

Computer Vision

Introduction

Filippo Bergamasco (<u>filippo.bergamasco@unive.it</u>)
http://www.dais.unive.it/~bergamasco
DAIS, Ca' Foscari University of Venice
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About this course

Official UNIVE course page:

http://www.unive.it/data/course/230457

Unofficial course page:

http://www.dais.unive.it/~bergamasco/computer_vision_2017_2018.html

- 30 hours frontal lessons (lab included)
- Timetable: Monday and Tuesday 12:15 1:45 pm
- Code: CM0193



About the teacher



Filippo Bergamasco

http://www.dais.unive.it/~bergamasco

- I'm currently a postdoctoral research fellow here at DAIS
- KiiS research group. I work on 3D reconstruction, camera calibration, ego-motion estimation, photogrammetry etc.



About the teacher



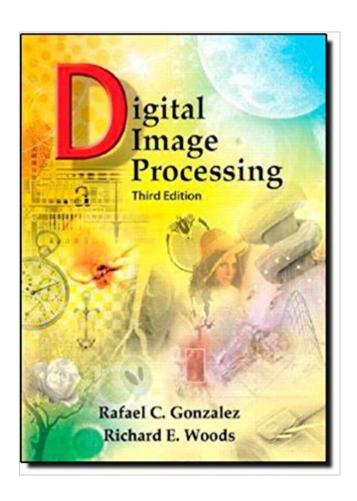
Filippo Bergamasco

http://www.dais.unive.it/~bergamasco/

For any question regarding this course, please contact me via mail: filippo.bergamasco@unive.it
Or come visit me at my office



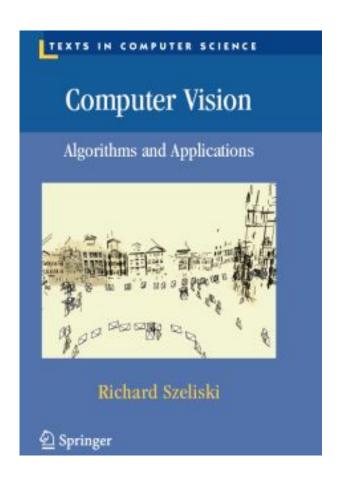
Referral Texts



R. C. Gonzalez e R.E. Woods. **Digital Image Processing** (3rd edition). Pretience Hall



Referral Texts

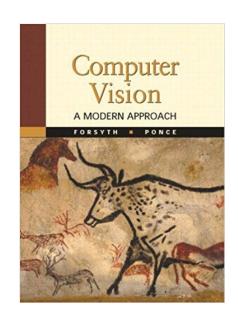


R. Szeliski. Computer Vision Algorithms and Applications. Springer

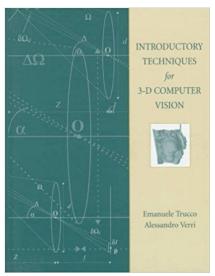


Additional useful readings

D. Forsyth and J. Ponce.
 Computer Vision. A Modern
 Approach. Prentice-Hall, 2002



E. Trucco and A. Verri.
 Introductory Techniques for
 3D Computer Vision.
 Prentice-Hall, 1998





About the exam...

Final exam consists in the development of a software project to be performed individually.

Project submission information: (more will follow)

- Send me the full source-code and data at least 1 week before the exam
- During the exam, I'll ask you to show me your project and explain its details (choice made, trade-offs, strengths and limitations)
- The goal is to evaluate your knowledge of the most common computer vision techniques and theories



About the exam...

This year project will be about the development of a computer vision system to cheat at Blackjack

Requirements:

- The project must be developed in C++ and can use the OpenCV Computer Vision Library
- I'm not particularly interested in the coding style, but a clean object-oriented organization of the code is considered a plus
- Laboratory lessons are designed to let you develop useful core parts for your project (so use the lab time wisely!)



About the exam...

This year project will be about the development of a computer vision system to cheat at Blackjack

DEMO TIME!



Are you interested in a challenging computer vision thesis?



WE WANT YOU!

