



Color video segmentation by dissimilarity based on edges

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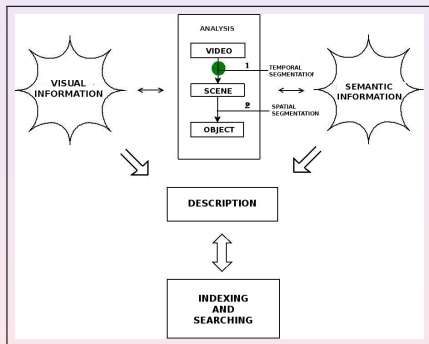
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 - Motivation
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 - General scheme
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Introduction > Motivation

Content-based multimedia retrieval



- Manual processes
 - Time-consuming
 - There are no uniform criteria

Solution

Automate

Introduction > Objectives

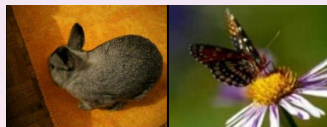
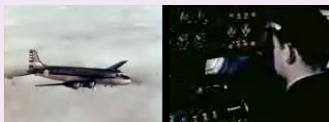
- Study of dissimilarity measures based on edges
 - Space domain
 - Frequency domain
 - Improved extensions to color-space
- Experimentation
 - Experiments design
 - Results

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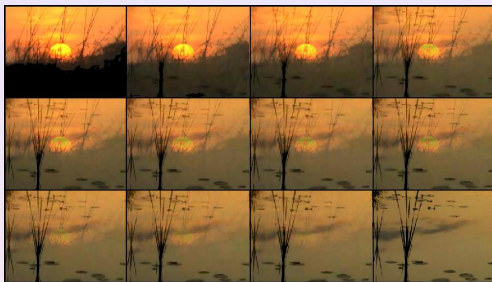
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Shot change types

- Sudden, occurring an abrupt change between two consecutive frames
 - Cut

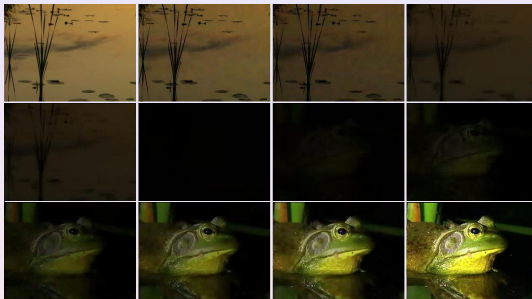


- Gradual, occurring slowly over successive frames



Dissolve

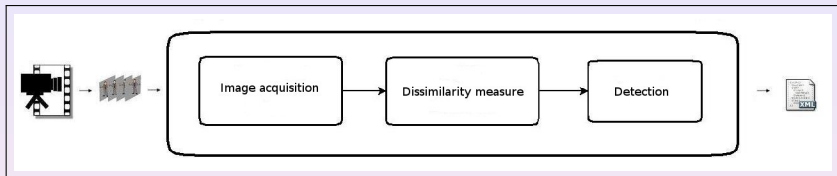
Frames of different scenes are superimposed, fading of the previous scene to defining the extent of the new.



Fade

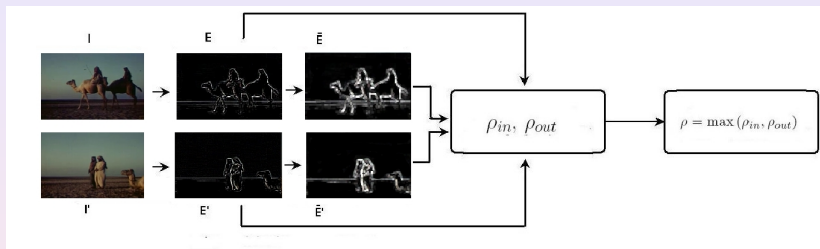
Special case of dissolve where a monochrome frame is intercalated between consecutive scenes.

