

MALAY ARTICULATION SYSTEM FOR EARLY SCREENING DIAGNOSTIC
USING HIDDEN MARKOV MODEL AND GENETIC ALGORITHM

MOHD NIZAM BIN MAZENAN

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Dedicated to Papa and Mama,
my Sister (Mimi) and Brother (Agung)

Haziq and Maisarah
and my beloved family.

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ABSTRACT

Speech recognition is an important technology and can be used as a great aid for individuals with sight or hearing disabilities today. There are extensive research interest and development in this area for over the past decades. However, the prospect in Malaysia regarding the usage and exposure is still immature even though there is demand from the medical and healthcare sector. The aim of this research is to assess the quality and the impact of using computerized method for early screening of speech articulation disorder among Malaysian such as the omission, substitution, addition and distortion in their speech. In this study, the statistical probabilistic approach using Hidden Markov Model (HMM) has been adopted with newly designed Malay corpus for articulation disorder case following the SAMPA and IPA guidelines. Improvement is made at the front-end processing for feature vector selection by applying the silence region calibration algorithm for start and end point detection. The classifier had also been modified significantly by incorporating Viterbi search with Genetic Algorithm (GA) to obtain high accuracy in recognition result and for lexical unit classification. The results were evaluated by following National Institute of Standards and Technology (NIST) benchmarking. Based on the test, it shows that the recognition accuracy has been improved by 30% to 40% using Genetic Algorithm technique compared with conventional technique. A new corpus had been built with verification and justification from the medical expert in this study. In conclusion, computerized method for early screening can ease human effort in tackling speech disorders and the proposed Genetic Algorithm technique has been proven to improve the recognition performance in terms of search and classification task.

ABSTRAK

Pada masa kini, aplikasi pengecaman pertuturan sudah menjadi tidak asing lagi sebagai satu teknologi yang mampu membantu individu kurang upaya. Terdapat kajian yang meluas dan pembangunan yang pesat telah dijalankan sejak beberapa dekad yang lalu. Namun begitu, bagi prospek kajian di Malaysia, ianya masih baru dan belum matang dari aspek penggunaan dan pendedahan walaupun terdapat permintaan yang tinggi daripada sektor perubatan dan penjagaan kesihatan. Tujuan utama kajian ini adalah untuk menilai kualiti dan kesan penggunaan kaedah komputeran bagi tujuan peringkat awal diagnosis bagi mengatasi masalah gangguan pertuturan artikulasi untuk konteks di Malaysia. Di dalam kajian ini, kaedah kebarangkalian statistik menggunakan Hidden Markov Model (HMM) telah diadaptasi dengan reka bentuk corpus bahasa Melayu baru untuk permasalahan artikulasi berpandukan skrip SAMPA dan garis panduan yang telah ditetapkan oleh IPA bersama-sama pengesanan daripada terapis pertuturan di Hospital. Penambahbaikan telah dilakukan bermula dari peringkat pemprosesan bahagian depan bagi pemilihan vektor ciri dengan menerapkan algoritma kalibrasi kawasan tanpa isyarat suara untuk pengesanan titik permulaan dan akhir. Pengkelas yang digunakan turut ditambah baik dengan menggunakan teknik penghibridan carian Viterbi dengan genetik algoritma (GA) bagi memperoleh ketepatan yang tinggi dalam keputusan pengecaman dan juga proses klasifikasi unit leksikal. Keputusan kajian dinilai berpandukan tanda aras oleh NIST. Keputusan kajian menunjukkan ketepatan pengecaman telah meningkat kira-kira 30% hingga 40% dengan menggunakan teknik hibridisasi Viterbi dan genetik algoritma berbanding teknik konvensional. Kesimpulannya, pembentukan corpus baru telah berjaya dibina dengan pengesanan dan justifikasi oleh pakar perubatan. Oleh yang demikian, penggunaan komputer untuk pemeriksaan awal, dapat meringankan usaha manusia dalam diagnosis gangguan pertuturan dan teknik genetik algoritma yang diperkenalkan berjaya meningkatkan prestasi pengecaman dalam aspek tugas pencarian dan pengelasan.

