

Advanced Course Computer Science

## Music Processing

Summer Term 2010

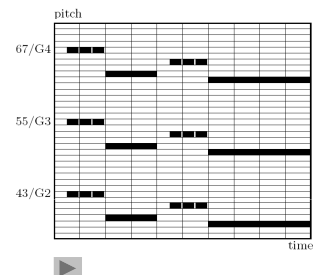
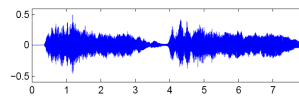
**Meinard Müller**

Saarland University and MPI Informatik  
meinard@mpi-inf.mpg.de

## Music Synchronization



## Music Data



## Music Data

Various interpretations – Beethoven's Fifth

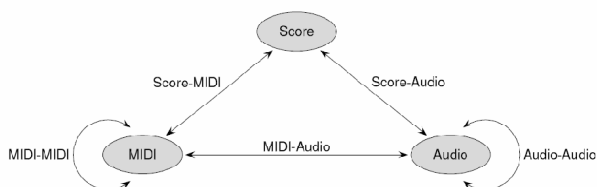
Bernstein	▶
Karajan	▶
Scherbakov (piano)	▶
MIDI (piano)	▶

## General Goals

- Automated organization of complex and inhomogeneous music collections
- Generation of annotations and cross-links
- Tools and methods for multimodal search, navigation and interaction

## Music Information Retrieval (MIR)

## Music Synchronization



Schematic view of various synchronization tasks

## Music Synchronization (Audio Alignment)

- Turetsky/Ellis (ISMIR 2003)
- Soulez/Rodet/Schwarz (ISMIR 2003)
- Arifi/Clausen/Kurth/Müller (ISMIR 2003)
- Hu/Dannenberg/Tzanetakis (WASPAA 2003)
- Müller/Kurth/Röder (ISMIR 2004)
- Raphael (ISMIR 2004)
- Dixon/Widmer (ISMIR 2005)
- Müller/Mattes/Kurth (ISMIR 2006)
- Dannenberg /Raphael (Special Issue ACM 2006)
- Kurth/Müller/Fremerey/Chang/Clausen (ISMIR 2007)
- Fujihara/Goto (ICASSP 2008)
- Wang/Iskandar/New/Shenoy (IEEE-TASLP 2008)
- Ewert/Müller/Grosche (ICASSP 2009)

## Music Synchronization: Audio-Audio

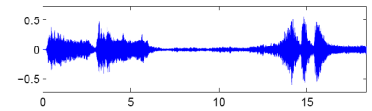
**Given:** Two different audio recordings of the same underlying piece of music.

**Goal:** Find for each position in one audio recording the **musically** corresponding position in the other audio recording.

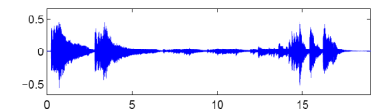
## Music Synchronization: Audio-Audio

### Beethoven's Fifth

Karajan



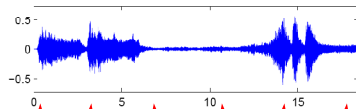
Scherbakov



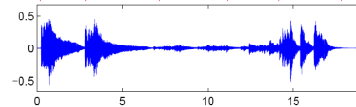
## Music Synchronization: Audio-Audio

### Beethoven's Fifth

Karajan



Scherbakov

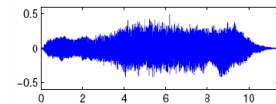


Synchronization: Karajan → Scherbakov

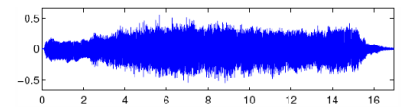
## Music Synchronization: Audio-Audio

### Bach Toccata

Koopman



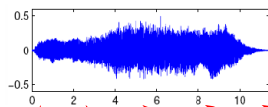
Ruebsam



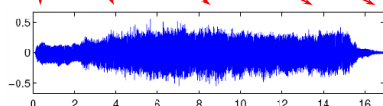
## Music Synchronization: Audio-Audio

### Bach Toccata

Koopman



Ruebsam



Synchronization: Koopman → Ruebsam

## Music Synchronization: Audio-Audio

- Transformation of audio recordings into sequences of **feature vectors**
  - $\rightsquigarrow V := (v^1, v^2, \dots, v^N)$
  - $\rightsquigarrow W := (w^1, w^2, \dots, w^M)$
- Fix **cost measure**  $c$  on the feature space
- Compute  $N \times M$  **cost matrix**  $C(n, m) := c(v^n, w^m)$
- Compute cost-minimizing warping path from  $C$

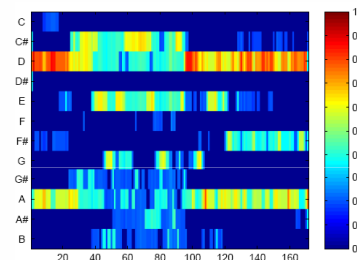
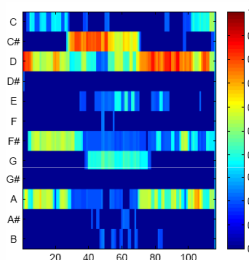
## Chroma Features

