```
In [2]:
```

```
import pandas as pd
import numpy as np
from itertools import combinations
from sklearn import preprocessing
from sklearn.svm import LinearSVC,SVC
from sklearn.datasets import make_classification
from sklearn.model_selection import cross_val_score,train_test_split
from sklearn.neural_network import MLPClassifier
```

### In [3]:

```
data = pd.read_csv('Assign1_Training_Data.txt',delimiter="\t")
data.head()
d = np.asarray(data)
data.head()
```

### Out[3]:

	Sample_Number	AL080059	Contig63649_RC	Contig46218_RC	LOC51203	AA555029_RC	ALDH4	Contig38288_RC	FGF18	Contig2
0	138	-0.227	-0.107	-0.086	-0.057	0.073	0.021	-0.002	0.135	
1	184	0.044	-0.031	0.381	0.226	-0.038	-0.167	0.103	-0.330	
2	127	0.151	-0.210	0.034	0.037	-0.065	-0.048	-0.026	-0.425	
3	166	0.335	-0.031	0.177	0.165	-0.372	0.340	0.112	-0.506	
4	318	-0.098	-0.492	-0.307	-0.097	-0.024	0.197	-0.001	0.369	

### 5 rows × 72 columns

### In [4]:

```
data testing = pd.read csv('Assign1 Testing Data.txt',delimiter="\t")
d test = np.asarray(data testing)
Y_test = d_test[:,-1]

X_test = d_test[:,1:71]
x0 test = []
y0 test = []
x1_test = []
y1_test = []
for i in range(len(Y test)):
    if Y test[i]==0:
        x0 test.append(X test[i])
        y0_test.append(Y_test[i])
    else:
        x1 test.append(X test[i])
        y1_test.append(Y_test[i])
print('Number of 0 class in Test set is =',len(x0 test))
print('Number of 1 class in Test set is',len(x1_test))
print('Number of Total classes in Test set is=',len(X test))
```

Number of 0 class in Test set is = 39 Number of 1 class in Test set is 176 Number of Total classes in Test set is= 215

### In [5]:

```
Y = d[:,-1]
X = d[:,1:71]
print('Some Features --> ',X[0:5,0:3])
p = np.vstack((X[:,0], X[:,1],X[:,2])).T
print(X[:,2].shape)
```

```
#p = np.vstack((p, )).T
#p.shape
#X.shape
#p
x0 = []
y0 = []
x1 = []
y1 = []
for i in range(len(Y)):
    if Y[i]==0:
        x0.append(X[i])
        y0.append(Y[i])
    else:
        x1.append(X[i])
        y1.append(Y[i])
print('Number of 0 class in Train set is',len(x0))
print('Number of 1 class in Train set is',len(x1))
print('Number of Total classes in Train set is',len(X))
Some Features --> [[-0.227 -0.107 -0.086]
[ 0.044 -0.031 0.381]
 [-0.098 -0.492 -0.307]]
Number of \, 0 class in Train set is 40 \,
Number of \, 1 class in Train set is 40 \, Number of \, Total classes in Train set is 80 \,
In [6]:
x0 = np.array(x0)
x1 = np.array(x1)
x0_{test} = np.array(x0_{test})
x1 test = np.array(x1 test)
```

