

# PureChain-Based Intrusion Detection and Prevention System for Secure Industrial Internet of Things

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**Abstract**—The Industrial Internet of Things (IIoT) intrusions require beyond-detection systems due to their damaging effects. There is a need for a combined intrusion detection and prevention (IDPS) mechanism to ensure complete protection of these critical systems. This study proposes a PureChain-based framework for IIoT to maintain an immutable ledger and a proactive, real-time threat detection and mitigation. The experiments demonstrate low latency, high throughput, minimal resource utilization, and effective attack prevention, showcasing the scalability and suitability of the PureChain for secure IIoT environments.

**Index Terms**—Blockchain, Industrial IoT, Intrusion Prevention System, PureChain, PoA<sup>2</sup>, Security.

## I. INTRODUCTION

Securing Industrial Internet of Things (IIoT) devices remains challenging due to their resource constraints and growing exposure to complex cyber threats [1]. Traditional centralized models fall short, highlighting the need for lightweight, decentralized, and resilient protection frameworks [2]. Blockchain offers a promising foundation by ensuring immutable integrity, transparency, and distributed trust, making it well-suited for intrusion detection systems (IDS) [3]. However, detection alone is not enough; coupling IDS with intrusion prevention systems (IPS) enables both the identification and real-time mitigation of threats, thereby strengthening IIoT security [4].

Research shows a move from reactive IDS to proactive IPS, highlighting the need for adaptive, automated defenses [5]. Hybrid approaches combining deep learning and blockchain [4], as well as low-latency DL-based IPS for 5G/6G networks [6], set the stage. Building on this, this study integrates PureChain with IPS, offering a secure, efficient, and blockchain-verifiable solution for IIoT. Using Proof of Authority and Association (PoA<sup>2</sup>) consensus applied at the transaction level [7], PureChain eliminates block delays, ensures deterministic finality, and delegates heavy validation to trusted nodes, addressing the blockchain “quad-lemma” of decentralization, security, scalability, and cost [8]. This design supports scalable, energy-efficient IIoT IDPS deployments.

## II. METHODOLOGY

Figure 1 shows the PureChain with PoA<sup>2</sup> consensus mechanism to validate IIoT security logs and create immutable records. Alerts flow to a proactive IPS, which evaluates threats and executes smart contract-driven mitigation actions,

such as device isolation or IP blocking, while logging all actions on the PureChain. This design ensures transparent accountability, fast response, and robust protection for IIoT edge devices. For evaluation, we used IoTForge Pro [9], which simulates synthetic and real-world attacks. Security events  $E_k$ , represented as  $E_k = (\tau, H(E_{k-1}), y_t, p_t, \text{metadata}, \sigma_{V_i})$  encode timestamp  $\tau$ , previous event hash output, metadata, and validator signature which are selected based on reputation of prior event via PoA<sup>2</sup> protocol validator as  $V = \{V_1, V_2, \dots, V_m\}$  for each event  $E_k$ . A validator  $V_i$  is selected through a deterministic function  $f$  that incorporates validator reputation  $R(V_i)$  and the previous event hash  $V_i \sim f(R(V_1), R(V_2), \dots, R(V_m), H(E_{k-1}))$ . The designated validator verifies the event and signs it  $\sigma_{V_i} = \text{Sign}_{SK_{V_i}}(H(E_k))$ . Logging latency remains in milliseconds, enabling near real-time validation. The decision function  $\Delta$  maps the state of the system and the output to actions  $a_t$  from the permissible set  $A = \{\text{allow}, \text{quarantine}, \text{shutdown}, \text{throttle}\}$ :  $a_t = \Delta(S_t, (y_t, p_t), \Pi)$ , where the governing policy  $\Pi$  is implemented as threshold-based rules in Equation 1.

$$a_t = \begin{cases} \text{quarantine}, & \text{if } (y_t = c_{\text{malicious}}) \wedge (p_t^{(c_{\text{malicious}})} \geq \tau_{\text{high}}) \\ \text{throttle}, & \text{if } (y_t = c_{\text{suspicious}}) \wedge (\tau_{\text{low}} \leq p_t^{(c_{\text{malicious}})} < \tau_{\text{high}}) \\ \text{allow}, & \text{otherwise} \end{cases} \quad (1)$$

Using calibrated thresholds  $\tau_{\text{high}}$  and  $\tau_{\text{low}}$  to balance false positives and negatives. End-to-end latency  $L_{E2E} = t_{\text{response}} - t_{\text{occurrence}}$  covers detection, logging, and mitigation, ensuring secure and timely responses in IIoT deployments.

