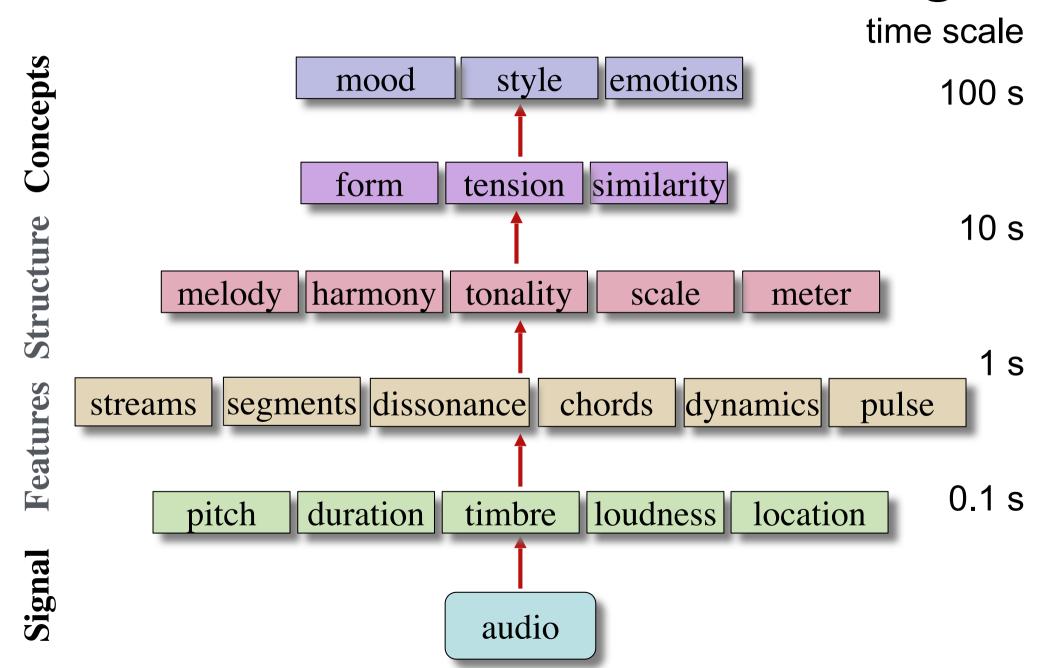
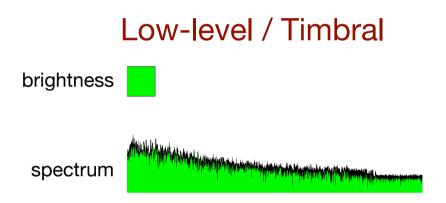
Why is window length important (perceptually)?

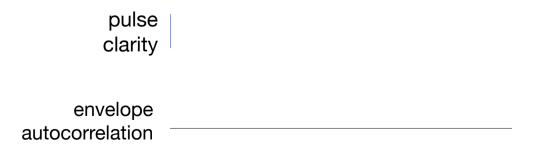
Levels of Music Processing



Musical features: Examples

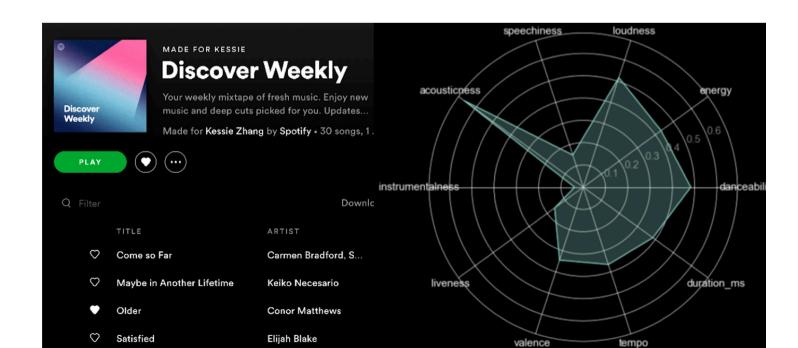


Mid/High-level / Rhythmic



Feature Extraction from music

- features in music evolve continuously
- feature extraction relies on summarising this evolution (means, std)

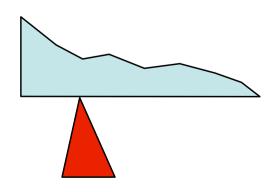


Which sounds brighter? (spectral centroid)



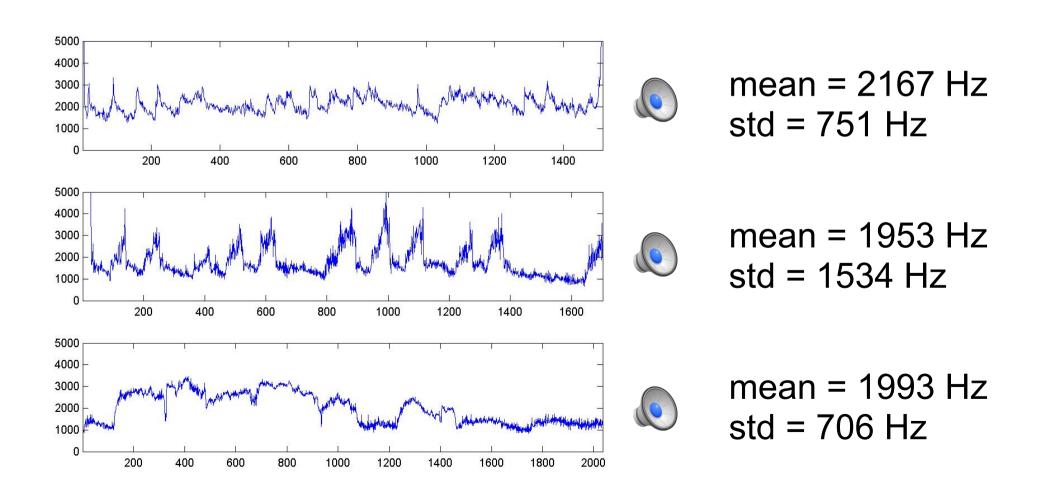






Example of frame-based analysis

spectral centroid of three excerpts



Features Overview

- Dynamics
- Pitch
- Timbre
- Tempo/rhythm
- Tonality
- Structure

Timbre

Acoustic features

Temporal

- zero-crossing rate
- low energy

Spectrotemporal

- spectral flux
- sub-band flux

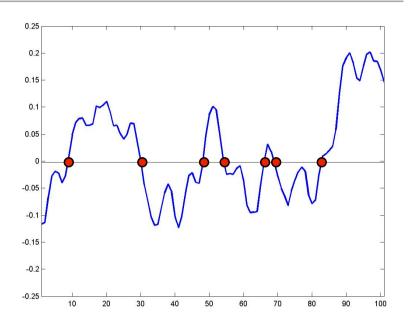
Spectral

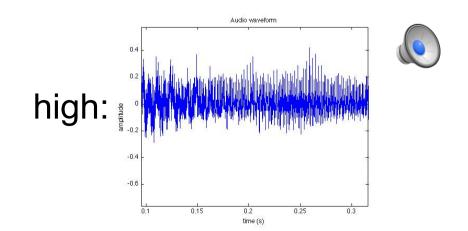
- centroid
- high energy-low energy ratio
- entropy
- roll-off 85
- MFCC

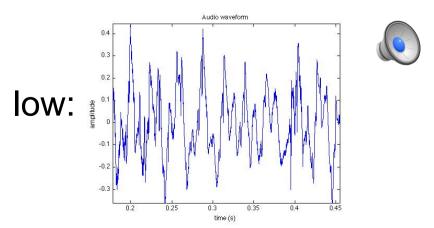
Identify features that might be useful for genre classification based on perceptual relevance

Zero-crossing rate

 number of time-domain zero-crossings of the signal per time unit



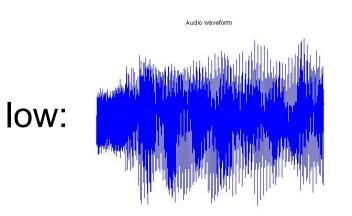




Low Energy

proportion of signal frames whose energy is below average energy

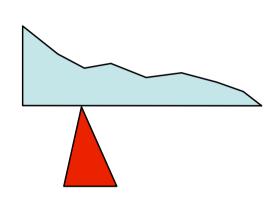


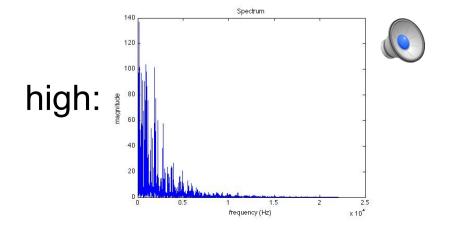


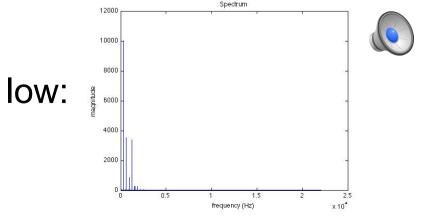
Spectral Centroid

Center of mass of the spectrum

$$sc = \frac{\sum a_i f_i}{\sum a_i}$$





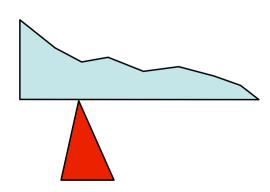


Which sounds brighter?



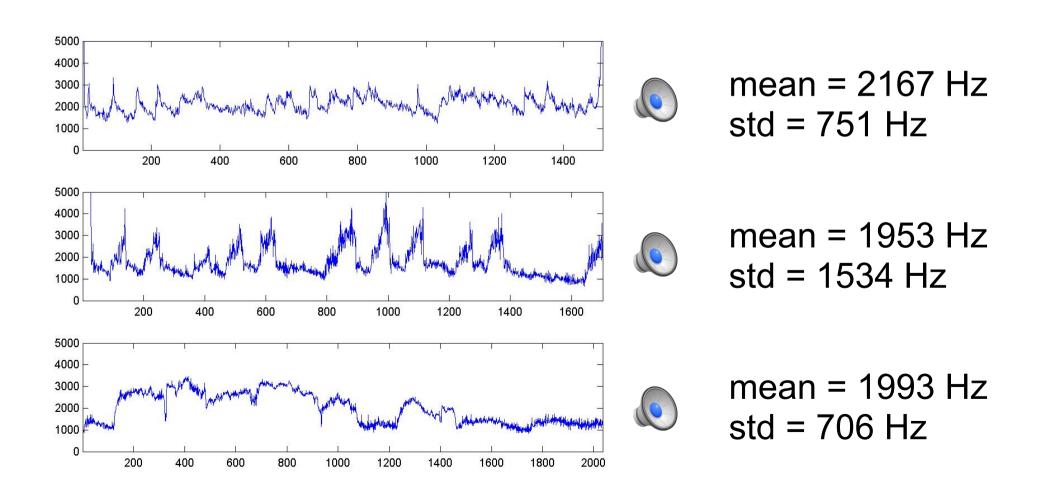






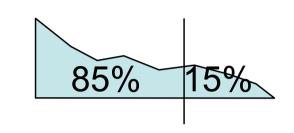
Example of frame-based analysis

spectral centroid of three excerpts



Spectral Roll-Off

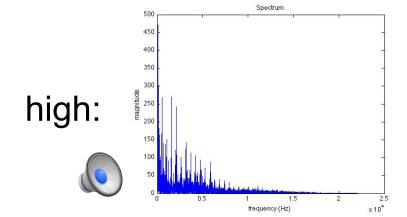
 Frequency, below which a certain fraction (usually 85%) of spectral energy

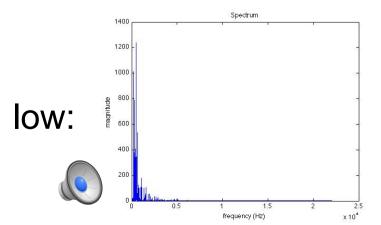


R such that

$$\sum_{1}^{R} a_i = 0.85 \sum_{1}^{N} a_i$$

Measure of spectral shape

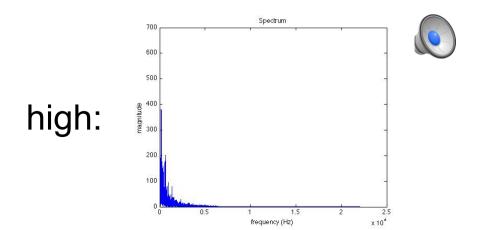


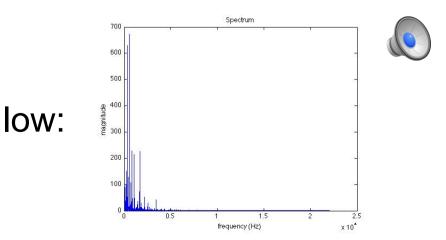


Spectral Irregularity

measure of "jaggedness" of spectrum (Jensen, 1999)

