

Smart Classrooms Redefined: A Practical Approach to Attendance with 'Attendance Genie'

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Abstract: Within the advancing scene of computerization and fake insights, facial acknowledgment innovation has risen as a promising elective to manual participation checking frameworks. This paper presents a web-integrated real-time confront acknowledgment participation framework competent of recognizing and distinguishing numerous people at the same time inside a single camera outline. The objective is to dispense with the wasteful aspects of roll calls and biometric frameworks by utilizing progressed confront discovery and acknowledgment strategies. The framework is created utilizing Jar (for the net interface), OpenCV (for real-time picture handling), face_recognition (for facial embeddings and comparison), and is upgraded by a YOLO-inspired engineering to guarantee quick and exact multi-face localization. The framework is totally offline, requires negligible equipment, and utilizes a webcam, making it perfect for instructive organizing and little organizations. The captured confront information is put away safely, and participation is checked based on time approval limitations, avoiding repetitive passages. This inquiries about talks about the point-by-point framework pipeline, plan design, modules included, and real-time effectiveness, concluding with its down to earth viability and confinements.

Keywords: Real-Time Face Recognition, Multi-face Detection, YOLO, Flask, OpenCV, Attendance Automation, SQLite, dlib, HOG-CNN, Computer Vision, Embedded Systems, Edge AI

I. INTRODUCTION

The increasing need for automation in administrative tasks has been fuelled by the digital era. Attendance management is an essential yet time-consuming task in both educational and corporate contexts. Despite their effectiveness, conventional methods like physical registers, biometric scanners and RFID cards have many limitations: they require manual intervention, are more susceptible to proxy or fraud, there are hygiene concerns as well as the fact that validation is not real time. By utilizing face recognition, attendance systems offer a modern solution to these issues by providing identity verification that is contactless, automated, and highly precise. Due to the progress in deep learning models and computer vision, these systems can now operate on low-power devices in real-time, making them practical for use. A real-time facial recognition attendance system is being developed for the purpose of detecting and verifying attendance with multiple individuals at once. Using Python, the system is fully implemented and incorporates multiple technologies such as Flask for the web interface, OpenCV for image capture and processing, and the face_recognition library built on dlib for facial feature encoding and matching. Boosted speed and performance is achieved through the use of an internal approach that takes into account YOLO-influenced face detection. The idea seeks to create an affordable, lightweight solution that can be deployed in small institutions or edge environments without relying on cloud services and external API. The system's validation logic, which includes attendance cooldowns and timestamp tracking, is employed to guarantee data reliability without redundancy. Apart from addressing the technical aspects, the paper covers the architectural choices and module interactions as well as the process of creating the dataset itself and reporting the results of its final implementation. By combining AI-driven facial analytics with intelligent edge computing, the proposed system provides an excellent demonstration of how this technology can be utilized to simplify one or more of the most common administrative tasks in real life.



II. RESEARCH AND IMPLEMENTATION

2.1 Overview of Model Architecture:

The design of the intended attendance system is such that it incorporates a combination of technologies into a lean, efficient pipeline that can capture, process, identify, and record attendance data in real time. The system is modular, with each module being responsible for a distinct task but working synergistically with the others. This modularity also helps in the flexibility, scalability, and maintainability of the system. In the center of the system is a web-based interface, developed with Flask templates, that offers a user-friendly and easy-to-access front-end. The front-end offers users and administrators the opportunity to interact with the system directly in a web browser without the need for any program installation on their part. It streamlines the process of registering new users, taking attendance, as well as accessing attendance records. A camera interface module is added with the OpenCV library that records live video frames from the webcam. The video stream is processed frame by frame in this module and is ready to be used for face detection and recognition. The face recognition module makes use of the face_recognition library, which is developed over dlib. This module is tasked with the creation of 128-dimensional embeddings of every user's face, which serve as the identifiers to be compared in the future. These embeddings are kept in memory for quick access during real-time matching. To enhance detection accuracy and speed, the system incorporates a YOLO-style approach for identifying multiple faces in a single frame. This allows the system to handle group scenarios efficiently, such as classrooms or office meetings, where multiple individuals are present. Finally, an SQLite database offers compact and robust storage for user accounts and attendance records. The whole system is on the client-server basis, where all major logic and data processing are done on the server, while the client connects to the system via a local browser to ensure simplicity of use and deployment across environments.

