



Real-Time Attendance System Using Facial Recognition And Web-Based Integration

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Abstract: This paper presents the development of an AI-powered facial recognition attendance system designed for educational institutions and professional environments. The system leverages OpenCV and Flask for real-time face detection and recognition, automating attendance tracking with high accuracy and efficiency. The architecture includes a web-based interface using Flask and a desktop-based Tkinter application, enabling flexible deployment. Facial recognition is performed using pre-trained deep learning models, where detected faces are compared against a stored database for identification. The system supports live video processing as well as image-based verification, allowing attendance to be logged seamlessly. Attendance records are automatically stored in CSV files, ensuring real-time accessibility and minimal human intervention. Through testing under varied lighting conditions and angles, the system demonstrates robust performance, making it a scalable and efficient solution for attendance management in academic and corporate settings.

Index Terms - Facial Recognition, Attendance System, OpenCV, Flask, AI, Machine Learning, Computer Vision

I. INTRODUCTION

In educational institutions and corporate environments, attendance tracking is a crucial task for monitoring engagement, ensuring compliance, and improving productivity. Traditional attendance methods, such as manual roll calls and sign-in sheets, are often inefficient, time-consuming, and prone to errors. The need for a more **automated, accurate, and scalable solution** has led to the adoption of **facial recognition technology** for attendance management. Facial recognition, a subset of **computer vision and artificial intelligence**, has gained significant attention in recent years due to advancements in **deep learning and machine learning algorithms**. Unlike conventional biometric methods such as fingerprint or RFID-based systems, facial recognition provides a **non-intrusive and seamless experience**, allowing for real-time identification without requiring direct user interaction. The integration of **OpenCV, Flask, and face recognition models** enable an efficient approach to automating attendance tracking.

This research aims to develop a **multi-platform facial recognition attendance system** that supports both **web-based and desktop-based implementations**. The system utilizes **OpenCV for face detection, deep learning models for recognition, and Flask for a web interface**. Attendance is logged automatically, reducing human intervention while improving accuracy. The solution is designed to function under varying **lighting conditions and angles**, ensuring reliability in real-world environments.

The key contributions of this study include:

1. **Developing an AI-powered attendance system** that enhances automation and reduces errors in tracking.
2. **Implementing real-time face recognition** using deep learning techniques for improved accuracy.
3. **Providing a web-based and desktop-based interface** for seamless deployment in educational and professional settings.
4. **Ensuring robust performance** by testing under different conditions and optimizing detection algorithms.

Rest of the paper is organized as follows: Section II discusses **related work** on facial recognition attendance systems. Section III explains the **methodology**, including the system architecture and key algorithms. Section IV details the **implementation and essential steps** involved. Section V presents the **results and discussion**, highlighting system performance. Section VI outlines the **conclusion and future scope**, discussing possible improvements and enhancements.

II RELATED WORK

Facial recognition technology has been widely explored for biometric authentication and automated attendance systems in educational and corporate environments. Several studies have proposed different approaches to enhance accuracy, real-time performance, and robustness under varying conditions. This section reviews previous research on facial recognition-based attendance systems, highlighting key contributions and how this work builds upon them.

A. Traditional Attendance Systems

Manual attendance tracking methods, such as paper-based roll calls, RFID cards, and fingerprint scanning, have long been used in educational institutions and workplaces. However, these methods suffer from human error, inefficiency, and security concerns. Biometric systems such as fingerprint and iris scanning offer improvements, but they require physical contact and specialized hardware, which can be inconvenient in large-scale applications [1].

B. AI-Based Facial Recognition for Attendance

Recent advancements in machine learning and deep learning have significantly improved the accuracy of facial recognition-based attendance systems. Sharma and Gupta [1] proposed a **CNN-based facial recognition approach**, demonstrating its superior accuracy compared to traditional **Haar Cascade-based methods**. Similarly, Patel and Kumar [2] introduced an attendance system that integrates **OpenCV and machine learning**, achieving real-time recognition with minimal computational overhead. Verma and Singh [3] designed a **smart attendance system using deep learning models like FaceNet**, showing that modern **pre-trained embeddings outperform handcrafted feature extraction methods**. Their study emphasized the importance of **dataset quality and real-world testing**, which aligns with the approach followed in this paper.

