

Computer Vision

Introduction

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About the teacher



Filippo Bergamasco

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

- Fixed-term researcher (RTDA)
- 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: <u>filippo.bergamasco@unive.it</u>

Or come visit me at my office Z.B08



About this course

Official UNIVE course page:

http://www.unive.it/data/course/255294

Unofficial course page:

http://www.dais.unive.it/~bergamasco/computer_vision_2018_2019.html

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



Changelog...

- Computer Vision 2016 2017 [CM0193]
 - 30 hours
- Computer Vision 2017 2018 [CM0193]
 - 30 hours
- Computer Vision 2018 2019 [CM0193]
 - 48 hours
- Geometric and 3D Computer Vision 2019-2020 [CM0526]
 - 48 hours

This year I'll introduce some of the concepts related to 3D reconstruction and stereo geometry Most of the course still focused on low/mid level vision

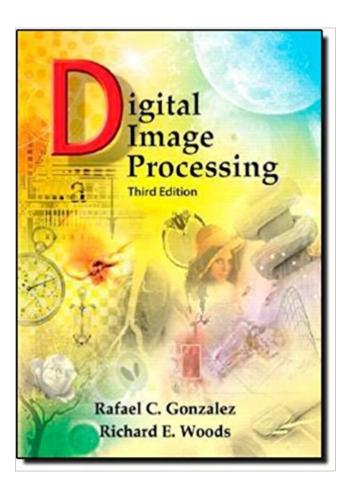


Check your study plan!

- Did you enroll before a.y. 2017?
 - Good, you can follow this course now
- Did you enroll in a.y. 2017-2018?
 - Good, you can follow this course now
- Did you enroll in a.y. 2018-2019?
 - You should have "Geometric and 3D Computer Vision" in your plan. You cannot give the exam this year.
 - Why?
 - Course codes are not compatible
 - Syllabus will be subject to changes



Referral Texts



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

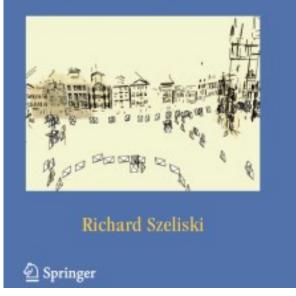


Referral Texts

TEXTS IN COMPUTER SCIENCE

Computer Vision

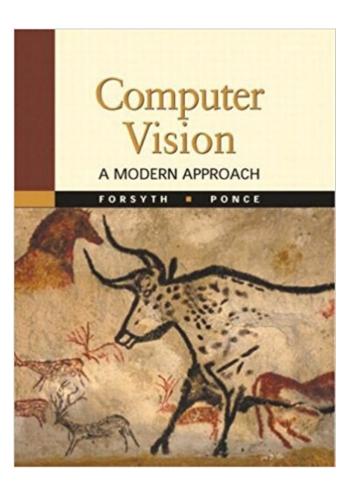
Algorithms and Applications



R. Szeliski. Computer Vision Algorithms and Applications. Springer



Referral Texts

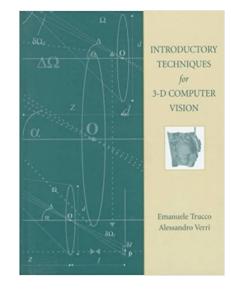


D. Forsyth and J. Ponce. **Computer Vision. A Modern Approach**. Prentice-Hall, 2002



Additional reading

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





About the exam...

Final exam consists in the development of a software project that must be performed <u>individually</u>.

Project submission information: (more will follow)

- Submit via moodle 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 (algorithms, choice made, tradeoffs, strengths and limitations)
- The goal is to evaluate your knowledge of the most common computer vision techniques and theories



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 few weeks (deadline is usually not strict)

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



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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.



Get involved

digitalmetrix



Start-up & spin-off focused on the development of structured-light based 3D scanning solutions Spin-off focused on the development of optical vision-based solutions for industrial quality inspection



Our topics...

- Camera-projector and camera-camera calibration with non-pinhole camera models
- Stochastic phase-unwrapping for robust structured-light scanning of complex materials like metal, shiny plastic, etc
- Deep learning for the identification of aesthetic defects in production pipelines
- Automated 2D/3D assessment of serigraphic printings
- Industrial metrology
- Multi-spectral and polarimetric imaging



Questions?



