

Thinning

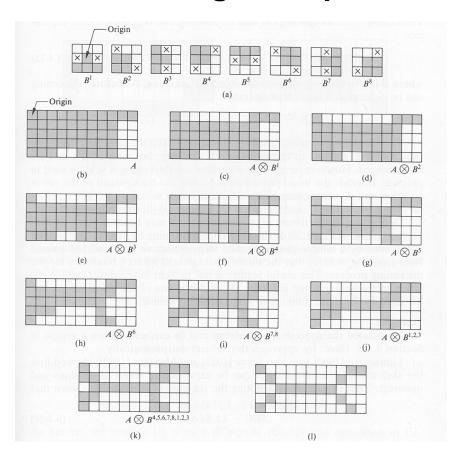
 The thinning of a set A by structuring element B, denoted A⊗B, can be defined in terms of the hit-or-miss transform

$$A \otimes B = A - (A \otimes B)$$
$$= A \cap (A \otimes B)^{c}$$

- The usual process is to thin A using a sequence of structuring elements B¹,...Bⁿ
- In other words, A is thinned by successive passes of structuring elements B¹, B²,...
- The entire process is repeated until no further change occurs



Thinning Example



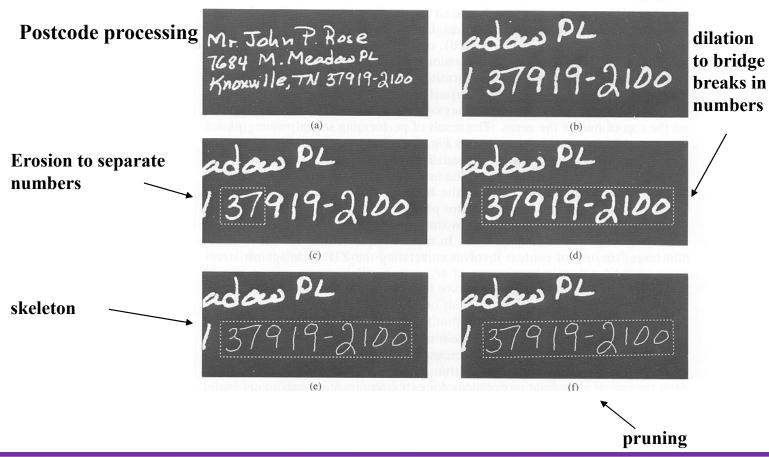


Other Morphological Operations

- We also have definitions for the following operations
 - Thickening
 - make lines thicker
 - Skeletonization
 - extract morphological skeleton
 - Pruning
 - extract parasitic components after skeletonization
- See Gonzalez and Woods, "Digital Image Processing," pp 518-545 for details



Application





Grayscale Morphology

- Many binary morphological operations are simply extended to grayscale images
- Here we regard the grayscale image as a surface that is eroded and dilated
- Often it is simpler to illustrate the process with 1-D functions, since the extension to 2-D is trivial.



Grayscale Dilation

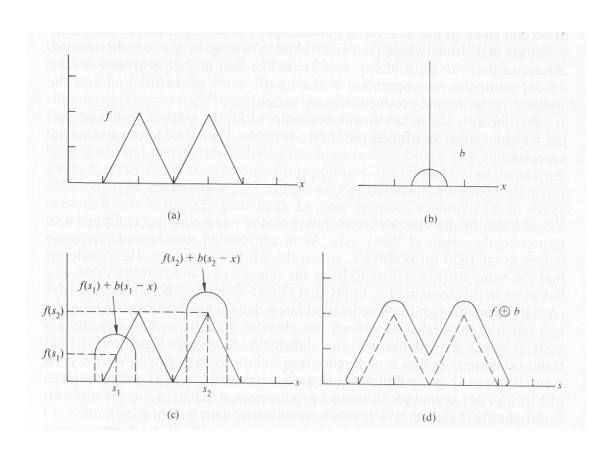
Grayscale dilation of f by b, denoted f⊕b, is defined by

$$(f \oplus b)(s,t) = \max\{f(s-x,t-y) + b(x,y) \mid (s-x), (t-y) \in D_f; (x,y) \in D_b\}$$

- In other words, we find the maximum of the function f+b in a neighborhood defined by the structuring element b as we slide b over f.
- This is illustrated graphically on the next slide for a 1-D function



Dilation Example





Grayscale Erosion

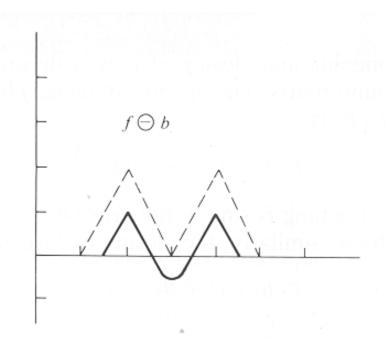
Similarly, Grayscale erosion of f by b, denoted f⊖b, is defined by

$$(f \ominus b)(s,t) = \min\{f(s-x,t-y) + b(x,y) \mid (s-x), (t-y) \in D_f; (x,y) \in D_b\}$$

In other words, we find the minimum of the function f+b in a
neighbourhood defined by the structuring element b as we slide b over f.



