

# Home Credit Default Risk : Flask API Deployment Pipeline

## 1. Importing the Necessary Libraries

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In [2]: import warnings
warnings.filterwarnings("ignore")

import pandas as pd
import matplotlib.pyplot as plt
import re
import time
import numpy as np
import gc
import xgboost as xgb
import lightgbm as lgb
import seaborn as sns
import math
import pickle
import os

from lightgbm import LGBMClassifier
from sklearn.metrics import roc_auc_score
from scipy.stats import randint as sp_randint
from sklearn.model_selection import KFold, StratifiedKFold
from prettytable import PrettyTable
from sklearn.metrics import roc_curve, auc
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import normalize
from sklearn.feature_selection import SelectKBest
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from sklearn.feature_selection import f_classif
from sklearn.metrics import confusion_matrix
from sklearn.metrics.classification import accuracy_score, log_loss
from sklearn.linear_model import SGDClassifier
from collections import Counter
from scipy.sparse import hstack
from sklearn.calibration import CalibratedClassifierCV
from sklearn.model_selection import train_test_split
from sklearn.model_selection import GridSearchCV
from sklearn.model_selection import RandomizedSearchCV
from sklearn.preprocessing import StandardScaler
from sklearn.ensemble import RandomForestClassifier
from bayes_opt import BayesianOptimization
from sklearn import model_selection
from sklearn.linear_model import LogisticRegression
from datetime import datetime

```

## 2. Class which has all the Necessary Functions Defined

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In [1]: class initial_function_definition:

        def reduce_memory_usage(df):

            start_mem = df.memory_usage().sum() / 1024**2
            print('Memory usage of dataframe is {:.2f} MB'.format(start_mem
            ))

            for col in df.columns:
                col_type = df[col].dtype

                if col_type != object:
                    c_min = df[col].min()
                    c_max = df[col].max()
                    if str(col_type)[:3] == 'int':
                        if c_min > np.iinfo(np.int8).min and c_max < np.iinfo(
np.int8).max:
                            df[col] = df[col].astype(np.int8)

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        elif c_min > np.iinfo(np.int16).min and c_max < np.
iinfo(np.int16).max:
            df[col] = df[col].astype(np.int16)
        elif c_min > np.iinfo(np.int32).min and c_max < np.
iinfo(np.int32).max:
            df[col] = df[col].astype(np.int32)
        elif c_min > np.iinfo(np.int64).min and c_max < np.
iinfo(np.int64).max:
            df[col] = df[col].astype(np.int64)
        else:
            if c_min > np.finfo(np.float16).min and c_max < np.
finfo(np.float16).max:
                df[col] = df[col].astype(np.float16)
            elif c_min > np.finfo(np.float32).min and c_max < n
p.finfo(np.float32).max:
                df[col] = df[col].astype(np.float32)
            else:
                df[col] = df[col].astype(np.float64)

    end_mem = df.memory_usage().sum() / 1024**2
    print('Memory usage after optimization is: {:.2f} MB'.format(en
d_mem))
    print('Decreased by {:.1f}%'.format(100 * (start_mem - end_mem)
/ start_mem))

    return df

def fix_nulls_outliers(data):

    #Replace NA with the most frequently occurring class for Count o
f Client Family Members
    data['CNT_FAM_MEMBERS'].fillna(data['CNT_FAM_MEMBERS'].value_co
unts().idxmax(), \
                                   inplace=True)
    data.replace(max(data['DAYS_EMPLOYED'].values), np.nan, inplace
=True)
    data['NAME_FAMILY_STATUS'].fillna('Data_Not_Available', inplace
=True)

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data['NAME_HOUSING_TYPE'].fillna('Data_Not_Available', inplace=
True)
data['FLAG_MOBIL'].fillna('Data_Not_Available', inplace=True)
data['FLAG_EMP_PHONE'].fillna('Data_Not_Available', inplace=Tru
e)
data['FLAG_CONT_MOBILE'].fillna('Data_Not_Available', inplace=T
rue)
data['FLAG_EMAIL'].fillna('Data_Not_Available', inplace=True)
data['OCCUPATION_TYPE'].fillna('Data_Not_Available', inplace=Tr
ue)

    #Replace NA with the most frequently occurring class for Count o
f Client Family Members
    data['CNT_FAM_MEMBERS'].fillna(data['CNT_FAM_MEMBERS'].value_co
unts().idxmax(), \
                                inplace=True)
data.replace(max(data['DAYS_EMPLOYED'].values), np.nan, inplace
=True)

data['CODE_GENDER'].replace('XNA', 'M', inplace=True)
data['AMT_ANNUITY'].fillna(0, inplace=True)
data['AMT_GOODS_PRICE'].fillna(0, inplace=True)
data['NAME_TYPE_SUITE'].fillna('Unaccompanied', inplace=True)
data['NAME_FAMILY_STATUS'].replace('Unknown', 'Married', inplace
=True)
data['OCCUPATION_TYPE'].fillna('Data_Not_Available', inplace=Tr
ue)

data['EXT_SOURCE_1'].fillna(0, inplace=True)
data['EXT_SOURCE_2'].fillna(0, inplace=True)
data['EXT_SOURCE_3'].fillna(0, inplace=True)

    return data

def FE_application_data(data):

    data['CREDIT_INCOME_PERCENT'] = data['AMT_CREDIT'] / data['AMT_
INCOME_TOTAL']

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data['ANNUITY_INCOME_PERCENT'] = data['AMT_ANNUITY'] / data['AMT_INCOME_TOTAL']
data['CREDIT_ANNUITY_PERCENT'] = data['AMT_CREDIT'] / data['AMT_ANNUITY']

data['FAMILY_CNT_INCOME_PERCENT'] = data['AMT_INCOME_TOTAL'] / data['CNT_FAM_MEMBERS']
data['CREDIT_TERM'] = data['AMT_ANNUITY'] / data['AMT_CREDIT']
data['BIRTH_EMPLOYED_PERCENT'] = data['DAYS_EMPLOYED'] / data['DAYS_BIRTH']
data['CHILDREN_CNT_INCOME_PERCENT'] = data['AMT_INCOME_TOTAL'] / data['CNT_CHILDREN']

data['CREDIT_GOODS_DIFF'] = data['AMT_CREDIT'] - data['AMT_GOODS_PRICE']
data['EMPLOYED_REGISTRATION_PERCENT'] = data['DAYS_EMPLOYED'] / data['DAYS_REGISTRATION']
data['BIRTH_REGISTRATION_PERCENT'] = data['DAYS_BIRTH'] / data['DAYS_REGISTRATION']
data['ID_REGISTRATION_DIFF'] = data['DAYS_ID_PUBLISH'] - data['DAYS_REGISTRATION']

data['ANNUITY_LENGTH_EMPLOYED_PERCENT'] = data['CREDIT_TERM'] / data['DAYS_EMPLOYED']

data['AGE_LOAN_FINISH'] = data['DAYS_BIRTH'] * (-1.0/365) + \
    (data['AMT_CREDIT'] / data['AMT_ANNUITY']) * (1.0 / 12)
     #(This basically refers to the client's age when he/she finishes loan repayment)

data['CAR_AGE_EMP_PERCENT'] = data['OWN_CAR_AGE'] / data['DAYS_EMPLOYED']
data['CAR_AGE_BIRTH_PERCENT'] = data['OWN_CAR_AGE'] / data['DAYS_BIRTH']
data['PHONE_CHANGE_EMP_PERCENT'] = data['DAYS_LAST_PHONE_CHANGE'] / data['DAYS_EMPLOYED']
data['PHONE_CHANGE_BIRTH_PERCENT'] = data['DAYS_LAST_PHONE_CHANGE'] / data['DAYS_BIRTH']

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income_by_contract = data[['AMT_INCOME_TOTAL', 'NAME_CONTRACT_T
YPE']].groupby('NAME_CONTRACT_TYPE').median()['AMT_INCOME_TOTAL']
data['MEDIAN_INCOME_CONTRACT_TYPE'] = data['NAME_CONTRACT_TYPE'
].map(income_by_contract)

income_by_suite = data[['AMT_INCOME_TOTAL', 'NAME_TYPE_SUITE']]
.groupby('NAME_TYPE_SUITE').median()['AMT_INCOME_TOTAL']
data['MEDIAN_INCOME_SUITE_TYPE'] = data['NAME_TYPE_SUITE'].map(
income_by_suite)

income_by_housing = data[['AMT_INCOME_TOTAL', 'NAME_HOUSING_TYP
E']].groupby('NAME_HOUSING_TYPE').median()['AMT_INCOME_TOTAL']
data['MEDIAN_INCOME_HOUSING_TYPE'] = data['NAME_HOUSING_TYPE'].
map(income_by_housing)

income_by_org = data[['AMT_INCOME_TOTAL', 'ORGANIZATION_TYPE']]
.groupby('ORGANIZATION_TYPE').median()['AMT_INCOME_TOTAL']
data['MEDIAN_INCOME_ORG_TYPE'] = data['ORGANIZATION_TYPE'].map(
income_by_org)

income_by_occu = data[['AMT_INCOME_TOTAL', 'OCCUPATION_TYPE']].
groupby('OCCUPATION_TYPE').median()['AMT_INCOME_TOTAL']
data['MEDIAN_INCOME_OCCU_TYPE'] = data['OCCUPATION_TYPE'].map(i
ncome_by_occu)

income_by_education = data[['AMT_INCOME_TOTAL', 'NAME_EDUCATION
_TYPE']].groupby('NAME_EDUCATION_TYPE').median()['AMT_INCOME_TOTAL']
data['MEDIAN_INCOME_EDU_TYPE'] = data['NAME_EDUCATION_TYPE'].ma
p(income_by_education)

data['ORG_TYPE_INCOME_PERCENT'] = data['MEDIAN_INCOME_ORG_TYPE'
]/data['AMT_INCOME_TOTAL']
data['OCCU_TYPE_INCOME_PERCENT'] = data['MEDIAN_INCOME_OCCU_TYP
E']/data['AMT_INCOME_TOTAL']
data['EDU_TYPE_INCOME_PERCENT'] = data['MEDIAN_INCOME_EDU_TYPE'
]/data['AMT_INCOME_TOTAL']

data= data.drop(['FLAG_DOCUMENT_2', 'FLAG_DOCUMENT_4', 'FLAG_DOCU
MENT_5', 'FLAG_DOCUMENT_6', 'FLAG_DOCUMENT_7',
'FLAG_DOCUMENT_8', 'FLAG_DOCUMENT_9', 'FLAG_DOCUMENT_10', 'FLAG_D

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DOCUMENT_11', 'FLAG_DOCUMENT_12', 'FLAG_DOCUMENT_13',
    'FLAG_DOCUMENT_14', 'FLAG_DOCUMENT_15', 'FLAG_DOCUMENT_16', 'FLAG_
DOCUMENT_17', 'FLAG_DOCUMENT_18', 'FLAG_DOCUMENT_19',
    'FLAG_DOCUMENT_20', 'FLAG_DOCUMENT_21'], axis=1)

    cat_col = [category for category in data.columns if data[category].dtype == 'object']
    data = pd.get_dummies(data, columns= cat_col)

    return data

def one_hot_encode(df):

    original_columns = list(df.columns)
    categories = [cat for cat in df.columns if df[cat].dtype == 'object']
    df = pd.get_dummies(df, columns= categories, dummy_na= True) #one_hot_encode the categorical features
    categorical_columns = [cat for cat in df.columns if cat not in original_columns]
    return df, categorical_columns

def generate_credit_type_code(x):

    if x == 'Closed':
        y = 0
    elif x == 'Active':
        y = 1
    else:
        y = 2
    return y

def FE_bureau_data_1(bureau_data):

    bureau_data['CREDIT_DURATION'] = -bureau_data['DAYS_CREDIT'] + bureau_data['DAYS_CREDIT_ENDDATE']
    bureau_data['ENDDATE_DIFF'] = bureau_data['DAYS_CREDIT_ENDDATE']

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] - bureau_data['DAYS_ENDDATE_FACT']
    bureau_data['UPDATE_DIFF'] = bureau_data['DAYS_CREDIT_ENDDATE']
- bureau_data['DAYS_CREDIT_UPDATE']
    bureau_data['DEBT_PERCENTAGE'] = bureau_data['AMT_CREDIT_SUM']
/ bureau_data['AMT_CREDIT_SUM_DEBT']
    bureau_data['DEBT_CREDIT_DIFF'] = bureau_data['AMT_CREDIT_SUM']
- bureau_data['AMT_CREDIT_SUM_DEBT']
    bureau_data['CREDIT_TO_ANNUITY_RATIO'] = bureau_data['AMT_CREDIT_SUM'] / bureau_data['AMT_ANNUITY']
    bureau_data['DEBT_TO_ANNUITY_RATIO'] = bureau_data['AMT_CREDIT_SUM_DEBT'] / bureau_data['AMT_ANNUITY']
    bureau_data['CREDIT_OVERDUE_DIFF'] = bureau_data['AMT_CREDIT_SUM'] - bureau_data['AMT_CREDIT_SUM_OVERDUE']

    #Refer :- https://www.kaggle.com/c/home-credit-default-risk/discussion/57750
    #Calculating the Number of Past Loans for each Customer
    no_loans_per_customer = bureau_data[['SK_ID_CURR', 'SK_ID_BUREAU']].groupby(by = \
                                                                                                     ['SK_ID_CURR'])['SK_ID_BUREAU'].count()
    no_loans_per_customer = no_loans_per_customer.reset_index().rename(columns={'SK_ID_BUREAU': 'CUSTOMER_LOAN_COUNT'})
    bureau_data = bureau_data.merge(no_loans_per_customer, on='SK_ID_CURR', how='left')

