.12000000e+01
                                       6.82300000e+01
                                                        1.18000000e+00
   1.89000000e+01
                     3.45000000e+01]
   3.00000000e-02
                     5.90000000e+00
                                       6.96500000e+01
                                                        7.10000000e+00
   1.62000000e+01
                     1.75000000e+01]]
```

#### b. A

{'Ni': Ttest\_indResult(statistic=2.1832777249025574, pvalue=0.044256486360899815), 'Fe': Ttest\_indResult(statistic=1.4593669470171782, pvalue=0.16371947301978981), 'C': Ttest\_indResult(statistic=-1.4370077964097883, pvalue=0.16872848917102379), 'Mn': Ttest\_indResult(statistic=-0.555911128003719, pvalue=0.58332167105335986), 'Cr': Ttest\_indResult(statistic=-0.36725489505964665, pvalue=0.7164182125651759)}

# c. Classification error with top 2: [0.075757575757576]



## d. A

Classification error with top 3:[0.0694444444444445]

Classification error with top 4:[0.06818181818181818]

Classification error with top 5:[0.06781414141414141]

As we increase number of predictors, the classification error will decrease, because we will have a better classifier.

#### Code:

```
Assignment 1
       #!/usr/bin/env python3
       # -*- coding: utf-8 -*-
       Created on Fri Oct 6 21:32:11 2017
       @author: jianfengsong
       import numpy as np
       import matplotlib.pyplot as plt
       import scipy.integrate as integrate
       import scipy.stats as ns
       import math
       infn=np.inf
       def var(x1,y1):
         h=-np.sqrt(2)*x1*np.sqrt(1+y1)
         return 1/h
       # return -math.pow(h,-1)
       #def snrv(x):
```

```
# return [(2*np.pi)**0.5]*{integrate.quad(np.exp(-(x**2))/2,lambda x :-infn,lambda)}
x:x
x=np.arange(0.001,10.0,0.001)
y=np.arange(0.4,1.0,0.2)
plt.figure(1)
#for a in y:
# for b in x:
plt.plot(x,ns.norm.cdf(var(x,0)),label='0')
plt.plot(x,ns.norm.cdf(var(x,0.4)),'r',label='0.4')
plt.plot(x,ns.norm.cdf(var(x,0.6)),'b',label='0.6')
plt.plot(x,ns.norm.cdf(var(x,0.8)),'g',label='0.8')
plt.plot(x,ns.norm.cdf(var(x,1.0)),'y',label='1.0')
    print(x, var(x, a))
plt.legend()
plt.show()
#print (snrv(1))
c.
   #!/usr/bin/env python3
   # -*- coding: utf-8 -*-
   Created on Fri Oct 6 21:32:11 2017
   @author: jianfengsong
   import numpy as np
   import matplotlib.pyplot as plt
   import scipy.integrate as integrate
   import scipy.stats as ns
   import math
   import sympy as sym
   #import scipy as sym
   x=sym.Symbol('x')
   p=np.array([[1,0.2],[0.2,1]])
   pt=p**-1
   u1=np.array([[1,1]])
   u1t=np.matrix.transpose(u1)
   u0 = np.array([[0,0]])
   u0t=np.matrix.transpose(u0)
   x1ur=0
   x2ur=0
   v1ur=0
```

