# CAS CS 585 Image and Video Computing

Lecture by Margrit Betke

Topics: Image and Video Formats, Color, Projections, Multi-object Labeling (recursive & sequential), and Face Detection

### Digital Image File Formats



Computer Science

#### Image:

#### Header

Table
= Map = Raster
of Pixel Values

Size of table, color, compression scheme

Gray-scale images: generally
1 byte per pixel
Medical images: typically
2 bytes per pixel

Color images: 3 numbers (each 1 byte) per pixel

# Example: PGM Image



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#### Image file

### Image ??

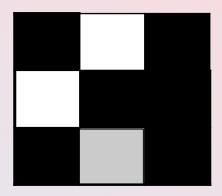
# Example: PGM Image



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### Image file

### **Image**



### Light: Electromagnetic Waves



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#### Wavelength λ

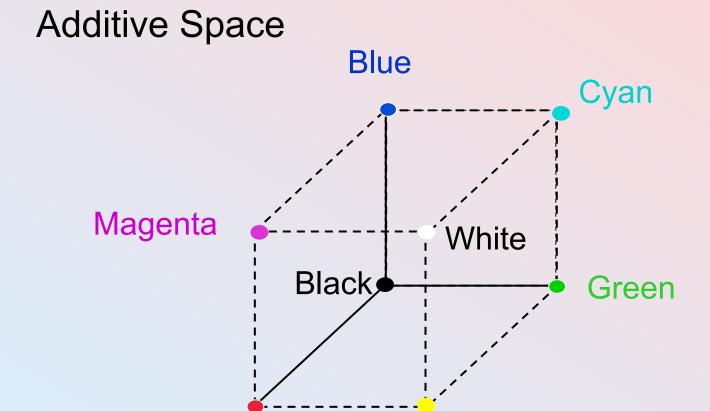
#### Visible Range



### **RGB Color Space**

Red





## **Example: PPM Image**



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### Image file

P3
3 3 255

0 0 0 255 0 0 0 0 0
0 255 0 0 0 255 255 0
0 0 0 0 255 0 0 0

Image ??

### **Example: PPM Image**



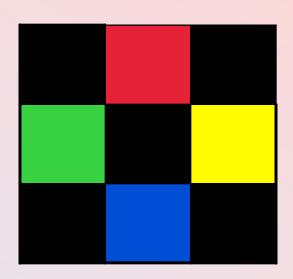
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#### Image file

P3 3 3 255

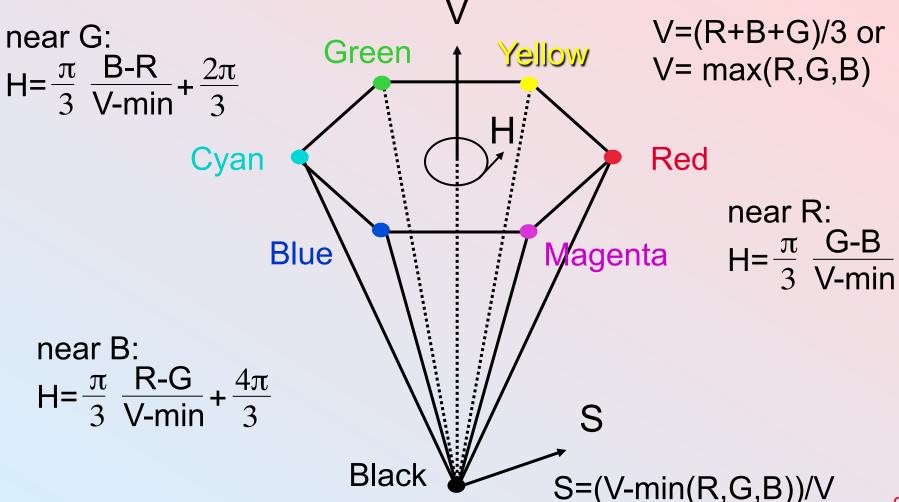
0 0 0 255 0 0 0 0 0 0 255 0 0 0 0 255 255 0 0 0 0 0 0 255 0 0

### **I**mage



### Hue-Saturation-Value (HSV) Color Space





### **Object Detection**



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1. Find bounding box around black object in grayscale background

Background may contain black pixels



Algorithm ??

# Using Black Color & Projections for Object Detection



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1. Find bounding box around black object in grayscale background

Background may contain black pixels



### Algorithm:

- Count number of black pixels in each row and column
- Analyze these histograms or projections of black pixels onto x- and y-axes.

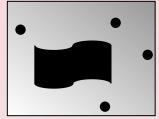
# Using Black Color & Projections for Object Detection



Computer Science

1. Find bounding box around black object in grayscale background

Background may contain black pixels



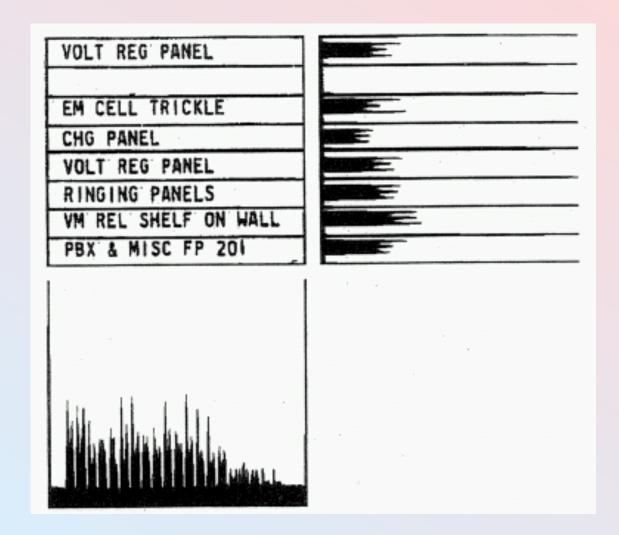
### Algorithm:

- Count number of black pixels in each row and column
- Analyze these histograms or projections of black pixels onto x- and y-axes.

P1(y)= 
$$\Sigma_x B(x,y)$$
, P2(x)=  $\Sigma_y B(x,y)$ 

### **Examples of Projections**





### **Examples of Projections**

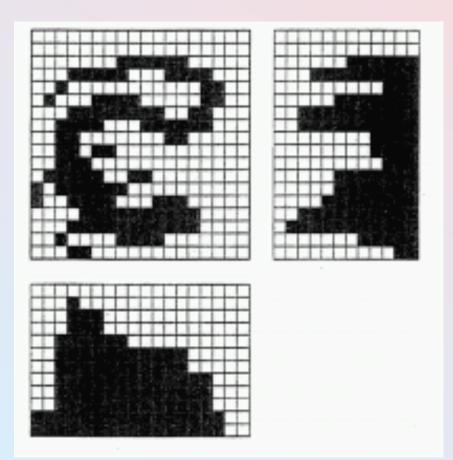




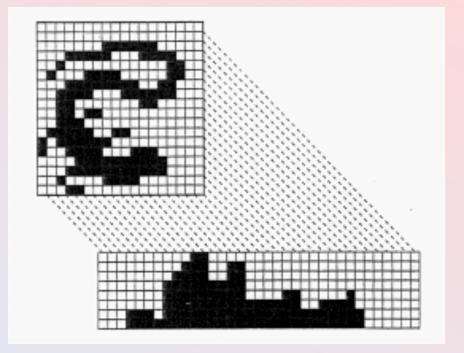
# **Examples of Projections**



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From Machine Vision by Jain et al.



# Object Detection: Alternative Algorithm: "Flood Fill"

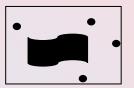


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1. Convert to binary image:



-->



Black = "-1" pixels White = "1" pixels

- Scan row by row until first "-1" pixel is reached and label it "object 1."
- 3. Find all neighbors of the current pixel that are "-1" and assign the object label of the current object in a recursive, depth first search manner.
- 4. When there are no more "-1" neighbors, continue scanning the image until the next "-1" pixel is reached and label it with the next object label and go to step 3.

# Sequential Multi-Object Labeling Algorithm



- □ Horn, page 69
- More explanation with an example next time

D C B A

### **Sequential Labeling Algorithm**



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Case 1:

Case 2:

Case 3:

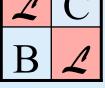
Case 4:

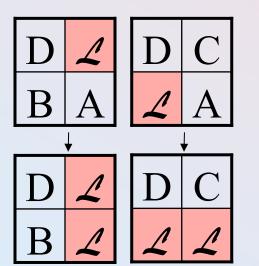
D labeled B, C not

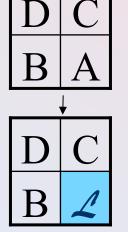
Either
B or C
labeled

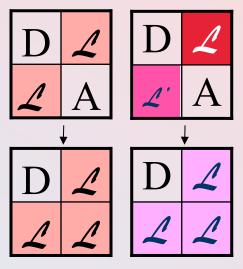
Neither B,C, or D labeled B, C labeled same different

∠ C B A









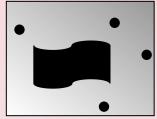
# Back to Algorithm that uses Black Color & Projections for Object Detection



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# 1. Find bounding box around black object in grayscale background

Background may contain black pixels



### Algorithm:

- Count number of black pixels in each row and column
- Analyze these histograms or projections of black pixels onto x- and y-axes.

P1(y)= 
$$\Sigma_x B(x,y)$$
, P2(x)=  $\Sigma_y B(x,y)$ 

# Using Skin Color for Face Detection



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2. Find bounding box around face in color image
Algorithm ??



# Using Skin Color for Face Detection



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# 2. Find bounding box around face in color image

#### Algorithm: Same Idea

- Find all pixels with skin color
- Count number of skin color pixels in each row and column
- Analyze these histograms or projections of skin-color pixels onto x- and y-axes.



Normalized Green

0.7

# Using skin color for face detection

face 9.2 ▲ Figure 6.12 Skin color clusters obtained from training: the horizontal axis is Rnorm and the vertical axis is Gnorm. The cluster labeled as t\_4 is the primary face color, cluster t 5 and t 6 are secondary face clusters associated with shadowed or bearded areas of a face. (Figure from V. Bakic.) ▼ Figure 6.13 Face extraction examples: (left) input image; (middle) labeled image; and (right) boundaries of the extracted face region. (Images from V.Bakic.)

From Computer Vision by Shapiro & Stockman



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### facial shadows

Normalized Red



beard