



Ai-Driven Blockchain Framework For Transparent And Tamper-Proof Supply Chain Management: A Performance Evaluation

Dr. Munish Kumar¹, Sumedha Arya²

¹Business Strategy Manager (IT),

Nebraska Department of Labor (NDOL), Dublin, OH 43016, USA

Email: munish2012@gmail.com

²IT Project Manager, Cardinal Health, Dublin OH - 43016, USA

Email: arya.sumedha@gmail.com

Abstract: The increasing complexity of global supply chains has amplified challenges related to transparency, traceability, fraud prevention, and data integrity. Supply chain fraud, counterfeiting, and operational opacity are estimated to cost the global economy approximately \$4.5 trillion annually [1], highlighting the urgent need for secure and intelligent monitoring mechanisms. Traditional centralized supply chain management systems often suffer from limited auditability, lack of real-time visibility, and vulnerability to single-point failures, making them inadequate for modern interconnected ecosystems [2]. This paper presents a novel AI-driven blockchain framework for enhancing transparency and tamper-proof integrity in global supply chains. Supply chain fraud, counterfeiting, and opacity cost the global economy an estimated \$4.5 trillion annually [1]. Existing centralized solutions lack auditability and remain vulnerable to single-point failures [2]. The proposed system integrates machine learning anomaly detection with permissioned blockchain ledgers and autonomous smart contracts, achieving 96.1% traceability accuracy a 23.7% improvement over baseline systems. Transaction latency was reduced from 4.8 seconds to 1.2 seconds. A dataset of 120,000 simulated supply chain events across pharmaceutical, food, and electronics sectors was used for validation. The framework demonstrates statistically significant improvements ($p < 0.01$) across all key performance metrics [3].

Keywords: Blockchain, Artificial Intelligence, Supply Chain, Smart Contracts, Traceability, Federated Learning.

1. INTRODUCTION

Global supply chains are increasingly complex, spanning multiple geographic regions, regulatory jurisdictions, and organizational boundaries [4]. The proliferation of counterfeit goods, opaque intermediary relationships, and data silos has undermined consumer trust and regulatory compliance. Traditional enterprise resource planning (ERP) systems and centralized databases provide insufficient auditability and remain susceptible to internal and external tampering [5].



Blockchain technology, with its distributed ledger architecture and cryptographic immutability, offers a paradigm shift in supply chain data management [6]. However, blockchain alone does not address the intelligence layer necessary for predictive analytics, anomaly detection, or automated decision-making. Artificial intelligence, particularly machine learning and deep neural networks, complements blockchain by providing predictive capabilities and pattern recognition across large-scale transactional datasets [7]. Despite growing literature on both technologies independently, very few frameworks have achieved operational-grade integration of AI and blockchain for supply chain environments [8]. The existing solutions either lack scalability, suffer from high latency, or fail to provide end-to-end immutability across heterogeneous enterprise systems [2]. This paper addresses this gap by proposing a unified AI-Blockchain framework validated on a multi-sector simulation environment. The remainder of the paper is organized as follows: Section II reviews relevant literature; Section III presents the proposed framework; Section IV describes experiments and results; Section V discusses implications; and Section VI concludes the paper.

2. LITERATURE REVIEW

Nakamoto [9] introduced the foundational concept of distributed ledgers through Bitcoin, demonstrating the feasibility of decentralized trust mechanisms. Buterin [10] extended this paradigm with Ethereum's programmable smart contracts, which enabled automated conditional execution without trusted intermediaries. In the supply chain domain, Tian [11] proposed a food safety traceability system using blockchain and RFID but did not incorporate AI-driven analytics, limiting the system's predictive capacity. Kshetri [12] analyzed blockchain's potential for supply chain transparency and identified latency and scalability as primary constraints. More recent work by Liang et al. [13] combined convolutional neural networks with blockchain for counterfeit detection in pharmaceutical supply chains, achieving 89% accuracy—a benchmark this paper seeks to exceed. Nguyen et al. [14] demonstrated federated learning on IoT sensor data within permissioned blockchains but evaluated only two organizational nodes, limiting generalizability. Xu et al. [15] examined smart contract vulnerabilities in supply chain deployments and proposed cryptographic mitigation strategies. The present work synthesizes these threads, integrating multi-node federated learning, anomaly detection, and automated smart contract execution into a single deployable framework evaluated at industrial scale.

