



3D Computer Vision Introduction

Guido Gerig

CS 6320, Spring 2013

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Acknowledgements: some slides from Marc Pollefeys and Prof.
Trevor Darrell, trevor@eecs.berkeley.edu



Administrivia

- Classes: M & W, 1.25-2:45
Room WEB L126
- Instructor: Guido Gerig
gerig@sci.utah.edu
- TA: tbd
- Prerequisites: CS 6640 ImProc (or equiv.)
- Textbook:

“Computer Vision: A Modern Approach” by Forsyth & Ponce

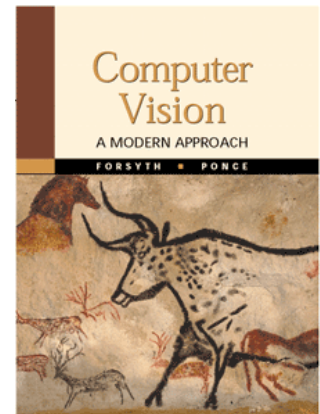
- Organization:

Admin/Grading/Uploads:

UofU [canvas](#)

Slides, documents and assignments:

[Course Website](#)

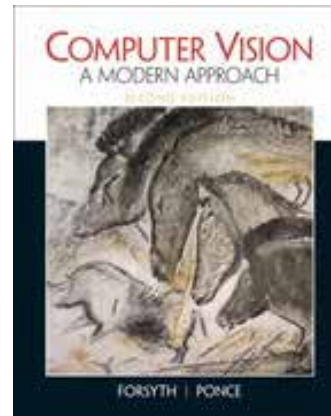
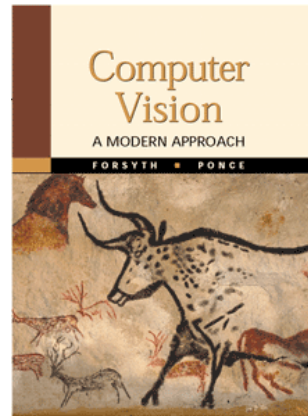


Administrivia

- Textbook:

“Computer Vision: A Modern Approach” by Forsyth & Ponce

Version 1 Version 2e 2012



- The Version 1 is sufficient for this course, but you can also buy the new updated version.
- Electronic version:

http://www.coursesmart.com/IR/5316068/9780132571074?__hdv=6.8



Web-Site

- Linked to canvas CS 6320-001 home page
- Linked to UofU **Spring 2013 Class Schedule**
- Linked to my home page:

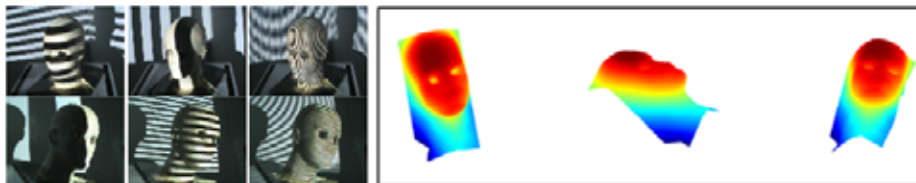
http://www.sci.utah.edu/~gerig/CS6320-S2013/CS6320_3D_Computer_Vision.html



CS6320 3D Computer Vision, Spring 2013

Computing properties of our 3-D world from passive and active sensors

[Syllabus](#), **Guido Gerig ([home](#))**



Goal and Objectives:

- To introduce the fundamental problems of 3D computer vision.
- To introduce the main concepts and techniques used to solve those.
- To enable participants to implement solutions for reasonably complex problems.
- To enable participants to understand basic methodology that is discussed in the computer vision literature.



TA / SW Tools / Office Hours

- TA: tbd
- HW/SW: Matlab+ ev. Imaging Toolbox
CADE lab WEB 130
<http://www.cade.utah.edu/>
- Office Hours TA office Hours: tbd
send email for other appointments
- Office Hours instructor: Mo/Wed 3-5pm after class



Prerequisites

- General Prerequisites:
 - Data structures
 - A good working knowledge of MATLAB programming (or willingness and time to pick it up quickly!)
 - Linear algebra
 - Vector calculus
- Assignments include theoretical paper questions and programming tasks (ideally Matlab or C++).
- Image Processing CS 6640 (or equivalent).
- Students who do not have background in signal processing / image processing: Eventually possible to follow class, but requires significant special effort to learn some basic procedures necessary to solve practical computer problems.



Grading - Weights

- Assignments (4-5 theory/prog.): 60%
- Final project (incl. design, proposal, demo, presentation, report): 30%
- Class participation (active participation in summaries and discussions): 10%
- Final project replaces final exam
- Successful final project required for passing grade



Assignments & Projects

- Assignments: Theoretical and Practical Part: [Example](#)
- Strict deadlines: 10% deduction per day late, no more accepted after 3 days
- Assignments solutions include:
 - Solutions to [theoretical parts](#) (can be handwritten and scanned)
 - Detailed report on [practical solution](#) (pdf document)
 - Code used to solve practical part
- Important:
 - Be creative with own images and experiments with your code. Try different scenarios and discuss pro's and con's.
 - Report needs to include description of what you did, description of results, and critical assessment of results, your code, limitations, etc.



Other Resources

- Cvonline:
<http://homepages.inf.ed.ac.uk/rbf/CVonline/>
- A first point of contact for explanations of different image related concepts and techniques. CVonline currently has about 2000 topics, 1600 of which have content.
- See list of other relevant books in syllabus.



Some Basics

- Instructor and TA do not use email as primary communication tool.
- It will be your responsibility to regularly read the Announcements on canvas.
- We don't need a laptop for the class, please keep them closed !!!!!
- Please interact, ask questions, clarifications, input to instructor and TA.
- Cell phones, you surely know.



Syllabus

- See separate syllabus (linked to web-site and included on canvas).
- [Document](#)



Goal and objectives

- To introduce the fundamental problems of computer vision.
- To introduce the main concepts and techniques used to solve those.
- To enable participants to implement solutions for reasonably complex problems.
- To enable the student to make sense of the literature of computer vision.



CV: What is the problem?

Image Formation: From World to Image

- Camera model (optics & geometry): From points in 3D scene to points on 2D image.
- Photometry: From lights and surfaces in scene to intensity (brightness) and color in image.

Vision: From Image to (Knowledge of the) World

- Reconstruct scene (**world model**) from images.
- Extract sufficient **information** for detection/control **task**.



CV: A Hard Problem

- Under-constrained inverse problem – 3D world from 2D image.
- Images are **noisy** – shadows, reflections, focus, (ego-)motion blur – and noise is hard to model.
- Appearances – shape, size, color – of objects change with pose and lighting conditions.
- Image understanding requires cognitive ability (“AI-complete”).
- Robotics & Control: massive data rate, real-time requirements.

What is Computer Vision?

- Automatic understanding of images and video
 - Computing properties of the 3D world from visual data (*measurement*)
 - Algorithms and representations to allow a machine to recognize objects, people, scenes, and activities. (*perception and interpretation*)



Vision and graphics



Images



Vision



Model



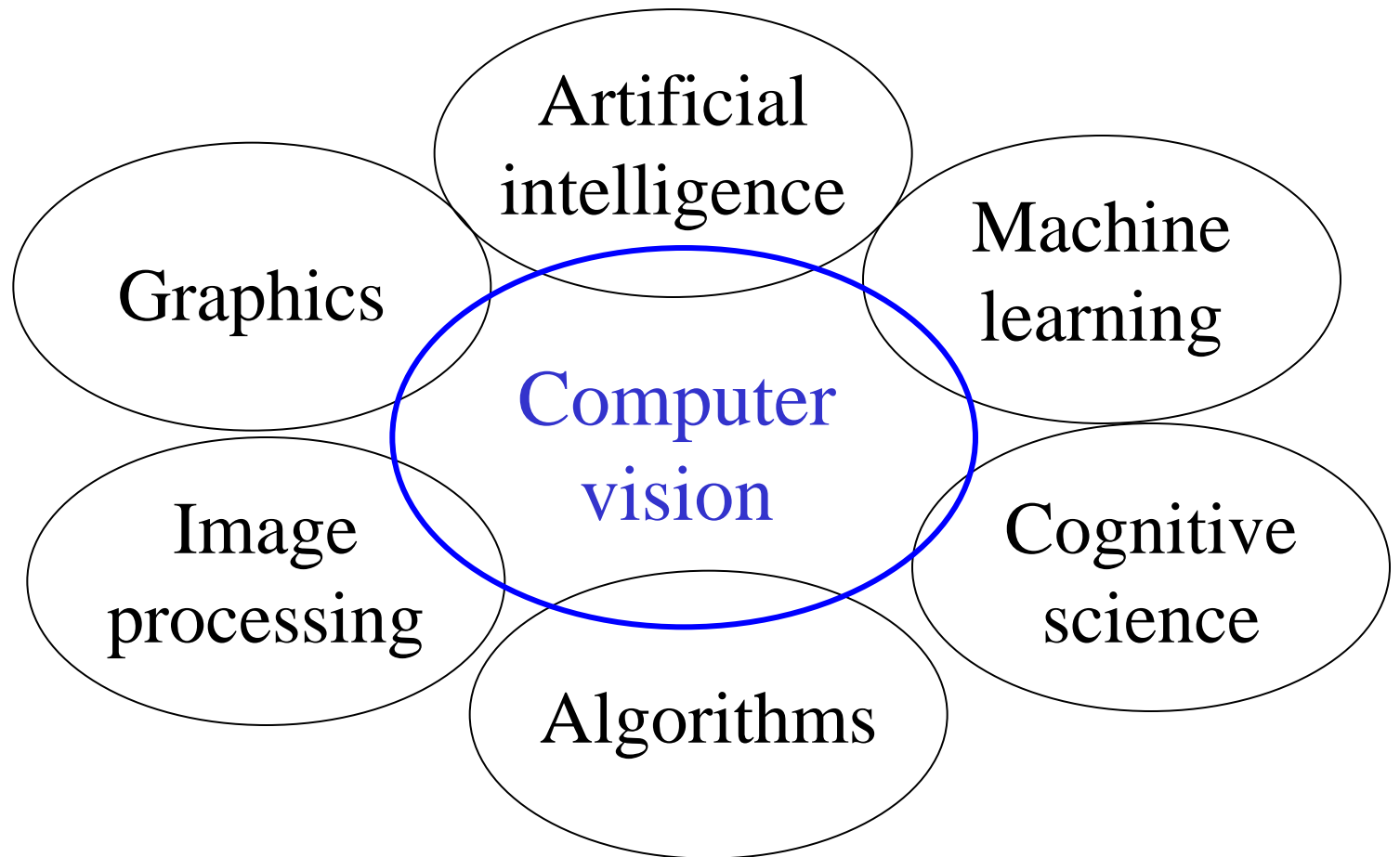
Graphics



Inverse problems: analysis and synthesis.



