
Lecture
Music Processing

Audio Structure Analysis

Meinard Müller

International Audio Laboratories Erlangen
meinard.mueller@audiolabs-erlangen.de

Music Structure Analysis

- Music segmentation
 - pitch content (e.g., melody, harmony)
 - music texture (e.g., timbre, instrumentation, sound)
 - rhythm
- Detection of repeating sections, phrases, motives
 - song structure (e.g., intro, versus, chorus)
 - musical form (e.g., sonata form, rondo form)
- Detection of other hidden relationships

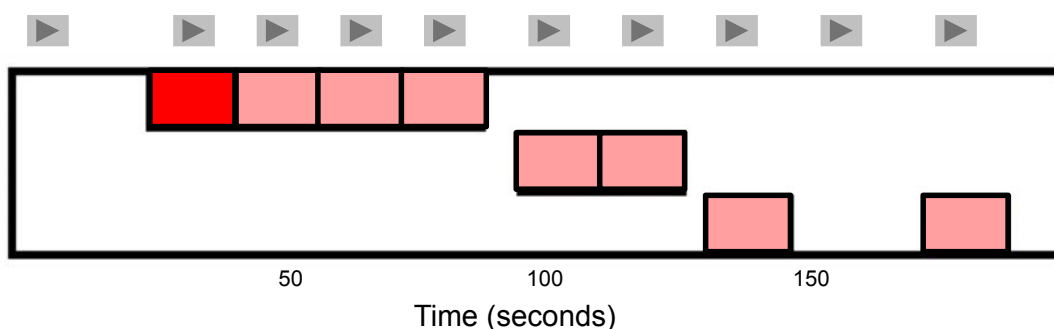
Repetition-Based Audio Structure Analysis

- Extract the **repetitive structure** of a given audio recording
- Often corresponds to **musical form** of the underlying piece
- The **thumbnail** is the most repetitive segment

Repetition-Based Audio Structure Analysis

- Extract the **repetitive structure** of a given audio recording
- Often corresponds to **musical form** of the underlying piece
- The **thumbnail** is the most repetitive segment

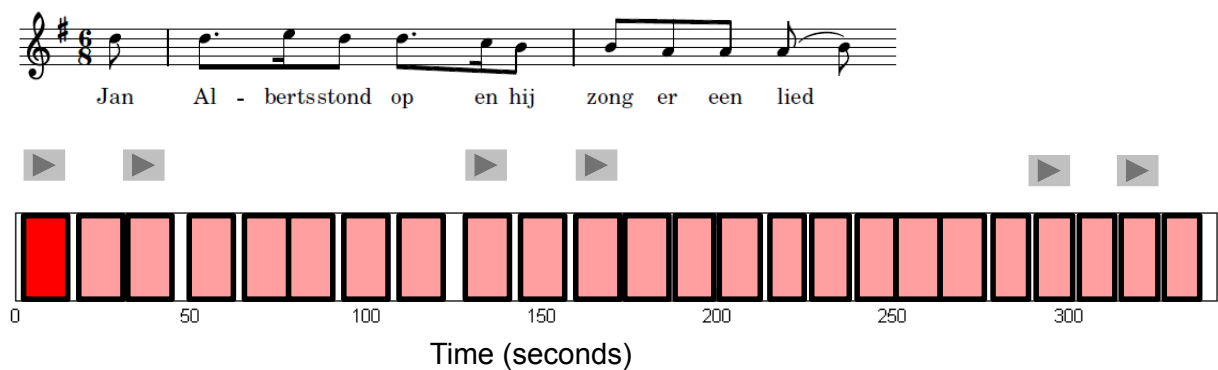
Example: Zager & Evans “In The Year 2525”



Repetition-Based Audio Structure Analysis

- Extract the **repetitive structure** of a given audio recording
- Often corresponds to **musical form** of the underlying piece
- The **thumbnail** is the most repetitive segment

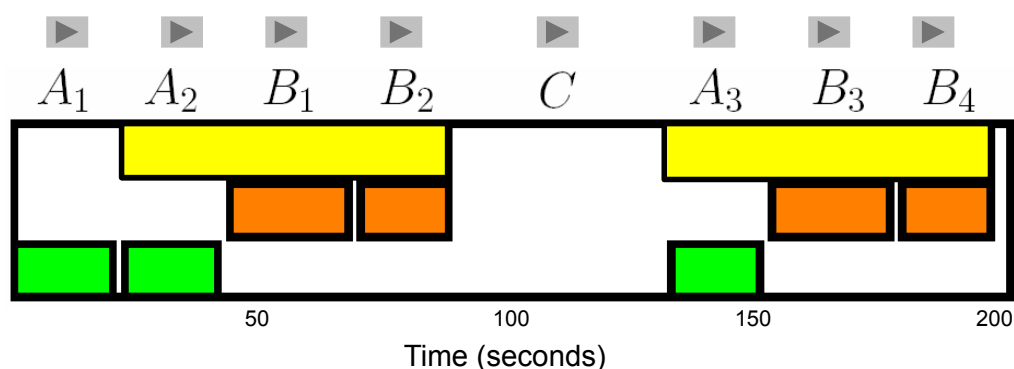
Example: Folk Song Field Recording (Nederlandse Liederbank)



Repetition-Based Audio Structure Analysis

- Extract the **repetitive structure** of a given audio recording
- Often corresponds to **musical form** of the underlying piece
- The **thumbnail** is the most repetitive segment

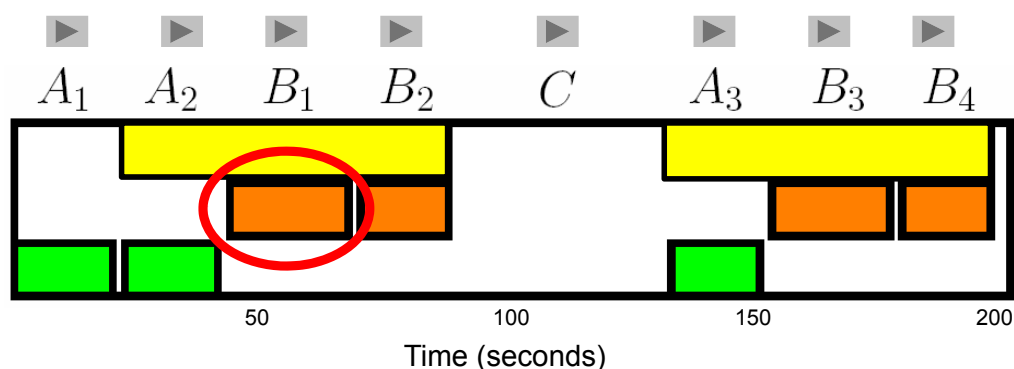
Example: Brahms Hungarian Dance No. 5 (Ormandy)



Repetition-Based Audio Structure Analysis

- Extract the **repetitive structure** of a given audio recording
- Often corresponds to **musical form** of the underlying piece
- The **thumbnail** is the most repetitive segment

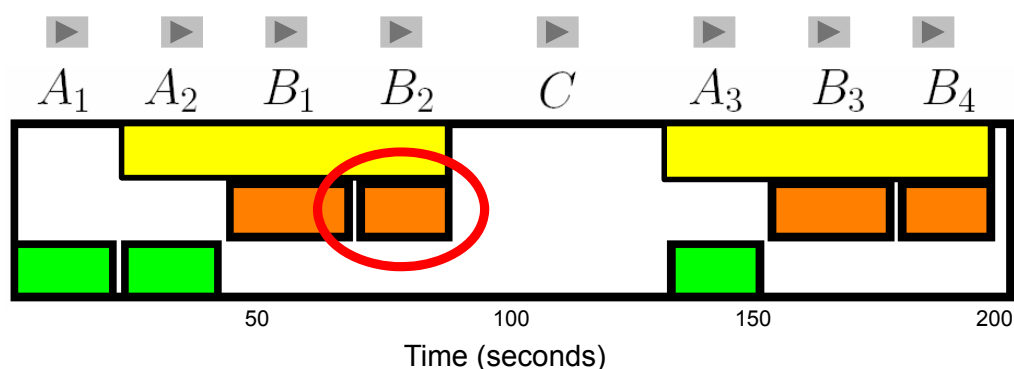
Example: Brahms Hungarian Dance No. 5 (Ormandy)



Repetition-Based Audio Structure Analysis

- Extract the **repetitive structure** of a given audio recording
- Often corresponds to **musical form** of the underlying piece
- The **thumbnail** is the most repetitive segment

Example: Brahms Hungarian Dance No. 5 (Ormandy)



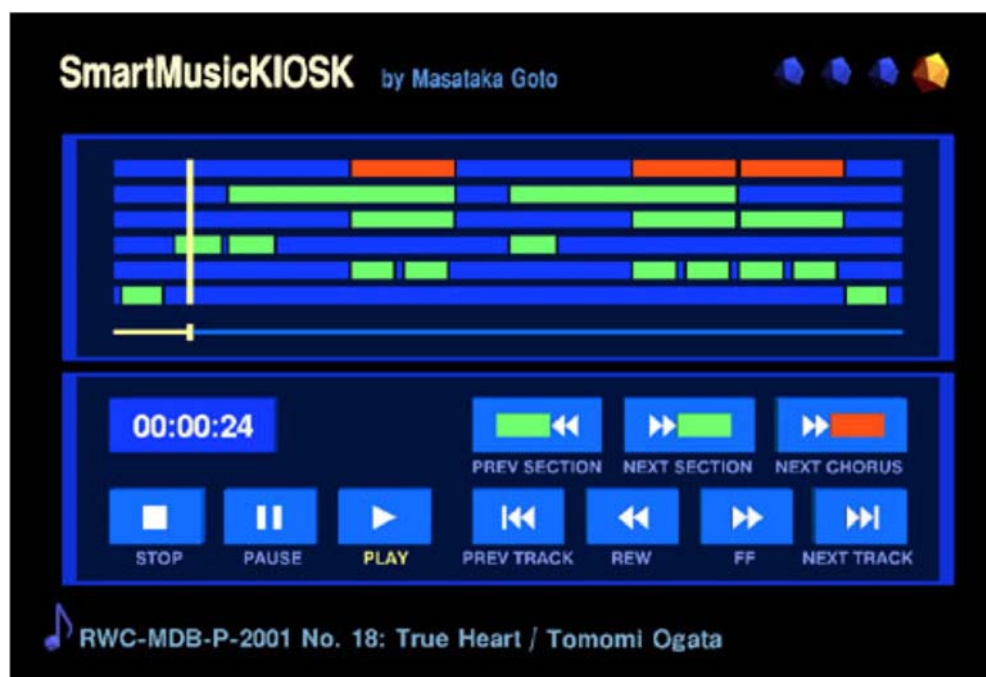
Repetition-Based Audio Structure Analysis

- Extract the **repetitive structure** of a given audio recording
- Often corresponds to **musical form** of the underlying piece
- The **thumbnail** is the most repetitive segment

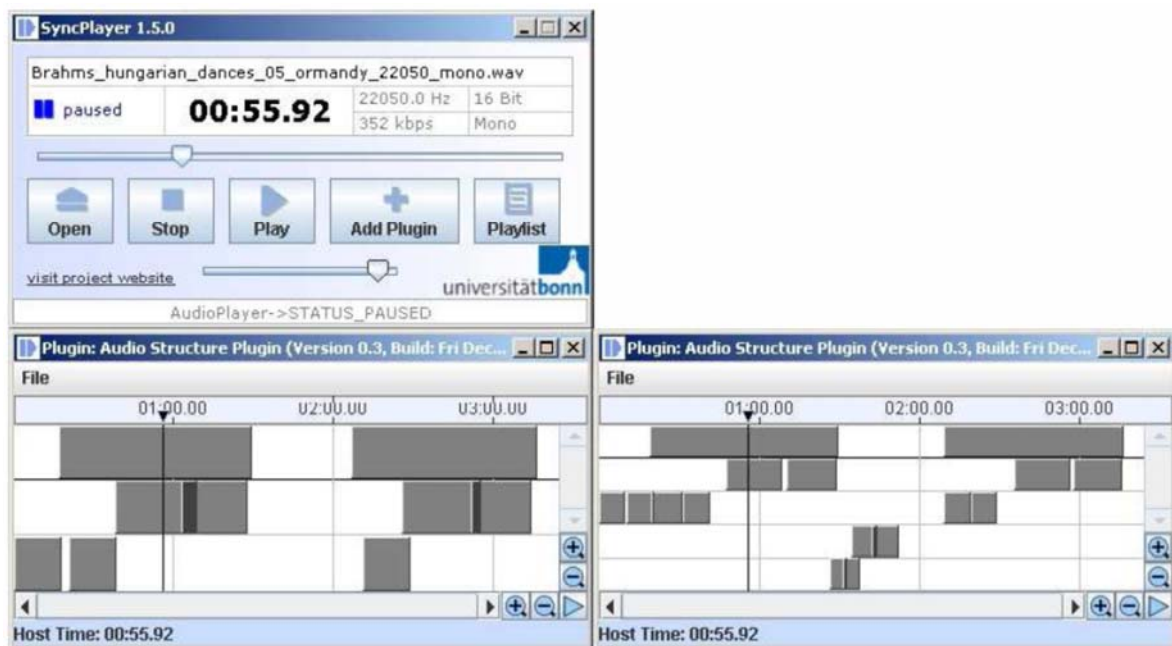
Lots of previous work such as:

- Dannenberg/Hu (ISMIR 2002)
- Peeters/Burthe/Rodet (ISMIR 2002)
- Cooper/Foote (ISMIR 2002)
- Goto (ICASSP 2003)
- Chai/Vercoe (ACM Multimedia 2003)
- Lu/Wang/Zhang (ACM Multimedia 2004)
- Bartsch/Wakefield (IEEE Trans. MM 2005)
- Goto (IEEE Trans. Audio 2006)
- Müller/Kurth (EURASIP 2007)
- Rhodes/Casey (ISMIR 2007)
- Peeters (ISMIR 2007)
- Paulus/Klapuri (IEEE TASLP 2009)
- Paulus/Müller/Klapuri (ISMIR 2010)
- Müller/Grosche/Jiang (ISMIR 2011)
- ...

System: SmartMusicKiosk (Goto)



System: SyncPlayer/AudioStructure



Basic Procedure

- Audio features
- Cost measure and cost matrix
 - ↪ self-similarity matrix
- Path extraction (pairwise similarity of segments)
- Global structure (clustering, grouping)

Basic Procedure

- Audio $\rightsquigarrow V := (v^1, v^2, \dots, v^N)$
- $v^n = 12$ -dimensional normalized chroma vector
- Local cost measure $c : \mathbb{R}^{12} \times \mathbb{R}^{12} \rightarrow \mathbb{R}$

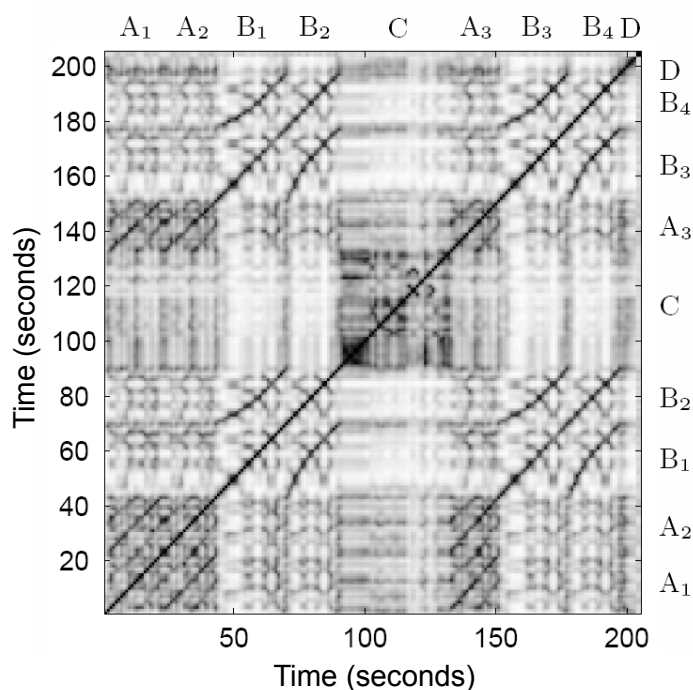
$$c(v^n, w^m) := 1 - \langle v^n, w^m \rangle$$

- $N \times N$ cost matrix $C(n, m) := c(v^n, w^m)$

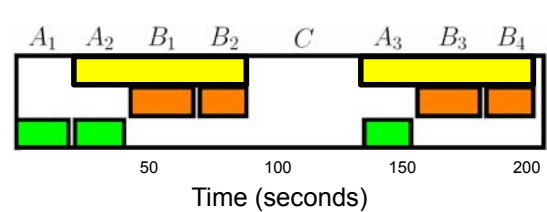
\rightsquigarrow quadratic self-similarity matrix