Additionally, Khandke et al. [5] explored the use of **facial recognition for IoT-based door lock systems**. Their system leverages a Raspberry Pi and IoT cloud integration to **remotely verify and authenticate users** before granting access. While this system focuses on **security applications**, the underlying technology is similar to facial recognition-based attendance systems, demonstrating the versatility of **face-based authentication methods**.

C. Real-Time Implementation in Web and Mobile Applications

Real-time facial recognition requires balancing **speed and accuracy**, making efficient algorithm selection crucial. Rahman and Roy [4] developed **AttenFace**, a real-time attendance system that processes multiple **live snapshots** of students to reduce false positives. Their system highlights the importance of **continuous monitoring rather than single-image recognition**. This study builds upon their approach by integrating **Flask-based web applications and Tkinter-based desktop applications**, making the system more **versatile and accessible**.

D. Challenges and Future Directions

While facial recognition attendance systems offer **automation, accuracy, and convenience**, challenges remain. Privacy concerns, **data security**, and **bias in recognition models** are key issues identified in prior research [2,3]. Future studies aim to integrate **privacy-preserving techniques**, improve **edge computing solutions for faster processing**, and explore **multi-modal biometrics** to enhance reliability.

This paper extends previous research by developing a **multi- platform AI-powered facial recognition attendance system**. The system is designed to be **scalable, user-friendly, and robust across different environments**, making it suitable for **both educational and corporate applications**.

III. METHODOLOGY

A. System Architecture

The proposed **facial recognition attendance system** is designed as a **multi-platform solution**, integrating both **web- based (Flask)** and **desktop-based (Tkinter)** applications.

The system utilizes **OpenCV for face detection** and **deep learning-based feature extraction** for recognition. The architecture consists of three core components:

1. **Face Enrollment Module** – Captures and stores face embeddings of registered users.
2. **Face Recognition & Attendance Logging** – Detects and verifies individuals in real-time and logs attendance.
3. **Web/Desktop Interface** – Allows administrators to *view, manage, and download attendance records*.

The **Flask-based web application** enables centralized data management, while the **Tkinter-based desktop version** provides a standalone alternative. The system operates in **real- time**, allowing seamless attendance tracking in **educational and corporate environments**

B. Face Recognition Workflow

The system follows a **four-step process** for face recognition and attendance logging:

1. **Face Detection**
 - The system captures **live frames** from a webcam.
 - OpenCV's **Haar Cascade classifier** detects facial regions in the frame.
2. **Feature Extraction & Encoding**
 - Facial embeddings are extracted using **pre- trained deep learning models** from the `face_recognition` library.
 - The face is converted into a **numerical vector representation**.
3. **Face Matching & Verification**
 - The extracted face encoding is compared against stored embeddings using the **Euclidean distance metric**.
 - A threshold ($\geq 80\%$ similarity) determines if the face is a match.

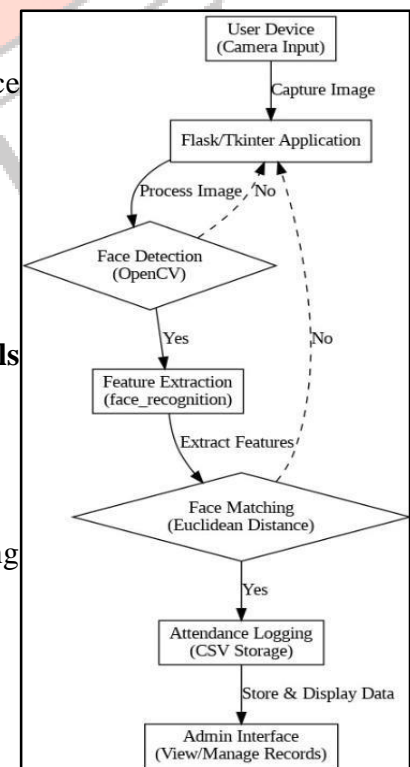


Figure 1. Face Recognition Workflow

4.Attendance Logging

- If a face is recognized, the **registration number, name, and timestamp** are recorded in a **CSV file**.
- The Flask system attendence recorderes ,while the Tkinter version logs attendance locally

