



Faculty of Engineering

PERSON IDENTIFICATION BASED ON MULTIMODAL BIOMETRIC RECOGNITION

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Final Year Project Report

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PERSON IDENTIFICATION BASED ON MULTIMODAL BIOMETRIC RECOGNITION

ALEX NG HO LIAN

A final year project report submitted in partial fulfilment of
the requirement for the degree of
Bachelor of Engineering (Hons) in Electronics (Telecommunications)

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ABSTRACT

Unimodal biometric systems have limited effectiveness in identifying people, mainly due to their susceptibility to changes in individual biometric features and presentation attacks. The identification of people using multimodal biometric systems attracts researchers' attention due to their advantages, such as greater recognition efficiency and greater security compared to the unimodal biometric system. A multimodal biometric system can overcome various unimodal biometric systems' limitations, so it is suitable and recommended use for this society. In this project, face and fingerprint recognition are used to develop a multimodal biometric system. In the process of face recognition, Classic Convolutional Neural Network (CNN) is used for training face datasets. After done training face dataset, the testing process is needed to recognize a face with face dataset. In the process of fingerprint recognition, the ORB algorithm is recommended to use in feature matching. ORB (Oriented FAST and Rotated BRIEF) algorithm consists of 3 stages: feature point extraction, defining feature point descriptors, and computing feature point matching. For these three stages, the fingerprint image is matching with the fingerprint database. For the process of fusion of face and fingerprint recognition, two features are fused by match score level fusion based on Weighted Sum-Rule. If the fusion score is higher than the threshold level is given, then the verification process is matched. The result of accuracy is displayed if the user selects the same biometric characteristics for both recognition. If the fusion score is less than the threshold level, then the verification process indicates a mismatch. The result of accuracy will not be displayed if the user selects different biometric characteristics for both recognition.

ABSTRAK

Sistem biometrik unimodal mempunyai keberkesanan terhad dalam mengenal pasti orang, terutamanya kerana kerentanan mereka terhadap perubahan ciri biometrik individu dan serangan persembahan. Pengenalpastian orang yang menggunakan sistem biometrik multimodal menarik perhatian penyelidik kerana kelebihanannya, seperti kecekapan pengiktirafan dan keselamatan yang lebih besar berbanding sistem biometrik unimodal. Sistem biometrik multimodal dapat mengatasi pelbagai batasan sistem biometrik unimodal, jadi penggunaannya sesuai dan disarankan untuk masyarakat ini. Dalam projek ini, pengecaman wajah dan cap jari digunakan untuk mengembangkan sistem biometrik multimodal. Dalam proses pengecaman wajah, Rangkaian Neural Konvolusional Klasik (CNN) digunakan untuk melatih set data wajah. Setelah selesai membuat set data wajah, proses ujian diperlukan untuk mengenali set data wajah dengan wajah. Dalam proses pengecaman cap jari, algoritma ORB disarankan untuk digunakan dalam pepadanan ciri. Algoritma ORB (Oriented FAST and Rotated BRIEF) terdiri daripada 3 peringkat: pengekstrakan titik ciri, menentukan deskriptor titik ciri, dan pepadanan titik ciri pengkomputeran. Untuk ketiga tahap ini, gambar cap jari sepadan dengan pangkalan data cap jari. Untuk proses penggabungan wajah dan pengecaman cap jari, dua ciri digabungkan dengan gabungan skor skor pertandingan berdasarkan Jumlah Berat Peraturan. Sekiranya skor gabungan lebih tinggi daripada tahap ambang yang diberikan, maka proses pengesahan sesuai. Hasil ketepatan ditunjukkan jika pengguna memilih ciri biometrik yang sama untuk kedua-dua pengiktirafan. Sekiranya skor peleburan kurang dari tahap ambang, maka proses pengesahan menunjukkan ketidakcocokan. Hasil ketepatan tidak akan ditampilkan jika pengguna memilih ciri biometrik yang berbeza untuk kedua-dua pengiktirafan.

