

Major Project Report

Title:

Online Payments Fraud Detection using Machine Learning

Objective:

To build a machine learning model that classifies online transactions as either fraudulent or non-fraudulent using a real-world dataset. This solution can help financial institutions automatically detect suspicious activity and reduce fraud losses.

Dataset Description:

The dataset used for this project was sourced from Kaggle and contains historical data of transactions. It includes the following columns:

1. **step** – Represents a unit of time (in hours)
 2. **type** – Type of transaction (TRANSFER, CASH_OUT, etc.)
 3. **amount** – Transaction amount
 4. **nameOrig** – Customer initiating the transaction
 5. **oldbalanceOrig** – Sender's balance before transaction
 6. **newbalanceOrig** – Sender's balance after transaction
 7. **nameDest** – Receiver of the transaction
 8. **oldbalanceDest** – Receiver's balance before transaction
 9. **newbalanceDest** – Receiver's balance after transaction
 10. **isFraud** – Label (1 if fraudulent, 0 otherwise)
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Methodology:

Step 1: Data Loading & EDA

- Loaded data using Pandas
- Checked shape, null values, and class distribution

Step 2: Feature Engineering

- Derived new feature: `hour` from `step`
- Encoded `type` using LabelEncoder
- Dropped identifiers like `nameOrig` and `nameDest`

Step 3: Handling Class Imbalance

- Fraudulent transactions were highly underrepresented
- Applied **SMOTE** (Synthetic Minority Oversampling Technique) to balance the dataset

Step 4: Model Training

- Used **Random Forest Classifier** for prediction
- Split data into 70% training and 30% testing

Step 5: Model Evaluation

- Evaluated model using:
 - Confusion Matrix
 - Classification Report (Precision, Recall, F1-score)
 - Visualized **feature importances** using a horizontal bar chart
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Results:

- Model accurately classifies fraudulent transactions with high recall and precision
 - Top contributing features include: `amount`, `oldbalanceOrg`, and `type`
 - Feature importance graph provides explainability for model decisions
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Tools & Technologies Used:

- Python 3
 - Pandas, NumPy
 - Scikit-learn
 - imbalanced-learn (SMOTE)
 - Matplotlib, Seaborn
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Future Work:

- Deploy the model using Streamlit or Flask as a real-time web app
 - Integrate time-series and geolocation-based features
 - Use deep learning techniques for anomaly detection
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Conclusion:

This project demonstrates how machine learning can be used effectively to combat fraud in online payments. With further enhancements, it can be deployed as a full-scale fraud monitoring tool.

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