

**DR. VISHWANATH KARAD MIT WORLD PEACE
UNIVERSITY, PUNE**

Department of Computer Engineering & Technology
School of Computer Science & Engineering
Seminar
Year B. Tech, Semester 5

**COMPARISON BETWEEN FACE RECOGNITION
ALGORITHMS AND TECHNIQUES**

SEMINAR REPORT

**Under the Guidance of
Dr. Sharmishta Desai**

Prepared By

Krishnaraj Thadesar, PA10, 1032210888

May 5, 2025

List of Figures

3.1	Cropped Faces using OpenCV Haar Cascades	10
3.2	Cropped Faces using OpenCV Haar Cascades	10
3.3	Cropped Faces using OpenCV Haar Cascades	11
3.4	Krishnaraj's Segregated Images	11
3.5	Saubhagya's Segregated Images	12
3.6	Parth's Segregated Images	12
3.7	Sourab's Segregated Images	13
3.8	Abhijeet's Segregated Images	13
3.9	Results identifying 3 of the 4 faces. Empirical results show that the model is working, with accuracy of around 75 %	14
3.10	Accuracy per Person (Class) - LBPH Face Recognizer	20

List of Tables

3.1 Training Data Images	8
------------------------------------	---

Abbreviations

1. **PCA** - Principal Component Analysis
2. **LDA** - Linear Discriminant Analysis
3. **SVM** - Support Vector Machine
4. **KNN** - K-Nearest Neighbors
5. **CNN** - Convolutional Neural Network
6. **DNN** - Deep Neural Network
7. **ANN** - Artificial Neural Network
8. **ML** - Machine Learning
9. **DL** - Deep Learning
10. **AI** - Artificial Intelligence

Acknowledgment

I would like to express my deepest appreciation to all those who provided me the possibility to complete this report. A special gratitude I give to our mentor, Dr. Sharmishta Desai, whose contribution in stimulating suggestions and encouragement, helped me to coordinate my project especially in writing this report.

Furthermore, I would also like to acknowledge with much appreciation the crucial role of the staff of MIT WPU, who gave the permission to use all required equipment and the necessary materials to complete the task. A special thanks goes to my team mates, who helped me enormously to assemble the parts and gave suggestion about the task of using the techniques of measurements.

I have to appreciate the guidance given by other supervisor as well as the panels especially in our project presentation that has improved our presentation skills thanks to their comment and advices.

I would also like to thank my parents for their wise counsel and sympathetic ear. You are always there for me. Finally, I wish to thank my friends for their support and encouragement throughout my study.

Name of Student

1. Parth Zarekar, 1032210846

Contents

0.1 Abstract	7
0.2 Keywords	7
1 Introduction	1
1.0.1 Problem Statement	1
1.0.2 Need of the Project	1
2 Literature Survey	2
2.1 Paper 1	2
2.1.1 Positives and Learnings from this Paper	2
2.1.2 Identified Research Gaps	2
2.2 Paper 2	2
2.2.1 Positives and Learnings from this Paper	2
2.2.2 Identified Research Gaps	3
2.3 Paper 3	3
2.3.1 Positives and Learnings from this Paper	3
2.3.2 Identified Research Gaps	3
2.4 Paper 4	3
2.4.1 Positives and Learnings from this Paper	3
2.4.2 Identified Research Gaps	4
2.5 Paper 5	4
2.5.1 Positives and Learnings from this Paper	4
2.5.2 Identified Research Gaps	4
3 Methodology and Implementations	5
3.1 Methodology	5
3.2 Advantages	5
3.3 Disadvantages	6
3.4 Evaluation Metrics	6
3.5 Applications	6
3.6 Implementations	7
3.7 Platform	7
3.7.1 Training Data	7
3.7.2 Creation of Dataset	9
3.7.3 Preliminary Results	14
3.7.4 Accuracy per Person (Class) - LBPH Face Recognizer	14
4 Future Prospects for the Attendance Assistant	21
4.1 Integration of advanced Deep-Learning Models	21
4.2 Dataset Expansion and Synthetic Augmentation	21
4.3 Edge Deployment and Resource Optimization	21
4.4 Privacy, Security, and Ethical Considerations	22
4.5 Multi-Modal and Context-Aware Fusion:	22
4.6 Broader Applications and Commercialization	22

5 Conclusion **23**

Bibliography **24**

Abstract and Keywords

0.1 Abstract

Face Recognition is a biometric method of identifying an individual by comparing live capture or digital image data with the stored record for that person. It is a widely used technology in security systems and can be compared to other biometrics such as fingerprint or eye iris recognition systems. This report will discuss the various algorithms and techniques used in face recognition and compare them based on their performance and accuracy. The report will also discuss the implementation of these algorithms and techniques in real-world applications.

Methods used in Face Recognition by most commonly used python libraries like OpenCV, Dlib, etc. will be discussed in this report. The report will also discuss the various challenges faced in face recognition and how these challenges can be overcome. The report will also discuss the future of face recognition technology and how it can be used in various applications.

0.2 Keywords

Face Recognition, Biometric, Algorithms, Techniques, OpenCV, Dlib, Python, Machine Learning, Deep Learning, Artificial Intelligence, Security Systems, Applications, Attendance System, Surveillance System.

Chapter 1

Introduction

Face recognition is a biometric technology that utilizes distinctive features of the face to identify individuals. Widely employed in security systems, it serves various applications such as access control, attendance tracking, and surveillance. While face recognition has existed for decades, recent advancements in machine learning and computer vision have significantly enhanced its accuracy and reliability. Numerous algorithms and techniques are available, each with unique strengths and weaknesses. In this seminar, we aim to compare popular face recognition algorithms and assess their performance using a standardized dataset.

1.0.1 Problem Statement

We need to compare the various face recognition algorithms and techniques to determine which one is the most accurate and efficient. We also need to discuss the implementation of these algorithms in real-world applications.

1.0.2 Need of the Project

- The motivation for this topic came from impending research for a Project titled "Machine Learning Powered Automated Facial Attendance Tracking System".
- The project aims to develop a system that can automatically track attendance using facial recognition technology.
- To achieve this goal, it is essential to understand the different face recognition algorithms and techniques available and evaluate their performance to identify the most suitable approach for the project.
- By comparing the performance of different face recognition algorithms and techniques, we can gain insights into their strengths and weaknesses and make informed decisions about which approach to use for the project.
- This seminar will provide a comprehensive overview of the most popular face recognition algorithms and techniques and evaluate their performance on a common dataset to help guide the development of the attendance tracking system.

