

tags: DIP

# Principles and Applications of Digital Image Processing

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Fall, 2021

for better reading: <https://hackmd.io/@tohow06/BJsMY6nIY>

## Homework 3

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### Part 1:

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#### 3.22

(a)

$$\omega = vw^T = \begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix} [2 \ 1 \ 1 \ 3] = \begin{bmatrix} 2 & 1 & 1 & 3 \\ 4 & 2 & 2 & 6 \\ 2 & 1 & 1 & 3 \end{bmatrix}$$

Yes, it is separable. It can be separated to two column vectors' (v and w) outer product.

(b)

$$\begin{bmatrix} 1 & 3 & 1 \\ 2 & 6 & 2 \end{bmatrix} = \begin{bmatrix} 1 \\ 2 \end{bmatrix} [1 \ 3 \ 1]$$

$$w_1 = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \quad w_2 = \begin{bmatrix} 1 \\ 3 \\ 1 \end{bmatrix}$$

#### 3.27

(a)

According (3-38), (3-39) in textbook, the single kernel size is

$$W_v = W_h = 4 \times (3 - 1) + 3 = 11$$

kernel size is 11x11.

**(b)**

According Table 3.6, standard deviation of convolution of two Gaussian functions is

$$\sigma = \sqrt{\sigma_1^2 + \sigma_2^2}$$

So the standard deviation of the single kernel will be

$$\sigma = \sqrt{1^2 + 1^2 + 1^2 + 1^2} = 2$$

## 3.38

If the order is reversed, the detail and the noise will be enhanced first by Laplacian kernel, then the smoothing kernel can only blur all the detail and noise. With the first order we can get clean and fine detailed image, however, with the reversed order, we get blurred image with somewhat mess.

## 4.3

**(a)**

Convolution in the frequency domain is analogous to multiplication in the spatial domain, so I first transform two functions

$$\begin{aligned}\mathcal{J}\{\delta(t)\} &= 1 \\ \mathcal{J}\{\delta(t - t_0)\} &= e^{-j2\pi\mu t_0}\end{aligned}$$

Because the first one equals 1, the result of convolution is  $\delta(t - t_0)$ .

**(b)**

Same process with (a)

$$\begin{aligned}\mathcal{J}\{\delta(t + t_0)\} &= e^{j2\pi\mu t_0} \\ \mathcal{J}\{\delta(t - t_0)\} &= e^{-j2\pi\mu t_0} \\ e^{j2\pi\mu t_0} \times e^{-j2\pi\mu t_0} &= 1 \\ \mathcal{J}^{-1}\{1\} &= \delta(t)\end{aligned}$$

The result of convolution is  $\delta(t)$ .

## 4.32

**(a)**

Origin sequence of array is even. Insert 0's in center of array become

$$\{a, b, c, 0, 0, 0, 0, 0, 0, 0, 0, 0, c, b\}$$

The new array still has evenness.

**(b)**

Origin sequence of array is odd. Insert 0's in center of array become

$$\{0, -b, -c, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, c, b\}$$

The new array still has oddness.

**(c)**

Origin sequence of array is even. Insert 0's become

$$\{a, b, c, 0, 0, 0, 0, 0, d, 0, 0, 0, 0, 0, 0, c, b\}$$

The new array still has evenness.