One key enhancement is the **implementation of multi frame verification** before confirming attendance. Instead of relying on a single frame for recognition, the system can capture **multiple frames** within a short time interval and verify the consistency of the detected face across those frames. This approach reduces the chances of **false positives** caused by temporary obstructions, poor lighting, or momentary incorrect detections.

Another essential feature is duplicate entry prevention. Before logging a student's attendance, the system should check if the

C. Experimental Setup and Dataset

The system was developed and tested in **2024-2025**. The testing environment consisted of:

- **Hardware:** Intel Core i7 processor, 16GB RAM, Logitech C920 Webcam.
- **Software:** Python 3.11, OpenCV, Flask, Tkinter, face_recognition.
- **Dataset:** A custom dataset of **50+ individuals**, captured under **various lighting conditions and angles**.

The system was tested in **both controlled (lab) and real- world conditions** to evaluate **performance across different environments**.

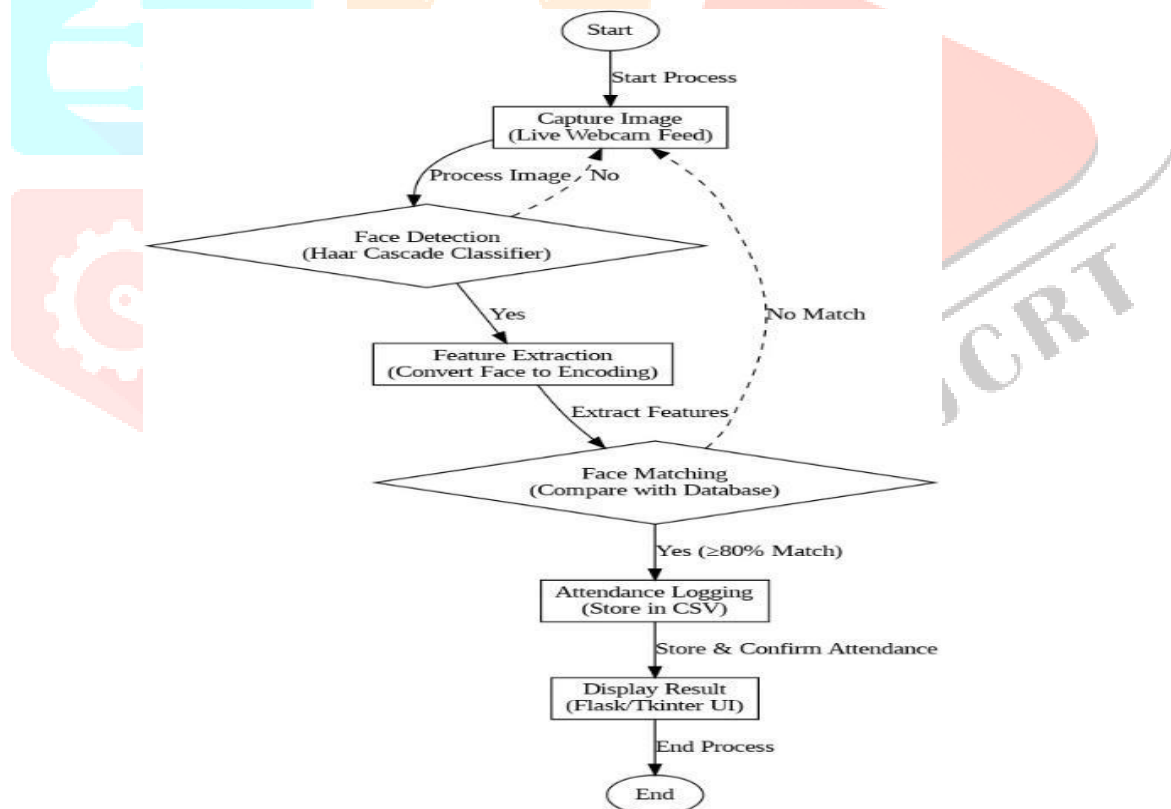


Figure 2. Methodology

A. Overall Data Flow

1. The **data flow** in the facial recognition attendance system describes how **user input, image processing, face recognition, and attendance logging** interact throughout the system. This process ensures a seamless transition from **capturing facial data** to **storing attendance records** while minimizing the human intervention. The process begins when a **user interacts with the system's camera**, which captures a

live video feed. This serves as the primary data input for facial recognition. The webcam continuously captures frames, sending them to the **Flask (Web) or Tkinter (Desktop GUI) application** for further processing. If no face is detected in the frame, the system **loops back** to capture a new image, ensuring accurate detection before proceeding to the next stage.