    #Calculating the Past Credit Types per Customer
    credit_types_per_customer = bureau_data[['SK_ID_CURR', 'CREDIT_TYPE']].groupby(by=['SK_ID_CURR'])['CREDIT_TYPE'].nunique()
    credit_types_per_customer = credit_types_per_customer.reset_index().rename(columns={'CREDIT_TYPE': 'CUSTOMER_CREDIT_TYPES'})
    bureau_data = bureau_data.merge(credit_types_per_customer, on='SK_ID_CURR', how='left')

    #Average Loan Type per Customer
    bureau_data['AVG_LOAN_TYPE'] = bureau_data['CUSTOMER_LOAN_COUNT'] / bureau_data['CUSTOMER_CREDIT_TYPES']

    bureau_data['CREDIT_TYPE_CODE'] = bureau_data.apply(lambda x:\
                                                         initial_function_definition.gen

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erate_credit_type_code(x.CREDIT_ACTIVE), axis=1)

    customer_credit_code_mean = bureau_data[['SK_ID_CURR', 'CREDIT_T
YPE_CODE']].groupby(by=['SK_ID_CURR'])['CREDIT_TYPE_CODE'].mean()
    customer_credit_code_mean.reset_index().rename(columns={'CREDIT
_TYPE_CODE': 'CUSTOMER_CREDIT_CODE_MEAN'})
    bureau_data = bureau_data.merge(customer_credit_code_mean, on=
'SK_ID_CURR', how='left')

    #Computing the Ratio of Total Customer Credit and the Total Cus
tomer Debt
    bureau_data['AMT_CREDIT_SUM'] = bureau_data['AMT_CREDIT_SUM'].f
illna(0)
    bureau_data['AMT_CREDIT_SUM_DEBT'] = bureau_data['AMT_CREDIT_SU
M_DEBT'].fillna(0)
    bureau_data['AMT_ANNUITY'] = bureau_data['AMT_ANNUITY'].fillna(
0)

    credit_sum_customer = bureau_data[['SK_ID_CURR', 'AMT_CREDIT_SU
M']].groupby(by=['SK_ID_CURR'])['AMT_CREDIT_SUM'].sum()
    credit_sum_customer = credit_sum_customer.reset_index().rename(
columns={'AMT_CREDIT_SUM': 'TOTAL_CREDIT_SUM'})
    bureau_data = bureau_data.merge(credit_sum_customer, on='SK_ID_
CURR', how='left')

    credit_debt_sum_customer = bureau_data[['SK_ID_CURR', 'AMT_CREDI
T_SUM_DEBT']].groupby(by=['SK_ID_CURR'])['AMT_CREDIT_SUM_DEBT'].sum()
    credit_debt_sum_customer = credit_debt_sum_customer.reset_index
().rename(columns={'AMT_CREDIT_SUM_DEBT': 'TOTAL_DEBT_SUM'})
    bureau_data = bureau_data.merge(credit_debt_sum_customer, on='S
K_ID_CURR', how='left')
    bureau_data['CREDIT_DEBT_RATIO'] = bureau_data['TOTAL_CREDIT_SU
M']/bureau_data['TOTAL_DEBT_SUM']

    return bureau_data

def FE_bureau_data_2(bureau_data, bureau_balance, bureau_data_columns
, bureau_balance_columns):

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bureau_balance_agg = {'MONTHS_BALANCE': ['min', 'max', 'mean', 'size']}

for column in bureau_balance_columns:
    bureau_balance_agg[column] = ['min', 'max', 'mean', 'size']
    bureau_balance_final_agg = bureau_balance.groupby('SK_ID_BUREAU').agg(bureau_balance_agg)

col_list_1 = []

for col in bureau_balance_final_agg.columns.tolist():
    col_list_1.append(col[0] + "_" + col[1].upper())

bureau_balance_final_agg.columns = pd.Index(col_list_1)
bureau_data_balance = bureau_data.join(bureau_balance_final_agg, how='left', on='SK_ID_BUREAU')
bureau_data_balance.drop(['SK_ID_BUREAU'], axis=1, inplace=True)

del bureau_balance_final_agg
gc.collect()

numerical_agg = {'AMT_CREDIT_SUM_DEBT': ['mean', 'sum'], 'AMT_CREDIT_SUM_OVERDUE': ['mean', 'sum'],
                  'DAYS_CREDIT': ['mean', 'var'], 'DAYS_CREDIT_UPDATE': ['mean', 'min'],
                  'CREDIT_DAY_OVERDUE': ['mean', 'min'],
                  'DAYS_CREDIT_ENDDATE': ['mean'], 'CNT_CREDIT_PROLONG': ['sum'],
                  'MONTHS_BALANCE_SIZE': ['mean', 'sum'],
                  'AMT_CREDIT_SUM_LIMIT': ['mean', 'sum'], 'AMT_CREDIT_MAX_OVERDUE': ['mean', 'max'],
                  'AMT_ANNUITY': ['max', 'mean', 'sum'], 'AMT_CREDIT_SUM': ['mean', 'sum', 'max']}

categorical_agg = {}

for col in bureau_data_columns:
    categorical_agg[col] = ['mean']
    categorical_agg[col] = ['max']

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        for col in bureau_balance_columns:
            categorical_agg[col + "_MEAN"] = ['mean']
            categorical_agg[col + "_MIN"] = ['min']
            categorical_agg[col + "_MAX"] = ['max']

        bureau_data_balance_2 = bureau_data_balance.groupby('SK_ID_CUR
R').agg(**numerical_agg,\
**categorical_agg)
        col_list_2=[]

        for col in bureau_data_balance_2.columns.tolist():
            col_list_2.append('BUREAU_'+col[0]+'_'+col[1])
        bureau_data_balance_2.columns = pd.Index(col_list_2)

        bureau_data_balance_3 = bureau_data_balance[bureau_data_balance
['CREDIT_ACTIVE_Active'] == 1]
        bureau_data_balance_3_agg = bureau_data_balance_3.groupby('SK_I
D_CURR').agg(numerical_agg)

        col_list_3=[]

        for col in bureau_data_balance_3_agg.columns.tolist():
            col_list_3.append('A_'+col[0]+'_'+col[1].upper())

        bureau_data_balance_3_agg.columns = pd.Index(col_list_3)
        b3_final = bureau_data_balance_2.join(bureau_data_balance_3_agg
, how='left', on='SK_ID_CURR')

        bureau_data_balance_4 = bureau_data_balance[bureau_data_balance
['CREDIT_ACTIVE_Closed'] == 1]
        bureau_data_balance_4_agg = bureau_data_balance_4.groupby('SK_I
D_CURR').agg(numerical_agg)

        col_list_4 =[]

        for col in bureau_data_balance_4_agg.columns.tolist():
            col_list_4.append('C_'+col[0]+'_'+col[1].upper())

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        bureau_data_balance_4_agg.columns = pd.Index(col_list_4)
        bureau_data_balance_final = bureau_data_balance_2.join(bureau_data_balance_4_agg, how='left', on='SK_ID_CURR')

    del bureau_data_balance_3, bureau_data_balance_4_agg
    gc.collect()

    return bureau_data_balance_final

def preprocess_previous_application(data):

    data['DAYS_FIRST_DRAWING'].replace(max(data['DAYS_FIRST_DRAWING'].values), np.nan, inplace=True)
    data['DAYS_FIRST_DUE'].replace(np.nan, 0, inplace=True)
    data['DAYS_FIRST_DUE'].replace(0, np.nan, inplace=True)
    data['DAYS_FIRST_DUE'].replace(max(data['DAYS_FIRST_DUE'].values), np.nan, inplace=True)

    data['DAYS_LAST_DUE_1ST_VERSION'].replace(np.nan, 0, inplace=True)
    data['DAYS_LAST_DUE_1ST_VERSION'].replace(0, np.nan, inplace=True)
    data['DAYS_LAST_DUE_1ST_VERSION'].replace(max(data['DAYS_LAST_DUE_1ST_VERSION'].values), np.nan, inplace=True)

    data['DAYS_LAST_DUE'].replace(np.nan, 0, inplace=True)
    data['DAYS_LAST_DUE'].replace(0, np.nan, inplace=True)
    data['DAYS_LAST_DUE'].replace(max(data['DAYS_LAST_DUE'].values), np.nan, inplace=True)

    data['DAYS_TERMINATION'].replace(np.nan, 0, inplace=True)
    data['DAYS_TERMINATION'].replace(0, np.nan, inplace=True)
    data['DAYS_TERMINATION'].replace(max(data['DAYS_TERMINATION'].values), np.nan, inplace=True)

    return data

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def FE_previous_application(previous_application):
    prev_app, previous_application_columns = initial_function_definition.one_hot_encode(previous_application)

    prev_app['APPLICATION_CREDIT_DIFF'] = prev_app['AMT_APPLICATION'] - prev_app['AMT_CREDIT']
    prev_app['APPLICATION_CREDIT_RATIO'] = prev_app['AMT_APPLICATION'] / prev_app['AMT_CREDIT']
    prev_app['CREDIT_TO_ANNUITY_RATIO'] = prev_app['AMT_CREDIT'] / prev_app['AMT_ANNUITY']
    prev_app['DOWN_PAYMENT_TO_CREDIT'] = prev_app['AMT_DOWN_PAYMENT'] / prev_app['AMT_CREDIT']

    total_payment = prev_app['AMT_ANNUITY'] * prev_app['CNT_PAYMENT']
    prev_app['SIMPLE_INTERESTS'] = (total_payment / prev_app['AMT_CREDIT'] - 1) / prev_app['CNT_PAYMENT']

    prev_app['DAYS_LAST_DUE_DIFF'] = prev_app['DAYS_LAST_DUE_1ST_VERSION'] - prev_app['DAYS_LAST_DUE']

    numerical_agg_prev = {'AMT_ANNUITY': ['max', 'mean'], 'AMT_APPLICATION': ['max', 'mean'], \
                          'AMT_CREDIT': ['max', 'mean'], 'AMT_DOWN_PAYMENT': ['max', 'mean'], \
                          'AMT_GOODS_PRICE': ['mean', 'sum'], 'HOUR_APPR_PROCESS_START': \
                          ['max', 'mean'], 'RATE_DOWN_PAYMENT': ['max', 'mean'], \
                          'RATE_INTEREST_PRIMARY': \
                          ['max', 'mean'], 'RATE_INTEREST_PRIVILEGED': ['max', 'mean'], \
                          'DAYS_DECISION': ['max', 'mean'], 'CNT_PAYMENT': ['mean', 'sum'], \
                          'DAYS_FIRST_DRAWING': ['max', 'mean'], 'DAYS_TERMINATION': ['max', 'mean'], \
                          'APPLICATION_CREDIT_RATIO': ['max', 'mean'], 'DOWN_PAYMENT_TO_CREDIT': \
                          ['max', 'mean'], 'DAYS_LAST_DUE_DIFF': ['max', 'mean']}

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n' ]}

categorical_agg_prev = {}

for column in previous_application_columns:
    categorical_agg_prev[column] = ['mean']

prev_app_agg1 = prev_app.groupby('SK_ID_CURR').agg(**numerical_agg_prev, **categorical_agg_prev)

col_list_5 = []

for col in prev_app_agg1.columns.tolist():
    col_list_5.append('PREV_' + col[0] + '_' + col[1].upper())

prev_app_agg1.columns = pd.Index(col_list_5)

prev_app_cs_approved = prev_app[prev_app['NAME_CONTRACT_STATUS_
Approved']==1]
prev_app_agg2 = prev_app_cs_approved.groupby('SK_ID_CURR').agg(
numerical_agg_prev)

col_list_6 = []

for col in prev_app_agg2.columns.tolist():
    col_list_6.append('CS_APP_' + col[0] + '_' + col[1].upper
())

prev_app_agg2.columns = pd.Index(col_list_6)
prev_app_agg1_join = prev_app_agg1.join(prev_app_agg2, how='lef
t', on='SK_ID_CURR')

prev_app_cs_refused = prev_app[prev_app['NAME_CONTRACT_STATUS_R
efused']==1]
prev_app_agg3 = prev_app_cs_refused.groupby('SK_ID_CURR').agg(n
umerical_agg_prev)

col_list_7 = []

for col in prev_app_agg3.columns.tolist():

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        col_list_7.append('CS_REF_' + col[0] + '_' + col[1].upper
    ())

    prev_app_agg3.columns = pd.Index(col_list_7)
    prev_app_agg_final = prev_app_agg1_join.join(prev_app_agg3,how=
'releft', on='SK_ID_CURR')

    del prev_app_agg1_join, prev_app_agg3, prev_app_cs_refused, pre
v_app_agg1, prev_app_agg2,prev_app_cs_approved
    gc.collect()

    return prev_app_agg_final

def FE_previous_application_days_decision(data,data_temp,previous_a
pplication):

    temp_1 = initial_function_definition.FE_previous_application(in
itial_function_definition.reduce_memory_usage(previous_application))
    data = data_temp.merge(temp_1, how='left', on='SK_ID_CURR')
    del temp_1
    gc.collect()

    temp_2 = initial_function_definition.reduce_memory_usage(previo
us_application[previous_application['DAYS_DECISION']>=-365].reset_index
())
    temp_2.drop(['index'], axis=1, inplace=True)
    temp_2 = initial_function_definition.FE_previous_application(te
mp_2)
    data = data.join(temp_2, how='left', on='SK_ID_CURR',rsuffix='_
year')
    del temp_2
    gc.collect()

    temp_3 = initial_function_definition.reduce_memory_usage(previo
us_application[previous_application['DAYS_DECISION']>=-182].reset_index
())
    temp_3.drop(['index'], axis=1, inplace=True)
    temp_3 = initial_function_definition.FE_previous_application(te

