



jMIR Overview

Cory McKay

Marianopolis College and CIRMMT

Montreal, Canada

Lecture contents

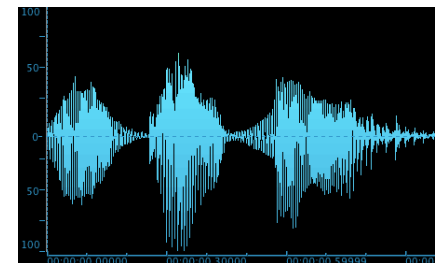
- Introduction to automatic music classification
- Introduction to the jMIR software
- The jMIR components
 - One by one
- Paper ideas
 - Previous collaborations

Goal of automatic music classification

- Learn some way of mapping “**features**” extracted from an “**instance**” to one or more “**classes**”
 - Instance: an item to be classified
 - e.g. a song
 - Features: representative information extracted from an instance
 - e.g. amount or chromatic motion in a song
 - Class: a category of interest
 - e.g. a genre, mood, artist, composer, instrument, etc.
 - Ideally organized into some kind of class **ontology**
- This mapping is typically learned using some form of **pattern recognition** and **machine learning**

Main sources of information

- Symbolic recordings
 - e.g. MIDI
- Audio recordings
 - e.g. MP3
- Cultural data
 - e.g. web data, listener statistics, metadata tags, etc.
- Lyrics
- Others
 - Album art, videos, etc.



Automatic music classification

- Typical procedure:
 - Collect annotated training / testing data
 - Extract features
 - Reduce feature dimensionality
 - Train a classification model
 - Typically supervised
 - Validate the model
- Some significant challenges:
 - Acquiring a sufficiently large dataset with sufficiently high-quality annotations
 - Designing features that encapsulate relevant data

Overview of the jMIR software

- jMIR is software suite designed for performing research in automatic music classification
- Primary tasks performed:
 - Dataset management
 - Acquiring, correcting and organizing metadata
 - Feature extraction
 - Machine learning
 - Data storage file formats

Some characteristics of jMIR

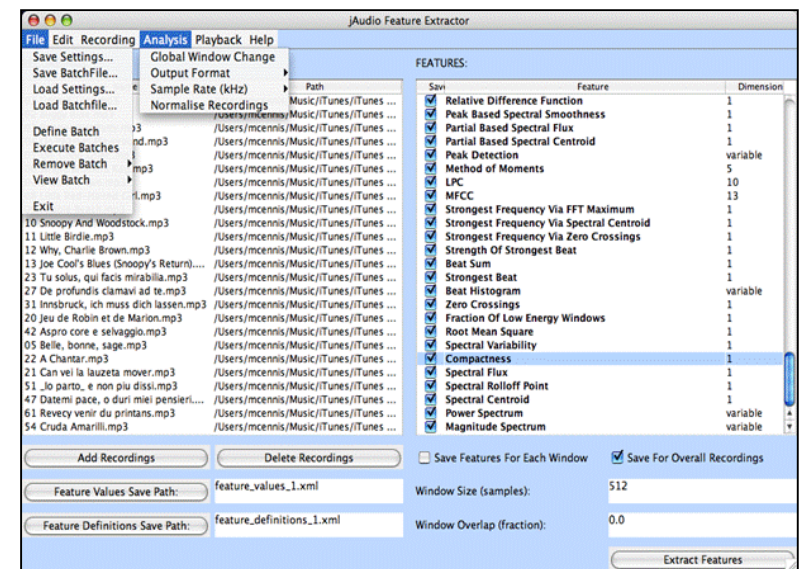
- Has a **separate software component** to address each important aspect of automatic music classification
 - Each component can be used independently
 - jMIR can also be used as an integrated whole
- Free and **open source**
- Architectural emphasis on providing an **extensible platform** for iteratively developing new techniques and algorithms
- Interfaces designed for both **technical** and **non-technical** users
- Facilitates **multimodal** research

jMIR components

- **jAudio**: Audio feature extraction
- **jSymbolic**: Feature extraction from MIDI files
- **jWebMiner**: Cultural feature extraction
- **jLyrics**: Extracts features from lyrical transcriptions
- **ACE**: Meta-learning classification engine
- **ACE XML**: File formats
 - Features, feature metadata, instance metadata and ontologies
- **lyricFetcher**: Lyric mining
- **Codaich**, **Bodhidharma MIDI** and **SLAC**: datasets
- **jSongMiner**: Metadata harvesting
- **jMusicMetaManager**: Metadata management
- **jMIRUtilities**: Infrastructure for conducting experiments

