Computer Vision – T2.1 Segmentation

MAP-I Doctoral Programme

Miguel Tavares Coimbra



Outline

- Thresholding
- Region-based Segmentation
- Morphological Filters

Acknowledgements: Most of this course is based on the excellent courses offered by Prof. Shree Nayar at Columbia University, USA and by Prof. Srinivasa Narasimhan at CMU, USA. Please acknowledge the original source when reusing these slides for academic purposes.

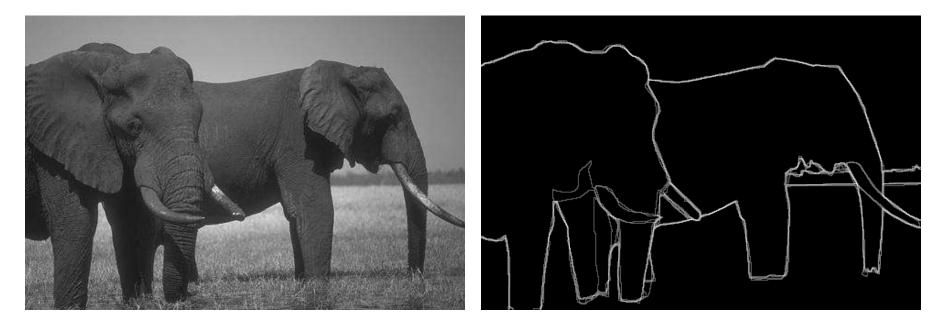


Topic: Thresholding

- Thresholding
- Region-based Segmentation
- Morphological Filters



Boundaries of Objects

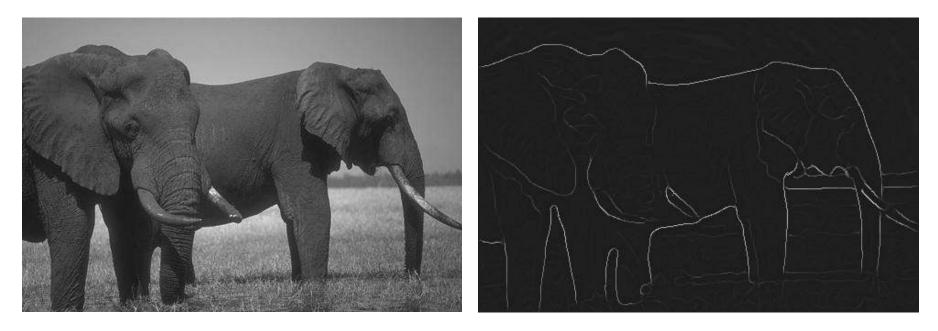


Marked by many users

http://www.eecs.berkeley.edu/Research/Projects/CS/vision/grouping/segbench/bench/html/images.html



Boundaries of Objects from Edges

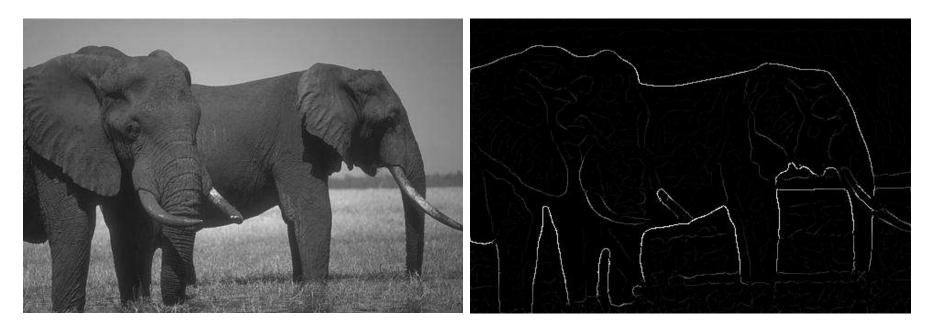


Brightness Gradient (Edge detection)

