

Image Processing

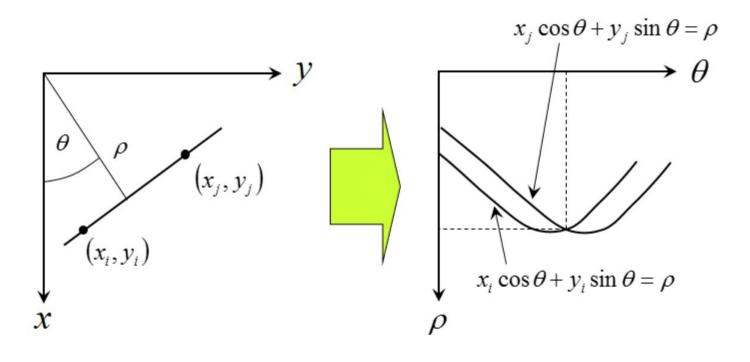
Image Segmentation (Part II)

Pattern Recognition and Image Processing Laboratory (Since 2012)

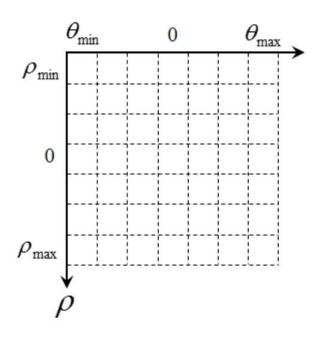


One approach that can be used to find and link line segments in an image is the Hough transform.

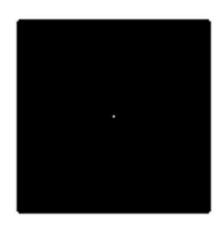




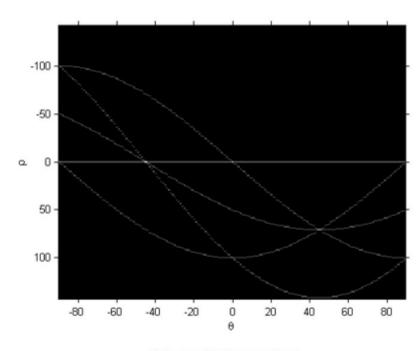








Binary image with five dots



Hough transform

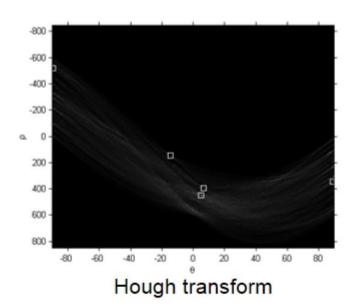
ex_hough_5dots.m



 Line segments corresponding to the Hough transform peaks



Edge detection image

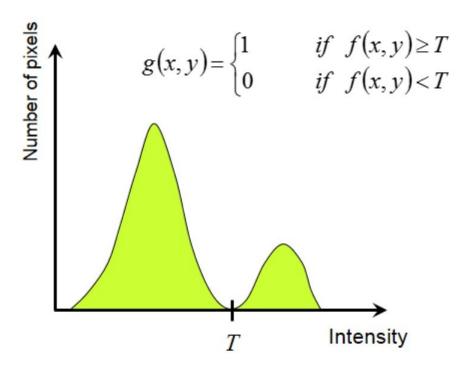


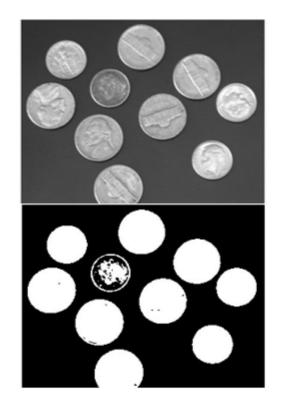


 Line segments corresponding to the Hough transform peaks







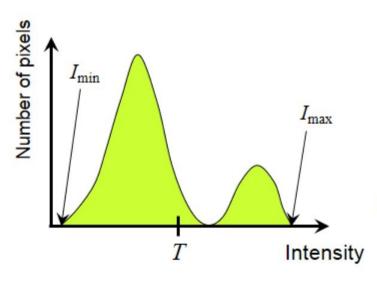


T is a specified threshold.



Global Thresholding

For choosing a threshold automatically, Gonzalez and Woods describe the following iterative procedure.



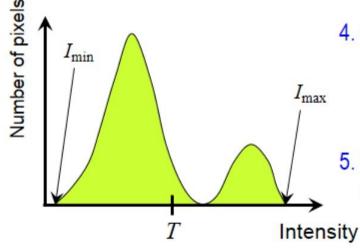
1. Select an initial estimate for T

$$T = \frac{I_{\text{max}} + I_{\text{min}}}{2}$$

2. Segment the image using T



Global Thresholding



- 3. Compute the average intensity values μ_1 and μ_2 for the pixels in regions G_1 and G_2
- 4. Compute a new threshold T

$$T = \frac{\left(\mu_1 + \mu_2\right)}{2}$$

5. Repeat steps 2-4 until *T* is not change or less than a specified value.

Global Thresholding

```
>> f = imread('rice.tif');
>> T = 0.5*(double(min(f(:))) + double(max(f(:))));
>> done = false;
>> while ~done
\Rightarrow q = f \Rightarrow T;
>> Tnext = 0.5*(mean(f(q)) + mean(f(~q)));
\rightarrow done = abs(T-Tnext) < 0.5;
>> T = Tnext;
>> end
>> clc
>> T
>> To =graythresh(f)*255
```

Local Thresholding

Local threshold

$$g(x,y) = \begin{cases} 1 & \text{if } f(x,y) \ge T(x,y) \\ 0 & \text{if } f(x,y) < T(x,y) \end{cases}$$

where
$$T(x,y) = f_o(x,y) + T_o$$

Morphological opening of f

Automatic threshold

Local Thresholding: MATLAB code

```
>> f = imread('rice.tif');
>> figure(1); imshow(f);
>> se =strel('disk', 10);
>> fo =imopen(f, se);
>> figure(2); imshow(fo);
>> To =graythresh(fo);
>> T = fo + (To*255);
>> figure(3); imshow(T);
>> [m, n] = size(f);
>> out = zeros(m, n);
>> out idx = find(f >= T);
>> out(out idx)) = 1;
>> figure(4); imshow(out);
```



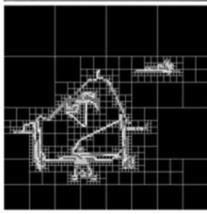
Region-Based Segmentation

Basic Formulation

Let R representation the entire image region. We may view segmentation as a process that partitions R into n subregions, $R_1, R_2, ..., R_n$, such that







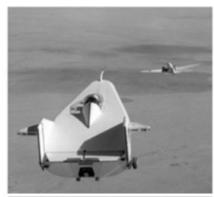


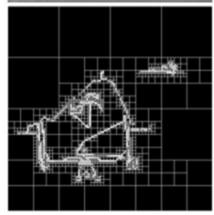
Region-Based Segmentation

Basic Formulation

$$1 \quad \bigcup_{i=1}^n R_i = R$$

- **2** R_i is a connected region, i = 1, 2, ..., n
- **3** $R_i \cap R_j = \emptyset$ for all i and j, $i \neq j$
- **4** $P(R_i) = \text{TRUE for } i = 1, 2, ..., n$
- **5** $P(R_i \cup R_j) = \text{FALSE for any adjacent regions } R_i \text{ and } R_j.$





 $P(R_i)$ is a logical predicate.







 Watershed Segmentation Using the Distance Transform

It is the distance from every pixels to the nearest nonzero-valued pixel.



Example: Segmenting a binary image using the distance and Watershed Transforms.

>> watershed_dt.m % See demo



Watershed Segmentation Using Gradients

The key concept of this method is that the gradient magnitude is used often to preprocess a gray-scale image prior to using the Watershed Transform for segmentation.



Example: Segmenting a gray-scale image using gradients and the Watershed Transform.

>> watershed_g.m % See demo



Marker Controlled Watershed Segmentation

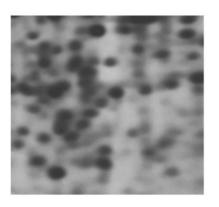


Direct application of the watershed transform to a gradient image usually leads to over-segmentation due to noise and other local irregularities of the gradient.

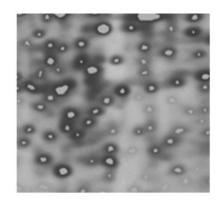


Marker Controlled Watershed Segmentation

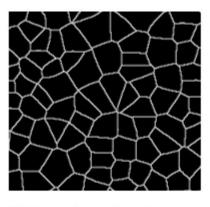
An approach used to control over-segmentation is based on the concept of markers



Original image



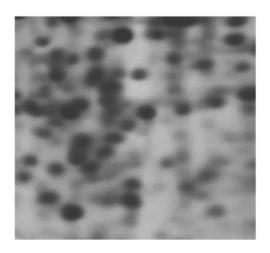
Internal marker image



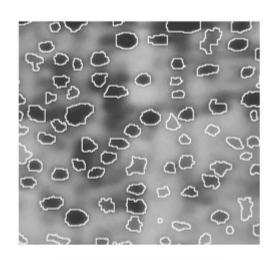
External marker image



Marker Controlled Watershed Segmentation



Original image



Segmented image



```
%% line segment.m
f = imread('building.tif');
g canny best = edge(f, 'canny', [0.04 \ 0.10], 1.5);
figure(1); imshow(g canny best)
[H, theta, rho] = hough(g canny best, 0.5);
figure(2);
imshow(H,'XData', theta, 'YData',rho); axis on, axis normal;
xlabel('\theta'), ylabel('\rho');
[r, c] = houghpeaks(H, 10);
hold on
plot(theta(c), rho(r), 'linestyle', 'none', 'marker', ...
    's', 'color', 'w');
lines = houghlines(g canny best, theta, rho, r, c);
figure(3); imshow(g canny best), hold on
for k = 1:length(lines)
    xy = [lines(k).point1 ; lines(k).point2];
    plot(xy(:, 2), xy(:, 1), 'LineWidth', 4, 'Color', ...
        [.6 .6 .6]);
end
```

```
%% watershed dt.m
```

88 -----

```
clear all
close all

f = imread('bwdowel.tif');
figure, imshow(f);

gc = ~f;
figure, imshow(gc);

D = bwdist(gc);
figure, imshow(mat2gray(D));

L = watershed(-D);
figure, imshow(~L);

w = (L==0);
g2 = f & ~w;
figure, imshow(g2);
```

```
%% watershed g.m
clear all
close all
f = imread('blobs.tif');
figure(1), imshow(f);
h = fspecial('sobel');
fd = double(f);
g = sqrt(imfilter(fd, h, 'replicate').^2 + ...
         imfilter(fd, h', 'replicate').^2);
figure(2), imshow(mat2gray(g));
L = watershed(q);
wr = (L==0);
figure(3), imshow(wr);
g2 = imclose(imopen(g, ones(3,3)), ones(3,3));
L2 = watershed(q2);
wr2 = (L2 == 0);
figure(4), imshow(wr2);
```

```
%% last ex.m
88 -----
clear all
close all
f = imread('gel.tif');
figure(1), imshow(f)
h = fspecial('sobel');
fd = double(f);
g = sqrt(imfilter(fd, h, 'replicate').^2 + ...
        imfilter(fd, h', 'replicate').^2);
g1 = log2(1+double(g));
figure, imshow(mat2gray(g1));
L = watershed(g);
wr = L==0;
figure, imshow(wr);
rm = imregionalmin(g);
figure, imshow(rm)
im = imextendedmin(f, 2);
fim = f;
fim(im) = 175;
figure, imshow(fim)
Lim = watershed(bwdist(im));
em = Lim == 0;
figure, imshow(em);
%g2=imimposemin(f,mark)
gg = mat2gray(g1);
g2 = imimposemin(gg, em | im);
% g22 = log(1+double(g2));
% figure, imshow(mat2gray(g22))
figure, imshow(g2)
L2 = watershed(g2);
f2 = f;
f2(L2 == 0) = 255;
figure, imshow(f2)
```