Basic Procedure

Self-similarity matrix

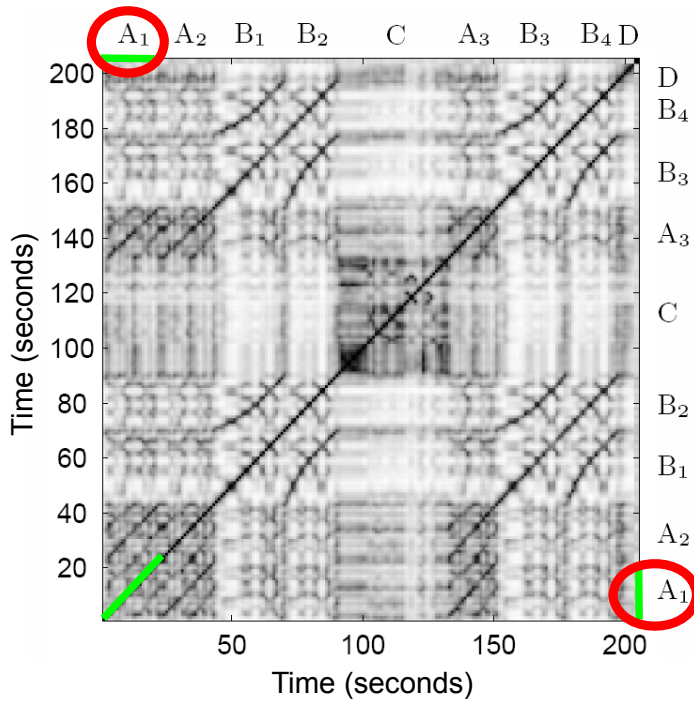


Similarity structure

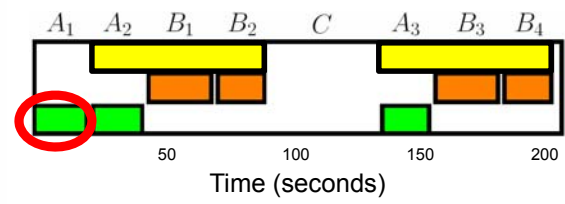


Basic Procedure

Self-similarity matrix

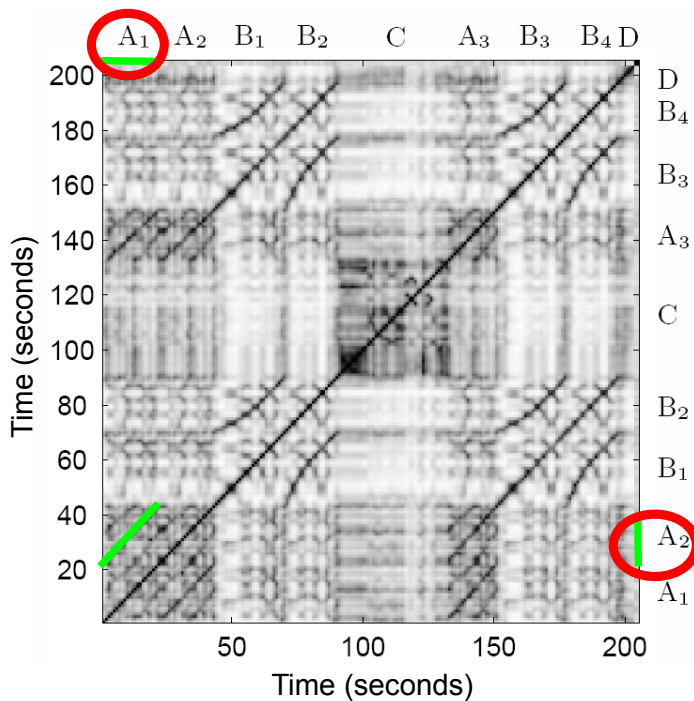


Similarity structure

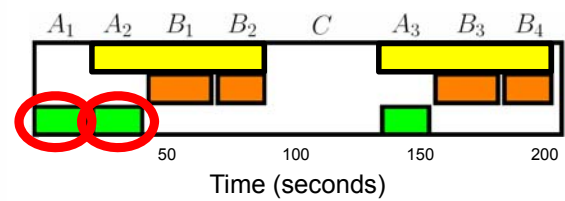


Basic Procedure

Self-similarity matrix

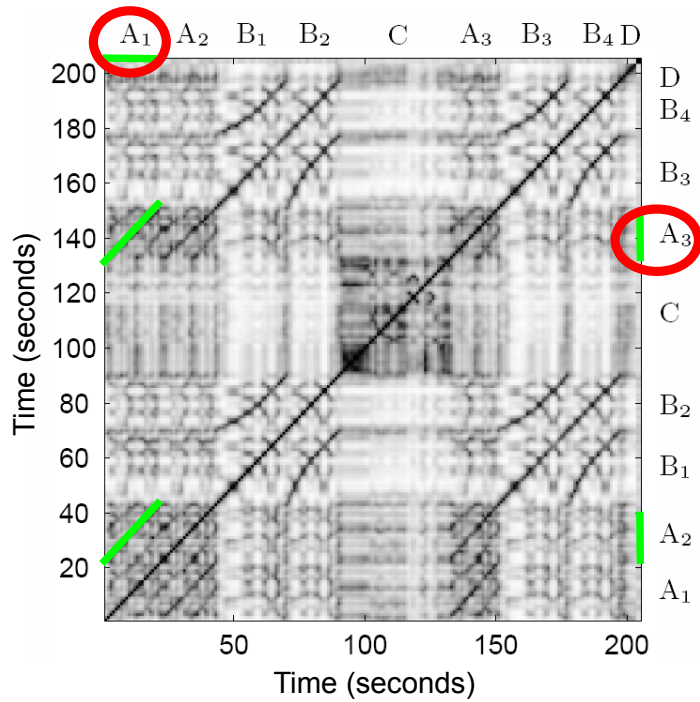


Similarity structure

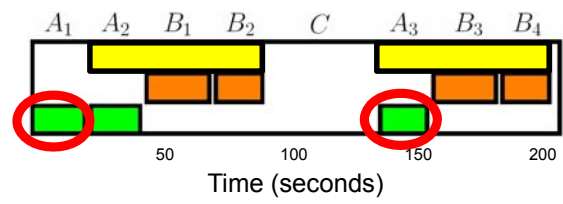


Basic Procedure

Self-similarity matrix

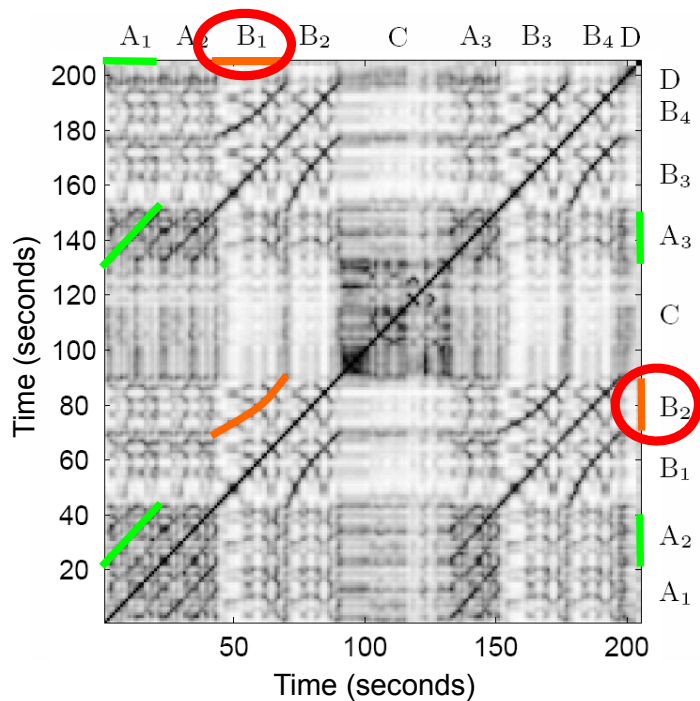


Similarity structure

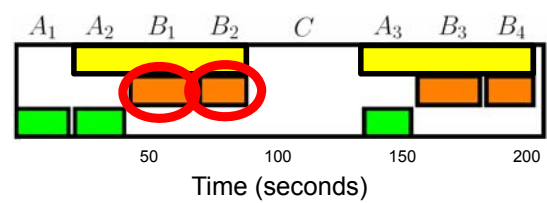


Basic Procedure

Self-similarity matrix

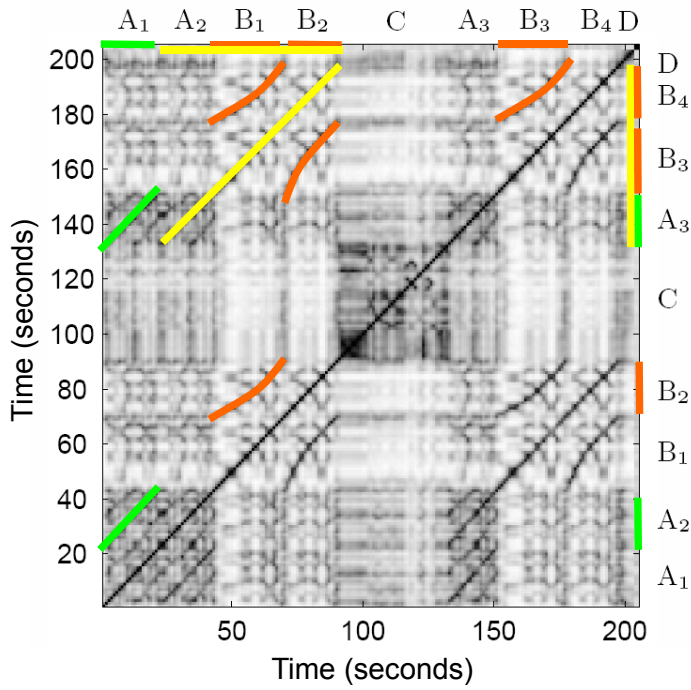


Similarity structure

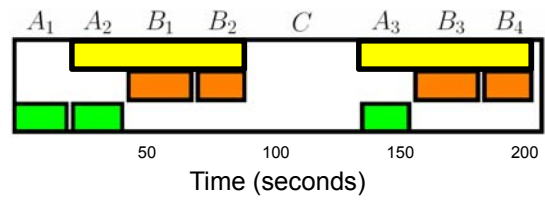


Basic Procedure

Self-similarity matrix

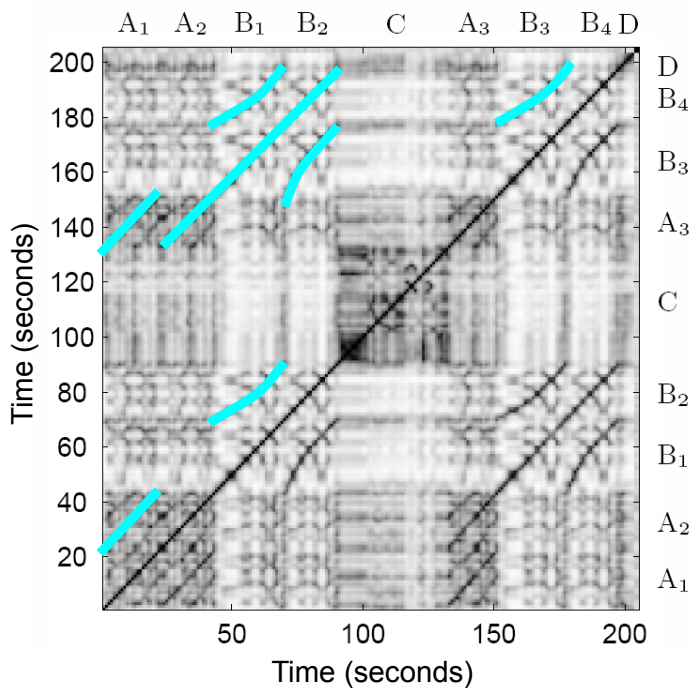


Similarity structure

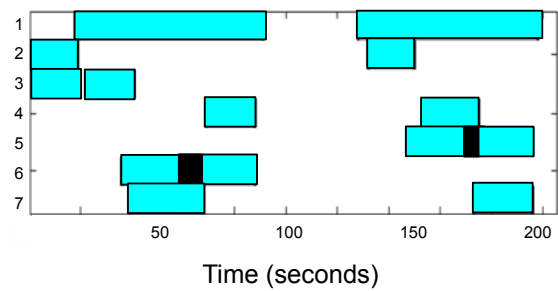


Basic Procedure

Self-similarity matrix

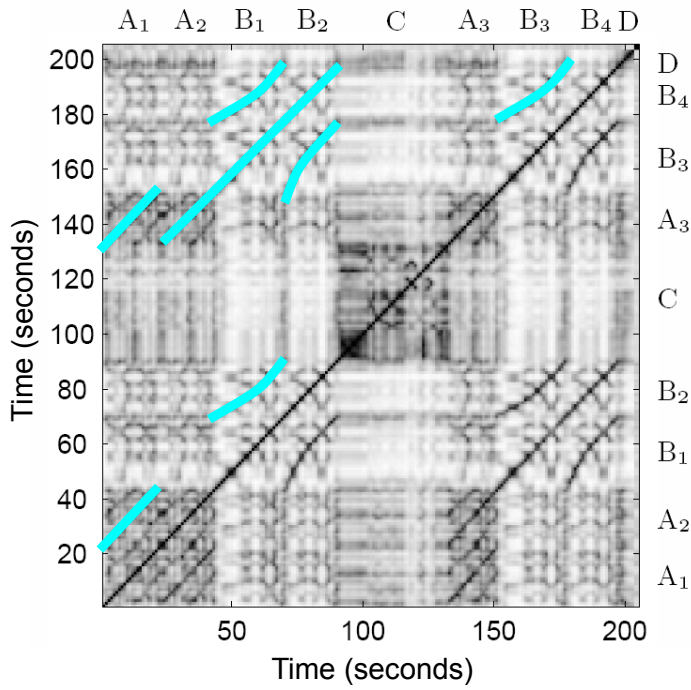


Path relations

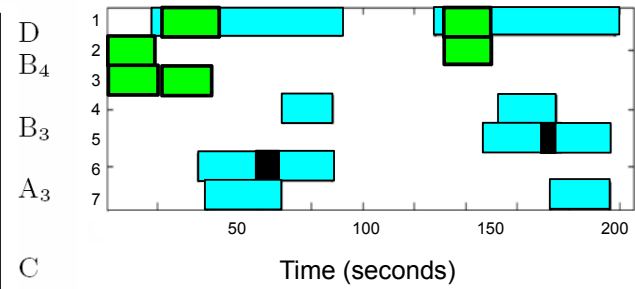


Basic Procedure

Self-similarity matrix



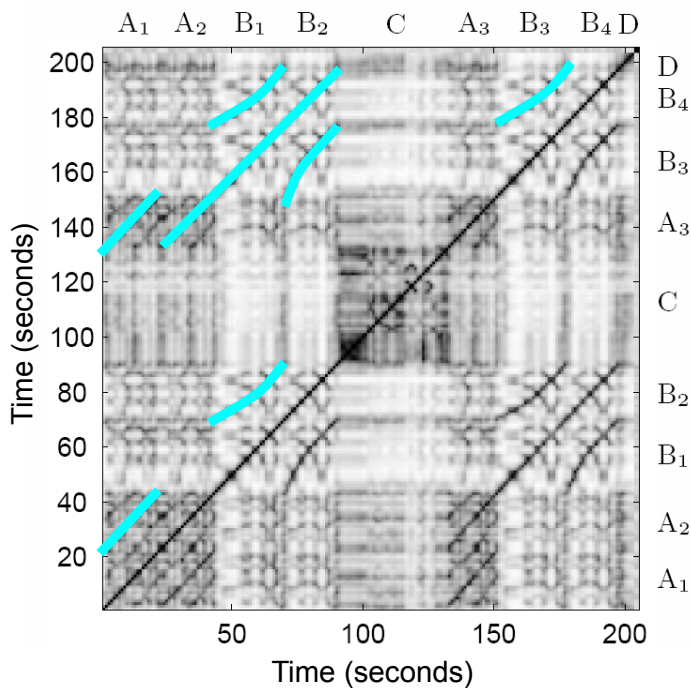
Path relations



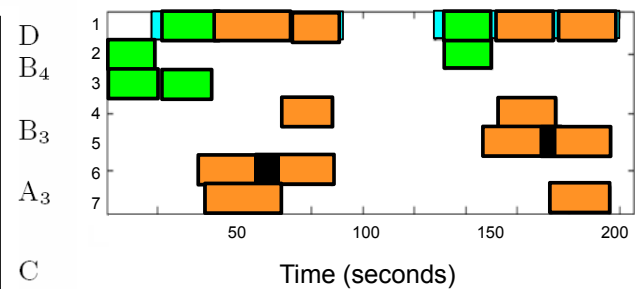
Grouping / Transitivity

Basic Procedure

Self-similarity matrix



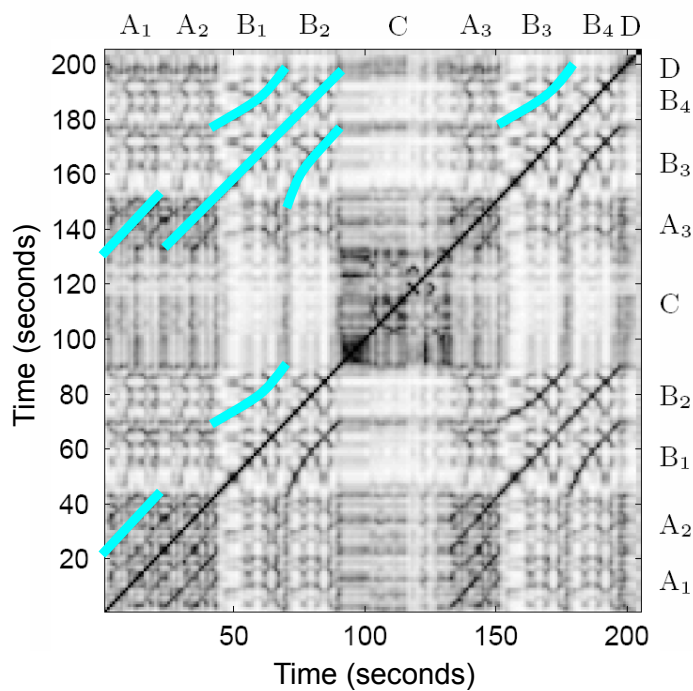
Path relations



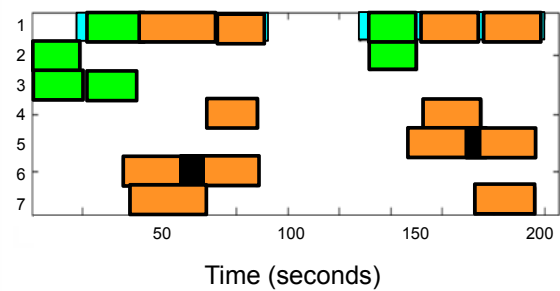
Grouping / Transitivity

Basic Procedure

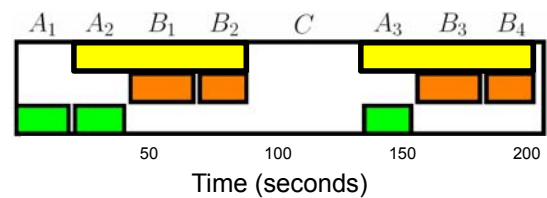
Self-similarity matrix



Path relations



Grouping / Transitivity



Matrix Enhancement

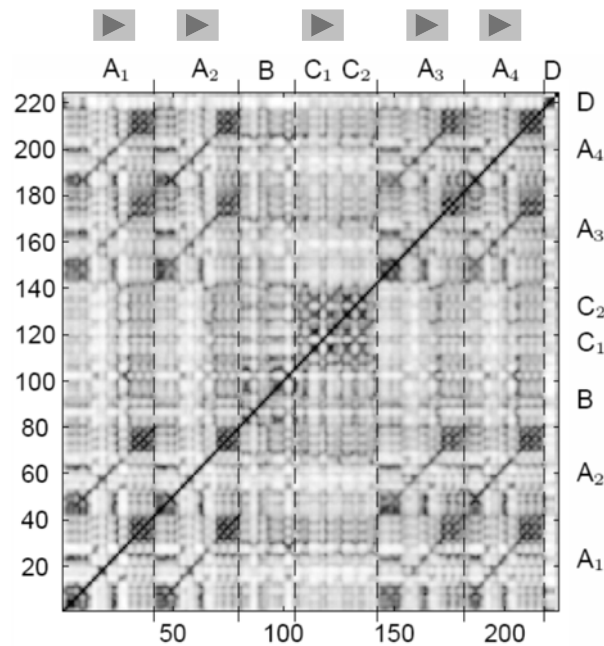
Challenge: Presence of musical variations

- Fragmented paths and gaps
- Paths of poor quality
- Regions of constant (low) cost
- Curved paths

Idea: Enhancement of path structure

Matrix Enhancement

Shostakovich Waltz 2, Jazz Suite No. 2 (Chailly)



Matrix Enhancement

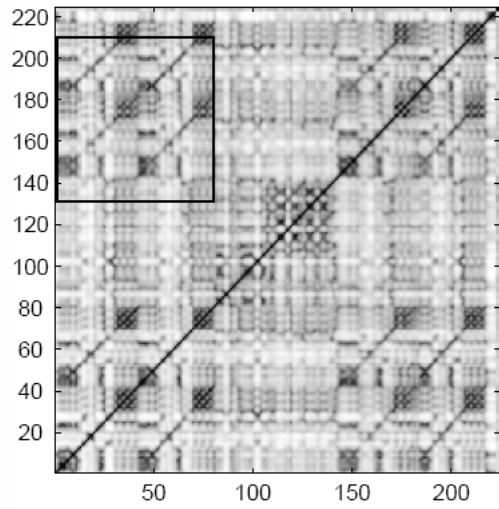
Idea: Usage of contextual information (Foote 1999)

$$C_L(n, m) := \frac{1}{L} \sum_{\ell=0}^{L-1} c(v_{n+\ell}, v_{m+\ell})$$

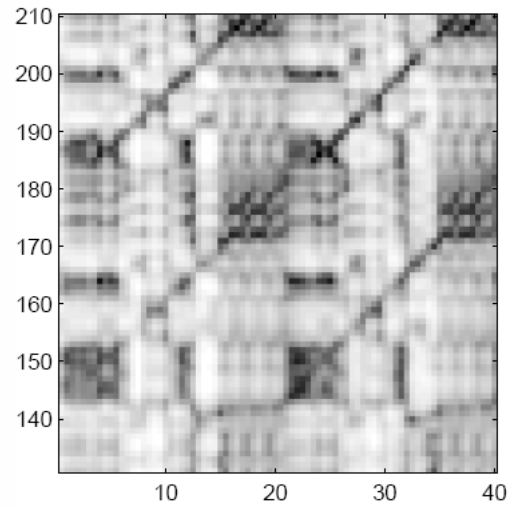
- Comparison of entire sequences
- L = length of sequences
- C_L = enhanced cost matrix

↪ smoothing effect

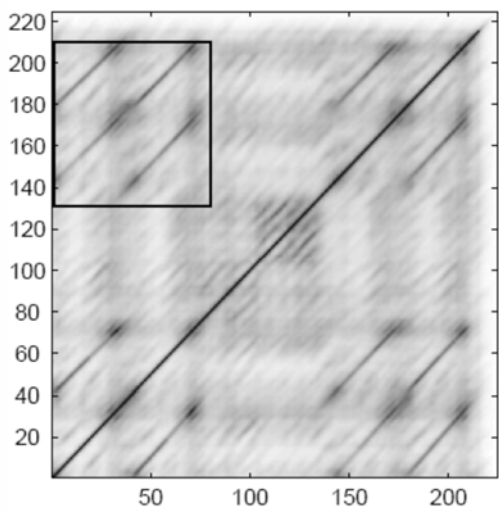
Matrix Enhancement (Shostakovich)



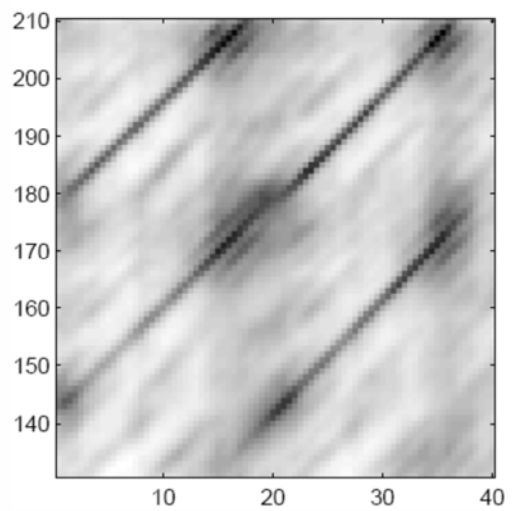
Cost matrix C



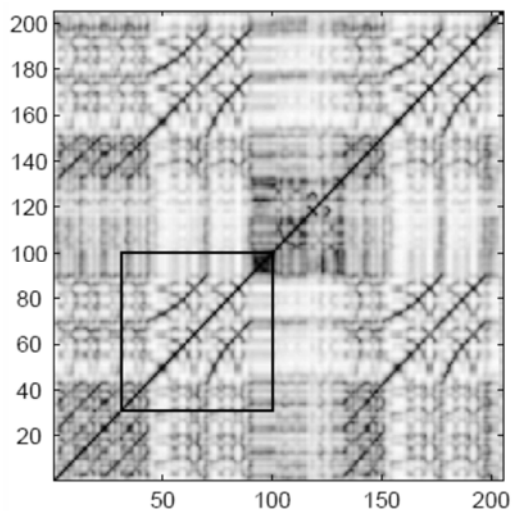
Matrix Enhancement (Shostakovich)



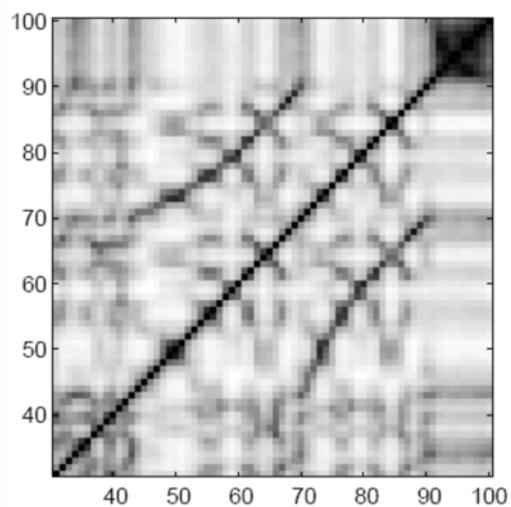
Enhanced cost matrix C_L



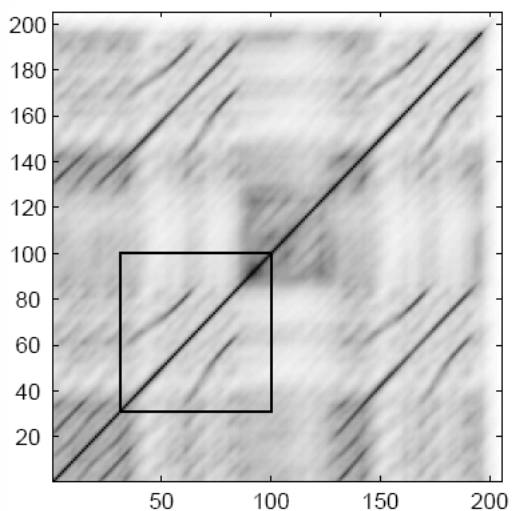
Matrix Enhancement (Brahms)



