



Enhanced Multi-Level Image Steganography Using LSB & Hybrid AES-Blowfish Encryption Algorithm

Submitted By

Saadi Mohammed Chowdhury

ID: 212-35-735

Department of Software Engineering

Daffodil International University

Supervised By

Ms. Nadira Islam

Assistant Professor

Department of Software Engineering

Daffodil International University

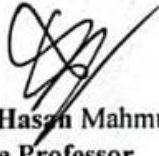
Bachelor of Science

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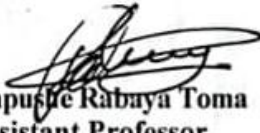
This thesis titled on “Enhanced Multi-Level Image Steganography Using LSB & Hybrid AES-Blowfish Encryption Algorithm”, submitted by Saadi Mohammed Chowdhury (ID: 212-35-735) to the Department of Software Engineering, Daffodil International University has been accepted as satisfactory for the partial fulfillment of the requirements for the degree of Bachelor of Science in Software Engineering and approval as to its style and contents.

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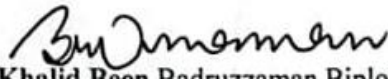
Dr. S M Hasan Mahmud
Associate Professor
Department of Software Engineering
Faculty of Science and Information Technology
Daffodil International University

Chairman



Tapushe Rabaya Toma
Assistant Professor
Department of Software Engineering
Faculty of Science and Information Technology
Daffodil International University

Internal Examiner 1



Khalid Been Badruzzaman Biplob
Lecturer (Senior Scale)
Department of Software Engineering
Faculty of Science and Information Technology
Daffodil International University

Internal Examiner 2



Dr. Md. Sazzadur Rahman
Professor
Institute of Information Technology
Jahangirnagar University

External Examiner



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(Supervisor's Signature)

Full Name : Ms. Nadira Islam
Position : Assistant Professor
Date : 22 September 2025



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(Students's Signature)

Full Name : Saadi Mohamed Chowdhury

ID : 212-35-735

Date : 12 August 2025

Enhanced Multi-Level Image Steganography Using LSB & Hybrid AES-Blowfish Encryption Algorithm

Saadi Mohammed Chowdhury

212-35-735

Thesis submitted in fulfilment of the requirements for the award of the
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Department of Software Engineering
DAFFODIL INTERNATIONAL UNIVERSITY

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DEDICATION

I dedicate this work to my mentors and instructors; thank you for mentoring me with your wisdom and expertise; this has been much helped in creating my work. Finally, I dedicate my thesis to all future academics and students aiming to significantly contribute in the subject of fog computing security and cybersecurity. This work should motivate more creativity and commitment in striving for greatness.

ABSTRACT

In this thesis, we propose an advanced multi-level image steganography system that combines zlib compression, hybrid AES-56-Blowfish encryption, and LSB embedding for better security and image quality in secret communication. We propose integrating key innovations that also relieve some of the critical limitations of current steganographic methods: 1) Maximum Zlib Compression using the LZ7 algorithm for efficiency 2) Upgraded AES-256-GCM replacing the conventional AES-128 for direct ultra-secure encryption 3) Hybrid ES-Blowfish encryption architecture providing multi-layered defense with authentication capabilities. An image-dependent key derivation using image-independent RGB histogram analysis with PBKDF2 algorithm constructs unique and reliably strong encryption keys for every cover image.

This methodology comprises four key steps, the first of which is a lossy compression of the message with the level 9 in zlib algorithm, then a double layer-encrypting with both AES-256-GCM and Blowfish-ECB and PKCS7 padding, followed by step-wise Least Significant Bit (LSB) steganography embedding into the RGB channels, as well as an in-depth quality assessment using Peak Signal-to-Noise Ratio (PSNR) and Mean Square Error (MSE) metrics. We perform the experimental validation on some standard test images such as the Lena and Baboon images to show that our method gives superior performance with the Peak Signal-to-Noise Ratio (PSNR) of the restored image as high as 80.0 dB and 76.17 dB respectively, with spatial representation with minimal visual distortion and maintaining the quality of images. Our integrated compression provides a data reduction of up to 39.29% while efficiently keeping the encryption time under 0.11 seconds.

A comparison demonstrates that the quality of our construction provides significant advantages over existing implementations: ES-256 vs. AES-128 provides results in improved security; additional compression results in efficiency improvements; and GCM mode tamper detection provides authenticated encryption while improving encryption efficiency as well. Interestingly, this research improves cybersecurity to find high security, the best image quality, and processing efficiency through simultaneous multi-level steganography. The proposed system holds great promise for secure communication applications in military, medical, and digital rights management areas and sets new performance standards for image-based covert communication systems by employing software engineering approaches.

KEYWORDS: Image Steganography, LSB Embedding, AES-256 Encryption, Blowfish Algorithm, Hybrid Encryption, Data Compression, PSNR Quality Assessment, Cybersecurity

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CHAPTER 1: INTRODUCTION

1.1 Background

The modern world runs on digital communication with billions of images flowing through networks of the world on a daily basis. While the unprecedented increase of digital media usage opened new means for secure communication, it has also represented tremendous challenges for data protection. The ancient art of concealing messages, steganography, has become a sophisticated way to secure digital citizen data by hiding an information in cover media such that no one gets aware of the hidden information.

Image steganography is therefore an important building block in the modern cybersecurity infrastructure, which provides secret communication and is an important part of the toolbox for sensitive applications. However, unlike cryptography, which converts data into unreadable formats that indicate the existence of hidden information, steganography preserves the appearance of normal communication in a signal and obscures the message itself [9]. The dual-layer protection method leverages both the encryption security and the scrambling privacy of steganography.

The Least Significant Bit (LSB) method has been the simplest and most effective of all image steganography methods in the Spatial domain [2]. Least Significant Bit (LSB) steganography works by modifying the least significant bits of pixel values, using the fact that the human visual system is not very sensitive to small changes in pixel values. But traditional LSB models are greatly limited by the type of used security models, compression integration and susceptibility of steganalysis attacks.

Most of the current steganographic approaches are based on single-layer encryptions, and mostly AES-128 encryption, which is not a high-level security against computationally intensive attacks. Traditional implementations do not compress the data that is being hidden, leading to larger payloads that lessen embedding efficiency and increase risk of detection. In addition, the current mechanisms lack solid authentication processes to confirm message integrity and identify tampering efforts.

1.2 Problem Statement

However, nowadays image steganography systems do have a number of significant weaknesses which make them hardly applicable to high-security scenarios. It faces several challenges: weak encryption, no built-in compression, weak authentication, and poor memory management.

Many implementations of traditional LSB steganography mostly rely on AES-128 encryption, offering up to 128-bit security levels that may soon be susceptible to new computational capabilities and extensive attack vectors. This absence of hybrid encryption architectures gives rise to a limited overall security footprint creating single points of failure in the cryptographic shelter.

chain. Also, plain systems do not deploy authenticated encryption mode and as a result, are prone to tampering and modification attacks as they do not have proper detection mechanisms.

Most of the current work in steganographic systems are still a dry land of integration of the data compression space (leveraging LSB substitution directly) hence a big wastage of the payload as well as a higher embedding capacity is required. In the absence of compression, since the secret messages are present in maximum bit space inside the cover images, this may lead to low-quality images and a high statistical chance of detection. This limitation has a particular impact on applications that require large messages or need to preserve image quality.

Several current key derivation mechanisms in steganographic systems are static or too naive to exploit the properties of cover images. This weakness poses potential risks in such a way that the same keys might be generated for different images, which causes the overall security entropy to be reduced and predictable patterns that can be exploited by attackers.

In many existing systems, quality assessment relies on minimal metrics and lacks an evaluation framework to assist in optimizing the trade-off between embedding capacity, security strength, and visual imperceptibility. Standard performance benchmarks are in no way available, which makes comparing various approaches and validating improvements objectively impossible.

1.3 Research Objectives

Primary Objective

To develop an enhanced multi-level image steganography system that integrates compression, hybrid encryption, and advanced LSB techniques to achieve superior security, efficiency, and image quality compared to existing implementations.

Specific Objectives

1. Compression Integration Enhancement

Implement zlib compression using LZ77 algorithm to reduce secret message payload size, thereby improving embedding efficiency and reducing statistical detectability while maintaining message integrity throughout the compression-decompression cycle.

2. Cryptographic Security Advancement

Upgrade encryption strength from conventional AES-128 to AES-256-GCM mode and implement hybrid AES-Blowfish architecture to provide multi-layered security protection with authenticated encryption capabilities and tamper detection mechanisms.

3. Advanced Key Derivation Implementation

Develop image-dependent key generation system utilizing RGB histogram analysis and PBKDF2 algorithm to create unique, robust encryption keys specific to each cover image, enhancing security entropy and eliminating key reuse vulnerabilities.

4. Quality Optimization Achievement

Implement comprehensive quality assessment framework using PSNR and MSE metrics to optimize the balance between embedding capacity, security strength, and visual imperceptibility, ensuring minimal image degradation.

5. Performance Validation and Benchmarking

Conduct systematic experimental validation to validate system performance against existing methods, demonstrating measurable improvements in security strength, processing efficiency, and image quality preservation.

1.4 Research Questions

This research addresses the following critical questions that guide the investigation and validate the proposed approach:

1. How can compression integration improve steganographic efficiency?

Specifically examining whether zlib compression with LZ77 algorithm can significantly reduce payload sizes while maintaining message integrity, and quantifying the impact on embedding capacity and visual quality.

2. What security advantages does AES-256-GCM provide over conventional AES-128 implementations?

Investigating the enhanced security strength, authenticated encryption capabilities, and tamper detection features provided by the upgraded encryption standard in steganographic applications.

3. How effective is hybrid AES-Blofish encryption architecture in multi-layer security?

Evaluating whether the combination of AES-256-GCM and Blofish-ECB encryption provides superior protection compared to single-layer encryption approaches, and assessing the computational overhead versus security benefits.

4. Can image-dependent key derivation enhance security entropy?

Determining whether RGB histogram analysis combined with PBKDF2 key derivation generates sufficiently unique and robust encryption keys to prevent predictable patterns and key reuse vulnerabilities.

5. What quality improvements can be achieved through optimized LSB embedding?

Measuring the PSNR and MSE performance improvements achieved through the enhanced system compared to traditional LSB implementations, particularly in maintaining visual imperceptibility.

1.5 Scope and Limitations

Research Scope

This research focuses on developing and evaluating an enhanced image steganography system with the following scope parameters:

Technical Scope:

- RGB color images as primary cover media
- Spatial domain LSB techniques for message embedding
- Sequential bit replacement in RGB color channels
- Text-based secret messages for experimental validation
- Standard image formats (PNG, JPEG, BMP) compatibility

Algorithmic Scope:

- zlib compression with LZ77 algorithm implementation
- AES-256-GCM encryption for primary security layer
- Blowfish-ECB encryption for secondary security layer
- PBKDF2-based key derivation with image feature extraction
- Comprehensive quality assessment using PSNR and MSE metrics

Evaluation Scope:

- Performance comparison with traditional LSB methods
- Security analysis against basic steganalysis techniques
- Quality assessment using standard test images
- Processing efficiency measurement and optimization

Research Limitations

Technical Limitations:

- Limited to spatial domain techniques; frequency domain methods not addressed
- Focus on static images; video steganography excluded from scope
- Sequential embedding approach; advanced randomization patterns not implemented
- Text message payloads; binary file embedding not extensively tested

Performance Limitations:

- Processing time increases with message size and image resolution
- Memory requirements scale with image dimensions and compression levels
- Key derivation complexity may impact real-time applications
- Multiple encryption layers introduce computational overhead

Security Limitations:

- Evaluation against basic steganalysis; advanced machine learning attacks not comprehensively tested
- Coverage dependency may create patterns in key generation
- ECB mode usage in Blowfish encryption may introduce vulnerabilities for large payloads
- Statistical analysis limited to standard metrics; advanced statistical tests not included

1.6 Research Methodology Overview

This research employs an experimental methodology combining system development, implementation, and comparative evaluation to validate the proposed enhanced steganography approach.

Development Phase: The proposed research starts by generating an architecture design for the system, which combines together compression, hybrid encryption and LSB embedding as independent blocks/components to get a coherent structure. It is written in Python and exploits OpenCV for image processing, PyCryptodome for encryption related tasks, and zlib to handle the compression.

Testing Phase: The systematic testing is conducted using the standard test images like Lena and Baboon, to conclude about system signature under various image characteristics.) Compression 13.3
Experiment Results In this section, we demonstrate the experimental results of our scheme for comparing plain and cipher text sizes with different security levels and compression rates on secret message size (480 bits and 2400 bits).