General scheme of temporal video segmentation



- Acquire successive frames composing the video sequence
 - Grayscale representation
 - Color representation
- Obtain edge-based distance
 - Dissimilarity on space domain
 - Dissimilarity on frequency domain
- Decide the existence of shot changes
 - Fixed threshold
 - Adaptive threshold

Dissimilarity measures based on edges > Space domain



$$\rho_{out} = 1 - \frac{\sum_{x,y} E[x,y] \bar{E}'[x,y]}{\sum_{x,y} E[x,y]}$$

$$\rho_{in} = 1 - \frac{\sum_{x,y} \bar{E}[x,y] E'[x,y]}{\sum_{x,y} E[x,y]}$$

Proportion of exiting pixels

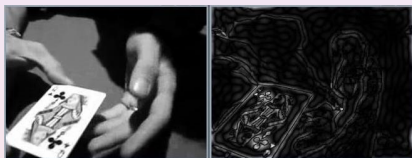
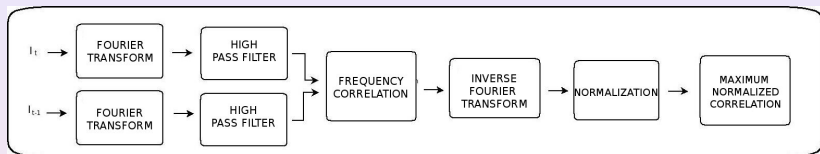
Elevated for a cut, a fade-out or in the first half of a dissolve

Proportion of entering pixels

Elevated for a cut, a fade-in or in the second half of a dissolve

Scene breaks are detected by looking for peaks in ρ

Dissimilarity measures based on edges > Frequency domain



Ideal high-pass filter

Enhance the contribution of the regions that exhibit the characteristics of changes

Normalized correlation on frequency domain

Measuring the similarity between two consecutive frames. The peak represents the best match

$$\rho(\xi) = \frac{TF^{-1} \{ \hat{x}_1(\omega) \hat{x}_2^*(\omega) \}}{\sqrt{\int |\hat{x}_1(\omega)|^2 d\omega \cdot \int |\hat{x}_2(\omega)|^2 d\omega}}$$

Improved extensions > Color dissimilarity on space domain

Objective

Get edges due to color variations



Adapt edges extractor

Canny

- Smoothing
- Derivation
- Non-maximum suppression
- Hysteresis

Improved extensions > Color dissimilarity on space domain

RGB space

Not capture all possible color differences

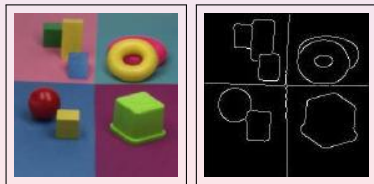
Ratios of Gevers

For calculating perceptual color differences independently of the luminosity

$$m_i \left(\vec{c}_1^{k_1}, \vec{c}_1^{k_2}, \vec{c}_2^{k_1}, \vec{c}_2^{k_2} \right) = \frac{c_1^{k_1} c_2^{k_2}}{c_1^{k_2} c_2^{k_1}}$$

Derivation

- ★ Finite difference between the red-green, red-blue and green-blue channels



Improvements > Color dissimilarity on frequency domain

Objective

Correlation based on color variations



Adapting high-pass filter

Fourier Transform works with a complex representation

Color space \Rightarrow Several dimensions

- RGB
 - ★ Get 3 complex transformations and combine processed
 - ★ Luminosity can not be separated

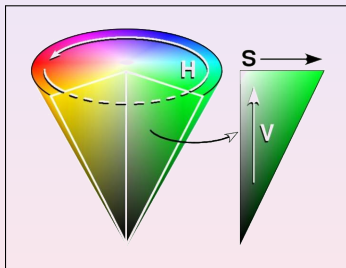
Improvements > Color dissimilarity on frequency domain

HSV space

- Complex representation of the hue-saturation space

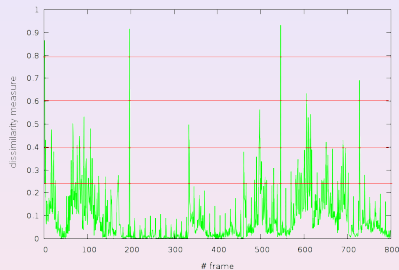
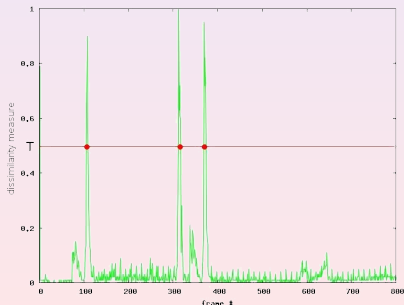
$$b(x, y) = S(x, y) \cdot e^{iH(x, y)}$$

- The lightness component is discarded



Improvements > Adaptative threshold

- Fixed threshold
 - It is based on experimental results
 - Remains fixed throughout the sequence

