

# Attendance System with Face Recognition using OpenCV

Project report submitted in partial fulfillment of the requirement  
for the degree of Bachelor of Technology

In

## Computer Science and Engineering

By

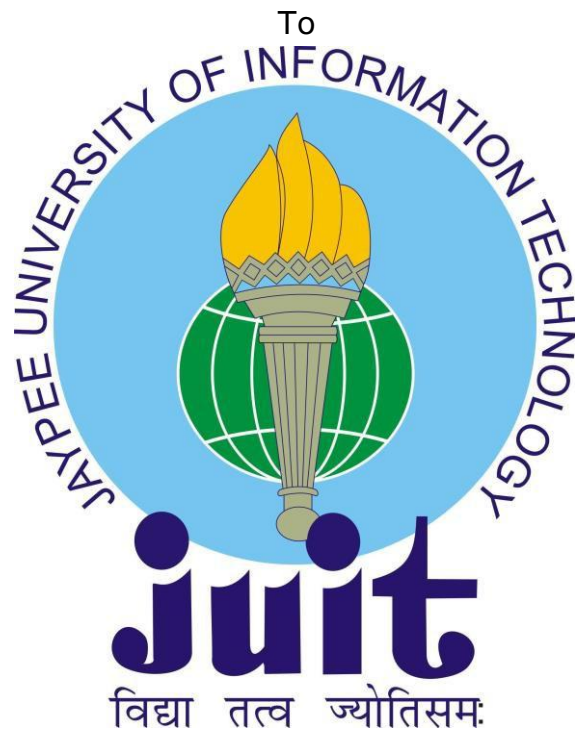
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To



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## **Candidate's Declaration**

We hereby declare that the work presented in this report entitled “**Attendance System with Face Recognition using OpenCV**” in partial fulfillment of the requirements for the award of the degree of Bachelor of Technology in Computer Science and Engineering Technology submitted to the department of Computer Science & Engineering and Information Technology, Jaypee University of Information Technology Waknaghat is an authentic record of my own work carried out over a period from Jan 2021 to May 2021 under the supervision of **Dr.Rakesh Kanji**. The matter embodied in the report has not been submitted for the award of any other degree or diploma.

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**Prikshit Dogra (161226)**

This is to certify that the above statement made by the candidate is true to the best of my knowledge.

**Dr. Rakesh Kanji**  
**Assistant Professor**  
**Department of Computer Science**

## Acknowledgement

It is our privilege to express our sincerest regards to our project coordinator, Dr. Rakesh Kanji, for his valuable inputs, able guidance, encouragement, whole-hearted cooperation and constructive criticism throughout the duration of our project and without his support and guidance it would not have been possible for us to successfully implement our project. We deeply express our sincere thanks to the Department of Computer Science and technology for encouraging and allowing us to present the project on the topic "**Attendance System with Face Recognition using OpenCV**" providing us the necessary opportunities for the completion of our project and owe our debt to them for their invaluable help and guidance.

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## **Abstract**

The objective of this venture is to make a proficient participation framework which uses face acknowledgment instead of biometric strategies. The objective is to make the frontend for educators as theoretical and easy to use as could really be expected. At the point when the instructor begins the participation system, the backend script runs which utilizes a web camera or some other picture catching gadget and distinguishes the individual present in the casing, if the individual is introduced in the edge is additionally present in the information base at the backend then participation is made. Nearby parallel example histograms are utilized for its high effective outcomes. Clients of the application can snap a photo under various conditions, the pre-owned picture acknowledgment calculation needs to recognize faces in outline under various conditions like low light, picture obscure and so on

For the frontend, NodeJS is utilized because of its non concurrent execution, which assists with a client - cordial front end for instructors which is straightforward.

# Chapter 1: Introduction

## **(1.1) Project Introduction:**

Face acknowledgment is described as discovering the circumstance of the substance of an individual. By the day's end it might be portrayed as discovering the face region in an image. In the wake of recognizing the substance of people its facial features are isolated and have a wide extent of uses like visible presentation affirmation, face affirmation, discernment systems, human PC interface, and so forth... Distinguishing faces in an image of a single individual is straightforward yet when we consider a social event image of an image containing various faces, the endeavor gets irksome. For the utilization of face affirmation, area of face is critical and the underlying advance. Ensuing to recognizing face the face affirmation computation should be utilitarian. Face acknowledgment itself incorporates a couple of intricacies for example climate, positions, brightening, etc. There are some current strategies for acknowledgment of face. Some of them are skin concealing based, brand name or feature based (counting mouth, nose and eyes) and neural framework based. Face ID and affirmation isn't new in our overall population we live in. The constraint of the human character to see explicit individuals is stunning. It is astonishing how the human character can regardless suffer in the ID of explicit individuals even through the movement of time, disregarding slight changes by all accounts.

## **(1.2) Statement of Problem:**

This undertaking is being done on account of the concerns that have been included on the methodologies which talks use to measure investment during addresses. The usage of clickers, ID cards swiping and actually recording names on a piece of paper as a technique to follow understudy orderlies has affected this endeavor to be finished. This isn't in any way to reproach the various methodologies used for understudy cooperation, anyway to fabricate a system that will separate the amount of faces present in a homeroom similarly as recollecting that them. Moreover, a teacher will have the alternative to tell if an understudy was clear as these procedures referred to can be used by anyone for support records, anyway with the face disclosure and affirmation structure set up, it will be anything besides hard to recognize whether an understudy is truly present in the homeroom or not. This system will not simply improve homeroom control during addresses, it will moreover possibly perceive faces for understudy

support purposes. We will use Python and NodeJS with its different libraries to amass and realize this system.