Table of contents

	Page
ACKNOWLEDGMENT	ii
ABSTRACT	iii
ABSTRAK	iv
Table of contents	v
List of Tables	ix
List of Figures	x
List of Equations	xiii
List of Abbreviations	xiv
CHAPTER 1	1
INTRODUCTION	1
1.1 Research Background	1
1.2 Problem Statement	6
1.3 Objective	8
1.4 Project Scopes	8
1.5 Project Outline	9
CHAPTER 2	10
LITERATURE REVIEW	10
2.1 Introduction	10
2.2 Biometric System	10
2.3 Face Recognition	15
2.4 Face Detection with Different Techniques	16
2.4.1 Spatio-Temporal Filtering in Face Detection	16

2.4.2	Real-Time Face Detection	17
2.4.3	Face Detection Using Templates	17
2.4.4	Face Detection using Fuzzy Pattern	18
2.4.5	3D object detection	18
2.4.6	Cascade Object Detection	18
2.4.7	Canny Edge Detection	19
2.5	Image Pre-processing	20
2.5.1	Image Preprocessing in Different Techniques	20
2.5.2	Preprocessing using elastic Bunch Graph with Gabor filters	24
2.6	Detection of Singular Points in Fingerprint Images.	25
2.7	Feature Extraction	27
2.7.1	Principal component analysis (PCA)	27
2.7.2	Independent Component Analysis	30
2.7.3	Curvelet transforms	30
2.7.4	Linear Discriminant Analysis (LDA)	33
2.7.5	Incremental Linear Discriminant Analysis (ILDA)	37
2.8	Support Vector Machine (SVM) Classifier	38
2.9	Knowledge Transfer	39
2.10	KNN Classification	41
2.11	Conclusion	43
CHAPTER 3		44
METHODOLOGY		44
3.1	Introduction	44
3.2	Project Overview	44
3.3	Multimodal Biometric Recognition	45
3.4	Software	46
3.5	Hardware	46
3.6	Image Dataset	47

3.7	Face Recognition	47
3.7.1	Face Images Collection and Pre-processing	48
3.7.2	Convolution Neural Network and Softmax Classifier	48
3.7.2.1	Convolution Layer	49
3.7.2.2	Pooling Layer	49
3.7.2.3	ReLU Layer	49
3.7.2.4	Fully Connected Layer	50
3.7.3	Summary	50
3.8	Fingerprint Recognition	50
3.8.1	Pre-processing in Fingerprint Recognition	51
3.8.2	Fingerprint Enhancement	52
3.8.3	Binarization	54
3.8.4	Thinning	55
3.8.5	Fingerprint Matching with OcomRB (Oriented FAST and Rotated BRIEF)	55
3.9	Matching Score Level Fusion	57
3.10	Conclusion	58
CHAPTER 4		60
RESULTS AND DISCUSSION		60
4.1	Introduction	60
4.2	Face Recognition	60
4.2.1	Convolutional Neural Network (CNN's)	61
4.2.2	Comparison of ILDA with SVM and CNN's model	63
4.3	Fingerprint Recognition	66
4.3.1	ORB algorithm (Oriented FAST and Rotated BRIEF)	66
4.3.2	Comparison of Resnet50 and ORB algorithm	68
4.4	Multimodal Biometric Recognition (Face and Fingerprint)	71
4.5	Summary	76
CHAPTER 5		77

CONCLUSIONS AND RECOMMENDATIONS	77
5.1 Overview	77
5.2 Conclusion	77
5.3 Limitation and Recommendation	78
REFERENCES	80
APPENDIX	86

List of Tables

Table		Page
Table 2.2	Some Advantages and Disadvantages of Biometric System	15
Table 2.7.4	The difference between LDA and PCA	36
Table 3.8.2	The process of Minutiae-based features	54
Table 4.2.2.1	ILDA with SVM and CNN's Accuracy	63
Table 4.2.2.2	The result of Student's T-Test	64
Table 4.3.2.1	Highest accuracy for Resnet50 and ORB	68
Table 4.3.2.2	The result of Student's T-Test	69
Table 4.4	The result of 2 samples after fusion	75

