

## 10.1 Introduction

- *Morphology* is a branch of image processing that is particularly useful for analyzing **shapes** in images
- We will develop basic morphological tools for investigation of **binary images** and then show how to extend these tools to **grayscale images**

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## 10.2 Basic Ideas

### • 10.2.1 Translation

$$A_w = \{(a, b) + (x, y) : (a, b) \in A\}$$

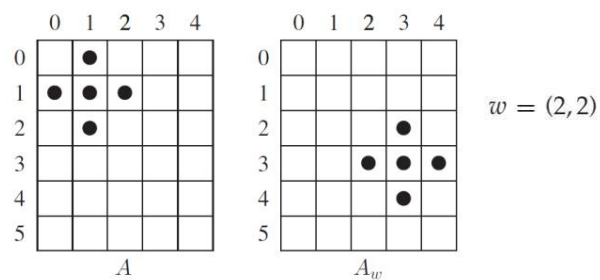


FIGURE 10.1 Translation.

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## FIGURE 10.2

### • 10.2.2 Reflection

$$\hat{A} = \{(-x, -y) : (x, y) \in A\}$$

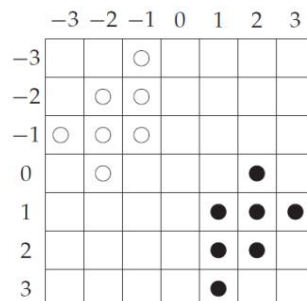


FIGURE 10.2 Reflection.

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## 10.3 Dilation and Erosion

### • 10.3.1 Dilation

$$A \oplus B = \bigcup_{x \in B} A_x$$

✓  $A$  and  $B$  are sets of pixels

$$A \oplus B = \{(x, y) + (u, v) : (x, y) \in A, (u, v) \in B\}$$

✓ Also known as **Minkowski addition**

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## FIGURE 10.3

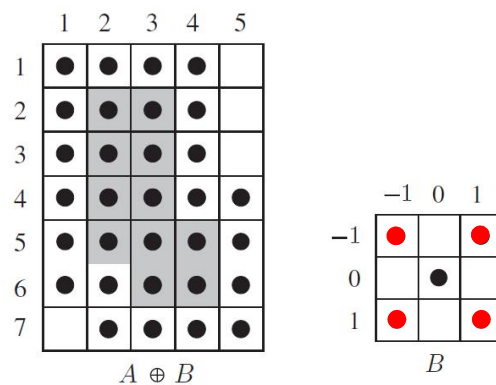


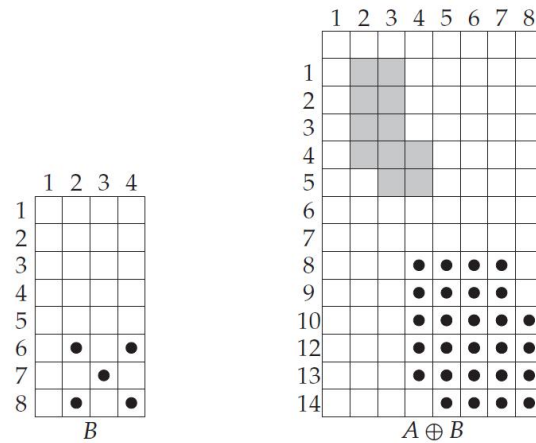
FIGURE 10.3 Dilation.

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FIGURE 10.4

FIGURE 10.4 A dilation for which  $A \not\subseteq A \oplus B$ .

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FIGURE 10.5

```
>> t=imread('text.tif');
>> sq=ones(3,3);
>> td=imdilate(t,sq);
>> subplot(1,2,1),imshow(t)
>> subplot(1,2,2),imshow(td)
```

Cross-Correlation Used  
To Locate A Known  
Target in an Image

Text Running  
In Another  
Direction

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(a)

Cross-Correlation Used  
To Locate A Known  
Target in an Image

Text Running  
In Another  
Direction

(b)

FIGURE 10.5 Dilation of a binary image. (a) Text image. (b) Result of dilation.

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## 10.3 Dilation and Erosion

### • 10.3.1 Erosion

$$A \ominus B = \{w : B_w \subseteq A\}$$

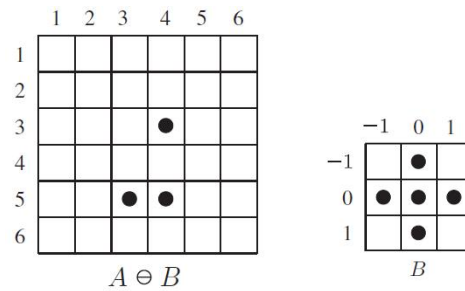


FIGURE 10.6 Erosion with a cross-shaped structuring element.

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## FIGURE 10.7

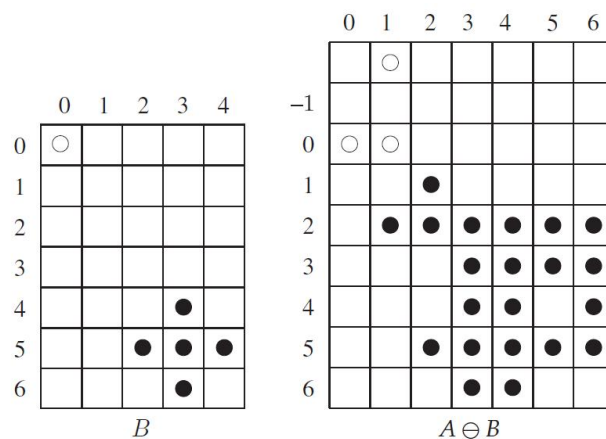


FIGURE 10.7 Erosion with a structuring element not containing the origin.

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FIGURE 10.8

```
>> c=imread('circbw.tif');
>> ce=imerode(c,sq);
>> subplot(1,2,1),imshow(c)
>> subplot(1,2,2),imshow(ce)
```

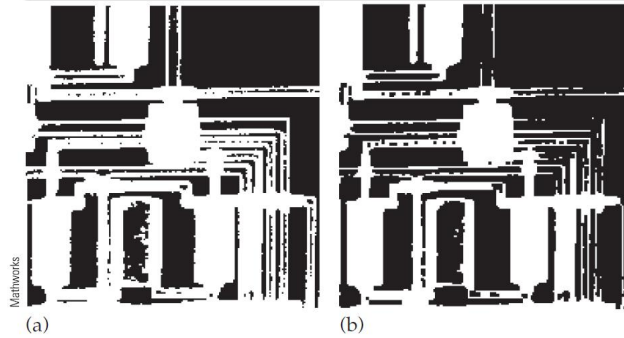


FIGURE 10.8 Erosion of a binary image. (a) Original image. (b) Result of erosion.

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## 10.3 Dilation and Erosion

### • RELATIONSHIP BETWEEN EROSION AND DILATION

$$\overline{A \ominus B} = \overline{A} \oplus \hat{B}$$

$$\overline{A \oplus B} = \overline{A} \ominus \hat{B}$$

```
>> lhs=~imerode(t,sq);
```

```
>> rhs=imdilate(~t,sq);
```

```
>> all(lhs(:)==rhs(:))
```

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## 10.3 Dilation and Erosion

### • 10.3.3 An Application: Boundary Detection

✓ If  $A$  is an image and  $B$  a small structuring element

- (i)  $A - (A \ominus B)$  internal boundary
- (ii)  $(A \oplus B) - A$  external boundary
- (iii)  $(A \oplus B) - (A \ominus B)$  morphological gradient

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## FIGURE 10.9

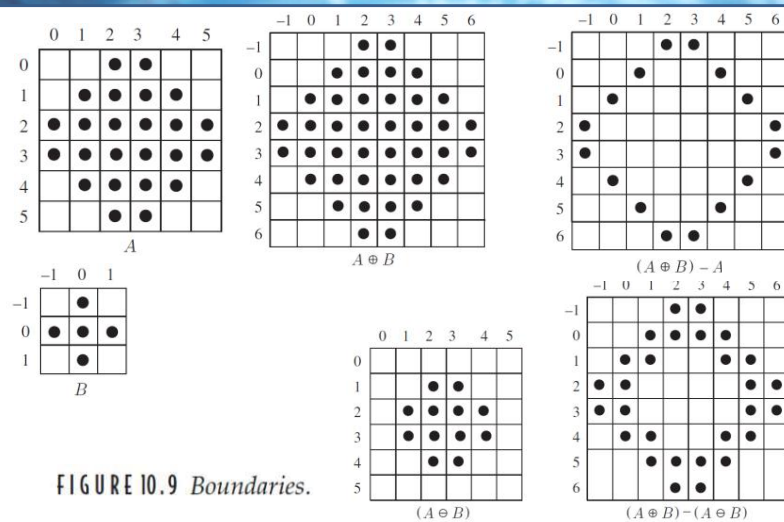


FIGURE 10.9 Boundaries.

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FIGURE 10.10

```
>> re=imerode(r,sq);
>> r_int=r&~re;
>> subplot(1,2,1),imshow(r)
>> subplot(1,2,2),imshow(r_int)
```

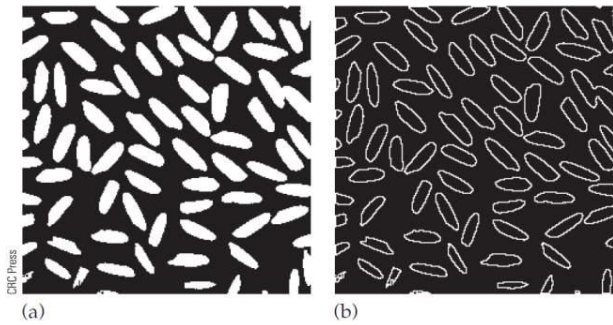


FIGURE 10.10 Morphological edge detection. (a) The rice grains image. (b) The internal boundary.

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FIGURE 10.11

```
>> rd=imdilate(r,sq);
>> r_ext=rd&~r;
>> r_grad=rd&~re;
>> subplot(1,2,1),imshow(r_ext)
>> subplot(1,2,2),imshow(r_grad)
```

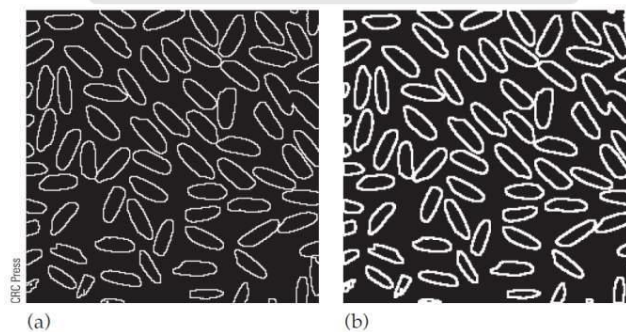


FIGURE 10.11 More morphological edge detection. (a) External boundary. (b) Morphological gradient.

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## 10.4 Opening and Closing

### • 10.4.1 Opening (function: imopen)

$$A \circ B = (A \ominus B) \oplus B$$

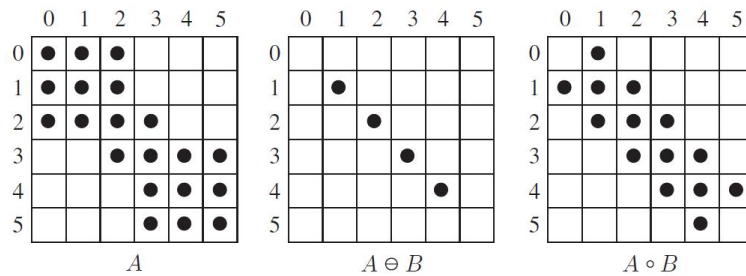


FIGURE 10.12 Opening.

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## 10.4 Opening and Closing

- ✓  $(A \circ B) \subseteq A$ . Note that this is not the case with erosion. As we have seen, an erosion may not necessarily be a subset
- ✓  $(A \circ B) \circ B = A \circ B$ . That is, an opening can never be done more than once (**idempotence**, 멍등법칙)
- ✓ If  $A \subseteq C$ , then  $(A \circ B) \subseteq (C \circ B)$
- ✓ Opening tends to smooth an image, to break narrow joins, and to remove thin protrusions (돌출부)

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## 10.4 Opening and Closing

- **10.4.2 Closing** (function: `imclose`)

$$A \bullet B = (A \oplus B) \ominus B$$

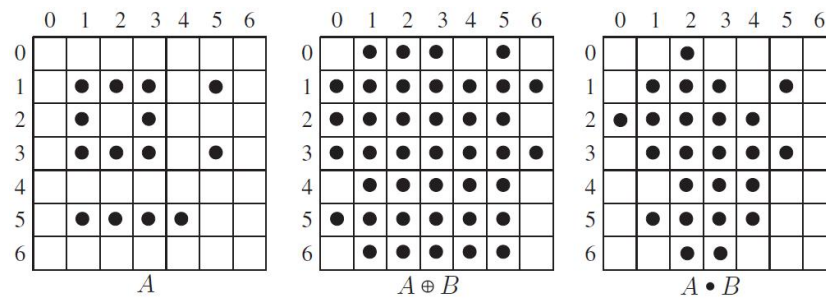


FIGURE 10.13 Closing.

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## 10.4 Opening and Closing

- ✓  $A \subseteq (A \bullet B)$
- ✓  $(A \bullet B) \bullet B = A \bullet B$ ; that is, closing, like opening, is **idempotent**
- ✓ If  $A \subseteq C$ , then  $(A \bullet B) \subseteq (C \bullet B)$
- ✓ Closing also tends to smooth an image, but it fuses narrow breaks and thin gulfs and eliminates small holes

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FIGURE 10.14

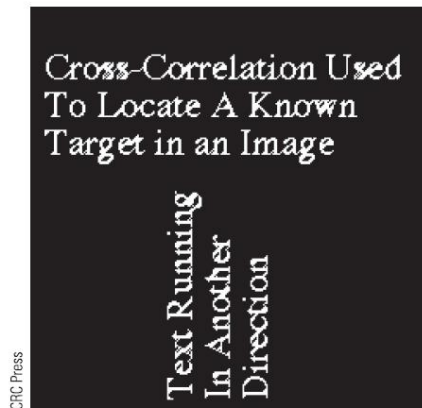


FIGURE 10.14 An example of closing.