Related disciplines



Object recognition (in mobile phones)



- This is becoming real:
 - **Lincoln** Microsoft Research
 - Point & Find, Nokia
 - SnapTell.com (now amazon)

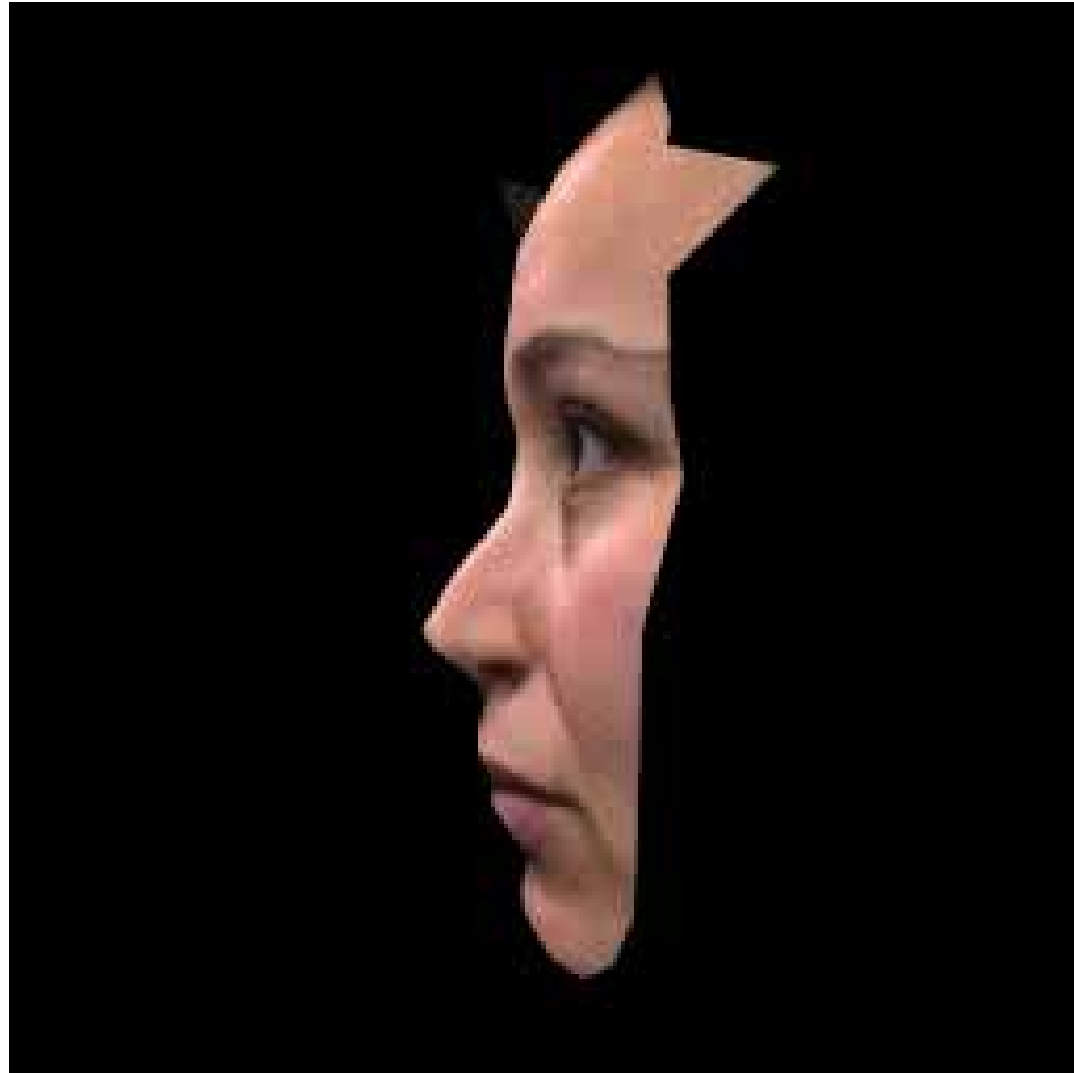
Smart cars

The screenshot displays the Mobileye website interface. At the top, there are navigation tabs for 'manufacturer products' and 'consumer products'. The main headline reads 'Our Vision. Your Safety.' Below this, a top-down view of a car is shown with four camera fields of view: 'rear looking camera', 'forward looking camera', and 'side looking camera'. The bottom section features three product highlights: 'EyeQ Vision on a Chip' with an image of the chip, 'Vision Applications' showing a pedestrian detection box, and 'AWS Advance Warning System' with a car icon and a distance reading of '0.6'. On the right side, there is a 'News' section with two articles about Volvo's collision warning system and an 'Events' section with two entries about Mobileye at Equip Auto in Paris and at SEMA in Las Vegas.



- [Mobileye](#)
 - Vision systems currently in high-end BMW, GM, Volvo models
 - By 2010: 70% of car manufacturers.
 - [Video demo](#)
 - [YouTube](#), [TestMovie](#)

Modeling 3D Structure from Pictures or 3D Sensors



Modeling ctd.





Main topics

- Shape (and motion) recovery
“What is the 3D shape of what I see?”
- Segmentation
“What belongs together?”
- Tracking
“Where does something go?”
- Recognition
“What is it that I see?”



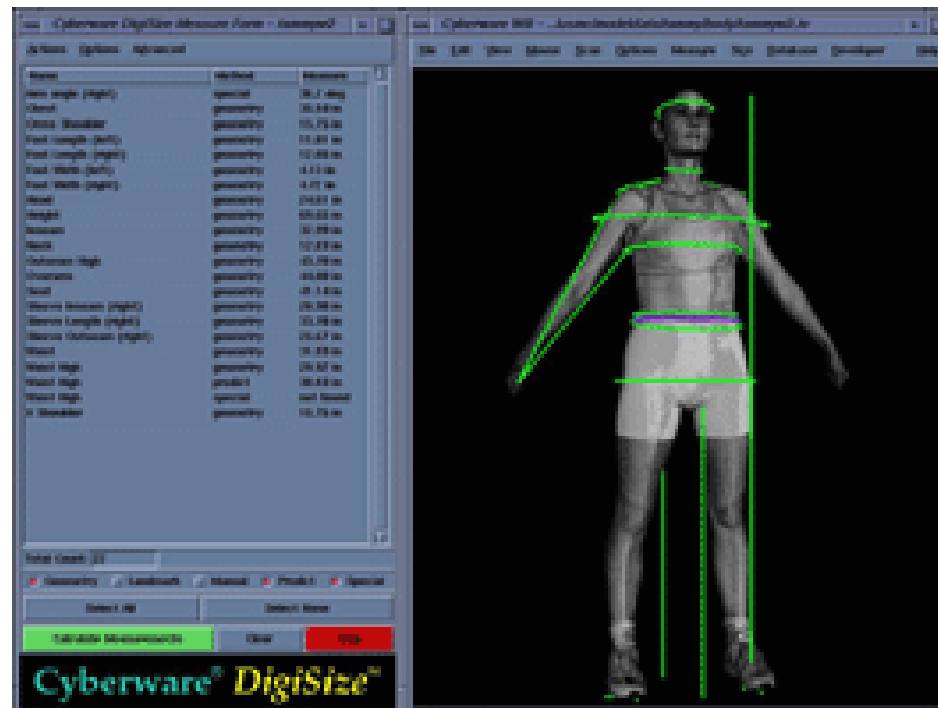
Why study Computer Vision?

- Images and movies are everywhere
- Fast-growing collection of useful applications
 - building representations of the 3D world from pictures
 - automated surveillance (who's doing what)
 - movie post-processing
 - CAM (computer-aided manufacturing)
 - Robot navigation
 - face finding
- Various deep and attractive scientific mysteries
 - how does object recognition work?
- Greater understanding of human vision

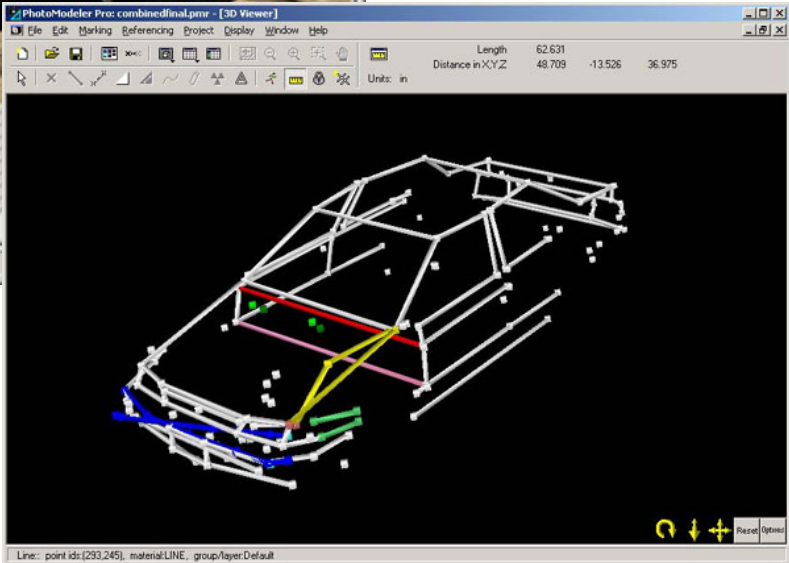
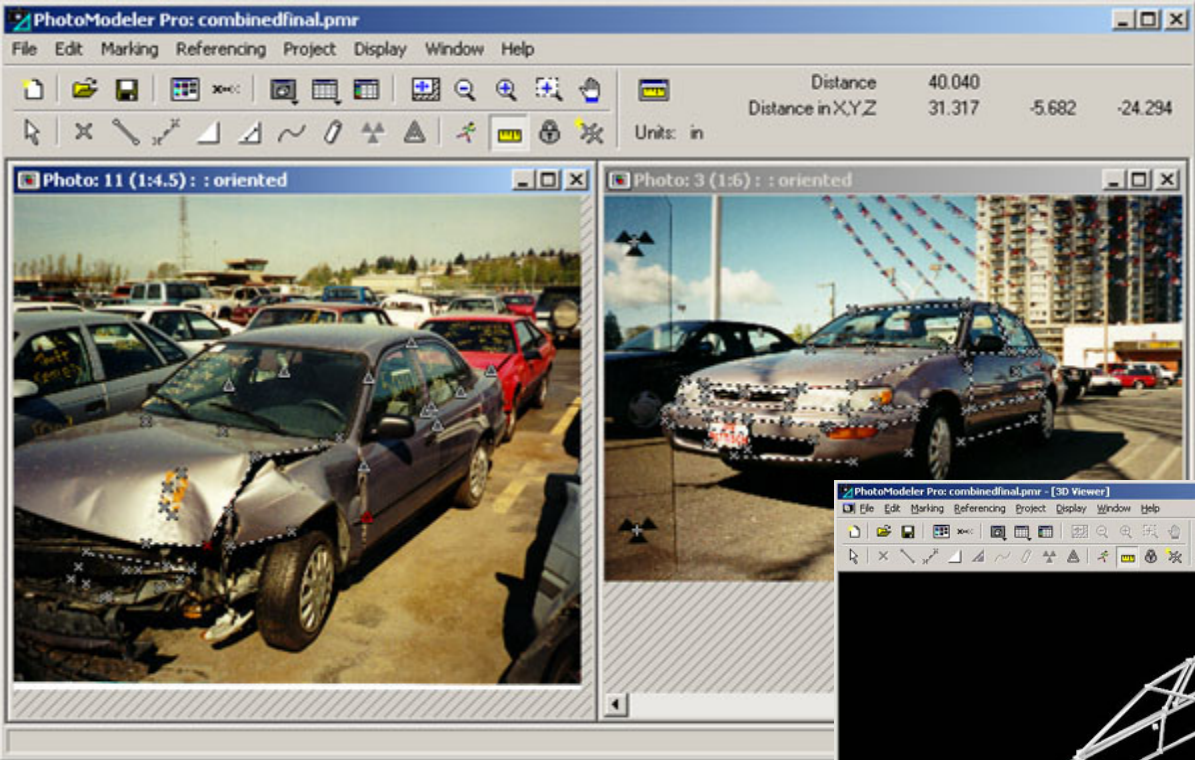


