SEGMENTATION OF TV SHOWS INTO SCENES

USING SPEAKER DIARIZATION AND SPEECH RECOGNITION

Context

- Exponential growth of video content
 - Mostly TV and web
- Content-based video indexing
 - Query/browse by people
 - REPERE challenge
 - Query/browse by semantic concept
 - GdR ISIS *IRIM* project
 - NIST TRECVid Semantic Indexing task
- Make content consumption easier
 - Automatic summarization
 - PhD thesis with IRIT (Ph. Ercolessi)

(More) Context

- Automatic summarization of TV series
 - At the episode level
 - Shot segmentation
 - Scene segmentation
 - Plot (or substory) deinterlacing
 - Episode summarization
 - « Previously, on Lost... »
 - Browse by plot
 - At the collection level
 - Cross-episode plot
 - Episode summarization wrt. whole collection

Shot, scene, sequence and plot

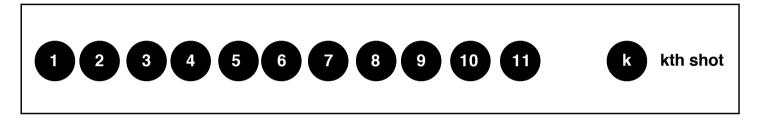
- Shot
 - a part of a film between two camera cuts
- Scene
 - (each author working on scene segmentation uses its own definition)
 - group of consecutive shots
 - temporal continuity / unity of time
 - semantic coherence / unity of action
- Sequence
 - group of consecutive scenes
 - semantic coherence
 - aka story or topic in TV news
 - made of multiple stories introduced by the anchor
- Plot
 - group of (not necessarily consecutive) sequences
 - modern TV show episodes usually have multiple interlaced plots
 - two plots can overlap

Outline

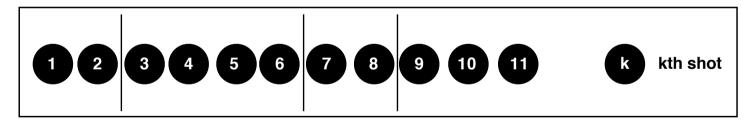
- Context
- Definitions & notations
- Principle
 - Scene transition graph with color histograms
 - Generalized STG
- Multimodal extension
 - Speaker diarization & speech recognition
 - Multimodal fusion

Scene segmentation

Input: shot boundaries

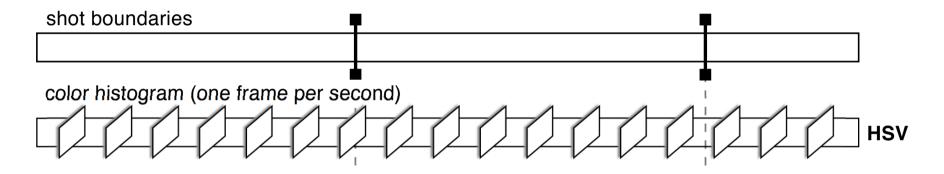


Output: scene boundaries



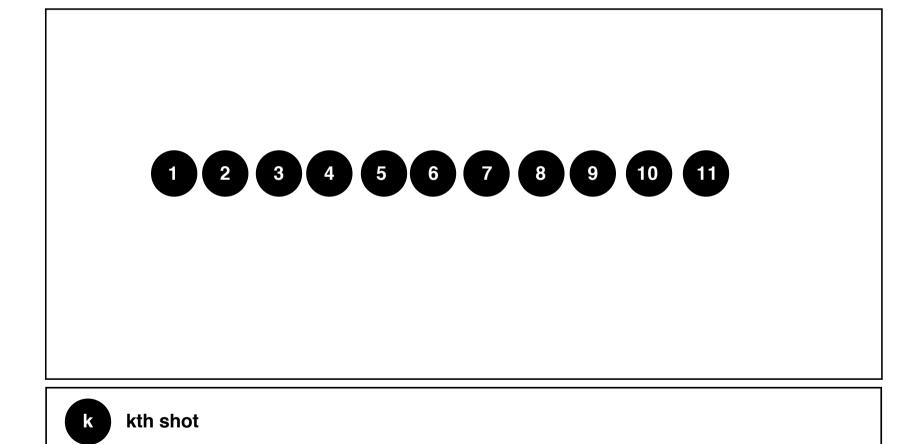
- Classification problem on shot boundaries
 - precision, recall, F₁-measure

Color



- Multiple frame per shot, one color histogram per frame
- $d_{ij}^{
 m HSV}$: minimum distance between all possible pairs of histograms from shots i and j

Segmentation of Video by Clustering and Graph Analysis / Yeung (1998)