What is Computer Vision?

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



What is Computer Vision?

Different levels of vision:

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

Image ——— Image

- 1. Medium
 - a. Segmentation
 - b. Shape recognition

Images ----- attributes/features

1. High

a. Scene understanding

Images ------ concepts



What is Computer Vision?

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



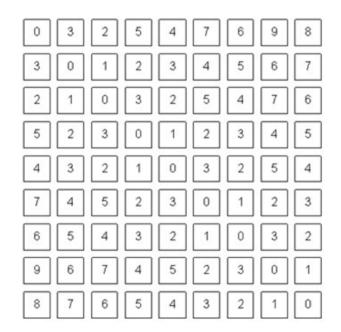
Easy to distinguish the flower in the foreground from the leaves in the background

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



What is Computer Vision?





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"

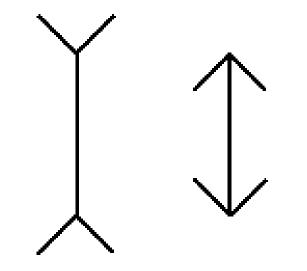
Humans and animals can interpret images so effortlessly while computer vision algorithms are still error prone:

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



Human vision

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

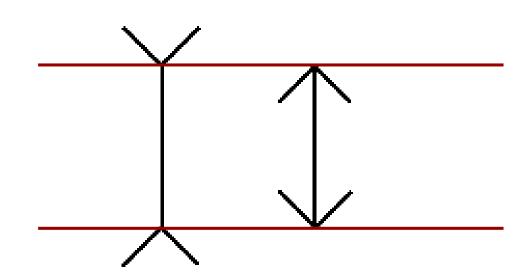


Which one is longer?

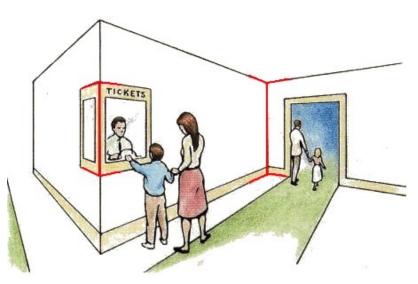
Müller-Lyer Illusion



Human vision

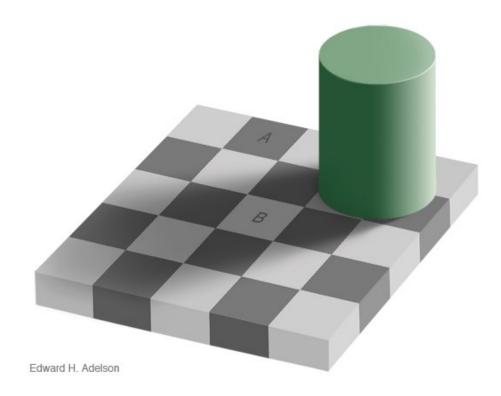


The Müller-Lyer illusion occurs because the visual system processes that judge depth and distance assume in general that the "angles in" configuration corresponds to an object which is closer, and the "angles out" configuration corresponds to an object which is far away.





Human vision



Which one is brighter?

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



Human vision

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)



Human vision

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





Computer Vision applications

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



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Computer Vision applications

Optical Character Recognition (OCR)

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ACHEARDECDUMA

56789 ABCDEHIJKLMN 979 89 59669687857 ACFEAFIECDLMA





Computer Vision applications

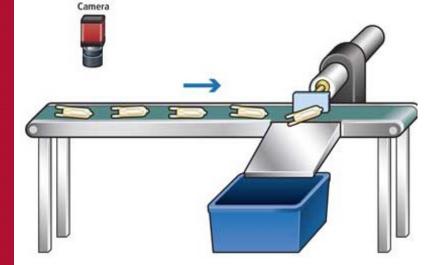
Optical Character Recognition (OCR)





