

## Kelompok 3 C2

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Data NPL untuk mendeteksi nasabah yang kreditanya akan macet. Field yang akan diprediksi adalah "flag\_kredit\_macet".

### IMPORT LIBRARY

```
In [1]: # display
%matplotlib inline
# numerik
import numpy as np
# analisis
import pandas as pd
# plot
import matplotlib.pyplot as plt
# visualisasi
import seaborn as sns
import warnings
warnings.filterwarnings("ignore")
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import classification_report
from sklearn.metrics import accuracy_score
from xgboost import XGBClassifier
from sklearn.tree import DecisionTreeClassifier
from sklearn.naive_bayes import GaussianNB
```

### READ FILE

```
In [2]: data=pd.read_csv("npl_train.txt")
data.head(3)
```

```
Out[2]:
```

|   | X | jumlah_kartu | outstanding | limit_kredit | tagihan   | total_pemakaian_tunai | total_pemakaian_retail | sis_a_tagihan_tidak_terbayar | kode_cabang | rasio_pembayaran | ... | flag_kredit_macet | jumlah_tahun |
|---|---|--------------|-------------|--------------|-----------|-----------------------|------------------------|------------------------------|-------------|------------------|-----|-------------------|--------------|
| 0 | 1 | 2            | 36158       | 7000000.0    | 23437.0   | 0.0                   | 94.0                   | 26323.0                      | I           | 102.19           | ... | 0                 |              |
| 1 | 2 | 2            | 268691      | 10000000.0   | 254564.0  | 0.0                   | 1012.0                 | 0.0                          | A           | 0.00             | ... | 0                 |              |
| 2 | 3 | 3            | 6769149     | 28000000.0   | 4159779.0 | 0.0                   | 0.0                    | 0.0                          | A           | 100.00           | ... | 0                 |              |

3 rows × 24 columns

```
In [3]: data.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 15493 entries, 0 to 15492
Data columns (total 24 columns):
#   Column              Non-Null Count  Dtype  
---  --
0   X                    15493 non-null  int64  
1   jumlah_kartu         15493 non-null  int64  
2   outstanding           15493 non-null  int64  
3   limit_kredit         15493 non-null  float64 
4   tagihan              15493 non-null  float64 
5   total_pemakaian_tunai 15493 non-null  float64 
6   total_pemakaian_retail 15493 non-null  float64 
7   sis_a_tagihan_tidak_terbayar 15493 non-null  float64 
8   kode_cabang          15393 non-null  object  
9   rasio_pembayaran     15493 non-null  float64 
10  persentase_overlimit  15493 non-null  float64 
11  rasio_pembayaran_3bulan 15493 non-null  float64 
12  rasio_pembayaran_6bulan 15493 non-null  float64 
13  skor_delikuenasi      15493 non-null  int64  
14  flag_kredit_macet     15493 non-null  int64  
15  jumlah_tahun_sejak_pembukaan_kredit 15493 non-null  float64 
16  total_pemakaian      15493 non-null  float64 
17  sis_a_tagihan_per_jumlah_kartu 15493 non-null  float64 
18  sis_a_tagihan_per_limit 15493 non-null  float64 
19  total_pemakaian_per_limit 15493 non-null  float64 
20  pemakaian_3bln_per_limit 15493 non-null  float64 
21  pemakaian_6bln_per_limit 15493 non-null  float64 
22  utilisasi_3bulan      15493 non-null  float64 
23  utilisasi_6bulan      15493 non-null  float64 
dtypes: float64(18), int64(5), object(1)
memory usage: 2.8+ MB
```

### Menghapus Kode Cabang

kode\_cabang dan X dihapus karena tidak memiliki kaitan dengan hasil akhir.

```
In [4]: data = data.drop(["kode_cabang","X"], axis=1)
```

```
In [5]: data["flag_kredit_macet"].value_counts()
```

```
Out[5]:
0    14134
1     1359
Name: flag_kredit_macet, dtype: int64
```