A thesis activity can substitute the proposed project and assignments.

If interested, contact me as soon as possible so we can discuss the details together.



About the lab

We will spend some hours in the lab so you can put your hands on some of the algorithms discussed during the course

After each lab session you will be asked to complete an assignment within 2 weeks (deadline can be slightly postponed in exceptional cases)

Completing all the assignments is required (not sufficient) to obtain the full-marks (lode)



Computer vision is about building an automatic system that "sees"

 Difficult to give a concise definition since the area spans multiple different problems (we will see some examples in a minute...)

How we'll consider cv in this course:

"A set of computational techniques aiming at estimating or making explicit the **geometric** and **dynamic properties** of the **3D world** from **digital images**"

Introductory techniques for 3D computer vision - E. Trucco, A. Verri



Different levels of vision:

- 1. Low (Image processing)
 - a. Image restoration
 - b. Contrast enhancement
 - c. Noise reduction

Image Image

2. Medium

- a. Segmentation
- b. Shape recognition

Images attributes/features

3. High

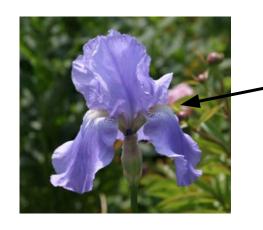
a. Scene understanding

Images — concepts



The purpose of a machine vision system is to produce a symbolic description of what is being imaged

As human, we perceive the three-dimensional structure of the world around us with apparent ease

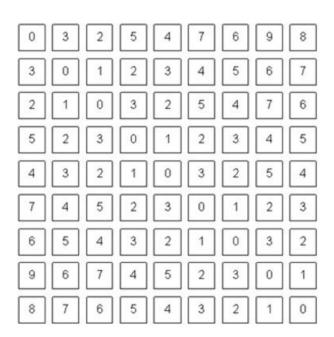


Easy to distinguish the flower in the foreground from the leaves in the background

From a signal point of view, there are just variations in color and brightness







What we see

What a computer sees



A difficult business

Despite the remarkable advances in the last decades...

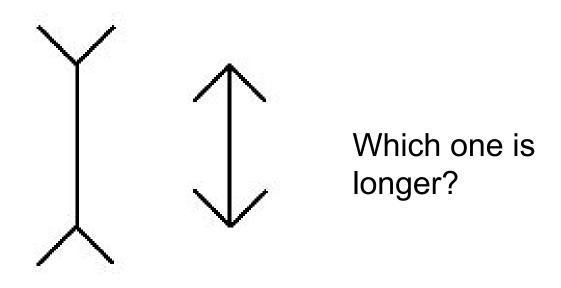
"the dream of having a computer interpret an image at the same level as a two-year old (for example, counting all of the animals in a picture) remains elusive"

Humans and animals can interpret images so effortlessly while computer vision algorithms are still error prone:

 Perceptual components of artificial intelligence (such as vision) is probably as difficult as cognitive ones (logic proving, planning)

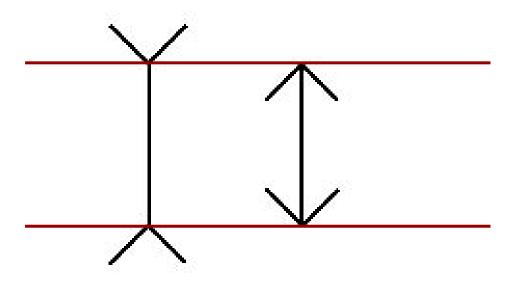


Optical illusions give us interesting clues of how the human visual system works (and the inherent assumptions made by our brain)

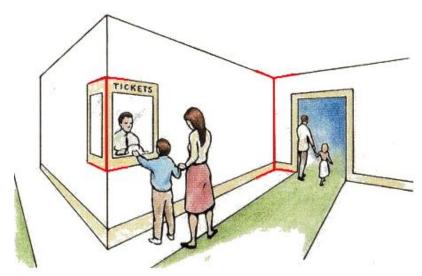


Müller-Lyer Illusion

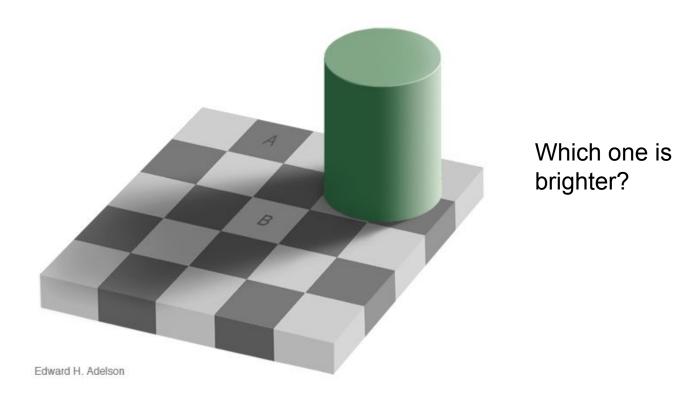




The Müller-Lyer illusion occurs because the visual system processes that judge depth and distance assume in general that the "angles in" configuration corresponds to an object which is closer, and the "angles out" configuration corresponds to an object which is far away.







The percept is due to brightness constancy, the visual system's attempt to discount illumination when interpreting colors



Rotating mask illusion



When determining the shape of an object only from shading (with no stereo vision), the brain assumes a convex shape (especially for faces)



Optical Perception





Why is difficult?

Giving a symbolic description of the contents of an image is difficult because is intrinsically an **inverse problem**:

- We want to recover some unknowns given insufficient information to fully specify the solution
- We must resort to physic-based and/or probabilistic-based models to disambiguate potential solutions



Computer Vision vs. Graphics

Computer graphics

3d to 2d implies information loss





Computer Vision

Need for models

Raster image

- A house
- Some polygons
- Lines, edges



Computer Vision vs. Graphics

The **models** we use in computer vision are usually developed by:

- Physics: radiometry, optics, etc.
- Computer graphics

And describe how:

- Objects move or animate
- Light reflects from object surfaces
- Shapes get projected to the sensor image plane
- etc.



Computer Vision vs. Graphics

In limited contexts, computer graphics nowadays is so advanced that the illusion of reality is almost perfect





So, what are the most common applications in which computer vision is used today?



Optical Character Recognition (OCR)





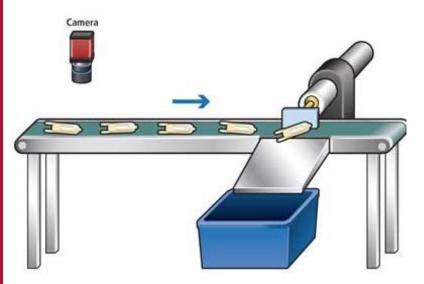


Optical Character Recognition (OCR)