Computer Vision applications

Machine inspection



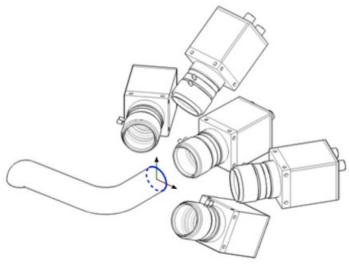




Computer Vision applications

Machine inspection



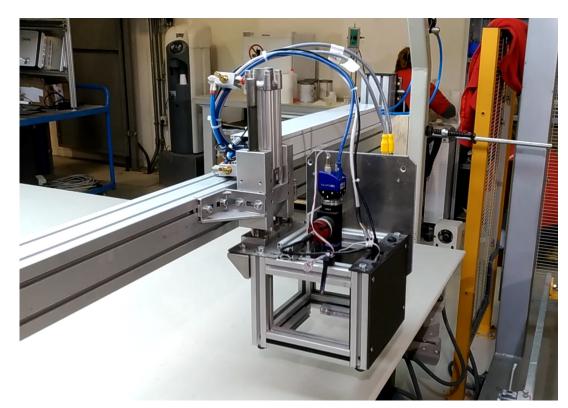


Pipe inspection device developed in our lab



Computer Vision applications

Machine inspection



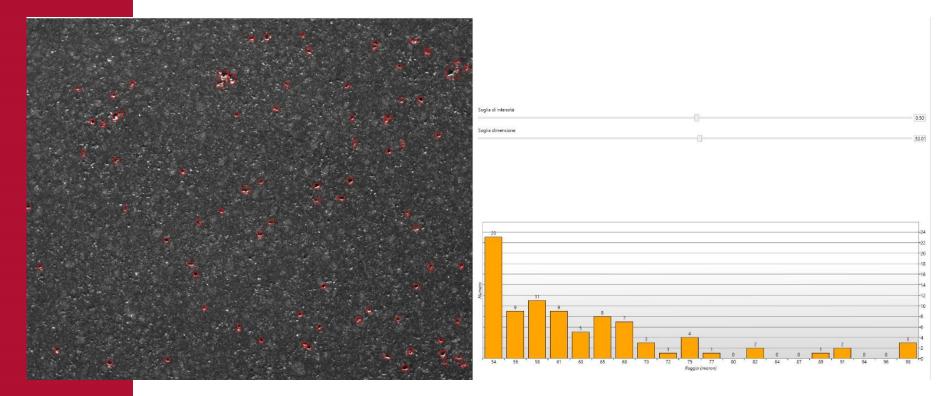
Kitchen countertops inspection device developed in our lab





Computer Vision applications

Machine inspection



Kitchen countertops inspection device developed in our lab





Computer Vision applications

Vision-based industrial metrology



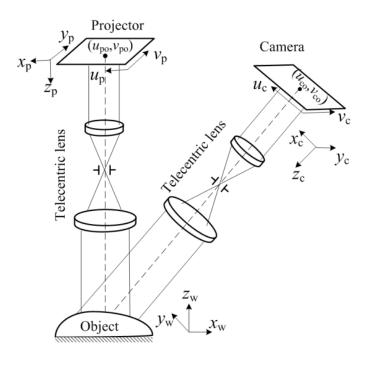
Gear' surface reconstruction device developed in our lab





Computer Vision applications

Vision-based industrial metrology





Gear' surface reconstruction device developed in our lab





Ca' Foscari

Venezia

Computer Vision applications

Surveillance and tracking







Computer Vision applications

Surveillance and tracking



Boat tracking device developed in our lab





Computer Vision applications

Surveillance and tracking



Boat tracking device developed in our lab





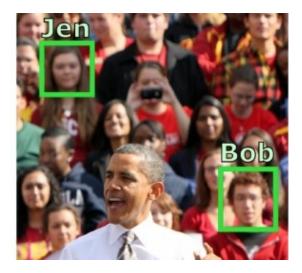
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Computer Vision applications