Missing edge continuity, many spurious edges



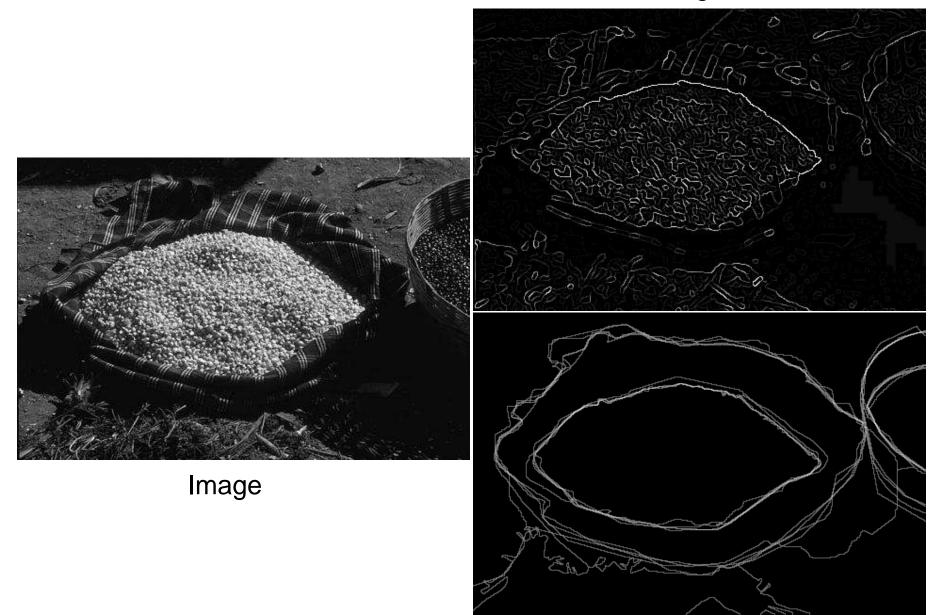
Boundaries of Objects from Edges



Multi-scale Brightness Gradient

• But, low strength edges may be very important

Machine Edge Detection



Human Boundary Marking

Boundaries in Medical Imaging

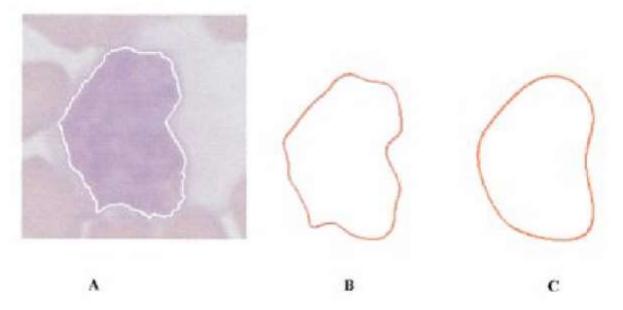


Fig. 2. Representation of a closed contour by elliptic Fourier descriptors. (a) Input. (b) Series truncated at 16 harmonics. (c) Series truncated to four harmonics.

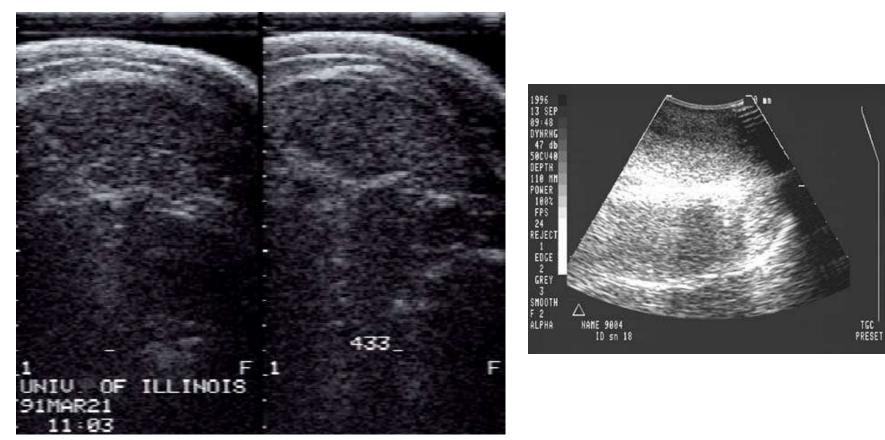
Detection of cancerous regions.



Computer Vision - T2.1 - Segmentation

[Foran, Comaniciu, Meer, Goodell, 00]

Boundaries in Ultrasound Images



Hard to detect in the presence of large amount of speckle noise





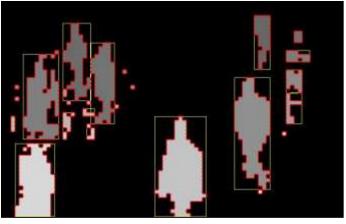
Sometimes hard even for humans!

What is 'Segmentation'?

- Separation of the image in different areas.
 - Objects.
 - Areas with similar visual or semantic characteristics.

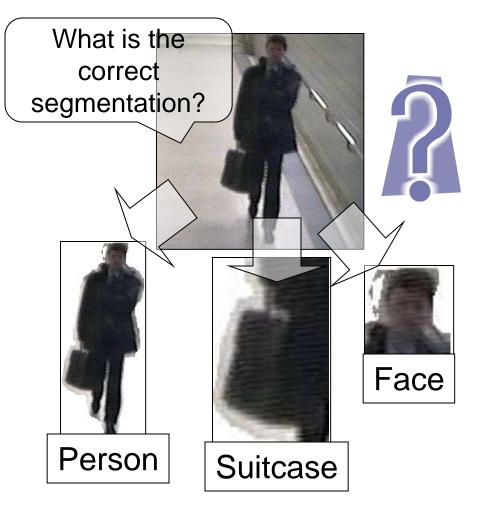
Not trivial! It is the holy grail of most computer vision problems!





Subjectivity

- A 'correct' segmentation result is only valid for a specific context.
 - Subjectivity!
 - Hard to implement.
 - Hard to evaluate.

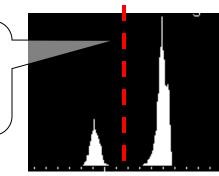


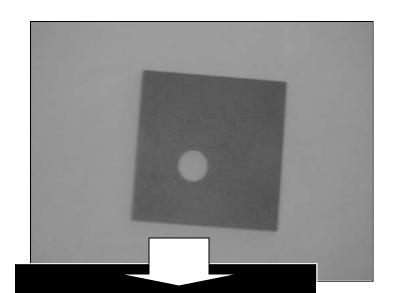


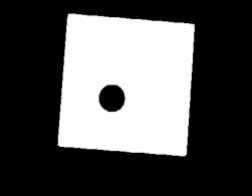
Core Technique: Thresholding

- Divide the image into two areas:
 - 1, if f(x,y) > K
 - 0, if f(x,y) <= K
- Not easy to find the ideal **k** magic number.
- Core segmentation technique
 - Simple
 - Reasonably effective