Notations

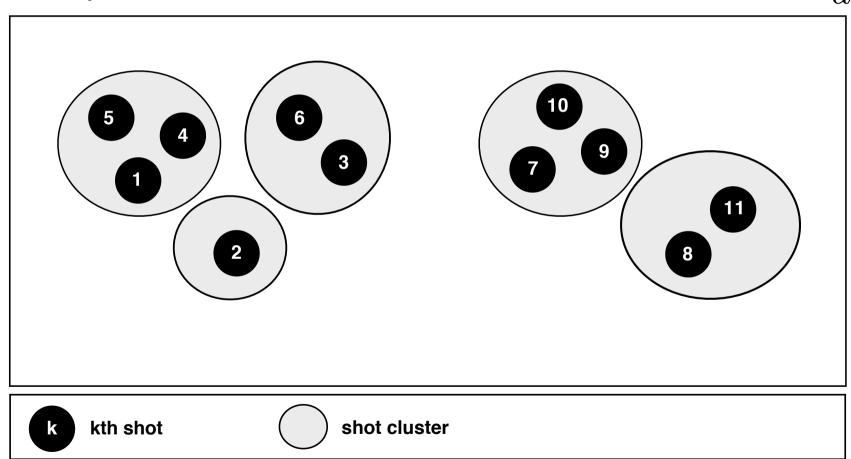
 d_{ij} dissimilarity between shots i and j t_{ij} temporal distance between shots i and j D_{ij} combined distance between shots i and j

$$D_{ij} = \begin{cases} d_{ij} & \text{if } t_{ij} < \Delta_t \\ +\infty & \text{otherwise} \end{cases}$$

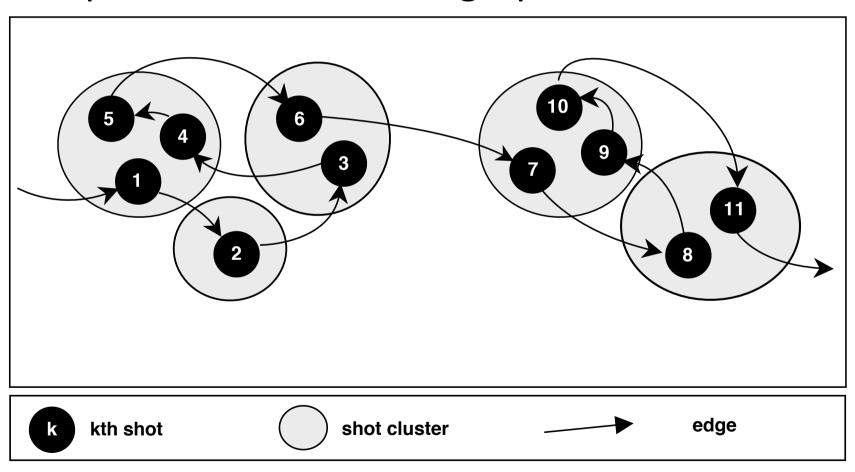
 Δ_t temporal distance threshold Δ_d combined distance threshold

• Step 1: complete-link agglomerative clustering

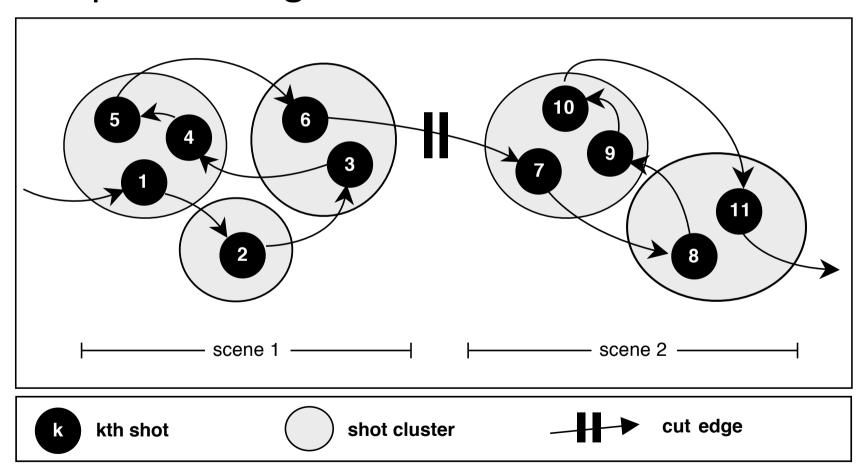
 Δ_d



• Step 2: scene transition graph



• Step 3: cut-edge detection



HSV/STG Results

Corpus

- First eight episodes of Ally McBeal TV shows
- 5 hours of videos, 5564 shots and 306 scenes