Face detection and recognition

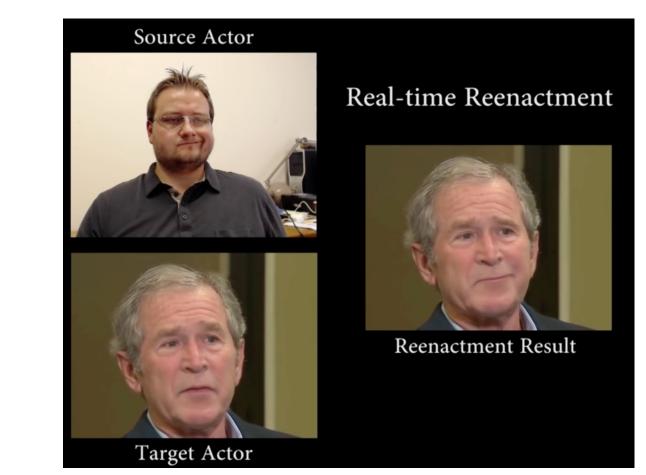






Computer Vision applications

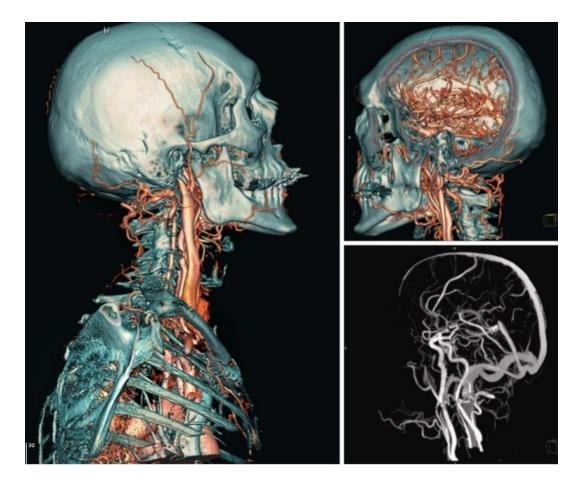
Face capture and reenactment





Computer Vision applications

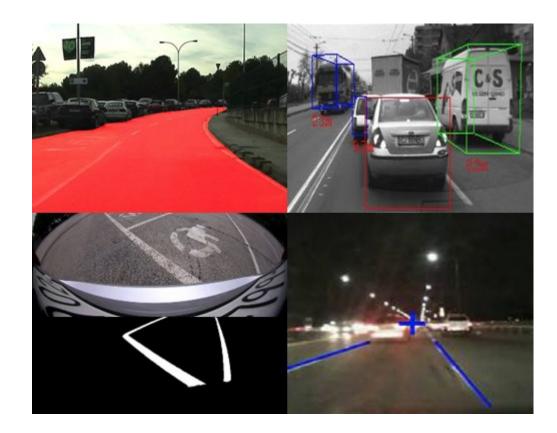
Medical image analysis





Computer Vision applications

Driving assistance





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Computer Vision applications

Driving autopilot (autonomous vehicles)