```
y2ur=0
plt.figure(1)
x11=list()
x2l=list()
sample size=10
x1 = np.random.multivariate normal([1,1], p,sample size)
x2= np.random.multivariate normal([0,0], p,sample size)
 an=np.dot(pt,(u1t-u0t))
 bn=(-1/2)*np.dot(np.dot([np.matrix.transpose(u1t-u0t)],p**-1),(u1t+u0t))
plt.plot(x1[:,0],x1[:,1],'x')
plt.plot(x2[:,0],x2[:,1],'o')
sumx=0
sumy=0
for a in x1:
           sumx=a+sumx
sum1=sumx/sample size #sum1 is mean of u1(1,1)
 for a in x2:
           sumy=a+sumy
sum0=sumy/sample size#sum0 is mean of u0(0,0)
pn1=(np.dot((x1-sum0),T,(x1-sum0))+np.dot((x2-sum1),T,(x2-sum0))+np.dot((x2-sum1),T,(x2-sum0))+np.dot((x2-sum1),T,(x2-sum0))+np.dot((x2-sum1),T,(x2-sum0))+np.dot((x2-sum1),T,(x2-sum0))+np.dot((x2-sum1),T,(x2-sum0))+np.dot((x2-sum1),T,(x2-sum1),T,(x2-sum1))+np.dot((x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-sum1),T,(x2-
 sum1)))/(2*sample size-2)
an1=np.dot(pn1**-1,(sum1-sum0))
bn1=(-1/2)*np.dot(np.dot((sum1-sum0).T,pn1**-1),(sum1-sum0))
print(an1)
print(bn1)
x1=np.arange(-4,5,0.1)
x2p=-bn1/an1[1]-an1[0]*x1/an1[1]
plt.plot(x1,x2p,'b',label='real')
x12=np.arange(-4,5,0.1)
bn2=(-1/2)*np.dot(np.dot((u1t-u0t).T,p**-1),(u1t-u0t))
x22p=-bn2/an[1]-an[0]*x12/an[1]
plt.plot(x12,x22p.T,'r',label='optimal')
plt.legend()
plt.show()
#optimal
\#pn2 = (np.dot((x1-u0.T).T,(x1-u0.T)) + np.dot((x2-u1.T).T,(x2-u0.T)) + np.dot((x2-u1.T).T,(
u1.T))/(2*sample size-2)
\#an2=np.dot(pn2**-1,(u1-u0))
\#bn2 = (-1/2)*np.dot(np.dot((u1-u0).T,pn2**-1),(u1-u0))
##print(an1)
##print(bn1)
\#x12=np.arange(-4,5,0.1)
\#x22p=-bn2/an2[1]-an2[0]*x1/an2[1]
```

```
\#plt.plot(x12,x22p,'r')
d.
        #!/usr/bin/env python3
        # -*- coding: utf-8 -*-
         Created on Sun Oct 8 15:56:58 2017
         @author: jianfengsong
         import numpy as np
         import matplotlib.pyplot as plt
         import scipy.stats as ns
         import math
         from scipy.linalg import det
        p=np.array([[1,0.2],[0.2,1]])
         u0 = np.array([[0,0]])
         u1=np.array([[1,1]])
         sample size=np.array([20,30,40,50])
         LDA error set=list()
         error set=list()
         for a in sample size:
               x1 = np.random.multivariate normal([1,1], p,a)
               x2= np.random.multivariate normal([0,0], p,a)
               x3 = np.random.multivariate normal([1,1],p,250)
               x4= np.random.multivariate normal([0,0],p,250)
               sumx 1=0
               sumx2=0
               for b in x1:
                      sumx1=b+sumx1
               mean x1=sumx1/a
               for c in x2:
                     sumx2=c+sumx2
               mean x0=sumx2/a
               cov=(np.dot((x1-mean x0).T,(x1-mean x0))+np.dot((x2-mean x1).T,(x2-mean x0))+np.dot((x2-mean x0).T,(x2-mean x0))+np.dot((x2-mean x0).T,(x2-mean x0))+np.dot((x2-mean x0).T,(x2-mean x0).T,(x3-mean x0))+np.dot((x3-mean x0).T,(x3-mean x0).T,(x3-mea
        mean x1))/(2*a-2)
               an=np.dot(cov**-1,(mean x1-mean x0))
               bn=(-1/2)*np.dot(np.dot((mean x1-mean x0).T,cov**(-1)),(mean x1-mean x0))
               var x0=(np.dot(an,mean x0.T)+bn)/math.sqrt(np.dot(np.dot(an,p),an.T))
               var x1 = (np.dot(an,mean x1.T) + bn)/math.sqrt(np.dot(np.dot(an,p),an.T))
               LDA error=1/2*(ns.norm.cdf(var x0)+ns.norm.cdf(-var x1))
               LDA error set.append(LDA error)
               clas x3 y=-bn/an[1]-an[0]*x3[:,0]/an[1]
               clas x4 y=-bn/an[1]-an[0]*x4[:,0]/an[1]
               error time=0
```

```
for t in range(250):
               if x3[t,1] < clas x3 y[t]:
                  error time=error time+1
               if x4[t,1] > clas x4 y[t]:
                  error time=error time+1
             error set.append(error time/(500))
           plt.plot(sample size*2,error set,'y',label='formula')
           plt.plot(sample size*2,LDA error set,'r',label='test')
           plt.legend()
           plt.show()
Assignment 2
       a. A
           #!/usr/bin/env python3
           # -*- coding: utf-8 -*-
           Created on Sun Oct 8 20:04:49 2017
           @author: jianfengsong
           import xlrd as xl
           import numpy as np
           import matplotlib.pyplot as plt
           import scipy.integrate as integrate
           import scipy.stats as ns
           import math
           import random
           import xlwt as xlw
           rows value=list()
           cols value=list()
           excel=xl.open workbook('SFE Dataset.xlsx')
           data table=excel.sheet by index(0)
           rows=data table.nrows
           cols=data table.ncols
           for a in range(1,rows,1):
             if data table.row values(a,cols-1)[0]>=45 or data table.row values(a,cols-
           1)[0] <= 35:
               rows value.append(data table.row values(a))
           #for a in range(1,cols,1):
              if data table.col values(cols-1):
                 cols value.append(data table.col values(a))
           #for a in range(rows):
           ele 0=list()
```

```
num 0=0
for a in range(0,cols,1):
  for b in range(0,len(rows value),1):
    if rows_value[b][a]<=0.00000001:
       num 0=num 0+1
  if num 0/len(rows value)>=0.4:
      print(len(rows value))
    num 0=0
    ele 0.append(a)
for row in rows value:
  num del=0
  for col in ele 0:
    col=col-num del
    row.remove(row[col])
    num del=num del+1
num del row=list()
num 0 \text{ row}=0
#delete row value that are 0
for row in rows value:
   print (row)
  for a in row:
#
      print(a)
    if a==0:
       num 0 row=num 0 row+1
  if num 0 \text{ row} > 0:
    num 0 \text{ row}=0
    num del row.append(row)
for a in range(0,len(num del row),1):
  num=0
  a=a-num
  rows value.remove(num_del_row[a])
  num=num+1
#random choose value for test set
numtrain=list()
numtest=list()
test set=list()
train set=list()
status=True
go=0
#num=0
while (status):
  num=0
  go=go+1
  print(go)
```

```
random set=random.sample(range(len(rows value)),len(rows value))
      for a in range(int(len(rows value)*0.2)):
        numtrain.append(random set[a])
      for a in range(len(numtrain)):
        train set.append(rows value[numtrain[a]])
      for a in range(len(rows value)-int(len(rows value)*0.2)):
        numtest.append(random set[a+int(len(rows value)*0.2)])
      for a in range(len(numtest)):
        test set.append(rows value[numtest[a]])
      for row in train set:
        if row[len(row)-1] \le 35:
           num=num+1
      if num/len(train set)>=0.55 or num/len(train set)<=0.45:
        numtrain=list()
        numtest=list()
        test set=list()
        train set=list()
        print("not yet")
   # if num/len(train_set)<0.55 and num/len(train_set)<0.45:
      else:
        status=False
        print("you are good to go")
   data final=xlw.Workbook()
   sheet1=data final.add sheet('sheet1',cell overwrite ok=True)
   for i in range(len(train set)):
      for j in range(len(train set[i])):
        sheet1.write(i,j,train set[i][j])
   data final.save('data final.xls')
   test set1=np.asarray(test set)
   train set0=np.asarray(train_set)
   print(test set1)
   print(train set0)
b. A
   import xlrd as xl
   import numpy as np
   import matplotlib.pyplot as plt
   import scipy.integrate as integrate
   import scipy.stats as ns
   import math
   import random
   rows value=list()
   cols value=list()
   excel=xl.open workbook('SFE Dataset.xlsx')
   data table=excel.sheet by index(0)
```

```
rows=data table.nrows
cols=data table.ncols
for a in range(1,rows,1):
  if data table.row values(a,cols-1)[0]>=45 or data table.row values(a,cols-
1)[0] <= 35:
    rows value.append(data_table.row_values(a))
ele 0=list()
num 0=0
for a in range(0, cols, 1):
  for b in range(0,len(rows value),1):
    if rows value[b][a]\leq=0.00000001:
       num 0=num 0+1
  if num 0/len(rows value)>=0.4:
    num 0=0
     ele 0.append(a)
for row in rows value:
  num del=0
  for col in ele 0:
     col=col-num del
    row.remove(row[col])
    num del=num del+1
num del row=list()
num 0 \text{ row}=0
for row in rows value:
  for a in row:
    if a==0:
       num 0 row=num 0 row+1
  if num 0 \text{ row} > 0:
    num 0 \text{ row}=0
    num del row.append(row)
for a in range(0,len(num del row),1):
  num=0
  a=a-num
  rows value.remove(num del row[a])
  num=num+1
#random choose value for test set
numtrain=list()
numtest=list()
test set=list()
train set=list()
status=True
g_0=0
while (status):
  num=0
```

