

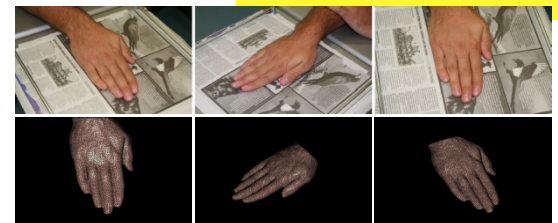
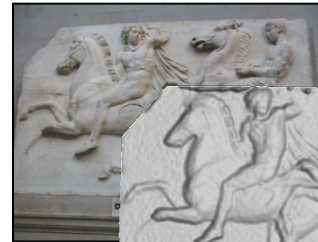
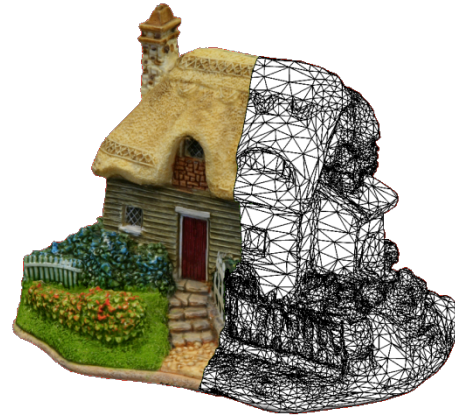
Reconstruction

CSE 576

Ali Farhadi

3d model

- “Digital copy” of real object
- Allows us to
 - Inspect details of object
 - Measure properties
 - Reproduce in different material
- Many applications
 - Cultural heritage preservation
 - Computer games and movies
 - City modelling
 - E-commerce



Applications: cultural heritage

SCULPTEUR European project



Applications: art



Block Works Precipitate III 2004
Mild steel blocks 80 x 46 x 66 cm



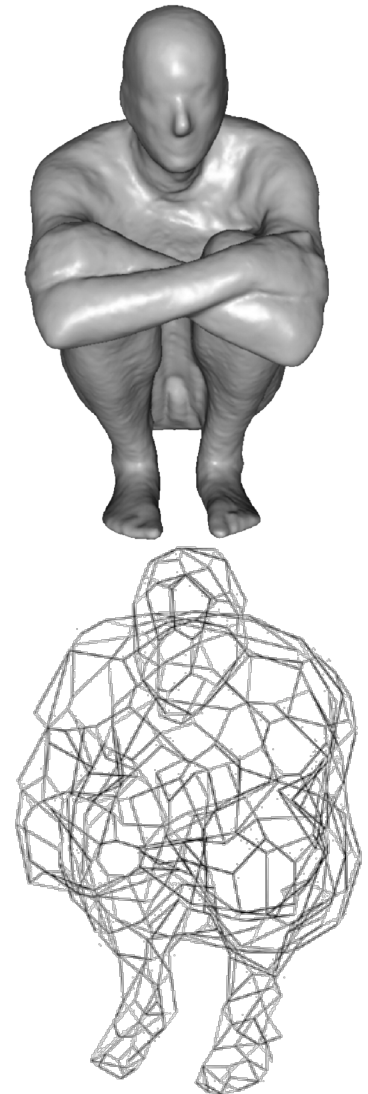
Domain Series Domain VIII Crouching
1999 *Mild steel bar 81 x 59 x 63 cm*



Applications: structure engineering



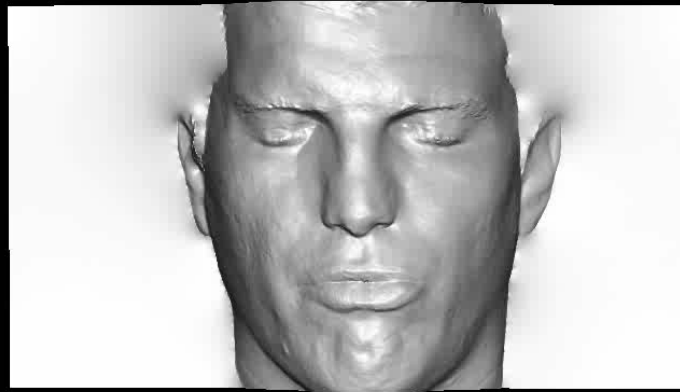
BODY / SPACE / FRAME, Antony Gormley, Lelystad, Holland



Applications: art



Applications: computer games



RECORDING
Frame rate: 5145.217578

Applications: 3D indexation

The image displays a 3D indexing application interface. It features a large grid of small 3D model thumbnails. A large central window shows a detailed 3D model of a female torso sculpture. To the right, a photo of a museum gallery is overlaid, showing various artifacts on display. Below the photo, a table lists the top six search results for the query, including their IDs and similarity scores.

