



# EECE 4353 Image Processing

Lecture Notes on Mathematical Morphology:  
Binary Images

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# What is Mathematical Morphology?

*It is:*

- nonlinear,
- built on Minkowski set theory,
- part of the theory of finite lattices,
- for image analysis based on shape,
- extremely useful, yet not often used.



# Uses of Mathematical Morphology

- image enhancement
- image segmentation
- image restoration
- edge detection
- texture analysis
- particle analysis
- feature generation
- skeletonization
- shape analysis
- image compression
- component analysis
- curve filling
- general thinning
- feature detection
- noise reduction
- space-time filtering



# Notation and Image Definitions

An image is a mapping,  $\mathbf{I}$ , from a set,  $S_p$ , of pixel coordinates to a set,  $G$ , of values such that for every coordinate vector,  $\mathbf{p} = (r,c)$  in  $S_p$ , there is a value  $I(\mathbf{p})$  drawn from  $G$ .  $S_p$  is also called the *image plane*.

A *binary image* has only 2 values. That is,  $G = \{v_{fg}, v_{bg}\}$ , where  $v_{fg}$  is called the *foreground* value and  $v_{bg}$  is called the *background* value.

Often, in the literature, the foreground value is  $v_{fg} = 0$  and the background is  $v_{bg} = -\infty$ . (That is to be consistent with the mathematical morphology of intensity images.) Other possibilities are  $\{v_{fg}, v_{bg}\} = \{0, \infty\}$ ,  $\{0, 1\}$ ,  $\{1, 0\}$ ,  $\{0, 255\}$ , and  $\{255, 0\}$ .

In this lecture  $\{v_{fg}, v_{bg}\} = \{255, 0\}$ , or  $\{v_{fg}, v_{bg}\} = \{0, 255\}$ , although the foreground is often displayed in different colors for contrast.



# Notation and Image Definitions

The foreground of binary image  $\mathbf{I}$  is

$$FG\{\mathbf{I}\} = \left\{ \mathbf{I}(\mathbf{p}), \mathbf{p} = (r, c) \in S_p \mid \mathbf{I}(\mathbf{p}) = v_{fg} \right\},$$

$S_p$  is the set of all pixel locations in the image, i.e., the pixel grid, or the support of the image.

i.e. the set of locations,  $\mathbf{p}$ , where  $\mathbf{I}(\mathbf{p}) = v_{fg}$ . Similarly, the background is

$$BG\{\mathbf{I}\} = \left\{ \mathbf{I}(\mathbf{p}), \mathbf{p} = (r, c) \in S_p \mid \mathbf{I}(\mathbf{p}) = v_{bg} \right\}.$$

The two sets form a partition of the image:

$$FG\{\mathbf{I}\} \cup BG\{\mathbf{I}\} = \mathbf{I} \text{ and } FG\{\mathbf{I}\} \cup BG\{\mathbf{I}\} = \emptyset.$$

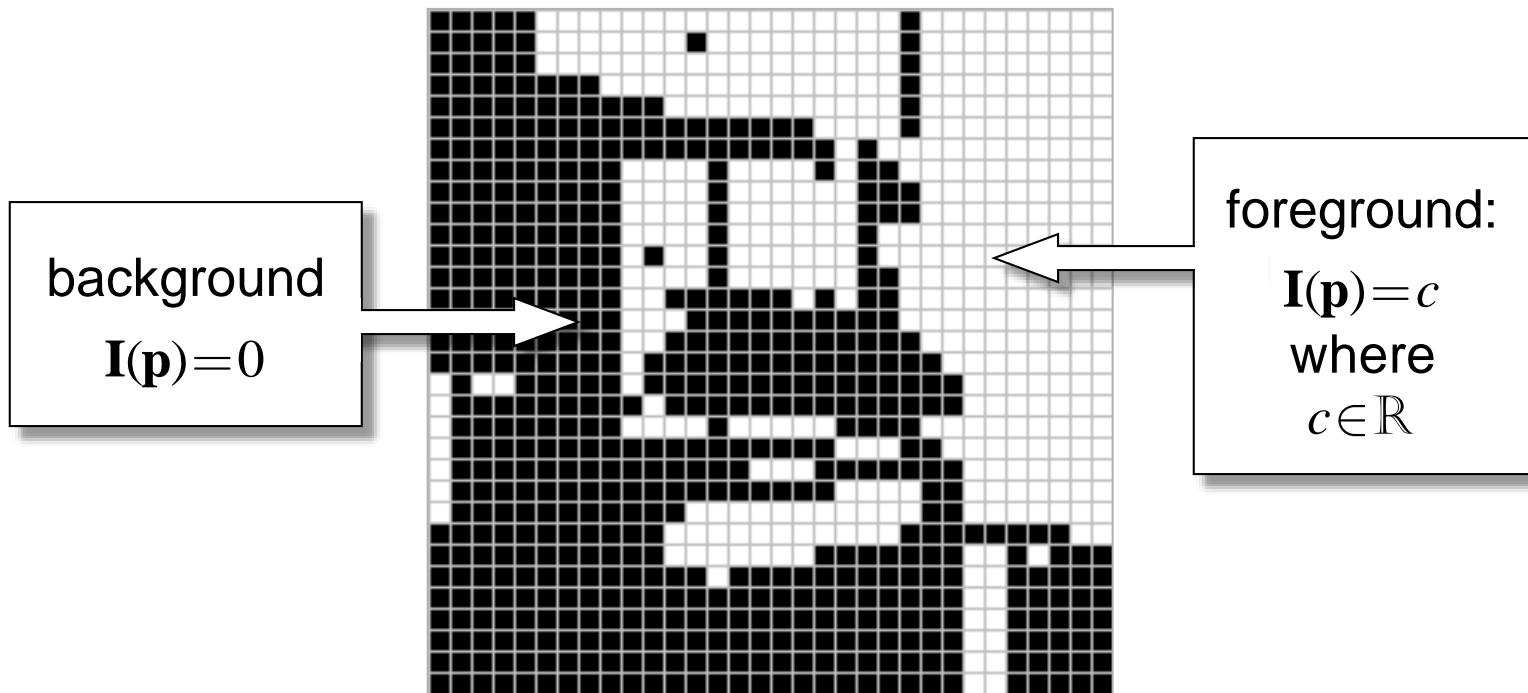
Moreover,

$$BG\{\mathbf{I}\} = \{FG\{\mathbf{I}\}\}^c \text{ and } FG\{\mathbf{I}\} = \{BG\{\mathbf{I}\}\}^c.$$

The background is the complement of the foreground and vice-versa.



# A Binary Image



This represents a digital image. Each square is one pixel.



# Support of an Image

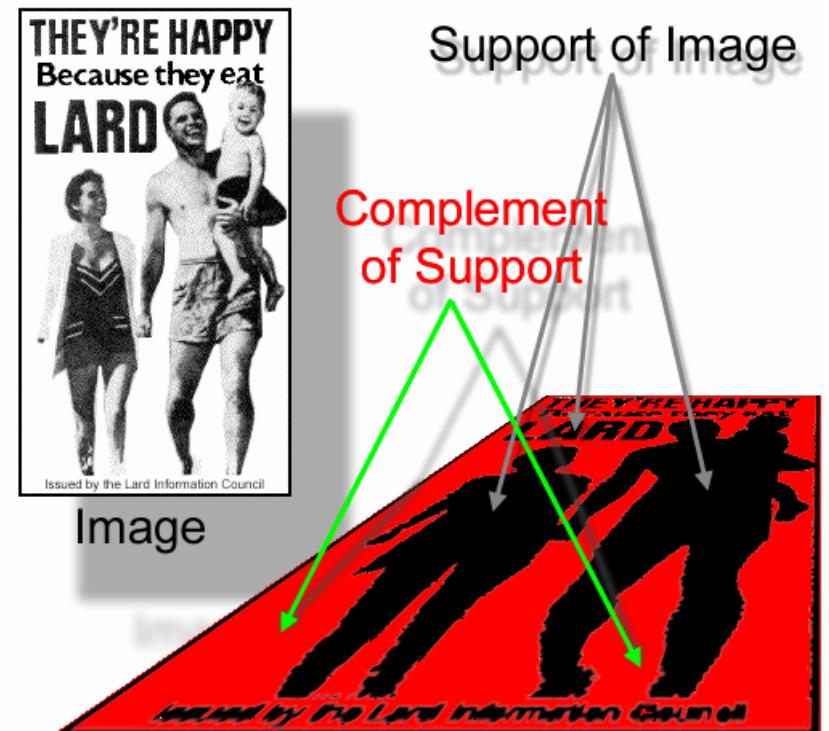
The support of a binary image is often defined as the set of locations of the foreground pixels. *I.e.*,

$$\text{supp}\{\mathbf{I}\} = \left\{ \mathbf{p} = (r, c) \in S_p \mid \mathbf{I}(\mathbf{p}) = v_{fg} \right\}.$$

In that case, the complement of the support is the set of background pixel locations within the image plane.

$$\{\text{supp}\{\mathbf{I}\}\}^c = \left\{ \mathbf{p} = (r, c) \in S_p \mid \mathbf{I}(\mathbf{p}) = v_{bg} \right\}.$$

But other times, the support is defined as the set of *all* pixel locations in the image both foreground and background.

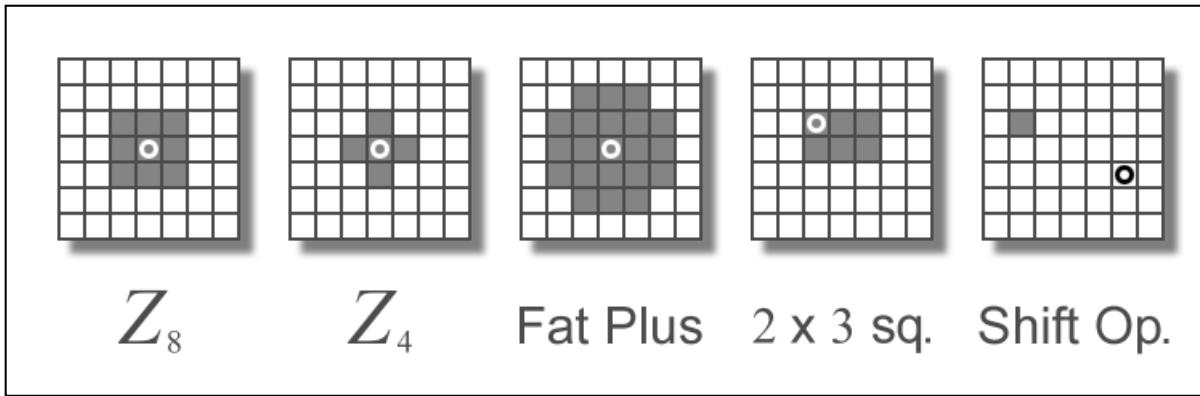




# Structuring Element (SE)

A structuring element is a small image – used as a moving window – whose support delineates pixel neighborhoods in the image plane.

Example SEs



It can be of any shape, size, or connectivity (more than 1 piece, have holes). In the figure the circles mark the location of the structuring element's origin which can be placed anywhere relative to its support.



# Structuring Element

Let  $\mathbf{I}$  be an image and  $\mathbf{Z}$  a SE.

$\mathbf{Z} + \mathbf{p}$  means that  $\mathbf{Z}$  is moved so that its origin coincides with location  $\mathbf{p}$  in  $S_p$ .

$\mathbf{Z} + \mathbf{p}$  is the *translate* of  $\mathbf{Z}$  to location  $\mathbf{p}$  in  $S_p$ .

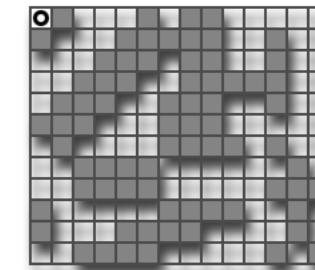
The set of locations in the image delineated by  $\mathbf{Z} + \mathbf{p}$  is called the **Z-neighborhood** of  $\mathbf{p}$  in  $\mathbf{I}$  denoted  $N\{\mathbf{I}, \mathbf{Z}\}(\mathbf{p})$ .

$S_p$  is the set of all pixel locations in the image.

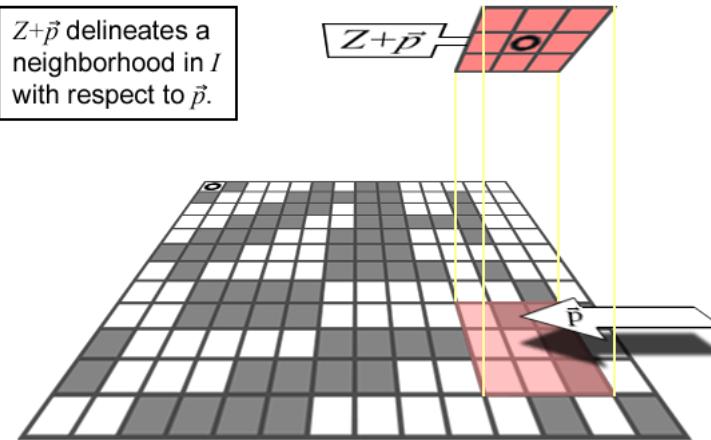
Image,  $\mathbf{I}$ .  
Origin is marked o.



Structuring Element,  $\mathbf{Z}$ .  
Origin is marked o.



$\mathbf{Z} + \vec{p}$  delineates a neighborhood in  $\mathbf{I}$  with respect to  $\vec{p}$ .





