

## Lecture 7-1: Spatial Filtering

**Yuyao Zhang, Xiran Cai PhD**

[zhangyy8@shanghaitech.edu.cn](mailto:zhangyy8@shanghaitech.edu.cn) [caixr@shanghaitech.edu.cn](mailto:caixr@shanghaitech.edu.cn)

SIST Building 2 302-F/302-C

Course piazza link: [piazza.com/shanghaitech.edu.cn/spring2021/cs270spring2021](https://piazza.com/shanghaitech.edu.cn/spring2021/cs270spring2021)

# Outline

## ➤ Spatial filtering definition

## ➤ Smoothing (平滑)

- Linear filter
- Non-linear filter

## ➤ Sharpening (锐化)

- Spatial differentiation
- Laplace filter

# Spatial Filtering

- A Spatial filter is directly applied on the image
- A Spatial filter is also called spatial masks (掩模)、kernels (核)、templates (模板)、windows (窗口)
- A Spatial filter consists of
  - 1) neighborhood
  - 2) a predefined operation
- A Spatial filter can be linear and nonlinear
  - Linear spatial filter corresponds to spectral filter in frequency domain
  - Nonlinear spatial filter cannot be accomplished in frequency domain

# Time-domain Convolution

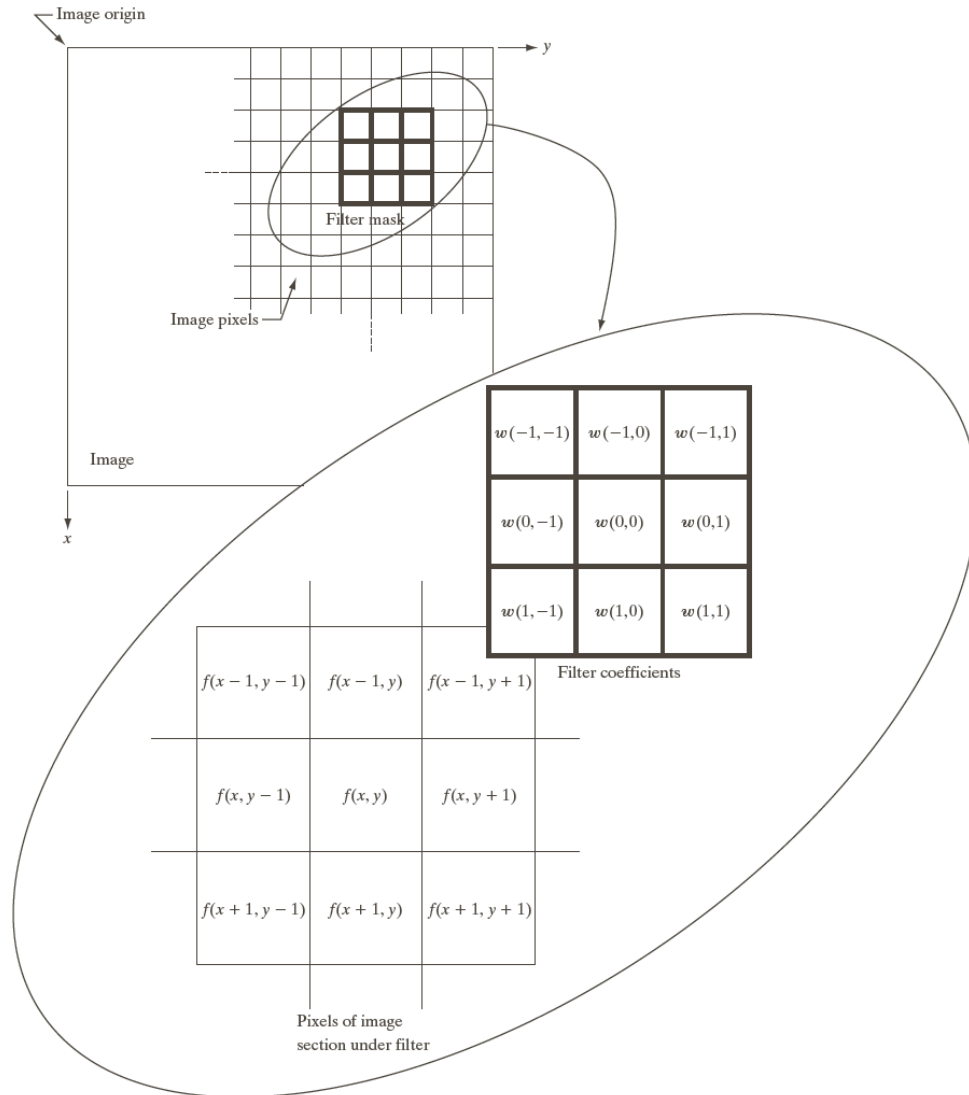
- Convolution of two signals  $x(t)$  and  $h(t)$ , denoted by  $x(t) * h(t)$ , is defined by