## III. EVALUATION PERFORMANCE

Table I shows PureChain achieving 24.56 TPS with 0.068 s latency using 7 validators (quorum of 5). Validators required just 2.07% CPU, 64 MB memory, and 1.85 W power, keeping total system use at 14.5% CPU, 448 MB, and 13 W. Across 278 seconds and 6,802 events, PureChain blocked 100% of attacks with a resilience score of 1.0. Table II highlights that, unlike prior works focused mainly on accuracy or modest scalability, PureChain combines complete attack blocking, blockchain-backed performance (24.56 TPS, 0.068 s

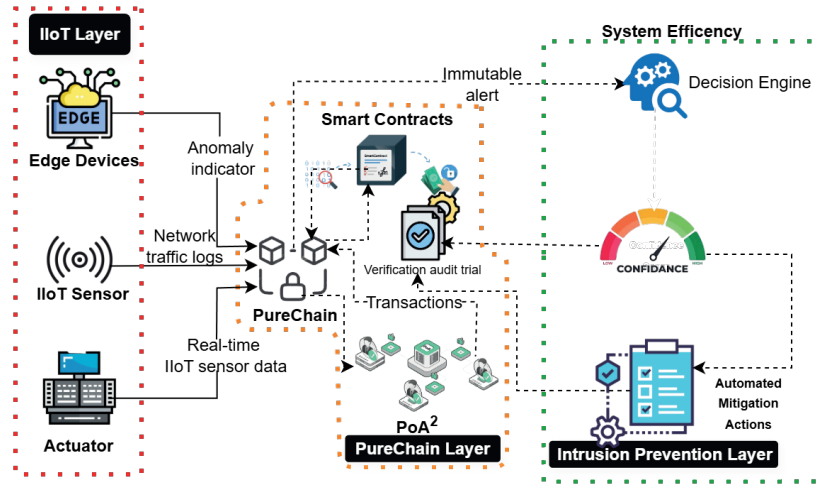


Fig. 1. PureChain Pipeline Process

latency), and low resource costs, offering real-time prevention, scalability, and resilience for IIoT.

TABLE I  
SYSTEM METRICS

| Category           | Metric                         | Result                                   |
|--------------------|--------------------------------|--|
| PureChain Metrics  | Throughput (TPS)               | 24.56                                    |
|                    | Latency (Avg Commit Time)      | ~0.068 s per block                       |
|                    | Validators and Quorum          | 7 validators, quorum = 5                 |
|                    | Scalability & Fault Resilience | Supported                                |
| Resource Estimates | Per-validator Load (CPU)       | 2.07%                                    |
|                    | Per-validator Memory           | 64 MB                                    |
|                    | Per-validator Power            | 1.85 W                                   |
|                    | Total System Load (CPU)        | ~14.5%                                   |
|                    | Total System Memory            | 448 MB                                   |
|                    | Total System Power             | ~13 W                                    |
| System Metrics     | Simulation Duration            | ~278 s                                   |
|                    | Logged Events                  | 6802                                     |
|                    | Security Outcome               | 100% attacks blocked,<br>0% success rate |
|                    | Resilience Proxy               | 1.0                                      |

TABLE II  
COMPARATIVE PERFORMANCE OF IDS/IPS APPROACHES

| Ref.                        | Accuracy (%)                 | Scalability                               | Efficiency                                | Mitigation / Latency                   | Real-Time  |
|-----------------------------|------------------------------|---|---|--|------------|
| [6]                         | 99.9                         | Not reported                              | Not reported                              | Optimized processing speed             | Partial    |
| [4]                         | 96.8%                        | 92.5%                                     | 91.7%                                     | 90.8%                                  | Yes        |
| <b>Proposed (PureChain)</b> | <b>100 (attacks blocked)</b> | <b>24.56 TPS; 7 validators (quorum=5)</b> | <b>CPU 2.07%; Mem 64 MB; Power 1.85 W</b> | <b>~0.068s; Resilience Proxy = 1.0</b> | <b>Yes</b> |

#### IV. CONCLUSION

We presented a PureChain-based intrusion detection and prevention framework for securing IIoT systems. The system utilizes PoA<sup>2</sup> consensus mechanism for tamper-proof logging with low latency and high throughput, while the IDPS layer enables real-time detection, automated mitigation, and auditable responses. The results show minimal resource overhead, 100% attack blocking, and a resilience score of 1.0. Future work will scale the framework to larger IIoT networks, integrate adaptive approaches, and validate it on real-world testbeds.

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