

The software which we will obtain to identify faces has to realize faces irrespective of the facial expression or their facial orientation. This is possible through face landmark estimation; the procedure is to spot 68 points on the face. The points are so spotted that, they feature the eyebrows, circle the eyes, align the nose and lips. When this outlining is identified during the facial recognition time, the image is revolved or scaled such that the parallel lines are preserved and centered face is obtained to be correlated with the data sets.

The obtained unknown face from the outcome of the camera is now to be compared with

our huge datasets. Every organization or institution has numerous populations in which people might resemble each other's features. It is important to consider the factor of accuracy and misinterpretation of one person's face with another. For this problem, encoding is the solution; the face is encoded in 128 measurements considering the distance between eyebrows, length of the nose, etc. as shown in fig-2.2a and fig-2.2b. From the encoded image the name of the person is found by classifying the coded image through the dataset. The closest match is found to retain the accuracy. This utilizes the Eigen face approach to compare the dataset with the input image.



Fig – 2.2a: Input Image

```
[ -0.23259937 0.89573787 0.86281026 0.80185741 -0.83042841 -0.85549094
-0.81326437 -0.21617742 0.24921178 -0.87483435 0.39938849 0.86723037
-0.28134989 -0.28522881 -0.85428499 0.88758651 -0.21798581 -0.24883234
-0.84828288 -0.84918039 0.8196441 -0.83582587 0.81886629 0.83737824
-0.28749789 -0.22646874 -0.80757743 -0.21097047 0.87481185 -0.86782852
-0.8104848 -0.80912232 -0.7248764 -0.80243272 -0.86952283 0.35134224
0.82926545 0.80318297 0.20762389 -0.82858068 -0.28136253 -0.88988737
0.83071981 0.25188899 0.3662895 -0.81420718 0.83275681 0.83513227
0.8937149 -0.25187777 0.87617582 0.38818593 0.38933372 0.82642423
0.88179985 -0.24235686 -0.81446182 0.2882465 -0.2829788 -0.89289866
0.88149622 -0.87536867 -0.87223525 -0.88893645 0.27889545 0.27136425
-0.84139118 -0.21966112 0.24248982 -0.26828225 -0.83286726 0.86287528
-0.83863189 -0.22643731 -0.27338332 0.8628988 0.2272213 0.25817285
-0.25675418 0.86798884 -0.88846641 0.81248589 0.84172584 -0.83832173
-0.23989889 0.87529542 -0.24898888 -0.82184182 0.27734792 0.23985281
-0.85941385 0.24232861 0.88883187 0.84017748 0.86636186 -0.81888764
-0.87871819 0.88888775 -0.21447721 0.8182787 0.85812545 -0.8879484
0.85342542 0.87396541 -0.89088877 0.87536927 0.81186154 0.8366048
0.88879432 0.21678988 -0.23486135 -0.84884982 0.87790545 -0.28888234
0.38147621 0.21922574 0.88882876 0.25158772 0.28888333 0.83949629
0.82893947 0.81135381 -0.89548854 -0.8863687 0.8597992 -0.33128859
0.22187879 0.21246784]
```

Fig – 2.2b: 128 measurements of input image

Eigen face approach:

Eigen face scenario is the facial recognition algorithm that moves toward linear algebra as well as dimensionality contraction as the principal component analysis. The training images database is converted into vectors and then the average of all these face vectors is evaluated to subtract it from individual vectors. A matrix of face vectors is so made to locate the covariance matrix which is to be utilized to build Eigen-values and vectors. The Eigen-face coefficients depict the training database which is used to distinguish the portraits from the CCTV. The linear mixture of Eigen-faces is obtained by casting the normalized vector into the Eigen-space. The obtained result

vector is subtracted from the train vectors to obtain the least length between test and train, the lesser the displacement accurate is the prediction.

III. LITERATURE SURVEY

The prominent procedure utilized in the face recognition system is primarily the front face of the human being, some of the techniques include neural networks, hidden Markov model, face matching is done geometrically, and template matching. The most widely used tool is Eigen-face is also called the principal element in mathematical terms. The eigenvectors are used to represent various proportions of the variations in the faces.

An ANN (artificial neural network) was

used in face recognition which contained a single layer, which shows adaptiveness in crucial face recognition networks. The face verification is done using a double layer of WIZARD in neural networks. The other option is graph matching. By using this method we can formulate object as well as face recognition.

