In [4]:

```
## import required packages ##
import numpy as np
from sklearn.metrics import roc_auc_score
import pandas as pd
from xgboost.sklearn import XGBClassifier
from sklearn.ensemble import RandomForestClassifier
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import KFold
import tensorflow as tf
from sklearn.preprocessing import OneHotEncoder
from tqdm import tqdm_notebook
import warnings
warnings.filterwarnings("ignore")

### define a function (ses_spe) to calculte sensitivity, specificity,
#and accuracy, and return these results ###
def ses_spe (real_value,pre_value,cut_value=0.5):
    data1=np.concatenate((real_value,pre_value),axis=1)
    real_pred=pd.DataFrame(data1,columns=['real','prediction'])
    real_pred['prediction2']=real_pred.prediction.apply(lambda x: 1 if x>cut_value else 0)
    TP_df=real_pred[real_pred.real==1][real_pred.prediction2==1]
    FP_df=real_pred[real_pred.real==0][real_pred.prediction2==1]
    TN_df=real_pred[real_pred.real==0][real_pred.prediction2==0]
    FN_df=real_pred[real_pred.real==1][real_pred.prediction2==0]
    TP=TP_df.count()[0]
    FP=FP_df.count()[0]
    TN=TN_df.count()[0]
    FN=FN_df.count()[0]
    ses=TP/(TP+FN)
    spe=TN/(TN+FP)
    ac=(TP+TN)/(TP+TN+FP+FN)
    return ses,spe,ac

## load the dataset ##
dataset=np.loadtxt("direction of the dataset",delimiter=',',skiprows=1)
## one-hot one-hot coding ##
enc = OneHotEncoder()
## position of the categorical independent variables that need to be transformed ##
enc.fit(dataset[:,[1,6,8,11]])
trans=enc.transform(dataset[:,[1,6,8,11]]).toarray()
## combine with the rest of the dataset ##
dataset_final=np.concatenate((dataset[:,[0,2,3,4,5,7,10]],trans),axis=1)

## four empty list to save the results for each ML model ##
result_dnn=[]
result_xgb=[]
result_rf=[]
result_lr=[]

## start loop 10 times 5-fold cross-validation ##
for i in tqdm_notebook(range(10),desc='repeat 10 times',leave=False):
    ## split the dataset into 5 subsets##
    ## random_state is defined so that the results are repeatable ##
    fold=KFold(n_splits=5,shuffle=True,random_state=i)
    for train_index,test_index in fold.split(dataset_final):
        ## train set ##
        train=dataset_final[train_index]
        ## test set ##
        test=dataset_final[test_index]
        ## independent variables in the train set ##
        x_train=train[:,1:15]
        ## independent variables in the test set ##
        x_test=test[:,1:15]
        ## dependent variables in the train set ##
        y_train=train[:,0][:,np.newaxis]
        ## dependent variables in the test set ##
        y_test=test[:,0][:,np.newaxis]
        ##DNN model##
        tf.reset_default_graph()
        ## random_seed is defined so that the results are repeatable ##
        tf.set_random_seed(i)
        ## length of the array ##
        n_variables = 14
        ## number of nodes ##
        n = 12
        x=tf.placeholder(tf.float32,[None,n_variables])
        y=tf.placeholder(tf.float32,[None,1])
        prob=tf.placeholder(tf.float32)
        ## learning rate ##
        lr=tf.Variable(0.01,dtype=tf.float32)
        ## define the structure of the newtwork ##
        w1=tf.Variable(tf.truncated_normal([n_variables,n],stddev=0.1))
        b1=tf.Variable(tf.zeros([1,n])+0.1)
        l1=tf.nn.leaky_relu(tf.matmul(x,w1)+b1)
        ## dropput is used ##
        l1_drop=tf.nn.dropout(l1,prob)

        w2=tf.Variable(tf.truncated_normal([n,n],stddev=0.1))
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b2=tf.Variable(tf.zeros([1,n])+0.1)
l2=tf.nn.leaky_relu(tf.matmul(l1_drop,w2)+b2)
## dropput is used ##
l2_drop=tf.nn.dropout(l2,prob)

w3=tf.Variable(tf.truncated_normal([n,n],stddev=0.1))
b3=tf.Variable(tf.zeros([1,n])+0.1)