Adequate threshold

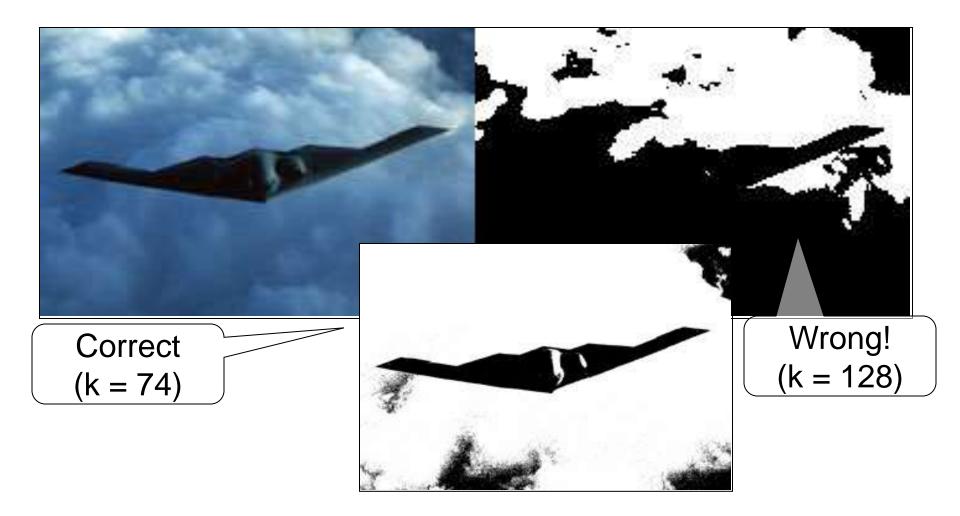








Finding the 'magic number'





Sonnet for Lena

O dear Lena, your beauty is so vast It is hard sometimes to describe it fast. I thought the entire world I would impress If only your portrait I could compress. Alas! First when I tried to use VQ I found that your checks belong to only you. Your silky hair contains a thousand lines Hard to match with sums of discrete cosines. And for your lips, sensual and tactual Thirteen Crays found not the proper fractal. And while these sotbacks are all quite severe I might have fixed them with backs here or there But when filters took sparkle from your eyes I said. 'Dumn all this. Til just digitize.'

Thomas Colthurst

Sonnet let ε

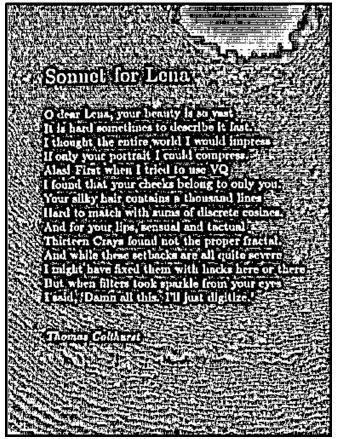
O dear k of It is hard small the state I thought the state If only your port and have had that your checks belong to only you by hair contains a thousand lines in with mins of discrete cosines is present the proper fractal.

Global thresholds are not always adequate...

http://www.cee.hw.ac.uk/hipr/html/adpthrsh.html

Adaptive Thresholding

- Adapt the threshold value for each pixel.
- Use characteristics of nearby pixels.
- How?
 - Mean
 - Median
 - Mean + K



Mean of 7x7 neighborhood



Sonnet for Lena

O dear Lena, your beauty is so vast it is hard sometimes to describe it fast. I thought the entire world I would impress If only your portrait I could compress. Alas! First when I tried to use VQ I found that your checks belong to only you. Your silky hair contains a thousand lines Hard to match with sums of discrete cosines. And for your lips, sensual and tactual Thirteen Crays found not the proper fractal. And while these setbacks are all quite severe I might have fixed them with backs here or there But when filters took sparkle from your eyes I said, 'Dann all this. TB just digitize.'

Thomas Calthurst

Sonnet for Lena

O dear Lens, your beauty is so vart. It is hard sometimes to describe it fast. I thought the entire world I would impress if only your portrait I could compress. Alas! First when I tried to use VQ I found that your cheeks belong to only you. Your silky hair contains a thousand lines Hard to match with sums of discrete cosines. And for your lips, sensual and tactual Thirteen Crays found not the proper fractal. And while these setbacks are all quite severe I might have fixed them with backs here or there But when filters took sparkle from your eyes I said. 'Dann all this. I'll just digitize.'

Thomas Callburst

7x7 window; K = 7

75x75 window; K = 10

Topic: Region-based Segmentation

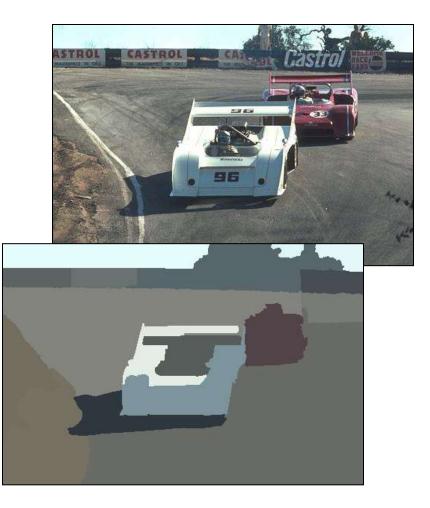
- Thresholding
- Region-based Segmentation
- Morphological Filters

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Why Region-Based Segmentation?

- Segmentation
 - Edge detection and Thresholding not always effective.
- Homogenous regions
 - Region-based segmentation.
 - Effective in noisy images.

