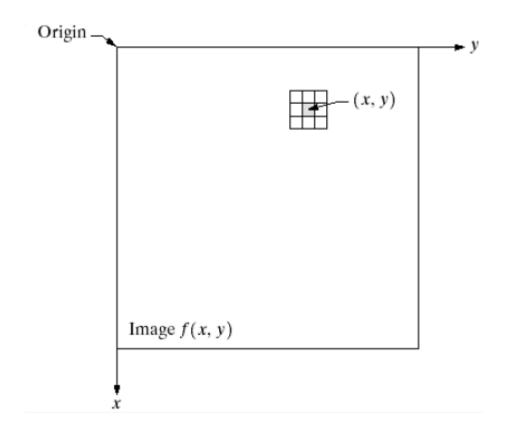
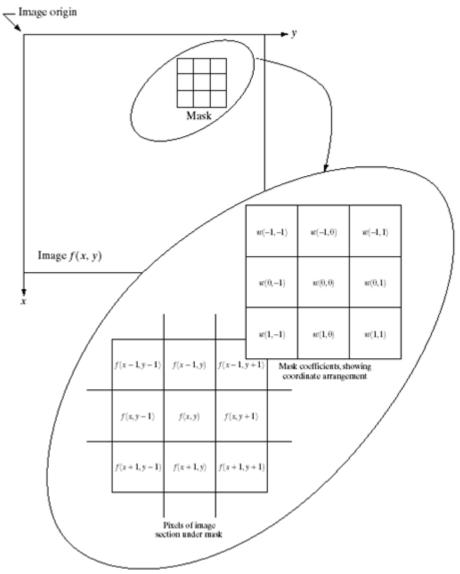
# CSE 365: Computer Vision

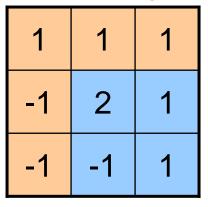
# Image Enhancement in the Spatial Domain

