HW2_2trial2

November 1, 2018

In [2]:
```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
f=open('SFE_Test_Data.txt','r')
SFE_testf=f.read()
f.close

f=open('SFE_Train_Data.txt','r')
SFE_trainf=f.read()
f.close
```

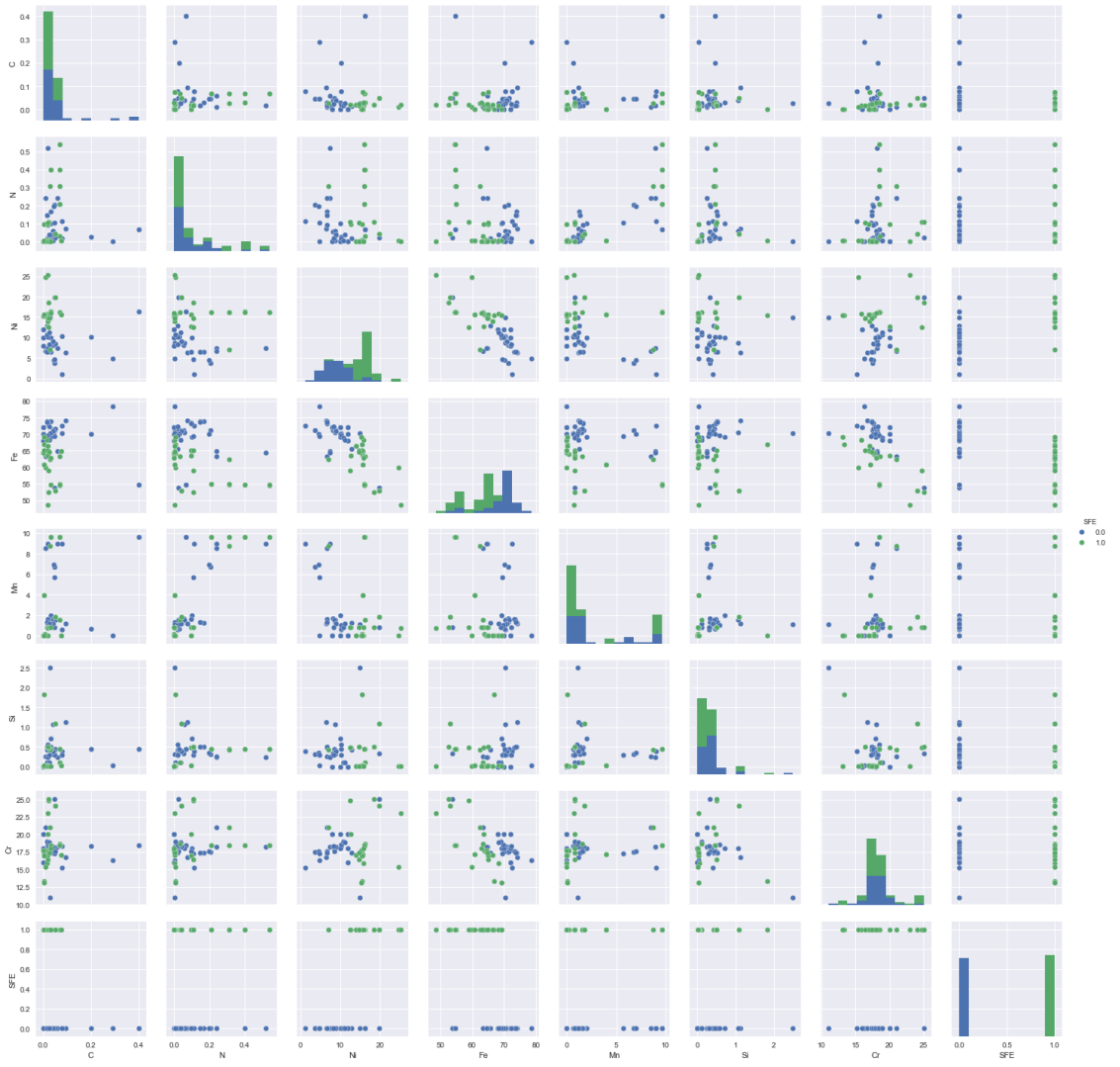
Out[2]:
```
<function TextIOWrapper.close()>
```

