

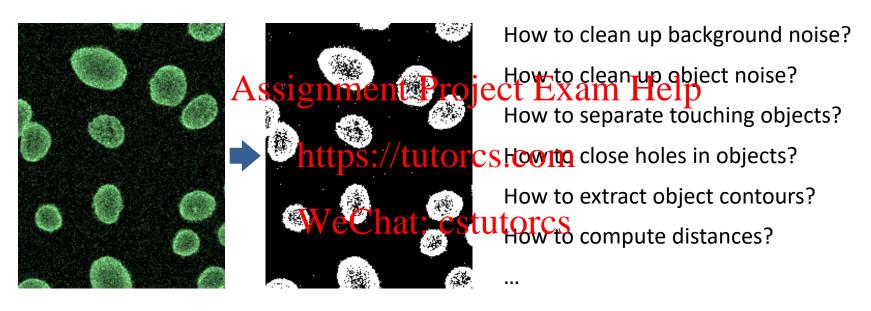
COMP9517: Computer Vision WeChat: cstutorcs

Image Segmentation

Part 2

#### How to improve image segmentation results?

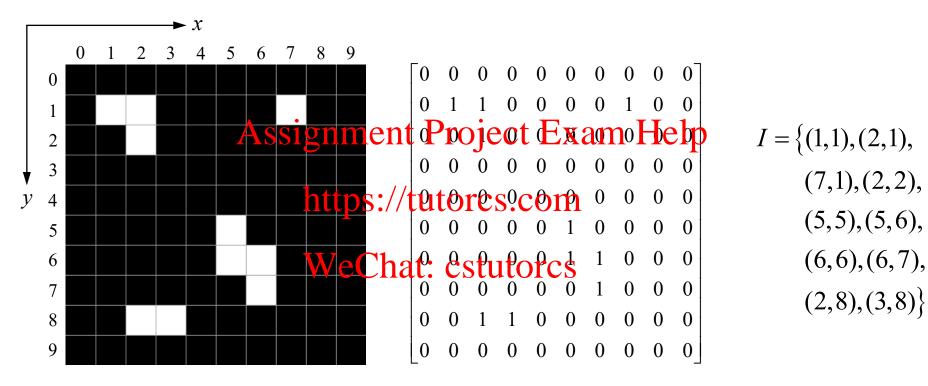
#### Processing using mathematical morphology



- Binary mathematical morphology (applies to binary images)
- Gray-scale mathematical morphology (applies to gray-scale images)

Nonlinear image processing based on set-theory rather than on calculus

## **Binary image representations**



#### **Binary image**

white = foreground black = background

#### Image as matrix

1 = foreground0 = background

#### Image as set

## **Basic set operations**

# Given sets $A = \{\mathbf{a}_i\}_{i=1}^{N_A}$ and $B = \{\mathbf{b}_i\}_{i=1}^{N_B}$ , with $\mathbf{a}_i, \mathbf{b}_i \in \mathbf{Z}^n$ , we have

Translation: 
$$A_{\mathbf{x}} = \{\mathbf{c}_i \mid \mathbf{c}_i = \mathbf{a}_i + \mathbf{x}, \, \mathbf{a}_i \in A\}$$
 for any given  $\mathbf{x} \in \mathbf{Z}^n$ 

Reflection: 
$$A^r = \{\mathbf{x}_i \mid \mathbf{x}_i = -\mathbf{a}_i, \mathbf{a}_i \in A\}$$

Complement: 
$$A^c = \{\mathbf{x}_i \mid \mathbf{x}_i \notin A\}$$



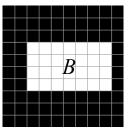


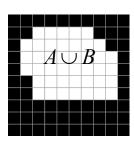
Union: 
$$A \cup B = \{\mathbf{x}_i \mid \mathbf{x}_i \in A \lor \mathbf{x}_i \in B\}$$

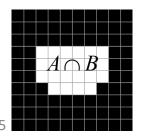
Intersection: 
$$A \cap B = \{ \mathbf{x}_i \mid \mathbf{x}_i \in A \land \mathbf{x}_i \in B \}$$

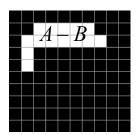
Difference: 
$$A - B = \{ \mathbf{x}_i \mid \mathbf{x}_i \in A \land \mathbf{x}_i \notin B \}$$

Cardinality: 
$$|A| = N_A$$
 and  $|B| = N_B$ 









#### **Dilation of binary images**

# **Definition of binary dilation:** $I \oplus S = \{ \mathbf{x} \mid S_{\mathbf{x}}^r \cap I \neq \emptyset \}$

That is, the set of points  $\mathbf{x} \in \mathbf{Z}^n$  for which the intersection of the image I with

the reflected version of a structuring element S translated to  $\mathbf{x}$ , is not empty  $\underset{\mathbf{Assignment}}{\mathbf{Assignment}}$  Project Exam Help

