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
     from sklearn.utils import shuffle
```

```
[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, □
        →round(tsec, 2)))
```

```
[3]: train = pd.read_csv('trainLR.csv')
     train = shuffle(train)
     X_train = train.iloc[:,1:25]
     Y_train = train.loc[:, train.columns == 'Y']
     test = pd.read_csv('testLR.csv')
     test = shuffle(test)
     X \text{ test} = \text{test.iloc}[:,1:25]
     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, 24)
[5]: print(X_test.shape)
    (65995, 24)
[6]: X_train.head()
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     [5 rows x 24 columns]
```

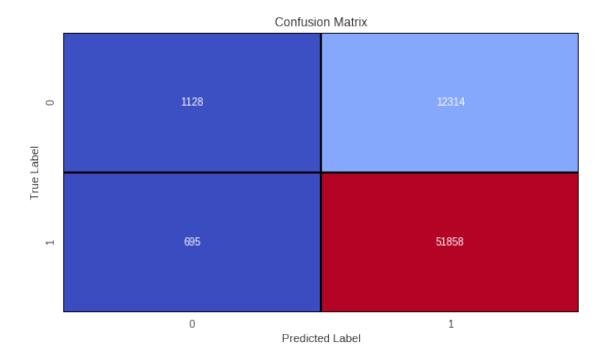
Voting Ensemble

```
[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,
```

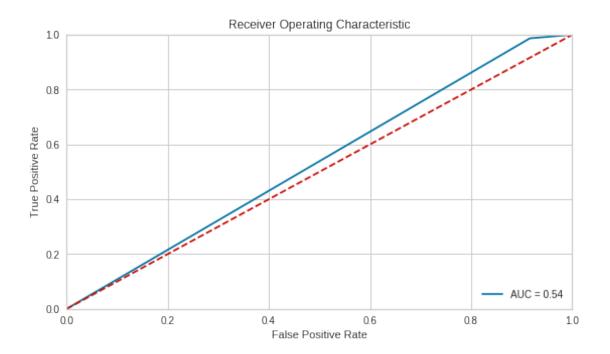
```
colsample_bynode=0.1, colsample_bytree=1, gamma=1,
              learning_rate=0.1, max_delta_step=1, max_depth=3,
              min_child_weight=1, missing=None, n_estimators=300, n_jobs=4,
              nthread=None, objective='reg:squarederror', random_state=1,
              reg_alpha=0, reg_lambda=10, scale_pos_weight=1, seed=None,
              silent=None, subsample=1, verbosity=1)
model = VotingClassifier(estimators=[('bg', model1), ('dtc', model2), ('ada', u
→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)
```

Tempo Necessário: O minutos and 58.53 segundos.

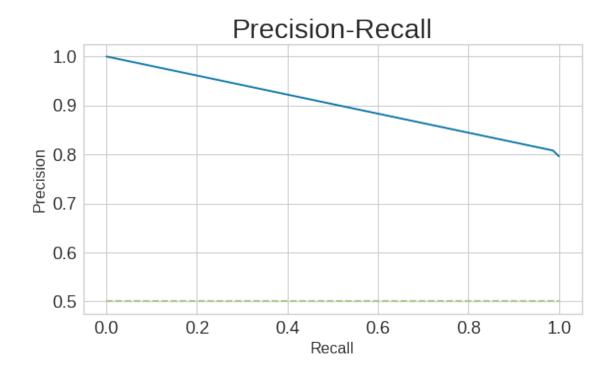
<Figure size 432x288 with 0 Axes>



<Figure size 432x288 with 0 Axes>



<Figure size 432x288 with 0 Axes>



Classification Report

	precision	recall	f1-score	support
0	0.62	0.08	0.15	13442
1	0.81	0.99	0.89	52553
accuracy			0.80	65995
macro avg	0.71	0.54	0.52	65995
weighted avg	0.77	0.80	0.74	65995

Acurácia do Modelo Cross Validation 0.8000490985961202

[]: