|  |  |
| --- | --- |
| https://lh4.googleusercontent.com/e7FOs35jOgD2La7Ja45za5gKcpt9lWe21HAC5TdNC1moeBiFyc4rHYK_PWirw8bcV8MhaXDoL-l4SD2EF1lgRD8dBzVoCr4Plai4ZuW4K4nIqLnuGrU | **The Hong Kong Polytechnic University** **Department of Computing** **COMP55422 Multimedia Computing, Systems and Applications** **Class of 2011-2012** |

**Group Assignment**   
  
  
  
**Topic: Face Recognition**

|  |  |  |
| --- | --- | --- |
| **Date of Submission:** | 16 April 2012 |  |
| **Group 6:** | NG Check Kiu Jay | 10629203G |
|  | CHUNG Hau Ying, Janice | 11521450G |
|  | LAM Wai Chung | 10634631G |
|  | WONG Kwan Shing | 10604036G |
|  | QING Pei | 11500811G |

# Abstract

Numbers of general face recognition issues has been a topic which addressed by analyzing the facial feature for many years. It sparked an explosion of interest in the topic of face recognition.

The ultimate challenge of face recognition is to effectively identify human faces. In this project, two methods are adapted to enhance the ability of image reorganization. The characteristics of these techniques might be suitable for many applications. These heuristics might put an effort to predict human face variants behavior under different conditions.

**Table of Content (TBC)**

# Introduction

**Background**

Since the past decade, face recognition gained a significant attention and become an active research area of computer vision. Face Recognition is a computer application for automatically identifying or verifying a person by using digital image or video source. This automatic capability can be used in law enforcement, security, biometric authentication and surveillance, etc, whereas it can perform effortlessly and routinely in our daily life. According to Zhao & Chellappa (2002), this is a useful tool to secure our asset and protect our privacy without losing the identity.

According to Li & Jain (2011), Takeo Kanade developed the first automated face recognition system for his Ph.D thesis work in 1973. He demonstrated a simple neural net which can perform face recognition for aligned and normalized face images. By approximating the eigenvectors of face image's autocorrelation matrix, the face description is computed. Such eigenvectors are now known as “Eigenfaces”.

In 1989, Kirby and Sirovich worked on low dimensional face representation by using the Karhunen-Loeve transform or Principal Component Analysis (PCA). They planned to make it in an easier way to directly calculate the Eigenfaces.

Regarding to the research of Turk & Pentland (1991), Turk & Pentland reinvigorated face recognition technology. At that time, when using eigenfaces, there was still residual error by detecting the face in cluttered natural imagery and determining the precise location and sale of faces in an image. Therefore, they derived a method to detect and localize faces with eigenface recognition. It could achieve reliable, real-time recognition of faces in a minimally constrained environment. In addition, this is relatively simple, and in real-time pattern.

Fisher face method also created interest to develop face recognition technology. According to the studies of Belhumeur (1997), they applied Linear Discriminant Analysis (LDA) adter a PCA setup. This suggestion can help to achieve higher accuracy. Effective facial features can be achieved by using local filters, for example Gabor jets. This is the milestone of real time face detection.

**Usage of Existing Face Recognition Technology**

Face recognition technology is commercially available as there is a strong demand on using Face Recognition technology as it can increase security in a convenient way. Suggested by Philips, Rauss & Der (1996), there are numbers of security and law enforcement tasks using face recognition technology. In monitoring areas (including airports, border crossings and secure manufacturing facilities and staff only areas), key and identity card are not necessary and the face recognize system only need to scan the users’ face. The technology also widely used in verifying identity. For example, users can search the photo record in a database for fraud detection.

3. Search and planning of face recognition (3 pages, Edward, Jay & Eric)

{Prior Research [Face Recognition/Pgm] - Findings , what is important to understand , it has been found that the processes involved are generally quite simple.

Brief introduction to our model - Algorithms, system workflow and diagrams }

3a) Illumination (upto 3 pages, Jay) – {prerequisite,assumption for the function design}

3b) Head pose (upto 3 pages, Eric) – {prerequisite,assumption for the function design}

# Illumination

Illumination processing is a necessary step for face recognition to be useful.   
In this section, we will focus on the influence on face recognition caused by illumination.[ <http://cdn.intechopen.com/pdfs/10200/InTech-Illumination_processing_in_face_recognition.pdf>]

Illumination variation can be removed: the input face images can be normalized to some state where comparisons are more reliable. [http://www.intechopen.com/download/pdf/pdfs\_id/10200] We are going to introduce two popular algorithms to do the normalization on the image, they are based on the *Retinex* theory. The SSR (Single-Scale Retinex) and the MSR (Multi-Scale Retinex) could probably solve the problem of illumination of the images. They will be discussed in details in this section. In fact there are many different kind of algorithms doing similar work such as *“adaptive single scale retinex”, “DCT based normalization”, “wavelet based normalization”, “anisotropic diffusion based normalization”* etc. We are going to propose the most common algorithms in this report.

# SSR (Single-Scale Retinex)

The single scale retinex (SSR) algorithm was proposed by Jobson et al. It is normalization techniques based on retinex theory. Each output value of the function is depended by the corresponding input value and its neighbourhood. The surround function is a Gaussian function. The equation of the single-scale Retinex (SSR) is given by



where I is the input image, R is the Retinex output. F is a Gaussian filter (surround or kernel) defined by  <http://dragon.larc.nasa.gov/background/pubabs/papers/gspx1.pdf>



Below is the example of applying SSR to the images.



# MSR (Multi-Scale Retinex)

The multi-scale retinex (MSR) algorithm is an extension of the single scale

retinex algorithm again proposed by Jobson et al. The mathematical form of the multi-scale Retinex (MSR) is given by

where Rn(x,y) are different scale SSR (obtained with different Gaussian functions) and Wn is the weight of each SSR. Usually, these weights are taken to be equal. <http://www.ee.bgu.ac.il/~Orly_lab/Publications/Marius_paper1_SPIE.pdf>

Below is the example of applying MSR to the images.



# Finalize

From the above findings, we could say that MSR should be better but it is more complicated. In our project, we chose to apply the SSR instead of MSR because it is easier to implement and we believed that the processing time would be faster.

The function of SSR in our project would receive an INPUT of a grey- scale image and the OUTPUT should be a grey-scale image processed with the SSR algorithm.

# Head Pose

The proportion of human head varies from person to person and growth change may affect the facial outlook too. In general, it includes general size, shape and position of facial features in front face view. In this section, we mainly focus to head pose variations detection and attempt to correct the face angle if high confidence level can be obtained by using our algorithm.

Left

Mask

Right

Mask

Top

Mask

Bottom

Mask



Left

Mask

Right

Mask

Top

Mask

Bottom

Mask

Left

Mask

Right

Mask

Top

Mask

Bottom

Mask

Left

Mask

Right

Mask

Top

Mask

Bottom

Mask

Left

Mask

Right

Mask

Top

Mask

Bottom

Mask

Figure – Various head pose conditions to be tracked

In our project, we adopt most common facial features and assigned sample image set of faces[[1]](#footnote-1) taken by Olivetti Research Laboratory in Cambridge, UK between April 1992 and April 1994. The key workflows are 1)Skin extraction, 2)noise removal, 3)inverse skin for 4)face components detection and 5)eye orientation angle determination.

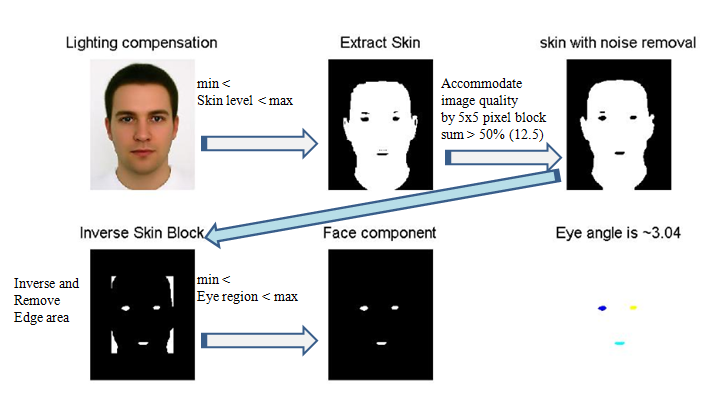


Figure - Workflow of headpose detection

The following outline depicts the human head features and the characteristics are being derived.

a) The width of human head is approximately two thirds of its height.

b) The front view of human face can be divided into three sections, namely the top, middle and the bottom parts. Each section is approximately one third of the face view section.

|  |  |
| --- | --- |
| Top Section  Bottom  Section  Middle Section | * Top section : cover most of the hair feature (head down to the hairline) |
| * Middle section: cover eyes, nose and ears. (hairline down to eyes, usually same level with the ears and nose a the lower part aligning the ear lobes. |
| * Bottom section : include mouth (bottom up from chin to the mouth position) |

The proportions will varies if the face view turns, head from above or below. We assume the eyes as the focal point of head image, and the eyebrows are situated approximately from 1/3 down the head without head turn above or below. These distances work out at approximately two third of the width of the face. Also the eye proportion is between 4% and 12.5% image width and between 1.6% and 6.6% image height..

Eyemin = round(ImageWidth/25 \* ImageHeight/60);

Eyemax = round(ImageWidth/8 \*ImageHeight/15);

As you can see from the illustration below, the two eyes focus areas bisects the head of the vertical line.

|  |  |  |
| --- | --- | --- |
| Left  Mask  Right  Mask  Top  Mask  Bottom  Mask  Head pose with vertical line |  | Left  Mask  Right  Mask  Top  Mask  Bottom  Mask  1  6  1  6  1  6  1  6  1  6  1  6  1  6  1  6  Image layout with grid lines |

In many cases, eye is the only facial feature which is easily and effectively identifiable from facial skin. We detect eyes depending basically on the intensity level variance between Skin level sampled. The Skin intensity level sampling are taken by introducing three circular line (CL[1:3,1:12]) tracks starting from centre of the image, each with approximately one third of the image width.

|  |  |
| --- | --- |
| Figure - Circular-Line tracking of skin level | The skin intensity level samples are taken in 30 degree increment clockwise starting from Y axis and our observation tends to use the mean value taken from the two inner circular lines for our setup. By making an assumption of intensity variance of face skin level with a variations upto 50%. |

skinlevelmin = min( mean(CL1(1,:)), mean(CL1(2,:)) ) \* 0.5;

skinlevelmax = max( mean(CL1(1,:)), mean(CL1(2,:)) ) \*1.5;

However, we do not consider complex conditions such as different head view projections as they will fail under our imaging conditions.

[SkinIndexRow,SkinIndexCol] =find(skinlevelmin<I & I<skinlevelmax);

for i=1:length(SkinIndexRow)

S(SkinIndexRow(i),SkinIndexCol(i))=1;

end

After the extraction of facial skin region, a simple noise removal routine by examining a 5x5 pixels block each with 50% intensity level ‘1’. Then by using Matlab function “regionprops(CC,'Area');” and “L = labelmatrix(CC); “, we attempt to keep area with our pre-determined eye area ranges, and also apply a peripheral mask so as to focus our eye components inside face Skin area.

|  |  |  |
| --- | --- | --- |
| Figure - Region mask layout | TopArea =>between 0 and 20% Height  BottomArea=>between 80-100% Height  LeftArea => between 0-20% Width  RightArea =>between 80-100% Width | |
| By utilizing the Matlab function regionprops (Measure properties of image regions) with 'BoundingBox' option, the coordinate of the two eye components, if it can be detected, and the associated head pose angle turn can be determined. | | Left  Mask  Right  Mask  Top  Mask  Bottom  Mask  θ°  Figure – head pose turn angle determinion |

Finally, we attempt to rotate the head angle and extract the two eye regions, its similarity of both sides, for further comparison by checking the mean difference of the intensity level of each eye regions. We suggest applying the head pose correction if the confidence level is greater than 90%.

Because of noise in the images due to lighting condition, facial expression or wearing, the successful detection rate (i.e. images with correctly detected eye/nose region relative to the whole set of facial database) is not so high, but it is reasonable.

The efficiency of the method is evaluated with ORL databases (40 sets variety of images). It has been shown by experimental results that 18 out of 400 images can

Be detected by the proposed method. Also 7 out of 11 additional images can be detected and we consider the outcomes are acceptable.

The followings are the seven samples by implementing rgb2gray transformation from colour images.

|  |  |
| --- | --- |
| Rgb2gray transformation | Sample #1 |
| Sample #2 | Sample #3 |
| Sample #4 | Sample #5 |
| Sample #6 | Sample #7 |

4. System Testing / Result (Edward & Chris, procedure and test summary, wait until program completed)

{ The Facial recognition method was implemented in MATLAB and examines with set of different images .....

Ready Images sample baseline and training set building for program library, prepare test image for program test, comparison result}

# Testing Result

|  |  |  |  |  |
| --- | --- | --- | --- | --- |
| **ImageID** | **Headpose** | **No Feature** | **All Features** | **Image\_enhanced** |
| 1 | 0 | 0 | 0 | 0 |
| 2 | 1 | 1 | 1 | 1 |
| 3 | 1 | 1 | 1 | 1 |
| 4 | 1 | 1 | 1 | 1 |
| 5 | 1 | 1 | 1 | 1 |
| 6 | 1 | 1 | 1 | 1 |
| 7 | 1 | 1 | 0 | 0 |
| 8 | 1 | 1 | 1 | 1 |
| 9 | 1 | 1 | 1 | 1 |
| 10 | 1 | 1 | 0 | 0 |
| 11 | 1 | 1 | 1 | 1 |
| 12 | 1 | 1 | 1 | 1 |
| 13 | 1 | 1 | 1 | 1 |
| 14 | 1 | 1 | 1 | 1 |
| 15 | 1 | 1 | 1 | 1 |
| 16 | 1 | 1 | 1 | 1 |
| 17 | 1 | 1 | 1 | 1 |
| 18 | 1 | 1 | 1 | 1 |
| 19 | 1 | 1 | 1 | 1 |
| 20 | 1 | 1 | 1 | 1 |
| 21 | 1 | 1 | 1 | 1 |
| 22 | 1 | 1 | 1 | 1 |
| 23 | 1 | 1 | 1 | 1 |
| 24 | 0 | 0 | 0 | 0 |
| 25 | 0 | 0 | 0 | 0 |
| 26 | 1 | 1 | 1 | 1 |
| 27 | 1 | 1 | 1 | 1 |
| 28 | 1 | 1 | 1 | 1 |
| 29 | 1 | 1 | 1 | 1 |
| 30 | 1 | 1 | 1 | 1 |
| 31 | 1 | 1 | 1 | 1 |
| 32 | 1 | 1 | 1 | 1 |
| 33 | 1 | 1 | 1 | 1 |
| 34 | 1 | 1 | 1 | 1 |
| 35 | 1 | 1 | 1 | 1 |
| 36 | 0 | 0 | 0 | 0 |
| 37 | 1 | 1 | 1 | 1 |
| 38 | 1 | 1 | 1 | 1 |
| 39 | 1 | 1 | 1 | 1 |
| 40 | 1 | 1 | 1 | 1 |
| 41 | 1 | 1 | 1 | 1 |
| 42 | 1 | 1 | 1 | 1 |
| 43 | 1 | 1 | 1 | 1 |
| 44 | 1 | 1 | 1 | 1 |
| 45 | 1 | 1 | 1 | 1 |
| 46 | 1 | 1 | 1 | 1 |
| 47 | 1 | 1 | 1 | 1 |
| 48 | 1 | 1 | 1 | 1 |
| 49 | 1 | 1 | 1 | 1 |
| 50 | 1 | 1 | 0 | 0 |
| 51 | 1 | 1 | 1 | 1 |
| 52 | 1 | 1 | 1 | 1 |
| 53 | 1 | 1 | 1 | 1 |
| 54 | 1 | 1 | 1 | 1 |
| 55 | 1 | 1 | 1 | 1 |
| 56 | 1 | 1 | 1 | 1 |
| 57 | 1 | 1 | 1 | 1 |
| 58 | 1 | 1 | 1 | 1 |
| 59 | 1 | 1 | 1 | 1 |
| 60 | 1 | 1 | 1 | 1 |
| 61 | 1 | 1 | 1 | 1 |
| 62 | 0 | 0 | 1 | 1 |
| 63 | 1 | 1 | 1 | 1 |
| 64 | 1 | 1 | 1 | 1 |
| 65 | 1 | 1 | 1 | 1 |
| 66 | 1 | 1 | 1 | 1 |
| 67 | 1 | 1 | 1 | 1 |
| 68 | 1 | 1 | 1 | 1 |
| 69 | 1 | 1 | 1 | 1 |
| 70 | 1 | 1 | 1 | 1 |
| 71 | 1 | 1 | 1 | 1 |
| 72 | 1 | 1 | 1 | 1 |
| 73 | 1 | 1 | 1 | 1 |
| 74 | 1 | 1 | 1 | 1 |
| 75 | 1 | 1 | 1 | 1 |
| 76 | 1 | 1 | 1 | 1 |
| 77 | 1 | 1 | 1 | 1 |
| 78 | 1 | 1 | 1 | 1 |
| 79 | 1 | 1 | 1 | 1 |
| 80 | 1 | 1 | 1 | 1 |
| 81 | 1 | 1 | 1 | 1 |
| 82 | 1 | 1 | 1 | 1 |
| 83 | 1 | 1 | 1 | 1 |
| 84 | 1 | 1 | 1 | 1 |
| 85 | 1 | 1 | 1 | 1 |
| 86 | 1 | 1 | 1 | 1 |
| 87 | 1 | 1 | 1 | 1 |
| 88 | 1 | 1 | 1 | 1 |
| 89 | 1 | 1 | 1 | 1 |
| 90 | 1 | 1 | 1 | 1 |
| 91 | 1 | 1 | 1 | 1 |
| 92 | 1 | 1 | 1 | 1 |
| 93 | 1 | 1 | 1 | 1 |
| 94 | 1 | 1 | 1 | 1 |
| 95 | 1 | 1 | 1 | 1 |
| 96 | 1 | 1 | 1 | 1 |
| 97 | 1 | 1 | 1 | 1 |
| 98 | 1 | 1 | 1 | 1 |
| 99 | 1 | 1 | 1 | 1 |