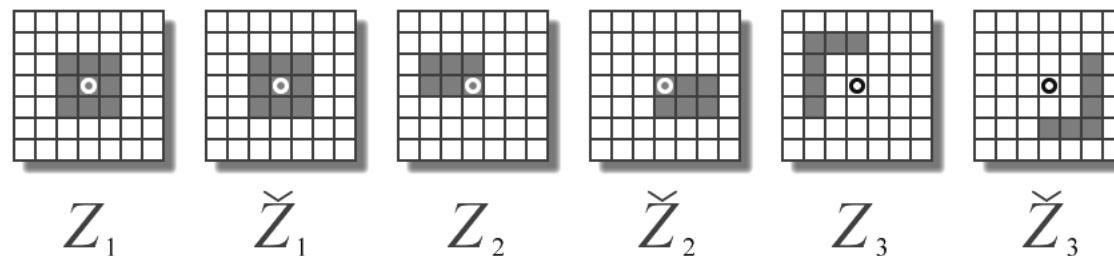
# Reflected Structuring Elements

Let  $\mathbf{Z}$  be a SE and let  $\mathcal{S}$  be the square of pixel locations that contains the set  $\{(r,c), (-r,-c) \mid (r,c) \in \text{supp}(\mathbf{Z})\}$ . Then

$$\breve{\mathbf{Z}}(\rho, \chi) = \mathbf{Z}(-\rho, -\chi) \quad \text{for all } (\rho, \chi) \in \mathcal{S}.$$

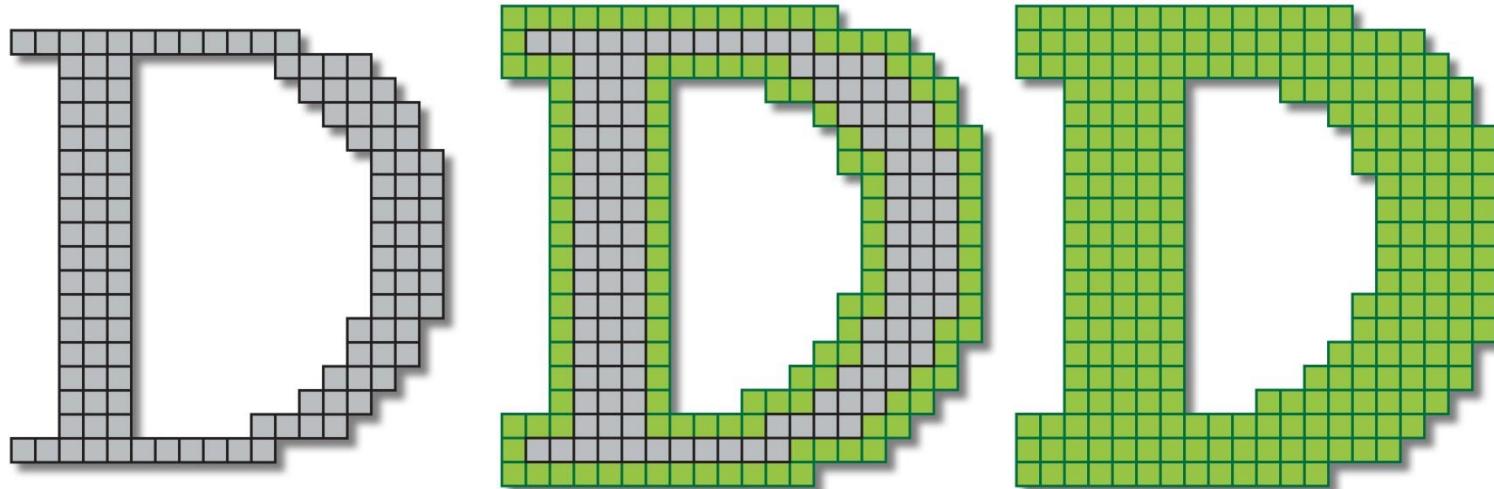
is the reflected structuring element.

$\breve{\mathbf{Z}}$  is  $\mathbf{Z}$  rotated by  $180^\circ$  around its origin.





# Dilation





# Dilation of Binary Images

There are a number of equivalent definitions of dilation. Three of them that apply to binary images are:

---

$$\mathbf{I} \oplus \mathbf{Z} = \left\{ \mathbf{p} \in S_p \left| \left[ (\check{\mathbf{Z}} + \mathbf{p}) \cap \text{fg}\{\mathbf{I}\} \right] \neq \emptyset \right. \right\}.$$

The set of all pixel locations,  $\mathbf{p}$ , in the image plane where the intersection of  $\check{\mathbf{Z}} + \mathbf{p}$  with the foreground of  $\mathbf{I}$  is not empty.

---

$$\mathbf{I} \oplus \mathbf{Z} = \bigcup_{\mathbf{p} \in \text{fg}\{\mathbf{I}\}} (\mathbf{Z} + \mathbf{p}).$$

The union of copies of the SE, one translated to each pixel location in the foreground of the image.

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$$\mathbf{I} \oplus \mathbf{Z} = \bigcup_{\mathbf{p} \in \text{fg}\{\mathbf{Z}\}} (\mathbf{I} + \mathbf{p}).$$

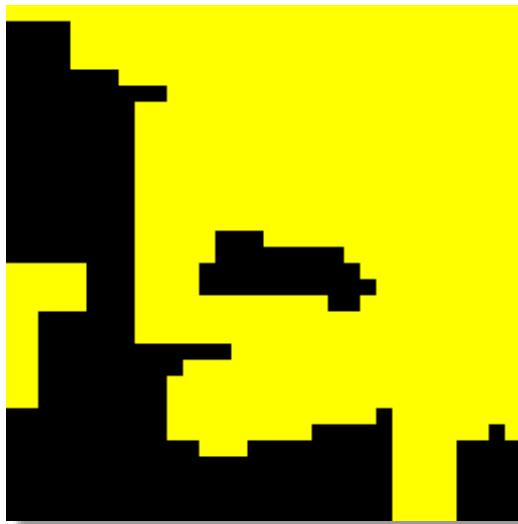
The union of copies of the image, one translated to each pixel location in the foreground of the SE.

---



# Dilation

The locus of pixels  $\mathbf{p} \in S_P$  such that  $(\mathbf{Z} + \mathbf{p}) \cap I \neq \emptyset$ .



dilated image



original / dilation



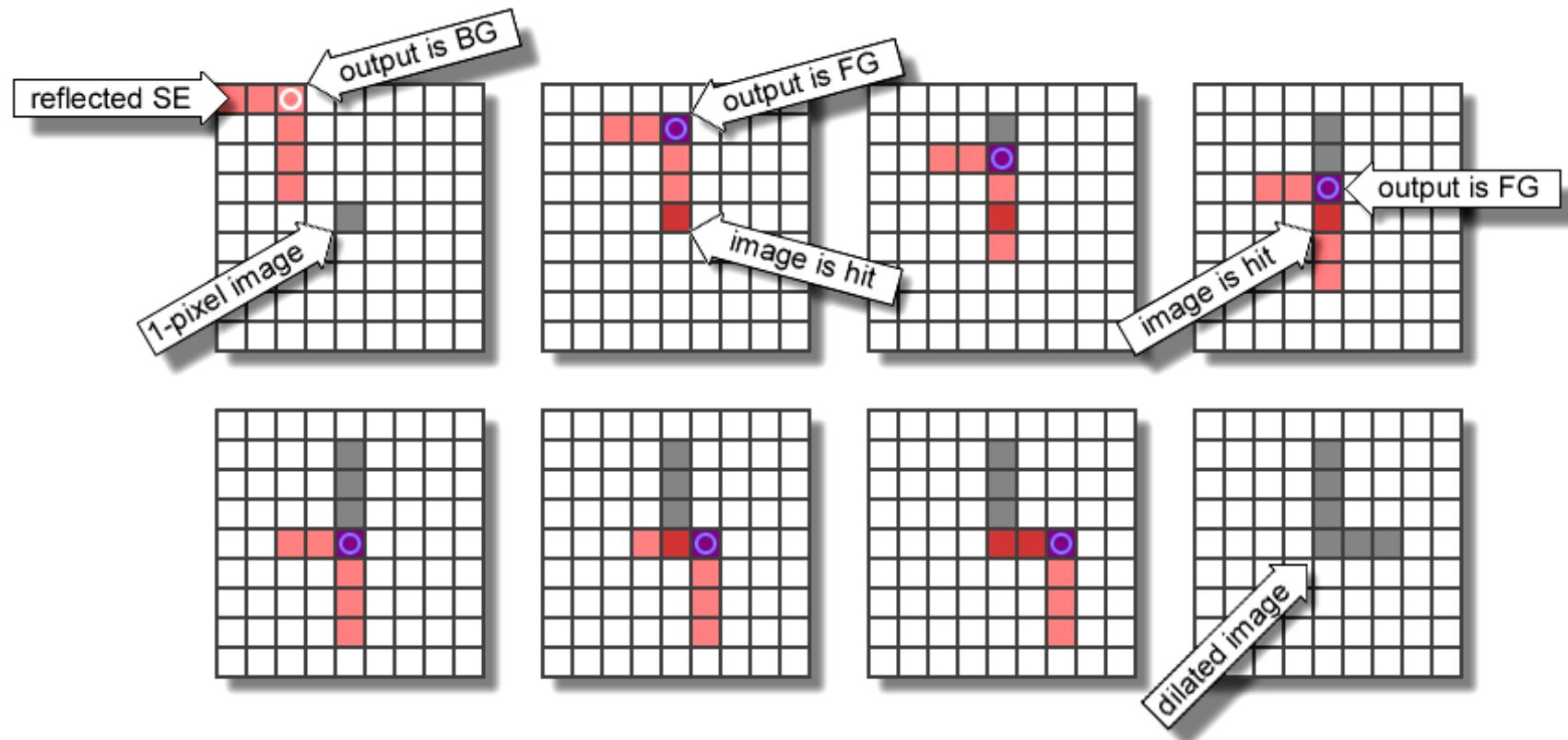
original image

$$SE = Z_8$$

This is a piece of a larger image. Boundary effects are not apparent



# Dilation using a Reflected SE





# Fast Computation of Dilation

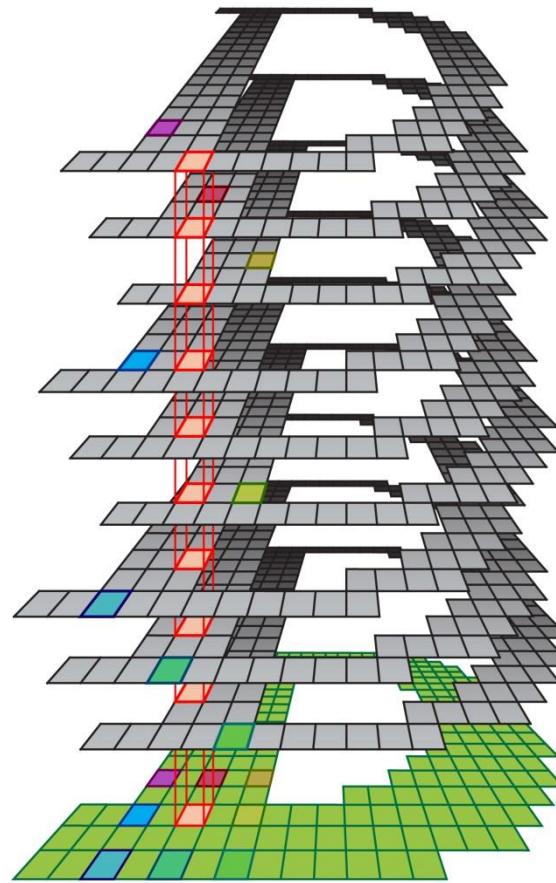
The fastest way to compute *binary* dilation is to use the union-of-translates-of-the-image definition. That is, use

$$\mathbf{J} = \mathbf{I} \oplus \mathbf{Z} = \bigcup_{\mathbf{q} \in \mathbf{Z}} \mathbf{I} + \mathbf{q}.$$

Assume the dimensions of  $\mathbf{I}$  are  $R \times C$ , the dimensions of  $\mathbf{Z}$  are  $N \times M$ , and  $\mathbf{Z}$ 's origin is offset from the upper left hand corner (ULHC) by  $\rho$  rows and  $\chi$  columns. Allocate a scratch image,  $\mathbf{T}$ , that is  $(R+N-1) \times (C+M-1)$  and initialized to zeros. Then, for each FG pixel loc  $(v, u)$  in  $\mathbf{Z}$  (measured from the ULHC of  $\mathbf{Z}$ ) perform a logical OR between  $\mathbf{I}+(v, u)$  and  $\mathbf{T}$ . Put the results in  $\mathbf{T}$ . When done, copy to  $\mathbf{J}$  the  $R \times C$  subarray of  $\mathbf{T}$  starting at  $(\rho, \chi)$ .