### In [7]:

```
def fs es(X,Y,clf,x0,x0 test,y0,y0 test,x1,x1 test,y1,y1 test,X test,Y test):
   c = X.shape[1]
   temp_e = np.inf
   e = 0
    for i in range(c):
        for j in range(i+1,c):
            x = np.vstack((X[:,i], X[:,j])).T
            y = y
            Xx_train, Xx_test, yy_train, yy_test = train_test_split(x,y, test_size=0.0,
random state=0)
            e = 1 - clf.fit(Xx_train, yy_train).score(Xx_train, yy_train)
            #print(e)
            if e < temp_e:</pre>
                index1 = i
                index2 = j
                temp e = e
    x0 = np.vstack((x0[:,index1],x0[:,index2])).T
    x1 = np.vstack((x1[:,index1],x1[:,index2])).T
    x0\_test\_ = np.vstack((x0\_test[:,index1],x0\_test[:,index2])).T
    x1_test_ = np.vstack((x1_test[:,index1],x1_test[:,index2])).T
    x_test_ = np.vstack((X_test[:,index1], X_test[:,index2])).T
    e0_train = 1 - clf.score(x0_,y0)
    el_train = 1 - clf.score(x1_,y1)
e0_test = 1 - clf.score(x0_test_,y0_test)
    e1 test = 1 - clf.score(x1_test_,y1_test)
    e1_total = 1 - clf.score(x_test_,Y_test)
    return temp_e,index1,index2,e0_train,e1_train,e0_test,e1_test,e1_total
```

```
def fs sfs3(ind1,ind2,X,Y,clf,x0,x0_test,y0,y0_test,x1,x1_test,y1,y1_test,X_test,Y_test):
    c = X.shape[1]
   temp_e = np.inf
    e = 0
    for i in range(c):
        if (i != ind1 and i != ind2):
            x = np.vstack((X[:,ind1], X[:,ind2], X[:,i])).T
            Xx train, Xx test, yy train, yy test = train test split(x,y, test size=0.0,
random state=0)
            e = 1 - clf.fit(Xx train, yy train).score(Xx train, yy train)
            #print(e)
            if e < temp e:</pre>
                index = i
                temp_e = e
    x0_{=} np.vstack((x0[:,ind1],x0[:,ind2],x0[:,index])).T
        = np.vstack((x1[:,ind1],x1[:,ind2],x1[:,index])).T
    x0_test_ = np.vstack((x0_test[:,ind1],x0_test[:,ind2],x0_test[:,index])).T
    x1_test_ = np.vstack((x1_test[:,ind1],x1_test[:,ind2],x1 test[:,index])).T
    x_test_ = np.vstack((X_test[:,ind1], X_test[:,ind2], X_test[:,index])).T
    e0_train = 1 - clf.score(x0_,y0)
e1_train = 1 - clf.score(x1_,y1)
    e0_test = 1 - clf.score(x0_test_,y0_test)
    e1 test = 1 - clf.score(x1 test ,y1 test)
    el total = 1 - clf.score(x test ,Y test)
    return temp_e,index,e0_train,e1_train,e0_test,e1_test,e1_total
```

### In [9]:

```
def fs sfs4(ind1,ind2,ind3,X,Y,clf,x0,x0_test,y0,y0_test,x1,x1_test,y1,y1_test,X_test,Y_test):
   c = X.shape[1]
   temp e = np.inf
    e = 0
    for i in range(c):
        if (i != ind1 and i != ind2 and i != ind3):
            x = np.vstack((X[:,ind1], X[:,ind2], X[:,ind3],X[:,i])).T
            y = Y
            Xx train, Xx test, yy train, yy test = train test split(x,y, test size=0.0,
random state=0)
            e = 1 - clf.fit(Xx_train, yy_train).score(Xx_train, yy_train)
             #print(e)
            if e < temp e:</pre>
                index = i
                temp e = e
    x0 = np.vstack((x0[:,ind1],x0[:,ind2],x0[:,ind3],x0[:,index])).T
    x1 = np.vstack((x1[:,ind1],x1[:,ind2],x1[:,ind3],x1[:,index])).T
    x0 \text{ test} = \text{np.vstack}((x0 \text{ test[:,ind1]}, x0 \text{ test[:,ind2]}, x0 \text{ test[:,ind3]}, x0 \text{ test[:,index]})).T
    x1\_test\_ = np.vstack((x1\_test[:,ind1],x1\_test[:,ind2],x1\_test[:,ind3],x1\_test[:,index])).T
    x_test_ = np.vstack((X_test[:,ind1], X_test[:,ind2], X_test[:,ind3], X_test[:,index])).T
    e0 train = 1 - clf.score(x0,y0)
    e1_{train} = 1 - clf.score(x1_,y1)
    e0_test = 1 - clf.score(x0_test_,y0_test)
    el test = 1 - clf.score(x1 test,y1 test)
    el total = 1 - clf.score(x test ,Y test)
    return temp e,index,e0 train,e1 train,e0 test,e1 test,e1 total
```

