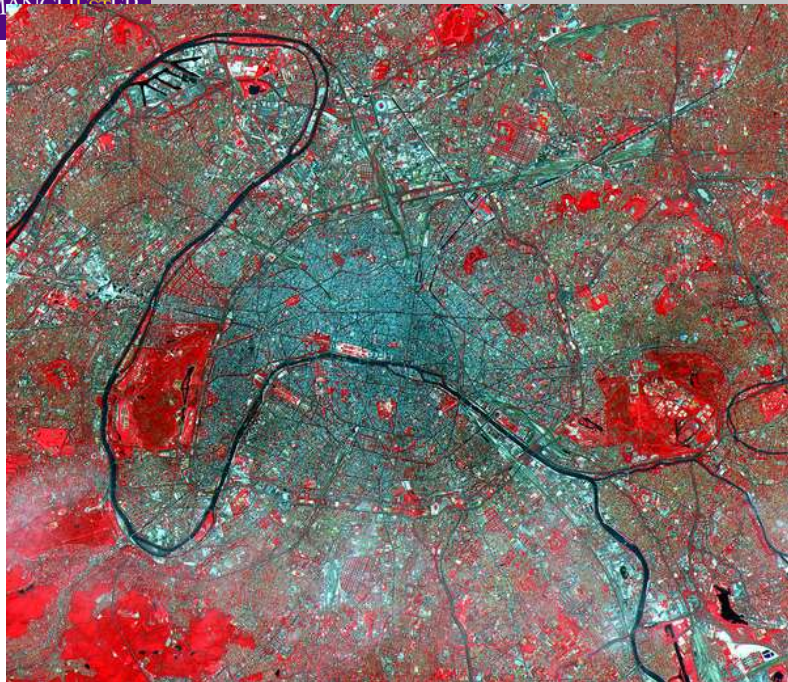


Computer Graphics and Image Processing

Lecture B4
Region Processing (2)



- n bands of input data
 - Within and beyond visible spectrum
- 3 bands of output data
 - An input band
 - Combine sets of input
 - Ratio of input bands
- Calibration with ground truth suggests what mappings are useful

Edge Definition

An edge is an extended, significant, local change in image intensity.

First Derivative, Gradient Edge Detection

- If an edge is a discontinuity
- Can detect it by differencing
- Convolve with appropriate templates
 - Suggestions?

Delta X and Delta Y

- Subtract horizontally adjacent pixels – Δx
- Subtract vertically adjacent pixels – Δy
- Can these be combined to give the correct edge strength?

Roberts Cross Edge Detector

-1	0
0	1

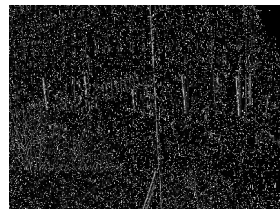
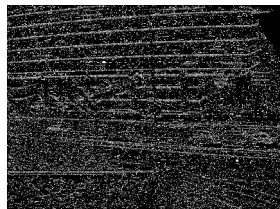
0	-1
1	0

- Simplest edge detector
- Awkward localisation
 - On the joint between the four pixels
- Noise sensitive
 - If one pixel is corrupted, the edge strength is equally corrupted

Prewitt/Sobel Edge Detector

-1	-1	-1
0	0	0
1	1	1

-1	0	1
-1	0	1
-1	0	1



Location and Noise

- Location estimate is at the centre pixel
- More robust against noise
 - Averaging of pixels either side of edge location
 - Noise magnitude reduced by $\sqrt{3}$ or $\sqrt{4}$

Edge Detection

- Combine horizontal and vertical edge estimates

$$Mag = \sqrt{h^2 + v^2}$$

$$\vartheta = \tan^{-1} \frac{v}{h}$$

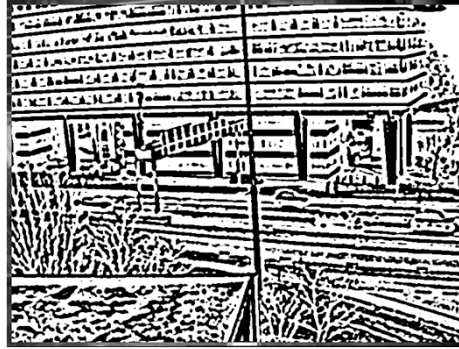
Problems

- Images are noise corrupted
 - Edges are noise corrupted
 - problems with detection and localisation
 - Can be improved by smoothing
- Scale
 - What is “local”?
 - Can be investigated by size of smoothing template

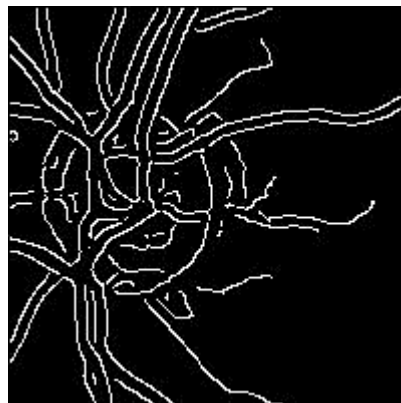
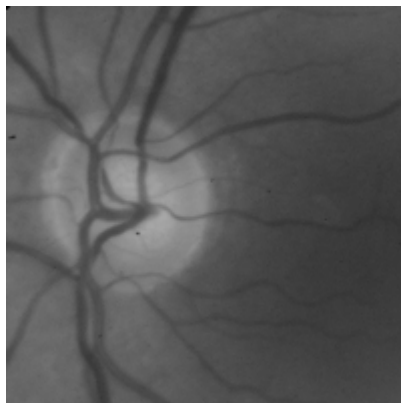
Canny/Deriche Edge Detector

- Require
 - edges to be detected
 - accurate localisation
 - single response to an edge
- Solution
 - Convolve image with Difference of Gaussian (DoG)
 - Template?

Example Results



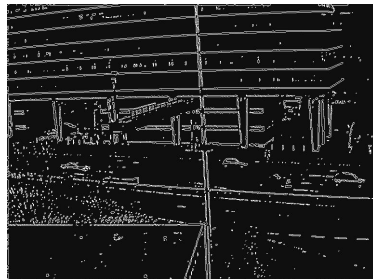
Canny applied to eye image



Second Derivative Operators Zero Crossing

- Models HVS
- Can locate edge to subpixel accuracy
- Convolve image with Laplacian of Gaussian (LoG)
 - Template?
- Edge location at crossing of zero axis

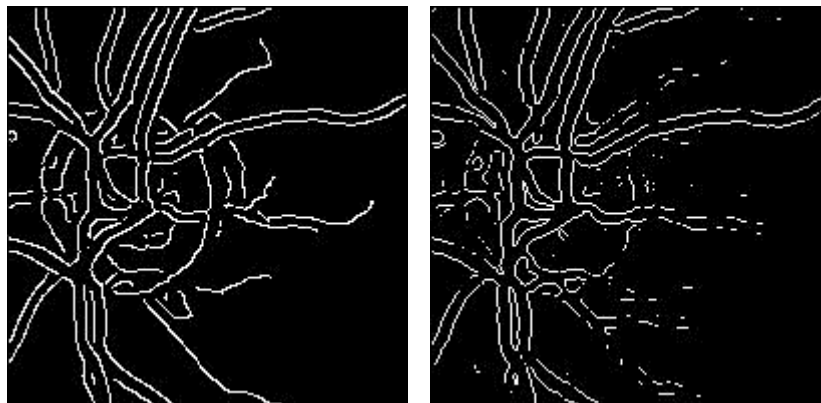
Example Results



LoG applied to eye image



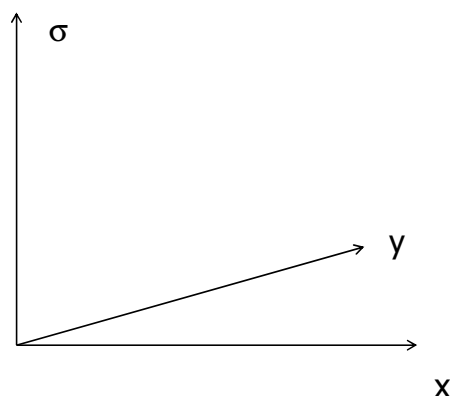
Comparison of Canny and LoG



Parameter Choices

- Width of Gaussian controlled by σ
- Large σ
 - More smoothing before edge detection
 - Small scale edges are blurred out
- What is a good value for σ ?

Scale Space





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Template matching

- Technique to measure similarities – hence find things
- Define a template
 - a model of the object to be recognised
- Define a measure of similarity
 - between template and similar sized image region

Aside: How to measure similarity and why use convolution

Measure dissimilarity between image $f[i,j]$ and
template $g[i,j]$

Place template on image and compare
corresponding intensities

Need a measure of dissimilarity

$$\max_{[i,j] \in R} |f - g| \quad \sum_{[i,j] \in R} |f - g| \quad \sum_{[i,j] \in R} (f - g)^2$$

Last is best....

...easiest to manipulate

Expanding

$$\sum_{[i,j] \in R} (f - g)^2 = \sum_{[i,j] \in R} f^2 + \sum_{[i,j] \in R} g^2 - 2 \sum_{[i,j] \in R} fg$$

If f and g fixed (is this reasonable?)

$-\sum fg$ a good measure of mismatch

$\sum fg$ a good measure of match

Compute match between template and image
with cross-correlation

$$M[i, j] = \sum_{k=-m}^{k=m} \sum_{l=-n}^{l=n} g[k, l] f[i + k, j + l]$$

Compare this to expression for convolution

g is constant, f varies and so influences M

Normalisation

$$C[i, j] = \frac{\sum_{k=-m}^{k=m} \sum_{l=-n}^{l=n} g[k, l] f[i + k, j + l]}{\sqrt{\left(\sum_{k=-m}^{k=m} \sum_{l=-n}^{l=n} f^2[i + k, j + l] \right)}}$$

C is maximum where f and g are same.

Limitations

- number of templates required
- rotation and size changes
- partial views

Input

Have begun with the Book of the Fall of Camelot.

When that April, with his quiver, rode,
The drapery of March, each jewel to the side,
And faded every scene in each corner,
Of which virtue engaged to the floor,
When April, too, with his quiver, rode,
The drapery of March, each jewel to the side,
And faded every scene in each corner,
Of which virtue engaged to the floor,
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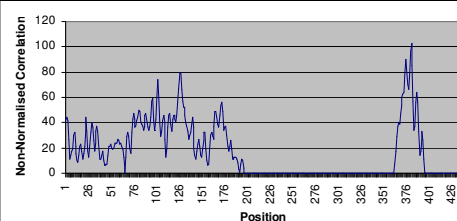
Template

t)

Output

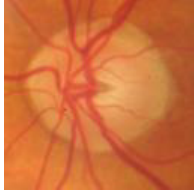
Have begun with the Book of the Fall of Camelot.


When that April, with his quiver, rode,
The drapery of March, each jewel to the side,
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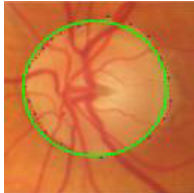
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Source


Texture based classification


Template


Result

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Summary

- Sections 3.3, 2.4, 5.5, 7.2

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Computers in the future may weigh no more
than 1 ½ tons
Popular Mechanics, 1949