Example: Bach Toccata

Koopman



Ruebsam



Feature resolution: 10 Hz

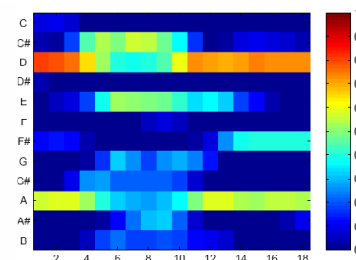
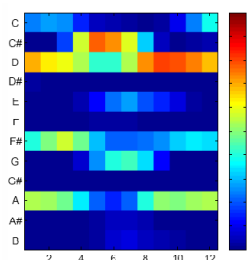
## Chroma Features

Example: Bach Toccata

Koopman



Ruebsam

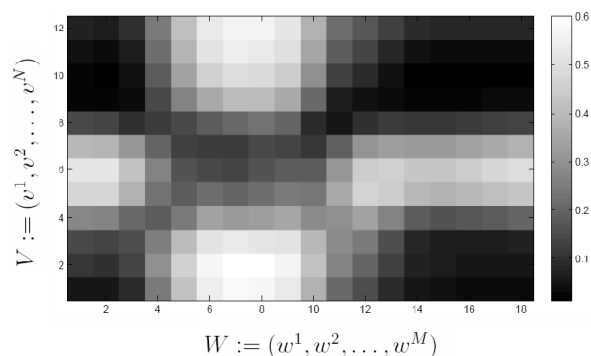


Feature resolution: 1 Hz

## Music Synchronization: Audio-Audio

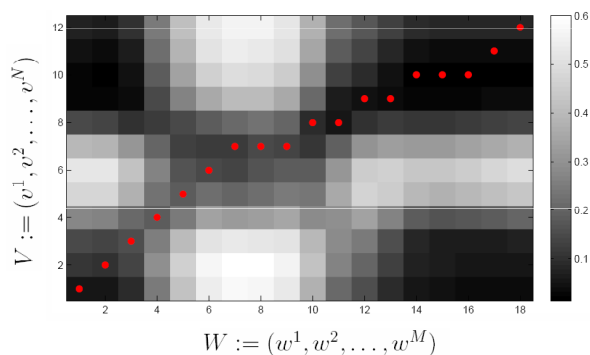
- Koopman  $\rightsquigarrow V := (v^1, v^2, \dots, v^N)$   $N = 12$
- Ruebsam  $\rightsquigarrow W := (w^1, w^2, \dots, w^M)$   $M = 18$
- $v^n, w^m = 12$ -dimensional normalized chroma vectors
- Local cost measure  $c: \mathbb{R}^{12} \times \mathbb{R}^{12} \rightarrow \mathbb{R}$   
 $c(v^n, w^m) := 1 - \langle v^n, w^m \rangle$
- $N \times M$  cost matrix  $C(n, m) := c(v^n, w^m)$

## Music Synchronization: Audio-Audio



## Music Synchronization: Audio-Audio

Cost-minimizing warping path



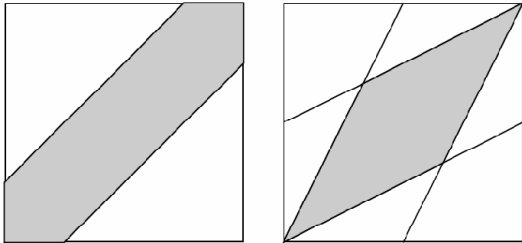
## Cost-Minimizing Warping Path

- Computation via dynamic programming  
 $\rightsquigarrow$  Dynamic Time Warping (DTW)
- Memory requirements and running time:  $O(NM)$
- **Problem: Infeasible for large  $N$  and  $M$**
- Example: Feature resolution 10 Hz, pieces 15 min  
 $\Rightarrow N, M \sim 10,000$   
 $\Rightarrow N \cdot M \sim 100,000,000$

### Strategy: Global Constraints

Sakoe-Chiba band

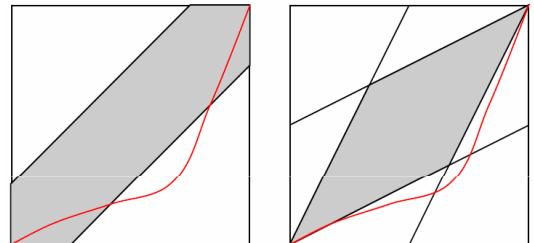
Itakura parallelogram



### Strategy: Global Constraints

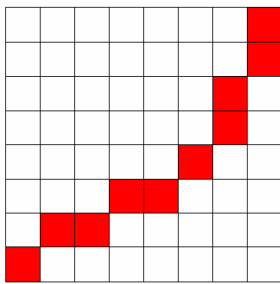
Sakoe-Chiba band

Itakura parallelogram



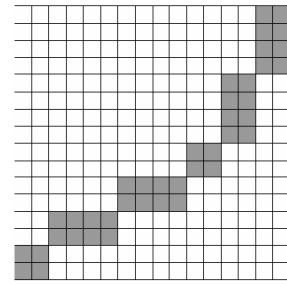
Problem: Optimal warping path not in constraint region

### Strategy: Multiscale Approach



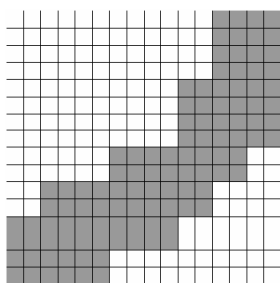
Compute optimal warping path on coarse level