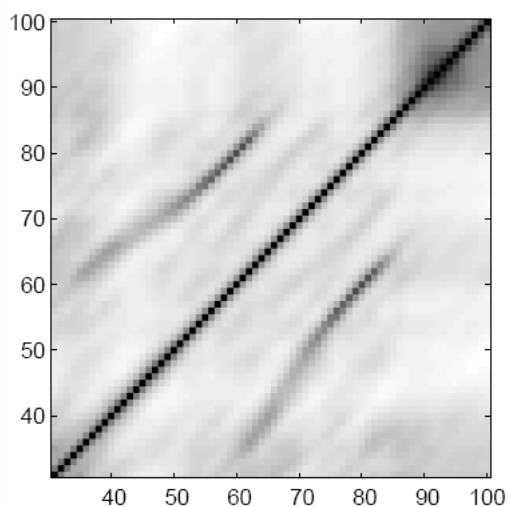
Cost matrix C



Matrix Enhancement (Brahms)



Enhanced cost matrix C_L



Problem: Relative tempo differences are smoothed out

Matrix Enhancement

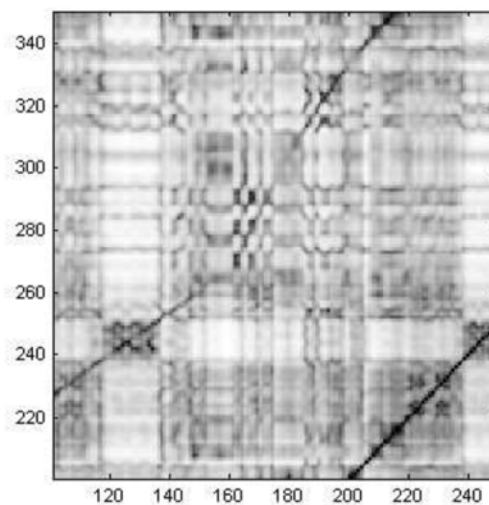
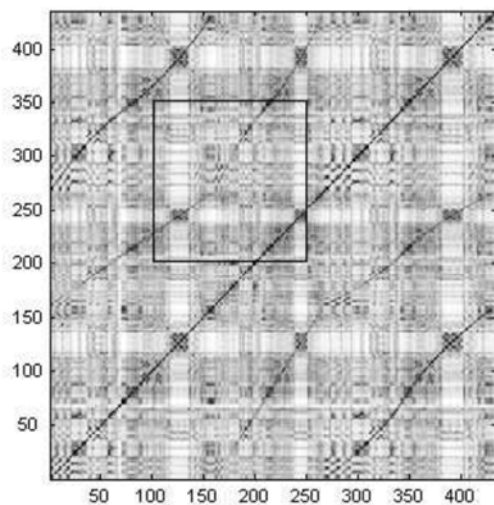
Idea: Smoothing along various directions
and minimizing over all directions

$$C_L^{\min}(n, m) := \min_k C_L^{\text{slope}_k}(n, m)$$

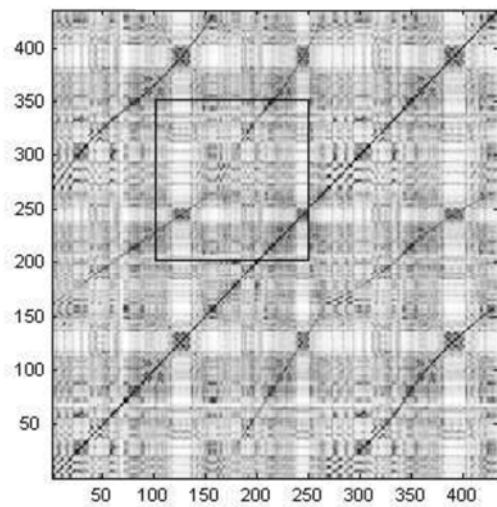
- $\text{slope}_k = k$ th direction of smoothing
- $C_L^{\text{slope}_k} =$ enhanced cost matrix w.r.t. slope_k
- Usage of eight slope values

↪ tempo changes of -30 to +40 percent

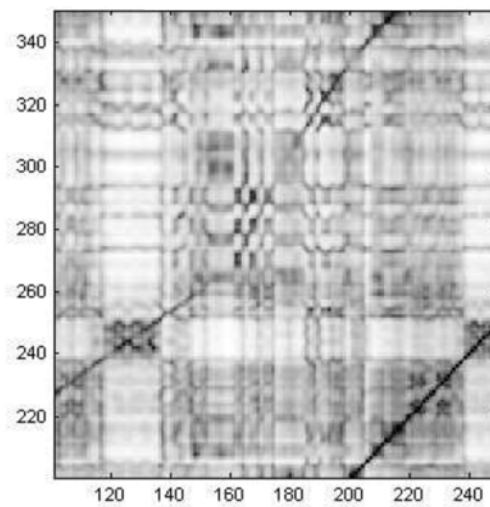
Matrix Enhancement



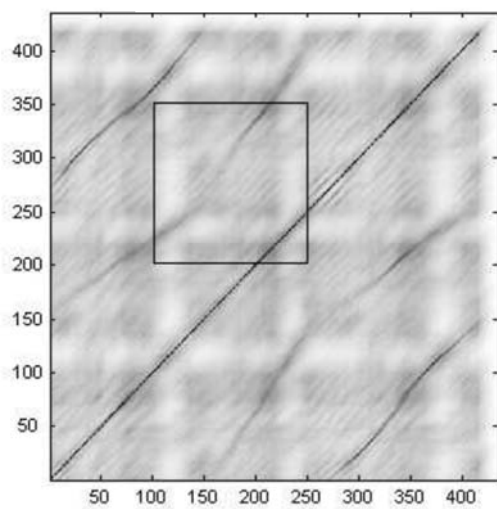
Matrix Enhancement



Cost matrix C

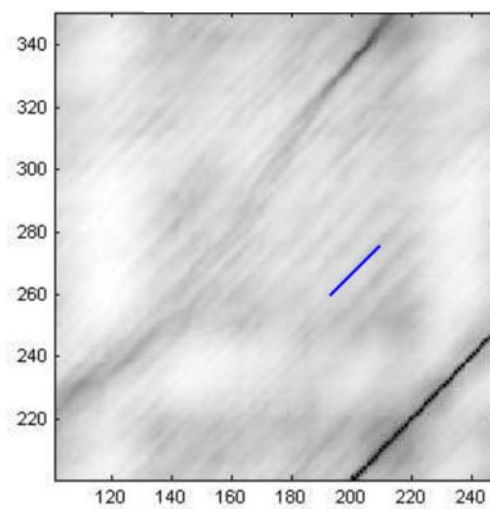


Matrix Enhancement

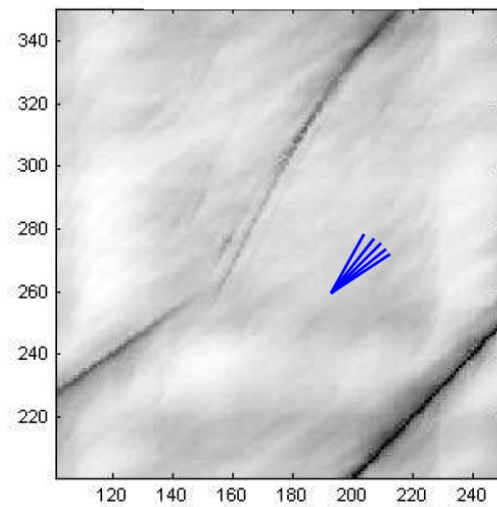
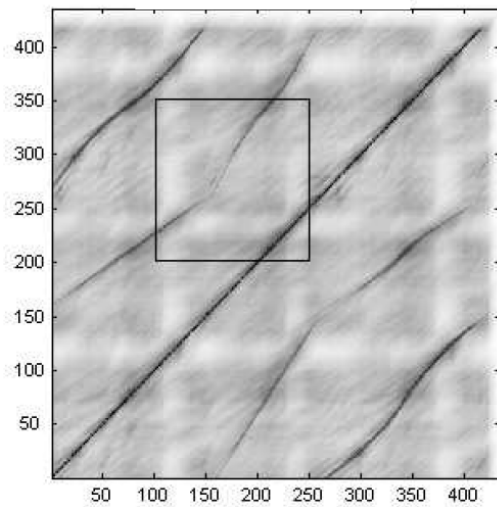


Cost matrix C_L with $L = 20$

Filtering along main diagonal



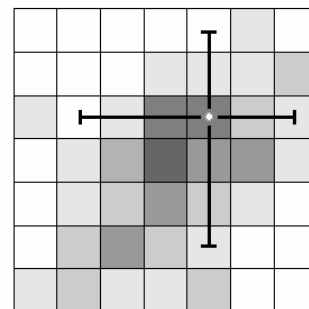
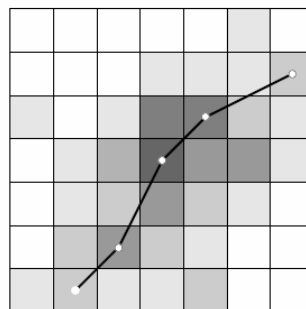
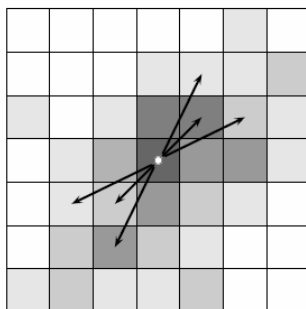
Matrix Enhancement



Cost matrix C_L^{\min} with $L = 20$

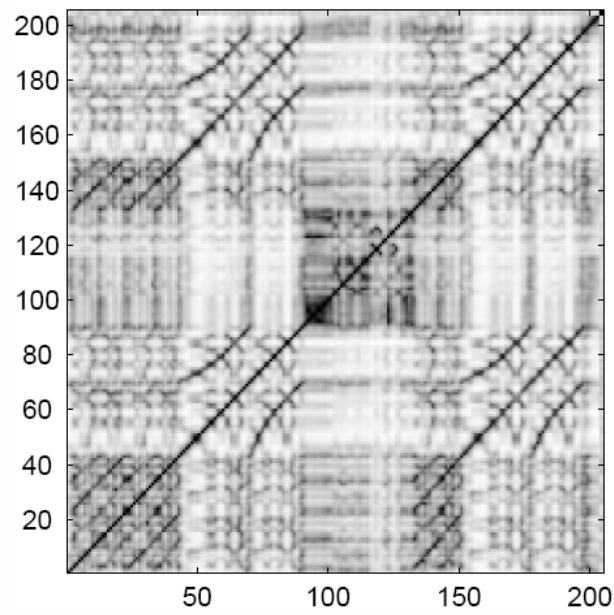
Filtering along 8 different directions and minimizing

Path Extraction



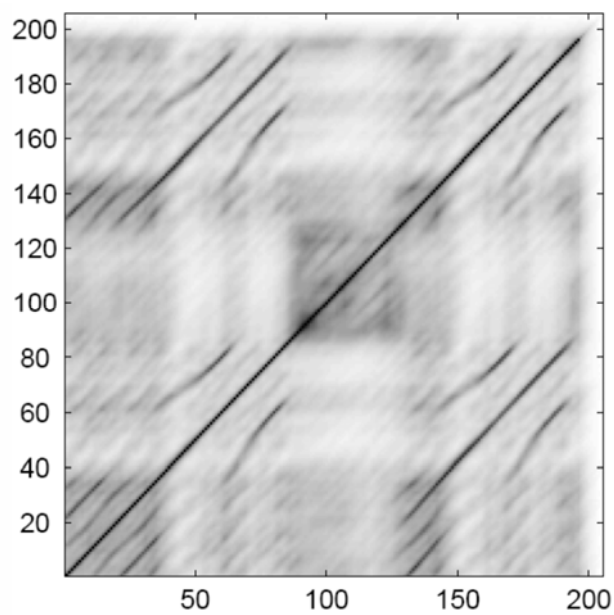
- Start with initial point
- Extend path in greedy fashion
- Remove path neighborhood

Path Extraction



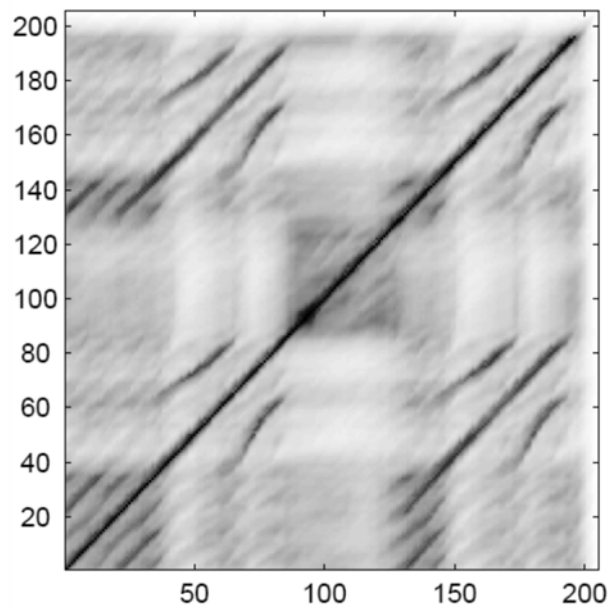
Cost matrix C

Path Extraction



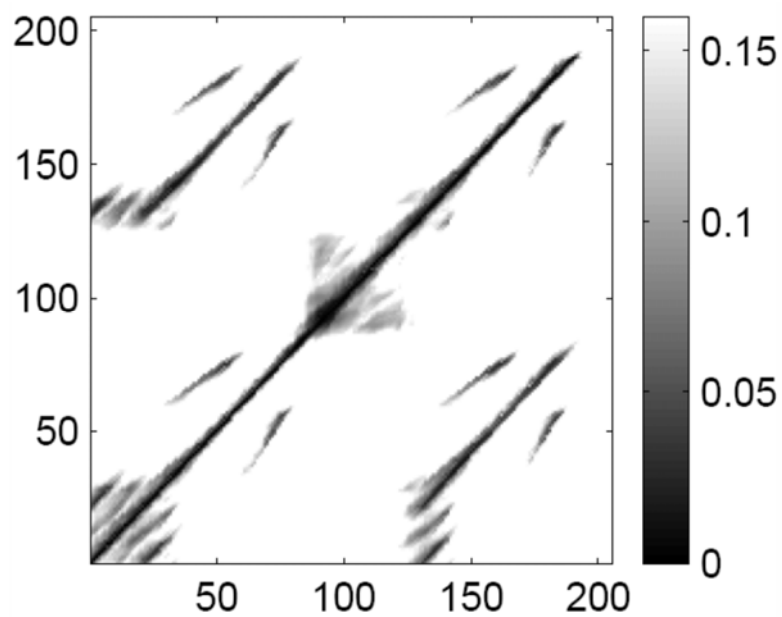
Enhanced cost matrix C_L

Path Extraction



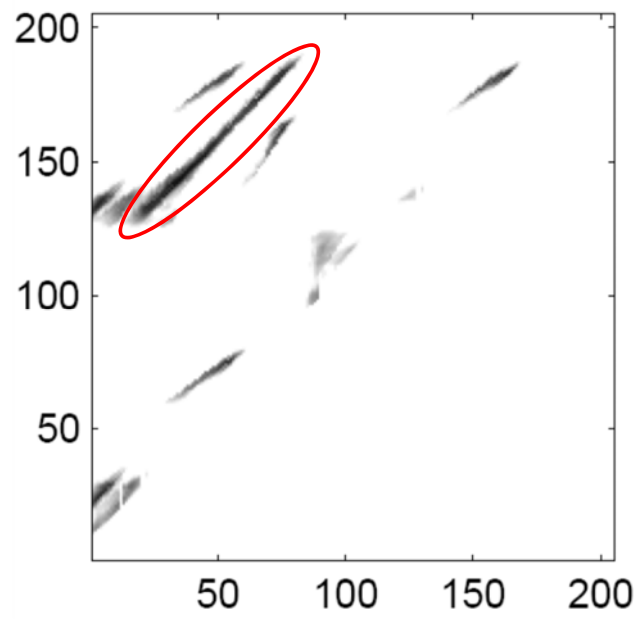
Enhanced cost matrix C_L^{\min}

Path Extraction



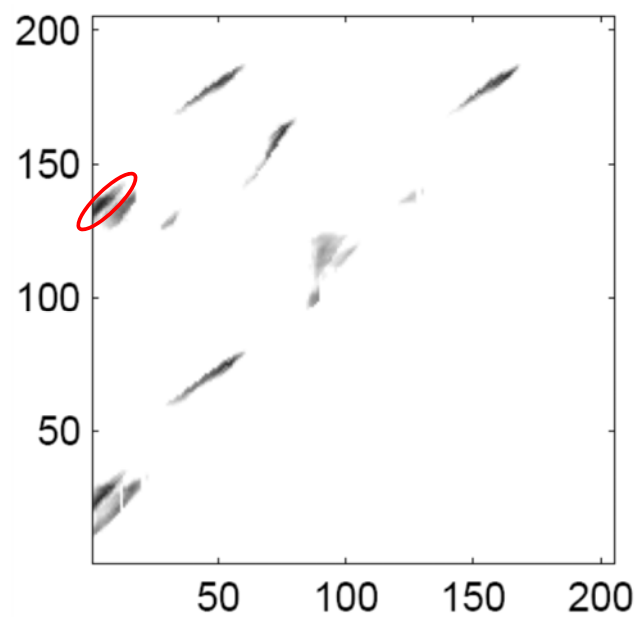
Thresholded C_L^{\min}

Path Extraction



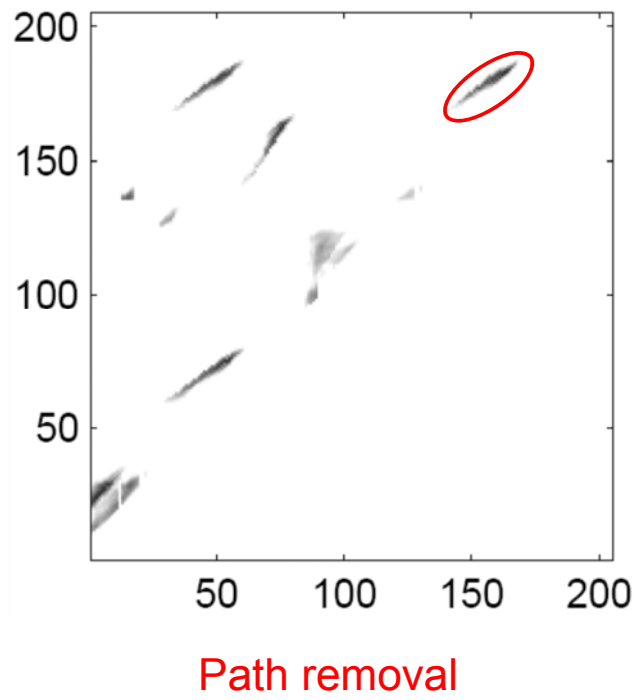
Thresholded C_L^{\min} , upper left

Path Extraction

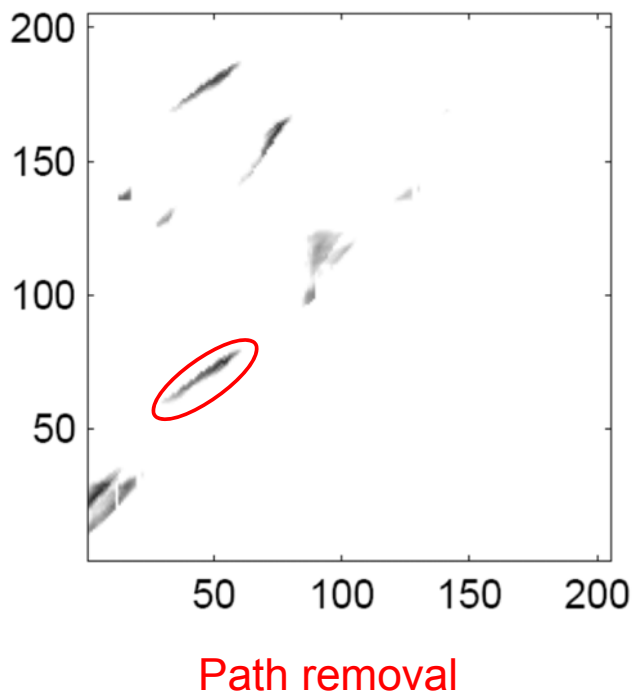


Path removal

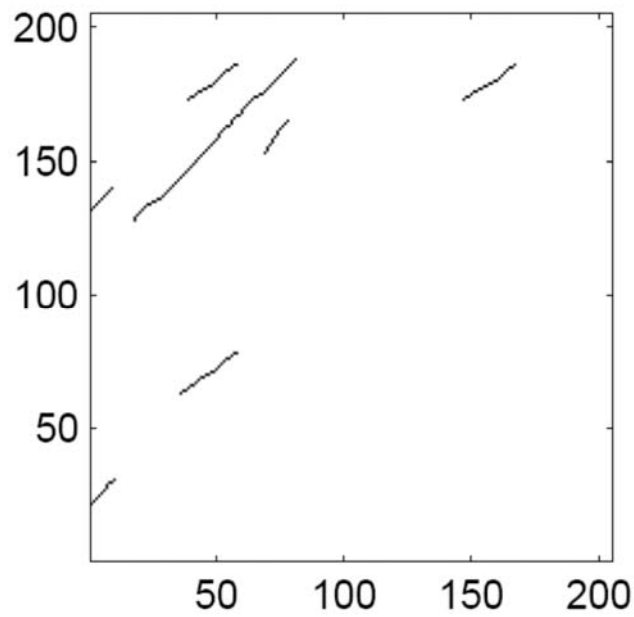
Path Extraction



Path Extraction

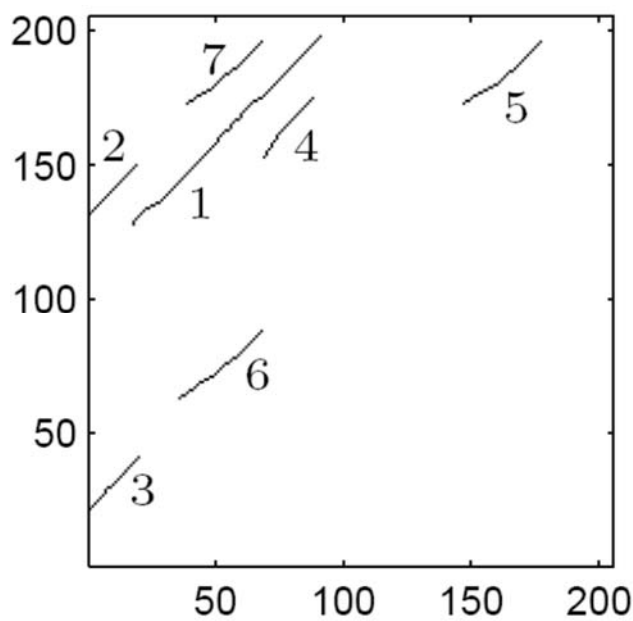


Path Extraction



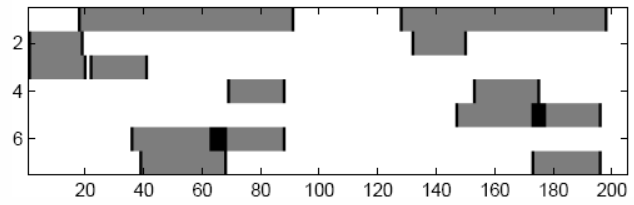
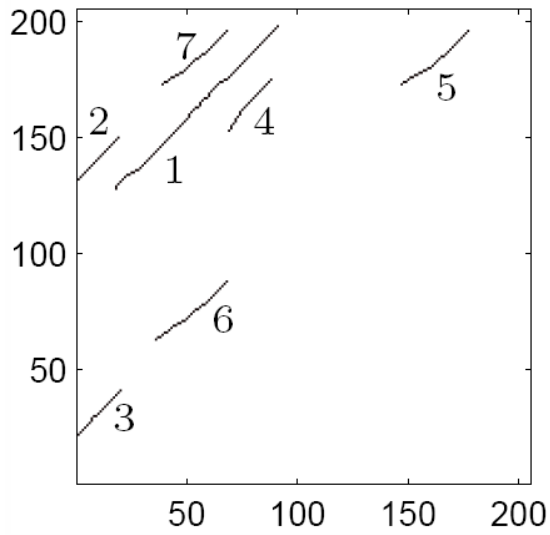
Extracted paths

