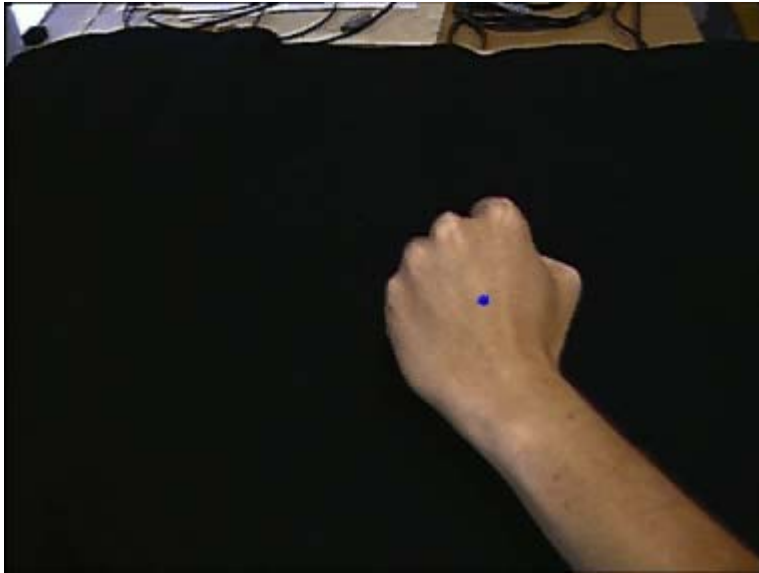


Segmentation in video

Advanced Image Processing
VGIS8 – Spring 2009

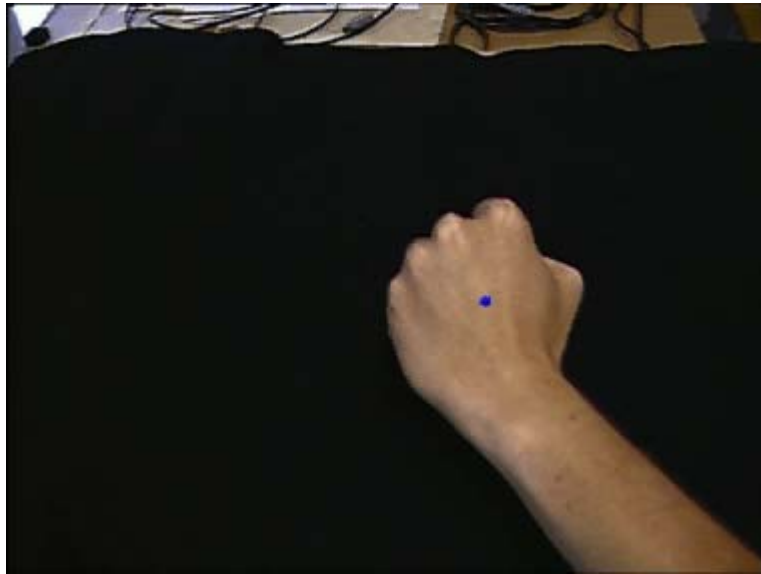
Segmentation in video

Basic idea: detect objects of interest



Segmentation in video

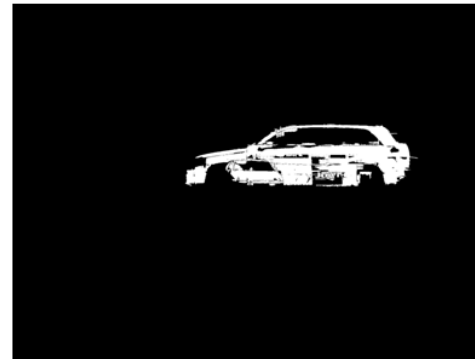
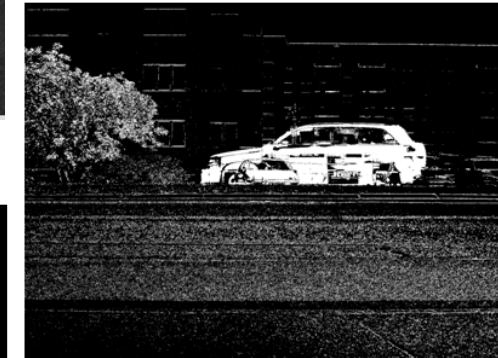
Basic idea: detect objects of interest



