FACE RECOGNITION BASED ATTENDANCE SYSTEM USING PYTHON GUI

Submitted in partial fulfillment of the requirements for the award of Bachelor of Engineering degree in

INFORMATION TECHNOLOGY

By

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SATHYABAMA

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BONAFIDE CERTIFICATE

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ABSTRACT

FACE RECOGNITION BASED ATTENDANCE SYSTEM USING PYTHON GUI

Face recognition systems are part of facial image processing applications and theirsignificanceasaresearchareaareincreasingrecently. They use biometric information of the humans and are applicable easily instead of finger print, iris, signature etc., because these types of biometrics are not much suitable for non-collaborative people. Face recognition systems are usually applied and preferred for people and security camerasin metropolitan life. These systems can be

usedforcrimeprevention, videosurveillance, personverification, and similar security activiti es. We describe a face recognition-based automated attendance system utilizing a Python GUI in this work. This technique has a lot of applications in everyday life, notably in school and college. Scaling of the image size is conducted at the first phase, or pre-processing stage, to avoid or minimize information loss. HAAR CASCADE and XGBOOST are the algorithms involved. Overall, we created a Python programme that accepts an image from a database, performs all necessary conversions for picture identification, then confirms the image in video or real time using a user-friendly interface by accessing the camera. The name and time of the successful match is then recorded. Face detection and recognition are two of the most demanding computer vision applications and outcomes. This area has traditionally been a prominent focus of image analysis research because to its function as the primary identification strategy for human faces. It's both exciting, as well as biometrics, pattern recognition Also, teaching a machine to accomplish this is challenging. Face recognition is another one of the toughest problems in computer vision. Recognizing and detecting faces and computer vision, are all hot issues in the medical and research industries. There are several software programmes or technologies that have improved to the point where even blurry images may be reconstructed and analysed to understand more about a person's personality. Facial recognition technology is a framework or programme that analyses an image or video footage to recognise people's faces and authenticate their identification. The face is a one-of-a-kind reflection of a person's personality. Face recognition is a biometric approach that involves matching a real-time image with previously stored photographs of the same individual in a database to identify a person

Keywords:HAAR CASCADE, XGBOOST, Face Recognition, Face Detection, Attendance automation.

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LISTOFABBREVATION

ACRONYM	ABBREVIATION
RGB	Red-Green-Blue
YCbCr	Luminance-blue Difference Chroma-red Difference Chroma
HSV	Hue-Saturation-Value
YUV	Luminance-Blue Luminance Difference-Red Luminance Difference
FFNN	Feed Forward Neural Network
SOM	Self Organizing Map
NN	Neural Network
PCA	Principal Component Analysis
SVM	Support Vector Machines
BP	Back Propagation
DCT	Discrete Cosine Transform
RBNN	Radial Basis Neural Network
LDA	Linear Discriminant Analysis
ICA	Independent Component Analysis
НММ	Hidden Markov Model
PTZ	Pan/Tilt/Zoom
	_aplacian of Gaussian

CHAPTER 1

INTRODUCTION

1.1 GENERAL

Facerecognitionsystemisacompleximage-

processingprobleminrealworldapplications with complex effects of illumination, occlusion. and imaging conditionontheliveimages. It is a combination of face detection and recognition technique sin image analyzes. Detection application is used to find position of the faces in agiven image. Recognition algorithm is used to classify given images with knownstructured properties, which are used commonly in most of the computer visionapplications. These images have some known properties like; same resolution, incl uding same facial feature components, and similar eye alignment. These imageswill be refered as "standard image" in the further sections. Recognition applicationsuses standard images, and detection algorithms detect the faces and extract faceimages which include eyes, eyebrows, nose, and mouth. That makes the algorithmmore complicated than single detection or recognition algorithm. The first step forface recognition system is to acquire an image from a camera. Second step is facedetection from theacquired image.As a third step, face recognition thattakes theface images from output of detection part. Final step is person identity as a result of recognition part. An illustration of the steps for the face recognition system is giveninFigure1.

Acquiringimagestocomputerfromcameraandcomputationalmedium(environment)via framegrabberisthefirststepinfacerecognitionsystemapplications. The input image, in the form of digital data, is sent to face detectionalgorithm part of a software for extracting each face in the image. Many methods areavailable for detecting faces in the images in the literature [1 - 29]. The availablemethods could be classified into two main groups as; knowledge-based [1 - 15] andappearance-based [16 - 29] methods. Briefly, knowledge-based methods are derivedfrom human knowledge for features that makes a face. Appearance-based methodsare derived from

training and/or learning methods to find faces. The details about themethodswill besummarized in thenext chapter.

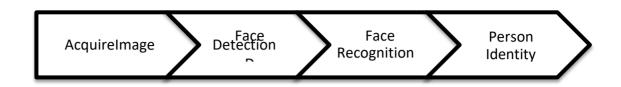


Figure 1.1: Steps of FaceRecognitionSystemApplications

After faces are detected, the faces should be recognized to identify the persons in thefaceimages.Intheliterature,mostofthemethodsusedimagesfromanavailableface library, which is made of standard images [30 - 47]. After faces are detected,standardimagesshouldbecreatedwithsomemethods.Whilethestandardimage sarecreated,thefacescouldbesenttorecognitionalgorithm.Intheliterature,methodscanbe dividedintotwogroupsas2Dand3Dbasedmethods.In2Dmethods,2Dimagesareusedasin putandsomelearning/trainingmethodsareusedtoclassifytheidentificationofpeople[30-43].In3Dmethods,thethree-dimensional data of face are used as an input for recognition. Different approachesare used for recognition, i.e. using corresponding point measure, average half face,and 3D geometric measure [44 - 47]. Details about the methods will be explained inthenextsection.

Methodsforfacedetectionandrecognitionsystemscanbeaffectedbypose, presence or of components, facial expression, absence structural occlusion, imageorientation, imaging conditions, and timedelay (for recognition). Available application by researchers can usually handle one or two S developed effects only, therefore they have limited capabilities with focus on some well-structured application. A robust face recognition system is difficult to develop which worksunderall conditions with awidescopeofeffect.

1.2. STATEMENT OF THE PROBLEM

Main problem of thesis is to design and implement a face recognition system in theRobot Vision Laboratory of the Department of Mechatronics Engineering at theAtılım University. The system should detect faces in live acquired images inside thelaboratory, and detected faces should be recognized. This system will be integrated to projects of guide robot, guard robot, office robot, etc. Later on, this thesis will be part of humanoid robot project.

1.3. SCOPE AND OUTLINE OF THESIS

Scope of the thesis is stated as follows:

- Facerecognitionsystemwilldetect,extractandrecognizefrontalfacesfromacquiredli veimages in laboratoryenvironment.
- Systemshouldworkunderchanginglightingconditionsinthelaboratory.
- Systemshouldrecognize50peopleatleast.
- Systemshouldnot extractfaces iftheyusesunglasses.
- Systemwillnotdetectprofileandnon-frontalfaceimages.
- Outlineofthesis is stated as follows:
- Chapter2introducesfacedetection,facerecognition,andfacerecognitionsystemap plications that exist inliterature.
- Chapter3describestheoryoffacerecognitionsystemthatisbasedonproblemstatem ent ofthesis.
- Chapter4summarizestheexperimentsperformedandtheirresultsusingtheface recognition system.
- Chapter5discussesandconcludesthesis.
- Chapter6givessomefutureworksonthethesistopic.

CHAPTER 2

LITERATURE SURVEY

Although face recognition systems are known for decades, there are many activeresearchwork on thetopic. Thesubjectcan bedivided into three parts;

- 1. Detection
- 2. Recognition
- 3. Detection&Recognition

Face detection is the first step of face recognition system. Output of the detection canbe location of face region as a whole, and location of face region with facial features (i.e.eyes, mouth, eyebrow, nose etc.). Detection methods in the literature are difficult to classify strictly, because most of the algorithms are combination of methods for detecting faces to increase the accuracy. Mainly, detection can be classified into two groups as Knowledge-Based Methods and Image-Based Methods. The methods for detection are given in Figure 2.

Knowledge-Based methods use information about Facial Features, Skin Color orTemplate Matching. Facial Features are used to find eyes, mouth, nose or other facialfeatures to detect the human faces. Skin color is different from other colors andunique, and its characteristics do not change with respect to changes in pose andocclusion. Skin color is modeled in each color spaces like RGB (Red-Green-Blue),YCbCr (Luminance-Blue Difference Chroma-Red Difference Chroma), HSV (Hue-Saturation-Value), YUV (Luminance-Blue Luminance Difference-Red LuminanceDifference), and in statistical models. Face has a unique pattern to differentiate fromotherobjects andhencea templatecanbegeneratedto scanand detectfaces.

Facialfeaturesareimportantinformationforhumanfacesandstandardimagescanbe generated using these information. In literature, many detection algorithms basedonfacial features areavailable[1 -6]. Zhi-fang et al. [1] detect faces and facial features by extraction of skin like regionwithYCbCrcolorspaceandedgesaredetectedintheskinlikeregion.Then,eyesar e found with Principal Component Analysis (PCA) on the edged region. Finally,Mouth is found based on geometrical information. Another approach extracts skinlikeregionwithNormalizedRGBcolorspaceandfaceisverifiedbytemplatematching.

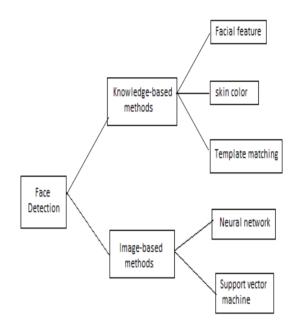


Figure 1.2: Methods for FaceDetection

Tofindeyes, eyebrows and mouth, colors nakes are applied to verified face image [2]. Ruan and Yin [3] segment skin regions in YCbCr color space and faces are verified with Linear Support Vector Machine (SVM). For final verification of face, eyes and mouth are found with the information of Cb and Cr difference. Foreye region Cb value is greater than Cr value and for mouth region Cr value is greater than Cb value. Another application segments skin like regions with statistical model.

StatisticalmodelismadefromskincolorvaluesinCbandCrchannelinYCbCr color space. Then, face candidates are chosen with respect to rectangular ratio ofsegmented region. Finally, the candidates are verified with eye & mouth map [4].Also, RGB color space can be used to segment skin like region and skin color likeregionisextractedtobe facecandidate.Candidate isverifiedby findingfacial features. Eyes and mouth are found based on isosceles triangle property. Two eyesand one mouth create an isosceles triangle and also distance between two eyes and distance from mid-point of eyes to mouth are equal. After eyes and mouth is found, FeedForward Neural Network (FFNN) is used for final verification of face candidate[5]. Bebar et al. [6] segment with YCbCr color space & mouth and eyes are found onthecombination of segmented image and edged image. For final verification, horizontal and vertical profiles of the images are used to verify the position of theeyes and mouth. All the methods are using firstly skin segmentation to eliminatenonfaceobjects in theimages to savecomputationaltime.

Skin color is one of the most significant features of human face. Skin color can bemodeled with parameterized or non parameterized methods. Skin color region can beidentified in terms of threshold region, elliptical modeling, statistical modeling (i.e.Gaussian Modeling), or Neural Network. Skin color is described in all color spaceslike RGB, YCbCr, and HSV. RGB is sensitive to light changes but YCbCr and HSVare not sensitive to light changes. The reason is that these two color spaces haveseparate intensity and color channel. In literature many algorithms based on skincolor available [7 - 13]. Kherchaoui and Houacine [7] modeled skin color usingGaussian Distribution Model with Cb and Cr channel in YCbCr color space. Thenskinlikeregionischosenasafacecandidatewithrespecttotheboundingboxratioof the and candidates are verified with template matching. region Another methodpreprocesses the given image to remove background part as a first step. It is done by applying edge detection on the Y component of YCbCr color space. Then, the closedregion is filled to take it as foreground part. After that, skin segmentation is done onYCrCb color space with conditions. The segmented parts are taken as candidate

andverificationisdonebycalculatingtheentropyofthecandidateimageandusethresholdin g to verify face candidate [8]. Qiang-rong and Hua-lan [9] applied whitebalancecorrectionbeforedetectingfaces.Thecolorvalueisimportantforsegmentatio n,sowhileacquiringtheimagecolorsmayreflectfalsecolor.Toovercomethis,whitebalancec orrectionshouldbedoneasafirststep.