Path Extraction

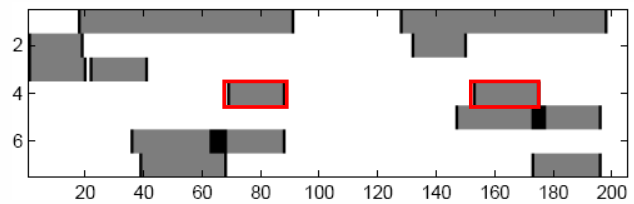
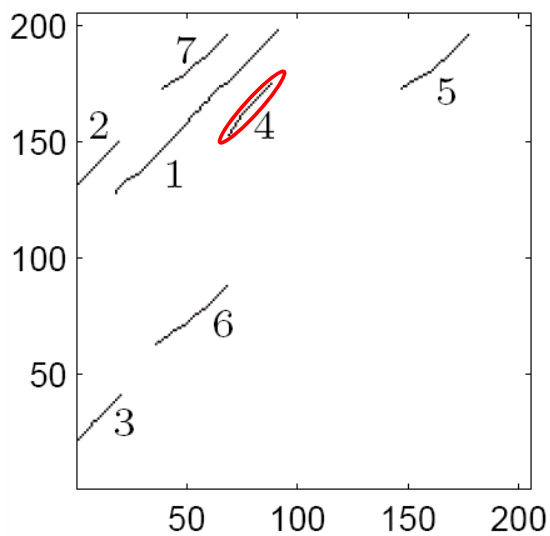


Extracted paths after postprocessing

Global Structure



Global Structure



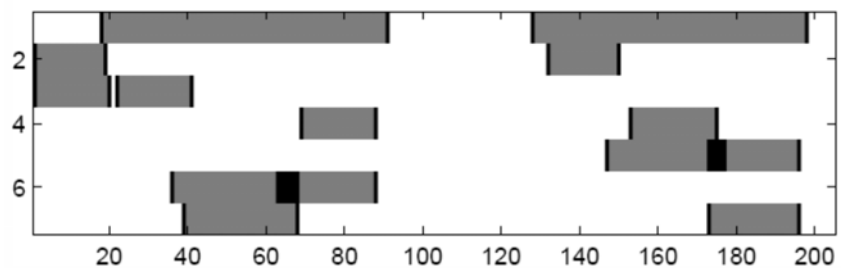
How can one derive the global structure from pairwise relations?

Global Structure

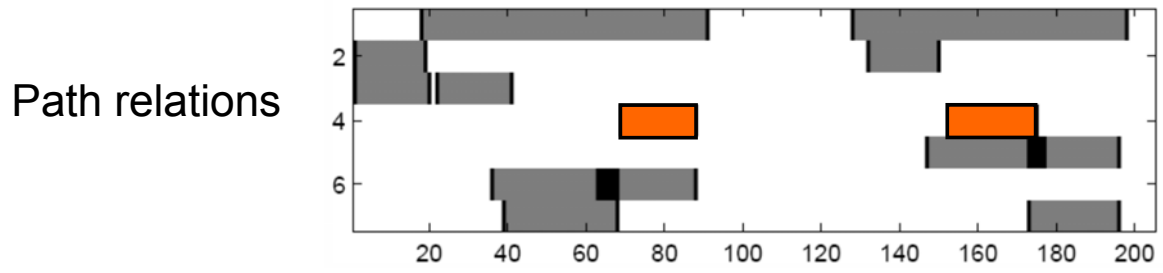
- Taks: Computation of similarity clusters
- Problem: Missing and inconsistent path relations
- Strategy: Approximate “transitive hull”

Global Structure

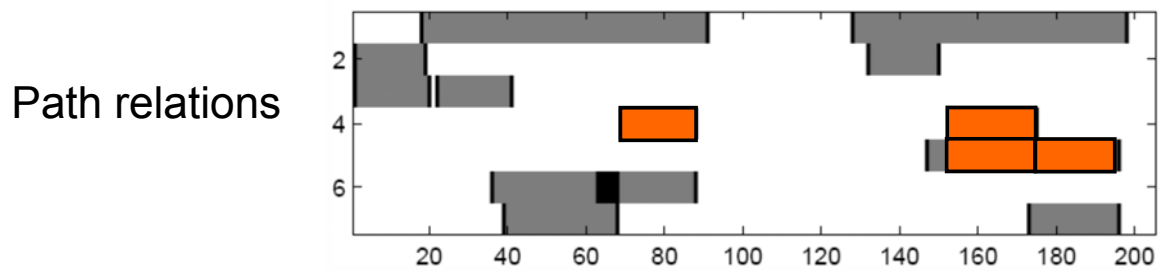
Path relations



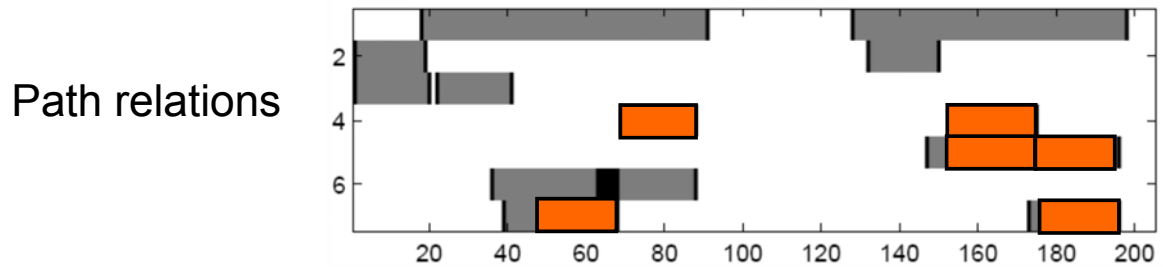
Global Structure



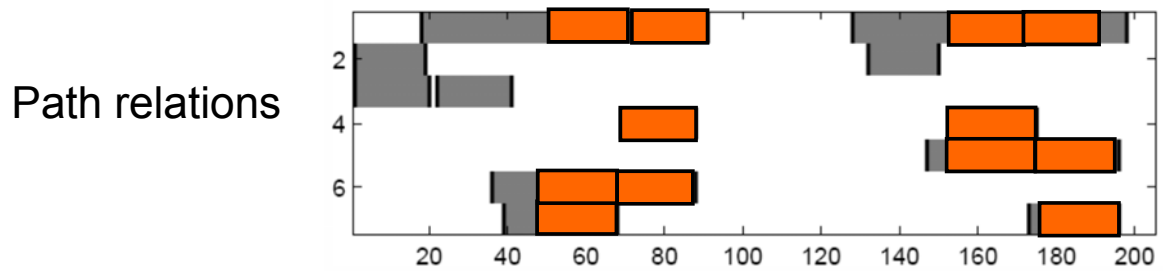
Global Structure



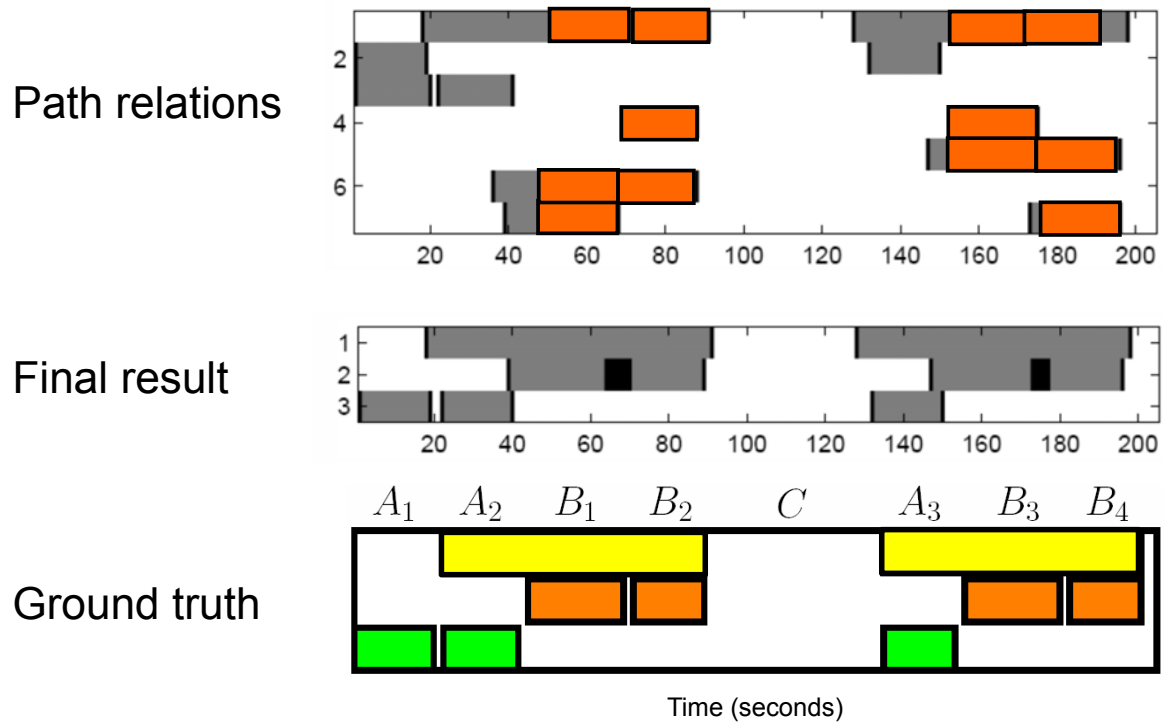
Global Structure



Global Structure

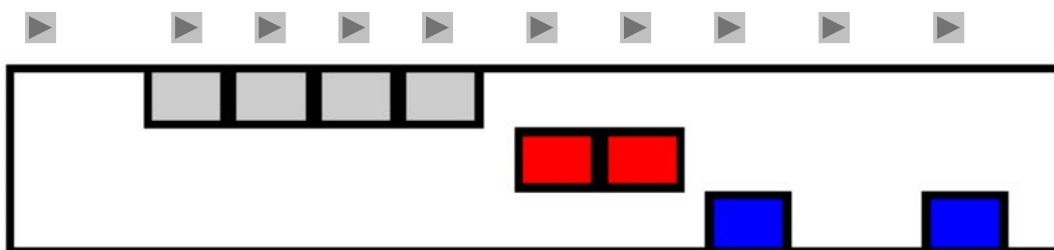


Global Structure



Transposition Invariance

Example: Zager & Evans "In The Year 2525"



Transposition Invariance

Goto (ICASSP 2003)

- Cyclically shift chroma vectors in one sequence
- Compare shifted sequence with original sequence
- Perform for each of the twelve shifts a separate structure analysis
- Combine the results

Transposition Invariance

Goto (ICASSP 2003)

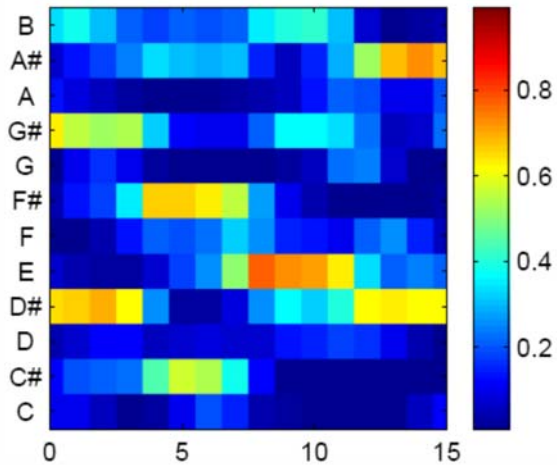
- Cyclically shift chroma vectors in one sequence
- Compare shifted sequence with original sequence
- Perform for each of the twelve shifts a separate structure analysis
- Combine the results

Müller/Clausen (ISMIR 2007)

- Integrate all cyclic information in one **transposition-invariant self-similarity matrix**
- Perform **one** joint structure analysis

Transposition Invariance

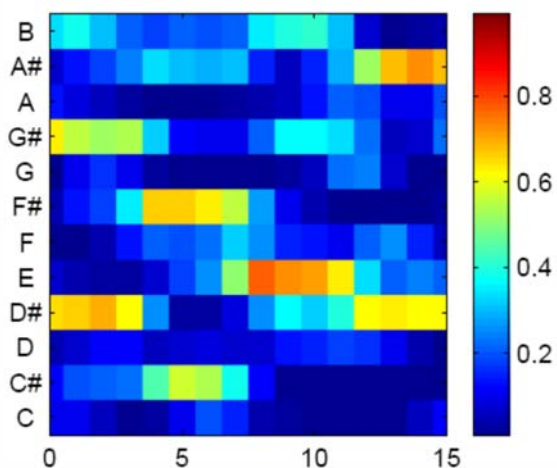
Example: Zager & Evans "In The Year 2525"



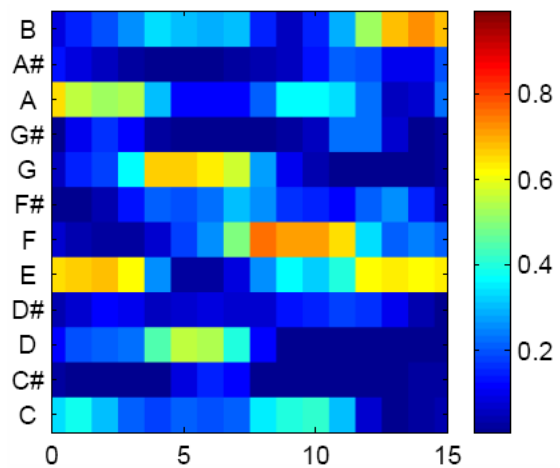
Original: (v^1, \dots, v^N)

Transposition Invariance

Example: Zager & Evans "In The Year 2525"

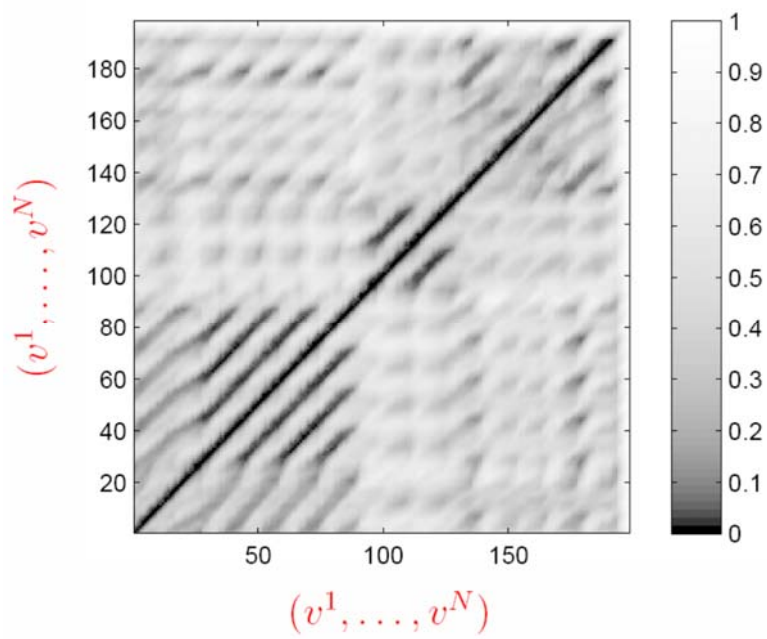


Original: (v^1, \dots, v^N)

