#### Announcements

- Project 3 questions
- Photos after class

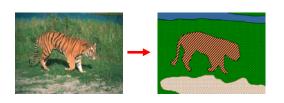
# **Image Segmentation**



#### Today's Readings

- Shapiro, pp. 279-289
  - http://www.dai.ed.ac.uk/HIPR2/morops.htm
  - Dilation, erosion, opening, closing

## From images to objects



#### What Defines an Object?

- Subjective problem, but has been well-studied
- Gestalt Laws seek to formalize this
  - proximity, similarity, continuation, closure, common fate
  - see  $\underline{\text{notes}}$  by Steve Joordens, U. Toronto

## Image Segmentation

We will consider different methods

Already covered:

• Intelligent Scissors (contour-based, manual)

Today—automatic methods:

- K-means clustering (color-based)
- Normalized Cuts (region-based)

## Image histograms



How many "orange" pixels are in this image?

- This type of question answered by looking at the histogram
- A histogram counts the number of occurrences of each color
  - Given an image

 $F[x, y] \rightarrow RGB$ 

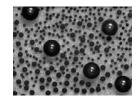
- The histogram is defined to be

$$H_F[c] = |\{(x, y) \mid F[x, y] = c\}|$$

- What is the dimension of the histogram of an NxN RGB image?

## What do histograms look like?

Photoshop demo





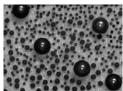
How Many Modes Are There?

• Easy to see, hard to compute

## Histogram-based segmentation

#### Goal

- Break the image into K regions (segments)
- Solve this by reducing the number of colors to K and mapping each pixel to the closest color
  - photoshop demo

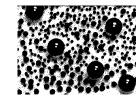


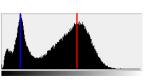


## Histogram-based segmentation

#### Goa

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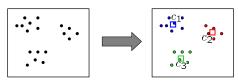


Here's what it looks like if we use two colors

## Clustering

How to choose the representative colors?

· This is a clustering problem!



#### Objective

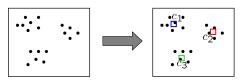
· Each point should be as close as possible to a cluster center - Minimize sum squared distance of each point to closest center

$$\sum_{\text{clusters } i} \quad \sum_{\text{points p in cluster } i} \|p - c_i\|^2$$

## Break it down into subproblems

Suppose I tell you the cluster centers c<sub>i</sub>

- Q: how to determine which points to associate with each c;?
- A: for each point p, choose closest  $c_{i}$



Suppose I tell you the points in each cluster

- Q: how to determine the cluster centers?
  A: choose c<sub>i</sub> to be the mean of all points in the cluster

## K-means clustering

K-means clustering algorithm

- 1. Randomly initialize the cluster centers,  $c_1$ , ...,  $c_K$
- 2. Given cluster centers, determine points in each cluster
  - For each point p, find the closest  $c_i$ . Put p into cluster i
- 3. Given points in each cluster, solve for c<sub>i</sub>
- Set c<sub>i</sub> to be the mean of points in cluster i
- 4. If c, have changed, repeat Step 2

Java demo: http://www.elet.polimi.it/upload/matteucc/Clustering/tutorial\_html/AppletKM.html

#### Properties

- Will always converge to some solution
- Can be a "local minimum"
  - does not always find the global minimum of objective function:

$$\sum_{\text{clusters } i} \quad \sum_{\text{points p in cluster } i} \|p - c_i\|^2$$

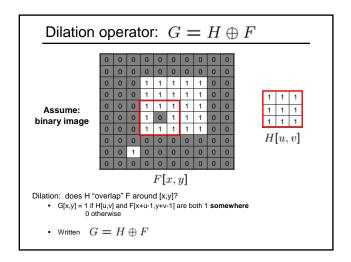
### Cleaning up the result

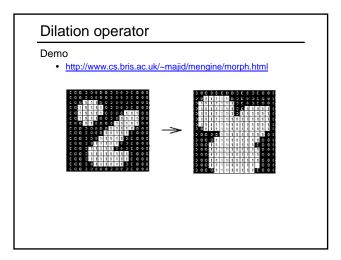
#### Problem:

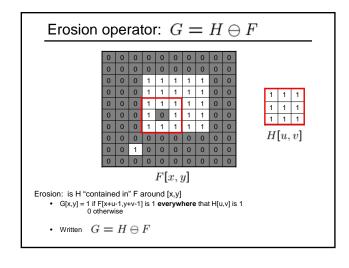
- Histogram-based segmentation can produce messy regions
  - segments do not have to be connected
  - may contain holes

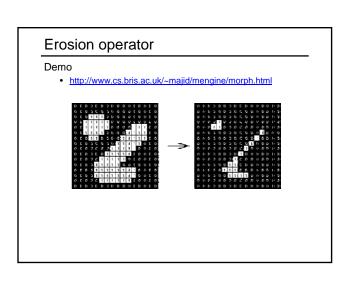
How can these be fixed?

photoshop demo









#### Nested dilations and erosions

What does this operation do?

$$G = H \ominus (H \oplus F)$$

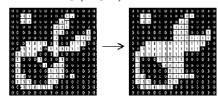


• this is called a **closing** operation

#### Nested dilations and erosions

What does this operation do?

$$G = H \ominus (H \oplus F)$$



• this is called a closing operation

Is this the same thing as the following?

$$G=H\oplus (H\ominus F)$$

#### Nested dilations and erosions

What does this operation do?

$$G = H \oplus (H \ominus F)$$

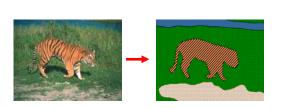
- this is called an **opening** operation
- http://www.dai.ed.ac.uk/HIPR2/open.htm

You can clean up binary pictures by applying combinations of dilations and erosions

Dilations, erosions, opening, and closing operations are known as **morphological operations** 

• see <a href="http://www.dai.ed.ac.uk/HIPR2/morops.htm">http://www.dai.ed.ac.uk/HIPR2/morops.htm</a>

## Graph-based segmentation?



## Images as graphs





#### Fully-connected graph

- · node for every pixel
- link between every pair of pixels, p,q
- cost cpq for each link

# Segmentation by Graph Cuts

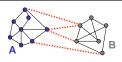




#### Break Graph into Segments

- · Delete links that cross between segments
- · Easiest to break links that have low cost (similarity)
  - similar pixels should be in the same segments
  - dissimilar pixels should be in different segments

## Cuts in a graph



#### Link Cut

- set of links whose removal makes a graph disconnected
- · cost of a cut:

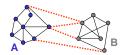
$$cut(A, B) = \sum_{p \in A, q \in B} c_{p,q}$$

## Find minimum cut

- gives you a segmentation fast algorithms exist for doing this

# But min cut is not always the best cut... Min-cut 2 • nl Min-cut 1 better cut -

# Cuts in a graph



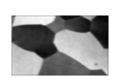
#### Normalized Cut

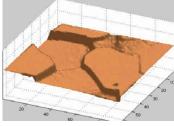
- a cut penalizes large segments
- fix by normalizing for size of segments

$$Ncut(A,B) = \frac{cut(A,B)}{volume(A)} + \frac{cut(A,B)}{volume(B)}$$

• volume(A) = sum of costs of all edges that touch A

# Interpretation as a Dynamical System





Treat the links as springs and shake the system

• elasticity proportional to cost

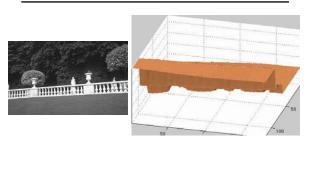
• vibration "modes" correspond to segments

- can compute these by solving an eigenvector problem

- for more details, see

\* J. Shi and J. Malik, Normalized Cuts and Image Segmentation, CVPR, 1997

# Interpretation as a Dynamical System



# Color Image Segmentation

