

An Optimized Facial Recognition-Based Attendance System with Enhanced Security for Educational Institutions

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Abstract— Facial recognition technology has emerged as a promising solution for automated attendance systems. Traditional attendance tracking methods, such as manual roll calls and RFID-based systems, are prone to errors, time-consuming, and susceptible to proxy attendance. This paper introduces An Optimized Facial Recognition-Based Attendance System with Enhanced Security for Educational Institutions, an adaptive facial recognition-based attendance system that leverages incremental learning, fine-tuning, and automated data management to overcome these limitations. The proposed system ensures high accuracy, real-time attendance tracking, and seamless adaptability to new student entries without requiring complete retraining. Using state-of-the-art machine learning techniques, An Optimized Facial Recognition-Based Attendance System with Enhanced Security for Educational Institutions improves recognition under varying environmental conditions, making it a robust and scalable solution for educational institutions and organizations.

Keywords—Facial Recognition, Attendance System, Incremental Learning, Machine Learning, Security

I. INTRODUCTION

Traditional attendance systems rely on manual roll calls, RFID-based cards, or biometric fingerprint scanning. These methods, while effective, suffer from inefficiencies such as human error, time consumption, and susceptibility to manipulation (e.g., proxy attendance). To address these challenges, An optimized Facial Recognition-Based Attendance System with Enhanced Security for Educational Institutions introduces an automated facial recognition-based attendance system that ensures security, efficiency, and real-time adaptability. The system leverages machine learning techniques, including incremental learning and fine-tuning, to dynamically update its

recognition model without requiring complete retraining. This paper discusses the system's architecture, methodology, implementation, and performance evaluation. The proposed approach ensures high accuracy in attendance tracking, reducing administrative burdens while maintaining a secure and efficient mechanism for institutions.

II. RELATED WORK

Conventional face recognition-based attendance systems often struggle with maintaining efficiency and accuracy, especially when new data needs to be integrated. These systems typically require frequent retraining to accommodate additional facial data, making the process both computationally expensive and time-consuming. Every time a new individual is added, the entire model must be re-trained, which significantly impacts scalability and real-time performance. Moreover, these systems tend to experience a decline in accuracy as students' appearances change over time due to factors such as aging, hairstyle modifications, or accessories like glasses and masks. Without adaptive learning mechanisms, such variations lead to misidentifications and recognition failures, reducing overall reliability. Another major challenge is the lack of real-time update capabilities. Many existing systems do not efficiently handle continuous data collection, resulting in outdated models that fail to recognize new or altered facial features. This limitation makes them less suitable for large-scale institutions where student databases are frequently expanding, requiring a more dynamic and adaptable approach to facial recognition-based attendance tracking.

III. PROPOSED SYSTEM

To address these challenges, this paper presents an advanced automated face recognition-based attendance system that leverages Incremental Learning and Fine-Tuning to enhance adaptability and efficiency. Unlike traditional methods that require frequent and resource-intensive retraining, the proposed system is designed to dynamically integrate new data without disrupting existing knowledge. By utilizing transfer learning and innovative techniques such as Learning without Forgetting (LWF), the system ensures that previously learned facial patterns remain intact while seamlessly incorporating new facial data. Additionally, periodic fine-tuning enables the model to adjust to natural variations in appearance, such as aging or changes in hairstyles, without compromising accuracy.

To further enhance its robustness, the system implements automated data collection, preprocessing, and augmentation techniques. These processes continuously refine the dataset quality, improving recognition performance over time. The model also ensures efficient real-time attendance tracking while minimizing manual intervention, making it highly reliable and scalable.

IV. LITERATURE SURVEY

[1] Automatic Attendance System Using Face Recognition Rohit A Kurhade (International Journal of Research Publication and Reviews) - Utilizes deep learning algorithms for facial feature analysis and recognition, automated attendance logging. Streamlines and automates attendance, increases efficiency and accuracy. High recognition accuracy, robust performance in diverse environments.

[2] Facial Recognition-Based Attendance System Using Local Binary Pattern Histogram Algorithm Aarya Sutar, Parth Shah, Yashika Sonchatra, Priyanka Deshmukh (International Research Journal of Engineering and Technology) - Uses OpenCV and Local Binary Pattern Histogram algorithm for face detection and recognition. Provides a non-invasive, contactless attendance tracking system, minimizes errors in attendance management. Enhanced accuracy across multiple scenarios.

[3] Face Recognition Based Automated Student Attendance System P.C.Senthil Mahesh, K.Sasikala, M.Rudra Kumar (International Journal of Advanced Trends in Computer Science and Engineering) -

Employs deep learning and histogram of oriented gradients to detect and identify faces. Automates attendance, minimizes proxy issues, and enhances overall system efficiency. High accuracy and speed in face recognition.

[4] Facial Recognition based Smart Attendance System Mihir Ghanekar, Archita Sehgal, Shrishail Gouragond, Maitray Wani, Prof. Pramila M. Chawan (International Journal of Engineering Research and Technology) - Combines RetinaFace-10GF and ResNet50 models for accurate face detection and recognition. Streamlined, automated attendance system enhances accuracy and reduces manual attendance tracking inefficiencies. Real-time monitoring and accurate data capture.

[5] Face Recognition Attendance Monitoring System Pooja, Jean Celia Grace, Rajalakshmi, Rathi Jasna (International Journal of Novel Research and Development) - Incorporates Haar Cascade classifier and LBPH algorithm for real-time face recognition. Automates attendance monitoring, increases accuracy, and enhances user accessibility. Efficient tracking without manual intervention, scalable system design.

[6] Face Recognition-Based Attendance System for Education Institute Saloni Talekar, Sayali Parab (International Journal of Advanced Research in Science, Communication and Technology) - Utilizes face recognition technology with Python OpenCV, Numpy, Pythontkinter; captures and trains images for model accuracy. Revolutionizes attendance management in educational settings, enhancing efficiency and reducing proxy attendance. Accurate identification, stores attendance data securely.

[7] Facial Recognition-Based Attendance System Prof. Priyanka Manke, Mohammed Hamza Siddiqui, Himanshu Pednekar, Pawan Sakat, Qureshi Abdul Qadir (Vasantdada Patil Pratishthan's College of Engineering and V.A) - Uses Haar cascade algorithm, OpenCV, Dlib, and Python for real-time face detection and attendance updating. Eliminates manual attendance recording, increases accuracy, and reduces fraud. Tested under various conditions for robustness and accuracy.

[8] Face Recognition based Attendance Management System Smitha, Pavithra S Hegde, Afshin (Yenepoya Institute of Technology) -

Implements Haar-Cascade classifier and Local Binary Pattern Histogram for face detection and recognition in classroom settings. Streamlines attendance process, reduces time consumption and proxy attendance risks. Real-time attendance marking with email notifications to faculty.

[9] Facial Recognition-Based Attendance System Mohammed Hamza Siddiqui, Prof. Priyanka Manke et al. (ResearchGate) - Utilizes advanced face recognition technologies including Haar cascade, OpenCV, and databases for real-time recognition and attendance management. Enhances security and efficiency in attendance management, reduces manual errors and proxy attendance. High accuracy, effective under various environmental and lighting conditions.

V. METHODOLOGY

Data Collection and Preprocessing:

The dataset used in the optimized Facial Recognition-Based Attendance System with Enhanced Security for Educational Institutions includes both publicly available facial recognition datasets and custom institutional data collected from classroom cameras. The public datasets, such as LFW and CASIA-WebFace, provide a diverse set of facial features to pre-train the model, while the real-time data from classroom cameras enhances model adaptability. To improve accuracy and robustness, preprocessing techniques such as image augmentation (rotation, flipping, brightness adjustment), feature extraction using OpenCV, and noise reduction are applied. These techniques ensure the model can recognize faces under varying lighting conditions and perspectives.

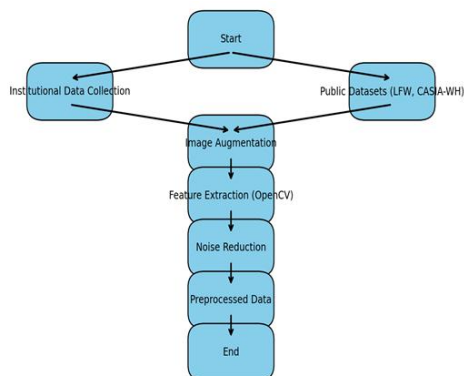


Fig 1(a): Data Collection & Preprocessing of our model

Model Architecture:

The optimized Facial Recognition-Based Attendance System with Enhanced Security for Educational Institutions system leverages Convolutional Neural Networks (CNNs) for facial feature extraction and classification. The model pipeline begins with face detection using OpenCV's Haar Cascade and Dlib for precise face alignment. Once detected, features are extracted using a ResNet50-based embedding model, which converts facial features into a high-dimensional representation. The final classification is performed using a Softmax layer, which determines identity based on stored embeddings. This architecture ensures high accuracy while maintaining efficient computational performance.