Machine inspection

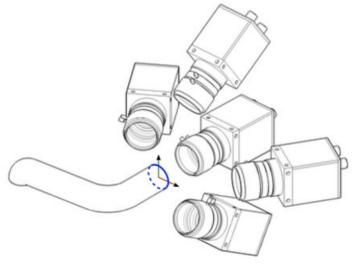






Machine inspection

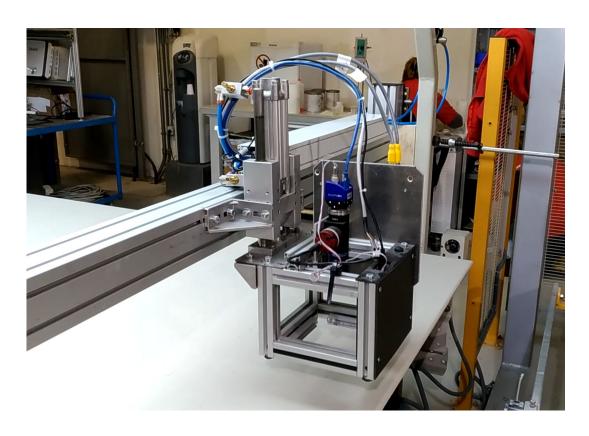




Pipe inspection device developed in our lab



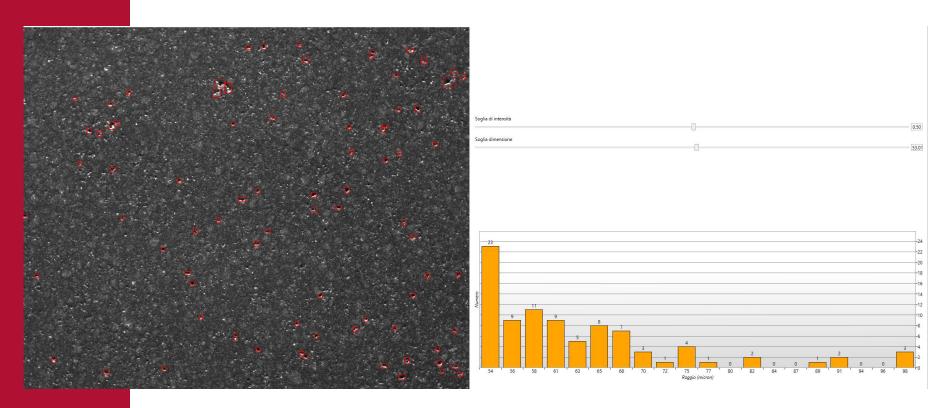
Machine inspection



Kitchen countertops inspection device developed in our lab



Machine inspection



Kitchen countertops inspection device developed in our lab



Surveillance