List of Figures

Figure		Page
Figure 1.1	Biometric System	2
Figure 1.1.2	Partial Fingerprint in Feature Matching	6
Figure 2.2.1	Three Phases of Biometric System	11
Figure 2.2.2	General Biometric System	13
Figure 2.2.3	Biometric System Error	14
Figure 2.3	Block Diagram of Face Recognition System	16
Figure 2.4.7	Comparison between the Original Images with the Edge Image	19
Figure 2.5.1.1	Block Diagram of Image Preprocessing	20
Figure 2.5.1.2	Impact of image cropping on FR rate	21
Figure 2.5.1.3	Impact of normalization on FR rate	22
Figure 2.5.1.4	Recognition Rate after Image Cropping	23
Figure 2.5.1.5	Resized Images by scale 0.3	23
Figure 2.5.1.6	Recognition Rate after Brightness Increases and Decreases	24
Figure 2.6.1	Proposed Result	25
Figure 2.6.2	Oriented Picture Computation	26
Figure 2.7.1.1	Region Processing by PCA	27
Figure 2.7.1.2	The result of face Recognition by using PCA	29
Figure 2.7.3.1	Curvelet Transform	31
Figure 2.7.3.2	Block Diagram of Blood Vessel Retina	32
Figure 2.7.3.3	Proposed Result	33
Figure 2.7.4.1	Transformation of LDA	34
Figure 2.7.4.2	The process of LDA separation	34
Figure 2.7.4.3	LDA Separation	35
Figure 2.7.4.4	Different Classes between 2 Persons	36
Figure 2.8	Linearly Separable Samples of SVM	38
Figure 2.9	Knowledge Transfer	40
Figure 2.10.1	Training error rate with varying value of K	41
Figure 2.10.2	The workflow of K nearest neighbor	42

Figure 3.3	Block Diagram of a Multimodal Biometric System	45
Figure 3.5	Python GUI for Multimodal Biometric Recognition	46
Figure 3.7	General Structure of Face Recognition	47
Figure 3.7.2	Design of Convolution Neural Network	48
Figure 3.8	General Structure of Fingerprint Recognition	50
Figure 3.8.1.1	Ridges and Valleys of Fingerprint	51
Figure 3.8.2	Left side) Original Image Right Side) Enhanced Image	53
Figure 3.8.3	Result of binarization of fingerprint	54
Figure 3.8.4	Result of Thinning of Fingerprint	55
Figure 3.8.5.1	Block Diagram of ORB	55
Figure 3.8.5.2	Result of Brute Force Matcher	56
Figure 3.9	Block Diagram of Score Level Fusion	57
Figure 4.2	Labelling face datasets in folder 1 to 20 of each person	60
Figure 4.2.1.1	CNN's model training process	61
Figure 4.2.1.2	Sample 1 of testing result	61
Figure 4.2.1.3	Sample 2 of testing result	62
Figure 4.2.1.4	Sample 11 of testing result	62
Figure 4.2.2	T Distribution Table	65
Figure 4.3	Labelling fingerprint datasets in folder 1 to 20 of each person	66
Figure 4.3.1.1	ORB algorithm matching process	66
Figure 4.3.1.2	The result of Sample 1	67
Figure 4.3.1.3	The result of Sample 3	67
Figure 4.3.1.4	The result of Sample 15	68
Figure 4.3.2	T Distribution Table	70
Figure 4.4.1	The result of Sample 1 after fusion	71
Figure 4.4.2	The result of Sample 8 after fusion	72
Figure 4.4.3	The result of Sample 11 after fusion	72
Figure 4.4.4	The result of Sample 18 after fusion	73

Figure 4.4.5	Both traits are not same person	74
Figure 4.4.6	Both traits are not same person	74

List of Equations

Equation Page

(1)	LDA Separation	35
(2)-(14)	ILDA algorithm	37-38
(15)	KNN Classification	41
(16)-(19)	Weighted Sum-Rule	57-58
(20)-(22)	Calculation between ILDA + SVM and CNN's based on Student's T-Test	64
(23)-(25)	Calculation between Restnet50 and ORB based on Student's T-Test	69

List of Abbreviations

AI	-	Artificial Intelligence
CNN	-	Convolution Neural Network
ORB	-	Oriented FAST and Rotation Briefing
UBS	-	Unimodal Biometric System
MBS	-	Multimodal Biometric System
GUI	-	Graphic User Interface
ERR	-	Equal Error Rate
FAR	-	False Accept Ratio
FRR	-	False Reject Ratio
DCT	-	Discrete Cosine Transform
PCA	-	Principal Component Analysis
ICA	-	Independent Component Analysis
LDA	-	Linear Discriminant Analysis
ILDA	-	Incremental Linear Discriminant Analysis
SVM	-	Support Vector Machine
BGR	-	Blue Green Red
RGB	-	Red Green Blue
FCL	-	Fully Connected Layer
MLP	-	Multi-layer perception

CHAPTER 1

INTRODUCTION

1.1 Research Background

Nowadays, the security system becomes more important than ever because this is an era of industry 4.0. However, some security systems do not use the latest technology, making it easy for unauthorized people to access or invade the system, so it is called traditional security. In the past few years, it has been confirmed that the use of traditional security systems will lead to a rising crime rate of cyber-attacks. These attacks now target industrial espionage, financial crime, and even government and critical infrastructure from time to time. It also has many flaws such as passwords that may be lost, hacked or passed, plastic IDs may be duplicated, and RFID cards can be copied. All these flaws occur because it does not provide strong security in data protection. Therefore, all these problems will be solved in the generation of Industry 4.0. Industry 4.0 generation's focus has shifted to artificial intelligence (AI) applications such as neural networks, machine learning, and deep learning, which define new working cultures in almost all industries. For AI applications, biometric systems can make cybersecurity more effective and respond to growing threats and improve an organization's cybersecurity posture [1]. A biometric system is a technical system used to identify a person with information about him. It provides high accuracy and confident automatic recognition to identify the person based on biometric traits such as hand, finger-vein, iris, face, voice, fingerprint, signature, and keystroke, as shown in Figure 1.1.1. The advantages of using Biometric systems were, it only can be accessed when the user to identify with the recognition devices. It provides a high accessibility system for the user without memorizing the passwords or codes. It makes it difficult for cybercriminals because they really cannot easily access the entry points and devices of the network [2].

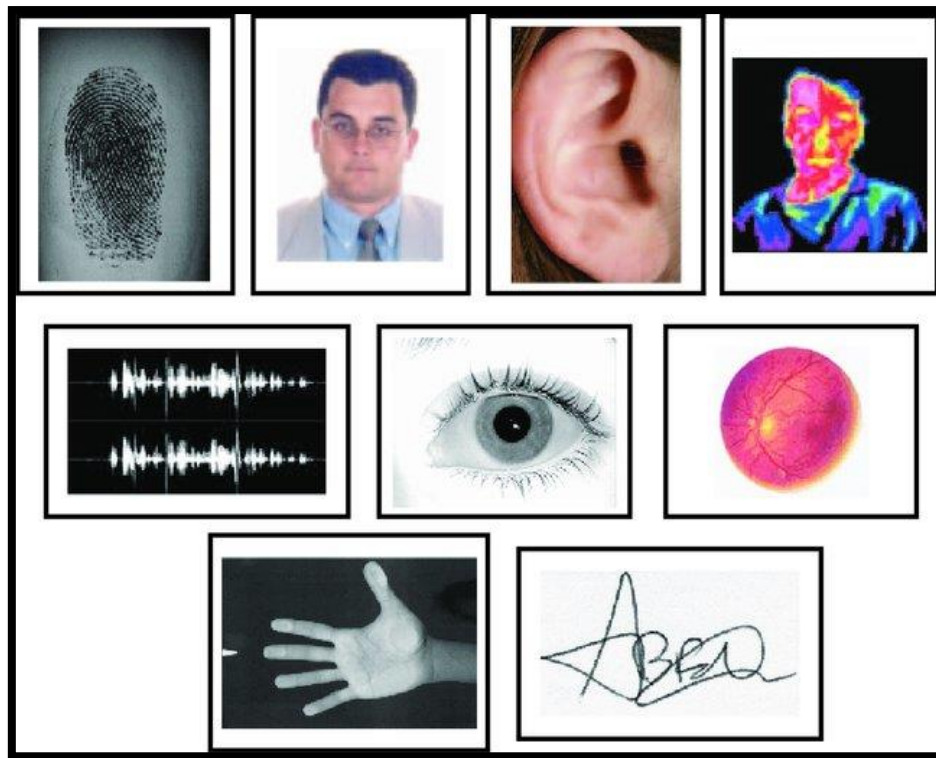


Figure 1.1.1: Biometric System [1]

As shown in Figure 1.1.1, the biometric system has two main characteristics: physiological and behavioral. Physiological characteristics refer to human traits, such as fingerprint, face, hand, iris recognition, etc. For behavior, it refers to keystrokes, signatures, sounds. It relates to human behavior. However, there are many types of biometric systems, namely Unimodal biometric systems and Multimodal biometric systems. For unimodal biometric systems, they use only one identification function. In contrast, multimodal biometric recognition systems use more than one feature for recognition, involving multiple sensors [2].