This project will help in understanding the various face recognition algorithms and techniques and how they can be implemented in real-world applications. It will also help in understanding the challenges faced in face recognition and how these challenges can be overcome.

To use the correct method and library in finding attendance, so as to reduce time and cost, while also maintaining high levels of accuracy, it was necessary to compare the various face recognition algorithms and techniques.

Chapter 2

Literature Survey

2.1 Paper 1

Title: "A Comparative Study of Facial Recognition Techniques: With focus on low computational power." Author: Schenkel, T., Ringhage, O. and Branding, N. [7]

2.1.1 Positives and Learnings from this Paper

1. The publication compares five performance metrics, including recall and F-score, providing a comprehensive evaluation of facial recognition techniques.
2. It addresses the importance of balancing low computational time and prediction ability for security systems, offering practical guidelines for implementation.
3. The research questions are clearly defined, focusing on significant differences in performance, training time, and prediction time among different facial recognition techniques and classifiers.

2.1.2 Identified Research Gaps

1. The document lacks detailed information on the specific facial recognition techniques and classifiers used in the experiments.
2. It does not provide a detailed breakdown of the dataset used for training and testing the facial recognition models.
3. While the document mentions the comparison of results, it does not delve into the specific findings or implications of these comparisons.

2.2 Paper 2

Title: "A Comparative Study on Facial Recognition Algorithms" Author: Sanmoy Paul and Sameer Acharya [8]

2.2.1 Positives and Learnings from this Paper

1. Comparative Analysis: The study provides a comparative analysis of different facial recognition algorithms, allowing developers to make informed choices based on recognition accuracies.
2. Algorithm Selection: By studying the advantages and disadvantages of various algorithms, developers can select the best facial recognition algorithm for their specific implementation needs.

3. Future Improvements: The research suggests future efforts to test on a larger set of images to enhance the accuracy of CNN and explore combining multiple machine learning classification algorithms for increased recognition accuracy and handling large datasets.

2.2.2 Identified Research Gaps

1. The document lacks detailed discussion on the specific methodologies used for training and testing the algorithms, which could provide more clarity on the experimental setup.
2. There is no mention of the computational resources or hardware specifications used for running the experiments, which could impact the reproducibility and scalability of the results.
3. The publication does not delve into the potential biases or limitations in the dataset used for training and testing the facial recognition models, which could affect the generalizability of the findings.

2.3 Paper 3

Title: "*A comparison of facial recognition algorithms.*" Author: *Delbiaggio, Nicolas.* [9]

2.3.1 Positives and Learnings from this Paper

1. Thesis covers a comprehensive comparison of facial recognition algorithms like Eigenfaces, Fisherfaces, LBPH, and OpenFace.
2. The study includes a detailed explanation of each algorithm, their strengths, weaknesses, and performance in a test case scenario.
3. The findings highlight OpenFace as the most accurate algorithm for facial recognition, providing valuable insights for further research in the field.

2.3.2 Identified Research Gaps

1. Lack of Exploration of Real-World Applications: The paper focuses on comparing facial recognition algorithms in a controlled setting. However, it does not delve into the practical applications of these algorithms in real-world scenarios.
2. Limited Discussion on Algorithm Limitations: While the strengths of the algorithms are discussed, there is a lack of emphasis on the limitations of each algorithm.
3. Absence of Future Research Directions: The paper concludes with the identification of the most accurate algorithm but fails to suggest potential future research directions in the field of facial recognition.

2.4 Paper 4

Title: "*Evaluating impact of race in facial recognition across machine learning and deep learning algorithms.*" Author: *Coe, James, and Mustafa Atay.* [10]

2.4.1 Positives and Learnings from this Paper

1. The paper provides a detailed comparison of various facial recognition algorithms, including Eigenfaces, Fisherfaces, Local Binary Pattern Histogram, deep convolutional neural network algorithm, and OpenFace.
2. It highlights the efficiency and accuracy of these algorithms in real-life settings, with OpenFace being identified as the algorithm with the highest accuracy in identifying faces.

3. The study's findings offer valuable insights for practitioners in selecting the most suitable algorithm for facial recognition applications and suggest ways for academicians to enhance the current algorithms' accuracy further.

2.4.2 Identified Research Gaps

1. The paper focuses on a few specific facial recognition algorithms like Eigenfaces, Fisherfaces, and Local Binary Pattern Histograms. It lacks exploration of a wider range of algorithms available in the field, potentially missing out on newer, more accurate models.
2. While the study evaluates the algorithms' accuracy, it does not delve into their performance in real-life settings or practical applications. This gap could impact the algorithms' effectiveness when deployed in scenarios beyond controlled test environments.
3. The paper mentions the use of a custom dataset for testing the algorithms but does not elaborate on the dataset's diversity or size.

2.5 Paper 5

Title: "A Comparative Study of Facial Recognition Techniques: With focus on low computational power." Author: Schenkel, T., Ringhage, O. and Branding, N. [11]

2.5.1 Positives and Learnings from this Paper

1. Efficiency Evaluation: The paper provides a detailed comparison of popular open source facial recognition algorithms, highlighting the efficiency and accuracy of each in real-life settings.
2. Practical Implications: The findings of the study offer valuable insights for practitioners in selecting the most suitable algorithm for facial recognition applications, enhancing decision-making processes.
3. Academic Contribution: The research contributes to the academic field by emphasizing the importance of improving the accuracy of existing algorithms, paving the way for further advancements in facial recognition technology.

2.5.2 Identified Research Gaps

1. The paper focuses on comparing a few facial recognition algorithms like Eigenfaces, Fisherfaces, and Local Binary Pattern Histogram. However, it lacks a comparison with a wider range of algorithms to provide a more comprehensive analysis.
2. While the paper evaluates the algorithms' performance in a controlled environment using test datasets, it doesn't discuss the practical implementation challenges or results in real-life scenarios, which could be a crucial research gap.
3. The paper does not delve into the scalability and efficiency aspects of the facial recognition algorithms studied. Understanding how these algorithms perform with larger datasets or in real-time applications could be a significant research gap to address.