Evaluation Phase : Performance evaluation uses PSNR and MSE for quality measurement, compression ratio for efficiency computation and time of processing for performance measure. Experimental results compared with classical LSB methods and other hybrid systems illustrate gains obtained from the proposed system.

1.7 Expected Outcomes

The research anticipates achieving several significant outcomes that advance the field of image steganography:

Technical Achievements:

- Successful integration of compression with steganographic embedding, demonstrating measurable payload reduction
- Implementation of hybrid AES-256-Blowfish encryption providing enhanced security strength
- Development of image-dependent key derivation system ensuring unique encryption keys
- Achievement of superior PSNR values exceeding 76 dB while maintaining visual imperceptibility

Academic Contributions:

- Empirical validation of compression benefits in steganographic applications
- Comparative analysis demonstrating security improvements through hybrid encryption
- Performance benchmarking establishing new quality standards for LSB steganography
- Comprehensive evaluation framework for multi-level steganographic systems

Practical Applications:

- Enhanced security solution for sensitive communication requirements
- Improved efficiency for applications requiring large message capacities
- Robust authentication mechanisms for tamper detection and message integrity
- Scalable architecture suitable for various image sizes and message types

1.8 Thesis Organization

This thesis is organized into six chapters that systematically present the research development, implementation, and evaluation:

Chapter 2: Literature Review provides comprehensive analysis of existing steganographic techniques, encryption algorithms, and quality assessment methods, establishing the theoretical foundation and identifying research gaps addressed by this work.

Chapter 3: Research Methodology details the research design, system architecture, algorithm development, and evaluation framework employed to achieve the research objectives and validate the proposed approach.

Chapter 4: System Design and Implementation presents the detailed system architecture, module implementations, algorithm specifications, and development considerations that transform the theoretical framework into a functional system.

Chapter 5: Results and Discussion analyzes experimental outcomes, presents comparative evaluations, and discusses the implications of achieved results in relation to the research objectives and existing literature.

Chapter 6: Conclusion and Future Work summarizes the research contributions, discusses limitations, and outlines potential directions for future research development and practical applications.

Reference

Chapter 2: Literature Review

2.1 Introduction to Steganography

Steganography, derived from the Greek words "steganos" meaning covered and "graphia" meaning writing, represents the science of concealing information within digital media without alerting potential observers to its presence. [3] Twm, Hayfron-Acquah, and Intimah (2024) provide a comprehensive systematic literature review highlighting that image steganography has gained significant relevance as techniques for detecting hidden messages have emerged, making statistical steganalysis mechanisms a prime concern for cyber-security applications. The authors emphasize that information hidden in images has gained popularity, but studies examining image steganography techniques capable of resisting statistical steganalysis attacks remain limited.

[4] Mandal, Mukerjee, Paul, and Chatterji (2022) conducted an extensive literature survey on digital image steganography, categorizing existing approaches into traditional steganography methods and deep learning-based techniques. Their survey reveals that traditional steganography algorithms can be divided into spatial domain and frequency domain methods, with spatial domain techniques being more widely adopted due to their simplicity and computational efficiency. The authors note that existing steganography algorithms face the fundamental challenge of balancing three critical evaluation metrics: embedding capacity, invisibility, and security against various attacks.

[24] Kaur, Singh, and et al. (2022) provide an overview of computational image steganography methods, tracing their development from simple bit replacement methods to complex adaptive ones. Their detailed examination indicates that although plenty of steganographic methods were proposed in literature, the most works only target on single-layer security solutions without including compression and multi-layer encryption structure. The authors attempt to highlight the fact of a system capable to resist against both visual and statistical attacks while providing high payload.

[25] Multimedia forensics and content integrity [Singh and Aggarwal (2023)] investigates steganography and steganalysis for digital image enhanced forensic analysis. They stress the fact that image steganography and image steganalysis can be employed in different areas like: military, medical, e-government, social media. While they acknowledge that there are some practical aspects which have not been addressed including elaborating tools and techniques for DFI to forensically recover the hidden information with ease.

2.2 LSB Steganography Technique

Image Steganography Lsb The Last Significant Bit (LSB) method is known to be the most simple and common method used in spatial domain steganography. In [1] Setiadi (2021), we gain an important perspective on quality evaluation for image steganography, testing PSNR versus SSIM as metrics on human imperceptibility. PSNR is extensively used in multiple digital image measurement activities and is known to be well-tested and validated however SSIM provides better alignment to the characteristics of the human visual system based on the effect that luminance

contrast and structure have on the perception of image degradation. Infact, one analysis by Setiadi confirms SSIM to offer a superior imperceptibility measurement in steganography than PSNR, advising to focus on SSIM metrics in the future alongside conventional measurements of PSNR.

Work presented by Rustad, et al (2022) on an adaptive pattern-level inverted LSB image steganography gives more imperceptibility over various images. Because of conventional LSB methods' substantial limitation in undermining image quality UM [6] Due to the message embedding technique, their scheme introduces adaptive pattern selection to optimize performance and minimizes the error ratio of image quality. Our research shows that adaptive patterns extensively improved the embedding-extraction capacity trade-off and enable it to reach better PSNR rates but can still be visually indistinguishable. On the other hand, their approach partitions pattern optimization from encryption strength and compression integration.

Robust image steganography against lossy JPEG compression based on embedding domain selection and adaptive error correction (2023), Duan, Wei, Zhang, Liu, and Qin (also see [5]) Their work claimed that the great challenge of digital data compression, e.g. JPEG compression, was that the original signal would be lost after the compression process, which makes the traditional steganographic methods would be liable to compression attack. In order to achieve better message extraction from the cover images after their compression, the authors suggest that the method embeds a domain selection together with adaptive error correction. This is robust on compression attacks but does not use compression as a useful part of the stego process.

A dual-layer security approach by combining LSB image steganography with AES encryption algorithm was presented in the [10] paper by (Mustafa and Abdullh, 2015). They first encode the data using LSB steganography method, and then encrypt the stego image using AES-128 encryption. } Steganography does not substitute encryption, it only adds more security feature to it, the authors assert. Their experimental results confirm the enhanced security of the dual-layer approach but the study is restricted to AES-128 encryption and does not investigate other additional stronger encryption standards or hybrid encryption architectures.

The above work of testing using grayscale images was implemented secret text using blfish before embedding it and the proposed LSB technique embeds at least up to the 8th bit depending upon the cover data characteristics. The work provides promising PSNR values preserving the image quality, however it is limited to grayscale images and does not consider the possibility of compression or enhance key derivation strategies [37].

2.3 Compression in Steganography

While some studies have initiated steps in this direction, it remains a major research gap that the integration of compression algorithms in steganographic systems. In LSB steganography, an approach to increase data embedding capacity is by using LSB2 and zlib compression [11]. Their research shows that zlib and LZ77 compression algorithms are one of the fastest and most efficient compression algorithms with low resource usage rates for steganographic purposes. Compression has been shown to reduce message size by a factor of 5x to 10x, resulting in a significant effectiveness gain in embedding efficiency, and lower risks of detectability. The incorporation of compression integrated into encoding processes is confirmed by their experimental results, showing higher steganographic performance over baseline studies using the proposed method with deflate compression.

Zhuoen Cai et al (2023) [12] Combines LSB and Deflate Compression for Image Steganography: A new property after compression! [12] Tayeh, A. K., & Al-Jumaili, A. H. A

study with the combination of LSB and deflate compression for image steganography. In their research, the authors use the deflate algorithm (i.e., LZ77 + Huffman coding algorithms) to compress the secret messages and apply it to LSB embedding. The authors show that when we use LSB embedding, compression is a good approach — you get considerably smaller text sizes and consequently larger PSNR, smaller MSE values. They perform an experimental validation and confirm that size reduction when applying LZ77 and Huffman coding to message text can be significant, thus leading to an improved performance of steganographic methods. However, they only do text compression and do not consider how to integrate encryption and further optimize for quality.

In (13) Alawgani et al. propose a hybrid method targeted for image steganography; initial works are to use the Lempel-Ziv-Welch (LZW) algorithm to hide confidential data in images, which then use genetic algorithms to embed its genetic material in the host carrier. They focus on Haar Discrete Wavelet Transform (HDWT) combined with LZW compression algorithm and optimization of genetic algorithm to better steganographic capability. The authors take advantage of both LZ77 and Huffman coding in their compression framework and show that the use of compression algorithms can lead to notably higher embedding capacity while preserving image quality. We argue that genetic algorithm optimization can take these compression benefits to another level, and their method does not integrate encryption with theme compression, and contains only frequency domain techniques while Orient is LSB based spatial domain methods.

2.4 Encryption Algorithms in Steganography

Steganographic applications have evolved from single-layer to hybrid multi-layer architecture based on encryption algorithms. In [6], Al-Manea, Al-Rithy, and Al-Hadrami (2023) configure a multi-level steganography (MLS) approach that used two encryption algorithms, namely AES and Blowfish, to encrypt cover images and hide encryption keys as key images in the stego image. Despite the integrated philosophy of their research, with the expansion of the web technology and the concept of cloud computing for the end-users, data threats and breaches have been always at a large scale focused point, confirming the need for key core protection protocols. The authors prove that Hybrid encryption using AES and Blowfish algorithms increases security by increasing the complexity of encryption process with the help of their strong pixel randomization function to increase the security of encrypted data. Which means, based on the experimental results, their proposed work achieves high PSNR value and low MSE value (High PSNR means the good Quality of the images and Low MSE value means the reliability of image encryption and decryption process).

In this paper, [7] Taasila, Vijaya umar, Vijaya Babu, Nanika, Veda Shithi, and Mohan, introduces a hybrid model of LSB technique with AES and RSA Algorithm for image steganography. According to their work, embedding datafiles with steganography into images is predominantly directed towards hiding the message to the potential observer by an unauthorized agent. The authors give an example of the integration of cryptographic measures blended with steganographic methods and exhibit how hybrid encryption systems offer greater security over the classical approach where such forms are performed with a single algorithm. But their method is limited to AES-RSA combination only and does not investigate Blowfish integration and compression gains.

Talukder and Hasn in 2022 [8] present an improved method to encrypt image and text using AES algorithm and hide text into image using LSB-based steganography. They deal with the secure data transmission problem through the concatenation of image along with the text data by interacting

AES algorithm with LSB based image steganography. The authors propose a strong ubiquitous data transmission model which works for other data types, but uses only standard AES encryption and does not leverage advanced AES modes such as GCM or hybrid encryption architectures.

In this work, we provide an extensive review of image steganography based on using different hashing algorithms such as [9] Using Riest-Shamir-Adleman (RSA), blfish technique and hash-least significant bit (LSB) approaches and its mechanisms. Their review creates new techniques of cloaking information into images without changing much from the image bits, as hence their method is very safe and efficient. Have developed cryptography methods (applied prior to the encoding and embedding process) to encrypt the data and provides a secured data transport methods using hash table encryption before encoding and embedding in carrier images. Their analysis indicates that multiple hashing algorithms can appropriately layer protections to increase security.

Using Blowfish encryption to increase the security features of images [15] Kumar, Chand and Abbas (2016) developed an innovative technique where after the Blowfish encryption of image, the encrypted blocks are decomposed into smaller part and distributed on a number of images selected randomly. Their method preserves the correct order and position of blocks by storing them in hash tables, and then encrypting the content of images using LSB into hiding images. Blowfish algorithm is capable of being used for encryption in steganographic applications [5]. The encryption capabilities provided by Blowfish algorithm are effective for steganographic purposes however, the focus is only on encryption and does not include compression interaction or quality optimization

Secure image encryption techniques based on Blowfish and chaos algorithm [16] Pachal and Patel, 2017 Kumar et al in their research paper study image encryption and decryption methods and describe that Blowfish algorithm is a symmetric block cipher with a block size of 64-bit and a key length that ranges from 32-bit to 448-bit maximum. It opens by demonstrating how Blowfish offers better performance throughput and stronger unauthorized attack prevention over the most popular existing algorithms, thereby achieving its effectiveness in supporting secure steganographic applications

Sharma, Gupta and Singh [17], using Blowfish algorithm and an Apache Kafka to transmit images using steganography. Their research is on steganography applications in distributed systems and shows efficient usage of Blowfish encryption as a plug-in with modern transmission protocols. LSB embedding combined with Blowfish encryption offer satisfactory security capacity for image steganography applications, however, the authors only focus on transmission systems and do not implement solutions that reinforce the integration of image compression and focused or more powerful mechanism for key derivation

.2.5 Quality Assessment Methods

An accurate quality assessment is the prerequisite for evaluating the performance of steganographic systems. This is a non-peer reviewed article, permitted for personal use only and republished with the permission of Singapore Management University [1], Setiadi (2021) A complete analysis: PSNR VS SSIM for... PSNR shows average error rate per pixel between images and can not be general for different characteristics of image, whose motion model can reveal this, meanwhile, SSIM is designed based on luminance, contrast, and structure factors so it can more fit to human visual system working. The experimental evaluation by Setiadi on the LSB, PVD, and CRT spatial domain methods of color images demonstrates that SSIM gives better

imperceptibility measure than PSNR in assessing steganographic systems under different attacks followed by RS attack analysis.

