



Faculty of Engineering

**MULTIMODAL PERSON RECOGNITION SYSTEM USING
CONVOLUTIONAL NEURAL NETWORK**

Jagathis A/L Karunakaran

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MULTIMODAL PERSON RECOGNITION SYSTEM USING CONVOLUTIONAL NEURAL NETWORK

JAGATHIS A/L KARUNAKARAN

A dissertation submitted in partial fulfilment
of the requirement for the degree of
Bachelor of Engineering
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ABSTRACT

Multimodal person recognition system in the current age is an alternative to unimodal biometric system. Recognition systems routinely used for person identification in various segments and commonly resort to daily life use. In this project, two pretrained Convolutional Neural Networks were trained, tested and evaluated using MATLAB R2021b. The networks are AlexNet with 25 layers and VGG16 with 16 layers. Then an optimal network was opted, AlexNet and used to train face and fingerprint dataset separately with different variation of hyperparameter. The face and fingerprint dataset used to train and test the networks are self-created face dataset and NIST Special Database 302 fingerprint dataset. The accuracy of the AlexNet for face and fingerprint were 95.00% and 98.67% respectively. The AlexNet model was evaluated with an accuracy of 98.67% and 100% in the 5-fold Validation test. The accuracy of the confusion matrix was the same for the AlexNet networks of face and fingerprint. The face and fingerprint networks were later fused in decision-level fusion to produce an overall multimodal recognition network. AlexNet has an average high accuracy of 96.84% setting a high standard for future work in multimodal person recognition suggesting AlexNet as an effective pretrained network for classification.

ABSTRAK

Sistem pengecaman orang multimodal pada zaman sekarang adalah alternatif kepada sistem biometrik unimodal. Sistem pengecaman yang digunakan secara rutin untuk mengenal pasti orang dalam pelbagai segmen dan biasanya menggunakan penggunaan kehidupan harian. Dalam projek ini, dua Rangkaian Neural Konvolusi yang telah dilatih telah dilatih, diuji dan dinilai menggunakan MATLAB R2021b. Rangkaian tersebut ialah AlexNet dengan 25 lapisan dan VGG16 dengan 16 lapisan. Kemudian rangkaian optimum telah dipilih, AlexNet dan digunakan untuk melatih set data muka dan cap jari secara berasingan dengan variasi hiperparameter yang berbeza. Set data muka dan cap jari yang digunakan untuk melatih dan menguji rangkaian ialah set data muka ciptaan sendiri dan set data cap jari Pangkalan Data Khas NIST 302. Ketepatan AlexNet untuk muka dan cap jari masing-masing adalah 95.00% dan 98.67%. Model AlexNet dinilai dengan ketepatan 98.67% dan 100% dalam ujian Pengesahan 5 kali ganda. Ketepatan matriks kekeliruan adalah sama untuk rangkaian muka dan cap jari AlexNet. Rangkaian muka dan cap jari kemudiannya digabungkan dalam gabungan peringkat keputusan untuk menghasilkan rangkaian pengecaman multimodal keseluruhan. AlexNet mempunyai purata ketepatan tinggi sebanyak 96.84% menetapkan standard yang tinggi untuk kerja masa hadapan dalam pengiktirafan orang berbilang mod yang mencadangkan AlexNet sebagai rangkaian pralatihan yang berkesan untuk pengelasan.

TABLE OF CONTENTS

Acknowledgement		i
Abstract		ii
Abstrak		iii
Table of contents		iv
List of tables		ix
List of figures		x
List of abbreviations		xiii
Chapter 1	INTRODUCTION	1
1.1	Background	1
1.2	Problem Statement	4
1.3	Objectives	5
1.4	Project Motivation	5
1.5	Project Scope	5
1.6	Chapter Outline	7
1.6.1	Chapter 1: Introduction	7
1.6.2	Chapter 2: Literature Review	7
1.6.3	Chapter 3: Methodology	7
1.6.4	Chapter 4: Result and Discussion	8
1.6.5	Chapter 5: Conclusion	8
Chapter 2	LITERATURE REVIEW	9
2.1	Overview	9
2.2	Unimodal & Multimodal Biometric System	9
2.3	Basic Framework of Multimodal Person Recognition System	10
2.4	Person Recognition System	12
2.4.1	Fingerprint Recognition system	12

