Voting_classifier_teste

November 4, 2019

```
[1]: import pandas as pd
     import numpy as np
     import seaborn as sns
     from datetime import datetime
     import matplotlib.pyplot as plt
     %matplotlib inline
     from sklearn.model_selection import StratifiedKFold, GridSearchCV
     from sklearn.ensemble import RandomForestClassifier, __
     →GradientBoostingClassifier, AdaBoostClassifier, VotingClassifier, U
     →BaggingClassifier
     from sklearn.metrics import roc_auc_score, roc_curve, auc, u
     →precision_recall_curve
     from sklearn.metrics import classification_report, confusion_matrix
     from xgboost import XGBClassifier
     from mlxtend.plotting import plot_learning_curves
     from yellowbrick.model_selection import LearningCurve
     import matplotlib.gridspec as gridspec
     import itertools
     from sklearn.model_selection import cross_val_score, train_test_split
     from sklearn.tree import DecisionTreeClassifier
     from sklearn.metrics import accuracy_score
     from sklearn.linear model import LogisticRegression
     from sklearn import tree
     from sklearn.naive_bayes import MultinomialNB
     from sklearn.preprocessing import LabelEncoder, OrdinalEncoder
```

```
[3]: train = pd.read_csv('trainAG.csv', dtype={'id': np.int32, 'target': np.int8})
X_train = train.iloc[:,1:105]
```

```
Y_train = train.loc[:, train.columns == 'Y']
     test = pd.read_csv('testAG.csv', dtype={'id': np.int32})
     X_{\text{test}} = \text{test.iloc}[:,1:105]
     Y_test = test.loc[:, test.columns == 'Y']
     \#all\_features = [x for x in train.drop(['ID', 'Y'], axis=1).columns]
[4]: print(X_train.shape)
    (109992, 104)
[5]: print(X_test.shape)
    (65995, 104)
[6]: X train.head()
[6]:
        v1
            v2
                v4
                    v5
                        v6
                             v9
                                 v11
                                     v13
                                           v14
                                                v15
                                                         v50
                                                              v54 v56 v66
                                                                             v68
                         0
                                   0
                                             0
                                                         0.0
                                                              0.0 0.0
                                                                        0.0
                                                                              0.0
     1
         0
             0
                 0
                     0
                         0
                              0
                                   0
                                        0
                                             0
                                                         0.0
                                                              0.0 0.0 0.0
                                                                              0.0
                                                  0
                                                     •••
     2
         0
             0
                 0
                     0
                         0
                              0
                                   0
                                        0
                                             0
                                                  0
                                                         0.0
                                                              0.0 0.0
                                                                        0.0
                                                                              0.0
                         0
                                   0
                                             0
                                                         0.0
                                                              0.0 0.0 0.0
     3
         0
             0
                 0
                     0
                              0
                                        0
                                                  0
                                                                             0.0
                                   0
         0
             0
                 0
                     0
                         0
                              0
                                        0
                                             0
                                                         0.0
                                                              0.0 0.0 0.0 0.0
        v70
                v72
                        v74 v76
                                      v78
     0.0
                0.0
                        0.0 0.0
                                      0.0
     1 0.0
                0.0
                        0.0 0.0
                                      0.0
     2 0.0
                        0.0 0.0
                                      0.0
                0.0
     3 0.0 4330.0
                     2260.0 0.0 1800.0
     4 0.0
                        0.0 0.0
                                      0.0
                0.0
     [5 rows x 104 columns]
```

Voting Ensemble

```
[7]: from sklearn.ensemble import VotingClassifier
     from sklearn.svm import SVC
     starttime = timer(None)
     start_time = timer(None)
     model1 = RandomForestClassifier(n_estimators=200, max_depth=2,_
     ⇔criterion='gini', n_jobs=4, random_state=1)
     model2 = DecisionTreeClassifier(class_weight=None, criterion='entropy',__
     →max_depth=None, max_features=None, max_leaf_nodes=None,)
     model3 = AdaBoostClassifier(n_estimators=200, random_state=1, learning_rate=1)
     model4 = XGBClassifier(base score=0.2, booster='gbtree', colsample bylevel=1,
                   colsample_bynode=0.1, colsample_bytree=1, gamma=1,
                   learning rate=0.1, max delta step=1, max depth=3,
                   min_child_weight=1, missing=None, n_estimators=300, n_jobs=4,
```

```
nthread=None, objective='reg:squarederror', random_state=1,
              reg_alpha=0, reg_lambda=10, scale_pos_weight=1, seed=None,
              silent=None, subsample=1, verbosity=1)
model = VotingClassifier(estimators=[('bg', model1), ('dtc', model2), ('ada', __
→model3), ('xgb', model4)], voting='hard')
model.fit(X train, Y train.values.ravel())
model.score(X_test,Y_test.values.ravel())
predictions_model = model.predict(X_test)
scores = cross_val_score(model, X_train, Y_train.values.ravel(), cv=3)
timer(start_time)
false_positive_rate, true_positive_rate, thresholds = roc_curve(Y_test,__
→predictions_model)
roc_auc = auc(false_positive_rate, true_positive_rate)
matrix_model = confusion_matrix(Y_test, predictions_model)
plt.figure(1)
plt.figure(figsize=(9,5))
model_heatmap = sns.heatmap(matrix_model,annot=True, cbar=False, fmt="d", cmap_
→='coolwarm', linecolor ='black', linewidths = 1)
bottom, top = model heatmap.get ylim()
model_heatmap.set_ylim(bottom + 0.5, top - 0.5)
plt.ylabel('True Label')
plt.xlabel('Predicted Label')
plt.title('Confusion Matrix')
plt.show()
plt.figure(2)
plt.figure(figsize=(9,5))
plt.title('Receiver Operating Characteristic')
plt.plot(false_positive_rate, true_positive_rate, 'b',
label='AUC = %0.2f'% roc_auc)
plt.legend(loc='lower right')
plt.plot([0,1],[0,1],'r--')
plt.xlim([-0.0,1.0])
plt.ylim([-0.0,1.0])
plt.ylabel('True Positive Rate')
plt.xlabel('False Positive Rate')
plt.show()
plt.figure(3)
precision, recall, thresholds = precision_recall_curve(Y_test,__
→predictions_model)
plt.figure(figsize = (9,5))
plt.plot(recall, precision)
```

```
plt.plot([0, 1], [0.5, 0.5], linestyle = '--')
plt.xlabel('Recall', fontsize = 16)
plt.ylabel('Precision', fontsize = 16)
plt.xticks(size = 18)
plt.yticks(size = 18)
plt.title('Precision-Recall', fontsize = 28)
plt.show();

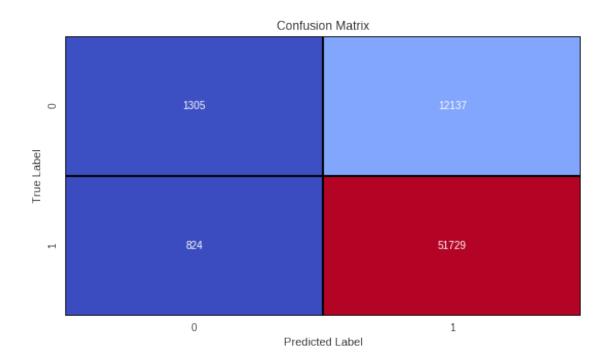
print("Classification Report")
print(classification_report(Y_test, predictions_model))

print("Acurácia do Modelo")
print(accuracy_score(Y_test, predictions_model))

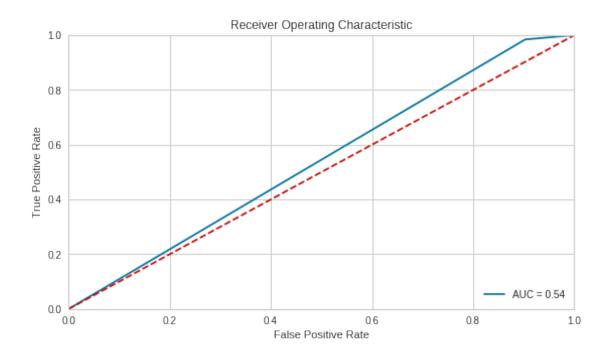
print("Acurácia do Modelo Cross Validation")
print(scores.mean())
```

Tempo Necessário: 3 minutos and 5.96 segundos.

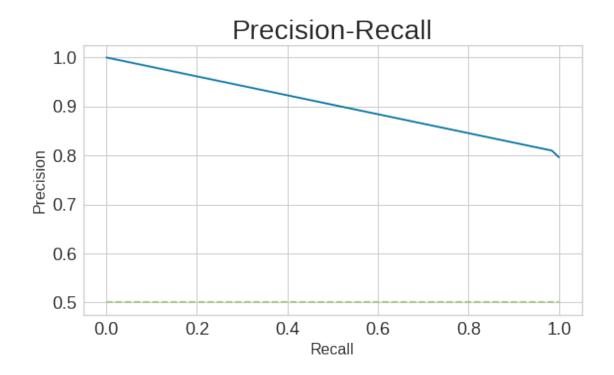
<Figure size 432x288 with 0 Axes>



<Figure size 432x288 with 0 Axes>



<Figure size 432x288 with 0 Axes>



Classification Report

	precision	recall	f1-score	support
0	0.61	0.10	0.17	13442
1	0.81	0.98	0.89	52553
accuracy			0.80	65995
macro avg	0.71	0.54	0.53	65995
weighted avg	0.77	0.80	0.74	65995

Acurácia do Modelo Cross Validation 0.7995854245433551

[]: