

## **Customer Behavioural Spending Patterns for Early Financial Stress Detection (SQL-based Analysis)**

### **Business question:**

- Can we identify early signs of customer financial stress by analyzing transaction behavior using SQL?

### **Use cases:**

- Banks/Fintech corporations need to identify financial stress of the customers **early** and not after the payment default.
- Internal Collections and risk teams can identify customers and **prioritize** efforts based on their behaviour patterns.

### **Dataset used:**

- Personal Financial transactions data over 365 days (CSV file)
- Fields are as below,
  - o Transaction\_date
  - o Weekday
  - o Amount
  - o Time\_of\_day
  - o Merchant\_type
  - o Month

### **Assumption:**

- Dataset comprises anonymized customer spending behavior. No real customer data is used.

**Analytics:** Four behavioral KPIs were used to identify early financial stress signals.

- **Spend Volatility:** Higher fluctuations would indicate unstable cash flow.
  - o SQL query used:
  - o select
  - o avg(amount) as avg\_spend,
  - o stddev(amount) as spend\_volatility
  - o from finance\_transactions;
  - o **Insight** – Customers with high spending volatility may have unstable cash flows, even if their average spending appears normal.
- **Spend by Day (Behavioural Timing):**
  - o SQL query used:
  - o select weekday,
  - o round(sum(amount),2) as total\_spend
  - o from finance\_transactions
  - o group by weekday
  - o order by total\_spend desc;
  - o **Insight** – Would provide the day where customers spend the most, implying impulsive spending or financial pressure.

- **Spend by Category (Merchant split up spend):**
  - o SQL query used:
  - o select merchant\_type,
  - o round(sum(amount),2) as total\_spend
  - o from finance\_transactions
  - o group by merchant\_type
  - o order by total\_spend desc;
  - o **Insight** – Would provide the area where customer spends the most, indicating their priorities and obligations.
- **Early Warning Signal (Daily spend):**
  - o SQL query used:
  - o select transaction\_date,
  - o weekday,
  - o sum(amount) as daily\_spend
  - o from finance\_transactions
  - o group by transaction\_date, weekday
  - o order by daily\_spend desc;
  - o **Insight** – Spikes in daily spending may reflect emergency expenses or poor cash flow planning.

#### **Findings:**

- Spending pattern shows significant day level volatility even with stable averages.
- Certain weekdays consistently show higher customer spend.
- Services and Subscriptions (Merchant type) dominate peak spend days; retail is where the customers spend the least indicating the trend towards online platforms.

#### **Business Implications:**

- Highly volatile customers can be flagged early preventing default risks.
- Peak spend days can help internal teams to increase outreach and communication effectively improving efficiency.
- Classification based on categorical spending helps identify risks and priorities associated with respective teams ensuring holistic approach.

#### **Conclusion:**

- Transaction-level data provides strong early indicators of financial stress.
- Spending volatility and timing-based patterns can help risk and collections teams take proactive action even in the absence of traditional credit data.