```
go=go+1
# print(go)
  random set=random.sample(range(len(rows value)),len(rows value))
  for a in range(int(len(rows value)*0.2)):
     numtrain.append(random set[a])
  for a in range(len(numtrain)):
     train set.append(rows value[numtrain[a]])
  for a in range(len(rows value)-int(len(rows value)*0.2)):
     numtest.append(random set[a+int(len(rows value)*0.2)])
  for a in range(len(numtest)):
     test set.append(rows value[numtest[a]])
  for row in train set:
     if row[len(row)-1] \le 35:
       num=num+1
  if num/len(train set)>=0.55 or num/len(train set)<=0.45:
     numtrain=list()
     numtest=list()
     test set=list()
     train set=list()
      print("not yet")
# if num/len(train_set)<0.55 and num/len(train_set)<0.45:
     status=False
      print("you are good to go")
#save excel doc
data final=xlw.Workbook()
sheet1=data final.add sheet('sheet1',cell overwrite ok=True)
for i in range(len(train set)):
  for j in range(len(train set[i])):
     sheet1.write(i,j,train set[i][j])
data final.save('data final.xls')
#Assignment2(b)
train set35=list()
train set45=list()
length train=len(train set)
for a in range(0,length train,1):
  if train set[a][len(train set[a])-1] \le 35:
     train set35.append(a)
  else:
     train set45.append(a)
train35=list()
train45=list()
for a in train set35:
  train35.append(train set[a])
for a in train set45:
```

```
train45.append(train set[a])
   #train 35=list()
   #train 45=list()
   #for a in range(len(train35[1])):
       for b in range(len(train35)):
           train 35.append(train35[b][a])
   #for a in range(len(train45[1])):
       for b in range(len(train45)):
           train 45.append(train45[b][a])
   train 35=[[] for i in range(len(train35[1]))]
   train 45=[[] for i in range(len(train45[1]))]
   Tset=list()
   for a in range(len(train35[1])):
      for b in range(len(train35)):
          train 35[a].append(train35[b][a])
   for a in range(len(train45[1])):
      for b in range(len(train45)):
          train 45[a].append(train45[b][a])
   for a in range(len(train 35)-1):
      h=ns.ttest ind(train 35[a],train 45[a],equal var=False)
      Tset.append(h)
   Tset sta=list()
   for a in range(len(Tset)):
      Tset sta.append(abs(Tset[a][0]))
   #print (Tset sta)
   Tset sta name={'C':Tset sta[0],'Ni':Tset sta[1],'Fe':Tset sta[2],'Mn':Tset sta[3],'Cr':
   Tset sta[4]}
   Tset_name={'C':Tset[0],'Ni':Tset[1],'Fe':Tset[2],'Mn':Tset[3],'Cr':Tset[4]}
   Tset name0=np.asarray(Tset name)
   print(Tset name0)
   import operator
   sorted tset = sorted(Tset sta name.items(), key=operator.itemgetter(1),reverse=True)
   #print(sorted tset)
c. A
   import xlrd as xl
   import numpy as np
   import matplotlib.pyplot as plt
   import scipy.integrate as integrate
   import scipy.stats as ns
   import math
   import random
   import operator
   rows value=list()
```

```
cols value=list()
excel=xl.open workbook('SFE Dataset.xlsx')
data table=excel.sheet by index(0)
rows=data table.nrows
cols=data table.ncols
for a in range(1,rows,1):
  if data table.row values(a,cols-1)[0]>=45 or data table.row values(a,cols-
1)[0] <= 35:
     rows value.append(data table.row values(a))
ele 0=list()
num 0=0
for a in range(0, cols, 1):
  for b in range(0,len(rows_value),1):
    if rows value[b][a]\leq=0.00000001:
       num 0=num 0+1
  if num 0/len(rows value)>=0.4:
    num 0=0
     ele 0.append(a)
for row in rows value:
  num del=0
  for col in ele 0:
     col=col-num del
    row.remove(row[col])
    num del=num del+1
num del row=list()
num 0 \text{ row}=0
for row in rows value:
  for a in row:
    if a==0:
       num 0 row=num 0 row+1
  if num 0 \text{ row} > 0:
    num 0 \text{ row}=0
    num del row.append(row)
for a in range(0,len(num_del_row),1):
  num=0
  a=a-num
  rows value.remove(num del row[a])
  num=num+1
#random choose value for test set
numtrain=list()
numtest=list()
test set=list()
train set=list()
status=True
```

