BSB 663 Image Processing Feb. 26, 2013

Erkut Erdem Hacettepe University

Introduction

Instructor and Course Schedule

- Dr. Erkut ERDEM
- <a>erkut@cs.hacettepe.edu.tr
- Office: II4
- Tel: 297 7500 / 149
- Lectures (BSB663): Tuesday, 18:15-21:00@D7

About BSB663

- A course about the fundamentals of image processing.
- The aim of this course is to provide an introduction to students who wish to specialize in interrelated disciplines like image processing, computer vision.
- The students will also be expected to gain hand-on experience via a set of programming assignments.

Communication

• The course webpage will be updated regularly throughout the semester with lecture notes, programming and reading assignments and important deadlines.

http://web.cs.hacettepe.edu.tr/~erkut/bsb663.s13

 All other communications will be carried out through Piazza. Please enroll it by following the link <u>https://piazza.com/hacettepe.edu.tr/spring2013/</u> <u>bsb663</u>

Prerequisites

- Programming skills (Matlab)
- Good math background (Calculus, Linear Algebra, Statistical Methods)
- Students are not expected to have any prior knowledge of image processing techniques.

Reading Material

- Lecture notes and handouts
- Papers and journal articles

Textbooks



Rafael C. Gonzalez Richard E. Woods Computer Vision: Algorithms and Applications, Richard Szeliski, Springer, 2010

Digital Image Processing,
R. C. Gonzalez, R. E. Woods,
3rd Edition, Prentice Hall, 2008

Grading Policy

- BSB663 Image Processing
 - Class participation, 5%
 - Reading Assignments, 5%
 - Programming Assignments, 20%
 - Midterm Exam, 30%
 - Final Exam, 40%

Programming Assignments

- Three assignments related to the topics covered in the class.
- Each assignment will involve implementing an algorithm, carrying out a set of experiments to evaluate it, and writing up a report on the experimental results.
- <u>All assignments have to be done individually, unless</u> <u>stated otherwise.</u>

Important Dates

- Programming Assignment I
- Programming Assignment 2
- Programming Assignment 3
- Midterm exam
- Final exam

26 March 2013 4 April 2013 14 May 2013

4 April 2013 To be announced later

Tentative Outline

- (I week) Image formation and the digital camera
- (I week) Color perception and color spaces
- (I week) Point operations
- (I week) Spatial filtering
- (I week) Fourier Transform
- (I week) Image pyramids and wavelets
- (I week) Gradients, edges, contours
- (I week) Image smoothing
- (2 weeks) Image segmentation
- (I week) Advanced topics

Image Processing



What does it mean, to see?

- "The plain man's answer (and Aristotle's, too) would be, to know what is where by looking. In other words, vision is the process of discovering from images what is present in the world, and where it is." David Marr, Vision, 1982
- Our brain is able to use an image as an input, and interpret it in terms of objects and scene structures.



What does Salvador Dali's Study for the Dream Sequence in Spellbound (1945) say about our visual perception?

We see a two dimensional image

But, we perceive depth information

light reflected on the retina

converging lines

shadows of the eye

Why does vision appear easy to humans?

- Our brains are specialized to do vision.
- Nearly half of the cortex in a human brain is devoted to doing vision (cf. motor control ~20-30%, language ~10-20%)
- "Vision has evolved to convert the <u>ill-posed problems</u> into solvable ones by adding premises: <u>assumptions</u> about how the world we evolved in is, on average, put together"
 Steven Pinker, How the Mind Works, 1997
- Gestalt Theory (Laws of Visual Perception), Max Wertheimer, 1912





Figures: Steven Pinker, How the Mind Works, 1997

Computer Vision

- "Vision is a process that produces from images of the external world a description that is useful to the viewer and not cluttered with irrelevant information" ~David Marr
- <u>The goal of Computer Vision</u>: To develop artificial machine vision systems that make inferences related to the scene being viewed through the images acquired with digital cameras.