Clothing

- Scan a person, custom-fit clothing



Forensics



PhotoModeler



3D urban modeling



drive by modeling in Baltimore

Earth viewers (3D modeling)

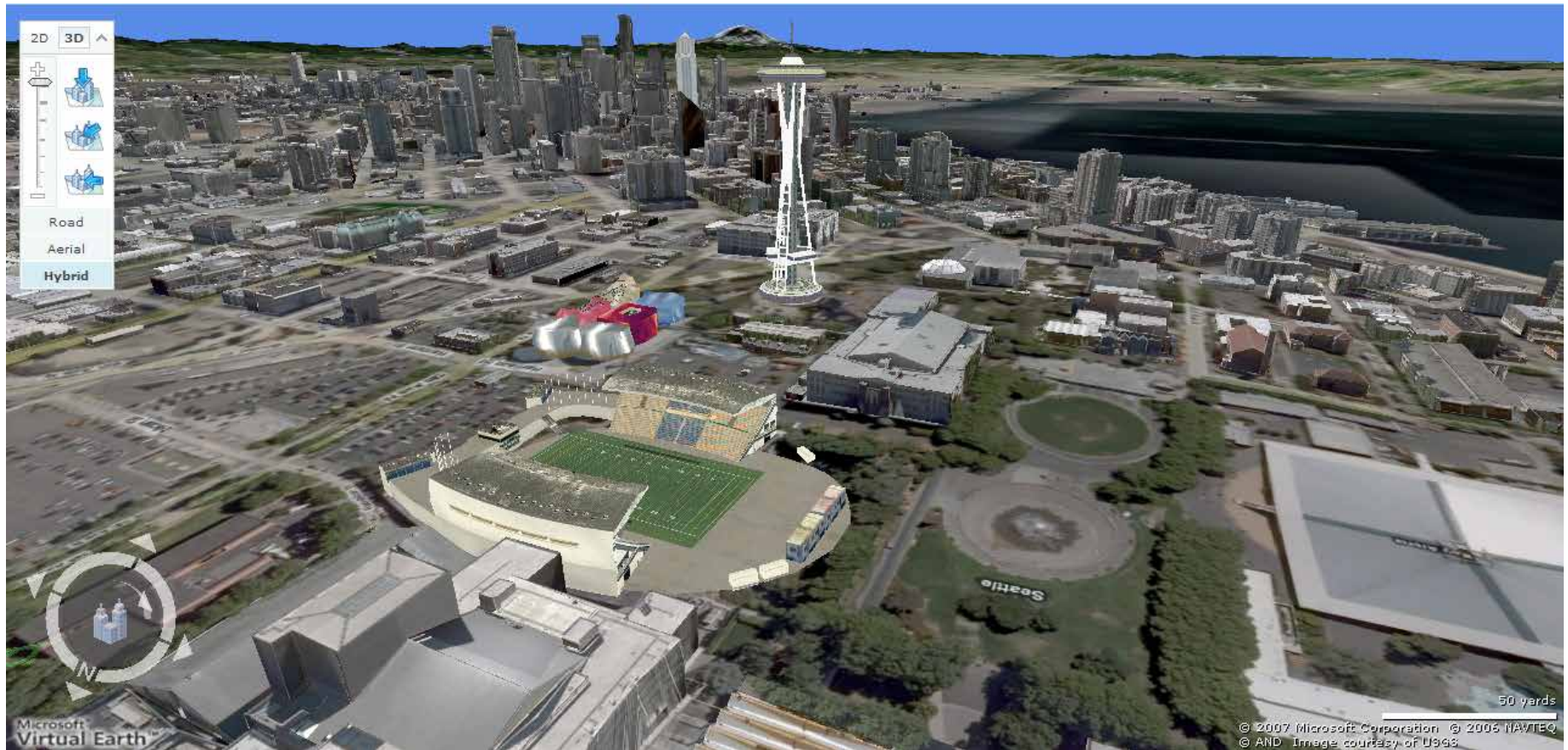
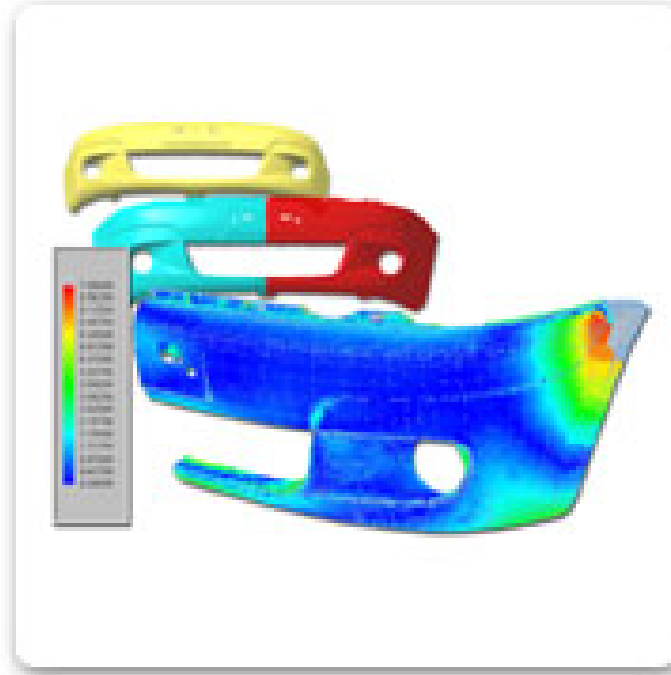


Image from Microsoft's [Virtual Earth](#)
(see also: [Google Earth](#))



Industrial inspection

- Verify specifications
- Compare measured model with CAD



Scanning industrial sites



as-built 3D model of off-shore oil platform



Vision in space



[NASA'S Mars Exploration Rover Spirit](#) captured this westward view from atop a low plateau where Spirit spent the closing months of 2007.

Vision systems (JPL) used for several tasks

- Panorama stitching
- 3D terrain modeling
- Obstacle detection, position tracking
- For more, read “[Computer Vision on Mars](#)” by Matthies et al.