Then Skin, color like regions are segmented using elliptical model in YCbCr. After skin

regionsarefound,theycombinedwithedgedimagestograyscaleimage.Finally,thecomb ined regions are verified as face by checking bounding box ratio and area insidethebounding

box.Anotherapplicationsegmentsskinlikeregionwiththresholdvalue in Cb, Cr, Normalized r and Normalized g. Then candidate for face is chosenwith respect to bounding box ratio, ratio of area inside and area bounding box, andminimum area of the region. After candidates are found, then AdaBoosting method isapplied to find face candidates. The verification is done with combining both resultsfrom skin like region and AdaBoosting [10]. Also, skin color can be modeled inelliptical region in Cb and Cr channel in YCbCr color space. Skin like region issegmentedifthecolorvalueisinsideellipticregionandcandidateregionsareverified using template matching [11]. Peer et al. [12] detect faces using only skinsegmentationinYCbCrcolorspaceandresearchersgeneratetheskincolorcondition s in RGB color space as well. Another approach for skin color modeling isdone by Self Organizing Map (SOM) Neural Network (NN). After skin segmentationis applied, each segment is taken as candidate and verified if they can fit into ellipticregion ornot [13].

Another significant information about the human face detection is pattern of humanface.Templatematchingcanbeappliedoverwindowscanningtechniqueorsegm ented region. Scanning technique is applied with small size window like 20x20or 30x30 pixel window. This approach scans all over the original image, and thendecreases the image size with some iteration suitable to re-scanning. Decreasing thesize is important to locate the large or medium size faces. However,

requiresexcessivecomputationaltimetolocatefaces.Templatematchinginsegmentedr egionregiuresmuchlesscomputationaltimethanscanning, because itonly considers the matchingofsegmentedpart.Inliteraturemanyapplicationsareavailable by using template matching, i.e., [14], [15]. Chen et al. [14], use half-facetemplate, instead of full-face template. This method decreases computational time. And this halffacecanbeadoptedtofaceorientations. Anotherapproachuses abstract templates, that are notimage like but composed of some parameters (i.e.size, shape, color, and position). Skin like regions are segmented with respect to YCbCr color space. templates Then, eve and eye pair abstract are applied to thesegmented region. First templates locate the region of eyes and second templates locate the each eye. Second template also determines the orientation of eyes. ThenTexturetemplateis applied toverifythefacecandidateregion [15].

Image-Based methods use training/learning methods to make comparison betweenface and non-face images. For these methods, large number of images of face andnon-face should be trained to increase the accuracy of the system. AdaBoost [16],EigenFace [17 - 19], Neural Networks [20 - 25] and Support Vector Machines [26 -29] are kind of methods that are used commonly in face detection algorithms. Faceand non-face images are described in terms of wavelet feature in AdaBoost method.Principal Component Analysis (PCA) is used to generate the feature vector of faceand non-face image in EigenFace method. Also, PCA is used giveninformationvector.Kernelfunctioniscreatedtodescribefaceandnon-

faceimagesin Support Vector Machines (SVM). Face and non-face images are also classified byartificialneuron structureinNeural Networks (NN).

AdaBoost is an algorithm that constructs strong classifier from weak classifiers. FacecandidatesarefoundbyapplyingAdaBoostalgorithm.Then,verificationisdonewith Cascade Classifier. This algorithm can handle of faces; left, left+45,front, andright+45,right pose.

Some performance statistics of algorithms are given in Appendix section.Detection/Recognitionratesshowtheperformanceofcorrectdetectionoffacesi nthe given image or recognition of given face image. Miss rate shows percentage ofmissed face in the given image. False rate gives the percentage of wrong detectedfaceorwrong classified face.

Methodsforfacerecognitionsystemareinvestigatedandpossiblesolutionsarestudied extensively. Selected face detection method is skin segmentation and facecandidateisverifiedwitheyesandmouthfinding.Then,extractedfacesareclassified with FFNN. The detail about the face recognition system will be explained innext chapter.

CHAPTER3

DESIGN OF A FACE RECOGNITION SYSTEM

Research papers on Face recognition systems are studied and state of the currenttechnology are reviewed and summarized in the previous chapter, results of whichwillguideustodesignafacerecognitionsystemforfuturehumanoidand/orguide/gu ardrobot.Athroughoutsurveyhasrevealedthatvariousmethodsandcombinationofthes emethodscanbeappliedindevelopmentofanewfacerecognition system. Among the many possible approaches, we have decided to use acombination of knowledge-based methods for face detection part and neural networkapproach for face recognition part. The main reason in this selection is their smoothapplicability and reliability issues. Our face recognition system approach is given inFigure4.

3.1. INPUT PART:

Input part is prerequisite for face recognition system. Image acquisition operation isperformedinthispart.Livecapturedimagesareconvertedtodigitaldataforperforming image-processing computations. These captured images are sent to facedetectionalgorithm.

3.2. FACE DETECTION PART:

Facedetectionperformslocatingandextractingfaceimageoperationsforfacerecognitionsystem.FacedetectionpartalgorithmisgiveninFigure5.Our experiments reveal that skin segmentation, as a first step for facedetection,reducescomputationaltimeforsearchingwholeimage.Whilesegmentationisapplied, only segmented region is searched weather the segment includes any faceornot.

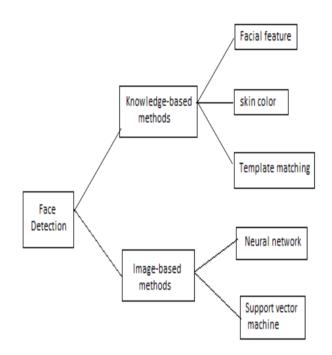


Figure 3.1: AlgorithmofFaceDetectionPart

For this reason, skin segmentation is applied as a first step of detection part. RGBcolor space is used to describe skin like color [12], and also other color spaces

areexaminedforskinlikecolors, i.e. HSV&YCbCr[54], HSV[55], and RGB&YCbCr&HSV[56]. However, bestresults give RGB colorspaces kin segmentation. The results of skin segmentation on different color spaces are given inthenext chapter. Skincolorlike pixel conditions are given below [12]:

r>95	r-g >15
g>40	r>g
b>20	r>b

max(r,g,b)-min(r,g,b)>15

"r", "g", and "b" parameters are red, green and blue channel values of pixel. If theseseven conditions are satisfied, then pixel is said to be skin color and binary image iscreatedfrom satisfied pixels.

Whitebalanceofimagesdiffersduetochangeinlightingconditionsoftheenvironmentwhil eacquiringimage. Thissituationcreates non-skinobjects that belong to skin objects. Therefore, white balance of the acquired image should becorrected before segmenting it. The implemented white balance algorithm is given below [57]:

- Calculateaveragevalueofredchannel(R_{av}),greenchannel(G_{av}),andbluec hannel(B_{av})
- CalculateaveragegrayGray_{av}=(R_{av}+G_{av}+B_{av})/3
- Then,K_R=Gray_{av}/R_{av},K_G=Gray_{av}/G_{av},andK_B=Gray_{av}/B_{av}
- Generate new image (Newl) from original image (Orjl) byNew(R)=K_R*Orj(R),New(G)=K_G*Orj(G),andNew(B)=K_B*Orj(B)

White balance algorithm, as a brief, makes image hotter if image is cold, and makescolder if image is hot. If image appears as blue, then image is called as cold. If imageappears as red or orange, then image is called as hot. Lighting conditions in

thecaptureareaarealwayschanging,duetochangeinsunlightdirection,indoorlighting, and other light reflections. Generally, taken pictures are hotter than theyshould be. Figure 6 shows hotter image that is taken in capture area and skin colorsegmentationto hotterimage, and whitebalancecorrectedimage.

If the image is not balanced, then some part of the wall will be taken as skin color asin Figure 6. Under some lighting conditions, acquired image can be colder. Then, the colder image will be balanced to hotter image.

On the contrary, this process will generate unwanted skin color like regions. To getrid of this problem and create final skin image, logical "and operation" is applied onboth segmented originalimage and white balance corrected. This operation willeliminate change of color value due to change of lighting condition. Also, bad results of segmentation on uncorrected image and good results on corrected image

are givenin Figure 7. In uncorrected image, distinguishing of face part is hard and face partseemsto bepart ofbackground.



a.)OriginalImage (OI)b.)Skin Segmentation on OI



n Ol

C)WhiteBalanceCorrectiononOl(WBI) d.) Skin Segmentation on WBIFigure4

FIGURE 3.2:

Exampleoftaken/whitebalancecorrectedimageandskincoloursegmentation

After"andoperation"isappliedonsegmentedimages, some morphological operations are applied on final skin image to search face candidate. Noisy like smallregions, that are less than 100 pixel square area. are eliminated. Then, morphological closing operation is applied to merge gaps with 3-by-3 square structure. Applyingdilation operation and then applying erosion operation are considered as closingoperation. After these two morphological operations, face candidate regions can bedetermined. To select candidate, each 1's are labeled. On each label two conditions are concerned to be face candidate. First condition is ratio of bounding box, which covers the label. The ratio of bounding box, width over height, should lie between 0.3 and 1.5. The limits determined experimentally. Lower limit is taken to be as low as possible, to get facial part that include neck or some part of chest. Other condition is to cover some gaps inside the region. This property will distinguish face from oth er body part, i.e. hand. Segmentation on hand will have no gap which maked ifferent from face.



Figure 3.3:Results of Segmentation on Uncorrected (Left) and Corrected Image(Right)

Based on these conditions, face candidates are extracted frominput image withmodifiedboundingboxfromoriginalboundingbox.Asitismentioned, withlowering neck could be boundina box limit, chest or included the the in candidate.thoughchestorneckshouldbediscarded.Theheightofboundingboxmodified as1.28 times bigger than width of bounding box because width of face candidate doesnotchange if candidateincludeschest/neckornot.Thismodificationvalue havebeen determined experimentally. After this modification, new bounding box coversonly face. These face candidates will be sent to facial feature extraction part tovalidatethecandidates.

Face candidates are found after white balance correction, skin like color detection,morphologicaloperation,andfacecandidatesearching.Forfinalverification of candidateandfaceimageextraction,facialfeatureextractionisapplied.Facialfeature is one of the most significant features of face. Facial features are eyebrows,eyes, mouth, nose, nose tip, cheek, etc. If some of the features can be found in

thecandidate,thenthecandidatewillbeconsideredasface.Twoeyesandmouthgenerat e isosceles triangle, and distance between eye to eye and mid point of eyesdistance to mouth is equal [5]. On the other hand, candidate facial feature should beextracted from face candidate image, because it is difficult to determine the features.Some filtering operations are applied to extract feature candidates and steps are listed below:

- □ LaplacianofGaussianFilteronRedchannelofcandidate
- □ Contrastcorrectionto improvevisibilityoffilterresult
- □ Averagefilteringtoeliminatesmallnoises
- □ Convertingto binaryimage
- □ Removingsmall noisesfrom binaryimages
- □ Instead of Laplacian of Gaussian (LoG) Filter, binary thresholding is applied inprevious application. Binary thresholding is sensitive to lighting. If shadow appearson some part of the face, some facial feature components can be eliminated. In sometrials, left eye has eliminated due to shadowing on left part of face. Also, beard onface can eliminate mouth while thresholding. Due to these problems, Sobel edgedetection method is tried to eliminate thresholding problem. Edge detection canreveal facial components better than thresholding and not sensitive to light change orshadows. On the other hand, edge detection gives higher response than LoG, and eyepart is not clear as LoG result. Due to this reasons, LoG filter is used to extract facialfeatures.
- Result of some filtering operations on face candidate and face candidate is given inFigure8.



a) Face Candidate Imageb.) Face Image After Filtering
 Figure3.4: Result offiltering
 operationsonfacecandidate

Figure 8 shows that, facial features can be selected easily. Eyes, mouth line can beselected and with some operations, it may be feasible for computers as well. Afterobtaining filtered image, labeling operation is applied to determine which labels arepossible to be facial features. Then, filtered image is divided into three regions whichisillustrated in Figure9.

In Figure 9, R denotes right region, L denotes left region, and D denotes down regionofface.

Criteriacheckingareappliedoneachlabeltodetermineleftand righteyes.

Criteriaarelistedbelow:

- 1. Width denotes width of face candidate image and height denotes height offace candidateimage
- 2. ypositionofleft/righteyeshouldbelessthan0.5*height
- 3. xpositionofrighteyeshouldbeinregion of 0.125*widthto0.405*width
- 4. xpositionoflefteyeshouldbeinregion of 0.585* width to 0.875* width
- 5. Areashouldbegreaterthan100 pixelsquare
- 6. BoundingBoxratiooflabel shouldbein theregion of 1.2 to 4

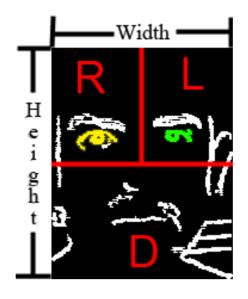


Figure 3.5: Regions of Filtered Image

Ifalabelisin regionRand satisfiescriteria2,3,5 and 6,then itissaid to becandidate to right eye. Yellow label is a right eye candidate (Figure 9). If a label is inregion L and satisfies 2,4,5 and 6, then it is said to be candidate to left eye. Greenlabel is a left eye candidate (Figure 9). A right eye candidate is said to be right eye ifits distance to center point of image is minimum in all right eye candidates. Also, aleft eye candidate is said to be left eye if its

distance to center point of image isminimum in all left eye candidates. Left and right eye are mostly found correctly butsometimes bottom eyelid is found falsely. If left and right eyes are detected, thenmouthfindingapplication can beapplied.