2.4.2	Face Recognition system	13
2.5	Image Pre-processing	14
2.6	Image Feature Extraction	15
2.6.1	Minutiae Extraction	15
2.6.2	Local Binary Pattern	16
2.6.3	2D Gabor filter	17
2.6.4	Principal Component Analysis (PCA)	17
2.6.5	Comparison between the image feature extraction techniques	18
2.7	Image Classification	20
2.7.1	Neural Network	20
2.7.2	Support Vector Machine	21
2.7.3	Probabilistic Neural Network (PNN)	22
2.7.4	Comparison between the classification techniques	22
2.8	Dataset of Fingerprint & Face	23
2.8.1	Fingerprint dataset	23
2.8.2	Face dataset	24
2.9	Deep Learning	25
2.9.1	Introduction to CNN	25
2.9.2	AlexNet	27
2.9.3	VGGNet	27
2.9.4	ResNet	28
2.9.5	GoogLeNet	29
2.9.6	Comparison between CNN Architectures	29
2.10	Fusion Techniques in Multimodal Recognition System using deep learning	31
2.10.1	Sensor level fusion	31

	2.10.2 Feature level fusion	32
	2.10.3 Score level fusion	32
	2.10.4 Decision level fusion	33
2.11	Summary	33
Chapter 3	METHODOLOGY	35
3.1	Overview	35
3.2	Introduction	35
3.3	Planning Phase	37
	3.3.1 Hardware	38
	3.3.2 Software	38
	3.3.3 Face & Fingerprint Dataset	38
	3.3.4 Network Design	39
	3.3.5 Network Training	40
	3.3.6 Network Testing	40
3.4	Analysis Phase	40
	3.4.1 Dataset Analysis	41
	3.4.2 Pretrained Network Analysis	42
	3.4.3 Use Case Diagram	42
	3.4.4 Sequence Diagram	43
3.5	Design Phase	44
	3.5.1 Multimodal Person Recognition System Architecture	44
	3.5.2 The Project Flowchart for Algorithm Development	45
	3.5.3 Deep Learning Approach for Multimodal Person Recognition System	46
3.6	Implementation Phase	47
	3.6.1 Project Implementation Flowchart	47

	3.6.2	MATLAB	48
	3.6.3	Deep Network Designer	49
	3.7	Testing Phase	50
	3.7.1	K-Fold Cross Validation	50
	3.7.2	Confusion Matrix	51
	3.8	Gantt Chart	52
	3.9	Summary	54
Chapter 4		RESULTS AND DISCUSSION	55
	4.1	Introduction	55
	4.2	Graphical User Interface	55
	4.3	Design of Convolutional Neural Network	56
	4.3.1	AlexNet	56
	4.3.2	VGG16	59
	4.3.3	Similarity between the networks	61
	4.3.4	Comparison between properties of these networks	62
	4.4	Completed Training of Convolutional Neural Networks	64
	4.4.1	Hyperparameters for the networks	64
	4.4.2	AlexNet	65
	4.4.3	VGG16	66
	4.5	Completed Training of Face & Fingerprint Network	68
	4.5.1	Face Training Results	68
	4.5.2	Fingerprint Training Results	70
	4.5.3	Comparison between different hyperparameters	71
	4.6	Decision level fusion	74
	4.7	Network Testing	75
	4.7.1	Image testing for all the four classes	75

	4.7.2	Image testing containing noise, occlusion, and illumination	77
	4.7.3	Fusion Testing Outcome	80
	4.8	Performance Evaluation of Network	82
	4.8.1	K-Fold Cross-Validation	82
	4.8.2	Confusion Matrix	83
	4.9	Summary	88
Chapter 5		CONCLUSIONS	89
	5.1	Introduction	89
	5.2	Contribution	89
	5.3	Limitation and Challenges	91
	5.4	Future Works and Recommendations	92
	5.5	Summary	93
		REFERENCES	94

LIST OF TABLES

Table	Page
Table 2.1: Comparison between the image feature extraction techniques	18
Table 2.2: Comparison between classification techniques	22
Table 2.3: Comparison between FVC 2004 database, NIST Special Database 4 and LivDet2015 database	24
Table 2.4: Comparison between YALE database, ORL face database, FERET database, and Labeled Faces in the Wild (LFW)	25
Table 2.5: Comparison between CNN architectures	29
Table 3.1: Confusion matrix characteristics	52
Table 4.1: Network property comparison	62
Table 4.2: Comparison of hyperparameters between networks	64
Table 4.3: Comparison of hyperparameters between various settings of face network	71
Table 4.4: Comparison of hyperparameters between various settings of fingerprint network	72
Table 5. 1: Comparison of final accuracy between previous and proposed work	90
Table 5. 2: Objective and accomplishments of the proposed work.	91