```

```

mp_3)
    data = data.join(temp_3, how='left', on='SK_ID_CURR', rsuffix=
'_half_year')
    del temp_3
    gc.collect()

    temp_4 = initial_function_definition.reduce_memory_usage(previo
us_application[previous_application['DAYS_DECISION']>=-90].reset_index
())
    temp_4.drop(['index'], axis=1, inplace=True)
    temp_4 = initial_function_definition.FE_previous_application(te
mp_4)
    data = data.join(temp_4, how='left', on='SK_ID_CURR', rsuffix=
'_quarter')
    del temp_4
    gc.collect()

    temp_5 = initial_function_definition.reduce_memory_usage(previo
us_application[previous_application['DAYS_DECISION']>=-30].reset_index
())
    temp_5.drop(['index'], axis=1, inplace=True)
    temp_5 = initial_function_definition.FE_previous_application(te
mp_5)
    data = data.join(temp_5, how='left', on='SK_ID_CURR', rsuffix=
'_month')
    del temp_5
    gc.collect()

    temp_6 = initial_function_definition.reduce_memory_usage(previo
us_application[previous_application['DAYS_DECISION']>=-14].reset_index
())
    temp_6.drop(['index'], axis=1, inplace=True)
    temp_6 = initial_function_definition.FE_previous_application(te
mp_6)
    data = data.join(temp_6, how='left', on='SK_ID_CURR', rsuffix=
'_fortnight')
    del temp_6
    gc.collect()

    temp_7 = initial_function_definition.reduce_memory_usage(previo