Each label inside down region chooses as mouth candidate and candidate propertyvectoriscalculated.Euclidiandistanceofrighteyetomouthcandidate(right-

distance) and Euclidian distance of left eye to mouth candidate (left-distance) arecalculated. Also, Euclidian distance between two eyes (eye-distance) and Euclidiandistancebetweenmidpointofeyestomouthcandidate(center-

distance)arecalculated.Then, propertyvectoris created byusingthedistances. Propertyvector:

- □ Labelnumberofthemouth candidate
- □ Absolutedifferencebetweenleft-distanceandright-distance(error1)
- □ Absolutedifferencebetweeneye-distanceandcenter-distance(error2)
- □ Summationoferror1and error2(error-sum)

If error1 and error2 are smaller than 0.25*eye-distance, then candidate is possibly

amouth. Minimum error-sum inside possible mouths is considered as mouth. Requiredfacial features are found which are right eye, left eye and mouth. Face image can

beextracted which coverst woeyes and mouth. Face covering is created with a rectangle in which corner positions are;

- □ Rightupcorner:0.3*eye-distanceupandleftfromrighteyelabelcentroid,
- □ Leftupcorner:0.3*eye-distanceupandrightfromlefteyelabelcentroid,
- □ Right down corner: 0.3*eye-distance from left from right eye label
- □ centroidanddown from mouth label centroid,
- □ Left downcorner: 0.3*eye-distancefromright from lefteye labelcentroidanddown from mouth label centroid,

After face cover corner points are calculated, face image can be extracted. Facialfeatureextraction, covering and face image extraction are given in Figure 10.



Figure 3.6: Facial Feature Extractions (Left) and FaceImage (Right) for the author

Up to here, face detection partis completed, and face images are found in theacquired images. This algorithm is implemented using MATLAB and tested for morethanhundredimages. This algorithm detects not only one face but also more than one face. Small amount of oriented face are acceptable. Results are satisfactory for all purpose.

3.3 FACERECOGNITIONPART

Modified face image which is obtained in the Face recognition system, should to beclassified to identify the person in the database. This is face recognition part of aFace Recognition System. Face recognition part is composed of preprocessing faceimage, vectorizing image matrix, database generation, and then classification. The classification is achieved by using FeedForward Neural Network (FFNN) [39]. Facerecognition part algorithm is given.

Beforeclassifyingthefaceimage,itshouldbepreprocessed.Preprocessingoperations are histogram equalizing of grayscale face image, resizing to 30-by-30pixels, and finally vectorizing the matrix image. Histogram equalizing is used forcontrast adjustment. After histogram equalization is applied, input face image issimilar to faces in database. Input face image has a resolution about 110-by-130pixels which is large for computation of classifier. So, dimension reduction is madewithresizing imagesto30-by-30pixelsimagetoreducecomputationaltimeinclassification. After resizing, image matrix should be converted to vector becauseclassifier does not work with two-dimensional input. Input vector size will be 900-by-1vectorto classifier.

Neural Network is used to classify given images. Neural Network is a mathematicalmodelthatisinspiredfrombiologicalneuralnetworksystem.Neuralnetwor kconsists of neurons, weights, inputs and output. A simple neuron model is given inFigure 12. Inside neuron, summation (Σ) and activation function (f) operations areapplied. 'x' denotesinput of neuron, 'w' denotes weight of input, 'I' denotes output of summation operation, and 'y' denotes output of neuron or output of activationfunction. Equations of I and y is given in Eq.1 and Eq.2. Network structure may bemultilayered(Figure13).

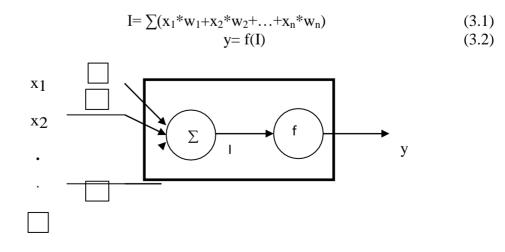


Figure 3.7: NeuronModel

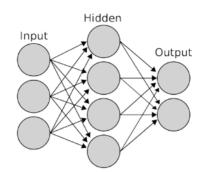


Figure 3.8: MultilayerNetworkstructure

Output value range shows a difference with respect to selected activation function.Common activation functions are threshold, linear and sigmoid functions, and showninFigure14.

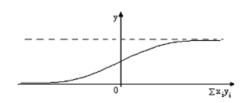


Figure 3.9: (Activation Functions – Sigmoid)

Also,many

differenttypeofnetworkstructuresexistinliterature.Inclassifier,FeedForward Neural Network (FFNN) is used. FFNN is the simplest structure in theneural network. Figure 13 is a kind of FFNN. Information flows through input tooutput and does not perform any loop or cycle operations. Two-layer with sigmoidtransfer function FeedForward Neural Network is used for classification operation. This type of network structure is generally used for pattern recognition applications.Systemnetwork properties are:inputlayer has 900 inputs, hidden layer has 41 neurons and output layer has 26 neurons. Output layer has 26 neuron since thenumberofpeoplein databaseis 26.

After structure is generated, then network should be trained to classify the givenimages with respect to face database. Therefore, face database is created before anytests. A database is created for 26 people with 4 samples for each

person. This results104 training sample. Due to that, 900-by-104 size matrix will be training matrix. Training matrix vector element is arranged with four groups due to the number of samples for each person. Though, first 26 vector element belongs to first samples of 26 people, and it continues. Training matrix's columns are made from preprocessing age and then vectorizing to face image which generate database. Then, target matrix is generated to inform network vectors belonging to persons. Target vector iscreated with putting '1' with respect to order number of the name in database and other elements '0' for target vector. Due to vector arrangement of training matrix, target matrix is combination of 4 horizontally concatenated identity matrix with sizeof26.

Aftertrainingmatrixandtargetmatrixiscreated, then training of NN can be performed.

Training means that configuring the values of the weight to make bestrelationship between training matrix and target matrix. When weights are configuredwell, classification of the given new face image will be classified correctly. Backpropagation is used to train the network. Back propagation has two phases, that arepropagation and weight update. In propagation phase, first flowing input throughoutput to see result. Then, going back output to input with calculating errors for eachweights. After that, weights are updated with respect to their error values. Trainingperformance andgoalerrorsaresetto1e-17 toclassifygivenimage correctly. When training is completed, network can be used to classify new faces that are fedfromfacedetectionpart. Thepersonnameselectionisbasedonoutputofthenetwork. The row, which has maximum value in output vector, is matched with orderofthepeoplein databaseand nameis printed.

3.4.OUTPUT PART

This part is final step of face recognition system. Person name is determined with respect to output of face recognition. Output vector of neural network is used to identify person name. The row number which has maximum value is match with same row number in the database name order.

3.5.CHAPTER SUMMARY AND DESCRIPTION

Face recognition system has four main steps, which are input, detection, recognition. and output Input performs image acquisition part, which converts live capturedimage to digital image data. Detection part composed of white balance

correction toacquired image, skin like region segmentation, facial feature extraction, and faceimage extraction. White balance correction is an important step to eliminate colorchages of acquired imade due to illumination conditions change. Skin like

regionsegmentationperformancecanbeimprovedwithintegratingwhitebalancecorrect ion before segmenting. Skin color like region segmentation decreases searchtime for possible face region since only segmented regions are considered as regionmay contain face. Facial feature extraction is important to extract face image whichwill be standard face image. LoG filter gives best results to extract facial featureswith respect to black and white convertion. Facial features are found with property oftwoeyes andamouth creates isosceles triangle.

Face image is extracted based on facial features positions, and image preprocessoperation is performed to eliminated illuminate changes and prepared to be input toclassifier. Classifier is a key point of recognition part. FFNN is good at patternrecognition problem. Decreasing gradient value of performance, increase accuracy

ofclassification.Outputofnetworkdeterminespersonidentity.Eachpersonhasidentity number and the row with a maximum value in the output vector matcheswithidentitynumber.Thismatchingestablishconnectionbetweenoutputofclas sificationand personnames.

This face recognition system algorithm will perform fast and accurate person nameidentification. Performance of skin segmentation is improved with white balancecorrection and facial feature extraction performance is improved with LoG filter withrespect to Lin's implementation [5]. Accuracy of classification is improved withdecreasinggradient valueofperformancevalue.

CHAPTER-4

EXPERIMENTS AND RESULTS

A complete hardware and software system is designed and implemented in the RobotVision Laboratory of the Department of Mechatronics Engineering at the AtılımUniversity. The ultimate goal of the larger project (umbrella project) is to develop ahumanoid robot with a narrower application like Guide robot, Guard robot, Officerobot, etc. The developed system has been tested for many live acquired images andresults are satisfactory for such a pioneering work in the department. Improvementsare required for better performance. System description and possible improvementsarediscussed in this chapter.

4.1.SYSTEM HARDWARE:

Systemhasthreemainhardwareparts.Theyarecomputer,framegrabber,andcamera. Computer is brain of system, which processes acquired image, analyzesimageanddeterminesthepersonsname.Thecomputerusedinthetestisatypic alPCwith thefollowingspecifications:

- IntelCore2Duo3.0GHz
- 3.0GbRAM
- OnBoardGraphicCard

Sony EVI-D100P camera is used in Face recognition system (Figure 15). Camera hashigh quality CCD sensor with remote Pan/Tilt/Zoom (PTZ) operations. Camera has10x optic and 4x digital zoom, so totally 40x zoom capability. Optically, it has 3.1mm wide angle to 31 mm tele angle focal length and 1.8 to 2.9 minimum aperturecapacities. Resolution of CCD sensor is 752x582 pixel. Pan/Tilt capacity is $\pm 100^{\circ}$ forpan and $\pm 25^{\circ}$ for tilt operations. Camera has RS232 serial communication. Camerasettings and PTZ operation can be performed via serial communication. Camerasettings are shutter time, aperture value, white balance selection, etc. Camera videooutputsignalsareS-VideoandCompositesignals.Compositevideosignalisused.



Figure-4.1: SonyEVI-D100p

Image acquisition from camera is performed by frame grabber. Imagenation PXC200A frame grabber from CyberOptics is used (Figure 16). This grabber is connectedtocomputerviaPCIconnection.Upto4compositeand1s-

videosignalstypecamera can be connected. YCbCr, RGB and Monochrome color channel can beselected, and in our system RGB color channel is used. Resolution

of grabber is up to768x576 pixel.



Figure-4.2: PXC200AFrameGrabber

4.2.SYSTEM SOFTWARE:

Algorithm of system is implemented on MATLAB R2011a software. MATLAB is aproductionofMathWorksCo.andcanperformalgorithmdevelopment,datavisualizatio n,dataanalysis,andnumericcomputationwithtraditionalprogramming language i.e. C. Signal processing, image processing, controller design, mathematicalcomputation, etc. may be implemented easily with MATLAB that includes

manytoolboxeswhichsimplifiesgenerationofalgorithmmorepowerfully.ImageAcquisiti on Toolbox, Image Processing Toolbox, and Neural Network Toolbox areusedwhilegeneratingalgorithm ofFacerecognition system.

Image Acquisition Toolbox enables image acquiring from frame grabber or otherimaging system that MATLAB supports. This toolbox supports acquiring resolution of frame grabber, triggering specification, color space, number of acquired image attriggering, region of interest while acquiring, etc. This toolbox will bridge betweenframe grabberand MATLABenvironment.

ImageProcessingToolboxprovidesmanyreferencealgorithms,graphicaltools,analysis , etc. Reference algorithm provides fast development of algorithms.

Filters, transforms, enhancements, etc. are ready to use functions which simplify to codegeneration. This toolbox is used in face detection and some part of face recognitionsections.

NeuralNetworkToolboxprovidesdesigning,implementing,visualizing,andsimulating neuralnetworks.Patternrecognition,clustering,datafittingtoolsaresupported.Supervis edlearningnetworks,i.e.FeedForward,RadialBasis,TimeDelay,etc.,andunsupervised learningnetworks,i.e.SelfOrganizingMap,Competitive Layer, are also supported. For classifying of face images are performedbyPatternRecognitionToolwithtwolayer,sigmoid,FeedForwardNeuralNet work.