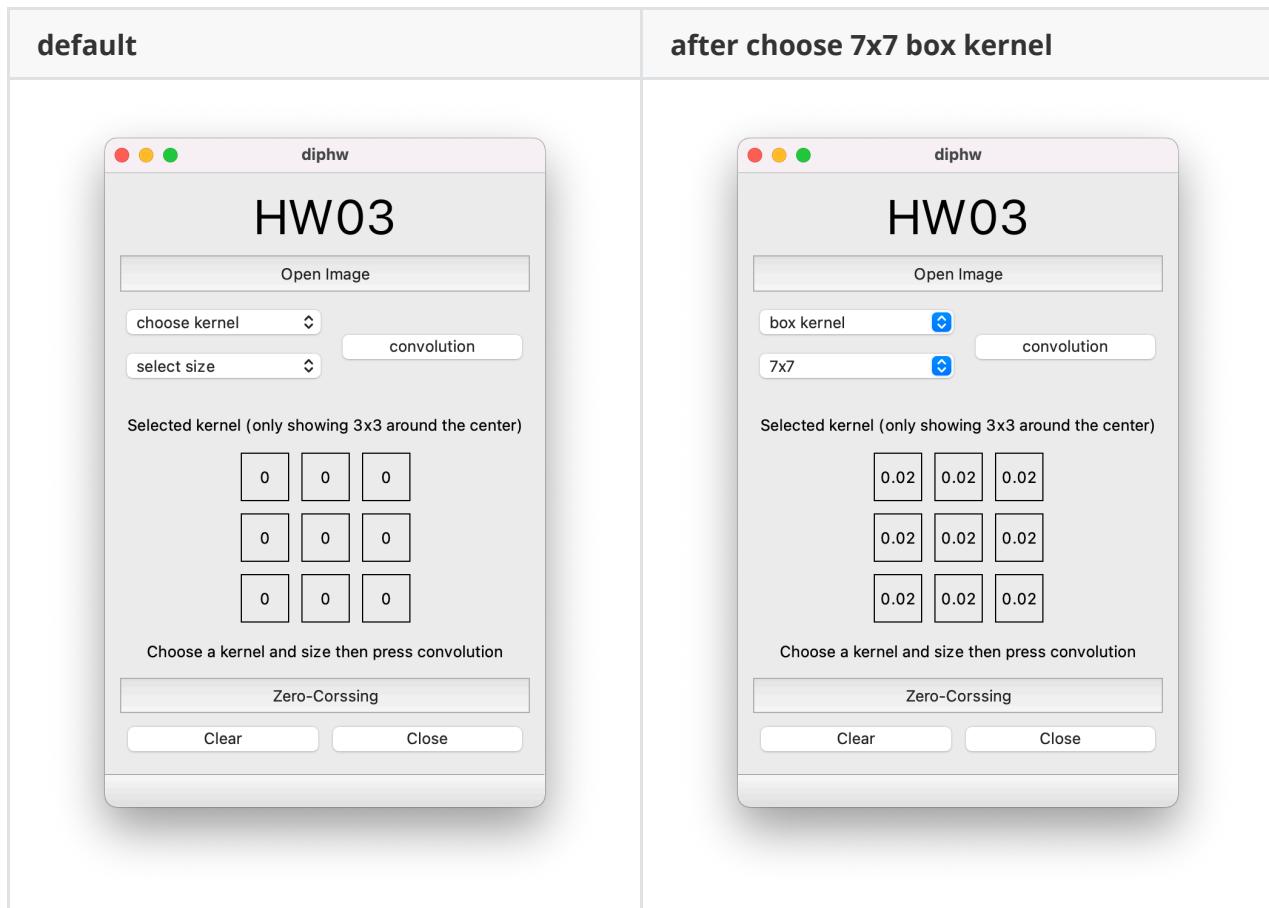
**(d)**

Origin sequence of array is odd. Insert 0's in center of array become

$$\{0, -b, -c, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0, c, b\}$$

The new array still has oddness.

## Program Usage



1. **open image**

2. choose kernel type

3. choose kernel size

4. **convolution**

- o execution time will show in third-to-last line

5. **Clear** to close image
6. **Close** to exit the program

I do the preprocess to map rpg image to gray scale and add suitable padding before convolution

## Part 2:

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### result

have various types of masks to use

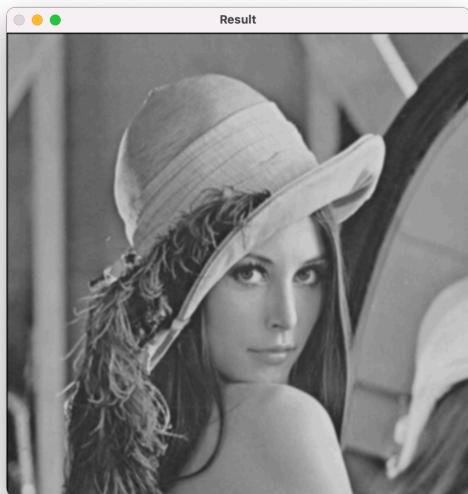
- box
- Gaussian
- LoG
- Marr-Hildreth
- Sobel
- max
- median

and can use with several sizes

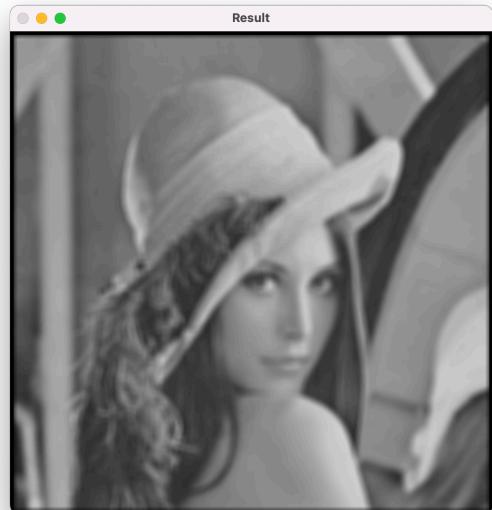
- 3, 5, 7, 9, 11, 21, 25, 43, 85

Take box filter for example

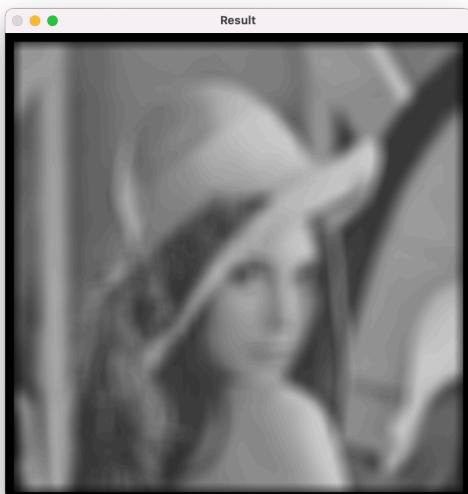
**3x3 box execution time: 40ms**



**9x9 box execution time: 256ms**



**21x21 box excution time: 1390ms**



**43x43 box execution time: 5750ms**



## Discuss

when mask's size bigger

1. the filter is more powerful
  - although it's not obvious when using Gaussian
2. the execution time will be longer

## Part 3:

# method

Marr-Hildreth

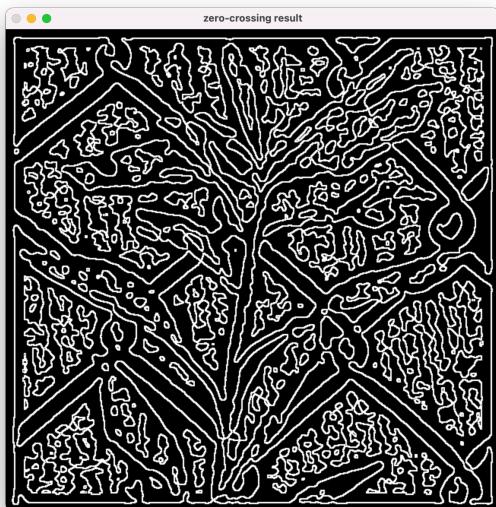
1. convert to grayscale
2. scale to [0 1]
3. conv2 25x25 Gaussian and Laplacian( $1\ 1\ 1\ 1\ -8\ 1\ 1\ 1$ )
4. scale to [0 255]
5. zero-crossing
  - find 3x3 neighbors which have different sign and difference > threshold

# result

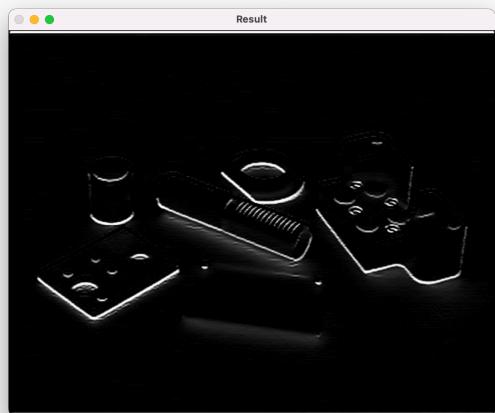
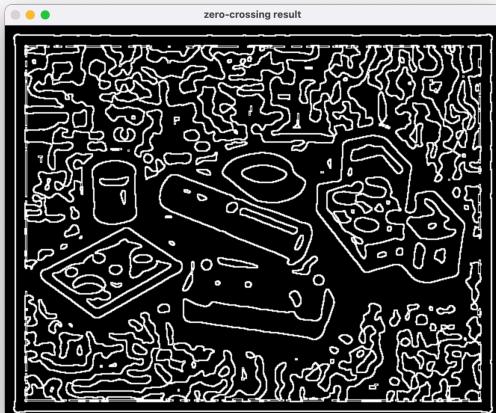
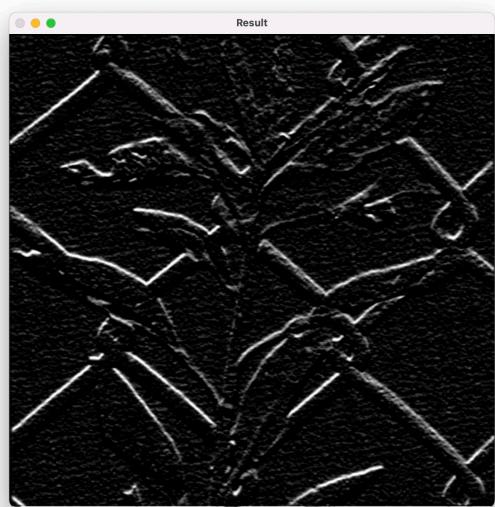
Marr-Hildreth Sobel comparison

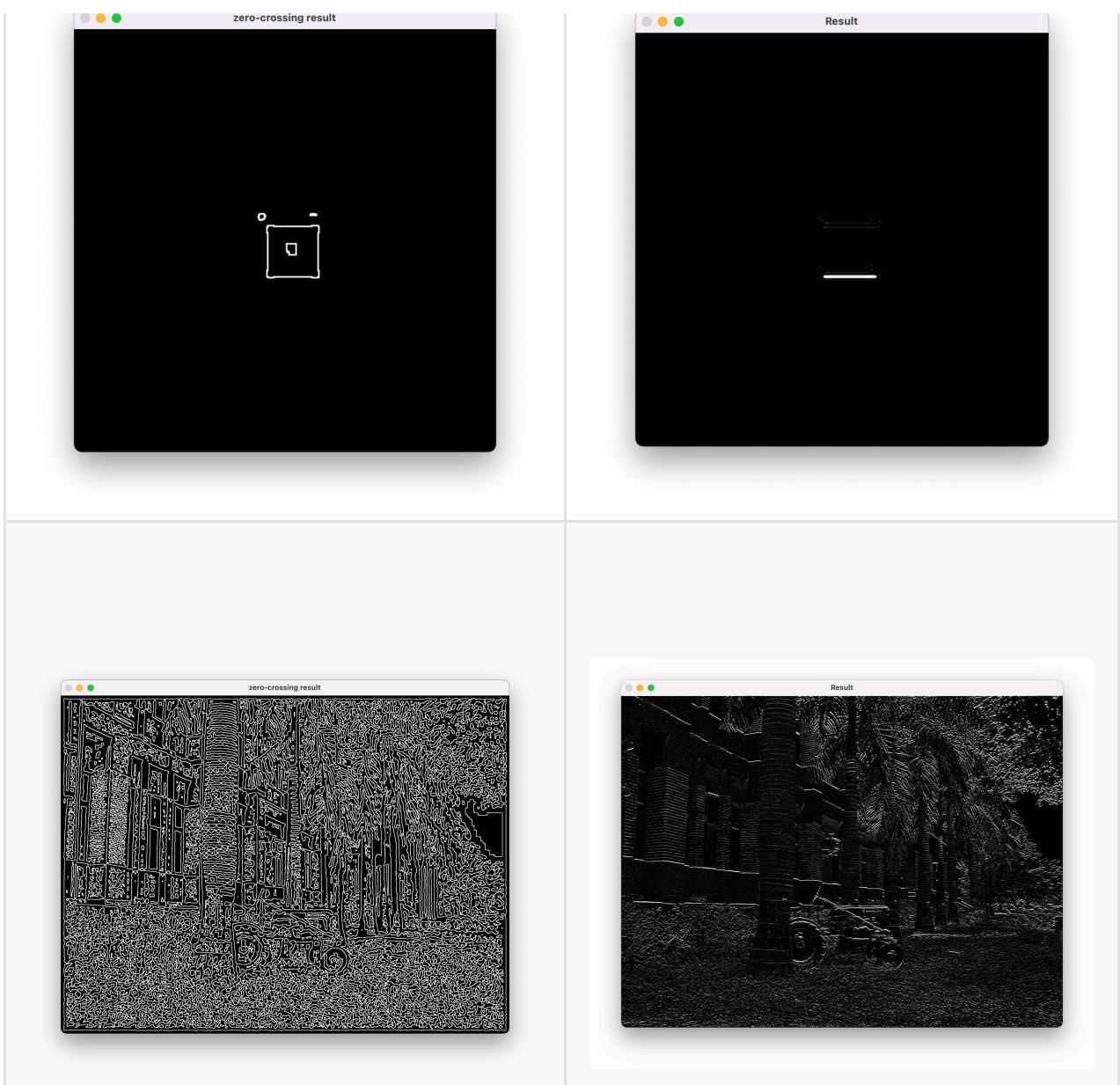
zero-crossing threshold = 0

**Marr-Hildreth**

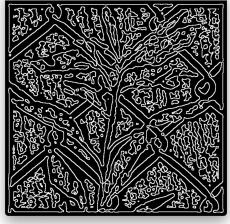
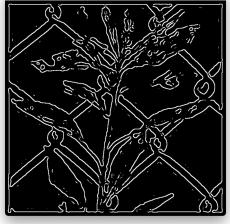
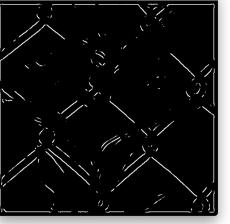


**Sobel**





zero-crossing threshold with Marr-Hildreth comparison

thres = 0% of max value	25%	50%	75%
			

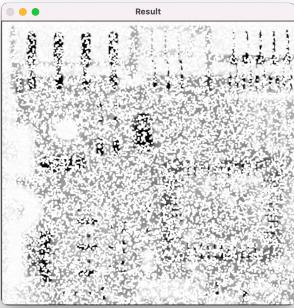
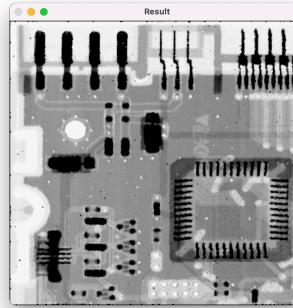
## Discussion

With 0% threshold, all edges can be detected, however, some noise will also show in image.  
In this example, with 25%, some noise we don't want will be removed, and show a nice result.  
When threshold more than 50%, almost everything were removed.

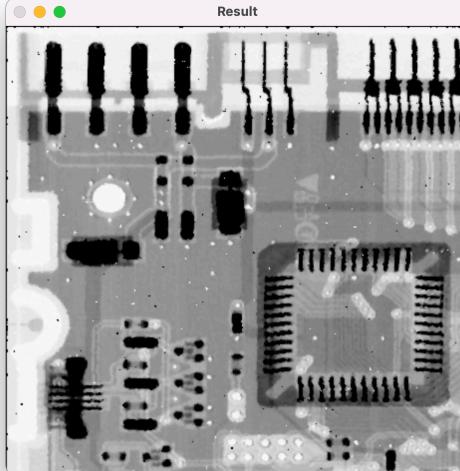
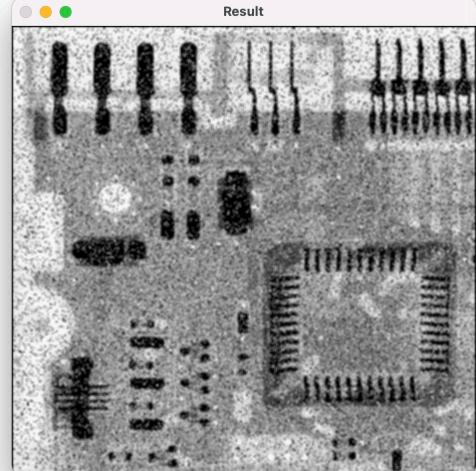
## Part 4:

### result

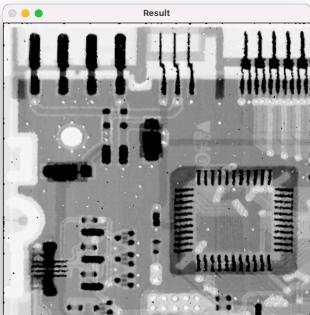
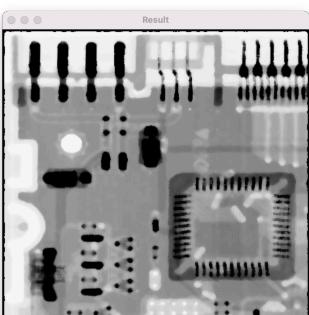
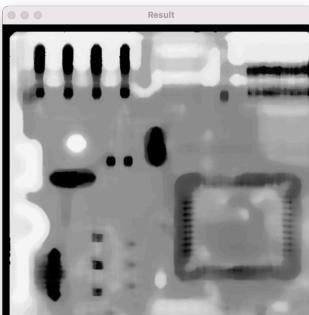
max median min filter comparison

3x3 max	3x3 median	3x3 min
		

median Gaussian filter comparison

3x3 median	3x3 Gaussian sigma=1
	

different size median filter comparison

<b>3x3 median</b>	<b>7x7 median</b>	<b>21x21 median</b>
		

## Discuss

1. Comparing median to Gaussian filter can find out that median filter can remove the noise points which spread on monotonous blocks of the image, however Gaussian filter smoothed the entire image.
2. Comparing the filter size, again, the larger filter will cause stronger effect. If we want to reduce the noise, 21x21 median may be too large.