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## 10.4 Opening and Closing

### • AN APPLICATION: NOISE REMOVAL

#### ✓ Morphological filtering

```
>> c=imread('circles.tif');
>> x=rand(size(c));
>> d1=find(x<=0.05);
>> d2=find(x>=0.95);
>> c(d1)=0;
>> c(d2)=1;
>> imshow(c)
```

```
>> cf1=imclose(imopen(c,sq),sq);
>> figure,imshow(cf1)
>> cf2=imclose(imopen(c,cr),cr);
>> figure,imshow(cf2)
```

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FIGURE 10.15

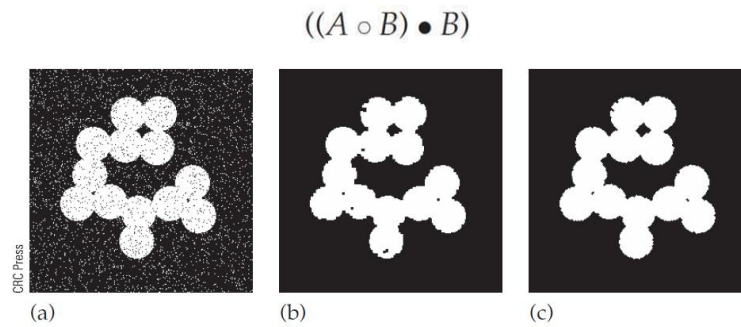


FIGURE 10.15 A noisy binary image and results after morphological filtering with different structuring elements. (a) Binary noise. (b) Using the square kernel. (c) Using the cross kernel.

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## 10.4 Opening and Closing

### • RELATIONSHIP BETWEEN OPENING AND CLOSING

$$\overline{A \bullet B} = \overline{A} \circ \hat{B}$$

$$\overline{A \circ B} = \overline{A} \bullet \hat{B}$$

✓ see Haralick and Shapiro [11] for a formal proof

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## 10.5 The Hit-or-Miss Transform

✓  $B$  is the square structuring element

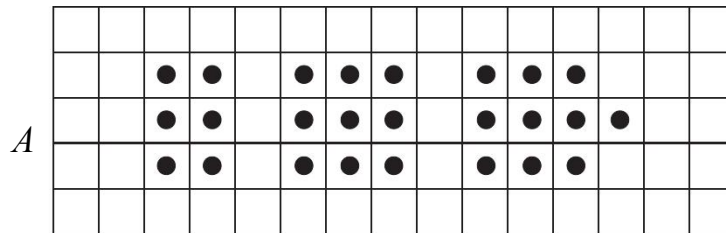


FIGURE 10.16 An image  $A$  containing a shape to be found.

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## FIGURE 10.17 & 18

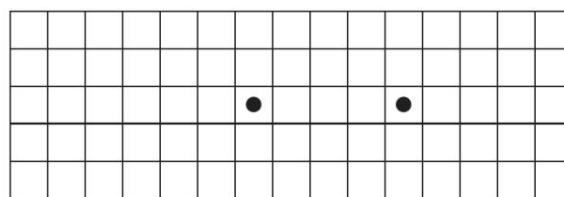


FIGURE 10.17 The erosion  $A \ominus B$ .

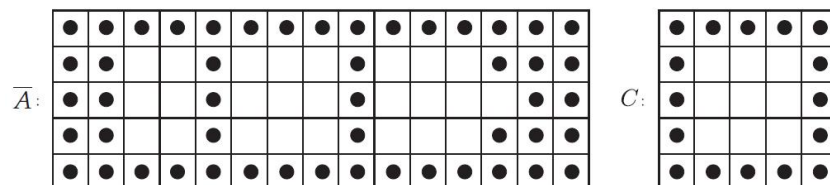


FIGURE 10.18 The complement and the second structuring element.

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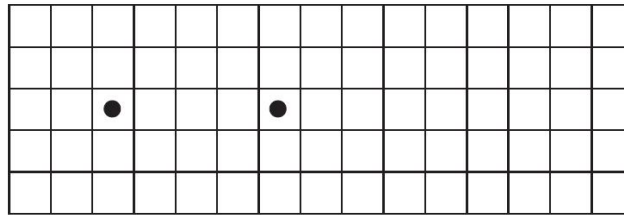
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FIGURE 10.19

FIGURE 10.19 The erosion  $\bar{A} \ominus C$ .

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## 10.5 The Hit-or-Miss Transform

$$A \otimes B = (A \ominus B_1) \cap (\bar{A} \ominus B_2)$$

```
>> b1=ones(1,6);
>> b2=[1 1 1 1 1 1 1 1;1 0 0 0 0 0 0 1; 1 1 1 1 1 1 1 1];
>> tb1=imerode(t,b1);
>> tb2=imerode(~t,b2);
>> hit_or_miss=tb1&tb2;
>> [x,y]=find(hit_or_miss==1)
```

```
>> tb1=imerode(t,b1);
```

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FIGURE 10.20

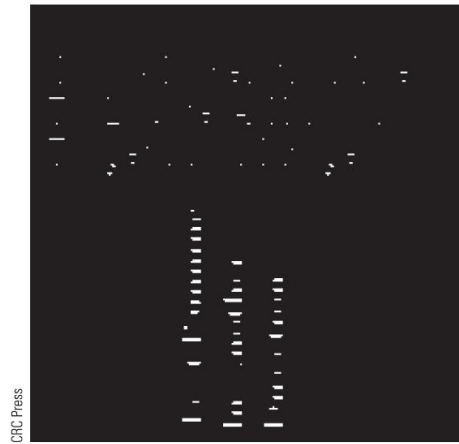


FIGURE 10.20 Text eroded by a hyphen-shaped structuring element.

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## 10.6 Some Morphological Algorithm

### • 10.6.1 Region Filling

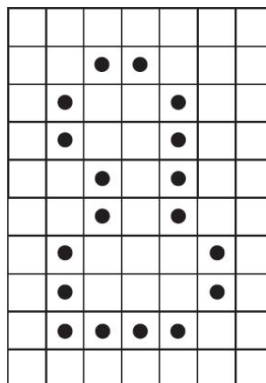


FIGURE 10.21 An 8-connected boundary of a region to be filled.

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## 10.6 Some Morphological Algorithm

- Given a pixel  $p$  within the region, we wish to fill up the entire region
- we start with  $p$  and **dilate** as many times as necessary with the cross-shaped structuring element  $B$

$$\{p\} = X_0, X_1, X_2, \dots, X_k = X_{k+1},$$

$$X_n = (X_{n-1} \oplus B) \cap \bar{A}$$

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### FIGURE 10.22

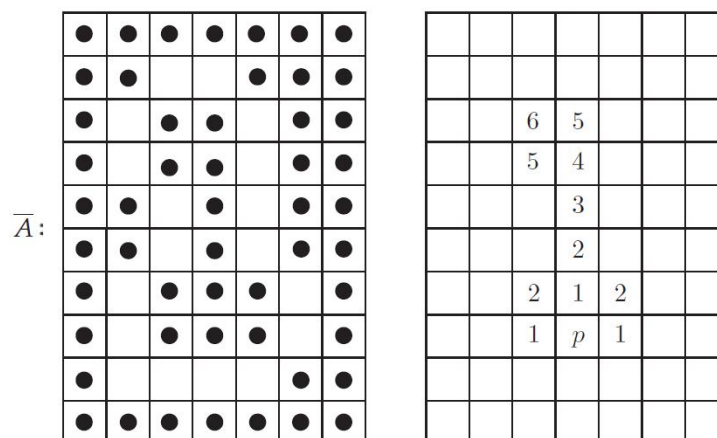


FIGURE 10.22 The process of filling a region.

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## 10.6 Some Morphological Algorithm

### • 10.6.2 Connected Components

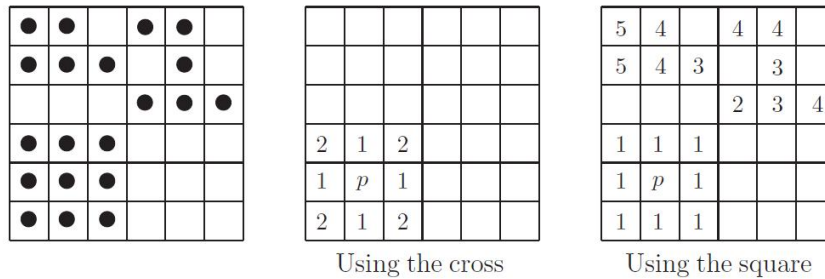


FIGURE 10.23 Filling connected components.

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## FIGURE 10.24

```
function out=regfill(im,pos,kernel)
% REGFILL(IM,POS,KERNEL) performs region filling of binary
% image IMAGE, with kernel KERNEL, starting at point with
% coordinates given by POS.
% Example:
%     n=imread('nicework.tif');
%     nb=n~imerode(n,ones(3,3));
%     nr=regfill(nb,[74,52],ones(3,3));
%
current=zeros(size(im));
last=zeros(size(im));
last(pos(1),pos(2))=1;
current=imdilate(last,kernel)&~im;
while any(current(:)~=last(:)),
    last=current;
    current=imdilate(last,kernel)&~im;
end;
out=current;
```

FIGURE 10.24 A simple program for filling regions.

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FIGURE 10.25

```
>> n=imread('nicework.tif');
>> imshow(n),pixval on
>> nb=n&~imerode(n,sq);
>> figure,imshow(nb)
>> nf=regfill(nb,[74,52],sq);
>> figure,imshow(nf)
```

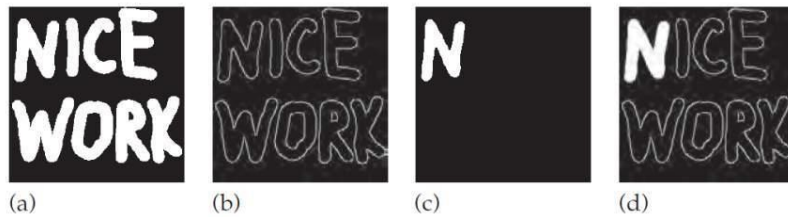


FIGURE 10.25 Region filling.

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FIGURE 10.26

```
function out=components(im,pos,kernel)
% COMPONENTS(IM,POS,KERNEL) produces the connected component
% of binary image IMAGE which includes the point with coordinates given
% by POS, using kernel KERNEL.
%
% Example:
%         n=imread('nicework.tif');
%         nc=components(nb,[74,52],ones(3,3));
%
current=zeros(size(im));
last=zeros(size(im));
last(pos(1),pos(2))=1;
current=imdilate(last,kernel)&im;
while any(current(:)~=last(:)),
    last=current;
    current=imdilate(last,kernel)&im;
end;
out=current;
```

FIGURE 10.26 A simple program for connected components.

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FIGURE 10.27

```
>> sq2=ones(11,11);
>> nc=components(n,[57,97],sq);
>> imshow(nc)
>> nc2=components(n,[57,97],sq2);
>> figure,imshow(nc2)
```

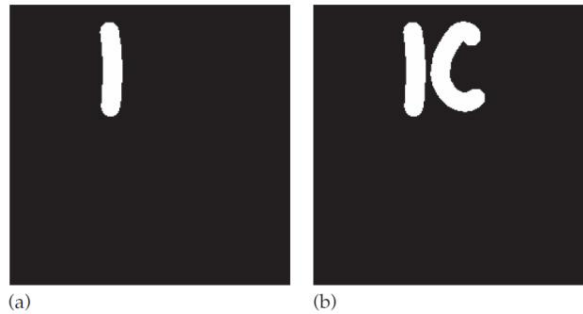


FIGURE 10.27 Connected components.

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## 10.6 Some Morphological Algorithm

### • 10.6.3 Skeletonization

TABLE 10.1 Operations used to construct the skeleton.

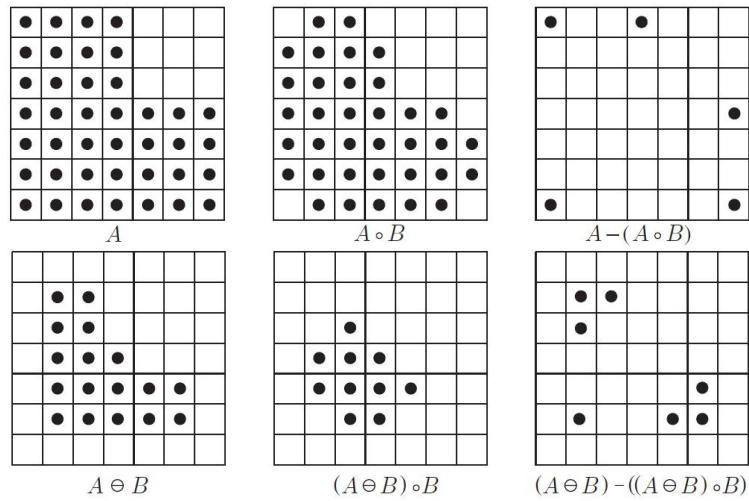
Erosions	Openings	Set differences
$A$	$A \circ B$	$A - (A \circ B)$
$A \ominus B$	$(A \ominus B) \circ B$	$(A \ominus B) - ((A \ominus B) \circ B)$
$A \ominus 2B$	$(A \ominus 2B) \circ B$	$(A \ominus 2B) - ((A \ominus 2B) \circ B)$
$A \ominus 3B$	$(A \ominus 3B) \circ B$	$(A \ominus 3B) - ((A \ominus 3B) \circ B)$
$\vdots$	$\vdots$	$\vdots$
$A \ominus kB$	$(A \ominus kB) \circ B$	$(A \ominus kB) - ((A \ominus kB) \circ B)$

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FIGURE 10.28



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FIGURE 10.28

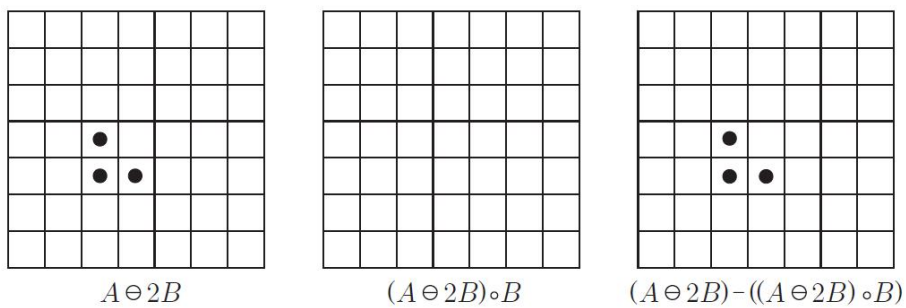


FIGURE 10.28 Skeletonization.

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## FIGURE 10.29

This method of skeletonization is called **Lantuéjoul's method**. For details see Serra [33]

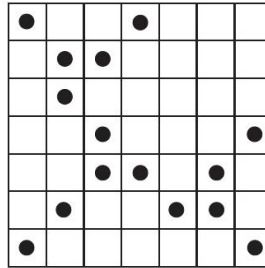


FIGURE 10.29 The final skeleton.

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## FIGURE 10.30

```
function skel = imskel(image,str)
% IMSKEL(IMAGE,STR) - Calculates the skeleton of binary image IMAGE using
% structuring element STR. This function uses Lantejoul's algorithm.
%
skel=zeros(size(image));
e=image;
while (any(e(:))),
    o=imopen(e,str);
    skel=skel | (e&~o);
    e=imerode(e,str);
end
```

FIGURE 10.30 A simple program for computing skeletons.

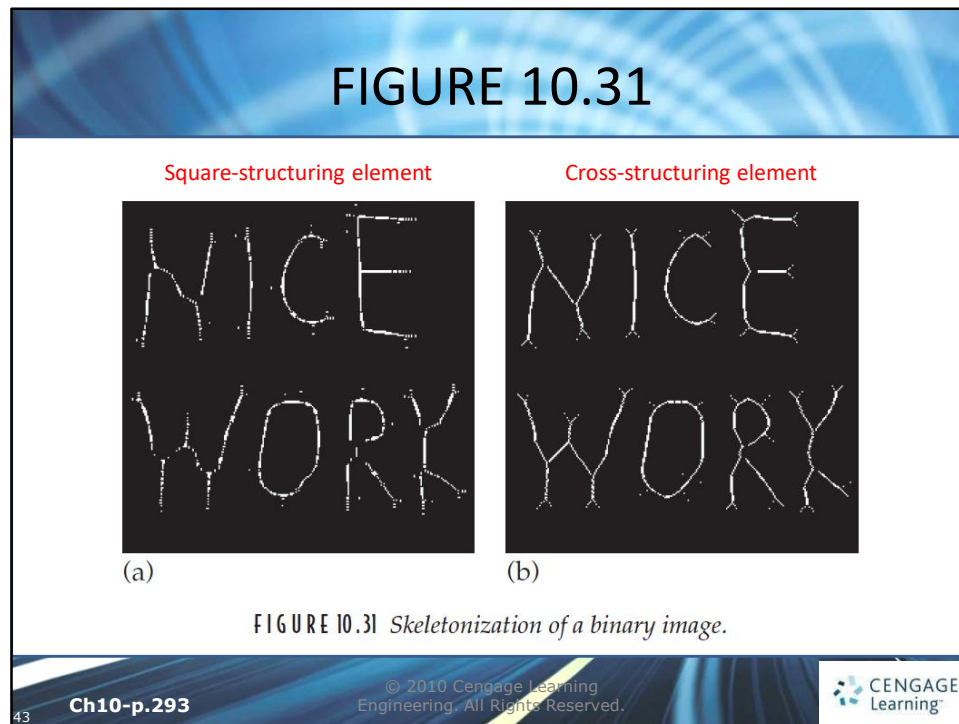
```
>> nk=imskel(n,sq);
>> imshow(nk)
>> nk2=imskel(n,cr);
>> figure,imshow(nk2)
```

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## 10.7 A Note on MATLAB's bwmorph Function

- Based on **lookup tables** (ch11)
  - ✓ Consider the  $3 \times 3$  neighborhood of a pixel
  - ✓ Since each pixel in the neighborhood can have only two values, there are  $2^9 = 512$  different possible neighborhoods
  - ✓ Define a morphological operation to be a function that maps these neighborhoods to the values 0 and 1

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## 10.8 Grayscale Morphology

- **Dilation**

- ✓ 1. Find the  $3 \times 3$  neighborhood  $N_p$  of  $p$
- ✓ 2. Compute the matrix  $N_p + B$
- ✓ 3. Find the maximum of that result

- **Summary**

$$(A \ominus B)(x, y) = \min\{A(x + s, y + t) - B(s, t), (s, t) \in D_B\},$$

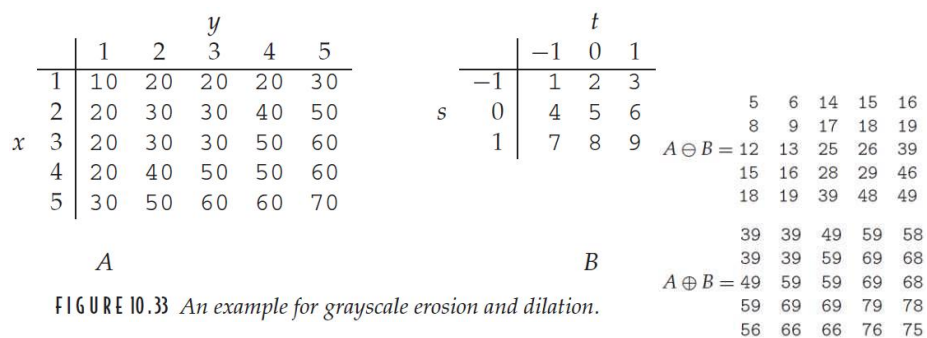
$$(A \oplus B)(x, y) = \max\{A(x + s, y + t) + B(s, t), (s, t) \in D_B\}.$$

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### FIGURE 10.33



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## 10.8 Grayscale Morphology

- The arbitrary parameter of `strel` allows us to create a structuring element containing any values we like
- `ones(3,3)` provides the neighborhood; the second matrix provides the values

```
>> str=strel('arbitrary',ones(3,3),[1 2 3;4 5 6;7 8 9])
str =
Nonflat STREL object containing 9 neighbors.
Neighborhood:
    1    1    1
    1    1    1
    1    1    1
Height:
    1    2    3
    4    5    6
    7    8    9
```