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LIST OF ABBREVIATIONS

AAC	-	Alternatives And Augmentative Communication
AI	-	Artificial Intelligence
ASR	-	Automatic Speech Recognition
FE	-	Feature Extraction
FFT	-	Fast Fourier Transform
FSM	-	Finite State Machine
GA	-	Genetic Algorithm
GMM	-	Gaussian Mixture Model
HMM	-	Hidden Markov Model
HSA	-	Hospital Sultanah Aminah
IPA	-	International Phonetic Association
SAMPA	-	Speech Assessment Methods Phonetic Alphabet
SKPK	-	Sekolah Kebangsaan Pendidikan Khas
STE	-	Short-Time Energy
KKM	-	Kementerian Kesihatan Malaysia
KPM	-	Kementerian Pengajian Malaysia
LPC	-	Linear Predictive Coding
MFCC	-	Mel-Frequency Cepstral Coefficients
NIST	-	National Institute of Standards and Technology
OOV	-	Out-of-Vocabulary
SLP	-	Speech Language Pathologists
ST	-	Speech Therapist
STT	-	Speech-to-Text
WER	-	Word Error Rate
WHO	-	World Health Organization
ZCR	-	Zero Crossing Rate

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CHAPTER 1

INTRODUCTION

1.1 Introduction

Individuals who are diagnosed with speech disorder have problems creating or forming speech sounds to communicate with others. This may lead to several speech impairments such as articulation disorders, fluency disorders and voice disorders. Statistic by World Health Organization (WHO) shows that, speech and language disorder affects at least 3.5% of human population communication skills. In Malaysia, its affect 5-10% of children with manifest speech and language problems totaling about 10,223 children were reported to have learning disabilities (KPM 2002, Woo and Teoh 2007, Rafflesia 2008). This research is written to highlight and focus more on the main issue in medical speech that is related to early screening of articulation disorder case. Previous research shows that, early speech therapy intervention is critically important as well as other developmental concerns as a part of knowing what's "normal" and what's "not" in speech and language development. This can help parents or guidance especially Speech Therapist (ST) to figure out the next concern on right schedule and to prevent possible future problems (Fey 1986, Nelson 1989, Warren and Yoder 1996, Trevino-Zimmerman 2008, Navarro-Newball, Loaiza et al. 2014, Rubin, Kurniawan et al. 2014, Sparreboom, Langereis et al. 2015).

Articulation disorders are errors in the production of speech sounds that may be related to anatomical or physiological limitations in the skeletal, muscular, or neuromuscular support for speech production (Gargiulo 2006) which may also interfere with intelligibility (Association 1993). These disorders include omissions (bo for boat), substitutions (wabbit for rabbit), additions (animamal for animal) and

distortions (shlip for sip) (Bauman-Waengler 2012). The degrees of disorders vary from mild impairments like pronunciation errors to more severe ones including hearing-loss, aphasia, and craniofacial anomalies. Some children or even adults, have trouble saying certain words, sounds or sentences where others have trouble understanding what they are trying to say. However, there are special therapies such as Speech Therapist (ST) and Speech Language Pathology (SLP) that can help these patients to go through the early screening diagnosis, assessment, and therapeutic assistance to help this patient to improve their speech.

The advanced research in speech and signal processing technology has open massive development of computer-assisted methods for speech therapy or called assistive device (AD). Speech-to-Text recognition system can be applied to help people who have speech impairment. It is to assist the speech therapist in diagnosing the patient. This idea has been well mentioned where the Automatic Speech Recognition (ASR) systems may be used for therapeutic applications, assessment applications, report writing and as a mode of alternatives and augmentative communication (AAC) (Griffin, Wilson et al. 2000). AD has widely been used for much therapeutic assistance for ST and few research in this area in Malaysia include Malay speech therapy assistance tools (MSTAT) (Tan, Ariff et al. 2007) and Computer-based Malay speech analysis system (Liboh 2011). Communication disorder by nature is very complex and those existing diagnostic tool is still incomplete and imprecise which need a lot more improvement. The research will focus more on the design and the development of Malay articulation disorder early screening system to assist in speech communication by using speech recognition engine that meet specific criteria. Speech therapy early screening diagnostic has been believed to help ST by doing the quick assessment to verify the correct level of disorder of the speech disorder patient (Misman 2014).

Speech-to-Text or often called speech recognition is a translation of spoken words into text (Rabiner and Juang 1993). Speech signal processing includes the acquisition, manipulation, storage, transfer and output of human vocal utterances by a computer (Gurjar and Ladhake 2008). Speech feature extraction work in front end and play an important role in the conversion and extraction of useful meaning from speech. The extracted speech features are used as input for speech recognition.

Feature extraction is a general term for methods of constructing combinations of the variables to get around these problems in manipulation of the collected speech vector while still describing the data with sufficient accuracy. To produce the right speech recognition system specifically as a speech diagnosis tool that covered articulation case, there are few criteria need to be met such as the design of the Malay corpus, segmentation of speech utterance to specific lexical unit, and the robust pattern matching algorithm to spot and classifies the error in the speech utterance sample. Malay corpus-based has been designed with the collaboration of ST at Hospital Sultanah Aminah (HSA) and Sekolah Kebangsaan Pendidikan Khas (SKPK) whereby the process include mapping the word list into table of phoneme and manner of articulation including the pictogram rule specified by ST worldwide (Bell and Lindamood 1991, Gambrell and Jawitz 1993). This method of designing corpus will be based on human sound production and specifically to cover the process on human speech. Another aspect is the segmentation method of the speech utterance smallest lexical unit where most of the process been applied using Hidden Markov Model (HMM)-based speech recognition (Rabiner 1989). Following this method such as wavelet spectrum method (Ziółko, Manandhar et al. 2006) and segment features approach (Rybach, Gollan et al. 2009) which is the process to analyze and validate a computational model for the speech feature. The main purpose for the segmentation in speech recognition is to identify the boundaries between words, syllables, or phonemes in spoken natural language. Since speech utterance will be recognized by phoneme classifier, this utterance need to be segmentized into the smallest lexical unit to achieve higher recognition output. Back-end classification is another important process in this research which include acoustic modeling, pronunciation dictionary and language model as the ASR components. The statistical methods based on HMM is to produce a model from the speech training data which later on be use to predicts the target values of the speech testing data (Suuny, David Peter et al. 2013) using best pattern matching algorithm as the classifier which is treated as a search problem in this research.

1.2 Background of the Problem

Computer-based Malay speech early diagnostic system is still new in Malaysia and most of the systems available now are in foreign language such as English (Ting, Yunus et al. 2003). Most of the diagnostic system is not using the corpus specifically designed and verified by expert in this area such as ST. There is no specific or standard formatting in creating the corpus whether it is for early screening or diagnosis, treatment or training. All current wordlist design used by ST (especially in HSA and SKPK) are using the old wordlist developed by Ministry of Health Malaysia (KKM 2014, Mismam 2014) in which most of the words may not be accurate enough to be used. Therefore the development of the corpus will follow specific rule to address articulation disorder case based on verification by ST and SLP. This is important aspect as the main purposed is to overcome the problem regarding articulation disorder early screening diagnosis. Another important aspect that may lead to the improvement in development of high quality speech-to-text system which is the segmentation of the acoustic utterance selected from the large-scaled corpus of isolated words that been specifically designed to fit Malay language that cover wide range of articulation words. Thus, a robust segmentation is needed to handle this sample by considering the aspect of right FE technique and the improvement of silence region calibration algorithm (Bhattacharjee 2013, Shanthi Therese and Lingam 2013). The effect of ignoring this aspect may lead to problem where many short silence tags are unreliable and hence lead to merging with neighboring segments (Hain, Johnson et al. 1998) which may reduce the recognition accuracy. After the useful feature has been extracted, segmentized and model, the pattern matching algorithm will take place as the speech recognizer to define each model to the right place according to the artic sample utterance by the patient of speech disorder.

1.3 Problem Statement

- i) ST is deals with speech disorder patient where the process of early screening diagnosis can be conducted by using computerized speech recognition system to provide a way to assess the speech problem of the patient (Glykas and Chytas 2004). Therefore, it is an issue here on how to create the right speech modeling that can recognizer speech disorder unknown utterance by achieving high accuracy of recognition speech.
- ii) The way of representing the speech output and transcriptions result must be effective to show the right disorder categories with met additional criteria such as quicken the process of therapy screening session but maintain the accuracy result. This is the most important element in a speech therapy program but the most difficult for a therapist to carry out, due to time limits (Bälter, Engwall et al. 2005). Therefore computational time of the system, accuracy, and intelligibility is an issue here.
- iii) The wordlist of the Malay speech corpus design must contain sufficient prosodic and acoustic variations with significant unit following the right design and rule for articulation disorder early diagnosis. But, the current Malay speech database design is lack of the verification from medical expert and some of the target selection word not following “stimulable” therapy word. Therefore, it is an issue to develop the robust corpus that is based on the interwoven ideas of developmental readiness, ease of learning and understanding and easier for the clinician to teach (Van Riper and Irwin 1958, Van Riper 1978, Hodson 2007).
- iv) Research been done in Malaysia shows that, this traditional method (hearing and experience judgment) is still in use widely in Malaysia hospital or speech center. There are no wrong doing by manual technique but based on observation to the hospital or clinic that still using this technique, the method sometimes lack of accuracy, time consuming (Ai and Yunus 2007) and require high number of ST for each session (Saz, Yin et al. 2009). The common practice of this process going through, the "tool" that only been rely by the ST is their hearing and experience judgment. That would go on for an hour and sometimes it would lead to error elimination during the process (Secord, Boyce et al. 2007).

1.4 Objectives of the Research

This research study is aiming to solve related problems in building Malay early screening computerized system for articulation disorder diagnosis. Therefore the objectives are:

- i) To develop high quality Malay articulation early screening computerized system (in term of providing higher speech recognition accuracy and robust corpus design) for articulation disorder diagnosis use as therapeutic assistance for ST.
- ii) To design a new set of Malay SAMPA database that specifically for computer-readable script that accommodate Malay articulation pronunciation phoneme applied in speech-to-text system.
- iii) To implement and improve acoustic phonetic segmentation in speech recognition boundaries identification between smallest lexical units in spoken natural language with silence region calibration algorithm.
- iv) To improve Viterbi searching algorithm by hybridization with Genetic Algorithm (GA) which work in back-end phase that provide optimal solution when solving artic lexical unit selection task.

1.5 Scope of the Research

The current corpus from Kementerian Kesihatan Malaysia (KKM) is based on Malay language for speech disorder diagnosis by manual technique that contains 108 target words with no specific or standard formatting in creating the corpus whether it's for early screening or diagnosis, treatment or training. A new speech database is built specifically for articulation disorder early screening followed specific rule in qualitative approach and target selection in phonological intervention. The target word has been simplified to 76 words or equivalent to 36480 training sampling sum up by $80 \times 76 \times 6$ which considering the total 80 patient with 6 times repetition for each word. The subject patient is from 2 main races in Malaysia (Malay and Chinese), with both male and female, ranging from normal and articulation disorder patient. The age coverage is starting from the age of 9 until 23 years old.

The statistical engine used in this research is based on HMM as the auto-segment of the recorded speech sample to cover three lexical units of phoneme, syllable and word based, and also work as probabilistic rule for the speech modeling phase. The speech recognition processing module includes the segmentation, silence region calibration, feature extractor (based on MFCC) and pattern matching engine. These methods are also includes Genetic Algorithm (GA), HMM probabilistic method and the hybridization of Viterbi search and GA. The evaluation of the recognition system was tested based on Word Error Rate (WER), Out-of-Vocabulary (OOV), %Correct and Recognition Rate accuracy which been used as objective evaluation of the recognition system based on NIST benchmarking. This can measure the recognition performance, vocabulary identification and intelligibility of the output speech utterances.

1.6 Significance of the Study

This study implements speech recognition technology with improvements of HMM as the speech engine, silence region calibration algorithm in acoustic segmentation phase and hybridization of Viterbi search with GA for the classification process. The articulation early screening case is often referred to the problem of producing speech sounds related to smallest lexical unit such as phoneme and syllable. therefore, the purposed of solving this case is by tackling the phonology dimension in term of articulatory phonetics (transmission of sounds) by following the linguistic guideline based on the table of placed and articulation manner for human speech production. It also essential to preoccupied sound as a system that carrying information where is the task is to identify and recognize the correct phonemes. The evaluations of the recognition system were tested based on values of objective functions obtained, accuracy of the recognition and comparison test. By doing so, the advantage and disadvantage of the improvement methods used are known. Besides, new algorithm of silence region calibration in segmentation phase and the hybridization of Viterbi search and GA in classification process can be maintained the advantageous of both method. This new algorithm is able to improve the performance and obtain good solution in lexical unit recognition process that

definitely makes contribution in the area of speech recognition specifically for articulation disorder early screening diagnosis.

1.7 Research Methodology

The design of Malay Speech-to-Text (STT) system is divided into three steps as shown in Figure 1.1. The system starts with first step of Malay speech corpus design or database development with specific rule verify by ST. The second step is the front-end processing component that process sampled speech signal into parametric representation by FE. This process also includes the silence region calibration, segmentation of lexical unit, and pattern modeling of the training data. The pattern of phoneme will be the templates for the next phase that was pattern recognition. The final step or back-end processing will deal with the classifier and decoder component for the recognition of unknown speech signal by the patient or normal speaker. The whole system is also been evaluated in this phase based on objective evaluation and also comparison test.