| 100 | 0 | 0 | 0 | 0 |
| 101 | 1 | 1 | 1 | 1 |
| 102 | 1 | 1 | 1 | 1 |
| 103 | 1 | 1 | 1 | 1 |
| 104 | 1 | 1 | 1 | 1 |
| 105 | 1 | 1 | 1 | 1 |
| 106 | 1 | 1 | 1 | 1 |
| 107 | 1 | 1 | 1 | 1 |
| 108 | 1 | 1 | 1 | 1 |
| 109 | 1 | 1 | 1 | 1 |
| 110 | 1 | 1 | 1 | 1 |
| 111 | 1 | 1 | 1 | 1 |
| 112 | 1 | 1 | 1 | 1 |
| 113 | 1 | 1 | 1 | 1 |
| 114 | 1 | 1 | 1 | 1 |
| 115 | 1 | 1 | 1 | 1 |
| 116 | 1 | 1 | 1 | 1 |
| 117 | 1 | 1 | 1 | 1 |
| 118 | 1 | 1 | 1 | 1 |
| 119 | 1 | 1 | 1 | 1 |
| 120 | 1 | 1 | 1 | 1 |
| 121 | 1 | 1 | 1 | 1 |
| 122 | 1 | 1 | 1 | 1 |
| 123 | 1 | 1 | 1 | 1 |
| 124 | 1 | 1 | 1 | 1 |
| 125 | 1 | 1 | 1 | 1 |
| 126 | 1 | 1 | 1 | 1 |
| 127 | 1 | 1 | 1 | 1 |
| 128 | 1 | 1 | 1 | 1 |
| 129 | 1 | 1 | 1 | 1 |
| 130 | 1 | 1 | 1 | 1 |
| 131 | 1 | 1 | 1 | 1 |
| 132 | 1 | 1 | 1 | 1 |
| 133 | 1 | 1 | 1 | 1 |
| 134 | 1 | 1 | 1 | 1 |
| 135 | 1 | 1 | 1 | 1 |
| 136 | 1 | 1 | 1 | 1 |
| 137 | 1 | 1 | 1 | 1 |
| 138 | 1 | 1 | 1 | 1 |
| 139 | 1 | 1 | 0 | 0 |
| 140 | 1 | 1 | 1 | 1 |
| 141 | 1 | 1 | 1 | 1 |
| 142 | 1 | 1 | 1 | 1 |
| 143 | 1 | 1 | 1 | 1 |
| 144 | 1 | 1 | 1 | 1 |
| 145 | 1 | 1 | 1 | 1 |
| 146 | 1 | 1 | 1 | 1 |
| 147 | 1 | 1 | 1 | 1 |
| 148 | 1 | 1 | 0 | 0 |
| 149 | 1 | 1 | 1 | 1 |
| 150 | 1 | 1 | 1 | 1 |
| 151 | 1 | 1 | 1 | 1 |
| 152 | 1 | 1 | 1 | 1 |
| 153 | 1 | 1 | 1 | 1 |
| 154 | 1 | 1 | 1 | 1 |
| 155 | 1 | 1 | 1 | 1 |
| 156 | 1 | 1 | 1 | 1 |
| 157 | 1 | 1 | 1 | 1 |
| 158 | 1 | 1 | 1 | 1 |
| 159 | 1 | 1 | 1 | 1 |
| 160 | 0 | 0 | 0 | 0 |
| 161 | 1 | 1 | 1 | 1 |
| 162 | 1 | 1 | 1 | 1 |
| 163 | 1 | 1 | 1 | 1 |
| 164 | 1 | 1 | 1 | 1 |
| 165 | 1 | 1 | 1 | 1 |
| 166 | 1 | 1 | 1 | 1 |
| 167 | 1 | 1 | 1 | 1 |
| 168 | 1 | 1 | 1 | 1 |
| 169 | 1 | 1 | 1 | 1 |
| 170 | 1 | 1 | 1 | 1 |
| 171 | 1 | 1 | 1 | 1 |
| 172 | 1 | 1 | 1 | 1 |
| 173 | 1 | 1 | 1 | 1 |
| 174 | 1 | 1 | 1 | 1 |
| 175 | 1 | 1 | 1 | 1 |
| 176 | 1 | 1 | 1 | 1 |
| 177 | 1 | 1 | 1 | 1 |
| 178 | 1 | 1 | 1 | 1 |
| 179 | 1 | 1 | 1 | 1 |
| 180 | 0 | 0 | 0 | 0 |
| 181 | 1 | 1 | 1 | 1 |
| 182 | 1 | 1 | 1 | 1 |
| 183 | 1 | 1 | 1 | 1 |
| 184 | 1 | 1 | 1 | 1 |
| 185 | 1 | 1 | 1 | 1 |
| 186 | 1 | 1 | 1 | 1 |
| 187 | 1 | 1 | 1 | 1 |
| 188 | 1 | 1 | 1 | 1 |
| 189 | 0 | 0 | 0 | 0 |
| 190 | 1 | 1 | 0 | 0 |
| 191 | 1 | 1 | 1 | 1 |
| 192 | 1 | 1 | 1 | 1 |
| 193 | 1 | 1 | 1 | 1 |
| 194 | 1 | 1 | 1 | 1 |
| 195 | 1 | 1 | 1 | 1 |
| 196 | 1 | 1 | 1 | 1 |
| 197 | 1 | 1 | 1 | 1 |
| 198 | 1 | 1 | 1 | 1 |
| 199 | 1 | 1 | 1 | 1 |
| 200 | 1 | 1 | 1 | 1 |
| 201 | 1 | 1 | 1 | 1 |