Robotics

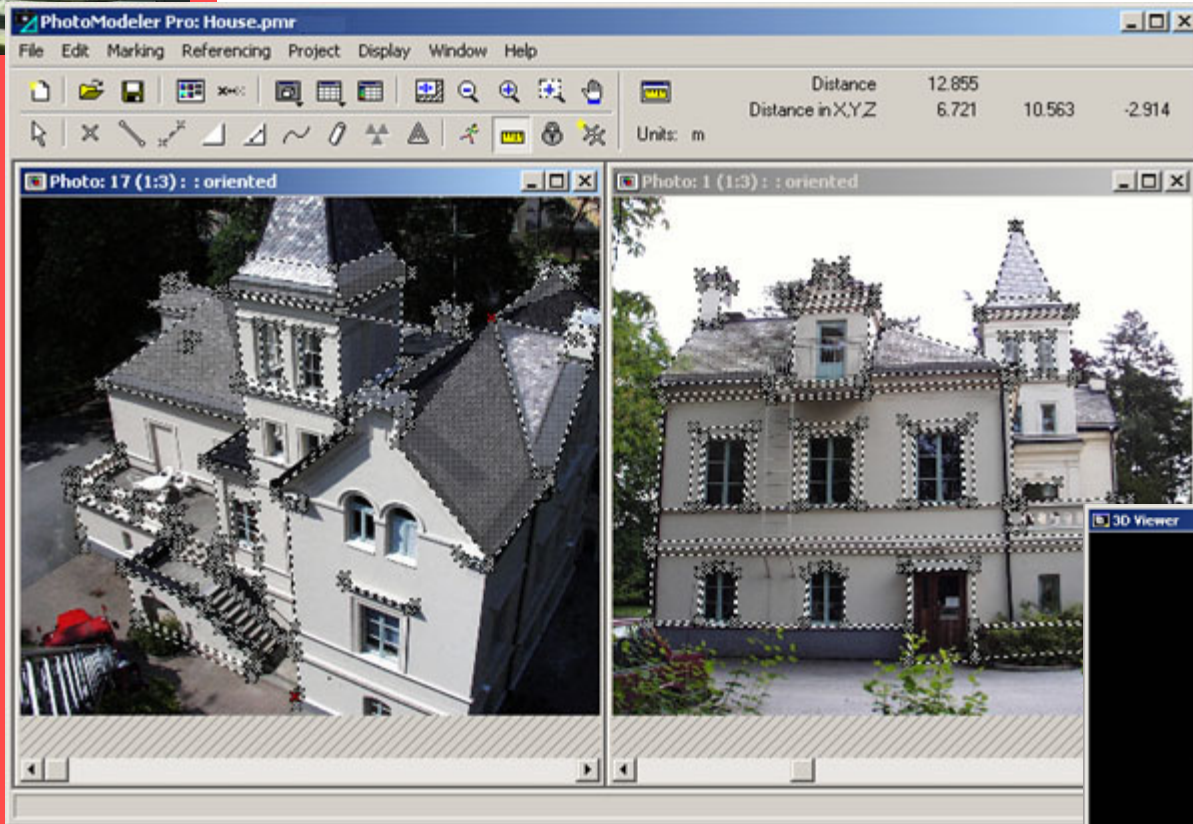


NASA's Mars Spirit Rover
http://en.wikipedia.org/wiki/Spirit_rover



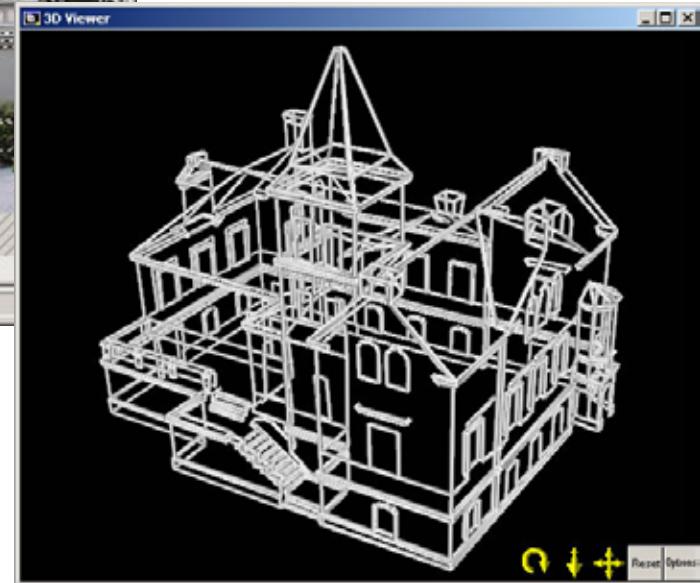
<http://www.robocup.org/>

Architecture

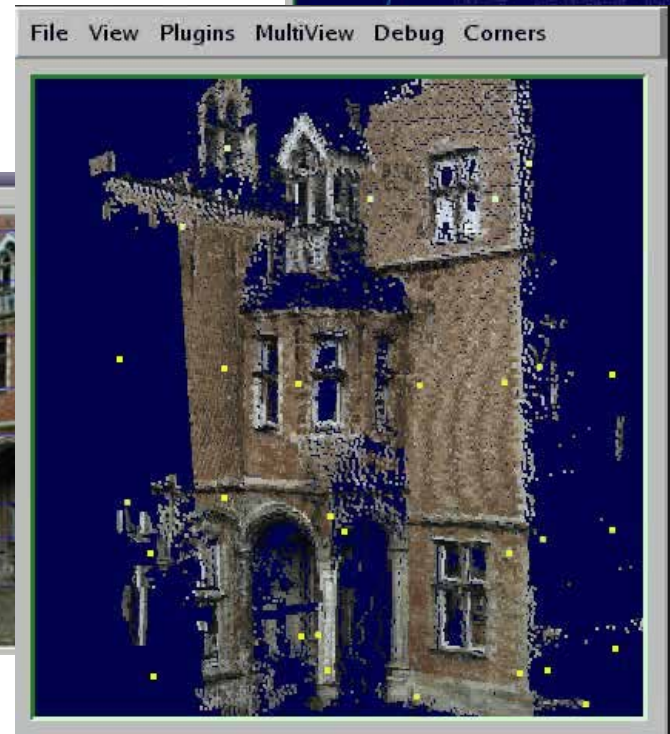
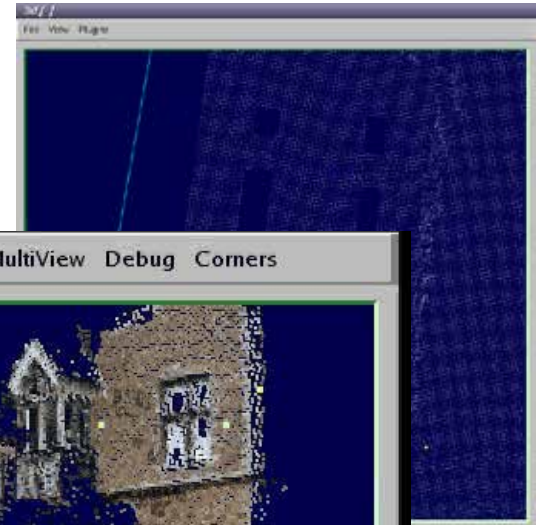
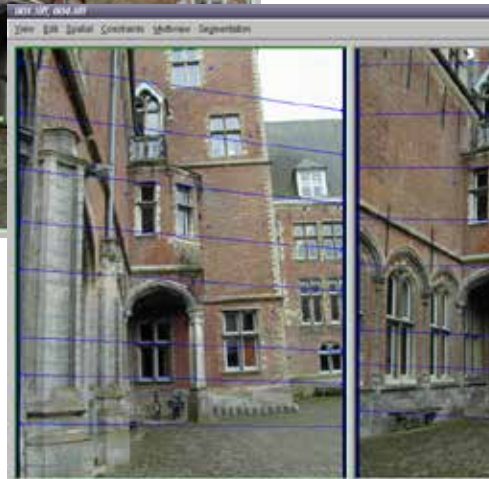
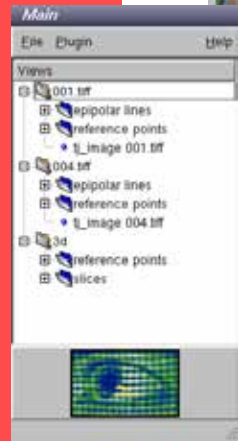


Survey
Stability analysis
Plan renovations

PhotoModeler



Architecture

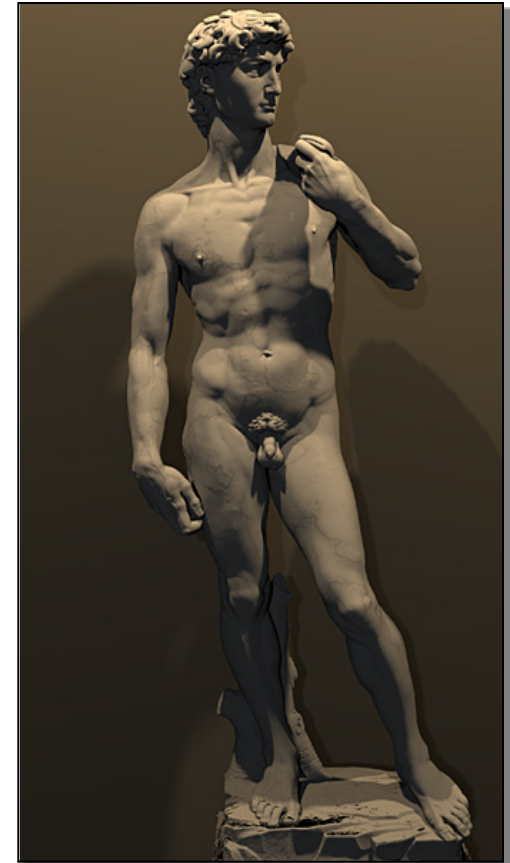


Survey
Stability analysis
Plan renovations

Cultural heritage



Stanford's Digital Michelangelo

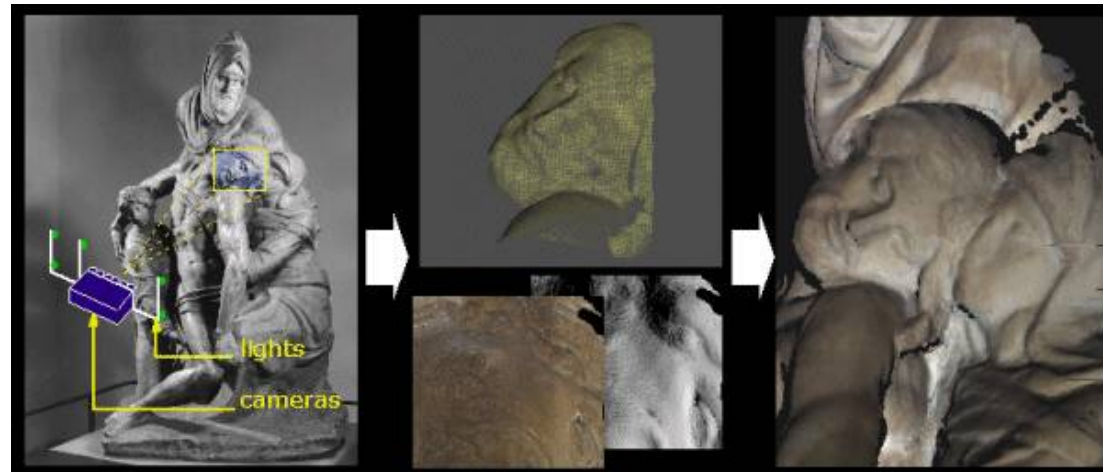


Digital archive
Art historic studies



IBM's pieta project

Photometric stereo + structured light



more info:

http://researchweb.watson.ibm.com/pieta/pieta_details.htm



Archaeology



accuracy ~1/500 from DV video
(i.e. 140kb jpegs 576x720)



Visual Cues: Stereo and Motion

Disparity map from Stereo

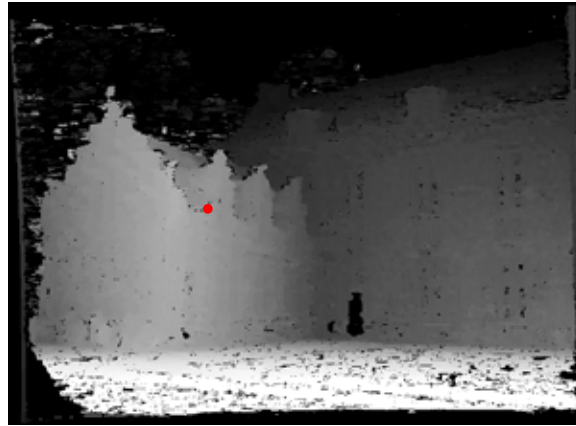


image $I(x,y)$



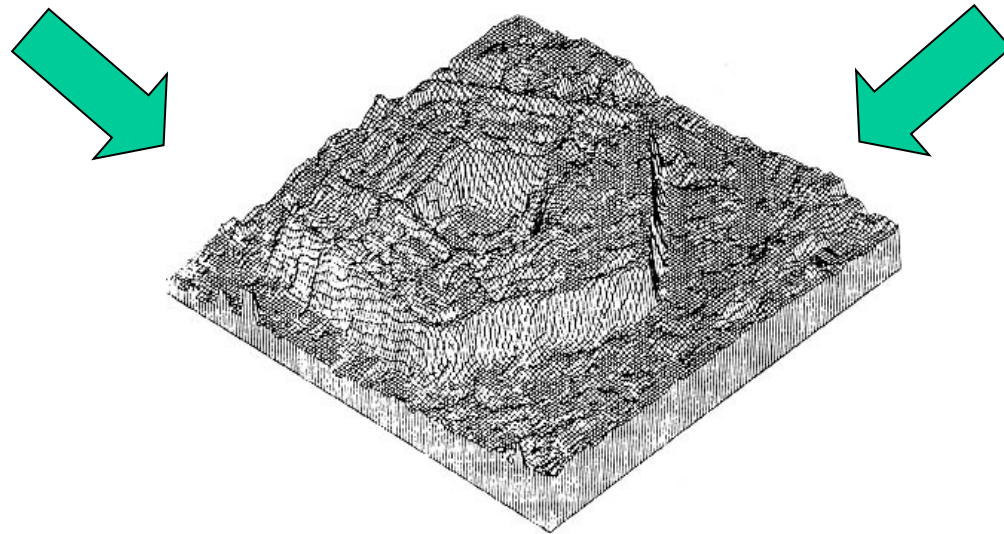
Disparity map $D(x,y)$

image $I'(x',y')$



$$(x',y')=(x+D(x,y),y)$$

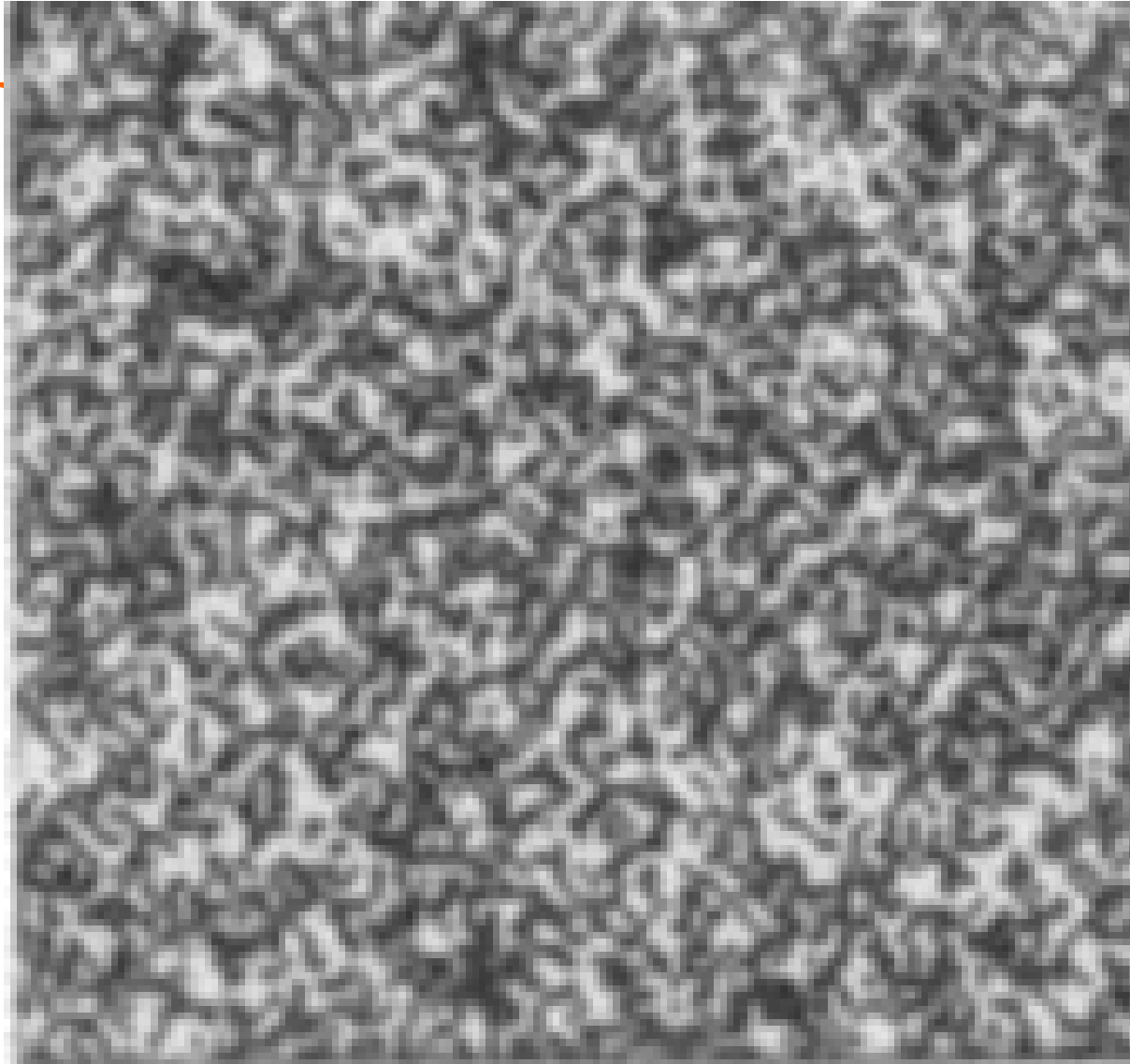
Dynamic Programming (Ohta and Kanade, 1985)