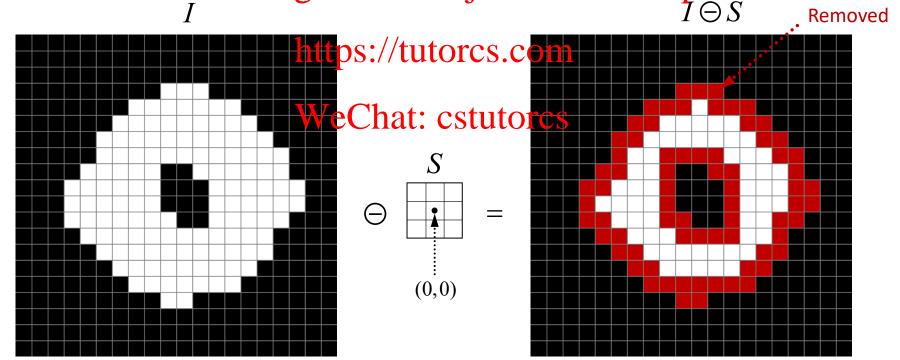
Interest of the second second

#### **Erosion of binary images**

# **Definition of binary erosion:** $I \ominus S = \{ \mathbf{x} \mid S_{\mathbf{x}} \subseteq I \}$

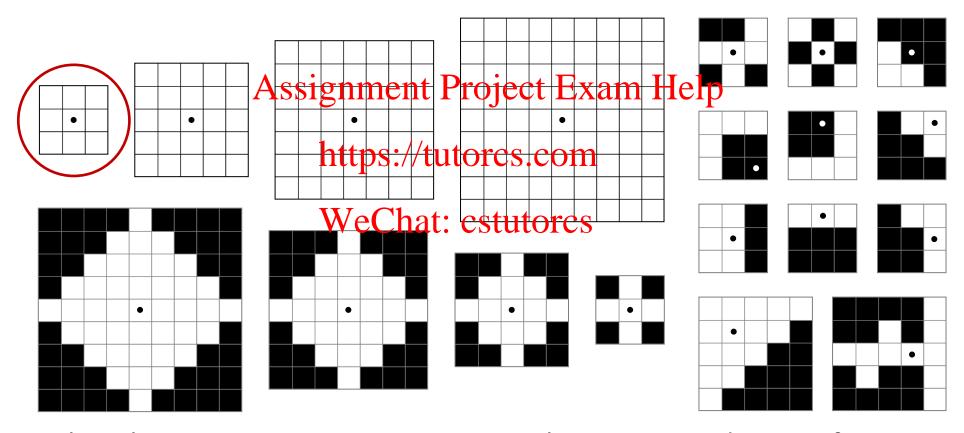
That is, the set of points  $\mathbf{x} \in \mathbf{Z}^n$  for which the structuring element S

translated over  $\mathbf{x}$  is completely contained in the image  $I_{\mbox{Assignment Project Exam Help}}$ 



## **Structuring elements**

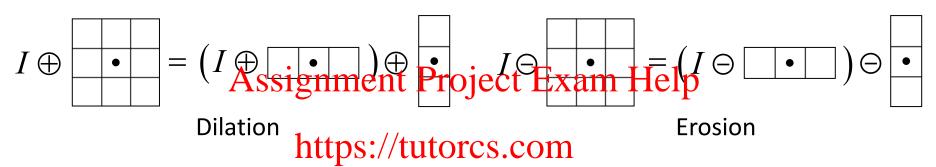
In principle a structuring element can have any shape ...

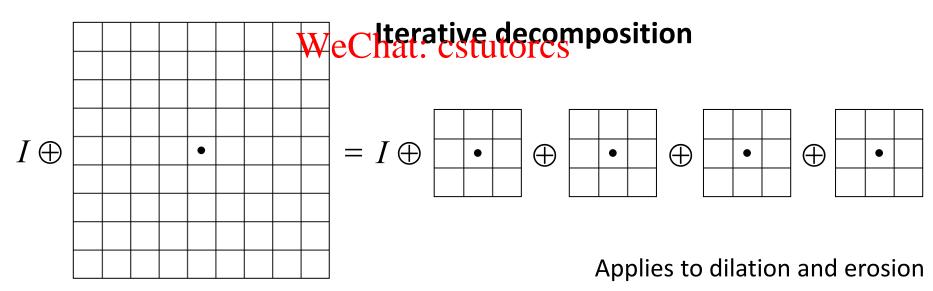


... but the symmetric 3 x 3 structuring element is used most often

#### **Dimensional decomposition**

#### **Decomposition of the basic structuring element**



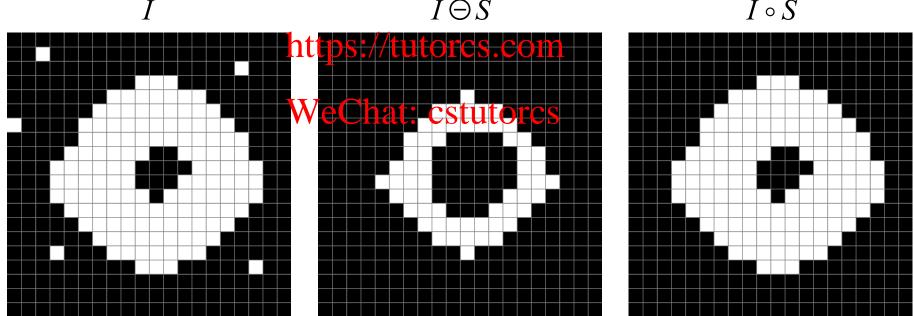


#### **Opening of binary images**

## **Definition of binary opening:** $I \circ S = (I \ominus S) \oplus S$

That is, an erosion followed by a dilation using the same structuring element

Example using the basic 3 x 3 structuring element:
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Eliminates details smaller than the structuring element *outside* the main object

## **Closing of binary images**

# **Definition of binary closing:** $I \bullet S = (I \oplus S) \ominus S$

That is, a dilation followed by an erosion using the same structuring element

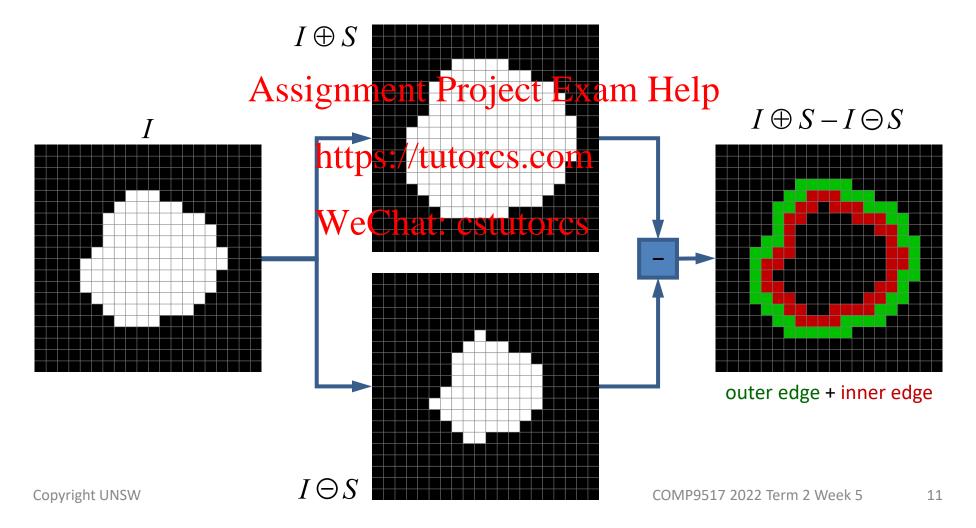
Example using the basic 3 x 3 structuring element:
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Eliminates details smaller than the structuring element *inside* the main object