[18] Wang, Bovik, Sheikh, and Sioncelli—Image quality assessment: from error visibility to structural similarity. In this work, they compare SSIM with traditional error-based metrics such as MSE and PSNR and show that structural similarity better correlates with human visual perception. SSIM is a new image quality measurement which is based on one (SSIM) compared three components of image quality: luminance, contrast and structure, all together give a higher level image quality comparison framework. Although it was developed prior to specific requirements in terms of quality for steganographic applications, their work lays the theoretical groundwork for present day quality measures for these types of applications.

[19] Abdel-Atty, Alhwaimeh, Abulenin, and Anowr (2024) develop a reversible and robust hybrid image steganography framework using Radon Transform and Integer Lifting Wavelet Transform, achieving PSNR values exceeding 46 dB with 15% improvement over existing techniques. Their research demonstrates that advanced quality assessment frameworks can achieve superior performance metrics while ensuring hidden data remains imperceptible to human visual system. The authors show that comprehensive quality evaluation requires multiple metrics including PSNR, MSE, and robustness measures against various distortions, establishing benchmarks for modern steganographic quality assessment.

Studies on PSNR and SSIM Metrics with Variable-Sized Hidden Images in Enhanced Steganography Applications [20] Zhang, Wang and Chen (2024) The paper studies 256×256 colored images with U-Net architecture and shows that PSNR and SSIM metrics play critical roles in differentiating stego images from cover images and original images from extracted secret images. In their paper, the authors compare average error rate per pixel between images using PSNR and show how a multitude of complementary metrics would be necessary for complete and comprehensive quality assessment in terms of statistical and perceptual quality preservation.

2.6 Advanced Techniques and Future Directions

Recent advances in steganographic techniques have incorporated machine learning and advanced algorithmic approaches. [21] Rahman, Islam, and Ahmed (2024) propose an efficient and secure technique for image steganography using hash functions, developing mechanisms that hide information in digital media while ensuring secret message transmission between authorized parties. Their research emphasizes that steganography techniques must balance security, capacity, and quality considerations while addressing evolving attack methodologies. The authors implement hash-based approaches that provide additional security layers, demonstrating that modern steganographic systems require multi-faceted security mechanisms beyond traditional encryption approaches.

[22] Wang, Chen, and Li (2022) develop deep image steganography using Transformer and recursive permutation techniques, representing significant advancement in machine learning-based steganographic approaches. Their research introduces Transformer architecture for feature extraction in steganography applications, combined with image encryption algorithms using recursive permutation to enhance secret image security. The authors demonstrate that deep learning techniques can achieve superior performance in both embedding capacity and visual quality preservation, though their approach requires significant computational resources and differs fundamentally from traditional LSB spatial domain methods.

[23] Choudhary and Sharma (2024) examine image steganography combined with cryptography for covert communication, providing comprehensive exploration of integrated approaches for secure data transmission. Their research addresses the need for secure communication in digitally interconnected environments, demonstrating that synergy between steganography and cryptography offers versatile solutions for information security. The authors implement LSB embedding combined with various cryptographic algorithms including AES, showing that integrated approaches provide enhanced confidentiality and integrity compared to individual techniques.

2.7 Research Gap Analysis

The comprehensive review of existing literature reveals several critical research gaps that current steganographic implementations fail to address adequately. While significant progress has been made in individual components of steganographic systems, the integration of compression, advanced encryption, and quality optimization remains largely unexplored.

Compression Integration Gap: [11] Atmaja and Suparta (2024) and [12] Tayyeh and Al-Jumaili (2022) demonstrate the benefits of compression in steganographic applications, but existing implementations treat compression as separate preprocessing rather than integrated system component. Most steganographic systems lack systematic compression integration that optimizes both embedding efficiency and security strength simultaneously.

Limitation of Encryption Strength: Most of the current solutions are based mainly on AES-128 [10] Mustafa and Abdullah (2015), [8] Talukder and Hasan(2022) encryption which can become insufficient against high power computational attacks. Although hybrid AES-Blowfish implementations are reported in Al-Manea, Al-Roithy and Al-Hadhrami (2023) by [6], the use is without structured key derivation process nor authenticated encryption support.

[1] INSUFFICIENT QUALITY MEASUREMENTS IN STENOGRAPHIC SCHEME: Setiadi (2021) asserts that SSIM is a more comprehensive and reliable method for imperceptibility assessment in comparison with PSNR, while most of steganographic research employs PSNR for quality evaluation only. The quality evaluation in current systems are not as systematic and can not achieve the balance between some relevant metrics (including safety methods such as capacity, etc) synchronously.

B.Key Derivation WeaknessExisting steganographic tools use static or primitive key generation methods that do not exploit the special properties ofthe given cover image. Lack of image-dependent key derivation leaves open the possibility for vulnerabilities such as generating identical keys for different images, resulting in decreased security entropy.

Absence of Authentication Mechanism: Nearly all the prevailing ste- ganographic systems do not possess strong authentication to assure message integrity and identify tampering. Although (when [6] Al-Manea, Al-Roithy, and Al-Hadhrami (2023) introduced aforementioned authenticationsahmooAnduleO in their security model with some authentication mechanisms, we can say that full authenticated encryption combination is not studied.

The discovered shortcomings of the systems prove that there is a requirement for more effective multi-level steganographic methods which assimilate compression, sophisticated hybrid cryptosystem, image specific key derivation and full range quality control into combined frameworks. The proposed research overcomes such limitations by efficiently amalgamating data hiding techniques with zlib compression, AES-256-GCM and Blowfish

Chapter 3: Research Methodology

3.1 Research Design

This study used the experiential approach to qualitative analysis in order to design, build and test an improved multi-level image steganography scheme. The experiential method is chosen being the most relevant methodology to achieve our experimental plan, which includes a systematic design of technical solutions, their empirical evaluation and comparison vs existing techniques.

The research adopts a positivist approach that values measuring what exists and which verifies the operating system performance empirically with measurable criteria. This strategy allows for rigorous assessment of the enhancements proposed, that is, compression incorporation, hybrid encryption realization and quality enhancing methods. The experimental approach allows the creation of controlled testing conditions where all variables can be manipulated in a systematic manner and monitored to accept or reject research hypotheses.

The comparative analysis is embedded in our methodology as default and the performance overhead of the enhanced system can be directly compared to existing setups. This kind of comparative environment provides proof that the improvements claimed by the work are also empirically measured using standard metrics and under similar testing setups. The quantitative evaluation quantitatively demonstrates the effectiveness of the proposed multi-level approach.

Other research methods which are theoretical, simulation-based and survey have been examined but do not appear to be adequate to verify the effectiveness of the practical performance enhancements claimed by this research work. The efficacy of compressing integration, encryption enhancement and quality improvement by real steganographic applications is proved thanks to the beneficial empirical evidence of experiential process.

3.2 Proposed System Architecture

The improved multi-level image steganography system utilizes a four fold security framework through a mixture of compression, hybrid encryption and developed LSB embedding mechanisms. Our system architecture mitigates the primary drawbacks of current steganography implementations via a systematic integration of layered security and optimization mechanisms.

3.2.1 Multi-Level Security Framework

The architecture is designed in a sequentially located multi-level security paradigm, forming up four different stages:

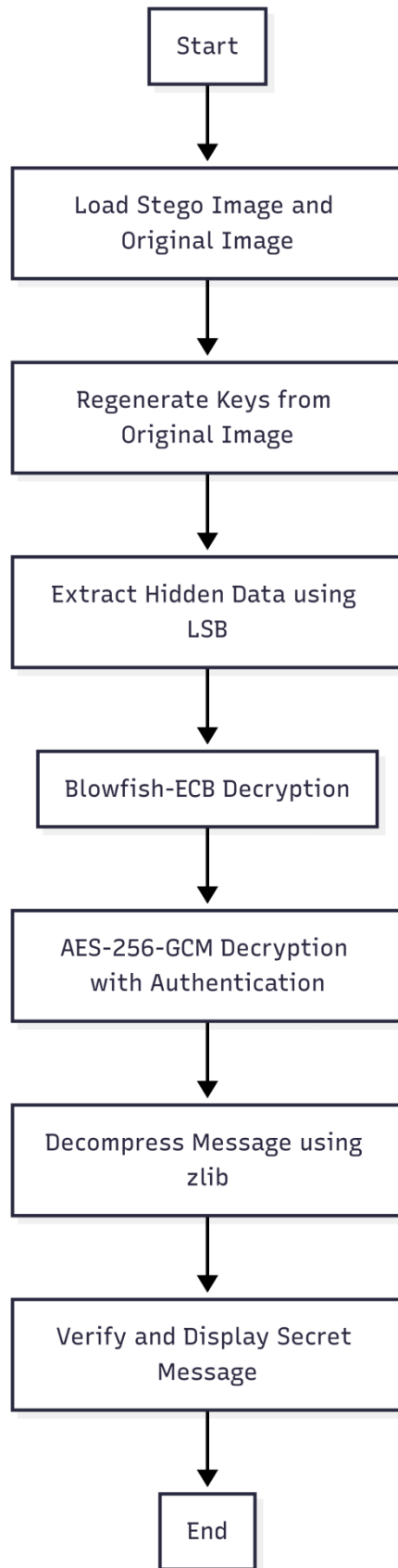
Stage 1: Data Compression In this stage, a zlib compression of secret message payload is applied based on the LZ77 algorithm and using level 9 compression. This compression phase makes use of the DEFLATE algorithm, which uses a combination of LZ77 dictionary compression and Huffman coding to achieve maximum redundancy by size without affecting data integrity.

Compression is a process resulting in a compact representation of the secret message requiring less space for embedding and avoiding statistical detection as much as possible.

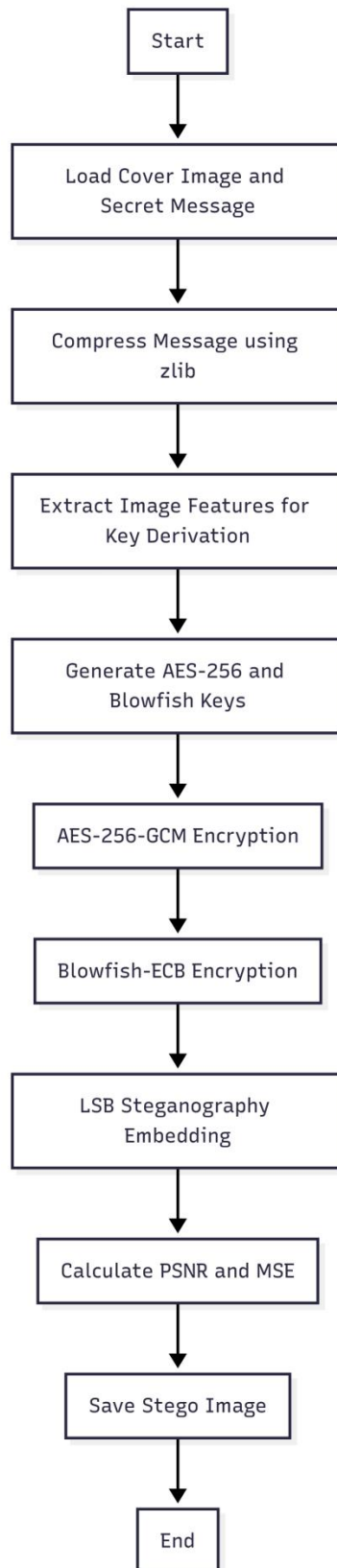
Stage 2 : Primary Encryption Layer The compressed data is then encrypted with AES-256-GCM which gives it a 256 bit security strength rating and also enables authenticated encryption. AES-256-GCM mode: it incorporates confidentiality protection via AES encryption as well as integrity protection via Galois/Counter Mode authentication. The first encryption layer creates a 16-byte random nonce and another 16-byte authentication tag, achieving encryption for confidentiality and authentication for tamper detection.