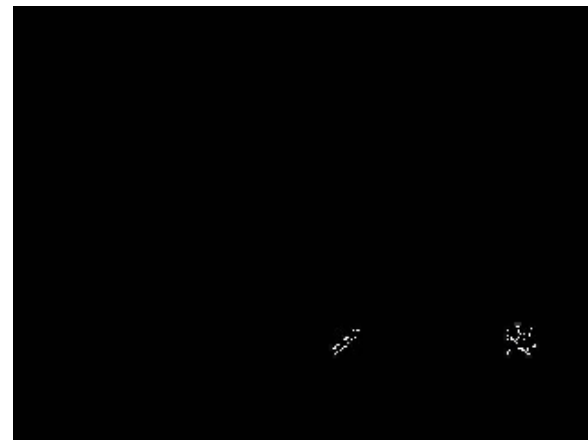
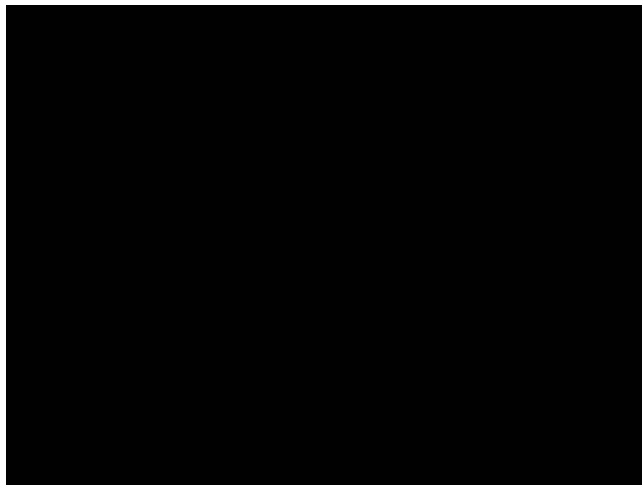
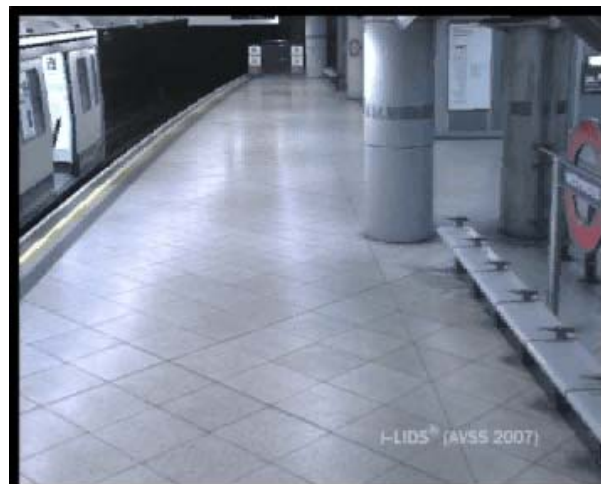
Result: binary image

General algorithm

1. Save reference image
2. Capture current image
3. Perform image subtraction
4. Thresholding
5. Filter noise



Results - General algorithm



Today's topics

- Background subtraction
- Constructing the reference image
- Updating the reference image
- Global and local thresholds
- Image differencing
- Noise filtering
- Advanced segmentation

1. Acquire reference image
2. Capture current image
3. Perform image subtraction
4. Thresholding
5. Filter noise

Background subtraction

- First frame as reference image

1. Acquire reference image
2. Capture current image
3. Perform image subtraction
4. Thresholding
5. Filter noise

$r(x,y)$:



$f(x,y)$:



- Pixel-wise subtraction

$$g(x,y) = f(x,y) - r(x,y)$$

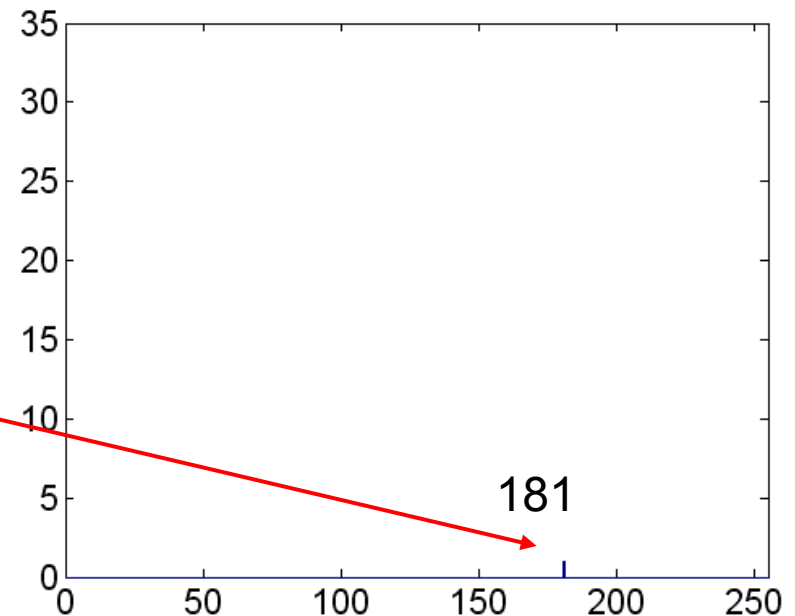
- Apply absolute value

$$g(x,y) = |f(x,y) - r(x,y)|$$



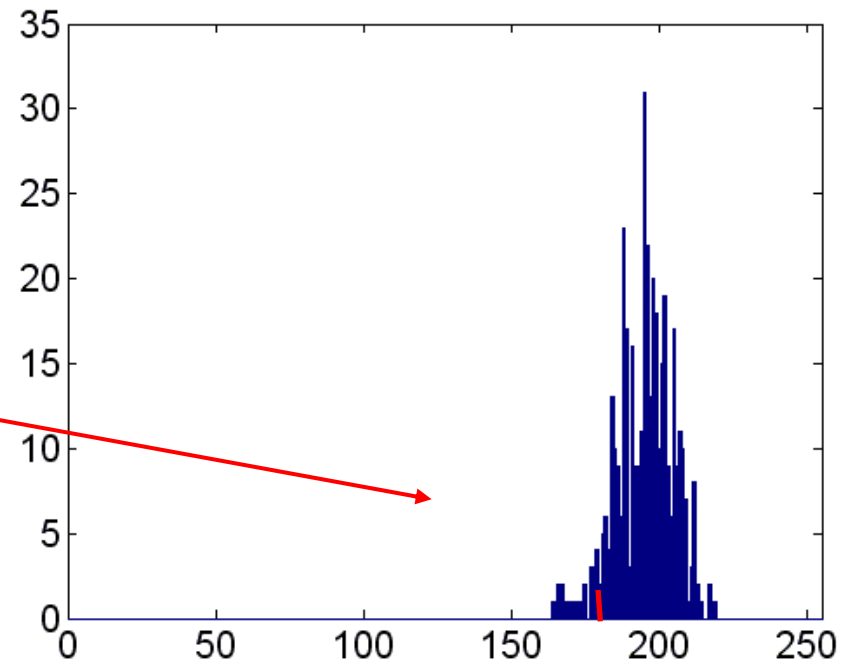
Constructing the reference image

- With non-stable background
 - e.g. camera noise, changes in light, small motion from wind gusts
- Then learn the background