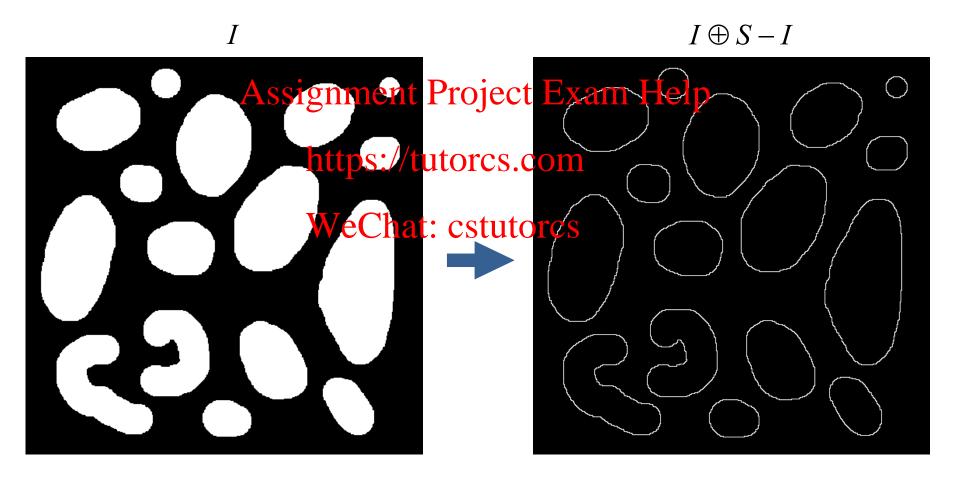
## Morphological edge detection

#### Difference between the dilated and the eroded image



## **Binary object outlines**

#### How to get a one-pixel thick outline of all objects in the image?

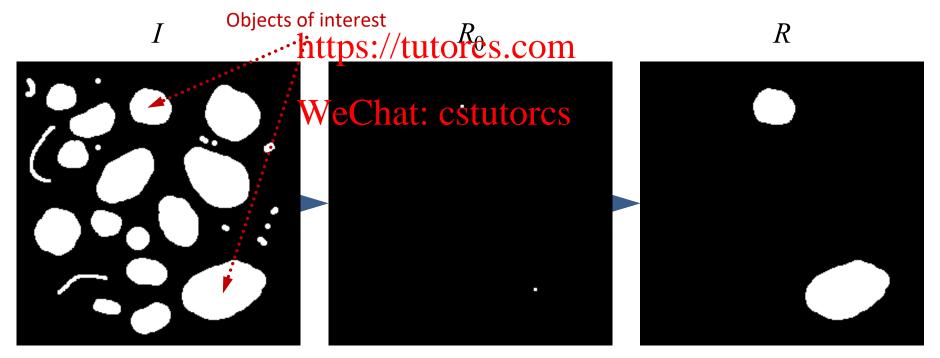


#### **Reconstruction of binary objects**

#### How to create an image containing selected objects only?

Create a marker image  $R_0$  containing seed pixels from each selected object

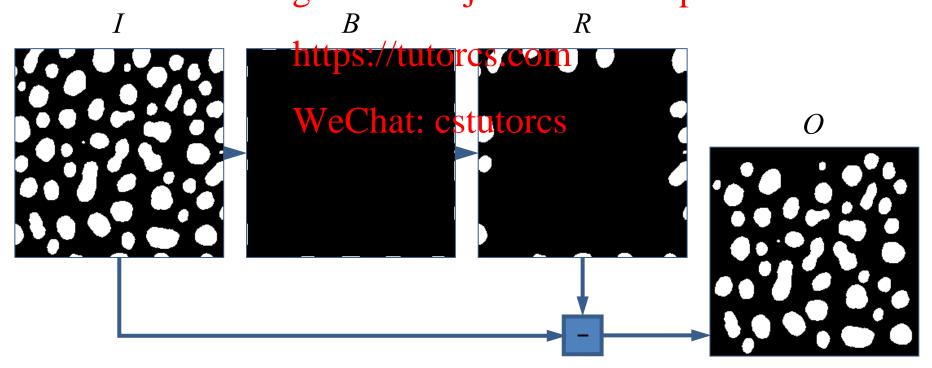
of image I and then iteratively compute  $R_i = (R_{-1} \oplus S) \cap I$  until  $R_i = R_{i-1}$  and  $R_i = R_{i-1}$ 



#### **Reconstruction of binary objects**

#### How to remove objects that are only partly in the image?

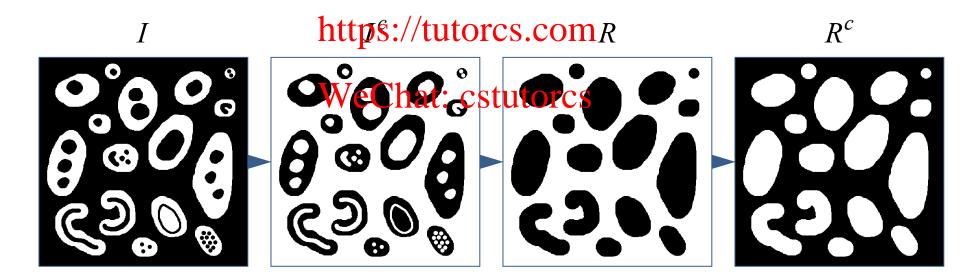
Take the boundary pixels B of input image I as the seeds, compute the reconstruction R from those seeds, and subtract the result from the input Assignment Project Exam Help



#### **Reconstruction of binary objects**

#### How to fill all holes in all of the objects in the image?

Take the complement  $I^c$  of image I, take the boundary pixels of  $I^c$  as seeds, compute reconstruction R of  $I^c$  from those seeds, take the complement  $R^c$  Assignment Project Exam Help

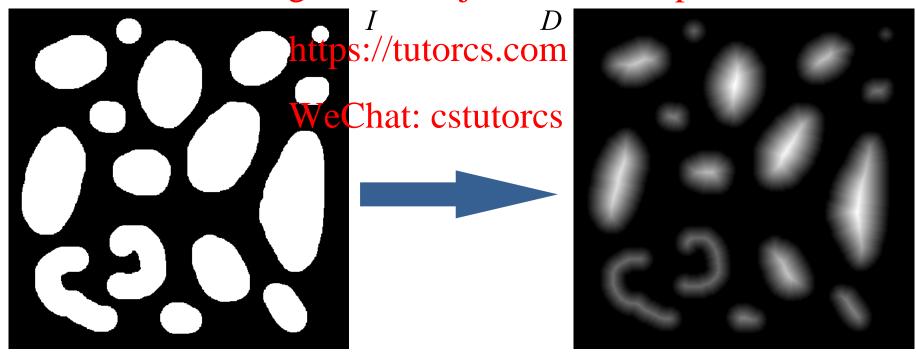


## Distance transform of binary images

#### How to compute the distance of object pixels to the background?

Denote input image I as  $I_0$  and iteratively compute  $I_i = I_{i-1} \Theta S$  for i = 1, 2, ...

while setting all pixels eroded in iteration i to value i in the output image D Assignment Project Exam Help

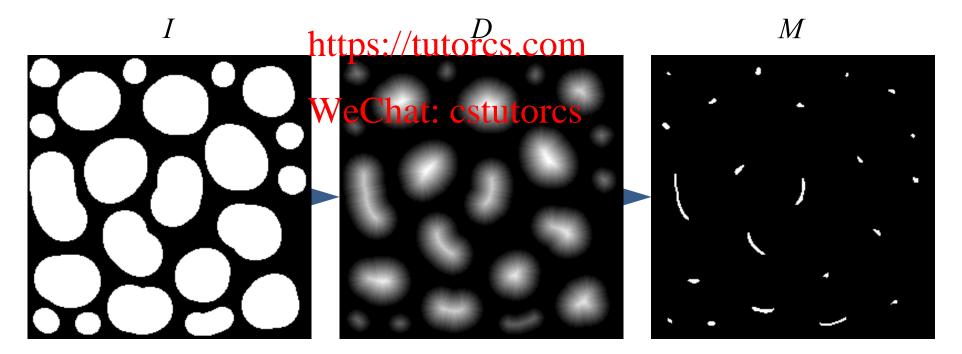


#### Ultimate erosion of binary images

#### How to find representative center points for all the objects?

Compute the distance transform of the image and find all the local maxima

This is the same as keeping only the last object pixels before final erosion ASSIGNMENT Project Exam Help

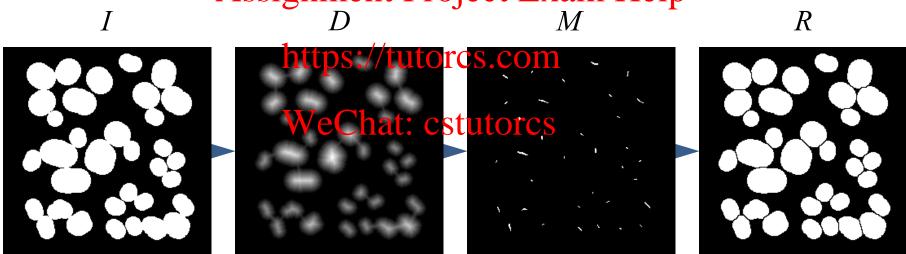


#### Ultimate erosion and reconstruction

#### How to separate objects that are touching each other?

Perform ultimate erosion and then perform a reconstruction of the result

with the additional constraint that objects may not merge Assignment Project Exam Help



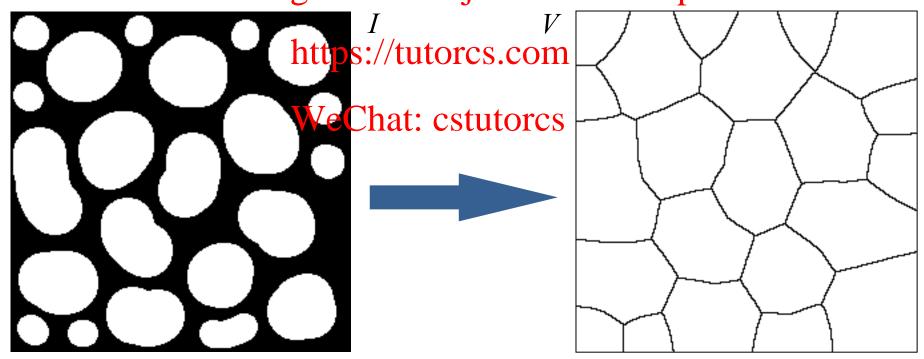
Since elongated objects may result in multiple local maxima this approach works best for objects that are more or less circular

#### Ultimate dilation of binary images

#### How to find the background points equidistant to the objects?

Iteratively dilate the image while imposing the non-merging constraint

The result is called the Voronoi (or Dirichlet) tessellation of the objects Assignment Project Exam Help

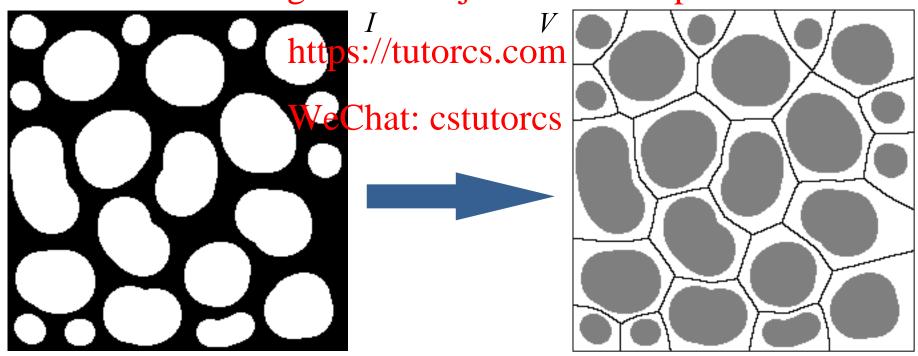


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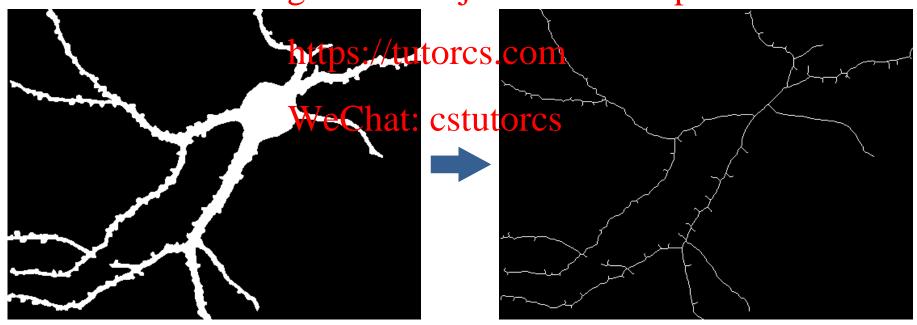


## **Skeletonization of binary images**

#### How to find a representative centerline of the objects?

Iteratively apply conditional erosion (thinning) that does not break the

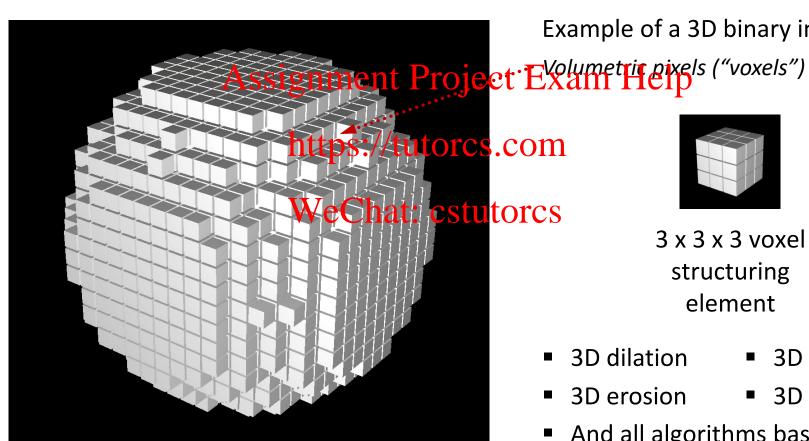
connectivity of the result and does not remove single pixels or end-pixels ASSIGNMENT Project Exam Help