```
go=0
while (status):
  num=0
  go=go+1
# print(go)
  random set=random.sample(range(len(rows value)),len(rows value))
  for a in range(int(len(rows value)*0.2)):
     numtrain.append(random set[a])
  for a in range(len(numtrain)):
     train set.append(rows value[numtrain[a]])
  for a in range(len(rows value)-int(len(rows value)*0.2)):
     numtest.append(random set[a+int(len(rows value)*0.2)])
  for a in range(len(numtest)):
     test set.append(rows value[numtest[a]])
  for row in train set:
     if row[len(row)-1]<=35:
       num=num+1
  if num/len(train set)>=0.55 or num/len(train set)<=0.45:
     numtrain=list()
     numtest=list()
     test set=list()
     train set=list()
  else:
     status=False
#save excel doc
#data final=xlw.Workbook()
#sheet1=data final.add sheet('sheet1',cell overwrite ok=True)
#for i in range(len(train set)):
   for j in range(len(train set[i])):
      sheet1.write(i,j,train set[i][j])
#data final.save('data final.xls')
#Assignment2(b)
train set35=list()
train set45=list()
length train=len(train set)
for a in range(0,length train,1):
  if train set[a][len(train set[a])-1] \le 35:
     train set35.append(a)
  else:
     train set45.append(a)
train35=list()
train45=list()
for a in train set35:
  train35.append(train set[a])
for a in train set45:
```

```
train45.append(train set[a])
train 35=[[] for i in range(len(train35[1]))]
train 45=[[] for i in range(len(train45[1]))]
Tset=list()
for a in range(len(train35[1])):
  for b in range(len(train35)):
      train 35[a].append(train35[b][a])
for a in range(len(train45[1])):
  for b in range(len(train45)):
      train 45[a].append(train45[b][a])
for a in range(len(train 35)-1):
  h=ns.ttest ind(train 35[a],train 45[a],equal var=False)
  Tset.append(h)
Tset sta=list()
for a in range(len(Tset)):
  Tset sta.append(abs(Tset[a][0]))
#print (Tset sta)
Tset sta name={'C':Tset sta[0],'Ni':Tset sta[1],'Fe':Tset sta[2],'Mn':Tset sta[3],'Cr':
Tset sta[4]}
sorted tset = sorted(Tset sta name.items(), key=operator.itemgetter(1),reverse=True)
print(sorted tset)
#Assignment2(c)
top=list()
#for a in range(len(sorted tset)):
# top.append(sorted tset[a][1])
\#x1=[[] for i in range(len(train35))]
\#x2=[[] for i in range(len(train45))]
x11=list()
x21=list()
#for a in range(len(x1)):
#for a in range(len(train35)):
first=Tset sta.index(sorted tset[0][1])
second=Tset sta.index(sorted tset[1][1])
for b in range(len(train35)):
  x11.append([train35[b][1],train35[b][2]])
for b in range(len(train45)):
  x21.append([train45[b][1],train45[b][2]])
###different
#for b in range(len(train35)):
# x11.append([train35[b][first],train35[b][second]])
#for b in range(len(train45)):
   x21.append([train45[b][first],train45[b][second]])
sumx1=0
sum x 2 = [0,0]
x1=np.asarray(x11)
```