Constructing the reference image

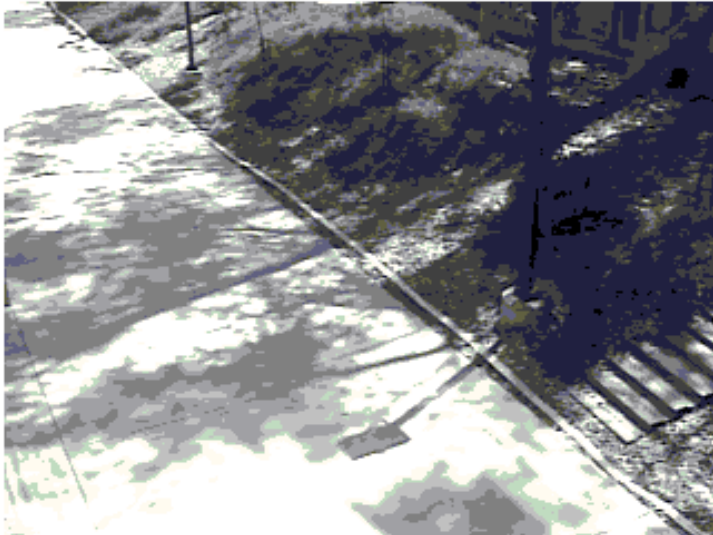
- With non-stable background
 - e.g. camera noise, changes in light, small motion from wind gusts
- Then learn the background



Constructing the reference image

- Learning the background
 - Capture N empty images and calculate the average background image

$$r(x,y) = \sum_{i=1:N} f_i(x,y) / N$$



1. Calculate average reference image
2. Capture current image
3. Perform image subtraction
4. Thresholding
5. Filter noise

Updating the reference image

- With gradual changes during processing
 - e.g. light changes during a day
- Then update the reference image



Updating the reference image

- With gradual changes during processing
 - e.g. light changes during a day
- Then update the reference image
 - Update speed α

$$r_{new}(x,y) = \alpha \cdot r_{old}(x,y) + (1 - \alpha) \cdot f(x,y)$$



Update speed
 $\alpha = 0.98$



Global vs. local thresholds

1. Acquire reference image
2. Capture current image
3. Perform image subtraction
4. Thresholding
5. Filter noise

- So far:

- Image subtraction

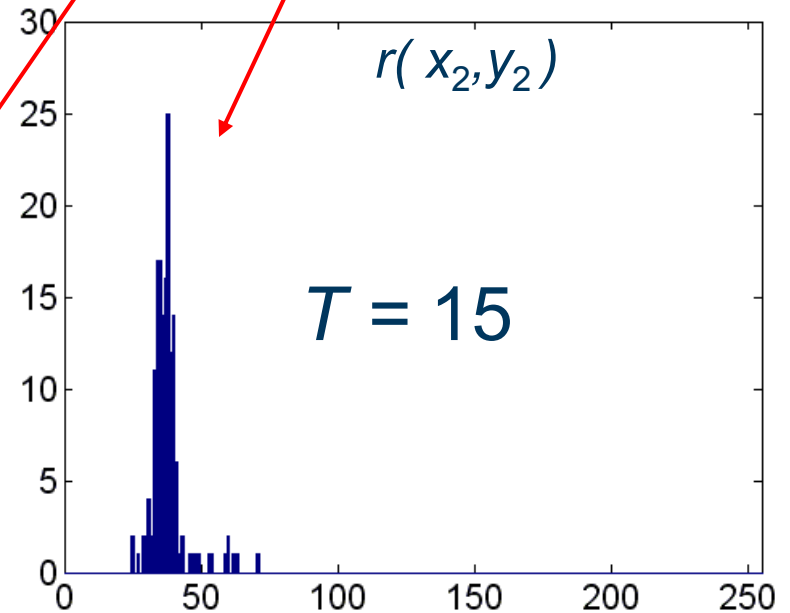
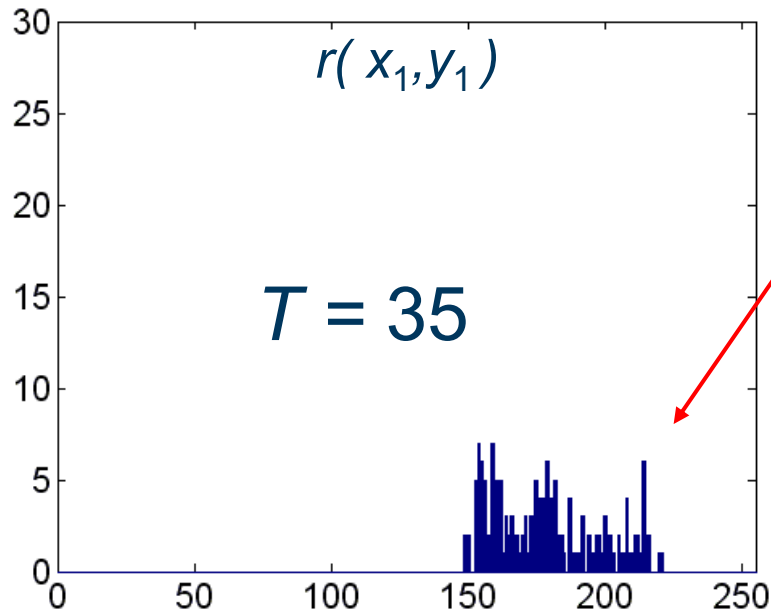
$$g(x,y) = |f(x,y) - r(x,y)|$$