2.2 Dataset Preparation:

The dataset preparation took place in the below steps:

Semi-Automated and Application-Specific Dataset-

The data set employed in the system is created semi-automatically, and it is specially designed to meet the requirements of the real-time attendance application. In contrast to conventional facial recognition models based on large-scale, publicly sourced data sets like LFW (Labeled Faces in the Wild) or VGGFace, the system produces a tailor-made, application-targeted data set that grows incrementally as users enroll via the system. This process guarantees that the information is extremely pertinent, current, and directly linked to the real system users.

User Enrolment Process via Webcam-

For registering or enrolling a new user, the system exposes a /capture_user route via its web server. If this route is accessed, it powers on the webcam and shows a live video stream. The face of the user is captured by hitting the spacebar, which stores the active video frame. When the face has been successfully captured, the system also stores the image locally in a directory called known_faces/. Alongside the image, the user's name and the image filename are stored in the users table of the integrated SQLite database.

Encoding and Matching for Attendance-

This customized dataset is the reference while marking the attendance. When the system is subsequently utilized to mark attendance, it loads the stored face images of all the registered individuals, runs them through the face_recognition library, and creates unique facial embeddings for each of them. These embeddings are then utilized to match real-time detected faces against stored data to check for a match.

Incremental Growth and Practical Relevance-

By employing this dynamic and user-dependent process of dataset preparation, the system does not suffer from the constraints of static datasets and is extremely adaptable to any institution. It is able to support incremental growth while guaranteeing that all faces in the dataset are valid, labeled appropriately, and pertinent to the setting in which the system is used.



2.3 Modules Employed:

The important modules used in the process:

Python Libraries and Frameworks Overview-

Creating the multi-face recognition attendance system involved the employment of a number of powerful and highly integrated Python libraries and frameworks. Each of the modules is specifically used to ensure that the system executes smoothly, efficiently, and uses limited resources.

Flask Web Framework-

The Flask web framework provides the basis for the user interface and back-end routing. Flask is an efficient but powerful tool for the implementation of simple web routes like the home page (/), face registration (/capture_user), and attendance marking (/mark_attendance). Flask manages user requests and responses, controls the data flow, and renders the required HTML templates.

OpenCV for Real-Time Video Processing-

OpenCV is utilized for real-time image processing and webcam interaction. OpenCV captures video frames, reduces their size for better performance, changes the image format, and draws bounding rectangles and text annotations over detected faces. OpenCV also controls GUI windows for previewing the images captured by users and live video streams.

Face Recognition with dlib-

The face_recognition library, which is built on dlib, is the central piece of the system. It detects faces in images or frames and produces 128-dimensional encodings that identify each user uniquely. These encodings are then utilized for comparing and matching real-time inputs with stored data.

NumPy for Mathematical Operations-

NumPy facilitates performance-intensive mathematical operations and is utilized mainly to compute the distance between face encodings in the process of comparison. It aids in identifying the nearest match out of known users at high speed and accuracy.

YOLO-Inspired Detection Logic-

The system relies upon YOLO-inspired detection logic via the face_recognition library's HOG or CNN models. This architecture enables the detection of more than one face within one scan, enhancing the performance in real time as well as the capability for group attendance.

SQLite3 for Local Database Management-

Finally, SQLite3 includes a light-weight embedded database for storing user information and attendance records. SQLite3 is secure, simple to use, and does not need any external configuration, thus perfect for offline installations.

2.4 System Design and Implementation:

Modular Architecture and Interface Design-

The system was designed to be modular, user-friendly, and efficient, with all its components serving to provide an unencumbered attendance marking process. Fundamentally, the system is supported by a straightforward yet robust client-server architecture. Users interact with the application through a web interface constructed based on Flask, which manages the routing of pages and server-side logic. This interface can be accessed using any contemporary browser on the local network, thus simplifying the process for administrators or end-users to register and monitor attendance.