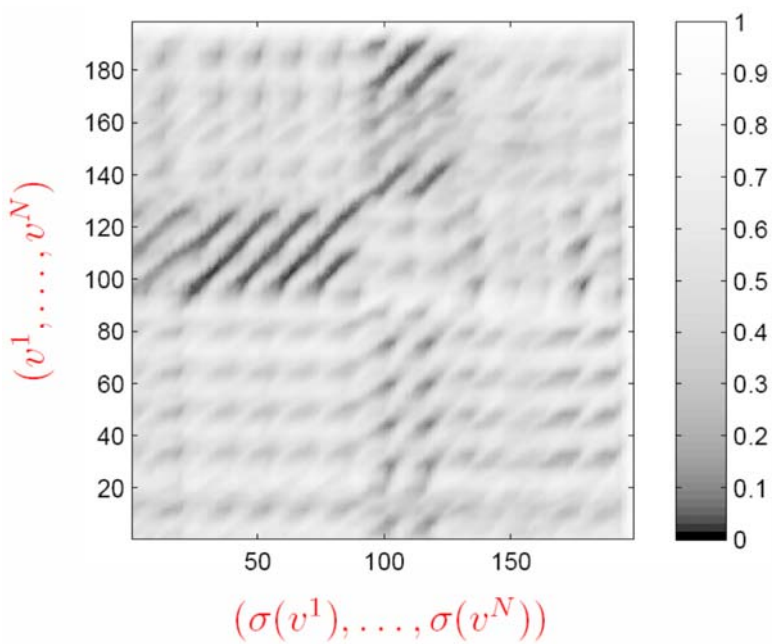


Shifted: $(\sigma(v^1), \dots, \sigma(v^N))$

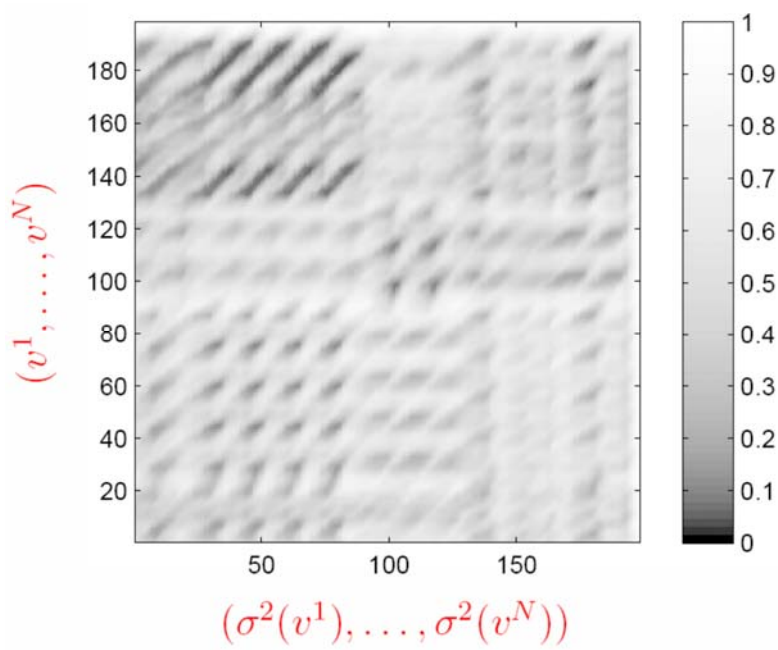
Transposition Invariance



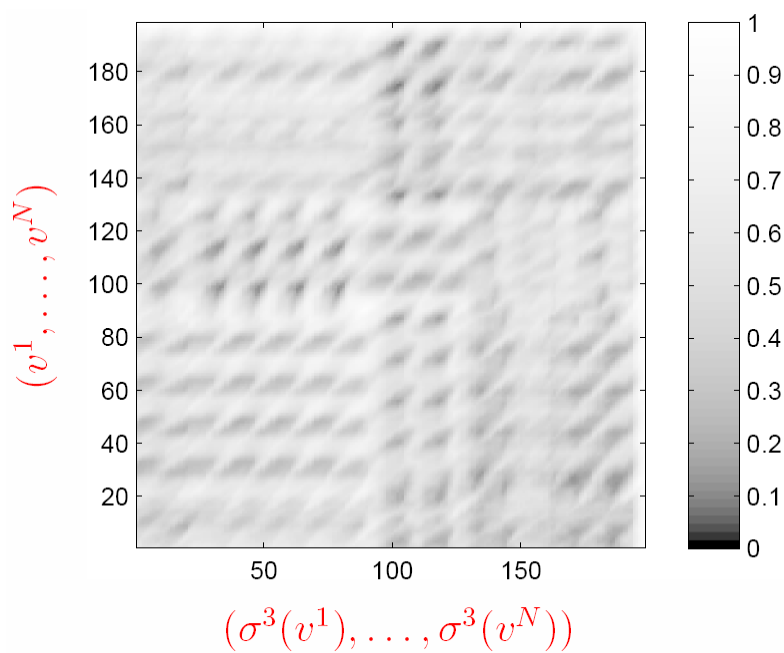
Transposition Invariance



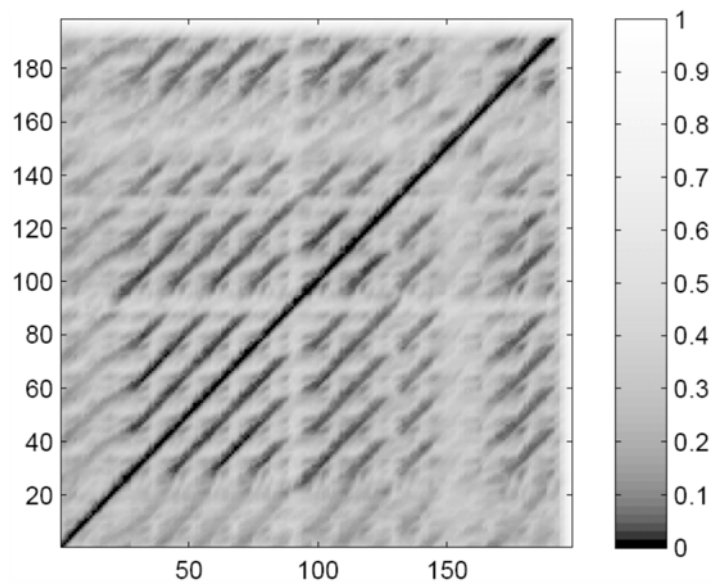
Transposition Invariance



Transposition Invariance

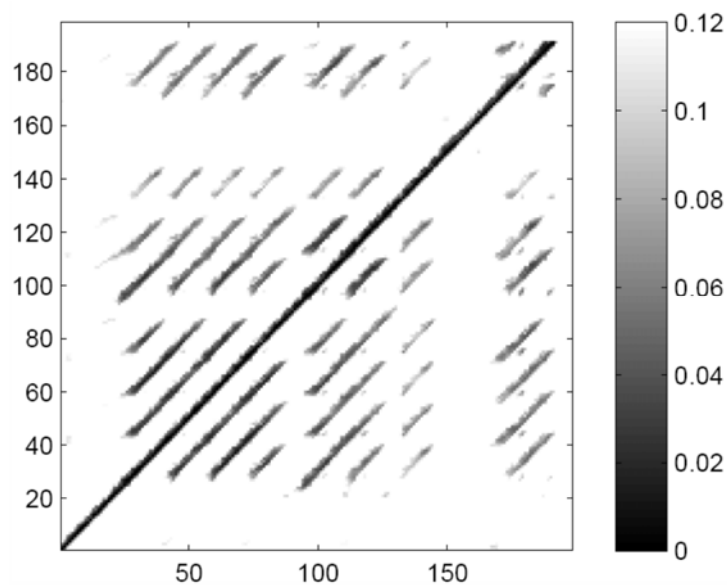


Transposition Invariance



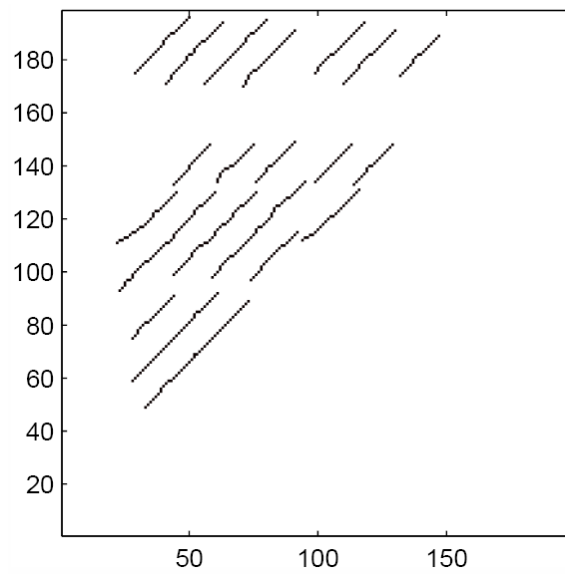
Minimize over all twelve matrices

Transposition Invariance



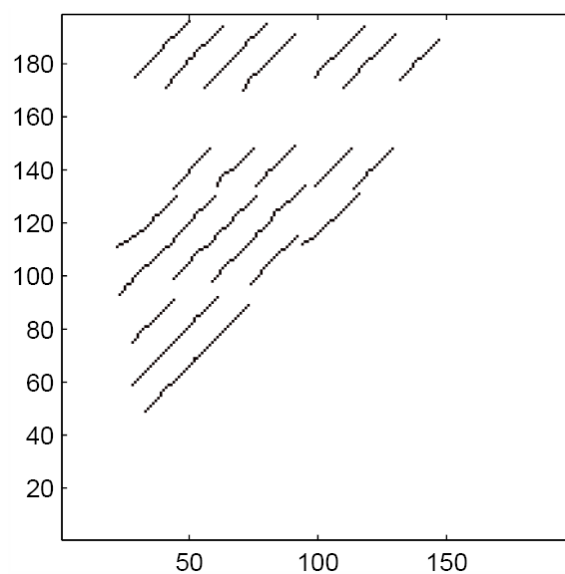
Thresholded self-similarity matrix

Transposition Invariance

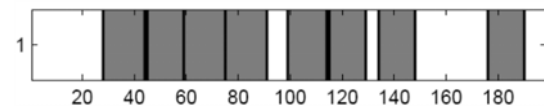


Path extraction

Transposition Invariance



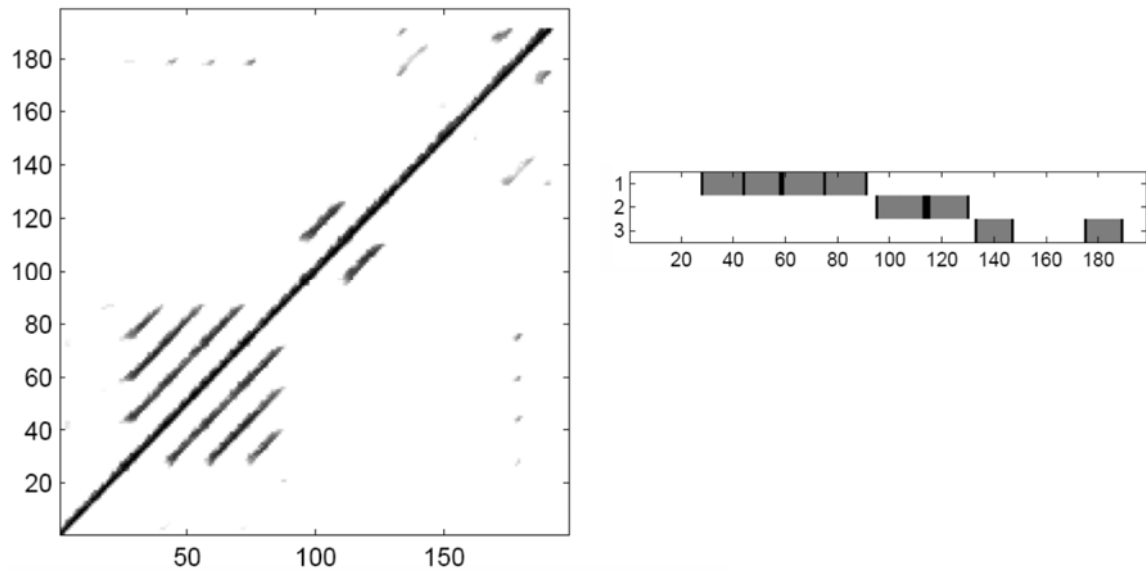
Path extraction



Computation of similarity clusters

Transposition Invariance

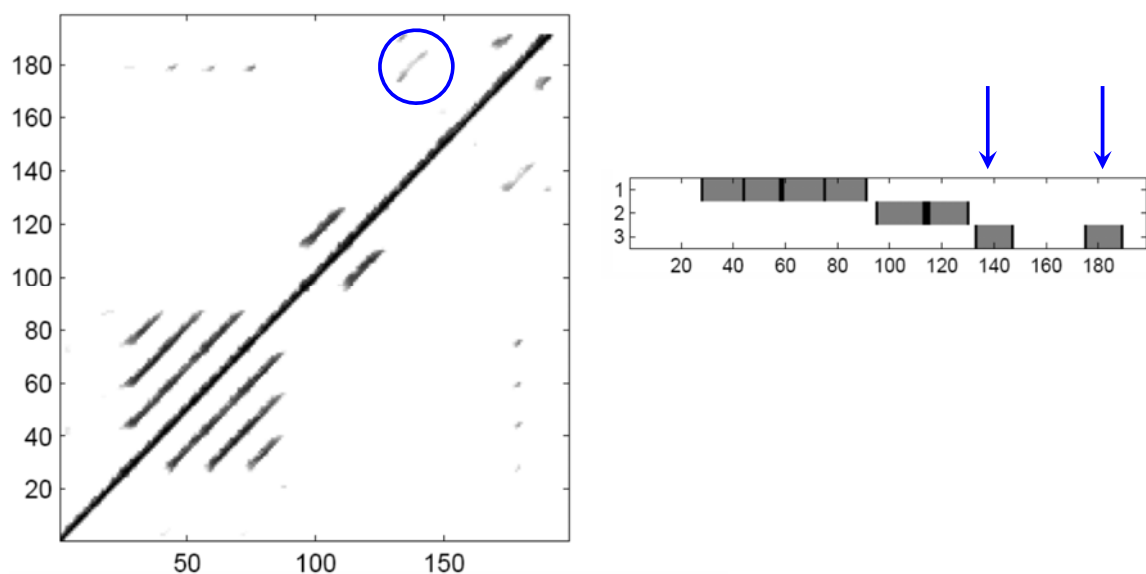
Stabilizing effect



Self-similarity matrix (thresholded)

Transposition Invariance

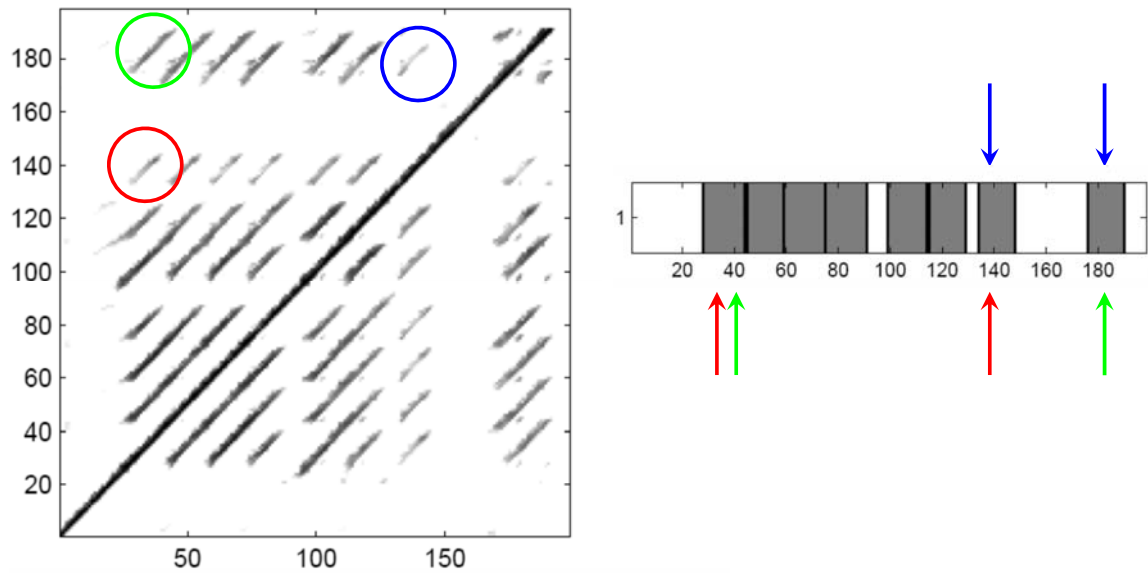
Stabilizing effect



Self-similarity matrix (thresholded)

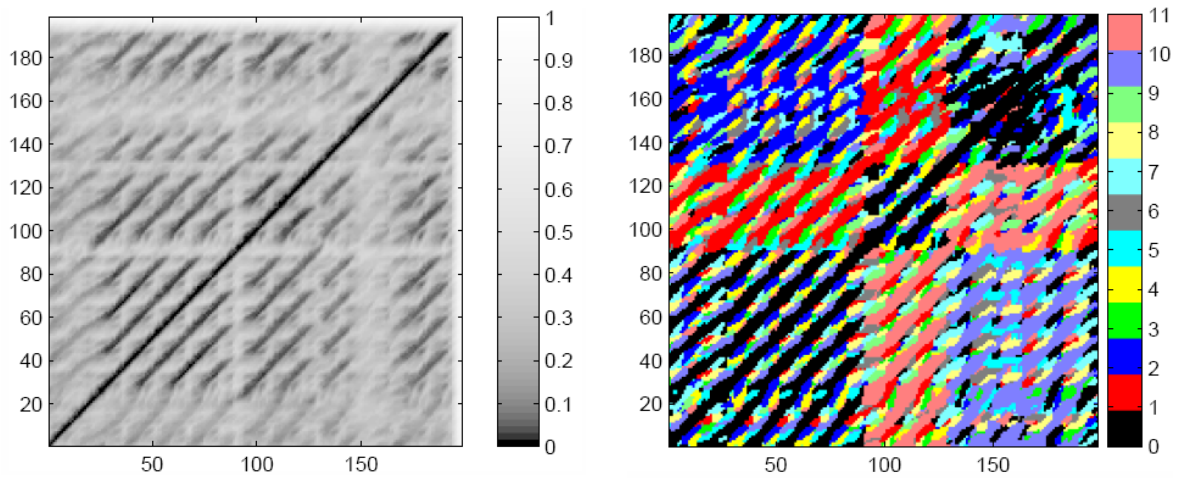
Transposition Invariance

Stabilizing effect



Transposition-invariant self-similarity matrix (thresholded)

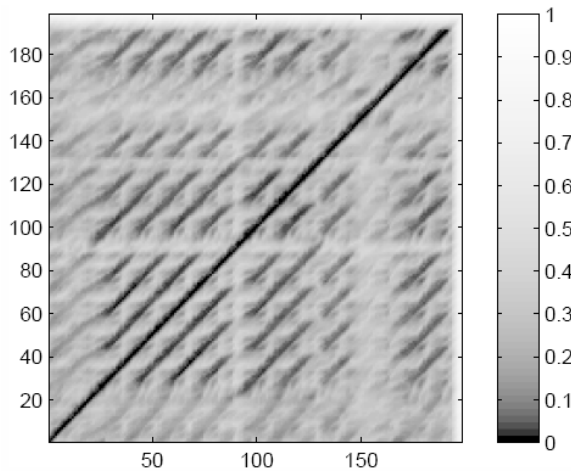
Transposition Invariance



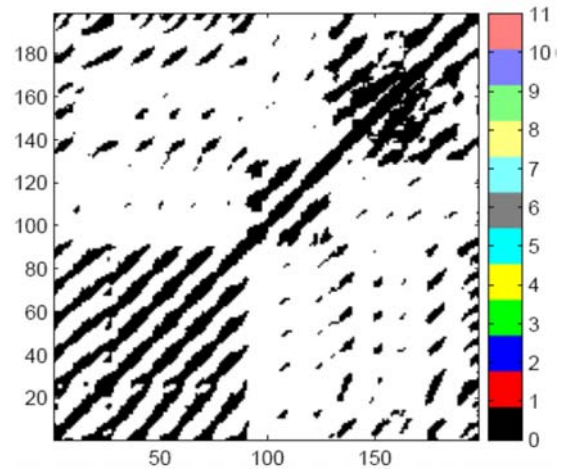
Transposition-invariant matrix

Minimizing shift index

Transposition Invariance

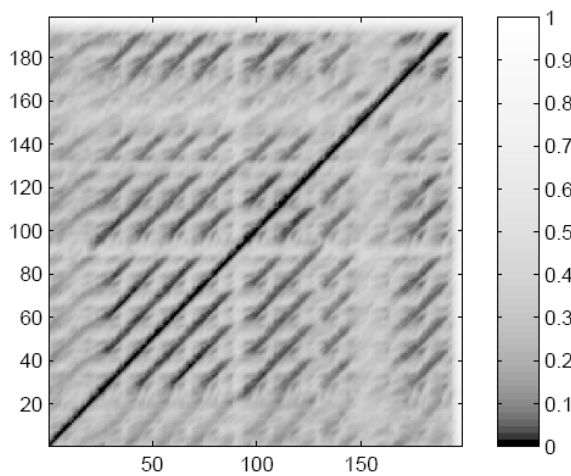


Transposition-invariant matrix

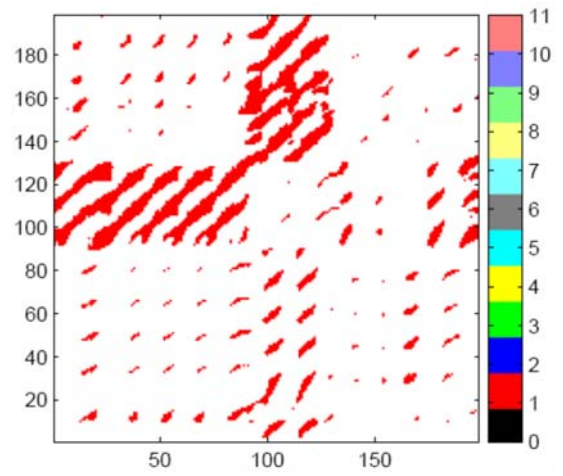


Minimizing shift index = 0

Transposition Invariance

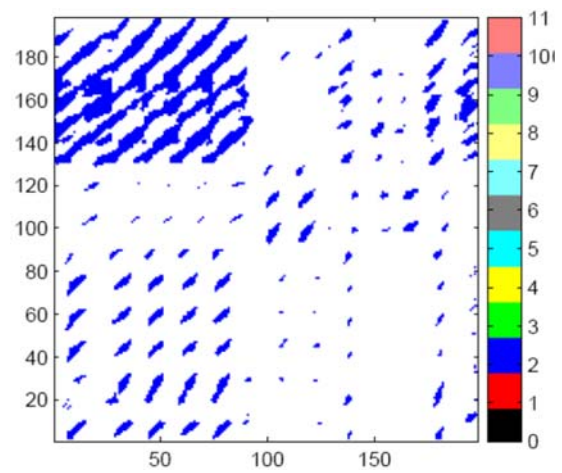
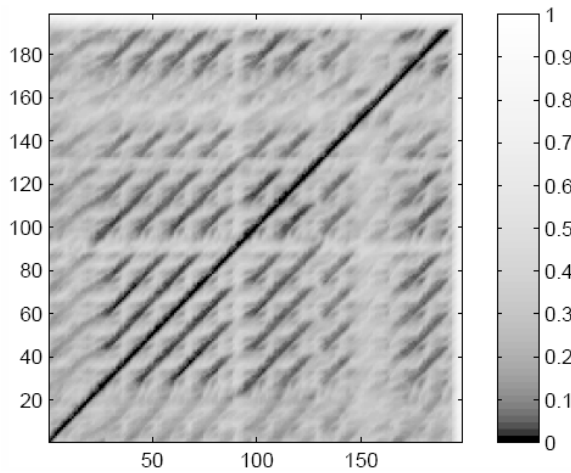


Transposition-invariant matrix



Minimizing shift index = 1

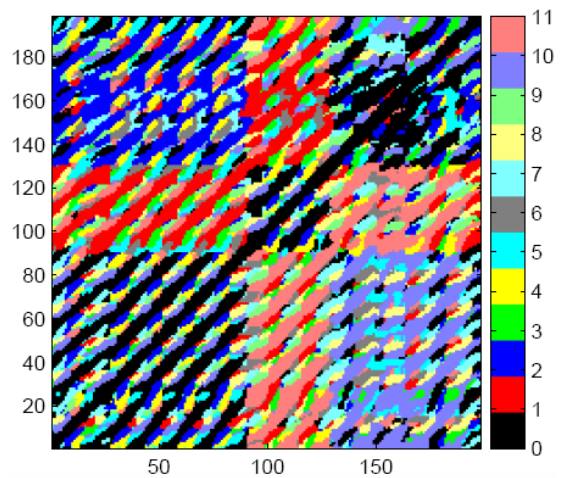
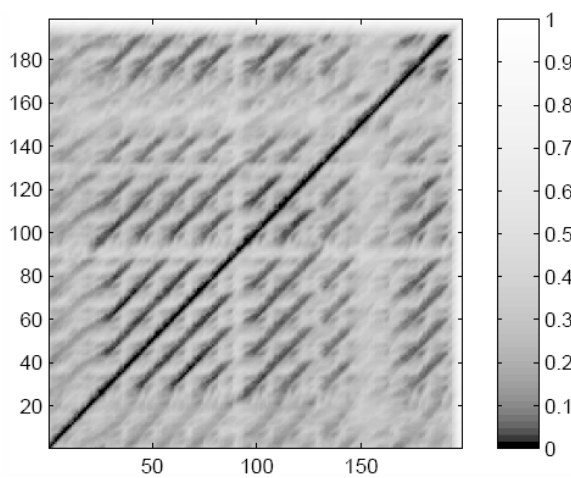
Transposition Invariance



Transposition-invariant matrix

Minimizing shift index = 2

Transposition Invariance

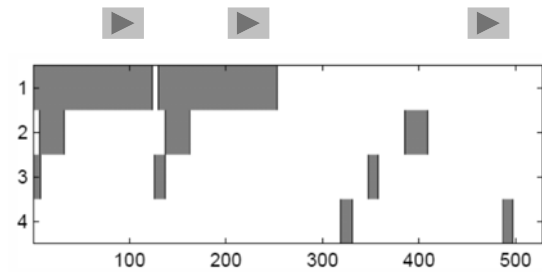
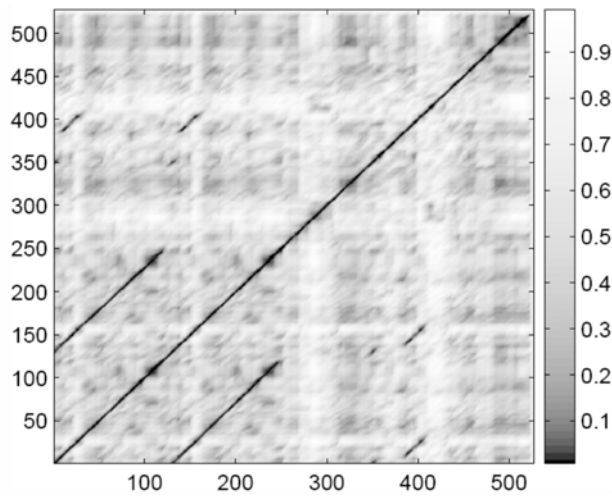


Serra/Gomez (ICASSP 2008): Used for Cover Song ID

Discrete structure \rightsquigarrow suitable for indexing?