l3=tf.nn.leaky_relu(tf.matmul(l2_drop,w3)+b3)
## dropput is used ##
l3_drop=tf.nn.dropout(l3,prob)

w4=tf.Variable(tf.truncated_normal([n,n],stddev=0.1))
b4=tf.Variable(tf.zeros([1,n])+0.1)
l4=tf.nn.leaky_relu(tf.matmul(l3_drop,w4)+b4)
## dropput is used ##
l4_drop=tf.nn.dropout(l4,prob)

w5=tf.Variable(tf.truncated_normal([n,n],stddev=0.1))
b5=tf.Variable(tf.zeros([1,n])+0.1)
l5=tf.nn.leaky_relu(tf.matmul(l4_drop,w5)+b5)
## dropput is used ##
l5_drop=tf.nn.dropout(l5,prob)

w6=tf.Variable(tf.truncated_normal([n,1],stddev=0.1))
b6=tf.Variable(tf.zeros([1,1])+0.1)
prediction=tf.nn.sigmoid(tf.matmul(l5_drop,w6)+b6)

## define the loss function and optimizer ##
loss=tf.reduce_mean(tf.nn.sigmoid_cross_entropy_with_logits(labels=y, logits=(tf.matmul(l5_drop,w6)+b6)
train=tf.train.AdamOptimizer(lr).minimize(loss)
init=tf.global_variables_initializer()

## start model fitting ##
with tf.Session() as sess:
    sess.run(init)
    ## 5000 iterations ##
    for step in range(5001):
        sess.run(train,feed_dict={x:x_train,y:y_train,prob:0.9})
        ## when comes to last iteration calculate the auc and predicted probability##
        if step%5000 == 0:
            predic_test=sess.run(prediction,feed_dict={x:x_test,prob:1.0})
            predic_train=sess.run(prediction,feed_dict={x:x_train,prob:1.0})
            auc_test = roc_auc_score(y_test,predic_test)
            auc_train = roc_auc_score(y_train,predic_train)
        ## for every 500 iterations, reduce learning rate ##
        if step%500 == 0:
            sess.run(tf.assign(lr,0.01*(0.95**(step//500))))
## calculate the performance on the train set ##
ses_train,spe_train,ac_train=ses_spe(y_train,predic_train)
## calculate the performance on the test set ##
ses_test,spe_test,ac_test=ses_spe(y_test,predic_test)
## save the results ##
result_dnn.append([auc_train,ses_train,spe_train,ac_train,auc_test,ses_test,spe_test,ac_test])
## print results when finished ##
print('ANN-'+str(i),'Train AUC:',auc_train,'Test AUC:',auc_test)
##XGBoost model##
## seed is defined so that the results are repeatable ##
xgb1 = XGBClassifier(learning_rate =0.1,n_estimators=120,n_jobs=6,max_depth=4,
                     min_child_weight=10,seed=i,gamma=0.1,subsample=1.0,
                     colsample_bytree=0.8,
                     objective= 'binary:logistic',reg_alpha=0.5,reg_lambda=2.5)
## fit the model using train set ##
xgb1.fit(x_train,y_train)
## calculate the predicted probability ##
predic_train=xgb1.predict_proba(x_train)[:,1]
predic_train=predic_train[:,np.newaxis]
## calculate the predicted probability ##
predic_test=xgb1.predict_proba(x_test)[:,1]
predic_test=predic_test[:,np.newaxis]
## calculate the auc ##
auc_train=roc_auc_score(y_train,predic_train)
auc_test=roc_auc_score(y_test,predic_test)
## calculate the performance on the train set ##
ses_train,spe_train,ac_train=ses_spe(y_train,predic_train)
## calculate the performance on the test set ##
ses_test,spe_test,ac_test=ses_spe(y_test,predic_test)
## save the results ##
result_xgb.append([auc_train,ses_train,spe_train,ac_train,auc_test,ses_test,spe_test,ac_test])
## print results when finished ##
print('XGBoost-'+str(i),'Train AUC:',auc_train,'Test AUC:',auc_test)
##Random Forest model##
## random_state is defined so that the results are repeatable ##
rf=RandomForestClassifier(n_estimators=320,criterion='gini',random_state=i,
                          min_samples_leaf=80,min_samples_split=40,n_jobs=6)
## fit the model using train set ##
rf.fit(x_train,y_train)
## calculate the predicted probability ##
predic_test=rf.predict_proba(x_test)[:,1]
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        predic_train=rf.predict_proba(x_train)[:,1]