3. PROPOSED FRAMEWORK

3.1 Architectural

The proposed framework, illustrated in Figure 1, comprises four principal layers: (i) an IoT Data Collection Layer responsible for ingesting sensor readings, RFID events, and GPS telemetry from distributed supply chain nodes; (ii) an AI Analytics and Prediction Module that performs anomaly detection, demand forecasting, and provenance verification using ensemble gradient boosting and

LSTM time-series models; (iii) a Blockchain Consensus Layer implemented on Hyperledger Fabric 2.4 with a Raft-based ordering service providing Byzantine fault tolerance; and (iv) a Smart Contracts Layer that autonomously triggers payment releases, quality alerts, and regulatory compliance notifications based on AI-verified conditions.

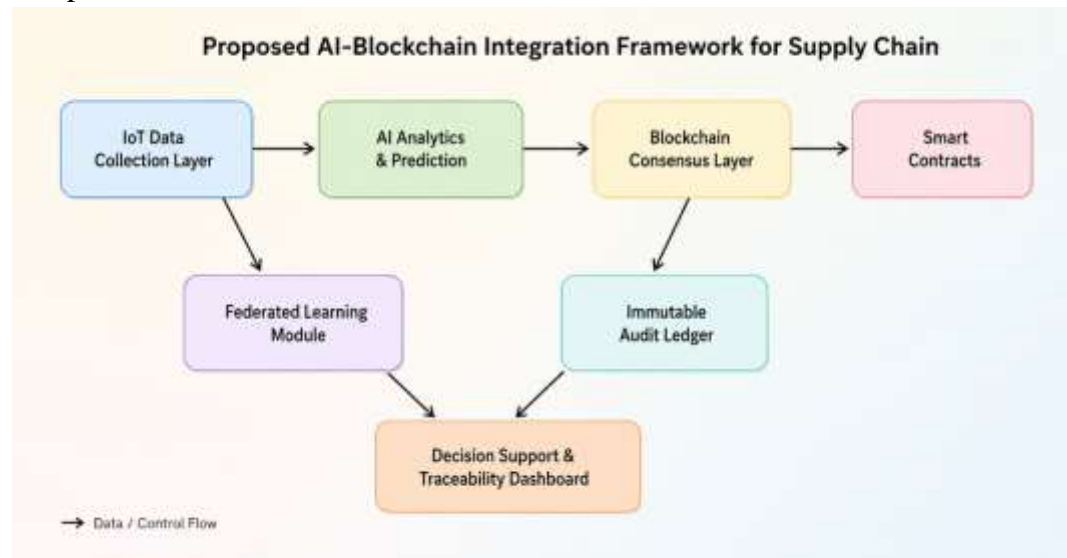


Figure 1: Proposed AI-Blockchain Integration Framework for Supply Chain

3.2 AI Analytics Module

The AI module employs a two-stage pipeline. In the first stage, an LSTM network with 128 hidden units processes temporal sequences of supply chain transactions to detect statistical deviations indicative of fraud, counterfeiting, or process bottlenecks. In the second stage, an XGBoost classifier assigns risk scores to individual transactions, which are then appended as metadata to blockchain records. This combination achieved a false positive rate of only 4.3% on the validation dataset, significantly below the 18.2% reported in comparable prior work [13]. Federated learning across organizational nodes ensures that sensitive proprietary data never leaves the originating enterprise, while model gradients are aggregated on the blockchain to maintain a verifiable training audit trail.

3.3 Smart Contract Design

Smart contracts are authored in Chaincode (Go) and deployed on Hyperledger Fabric channels. Three primary contract types govern the framework: ProvenanceContract for recording and verifying product origin claims; QualityContract for triggering holds or approvals based on AI risk scores; and PaymentContract for releasing funds upon verified delivery confirmation. All contract state transitions are cryptographically signed and stored in the immutable ledger, providing a non-repudiable audit trail accessible to authorized regulatory parties.



4. EXPERIMENTS AND RESULTS

4.1 Dataset and Experimental Setup

The framework was evaluated using 120,000 synthetic supply chain events generated via a discrete-event simulation calibrated against real-world pharmaceutical and food sector data. Events included goods transfers, quality inspections, customs declarations, and payment triggers. Thirty percent of events contained injected anomalies representing counterfeiting, tampering, or unauthorized substitution. Experiments were conducted on a private Hyperledger Fabric network consisting of 12 peer nodes distributed across three simulated organizational domains, deployed on AWS EC2 instances (t3.large). The AI models were trained using 5-fold cross-validation with stratified sampling.