With the development of the technological world, the use of biometric technology to implement secure personal identification protocols is facing more and more challenges. In almost every market in the world, there is an increasing demand for targeted identification. The biometric management market has discovered that using a single hardware model for identification in many industries may no longer be the wisest choice. The unimodal biometric identification system must deal with various issues, such as noisy data, non-universality, mutations within the class, uniqueness, and

deceptive attacks. By deploying a multimodal biometric system that integrates multiple biometric patterns in a single scan, some of these limitations are solved, thereby solving an unimodal system's challenge [3].

A multimodal biometric system is a system that can use multiple physiological or behavioral characteristics for registration, identification, and verification. Multimodal biometric recognition has become the latest technology in human attention, and one of the most important reasons for combining different modes is to improve recognition accuracy. There are other reasons for combining two or more biometrics, for example, different biometrics are more suitable for various deployment scenarios, or this fact becomes a reality when security is essential to protect sensitive data [4].

The multimodal biometric system obtains input from multiple biometric identification devices to measure different biometrics traits. For example, combining faces and fingerprints for biometric identification is called a multimodal biometric system, even if different facial and fingerprint images are acquired through different devices or the same biometric identification. There is no need to combine various indicators mathematically, because the biological characteristics are always independent of each other, which can improve the accuracy of identifying people [3].

There are some advantages of multimodal biometric system as shows [2]:

- Accuracy: Multi-pattern biometrics uses multiple patterns to identify people who can ensure higher accuracy.
- Security: The multi-mode biometric identification system improves security by eliminating the possibility of spoofing. It is impossible for one to deceive multiple biological characteristics at once.
- Activity detection: The multi-mode biometric identification system requires end users to randomly submit multiple biometrics to ensure reliable activity detection to prevent fraud or hacking.
- Universality: The multimodal biometric system is universal; even if the disabled person cannot provide some form of biometric information, other forms of biometric information can also be used for identity verification.

- Cost-effective: Multi-mode biometric systems reduce the risk of violations or criminal attacks by providing a higher level of security, which is cost-effective.

When the input of the system is destroyed, problems will occur in the unimodal biometric system. The performance of the system will also decrease. Therefore, the multi-mode biometric system can solve the problem and improve the accuracy, and the system will be more difficult to breakthrough. In this project, face and fingerprint recognition will be selected for multi-modal biometric recognition [5].

Face recognition is to identify people related to the face through the scanning process. Because of its use in security systems, video surveillance, access control, business, and even Facebook and other social networks, it has become an essential tool for human-computer interaction. With the development of artificial intelligence, face recognition has attracted people's attention compared with other recognition. It is the primary method of human personal identification. In an uncontrolled environment, face recognition is easily checked, even if the subject cannot understand [5].

Researchers can find that the problem has been solved [5]. Traditional learning methods always faces challenges such as posture changes, facial camouflage, scene lighting, image background complexity, and facial expression changes. Methods based on shallow learning only use some features of the image and rely on artificial experience to extract sample characteristics. Deep learning methods can extract more complex face characteristics. Deep learning has made significant progress in solving the problems that have plagued the AI industry's best attempts for many years. The practice has proved that it is very good at revealing the intricate structure in high-dimensional data, so it is suitable for many fields such as science, business, and government. It uses one or several algorithms to solve the problem of learning hierarchical representation and performs well in natural language processing, semantic segmentation, image recognition, and many other practical scenarios. Deep learning involves some methods, such as deep belief networks (DBN) [6], stacked auto encoders [7], and convolutional neural networks (CNN).

CNN is an algorithm that mostly used in face recognition. CNN is an artificial neural network that uses a convolution technique to extract features from input data.

CNN was first proposed by LeCun and first applied to the recognition of handwriting. His network was the origin of many modern constructions and the real inspiration of many scientists. Researchers published their production in the ImageNet contest [8]. It is known as one of the most influential publication figures in the field of computer vision. It also shows that the performance of recognition on CNN is better than that based on manual methods. Besides, with the help of the computing power of the graphics processing unit (GPU), CNN achieved extraordinary cutting-edge results in many fields, including scene recognition, image recognition, edge detection, and semantic segmentation [8].

The overall structure of the face recognition process in this project consists of three steps. For the preprocessing step, it involves color space conversion and image resizing, and then continues to extract facial features, and then classifies the feature set has been extracted. At last, Softmax Classifier is the last stage of classifier the dataset file's name based on the facial features extracted from CNN. Hence, the CNN algorithm will be used in face recognition.

Fingerprints are the most commonly used biometric function in judicial and law worldwide [9]. Today, embedded fingerprint chips used in mobile devices such as smartphones have become common. Fingerprints chips can provide excellent security for various information and memory they save on mobile devices. Nevertheless, the chip size has its limitation, such as cost and space. It can usually only collect a small number of fingerprints at a time.

