Hw1_2

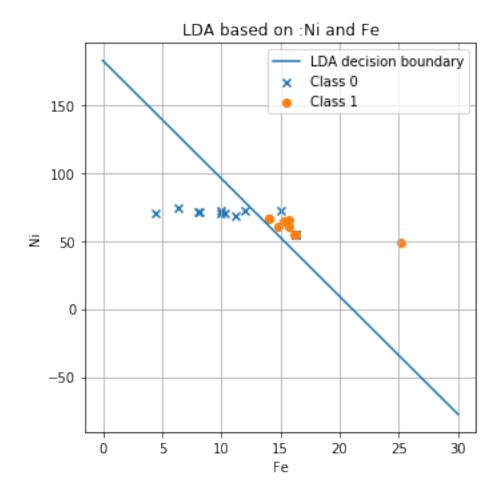
October 16, 2018

```
In [24]: import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         #import excel table
         SFE_data=pd.read_excel('SFE_Dataset.xlsx')
         col=SFE_data.columns
         print(col)
         #preprocessing
         #remove columns with zero elements
         nonzero_ratio=np.zeros(len(col))
         counter=0
         for col_i in col:
             n_zeros=0
             #print('column : ',col_i)
             #import pdb; pdb.set trace()
             for j in range(0,len(SFE_data[col_i])):
                 if(SFE_data[col_i][j]==0):
                     n_zeros+=1
             nonzero_ratio[counter]=1-n_zeros/len(SFE_data[col_i])
             counter+=1
         print('non zero ratio : ',nonzero_ratio)
         drop_cols=[]
         for i in range(0,len(SFE_data.columns)-1):
             if (nonzero_ratio[i]<.6):</pre>
                 #print('non zero ratio',nonzero_ratio[i])
                 drop_cols.append(col[i])
         SFE_data_cleancols=SFE_data.drop(drop_cols,axis=1)
         new_col=SFE_data_cleancols.columns
         print(len(new col),len(col))
         #import pdb; pdb.set_trace()
```

```
Index(['C', 'N', 'P', 'S', 'V', 'Ni', 'Nb', 'Al', 'Ti', 'Fe', 'Hf', 'Mo', 'Mn',
       'Co', 'Si', 'Cr', 'Cu', 'SFE'],
      dtype='object')
non zero ratio: [0.89006342 0.60887949 0.36997886 0.40169133 0.00422833 0.75052854
0.01268499 0.09513742 0.01479915 1.
                                             0.00211416 0.39957717
0.80338266 0.00634249 0.60887949 0.8372093 0.05496829 1.
                                                                   1
8 18
In [25]: #SFE into high and low, remove rows
         #high and low SFE
         SFE=SFE_data_cleancols['SFE']
         non SFE=[]
         for i in range(0,len(SFE)):
             if(SFE[i] <= 35):</pre>
                 SFE_data_cleancols['SFE'][i]=0
             elif(SFE[i]>=45):
                 SFE_data_cleancols['SFE'][i]=1
             elif(SFE[i]<45 and SFE[i]>35):
                 non SFE.append(i)
         SFE_data_cleanrows=SFE_data_cleancols.drop(non_SFE)
         SFE=SFE data cleanrows['SFE']
         #print(SFE)
         #SFE data cleanrows.head
         #SFE data cleancols.head
         #import pdb; pdb.set trace()
         # for resetting the index after removing the columns or rows
         SFE_data_cleanrows=SFE_data_cleanrows.reset_index(drop=True)
         [m,n]=SFE_data_cleanrows.shape
         col=SFE_data_cleanrows.columns
         #clean rows that have any zeros
         zer=[]
         ncol=len(col)
         for i in range(0,m):
             #import pdb; pdb.set_trace()
             n zeros=0
             for j in range (0,ncol-1):
                 if(SFE data cleanrows.at[i, col[j]]==0):
                     n zeros+=1
             if(n_zeros>0):
                 zer.append(i)
         #import pdb; pdb.set_trace()
```

```
In [26]: SFE_data_cleanrows2=SFE_data_cleanrows.drop(zer)
                    # for resetting the index after removing the columns or rows
                    SFE_data_cleanrows2=SFE_data_cleanrows2.reset_index(drop=True)
In [27]: # randomly sample into training data and then reject data with over 55% of any one
                    [m,n]=SFE_data_cleanrows2.shape
                    import random
                    import time
                    random.seed(time.time())
                    okay_sample_flag=0
                    while(okay_sample_flag==0):
                             train_set=SFE_data_cleanrows2.sample(frac=0.2,random_state=random.randint(1,100))
                             test_set=SFE_data_cleanrows2.drop(train_set.index)
                             train_set=train_set.reset_index(drop=True)
                             test_set=test_set.reset_index(drop=True)
                             #remove samples for more than 55%
                             n_ones=0
                             for i in range(0,len(train_set)):
                                      if (train_set.at[i,'SFE']==1):
                                               n ones += 1
                             if(n_ones/len(train_set)>.45 and n_ones/len(train_set)<.55):</pre>
                                      okay_sample_flag+=1
                               import pdb; pdb.set_trace()
In [36]: train_set.head()
Out [36]:
                                        C
                                                                                                                                                 Cr SFE
                                                         N
                                                                       Ni
                                                                                              Fe
                                                                                                               Mn
                                                                                                                                  Si
                    0 0.04100 0.0540
                                                                8.10 70.87500 1.7100 0.3300 18.20 0.0
                    1 0.00400 0.0030 15.60 64.31700 0.0300 0.0200 17.50 1.0
                    2 0.07000 0.5400 16.13 54.67800 9.6400 0.4500 18.48 1.0
                    3 0.40000 0.0660 16.30 54.66000 9.6400 0.4500 18.48 0.0
                    4 0.00004 0.0001 12.00 71.99952 0.0001 0.0002 16.00 0.0
In [28]: from scipy import stats
                    col=train_set.columns
                    ttest=[]
                    for i in range (0,len(col)-1):
                             ttest.append([stats.ttest_ind(train_set[train_set['SFE'] == 0] [col[i]], train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[train_set[tra
In [29]: ttest_sorted=sorted(ttest,key=lambda x:x[:][0][1])
                    print(ttest_sorted[0][1])
                    col1=ttest_sorted[0][1]
                    col2=ttest_sorted[1][1]
                    col3=ttest_sorted[2][1]
                    col4=ttest_sorted[3][1]
                    col5=ttest_sorted[4][1]
```

