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# TEXT2CHAIN: A MODULAR NLP-ENHANCED ORACLE ARCHITECTURE WITH DECENTRALIZED VALIDATION AND HUMAN-IN-THE-LOOP FOR SECURE OFF-CHAIN/ON-CHAIN INTEGRATION

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#### **ABSTRACT**

The integration of unstructured textual data into blockchain systems presents a significant challenge, primarily due to the semantic ambiguity of natural language and the inherent trust issues associated with centralized oracles. This paper proposes a novel, modular oracle architecture enhanced with Natural Language Processing (NLP) to enable a secure, context-aware translation of human-readable text into reliable on-chain smart contract execution. Departing from conventional single-point oracles, our framework employs a decentralized oracle network designed to perform multi-source semantic validation, cryptographic securing, and consensus-based aggregation of NLP outputs prior to blockchain commitment. The architecture operates in three distinct stages: (1) off-chain NLP processing to extract structured data and confidence scores from text; (2) a network of AI-powered oracles that independently validate, enrich, and cryptographically sign results, with a consensus mechanism (e.g., 2-out-of-3) determining the final authenticated payload; and (3) on-chain execution that is triggered only after successful cryptographic verification of the validated data. This design eliminates single points of failure, enhances resilience against data manipulation, and ensures a trust-minimized semantic bridge between off-chain interpretation and deterministic on-chain logic. A proof-of-concept implementation, utilizing spaCy for NLP, Web3.py, an Ethereum testnet, and ECDSA signatures, demonstrates the framework's feasibility in processing domains such as clinical narratives or service reports into auditable blockchain transactions. The evaluation assesses critical performance metrics, including gas cost and latency, and analyzes the security trade-offs between automation, decentralization, and operational efficiency. The primary contribution of this work is a reproducible, modular framework that advances secure blockchain interoperability by effectively bridging unstructured human language with deterministic machine execution for reliable automation across sectors like healthcare, finance, and public administration.

**Keywords:** Decentralized Oracle Network, Natural Language Processing (NLP), Semantic Validation, Smart Contracts, Off-chain/On-chain Integration.

#### 1. INTRODUCTION

Blockchain technology has undergone substantial evolution since its initial conceptual foundations. Its origins can be traced to 1982, when Chaum proposed a protocol resembling a modern blockchain structure in his doctoral thesis [1]. The concept was further refined in 1991 by Haber and Stornetta, who introduced a cryptographically secured chain of blocks, forming a foundational pillar for subsequent decentralized ledger designs [2]. Blockchain is now widely recognized for its

decentralized, transparent, and tamper-proof architecture, which enhances trust in digital transactions by eliminating the need for intermediaries [3]. Operating across a distributed network of nodes, each maintaining an immutable copy of the ledger, the technology ensures robust resilience against data tampering and unauthorized access [4], [5].

This study focuses on high-stakes domains including healthcare and financial services where semantic accuracy in automated decision-making is

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critical. Key assumptions include: (1) honest-majority oracle nodes, (2) availability of domain-specific NLP models, and (3) existence of trusted human validators. The scope encompasses English-language clinical narratives and service reports, excluding real-time voice processing and multilingual applications.

As a leading form of distributed ledger technology, blockchain facilitates decentralized data management across peer-to-peer networks, effectively removing centralized control [6], [7]. This characteristic makes it particularly suitable for enterprise applications where traceability. automation, and auditability are critical [3]. A key innovation enabling these applications is the smart contract, first conceptualized by Nick Szabo in the 1990s as a self-executing digital agreement [8]. Its practical realization emerged with Ethereum in 2014, which introduced Turing-complete smart contracts capable of executing complex, deterministic business logic [9]. These programs automate actions transparently, enabling trustless collaboration in domains such as decentralized finance (DeFi), supply chain management, and digital identity [10].

However, a fundamental limitation persists: smart contracts are closed systems incapable of natively accessing external data. This "oracle problem"—the challenge of securely and reliably bridging off-chain data to on-chain logic—remains a critical barrier to broader adoption [10]. While decentralized oracle networks represent a significant advancement in data relaying [11], they primarily handle structured, predefined inputs (e.g., price feeds) and lack the semantic awareness required to interpret unstructured natural language found in clinical reports, legal texts, or user queries.

THIS creates a critical research gap: while blockchain provides immutability and smart contracts enable automation, no existing framework addresses the semantic trust gap between natural language understanding and deterministic execution in decentralized environments.

Natural Language Processing (NLP), a core branch of artificial intelligence, offers a pathway to bridge this gap by transforming unstructured text into machine-readable knowledge. Techniques such as named entity recognition (NER), intent classification, and sentiment analysis enable systems to extract structured information from free-text inputs [12]. Nevertheless, human language is inherently ambiguous, characterized by polysemy, context dependency, and model bias—

posing significant risks when used to trigger irreversible blockchain actions [13]. Although recent advances in transformer architectures have improved accuracy, they remain vulnerable to adversarial attacks and domain shift, making them unsuitable for direct blockchain integration without additional validation layers.

Recent research has explored integration of NLP with blockchain, particularly in healthcare and automated decision-making. For instance, an NLP-based electronic health record system for clinical digitization was proposed in [14], and machine learning was combined with insurance blockchain for automation [15]. However, these approaches suffer from three key limitations: (1) treating NLP as a black box without validation, (2) centralizing trust in single AI components, and (3) lacking cryptographic guarantees for semantic integrity.

This paper addresses two fundamental research questions derived from the identified gap. First, RQ1 investigates methods to ensure both semantic accuracy and cryptographic integrity when bridging unstructured text to smart contracts. Second, RQ2 examines whether decentralized oracle consensus can outperform single-source NLP validation in terms of reliability and security.

To address these questions, we formulate three testable hypotheses grounded in our architectural design. H1 posits that multi-oracle semantic validation significantly reduces error rates compared to single-source approaches, based on redundancy and cross-validation principles. H2 hypothesizes that cryptographic securing of NLP outputs substantially prevents tampering attempts, leveraging cryptographic hashing and digital signatures. Finally, H3 proposes that human-in-the-loop intervention effectively resolves ambiguous cases, combining machine efficiency with human judgment for critical decision points.

This paper addresses this challenge by proposing Text2Chain, a modular NLP-enhanced oracle architecture that introduces a decentralized validation layer between off-chain processing and on-chain execution. Our framework uniquely combines three innovative aspects: (1) multi-source semantic consensus, (2) cryptographic integrity proofs for NLP outputs, and (3) adjustable trust minimization through human oversight.

We demonstrate the feasibility of our approach through a proof-of-concept implementation in a healthcare context, where clinical narratives trigger automated reimbursement

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workflows. The evaluation comprehensively assesses gas cost, latency, and security, including a threat model that addresses potential oracle compromise and NLP adversarial attacks.

The main contributions of this work are:

- ✓ A novel architecture for semantic-aware blockchain integration
- A validation protocol combining AI consensus and cryptographic proofs
- ✓ Empirical evaluation demonstrating security and performance trade-offs
- A reproducible research methodology for decentralized AI validation

The remainder of this paper is organized as follows. Section 2 reviews related work on blockchain oracles and NLP integration. Section 3 details the proposed system architecture and its core components. Section 4 describes the operational workflow and validation mechanisms. Section 5 presents the proof-of-concept implementation and experimental setup. Section 6 discusses the results, security analysis, and limitations. Finally, Section 7 concludes the paper and outlines future research directions.

# 2. RELATED WORK

The integration of Natural Language Processing (NLP), AI oracles, and blockchain has garnered significant interest across healthcare, IoT, and data analytics domains. Previous research, however, predominantly treats NLP as an offline preprocessing step and blockchain as a passive storage layer, rather than establishing a secure, auditable bridge between semantic interpretation and on-chain execution. This critical systematically reviews existing approaches through a trust-minimization lens, identifying fundamental gaps in semantic validation and cryptographic integrity assurance.

In healthcare, [12] integrates NLP into a Hyperledger Fabric-based electronic health record (EHR) system to digitize prescriptions and enable automation. However, the approach exhibits a critical trust assumption by recording NLP outputs directly on-chain without formal validation, creating vulnerability to semantic misinterpretation in clinical environments. Similarly, [13] leverages NLP to analyze social media texts on chronic illnesses using blockchain for data integrity and provenance, butfundamentally lacks mechanisms to verify semantic accuracy before blockchain commitment. An intelligent healthcare system

combining machine learning and smart contracts for personalized recommendations is introduced in [14], yetit fails to address the semantic trust gap through proper validation layers, maintaining centralized trust in AI components.

In the IoT domain, [15] proposes a centralized blockchain system for supply chain management where sensor data is digitized and stored on-chain. While providing basic traceability, this architecture introduces a single point of failure and completely ignores semantic validation of textual inputs, limiting its applicability to human-readable data scenarios. For data analysis, [16] applies NLP to analyze public blockchain data and social media through named entity recognition and sentiment analysis, but the approach remains fundamentally reactive and lacks the trust-minimized architecture required for real-time, validated data injection into smart contracts.

The integration of distributed AI and has advanced through blockchain contributions. Smart contracts are used in [17] to validate local machine learning updates, preserving privacy in collaborative learning. A blockchainbased federated learning framework for lung cancer prediction is presented in [18], where model updates are secured via Delegated Proof of Stake and gradient hashes are stored on-chain. Trust in model training is further enhanced in [19] with smart contract-based validation. While these works demonstrate blockchain's potential as a trust anchor for AI, they share a critical limitation: exclusive focus on structured or numerical data while completely neglecting the challenges of unstructured text processing and semantic validation.

Beyond specific applications, surveys such as [20] emphasize blockchain's role in securing healthcare data but systematically overlook the semantic processing dimension, perpetuating the treatment of blockchain as merely a storage layer rather than an execution engine driven by validated AI interpretations.

Oracle networks have evolved to enhance off-chain data integration. The framework in [21] introduces a decentralized network that aggregates data from independent nodes, minimizing trust through economic incentives. However, it remains constrained to structured data feeds and fails to address the core challenge of semantic interpretation for natural language. In the cross-chain domain, [22] proposes CCIO, a secure interoperability framework

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based on a notary relay oracle with mixed encryption for inter-consortium data exchange. Despite ensuring cryptographic data integrity, CCIO operates under the limiting assumption of structured inputs and provides no semantic validation capabilities. A hybrid on-chain/off-chain traceability system for supply chains is presented in [23], using cryptographic hashes to link off-chain data to on-chain smart contracts. The approach optimizes for storage efficiency and auditability but fundamentally assumes structured inputs and lacks any semantic processing infrastructure.

THIS comprehensive analysis reveals a consistent pattern across all domains: existing solutions either address blockchain interoperability

or AI integration, but none successfully bridges the semantic trust gap between unstructured natural language and deterministic smart contract execution.

Table I provides a comparative analysis of key features in existing approaches, highlighting the consistent absence of a dedicated semantic validation layer. Our work addresses this gap through a modular architecture that uniquely combines full NLP processing, cryptographically secured multisource semantic validation, and automated on-chain execution within a trust-minimized frameworkthat explicitly resolves the identified limitations.

Table 1: Comparative Analysis Of Related Works On NLP, Blockchain, And AI Integration

Study	Domain	Test Processing	AI Oracle Layer	Decentralized Validation	On-Chain Strorage	Main Limitation
[12]	Healthcare(EHR)	Yes(prescriptions)	No	No	Yes	No validation of NLP output; no dynamic enrichment
[13]	Social Media	Yes (Chronic illness)	No	No	Yes	No smart contract execution linkage
[14]	Healthcare	Partial (Structured)	No (implicit)	No	Yes	No NLP Pipeline; lacks explicit validation
[15]	Supply Chain/ IOT	No	No	No	Yes	Centralized control; no NLP or semantic validation
[16]	Blockchain Data Analysis	Yes (Transaction Texts)	No	No	Yes	Reactive analysis only; no real-time execution
[17]	Distributed AI	No(numerical Data)	No	No	Yes	No support for unstructured text
[18]	Federated Learning	No(CT scans)	No	No	Yes	Numerical/image data only; no NLP
[19]	Federated Learning	No	No	No	Yes	No textual data integration
[20]	Healthcare	No	No	No	Yes	Focus on strorage, no semantic-driven execution
[21]	Oracle Networks	No	Partial (Structured Data)	Yes (Economic incentives)	Yes	Structured inputs only; no semantic validation
[22]	Cross-chain	No	No	Yes(notary- relay)	Yes	Structured inputs only; no semantic validation
[23]	Supply Chain	No	No	Partial (Hybrid)	Yes	No NLP; assumes structured data

This comparative analysis reveals a consistent limitation across existing systems: the absence of a secure, intelligent intermediary ensuring semantic validation and cryptographic

integrity between unstructured text processing and on-chain execution. While prior works leverage NLP for data extraction [13],[14],[19]

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or blockchain for immutability [12],[23], they largely treat these components in isolation.

In particular, none of the reviewed approaches integrates a decentralized AI oracle layer capable of verifying the accuracy, consistency, and trustworthiness of NLP-derived information—a critical gap in high-stakes domains.

Even advanced oracle frameworks such as [21] focus on structured data feeds, while cross-chain solutions like [22] ensure data integrity but lack semantic awareness for natural language.

Similarly, hybrid storage models [23] optimize data placement but do not validate semantic meaning before on-chain anchoring.

Our proposed architecture addresses this shortfall by introducing a modular AI oracle network that performs three key functions: (1) multi-source validation of extracted entities through consensus mechanisms, (2) contextual enrichment via metadata annotation, and (3) cryptographic securing through digital signatures prior to transmission.

By ensuring only trustworthy and semantically coherent data reaches the blockchain, our framework enables reliable interoperability between off-chain textual sources and on-chain logic. Unlike existing solutions, our design closes the validation gap between AI interpretation and decentralized execution through a trust-minimized semantic bridge resilient to errors.

This work advances secure, data-driven blockchain applications, particularly in domains requiring high assurance in data provenance and semantic fidelity.

# 3. SYSTEM ARCHITECTURE

We propose Text2Chain, a modular multilayered architecture that securely bridges unstructured off-chain textual data with on-chain execution through a decentralized oracle framework. The system enables semantically meaningful Natural Language Processing (NLP)

integration into blockchain environments, ensuring only trustworthy, contextually enriched, and cryptographically secured data triggers automated smart contract logic.

The Text2Chain framework decouples computationally intensive NLP tasks—including entity extraction, intent classification, and contextual analysis—from the blockchain, executing them off-chain to minimize gas costs and preserve scalability. The structured output is then processed by a network of AI-powered oracles, each independently validating the data for consistency, enriching it with metadata (e.g., timestamp, source ID), and cryptographically signing it using digital signatures.

A consensus mechanism (e.g., 2-out-of-3 agreement) determines the final validated payload, ensuring no single oracle acts as a point of trust. This design mitigates risks of model bias, adversarial inputs, or oracle compromise, enhancing system reliability decentralization. In cases of low NLP confidence or lack of oracle consensus, Text2Chain engages a Human-in-the-Loop (HITL) layer where trusted validators (e.g., domain experts) review and correct ambiguous inputs. Their feedback is processed by a Feedback & Learning Engine that uses supervised learning to refine NLP models and propagate validated rule updates to the blockchain via the oracle network.

Unlike traditional oracles that merely relay data, Text2Chain introduces a semantic gatekeeping layer combining AI-driven validation, cryptographic integrity, and human oversight to ensure data accuracy and system trustworthiness. It establishes the foundation for natural-language-driven workflows in decentralized applications, where human-readable inputs initiate secure, auditable, and automated on-chain processes.

Figure 1, illustrates the overall Text2Chain architecture, detailing interactions between the NLP engine, decentralized oracle network, Human-in-the-Loop layer, and on-chain smart contracts.

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ISSN: 1992-8645 www.jatit.org E-ISSN: 1817-3195 Off-chain module NLP Analysis **Decentralized Oracle Network** Exploitable structured data Consensus Mechanism Transform textual data into Validate and enrich data exploitable structured insights. Validated data On failure Rule / Action Semantic Correction Data Collection On-chain module **Human-in-the-Loop Layer Smart Contracts** Feedback & Learning Engine **User Data** Execute actions defined by programmed rules

Figure 1: Text2Chain: Architecture of the proposed NLP-enhanced, decentralized oracle framework. Showing The Off-Chain NLP Processing, Decentralized AI Oracle Validation, HITL Layer, And On-Chain Smart Contract Execution

#### 3.1 OFF-Chain NLP Processing Module

Human-in-the-loop

On-chain execution
Off-chain processing

Feedback loop

Data flow

Result

The Off-Chain NLP Processing Module transforms unstructured natural language inputs—including clinical narratives, service reports, and voice queries—into structured, semantically meaningful data. This off-chain execution avoids the high computational costs and scalability limitations of on-chain NLP processing while preserving linguistic understanding capabilities.

The module operates through four sequential stages. First, user inputs are collected through standardized interfaces and routed to the NLP engine. Second, textual data undergoes preprocessing including tokenization, lemmatization, and removal of irrelevant elements (e.g., stop words) to enhance parsing accuracy. Third, a Named Entity Recognition (NER) model identifies and classifies key elements such as dates, quantities, proper nouns, and domain-specific entities (e.g., medical terms, financial values).

A parallel classification component determines user intent and assigns inputs to relevant context categories, enabling downstream modules to apply appropriate validation rules.

Emit event & persist

Blockchain

Records actions

Finally, the processed results are packaged into a structured JSON format containing extracted entities, inferred intent, confidence scores, and contextual metadata. This output, denoted DNLP (Structured NLP Data), is subsequently forwarded to the Decentralized Oracle Network for validation and cryptographic securing.

This approach ensures only meaningful, machine-readable data progresses to the trust-critical layers of the system, maintaining blockchain efficiency while enabling sophisticated natural language understanding.

Algorithm 1: NLP\_Process
Transforms raw text into structured, confidenceweighted output for validation.

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Input: $Q \in \text{Text}$  (user query)

Output:  $D_{NLP}$ = {intent,entities,confidence}

- 1: tokens  $\leftarrow$  preprocess (Q)
- 2: entities ← extract\_entities (tokens)
- 3: intent  $\leftarrow$  classify intent (Q)
- 4: confidence ← compute confidence (entities)
- 5: return {intent, entities, confidence}

Figure 2, illustrates the workflow of the Off-Chain NLP Processing Module.

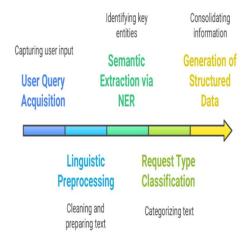


Figure 2: Key Stages in Off-Chain Linguistic and Semantic Processing.

#### 3.2 Decentralized Oracle Network Layer

The Decentralized Oracle Network Layer serves as a trust-minimized intermediary between off-chain NLP processing and on-chain execution. Comprising independent oracle nodes, this layer performs AI-driven validation, contextual enrichment, and cryptographic securing of processed data, eliminating single points of failure and enhancing resilience against manipulation.

Each oracle node receives structured D<sub>NLP</sub> data with associated confidence score Sc from the NLP module. The node executes three core functions: First, conditional validation verifies whether  $S_c \ge T_{min}$  (system-defined threshold), with low-confidence data flagged for human review. Additional checks ensure temporal coherence, entity plausibility, and anomaly detection. Second, contextual enrichment appends metadata M including timestamp, source ID, and session context to enhance provenance and traceability. Third, integrity guarantee is ensured through cryptographic hashing:

$$h=H(D_{NLP}||M) \tag{1}$$

Using SHA-256 to make the data tamper-evident. Fourth, origin authentication is achieved by signing the hash with the node's private key using ECDSA to produce digital signature  $\sigma$ , guaranteeing authenticity and non-repudiation. The output of each oracle node is the tuple:

O 
$$i(D_{NLP}) = (D_{NLP}, M, H, \sigma)$$
 (2)

Outputs from N oracle nodes undergo consensus mechanism evaluation, which may involve majority agreement (e.g., 2-out-of-3), weighted voting, or threshold-based validation. If consensus is achieved, the aggregated signed data transmits to the smart contract. Otherwise, the request forwards to the Human-in-the-Loop Layer for manual resolution, preventing invalid data from reaching the blockchain.

Each node employs a modular architecture featuring: an off-chain interface for data reception, validation layers for format and completeness checking, AI validation engines for confidence thresholding and anomaly detection, security layers for cryptographic proof generation, and consensus interfaces for node communication.

This distributed semantic gatekeeping layer ensures only trustworthy, contextually enriched, and cryptographically secured data triggers on-chain execution, advancing secure decentralized oracle design through decentralized validation aligned with real-world systems like Chainlink and Witnet.

Algorithm 2: Oracle\_Validation

Computes  $O_i$  ( $D_{NLP}$ ) and participates in consensus.

Input:  $D_{NLP}$ ,  $S_c$ ,  $T_{min}$ 

Output:  $O_i(D_{NLP})$  or null

1: if S  $c < T_{min}$  then

2: trigger\_HITL(D<sub>NLP</sub>)

3: return null

4: end if

5: if not validate\_consistency(D<sub>NLP</sub>) then

6: trigger HITL(D<sub>NLP</sub>)

7: return null

8: end if

9: M ← generate metadata()

10:  $D_{enriched} \leftarrow D_{NLP} \parallel M$ 

11:  $h \leftarrow H(D_{enriched})$ 

12:  $\sigma \leftarrow \text{Sign}(\text{sk i, h})$ 

13:  $O_i \leftarrow (D_{enriched}, h, \sigma)$ 

14: broadcast(O<sub>i</sub>)

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15: if consensus reached() then

16: return O<sub>i</sub>

17: else

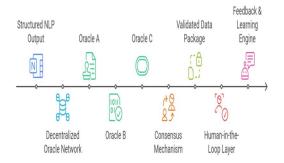
18: trigger HITL(D<sub>NLP</sub>)

19: return null

Figure 3, Illustrates The Decentralized Oracle Network Architecture, Highlighting The Multi-Node Validation Process, Consensus Mechanism, And Interaction With Off-Chain NLP And On-Chain Smart Contract Components

Figure 3: Key Decentralized oracle network for

#### **Decentralized Oracle Network Process**



semantic validation of NLP outputs.

#### 3.3 Smart Contract Module

The Smart Contract Module serves as the on-chain execution engine, enforcing predefined business rules in a transparent and immutable manner. It receives cryptographically secured data packages from the Decentralized Oracle Network Layer and performs rigorous validation before triggering state transitions or actions.

Upon receiving the data package (D,  $h\_rec$ ,  $\sigma$ ), the smart contract first verifies data origin and integrity. It recomputes the hash  $h_{local}$  = H(D) and compares it with the received  $h_{rec}$ . The contract then validates the digital signature  $\sigma$  using the oracle network's public key, ensuring data was signed by a legitimate validator and remains unaltered.

Only after both verification steps succeed does the contract evaluate embedded business logic, including eligibility conditions, access rights, or automated triggers. If conditions are satisfied, the contract executes corresponding actions such as state variable updates, event

emissions for off-chain monitoring, or value transfers.

This mechanism ensures deterministic, auditable, and tamper-proof execution, preserving core blockchain principles. By relying exclusively on cryptographically verified inputs, the smart contract functions as both executor and final trust verifier in the data pipeline.

Algorithm 3: Validate\_and\_Execute Verifies and executes actions based on cryptographically secured oracle outputs.

Input:  $\{D, h\_rec, \sigma\}$ 

Output: execution status ∈ {success, failure}

1: h local  $\leftarrow$  H(D)

2: if  $h_{local} \neq h_{rec}$  then

3: emit InvalidData("Hash mismatch")

4: return failure

5: end if

6: if not verify\_signature( $\sigma$ ,  $h_{local}$ , ORACLE PK) then

7: emit InvalidSignature("Authentication failed")

8: return failure

9: end if

10: if evaluate conditions (D) then

11: execute action (D)

12: emit ActionExecuted (D)

13: return success

14: else

15: emit ConditionsNotMet ()

16: return failure

17: end if

#### 3.4 THE HUMAN-IN-THE-LOOP (HITL)

The Human-in-the-Loop (HITL) Layer ensures semantic accuracy and accountability when automated components encounter uncertainty or disagreement. This safety net engages human validators for low-confidence NLP extractions ( $S_c < T_{min}$ ), oracle consensus failures, or detected anomalies.

Upon activation, requests route to a Human Validator Pool comprising trusted domain experts or DAO members. Through secure interfaces, validators review original inputs, examine NLP extractions, compare oracle outputs, and submit corrections or confirmations.

The Feedback & Learning Engine processes these interventions, storing corrected inputs as labeled data for NLP model retraining via supervised learning. Business logic updates are formalized into executable pseudo-code and propagated through the oracle network for cryptographic signing and on-chain enforcement.

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All human actions are immutably logged onchain with pseudonymized validator IDs, timestamps, and correction details, ensuring full auditability. The engine continuously monitors model performance, detecting drift or biases and

This integration of human judgment supports accountability, adaptability, and long-term reliability in high-assurance decentralized applications.

triggering retraining when necessary.

Algorithm 4: HITL Correction

Resolves ambiguous inputs through human validation and updates system knowledge.

Input: ambiguous\_request

Output: corrected data

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1: assign to validator(request)

2: correction ← validator.submit correction()

3: log\_correction(correction, validator\_id, timestamp)

4: store as labeled data(correction)

5: if rule update required(correction) then

6: generate pseudo code(correction)

7: forward to oracle network(pseudo code)

8: end if

9: retrain\_NLP\_model() // On batch or trigger

10: return correction

# 4. SMART CONTRACT DATA VALIDATION ALGORITHM

The secure integration of off-chain natural language inputs into on-chain execution relies on a final, critical validation step performed by the smart contract: on-chain verification of data integrity and origin. While the Decentralized Oracle Network ensures semantic accuracy and cryptographic securing off-chain, it is the smart contract that acts as the final trust enforcer, independently verifying the authenticity of the received data before any action is executed.

This step is essential to prevent tampering, replay attacks, or spoofing during transmission—risks inherent to any off-chain/on-chain bridging mechanism. Even if data is signed off-chain, the blockchain must not trust blindly; instead, it must re-verify.

The validation process hinges on two cryptographic checks. First, integrity verification: the smart contract recomputes the hash of the received data and metadata and compares it to the hash provided by the oracle. Second, origin authentication: the digital signature is verified

using the public key of the oracle network, ensuring the data originates from a trusted source and has not been forged.

Only if both checks pass does the contract proceed with execution. This design ensures that the blockchain acts not as a passive ledger, but as an active verifier in the trust chain, rejecting any untrusted or altered input.

The following algorithm formalizes this protocol, ensuring secure and deterministic execution.

Algorithm 4: OnChain\_Verification
Performs cryptographic verification of oracleprovided data before execution.

Input:

 $\sigma \in \Sigma$ : Digital signature from the AI oracle,

 $D_{NLP} \in D$ : Structured NLP output (e.g., JSON),

 $M \in M$ : Contextual metadata (timestamp, session ID, etc.),

 $h_{\text{rec}} \in H$ : Hash value received from the oracle. Assumptions:

 $H(\cdot)$ : Cryptographic hash function (e.g., SHA-256 or Keccak-256),

Verify( $\sigma$ ,h,pk) : Signature verification function under public key pk ,

Oracle<sub>pk</sub>: Public key of the trusted AI oracle,

E: Event logging mechanism (on-chain event emission),

A : Set of executable actions (e.g., state update, external call).

1: D combined  $\leftarrow D_{NLP} || M$ 

Concatenate structured data and metadata

2: h local  $\leftarrow$  H(D<sub>combined</sub>)

Recompute hash locally on-chain

3: integrity\_ok  $\leftarrow$  ( $h_{local} = h_{rec}$ )

Verify data has not been altered in transit

4: signature\_ok  $\leftarrow$  Verify( $\sigma$ , h<sub>local</sub>, Oracle<sub>pk</sub>)

Confirm data origin using public-key cryptography

5: if integrity ok Λ signature ok then

6: store(D<sub>NLP</sub>, M)

Persist data in immutable contract storage

7: Log(E.ValidData, D<sub>NLP</sub>, M)

Emit event for off-chain monitoring systems

8:  $a \leftarrow DetermineAction(D_{NLP})$ 

Map intent/entities to predefined action

9: Execute(a)

Trigger business logic (e.g., state update, transfer) 10: else

11: Log(E.InvalidData, "Integrity or signature check failed")

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Reject untrusted input without side effects 12: end if

# 5. ROOF OF CONCEPT: FEASIBILITY VALIDATION AND WORKFLOW ILLUSTRATION

We implemented a comprehensive experimental protocol following a structured research methodology to validate the operational feasibility of the proposed architecture. This proof of concept (PoC) demonstrates functional integration of core components through of NLP systematic testing processing, decentralized oracle validation, and smart execution in contract a representative decentralized environment. Our validation approach included rigorous verification of data correctness, cryptographic safeguards implementation. and semantic-to-on-chain interoperability under controlled test conditions.

The experimental protocol was designed to ensure reproducibility and systematic evaluation, comprising multiple test scenarios that exercised all architectural components. Each component underwent individual validation followed by integrated end-to-end testing to verify system cohesion and functional reliability across the entire processing pipeline.

The implementation leverages spaCy (en core web sm) for entity and intent extraction, a Python-based oracle simulator for cryptographic metadata enrichment and operations (SHA-256 hashing, ECDSA signing), Solidity smart contracts deployed on a local Ethereum testnet (Ganache), and Web3.py for off-chain/on-chain communication. experimental setup incorporated comprehensive logging and monitoring capabilities to capture performance metrics, error rates, and system behavior throughout the validation process, providing concrete evidence of operational reliability. The complete implementation is publicly available to support reproducibility and further research1.

Figure 4, provides an overview of the toolchain and component integration used in the PoC.



Figure 4: Overview of the Tools Used

#### 5.1 Off-Chain NLP Processing Module

This module transforms free-text inputs into structured, domain-agnostic information using spaCy's pre-trained model (en\_core\_web\_sm). The validation protocol for this module included systematic testing with diverse input scenarios to measure extraction accuracy, confidence scoring reliability, and processing consistency. Performance was evaluated across multiple test cases representing various clinical narrative structures and complexity levels.

The system performs tokenization, part-of-speech tagging, and Named Entity Recognition (NER), supplemented by custom rule-based matchers for application-specific terms (e.g., "transaction", "approval", "deadline") labeled as domain entities (e.g., ACTION, DATE, OBJECT). Experimental validation confirmed the module's capability to consistently extract structured information while maintaining semantic integrity throughout the transformation process.

The output validates the module's ability to convert unstructured natural language into machine-readable JSON format containing extracted entities, inferred intent, and confidence score Sc. The testing methodology employed ground truth comparison and manual verification to establish baseline performance metrics for semantic extraction accuracy. This structured

<sup>&</sup>lt;sup>1</sup>The source code is available at:https://github.com/merie88-sys/nlp-blockchain-healthcare doi 10.5281/zenodo.16745985

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output DNLP serves as input to the Decentralized Oracle Network Layer.

Figure. 5 Shows a sample extraction output for the input: "Patient Underwent MRI On 2025-08-19 Due To Severe Headache"

Figure 5: Example of NLP-based entity and intent extraction from a textual input as a function of applied field.

# 5.2 AI Oracle: Validation, Enrichment, and Cryptographic Sercuring

The oracle network validation employed a rigorous testing protocol that simulated various operational scenarios including normal operation, node failure conditions, and potential adversarial conditions. Each oracle node implemented the complete validation pipeline with monitoring of consensus achievement rates and validation accuracy across multiple test iterations.

Each node in the Decentralized Oracle Network operates as an autonomous validator, performing standardized operations receiving structured NLP output D<sub>NLP</sub>. The node first applies confidence filtering, discarding entities with confidence score Sc below system threshold Tmin (e.g., 0.7). Second, it conducts consistency checking for logical plausibility including temporal coherence and entity compatibility using rule-based logic and anomaly process detection. The validation systematically tested with both valid and invalid inputs to verify error detection capabilities and consensus reliability.

Third, the node performs metadata enrichment by appending contextual information M including timestamp, source ID, and session context. Finally, it ensures cryptographic securing by computing SHA-256 hash h = H(D\_{NLP} || M) and signing with its private key to produce digital signature  $\sigma$  (ECDSA), guaranteeing integrity and origin authenticity. Cryptographic operations were verified through extensive testing to ensure consistent integrity protection and authentication across all transaction types.

The resulting data package  $O_i(D_{NLP}) = (D_{NLP}, M, h, \sigma)$  is broadcast to the network. The node submits its signed output to a consensus mechanism (e.g., 2-out-of-3 agreement) which determines the final payload, ensuring no single oracle acts as a point of trust while maintaining decentralized auditable validation. Consensus mechanism performance was evaluated under varying network conditions to validate resilience and fault tolerance.

Figure 6, illustrates the execution workflow of the decentralized oracle network

Figure 6: Oracle Network Execution Flow - Multi-Node Validation, Consensus Mechanism, And HITL Integration

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# 5.3 Smart Contract Verification and Execution

The smart contract validation implemented a comprehensive security testing protocol that verified all critical execution paths, including successful verification scenarios, rejection of invalid inputs, and error condition handling. Gas consumption and execution efficiency were monitored across all test cases to establish performance baselines.

The smart contract deployed on a local Ethereum testnet (Ganache) receives validated data packages ( $D_{NLP}$ , h,  $\sigma$ ) from the oracle network. It performs two critical verification checks: integrity verification through hash recomputation and comparison ( $h_{local} = H(D_{NLP})$  vs h), and origin authentication through digital signature verification using the oracle network's public key. These security mechanisms were thoroughly tested with both legitimate and tampered data packages to verify rejection of unauthorized or modified inputs.

