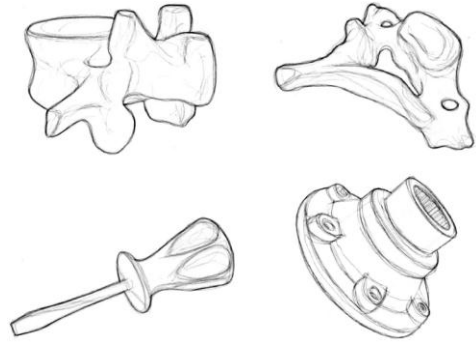


Edge Detection

CSE P 576

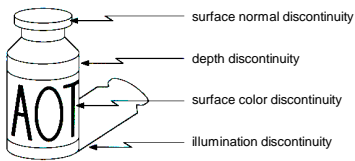
Larry Zitnick (larryz@microsoft.com)

What is an edge?



Cole et al. Siggraph 2008, results of 107 humans.

Origin of edges



Edges are caused by a variety of factors

Illusory contours

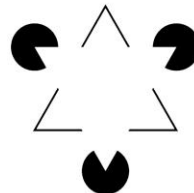


Image gradient



- The gradient of an image:

$$\nabla f = \left[\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right]$$

- The gradient points in the direction of most rapid change in intensity



$$\nabla f = \left[\frac{\partial f}{\partial x}, 0 \right]$$



$$\nabla f = \left[0, \frac{\partial f}{\partial y} \right]$$



$$\nabla f = \left[\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right]$$

Image gradient



$$\nabla f = \left[\frac{\partial f}{\partial x}, \frac{\partial f}{\partial y} \right]$$

$$\frac{\partial f}{\partial x} = f(x+1, y) - f(x, y)$$

How would you implement this as a filter?

The gradient direction is given by:

$$\theta = \tan^{-1} \left(\frac{\partial f / \partial y}{\partial f / \partial x} \right)$$

How does this relate to the direction of the edge?

The *edge strength* is given by the gradient magnitude

$$\|\nabla f\| = \sqrt{\left(\frac{\partial f}{\partial x}\right)^2 + \left(\frac{\partial f}{\partial y}\right)^2}$$

Sobel operator

In practice, it is common to use:

$$g_x = \begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix}$$

$$g_y = \begin{bmatrix} -1 & -2 & -1 \\ 0 & 0 & 0 \\ 1 & 2 & 1 \end{bmatrix}$$

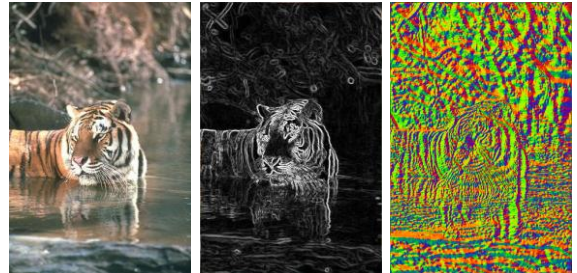
Magnitude:

$$g = \sqrt{g_x^2 + g_y^2}$$

Orientation:

$$\theta = \tan^{-1} \left(\frac{g_y}{g_x} \right)$$

Sobel operator



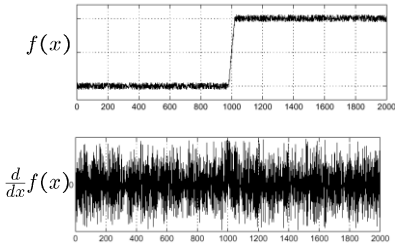
Original

Magnitude

Orientation

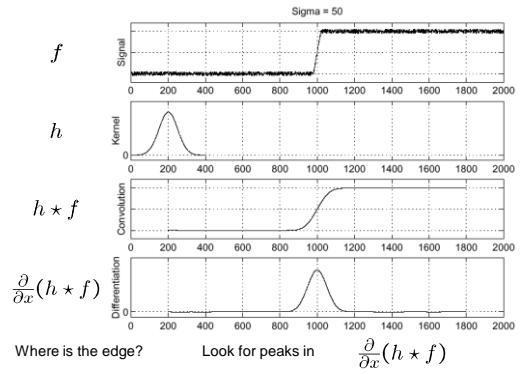
Effects of noise

- Consider a single row or column of the image
 - Plotting intensity as a function of position gives a signal

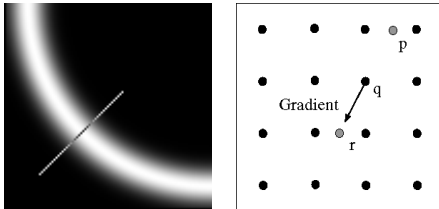


Where is the edge?

Solution: smooth first



Non-maximum suppression



- Check if pixel is local maximum along gradient direction
 - requires checking interpolated pixels p and r

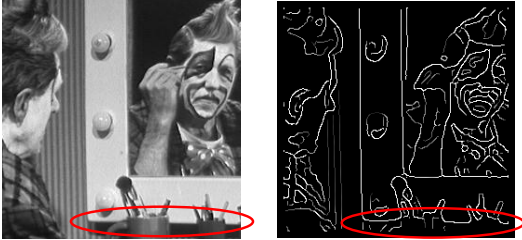
Effect of σ (Gaussian kernel spread/size)



original Canny with $\sigma = 1$ Canny with $\sigma = 2$

- The choice of σ depends on desired behavior
- large σ detects large scale edges
 - small σ detects fine features

An edge is not a line...

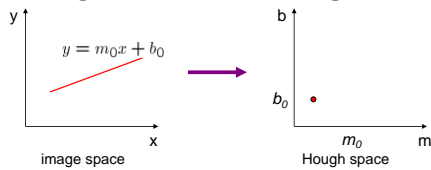


How can we detect *lines* ?

Finding lines in an image

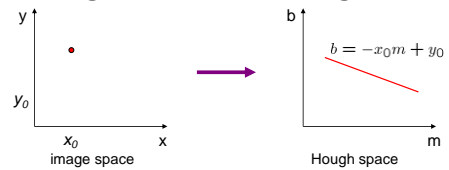
- Option 1:
 - Search for the line at every possible position/orientation
 - What is the cost of this operation?
- Option 2:
 - Use a voting scheme: Hough transform

Finding lines in an image



- Connection between image (x,y) and Hough (m,b) spaces
 - A line in the image corresponds to a point in Hough space
 - To go from image space to Hough space:
 - given a set of points (x,y) , find all (m,b) such that $y = mx + b$

Finding lines in an image



- Connection between image (x,y) and Hough (m,b) spaces
 - A line in the image corresponds to a point in Hough space
 - To go from image space to Hough space:
 - given a set of points (x,y) , find all (m,b) such that $y = mx + b$
 - What does a point (x_0, y_0) in the image space map to?
 - A: the solutions of $b = -x_0m + y_0$
 - this is a line in Hough space

Hough transform algorithm

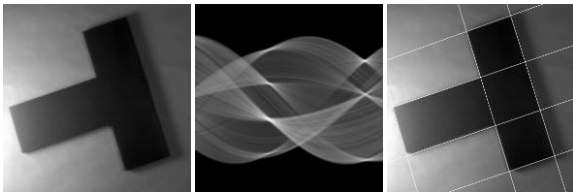
- Typically use a different parameterization

$$d = x \cos \theta + y \sin \theta$$
 - d is the perpendicular distance from the line to the origin
 - θ is the angle
 - Why?

Hough transform algorithm

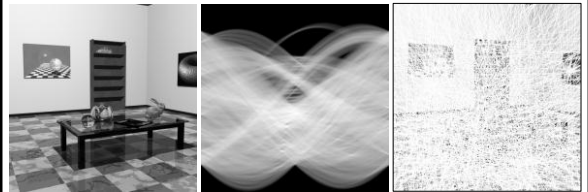
- Basic Hough transform algorithm
 1. Initialize $H[d, \theta] = 0$
 2. for each edge point $I[x, y]$ in the image
 for $\theta = 0$ to 180
 $d = x \cos \theta + y \sin \theta$
 $H[d, \theta] += 1$
 3. Find the value(s) of (d, θ) where $H[d, \theta]$ is maximum
 4. The detected line in the image is given by $d = x \cos \theta + y \sin \theta$
- What's the running time (measured in # votes)?

Hough transform algorithm



<http://www.cs.utah.edu/~vpegorar/courses/cs7966/>

Hough transform algorithm



<http://www.cs.utah.edu/~vpegorar/courses/cs7966/>

Extensions

- Extension 1: Use the image gradient
 1. same
 2. for each edge point $I[x,y]$ in the image
 - compute unique (d, θ) based on image gradient at (x,y)
 - $H[d, \theta] += 1$
 3. same
 4. same
- What's the running time measured in votes?
- Extension 2
 - give more votes for stronger edges
- Extension 3
 - change the sampling of (d, θ) to give more/less resolution
- Extension 4
 - The same procedure can be used with circles, squares, or any other shape