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In [50]: for i in range (0,len(ttest_sorted)):
             print(ttest_sorted[i][0:3])
[Ttest_indResult(statistic=-3.6652244046944773, pvalue=0.0013486369196097205), 'Ni']
[Ttest indResult(statistic=2.359906793878512, pvalue=0.027438300016957032), 'Fe']
[Ttest_indResult(statistic=2.1848241668704795, pvalue=0.04179478729923675), 'Si']
[Ttest_indResult(statistic=1.4943828075999175, pvalue=0.15926508688174165), 'C']
[Ttest_indResult(statistic=-1.0278205792662476, pvalue=0.31477403802244214), 'Cr']
[Ttest_indResult(statistic=0.80913241119481, pvalue=0.4267305733623522), 'Mn']
[Ttest_indResult(statistic=0.2829267844867566, pvalue=0.7800377202267232), 'N']
In [121]: train_set_0=train_set[train_set['SFE']==0]
          train_set_1=train_set[train_set['SFE']==1]
          train_set_0=train_set_0.reset_index(drop=True)
          train_set_1=train_set_1.reset_index(drop=True)
          smean1=np.array([float(train_set_0.mean(0)[col1]),float(train_set_0.mean(0)[col2])])
          smean2=np.array([float(train_set_1.mean(0)[col1]),float(train_set_1.mean(0)[col2])])
          cov=train_set.cov()
          #import pdb; pdb.set_trace()
          scov=np.zeros((2,2))
          scov=[[float(cov[col1][col1]),float(cov[col1][col2])],[float(cov[col2][col1]),float(
          scov_inv=np.linalg.inv(scov)
          a_lda=scov_inv.dot((smean2-smean1))
          b_lda=-0.5*((smean2-smean1).dot(scov_inv)).dot((smean1+smean2))
          x1_lda=np.linspace(0,30,num=20)
          y_1da=-a_1da[0]/a_1da[1]*x1_1da-b_1da/a_1da[1]
In [74]: fig, ax = plt.subplots(figsize=[5,5])
         plt.scatter(train_set_0[col1],train_set_0[col2],marker='x',label='Class 0')
         plt.scatter(train_set_1[col1],train_set_1[col2],marker='o',label='Class 1')
         plt.plot(x1_lda,y_lda,label='LDA decision boundary')
         plt.title('LDA based on :'+ col1 +' and '+ col2)
         plt.ylabel(col1)
         plt.xlabel(col2)
         fig.tight_layout()
         ax.legend()
         plt.grid(True)
         plt.show
         fig.savefig('hw2c.png')
```



