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Face Recognition Attendance System Using K-Nearest Neighbors Algorithm

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Abstract: This research paper presents the design and implementation of a real-time, contactless attendance monitoring system based on face recognition using the K-Nearest Neighbor (KNN) algorithm. With the growing need for secure, automated, and hygienic attendance systems—especially in educational institutions and workplaces—biometric solutions have become highly relevant. The proposed system utilizes a webcam to capture facial images of individuals, which are then pre-processed by converting them into grayscale and resizing them to a standard format. These images are used to train a KNN classifier that learns to distinguish between different individuals based on facial features.

The system employs Haar cascade classifiers for face detection, and during execution, it continuously scans and recognizes faces in real-time, logging attendance data with timestamps. All recognized faces are stored systematically in a text-based log file organized by individual and date. The solution was tested under various environmental conditions including well-lit indoor, low-light, and outdoor scenarios. The experimental results indicate a high level of accuracy, particularly in controlled lighting environments, with performance gracefully degrading in low-light scenarios.

The simplicity and efficiency of the KNN algorithm make it well-suited for small- to medium-scale applications, offering a cost-effective and scalable alternative to more complex deep learning models. This system serves as a practical demonstration of how traditional machine learning methods can still offer viable solutions for real-world biometric identification tasks.

Keywords: Face Recognition, Attendance System, KNN, OpenCV, Machine Learning, Pattern Recognition

I. INTRODUCTION

Attendance management is an essential administrative task in educational institutions, workplaces, and other formal environments. Traditional methods of attendance recording, such as paper-based registers or RFID cards, are not only time-consuming but also susceptible to human error and manipulation, including proxy attendance. These limitations highlight the need for a more robust, efficient, and secure solution. Biometric-based attendance systems, particularly those utilizing facial recognition technology, have emerged as a promising alternative. Unlike fingerprint or iris recognition, face recognition is non-intrusive and does not require physical contact, making it a safer and more hygienic solution—especially important in the post-pandemic world. It also offers the convenience of automatic identification without requiring users to carry cards or remember login credentials.

In this project, a real-time face recognition-based attendance system is proposed, using the K-Nearest Neighbors (KNN) algorithm. KNN is a simple yet effective machine learning algorithm known for its accuracy in classification tasks. The system captures face images using a webcam, processes them using OpenCV, and classifies them with a trained KNN model. Recognized individuals are logged into attendance records along with timestamps.

The system aims to provide a cost-effective, scalable, and user-friendly alternative to traditional methods. It is particularly suitable for educational institutions and small organizations seeking to adopt biometric solutions without investing in complex or expensive infrastructure. Through this project, we demonstrate the practical application of classic machine learning techniques in solving real-world problems related to security and identity verification.

II. LITERATURE SURVEY

A literature survey provides insight into existing face recognition systems and classification algorithms, helping to identify their strengths, limitations, and areas for improvement. The field has evolved from basic image processing methods to complex machine learning and deep learning models. This section highlights significant contributions in face recognition and justifies the use of the K-Nearest Neighbours (KNN) algorithm in the proposed system.

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VI. IMPLEMENTATION

A. Data Collection

Using OpenCV, faces are detected via the webcam and saved after converting to grayscale and resizing to 100x100 pixels:

```
cv2.CascadeClassifier(cv2.data.harcascades + "haarcascade_frontalface_default.xml")
cv2.VideoCapture(0)
cv2.imwrite("dataset/person_name/image.jpg", resized_face)
```

B. Feature Extraction & Training

All images are flattened and labeled. A KNN model with `k=3` is trained using scikit-learn:

```
knn = KNeighborsClassifier(n_neighbors=3)
knn.fit(X, y)
```

