



Enhancing Attendance Management with CNN-Based Face Recognition: A Secure and Efficient Approach

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Abstract

The Facial Recognition Attendance System is an automated solution designed to streamline attendance marking using facial recognition technology. The system employs a webcam to capture real-time images, which are processed by a Convolutional Neural Network (CNN)-based deep learning model for accurate identification. Upon successful recognition, the system records the individual's attendance along with the date and time in an Excel sheet. The application is built using the Flask web framework, providing a user-friendly interface for seamless attendance tracking without manual intervention. This system is particularly beneficial for educational institutions and organizations where efficient and accurate attendance management is essential.

Keywords: Facial Recognition, Attendance System, Convolutional Neural Network (CNN), Flask, Data Augmentation, Image Processing, Automation, Deep Learning, Python, Web-Based Application.

1. Introduction

In today's fast-paced world, traditional attendance marking methods such as pen-and-paper or manual entry are becoming increasingly inefficient and error prone. These approaches are time-consuming, susceptible to inaccuracies, and vulnerable to fraudulent entries. To overcome these challenges, automated attendance systems have emerged as an essential solution, enhancing both efficiency and accuracy. One such system is the Facial Recognition Attendance System, which leverages computer vision and machine learning to automatically identify individuals and mark their attendance. Facial recognition technology enables identity verification based on facial features, eliminating the need for manual identification or biometric devices. This makes the process faster, more reliable, and less intrusive. At the core of this system is a Convolutional Neural Network (CNN), a deep learning model widely recognized for its effectiveness in image classification and facial recognition. CNNs can automatically learn hierarchical patterns in visual data, enabling high accuracy recognition. The model was trained on images of classmates, ensuring precise identification

in a controlled environment. Testing results show an impressive 99% accuracy, demonstrating the system's reliability. The Facial Recognition Attendance System is integrated into a Flask-based web application, making it accessible through a browser interface. Users can interact with the system to start or stop the webcam, process facial recognition in real-time, and generate attendance reports. Once a person is successfully identified, their attendance is automatically logged in an Excel sheet, along with the date and time. This ensures that attendance records are efficiently stored and easily retrievable.

2. Literature Review

Facial recognition-based attendance systems have been a topic of extensive research over the past few decades. Researchers have explored various machine learning and deep learning techniques to improve the accuracy, efficiency, and reliability of these systems. This section presents a review of relevant studies that have contributed to the development of automated attendance systems using facial recognition. Kar et al. (2012) proposed an automated student attendance system integrating face recognition technology based on the Principal Component Analysis (PCA)

algorithm. The system aimed to replace traditional attendance methods by automatically recording students' attendance in a classroom setting. It maintained a log of students' clock-in and clock-out times, enabling faculty members to access attendance records easily. The study emphasized the importance of biometric verification for authentication in computer-based communication systems [1]. Turk and Pentland (1991) introduced the Eigenfaces approach, a pioneering method for face recognition. Their research treated facial recognition as a two-dimensional pattern recognition problem, projecting face images onto a lower-dimensional feature space called 'face space.' The approach utilized eigenvectors to encode variations among different face images, allowing for real-time face recognition. This study laid the foundation for subsequent advancements in facial recognition by demonstrating that new faces could be learned in an unsupervised manner [2]. Tharanga et al. (2013) developed a smart attendance system that combined face detection with Support Vector Machines (SVM) and Convolutional Neural Networks (CNN). Their approach aimed to improve the reliability of automated attendance tracking in educational institutions by employing multiple cameras to capture student images from 360 degrees, minimizing occlusion-related recognition errors. The system utilized Local Binary Patterns Histograms (LBPH) to extract contextual, lighting, and positional variations before employing CNN and SVM for face detection. Additionally, the AdaBoost classification method was used to filter duplicate images and improve accuracy. This multi-technique approach contributed to enhanced precision in student identification and attendance management [3]. Kanti and Sharma (2012) proposed an automated attendance system that utilized PCA for global feature extraction and an Artificial Neural Network (ANN) with backpropagation for face detection. Their model set a threshold value to reject unrecognized faces, ensuring that only registered individuals were marked present. The use of neural networks allowed the system to learn and adapt to variations in facial features over time. This study highlighted the potential of combining PCA and ANN to develop a robust and scalable attendance

management system [4]. These studies collectively demonstrate the evolution of facial recognition-based attendance systems from traditional biometric methods to deep learning powered solutions. The integration of CNNs, SVMs, and neural networks has significantly improved recognition accuracy and system robustness, making automated attendance management more practical and efficient in real-world scenarios. The present study builds upon these foundations by implementing a Convolutional Neural Network (CNN)-based model trained on real time classroom data, ensuring high accuracy and ease of use through a web-based interface.

3. Proposed System

The Facial Recognition Attendance System is designed to automate attendance tracking using deep learning based facial recognition. The system captures images through a webcam, processes them using a Convolutional Neural Network (CNN) model, and logs the recognized individual's attendance into an Excel sheet. This eliminates the need for manual attendance marking, ensuring efficiency and accuracy. The system follows a structured workflow, including image acquisition, preprocessing, model training, and real-time attendance marking. A trained CNN model is used to recognize faces in real time, and a Flask-based web application provides an interface for users to manage attendance records, Figure 1. The system ensures high accuracy and reliability, making it suitable for various environments such as classrooms and offices.