In [3]:
```
def list_concat(word):
    out=''
    for i in word:
        out+=i
    return(out)
```

In [4]:
```
temp=[]
SFE_train=[]
for t in SFE_trainf:
    if(t!='\n' and t!='\t'):
        temp.append(t)
    else:
        SFE_train.append(temp)
        temp=[]
temp=[]
SFE_test=[]
for t in SFE_testf:
    if(t!='\n' and t!='\t'):
        temp.append(t)
    else:
        SFE_test.append(temp)
        temp=[]
```

In [5]:
```
SFE_train1=[]
for i in SFE_train:
```

```
            SFE_train1.append(list_concat(i))
        SFE_test1=[]
        for i in SFE_test:
            SFE_test1.append(list_concat(i))

In [6]: cols=SFE_train1[0:8]
        train=SFE_train1[8:len(SFE_train1)]

        cols=SFE_test1[0:8]
        test=SFE_test1[8:len(SFE_test1)]

In [7]: train3=[]
        for i in train:
            if (i=='High'):
                train3.append(float(1))
            elif (i=='Low'):
                train3.append(float(0))
            else:
                train3.append(float(i))

        test3=[]
        for i in test:
            if (i=='High'):
                test3.append(float(1))
            elif (i=='Low'):
                test3.append(float(0))
            else:
                test3.append(float(i))

In [8]: train1=np.array(train3)
        test1=np.array(test3)

In [9]: train4=np.reshape(train1,(25,8))
        test4=np.reshape(test1,(int(len(test1)/8),8))

In [10]: len(test1)/8

Out[10]: 98.0

In [11]: traindf=pd.DataFrame(data=train4,columns=cols)
         testdf=pd.DataFrame(data=test4,columns=cols)

In [12]: testdf
         #scatterplot
         sns.set()
         sns.pairplot(testdf, size = 2.5,hue='SFE')
         plt.show();
```




```
In [13]: import numpy as np
         import matplotlib.pyplot as plt
         from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
         from sklearn.neighbors import KNeighborsClassifier
         from itertools import combinations

         #Exhaustive search
         #Obtain combinations of size r
         train_cols=traindf.columns[0:7]
         test_cols=testdf.columns[0:7]

         all_combs=[]
         all_combs_feat=[]
         for ncomb in range(1,6):
```

```
            #combinations of features
            col_comb = list(combinations(train_cols, ncomb))
            count=0
            score_combs=np.zeros([len(col_comb),4])
            sel_feat=[]
            print('N features', ncomb)
            #import pdb; pdb.set_trace()
            for col_i in col_comb :

                X_train=traindf.loc[:,col_i]
                Y_train=traindf.loc[:,'SFE']

                X_test=testdf.loc[:,col_i]
                Y_test=testdf.loc[:,'SFE']

                #model1
                model1 = LinearDiscriminantAnalysis()
                model1.fit(X_train, Y_train)
                LinearDiscriminantAnalysis(n_components=None, priors=None, shrinkage=None,
                solver='svd', store_covariance=False, tol=0.0001)

                #model2
                model2 = KNeighborsClassifier(n_neighbors=3)
                model2.fit(X_train, Y_train)
                KNeighborsClassifier(...)

                #acccuracy for this iteration
                score_combs[count,0]=model1.score(X_train,Y_train)
                score_combs[count,1]=model2.score(X_train,Y_train)
                score_combs[count,2]=model1.score(X_test,Y_test)
                score_combs[count,3]=model2.score(X_test,Y_test)
                #print('Count score',score_combs)
                count+=1
                sel_feat.append(col_i)
            all_combs_feat.append(sel_feat)
            all_combs.append(score_combs)
```

```
N features 1
N features 2
N features 3
N features 4
N features 5
```

```
In [14]: x=all_combs[0][:,0]
```

```
In [15]: all_combs_feat[0]
```

```
Out[15]: [('C',), ('N',), ('Ni',), ('Fe',), ('Mn',), ('Si',), ('Cr',)]
```

```
In [16]: from operator import itemgetter
         maxscore=[]
         maxscore_feat=[]
         maxscore.append(max(enumerate(x), key=itemgetter(1))[1])
         maxscore_feat.append(all_combs_feat[0][max(enumerate(x), key=itemgetter(1))[0]])

In [17]: type(maxscore_feat)

Out[17]: list

In [18]: score_max=[]
         feat_set=[]
         #number of features to be considered
         for nfeat in range (0,5):
             maxscore=[]
             best_feat=[]
             #different models and err types
             for i in range (0,4):
                 x=all_combs[nfeat][:,i]
                 maxscore.append(max(enumerate(x), key=itemgetter(1))[1])
                 best_feat.append(all_combs_feat[nfeat][max(enumerate(x), key=itemgetter(1))[0]
             score_max.append(maxscore)
             feat_set.append(best_feat)

In [36]: feat_set[0]

Out[36]: [('Fe',), ('Mn',), ('Ni',), ('Ni',)]

In [39]: test_errlda=[]
         test_errknn=[]
         for nfeat in range(0,5):
             #two models
             err1=[]
             err2=[]
             for i in range(0,2):
                 col_i=feat_set[nfeat][i]

                 X_train=traindf.loc[:,col_i]
                 Y_train=traindf.loc[:,'SFE']

                 X_test=testdf.loc[:,col_i]
                 Y_test=testdf.loc[:,'SFE']

                 #model1
                 model1 = LinearDiscriminantAnalysis()
                 model1.fit(X_train, Y_train)
                 LinearDiscriminantAnalysis(n_components=None, priors=None, shrinkage=None,
                 solver='svd', store_covariance=False, tol=0.0001)
```

```
                #model2
                model2 = KNeighborsClassifier(n_neighbors=3)
                model2.fit(X_train, Y_train)
                KNeighborsClassifier(...)

                #acccuracy for this iteration
                #import pdb; pdb.set_trace()
                if (i==0):
                    err1.append(1-model1.score(X_test,Y_test))
                else:
                    err2.append(1-model2.score(X_test,Y_test))
            test_errlda.append(err1)
            test_errknn.append(err2)
```

```
In [47]: test_errlda[0]
```

```
Out[47]: [0.1428571428571429]
```

```
In [43]: varlda_app=[]
         varknn_app=[]
         errlda_app=[]
         errlda_test=[]
         errknn_app=[]
         errknn_test=[]
         for i in range(0,5):
             varlda_app.append(feat_set[i][0])
             varknn_app.append(feat_set[i][2])
             errlda_app.append(1-score_max[i][0])
             errlda_test.append(test_errlda[i])
             errknn_app.append(1-score_max[i][2])
             errknn_test.append(test_errknn[i])
```

```
In [51]: errlda_app[0]
```

```
Out[51]: 0.12
```

```
In [52]: dat=[]
         for i in range(0,6):
             dat.append([])

         dat[0].append('LDA Based Features')
         dat[1].append('LDA best apparent Error')
         dat[2].append('LDA test Error')
         dat[3].append('KNN Based Features')
         dat[4].append('KNN apparent Error')
         dat[5].append('KNN_test Error')

         for i in range(0,5):
```

```
            dat[0].append(varlda_app[i])
            dat[1].append(errlda_app[i])
            dat[2].append(errlda_test[i])
            dat[3].append(varknn_app[i])
            dat[4].append(errknn_app[i])
            dat[5].append(errknn_test[i])
```