MASSACHUSETTS INSTITUTE OF TECHNOLOGY PROJECT MAC

Artificial Intelligence Group Vision Memo. No. 100. July 7, 1966

THE SUMMER VISION PROJECT

Seymour Papert

The summer vision project is an attempt to use our summer workers effectively in the construction of a significant part of a visual system. The particular task was chosen partly because it can be segmented into sub-problems which will allow individuals to work independently and yet participate in the construction of a system complex enough to be a real landmark in the development of "pattern recognition".

Origins of computer vision

-23-4445(a-d)



(a) Original picture.





(c) Line drawing.

L. G. Roberts, *Machine Perception of Three Dimensional Solids,* Ph.D. thesis, MIT Department of Electrical Engineering, 1963.

(d) Rotated view.

Slide credit: S. Lazebnik

Marr's observation: Studying vision at 3 levels

- Vision as an information processing task [David Marr, 1982]
- Three levels of understanding:
- I. Computational theory
 - What is computed? Why it is computed?
- 2. Representation and Algorithm
 - How it is computed?
 - Input, Output, Transformation
- 3. Physical Realization
 - Hardware

Visual Modules and the Information Flow



- Visual perception as a data-driven, bottom-up process (traditional view since D. Marr)
- Unidirectional information flow
- Simple low-level cues >> Complex abstract perceptual units

Visual Modules and the Information Flow



- Vision modules can be categorized into three groups according to their functionality:
 - Low-level vision: filtering out irrelevant image data
 - Mid-level vision: grouping pixels or boundary fragments together
 - High-level vision: complex cognitive processes

Fundamentals of Image Processing



• How we process images?

Image Formation



- What is measured in an image location?
 - brightness
 - color



Figures: Francis Crick, The Astonishing Hypothesis, 1995

Image Formation



- Discretization
 - in image space sampling
 - In image brightness quantization

Figures: Gonzalez and Woods, Digital Image Processing, 3rd Edition, 2008

Image Representation

- **Digital image:** 2D discrete function f
- **Pixel**: Smallest element of an image f(x,y)



Figure: M. J. Black

Image Representation

- **Digital image:** 2D discrete function f
- **Pixel**: Smallest element of an image f(x,y)





- Two types of receptor cells in retina:
 - Cone Receptor cells: 6-7 million → function in bright light, color sensitive, fine detail
 - Rod receptor cells: 75-150 million → function in dim light, color insensitive, coarse detail
- A recent discovery: Photosensitive retinal ganglion cells \rightarrow sensitive to blue light

Figures: Giomzelezrancis/Vorocks, Digitastoniastengrobyessimesis, d19005ion, 2008

Hierarchy of Visual Areas

• There are many different neural connections between different visual areas.



Figures: Nikos K. Logothetis, Vision: A Window on Consciousness, SciAm, Nov 1999F (on the left) Felleman & van Essen, 1991 (on the right)

Visual Modules and the Information Flow



- Vision modules can be categorized into three groups according to their functionality:
 - Low-level vision: filtering out irrelevant image data
 - Mid-level vision: grouping pixels or boundary fragments together
 - High-level vision: complex cognitive processes

Sample Problems

- Edge Detection
- Image Denoising
- Image Smoothing
- Image Segmentation
- Image Registration
- Image Inpainting
- Image Retargeting
- •

Image Filtering

• Filtering out the irrelevant information



- Image denoising, image sharpening, image smoothing, image deblurring, etc.
- Edge detection

Edge Detection





Canny edge detector

- Edges: abrupt changes in the intensity
 - Uniformity of intensity or color
- Edges to object boundaries

Image Filtering

• **Difficulty:** Some of the irrelevant image information have characteristics similar to those of important image features



Image Smoothing - A Little Bit of History

- Gaussian Filtering / linear diffusion
 - the most widely used method

$$\frac{\partial u}{\partial t} = \nabla \cdot (\nabla u) = \nabla^2 u$$



- mid 80's unified formulations
 - methods that combine smoothing and edge detection
 - Geman & Geman'84, Blake & Zisserman'87, Mumford & Shah'89, Perona & Malik'90

Image Denoising

• Images are corrupted with 70% salt-and-pepper noise



R. H. Chan, C.-W. Ho, and M. Nikolova, Salt-and-Pepper Noise Removal by Median-Type Noise Detectors and Detail-Preserving Regularization. IEEE TIP 2005

Non-local Means Denoising









Figure 1. Scheme of NL-means strategy. Similar pixel neighborhoods give a large weight, w(p,q1) and w(p,q2), while much different neighborhoods give a small weight w(p,q3).