### **(1.3) Objective:**

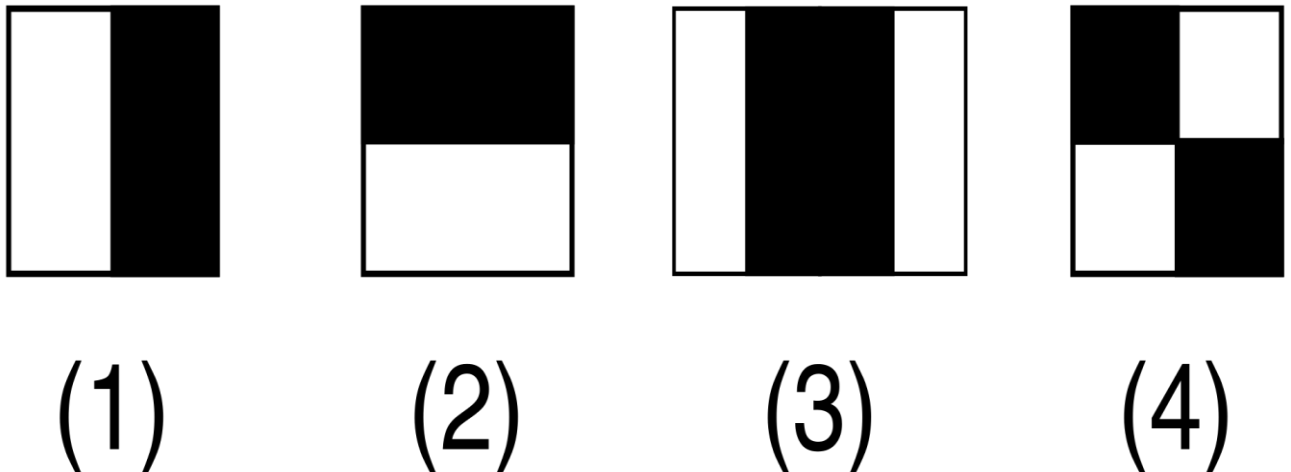
Perceive understudies put away on a data set of countenances by coordinating with them to pictures on a data set with a precision inside half. The framework ought to have the option to coordinate with identified understudies' appearances edited from a picture to those on a data set on the framework. The framework ought to have the option to handle a picture inside 10 minutes to have the option to accomplish the goal of acknowledgment before the finish of a talk. for example 5 names each hour for every talk. The calculation carried out for the framework's usefulness will accomplish framework exactness inside 50%.The positive expectation ought to be inside half. The framework planned will be easy to use with a Graphical UI that will fill in as an admittance to the functionalities of the framework

### **(1.4) Methodology:**

In our proposed framework, the framework is started up by the PC. After it starts then the framework begins preparing the picture of the understudies for which we need to stamp the participation. Picture Catching stage is one in which we catch the picture of the understudies. This is the fundamental stage from which we begin introducing our framework. We catch a picture from our camera which overwhelmingly checks for specific limitations like lightning, dispersing, thickness, looks and so forth The caught picture is unflinching as indicated by our necessities. When it is unfaltering, we ensure it is either in .png or .jpeg design. We take diverse front facing stances of an individual so the exactness can be accomplished the most extreme degree. For the caught picture, from each article we recognize just front facing faces. This identifies just faces and eliminates each and every other part. These recognized appearances are put away some place in the information base for additional enquiry. Highlights are removed in the extraction stage.

## Chapter 2: Literature Survey

### (2.1) Face identification:



**Fig 2.1 Haar features**

Item acknowledgment is one of the PC propels, which is related with the image taking care of and PC vision, collusion with various events of an article, for example faces, structures, trees, vehicles, and so forth The essential piece of face arrange calculation is to choose whether there is a human face in the picture or not.

Face ID is a technique of discovering face inside an image layout, paying little brain to the character of that face. Prior to seeing a face, it is the primary principal to perceive and eliminate the appearances from the main pictures. Face ID/location centers around finding the countenances in an image and focusing them to be utilized by the identification of face affirmation estimation. Recently, various procedures are liked for distinguishing proof of a face in a picture. In face recognition strategies, the people who are depending after getting ready sets to get the huge lopsidedness in facial features have charmed a great deal of thought and given the best results. Generally these procedures channel the data picture and give the yield of a presence of face or its nonattendance. Viola and Jones showed an effective area strategy using Haar-like features and AdaBoost as a lively planning calculation. For recognizing a face, the calculations take a gander at just the faces present. Another constituent in the picture that isn't important for a face are dismissed or unapproachable.

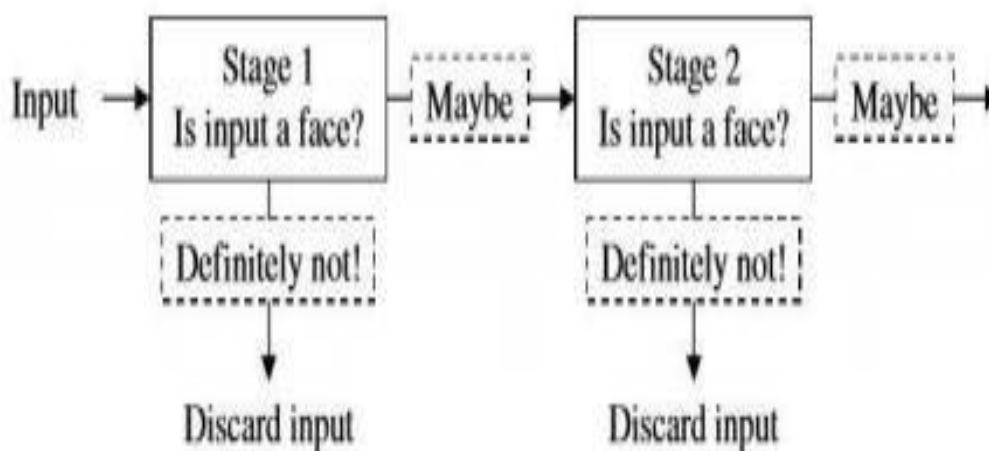
Mainly Classification(Face) is based on four kinds of features:

1. Edge Feature
2. Edge Feature
3. Line Feature
4. Four Rectangle Features

Each point corresponds to each feature in **figure 2.1** above.

### **(2.1.1) Haar Cascade classifiers and Haar Features**

The essential state of the Viola Jones face distinguishing proof is to inspect the locater effectively with the utilization of a comparable picture. Another size is inspected at each time. Whether or not a picture should contain at any rate a solitary example of face is by no uncertainty an irrational course of bogus condition i.e Nonattendance of a face. This acknowledgment flags an alternate issue: Rather than recognizing faces, the assessment should dispose of parts with nonattendance of appearances. It is speedier to dispose of nonattendance of a face as opposed to recognizing a face. With regards to a locater comprising of one (in number) classifier gives off an impression of being inefficient, the calculation time is consistent notwithstanding. In this manner the need for a full classifier launches. The full classifier is built of different various stages. Each stage is to finish up whether a given picture part is certainly missing of face or presence of a face. At the point when a sub picture is assembled to be missing of face by a given stage, at that point it is being disposed of. In this manner, a sub picture shows a potential face in picture is given to the resulting stages in the course. It seeks after bigger the quantity of stages a sub picture goes through.

