

# FINANCIAL RISK ANALYSIS WITH PYTHON

## – BARCLAYS

### PROJECT OVERVIEW

This project focuses on financial transaction analysis and risk identification using real-world banking transaction data from Barclays. The objective is to understand customer transaction behavior, identify financially risky accounts, and derive data-driven insights that can support risk monitoring and customer management decisions.

The dataset contains detailed information on customers, accounts, transaction types, dates, balances, inflows, outflows, and risk indicators. Using Python-based data analysis and statistical techniques, the project performs exploratory data analysis (EDA), behavioral segmentation, financial risk detection, and hypothesis testing.

### OBJECTIVES

The main objectives of this project are to analyze transactional banking data and extract meaningful insights related to customer behavior and financial risk. Specifically, the project aims to:

- Analyze customer transaction patterns across accounts, products, and time periods
- Identify dormant and low-activity accounts based on transaction gaps and frequency
- Detect accounts with financial risk, including overdrafts, negative balances, and high balance volatility
- Track credit and debit trends over time to understand cash flow behavior
- Detect anomalous transactions using statistical techniques such as IQR and Z-score methods
- Segment customers based on average balance and transaction volume
- Test hypotheses to validate whether transaction volume impacts account balance levels
- Provide data-backed recommendations for customer engagement and risk monitoring

### DATA CLEANING & VALIDATION

Before performing analysis, the dataset was thoroughly cleaned and validated to ensure accuracy and reliability of insights.

#### 1. Special Character & Numeric Validation

- Special characters in numeric columns: **0 found**
- Currency symbols detected: **0**
- Decimal separator verified: **(point) – correct**
- All **800 rows** successfully converted to **float64**

#### 2. Negative Values Check

- Negative transactions identified: **21**
  - Interpreted as valid payments/refunds
- Negative balances identified: **14**
  - Treated as valid overdraft cases

#### 3. Null Value Handling

- Null values across all columns: **0**
- No rows dropped or imputed

#### 4. Date Validation & Transformation

- Original format converted: DD-MM-YYYY → datetime64[ns]
- Conversion success rate: **800 / 800 (100%)**
- Invalid or corrupted dates: **0**
- Extracted components:
  - Year
  - Month
  - Day
  - Day of Week
  - Week of Year

#### 5. Date Range Confirmation

- Start Date: 2023-01-02
- End Date: 2024-06-22
- Total Duration: 537 days

#### 6. Standardization of Categorical Columns

- **Account Type:** 4 categories
  - Savings, Credit, Loan, Current
- **Transaction Type:** 4 categories
  - Deposit, Payment, Withdrawal, Transfer
- **Product:** 5 categories
  - Credit Card, Mutual Fund, Personal Loan, Home Loan, Savings Account
- Extra spaces removed: **0**
- Inconsistent labels found: **0**

The dataset is **fully clean, standardized, and analysis ready.**

## EXPLORATORY DATA ANALYSIS (EDA)

The EDA reveals clear patterns:

- Customer activity is decreasing over time.
- Majority of accounts are low-activity but moderate balance holders.
- Significant risk indicators exist (overdrafts, high CV accounts, anomaly transactions).
- No statistical link exists between frequency of transactions and account balances.
- Dormancy is widespread → immediate operational attention required.

## INSIGHTS & FINDINGS

This section summarizes the key analytical findings derived from the cleaned and processed Barclays financial dataset (800 transactions across 15 variables). Insights are organized by analytical tasks for clarity.

### 1. Data Quality Insights

- **No missing values** in the dataset, ensuring reliability of analysis (0 nulls across all columns).
- **No special characters** or corrupted entries found in numeric fields such as TransactionAmount and AccountBalance.
- All dates successfully converted, enabling accurate time-series analysis.
- **14 accounts** were found with **negative balances**, representing overdrafts—important early indicators of financial stress.

- Numeric ranges validated:
  - TransactionAmount ranges from **-59,669** to **166,083**
  - AccountBalance ranges from **-30,766** to **184,008**

#### **Overall Finding:**

The data is clean, structured, and trustworthy for financial risk analytics.

## **2. Transaction Behaviour Insights**

### **Monthly Trends**

- Highest monthly transaction activity occurred in **June 2023 (41 transactions, ₹18.9 lakh)**.
- Year-wise totals:
  - **2023:** 209 transactions (₹1.14 crore)
  - **2024:** 113 transactions (₹55.0 lakh)

#### **Insight:**

Customer engagement decreased by ~46% from 2023 to 2024, indicating a possible decline in activity or user base.

