Report

Credit Card Fraud Detection Using Anomaly Detection

PROJECT TEAM:

Group Members:

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EXECUTIVE SUMMARY:

This project implements Isolation Forest algorithm for credit card fraud detection on a highly imbalanced dataset of 284,807 transactions. Our optimized model achieves 57% fraud recall with ROC AUC of 0.947, demonstrating strong anomaly detection capabilities. The analysis reveals critical temporal patterns where fraud rates peak at 1.4% during early morning hours (1-4 AM), providing actionable insights for real-time fraud prevention systems.

1. DATASET OVERVIEW:

Source: European credit card transactions (September 2013)

Size: 284,807 transactions with 31 features Time Period: 48 hours of transaction data

Class Distribution:

- Normal: 284,315 (99.83%)

- Fraud: 492 (0.17%)

Features:

- V1-V28: PCA-transformed features (privacy protection)

- Time: Seconds elapsed between transactions

- Amount: Transaction value in dollars

- Class: Binary target (0=Normal, 1=Fraud)

Data Quality: Zero missing values, clean dataset ready for modeling.

2. KEY FINDINGS FROM EXPLORATORY ANALYSIS:

- 2.1 Transaction Amount Patterns
- Average transaction: \$88.47
- Median transaction: \$22.00
- Distribution: Highly right-skewed with 75% of transactions under \$77.51
- Fraud vs Normal: Fraudulent transactions show more variability in amount distribution
- 2.2 Temporal Fraud Patterns (Critical Discovery)
- High-Risk Period: Hours 0-5 (Midnight to 5 AM)
- Peak fraud rate: 1.4% at 1 AM
- Why it matters: 8x higher than daytime average (0.17%)
- Business impact: Targeted monitoring during off-peak hours
- Low-Risk Period: Hours 9-23 (Day and Evening)
- Fraud rate: Stable at 0.05-0.2%
- Insight: Normal business hours show baseline fraud activity
- 2.3 Class Imbalance Challenge
- Extreme imbalance (492 frauds in 284,807 transactions) requires anomaly detection techniques

3. ANOMALY DETECTION MODEL: ISOLATION FOREST:

Why Isolation Forest?

- Unsupervised approach: Doesn't require balanced training data
- Anomaly-focused: Specifically designed to isolate outliers
- Scalability: Efficient on large datasets (300 trees trained)
- Interpretability: Provides anomaly scores for threshold tuning

Model Configuration:

- Algorithm: Isolation Forest
- Estimators: 300 trees
- Contamination: Tuned from 0.00167 to 0.01
- Features: 30 (V1-V28 + scaled Amount + scaled Time)

4. PERFORMANCE RESULTS:

Baseline Model (Contamination = 0.00167)

- Precision: 0.239- Recall: 0.239

- F1-Score: 0.239

- ROC AUC: 0.947

Confusion Matrix:

- TN: 282,893, FP: 360, FN: 360, TP: 113

Optimized Model (Contamination = 0.01)

- Precision: 0.096- Recall: 0.575- F1-Score: 0.164- ROC AUC: 0.947

Business Impact:

- 272 frauds detected (vs 113 baseline)
- 159 additional frauds caught = significant loss prevention
- Trade-off: More false alarms require investigation resources

Threshold Analysis:

- 40% Recall → 14.5% Precision
- 50% Recall → 10.4% Precision (Recommended)
- 60% Recall \rightarrow 10.0% Precision

5. VISUAL INSIGHTS:

- PCA Visualization: Fraudulent transactions separate from normal clusters.
- Precision-Recall Curve: Demonstrates recall-precision trade-off.

6. BUSINESS RECOMMENDATIONS:

Implementation:

- Deploy optimized model (contamination=0.01)
- Focus monitoring during 0-5 AM high-risk hours
- Two-tier detection: automatic blocks + manual review

ROI:

Baseline detection: \$28,250 savedOptimized detection: \$68,000 saved

- Additional prevention: \$39,750 per 48 hours

Operational Considerations:

- False positive rate: ∼1%
- Review workload: ~2,800 flagged transactions / 48 hrs

7. TECHNICAL METHODOLOGY:

- Preprocessing: Scaling, duplicate removal

- Model Training: Full dataset (unsupervised)

- Evaluation Metrics: Recall, Precision, ROC AUC

8. LIMITATIONS:

Current Limitations:

- Recall ceiling: 57%

- Precision trade-off: Higher detection = more false positives

- Model trained on 48-hour dataset

9. CONCLUSION:

Isolation Forest achieves 0.947 ROC AUC and 57% fraud recall.

Key discoveries: Temporal fraud patterns, tunable system, actionable recommendations.

Potential savings: \$39,750 per 48 hours of fraud prevention.