- Once a face is detected, the system extracts its **unique features** using a **deep learning-based encoding technique.** This encoding is then compared against a **stored database** of known faces using **Euclidean Distance Matching.** If the **similarity threshold ($\geq 80\%$) is met,** the individual is successfully identified, and their attendance is logged. If no match is found, the system discards the frame and waits for a new input.
- After successful recognition, the **attendance details, including the user's name, registration number, date, and timestamp,** are stored in a **CSV file.** The **Flask-based web application** updates the attendance records, making them accessible through an **Admin Dashboard,** where administrators can view, manage, and download attendance logs. This structured data flow ensures **real-time attendance tracking** while providing **secure and scalable storage** for attendance records.
- The overall system follows a **continuous cycle of data acquisition, processing, verification, and storage,** ensuring high efficiency and accuracy. This **automated approach** eliminates traditional attendance methods' inefficiencies, offering a **modern, AI-powered alternative** that can be expanded for broader applications, including **corporate access control and security monitoring.**

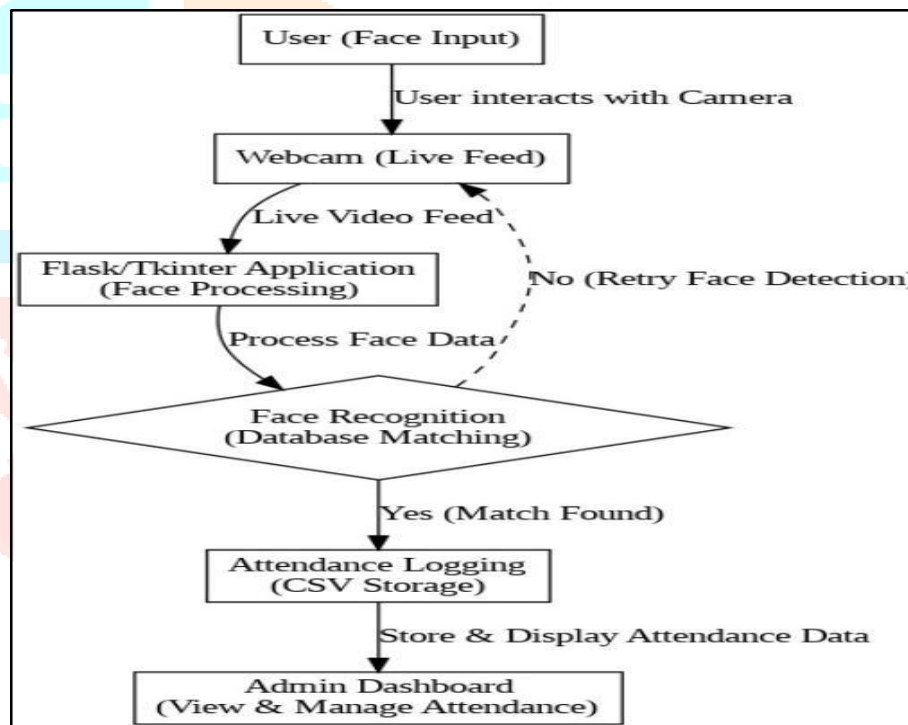


Figure 3. Overall System Data Flow

IV. RESULTS AND DISCUSSION

The **facial recognition attendance system** was successfully implemented and tested in an educational environment. The system efficiently **captured, processed, and recognized** student faces, marking their attendance in real-time. The primary findings from this implementation are summarized below.