Optical flow



Wr



Optical flow





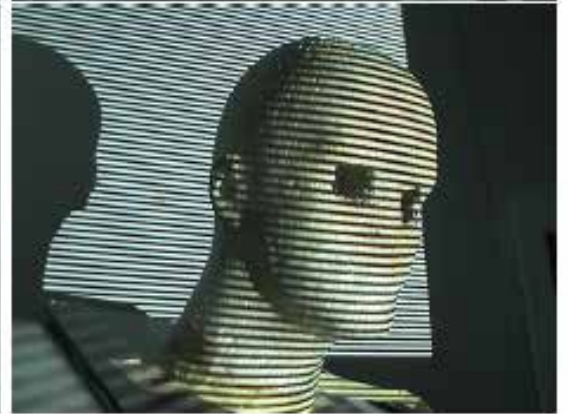
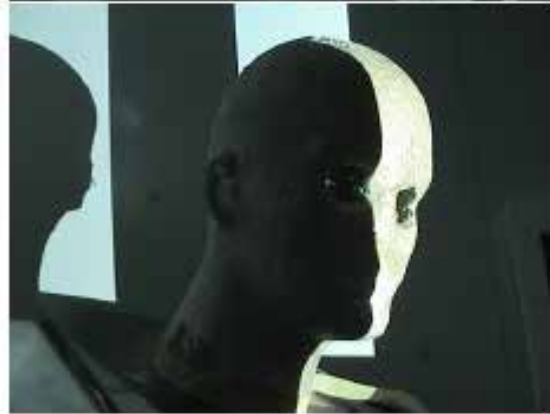
Results



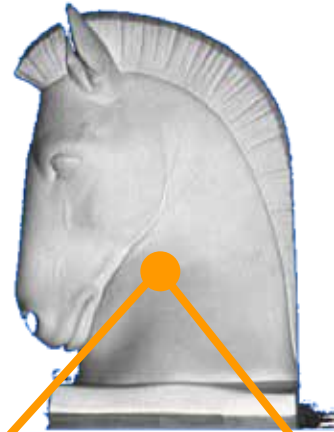


Active Vision: Structured Light

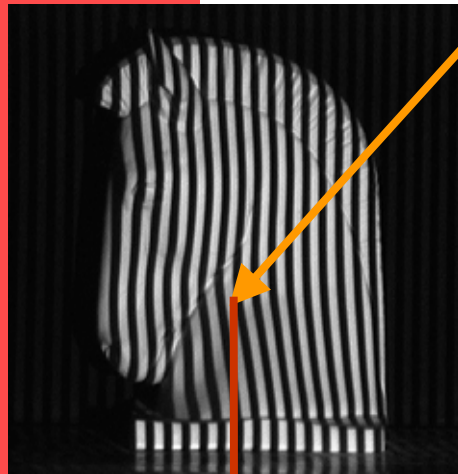
Active Vision: Structured Light



Binary Coding



Example: 7
binary patterns
proposed by
Posdamer &
Altschuler



...

Pattern 3

Pattern 2

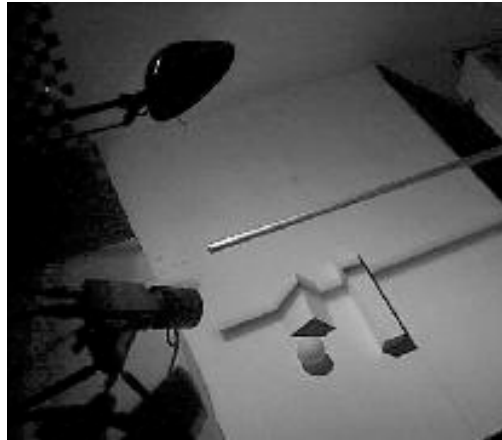
Pattern 1

Projected
over time



**Codeword of this píxel: 1010010 à
identifies the corresponding pattern stripe**

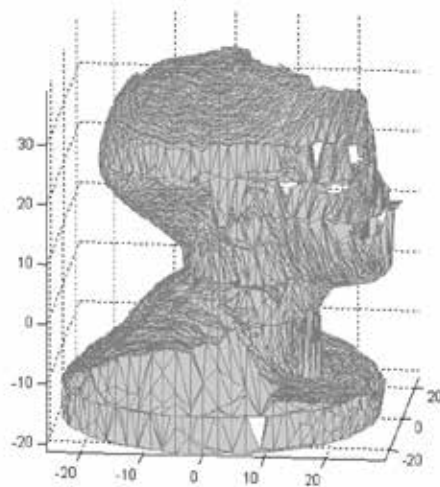
"Cheap and smart" Solution



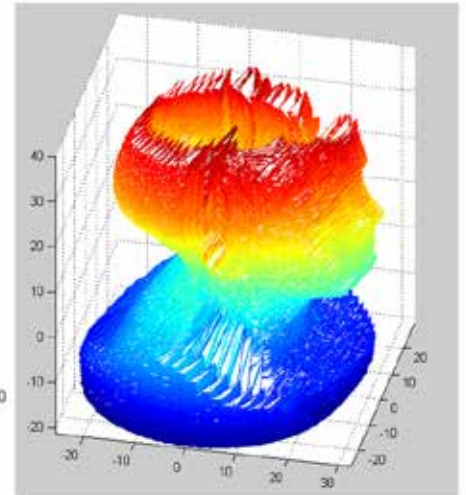
Example:
Bouguet and
Perona,
ICCV'98

Structured Light Using a Rotating Table

James Clark, 3D CV F2009



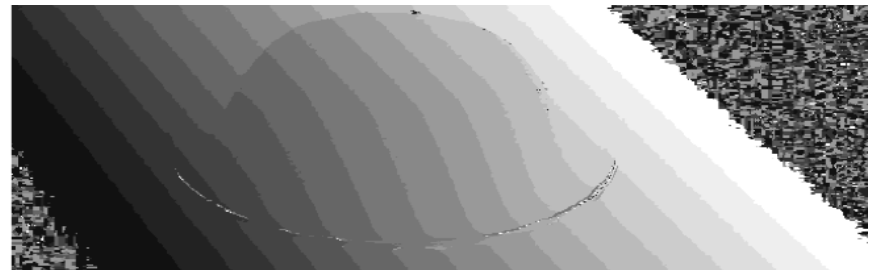
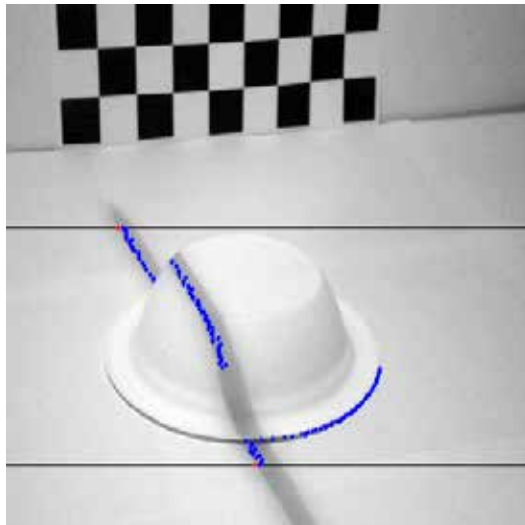
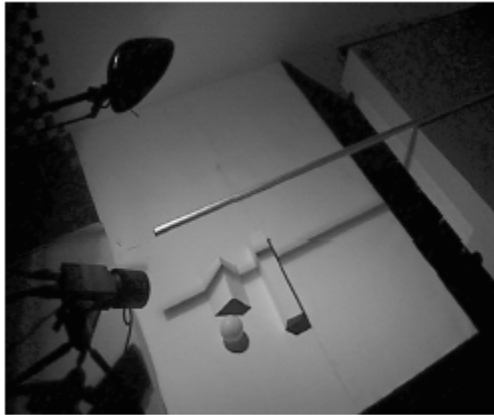
Height Strip Mesh



Localized Mesh

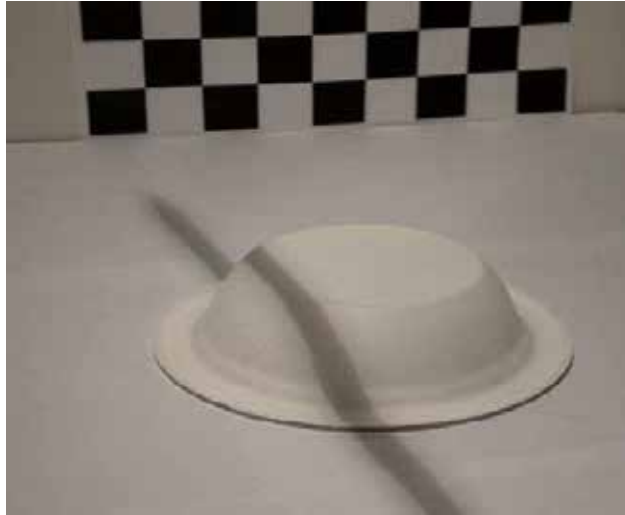
Structured Light

Anuja Sharma, Abishek Kumar

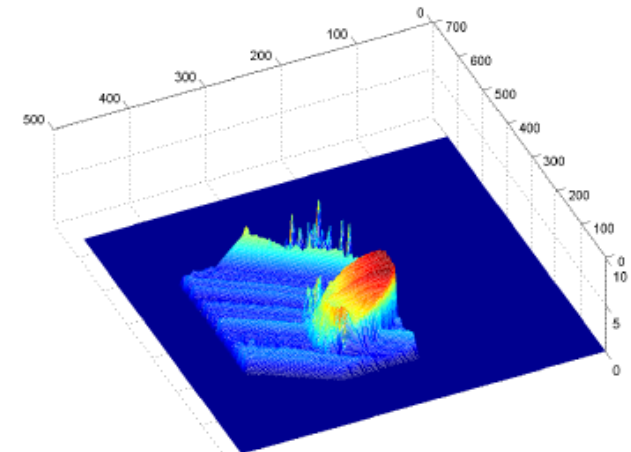


Structured Light

Anuja Sharma, Abishek Kumar



3D plot 1



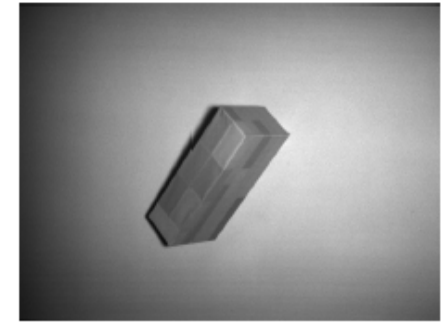
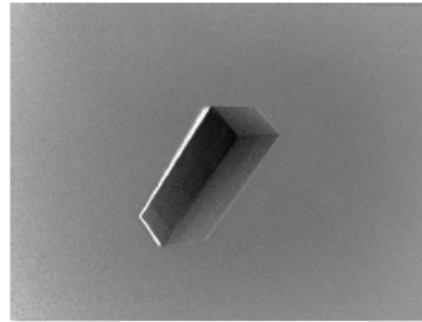


