

Importing Data dan Inspection

Membaca dataset dari berkas .csv, dan melihat informasi dari dataset tersebut.

```
import pandas as pd

df = pd.read_csv('data_retail.csv', sep=';')

print('Lima data teratas:')
print(df.head())

print('\nInfo dataset:')
print(df.info())

Lima data teratas:
   no  Row_Num  Customer_ID Product  First_Transaction
Last_Transaction \
0    1         1        29531  Jaket      1466304274396
1538718482608
1    2         2        29531  Sepatu     1406077331494
1545735761270
2    3         3       141526    Tas      1493349147000
1548322802000
3    4         4       141526  Jaket      1493362372547
1547643603911
4    5         5       37545  Sepatu     1429178498531
1542891221530

   Average_Transaction_Amount  Count_Transaction
0                  1467681                 22
1                  1269337                 41
2                  310915                  30
3                  722632                  27
4                 1775036                 25

Info dataset:
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100000 entries, 0 to 99999
Data columns (total 8 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   no              100000 non-null   int64  
 1   Row_Num         100000 non-null   int64  
 2   Customer_ID     100000 non-null   int64  
 3   Product          100000 non-null   object  
 4   First_Transaction 100000 non-null   int64  
 5   Last_Transaction 100000 non-null   int64  
 6   Average_Transaction_Amount 100000 non-null   int64  
 7   Count_Transaction 100000 non-null   int64
```

```
dtypes: int64(7), object(1)
memory usage: 6.1+ MB
None
```

Data Cleansing

Dua kolom yang menunjukkan terjadinya transaksi tidak bertipe datetime, maka perlu untuk mengubah kedua kolom tersebut ke tipe data datetime.

```
import pandas as pd
df = pd.read_csv('data_retail.csv', sep=';')

# Kolom First_Transaction
df['First_Transaction'] = pd.to_datetime(df['First_Transaction']/1000,
unit='s', origin='1970-01-01')
# Kolom Last_Transaction
df['Last_Transaction'] = pd.to_datetime(df['Last_Transaction']/1000,
unit='s', origin='1970-01-01')

print('Lima data teratas:')
print(df.head())

print('\nInfo dataset:')
print(df.info())

Lima data teratas:
   no  Row_Num  Customer_ID Product      First_Transaction \
0   1          1        29531  Jaket 2016-06-19 02:44:34.3960000000
1   2          2        29531  Sepatu 2014-07-23 01:02:11.493999872
2   3          3       141526    Tas 2017-04-28 03:12:27.0000000000
3   4          4       141526  Jaket 2017-04-28 06:52:52.546999808
4   5          5        37545  Sepatu 2015-04-16 10:01:38.530999808

                                         Last_Transaction  Average_Transaction_Amount
Count_Transaction
0 2018-10-05 05:48:02.6080000000                      1467681
22
1 2018-12-25 11:02:41.269999872                     1269337
41
2 2019-01-24 09:40:02.0000000000                      310915
30
3 2019-01-16 13:00:03.911000064                     722632
27
4 2018-11-22 12:53:41.529999872                     1775036
25

Info dataset:
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100000 entries, 0 to 99999
Data columns (total 8 columns):
```

```

#   Column           Non-Null Count  Dtype  
---  --  
0    no              100000 non-null   int64  
1    Row_Num         100000 non-null   int64  
2    Customer_ID    100000 non-null   int64  
3    Product         100000 non-null   object  
4    First_Transaction 100000 non-null   datetime64[ns] 
5    Last_Transaction 100000 non-null   datetime64[ns] 
6    Average_Transaction_Amount 100000 non-null   int64  
7    Count_Transaction 100000 non-null   int64  
dtypes: datetime64[ns](2), int64(5), object(1)
memory usage: 6.1+ MB
None

```

Churn Customers

Untuk menentukan churn customers sesuai definisi yang telah diberikan, maka mencari

- transaksi paling terakhir kapan dilakukan
- mengklasifikasikan mana customer yang berstatus churn dan mana yang tidak.

```

import pandas as pd
df = pd.read_csv('data_retail.csv', sep=';')
df['First_Transaction'] = pd.to_datetime(df['First_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Last_Transaction'] = pd.to_datetime(df['Last_Transaction']/1000,
unit='s', origin='1970-01-01')

# Pengecekan transaksi terakhir dalam dataset
print(max(df['Last_Transaction']))

# Klasifikasikan customer yang berstatus churn atau tidak dengan
# boolean
df.loc[df['Last_Transaction'] <= '2018-08-01', 'is_churn'] = True
df.loc[df['Last_Transaction'] > '2018-08-01', 'is_churn'] = False

print('Lima data teratas:')
print(df.head())

print('\nInfo dataset:')
print(df.info())

2019-02-01 23:57:57.286000128
Lima data teratas:
   no  Row_Num  Customer_ID Product      First_Transaction 
0   1          1        29531  Jaket 2016-06-19 02:44:34.396000000
1   2          2        29531  Sepatu 2014-07-23 01:02:11.493999872
2   3          3       141526   Tas 2017-04-28 03:12:27.000000000
3   4          4       141526  Jaket 2017-04-28 06:52:52.546999808
4   5          5        37545  Sepatu 2015-04-16 10:01:38.530999808