### Strategy: Multiscale Approach



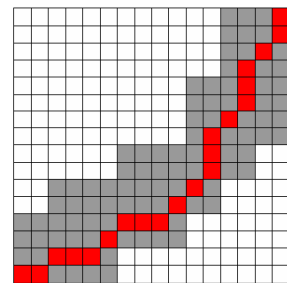
Project on fine level

### Strategy: Multiscale Approach



Specify constraint region

### Strategy: Multiscale Approach



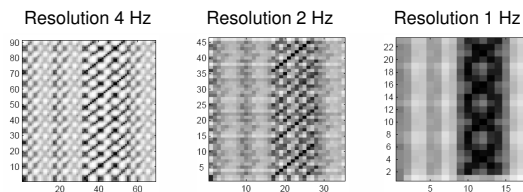
Compute *constrained* optimal warping path

## Strategy: Multiscale Approach

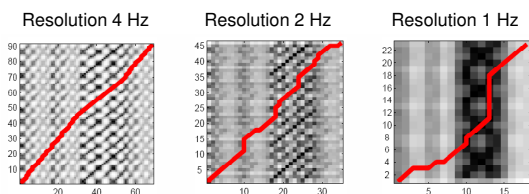
- Suitable features?
- Suitable resolution levels?
- Size of constraint regions?

Good trade-off between efficiency and robustness?

## Strategy: Multiscale Approach



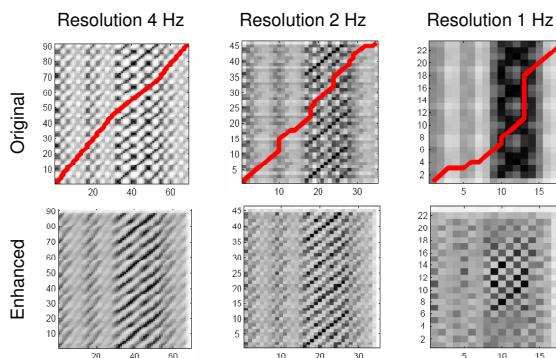
## Strategy: Multiscale Approach



Problem: Cost matrix may degenerate  
 ↪ useless warping path

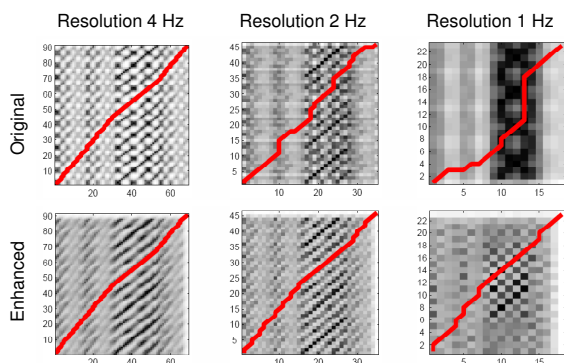
## Strategy: Multiscale Approach

Improve robustness by enhancing cost matrix



## Strategy: Multiscale Approach

Improve robustness by enhancing cost matrix



## Strategy: Multiscale Approach

Chroma features at three levels: 0.33 Hz / 1 Hz / 10 Hz

Recording 1	length [sec]	Recording 2	length [sec]	$t_{DTW}$ [sec]	$t_{MsDTW}$ [sec]	[%]
Beet9Bern	1144.9	Beet9Kar	1054.8	31.18	1.08	3.46

## Strategy: Multiscale Approach

Chroma features at three levels: 0.33 Hz / 1 Hz / 10 Hz

Recording 1	length [sec]	Recording 2	length [sec]	$t_{DTW}$ [sec]	$t_{MsDTW}$ [sec]	[%]
Beet9Bern	1144.9	Beet9Kar	1054.8	31.18	1.08	3.46

Number of matrix entries needed for DTW and MsDTW:

	DTW	MsDTW	%
Level 1	120,808,050	2,117,929	1.75
Level 2	1,209,030	17,657	1.46
Level 3	134,464	134,464	100

## Music Synchronization: Audio-Audio

### Conclusions

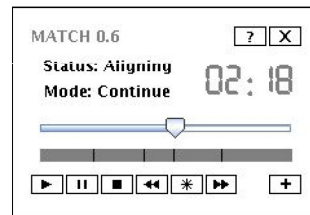
- Chroma features
  - ↔ suited for harmony-based music
- Relatively coarse but good global alignments
- Multiscale approach: simple, robust, fast

## Music Synchronization: Audio-Audio

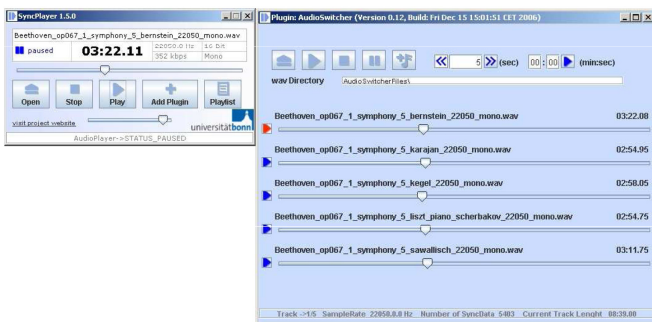
### Applications

- Efficient music browsing
- Blending from one interpretation to another one
- Mixing and morphing different interpretations
- Tempo studies

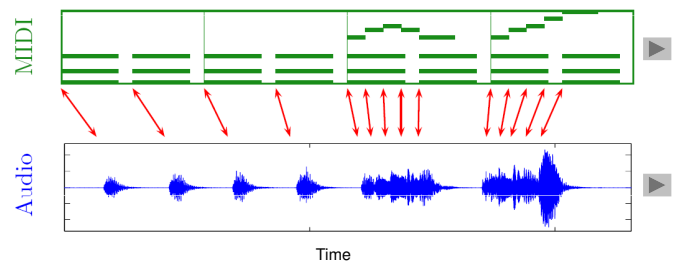
## System: Match (Dixon)



