

# **HOME LONE DATA ANALYSIS**

## **Abstract:**

In the domain of financial lending, ensuring a safe and secure lending experience requires accurate prediction models to assess the likelihood of loan defaults. This project focuses on the development of a deep learning model aimed at predicting the probability of default for future loans based on historical loan data. The dataset used in this project is highly imbalanced, with a disproportionate number of default versus non-default instances, adding complexity to the predictive task. Additionally, the dataset contains a large number of features, increasing the challenge of effectively capturing relevant patterns while mitigating overfitting. The proposed solution employs advanced deep learning techniques, such as neural networks, to handle class imbalance and leverage the rich feature set for accurate predictions. Evaluation metrics, such as precision, recall, and the area under the ROC curve, are utilized to assess the model's performance in real-world scenarios, ensuring that it provides robust and reliable predictions for future loan default risks.

## **Objective:**

Creating a model that predicts whether or not an applicant will be able to repay a loan using historical data

## **Domain:**

Finance

## **Analysis to be done:**

Perform data preprocessing and build a deep learning prediction model

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homeloan.ipynb x +
+ - Code Python 3 [3.10]

[1]: import pandas as pd
import sklearn
import numpy as np
import matplotlib.pyplot as plt
import os
import warnings
import seaborn as sns
from sklearn.preprocessing import OneHotEncoder
from sklearn.datasets import make_blobs
from sklearn.impute import SimpleImputer
from sklearn.pipeline import Pipeline
from sklearn.compose import ColumnTransformer
from sklearn.preprocessing import StandardScaler
from sklearn.svm import LinearSVC
from sklearn.metrics import roc_auc_score
from sklearn.linear_model import LogisticRegression
from sklearn.metrics import roc_auc_score
from sklearn.calibration import CalibratedClassifierCV
from sklearn.metrics import confusion_matrix
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score
from sklearn.linear_model import SGDClassifier
import plotly.offline as py
import plotly.graph_objs as go
from plotly.offline import init_notebook_mode, iplot
from sklearn.model_selection import train_test_split
init_notebook_mode(connected=True)

Simple 0 0 Python 3 [3.10] | Idle Mode: Command Ln 3, Col 18 homeloan.ipynb
```

```
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homeloan.ipynb x +
+ - Code Python 3 [3.10]

import pickle
import gc
warnings.filterwarnings('ignore')
%matplotlib inline

[2]: # Load the dataset
house_loan = pd.read_csv('loan_data (1).csv')
house_loan.describe()

[2]:
```

	SK_ID_CURR	TARGET	CNT_CHILDREN	AMT_INCOME_TOTAL	AMT_CREDIT	AMT_ANNUITY	AMT_GOODS_PRICE	REGION_POPULATION_RELATIVE
count	307511.000000	307511.000000	307511.000000	3.075110e+05	3.075110e+05	307499.000000	3.072330e+05	307511.000000
mean	278180.518577	0.080729	0.417052	1.687979e+05	5.990260e+05	27108.573909	5.383962e+05	0.020868
std	102790.175348	0.272419	0.722121	2.371231e+05	4.024908e+05	14493.737315	3.694465e+05	0.013831
min	100002.000000	0.000000	0.000000	2.565000e+04	4.500000e+04	1615.500000	4.050000e+04	0.000290
25%	189145.500000	0.000000	0.000000	1.125000e+05	2.700000e+05	16524.000000	2.385000e+05	0.010006
50%	278202.000000	0.000000	0.000000	1.471500e+05	5.135310e+05	24903.000000	4.500000e+05	0.018850
75%	367142.500000	0.000000	1.000000	2.025000e+05	8.086500e+05	34596.000000	6.795000e+05	0.028663
max	456255.000000	1.000000	19.000000	1.170000e+08	4.050000e+06	258025.500000	4.050000e+06	0.072508

8 rows x 106 columns

```
Simple 0 0 Python 3 [3.10] | Idle Mode: Command Ln 1, Col 1 homeloan.ipynb
```

```
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homeloan.ipynb x +
Python 3 [3.10]