```

```
      Last_Transaction  Average_Transaction_Amount \
0 2018-10-05 05:48:02.608000000                1467681
1 2018-12-25 11:02:41.269999872                1269337
2 2019-01-24 09:40:02.000000000                310915
3 2019-01-16 13:00:03.911000064                722632
4 2018-11-22 12:53:41.529999872                1775036
```

```
Count_Transaction is_churn
0            22    False
1            41    False
2            30    False
3            27    False
4            25    False
```

Info dataset:

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 100000 entries, 0 to 99999
Data columns (total 9 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   no              100000 non-null   int64  
 1   Row_Num         100000 non-null   int64  
 2   Customer_ID     100000 non-null   int64  
 3   Product          100000 non-null   object  
 4   First_Transaction 100000 non-null   datetime64[ns]
 5   Last_Transaction 100000 non-null   datetime64[ns]
 6   Average_Transaction_Amount 100000 non-null   int64  
 7   Count_Transaction 100000 non-null   int64  
 8   is_churn        100000 non-null   object  
dtypes: datetime64[ns](2), int64(5), object(2)
memory usage: 6.9+ MB
None
```

Menghapus kolom yang tidak diperlukan

```
import pandas as pd
df = pd.read_csv('data_retail.csv', sep=';')
df['First_Transaction'] = pd.to_datetime(df['First_Transaction']/1000,
                                         unit='s', origin='1970-01-01')
df['Last_Transaction'] = pd.to_datetime(df['Last_Transaction']/1000,
                                         unit='s', origin='1970-01-01')

# Hapus kolom-kolom yang tidak diperlukan
del df['no']
del df['Row_Num']

# Cetak lima data teratas
print(df.head())

      Customer_ID Product           First_Transaction \
0            29531    Jaket 2016-06-19 02:44:34.396000000
```

```

1      29531 Sepatu 2014-07-23 01:02:11.493999872
2      141526 Tas    2017-04-28 03:12:27.000000000
3      141526 Jaket   2017-04-28 06:52:52.546999808
4      37545 Sepatu 2015-04-16 10:01:38.530999808

```

	Last_Transaction	Average_Transaction_Amount
Count_Transaction		
0	2018-10-05 05:48:02.608000000	1467681
22		
1	2018-12-25 11:02:41.269999872	1269337
41		
2	2019-01-24 09:40:02.000000000	310915
30		
3	2019-01-16 13:00:03.911000064	722632
27		
4	2018-11-22 12:53:41.529999872	1775036
25		

Customer acquisition by year

Mem-visualisasikan trend jumlah customer per tahunnya dengan menggunakan bar chart.

```

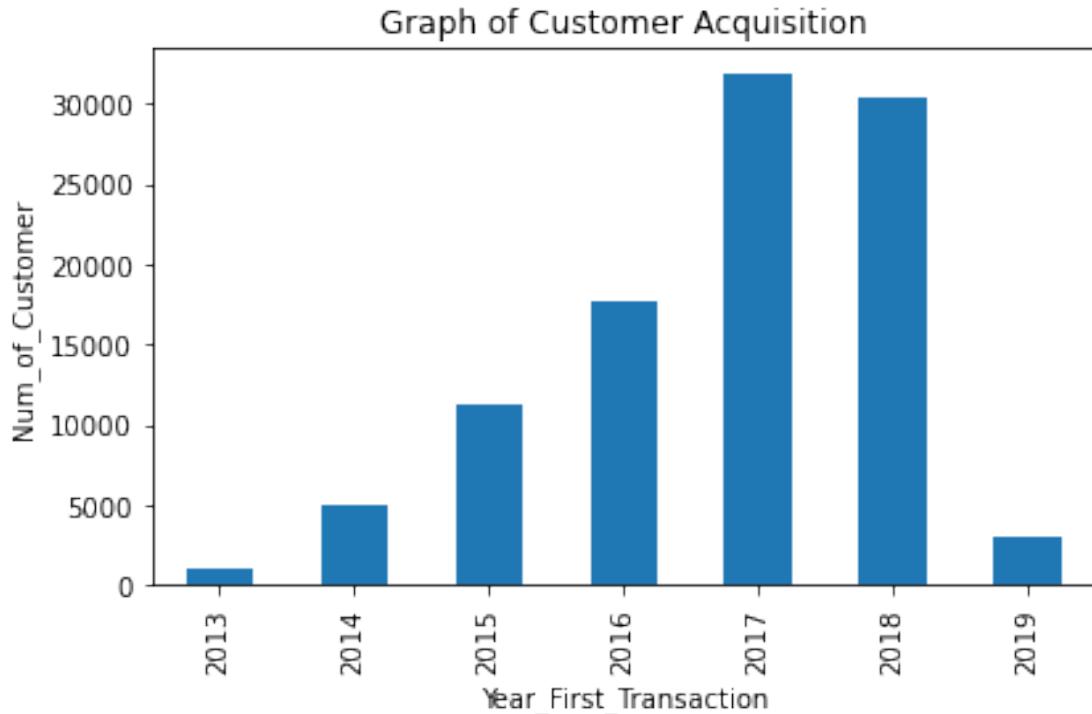
import pandas as pd
df = pd.read_csv('data_retail.csv', sep=';')
df['First_Transaction'] = pd.to_datetime(df['First_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Last_Transaction'] = pd.to_datetime(df['Last_Transaction']/1000,
unit='s', origin='1970-01-01')

import matplotlib.pyplot as plt

# Kolom tahun transaksi pertama
df['Year_First_Transaction'] = df['First_Transaction'].dt.year
# Kolom tahun transaksi terakhir
df['Year_Last_Transaction'] = df['Last_Transaction'].dt.year

df_year = df.groupby(['Year_First_Transaction'])
['Customer_ID'].count()
df_year.plot(x='Year_First_Transaction', y='Customer_ID', kind='bar',
title='Graph of Customer Acquisition')
plt.xlabel('Year_First_Transaction')
plt.ylabel('Num_of_Customer')
plt.tight_layout()
plt.show()

```



Transaction by year

Mem-visualisasikan trend jumlah transaksi per tahunnya dengan menggunakan bar chart.

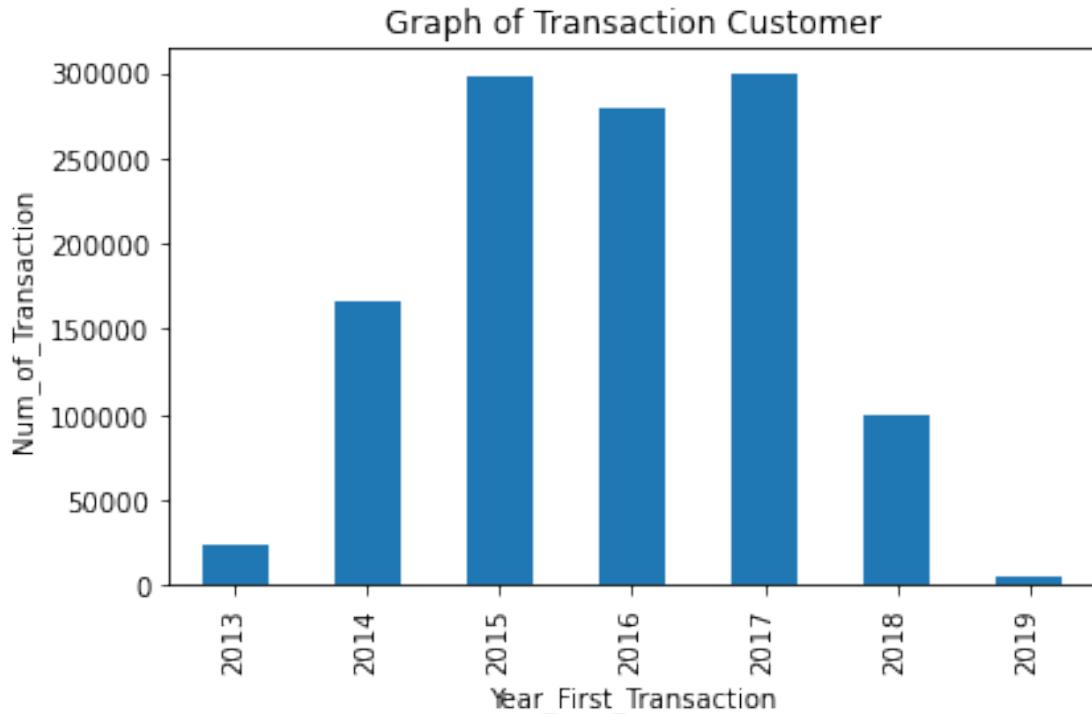
```

import pandas as pd
df = pd.read_csv('data_retail.csv', sep=';')
df['First_Transaction'] = pd.to_datetime(df['First_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Last_Transaction'] = pd.to_datetime(df['Last_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Year_First_Transaction'] = df['First_Transaction'].dt.year
df['Year_Last_Transaction'] = df['Last_Transaction'].dt.year

import matplotlib.pyplot as plt

plt.clf()
df_year = df.groupby(['Year_First_Transaction'])
['Count_Transaction'].sum()
df_year.plot(x='Year_First_Transaction', y='Count_Transaction',
kind='bar', title='Graph of Transaction Customer')
plt.xlabel('Year_First_Transaction')
plt.ylabel('Num_of_Transaction')
plt.tight_layout()
plt.show()

```



Average transaction amount by year

Mem-visualisasikanlah tren dari tahun ke tahun rata-rata jumlah transaksi untuk tiap-tiap produknya dengan menggunakan seaborn pointplot.

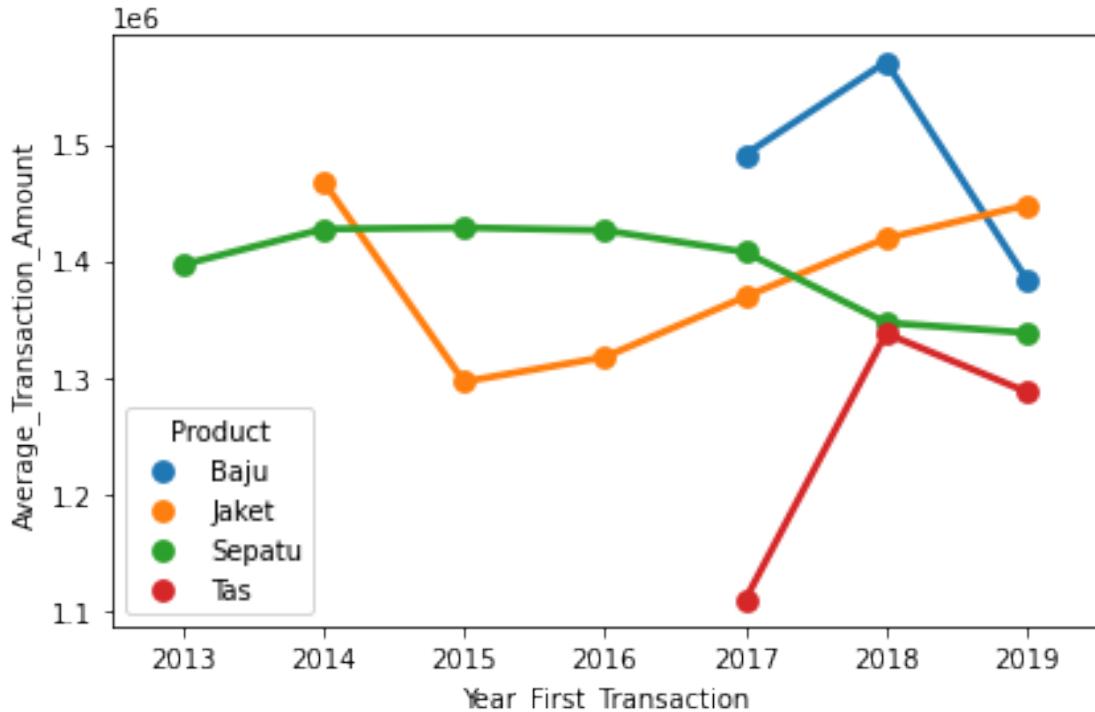
```

import pandas as pd
df = pd.read_csv('data_retail.csv', sep=';')
df['First_Transaction'] = pd.to_datetime(df['First_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Last_Transaction'] = pd.to_datetime(df['Last_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Year_First_Transaction'] = df['First_Transaction'].dt.year
df['Year_Last_Transaction'] = df['Last_Transaction'].dt.year

import matplotlib.pyplot as plt
import seaborn as sns

plt.clf()
sns.pointplot(data = df.groupby(['Product',
'Year_First_Transaction']).mean().reset_index(),
               x='Year_First_Transaction',
               y='Average_Transaction_Amount',
               hue='Product')
plt.tight_layout()
plt.show()

```



Proporsi churned customer untuk setiap produk

Dari sisi churned customer, khususnya untuk melihat seberapa besar proporsi churned customer untuk tiap-tiap produk dapat diketahui insight-nya melalui pie chart.

```

import pandas as pd
df = pd.read_csv('data_retail.csv', sep=';')
df['First_Transaction'] = pd.to_datetime(df['First_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Last_Transaction'] = pd.to_datetime(df['Last_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Year_First_Transaction'] = df['First_Transaction'].dt.year
df['Year_Last_Transaction'] = df['Last_Transaction'].dt.year
df.loc[df['Last_Transaction'] <= '2018-08-01', 'is_churn'] = True
df.loc[df['Last_Transaction'] > '2018-08-01', 'is_churn'] = False