Erosion Example

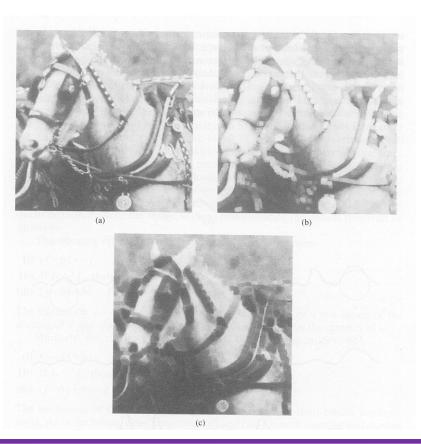


Same example as before



Image Example

Original



Dilated

Eroded



Grayscale Opening and Closing

- The expression for grayscale opening and closing is the same as for binary
- The expression for opening is

$$f \circ b = (f \ominus b) \oplus b$$

which is simply erosion followed by dilation

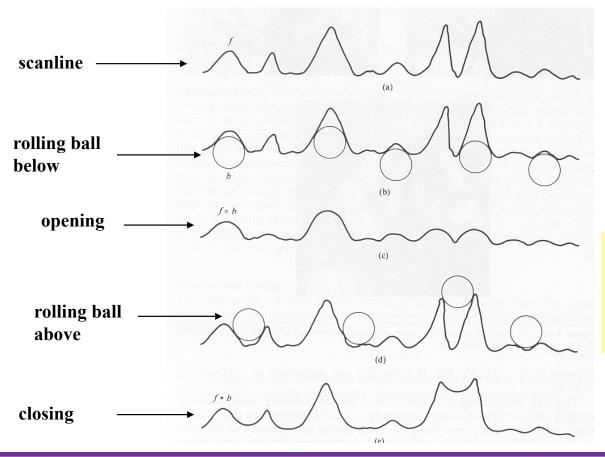
The expression for closing is

$$f \bullet b = (f \oplus b) \ominus b$$

which is simply dilation followed by erosion



Geometric Interpretation



Rolling

Ball

Interpretation



Comments

- Openings are used to remove small light details, while leaving the overall gray levels and larger bright features relatively undisturbed
- Closing is generally used to remove dark details from an image while leaving bright features relatively undisturbed



Applications

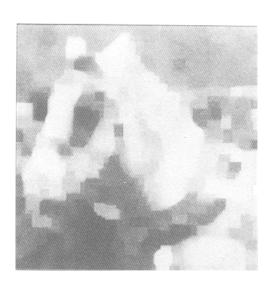
- Morphological smoothing
 - Opening followed by a closing
 - removes or attenuates both bright and dark artifacts or noise
- Morphological gradient
 - This is the difference between dilation and erosion

$$g = (f \oplus b) - (f \ominus b)$$

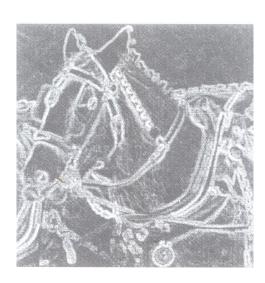
- highlights sharp gray level transitions in the image
- depends less on edge direction than Sobel etc



Examples



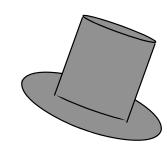
Morphological Smoothing



Morphological Gradient



Applications



- Top Hat Transformation
 - The Morphological top-hat transformation is defined by

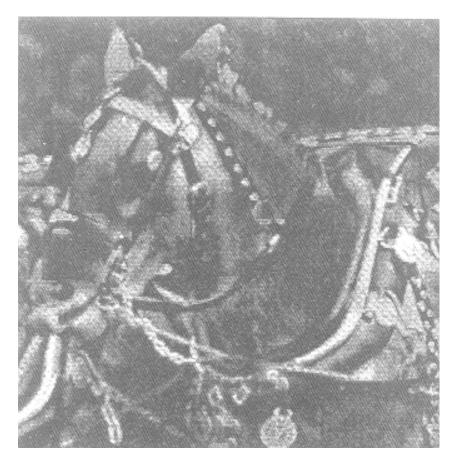
$$h = f - (f \circ b)$$

- That is the image f minus its opening with a structuring element b, which is often of the form of a "top-hat." That is, a cylinder attached to a disk.
- This transformation is useful for enhancing detail in the presence of shading
- Also good in 1D for finding peaks that are, say, greater than a certain width and more than a certain depth (significant peaks)