Chapter 3

Methodology and Implementations

3.1 Methodology

Libraries Tested

These are the libraries that were used to train and test a model.

1. OpenCV
2. face_recognition
 - face_recognition is a Python library that provides a simple interface for face recognition tasks.
 - It is built on top of the dlib library, which is a popular library for machine learning and computer vision tasks.
 - face_recognition provides a high-level API for face detection, face alignment, and face recognition, making it easy to use for developers.
 - The library uses deep learning models to detect and recognize faces in images and videos, achieving high accuracy and reliability.
 - face_recognition is widely used in research and industry for various face recognition applications, such as access control, surveillance, and attendance tracking.

3.2 Advantages

1. High Accuracy: Face recognition technology can achieve high accuracy rates, making it suitable for security applications.
2. Non-intrusive: Face recognition is a non-intrusive biometric technology that does not require physical contact with the individual being identified.
3. Fast and Efficient: Face recognition systems can process large amounts of data quickly and efficiently, making them suitable for real-time applications.
4. Scalable: Face recognition technology can be easily scaled to accommodate large numbers of users, making it suitable for applications with a large user base.
5. Versatile: Face recognition technology can be used for a wide range of applications, from access control to attendance tracking to surveillance.

3.3 Disadvantages

1. Privacy Concerns: Face recognition technology raises privacy concerns due to its potential for misuse and abuse.
2. Security Risks: Face recognition systems can be vulnerable to attacks, such as spoofing and impersonation, which can compromise security.
3. Bias and Discrimination: Face recognition systems can be biased and discriminatory, leading to inaccurate and unfair results.
4. Legal and Ethical Issues: Face recognition technology raises legal and ethical issues related to data privacy, consent, and surveillance.
5. Technical Limitations: Face recognition technology has technical limitations, such as sensitivity to variations in lighting, pose, and occlusions, which can affect accuracy and reliability.
- 6.

3.4 Evaluation Metrics

1. Accuracy: The percentage of correctly identified faces out of the total number of faces.
2. Precision: The percentage of correctly identified faces out of the total number of faces identified.
3. Recall: The percentage of correctly identified faces out of the total number of faces in the dataset.
4. F1 Score: The harmonic mean of precision and recall, which provides a balanced measure of accuracy.
5. Time taken: The time taken to process the dataset and identify the faces, which measures the efficiency of the algorithm.
6. False Positive Rate: The percentage of incorrectly identified faces out of the total number of faces identified.
7. False Negative Rate: The percentage of correctly identified faces out of the total number of faces not identified.

3.5 Applications

1. Access Control: Face recognition technology can be used for access control in buildings, vehicles, and devices.
2. Attendance Tracking: Face recognition technology can be used to track attendance in schools, colleges, and workplaces.
3. Surveillance: Face recognition technology can be used for surveillance in public spaces, airports, and other high-security areas.
4. Personalization: Face recognition technology can be used for personalization in devices, such as smartphones and smart home devices.
5. Healthcare: Face recognition technology can be used in healthcare for patient identification and monitoring.

3.6 Implementations

3.7 Platform

Operating System: Windows 11 Pro

IDEs or Text Editors Used: Visual Studio Code

Compilers or Interpreters: Python 3.10.1

3.7.1 Training Data

- The dataset used for training and testing the face recognition algorithms is a collection of images of individuals, each labeled with their name.
- The dataset contains images of different individuals taken under various lighting conditions, angles, and expressions to ensure robustness in face recognition.
- The dataset is divided into two parts: a training set used to train the face recognition model and a testing set used to evaluate the model's performance.

Serial Number	Name	Image		
1	Saubhagya			
2	Avishkar			
3	Karad			
4	Krish			
5	Parth			

Table 3.1: Training Data Images

3.7.2 Creation of Dataset

Cropped Faces using openCV-haar-cascades

The following code generated 18,832 possible faces (245x245px) each. The code uses the OpenCV library to detect faces in images and crop them. The cropped faces are then saved in a specified output folder. The code uses the Haar Cascade classifier for face detection, which is a popular method for real-time face detection.

```

1 import cv2
2 import os
3 import numpy as np
4
5 def detect_and_crop_faces(input_folder, output_folder, padding=10):
6     if not os.path.exists(output_folder):
7         os.makedirs(output_folder)
8
9     face_cascade = cv2.CascadeClassifier(cv2.data.haarcascades + \
10                                         'haarcascade_frontalface_default.xml')
11
12    for filename in os.listdir(input_folder):
13        if filename.lower().endswith(('png', 'jpg', 'jpeg', 'webp')):
14            image_path = os.path.join(input_folder, filename)
15            image = cv2.imread(image_path)
16
17            if image is None:
18                continue
19
20            gray = cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)
21            faces = face_cascade.detectMultiScale(gray, scaleFactor=1.1, minNeighbors=5,
22                                               minSize=(30, 30))
23
24            for i, (x, y, w, h) in enumerate(faces):
25                x1 = max(x - padding, 0)
26                y1 = max(y - padding, 0)
27                x2 = min(x + w + padding, image.shape[1])
28                y2 = min(y + h + padding, image.shape[0])
29
30                face_crop = image[y1:y2, x1:x2]
31                output_path = os.path.join(output_folder, f"{os.path.splitext(filename)[0]}_face_{i}.jpg")
32                cv2.imwrite(output_path, face_crop)
33                print(f"Saved cropped face: {output_path}")
34
35 input_folder = os.path.join(os.getcwd(), "input_images")
36 output_folder = os.path.join(os.getcwd(), "output_images")
37
38 detect_and_crop_faces(input_folder, output_folder)

```

Results of the above code are shown in the figure below. The images are cropped and saved in the output folder.

