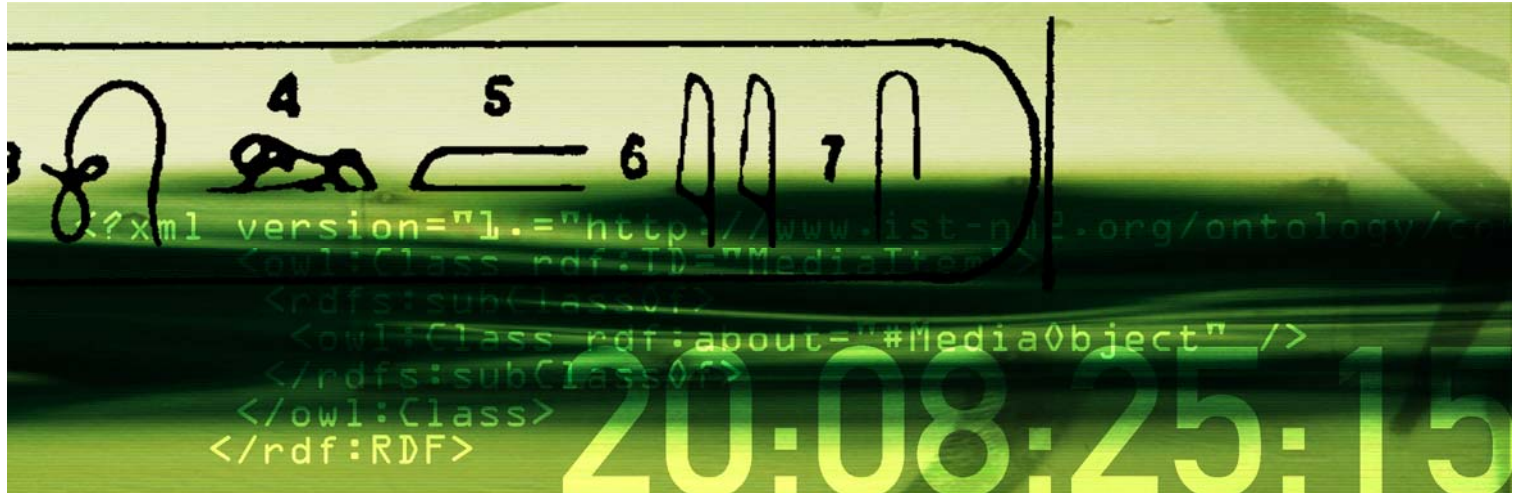
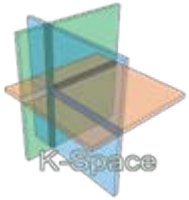


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## Eliminating the Back-Tracking Step in the Longest Common Subsequence (LCSS) Algorithm for Video Sequence Matching

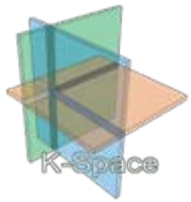
Werner Bailer  
SAMT 2008, Koblenz, 2008-12-04



# Outline

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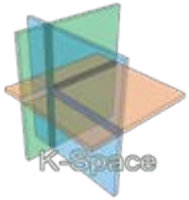
- Motivation
  - detecting repeated video content
- Core Problem: distance between video sequences
- Proposed distance measure
- Optimisation
- Demo



# Core Problem: Measuring Distance between Video Sequences

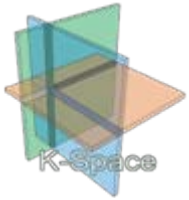
- Requirements
  - support any type of features (audio/video, discrete or continuous values, scalar/vector, custom distance functions between two feature values/vectors)
  - support partial match of sequences
  - different timing of sequences
  - support gaps and insertions in the matching parts
  - minimum length of useful match
- Approaches
  - Dynamic Time Warping: assign each element of sequence A to nearest element of sequence B
  - String edit distance: identical, replace, insert, delete

# Distance Measure for Video Sequences (1)



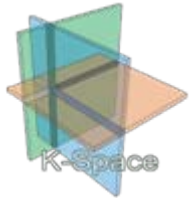
- match sequences of feature vectors extracted from the input videos
- based on Longest Common Subsequence (LCSS) algorithm
  - variant of string edit distance
  - build matrix from matches
  - find longest matching subsequence which may have gaps by back-tracking
  - original LCSS algorithm assumes discrete input values, match := equality of input values
- LCSS for vectors of continuous values
  - proposed for 2D trajectories [Vlachos et al., 2002]
  - match := Euclidian distance between elements  $< \varepsilon$  and offset between elements  $< \delta$

# Distance Measure for Video Sequences (2)



- for matching feature sequences of videos
  - replace  $\varepsilon$  by a vector of thresholds  $\theta_{\text{sim}} = \{\varepsilon_1, \dots, \varepsilon_F\}$  for  $m$  features  $1, \dots, F$ , which are weighted by weights  $\{w_1, \dots, w_F\}$
  - match features  $f$  with appropriate distance functions
  - discard  $\delta$ , absolute temporal distance of feature vectors in the sequence is irrelevant
  - accept all matches longer than a minimum length  $\theta_{\text{len}}$
  - introduce maximum gap  $\gamma$  between two consecutive matching feature vectors
  - consequence of gap constraint:
    - not just *longest* common subsequence (might have gaps)
    - but all *sufficiently long* ( $> \theta_{\text{len}}$ ) subsequences
  - similarity := length of match / min(lengths of input sequences)