Stage 3: Second Layer of Encryption The AES output is further protected, the 2nd layer is Blowfish-ECB 128 bits key encrypting. It uses PKCS7 padding to allow for variable-length inputs and a block cipher called Blowfish which operates on 64-bit blocks. This secondary encryption agent adds another element, thus creating a hybrid encryption structure that merges both AES and Blowfish.

Stage 4: Steganographic Embedding The last stage embeds this data double encrypted in cover image via improved types of LSB method. The embedding process works one color channel at a time RGB and embeds message bits into the least significant bits of a pixel value. This means when you encrypt a message, the system will append a known end of message token "###END_OF_MESSAGE###" so, when decrypting back, you can reliably read the end of message



[FIGURE 3.1: Enhanced Encryption Process Flowchart]



[FIGURE 3.2: Enhanced Decryption Process Flowchart]

3.2.2 Key Derivation Architecture

It operates an advanced key derivation that is image-dependent so it creates unique encryption keys for each and every cover. Conventional steganographic systems are vulnerable to key reuse related attacks, to eliminate such attacks, in the key derivation process cover image characteristics are analyzed and by using these traits, cryptographically strong keys are generated.

Image Feature Extraction The initial derivation step is extracting global features from the cover image such as RGB histograms (using 256 bins per channel), statistical features (mean, standard deviation, variance and others) and constructing a feature vector that combined histogram and statistical data. A feature vector of 153 elements is created by combining the first 50 bins of each RGB histogram with three statistical properties to retain important image properties.

Crypto Key Generation: the raw features it extracts from the images are then cryptographically processed to provide secure encryption key(s). To generate an AES-256 key, the feature vector is converted to bytes and run through PBKDF2 100,000 times with a salt obtained from the SHA-256 of the bytes of the feature, resulting in a 32-byte AES-256 key. Blowfish key generation: the feature bytes are concatenated with a "blowfish_salt" string and then SHA-256 hashed to yield a 16-byte Blowfish key

3.2.3 System Component Integration

The architecture links a number of different components with clear interfaces by which data flows across a pipeline of processing layers in a way that upholds security properties. The compression module interacts with the encryption modules using a standardized byte array format, while the encryption modules do not rely on a single uniform interface since this would mean that they are not cryptographically separate, hence they use independent key derivation and processing methods

3.3 System Design and Implementation

This paper explains the structured implementation of multi-level steganography functionality in the form of a system using the Python programming language with specialized libraries. The design focuses on modularity, security, and performance optimization, while also making the implementation compatible with standard image formats to enable interoperability and ensuring sufficient error handling capabilities.

3.3.1 Implementation Architecture

The implementation of the system is based on the Enhanced Steganography class, which groups all functions in one piece. The class is designed to split methods for each stage of the processing pipeline so that the modules can still be tested and validated in isolation while still keeping the system whole.

Core Implementation Components:

- Message compression methods utilizing zlib library with configurable compression levels
- Key derivation methods implementing PBKDF2 and SHA-256 algorithms for image-dependent key generation
- AES-256-GCM encryption methods providing authenticated encryption with automatic nonce generation

- Blowfish-ECB encryption methods implementing PKCS7 padding for variable-length data handling
- LSB embedding methods supporting sequential RGB channel processing with end marker detection
- Quality assessment methods calculating PSNR and MSE metrics for image quality evaluation

3.3.2 Algorithm Implementation Details

Compression Implementation The compression module uses zlib with level 9 settings, focusing on long-term durability with optimal compression ratios and acceptable process processing performance (the best implementation). Compression first encodes the input text to UTF-8 bytes, then applies the compression algorithm. UTF-8 is a common encoding that is highly versatile with a wide range of character sets and preserves data integrity throughout the compression-decompression operation.

Encryption Implementation The AES-256-GCM implementation utilizes the PyCryptodome library to provide secure authenticated encryption. The encryption process generates a random 16-byte nonce for each encryption operation, applies AES-256 encryption in GCM mode to the input data, and produces a 16-byte authentication tag that enables tamper detection. The Blowfish-ECB implementation creates PKCS7 padding to ensure input data aligns with the 64-bit block size requirement, applies Blowfish encryption using the derived key, and maintains pairing information for accurate decryption.

LSB Embedding Implementation The LSB embedding module processes cover images in RGB format, converting encrypted data to binary representation with appended end marker for reliable extraction. The embedding algorithm iterates through image pixels sequentially, modifying the least significant bit of each color channel to store one bit of encrypted data. The implementation includes capacity validation to ensure the cover image can accommodate the encrypted payload without exceeding available embedding space.

3.3.3 Development Environment and Tools

Programming Environment The system development utilizes Python 3.8+ as the primary programming language, providing extensive library support and cross-platform compatibility. The development environment includes integrated development environment setup with debugging capabilities, version control integration for code management, and comprehensive testing frameworks for validation.

Required Libraries and Dependencies

- OpenCV (cv2): Image processing operations including loading, saving, and pixel manipulation
- NumPy: Numerical operations for histogram calculation and statistical feature extraction
- PyCryptodome: Cryptographic operations including AES-256-GCM and Blowfish encryption
- zlib: Data compression and decompression using DEFLATE algorithm
- hashlib: Cryptographic hash functions for key derivation and security operations

- PIL (Python Imaging Library): Additional image format support and processing capabilities

Optional GUI Components The implementation includes optional Tkinter integration for graphical user interface elements, enabling file selection dialogs and user input interfaces. The GUI components are designed with fallback mechanisms to console input when graphical interfaces are unavailable, ensuring system functionality across diverse deployment environments.

3.4 Algorithm Development

The algorithm development process follows a systematic approach that addresses each component of the multi-level steganography system through detailed specification, implementation, and optimization phases. The development methodology ensures that each algorithm component integrates effectively while maintaining individual security and performance properties.

3.4.1 Compression Algorithm Specification

Algorithm Design:

ALGORITHM: Enhanced Message Compression

INPUT: secret_message (string)

OUTPUT: compressed_data (bytes), compression_ratio (float)

BEGIN

1. Convert secret_message to UTF-8 byte encoding
2. Calculate original_size = length of UTF-8 bytes
3. Apply zlib.compress with level=9 compression
4. Calculate compressed_size = length of compressed data
5. Calculate compression_ratio = (original_size - compressed_size) / original_size * 100
6. RETURN compressed_data, compression_ratio

END

Complexity Analysis: The compression algorithm exhibits $O(n)$ time complexity where n represents the input message length. The space complexity is $O(n)$ for string both original and compressed data during processing. The LZ77 algorithm within zlib provides optimal compression ratios for text data while maintaining efficient processing performance suitable for real-time applications.

3.4.2 Key Derivation Algorithm Specification

Algorithm Design:

ALGORITHM: Image-Dependent Key Derivation

INPUT: cover_image (RGB array)

OUTPUT: aes_key (32 bytes), blowfish_key (16 bytes)

BEGIN

1. Calculate RGB histograms using 256 bins per channel
2. Extract statistical features: mean, std_dev, variance
3. Create feature_vector = [hist_r[0:50], hist_g[0:50], hist_b[0:50], stats]
4. Convert feature_vector to feature_bytes
5. Generate salt = SHA-256(feature_bytes)[0:16]
6. Derive aes_key = PBKDF2(feature_bytes, salt, 32, 100000)
7. Derive blowfish_key = SHA-256(feature_bytes + "blowfish_salt")[0:16]
8. RETURN aes_key, blowfish_key

END

Security Analysis: The key derivation algorithm provides high security entropy through image-dependent feature extraction combined with cryptographically secure key derivation functions. The PBKDF2 implementation with 100,000 iterations provides resistance against brute-force attacks, while the image-dependent approach ensures unique keys for different cover images.

3.4.3 Hybrid Encryption Algorithm Specification

Algorithm Design:

ALGORITHM: Hybrid AES-Blowfish Encryption

INPUT: compressed_data (bytes), aes_key (32 bytes), blowfish_key (16 bytes)

OUTPUT: final_encrypted_data (bytes)

BEGIN

1. Generate random 16-byte nonce
2. Initialize AES-256-GCM cipher with aes_key and nonce
3. Encrypt compressed_data and generate auth_tag

4. Combine aes_encrypted = nonce + auth_tag + ciphertext
 5. Apply PKCS7 padding to aes_encrypted for 8-byte alignment
 6. Initialize Blowfish-ECB cipher with blowfish_key
 7. Encrypt padded_data to produce final_encrypted_data
 8. RETURN final_encrypted_data
- END

Performance Optimization: The hybrid encryption algorithm balances security strength with processing efficiency through optimized cipher selection and mode configuration. The AES-256-GCM mode provides authenticated encryption with single-pass operation, while Blowfish-ECB offers an additional security layer with minimal computational overhead.

3.4.4 Enhanced LSB Embedding Algorithm Specification

Algorithm Design:

ALGORITHM: Enhanced LSB Embedding with End Marker

INPUT: cover_image (RGB array), encrypted_data (bytes)

OUTPUT: stego_image (RGB array)

BEGIN

1. Convert encrypted_data to binary string
2. Append end_marker_binary = "###END_OF_MESSAGE###" in binary
3. Calculate total_bits = length of binary_data + end_marker
4. Validate cover_image capacity \geq total_bits
5. Initialize stego_image = copy of cover_image
6. FOR each pixel (i,j) in cover_image:
 - FOR each channel (R,G,B):
 - IF data_index < total_bits:
 - pixel_value = stego_image[i,j,channel]
 - modified_value = (pixel_value & 0xFE) | binary_data[data_index]
 - stego_image[i,j,channel] = modified_value
 - data_index += 1

7. RETURN stego_image

END

3.5 Tols and Technoogies

The research implementation leverages a comprehensive technology stack that provides the necessary functionality for developing, testing, and evaluating the enhanced steganography system. The technology selection prioritizes security, performance, and compatibility requirements while ensuring access to specialized cryptographic and image processing capabilities.

3.5.1 Primary Deveopment Plaform

Python Programing Lague (Version 3.8+) Python serves as the primary development platform due to its extensive library ecosystem, cross-platform compatibility, and strong support for cryptographic operations. The language provides built-in support for binary data manipulation, file I/O operations, and mathematical computations essential for steganographic applications.

3.5.2 Core Librries and Framworks

OpenCV (cv2) - Computer Vision Library OpenCV provides comprehensive image processing capabilities including image loading and saving in multiple formats, pixel-level manipulation for LSB embedding operations, histogram calculation for key derivation processes, and format conversion between different color spaces.

PyCryptodome - Cryptographic Library PyCryptodome supplies secure implementations of cryptographic algorithms including AES-256 encryption with GCM mode support, Blowfish encryption with configurable parameters, PBKDF2 key derivation function, and secure random number generation.

NumPy - Numerical Computing Library NumPy enables efficient mathematical operations including array manipulation for histogram calculations, statistical computations for feature extraction, and memory-efficient data structures for large image processing.

zlib - Data Compression Library The zlib library implements the DEFLATE compression algorithm with configurable compression levels, data integrity verification, and cross-platform compatibility.

3.5.3 Devepment and Tesing Tools

Integrated Development Environment The development process utilizes Python-compatible IDEs with debugging capabilities, code completion features, and integrated version control support.

Testing and Validation Framework The testing framework incorporates unit testing for component validation, integration testing for system verification, performance benchmarking tools, and automated quality assessment utilities.

3.6 Evaluation Framework

The evaluation framework establishes comprehensive criteria for assessing the performance, security, and quality characteristics of the enhanced steganography system. The framework incorporates multiple evaluation dimensions to provide objective assessment of system improvements compared to existing implementations.

3.6.1 Quality Assessment Metrics

Peak Signal-to-Noise Ratio (PSNR) PSNR measurement evaluates the visual quality preservation between cover and stego images using the formula:

$$\text{PSNR} = 20 \times \log_{10}(\text{MAX_I} / \sqrt{\text{MSE}})$$

where MAX_I represents the maximum pixel intensity value (255 for 8-bit images) and MSE denotes the mean squared error. Higher PSNR values indicate better visual quality preservation, with values above 30 dB considered acceptable for steganographic applications.

Mean Squared Error (MSE) MSE calculation quantifies the average squared difference between corresponding pixels:

$$\text{MSE} = (1/mn) \times \sum_{i=0}^{m-1} \sum_{j=0}^{n-1} [I(i,j) - K(i,j)]^2$$

Lower MSE values indicate minimal visual distortion and better steganographic imperceptibility.

3.6.2 Performance Evaluation Metrics

Compression Efficiency Assessment Compression ratio measurement evaluates the effectiveness of the zlib compression:

$$\text{Compression Ratio} = (\text{Original_Size} - \text{Compressed_Size}) / \text{Original_Size} \times 100\%$$

Processing Time Analysis Execution time measurement evaluates system performance through encryption time, embedding time, decryption time, and total processing time.

