#### **Computer Vision**

# CSE 455 Color and Texture

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#### We don't really understand vision

- Visual cortex highly studied part of the brain
- Only rough idea of what different components do
- New discoveries in vision all the time
  - Eye uses blinking to reset its rotational orientation
  - Visual cortex can make some "high-level" decisions



### Is vision easy or hard for humans?

- What do you think?

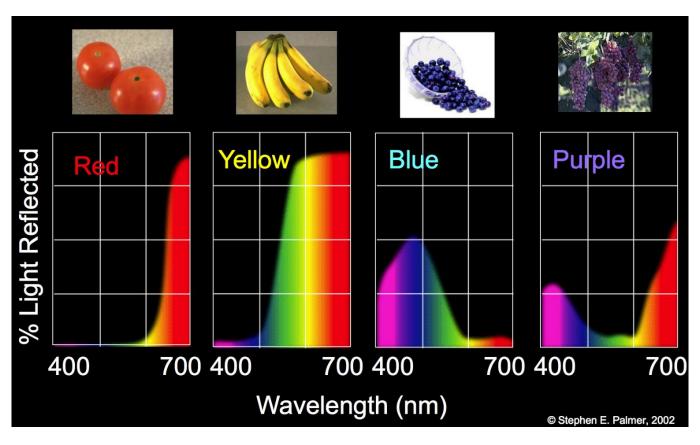
### Is vision easy or hard for humans?

What do you think now?

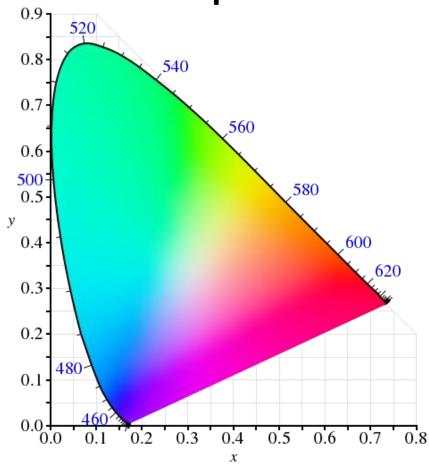




### Objects reflect only some light



## We can make a map of color

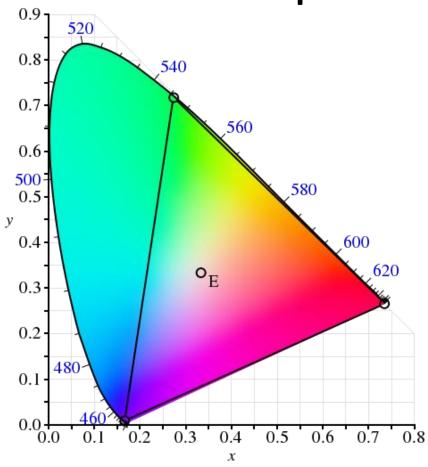


#### Linear colorspace

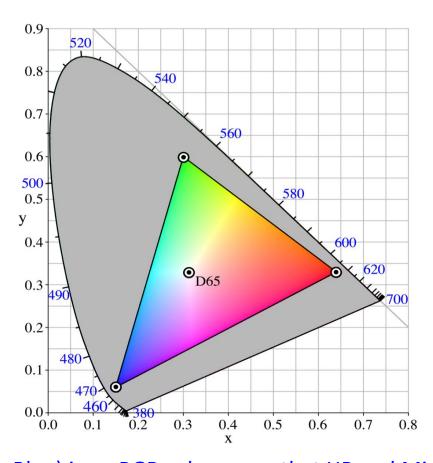
- Pick some primaries
- Can mix those primaries to match any color inside the triangle
- There is a commission that studies color!

 Commission internationale de l'éclairage (CIE) is a 100 year old organization that creates international standards related to light and color.

## "Theoretical" CIE RGB primaries



### Practical sRGB primaries, MSFT 1996



sRGB (standard Red Green Blue) is an RGB color space that HP and Microsoft created cooperatively in 1996 to use on monitors, printers, and the Internet. It was subsequently standardized by the IEC (International Electrotechnical Commission) as IEC 61966-2-1:1999.

### What does this mean for computers?

- We represent images as grids of pixels
- Each pixel has a color, 3 components: RGB
- Not every color can be represented in RGB!
  - Have to go out in the real world sometimes
- RGB is not fully accurate
- We can represent a color with 3 numbers
  - #ff00ff; (1.0, 0.0, 1.0); 255,0,255
  - WHAT COLOR IS THAT?



