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BANK FRAUD

DETECTION SYSTEM



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1. Introduction

This project simulates a **bank fraud detection system** using a combination of OOP and machine learning. The system models **users**, **transactions**, **banks**, and **analysts**, where an AI model predicts the likelihood of a transaction being fraudulent.

The main goal is to **demonstrate ML integration into a real-world scenario** and to create a modular, testable Python project.

2. Project Overview

The project consists of:

- Users performing deposits, withdrawals, and transfers.
- Banks coordinating transactions and calling analysts for approval.
- Analysts using a trained **ML model** to assess transaction risk.
- Transactions that log all activities for traceability.

Additionally, the project includes a **FastAPI interface** to expose the ML model for external clients.

3. Features

- **User Management:** Create users, check balances, perform transactions.
- **Fraud Detection:** ML model predicts fraud probability (`p_fraud`) and fraud status (`is_fraud`).
- **Manual Review:** Analysts can approve or block suspicious transactions.
- **Logging:** Full traceability of transactions, decisions, and ML predictions.
- **API:** FastAPI endpoint to submit transaction data and receive predictions.
- **Unit & Integration Testing:** Pytest tests covering classes and API endpoints.

4. Project Structure

Bank Fraud Detection System/

- **README.pdf**
- **requirements.txt** # Needed libraries and packages
- **main.py** # Demo program
- **api.py** # API's configuration
- **data/** # Needed data for ML model training
 - **raw_data.csv**
 - **data_cleaner.py** # Cleans raw_data.csv y returns cleaned_data.csv
 - **cleaned_data.csv**
 - **add_concept_to_data.csv** # Adds concepts and Is_fraud columns to cleaned_data.csv and returns final_data.csv
 - **final_data.csv** #Data ready for ML model training
- **ML/**
 - **train_model.py** # Trains ML model and saves the model as model.pkl
 - **model.pkl** # Trained model
 - **predict.py** # Uses model.pkl in order to make predictions
 - **utils.py** # Auxiliar functions of predict.py
- **bank_system/**
 - **user.py** # Class User
 - **bank.py** # Class Bank
 - **analyst.py** # Class Analyst
 - **transaction.py** # Class Transaction
- **tests/** #unit testing for bank_system and API's
 - **test_user.py**
 - **test_bank.py**
 - **test_analyst.py**
 - **test_transaction.py**
 - **test_api.py**

5. Installation

1. Clone the repository:

```
git clone <your-repo-url>
```

```
cd project_root
```

2. Create a virtual environment:

```
python -m venv venv
```

```
source venv/bin/activate # Linux/Mac
```

```
venv\Scripts\activate # Windows
```

3. Install dependencies:

```
pip install -r requirements.txt
```

6. Running the Project

- **Demo (CLI):** Run main.py to simulate transactions locally.
- **Train ML Model:** Execute train_model.py to train and save the model.
- **FastAPI:** Run api.py to start the API server: uvicorn api:app --reload
- Access the API documentation at <http://127.0.0.1:8000/docs>.

7. Machine Learning Model

- Model: Logistic Regression with TfidfVectorizer for the Concept column and StandardScaler for numeric features.
- Features: Amount, time_sin, time_cos, Is_Night, Concept.
- Target: Is_Fraud (1 = fraud, 0 = safe).
- Dataset: final_data.csv with probabilistic assignment of fraud based on amount, time, and concept.
- Metrics Achieved:
Accuracy: 0.77
ROC-AUC: 0.82
Precision / Recall / F1-scores vary for fraud and safe classes.

8. API Usage

Endpoint: /predict (POST)

Request body(.json):

```
{  
  "amount": 51000,  
  "concept": "urgent crypto investment",  
  "local_time": "22:30:00"  
}
```

Response body(.json):

```
{  
  "is_fraud": 1,  
  "p_fraud": 0.92  
}
```

- The API allows external clients to receive ML predictions without needing the full project.
- local_time is sent as a string in "HH:MM:SS" format.

9. Logging

- **Purpose:** Track all transactions, analyst decisions, and ML predictions.
- **Logging Levels:**
 - INFO: Transaction creation, approvals, and rejections.
 - WARNING: Manual review required.
- **Example:** In transaction.py, logger.info records each new transaction.
- Logs are saved in logs/bank_system.log and include timestamps.

10. Testing

- **Unit Tests:** For User, Bank, Analyst, Transaction.
- **Integration Tests:** Full bank transaction flows.
- **API Tests:** Using fastapi.testclient to validate /predict endpoint.
- Tests are runned with “python -m pytest” on a cmd on the main directory of the project