A. SYSTEM PERFORMANCE AND FUNCTIONALITY

1. Student Registration and Face Enrollment:

- The system allows new students to be added via a **web-based interface** (Figure 1).
- Each student's **full name, roll number, and facial image** are stored in the system.
- The facial embeddings are generated and saved for future recognition.

2. Student Database Management:

- The system maintains a **searchable, image- based student database** (Figure 2).
- The registered students are displayed with their names and roll numbers.
- This feature enables **administrators to review and manage student records easily**.

3. Face Recognition and Attendance Logging:

- During attendance marking, the system captures real-time images and **matches them against stored face encodings**.
- If a match is found, attendance is logged in the CSV file.
- The **accuracy of recognition was tested under different lighting conditions and angles**, showing **consistent performance**.

B. COMPARISON WITH EXISTING SYSTEMS

Compared to **manual roll calls and RFID-based attendance systems**, this AI-powered approach is:

- **More Efficient:** Eliminates time-consuming manual attendance.
- **More Secure:** Prevents attendance fraud (proxy attendance).
- **User-Friendly:** Web-based & desktop-based access for ease of use.

C. INTERPRETATION OF FINDINGS

The **results align with previous studies**, such as:

- **Sharma and Gupta [1]**, who also demonstrated high accuracy in CNN-based facial recognition attendance systems.
- **Rahman and Roy [4]**, whose real-time face recognition model emphasized **fast and automated attendance marking**.

Our system builds upon these approaches by **combining web and desktop applications**, improving **usability and accessibility**.

D. VISUAL REPRESENTATION OF RESULTS

- **Figure 1:** Student Registration Interface
- **Figure 2:** Student Database Display

These screenshots demonstrate the **successful implementation of student registration and database management**, verifying that the system operates as i