In [1]:
```
feat_set[0][0]
```

```
---------------------------------------------------------------------------

NameError                                 Traceback (most recent call last)

<ipython-input-1-17e655e56012> in <module>()
----> 1 feat_set[0][0]


NameError: name 'feat_set' is not defined
```

In [53]:
```
labs=['Categories','1 Feature','2 Features','3 Features','4 Features','5 Features']
df1=pd.DataFrame(data=dat,columns=labs)
df1
```

Out[53]:

	Categories	1 Feature	2 Features	3 Features	4 Features	5 Features
0	LDA Based Features	(Fe,)	(C, Fe)	(C, Ni, Fe)	(C, N, Fe, Mn)	(N, Ni, Fe, Si, Cr)
1	LDA best apparent Error	0.12	0.04	0.04	0.04	0
2	LDA test Error	[0.1428571428571429]	[0.12244897959183676]	[0.061224489795918324]	[0.11224489795918369]	[0.16326530612244894]
3	KNN Based Features	(Ni,)	(N, Ni)	(C, Ni, Fe)	(C, N, Ni, Si)	(C, Fe, Mn, Si, Cr)
4	KNN apparent Error	0.122449	0.0714286	0.0612245	0.0510204	0.0408163
5	KNN_test Error	[0.2551020408163265]	[0.23469387755102045]	[0.23469387755102045]	[0.061224489795918324]	[0.061224489795918324]