4.3FACE DETECTION:

First implementation of system is performed on detection of faces in acquired image.Therefore,facedetectionhasstartedwithskinlikeregionsegmentation.Manymet hods have been tried to select which segmentation algorithm works best on ourimage acquisition area. Based on RGB [12], HSV&YCbCr [54], HSV [55], andRGB&YCbCr&HSV[56]colorchannelsskinlikesegmentationaretestedonacquiredi magesandbestresultsaretakenfromRGBcolorspace.RGB&YCbCr&HSV are not performed well, based on our acquired images. Resultsofperformed skin likesegmentation aregiveninFigure17-19.



Figure4.3:OriginalImage (Left) & RGBSkinSegmentation(Right)



Figure 4.4: Skin Segmentation on Original Image with HS (Left) and CbCr Channel(Right)



Figure 4.5: SkinSegmentation on Original image with HCbCrCombinations

Besides RGB gives the best result, colors of wall inside laboratory can be skin like color due to white balance value of camera. Unwanted skin like color regions canaffectdetectionanddistortfaceshape. This color problem can be eliminated by white balance correction of acquired image. The implementation of white balance correction is given Figure 20. Wardrobe color is white in real (Figure 20). On theother hand, color in acquired image (left image) is cream and also wall color lookslikeskin color and it affects segmentation results. Figure 21 shows the results of

segmentation on acquired image and white balance corrected image. Results

showsthatwhitebalance correction should beapplied afterimageis acquired.



Figure 4.6: AnImageWithout(Left)and With (Right)WhiteBalanceCorrection

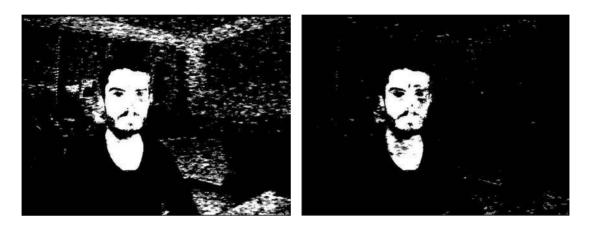


Figure 4.7: SkinSegmentationResultsonAcquired(Left) &Corrected Image (Right)

To guarantee color correction both segmentation are performed on acquired and corrected image, then logical "and operation" is performed. The reason is that correction n operation make hotter image if image is cold and make colder if image is hot. When image is hotter, color of objects in the laboratory becomes similar to skincolor, i.e. wall, floor, wardrobe. After segmentation is performed, morphological operations and candidate search is achieved as described in the previous chapter. Selection of face candidate followed by facial feature extraction and face verification of candidate.

Facial feature extraction is one of the most important parts of face detection sectionbecause this part makes bridging between detection and recognition. First trial

onextractionismadewithprofileextractionoffacecandidate.Verticalprofileofcandidate is performed by taking mean of each row of image matrix. Then, localminimum shows possible positions of eyebrow, eye, nose tip, mouth, and chin. Aftereyepositionisdeterminedinverticalprofile,thenhorizontalprofileisextractedto determine eyes positions. Vertical and horizontal profiles of four test face images aregiven in Figure 22 - 25. Determination of exact position of eye position and mouthposition is difficult to determine in vertical profile. Also, it is difficult to determinepositionofeyesinhorizontalprofileeventheveritcalpositionofeyesaredetermi nedin vertical profile.

Due to difficulty in determination files infacecandidate,faceprofileextractionisdiscardedandconvertingBlack-Whiteimage to find facial feature is performed. Some experiments are performed andresultsaregiven inFigure26 and 27.



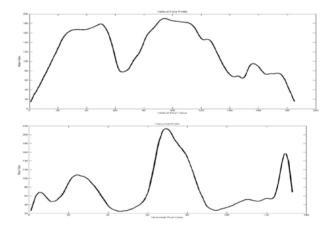


Figure 4.8: Test Image 1 (Left) & Vertical (Right-Top) - Horizontal (Right-Bottom)

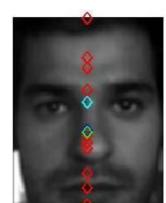
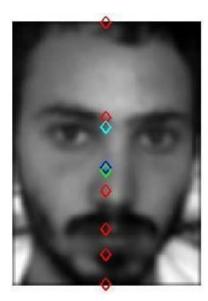




Figure 4.9: Test Image 2 (Left) & Vertical (Right-Top) - Horizontal (Right-Bottom)Profiles



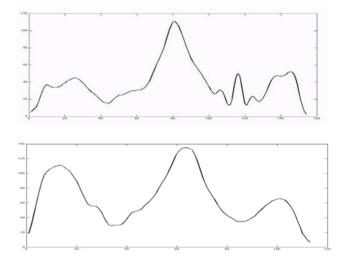
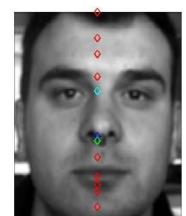


Figure 4.10: Test Image 3 (Left) & Vertical (Right-Top) - Horizontal (Right-Bottom)Profiles



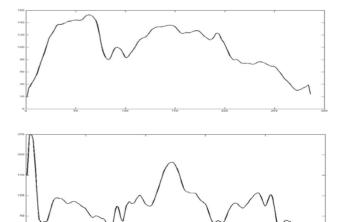


Figure 4.11: Test Image 4 (Left) & Vertical (Right-Top) - Vertical (Right-Bottom)Profile



Figure 4.12:TestImage5(Left)&Black-WhiteConversiononthe5(Right)



Figure4.13: TestImage6(Left)&Black-WhiteConversiononthe6(Right)

RighteyeisisolatedbutlefteyeiscombinedwitheyebrowinFigure26.Also,mouth is nearly erased. On the other hand, right eye and mouth are combined withbackgroundinFigure27.Thatmakesdifficulttofindeyeandmouth.Thecombination problem is due to lighting conditions while acquiring the image. Since,Black-White conversion is sensitive to light condition/changes, this approaches cannot be applicable easily. So, approaches that are not much sensitive to light shouldbeprefered.

Edge detection methods can be applicable on this problem because they are nearlyinsensitive to light change. Sobel edge detector is used to extract features. Figure 28showsresults of edge detection on test image 5 and 6.



Figure4.14: EdgeDetectionResultsonTestImage 5(Left)and6(Right)

Results show that, edge detection is not sensitive to light condition as Black-Whiteconversion. On both images, eyes and mouth can be selected with human eyes butmouth can be difficult to extract on the images and eye parts also vary on shapes.Also,edgedetection hashigh responses.

In order to use edge detection, Laplacian of Gaussian (LoG) filter can be used. LoGfilter has low responses than edge detection. Itmakes usefull enhancements on facialfeatures.Figure29shows resultsofLoGfilterontest image5and 6.



Figure 4.15 LaplacianofGaussianFilterResultsonTestImage5(Left)and6(Right)

Results of LoG filter are better than previous three trials. Mouth is more significant than others and eyes can be selected more accurately.

4.4FACE RECOGNITION:

With addition of isosceles triangle approach, as described in the previous chapter, eyesand mouth are found and cropped face image. Then, database of face image escan be generated. Database is generated from 26 people with 4 samples image for each person. Database is created from face detection part. Sample images of 26 people are given in Figure 30.





Figure 4.16:26ParticipantsFaceImagesFromDatabase.

Name of the Participant are: Natasha, Thor, Bruce, Tony, Steve, Clint, Lizzy, Nick, Clint, Tom, Peter, Harry, Nick, Ari, Jessy, Eric, Damon, Klaus, Stefan, Enzo, Mathew, Loki, Tessa, Mike, Dustin, William, Loki, Hari, Henry, Robert, Christopher, Scarlet. While generating database, four different sample images are stored for each persons.Thereasonisthatacquisitionoffaceimagemaydiffereachtimetheimagetaken. For example, shaved and no-shaved faces are included for my samples (Left image inFigure). Also, differentcaptured face framesare added (Rightimagein Figure).



Figure4.17DifferentFaceSamples

900x104 size training matrix is generated to train neural network which will be usedtoclassify

givennewfaceimage.PatternRecognitionToolinNeuralNetworkToolboxisusedtogen erateandtrainneuralnetwork.Thegeneratednetworkconsists of 2 layers with sigmoid transfer function. 2 layers are hidden layer andoutput layer. Output layer has 26 neurons due to number of persons in the facedatabase. Hidden layer number is an approach that is applied in [22]. The approachproposes that to guess initial neuron number, use Eq. 4.1, then train with this neuronnumber and record training time. After that, increase neuron number until trainingtime remain constant. When starting point of remaining constant, this will be numberofhidden neurons in thenetwork.

$n = \log_2 N$

Nisnumberofinputlayer, and nisthen umberof neurons inhidden layer. The initial quess is 9.81 for 900 inputs. So, start initially as 10. Figure 32 shows graph of number of neurons vs. training time. The graphshows that at 41 neurons trainingtime is 4 s and after this neuron number training time remain constant in 5 seconds. Therefore, 41 neurons are used in our system. Databasing and training networkcodeis of given in App 2.Also,performanceofclassificationisaffectedbytrainingparameterwhichisgradient value. Gradient value is related with error of target value and output value. Tests show that aminimum gradient valueresults in more accurate classification. The low gradient value causes false selection in the database. The comparison of gradient values for errors 1e-6 and 1e-17 for the same input image is given TestImage2 (Table1).

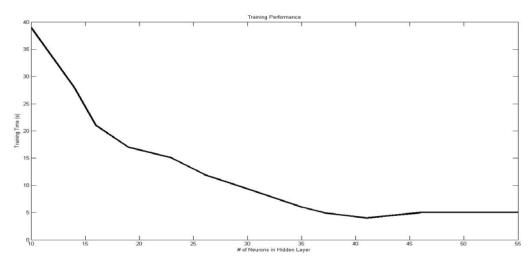


Figure4.18:NumberofNeurons vs.TrainingTime

Table1showsthatminimumgradientvalueshouldbeconsideredinsystemtogetmore accurateresults.

1e-6	1e-17
0.0076	0.0000
0.0041	0.0000
98.0832	99.99
	25
0.6032	0.0049
0.0040	0.0000

Table4.1GradientValueEffectsonClassification

0.0000
0.0000
0.0000
0.0000
0.0000
0.0000
0.0000
0.0000
0.0000
0.0000
0.0000
0.0000
0.0000
0.0000
0.1210
0.0000
0.0000
0.0000
0.0000
0.0000
0.0004

4.5FACE RECOGNITION SYSTEM:

Finally,facedetectionandrecognitionpartsaremergedtoimplementfacerecognition system. System can also handle more than one faces in the acquiredimage. Code is generated on MATLAB environment and given in App. 3 Results are shown Figure 33 - 43.

Fiveskinlikeregionsareextractedandlabeled.Label4&5istakenasfacecandidate.

Facial feature extraction operation is performed and eyes and mouth arefound, then faces are validated. Validated faces are classified, output results are; firstface belongs to Ayça and second face belongs to Cahit. Output result of system givescorrect results. Experiment and results are shown that algorithm can find multiplefaces in the acquired image and classify correctly. This results important, since some methods can only detecton eface in agiven image.

Fourskinlikeregionsarefoundinacquiredimage2.Thirdlabelisconsideredasface candidate. After LoG filter is applied, eyelashes appear clearly. Then,

algorithmconsiders eyelashes are eyes. Extracted face image classified correctly. Above resultsshow thatalgorithmcan recognizewheneyesareclosedwith 99.0498networkoutput.

Thistime, if personstands far from camera, can system detect and recognize correctly or not is tested. Five skin like regions are labeled. Then, label three is taken as face candidate. Taking height of face candidate as 1.28 times bigger than widtheliminates neck of person. Since, only face part is considered as face candidate. LoG filter performs well to extract facial feature regions. Due to low resolution, both left right, eye and eyebrow are merged but centroids of merged region do not affect the results. Low resolution of extracted face image is classified correctly. Above resolution of extracted face image can be recognized correctly even resolution is not large.

Face image is identified if maximum network output value is greater than 90 percent. Classification result for right image (Figure 40) is maximum 52.7422 because

thisfaceisnotinthefacedatabase.Therefore,algorithmsaysthat'Personisnotrecognize d'.

Since thisface isnotinthe face database, network resultismaximum 1.3872. Therefore, answerof algorithm is 'Person isnot recognized'.

Manyexperimentsareperformedonliveacquiredimages.Facedetectionandrecognitio n parts performed well. Skin segmentation both decrease computationaltime and search area for face. Experiments show that connection is established wellbetweendetectionandrecognitionparts.Thenetworkcancorrectlyclassifywhen eye/eyes are closed, eyebrows are moved and face is smiled or showed teeth. Also,number of people in database can be increased and most probably will correctlyclassifyfaces.