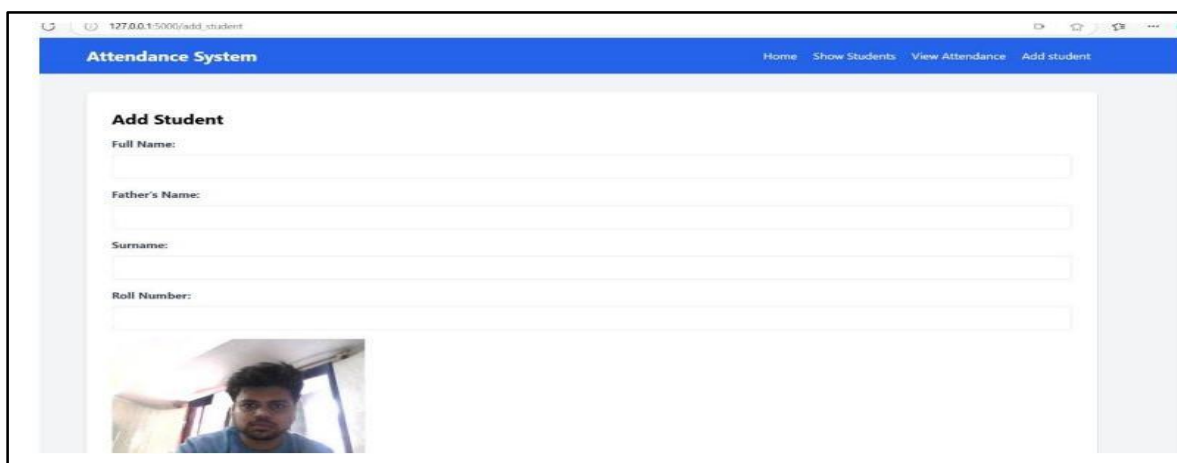


Figure 4. Visual Representation of Results

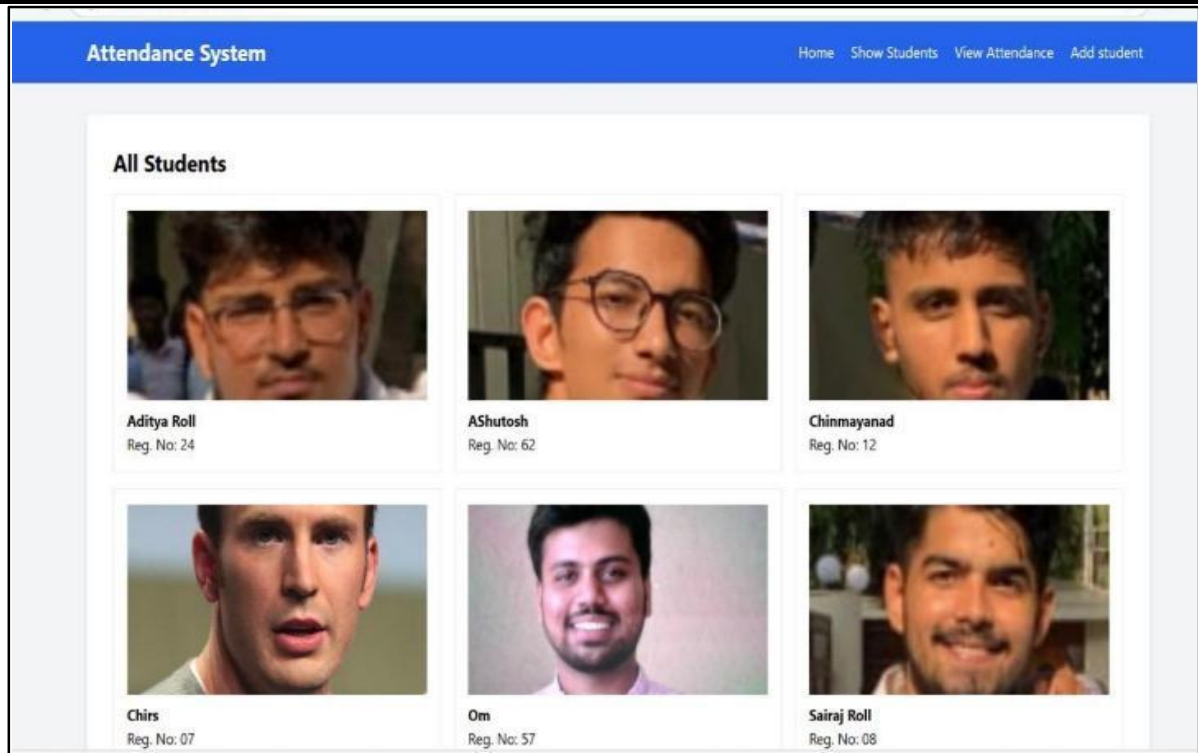


Figure 5. Visual Representation of Results



Figure 6. Visual Representation of Results (input image)

