

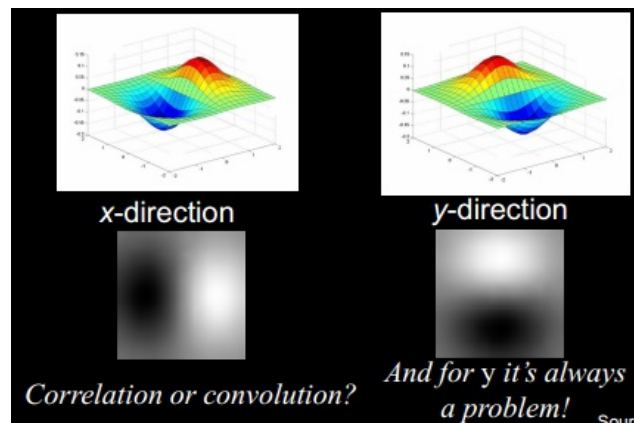
2A-L6 Edge detection - 2D

2017/11/11 13:41

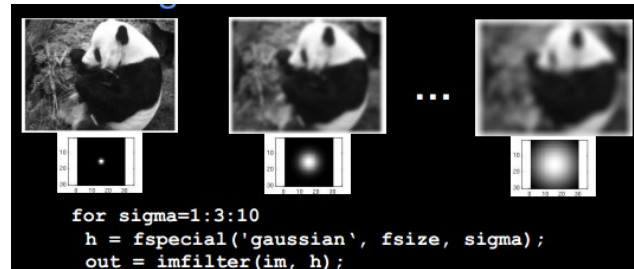
1. SUM
 - a. the procedure to find the edge
 - i. smooth the filter
 1. sigma effects the final egde, the bigger, the more significant
 - ii. gradient the filter
 1. Laplacian for 2D
 - iii. apply the filter to the image to get the mag and dir of gradient
 - iv. threshold
 - v. thin
 - vi. connect
 - b. filters
 - i. canny
 1. key ideas
 - a. thinning
 - b. connecting
 2. `edge(img, 'canny')`
 3. works better than sobel and others
 - c. edge detector matlab
 - i. `edge(img, 'type', threshold, direction, thinning)`
2. Intro
 - a. as mentioned before, we can compute the derivative of the filter first and then apply it to the images
 - b. reasons
 - i. Since h is typically smaller we take fewer derivatives so it's faster.
 - ii. The smoothed derivative operator is computed once and you have it to use repeatedly.
 - c. why can we do this

$$(I \otimes g) \otimes h_x = I \otimes (g \otimes h_x)$$

3. How to choose the size of Gaussian Filter 2D



- a. it's the correlation, ATT to the positive direction

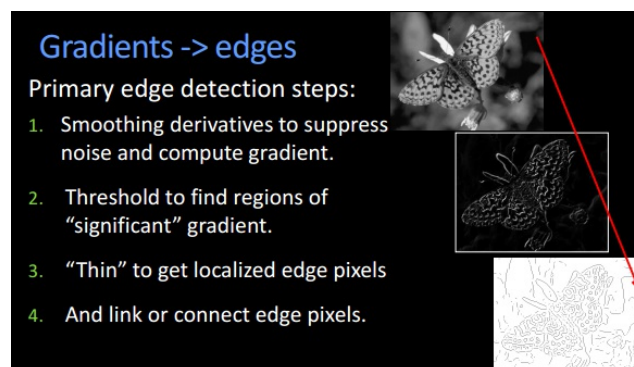


- b. the bigger size of Gaussian filter, the stronger the smoother and thus the sharper edge remains and the shallow edges are just smooth out.

- i. Smaller values: finer features detected
- ii. Larger values: larger scale edges detected

1. How to find the edges

- a. Primary edge detection steps:
 - i. smoothing the derivatives of filters to suppress noise and compute gradient
 - ii. threshold to find regions of "significant" gradients
 - iii. "thin" to get localized edge pixels
 - iv. link or connect edge pixels



- b. MATLAB: `edge(image, 'canny')`

1.

[https://de.mathworks.com/help/images/ref/edge.html?](https://de.mathworks.com/help/images/ref/edge.html?searchHighlight=edge&s_tid=doc_srchttitle)

[searchHighlight=edge&s_tid=doc_srchttitle](https://de.mathworks.com/help/images/ref/edge.html?searchHighlight=edge&s_tid=doc_srchttitle)

2. `BW = edge(I)`

- a. returns a binary image BW

containing **1s** where the function finds **edges** and **0s elsewhere**.

b. The input image **I** is an **intensity or a binary image**.

c. **BW** is the same size as **I**.

3. **BW** = `edge(I, 'canny')`

a. By default, `edge` uses the Sobel `edge` detection method, but we can specify Canny (or a Canny approximation), Laplacian of Gaussian (log), Prewitt, Roberts, or Zero-crossings.

b. **canny** is better than **sobel**

4. **BW** = `edge(I, 'canny', threshold)`

a. return all **edges** that are stronger than `threshold`. if not specified, `edge` chooses the value automatically.

5. **BW** =

`edge(I, 'canny', threshold, direction)`

a. specify the direction in which the function looks for **edges** in the image: 'horizontal', 'vertical', or 'both'.

6. **BW** =

`edge(I, 'canny', threshold, direction, 'nothinning')`

a. 'nothinning' specify to skip the additional **edge**-thinning stage, The default value is 'thinning'.

7. [**BW**, **threshOut**] =

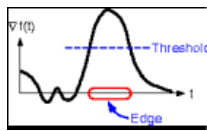
`edge(I, 'canny', ____)`

a. **eturns** the threshold value.

1. Canny Edge Detector

a. **filter derivative**: Filter image with derivative of Gaussian

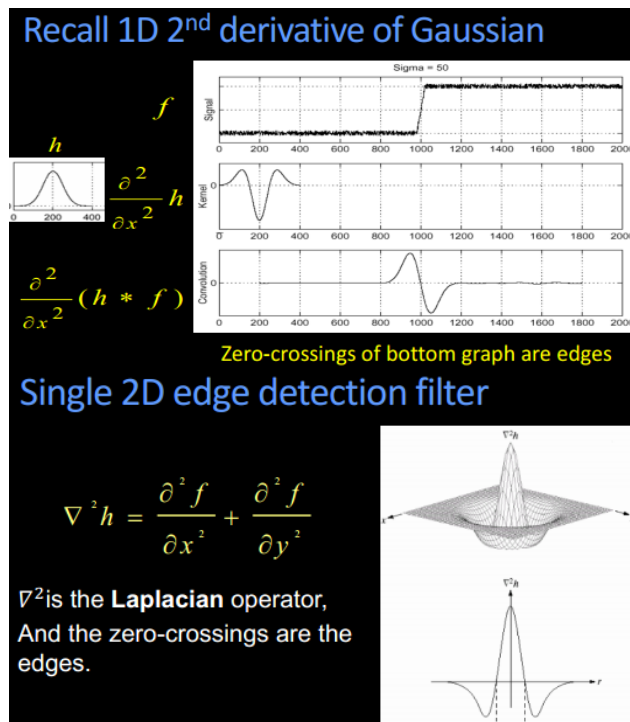
- b. **gradient**: Find magnitude and orientation of gradient
- c. **thresholding**: to keep the 'significant' edges
- d. **Non-maximum suppression**: -- thinning
 - i. Thin multi-pixel wide "ridges" down to single pixel width
 - ii. Check and only keep the pixel with local maximum along gradient direction
- e. Linking and thresholding (hysteresis):
 - i. Core idea: Define two thresholds: low and high. Use the high threshold to start edge curves and the low threshold to continue them
 - ii. procedure
 1. Apply a high threshold to detect strong edge pixels.



- a. problem: some pixel along the edge can't survive the thresholding

1. Link those strong edge pixels to form strong edges.
2. Apply a low threshold to find weak but plausible edge pixels.
3. Extend the strong edges to follow weak edge pixels
 - a. the assumption here: the important edges have some strong gradient pixels. thus we only keep the edges with strong-edge pixels

1. Canny Results
 - a. with canny, we get better results than with others
 - b. but actually, it's hard to know when an edge image is good since it has the meaning when it is put into the specific application domain
 - c. size of sigma influence the granurity of the edges
2. Quiz: Canny Edge Quiz
 - a. The Canny edge operator is probably quite sensitive to noise
 - i. Mostly false - it depends upon the σ chose
3. Single 2D Edge Detection Filter
 - a. we use the Laplacian operator for the 2nd derivative for 2D



9. Side

a. image types

- i. A binary image is a digital image that has only two possible values for each pixel.
- ii. An **intensity image** is a data matrix, I , whose values represent intensities within some range, e.g. $[0, 255]$. with each element of the matrix corresponding to one image pixel

1. grayscale image

iii. **Indexed Images**

1. An indexed image consists of a data matrix, X , and a colormap matrix, map . map is an m -by-3 array of class `double` containing floating-point values in the range $[0, 1]$.
2. Each row of map specifies the red, green, and blue components of a single color

iv. **RGB (Truecolor) Images**

1. stored as an m -by- n -by-3 data array that defines red, green, and blue color components for each individual pixel.
2. Graphics file formats store RGB images as 24-bit images, where the red, green, and blue components are 8 bits each.
3. can be of class `double`, `uint8`, or `uint16`.