

CS 369: Introduction to Robotics

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Spring 2026



Haverford
COLLEGE

Admin

- Lab 2 grades posted on Moodle

Outline for today

- Robot perception
- Computer vision

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- Computer vision

Robot sensors



Gyroscope



Lever Switch



Linear Encoder



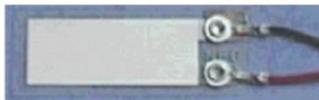
GPS



Camera



Laser Rangefinder



Piezo Bend



Rotary Encoder



Accelerometer



Sonar Ranging



PIR



Resistive Bend



Metal Detector



Pendulum Resistive Tilt



Gas



Pyroelectric Detector



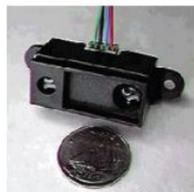
Pressure



Resistive Bend



UV Detector



Infrared Ranging



CDS Cell



Compass



Radiation



Magnetometer



IR Modulator Receiver



Microphone



Magnetic Reed Switch

Sensor data

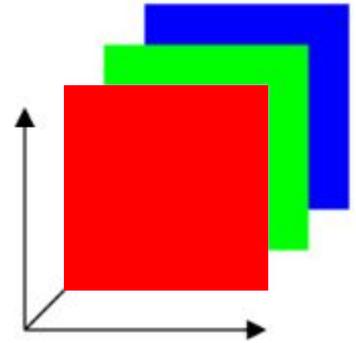
- Audio
- Distance
- Tactile
- Visual
- etc.



Images

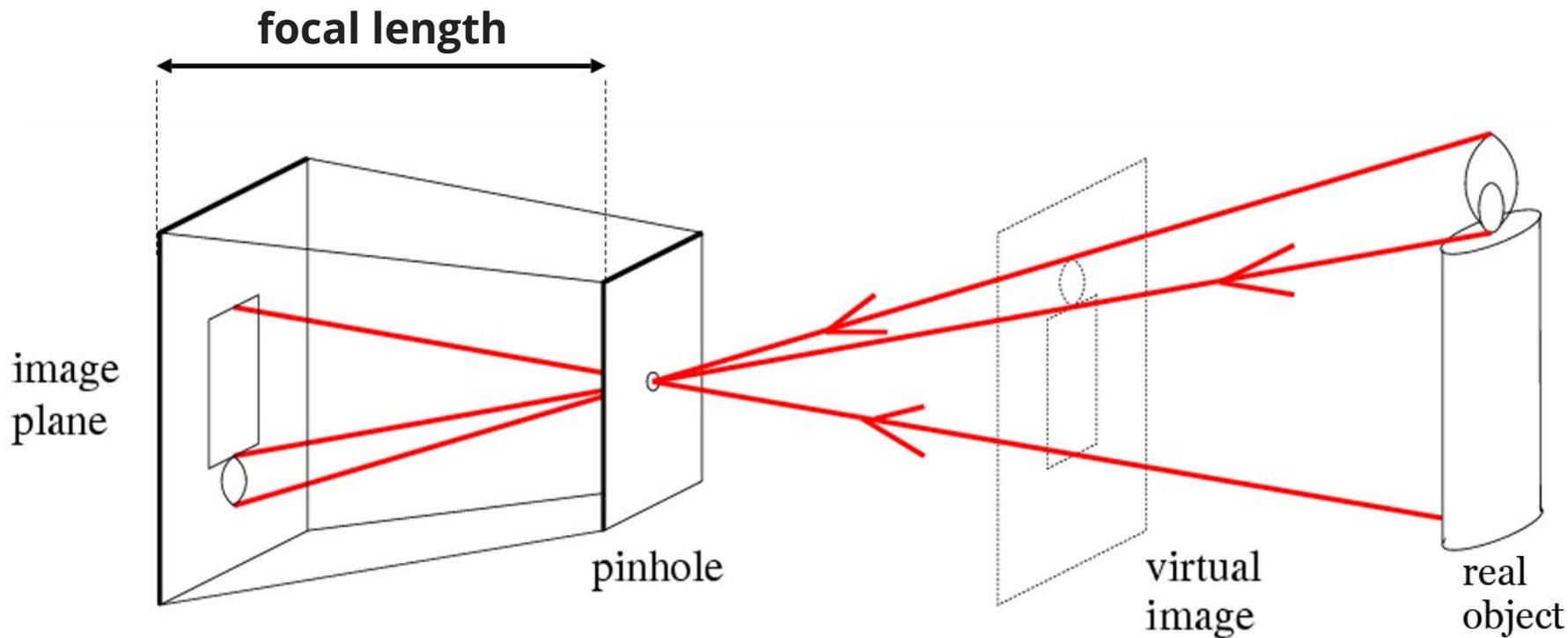
- Discrete representation of a continuous function
 - An image is a two dimensional array of pixels
 - Pixel: cell with numeric value representing a uniform portion of an image
- Resolution
 - Number of pixels across the horizontal dimension
 - Number of pixels in the vertical dimension
 - Number of layers used for color

