

jMIR Overview

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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





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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



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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 nontechnical users

Facilitates multimodal research







jMIR components

Audio: Audio feature extraction Symbolic: Feature extraction from MIDI files WebMiner: 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 SongMiner: Metadata harvesting MusicMetaManager: Metadata management MIRUtilities: 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.

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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

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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" cooccur 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

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lyricFetcher: Lyric miner

lyricFetcher automatically harvests lyrics from online 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 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 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





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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







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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 Eujin

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

jlmage

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



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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/mck ay10dissertation.pdf

Other jMIR-related publications can also be found on my web page: www.music.mcgill.ca/~cmckay/

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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?



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