## System: SyncPlayer/AudioSwitcher



## Music Synchronization: MIDI-Audio



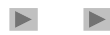
## Music Synchronization: MIDI-Audio

MIDI = meta data

Automated annotation

Audio recording

Sonification of annotations



## Music Synchronization: MIDI-Audio

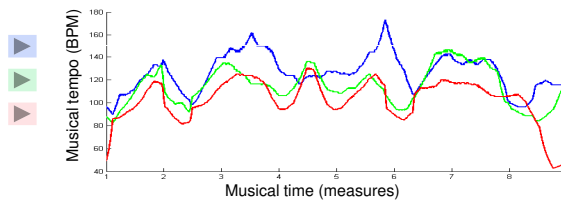
MIDI = reference (score)

Tempo information

Audio recording

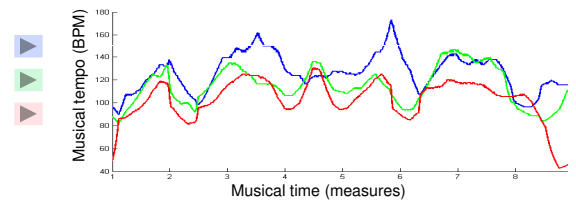
## Performance Analysis: Tempo Curves

Schumann:  
Träumerei



## Performance Analysis: Tempo Curves

What can be done if no reference is available?

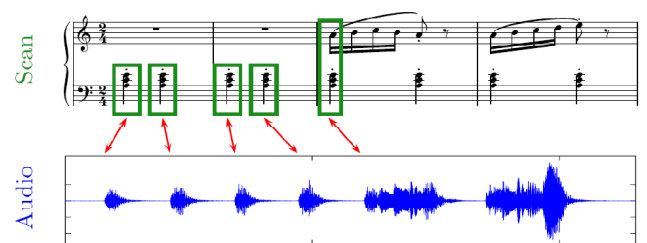


## Music Synchronization: MIDI-Audio

### Applications

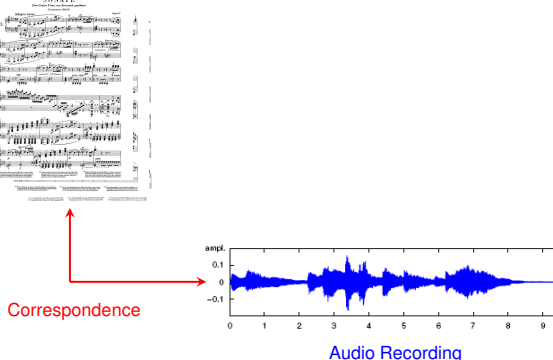
- Automated audio annotation
- Accurate audio access after MIDI-based retrieval
- Automated tracking of MIDI note parameters during audio playback
- Performance Analysis