Images

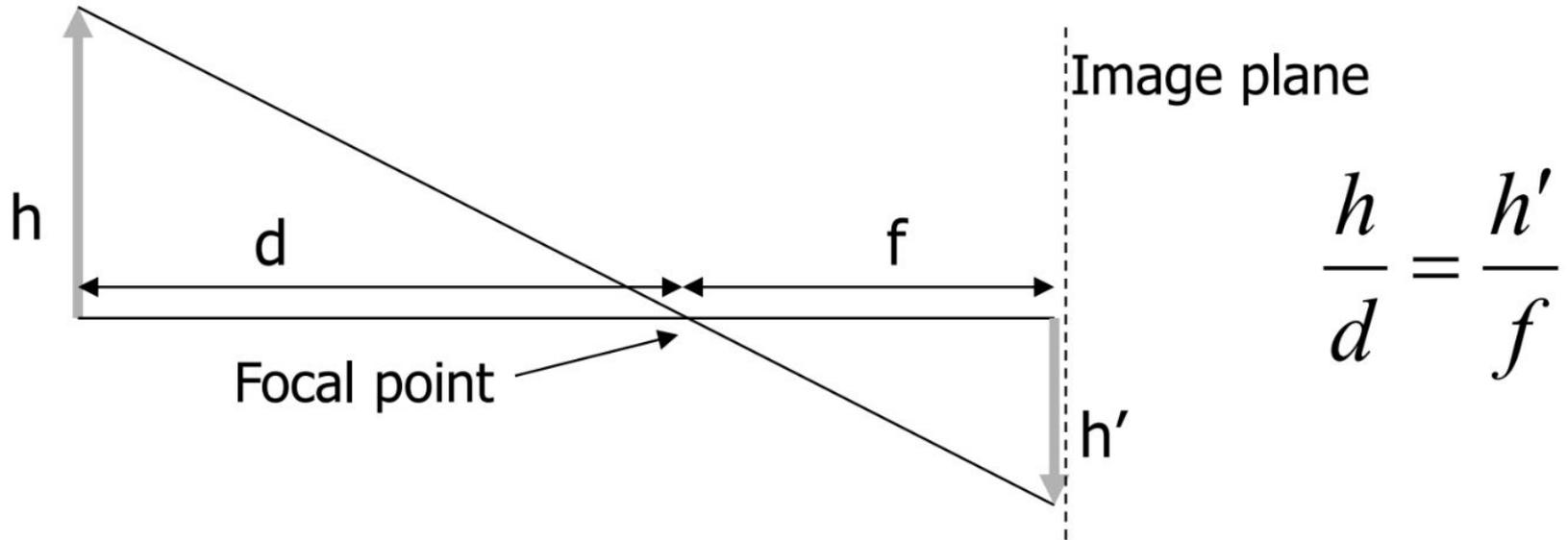


- Binary: two color image
 - Pixel is only one bit of information
 - Indicates if the intensity of color is above or below some nominal value
- Grayscale
 - All pixels represent the intensity of light in an image
 - Intensity of light in a pixel is stored as a number, generally 0 to 255 inclusive
- Color
 - Three grayscale images layered on top of each other with each layer indicating the intensity of a specific color light — generally red, green, and blue (RGB)
 - Third dimension in a digital image