C. Recognition & Attendance Logging

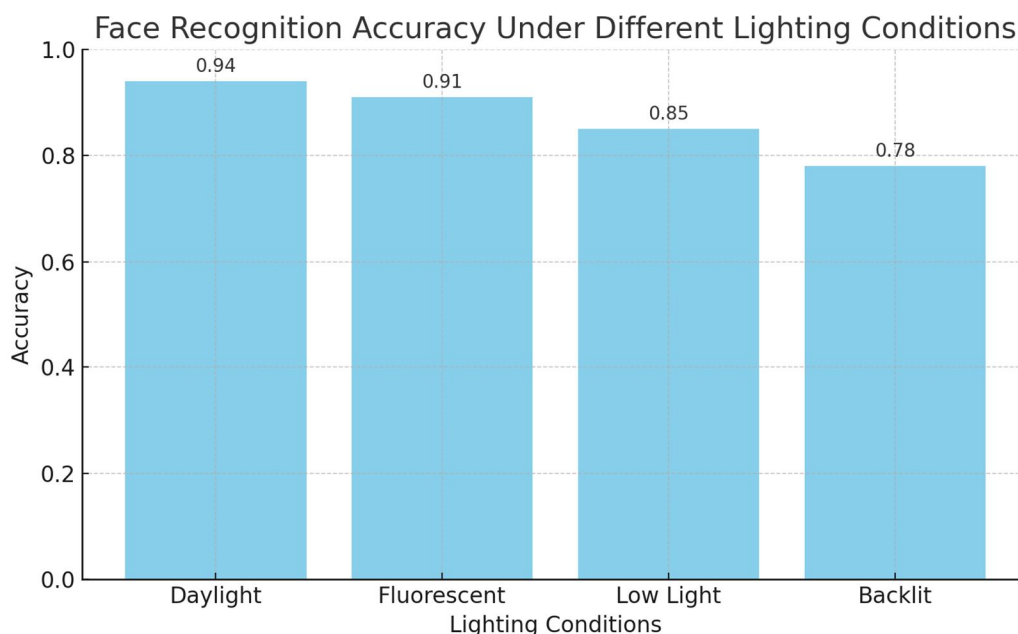
The trained model predicts face identities in real time and logs them into timestamped text files:

```
pred = knn.predict(face_flat)
with open("attendance/name/date.txt", "a") as f:
    f.write(f'Recognized: {name} at {timestamp}\n')
```

VII. RESULTS

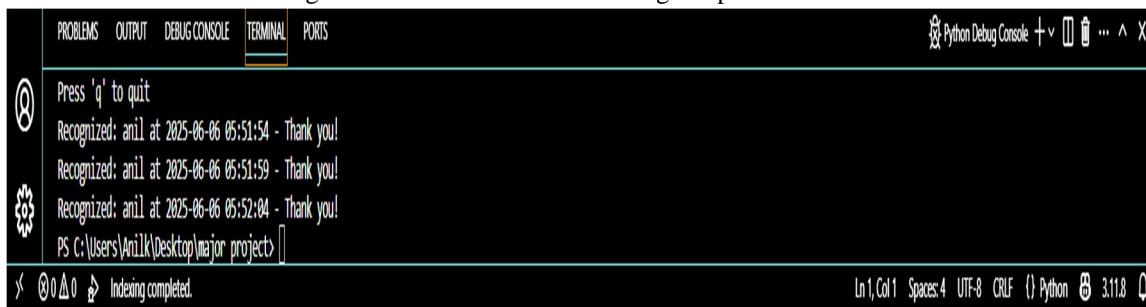
The system was tested using a dataset created by capturing face images of multiple users under varied lighting conditions. Recognition accuracy remained high under standard indoor environments, with some degradation observed under poor lighting.

Below is a graph illustrating the face recognition accuracy under different lighting conditions:



This demonstrates that while the model performs reliably in most scenarios, enhancements such as lighting normalization or deep learning techniques can further improve robustness.

Figure 2: Real-time Attendance Log Output in Terminal



This demonstrates that while the model performs reliably in most scenarios, enhancements such as lighting normalization or deep learning techniques can further improve robustness.

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