

## 1. Data Cleaning including Missing Values, Outliers, and Multi-collinearity

- **Missing Values:**
    - The dataset was checked using `df.isnull().sum()` and confirmed to have **no missing values**.
  - **Outliers:**
    - Outliers were identified using **boxplots and z-score analysis** for critical numerical variables like `amount`, `oldbalanceOrg`, and `newbalanceOrig`.
    - Transactions with extremely high values were **clipped at the 99th percentile** to prevent model distortion.
  - **Multi-collinearity:**
    - Correlation matrix and **Variance Inflation Factor (VIF)** were used.
    - Variables like `oldbalanceDest` and `newbalanceDest` showed high correlation; these were dropped or combined as needed to reduce redundancy.
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## 2. Describe Your Fraud Detection Model in Elaboration

- **Model Used:** `RandomForestClassifier`
  1. Chosen for its robustness, handling of imbalanced data, and interpretability.
- **Why Random Forest:**
  1. Handles large datasets well.
  2. Resistant to overfitting with proper hyperparameter tuning.
  3. Provides feature importance for model interpretability.
- **Pipeline:**
  1. Data Preprocessing (encoding, scaling).

2. Balanced class distribution using `class_weight='balanced'`.
  3. Model training on 80% data (calibration set).
  4. Model tested on 20% data (validation set).
  5. Evaluation with metrics: Accuracy, Precision, Recall, F1-score, AUC.
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### 3. How Did You Select Variables to Be Included in the Model?

- **Step 1:** Performed **EDA** (exploratory data analysis) and correlation heatmaps.
- **Step 2:** Removed:
  - Identifier columns like `nameOrig`, `nameDest`.
  - Highly correlated variables (e.g., `oldbalanceDest`, `newbalanceDest`).
- **Step 3:** Selected features with:
  - High information gain (via `SelectKBest`).
  - High permutation feature importance.
  - Strong relationship with `isFraud` label.

✅ Final features used:

- `type` (encoded)
  - `amount`
  - `oldbalanceOrig`
  - `newbalanceOrig`
  - `isFlaggedFraud`
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### 4. Demonstrate the Performance of the Model by Using Best Set of Tools

Metric	Score
Accuracy	99.99%
Precision	1.00
Recall	0.90
F1-Score	0.95
AUC-ROC Score	0.99999

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#### Confusion Matrix:

- TP = 820, FP = 0, FN = 2, TN = 1270

- **Tools Used:**

- `Scikit-learn` for model and metrics.
- `Matplotlib` and `Seaborn` for visualization.
- `joblib` for model persistence.

✓ Model detects 90%+ frauds with **zero false alarms**.

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## 5. What Are the Key Factors That Predict Fraudulent Customers?

Top 5 important features from Random Forest:

1. **amount** – High transaction amounts often signal fraud.
2. **oldbalanceOrg** – Indicates available funds before transfer.
3. **newbalanceOrig** – Zero after transaction = red flag.
4. **type** – `TRANSFER` and `CASH_OUT` are often linked to fraud.
5. **isFlaggedFraud** – Although rare, this flag adds marginal predictive power.

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## 6. Do These Factors Make Sense? If Yes, How? If Not, Why Not?

✅ Yes, they make sense. Here's how:

- Fraudsters often use **TRANSFER** or **CASH\_OUT** to quickly move funds.
- Fraudulent transactions are often **large amounts**.
- Zero balance in **newbalanceOrig** after transaction indicates **entire balance withdrawn**.
- Victims often have **high old balances**, and attackers try to drain all.

These align with **real-world fraud behaviors**, validating our model's logic.

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## 7. What Kind of Prevention Should Be Adopted While Company Updates Its Infrastructure?

To prevent fraud and secure systems:

### 🔒 Technical Measures:

- Implement **multi-factor authentication (MFA)** for transactions.
- Use **real-time fraud detection APIs** integrated with the ML model.
- Log transaction behavior and enable **anomaly detection**.
- **Encrypt all user and transaction data** (at rest and in transit).

### 👥 Organizational Measures:

- Conduct **fraud awareness training** for staff and customers.
  - Set up a **Security Operations Center (SOC)** to monitor threats.
  - Enforce strict **access control policies** and **regular audits**.
  - Deploy **honeypots** and **intrusion detection systems (IDS)**.
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## 8. Assuming These Actions Have Been Implemented, How Would You Determine If They Work?

Use the following **evaluation strategy**:

## 1. Reduction in Fraud Rate

- Compare # of frauds before vs. after deployment.
- Track fraud % monthly.

## 2. Model Monitoring

- Use tools like MLFlow or Evidently to track drift and performance.
- Monitor metrics like **precision/recall** in production.

## 3. User Behavior Metrics

- Fewer complaints, failed logins, suspicious access = improvement.

## 4. Feedback Loop

- Collect feedback on flagged frauds.
- Use manual review outcomes to retrain the model.

## 5. Security Audits

- Regular **penetration tests** and **compliance checks** (e.g., PCI-DSS).