```



```

us_application[previous_application['DAYS_DECISION']>=-7].reset_index
())
    temp_7.drop(['index'], axis=1, inplace=True)
    temp_7 = initial_function_definition.FE_previous_application(temp_7)
    data = data.join(temp_7, how='left', on='SK_ID_CURR', rsuffix='_week')
    del temp_7
    gc.collect()

    return data

def FE_pos_cash_balance(pos_cash_balance):

    pos_balance_data, pos_balance_columns = initial_function_definition.one_hot_encode(pos_cash_balance)

    pos_balance_data['LATE_PAYMENT'] = pos_balance_data['SK_DPD'].apply(lambda x:1 if x>0 else 0)

    numerical_agg_pos_balance = {'SK_DPD_DEF': ['max', 'mean', 'min'], 'SK_DPD': ['max', 'mean', 'min'],
    'MONTHS_BALANCE': ['max', 'mean', 'size'], 'CNT_INSTALLMENT': ['max', 'size'],
    'CNT_INSTALLMENT_FUTURE': ['max', 'size', 'sum']}

    categorical_agg_pos_balance = {}

    for col in pos_balance_columns:
        categorical_agg_pos_balance[col] = ['mean']

    pos_balance_agg = pos_balance_data.groupby('SK_ID_CURR').agg(*numerical_agg_pos_balance, **categorical_agg_pos_balance)

    col_list_8=[]

    for col in pos_balance_agg.columns.tolist():
        col_list_8.append('POS_'+col[0] + '_' + col[1].upper())