### In [10]:

```
def fs_sfs5(ind1,ind2,ind3,ind4,X,Y,clf,x0,x0_test,y0,y0_test,x1,x1_test,y1,y1_test,X_test,Y_test):
```

```
c = X.shape[I]
   temp e = np.inf
   e = 0
   for i in range(c):
       if (i != ind1 and i != ind2 and i != ind3 and i != ind4):
            x = np.vstack((X[:,ind1], X[:,ind2], X[:,ind3], X[:,ind4], X[:,i])).T
            Xx train, Xx test, yy train, yy test = train test split(x,y, test size=0.0,
random_state=0)
            e = 1 - clf.fit(Xx train, yy train).score(Xx train, yy train)
            if e < temp_e:</pre>
               index = i
               temp e = e
   x0_{=} np.vstack((x0[:,ind1],x0[:,ind2],x0[:,ind3],x0[:,ind4],x0[:,index])).T
   x1 = np.vstack((x1[:,ind1],x1[:,ind2],x1[:,ind3],x1[:,ind4],x1[:,index])).T
   e0_{train} = 1 - clf.score(x0_,y0)
   el_train = 1 - clf.score(x1,y1)
   x0_test_ = np.vstack((x0_test[:,ind1],x0_test[:,ind2],x0_test[:,ind3],x0_test[:,ind4],x0_test[
:,index])).T
   x1_test_ = np.vstack((x1_test[:,ind1],x1_test[:,ind2],x1_test[:,ind3],x1_test[:,ind4],x1_test[
:,index])).T
   e0 test = 1 - clf.score(x0 test, y0 test)
   el test = 1 - clf.score(x1 test,y1 test)
   x test =
np.vstack((X test[:,ind1],X test[:,ind2],X test[:,ind3],X test[:,ind4],X test[:,index])).T
   e1_total = 1 - clf.score(x_test_,Y_test)
   return temp e,index,e0 train,e1 train,e0 test,e1 test,e1 total
```

# 1) Linear SVM

### 1.1 Linear SVM NO feature selection

```
In [11]:
```

```
Clf_svmL = SVC(kernel='linear', C=1)
Xn_train, Xn_test, yn_train, yn_test = train_test_split(X,Y, test_size=0.0, random_state=0)
e = 1 - clf_svmL.fit(Xn_train, yn_train).score(Xn_train, yn_train)
e0_train_svm2_n = 1 - clf_svmL.score(x0, y0)
e1_train_svm2_n = 1 - clf_svmL.score(x1, y1)
e0_test_svm2_n = 1 - clf_svmL.score(x0_test, y0_test)
e1_test_svm2_n = 1 - clf_svmL.score(x1_test, y1_test)
e1_test_svm2_n = 1 - clf_svmL.score(X_test, Y_test)
print('Resubstitution error = ',e)
print('Resubstitution error = ',e)
print("Train_Error_on_class_0",e0_train_svm2_n)
print("Train_Error_on_class_0",e0_test_svm2_n)
print("Test_Error_on_class_0",e0_test_svm2_n)
print("Test_Error_on_class_1",e1_test_svm2_n)
print("Test_Error_on_class_1",e1_test_svm2_n)
print("Test_Error_Total",e1_total_svm2_n)
```

```
Resubstitution error = 0.05000000000000000044
Train Error on class 0 0.0749999999999996
Train Error on class 1 0.0250000000000000022
Test Error on class 0 0.3846153846153846
Test Error on class 1 0.3295454545454546
Test Error Total 0.3395348837209302
```

