



# MULTI-FACE RECOGNITION ATTENDANCE SYSTEM USING MACHINE LEARNING

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**Abstract:** In the quest for efficient and precise attendance management, traditional methods often fall short due to their manual nature and susceptibility to errors. To address these challenges, we present a Multi-Face Recognition Attendance System that harnesses the power of Machine Learning to automate and enhance the attendance tracking process. Central to our system are two advanced techniques: the Haar Cascade classifier for face detection and the Local Binary Patterns Histograms (LBPH) classifier for face recognition. The Haar Cascade classifier is adept at identifying and locating faces in images or live video feeds, even under varying conditions and angles. Once faces are detected, the LBPH classifier steps in to match these faces against a pre-existing database, ensuring accurate identification and verification of individuals. This approach is particularly effective in environments where multiple individuals need to be recognized simultaneously, such as classrooms or corporate settings. The system is implemented as a web application using Streamlit, known for its simplicity and efficiency in building interactive applications. Streamlit provides a user-friendly interface where users can upload images or connect live cameras to capture real-time data for attendance tracking. The application not only simplifies the process of recording attendance but also offers easy management and review of attendance records. Our system is designed to be scalable, accommodating a growing number of users and faces, and can be integrated with existing data management infrastructure. Future improvements will aim at refining the accuracy of face detection and recognition, incorporating real-time monitoring features, and expanding compatibility with other biometric systems. By leveraging these state-of-the-art techniques within a practical and accessible platform, our Multi-Face Recognition Attendance System offers a modern solution to the challenges of attendance management, promising both increased efficiency and reliability.

**Index Terms - Multi-Face Recognition, Machine Learning, Haar Cascade, LBPH Classifier, Streamlit Web App, Attendance Management.**

## I. INTRODUCTION

In today's fast-paced world, organizations and institutions face increasing pressure to manage attendance efficiently and accurately. Traditional methods, such as manual attendance taking and paper records, have proven to be time-consuming, cumbersome, and often unreliable. These systems are prone to human error and fraud, leading to significant challenges when dealing with large groups or dynamic environments. As a result, the gap between the need for precise attendance tracking and the limitations of conventional systems has become more pronounced.

To address this gap, technology-driven solutions are emerging as vital tools for improving attendance management. The integration of advanced systems offers the promise of reducing the inefficiencies associated with manual processes while enhancing the overall accuracy of attendance records. This is particularly important in environments where real-time monitoring and authentication are crucial. Organizations are increasingly looking to leverage automation and machine learning technologies to streamline these tasks, ensuring more reliable outcomes. One such innovative solution is the Multi-Face Recognition Attendance System, which harnesses the power of machine learning to transform the way attendance is tracked. This system utilizes the Haar Cascade classifier for efficient face detection and the Local Binary Patterns Histograms (LBPH) classifier for precise face recognition. These advanced techniques enable the system to accurately identify and authenticate individuals in real-time, offering a highly reliable solution for attendance management.

The Multi-Face Recognition Attendance System is implemented as a web-based application using Streamlit, ensuring a seamless and user-friendly interface for both administrators and users. By automating attendance tracking, this solution reduces the burden on human resources, minimizes errors, and provides a scalable approach that can be adapted to various organizational needs. It not only meets the demands of today's environments but also lays the foundation for future advancements in attendance management technology. This system represents a significant leap forward from outdated manual methods. By combining machine learning with modern web-based interfaces, it addresses the inefficiencies of traditional systems while offering a more robust, scalable, and accurate solution.