Prof. Mahmoud Khalil



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



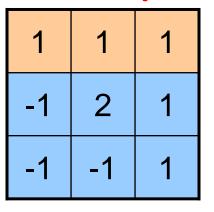


2	2	2	3
2	1	3	3
2	2	1	2
1	3	2	2



5		

f\*h



2	2	2	3
2	1	3	3
2	2	1	2
1	3	2	2

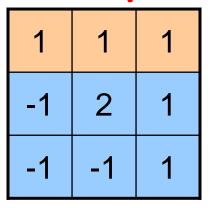
h

1	1	1	
-2	4	2	3
-2	-1	3	3
2	2	1	2
1	3	2	2



5	4	

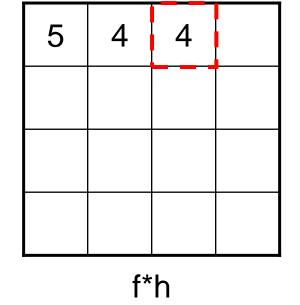
f\*h



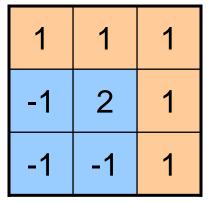
2	2	2	3
2	1	3	3
2	2	1	2
1	3	2	2

h			
	1	1	1
2	-2	4	3
2	-1	-3	3
2	2	1	2
1	3	2	2
f			





h



2	2	2	3
2	1	3	3
2	2	1	2
1	3	2	2

		1	1	1	
2	2	-2	6	1	
2	1	-3	-3	1	
2	2	1	2		
1	3	2	2		

5	4	4	-2

f\*

1	1	1
-	2	1
-1	-1	1

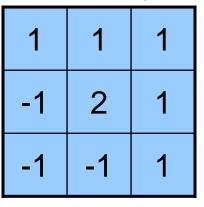
2	2	2	3
2	1	3	3
2	2	1	2
1	3	2	2

h

1	2	2	2	3
-1	4	1	3	3
-1	-2	2	1	2
	1	3	2	2



5	4	4	-2
9			

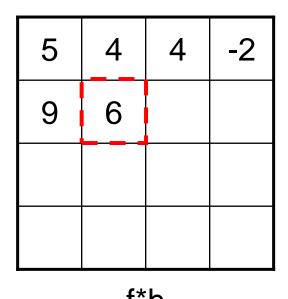


2	2	2	3
2	1	3	3
2	2	1	2
1	3	2	2

h

2	2	2	3
-2	2	3	3
-2	-2	1	2
1	3	2	2

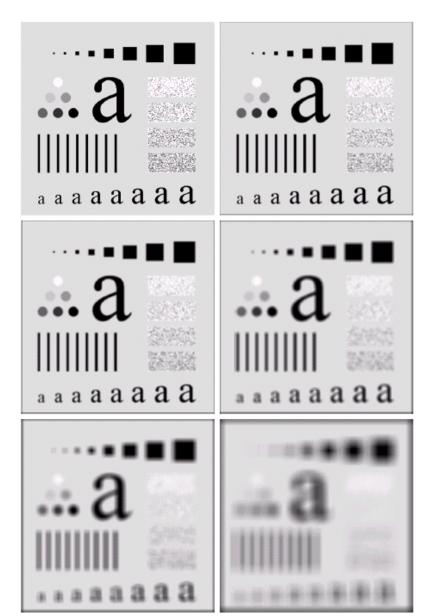




f

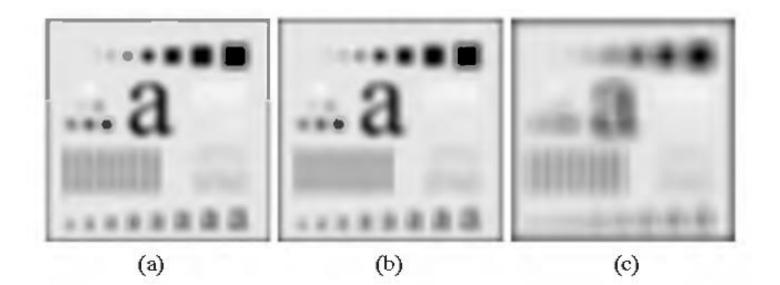
	1	1	1
$\frac{1}{9}$ ×	1	1	1
	1	1	1

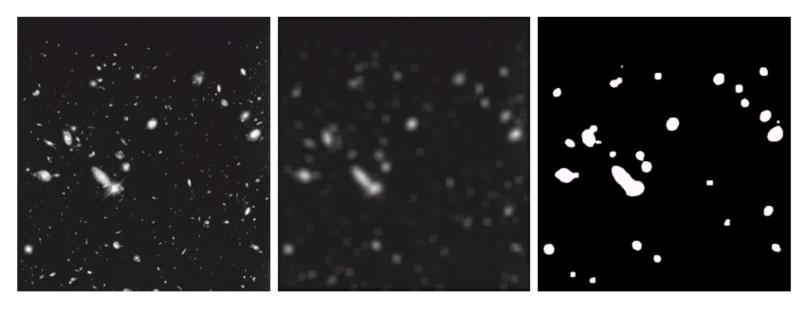
	1	2	1
$\frac{1}{16}$ ×	2	4	2
	1	2	1



#### Problem 3.21

The three images shown were blurred using square averaging masks of sizes n = 23,25, and 45, respectively. The vertical bars on the left lower part of (a) and (c) are blurred, but a clear separation exists between them. However, the bars have merged in image (b), in spite of the fact that the mask that produced this image is significantly smaller than the mask that produced image (c). Explain this.





a b c

**FIGURE 3.36** (a) Image from the Hubble Space Telescope. (b) Image processed by a 15 × 15 averaging mask. (c) Result of thresholding (b). (Original image courtesy of NASA.)

#### Problem 3.22

Consider an application such as the one shown in Fig. 3.34, in which it is desired to eliminate objects smaller than those enclosed in a square of size q X q pixels.