LIST OF FIGURES

Figure	Page
Figure 1.1: The multimodal person recognition system's overall structure [1]	2
Figure 1.2: Convolutional Neural Network.....	3
Figure 2.1: The general architecture of multimodal person recognition system [1]	11
Figure 2.2: CNN for fingerprint recognition system [14]	13
Figure 2.3: The general layout of face recognition algorithm [17]	14
Figure 2.4: Ridge Ending & Bifurcation [22]	15
Figure 2.5: LBP operator in a 3×3 window	16
Figure 2.6: Layers in Neural Network	21
Figure 2.7: Convolutional Neural Network with three layers structure	26
Figure 2.8: AlexNet Architecture [44]	27
Figure 2.9: VGGNet Architecture	28
Figure 2.10: ResNet-50 Architecture [46]	28
Figure 2.11: Inception modules present in GoogLeNet [47]	29
Figure 2.12: The four different level of fusion in multimodal person recognition system	31
Figure 3.1: Agile model with 5 phases and iteration	36
Figure 3.2: Dataset classification diagram	39
Figure 3.3: Flowchart of dataset analysis	41
Figure 3.4: Flow of pretrained network analysis	42
Figure 3.5: Multimodal Person Recognition system's use case diagram	43
Figure 3.6: Multimodal Person Recognition system's sequence diagram	44
Figure 3.7: Multimodal Person Recognition System	45
Figure 3.8: General project flowchart	46
Figure 3.9: Flowchart of algorithm multimodal person recognition system	47
Figure 3.10: Implementation steps flowchart	48
Figure 3.11: MATLAB R2021b User Interface	49
Figure 3.12: Deep Network Designer User Interface	49

Figure 3.13: K-Fold Cross Validation Method	51
Figure 3.14: Gantt Chart of Final Year Project 1	53
Figure 3.15: Gantt Chart of Final Year Project 2	53
Figure 4.1: Multimodal Person Recognition System Interface	56
Figure 4.2: First half analysis of AlexNet network of face	57
Figure 4.3: Second half analysis of AlexNet network of face.....	57
Figure 4.4: First half analysis of AlexNet network of fingerprint.....	58
Figure 4.5: Second half analysis of AlexNet network of fingerprint	58
Figure 4.6: First half analysis of VGG16 network of face	59
Figure 4.7: Second half analysis of VGG16 network of face.....	59
Figure 4.8: First half analysis of VGG16 network of fingerprint.....	60
Figure 4.9: Second half analysis of VGG16 network of fingerprint	60
Figure 4.10: Common CNN architecture demonstration	61
Figure 4.11: Completed training of AlexNet for face	65
Figure 4.12: Completed training of AlexNet for fingerprint.....	65
Figure 4.13: Completed training of VGG16 for face	66
Figure 4.14: Completed training of VGG16 for fingerprint.....	66
Figure 4.15: Training Setting 1 of AlexNet: Face Network.....	68
Figure 4.16: Training Setting 2 of AlexNet: Face Network.....	69
Figure 4.17: Training Setting 3 of AlexNet: Face Network.....	69
Figure 4.18: Training Setting 1 of AlexNet: Fingerprint Network	70
Figure 4.19: Training Setting 2 of AlexNet: Fingerprint Network	70
Figure 4.20: Training Setting 3 of AlexNet: Fingerprint Network	71
Figure 4.21: Successful recognition through decision-level fusion	74
Figure 4.22: Failed recognition through decision-level fusion	74
Figure 4.23: Face and fingerprint images of Jolie being tested through system interface	75
Figure 4.24: Face and fingerprint images of MorganFreeman being tested through system interface.....	76
Figure 4.25: Face and fingerprint images of Rihanna being tested through system interface.....	76
Figure 4.26: Face and fingerprint images of ShahRukhKhan being tested through system interface.....	77