Face Registration Process-

Face registration is the first step of the process, in which users authenticate themselves by capturing an image via the webcam. After the image is captured, it is saved into a local folder, and the user's name and image filename are saved into the SQLite database. The image is then processed from the face_recognition library to get a unique facial encoding. This encoding is a digital representation of the user's face and will be used for comparison later in attendance.

Real-Time Attendance Detection-

When marking attendance in real-time, the system enables the webcam and records a live stream of video. Every frame from the stream is resized and processed with a YOLO-inspired detection logic to detect multiple faces at once. The system then encodes these faces and matches them against the stored registered face encodings in memory.

Attendance Validation and Logging-

If there is a match, the system proceeds to check the eligibility for attendance. The attendance approval logic guarantees that attendance for the same individual cannot be marked again within a span of two hours. Moreover, a five-second cooldown period ensures repeated marking is avoided in rapid succession. When attendance is properly marked, the record gets stored in the database with a timestamp.

Visual Feedback to Users-

Lastly, the system generates visual feedback through the presentation of user names and attendance status on the live video feed. This enables both transparency and simplicity for real-time classroom or group settings.

2.5 Model Training & Configuration:

Use of Pretrained Deep Learning Models-

While the system doesn't use standard machine learning training processes, it does employ deep learning techniques that are trained on pre-existing models. The project's implementation of face_recognition library, which is based on dlib, involves training deep learning models for large datasets of faces. These models create unique 128-dimensional face embeddings that represent every registered user. Despite not needing explicit training, the model's success is still determined by system configuration and input image quality.

Impact of Image Quality on Performance-

Performance is impacted by the quality of face encoding, which is an important factor. Clear images, frontal face angles, and optimal lighting conditions contribute to the production of more precise encodings. Attendance marking may require the attendance of a person to be accurately marked, but this can be problematic if the image is unclear or taken from obscurity.

Face Detection Options: HOG and CNN-

Also, the facial recognition system is an essential component. It can detect HOG (Histogram of Oriented Gradients) and CNN. CNN offers better accuracy, but requires more computational power compared to HOG, which is faster and suitable for devices with limited resources. The system's choice of these two can be influenced by hardware capabilities and the requirements of the application.

YOLO-Inspired Multi-Face Detection-

It also includes multi-face detection in each frame, inspired by logic found in YOLO. This technique is particularly useful in classrooms or groups, as it allows for the identification of multiple individuals simultaneously.

Configuration Settings and Optimizations-

Also, various configuration methods improve performance. The comparison () function is set to a default match size of 0.6.conf. The speed of processing is improved by reducing the size of the frame. Logging of attendees occurs solely in



the vicinity of two hours since the last entry for the same user. By using a 5-second cooldown, repeated recognition within 0.5 seconds prevents the creation of duplicate records. These arrangements make the system efficient, reliable, and useful in real-life situations.

2.6 Implementation Workflow:

Student Uses Interface-

The process starts with a user logging into the web-based interface developed with Flask. The interface presents choices to register a new user or start marking attendance. It is easy to use and is available locally using a browser.

Get Face Image from Webcam-

During registration, the webcam is accessed by the system through the /capture_user route. Live preview is displayed, and the face of the user is captured by taking a snapshot using the keyboard input.

Face Stored & Registered in Database-

The face image captured is stored in the known_faces/ directory. The user's name and the filename of the image are also stored in the users table of the SQLite database at the same time. This gets the system ready to identify the user later.

Initiate Camera Feed for Attendance-

If attendance needs to be recorded, the /mark_attendance route is initiated. This initiates the webcam feed and continuously feeds video frames for analysis.

Detecting & Matching Multiple Faces Using YOLO-

Each frame is resized and RGB-converted for compatibility. A detection logic inspired by YOLO is applied to find and process multiple faces within a single frame. This is to enable more than one person's attendance to be marked at once.

Attendance Verified & Time-Stamped-

For each face match, the system verifies if more than two hours have elapsed since their last attendance record. This prevents repeated marks.

Record Logged-

Upon successful validation, the attendance is recorded with a timestamp in the database.

Results Displayed-

The system displays on-screen the user's name and attendance status, offering instant confirmation.