Range Sensor Data Processing to get 3D Shapes





Input Data: Depth Maps



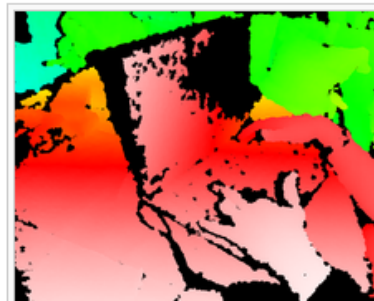
Range Image (left) and gray level image (right)



This infrared image shows the laser grid Kinect uses to calculate depth



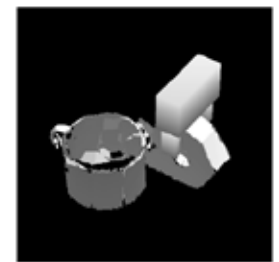
A slide from Microsoft's E3 Conference showing a diagram of the technologies in Kinect



The depth map is visualized here using color gradients from white (near) to blue



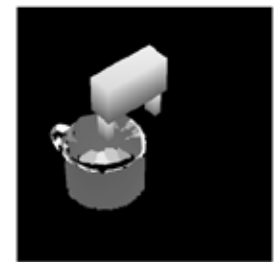
(e)



(f)



(g)



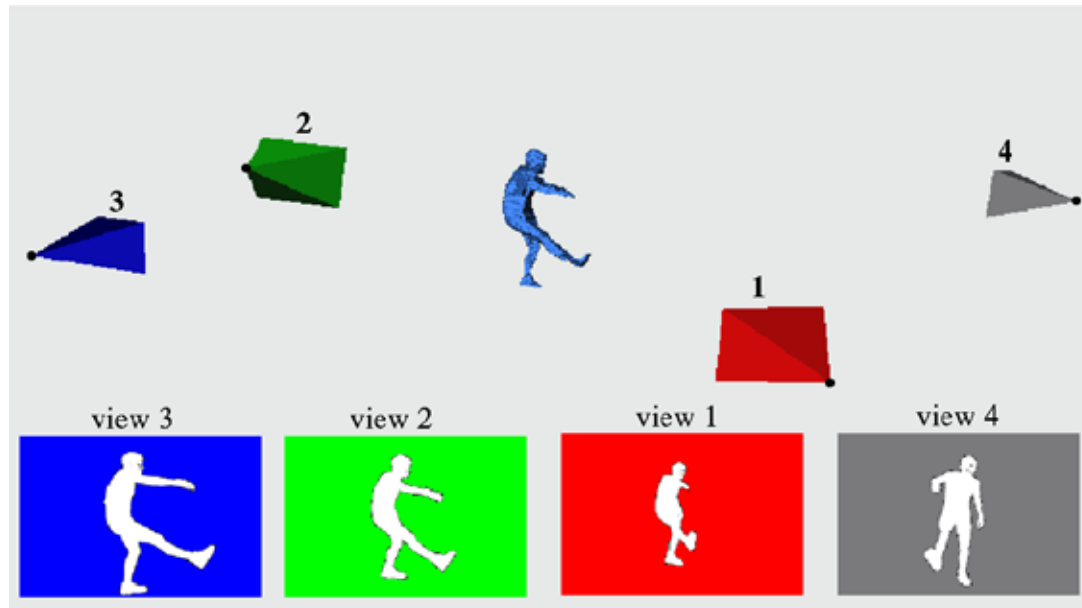
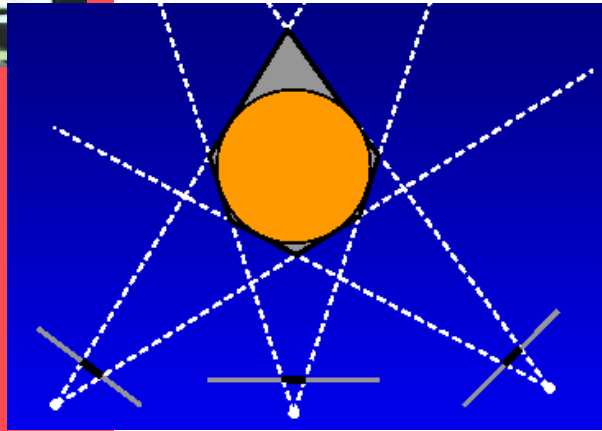
(h)

Figure 9: Continuation of the example scene consisting of four objects. (e) and (f) grasping the Scotch tape roller, and (g) and (h) grasping the coffee cup.

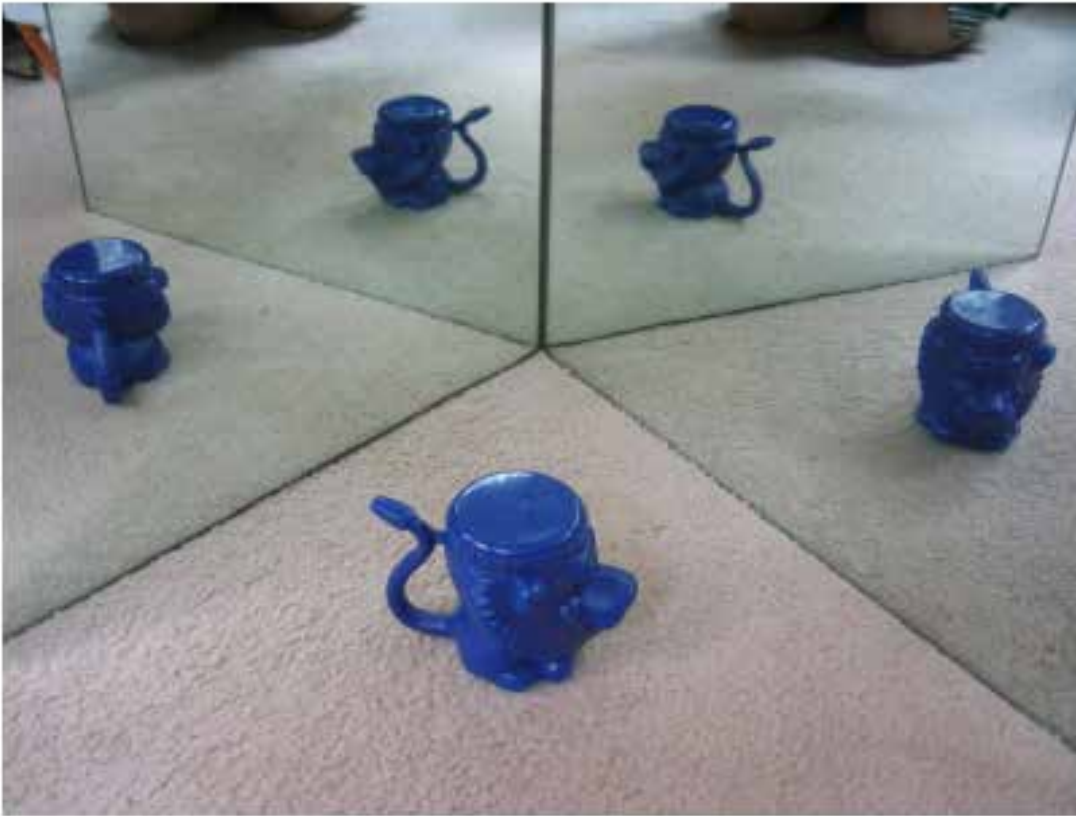


3D Shape Cues: Shape from Silhouettes

3D Shape from Silhouettes



3D shape from silhouettes: Two Mirrors and uncalibrated camera

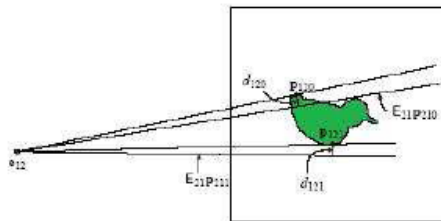


Forbes et al.,
ICCV2005

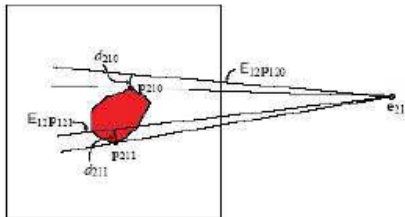
Christine Xu,
Computer Vision
Student Project



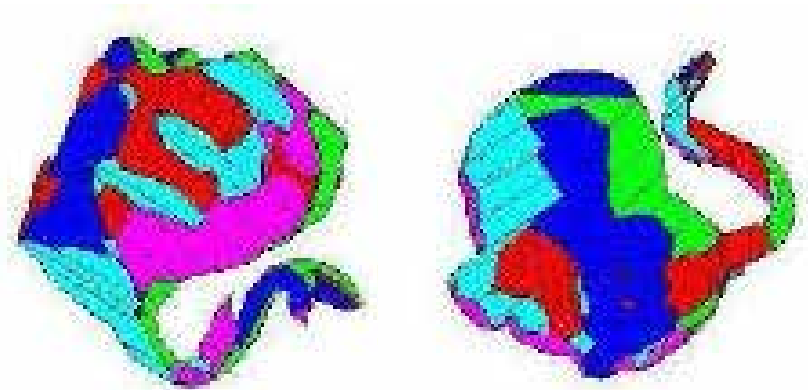
3D shape from silhouettes



(a)