Figure 4.27: MorganFreeman face image test wearing sunglasses	78
Figure 4.28: ShahRukhKhan face image test containing unclassified object	78
Figure 4.29: Rihanna face image test with a bright illumination	79
Figure 4.30: Jolie fingerprint image test of incomplete fingerprint structure	79
Figure 4.31: ShahRukhKhan fingerprint image test containing noise	80
Figure 4.32: Successful person recognition of Jolie.....	81
Figure 4. 33: Failure to recognize person.....	81
Figure 4.37: 5th iteration of K-Fold Cross Validation test for face	82
Figure 4.38: 5th iteration of K-Fold Cross Validation test for fingerprint.....	83
Figure 4.34: Confusion Matrix of AlexNet for Face	84
Figure 4.35: Confusion Matrix of AlexNet for Fingerprint	85
Figure 4.36: Confusion Matrix of AlexNet for both face & fingerprint networks combined	86

LIST OF ABBREVIATIONS

2D	–	Two-Dimensional
3D	–	Three-Dimensional
ANN	–	Artificial Neural Network
CNN	–	Convolutional Neural Network
FC	–	Fully Connected
FYP	–	Final Year Project
GPU	–	Graphics Processing Unit
GUI	–	Graphical User Interface
IEEE	–	Institute of Electrical and Electronics Engineers
LBP	–	Local Binary Pattern
MATLAB	–	Matrix Laboratory
NN	–	Neural Network
PIN	–	Personal Identification Number
PNN	–	Probabilistic Neural Network
PCA	–	Principal Component Analysis
ReLU	–	Rectified Linear Unit
ResNet	–	Residual Network
SVM	–	Support Vector Machine
VGG16	–	Visual Geometry Group 16
VGGNet	–	Visual Geometry Group Network

Chapter 1

INTRODUCTION

1.1 Background

Since the dawn of modern computer application, the concepts of digital authenticators and person recognition system have been a fundamental part of digital security evolution. Secure access has been enabled in many applications, such as smartphone, banks, and government related procedures due to security related policies and reasons to preserve information and details of a personal as per law implementation. In conjunction with this, person recognition system was introduced during industrial revolution and further elevated due to its prominence for users and by users. In simple words, person recognition system is an application used to recognise and identify a person using certain authentication techniques.

In the older norms, passwords, PIN, and patterns were used as recognition technique in unlocking digital devices. However, existing person recognition system was upgraded with the use human physical traits such as fingerprints of human that is also known as biometric authentication. The biometric were later alternated with physical traits of human such as face, iris, and voice. The existing person recognition systems were labelled as unimodal recognition system due to its singular use of human biometric. The existing unimodal systems are widely utilised in commercial applications, government administrations, and forensic applications such as biometric authentication on mobile phones, national ID cards, and identifying criminals respectively [1].

Unimodal systems were received greatly by many parties; however, the technology evolution has made the system weak, the unimodal biometric system fails due to several flaws and limitations [1]. The biometric(s) such as fingerprint and voice were easily imitated by offenders. Even though the computer and software can decipher the biometric, but it was easily exploited with third party tool for identity bypass related purposes. The implementation of unimodal system could not withstand duplicates due to only single

biometric use, and security breaches and scams rise. This is when multimodal person recognition system was introduced.

A multimodal recognition system integrates two or more biometric features with the raw data of the multiple biometrics traits [1]. The multimodal person recognition works similarly to unimodal system, but it has two or more input data and the algorithm developed has more precision. Biometric identification can be either one-dimensional or multi-dimensional. Thus, combining various sources of biometric data creates a multibiometric recognition system that could overcome certain weakness exists in unimodal systems [2]. Multimodal systems are greater due to high performance, greater robustness, and higher resistance to spoofing [3]. In current generation, the multimodal person recognition systems are vastly used. Thus, a multimodal system that uses face and fingerprint could play a major role in recognising patterns and image processing. Facial and fingerprint recognition technology are the techniques used to identify an individual based on their facial characteristics and fingerprints respectively. In identity verification, facial recognition is the most inherent and accepted method [3], whereas fingerprint has great accuracy and user friendly. A general multimodal person recognition system has the following architecture as depicted in Figure 1.1.

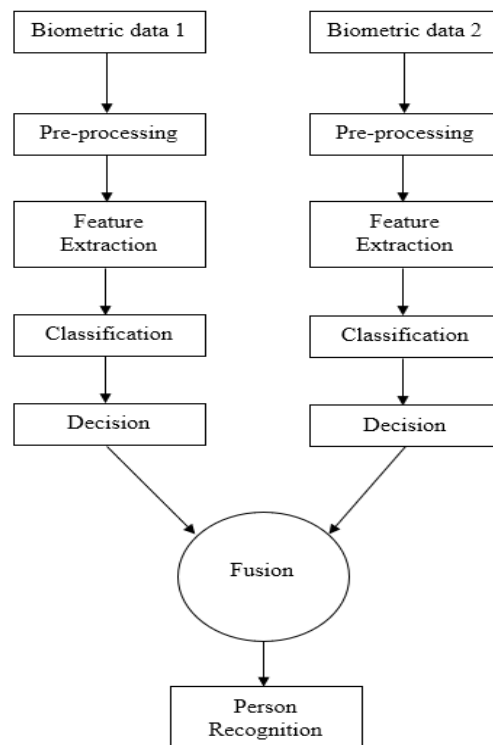


Figure 1.1: The multimodal person recognition system's overall structure [1]

