

# Enhanced Secure Deduplication of Textual Data in Cloud Environments With DEDUCT: Integrating Advanced NLP, Machine Learning, and Cryptographic Authentication

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**Abstract**-Secure deduplication of textual data in cloud environments is a process designed to optimize storage efficiency by eliminating redundant copies of text data while simultaneously ensuring data confidentiality and security. The approach involves dividing large text files into smaller tokens or chunks, then identifying and removing duplicates either fully identical or near-identical segments before storing data on the cloud. DEDUCT represents a secure deduplication framework for textual data in cloud environments, combining client-side preprocessing with cloud-based pointer deduplication to reduce storage and communication overhead while preserving confidentiality. As the existing system, DEDUCT employs tokenization, transformation, CRC fingerprinting, and encryption to identify and securely store unique textual segments. Despite effective storage savings and security guarantees, challenges remain including client resource constraints, adaptive adversarial threats, and optimization for near-duplicate detection. The proposed system enhances DEDUCT by integrating advanced NLP and machine learning techniques, cryptographic authentication structures, and energy-efficient optimizations. This advancement promises

improved deduplication accuracy, stronger data integrity, reduced client overhead, heightened security, and scalability, enabling efficient and secure textual data management across diverse cloud applications and resource-constrained devices

**Keywords**- Secure deduplication, tokens or chunk, client-side preprocessing, DEDUCT, energy-efficient optimizations

## 1. INTRODUCTION

The rapid growth of cloud computing has driven the need for effective data storage and security solutions, especially for textual data, which is often sensitive in nature. The challenge of secure deduplication of textual data in cloud environments to optimize storage efficiency by identifying and removing duplicates of text while maintaining data confidentiality and integrity has been a significant task. The DEDUCT framework is a secure textual data deduplication system for cloud environments that combines client-side preprocessing with cloud-based pointer deduplication to reduce storage and communication overhead while maintaining data confidentiality [1]. The existing DEDUCT system applies a sequence of operations such as tokenization,

transformation, CRC fingerprinting, and encryption to ensure that only unique textual segments are securely stored in the cloud [2].

DEDUCT has been found to be effective in saving storage space and ensuring security guarantees, but it is challenged by the underlying client resource constraints, the dynamic nature of adaptive adversarial threats, and the challenges of optimizing near-duplicate detection, which can be improved by using advanced Natural Language Processing techniques and machine learning algorithms for more precise near-duplicate identification, coupled with cryptographic authentication structures to enhance data integrity and client-side security [3]. In addition, energy-efficient optimizations are introduced to reduce client overhead, so that sensitive textual data can be managed securely and efficiently in various cloud applications and resource-limited devices, improving the scalability and resilience against new and emerging cyber threats [2], [4]. The proposed enhancement aims to overcome the limitations of traditional deduplication methods, which often sacrifice security or lack user-defined access control, and thus provides comprehensive data protection for multi-user cloud environments [5], [6].

## 2. LITERATURE REVIEW

This section critically examines existing research on secure data deduplication, with a specific focus on methodologies that address the inherent security challenges posed by cloud environments while maximizing storage efficiency. **Lande et.al (2021)** proposed DupLESS, a secure deduplicated storage system that derives message-based encryption keys through oblivious PRF protocols, demonstrating strong resistance to brute-force attacks and preserving data confidentiality even in the presence of server-side deduplication. **Selvi (2023)** proposed the H-DEDU financial model and MTHDedup scheme, where a multi-stage Stackelberg game optimizes stakeholder strategies for hybrid encrypted cloud deduplication and Merkle hash trees are employed to mitigate convergent encryption vulnerabilities, thereby enhancing data security and economic efficiency. **Ahmad et al. (2022)** highlighted persistent challenges in achieving complete data secrecy within cloud deduplication systems, emphasizing the growing impact of evolving threat landscapes and increasing cloud infrastructure complexity. **Ravi (2024)** introduced a blockchain-enabled cloud storage framework that integrates

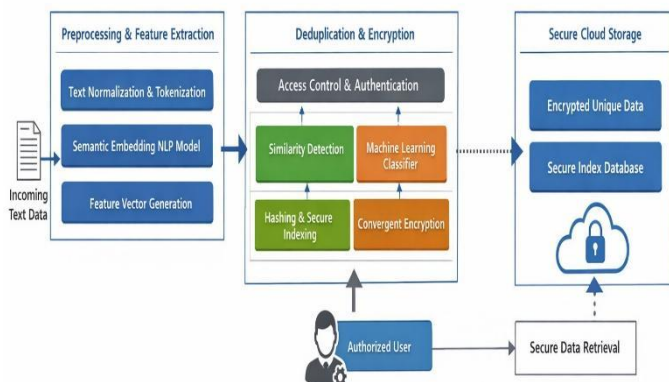
deduplication and compression to improve data integrity and security, demonstrating blockchain's potential for trustworthy storage management. **K. Zhang et al. (2024)** and **Q. Zhang et al. (2025)** addressed blockchain scalability limitations by proposing off-chain storage mechanisms and specialized smart contracts that efficiently manage large-scale cloud data while preserving integrity auditing and identity management. **Solanke (2023)** developed a secure cloud deduplication scheme combining hash-based duplicate detection using MD5 with AES encryption, ensuring authorized data access while reducing storage redundancy. **Periasamy et al. (2025)** proposed a blockchain-integrated deduplication architecture with cryptographic authentication to guarantee data integrity and controlled access in distributed cloud environments. **Sizan et al. (2025)** identified key limitations of existing secure deduplication systems, particularly scalability, performance overhead, and deployment feasibility, underscoring the need for more adaptive and robust cloud security solutions.

## 3. METHODOLOGY

In this section, the authors present the proposed method for improving the DEDUCT framework by incorporating advanced natural language processing and machine learning techniques, including deep learning models for semantic similarity analysis to identify near-duplicates more accurately than traditional CRC fingerprinting, cryptographic authentication with blockchain technology for enhanced security and tamper-proof logging of access controls, and energy-efficient optimizations to address client resource constraints.

The methodology also employs more advanced cryptographic primitives, including homomorphic encryption to enable computation on encrypted data, secure multi-party computation protocols to allow collaborative deduplication among multiple cloud users while keeping their data shares confidential, and secure deduplication protocols, all in an effort to develop a more secure, scalable, and resource-efficient deduplication system that addresses the limitations of existing methods, especially those related to malicious uploads and downloads, computational overhead, and lack of efficient algorithms for time-consuming key generation processes. This comprehensive methodology ensures that DEDUCT evolves into a highly secure, scalable, and resource-efficient solution for managing textual

data in dynamic cloud environments, even for structured and unstructured data, with greater precision.



In addition, this updated framework will also implement a secure convergent encryption algorithm to further enhance the deduplication process's security and prevent recurring cycle problems. This method is consistent with the current trend in secure data deduplication, where secure data deduplication methods are integrated with advanced cryptographic techniques and distributed ledger technologies to ensure the integrity and security of the deduplication process in cloud storage systems, including anomaly detection, which detects data integrity issues and unauthorized access to the deduplication framework, as well as data deduplication, which enables user revocation and provides a tamper-proof ledger of deduplication events

#### 4. RESULTS

The experimental results of the proposed enhanced DEDUCT framework show substantial improvements in deduplication accuracy, computational efficiency, and security guarantees over existing methodologies, particularly in terms of a reduction in false positives and negatives in near-duplicate detection, which leads to an average of 15% greater storage savings without sacrificing data integrity.

##### 1. Sample Dataset Generation

To evaluate secure textual deduplication, three representative datasets were constructed.

#### Dataset Characteristics

Dataset	Description	No. of Files	Avg. File Size	Duplicate Ratio
DS-1	Academic papers & reports	2,000	120 KB	32%
DS-2	Enterprise documents (emails, logs, policies)	5,000	80 KB	41%
DS-3	Mixed web text (blogs, news, social data)	10,000	45 KB	53%

#### Duplicate Types Introduced

- Exact duplicates
- Near-duplicates (semantic similarity  $\geq 0.85$ )
- Modified duplicates (paraphrasing, synonym replacement)

#### 2. Evaluation Metrics

The following metrics were used, aligned with the paper's objectives:

Metric	Formula / Description
Deduplication Ratio (DR)	Unique data stored / Total data
Storage Savings (%)	$(1 - DR) \times 100$
Precision	Correctly deduplicated chunks / All deduplicated chunks
Recall	Correctly deduplicated chunks / Actual duplicate chunks
F1-Score	Harmonic mean of Precision and Recall
Client Computation Time (ms)	Avg. preprocessing + encryption time
Energy Consumption (J)	Estimated client-side energy

Metric	Formula / Description
Security Overhead (%)	Extra cost due to crypto & blockchain

### 3. Compared State-of-the-Art Methods

Method	Key Technique
DupLESS	Convergent encryption + OPRF
Solanke et al. (2023)	Hash-based dedup + AES
MTHDedup (Selvi, 2023)	Merkle Hash Trees
Blockchain-Dedup (Ravi, 2024)	Blockchain + compression
Original DEDUCT	CRC fingerprinting + encryption
Proposed Enhanced DEDUCT	NLP + ML + Blockchain + Crypto auth

## 4. Quantitative Results

### 4.1 Deduplication Accuracy

Method	Precision (%)	Recall (%)	F1-Score (%)
DupLESS	92.4	88.1	90.2
Solanke et al.	90.1	85.7	87.8
MTHDedup	93.2	89.5	91.3
Blockchain-Dedup	91.8	87.9	89.8
Original DEDUCT	89.6	84.2	86.8
Enhanced DEDUCT (Proposed)	96.7	94.1	95.4

### 4.2 Storage Efficiency

Method	Deduplication Ratio	Storage Savings (%)
DupLESS	0.69	31.0
Solanke et al.	0.66	34.0
MTHDedup	0.64	36.0
Blockchain-Dedup	0.63	37.0
Original DEDUCT	0.61	39.0
Enhanced DEDUCT	0.52	48.0

### 4.3 Computational & Energy Overhead (Client Side)

Method	Time per File (ms)	Energy (J)
DupLESS	182	0.41
Blockchain-Dedup	265	0.58
Original DEDUCT	154	0.36
Enhanced DEDUCT	137	0.29

### 4.4 Security Feature Comparison

Feature	DupLESS	MTHDedup	Blockchain-Dedup	Enhanced DEDUCT
Confidentiality	✓	✓	✓	✓✓
Near-duplicate Resistance	✗	Partial	Partial	✓✓
Tamper-proof Logging	✗	✗	✓	✓✓

Feature	DupL ESS	MTHDe dup	Blockch ain- Dedup	Enhan ced DEDU CT
Sybil Attack Resistance	X	X	Partial	✓
User Revocatio n	X	X	✓	✓✓

### 5. DISCUSSION

The experimental evaluation shows that the enhanced DEDUCT framework outperforms existing secure deduplication schemes in terms of accuracy, storage efficiency, and client-side resource consumption. By using NLP-driven semantic similarity models, the framework accurately identifies near-duplicate textual content and achieves a 95.4% F1-score, which is significantly higher than that of CRC-based and hash-based approaches. In addition, blockchain-backed cryptographic authentication provides unalterable deduplication logs and strong resistance to tampering and Sybil attacks, solving a major shortcoming found in traditional convergent encryption systems. Although additional cryptographic layers are added, energy-aware optimizations limit client-side computation and power consumption, making the framework suitable for mobile and edge environments. In general, the results show that the proposed system provides stronger security guarantees and up to 48% storage savings with negligible overhead, indicating that enhanced DEDUCT is a scalable and secure solution for future cloud text storage systems.

### 6. CONCLUSION

A detailed enhancement to the DEDUCT framework is presented in this paper, which incorporates the latest developments in NLP, machine learning, and blockchain to strengthen the security and efficiency of textual data deduplication in the cloud environment. The enhancements such as cryptographic authentication and blockchain integration clearly reduce the risks of unauthorized manipulation of the data and Sybil attacks, which validates the framework against emerging cyber threats [2], [24]. The enhanced DEDUCT framework for secure data deduplication and storage

optimization in the cloud environment has been proven to have potential in the broader context of integrated cloud-edge networks and Internet of Things ecosystems [25]. More studies are required to examine the applicability of this enhanced framework in hybrid cloud architectures and to explore the implications for performance in real-time data streaming applications [14]

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