        predic_test=predic_test[:,np.newaxis]
        predic_train=predic_train[:,np.newaxis]
        ## calculate the auc ##
        auc_test=roc_auc_score(y_test,predic_test)
        auc_train=roc_auc_score(y_train,predic_train)
        ## calculate the performance on the train set ##
        ses_train,spe_train,ac_train=ses_spe(y_train,predic_train)
        ## calculate the performance on the test set ##
        ses_test,spe_test,ac_test=ses_spe(y_test,predic_test)
        ## save the results ##
        result_rf.append([auc_train,ses_train,spe_train,ac_train,auc_test,ses_test,spe_test,ac_test])
        ## print results when finished ##
        print('RF-'+str(i),'Train AUC:',auc_train,'Test AUC:',auc_test)
        ##Logistice regression model##
        ## random_state is defined so that the results are repeatable ##
        lr=LogisticRegression(random_state=i,n_jobs=6,penalty='none')
        ## fit the model using train set ##
        lr.fit(x_train,y_train)
        ## calculate the predicted probability ##
        predic_test=lr.predict_proba(x_test)[:,1]
        predic_train=lr.predict_proba(x_train)[:,1]
        predic_test=predic_test[:,np.newaxis]
        predic_train=predic_train[:,np.newaxis]
        ## calculate the auc ##
        auc_test=roc_auc_score(y_test,predic_test)
        auc_train=roc_auc_score(y_train,predic_train)
        ## calculate the performance on the train set ##
        ses_train,spe_train,ac_train=ses_spe(y_train,predic_train)
        ## calculate the performance on the test set ##
        ses_test,spe_test,ac_test=ses_spe(y_test,predic_test)
        ## save the results ##
        result_lr.append([auc_train,ses_train,spe_train,ac_train,auc_test,ses_test,spe_test,ac_test])
        ## print results when finished ##
        print('LR-'+str(i),'Train AUC:',auc_train,'Test AUC:',auc_test)
```

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ANN-0 Train AUC: 0.7449184999138322 Test AUC: 0.7305673132495837
XGBoost-0 Train AUC: 0.7715926607069278 Test AUC: 0.7438529331549248
RF-0 Train AUC: 0.7430629853013933 Test AUC: 0.7298707129544025
LR-0 Train AUC: 0.6340363816960541 Test AUC: 0.6293938740526974
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XGBoost-0 Train AUC: 0.7706127761983608 Test AUC: 0.7412348303658428
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ANN-1 Train AUC: 0.73275792163817 Test AUC: 0.7398154572968184
XGBoost-1 Train AUC: 0.7679542919574929 Test AUC: 0.7579879924214463
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ANN-1 Train AUC: 0.7406628803959281 Test AUC: 0.7200957444085935
XGBoost-1 Train AUC: 0.7742934119873555 Test AUC: 0.7362439390747825
RF-1 Train AUC: 0.7453851979824604 Test AUC: 0.7223392464471118
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ANN-1 Train AUC: 0.7307542859653101 Test AUC: 0.7276199659012159
XGBoost-1 Train AUC: 0.7687723207171817 Test AUC: 0.7593594468594469
RF-1 Train AUC: 0.7383904276508827 Test AUC: 0.7452971437346437
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ANN-1 Train AUC: 0.7449332574737554 Test AUC: 0.7142972876472834
XGBoost-1 Train AUC: 0.773425950668391 Test AUC: 0.7283293729454206
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LR-1 Train AUC: 0.6313151468422928 Test AUC: 0.6335967299097525
ANN-2 Train AUC: 0.7397802972478678 Test AUC: 0.7343877340105119
XGBoost-2 Train AUC: 0.7716151364511755 Test AUC: 0.7497632827417862
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RF-2 Train AUC: 0.7412571407477135 Test AUC: 0.7379382770485736
LR-2 Train AUC: 0.6334869613099907 Test AUC: 0.6297663736289694
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XGBoost-2 Train AUC: 0.7725901663954883 Test AUC: 0.7258260725724903