4.6LIMITATIONS OF THE SYSTEM:

Somelimitationsofthedesignedsystemaredeterminedafterthe experiments:

Skincolorlikeclothes: Skin color like clothes are segmented at skin

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segmentationstagebutitwillaffectresultsoffacecandidate.Mostofexperimentswithski ncolorlikeclothesshowthat,faceandclothsegmentsaremerged.Thesoftwaredoesnotr ecognizethemastwoseparatebodyregions.Thus,facialfeatureextractionoperation can be resulted, as candidate is not a face or wrong face image extraction.

Presenceofobjectonface:Glassesonthefacemayaffectfacialfeatureextractionresul ts.lfglasseshavereflection,LoGfiltercannotperformwelloneyeregion.Also,sunglasse swillcovereyeregionandeyescannotbedetectedbasedontheproposedalgorithm.

Contrast of face candidate image:Contrast value of image affects results of filter.Less contrast image has fewer edges on the image. Thus, facial component could notbe visible after filtered. Therefore, face image can not be extracted if candidatecontainsface.

*Systemworkingrange:*Systemcandetectandrecognizethepersonifpersonstandingr angeisbetween50cmto180cm.Thisrangeisacquiredwithpropertyof3.1 mm focal length and 768x576 pixels resolution. Thus, working range can bechangedbycameraproperty.

*Skin color range:*RGB skin color segmentation work well in the range of light tonetodark tone.

*Headpose:*Frontalheadposescanbedetectedandextractedcorrectly.Smallamountof roll andyawrotation canbeacceptableforsystem.

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CHAPTER 5 CONCLUSION AND DISCUSSION

5.1.DISCUSSION

This thesis study focuses on designing and implementing of a face recognition system.System is composed of image acquisition, face detection part, face recognition part, and identification of person name.

Image acquisition is prerequisite for system. Frame grabber is used to capture framefrom video camera device, and digitalize it to be processed in the algorithm. VideocameradeviceisSONYEVI-100Pwhichhascapabilityofpan,tiltandzoom.Therefore,livecapturedimageshav edifferentillumination,background,whitebalance difference, and position and size of human face. Image acquisition process isperformedbyMATLAB software withImageAcquisitionToolbox.

Knowledge-Based face detection method is performed for face detection part withcombination of skin color and facial feature methods. Methods are combined due todecreasing computational time and increasing accuracy of detection part. Before skincolor segmentation white balance correction is performed to overcome color problemwhileacquiringimage.Colorproblemaffectsskincolorsegmentation,bec auseirrelevantobjects can haveskin colorvalue.

RGB skin color segmentation is performed in the algorithm. Many methods useYCbCrcolorspace.Also,YCbCrandHSVcolorspacesaretestedbutbestresult sareobtainedwithRGBcolorspace.RGBcolorspaceworkswellatindoorcondition s but performance is not tested for outdoor condition. If modelling of skincolor is performed with statistical model, skin color segmentation may be moreaccurate. Segmentationis performed on bothacquired image andwhite balancecorrected in acquired image. Then, logical "and operation" is

performed on bothsegmentedimages to reducecolorproblem.

Face candidates are chosen from segments and facial feature extraction operation isperformed to verify face candidate and extract face image. LoG filter is performed

toshowfacialcomponentsclearly.BeforeLoGfilterisperformed,blackwhiteconversion and edge detection are performed. Black-white conversion is sensitive tolight changes and some components can be eliminated due to shadowing on face. On the other hand,edge detection is not sensitive to light changes but shapes are notclear as LoG filter. Facial components can be selected clearly with eyes after appliedLoG filter. Two eyes and mouth are found with property of two eyes and a mouthcreate isosceles triangle. Eyes and mouth are found with this property and face imageis extracted based on positions of facial components. On the other hand, componentsarefound by estimation but can befound more accurately.

Withextractionoffacialcomponents,facedetectionpartiscompletedandfaceima ge is ready to be classified. Before sending to classifier, histogram equalization,resizing and vectorizing operations are performed. Histogram equalization is appliedto eliminate light changes on image and equalizing contrast of image. Finally, faceimageisreadytobeclassified.ClassificationisperformedbytwolayersFeedF orward Neural Network. Sigmoid function is used for activation function inneurons. This type of network structure and activation function are good at patternrecognition problem since face recognition is a kind of pattern recognition

problem.Inthehiddenlayer,41neuronsareusedandtrainingtimevs.numberofneu rongraph is given in Figure 32. Best performance is achieved by 41 neurons. Outputlayerneuron numberis determined bynumberofpeoplein thedatabase.

Output of network gives classification result. The row with a maximum value givesorder number of names in database. Classification result is affected by performancevalue of network. The minimum gradient value gives more accurate result. Gradientvalue for system while training is taken as 1e-17. Performance with lower gradientvalueis given in Table1.

Algorithm is developed on MATLAB environment and it gives capability to detectmultiplefacesinacquiredimage.Personnamingisachievedwhenmaximu mvalueof row is greather than 90%. If it is lower, output is "Person is not recognized". Thesystemhasacceptableperformancetorecognizefaces withinintendedlimits.

5.2.CONCLUSION

Face recognition systems are part of facial image processing applications and theirsignificance as a research area are increasing recently. Implementations of system arecrimeprevention,videosurveillance,personverification,andsimilarsecuritya ctivities.Thefacerecognitionsystemimplementationwillbepartofhumanoidrobot project at Atılım University.

Main goal of thesis is to design and implement face recognition system in the RobotVisionLaboratory oftheDepartmentofMechatronicsEngineering.Thegoalisreached by face detection and recognition methods. Knowledge-Based face detectionmethods are used to find, locate and extract faces in acquired images.

Implemented methods are skincolor and facial features. Neural network is used for face recognition.

RGB color space is used to specify skin color values, and segmentation decreasessearching time of face images. Facial components on face candidates are appeared with implementation of LoG filter. LoG filter shows good performance on extracting facial components under different illumination conditions.

FFNNisperformedtoclassify

tosolvepatternrecognitionproblemsincefacerecognitionisakindofpatternrecognition.Classificationresultisaccurate.Classificationisalsoflexibleandcorrectwh

enextractedfaceimageissmalloriented,closedeye,andsmallsmiled.Proposedal gorithmiscapableof detectmultiplefaces, andperformanceofsystem hasacceptablegood results.

Proposedsystemcanbeaffectedbypose,presenceorabsenceofstructuralcomp onents,facialexpression, imagingcondition, and strongillumination.

CHAPTER 6

FUTURE WORKS

Face recognition system is designed, implemented and tested. Test results show thatsystemhasacceptableperformance.Ontheotherhand,systemhassomefuturewor ksforimprovementsand implementation onhumanoid robot project.

Future works will be stated in the order of algorithm. First future work can beapplied on camera device to improve imaging conditions. Sony camera that is used inthesis, can communicate with computer. Camera configurations can be changed viacomputer and these changescan improve imaging conditions. Exposure values canbe fixed to capture all frames with same brightness value / similar histogram. Also, fixing white balance value can improve performance of skin segmentation which willlead to eliminate non-skin objects. Maybe, white balance correction section notbe needed any later may more. For implementations, pan, tilt and zoom actuators can becontrolled.Cameraiscontrolled viaremotecontrollerinthetestofthesis.

Then skin color modelling can be improved. In the thesis work, some conditions areused to describe skin color. On the other hand, broad skin color modelling can beachieved by use of statistical modelling. Dark skins, skins under shadow or brightlight can be modelled and more skin region segmentation can be achieved. Skin colorsegmentation is an important step for algorithm of system. If more correct skinregions are segmented, more faces can be detected. Also, instead of RGB,

YCbCrskincolormodellingwithstatisticalmodelcanbeperformed,sinceCbandCrchan nelsvaluesarenot sensitiveto light changes.

On the other hand, some improvements can be applied on facial feature

extractionsection in face detection part. Computational volume is the biggest with respect toother sections in the algorithm. Computations of facial feature extraction can bereduced.Otherpointis thattocalculateeyeorientation,whichwillbeusedtoreorient face candidate and extract horizontally oriented face image. This operationwilldecreaseworkinglimitations ofdetection part.

Some improvements can be performed on recognition part. Firstly, number of peoplein the face database can be increased. However, increasing face images may createproblem in classification performance due to less number of sample image for eachpeople. Therefore, sample number could be increased.Later, input neuron numbersof neural network can be decreased with use of feature extraction method

inputfaceimage. This feature extraction method will decrease computational time of network. Possible extraction methods can be PCA, ICA, DCT or LDA. Also, iffeature extraction will be applied, face image database should be generated with feature extracted face images.

Later on, this system will be integrated to humanoid robot or narrower applications. Therefore this algorithm should be designed and implemented on embedded

systems.DigitalSignalProcessorsorFieldProgrammableGateArrayscanbeusedfore mbeddedsystems.Withuseofembeddedsystems,realtimefacerecognitionsystemcan beachieved.

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CHAPTER 7

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CHAPTER 8

SOURCE CODE

% MAIN MATLAB code for main.fig

% MAIN, by itself, creates a new MAIN or raises the existing

% singleton*.

%

% H = MAIN returns the handle to a new MAIN or the handle to

% the existing singleton*.

%

% MAIN('CALLBACK',hObject,eventData,handles,...) calls the local

% function named CALLBACK in MAIN.M with the given input arguments.

%

% MAIN('Property','Value',...) creates a new MAIN or raises the

% existing singleton*. Starting from the left, property value pairs are

54

% applied to the GUI before main_OpeningFcn gets called. An

% unrecognized property name or invalid value makes property application

% stop. All inputs are passed to main_OpeningFcn via varargin.

%

*See GUI Options on GUIDE's Tools menu. Choose "GUI allows only one

% instance to run (singleton)".

%

% See also: GUIDE, GUIDATA, GUIHANDLES

% Edit the above text to modify the response to help main

% Begin initialization code - DO NOT EDIT

gui_Singleton = 1;

gui_State = struct('gui_Name', mfilename, ...

'gui_Singleton', gui_Singleton, ...

'gui_OpeningFcn', @main_OpeningFcn, ...

'gui_OutputFcn', @main_OutputFcn, ...

'gui_LayoutFcn', [], ...

'gui_Callback', []);

if nargin && ischar(varargin{1})

```
gui_State.gui_Callback = str2func(varargin{1});
```

end

if nargout

[varargout{1:nargout}] = gui_mainfcn(gui_State, varargin{:});

else

gui_mainfcn(gui_State, varargin{:});

end

% End initialization code - DO NOT EDIT

% --- Executes just before main is made visible.

function main_OpeningFcn(hObject, eventdata, handles, varargin)

% This function has no output args, see OutputFcn.

% hObject handle to figure

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

% varargin command line arguments to main (see VARARGIN)

% Choose default command line output for main

handles.output = hObject;

% Update handles structure

guidata(hObject, handles);

% UIWAIT makes main wait for user response (see UIRESUME)

% uiwait(handles.figure1);

% --- Outputs from this function are returned to the command line.
function varargout = main_OutputFcn(hObject, eventdata, handles)
% varargout cell array for returning output args (see VARARGOUT);
% hObject handle to figure
% eventdata reserved - to be defined in a future version of MATLAB
% handles structure with handles and user data (see GUIDATA)

% standard size of image is 300 *300

global co

clc

warning off

st = version;

```
if str2double(st(1)) < 8
```

beep

hx = msgbox('PLEASE RUN IT ON MATLAB 2013 or Higher','INFO...!!!','warn','modal');

pause(3)

delete(hx)

close(gcf)

return

end

```
co = get(hObject,'color');
```

addpath(pwd,'database','codes')

```
if size(ls('database'),2) == 2
```

```
% delete('features.mat');
```

```
% delete('info.mat');
```

end

% Get default command line output from handles structure

varargout{1} = handles.output;

function edit1_Callback(hObject, eventdata, handles)

% hObject handle to edit1 (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

% Hints: get(hObject,'String') returns contents of edit1 as text

% str2double(get(hObject,'String')) returns contents of edit1 as a double

% --- Executes during object creation, after setting all properties.

function edit1_CreateFcn(hObject, eventdata, handles)

% hObject handle to edit1 (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles empty - handles not created until after all CreateFcns called

% Hint: edit controls usually have a white background on Windows.

