



UNIVERSITI PUTRA MALAYSIA

DESIGN OF INTELLIGENT QIRA'AT IDENTIFICATION ALGORITHM

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DESIGN OF INTELLIGENT QIRA'AT IDENTIFICATION ALGORITHM

By

NORAZIAHTULHIDAYU BINTI KAMARUDIN

**Thesis submitted to the School of Graduate Studies, Universiti Putra Malaysia,
in Fulfillment of the Requirements for the Degree of Doctor of Philosophy**

December 2017

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DEDICATION

Firstly, everything would not be possible without the support of the higher education ministry of Malaysia(MOE) for the financial support of my PhD study in Universiti Putra Malaysia(UPM)so, great thanks to both the MOE and UPM.



Abstract of thesis presented to the Senate of Universiti Putra Malaysia in fulfillment of the requirement for the degree of Doctor of Philosophy

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By

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December 2017

Chairman : Associate Professor Syed Abdul Rahman Al Haddad, PhD
Faculty : Engineering

The speaker's native dialect, accents and the socioeconomic background are few factors that influence the speaking style. The mixing of Qira'at is considered forbidden in Islam, especially during salat prayer. The identification of threats that could influence the accuracy of voice recognition and influence decisions in recitation recognition performance of accents recognition. On the other hand, only few studies focus on research of the performance factor or accuracy in the reverberant environment and none yet focusing on the factors that would affect Qira'at speech signals and identification.

The main objective of this thesis is to propose the identification process of Quranic recitation but oriented to the identification of various Qira'at with the emphasis on recognition without being affected by echo or noise during live recitation or in recordings. Sequential Windowing Parameterizing of Affine Projection Algorithm (SPAP) is proposed to improve windowing parameterizing during echo cancellation, while recognition accuracy factors are taken into account for further improvement. The process of the SPAP Algorithm is to extend parameters of the Affine Projection Block with two different selections of windowing length that affect the final accuracy on pattern classification.

The usage of Feature Selection (FS) contributes to simplify and enhance the quality of the dataset used by selecting significant features. Qira'at audio files for Surah Ad-Dhuha are used in this study to re-sample an audio sample. Clean audio signals from AEC are used with proposed feature selection technique called as X-Ant Colony Optimization, that utilizes the concept of Ant Colony Optimization, and can enhance feature extraction. For the feature vectors that are collected from feature extraction (MFCC) and feature selection (X-ACO), the feature vectors are used as input for the

classification phase. A combination of Principal Component Analysis (PPCA) and Gaussian Mixture Model (GMM) is proposedly in used for the classification phase as it is able to reduce any redundancy from the latent variables and carries only the most important information through dispersion of entropy.

To evaluate the algorithm, 350 samples for 10 types of Qira'at recitation are in used, and for justifying the best pattern classification, few algorithms are tested in the early preliminary evaluation with K-Nearest Neighbour, GMM and PPCA. And in the final evaluation for PPCA, it achieved high accuracy with 95.15%, while WER and EER are around 7.63%. The current evaluation for SPAP tests another echo database of a Lecture Room that presents a reduction in the accuracy rate of around 92.1% while the WER and EER are around 7.53%. But both of the results are significant compared to achieve results for MFCC without SPAP feature selection technique that just acquired 89.1% in early preliminary test. It proves that the current proposed algorithm achieves better results in Echo Greathall comparable to Echo Lecture Room and finally, these results will be used as foundation for any upcoming related research that may improve the understanding of Qira'at among the Muslim.

Abstrak tesis dikemukakan kepada Senat Universiti Putra Malaysia sebagai memenuhi keperluan untuk ijazah Doktor Falsafah

REKABENTUK ALGORITMA KECERDIKAN IDENTIFIKASI QIRA'AT

Oleh

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Dialek bahasa asli, cara percakapan dan latar belakang sosioekonomi adalah beberapa faktor yang mempengaruhi gaya bercakap. Pencampuran Qira'at adalah dilarang dalam Islam, terutamanya semasa pembacaan sewaktu solat. Pengenalpastian ancaman yang boleh mempengaruhi ketepatan pengenalan suara dan keputusan mempengaruhi pengenalan bacaan itu sendiri belum lagi dibincangkan secara meluas, terutamanya dengan pelbagai kesan yang berlaku dalam bunyi bising dan kekeliruan latar belakang, prestasi pengecaman cara percakapan dalam persekitaran yang mempunyai bunyi bising dan gema akan menurunkan kualiti audio secara dramatik juga. Namun begitu, hanya sedikit kajian yang menumpukan pada penyelidikan faktor prestasi atau ketepatan dalam persekitaran gema dan masih lagi belum ada yang mengkaji faktor-faktor tersebut pada kesannya di dalam sampel Qira'at dan pengelasannya.

Tujuan utama tesis ini adalah untuk mencadangkan proses pengenalan bacaan Al-Quran tetapi berorientasikan identifikasi pelbagai Qira'at dengan penekanan atas pengiktirafan tanpa dipengaruhi oleh gema atau bunyi semasa bacaan secara langsung atau dalam rakaman. Melalui teknik Turutan Tetingkap Parameter melalui teknik Affine Projection (SPAP) dicadangkan untuk meningkatkan tetingkap parameter semasa pembatalan gema, manakala faktor ketepatan pengiktirafan diambil kira untuk penambahbaikan selanjutnya. Proses Algoritma SPAP adalah untuk mendapatkan parameter Blok Perangkaan Affin dengan dua pilihan panjang tetingkap yang mempengaruhi ketepatan corak pengelasan akhir identifikasi.

Penggunaan Pemilihan Ciri (FS) menyumbang untuk memudahkan dan meningkatkan kualiti dataset yang digunakan dengan memilih ciri-ciri penting. Fail-fail audio Qira'at untuk Surah Ad-Dhuha digunakan dalam kajian ini untuk menyusun semula sampel

audio. Isyarat audio bersih dari AEC digunakan dengan teknik pemilihan ciri yang dicadangkan dipanggil sebagai Pengoptimuman Koloni X-Semut, yang menggunakan konsep Pengoptimuman Koloni Semut, dan dapat meningkatkan pengekstrakan ciri. Untuk vektor ciri yang dikumpulkan dari pengekstrakan ciri (MFCC) dan pemilihan ciri (X-ACO), vektor ciri digunakan sebagai input untuk fasa klasifikasi. Gabungan Analisis Komponen Utama (PPCA) dan *Gaussian Mixture Model* dicadangkan untuk digunakan untuk fasa klasifikasi kerana ia dapat mengurangkan sebarang redundansi daripada pembolehubah *laten* dan hanya membawa maklumat yang paling penting melalui penyebaran entropi.

Untuk menilai algoritma, 350 sampel untuk 10 jenis bacaan Qira'at digunakan, dan untuk membenarkan klasifikasi corak yang terbaik, beberapa algoritma diuji pada awal penilaian awal dengan *K-Nearest Neighbor*, GMM dan PPCA. Dan dalam penilaian akhir untuk PPCA, ia mencapai ketepatan yang tinggi dengan 95.15%, manakala WER dan EER adalah sekitar 7.63%. Penilaian semasa untuk SPAP menguji satu lagi pangkalan data gema Bilik Kuliah yang menunjukkan pengurangan kadar ketepatan sekitar 92.1% manakala WER dan EER adalah sekitar 7.53%. Tetapi kedua-dua keputusan adalah signifikan berbanding dengan mencapai keputusan untuk MFCC tanpa teknik pemilihan ciri SPAP yang hanya memperoleh 89.1% pada awal ujian awal. Ia membuktikan bahawa algoritma yang dicadangkan semasa mencapai hasil yang lebih baik dalam Echo Greathall setanding dengan Bilik Kuliah Echo dan akhirnya, keputusan ini akan digunakan sebagai asas bagi sebarang penyelidikan berkaitan yang akan meningkatkan pemahaman Qira'at di kalangan umat Islam.

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I certify that a Thesis Examination Committee has met on 8 December 2017 to conduct the final examination of Noraziahtulhidayu binti Kamarudin on her thesis entitled "Design of Intelligent Qira'at Identification Algorithm" in accordance with the Universities and University Colleges Act 1971 and the Constitution of the Universiti Putra Malaysia [P.U.(A) 106] 15 March 1998. The Committee recommends that the student be awarded the Doctor of Philosophy.

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LIST OF ABBREVIATIONS

AC	Accents Conversion
AEC	Acoustic Echo Cancellation
AP	Affine Projection(AP)
ALGASD	ALGerian Arabic Speech Database
ADC	analog to digital conversion
ACO	Ant Colony Optimization
ANN	Artificial Neural Network
ASR	Automatic Speech Recognition
DFT	Discrete Fourier Transform
DTW	Dynamic Time Warping
ERLE	Echo Return Loss Enhancement
EER	Equal Error Rate
FFT	Fast Fourier Transform
FRLS	Fast Recursive Least Squares
GMM	Gaussian Mixture Model
HMM	Hidden Markov Model
LI	Language Identification
LMS	Least Mean Squares
LS	Least Squares
LDA	Linear Discriminant Analysis
LP	Linear Predictive
LPCC	Linear Predictive Cepstral Coefficients
LPC	Linear Predictive Coding
MLLR	Maximum Likelihood Linear Regression
MFCC	Mel Frequency Cepstral Coefficients
MSA	Modern Standard Arabic
MSE	Mean Square Error
MLP	Multilayer Perceptron
NLMS	Normalized Least Mean Square
PLP	Perceptual Linear Prediction
PSD	Power Spectral Density
PSNR	Peak to Signal Noise Ratio
PCA	Principal Component Analysis
PPCA	Probabilities Principal Component Analysis
RANSAC	RANdom Sample And Consensus
RASTA	Relative Spectra Filtering
RNN	Recurrent Neural Network (RNN)
RLS	Recursive Least Squares (RLS)
RASTA-PLP	Relative Spectral Transform - Perceptual Linear Prediction
RIR	Room Impulse Response (RIR)
SVM	Support Vector Machine (SVM)
VQ	Vector Quantization (VQ)
WER	Word Error Rate (WER)
ZCR	Zero Crossing Rate(ZCR)

والضحى	Wadh-dhuha
سجى	Saja
فترضى	Fa-tardha
فاوى	Fa-aawa
فهدى	Fa-hada
فاغنى	Fa-aghna
Imam Abu Amru al Basri	Al Douri from Abu Amru Al Susi Abu Amru
Imam Asim al Kufi	Hafs an Asim Shoba from Asim
al Kisai Al Kufi	Abu al Hareth al Kisaie Ad-Duri al Kisaie
Imam Nafi' al Madani	Warsh, Qaloon
Abu Ja'far al Madani	Wardan, Ibn Jammaz
Hamzah al-Kufi	Khallad from Al Hamzah Khallaf from al Hamzah
Ibnu Kathir al-Makki	Qunbul Ibnu Katheer, Albazzi Ibnu Kathir
Khalaf al-bazar	Edrees Al Haddad, Khalf Ishak Al Bazaar
Imam Ibnu Amir As-Syami	Hesham from Ibn Aamer, Ibnzakwan
Imam Yaaqub Al Basri	Rauh from Yaacob, Ruweis from Yaacob

CHAPTER 1

INTRODUCTION

1.1 Overview

The understanding of accents relates to a person's language while dialects mostly apply to the varieties of vocabularies, idioms, grammars and pronunciation of a specific language within certain community, often due to geographical differences. Accents for a selected language vary between speakers (Han-ping Shen, et al,2015), but they are still understandable within the same scope of speakers. But, if it is applied within speech processing field; it is a requirement to have certain standard of accent identification; in terms of improving the execution of speech, speaker recognition systems, human machine interaction and developing regional dialect spoken query systems (Borah and Sharma, 2016). This will create difficulties to model input speech for the development of speaker-independent systems. Factors of rate recognition accuracy is affected due to the spectrum changes for Quranic signals that are prone to additive noise or echoes from recordings.

Mixing different types of Qira'at recitation is considered forbidden in Islam, especially during Solat. As Quran is written in Arabic Language and Arabic is one of the languages that are often described as morphologically complex and the problem of language for Arabic are multipart by the variation of dialectal (Almeman K, 2015; N. Zainon et al,2012 ;Noor Jamaliah,2011).

Al Quran is the Holy Scripture given by Allah swt to Muslims worldwide and recited with the Arabic Language, the only language it was revealed in. All Muslims believe that the Quran's content is neither corrupted nor changed and hence the value, and the original text is protected and maintained. Therefore, it is forbidden to recite Holy Quran in other languages except Arabic with neither additions nor subtractions (Mohammed, Shahrizal, and Salam, 2015).

Although recited the Quran is not used in communication, it is extremely important to teach the pronunciation of classical Arabic sounds (language of the holy Quran). Moreover, teaching how to pronounce Quranic sounds is indispensable to correctly read the holy Quran, either for memorizing it or for reciting it in Islamic worship, such as prayers (Elhadj, Alghamdi, and Alkanhal, 2013). Basically, there are 10 different Qira'at recitations (7 types known as Mutawatirrah and 3 Mashuur) (Rauf,2014), which use the same Mushaf Uthmani and the differences between those Qira'at happened due to the location of where the Holy Quran was revealed (eg; Nafi' Ibn Kathir, Abu 'Amr ibn al-'Ala', Ibn 'amir, 'Asim, Hamza and al-Kisa'i). The sensitivity of recitation and pronunciation of the same vowel and different words by a particular speaker may affect the sensitivity or accuracy of automatic identification for language accents and this may happen with Quranic recitations too. Figure 1.0 shows

the coverage of different Qiraat based on geographical location. The most dominant Qira'at is Hafs An Asim, which is widely recited by most Muslims worldwide, then followed by the Qira'at of Warsh, Ad-Douri and Qaloon.

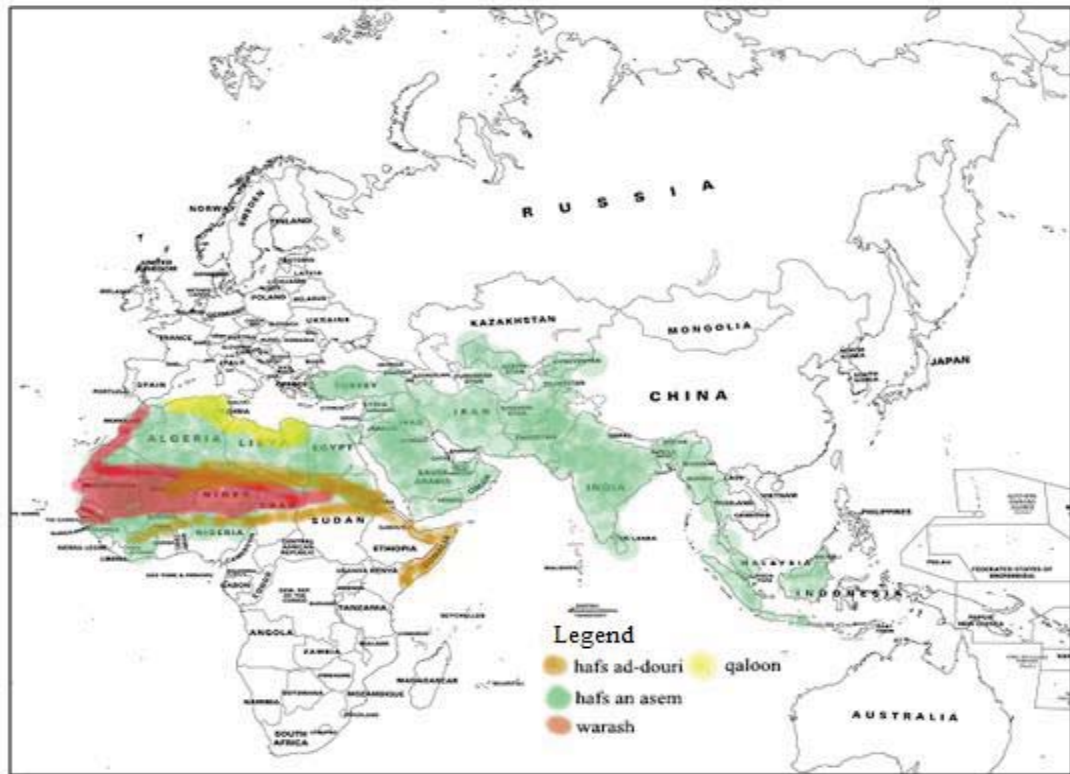


Figure 1.0 : World map with different Qira'at Recitations
(Ibnamin,2017)

As the Holy Quran was revealed to Muslims in the Arabic Language, it is an obligation for Muslims to learn and mastered it. Arabic is the official dialect for more than 22 nations. This study, the focus is to identify the Qira'at, which therefore would help to identify the Qira'at signals based on the stored recorded signal. Throughout this study, some preliminary discovery has been done on related scholars that focus on similar topics like Tajweed Automatic Identification, Language with Accents Automatic Identification and Music Genre Identification. The previous scholars have successfully created an application, which is able to detect errors in recitation by using Mel Frequency Cepstral Coefficients(MFCC) for feature extraction and also Hidden Markov Model (HMM) as pattern classification for identifying error in Quranic Recitation(Tajweed) especially during recitation. The best results achieved by them were within 86% to 92 % for both different researchers (Noor Jamaliah et al,2013;Elhadj et al, 2014). The rate of recognition accuracy is considered to be less due to the spectrum of the Quranic signals, which may be prone to additive noise or echoes from the recordings. In any case, the latest speech databases have representation of distinctive levels and range from the phonemes to sounds.

Furthermore, it will interpret the speech; and the interpretation is adjusted to the speech acoustic signal(Auran et al, 2004). But the process of transcribing Quran verse surely will take a long time and should be done carefully as it involves a Holy Surah. Therefore, this research embarks new process in identifying the Qira'at only not to be used for a specific Surah but can be applied to the whole Holy Quran. Errors can be missing words, verse, misreading Harakat (articulation, accentuations and diacriticization) (WaqarMirza et al, 2010).

From Waqar Mirza Muhammad et al (2010), E-Hafiz system is developed for answering these problems that happen due to noise in the environment for detecting speech recognition while reciting a Surah. The application is produced using Mel-Frequency Cepstral Coefficient (MFCC) strategy to concentrate voice features from Quranic verse recitation and maps them with the information gathered amid the preparation stage. Any dissimilarity of Tajweed recognition is pointed out for the recitors. Testing results of short verses of the Quran utilizing the E-Hafiz framework are exceptionally promising as what's being done by the researcher of the E - Hafiz system (Waqar Mirza Muhammad et al, 2010). For those Quranic research done, they have used multiple wave files based on collected databases, which varies by gender and age. In this specific research of Qira'at identification, only expert scholars shall be collected for the Qira'at database in terms of feature extraction and pattern classification purpose.

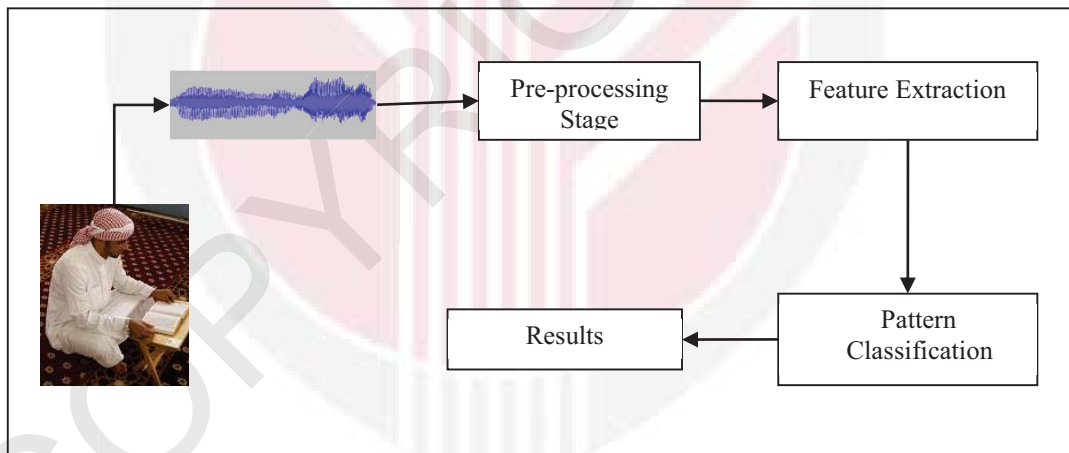


Figure 1.1 : Basic flow of Qiraat Identification

Quranic signal is an audio signal under study to investigate the Qira'at classification. Therefore, Quranic signals are closely related to music signals in terms of pitch and skewness. For the targeted Surah of Ad-Dhuha, which also contains *imalah*, *taqlil*, *naqal* and *badal*; these attributes contribute to the parameters of Qira'at identification. The interruption of echo that always occur in pre-processing stages as shown in Figure 1.1, may reduce the accuracy rate achieve during recording, or in the live environment. Therefore, methods of applying suitable adaptive algorithms are investigated with the robust feature selection algorithms that can give a higher value of accuracy during the

feature extraction process. The processes focused in this study will involve the main stages of Pre-Processing and Feature Extraction, while the whole processes of identification will still cover till pattern classification.

Figure 1.2 illustrates the accuracy results achieved by different implementation of research done by previous researchers, those implementations vary with the accuracy of Arabic speech recognition, automatic error (Tajweed) identification for Quranic recitation, songs and singer voice classification. The investigation done by those researchers was based on identifying on the accuracies factor and Word Error Rate (WER) and the results were presented in percentage values. Among that research the highest accuracy of 94.82% on songs classification accuracy was done by Ghosal et al (2012). Feature extraction techniques used were solely using MFCC for the feature vector while wavelet was used in songs identification.

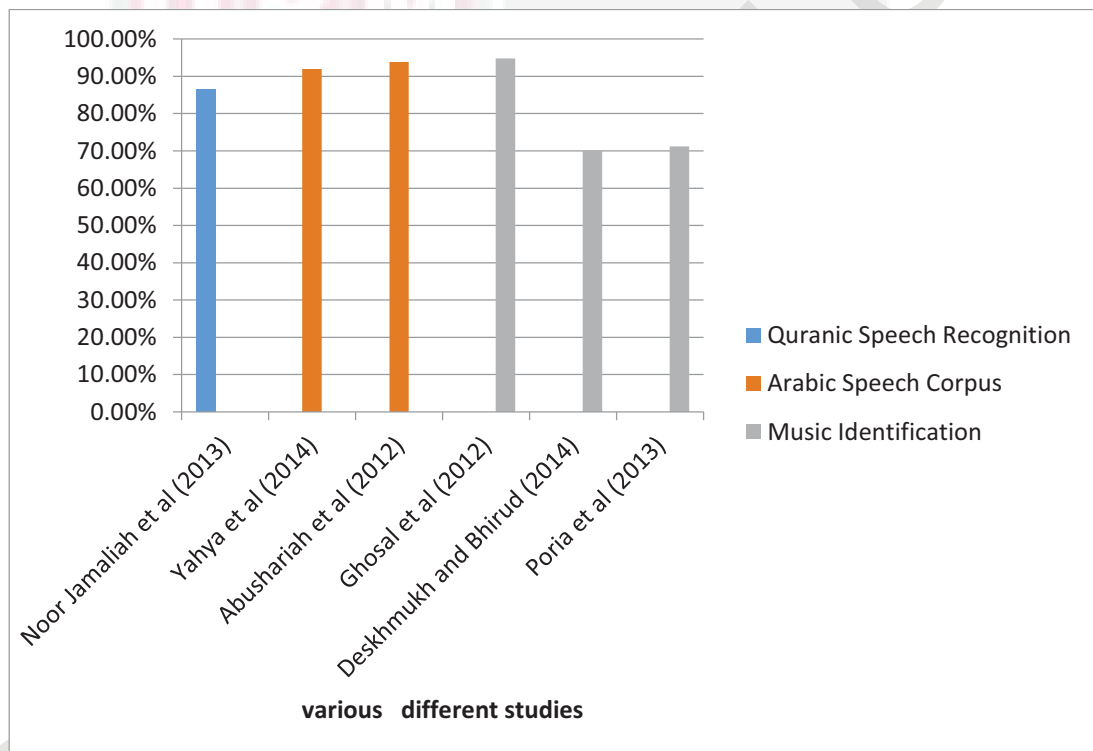


Figure 1.2 : Accuracies achieved by Music Identification, Arabic Speech and Quranic Speech

Throughout this studies, preliminary test are already done to verify the suitable algorithms to be used for the current research as presented in Literature Review.

1.2 Problem Statement

In solving the problems of Qira'at identification and to avoid from wrong recitation during salat; this study of Qira'at identification considered new in the current research area. It is important to understand the correct concept and flows of identification that may involved by first extracting the significant cleaned signal as an input for feature vector while the second one is the deep emphasis on the feature selection process.

Acoustic noise from ambience and sound systems distorted the real signals. It happens from frequency response and masking produces unusually large peaks as a result music recording are prone to distraction and the identification would be a problem. Noise is omnipresent in almost all acoustical environments(Utpal et al, 2016;Hom K,2013; Gil-Cacho, J,2013). Amongst popular solution are Least Mean Square(Zhang et al, 2015;Deepika M,2013;Longbiao et al,2012) and Recursive Least Square(Rakesh P, Kumar T, 2015;Adapa N,Bollu S,2012;PSR Diniz, 1984). But the techniques used to have slow convergence rate(Gonzalez A et al, 2012), high computational complexity taken (Roshahliza et al, 2012;Ranbeer T, Dheeraj A,2012; Nielsen JK, Jensen SH, 2012)based on Mean Square Error (Weiyang Cai,2013;Nurul et al,2013, Arablouei R, Do K,2012)and Peak to Signal Noise Ratio(Verma A, Verma N,2016;Santosh D et al,2012;Kun Han et al,2014).

Interruptions by echoes from sound systems, their frequency response and masking produces unusually large peaks. Noise appears in music recordings such as country music; and few studies with different individual have shown that they may be prone to distraction and for gender difference especially during indexing process(Jing, et al, 2012), and this may happen also to Quranic recordings (Adapa and Bollu, 2012; Razak et al,2008). The performance of the accuracy prediction model are severely influenced by noise sources from a computer, Air Conditioner (AC) duct, fan etc and always depends on number model data, noisiness and target classes(Julian Teryzk, 2014; Lerch A,2012). While, for hands free telephony inside a car, the noise source could be from car engine noise, outside vehicle and ambient wind. It is even possible that surrounding noise sources are captured at a higher level than the speech signal (Ganesan and Manoharan, 2015). The detection may drop for some of Qira'at due to appearance of noise/echo, the identification shall be robust enough to face possible environment variability and shall be able to achieve a high accuracy for short input speech samples (Safavi, 2015).

Secondly, in finding significant features vector it has to find the best feature coefficient frame selections and also the proper technique of feature extraction. The results later can be used in pattern classification for Qira'at identification but on the other hand it may results over fitting of training irrelevant data (Tabakhi, et al, 2014; Rashno A et al,2013;Unler et al,2011).

Thirdly, the identification shall be robust enough to face possible, environment variability and shall be able to achieve a high accuracy using selected pattern classification (Safavi, 2015; Gholiput A et al,2012; Rao KS, Koolagudi SG,2013) of Probabilistic Principal Component Analysis(Xiao-chun L et. al,2012; Xiaodong Cui MA and BZ,2012; Ismail MNI, Muse ME-HM,2014).

1.3 Objectives

The main objectives of this research is to propose an intelligent Qira'at identification algorithm. The specific objectives are as follow;

- a. To design dynamical sequential windowing selection for Acoustic Echo Cancellation using Affine Projection (SPAP) from the recording sound on Qira'at speech signals, within best parameters and experimental setup.
- b. To optimize new feature selection using X-Ant Colony Optimization(X-ACO) on current feature extraction technique with Mel Frequency Cepstral Coefficients(MFCC),
- c. To design Qira'at Identification from speech signals using Probabilistic Principal Component Analysis and Gaussian Mixture Model.

1.4 Thesis Scope

This research presents the analyzation of Qira'at Identification based on digital speech audio recordings to improve the recognition performance and verification on the accents classifications and finally would solve the problems of Qira'at identification. The performance of recognition for accuracies is easily degraded when noise from surroundings and echoes (reverberant) corrupt the live or recorded Qira'at speech. Time is wasted to find the significant data for feature vectors to be used in the classification phase.