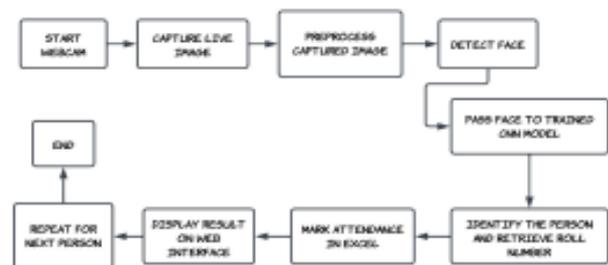


Figure 1 Functional Workflow

4. System Implementation

4.1. Data Collection and Pre-processing

The system begins with image acquisition, where facial images are collected and stored in directories

based on roll numbers. These images are resized to 100x100 pixels and normalized to maintain consistency. The LabelEncoder is used to convert roll numbers into numerical labels, and the dataset is split into training (80%) and testing (20%) sets.

4.2. CNN Model Architecture and Training

A CNN model is designed with multiple convolutional layers using ReLU activation functions to extract facial features. Max-pooling layers reduce dimensionality, while a fully connected layer classifies the recognized face. The final softmax activation function assigns probabilities to each class (roll number). The model is trained using the Adam optimizer and sparse categorical cross-entropy loss. To improve generalization, data augmentation techniques such as rotation, width and height shifting, shearing, zooming, and horizontal flipping are applied. The model is trained for 20 epochs, achieving an accuracy of 99%.

4.3. Real-Time Face Detection and Recognition

The trained model is integrated with a Flask-based web application. The system captures real-time images using a webcam and detects faces with OpenCV. The detected face is processed and passed through the trained CNN model for recognition. If a match is found, the corresponding roll number is retrieved.

4.4. Attendance Marking and Logging

Once a student is identified, their attendance is logged in an Excel sheet with a timestamp. The system ensures that each student is marked present only once per session, preventing duplicate entries.

4.5. Web Interface using Flask

A Flask-based web application provides a user-friendly interface to manage attendance. Users can start and stop facial recognition, view attendance records, and monitor the system through a browser-based dashboard.

4.6. Model Evaluation and Performance

The model's performance is evaluated using accuracy. The system demonstrates high reliability in recognizing faces even under varying lighting conditions and angles. The 99% accuracy rate ensures that the model performs efficiently in real-world scenarios.

5. Results

The Convolutional Neural Network (CNN) trained for facial recognition achieved an accuracy of 99% on the test dataset. The high accuracy demonstrates the effectiveness of the model in correctly identifying individuals based on their facial features. The system was tested in real-world conditions using a webcam to capture and recognize faces. The recognition process operates in real-time, identifying individuals in less than a second. The use of OpenCV for face detection and the CNN for classification ensures smooth and fast performance, making the system practical for daily attendance tracking, shown in Figure 2 & 3.

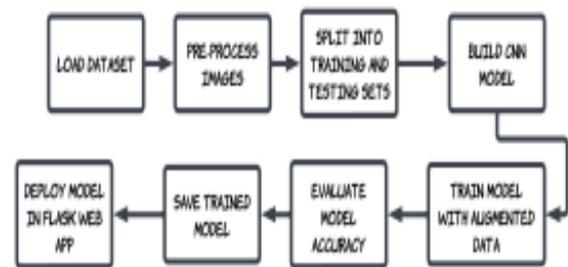


Figure 2 System Implementation



Figure 3 Recognizing Roll Number

The attendance marking system successfully recorded student attendance with no duplicate entries and ensured that each student was marked present only once per session. The generated attendance logs were accurate and matched manually recorded attendance, validating the system's reliability. The integration of Excel sheets for storing attendance data provided an easy way to manage and retrieve attendance records.



	A	B	C	D	E	F
1	Roll No	Date	Time			
2	2309C28	02-02-2025	12:12 PM			

Figure 4 Excel Log

Figure 4, To test the robustness of the system, experiments were conducted under varying lighting conditions and facial orientations. The system performed well even in dim lighting and at different angles, showing its ability to generalize across different scenarios. The Flask-based web interface was tested for ease of use and functionality. Users could easily start and stop facial recognition and view attendance reports with minimal training. The interface was responsive, and the system worked seamlessly on different devices with a webcam. The simplicity of deployment, requiring only a webcam and a browser, makes it a scalable solution for classrooms and organizations. Compared to manual attendance marking and RFID based systems, the Facial Recognition Attendance System significantly reduced the time required for attendance tracking. While manual methods took several minutes per session, the proposed system marked attendance almost instantly. Additionally, unlike fingerprint-based systems, this approach is contactless, reducing hygiene concerns and making it more suitable for large-scale implementation.

Conclusion

The Facial Recognition Attendance System developed in this project demonstrates an efficient, automated, and reliable method for tracking student attendance. By leveraging Convolutional Neural Networks (CNNs) and OpenCV, the system achieves a high accuracy of 99%, ensuring precise identification of individuals in real time. The Flask-based web interface provides an easy to-use platform for managing attendance records, making the system practical for classroom and institutional use. Compared to traditional attendance methods, the proposed system significantly reduces manual effort, minimizes errors, and eliminates the possibility of proxy attendance. The use of real-time face recognition ensures that attendance is marked instantly, improving efficiency and accuracy.

Additionally, the contactless nature of the system makes it a hygienic and scalable alternative to fingerprint-based attendance systems. While the system performs well under various conditions, challenges such as face occlusion, extreme lighting variations, and similar facial features among individuals could impact recognition accuracy. Future enhancements may include multi-angle face recognition, integration with cloud storage for better scalability, and improved model training with a larger dataset to further enhance accuracy and reliability. Overall, the Facial Recognition Attendance System provides a secure, fast, and automated approach to attendance tracking, making it a valuable solution for educational institutions and workplaces.

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