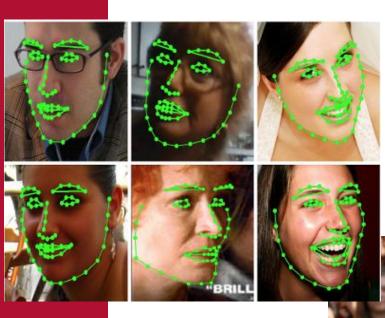


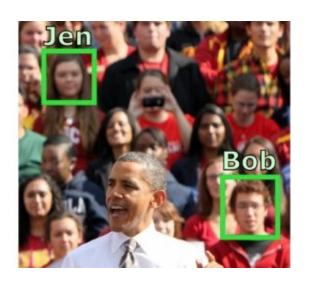




Face detection and recognition

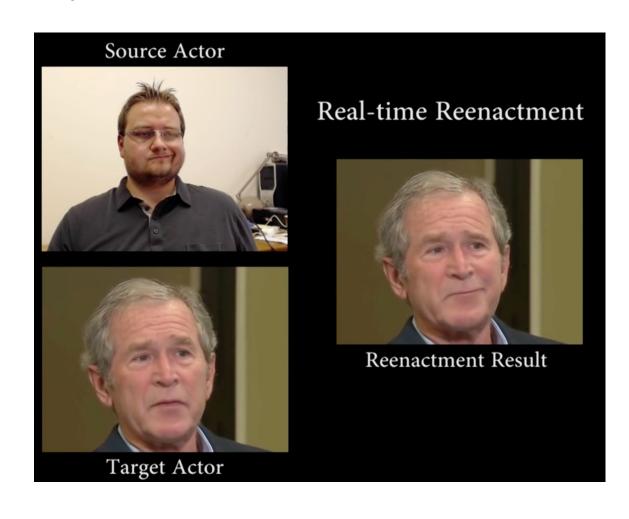
Barack Obama





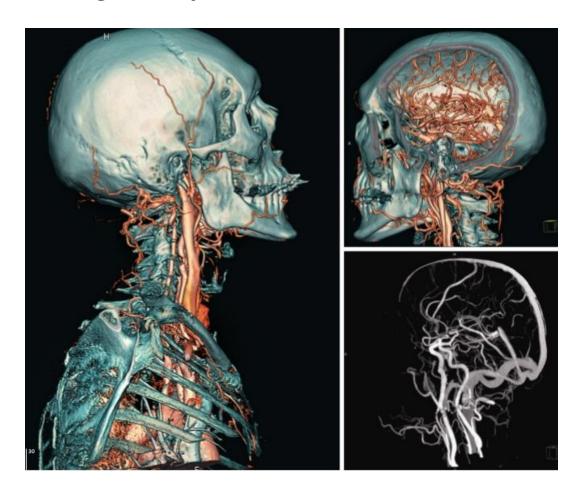


Face capture and reenactment





Medical image analysis





Driving assistance



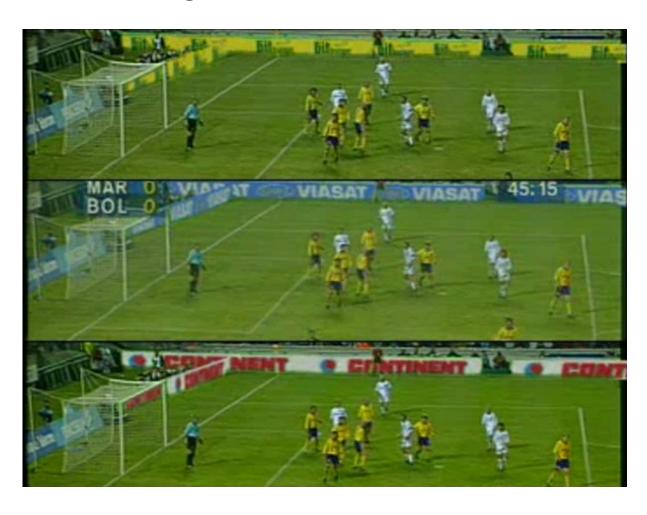


Driving autopilot (autonomous vehicles)