```
In [122]: #estimating error on test set
                                       test_set_0=test_set[test_set['SFE']==0]
                                       test_set_1=test_set[test_set['SFE']==1]
                                       test_set_0=test_set_0.reset_index(drop=True)
                                       test_set_1=test_set_1.reset_index(drop=True)
                                       errtest_set_0=np.zeros((len(test_set_0),2))
                                       errtest_set_1=np.zeros((len(test_set_1),2))
                                       for i in range (0,len(test_set_0)):
                                                       errtest_set_0[i][0]=test_set_0[col1][i]
                                                       errtest_set_0[i][1]=test_set_0[col2][i]
                                       for i in range (0,len(test_set_1)):
                                                       errtest_set_1[i][0]=test_set_0[col1][i]
                                                       errtest_set_1[i][1]=test_set_0[col2][i]
                                                       \#[test\_set[test\_set['SFE'] == 0][col1], test\_set[test\_set['SFE'] == 0][col2]]
                                                       \#test\_set\_1[i] = [test\_set[test\_set['SFE'] == 1][col1], test\_set[test\_set['SFE'] == 1][test\_set[test\_set['SFE'] == 1][test\_set['SFE'] == 1][test\_set['SF
                                        #import pdb; pdb.set_trace()
```

```
TP=0
          FP=0
          TN=0
          FN=0
          g1 = 0
          g2 = 0
          for i in range(0,len(errtest_set_0)):
              g1=a_lda.dot(errtest_set_0[i])+b_lda
              if (g1>0):
                  err1+=1
                  FN+=1
              if (g1<=0):
                  TN+=1
          err2=0
          for i in range(0,len(errtest_set_1)):
              g2=a_lda.dot(errtest_set_1[i])+b_lda
              if (g2<=0):
                  err2+=1
                  FP+=1
              if (g2>0):
                  TP+=1
          err=(err1+err2)/(len(errtest_set_0)+len(errtest_set_1))
          conf_mat=np.array(([TP,FP],[FN,TN]))
          print(err)
0.49504950495049505
In [105]: err
          print(conf_mat)
[[ 4 46]
[ 4 47]]
In [107]: pd.DataFrame(conf_mat,[' Predicted 1','Predicted 0'],['Actual 1','Actual 0'])
Out[107]:
                        Actual 1 Actual 0
           Predicted 1
                               4
                                         46
          Predicted 0
                               4
                                         47
In [119]: #now repeat for the top 3,4 and 5 predictors
          #first we do top 3 predictors
          train_set_0=train_set[train_set['SFE']==0]
```

err1=0