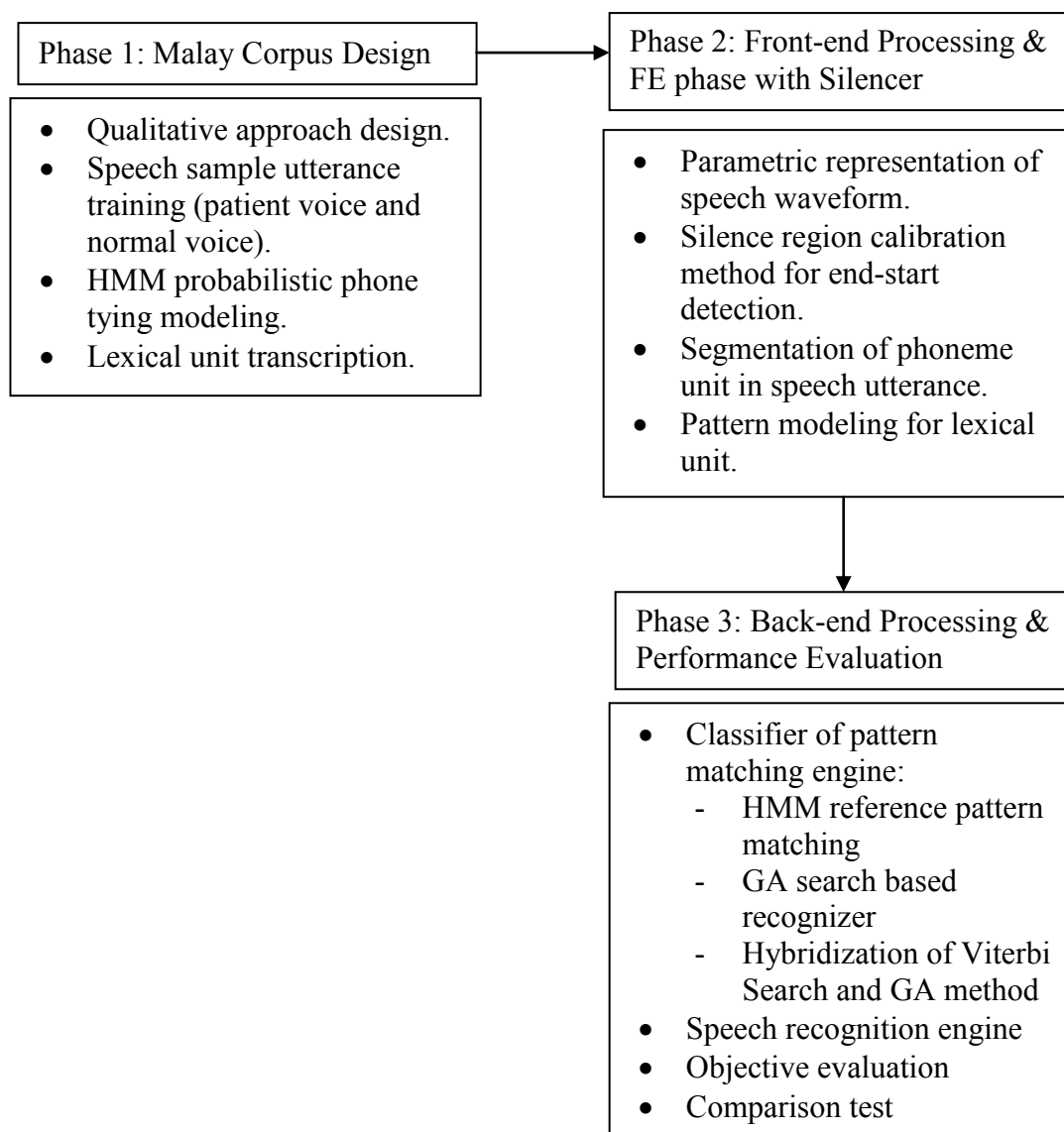


Figure 1.1. Mapping of procedures in this study

1.8 Thesis Layout

This thesis is divided into five chapters. Chapter 1 includes introduction, background of the problem, objectives of the research and scope of the thesis. The purpose is to show the initiative in using the proposed recognition method rather than current conventional or traditional method.

Chapter 2 will provide a literature review of this study. It includes reviews from previous study of the Text-to-Speech system, speech recognition and few methods in developing the recognizer machine. The focus is to develop the robust

corpus for early screening diagnosis system of articulation speech disorder, the segmentation of smallest lexical unit method and the pattern matching in classifier phase. This chapter also included review of standard method in feature extraction, segmentation technique, silence region calibration, HMM recognizer engine, Viterbi search and GA for pattern matching.

Chapter 3 is the methodology used in this study. It describes the Malay early screening diagnostic system and the implementation specifically for articulation speech disorder. This chapter discussed the overall implementation, design, and the development of Malay speech corpus, cost function used and objective evaluation of the target cost. It also describes the procedure to implement the silence region calibration algorithm before segmentation phase for the front-end. The performance will be compared with previous method (without silencer and with silencer). Then the segmentation will be done by using the correct MFCC adjustment for the training sample. The performance evaluation methods are based on the accuracy percentage of word spotting. Another method discussed is regarding the procedure to implement HMM recognizer emphasizing the Viterbi search as a classifier for reference pattern matching. Then the procedure of hybridization with GA as a search based recognizer for the pattern matching and word spotting in decoder. The performance will be compared with conventional method.

Chapter 4 is the result and discussion section. It shows the result accuracy achieved by the recognizer and also describes the performance evaluation of all method including corpus design testing, smallest lexical unit segmentation, silence region calibration, FE adjustment, and the pattern matching in recognition phase. Their performance will be based on the evaluation of the recognition system was tested based on Word Error Rate (WER), Out-of-Vocabulary (OOV), %Correct and Recognition Rate accuracy which been used as objective evaluation of the recognition system. The entire test is based on NIST standard for ASR system.

Chapter 5 outlined the conclusion for the proposed technique and method which reflect the novelties of the research. This chapter will also include the future plan and the recommendation for further improvement of this study.

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