[3]: house_loan.columns

[3]: Index(['SK_ID_CURR', 'TARGET', 'NAME_CONTRACT_TYPE', 'CODE_GENDER',
        'FLAG_OWN_CAR', 'FLAG_OWN_REALTY', 'CNT_CHILDREN', 'AMT_INCOME_TOTAL',
        'AMT_CREDIT', 'AMT_ANNUITY',
        ...,
        'FLAG_DOCUMENT_18', 'FLAG_DOCUMENT_19', 'FLAG_DOCUMENT_20',
        'FLAG_DOCUMENT_21', 'AMT_REQ_CREDIT_BUREAU_HOUR',
        'AMT_REQ_CREDIT_BUREAU_DAY', 'AMT_REQ_CREDIT_BUREAU_WEEK',
        'AMT_REQ_CREDIT_BUREAU_MON', 'AMT_REQ_CREDIT_BUREAU_QRT',
        'AMT_REQ_CREDIT_BUREAU_YEAR'],
        dtype='object', length=122)

[4]: house_loan.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 307511 entries, 0 to 307510
Columns: 122 entries, SK_ID_CURR to AMT_REQ_CREDIT_BUREAU_YEAR
dtypes: float64(65), int64(41), object(16)
memory usage: 286.2+ MB

[5]: house_loan.isnull().sum()

[5]: SK_ID_CURR      0
     TARGET        0
     NAME_CONTRACT_TYPE  0
```

```
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homeloan.ipynb x +
Python 3 [3.10]

[5]: SK_ID_CURR      0
     TARGET        0
     NAME_CONTRACT_TYPE  0
     CODE_GENDER      0
     FLAG_OWN_CAR      0
     ...
     AMT_REQ_CREDIT_BUREAU_DAY  41519
     AMT_REQ_CREDIT_BUREAU_WEEK  41519
     AMT_REQ_CREDIT_BUREAU_MON  41519
     AMT_REQ_CREDIT_BUREAU_QRT  41519
     AMT_REQ_CREDIT_BUREAU_YEAR  41519
     Length: 122, dtype: int64

[6]: house_loan.head()

[6]: SK_ID_CURR TARGET NAME_CONTRACT_TYPE CODE_GENDER FLAG_OWN_CAR FLAG_OWN_REALTY CNT_CHILDREN AMT_INCOME_TOTAL AMT_CREDIT AMT_REQ_CREDIT_BUREAU_YEAR
0      100002      1      Cash loans      M      N      Y      0      202500.0      406597.5      41519
1      100003      0      Cash loans      F      N      N      0      270000.0      1293502.5      41519
2      100004      0      Revolving loans      M      Y      Y      0      67500.0      135000.0      41519
3      100006      0      Cash loans      F      N      Y      0      135000.0      312682.5      41519
4      100007      0      Cash loans      M      N      Y      0      121500.0      513000.0      41519

5 rows x 122 columns
```

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homeloan.ipynb Python 3 [3.10]

```
SK_ID_CURR
TARGET
NAME_CONTRACT_TYPE
CODE_GENDER
FLAG_OWN_CAR
...
AMT_REQ_CREDIT_BUREAU_DAY 41519
AMT_REQ_CREDIT_BUREAU_WEEK 41519
AMT_REQ_CREDIT_BUREAU_MON 41519
AMT_REQ_CREDIT_BUREAU_QRT 41519
AMT_REQ_CREDIT_BUREAU_YEAR 41519
Length: 122, dtype: int64
```

```
[6]: house_loan.head()
```

```
[6]:
```

	SK_ID_CURR	TARGET	NAME_CONTRACT_TYPE	CODE_GENDER	FLAG_OWN_CAR	FLAG_OWN_REALTY	CNT_CHILDREN	AMT_INCOME_TOTAL	AMT_CREDIT	AMT
0	100002	1	Cash loans	M	N	Y	0	202500.0	406597.5	
1	100003	0	Cash loans	F	N	N	0	270000.0	1293502.5	
2	100004	0	Revolving loans	M	Y	Y	0	67500.0	135000.0	
3	100006	0	Cash loans	F	N	Y	0	135000.0	312682.5	
4	100007	0	Cash loans	M	N	Y	0	121500.0	513000.0	

5 rows x 122 columns

Simple 0 Python 3 [3.10] | Idle Mode: Command Ln 1, Col 1 homeloan.ipynb

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homeloan.ipynb Python 3 [3.10]

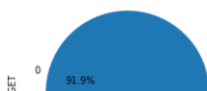
```
5 rows x 122 columns
```

