

# **Software Requirements Specification (SRS)**

**Project Name:** Fraud Detection System for Financial Transactions

**Author:** SANKET RAVINDRAKUMAR SHRIVASTAV

**Date:** 03-Feb-2026

**Version:** 1.0

---

## **1. Introduction**

### **1.1 Purpose**

The purpose of this system is to provide a **real-time fraud detection mechanism** for financial transactions. It uses **machine learning** to identify unusual or suspicious transaction patterns and flag them for review. This helps banks, payment platforms, and financial institutions reduce financial risks.

### **1.2 Scope**

- Web-based application using Python and Flask.
- Users can input transaction details (amount, location, account type, etc.).
- The system predicts whether a transaction is **fraudulent** or **legitimate**.
- The system is **extensible**: it can be connected to a database or real-time transaction feed in the future.

### **1.3 Intended Audience**

- Banks and financial institutions.
- Project reviewers on LMS.
- Developers and stakeholders interested in ML applications in finance.

### **1.4 Definitions / Acronyms**

- **ML:** Machine Learning
  - **CSV:** Comma-Separated Values
  - **SRS:** Software Requirements Specification
  - **API:** Application Programming Interface
-

## **2. Overall Description**

### **2.1 Product Perspective**

- This is a standalone web application.
- Can be extended to connect with **real-time banking APIs**.
- Uses **Python ML libraries** (scikit-learn, pandas, numpy).

### **2.2 Product Functions**

1. Accept user input for transaction details.
2. Apply ML model to predict fraudulent transactions.
3. Display the prediction result instantly on the web page.
4. Optional: Export predictions to CSV (future extension).

### **2.3 User Classes and Characteristics**

- **End User:** Bank employees, financial auditors.
- **Developer:** Anyone maintaining or enhancing the system.
- **Admin:** Can modify ML model or extend system functionality.

### **2.4 Operating Environment**

- Windows 10/11, Linux, or Mac OS.
- Python 3.10+ environment.
- Local CMD or terminal to run Flask app.

### **2.5 Design and Implementation Constraints**

- Must use Python and Flask.
  - Must run in a virtual environment (venv).
  - ML model must be pre-trained or trained on sample dataset.
-

### **3. System Features**

<b>Feature</b>	<b>Description</b>	<b>Priority</b>
Transaction Input	User enters transaction amount, location, account type, etc.	High
Fraud Prediction	ML model predicts fraud / legitimate transactions	High
Result Display	Shows prediction instantly on the web page	High
Data Export	Save predictions to CSV (optional)	Medium
Security	No sensitive information stored	High

---

### **4. Functional Requirements**

1. User can open a browser and access the web app.
  2. User can enter transaction details via a web form.
  3. ML model predicts fraud or legitimate status.
  4. Prediction result is displayed immediately.
  5. System should handle invalid inputs gracefully.
- 

### **5. Non-Functional Requirements**

#### **Requirement Description**

Performance	Results displayed in under 2 seconds.
Reliability	System should not crash on normal inputs.
Scalability	Can handle multiple transaction inputs in the future.
Security	No sensitive data is stored; only simulated inputs.
Usability	Easy-to-use web interface.

---

## **6. Technical Requirements**

- **Programming Language:** Python 3.10+
  - **Framework:** Flask
  - **Libraries:** pandas, scikit-learn, numpy, textblob (optional)
  - **Web:** HTML, CSS for frontend
  - **IDE:** CMD / VS Code / PyCharm
  - **OS:** Windows/Linux
- 

## **7. System Design Approach**

### **1. Frontend:**

- HTML form for user input.
- CSS for basic styling.

### **2. Backend:**

- Flask routes handle / (home page) and /predict.
- ML model loaded in Python (scikit-learn) predicts the transaction.

### **3. Data Flow:**

4. User input → Flask backend → ML model → Prediction → Display on webpage

### **5. ML Model:**

- Pre-trained Random Forest / Logistic Regression model (sample dataset).
  - Can be replaced with real banking data in future.
- 

## **8. Testing Plan**

- **Unit Testing:** Verify individual functions like ML prediction, input validation.
  - **Integration Testing:** Ensure form input, prediction, and result display work together.
  - **User Testing:** Test with sample transactions for correctness.
  - **Edge Cases:** Empty input, extremely high values, unusual location.
-

## 9. Deliverables

1. Fully functional Flask app (app.py)
  2. Templates and static files (index.html, style.css)
  3. requirements.txt for dependencies
  4. SRS document (this document)
  5. README.md for GitHub / LinkedIn submission
- 

## 10. References

- Flask Documentation: <https://flask.palletsprojects.com/>
  - scikit-learn Documentation: <https://scikit-learn.org/stable/>
  - pandas Documentation: <https://pandas.pydata.org/>
- 

Save this as **SRS.docx** or **SRS.pdf** in your project folder:

fraud\_detection\_system/

```
|  
|   └── SRS.docx  
|  
|   └── app.py  
|  
|   └── requirements.txt  
|  
|   └── README.md  
|  
|   └── templates/  
|       └── static/
```

---