- Global threshold T to obtain binary image b

$$b(x,y) = \begin{cases} 0, & \text{if } g(x,y) < T \\ 255, & \text{else} \end{cases}$$

Global vs. local thresholds

- With local pixel variation
- Then apply local thresholds

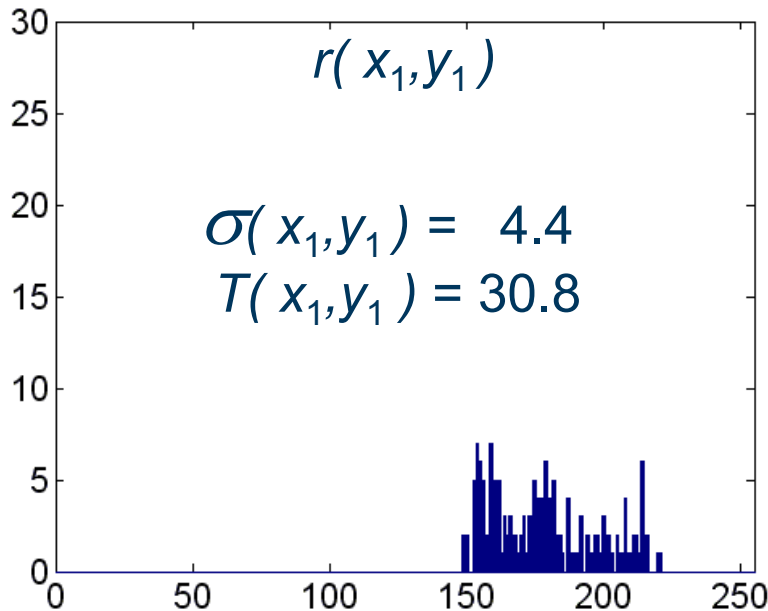


Global vs. local thresholds

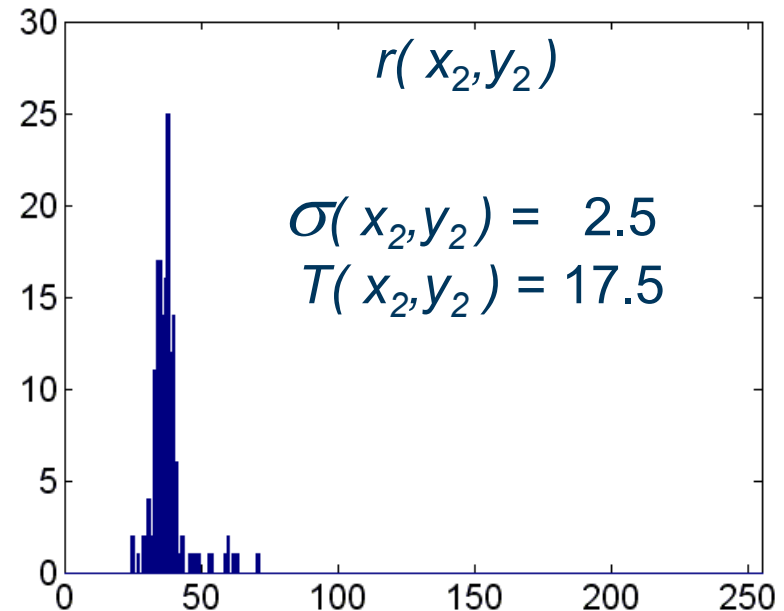
- Apply local threshold
 - Calculate mean and standard variation σ
 - Define T in terms of σ

$$T(x,y) = \beta \cdot \sigma(x,y)$$

- Choose (global) β



$$\beta = 7$$



Global vs. local thresholds

- Apply local threshold

1. Calculate mean and standard variation for each pixel
2. Capture current image
3. Perform image subtraction
4. Threshold according to standard variation
5. Filter noise

