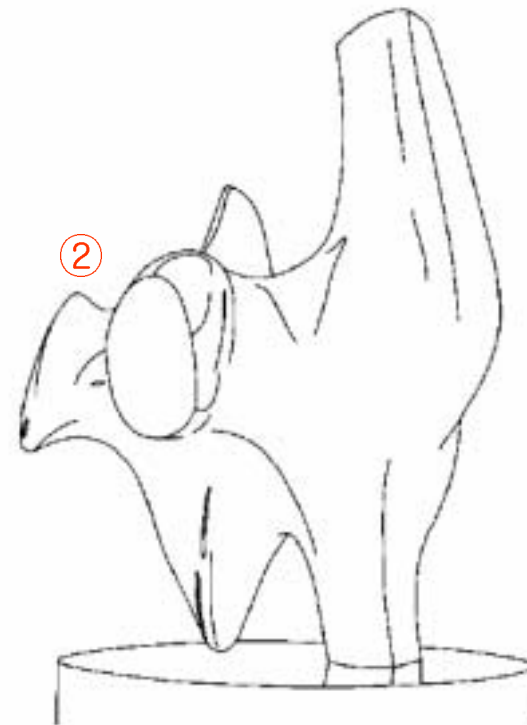


Detecting Edges

 Hoon Yoo, Ph.D.

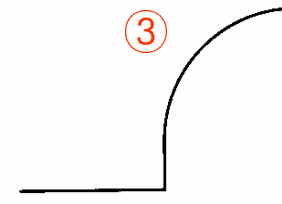
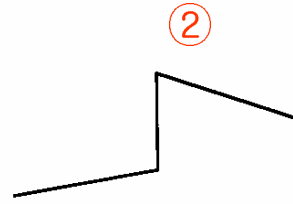
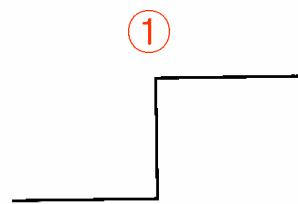
Edge detection

- Convert a 2D image into a set of curves
 - Extracts salient features of the scene
 - More compact than pixels

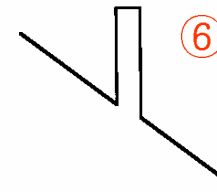
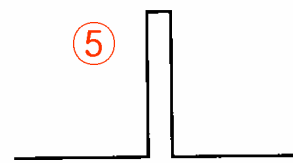
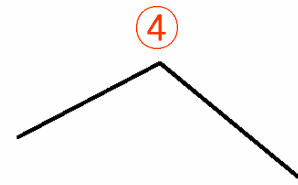


Profiles of image intensity edges

- How can you tell that a pixel is on an edge?



Step Edges

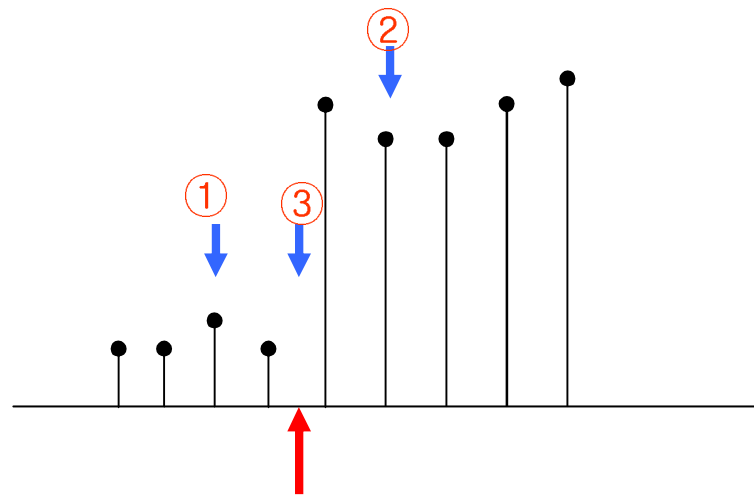


Roof Edge

Line Edges

Edge is Where Change Occurs

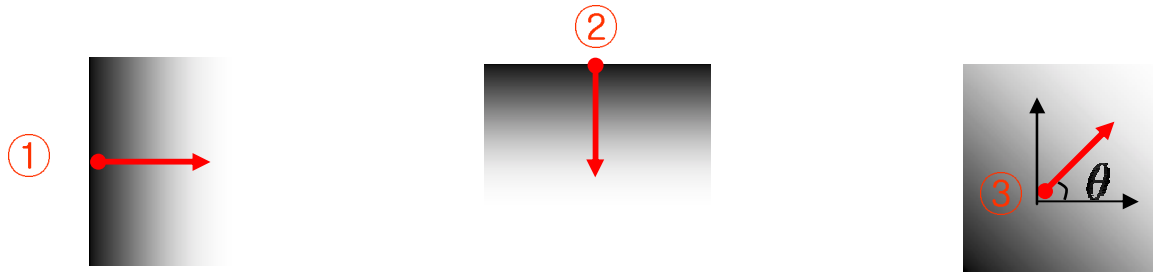
- Change is measured by derivative in 1D
- Biggest change, derivative has maximum magnitude
- Or 2nd derivative is zero.



Edge = Large change

Derivative = Difference between pixels

Image gradient



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

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

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

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

- The gradient direction is given by:

$$\textcircled{4} \quad \theta = \tan^{-1} \left(\frac{\partial f / \partial y}{\partial f / \partial x} \right)$$

– how does this relate to the direction of the edge?

- The *edge strength* is given by the gradient magnitude

$$\textcircled{5} \quad \|\nabla f\| = \sqrt{\left(\frac{\partial f}{\partial x} \right)^2 + \left(\frac{\partial f}{\partial y} \right)^2}$$

The discrete gradient

- How can we differentiate a *digital* image $f[x,y]$?

$$\frac{\partial f}{\partial x}[x, y] \approx \underset{\textcircled{1}}{f[x + 1, y]} - \underset{\textcircled{2}}{f[x, y]}$$

High Pass Filter

The Sobel operator

- Better approximations of the derivatives exist
 - The *Sobel* operators below are very commonly used

①

$$\frac{1}{8} \begin{bmatrix} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{bmatrix}$$

s_x

②

$$\frac{1}{8} \begin{bmatrix} 1 & 2 & 1 \\ 0 & 0 & 0 \\ -1 & -2 & -1 \end{bmatrix}$$

s_y

Gradient operators

$$\begin{array}{cc} \Delta_1 & \Delta_2 \\ \begin{array}{cc} 0 & 1 \\ -1 & 0 \end{array} & \begin{array}{cc} 1 & 0 \\ 0 & -1 \end{array} \end{array}$$

(a)

$$\begin{array}{cc} \Delta_1 & \Delta_2 \\ \begin{array}{ccc} -1 & 0 & 1 \\ -1 & 0 & 1 \\ -1 & 0 & 1 \end{array} & \begin{array}{ccc} 1 & 1 & 1 \\ 0 & 0 & 0 \\ -1 & -1 & -1 \end{array} \end{array}$$

(b)

$$\begin{array}{cc} \Delta_1 & \Delta_2 \\ \begin{array}{ccc} -1 & 0 & 1 \\ -2 & 0 & 2 \\ -1 & 0 & 1 \end{array} & \begin{array}{ccc} 1 & 2 & 1 \\ 0 & 0 & 0 \\ -1 & -2 & -1 \end{array} \end{array}$$

(c)

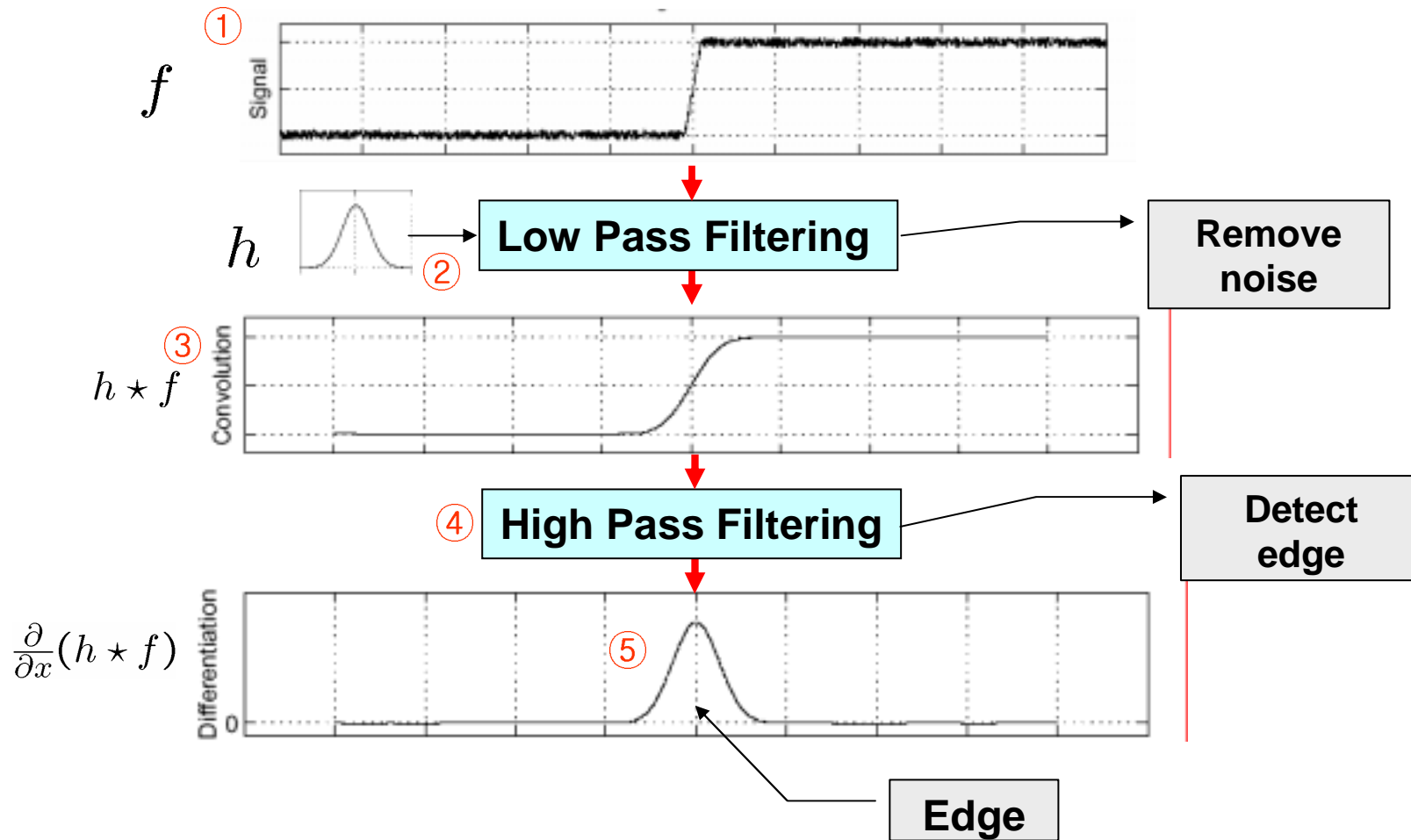
$$\begin{array}{cc} \Delta_1 & \Delta_2 \\ \begin{array}{cccc} -3 & -1 & 1 & 3 \\ -3 & -1 & 1 & 3 \\ -3 & -1 & 1 & 3 \\ -3 & -1 & 1 & 3 \end{array} & \begin{array}{cccc} 3 & 3 & 3 & 3 \\ 1 & 1 & 1 & 1 \\ -1 & -1 & -1 & -1 \\ -3 & -3 & -3 & -3 \end{array} \end{array}$$

(d)

(a): Roberts' cross operator (b): 3x3 Prewitt operator
(c): Sobel operator (d) 4x4 Prewitt operator

Solution: smooth first

- Where is the edge?



Optimal Edge Detection: Canny

- Assume:
 - Linear filtering
 - Additive iid Gaussian noise
- Edge detector should have:
 - Good Detection. Filter responds to edge, not noise.
 - Good Localization: detected edge near true edge.
 - Single Response: one per edge.

Summary

- Define Edges
- Edge detector
 - Low pass filter + High pass filter