Evaluation

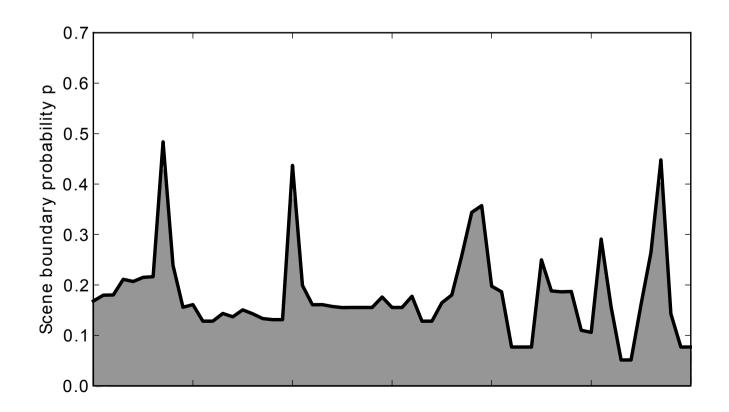
- Leave-one-episode-out cross validation
- Two thresholds $\,\Delta_t\,$ and $\,\Delta_d\,$

	Precision	Recall	F-Measure	# Scenes
HSV (STG)	0.256	0.533	0.449	461

- Limitation:
 - Every pair (Δ_t, Δ_d) leads to a different set of detected scene boundaries.
 - The optimal values are very dependent on the video
- Proposition (Sidiropoulos, 2011):
 - Generalized STG

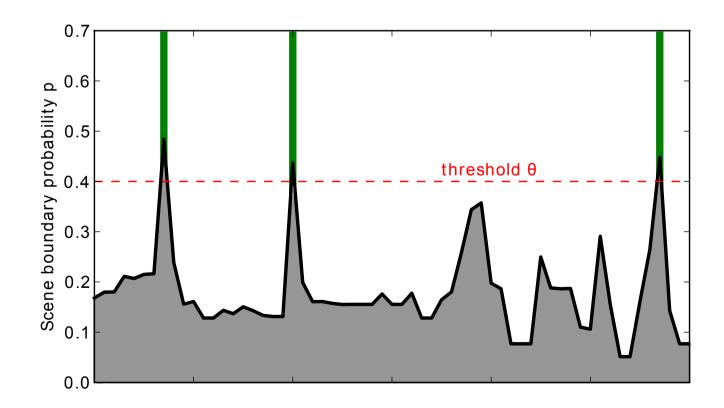
Generalized STG

- Large set of STGs by selecting random Δ_t and Δ_d
- Scene boundary probability



Generalized STG

- Large set of STGs by selecting random Δ_t and Δ_d
- Scene boundary probability
- Unique threshold $\, heta$



HSV/GSTG Results

Corpus

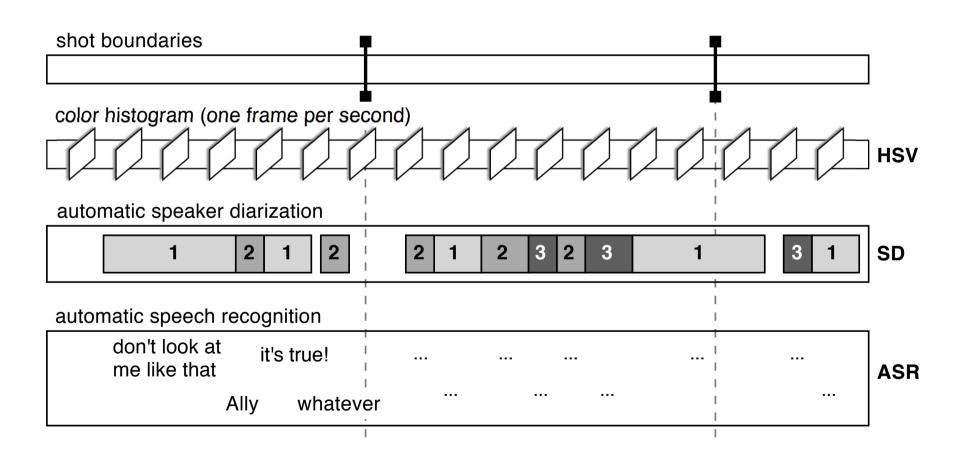
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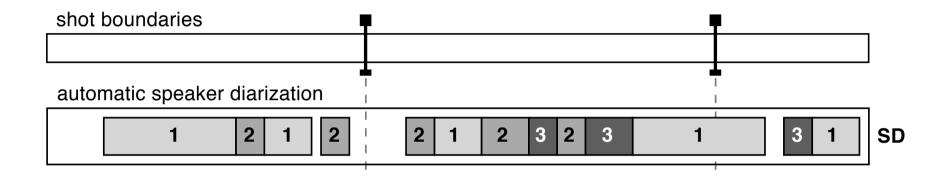
Leave-one-episode-out cross validation