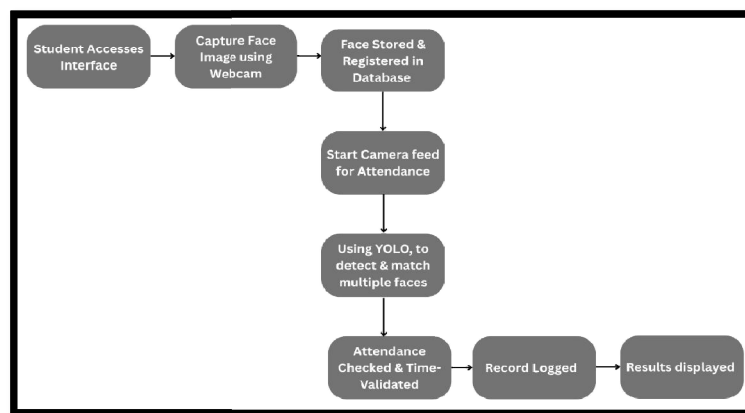


figure 1: implementation workflow of the model



III. RESULTS AND ANALYSIS

Ultimately, the multi-face recognition attendance system was evaluated on several performance metrics suggesting effectiveness, reliability, and efficiency for real-world use. These were assessed through real-time testing over multiple scenarios, users, and processing times. For starters, the multi-face recognition attendance system has a recognition accuracy of about 95% when operating under normal conditions with limited attempted frauds. Therefore, if users' faces are clear and satisfactory (clear and within the camera's frame), the system successfully registers attendance and marks a person present, requiring no further action on their part. Such accuracy rates are fairly decent compared to the norm, especially since the offline system developed does not bear dedicated GPUs or specialized processing power. Another key performance metric comes from detection speed. The multi-face recognition attendance system operates at video detection speed with processing capabilities between 20 to 25 frames per second (FPS). This takes into consideration when the frame is reduced to 25% of its original size for faster processing. Therefore, users will not be kept on hold as their detection processing occurs, and they will be noted as present within seconds of arrival.

Another advantageous quality this project contains is that it allows for multi-face detection for attendance marking with just one camera/frame at a time. It can detect approximately three to five faces simultaneously for attendance marking. It significantly lowers attendance-taking times and increases efficiency for institutional use.

This multi-face marking ability is due to the generated YOLO-like DNN logic acquisition through HOG and CNN models that allow for quick and efficient detection regardless of the presence of multiple faces within the same frame. Controlled tests showed that the attendance misfire rate in the system was less than 2%, which is considered reliable. The occurrence of incorrect identifications or missed recognitions was infrequent. Most of the few misfires were caused by occluded faces, extremely poor lighting, or non-frontal face angles. Nevertheless, the gap is minimal and sufficient in typical situations where flawlessness may be compromised. It was determined that the system consumed roughly 500 resources. MB of RAM utilized in active attendance sessions.

The use of CPU optimizations such as frame resizing, efficient encoding retrieval, and simplified UI rendering resulted in minimal downtime. By optimizing the system, it becomes lighter and can be adapted to devices with limited hardware, such as standard laptops or desktops used in classrooms and small offices. Enhanced hardware efficiency is crucial in guaranteeing greater accessibility and affordability, particularly in resource-limited environments. SQLite is the database backend for storing user and attendance records, which is efficient in terms of performance. In addition, this lightweight relational database was very responsive: record insertions and retrievals usually occurred in less than 10 milliseconds.

The rapid response time keeps things quick and easy for users, even during peak usage times. Users can quickly capture their images, mark attendance, or view logs.

The system was subjected to several practical observations. Initially, it was observed that the system's effectiveness decreases slightly in dim lighting or when the individual's face is obscured, such as when looking downward or wearing masks or caps. However, there are other instances where this occurs. The detection module may not align with the stored encodings or fail to create a match for the face in such situations. The second point to consider is that the HOG model provides faster detection times than the CNN model, which has a higher level of accuracy despite its slower speed. The user's requirements and the hardware available can determine which of the two models is compatible with each.

