



3D Computer Vision Introduction

Guido Gerig

CS 6320, Spring 2012

gerig@sci.utah.edu



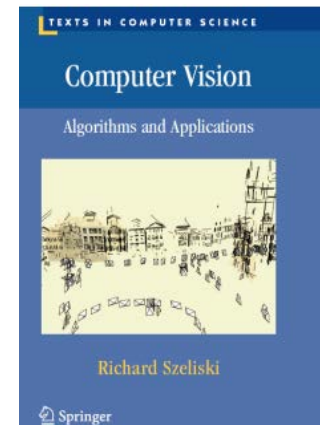
Administrivia

- Classes: M & W, 1.25-2:45
Room WEB L126
- Instructor: Guido Gerig
gerig@sci.utah.edu
(801) 585 0327
- Prerequisites: CS 6640 ImProc (or equiv.)
- Textbook:

[Computer Vision:](#)

Algorithms and Applications“
by Richard Szeliski

- Organization
web-site
(slides, documents and assignments)





Teaching Assistant

- TA: Kathlea Quebbeman, SoC
quebbeka@cs.utah.edu
kathlea.quebbeman@slcc.edu
- HW/SW: Matlab+ev. Imaging Toolbox
CADE lab WEB 130
<http://www.cade.utah.edu/>
- Office Hours: tbd



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.

Textbook



1.



Computer Vision: Algorithms and Applications (Texts in Computer Science)
by Richard Szeliski (Nov 24, 2010)

Buy new: ~~\$89.95~~ **\$62.74**

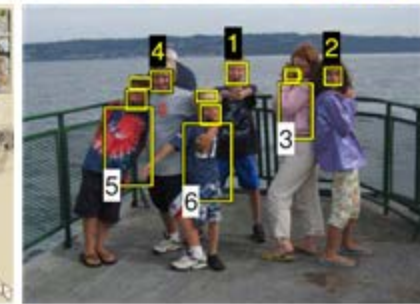
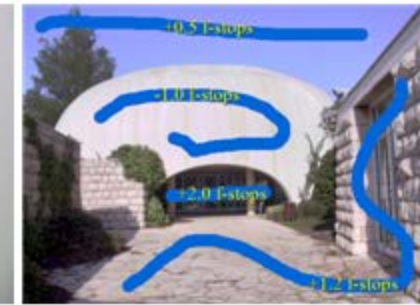
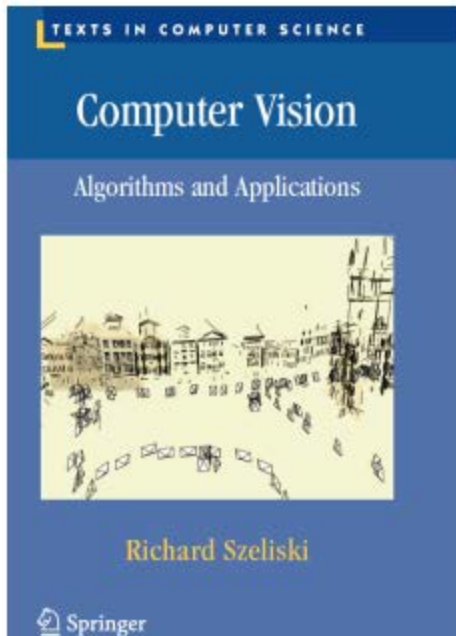
19 new from **\$62.74**

22 used from **\$59.60**

Get it by **Tuesday, Jan 10** if you order in the next **31 hours** and choose one-day shipping.

Computer Vision: Algorithms and Appl

© 2010 [Richard Szeliski](#), Microsoft Research





Grading

- Assignments (4-6 theory/prog.): 50%
- Final project (incl. proposal and presentation): 20%
- Class participation: 10%
- Final project replaces final exam



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 WebCT email list (since most students use individual emails)
- It will be your responsibility to regularly read the Announcements on WebCT
- We don't need a laptop for the class, please keep them closed !!!!!
- Please interact, ask questions, clarifications, input to instructor and TA



Syllabus

- See separate syllabus (on blackboard and linked to UofU class list).
- [Document](#)



Goal and Objectives

From Snapshots, a 3-D View

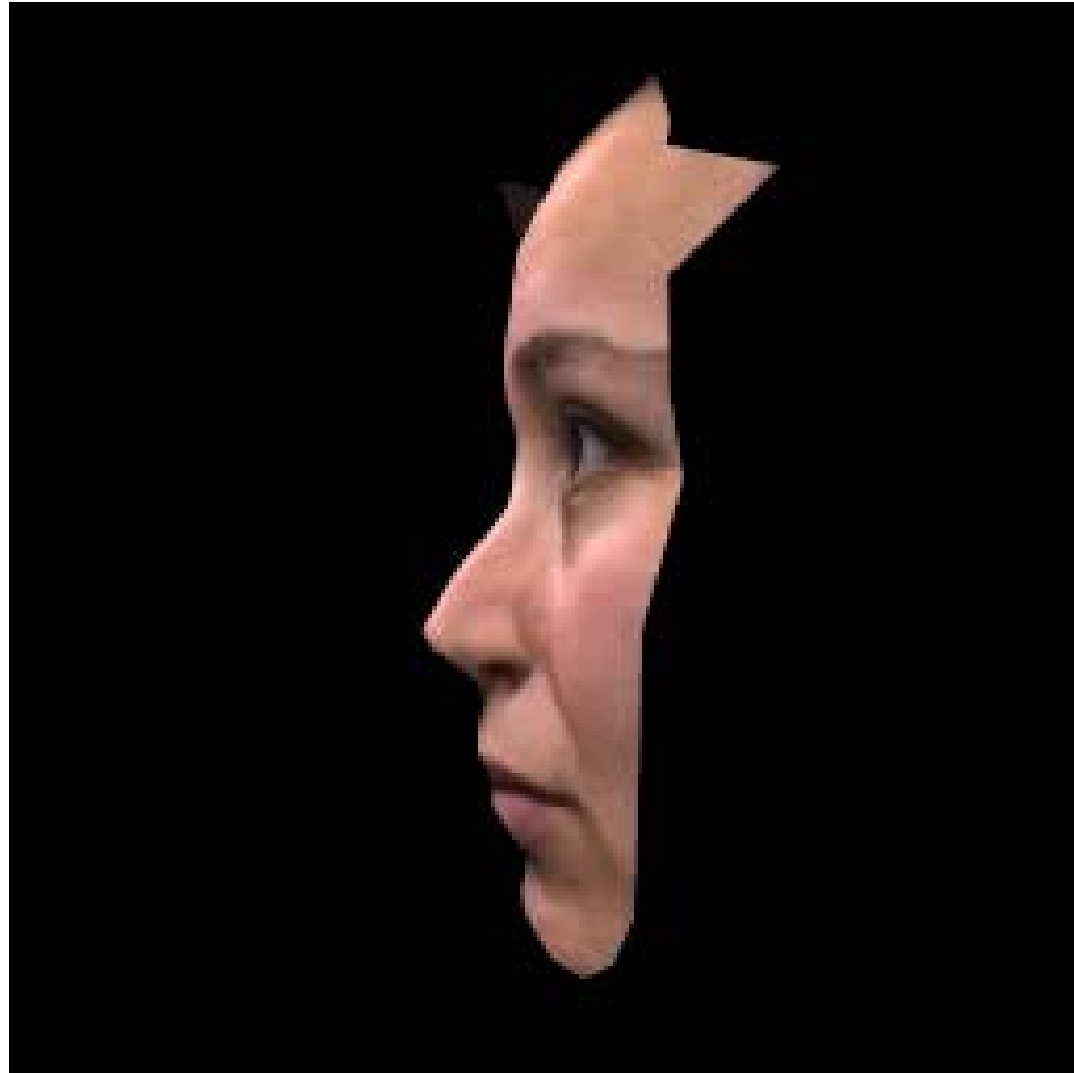
NYT, August 21, 2008, Personal Tech

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



Stuart Goldenberg

Modeling 3D Structure from Pictures or 3D Sensors



Modeling ctd.





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.



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



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.



Properties of Vision

- 3D representations are easily constructed
 - There are many different cues.
 - Useful
 - to humans (avoid bumping into things; planning a grasp; etc.)
 - in computer vision (build models for movies).
 - Cues include
 - multiple views (motion, stereopsis)
 - texture
 - shading



Properties of Vision

- People draw distinctions between what is seen
 - “Object recognition”
 - This could mean “is this a fish or a bicycle?”
 - It could mean “is this George Washington?”
 - It could mean “is this poisonous or not?”
 - It could mean “is this slippery or not?”
 - It could mean “will this support my weight?”
 - Great mystery
 - How to build programs that can draw useful distinctions based on image properties.



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?”



Main topics

- Camera & Light
 - Geometry, Radiometry, Color
- Digital images
 - Filters, edges, texture, optical flow
- Shape (and motion) recovery
 - Multi-view geometry
 - Stereo, motion, photometric stereo, ...
- Segmentation
 - Clustering, model fitting, probabilistic
- Tracking
 - Linear dynamics, non-linear dynamics
- Recognition
 - templates, relations between templates



Camera and lights

- How images are formed
 - Cameras
 - What a camera does
 - How to tell where the camera was
 - Light
 - How to measure light
 - What light does at surfaces
 - How the brightness values we see in cameras are determined
 - Color
 - The underlying mechanisms of color
 - How to describe it and measure it



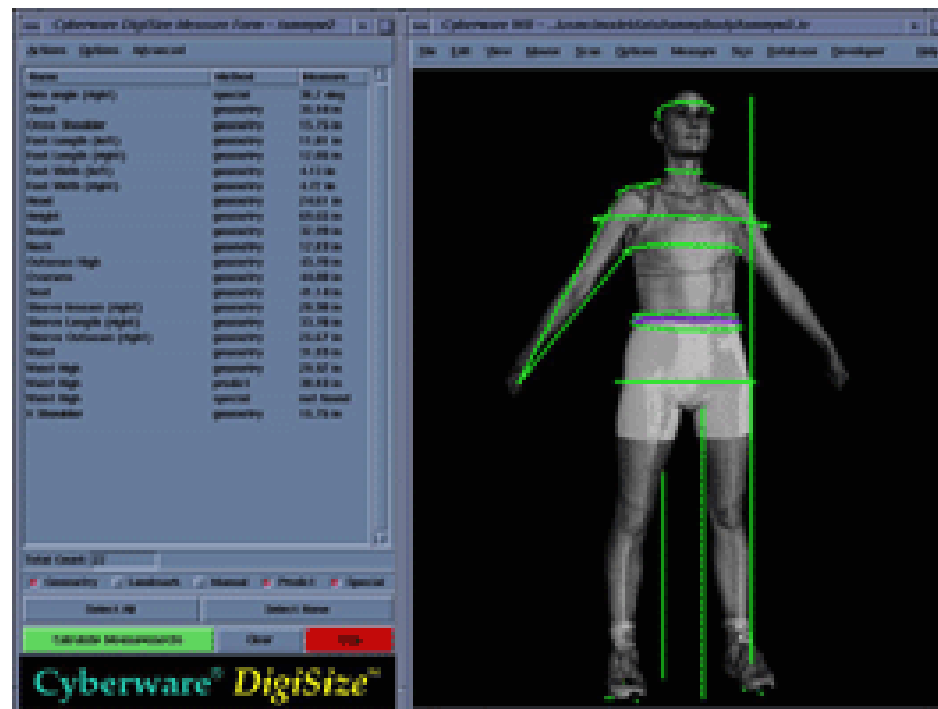
Motivation

- (some slides modified from Marc Pollefeys, UNC Chapel Hill & ETH Zurich)

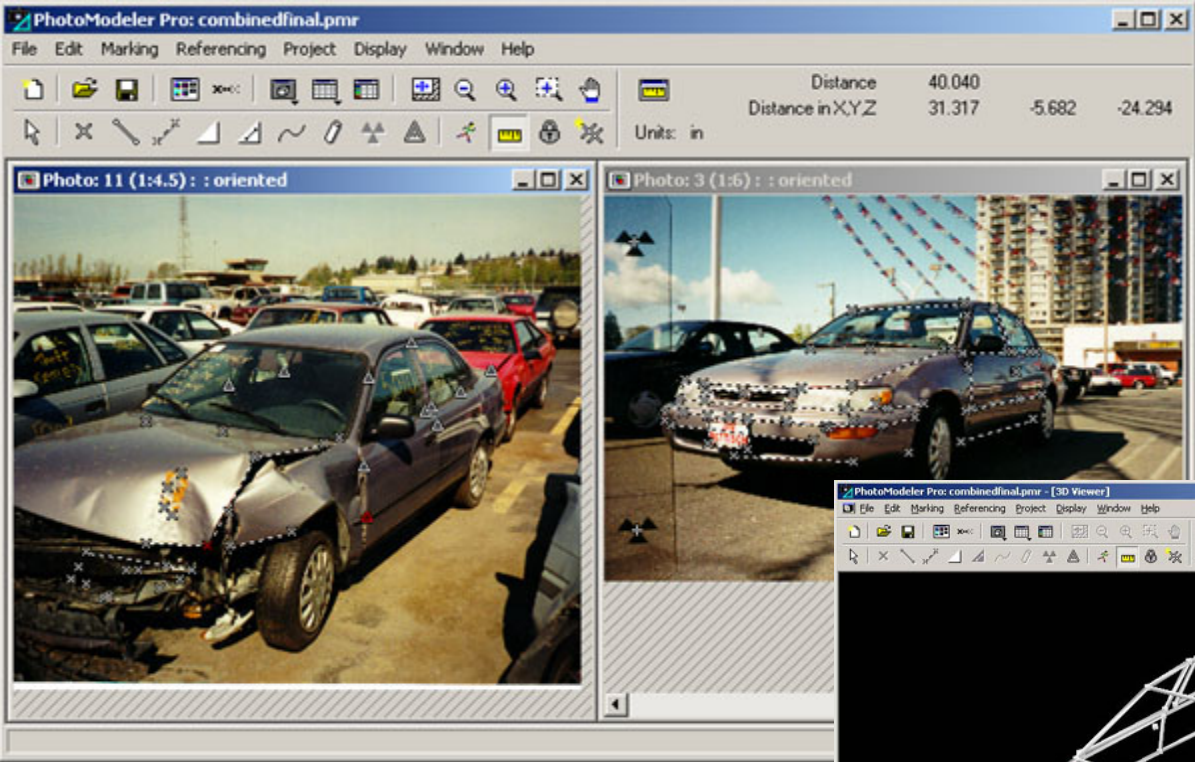


Clothing

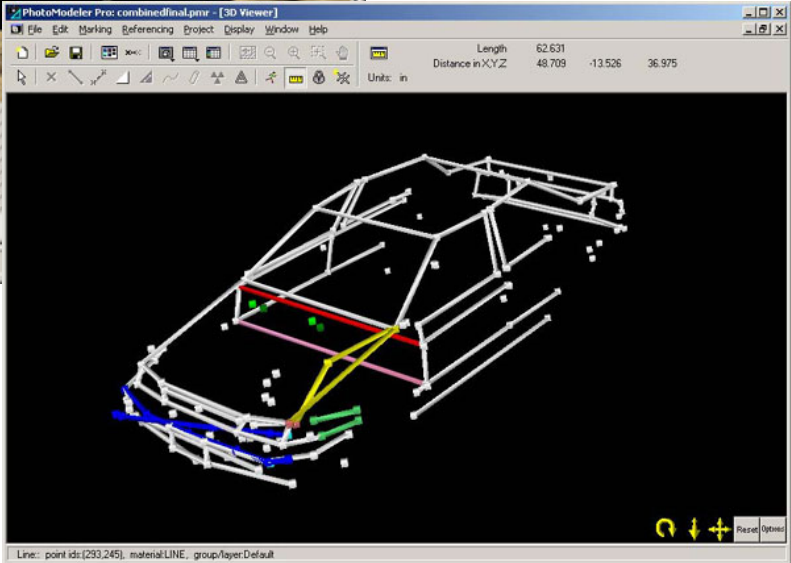
- Scan a person, custom-fit clothing



Forensics



PhotoModeler



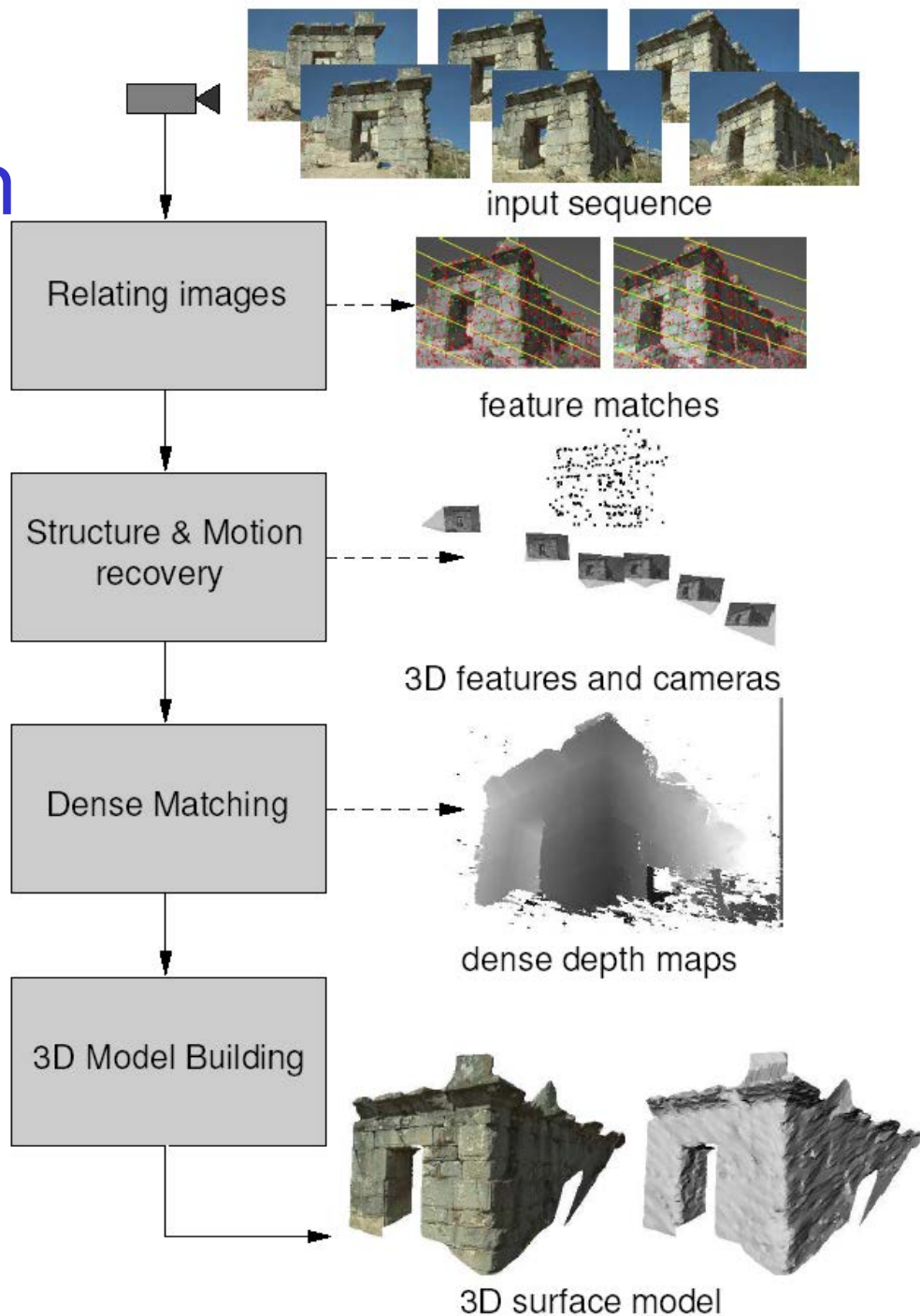


3D urban modeling



drive by modeling in Baltimore

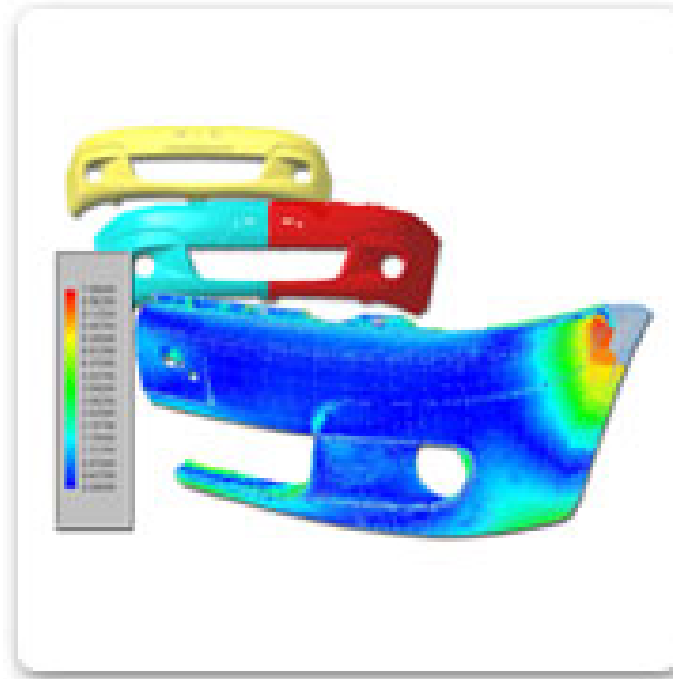
Structure from Motion





Industrial inspection

- Verify specifications
- Compare measured model with CAD



Scanning industrial sites



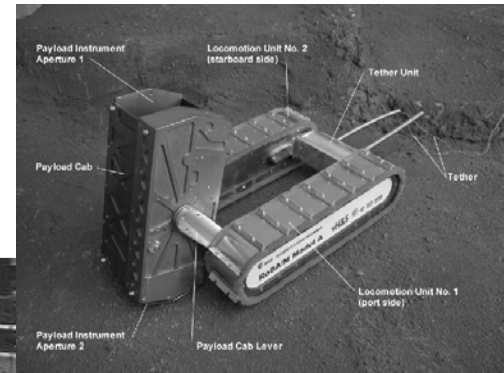
as-built 3D model of off-shore oil platform