## **3. Net Inflow Performance**

### **Top Net Inflow Accounts**

- Highest positive net inflow:  
**ACC76549 – ₹1,78,514**
- Other strong performers include:  
**ACC39544 (₹1,42,078), ACC64022 (₹1,20,503), ACC95164 (₹1,16,682)**

### **Bottom/Zero Net Inflow Accounts**

- Nearly **70% of accounts** show **no positive inflow** ( $\text{Total\_Inflow} = \text{Total\_Outflow} = 0$ ).
- Many accounts appear **inactive or dormant**.

#### **Insight:**

A major portion of customers are not actively adding funds, suggesting low engagement or dormant status.

## **4. Dormancy & Activity Analysis**

- Accounts with **> 60-day inactivity** flagged as dormant:
  - Examples: **ACC10117 (298 days), ACC11062 (235 days), ACC15228 (329 days)**.
- Activity segmentation:
  - **100% of the sample shown** falls under **Low Activity** (<10 transactions).

#### **Insight:**

Dormancy risk is very high; multiple customers show long gaps between transactions.

## **5. Segmentation Findings (Balance & Volume)**

- Majority of customers fall under:

- **Medium Balance (₹50k–₹1 lakh)**
- **Low Volume (<10 transactions)**
- Example segments:
  - **High Balance, Low Volume:** CUST1555 (₹1.03 lakh), CUST1749 (₹1.25 lakh)
  - **Low Balance, Low Volume:** CUST1467 (₹44,806), CUST1547 (₹42,838)

**Insight:**

Even high-balance customers remain low-engagement, indicating missed cross-sell or up-sell opportunities.

## 6. Financial Risk Insights

### Overdraft & Negative Balance Accounts (14 identified)

Examples:

- ACC42467: -₹30,766
- ACC19156: -₹9,649
- ACC72197: -₹14,934

These accounts require monitoring due to potential credit risk.

### Volatility Findings (Coefficient of Variation – CV)

- High volatility accounts ( $CV > 0.7$ ):
  - ACC11285 ( $CV = 1.12$ )
  - ACC12334 ( $CV = 0.74$ )
  - ACC19156 ( $CV = 0.95$ )

**Insight:**

These accounts show unstable balances → risk of sudden negative turns.

## Anomaly Detection

- **7 IQR anomalies & 4 Z-score anomalies** detected.
- Very high deviation examples:
  - Debit of **-₹59,669**
  - Transfer of **₹1,66,083**
  - Credit of **₹1,42,081**

**Insight:**

Irregularly large movements indicate possible unusual or risky transactions.

## 7. Hypothesis Testing Insights

### Test 1: Do high-volume accounts have higher balances?

- p-value = **0.869**
- **Result:** Fail to reject H<sub>0</sub>  
→ No evidence that transaction volume increases balance.

## Test 2: Segment Comparison (Low vs Medium vs High balance groups)

- p-values: **0.56–0.57**
- **Result:** No significant difference found.

### Insight:

Balance levels and transaction volume are **not statistically related** in this dataset.

## FINAL SUMMARY OF FINDINGS

- Customer transaction activity is **decreasing significantly** over time.
- Most accounts display **very low transaction volume**, indicating disengagement.
- **Dormancy risk is high** (multiple accounts inactive >200 days).
- **Overdrafts and high-volatility accounts** highlight financial risk pockets.
- **Net inflow performance** is heavily skewed — only a few accounts contribute positively.
- High-value customers are often **low-engagement**, showing cross-sell opportunity.
- There is **no statistical relationship** between transaction volume and balance.

## FINAL RECOMMENDATIONS

- Proactively engage low-activity accounts with reminders, offers, and targeted nudges.
- Introduce periodic alerts for accounts showing long inactivity (>200 days) to reduce dormancy risk.
- Monitor overdraft and high-volatility accounts with stricter controls and automated risk flags.
- Strengthen credit controls for customers frequently touching negative balances.
- Prioritize retention campaigns for high-balance but low-engagement customers (cross-sell opportunity).
- Improve transaction monitoring dashboards to track monthly inflow, outflow, and anomalies.
- Encourage digital payments and auto-debit setups to boost transaction frequency.
- Review the customer journey for inactive users to identify friction points and service gaps.

## Video Link:

[https://drive.google.com/file/d/1ItXhXRkqWyfsF2B0iFwEsTPG\\_s-HqPMF/view?usp=sharing](https://drive.google.com/file/d/1ItXhXRkqWyfsF2B0iFwEsTPG_s-HqPMF/view?usp=sharing)