```

```

pos_balance_agg.columns = pd.Index(col_list_8)

sort_pos_balance = pos_balance_data.sort_values(by=['SK_ID_PREV', 'MONTHS_BALANCE'])
pos_group = sort_pos_balance.groupby('SK_ID_PREV')

pos_final_df = pd.DataFrame()
pos_final_df['SK_ID_CURR'] = pos_group['SK_ID_CURR'].first()
pos_final_df['MONTHS_BALANCE_MAX'] = pos_group['MONTHS_BALANCE'].max()

pos_final_df['POS_LOAN_COMPLETED_MEAN'] = pos_group['NAME_CONTRACT_STATUS_Completed'].mean()
pos_final_df['POS_COMPLETED_BEFORE_MEAN'] = pos_group['CNT_INSTALLMENT'].first() - pos_group['CNT_INSTALLMENT'].last()

pos_final_df['POS_COMPLETED_BEFORE_MEAN'] = pos_final_df.apply(
    lambda x: 1 if x['POS_COMPLETED_BEFORE_MEAN'] > 0 and x['POS_LOAN_COMPLETED_MEAN'] > 0 else 0, axis=1)

pos_final_df['POS_REMAINING_INSTALLMENTS'] = pos_group['CNT_INSTALLMENT_FUTURE'].last()
pos_final_df['POS_REMAINING_INSTALLMENTS_RATIO'] = pos_group['CNT_INSTALLMENT_FUTURE'].last()/pos_group['CNT_INSTALLMENT'].last()

pos_final_df_groupby = pos_final_df.groupby('SK_ID_CURR').sum().reset_index()
pos_final_df_groupby.drop(['MONTHS_BALANCE_MAX'], axis=1, inplace=True)
pos_final_agg = pd.merge(pos_balance_agg, pos_final_df_groupby, on='SK_ID_CURR', how='left')

del pos_balance_agg, pos_final_df_groupby, pos_group, sort_pos_balance
gc.collect()
return pos_final_agg

def FE_pos_cash_balance_months_balance(data, data_temp, pos_cash_ba

