

Introduction to Audio Content Analysis

Module 13: Music Similarity

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corresponding textbook section

Section 13

overview

lecture content

- music similarity and its relation to musical genre
- clustering and visualization of feature space

learning objectives

- describe potential issues with algorithms for measuring music similarity
- understand common approaches for timbre similarity



Module 13: Music Similarity 1 /

corresponding textbook section

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Module 13: Music Similarity 1 /

2 / 9

- **perception** of music similarity
 - multi-dimensional (melodic, rhythmic, sound quality, ...)
 - user dependent
 - associative, may also depend on editorial data
 - may be context dependent
- genres are clusters of musical similarity
- ⇒ genre classification is a *special case* of audio similarity measures
- instead of assigning (genre) labels, the similarity/distance between (pairs) of files is measured

Module 13: Music Similarity

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Module 13: Music Similarity 2 / 9

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Module 13: Music Similarity 2 / 9

- commonalities with genre classification
 - similar set of features
 - ambiguous 'ground truth'
 - unclear value/impact of low level and high level features
- differences to genre classification
 - distance/similarity measure instead of categorical classification
- tasks
 - playlist generation
 - music discovery

music similarity

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music similarity introduction

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■ standard approach

- define set of features and aggregate them appropriately
- use vector distance in latent space as similarity approximation

■ modern approach

- train 'meaningful' representation (e.g., VAE etc.)
- use vector distance in latent space as similarity approximation

audio similarity visualization in a 2D space

problem

- feature space is high-dimensional
- \rightarrow cannot be visualized
- find **mapping** to 2D "preserving" (high-dimensional) distance metrics example:
 - Self-Organizing Maps

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- visualization example: SOM 1/2
 - 1 create a map with 'neurons'
 - 2 train
 - for each training sample find BMU (best matching unit)
 - adapt BMU and neighbors toward training sample

$$W_{\nu}(t+1) = W_{\nu}(t) + \theta(u, \nu, t)\alpha(t)(D(t) - W_{\nu}(t))$$

- \blacktriangleright $\theta(u, v, t)$: depends on neighborhood distance from BMU
- $ightharpoonup \alpha(t)$: learning restraint
- $\triangleright D(t)$ training sample



audio similarity SOM 2/2





 $from^1$

 $^{^{1}\}mathrm{E.}$ Pampalk, "Islands of Music," Diploma Thesis, Technische Universität Wien, 2001.



- as the task is not clearly defined, it often can only be indirectly evaluated
 - evaluation through genre/artist/album labels
 - evaluation through listening surveys
 - qualitative evaluation (picking pairs of samples and discussing them)
- lack of ground truth and lack of established metrics

- **■** music similarity
 - even less clearly defined than music genre
- processing steps
 - 1 extract features
 - 2 define some distance metric in feature space
- clustering algorithms
 - work to a certain degree with traditional features