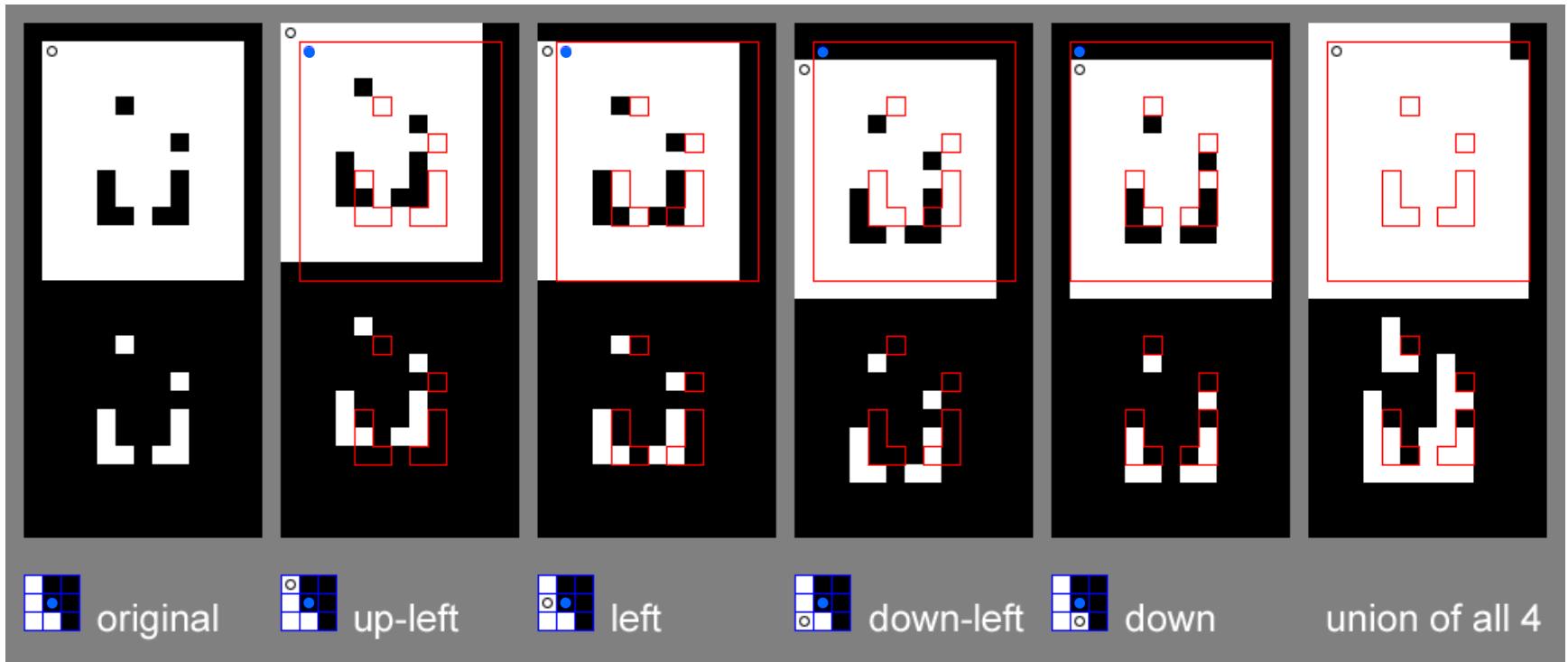


# Dilation through Image Shifting





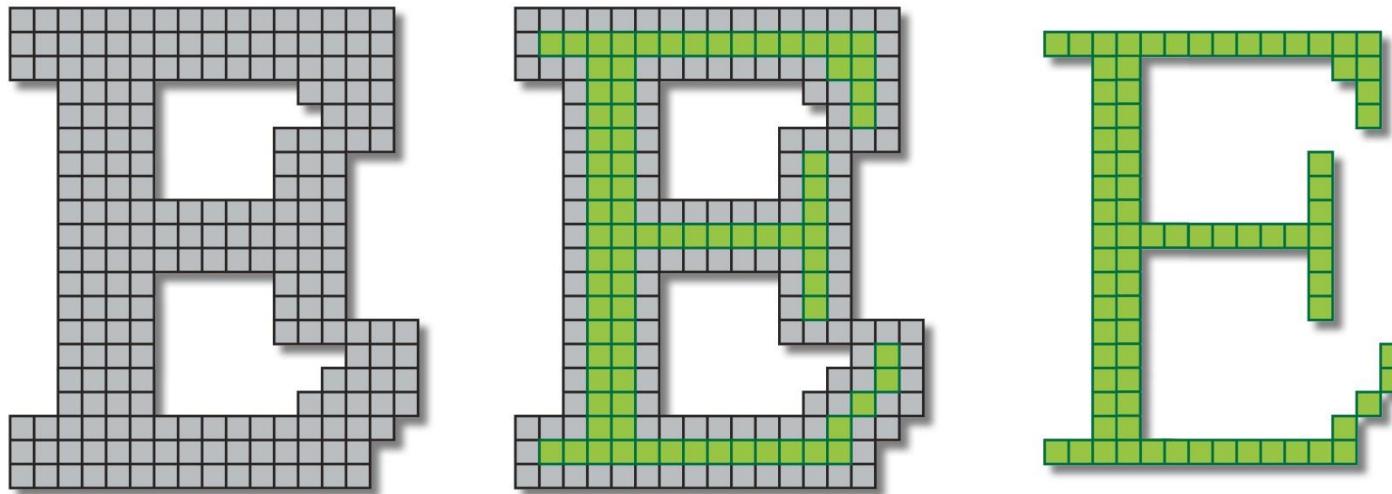
# Dilation through Image Shifting



The red outlines indicate the positions of the features in the original images.



# Erosion





# Erosion of Binary Images

There are a number of equivalent definitions of erosion. Three of them that apply to binary images are:

---

$$\mathbf{I} \ominus \mathbf{Z} = \left\{ \mathbf{p} \in S_p \mid (\mathbf{Z} + \mathbf{p}) \in \text{fg}\{\mathbf{I}\} \right\}.$$

The set of all pixel locations,  $\mathbf{p}$ , in the image plane where  $\mathbf{Z} + \mathbf{p}$  is contained in  $\text{fg}\{\mathbf{I}\}$ .

---

$$\mathbf{I} \ominus \mathbf{Z} = \bigcap_{\mathbf{p} \in \text{fg}\{\mathbf{I}\}} (\check{\mathbf{Z}} + \mathbf{p}).$$

The intersection of copies of the refl. SE, one translated to each pixel location in the foreground of the image.

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$$\mathbf{I} \ominus \mathbf{Z} = \bigcap_{\mathbf{p} \in \text{fg}\{\check{\mathbf{Z}}\}} (\mathbf{I} + \mathbf{p}).$$

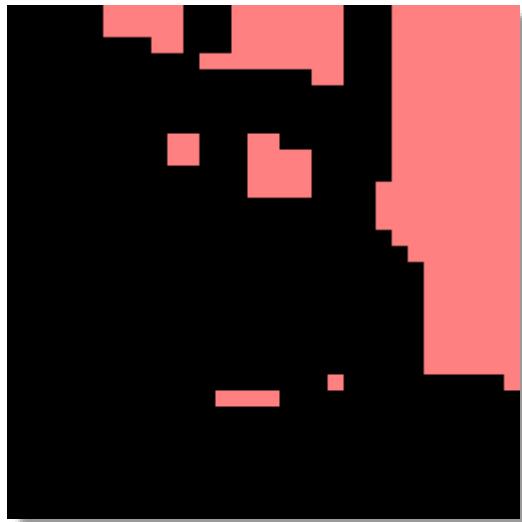
The intersection of copies of the image, one translated to each pixel location in the foreground of the refl. SE.

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# Erosion

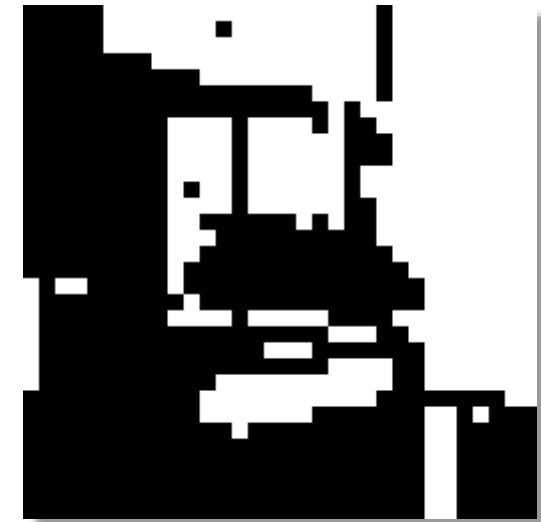
The locus of pixels  $\mathbf{p} \in S_p$  such that  $\mathbf{Z} + \mathbf{p} \subset I$ .



eroded image



erosion / original



original image

$$SE = \mathbf{Z}_8$$

This is a piece of a larger image. Boundary effects are not apparent



# Fast Computation of Erosion

The fastest way to compute *binary* erosion is to use the intersection-of-translates-of-the-image definition. That is, use

$$\mathbf{J} = \mathbf{I} \ominus \mathbf{Z} = \bigcap_{\mathbf{q} \in \text{fg}\{\check{\mathbf{Z}}\}} \mathbf{I} + \mathbf{q}.$$

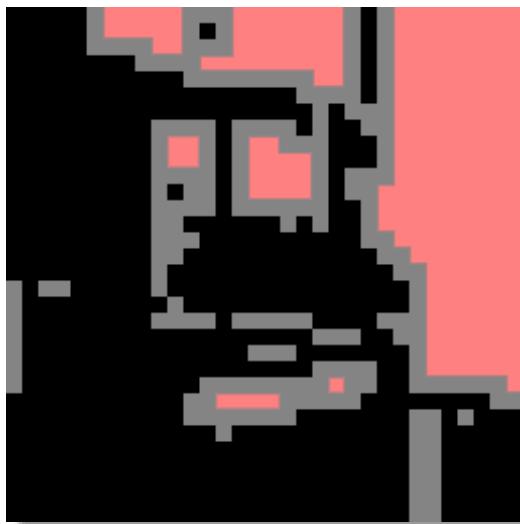
Assume the dimensions of  $\mathbf{I}$  are  $R \times C$ , the dimensions of  $\mathbf{Z}$  are  $N \times M$ , and  $\mathbf{Z}$ 's origin is offset from the upper left hand corner (ULHC) by  $\rho$  rows and  $\chi$  columns. Allocate a scratch image,  $\mathbf{T}$ , that is  $(R+N-1) \times (C+M-1)$  and initialized to  $\mathbf{I}$ . Rotate<sup>1</sup>  $\mathbf{Z}$  by 180°. Then, for each FG pixel loc  $(v, u)$  in  $\check{\mathbf{Z}}$  (measured from the ULHC of  $\check{\mathbf{Z}}$ ) perform a logical AND between  $\mathbf{I}+(v, u)$  and  $\mathbf{T}$ . Put the results in  $\mathbf{T}$ . When done, copy to  $\mathbf{J}$  the  $R \times C$  subarray of  $\mathbf{T}$  starting at  $(N-\rho, M-\chi)$ .

<sup>1</sup>In matlab the fastest way to rotate Z by 180° is  
`» Zrefl = flipud(fliplr(Z));`



# Comparison of Erosion and Dilation

original contains erosion



erosion / original

SE =  $Z_8$

This is a piece of a larger image. Boundary effects are not apparent

dilation contains original



erosion / original / dilation

original / dilation



# Erosion from Dilation / Dilation from Erosion

Dilation and erosion are duals of each other with respect to complementation:

$$\mathbf{I}^C \oplus \check{\mathbf{Z}} = \{\mathbf{I} \ominus \mathbf{Z}\}^C \quad \text{and} \quad \mathbf{I}^C \ominus \check{\mathbf{Z}} = \{\mathbf{I} \oplus \mathbf{Z}\}^C$$

That is, dilation with the reflected SE of the complement of a binary image is the complement of the erosion. Erosion with the reflected SE of the complement of the image is the complement of the dilation. It follows that,

$$\mathbf{I} \ominus \mathbf{Z} = \{\mathbf{I}^C \oplus \check{\mathbf{Z}}\}^C \quad \text{and} \quad \mathbf{I} \oplus \mathbf{Z} = \{\mathbf{I}^C \ominus \check{\mathbf{Z}}\}^C$$

erosion can be performed with dilation and vice versa. That implies that only one or the other must be implemented directly.

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# Opening and Closing

Opening is erosion by  $\mathbf{Z}$  followed by dilation by  $\mathbf{Z}$ .

$$\mathbf{I} \circ \mathbf{Z} = (\mathbf{I} \ominus \mathbf{Z}) \oplus \mathbf{Z}$$

The opening is the best approximation of the image FG that can be made from copies of the SE, given that the opening is contained in the original.  $\mathbf{I} \circ \mathbf{Z}$  contains no FG features that are smaller than the SE.

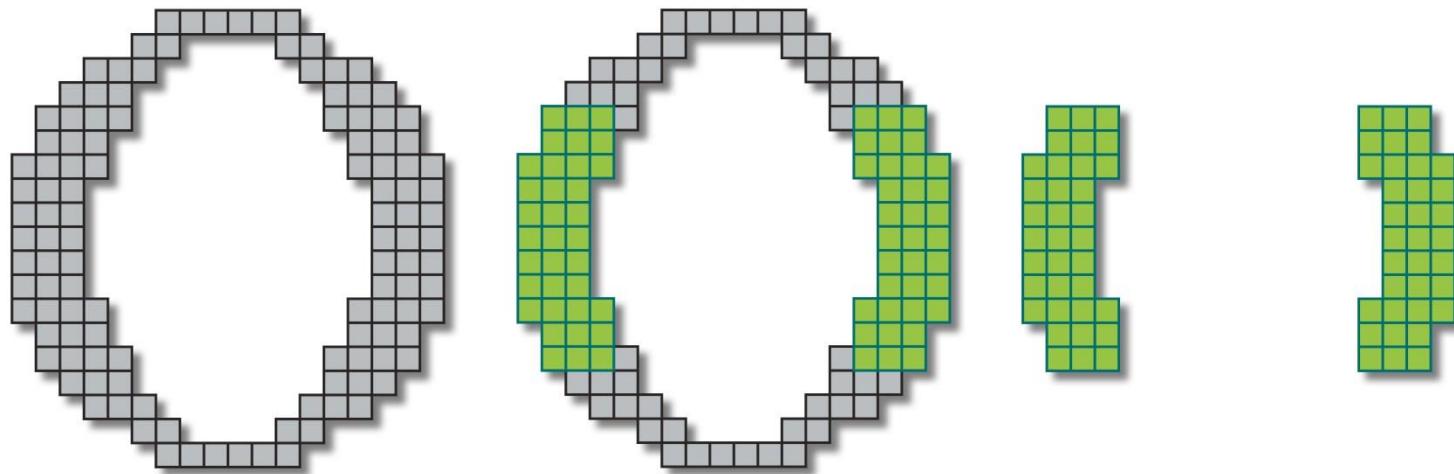
Closing is dilation by  $\check{\mathbf{Z}}$  followed by erosion by  $\check{\mathbf{Z}}$ .

$$\mathbf{I} \bullet \mathbf{Z} = (\mathbf{I} \oplus \check{\mathbf{Z}}) \ominus \check{\mathbf{Z}}$$

The closing is the best approximation of the image BG that can be made from copies of the SE, given that the closing is contained in the image BG.  $\mathbf{I} \bullet \mathbf{Z}$  contains no BG features that are smaller than the SE.



