## **Image Processing**

- An image can be represented by functions of two *spatial* variables *f*(*x*,*y*), where *f*(*x*,*y*) is the *brightness* of the gray level of the image at a spatial coordinate (*x*,*y*)
- A multispectral image is a **f** is a vector-valued function with components (*f*<sub>1</sub>, *f*<sub>2</sub>, ..., *f*<sub>n</sub>); a special case is a color image in which the components measure the brightness values of each of three wavelengths, that is:

$$\mathbf{f}(\mathbf{x}) = \left\{ f_{red}(\mathbf{x}), f_{green}(\mathbf{x}), f_{blue}(\mathbf{x}) \right\}$$
$$\mathbf{x} = (x, y)$$



#### RGB planes decomposed...



# **Point Operations**

• In a point operation each pixel in the output image is a function of the grey-level (or color value) of the pixel at the corresponding position in the input image

out(x,y) = f(in(x,y))

• For example: photometric decalibration, contrast stretching, thresholding, background subtraction...

# Histogram

- A grey level histogram is a function that gives the frequency of occurrence of each gray level in the image
- If the gray levels are quantized in *n* values (usually 256), the value of the histogram at a particular gray level *p*, *h(p)*, is the number of pixels in the image with that gray level
- Often it is expressed in terms of *fraction* of pixels





(image 512x512)

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## How do we compute the histogram

function histo=computeHisto(A)

histo=zeros(1,256);

R=size(A,1); C=size(A,2);

```
for r=1:R
for c=1:C
index=A(r,c);
histo(index+1)=histo(index+1)+1;
end
end
```

#### Some characteristic histograms...





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# Negative





function S=negative(A)

R=size(A,1); C=size(A,2);

%prepare image S=zeros(R,C); ... for r=1:R for c=1:C S(r,c)=255-double(A(r,c)); end end

• • •

# Threshold

- Produces a two-level image
- We pick a threshold t, we set to 255 all pixels whose value > t, 0 all the others





### Histogram Stretch

- From the histogram it is possible to see if there are levels in the image that are not used
- We can map the levels of the image to expand the histogram



### Histogram stretch: sample code

function S=stretchHisto(A, min, max)

```
%%%%% build look up table

lut=zeros(1,256);

for i=0:255

    if (i<min)

        lut(i+1)=0;

    elseif (i>max)

        lut(i+1)=255;

    else

        lut(i+1)=(i-min)*255/(max-min);

    end

end

%%%%%%
```

. . . .

...
R=size(A,1);
C=size(A,2);
%prepare image
S= zeros(R,C);
for r=1:R
for c=1:C
 index= A(r,c)+1;
 S(r,c)=lut(index);
 end
end

### Histogram equalization

- Equally use all gray levels
- Find a transformation y=f(x) to "flatten" the histogram



#pixels is unchanged:  $p(y) \cdot dy = p(x) \cdot dx$  Assume x and y between 0 and 1 We would like p(y) = constant = 1

$$p(y) \cdot dy = p(x) \cdot dx$$
$$dy = p(x) \cdot dx \Longrightarrow \frac{dy}{dx} = p(x)$$
$$y(x) = \int_0^x p(u) du$$

cumulative probability distribution



- low  $p(x) \rightarrow \text{smooth } f(x) \rightarrow \text{narrow } d(y)$
- large  $p(x) \rightarrow \text{steep } f(x) \rightarrow \text{large } d(y)$

#### Discrete case

• x and y assume discrete values between [0,L-1] (often L=256)

$$y(x) = \int_0^x p(u) du \longrightarrow y'(x) = \sum_0^x P_i$$
  
number of pixels that have value i  
$$P_i = \frac{n_i}{N} \longleftarrow \text{ total number of pixels (HxW)}$$

• y' assumes values between  $[0,1] \rightarrow$  needs to be scaled to [0, L-1]:

$$y = \left\lfloor \frac{y' - y'_{\min}}{1 - y'_{\min}} (L - 1) + 0.5 \right\rfloor$$
 integral part of a real number

### Example



Note: the resulting histogram is not perfectly flat because we cannot separate pixels having the same gray level, the resulting cumulative distribution, however, would approximate a linear ramp







# **Detect Changes**

- Take the difference between each pixel in two images A and B (grayscale): B="background"
  - A=new image
  - D=abs(A-B)
- Extend the concept to a sequence of images
- At each instant in time we take the difference between the current frame and the previous one: D=abs(A(t)-A(t-1))

Detection can be done by thresholding: *Out=threshold(D,th);* 

### Image Difference

function imageDiff(basename, start, last)

cFrame=sprintf('%s%d.ppm', basename, start); A=imRead(cFrame); PREV=rgb2Gray(A);

for i=start:last
 cFrame=sprintf('%s%d.ppm', basename, i);
 % read new image
 A=imRead(cFrame);

% convert to grayscale G=rgb2Gray(A);

% take the difference between the current frame and the previous one D=double(G)-double(PREV); % compute the abs value D=abs(D); % threshold diff\_th=im2bw(uint8(D),50/255);

% store frame PREV=G;

%%%% PLOT figure(1), subplot(1,2,1), imShow(uint8(A)), drawnow; figure(1), subplot(1,2,2), imShow(uint8(255\*diff\_th)), drawnow;

pause(0.05); %%wait some time end



# Another option

• Model the background by taking into account more than a single frame:

B=a\*A(t-1)+(a-1)\*B D=abs(A(t)-B)

*a* determines how fast we update the background:  $a=1 \rightarrow$  image difference  $a=0 \rightarrow$  persistent background (never updated)

## Color Histograms

- Count the color of the pixels of the images
- It is a statistical description of the color of the image, useful to characterize a particular object
- Appealing because invariant to translation and rotation, slowly changing with scale and view point
  - r,g,b → 3D function, intensity dependent, easily too large (es: 256x256x256x32 ~ 64MB)
  - discard luminance, use H,S or r,g  $\rightarrow$  2D

### Color Histogram: examples









bin size: 16x16

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## **Comparing Histograms**

- Suppose we want to compare two histograms I and M, each with n bins
- Useful to solve the *identification problem*: compare two images M and I and decide if they are similar
- Intersection, the number of pixels from the model that correspond to pixels of the same color in the image, formally:

 $\sum_{j=1}^{n} \min(I_j, M_j)$ 

• Normalize by the number of pixels in the histogram M:

$$H(I,M) = \frac{\sum_{j=1}^{n} \min(I_{j}, M_{j})}{\sum_{j=1}^{n} M_{j}}$$

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Swain and Ballard 1991



### Histogram Backprojection

- Assume we have a model of an object (its color histogram)
- Localization problem: where in the image are the colors of the object being looked for?
- The histogram gives the probability of occurrence of the colors of the object, or *p(color/object)*
- We can approximate:



• Similar approach, compute the "ratio histogram" (Swain and Ballard, 1991):

$$R_i = \min\left(\frac{M_i}{I_i}, 1\right)$$

- Perform backprojection of R into the image
- Heuristic to deemphasize colors that are not in the object looked for (for which I>M)
- Search for a uniform region whose size matches the one of the object



Compute histogram





Backprojection (ratio histogram)



### Examples:

- Swain and Ballard 1991, use color histograms to recognize objects
- Skin detection, preprocessing for face detection...
  - Example (Peer 2003)

Assume (r,g,b) space (and daylight illumination) classify (r,g,b) as skin if:

r > 95 and g > 40 and b > 20,  $Max\{r,g,b\} - min\{r,g,b\} > 15$ , and |r-b| > 15 and r > g and r > b