Build 3D model

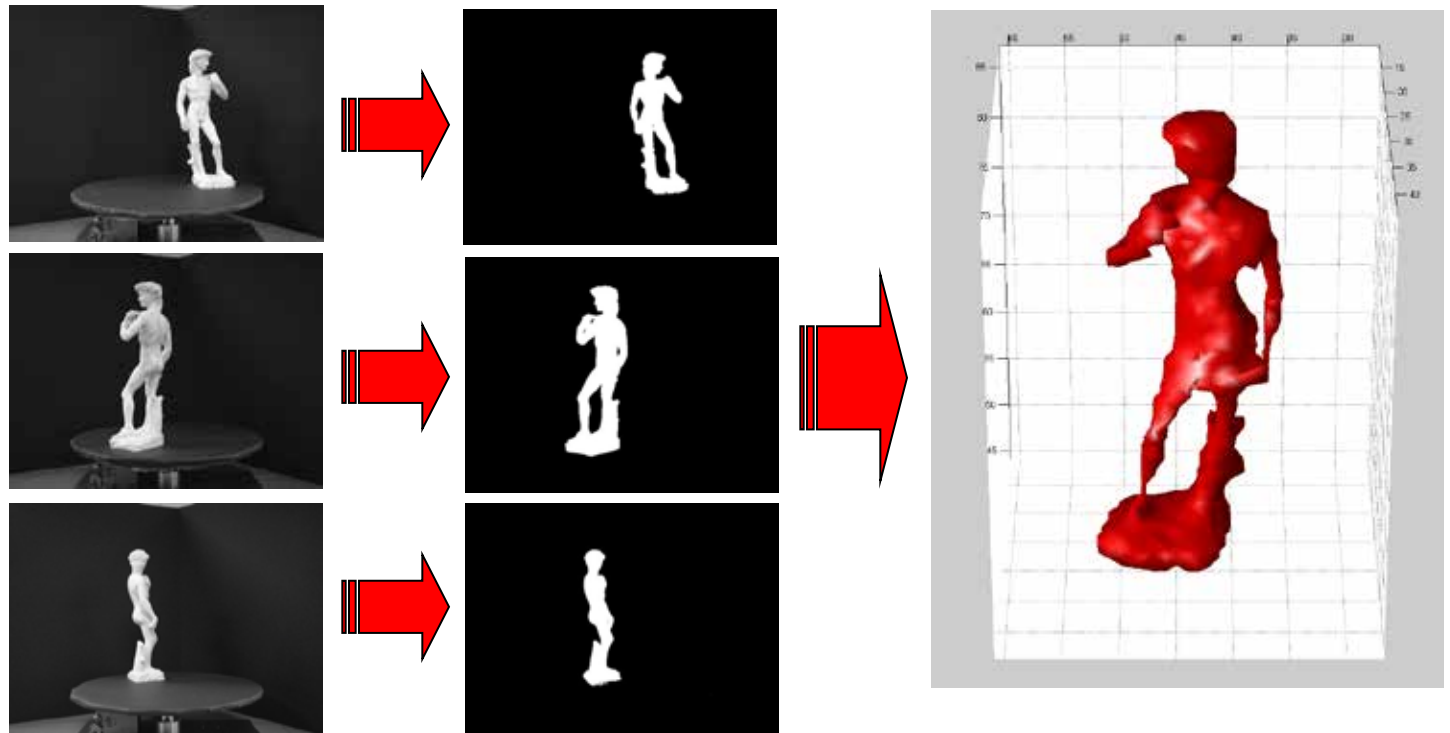


Visualize 3D model from arbitrary viewing angles



Example

- Compute visual hull with silhouette images from multiple calibrated cameras
- Compute Silhouette Image
- Volumetric visual hull computation
- Display the result





Shape from Shading

Photometric Stereo Christopher Bireley



Bandage Dog



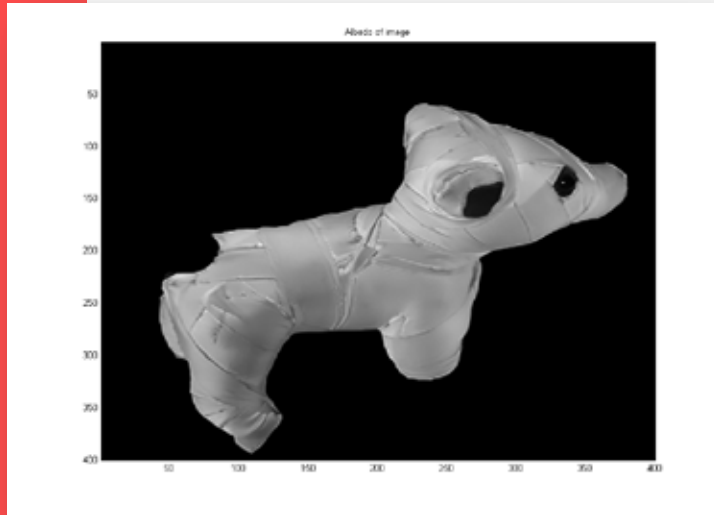
Imaging Setup

Preprocessing

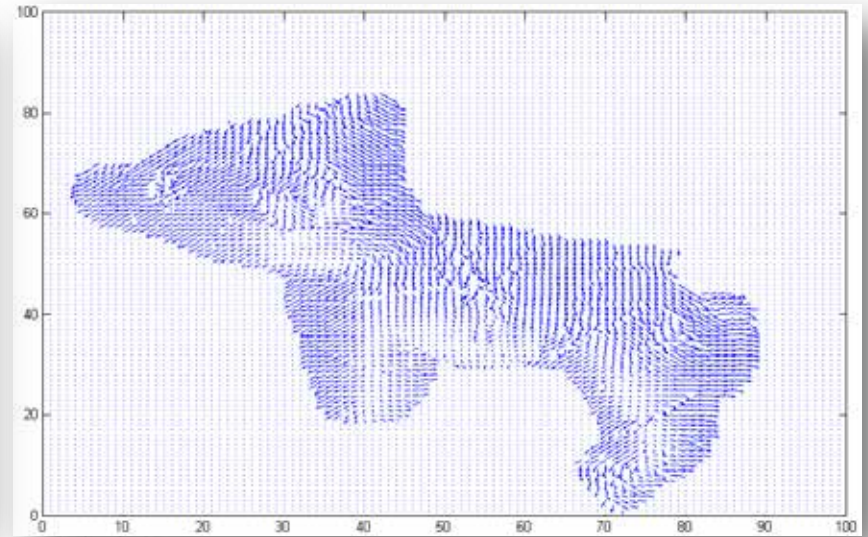
- Remove background to isolate dog
- Filter with NL Means



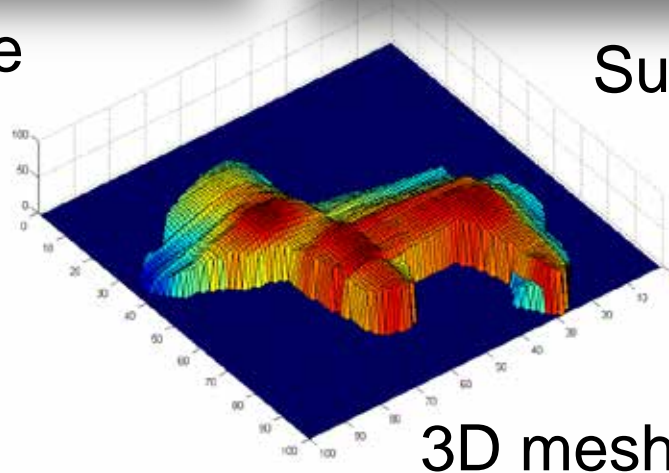
Photometric Stereo Christopher Bireley



Albedo image



Surface Normals



3D mesh

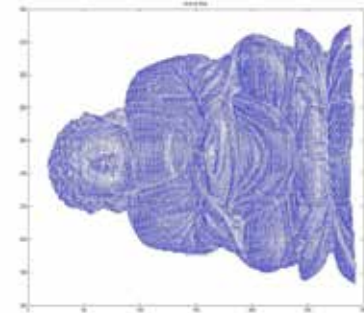
Results – Lord Buddha Images – Pre-Processed Images Guozhen Fan and Aman Shah



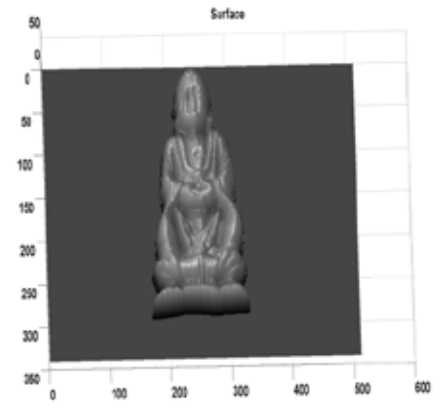
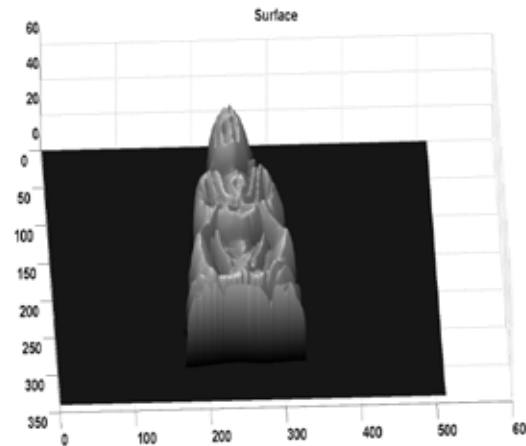
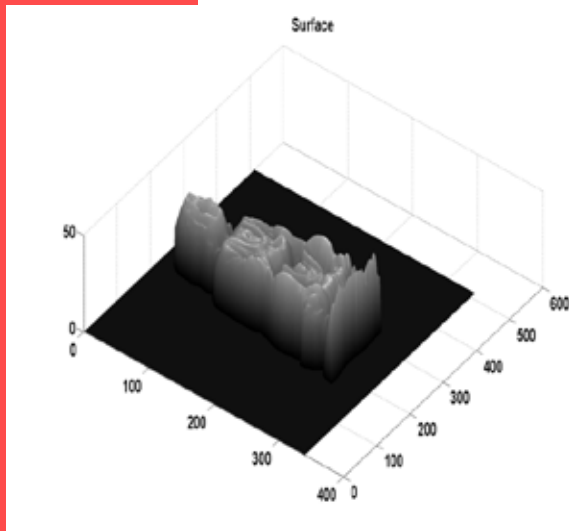
Original Image



Albedo Map



Surface Normals




Obtained Surfaces from different angles



Object Tracking

Object Tracking: Using Deformable Models in Vision



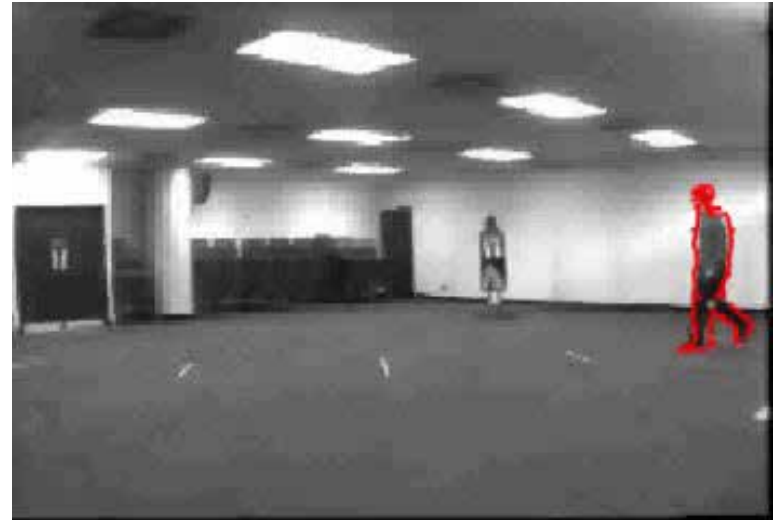


Object Tracking: Using Deformable Models in Vision: II

**Unifying Boundary and
Region-based information for
Geodesic Active Tracking**



Object Tracking III



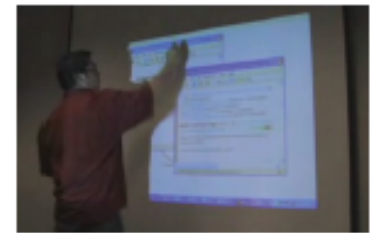
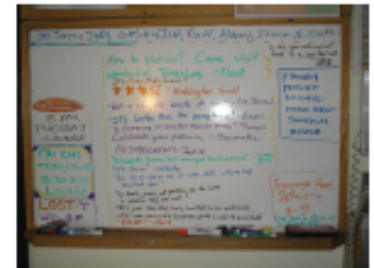