# Opening



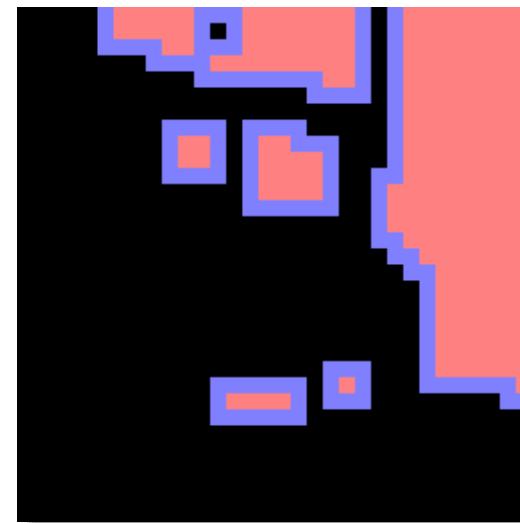


# Opening is Erosion Followed by Dilation

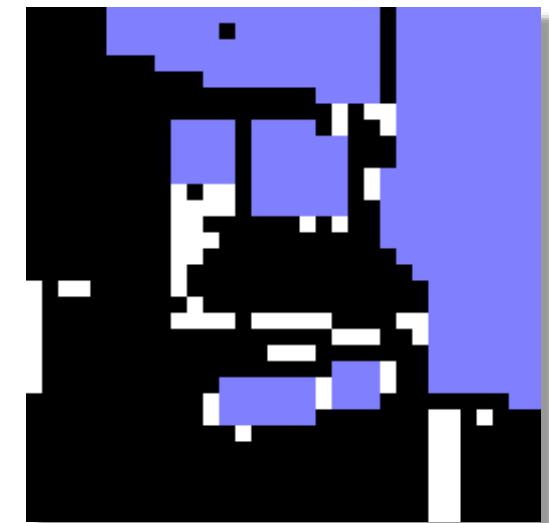
erode the original



dilate the erosion



dilated erosion



erosion / original

erosion / opening

opening / original

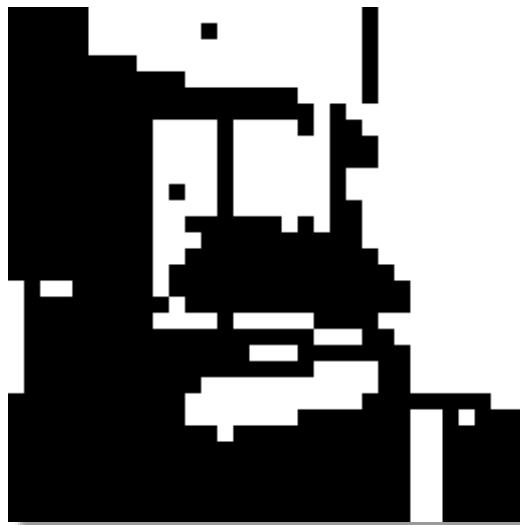
$$SE = Z_8$$

This is a piece of a larger image. Boundary effects are not apparent



# Opening is Erosion Followed by Dilation

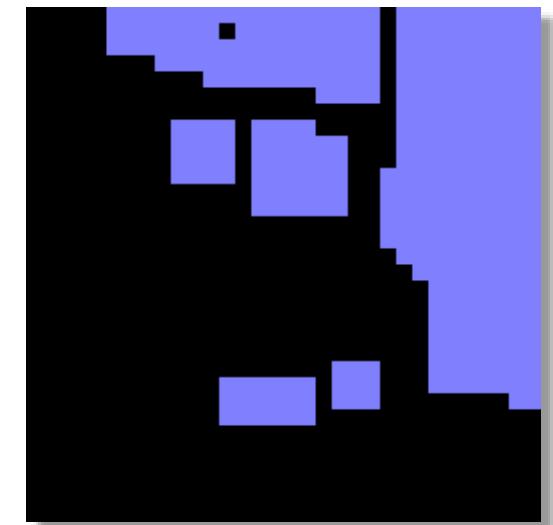
original image



eroded image



dilated erosion



original

erosion

opening

$$SE = Z_8$$

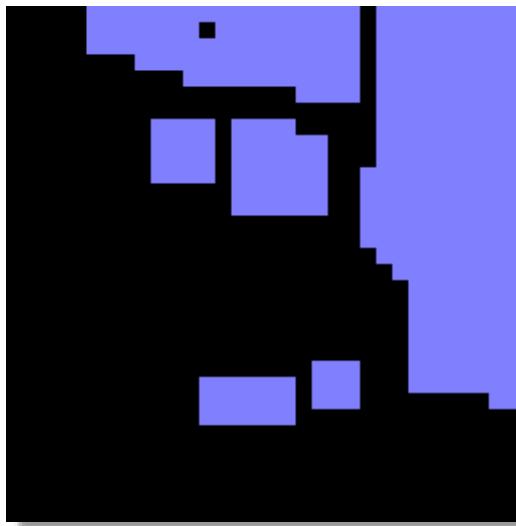
This is a piece of a larger image. Boundary effects are not apparent



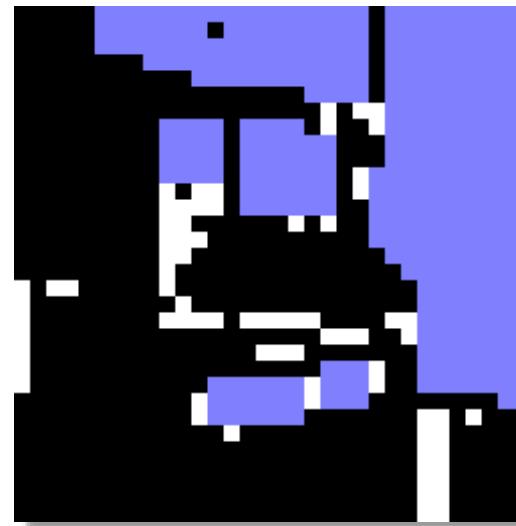
# Opening

$$SE = Z_8$$

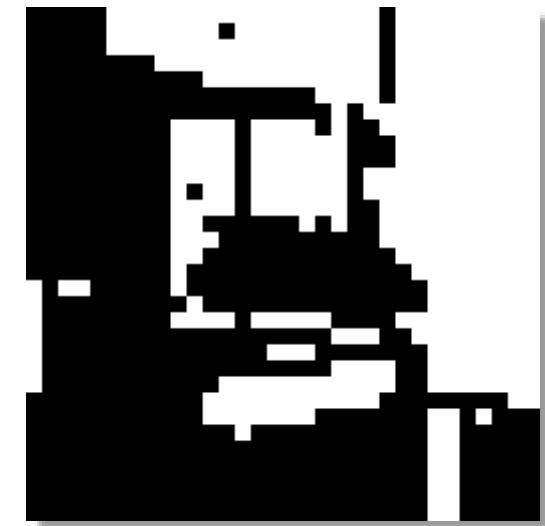
The union of translates of  $Z$  such that  $Z + p \subset I$ .



open image



opening / original

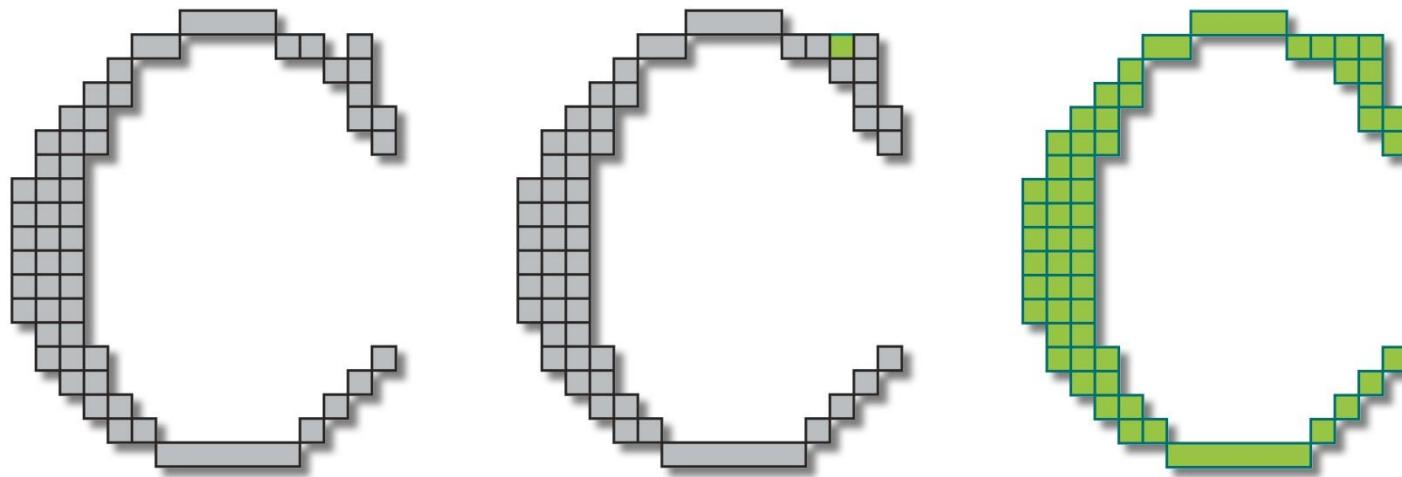


original image

The opening of  $I$  by  $Z$  is the best approximation of  $I$  that can be made by taking the union of translated copies of  $Z$ , subject to the constraint that the opening be contained by the original image.



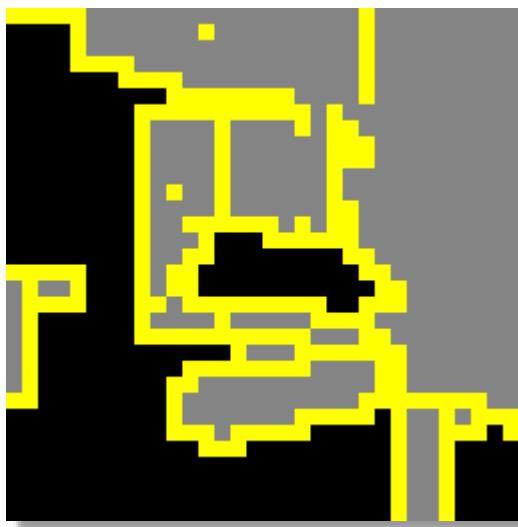
# Closing



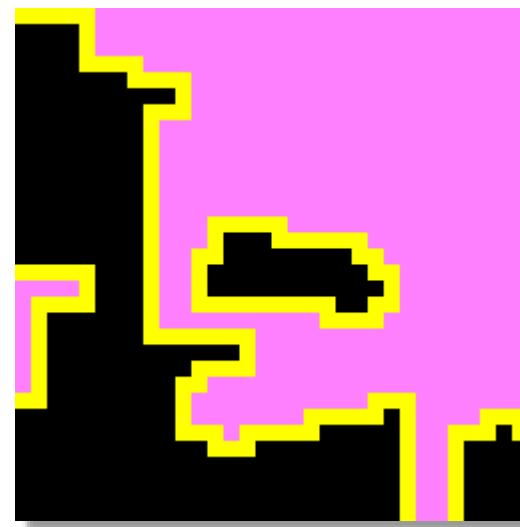


# Closing is Dilation Followed by Erosion<sup>1</sup>

original image



erode the dilation



to get the closing



original / dilation

$$SE = Z_8$$

closing / dilation

<sup>1</sup>using the reflected SE,  $\check{Z}$

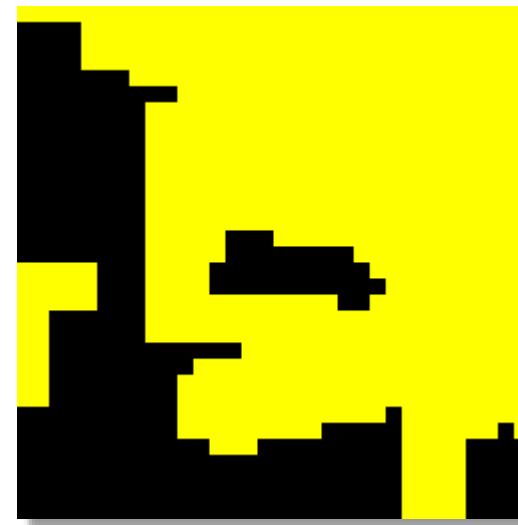


# Closing is Dilation Followed by Erosion<sup>1</sup>

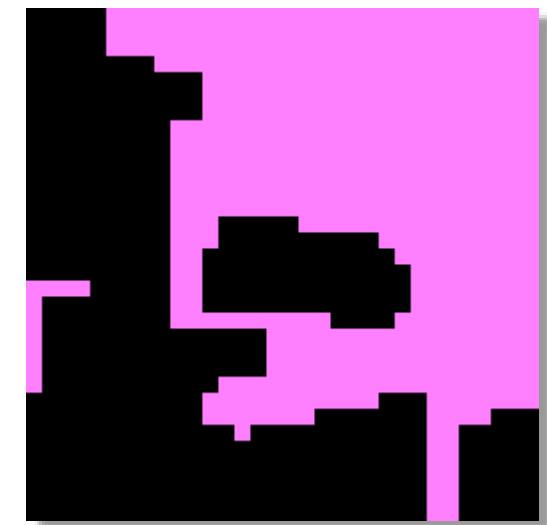
original image



dilated image



eroded dilation



original

dilation

closing

$$SE = Z_8$$

<sup>1</sup>using the reflected SE,  $\check{Z}$



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# Duality Relationships

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Erosion in terms of dilation:  $I \ominus Z = [I^C \oplus \check{Z}]^C$

Dilation in terms of erosion:  $I \oplus Z = [I^C \ominus \check{Z}]^C$

Opening in terms of closing:  $I \circ Z = [I^C \bullet Z]^C$

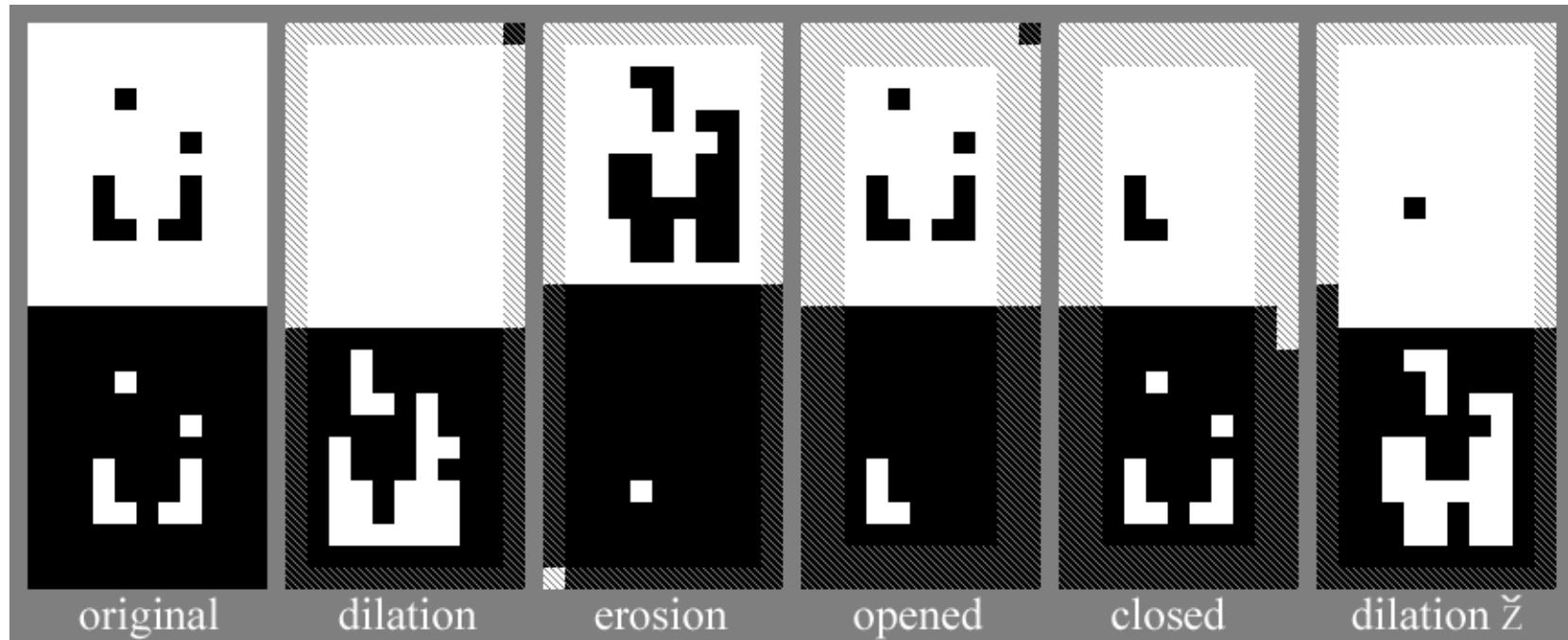
Closing in terms of opening:  $I \bullet Z = [I^C \circ Z]^C$

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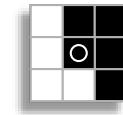
$I^C$  is the complement of  $I$  and  $\check{Z}$  is the reflected SE.