% See ISPC and COMPUTER.

if ispc && isequal(get(hObject,'BackgroundColor'), get(0,'defaultUicontrolBackgroundColor'))

set(hObject,'BackgroundColor','white');

end

% --- Executes on button press in pushbutton1.

function pushbutton1_Callback(hObject, eventdata, handles)

% hObject handle to pushbutton1 (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

p = get(handles.edit1,'UserData');

if strcmp(p,'123') == 1

delete(hObject);

delete(handles.pushbutton2)

delete(handles.edit1);

delete(handles.text2);

delete(handles.text3);

delete(handles.text1);

delete(handles.text4);

msgbox('WHY DONT U READ HELP BEFORE

STARTING', 'HELP....!!!', 'help', 'modal')

set(handles.AD_NW_IMAGE,'enable','on')

set(handles.DE_LETE,'enable','on')

set(handles.TRAIN_ING,'enable','on')

set(handles.STA_RT,'enable','on')

set(handles.RESET_ALL,'enable','on')

set(handles.EXI_T,'enable','on')

set(handles.HE_LP,'enable','on')

set(handles.DATA_BASE,'enable','on')

set(handles.text5,'visible','on')

else

msgbox('INVALID PASSWORD FRIEND... XX','WARNING....!!!','warn','modal') end

% --- Executes on button press in pushbutton2.

function pushbutton2_Callback(hObject, eventdata, handles)

% hObject handle to pushbutton2 (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

close gcf

% -----

function AD_NW_IMAGE_Callback(hObject, eventdata, handles)

% hObject handle to AD_NW_IMAGE (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

% -----

function DE_LETE_Callback(hObject, eventdata, handles)

% hObject handle to DE_LETE (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

% -----

function TRAIN_ING_Callback(hObject, eventdata, handles)

% hObject handle to TRAIN_ING (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

% -----

function STA_RT_Callback(hObject, eventdata, handles)

% hObject handle to STA_RT (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

% -----

function DATA_BASE_Callback(hObject, eventdata, handles)

% hObject handle to DATA_BASE (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

% -----

function RESET_ALL_Callback(hObject, eventdata, handles)

% hObject handle to RESET_ALL (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

% -----

function EXI_T_Callback(hObject, eventdata, handles)

% hObject handle to EXI_T (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

% -----

function HE_LP_Callback(hObject, eventdata, handles)

% hObject handle to HE_LP (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

% -----

function READ_ME_Callback(hObject, eventdata, handles)

% hObject handle to READ_ME (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

winopen('help.pdf')

% -----

function PRE_CAP_Callback(hObject, eventdata, handles)

% hObject handle to PRE_CAP (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

```
if exist('features.mat','file') == 0
```

msgbox('FIRST TRAIN YOUR DATABASE','INFO...!!!','MODAL')

return

end

ff = dir('database');

if length(ff) == 2

h = waitbar(0,'Plz wait Matlab is scanning ur database...','name','SCANNING IS
IN PROGRESS');

for k = 1:100

```
waitbar(k/100)
```

pause(0.03)

end

close(h)

```
msgbox({'NO IMAGE FOUND IN DATABASE';'FIRST LOAD YOUR
DATABASE';'USE "ADD NEW IMAGE"
MENU'},'WARNING....!!!','WARN','MODAL')
```

return

end

```
fd = vision.CascadeObjectDetector();
```

```
[f,p] = uigetfile('*.jpg','PLEASE SELECT AN FACIAL IMAGE');
```

if f == 0

return

end

```
p1 = fullfile(p,f);
```

```
im = imread(p1);
```

bbox = step(fd, im);

vo = insertObjectAnnotation(im,'rectangle',bbox,'FACE');

```
r = size(bbox, 1);
```

if isempty(bbox)

```
axes(handles.axes1)
```

imshow(vo);

```
msgbox({'NO FACE IN THIS PIC';'PLEASE SELECT SINGLE FACE
IMAGE'},'WARNING...!!!','warn','modal')
```

uiwait

```
cla(handles.axes1); reset(handles.axes1);
```

```
set(handles.axes1,'box','on','xtick',[],'ytick',[])
```

return

elseif r > 1

```
axes(handles.axes1)
```

imshow(vo);

```
msgbox({'TOO MANY FACES IN THIS PIC';'PLEASE SELECT SINGLE FACE IMAGE'},'WARNING...!!!','warn','modal')
```

uiwait

```
cla(handles.axes1); reset(handles.axes1);
set(handles.axes1,'box','on','xtick',[],'ytick',[])
```

return

end

```
axes(handles.axes1)
```

image(vo);

```
set(handles.axes1,'xtick',[],'ytick',[],'box','on')
```

```
bx = questdlg({'CORRECT IMAGE IS SELECTED';'SELECT OPTION FOR FACE
EXTRACTION'},'SELECT AN OPTION','MANUALLY','AUTO','CC');
```

```
if strcmp(bx,'MANUALLY') == 1
```

while 1

fhx = figure(2);

set(fhx,'menubar','none','numbertitle','off','name','PREVIEW')

imc = imcrop(im);

```
bbox1 = step(fd, imc);
```

```
if size(bbox1,1) ~= 1
```

```
msgbox({'YOU HAVENT CROPED A FACE';'CROP AGAIN'},'BAD
ACTION','warn','modal')
```

uiwait

else

close gcf

break

end

close gcf

end

```
imc = imresize(imc,[300 300]);
```

image(imc)

text(20,20,'\bfUr Precaptured image.','fontsize',12,'color','y','fontname','comic sans ms')

```
set(handles.axes1,'xtick',[],'ytick',[],'box','on')
```

end

```
if strcmp(bx,'AUTO') == 1
```

```
imc = imcrop(im,[bbox(1)-50 bbox(2)-250 bbox(3)+100 bbox(4)+400]);
```

fhx = figure(2);

```
set(fhx,'menubar','none','numbertitle','off','name','PREVIEW')
```

imshow(imc)

qx = questdlg({'ARE YOU SATISFIED WITH THE RESULTS?';' ';'IF YES THEN
PROCEED';' ';'IF NOT BETTER DO MANUAL
CROPING'},'SELECT','PROCEED','MANUAL','CC');

```
if strcmpi(qx,'proceed') == 1
```

close gcf

```
imc = imresize(imc,[300 300]);
```

```
axes(handles.axes1)
```

image(imc)

```
text(20,20,'\bfUr Precaptured image.','fontsize',12,'color','y','fontname','comic sans ms')
```

```
set(handles.axes1,'xtick',[],'ytick',[],'box','on')
```

```
elseif strcmpi(qx,'manual') == 1
```

while 1

```
fhx = figure(2);
```

set(fhx,'menubar','none','numbertitle','off','name','PREVIEW')

```
imc = imcrop(im);
```

```
bbox1 = step(fd, imc);
```

if size(bbox1,1) ~= 1

msgbox({'YOU HAVENT CROPED A FACE';'CROP AGAIN'},'BAD ACTION','warn','modal')

uiwait

else

break

end

close gcf

end

close gcf

```
imc = imresize(imc,[300 300]);
```

```
axes(handles.axes1)
```

image(imc)

```
text(20,20,'\bfUr Precaptured image.','fontsize',12,'color','y','fontname','comic sans ms')
```

```
set(handles.axes1,'xtick',[],'ytick',[],'box','on')
```

else

end

end

```
immxx = getimage(handles.axes1);
```

```
zz = findsimilar(immxx);
```

zz = strtrim(zz);

```
fxz = imread(['database/' zz]);
```

```
q1= ehd(immxx,0.1);
```

q2 = ehd(fxz, 0.1);

```
q3 = pdist([q1 ; q2]);
```

disp(q3)

```
if q3 < 0.5
```

```
axes(handles.axes2)
```

image(fxz)

```
set(handles.axes1,'xtick',[],'ytick',[],'box','on')
```

text(20,20,'\bfUr Database Entered

```
Image.','fontsize',12,'color','y','fontname','comic sans ms')
```

```
set(handles.axes2,'xtick',[],'ytick',[],'box','on')
```

```
xs = load('info.mat');
```

xs1 = xs.z2;

```
for k = 1:length(xs1)
```

 $st = xs1\{k\};$

 $stx = st{1};$

```
if strcmp(stx,zz) == 1
```

```
str = st{2};
```

break

end

end

```
fid = fopen('attendence_sheet.txt','a');
```

fprintf(fid,'%s %s %s %s\r\n\n', 'Name','Date','Time', 'Attendence');

c = clock;

if c(4) > 12

```
s = [num2str(c(4)-12) ,':',num2str(c(5)), ':', num2str(round(c(6))) ];
```

else

```
s = [num2str(c(4)), ':', num2str(c(5)), ':', num2str(round(c(6)))];
```

end

fprintf(fid,'%s %s %s %s\r\n\n', str, date,s,'Present'); fclose(fid);

set(handles.text5,'string',['Hello ' str ' ,Your attendence has been Marked.'])

try

```
s = serial('com22');
```

fopen(s);

fwrite(s,'A');

pause(1)

fclose(s);

clear s

catch

```
msgbox({'PLZ CONNECT CABLE OR';'INVALID COM PORT
SELECTED'},'WARNING','WARN','MODAL')
```

uiwait

delete(s)

clear s

end

else

```
msgbox('YOU ARE NOT A VALID PERSON', 'WARNING','WARN','MODAL')
cla(handles.axes1)
reset(handles.axes1)
cla(handles.axes2)
reset(handles.axes2)
set(handles.axes1,'box','on','xcolor','w','ycolor','w','xtick',[],'ytick',[],'color',[0.0431
```

```
0.5176 0.7804],'linewidth',1.5);
```

```
set(handles.axes2,'box','on','xcolor','w','ycolor','w','xtick',[],'ytick',[],'color',[0.0431 0.5176 0.7804],'linewidth',1.5)
```

end

```
% -----
```

```
function LIVE_CAM_Callback(hObject, eventdata, handles)
```

```
% hObject handle to LIVE_CAM (see GCBO)
```

```
% eventdata reserved - to be defined in a future version of MATLAB
```

```
% handles structure with handles and user data (see GUIDATA)
```

global co

if exist('features.mat','file') == 0

msgbox('FIRST TRAIN YOUR DATABASE','INFO...!!!','MODAL')

return

end

```
ff = dir('database');
```

```
if length(ff) == 2
```

h = waitbar(0,'Plz wait Matlab is scanning ur database...','name','SCANNING IS
IN PROGRESS');

for k = 1:100

waitbar(k/100)

pause(0.03)

end

close(h)

msgbox({'NO IMAGE FOUND IN DATABASE';'FIRST LOAD YOUR DATABASE';'USE "ADD NEW IMAGE" MENU'},'WARNING....!!!','WARN','MODAL')

return

end

if isfield(handles,'vdx')

vid = handles.vdx;

stoppreview(vid)

delete(vid)

handles = rmfield(handles,'vdx');

```
guidata(hObject,handles)
```

cla(handles.axes1)

```
reset(handles.axes1)
```

set(handles.axes1,'box','on','xcolor','w','ycolor','w','xtick',[],'ytick',[],'color',[0.0431 0.5176 0.7804],'linewidth',1.5)

cla(handles.axes2)

```
reset(handles.axes2)
```

```
set(handles.axes2,'box','on','xcolor','w','ycolor','w','xtick',[],'ytick',[],'color',[0.0431 0.5176 0.7804],'linewidth',1.5)
```

end

```
info = imaqhwinfo('winvideo');
```

```
did = info.DeviceIDs;
```

if isempty(did)

```
msgbox({'YOUR SYSTEM DO NOT HAVE A WEBCAM';' ';'CONNECT A ONE'},'WARNING....!!!!','warn','modal')
```

return

end

```
fd = vision.CascadeObjectDetector();
```

did = cell2mat(did);

for k = 1:length(did)

devinfo = imaqhwinfo('winvideo',k);

na(1,k) = {devinfo.DeviceName};

sr(1,k) = {devinfo.SupportedFormats};

end

[a,b] = listdlg('promptstring', 'SELECT A WEB CAM

DEVICE','liststring',na,'ListSize', [125, 75],'SelectionMode','single');

```
if b == 0
```

return

```
end
```

```
if b ~= 0
```

```
frmt = sr{1,a};
```

[a1,b1] = listdlg('promptstring','SELECT RESOLUTION','liststring',frmt,'ListSize', [150, 100],'SelectionMode','single');

if b1 == 0

return

end

end

```
frmt = frmt{a1};
```

```
I = find(frmt == '_');
```

```
res = frmt(l+1 : end);
```