```
[7]: defaulters=(house_loan.TARGET==1).sum()
payers=(house_loan.TARGET==0).sum()
print((defaulters/payers)*100)
8.781828601345662
```

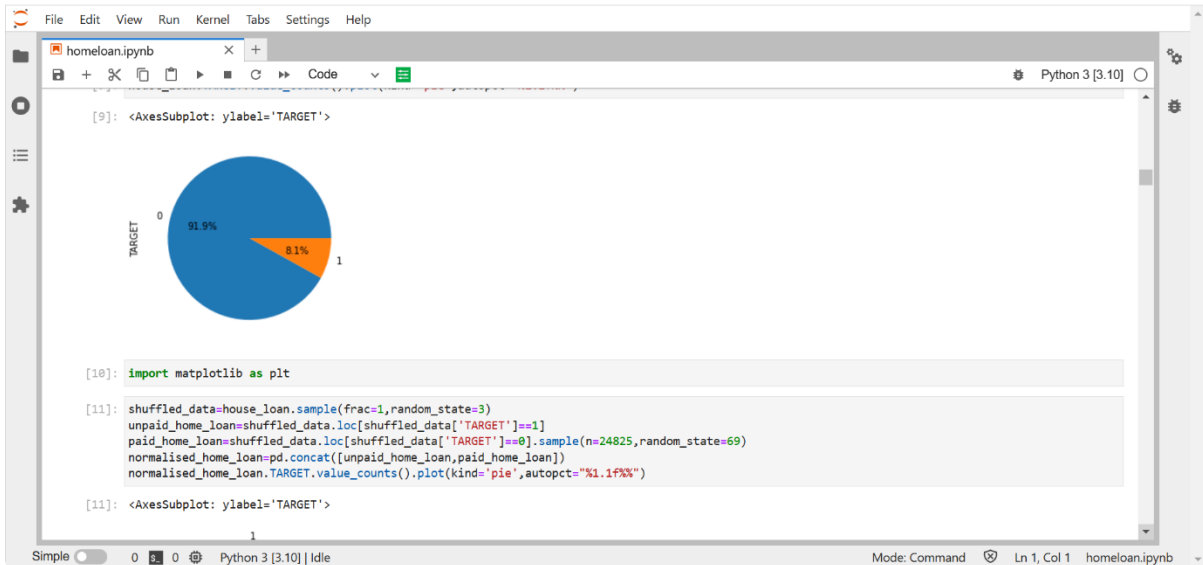
```
[8]: without_id=[column for column in house_loan.columns if column!='SK_ID_CURR']
#check for duplicate values
na=house_loan[house_loan.duplicated(subset=without_id,keep=False)]
print("Duplicates are: ",na.shape[0])
Duplicates are: 0
```

```
[9]: house_loan.TARGET.value_counts().plot(kind='pie',autopct='%1.1f%%')
```

```
[9]: <AxesSubplot: ylabel='TARGET'>
```



Simple 0 Python 3 [3.10] | Idle Mode: Command Ln 1, Col 1 homeloan.ipynb



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homeloan.ipynb Python 3 [3.10]

```
[53]: import tensorflow as tf
```

```
[13]: normalised_home_loan.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 49659 entries, 207339 to 121862
Columns: 122 entries, SK_ID_CURR to AMT_REQ_CREDIT_BUREAU_YEAR
dtypes: float64(65), int64(41), object(16)
memory usage: 46.6+ MB
```

```
[14]: normalised_home_loan.head
```

```
[14]: <bound method NDFrame.head of
207339 340318 1 Cash loans F N
8756 110186 1 Cash loans M Y
230344 366811 1 Cash loans F N
178329 306645 1 Cash loans M Y
55586 164407 1 Cash loans M N
... ..
130947 251878 0 Cash loans F Y
40467 146875 0 Cash loans F N
187004 316791 0 Cash loans M N
131755 252811 0 Cash loans F N
121862 241287 0 Cash loans M N
```

```
FLAG_OWN_REALTY CNT_CHILDREN AMT_INCOME_TOTAL AMT_CREDIT \
207339 N 0 112500.0 405000.0
```

Simple 0 Python 3 [3.10] | Idle Mode: Command Ln 1, Col 1 homeloan.ipynb