| 202 | 1 | 1 | 1 | 1 |
| 203 | 1 | 1 | 1 | 1 |
| 204 | 1 | 1 | 1 | 1 |
| 205 | 1 | 1 | 1 | 1 |
| 206 | 1 | 1 | 1 | 1 |
| 207 | 1 | 1 | 1 | 1 |
| 208 | 1 | 1 | 1 | 1 |
| 209 | 1 | 1 | 1 | 1 |
| 210 | 1 | 1 | 1 | 1 |
| 211 | 1 | 1 | 1 | 1 |
| 212 | 0 | 0 | 1 | 1 |
| 213 | 1 | 1 | 1 | 1 |
| 214 | 1 | 1 | 1 | 1 |
| 215 | 0 | 1 | 1 | 1 |
| 216 | 1 | 1 | 0 | 0 |
| 217 | 1 | 1 | 1 | 1 |
| 218 | 1 | 1 | 1 | 1 |
| 219 | 1 | 1 | 1 | 1 |
| 220 | 1 | 1 | 1 | 1 |
| 221 | 1 | 1 | 1 | 1 |
| 222 | 1 | 1 | 1 | 1 |
| 223 | 1 | 1 | 1 | 1 |
| 224 | 1 | 1 | 1 | 1 |
| 225 | 1 | 1 | 1 | 1 |
| 226 | 1 | 1 | 1 | 1 |
| 227 | 1 | 1 | 1 | 1 |
| 228 | 1 | 1 | 1 | 1 |
| 229 | 1 | 1 | 1 | 1 |
| 230 | 1 | 1 | 1 | 1 |
| 231 | 1 | 1 | 1 | 1 |
| 232 | 1 | 1 | 1 | 1 |
| 233 | 1 | 1 | 1 | 1 |
| 234 | 1 | 1 | 1 | 1 |
| 235 | 1 | 1 | 0 | 0 |
| 236 | 1 | 1 | 1 | 1 |
| 237 | 1 | 1 | 1 | 1 |
| 238 | 1 | 1 | 1 | 1 |
| 239 | 1 | 1 | 1 | 1 |
| 240 | 1 | 1 | 1 | 1 |
| 241 | 1 | 1 | 1 | 1 |
| 242 | 0 | 1 | 1 | 1 |
| 243 | 1 | 1 | 1 | 1 |
| 244 | 1 | 1 | 1 | 1 |
| 245 | 1 | 1 | 1 | 1 |
| 246 | 1 | 1 | 1 | 1 |
| 247 | 1 | 1 | 1 | 1 |
| 248 | 1 | 1 | 1 | 1 |
| 249 | 1 | 1 | 1 | 1 |
| 250 | 1 | 1 | 1 | 1 |
| 251 | 1 | 1 | 1 | 1 |
| 252 | 1 | 1 | 0 | 0 |
| 253 | 1 | 1 | 1 | 1 |
| 254 | 1 | 1 | 1 | 1 |
| 255 | 1 | 1 | 1 | 1 |
| 256 | 1 | 1 | 1 | 1 |
| 257 | 1 | 0 | 0 | 0 |
| 258 | 1 | 1 | 1 | 1 |
| 259 | 0 | 0 | 1 | 1 |
| 260 | 1 | 1 | 1 | 1 |
| 261 | 1 | 1 | 1 | 1 |
| 262 | 1 | 1 | 1 | 1 |
| 263 | 1 | 1 | 1 | 1 |
| 264 | 1 | 1 | 1 | 1 |
| 265 | 1 | 1 | 1 | 1 |
| 266 | 1 | 1 | 1 | 1 |
| 267 | 1 | 1 | 1 | 1 |
| 268 | 1 | 1 | 1 | 1 |
| 269 | 1 | 1 | 1 | 1 |
| 270 | 0 | 0 | 0 | 0 |
| 271 | 1 | 1 | 0 | 0 |
| 272 | 1 | 1 | 1 | 1 |
| 273 | 1 | 1 | 1 | 1 |
| 274 | 1 | 1 | 1 | 1 |
| 275 | 1 | 1 | 1 | 1 |
| 276 | 1 | 1 | 1 | 1 |
| 277 | 1 | 1 | 1 | 1 |
| 278 | 1 | 0 | 0 | 0 |
| 279 | 1 | 1 | 1 | 1 |
| 280 | 1 | 1 | 0 | 0 |
| 281 | 1 | 1 | 1 | 1 |
| 282 | 1 | 1 | 1 | 1 |
| 283 | 1 | 1 | 1 | 1 |
| 284 | 1 | 1 | 1 | 1 |
| 285 | 1 | 1 | 1 | 1 |
| 286 | 1 | 1 | 1 | 1 |
| 287 | 1 | 1 | 1 | 1 |
| 288 | 1 | 1 | 1 | 1 |
| 289 | 1 | 1 | 1 | 1 |
| 290 | 1 | 1 | 1 | 1 |
| 291 | 1 | 1 | 1 | 1 |
| 292 | 1 | 1 | 1 | 1 |
| 293 | 1 | 1 | 1 | 1 |
| 294 | 1 | 1 | 1 | 1 |
| 295 | 1 | 1 | 1 | 1 |
| 296 | 1 | 1 | 1 | 1 |
| 297 | 1 | 1 | 1 | 1 |
| 298 | 1 | 1 | 0 | 0 |
| 299 | 1 | 1 | 1 | 1 |
| 300 | 1 | 1 | 1 | 1 |
| 301 | 1 | 1 | 1 | 1 |
| 302 | 1 | 1 | 1 | 1 |