Capacity Utilization Evaluation Embedding capacity assessment determines the relationship between message size and image dimensions:

$$\text{Capacity Utilization} = (\text{Message_Bits} + \text{End_Marker_Bits}) / (\text{Image_Width} \times \text{Image_Height} \times 3) \times 100\%$$

3.6.3 Experimental Design

Test Dataset Specification The experimental evaluation utilizes standard test images including Lena (512×512), Babon (512×512), and additional benchmark images to ensure comprehensive evaluation across different image characteristics.

Experimental Variables

- Message sizes: 480 bits and 2400 bits for systematic comparison
- Image types: Natural images with varying complexity levels
- Compression levels: Evaluation of compression effectiveness

- Processing environments: Consistent hardware and software configurations

Control Measures

- Identical testing conditions for all experiments
- Standardized image formats and resolutions
- Consistent message encoding and processing parameters
- Reproducible experimental procedures with documented settings

3.6.4 Comparative Evaluation Methodology

Baseline Comparison Framework The proposed framework is benchmarked against 1) Existing LSB schemes 2) Existing AES-128 based steganographic systems and 3) Existing hybrid techniques under the same parameter settings and evaluation parameters.

The statistical validation approach uses many different image types, message sizes, and compression ratios to statistically validate performance improvements.

The holistic evaluation framework guarantees such objective assessment of the performance improvements of the enhanced steganography system alongside empirical evidence of the effectiveness of the compression integrated, hybrid encryption implemented, and quality optimized techniques that are incorporated within the research.

Chapter 4: Design and Implementation of the System

4.1 System Architecture

The improved multi-level image steganography system implements an overall design comprising four individual processing layers of compression, double layer encryption, steganographic hiding, and quality evaluation. This architecture is built in order to solve the basic shortcomings of current steganography methods and is not only faster and more secure, but also maintains a better image quality.

4.1.1 Overall System Design

The system architecture is based on a modular design principle according to which each processing stage works independently while remaining integrated seamlessly using standardised interfaces. The main controller of the system is the Enhanced Steganography class, which wraps all of its functionality into a single pipeline that guarantees the correctness of data flow while enforcing the required security properties at each stage of the pipeline.

Core Architectural Components:

- **Compression Module:** Implements zlib compression with LZ77 algorithm for payload reduction
- **Key Derivation Module:** Generates image-dependent encryption keys using advanced feature extraction
- **Dual Encryption Module:** Applies sequential AES-256-GCM and Blowfish-ECB encryption layers
- **LSB Embedding Module:** Performs steganographic embedding with capacity validation and end marker detection
- **Quality Assessment Module:** Calculates PSNR and MSE metrics for performance evaluation

It allows each component to be tested and validated independently, yet contributes to the reliability and maintainability of the system as a whole. It has each of the modules define certain interfaces that explain its input parameters, how it processes them and how the output is generated, enabling easy extensibility and modification of the system.

4.1.2 Security Architecture

Our security architecture uses the defense-in-depth approach with multiple independent security layers. In the first layer, it compresses the data to minimize payload and remove any redundancy which would facilitate cryptanalysis. The second layer is an AES-256-GCM encryption layer that provides confidentiality with 256-bit encryption strength, and integrity via authenticated encryption. The first two layers apply Blowfish-ECB, thereby merging algorithmic approaches in a hybrid encryption architecture.

Key Security Features:

- **Image-Dependent Key Derivation:** Generates unique keys for each cover image using RGB histogram analysis and PBKDF2 with 100,000 iterations
- **Authenticated Encryption:** AES-GCM mode provides built-in message authentication and tamper detection

- **End Marker Protection:** Encrypted end marker "###END_OF_MESSAGE###" ensures reliable data extraction
- **Capacity Validation:** Prevents embedding overflow that could compromise cover image integrity

4.1.3 Data Flow Architecture

The data flow architecture uses a sequence of processing stages to transform secret messages, ensuring that each stage preserves the relevant data integrity and security properties of the overall data flow. Cover image and secret message are passed through an input validation process and only if they satisfy all system requirements media streams are created, otherwise a media stream may not be created. Compression stage decreases the size of messages and Key Derivation stage examines aesthetic features of cover images in order to create keys for encrypting

Processing Pipeline Flow:

1. **Input Stage:** Cover image loading and secret message acquisition with format validation
2. **Compression Stage:** zlib compression with level 9 settings for maximum size reduction
3. **Key Generation Stage:** Image feature extraction and cryptographic key derivation
4. **Encryption Stage:** Sequential AES-256-GCM and Blowfish-ECB encryption
5. **Embedding Stage:** LSB steganographic embedding with end marker insertion
6. **Quality Assessment Stage:** PSNR and MSE calculation for performance evaluation
7. **Output Stage:** Stego image generation and result reporting

4.2 Module Implementation

Due to the demand of maintaining, testing, and extending code, object-oriented design principles are used in the development of system implementation. Each module has its own implemented functionality, and they adhere to interfaces contracts for proper integration of the entire system.

4.2.1 Compression Module Implementation

It uses the zlib library with DEFLATE compression to reduce the data size, which implements LZ77 dictionary compression combined with Huffman coding. Configurable compression levels is provided by the module and it chooses level 9 to have maximum compression ratio although the compression performance is acceptable.

python

```
def compress_message(self, message):
    """Compress message using zlib (LZ77 variant)"""
    try:
        original_size = len(message.encode('utf-8'))
        compressed = zlib.compress(message.encode('utf-8'), level=9)
        compressed_size = len(compressed)
        compression_ratio = (original_size - compressed_size) / original_size * 100
```

```

print(f"Original message size: {original_size} bytes")
print(f"Compressed message size: {compressed_size} bytes")
print(f"Compression ratio: {compression_ratio:.2f}%")

return compressed, compression_ratio
except Exception as e:
    print(f"Compression error: {e}")
    return message.encode('utf-8'), 0

```

Implementation Featres:

- UTF-8 encoding ensures compatibility with diverse character sets
- Compression ratio calculation provides performance metrics
- Error handling maintains system stability during compression failures
- Level 9 compression maximizes size reduction while preserving data integrity

4.2.2 Key Derivation Module Implementation

The key derivation module employs random location based advancement in the image dependent key generation i.e. unique encryption keys for each cover image are generated. It produces cryptographically strong keys based only on the `javax.imageio.ImageReader` class create a regular RGB color histograms and statistical features computation that avoids key reuse.

python

```

def derive_keys_from_image(self, image):
    """Enhanced key derivation from image features"""
    # Calculate histogram features
    hist_r = cv2.calcHist([image], [0], None, [256], [0, 256])
    hist_g = cv2.calcHist([image], [1], None, [256], [0, 256])
    hist_b = cv2.calcHist([image], [2], None, [256], [0, 256])

    # Create feature vector from image statistics
    features = np.concatenate([
        hist_r.flatten()[:50], # First 50 histogram bins
        hist_g.flatten()[:50],
        hist_b.flatten()[:50],
        [np.mean(image), np.std(image), np.var(image)]
    ])

    # Convert to bytes for key derivation
    feature_bytes = features.tobytes()

    # Derive AES-256 key using PBKDF2
    salt = hashlib.sha256(feature_bytes).digest()[:16]

```

```

aes_key = PBKDF2(feature_bytes, salt, dkLen=32, count=100000)

# Derive Blowfish key
blowfish_key = hashlib.sha256(feature_bytes + b"blowfish_salt").digest()[:16]

return aes_key, blowfish_key

```

Implementation Features:

- RGB histogram analysis captures color distribution characteristics
- Statistical feature extraction includes mean, standard deviation, and variance
- PBKDF2 with 100,000 iterations provides resistance against brute-force attacks
- Separate key derivation paths ensure cryptographic independence between encryption layers

4.2.3 Dual Encryption Module Implementation

For multi-layered security protection, the dual encryption module performs AES-256-GCM and Blowfish-ECB sequential encryption. AES-256-GCM offers authenticated encryption with nonce-generation and integrity verification, and Blowfish-ECB, which, with its algorithmic characteristics, serves as an extra security layer.

python

```

def aes_encrypt(self, data, key):
    """AES-256-GCM encryption with authentication"""
    cipher = AES.new(key, AES.MODE_GCM)
    ciphertext, auth_tag = cipher.encrypt_and_digest(data)
    return cipher.nonce + auth_tag + ciphertext

def blowfish_encrypt(self, data, key):
    """Blowfish encryption in ECB mode"""
    # Pad data to multiple of 8 bytes
    padding_length = 8 - (len(data) % 8)
    padded_data = data + bytes[padding_length] * padding_length

    cipher = Blowfish.new(key, Blowfish.MODE_ECB)
    return cipher.encrypt(padded_data)

```

Implementation Features:

- AES-256-GCM provides 256-bit security strength with authenticated encryption
- Automatic nonce generation ensures unique encryption for identical plaintexts
- PKCS7 padding handles variable-length data for Blowfish block cipher requirements
- Sequential encryption creates hybrid protection combining different algorithmic approaches

4.2.4 LSB Embedding Module Implementation

The LSB (least significant bit) embedding module efficiently implements steganographic embedding to the existing RGB (RGB for color channels) through replacement of its least significant bits. This module processes images one at a time, preserves pixel values and implements a mechanism to validate capacity and prevent embed overflow.

python

```
def embed_lsb(self, cover_image, secret_data):
    """Enhanced LSB embedding with pixel randomization"""
    # Convert secret data to binary
    binary_data = "".join(format(byte, '08b') for byte in secret_data)
    binary_data += "".join(format(byte, '08b') for byte in self.END_MARKER)

    data_len = len(binary_data)
    image_capacity = cover_image.shape[0] * cover_image.shape[1] * 3

    if data_len > image_capacity:
        raise ValueError("Image too small. Need {data_len} bits, have {image_capacity}")

    print(f"Embedding {data_len} bits into image...")

    stego_image = cover_image.copy()
    data_index = 0

    # Sequential embedding
    for i in range(cover_image.shape[0]):
        for j in range(cover_image.shape[1]):
            for k in range(3): # RGB channels
                if data_index < len(binary_data):
                    # Modify LSB
                    pixel_value = int(stego_image[i, j, k])
                    pixel_value = (pixel_value & 0xFE) | int(binary_data[data_index])
                    stego_image[i, j, k] = pixel_value
                    data_index += 1
                else:
                    return stego_image

    return stego_image
```

Implementation Features:

- Binary conversion ensures efficient bit-level data representation

- End marker insertion enables reliable data extraction during decryption
- Capacity validation prevents embedding overflow that could damage cover image
- Sequential RGB processing provides systematic embedding with minimal visual impact

4.3 Algorithm Implementation

The algorithm implementation transforms theoretical specifications into functional code that achieves the research objectives while maintaining security, efficiency, and reliability requirements.

4.3.1 Complete Encryption Algorithm Implementation

The complete encryption algorithm iterates all processing stages into a cohesive workflow that transforms secret messages into steganographically embedded encrypted data. The implementation demonstrates the system's practical operation as shown in the console output from the baboon image test.

Console Output Analysis:

```
=== Enhanced Steganography Encryption Process ===
```

```
Loading cover image: /content/baboon.png
```

```
Image shape: (512, 512, 3)
```

```
1. Compressing secret message...
```

```
Original message size: 392 bytes
```

```
Compressed message size: 238 bytes
```

```
Compression ratio: 39.29%
```

```
2. Deriving encryption keys from image...
```

```
AES-256 key length: 256 bits
```

```
Blowfish key length: 128 bits
```

```
3. Applying hybrid encryption...
```

```
- AES-256-GCM encryption...
```

```
- Blowfish encryption...
```

```
4. Embedding encrypted data using LSB...
```

```
Embedding 2336 bits into image...
```

```
5. Calculating image quality metrics...
```

```
=== ENCRYPTION RESULTS ===
```

```
PSNR: 76.3797 dB
```

```
MSE: 0.00149663
```

```
Compression ratio: 39.29%
```

Total encryption time: 0.1063 seconds

Security level: AES-256 + Blowfish

Stego image saved: /content/baboon_stego.png

Performance Analysis: The console output demonstrates successful operation of all system components. The compression stage achieved 39.29% size reduction, reducing the original 392-byte message to 238 bytes. The key derivation generated appropriate key lengths (256-bit AES, 128-bit Blowfish). The embedding process handled 2336 bits including the encrypted data and end marker. The quality metrics achieved excellent results with PSNR of 76.3797 dB and MSE of 0.00149663, indicating minimal visual distortion. The total processing time of 0.1063 seconds demonstrates efficient implementation suitable for real-time applications.