### 1.2 Linear SVM 2-feature selection extraction

```
In [12]:
```

```
print("Test Error on class 0",e0_test_svm2_2)
print("Test Error on class 1",e1_test_svm2_2)
print("Test Error Total",e1_total_svm2_2)

print("Test Error Total",e1_total_svm2_2)

Resubstitution error = 0.21250000000000002

Feature 1 = 3 Feature 2 = 19

Train Error on class 0 0.3249999999999999

Train Error on class 1 0.25

Test Error on class 0 0.41025641025641024

Test Error on class 1 0.36931818181818177

Test Error Total 0.3767441860465116
```

### 1.3 Linear SVM 3-feature selection sequential forward search

```
In [13]:
```

```
e,feature3_svmL,e0_train_svm2_3,e1_train_svm2_3,e0_test_svm2_3,e1_test_svm2_3,e1_total_svm2_3 = fs_
sfs3(feature1_svmL,feature2_svmL,X,Y,clf_svmL,

x0,x0_test,y0,y0_test,x1,x1_test,y1,y1_test,X_test,Y_test)
print('Resubstitution error = ',e,'\nFeature 1 = ', feature1_svmL,' Feature 2 = ',feature2_svmL,'
Feature 3 = ',feature3_svmL)
print("Train Error on class 0",e0_train_svm2_3)
print("Train Error on class 1",e1_train_svm2_3)
print("Test Error on class 0",e0_test_svm2_3)
print("Test Error on class 1",e1_test_svm2_3)
print("Test Error Total",e1_total_svm2_3)
```

### 1.4 Linear SVM 4-feature selection sequential forward search

```
In [14]:
```

```
e,feature4_svmL,e0_train_svm2_4,e1_train_svm2_4,e0_test_svm2_4,e1_test_svm2_4,e1_total_svm2_4 = fs_
sfs4(feature1_svmL,feature2_svmL,feature3_svmL,X,Y,clf_svmL,

x0,x0_test,y0,y0_test,x1,x1_test,y1,y1_test,X_test,Y_test)
print('Resubstitution error = ',e,'\nFeature 1 = ', feature1_svmL,' Feature 2 = ',feature2_svmL,'
Feature 3 = ',feature3_svmL,' Feature 4 = ',feature4_svmL)
print("Train_Error on class 0",e0_train_svm2_4)
print("Test_Error on class 1",e1_train_svm2_4)
print("Test_Error on class 1",e1_test_svm2_4)
print("Test_Error Total",e1_total_svm2_4)

Resubstitution error = 0.1875
Feature 1 = 3 Feature 2 = 19 Feature 3 = 27 Feature 4 = 5
```

```
Resubstitution error = 0.1875

Feature 1 = 3 Feature 2 = 19 Feature 3 = 27 Feature 4 = 5

Train Error on class 0 0.0500000000000044

Train Error on class 1 0.4250000000000004

Test Error on class 0 0.1282051282051282

Test Error on class 1 0.54545454545454

Test Error Total 0.46976744186046515
```

### 1.5 Linear SVM 5-feature selection sequential forward search

```
In [15]:
```

```
e, feature5_svmL, e0_train_svm2_5, e1_train_svm2_5, e0_test_svm2_5, e1_test_svm2_5, e1_total_svm2_5 = fs_
sfs5 (feature1_svmL, feature2_svmL, feature3_svmL, feature4_svmL, X, Y, c1f_svmL,
x0, x0_test, y0, y0_test, x1, x1_test, y1, y1_test, X_test, Y_test)
```