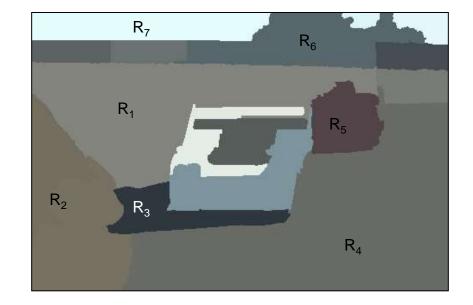


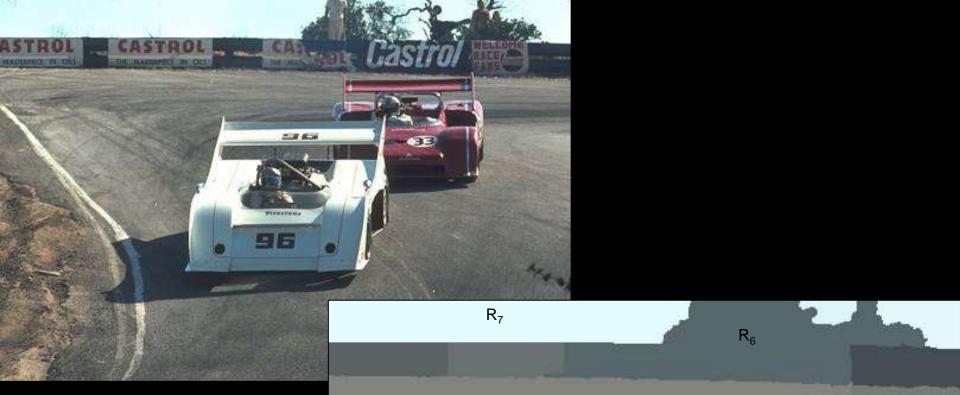


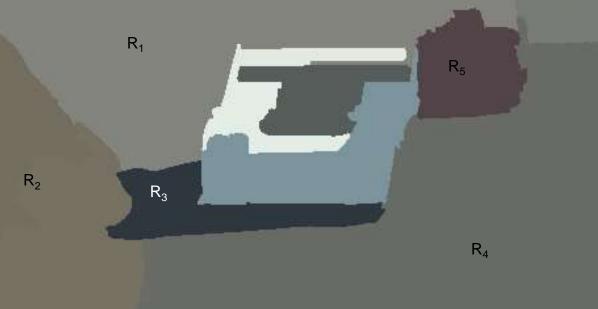
Definitions

- Based on *sets*.
- Each image R is a set of regions R_i.
 - Every pixel belongs to one region.
 - One pixel can only belong to a single region.

$$R = \bigcup_{i=1}^{S} R_i \qquad R_i \bigcap R_j \neq 0$$







Basic Formulation

Let R represent the entire image region. Segmentation partitions R into n subregions, R₁, R₂, ..., R_n, such that:

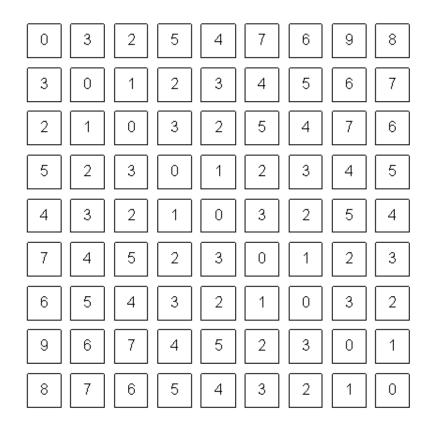
a)
$$\bigcup_{i=1}^{n} R_i = R$$

- b) R_i is a connected region, i = 1, 2, ..., n.
- C) $R_i \cap R_j = \phi$ for all *i* and $j, i \neq j$
- d) $P(R_i) = TRUE \text{ for } i = 1, 2, ..., n.$
- e) $P(R_i \bigcup R_j) = FALSE$ for $i \neq j$.

- a) Every pixel must be in a region
- b) Points in a region must be connected.
- c) Regions must be disjoint.
- All pixels in a region satisfy specific properties.
- e) Different regions have different properties.

How do we form regions?

- Region Growing
- Region Merging
- Region Splitting
- Split and Merge
- Watershed



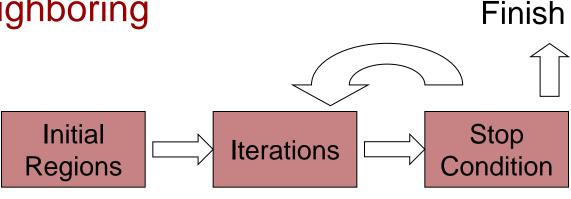
What a computer sees



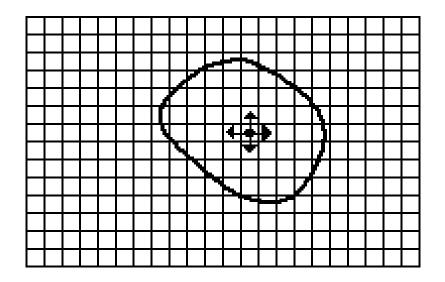
Region growing

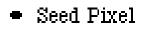
- Groups pixels into larger regions.
- Starts with a **seed** point.
- Grows region by merging neighboring pixels.

- Iterative process
 - How to start?
 - How to iterate?
 - When to stop?



