

CAS CS 585

Image and Video Computing

Lecture by Margrit Betke

Topics: Face Detection, Multi-level Approach,
Template Matching (SSD, NCC), Motion
Detection, Binary Image Analysis

Finding the Head and its Movement by Detecting Pixels with Skin Color



Computer Science



Finding the Head and its Movement by Detecting Pixels with Skin Color



Computer Science



(a) Input

(b) Color

*Common Trick in
Computer Vision:*

Use

“**image pyramid**” = input
image at difference scale.

Here: 6 levels, reduction
in x, y by $\frac{1}{4}$ (other
schemes possible)

Then process result
pyramid. Why?

Finding Movement by Detecting Pixels with Brightness Changes



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(a) Input



(c) Motion

- ❑ $DifferenceImage(x,y,t) = Image(x,y,t) - Image(x,y,t-1)$
- ❑ Or $DifferenceImage(x,y,t) = Image(x,y,t) - Image(x,y,t-k)$
- ❑ Or $DifferenceImage(x,y,t) = |Image(x,y,t) - Image(x,y,t-k)|$

Finding the Face and its Movement by Locating the Best Match of a Face Template



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(a) Input



(d) Correlation

Finding the Face and its Movement by Locating the Best Match of a Face Template



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Use Jingbin's face as a template



Visualization
of Match Values

Template Matching with Sum-Squared Difference (SSD)



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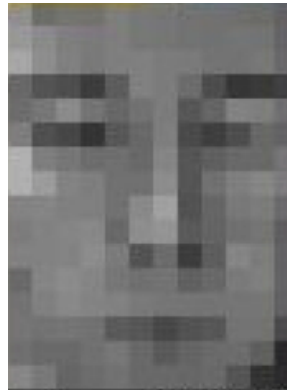
- ❑ *Scene subimage s , template image m*
- ❑ *same size images = n pixels*
- ❑ *s_i = i th pixel in subimage of scene*
- ❑ *m_i = i th pixel in template image m*
- ❑ $SSD = \sum_{(i=1 \text{ to } n)} (s_i - m_i)^2$
- ❑ *Template matching = exhaustive search algorithm for position of scene subimage that best matches the template (where SSD is smallest)*

Finding the Face and its Movement by Locating the Best Match of a Face Template



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Used average face as a template



Visualization
of Match Values

Template Matching via with Normalized Correlation



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- ❑ Scene subimage s , template *image* m

- ❑ *Normalized correlation coefficient*

$$r = 1/n \sum_i ((s_i - \text{mean}(s)) * (m_i - \text{mean}(m)) / (\sigma_s \sigma_m)) \text{ where}$$

s_i and m_i are respective brightness values of the i th pixel

$\text{mean}(m)$ and σ_m are mean and standard deviation of all pixels in the template

$\text{mean}(s)$ and σ_s are mean and standard deviation of all pixels in the subimage of the scene

- ❑ Template matching = exhaustive search for position of subimage that produces highest r

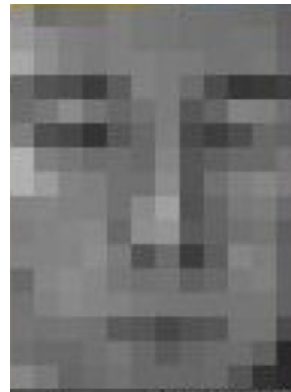
- ❑ r can be between -1 and 1

Finding the Face and its Movement by Locating the Best Match of a Face Template



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Used average face as a template



Visualization
of Normalized
Correlation
Coefficient
Match Values

1D Discrete Convolution



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$$\begin{aligned}(f * g)[n] &\stackrel{\text{def}}{=} \sum_{m=-\infty}^{\infty} f[m] g[n - m] \\ &= \sum_{m=-\infty}^{\infty} f[n - m] g[m].\end{aligned}$$

2D generalization:

f = input image, g = template image
(or CNN function)

2D Convolution Example



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1 _{x1}	1 _{x0}	1 _{x1}	0	0
0 _{x0}	1 _{x1}	1 _{x0}	1	0
0 _{x1}	0 _{x0}	1 _{x1}	1	1
0	0	1	1	0
0	1	1	0	0

Image

4		

Convolved
Feature

2D Convolution Example



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0	0	0	0	0	0	0
0	1	1	1	1	0	0
0	1	0	0	1	0	0
0	1	1	1	1	0	0
0	1	0	0	1	0	0
0	1	1	1	1	0	0
0	0	0	0	0	0	0

Input: black and white image

1	3	1
3	2	3
1	3	1

Weight
matrix

0	0x1	0x3	0x1	0	0	0
0	1x3	1x2	1x3	1	0	0
0	1x1	0x3	0x1	1	0	0
0	1	1	1	1	0	0
0	1	0	0	1	0	0
0	1	1	1	1	0	0
0	0	0	0	0	0	0

Applying
weight matrix

8	9			

Output: convolved feature

Convolution example



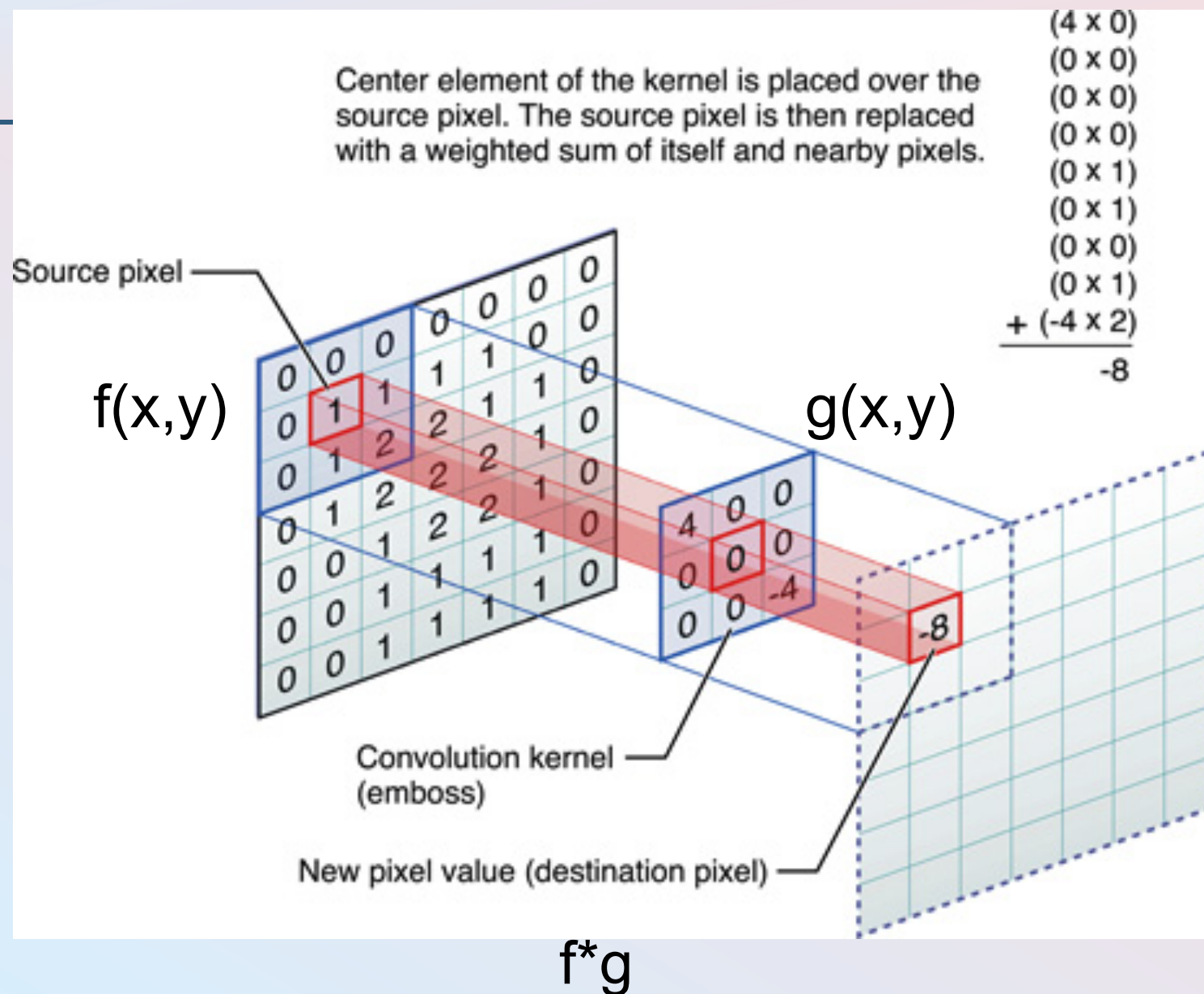
Computer Science

1 _{x1}	1 _{x0}	1 _{x1}	0	0
0 _{x0}	1 _{x1}	1 _{x0}	1	0
0 _{x1}	0 _{x0}	1 _{x1}	1	1
0	0	1	1	0
0	1	1	0	0

Image

4		

Convolved
Feature





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Why is Convolution Powerful?



Signal Processing:

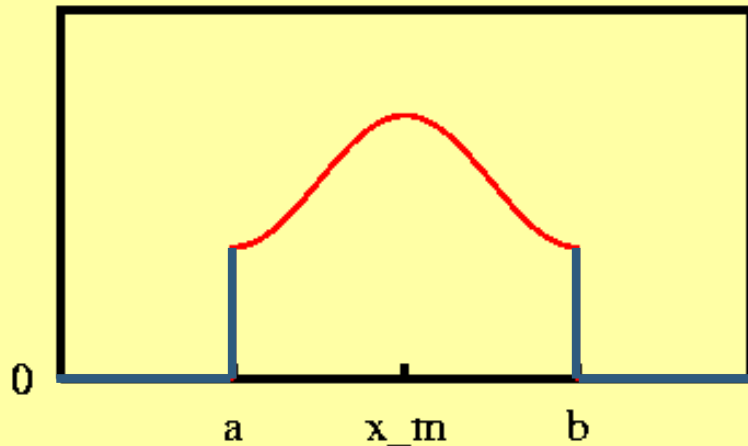
**Convolution is used to define a
“matched filter” for locating
“targets” in time signals**

(optimal if Gaussian noise)

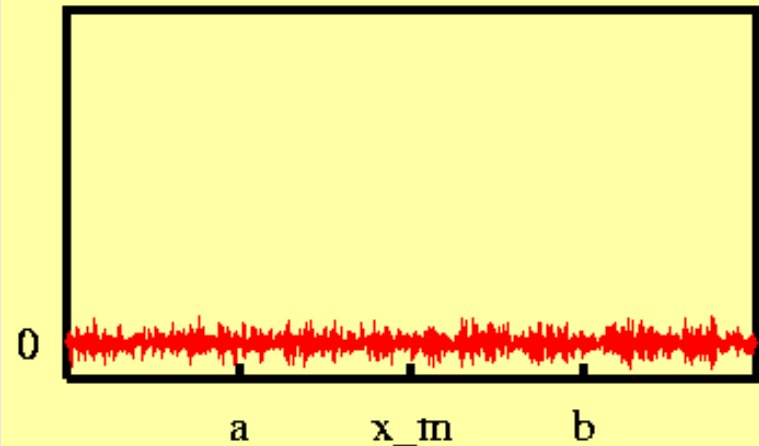
1D Position Estimation: $\sum \text{object} * \text{background}$