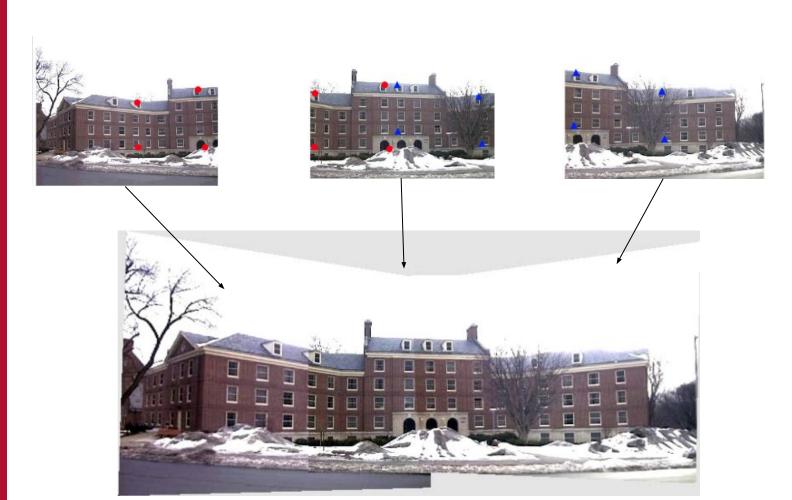


Smart advertising



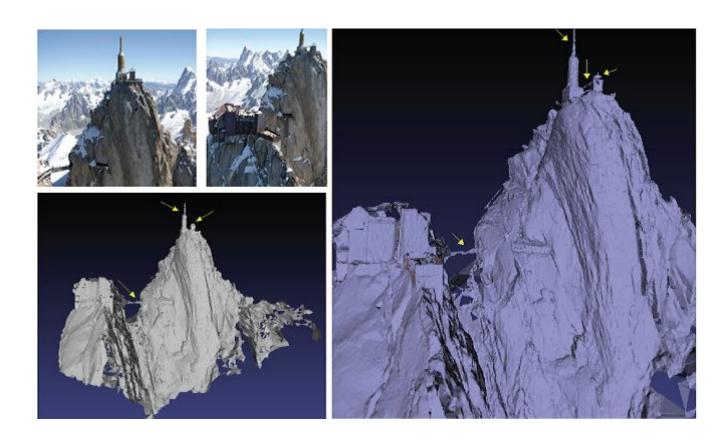


Panorama stitching



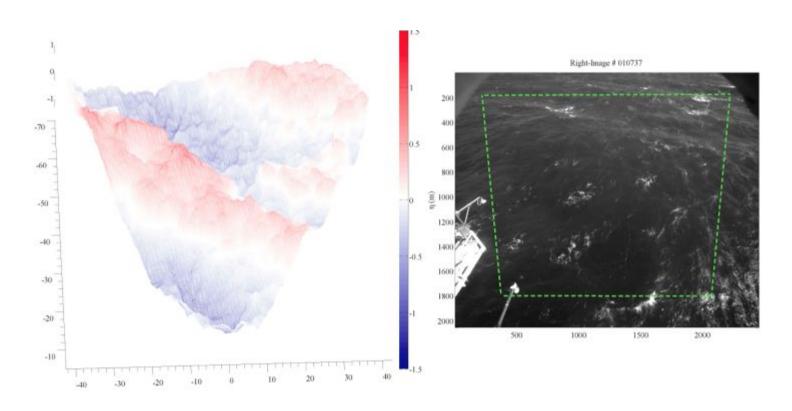


Stereo 3D reconstruction





Stereo 3D reconstruction



Sea waves 3D reconstruction pipeline http://www.dais.unive.it/wass



3D reconstruction (structure from motion)



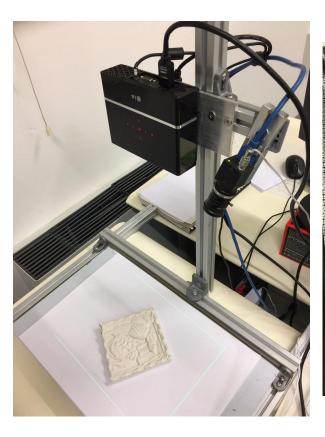


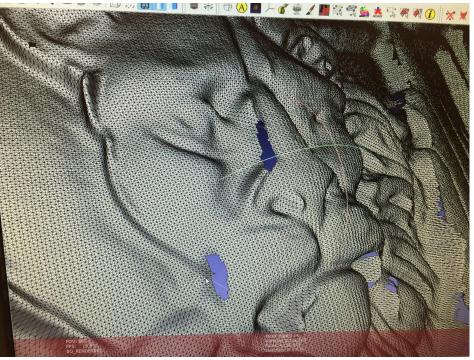
3D reconstruction (structure from motion)





Structured-light scanning





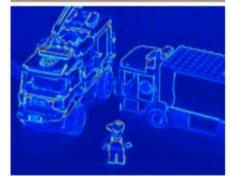


And much, much more...











Projective invariants

Factorization

Physics-based vision

Graph cuts

1970 1980 1990 2000

texture, and focus Digital image processing Blocks world, line labeling Generalized cylinders Pictorial structures Stereo correspondence Intrinsic images Optical flow Structure from motion Image pyramids Scale-space processing Physically-based modeling Regularization Markov Random Fields Kalman filters 3D range data processing Shape from shading,

Particle filtering
Energy-based segmentation
Face recognition and detection
Subspace methods
Image-based modeling
and rendering
Computational photography
Feature-based recognition
MRF inference algorithms
Category recognition
Learning

1970: What started distinguishing computer vision from image processing were early attempts to infer3D structure from images:

Lines extraction and labelling, stereo correspondences, optical flow, structure from motion



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Learning

1980: attention was focused on more sophisticated mathematical techniques for performing quantitative image and scene analysis:

Variational optimization, MRFs, Image Pyramids, 3D scanning



Physics-based vision

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Learning

1990: While a lot of the previously mentioned topics continued to be explored, a few of them became significantly more active:

Sfm, bundle adjustment, projective invariants, multi-view stereo, image segmentation (GC)



Graph cuts

Particle filtering

Energy-based segmentation Face recognition and detection Subspace methods

Image-based modeling

and rendering

Texture synthesis and inpainting

Learning

MRF inference algorithms

Category recognition

Feature-based recognition

Computational photography

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2000 - today: Continuous advances of all the previous topics:

SIFT features, texture synthesis, computational photography, learning



Graph cuts

Physics-based vision

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2012 - today: Deep learning