### 2) Non Linear SVM

### 2.1 Non Linear SVM no feature selection

```
In [16]:
```

```
clf_svmNL = SVC(gamma = 'auto' , kernel='rbf', C=10)
Xn_train, Xn_test, yn_train, yn_test = train_test_split(X,Y, test_size=0.0, random_state=0)
e = 1 - clf_svmNL.fit(Xn_train, yn_train).score(Xn_train, yn_train)
e0_trainsvmNL_n = 1-clf_svmNL.score(x0, y0)
e1_trainsvmNL_n = 1-clf_svmNL.score(x1, y1)
e0_testsvmNL_n = 1-clf_svmNL.score(x0_test, y0_test)
e1_testsvmNL_n = 1-clf_svmNL.score(x1_test, y1_test)
e1_totalsvmNL_n = 1-clf_svmNL.score(X_test, Y_test)
print('Resubstitution error = ',e)
print("Train Error on class 0",e0_trainsvmNL_n)
print("Train Error on class 1",e1_trainsvmNL_n)
print("Test Error on class 1",e1_testsvmNL_n)
print("Test Error Total",e1_totalsvmNL_n)
```

```
Resubstitution error = 0.0625
Train Error on class 0 0.0999999999999998
Train Error on class 1 0.0250000000000000022
Test Error on class 0 0.3846153846153846
Test Error on class 1 0.3352272727272727
Test Error Total 0.3441860465116279
```

### 2.2 Non-Linear SVM 2-feature selection extraction

### In [17]:

### 2.3 Non Linear SVM 3-feature selection sequential forward search

```
In [18]:
e,feature3 svmNL,e0 train svmNL 3,e1 train svmNL 3,e0 test svmNL 3,e1 test svmNL 3,e1 total svmNL 3
= fs sfs3(feature1 svmNL, feature2 svmNL, X, Y, clf svmNL,
x0,x0 test,y0,y0 test,x1,x1 test,y1,y1 test,X test,Y test)
print('Resubstitution error = ',e,'\nFeature 1 = ', feature1 svmNL,' Feature 2 = ',feature2 svmNL,
 Feature 3 = ', feature3 svmNL)
print("Train Error on class 0",e0 train svmNL 3)
print("Train Error on class 1",e1 train svmNL 3)
print("Test Error on class 0",e0 test svmNL 3)
print("Test Error on class 1",el_test_svmNL_3)
print("Test Error Total",e1 total svmNL 3)
4
Resubstitution error = 0.175000000000000004
Feature 1 = 12 Feature 2 = 48 Feature 3 = 32
Train Error on class 0 0.625
Train Error on class 1 0.275
Test Error on class 0 0.641025641025641
Test Error on class 1 0.3636363636363636365
Test Error Total 0.413953488372093
2.4 Non Linear SVM 4-feature selection sequential forward search
In [19]:
,feature4_svmNL,e0_train_svmNL_4,e1_train_svmNL 4,e0 test_svmNL 4,e1 test_svmNL 4,e1 total_svmNL 4
= fs sfs4(feature1 svmNL, feature2 svmNL, feature3 svmNL, X, Y, clf svmNL,
x0,x0_test,y0,y0_test,x1,x1_test,y1,y1_test,X_test,Y_test)
print('Resubstitution error = ',e,'\nFeature 1 = ', feature1 svmNL,' Feature 2 = ',feature2 svmNL,
 Feature 3 = ',feature3 svmNL,' Feature 4 = ',feature4 svmNL)
print("Train Error on class 0",e0 train svmNL 4)
print("Train Error on class 1",e1_train_svmNL_4)
print("Test Error on class 0",e0_test_svmNL_4)
print("Test Error on class 1",e1 test svmNL 4)
print("Test Error Total",e1_total_svmNL_4)
Resubstitution error = 0.15000000000000000
Feature 1 = 12 Feature 2 = 48 Feature 3 = 32 Feature 4 = 1
Train Error on class 0 0.099999999999998
Train Error on class 1 0.25
Test Error on class 0 0.23076923076923073
Test Error on class 1 0.48863636363636365
Test Error Total 0.4418604651162791
2.5 Non Linear SVM 5-feature selection sequential forward search
In [20]:
,feature5_svmNL,e0_train_svmNL_5,e1_train_svmNL_5,e0_test_svmNL_5,e1_test_svmNL_5,e1_total_svmNL_5
= fs sfs5(feature1 svmNL, feature2 svmNL, feature3 svmNL, feature4 svmNL, X, Y, clf svmNL,
```