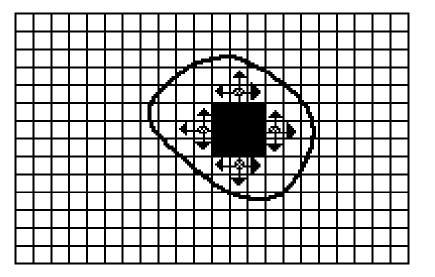


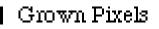




† Direction of Growth

(a) Start of Growing a Region





Pixels Being
 Considered

(b) Growing Process After a Few Iterations

Region merging

- Algorithm
 - Divide image into an initial set of regions.
 - One region per pixel.
 - Define a similarity criteria for merging regions.
 - Merge similar regions.
 - Repeat previous step until no more merge operations are possible.

Similarity Criteria

- Homogeneity of regions is used as the main segmentation criterion in region growing.
 - gray level
 - color, texture
 - shape
 - model
 - etc.

Choice of criteria affects segmentation results dramatically!

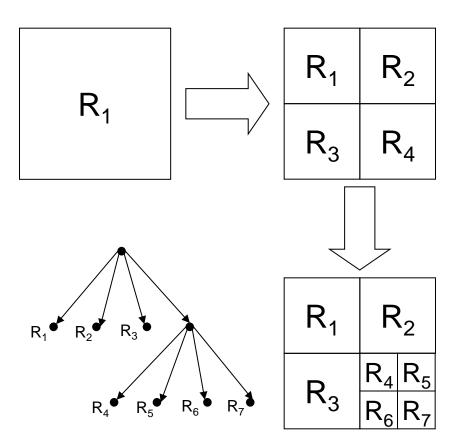
Gray-Level Criteria

- Comparing to Original Seed Pixel

 Very sensitive to choice of seed point.
- Comparing to Neighbor in Region
 - Allows gradual changes in the region.
 - Can cause significant drift.
- Comparing to Region Statistics
 Acts as a drift dampener.
- Other possibilities!

Region splitting

- Algorithm
 - One initial set that includes the whole image.
 - Similarity criteria.
 - Iteratively **split** regions into sub-regions.
 - Stop when no more splittings are possible.



The segmentation problem

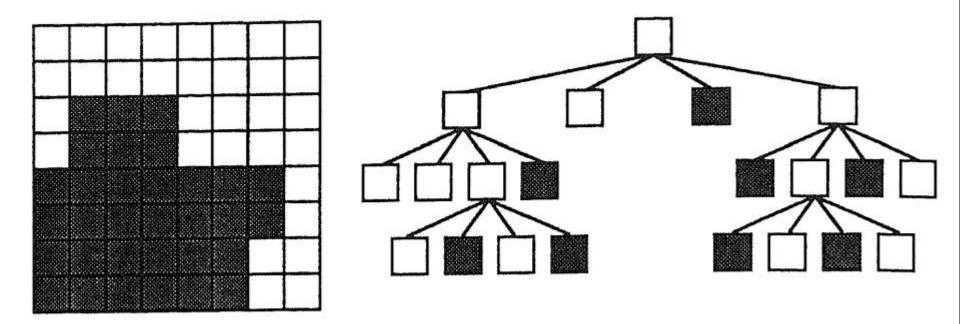
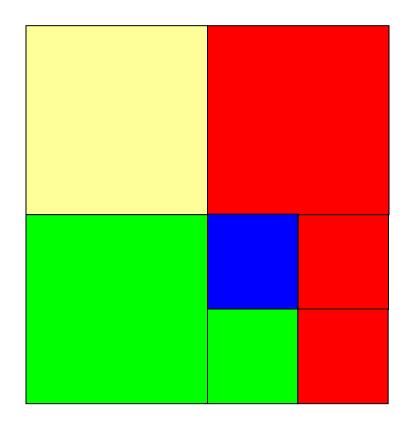


Figure 5.23 A quad-tree representation of an 8×8 binary image.

[Machine Vision; David Vernon]

Split and Merge

- Combination of both algorithms.
- Can handle a larger variety of shapes.
 - Simply apply previous algorithms consecutively.

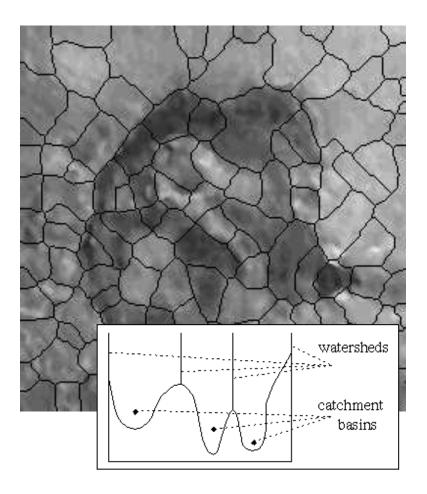




The Watershed Transform

• Geographical inspiration.

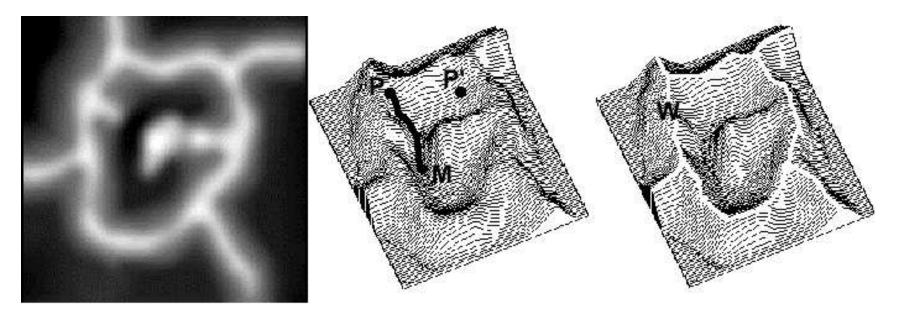
- Shed water over rugged terrain.
- Each lake corresponds to a region.
- Characteristics
 - Computationally complex.
 - Great flexibility in segmentation.
 - Risk of over-segmentation.