Transposition Invariance

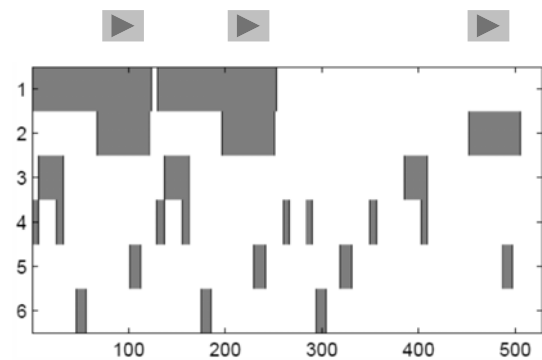
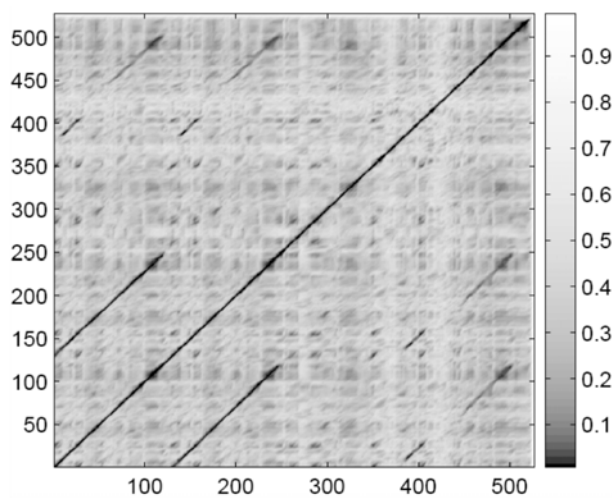
Example: Beethoven "Tempest"



Self-similarity matrix

Transposition Invariance

Example: Beethoven "Tempest"



Transposition-invariant self-similarity matrix

Conclusions: Audio Structure Analysis

Challenge: Musical variations

- Timbre, dynamics, tempo
- Musical key \rightsquigarrow cyclic chroma shifts
- Major/minor
- Differences at note level / improvisations

Conclusions: Audio Structure Analysis

Strategy: Matrix enhancement

- Filtering techniques / contextual information
 - Cooper/Foote (ISMIR 2002)
 - Müller/Kurth (ICASSP 2006)
- Transposition-invariant similarity matrices
 - Goto (ICASSP 2003)
 - Müller/Clausen (ISMIR 2007)
- Higher-order similarity matrices
 - Peeters (ISMIR 2007)

Novel Approach for Audio Thumbnailing

Original approach: Two steps

1. Path extraction
 - Paths of poor quality (fragmented, gaps)
 - Regions of constant (low) cost
 - Curved paths
2. Grouping:
 - Noisy relations (missing, distorted, overlapping)
 - Transitivity computation difficult

Both steps are problematic!

Our main idea: Do both, path extraction and grouping, jointly

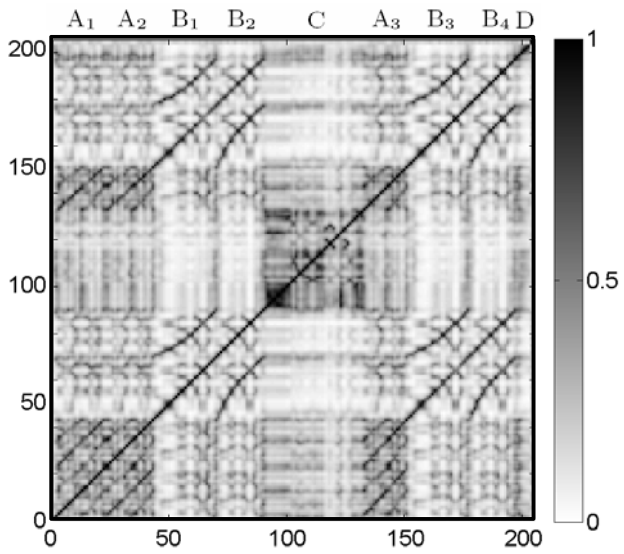
- One optimization scheme for both steps
- Stabilizing effect
- Efficient

Novel Approach for Audio Thumbnailing

Our main idea: Do both path extraction and grouping jointly

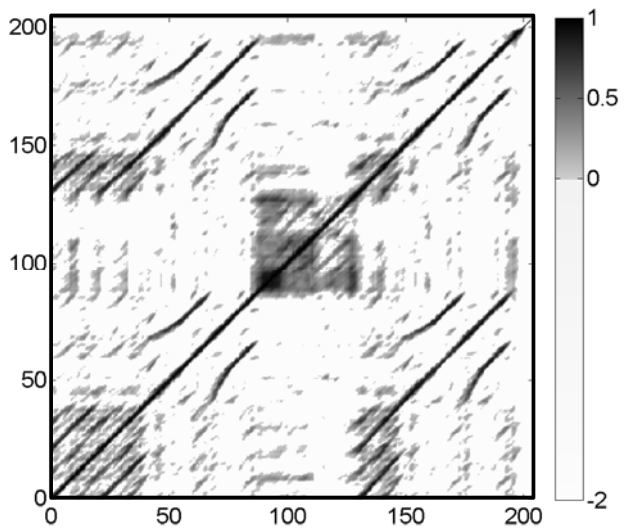
- For each audio **segment** we define a **fitness** value
- This fitness value expresses “how well” the segment explains the entire audio recording
- The segment with the highest fitness value is considered to be the **thumbnail**
- As main technical concept we introduce the notion of a **path family**

Fitness Measure



Self-similarity matrix

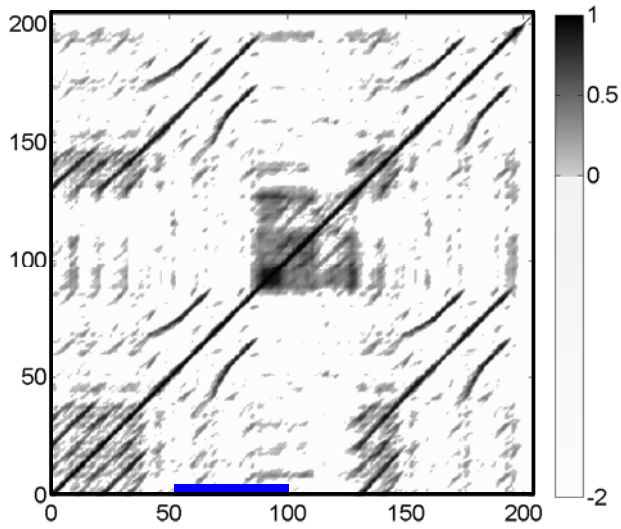
Fitness Measure



Self-similarity matrix

- Smoothing
- Transposition-Invariance
- Normalization
- Thresholding
- Negative score

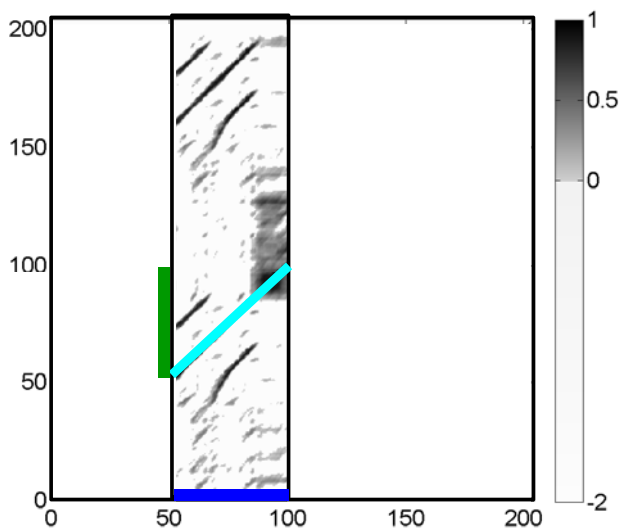
Fitness Measure



Path over segment

- Consider a fixed **segment**

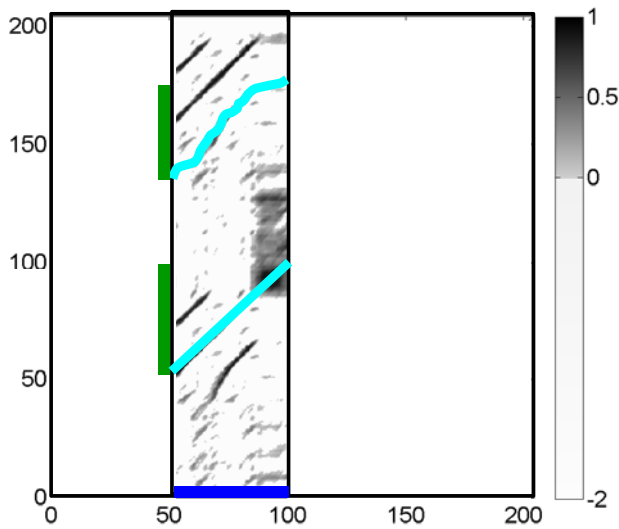
Fitness Measure



Path over segment

- Consider a fixed **segment**
- Path** over **segment**
- Induced segment**
- Score is high

Fitness Measure

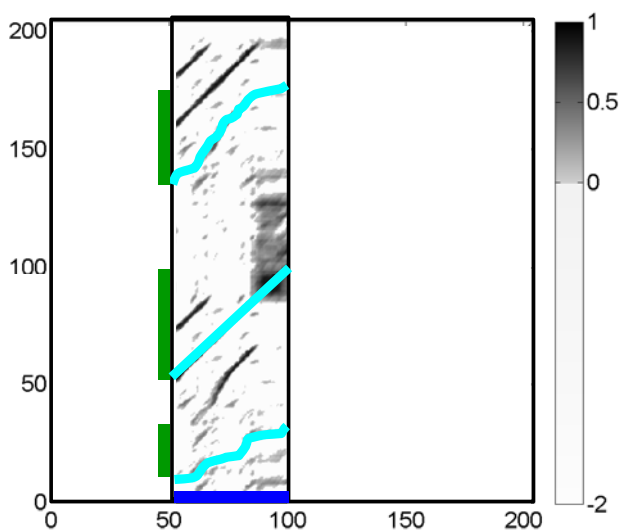


Path over segment

- Consider a fixed **segment**
- **Path** over **segment**
- **Induced segment**
- Score is high

- **A second path** over **segment**
- **Induced segment**
- Score is not so high

Fitness Measure



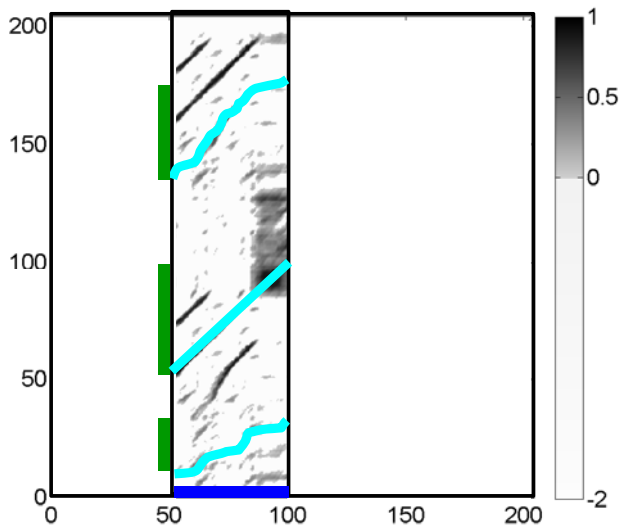
Path over segment

- Consider a fixed **segment**
- **Path** over **segment**
- **Induced segment**
- Score is high

- **A second path** over **segment**
- **Induced segment**
- Score is not so high

- **A third path** over **segment**
- **Induced segment**
- Score is very low

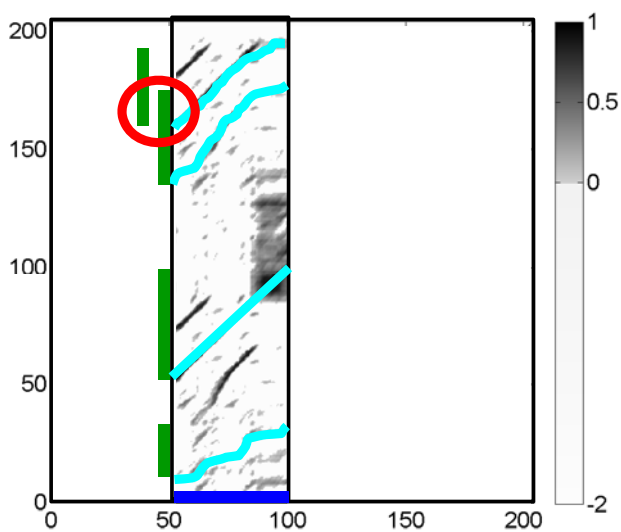
Fitness Measure



Path family

- Consider a fixed **segment**
- A path family over a **segment** is a family of paths such that the **induced segments** do **not overlap**.

Fitness Measure

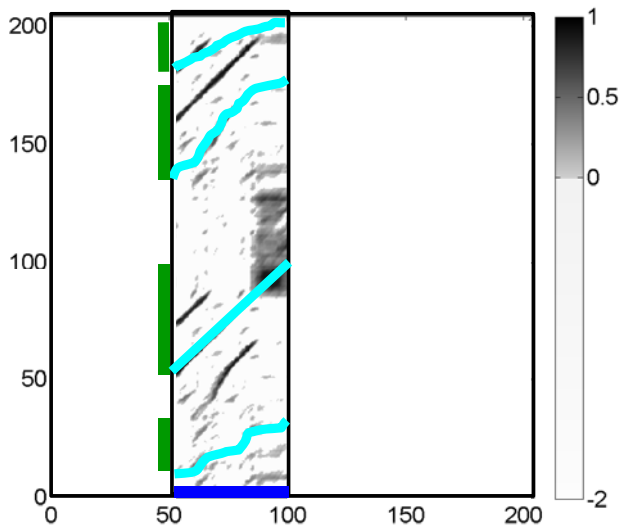


Path family

- Consider a fixed **segment**
- A path family over a **segment** is a family of paths such that the **induced segments** do **not overlap**.

This is **not** a path family!

Fitness Measure



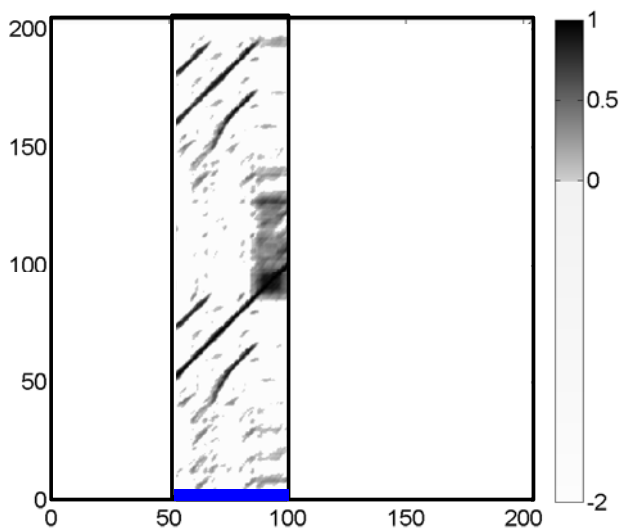
Path family

- Consider a fixed **segment**
- A path family over a **segment** is a family of paths such that the **induced segments** do **not overlap**.

This is a path family!

(Even though not a good one)

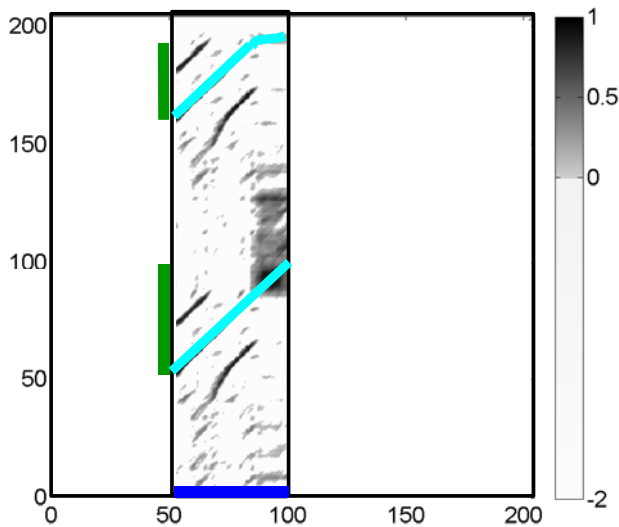
Fitness Measure



Optimal path family

- Consider a fixed **segment**

Fitness Measure

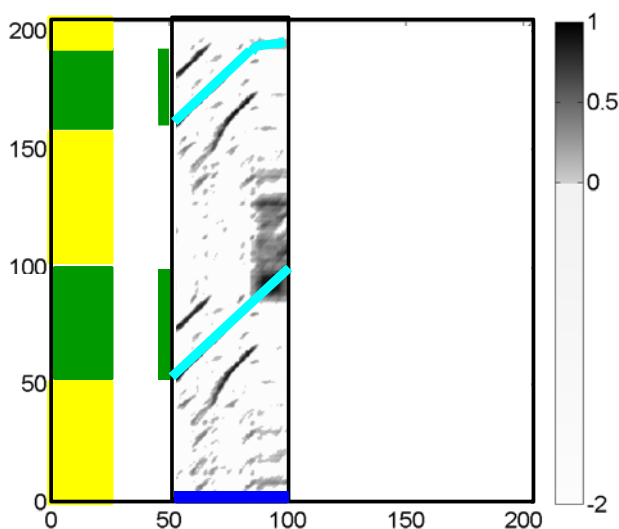


Note: This optimal path family can be computed using dynamic programming.

Optimal path family

- Consider a fixed **segment**
- Consider over the **segment** the **optimal path family**, i.e., the path family having maximal overall score.
- Call this value:
 $\text{Score}(\text{segment})$

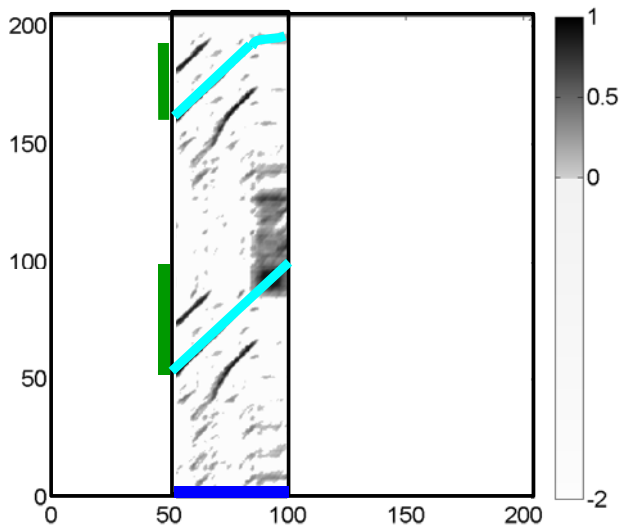
Fitness Measure



Optimal path family

- Consider a fixed **segment**
- Consider over the **segment** the **optimal path family**, i.e., the path family having maximal overall score.
- Call this value:
 $\text{Score}(\text{segment})$
- Furthermore consider the amount covered by the **induced segments**.
- Call this value:
 $\text{Coverage}(\text{segment})$

Fitness Measure

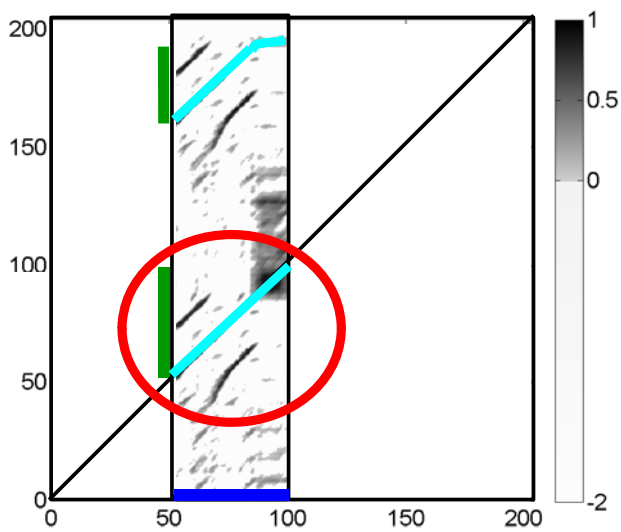


P := Score(segment)
R := Coverage(segment)

Fitness

- Consider a fixed segment

Fitness Measure

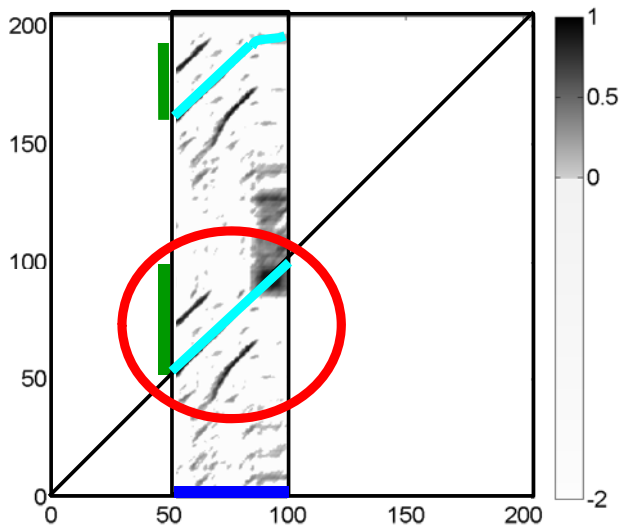


P := Score(segment)
R := Coverage(segment)

Fitness

- Consider a fixed segment
- Self-explanation are trivial!