In general, the basic unimodal recognition system using neural network has three stages known as pre-processing, feature extraction and classification. However, based on Figure 1.1, it can be seen a person recognition system with two biometrics can be identified through two separate detection and feature extraction for face and fingerprint respectively, moving onto classification and then fused together in through implying suitable fusion technique. The features are classified from picture sequences or static images, then the person is recognised as an output. In last few years' time, deep learning approaches have been implemented to summarise such complicated and technical procedures to form an algorithm.

Lately, convolutional neural network (CNN) which is the subset of deep learning method is used to construct multimodal recognition system [2]. In contrast to ordinary neural networks, image-based neural networks use pixels of an input image [3]. In this paper, the aim to implement a multimodal person recognition system by applying deep learning approach that could analyse pixel input for image processing and recognition.

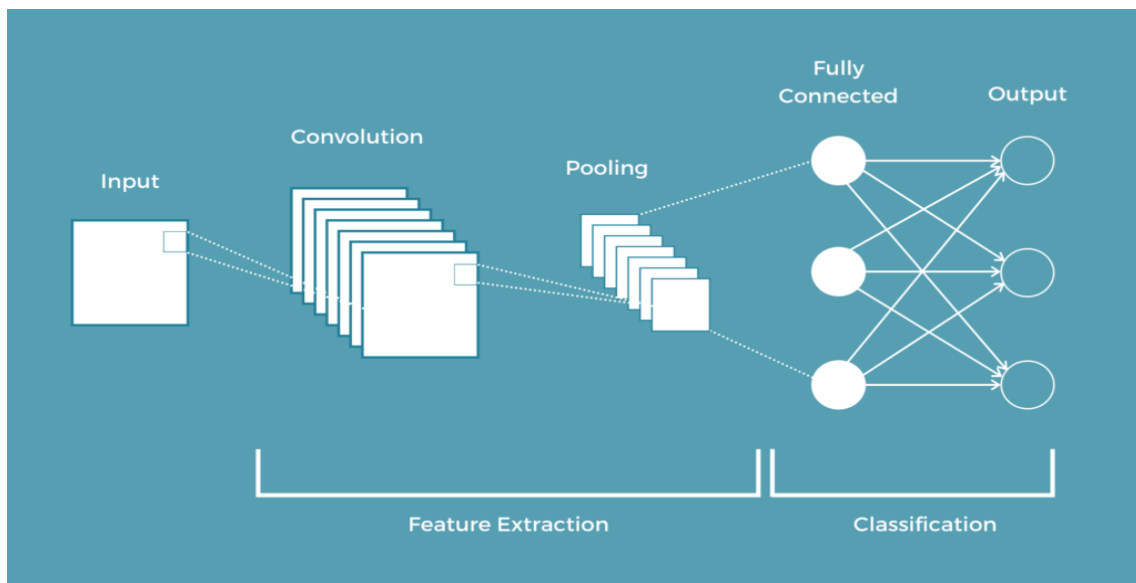


Figure 1.2: Convolutional Neural Network

In this research, a multimodal person recognition system based on computer and sophisticated technique of CNN, will evaluate faces and fingerprints in photos through mathematical algorithms. The artificial neural network is invented with numerous levels or layers, each of which can include one or more neurons [3], and uses convolution to extract characteristics from pictures while keeping spatial correlation [4].

1.2 Problem Statement

Recognising people for authentication purpose are important and needs accuracy for true recognition. Multiple types of recognition system exist in current electronic driven century involving traditional authenticator, digital authenticator, and systems through human-computer interaction (HCI). However, there is always a different accuracy between digital authenticator based on physiological or behavioural characteristics and conventional ways of passwords, signature, and fingerprint.