### SPLIT DATA

```
In [6]: from sklearn.model_selection import train_test_split
```

```
In [7]: data_train, data_test=train_test_split(data,test_size=0.4,random_state=1)
```

```
In [8]: data_train.head(5)
```

```
Out[8]:
```

|       | jumlah_kartu | outstanding | limit_kredit | tagihan    | total_pemakaian_tunai | total_pemakaian_retail | sis_a_tagihan_tidak_terbayar | rasio_pembayaran | persentase_overlimit | rasio_pembayaran_3bulan |
|-------|--------------|-------------|--------------|------------|-----------------------|------------------------|------------------------------|------------------|----------------------|-------------------------|
| 10502 | 4            | 24962677    | 23000000.0   | 22575840.0 | 0.0                   | 2957908.0              | 25136225.0                   | 0.0              | 8.62                 | 7.88                    |
| 14010 | 2            | 7474329     | 7000000.0    | 2961653.0  | 0.0                   | 0.0                    | 2133750.0                    | 60.9             | 2.07                 | 83.90                   |
| 5089  | 3            | 596113      | 44000000.0   | 530033.0   | 0.0                   | 0.0                    | 0.0                          | 0.0              | 0.00                 | 18.10                   |
| 7683  | 3            | 4851906     | 10000000.0   | 4748881.0  | 0.0                   | 1130000.0              | 4746073.0                    | 23.1             | 0.00                 | 30.50                   |
| 5692  | 2            | 76966       | 80000000.0   | 78000.0    | 0.0                   | 196.0                  | 0.0                          | 0.0              | 0.00                 | 25.20                   |

5 rows × 22 columns

```
In [9]: data_test=data_test.drop("flag_kredit_macet", 1)
```

```
In [10]: data_train.info()
```

```
<class 'pandas.core.frame.DataFrame'>
Int64Index: 9295 entries, 10502 to 13349
Data columns (total 22 columns):
#   Column              Non-Null Count  Dtype  
---  --
0   jumlah_kartu         9295 non-null  int64  
1   outstanding           9295 non-null  int64  
2   limit_kredit         9295 non-null  float64 
3   tagihan              9295 non-null  float64 
4   total_pemakaian_tunai 9295 non-null  float64 
5   total_pemakaian_retail 9295 non-null  float64 
6   sis_a_tagihan_tidak_terbayar 9295 non-null  float64 
7   rasio_pembayaran     9295 non-null  float64 
8   persentase_overlimit  9295 non-null  float64 
9   rasio_pembayaran_3bulan 9295 non-null  float64 
10  rasio_pembayaran_6bulan 9295 non-null  float64 
11  skor_delikuenasi      9295 non-null  int64  
12  flag_kredit_macet     9295 non-null  int64  
13  jumlah_tahun_sejak_pembukaan_kredit 9295 non-null  float64 
14  total_pemakaian      9295 non-null  float64 
15  sis_a_tagihan_per_jumlah_kartu 9295 non-null  float64 
16  sis_a_tagihan_per_limit 9295 non-null  float64 
17  total_pemakaian_per_limit 9295 non-null  float64 
18  pemakaian_3bln_per_limit 9295 non-null  float64 
19  pemakaian_6bln_per_limit 9295 non-null  float64 
20  utilisasi_3bulan      9295 non-null  float64 
21  utilisasi_6bulan      9295 non-null  float64 
dtypes: float64(18), int64(4)
memory usage: 1.6 MB
```

### Kelas Target

```
In [11]: X=data_train.drop("flag_kredit_macet",1)
```

```
In [12]: X.head(5)
```

```
Out[12]:
```

|       | jumlah_kartu | outstanding | limit_kredit | tagihan    | total_pemakaian_tunai | total_pemakaian_retail | sis_a_tagihan_tidak_terbayar | rasio_pembayaran | persentase_overlimit | rasio_pembayaran_3bulan |
|-------|--------------|-------------|--------------|------------|-----------------------|------------------------|------------------------------|------------------|----------------------|-------------------------|
| 10502 | 4            | 24962677    | 23000000.0   | 22575840.0 | 0.0                   | 2957908.0              | 25136225.0                   | 0.0              | 8.62                 | 7.88                    |
| 14010 | 2            | 7474329     | 7000000.0    | 2961653.0  | 0.0                   | 0.0                    | 2133750.0                    | 60.9             | 2.07                 | 83.90                   |
| 5089  | 3            | 596113      | 44000000.0   | 530033.0   | 0.0                   | 0.0                    | 0.0                          | 0.0              | 0.00                 | 18.10                   |
| 7683  | 3            | 4851906     | 10000000.0   | 4748881.0  | 0.0                   | 1130000.0              | 4746073.0                    | 23.1             | 0.00                 | 30.50                   |
| 5692  | 2            | 76966       | 80000000.0   | 78000.0    | 0.0                   | 196.0                  | 0.0                          | 0.0              | 0.00                 | 25.20                   |