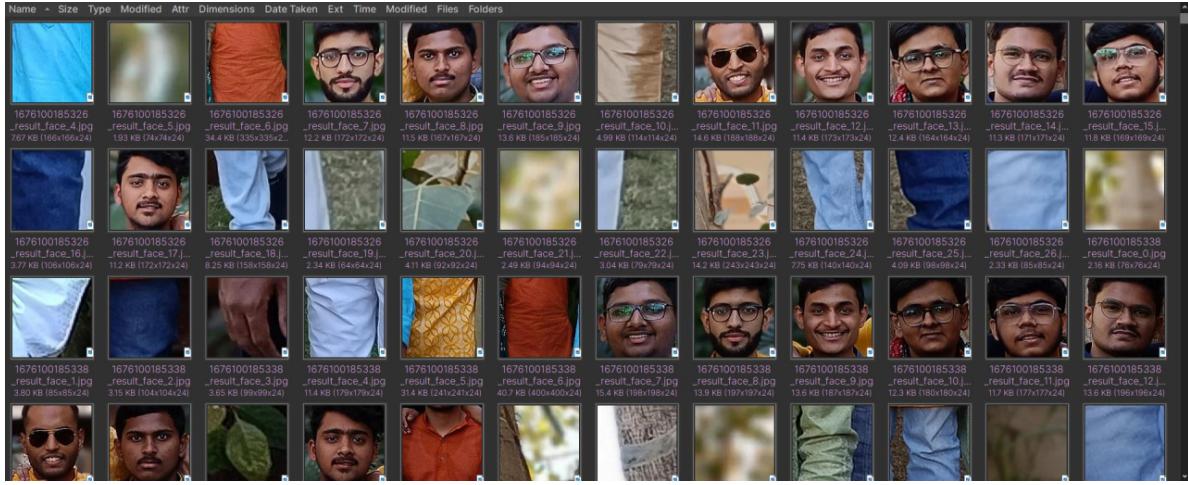


Figure 3.1: Cropped Faces using OpenCV Haar Cascades

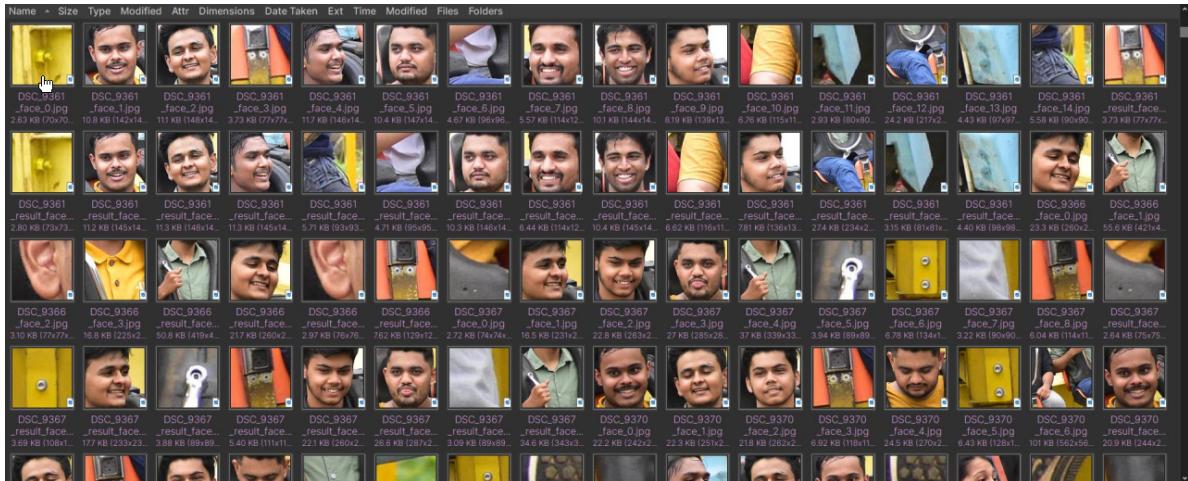


Figure 3.2: Cropped Faces using OpenCV Haar Cascades



Figure 3.3: Cropped Faces using OpenCV Haar Cascades

Manually Segregated photos for 5 people and labelled them



Figure 3.4: Krishnaraj's Segregated Images

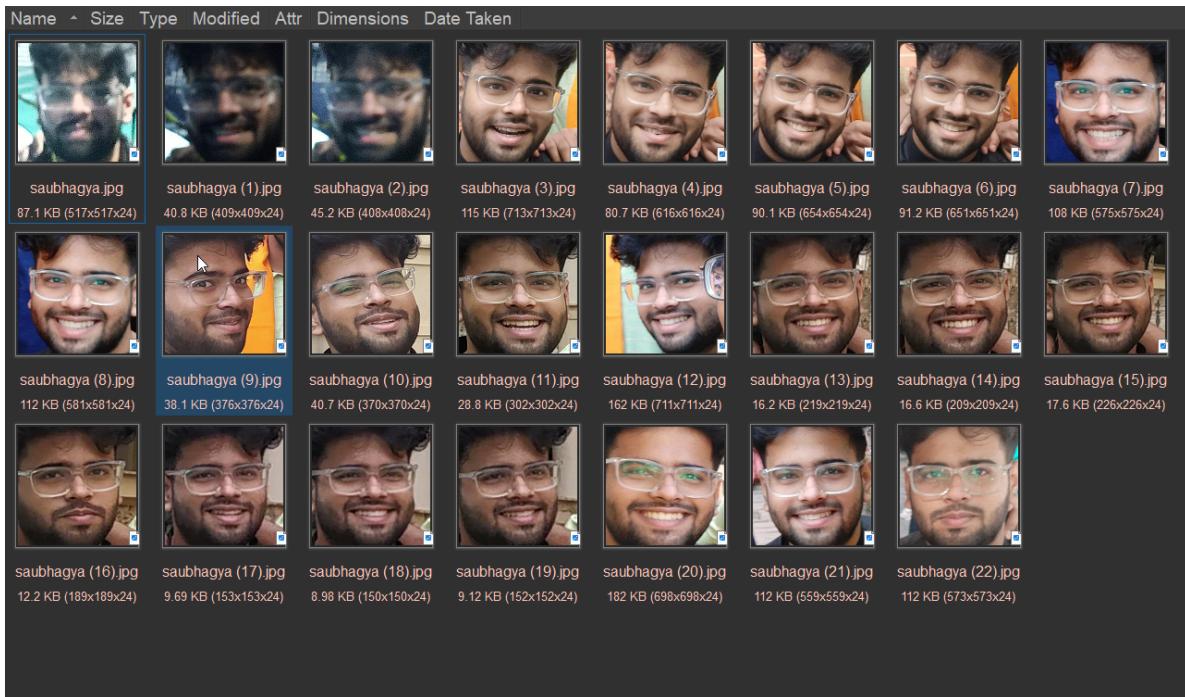