People are still using common authenticator such as password, pin number and pattern recognition. For example, computer logins are still highly dependent on password and pin number. The credentials are easily guessed, weak and does not provide sufficient protection against vulnerabilities. The vast growth of technology introduced the biometric recognition system that involves fingerprint, voice, and iris. The government and any legal procedure that requires identity verification are still highly using fingerprint as recognition system.

Problem arises when common authenticators and biometrics such as fingerprint and voice that computers and software can decipher are being exploited for use of identity bypass related purposes. What are some of the ways that developing a person recognition system might benefit a variety of sectors? Firstly, a recognition system built with convolutional neural network enhances the security measures through automated recognition with increased efficiency and accuracy due to complex nature of the network making it achieve sophistication through network training. Even with fast and easy recognition systems available, the necessity for well-secured identity and personal recognition technology is increasingly obvious as the number of security breaches and scams rises.

Next, the algorithm developed avoids imitations of a personal's features such as fingerprint, signature, and voice that are easily imitated through implementation of unimodal biometric systems. Unimodal biometric system only has single metric of accuracy delivered where it can vary from low to high, making the system less dependable. Thus, through multimodal recognition, the individuality of a person makes it difficult to discover another person with similar traits that would match your digital print and have second metric of accuracy that could compensate if the first biometric could not produce high accuracy. It is also important to create a multimodal system without

complicated structures through integration of various techniques. This is due to available multimodal person recognition that is not completely based on deep learning are built through various integration of separately introduced third party techniques. The architecture varies widely which has an occurrence of relatively different recognition rates and inconsistent accuracy that could be enhanced through deep learning.

1.3 Objectives

The main goal of this study is to create an algorithm that can recognise person using multimodal biometrics. The following research aims will help to achieve this goal:

- To investigate the existing works related to person recognition system
- To develop an algorithm of multimodal person recognition system using convolutional neural network
- To validate the algorithm and compare to an existing person recognition system

1.4 Project Motivation

The motivation of the project is to develop a multimodal person recognition system using deep learning method by eliminating the problems faced in existing algorithms. Besides, the project motivates to solve recognition issues faced by users in a daily life use of available recognition systems widely. Throughout the project, it is important to understand a user's perspective while implementing an algorithm that is better than previous works in order future works are based on this project's evolution.

1.5 Project Scope

The overall project is about a multimodal person recognition system that can be used by personals associated for secure access requiring personal identification in various segments and sectors. The ability of the system is to recall information or a person's characteristics in order to later identify the individual, who has saved their information in the database. This multimodal person recognition system uses two physical biometric traits of a human, thus face and fingerprint are intended to be included in this system expansion. The algorithm is developed using convolutional neural network, a deep learning approach with the use of MATLAB and C++ programming language.

The project is expected to provide various outcomes, one such is the extensive literature study. Literature study is important to any research and this context all the related works from 2017-2021 is thoroughly studied and reviewed to provide a bigger picture of all aspects involved in this project. Through algorithm design and implementation, an algorithm for the multimodal person recognition system is expected. As mentioned earlier, the algorithm is on the grounds of deep learning approach with complex computational techniques. The algorithm must be able to extract features from inserted image accurately and matched with the system's database. Moreover, the efficiency is prioritised too, where it should work better than existing available algorithms.

Next, there are certain acceptance criteria to be ticked before the project is declared successful. Firstly, the algorithm could detect the biometric inputs, such as face and fingerprint then extract their features. This is very crucial as without achieving the basic function of the algorithm, the project will not progress to next category of algorithm development. The algorithm produced must has higher recognition rate in term of accuracy and duration. Accuracy is highly anticipated for this recognition system as it acts as the core of the system, then moving on duration, the recognition process should only take seconds. This recognition duration includes time of input, system processing time and the time to display recognition result. The multimodal person recognition system should be user-friendly, so that users are convenient to use anytime and anywhere if required without much of hassle.

A project will have many exclusions to avoid inefficient methods and mistakes highlighted through literature review. In this case, the development of algorithm did not implement traditional authenticators. The authenticators such as pin and password are outdated for current industrial revolution, instead human physical traits are integrated for much secure and reliable authentication. Next, the existence of neural network is widely exploited, and initial recognition systems use complicated neural network structure. There were combinations of multiple techniques of pre-processing, feature extraction, and classification. However, this issue is not expected to recur in this project due to advanced neural networks that are simplified and open-ended.

Finally, as the project's design and implementation are expected to take off, there are few constraints to be faced and requires a fix. The algorithm requires face and fingerprint dataset for network training and test, however there are limited resources in the online