Hidden Markov Models (HMM) the model applied to human face recognition where the faces get divided into parts such as the eyes, nose, ears, etc. The face recognition and correct matching are 87% correct, matching always gives out the best and right choice of face detection through the stored dataset. Geometrical feature matching is a technique that is based on the geometrical shapes of the face. The geometrical face configuration has a sufficient dataset for face detection and recognition system. Template matching is one of the processes through which the test image is represented as a two-dimensional array of values which can be compared using Euclidean distance with an individual template representing the whole face. This method can also use more than one face template from several points of view to depict an individual face.

IV. ADAVANTAGES

- The execution of this system is simple and cost-effective.
- It saves time and human effort.
- We don't need any additional data resources.
- It can be used in multiple industries for multiple purposes.
- It can detect many people at the same time.
- Obtained data can be further manipulated and analyzed.

V. DISADVANTAGES

- The chance of fraudulence is not completely precise.
- The distance of the person from the camera is limited.
- There will be a problem in recognizing identical twins.

SOURCE CODE:

```
import os
import datetime
import numpy as np
import cv2
import face_recognition
# from PIL import ImageGrab
path = 'ImagesAttendance'
images = []
classNames = []
```

```
myList = os.listdir(path)

print(myList)
for cl in myList:
    curImg = cv2.imread(f'{path}/{cl}')
    images.append(curImg)
    classNames.append(os.path.splitext(cl)[0])
print(classNames)

def findEncodings(images):
    encodeList = []
    for img in images:
        img = cv2.cvtColor(img,
            cv2.COLOR_BGR2RGB)
        encode =
            face_recognition.face_encodings(img)[0]
        encodeList.append(encode)
    return encodeList

def markAttendance(name):

    with open('Attendance.csv', 'r+') as f:

        myDataList = f.readlines()

        nameList = []
        for line in myDataList:
            entry = line.split(',')
            nameList.append(entry[0])

            if name not in nameList:
                now = datetime.datetime.now()
                dt = datetime.date.today()
                dtString = now.strftime('%H:%M')
                f.writelines(f'\n{name},{dt},{dtString}')

encodeListKnown = findEncodings(images)
print('Encoding Complete')

cap = cv2.VideoCapture(0)

while True:
    success, img = cap.read()

    imgS = cv2.resize(img, (0, 0), None, 0.25, 0.25)
    imgS = cv2.cvtColor(imgS,
        cv2.COLOR_BGR2RGB)

    facesCurFrame =
        face_recognition.face_locations(imgS)
    encodesCurFrame =
        face_recognition.face_encodings(imgS,
            facesCurFrame)

    for encodeFace, faceLoc in
        zip(encodesCurFrame, facesCurFrame):
```

```

matches =
face_recognition.compare_faces(encodeListKnown
, encodeFace)
faceDis =
face_recognition.face_distance(encodeListKnown,
encodeFace)

matchIndex = np.argmin(faceDis)
if matches[matchIndex]:
name =
classNames[matchIndex].upper()
y1, x2, y2, x1 = faceLoc
y1, x2, y2, x1 = y1 * 4, x2 * 4, y2 * 4, x1 * 4

cv2.rectangle(img, (x1, y1), (x2, y2), (0, 255, 0),
2)
cv2.rectangle(img, (x1, y2 - 35), (x2, y2), (0, 255,
0), cv2.FILLED)
cv2.putText(img, name, (x1 + 6, y2 - 6),

```

```

cv2.FONT_HERSHEY_COMPLEX, 1, (255, 255,
255), 2)
markAttendance(name)
print(name)
print(faceDis)
cv2.imshow('Webcam', img)

cv2.waitKey(1

```

VI. RESULTS

The results of the project, executed using the above source code are as shown in the fig-6.1, fig-6.2, fig-6.3. The results show the facial recognition of the student enabled with face tracking to monitor the movements of the student in the class. We have also successfully managed to mark attendance of the student recognized by algorithm automatically in an excel sheet in CSV format.



Fig-6.1



Fig-6.2



Fig-6.3

VII. CONCLUSION

The facial recognition system is employed in various other sectors like banking, security purposes, industries, crime detection, etc. Using this strategy in universities helps students and lecturers save time and perform analyses of attendees as data can be stored and manipulated through algorithms in the system. Further, the project can be taken to analyze emotions, their face time towards the lecture which helps in calculating student's individual activity. When the classroom size is huge, the class can be virtually divided into a matrix to find the density of population through heat mapping. Through this, we can control the electrical appliances like lights and fans in the classroom autonomously.

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