## Music Synchronization: Scan-Audio



## Music Synchronization: Scan-Audio

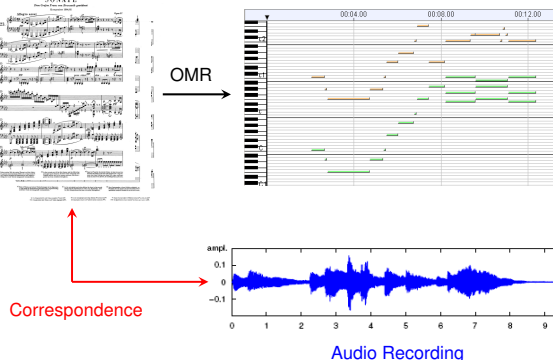
Scanned Sheet Music



## Music Synchronization: Scan-Audio

Scanned Sheet Music

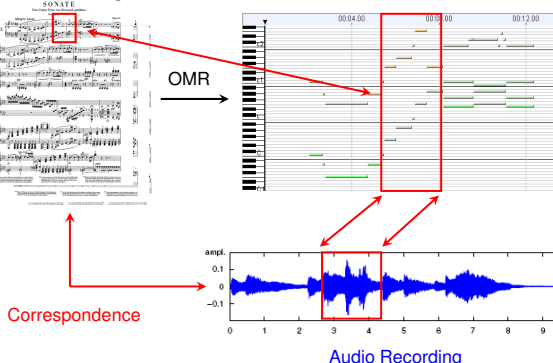
Symbolic Note Events



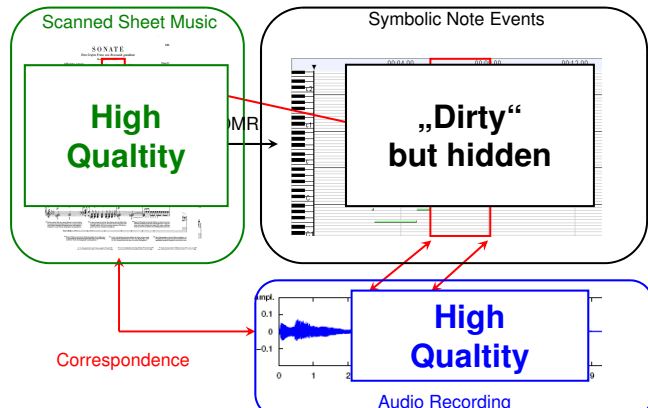
## Music Synchronization: Scan-Audio

Scanned Sheet Music

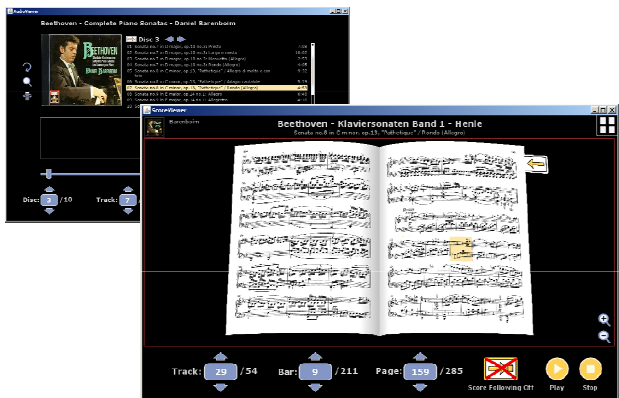
Symbolic Note Events



## Music Synchronization: Scan-Audio

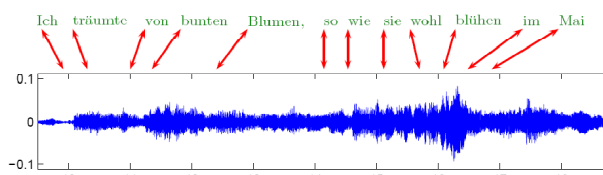


## Application: Score Viewer



[ECDL 08, ICM1 08]

## Music Synchronization: Lyrics-Audio

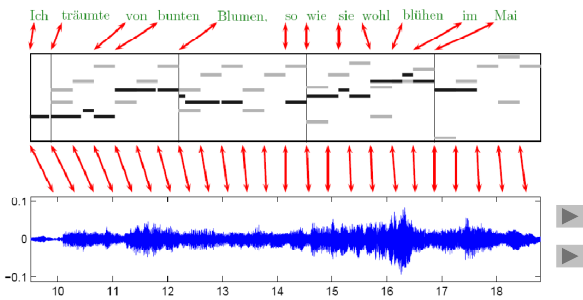


Difficult task!



## Music Synchronization: Lyrics-Audio

Lyrics-Audio → Lyrics-MIDI + MIDI-Audio



## System: SyncPlayer/LyricsSeeker



## High-Resolution Music Synchronization

- Normalized chroma features
  - robust to changes in instrumentation and dynamics
  - robust synchronization of reasonable overall quality
- Drawback: low temporal alignment accuracy
- Idea: Integration of note onset information

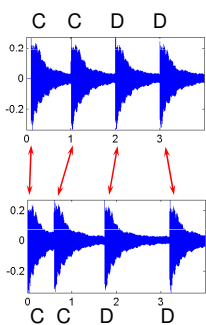
## High-Resolution Music Synchronization

- Normalized chroma features
  - robust to changes in instrumentation and dynamics
  - robust synchronization of reasonable overall quality
- Drawback: low temporal alignment accuracy
- Idea: Integration of note onset information
- Example: MIDI-Audio synchronization

Chroma-Chroma:    
 Chroma-Chroma + onset information:  

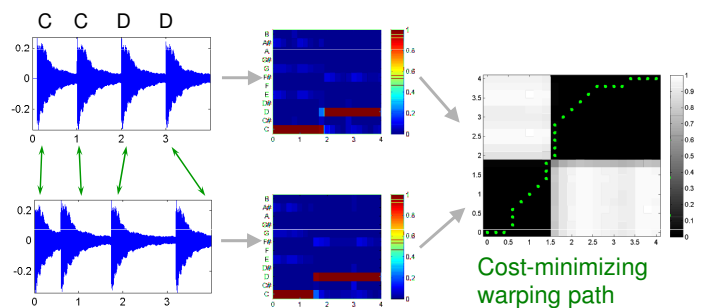
## High-Resolution Music Synchronization

Example: C – C – D – D



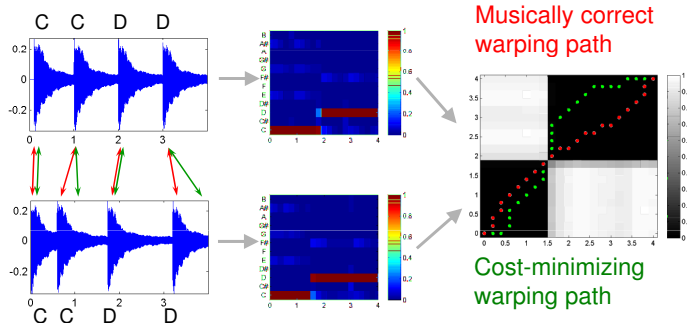
## High-Resolution Music Synchronization

Example: C – C – D – D



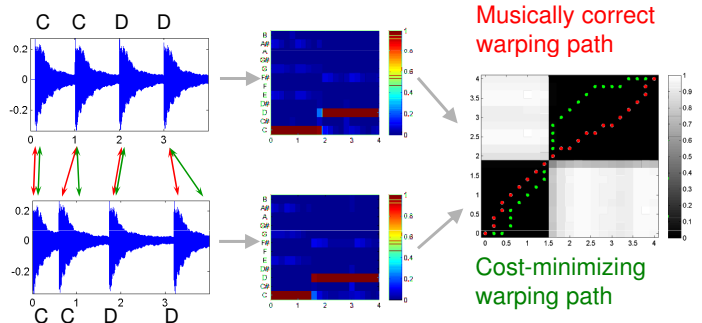
## High-Resolution Music Synchronization