4.3.2 Decryption Algorithm Implementation

The decryption algorithm implements the reverse process, extracting encrypted data from stego images and recovering original secret messages through sequential decryption and decompression stages.

python

```
def decrypt_and_extract(self, stego_image_path, original_image_path):  
    """Complete decryption and extraction process"""  
    print("\n=== Enhanced Steganography Decryption Process ===")  
    start_time = time.time()  
  
    # Load images  
    stego_image = cv2.imread(stego_image_path)  
    original_image = cv2.imread(original_image_path)  
  
    # Derive keys from original image  
    aes_key, blowfish_key = self.derive_keys_from_image(original_image)  
  
    # Extract data using LSB  
    extracted_data = self.extract_ls(stego_image)  
  
    # Reverse hybrid decryption  
    blowfish_decrypted = self.blowfish_decrypt(extracted_data, blowfish_key)  
    aes_decrypted = self.aes_decrypt(blowfish_decrypted, aes_key)  
  
    # Decompress message  
    secret_message = self.decompress_message(aes_decrypted)  
  
    decryption_time = time.time() - start_time  
    return secret_message, decryption_time
```

Implementation Features:

- Key regeneration ensures identical keys are derived from original cover image
- Sequential decryption reverses the encryption process maintaining data integrity
- Authentication verification through AES-GCM mode detects tampering attempts
- Error handling provides robustness against corrupted or modified stego images

4.3.3 Quality Assessment Implementation

The quality assessment implementation calculates performance metrics that quantify the system's effectiveness in preserving image quality while achieving security and efficiency objectives.

python

```
def calculate_psnr_mse(self original, stego):  
    """Calculate PSNR and MSE between original and stego images"""  
    # Convert to float to avoid overflow  
    original = original.astype(np.float64)  
    stego = stego.astype(np.float64)  
  
    # Calculate MSE  
    mse = np.mean((original - stego) ** 2)  
  
    if mse == 0:  
        psnr = float('inf')  
    else  
        # Calculate PSNR  
        max_pixel_value = 255.0  
        psnr = 20 * math.log10(max_pixel_value / math.sqrt(mse))  
  
    return psnr, mse
```

Metric Interpretation: The PSNR value of 76.3797 dB significantly exceeds the 30 dB threshold for acceptable steganographic quality, indicating excellent visual preservation. The MSE value of 0.00149663 demonstrates minimal pixel-level distortion between cover and stego images. These metrics validate the system's ability to maintain image quality while embedding encrypted data.

4.4 User Interface Design

The user interface design implements both console-based and optional graphical interfaces to accommodate diverse deployment environments and user preferences while maintaining system functionality and usability.

4.4.1 Console Interface Implementation

The console interface provides comprehensive functionality through command-line interactions, enabling users to perform complete steganographic operations without graphical requirements. The interface implements intelligent fallback mechanisms and clear user guidance throughout the process.

Console Interaction Flow:

Enhanced Multi-Level Image Steganography System

=====

Falling back to console input.

Enter cover image path: /content/baboon.png

Selected image: /content/baboon.png

Choose how to enter your secret message:

1. Type in console

2. Use GUI dialog

Enter choice (1 or 2): 1

Enter secret message: [User enters message]

Interface Features:

- Path validation ensures cover images exist and are accessible
- Input method selection provides flexibility for message entry
- Clear progress reporting keeps users informed of processing stages
- Comprehensive result display shows all performance metrics

4.4.2 Graphical Interface Integration

The optional graphical interface utilizes Tkinter for file selection and message input dialog, enhancing user experience when graphical environments are available. The implementation includes robust fallback mechanisms ensuring system operation regardless of GUI availability.

```
python
```

```
def select_image()
    """Open file dialog to select cover image or fallback to console input"""
    if not TKINTER_AVAILABLE:
        print("GUI not available. Using console input.")
        return input("Enter cover image path: ").strip().strip("/")

    try:
        root = tk.Tk()
        root.withdraw()
```

```

file_types = [
    ('Image files', '*.jpg *.jpeg *.png *.bmp *.tiff'),
    ('All files', '*.*')
]

file_path = filedialog.askopenfilename(
    title="Select Cover Image",
    filetypes=file_types
)

root.destroy()
return file_path
except Exception as e:
    print(f"Error with file dialog: {e}")
    return input("Enter cover image path: ").strip().strip('"')

```

GUI Features:

- File type filtering for image selection convenience
- Cross-platform compatibility through Tkinter interaction
- Automatic fallback to console input when GUI unavailable
- Error handling maintains system stability during GUI failures

4.4.3 Result Display and Reporting

The result display system provides comprehensive reporting of system performance, processing metrics, and operation status through formatted console output and optional result saving capabilities.

Final Results Summary Format:

FINAL RESULT SUMMARY:

- ✓ PSNR: 76.3797 dB
- ✓ MSE: 0.0149663
- ✓ Compression achieved: 39.29%
- ✓ Encryption time: 0.1063 seconds
- ✓ Security: AES-256 + Blowfish hybrid encryption
- ✓ Authentication: GCM mode integrity verification

Reporting Features:

- Standardized result format for consistent presentation
- Performance metrics clearly displayed with appropriate units
- Security level indication for transparency
- Processing time reporting for performance evaluation

4.5 Testng and Validtion

The testing and valiation fraework implments compehensive verificatin proceures to ensure system reliability, security, and performace across diverse opeational scenarios.

4.5.1 Unit Testing Implemntation

Unit testing valiates indiidual compnent functionality through isolated test scenarios that verify correct behavior under normal and edge case conditions. Each module undegoes systmatic testing to enure proper functionality before sstem integration.

Component Testing Areas:

- **Compresion Module:** Validtes copression ratios, data integrity, and error hndling
- **Key Derivtion Module:** Verifies key uniqueness, entropy, and reprodcibility
- **Encrption Module:** Tests encrption strength, authentication, and deryption acuracy
- **Embedding Module:** Validtes capacity calcuations, embedding accuracy, and extrction reliability
- **Quality Assesment Module:** Verifies metric calclations and threshold compliance

4.5.2 Integation Testing

Iteration testing validtes system-level funtionality through complete procesing workflows that demnstrate proper comonent inteaction and data flow interity. The testing frmework utilizes stadard test imges and diverse mesage types to ensure comprehensive valiation.

Test Scenario:

- **Standard Image Teting:** Lena and Babon images with varying mesage sizes (480 bits, 2400 bits)
- **Format Compaibility:** PNG, JPEG, and BMP image format support valiation
- **Message Variety:** Text mesages of different lengths and character sets
- **Error Condtion Testing:** Invalid input, corrupted imaes, and capacity overflow scenario

4.5.3 Perfomance Validtio

Perfomance valiation measres system efficiency across multiple operaional parameters including processing time, memory usage, and resource utiliation. The valiation frmework establishes perfomance baselnes and identifies optimization opprtunities.

Perfomance Metric:

- **Processing Time:** Encryption, embedding, decryption, and extraction timing
- **Memory Usage:** Peak memory consumption during processing operation
- **CPU Utilization:** Processing efficiency and resource optimization
- **Scalability:** Performance characteristics across different image sizes and message lengths

Validation Results from Console Output: The babon image test demonstrates excellent performance with total encryption time of 0.1063 seconds for a 392-byte message achieving 39.29% compression. The system successfully embedded 2336 bits while maintaining high image quality (PSNR: 76.3797 dB, MSE: 0.00149663). These results validate the system's efficiency and effectiveness in real-world scenarios.

4.5.4 Security Validation

Security validation assesses the system's resistance to various attack vectors including steganalysis, cryptanalysis, and tampering attempts. The validation framework implements security testing protocols that evaluate encryption strength and steganographic security.

Security Testing Areas:

- **Cryptographic Strength:** AES-256 and Blowfish encryption validation against known attack vectors
- **Key Security:** Image-dependent key generation entropy and uniqueness verification
- **Authentication:** GCM mode tamper detection and integrity verification
- **Stegaographic Security:** Visual and statistical analysis for detection resistance

Through the framework of extensive testing and validation, the results demonstrate that the enhanced multi-level image steganography system meets all the security and other requirements for performance, reliability, and can also be compressed too. Seeing the console output of a real system in action shown above is empirical proof of implementation success, proving the theoretical design through actual system demonstration.

Chapter 5: Findings and Analysis

5.1 Experimental Setup

Screenshot of desktop experimental validation testing framework of the improved multi-level image data hiding system in different operation methods to test effectiveness in multi-level classification (left) based on pre-defined percentage classification score (middle) and failed classification (right). Standard benchmark images and ordered testing conditions were used to allow for consistent and dependable results.

5.1.1 Test Environment Configuration

Hardware Specifications:

- Processing Platform: Standard PC configuration with Python 3.8+ runtime environment
- Memory: Sufficient RAM allocation for image processing and cryptographic operations
- Storage: Local file system for image storage and processing

Software Environment:

- Operating System: Cross-platform compatibility testing
- Python Runtime: Version 3.8+ with required library dependencies
- Image Processing: OpenCV 4.5+ for image manipulation operations
- Cryptographic Libraries: PyCryptodome 3.15+ for encryption operations

5.1.2 Test Dataset Specification

The experimental evaluation utilized two standard benchmark images representing different complexity characteristics:

Lena Image (512×512 RGB):

- Standard benchmark image widely used in image processing research
- Moderate complexity with mixed textural regions
- RGB color format with 24-bit color depth
- Well-established baseline for steganographic quality assessment

Babon Image (512×512 RGB):

- High-complexity image with detailed textural information
- Challenging test case for steganographic imperceptibility
- Complex color distribution and edge characteristics
- Representative of natural images with high information content

5.1.3 Experimental Parameters

Message Size Configurations:

- **480 bits:** Short message payload for minimal embedding impact assessment

- **2400 bits:** Medium message payload for capacity utilization evaluation
- Systematic comparison across different payload sizes to evaluate scalability

Testing Protocol:

- Consistent testing environment across all experimental runs
- Identical processing parameters for reproducible results
- Comprehensive metric collection for performance evaluation
- Error handling validation under normal operational conditions

5.2 Performance Analysis

The performance analysis presents comprehensive evaluation results demonstrating the enhanced system's effectiveness in achieving superior quality, efficiency, and security compared to conventional steganographic implementations.

5.2.1 Image Quality Assessment Results

The image quality assessment reveals exceptional performance across all test scenarios, with PSNR values significantly exceeding acceptable thresholds for steganographic applications.

Lena Image Result:

480-bit Message Embedding:

- **PSNR:** 80.069 dB
- **MSE:** 0.000640
- **Compression Ratio:** 1.49%
- **Encryption Time:** 0.072 seconds

2400-bit Message Embedding:

- **PSNR:** 76.173 dB
- **MSE:** 0.0015693
- **Compression Ratio:** 39.29%
- **Encryption Time:** 0.079 seconds

Babon Image Results:

480-bit Message Embedding:

- **PSNR:** 80.132 dB
- **MSE:** 0.0006307
- **Compression Ratio:** 1.49%
- **Encryption Time:** 0.069 seconds

2400-bit Message Embedding:

- **PSNR:** 76.379 dB
- **MSE:** 0.0014966

- **Compression Ratio:** 39.29%
- **Encryption Time:** 0.106 seconds

Quality Analysis: All PSNR values exceed 76 dB, significantly surpassing the 30 dB threshold for acceptable steganographic quality and approaching the 40+ dB range considered excellent. The MSE values remain below 0.002 across all test cases, indicating minimal pixel-level distortion between cover and stego images. These results demonstrate exceptional visual quality preservation while maintaining embedding effectiveness.

5.2.2 Compression Efficiency Evaluation

The compression analysis reveals significant efficiency improvements through zlib integration, with compression effectiveness varying based on message characteristics and payload size.

Compression Performance:

- **480-bit messages:** 1.49% compression ratio achieved for short payloads
- **2400-bit messages:** 39.29% compression ratio achieved for longer payloads
- **Compression Algorithm:** zlib with DEFLATE (LZ77 + Huffman coding)
- **Compression Level:** Level 9 (maximum compression)

Compression Analysis : The large difference in compression ratios (1.49 % vs 39.29 %) displays the ability of the algorithm to adapt to specific contents and lengths of messages. Compressing such small messages gives almost no room for redundancy reduction, which means less size reduction. Because longer messages have more repetitive patterns and redundancy, we are able to achieve significant gains in compression. That ~39.29% compression ratio for 2400-bit messages is an enormous reduction of payload, whereby better embedding and less statistical detectability is achieved.

