

Transactional Fraud Detection Report

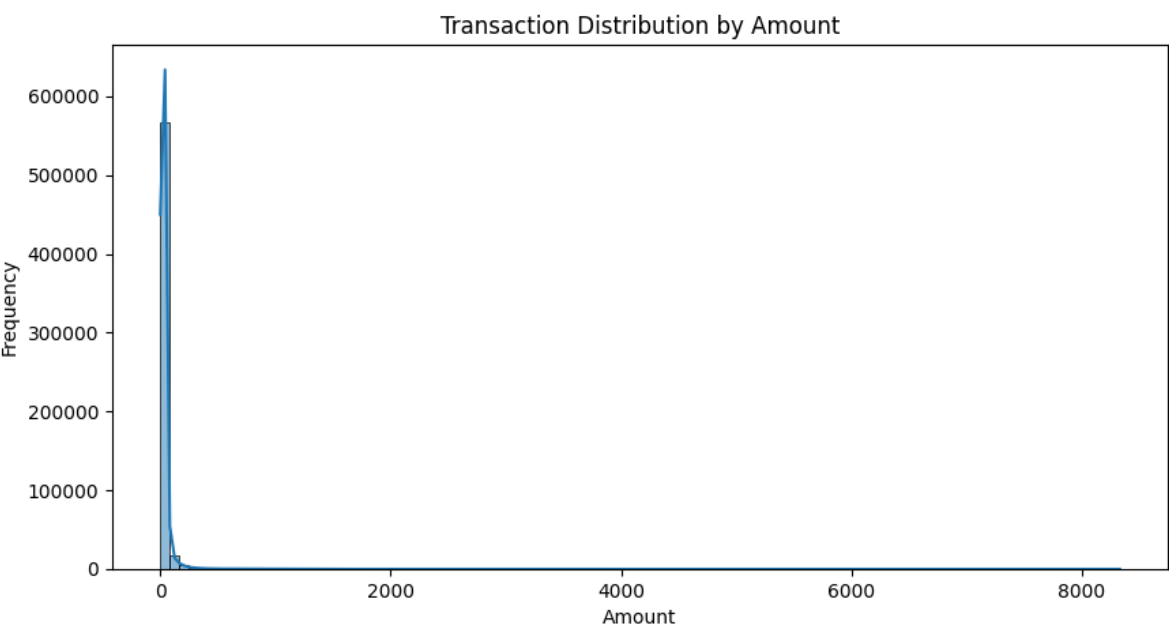
Project Objectives

1. Data Handling & Processing: Work with large transaction datasets, clean, and preprocess the data.
 - Read and write data using Pandas
 - Handle missing values and data inconsistencies
 - Convert timestamps into useful formats (hour, day of the week)
 - Extract high-risk locations
2. Fraud Analysis: Identify high-risk transactions based on anomalies in spending patterns.
 - Compute fraud rate (%)
 - Identify high-risk transactions (e.g., over \$10,000 or odd hours)
 - Find the most fraud-prone regions
3. Statistical Insights: Compute fraud percentages, transaction distributions, and high-risk regions.
4. Data Visualization: Create fraud detection dashboards using Matplotlib & Seaborn.
 - Transaction Distribution by Amount
 - Fraud vs Non-Fraud Transactions
 - Hourly Fraud Trend
 - Fraud-Prone Locations
5. Anomaly Detection Using Rules: Define fraud detection rules based on transaction behaviour.
6. Interactive Dashboard: Display fraud trends and allow filtering by time, amount, and location.

Console Output

```
(.venv) PS D:\python_lab_work\transactional_fraud_detection_system> python .\script.py
Total Transactions: 594643
Total Frauds: 7200
Fraud Rate (%): 1.21
High-Value Frauds (>10K): 0
Odd-Hour Frauds: 2145
Top Fraud-Prone Merchants: {'M480139044': 1634, 'M980657600': 1472, 'M732195782': 518, 'M1198415165': 358, 'M3697346': 290, 'M855959430': 280, 'M1888755466': 228, 'M1873032707': 216, 'M2122776122': 200, 'M1741626453': 196}
(.venv) PS D:\python_lab_work\transactional_fraud_detection_system> █
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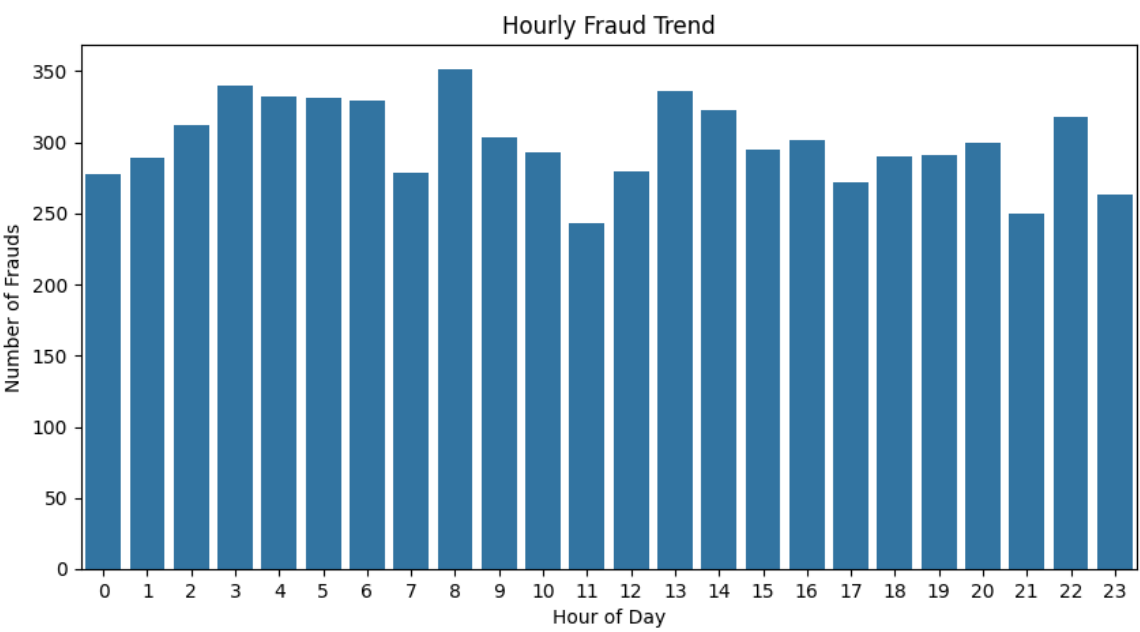
Transaction Distribution by Amount



Fraud vs Non-Fraud Transactions



Hourly Fraud Trend



Top Fraud-Prone Merchants