Example: C – C – D – D



## High-Resolution Music Synchronization

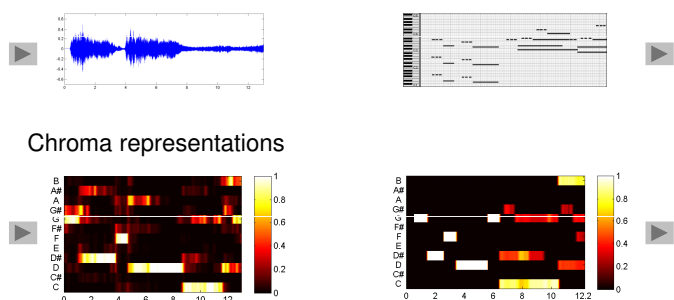
Example: C – C – D – D



Problem: note onsets are not captured in feature representation

## High-Resolution Music Synchronization

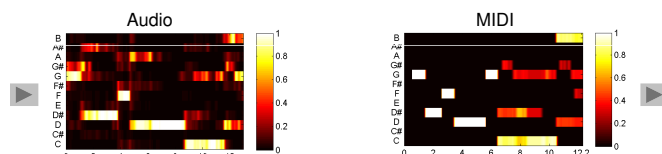
Example: Beethoven's Fifth



Problem: note onsets are not captured in feature representation

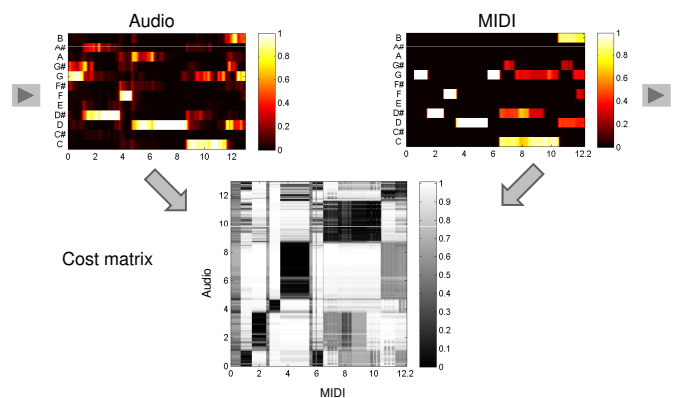
## High-Resolution Music Synchronization

Example: Beethoven's Fifth



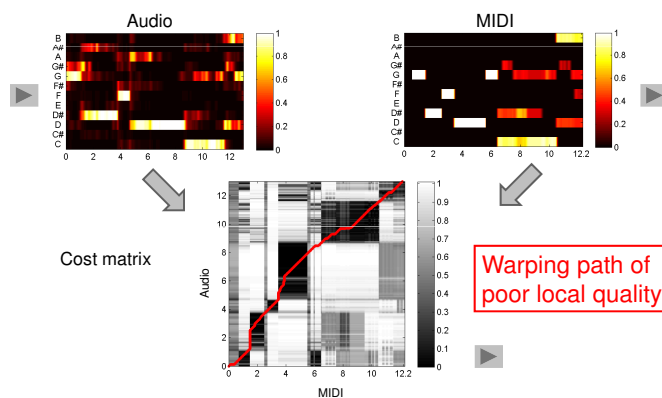
## High-Resolution Music Synchronization

Example: Beethoven's Fifth



## High-Resolution Music Synchronization

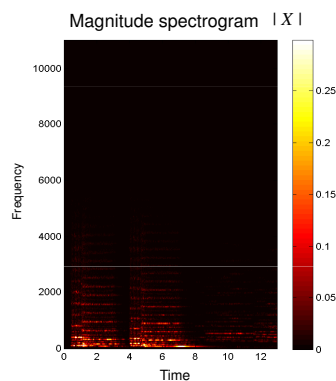
Example: Beethoven's Fifth



## Onset Detection

- General goal: Detection of onsets of musical notes
- Typical signal properties at note onset positions:
  - increase in energy
  - change of pitch
  - change of spectral content
  - high frequency content
- Idea: locate note onset candidates by measuring changes in spectral content

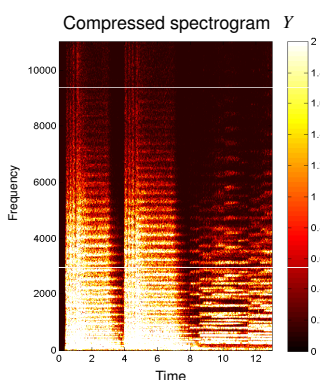
## Onset Detection



### Steps:

- Spectrogram

## Onset Detection



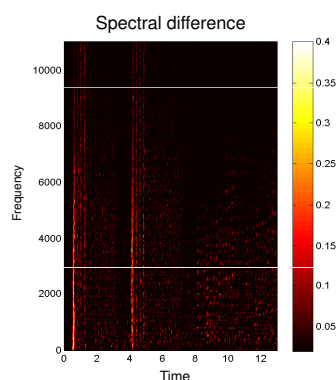
### Steps:

- Spectrogram
- Logarithmic compression

$$Y = \log(1 + C \cdot |X|)$$

- human sensation
- enhances low intensity values
- high frequency content
- reduces influence of amplitude modulation