Therefore, multiple templates are needed combining in the fingerprint enrollment process, so each template includes different parts of the fingerprint. Traditional minutiae fingerprint image requires a large amount of overlapping two fingerprint images in merging technology. However, the fingerprint sensor chip extracts insufficient details at a size of $160 * 160$ and dpi 500. Therefore, detail-based technology requires touch sensors to capture many images. At this moment, due to insufficiently detailed features such as endpoints, triangle points, and some detailed comparison algorithms are applied to compare parts of the fingerprint, which significantly reduces the accuracy. Previous researchers have found solutions to combine feature details and texture features for various alignment algorithms [10].

Reseachers	Features	Database	Types
Jea et al.(2005)[2]	Secondary features around minutiae	FVC 2002	Partial to full
Chen et al.(2007)[3]	Level 3 features	NIST sd30	Partial to full
Fang et al.(2007)[4]	Ridge points and minutiae	FVC 2002	Partial to partial
Zanganeh et al.(2014)[5]	pixel-wise correlation	FVC 2002	Partial to partial
Yamazaki et al.(2015)[6]	SIFT	Private Database	Partial to partial
Surbhi et al.(2016)[7]	AKAZE	FVC 2002	Partial to partial

Figure 1.1.2: Partial Fingerprint in Feature Matching[11]

In Figure 1.1.2, most of the fingerprint algorithms in feature matching are mainly based on multiscale texture descriptors. Besides, SIFT [12] and SURF [13] algorithms are the most popular two-dimensional multiscale description and feature detection. However, ORB is the algorithm that is used in fingerprint recognition in this project. ORB (Oriented FAST and Rotation Briefing) is an algorithm that extracting and describing feature points very quickly. ORB algorithm was proposed by Vincent Rabaud et al. [14]. The ORB algorithm improves and optimizes the datasets by combining the detection between FAST feature points and Brief descriptors. Also, the process of the ORB algorithm is faster than SIFT and SURF.

After used CNN algorithm in face recognition and ORB algorithm in fingerprint recognition. Both recognition are fused by using score level fusion based on weighted sum-rule. Graphical User Interface (GUI) will be developed as a system to combine face and fingerprint recognition and calculate the accuracy after both are merged.

1.2 Problem Statement

In recent years, a non-biometric system such as a password or a code system has been used to make a payment or unlock a phone a thing of the past. In the age of information technology, people are already switching to biometric systems such as face

and fingerprint recognition. It was used to maintain the security system resources and to manage the identity of the user.

Semi-automated biometric devices such as face recognition were developed at the 1960s. Then, fingerprint and face recognition were widely used in law enforcement at 1969s. In 1985, the idea was proposed that was unique to everyone, including fingerprints, irises. In 1994, the first iris recognition algorithm was patented. Then, the first real-time recognition was developed in 1991. Hundred recognition algorithms were patented and functional by 2000s [15]. Apple Company also developed fingerprint to unlock the iPhone in 2013. All these biometric histories were helped to overcome the limitation of a non-biometric system. Therefore, the biometric system has been gaining popularity since the last two decades [16]. The efficiency of this system demonstrates the robustness of the security of the system.

The biometric system has two main features: physiological or behavioral, and controls the identity of a person and ensures much higher security than traditional person identification system. Physiological features of the biometrics are easy to use compared to behavioral features such as keystroke, voice or even signature. These features have categorized into Unimodal Biometric System and Multimodal Biometric System. For Unimodal Biometric System, it involve the face or fingerprint. It can store information in the data, but there are limitations. Although Unimodal has been chosen to recognize the face or fingerprints, there are a lot of tasks require to complete this project. In-person identification will not fully match when the system extracts data from the same biometric features because people are constantly changing over time. For example, a facial structure such as hairstyle, facial expression, facial angles and age changes will be a problem in facial recognition. In recognition of fingerprints, the valleys and ridges of the fingerprint will change due to human age. It is also a serious problem if there is a biometric trait disfigured. For example, it only can require one single task, and it has does not provide high accuracy for identification due to noise data, spoof attack, and so on [3].

Therefore, the Multimodal Biometric System (MBS) can perform two or more biometric features that used to overcome this limitation of a unimodal biometric system. The MBS development system has enhanced the recognition rate. The multimodal