```
data = pd.read csv("/Users/user/documents/Class/BMEN415/GitHub/BMEN415FinalProject/Dat
 #Converted the group to dummy variables
 data.loc[data.Group=='Nondemented', 'Group'] = 0
 data.loc[data.Group=='Demented', 'Group'] = 1
 data.loc[data.Group=='Converted', 'Group'] = 2
 #Converted the M/F to dummy variables
 data.loc[data.Sex=='M', 'Sex'] = 0
 data.loc[data.Sex=='F', 'Sex'] = 1
 #Fill in SES and Mini Mental State missing sample values
 #Use the mean of that category for the value
 # Get rid of SES data
 data
                                                                                  Mini
                                                                                         Clinica
                                               MR
      Subject ID
                          MRI ID Group Visit
                                                    Sex Hand Age EDUC SES Mental Dementi-
                                             Delay
                                                                                 State
                                                                                          Rating
  0 OAS2_0001 OAS2_0001_MR1
                                     0
                                           1
                                                 0
                                                      0
                                                            R
                                                                87
                                                                       14
                                                                            2.0
                                                                                  27.0
                                                                                             0.1
   1 OAS2_0001 OAS2_0001_MR2
                                     0
                                           2
                                                      0
                                                                            2.0
                                                                                  30.0
                                               457
                                                            R
                                                                88
                                                                       14
                                                                                             0.1
   2 OAS2_0002
                 OAS2_0002_MR1
                                     1
                                           1
                                                 0
                                                      0
                                                            R
                                                                75
                                                                       12 NaN
                                                                                  23.0
                                                                                             0.
   3 OAS2_0002 OAS2_0002_MR2
                                           2
                                               560
                                                                76
                                                                          NaN
                                                                                  28.0
                                                      0
                                                            R
                                                                       12
                                                                                             0.
   4 OAS2_0002 OAS2_0002_MR3
                                                                       12 NaN
                                     1
                                           3
                                              1895
                                                      0
                                                            R
                                                                80
                                                                                  22.0
                                                                                             0.
                 OAS2_0185_MR2
368 OAS2_0185
                                     1
                                           2
                                               842
                                                      0
                                                            R
                                                                82
                                                                       16
                                                                            1.0
                                                                                  28.0
                                                                                             0.
                                              2297
369 OAS2_0185
                OAS2_0185_MR3
                                                                                  26.0
                                                            R
                                                                86
                                                                       16
                                                                            1.0
                                                                                             0.
370 OAS2_0186
                 OAS2_0186_MR1
                                     0
                                           1
                                                 0
                                                      1
                                                            R
                                                                61
                                                                       13
                                                                            2.0
                                                                                  30.0
                                                                                             ٥.١
                                           2
                                               763
 371 OAS2_0186
                 OAS2_0186_MR2
                                                            R
                                                                63
                                                                            2.0
                                                                                  30.0
                                                                                             0.
                                                      1
                                                                       13
                                              1608
 372 OAS2_0186
                 OAS2_0186_MR3
                                     0
                                           3
                                                      1
                                                            R
                                                                65
                                                                       13
                                                                            2.0
                                                                                  30.0
                                                                                             0.1
373 rows × 15 columns
 import numpy as np
```

## #It is possible to use categorical and continuous predictors

In [4]:

SES

Out[8]: MR Delay

Sex

Age

10

EDUC SES

Mini Mental State

Atlas Scaling Factor Name: 355, dtype: object

data drop.loc[354]

Mini Mental State

Atlas Scaling Factor

Clinical Dementia Rating

Name: 354, dtype: object

Clinical Dementia Rating

Estimated total Intracranial Volume

Normalize Whole Brain Volume

Estimated total Intracranial Volume

In [9]: explained\_variance = pca.explained\_variance\_ratio\_

Normalize Whole Brain Volume

print(len(explained\_variance)) print(explained\_variance)

from sklearn.model selection import train test split

#PCA might be a good technique to select predictors

#dummy variables for the binary predictors.

#Variables that we will need to deal with:

# Hand, Visit, Subject ID, MRI ID

#Attempting PCA on data

#note that PCA performs best when data is normalized (range b/w 0 and 1)

data drop = data.drop(['Hand','Visit','Subject ID','MRI ID'], axis = 1) #axis = 1 mean

#get rid of row 360 and 359 bc they are missing alot of data (both SES and MMS)

#for a regression problem. My understanding is you need to make

#Hand is completely useless as it is identical for all samples

import pandas as pd

from sklearn.metrics import mean squared error

```
data_drop = data_drop.drop([360, 359])
         #delete all data points that dont have SES in them (this is where they have NaN)
         data_drop = data_drop.dropna()
         #dementia status is what we want to predict - change this to single target
         group = data drop[['Group']]
         data drop = data drop.drop(['Group'], axis = 1) #axis = 1 means to drop column not rot
        # give it a label type
         group = group.astype('int')
         #get a list of columns in pandas object
         names of data = data drop.columns.tolist()
         #shuffle = false prevents data split being different everytime
         X train, X test, y train, y test = train test split(data drop, group, test size=0.2, s
         #split test into validate and test, again making sure the data is always the same for
         X_train, X_val, y_train, y_val = train_test split(X train, y train, test size=0.05, sl
         #Normalizing the data
         from sklearn.preprocessing import StandardScaler
         sc = StandardScaler()
         X train = sc.fit transform(X train)
         X test = sc.transform(X test)
         #running the actual PCA
         from sklearn.decomposition import PCA
         pca = PCA()
         X train = pca.fit transform(X train)
         X test = pca.transform(X test)
         #relief f algorithm - sorting features
In [7]: data_drop.loc[355]
                                                  652
Out[7]: MR Delay
                                                   0
        Sex
                                                   81
        Age
        EDUC
                                                   20
```

1 26

> 0 0

79

20

1

26

0.5

1548

0.711

1.134

[0.28391839 0.21461209 0.15304569 0.1266165 0.08687837 0.05114889

0.5

1556

0.691 1.128

## from sklearn.neural\_network import MLPClassifier MLPAlz = MLPClassifier(random\_state = 1, max\_iter = 100000, hidden\_layer\_sizes=(256,51 history = MLPAlz.fit(X train, y train.values.ravel())

From now on is Multi Layer Perceptron

0.03240389 0.02838197 0.02191061 0.00108361]

predictedMLPAlz	<pre>= MLPAlz.predict(X_test)</pre>	
predictedMLPAlz		

accuracy\_score(y\_test, predictedMLPAlz)

print(classification\_report(y\_test, predictedMLPAlz)) results = classification\_report(y\_test, predictedMLPAlz)

In [ ]: from sklearn.metrics import classification report, confusion matrix, accuracy score print(confusion\_matrix(y\_test, predictedMLPAlz))