

Sudan University of Science and Technology College of Graduate Studies

SOUND RECOGNITION FOR SELECTED VERSES OF THE HOLY QURAN

التعرف الصوتي على آيات مختارة من القرآن الكريم

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Dedication:

To my mother.

Abstract

In this study of sound recognition for selected verses of the Holy Quran, five verses selected from the Holy Quran. This done as a small number of verses contains all the Arabic phonemes. One hundred recitations of each verse were prepared in wave files. Verses had been recited by famous certified readers of the Holy Quran. A file created from each recite by extracting only the first word of the verse. A wave file of noise is set by the researcher's voice, for testing purpose. All wave files recorded at Sampling rate =22.050 kHz, PCM signed 16 bit mono.

Three types of coefficients are extracted from each wave file to represent features of speech. They are Mel Frequency Cepstrum Coefficients MFCC, Power Spectral Density PSD and Reflection Coefficients RC.

Also three techniques of speech recognition are used. They are Hidden Markov Models, Dynamic time warping and Artificial Neural Networks.

Test were done at two levels. The first stage was applied at the full verse level. All the three techniques mentioned were used. HMMs and ANNs trained by the first 30 samples and test done by all the 100 sample of each verse.

The second stage was applied in the same way, but the used samples were of the first word of the verse. HMMs technique only was selected to be used in recognition. That due to high recognition rates scored at the first stage. This stage repeated in the same way, but used the first 50 samples in training instead of the first 30 samples.

Mainly, HMMs scored high rates of recognition to coefficients used(MFCC, PSD and RC). Low recognition rates with high confusion scored by ANNs and DTW at

verse level. For all coefficients used, high scores of recognition rates with low confusion rate concentrated in HMMs with MFCC. MFCC scored higher recognition rates than PSD and RC.

الخلاصة

في هذه الدراسة للتعرف الحاسوبي الصوتي على آيات من القرآن الكريم، تم اختيار خمس آيات من القرآن الكريم وهو أقل عدد من الايات يمكن تحتوي على جميع الوحدات الصوتية العربية ويكون زمن التلاوة أقصر زمن ممكن ليكون حجم البيانات أقل وتركيز الخصائص الصوتية المطلوب أكبر.

أعدت مائة ملف صوت لكل آية بأصوات قراء مشهورين ومجازين. اعد من كل منها ملف صوت اخر يحتوي فقط على الكلمة الاولى . اعد ملف بصوت الباحث يحتوي ضجيج، للتحكم أو معرفة مدى قرب النتائج منه وذلك بوضعة كنموذج للمقارنة.ملفات الصوت مسجلة بمعدل ٢٢,٠٥٠ كيلو هرتز ١٦ بت مونو.

استخرجت ثلاثة انواع من المعاملات لتمثل الخصائص الصوتية من هذه الملفات. والمعاملات هي معاملات السبسترم وكثافة الطاقة ومعاملات الانعكاس. واستخدمت ثلاثة تقنيات للتعرف على الكلام هي نماذج ماركوف المخفية والبرمجة الديناميكية والشبكات العصبية الاصطناعية.

أجريت الاختبارات على مرحلتين. المرحلة الأولى على مستوى الآية ، باستخدام التقنيات الثلاث المذكورة للتعرف على الكلام. نماذج ماركوف المخفية والشبكات العصبية الاصطناعية تم تدريبها على العينات ال ٣٠ الأولى من كل آية، والاختبار استخدمت فيه جميع العينات ال ١٠٠ من كل آية.

المرحلة الثانية بنفس الطريقة للمرحلة الأولى ، ولكن على مستوى الكلمة (الكلمة الأولى من الآية). استخدمت فقط نماذج ماركوف المخفية في الاختبارات. نظرا لانها سجلت ارتفاع في معدلات التعرف على الكلام في المرحلة الأولى. اعيدت المرحلة الأخيرة بنفس الطريقة، واستخدم فيها أول ٥٠ عينة في التدريب بدلا من ٣٠ عينة.

بخلاصة النتائج وجد الاتي:

- نماذج ماركوف المخفية (HMMs) سجلت أكبر معدل تعرف مع معاملات MFCC وتميزت عموما
 بالاستقرار والاتزان في ادائها.
- الشبكات العصبية و تقنية DTW سجلت أقل معدلات تعرف ما عد في الاية الاولى مع MFCC والاية الرابعة مع PSD وتميزت بالتباين في ادائها لذلك لم تستخدم في المرحلة التالية في البحث.
 - في نماذج ماركوف عموما سجلت معاملات MFCC معدلات تعرف أعلى من PSD وRC .
 - بعض الكلمات سجلت معدلات تعرف عليها بصورة اكبر من غيرها.

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