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# -*- coding: utf-8 -*-
"""Copy of Copy of Project_Main.ipynb
Automatically
generated by Colaboratory.
Original file is located at
https://colab.research.google.com/drive/1VBAbhTMwx5D-rT8fwX1z9fjKRMho2_KH
f
rom google.colab import drive
drive.mount('/content/drive')
import numpy as np
import pandas
as pd
import os
import matplotlib.pyplot as plt
import seaborn as sns
data_path =
'/content/drive/MyDrive/data science '
train_path = os.path.join(data_path,
'train.csv')
test_path = os.path.join(data_path, 'test.csv')
my_data =
pd.read_csv(train_path)
my_data
mis_val = my_data.isnull().sum()
mis_val = mis_val[mis_val
> 0]
mis_val
after_drop = my_data.drop(columns=['SellerID',
'ExpressID'])
after_drop
col_datatype =
after_drop.dtypes
print(col_datatype)
after_drop['ExportationCountry'].unique()
after drop.nu
nique()
after_drop.count()
# checking the unique in the column
'ProcessType'
after_drop['ProcessType'].unique()
# to know how many times each value occurs in
the 'PaymentType' column
after_drop['ProcessType'].value_counts()
# to know how many times
each value occurs in the 'PaymentType' column
after_drop['PaymentType'].value_counts()
after_drop['ProcessType'].value_counts()
checking the unique in the column 'ProcessType'
after_drop['DisplayIndicator'].unique()
```

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#t.o
know how many times each value occurs in the 'DisplayIndicator' column
after_drop['DisplayIndicator'].value_counts()
### checking the unique in the column
'TransactionNature'
after_drop['TransactionNature'].value_counts()
after_drop['Type'].value_
counts()
after_drop['PaymentType'].value_counts()
after_drop['BorderTransportMeans'].value_co
unts()
after_drop['IssueDateTime'].value_counts()
after_drop['Fake'].value_counts()
after_dr
op['TaxRate'].unique()
after_drop[after_drop.TaxRate ==
'0.000e+00']
sns.countplot(x='TaxRate', hue='Fake', data=after_drop)
#important
final_data =
after_drop.copy()
final_data.corr()
final_data['IssueDateTime'] =
pd.to_datetime(final_data['IssueDateTime']).dt.month
final_data['ClassificationID'] =
final_data['ClassificationID'].astype(str)
final_data.loc[final_data['ClassificationID'].str.le
n() == 10, 'ClassificationID'] =
final_data['ClassificationID'].str[:2]
final_data.loc[final_data['ClassificationID'].str.len()
== 9, 'ClassificationID'] = final_data['ClassificationID'].str[0]
encode_label =
['DutyRegime', 'ClassificationID', 'ExportationCountry', 'OriginCountry']
col_unique_dict =
{col: list(final_data[col].unique()) for col in encode_label}
encode_target =
['IssueDateTime',
'DisplayIndicator', 'BorderTransportMeans'
,'PaymentType','DeclarationOfficeID']
for col in
encode_target:
  final_data[col] =
final_data.groupby(col)["Fake"].transform("mean")
for col in encode_label:
 final_data[col] = pd.factorize(final_data[col], sort=True)[0]
col_num_dict = {col:
list(final_data[col].unique()) for col in encode_label}
dropped_cols = ['TransactionNature',
'DeclarerID', 'ImporterID', 'Type', 'ProcessType']
final_data =
final_data.drop(columns=dropped_cols)
feature = final_data.columns[1:-1]
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labl =
'Fake'
feature
from sklearn.model_selection import train_test_split
X_train, X_test, y_train,
y test = train test split(final data[feature], final data[labl], test size=0.2,
random_state=42)
from sklearn.model_selection import GridSearchCV
from sklearn.neighbors
import KNeighborsClassifier
prediction = KNeighborsClassifier()
param_search = {'n_neighbors' :
list(range(300, 305)), 'weights' : ['distance']}
grid_search = GridSearchCV(prediction ,
param_search, cv=5,
scoring='accuracy',return_train_score=True)
grid_search.fit(final_data[feature],
final_data[labl])
res = grid_search.cv_results_
params_score_list = [(params, outcome) for
(params, outcome) in zip(res['params'], res['mean_test_score'])]
outcomes = [outcome for
(params, outcome) in params_score_list] # get the scores for plotting with neighbours
neighbors
= [params['n_neighbors'] for (params, outcome) in zip(res['params'], res['mean_test_score'])] #
get the neighbors for plotting with scores
for (params, outcome) in params_score_list:
print('K:', params['n_neighbors'], ': Accuracy:', outcome)
av_score =
np.mean(outcomes)
grid_search.best_estimator_,
grid_search.best_score_
params_score_list
outcomes
neighbors
outcomes
# neighbors versus
accuracy
plt.plot(neighbors, outcomes)
plt.title('K versus score
plot')
plt.xlabel('K')
plt.ylabel('accuracies')
plt.show()
final_data
data test =
pd.read_csv(test_path)
test_after = data_test.drop(columns=['SellerID',
'ExpressID'])
test_final = test_after.copy()
test final['IssueDateTime'] =
pd.to datetime(test final['IssueDateTime']).dt.month
test_final['ClassificationID'] =
(test_final['ClassificationID'] // (1e8)).astype(int)
for col in encode_target:
test_final[col] = final_data.groupby(col)["Fake"].transform("mean")
for col
in encode_label:
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test_final[col] = test_final[col].apply(lambda x:
col_num_dict[col][col_unique_dict[col].index(x)] if x in col_unique_dict[col] else
-1)
test_final = test_final.drop(columns=dropped_cols)
test_final.columns
feature =
test_final.columns[1:]
predic =
grid_search.best_estimator_.predict(test_final[feature])
result = pd.DataFrame({'ID':
test_final['ID'], 'Fake': prediction})
filename = f"result_{prediction}_{int(av_score *
100)}.csv"
output_directory = "/content/drive/MyDrive/data science "
output_path
= os.path.join(output_directory, filename)
result.to_csv(output_path, index=False)
```