Computer Vision applications

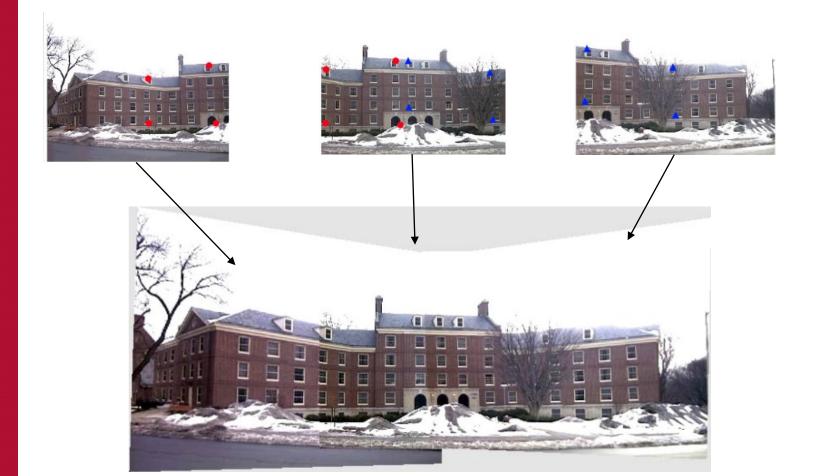
Smart advertising





Computer Vision applications

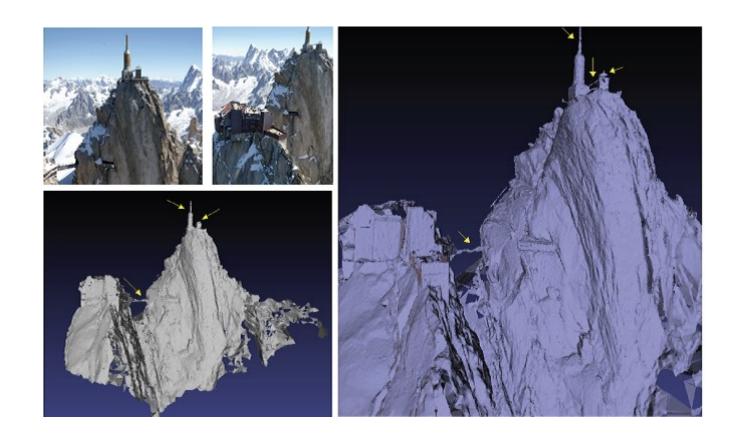
Panorama stitching





Computer Vision applications

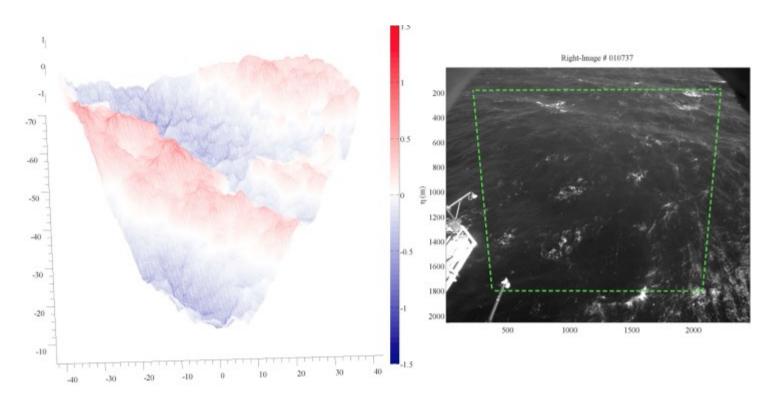
Stereo 3D reconstruction





Computer Vision applications

Stereo 3D reconstruction

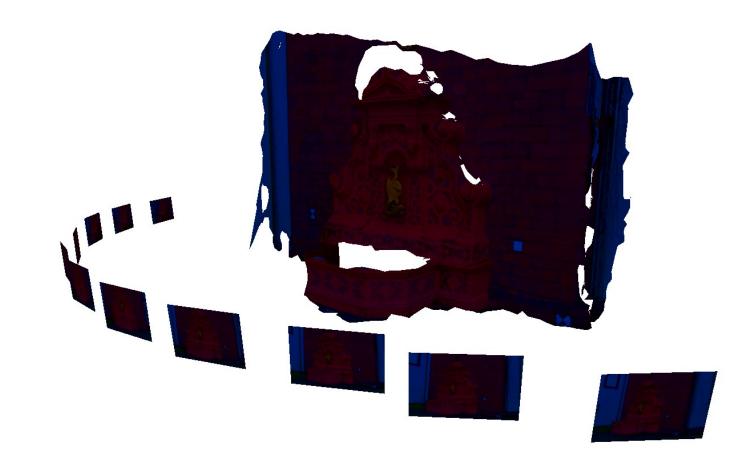


Sea waves 3D reconstruction pipeline <u>http://www.dais.unive.it/wass</u>



Computer Vision applications

3D reconstruction (structure from motion)





Computer Vision applications

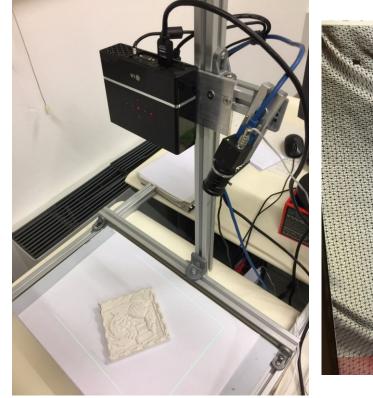
3D reconstruction (structure from motion)