## Onset Detection

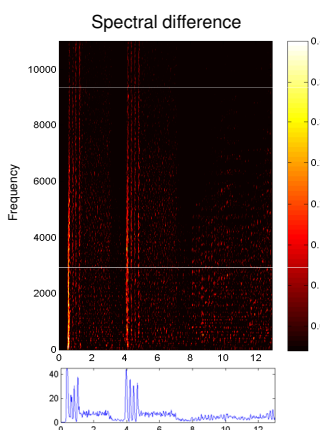


### Steps:

- Spectrogram
- Logarithmic compression
- Differentiation

- energy increase to be captured
- only positive values considered

## Onset Detection



### Steps:

- Spectrogram
- Logarithmic compression
- Differentiation
- Accumulation

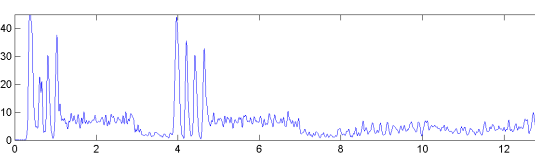
Novelty Curve

## Onset Detection

### Steps:

- Spectrogram
- Logarithmic compression
- Differentiation
- Accumulation

### Novelty Curve



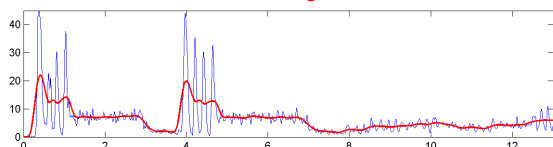
## Onset Detection

### Steps:

1. Spectrogram
2. Logarithmic compression
3. Differentiation
4. Accumulation
5. Normalization

Novelty Curve

Substraction of local average

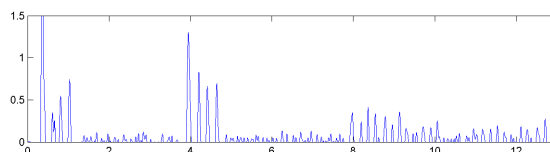


## Onset Detection

### Steps:

1. Spectrogram
2. Logarithmic compression
3. Differentiation
4. Accumulation
5. Normalization

Normalized novelty curve

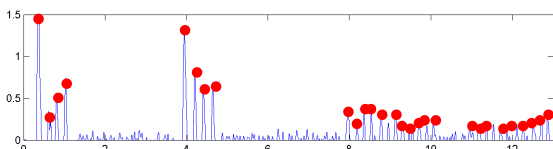


## Onset Detection

### Steps:

1. Spectrogram
2. Logarithmic compression
3. Differentiation
4. Accumulation
5. Normalization
6. Peak picking

Normalized novelty curve

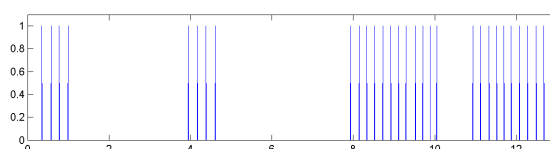


## Onset Detection

### Steps:

1. Spectrogram
2. Logarithmic compression
3. Differentiation
4. Accumulation
5. Normalization
6. Peak picking

Impulses

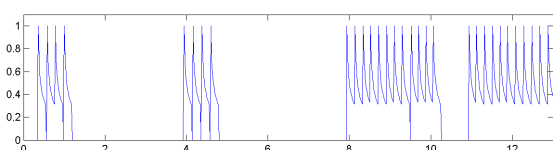


## Onset Detection

### Steps:

1. Spectrogram
2. Logarithmic compression
3. Differentiation
4. Accumulation
5. Normalization
6. Peak picking
7. Decay Filter

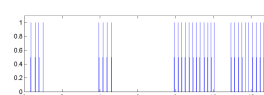
Decaying impulses



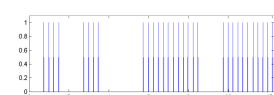
## High-Resolution Music Synchronization

Cost matrix based on impulses

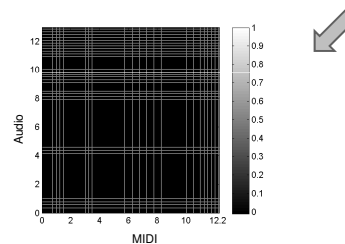
Audio



MIDI

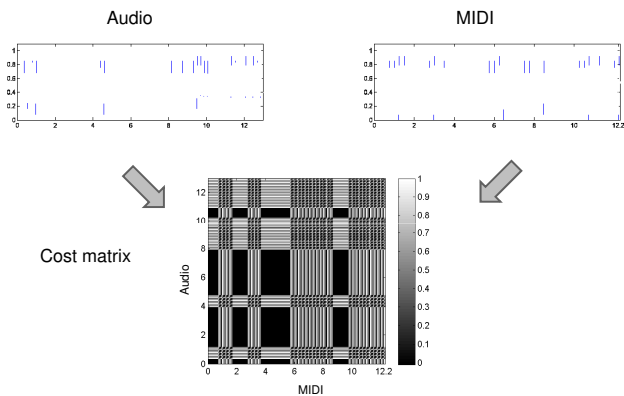


Cost matrix



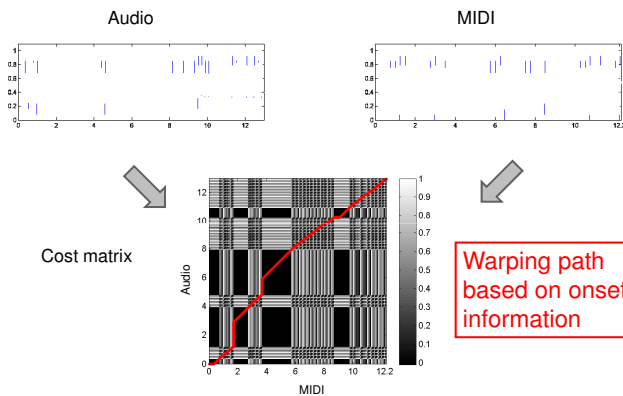
## High-Resolution Music Synchronization