Fitness Measure



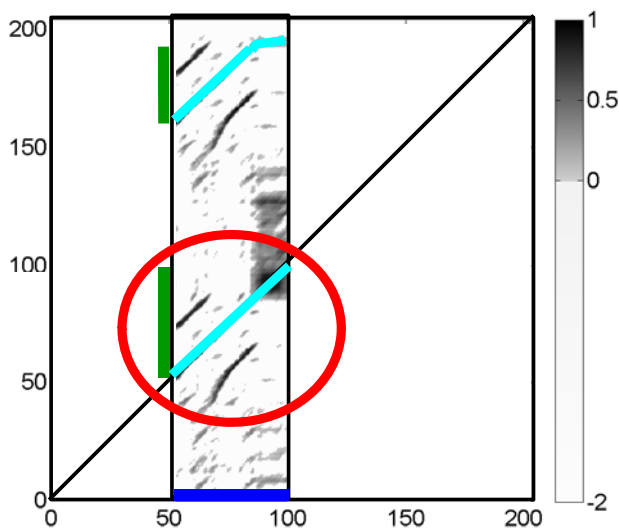
$P := \text{Score}(\text{segment}) - \text{length}(\text{segment})$

$R := \text{Coverage}(\text{segment}) - \text{length}(\text{segment})$

Fitness

- Consider a fixed **segment**
- Self-explanation are trivial!**
- Substract length of **segment**

Fitness Measure



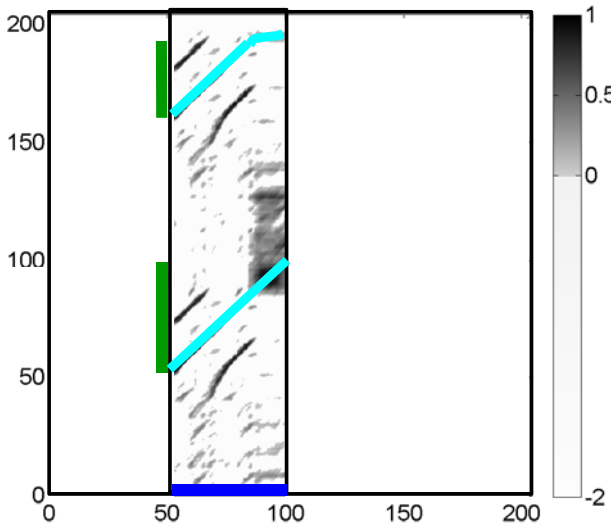
$P := \text{Normalize}(\text{Score}(\text{segment}) - \text{length}(\text{segment})) \in [0,1]$

$R := \text{Normalize}(\text{Coverage}(\text{segment}) - \text{length}(\text{segment})) \in [0,1]$

Fitness

- Consider a fixed **segment**
- Self-explanation are trivial!**
- Substract length of **segment**
- Normalization

Fitness Measure



Fitness

- Consider a fixed **segment**

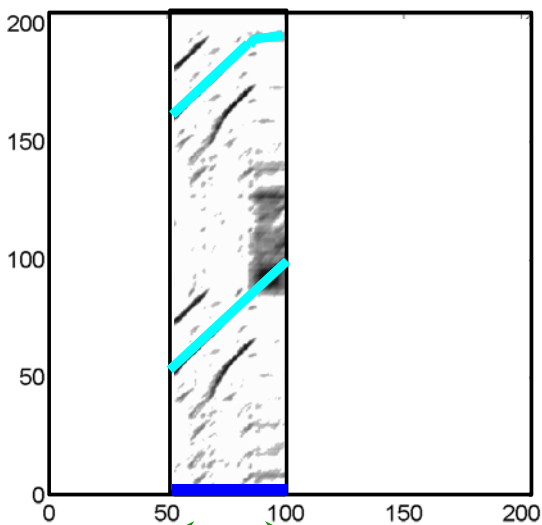
Fitness(segment)

$$F := 2 \cdot P \cdot R / (P + R)$$

$P := \text{Normalize}(\text{Score}(\text{segment}) - \text{length}(\text{segment})) \in [0,1]$

$R := \text{Normalize}(\text{Coverage}(\text{segment}) - \text{length}(\text{segment})) \in [0,1]$

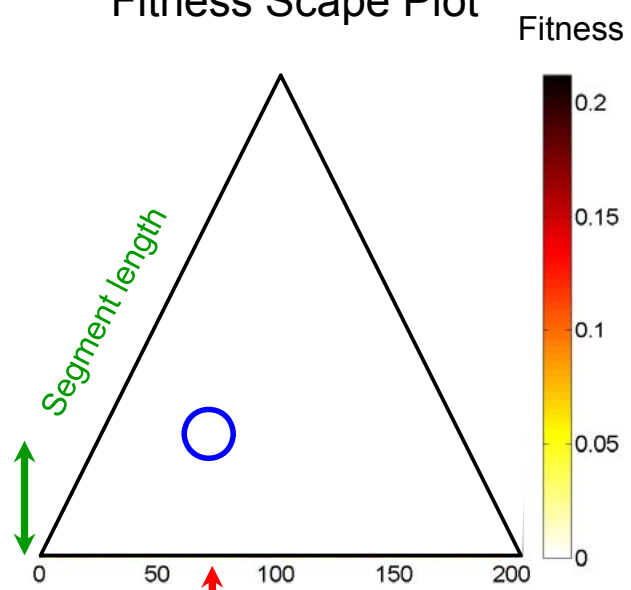
Thumbnail



Segment length

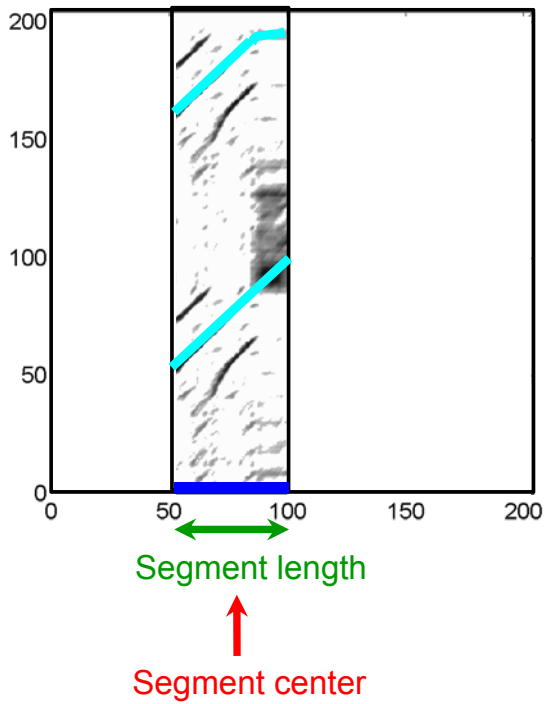
Segment center

Fitness Scape Plot

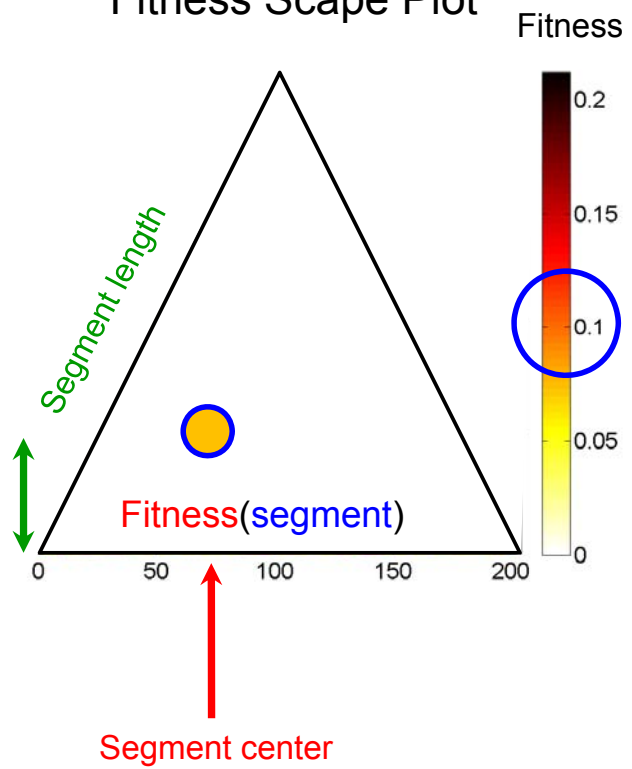


Segment center

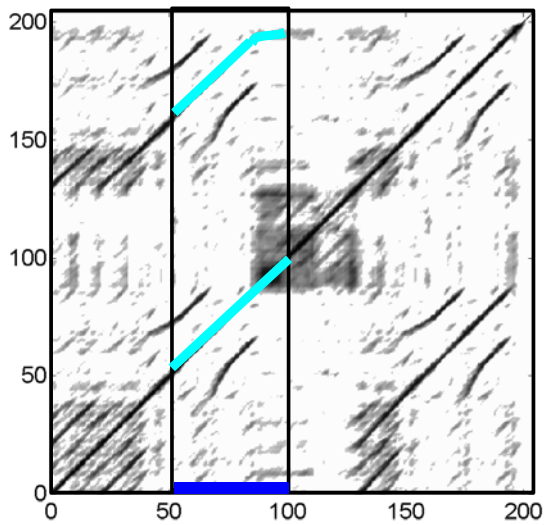
Thumbnail



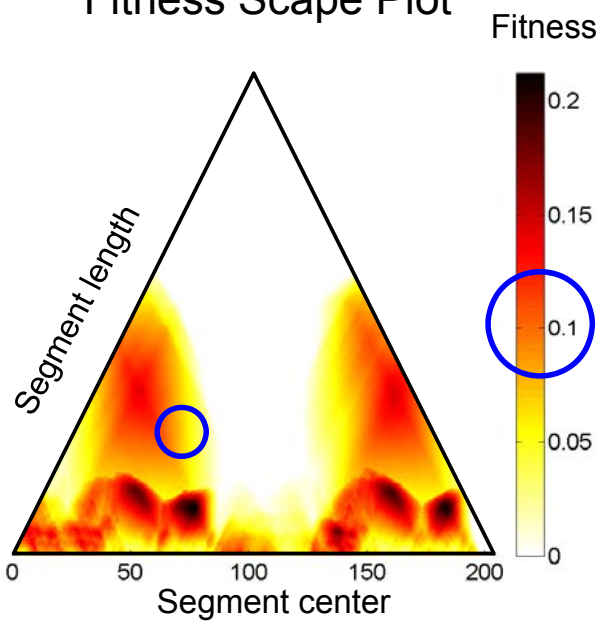
Fitness Scape Plot



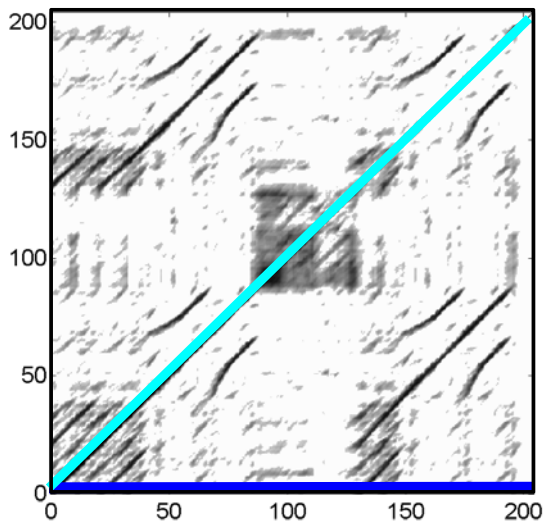
Thumbnail



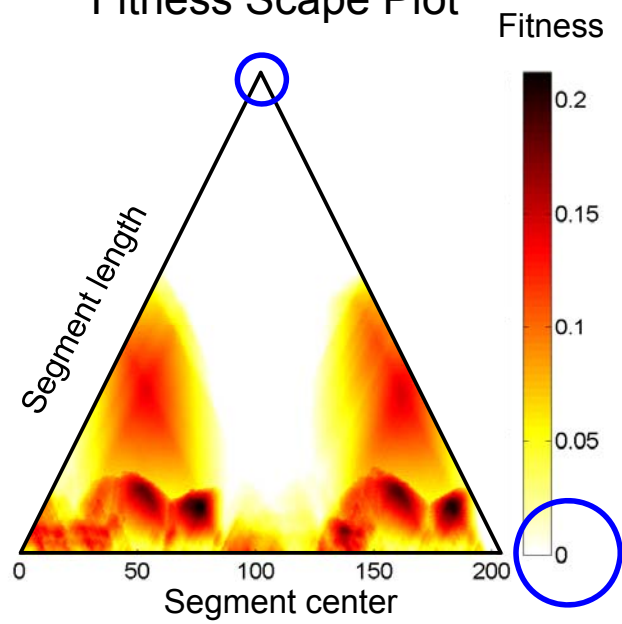
Fitness Scape Plot



Thumbnail

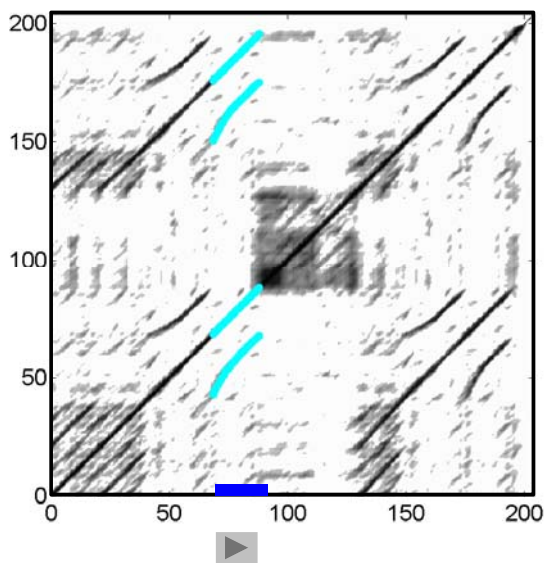


Fitness Scape Plot

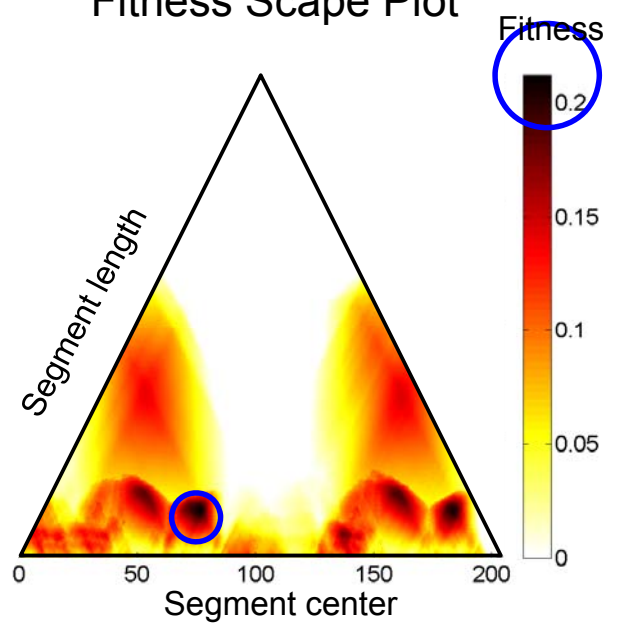


Note: Self-explanations are ignored → fitness is zero

Thumbnail

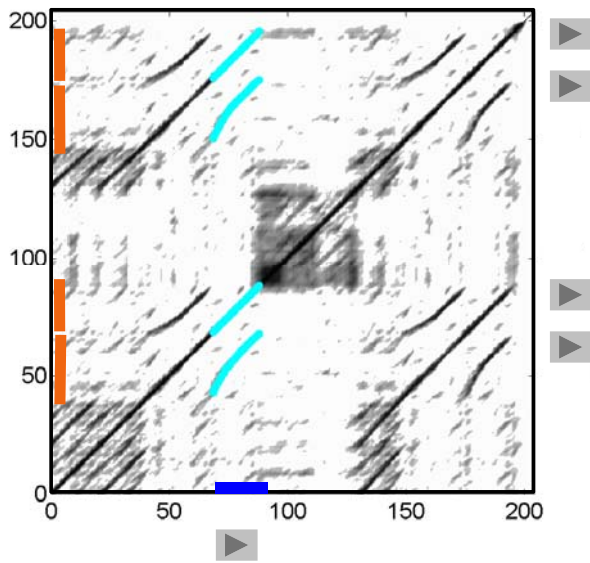


Fitness Scape Plot

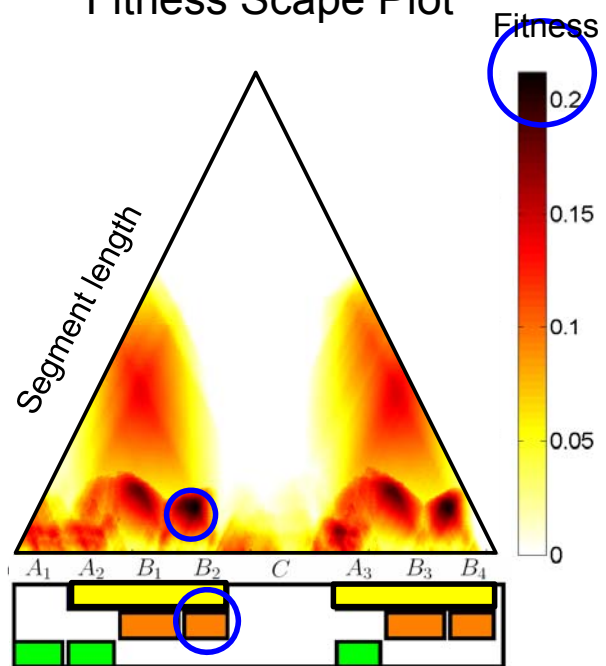


Thumbnail := segment having the highest fitness

Thumbnail

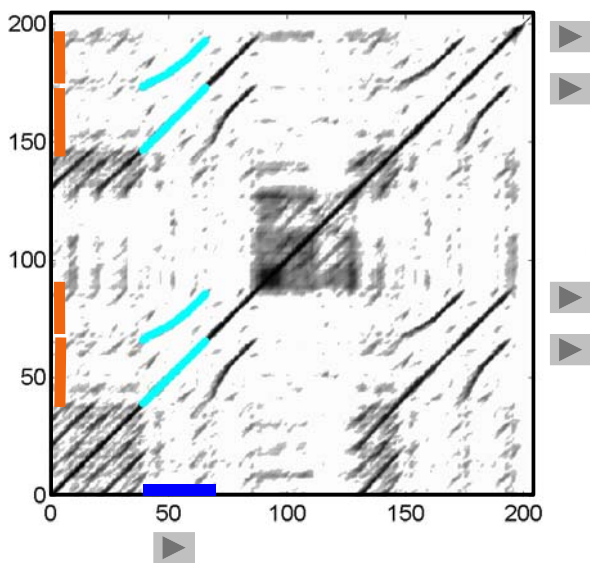


Fitness Scape Plot

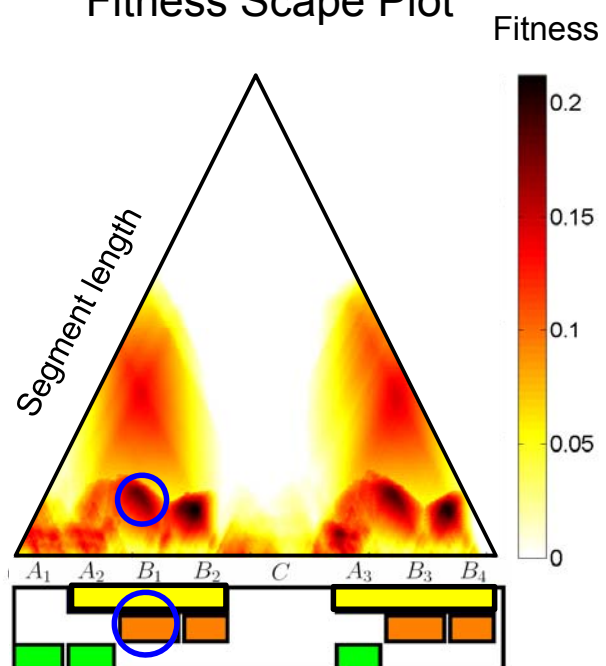


Example: Brahms Hungarian Dance No. 5 (Ormandy)

Thumbnail

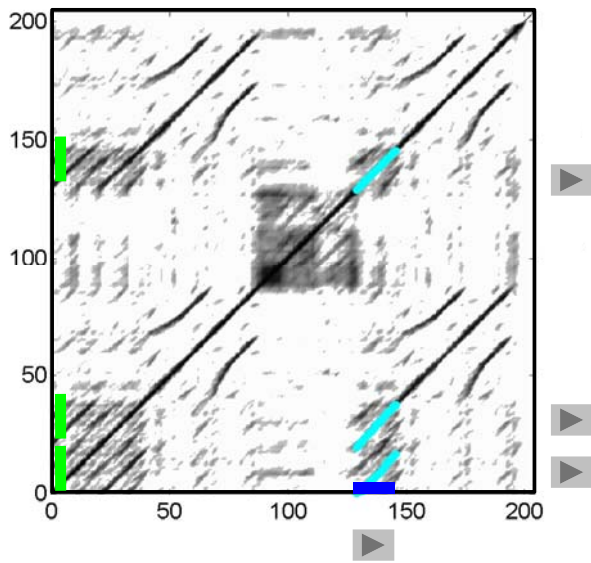


Fitness Scape Plot

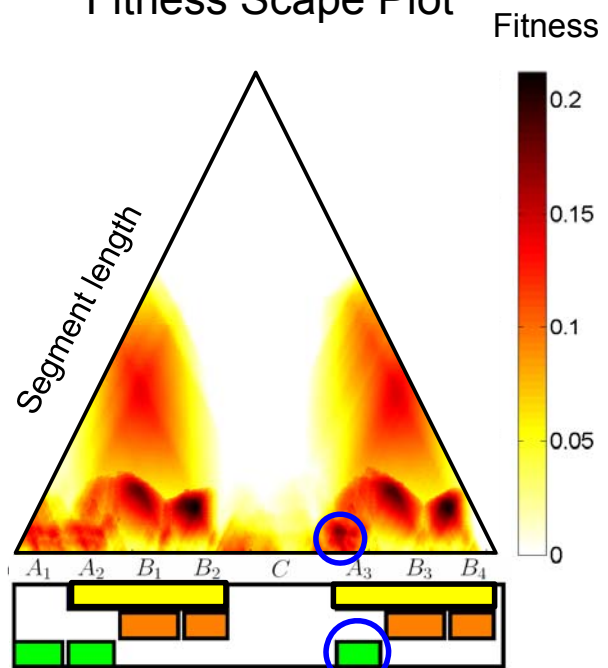


Example: Brahms Hungarian Dance No. 5 (Ormandy)