In [54]:
```
#Sequential forward search
train_cols=traindf.columns[0:7]
test_cols=testdf.columns[0:7]

#initialize variables for storing features and scores
all_combs=[]
all_combs_feat=[]
chosen_feat=[]
```

```
score_max=[]
for i in range(0,4):
    chosen_feat.append([])
    score_max.append([])
feat_set=[]


#keep adding one feature at a time
for nfeat in range(1,6):
    #combinations of features
    #keep track of remaining columns separately for each path
    rem_cols=[]
    for j in range(0,4):
        rem_cols.append([])
    for i in range(0,4):
        for t in train_cols:
            if (t not in chosen_feat[i]):
                rem_cols[i].append(t)

    score_combs=np.zeros([len(rem_cols[0]),4])
    sel_feat=[]
    for i in range(0,4):
        sel_feat.append([])
    #for each of the four paths
    for path in range (0,4) :
        count=0
        for col_i in rem_cols[path]:
            X_train=traindf.loc[:,chosen_feat[path]+list([col_i])]
            Y_train=traindf.loc[:,'SFE']

            X_test=testdf.loc[:,chosen_feat[path]+list([col_i])]
            Y_test=testdf.loc[:,'SFE']
            #import pdb; pdb.set_trace()
            #model1
            model1 = LinearDiscriminantAnalysis()
            model1.fit(X_train, Y_train)
            LinearDiscriminantAnalysis(n_components=None, priors=None, shrinkage=None
            solver='svd', store_covariance=False, tol=0.0001)

            #model2
            model2 = KNeighborsClassifier(n_neighbors=3)
            model2.fit(X_train, Y_train)
            KNeighborsClassifier(...)

            #acccuracy for this iteration
            if (path==0):
                score_combs[count,0]=model1.score(X_train,Y_train)
            elif(path==1):
```

```
                    score_combs[count,1]=model2.score(X_train,Y_train)
                elif(path==2):
                    score_combs[count,2]=model1.score(X_test,Y_test)
                else:
                    score_combs[count,3]=model2.score(X_test,Y_test)
                count+=1
                sel_feat[path].append(col_i)
            #print('Chosen Features ',path,chosen_feat,'\n',score_combs)
        all_combs.append(score_combs)
        for i in range (0,4):
            x=all_combs[nfeat-1][:,i]
            maxscore=(max(enumerate(x), key=itemgetter(1))[1])
            best_feat=(sel_feat[i][max(enumerate(x), key=itemgetter(1))[0]])
            score_max[i].append(maxscore)
            chosen_feat[i].append(best_feat)
        #all_combs_feat[i].append(chosen_feat)
        #import pdb; pdb.set_trace()
```

In [55]: score_max

```
Out[55]: [[0.88, 0.96, 0.96, 0.96, 0.96],
 [0.96, 0.96, 0.96, 0.96, 0.92],
 [0.8775510204081632,
  0.9285714285714286,
  0.9285714285714286,
  0.9489795918367347,
  0.9081632653061225],
 [0.9081632653061225,
  0.9387755102040817,
  0.9387755102040817,
  0.9387755102040817,
  0.9387755102040817]]
```