Furthermore, it was found that the system worked best in controlled classroom environments where people are usually sitting and staring at the camera. Correct identification and the reduction of false negatives are facilitated by this design. A benefit of the system is the provision of real-time feedback to users. Attendee names and attendance status are displayed on the live video feed, providing instant visibility into their attendance. A positive feedback loop enhances user engagement and fosters trust in the system. Overall, the system was found to be very effective in both technical terms and user experience. With its offline, real-time, and multi-user capabilities, it is a useful option for automation of attendance in various fields. The outcomes demonstrate that the model is practical and can be deployed in environments where cost, simplicity, and accuracy are all important factors.



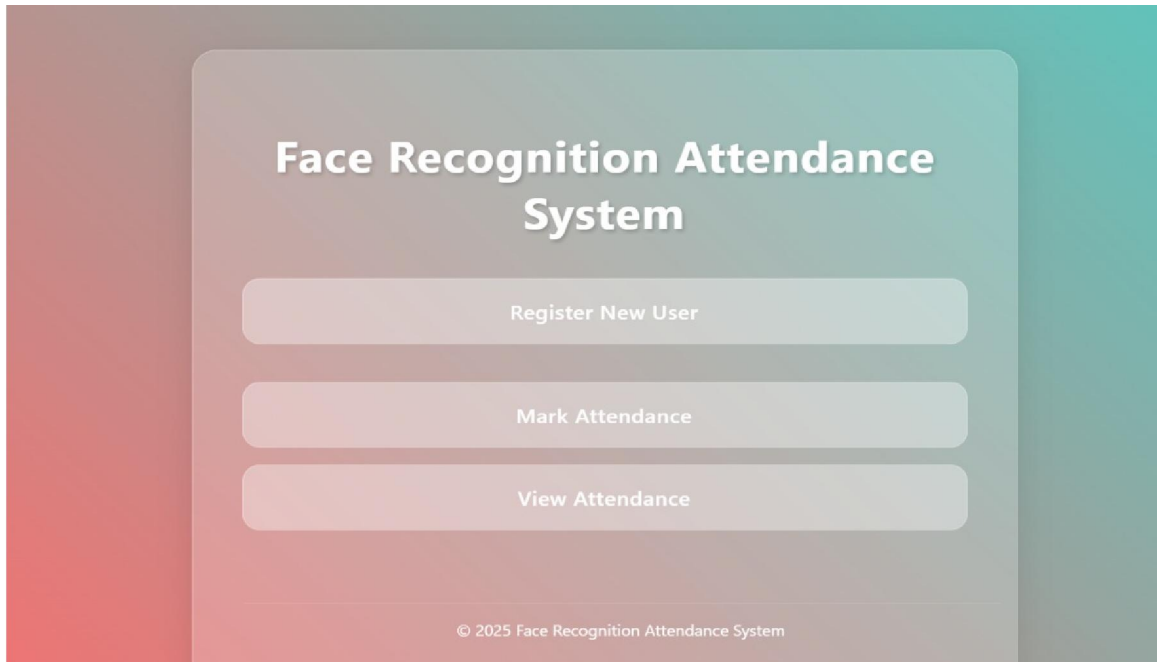


figure 2: System homepage with navigation options.