I = find(res == 'x');

```
res1 = str2double(res(1: I-1));
```

```
res2 = str2double(res(l+1 : end));
```

```
axes(handles.axes1)
```

```
vid = videoinput('winvideo', a);
```

```
vr = [res1 res2];
```

```
nbands = get(vid,'NumberofBands');
```

h2im = image(zeros([vr(2) vr(1) nbands], 'uint8'));

```
preview(vid,h2im);
```

```
handles.vdx = vid;
```

guidata(hObject,handles)

tx = msgbox('PLZ STAND IN FRONT OF CAMERA STILL','INFO.....!!!');

pause(1)

delete(tx)

kx = 0;

while 1

```
im = getframe(handles.axes1);
```

im = im.cdata;

bbox = step(fd, im);

vo = insertObjectAnnotation(im,'rectangle',bbox,'FACE');

axes(handles.axes2)

imshow(vo)

```
if size(bbox, 1) > 1
```

```
msgbox({'TOO MANY FACES IN FRAME';' ';'ONLY ONE FACE IS ACCEPTED'},'WARNING.....!!!','warn','modal')
```

uiwait

```
stoppreview(vid)
```

delete(vid)

```
handles = rmfield(handles,'vdx');
```

guidata(hObject,handles)

```
cla(handles.axes1)
```

```
reset(handles.axes1)
```

```
set(handles.axes1,'box','on','xtick',[],'ytick',[],'xcolor',[1 1 1],'ycolor',[1 1 1],'color',co,'linewidth',1.5)
```

cla(handles.axes2)

```
reset(handles.axes2)
```

```
set(handles.axes2,'box','on','xtick',[],'ytick',[],'xcolor',[1 1 1],'ycolor',[1 1 1],'color',co,'linewidth',1.5)
```

return

end

kx = kx + 1;

if kx > 10 && ~isempty(bbox)

break

end

```
end
```

```
imc = imcrop(im, [bbox(1)+3 bbox(2)-35 bbox(3)-10 bbox(4)+70]);
```

```
imx = imresize(imc,[300 300]);
```

```
axes(handles.axes1)
```

```
image(imx)
```

```
text(20,20,'\bfUr Current image.','fontsize',12,'color','y','fontname','comic sans ms')
```

```
set(handles.axes1,'xtick',[],'ytick',[],'box','on')
```

immxx = imx;

```
zz = findsimilar(immxx);
```

```
zz = strtrim(zz);
```

```
fxz = imread(['database/' zz]);
```

```
q1 = ehd(immxx, 0.1);
```

q2 = ehd(fxz, 0.1);

q3 = pdist([q1; q2]);

```
disp(q3)
```

```
if q3 < 0.5
```

axes(handles.axes2)

image(fxz)

set(handles.axes1,'xtick',[],'ytick',[],'box','on')

text(20,20,'\bfUr Database Entered

Image.','fontsize',12,'color','y','fontname','comic sans ms')

```
set(handles.axes2,'xtick',[],'ytick',[],'box','on')
```

```
xs = load('info.mat');
```

```
xs1 = xs.z2;
```

```
for k = 1:length(xs1)
```

 $st = xs1\{k\};$

 $stx = st\{1\};$

```
if strcmp(stx,zz) == 1
```

```
str = st{2};
```

break

end

end

```
fid = fopen('attendence_sheet.txt','a');
```

fprintf(fid,'%s %s %s %s\r\n\n', 'Name', 'Date', 'Time', 'Attendence'); c = clock;if c(4) > 12s = [num2str(c(4)-12), ':', num2str(c(5)), ':', num2str(round(c(6)))]; else s = [num2str(c(4)), ':', num2str(c(5)), ':', num2str(round(c(6)))]; end fprintf(fid,'%s %s %s %s\r\n\n', str, date,s,'Present'); fclose(fid); set(handles.text5,'string',['Hello ' str ' ,Your attendence has been Marked.']) try s = serial('com22'); fopen(s); fwrite(s,'A'); pause(1) fclose(s); clear s catch msgbox({'PLZ CONNECT CABLE OR';'INVALID COM PORT

SELECTED'},'WARNING','WARN','MODAL')

uiwait

delete(s)

clear s

end

else

msgbox('YOU ARE NOT A VALID PERSON', 'WARNING', 'WARN', 'MODAL')

cla(handles.axes1)

reset(handles.axes1)

cla(handles.axes2)

reset(handles.axes2)

set(handles.axes1,'box','on','xcolor','w','ycolor','w','xtick',[],'ytick',[],'color',[0.0431 0.5176 0.7804],'linewidth',1.5);

set(handles.axes2,'box','on','xcolor','w','ycolor','w','xtick',[],'ytick',[],'color',[0.0431 0.5176 0.7804],'linewidth',1.5)

end

```
% -----
```

function SINGL_PIC_Callback(hObject, eventdata, handles)

% hObject handle to SINGL_PIC (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

flist = dir('database');

if length(flist) == 2

msgbox('NOTHING TO DELETE','INFO','modal');

return

end

```
cd('database')
```

```
[f,p] = uigetfile('*.jpg','SELECT A PIC TO DELETE IT');
```

if f == 0

cd ..

return

end

p1 = fullfile(p,f);

delete(p1)

flist = dir(pwd);

if length(flist) == 2

cd ..

return

end

```
for k = 3:length(flist)
```

```
z = flist(k).name;
```

z(strfind(z, '.') : end) = [];

```
nlist(k-2) = str2double(z);
```

end

```
nlist = sort(nlist);
```

h = waitbar(0,'PLZ WAIT, WHILE MATLAB IS RENAMING','name','PROGRESS...');

for k = 1:length(nlist)

if k ~= nlist(k)

```
p = nlist(k);
```

```
movefile([num2str(p) '.jpg'] , [num2str(k) '.jpg'])
```

```
waitbar((k-2)/length(flist),h,sprintf('RENAMED %s to %s',[num2str(p)
'.jpg'],[num2str(k) '.jpg']))
```

pause(.5)

end

close(h)

cd ..

```
% ------
```

function MULTI_PIC_Callback(hObject, eventdata, handles)

```
% hObject handle to MULTI_PIC (see GCBO)
```

% eventdata reserved - to be defined in a future version of MATLAB

```
% handles structure with handles and user data (see GUIDATA)
```

```
flist = dir('database');
```

```
if length(flist) == 2
```

msgbox('NOTHING TO DELETE','INFO','modal');

return

end

```
for k = 3:length(flist)
```

```
na1(k-2,1) = {flist(k).name};
```

```
[a,b] = listdlg('promptstring','SELECT FILE/FILES TO DELETE','liststring',na1,'listsize',[125 100]);
```

if b == 0

return

end

```
cd ('database')
```

for k = 1:length(a)

str = na1{k};

delete(str)

end

cd ..

```
flist = dir('database');
```

```
if length(flist) == 2
```

```
msgbox({'NOTHING TO RENAME';'ALL DELETED'},'INFO','modal');
```

return

end

```
cd('database')
```

flist = dir(pwd);

```
for k = 3:length(flist)
```

```
z = flist(k).name;
```

z(strfind(z, '.') : end) = [];

nlist(k-2) = str2double(z);

```
nlist = sort(nlist);
```

h = waitbar(0, 'PLZ WAIT, WHILE MATLAB IS

RENAMING','name','PROGRESS...');

```
for k = 1:length(nlist)
```

```
if k ~= nlist(k)
```

```
p = nlist(k);
```

```
movefile([num2str(p) '.jpg'] , [num2str(k) '.jpg'])
```

```
waitbar((k-2)/length(flist),h,sprintf('RENAMED %s to %s',[num2str(p)
'.jpg'],[num2str(k) '.jpg']))
```

end

```
pause(.5)
```

end

```
close(h)
```

cd ..

% -----

```
function BR_OWSE_Callback(hObject, eventdata, handles)
```

```
% hObject handle to BR_OWSE (see GCBO)
```

```
% eventdata reserved - to be defined in a future version of MATLAB
```

```
% handles structure with handles and user data (see GUIDATA)
```

```
[f,p] = uigetfile('*.jpg','PLEASE SELECT AN FACIAL IMAGE');
```

if f == 0

return

end

```
p1 = fullfile(p,f);
```

```
im = imread(p1);
```

```
fd = vision.CascadeObjectDetector();
```

```
bbox = step(fd, im);
```

vo = insertObjectAnnotation(im,'rectangle',bbox,'FACE');

```
r = size(bbox, 1);
```

if isempty(bbox)

fhx = figure(2);

set(fhx,'menubar','none','numbertitle','off','name','PREVIEW')

imshow(vo);

```
msgbox({'WHAT HAVE U CHOOSEN?';'NO FACE FOUND IN THIS
PIC,';'SELECT SINGLE FACE IMAGE.'},'WARNING...!!!','warn','modal')
```

uiwait

```
delete(fhx)
```

return

```
elseif r > 1
```

fhx = figure(2);

set(fhx,'menubar','none','numbertitle','off','name','PREVIEW')

imshow(vo);

msgbox({'TOO MANY FACES IN THIS PIC';'PLEASE SELECT SINGLE FACE

IMAGE'},'WARNING...!!!','warn','modal')

uiwait

delete(fhx)

return

end

bx = questdlg({'CORRECT IMAGE IS SELECTED';'SELECT OPTION FOR FACE EXTRACTION'},'SELECT AN OPTION','MANUALLY','AUTO','CC');

```
if strcmp(bx,'MANUALLY') == 1
```

while 1

fhx = figure(2);

set(fhx,'menubar','none','numbertitle','off','name','PREVIEW')

imc = imcrop(im);

bbox1 = step(fd, imc);

if size(bbox1,1) ~= 1

```
msgbox({'YOU HAVENT CROPED A FACE';'CROP AGAIN'},'BAD
ACTION','warn','modal')
```

uiwait

else

break

end

close gcf

end

close gcf

```
imc = imresize(imc,[300 300]);
```

cd ('database');

I = length(dir(pwd));

n = [int2str(l-1) '.jpg'];

imwrite(imc,n);

cd ..

while 1

```
qq = inputdlg('WHAT IS UR NAME?','FILL');
```

if isempty(qq)

```
msgbox({'YOU HAVE TO ENTER A NAME';' ';'YOU CANT CLICK
CANCEL'},'INFO','HELP','MODAL')
```

uiwait

else

break

end

end

 $qq = qq\{1\};$

if exist('info.mat','file') == 2

load ('info.mat')

r = size(z2, 1);

 $z2{r+1,1} = {n, qq};$

save('info.mat','z2')

else

```
z2{1,1} = {n,qq};
```

```
save('info.mat','z2')
```

end

```
if strcmp(bx,'AUTO') == 1
```

```
imc = imcrop(im,[bbox(1)-50 bbox(2)-250 bbox(3)+100 bbox(4)+400]);
```

fhx = figure(2);

```
set(fhx,'menubar','none','numbertitle','off','name','PREVIEW')
```

imshow(imc)

qx = questdlg({'ARE YOU SATISFIED WITH THE RESULTS?';' ';'IF YES THEN
PROCEED';' ';'IF NOT BETTER DO MANUAL
CROPING'},'SELECT','PROCEED','MANUAL','CC');

```
if strcmpi(qx,'proceed') == 1
```

```
imc = imresize(imc,[300 300]);
```

cd ('database');

```
I = length(dir(pwd));
```

```
n = [int2str(l-1) '.jpg'];
```

imwrite(imc,n);

cd ..

while 1

```
qq = inputdlg('WHAT IS UR NAME?','FILL');
```

if isempty(qq)

```
msgbox({'YOU HAVE TO ENTER A NAME';' ';'YOU CANT CLICK
CANCEL'},'INFO','HELP','MODAL')
```

```
uiwait
else
break
end
end
qq = qq{1};
```

```
if exist('info.mat','file') == 2
```

load ('info.mat')

```
r = size(z2,1);
```

 $z2\{r{+}1{,}1\}=\{n\ ,\ qq\};$

save('info.mat','z2')

else

```
z2{1,1} = {n,qq};
```

```
save('info.mat','z2')
```

end

close gcf

```
elseif strcmpi(qx, 'manual') == 1
```

while 1

```
fhx = figure(2);
```

set(fhx,'menubar','none','numbertitle','off','name','PREVIEW')

imc = imcrop(im);

```
bbox1 = step(fd, imc);
```

```
if size(bbox1,1) ~= 1
```

msgbox({'YOU HAVENT CROPED A FACE';'CROP AGAIN'},'BAD

ACTION', 'warn', 'modal')

uiwait

else

break

end

close gcf

end

close gcf

```
imc = imresize(imc,[300 300]);
```

cd ('database');

I = length(dir(pwd));

n = [int2str(l-1) '.jpg'];

imwrite(imc,n);

cd ..

while 1

qq = inputdlg('WHAT IS UR NAME?','FILL');

if isempty(qq)

msgbox({'YOU HAVE TO ENTER A NAME';' ';'YOU CANT CLICK CANCEL'},'INFO','HELP','MODAL')

uiwait

else

break

end

 $qq = qq\{1\};$

if exist('info.mat','file') == 2

load ('info.mat')

r = size(z2, 1);

 $z2{r+1,1} = {n, qq};$

save('info.mat','z2')

else

 $z2{1,1} = {n,qq};$

save('info.mat','z2')

end

else

return

end

end

% -----

function FRM_CAM_Callback(hObject, eventdata, handles)