Suppose that we want to reduce the average gray level of those objects to one-tenth of their original average gray level. In this way, those objects will be closer to the gray level of the background and they can then be eliminated by thresholding. Give the (odd) size of the smallest averaging mask that will accomplish the desired reduction in average gray level in only one pass of the mask over the image.

#### Problem 3.20

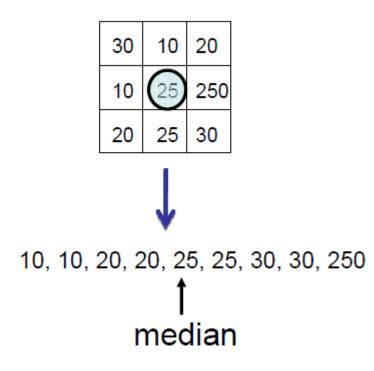
- (a) In a character recognition application, text pages are reduced to binary form using a thresholding transformation function of the form shown in Fig. 3.2(b). This is followed by a procedure that thins the characters until they become strings of binary 1's on a background of 0's. Due to noise, the binarization and thinning processes result in broken strings of characters with gaps ranging from 1 to 3 pixels. One way to "repair" the gaps is to run an averaging mask over the binary image to blur it, and thus create bridges of nonzero pixels between gaps. Give the (odd) size of the smallest averaging mask capable of performing this task.
- (b) After bridging the gaps, it is desired to threshold the image in order to convert it back to binary form. For your answer in (a), what is the minimum value of the threshold required to accomplish this, without causing the segments to break up again?

#### **Order-Statistics Filters**

Median Filter

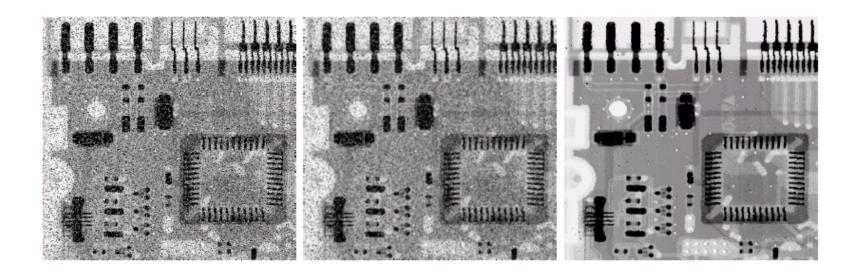


Minimum Filter



#### **Order-Statistics Filters**

a b c



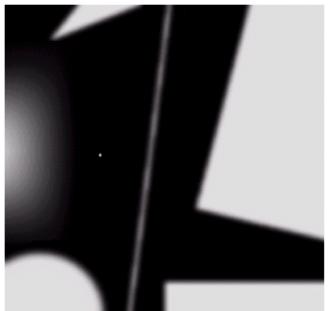
**FIGURE 3.37** (a) X-ray image of circuit board corrupted by salt-and-pepper noise. (b) Noise reduction with a 3 × 3 averaging mask. (c) Noise reduction with a 3 × 3 median filter. (Original image courtesy of Mr. Joseph E. Pascente, Lixi, Inc.)

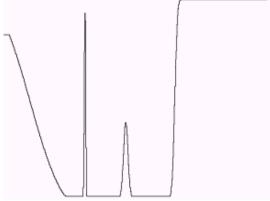
## **Sharpening Spatial Filters**

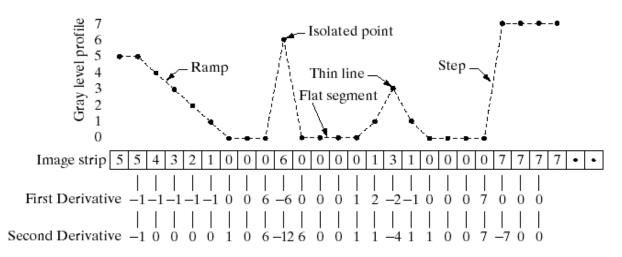
a b

#### FIGURE 3.38

(a) A simple image. (b) 1-D horizontal gray-level profile along the center of the image and including the isolated noise point.
(c) Simplified profile (the points are joined by dashed lines to simplify interpretation).







