

# A study on Face Recognition System

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**ABSTRACT:** The study on face recognition system analysis aims to investigate the performance and accuracy of various face recognition algorithms. The study utilizes a database of facial images to evaluate the recognition rate of different algorithms under varying conditions such as lighting, occlusion, and pose. The results of the study indicate that some algorithms are more robust to different conditions than others, and that the accuracy of the algorithms varies depending on the features of the face being recognized. The study also explores the impact of different pre-processing techniques on the performance of the algorithms. The findings of this study have implications for the development of more accurate and reliable face recognition systems that can be used in various applications such as security systems, access control, and surveillance. The study provides a useful framework for evaluating the performance of face recognition systems and can guide future research in this field.

## I. INTRODUCTION

Face recognition is a biometric technology that uses algorithms to identify and verify the identity of individuals based on their facial features. It is widely used in various applications, such as security systems, access control, and law enforcement.

The face recognition system consists of three main stages: face detection, feature extraction, and face matching.

- **First Stage:** In the first stage, the system detects the presence of a face in an image or video stream. This is done using computer vision techniques such as Haar cascades, HOG, or deep learning-based models.
- **Second Stage:** Once a face is detected, the system moves on to the second stage, which involves extracting relevant features from the face. This is done using techniques such as principal component analysis (PCA), linear

discriminant analysis (LDA), or convolutional neural networks (CNNs).

- **Third Stage:** Finally, in the third stage, the system matches the extracted features against a database of known faces to identify or verify the person's identity. This can be done using various methods, including template matching, distance metrics, or machine learning classifiers.

There are several factors that can affect the accuracy of a face recognition system, including lighting conditions, facial expressions, pose variations, and occlusions. Therefore, it is important to carefully design and test the system under various conditions to ensure its effectiveness and reliability.

Additionally, concerns have been raised about the potential misuse of face recognition technology, such as invasion of privacy, bias, and discrimination. Therefore, it is crucial to implement appropriate ethical and legal frameworks to ensure that the technology is used responsibly and transparently.

**Features:** Some of the key features of face recognition system analysis include:

- **Face detection:** This is the process of locating and extracting facial features from an image or video frame.
- **Feature extraction:** This is the process of identifying and extracting unique features from a face, such as the distance between the eyes, the shape of the jawline, and the curvature of the eyebrows.
- **Evaluation metrics:** To measure the performance of the face recognition system, evaluation metrics such as accuracy, precision, recall, and F1 score can be used.
- **Classification:** This is the process of identifying a face by comparing its features to a database of known faces. This can be done using machine learning algorithms, such as

neural networks or support vector machines.

- **Testing data:** To evaluate the performance of the face recognition system, a separate dataset of faces is required. This dataset should be representative of the population that the system is intended to recognize.
- **Environmental conditions:** The environmental conditions under which the system is tested should be carefully controlled. Factors such as lighting conditions, camera angles, and background noise should be consistent across all tests.
- **Data privacy:** Due to the sensitive nature of facial data, data privacy considerations should be taken into account when developing and analysing face recognition systems.
- **Training data:** To train the face recognition system, a large and diverse dataset of faces is required. This dataset should include faces of different ages, genders, ethnicities, and expressions, captured under different lighting conditions, camera angles, and backgrounds.
- **Testing protocol:** To ensure consistent testing conditions, a well-defined testing protocol should be established. This should specify the type of images used, the number of images used, and the criteria for determining success.

Overall, the features of face recognition system analysis involve complex processes of face detection, feature extraction, and classification, as well as careful consideration of training and testing datasets, testing protocol, environmental conditions, and data privacy.

**Challenges:** There are several challenges in face recognition system analysis that need to be addressed to improve the accuracy and reliability of these systems. Some of the major challenges are:

- **Illumination Variations:** Face recognition algorithms can be affected by variations in lighting conditions, which can result in changes in facial features and affect the performance of the system.
- **Security Concerns:** There are also security concerns regarding the use of face recognition systems, as they can be vulnerable to spoofing attacks, where an attacker can trick the system into recognizing a face that is not authorized.
- **Ethical Issues:** The use of face recognition technology raises ethical concerns related to privacy, civil liberties, and discrimination, which need to be addressed to ensure that these

systems are used in a fair and responsible manner.