Computer Vision Systems

Webcam Based Virtual Whiteboard

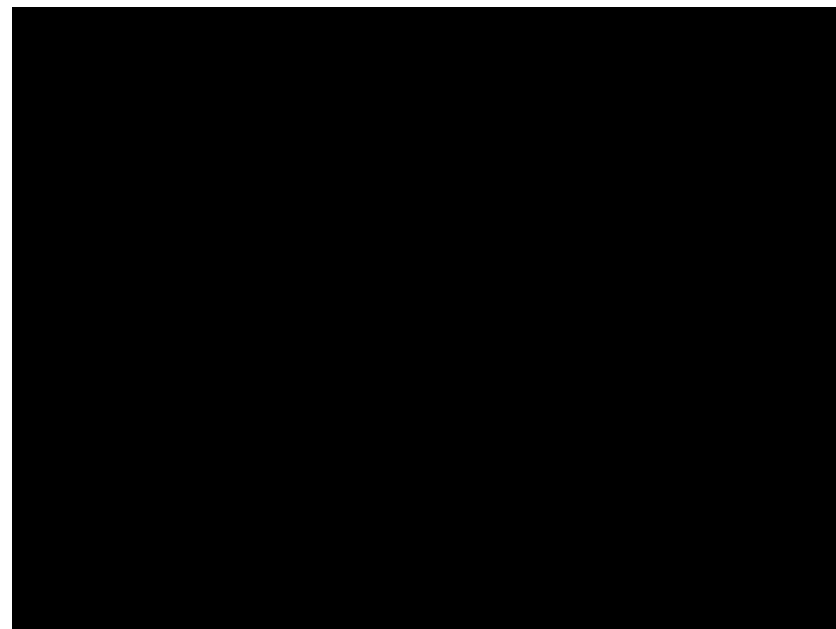
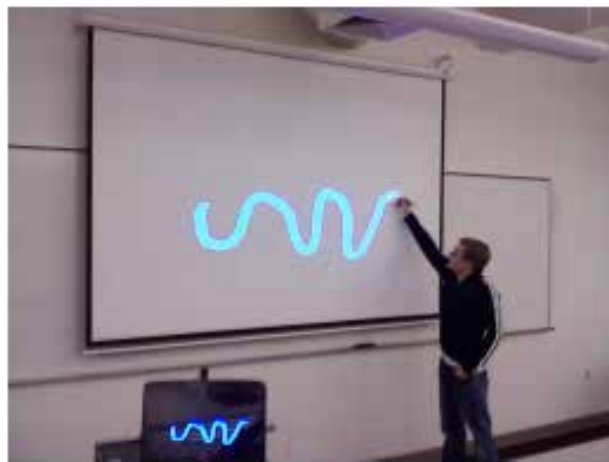
Jon Bronson James Fishbaugh

- Blackboards came first
- Whiteboards eventually followed
- Virtual Whiteboards are coming
- Basic Idea:
 - Write on any surface
 - Use no ink/chalk
 - Store all information to disk



Webcam Based Virtual Whiteboard

Jon Bronson James Fishbaugh



Real-Time 3D Glowstick Detection

Computer Vision Project 2009

Andrei Ostanin



Detecting the 3D position of glowsticks in real-time using two cameras.

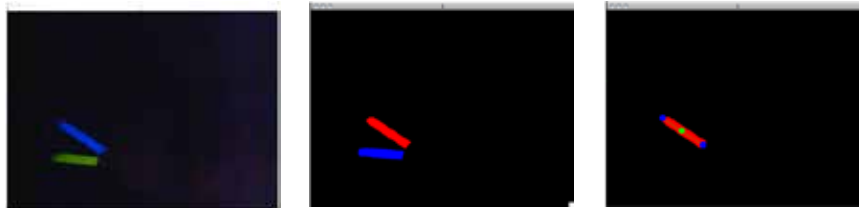
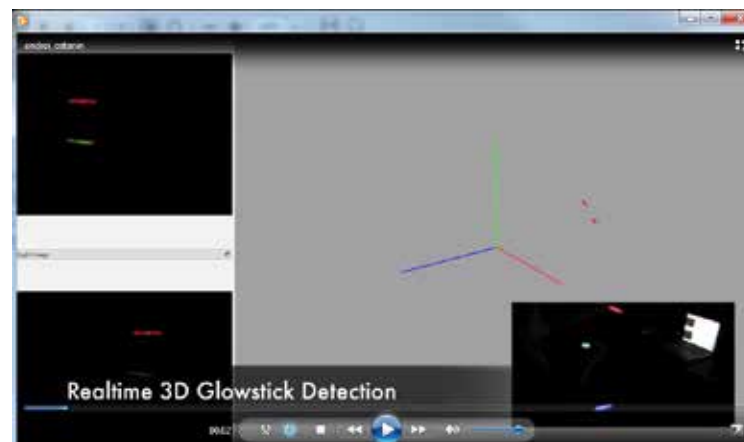
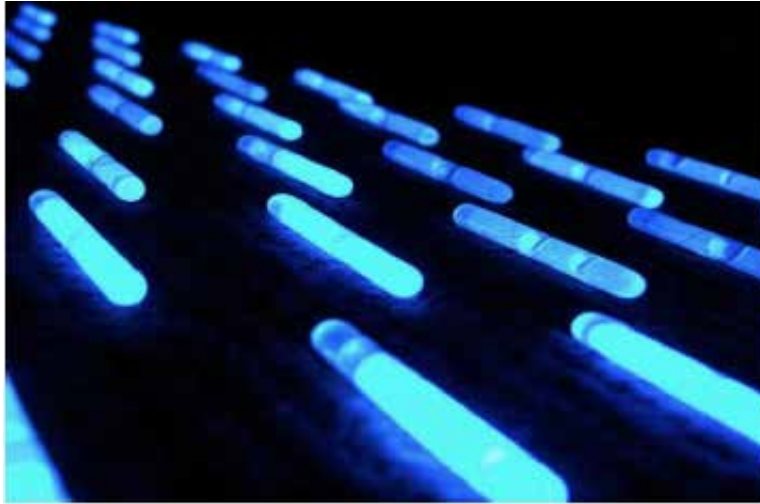


Figure 2: Camera input images

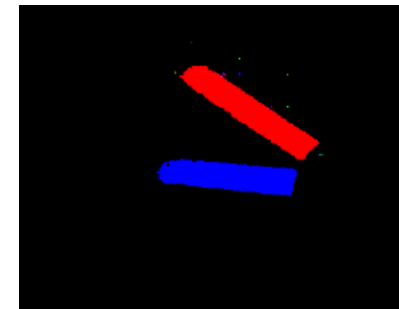


Realtime Glowstick Detection

Andrei Ostanin



- ▶ Capture the 3D position of glowsticks in real-time using two webcams
- ▶ Environment dark enough that glowsticks are easily segmented out
- ▶ Prefer speed over correctness



[movie](#)

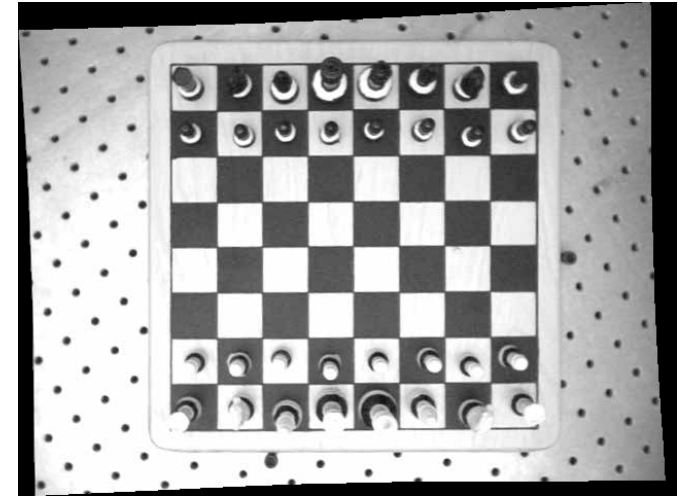
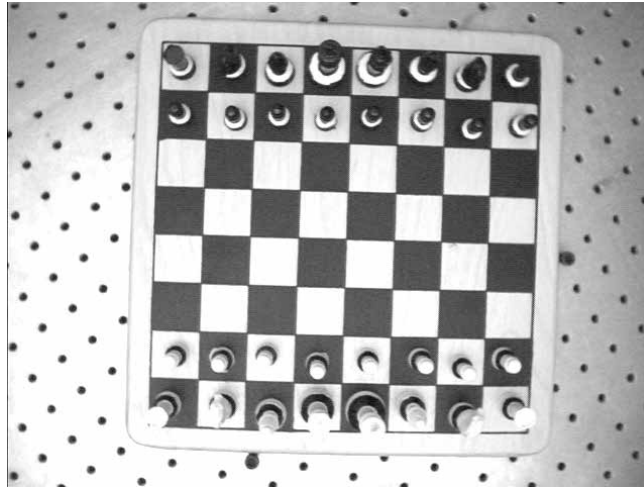


Student Project: Playing Chess, Recognition and Simulation

- Track individual chess pieces
- Maintain state of board
- Graphically represent state changes and state
- D. Allen, D. McLaurin
UNC
- Major ideas:
 - 3D from stereo
 - detect and describe changes
 - Use world knowledge (chess)



Calibration, Rendering & Replay



Movie





Goal and objectives

From Snapshots, a 3-D View

NYT, August 21, 2008, Personal Tech

<http://www.nytimes.com/2008/08/21/technology/personaltech/21pogue.html>



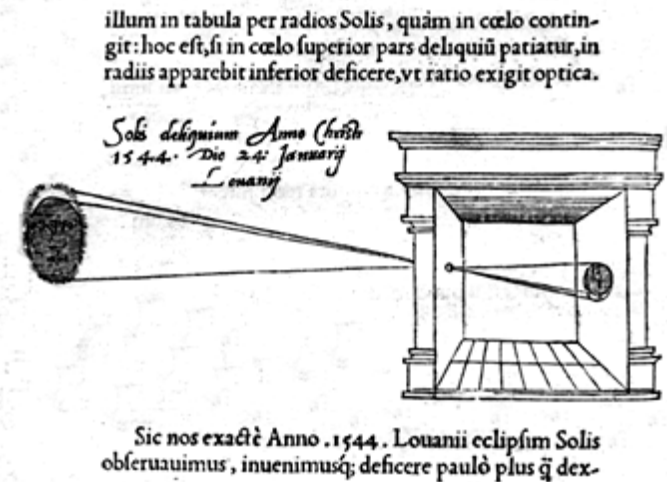
Stuart Goldenberg



Next class: Image Formation

Chapter 1: Cameras

- Please find pdf copies of Chapters 1&2, Forsyth&Ponce, on the website.
- Purchase the course book on your own.



Assignment:

- Read Chapter 1: Cameras, Lenses and Sensors: See Course [home page](#)