5 rows × 21 columns

```
In [13]: y=data_train[["flag_kredit_macet"]]
```

```
In [14]: y.head(5)
```

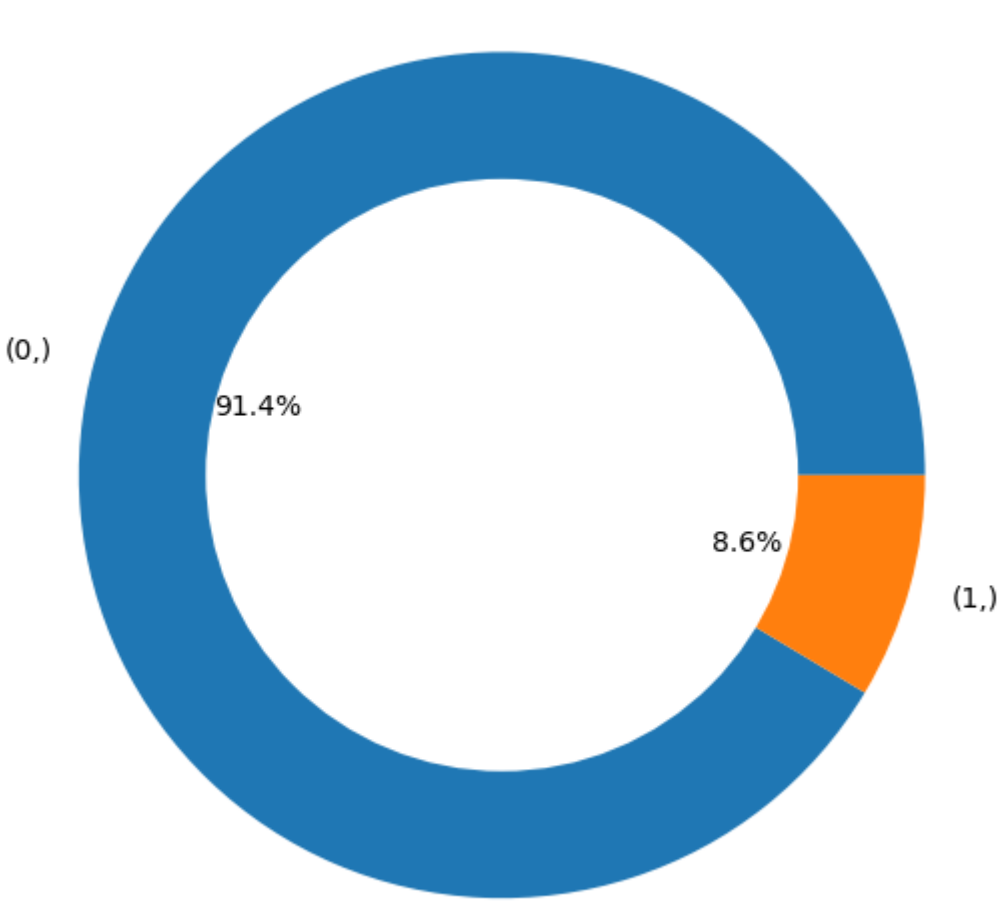
```
Out[14]:
```

|       | flag_kredit_macet |
|-------|-------------------|
| 10502 | 0                 |
| 14010 | 0                 |
| 5089  | 1                 |
| 7683  | 0                 |
| 5692  | 1                 |

### DATA IMBALANCE

```
In [15]: plt.figure(figsize=(15,8))
plt.subplot(1,2,1)
y.value_counts().plot.pie(autopct='%1.1f%%')
centre=plt.Circle((0,0),0.7,fc='white')
fig=plt.gcf()
fig.gca().add_artist(centre)
```

```
Out[15]: <matplotlib.patches.Circle at 0x230ef80c110>
```