```
File Edit View Run Kernel Tabs Settings Help
homeloan.ipynb Python 3 [3.10]
8756 0 0.0 0.0
230344 0 NaN NaN
178329 0 NaN NaN
55586 0 0.0 0.0
... ..
130947 0 0.0 0.0
40467 0 0.0 0.0
187004 0 0.0 0.0
131755 0 0.0 0.0
121862 0 0.0 0.0

AMT_REQ_CREDIT_BUREAU_WEEK AMT_REQ_CREDIT_BUREAU_MON \
207339 0.0 0.0
8756 0.0 0.0
230344 NaN NaN
178329 NaN NaN
55586 0.0 0.0
... ..
130947 0.0 1.0
40467 0.0 0.0
187004 0.0 0.0
131755 0.0 0.0
121862 0.0 0.0

AMT_REQ_CREDIT_BUREAU_QRT AMT_REQ_CREDIT_BUREAU_YEAR
207339 0.0 3.0
8756 0.0 0.0
230344 NaN NaN
```

```
File Edit View Run Kernel Tabs Settings Help
homeloan.ipynb Python 3 [3.10]
[49650 rows x 122 columns]

[15]: normalised_home_loan.dropna(axis=0)
normalised_home_loan.info()

<class 'pandas.core.frame.DataFrame'>
Int64Index: 49650 entries, 207339 to 121862
Columns: 122 entries, SK_ID_CURR to AMT_REQ_CREDIT_BUREAU_YEAR
dtypes: float64(65), int64(41), object(16)
memory usage: 46.6+ MB

[16]: normalised_home_loan.isnull().sum()

[16]: SK_ID_CURR 0
TARGET 0
NAME_CONTRACT_TYPE 0
CODE_GENDER 0
FLAG_OWN_CAR 0
...
AMT_REQ_CREDIT_BUREAU_DAY 7648
AMT_REQ_CREDIT_BUREAU_WEEK 7648
AMT_REQ_CREDIT_BUREAU_MON 7648
AMT_REQ_CREDIT_BUREAU_QRT 7648
AMT_REQ_CREDIT_BUREAU_YEAR 7648
Length: 122, dtype: int64

[17]: #print(normalised_home_loan.apply())

[18]: print(pd.unique(normalised_home_loan.AMT_REQ_CREDIT_BUREAU_DAY))
```

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homeloan.ipynb Python 3 [3.10]

```
[17]: #print(normalised_home_loan.apply())

[18]: print(pd.unique(normalised_home_loan.AMT_REQ_CREDIT_BUREAU_DAY))
print(pd.unique(normalised_home_loan.AMT_REQ_CREDIT_BUREAU_WEEK))
print(pd.unique(normalised_home_loan.AMT_REQ_CREDIT_BUREAU_MON))
print(pd.unique(normalised_home_loan.AMT_REQ_CREDIT_BUREAU_QRT))
print(pd.unique(normalised_home_loan.AMT_REQ_CREDIT_BUREAU_YEAR))

[ 0. nan 1. 2. 4. 3. 9.]
[ 0. nan 1. 2. 4. 3. 5. 6.]
[ 0. nan 1. 3. 5. 9. 2. 6. 8. 4. 11. 12. 7. 13. 10. 17. 15. 14.
 16. 18. 27.]
[ 0. nan 2. 3. 1. 4. 5. 6. 19. 7.]
[ 3. 0. nan 1. 5. 4. 2. 6. 7. 8. 9. 10. 14. 13. 12. 11. 22. 16.
 23. 17.]

[19]: normalised_home_loan.dropna(axis=0)

[19]:
```

	SK_ID_CURR	TARGET	NAME_CONTRACT_TYPE	CODE_GENDER	FLAG_OWN_CAR	FLAG_OWN_REALTY	CNT_CHILDREN	AMT_INCOME_TOTAL	AMT_CREDIT
279124	423360	1	Cash loans	M	Y	N	1	157500.0	1125000.0
216116	350411	1	Cash loans	M	Y	N	0	112500.0	225000.0
133687	255050	1	Cash loans	M	Y	N	1	337500.0	704844.0
4159	104863	1	Cash loans	M	Y	N	0	265500.0	521280.0

Simple 0 Python 3 [3.10] | Idle Mode: Command Ln 1, Col 1 homeloan.ipynb

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homeloan.ipynb Python 3 [3.10]