Robot navigation

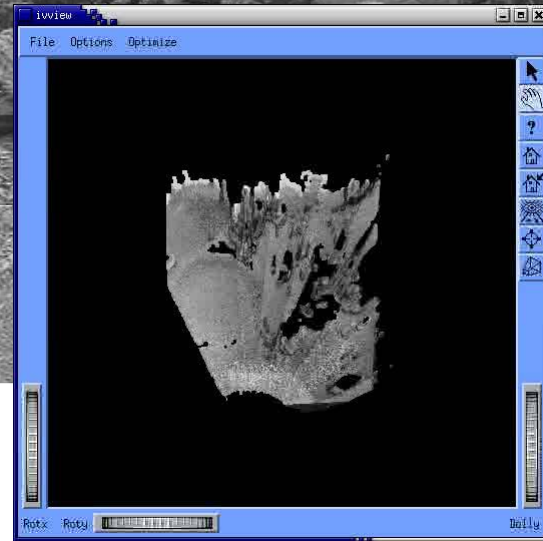
ESA project
our task: Calibration + Terrain modelling + Visualization



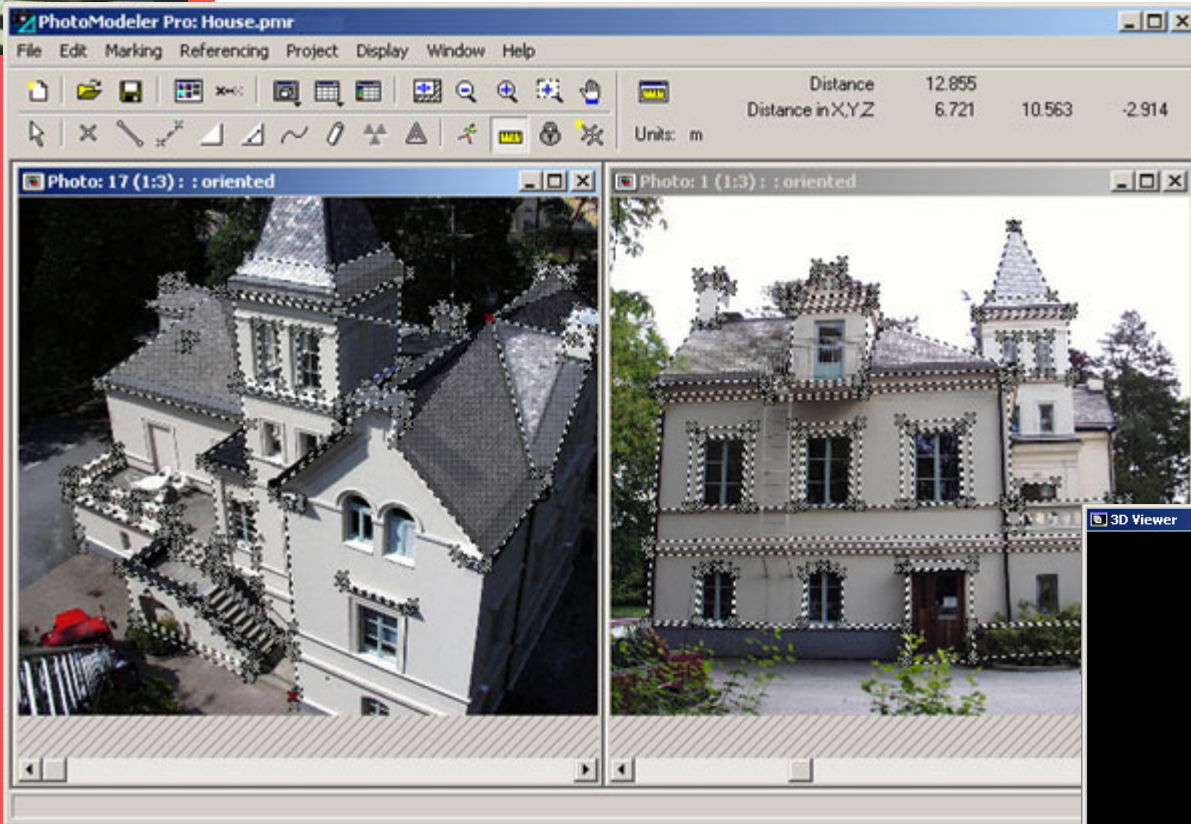
small tethered rover



pan/tilt stereo head

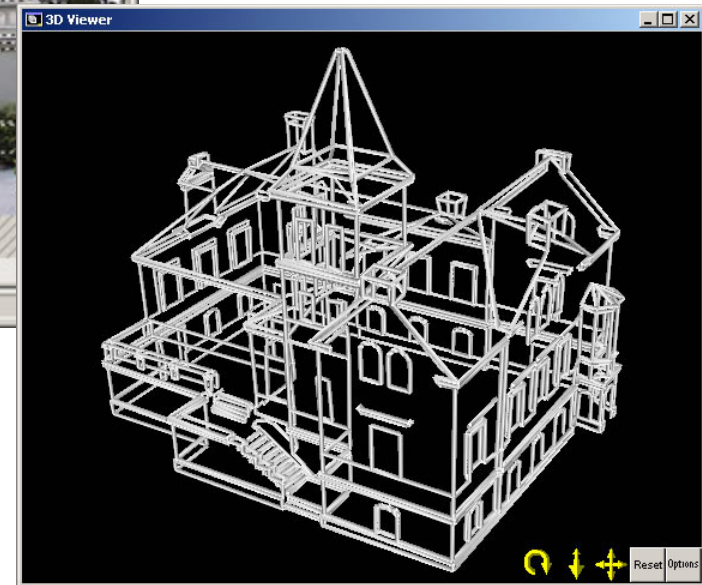


Architecture

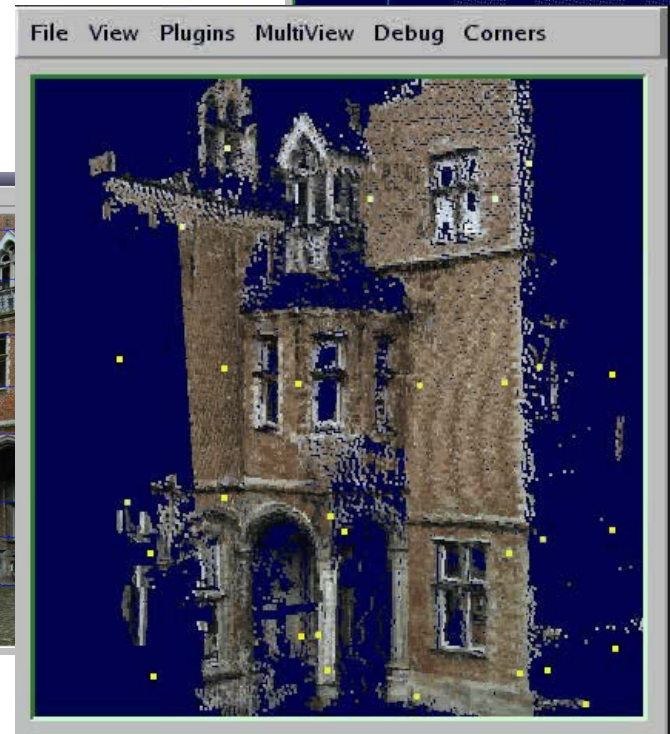
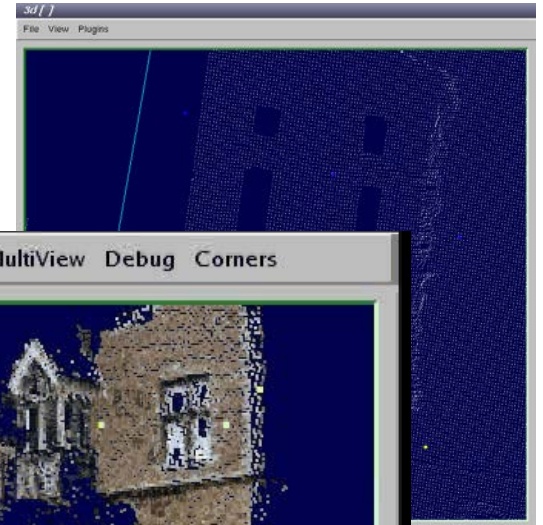
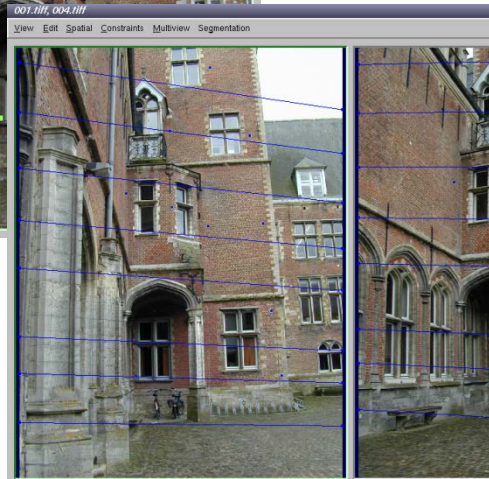
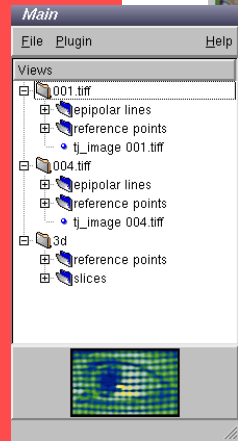


