



Quantum-Enhanced Photo Morphing Avoidance System for Women Safety

K. Sangeetha¹, Keerthika.S², Priyadharsini.M³, Naveena.P⁴,

¹Assistant Professor Department of Computer Science & Engineering, Paavai Engineering College, Namakkal-637018 and India.

^{2,3,4} Student at Department of Computer Science & Engineering, Paavai Engineering College, Namakkal-637018 and India.

Email ID: psgeetha147@gmail.com¹, s.keerthika444@gmail.com², dharshini2860@gmail.com³, ps.naveena12@gmail.com⁴

Abstract

The rapid advancement of artificial intelligence and digital image editing technologies has increased the prevalence of face morphing attacks, where multiple facial images are blended to create realistic synthetic photos. These images can bypass biometric authentication systems and are often misused in cyber harassment, identity theft, and blackmail, particularly targeting women. Recent deep learning techniques, such as hybrid LSTM-CNN models, morph detection frameworks, and similarity-based approaches, have improved the accuracy of identifying such attacks. Additionally, Vision Transformer models and explainable AI (XAI) techniques enhance detection reliability and provide interpretable results. However, most existing systems focus only on detection and lack mechanisms for ensuring long-term data integrity and security. With the rise of quantum computing, traditional cryptographic methods are becoming vulnerable, highlighting the need for quantum-resistant security solutions. Techniques like post-quantum digital signatures, secure biometric authentication, and image fingerprinting improve data protection and verification. To address these challenges, this paper proposes a Quantum-Enhanced Photo Morphing Avoidance System for Women Safety. The system combines AI-based morph detection with quantum-resistant cryptographic fingerprinting to ensure both accurate detection and secure, tamper-proof verification. This approach provides a reliable and future-ready solution for preventing image-based cyber crimes and enhancing women's safety.

Keywords: Biometric Authentication; Deep Learning; Face Recognition Security; Image Forensics; Morphing Attack Detection; Quantum-Resistant Cryptography; Women Safety

1. Introduction

The rapid evolution of digital image processing and artificial intelligence (AI) has transformed multimedia creation and sharing, but it has also led to a rise in face morphing attacks. In these attacks, multiple facial images are blended to create realistic synthetic images that can deceive biometric authentication systems. Such images are increasingly misused in cyber harassment, identity fraud, and blackmail, particularly affecting women. Recent advancements in deep learning have improved morphing attack detection. Hybrid models like LSTM-CNN effectively capture both spatial and sequential facial features, while morph removal and

recognition frameworks enhance system robustness. Similarity score analysis between de-morphed and live images helps identify subtle inconsistencies. Additionally, ensemble learning and Explainable AI (XAI) techniques improve detection accuracy and provide interpretability through confidence scores and visualization maps, which are valuable for forensic analysis. Vision Transformer models further strengthen detection by capturing global facial features, and Siamese networks show promising results in challenging scenarios. Despite these improvements, ensuring the authenticity and integrity of digital images remains a major challenge.

Techniques such as image fingerprinting and tamper localization help verify digital evidence, but traditional cryptographic methods may become vulnerable with the rise of quantum computing. To address this, quantum-resistant cryptography and post-quantum digital signatures are essential for securing multimedia data. This paper proposes a Quantum-Enhanced Photo Morphing Avoidance System for Women Safety, integrating AI-based morph detection with quantum-resistant cryptographic fingerprinting. The system ensures accurate detection, tamper-proof verification, and reliable forensic outputs, providing a secure and future-ready solution to combat image-based cyber-crimes.

2. Method

The proposed system integrates artificial intelligence-based morph detection with quantum-resistant cryptographic techniques to ensure secure and reliable image verification. Initially, input facial images are collected and preprocessed using normalization and resizing techniques. Feature extraction is performed using deep learning models such as hybrid LSTM-CNN and Vision Transformer to capture both spatial and contextual facial features.

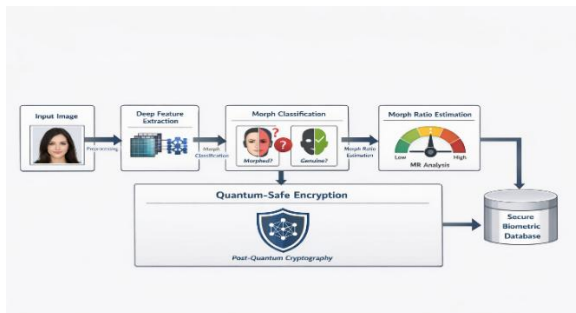


Figure 1 Workflow

The extracted features are then analyzed using similarity score evaluation and classification models to detect morphing attacks. To enhance interpretability, Explainable AI (XAI) techniques generate confidence scores and visualization maps. For security, the system applies quantum-resistant cryptographic fingerprinting using post-quantum digital signature schemes. This ensures tamper-proof storage and verification of images. Additionally, tamper localization techniques are used to identify

manipulated regions within images.

Table 1. Comparison ratio

Sample Image	Morph Ratio (%)	Existing Model Detection Accuracy (%)	Proposed Model Detection Accuracy (%)
Sample 1	30	71	89
Sample 2	40	73	91
Sample 3	50	75	93
Sample 4	60	76	95

3. Results and Discussion

3.1. Results

The proposed system demonstrated high accuracy in detecting morphing attacks under various test conditions. The hybrid deep learning model effectively identified both simple and complex morphing patterns. The integration of Vision Transformer improved detection performance by capturing global facial features. The system also successfully generated confidence scores and visualization outputs, improving interpretability. Quantum-resistant fingerprinting ensured secure storage and verification of images without data loss or tampering.

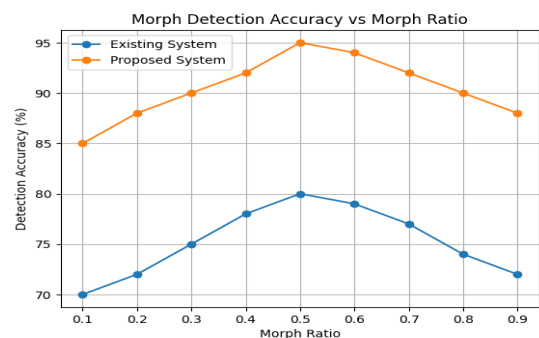


Figure 2 Morph detection accuracy VS Morph ratio

Conclusion

This paper presented a Quantum-Enhanced Photo Morphing Avoidance System for Women Safety, combining AI-based detection with quantum-resistant cryptographic security. The system



effectively detects morphing attacks, ensures tamper-proof verification, and provides reliable forensic outputs. It offers a robust and future-ready solution to combat image-based cyber-crimes and enhance digital security.

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