```
[20]: print(normalised_home_loan.info())
print(normalised_home_loan.isnull().sum())

<class 'pandas.core.frame.DataFrame'>
Int64Index: 49659 entries, 207339 to 121862
Columns: 122 entries, SK_ID_CURR to AMT_REQ_CREDIT_BUREAU_YEAR
dtypes: float64(65), int64(41), object(16)
memory usage: 46.6+ MB
None
SK_ID_CURR          0
TARGET              0
NAME_CONTRACT_TYPE  0
CODE_GENDER         0
FLAG_OWN_CAR        0
...
AMT_REQ_CREDIT_BUREAU_DAY    7648
AMT_REQ_CREDIT_BUREAU_WEEK  7648
AMT_REQ_CREDIT_BUREAU_MON   7648
AMT_REQ_CREDIT_BUREAU_QRT   7648
AMT_REQ_CREDIT_BUREAU_YEAR  7648
Length: 122, dtype: int64

[21]: normalised_home_loan.TARGET.value_counts().plot(kind='pie', autopct='%1.1f%%')

[21]: <AxesSubplot: ylabel='TARGET'>
```

Simple 0 Python 3 [3.10] | Idle Mode: Command Ln 1, Col 1 homeloan.ipynb

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homeloan.ipynb Python 3 [3.10]

```
[20]: print(normalised_home_loan.info())
      print(normalised_home_loan.isnull().sum())

<class 'pandas.core.frame.DataFrame'>
Int64Index: 49650 entries, 207339 to 121862
Columns: 122 entries, SK_ID_CURR to AMT_REQ_CREDIT_BUREAU_YEAR
dtypes: float64(65), int64(41), object(16)
memory usage: 46.6+ MB
None
SK_ID_CURR                0
TARGET                    0
NAME_CONTRACT_TYPE        0
CODE_GENDER               0
FLAG_OWN_CAR              0
...
AMT_REQ_CREDIT_BUREAU_DAY  7648
AMT_REQ_CREDIT_BUREAU_WEEK 7648
AMT_REQ_CREDIT_BUREAU_MON  7648
AMT_REQ_CREDIT_BUREAU_QRT  7648
AMT_REQ_CREDIT_BUREAU_YEAR 7648
Length: 122, dtype: int64

[21]: normalised_home_loan.TARGET.value_counts().plot(kind='pie', autopct='%1.1f%%')

[21]: <AxesSubplot: ylabel='TARGET'>
```

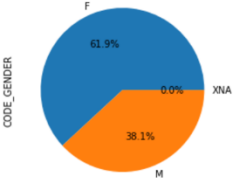
Simple 0 Python 3 [3.10] | Idle Mode: Command Ln 1, Col 1 homeloan.ipynb

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homeloan.ipynb Python 3 [3.10]

```
[23]: normalised_home_loan.CODE_GENDER.value_counts().plot(kind='pie', autopct='%1.1f%%')
      #roughly equal amount


[23]: <AxesSubplot: ylabel='CODE_GENDER'>
```



CODE_GENDER	Percentage
F	61.9%
M	38.1%
XNA	0.0%

```
[24]: normalised_home_loan.FLAG_OWN_CAR.value_counts().plot(kind='pie', autopct='%1.1f%%')

