```
import pandas as pd
import numpy as np
import random
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy score, confusion matrix, auc
from sklearn.ensemble import RandomForestClassifier
from sklearn.ensemble import BaggingClassifier
from sklearn.tree import DecisionTreeClassifier
from sklearn import svm
from sklearn.linear_model import LogisticRegression
import matplotlib.pyplot as plt
from sklearn.metrics import roc curve
data = pd.read csv('C:/Data analysis Project/practical/default.csv')
data['industry'] = data['industry'].str.replace('Other Services \((except Public Administratio)\)
data['industry'] = data['industry'].str.replace('Accomodation','Accommodation')
data['industry'] = data['industry'].str.replace('Fishing or Hunting','Fishing and Hunting')
data['industry'] = data['industry'].fillna('Other Services')
data['industry'] = data['industry'].str.strip()
test = data.groupby('industry').size()
data = data.dropna(subset=['years_of_operation'], how='any')
data = data.dropna(subset=['fico score'], how='any')
data = data[data['fico_score'] != 0]
data = data.dropna(subset=['business employee count'], how='any')
data = data.dropna(subset=['Homeowner'], how='any')
data['principal'] = data['principal'].str.replace('$','').str.replace(',','')
data['principal'] = data['principal'].astype(int)
data = data[data['status'] != 'current']
data['status'] = np.where(data['status'] == 'defaulted', 1, 0)
data.head(5)
```



#Descriptive Statistics data.describe()

term

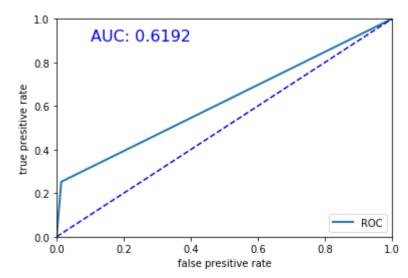
```
4709.000000
                                                4709.000000
                                                             4709.000000 4709.000000
count
               4709.000000
                                                                                         4709.00
                                                   0.141907
                 10.649529
                             138240.356764
                                                                40.885114
                                                                               0.086855
                                                                                          706.1°
mean
 std
                  7.991417
                             104803.995393
                                                   0.044720
                                                                14.908321
                                                                               0.281652
                                                                                           44.5
min
                   0.000000
                              25000.000000
                                                   0.049900
                                                                 6.000000
                                                                               0.000000
                                                                                          604.00
25%
                   5.000000
                              60000.000000
                                                   0.111900
                                                                36.000000
                                                                               0.000000
                                                                                          672.00
50%
                  8.600000
                             100000.000000
                                                   0.139900
                                                                36.000000
                                                                               0.000000
                                                                                          703.00
75%
                 13.800000
                             200000.000000
                                                   0.162900
                                                                60.000000
                                                                               0.000000
                                                                                          738.00
                 67.000000
                             500000.000000
                                                                60.000000
max
                                                   0.287900
                                                                               1.000000
                                                                                          843.00
```

```
#dummy
data = pd.get_dummies(data,columns=['credit_rating'],prefix=['credit_rating'],prefix_sep="_",
data = pd.get dummies(data,columns=['industry'],prefix=['industry'],prefix sep=" ",dummy na=F
data = data.reset index(drop=True)
#split into train and test
random.seed(110)
train,test=train_test_split(data,test_size=0.3)
train_y = train['status']
train_x = train.drop(['status','state'],axis=1)
test y = test['status']
test_x = test.drop(['status','state'],axis=1)
#Logistic regression
lm = LogisticRegression(solver='liblinear')
lm.fit(train x,train y)
predict_lm = lm.predict(test_x)
accuracy lm = accuracy score(test y, predict lm)
tn_lm, fp_lm, fn_lm, tp_lm = confusion_matrix(test_y, predict_lm).ravel()
sensitivity_lm = tp_lm / (tp_lm+fn_lm)
print("Accuracy is %s" %accuracy lm)
print("Sensitivity is %s" %sensitivity_lm)
print(confusion matrix(test y, predict lm))
     Accuracy is 0.9285208775654635
     Sensitivity is 0.2522522522522523
     [[1284
              18]
        83
              28]]
      П
#Logistic regression ROC
fpr lm,tpr lm,thresholds = roc curve(test y, predict lm,pos label=1)
auc_lm = round(auc(fpr_lm, tpr_lm),4)
plt.plot(fpr lm,tpr lm,linewidth=2,label="ROC")
plt.xlabel("false presitive rate")
plt.ylabel("true presitive rate")
```

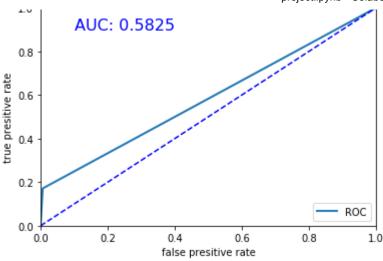
```
plt.yllm(0,1)
plt.xlim(0,1)
plt.plot([0, 1], [0, 1], '--', color=(0, 0, 1))
plt.legend(loc=4)
plt.text(0.1,0.9,'AUC: %s' %auc_lm,fontdict={'size':'16','color':'b'})
plt.show()
```