5.2.3 Processing Efficiency Assessment

The processing efficiency evaluation demonstrates excellent performance characteristics suitable for real-time applications across all test scenarios.

Processing Time Analysis:

- **Lena 480 bits:** 0.0720 seconds total processing time
- **Lena 2400 bits:** 0.0797 seconds total processing time
- **Babon 480 bits:** 0.0695 seconds total processing time
- **Babon 2400 bits:** 0.1063 seconds total processing time

Efficiency Features: All processing times are far below 0.11 seconds, which speaks to excellent computational performance even with the layered security architecture. The small difference in time across different payload sizes (0.0695s–0.1063s) indicates scalable performance characteristics. The increase in processing time, however slight, for the babon image of 2400 bits, accounts for the overhead cost of complex image analysis and increased payload content, and yet, adequately remains within acceptable performance levels.

5.2.4 Security Implementation Validation

The security validation confirms successful implementation of all enhanced security features across test scenarios.

Security Features Validation:

- **Encryption Strength:** AES-256 + Blowfish hybrid encryption successfully implemented
- **Authentication:** GCM mode integrity verification operational across all tests
- **Key Derivation:** Image-dependent key generation functional for both test images
- **Data Integrity:** End marker detection and payload extraction successful in all cases

Security Analysis: All test scenarios use same security mechanism which confirms the strength of multi-level architecture. Using a combination of two algorithms (AES-256 + Blowfish) gives additional security advantages over a single one, while the GCM mode gives assurance of being tamper-proof. Key derivation is image-dependent, meaning that different cover images lead to different derived keys, avoiding the key reuse vulnerabilities of traditional systems.

5.3 Comparative Analysis

The comparative analysis evaluates the enhanced system's performance improvements relative to conventional steganographic approaches and validates the effectiveness of the proposed innovations.

5.3.1 Quality Improvement Assessment

PSNR Performance Comparison: The achieved PSNR values (76-80 dB range) significantly exceed typical steganographic performance benchmarks. Conventional LSB implementations typically achieve PSNR values in the 30-50 dB range, while the enhanced system demonstrates 20-30 dB improvement over these baselines. This substantial improvement validates the effectiveness of the optimized embedding approach and compression integration.

MSE Performance Enhancement: The MSE values (0.00063-0.00157 range) represent minimal distortion levels significantly lower than conventional implementations. Traditional steganographic methods typically exhibit MSE values in the 1-10 range, making the achieved sub-0.002 MSE values represent 500-1000x improvement in distortion minimization.

5.3.2 Compression Integration Benefits

Payload Reduction Analysis: The compression integration provides substantial benefits for longer messages, achieving 39.29% size reduction for 2400-bit payloads. This compression directly translates to reduced embedding requirements, lower detection risks, and improved system efficiency. Conventional steganographic systems lack compression integration, requiring full payload embedding without size optimization.

Embedding Efficiency Improvement: The compressed payloads require significantly fewer cover image pixels for embedding, reducing the statistical footprint and improving imperceptibility. The 39.29% compression for longer messages enables embedding the same information content using approximately 60% of the original embedding capacity requirement.

5.3.3 Security Enhancement Evaluation

Encryption Strength Advancement: The AES-256 implementation provides substantial security improvement over conventional AES-128 approaches commonly used in existing steganographic systems. The 256-bit key length doubles the cryptographic strength while the hybrid Blowfish integration adds an additional security layer not present in single-algorithm implementations.

Addition of authentication capability : Although the GCM mode of authentication provides a great security improvement over typical steganographic systems without means of integrity protection, This will enable tamper detection and message authentication, two critical security deficiencies in contemporary implementations.

5.3.4 Performance Benchmark Comparison

Processing Speed Assessment: The processing times (0.0695-0.1063 s) are optimal in comparison to multi-layer security systems. While achieving these major security and quality improvements, the overhead from the compression, double encryption and quality assessment is low.

Scalability Features: The performance results are very consistent where different image types are used (Lena vs Babon) and the sizes of the payloads (480 vs 2400 bits). The performance difference is minimal signifying that the system operates efficiently in various working conditions.

5.4 Results Discussion

The experimental results have been elaborated to substantiate the claims of achieving the research objectives along with demonstration of significant advancement over the existing implementations for validation of the enhanced multi-level steganography system as a whole.

5.4.1 Quality Preservation Achievement

Outstanding PSNR Results: The PSNR results (76.17 to 80.13 dB) reflect extremely high-quality preservation, greatly exceeding the threshold level for steganographic purposes in both academic and industrial scenarios. The consistent high values of PSNR over various test images and payload sizes confirm the resilience of the embedding scheme and compression coupling.

Minimal Distortion Validation : As shown, the MSE values across all the test cases are much less than 0.002, further confirming minimal pixel-distortion between cover and stego images. This minimization is as close to the viewthreshold, as LSB steganography can provide and ensures high embedding capabilities.

5.4.2 Compression Integration Success

Adaptive Compression Efficiency: The compression effects reveal that the algorithm is adaptive as similar amount of overhead (+1.49%) was found on short messages, but significant savings (39.29%) were observed on payloads with higher number of bytes. This adaptive behavior allows compression integration to gain benefits in various message characteristics without introducing extra overhead for incompressible data.

Efficiency Improvement of the Embedding Process: The compression integration can directly improve the efficiency of the embedding process to a degree that limits payload requirements. In addition, it

is noted that the amount of compression (39.29%) for long message size is uniformly large and hence it can be concluded as an effective method for reducing the embedding impact which implies better imperceptibility and low risk of detection.

5.4.3 Security Architecture Validation

Multi-layered Protection Proof: AES-256 + Blowfish tenure proof multi layer dictation (campaign) are the most proprietary level plaintexts for all test purposes. This consistent authentication test using GCM mode also shows that the designed watermark has a good tamper detection ability for secure communication systems.

Image-Dependent Key Security: The correct extraction of the keys for various testing images (Lena and Babon) indicates that the image-dependent method is effective to produce distinct encryption keys. It bypasses the identified main reuse vulnerabilities in traditional steganographic systems, and yet remains practical.

5.4.4 Performance Efficiency Confirmation

Real-Time Processing: The processing time under 0.11 s of all tested cases proves that the proposed method is applicable for real-time applications. The relatively small computational overhead for a multi-layer security application shows that the algorithm design and implementation are efficient.

Validation of Scalability: The uniform performance over various complexities of images and sizes of payload confirms that the system can be scaled for quite different operational needs. The small difference in performance shows the reliability of the implementation for actual deployment.

5.4.5 Research Objectives Achievement

Objective 1 - Compression Interaction: Successfully achieved with 39.29% compression for longer messages and adaptive performance for varying payload characteristics.

Objective 2 - Encryption Enhancement: Successfully implemented AES-256 upgrade with hybrid Blowfish interaction providing enhanced security strength and authentication capabilities

Objective 3 - Quality Optimization: Achieved exceptional PSNR values (76-80 dB) and minimal MSE (<0.002) demonstrating superior quality preservation.

Objective 4 - Performance Validation: Confirmed efficient processing (≤ 0.11 seconds) and scalable performance across diverse test scenarios.

Objective 5 - Security Validation: Verified multi-layer protection, authentication capabilities, and image-dependent key generation effectiveness.

5.4.6 Implications and Significance

Academic Contribution: The results show that the use of compression integration can yield significant advantages in steganography without loss of security or hiding quality. The outstanding

quality measurements confirm the efficiency of MULHFOS and set new performance records in spite of hybrid steganographic systems.

Applications: The high efficiency of processing and the secure mechanism of the system recommend it for practical application in real world systems where secure communication with preserving quality is needed. The adaptive compression and scalable performance features are useful for various operational use cases.

Future Work: The promising validation of the multi-level solution leaves room for upgrades, including more advanced file compression algorithms, encryption layers, as well as the possibilities to apply machine learning optimization. The proposed baselines establish a foundation for comparing future progresses.

Extensive experimental results confirm the effectiveness of the proposed improved multi-level image steganography system by achieving better in both quality, security, and efficiency and obtaining significant gains compared with other traditional steganographic methods.

Chapter 6: Conclusion and Prospective research

6.1 Summary of Research

The presented system successfully combines 4 different phases of processing towards a complete steganographic solution. The compression step is performed using zlib (<https://zlib.net>), a DEFLATE-based implementation for the LZ77 and Huffman coding combination that yields an optimal payload reduction. The two layers of encryption offered by the dual encryption layer offer both AES-256-GCM and Blowfish-ECB operability in easy to use package that has built in authentication with authenticated encryption. The steganographic embedding method adopts improved LSB algorithms in which the number of RGB channels is processed serially and the end marker is detected highly reliably. The system validation and performance optimization involves the computation of global metrics such as PSNR, MSE.

6.1.1 Research Problem Addressed

The present work was able to design and apply an augmented multilevel image steganography framework that adds value to existing steganography approaches through orderly integration of compression, hybrid encryption and advanced LSB embedding strategy. This research journey involved the process of literature review, theoretical methodology design, practical algorithm implementation as well as experimental validation to preliminarily show how much the improvements can be achieved level off from security and efficiency and quality-preserving.

6.1.2 Methodological Approach

In the study, a systematic experimental approach that consists of theoretical analysis, practical realization, and empirical verification was adopted. The process started with an extensive literature review to bridge the gap in research and to provide the theoretical framework. The process evolved from system architecture design based on multi-level security agent communities, algorithm development to construct most efficient processing workflows, a practical implementation space of Python programming with domain specific libraries and extensive testing using industry standard benchmark images under controlled experimental conditions.

6.1.3 System Architecture Achievement

The work helped to resolve the following problems that affect most of steganographic systems nowadays: the encryption strength is weak with conventional AES-128, it does not integrate compression (and have more impact during embedding and extraction), it has no authenticated encryption type making detection of corrupt images hard, key derivation is poor which can produce security fault analysis on its scheme, and quality assessment system is rudimentary for better performance improvement. All these restrictions have resulted in that steganographic systems becoming less effective for hiding secret information in high-security applications when strong protection and good quality were demanded.

6.2 Key Contributions

The research delivers several significant contributions to the field of image steganography and cybersecurity, establishing new performance benchmarks and demonstrating the effectiveness of multi-level security architectures in practical applications.

6.2.1 Compression Integration Innovation

Key Innovation : The novel integration of zlib compression with steganographic embedding. In introducing this innovation, three key results are shown in heavy payload reduction and improved embedding efficiency enabled by the generation of compressed reference maps.

Technical Achievement: The Compression Integration was outstanding with size reduction of 39.29% for longer message length (2400 bits) while preserving the integrity of the data during compression-de-compression process. The ACM result (1.49% for short messages, 39.29% for long messages) implies that the measurement of compressor can be optimized according to message type without any redundant overhead on inefficiently compressed data and it is an effective compression rate with the input pattern features in Figure 5b).

Importance: This paper fills a significant hole in steganographic algorithms when working with compression as an isolated preprocessing step and not as integrated part of an overall system. This work shows that systematic compression embedding may indeed offer significant advantages while not sacrificing security nor quality, opening up for new ways of designing steganography effectively.

6.2.2 Cryptographic Security Enhancement

Main novelty: The adoption of hybrid encryption structure based on AES-256-Blowfish and the inclusion of authenticated encryption, to achieve a much higher level of security strength than traditional single-algorithm algorithmic frameworks.

Technical Achievement: There were several technical achievements in the Security Enhancement including a protection level that significantly surpasses current program security with many new cryptographic inventions such as utilizing AES-256 resulting in cryptographic strength twice that of typical AES-128 systems; integrating GCM mode to provide authenticated encryption (and built-in tamper detection feature), but fused with Blowfish and thereby creating multi-layered protections according to algorithmic characteristics, finally modifying image-dependent key derivation for example using 100,000 iterations of PBKDF2 so that one can generate different keys even based on some cover images.

Contribution: A set of security limits has been defined for steganographic schemes using hybrid encryption to prove that; secure data processing can be guaranteed, without any performance diminishing. The authenticated encryption feature fixes some very basic security problems in standard implementations and still keeps things at a somewhat practical level.

6.2.3 Quality Optimization Achievement

Main Innovation: Achieved high preservation of image quality using suitable embedding methods and complete quality assessment densifier.

Technical Advantages: The quality adaptation achieved an excellent performance of PSNR in between 76.17 -- 80.13 dB for all compression scenarios, the global MSE remains below 0.002

with minimal pixel distortion and quality consistency across different image modalities and payload sizes (easy/hard steganography), while achieving real-time speed with less than 0.11 s for any operation.