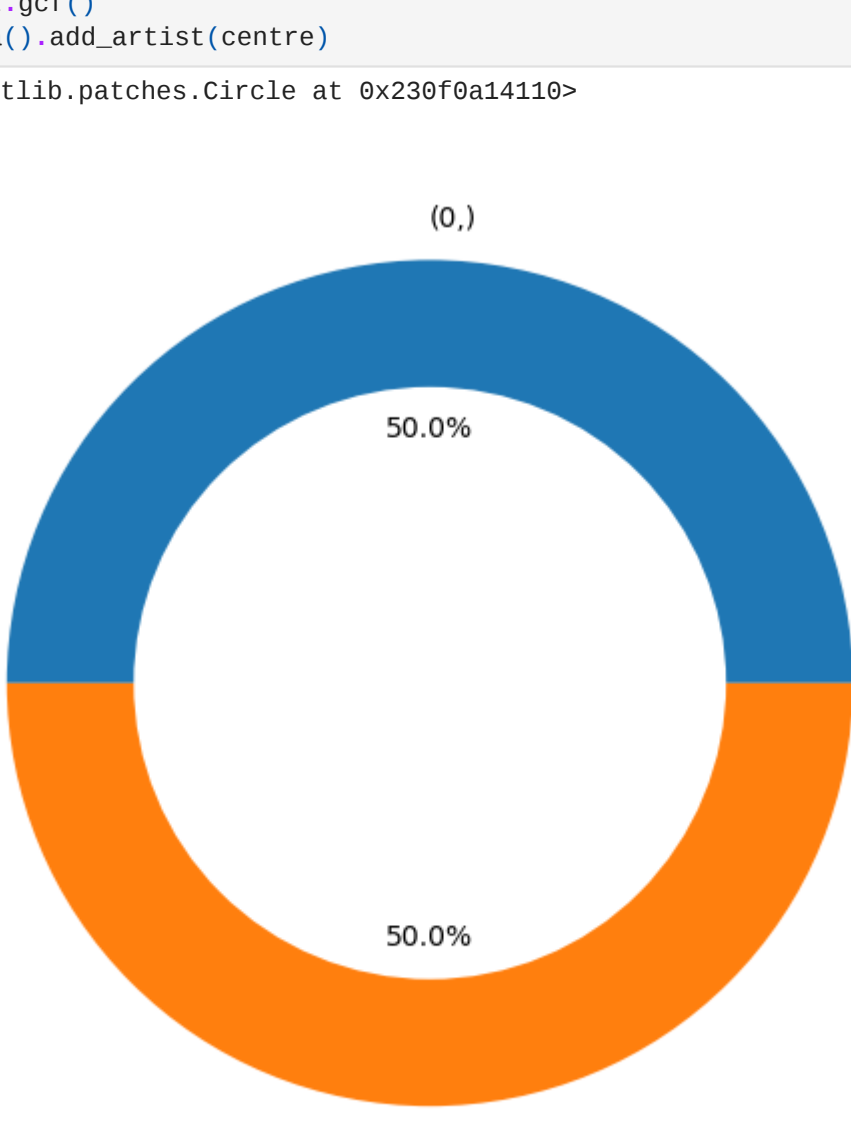


```
In [16]: from imblearn.over_sampling import SMOTE
over=SMOTE()
X,y=over.fit_resample(X,y)
```

```
In [17]: x_train,x_cv,y_train,y_cv=train_test_split(X,y,test_size=0.2,random_state=1)
```

```
In [18]: plt.figure(figsize=(15,8))
plt.subplot(1,2,1)
y.value_counts().plot.pie(autopct='%1.1f%%')
centre=plt.Circle((0,0),0.7,fc='white')
fig=plt.gcf()
fig.gca().add_artist(centre)
```

```
Out[18]: <matplotlib.patches.Circle at 0x230f0a14110>
```



### METODE NAIVE BAYES

```
In [19]: nb_model = GaussianNB()
nb_model.fit(x_train, y_train)
Y_pred = nb_model.predict(x_cv)
acc = accuracy_score(y_cv, Y_pred)*100
```

```
print("Akurasi {}".format(acc))
print(classification_report(y_cv, Y_pred))
```

```
Akurasi 54.81023830538394
precision    recall  f1-score   support

   0       0.82     0.11     0.19     1680
   1       0.53     0.98     0.69     1719

 accuracy          0.68
 macro avg         0.67
weighted avg         0.67
```

### METODE DECISION TREE

```
In [20]: tree_model = DecisionTreeClassifier(random_state=1)
tree_model.fit(x_train,y_train)
pred_cv_tree=tree_model.predict(x_cv)
score_tree = accuracy_score(pred_cv_tree,y_cv)*100
print("Akurasi: ", score_tree)
print(classification_report(y_cv, pred_cv_tree))
```

```
akurasi: 86.70197116799059
precision    recall  f1-score   support

   0       0.87     0.85     0.86     1680
   1       0.86     0.88     0.87     1719

 accuracy          0.87
 macro avg         0.87
weighted avg         0.87
```