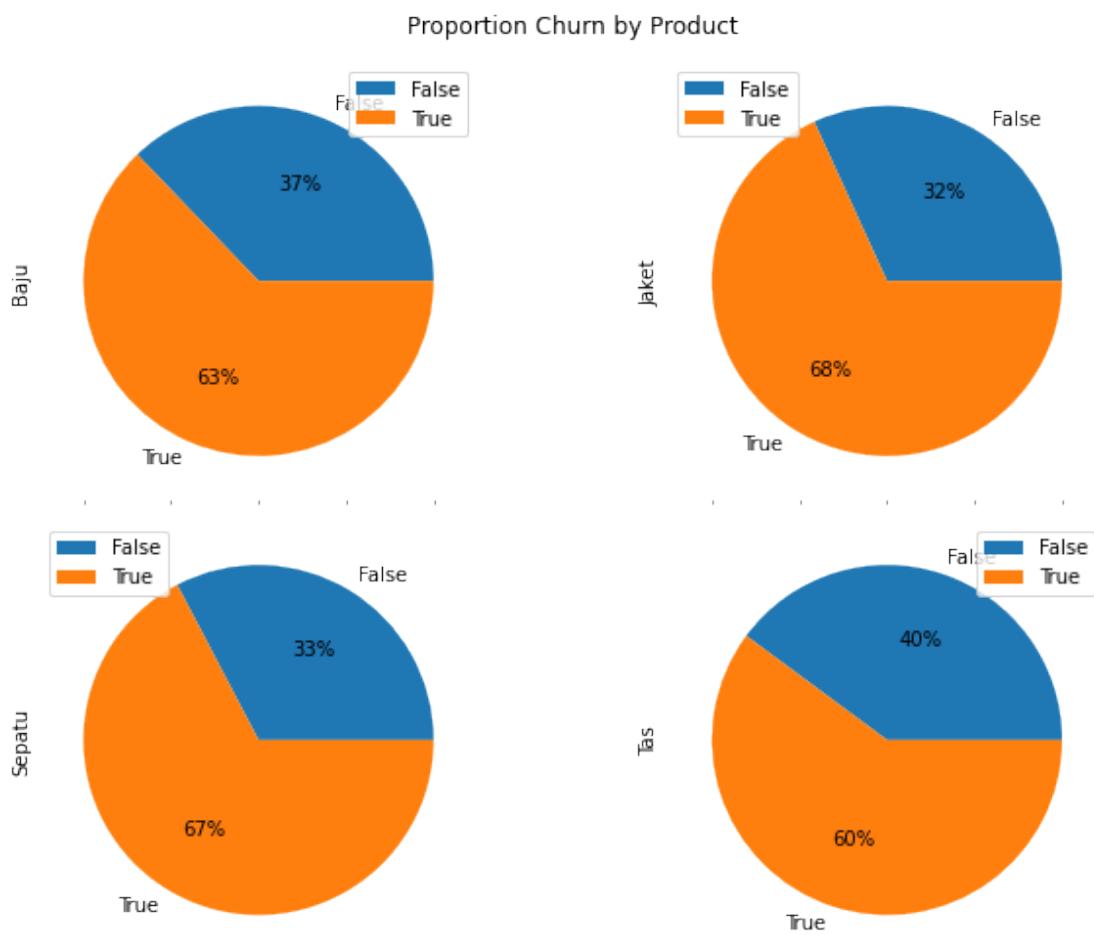
import matplotlib.pyplot as plt

plt.clf()
# Melakukan pivot data dengan pivot_table
df_piv = df.pivot_table(index='is_churn',
columns='Product',
values='Customer_ID',
aggfunc='count',
fill_value=0)
# Mendapatkan Proportion Churn by Product
plot_product =
df_piv.count().sort_values(ascending=False).head(5).index

```

```
# Plot pie chartnya
df_piv = df_piv.reindex(columns=plot_product)
df_piv.plot.pie(subplots=True,
                 figsize=(10, 7),
                 layout=(-1, 2),
                 autopct='%1.0f%%',
                 title='Proportion Churn by Product')
plt.tight_layout()
plt.show()
```

<Figure size 432x288 with 0 Axes>



Distribusi kategorisasi count transaction

Kategorisasi ini dilakukan dengan mengelompokkan jumlah transaksi

```
import pandas as pd
df = pd.read_csv('data_retail.csv', sep=';')
df['First_Transaction'] = pd.to_datetime(df['First_Transaction']/1000,
                                         unit='s', origin='1970-01-01')
df['Last Transaction'] = pd.to_datetime(df['Last Transaction']/1000,
```

```

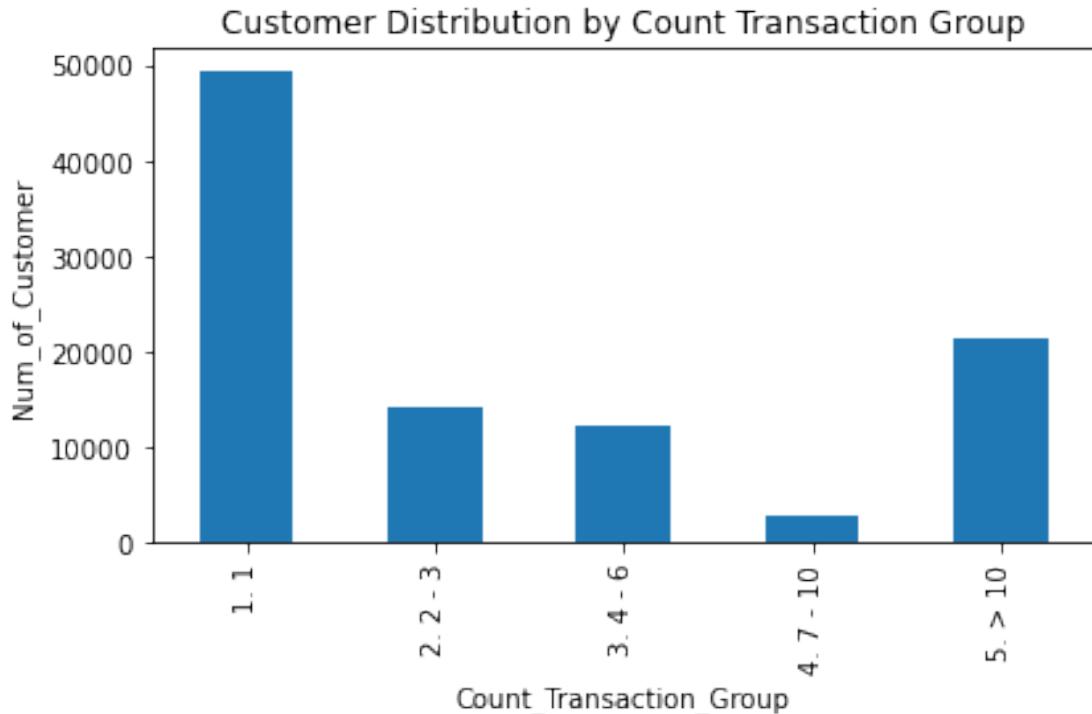
unit='s', origin='1970-01-01')
df['Year_First_Transaction'] = df['First_Transaction'].dt.year
df['Year_Last_Transaction'] = df['Last_Transaction'].dt.year
df.loc[df['Last_Transaction'] <= '2018-08-01', 'is_churn'] = True
df.loc[df['Last_Transaction'] > '2018-08-01', 'is_churn'] = False