#### **Linda Shapiro's Schedule: Spring 2020**

#### FF00AA



Time	Monday	Tuesday	Wednesday	Thursday	Friday
9:30-10:30					
10:00-11:20		CSE 455 online		CSE 455 online	
11:00-12:00			Office Hour	ECE Faculty Meeting	
12:30-2:00	Lunch	Multimedia Meeting	Lunch	Lunch	Lunch
1:30-3:30				CSE Faculty Meeting	
3:30-4:30				CSE Colloquium	
4:30-6:00				FERG Meeting	























#### Linda Shapiro's Schedule: Winter 2024

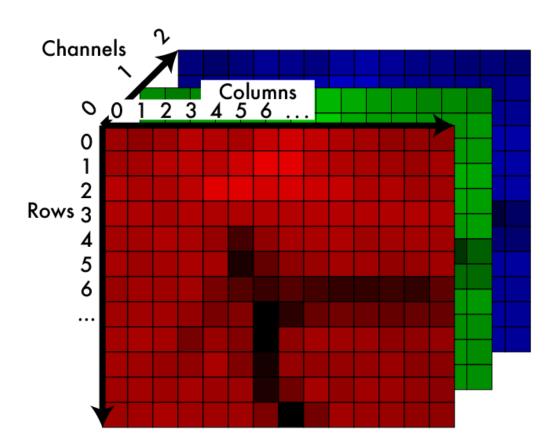
#### 0000FF



Time	Monday	Tuesday	Wednesday	Thursday	Friday
9:30-11:30		10:00-11:20 CSE 455		10:00-11:20 CSE 455	10:30 Grants Meeting
11:30-12:30	12:00 OH	Histology Big Meeting 2nd & 4th Tuesdays		ECE Faculty Meeting	
12:30-2:00		12:30 Multimedia Meeting		1:30 CSE Faculty Meeting	
2:00-3:00				2-3:30 CSE Faculty Meeting	
3:00-4:00	Amazon Meeting				

### Image: 2d array of color

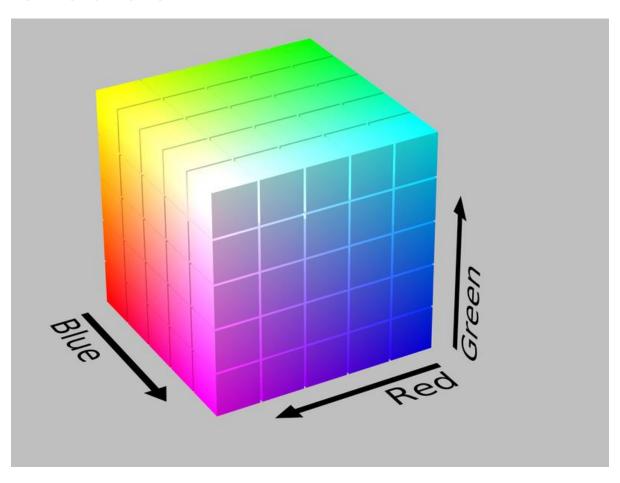
- Some range
  - [0,255]
  - 0.0 1.0
- We'll talk more about this later.



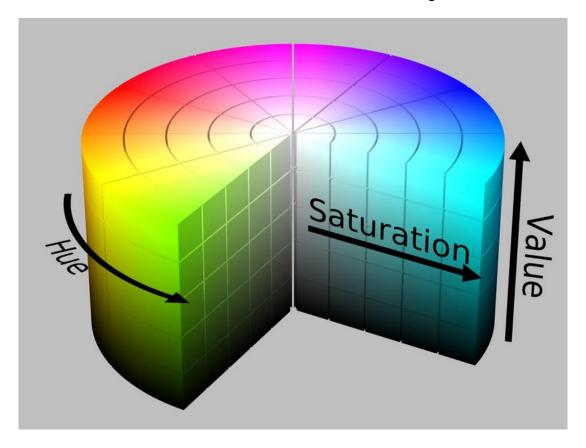
### Grayscale - making color images not

- We can simulate monochromatic images from RGB
- Want a good approximation of how "bright" the image is without color information
- (R+B+G/3) looks weird
- We should
  - Gamma decompress
  - Calculatight lightness
  - Gamma compress
- We can just operate on sRGB
  - Typically  $\sim .30R + .59G + .11B$

