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# -*- coding: utf-8 -*-
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Created on Fri May 8 14:45:10 2020
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@author: mrhaboon
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# -*- coding: utf-8 -*-
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Created on Fri May 8 00:37:07 2020
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@author: mrhaboon
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# -*- coding: utf-8 -*-
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Created on Wed May 6 21:21:40 2020
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@author: mrhaboon
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# -*- coding: utf-8 -*-
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Created on Wed May 6 13:23:47 2020
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@author: mrhaboon
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```
import csv
from sklearn import preprocessing
import numpy as np
from sklearn.neural_network import MLPClassifier
from sklearn.metrics import confusion_matrix
from itertools import combinations
from sklearn.model_selection import cross_val_score
from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
from sklearn.decomposition import PCA
from sklearn.preprocessing import PolynomialFeatures
from sklearn.model_selection import GridSearchCV
import os
```

```
os.chdir('C:/Users/mrhaboon/Desktop/usc_coursework/ee_559/Project')
```

```
#par=20
```

```
#par=7
```

```
#fold=3
```

```
fold=5
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```
poly_count=3
```

```
train1=[]
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```
test1=[]
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```
with open('D_Train1.csv') as csv_file:
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```
    reader=csv.reader(csv_file)
```

```
    j=0
```

```
    for i in reader:
```

```
        if j==0:
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```
            j+=1
```

```
            continue
```

```

        else:
            train1.append([float(x) for x in i])
        train1=np.array(train1)

j=0
with open('D_Test1.csv') as csv_file:
    reader=csv.reader(csv_file)
    for i in reader:
        if j==0:
            j+=1
            continue
        else:
            test1.append([float(x) for x in i])
    test1=np.array(test1)

###

train1_data=train1[:,1:]
label_train=train1[:,0]

test1_data=test1[:,1:]
label_test=test1[:,0]

scaler=preprocessing.StandardScaler().fit(train1_data)
std_train=np.concatenate((label_train[:,np.newaxis],scaler.transform(train1_data)),axis=1)
std_test=np.concatenate((label_test[:,np.newaxis],scaler.transform(test1_data)),axis=1)

normer=preprocessing.MinMaxScaler()
norm_train=np.concatenate((label_train[:,np.newaxis],normer.fit_transform(train1_data)),axis=1)
norm_test=np.concatenate((label_test[:,np.newaxis],normer.fit_transform(test1_data)),axis=1)

layers=[(5,),(5,2),(5,3),(15,),(15,2),(15,3),(50,),(50,2),(50,3),(100,),(100,2)]
param_grid=[{'activation':['identity', 'logistic', 'tanh', 'relu'],'solver':['lbfgs'],'alpha':np.linspace(1e-5,1e-1,10)}]

###
class find_best_mlp:
    def __init__(self,train,test,dim='none'):
        self.train=train
        self.train_data=self.train[:,1:]
        self.train_labels=self.train[:,0]
        self.test=test
        self.test_data=self.test[:,1:]
        self.test_labels=self.test[:,0]
        self.dim=dim
        if dim=='none':
            print('before gen')
            best_model=self.gen_model(self.train_data,self.train_labels)
            print('after gen')
            best_params=best_model[2]
            best_p_stats=best_model[1]
            best_dim_stats=best_model[0]

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self.best={'best_params':best_params,'best_p_stats':best_p_stats,'best_dim_stats':best_
self.final_model=best_model[-1]

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self.final_model.fit(self.train_data,self.train_labels)
self.performance(self.train_data,self.test_data)

elif dim=='perm':
    self.gen_perm()
    best_params=[]
    best_p_stats=(0,0)
    best_dim_stats=(0,0)

    for i in self.perm:
        tmp_train=self.dim_transform(self.train_data,i)
        curr_model=self.gen_model(tmp_train,self.train_labels)
        curr_score=(curr_model[0][0]+curr_model[1][0])/2
        if curr_score>(best_p_stats[0]+best_dim_stats[0]):
            best_params=curr_model[2]
            best_p_stats=curr_model[1]
            best_dim_stats=curr_model[0]
            best_model=curr_model[-1]
            best_dim=i

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self.best={'best_params':best_params,'best_p_stats':best_p_stats,'best_dim_stats':best_
self.final_model=best_model
new_train=self.dim_transform(self.train_data,best_dim)
new_test=self.dim_transform(self.test_data,best_dim)
self.final_model.fit(new_train,self.train_labels)
self.performance(new_train,new_test)

elif dim=='PCA':
    best_params=[]
    best_p_stats=(0,0)
    best_dim_stats=(0,0)

    for i in range(7):
        print('progress')
        tmp=PCA(n_components=i+1)
        tmp.fit(self.train_data)
        tmp_train=tmp.transform(self.train_data)
        print('before gen')

        curr_model=self.gen_model(tmp_train,self.train_labels)
        print('after gen')
        curr_score=(curr_model[0][0]+curr_model[1][0])/2
        if curr_score>(best_p_stats[0]+best_dim_stats[0]):
            best_params=curr_model[2]
            best_p_stats=curr_model[1]
            best_dim_stats=curr_model[0]
            best_model=curr_model[-1]
            best_dim=i+1

self.best={'best_params':best_params,'best_p_stats':best_p_stats,'best_dim_stats':best_
self.final_model=best_model