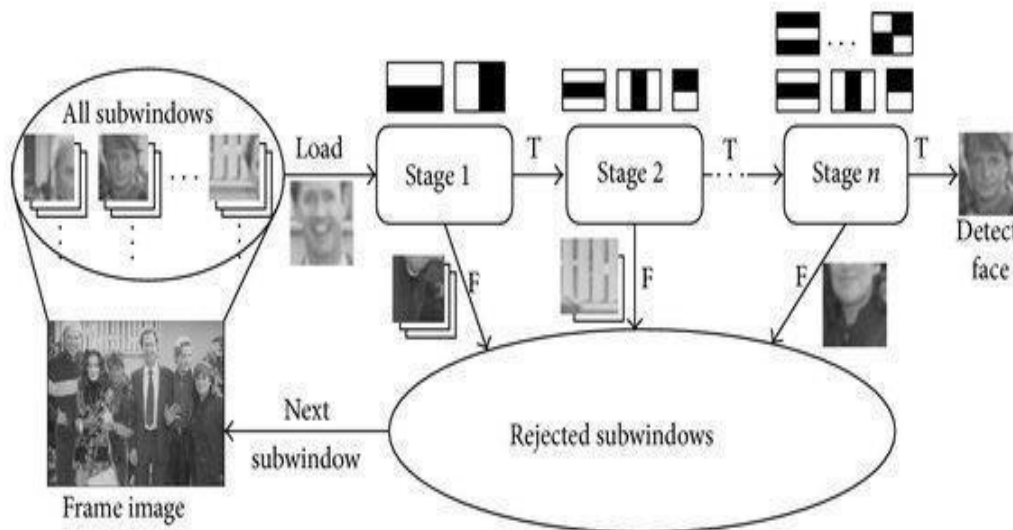


**Fig 2.2 Face Detection**

Bogus negative cases are acknowledged for diminishing bogus positive numbers and it is done at one phase of the classifier.

Bogus positives are not viewed as an obstruction or an issue inside the principal phase of classifier on the grounds that the accompanying phases of the classifier will characterize them. This is the motivation behind why in the Viola Jones calculation, bogus positives are acknowledged in starting phases of classifier. Following every one of the stages, in the last phase of classifier the quantity of bogus negatives are undeniably less. Course classifiers are likewise portrayed as attentional classifiers by Viola and Jones. Attentional is on the grounds that enormous consideration

is focused on the districts where there is high likelihood that face is available. May there be  $n$  negative examples, at that point while preparing the set at any stage, every one of the negatives should be bogus negatives which are made at stage past to  $n$  for example  $(n-1)$ .



**Fig 2.3 Channels & Sub Windows**

The thinking is passing these channels on the image, exploring the slightest bit (or window) at that point. By then, for each window, all the pixel forces of, independently, white and dull sections are added. Finally, the value acquired by taking away those two summations is the assessment of the component removed. Ideally, a phenomenal assessment of a part infers it is significant. A Haar-Highlight is similarly as a piece in CNN, on the other hand, really in a CNN, the assessments of the piece are directed by getting ready, while a Haar-Highlight is actually settled. At the point when Haar features are applied to pictures. Every segment achieves a singular worth which is dictated by deducting the total of pixels under white square shape from the entire of pixels under

dim square shape. Each Haar incorporates a kind of resemblance to perceive a piece of face. Viola Jones uses  $24 \times 24$  as base window size and finds out the above features wherever all through the image moving by 1 PX. In case we consider each and every comprehensible boundary of the haar features like position, scale and type we end up figuring about 160,000+ features. So we need to evaluate colossal plan of features for each  $24 \times 24$  PX.

```

45 <opencv_storage>
46 <cascade type_id="opencv-cascade-classifier"><stageType>BOOST</stageType>
47   <featureType>HAAR</featureType>
48   <height>24</height>
49   <width>24</width>
50   <stageParams>
51     <maxWeakCount>211</maxWeakCount></stageParams>
52   <featureParams>
53     <maxCatCount>0</maxCatCount></featureParams>
54   <stageNum>25</stageNum>
55   <stages>
56     <_>
57       <maxWeakCount>9</maxWeakCount>
58       <stageThreshold>-5.0425500869750977e+00</stageThreshold>
59       <weakClassifiers>
60         <_>
61           <internalNodes>
62             0 -1 0 -3.1511999666690826e-02</internalNodes>
63           <leafValues>
64             2.0875380039215088e+00 -2.2172100543975830e+00</leafValues></_>
65         <_>
66           <internalNodes>
67             0 -1 1 1.2396000325679779e-02</internalNodes>
68           <leafValues>
69             -1.8633940219879150e+00 1.3272049427032471e+00</leafValues></_>
70         <_>
71           <internalNodes>
72             0 -1 2 2.1927999332547188e-02</internalNodes>

```

**Fig 2.4 (haarcascade frontal face.xml) file code**

### **(2.1.2) Local Binary Pattern Histograms(LBPH)**

Local Binary Pattern or LBP for short is a basic however profitable strategy which names the pixels of a picture by isolating the space of each pixel returns an arranged pair. In view of its straightforward working, nearby double example strategy has become a standard methodology in different picture applications and surprisingly signal preparing. The most momentous property of the LBP strategy is its applications and its single calculation to plane lower the high changes undertaking, for example, by various enlightenments of pictures. Another imperative element is its basic computational ease, which makes it turn out in any event, for pictures with various setting, enlightenments, revolutions and so forth

The key methodology of Nearby Paired Example based face ID is as per the following: The face in the picture is isolated into neighborhood territories and its pixel identifier is delivered from each district unreservedly. The descriptors are then connected to outline an overall depiction of the face, as showed up in Fig 2.5. The above histogram shows the face with 3 particular degrees: LBP histogram from each square contains data of pixel esteems, names are added. u2 Let LBPP,R be the Nearby Twofold Example operator.(P,R) requested pair addresses the neighborhood of the administrator. Naming all examples aside from the uniform example and marking the rest with a solitary name is addressed by the variable u2.vThe picture is put away the in the capacity named fl(x,y) with a name l, histogram is created is produced utilizing the accompanying equation:

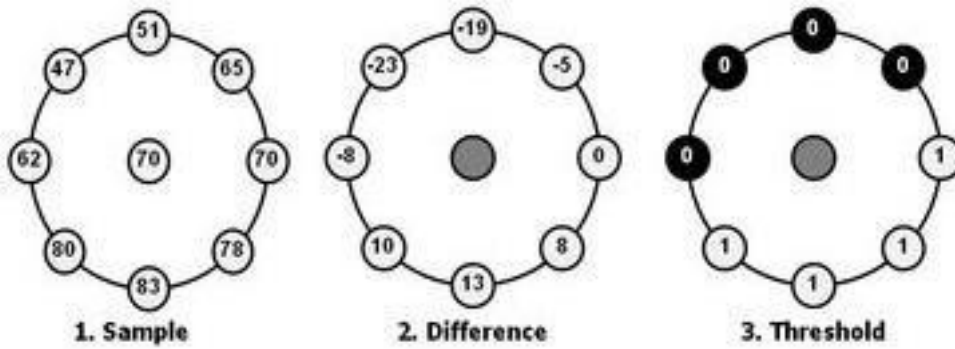
$$H_i = \sum_{x,y} I\{f_l(x,y) = i\}, i = 0, \dots, n - 1,$$

Where n addresses check of names created by the Nearby Twofold Example administrator LBP(P,R)u2. In the event that An is valid, l=1 else l=0. Histograms are standardized utilizing following:

$$N_i = \frac{H_i}{\sum_{j=0}^{n-1} H_j} .$$

The value of the LBP code of a pixel  $(x_c, y_c)$  is given by:

$$LBP_{P,R} = \sum_{p=0}^{P-1} s(g_p - g_c) 2^p \quad s(x) = \begin{cases} 1, & \text{if } x \geq 0; \\ 0, & \text{otherwise.} \end{cases}$$



$$1 \cdot 1 + 1 \cdot 2 + 1 \cdot 4 + 1 \cdot 8 + 0 \cdot 16 + 0 \cdot 32 + 0 \cdot 64 + 0 \cdot 128 = 15$$

4. Multiply by powers of two and sum

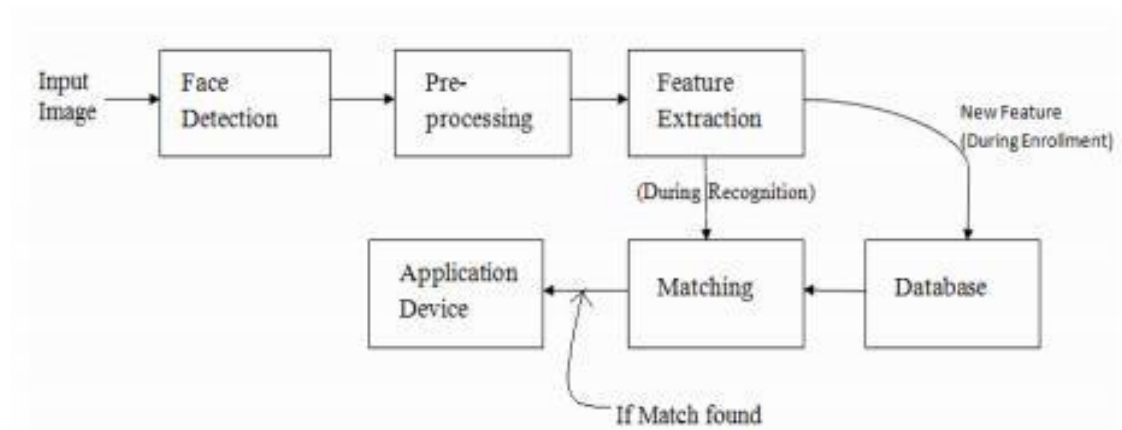
**Fig 2.5 LBP computation of a pixel**



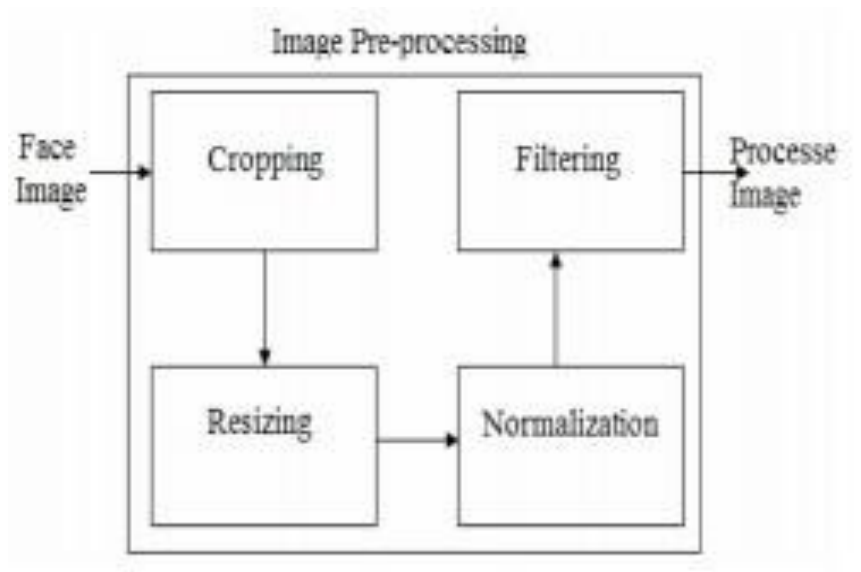
**Fig 2.6 (LBP Image)**



## (2.2) Pre-processing:



**Fig 2.7 Flow graph of Face recognition(plus identification)**



**Fig 2.8 Block diagram of Image Pre-processing**

### (2.2.1) Face detection and cropping:

Face editing is moreover a critical task to achieve high recognition rate. Altering ought to be conceivable using diverse face disclosure frameworks. Face ID incorporates perceiving a face from an image using all out (picture based strategy) or by recognizing in any event one feature from the image (Highlight based system, for instance, nose, eyes, lips, etc. Face disclosure ought to in

like manner be conceivable ward on unique shape models, for instance, discovering head limits.

### **(2.2.2) Image Normalization**

The primary issue which emerges in face acknowledgment is fluctuating light/brightening, establishes the greatest difficulties. Pictures that have changing lighting which are wild establish lopsided intricacy. For instance, the dissipating of levels which are dull are undefined. Histogram evening out system is utilized to adapt up to this issue. Given a  $M \times N$  picture or a network, Pixel evening out esteem i.e  $h$  on various  $l$  levels is discovered utilizing the CDF which is known as Combined Dispersion Capacity.