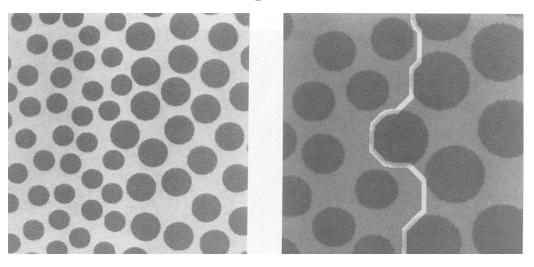
Example

Top hat transformation of image
Note the enhanced detail





Texture Segmentation



Method: Close with successively larger structuring elements until small dots disappear. Open remaining image with large structuring element and then threshold to determine textural boundary.

Consider interpretation with rolling ball – process resembles coin sorter

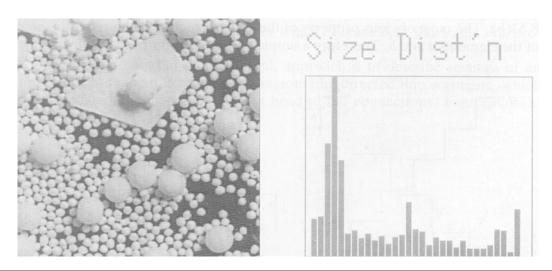


Granulometry

- Granulometry is a field that deals with the size distribution of particles
- In the example image (next slide), there are light objects of three different sizes
- The objects are overlapping and are too cluttered to detect individual objects



Example



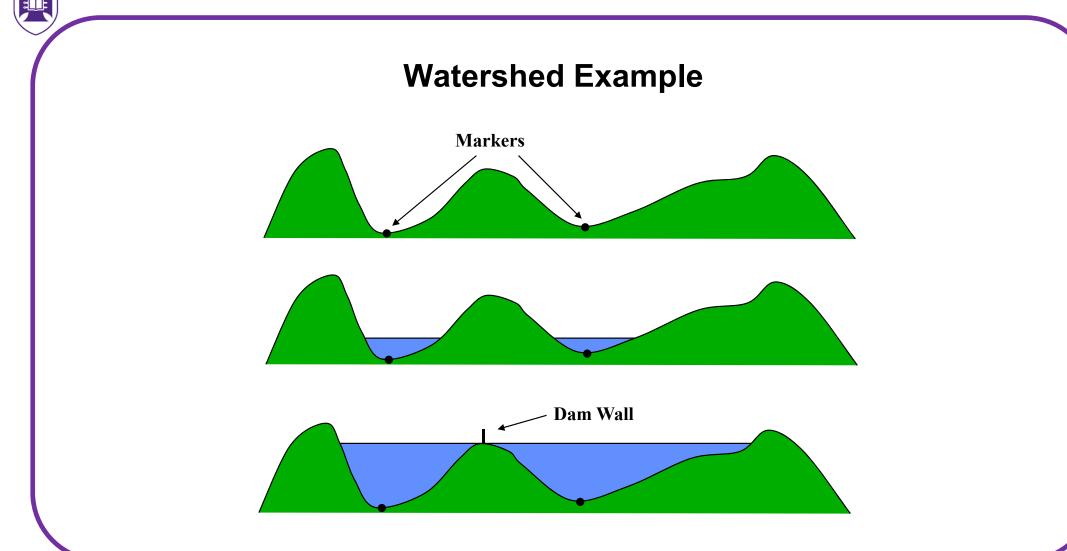
Method:

- opening operations with structuring elements of increasing size
- the difference between the original image and its opening is computed on each pass
- at the end of the process the differences are normalized and used to construct a histogram of particle size distribution



Watersheds

- Create markers on image (usually image gradient) to indicate regions of interest
- Visualize image as a topological surface (mountains and valleys)
- Flood object with a deluge of rain
- When waters from different regions meet, construct dams
- Once surface is completely flooded, the dam walls are our watershed segmentation.





Problems with Watersheds

- Usually operates on the gradient image rather than the image itself
 - -gradient accentuates noise
- Often difficult to determine appropriate markers in many applications
 - -may lead to poor segmentation
- Found to be unsuitable for cell image segmentation application