```

```

lance):

    temp_8 = initial_function_definition.FE_pos_cash_balance(initial_function_definition.reduce_memory_usage(pos_cash_balance))
    data = data_temp.merge(temp_8, how='left', on='SK_ID_CURR')
    del temp_8
    gc.collect()

    temp_9 = initial_function_definition.reduce_memory_usage(pos_cash_balance[pos_cash_balance['MONTHS_BALANCE']>=-12].reset_index())
    temp_9.drop(['index'], axis=1, inplace=True)
    temp_9 = initial_function_definition.FE_pos_cash_balance(temp_9)
    data = data.join(temp_9, how='left', on='SK_ID_CURR', rsuffix='_year')
    del temp_9
    gc.collect()

    temp_10 = initial_function_definition.reduce_memory_usage(pos_cash_balance[pos_cash_balance['MONTHS_BALANCE']>=-6].reset_index())
    temp_10.drop(['index'], axis=1, inplace=True)
    temp_10 = initial_function_definition.FE_pos_cash_balance(temp_10)
    data = data.join(temp_10, how='left', on='SK_ID_CURR', rsuffix='_half_year')
    del temp_10
    gc.collect()

    temp_11 = initial_function_definition.reduce_memory_usage(pos_cash_balance[pos_cash_balance['MONTHS_BALANCE']>=-3].reset_index())
    temp_11.drop(['index'], axis=1, inplace=True)
    temp_11 = initial_function_definition.FE_pos_cash_balance(temp_11)
    data = data.join(temp_11, how='left', on='SK_ID_CURR', rsuffix='_quarter')
    del temp_11
    gc.collect()

    temp_12 = initial_function_definition.reduce_memory_usage(pos_cash_balance[pos_cash_balance['MONTHS_BALANCE']>=-1].reset_index())

```

```

        temp_12.drop(['index'], axis=1, inplace=True)
        temp_12 = initial_function_definition.FE_pos_cash_balance(temp_
12)
        data = data.join(temp_12, how='left', on='SK_ID_CURR', rsuffix=
'_month')
        del temp_12
        gc.collect()

        return data

def FE_installments_payments(installments_payments):

    pay1 = installments_payments[['SK_ID_PREV', 'NUM_INSTALLMENT_NUM
BER'] + ['AMT_PAYMENT']]
    pay2 = pay1.groupby(['SK_ID_PREV', 'NUM_INSTALLMENT_NUMBER'])['A
MT_PAYMENT'].sum().reset_index()
    pay_final = pay2.rename(columns={'AMT_PAYMENT': 'AMT_PAYMENT_GR
OUPED'})
    payments_final = installments_payments.merge(pay_final, \
on=['SK_ID_PREV', 'NUM_INSTALLMENT_NUMBER'],
how='left')

    payments_final['PAYMENT_DIFFERENCE'] = payments_final['AMT_INST
ALMENT'] - payments_final['AMT_PAYMENT_GROUPED']
    payments_final['PAYMENT_RATIO'] = payments_final['AMT_INSTALLMEN
T'] / payments_final['AMT_PAYMENT_GROUPED']

    payments_final['PAID_OVER_AMOUNT'] = payments_final['AMT_PAYMEN
T'] - payments_final['AMT_INSTALLMENT']
    payments_final['PAID_OVER'] = (payments_final['PAID_OVER_AMOUN
T'] > 0).astype(int)

    payments_final['DPD'] = payments_final['DAYS_ENTRY_PAYMENT'] - \
        payments_final['DAYS_INSTALLMENT']
    payments_final['DPD'] = payments_final['DPD'].apply(lambda x: 0
if x <= 0 else x)

```

```

        payments_final['DBD'] = payments_final['DAYS_INSTALMENT'] - \
            payments_final['DAYS_ENTRY_PAYMENT']
        payments_final['DBD'] = payments_final['DBD'].apply(lambda x: 0
if x <= 0 else x)
        payments_final['LATE_PAYMENT'] = payments_final['DBD'].apply(la
mbda x: 1 if x > 0 else 0)

        payments_final['INSTALMENT_PAYMENT_RATIO'] = payments_final['AM
T_PAYMENT'] / payments_final['AMT_INSTALMENT']
        payments_final['LATE_PAYMENT_RATIO'] = payments_final.apply(lam
bda x: x['INSTALMENT_PAYMENT_RATIO'] if x['LATE_PAYMENT'] == 1 else 0,
axis=1)

        payments_final['SIGNIFICANT_LATE_PAYMENT'] = payments_final['LA
TE_PAYMENT_RATIO'].apply(lambda x: 1 if x > 0.05 else 0)

        payments_final['DPD_7'] = payments_final['DPD'].apply(lambda x:
1 if x >= 7 else 0)
        payments_final['DPD_15'] = payments_final['DPD'].apply(lambda x
: 1 if x >= 15 else 0)
        payments_final['DPD_30'] = payments_final['DPD'].apply(lambda x
: 1 if x >= 30 else 0)
        payments_final['DPD_60'] = payments_final['DPD'].apply(lambda x
: 1 if x >= 60 else 0)
        payments_final['DPD_90'] = payments_final['DPD'].apply(lambda x
: 1 if x >= 90 else 0)
        payments_final['DPD_180'] = payments_final['DPD'].apply(lambda
x: 1 if x >= 180 else 0)
        payments_final['DPD_WOF'] = payments_final['DPD'].apply(lambda
x: 1 if x >= 720 else 0)

        payments_final, pay_final_columns = initial_function_definition
.one_hot_encode(payments_final)

        numeric_agg_payments = {'LATE_PAYMENT': ['max', 'mean', 'min'], 'A
MT_PAYMENT': ['min', 'max', \
                'mean', 'sum'], 'NUM_INSTALMENT_VERSION': ['nuniq
ue'], \
                'NUM_INSTALMENT_NUMBER': ['max'], 'AMT_INSTALMENT'
: ['max', 'mean', 'sum'],