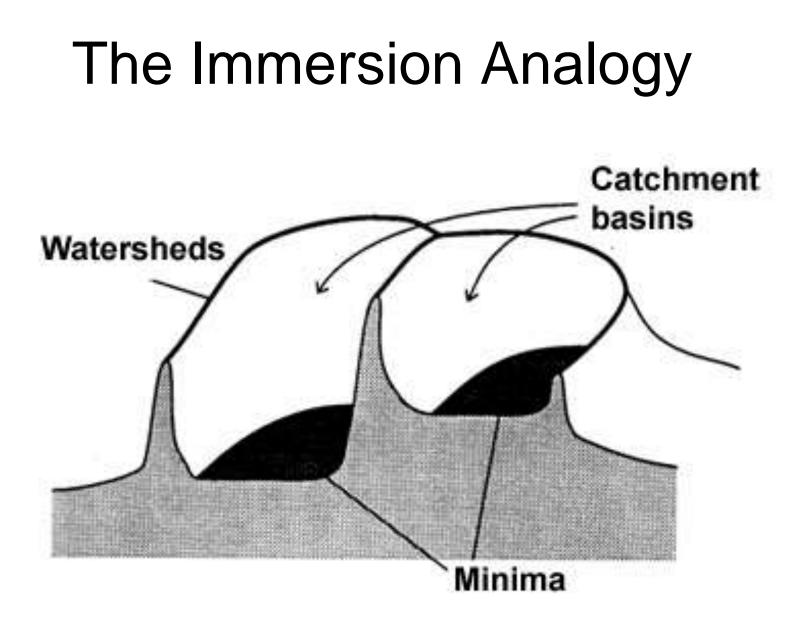
The Drainage Analogy

• Two points are in the same region if they drain to the same point.



Courtesy of Dr. Peter Yim at National Institutes of Health, Bethesda, MD







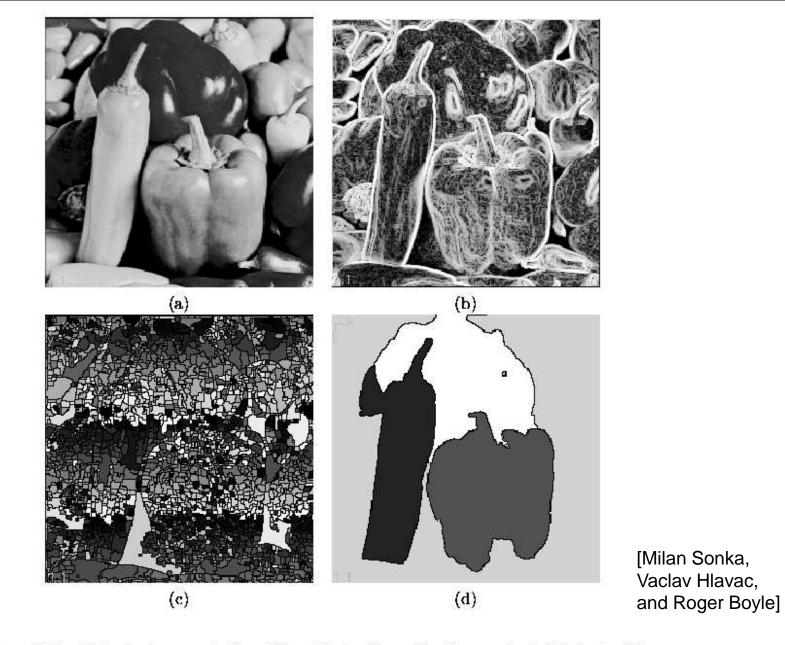


Figure 5.51: Watershed segmentation: (a) original; (b) gradient image, 3×3 Sobel edge detection, histogram equalized; (c) raw watershed segmentation; (d) watershed segmentation using region markers to control oversegmentation. Courtesy W. Higgins, Penn State University.

Over-Segmentation

- Over-segmentation.
 - Raw watershed segmentation produces a severely oversegmented image with hundreds or thousands of catchment basins.
- Post-Processing.
 - Region merging.
 - Edge information.
 - Etc.

Topic: Morphological Filters

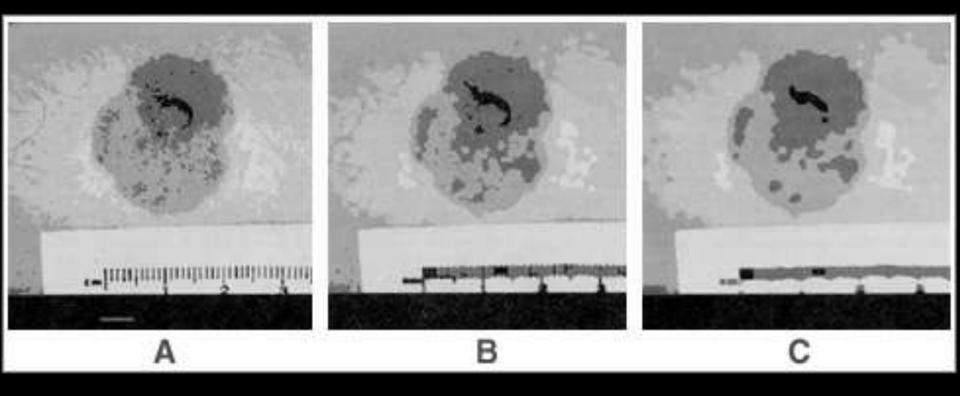
- Thresholding
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Mathematical Morphology

- Provides a mathematical description of geometric structures.
- Based on sets.
 - Groups of pixels which define an image region.