In this study, the effectiveness of applying accents identification is investigated for a recorded sample database of Quranic speech recitation. For the accuracy factor of the current investigation, three methods of performance. First using the technique of accuracy in percentage values, Word Error Rate and Equal Error Rate. On the segmentation part, current studies use 16kbit with Surah Ad-Dhuha verse segmentation that is done by manually for 10 Qira'at with a variety of recitation from selected trusted scholars with 2 faces of Qira'at recitation that by total covers 20 types of Qira'at. There are 350 speech segmented verses collected for Quranic speech audio, Sequential Parameterization Affine Projection for AEC. For feature extraction Mel Frequency Cepstral Coefficient (MFCC) together with X-Ant Colony Optimisation (X-ACO) on feature selection and Probabilities Principal Component Analysis (PPCA) is used for pattern classification. The effects of echo or reverberation in this study are also investigated using Aachen Impulse Response Database and the metrics that are used in evaluating for the whole signals are Mean Square Error and Peak to

Signal Noise Ratio. The data used are from Islamway.net since the audio samples are from trusted scholar (Mohamed, Hassanin, Taher, and Othman, 2014).

Sample data process based on 10 Qira'at includes; Riwayat Imam Abu Amru Al Basri (Al Douri from Abu Amru and Al Susi Abu Amru), Riwayat Imam Asim al Kufi (Hafs an Asim and Shoba from Asim), Riwayat al Kisai Al Kufi (Abu al Hareth al Kisaie, Ad-Duri al Kisaie), Riwayat Imam Nafi' al Madani (Warsh, Qaloon), Riwayat Abu Ja'far al Madani (Wardan, Ibn Jammaz), Riwayat Hamzah al-Kufi (Khallad from Al Hamzah, Khallaf from al Hamzah), Riwayat Ibnu Kathir al-Makki (Qunbul Ibnu Katheer, Albazzi Ibnu Kathir), Riwayat Khalaf al-bazar (Edrees Al Haddad, Khalf Ishak Al Bazaar), Riwayat Imam Ibnu Amir As-Syami (Hesham from Ibn Aamer, Ibn Zakwan) and Riwayat Imam Yaaqub Al Basri (Rauh from Yaacob, Ruwais from Yaacob).

And for this research, Surah Ad- Dhuha is in used for Quranic samples and only male offline recordings are collected and audio database are collected from Islamway.web while echo database from Aachen Impulse Response database.

1.5 Thesis Outlines

This thesis comprises of 5 chapters as follows:

Chapter 1 present the general overview of the research, which reflects on the significance of the study, problem statement, research objective, aim, and the background of study. This chapter gives general views of the continuous topics that are covered in the upcoming chapter.

In Literature Review for Chapter 2, the chapter explains a comprehensive understanding of recent achievement of previous research which relates to Quranic speech signals, the techniques and algorithms used for the feature extraction, and pattern classifications in identifying Quranic verse especially, performance of other research related, topics and issues, which arises for the automatic identification that are currently available. The issues of acoustic echo cancellation and parameters involved and the requirement of feature selection algorithms and the important parameters for feature extraction are also discussed in this chapter.

For Research Methodology in Chapter 3, the first section describes and discusses algorithms and all methods, including techniques and tools, which are involved in the sampling and pre-processing (basic feature extraction and pattern classification). For the second section an overview on sequential windowing length parameterization for acoustic echo cancellation is presented and the third section focuses on feature selection algorithms for the Quranic speech signals accents identification. Any parameters and metrics for evaluation, which supports this study are also investigated.

Furthermore, the Word Error Rates, Accuracies, Equal Error Rates given for recorded recitations from total signals are also evaluated in this study.

In Chapter 4, the overall results, which are implemented as described in Research Methodology, are presented. Therefore, the effectiveness based on accuracy achievement and Word Error Rate (WER) is evaluated too in this chapter. Overall discussion on the proposed algorithm is discussed in this chapter. The review results for a recorded recitation, is also covered within this chapter.

The final chapter 5, discusses the whole contributions of the currents studies, which includes; conclusions and a proposals for future work. In this chapter too, the trade off of this accents identification techniques with various language also presented together with music genre identification.

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