$$irreg = \frac{\sum_{2}^{N} (a_i - a_{i-1})^2}{\sum_{1}^{N} a_i^2}$$

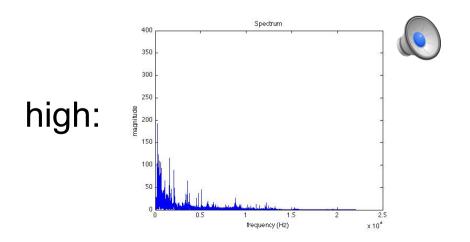


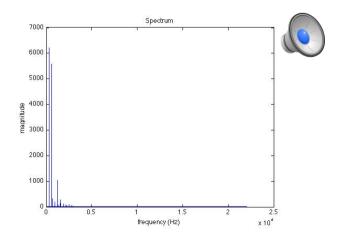


Spectral Entropy

information-theoretic measure of spectral energy distribution

 high entropy = even distribution of spectral energy (more noise-like?)



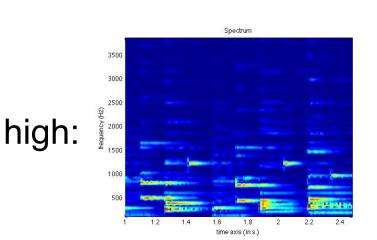


low:

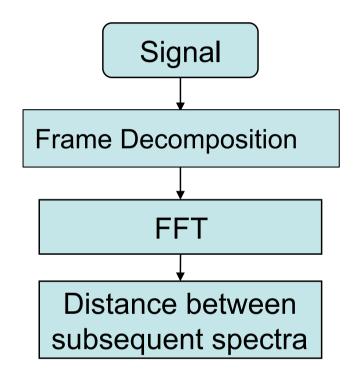
Spectral Flux

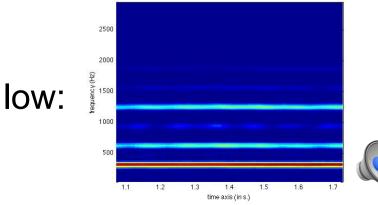
- Measure of change over time in spectrum
- Dissimilarity between subsequent spectral frames