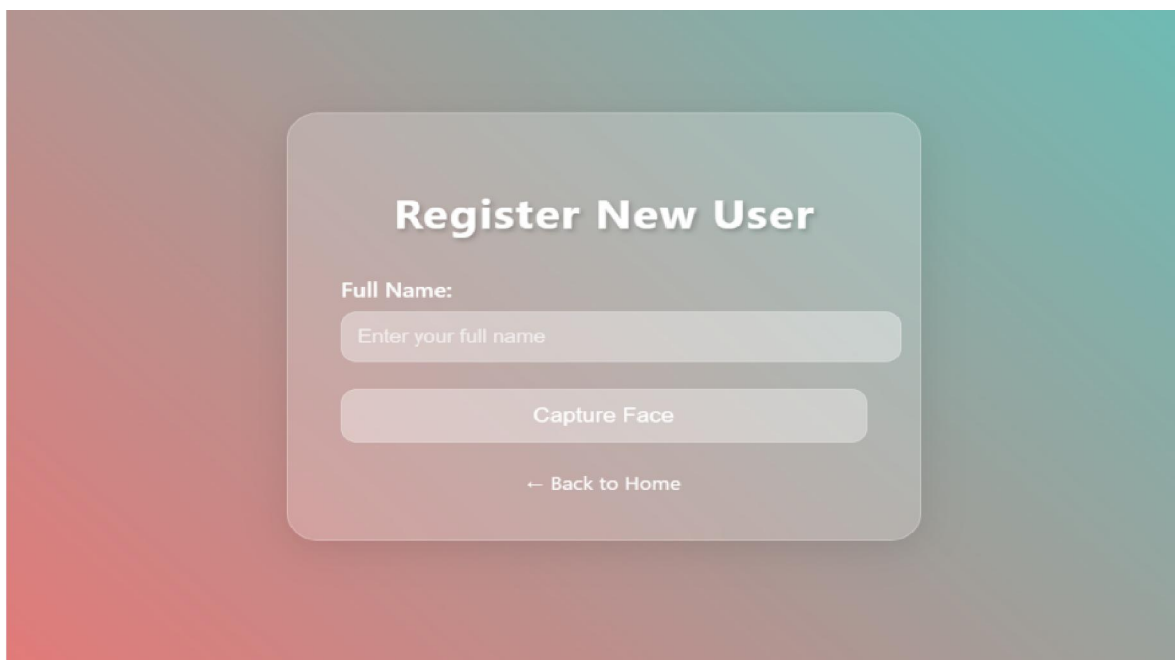


figure 3: User registration interface.



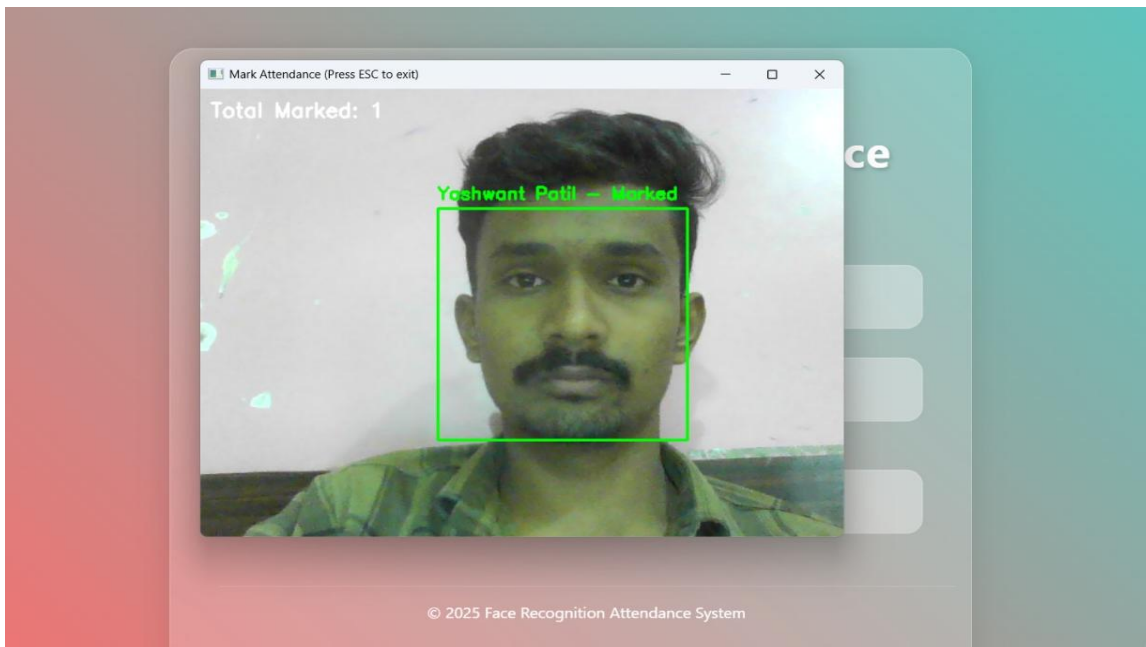


figure 4: Attendance marked for one user.