RF-2 Train AUC: 0.7464068646406588 Test AUC: 0.7092140359721509
LR-2 Train AUC: 0.6283772156096559 Test AUC: 0.6455646753359281
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XGBoost-2 Train AUC: 0.7687201777745184 Test AUC: 0.7495909917784918
RF-2 Train AUC: 0.7416678961531613 Test AUC: 0.7332880028192529
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XGBoost-6 Train AUC: 0.7727682597211547 Test AUC: 0.7385000185000186
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LR-7 Train AUC: 0.6331246316948479 Test AUC: 0.627593308923404
ANN-7 Train AUC: 0.7402921145781272 Test AUC: 0.7417832882609959
XGBoost-7 Train AUC: 0.7708321476227653 Test AUC: 0.7549380070637959
RF-7 Train AUC: 0.7419342096301007 Test AUC: 0.737515455132089
LR-7 Train AUC: 0.6329201133286009 Test AUC: 0.6316013733516578
ANN-7 Train AUC: 0.745244326118379 Test AUC: 0.7360596488947195
XGBoost-7 Train AUC: 0.7731304244990319 Test AUC: 0.7377170277393656
RF-7 Train AUC: 0.743883569530088 Test AUC: 0.7318599380680542
LR-7 Train AUC: 0.6296195697724541 Test AUC: 0.6428007560267426
ANN-8 Train AUC: 0.736906141531295 Test AUC: 0.7409594885877108
XGBoost-8 Train AUC: 0.7677503375265543 Test AUC: 0.7560198718516551
RF-8 Train AUC: 0.740965504367788 Test AUC: 0.7408531904213542
LR-8 Train AUC: 0.6325316688011766 Test AUC: 0.6313180972628222
ANN-8 Train AUC: 0.7461050732718431 Test AUC: 0.7102980445654432
XGBoost-8 Train AUC: 0.773041569777368 Test AUC: 0.7310226783951401
RF-8 Train AUC: 0.7460953215350647 Test AUC: 0.7097510406116602
LR-8 Train AUC: 0.6374600817115559 Test AUC: 0.6183177180067895
ANN-8 Train AUC: 0.740688385972751 Test AUC: 0.7444977725270128
XGBoost-8 Train AUC: 0.7689427179830054 Test AUC: 0.7502057643555097
RF-8 Train AUC: 0.7400739858136689 Test AUC: 0.744507371581875
LR-8 Train AUC: 0.6295644329863608 Test AUC: 0.6401973959487066
ANN-8 Train AUC: 0.742660879815843 Test AUC: 0.7237005026629226
XGBoost-8 Train AUC: 0.7741761089831953 Test AUC: 0.7392572655806313
RF-8 Train AUC: 0.7437347398393517 Test AUC: 0.7280831780659043
LR-8 Train AUC: 0.6306137010978856 Test AUC: 0.6353865543819864
ANN-8 Train AUC: 0.745080798757571 Test AUC: 0.722524755401268
XGBoost-8 Train AUC: 0.7754745006396475 Test AUC: 0.733065995358239
RF-8 Train AUC: 0.7444805055098443 Test AUC: 0.7216113273938758
LR-8 Train AUC: 0.6328292986165327 Test AUC: 0.6296107304271761
ANN-9 Train AUC: 0.7386601403320477 Test AUC: 0.7390334486463278
XGBoost-9 Train AUC: 0.769307177501687 Test AUC: 0.7488357917354754
RF-9 Train AUC: 0.7404762178703617 Test AUC: 0.7389460530164048
LR-9 Train AUC: 0.6315593874444156 Test AUC: 0.63403412319288
ANN-9 Train AUC: 0.7418151505281384 Test AUC: 0.7339019961022855
XGBoost-9 Train AUC: 0.7690022232741662 Test AUC: 0.7532207819051555
RF-9 Train AUC: 0.7417447349513392 Test AUC: 0.7304523711096675
LR-9 Train AUC: 0.6288121006015563 Test AUC: 0.6442550542333509
ANN-9 Train AUC: 0.7401534286261507 Test AUC: 0.7321000959622056
XGBoost-9 Train AUC: 0.7676988862310765 Test AUC: 0.7506151423439384
RF-9 Train AUC: 0.7416393346906163 Test AUC: 0.7354334292955391
LR-9 Train AUC: 0.6328796242741048 Test AUC: 0.6241140719962599
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XGBoost-9 Train AUC: 0.7713769994047353 Test AUC: 0.7401304357671039
RF-9 Train AUC: 0.7430722731264929 Test AUC: 0.7253571010160974
LR-9 Train AUC: 0.6338369344024511 Test AUC: 0.620462338628933
ANN-9 Train AUC: 0.7495290495395871 Test AUC: 0.7043034797445062
XGBoost-9 Train AUC: 0.776232656524765 Test AUC: 0.7174783673453304
RF-9 Train AUC: 0.7482315001674897 Test AUC: 0.7099446180398213
LR-9 Train AUC: 0.6366357335866764 Test AUC: 0.6299949169172089
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