

RandomForestServiceTime10FoldTrTest

November 19, 2018

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In [79]: import numpy as np
import matplotlib.pyplot as plt
from sklearn.model_selection import train_test_split
from sklearn import metrics
from sklearn.metrics import classification_report
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import precision_score, f1_score, confusion_matrix
import pandas as pd
from sklearn.model_selection import StratifiedKFold
from sklearn.metrics import roc_curve, auc
import matplotlib.patches as patches
from scipy import interp
```

```
import warnings
warnings.filterwarnings('ignore')
# Set random seed
np.random.seed(0)
```

```
In [68]: # Random Forest ML model for predicting Service Time (350-476 OR 476-600)
# 80-20% train test split
data = pd.read_csv("ServiceTimeBinaryBothFinal.csv")
data.head()
```

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X = data.ix[:,(0,1,2,3,4,6)].values
y = data.ix[:,5].values
```

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X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = .2, random_state=0)
clf = RandomForestClassifier(n_estimators=10, max_depth=4, random_state=0)
clf.fit(X_train, y_train)
```

```
Out[68]: RandomForestClassifier(bootstrap=True, class_weight=None, criterion='gini',
                                max_depth=4, max_features='auto', max_leaf_nodes=None,
                                min_impurity_decrease=0.0, min_impurity_split=None,
                                min_samples_leaf=1, min_samples_split=2,
                                min_weight_fraction_leaf=0.0, n_estimators=10, n_jobs=None,
                                oob_score=False, random_state=0, verbose=0, warm_start=False)
```

```
In [69]: # Collecting F-1 score, precision, recall
y_pred = clf.predict(X_test)
print(classification_report(y_test, y_pred))
```

	precision	recall	f1-score	support
1	1.00	0.93	0.97	30
2	0.90	1.00	0.95	18
micro avg	0.96	0.96	0.96	48
macro avg	0.95	0.97	0.96	48
weighted avg	0.96	0.96	0.96	48

```
In [70]: # Collecting Precision score (Accuracy)
print('Random Forest Model accuracy for train-test split: ', round(precision_score(y_test, y_pred), 2))
```

Random Forest Model accuracy for train-test split: 96.25 %

```
In [71]: print('Random Forest Model F-1 score for train-test split: ', round(f1_score(y_test, y_pred), 3))
```

Random Forest Model F-1 score for train-test split: 0.959

```
In [72]: print('Random Forest Model Confusion Matrix for train-test split: \n ', confusion_matrix(y_test, y_pred))
```

Random Forest Model Confusion Matrix for train-test split:

```
[[28  2]
 [ 0 18]]
```

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In [73]: # Random Forest ML model for predicting Service Time (350-476 OR 476-600)
# 10-fold cross validation
clf = RandomForestClassifier(n_estimators=100, max_depth=4, random_state=0)
cv = StratifiedKFold(n_splits=10, shuffle=False)
print('Random Forest Accuracy:', round(np.mean(cross_val_score(clf, X, y, cv=10)), 2) * 100)
```

Random Forest Accuracy: 95.5 %

```
In [82]: # ROC curve
data = pd.read_csv("ServiceTimeBinaryBothFinal.csv")
dict = {2:1, 1:0} # label = column name
data['Service Time'] = data['Service Time'].map(dict)

random_state = np.random.RandomState(0)
clf = RandomForestClassifier(n_estimators=100, max_depth=4, random_state=0)
```

```

cv = StratifiedKFold(n_splits=10,shuffle=False)

x = data.loc[:, data.columns != 'Service Time']
y = data.loc[:, 'Service Time']
fig1 = plt.figure(figsize=[12,12])
ax1 = fig1.add_subplot(111,aspect = 'equal')
ax1.add_patch(
    patches.Arrow(0.45,0.5,-0.25,0.25,width=0.3,color='green',alpha = 0.5)
)
ax1.add_patch(
    patches.Arrow(0.5,0.45,0.25,-0.25,width=0.3,color='red',alpha = 0.5)
)

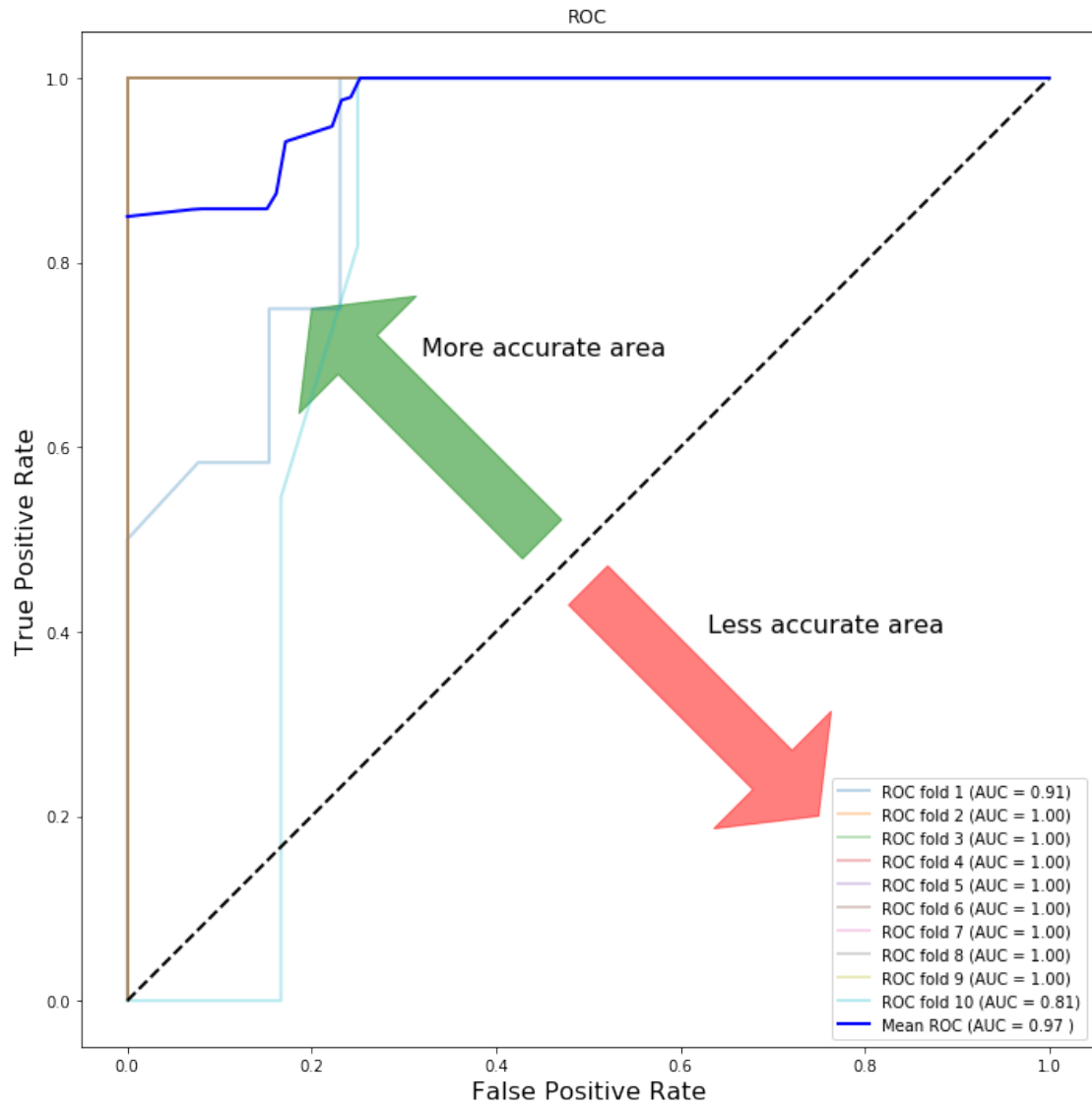
tprs = []
aucs = []
mean_fpr = np.linspace(0,1,100)

i = 1
for train,test in cv.split(x,y):
    prediction = clf.fit(x.iloc[train],y.iloc[train]).predict_proba(x.iloc[test])[:, 1]
    fpr, tpr, t = roc_curve(y[test], prediction)
    tprs.append(interp(mean_fpr, fpr, tpr))
    roc_auc = auc(fpr, tpr)
    aucs.append(roc_auc)
    plt.plot(fpr, tpr, lw=2, alpha=0.3, label='ROC fold %d (AUC = %0.2f)' % (i, roc_auc))
    i= i+1

plt.plot([0,1],[0,1],linestyle = '--',lw = 2,color = 'black')
mean_tpr = np.mean(tprs, axis=0)
mean_auc = auc(mean_fpr, mean_tpr)
plt.plot(mean_fpr, mean_tpr, color='blue',
        label=r'Mean ROC (AUC = %0.2f )' % (mean_auc),lw=2, alpha=1)

plt.xlabel('False Positive Rate',fontsize = 16)
plt.ylabel('True Positive Rate',fontsize = 16)
plt.title('ROC')
plt.legend(loc="lower right")
plt.text(0.32,0.7,'More accurate area',fontsize = 16)
plt.text(0.63,0.4,'Less accurate area',fontsize = 16)
plt.show()

```



In []: