

1 Summary of articles

1.1 Folk Music Classification

Chai begins by explaining how it would be useful for computers to be able to classify musical works, and by doing so we could learn how a human can identify a musical work quickly. They then explain the goals of their research being ”: (1) to explore whether there exists significant statistical difference among folk music from different countries based on their melodies; (2) to compare the classification performances using different melody representations; (3) to study how HMMs perform for music classification as a time series analysis problem.”

They used German and Irish folk music pieces for their data. They chose folk songs because of their clear monophonic melody line. They chose German and Irish because of the availability of data.

For the representation of the melody line they used four different methods.

- Absolute Pitch: where they have a value for each note in an octave.
- Absolute Pitch with Duration Representation: Same as absolute pitch but each note is repeated relative to the length of the smallest rhythmical note.
- Interval Representation: The melody is converted to a sequence of intervals.
- Contour Representation: The melody is converted to a sequence of contours, instead of using intervals it just states +/- for small change and ++/- for large change in the melody.

They used 16 different Hidden Markov Models to test how varying structures and number of states for the HMM’s performed in the melody classification. For the HMM’s they used the Baum-Welch reestimation method for training the markov model for each country. For the identification they used the Viterbi algorithm to decode the sequence and compute its log probabilities.

The data was split 70%/30% training/testing. After training the HMM’s the best representation achieved 63%-77% correctness in identifying the origin of the folk melody.

The results they concluded include that different melodic origins can have varying degrees of similarity making it more difficult to differentiate them. Additionally they discovered that although absolute pitch is a more objective way to represent the melodic line, interval representation performed better, which is consistent with how humans perceive melody lines.

Finally they conclude with stating how HMM’s can be used for melody classification for folk music and how folk music does have significant statistical difference between countries of origin. Ending with stating that melody is an important feature but classification can be greatly improved if other significant features (harmony, instrumentation, performance style etc.) are included in the features used.

1.2 Classification of Musical Genre: A Machine Learning Approach

This paper starts by describing musical genre and how it exists as a terms that define recurrences and similarities between two works of music that the community uses to identify musical events. Thus their analysis is focused on symbolic musical aspects to gain as much information about the dynamic genres, without additional noise. They used six different musical genres with 300 midi songs.

They used 5 different features of the musical work in their research:

- Melodic Intervals
- Instruments
- Instrument Classes and Drumkits
- Meter/Time Changes
- Note Extension

The experiments performed were all run within The Waikato Environment for Knowledge Analysis. They used 6 different Machine Learning Algorithms for their analysis which include:

- The Naive Bayes
- The Voting Feature Intervals
- J48
- The PART algorithm
- NNge
- RIPPER

2 References

Mllensiefen, Daniel, and Klaus Frieler. Evaluating Different Approaches to Measuring the Similarity of Melodies. *Studies in Classification, Data Analysis, and Knowledge Organization Data Science and Classification*, pp. 299306., doi:10.1007/3-540-34416-0_32.

Aloupis, Greg, et al. Algorithms for Computing Geometric Measures of Melodic Similarity. *Computer Music Journal*, vol. 30, no. 3, 2006, pp. 6776., doi:10.1162/comj.2006.30.3.67.

Chai, Wei, and Barry Vercoe. "Folk music classification using hidden Markov models." *Proceedings of international conference on artificial intelligence*. Vol. 6. No. 6.4. sn, 2001.