

## CREATION OF A SECURE PAYMENT SYSTEM INTEGRATED WITH ARTIFICIAL INTELLIGENCE USING BLOCKCHAIN TECHNOLOGY BASED ON JAVA

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**Abstract:** The rise of digital payments has transformed commerce, enabling seamless transactions across the globe. However, security concerns such as fraud, data breaches, and unauthorized access pose significant challenges. Blockchain technology, with its decentralized and immutable ledger, offers a robust foundation for secure payment systems. When integrated with Artificial Intelligence (AI), blockchain-based payment systems can leverage predictive analytics, anomaly detection, and fraud prevention to enhance security and efficiency. Java, with its robust libraries and cross-platform capabilities, is an ideal programming language for implementing such systems. This article explores the creation of a secure payment system integrated with AI using blockchain technology, implemented in Java, addressing methods, challenges, solutions, mathematical formulations, and key algorithms.

**Keywords:** Artificial Intelligence (AI), anomaly detection, Blockchain technology, security, Authentication and Authorization.

Developing a secure payment system using blockchain and AI involves integrating decentralized ledgers, cryptographic security, and intelligent analytics. Below are key methods, supported by Java libraries and mathematical formulations.

- **Transaction Validation:** Transactions are validated by consensus mechanisms like Proof of Work (PoW) or Proof of Stake (PoS). The computational cost of PoW is:

$$C_{PoW} = H \cdot T_{hash}$$

where  $C_{PoW}$  is the computational cost,  $H$  is the number of hashes, and  $T_{hash}$  is the time per hash.

- **Smart Contracts:** Smart contracts automate payment logic. The execution time is:

$$T_{exec} = \sum_{i=1}^n O_i \cdot T_{op}$$

where  $T_{exec}$  is execution time,  $O_i$  is the number of operations for instruction  $i$ , and  $T_{op}$  is the time per operation.

- Implementation: Use web3j to interact with Ethereum smart contracts in Java, ensuring secure transaction processing.

### AI-Driven Fraud Detection

AI enhances security by detecting fraudulent transactions in real-time using machine learning and deep learning.

- Anomaly Detection: Isolation Forest identifies unusual transaction patterns. The anomaly score is:

$$s(x, n) = 2^{-\frac{E(h(x))}{c(n)}}$$

where  $s(x, n)$  is the anomaly score,  $E(h(x))$  is the average path length, and  $c(n)$  is the average path length for  $n$  samples.

- Classification: Random Forest classifies transactions as legitimate or fraudulent. The classification accuracy is:

$$A = \frac{TP + TN}{TP + TN + FP + FN}$$

where  $TP$ ,  $TN$ ,  $FP$ ,  $FN$  are true positives, true negatives, false positives, and false negatives. Use Java libraries like Weka or Deeplearning4j for machine learning models.

### Cryptographic Security

Cryptography ensures data confidentiality, integrity, and authenticity in payment systems. • Encryption: AES encrypts transaction data. The encryption time is:

$$T_{enc} = \frac{D}{R_{enc}}$$

where  $T_{enc}$  is encryption time,  $D$  is data size, and  $R_{enc}$  is the encryption rate.

- Digital Signatures: ECDSA (Elliptic Curve Digital Signature Algorithm) ensures transaction authenticity. The signing time is:

$$T_{sign} = T_{gen} + T_{verify}$$

where  $T_{sign}$  is total signing time,  $T_{gen}$  is key generation time, and  $T_{verify}$  is verification time.

Java's `java.security` package supports AES and ECDSA.

### Authentication and Authorization

Strong authentication prevents unauthorized access, while authorization ensures users access only permitted resources.

• Multi-Factor Authentication (MFA): Combines passwords, biometrics, and tokens. The probability of unauthorized access is:

$$P_{unauth} = \prod_{i=1}^k P_i$$

where  $P_{unauth}$  is the probability of bypassing all  $k$  factors, and  $P_i$  is the failure probability of factor  $i$ .

Use Javas Auth0 library for OAuth-based authentication.

Blockchain scalability ensures high transaction throughput. Sharding and off-chain solutions like Lightning Network improve performance.

Throughput: Transaction throughput is:

$$\Theta = \frac{N_{tx}}{T}$$

where  $\Theta$  is throughput,  $N_{tx}$  is the number of transactions, and  $T$  is time.

Use Hyperledger Fabric with Java SDK for scalable blockchain solutions.

Blockchain systems often face scalability issues due to high transaction volumes.

High latency in transaction confirmation:

$$L_{tx} = T_{validate} + T_{consensus}$$

where  $L_{tx}$  is transaction latency,  $T_{validate}$  is validation time, and  $T_{consensus}$  is consensus time.

Implement sharding to distribute transactions across nodes:

$$T_{shard} = \frac{T_{total}}{N_{shards}}$$

where  $T_{shard}$  is sharded transaction time, and  $N_{shards}$  is the number of shards. Use Javas web3j with Ethereum sharding.

### Data Privacy

Payment systems handle sensitive financial data, requiring robust privacy measures.

• Problem: Centralized data storage risks breaches, with privacy loss:

$$\epsilon = \ln \left( \frac{P(M|D)}{P(M|D')} \right)$$

where  $\epsilon$  is the privacy budget,  $P(M|D)$  and  $P(M|D')$  are model output probabilities.

• Solution: Use zero-knowledge proofs (ZKPs) for private transactions and federated learning for AI models:

$$\Delta W = \sum_{i=1}^k \nabla L_i(W)$$

where  $\Delta W$  is the aggregated model update, and  $\nabla \text{Li}(W)$  is the gradient from device  $i$ . Implement ZKPs with zkSNARK libraries in Java.

### Computational Overhead

AI and blockchain are computationally intensive, increasing costs.

– Problem: High computational complexity:

$$C_{total} = C_{AI} + C_{blockchain}$$

where  $C_{total}$  is total complexity,  $C_{AI}$  is AI computation, and  $C_{blockchain}$  is blockchain computation.

– Solution: Optimize AI models with pruning techniques and use lightweight consensus algorithms. The optimized complexity is:

$$C_{opt} = C_{total} \cdot \beta$$

where  $\beta$  is a reduction factor. Use Java's DeepLearning4j for optimized neural networks.

### Key Algorithms for Secure Payment Systems

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#### Algorithm 1 Isolation Forest for Fraud Detection

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**Input:** Transaction data  $X = \{x_1, \dots, x_n\}$ , number of trees  $T$ , sample size  $s$

**Output:** Anomaly scores  $s(x, n)$

**for**  $t = 1$  to  $T$  **do**

Sample  $s$  points randomly from  $X$

Build isolation tree by recursive random splits

Compute path length  $h(x)$  for each  $x \in X$

**end for**

Compute average path length:  $E(h(x)) = \frac{1}{T} \sum_{t=1}^T h_t(x)$

Compute anomaly score:  $s(x, n) = 2^{-\frac{E(h(x))}{c(n)}}$

**Return:**  $s(x, n)$

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**Algorithm 2** AES Encryption for Transactions**Input:** Plaintext  $P$ , key  $K$ , block size  $B$ **Output:** Ciphertext  $C$ Initialize AES cipher with  $K$  in CBC modePad  $P$  to multiple of  $B$ Split  $P$  into blocks  $P_1, \dots, P_n$ **for** each block  $P_i$  **do**    Encrypt:  $C_i \leftarrow \text{AES}_K(P_i \oplus IV)$     Update IV:  $IV \leftarrow C_i$ **end for****Return:**  $C = C_1 || \dots || C_n$ **Algorithm 3** ECDSA for Digital Signatures**Input:** Message  $m$ , private key  $d$ , public key  $Q$ **Output:** Signature  $(r, s)$ Generate random  $k \in [1, n - 1]$ Compute point:  $R = kG$ , where  $G$  is the base pointCompute  $r = x_R \bmod n$ Compute  $s = k^{-1}(h(m) + d \cdot r) \bmod n$ , where  $h(m)$  is the hash of  $m$ **Return:**  $(r, s)$ 

Creating a secure payment system using blockchain and AI in Java combines decentralized ledgers, cryptographic security, and intelligent analytics to ensure robust transaction processing. Challenges like scalability, privacy, computational overhead, and user errors are mitigated through sharding, zero-knowledge proofs, model optimization, and user education. Mathematical formulations and algorithms, including Isolation Forest, AES, and ECDSA, provide a rigorous foundation for implementation. By leveraging Javas libraries and best practices, developers can build secure, scalable, and efficient payment systems, transforming digital commerce.

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