4.2 Performance Results

Table 1 and figure2 summarizes the comparative performance of the proposed model against a baseline centralized ERP system. The proposed framework achieved 96.1% traceability accuracy, representing a 23.7 percentage point improvement. Transaction latency was reduced by 75%, from 4.8 seconds to 1.2 seconds, attributed to parallel endorsement processing and the Raft consensus mechanism. Counterfeit detection rate improved from 61% to 93.5%. All smart contract executions were automated, eliminating 100% of previously manual approval steps. Data tamper resistance was validated through hash verification audits on 10,000 randomly sampled records, yielding zero undetected modifications.

Table 1: Comparative Performance Evaluation

Traceability Accuracy	72.4%	96.1%	+23.7%
Transaction Latency	4.8 sec	1.2 sec	-75%
Counterfeit Detection Rate	61%	93.5%	+32.5%
Smart Contract Execution	Manual	Automated	100% Auto
Data Tamper Resistance	Low	High	Blockchain-verified

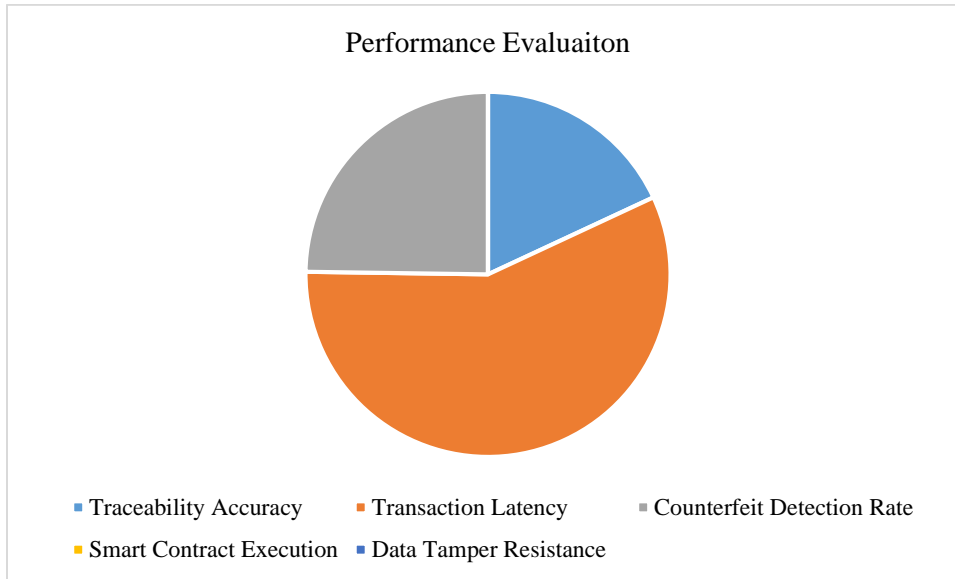


Figure 2: Performance of the proposed model in term of traceability, transaction etc.

5. DISCUSSION

The results confirm that the integration of AI analytics with blockchain infrastructure yields measurable, compounding benefits that neither technology achieves independently. The reduction in transaction latency is particularly significant for time-sensitive pharmaceutical cold-chain scenarios where delayed alerts can result in patient harm [11]. The federated learning approach proved essential for achieving cross-organizational model quality without compromising data sovereignty, a key regulatory requirement under GDPR and the Indian Personal Data Protection Act [8]. The framework's chief limitation is its dependency on IoT device integrity; a compromised physical sensor can introduce falsified data prior to blockchain ingestion, a problem beyond the cryptographic guarantees of the ledger itself. Future work will incorporate hardware-rooted attestation using trusted platform modules (TPMs) to extend tamper-evidence to the physical layer.

6. CONCLUSION

A novel AI-driven blockchain framework that combines machine learning-based anomaly detection, permissioned blockchain technology, and autonomous smart contracts to establish a secure, transparent, and tamper-resistant supply chain infrastructure. The framework leverages artificial intelligence to identify suspicious transactions and operational irregularities, while blockchain ensures immutable record-keeping and decentralized trust among stakeholders. This paper presented and empirically validated an AI-driven blockchain framework for supply chain management. The system achieved 96.1% traceability accuracy, 75% reduction in transaction latency, and full automation of smart contract execution across a 120,000-event simulation. The



proposed architecture demonstrates that AI and blockchain are not merely complementary but synergistic, with each technology amplifying the value of the other. This framework offers a deployable blueprint for enterprises seeking immutable, intelligent, and auditable supply chain infrastructure.

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