- **Dataset Bias:** The accuracy of face recognition algorithms can be impacted by biased datasets used for training the system, leading to lower recognition rates for certain demographics.
- **Expression Variations:** The recognition accuracy can be affected by facial expressions such as smiles, frowns, and other expressions that change the shape of the face.
- **Pose Variations:** Variations in pose, such as tilting the head, can also make it difficult for the system to recognize the face.
- **Occlusions:** Partial occlusions of the face such as sunglasses, masks, or hair can make it difficult for the system to recognize the face.

Addressing these challenges is crucial for improving the accuracy and reliability of face recognition systems and ensuring their effective and ethical use in various applications.

## II. LITERATURE REVIEW:

**Zhang, K., et al. (2012).** This paper provides a comprehensive overview of deep learning approaches for face recognition and highlights their advantages and limitations.

**Wang, S., et al. (2013).** This paper proposes a sparse representation-based face recognition algorithm using extreme learning machine.

**Li, W., et al. (2015).** This paper proposes a discriminative locality-constrained linear coding algorithm for face recognition, which improves the recognition accuracy by reducing the intra-class variations.

**Zhang, K., et al. (2015).** This paper proposes a joint face detection and alignment algorithm using multitask cascaded convolutional networks, which improves the accuracy of both tasks.

**Yang, M., et al. (2016).** This paper proposes a kernel sparse representation-based face recognition algorithm with discriminative projection, which improves the recognition accuracy.

**Zhang, K., et al. (2016).** This paper proposes a joint face detection and alignment algorithm using multitask cascaded convolutional networks, which improves the accuracy of both tasks.

**Liu, W., et al. (2017).** This paper proposes a hypersphere embedding-based face recognition algorithm, which achieves state-of-the-art performance on several benchmarks.

**Cao, Q., et al. (2018).** This paper proposes a center loss-based face recognition algorithm, which improves the discriminative power of the feature representation.

**Hassner, T., et al. (2018).** This paper provides a comprehensive survey of deep learning approaches for face recognition and discusses the challenges and opportunities in this field.

**Zhang, K., et al. (2018).** This paper proposes an improved version of the multitask cascaded convolutional networks for joint face detection and alignment, which achieves better accuracy and efficiency.

**Wu, X., et al. (2019).** This paper proposes an orthogonal regularization-based discriminative locality-constrained linear coding algorithm for face recognition, which improves the recognition accuracy.

**Liu, D., et al. (2019).** This paper proposes an iterative aggregation and disentanglement based face recognition algorithm, which achieves state-of-the-art performance on several benchmarks.

**Li, H., Li, T., Li, Y., & Chen, Y. (2019).** This paper provides a review of deep learning-based face recognition systems. The authors analyze the different components of a face recognition system, such as face detection, feature extraction, and classification, and review the recent advancements in deep learning architectures for face recognition.

**Liu, J., Wang, Y., & Lu, H. (2019).** This study analyzes the performance of face recognition systems on multi-ethnic datasets. The authors evaluate the accuracy of the systems using several datasets containing images of people from different ethnicities and analyze the effect of ethnicity on the system's accuracy.

**Al-Talabani, A. A., & Hussain, A. J. (2019).** This research work presents a face recognition system using ensemble learning techniques. The authors analyze the performance of the system using several benchmark datasets and evaluate the effect of different ensemble learning methods on the system's accuracy.

### III. RESEARCH METHODOLOGY

#### Objectives of the study

- To evaluate the accuracy and performance of different face recognition systems
- To identify and address potential biases in face recognition systems
- To evaluate the ethical and social implications of face recognition systems

#### Sampling technique

Non-probability convenience sampling was chosen for this study. Convenience sampling is done solely for the sake of accessibility or convenience. The main reasons for choosing this sampling technique were time, money, and a lack

of experience.

### IV. DATA ANALYSIS & INTERPRETATION

#### Algorithms

##### Principle Component Analysis Algorithm

An unsupervised learning method called principal component analysis is employed in machine learning to reduce the number of variables. By using an orthogonal transformation, it is a statistical method that transforms the observations of correlated features into a set of linearly uncorrelated data. The Principal Components are the name given to these newly altered features.