Thumbnail

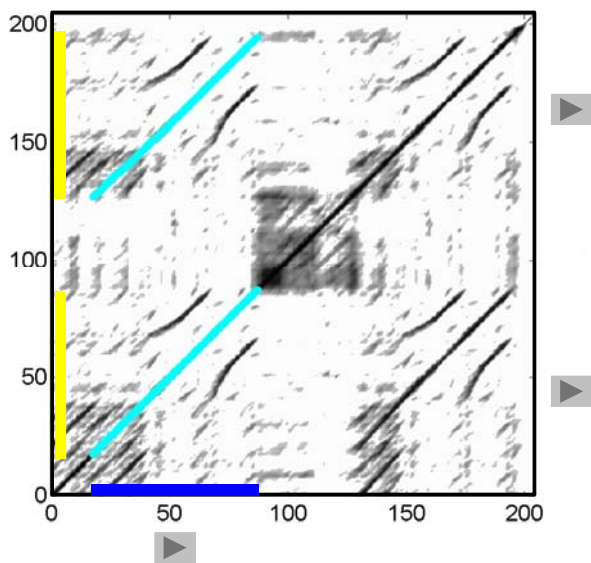


Fitness Scape Plot

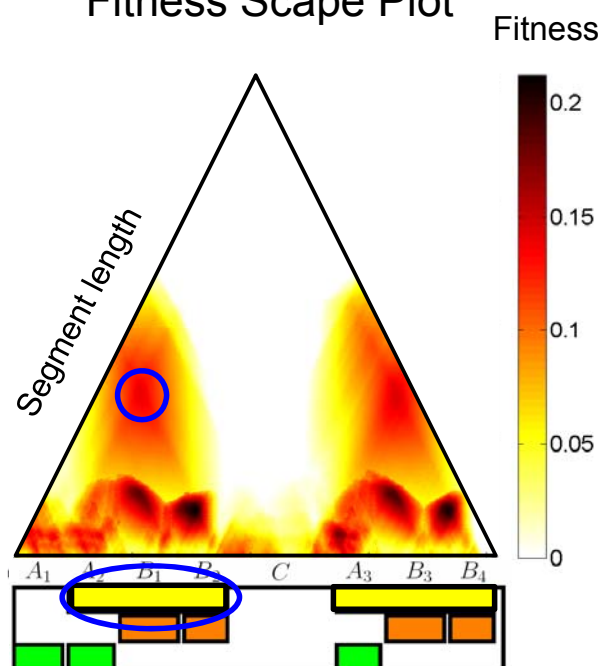


Example: Brahms Hungarian Dance No. 5 (Ormandy)

Thumbnail

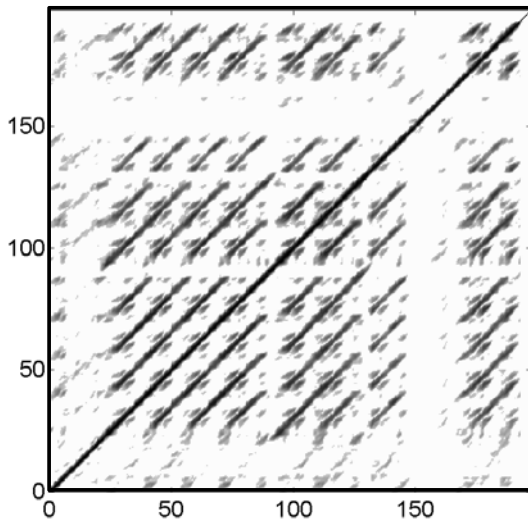


Fitness Scape Plot

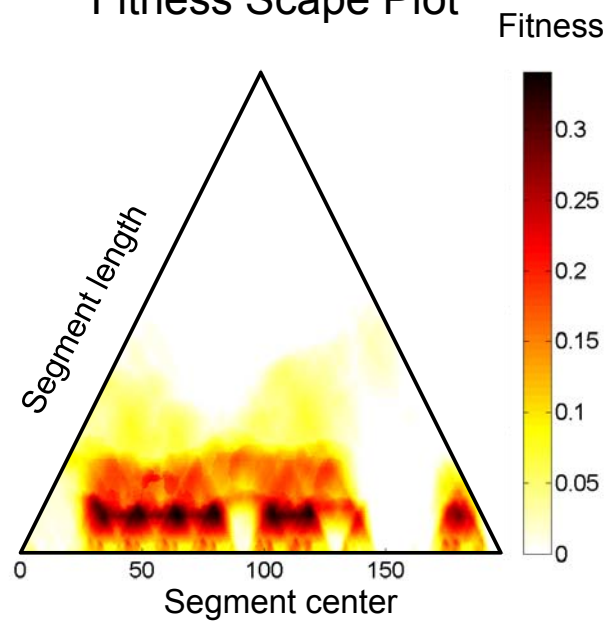


Example: Brahms Hungarian Dance No. 5 (Ormandy)

Thumbnail

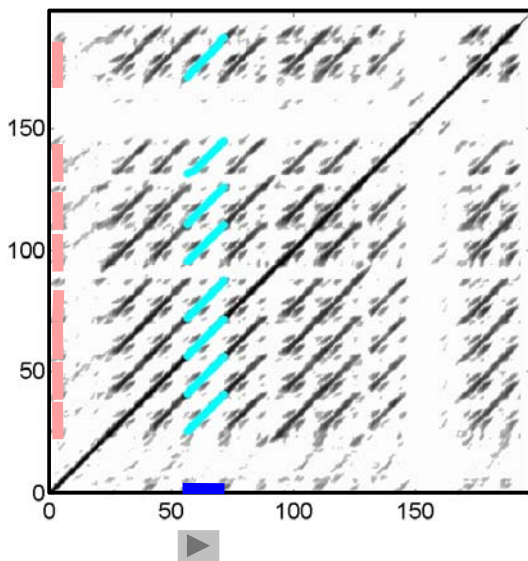


Fitness Scape Plot

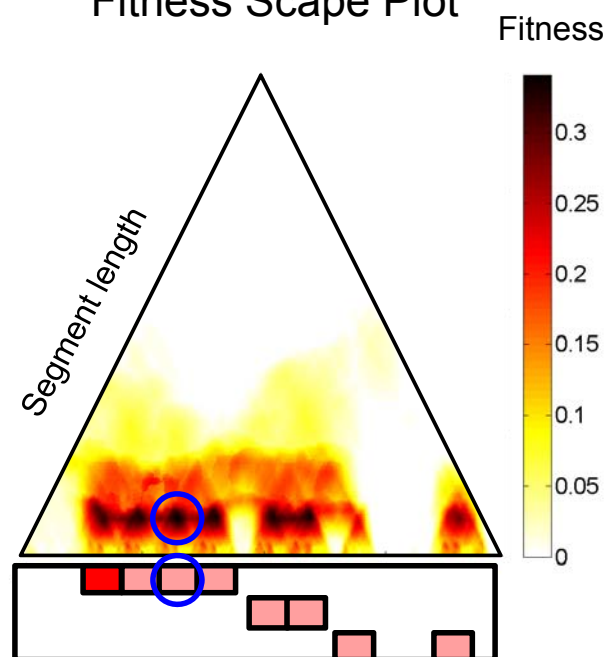


Example: Zager & Evans "In The Year 2525"

Thumbnail

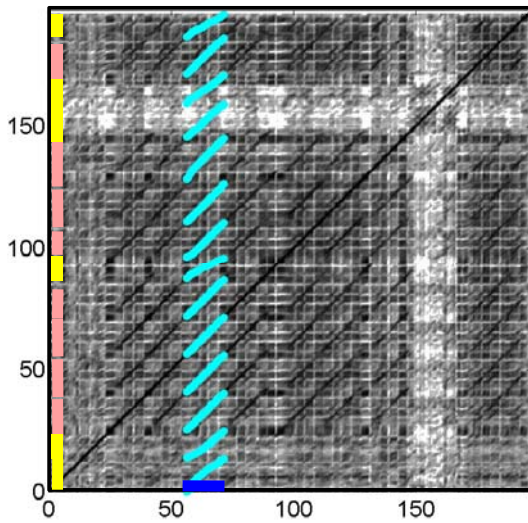


Fitness Scape Plot

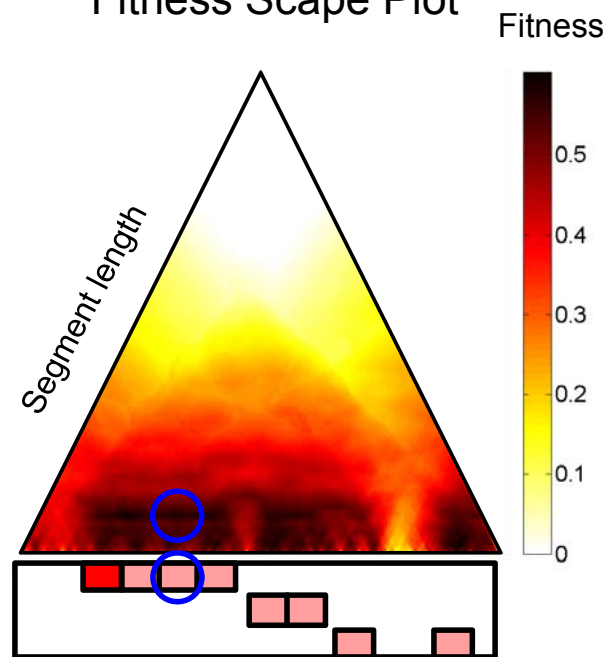


Example: Zager & Evans "In The Year 2525"

Thumbnail

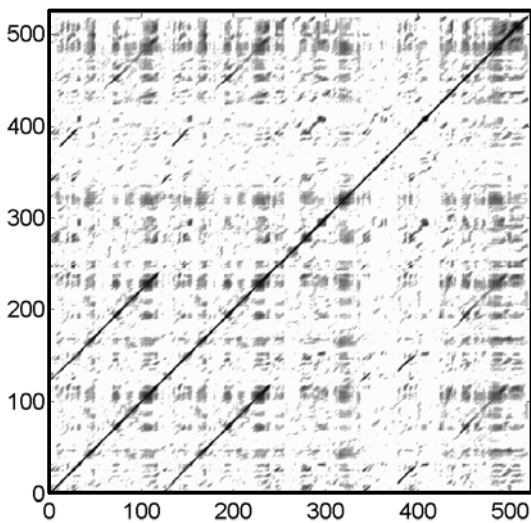


Fitness Scape Plot

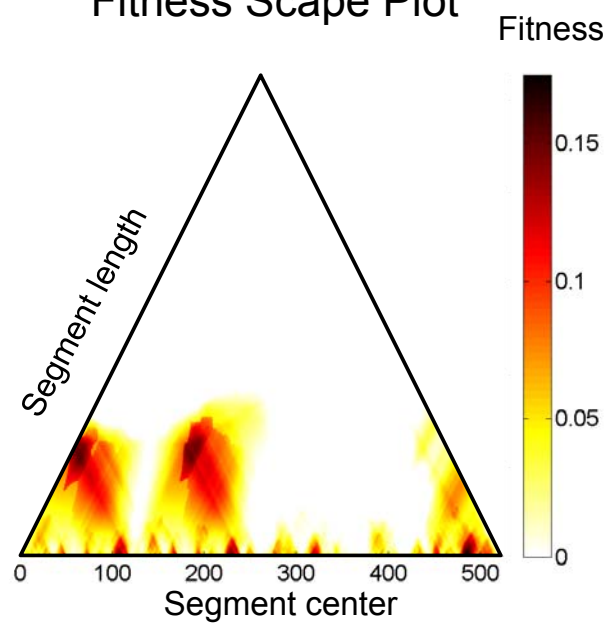


Example: Zager & Evans "In The Year 2525"

Thumbnail

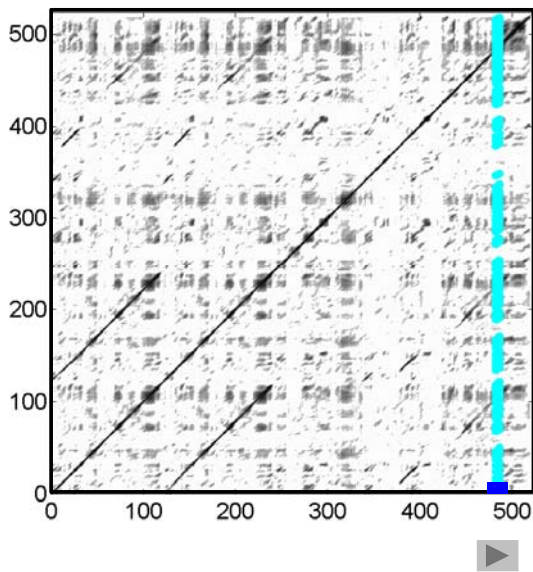


Fitness Scape Plot

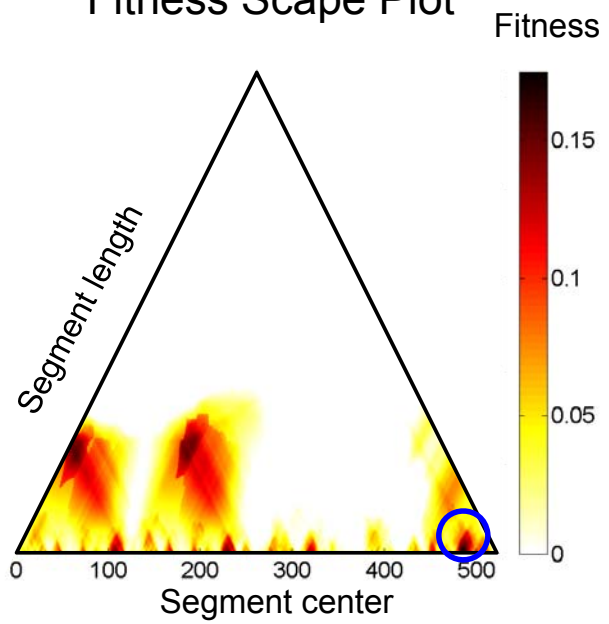


Example: Beethoven "Tempest", Pollini

Thumbnail

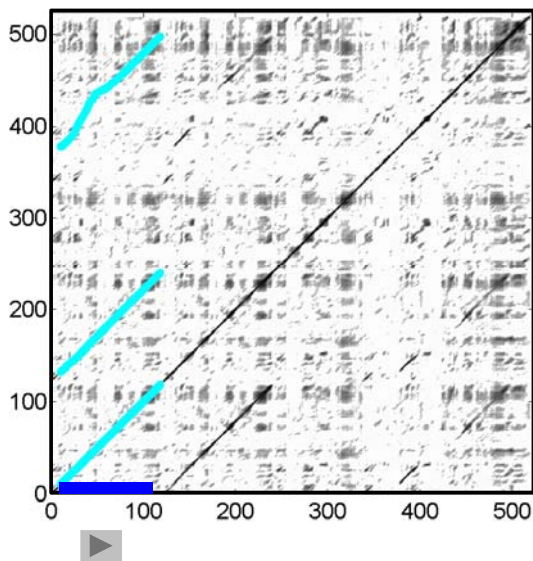


Fitness Scape Plot

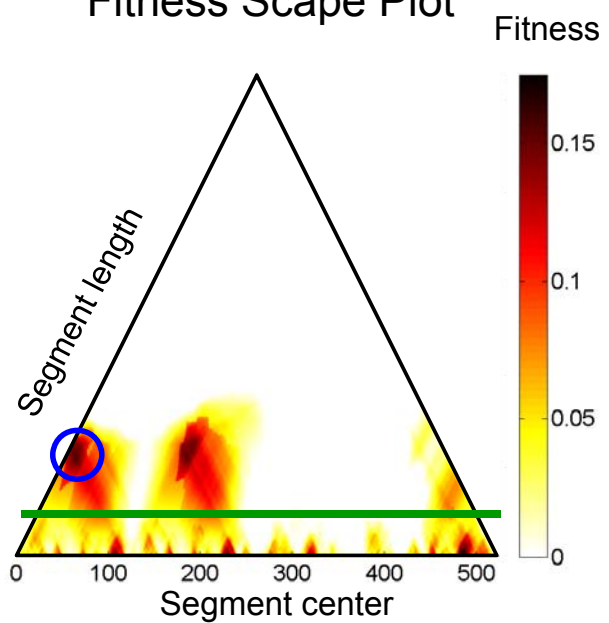


Example: Beethoven "Tempest", Pollini

Thumbnail



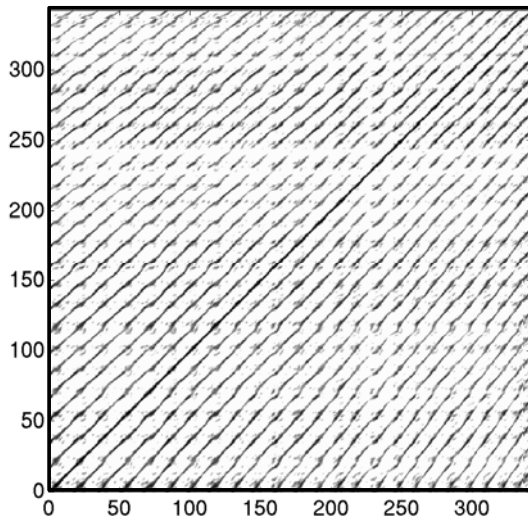
Fitness Scape Plot



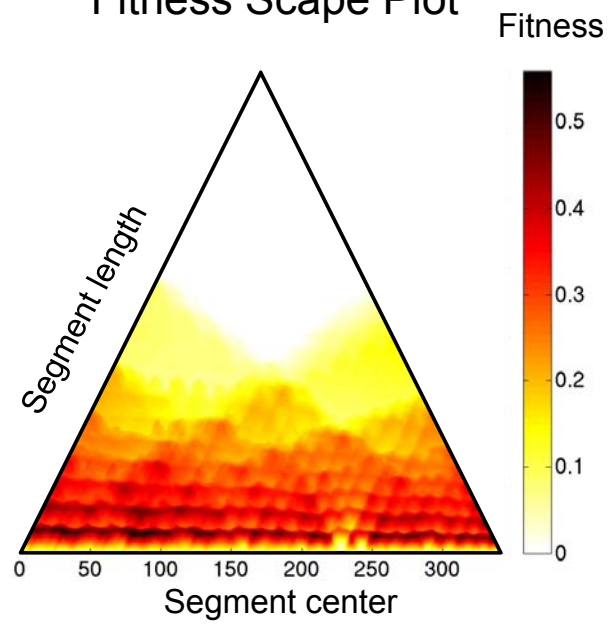
Example: Beethoven "Tempest", Pollini

Musical knowledge: Minimum length for thumbnail

Thumbnail

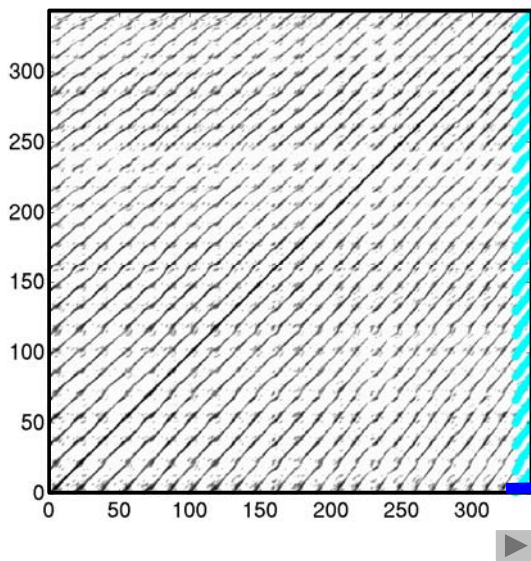


Fitness Scape Plot

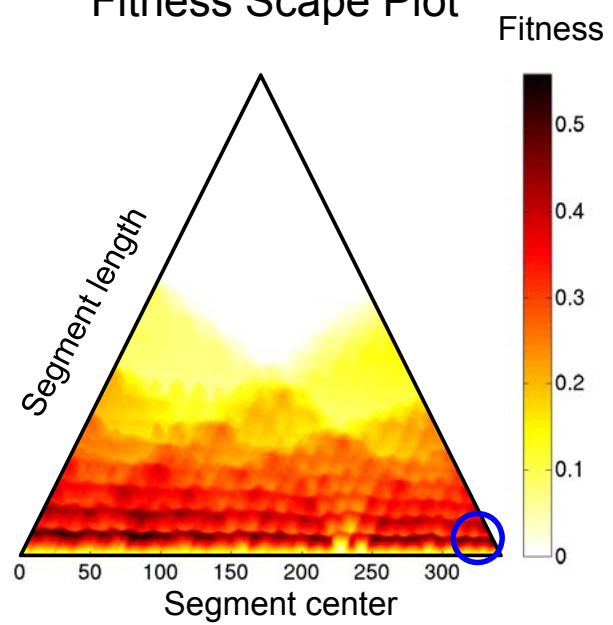


Example: NLB72246

Thumbnail



Fitness Scape Plot



Example: NLB72246

Conclusions

- **Path family:** Couples path extraction and grouping
- **Fitness:** Quality of segment in context of entire recording
 - Combination of score and coverage
 - Trivial self-explanation are disregarded
- **Thumbnail:** Segment of maximal fitness
- **Fitness scape plot:** Global structure visualization

Future work:

- **Multiscale approach**
- **Combination with novelty detection**
- **Interface for structure navigation**