Survey
Stability analysis
Plan renovations

PhotoModeler



Architecture



Survey
Stability analysis
Plan renovations



Cultural heritage

Virtual Monticello



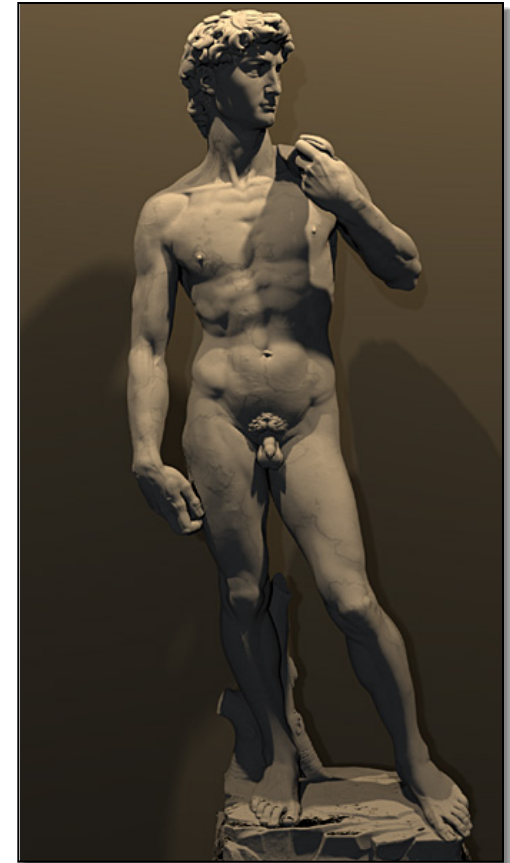
Allow virtual visits



Cultural heritage



Stanford's Digital Michelangelo

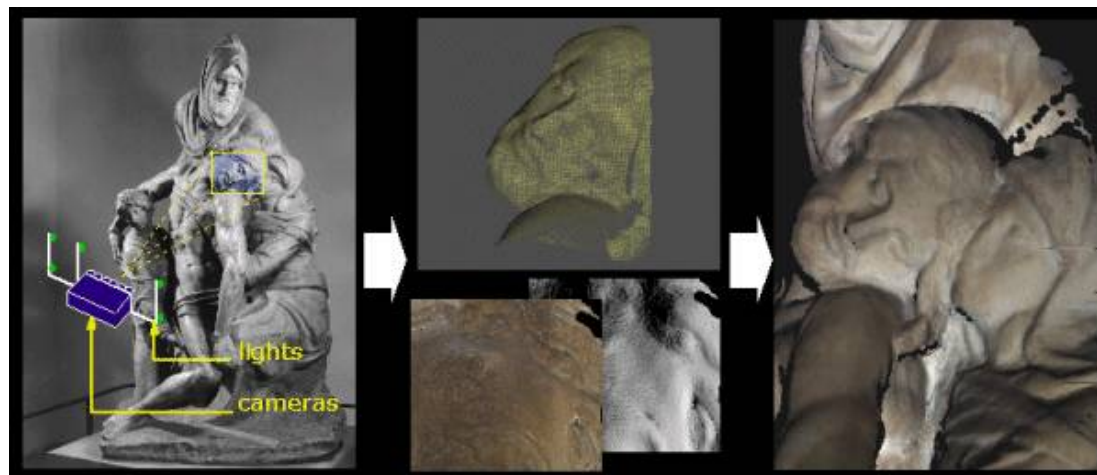


Digital archive
Art historic studies



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)



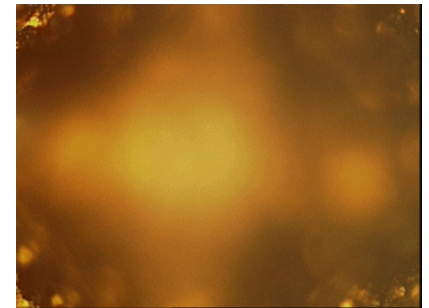
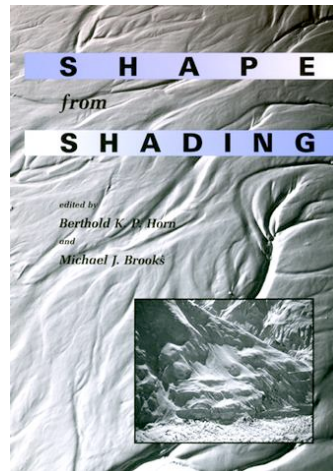
Sony's Eye Toy: Computer Vision for the masses



Background segmentation/
motion detection
Color segmentation
...

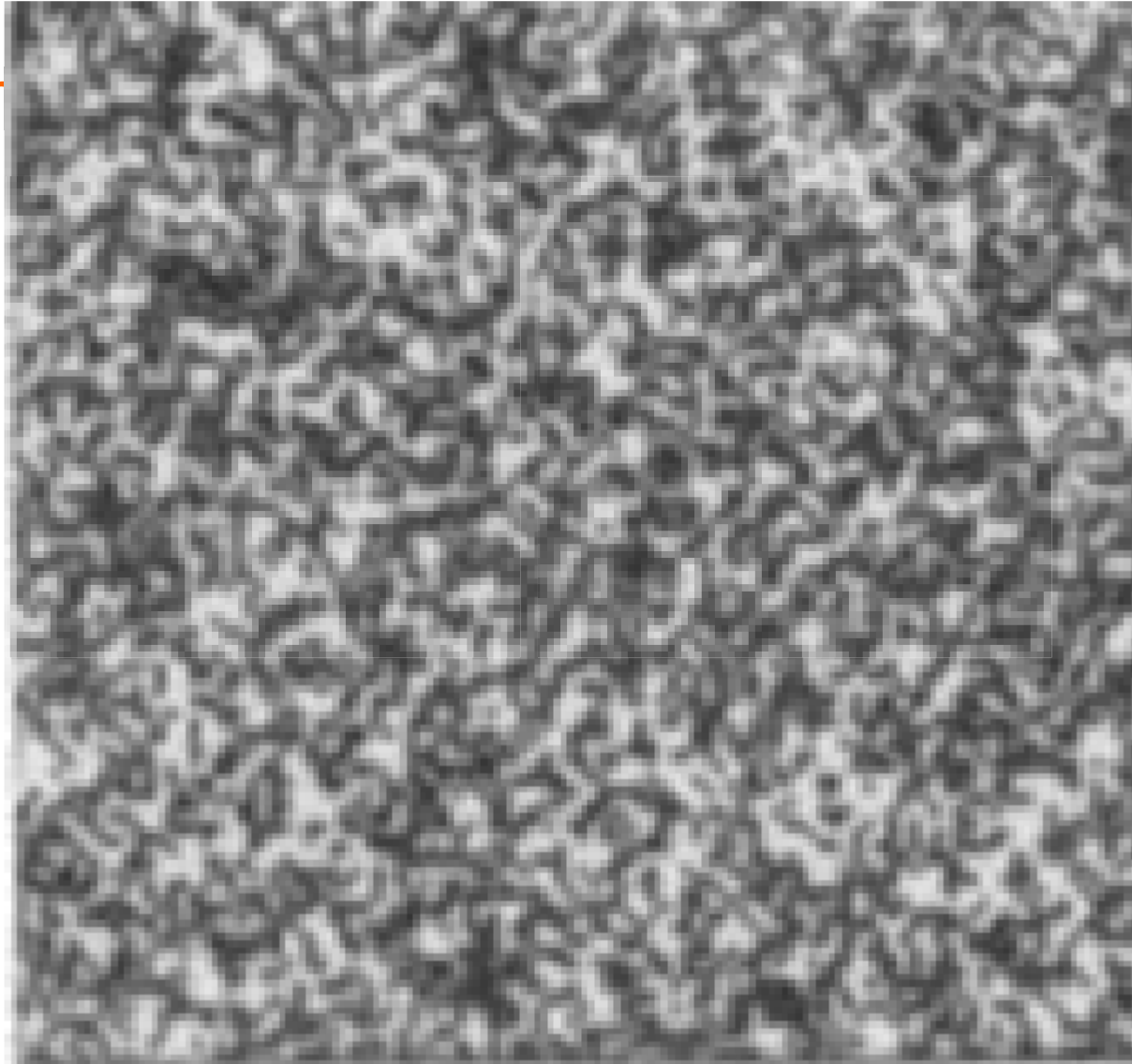
Shape from ...

many different approaches/cues



Optical flow

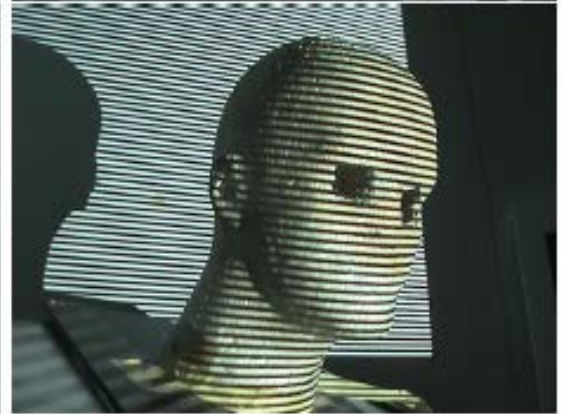
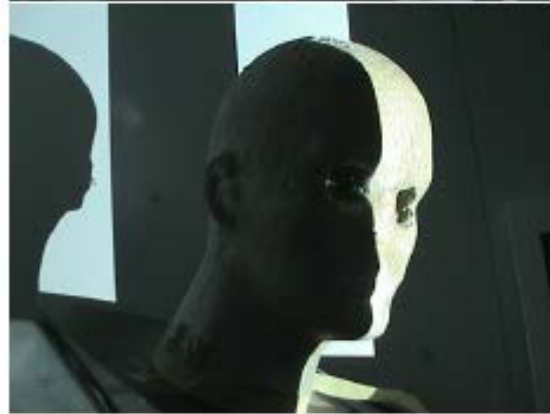
Wr



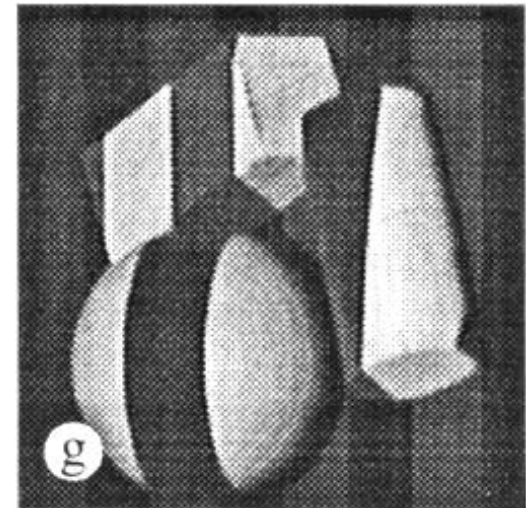
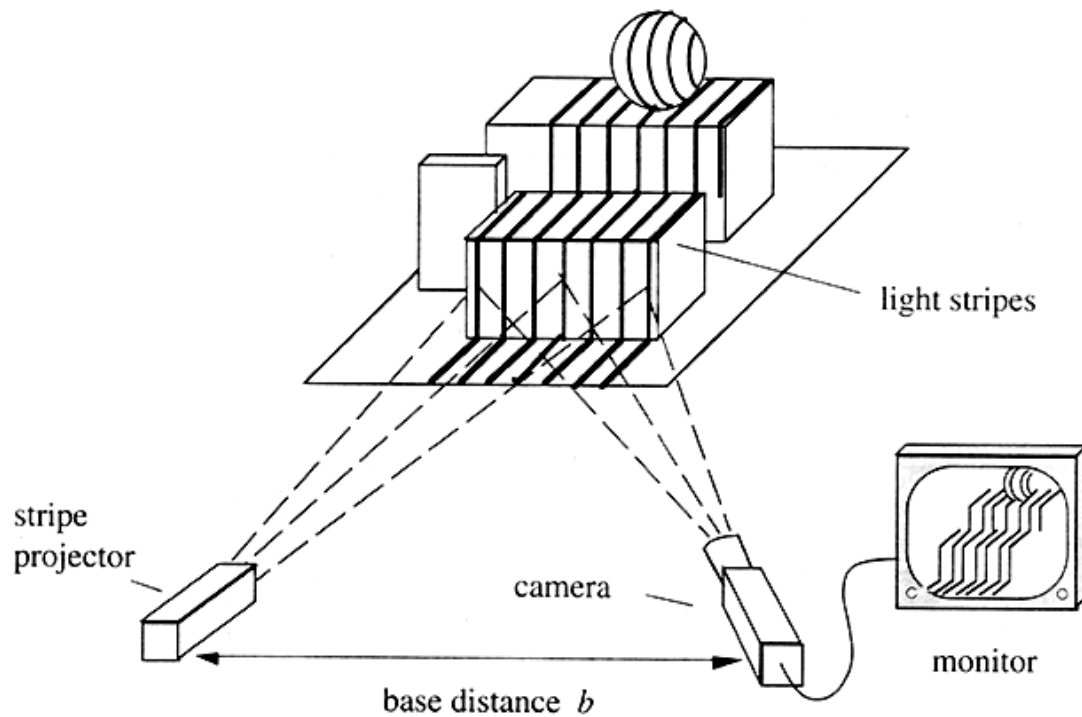
Optical flow



