

AUTOMATIC MUSIC TRANSCRIPTION 2017

Nora Huang

Department of Computer Science
University of Victoria
Victoria, B.C, Canada
norah@uvic.ca

Aazim Lakhani

Department of Computer Science
University of Victoria
Victoria, B.C, Canada
aazimlakhani@uvic.ca

Parul Parul

Department of Computer Science
University of Victoria
Victoria, B.C, Canada
pparul@uvic.ca

ABSTRACT

The objective of this project is to evaluate the accuracy of various models to perform automatic music transcription by extracting features from the audio music and applying different classifiers to get the note. Through this initiative we plan to focus at transcription on certain music instruments. Furthermore, only the main instrument of the polyphonic music will be transcribed.

1. INTRODUCTION

The dataset for this project will be collected through MIDI Aligned Piano Sounds (MAPS), Isophonics Annotations, KSN Database to name a few. The sources claim the data is used for transcriptions of both monophonic & polyphonic instruments. The datasets are annotated & we would evaluate the data to use music relevant for this project.

Then acoustics features such as Zero Crossing Rate, Autocorrelation, YIN metrics for time domain & FFT, STFT, Spectrogram, cepstrum metrics, spectral flux for the frequency domain would be used to extract notes from audio music.

The extracted features will be used as input to the classifiers for note transcription. Different classifiers, such as K-NN, Logistic Regression, Stochastic Gradient Descent, Support Vector Machines (SVM) would be evaluated for accuracy. Features would be combined together to give better accuracy on real world data.

Marsyas would be used as the framework for acoustic features extraction. Python & Scikit-learn would be used to prepare the learning model, which would eventually, transcribe.

The overall system diagram for the transcription is described in Figure 1

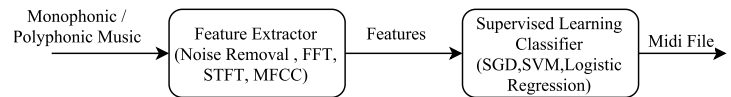


Figure 1. System Flow Diagram

1.1 Timeline

We make 4 milestones for the whole project, for each task there will be two responsible team member for it, one as primary while the other as secondary. Please find the detail of the milestone in Table 1.1

Table 1. Milestones

Milestones	Deadline	Tasks	Primary	Secondary
MileStone1	March 6	Dataset Features Classifier	Aazim Parul Nora	Nora Nora Aazim
MileStone2	March 13	Train	TBD	TBD
MileStone3	March 27	Testing	TBD	TBD
MileStone4	March 30	Report	TBD	TBD

1.2 Role of team member

Nora: Architecture design, team management, coding

Aazim: Training & testing with different combinations of classifiers & features

Parul: Work on audio processing & feature extraction.

2. DATASET

We obtained data for both monophonic and polyphonic sounds through MAPS. We're also in the process of having data from Isophonics Annotations & KSN Database.

3. ACOUSTIC FEATURES FOR TRAINING

Time Domain : Zero Crossing Rate, Autocorrelation, YIN metrics
Frequency Domain : FFT, STFT, Spectrogram, cepstrum metrics, spectral flux.



© Nora Huang, Aazim Lakhani, Parul Parul. Licensed under a Creative Commons Attribution 4.0 International License (CC BY 4.0). **Attribution:** Nora Huang, Aazim Lakhani, Parul Parul. "Automatic Music Transcription 2017", 17th International Society for Music Information Retrieval Conference, 2016.

4. CLASSIFIERS

Stochastic Gradient Descent, Logistic Regression, Support Vector Machines

5. RESULT

This would comprise of the collection of features, classifiers used to give the highest level accuracy. Various combinations would be tried, but only the one's that generalize well, would be mentioned in the result.

6. CONCLUSION

Basic on the result we should be able to figure out which combination is best for automatic music transcription.

7. REFERENCES