#### RGB is a cube...

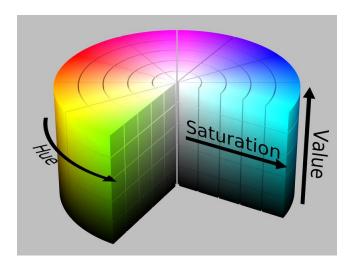


### Hue, Saturation, Value: cylinder!

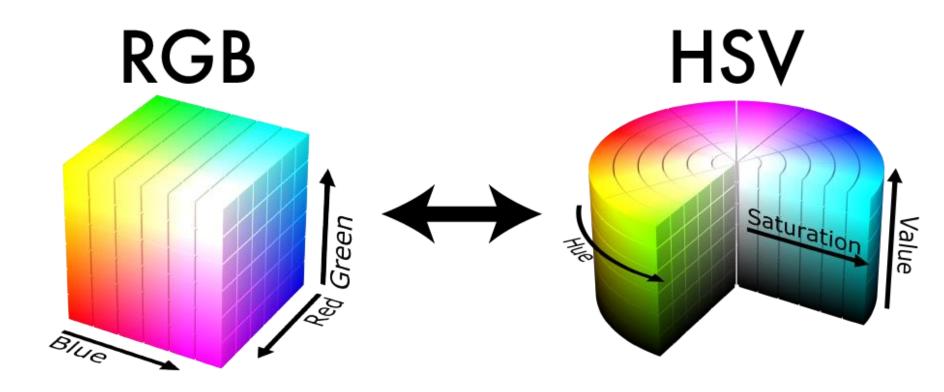


#### Hue, Saturation, Value

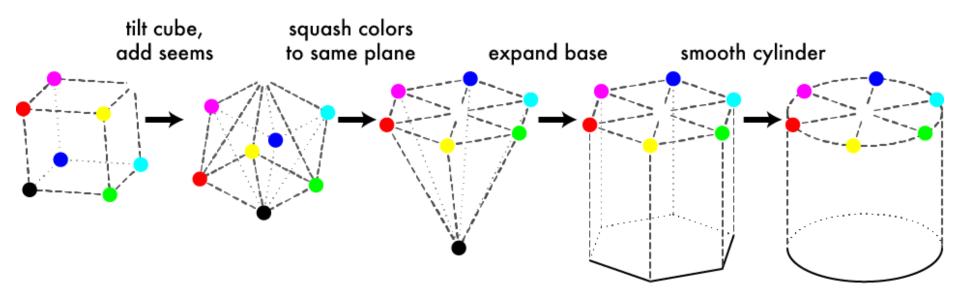
- Different model based on perception of light
- Hue: what color
- Saturation: how much color
- Value: how bright
- Allows easy image transforms
  - Shift the hue
  - Increase saturation



### Other colorspaces are fun!



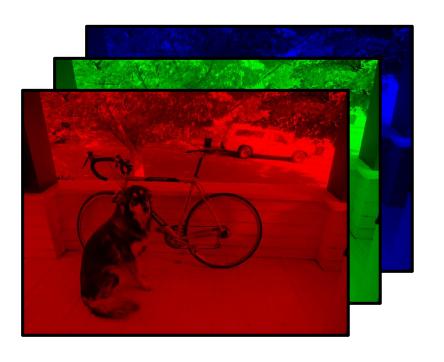
#### Geometric HSV to RGB:

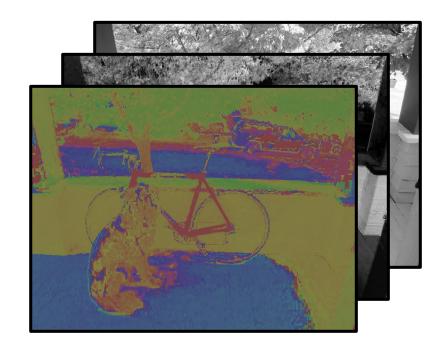


## An RGB Image

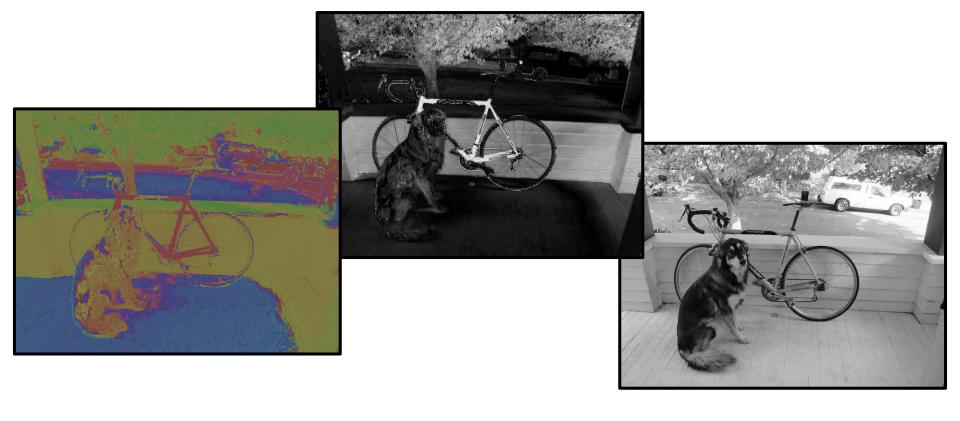


## Still 3d tensor, different info

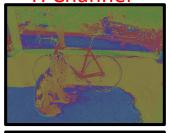


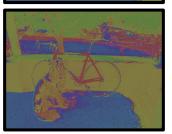


### Hue Saturation Value















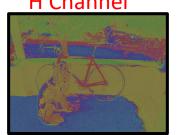


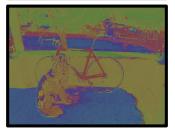






More value = lighter image



















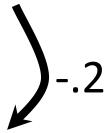


2x

## Shift hue = shift colors S Channel















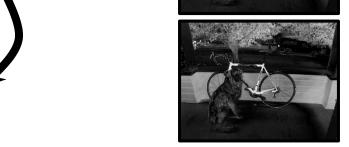
























Or pattern...





