

# Thresholding Based Segmentation

Robert B. Fisher  
School of Informatics  
University of Edinburgh

Slide 1/15

Slide credit: Bob Fisher

# Isolating flat parts

Isolate parts, then characterise later

Assume

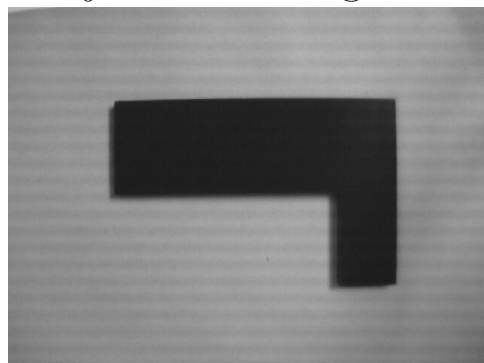
- Dark part
- Light background
- Reasonably uniform illumination – > distinguishable parts

Slide 2/15

Slide credit: Bob Fisher

# Motivating Example

Given this image, how might we label pixels as object and background?



Slide 3/15

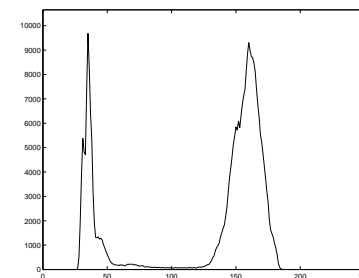
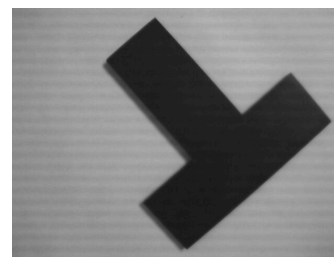
Slide credit: Bob Fisher

# Thresholding Introduction

Key technique: thresholding

Assume pixel values are separable

Part and typical distribution



Spread: not quite uniform illumination + part color variations + sensor noise

Slide 4/15

Slide credit: Bob Fisher

# Thresholding Algorithm

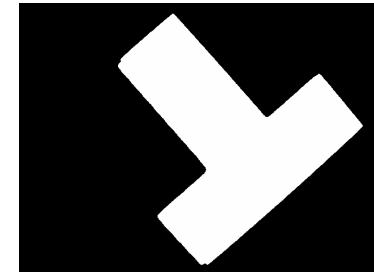
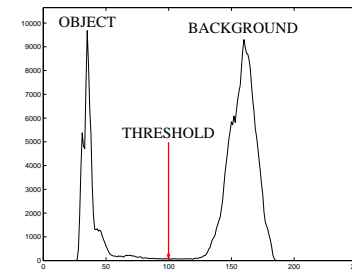
Thresholding: central technique

```
for row = 1 : height
  for col = 1 : width
    if value(row,col) < ThreshHigh % inside high bnd
      % & value(row,col) > ThreshLow % optional low bnd
      output(row,col) = 1;
    else
      output(row,col) = 0;
    end
  end
end
```

Slide 5/15

Slide credit: Bob Fisher

# Thresholding Example 1



Histogram

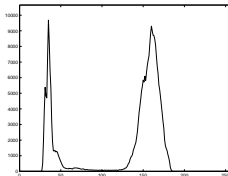
Thresholded Image

Slide 6/15

Slide credit: Bob Fisher

# Threshold Selection 1

Exploit bimodal distribution



But:

- Distributions broad and some overlap – > misclassified pixels
- Shadows dark so might be classified with object
- Distribution has more than 2 peaks

So: smooth histogram to improve shape for selection

Slide 7/15

Slide credit: Bob Fisher

# Convolution

General purpose image (and signal) processing function

Computed by a weighted sum of image data and a fixed mask

Linear operator:  $\text{conv}(a*B,C) = a*\text{conv}(B,C)$


Used in different processes: noise removal, smoothing, feature detection, differentiation, ...

Slide 8/15

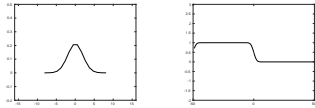
Slide credit: Bob Fisher

## Convolution in 1D

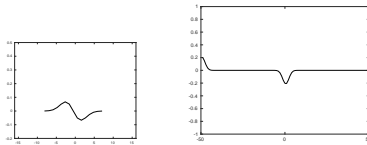
$$Output(x) = \sum_{i=-N}^N weight(i) * input(x - i)$$

Input: 

Gaussian Mask and Output:



Derivative of Gaussian Mask and Output:



Slide 9/15

Slide credit: Bob Fisher

## Histogram Smoothing for Threshold Selection

Histogram Smoothing (in findthresh.m)

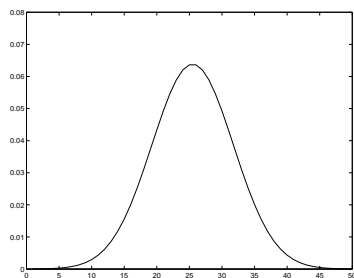
Convolve with a Gaussian smoothing window

```
filterlen = 50; % filter length
thefilter = gausswin(filterlen,sizeparam); % size=4
thefilter = thefilter/sum(thefilter); % unit norm
tmp2=conv(thefilter,thehist); % makes longer output
% select corresponding portion
offset = floor((filterlen+1)/2);
tmp1=tmp2(offset:len+offset-1);
```

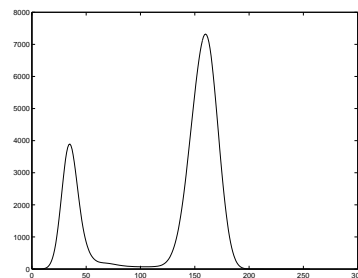
Slide 10/15

Slide credit: Bob Fisher

## Convolved Histogram Example



FILTER SHAPE



SMOOTHED HISTOGRAM

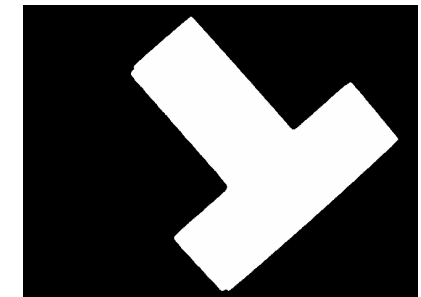
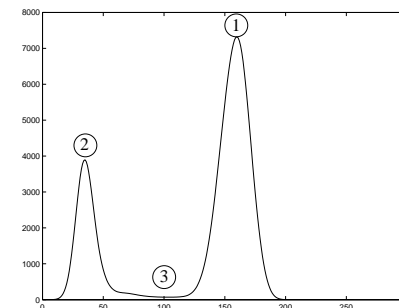
Slide 11/15

Slide credit: Bob Fisher

## Threshold Selection 2

Assume 2 big peaks, brighter background is higher:

1. Find biggest peak (background)
2. Find next biggest peak in darker direction
3. Find lowest point in trough between peaks



Slide 12/15

Slide credit: Bob Fisher

## Peak Pick Code

Omit special cases for ends of array and closing 'end's.

```
peak = find(tmp1 == max(tmp1)); % find largest peak

% find highest peak to left
xmaxl = -1;
for i = 2 : peak-1
    if tmp1(i-1) < tmp1(i) & tmp1(i) >= tmp1(i+1) ...
        & tmp1(i) > xmaxl
        xmaxl = tmp1(i);
        pk1 = i;
    end
end

% find deepest valley between peaks
```

Slide 13/15

Slide credit: Bob Fisher

```
xminl = max(tmp1)+1;
for i = pk1+1 : peak-1
    if tmp1(i-1) > tmp1(i) & tmp1(i) <= tmp1(i+1) ...
        & tmp1(i) < xminl
        xminl = tmp1(i);
        thresh = i;
    end
end
```

Slide 14/15

Slide credit: Bob Fisher

## Lecture Overview

1. Thresholding to differentiate object from a constant and simple background (not just white backgrounds: see also *bluescreening* or *chroma keying*)
2. 1D Convolution
3. Histogram smoothing & threshold selection

Slide 15/15

Slide credit: Bob Fisher