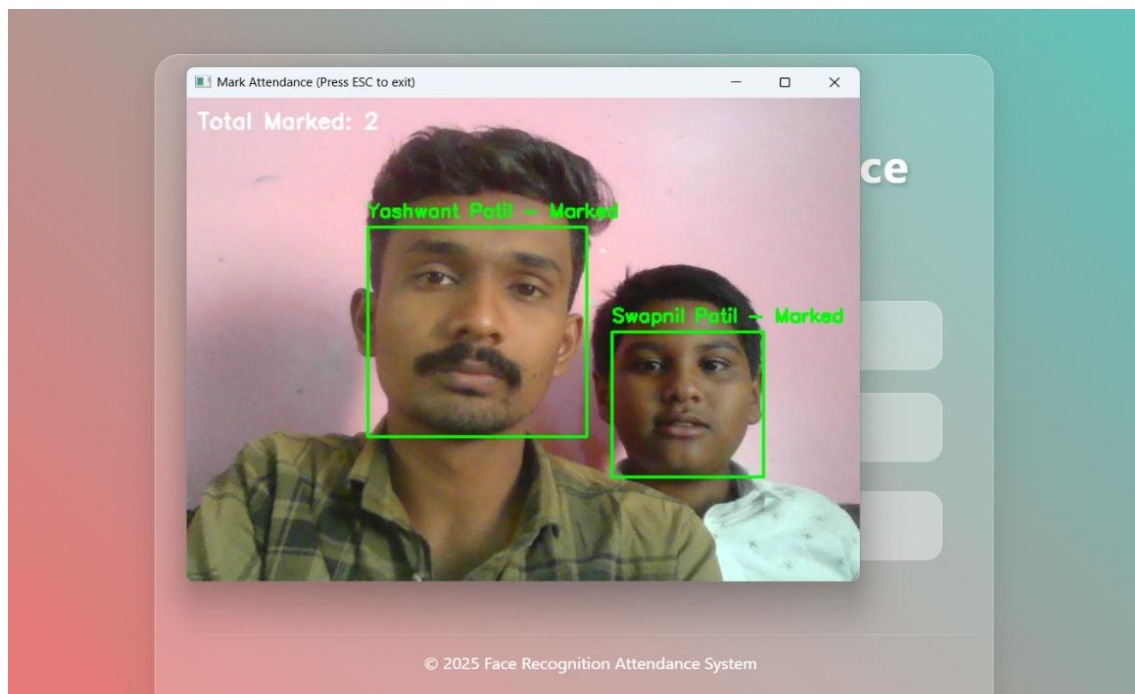


figure 5: Attendance marked for two users.