#### Sharpening Spatial Filters – Laplacian Filter

$$\frac{\partial f}{\partial x} = f(x+1) - f(x)$$

Second derivative 
$$\frac{\partial^2 f}{\partial x^2} = f(x+1) + f(x-1) - 2f(x)$$

$$\nabla^2 f(x,y) = \frac{\partial^2 f(x,y)}{\partial^2 x} + \frac{\partial^2 f(x,y)}{\partial^2 y}$$

$$\nabla^2 f(x,y) \approx f(x+1,y) + f(x-1,y) + f(x,y+1) + f(x,y-1) - 4f(x,y)$$

0	1	0
1	-4	1
0	1	0

#### **Sharpening Spatial Filters**

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



#### FIGURE 3.39

(a) Filter mask used to implement the digital Laplacian, as defined in Eq. (3.7-4). (b) Mask used to implement an extension of this equation that includes the diagonal neighbors. (c) and (d) Two other implementations of the Laplacian.

## **Sharpening Spatial Filters**

$$g(x,y) = f(x,y) \pm c \cdot \nabla^2 f(x,y)$$

use 
$$c = -1$$
 for

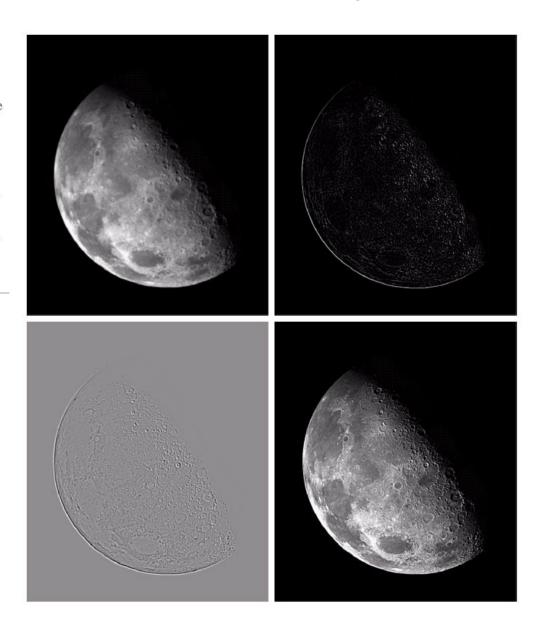
0	1	0	1	L	1	1
1	-4	1	1	L	-8	1
0	1	0	1	L	1	1

use 
$$c = +1$$
 for

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

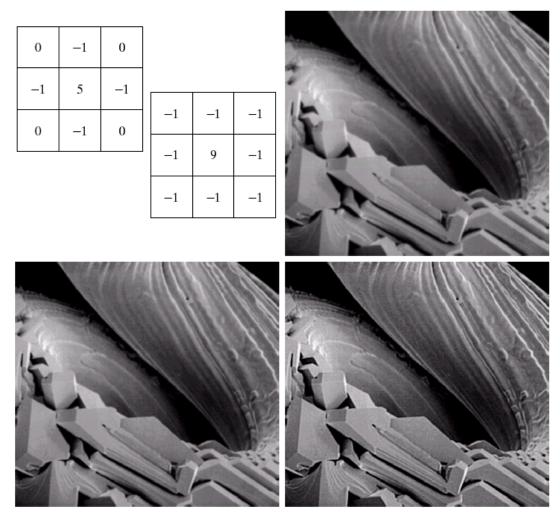
a b c d

FIGURE 3.40 (a) Image of the North Pole of the moon. (b) Laplacianfiltered image. (c) Laplacian image scaled for display purposes. (d) Image enhanced by using Eq. (3.7-5). (Original image courtesy of NASA.)



#### Problem 3.23

In a given application an averaging mask is applied to input images to reduce noise, and then a Laplacian mask is applied to enhance small details. Would the result be the same if the order of these operations were reversed?



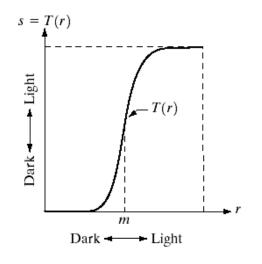
a b c d e

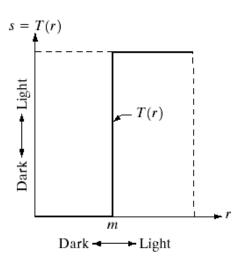
**FIGURE 3.41** (a) Composite Laplacian mask. (b) A second composite mask. (c) Scanning electron microscope image. (d) and (e) Results of filtering with the masks in (a) and (b), respectively. Note how much sharper (e) is than (d). (Original image courtesy of Mr. Michael Shaffer, Department of Geological Sciences, University of Oregon, Eugene.)

0	-1	0	-1	-1	-1
-1	A + 4	-1	-1	A + 8	-1
0	-1	0	-1	-1	-1

a b

**FIGURE 3.42** The high-boost filtering technique can be implemented with either one of these masks, with  $A \ge 1$ .





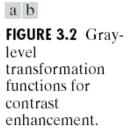
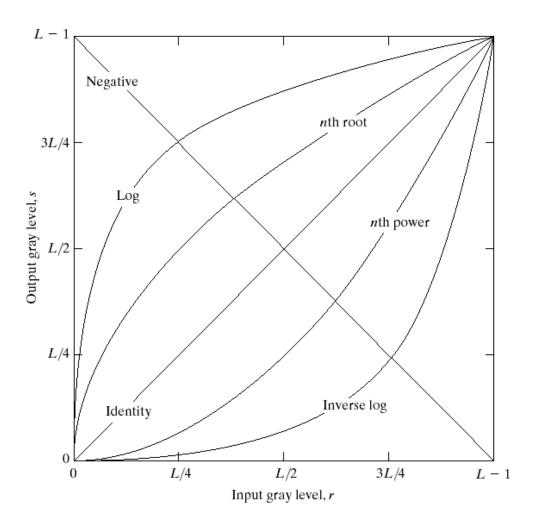
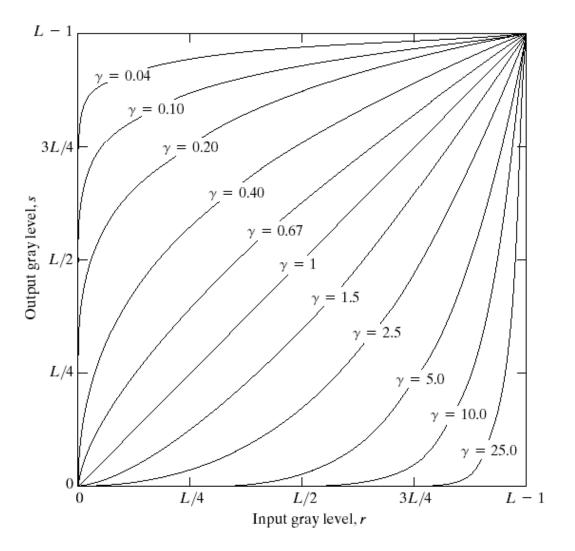


FIGURE 3.3 Some basic gray-level transformation functions used for image enhancement.



#### **Gamma Correction**



**FIGURE 3.6** Plots of the equation  $s = cr^{\gamma}$  for various values of  $\gamma$  (c = 1 in all cases).

#### **Gamma Correction**

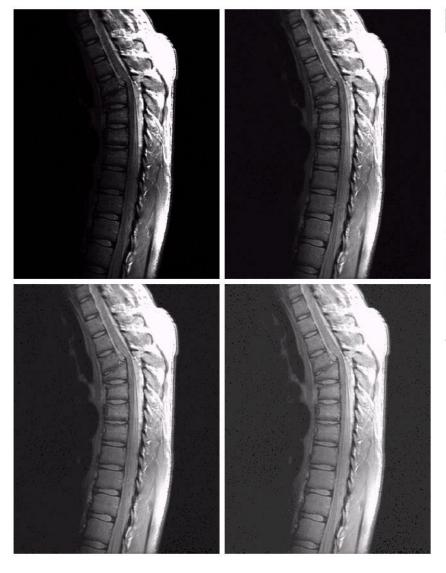


FIGURE 3.8
(a) Magnetic resonance (MR) image of a fractured human spine.
(b)–(d) Results of applying the transformation in Eq. (3.2-3) with c = 1 and  $\gamma = 0.6, 0.4$ , and 0.3, respectively. (Original image for this example courtesy of Dr. David R. Pickens, Department of Radiology and Radiological Sciences, Vanderbilt University Medical Center.)

#### **Gamma Correction**

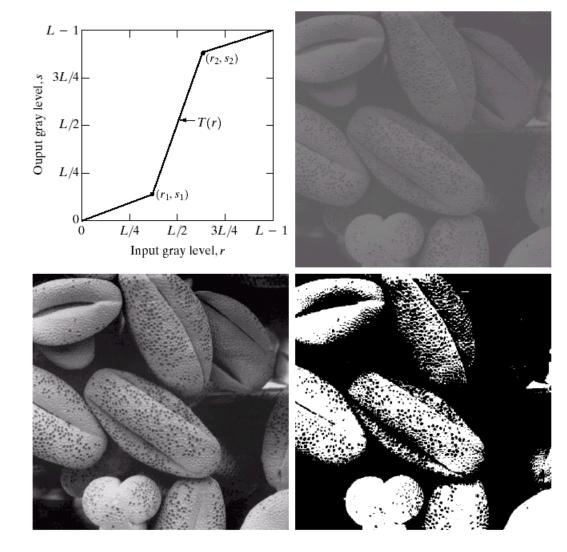
a b c d

#### FIGURE 3.9

(a) Aerial image. (b)–(d) Results of applying the transformation in Eq. (3.2-3) with c=1 and  $\gamma=3.0,4.0$ , and 5.0, respectively. (Original image for this example courtesy of NASA.)

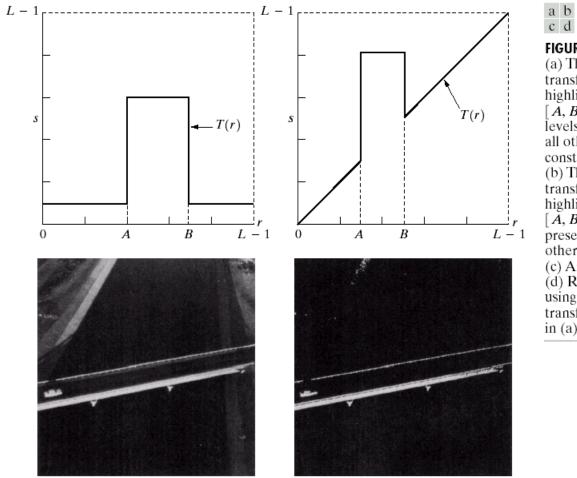


## **Contrast Stretching**



a b c d

FIGURE 3.10 Contrast stretching. (a) Form of transformation function. (b) A low-contrast image. (c) Result of contrast stretching. (d) Result of thresholding. (Original image courtesy of Dr. Roger Heady, Research School of Biological Sciences, Australian National University, Canberra, Australia.)





#### FIGURE 3.11

(a) This transformation highlights range [A, B] of gray levels and reduces all others to a constant level. (b) This transformation highlights range [A, B] but preserves all other levels. (c) An image. (d) Result of using the transformation in (a).

# Histogram Equalization