Active Vision: Structured Light



Static Light Pattern Projection



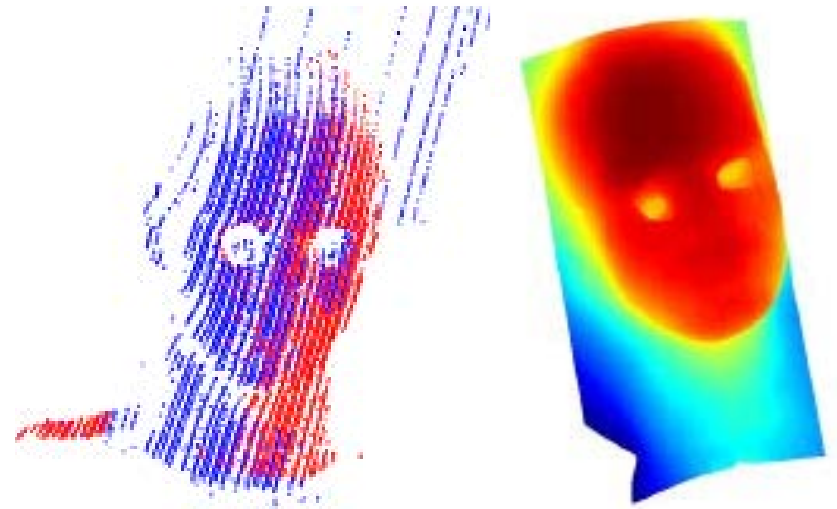
Segmentation of stripes from set of images

Geometry between projector, sensor and objects

Active Vision: Structured Light



Segmentation: Binarization
and coding of stripes



3D model extracted
from stripe pattern

Spatiotemporal Volumes

- Eric Bennet, UNC,
PhD thesis



Spatiotemporal Volumes



Figure 3.3: Visualization of a spatio-temporal volume and a spatio-temporal cut plane. On the left, a 10 second video is presented as a spatio-temporal volume. The front of the volume shows the first frame, the right side shows the right-most vertical line through time, and the top shows the top-most scanline through time. On the right, the volume has been rotated and been cut using two planar cuts. The first, parallel to the front face, has shortened the video. The second has revealed a different scanline which shows the motion of people walking during the duration of the video.

Video-sequence editing II

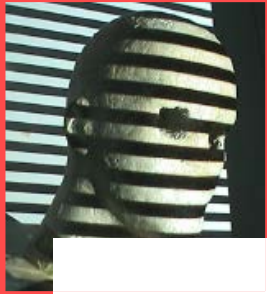


Figure 3.12: Sheared volume after the Frisbee has been stabilized. Three visualizations of the volume are shown above. The top-left image shows the sheared volume at a given time. The right image shows a fixed column through time and the bottom image shows a fixed scan line after the Frisbees has been stabilized.



Motion Tails

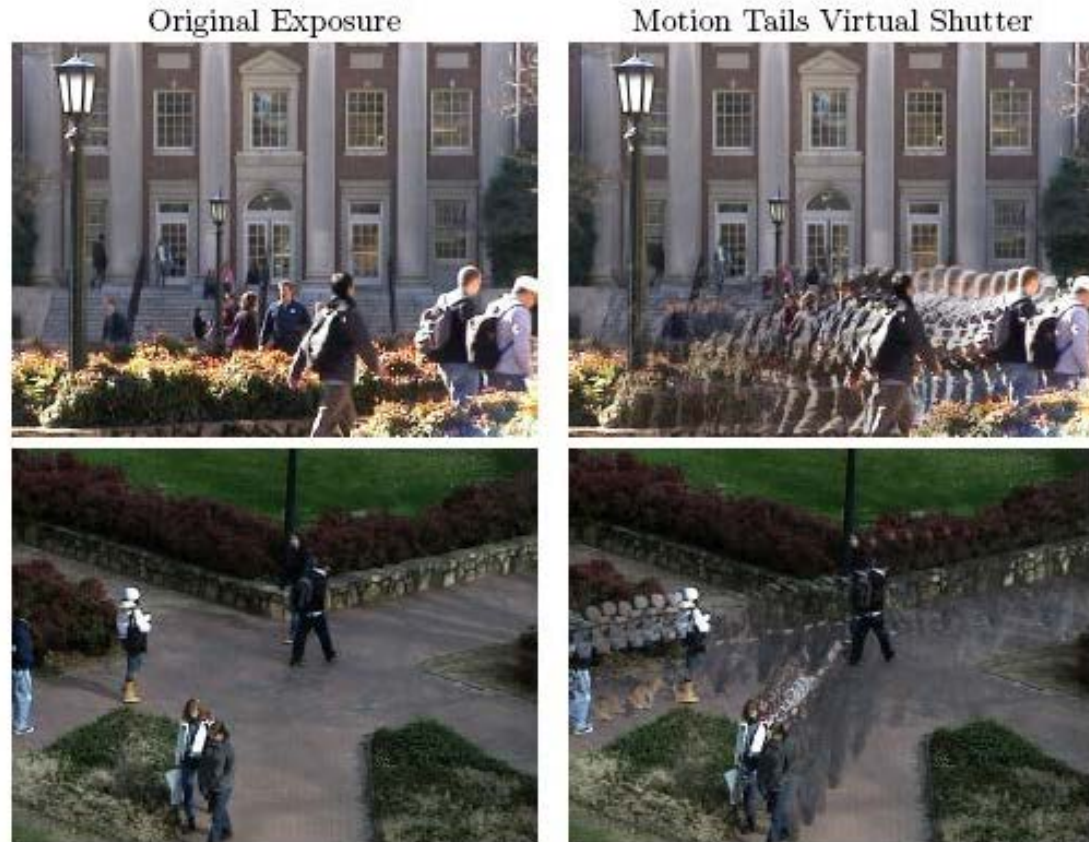



Figure 5.9: Two examples of using motion tails to depict dense motion paths between sampled time-lapse frames. The building front result (above) uses uniform sampling, while the crowded sidewalk (below) is non-uniformly sampled.

Object Tracking: Using Deformable Models in Vision



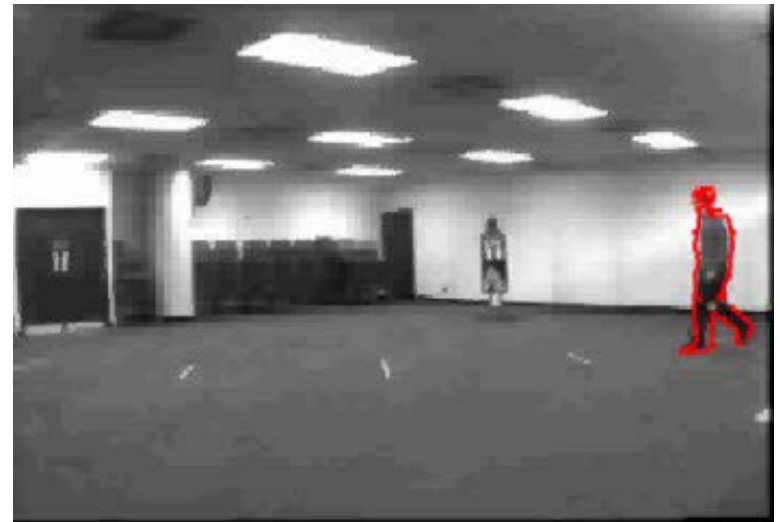


Object Tracking: Using Deformable Models in Vision: II

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

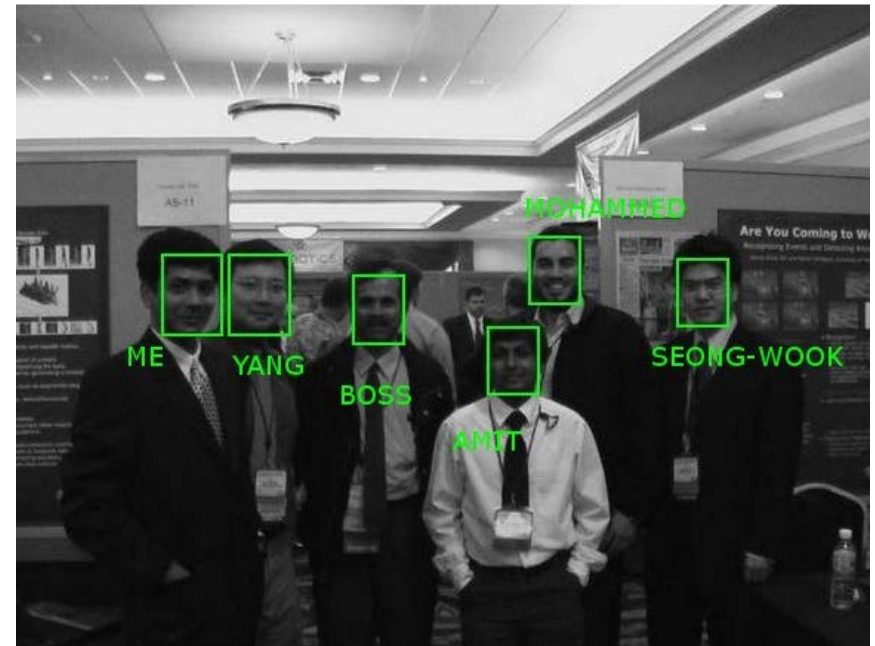


Object Tracking III





Face detection



<http://www.umiacs.umd.edu/~aswch/test.html>

<http://vasc.ri.cmu.edu/cgi-bin/demos/findface.cgi>



Examples of Student Projects

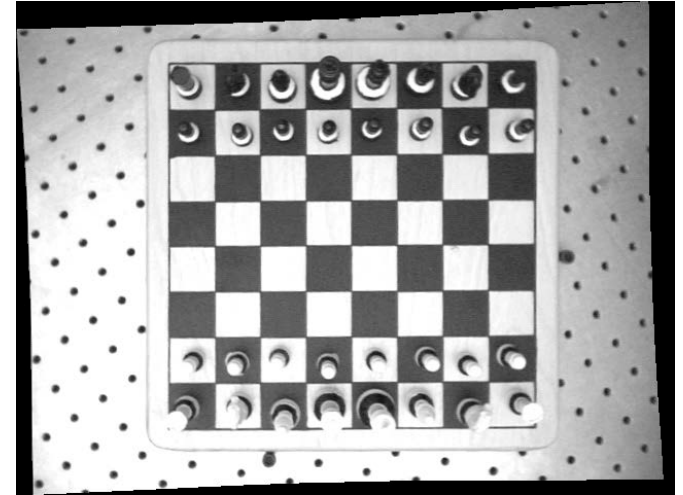
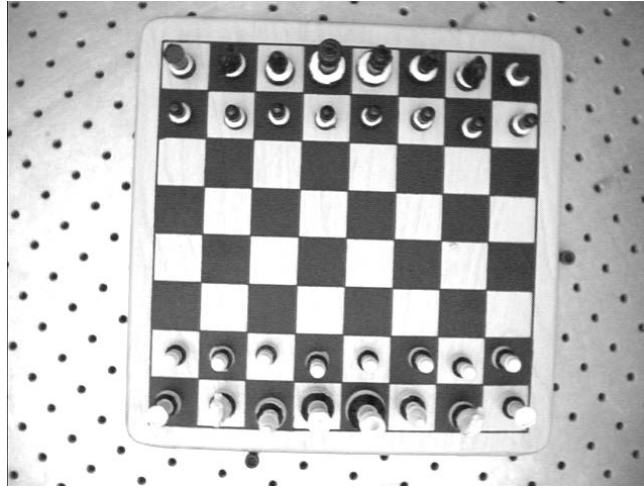


Student Project: Playing Chess, Recognition and Simulation

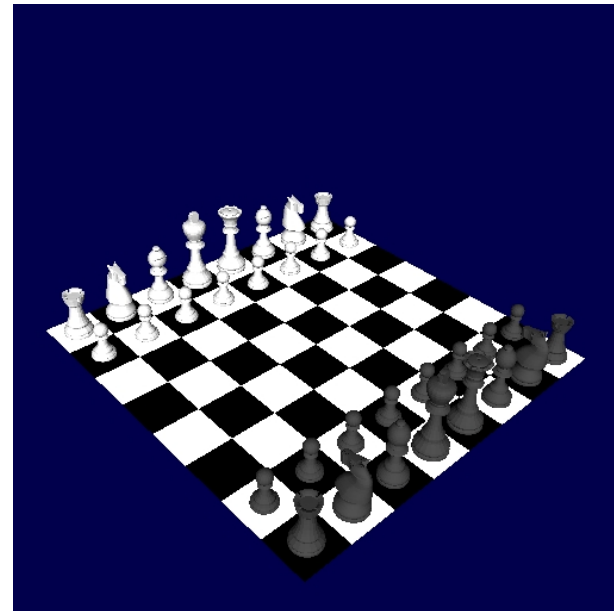
- Track individual chess pieces
- Maintain state of board
- Graphically represent state changes and state
- D. Allen, D. McLaurin
UNC
- Major ideas:
 - 3D from stereo
 - detect and describe changes
 - Use world knowledge (chess)



Calibration, Rendering & Replay



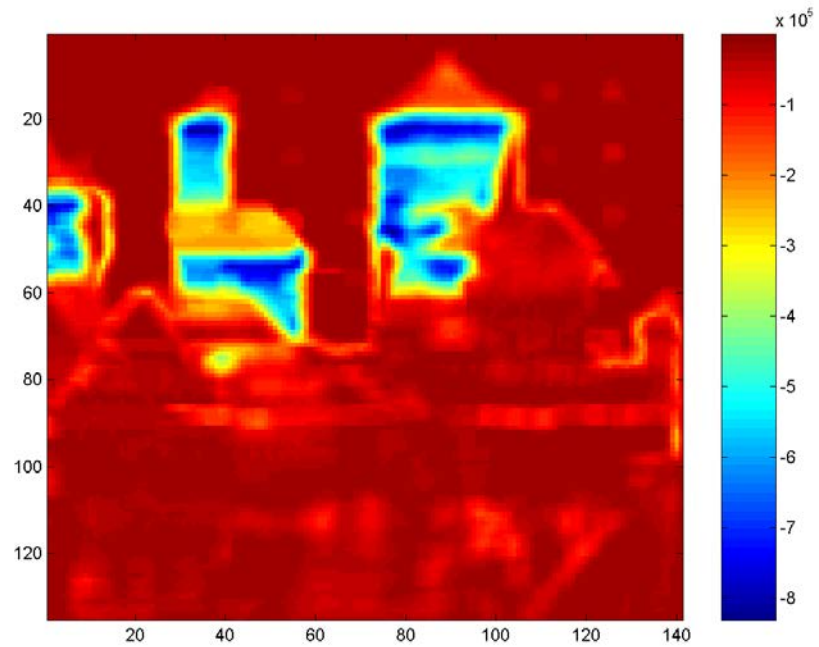
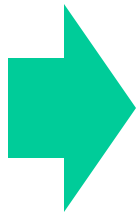
Movie



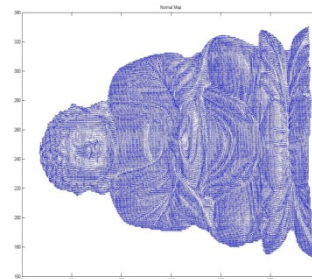
Enhanced Correlation for Stereo Vision



- Andrew Nashel
- Correlation map produced by precomputation method:



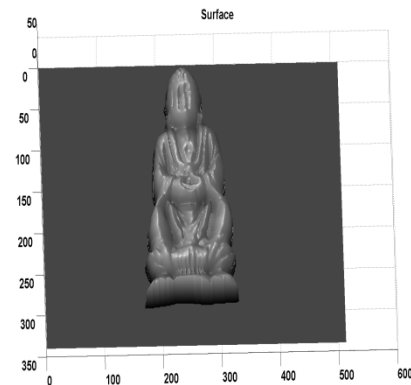
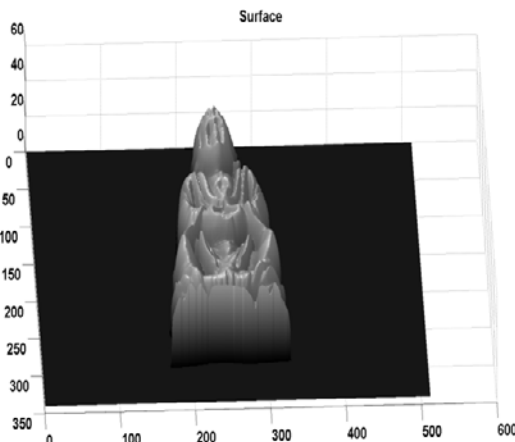
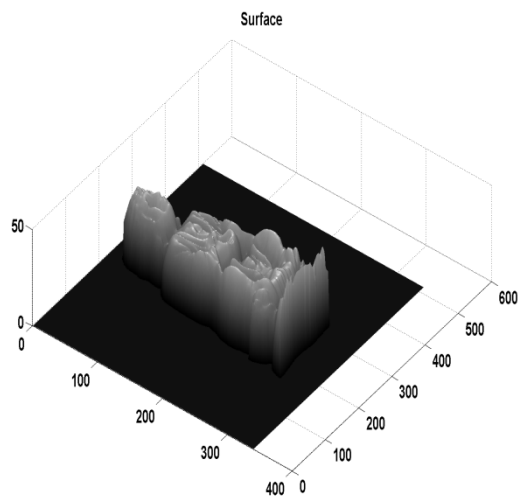
Results – Lord Buddha Images – Pre-Processed Images Guozhen Fan and Aman Shah



Original Image

Albedo

Surface Normals



Obtained Surfaces from different angles

Photometric Stereo Christopher Bireley



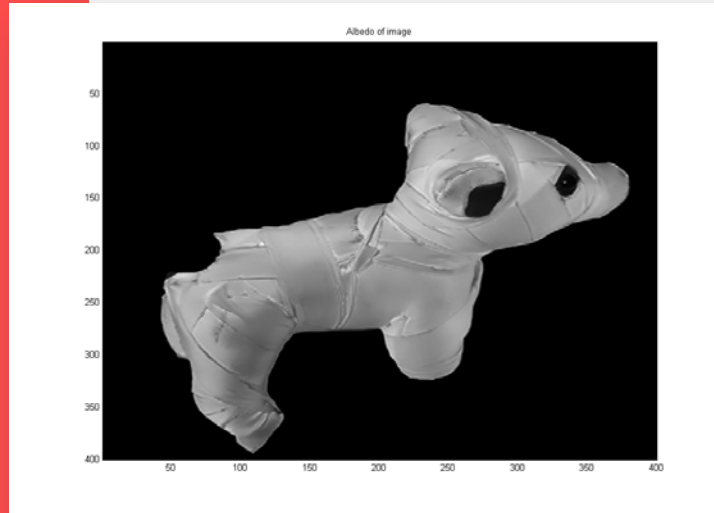
Bandage Dog



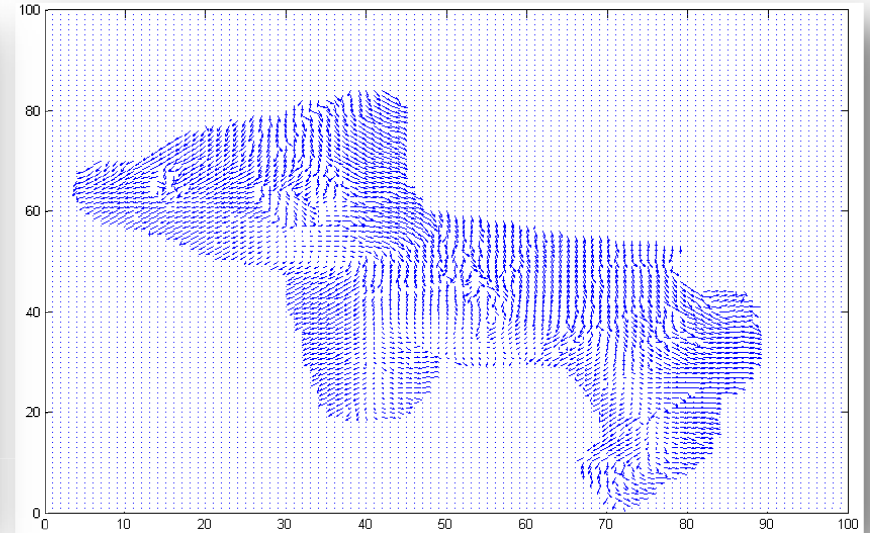
Imaging Setup

Photometric Stereo

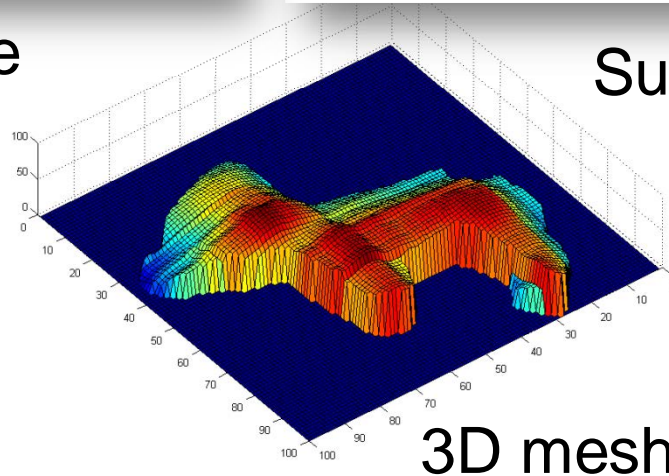
Christopher Bireley



Albedo image

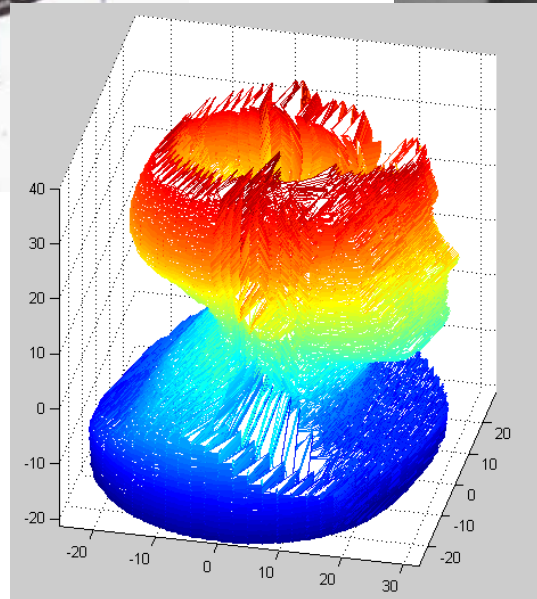
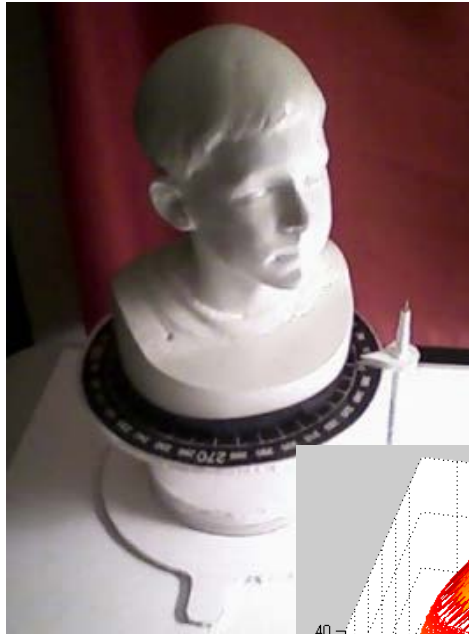


Surface Normals



3D mesh

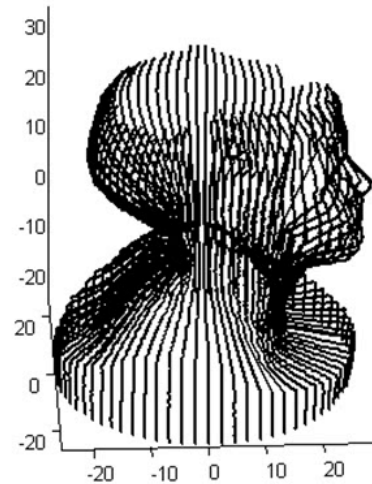
Structured Light James Clark



Result

**Positioned
camera with
object in view**

Structured Light ctd. James Clark



Webcam Based Virtual Whiteboard

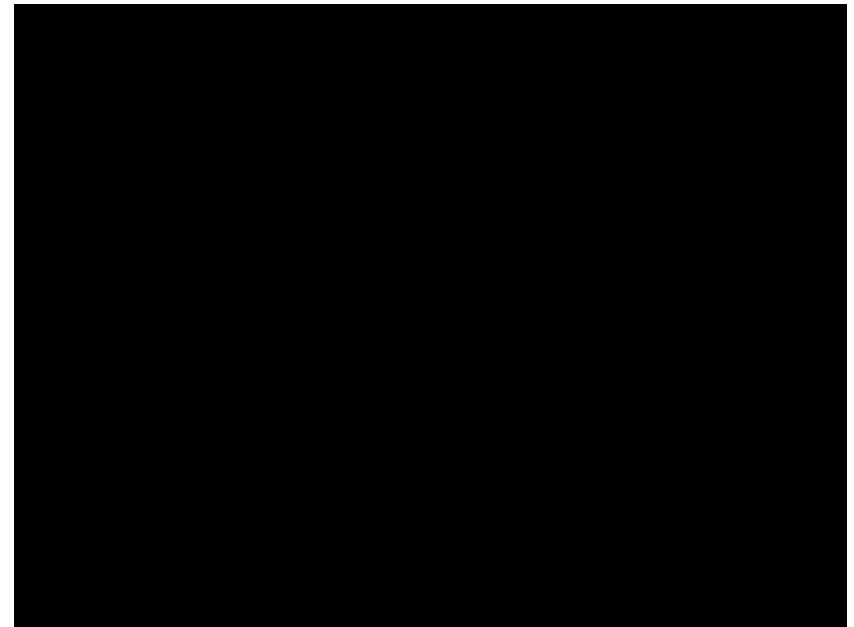
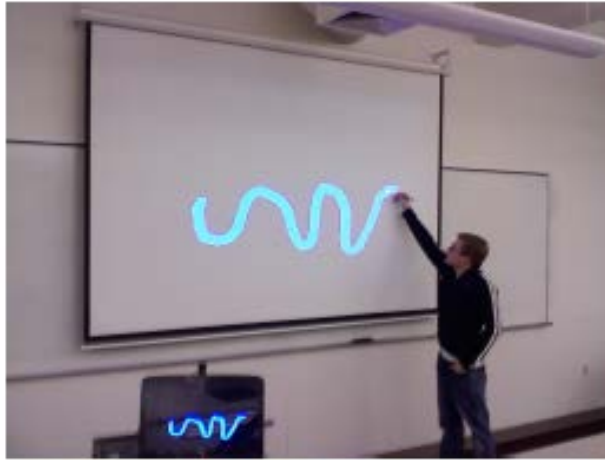
Jon Bronson James Fishbaugh

- 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



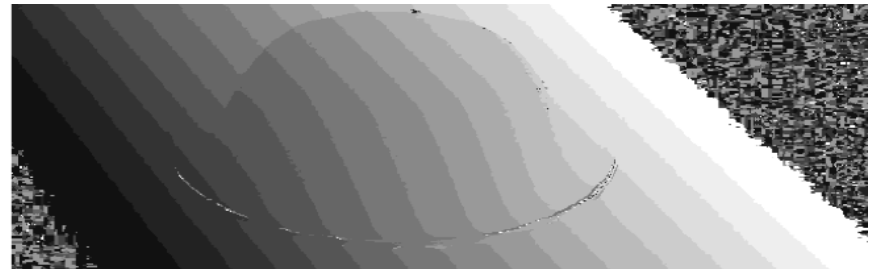
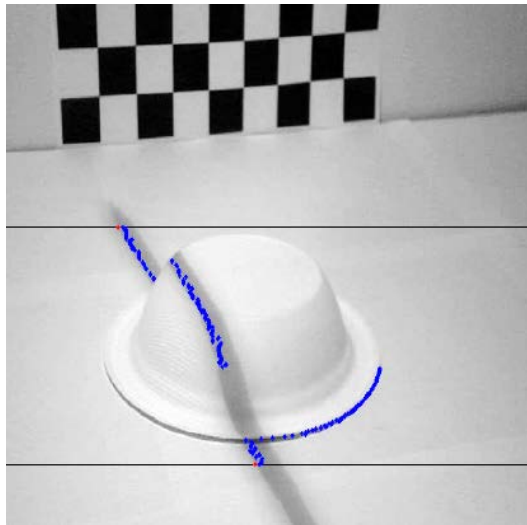
Webcam Based Virtual Whiteboard