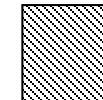
# Binary Ops with Asymmetric SEs



"L" shaped SE  
O marks origin



Foreground: white pixels  
Background: black pixels



Cross-hatched  
pixels are  
indeterminate.

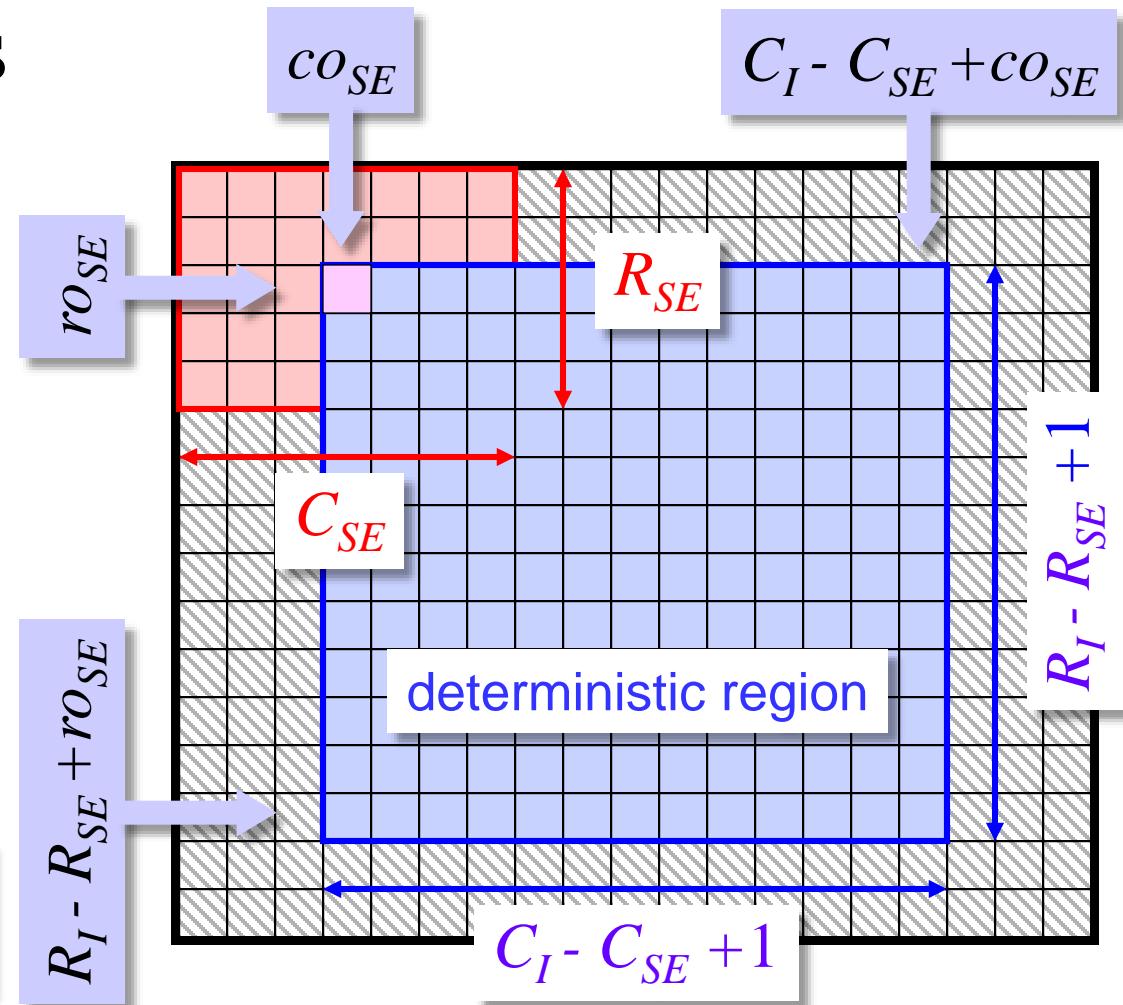


# Border Effects

## Erosion & Dilation

Since morph. ops. are neighborhood ops., there is a band of pixels around the border of the resultant image where the values are indeterminate.

The actual values of pixels in the indet. region depend on the specific algorithm used.

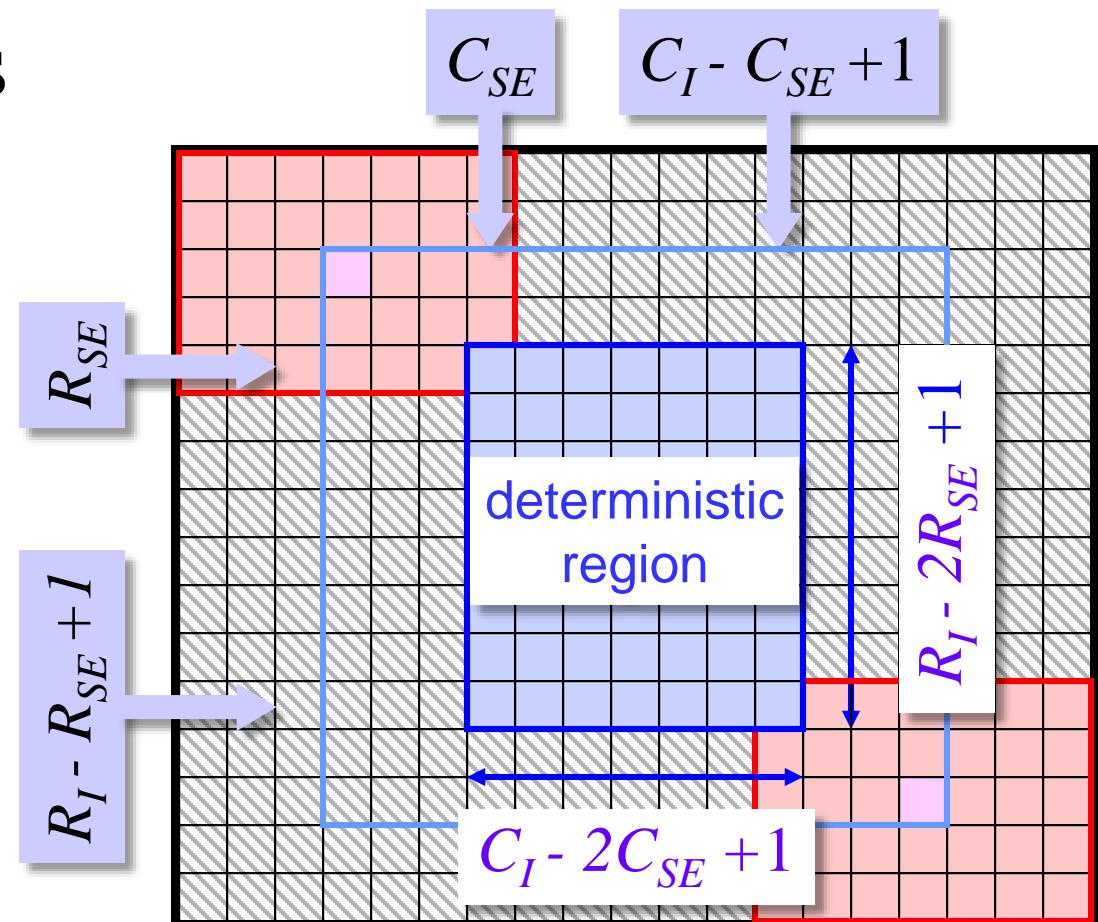




# Border Effects

## Opening & Closing

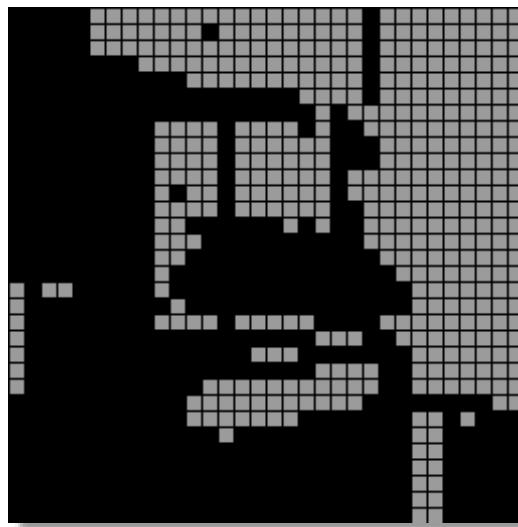
Since opening & closing iterate erosion & dilation, the boundaries of the deterministic region are 2x as far from the image border as are those of erosion or dilation.