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        tran=PCA(n_components=best_dim)
        tran.fit(self.train_data)
        new_train=tran.transform(self.train_data)
        new_test=tran.transform(self.test_data)
        self.final_model.fit(new_train,self.train_labels)
        self.performance(new_train,new_test)

elif dim=='Fisher':
    best_params=[]
    best_p_stats=(0,0)
    best_dim_stats=(0,0)

    for i in range(3):
        print('progress')
        tran=LinearDiscriminantAnalysis(n_components=i+1)
        tran.fit(self.train_data,self.train_labels)
        tmp_train=tran.transform(self.train_data)

        print('before gen')
        curr_model=self.gen_model(tmp_train,self.train_labels)
        print('after gen')
        curr_score=(curr_model[0][0]+curr_model[1][0])/2
        if curr_score>(best_p_stats[0]+best_dim_stats[0]):
            best_params=curr_model[2]
            best_p_stats=curr_model[1]
            best_dim_stats=curr_model[0]
            best_model=curr_model[-1]
            best_dim=i+1

    self.best={'best_params':best_params,'best_p_stats':best_p_stats,'best_dim_stats':best_dim}
    self.final_model=best_model

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        tran=LinearDiscriminantAnalysis(n_components=best_dim)
        tran.fit(self.train_data,self.train_labels)
        new_train=tran.transform(self.train_data)
        new_test=tran.transform(self.test_data)
        self.final_model.fit(new_train,self.train_labels)
        self.performance(new_train,new_test)

elif dim=='poly':
    best_params=[]
    best_p_stats=(0,0)
    best_dim_stats=(0,0)

    for i in range(poly_count):
        poly=PolynomialFeatures(i+1)
        tmp_train=poly.fit_transform(self.train_data)

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curr_model=self.gen_model(tmp_train,self.train_labels)
curr_score=(curr_model[0][0]+curr_model[1][0])/2
if curr_score>(best_p_stats[0]+best_dim_stats[0]):
    best_params=curr_model[2]
    best_p_stats=curr_model[1]
    best_dim_stats=curr_model[0]
    best_model=curr_model[-1]
    best_dim=i+1

self.best={'best_params':best_params,'best_p_stats':best_p_stats,'best_dim_stats':best_
self.final_model=best_model

```

```

poly=PolynomialFeatures(best_dim)
new_train=poly.fit_transform(self.train_data)
new_test=poly.fit_transform(self.test_data)
self.final_model.fit(new_train,self.train_labels)
self.performance(new_train,new_test)

def performance(self,train,test):
    self.train_conf=confusion_matrix(self.train_labels,self.final_model.predict(train))
    self.test_conf=confusion_matrix(self.test_labels,self.final_model.predict(test))
    self.acc=self.best
#     self.train_acc=
#     self.test_acc=

def gen_perm(self):
    self.perm=[]
    tmp=[]
    data_length=len(self.train_data[0])
    for i in range(data_length):
        tmp.append(i+1)
    for i in range(data_length):
        for i in combinations(tmp,i+1):
            self.perm.append(i)

def dim_transform(self,data,dim):
    result=[]
    for i in data:
        tmp_data=[]
        for j in dim:
            tmp_data.append(i[j-1])
        result.append(tmp_data)
    return result

def gen_model(self,data,labels):
    tmp=MLPClassifier()
    clf=GridSearchCV(tmp,param_grid)
    clf.fit(data,labels)
    best_index=clf.best_index_
    scores=[]
    for i in range(fold):
        print(self.dim)
        test_model=clf.best_estimator_
        scores.extend(list(cross_val_score(test_model,data,labels,cv=5)))

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        scores=np.array(scores)
        print('out')
        return ((scores.mean(),scores.std()*2),(clf.cv_results_['mean_test_score'][best_index],clf

#%%
#
#test=find_best_svm(train1,test1)

data_sets=[[std_train,std_test],[norm_train,norm_test]]
#data_sets=[[test1,test1],[std_test,std_test],[norm_test,norm_test]]
dim_choice=['none','Fisher','PCA']

file1=open('mlp_results.txt','w')

i=1
for k in data_sets:
    if i==0:
        file1.write('no std:\n')
    elif i==1:
        file1.write('std:\n')
    else:
        file1.write('norm:\n')
    for j in dim_choice:
        print('we movin')
        file1.write('---'+j+':\n')
        tmp=find_best_mlp(k[0],k[1],j)
        file1.write('All stats'+str(tmp.acc)+'\n')
        file1.write('train confusion matrix:'+str(tmp.train_conf)+'\n')
        file1.write('test confusion'+str(tmp.test_conf)+'\n')
    file1.write('-----'+'\n')
    i+=1

file1.close()

#%%
#test_model=SVC(kernel='rbf')
#scores=(cross_val_score(test_model,train1[:,1:],train1[:,0],cv=5))
##
###
###
#
#test=SVC()
#clf=GridSearchCV(test,param_grid)
#clf.fit(train1[:,1:],train1[:,0])
##
##clf.best_estimator_.score(train1[:,1:],train1[:,0])

#test=train1[:,1:]
#label=train1[:,0]
#

```