Preserve fine image details and texture during denoising

A. Buades, B. Coll, J. M. Morel, A non-local algorithm for image denoising, CVPR, 2005

Context-Guided Filtering

• Use local image context to steer filtering



Preserve main image structures during filtering

E. Erdem and S. Tari, Mumford-Shah Regularizer with Contextual Feedback, JMIV, 2009

Image Smoothing



L. Xu, C. Lu, Y. Xu, J. Jia, Image Smoothing via L0 Gradient Minimization, SIGGRAPH ASIA 2011

Image Smoothing



L. Karacan, E. Erdem, A. Erdem, work in progress

Image Segmentation

• Partition an image into meaningful regions that are likely to correspond to objects exist in the image



Grouping of pixels according to what criteria?

high-level object specific knowledge matters!

Figures: A. Erdem

Image Segmentation

- Boundary-based segmentation
- Region-based segmentation
- Unified formulations

Snakes

• Curve Evolution - parametric curve formulation



Fig. 3. Two edge snakes on a pear and potato. Upper-left: The user has pulled one of the snakes away from the edge of the pear. Others: After the user lets go, the snake snaps back to the edge of the pear.

Snakes

• Curve Evolution - parametric curve formulation



Non-rigid, deformable objects can change their shape over time, e.g. lips, hands...

M. Kass, A. Witkin, and D. Terzopoulos, Snakes: Active Contour Models, IJCV, 1988

Normalized Cuts

• A graph-theoretic formulation for segmentation



J. Shi and J. Malik, Normalized Cuts and Image Segmentation, IEEE Trans. Pattern Anal. Mach. Intel.

Normalized Cuts



Side credit: S. Lazebnik

From contours to regions

• <u>State-of-the-art:</u> gPb-owt-ucm segmentation algorithm



P. Arbelaez, M. Maire, C. Fowlkes and J. Malik, Contour Detection and Hierarchical Image Segmentation, IEEE Trans Pattern Anal. Mach. Intell. 33(5):898-916, 2011

From contours to regions

• <u>State-of-the-art:</u> gPb-owt-ucm segmentation algorithm



P. Arbelaez, M. Maire, C. Fowlkes and J. Malik, Contour Detection and Hierarchical Image Segmentation, IEEE Trans Pattern Anal. Mach. Intell. 33(5):898-916, 2011

Prior-Shape Guided Segmentation

 Incorporate prior shape information into the segmentation process



Registration

- Estimate a transformation function between
 - two images
 - two point sets
 - two shapes
 - ...

Registration



Fig. 5. Experiments on deformation. Each column represent one example. From left to right, increasing degree of deformation. *Top row*: warped template. *Second row*: template and target (same as the warped template). *Third row*: ICP results. *Bottom row*: RPM results.

H. Chui and A. Rangarajan, A new point matching algorithm for non-rigid registration, CVIU, 2003

Image Registration



Fig. 2. An example of a geodesic between images (original images taken from the Olivetti face database). The three intermediate images are generated by the optimization algorithm.



A tumor progressively appearing on a brain



Tumor: Reference image, registered target and deformation

(top) Alain Trouve and Laurent Younes, Metamorphoses Through Lie Group Action, Found. Comput. Math., 2005 (bottom) M. I. Miller and L. Younes, Group Actions, Homeomorphisms, and Matching: A General Framework, IJCV, 2001

Image Inpainting

• Reconstructing lost or deteriorated parts of images



M. Bertalmio, G. Sapiro, V. Caselles and C. Ballester, Image Inpainting, SIGGRAPH, 2000

Image Retargetting

 automatically resize an image to arbitrary aspect ratios while preserving <u>important image features</u>



How we define the importance?



S. Avidan and A. Shamir, Seam Carving for Content-Aware Image Resizing, SIGGRAPH, 2007

Image retargeting by Seam Carving with different importance maps