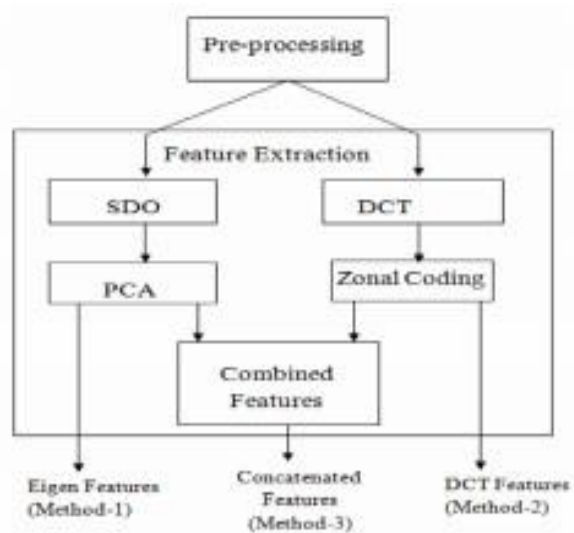
$$h = \left( \frac{CDF(V) - CDF_{min}}{MXN - CDF_{min}} \right) (L - 1)$$

---

### (2.2.3) Image Denoising and Filtering

Pictures are often as per usual have Gaussian commotion on account of light assortments. LPF of low pass channels is extremely helpful to eliminate the frequencies of high worth.

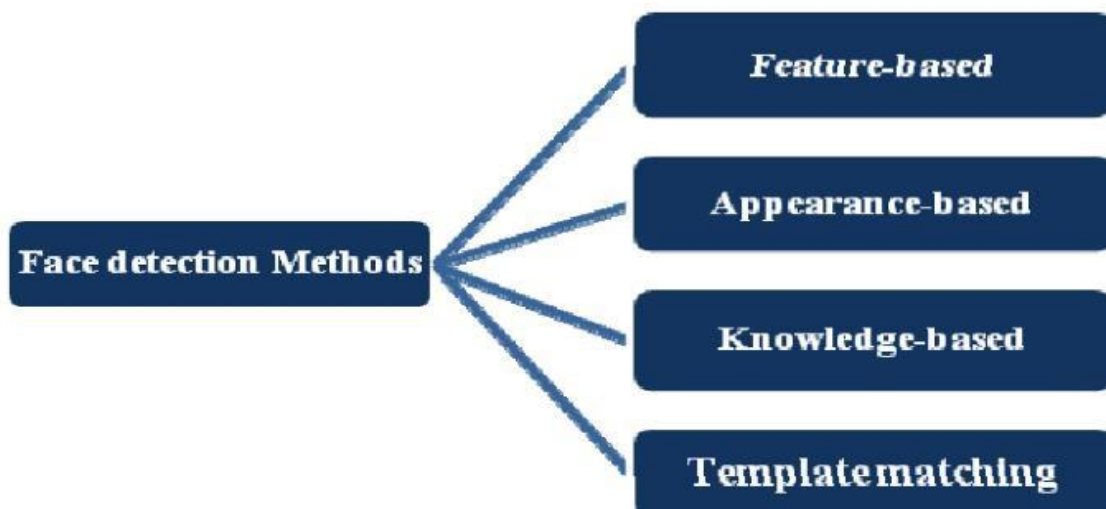
LPF passes data with a cut off.



**Fig 2.9 Block diagram for Pre-processing**

### (2.3) Detection:

To detect a face there are multiple stages, namely:



## **1. Knowledge-Based**

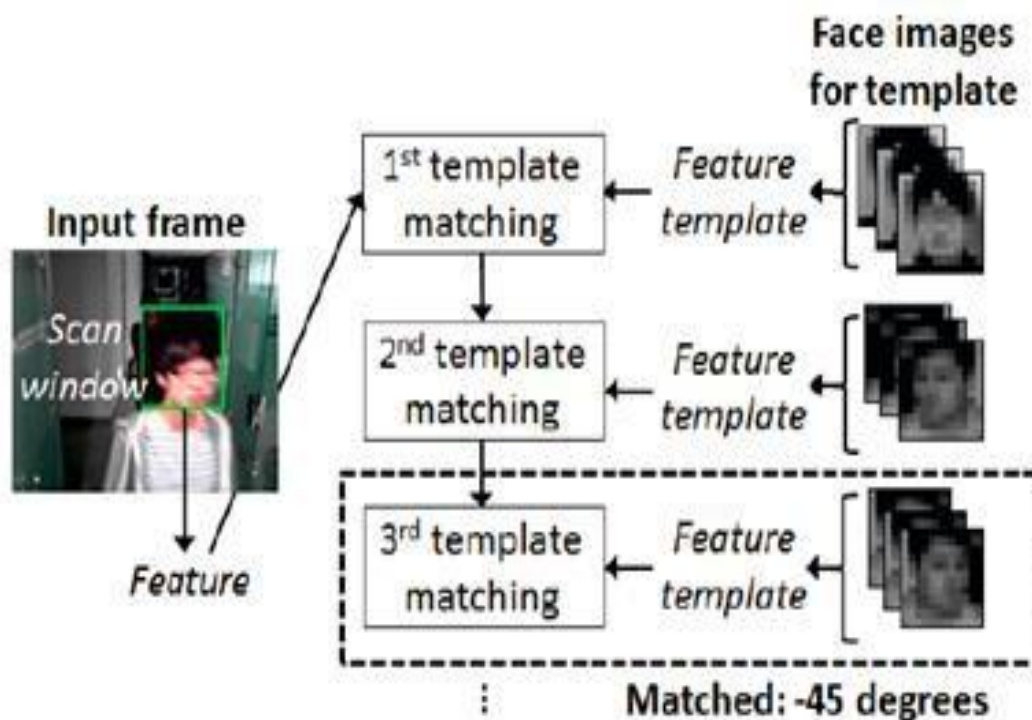
This method is subject to the game plan of rules. To distinguish a face, it requires some pre data that is the reason it is called Information based. It is unmistakable that a face comprises some regular highlights like eyes, mouth, nose and so on The gigantic issue with these procedures is the difficulty in building a fitting game plan of rules. This procedure autonomously isn't adequate to recognize different occurrences of countenances in numerous pictures.

## **2. Feature-Based**

This procedure disposes of assistant features of the face in the picture. At first it is characterized as a classifier at that point utilized as a separator among regions with faces and not appearances. The contemplation is to vanquish the marks of imprisonment of our normal data on faces. This strategy is isolated freely into various stages and furthermore functions admirably for pictures with various countenances. This procedure whenever applied well has an exactness of above 85%.