| 303 | 1 | 1 | 1 | 1 |
| 304 | 1 | 1 | 1 | 1 |
| 305 | 1 | 1 | 1 | 1 |
| 306 | 1 | 1 | 1 | 1 |
| 307 | 1 | 1 | 1 | 1 |
| 308 | 1 | 1 | 1 | 1 |
| 309 | 1 | 1 | 1 | 1 |
| 310 | 1 | 1 | 1 | 1 |
| 311 | 1 | 1 | 1 | 1 |
| 312 | 1 | 1 | 1 | 1 |
| 313 | 1 | 1 | 1 | 1 |
| 314 | 1 | 1 | 1 | 1 |
| 315 | 1 | 1 | 1 | 1 |
| 316 | 1 | 1 | 1 | 1 |
| 317 | 0 | 0 | 0 | 0 |
| 318 | 1 | 1 | 1 | 1 |
| 319 | 1 | 1 | 1 | 1 |
| 320 | 1 | 1 | 1 | 1 |
| 321 | 1 | 1 | 0 | 0 |
| 322 | 1 | 1 | 1 | 1 |
| 323 | 1 | 1 | 1 | 1 |
| 324 | 1 | 1 | 1 | 1 |
| 325 | 1 | 1 | 1 | 1 |
| 326 | 0 | 0 | 1 | 1 |
| 327 | 1 | 1 | 1 | 1 |
| 328 | 0 | 0 | 0 | 0 |
| 329 | 1 | 1 | 1 | 1 |
| 330 | 1 | 1 | 1 | 1 |
| 331 | 1 | 1 | 1 | 1 |
| 332 | 1 | 1 | 1 | 1 |
| 333 | 1 | 1 | 1 | 1 |
| 334 | 1 | 1 | 1 | 1 |
| 335 | 1 | 1 | 1 | 1 |
| 336 | 1 | 1 | 1 | 1 |
| 337 | 1 | 1 | 1 | 1 |
| 338 | 1 | 1 | 1 | 1 |
| 339 | 1 | 1 | 1 | 1 |
| 340 | 1 | 1 | 1 | 1 |
| 341 | 1 | 1 | 0 | 0 |
| 342 | 1 | 1 | 1 | 1 |
| 343 | 0 | 0 | 0 | 0 |
| 344 | 1 | 1 | 1 | 1 |
| 345 | 1 | 1 | 1 | 1 |
| 346 | 1 | 0 | 0 | 0 |
| 347 | 1 | 1 | 1 | 1 |
| 348 | 1 | 1 | 1 | 1 |
| 349 | 1 | 1 | 1 | 1 |
| 350 | 0 | 0 | 0 | 0 |
| 351 | 1 | 1 | 1 | 1 |
| 352 | 0 | 0 | 1 | 1 |
| 353 | 1 | 1 | 0 | 0 |
| 354 | 1 | 1 | 1 | 1 |
| 355 | 1 | 1 | 1 | 1 |
| 356 | 1 | 1 | 1 | 1 |
| 357 | 1 | 1 | 1 | 1 |
| 358 | 1 | 1 | 0 | 0 |
| 359 | 0 | 0 | 1 | 1 |
| 360 | 0 | 0 | 1 | 1 |
| 361 | 1 | 1 | 1 | 1 |
| 362 | 1 | 1 | 1 | 1 |
| 363 | 1 | 1 | 1 | 1 |
| 364 | 1 | 1 | 1 | 1 |
| 365 | 1 | 1 | 1 | 1 |
| 366 | 1 | 1 | 1 | 1 |
| 367 | 1 | 1 | 1 | 1 |
| 368 | 1 | 1 | 1 | 1 |
| 369 | 1 | 1 | 1 | 1 |
| 370 | 1 | 1 | 1 | 1 |
| 371 | 1 | 1 | 1 | 1 |
| 372 | 1 | 1 | 1 | 1 |
| 373 | 1 | 1 | 1 | 1 |
| 374 | 1 | 1 | 0 | 0 |
| 375 | 1 | 1 | 1 | 1 |
| 376 | 0 | 0 | 1 | 1 |
| 377 | 1 | 1 | 1 | 1 |
| 378 | 0 | 0 | 1 | 1 |
| 379 | 0 | 0 | 1 | 1 |
| 380 | 1 | 1 | 1 | 1 |
| 381 | 1 | 1 | 1 | 1 |
| 382 | 1 | 1 | 1 | 1 |
| 383 | 1 | 1 | 1 | 1 |
| 384 | 1 | 1 | 1 | 1 |
| 385 | 1 | 1 | 1 | 1 |
| 386 | 1 | 1 | 1 | 1 |
| 387 | 1 | 1 | 1 | 1 |
| 388 | 1 | 1 | 1 | 1 |
| 389 | 1 | 1 | 1 | 1 |
| 390 | 0 | 0 | 0 | 0 |
| 391 | 1 | 1 | 1 | 1 |
| 392 | 1 | 1 | 1 | 1 |
| 393 | 1 | 1 | 1 | 1 |
| 394 | 1 | 1 | 0 | 0 |
| 395 | 1 | 1 | 1 | 1 |
| 396 | 1 | 1 | 1 | 1 |
| 397 | 1 | 1 | 1 | 1 |
| 398 | 1 | 1 | 1 | 1 |
| 399 | 0 | 0 | 0 | 0 |
| 400 | 1 | 1 | 1 | 1 |
| **Success Rate** | **373 / 400** | **372 / 400** | **364 / 400** | **364 / 400** |

