### Assignment2\_p2

## In this Assignment I use SVM and ELM algorithm to train this EEG Dataset

First I use fixed-elm algorithm to train EEG, in order to see In order to compare the influence of the number of neurons on the network, I set up a for loop of 100, 200, 500, 800, 1000 neurons, recording accuracy.

```
In [51]: # -*- coding:utf-8 -*-
import elm

import load_data

number_neruon= [100,200,500,800,1000]
for i in number_neruon:
    elmc = elm.ELMClassifier(n_hidden=i,activation_func='sigmoid')
    data = load_data.read_data_sets(one_hot=True)

elmc.fit(data.train.data,data.train.labels)
    Accuracy = elmc.score(data.test.data,data.test.labels)

print("The number of ",i ,"neurons","Testing Accuracy is ",Accuracy)

The number of 200 neurons Testing Accuracy is 0.15419419212771815
The number of 500 neurons Testing Accuracy is 0.441336361134049
The number of 800 neurons Testing Accuracy is 0.44606729975227083
```

The number of 1000 neurons Testing Accuracy is 0.509496284062758

# Second, beacuse the performance of ELM is not good, so i use SVM Algorithm, And i use SVM with different kernel to see performance

```
In [52]: import pandas as pd import numpy as np

<font color='black' size=3 face="黑体">First i use 'linear' kernel to test this dataset </font>
```

```
In [53]: from sklearn.svm import SVC
         svclassifier = SVC(kernel='linear')
In [55]: import load data
         # return DataSet class
         data = load_data.read_data_sets(one_hot=False)
         # get train data and labels by batch size
         train_x, train_label = data.train.next_batch(n_samples)
         # get test data
         test_x = data.test.data
         # get test labels
         test labels = data.test.labels
         # get sample number
         n_samples = data.train.num_examples
In [56]: svclassifier.fit(train_x,train_label)
Out[56]: SVC(C=1.0, cache_size=200, class_weight=None, coef0=0.0,
             decision_function_shape='ovr', degree=3, gamma='auto_deprecated',
             kernel='linear', max iter=-1, probability=False, random state=None,
             shrinking=True, tol=0.001, verbose=False)
In [57]: y pred = svclassifier.predict(test x)
In [58]: from sklearn.metrics import classification report, confusion matrix
         print(confusion matrix(test labels,y pred))
         print(classification report(test labels, y pred))
         [[ 9097 7792 1549]
          [ 3720 14622
                       1398]
          [ 2302 1133 16515]]
                       precision recall f1-score
                                                        support
                    0
                            0.60
                                       0.49
                                                 0.54
                                                          18438
                            0.62
                                       0.74
                                                 0.68
                    1
                                                          19740
                    2
                            0.85
                                       0.83
                                                 0.84
                                                          19950
                                                 0.69
                                                          58128
             accuracy
            macro avq
                            0.69
                                       0.69
                                                 0.69
                                                          58128
         weighted avg
                            0.69
                                       0.69
                                                 0.69
                                                          58128
         <font color='black' size=3 face="黑体">Using SVM with 'linear' kernel
```

performace is: 69%

</font>

#### Second i use another kernel 'poly' to test this dataset

```
In [61]: from sklearn.svm import SVC
         svclassifier = SVC(kernel='poly',degree=8)
In [42]: # import load data
         # # return DataSet class
         # data = load data.read_data_sets(one_hot=False)
         # # get train data and labels by batch size
         # train x, train label = data.train.next batch(n samples)
         # # get test data
         \# test x = data.test.data
         # # get test labels
         # test labels = data.test.labels
         # # get sample number
         # n samples = data.train.num examples
In [62]: svclassifier.fit(train x,train label)
         /Library/Frameworks/Python.framework/Versions/3.6/lib/python3.6/site-pack
         ages/sklearn/svm/base.py:193: FutureWarning: The default value of gamma w
         ill change from 'auto' to 'scale' in version 0.22 to account better for u
         nscaled features. Set gamma explicitly to 'auto' or 'scale' to avoid this
         warning.
           "avoid this warning.", FutureWarning)
Out[62]: SVC(C=1.0, cache size=200, class weight=None, coef0=0.0,
             decision function shape='ovr', degree=8, gamma='auto deprecated',
             kernel='poly', max iter=-1, probability=False, random state=None,
             shrinking=True, tol=0.001, verbose=False)
In [63]: y pred = svclassifier.predict(test x)
```

```
In [64]:
         from sklearn.metrics import classification report, confusion matrix
         print(confusion matrix(test labels, y pred))
         print(classification_report(test_labels,y_pred))
         [[ 8690 8269
                         1479]
          [ 2495 15894
                         1351]
          [ 2458
                    816 16676]]
                        precision
                                      recall f1-score
                                                          support
                     0
                              0.64
                                        0.47
                                                  0.54
                                                            18438
                              0.64
                                        0.81
                                                   0.71
                     1
                                                            19740
                     2
                             0.85
                                        0.84
                                                  0.85
                                                            19950
                                                  0.71
                                                            58128
              accuracy
                             0.71
                                        0.70
                                                  0.70
                                                            58128
            macro avg
                                                  0.70
         weighted avg
                             0.71
                                        0.71
                                                            58128
```

Using SVM with 'poly' kernel performace is: 70%

#### Third i use another kernel 'rbf' to test this dataset

```
In [70]: from sklearn.metrics import classification_report, confusion_matrix
    print(confusion_matrix(test_labels,y_pred))
    print(classification_report(test_labels,y_pred))
```

[[10925	5714	1799]			
[ 1474 ]	17657	609]			
[ 685	1060	18205]]			
		precision	recall	f1-score	support
	0	0.83	0.59	0.69	18438
	1	0.72	0.89	0.80	19740
	2	0.88	0.91	0.90	19950
accuracy				0.80	58128
macro	avg	0.81	0.80	0.80	58128
weighted	avg	0.81	0.80	0.80	58128

Using SVM with 'rbf' kernel performace is: 80%

#### Conclusion

In this assigment, I use two algorithm to test this EEG, and especially, in terms of SVM I use several kernel to compare this performace

For Fixed-ELM, the performace is just 50% for 1K neruons, so we can see that is not good and it is not my expection. Then, I use SVM ,for SVM I use several kernels: 'linear' 'poly' 'rbf'

so compare those performace, I found that Using SVM with 'rbf' kernel the performace is best, it is 80%, which is much better than ELM and other algorithms

```
In [ ]:
```