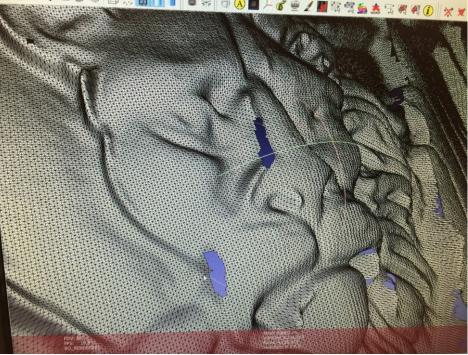




Computer Vision applications

Structured-light scanning

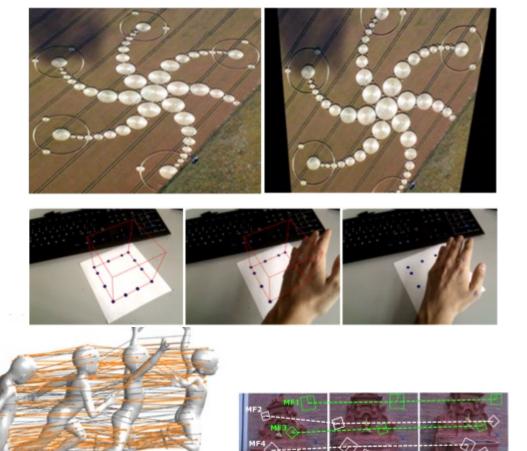






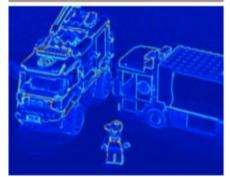
Computer Vision applications

And much, much more...











1970

A Brief History

1980

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1990

2000

Digital image processing Blocks world, line labeling Generalized cylinders Pictorial structures Stereo correspondence Intrinsic images Optical flow Structure from motion Image pyramids Structure from shading, texture, and focus Physically-based modeling Regularization Markov Random Fields Kalman filters 3D range data processing Projective invariants Projective invariants	Texture synthesis and inpainting Computational photography Feature-based recognition MRF inference algorithms Category recognition Learning
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1970: What started distinguishing computer vision from image processing were early attempts to infer**3D structure from images**:

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



1970

A Brief History

1980

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2000

Learning texture, and focus MRF inference algorithms 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 Projective invariants Factorization Physics-based vision Graph cuts Particle filtering Energy-based segmentation Face recognition and detection Subspace methods Image-based modeling and rendering Texture synthesis and inpainting Feature-based recognition Category recognition Shape from shading, Computational photography

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

Variational optimization, MRFs, Image Pyramids, 3D scanning



1970

Generalized cylinders Pictorial structures Stereo correspondence Intrinsic images Optical flow Structure from motion

Image pyramids

significantly more active:

Scale-space processing

texture, and focus

Shape from shading,

Physically-based modeling

view stereo, image segmentation (GC)

Regularization

Markov Random Fields

Kalman filters

3D range data processing

1990: While a lot of the previously mentioned topics

Sfm, bundle adjustment, projective invariants, multi-

continued to be explored, a few of them became

Digital image processing Blocks world, line labeling

A Brief History

1980

1990

Subspace methods Image-based modeling and rendering Texture synthesis and inpainting Computational photography

2000

MRF inference algorithms

Category recognition

Feature-based recognition

Learning

Particle filtering Energy-based segmentation Face recognition and detection

Factorization

Projective invariants Physics-based vision

Graph cuts



1970

Generalized cylinders Pictorial structures Stereo correspondence Intrinsic images Optical flow Structure from motion

Image pyramids

Scale-space processing

Digital image processing Blocks world, line labeling

A Brief History

1980

texture, and focus

Shape from shading,

Physically-based modeling

1990

Learning MRF inference algorithms Regularization Markov Random Fields Kalman filters 3D range data processing Projective invariants Factorization Physics-based vision Graph cuts Particle filtering Energy-based segmentation Face recognition and detection Subspace methods Image-based modeling and rendering Texture synthesis and inpainting Feature-based recognition Category recognition Computational photography

2000

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

SIFT features, texture synthesis, computational photography, learning



1970

A Brief History

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Learning Generalized cylinders Image pyramids exture, and focus Physically-based modeling Projective invariants Graph cuts MRF inference algorithms Digital image processing Blocks world, line labeling Pictorial structures Stereo correspondence Intrinsic images Optical flow Structure from motion Scale-space processing Regularization Markov Random Fields Kalman filters 3D range data processing Factorization Physics-based vision Particle filtering Energy-based segmentation Face recognition and detection Subspace methods Image-based modeling and rendering Texture synthesis and inpainting Feature-based recognition Category recognition Shape from shading, Computational photography

2012 - Today: Deep learning