- Local thresholds + reference image updating is a powerful background subtraction method – especially if you can choose your background

Image differencing

- With very dynamic scenes which can not be learned
- Then choose the previous frame as the reference frame



Image differencing

- With very dynamic scenes which can not be learned
- Then choose the previous frame as the reference frame

$$r_t(x, y) = f_{(t-1)}(x, y)$$

- Different reference image (depending on application)

$$r_t(x, y) = f_{(t-k)}(x, y), k = 2, 3, \dots$$

- Double differencing

$$g_t(x, y) = [f_t(x, y) - r_{1,t}(x, y)] \text{ AND } [f_t(x, y) - r_{2,t}(x, y)]$$



Image differencing vs. background subtraction

Image differencing	Background subtraction
<ul style="list-style-type: none">• Handles dynamic scenes<ul style="list-style-type: none">• Detects only motion• Detection results in ghosts	<ul style="list-style-type: none">• Requires static background• Detects stationary objects• Detects whole objects

Ghosts:

$f(x,y)$

$t=0$



$t=0$



$t=0$



$r(x,y)$

$t= -1$



$t= -3$



$t= -7$



$b(x,y)$

$t= -1$



$t= -3$

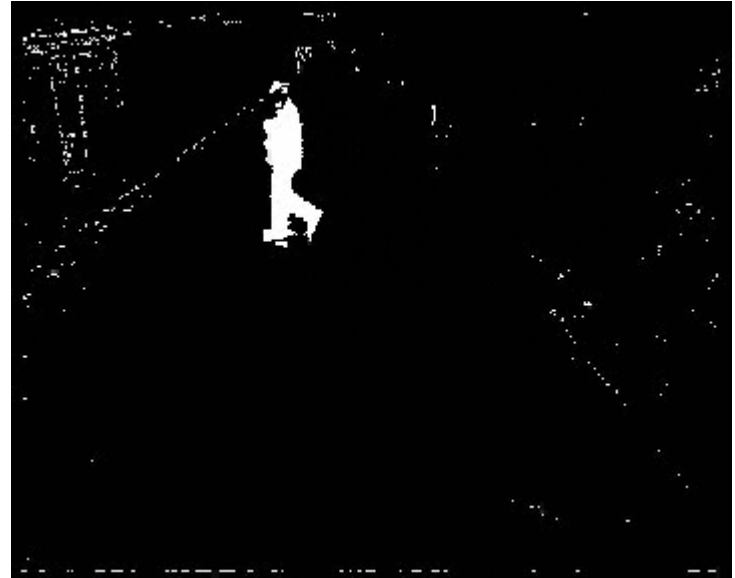


$t= -7$



Noise filtering

- Single pixels are likely to appear
 - Pixel-noise
- Apply Median filter
 - Depending on filter size, bigger spots can be erased
- Alternatives:
 - Morphology
 - Temporal filtering
 - Restrictions on size, shape



Summing up...

- The general algorithm for segmentation:

1. Acquire reference image
2. Capture current image
3. Perform image subtraction
4. Thresholding
5. Filter noise

- Learn reference image from a set a frames
- Update the reference image continuously
- Learn the standard variation and apply local thresholds
- Apply image differencing in dynamic scenes

Next topics – advanced segmentation

- Shadows and light changes
- Multi-modal background models
- Learning background with non-empty images
- Layered background models
- Optical flow for segmentation