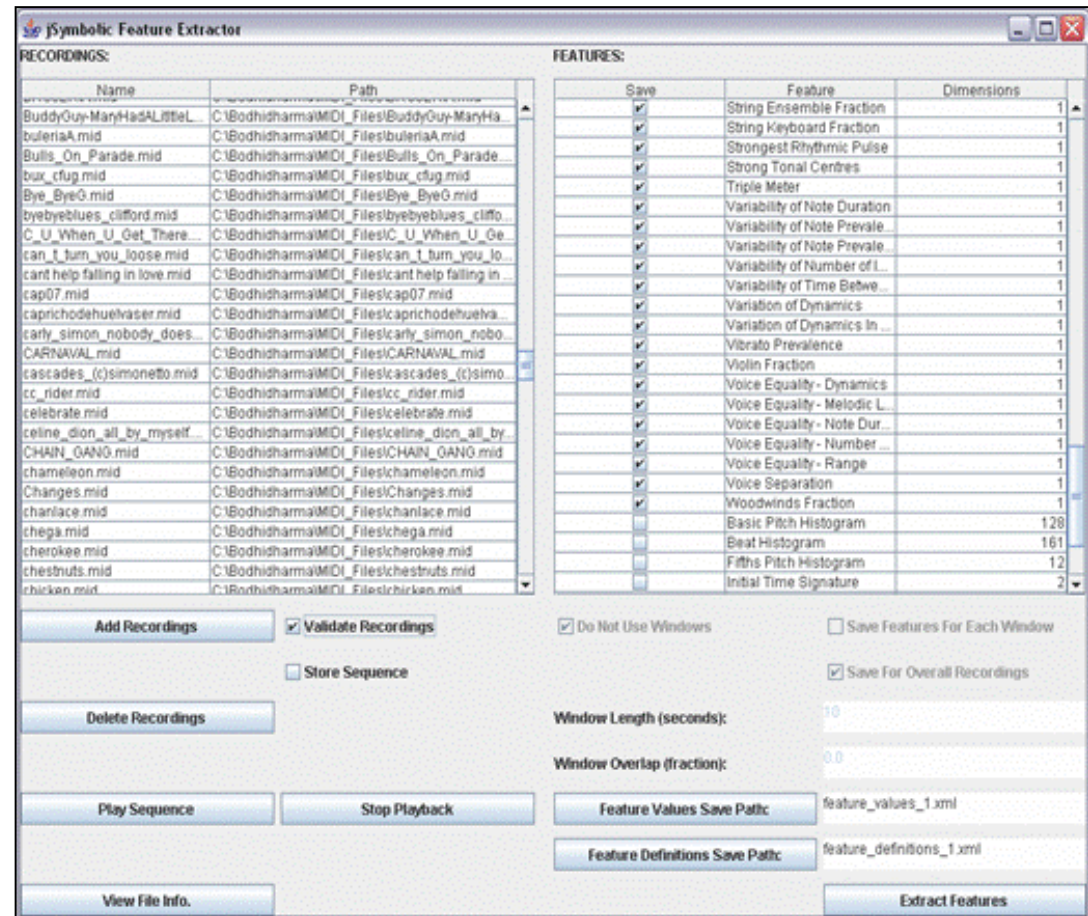
jAudio: Audio feature extractor

- Extracts features from audio files
 - MP3, WAV, AIFF, AU, SND, etc.
- 28 bundled core features
 - Mainly low-level, some high-level
- Can automatically generate new features using **metafeatures** and **aggregators**
 - e.g. the change in a feature value from window to window
- Includes tools for testing new features being developed
 - Synthesize audio, record audio, sonify MIDI, display audio, etc.



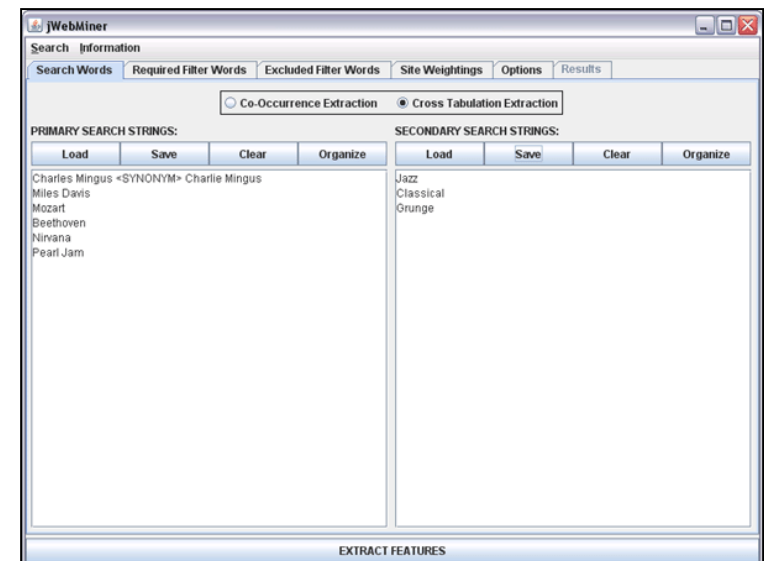
jSymbolic: Symbolic feature extractor

- Extracts features from MIDI files
- **111** implemented features
 - By far the largest existing symbolic feature catalogue
 - Many are original
- An additional **49** features are proposed but not yet implemented



jWebMiner: Cultural feature extractor

- Extracts cultural features from the web using search engine web services
- Calculates how often particular strings **co-occur** on the same web pages
 - e.g. how often does “J. S. Bach” co-occur on a web page with “Baroque”, compared to “Prokofiev”?
 - Results are processed to remove noise
- Additional options:
 - Can assign weights to particular sites
 - Can enforce filter words
 - Permits synonyms
- Also calculates features based on **Last.FM** user tags frequencies



lyricFetcher: Lyric miner

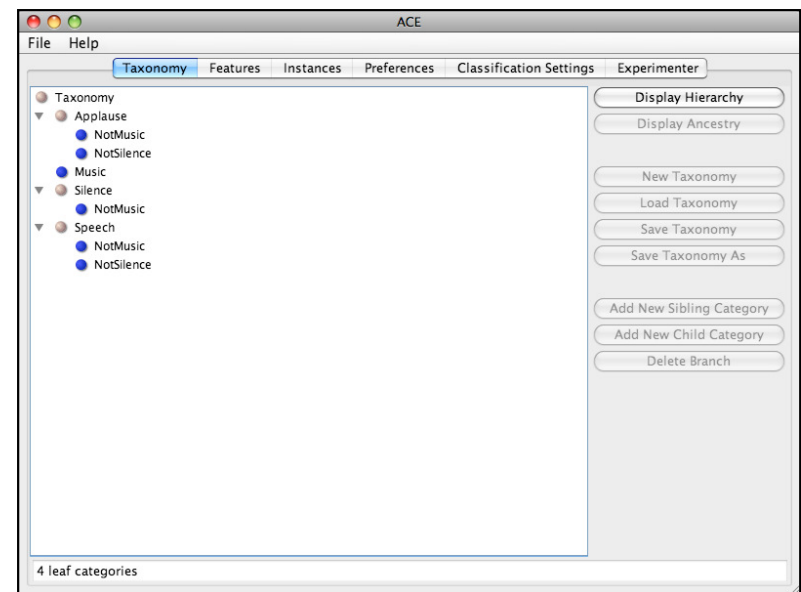
- lyricFetcher automatically **harvests lyrics** from on-line lyrics repositories
 - **LyricWiki** and **LyricsFly**
 - Queries based on lists of song titles and artist names
- **Post-processing** is applied to the lyrics in order to remove noise and make them sufficiently consistent for feature extraction
 - Deals with situations where sections of lyrics are abridged using keywords such as “chorus”, “bridge”, “verse”, etc.
 - Filters out keywords that could contaminate the lyrics

jLyrics: Lyrical feature extractor

- **Extracts features** from lyrics stored in text files:
 - Automated Readability Index
 - Average Syllable Count Per Word
 - Contains Words
 - Flesh-Kincaid Grade Level
 - Flesh Reading Ease
 - Function Word Frequencies
 - Letter-Bigram Components
 - Letter Frequencies
 - Letters Per Word Average
 - Letters Per Word Variance
 - Lines Per Segment Average
 - Lines Per Segment Variance
 - Number of Lines
 - Number of Segments
 - Number of Words
 - Part-of-Speech Frequencies
 - Punctuation Frequencies
 - Rate of Misspelling
 - Sentence Count
 - Sentence Length Average
 - Topic Membership Probabilities
 - Vocabulary Richness
 - Vocabulary Size
 - Word Profile Match
 - Words Per Line Average
 - Words Per Line Variance
- Can also automatically generate **word frequency profiles** for particular classes of training data is provided

ACE: Meta-learning engine

- Evaluates the **relative suitability** of different dimensionality reduction and classification algorithms for a given problem
 - Can also train and classify with manually selected algorithms
- Evaluates algorithms in terms of
 - Classification accuracy
 - Consistency
 - Time complexity
- Based on the Weka framework, so new algorithms can be added easily

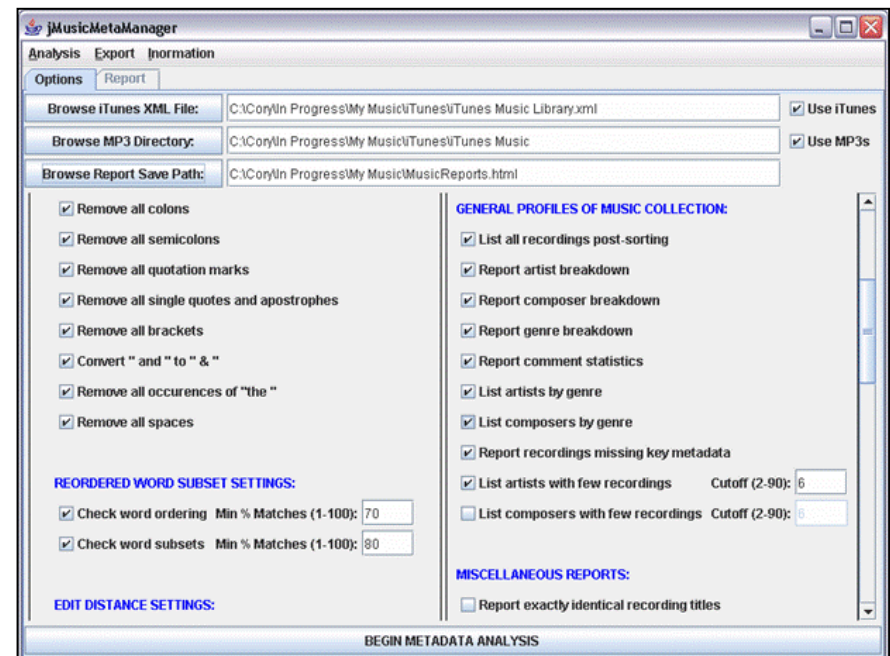


jMIR datasets

- **Codaich** is an MP3 research set
 - Carefully cleaned and labelled
 - The published 2006 version has 26,420 recordings
 - Belonging to 55 genres
 - Is constantly growing: currently over 45,000 MP3s
- **Bodhidharma MIDI** has 950 MIDI recordings
 - 38 genres of music
- **SLAC** consists of 250 matched audio recordings, MIDI recordings, lyrical transcriptions and metadata that can be used to extract cultural features
 - Useful for experiments on combining features from different types of data
 - 10 genres of music (in 5 pairs of similar genres)

jMusicMetaManager: Dataset manager

- Detects metadata errors/inconsistencies and redundant copies of recordings
- Detects differing metadata values that should in fact be the same
 - e.g. “Charlie Mingus” vs. “Mingus, Charles”
- Generates HTML inventory and profile reports
 - 39 reports in all
- Parses metadata from ID3 tags and iTunes XML



jSongMiner: Metadata miner

- Software for automatically acquiring formatted metadata about **songs, artists** and **albums**
- Designed for use with the **Greenstone** digital library software
 - May also be used for other purposes, such as cultural feature extraction
- Identifies music files
 - Uses Echo Nest **fingerprinting** functionality and **embedded metadata**
- Mines a wide range of metadata tags from the Internet and collates them in a standardized way
 - Data extracted from **The Echo Nest, Last.FM, MusicBrainz**, etc.
 - Over 100 different fields are extracted
 - Data may be formatted into unqualified and/or qualified **Dublin Core** fields if desired
- Saves the results in ACE XML or text
 - Can also be integrated automatically into a Greenstone collection

ACE XML: MIR research file formats

- Standardized file formats that can represent:
 - Feature values extracted from instances
 - Abstract feature descriptions and parameterizations
 - Instance labels and annotations
 - Class ontologies
- Designed to be flexible and extensible
 - Able to express types of information that are particularly pertinent to music
- Allow jMIR components to communicate with each other
 - Can also be adopted for independent use by other software
- ACE XML 2.0 provides even more expressivity
 - e.g. potential for integration into RDF ontologies