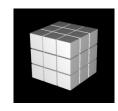
The resulting one-pixel thick structure is called the *skeleton* of the object

## Binary morphology of nD images

#### The presented concepts extend to any dimensionality



Example of a 3D binary image

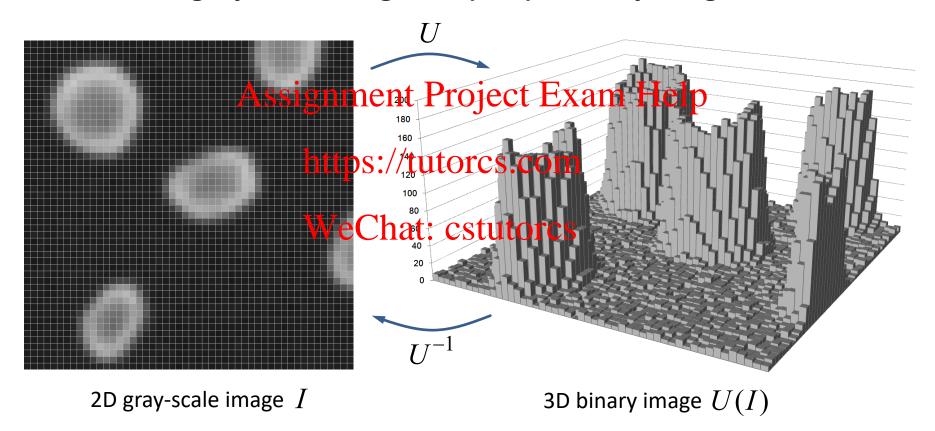


3 x 3 x 3 voxel structuring element

- 3D dilation
- 3D opening
- 3D erosion
- 3D closing
- And all algorithms based on it

## **Gray-scale mathematical morphology**

#### Consider *n*D gray-scale images as (*n*+1)D binary images...



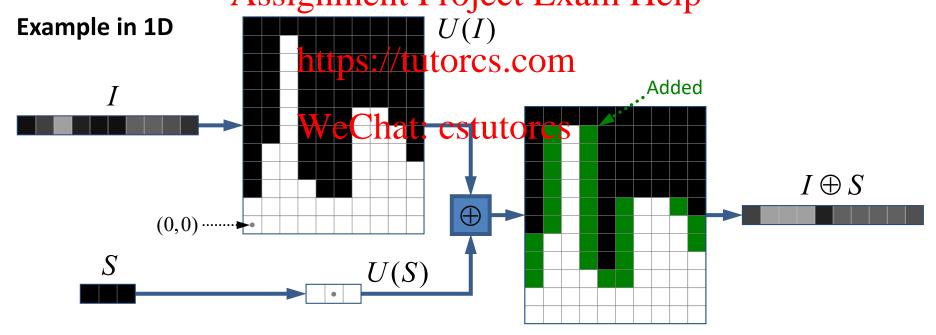
The landscape surface with the volume below is called the *umbra* of the image

#### **Dilation of gray-scale images**

# **Definition of gray-scale dilation:** $I \oplus S = U^{-1} [U(I) \oplus U(S)]$

That is, the binary dilation of the umbra  $\,U(I)\,$  of gray-scale image  $I\,$  with the

umbra U(S) of gray-scale structuring element S, turned back into gray-scale Assignment Project Exam Help

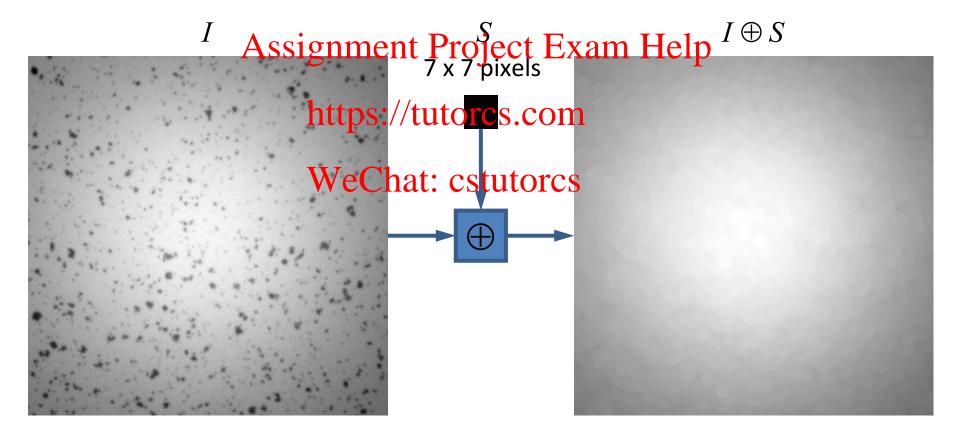


Any gray-scale S is possible but the flat one (as shown here) is used most often

## **Dilation of gray-scale images**

Equivalent definition: 
$$(I \oplus S)(\mathbf{x}) = \max_{\mathbf{p}} \{I(\mathbf{x} - \mathbf{p}) + S(\mathbf{p})\}$$

For a flat and symmetrical structuring element this is simply local max-filtering

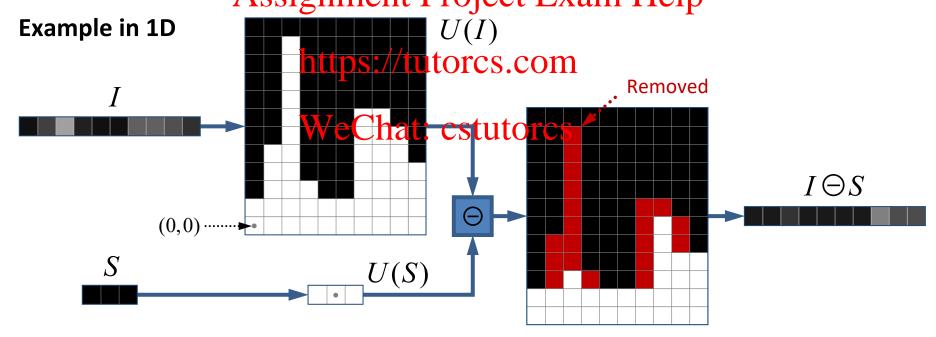


#### **Erosion of gray-scale images**

# **Definition of gray-scale erosion:** $I \ominus S = U^{-1} [U(I) \ominus U(S)]$

That is, the binary erosion of the umbra  $\,U(I)$  of gray-scale image  $\,I\,$  with the

umbra U(S) of gray-scale structuring element S turned back into gray-scale Assignment Project Exam Help