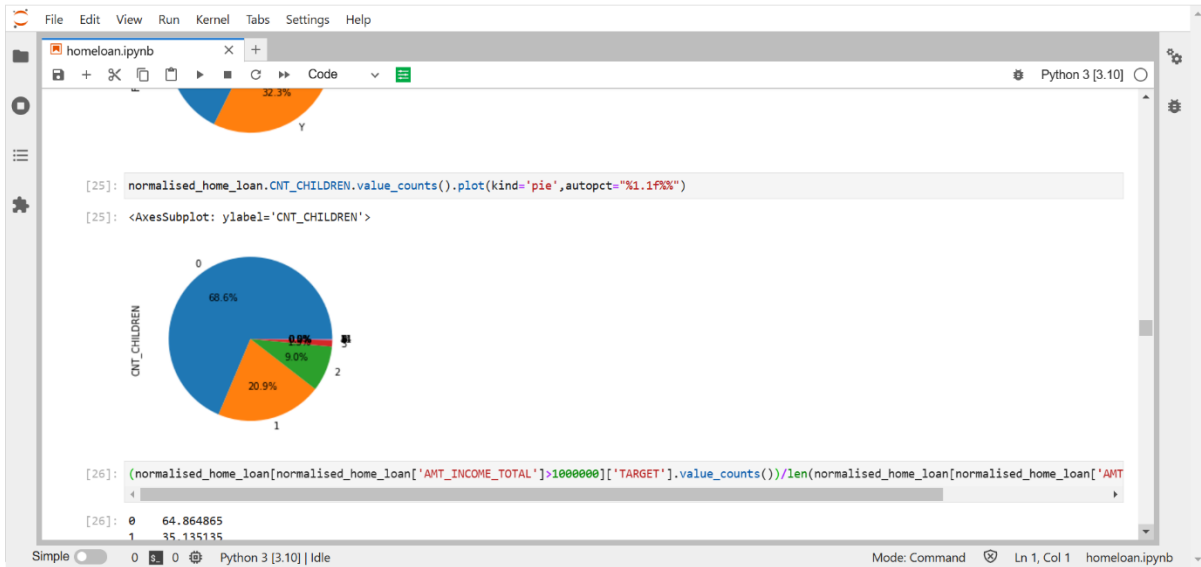
[24]: <AxesSubplot: ylabel='FLAG_OWN_CAR'>
```



FLAG_OWN_CAR	Percentage
N	67.7%
Y	32.3%

Simple 0 Python 3 [3.10] | Idle Mode: Command Ln 1, Col 1 homeloan.ipynb





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homeloan.ipynb Python 3 [3.10]

```
Name: TARGET, dtype: float64
[27]: #print((normalised_home_loan[normalised_home_loan['CNT_CHILDREN']>1]['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['CNT_CHILDREN']>1]))
print((normalised_home_loan[normalised_home_loan['CNT_CHILDREN']>2]['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['CNT_CHILDREN']>2]))
print((normalised_home_loan[normalised_home_loan['CNT_CHILDREN']>3]['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['CNT_CHILDREN']>3]))
#as number of children is increasing lone defaulters are increasing
1 57.047872
0 42.952128
Name: TARGET, dtype: float64
1 81.818182
0 18.181818
Name: TARGET, dtype: float64
[28]: print((normalised_home_loan[normalised_home_loan['FLAG_OWN_CAR']=='N']['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['FLAG_OWN_CAR']=='N']))
print((normalised_home_loan[normalised_home_loan['FLAG_OWN_CAR']=='Y']['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['FLAG_OWN_CAR']=='Y']))
1 51.350064
0 48.649936
Name: TARGET, dtype: float64
0 52.823962
1 47.176038
Name: TARGET, dtype: float64
[29]: print((normalised_home_loan[normalised_home_loan['CODE_GENDER']=='M']['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['CODE_GENDER']=='M']))
print((normalised_home_loan[normalised_home_loan['CODE_GENDER']=='F']['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['CODE_GENDER']=='F']))
```

Simple 0 Python 3 [3.10] | Idle Mode: Command Ln 1, Col 1 homeloan.ipynb

```
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homeloan.ipynb Python 3 [3.10]

Name: TARGET, dtype: float64

[27]: #print((normalised_home_loan[normalised_home_loan['CNT_CHILDREN']>1]['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['CNT_CHI
print((normalised_home_loan[normalised_home_loan['CNT_CHILDREN']>2]['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['CNT_CHI
print((normalised_home_loan[normalised_home_loan['CNT_CHILDREN']>5]['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['CNT_CHI
#as number of children is increasing lone defaulters are increasing

1 57.047872
0 42.952128
Name: TARGET, dtype: float64
1 81.818182
0 18.181818
Name: TARGET, dtype: float64

[28]: print((normalised_home_loan[normalised_home_loan['FLAG_OWN_CAR']=='N']['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['FLAG
print((normalised_home_loan[normalised_home_loan['FLAG_OWN_CAR']=='Y']['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['FLAG

1 51.350064
0 48.649936
Name: TARGET, dtype: float64
0 52.823962
1 47.176038
Name: TARGET, dtype: float64

[29]: print((normalised_home_loan[normalised_home_loan['CODE_GENDER']=='M']['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['CODE
print((normalised_home_loan[normalised_home_loan['CODE_GENDER']=='F']['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['CODE