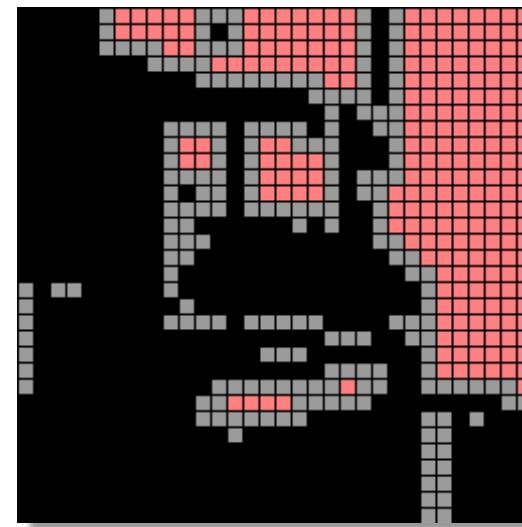
# Boundary Extraction

binary image



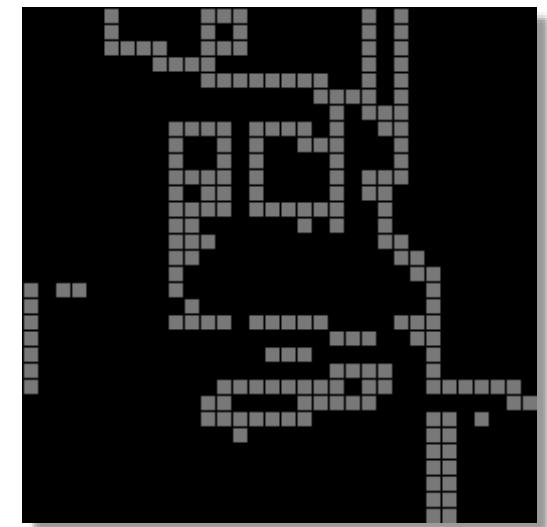
original

8-connected SE



erosion by square

4-conn inside bdry



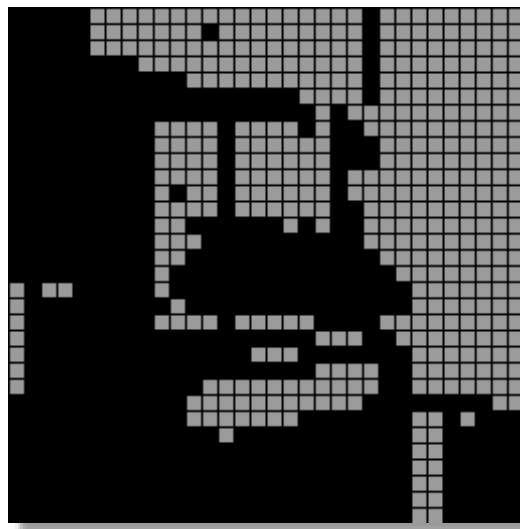
difference

This is a piece of a larger image. Boundary effects are not apparent

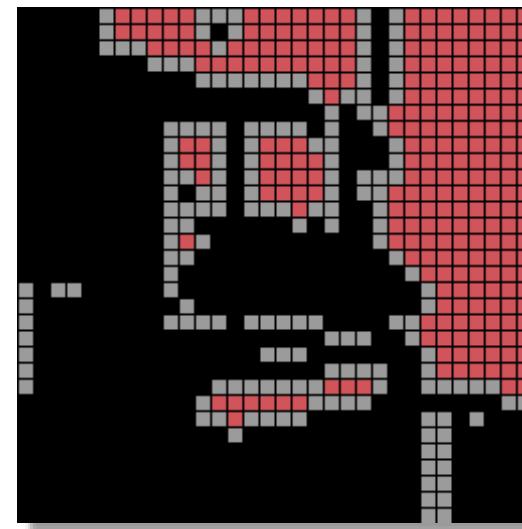


# Boundary Extraction

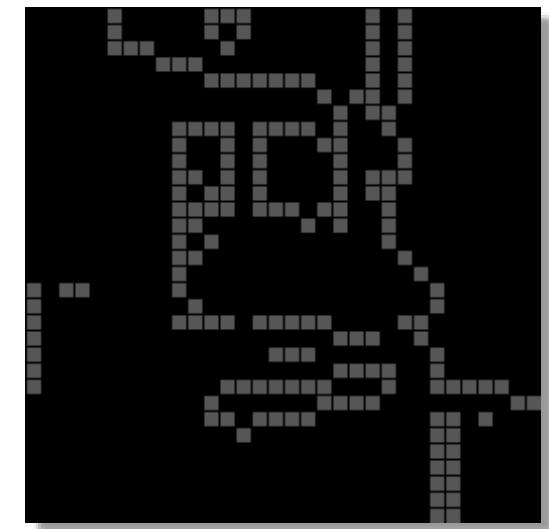
binary image



4-connected SE



8-conn inside bdry



original

erosion by plus

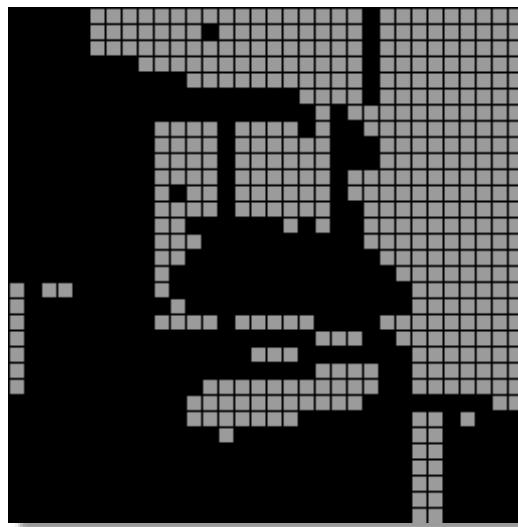
difference

This is a piece of a larger image. Boundary effects are not apparent



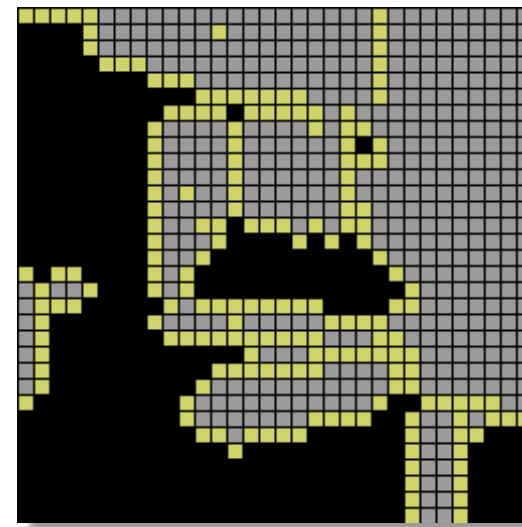
# Boundary Extraction

binary image



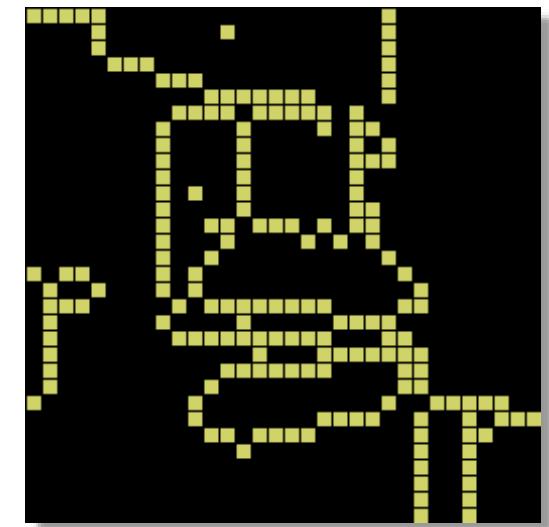
original

4-connected SE



dilation by plus

8-conn outside bdry



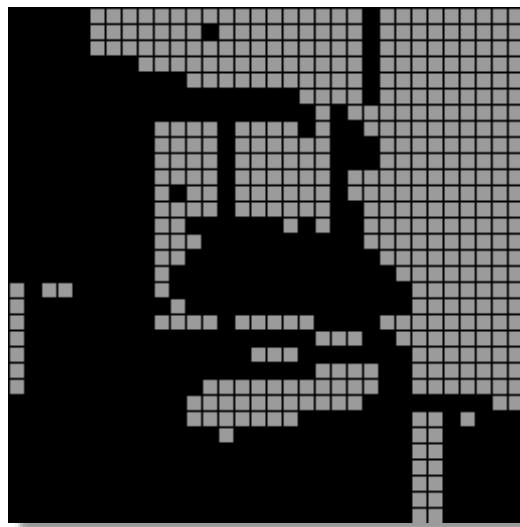
difference

This is a piece of a larger image. Boundary effects are not apparent



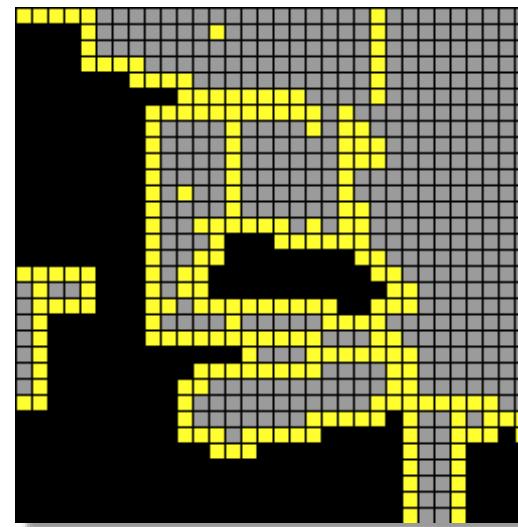
# Boundary Extraction

binary image



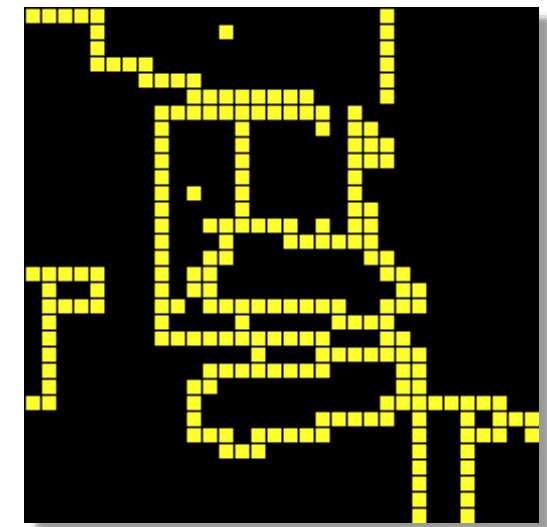
original

8-connected SE



dilation by square

4-conn outside bdry



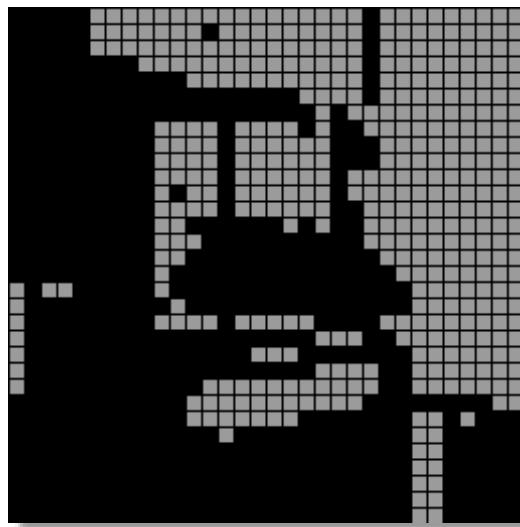
difference

This is a piece of a larger image. Boundary effects are not apparent



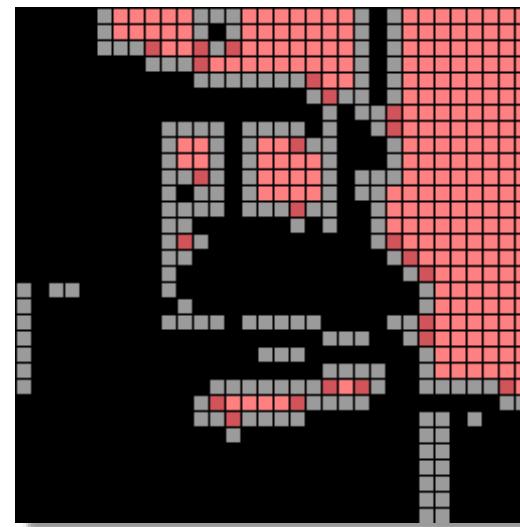
# Boundary Extraction

binary image



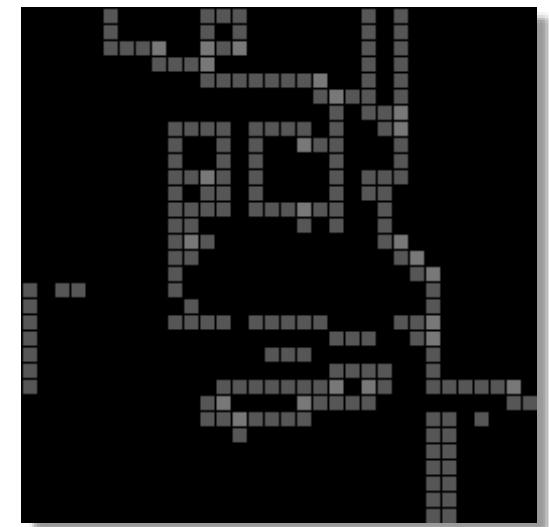
original

erosion by square is



in erosion by plus

8-conn inside bdry is



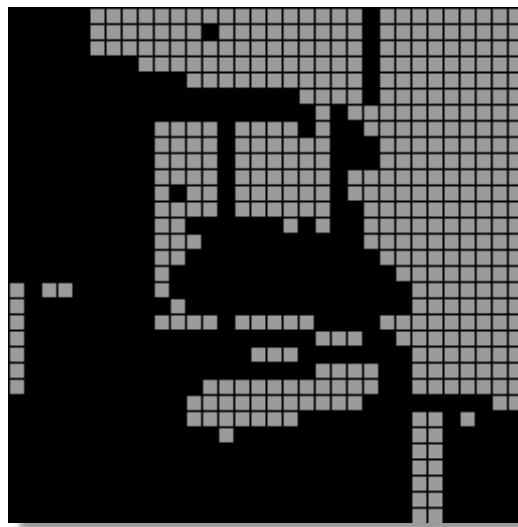
in 4-conn inside bdry

This is a piece of a larger image. Boundary effects are not apparent

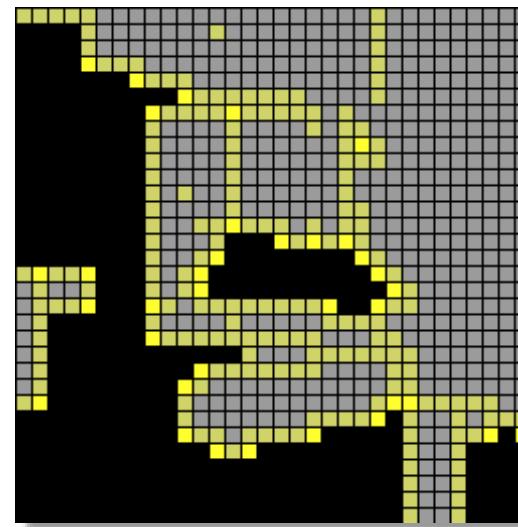


# Boundary Extraction

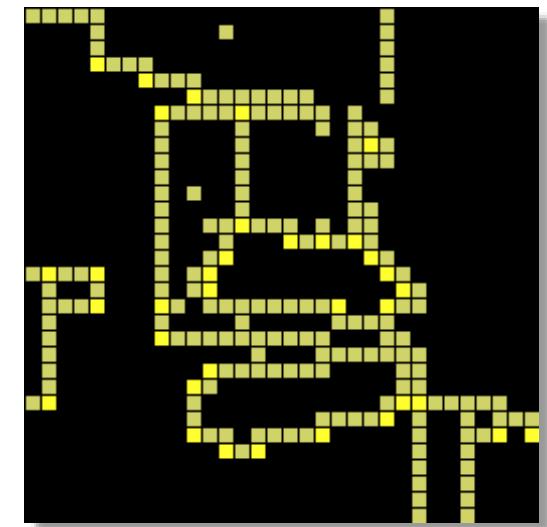
binary image



dilation by plus is



8-conn outside bdry is



original

in dilation by square

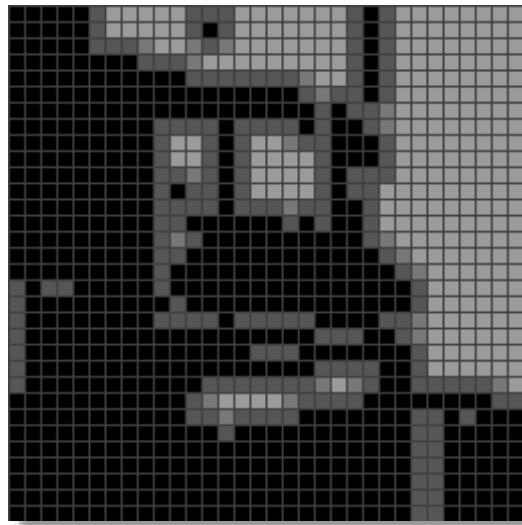
in 4-conn outside bdry

This is a piece of a larger image. Boundary effects are not apparent



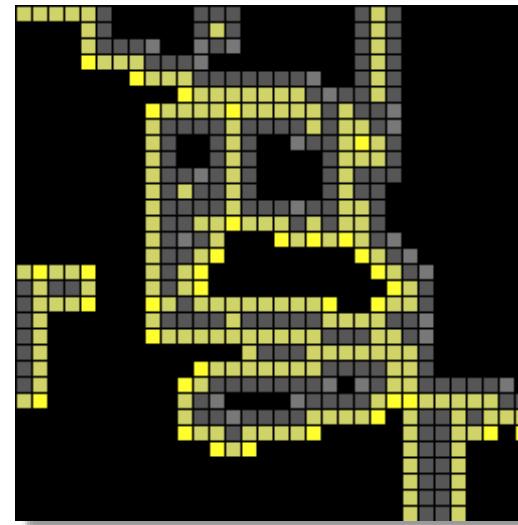
# Boundary Extraction

inside boundaries



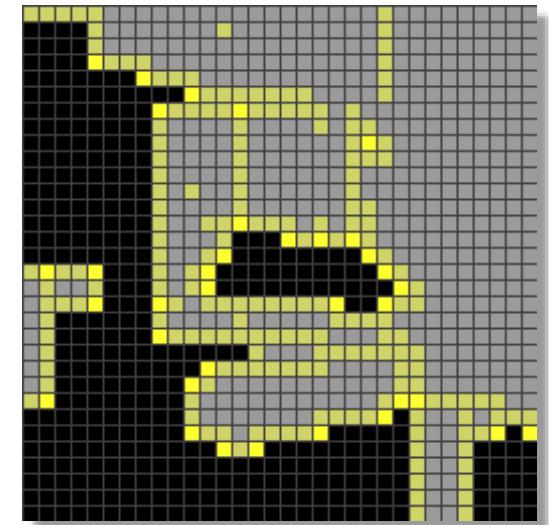
8-bdry/4-bdry/orig

are disjoint from



all 4 boundaries

outside boundaries



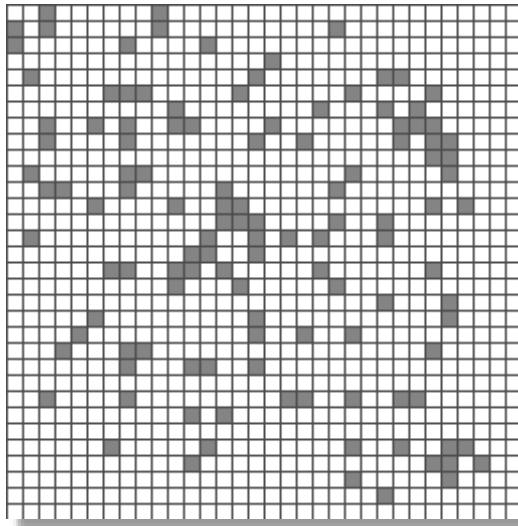
orig/8-bdry/4-bdry

This is a piece of a larger image. Boundary effects are not apparent

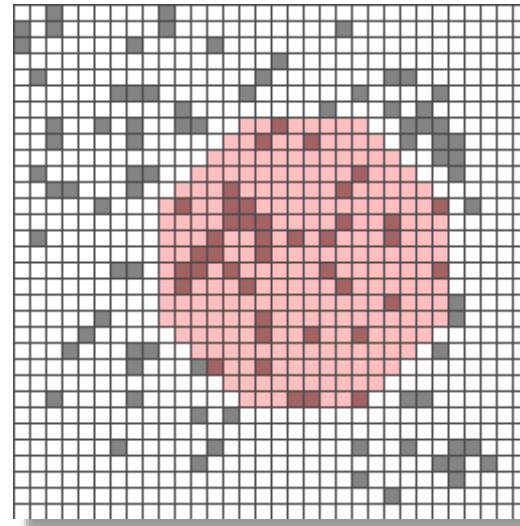


# Conditional Dilation

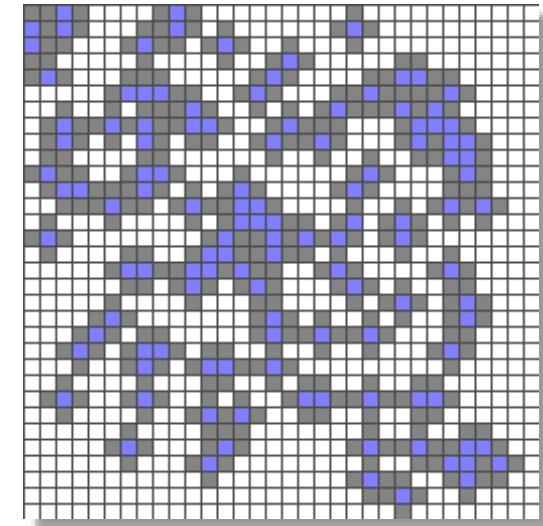
original image