```

```

        'PAYMENT_DIFFERENCE': ['max', 'mean', 'min', 'sum'], 'DAYS_ENTRY_PA
YMENT': ['max', \
        'mean', 'sum'], 'PAID_OVER_AMOUNT': ['max', 'mean', 'min']}]

    for col in pay_final_columns:
        numeric_agg_payments[col] = ['mean']

    payments_final_agg = payments_final.groupby('SK_ID_CURR').agg(n
umeric_agg_payments)
    col_list_9=[]

    for col in payments_final_agg.columns.tolist():
        col_list_9.append('INS_'+col[0]+'_'+col[1].upper())

    payments_final_agg.columns = pd.Index(col_list_9)
    payments_final_agg['INSTALLATION_COUNT'] = payments_final.group
by('SK_ID_CURR').size()

    del payments_final
    gc.collect()

    return payments_final_agg

def FE_installments_payments_days_instalment(data, data_temp, insta
llments_payments):

    installments_payments['DAYS_ENTRY_PAYMENT'].fillna(0, inplace=T
rue)
    installments_payments['AMT_PAYMENT'].fillna(0.0, inplace=True)

    temp_13 = initial_function_definition.FE_installments_payments(
initial_function_definition.reduce_memory_usage(installments_payments))
    data = data_temp.join(temp_13, how='left', on='SK_ID_CURR')
    del temp_13
    gc.collect()

    temp_14 = initial_function_definition.reduce_memory_usage(insta
llments_payments[installments_payments['DAYS_INSTALLMENT']>=-365].reset_
index())

```

```

        temp_14.drop(['index'], axis=1, inplace=True)
        temp_14 = initial_function_definition.FE_installments_payments(
temp_14)
        data = data.join(temp_14, how='left', on='SK_ID_CURR', rsuffix=
'_year')
        del temp_14
        gc.collect()

        temp_15 = initial_function_definition.reduce_memory_usage(instal
llments_payments[installments_payments['DAYS_INSTALLMENT']>=-182].reset_
index())
        temp_15.drop(['index'], axis=1, inplace=True)
        temp_15 = initial_function_definition.FE_installments_payments(
temp_15)
        data = data.join(temp_15, how='left', on='SK_ID_CURR', rsuffix=
'_half_year')
        del temp_15
        gc.collect()

        temp_16 = initial_function_definition.reduce_memory_usage(instal
llments_payments[installments_payments['DAYS_INSTALLMENT']>=-90].reset_i
ndex())
        temp_16.drop(['index'], axis=1, inplace=True)
        temp_16 = initial_function_definition.FE_installments_payments(
temp_16)
        data = data.join(temp_16, how='left', on='SK_ID_CURR', rsuffix=
'_quarter')
        del temp_16
        gc.collect()

        temp_17 = initial_function_definition.reduce_memory_usage(instal
llments_payments[installments_payments['DAYS_INSTALLMENT']>=-30].reset_i
ndex())
        temp_17.drop(['index'], axis=1, inplace=True)
        temp_17 = initial_function_definition.FE_installments_payments(
temp_17)
        data = data.join(temp_17, how='left', on='SK_ID_CURR', rsuffix=
'_month')
        del temp_17
        gc.collect()

```

```

        temp_18 = initial_function_definition.reduce_memory_usage(installments_payments[installments_payments['DAYS_INSTALLMENT'] >= -14].reset_index())
        temp_18.drop(['index'], axis=1, inplace=True)
        temp_18 = initial_function_definition.FE_installments_payments(temp_18)
        data = data.join(temp_18, how='left', on='SK_ID_CURR', rsuffix='_fortnight')
        del temp_18
        gc.collect()

        temp_19 = initial_function_definition.reduce_memory_usage(installments_payments[installments_payments['DAYS_INSTALLMENT'] >= -7].reset_index())
        temp_19.drop(['index'], axis=1, inplace=True)
        temp_19 = initial_function_definition.FE_installments_payments(temp_19)
        data = data.join(temp_19, how='left', on='SK_ID_CURR', rsuffix='_week')
        del temp_19
        gc.collect()

        return data

def FE_credit_card_balance(credit_card_balance):
    cc_balance_data, cc_balance_columns = initial_function_definition.one_hot_encode(credit_card_balance)
    cc_balance_data.rename(columns={'AMT_RECIVABLE': 'AMT_RECEIVABLE'}, inplace=True)

    cc_balance_data['LIMIT_USE'] = cc_balance_data['AMT_BALANCE'] / cc_balance_data['AMT_CREDIT_LIMIT_ACTUAL']
    cc_balance_data['PAYMENT_DIV_MIN'] = cc_balance_data['AMT_PAYMENT_CURRENT'] / cc_balance_data['AMT_INST_MIN_REGULARITY']
    cc_balance_data['LATE_PAYMENT'] = cc_balance_data['SK_DPD'].apply(lambda x: 1 if x > 0 else 0)

```



```

        cc_balance_data['DRAWING_LIMIT_RATIO'] = cc_balance_data['AMT_DRAWINGS_ATM_CURRENT'] / cc_balance_data['AMT_CREDIT_LIMIT_ACTUAL']

        cc_balance_data.drop(['SK_ID_PREV'], axis=1, inplace=True)
        cc_balance_data_agg = cc_balance_data.groupby('SK_ID_CURR').agg(['max', 'mean', 'sum', 'var'])

        col_list_9=[]

        for col in cc_balance_data_agg.columns.tolist():
            col_list_9.append('CR_'+col[0]+'_'+col[1].upper())

        cc_balance_data_agg.columns = pd.Index(col_list_9)

        cc_balance_data_agg['CREDIT_COUNT'] = cc_balance_data.groupby('SK_ID_CURR').size()

        del cc_balance_data, cc_balance_columns
        gc.collect()

        return cc_balance_data_agg

def FE_credit_card_balance_months_balance(data,data_temp,credit_card_balance):

    temp_20 = initial_function_definition.FE_credit_card_balance(initial_function_definition.reduce_memory_usage(credit_card_balance))
    data = data_temp.join(temp_20, how='left', on='SK_ID_CURR')
    del temp_20
    gc.collect()

    temp_21 = initial_function_definition.reduce_memory_usage(credit_card_balance[credit_card_balance['MONTHS_BALANCE']>=-12].reset_index())
    temp_21.drop(['index'], axis=1, inplace=True)
    temp_21 = initial_function_definition.FE_credit_card_balance(temp_21)
    data = data.join(temp_21, how='left', on='SK_ID_CURR', rsuffix=

```

```

'_year')
    del temp_21
    gc.collect()

    temp_22 = initial_function_definition.reduce_memory_usage(credit_card_balance[credit_card_balance['MONTHS_BALANCE']>=-6].reset_index())
    temp_22.drop(['index'], axis=1, inplace=True)
    temp_22 = initial_function_definition.FE_credit_card_balance(temp_22)
    data = data.join(temp_22, how='left', on='SK_ID_CURR', rsuffix='_half_year')
    del temp_22
    gc.collect()

    temp_23 = initial_function_definition.reduce_memory_usage(credit_card_balance[credit_card_balance['MONTHS_BALANCE']>=-3].reset_index())
    temp_23.drop(['index'], axis=1, inplace=True)
    temp_23 = initial_function_definition.FE_credit_card_balance(temp_23)
    data = data.join(temp_23, how='left', on='SK_ID_CURR', rsuffix='_quarter')
    del temp_23
    gc.collect()

    temp_24 = initial_function_definition.reduce_memory_usage(credit_card_balance[credit_card_balance['MONTHS_BALANCE']>=-1].reset_index())
    temp_24.drop(['index'], axis=1, inplace=True)
    temp_24 = initial_function_definition.FE_credit_card_balance(temp_24)
    data = data.join(temp_24, how='left', on='SK_ID_CURR', rsuffix='_month')
    del temp_24
    gc.collect()

    return data