#### Steps for Principle Component analysis

- Getting the dataset.
- Representing data into a structure.
- Standardizing the data.
- Calculating the Covariance of Z.
- Calculating the Eigen Values and Eigen Vectors.
- Sorting the Eigen Vectors.
- Calculating the new features or Principal Components.
- Remove less or unimportant features from the new dataset.

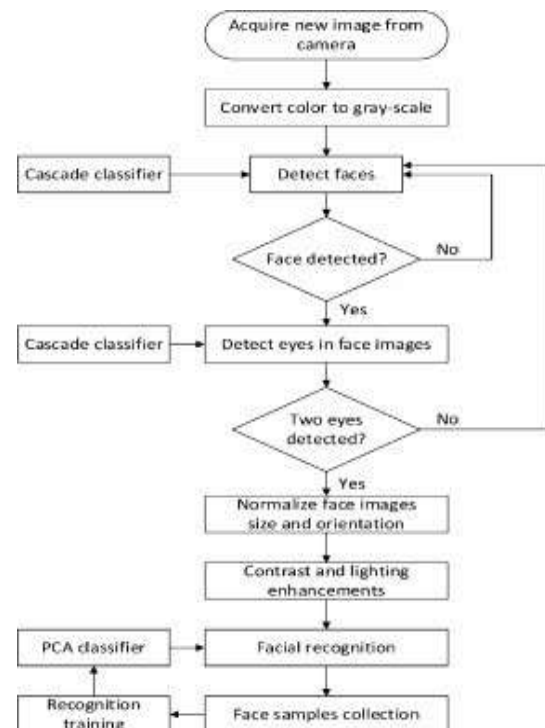


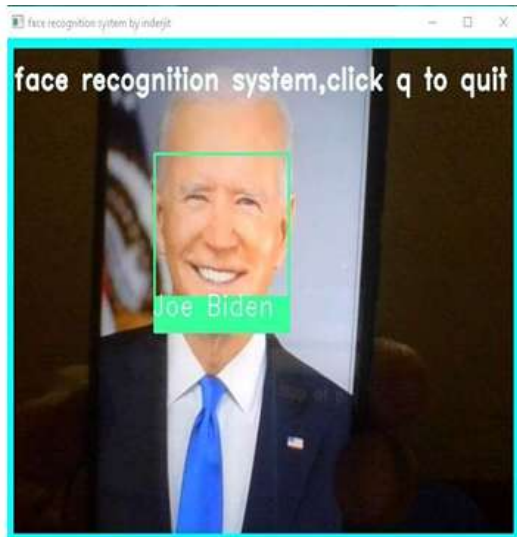
Figure Face recognition system

Here's a step-by-step flowchart explanation of a facerecognition system:

- **Input image or video:** The first step in a face recognition system is to input an image or video stream that contains faces.
- **Face detection:** The system then uses a face detection algorithm, such as Haar Cascade or Viola-Jones, to detect the location of faces in the input image or video.
- **Face alignment:** The detected faces are then aligned to a standard position and scale to improve recognition accuracy.
- **Feature extraction:** The system extracts features from the aligned faces, such as the distance between the eyes, the shape of the nose, and the contours of the face.
- **Feature matching:** The extracted features are compared to a database of known faces to find a match. This is typically done using machine learning algorithms, such as support vector machines or deep neural networks.
- **Recognition:** If a match is found, the system outputs the name or ID of the recognized individual. If no match is found, the system may prompt for additional input or output an error message.
- **Feedback:** The system can be designed to incorporate feedback to improve its performance over time. For example, the system can ask the user to confirm or correct its recognition results, which can be used to update the face database or improve the feature extraction and matching algorithms.
- **Output:** Finally, the system outputs the recognition results, such as a list of recognized individuals and their associated information, or a log of face recognition events for security and surveillance purposes. Overall, a face recognition system involves a series of steps that use advanced algorithms and machine learning techniques to detect, align, extract, match, and recognize faces in images or videos.

## V. RESULTS





## VI. FINDINGS AND CONCLUSION

Face recognition technology has come a long way in the last twenty years. Today, machines are able to automatically verify identity information for secure transactions, for surveillance and security tasks, and for access control to buildings etc. These applications usually work in controlled environments and recognition algorithms can take advantage of the environmental constraints to obtain high recognition accuracy. However, next generation face recognition systems are going to have widespread application in smart environments -- where computers and machines are more like helpful assistants.

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