	Precision	Recall	F-Measure	# Scenes
HSV/STG	0.256	0.533	0.449	461
HSV/GSTG	0.447	0.566	0.487	403

Multiple modalities, multiple d_{ij}

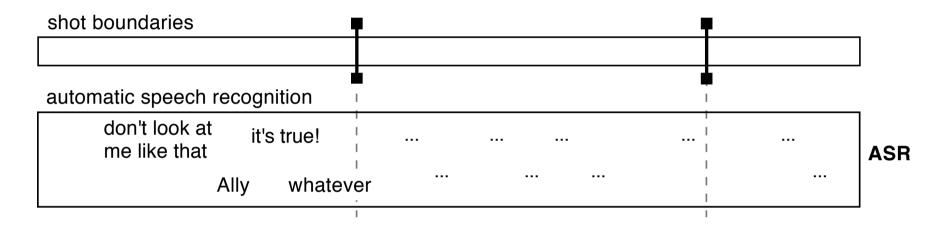


Speaker diarization



- Multiple speakers per shot, but only one descriptor
 - Term Frequency / Inverse Document Frequency
 - Speaker Frequency / Inverse Shot Frequency
- $d_{i\,i}^{\scriptscriptstyle{\mathrm{SD}}}$: cosine distance between TF-IDF vector

Automatic Speech Recognition



- Lemmatization (tree-tagger)
- $d_{ij}^{ ext{ASR}}$: cosine distance between TF-IDF vector

Monomodal GSTG Results

Corpus

- First eight episodes of Ally McBeal TV shows
- 5 hours of videos, 5564 shots and 306 scenes

Evaluation

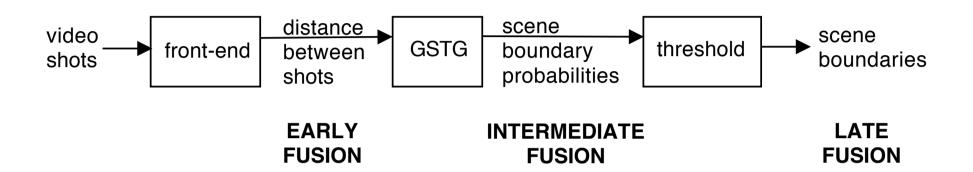
Leave-one-episode-out cross validation

	Precision	Recall	F-Measure	# Scenes
HSV (STG)	0.256	0.533	0.449	461
HSV	0.447	0.566	0.487	403
SD	0.157	0.562	0.240	1136
ASR	0.105	0.572	0.175	1751

Monomodal GSTG approaches

- Limitation:
 - One modality cannot solve the problem on its own
- Proposition:
 - Multimodal fusion

Multimodal Fusion



- Late fusion
 - intersection \cap or union \cup
- Early fusion

$$-d_{ij} = w_{\text{HSV}} \cdot d_{ij}^{\text{HSV}} + w_{\text{SD}} \cdot d_{ij}^{\text{SD}} + w_{\text{ASR}} \cdot d_{ij}^{\text{ASR}}$$

Intermediate fusion

$$= p = w_{\text{HSV}} \cdot p_{\text{HSV}} + w_{\text{SD}} \cdot p_{\text{SD}} + w_{\text{ASR}} \cdot p_{\text{ASR}}$$

Multimodal Results

Fusion	Precision	Recall	F-Measure
HSV (baseline)	0.447	0.566	0.487
$HSV \cap SD$	0.598	0.357	0.438
$\mathrm{HSV} \cap \mathrm{SD} \cap \mathrm{ASR}$	0.606	0.242	0.341
$\mathrm{HSV} \cup \mathrm{SD}$	0.180	0.770	0.288
$HSV \cup SD \cup ASR$	0.121	0.851	0.210
d(HSV) + d(SD)	0.445	0.599	0.499
d(HSV) + d(SD) + d(ASR)	0.445	0.599	0.499
p(HSV) + p(SD)	0.484	0.555	0.510
p(HSV) + p(SD) + p(ASR)	0.488	0.622	$\boldsymbol{0.539}$

Conclusion & future work

Graph-based approach to segmentation

- Using more modalities is better
 - Better use of ASR output
 - Add visual semantic concept detection

What's next?

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