```
x2=np.asarray(x21)
for b in x1:
       sumx1=b+sumx1
mean_x1=sumx1/len(x1)
for c in x2:
       sumx2=c+sumx2
mean x0=sumx2/len(x2)
cov = (np.dot((x1-mean x0).T,(x1-mean x0))+np.dot((x2-mean x1).T,(x2-mean x0))+np.dot((x2-mean x0).T,(x2-mean x0))+np.dot((x2-mean x0).T,(x2-mean x0).T,(x3-mean x0))+np.dot((x3-mean x0).T,(x3-mean x0
mean x1))/(min(len(train35),len(train45)-2))
an=np.dot(cov^{**}-1,(mean x1-mean x0))
bn=(-1/2)*np.dot(np.dot((mean x1-mean x0).T,cov**(-1)),(mean x1-mean x0))
plt.figure(1)
plt.plot(x1[:,0],x1[:,1],'x')
plt.plot(x2[:,0],x2[:,1],'o')
x1=np.arange(-10,100,1)
x2p = -bn/an[1] - an[0] * x1/an[1]
plt.plot(x1,x2p.T,'r')
#test set
test set35=list()
test set45=list()
length test=len(test set)
for a in range(0,length test,1):
       if test set[a][len(test set[a])-1] \le 35:
             test set35.append(a)
       else:
             test set45.append(a)
test35=list()
test45=list()
for a in test set35:
       test35.append(test_set[a])
for a in test set45:
       test45.append(test_set[a])
test 35=[[] for i in range(len(test35[1]))]
test 45=[[] for i in range(len(test45[1]))]
#Tset=list()
x112=list()
x212=list()
for a in range(len(test35[1])):
       for b in range(len(test35)):
                 test 35[a].append(test35[b][a])
for a in range(len(test45[1])):
       for b in range(len(test45)):
                 test 45[a].append(test45[b][a])
for b in range(len(test35)):
       x112.append([test35[b][1],test35[b][2]])
```

```
for b in range(len(test45)):
             x212.append([test45[b][1],test45[b][2]])
       sumx12=0
        sumx22 = [0,0]
       x12=np.asarray(x112)
       x22=np.asarray(x212)
       for b in x12:
             sumx12=b+sumx12
       mean x12=sumx12/len(x12)
       for c in x22:
             sumx22=c+sumx22
       mean x02=sumx22/len(x22)
        cov2 = (np.dot((x12-mean x02),T,(x12-mean x02)) + np.dot((x22-mean x12),T,(x22-mean x12),T,(x22-mean x12),T,(x22-mean x12),T,(x22-mean x12),T,(x22-mean x12),T,(x22-mean x12),T,(x12-mean x12),
       mean x12))/(min(len(test35),len(test45)-2))
        an2=np.dot(cov2**-1,(mean x12-mean x02))
       bn2=(-1/2)*np.dot(np.dot((mean x12-mean x02).T,cov2**(-1)),(mean x12-mean x02).T,cov2**(-1))
        mean x02)
        var x0=(np.dot(an2,mean x02.T)+bn2)/math.sqrt(np.dot(np.dot(an2,cov2),an2.T))
        var x1=(np.dot(an2,mean x12.T)+bn2)/math.sqrt(np.dot(np.dot(an2,cov2),an2.T))
        LDA error=1/2*(ns.norm.cdf(var x0)+ns.norm.cdf(-var x1))
        clas x12 y=-bn2/an2[1]-an2[0]*x12[:,0]/an[1]
        clas x22 v=-bn/an[1]-an[0]*x22[:,0]/an[1]
        error time=0
        error set=list()
        for t in range(min(len(test45),len(test35))):
             if x22[t,1] > clas x22 y[t]:
                   error time=error time+1
             if x12[t,1] < clas x12 y[t]:
                  error time=error time+1
        error set.append(error time/(len(test set)*len(test set[1])))
       print(error set)
       #print(LDA error)
       #LDA error set.append(LDA error)
d. For part d we just need to change two variable to calculate the result that we want.
        With 3 predictors:
              for b in range(len(test35)):
                   x112.append([test35[b][1],test35[b][2],test35[b][3]])
              for b in range(len(test45)):
                    x212.append([test45[b][1],test45[b][2],test45[b][3]])
        With 4 predictors:
             for b in range(len(test35)):
                  x112.append([test35[b][1],test35[b][2],test35[b][3],test35[b][0]])
             for b in range(len(test45)):
                  x212.append([test45[b][1],test45[b][2],test45[b][3],test45[b][0]])
```

```
With 5 predictors:
for b in range(len(test35)):
    x112.append([test35[b][1],test35[b][2],test35[b][3],test35[b][0],test35[b][4]])
for b in range(len(test45)):
    x212.append([test45[b][1],test45[b][2],test45[b][3],test45[b][0],test35[b][4]])
```