Figure 3.5: Saubhagya's Segregated Images

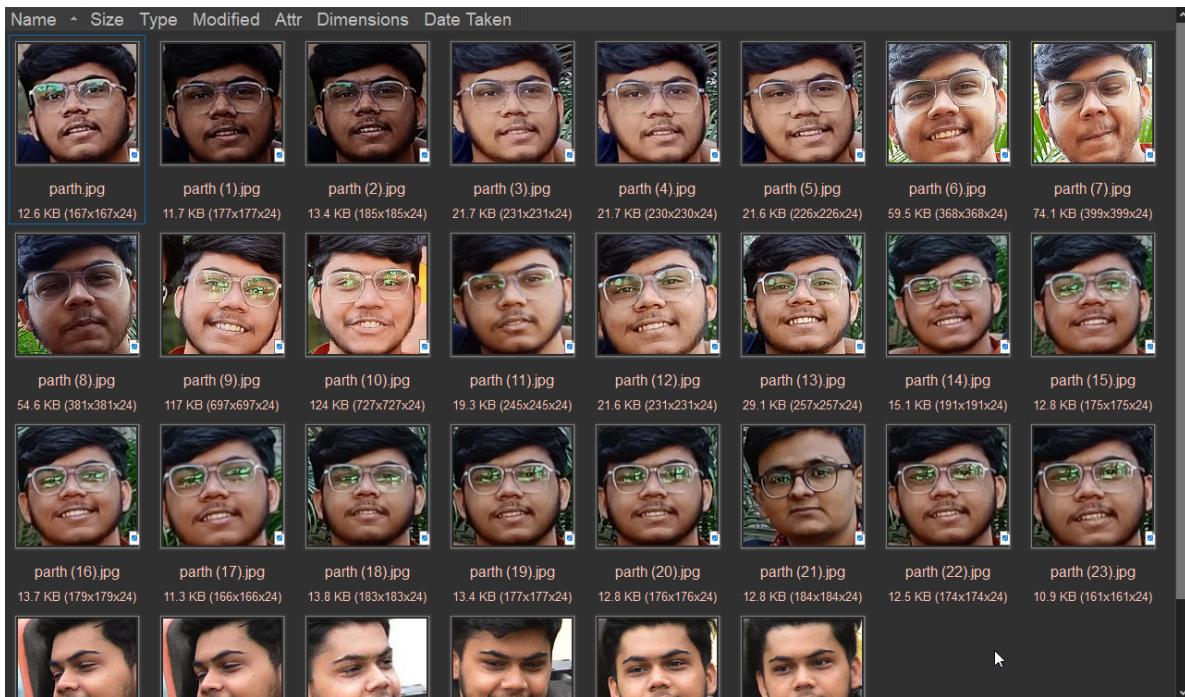


Figure 3.6: Parth's Segregated Images



Figure 3.7: Sourab's Segregated Images

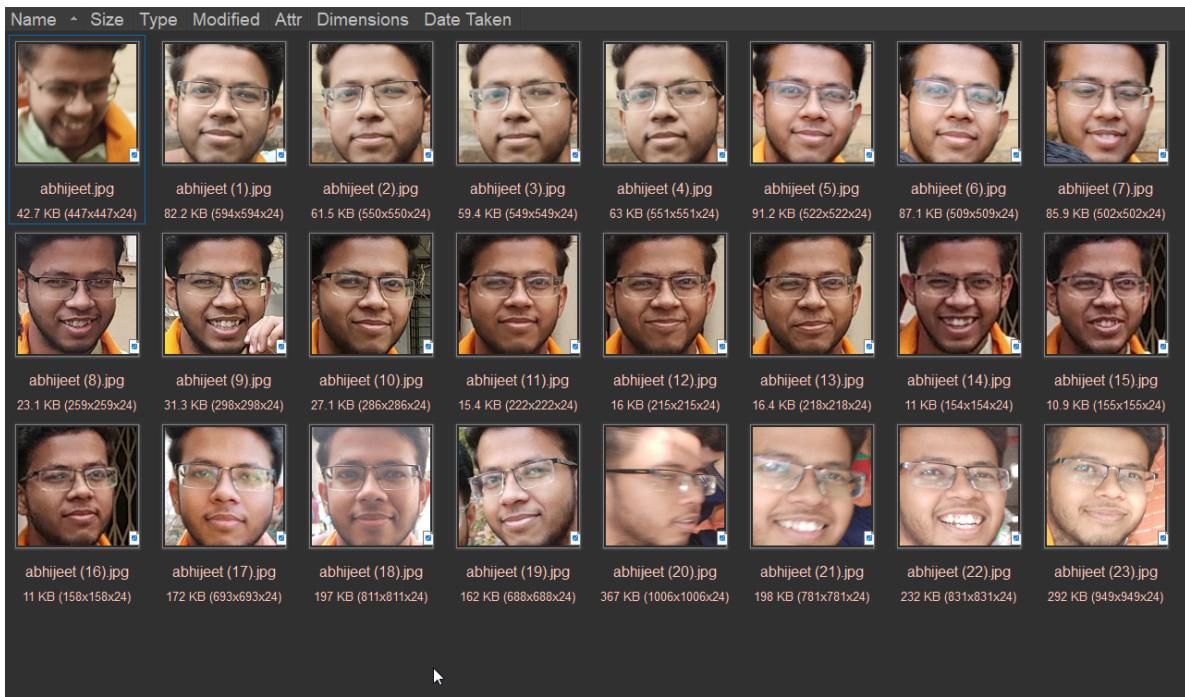


Figure 3.8: Abhijeet's Segregated Images

3.7.3 Preliminary Results



Figure 3.9: Results identifying 3 of the 4 faces. Empirical results show that the model is working, with accuracy of around 75 %

