

Financial Transaction Monitoring



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Problem Statement

Financial Transaction Monitoring





- Clean and flag potential issues, store processed stream in Delta Lake.
- Maintain infra-as-code (IaC) using Azure DevOps
 ARM/Bicep templates for Event Hub & Databricks setup.



ML Problem

- Use Isolation Forest or Autoencoder to detect fraudulent transactions.
- Alert integration using Azure Logic Apps + Power BI dashboards.

Procedure

3 4 5 **Load Dataset From Introduction: Step 1 & 2: Data Step 3: Ingest into** Step 4: Clean & Kaggle **End-to-End Fraud** Transform - Silver **Acquisition & Cloud Databricks – Bronze** (https://www.kaggle. **Detection Platform** Storage **Table Table** com/code/benroshan /transaction-fraud-d etection/input) 8 9 6 10 **Step 9: Visualize** Step 7: **Step 5: Feature Step 8: Real-time** Step 6: Build & **Fraud Trends in Orchestration** with **Engineering – Gold** Alerts via Azure Register ML Model **Apache Airflow Power BI Table Logic Apps**

Step 1 & 2: Data Acquisition & Cloud Storage

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Source Raw Transaction Data:
Obtain relevant financial transaction
data from Kaggle for initial analysis
and development.

Secure Cloud Data Lake: Upload dataset to Azure Storage Blob, establishing a scalable, accessible data source.

Foundation for Analytics: Centralized cloud storage provides a robust base for all subsequent data processing steps.

Step 3: Ingest into Databricks – Bronze Table

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Raw Data Ingestion Layer: Read raw

CSV from Azure Blob into
Databricks, creating the Bronze layer

Delta table.

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Immutable Data Capture: Bronze table acts as an unchangeable, versioned record of all raw input data.

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Foundation for Delta Lake: This is the crucial first step in building a reliable and scalable Delta Lake architecture.

Step 4: Clean & Transform – Silver Table



Enhance Data Quality: Clean missing values, cast data types, and standardize formats for better analysis.

Prepare for Analysis: Derive new features like 'hour from timestamp' and 'amount band' for deeper insights.

Governed Data Layer: The Silver layer provides a refined, trustworthy dataset for downstream applications.

Step 5: Feature Engineering – Gold Table



ML-Ready Data Creation: Enrich the Silver data with advanced features for machine learning models.



Business Insight Layer: Add
'is_suspicious' flags and
'fraud_risk_score' for actionable
business intelligence.



Optimized for Consumption: The Gold table is optimized for direct use by ML models and business intelligence dashboards.



Step 6: Build & Register ML Model

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Detect Anomalies with ML:
Train an Isolation Forest model
on the Silver data to identify
fraudulent transactions.

MLOps with MLflow: Log model experiments and versions using MLflow for reproducibility and management. Enhance Fraud Detection:
Machine learning adds an
intelligent layer for more
accurate and proactive fraud
identification.

Step 7: Orchestration with Apache Airflow



Automate Data Pipelines: Orchestrate all steps (Bronze to Gold, ML) as a DAG using Apache Airflow.

Reliable Workflow Management: Use
DatabricksSubmitRunOperator to trigger
notebooks, ensuring smooth, scheduled execution.

Continuous Processing: Airflow ensures the fraud detection pipeline runs consistently, keeping data fresh.

Step 8: Real-time Alerts via Azure Logic Apps

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Instant Fraud Notification:
Trigger alerts (email/Teams) from
Azure Logic Apps upon fraud
detection.

Configurable Alert Channels:
Flexibly send notifications to
various platforms based on defined
rules.

Accelerate Response Time:
Automated alerts enable rapid
response to minimize potential
fraud impact.

Step 9: Visualize Fraud Trends in Power Bl



Interactive Fraud Dashboard: Connect Power BI to the Gold Delta table for comprehensive fraud monitoring.

Key Metric Visualization: Visualize fraud trends, suspicious amounts, and risk by merchant and region.

Data-Driven Decisions: Empower analysts with actionable insights for strategic fraud prevention.

Thank You