## **3. Template Matching**

This procedure utilizes effectively recognized face layouts to distinguish the highlights of a face in a given picture by the relationship of the configurations and format pictures. For instance, Face can be isolated into various sub highlights like eyes, nose, mouth and so forth By utilizing edge recognizable proof strategies, a rough face could be found with given likelihood.



**Fig 2.10 Template matching**

#### **4. Appearance-Based**

This strategy relies on the preparation of appearances from numerous pictures. This method is for the most part better compared to some other strategy above. Appearance-put together procedures depend with respect to techniques from exact assessment, man-made intelligence to recognize the urgent highlights of face pictures. This technique similarly used in incorporate extraction for face affirmation.

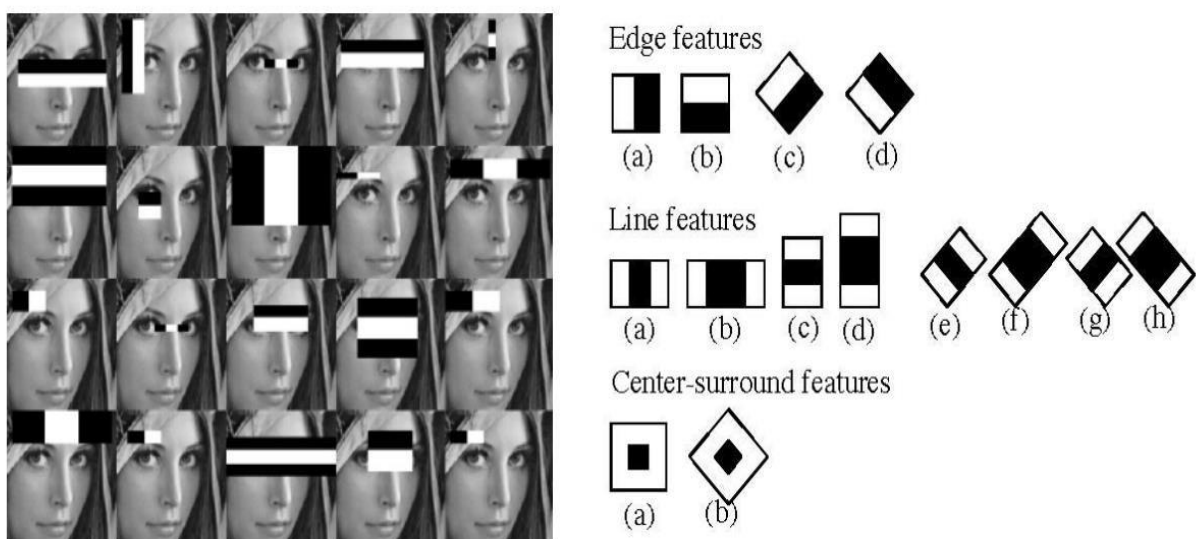
#### **How detection works:**

There are various frameworks to recognize faces, face acknowledgment should be possible with high precision utilizing these frameworks or methods. Every one of these strategies or techniques are by and large comparable working inside. For instance, OpenCV, Neural Organizations and so on depend on normal standards however carried out in an unexpected way. The fundamental objective of face discovery is to return whether a face is available or missing in a picture. We have

utilized OpenCV for our venture, face acknowledgment in addition to distinguishing proof utilizing OpenCV fills in as follows

Picture is perused/imported. At that point the picture is changed over to grayscale from RGB(Red Green Blue). At that point arrangement, trimming, resizing of the picture is finished.

Next stage utilizes Haar Highlights which is outwardly depicted in the picture beneath. These Haar highlights were characterized by Viola Jones. There are numerous normal similitudes in various faces, for example, the space of eye is hazier than rest of the space, additionally the nose region is lighter than different territories. This calculation characterized is called Viola Jones Calculations and it utilizes Haar course classifier.



**Fig 2.11 Haar Features Extraction**

Viola Jones calculation is moreover utilized for face include determination or highlight extraction for a specific component in the picture, help of edge acknowledgment, center area for perceiving other basic highlights like eyes, mouth, etc in the picture. It is utilized to pick the central features in an image and concentrate these features for face recognizable proof.

Following stages give the headings of different directions a,b,c, d which characterize a square shape, restrict the picture to show the space of the face. Then square shape is produced then restricts the region of the picture where face is recognized. Additionally, unique other area techniques that are

used together for recognizable proof, for eg, grin identification, state of mind acknowledgment, flicker acknowledgment, and some more.



```

import cv2
face_cascade = cv2.CascadeClassifier('cascades/data/haarcascade_frontalface_default.xml')
eye_cascade = cv2.CascadeClassifier('cascades/data/haarcascade_eye.xml')
cap = cv2.VideoCapture(0)
while 1:
    ret, img = cap.read()
    gray = cv2.cvtColor(img, cv2.COLOR_BGR2GRAY)
    faces = face_cascade.detectMultiScale(gray, 1.3, 5)

    for (x,y,w,h) in faces:
        cv2.rectangle(img, (x,y), (x+w,y+h), (255,255,0), 2)
        roi_gray = gray[y:y+h, x:x+w]
        roi_color = img[y:y+h, x:x+w]
        eyes = eye_cascade.detectMultiScale(roi_gray)

        for (ex,ey,ew,eh) in eyes:
            cv2.rectangle(roi_color, (ex,ey), (ex+ew,ey+eh), (0,127,255), 2)

    cv2.imshow('img',img)
    k = cv2.waitKey(30) & 0xff
    if k == 27:
        break
cap.release()
cv2.destroyAllWindows()

```

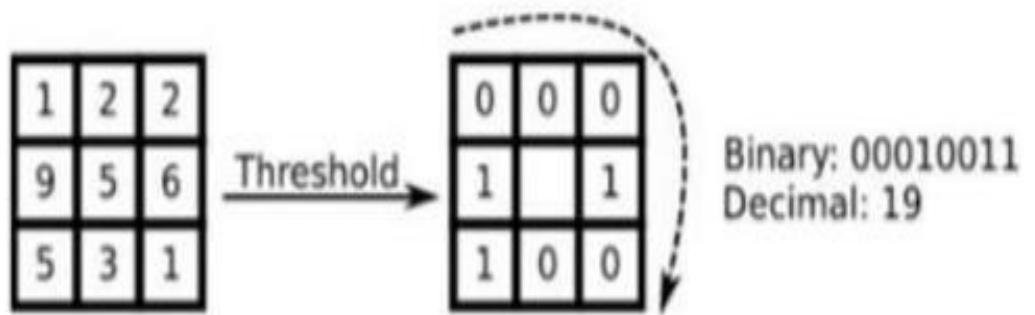
**Fig 2.12 Program for face feature detection using OpenCV**