In [290]: all_combs_feat

```
Out[290]: [[['Fe', 'C', 'Ni', 'Mn', 'N'],
  ['Mn', 'C', 'N', 'Si', 'Ni'],
  ['Ni', 'N', 'C', 'Si', 'Mn'],
  ['Ni', 'Fe', 'C', 'N', 'Mn']],
 [['Fe', 'C', 'Ni', 'Mn', 'N'],
  ['Mn', 'C', 'N', 'Si', 'Ni'],
  ['Ni', 'N', 'C', 'Si', 'Mn'],
  ['Ni', 'Fe', 'C', 'N', 'Mn']],
 [['Fe', 'C', 'Ni', 'Mn', 'N'],
  ['Mn', 'C', 'N', 'Si', 'Ni'],
  ['Ni', 'N', 'C', 'Si', 'Mn'],
  ['Ni', 'Fe', 'C', 'N', 'Mn']],
 [['Fe', 'C', 'Ni', 'Mn', 'N'],
  ['Mn', 'C', 'N', 'Si', 'Ni'],
```

```
            ['Ni', 'N', 'C', 'Si', 'Mn'],
            ['Ni', 'Fe', 'C', 'N', 'Mn']],
           [['Fe', 'C', 'Ni', 'Mn', 'N'],
            ['Mn', 'C', 'N', 'Si', 'Ni'],
            ['Ni', 'N', 'C', 'Si', 'Mn'],
            ['Ni', 'Fe', 'C', 'N', 'Mn']]]
```

```
In [220]: a=['p']
          b=['tt']
          c=list(b)
          a+c
```

```
Out[220]: ['p', 'tt']
```

```
In [62]: chosen_feat[0][0:1]
```

```
Out[62]: ['Fe']
```

```
In [63]: test_errlda=[]
         test_errknn=[]
         for nfeat in range(0,5):
             #two models
             err1=[]
             err2=[]
             for i in range(0,2):
                 col_i=chosen_feat[i][0:nfeat+1]

                 X_train=traindf.loc[:,col_i]
                 Y_train=traindf.loc[:,'SFE']

                 X_test=testdf.loc[:,col_i]
                 Y_test=testdf.loc[:,'SFE']

                 #model1
                 model1 = LinearDiscriminantAnalysis()
                 model1.fit(X_train, Y_train)
                 LinearDiscriminantAnalysis(n_components=None, priors=None, shrinkage=None,
                 solver='svd', store_covariance=False, tol=0.0001)

                 #model2
                 model2 = KNeighborsClassifier(n_neighbors=3)
                 model2.fit(X_train, Y_train)
                 KNeighborsClassifier(...)

                 #acccuracy for this iteration
                 #import pdb; pdb.set_trace()
                 if (i==0):
                     err1.append(1-model1.score(X_test,Y_test))
                 else:
```

```
                err2.append(1-model2.score(X_test,Y_test))
        test_errlda.append(err1)
        test_errknn.append(err2)
```

In [67]:
```
dat=[]
for i in range(0,6):
    dat.append([])

dat[0].append('LDA Based Features')
dat[1].append('LDA best apparent Error')
dat[2].append('LDA test Error')
dat[3].append('KNN Based Features')
dat[4].append('KNN best apparent Error')
dat[5].append('KNN_test Error')

for i in range(0,5):
    dat[0].append(varlda_app[i])
    dat[1].append(errlda_app[i])
    dat[2].append(errlda_test[i])
    dat[3].append(varknn_app[i])
    dat[4].append(errknn_app[i])
    dat[5].append(errknn_test[i])
```

In [68]:
```
labs=['Categories','1 Feature','2 Features','3 Features','4 Features','5 Features']
df1=pd.DataFrame(data=dat,columns=labs)
df1
```

Out[68]:

	Categories	1 Feature	2 Features	3 Features	4 Features	5 Features
0	LDA Based Features	(Fe,)	(C, Fe)	(C, Ni, Fe)	(C, N, Fe, Mn)	(N, Ni, Fe, Si, Cr)
1	LDA best apparent Error	0.12	0.04	0.04	0.04	0
2	LDA test Error	[0.1428571428571429]	[0.12244897959183676]	[0.061224489795918324]	[0.11224489795918369]	[0.16326530612244894]
3	KNN Based Features	(Ni,)	(N, Ni)	(C, Ni, Fe)	(C, N, Ni, Si)	(C, Fe, Mn, Si, Cr)
4	KNN best apparent Error	0.122449	0.0714286	0.0612245	0.0510204	0.0408163
5	KNN_test Error	[0.2551020408163265]	[0.23469387755102045]	[0.23469387755102045]	[0.061224489795918324]	[0.061224489795918324]

In [307]:
```
score_max[0]
```

Out[307]:
```
[0.88, 0.96, 0.96, 0.96, 0.96]
```