import matplotlib.pyplot as plt

plt.clf()
# Kategorisasi jumlah transaksi
def func(row):
    if row['Count_Transaction'] == 1:
        val = '1. 1'
    elif (row['Count_Transaction'] > 1 and row['Count_Transaction'] <= 3):
        val = '2. 2 - 3'
    elif (row['Count_Transaction'] > 3 and row['Count_Transaction'] <= 6):
        val = '3. 4 - 6'
    elif (row['Count_Transaction'] > 6 and row['Count_Transaction'] <= 10):
        val = '4. 7 - 10'
    else:
        val = '5. > 10'
    return val
# Tambahkan kolom baru
df['Count_Transaction_Group'] = df.apply(func, axis=1)

df_year = df.groupby(['Count_Transaction_Group'])
['Customer_ID'].count()
df_year.plot(x='Count_Transaction_Group', y='Customer_ID', kind='bar',
title='Customer Distribution by Count Transaction Group')
plt.xlabel('Count_Transaction_Group')
plt.ylabel('Num_of_Customer')
plt.tight_layout()
plt.show()

```



Distribusi kategorisasi average transaction amount

Kategorisasi ini dilakukan dengan mengelompokkan rata-rata besar transaksi

```

import pandas as pd
df = pd.read_csv('data_retail.csv', sep=';')
df['First_Transaction'] = pd.to_datetime(df['First_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Last_Transaction'] = pd.to_datetime(df['Last_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Year_First_Transaction'] = df['First_Transaction'].dt.year
df['Year_Last_Transaction'] = df['Last_Transaction'].dt.year
df.loc[df['Last_Transaction'] <= '2018-08-01', 'is_churn'] = True
df.loc[df['Last_Transaction'] > '2018-08-01', 'is_churn'] = False

import matplotlib.pyplot as plt

plt.clf()
# Kategorisasi rata-rata besar transaksi
def f(row):
    if (row['Average_Transaction_Amount'] >= 100000 and
row['Average_Transaction_Amount'] <= 250000):
        val ='1. 100.000 - 250.000'
    elif (row['Average_Transaction_Amount'] > 250000 and
row['Average_Transaction_Amount'] <= 500000):
        val ='2. >250.000 - 500.000'
    elif (row['Average_Transaction_Amount'] > 500000 and
row['Average_Transaction_Amount'] <= 750000):
        val ='3. >500.000 - 750.000'
    else:
        val ='4. >750.000'
    return val

