

UNIVERSITI TEKNIKAL MALAYSIA MELAKA

A STUDY OF HANDWRITTEN TEXT CHARACTER RECOGNITION USING NEURAL NETWORK



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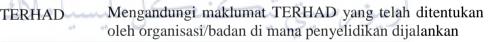
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APPROVAL

This report is submitted to the Faculty of Electrical and Electronic Engineering Technology of Universiti Teknikal Malaysia Melaka (UTeM) as a partial fulfilment of the requirements for the degree of Bachelor of Electronic Engineering Technology (Telecommunication) with Honours. The member of the supervisory is as follow:



ABSTRAK

Projek tajuk ini mencadangkan pendekatan yang efisien terhadap pengembangan sistem pengenalan teks tulisan tangan menggunakan Neural Network. Oleh kerana ketidakkonsistenan dalam gaya tulisan tangan, sistem pengenalan tulisan tangan yang canggih sering kali gagal memberikan prestasi yang memuaskan pada pelbagai jenis sampel tulisan tangan. Pendekatan yang tersedia untuk pengenalan tulisan tangan biasanya terdiri daripada pelbagai langkah yang merangkumi terutamanya 1.Prepemprosesan, 2. Pengekstrakan ciri, 3. Pengelasan dan 4. Pemprosesan pasca. Objektif utama kajian ini adalah untuk mengembangkan sistem pengenalan watak dan nombor tulisan tangan yang cekap untuk watak Inggeris berdasarkan Artificial Neural Network. Aksara tulisan tangan mungkin mengandungi huruf besar (huruf besar dan huruf kecil) watak Inggeris dengan 52 kelas (26 untuk huruf besar dan 26 untuk huruf kecil) termasuk dalam kajian ini untuk klasifikasi. Perisian pengiraan MATLAB dengan Image Processing Toolbox dan Neural Network Toolbox akan digunakan untuk menyelesaikan masalah pengecaman watak tulisan tangan yang ditentukan dari klasifikasi. Projek ini berfungsi untuk mengenali semua watak (Bahasa Inggeris) yang disediakan sebagai gambar input. Sekiranya input imej watak diberikan kepada program yang dicadangkan, watak input yang diberikan dalam gambar akan dikenali. Neural Network melakukan pengecaman dan pengelasan watak.

ونيوم سيتي تيكنيكل مليسيا ملاك

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ABSTRACT

This title project uses Neural Network as a method to develop recognition systems for handwritten text characters. The handwriting recognition systems frequently fail or unable to give sufficient results on different types of handwriting due to massive inconsistency styles of handwriting. Handwriting recognition systems comprise of Preprocessing, Segmentation, Feature Extraction and Classification. The main goal of this project is to propose a framework of text character recognition algorithm using Neural Network. The classification comprises 52 classes of English handwritten characters with 26 characters for capital letters and 26 characters for small letters. The handwritten character recognition will be using MATLAB software with Image Processing and Neural Network Toolbox. This project serves to recognize all characters (English) provided as input image. If the character image input is given to the proposed program, the input character given in the image will be recognized. Neural Network does recognition and classification of characters.

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DEDICATION

I dedicate this project to My Creator Allah S.W.T, my strong pillar, my source of inspiration, wisdom, knowledge and understanding. He has been the source of my strength throughout this project and on His wings only have I soared. I also dedicate this work to families; Hassan bin Mat Lihim, my dad and Zainab binti Ismail, my mom, who has encouraged me all the way and whose encouragement has made sure that I give it all it takes to finish that which I have started. To my sister, Ummi Umairah (Adik), and fellow friends who have been affected in every way possible by this quest. This project I also dedicate to my supervisor, Ts. Dr. Rostam Affendi bin Hamzah for his help to manage me to the accomplishment of undertaking for my degree. Thank you. My love for you all can never be quantified. Allah bless you.

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LIST OF SYMBOLS

g(x,y)	-	Grayscale Image
R(x,y)	-	Red Component of the Colour Image
G(x,y)	-	Green Component of the Colour Image
B(x,y)	-	Blue Component of the Colour Image
w ⁱ	-	Weight Vector of ith Layer
a ⁱ	-	Output of i th Layer
b ⁱ	-	Bias Vector for i th Layer



LIST OF ABBREVIATIONS

ANN	-	Artificial Neural Network
ACSII	-	American Standard Code for Information Interchange
OCR	-	Optical Character Recognition
ICR	-	Intelligent Character Recognition
RNN	-	Recurrent Neural Network
PDA	-	Personal Digital Assistant
LCD	-	Liquid Crystal Display
SVM	MALAYS	Support Vector Machine
MLP	and -	Multi Layered Perceptron
CPU	TEK	Central Processing Unit
MNIST	Far -	Modified National Institute of Standards and Technology
	Ainn	Dataset
SVHN	سيا ملاك	Street View House Numbers Dataset
SVM-RB	UNIVERSI	Support Vector Machine Radial Basis Function
HCCR	-	Handwritten Chinese Character Recognition
ICDAR	-	International Conference on Document Analysis and
		Recognition
CNN	-	Convolutional Neural Network
LSTM	-	Long Short Term Memory Network
MLP-BP	-	Multilayer Perceptron Network using Backpropagation
		Algorithm