Shadows and light changes

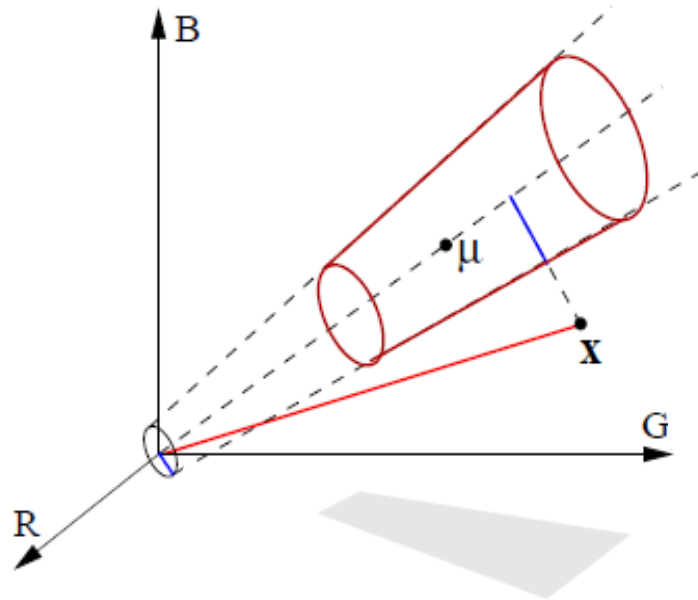
- Shadows and light changes cause false positives



- Solution: separate color and intensity
 - HSV-color space
 - Codebook background subtraction method
 - One method to handle advanced issues in segmentation

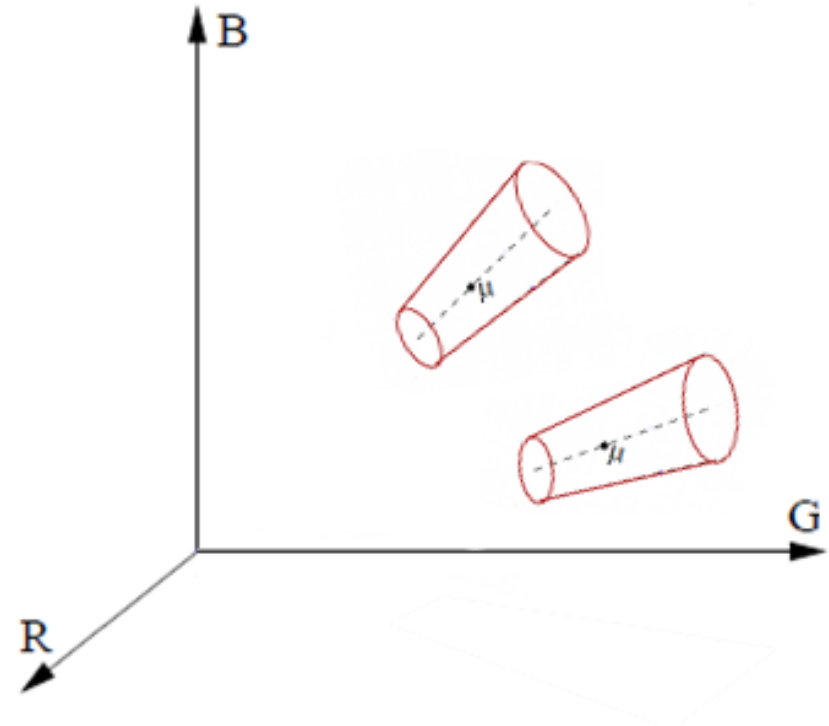
Shadows and light changes

- Color-model of codebook background subtraction:



Multi-modal background models

- Learn the background from N frames – $N \sim 300$
- Allow multiple codewords for each pixel



Learning background with non-empty images

- Learn the background from N frames – $N \sim 1000$



- Evaluate all codewords after training
 - True background vs. false background
 - True: pixel values appear periodically
 - False: pixel values appear in bursts

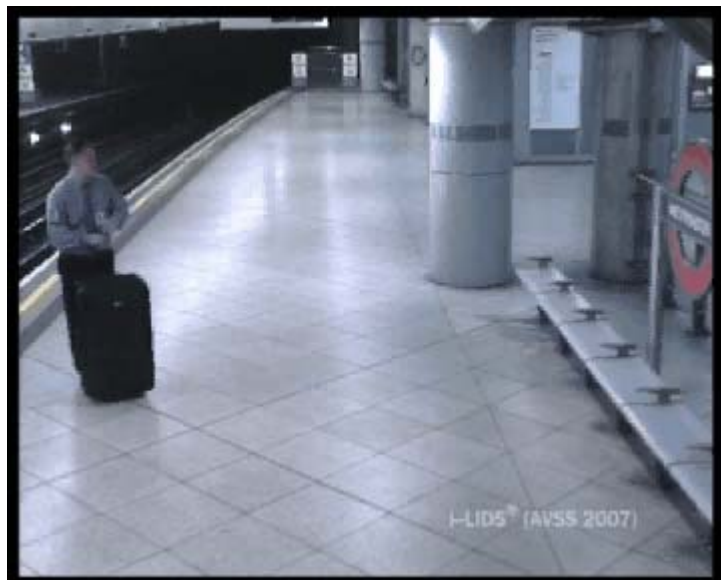
Layered background models

- Parts of the scene can become temporary background
 - e.g. parked cars, left behind luggage, the train
- Reference image updating adapts over time



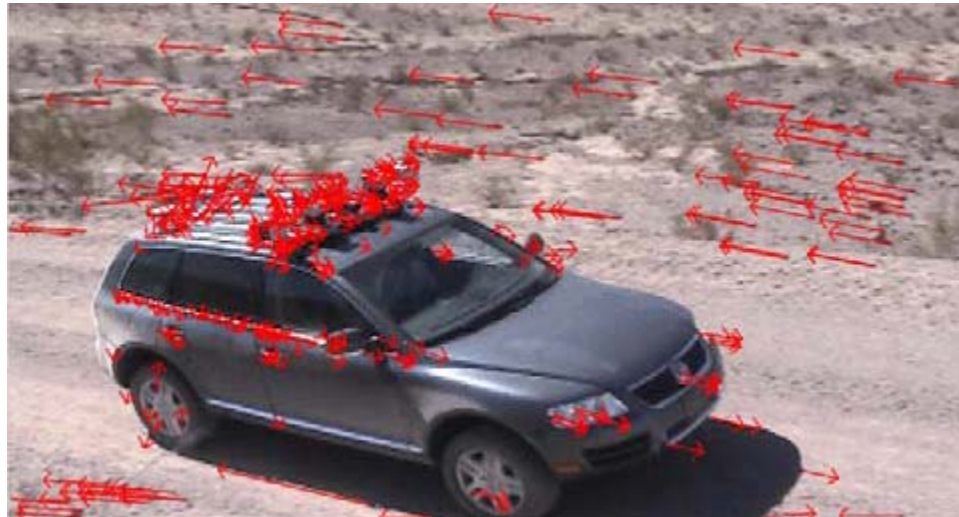
Layered background models

- Alternative: use multiple layers of background
 - Continuously evaluate all foreground regions
 - IF pixel values appear periodically for N frames
 - THEN add new codeword to the background model
 - Delete expired codewords



Optical flow for segmentation

- Moving cameras make reference images impossible (e.g. car-mounted cameras)
 - Both objects and background move
- Solution:
 - Estimate motion direction between frames
 - Objects with motion different than average motion can be identified



Summing up ...

- Advanced background subtraction addresses many real life problems
- ... but the problem of segmentation is not solved
- Still an area of research:
 - Combining color segmentation with edges
 - Applying regional information
 - Applying contextual information

Exercises

- Download Matlab script + videos

1. Acquire reference image
2. Capture current image
3. Perform image subtraction
4. Thresholding
5. Filter noise

- Segment hand.avi with different thresholds

- Implement:

- Learn reference image from a set a frames

- Use the first 100 frames in sun_and_trees.avi

- Update the reference image continuously

- Use the first 40 frames of office_light.avi for background

- Learn the standard variation and apply local thresholds

- Use the first 100 frames in sun_and_trees.avi

- Apply image differencing in dynamic scenes

- sun_and_trees.avi is a good test case

- Mini project

Mini-project

- To do (for each group)
 - Find information and learn the principle
 - Implement a working prototype
 - Hand-in report (3-5 pages) about: principle, usage, algorithm, your implementation and results
 - Present your work (10-20min)
- Hand-in report: 9/4 at 12.00
- Presentation: 13/4