Notebook Predictive analytics

January 24, 2023

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[99]: import numpy as np
       import pandas as pd
       import matplotlib as plt
       from sklearn.pipeline import Pipeline
       from sklearn.model_selection import GridSearchCV, train_test_split
       from sklearn.neighbors import KNeighborsClassifier
       from sklearn.multiclass import OneVsRestClassifier
       from sklearn.metrics import precision_recall_curve, roc_curve, accuracy_score,_
        →precision_score, recall_score
       from sklearn import datasets, neighbors
       from mlxtend.plotting import plot_decision_regions
       from sklearn.tree import DecisionTreeClassifier
       from sklearn.naive_bayes import MultinomialNB
       from sklearn.ensemble import RandomForestClassifier
       from sklearn.model_selection import RandomizedSearchCV
       from sklearn.naive_bayes import GaussianNB
[100]: df userdata = pd.read csv('final userdata min 40.csv')
       df_userdata = df_userdata.drop(df_userdata.columns[0:11],axis =1 )
       df_userdata = df_userdata.drop('liked_recipes',axis =1 )
       df_userdata.kitchen.unique()
[100]: array(['frans', 'aziatisch', 'hollands', 'mexicaans', 'italiaans',
              'mediterraan', 'amerikaans'], dtype=object)
[101]: for keuken in range(len(df_userdata)):
           if df_userdata['kitchen'].iloc[keuken] == 'aziatisch':
               df userdata['kitchen'].iloc[keuken] = 0
           elif df_userdata['kitchen'].iloc[keuken] == 'frans':
               df userdata['kitchen'].iloc[keuken] = 1
           elif df_userdata['kitchen'].iloc[keuken] == 'hollands':
               df_userdata['kitchen'].iloc[keuken] = 2
           elif df_userdata['kitchen'].iloc[keuken] == 'italiaans':
               df_userdata['kitchen'].iloc[keuken] = 3
           elif df_userdata['kitchen'].iloc[keuken] == 'mexicaans':
               df_userdata['kitchen'].iloc[keuken] = 4
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elif df_userdata['kitchen'].iloc[keuken] == 'mediterraan':
        df_userdata['kitchen'].iloc[keuken] = 5
    elif df_userdata['kitchen'].iloc[keuken] == 'amerikaans':
        df_userdata['kitchen'].iloc[keuken] = 6
df_userdata
/tmp/ipykernel_43015/3680158733.py:6: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-
docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
  df_userdata['kitchen'].iloc[keuken] = 1
/tmp/ipykernel_43015/3680158733.py:4: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-
docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
  df_userdata['kitchen'].iloc[keuken] = 0
/tmp/ipykernel_43015/3680158733.py:8: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-
docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
  df_userdata['kitchen'].iloc[keuken] = 2
/tmp/ipykernel_43015/3680158733.py:12: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-
docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
  df_userdata['kitchen'].iloc[keuken] = 4
/tmp/ipykernel_43015/3680158733.py:10: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-
docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
  df_userdata['kitchen'].iloc[keuken] = 3
/tmp/ipykernel_43015/3680158733.py:14: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-
docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
  df userdata['kitchen'].iloc[keuken] = 5
/tmp/ipykernel_43015/3680158733.py:16: SettingWithCopyWarning:
A value is trying to be set on a copy of a slice from a DataFrame
See the caveats in the documentation: https://pandas.pydata.org/pandas-
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docs/stable/user_guide/indexing.html#returning-a-view-versus-a-copy
df_userdata['kitchen'].iloc[keuken] = 6