$$y(t) = x(t) * h(t) = \int_{-\infty}^{+\infty} x(\tau)h(t - \tau)d\tau$$

- For discrete-time

$$y[n] = x[n] * h[n] = \sum_{k=-\infty}^{+\infty} x[k]h[n - k]$$

# Spatial Filter



$$g(x, y) = \sum_{s=-a}^a \sum_{t=-b}^b w(s, t) f(x + s, y + t)$$

- $f(x, y)$ : input image
- $g(x, y)$ : output filtered image
- $w(s, t)$ :  $m \times n$  spatial filter, where  $m = 2a + 1, n = 2b + 1$

# Spatial Filter

- For discrete-time convolution

$$y[n] = x[n] * h[n] = \sum_{k=-\infty}^{+\infty} x[k]h[n-k]$$

- Spatial filter

$w(-1,-1)$	$w(-1,0)$	$w(-1,1)$
$w(0,-1)$	$w(0,0)$	$w(0,1)$
$w(1,-1)$	$w(1,0)$	$w(1,1)$

$$g(x, y) = \sum_{s=-a}^a \sum_{t=-b}^b w(s, t) f(x + s, y + t)$$

$$g(x, y) = \sum_{s=-a}^a \sum_{t=-b}^b w(s, t) f(x - s, y - t)$$

# Correlation and Convolution (2D)

Figure 1 illustrates a 3x3 convolution operation. The figure is divided into four main parts: (a) Original image and kernel, (b) Padded image and correlation result, (c) Rotated kernel and correlation result, and (d) Full convolution result and cropped convolution result.

(a) Original image  $f(x, y)$  and kernel  $w(x, y)$ :

0	0	0	0	0
0	0	0	0	0
0	0	1	0	0
0	0	0	0	0
0	0	0	0	0

$w(x, y)$ :

1	2	3
4	5	6
7	8	9

(b) Padded  $f$  and correlation result:

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

(c) Rotated  $w$  and correlation result:

3	2	1
6	5	4
9	8	7

(d) Full convolution result and cropped convolution result:

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

(e) Cropped convolution result:

0	0	0	0	0
0	9	8	7	0
0	6	5	4	0
0	3	2	1	0
0	0	0	0	0

(f) Rotated  $w$  and correlation result:

9	8	7
6	5	4
3	2	1

(g) Full convolution result and cropped convolution result:

0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0
0	0	0	0	0	0	0	0	0	0

(h) Cropped convolution result:

0	0	0	0	0
0	1	2	3	0
0	4	5	6	0
0	7	8	9	0
0	0	0	0	0

# Equations

## Correlation

$$w(s, t) \star f(x, y) = \sum_{s=-a}^a \sum_{t=-b}^b w(s, t) f(x + s, y + t)$$

## Convolution

$$w(s, t) \star f(x, y) = \sum_{s=-a}^a \sum_{t=-b}^b w(s, t) f(x - s, y - t)$$

$w_1$	$w_2$	$w_3$
$w_4$	$w_5$	$w_6$
$w_7$	$w_8$	$w_9$



# Spatial Filter Masks

## ➤ Linear Spatial Filter (线性滤波器)

- $R = \frac{1}{9} \sum_{k=1}^9 Z_k$
- $h(x, y) = e^{-\frac{x^2+y^2}{2\sigma^2}}$