(a) Object

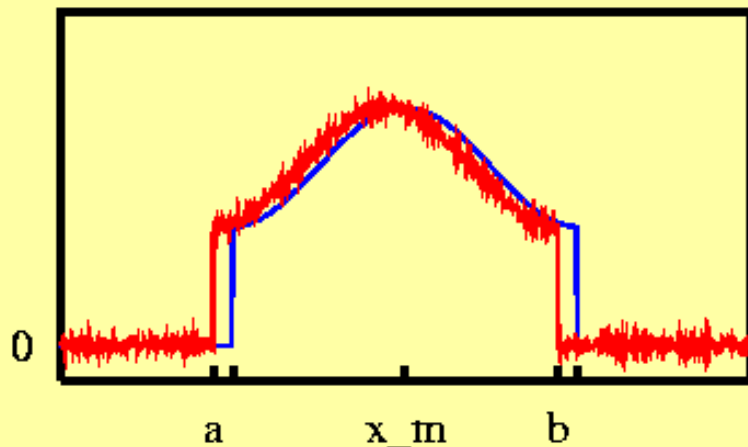


(b) Zero-mean Background

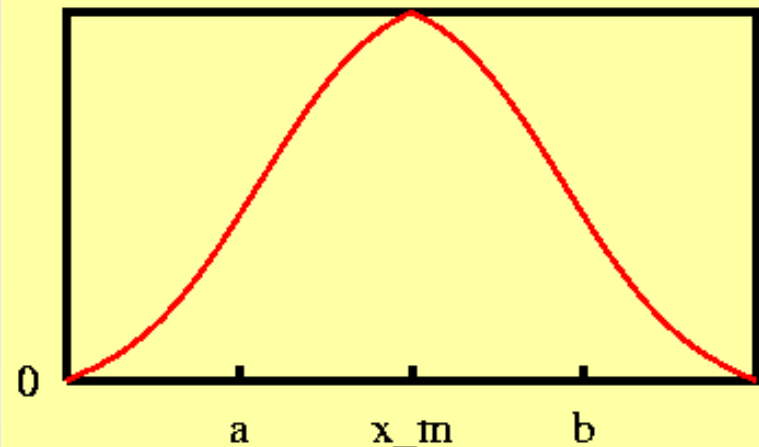


ance

(c) Object and Zero-mean Background



(d) Classical Matched Filter Output

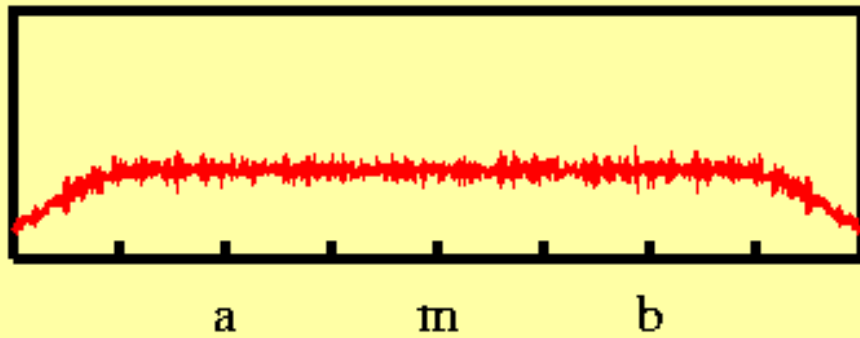


Another 1D convolution example:

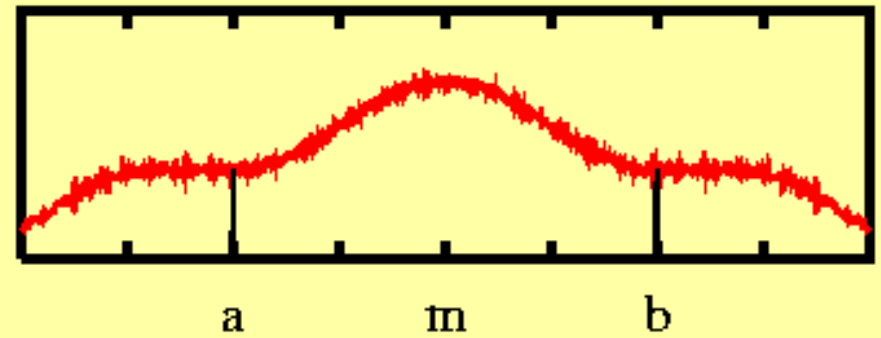


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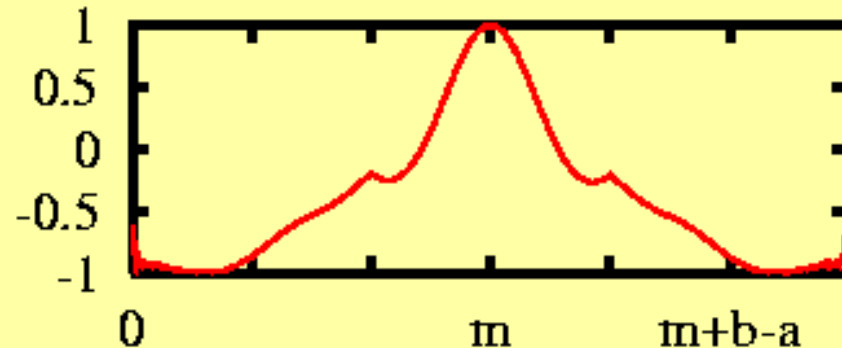
Nonzero-mean Background



Scene with Object



Norm. Correlation Coefficient



= convolution/st

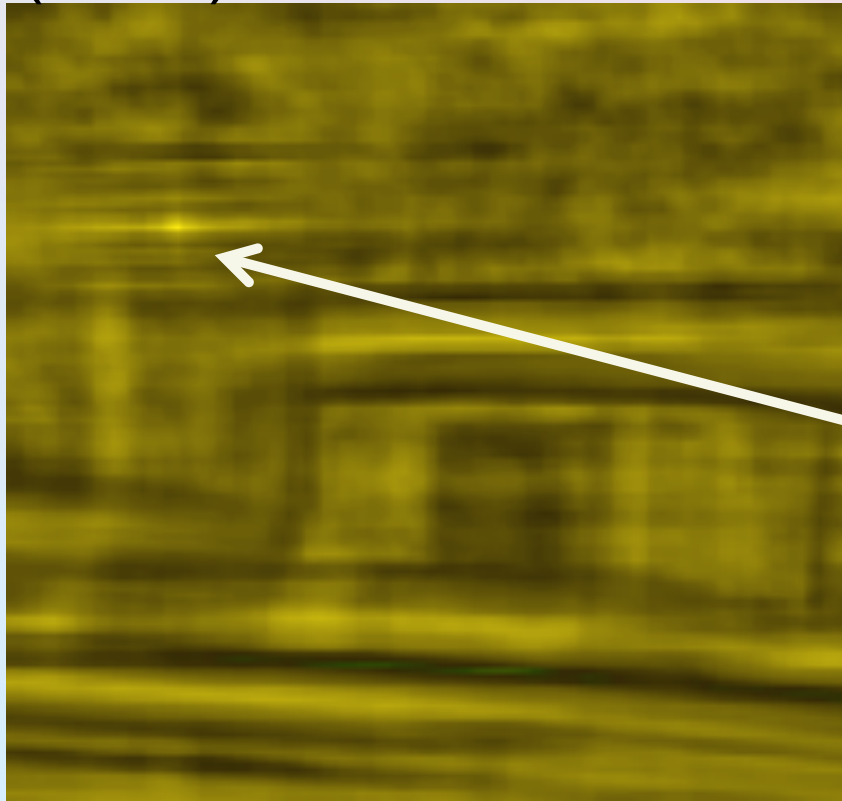
2D Position Estimation

Convolution of one-way sign with itself



2 D Position Estimation

Convolution of one-way sign with scene
(NCC)



Peak in
performance surface
(= negative loss fct)
at correct location

2 D Position Estimation

Convolution of one-way sign with scene (NCC)



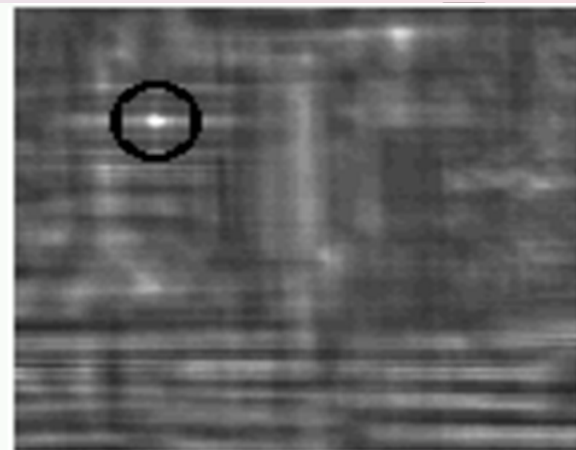
This performance surface is computed for **correct** size of one-way sign

Different surfaces for different sizes of object

Sample Performance Surfaces



complexity: 250
size: 73×27
max. cor. coef. 0.82
correct match



complexity: 33
size: 73×27
max. cor. coef. 0.64
incorrect match



(shown enlarged)
complexity: 25
size: 21×5
max. cor. coef. 0.70
incorrect match



Multi-Resolution Matching



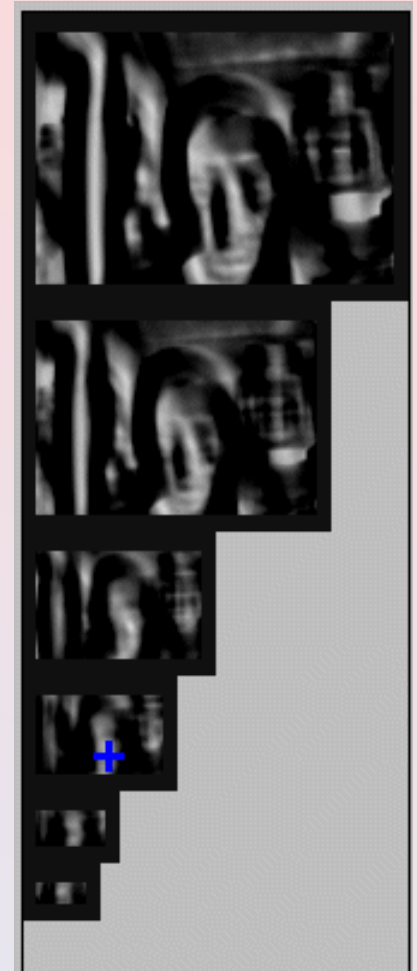
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Normalized correlation coefficient over multi-resolution search space:

$$r = \frac{1/n \sum_i (s_i - \text{mean}(s)) (m_i - \text{mean}(m))}{(\sigma_s \sigma_m)}$$



← Template
matched over all
resolutions →



Finding the Face and its Movement by Locating the Best Match of a Face Template



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(a) Input

You can apply template matching to a small version of your input image and use that search result to start searching for a match in the 2nd smallest images. Repeat until the original size is processed.



(d) Correlation

Multi-scale Pyramids



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Combine the results of color and motion detection to mask the regions of interest for correlation-based template matching



(a) Input

(b) Color

(c) Motion
(low-pass filter)

(d) Correlation

(e) Masked
Correlation

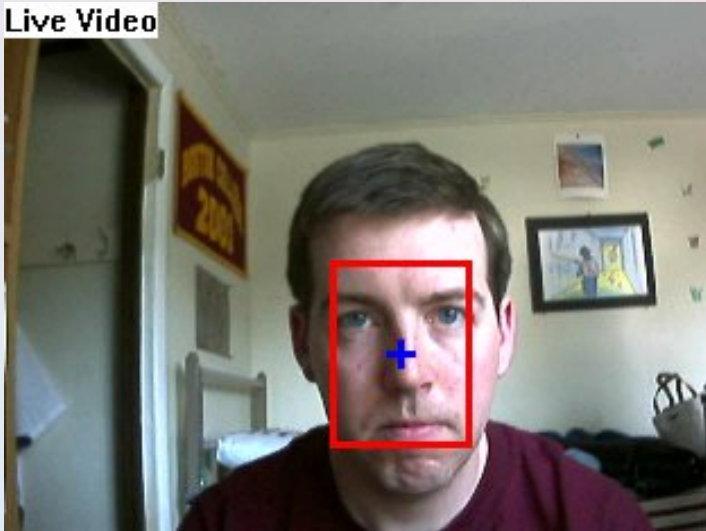
Face Detection



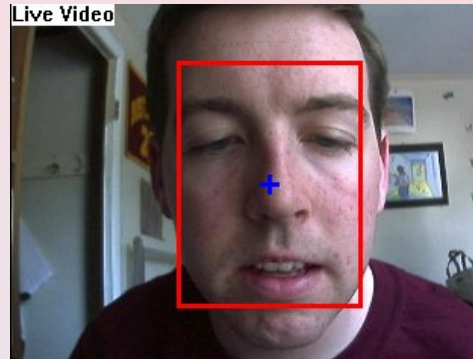
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Data Variability

Live Video

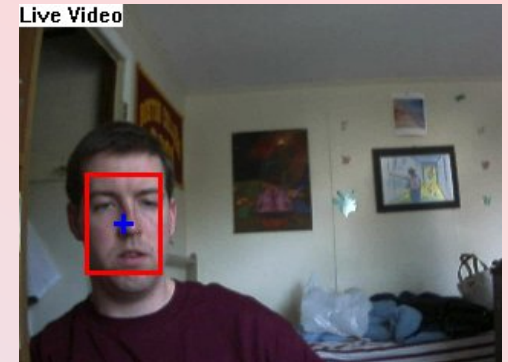


Live Video



Large Face

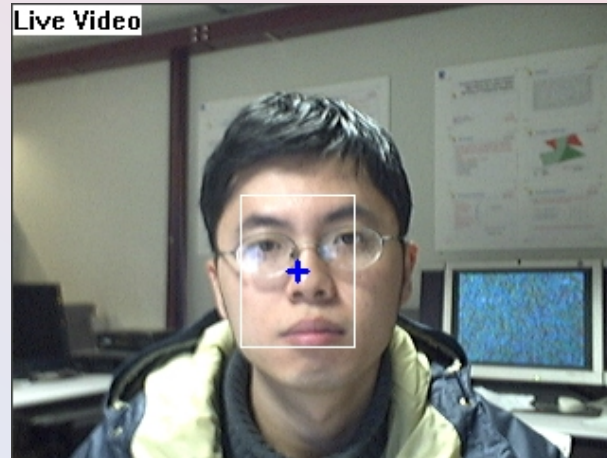
Live Video



Small Face

Shadows
Cluttered background

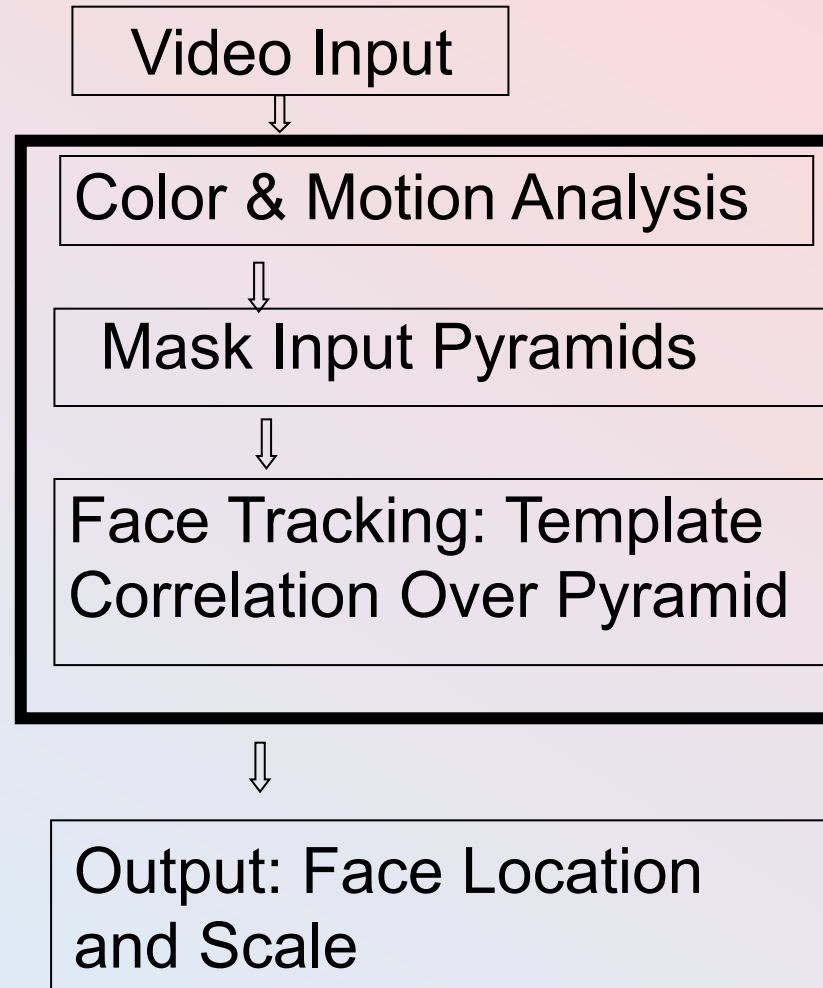
Live Video



Algorithm: Multi-scale Face Detection



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Face Detection Interface



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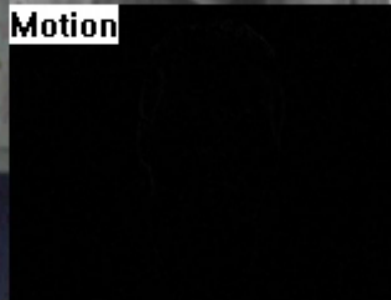
Live Video



B&W Video



Motion



Color



Correlation



Max Score: 193; Scale: 6; Location: (160, 120)

OK

Cancel

☒ Pyramid Display



Clo

Finding Movement in Video by Comparing to a Somewhat Static Background



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A) Establish background model:

Easy -- use first frame:

$$\text{BackgroundImage}(x,y,t) = \text{Image}(x,y,0)$$

Better: Dynamic Gaussian models of brightness changes. Filter

$$\text{if } k \sigma(x,y,t) < | \text{Image}(x,y,t) - \mu(x,y,t) |$$

for $k = 5\%$; mean μ and std. dev. σ updated in some time window

B) Compute difference to background:

$$\text{DifferenceImage}(x,y,t) = \text{Image}(x,y,t) - \text{BackgroundImage}(x,y,t)$$

Deal with negative values, e.g., use absolute difference or remap the range of values to 0-255.

C) Interpret difference image

Geometric Properties of Detected Object



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❑ Binary Image $B(x, y) = 1$ if $I(x, y) > \tau$,
 $= 0$ otherwise

❑ Size, area: $A = \sum_{x=0}^{x=\text{dim}-1} \sum_{y=0}^{y=\text{dim}-1} B(x, y)$

$A = M_{00}$ “0th moment”

❑ Position: use 1st moment

❑ Orientation: use 2nd moments