Above table shown the details success rate in different situation in 400 images, the comparison including “no features”, “headpose”, “both features” and “image enhanced”.

**Head pose**

From the result, the success rate of processing with “headpose” feature is a little bit higher than without any features. This is all depends on the training dataset as we have rotated some of the images to do the headpose testing. For example, ImageID 257 is one of them; it has been successfully recognized with headpose feature. It is meaning that the more images has been rotated, the more success rate of “headpose”.

**Image enhacement**

5 . Conclusion (Janice, any idea?)

{The initial goal of this research was to find if ....

The process involved was fairly simple. .......

It is recommended that any follow-up work focuses on using ...in order to predict human face as conditions change.}

# Conclusion

As our initial goal of this research is improving the effectiveness of face recognition technology, it is recommended to focus on improving illumination and detects the head pose variations. We suggested different algorithms to predict human face and improve the accuracy of face recognition while conditions change. From the testing result, it shown that ……………….by implementing the face recognition method in MATLAB with numbers of images.

# Reference

Belhumeur, P.N., hespanha, J.P., Kriegman, D.J.: Eigenfaces vs. Fisherfaces (1997): recognition using class specific linear projection. IEEE Trans. Pattern Anal. Mach. Intell. 19(7), 711-720

Li, S. Z. & Jain, A. K. (2011). Hand Book of Face Recognition. London: Springer-Verlag London Limited.

Philips, P. J., Rauss, P. J. & Der, S. Z. (October 1996). FERET (Face Recognition Technology) Recognition Algorithm Development and Test Results. U.S. Army Research Laboratory, 73.

Turk, M.A., Pentland, A.P. (1991): Eigenfaces for Recognition. J. Cogn. Necurosci. 3(1), 71-86

Zhao, W., and Chellappa, R. (2002), “Image-based Face Recognition: Issues and Methods”, in B. Javidi (ed.), *Image Recognition and Classification,* Mercel Dekker

<http://www.ee.bgu.ac.il/~Orly_lab/Publications/Marius_paper1_SPIE.pdf>

<http://203.72.2.115/Ejournal/AC01001602.pdf>

<http://cdn.intechopen.com/pdfs/10200/InTech-Illumination_processing_in_face_recognition.pdf>

<http://www.mathworks.com/matlabcentral/fileexchange/26523>

1. The ORL face database - This directory contains a set of faces taken between April 1992 and April 1994 at the Olivetti Research Laboratory in Cambridge, UK. There are 10 different images of 40 distinct subjects. For some of the subjects, the images were taken at different times, varying lighting slightly, facial expressions (open/closed eyes, smiling/non-smiling) and facial details glasses/no-glasses). All the images are taken against a dark homogeneous background and the subjects are in up-right, frontal position (with tolerance for some side movement) The files are in PGM format and can be conveniently viewed using the 'xv' program. The size of each image is 92x112, 8-bit grey levels. The images are organised in 40 directories (one for each subject) named as: sX where X indicates the subject number (between 1 and 40). In each directory there are 10 different images of the selected subject named as: Y.pgm where Y indicates which image for the specific subject (between 1 and 10). [↑](#footnote-ref-1)