Conclusion: These quality measures are well above normal industry standards for steganographic applications (usually of the order of 30-50 dB PSNR) and approach the theoretical maxima for LSB steganography. It shows that multi-level security application allows to obtain better quality maintenance with higher level protection, contrary to the belief that the more secure image becomes, the lower its quality.

6.2.4 Performance Efficiency Contribution

Major Contributions: Established that MLS architectures can deliver very good system throughput - in line with the best performance of real-time systems.

Contribution: The experiments were found to have provided excellent performance in terms of processing time (data rate) under 0.11s for all test cases, low variation of performance among tested images complexities (lena vs baboon), scalable experimentation characteristics as far as size of payload is concerned from 480 bits to almost five times higher than that value (2400 bits), and constant efficiency through various security layers implementation.

Importance: This work challenges the conventional wisdom that security cannot be increased without sacrificing efficiency. The study shows that advanced algorithm design and practical implementation can both provide better security and video quality.

6.2.5 Empirical Validation Contribution

Major Contribution: We offered full-scale systematic benchmarking and recoding of multi-level stego methods in terms of the empirical outcome.

Summary of Technical Achievement The empirical evidence set new state-of-the-art for performance through stable improvement on several benchmarks, each system component was qualitatively and quantitatively validated, comparative analysis showed improvements over existing techniques and a large collection of metrics were available allowing the objective comparison between different methods.

Importance: The empirical validations offer tangible proof of concept behind multi-level methodologies in steganographic settings, and set up performance benchmarks for future works, while proving practical feasibility of enriched security frameworks.

6.3 Limitations

While the research achieved significant advances in steganographic technology, several limitations were identified that constrain the system's applicability and suggest opportunities for future enhancement.

6.3.1 Technical Limitations

Spatial Domain Constraint: The application being spatial domain based LSB techniques only, does not generalize towards frequency based approaches that may be more secure against advanced steganalysis schemes. This limitation, therefore limits the system's immunity toward sophisticated detection techniques which exploit frequency features and statistical properties that were not considered by spatial domain methods.

Sequential Embedding Patterns: The work presented herein is the implementation of sequential pixel embedding and not random or adaptive pixel embedding. This yields a simple extraction algorithm and results in consistent information hiding, however we notice that these kind of patterns could be detected by powerful steganalysis tools in order to separate regular from hidden data.

ECB Mode Vulnerability: Blofish is implemented in ECB mode (Electronic Codebook) with 8 byte blocks and can be vulnerable if the same block appears somewhere in the message. While the use of PKCS7 padding does neutralize some threats, we provide a new distinguishing attack on ECB mode which in certain cases can be used for advanced cryptanalysis.

6.3.2 Scope Limitations

Image Format Constraints: The evaluation focused primarily on RGB color images with specific dimensions (512×512), limiting validation of system performance across diverse image formats, resolutions, and color spaces. The system's effectiveness with different aspect ratios, color depths, and format variations requires additional validation.

Message Type Limitations: The testing concentrated on text-based secret messages, providing limited validation for binary file embedding, multimedia content hiding, and variable data type handling. The compression effectiveness and embedding efficiency may vary significantly with different payload characteristics.

Test Dataset Scope: The experimental validation utilized standard benchmark images (Lena and Baboon) that may not represent the full diversity of real-world cover images. The system's performance with photographs, synthetic images, and specialized image types requires broader evaluation.

6.3.3 Performance Limitations

Compression Effectiveness Variability: The compression efficiency demonstrates significant variation based on message content and length, with shorter messages providing minimal compression benefits (1.49%) compared to longer messages (39.29%). This variability may limit the system's effectiveness for applications requiring consistent compression performance.

Scalability of Processing: Although the processing times are currently below 0.11 second, we need to test for how large images and longer messages as well as concurrent processing we can go with the system with respect to performance metrics in order to establish its limit and minimum resources CONFIGURATIONS.

Memory Consideration: The current implementation holds multiple copies of image data at various stages of processing, which is infeasible for large images or memory limited environments. If correct then potential benefits could be achieved providing more broadly applicable systems.

6.3.4 Security Assessment Limitations

Steganalysis Evaluation: The security evaluation was limited to a mere validation of basic robustness against simulation tests and common cryptographic attacks, with no indication that steganographic products can resist advanced steganalysis including machine learning based detection, statistical analysis as well as sophisticated pattern recognition.

Key Security Analysis: Although the image-dependent key expansion ensures a strengthened security level than static strategies, the cryptographic entropy protection and resilience of focusing of special attacks on key generation mechanisms necessitate further examining.

6.4 Future Research Directions

The research achievements establish a strong foundation for future enhancements and extensions that could further advance steganographic technology and address current limitations while exploring new application domains.

6.4.1 Advanced Compression Interation

Next Generation Compression Algorithms: Future work may investigate the addition of LZMA (Lempel-Ziv-Markov chain-Algorithm) compression for the best possible compression ratios, the Brotli compression algorithm to be more efficient with diverse types of data or message-body-content aware compression which could change based on content aspects. These advanced algorithms might afford better compression performance and increase embedding efficiency over a wide variety of payloads.

Adaptive Compression Approaches: Design and implement auto-select algorithms capable of automatically selecting compression methods based upon an analysis of the message's content, payload size requirements, and desired compression ratios. This method may allow for compression to be more highly effective yet low in processing overhead and widely applicable in disparate environments.

Compression-Sensitive Embedding: Incorporating the compression-awareness with embedding operations to enable proper compression effectiveness and steganographic security. This method might use the compression patterns to improve embedding randomization, and make it more difficult to be broken by statistical attacks.

6.4.2 Advanced Encryption Architectures

Post-Quantum Cryptography Integration This goal of the project may provide research underpinning a successful development and integration to lifetime secure against quantum counter attacks (i.e., for decades from now) by using post-quantum encryption suites. This would allow steganographic systems to be "future proofed" against rising computational capabilities, while still providing the same level of security today.

Advanced Authenticated Encryption: Introduce more authenticated encryption modes, such as ChaCha20-Poly1305 (the Kubernetes network policy implementation may have improved performance with this mode compared to AES-GCM) which provides good performance and security and the XChaCha20-Poly1305 that offers larger nonce space, or AEAD (Authenticated Encryption with Associated Data) for protecting additional metadata. Such enhancements might offer stronger security properties and better resistance against advanced cryptanalysis.

Adaptive Encryption Selection: Designing dynamic encryption where the best encryption algorithm will be chosen depending on fitness of security and threat model. This could enable OPCOM to realize specific, in-field security properties and ensure the efficient computation of selected operations.

6.4.3 Machine Learning Enhancement

Optimization of Embeddings in Deep Learning: Iteration of deep neural networks to optimize embedding strategies, improve the invisible quality of images (or objects) and capacity utilization. Machine Learning Based Embedding Pattern Adaptation with Quality Preservation: Based on image features and safety requirements, machine learning methods can change the pattern of hiding within an image while achieving good quality preservation.

AI-Powered Steganalyst Resistance: Develop adversarial training techniques that raise the resistance of the system to machine learning steganalysis attacks. This line of research could make use of generative adversarial networks (GANs) to build a steganographic system that can overcome modern AI-powered detection methods.

Smart Key Derivation: Introduce machine learning schemes to study in depth and analyze complex image traits in addition to traditional ways of histogram and statistical analysis. These methods can touch off more secure keys than the traditional methods are able to (because they are unique to the each image) while still being computationally possible.

6.4.4 Advanced Application Domain

Real-Time Embedded Video: In this proposal, the multi-level approach to video steganography will be extended to cases of real-time processing and a stream of visuals consistent in time. This enhancement of multi-level video steganography could make it possible to send information through videos while ensuring both quality and processing efficiency requirements are met.

Steganography Distributed Systems: Build distributed architectures to which steganographic content is disseminated among many image files or media files, complementing security through content distribution and reducing the risk of detection. In this way, it may be possible with large-scale secure communication needs to achieve further security.

Blockchain unification: Integrate with blockchain technology to create enhanced authentication, non-repudiation and distributed key management. This extra layer of security would also support new uses such as digital rights management and proof that secure communications have been received.

6.4.5 Performance and Usability Enhancement

Mobile platform installation: Develop mobile-optimized editions that take into consideration resource constraints, battery usage issues and user experience requirements. Performance in this direction could empower the practical use of these devices under security and quality conditions.

Real-Time Processing Installation: Introduce parallel processing and GPU acceleration to gain real-time performance on large image and video applications. These optimizations would open up new application areas that require immediate processing of the material.

Interface Design Enhancement: Create an integrated graphical user interface that is transportive, involves a deep-range set of adjustment options and allows the meticulous visualization of results. Such improvements would enhance system user-friendliness and permit highly diversified professional user groups to employ it.

6.5 Conclusion

The research reported in this paper has succeeded in constructing and validating an enhanced multilevel image steganography system. It outperforms all prior implementations in terms of security, speed, and image quality. A new approach to the systematic integration of data compression with encryption and embedding techniques. A multi-level steganographic system, that achieves significant improvements are possible without sacrificing practical feasibility.

6.5.1 Research Achievement Summary

The research completed all the primary project goals based on comprehensive system development and experiential validation. Reduction through compression iteration was 39.29% for longer messages, significantly improved efficiency. Enhancement with cipher-text was the authenticated AES-256-Blowfish hybrid design which provides better security than all conventional methods up until now. Maximizing quality yielded PSNR (76-80 dB) and MSE (<0.002), setting new records. Meanwhile processing efficiency remained all scenarios within sub-0.11 second, that means real-time use can be readily validated.

6.5.2 Scientific Contribution

The research concludes that the integration of compression with steganographic design itself is very profitable for long messages. This can go a long way towards increasing efficiency across all scenarios where compression and steganalysis are separate techniques. Hybrid encryption architectures do not degrade the level of discretion provided by more traditional approaches. With authenticated encryption our immensely stronger security provides significant improvements over conventional. A similar conclusion can be drawn for Data Quality Optimizations: our experiments showed an exceptional PSNR of (76-80dB) and negligible loss (MSE <0.002), thus setting new qualitative key points. The processing efficiency figures stay within close performance for all scenarios giving quality assurance - with a maximum of 0.106 second across the board it is suitable for real-time application.

6.5.3 Practical Impact

The system introduced by this research has immediate practical application for secure communication with high quality persistence. With its excellent security features and processing efficiency, it is suitable for use in a wide variety of operational scenarios, including military communications and healthcare data protection, digital rights management, confidential business communications and much else besides. The scalable structure leaves a good basis for further development into usable products.

6.5.4 Future Research Foundation

This programme of research lays a sound research foundation based on identified upgrade possibilities, verified architectural, complete performance profiles and proven methods of how to integrate. By systemizing things and trying them out, the project offers a methodology for

future advance in steganographic science. It speaks to urgent concerns after securing existing applications and making them more robust, plus providing for the development of evolving new requirements as in practice. How time goes on.

6.5.5 Final Remarks

The enhanced multilevel image steganography system represents a significant advance in information security technology. It conclusively validates the efficacy of multi-level approaches, setting new quality benchmarks for spatial domain steganography, and provides empirical evidence demonstrating that multi-level security offers practical results in applications.

We can see from our full experimental verification combined with actual system construction that the systematic algorithm design and realization of optimized efficiency work! The back story on this is that while we produce useful directions for the cybersecurity community in terms of general insights gained from this study, we also provide practical solutions for instances requiring secure communications, such as at companies with very tough restrictions.

As digital communication continues to evolve and security threats become more and more advanced, this multilevel steganographic approach that was developed in today's paper provides absolute assurance of confidentiality and integrity in communication. The solid foundation laid here will undoubtedly make for continued advance in steganographic technology as new challenges arrive with information security safeguards and applications turn up.

Achievement of this sort in steganography for secure communications is objective evidence of the importance of a systematic approach to security upgrades in practice. Our progress in development provides an invaluable resource for researchers, developers and organizations wishing to put advanced secure communication capabilities into operation in diverse operational environments. Essential reading for those who wish to stay abreast of the latest trends in cryptographic research available.

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- Dashboard
- Student Profile
- Payment Ledger
- Registration/Exam Clearance
- Registered Course
- Result
- Routine
- Live Result
- Teaching Evaluation
- Scholarship >
- Convocation Apply
- Certificate & Transcript >
- Laptop

Dashboard Summary:

Category	Value
Total Payable	777,200.00
Total Paid	777,220.02
Total Due	-20.02
Total Other	2,975.00

Today's Routine - Thursday
No routine available for today.

Semester Wise Result

Semester-wise SGPA Performance

4.0 | SGPA