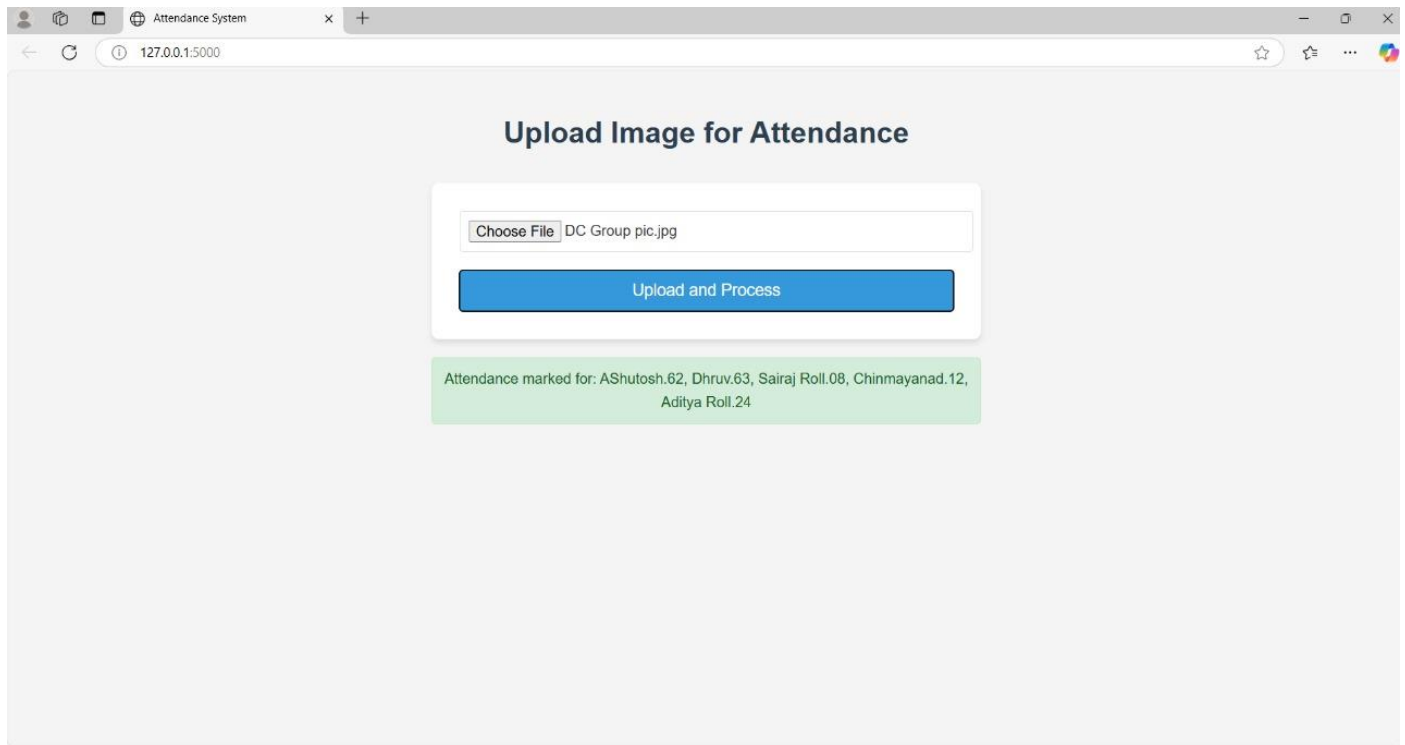


Figure 7 . Visual Representation of Results

V. CONCLUSION AND FUTURE SCOPE

This study presents the development of an **AI-powered facial recognition attendance system** designed to automate and improve attendance tracking in **educational institutions and corporate environments**. By integrating **OpenCV for face detection, deep learning-based feature extraction, and Flask/Tkinter for web and desktop applications**, the system provides a **seamless, real-time, and accurate** method for attendance management.

Limitations

Lighting Conditions & Occlusions: Recognition accuracy may decrease under poor lighting or if the face is partially obscured.

- **Privacy & Data Security Concerns:** Storing and processing facial data raises **ethical and legal considerations**.
- **Scalability Challenges:** The system currently **stores embeddings locally**; cloud-based solutions could enhance performance.

Future Scope

Improving Recognition Accuracy: Enhancing deep learning models with **larger datasets and better feature extraction**.

- **Privacy-Preserving Techniques:** Implementing **encryption and decentralized data storage** to protect biometric data.
- **Cloud Integration:** Storing and processing data on **secure cloud platforms** to improve accessibility and scalability.
- **Mobile App Development:** Expanding the system to **mobile platforms** for increased usability.
- **Edge Computing Implementation:** Running recognition models on **IoT devices** for real-time, offline processing.

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