3.7.4 Accuracy per Person (Class) - LBPH Face Recognizer

LBPH Face Recognizer code

```
1 #!/usr/bin/env python
2 # coding: utf-8
3
4 # we are testing 3 different algorithms
5
6 # In[1]:
7
8
9 pip install opencv-python numpy dlib face_recognition deepface facenet-pytorch
  scikit-learn mtcnn
10
11
12 # In[2]:
13
14
15 pip install opencv-python
16
17 # In[3]:
18
19
20 pip install dlib
21
22 # In[4]:
23
24
25 pip install cmake
26
27 # In[5]:
```

```
29 ! pip install dlib --no-cache-dir
30
31 # In[6]:
32
33
34 pip install --upgrade pip setuptools wheel
35
36
37 # In[7]:
38
39
40 pip install face_recognition
41
42 # In[8]:
43
44
45 ! pip install deepface
46
47 # In[9]:
48
49
50 %pip install opencv-python scikit-learn matplotlib seaborn tensorflow
51
52 # In[10]:
53
54
55 import cv2
56 print(cv2.__version__)
57
58 # In[18]:
59
60
61 pip uninstall opencv-python --yes
62
63 # In[1]:
64
65
66 pip install opencv-contrib-python jupyter matplotlib scikit-learn numpy
67
68 #
69
70 # In[8]:
71
72
73 import os
74 import cv2
75 import numpy as np
76 from sklearn.metrics import classification_report
77 from sklearn.preprocessing import LabelEncoder
78 base_path = "comprehensive_db\\comprehensive_db"
79
80 # === 1. Load Data ===
81 def load_faces_from_folder(base_path, split='train', image_size=(100, 100)):
82     faces = []
83     labels = []
84     person_names = []
85     for person_name in os.listdir(base_path):
86         person_folder = os.path.join(base_path, person_name, split)
87         if not os.path.isdir(person_folder):
```

```

88     continue
89
90     for filename in os.listdir(person_folder):
91         img_path = os.path.join(person_folder, filename)
92         img = cv2.imread(img_path, cv2.IMREAD_GRAYSCALE)
93         if img is None:
94             continue
95         img_resized = cv2.resize(img, image_size)
96         faces.append(img_resized)
97         labels.append(person_name) # use actual name for now
98         person_names.append(person_name)
99
100    return np.array(faces), np.array(labels)
101
102 # === 2. Encode Labels ===
103 def encode_labels(labels):
104     le = LabelEncoder()
105     numeric_labels = le.fit_transform(labels)
106     return numeric_labels, le
107
108
109 # In[14]:
110
111
112 X_train, y_train_names = load_faces_from_folder(base_path, 'train')
113 X_test, y_test_names = load_faces_from_folder(base_path, 'test')
114
115 # In[15]:
116
117
118 def train_and_evaluate_model(model, model_name, X_train, y_train, X_test, y_test,
119                             label_encoder):
120     model.train(X_train, y_train)
121     predictions = []
122
123     for face in X_test:
124         label_pred, _ = model.predict(face)
125         predictions.append(label_pred)
126
127     # Decode numeric labels back to names
128     y_test_names = label_encoder.inverse_transform(y_test)
129     y_pred_names = label_encoder.inverse_transform(predictions)
130
131     print(f"--- {model_name} ---")
132     print(classification_report(y_test_names, y_pred_names))
133     return classification_report(y_test_names, y_pred_names, output_dict=True)
134
135
136
137
138 y_train, label_encoder = encode_labels(y_train_names)
139 y_test = label_encoder.transform(y_test_names)
140
141 # Convert to correct format for OpenCV
142 X_train = [img for img in X_train]
143 X_test = [img for img in X_test]
144
145 # Create models

```

```

146 lbph_model = cv2.face.LBPHFaceRecognizer_create()
147 eigen_model = cv2.face.EigenFaceRecognizer_create()
148 fisher_model = cv2.face.FisherFaceRecognizer_create()
149
150 # Train and evaluate
151 lbph_results = train_and_evaluate_model(lbph_model, "LBPH Face Recognizer",
152                                         X_train, y_train, X_test, y_test, label_encoder)
152 eigen_results = train_and_evaluate_model(eigen_model, "EigenFace Recognizer",
153                                         X_train, y_train, X_test, y_test, label_encoder)
153 fisher_results = train_and_evaluate_model(fisher_model, "FisherFace Recognizer",
154                                         X_train, y_train, X_test, y_test, label_encoder)
155
155 # In[17]:
156
157
158 print(f"Unique classes in y_train: {np.unique(y_train)}")
159 print(f"Number of classes: {len(np.unique(y_train))}")
160
161 # In[21]:
162
163
164 import os
165 import cv2
166 import numpy as np
167 from sklearn.metrics import classification_report
168
169 base_path = "comprehensive_db\\comprehensive_db"
170 image_size = (100, 100)
171
172 def load_images(folder, label, max_images=None):
173     images = []
174     labels = []
175     for i, filename in enumerate(os.listdir(folder)):
176         if max_images and i >= max_images:
177             break
178         img_path = os.path.join(folder, filename)
179         img = cv2.imread(img_path, cv2.IMREAD_GRAYSCALE)
180         if img is not None:
181             img = cv2.resize(img, image_size)
182             images.append(img)
183             labels.append(label)
184     return images, labels
185
186 def train_individual_models(base_path):
187     person_dirs = [d for d in os.listdir(base_path) if os.path.isdir(os.path.join(base_path, d))]
188     for person in person_dirs:
189         print(f"\nTraining model for: {person}")
190
191         # Positive samples (label = 1)
192         pos_train_dir = os.path.join(base_path, person, 'train')
193         pos_test_dir = os.path.join(base_path, person, 'test')
194         pos_train_imgs, pos_train_labels = load_images(pos_train_dir, label=1)
195         pos_test_imgs, pos_test_labels = load_images(pos_test_dir, label=1)
196
197         # Negative samples (label = 0) from other people's train dirs
198         neg_train_imgs = []
199         neg_train_labels = []
200         neg_test_imgs = []