#random forest

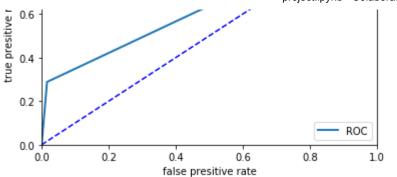


```
rf = RandomForestClassifier(n estimators=1000, random state=110)
rf.fit(train_x,train_y)
predict rf = rf.predict(test x)
accuracy_rf = accuracy_score(test_y, predict_rf)
tn_rf, fp_rf, fn_rf, tp_rf = confusion_matrix(test_y, predict_rf).ravel()
sensitivity_rf = tp_rf / (tp_rf+fn_rf)
print("Accuracy is %s" %accuracy_rf)
print("Sensitivity is %s" %sensitivity rf)
print(confusion_matrix(test_y, predict_rf))
     Accuracy is 0.9292285916489739
     Sensitivity is 0.17117117117117
     [[1294
               81
      92
              19]]
#random forest ROC
fpr_rf,tpr_rf,thresholds = roc_curve(test_y, predict_rf,pos_label=1)
auc rf = round(auc(fpr rf, tpr rf),4)
plt.plot(fpr_rf,tpr_rf,linewidth=2,label="ROC")
plt.xlabel("false presitive rate")
plt.ylabel("true presitive rate")
plt.ylim(0,1)
plt.xlim(0,1)
plt.plot([0, 1], [0, 1], '--', color=(0, 0, 1))
plt.legend(loc=4)
plt.text(0.1,0.9,'AUC: %s' %auc_rf,fontdict={'size':'16','color':'b'})
plt.show()
```



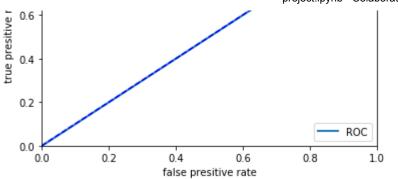
```
#bagging
tree = DecisionTreeClassifier(criterion='entropy', max_depth=None)
bag = BaggingClassifier(base estimator=tree, n estimators=1000, max samples=1.0, max features
bag.fit(train_x,train_y)
predict bag = bag.predict(test x)
accuracy_bag = accuracy_score(test_y, predict_bag)
tn_bag, fp_bag, fn_bag, tp_bag = confusion_matrix(test_y, predict_bag).ravel()
sensitivity bag = tp bag / (tp bag+fn bag)
print("Accuracy is %s" %accuracy_bag)
print("Sensitivity is %s" %sensitivity bag)
print(confusion_matrix(test_y, predict_bag))
     Accuracy is 0.9292285916489739
     Sensitivity is 0.2882882882882883
     [[1281
              21]
       79
              32]]
      Γ
#bagging ROC
fpr_bag,tpr_bag,thresholds = roc_curve(test_y, predict_bag,pos_label=1)
auc_bag = round(auc(fpr_bag, tpr_bag),4)
plt.plot(fpr_bag,tpr_bag,linewidth=2,label="ROC")
plt.xlabel("false presitive rate")
plt.ylabel("true presitive rate")
plt.ylim(0,1)
plt.xlim(0,1)
plt.plot([0, 1], [0, 1], '--', color=(0, 0, 1))
plt.legend(loc=4)
plt.text(0.1,0.9,'AUC: %s' %auc_bag,fontdict={'size':'16','color':'b'})
plt.show()
```





```
#svm
sv = svm.SVC(gamma='auto')
sv.fit(train_x,train_y)
predict svm = sv.predict(test x)
accuracy_svm = accuracy_score(test_y, predict_svm)
tn svm, fp svm, fn svm, tp svm = confusion matrix(test y, predict svm).ravel()
sensitivity_svm = tp_svm / (tp_svm+fn_svm)
print("Accuracy is %s" %accuracy_svm)
print("Sensitivity is %s" %sensitivity_svm)
print(confusion_matrix(test_y, predict_svm))
     Accuracy is 0.9200283085633404
     Sensitivity is 0.0
     [[1300
               2]
      [ 111
               011
#svm ROC
fpr_svm,tpr_svm,thresholds = roc_curve(test_y, predict_svm,pos_label=1)
auc_svm = round(auc(fpr_svm, tpr_svm),4)
plt.plot(fpr_svm,tpr_svm,linewidth=2,label="ROC")
plt.xlabel("false presitive rate")
plt.ylabel("true presitive rate")
plt.ylim(0,1)
plt.xlim(0,1)
plt.plot([0, 1], [0, 1], '--', color=(0, 0, 1))
plt.legend(loc=4)
plt.text(0.1,0.9,'AUC: %s' %auc_svm,fontdict={'size':'16','color':'b'})
plt.show()
```





#GITHUB LINK https://github.com/13065040/Project/blob/master/project.ipynb