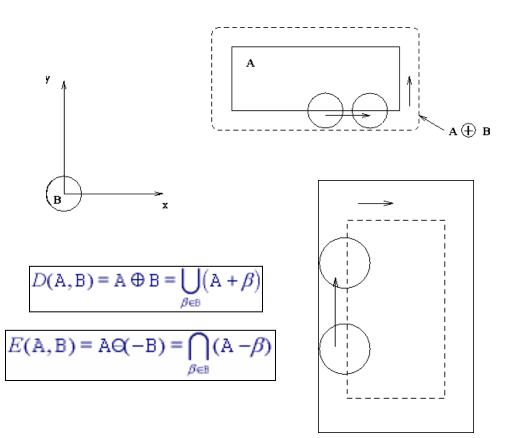
- What is this used for?
 - Binary images.
 - Can be used for postprocessing segmentation results!
- Core techniques
 - Erosion, Dilation.
 - Open, Close.



Tumor Segmentation using Morphologic Filtering

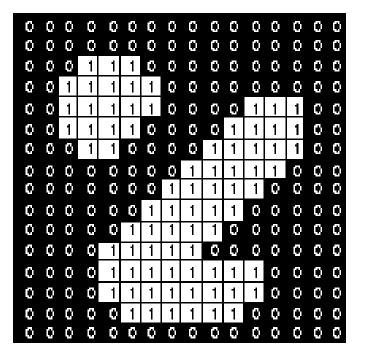
Dilation, Erosion

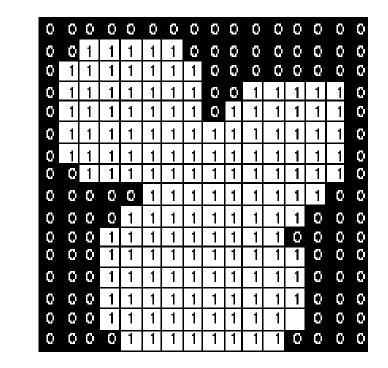
- Two sets:
 - Image
 - Morphological kernel.
- Dilation (D)
 - Union of the kernel with the image set.
 - Increases resulting area.
- Erosion (E)
 - Intersection.
 - Decreases resulting area.



Dilation

Example using a 3x3 morphological kernel

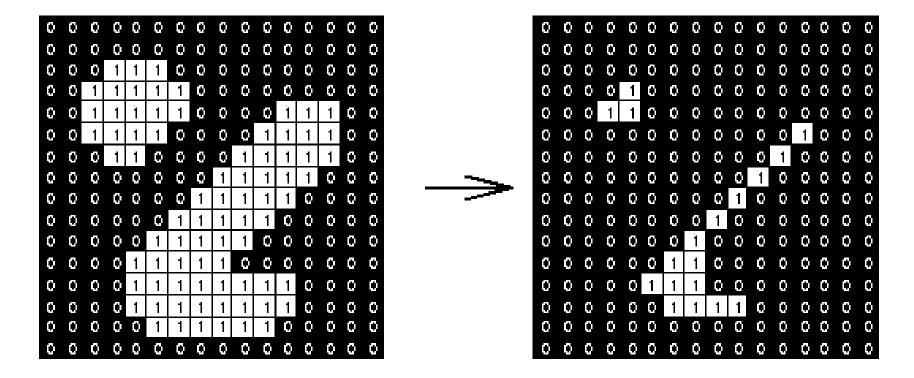






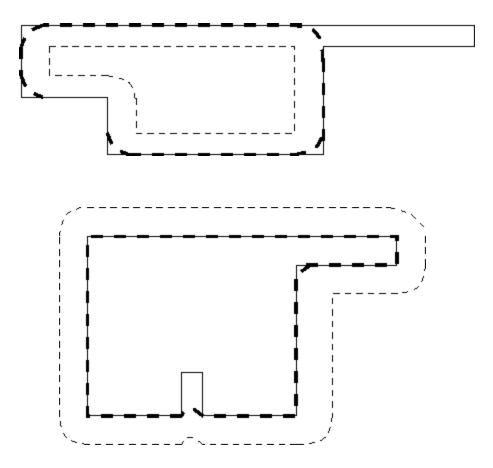
Erosion

Example using a 3x3 morphological kernel



Opening, Closing

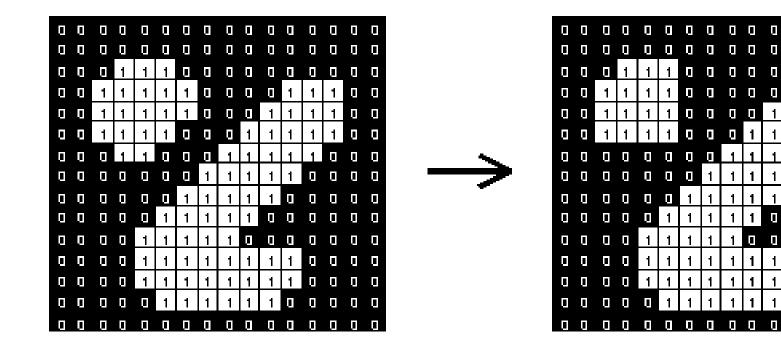
- Opening
 - Erosion, followed by dilation.
 - Less destructive than an erosion.
 - Adapts image shape to kernel shape.
- Closing
 - Dilation, followed by erosion.
 - Less destructive than a dilation.
 - Tends to close shape irregularities.