Cost matrix based on decaying impulses



## High-Resolution Music Synchronization

Cost matrix based on decaying impulses



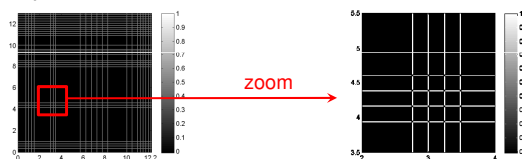
## High-Resolution Music Synchronization

Ideas:

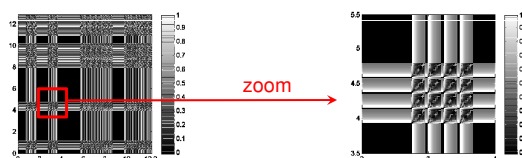
- Build up cost matrix with corridors of low cost
- Decaying strategy enforce corridor structure
- Each corridor corresponds to MIDI-audio pair of note onset candidates
- Warping path tends to run through corridors of low cost  
→ note onset positions are likely to be aligned

## High-Resolution Music Synchronization

Impulses

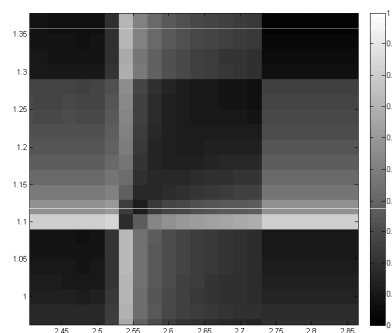


Decaying impulses



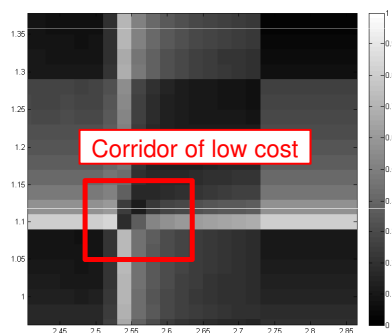
## High-Resolution Music Synchronization

Cost matrix for decaying impulses



## High-Resolution Music Synchronization

Cost matrix for decaying impulses

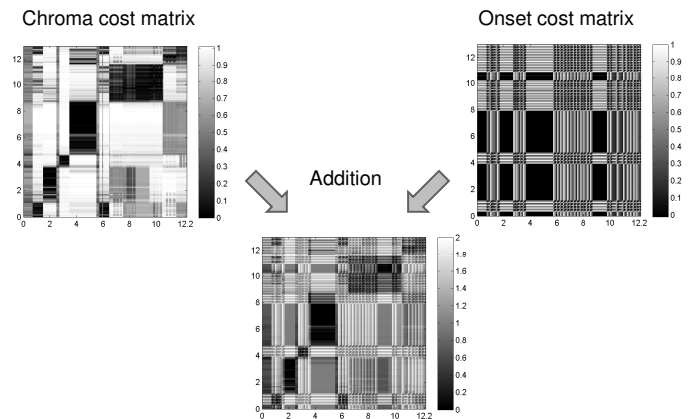


## High-Resolution Music Synchronization

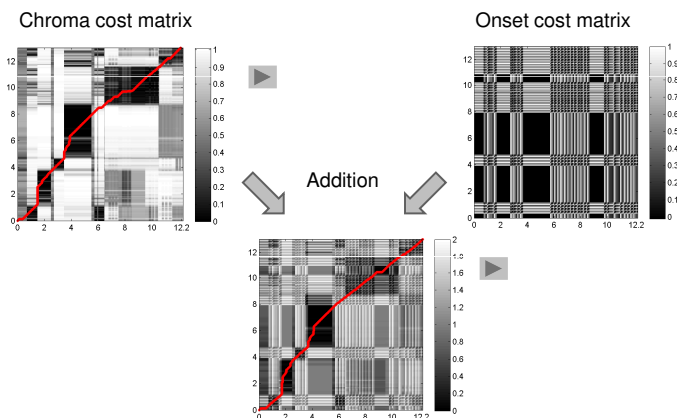
Combination of two different types of cost matrices:

- Cost matrix obtained from chroma features controls the global course of warping path
  - robust synchronization
- Cost matrix obtained from onset information controls the local course of warping path
  - accurate alignment

## High-Resolution Music Synchronization



## High-Resolution Music Synchronization



## Conclusions: Music Synchronization

### Various requirements

- Efficiency
- Robustness
- Accuracy
- Variability of music

## Conclusions: Music Synchronization

### Combination of various strategies

- Feature level
- Local cost measure level
- Global alignment level
- Evidence pooling using competing strategies

## Conclusions: Music Synchronization

### Offline vs. Online

- Online version: Dixon/Widmer (ISMIR 2005)
- Hidden Markov Models: Raphael (ISMIR 2004)
- Score-following
- Automatic accompaniment

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## Conclusions: Music Synchronization

### Presence of variations

- Instrumentation
- Musical structure
- Polyphony
- Musical key
- ...