Simple 0 0 Python 3 [3.10] | Idle Mode: Command Ln 1, Col 1 homeloan.ipynb
```

```
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homeloan.ipynb Python 3 [3.10]

[28]: print((normalised_home_loan[normalised_home_loan['FLAG_OWN_CAR']=='N']['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['FLAG
print((normalised_home_loan[normalised_home_loan['FLAG_OWN_CAR']=='Y']['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['FLAG

1 51.350064
0 48.649936
Name: TARGET, dtype: float64
0 52.823962
1 47.176038
Name: TARGET, dtype: float64

[29]: print((normalised_home_loan[normalised_home_loan['CODE_GENDER']=='M']['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['CODE
print((normalised_home_loan[normalised_home_loan['CODE_GENDER']=='F']['TARGET'].value_counts())/len(normalised_home_loan[normalised_home_loan['CODE

1 56.280372
0 43.719628
Name: TARGET, dtype: float64
0 53.867691
1 46.132309
Name: TARGET, dtype: float64

[30]: #men more Likely to default in payment of Loans

[31]: print((normalised_home_loan[normalised_home_loan['NAME_CONTRACT_TYPE']=='Cash_loans']['TARGET'].value_counts())/len(normalised_home_loan[normalised_
print((normalised_home_loan[normalised_home_loan['NAME_CONTRACT_TYPE']=='Revolving_loans']['TARGET'].value_counts())/len(normalised_home_loan[normal:
#cash Loans have a higher percent of defaulters

Simple 0 0 Python 3 [3.10] | Idle Mode: Command Ln 1, Col 1 homeloan.ipynb
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1 46.132309
Name: TARGET, dtype: float64

[30]: #men more Likely to default in payment of Loans

[31]: print((normalised_home_loan[normalised_home_loan['NAME_CONTRACT_TYPE']=='Cash loans']['TARGET'].value_counts())/len(normalised_home_loan[normalised_
print((normalised_home_loan[normalised_home_loan['NAME_CONTRACT_TYPE']=='Revolving loans']['TARGET'].value_counts())/len(normalised_home_loan[normal:
#Cash Loans have a higher percent of defaulters

Series([], Name: TARGET, dtype: float64)
Series([], Name: TARGET, dtype: float64)

[32]: normalised_home_loan=normalised_home_loan.sample(frac=1,random_state=5)

[33]: from sklearn.preprocessing import OrdinalEncoder
ordenc=OrdinalEncoder()
normalised_home_loan['NAME_CONTRACT_TYPE_CODE']=ordenc.fit_transform(normalised_home_loan[['NAME_CONTRACT_TYPE']])
print(normalised_home_loan[['NAME_CONTRACT_TYPE_CODE']].head(20))
print(normalised_home_loan['NAME_CONTRACT_TYPE_CODE'].value_counts())

NAME_CONTRACT_TYPE  NAME_CONTRACT_TYPE_CODE
302218      Cash loans                0.0
167526      Cash loans                0.0
159305      Cash loans                0.0
275427      Cash loans                0.0
8837        Cash loans                0.0
192094      Cash loans                0.0
```

```
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homeloan.ipynb x +
+ 🔍 📄 📄 ▶ ⏪ ⏩ Code Python 3 [3.10]

Name: NAME_CONTRACT_TYPE_CODE, dtype: int64

[34]: normalised_home_loan['CODE_GENDER_CODE']=ordenc.fit_transform(normalised_home_loan[['CODE_GENDER']])
print(normalised_home_loan[['CODE_GENDER','CODE_GENDER_CODE']].head(20))
print(normalised_home_loan['CODE_GENDER_CODE'].value_counts())

CODE_GENDER  CODE_GENDER_CODE
302218      M                1.0
167526      F                0.0
159305      M                1.0
275427      F                0.0
8837        M                1.0
192094      M                1.0
235115      F                0.0
79051      F                0.0
123267      M                1.0
5517        F                0.0
128624      M                1.0
187583      F                0.0
143193      M                1.0
288269      F                0.0
44320       F                0.0
256898      F                0.0
118237      F                0.0
5980        M                1.0
96475       F                0.0
249976      F                0.0
0.0      30716
1.0      18932
```

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homeloan.ipynb Python 3 [3.10]