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1266	0	0	0	1	0	0	
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       [281 rows x 171 columns]
[278]: model = RandomForestClassifier()
       model.fit(X_train,y_train)
       y_proba = model.predict(X_val)
[279]: print('Acc_score',accuracy_score(y_proba,y_val))
       print('prec score', precision score(y proba, y val, average = 'macro'))
       print('rec_score',recall_score(y_proba,y_val, average ='macro'))
      Acc_score 0.9644128113879004
      prec_score 0.9644599303135888
      rec_score 0.9649198332491933
[280]: #params = { 'criterion' : ['qini', 'entropy', 'log_loss'],
                   'max_depth': [3,5,8,10,15,20,25,30,40,50,60,70,80,90,100],
       #
                   'min_samples_split' : [5,6,7,8,9,10,20,30,40,50,60]
[281]: | #params = { 'leaf_size' : [1,2,3,4,5,6,7,8,9,10,20,30,40,50,60,70,80,90],
                   'n_neighbors' : [1,2,3,4,5,6,7,8,9,10],
       #
                   'weights' : ['uniform', 'distance'],
       #
                   'algorithm' : ['auto', 'ball_tree', 'kd_tree', 'brute']
[286]: params = { 'n_estimators' : [10,20,30,40,50,80,100,130,150,200,250,300,350,400],
                  'criterion' : ['gini', 'entropy', 'log_loss'],
                  'max_depth' : [3,5,8,10,15,20,25,30,40,50,60,70,80,90,100],
                  'min_samples_split' : [5,6,7,8,9,10,20,30,40,50,60]
[287]: randomsearch = RandomizedSearchCV(estimator= model ,param_distributions=__
        →params, n_iter=30, cv=20, random_state=42)
[288]: search = randomsearch.fit(X_train,y_train)
       search.best_estimator_
[288]: RandomForestClassifier(criterion='entropy', max_depth=20, min_samples_split=5,
                              n estimators=350)
[289]: model.score(X val, y val)
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[289]: 0.9644128113879004
[293]: | #model = KNeighborsClassifier(algorithm='brute', leaf_size=1, n_neighbors=8,__
       →weights='distance')
       #model = DecisionTreeClassifier(max depth=25, min samples split=20)
       #model = MultinomialNB()
      RandomForestClassifier(criterion='entropy', max_depth=20, min_samples_split=5,__
        \rightarrown_estimators=350)
[293]: RandomForestClassifier(criterion='entropy', max_depth=20, min_samples_split=5,
                             n estimators=350)
[294]: model.fit(X_train,y_train)
      y_proba = model.predict(X_test)
[295]: print('Acc_score',accuracy_score(y_proba,y_test))
      print('prec_score',precision_score(y_proba,y_test, average ='macro'))
      print('rec_score',recall_score(y_proba,y_test, average ='macro'))
      Acc_score 0.9786476868327402
      prec_score 0.9785714285714285
      rec_score 0.9791568754557977
[296]: from sklearn.metrics import confusion_matrix, classification_report
      print(confusion_matrix(y_test, y_proba))
      print(classification_report(y_test, y_proba))
      [[41 0 0 0 0 0 0]
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```

weighted avg [271]: model.score(X_test,y_test) [271]: 0.8540925266903915 [272]: from sklearn.utils.multiclass import unique_labels unique_labels(y_test) [272]: array([0, 1, 2, 3, 4, 5, 6]) [273]: def plot(Y_test, y_proba): labels= unique_labels(y_test) columns=[f"Predicted {label}" for label in labels] index=[f"Actual {label}" for label in labels] table= pd.DataFrame(confusion_matrix(y_test, y_proba), columns=columns,__ →index=index) return table [274]: plot(y_test, y_proba) [274]: Predicted 0 Predicted 1 Predicted 2 Predicted 3 Predicted 4 39 Actual 0 0 0 0 Actual 1 0 36 0 4 0 Actual 2 2 4 33 1 0 Actual 3 0 0 34 0 1 0 0 39 Actual 4 0 0 Actual 5 0 0 1 4 4 Actual 6 1 1 0 0 6 Predicted 5 Predicted 6 Actual 0 0 0 Actual 1 0 0

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[275]: import seaborn as sns
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[276]: def plot2(Y_test, y_proba):
           labels= unique_labels(y_test)
           columns=[f"Predicted {label}" for label in labels]
           index=[f"Actual {label}" for label in labels]
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table= pd.DataFrame(confusion_matrix(y_test, y_proba), columns=columns,udindex=index)

heatmap=sns.heatmap(table, annot=True, fmt="d", cmap= "Blues")
heatmap.set_xlabel('Predicted labels');
heatmap.set_ylabel('True labels');
heatmap.set_title('Confusion Matrix KNearest Neighbours');
heatmap.xaxis.set_ticklabels(["Aziatisch", "Frans", "Hollands",uditaliaans", "Mexicaans", "Mediterraan", "Amerikaans"]);
heatmap.yaxis.set_ticklabels(["Aziatisch", "Frans", "Hollands",uditaliaans", "Mexicaans", "Mediterraan", "Amerikaans"], rotation=0);
return heatmap
```

[277]: plot2(y_test, y_proba)

