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
```

```
[2]: def timer(start_time=None):
    if not start_time:
        start_time = datetime.now()
        return start_time
    elif start_time:
        tmin, tsec = divmod((datetime.now() - start_time).total_seconds(), 60)
        print('\n Tempo Necessário: %i minutos and %s segundos.' % (tmin, u round(tsec, 2)))
```

```
[3]: train = pd.read_csv('trainRF.csv', dtype={'id': np.int32, 'target': np.int8})
X_train = train.iloc[:,1:68]
```

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Y_train = train.loc[:, train.columns == 'Y']
     test = pd.read_csv('testRF.csv', dtype={'id': np.int32})
     X_test = test.iloc[:,1:68]
     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, 67)
[5]: print(X_test.shape)
    (65995, 67)
[6]: X train.head()
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     [5 rows x 67 columns]
```

Voting Ensemble

```
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)
```

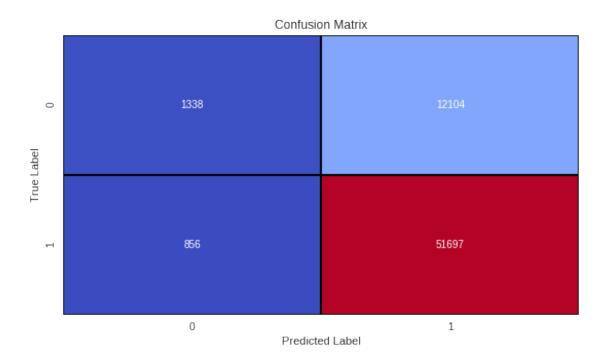
```
plt.plot([0, 1], [0.5, 0.5], linestyle = '--')
plt.xlabel('Recall', fontsize = 16)
plt.ylabel('Precision', fontsize = 16)
plt.xticks(size = 18)
plt.yticks(size = 18)
plt.title('Precision-Recall', fontsize = 28)
plt.show();

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

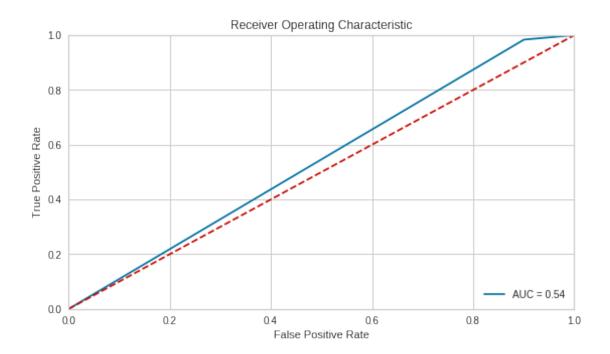
print("Acurácia do Modelo Cross Validation")
print(scores.mean())
```

Tempo Necessário: 2 minutos and 26.33 segundos.

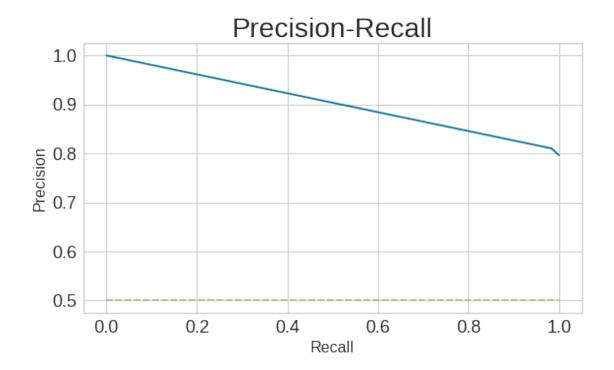
<Figure size 432x288 with 0 Axes>



<Figure size 432x288 with 0 Axes>



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Classification Report

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

Acurácia do Modelo Cross Validation 0.8005309322324199

[]:[