mask over original



dilated original

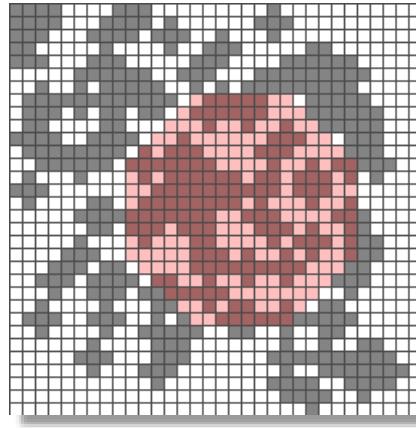


dilation inside a mask



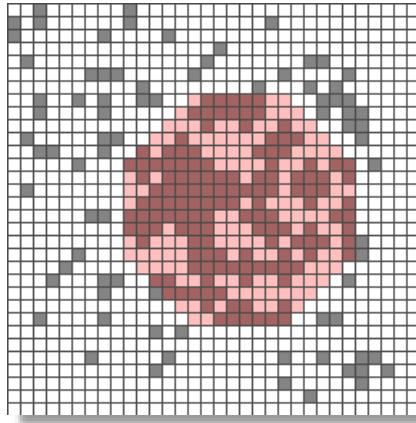
## Conditional Dilation

mask  
over  
dilated



masked  
dilated

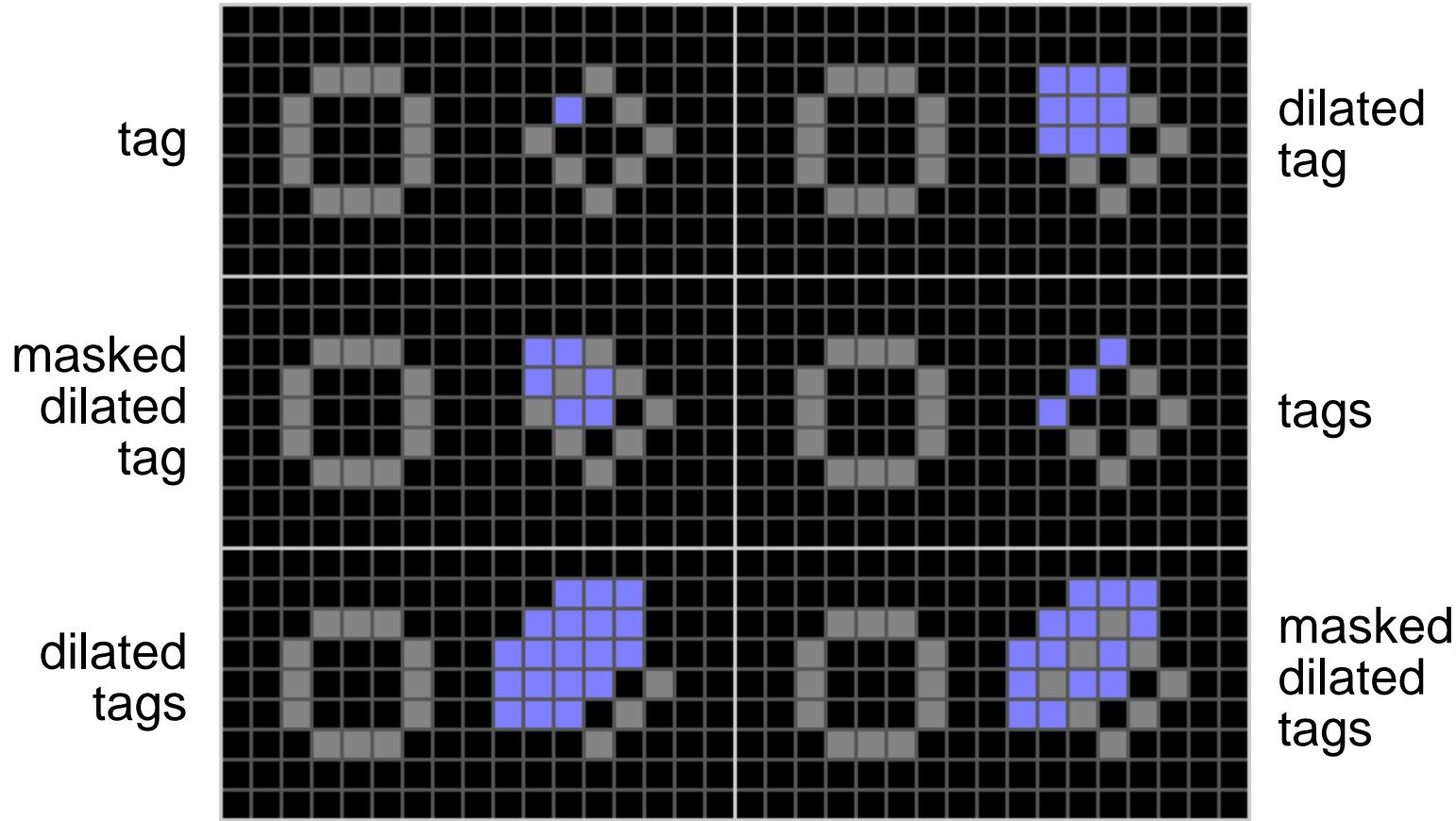
masked  
dilated  
union with  
original



conditionally  
dilated with  
respect to  
mask

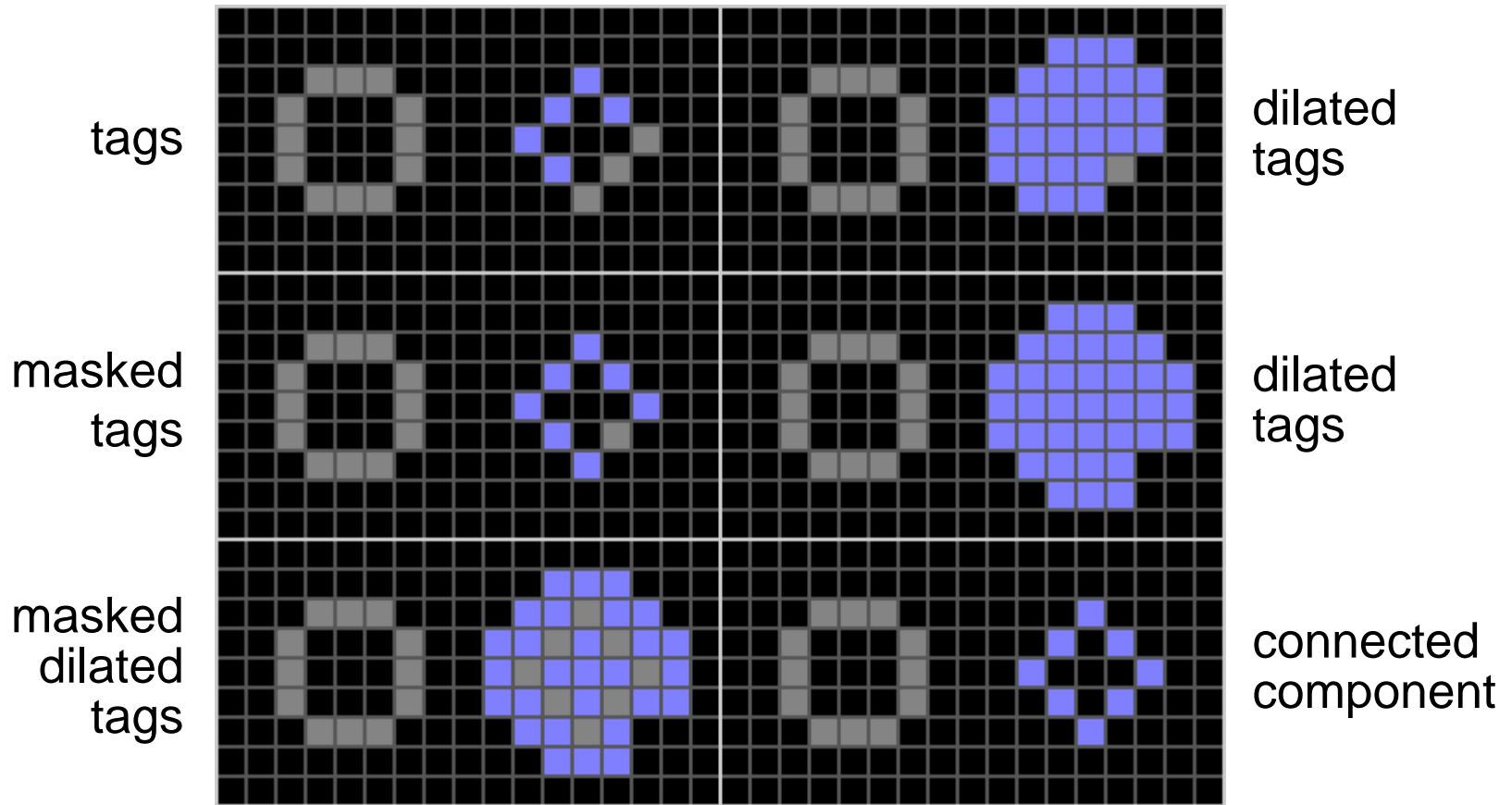


# Connected Component Extraction





# Connected Component Extraction





# Binary Reconstruction

Used after opening to *grow back* pieces of the original image that are connected to the opening.



original



opened



reconstructed

Removes of small regions that are disjoint from larger objects without distorting the small features of the large objects.



# Algorithm for Binary Reconstruction

1.  $J = I \circ Z$ , where  $Z$  is any SE.
2.  $T = J$ ,
3.  $J = J \oplus Z_k$ , where  $k=4$  or  $k=8$ ,
4.  $J = I \text{ AND } J$ , [*Take only those pixels from  $J$  that are also in  $I$ .*]
5. if  $J \neq T$  then go to 2,
6. else stop; [ *$J$  is the reconstructed image.*]

This is the same as connected component extraction with the opened image,  $J$ , containing the tags. The choice of  $Z_k$  determines the connectivity of the result.



# Algorithm for Binary Reconstruction

Usually a program for reconstruction will take both  $J$  and  $I$  as inputs. E.g,

1.  $J = I \circ Z$ , where
2.  $T = J$ ,
3.  $J = J \oplus Z_k$ , where  $k=4$  or  $k=8$ ,
4.  $J = I \text{ AND } J$ , [Take only those pixels from  $J$  that are also in  $I$ .]
5. if  $J \neq T$  then go to 2,
6. else stop; [  $J$  is the reconstructed image. ]

Then the algorithm starts at step 2.

This is the same as connected component extraction with the opened image,  $J$ , containing the tags. The choice of  $Z_k$  determines the connectivity of the result.



# Skeletonization

Let  $\text{Skel}(\mathbf{I}, r)$  be the set of pixels in  $\mathbf{I}$  such that if  $\mathbf{p} \subseteq \text{Skel}(\mathbf{I}, r)$  then  $D_p(r)$ , is a maximal disk of radius  $r$  in  $\mathbf{I}$ . That is,  $\text{Skel}(\mathbf{I}, r)$  is the locus of centers of maximal disks of radius  $r$  in  $\mathbf{I}$ . Then

$$S = \bigcup_{r=0}^{\infty} \text{Skel}(\mathbf{I}, r)$$

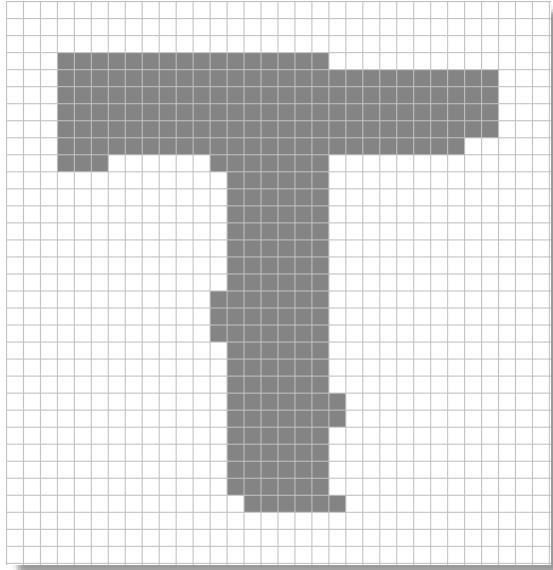
That is, the skeleton of  $\mathbf{I}$  is the union of all the sets of centers of maximal disks.