## II. LITERATURE SURVEY

In this paper [1], the authors address the critical challenges that face recognition systems encountered during the COVID-19 pandemic, particularly due to the widespread adoption of face masks in public spaces such as airports, railway stations, and payment platforms. Face recognition systems have become integral to modern authentication and security infrastructure because they offer a contactless and hygienic alternative to traditional touch-based biometrics. However, the mandatory use of masks disrupted these systems by occluding key facial features, making it difficult for algorithms to accurately identify individuals. This issue highlighted the limitations of conventional face recognition techniques, which rely heavily on complete or unobstructed facial data, and emphasized the need for new solutions capable of handling such occlusions. Existing literature has explored various methods to address the problem of partial occlusion, focusing on advanced techniques such as feature extraction from visible facial regions, multi-view face recognition, and the use of deep learning models. Multi-task Cascaded Convolutional Networks (MTCNN) have been widely recognized for their capability in robust face detection, even under challenging conditions. Similarly, VGGFace, a model built on deep convolutional neural networks, has proven effective in extracting facial features essential for recognition tasks. Researchers have also investigated alternative approaches, such as blending multiple biometric modalities (e.g., voice and iris recognition) to enhance authentication accuracy when faces are partially covered. However, these methods often involve higher implementation costs and increased system complexity. Considering these challenges, the authors of this paper propose a new algorithm that leverages both MTCNN and the VGGFace model to improve the accuracy of facial recognition under mask-wearing conditions. Their approach focuses on extracting and analyzing features from the upper half of the face, such as the eyes and forehead, which remain visible even when masks are worn. This targeted feature extraction method significantly enhances recognition performance, as the top half of the face carries unique patterns and structures that contribute to individual identification. By optimizing the integration of MTCNN for precise face detection and VGGFace for effective feature recognition, the proposed algorithm achieves a commendable 90% accuracy with partially masked faces, demonstrating its potential for real-world deployment. This study makes an important contribution to the field of face recognition by offering a practical solution that maintains security and authentication standards despite the occlusion challenges posed by masks. It also underscores the need for future research to develop more resilient biometric systems capable of adapting to unpredictable disruptions such as pandemics. The proposed algorithm not only improves the usability of existing face recognition systems but also reduces the need for redesigning entire authentication frameworks. As mask-wearing may continue to be prevalent in public health scenarios, this work highlights the importance of developing adaptive algorithms and hybrid models that can ensure the reliability of facial recognition technology in dynamic environments.

In this work [2], the authors address the growing need for automated understanding and evaluation of vast image and video datasets, driven by the rapid expansion of such data in recent years. Manual analysis of these datasets is impractical, highlighting the importance of smart frameworks capable of performing tasks such as face identification, emotion recognition, and head position evaluation. Unlike humans, who have limited capacity for accurately distinguishing subtle expressions and appearances, automated systems can process this data more efficiently. Facial recognition software, a key biometric tool, has emerged as a solution to these challenges by transforming the mathematical features of a person's face into a "face print," which is stored in a database for future identification and verification. This technology enables a deep learning system to compare a real-time digital image with previously stored data to confirm an individual's identity. The study emphasizes the importance of facial recognition not only in individual identification but also in various aspects of human interaction. Facial features play a critical role in interpersonal communication, and the ability to detect and recognize faces within digital images has numerous applications. Facial recognition software identifies the size, position, and orientation of a person's face, and its adoption spans both consumer markets and the security and surveillance industries. The COVID-19 pandemic has further accelerated the relevance of contactless biometric technologies like facial recognition, as people seek safer and more hygienic alternatives to touch-based systems. This shift has also brought increased attention to the importance of accuracy and reliability in facial recognition for various real-world applications, including online exam proctoring, where it ensures the integrity of remote assessments, and in the airline industry, where it allows seamless identification and verification at multiple touchpoints during a passenger's journey. Despite its advantages, one of the primary limitations of existing face recognition algorithms is their sensitivity to image quality, which can negatively impact recognition performance. The authors address this issue by focusing on improving image quality using advanced tools and algorithms. In this study, they implemented facial recognition methods using Python libraries such as OpenCV and the Face Recognition library, which provide robust frameworks for detecting and identifying faces in images. The research explores not only the technical implementation of these algorithms but also their practical applications in different industries.