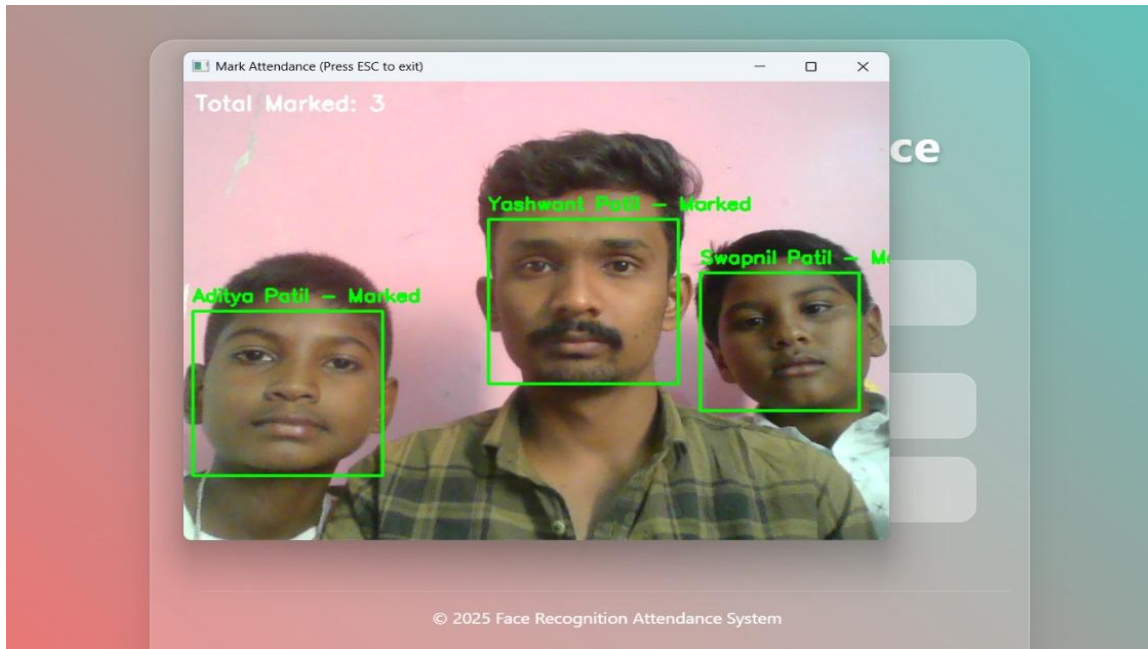


figure 6: Attendance marked for three users.

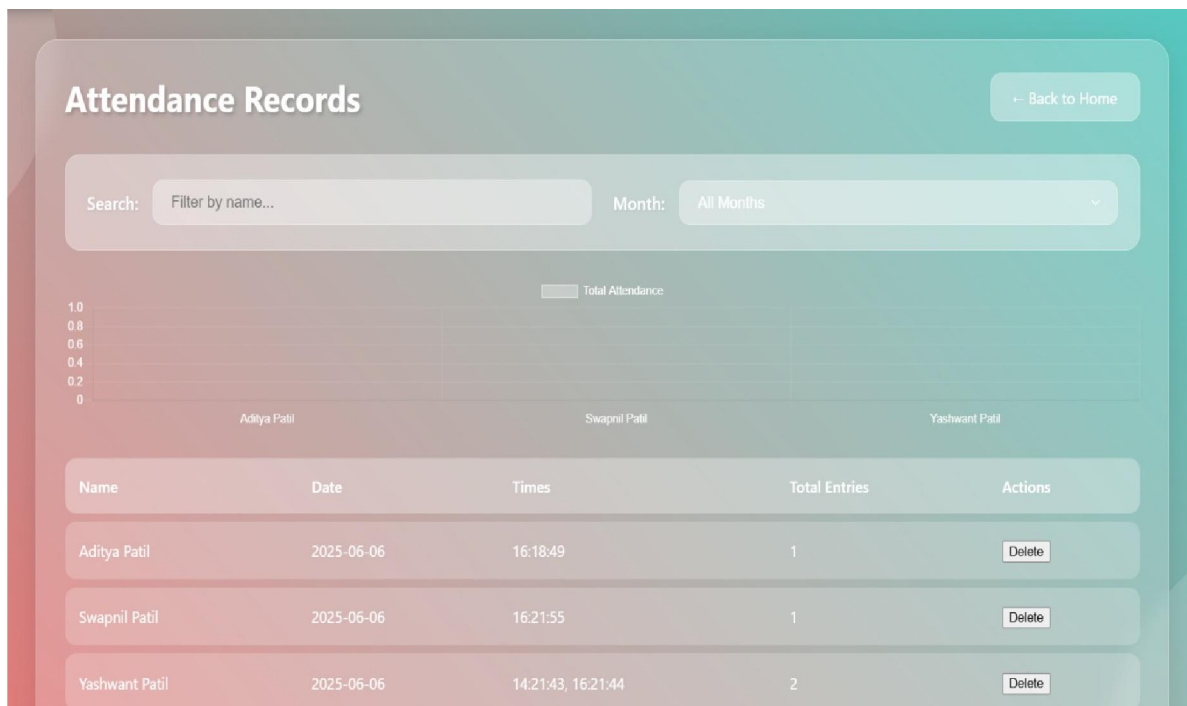


figure 7: Attendance records view.



IV. CONCLUSION

This research illustrates a robust, adaptable, and real-time face recognition attendance system competent of recognizing numerous individuals inside a single outline utilizing low-cost equipment and open-source innovations. By combining a YOLO-inspired face detection strategy with dlib's facial embedding framework and coordination it into a Flask-based web interface, we accomplished high execution without the require for GPU increasing speed or cloud framework. Such a framework holds solid potential for sending in educational teach, little enterprises, and government workplaces where fetched, exactness, and effectiveness are vital. Future improvements might incorporate portable integration, cloud reinforcement, anti-spoofing, and support for video-based attendance logging

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