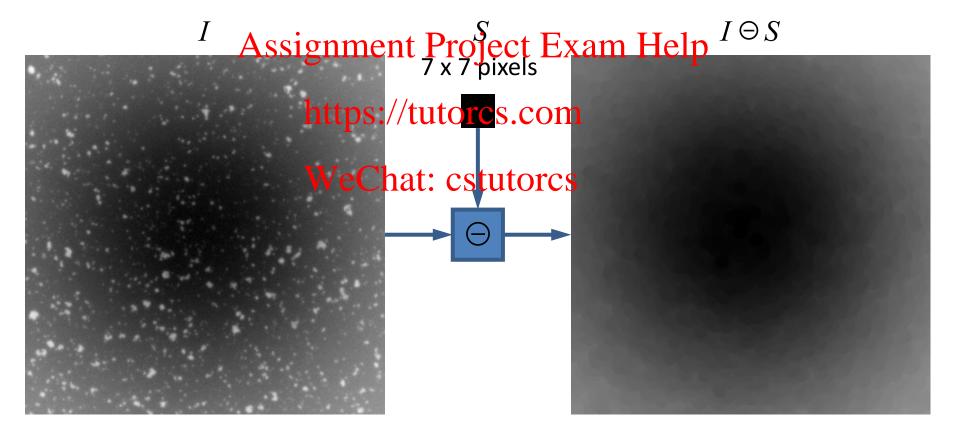


Any gray-scale S is possible but the flat one (as shown here) is used most often

#### **Erosion of gray-scale images**

Equivalent definition: 
$$(I \ominus S)(\mathbf{x}) = \min_{\mathbf{p}} \{I(\mathbf{x} + \mathbf{p}) - S(\mathbf{p})\}$$

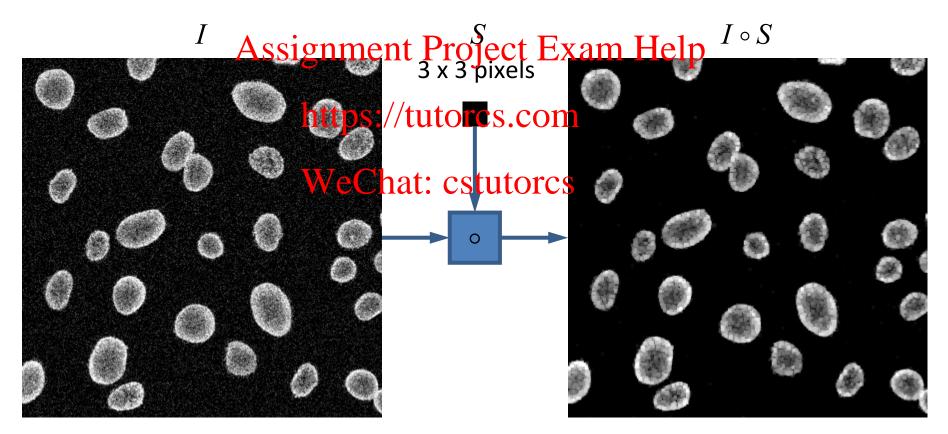
For a flat and symmetrical structuring element this is simply local min-filtering



## **Opening of gray-scale images**

## **Definition of gray-scale opening:** $I \circ S = (I \ominus S) \oplus S$

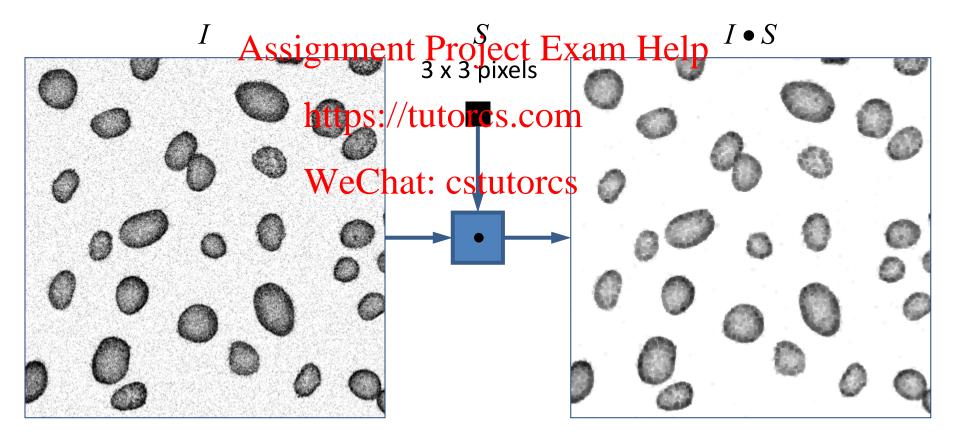
Gray-scale erosion and then gray-scale dilation with same structuring element



#### Closing of gray-scale images

# **Definition of gray-scale closing:** $I \bullet S = (I \oplus S) \ominus S$

Gray-scale dilation and then gray-scale erosion with same structuring element

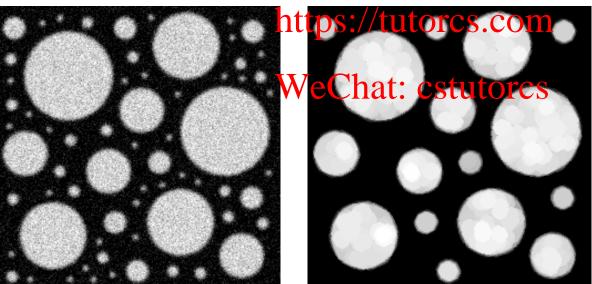


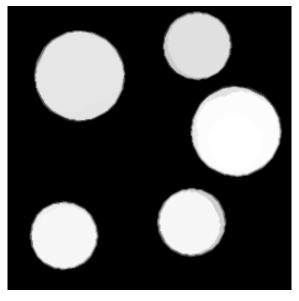
## Morphological smoothing of gray-scale images

#### Suppressing image structures of specific size (and shape)

- High-valued (bright) image structures are removed by gray-scale opening
- Low-valued (dark) image structures are removed by gray-scale closing Assignment Project Exam Help

 $I \circ S$  (radius = 7 pixels)  $I \circ S$  (radius = 25 pixels)

