

# Using Histograms to Detect and Track Objects in Color Video

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## The problem:

Static camera, but the background is not entirely stable

### Outdoor Scenes

-leaves and trees moving

### Indoor Scenes

-varying lighting conditions

-cluttered backgrounds

## Applications

Surveillance and safety:

- ◆ Security cameras, children, pedestrians

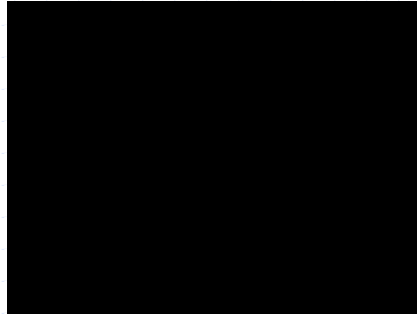


## Related work

Background subtraction on a pixel by pixel basis

- ◆ Very sensitive to any movements of the camera or the background scene
- ◆ Required to “learn” the background (statistics)
- ◆ Many different methods have been developed for background / foreground segmentation

## Results of previous work



## Motivations

Want to detect that something is happening in a video sequence without too many false positives.

Do not want to spend a lot of time “learning” the background (or doing post-processing)

Would like a fast and reliable method (accuracy is not paramount)

## Approach

- ◆ Compare color histograms computed for small regions of the “background” image and the current (new) image (reduced color/ 12 bit bit representation)
- ◆ Compare color edge histograms computed for small regions of the “background” image and the current image (36 bin quantization)

## Color Histograms

Reduced color representation =

$$C = (R/16) * 256 + (G/16)*16 + (B/16)$$

(This results in a 24 -> 12 bit color depth reduction)

This results in a 4096 bin histogram

- ◆ lowest 4 bits are less useful
- ◆ requires less storage
- ◆ faster implementation - easier to compare histograms

## Color Edge Histograms

For each frame:

- ◆ Use the Sobel edge detector to compute edges in each color band  $(r_x, r_y, g_x, g_y, b_x, b_y)$
- ◆ Combine the three color bands into the structure matrix,  $S$ , to compute the color edge response
- ◆ The edge strength is computed as the larger of the two eigenvalues of  $S$ , and the orientation is given by the corresponding eigenvector
- ◆ Histogram bin index is determined using edge orientation (36 bins total), and the bin count is incremented using the edge magnitude

## Color Edge Histograms (cont.)

$$S = \begin{pmatrix} r_x^2 + g_x^2 + b_x^2 & r_x r_y + g_x g_y + b_x b_y \\ r_x r_y + g_x g_y + b_x b_y & r_y^2 + g_y^2 + b_y^2 \end{pmatrix}$$

$S$  has two eigenvalues  $\lambda_1 \geq \lambda_2 \geq 0$

If  $\lambda_2 \approx 0$ , we have an edge

$\lambda_1$  is a squared edge magnitude and the edge direction is given by the corresponding eigenvector  $e_1$

## Histogram Matching

### ◆ Histogram Intersection

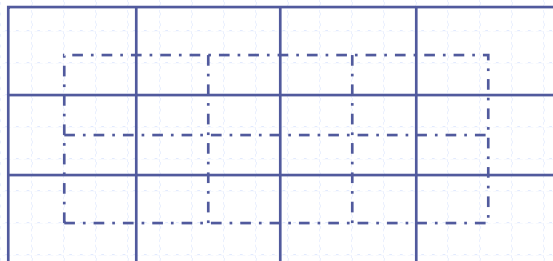
$$\frac{\sum_i \min\{h_c(i), h_b(i)\}}{\sum_i h_b(i)}$$

### ◆ Chi Squared Formula

$$X^2 = \sum_i 2 \frac{(h_c(i) - h_b(i))^2}{h_c(i) + h_b(i)}$$

## Overall control

- ◆ We divide each frame into  $40 \times 40$  pixel blocks
- ◆ To make sure that we do not miss objects on grid block boundaries we tile the frame by overlaying two grids, one of which is shifted by 20 pixels in  $x$  and  $y$  directions

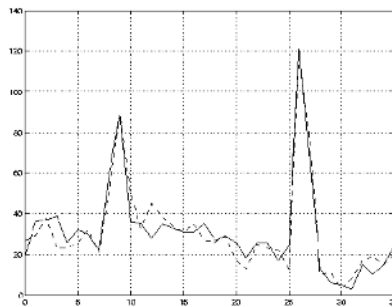


## Criteria for block activation

- ◆ On a block by block basis, similarity measures between background and foreground histograms are computed
- ◆ For histogram intersection: If the similarity is below a threshold,  $T$ , then the block contains a foreground object and is activated for display
- ◆ For chi squared: If the  $X^2$  measure is greater than a threshold,  $T$ , then the block contains a foreground object and is activated for display

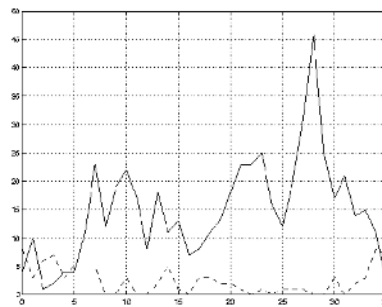
## Examples of edge histograms

similar histograms



**Similarity (inters.) = 92%**  
 **$X^2 = 61$**

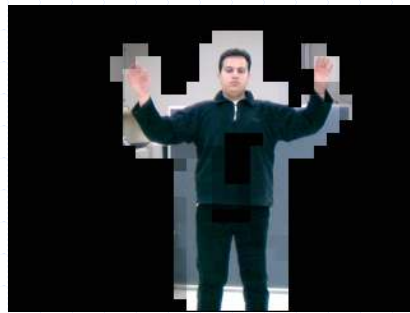
different histograms



**Similarity (inters.) = 22%**  
 **$X^2 = 828$**

## Results...

## Edge based detection

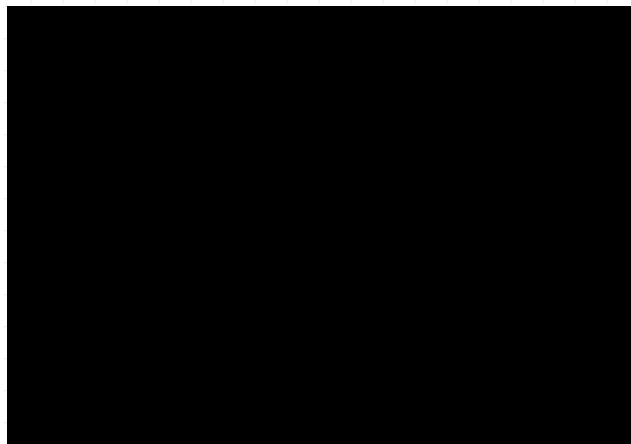




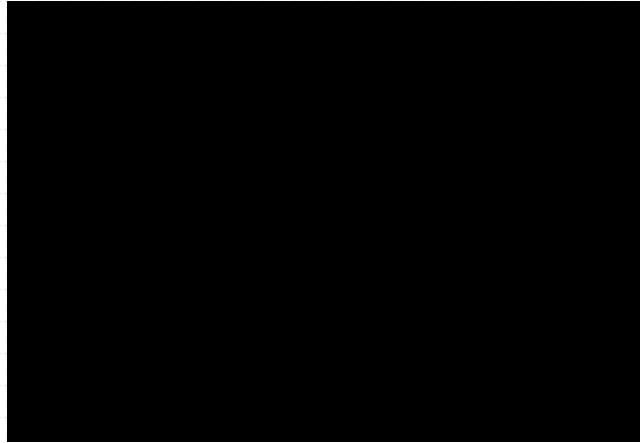
## Moving person in a cluttered scene



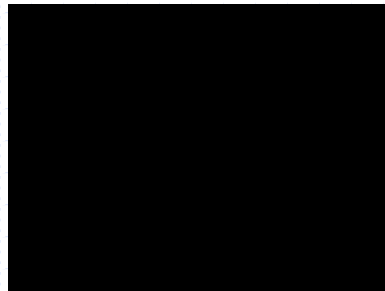
## Color histogram based detection



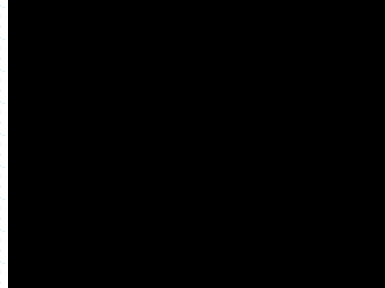
## Edge histogram based detection



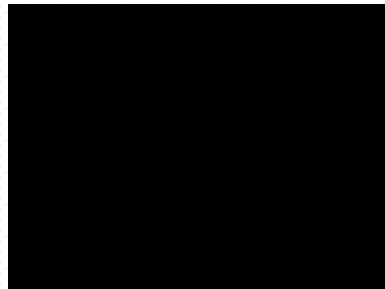
## Surveillance: dropping an object



## Surveillance: removing an object



## Surveillance: Interacting people



## Conclusions and future work

- ◆ Presented two histogram-based methods of detecting (tracking) objects in video
- ◆ The methods are fast and reliable and do not require prior training (learning of the scene)
- ◆ We are currently investigating:
  - Different histogram matching methods (unfolding histograms)
  - Tracking detected objects in video
  - Recognizing object classes from histograms of detected objects

Questions?