Opening

Example using a 3x3 morphological kernel





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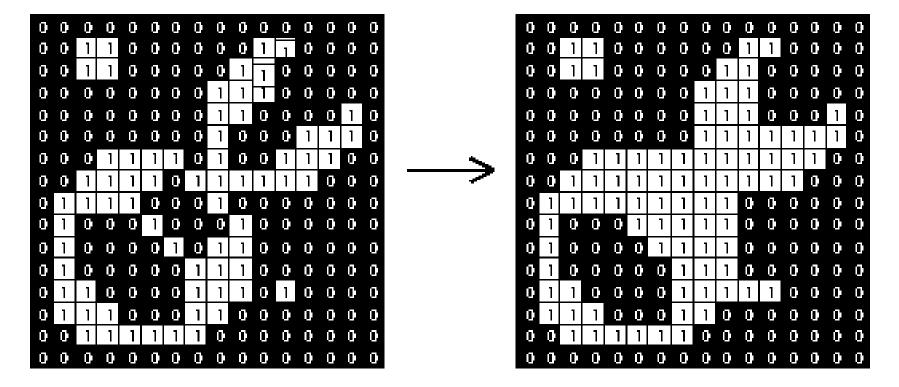
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Closing

Example using a 3x3 morphological kernel





Core morphological operators



Dilation



Closing



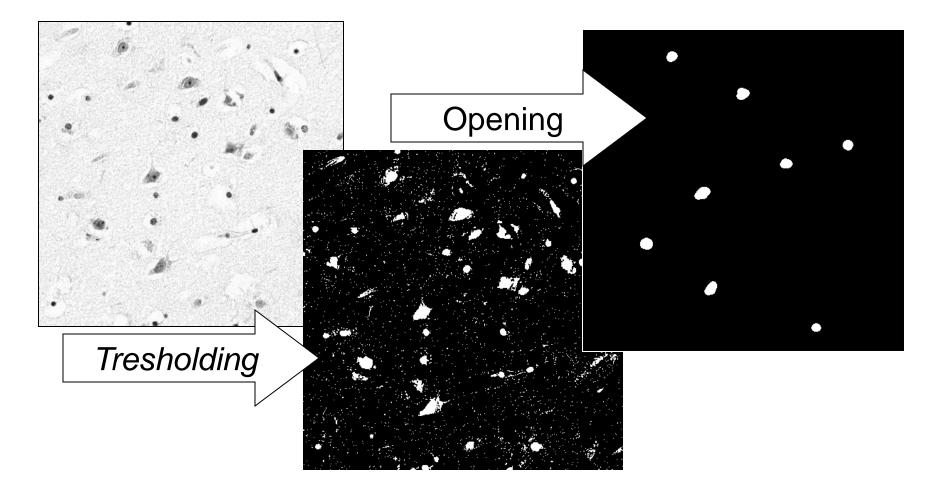
Erosion



Opening

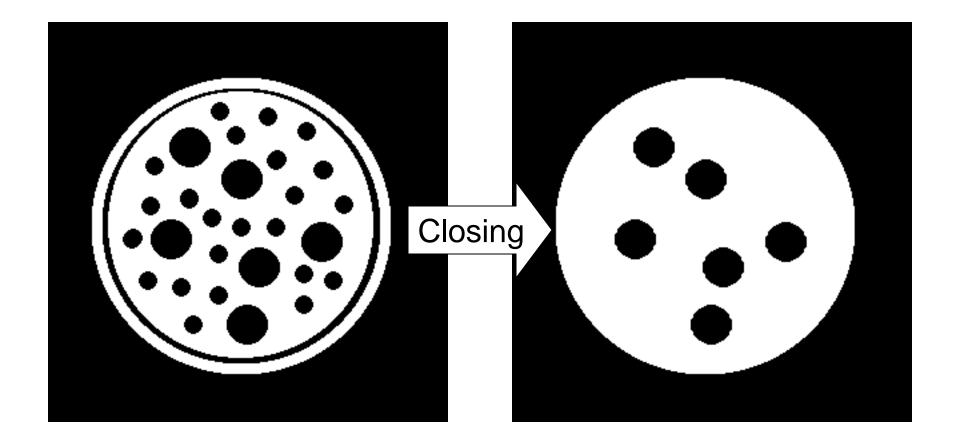


Example: Opening





Example: Closing

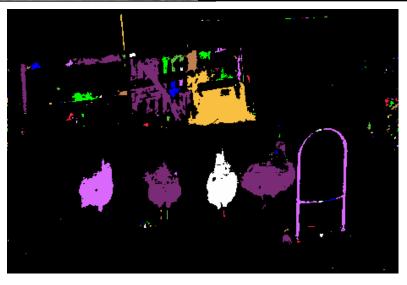




Connected Component Analysis

- Define 'connected'.
 - 4 neighbors.
 - 8 neighbors.
- Search the image for seed points.
- Recursively obtain all connected points of the seeded region.







Resources

- Gonzalez & Woods Chapter 7 and 8
- Russ Chapter 7
- N. Otsu, "A threshold selection method from gray-level histograms," IEEE Trans. Sys., Man., Cyber., vol. 9, pp. 62–66, 1979.