Jon Bronson James Fishbaugh



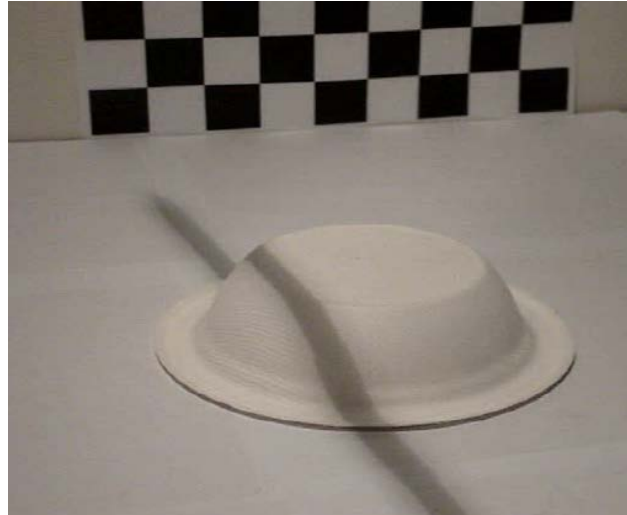
Structured Light

Anuja Sharma, Abishek Kumar

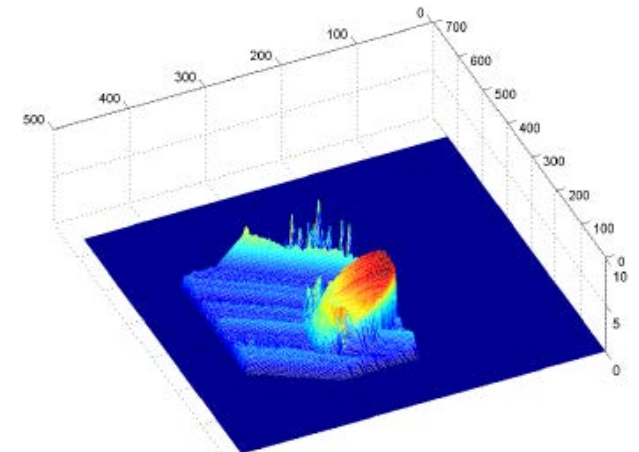


Structured Light

Anuja Sharma, Abishek Kumar

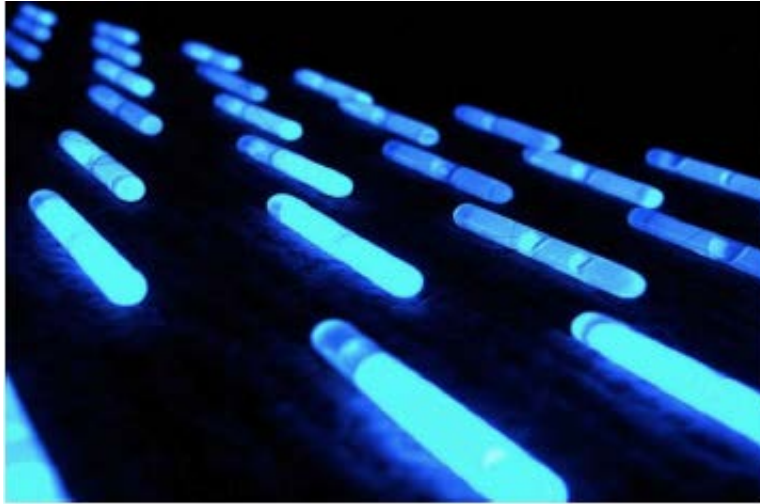


3D plot 1

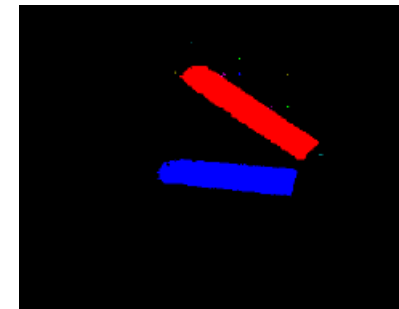


Realtime Glowstick Detection

Andrei Ostanin



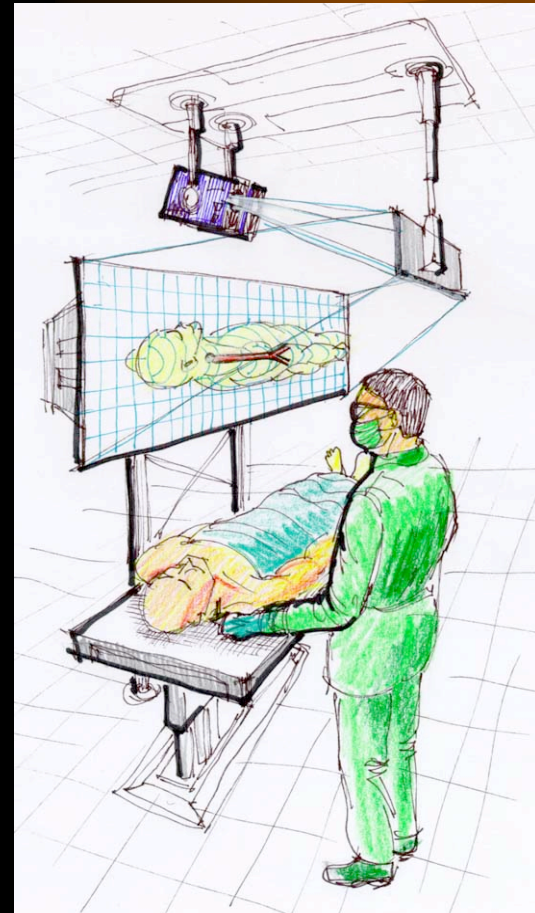
- ▶ Capture the 3D position of glowsticks in real-time using two webcams
- ▶ Environment dark enough that glowsticks are easily segmented out
- ▶ Prefer speed over correctness



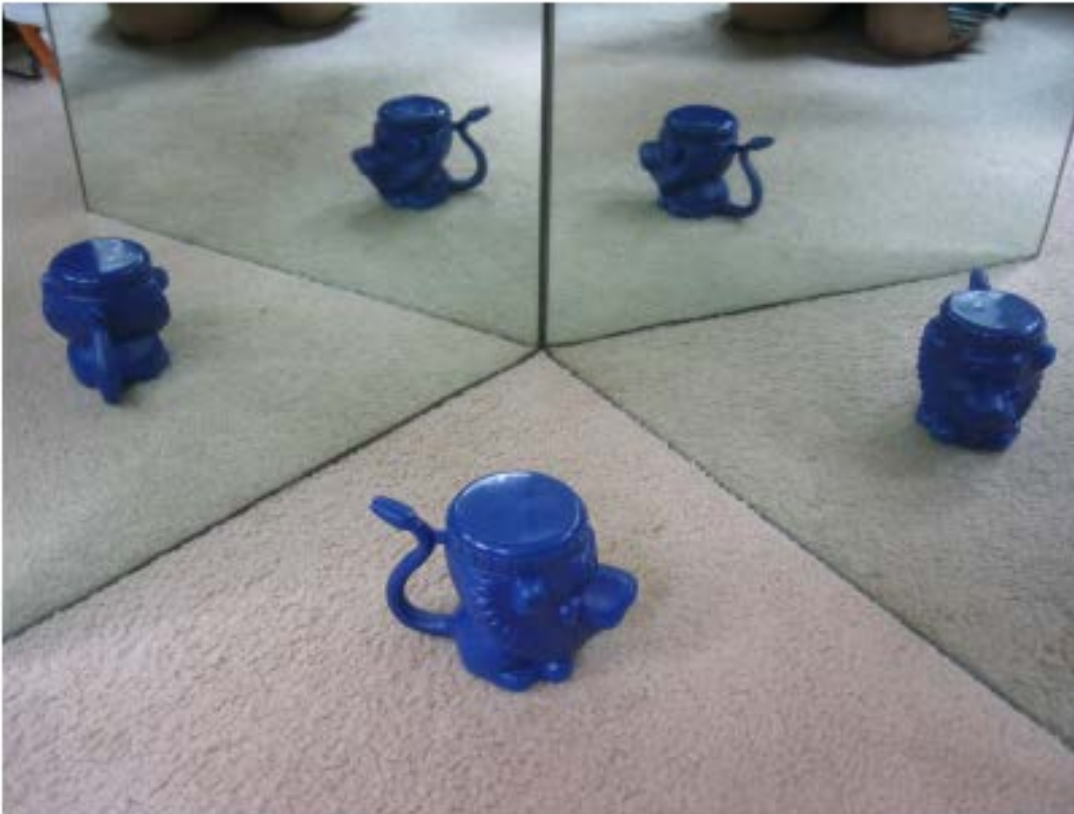
[movie](#)

Passive Object Tracking from Stereo Vision

- 3D visualization -
blood vessels, organs,
teapots
- Stereo displays require
head tracking
- Cumbersome trackers
are undesirable
- Michael Rosenthal



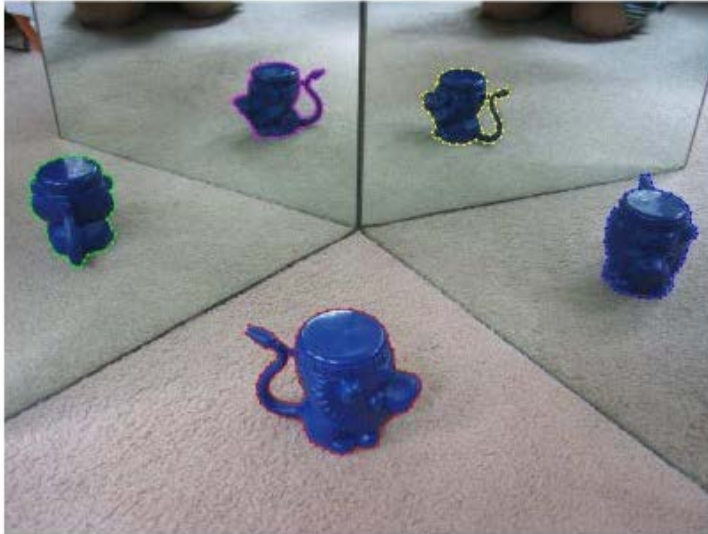
3D shape from silhouettes: Two Mirrors and uncalibrated camera



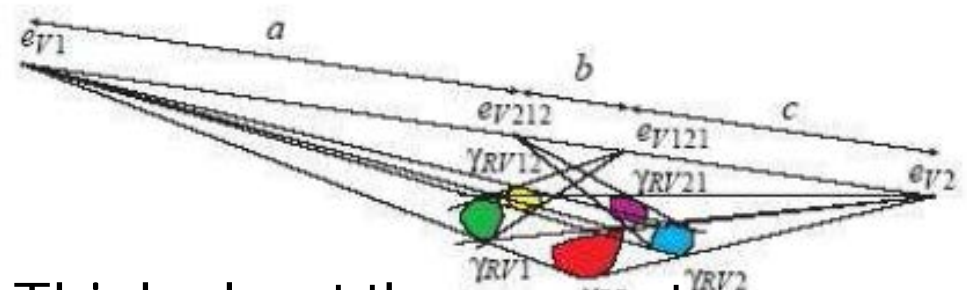
Forbes et al.,
ICCV2005

Christine Xu,
Computer Vision
Student Project

3D shape from silhouettes

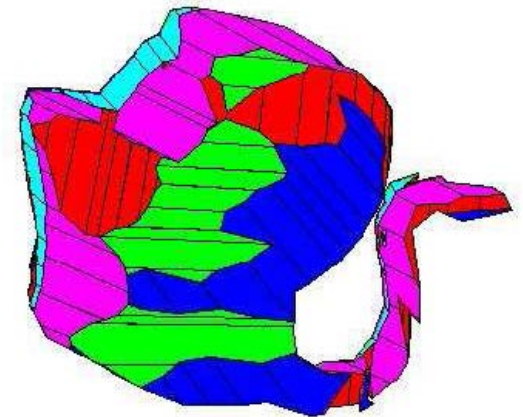


Segmentation of contours



Think about the geometry -> calculate relationship between silhouettes

Result: 3D Object



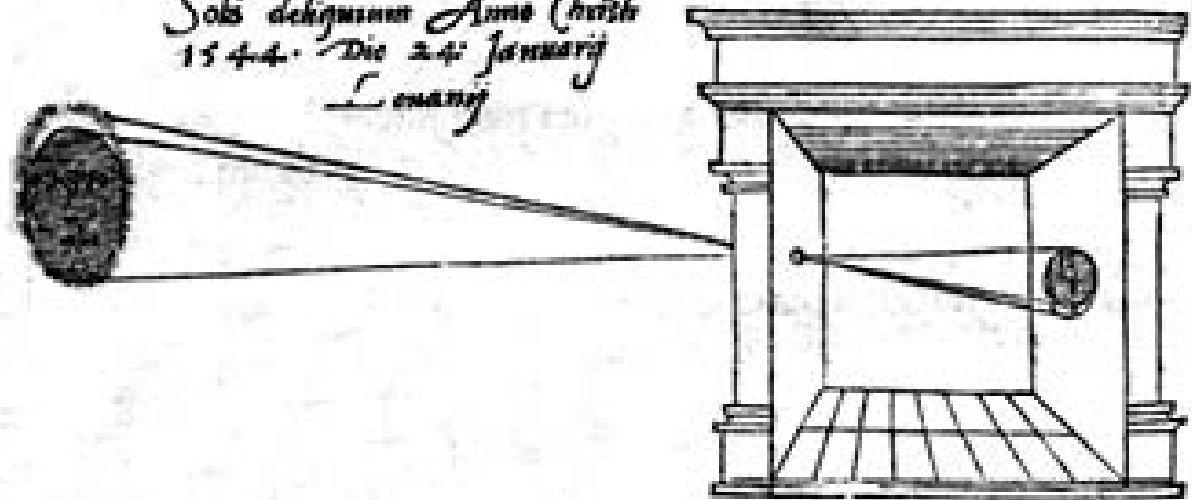


Next class: Cameras

Chapter 2: Image Formation

illum in tabula per radios Solis, quam in cœlo contingit: hoc est, si in cœlo superior pars deliquiū patiatur, in radiis apparebit inferior deficere, vt ratio exigit optica.

*Solis deliquium Anno Christi
1544. Die 24. Januarij
Louanij*



Sic nos exactè Anno .1544. Louanii eclipsim Solis obseruauimus, inuenimusq; deficere paulò plus q̄ dex-



Next class: Image Formation

Chapter 2: Textbook

- Please find pdf copies of Chapters 1&2 on the website.

Assignment:

- Read Chapter 1 for additional materials
- Read Chapter 2 for preparation of 2nd lecture