

# Credit Card Customer Analysis

BTPN Syariah Data Engineer



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#### About me

I am a 5th semester student majoring in informatics at Mulia University, I am very interested in learning data science and data visualization.

Therefore I am still learning to process data, analyze data, use data science tools, data visualization to become a great data scientist



Business Objective Data Exploration

Insight Presentation

Conclusion & Recommendations

### Ol Business Objective



Reducing the churn rate (termination of service)
of credit card customers by identifying and
understanding the profile of customers who
tend to abandon service.

 Increase customer retention by providing better service and being proactive in responding to customer needs.

#### A. Import library dan Input Datasets

```
[7] !pip install dataprep

# import library dataprep
from dataprep.eda import plot, plot_missing, plot_correlation, plot_diff, create_report

> [8] from dataprep.datasets import get_dataset_names, load_dataset

> # cek dataset yang get_dataset_names()
```

```
import pandas as pd
    df = pd.read excel(r"customer Data history.xlsx")
    print(df)
ⅎ
          CLIENTNUM idstatus Customer Age Gender Dependent count Educationid
           768805383
           818770008
          713982108
           769911858
           709106358
    10122 772366833
    10123 710638233
    10124 716506083
    10125 717406983
    10126 714337233
          Maritalid Income_Category card_categoryid Months on book \
                        $60K - $80K
                     Less than $40K
                                                                 44
                        $80K - $120K
                     Less than $40K
                         $60K - $80K
```

# **O2**Data Exploration

#### **B.** Datasets informations



```
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 10127 entries, 0 to 10126
Data columns (total 19 columns):
 # Column
                              Non-Null Count Dtype
    CLIENTNUM
                              10127 non-null int64
     idstatus
                               10127 non-null int64
                               10127 non-null int64
     Customer Age
     Gender
                               10127 non-null object
     Dependent count
                               10127 non-null int64
     Educationid
                               10127 non-null int64
     Maritalid
                               10127 non-null int64
     Income Category
                               10127 non-null object
    card categoryid
                               10127 non-null int64
     Months on book
                               10127 non-null int64
 10 Total Relationship Count 10127 non-null int64
 11 Months Inactive 12 mon
                              10127 non-null int64
 12 Contacts Count 12 mon
                               10127 non-null int64
 13 Credit Limit
                               10127 non-null float64
 14 Total Revolving Bal
                              10127 non-null int64
 15 Avg Open To Buy
                               10127 non-null float64
 16 Total_Trans_Amt
                              10127 non-null int64
 17 Total Trans Ct
                               10127 non-null int64
 18 Avg Utilization Ratio
                              10127 non-null float64
dtypes: float64(3), int64(14), object(2)
memory usage: 1.5+ MB
```

Provides detailed information about the DataFrame, including:

- The total number of entries (rows) in the DataFrame.
- The number and type of each column (column name, number of non-nulls, and data type).
- Memory usage by DataFrame.

#### C. Change data type



```
▶ #ubah tipe data ke category
     df["Income_Category"]=df["Income_Category"].astype("category")
     df["Gender"]=df["Gender"].astype("category")
     df["Customer_Age"]=df["Customer_Age"].astype("category")
[13] df["Total_Revolving_Bal"] = pd.to_numeric(df["Total_Revolving_Bal"], errors='coerce')
     numeric_df = df.select_dtypes(include=['number'])
[ ] #Returns column data types
     df.dtypes
     CLIENTNUM
                                     int64
     idstatus
                                     int64
     Customer Age
                                 category
     Gender
                                 category
     Dependent count
                                     int64
     Educationid
                                     int64
     Maritalid
                                     int64
     Income Category
                                  category
     card categoryid
                                     int64
     Months on book
                                    int64
     Total Relationship Count
                                    int64
     Months Inactive 12 mon
                                    int64
     Contacts Count 12 mon
                                    int64
     Credit Limit
                                   float64
     Total Revolving Bal
                                    int64
                                   float64
     Avg Open To Buy
     Total Trans Amt
                                     int64
     Total Trans Ct
                                    int64
     Avg Utilization Ratio
                                  float64
     dtype: object
```

#### Changed some data type columns

- Income\_Category column
- Gender Column
- Customer\_Age column

#### D. Checking for missing values in datasets





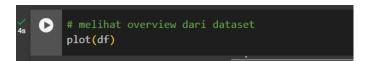
<b>√</b> 5s	0	<pre># memeriksa missing value plot_missing(df)</pre>					
		Computing isnull-73e05c0938b000e27f71c94c1d51e9f4: 0%  return func(*(_execute_task(a, cache) for a in args))				0/214 [00:00 ,</th	
		Stats	Bar Chart	Spectrum	Heat Map	Dendrogram	
		Missing Statistics					
		Missing Cells				0	
	Missing Cells (%)					0.0%	
		Missing Columns 0					_
		Missing Rows 0 Avg Missing Cells per Column 0.0					_
							_
		Avg Missing Cells per Row 0.0					
		<					

