

**AUTOMATIC TRANSCRIPTION AND PHONETIC LABELLING OF
DYSLEXIC CHILDREN'S READING IN BAHASA MELAYU**

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Abstrak

Pengecaman suara automatik (*ASR*) berpotensi untuk membantu kanak-kanak disleksia yang mengalami masalah pembelajaran. Kesalahan dalam penyebutan fonetik yang hampir sama oleh kanak-kanak disleksia amat tinggi sehingga memberi kesan kepada ketepatan pengecaman *ASR*. Oleh itu, objektif utama kajian ini adalah untuk menilai penerimaan ketepatan *ASR* dengan menggunakan transkripsi dan pelabelan fonetik automatik untuk kanak-kanak disleksia. Bagi mencapai matlamat utama tersebut, terdapat tiga objektif yang telah ditetapkan: pertama untuk menghasilkan transkripsi dan pelabelan fonetik manual; kedua untuk membina transkripsi dan pelabelan fonetik automatik menggunakan kaedah penajaran paksa; dan ketiga untuk membandingkan ketepatan di antara transkripsi dan pelabelan fonetik automatik dengan transkripsi dan pelabelan fonetik manual. Lantaran itu, untuk mencapai matlamat kajian ini beberapa kaedah telah digunakan, termasuk pelabelan ucapan dan segmentasi manual, penajaran paksa, *Hidden Markov Model (HMM)* dan Rangkaian Neural Buatan (*ANN*) untuk proses latihan, dan bagi mengukur ketepatan daripada *ASR*, Kadar Kesalahan Perkataan (*WER*) dan *False Alarm Rate* (*FAR*) digunakan. Sebanyak 585 fail ucapan telah digunakan untuk transkripsi manual, penajaran paksa dan juga proses latihan. Pengecaman yang dijana oleh *ASR* enjin yang menggunakan transkripsi dan pelabelan fonetik automatik telah mencapai keputusan yang paling optimum iaitu 76.04% dengan kadar *WER* serendah 23.96% dan *FAR* 17.9%. Keputusan ini adalah hampir sama dengan *ASR* enjin yang menggunakan transkripsi dan pelabelan fonetik manual iaitu 76.26%, *WER* serendah 23.97% dan *FAR* 17.9%. Kesimpulannya, ketepatan daripada transkripsi dan pelabelan fonetik automatik adalah diterima bagi membantu kanak-kanak disleksia belajar menggunakan *ASR* dalam Bahasa Melayu (BM).

Kata Kunci: Pembacaan kanak-kanak disleksia, Transkripsi manual, Transkripsi dan pelabelan fonetik automatik, Penajaran paksa, Pengukuran ketepatan *ASR* enjin.

Abstract

Automatic speech recognition (ASR) is potentially helpful for children who suffer from dyslexia. Highly phonetically similar errors of dyslexic children's reading affect the accuracy of ASR. Thus, this study aims to evaluate acceptable accuracy of ASR using automatic transcription and phonetic labelling of dyslexic children's reading in BM. For that, three objectives have been set: first to produce manual transcription and phonetic labelling; second to construct automatic transcription and phonetic labelling using forced alignment; and third to compare between accuracy using automatic transcription and phonetic labelling and manual transcription and phonetic labelling. Therefore, to accomplish these goals methods have been used including manual speech labelling and segmentation, forced alignment, Hidden Markov Model (HMM) and Artificial Neural Network (ANN) for training, and for measure accuracy of ASR, Word Error Rate (WER) and False Alarm Rate (FAR) were used. A number of 585 speech files are used for manual transcription, forced alignment and training experiment. The recognition ASR engine using automatic transcription and phonetic labelling obtained optimum results is 76.04% with WER as low as 23.96% and FAR is 17.9%. These results are almost similar with ASR engine using manual transcription namely 76.26%, WER as low as 23.97% and FAR a 17.9%. As conclusion, the accuracy of automatic transcription and phonetic labelling is acceptable to use it for help dyslexic children learning using ASR in Bahasa Melayu (BM).

Keywords: Dyslexic children's reading, Manual transcription, Automatic transcription and phonetic labelling, Forced alignment, Evaluation accuracy of ASR engine.

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List of Abbreviations

ANN	Artificial Neural Network
ART	Automatic Reading Tutor
ASCII	American Standard Code for Information Interchange
ASR	Automatic Speech Recognition
BM	Bahasa Melayu
C	Consonant
CV	Consonant Vowel
CALL	Computer-assisted language learning
CoLiT	Colorado Literacy Tutor
CSLU	Center for Spoken Language Understanding
FAR	False Alarm Rate
HMM	Hidden Markov Model
HTK	Hidden Markov Modelling Toolkit
IPA	International Phonetic Alphabet
IRT	Interactive Reading Tutor
LD	Learning Disability
MDR	Miscue Detection Rate
MS	Milliseconds
NN	Neural Network
SER	Sentences Error Rate
TTS	Text to Speech
V	Vowel
WER	Word Error Rate

CHAPTER ONE

INTRODUCTION

1.1 Introduction

Automatic speech recognition (ASR) has been an essential technology, and it has come to a stage where it has been actively applied in a lot of industrial and consumer applications. ASR research is still in early stage in Malaysia for Bahasa Melayu (BM). However, ASR can play an important role in the education field like to boost children's interest in learning. The availability of ASR technology gives opportunity to help children especially dyslexics to enhance their learning ability by using Automatic Reading Tutor (ART) or Interactive Reading Tutor (IRT). In order to develop ART and IRT using ASR technology, speech files of dyslexic children's reading aloud are used to perform transcription and phonetic labelling that serve as important basic elements for the construction of ASR engine (Athanaselis, Bakamidis, Dologlou, Argyriou, & Symvonis, 2014; Taileb, Al-Saggaf, Al-Ghamdi, Al-Zebaidi, & Al-Sahafi, 2013; Pedersen & Larsen, 2010; Husniza & Zulikha, 2009; Li, Deng, Ju, & Acero, 2008; Chuchiarini & Strik, 2003).

Since transcription and phonetic labelling are used for ASR engines, so the training and evaluation accuracy of it must be done by using standard methods and metrics (e.g. hybrids Hidden Markov Model (HMM) and Artificial Neural Network (ANN) for training; Word Error Rate (WER) and False Alarm Rate (FAR) for measuring accuracy). However, in this study the dyslexic children's speech presents a challenge to perform transcription and phonetic labelling due to dealing with highly

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