Note that for any actual image  $\mathbf{I}$ , the union will not be infinite, since  $\mathbf{I}$  is bounded (not infinite in extent).

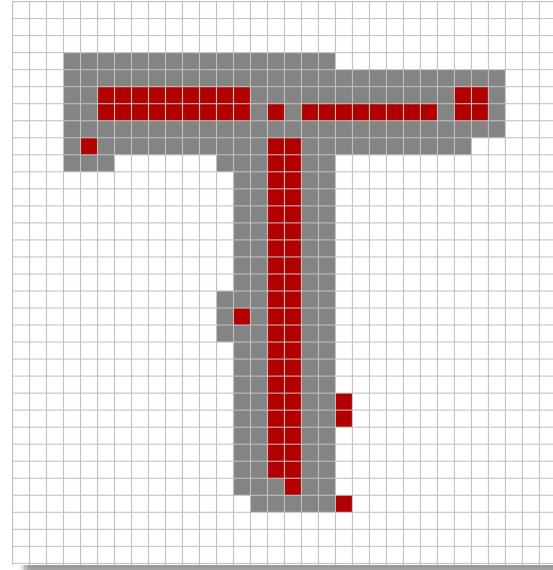


# Skeletonization

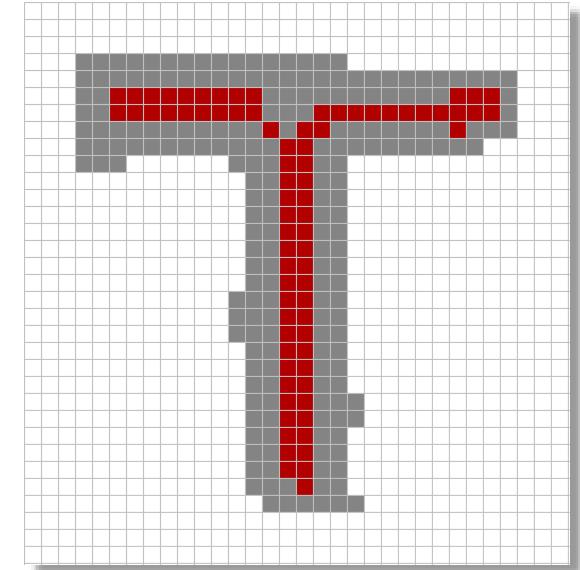
Original shape



Raw skeleton (red)



Pruned and connected



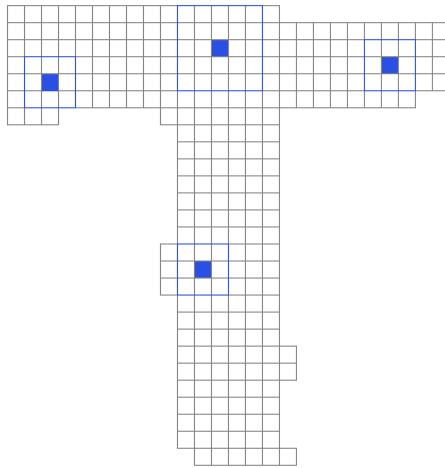
is the locus of centers of maximal disks.

skeleton

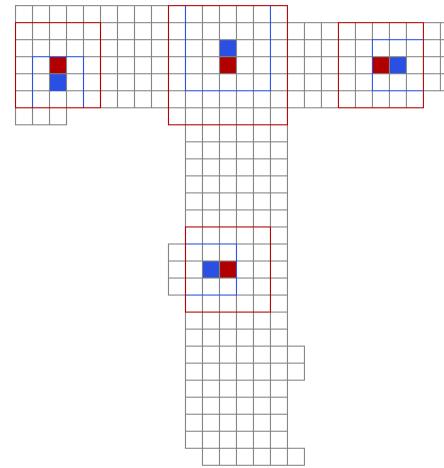


# Skeletonization: Maximal Disks

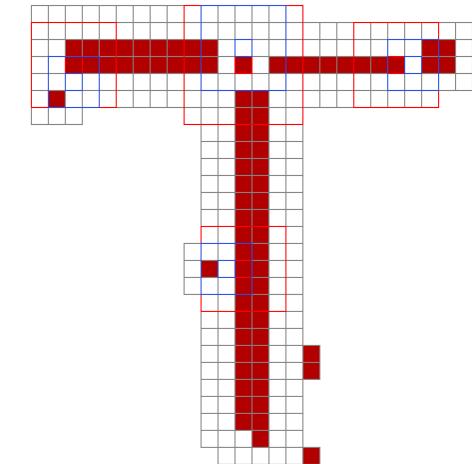
non maximal “disks”



maximal disks (red)



non max & max disks



“disks” are squares

non max disks (blue)

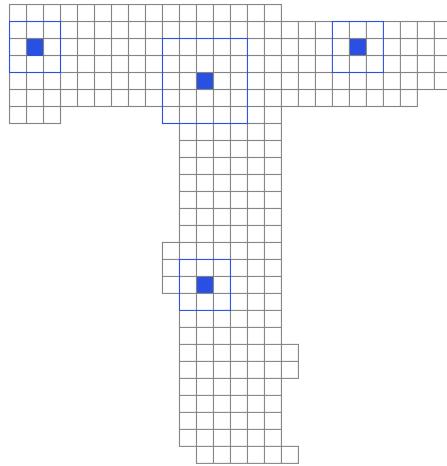
over skeleton

The maximal disk at pixel loc  $\mathbf{p}$  is the largest disk in the fg that includes  $\mathbf{p}$ .

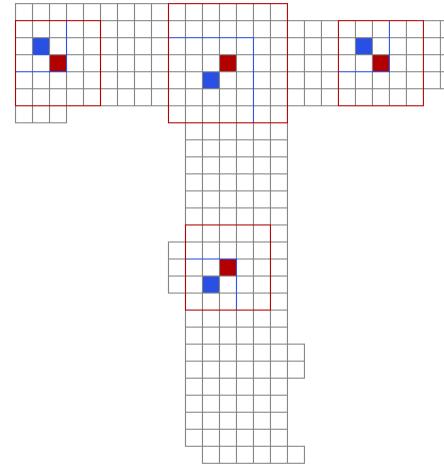


# Skeletonization: Maximal Disks

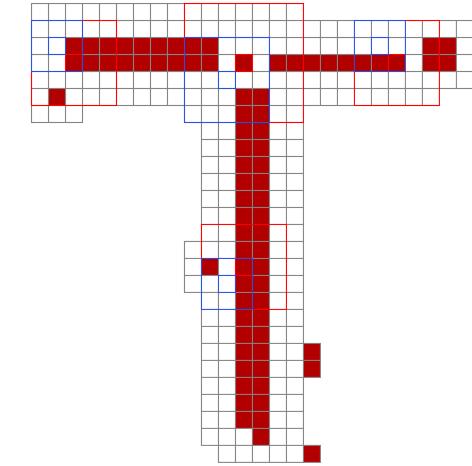
non maximal “disks”



maximal disks (red)



non max & max disks



“disks” are squares

non max disks (blue)

over skeleton

The maximal disk at pixel loc  $\mathbf{p}$  is the largest disk in the fg that includes  $\mathbf{p}$ .



# Computation of the Skeleton

- SE =  $Z_8 = \begin{smallmatrix} \textcolor{red}{\blacksquare} \\ \blacksquare \end{smallmatrix} = \text{Sq}(3)$
- $n = 0$ : SE = 1 pixel
- $n = 1$ : SE =  $\text{Sq}(3)$
- $n = 2$ : SE =  $\text{Sq}(5)$
- $n = 3$ : SE =  $\text{Sq}(7)$

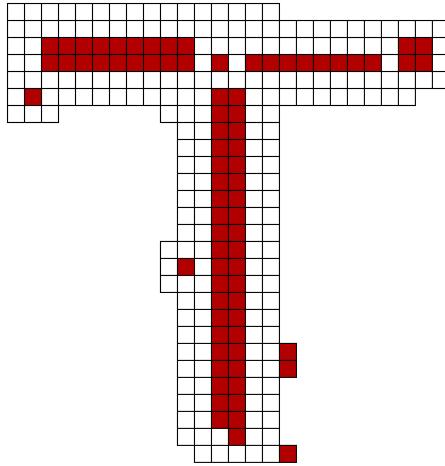
original		erode n=0		erode n=1		erode n=2		erode n=3	
skeleton		open above w/ Z (n=1)							
union of all 4 to the right		top - middle							
	$\text{Skel}(I, 0)$		$\text{Skel}(I, 1)$		$\text{Skel}(I, 2)$		$\text{Skel}(I, 3)$		

Note that the result is disconnected and has spurious points.



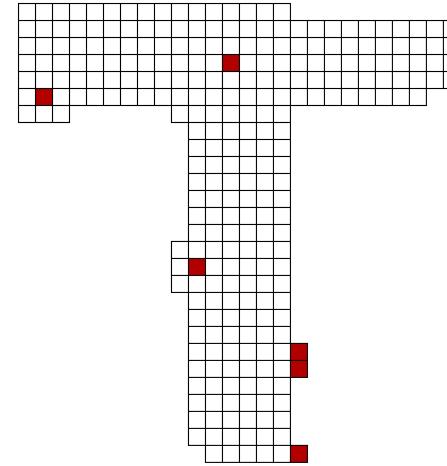
# Skeletonization: Delete Spurious Pixels

raw skeleton



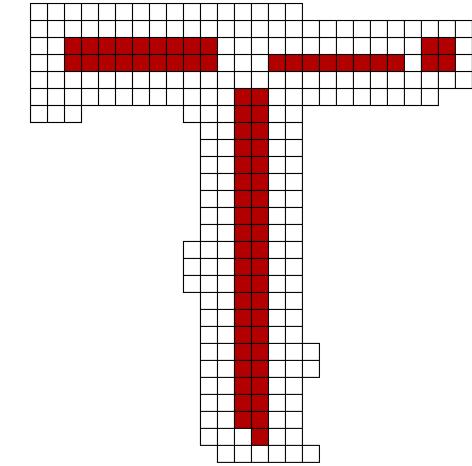
has spurious pixels

def. spurious pixels as  
conn. comp. of < 3 pix.



conn. comp. of < 3 pix.

pruned skeleton

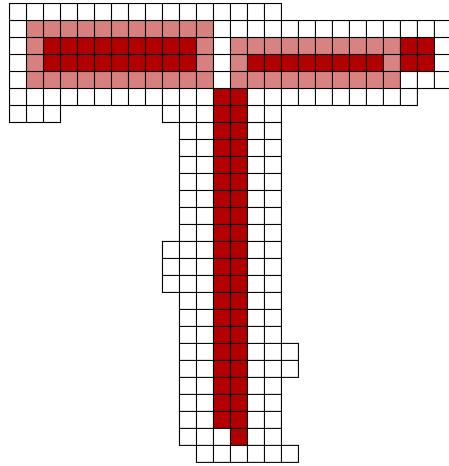


raw less spurious

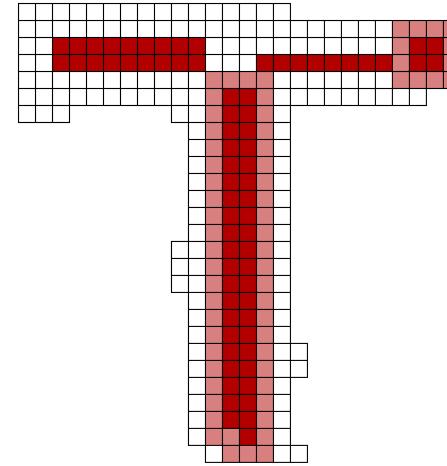


# Skeletonization: Reconnect Components

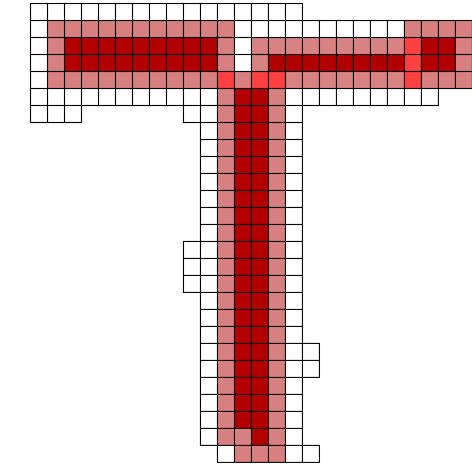
2 components of



2 other components



Intersection of



pruned skel. dilated.

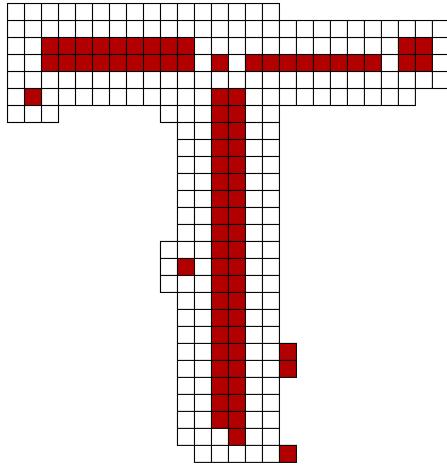
of pruned skel dilated.

dilated components.

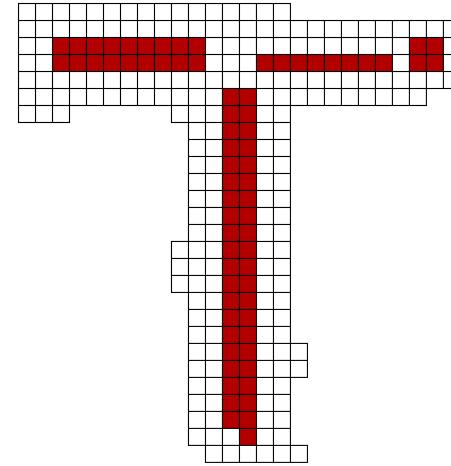


# Skeletonization

raw skeleton



pruned skeleton



reconnected skeleton