In this paper [3], the authors address the challenges and opportunities associated with the rapid expansion of image and video datasets, which demand automated systems for efficient data processing and evaluation. Manual analysis is impractical due to the scale of data, highlighting the importance of smart frameworks for tasks such as face detection, appearance recognition, head-pose estimation, and human-computer interaction. Facial recognition software, regarded as a biometric technology, plays a crucial role in these applications by converting a person's facial features into a mathematical representation, or "face print," which is stored in a database to authenticate identities. A deep learning system compares real-time or captured images to stored data, enabling quick and reliable identity verification. Facial recognition technology is essential in both personal and professional interactions, given the face's importance in communication and identification. It determines the size, orientation, and location of facial features within a digital image, making it useful across multiple sectors. The software has seen widespread adoption in consumer markets, such as mobile devices, and in security and surveillance systems, including public spaces and restricted areas. The COVID-19 pandemic has significantly accelerated the adoption of contactless biometric technologies, placing a greater focus on facial recognition as people seek safer alternatives to touch-based systems. This technology is now used in online exam proctoring to maintain integrity during remote assessments, ensuring that only authorized individuals participate. Similarly, the airline industry relies on facial recognition to streamline the passenger experience by providing rapid identification and verification at multiple checkpoints. Despite its growing use, the authors note that existing facial recognition algorithms face limitations, particularly with image quality, which directly affects recognition accuracy. Poor lighting, camera resolution, and environmental factors can degrade performance,

making it critical to develop more resilient solutions. This study addresses these challenges by leveraging advanced tools and algorithms, including OpenCV and the Face Recognition library in Python, to enhance the quality and accuracy of facial recognition systems. The proposed approach improves detection and identification by refining image processing techniques, ensuring better performance under varied conditions. This research not only discusses the technical aspects of facial recognition algorithms but also emphasizes their practical applications across industries. As contactless authentication becomes increasingly important, especially in safety-conscious environments, improving facial recognition systems is essential. The study underscores the need for continuous advancements in image handling and algorithm design to overcome existing challenges. By focusing on improving image quality, the research offers a robust framework for deploying facial recognition systems that meet the evolving demands of both public and private sectors. This work reinforces the importance of adaptive biometric technologies in facilitating seamless, secure, and hygienic interactions in a rapidly changing world.

