Assignment 2

• Donwload the credit dataset 'Credit.csv' from https://github.com/vahidpartovinia/ycbs255/

Setup:

```
In [12]:
```

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

path = "Desktop/"
filename = "Credit.csv"
data = pd.read_csv(path+filename)

data.head()
```

Out[12]:

	Unnamed:	Income	Limit	Rating	Cards	Age	Education	Gender	Student	Married
0	1	14.891	3606	283	2	34	11	Male	No	Yes
1	2	106.025	6645	483	3	82	15	Female	Yes	Yes
2	3	104.593	7075	514	4	71	11	Male	No	No
3	4	148.924	9504	681	3	36	11	Female	No	No
4	5	55.882	4897	357	2	68	16	Male	No	Yes

Create the attribute matrix 'X' as a subset of the data that includes only the qualitative attributes 'Income', 'Limit', 'Rating', 'Cards', 'Age', and 'Education'

```
In [13]:
```

```
X = data[['Income', 'Limit', 'Rating','Cards','Age','Education']]
X.head()
```

Out[13]:

	Income	Limit	Rating	Cards	Age	Education
0	14.891	3606	283	2	34	11
1	106.025	6645	483	3	82	15
2	104.593	7075	514	4	71	11
3	148.924	9504	681	3	36	11
4	55.882	4897	357	2	68	16

Create a discrete response variable, say 'y' by transforming 'Balance' to a binary output, which equals 1 if 'Balance > 1500' and equals 0 otherwise.

In [21]:

```
#To add y column to the original table
data['y'] = np.where(data['Balance']>1500, 1, 0)
data.head()
```

Out[21]:

	Unnamed: 0	Income	Limit	Rating	Cards	Age	Education	Gender	Student	Married
0	1	14.891	3606	283	2	34	11	Male	No	Yes
1	2	106.025	6645	483	3	82	15	Female	Yes	Yes
2	3	104.593	7075	514	4	71	11	Male	No	No
3	4	148.924	9504	681	3	36	11	Female	No	No
4	5	55.882	4897	357	2	68	16	Male	No	Yes

Fit 1- logistic regression, 2- linear discriminant, and 3- quadratic discriminant on the binary Balance as the output variable. Use 'Income', 'Limit', 'Rating', 'Cards', 'Age', and 'Education' as the input variables.

```
In [39]:
#Logistic regression:
from sklearn.linear model import LogisticRegression
from sklearn import metrics
predicators = data[['Income', 'Limit', 'Rating', 'Cards', 'Age', 'Education']]
prediction = data["y"]
logreg = LogisticRegression()
logreg.fit(predicators, prediction)
Out[39]:
LogisticRegression(C=1.0, class weight=None, dual=False, fit interce
pt=True,
          intercept scaling=1, max iter=100, multi class='ovr', n jo
bs=1,
          penalty='12', random state=None, solver='liblinear', tol=0
.0001,
          verbose=0, warm start=False)
In [55]:
#linear discriminant
from sklearn.discriminant analysis import LinearDiscriminantAnalysis
lda = LinearDiscriminantAnalysis()
lda.fit(predicators, prediction)
Out[55]:
LinearDiscriminantAnalysis(n components=None, priors=None, shrinkage
=None,
              solver='svd', store covariance=False, tol=0.0001)
In [56]:
#quadratic discriminant
from sklearn.discriminant_analysis import QuadraticDiscriminantAnalysis
qda = QuadraticDiscriminantAnalysis()
qda.fit(predicators, prediction)
Out[56]:
QuadraticDiscriminantAnalysis(priors=None, reg param=0.0,
               store covariance=False, store covariances=None, tol=0
.0001)
```

Find the probability of (Balance > 1500), for x1 and x2 below, using these three methods. Compare these probabilities and comment.

```
x1= 'Income' = 63, 'Limit' = 8100, 'Rating' = 600, 'Cards' = 4, 'Age' = 30, 'Education' = 14
x2= 'Income' = 186, 'Limit' = 13414, 'Rating' = 950, 'Cards' = 2, 'Age' = 41, 'Education' = 13
```

In [59]:

```
#Linear regression
linear reg pred = np.array([[63,8100,600,4,30,14],[186,13414,950,2,41,13]])
print(linear reg pred)
print(logreg.predict proba(linear reg pred))
print(logreg.predict(linear reg pred))
                600
                         4
                              30
     63
         8100
                                     14]
11
    186 13414
                950
                         2
                              41
                                    13]]
 ſ
[[ 0.93231523  0.06768477]
 [ 0.17298983  0.82701017]]
```

The probability that Balance > 1500 for x1 is 0.06768477, and for x2 is 0.06768477

In [60]:

[0 1]

```
#Linear discriminant
print(lda.predict_proba(linear_reg_pred))
print(lda.predict(linear_reg_pred))

[[ 0.94144572     0.05855428]
        [ 0.00721199     0.99278801]]
```

The probability that Balance > 1500 for x1 is 0.05855428, and for x2 is 0.99278801

In [62]:

[0 1]

[0 1]

```
#quadratic discriminant
print(qda.predict_proba(linear_reg_pred))
print(qda.predict(linear_reg_pred))

[[ 9.9999999e-01  1.24419207e-09]
  [ 7.83057752e-04  9.99216942e-01]]
```

The probability that Balance > 1500 for x1 is 1.24419207e-09, and for x2 is 9.99216942e-01