## ➤ Nonlinear Spatial Filter (非线性滤波器)

- Max filter (最大值滤波)
- Median filter (中值滤波)

# Smooth Filters (平滑滤波器)

- **Blurring – for preprocessing tasks**
- **Noise deduction**
  - **Linear filter : average filtering – lowpass filter in frequency domain**
  - **Nonlinear filter**

# Smooth Filters (平滑滤波器)

$$g(x, y) = \frac{\sum_{s=-a}^a \sum_{t=-b}^b w(s, t) f(x + s, y + t)}{\sum_{s=-a}^a \sum_{t=-b}^b w(s, t)}$$

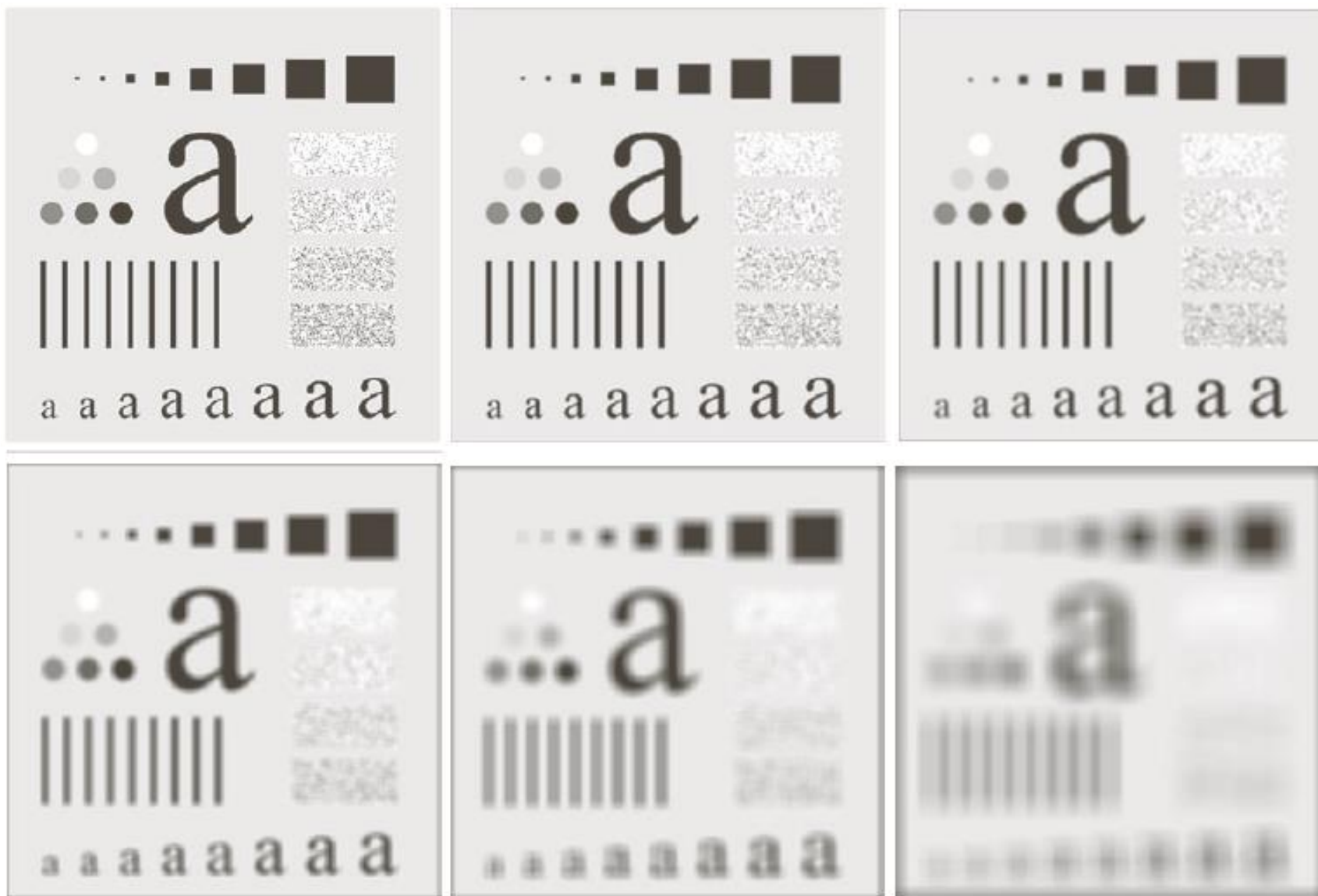
 $\frac{1}{9} \times$ 

1	1	1
1	1	1
1	1	1

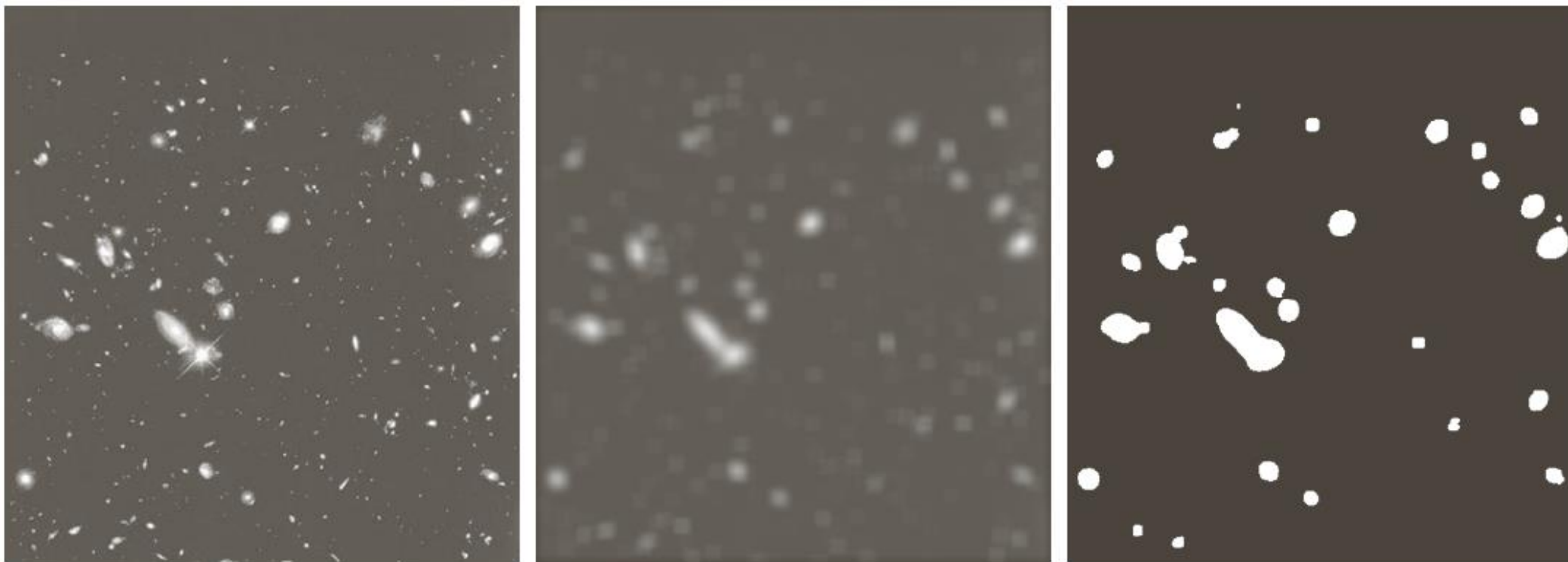
 $\frac{1}{16} \times$ 

1	2	1
2	4	2
1	2	1

# Filter size



# Smooth Filter and Thresholding(阈值处理)



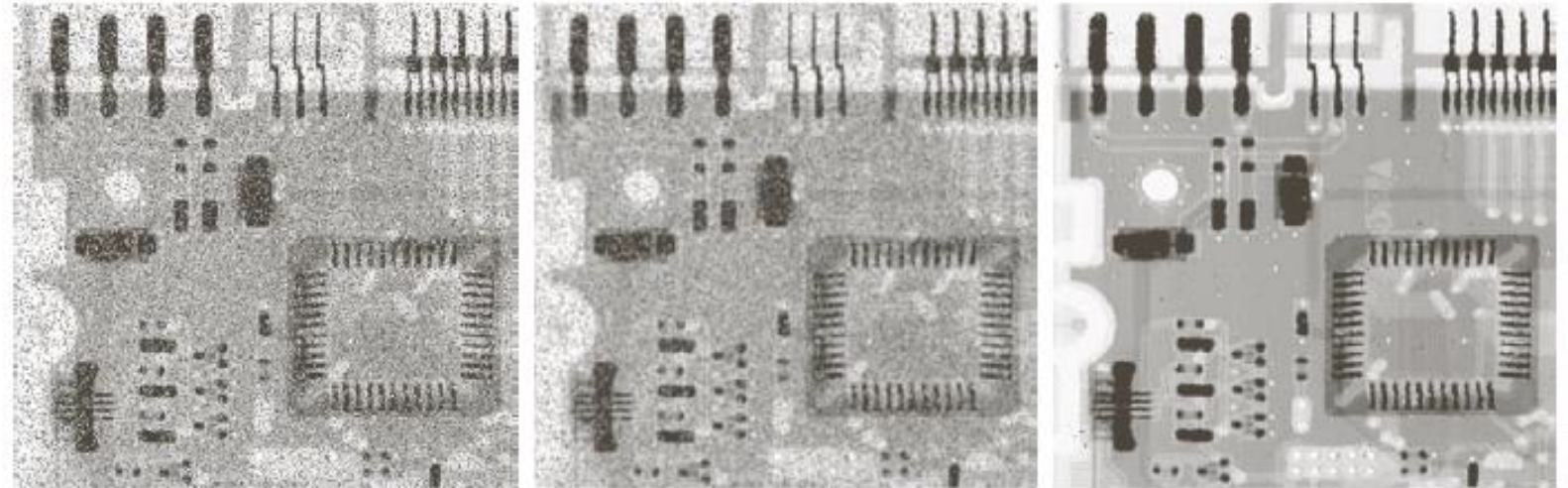
# Nonlinear Smooth Filters

➤ Order-statistic filter (统计排序滤波器)

Ex: median filter (中值滤波器)

$$g(x, y) = \text{median}\{m \times n \text{ pixel neighbouring around } I(x, y)\}$$

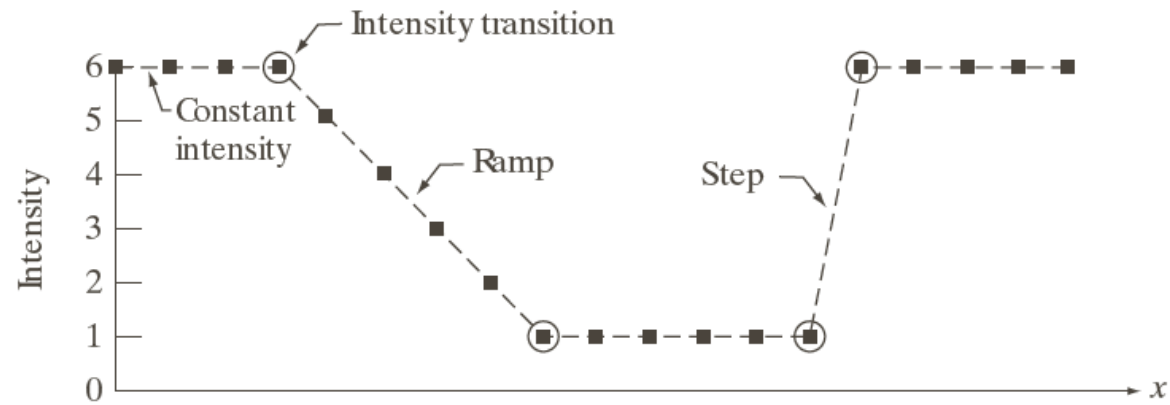
50	48	46	42
52	0	50	48
46	47	255	40
51	48	46	42



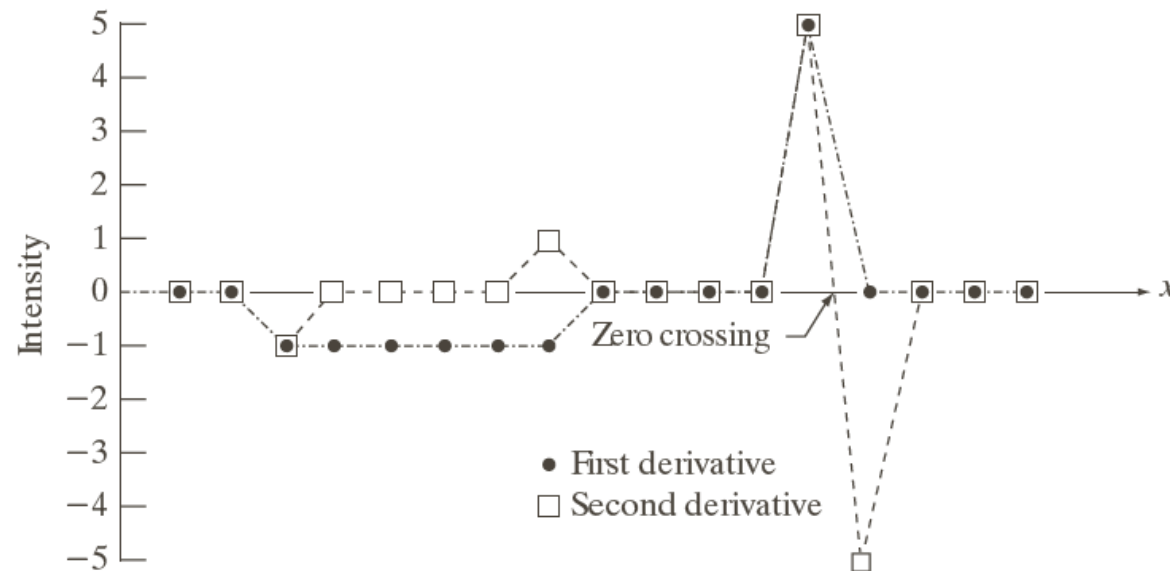
# Sharpening Filter

- Spatial differentiation (空间微分)
- Sharpening filter
  - Laplacian filtering (拉普拉斯算子)

# Derivative



Scan line	6	6	6	6	5	4	3	2	1	1	1	1	1	1	6	6	6	6	6	$x$
1st derivative	0	0	-1	-1	-1	-1	-1	0	0	0	0	0	0	5	0	0	0	0	0	
2nd derivative	0	0	-1	0	0	0	0	1	0	0	0	0	0	5	-5	0	0	0	0	





# Sharpening Filter

1. Zero in area of constant intensity
2. Nonzero at the onset of intensity step or ramp
3. (1) Nonzero along intensity ramp – 1<sup>st</sup> order derivative  
(2) Zero along intensity ramp with constant slope – 2<sup>nd</sup> order derivative

# Sharpening Filter

- To highlight transitions in intensity
- Accomplished by spatial differentiation
  - First-order derivative:  $\frac{\partial f}{\partial x} = f(x + 1) - f(x)$
  - Second-order derivative:  $\frac{\partial^2 f}{\partial x^2} = f(x + 1) + f(x - 1) - 2f(x)$

# Laplacian(拉普拉斯算子)

For an image function  $f(x, y)$ ,

$$\text{X direction: } \frac{\partial^2 f}{\partial x^2} = f(x+1, y) + f(x-1, y) - 2f(x, y)$$

$$\text{Y direction: } \frac{\partial^2 f}{\partial y^2} = f(x, y+1) + f(x, y-1) - 2f(x, y)$$

$$\begin{aligned}\nabla^2 f(x, y) &= \frac{\partial^2 f}{\partial x^2} + \frac{\partial^2 f}{\partial y^2} \\ &= f(x, y+1) + f(x, y-1) + f(x+1, y) + f(x-1, y) - 4f(x, y)\end{aligned}$$

# Laplacian Filter Masks

0	1	0	1	1	1	0	-1	0	-1	-1	-1
1	-4	1	1	-8	1	-1	4	-1	-1	8	-1
0	1	0	1	1	1	0	-1	0	-1	-1	-1

# Laplacian Filter Masks

$$g(x, y) = f(x, y) + c\nabla^2 f(x, y), \quad \text{where } c = \pm 1$$

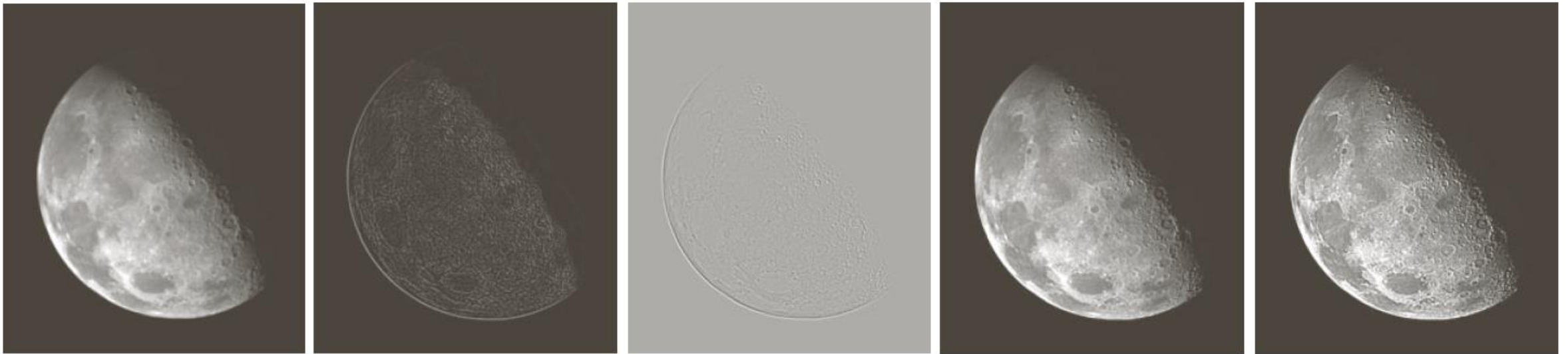
	1	
1	-4	1
	1	

$$c = -1$$

	-1	
-1	4	-1
	-1	

$$c = 1$$

# Image Sharpening with Laplacian



# Implementations in matlab

- **Low pass filter example:**

```
>> LP = 1/9 * [1,1,1;1,1,1;1,1,1];  
>> im3 = imfilter(com,LP);  
>> figure; imshow(im3,[]);
```

- **Sharpening filter example:**

```
>> f3 = [-1,-1,-1; -1,8,-1;-1,-1,-1];  
>> J1 = imfilter(im,f1);  
>> figure; imshow(J1,[]);
```

- **Median filter example:**

```
>> J2 = medfilt2(im,[3 3]);  
>> J4 = medfilt2(im,[6 6]);  
>> J3 = medfilt2(im,[11 1]);  
>> J5 = medfilt2(im,[1 11]);
```



# Take home message

- For image processing, the spatial domain processing is similar to 1-D signal processing in time domain.
- The spatial filter we discussed in this lecture is actually the calculation of correlation between the image and the filter. When using a diagonal symmetric filter, it is equivalent to convolution.
- Common spatial filters include smoothing filter and sharpening filter.