```

### 3. Computing the Probabilities on the Test Dataset

```
In [ ]: import warnings
warnings.filterwarnings("ignore")
import os
import os.path
import sqlite3
import flask

from flask import Flask, jsonify, request
from lightgbm import LGBMClassifier
from sqlalchemy import create_engine
from hcd_r_model import initial_function_definition

if os.path.isfile('pickles/test_data')==False:

    train_data = initial_function_definition.reduce_memory_usage(pd.read_csv('home-credit-default-risk/application_train.csv'))
    test_data = initial_function_definition.reduce_memory_usage(pd.read_csv('home-credit-default-risk/application_test.csv'))
    bureau_data = initial_function_definition.reduce_memory_usage(pd.read_csv('home-credit-default-risk/bureau.csv'))
    bureau_balance = initial_function_definition.reduce_memory_usage(pd.read_csv('home-credit-default-risk/bureau_balance.csv'))

    bureau_data_fe = initial_function_definition.FE_bureau_data_1(bureau_data)

    #One Hot Encoding the Bureau Datasets
    bureau_data, bureau_data_columns = initial_function_definition.one_hot_encode(bureau_data_fe)
    bureau_balance, bureau_balance_columns = initial_function_definition.one_hot_encode(bureau_balance)

    bureau_data_balance_final = initial_function_definition.FE_bureau_data_2(bureau_data, bureau_balance, bureau_data_columns, bureau_balance_columns)

    previous_application = initial_function_definition.reduce_memory_us
```

```

age(pd.read_csv('home-credit-default-risk/previous_application.csv'))
previous_application = initial_function_definition.preprocess_previous_application(previous_application)

pos_cash_balance = initial_function_definition.reduce_memory_usage(
pd.read_csv('home-credit-default-risk/POS_CASH_balance.csv'))
installments_payments = initial_function_definition.reduce_memory_usage(
pd.read_csv('home-credit-default-risk/installments_payments.csv'))
credit_card_balance = initial_function_definition.reduce_memory_usage(
pd.read_csv('home-credit-default-risk/credit_card_balance.csv'))

start = datetime.now()

test_data = initial_function_definition.fix_nulls_outliers(test_data)
test_data_temp_1 = initial_function_definition.FE_application_data(test_data)
bureau_data_balance_final = initial_function_definition.FE_bureau_data_2(
bureau_data, bureau_balance, bureau_data_columns, bureau_balance_columns)
test_data_temp_2 = test_data_temp_1.join(bureau_data_balance_final,
how='left', on='SK_ID_CURR')

test_data_temp_2 = initial_function_definition.FE_previous_application_days_decision(
test_data, test_data_temp_2, previous_application)
test_data_temp_2 = initial_function_definition.FE_pos_cash_balance_months_balance(
test_data, test_data_temp_2, pos_cash_balance)
test_data_temp_2 = initial_function_definition.FE_installments_payments_days_instalment(
test_data, test_data_temp_2, installments_payments)
test_data_temp_2 = initial_function_definition.FE_credit_card_balance_months_balance(
test_data, test_data_temp_2, credit_card_balance)

#Removing any duplicate features, if any are present in the final dataset
test_data = test_data_temp_2.loc[:, ~test_data_temp_2.columns.duplicated()]

```

```

        print("Time taken to run this cell :", datetime.now() - start)
    else:

        test_data = pd.read_pickle('pickles/test_data')

        features_top_df_train = pd.read_pickle('pickles/features_top_df_train.pkl')
        features_top_df_test = test_data[features_top_df_train.columns]
        features_top_df_test['SK_ID_CURR'] = test_data['SK_ID_CURR']
        features_top_df_test['TARGET'] = np.nan

    app = Flask(__name__)

    #home page
    @app.route('/', methods = [])
    def hello_world():
        return 'Hello World!'

    #prediction page
    @app.route('/index')
    def index():
        return flask.render_template('index.html')

    #results page
    @app.route('/predict', methods = ['POST'])
    def predict():

        conn = sqlite3.connect('Home_Credit_DB_Connection.db')
        sk_id_curr = request.form.to_dict()['SK_ID_CURR']
        sk_id_curr = int(sk_id_curr)

        test_datapoint = pd.read_sql_query(f'SELECT * FROM test_data_feats
        WHERE SK_ID_CURR == {sk_id_curr}', conn)
        test_datapoint = test_datapoint.replace([None], np.nan)

```

```

with open('lgbm/lgbm_model_500f_3.pickle','rb') as f:
    lgbm_model = pickle.load(f)

if os.path.isfile('lgbm/lgbm_best_threshold_500f_api.pkl')==False:

    feats = [f for f in features_top_df_train.columns if f not in [
'TARGET','SK_ID_CURR','SK_ID_BUREAU','SK_ID_PREV','index']]
    test_predict = np.zeros(features_top_df_test.shape[0])
    test_predict += lgbm_model.predict_proba(features_top_df_test[feats], num_iteration=lgbm_model.best_iteration_)[:, 1] / 5
else:

    with open('lgbm/lgbm_test_predict_500f.pkl','rb') as f:
        test_predict = pickle.load(f)

threshold = 0.3741018248484985

test_predict_rounded = np.round(test_predict,4)
predicted_class_label = np.where(test_predict_rounded > threshold,
1, 0)

select_index = list(np.where(test_data["SK_ID_CURR"] == sk_id_curr)
[0])
final_class_label = predicted_class_label[select_index[0]]
final_test_predict_rounded = test_predict_rounded[select_index[0]]

if final_class_label == 1:
    prediction = 'The customer with this ID is a Potential Defaulter with a probability of {}'.format(final_test_predict_rounded)
else:
    prediction = 'The customer with this ID is not a Potential Defaulter with a probability of {}'.format(final_test_predict_rounded)
    predicted_proba = 1 - final_test_predict_rounded

return jsonify({'prediction': prediction})

if __name__ == '__main__':
    app.debug=True
    app.run(host='0.0.0.0', port=8080)

```