## **(2.4) Algorithm:**

### **(2.4.1) LBPH(Local Binary Pattern Histogram)**

Neighborhood Double Example (LBP) is a fundamental yet capable surface manager which denotes the pixels of an image by thresholding the space of each pixel and considers the result an equal number. As a result of its discriminative force and computational ease, the LBP surface overseer has become a notable approach in various applications. It will in general be seen as a limiting together approach to manage the for the most part different quantifiable and essential models

of surface examination. Possibly the main property of the LBP overseer in obvious applications is its capacity to monotonic faint scale changes caused, for example, by lighting up assortments. Another huge property is its computational ease, which makes it possible to research pictures in testing progressing settings.



**Fig 2.13 Pixel LBP Conversion**

## **Advantages of LBPH algorithm**

- LBPH Technique is uncommon contrasted with other performing surface descriptors.
- The LBPH manager is hearty towards expanding or diminishing changes in dim scale.
- Fisher Appearances doesn't permit features of a person to get amazing, yet regardless of all things, it decides illumination assortment as a supportive part. In any case, light assortments are authentically not an accommodating component to eliminate as it's anything but a piece of the veritable face.
- Fisher faces requests enormous storing of face information and data and even more planning time in its affirmation.
- In LBPH each image is separated effectively, while the Eigen countenances and strategies of fisher faces looks at all the datasets.
- Techniques of LBPH will probably work better to fisher faces in various circumstances and distinctive light conditions. It generally relies on our arrangements and testing enlightening assortment of information.
- Additionally, it can collaborate with close by features present in the photos.
- Both side and front faces can be seen by LBPH.

## Chapter 3: System Development

### (3.1) Programming environment and framework

#### Programming Languages:

##### Python :

Python is an intuitive and a general item situated prearranging language. Python language is intended to be profoundly discernible by clients. It utilizes English words oftentimes then again different dialects utilize accentuation, and it has less syntactic developments in contrast with different dialects.

##### Characteristics:

- It upholds object arranged programming techniques just as utilitarian and organized programming strategies.
- It can be gathered as byte-code for growing huge estimated applications or can be used as a prearranging language.
- It upholds dynamic sort checking and gives us exceptionally undeniable level unique information types.
- Programmed assortment of trash is additionally upheld.
- It is effectively incorporated with java, C++, COM, ActiveX and CORBA,

##### Javascript (NodeJS):

JavaScript (JS) is a light, effectively deciphered, or in-time incorporated programming language with top of the line capacities. It is generally known as the prearranging language for Web pages, it is also used by numerous non-program

conditions ,specifically Adobe Gymnastic performer, Node.Js and Apache.  
JavaScript is a model based, multi-design, single-strung, powerful language,  
supporting article arranged, fundamental, and practical programming styles.

## XML:

XML full structure is Extensible Markup Language. XML is a markup language like HTML which is utilized to characterize information. XML is reasonable both by human and machine and it is versatile and simple to create. XML labels are not predefined in XML, accordingly, we need to characterize our own Labels. In Android XML, is utilized for making our designs since it is a lightweight language which doesn't make our format hefty.

## **Libraries and Frameworks:**

### OpenCV:

Open CV (Open Source PC Vision) is a free illustrations library by and large focused at continuous PC vision. Open CV was created by Intel to handle pictures. OpenCV was bound for picture handling as it is pre-stacked with numerous capacities and numerous calculations. As OpenCV is pre-stacked with these capacities and calculations it assists scientists with tackling vision-related issues. OpenCV gives applications that will aid the preparation of course classifiers. We have utilized the openCV for fostering our application.

### NodeJS:

Node.js is a worker side stage based on Google Chrome's JavaScript Motor (V8 Motor). Node.js is an open source, cross-stage runtime climate for creating worker side and systems administration applications. Node.js applications are written in JavaScript, and can be run inside the Node.js runtime on operating system X, Microsoft Windows, and Linux. Node.js additionally gives a rich library of different JavaScript modules which works on the advancement of web applications utilizing Node.js generally.

### **(3.2) Proposed Models**

Model of face acknowledgment which recommends the area and affirmation of the substitute appearances for implying their cooperation. The essential modules that are being used are :

1) Dataset Age structure preparing: This is the masterminded head where the face dataset of the customer is made, further in which 15-20 photos of each customer is taken and the features used are customer username and ID.

2) Face Identification: LBP-based face finder is utilized for the face recognition.

3) Pre-handling: Pre-preparing includes certain means:

a) Resizing: After the face is identified by an indicator ,face is resized to a fixed pixel goal.

b) Trimming: Just face outline is chosen and foundation is taken out from the picture.

c) Grayscale Change: The hue picture is changed over into a grayscale picture.

4) Component Extraction and Acknowledgment: Estimations dependent on histogram standard is used for acknowledgment and feature extraction. The direct LBPH estimations are picked for the correct consistent arrangement of data as its computational diverse nature is less and is progressively amazing diverged from the other face acknowledgment computations.



5) Picture Securing: Pictures are put away in a neighborhood registry or organizer with username and roll no initials and the enrolled understudies are put away inside a MySQL information base for later questioning.

6) Face Discovery: Face recognition is being adjusted also as speed with the usage of Haar like-highlights with the relationship of the Viola-Jones ID system for objects. One of the key compels in past boosting-related methodologies are the fortitude to illuminating and insufficient impediment of the face. To change the constraints, we utilize Neighborhood Parallel Example (LBP) features for recognizing the face on our applications.

#### Different steps for face Detection:

Step 1: Load the LBP cascade

Step 2: Instantiate the OpenCV Cascade Classifier

Step 3: For each video frame received, call the cascade classifier

#### Explanation of the code:

1) It stacks the nearby OpenCV library for using Python Programming interface. An occasions of Course Classifier are made, transient it the forenames of the record from where the classifier can be stacked.

2) After the stacking it begins catching the countenances and assuming the appearances identified exist in the data set, it yes it begins composing the name and move number in the json arrangement to a book record.