MLP-LM	-	Multilayer Perceptron Network using Lavenberg-Marquadt
		Algorithm
MDC	-	Minimum Distance Classifier
ANESP	-	Automated Numeral Extraction And Segmentation Program
k-NN	-	k-Nearest Neighbour
ML	-	Machine Learning
DT	-	Decision Tree
HMM	-	Hidden Markov Model
LM	-	Language Model
HOG	MALAY	Histogram of the Oriented Gradient
MKL	- ⁻	Multi Kernal Learning
HTML		Hypertext Markup Language
TIFF	Flore -	Tagged Image File Format
TXT	del (Text File
RTF	سيا ملاك	Rich Text Format
PDF	JNIVERS	Portable Document Format YSIA MELAKA
ROI	-	Region Of Interest
ICC	-	International Color Consortium Profile
DICOM	-	Digital Imaging And Communication in Medicine
LVQ	-	Learning Vector Quantization
NARX	-	Nonlinear Autoagressive
GPU	-	Graphics Processing Unit

CHAPTER 1

INTRODUCTION

1.1 Introduction

This chapter will further unfold a recap of this project regarding Handwritten Text Character Recognition using Neural Network. This chapter will review about background, statement of problem, objective, scope, and project significance. The thesis outline is mentioned at the end of this section.

1.2 Background

Handwritten Character recognition is an art of detection, segmentation, and identification of characters of image (Mori, S. et al, 1992, pp. 1029-1058). This hand-written character recognition enables human reading capabilities to be simulated thus the computer can interpret the input and work with text as human (Arica, N. & Yarman-Vural, F.T., 2001, pp. 216-233). According to (Mohapatra, H., 2009), the identification of handwriting has been one of the most important and difficult groundwork areas in recent years in the field of image processing and pattern recognition. It makes an important contribution to the automation process development, and in many applications strengthens the interface between man and machine. The main goal of this project is to propose a framework of text character recognition algorithm using Neural Network. To put it another way, a recognition using the Artificial Neural Network (ANN) approach can functionally identify a type of particular character.

Neural computing is a comparatively new area and therefore the design components are barely well defined compared to other frameworks. Data Parallelism applies to neural computers. Neural computers operate in a manner entirely different from standard computer performance. Neural computers are not programmed but trained to provide a certain starting state of data input, either classify the input data into one of the class numbers or cause the original data to evolve in a way that enhances some preferable property.

Recognition of characters consists of two types; online and offline. In online recognition, data is gathered in online character recognition during the writing process with the aid of a special pen on the electronic surface. Pre-written data written on a sheet of paper is scanned for offline recognition. Online approaches have been proven to be an expert on the identification of handwritten characters by their off-line counterparts, due to the temporal knowledge available to the former. According to (Bhattacharya, U. & Chaudhuri, B.B., 2008, pp. 444-457), the level of accuracy in recognition in off-line systems is comparably high. The handwritten characters include for classification in this analysis is a mix case (capital and small letters) of English characters consisting of 52 groups (26 for capital letters and 26 for smaller letters).

Based on Figure 1.1, recognition of characters consists of two sections; that is, section of training and test. The handwritten character images are scanned in both the training and testing section, followed by pre-processing process which includes noise filtering, smoothing, and normalization. Image rendering is ideal for segmentation in which the image is decomposed into sub-images. Meanwhile, Feature Extraction functions in improving identification rate and misclassification. The finishing process of the reconnaissance system is the simulation of the trained classifier. It produces

recognized representative characters by measuring equal of standardized ASCII character type, which means the Test Sample Recognition Index will be used.

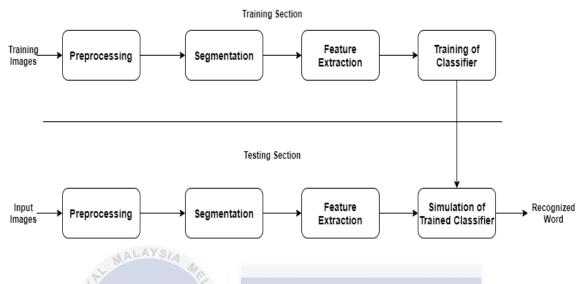


Figure 1.1 : Generic Character Recognition Model

Recently, electronic handwritten recognition has drawn the attention of researchers worldwide due to the increased use of handheld devices. Nowadays, many applications have included offline handwriting recognition systems, including mail sorting, bank processing, document reading and recognition of postal addresses. Consequently, the recognition of off-line handwriting continue to be an active area of research to explore the latest techniques that would improve recognition accuracy (Pal, U. et al, 2007, pp. 749-753) (Pal, U. & Chaudhuri, B.B., 2004, pp. 1887-1899).