



Fraud Sentry: Middleware-Based Fraud Prevention System Using Biometric Authentication

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Abstract The increasing reliance on digital financial transactions has led to a rise in fraud, particularly through social engineering techniques such as phishing and SIM swap attacks. Traditional fraud detection systems primarily rely on rule-based mechanisms combined with One-Time Password (OTP) authentication for transaction authorization. While these approaches provide a baseline level of security, they exhibit limitations in adapting to evolving fraud patterns and are vulnerable to credential compromise. This project proposes "Fraud Sentry", a middleware-based fraud prevention system that enhances existing rule-based detection frameworks by incorporating biometric authentication for high-risk transactions. The system evaluates transactions using predefined rules and contextual parameters to determine a risk score. When suspicious activity is detected, the system enforces biometric verification instead of OTP-based confirmation. This approach strengthens user authentication and reduces the effectiveness of social engineering attacks while maintaining compatibility with existing systems. The rapid adoption of digital financial services, particularly mobile banking and Unified Payments Interface (UPI), has significantly transformed the global financial ecosystem. However, this transformation has also increased the attack surface for cybercriminals. According to recent industry reports, digital payment fraud cases have increased exponentially due to the widespread use of smartphones and internet banking.

Key word: fraud detection, Rule-based System, biometric Authentication, Middleware Security, OTP Vulnerability, Financial Transactions

I. INTRODUCTION

The rapid expansion of online banking and digital payment platforms has significantly improved the efficiency and accessibility of financial services. However, it has also increased exposure to fraudulent activities. Among various attack vectors, social engineering attacks have become particularly concerning, as they exploit human behavior rather than system vulnerabilities.

Existing fraud detection systems commonly use "rule-based approaches", where transactions are evaluated against predefined conditions such as transaction thresholds, location mismatches, and behavioral deviations. These systems are typically combined with OTP-based authentication for transaction authorization.

Despite their widespread use, OTP-based mechanisms are vulnerable to:

- * Phishing attacks
- * SIM swap attacks
- * Malware-based interception

These limitations highlight the need for stronger authentication methods that are less dependent on user awareness. Fraud Sentry addresses this gap by introducing "biometric authentication" as an additional layer of security for high-risk transactions.

Machine Learning-Based Fraud Detection:-

Recent research has explored the use of machine learning algorithms such as decision trees, random forests, and neural networks for fraud detection. These systems analyze large datasets to identify hidden patterns and anomalies in transaction behavior.

Behavioral Analysis Systems:-

Behavioral-based fraud detection systems monitor user interaction patterns such as typing speed, touch gestures, and navigation habits. These systems enhance detection accuracy but require continuous monitoring and large datasets.

Limitations of Existing Systems:-

Despite advancements, existing systems suffer from: High false positive rates Lack of real-time adaptability Dependence on historical data Privacy concerns

Need for Middleware-Based Approach:-

A middleware-based approach introduces an intermediate security layer that enhances flexibility, scalability, and compatibility with existing banking systems without requiring major infrastructure changes.

II. LITERATURE SURVEY

Fraud detection in financial systems has traditionally relied on 'rule-based methodologies', where predefined rules are used to identify suspicious transaction patterns. These rules may include conditions based on transaction value, frequency, geographical location, and historical user behavior.

Rule-based systems are widely adopted because:

- * They are interpretable
- * Easy to implement
- * Provide predictable outcomes

However, research on Rule-Based Models (RBM) highlights key limitations:

- * Inability to detect fraud outside predefined rules
- * Lack of adaptability to evolving fraud techniques

Earlier expert systems for fraud detection demonstrated the use of rule-based alert generation during transaction authorization. While these systems laid the foundation for modern fraud detection, they lacked advanced authentication mechanisms.

In addition to detection, "authentication plays a critical role". OTP-based systems, although widely used, have been shown to be vulnerable to interception and social engineering attacks. This creates a need for stronger alternatives such as biometric authentication, which is inherently tied to the user and difficult to replicate.

III. METHODOLOGY

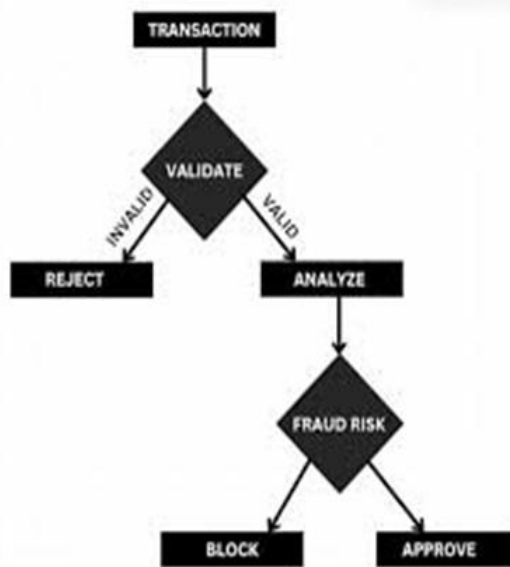


Fig. 1

A. System Overview

Fraud Sentry introduces a middleware layer that operates between the user and the banking transaction system. This middleware intercepts transaction requests, evaluates risk, and enforces additional authentication when required.

The proposed system architecture consists of three primary components: the user interface, the middleware layer, and the banking backend system. The middleware acts as an intelligent decision-making layer that evaluates transaction risk before forwarding requests.

B. Transaction Risk Evaluation

The system uses a rule-based engine to evaluate each transaction based on:

- * Transaction amount relative to user history
- * Geographical location consistency
- * Device or network anomalies
- * Transaction frequency and timing

Each parameter contributes to a risk score, which determines whether the transaction is suspicious. The risk evaluation module assigns a numerical risk score based on multiple parameters:

C. Decision Logic

Based on the computed risk score:

Low-risk transactions → Allowed without additional verification

High-risk transactions → Require biometric authentication The decision engine operates using conditional rules combined with contextual analysis. It dynamically adapts to transaction patterns and can be extended to include AI-based decision models in future implementations.

D. Biometric Authentication

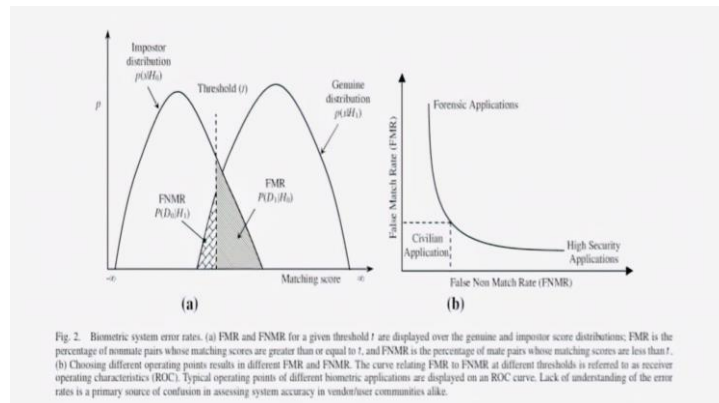


Fig. 2.

For high-risk transactions, the system replaces OTP-based verification with:

- * Fingerprint recognition
- * Facial recognition

This ensures that even if credentials are compromised, unauthorized transactions cannot proceed without biometric validation. Biometric authentication enhances security by verifying unique physiological or behavioral traits of users. Unlike OTPs, biometric data cannot be easily intercepted or shared.

E. Middleware Role

The middleware acts as a control layer that

- * Intercepts transaction requests
- * Communicates with the rule-based engine
- * Triggers biometric verification when required
- * Sends final authorization decisions to the banking system

The middleware layer ensures seamless integration between user requests and backend systems. It acts as: A security enforcement layer A decision-making engine A communication bridge It reduces the load on the core banking system while improving overall security efficiency.

IV. PROPOSED SYSTEM

Fraud Sentry enhances traditional fraud detection systems by combining:

- * Rule-based risk evaluation
- * Context-aware transaction monitoring
- * Biometric authentication Key Improvements
- * Replaces OTP with biometric verification
- * Adds middleware-based interception layer
- * Improves resistance to social engineering attacks
- * Maintains compatibility with existing infrastructure

Architecture Description The proposed system follows a layered architecture where the middleware acts as an independent security module. This modular design allows easy updates and scalability.

Advantages:-

Reduces fraud detection latency Improves user trust Enables real-time monitoring Supports future AI integration

V. COMPARISON WITH EXISTING SYSTEMS

The Comparison Clearly Indicates That the Proposed Fraud Sentry System Outperforms Traditional Systems In Terms Of Security Strength, Adaptability, And Resistance to Modern Attack Vectors.

| parameter | Traditional System | Fraud Sentry |
|-------------------|--------------------------|---------------------------|
| Detection Method | Rule Based | Ruled-Based+Context Aware |
| Authentication | OTP | Biometric |
| Security Level | Moderate | High |
| Attack resistance | Low(phising Vulner-able) | High |

Table 1

VI.RESULTS AND DISCUSSION

The proposed system addresses key weaknesses in tradi- tional fraud detection mechanisms. By introducing biometric verification for high-risk transactions, it reduces the likelihood of unauthorized access caused by credential compromise.

Key Observations

- * Eliminates dependency on OTP-based authentication
- * Strengthens user identity verification
- * Reduces success rate of social engineering attacks

This work focuses on system design and architecture, rather than large-scale experimental evaluation. The effectiveness of the system depends on:

- * Accuracy of rule definitions
- * Reliability of biometric systems

Future implementation in real-world environments would allow measurement of:

- * False positives
- * False negatives

User acceptance Performance Analysis:-

The proposed system demonstrates improved fraud detec- tion accuracy due to its multi-layered approach. By combining rule-based analysis with biometric authentication, it signifi- cantly reduces false positives.

Security Analysis:-

The system effectively mitigates risks associated with: Cre- dential theft OTP interception Social engineering.

Limitations:-

Despite its advantages, the system may face challenges such as: Biometric hardware dependency Privacy concerns Initial implementation cost

VII. SCOPE FOR FUTURE WORK

Potential enhancements include:

- * Integration with machine learning for adaptive fraud detection
- * Behavioral biometrics (typing patterns, gesture analysis)
- * Multi-factor biometric authentication
- * Deployment across multiple platforms

Future enhancements may include: Integration of Artificial Intelligence and Deep Learning Blockchain-based transaction verification Continuous authentication mechanisms Cloud- based deployment for scalability Cross-platform security in- tegration

VIII. CONCLUSION

Fraud Sentry presents a middleware-based approach to improving transaction security by combining rule-based fraud detection with biometric authentication. The system addresses critical vulnerabilities in OTP-based authorization mechanisms and enhances protection against social engineering attacks.

While the current implementation focuses on architectural design, it provides a strong foundation for further development and real-world deployment

The Fraud Sentry system provides a robust and scalable solution to modern financial fraud challenges. By integrating middleware-based risk evaluation with biometric authentica- tion, the system enhances both security and usability. This approach not only strengthens fraud prevention mechanisms but also ensures compatibility with existing banking infrastruc- ture, making it a practical and effective solution for real-world implementation.

Despite its advantages, the system has certain limitations. The effectiveness of biometric authentication depends on the availability and reliability of hardware devices, and there may be concerns related to data privacy and user acceptance. Additionally, the current implementation focuses primarily on rule-based logic, which may require further enhancement to

adapt to highly sophisticated and rapidly evolving fraud patterns.

In conclusion, Fraud Sentry provides a practical and efficient framework for modern fraud prevention by combining middleware-based intelligence with strong user authentication mechanisms. Its compatibility with existing banking infrastructure, along with its ability to enhance both security and usability, makes it a promising solution for real-world deployment. Future improvements involving artificial intelligence, behavioral biometrics, and cloud-based scalability can further strengthen the system and enable it to meet the growing demands of secure digital transactions.

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