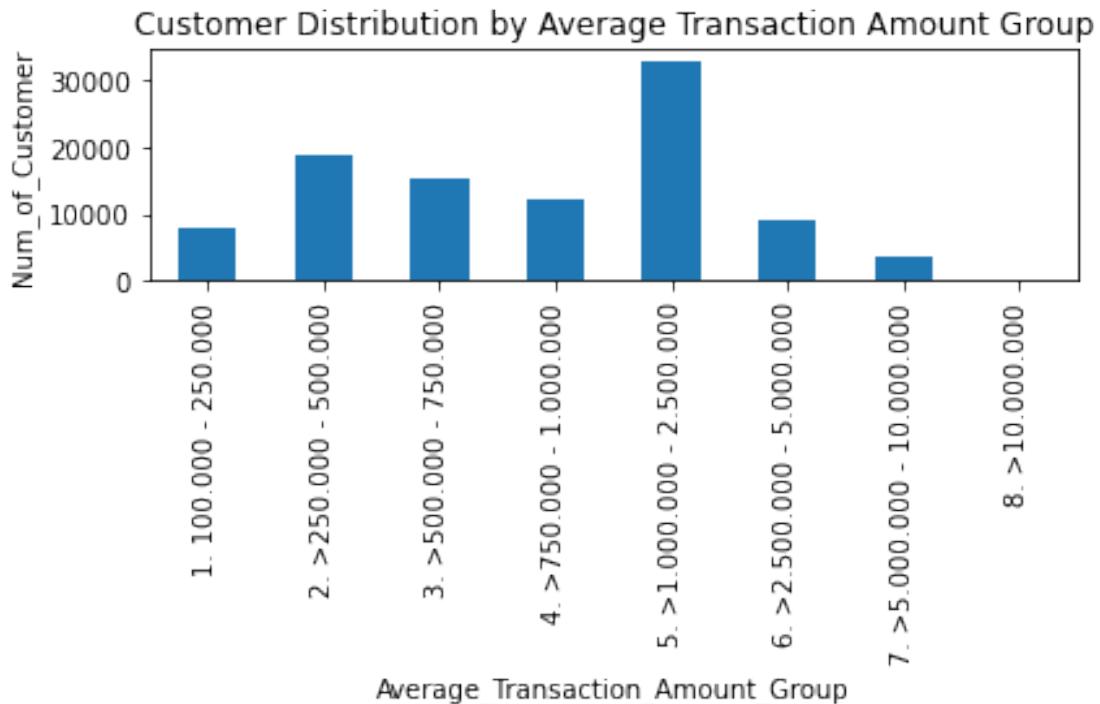
```

```

        val ='3. >500.000 - 750.000'
    elif (row['Average_Transaction_Amount'] > 750000 and
row['Average_Transaction_Amount'] <= 1000000):
        val ='4. >750.000 - 1.000.000'
    elif (row['Average_Transaction_Amount'] > 1000000 and
row['Average_Transaction_Amount'] <= 2500000):
        val ='5. >1.000.000 - 2.500.000'
    elif (row['Average_Transaction_Amount'] > 2500000 and
row['Average_Transaction_Amount'] <= 5000000):
        val ='6. >2.500.000 - 5.000.000'
    elif (row['Average_Transaction_Amount'] > 5000000 and
row['Average_Transaction_Amount'] <= 10000000):
        val ='7. >5.000.000 - 10.000.000'
    else:
        val ='8. >10.000.000'
    return val
# Tambahkan kolom baru
df['Average_Transaction_Amount_Group'] = df.apply(f, axis=1)

df_year = df.groupby(['Average_Transaction_Amount_Group'])
['Customer_ID'].count()
df_year.plot(x='Average_Transaction_Amount_Group', y='Customer_ID',
kind='bar', title='Customer Distribution by Average Transaction Amount
Group')
plt.xlabel('Average_Transaction_Amount_Group')
plt.ylabel('Num_of_Customer')
plt.tight_layout()
plt.show()

```



Feature Columns dan Target

Di bagian ini, selanjutnya akan menentukan feature columns dari dataset yang dimiliki, di sini dipilih kolom Average_Transaction_Amount, Count_Transaction, dan Year_Diff.

```
import pandas as pd
df = pd.read_csv('data_retail.csv', sep=';')
df['First_Transaction'] = pd.to_datetime(df['First_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Last_Transaction'] = pd.to_datetime(df['Last_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Year_First_Transaction'] = df['First_Transaction'].dt.year
df['Year_Last_Transaction'] = df['Last_Transaction'].dt.year
df.loc[df['Last_Transaction'] <= '2018-08-01', 'is_churn'] = True
df.loc[df['Last_Transaction'] > '2018-08-01', 'is_churn'] = False

# Feature column: Year_Diff
df['Year_Diff'] = df['Year_Last_Transaction'] -
df['Year_First_Transaction']

# Nama-nama feature columns
feature_columns = ['Average_Transaction_Amount', 'Count_Transaction',
'Year_Diff']

# Features variable
X = df[feature_columns]

# Target variable
y = df['is_churn']
```

Split X dan y ke dalam bagian training dan testing

Setelah variabel independent X dan variabel dependent y selesai dilakukan, maka pecahlah X dan y ke dalam bagian training dan testing. Bagian testing 25% dari jumlah entri data.

```
import pandas as pd
df = pd.read_csv('data_retail.csv', sep=';')
df['First_Transaction'] = pd.to_datetime(df['First_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Last_Transaction'] = pd.to_datetime(df['Last_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Year_First_Transaction'] = df['First_Transaction'].dt.year
df['Year_Last_Transaction'] = df['Last_Transaction'].dt.year
df.loc[df['Last_Transaction'] <= '2018-08-01', 'is_churn'] = True
df.loc[df['Last_Transaction'] > '2018-08-01', 'is_churn'] = False

df['Year_Diff']=df['Year_Last_Transaction']-
df['Year_First_Transaction']
feature_columns = ['Average_Transaction_Amount', 'Count_Transaction',
'Year_Diff']
```

```

X = df[feature_columns]
y = df['is_churn']

from sklearn.model_selection import train_test_split

X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.25, random_state=0)

```

Train, predict dan evaluate

Langkah selanjutnya akan membuat model menggunakan Logistic Regression, inisialisasilah model, fit, dan kemudian evaluasi model dengan menggunakan confusion matrix.