$$flux_i = \sum_{j=1}^{M} (a_{ij} - a_{(i-1)j})^2$$



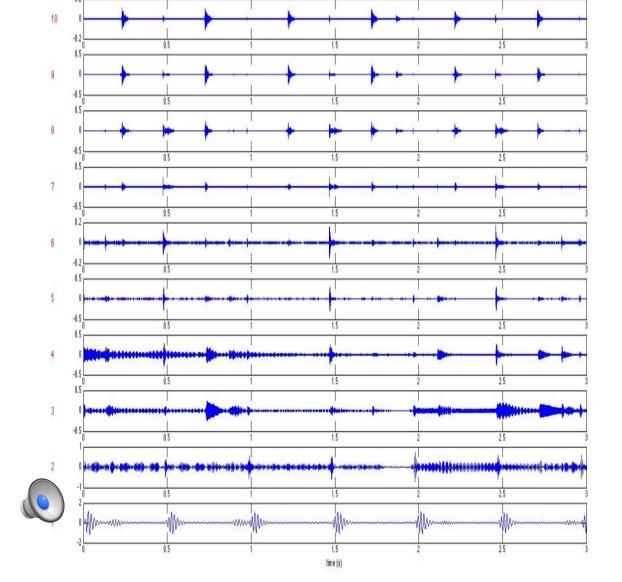






Sub-band Flux

- Octave-scaled spectrum
 - 50 hz
 - 10 bands
- Spectral Flux in each band



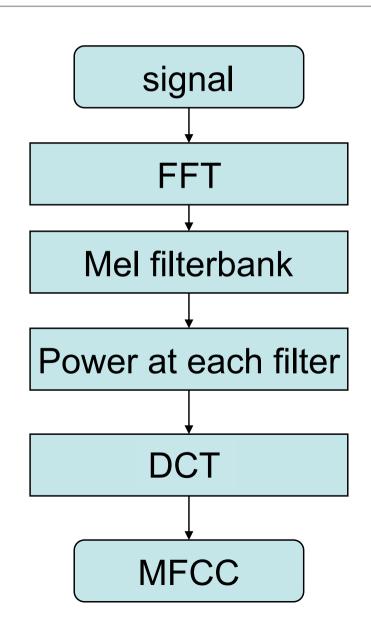
Audio waveform

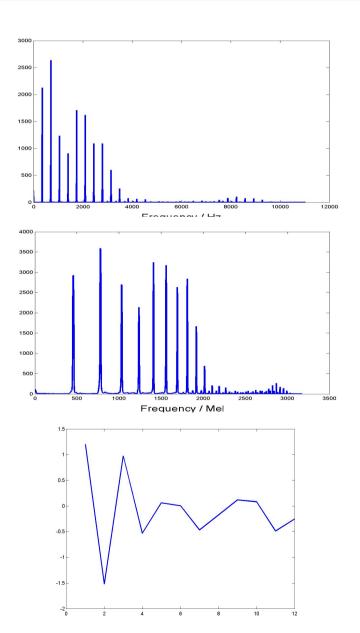


Mel-Frequency Cepstral Coefficients

- Descriptor of spectral shape based on perception
- widely used in speech research (e.g. speech recognition)

Mel-frequency Cepstral Coefficients





Significance of MFCC

 provide a representation of the sound spectrum that closely corresponds to perceived distances between timbres (DePoli and Prandoni, 1997; Eronen, 2001; Terasawa et al., 2005)

similarity in MFCC <-> similarity in perceived timbre

important in classification of genre, mood, emotion, semantics

Genre Classification

Identify features that might be useful for genre classification based on perceptual relevance:

Temporal

- zero-crossing rate
- low energy

Spectrotemporal

- roughness
- sub-band flux

Spectral

- centroid
- high energy-low energy ratio
- entropy
- roll-off 85
- MFCC



Acoustic features

Rhythm

- tempo
- pulse clarity

Tonality

- chromagram
- mode
- keystrength/keyclarity

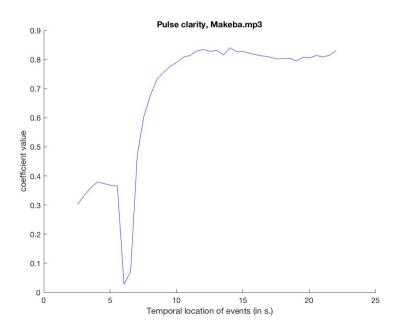
*typically extracted using longer time windows (contextdependent)



Tempo & Pulse Clarity

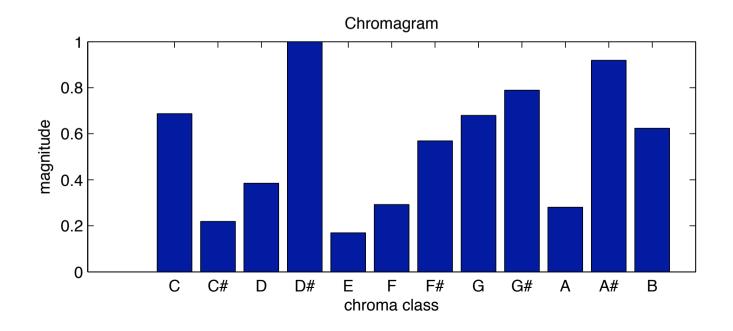
 tempo: estimate of how fast/slow the piece of music is

 pulse clarity/beat salience: how clear the beat is



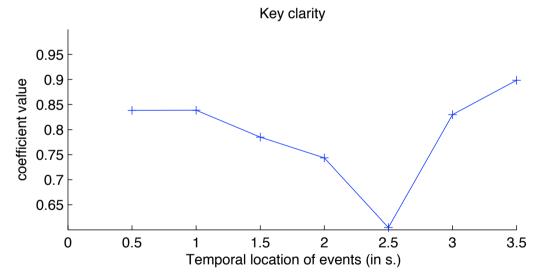
Acoustic features

 chromagram: Harmonic Pitch Class Profile, shows the distribution of energy along the pitches or pitch classes.



Acoustic features

keystrength: measure of the tonal clarity



mode: major or minor (roughly depicts "happy" or "sad")