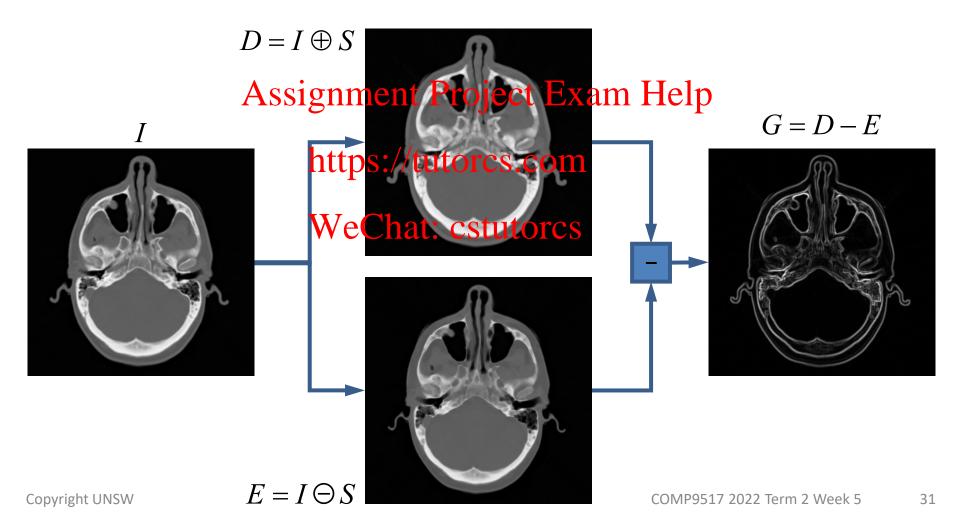




Behold the power of nonlinear filtering (not possible with linear filtering)

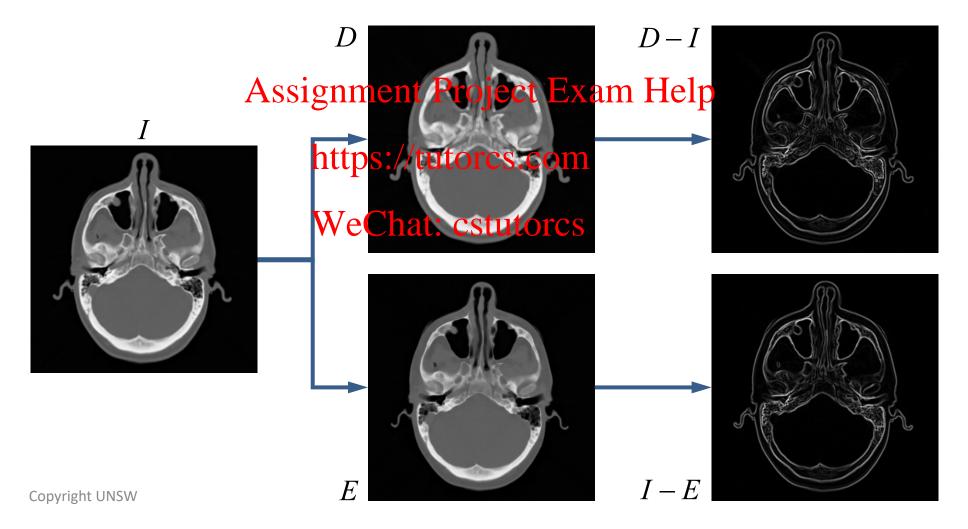
## Morphological gradient of gray-scale images

#### Difference between the dilated and the eroded image

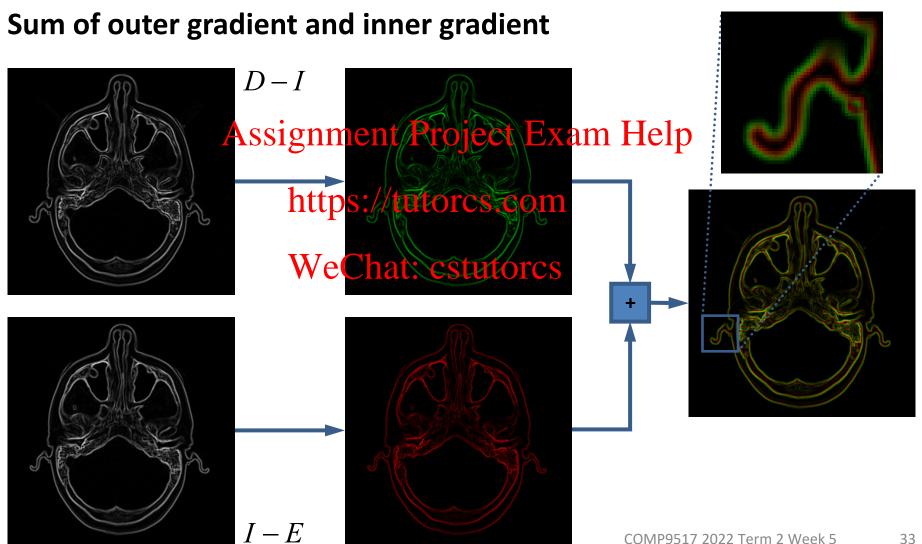


## Morphological gradient of gray-scale images

#### Outer gradient and inner gradient

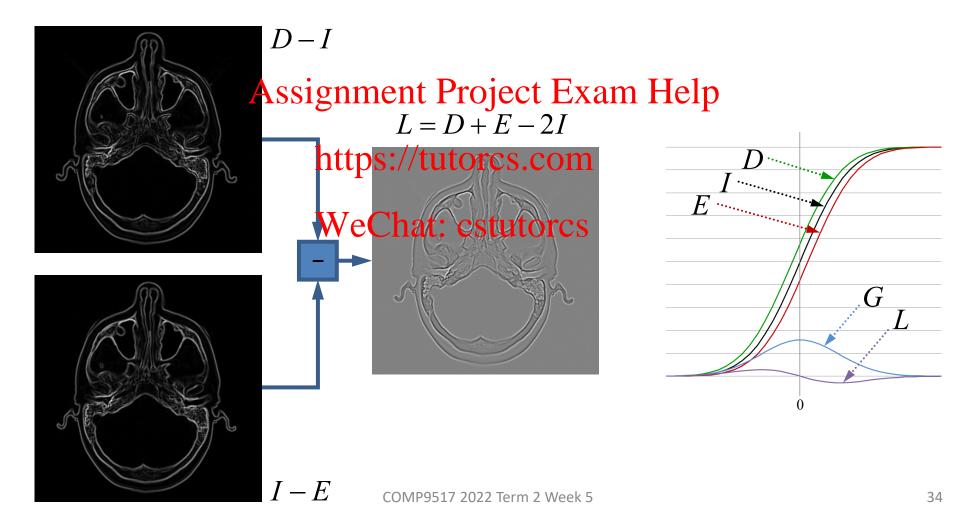


## Morphological gradient of gray-scale images

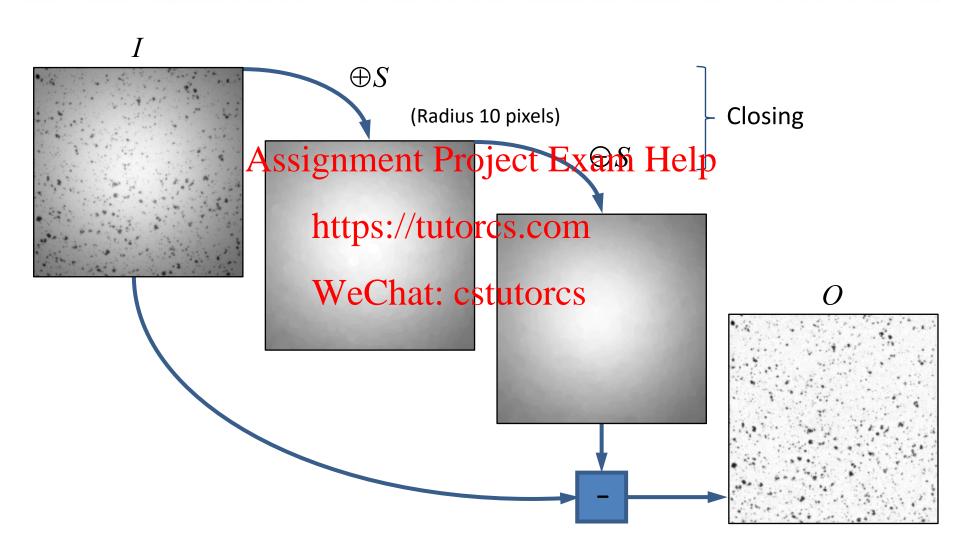


## Morphological Laplacean of gray-scale images

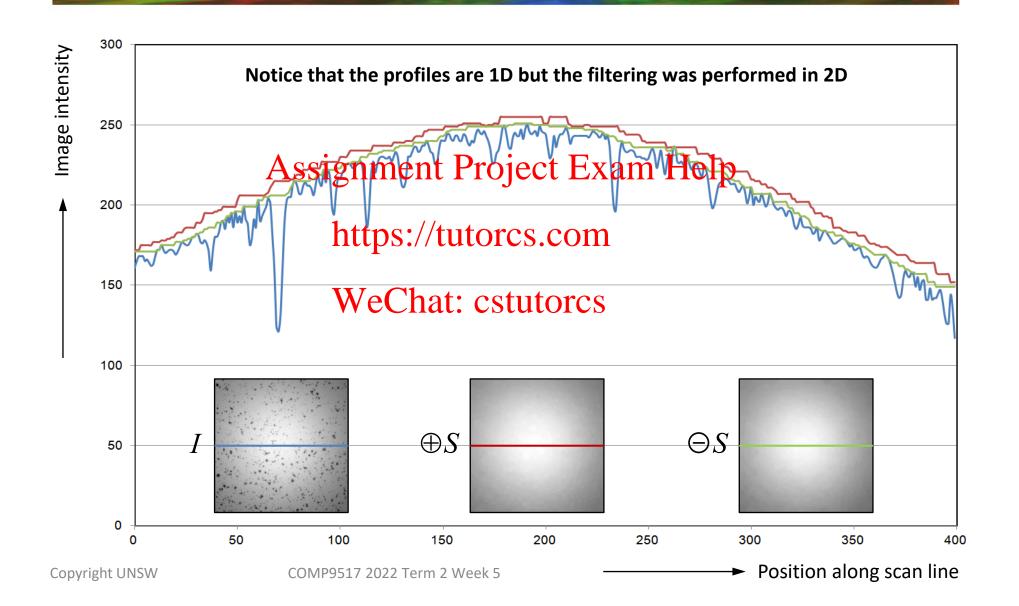
#### Difference between outer gradient and inner gradient



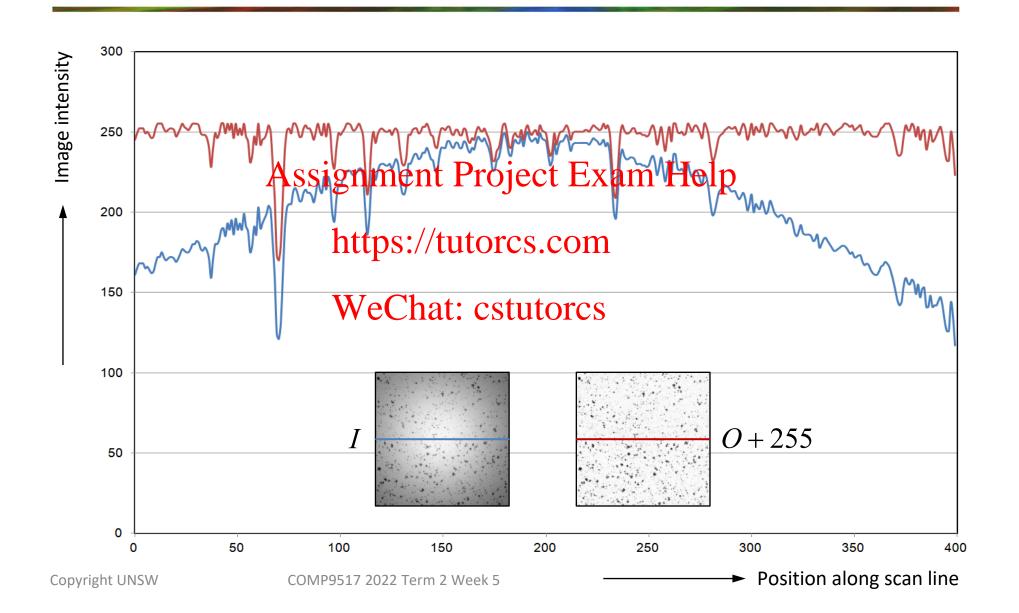
## Top-hat filtering of gray-scale images



## Top-hat filtering of gray-scale images



## Top-hat filtering of gray-scale images



## Summary of mathematical morphology

#### Powerful toolbox of methods around image segmentation

- Removal of background shading

  Removal of unwanted images in the content of the c
- Binary morphology for post placessing tutores
   Closing holes in objects and background
   Finding the inner or outer outline of objects
   Detecting and separating touching objects
   Extracting representative shapes of objects
   Computing distances within and between objects

After segmentation