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In [ ]: #Financial Data Analysis
In [ ]: # Import Libraries
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
In [ ]: # Loading CSV file
In [2]: df = pd.read_csv("Acquisition_Credit_Data.csv")
df.head()
Out[2]: account_id acquisition_channel open_date month credit_line balance revenue charge_off_amount charge_off_flag days_past_due acquisition_cost
0 A100001 Online 2023-07-07 2023-07-01 4900 723.83 19.09 0.0 0 0 130
1 A100002 Mail 2024-04-08 2024-04-01 4800 916.81 37.28 0.0 0 0 150
2 A100003 Online 2024-01-28 2024-01-01 8100 872.38 22.84 0.0 0 0 180
3 A100004 Partner 2023-05-19 2023-05-01 2300 600.00 5.28 0.0 0 0 130
4 A100005 Partner 2023-10-26 2023-10-01 5700 407.63 7.75 0.0 0 0 150
In [ ]: # Data Inspection
In [3]: print("Rows, Columns:", df.shape)
# Column types
df.info()
Rows, Columns: (600, 11)
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 600 entries, 0 to 599
Data columns (total 11 columns):
 #   Column          Non-Null Count  Dtype  
--- 
 0   account_id      600 non-null    object  
 1   acquisition_channel 600 non-null    object  
 2   open_date        600 non-null    object  
 3   month            600 non-null    object  
 4   credit_line      600 non-null    int64  
 5   balance           600 non-null    float64 
 6   revenue           600 non-null    float64 
 7   charge_off_amount 600 non-null    float64 
 8   charge_off_flag  600 non-null    int64  
 9   days_past_due    600 non-null    int64  
 10  acquisition_cost 600 non-null    int64  
dtypes: float64(3), int64(4), object(4)
memory usage: 51.7+ KB
In [ ]: # Summary Statistics
In [4]: df.describe(include='all')
Out[4]: account_id acquisition_channel open_date month credit_line balance revenue charge_off_amount charge_off_flag days_past_due acquisition_cost
count 600 600 600 600 600.000000 600.000000 600.000000 600.000000 600.000000 600.000000 600.000000
unique 600 3 419 24 NaN NaN NaN NaN NaN NaN NaN NaN NaN
top A100391 Online 2024-08-21 2024-06-01 NaN NaN NaN NaN NaN NaN NaN NaN
freq 1 310 5 34 NaN NaN NaN NaN NaN NaN NaN NaN NaN
mean NaN NaN NaN 5519.833333 2049.499483 33.939333 4.130017 0.003333 4.450000 128.683333
std NaN NaN NaN 2570.168525 916.801090 17.462905 73.095696 0.057687 13.638457 22.522844
min NaN NaN NaN 1000.000000 363.290000 -2.080000 0.000000 0.000000 0.000000 100.000000
25% NaN NaN NaN 3300.000000 1392.440000 21.595000 0.000000 0.000000 0.000000 120.000000
50% NaN NaN NaN 5500.000000 1899.605000 30.875000 0.000000 0.000000 0.000000 120.000000
75% NaN NaN NaN 7800.000000 2487.157500 43.130000 0.000000 0.000000 0.000000 150.000000
max NaN NaN NaN 10000.000000 5342.380000 110.750000 1503.970000 1.000000 90.000000 180.000000
In [ ]: # Checking for missing values
In [5]: df.isnull().sum()
Out[5]: account_id 0
acquisition_channel 0
open_date 0
month 0
credit_line 0
balance 0
revenue 0
charge_off_amount 0
charge_off_flag 0
days_past_due 0
acquisition_cost 0
dtype: int64
In [ ]: # Date Conversion
In [6]: df['open_date'] = pd.to_datetime(df['open_date'], errors='coerce')
df['month'] = pd.to_datetime(df['month'], format="%Y-%m")
In [ ]: # Revenue & Credit Utilization Metrics
In [7]: # Credit utilization rate
df['utilization_rate'] = df['balance'] / df['credit_line']

df[['account_id', 'credit_line', 'balance', 'utilization_rate']].head()
Out[7]: account_id credit_line balance utilization_rate
0 A100001 4900 723.83 0.147720
1 A100002 4800 916.81 0.191002
2 A100003 8100 872.38 0.107701
3 A100004 2300 600.00 0.260870
4 A100005 5700 407.63 0.071514
In [ ]: # Aggregation Example (Monthly Revenue)
In [8]: monthly_rev = (
    df.groupby('month')[['revenue']]
    .sum()
    .reset_index()
    .sort_values('month')
)
monthly_rev
Out[8]: month revenue
0 2023-01-01 651.78
1 2023-02-01 1143.94
2 2023-03-01 1037.06
3 2023-04-01 663.04
4 2023-05-01 755.91
5 2023-06-01 853.90
6 2023-07-01 805.06
7 2023-08-01 1057.89
8 2023-09-01 684.92
9 2023-10-01 963.98
10 2023-11-01 684.03
11 2023-12-01 693.17
12 2024-01-01 701.18
13 2024-02-01 861.66
14 2024-03-01 798.22
15 2024-04-01 1155.81
16 2024-05-01 595.42
17 2024-06-01 1173.86
18 2024-07-01 411.11
19 2024-08-01 878.33
20 2024-09-01 932.19
21 2024-10-01 1055.43
22 2024-11-01 1095.11
23 2024-12-01 710.60
In [ ]: # Channel Acquisition Summary
In [9]: channel_summary = df.groupby('acquisition_channel').agg({
    'account_id': 'count',
    'revenue': 'mean',
    'credit_line': 'mean',
    'balance': 'mean',
    'charge_off_flag': 'mean'
}).rename(columns={'account_id': 'num_accounts',
                  'charge_off_flag': 'charge_off_rate'})

channel_summary
Out[9]: num_accounts revenue credit_line balance charge_off_rate
acquisition_channel
Mail 169 35.392071 5345.562130 2042.106391 0.005917
Online 310 33.752774 5601.935484 2054.145935 0.000000
Partner 121 32.388264 5552.892562 2047.921240 0.008264
In [ ]: # Monthly Revenue Visualization
In [10]: plt.figure(figsize=(8,4))
plt.plot(monthly_rev['month'], monthly_rev['revenue'])
plt.xlabel("Month")
plt.ylabel("Total Revenue")
plt.title("Monthly Revenue Trend")
plt.xticks(rotation=45)
plt.show()

In [ ]: # Charge-Off Rate by Channel
In [11]: chargeoff = df.groupby('acquisition_channel')['charge_off_flag'].mean()
chargeoff.plot(kind='bar')
plt.xlabel("Charge-Off Rate by Acquisition Channel")
plt.ylabel("Charge-Off Rate")
plt.title("Charge-Off Rate by Acquisition Channel")
plt.show()

In [ ]:
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