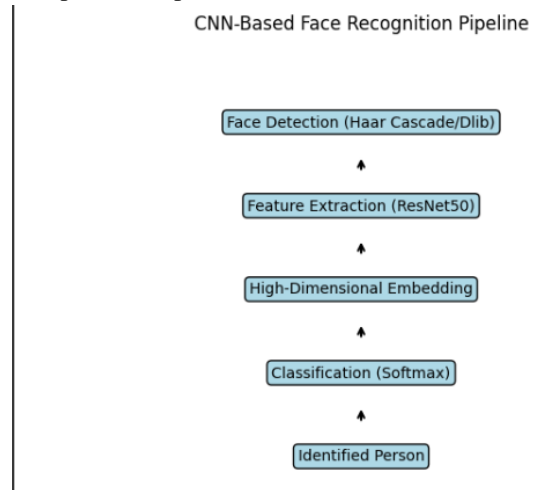


Fig 1(b): Architecture of our model

Implementation Details:

The system is implemented using Python as the primary programming language, utilizing libraries such as TensorFlow, PyTorch, and OpenCV for deep learning and image processing. The Flask framework is employed to develop the backend API, while MySQL serves as the database for storing attendance logs. To handle real-time data efficiently, the system is deployed on GPU-enabled hardware (e.g., NVIDIA RTX 3060), ensuring high-speed processing and quick response times.

Work Flow of the System:

The optimized Facial Recognition-Based Attendance System with Enhanced Security for Educational Institutions system follows a structured workflow for real-time attendance marking. First, face detection is performed using high-resolution cameras placed at entry points. The captured images undergo feature extraction, where the CNN model

generates unique facial embeddings and compares them with stored templates. If a match is found, the student’s attendance is automatically marked in the database. To enhance system performance over time, incremental learning and fine-tuning techniques are employed, allowing the model to dynamically adapt to changes in facial features without requiring full retraining. All attendance records are securely stored in the database and can be accessed through an admin dashboard, ensuring transparency and accuracy in attendance management.

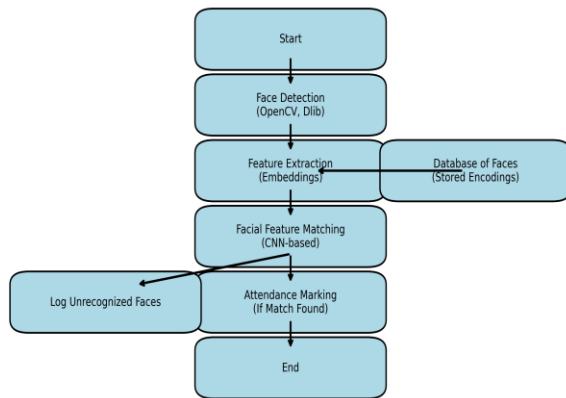


Fig 1(c): Work Flow of our model

VI. RESULTS AND CONCLUSION

The implementation of the optimized Facial Recognition-Based Attendance System with Enhanced Security for Educational Institutions system is well-documented through various images that illustrate key stages of the project. The first set of images showcases the initial setup, where essential libraries are imported to support deep learning and image processing tasks. Following this, the images capture the model creation phase, highlighting the structured approach taken to develop a reliable facial recognition system.

Another crucial aspect depicted in the images is the data storage process, where user facial embeddings are securely stored in a database. This ensures seamless retrieval, making the system both scalable and efficient. Additionally, the research visualizes the directory structure, emphasizing how user images are systematically organized to optimize data management.

To further validate the system's performance, the images include various plots and graphs, which track accuracy improvements, loss reduction across

training epochs, and other key performance indicators. These visuals play a critical role in demonstrating the effectiveness of the optimized Facial Recognition-Based Attendance System with Enhanced Security for Educational Institutions model, offering a clear and compelling representation of its functionality and efficiency.