# Distance Measure for Video Sequences (3)



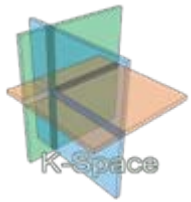
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## 1D Example

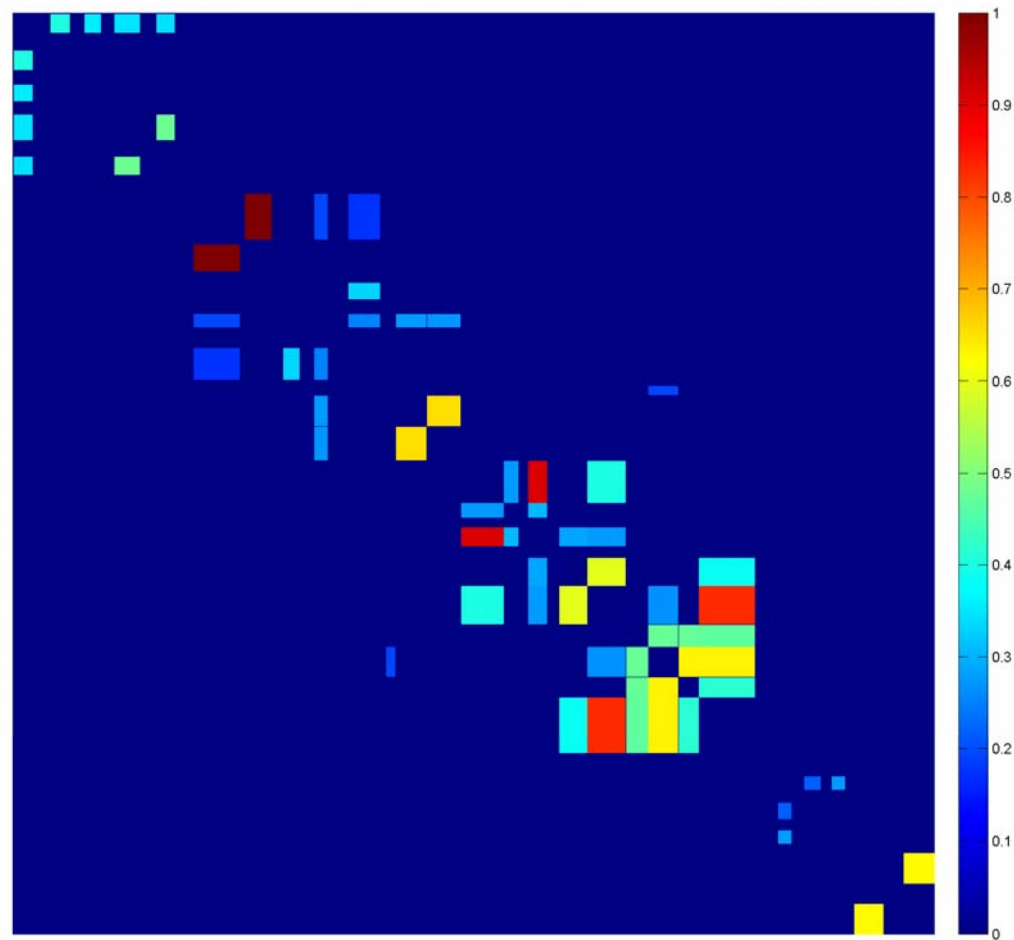
- input sequences top/left
- $\theta_{\text{sim}} = 0.5$
- $\theta_{\text{len}} = 3$
- $\gamma = 1$

	2.3	4.7	1.2	3.3	2.2	1.4
1.3	0	0	1	1	1	1
2.5	1	1	1	1	2	2
3.4	1	1	1	2	2	2
2.4	1	1	1	2	3	3
4.6	1	2	2	2	3	3
1.5	1	2	3	3	3	4
2.6	1	2	3	3	4	4

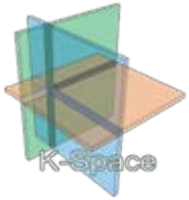
# Distance Measure for Video Sequences (4)



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# Distance Measure – Optimisation: Eliminating Back-Tracking

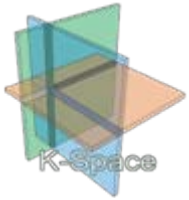


- proposed optimisation

- keep while building matrix
  - last match in each line and list of current sequences
- for each match
  - find nearest previous match (city block distance – 2)
  - search in  $-(\gamma + 1)$  lines, add to this sequence:  $O(1)$

	2.3	4.7	1.2	3.3	2.2	1.4
1.3	0	0	1	1	1	1
2.5	1	1	1	1	2	2
3.4	1	1	1	2	2	2
2.4	1	1	1	2	3	3
4.6	1	2	2	2	3	3
1.5	1	2	3	3	3	4
2.6	1	2	3	3	4	4





- Video Browsing Tool
- Rushes Skimming

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