```
Name: NAME_CONTRACT_TYPE_CODE, dtype: int64
```

```
[34]: normalised_home_loan['CODE_GENDER_CODE']=ordenc.fit_transform(normalised_home_loan[['CODE_GENDER']])
      print(normalised_home_loan[['CODE_GENDER','CODE_GENDER_CODE']].head(20))
      print(normalised_home_loan['CODE_GENDER_CODE'].value_counts())
```

	CODE_GENDER	CODE_GENDER_CODE
302218	M	1.0
167526	F	0.0
159385	M	1.0
275427	F	0.0
8837	M	1.0
192094	M	1.0
235115	F	0.0
79051	F	0.0
123267	M	1.0
5517	F	0.0
128624	M	1.0
187583	F	0.0
143193	M	1.0
288269	F	0.0
44320	F	0.0
256898	F	0.0
118237	F	0.0
5980	M	1.0
96475	F	0.0
249976	F	0.0
0.0	30716	
1.0	18932	

Simple 0 Python 3 [3.10] | Idle Mode: Command Ln 1, Col 1 homeloan.ipynb

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homeloan.ipynb Python 3 [3.10]

```
normalised_home_loan.loc[normalised_home_loan['CODE_GENDER_CODE']==2]
```

```
[35]:
```

SK_ID_CURR	TARGET	NAME_CONTRACT_TYPE	CODE_GENDER	FLAG_OWN_CAR	FLAG_OWN_REALTY	CNT_CHILDREN	AMT_INCOME_TOTAL	AMT_CREDIT	
83382	196708	0	Revolving loans	XNA	N	Y	1	135000.0	405000.0
189640	319880	0	Revolving loans	XNA	Y	Y	0	247500.0	540000.0

2 rows x 124 columns

```
[36]: normalised_home_loan['FLAG_OWN_CAR_CODE']=ordenc.fit_transform(normalised_home_loan[['FLAG_OWN_CAR']])
      print(normalised_home_loan[['FLAG_OWN_CAR','FLAG_OWN_CAR_CODE']].head(20))
      print(normalised_home_loan['FLAG_OWN_CAR_CODE'].value_counts())
```

	FLAG_OWN_CAR	FLAG_OWN_CAR_CODE
302218	N	0.0
167526	N	0.0
159385	N	0.0
275427	N	0.0
8837	N	0.0
192094	N	0.0
235115	N	0.0
79051	N	0.0
123267	N	0.0
5517	N	0.0
128624	N	0.0
187583	N	0.0

Simple 0 Python 3 [3.10] | Idle Mode: Command Ln 1, Col 1 homeloan.ipynb

```
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Python 3 [3.10]

[30]: normalised_home_loan=normalised_home_loan.sample(frac=1,random_state=45)

[39]: normalised_home_loan['TARGET'].value_counts()

[39]: 0    24825
      1    24825
      Name: TARGET, dtype: int64

[40]: y=normalised_home_loan.TARGET

[41]: #y=y.sample(frac=1,random_state=45)

[42]: normalised_home_loan_features=['SK_ID_CURR','NAME_CONTRACT_TYPE_CODE','CNT_CHILDREN_CODE','FLAG_OWN_CAR_CODE']

[43]: from sklearn.model_selection import train_test_split

[44]: X=normalised_home_loan[normalised_home_loan_features]

[45]: #X=X.sample(frac=1,random_state=45)

[46]: blobs_random_seed = 42
      centers = [(0,0), (5,5)]
      cluster_std = 1
      frac_test_split = 0.33
      num_features_for_samples = 2
      num_samples_total = 49650
      # Generate data

Simple 0 Python 3 [3.10] | Idle Mode: Command Ln 1, Col 1 homeloan.ipynb
```

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Python 3 [3.10]

[56]: import matplotlib.pyplot as plt
      import numpy
      from sklearn import metrics
      from sklearn import svm

[57]: clf=svm.SVC(kernel='linear')

[58]: clf=clf.fit(X_train,y_train)

[59]: predictions = clf.predict(X_test)

      # Generate confusion matrix
      matrix = plot_confusion_matrix(clf, X_test, y_test,

Simple 0 Python 3 [3.10] | Idle Mode: Command Ln 1, Col 1 homeloan.ipynb
```