3) This script runs for 5-10 minutes after which it exists by use of the set Timeout function provided in NodeJS.

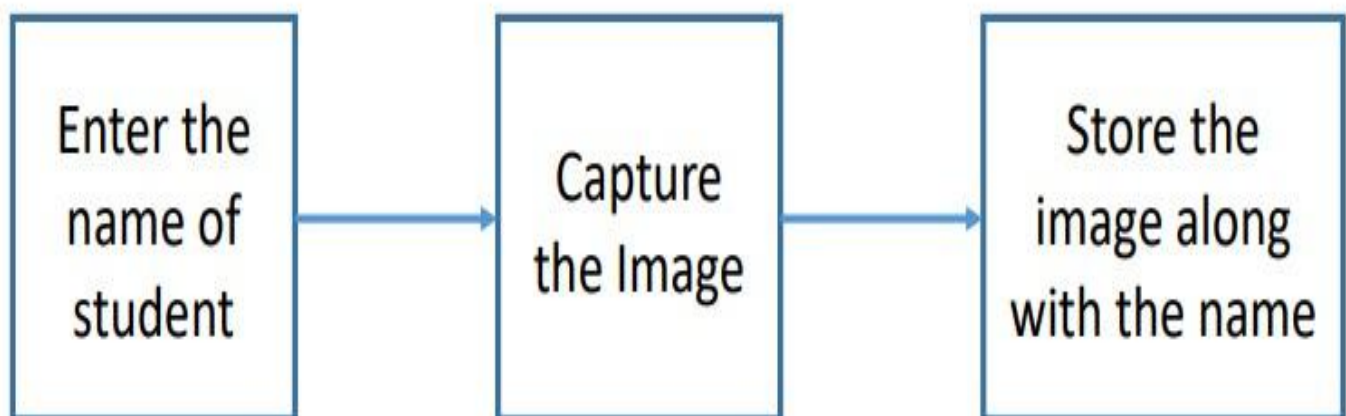
4) After this Hub Worker begins perusing that text record, assuming a similar client is available in excess of multiple times, it refreshes that individual's worth in the worth to be available else it proceeds to the following client. We have set the default worth to missing in the information base so if an individual couldn't be recognized he/she would be set apart as missing.

5) Finally attendance is reviewed and locked by the teacher.

## Chapter 4: Analysis Of Performance

### (4.1) Training activity

Nearby Paired Example estimations are used to set up the system in the method of snapping the photo of faces. XML records are made which obliges the data in regards to the face when you train the structure. This strategy is known as the pre-arranged classifier. Recognizable proof of the face, setting up the structure and treatment of the image are finished during the readiness activity of the applications.



**Fig. 4.1 Training Activity Process**

Executing face\_datasets.py will begin taking contribution for a specific understudy. Here it requests the client ID/move number and afterward his name. Subsequent to running it begins catching the picture with a webcam and partners the caught pictures with the understudy we enter. Next stage comes the preparation part.

As the preparation strategy begins it begins tossing the move number for which the webcam is catching at that case.

### Recognizing Face:

The script recognizes the faces in the frame with the confidence of 70% or more. Undetected faces are not written to the JSON file hence decreasing its size and improving the read time for the file.

When the training is done of the system, the testing of the system should be possible by the perceived movement. At whatever point the face is distinguished it is right off the bat changed into a vector image from 2D to 1D. The face identified is taken as a contribution for the Boolean capacity. Being a Boolean capacity it restores a genuine at whatever point a face gets a nearby match otherwise a false is returned. Moreover, when a nearby match is discovered the personality of the individual is removed by means of the perceived work.

### Drawbacks:

The application can't recognize and saw if the taken face isn't in the casings, as his/her face is defying downwards thus and we can't stamp his/her participation

On the off chance that a photo is pointed towards the webcam it additionally begins catching it which will bring about counterfeit intermediaries.

## **Chapter 5: Conclusion**

### **(5.1) Conclusion**

In this task, we read different calculations for the execution of face acknowledgment structure. Eigen faces and AI calculations were the fundamental need for setting up the framework subsequent to applying not many channels on the picture. Plus, the Eigen faces calculation grant the application to perceive the face continuously. Eigen faces were not sensitive towards change in the amount of subject during the primary stages, in any case, an expansion in the size of the preparation set assisted the calculation with rectifying its off-base expectations. An expansion in the informational index doesn't help in perceiving more subjects, still it transformed right expectations into wrong ones. Eigen faces were not exact in the subsequent stage. Fisher faces would be advised to bring about the main stage with a bigger amount of information or data. In any case, its activities were diverse in the resulting stage. In some cases with 25 pictures, the outcomes were nearly better, and now and then with 40 pictures the outcomes were comparative or more awful. This calculation was as a rule the unsuitable in the two phases. In addition in the second stage it had a low exactness. To distinguish a face the Neighborhood Paired plan course classifier is utilized. In the wake of testing the Haar course classifier, the speed of the discovery was very

low as diverged from LBP that consistently had practically 96% in the primary stage and is comparably better at computation. In second stage, this calculation had a continuous drop in the exactness or precision when contrasted with the primary stage. Rise or expansion in the amount of subjects forcefully changed its forecasts. In each stage, ascend in the preparation information had a positive effect or no effect.

### **(5.2) Disadvantages**

**Brightening Issue:** The enlightenment issue is that where a similar face seems diverse as a result of an adjustment of lighting. The movements provoked by enlightenment are consistently greater than the contrasts between individuals, causing structures dependent on contrasting pictures with misclassify input pictures.

**The Posture Issue in Face Acknowledgment:** It isn't amazing that the presentation of face acknowledgment frameworks drops inside and out when immense posture varieties are available in the info pictures. Exactly when light variety is likewise present, the endeavor of face acknowledgment turns out to be significantly more troublesome. Here we center around the out-of-plane revolution issue, since in-plane pivot is an unadulterated 2D issue and can be addressed substantially more without any problem.

**Detachment Issue:** Our face affirmation system crashes and burns if the division between the camera and the face outperforms numerous meters.

**Camera Issue:** Goal of a camera expects a huge occupation in seeing the image. Better the camera, even more definitely it will see the face. For example, we have attempted our face affirmation system with two webcams and we have found that the webcam with high goal camera is better than the other.

**Number of people:** Our face affirmation structure is obliged to only 7-8 people all at once. In case this number outperforms, our framework fails to see.

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