```
train_set_1=train_set[train_set['SFE']==1]
          train_set_0=train_set_0.reset_index(drop=True)
          train_set_1=train_set_1.reset_index(drop=True)
          #clean train set to only contain columns of our interest
          train set clean=train set[[col1,col2,col3]]
          train_set_0_clean=train_set_0[[col1,col2,col3]]
          train_set_1_clean=train_set_1[[col1,col2,col3]]
          smean1=np.array(train_set_0_clean.mean(0))
          smean2=np.array(train_set_1_clean.mean(0))
          cov=train_set_clean.cov()
          #import pdb; pdb.set_trace()
          cov_inv=np.linalg.inv(cov)
          a_lda=cov_inv.dot((smean2-smean1))
          b_lda=-0.5*((smean2-smean1).dot(cov_inv)).dot((smean1+smean2))
          x1_lda=np.linspace(0,40,num=20)
          y_1da=-a_1da[0]/a_1da[1]*x1_1da-b_1da/a_1da[1]
In [120]: #estimating error on test set
          test set 0=test set[test set['SFE']==0]
          test_set_1=test_set[test_set['SFE']==1]
          test_set_0=test_set_0.reset_index(drop=True)
          test_set_1=test_set_1.reset_index(drop=True)
          #consider only predictors of our importance
          test_set_clean=test_set[[col1,col2,col3]]
          test_set_0_clean=test_set_0[[col1,col2,col3]]
          test_set_1_clean=test_set_1[[col1,col2,col3]]
          errtest_set_0=np.array(test_set_0_clean)
          errtest_set_1=np.array(test_set_1_clean)
          #import pdb; pdb.set_trace()
          err1=0
          TP=0
          FP=0
          TN=0
          FN=0
          g1 = 0
          g2 = 0
          for i in range(0,len(errtest_set_0)):
              g1=a_lda.dot(errtest_set_0[i])+b_lda
```

```
if (g1>0):
                  err1+=1
                  FN+=1
              if (g1<=0):
                  TN+=1
          err2=0
          for i in range(0,len(errtest set 1)):
              g2=a_lda.dot(errtest_set_1[i])+b_lda
              if (g2 \le 0):
                  err2+=1
                  FP+=1
              if (g2>0):
                  TP+=1
          err=(err1+err2)/(len(errtest_set_0)+len(errtest_set_1))
          print(err)
          conf_mat=np.array(([TP,FP],[FN,TN]))
0.12871287128712872
In [110]: pd.DataFrame(conf_mat,[' Predicted 1','Predicted 0'],['Actual 1','Actual 0'])
Out[110]:
                        Actual 1 Actual 0
           Predicted 1
                              40
                                        10
                               3
          Predicted 0
                                        48
In [117]: #now we do for top 4 predictors
          train_set_0=train_set[train_set['SFE']==0]
          train set 1=train set[train set['SFE']==1]
          train_set_0=train_set_0.reset_index(drop=True)
          train_set_1=train_set_1.reset_index(drop=True)
          #clean train set to only contain columns of our interest
          train_set_clean=train_set[[col1,col2,col3,col4]]
          train_set_0_clean=train_set_0[[col1,col2,col3,col4]]
          train_set_1_clean=train_set_1[[col1,col2,col3,col4]]
          smean1=np.array(train_set_0_clean.mean(0))
          smean2=np.array(train_set_1_clean.mean(0))
          cov=train set clean.cov()
          #import pdb; pdb.set_trace()
          cov inv=np.linalg.inv(cov)
          a_lda=cov_inv.dot((smean2-smean1))
          b_lda=-0.5*((smean2-smean1).dot(cov_inv)).dot((smean1+smean2))
```