```

```

201     neg_test_labels = []
202     for other_person in person_dirs:
203         if other_person == person:
204             continue
205         other_train_dir = os.path.join(base_path, other_person, 'train')
206         other_test_dir = os.path.join(base_path, other_person, 'test')
207         imgs, labels = load_images(other_train_dir, label=0, max_images=len(
208             pos_train_imgs)//(len(person_dirs)-1))
209         neg_train_imgs += imgs
210         neg_train_labels += labels
211         imgs, labels = load_images(other_test_dir, label=0, max_images=len(
212             pos_test_imgs)//(len(person_dirs)-1))
213         neg_test_imgs += imgs
214         neg_test_labels += labels
215
216     # Combine positives and negatives
217     X_train = pos_train_imgs + neg_train_imgs
218     y_train = pos_train_labels + neg_train_labels
219     X_test = pos_test_imgs + neg_test_imgs
220     y_test = pos_test_labels + neg_test_labels
221
222     # Train LBPH model
223     model = cv2.face.LBPHFaceRecognizer_create()
224     model.train(X_train, np.array(y_train))
225
226     # Evaluate
227     predictions = []
228     for face in X_test:
229         label_pred, _ = model.predict(face)
230         predictions.append(label_pred)
231
232     print(classification_report(y_test, predictions, target_names=[ "Other",
233     person]))
234
235 train_individual_models(base_path)
236
237
238 # In[32]:
239
240 import matplotlib.pyplot as plt
241
242 # Accuracy data
243 accuracy_per_person = {
244     'abhijeet': 0.95,
245     'ashmi': 0.97,
246     'avishkar': 0.47,
247     'khare': 0.89,
248     'krish': 0.78,
249     'maitreyee': 0.87,
250     'mayur': 0.82,
251     'naman': 0.74,
252     'nishad': 0.75,
253     'parth': 0.69,
254     'prathamesh': 0.85,
255     'sahaj': 0.48,
256     'satyam': 0.84,
257     'saubhagya': 0.88,
258     'sourab': 0.92
259 }
```

```
257 }
258
259 # Prepare data
260 names = list(accuracy_per_person.keys())
261 accuracies = list(accuracy_per_person.values())
262 indices = list(range(1, len(names)+1)) # [1, 2, ..., N]
263
264 # Plot
265 plt.figure(figsize=(8, 6))
266 plt.bar(indices, accuracies, color='skyblue', edgecolor='black')
267 plt.ylim(0, 1.05)
268 plt.xlabel("Person Number")
269 plt.ylabel("Accuracy")
270 plt.title("Accuracy per Person (Class) - LBPH Face Recognizer")
271 plt.grid(axis='y', linestyle='--', alpha=0.7)
272 plt.xticks(indices)
273
274 # Build numbered name list for legend
275 legend_text = "\n".join([f"{i}. {name}" for i, name in enumerate(names, start=1)])
276
277 # Add text box to the plot
278 plt.text(len(indices)+1.7, 0.95, legend_text, fontsize=9, va='top', ha='left',
279         bbox=dict(facecolor='white', edgecolor='black', boxstyle='round', pad=0.5))
280
281 plt.tight_layout()
282 plt.show()
283
284
285 # In [ ]:
```

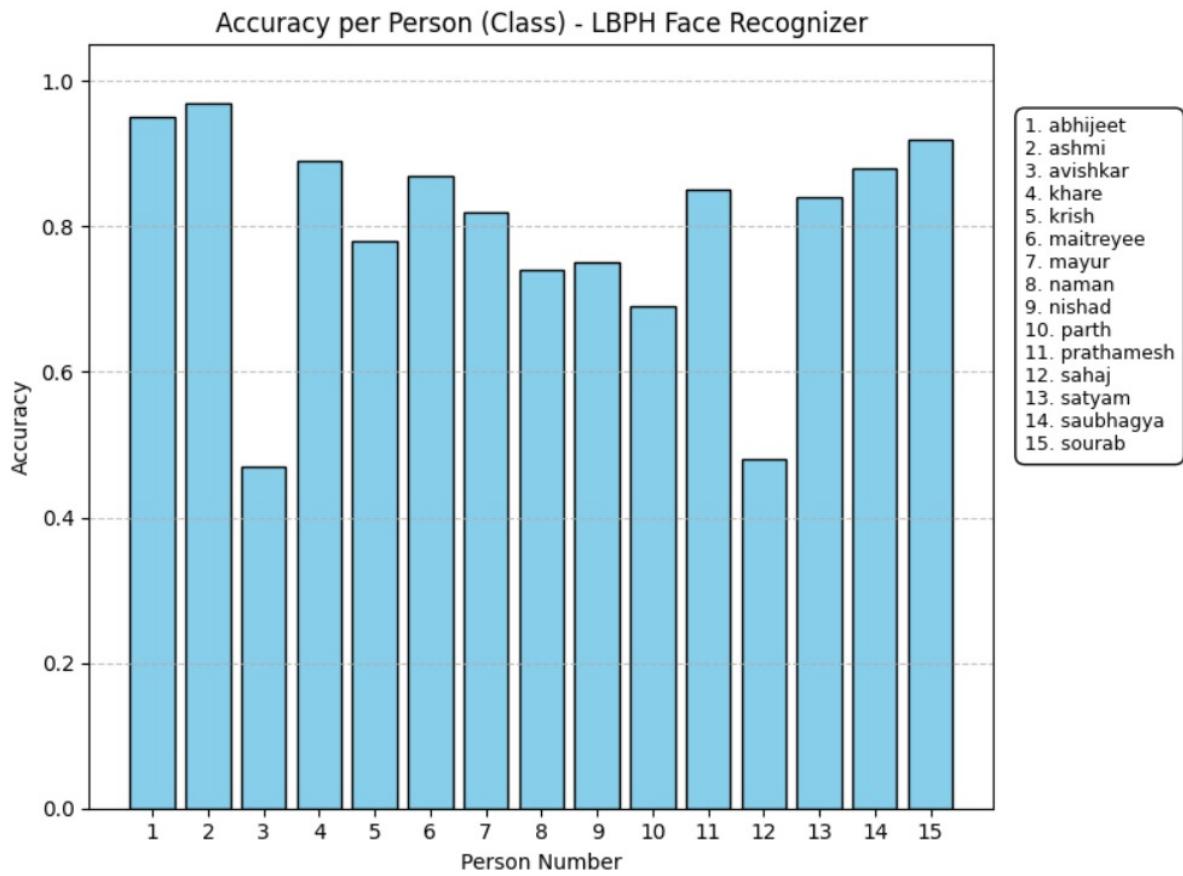


Figure 3.10: Accuracy per Person (Class) - LBPH Face Recognizer

Chapter 4

Future Prospects for the Attendance Assistant

Building on our implementation of an automated facial-recognition-based attendance assistant, several avenues can significantly enhance its performance, scalability, and real-world viability:

4.1 Integration of advanced Deep-Learning Models

- **Adopt Lightweight CNNs and Vision Transformers:** Replace or augment traditional feature-based methods (e.g., Eigenfaces, Fisherfaces) with compact convolutional neural networks (MobileNet, EfficientNet) or vision-transformer variants to boost accuracy under varied lighting and poses—while still enabling on-device inference.
- **Hybrid Pipeline Architecture:** Combine fast, classical face detection (e.g., OpenCV Haar cascades) with a secondary deep-learning re-identification stage to balance speed and precision in live classroom or office settings.

