

```

import pickle

def load_and_print_artifacts_dict(path):
    artifacts_dict = pickle.load(open(path, "rb"))

    print("Target encoder mapping:")
    print([ac for ac in artifacts_dict["encoder"].mapping])

    print("Columns to train:")
    print([ac for ac in artifacts_dict["columns_to_score"]])

if __name__ == "__main__":
    load_and_print_artifacts_dict("./Artifacts/artifacts_dict_file.pkl")

```

Target encoder mapping:

```
['City', 'State', 'Bank', 'BankState', 'RevLineCr', 'LowDoc', 'NewExist']
```

Columns to train:

```
['City_trg', 'State_trg', 'Zip_trg', 'Bank_trg', 'BankState_trg', 'NAICS_trg', 'NoEmp_trg', 'NewExist_trg', 'CreateJob_trg', 'RetainedJob_trg', 'FranchiseCode_trg', 'UrbanRural_trg', 'RevLineCr_trg', 'LowDoc_trg', 'DisbursementGross_trg', 'BalanceGross_trg', 'GrAppv_trg', 'SBA_Appv_trg', 'Zip', 'NAICS', 'NoEmp', 'CreateJob', 'RetainedJob', 'FranchiseCode', 'UrbanRural', 'DisbursementGross', 'BalanceGross', 'GrAppv', 'SBA_Appv', 'Log_DisbursementGross', 'Log_NoEmp', 'Log_GrAppv', 'Log_SBA_Appv', 'Log_BalanceGross', 'Disbursement_Bins', 'Loan_Efficiency', 'Guarantee_Ratio', 'Loan_Guarantee_Interaction', 'Disbursement_Squared']
```

```

from matplotlib import pyplot as plt
import shap
from sklearn.inspection import permutation_importance
from sklearn.metrics import confusion_matrix
from sklearn.metrics import roc_auc_score
import lightgbm as lgb
import warnings
import numpy as np
import pandas as pd
warnings.filterwarnings("ignore", category=Warning)

```

```
def scoring(data):
```

```
    """
```

*Function to score input dataset.*

*Input: dataset in Pandas DataFrame format*

*Output: Python list of labels in the same order as input records*

*Flow:*

```

- Load artifacts
- Transform dataset
- Score dataset
- Return labels

"""
artifacts_dict_file =
open("D:/Work/Gre/UTD/Courses/Fall/MIS6341/Softwares/Python/ml-fall-
2023/Project2/artifacts/artifacts_dict_file.pkl", "rb")
artifacts_dict = pickle.load(file=artifacts_dict_file)
artifacts_dict_file.close()
best_classifier = artifacts_dict["best_classifier"]
encoder = artifacts_dict["encoder"]
scaler = artifacts_dict["scaler"]
threshold = artifacts_dict["optimal_threshold"]
numerical_columns = artifacts_dict["numerical_columns"]
cat_cols = artifacts_dict["cat_cols"]
columns_to_score = artifacts_dict["columns_to_score"]

for i in data['RevLineCr']:
    if i not in ['Y', 'N']:
        data['RevLineCr'].replace(i, 'N', inplace=True)
for i in data['LowDoc']:
    if i not in ['Y', 'N']:
        data['LowDoc'].replace(i, 'N', inplace=True)
for i in data['NewExist']:
    if i not in [1, 2]:
        data['NewExist'].replace(i, None, inplace=True)

for column in cat_cols:
    data[column] = data[column].fillna(data[column].mode()[0])

data_encoded = encoder.transform(data)
data_encoded = data_encoded.add_suffix('_trg')
data_encoded = pd.concat([data_encoded, data], axis=1)
for column in cat_cols:
    data_encoded[column + "_trg"].fillna(data_encoded[column +
"_trg"].mean(), inplace=True)
data_encoded.drop(columns=cat_cols, inplace=True)

data_encoded['Log_DisbursementGross'] =
np.log1p(data_encoded['DisbursementGross'])
data_encoded['Log_NoEmp'] = np.log1p(data_encoded['NoEmp'])
data_encoded['Log_GrAppv'] = np.log1p(data_encoded['GrAppv'])
data_encoded['Log_SBA_Appv'] = np.log1p(data_encoded['SBA_Appv'])
data_encoded['Log_BalanceGross'] =
np.log1p(data_encoded['BalanceGross'])

data_encoded['Disbursement_Bins'] =
pd.cut(data_encoded['DisbursementGross'],

```

```

bins=[-np.inf, 50000,
150000, np.inf],
labels=['Low',
'Medium', 'High'])

data_encoded['Loan_Efficiency'] =
data_encoded['DisbursementGross'] / (data_encoded['CreateJob'] +
data_encoded['RetainedJob'] + 1) # Adding 1 to avoid division by zero

data_encoded['Guarantee_Ratio'] = data_encoded['SBA_Appv'] /
data_encoded['GrAppv']

data_encoded['Loan_Guarantee_Interaction'] =
data_encoded['SBA_Appv'] * data_encoded['GrAppv']

data_encoded['Disbursement_Squared'] =
data_encoded['DisbursementGross'] ** 2

data_encoded[numerical_columns] =
scaler.transform(data_encoded[numerical_columns])

y_prob =
best_classifier.predict_proba(data_encoded[columns_to_score])
y_pred = (y_prob[:,0] < threshold).astype(int)
d = {
    "index" : data_encoded.index,
    "label" : y_pred,
    "probability_0": y_prob[:,0],
    "probability_1": y_prob[:,1]
}

return pd.DataFrame(d)

import pandas as pd
df2 =
pd.read_csv("D:/Work/Gre/UTD/Courses/Fall/MIS6341/Softwares/Python/ml-
fall-2023/Project2/SBA_loans_project_2_holdout_students_valid.csv")
print(scoring(df2))

[LightGBM] [Warning] min_data_in_leaf is set=300, min_child_samples=20
will be ignored. Current value: min_data_in_leaf=300
[LightGBM] [Warning] feature_fraction is set=0.9, colsample_bytree=1.0
will be ignored. Current value: feature_fraction=0.9
[LightGBM] [Warning] lambda_l1 is set=9.408025110972025, reg_alpha=0.0
will be ignored. Current value: lambda_l1=9.408025110972025
[LightGBM] [Warning] lambda_l2 is set=3.9690665922792114e-08,
reg_lambda=0.0 will be ignored. Current value:
lambda_l2=3.9690665922792114e-08

```

[LightGBM] [Warning] bagging\_fraction is set=1.0, subsample=1.0 will be ignored. Current value: bagging\_fraction=1.0

[LightGBM] [Warning] bagging\_freq is set=5, subsample\_freq=0 will be ignored. Current value: bagging\_freq=5

|       | index | label | probability_0 | probability_1 |
|-------|-------|-------|---------------|---------------|
| 0     | 0     | 0     | 0.849864      | 0.150136      |
| 1     | 1     | 1     | 0.410310      | 0.589690      |
| 2     | 2     | 0     | 0.999752      | 0.000248      |
| 3     | 3     | 0     | 0.694401      | 0.305599      |
| 4     | 4     | 0     | 0.689322      | 0.310678      |
| ...   | ...   | ...   | ...           | ...           |
| 98904 | 98904 | 0     | 0.697117      | 0.302883      |
| 98905 | 98905 | 1     | 0.178386      | 0.821614      |
| 98906 | 98906 | 0     | 0.868871      | 0.131129      |
| 98907 | 98907 | 0     | 0.751461      | 0.248539      |
| 98908 | 98908 | 1     | 0.597262      | 0.402738      |

[98909 rows x 4 columns]