### III. METHODOLOGY

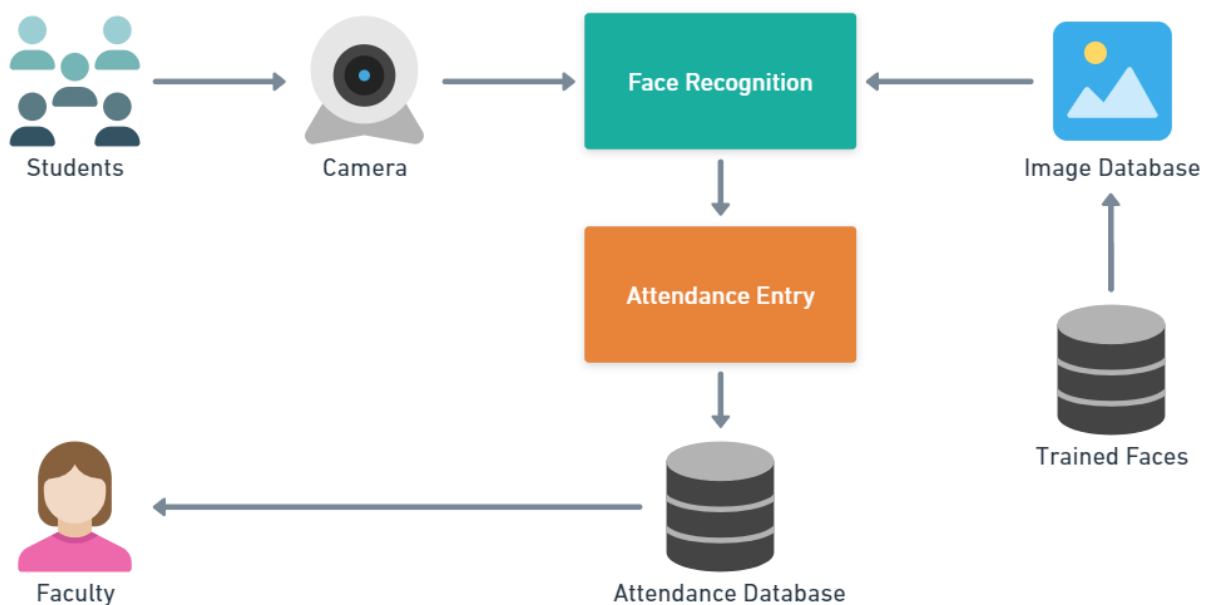


Figure 1 System Architecture

The proposed system involves a multi-step approach that combines advanced machine learning techniques with practical application development. Initially, the system employs the Haar Cascade classifier for face detection, which is trained to identify facial features and locate faces within images or video streams. This classifier uses a series of pre-trained models to detect faces with high accuracy, even under varying lighting conditions and angles. Once faces are detected, the images are processed for recognition using the Local Binary Patterns Histograms (LBPH) classifier. This classifier extracts distinctive facial features and compares them against a pre-existing database to accurately identify and authenticate individuals. The combination of these techniques ensures reliable face detection and recognition, crucial for effective attendance management.

The web application is developed using Streamlit, chosen for its ease of use and rapid development capabilities. Streamlit provides an intuitive interface that allows users to upload images or connect live camera feeds for real-time attendance tracking. The application integrates face detection and recognition algorithms, processing the data to record attendance automatically. Additionally, the system is designed to be scalable, accommodating a growing number of users and faces while maintaining performance and accuracy. Future enhancements will focus on optimizing algorithm efficiency, incorporating additional features such as real-time alerts, and ensuring seamless integration with existing data management systems. This comprehensive methodology ensures that the system is not only effective in addressing current needs but also adaptable to future advancements and requirements.

### VI. PROJECT MODULES

The five modules involved in Multi-Face Recognition Attendance System using Machine Learning:

- Face Detection using HAAR Cascade Module** - This module is responsible for identifying and locating faces within an image or video frame using the HAAR Cascade classifier, a machine learning-based approach. HAAR Cascade works by scanning the input image with multiple filters to detect specific patterns, such as the edges and contours of a face. It processes frames in real-time, making it well-suited for dynamic environments like classrooms or workplaces. This module ensures that only regions with detected faces proceed to the next steps for further processing.
- Face Preprocessing Module** - In this module, the detected face images undergo several preprocessing steps to enhance the accuracy of recognition. Preprocessing includes tasks like grayscale conversion, noise reduction, resizing, and alignment of the face to ensure uniformity across all images. Standardizing the input through such steps eliminates



variations caused by lighting, background, or face angles, creating consistent input for the feature extraction stage. This module plays a crucial role in minimizing errors by providing clean and normalized data for further analysis.

- c. **Feature Extraction and Recognition Module** - This module extracts key facial features and identifies individuals using machine learning algorithms, such as Local Binary Patterns Histograms (LBPH) or other deep learning models. Extracted features, such as the shape of eyes, nose, and lips, are compared against a pre-existing database to recognize the individual. This module is the core of the system, where recognition takes place based on the unique patterns identified in the face. It ensures quick and accurate identification, even with multiple faces detected simultaneously.
- d. **Attendance Recording Module** - Once individuals are recognized, this module automates the process of marking attendance. It tries to identify the recognized faces and records their presence, associating it with the relevant date and time. The system can also handle cases of multiple entries by ensuring that attendance is only marked once per session. This module eliminates the need for manual attendance tracking, reducing errors and ensuring efficiency in real-time environments like classrooms or workplaces.
- e. **Database Management Module** - The database management module is responsible for storing and organizing the data collected by the system, including registered users' facial data, attendance records, and timestamps. It ensures smooth data retrieval and management through structured queries, making it easy to generate reports or track attendance history. This module also handles updates, such as adding new users, modifying existing records, and managing backups to maintain data integrity and security.

## V. CONCLUSION

The Multi-Face Recognition Attendance System represents a significant advancement in automating and improving attendance management across diverse sectors. By leveraging the Haar Cascade and LBPH classifiers for accurate face detection and recognition and integrating these capabilities into a user-friendly Streamlit web application, the system offers a reliable, efficient, and scalable solution to traditional attendance challenges. Its ability to handle multiple faces in real-time and its adaptability to various environments make it an invaluable tool for educational institutions, corporate offices, events, healthcare facilities, and beyond. This innovative approach not only enhances accuracy and reduces manual effort but also sets a new standard for modern attendance management systems.

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