```

import pandas as pd
df = pd.read_csv('data_retail.csv', sep=';')
df['First_Transaction'] = pd.to_datetime(df['First_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Last_Transaction'] = pd.to_datetime(df['Last_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Year_First_Transaction'] = df['First_Transaction'].dt.year
df['Year_Last_Transaction'] = df['Last_Transaction'].dt.year
df.loc[df['Last_Transaction'] <= '2018-08-01', 'is_churn'] = True
df.loc[df['Last_Transaction'] > '2018-08-01', 'is_churn'] = False

df['Year_Diff']=df['Year_Last_Transaction']-
df['Year_First_Transaction']
feature_columns = ['Average_Transaction_Amount', 'Count_Transaction',
'Year_Diff']

X = df[feature_columns]
y = df['is_churn']

from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.25, random_state=0)

from sklearn.linear_model import LogisticRegression
from sklearn.metrics import confusion_matrix

# Inisiasi model logreg
logreg = LogisticRegression()

# fit the model with data
logreg.fit(X_train, y_train)

# Predict model
y_pred = logreg.predict(X_test)

```

```
# Evaluasi model menggunakan confusion matrix
cnf_matrix = confusion_matrix(y_test, y_pred)
print('Confusion Matrix:\n', cnf_matrix)
```

Visualisasi Confusion Matrix

Confusion matrix yang telah dihitung sebelumnya dapat divisualisasikan dengan menggunakan heatmap dari seaborn. Untuk itu tampilkanlah visualisasi dari confusion matrix ini.

```
import pandas as pd
df = pd.read_csv('data_retail.csv', sep=';')
df['First_Transaction'] = pd.to_datetime(df['First_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Last_Transaction'] = pd.to_datetime(df['Last_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Year_First_Transaction'] = df['First_Transaction'].dt.year
df['Year_Last_Transaction'] = df['Last_Transaction'].dt.year
df.loc[df['Last_Transaction'] <= '2018-08-01', 'is_churn'] = True
df.loc[df['Last_Transaction'] > '2018-08-01', 'is_churn'] = False

df['Year_Diff']=df['Year_Last_Transaction']-
df['Year_First_Transaction']
feature_columns = ['Average_Transaction_Amount', 'Count_Transaction',
'Year_Diff']

X = df[feature_columns]
y = df['is_churn']

from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.25, random_state=0)

from sklearn.linear_model import LogisticRegression
from sklearn.metrics import confusion_matrix

logreg = LogisticRegression()
logreg.fit(X_train, y_train)
y_pred = logreg.predict(X_test)
cnf_matrix = confusion_matrix(y_test, y_pred)

# import required modules
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

plt.clf()
# name of classes
class_names = [0, 1]
```

```

fig, ax = plt.subplots()

tick_marks = np.arange(len(class_names))
plt.xticks(tick_marks, class_names)
plt.yticks(tick_marks, class_names)

# create heatmap
sns.heatmap(pd.DataFrame(cnf_matrix), annot=True, cmap='YlGnBu',
fmt='g')
ax.xaxis.set_label_position('top')
plt.title('Confusion matrix', y=1.1)
plt.ylabel('Actual')
plt.xlabel('Predicted')
plt.tight_layout()
plt.show()

```

Accuracy, Precision, dan Recall

Kemudian, menghitung nilai accuracy, precision dan recall berdasarkan nilai target sesungguhnya dan nilai target hasil prediksi.

```

import pandas as pd
df = pd.read_csv('data_retail.csv', sep=';')
df['First_Transaction'] = pd.to_datetime(df['First_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Last_Transaction'] = pd.to_datetime(df['Last_Transaction']/1000,
unit='s', origin='1970-01-01')
df['Year_First_Transaction'] = df['First_Transaction'].dt.year
df['Year_Last_Transaction'] = df['Last_Transaction'].dt.year
df.loc[df['Last_Transaction'] <= '2018-08-01', 'is_churn'] = True
df.loc[df['Last_Transaction'] > '2018-08-01', 'is_churn'] = False

df['Year_Diff']=df['Year_Last_Transaction']-
df['Year_First_Transaction']
feature_columns = ['Average_Transaction_Amount', 'Count_Transaction',
'Year_Diff']

X = df[feature_columns]
y = df['is_churn']

from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y,
test_size=0.25, random_state=0)

from sklearn.linear_model import LogisticRegression
from sklearn.metrics import confusion_matrix

logreg = LogisticRegression()
logreg.fit(X_train, y_train)
y_pred = logreg.predict(X_test)

```

```
cnf_matrix = confusion_matrix(y_test, y_pred)

from sklearn.metrics import accuracy_score, precision_score,
recall_score

#Menghitung Accuracy, Precision, dan Recall
print('Accuracy :', accuracy_score(y_test, y_pred))
print('Precision:', precision_score(y_test, y_pred, average='micro'))
print('Recall    :', recall_score(y_test, y_pred, average='micro'))
```