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## 10.8 Grayscale Morphology

```
>> str2=strel('arbitrary',ones(3,3),[9 8 7;6 5 4;3 2 1])
str2 =
Nonflat STREL object containing 9 neighbors.
Neighborhood:
    1    1    1
    1    1    1
    1    1    1
Height:
    9    8    7
    6    5    4
    3    2    1
>> imdilate(A,str2)
ans =
    39    39    49    59    58
    39    39    59    69    68
    49    59    59    69    68
    59    69    69    79    78
    56    66    66    76    75
```

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FIGURE 10.34

```
>> c=imread('caribou.tif');
>> str=strel('square',5)
>> cd=imdilate(c,str);
>> ce=imerode(c,str);
>> imshow(cd),figure,imshow(ce)
```

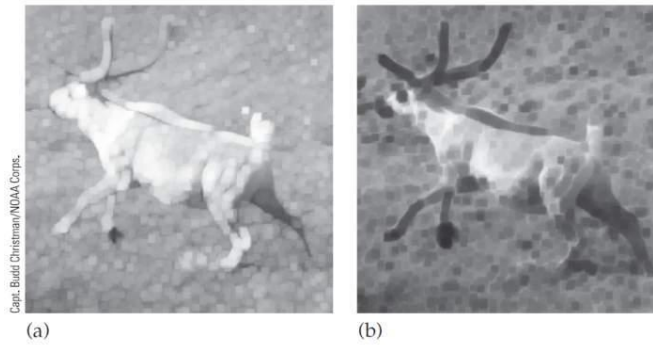


FIGURE 10.34 Morphology. (a) Dilation. (b) Erosion.

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FIGURE 10.35

### • OPENING AND CLOSING

```
>> co=imopen(c,str);
>> cc=imclose(c,str);
>> imshow(co),figure,imshow(cc)
```

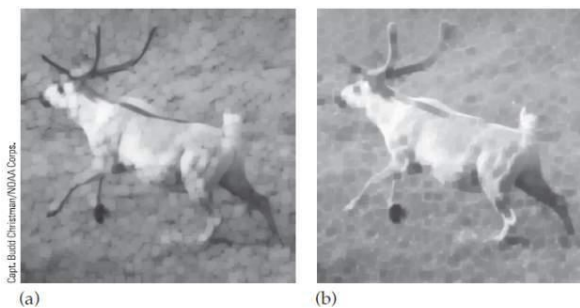


FIGURE 10.35 Grayscale opening and closing. (a) Opening. (b) Closing.

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## 10.9 Applications of Grayscale Morphology

### • 10.9.1 Edge Detection

✓ morphological gradient

$$(A \oplus B) - (A \ominus B)$$

```
>> str1=strel('square',3);
>> str2=strel('square',5);
>> ce1=imerode(c,str1);
>> ce2=imerode(c,str2);
>> cd1=imdilate(c,str1);
>> cd2=imdilate(c,str2);
>> cg1=imsubtract(cd1,ce1);
>> cg2=imsubtract(cd2,ce2);
>> imshow(cg1),figure,imshow(cg2)
```

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## FIGURE 10.36

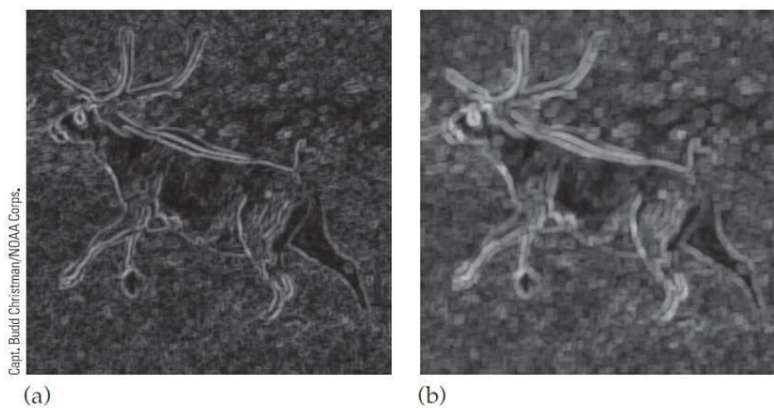


FIGURE 10.36 Use of the morphological gradient.

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## FIGURE 10.37

- **Noise Removal**

```
>> cn=imnoise(c,'salt & pepper');  
>> cf=imclose(imopen(cn,str),str);  
>> imshow(cn),figure,imshow(cf)
```

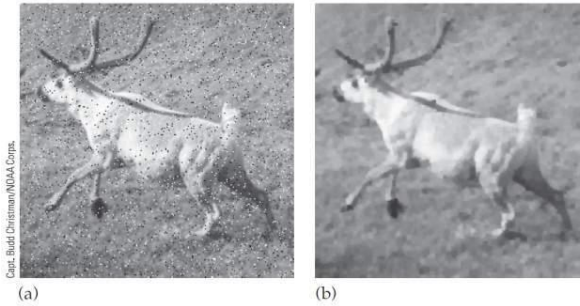


FIGURE 10.37 Use of morphological filtering to remove noise.