Tidak terdapat missing value pada datasets

#### D. View an overview of datasets







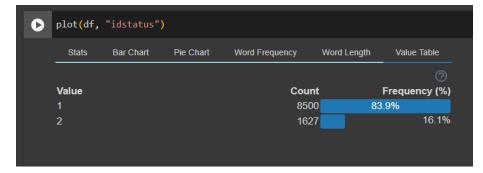


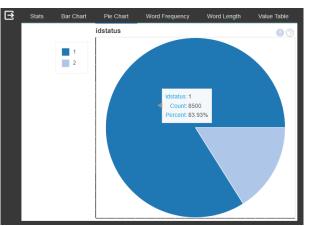
### 03

### Insight and Visualization

1. = Existing 2. = Attried







From the total data, 84% of customers still use credit card services, 16% of the data sample stopped using credit card services

#### **B.** Education







1. = Highschool
2. = Graduate
3. = Uneducated
4. = unknown
5. = Collage
6. = Post Graduate
7. = Doctorate

Customers with an educational background of 'uneducated' potentially show the highest likelihood of churn

#### C. Utilization Ratio

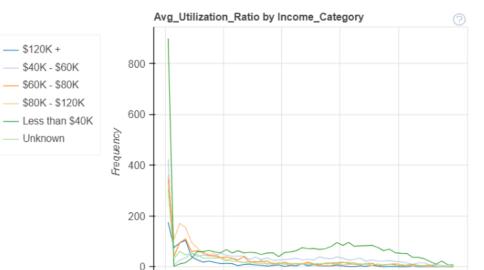
0

0.2









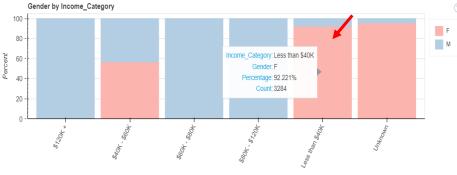
0.4

0.6

Avg Utilization Ratio

0.8

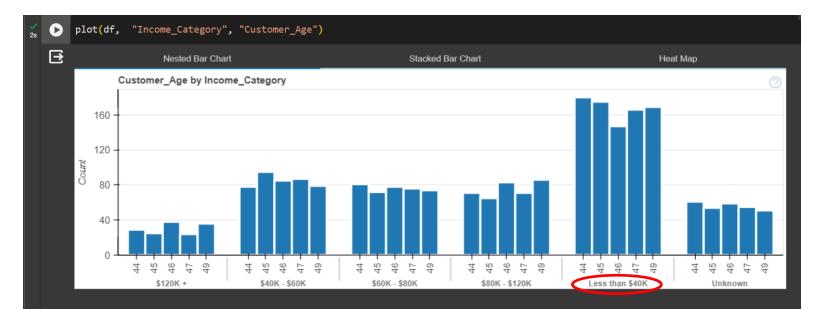
Customers in the income category of less than \$40K potentially show the highest likelihood of churn, 92% of customers are female



#### C. Utilization Ratio

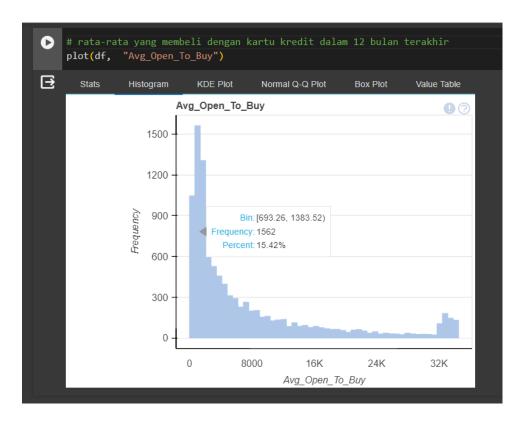






Based on this visualization, customers with an age range of 44 years - 49 years with an income category of "Less than \$40K" potentially show the highest likelihood of churn

#### D. Customer membeli dengan kartu kredit dalam 12 bulan terakhir



The majority of customers use credit cards to make transactions, namely the average number of customers who purchase with credit cards is 1,200 and 900.





## Conclusion & Recommendations

- Provide proactive recommendations to increase customer retention, such as loyalty programs, special offers, or customer service improvements.
- Develop strategies to identify customers who have the potential to abandon service and determine appropriate preventive measures.
- Present suggestions to improve service quality, based on findings from data analysis, and respond to customer feedback.
- Summarize key findings and present a clear action plan to increase customer retention and reduce churn rates.



## THANK YOU