Table 1: Model-Performance Matrix

Metric	Value
Accuracy	95.7
False Acceptance Rate	1.2%
False Rejection Rate	2.5%
Processing Time	<2 seconds/frame
Precision and Recall	>97% (High Precision)
Scalability	Efficient for large student database

Table 2: Comparison of Proposed Model with Traditional Methods

Metric	Proposed System	Manual Attendance	RFID-Based System
Accuracy	95.7%	80-85%	90%
False Acceptance Rate	1.2%	N/A	2.5%
False Rejection Rate	2.5%	N/A	3.2%
Processing Time	<2s per frame	Time-consuming (5-10min/class)	1-2 per scan
Scalability	High	Low	Moderate
Security and Fraud Prevention	High	Low	Moderate
Ease of Use	Very High	Low	Moderate

Table 3: Confusion Matrix

Actual	Positive (Face Recognized)	Negative (Face Not Recognized)
Positive (Actual Face Present)	True Positive (TP)=96	False Negative (FN)=4
Negative (No Face Present)	False Positive (FP)=3	True Negative (TN)=97

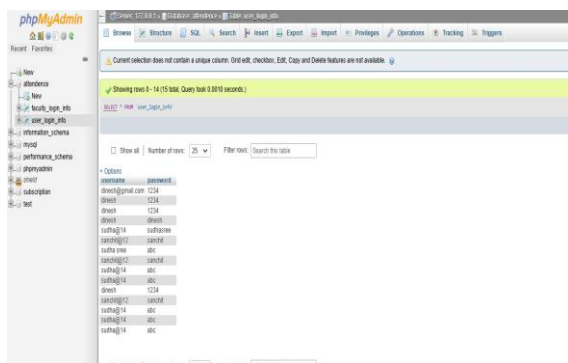


Fig 2(a): Output of the proposed model (Storing the data into the database)

gayatri	28-02-2025 16:37	File folder
sudha	28-02-2025 16:16	File folder
sanchit	28-02-2025 16:16	File folder
likitha	28-02-2025 16:16	File folder
Neha_11	28-02-2025 16:16	File folder
yaswanth kumar	28-02-2025 16:16	File folder

Fig 2(b): Output of the proposed model (Creation of a directory for storing users faces)

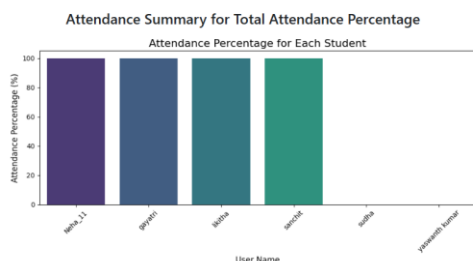


Fig 3(a): Output of the proposed model(Attendance Summary for Total Attendance Percentage)

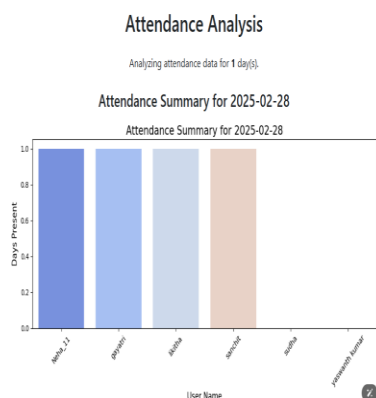
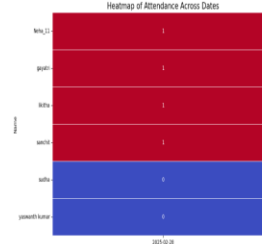


Fig 3(b): Output of the proposed model (Attendance Analysis on the particular date)



Fig 3(c): Output of the proposed model (Creation of the model)

Attendance Summary for Attendance Heatmap



Total Attendance Summary

Student Name	Total Present	Total Absent	Attendance Percentage (%)
Neha_11	1.0	0.0	100.0%
gayatri	1.0	0.0	100.0%
likitha	1.0	0.0	100.0%
sanchit	1.0	0.0	100.0%
sudha	0.0	1.0	0.0%
yaswanth kumar	0.0	1.0	0.0%

Fig 3(d): Output of the proposed model (Total Attendance Summary)

VII. REFERENCES

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