### METODE RANDOM FOREST

```
In [21]: forest_model = RandomForestClassifier(random_state=1,max_depth=10,n_estimators=50)
forest_model.fit(x_train,y_train)
pred_cv_forest=forest_model.predict(x_cv)
score_forest = accuracy_score(pred_cv_forest,y_cv)*100
print("Akurasi: ", score_forest)
print(classification_report(y_cv, pred_cv_forest))
```

```
akurasi: 89.14386584289497
precision    recall  f1-score   support

   0       0.92     0.86     0.89     1680
   1       0.87     0.92     0.90     1719

 accuracy          0.89
 macro avg         0.89
weighted avg         0.89
```

### METODE XGBoost

```
In [22]: xgb_model = XGBClassifier(n_estimators=50,max_depth=7)
xgb_model.fit(x_train,y_train)
pred_xgb=xgb_model.predict(x_cv)
score_xgb = accuracy_score(pred_xgb,y_cv)*100
print("Akurasi : ", score_xgb)
print(classification_report(y_cv, pred_xgb))
```

```
Akurasi : 93.14504265960576
precision    recall  f1-score   support

   0       0.95     0.91     0.93     1680
   1       0.92     0.95     0.93     1719

 accuracy          0.93
 macro avg         0.93
weighted avg         0.93
```

```
In [23]: pred_xgb_test=xgb_model.predict(data_test)
pred_xgb_test
```

```
Out[23]: array([0, 0, 0, ..., 0, 0, 1])
```

```
In [24]: df_pred_result=data_test
df_pred_result["Prediksi_kredit"]=pred_xgb_test
```

```
In [25]: df_pred_result
```

```
Out[25]:
```

|       | jumlah_kartu | outstanding | limit_kredit | tagihan    | total_pemakaian_tunai | total_pemakaian_retail | sis_a_tagihan_tidak_terbayar | rasio_pembayaran | persentase_overlimit | rasio_pembayaran_3bulan |
|-------|--------------|-------------|--------------|------------|-----------------------|------------------------|------------------------------|------------------|----------------------|-------------------------|
| 3576  | 2            | 8096704     | 8000000.0    | 4073734.0  | 0.0                   | 1901025.0              | 5213854.0                    | 65.5             | 1.24                 | 37.90                   |
| 8560  | 2            | 684568      | 8000000.0    | 47263.0    | 0.0                   | 538655.0               | 0.0                          | 100.0            | 0.00                 | 100.00                  |
| 13670 | 3            | 3617686     | 11000000.0   | 1933276.0  | 0.0                   | 586000.0               | 3155317.0                    | 100.0            | 0.00                 | 100.00                  |
| 13243 | 2            | 3110979     | 30000000.0   | 3007042.0  | 0.0                   | 100000.0               | 3097253.0                    | 10.1             | 3.84                 | 12.40                   |
| 804   | 2            | 139723      | 4000000.0    | 148961.0   | 0.0                   | 0.0                    | 0.0                          | 0.0              | 0.00                 | 39.46                   |
| ...   | ...          | ...         | ...          | ...        | ...                   | ...                    | ...                          | ...              | ...                  | ...                     |
| 8367  | 4            | 4732242     | 40000000.0   | 4505282.0  | 0.0                   | 165000.0               | 4766229.0                    | 46.7             | 18.90                | 17.10                   |
| 5691  | 2            | 2294072     | 4000000.0    | 1115000.0  | 0.0                   | 848587.0               | 1925453.0                    | 81.7             | 0.00                 | 98.80                   |
| 5995  | 6            | 22629584    | 18000000.0   | 20208542.0 | 0.0                   | 3927461.0              | 22643270.0                   | 40.1             | 11.90                | 58.00                   |
| 12277 | 2            | 8153311     | 8000000.0    | 3644008.0  | 0.0                   | 946400.0               | 7961240.0                    | 0.0              | 1.24                 | 97.00                   |
| 12844 | 2            | 8479101     | 7000000.0    | 8520816.0  | 0.0                   | 0.0                    | 8464004.0                    | 0.0              | 21.70                | 45.50                   |

6198 rows × 22 columns

```
In [26]: from sklearn.metrics import confusion_matrix
confusion_matrix(pred_xgb,y_cv)
```

```
Out[26]: array([[1528,    81],
               [ 152, 1638]], dtype=int64)
```