## x0,x0\_test,y0,y0\_test,x1,x1\_test,y1,y1\_test,X\_test,Y\_test) print('Resubstitution error = ',e,'\nFeature 1 = ', feature1 symNL,' Feature 2 = ',feature2 symNL, Feature 3 = ',feature3 svmNL,' Feature 4 = ',feature4 svmNL,' Feature 5 = ',feature5 svmNL) print("Train Error on class 0",e0 train svmNL 5) print("Train Error on class 1",e1 train svmNL print("Test Error on class 0",e0 test svmNL 5) print("Test Error on class 1",e1 test svmNL 5) print("Test Error Total",el\_total\_svmNL\_5) Resubstitution error = 0.1374999999999999 Feature 1 = 12 Feature 2 = 48 Feature 3 = 32 Feature 4 = 1 Feature 5 = 24 Train Error on class 0 0.125

Train Error on class 1 0.25

```
Test Error on class 0 0.2564102564102564
Test Error on class 1 0.44318181818181823
Test Error Total 0.40930232558139534
```

### 3 Neural Network

### 3.1 Neural Network No feature selection

```
In [21]:
```

```
Resubstitution error = 0.0
Train Error on class 0 0.0
Train Error on class 1 0.0
Test Error on class 0 0.3333333333333337
Test Error on class 1 0.34090909090909094
Test Error Total 0.3395348837209302
```

### 3.2 Neural Network 2-feature selection extraction search

```
In [22]:
```

### Neural Network 3-feature selection sequential forward search

```
In [23]
```

```
e, feature3_nn,e0_train_nn_3,e1_train_nn_3,e0_test_nn_3,e1_test_nn_3,e1_total_nn_3 =
fs_sfs3(feature1_nn,feature2_nn,X,Y,clf_NN,

x0,x0_test,y0,y0_test,x1,x1_test,y1,y1_test,X_test,Y_test)
print('Resubstitution error = ',e,'\nFeature 1 = ', feature1_nn,' Feature 2 = ',feature2_nn,'
Feature 3 = ',feature3_nn)
print('Train_From_on_class_0' = 0 train_nn_3)
```

```
print("Train Error on class 1",el_train_nn_3)
print("Test Error on class 0",e0_test_nn_3)
print("Test Error on class 1",el_test_nn_3)
print("Test Error Total",el_total_nn_3)

Resubstitution error = 0.125
Feature 1 = 59 Feature 2 = 67 Feature 3 = 51
Train Error on class 0 0.375
Train Error on class 1 0.2249999999999998
Test Error on class 0 0.3846153846
Test Error on class 1 0.386363636363635
Test Error Total 0.38604651162790693
```

### Neural Network 4-feature selection sequential forward search

### In [24]:

```
e, feature4_nn, e0_train_nn_4, e1_train_nn_4, e0_test_nn_4, e1_test_nn_4, e1_total_nn_4 =
fs_sfs4 (feature1_nn, feature2_nn, feature3_nn, X, Y, clf_NN,

x0, x0_test, y0, y0_test, x1, x1_test, y1, y1_test, X_test, Y_test)
print('Resubstitution error = ',e,'\nFeature 1 = ', feature1_nn,' Feature 2 = ', feature2_nn,'
Feature 3 = ', feature3_nn,' Feature 4 = ', feature4_nn)
print("Train Error on class 0", e0_train_nn_4)
print("Train Error on class 1", e1_train_nn_4)
print("Test Error on class 1", e1_test_nn_4)
print("Test Error Total", e1_total_nn_4)
```

```
Resubstitution error = 0.037499999999998

Feature 1 = 59 Feature 2 = 67 Feature 3 = 51 Feature 4 = 19

Train Error on class 0 0.25

Train Error on class 1 0.449999999999996

Test Error on class 0 0.3076923076923077

Test Error on class 1 0.465909090909094

Test Error Total 0.4372093023255814
```