```
x1_lda=np.linspace(0,40,num=20)
          y_lda=-a_lda[0]/a_lda[1]*x1_lda-b_lda/a_lda[1]
In [118]: #estimating error on test set
          test_set_0=test_set[test_set['SFE']==0]
          test_set_1=test_set[test_set['SFE']==1]
          test_set_0=test_set_0.reset_index(drop=True)
          test_set_1=test_set_1.reset_index(drop=True)
          #consider only predictors of our importance
          test_set_clean=test_set[[col1,col2,col3,col4]]
          test_set_0_clean=test_set_0[[col1,col2,col3,col4]]
          test_set_1_clean=test_set_1[[col1,col2,col3,col4]]
          errtest_set_0=np.array(test_set_0_clean)
          errtest_set_1=np.array(test_set_1_clean)
          #import pdb; pdb.set_trace()
          err1=0
          TP=0
          FP=0
          TN=0
          FN=0
          g1 = 0
          g2=0
          for i in range(0,len(errtest_set_0)):
              g1=a_lda.dot(errtest_set_0[i])+b_lda
              if (g1>0):
                  err1+=1
                  FN+=1
              if (g1 <= 0):
                  TN+=1
          err2=0
          for i in range(0,len(errtest_set_1)):
              g2=a_lda.dot(errtest_set_1[i])+b_lda
              if (g2 <= 0):
                  err2+=1
                  FP+=1
              if (g2>0):
                  TP+=1
          err=(err1+err2)/(len(errtest_set_0)+len(errtest_set_1))
          conf_mat=np.array(([TP,FP],[FN,TN]))
          print(err)
```

```
In [113]: pd.DataFrame(conf_mat,[' Predicted 1','Predicted 0'],['Actual 1','Actual 0'])
Out[113]:
                        Actual 1 Actual 0
           Predicted 1
                              44
          Predicted 0
                               3
                                        48
In [114]: #now we do for top 5 predictors
          train set 0=train set[train set['SFE']==0]
          train_set_1=train_set[train_set['SFE']==1]
          train_set_0=train_set_0.reset_index(drop=True)
          train_set_1=train_set_1.reset_index(drop=True)
          #clean train set to only contain columns of our interest
          train_set_clean=train_set[[col1,col2,col3,col4,col5]]
          train_set_0_clean=train_set_0[[col1,col2,col3,col4,col5]]
          train_set_1_clean=train_set_1[[col1,col2,col3,col4,col5]]
          smean1=np.array(train_set_0_clean.mean(0))
          smean2=np.array(train_set_1_clean.mean(0))
          cov=train_set_clean.cov()
          #import pdb; pdb.set trace()
          cov inv=np.linalg.inv(cov)
          a lda=cov inv.dot((smean2-smean1))
          b_lda=-0.5*((smean2-smean1).dot(cov_inv)).dot((smean1+smean2))
          x1_lda=np.linspace(0,40,num=20)
          y_1da=-a_1da[0]/a_1da[1]*x1_1da-b_1da/a_1da[1]
In [115]: #estimating error on test set
          test set 0=test set[test set['SFE']==0]
          test_set_1=test_set[test_set['SFE']==1]
          test_set_0=test_set_0.reset_index(drop=True)
          test_set_1=test_set_1.reset_index(drop=True)
          #consider only predictors of our importance
          test_set_clean=test_set[[col1,col2,col3,col4,col5]]
          test_set_0_clean=test_set_0[[col1,col2,col3,col4,col5]]
          test_set_1_clean=test_set_1[[col1,col2,col3,col4,col5]]
          errtest_set_0=np.array(test_set_0_clean)
          errtest_set_1=np.array(test_set_1_clean)
```

```
#import pdb; pdb.set_trace()
          err1=0
          TP=0
          FP=0
          TN=0
          FN=0
          g1=0
          g2=0
          for i in range(0,len(errtest_set_0)):
              g1=a_lda.dot(errtest_set_0[i])+b_lda
              if (g1>0):
                  err1+=1
                  FN+=1
              if (g1<=0):
                  TN+=1
          err2=0
          for i in range(0,len(errtest_set_1)):
              g2=a_lda.dot(errtest_set_1[i])+b_lda
              if (g2<=0):
                  err2+=1
                  FP+=1
              if (g2>0):
                  TP+=1
          err=(err1+err2)/(len(errtest_set_0)+len(errtest_set_1))
          conf_mat=np.array(([TP,FP],[FN,TN]))
          print(err)
0.07920792079207921
In [116]: pd.DataFrame(conf_mat,[' Predicted 1','Predicted 0'],['Actual 1','Actual 0'])
Out[116]:
                        Actual 1 Actual 0
           Predicted 1
                              45
                                         5
                               3
          Predicted 0
                                        48
```