% hObject handle to FRM_CAM (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

global co

```
if isfield(handles,'vdx')
```

vid = handles.vdx;

```
stoppreview(vid)
```

delete(vid)

```
handles = rmfield(handles,'vdx');
```

guidata(hObject,handles)

cla(handles.axes1)

```
reset(handles.axes1)
```

```
set(handles.axes1,'box','on','xcolor','w','ycolor','w','xtick',[],'ytick',[],'color',[0.0431 0.5176 0.7804],'linewidth',1.5)
```

```
cla(handles.axes2)
```

reset(handles.axes2)

```
set(handles.axes2,'box','on','xcolor','w','ycolor','w','xtick',[],'ytick',[],'color',[0.0431 0.5176 0.7804],'linewidth',1.5)
```

end

```
fd = vision.CascadeObjectDetector();
```

```
info = imaqhwinfo('winvideo');
```

```
did = info.DeviceIDs;
```

```
if isempty(did)
```

```
msgbox({'YOUR SYSTEM DO NOT HAVE A WEBCAM';' ';'CONNECT A ONE'},'WARNING....!!!!','warn','modal')
```

return

end

did = cell2mat(did);

for k = 1:length(did)

devinfo = imaqhwinfo('winvideo',k);

na(1,k) = {devinfo.DeviceName};

sr(1,k) = {devinfo.SupportedFormats};

end

[a,b] = listdlg('promptstring', 'SELECT A WEB CAM

DEVICE','liststring',na,'ListSize', [125, 75],'SelectionMode','single');

if b == 0

return

end

if b ~= 0

frmt = sr{1,a};

[a1,b1] = listdlg('promptstring','SELECT RESOLUTION','liststring',frmt,'ListSize', [150, 100],'SelectionMode','single');

if b1 == 0

return

end

end

```
frmt = frmt{a1};
```

```
I = find(frmt == '_');
```

res = frmt(l+1 : end);

I = find(res == 'x');

```
res1 = str2double(res(1: I-1));
res2 = str2double(res(I+1 : end));
axes(handles.axes1)
vid = videoinput('winvideo', a);
vr = [res1 res2];
nbands = get(vid,'NumberofBands');
h2im = image(zeros([vr(2) vr(1) nbands] , 'uint8'));
preview(vid,h2im);
handles.vdx = vid;
guidata(hObject,handles)
tx = msgbox('PLZ STAND IN FRONT OF CAMERA STILL','INFO.....!!!');
pause(1)
delete(tx)
kx = 0;
```

```
while 1
```

```
im = getframe(handles.axes1);
```

im = im.cdata;

bbox = step(fd, im);

```
vo = insertObjectAnnotation(im,'rectangle',bbox,'FACE');
```

```
axes(handles.axes2)
```

imshow(vo)

```
if size(bbox, 1) > 1
```

msgbox({'TOO MANY FACES IN FRAME';' ';'ONLY ONE FACE IS

ACCEPTED'},'WARNING.....!!!','warn','modal')

uiwait

```
stoppreview(vid)
```

delete(vid)

```
handles = rmfield(handles,'vdx');
```

```
guidata(hObject,handles)
```

cla(handles.axes1)

```
reset(handles.axes1)
```

```
set(handles.axes1,'box','on','xtick',[],'ytick',[],'xcolor',[1 1 1],'ycolor',[1 1
```

```
1],'color',co,'linewidth',1.5)
```

```
cla(handles.axes2)
```

```
reset(handles.axes2)
```

```
set(handles.axes2,'box','on','xtick',[],'ytick',[],'xcolor',[1 1 1],'ycolor',[1 1 1],'color',co,'linewidth',1.5)
```

return

end

```
kx = kx + 1;
```

```
if kx > 10 && ~isempty(bbox)
```

break

end

```
end
```

```
\mathsf{imc} = \mathsf{imcrop}(\mathsf{im}, [\mathsf{bbox}(1)+3 \quad \mathsf{bbox}(2)-35 \quad \mathsf{bbox}(3)-10 \quad \mathsf{bbox}(4)+70]);
```

```
imx = imresize(imc,[300 300]);
```

```
fhx = figure(2);
```

set(fhx,'menubar','none','numbertitle','off','name','PREVIEW')

imshow(imx)

```
cd ('database');
```

```
I = length(dir(pwd));
```

```
n = [int2str(l-1) '.jpg'];
```

```
imwrite(imx,n);
```

```
cd ..
```

while 1

```
qq = inputdlg('WHAT IS UR NAME?','FILL');
```

```
if isempty(qq)
```

```
msgbox({'YOU HAVE TO ENTER A NAME';' ';'YOU CANT CLICK
CANCEL'},'INFO','HELP','MODAL')
```

uiwait

else

break

end

end

 $qq = qq\{1\};$

```
if exist('info.mat','file') == 2
```

load ('info.mat')

r = size(z2, 1);

 $z2{r+1,1} = {n, qq};$

```
save('info.mat','z2')
```

else

```
z2{1,1} = {n,qq};
```

save('info.mat','z2')

end

close gcf

stoppreview(vid)

delete(vid)

```
handles = rmfield(handles,'vdx');
```

```
guidata(hObject,handles)
```

```
cla(handles.axes1)
```

```
reset(handles.axes1)
```

```
set(handles.axes1,'box','on','xtick',[],'ytick',[],'xcolor',[1 1 1],'ycolor',[1 1 1],'color',co,'linewidth',1.5)
```

```
cla(handles.axes2)
```

```
reset(handles.axes2)
```

```
set(handles.axes2,'box','on','xtick',[],'ytick',[],'xcolor',[1 1 1],'ycolor',[1 1 1],'color',co,'linewidth',1.5)
```

% --- Executes on key press with focus on edit1 and none of its controls.

function edit1_KeyPressFcn(hObject, eventdata, handles)

% hObject handle to edit1 (see GCBO)

% eventdata structure with the following fields (see UICONTROL)

% Key: name of the key that was pressed, in lower case

% Character: character interpretation of the key(s) that was pressed

% Modifier: name(s) of the modifier key(s) (i.e., control, shift) pressed

% handles structure with handles and user data (see GUIDATA)

```
pass = get(handles.edit1,'UserData');
```

```
v = double(get(handles.figure1,'CurrentCharacter'));
```

if v == 8

pass = pass(1:end-1);

set(handles.edit1,'string',pass)

elseif any(v == 65:90) || any(v == 97:122) || any(v == 48:57)

pass = [pass char(v)];

elseif v == 13

p = get(handles.edit1,'UserData');

```
if strcmp(p,'123') == true
```

delete(hObject);

delete(handles.pushbutton2)

delete(handles.pushbutton1);

delete(handles.text2);

delete(handles.text3);

delete(handles.text1);

delete(handles.text4);

msgbox('WHY DONT U READ HELP BEFORE

STARTING', 'HELP....!!!', 'help', 'modal')

set(handles.AD_NW_IMAGE,'enable','on')

set(handles.DE_LETE,'enable','on')

set(handles.TRAIN_ING,'enable','on')

set(handles.STA_RT,'enable','on')

set(handles.RESET_ALL,'enable','on')

set(handles.EXI_T,'enable','on')

set(handles.HE_LP,'enable','on')

set(handles.DATA_BASE,'enable','on')

set(handles.text5,'visible','on')

return

else

beep

msgbox('INVALID PASSWORD FRIEND...

```
XX','WARNING....!!!','warn','modal')
```

uiwait;

```
set(handles.edit1,'string',")
```

return

end

else

msgbox({'Invalid Password Character';'Can"t use Special Character'},'warn','modal')

uiwait;

set(handles.edit1,'string',")

return

end

set(handles.edit1,'UserData',pass)

```
set(handles.edit1,'String',char('*'*sign(pass)))
```

% -----

function VI_EW_Callback(hObject, eventdata, handles)

% hObject handle to VI_EW (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

```
f = dir('database');
```

if length(f) == 2

```
msgbox('YOUR DATA BASE HAS NO IMAGE TO DISPLAY', 'SORRY', 'modal')
```

return

end

I = length(f)-2;

while 1

a = factor(I);

if length(a) >= 4

break

l = l+1;

end

```
d = a(1: ceil(length(a)/2));
```

d = prod(d);

```
d1 = a(ceil(length(a)/2)+1 : end);
```

```
d1 = prod(d1);
```

zx = sort([d d1]);

figure('menubar','none','numbertitle','off','name','Images of Database','color',[0.0431 0.5176 0.7804],'position',[300 200 600 500])

for k = 3:length(f)

```
im = imread(f(k).name);
```

```
subplot(zx(1),zx(2),k-2)
```

imshow(im)

```
title(f(k).name,'fontsize',10,'color','w')
```

end

```
% -----
```

function Start_Training_Callback(hObject, eventdata, handles)

```
% hObject handle to Start_Training (see GCBO)
```

```
% eventdata reserved - to be defined in a future version of MATLAB
```

% handles structure with handles and user data (see GUIDATA)

ff = dir('database');

if length(ff) == 2

h = waitbar(0,'Plz wait Matlab is scanning ur database...','name','SCANNING IS
IN PROGRESS');

for k = 1:100

waitbar(k/100)

pause(0.03)

end

close(h)

msgbox({'NO IMAGE FOUND IN DATABASE';'FIRST LOAD YOUR DATABASE';'USE "ADD NEW IMAGE" MENU'},'WARNING....!!!','WARN','MODAL')

return

end

if exist('features.mat','file') == 2

bx = questdlg({'TRAINING HAS ALREDY BEEN DONE';' ';'WANT TO TRAIN DATABASE AGAIN?'},'SELECT','YES','NO','CC');

```
if strcmpi(bx,'yes') == 1
```

builddatabase

```
msgbox('TRAINING DONE....PRESS OK TO CONTINUE','OK','modal')
```

return

else

return

else

builddatabase

msgbox('TRAINING DONE PRESS OK TO CONTINUE', 'OK', 'modal')

return

end

% -----

function BYE_Callback(hObject, eventdata, handles)

% hObject handle to BYE (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

close gcf

% -----

function ATTENDENCE_Callback(hObject, eventdata, handles)

% hObject handle to ATTENDENCE (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

if exist('attendence_sheet.txt','file') == 2

```
winopen('attendence_sheet.txt')
```

else

msgbox('NO ATTENDENCE SHEET TO DISPLAY','INFO...!!!','HELP','MODAL') end

% -----

function DEL_ATTENDENCE_Callback(hObject, eventdata, handles)

% hObject handle to DEL_ATTENDENCE (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

```
if exist('attendence_sheet.txt','file') == 2
```

delete('attendence_sheet.txt')

msgbox('ATTENDENCE DELETED','INFO...!!!','MODAL')

else

msgbox('NO ATTENDENCE SHEET TO DELETE','INFO...!!!','HELP','MODAL') end

% -----

function Untitled_1_Callback(hObject, eventdata, handles)

% hObject handle to Untitled_1 (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

x = questdlg({'Resetting will Clear the followings: ';'1. Attendence_sheet';'2.
 Database';'3. features.mat';'4. Info.mat';'Do u want to continue?'},'Please select...!!');

```
if strcmpi(x,'yes') == 1
```

```
delete('attendence_sheet.txt')
```

```
delete('features.mat')
```

```
delete('info.mat')
```

```
cd ([pwd, '\database'])
```

```
f = dir(pwd);
```

```
for k = 1:length(f)
```

```
delete(f(k).name)
```

end

cd ..

```
cla(handles.axes1);
```

```
reset(handles.axes1);
```

```
set(handles.axes1,'box','on','xcolor','w','ycolor','w','xtick',[],'ytick',[],'color',[0.0431 0.5176 0.7804],'linewidth',1.5)
```

```
cla(handles.axes2);
```

```
reset(handles.axes2);
```

set(handles.axes2,'box','on','xcolor','w','ycolor','w','xtick',[],'ytick',[],'color',[0.0431

```
0.5176 0.7804],'linewidth',1.5)
```

```
set(handles.text5,'string',")
```

beep

msgbox('All Reset','Info','modal')

end

% -----

function Untitled_2_Callback(hObject, eventdata, handles)

% hObject handle to Untitled_2 (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

cla(handles.axes1);

```
reset(handles.axes1);
```

set(handles.axes1,'box','on','xcolor','w','ycolor','w','xtick',[],'ytick',[],'color',[0.0431 0.5176 0.7804],'linewidth',1.5)

```
cla(handles.axes2);
```

```
reset(handles.axes2);
```

set(handles.axes2,'box','on','xcolor','w','ycolor','w','xtick',[],'ytick',[],'color',[0.0431 0.5176 0.7804],'linewidth',1.5)

```
set(handles.text5,'string',")
```

% -----

function Untitled_3_Callback(hObject, eventdata, handles)

% hObject handle to Untitled_3 (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

% -----

function Untitled_4_Callback(hObject, eventdata, handles)

% hObject handle to Untitled_4 (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)

% -----

function Untitled_5_Callback(hObject, eventdata, handles)

% hObject handle to Untitled_5 (see GCBO)

% eventdata reserved - to be defined in a future version of MATLAB

% handles structure with handles and user data (see GUIDATA)