

Synthetic Dataset Generation Logic

Dataset Generation Logic Documentation

Overview

This document describes the logic for generating a synthetic credit card fraud dataset. The system simulates customer transactions, merchant profiles, and various types of fraudulent activities to generate a realistic mix of legitimate and fraudulent transactions, simulating real-world fraud patterns for training fraud detection models.

Core Components of Python Scripts

Utility Class (Utility)

Handles helper functions for:

- **IP Address Generation:** Creates random IPv4 addresses tied to country codes.
- **Date/Time Operations:** Converts timestamps, adjusts for holidays/weekends.
- **File Operations:** Exports data to CSV.

Customer Class (Customer)

Generates customer profiles with:

- **Income Levels:** Low, medium, high (affects transaction behavior).
- **Location:** Random (x, y) coordinates.
- **Transaction Behavior:** Mean amount, frequency, online/offline ratio.
- **Credit Cards:** Randomly assigned based on income.
- **IP Addresses:** Primary (home country) + secondary/tertiary (random countries).

Merchant Class (Merchant)

Generates merchant profiles with:

- **Categories:** From MCC (Merchant Category Codes).
- **Size:** Small/medium/large (affects POS terminals).
- **Store Credit Cards:** Large grocery stores more likely to issue them.
- **Fake Merchants:** Optionally generated for fraud scenarios.

Transaction Class (Transaction)

Generates transactions with:

- **Normal Transactions:** Based on customer habits.
- **Fraudulent Transactions:**
 - **CNP (Card Not Present) Testing:** Small transactions to validate stolen cards.
 - **CNP Monetization:** Larger fraudulent purchases after testing.
 - **CP (Card Present) Cloning:** Physical card fraud at unusual locations.

Fraud logic

Fraud Generation Overview

The system simulates two types of fraud: card present (CP) and card not present (CNP)

1. **CNP (Card Not Present)** – Small transactions to validate stolen card details in first phase, then large fraudulent purchases after a certain period of time if testing is successful
2. **CP (Card Present) Cloning** – Use cloned physical card fraud at unusual locations.

Fraudulent transactions are injected into legitimate transactions based on probabilistic rules.

Key Assumptions

1. Fraudsters test cards before monetizing.
2. CNP fraud is more common than CP fraud.
3. Fraudsters prefer holidays/weekends.
4. Physical card fraud involves distant POS terminals.

Fraud Lifecycle

Step 1: Compromising Customers

- **Daily Selection:**
 - Each day, `compromised_customer_nb_per_day` customers are randomly selected.
 - **CNP Fraud:** 2 customers/day.
 - **CP Fraud:** 1 customer/day (rarer).
- **Holiday Influence:**
 - Fraud probability increases during holidays (65% vs. normal 35%).

Step 2: Fraud Testing Phase (CNP & CP)

CNP Testing (Online Fraud)

- **Behavior:**
 - Small transactions (\$10–30) to check if the card is valid.
 - Transactions occur in quick succession (within 10–30 minutes).
- **IP Usage:**
 - 80%: Random IP (mimicking attacker).
 - 20%: Blacklisted IP (known fraudster IPs).

CP Cloning (Offline Fraud)

- **Behavior:**
 - Transactions at POS terminals **far from** the customer's usual location.
 - Uses **physical card-present** transactions.

Step 3: Monetization Phase (Large Fraud)

After a **testing phase gap** (default: 30 days for CNP, 1 day for CP), fraudsters execute large transactions:

- **CNP Monetization:**
 - 1 large transaction/day for **14 days**.
 - Amounts up to **\$3,500** (3x normal spending).
 - **CP Monetization:**
 - Shorter window (**7 days**).
 - Also at distant POS terminals.
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Fraud Transaction Mechanics

Fraudulent Transaction Generation

For each compromised customer:

1. **Determine Fraud Type** (cnp_testing, cnp_monetization, cp_cloning).
2. **Generate Transactions:**
 - **If cnp_testing:**
 - Small, rapid-fire transactions.
 - Always online (card not present).
 - **If cnp_monetization:**
 - Reuses the **same card** from testing.
 - Large amounts (scaled up from normal spending).
 - **If cp_cloning:**
 - Uses far-away POS terminals (unusual location).
3. **Adjust for Seasonality:**
 - **65% of frauds** are shifted to holiday periods.
 - **Weekend Bias:** 30% shifted to weekends.

Fake Merchants (Future Implementation)

- Used for QR code scams.
 - Transactions routed through fake merchant IDs.
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Generation of Dataset Output

Data Generation Workflow

Step 1: Generate Customers & Merchants

- **Customers:** n_customer profiles with randomized spending habits.
- **Merchants:** n_merchant profiles with MCC categories.

Step 2: Simulate Transactions

- **Normal Transactions:**
 - Generated daily per customer (Poisson-distributed counts).
 - Adjusted for weekends (20% shifted to weekends).
- **Fraudulent Transactions:**
 - **Testing:** 3–6 small transactions/day.
 - **Monetization:** 1 large transaction/day for 14 days (CNP) or 7 days (CP).

Step 3: Post-Processing

- **Holiday Adjustments:** 65% of frauds shifted to holiday periods.
- **Time Normalization:** Ensures transactions stay within the n_days period.

Output Dataset Structure

Column	Description
transaction_date	Timestamp of transaction
customer_id	Unique customer identifier
amount	Transaction amount
merchant_id	Merchant ID
pos_id	POS terminal ID (NULL for online)
IP_address	IP used (NULL for in-person)
type_of_credit_card_used	Card category
card_present_or_not	card present / card not present
is_fraud	1 (fraud) / 0 (legitimate)

Future Roadmap

1. **Dynamic Fraud Patterns:**
 - Geo-velocity checks (impossible travel detection).
2. **More Fraud Types:**
 - Account takeover
 - Fake merchant QR code scams.
3. **Code Optimization:**
 - Class structure to be optimized.
 - Replace iterrows with vectorized operations
4. **Enhanced Fraud Logic:**
 - Dynamic fraud patterns (e.g., geo-velocity checks).
5. **Unit Tests:**
 - Validate transaction/fraud generation functions for future reuse.