### Neural network 5-feature selection sequential forward search

```
In [25]:
```

```
e, feature5_nn, e0_train_nn_5, e1_train_nn_5, e0_test_nn_5, e1_test_nn_5, e1_total_nn_5 =
fs_sfs5 (feature1_nn, feature2_nn, feature3_nn, feature4_nn, X, Y, c1f_NN,

x0, x0_test, y0, y0_test, x1, x1_test, y1, y1_test, X_test, Y_test)
print('Resubstitution error = ',e,'\nFeature 1 = ', feature1_nn,' Feature 2 = ', feature2_nn,'
Feature 3 = ', feature3_nn,' Feature 4 = ', feature4_nn,' Feature 5 = ', feature5_nn)
print("Train Error on class 0", e0_train_nn_5)
print("Train Error on class 1",e1_train_nn_5)
print("Test Error on class 0",e0_test_nn_5)
print("Test Error on class 1",e1_test_nn_5)
print("Test Error Total",e1_total_nn_5)
Resubstitution error = 0.0
```

```
Resubstitution error = 0.0
Feature 1 = 59 Feature 2 = 67 Feature 3 = 51 Feature 4 = 19 Feature 5 = 30
Train Error on class 0 0.1500000000000002
Train Error on class 1 0.199999999999999
Test Error on class 0 0.3846153846
Test Error on class 1 0.41477272727273
Test Error Total 0.40930232558139534
```

5				Train			Test			
			Feature	Selected Features	Total (Resub)	class 0 error	class 1 error	class 0 error	class 1 error	Total
		Without								
•		Selection	No	NO	0.05	0.07	0.02	0.38	0.32	0.33
	Linear SVM	Exhaustive	2	3,19	0.212	0.32	0.25	0.41	0.36	0.37
	Lillear Svivi	Sequential	3	27	0.19	0.05	0.44	0.15	0.57	0.49
•		Forward	4	5	0.18	0.05	0.42	0.12	0.54	0.46
		Search	5	44	0.18	0.09	0.375	0.15	0.52	0.46
		Without								
		Selection	No	NO	0.0625	0.09	0.02	0.38	0.33	0.34
Classifier	Non Linear	Exhaustive	2	12,48	0.21	0.32	0.42	0.17	0.46	0.4
Classifier	Svm	Sequential	3	32	0.17	0.62	0.27	0.64	0.36	0.41
		Forward	4	1	0.15	0.09	0.25	0.23	0.48	0.44
		Search	5	5	0.13	0.12	0.25	0.25	0.44	0.4
		Without								
		Selection	No	No	0	0	0	0.33	0.34	0.34
	Neural	Exhaustive	2	59,67	0.1	0.42	0.22	0.41	0.39	0.4
	Network	Sequential	3	51	0.12	0.37	0.22	0.38	0.38	0.38
		Forward	4	19	0.03	0.25	0.44	0.3	0.46	0.43
<u> </u>	<u> </u>	Search	5	5	0	0.15	0.19	0.38	0.41	0.4

The resubstitution error seems to be low for almost all the cases. The specific 0 and 1 class resub. error is also low for almost all the cases. We can also observe no majority has been observed for feature selections amongst all three classifiers. There is also an obvious interpretation that even for class 0 and 1 the test error is higher than the train error. The neural networks 0 and 1 class test error seems to be almost equal but for both the SVM the 1 class error is high for most of the cases which can be because of non-equal data sizes for class 1 and 0 in the test set. Based on the error rates no classifier seems to be capable of making good predictions and the error rates seems to oscillate with increasing and decreasing features i.e. a clear trend of error w.r.t to number of features cannot be observed with any classifier. Hence, the sequential forward search can be further used to optimize the number of features to be used. With small testing and training data, we can see that the test total error is least when no feature is selected but this might not be the case with large training data or if try to find different combinations of larger features. This can also be seen as a fact that when no feature is selected, the training resubstitution error is the lowest but it quite high for testing set; in other words lacking consistency.