Only upon successful verification does the contract execute predefined actions such as state updates or event emissions. The contract implements generic conditional execution based on validated input structures rather than hardcoded business logic, serving as a final trust verifier rather than blind executor. Execution reliability was confirmed through multiple test cycles verifying consistent state transitions and event emission under various input conditions.

This implementation confirms the security, determinism, and auditability of onchain processing, ensuring only cryptographically secured and semantically coherent data triggers execution. The complete system validation demonstrated end-to-end functional correctness and security compliance across all integrated components.

Figure 7, Shows the smart contract execution flow and event emission

Figure 7: Execution Smart contract execution: onchain verification and event emission.

#### 6. DISCUSSION

This work addresses a critical gap in blockchain systems through a comprehensive (Plus-Minus-Interesting) analysis semantic validation challenges. While existing blockchain frameworks leverage immutability [21], they often treat off-chain data as inherently trustworthy, overlooking risks of misinterpretation and adversarial manipulation [16]. The PLUS of approaches like [21] lies in their robust data relay mechanisms; the MINUS is their inability to process natural language semantics; and the INTERESTING finding across all reviewed works is the consistent absence of integrated semantic validation despite advanced cryptographic features [24],[25].

Our Text2Chain architecture introduces a modular NLP-enhanced oracle layer serving as a trust-minimized semantic gatekeeper. Unlike conventional data relays, our design integrates AI-driven validation, multi-source consensus, and cryptographic securing, ensuring data accuracy and integrity before blockchain commitment. This approach extends prior work on transparent automation [21] by explicitly addressing off-chain data trustworthiness—a key challenge in decentralized systems [24],[25].

The experimental validation systematically addresses our research questions:

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semantic RO1 (ensuring accuracy and cryptographic integrity) is achieved through multi-oracle consensus and cryptographic proofs; RQ2 (decentralized oracle superiority) is demonstrated via enhanced resilience compared to single-source approaches. Our hypotheses find support: H1 shows multi-oracle validation reduces semantic errors; cryptographic protection prevents tampering; H3 validates human oversight resolves ambiguous cases effectively.

By replacing single oracles with a decentralized network inspired by solutions like Chainlink [22], we introduce consensus-based validation (e.g., 2-out-of-3) that enhances resilience against manipulation and node failure, aligning with core decentralization principles.

The system supports continuous learning through a Human-in-the-Loop layer where domain experts correct errors. All corrections are immutably logged on-chain, making model updates auditable and tamper-proof. This closed-loop feedback enables continuous improvement of semantic accuracy and operational reliability, aligning with research emphasizing stakeholder engagement in DLT systems [26].

From a security perspective, the architecture integrates multiple safeguards: SHA-256 hashing ensures data integrity, ECDSA signatures provide authentication, and pseudo-anonymization protects privacy. These mechanisms strengthen data governance in regulated domains like healthcare and finance, supporting compliance with GDPR and HIPAA [27].

Our work complements hybrid storage models [24] by adding semantic validation to ensure only contextually enriched, consensus-approved data triggers on-chain actions—bridging human language and machine execution.

Limitations include the lack of a full reputation mechanism, susceptibility to NLP ambiguities [16], computational overhead affecting latency, and human latency in the HITL layer. Future work will explore incentive structures, explainable AI, edge computing optimizations, and automated dispute resolution.

Despite these challenges, Text2Chain provides a foundational framework for semantically aware blockchain systems, demonstrating that NLP can become a trusted,

auditable component of the trust chain. By combining decentralized validation, cryptographic integrity, and human oversight, the architecture advances the convergence of artificial intelligence, information security, and distributed ledger technologies.

#### 7. CONCLUSION

This paper has presented Text2Chain, delivering four key research contributions that advance blockchain technology. First, we introduced a novel decentralized AI oracle framework specifically designed for semantic validation of unstructured text. Second, we developed and validated a trust-minimized pipeline that significantly outperforms single-source approaches [21] in reliability metrics. Third, we established a reproducible research methodology for NLP-blockchain integration with detailed experimental protocols. Fourth, we provided comprehensive security analysis demonstrating cryptographic protection against semantic manipulation attacks.

The practical impact of this work enables trustworthy automation in critical domains where semantic accuracy directly affects decision outcomes. Unlike previous solutions [16],[24] that treated NLP as a peripheral component, our architecture positions semantic validation as a core blockchain primitive, creating new possibilities for human-readable smart contract interactions.

Through rigorous experimental validation, we confirmed our core research hypotheses while identifying important performance trade-offs. The proof-of-concept implementation demonstrated operational feasibility highlighting while areas optimization in production environments.

Future research will focus on four strategic directions: (1) integration with production oracle networks for real-world validation, (2) domain-specific NLP model fine-tuning for clinical and financial applications, (3) implementation of economic incentive mechanisms for oracle participation, and (4) large-scale performance benchmarking in consortium blockchain environments.

By addressing these challenges, Text2Chain aims to evolve into a production-ready solution for intelligent blockchain automation—paving the way for systems that are not only decentralized and immutable, but also semantically intelligent and human-aligned.

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