Previous collaborations (1/2)

- Vigliensoni, McKay and Fujinaga (2010)
 - Addition of Last.FM functionality to jWebMiner
 - Empirical comparison of different kinds of cultural data
- Angeles, McKay and Fujinaga (2010)
 - Addition of MusicBrainz functionality to jMusicMetaManager
 - Empirical comparison of curated, noisy and automatically cleaned metadata
- McKay, Burgoyne, Hockman, Smith, Vigliensoni and Fujinaga (2010)
 - Development of jLyrics and lyricFetcher
 - Empirical comparison of different feature types
- Thompson, McKay, Burgoyne and Fujinaga (2009)
 - Improvements to ACE

Previous collaborations (2/2)

- McKay, Burgoyne, Thompson and Fujinaga (2009)
 - Improvements to ACE XML
- McEnnis, McKay and Fujinaga (2006)
 - Improvements to jAudio
- Fiebrink, McKay and Fujinaga (2005)
 - An empirical investigation of dimensionality reduction using ACE (and other technologies)
- Sinyor, McKay, Fiebrink, McEnnis and Fujinaga (2005)
 - Beatboxing classification using ACE and jAudio

Paper ideas: Improve components

- jSongMiner
 - Add more data sources
- jAudio
 - Develop more features, especially psychologically meaningful features
 - Improve interface
- jSymbolic
 - Develop more features or implement the remaining feature catalogue
- jWebminer
 - Take advantage of additional web services (e.g. Amazon) to add more features
- jLyrics
 - Develop features especially relevant to music
- ACE
 - Add more machine learning algorithms
 - Especially unsupervised learning
- jMusicMetaManager
 - Add correction functionality

Paper ideas: Develop new components

- OMEN
 - Reimplement and get working
- jImage
 - Extract features from album art, press photos, etc.
- jVideo
 - Extract features from music videos, concert videos, etc.
- jMusicVisualiser
 - Offer visual ways of exploring relationships between musical instances, features and classes
- jStructure:
 - Automatically segment audio streams in time, both in terms of partitioning separate pieces of music within a single stream and in terms of structural divisions within individual pieces
- jClassOntology
 - Use data mining to harvest class ontologies

Paper ideas: Apply jMIR

- There are many problems to which the jMIR components could be applied
 - Either directly or with specialized improvements (e.g. features developed especially for chord recognition)
- Consider the many MIREX application areas
 - www.music-ir.org/mirex/wiki/MIREX_HOME
- More ideas:
 - See the Future Research sections of my papers and dissertation

More information

- Overview, documentation and publications
 - jmir.sourceforge.net
- Code and (in most cases) manuals
 - sourceforge.net/projects/jmir/
 - sourceforge.net/projects/jaudio/
- My dissertation on jMIR
 - www.music.mcgill.ca/~cmckay/papers/musictech/mckay10dissertation.pdf
 - Other jMIR-related publications can also be found on my web page: www.music.mcgill.ca/~cmckay/
- E-mail
 - cory.mckay@mail.mcgill.ca

Acknowledgements

- Collaborators at McGill and Waikato Universities
- Funding, past and present:
 - CIRMMT
 - Centre for Open Software Innovation
 - Andrew W. Mellon Foundation
 - Social Sciences and Humanities Research Council of Canada
 - Canadian Foundation for Innovation
- Contact information:
 - jmir.sourceforge.net
 - cory.mckay@mail.mcgill.ca



Social Sciences and Humanities
Research Council of Canada

Conseil de recherches en
sciences humaines du Canada

Canada



McGill



Schulich School of Music
École de musique Schulich



Centre for Interdisciplinary Research
in Music Media and Technology

DDMAL



Centre for
Open Software
Innovation



More details?

- **jAudio**: Audio feature extraction
- **jSymbolic**: Symbolic feature extraction
- **jWebMiner**: Cultural feature extraction
- **lyricFetcher** and **jLyric**: Lyric harvesting and feature extraction
- **ACE**: Meta-learning classification engine
- **ACE XML**: File formats
 - Features, feature metadata, instance metadata, ontologies
- **Codaich**, **Bodhidharma MIDI** and **SLAC**: datasets
- **jMusicMetaManager** and **jSongMiner**: Metadata management and harvesting

- General questions?