4.2 Dataset Expansion and Synthetic Augmentation

- **Larger, More Divers Training Sets:** Scale beyond our initial 5-person dataset (=18k crops) by collecting images across multiple sessions, cameras, and demographics to reduce bias and improve generalization.
- **GAN-Based Augmentation:** Leverage generative adversarial networks to synthesize varied facial expressions, occlusions (masks, scarves), and lighting conditions—ensuring robust attendance capture even when subjects wear accessories or move dynamically.

4.3 Edge Deployment and Resource Optimization

- **Model Compression Techniques:** Apply quantization and pruning to shrink model size for deployment on low-power edge devices (e.g., Raspberry Pi, embedded IP cameras), minimizing latency in real-time roll-call scenarios.
- **Hardware Acceleration:** Utilize on-board NPUs or Coral TPUs to offload inference, enabling simultaneous multi-camera streams for large lecture halls or multiple office entrances.

4.4 Privacy, Security, and Ethical Considerations

- **Federated Learning and on-Device Training:** Implement a federated learning framework so endpoints (e.g., classroom tablets) collaboratively improve the recognition model without sharing raw images—protecting sensitive biometric data by sharing only encrypted weight updates.
- **Anti-Spoofing and liveness Detection:** Integrate texture analysis and micro-motion cues to distinguish live faces from photographs or video replays, safeguarding against presentation attacks
- **Bias Auditing:** Regularly evaluate performance metrics (accuracy, false positives/negatives) across gender, age, and skin-tone strata to identify and mitigate algorithmic bias.

4.5 Multi-Modal and Context-Aware Fusion:

- **Biometric Fusion:** Biometric Fusion Augment facial data with voice recognition or RFID badge readings to confirm identity when face confidence scores drop below a threshold (e.g., under poor lighting).
- **Environmental Sensing:** Dynamically adjust recognition thresholds based on scene context—bright vs. dim classrooms, seated vs. standing students—to maintain both high recall and precision.

4.6 Broader Applications and Commercialization

- **Enterprise Time-Tracking Systems:** Extend the attendance assistant to corporate environments, integrating with HR systems for automated timekeeping and employee verification at entrances.
- **Smart-campus and IoT Integration:** Link attendance data with campus access control, library entry logs, and canteen payments to create a unified student experience.
- **Analytics Dashboard:** Offer administrators real-time dashboards showing attendance trends, tardiness patterns, and automated alerts for absenteeism spikes.

By pursuing these directions—grounded in our core implementation and the comparative analyses documented earlier—our Attendance Assistant can evolve into a robust, privacy-preserving, and widely deployable solution for both educational institutions and enterprises alike.

Chapter 5

Conclusion

In this seminar, we have discussed the various face recognition algorithms and techniques used in the field of computer vision. We have compared the performance of these algorithms based on accuracy, efficiency, and scalability. We have also discussed the advantages and disadvantages of face recognition technology and its applications in real-world scenarios.

- Face recognition technology is a powerful biometric technology that can be used for a wide range of applications, from security to personalization.
- There are several face recognition algorithms and techniques available, each with its own strengths and weaknesses.
- By comparing the performance of different face recognition algorithms and techniques, we can gain insights into their suitability for different applications.
- The evaluation metrics provide a quantitative measure of the performance of face recognition algorithms and techniques, helping us identify the most suitable approach for a given application.
- Face recognition technology has the potential to revolutionize various industries and improve the quality of life for individuals by providing secure and personalized services.

Bibliography

- [1] Paul, Sanmoy and Acharya, Sameer Kumar, A Comparative Study on Facial Recognition Algorithms (December 21, 2020). e-journal - First Pan IIT International Management Conference – 2018, Available at SSRN: <https://ssrn.com/abstract=3753064> or <http://dx.doi.org/10.2139/ssrn.3753064>
- [2] Kaur, P., Krishan, K., Sharma, S.K. and Kanchan, T., 2020. Facial-recognition algorithms: A literature review. *Medicine, Science and the Law*, 60(2), pp.131-139.
- [3] Kukula EP, Elliott SJ. Evaluation of a facial recognition algorithm across three illumination conditions. *IEEE Aerospace and Electronic Systems Magazine*. 2004 Sep;19(9):19-23.
- [4] Kukula EP, Elliott SJ. Evaluation of a facial recognition algorithm across three illumination conditions. *IEEE Aerospace and Electronic Systems Magazine*. 2004 Sep;19(9):19-23.
- [5] Emami S, Suciu VP. Facial recognition using OpenCV. *Journal of Mobile, Embedded and Distributed Systems*. 2012 Mar 30;4(1):38-43.
- [6] Chen J, Jenkins WK. Facial recognition with PCA and machine learning methods. In 2017 IEEE 60th international Midwest symposium on circuits and systems (MWSCAS) 2017 Aug 6 (pp. 973-976). IEEE.
- [7] Schenkel T, Ringhage O, Branding N. A Comparative Study of Facial Recognition Techniques: With focus on low computational power.
- [8] Paul, S. and Acharya, S.K., 2020, December. A comparative study on facial recognition algorithms. In e-journal-First Pan IIT International Management Conference–2018.
- [9] Delbiaggio, N., 2017. A comparison of facial recognition's algorithms.
- [10] Coe, J. and Atay, M., 2021. Evaluating impact of race in facial recognition across machine learning and deep learning algorithms. *Computers*, 10(9), p.113.
- [11] Dirin, Amir, Nicolas Delbiaggio, and Janne Kauttonen. "Comparisons of facial recognition algorithms through a case study application." (2020): 121-133.