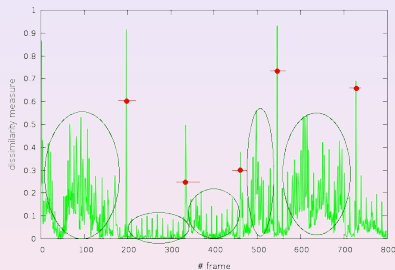


Problem

Difficult to establish the optimal threshold

Improvements > Adaptative threshold

- Adaptative threshold
 - It is based on the information contained in the video.
 - Fits over the sequence as the variability of the scene.



$$T_I = \mu + \alpha\sigma$$

It increases the rate of true positives without increasing false detections

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Dataset

- TRECVID 2002

| Total frames | Cuts | Fades | Dissolves | Others | Total |
|--------------|------|-------|-----------|--------|-------|
| 345306 | 1078 | 40 | 211 | 9 | 1338 |

- Heterogeneous in terms of cuts types
- Varied content
- Common cases of error in segmentation
 - Fast-moving objects
 - Camera operations
 - Changes in lighting
 - Gradual changes of different durations

Evaluation critery

- Quality detecion

$$Recall = \frac{T.P.}{T.P. + F.N.}$$

$$Precision = \frac{T.P.}{T.P. + F.P.}$$

Recall

A measure of the ability to detect all relevant items

Precision

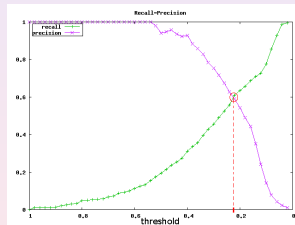
A measure of the ability to detect only relevant items

Experiments > Tuning parameter

- Representative subset of the data
- Select the parameter to optimize recall and precision

| Space domain | Canny | | | T | α | r |
|--------------|-------|-------|----------|------|----------|-----|
| | T_l | T_h | σ | | | |
| Grayscale | 20 | 60 | 5 | 0.15 | 4.6 | 1 |
| Color ratios | 15 | 50 | 5 | 0.20 | 5 | 1 |

| Frequency domain | T | α |
|------------------|------|----------|
| Grayscale | 0.43 | 4 |
| HSV | 0.28 | 4 |



Experiments > Results

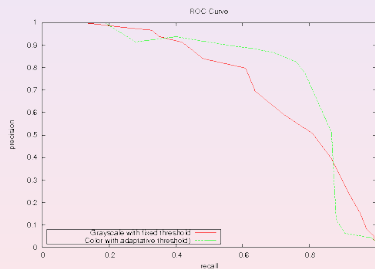
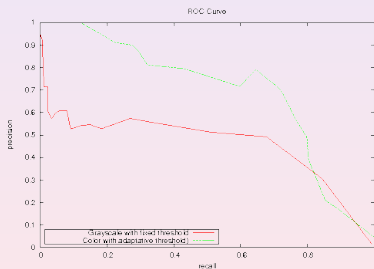
- Complete dataset
- Evaluation of each measure

Each cut type is considered separately

| | | Threshold | Recall | | | Recall | Precision |
|--------------------|--------------|------------|---------|-----------|----------|---------|-----------|
| | | | Cuts | Dissolves | Fades | | |
| Edges | Grayscale | Fixed | 52.75 % | 14.73 % | 71.50 % | 51.49 % | 41.91 % |
| | | Adaptative | 58.21 % | 28.04 % | 77.50 % | 59.01 % | 52.11 % |
| | Color ratios | Fixed | 71.07 % | 21.64 % | 93.54 % | 66.99 % | 57.29 % |
| | | Adaptative | 78.43 % | 41.18 % | 100.00 % | 70.38 % | 68.50 % |
| Correlation | Grayscale | Fixed | 74.22 % | 15.33 % | 86.91 % | 63.24 % | 65.75 % |
| | | Adaptative | 78.90 % | 30.02 % | 50.00 % | 64.17 % | 66.10 % |
| | HSV | Fixed | 82.80 % | 42.81 % | 94.29 % | 76.22 % | 78.70 % |
| | | Adaptative | 92.41 % | 50.00 % | 66.98 % | 82.60 % | 81.90 % |

Experiments > ROC Curves

- Comparison of different variants



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Conclusions

- Study of segmentation techniques
 - Influence of the frame representation
 - Influence of the detection threshold
 - Sensitivity to each cut type
 - Common error situations
- Contributions
 - Color extensions
 - ★ Gevers ratios on space domain
 - ★ HSV space on frequency domain
 - Adaptive threshold



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