S Channel





**V** Channel





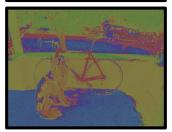






Increase and threshold saturation





















## More Details on Color Spaces

RGB

HSI/HSV

CIE L\*a\*b

YIQ

Opponent

standard for cameras

allows us to separate

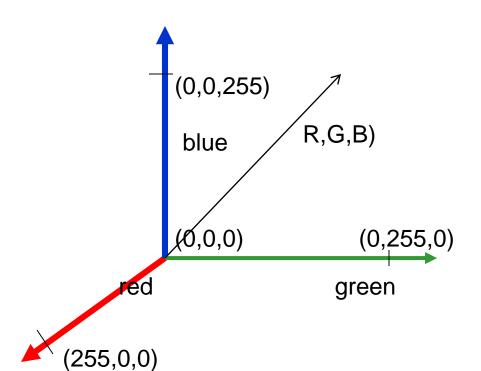
intensity plus 2 color channels

color TVs, Y is intensity

used in Swain & Ballard work

### **RGB Color Space**

#### **Absolute**



#### Normalized

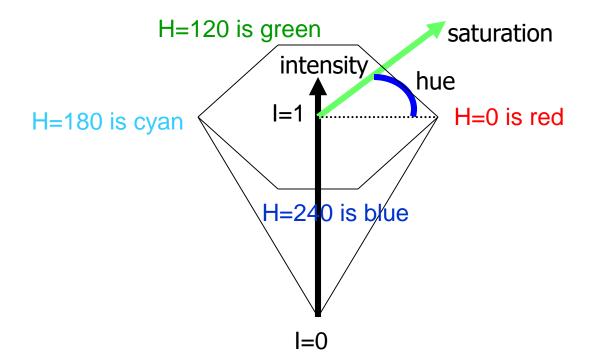
Normalized red r = R/(R+G+B)

Normalized green g = G/(R+G+B)

Normalized blue b = B/(R+G+B)

#### Color hexagon for HSI (HSV)

- Hue is encoded as an angle (0 to  $2\pi$ ).
- Saturation is the distance to the vertical axis (0 to 1).
- Intensity is the height along the vertical axis (0 to 1).



#### Conversion from RGB to YIQ

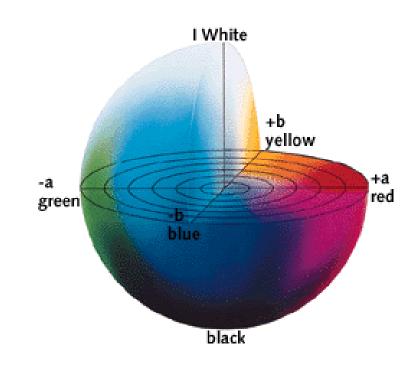
An approximate linear transformation from RGB to YIQ:

$$\begin{array}{rcl} luminance & Y & = & 0.30R \, + \, 0.59G \, + \, 0.11B \\ & R - cyan & I & = & 0.60R \, - \, 0.28G \, - \, 0.32B \\ magenta - green & Q & = & 0.21R \, - \, 0.52G \, + \, 0.31B \end{array}$$

We often use this for color to gray-tone conversion.

#### CIELAB, Lab, L\*a\*b

- One luminance channel (L) and two color channels (a and b).
- In this model, the color differences which you perceive correspond to Euclidian distances in CIELab.
- The a axis extends from green (-a) to red (+a) and the b axis from blue (-b) to yellow (+b). The brightness (L) increases from the bottom to the top of the three-dimensional model.



CIE L\*a\*b\* (CIELAB) is a <u>color space</u> specified by the <u>International Commission on</u> <u>Illumination</u> (French Commission internationale de l'éclairage, hence its CIE <u>initialism</u>).

#### Histograms

A histogram of a gray-tone image is an array
 H[\*] of bins, one for each gray tone.

 H[i] gives the count of how many pixels of an image have gray tone i.

 P[i] (the normalized histogram) gives the percentage of pixels that have gray tone i.

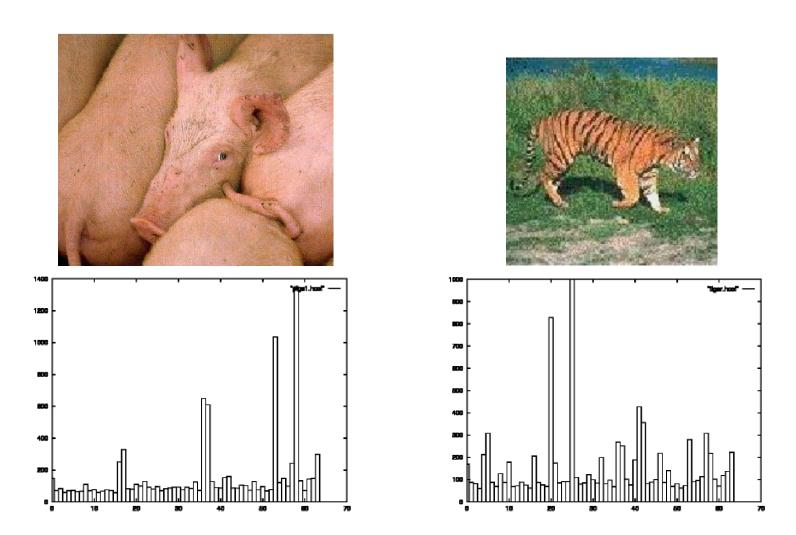
# Color histograms can represent an image

Histogram is fast and easy to compute.

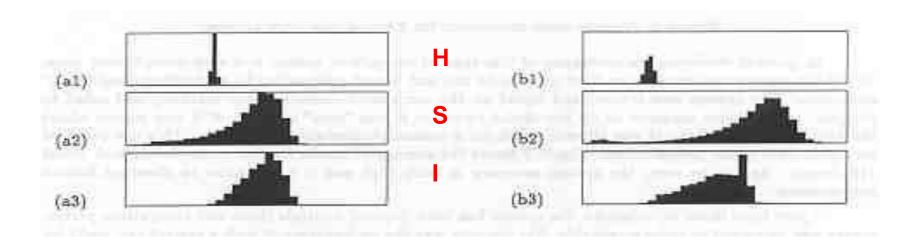
 Size can easily be normalized so that different image histograms can be compared.

 Can match color histograms for database query or classification.

# Histograms of two color images



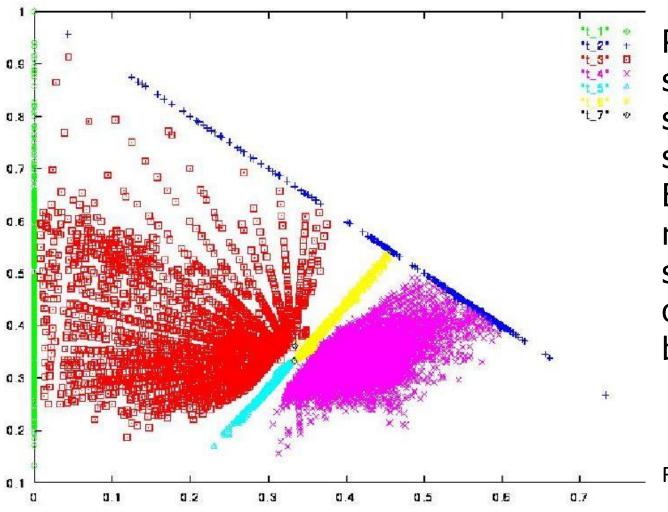
# **Apples versus Oranges**



Separate HSI histograms for apples (left) and oranges (right) used by IBM's VeggieVision for recognizing produce at the grocery store checkout station (see Ch 16).

# Skin color in Normalized R-G Space

#### **G** normalized

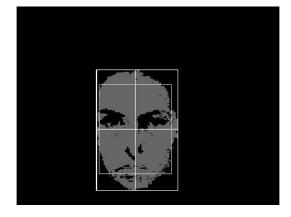


Purple region shows skin color samples from several people. Blue and yellow regions show skin in shadow or behind a beard.

# Finding a face in video frame







- (left) input video frame
- (center) pixels classified according to RGB space
- (right) largest connected component with aspect similar to a face (all work contributed by Vera Bakic)

# Swain and Ballard's Histogram Matching for Color Object Recognition (IJCV Vol 7, No. 1, 1991)

Opponent Encoding:

• 
$$wb = R + G + B$$

• 
$$rg = R - G$$

• by = 
$$2B - R - G$$

Histograms:  $8 \times 16 \times 16 = 2048$  bins

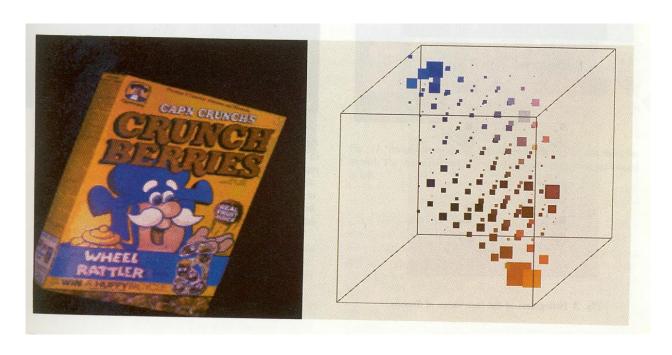
Intersection of image histogram and model histogram:

intersection(h(I),h(M)) = 
$$\sum_{j=1}^{\text{numbins}} \min\{h(I)[j],h(M)[j]\}$$

Match score is the normalized intersection:

$$\frac{\mathsf{match}(\mathsf{h}(\mathsf{I}),\mathsf{h}(\mathsf{M}))}{\mathsf{match}(\mathsf{h}(\mathsf{I}),\mathsf{h}(\mathsf{M}))} = \text{intersection}(\mathsf{h}(\mathsf{I}),\mathsf{h}(\mathsf{M})) / \sum_{j=1}^{\mathsf{numbins}} \mathsf{h}(\mathsf{M})[j]$$

### (from Swain and Ballard)

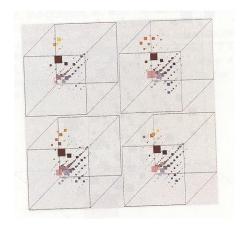


cereal box image

3D color histogram



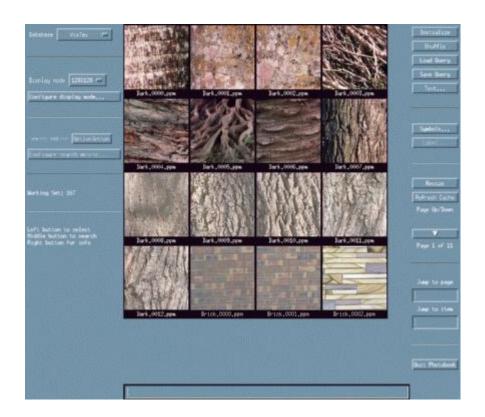
Four views of Snoopy



Histograms

### **Texture**

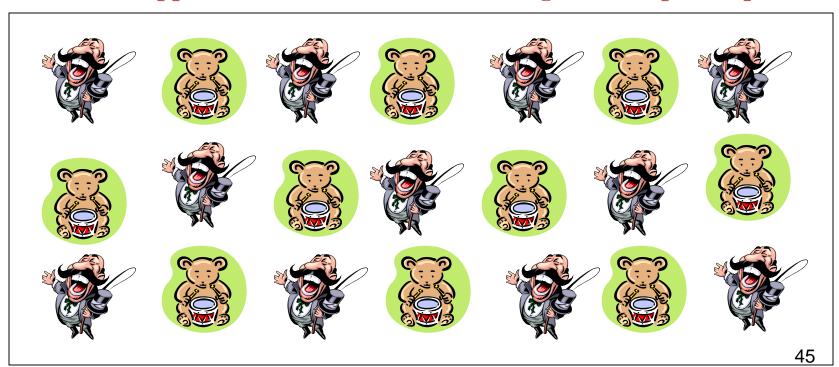
- Color is well defined.
- But what is texture?



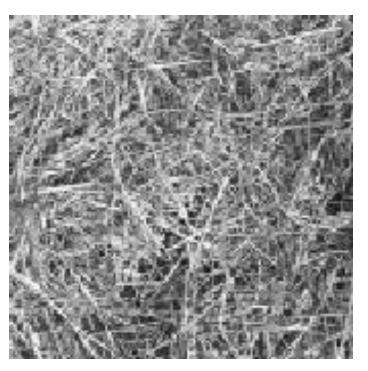
### Structural Texture

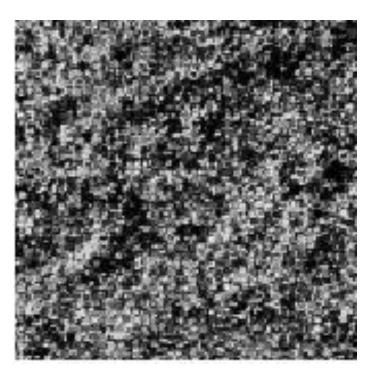
Texture is a description of the spatial arrangement of color or intensities in an image or a selected region of an image.

Structural approach: a set of texels in some regular or repeated pattern



## Natural Textures from VisTex





grass leaves

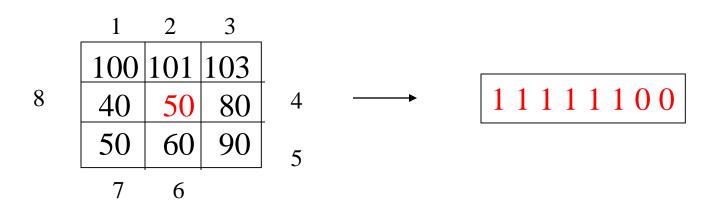
What/Where are the texels?

## The Case for Statistical Texture

- Segmenting out texels is difficult or impossible in real images.
- Numeric quantities or statistics that describe a texture can be computed from the gray tones (or colors) alone.
- This approach is less intuitive, but is computationally efficient.
- It can be used for both classification and segmentation.

# Local Binary Pattern Measure

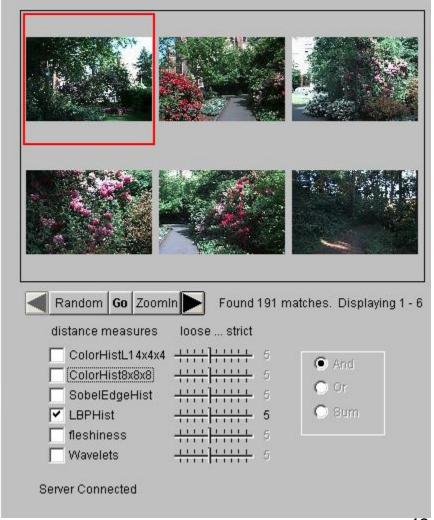
- For each pixel p, create an 8-bit number b<sub>1</sub> b<sub>2</sub> b<sub>3</sub> b<sub>4</sub> b<sub>5</sub> b<sub>6</sub> b<sub>7</sub> b<sub>8</sub>, where b<sub>i</sub> = 0 if neighbor i has value less than or equal to p's value and 1 otherwise.
- Represent the texture in the image (or a region) by the histogram of these numbers.



## Example

Fids (Flexible Image Database System) is retrieving images similar to the query image using LBP texture as the texture measure and comparing their LBP histograms

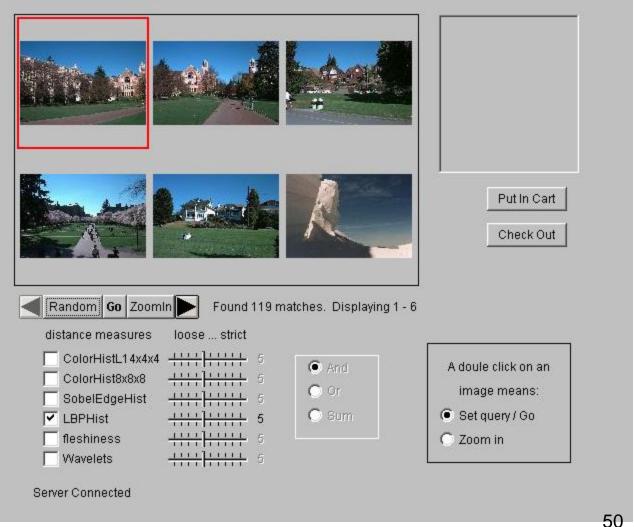
### Fids demo



## Example

Low-level measures don't always find semantically similar images.

### Fids demo



# What else is LBP good for?

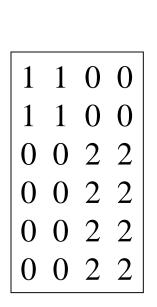
- We found it in a paper for classifying deciduous trees.
- We used it in a real system for finding cancer in Pap smears.
- We are using it to look for regions of interest in breast and melanoma biopsy slides.

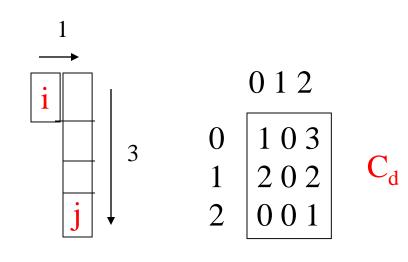
### Co-occurrence Matrix Features

A co-occurrence matrix is a 2D array C in which

- Both the rows and columns represent a set of possible image values.
- $C_d(i,j)$  indicates how many times value i co-occurs with value j in a particular spatial relationship d.
- The spatial relationship is specified by a vector  $\mathbf{d} = (\mathbf{dr}, \mathbf{dc})$ .

## Co-occurrence Example





$$d = (3,1)$$

co-occurrence matrix

gray-tone image

From  $C_d$  we can compute  $N_d$ , the normalized co-occurrence matrix, where each value is divided by the sum of all the values.

### Co-occurrence Features

### What do these measure?

$$Energy = \sum_{i} \sum_{j} N_d^2(i,j)$$
 (7.7)

$$Entropy = -\sum_{i} \sum_{j} N_d(i, j) log_2 N_d(i, j) \qquad (7.8)$$

$$Contrast = \sum_{i} \sum_{j} (i - j)^{2} N_{d}(i, j) \qquad (7.9)$$

$$Homogeneity = \sum_{i} \sum_{j} \frac{N_d(i,j)}{1 + |i-j|}$$

$$(7.10)$$

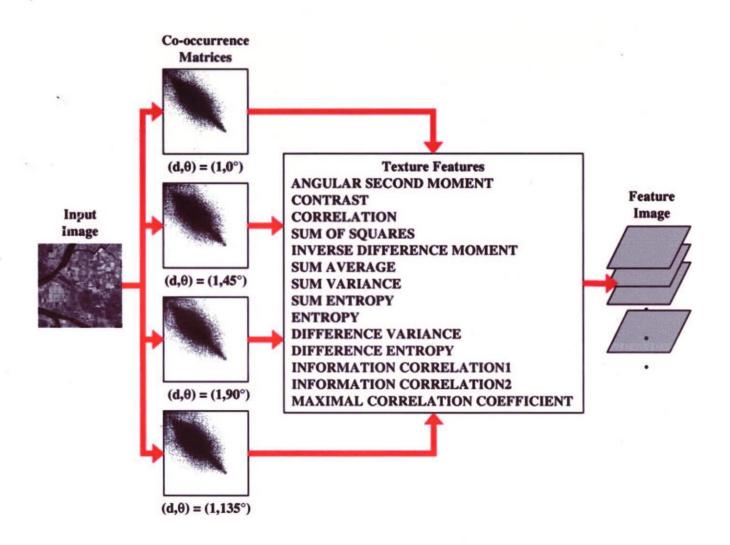
Correlation = 
$$\frac{\sum_{i} \sum_{j} (i - \mu_{i})(j - \mu_{j}) N_{d}(i, j)}{\sigma_{i} \sigma_{j}}$$
(7.11)

where  $\mu_i$ ,  $\mu_j$  are the means and  $\sigma_i$ ,  $\sigma_j$  are the standard deviations of the row and column sums.

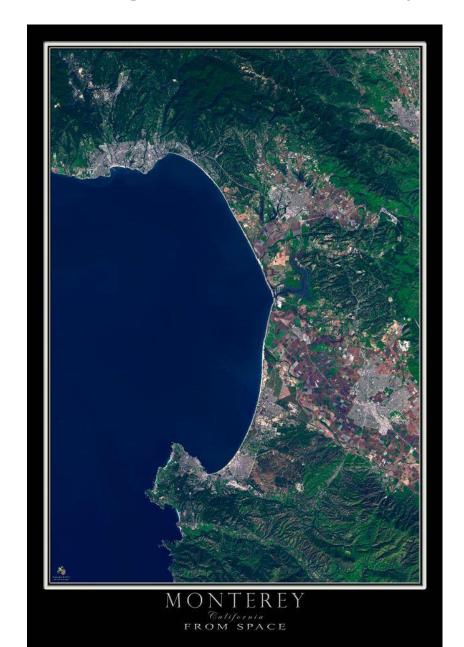
# What are Co-occurrence Features used for?

- They were designed for recognizing different kinds of land uses in satellite images.
- They are still used heavily in geospatial domains, but they can be added on to any other calculated features.

# Example



# Satellite Image of Monterey Bay



# Assignment 0

**FUN WITH COLOR!** 

# Image basics

Data structure for an image

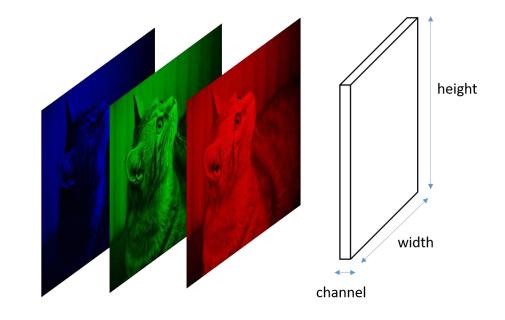
```
typedef struct{
    int h, w, c;
    float *data;
} image;
```

# Image basics

Data structure for an image

```
typedef struct{
    int h, w, c;
    float *data;
} image;
```

- -h = height
- -w = width
- -c = number of channels
  - c = 3 for RGB image; c = 1 for grayscale image



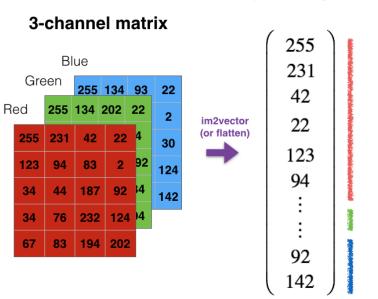
# Image basics

Data structure for an image

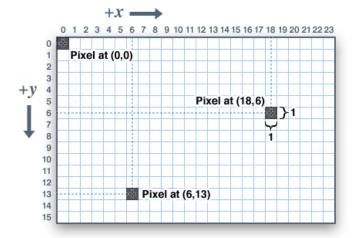
```
typedef struct{
    int h, w, c;
    float *data;
} image;
```

- data = array of floats
- floats in the range [0,1]
- 3D image matrix linearized into 1D array

#### reshaped image vector



- Fill in:
  - float get\_pixel(image im, int x, int y, int c)



- void set\_pixel(image im, int x, int y, int c,
  float v)
  - assign value v to im.data[coord(x,y,c)]

- Fill in:
  - image copy\_image(image im)
    - create a separate image that is a copy of the input image im and return the copy
    - Useful functions:
      - make\_image() in load\_image.c
      - memcpy()
    - You can use get\_pixel() and set\_pixel() functions you implemented from now on

- Fill in:
  - image rgb\_to\_grayscale(image im)
    - create a new image with 1 channel
    - Get R,G,B values from input image im
    - Compute Y = 0.299 R + 0.587 G + .114 B
    - Assign Y to new image and return that image

- Fill in:
  - void shift\_image(image im, int c, float v)
    - add value v to each pixel of im in channel c
    - change the input image in-place (do not create a separate image) wherever the return type of the function is void

- Fill in:
  - void clamp\_image(image im)
    - clamp the pixel values of input image im to be in the range [0,1]

- Fill in:
  - void rgb\_to\_hsv(image im)
    - Convert R,G,B values of image im pixelwise into H,S,V values using the given formulas
    - You can use the three\_way\_max() and three\_way\_min() functions provided

- Fill in:
  - void hsv\_to\_rgb(image im)
    - Convert H,S,V values of image im pixelwise into R,G,B values using the given table

# To Do #8 (extra credit)

### • Fill in:

- void scale\_image(image im, int c, float v)
  - multiply each pixel of im in channel c with value v
  - add to the lines in image.h and other files as necessary to test the function. No need to submit the other edited files; write a comment in CAPS at the top of process\_image.c if you attempt any extra credit

# To Do #9 (super extra credit)

- Fill in:
  - void rgb\_to\_lch(image im)
    - Convert R,G,B values of image im pixelwise into L,C,H
       values using the formulas in the given link

# Have Fun