Pinhole camera model



Projection on the image plane



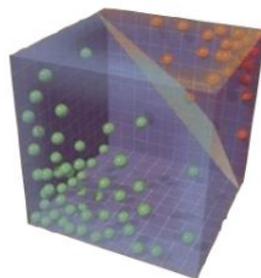
Outline for today

- Robot perception
- Computer vision

Computer vision

- Making sense of visual data
- Typical pipeline:

representation



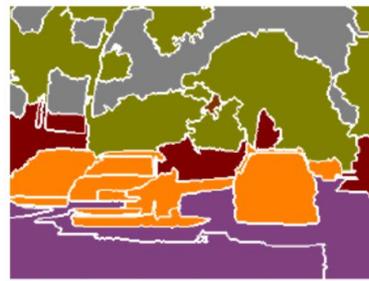
‘fancy math’



output



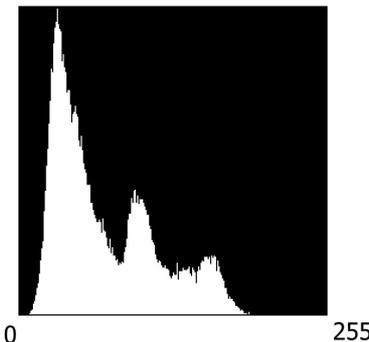
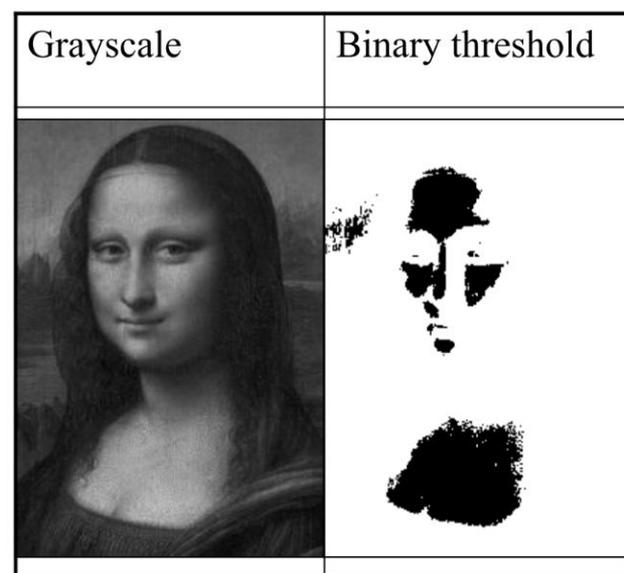
What should we look at?
(e.g. image features)



What can we understand?
(e.g. semantic segmentation)

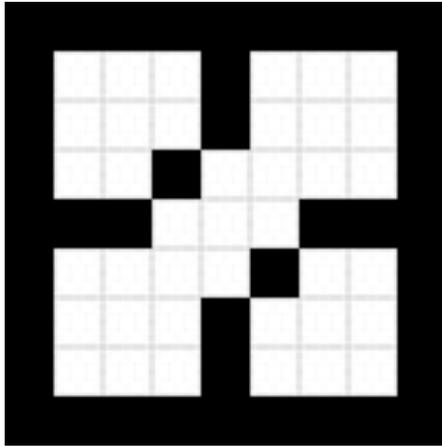
Thresholding

- Purpose
 - Find areas of high color intensity
 - Highlight locations of different features of the image
 - Image compression, use fewer bits to encode a pixel
- Process
 - Decide on a threshold value μ
 - Scan every pixel in the image
 - If it is greater than or equal to μ , make it 1
 - If it is less than μ , make it 0
 - Picking a good μ
 - Often 128 is a good value to start with
 - Use a histogram to determine values based on color frequency features

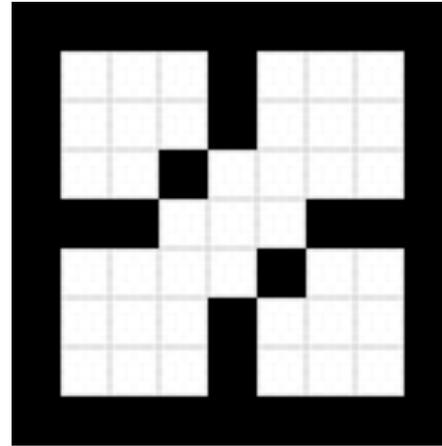


Segmentation: flood fill

Algorithm that determines and alters the area connected to a given node with some matching attribute.



Flood fill with 4-connectivity



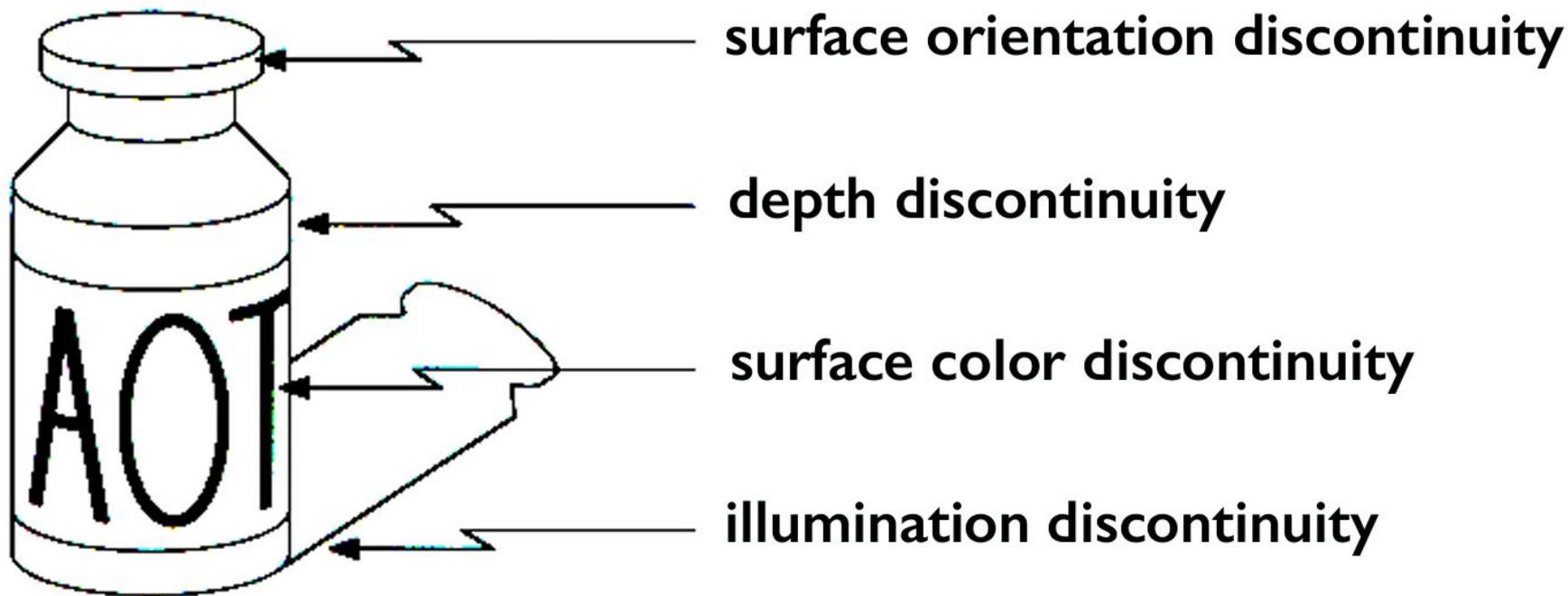
Flood fill with 8-connectivity

Edge detection

The process of identifying parts of a digital image with sharp changes (discontinuities) in image intensity.



Causes of edges



Characterizing edges

An edge is a place of rapid change in the image intensity function.

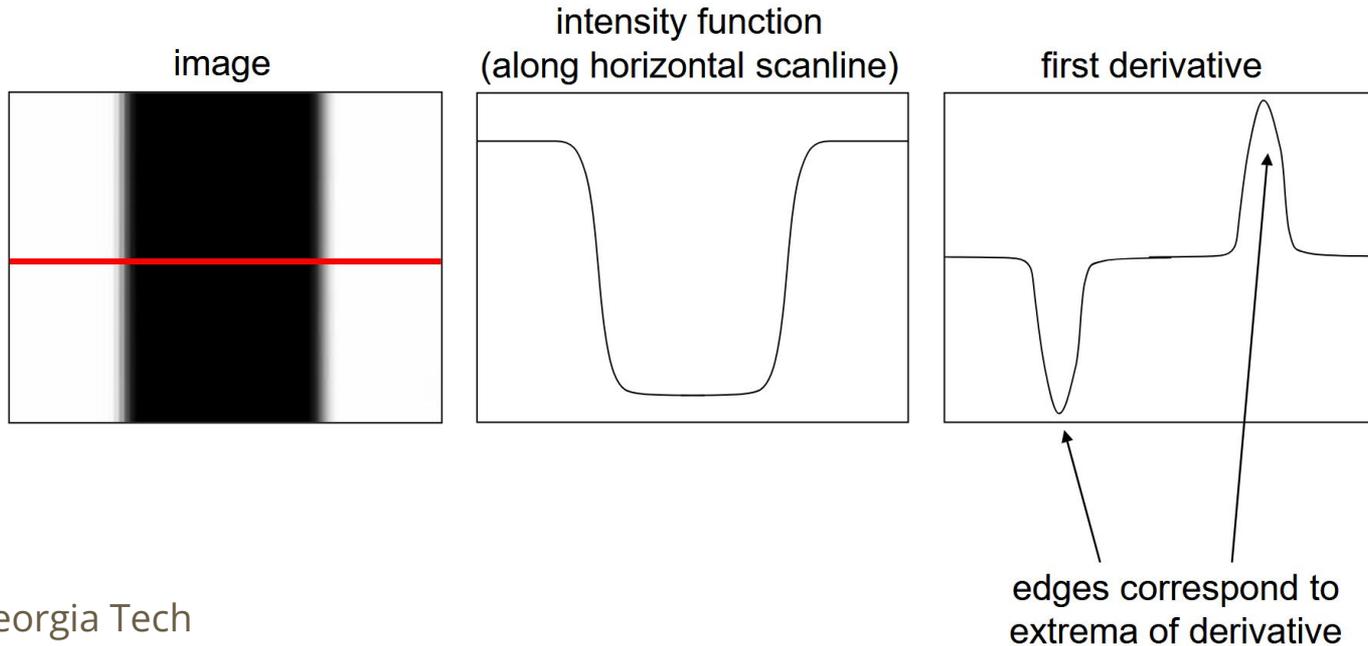
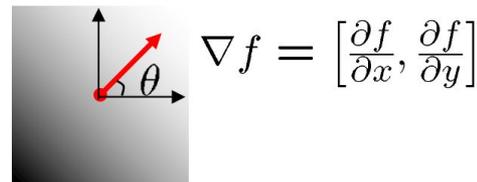
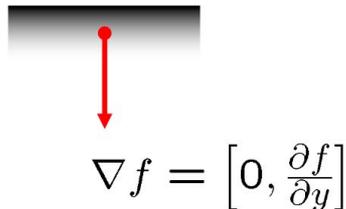
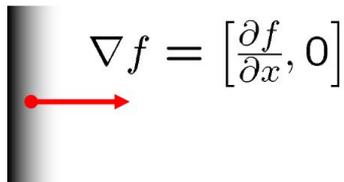


Image gradient

- Take discrete derivative (finite difference): $\frac{\partial f}{\partial x}[x, y] \approx F[x + 1, y] - F[x, y]$
- 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 increase in intensity



The gradient direction is given by: $\theta = \tan^{-1} \left(\frac{\partial f}{\partial y} / \frac{\partial f}{\partial x} \right)$

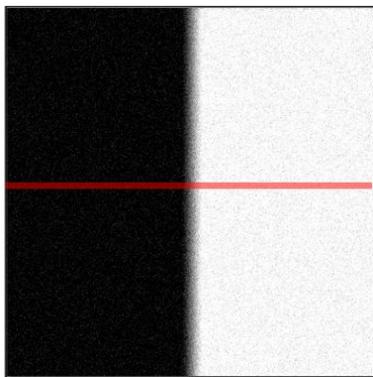
- how does this relate to the direction of the edge?

The *edge strength* is given by the gradient magnitude: $\|\nabla f\| = \sqrt{\left(\frac{\partial f}{\partial x}\right)^2 + \left(\frac{\partial f}{\partial y}\right)^2}$

Differentiation and convolution

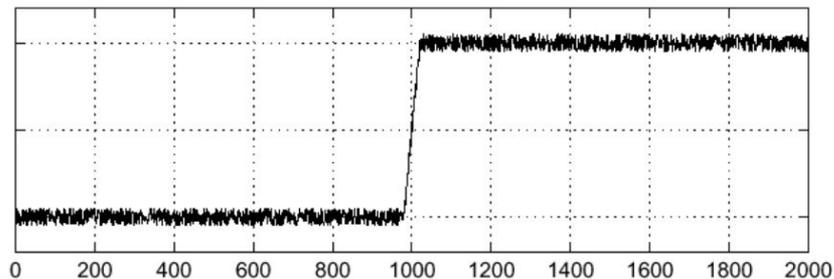
- Discrete derivative: $\frac{\partial f}{\partial x}[x, y] \approx F[x + 1, y] - F[x, y]$
- This is the dot product of the vector $[-1 \ 1]$ with each pixel in the image
$$[F[x,y] \ F[x+1,y]] \cdot [-1 \ 1]$$
- The operation of sliding a filter across an image is called convolution
- At the image boundary:
 - Extend the image using the pixel values at the boundary
 - Ignore pixels where the filter would extend beyond the image boundary
- Two-dimensional convolution: sum of element-wise products

Effects of noise

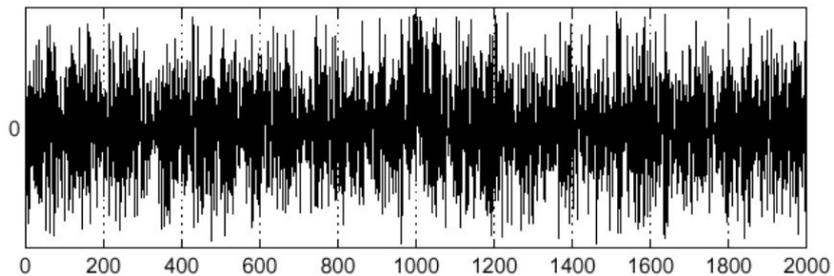


Noisy input image

$$f(x)$$

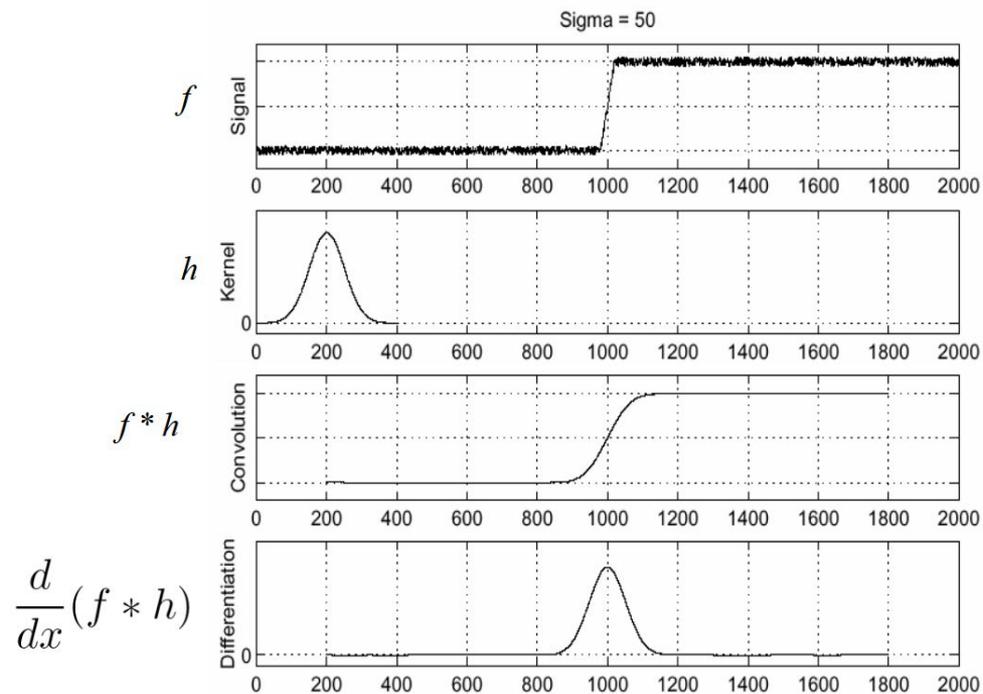


$$\frac{d}{dx} f(x)$$



Where is the edge?

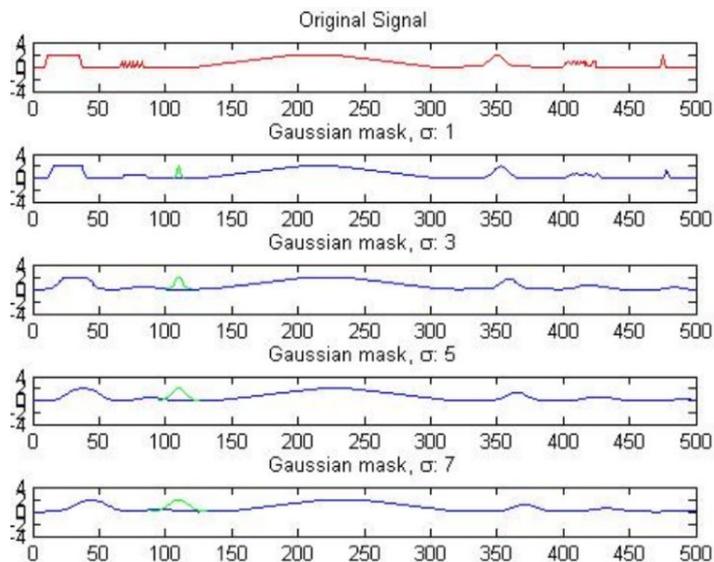
Solution: smoothing filter



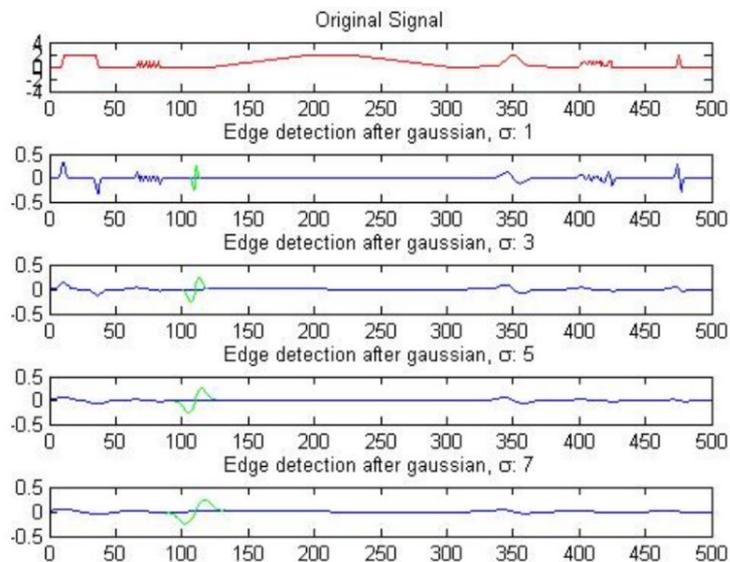
To find edges, look for peaks in $\frac{d}{dx}(f * h)$

Mask size

Wider masks lead to uncertainty about location of edge.



Convolution with Gaussian distribution



Edge detection on a signal that was convolved with a Gaussian

Canny edge detector

J. Canny, [A Computational Approach To Edge Detection](#),
IEEE Trans. Pattern Analysis and Machine Intelligence,
8:679-714, 1986

- Classic and very accurate edge detector
- Process
 - Filter image with derivative of Gaussian
 - Find magnitude and orientation of gradient
 - Non-maximum suppression: thin multi-pixel wide “ridges” down to single pixel width
 - Linking and thresholding (hysteresis):
 - Define two thresholds: low and high
 - Use the high threshold to start edge curves and the low threshold to continue them

Hysteresis thresholding



original image



high threshold
(strong edges)



low threshold
(weak edges)



hysteresis threshold