Index	Model ID	Similarity Score
0	deesse0	0.000000
1	deesse5	0.092800
5	deesse2	0.211000
6	ARCHI3203	0.236800

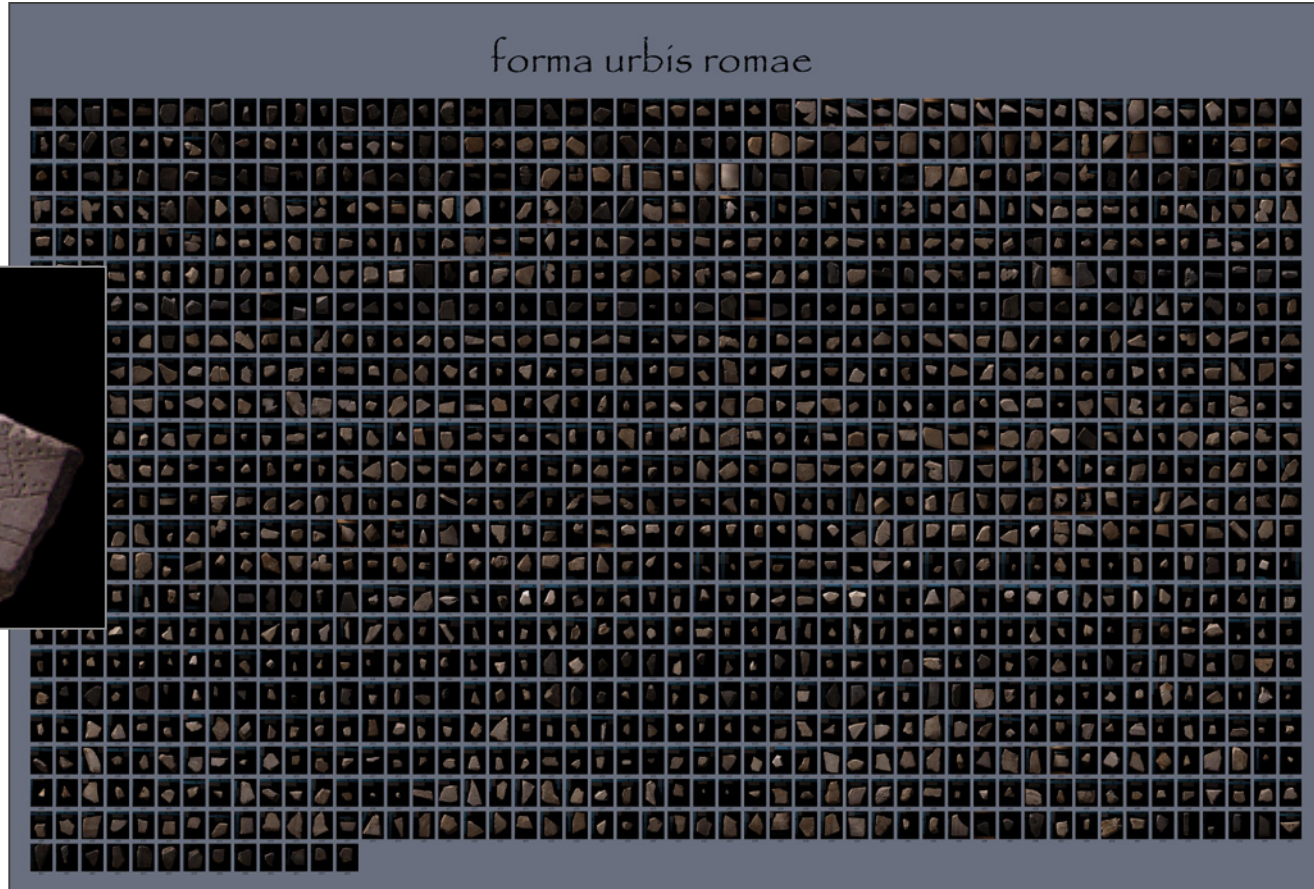
Applications: archaeology

- “forma urbis romae” project

Fragments of the City: Stanford's Digital Forma Urbis Romae Project

David Koller, Jennifer Trimble, Tina Najbjerg, Natasha Gelfand, Marc Levoy

*Proc. Third Williams Symposium
on Classical Architecture,
Journal of Roman Archaeology
supplement, 2006.*

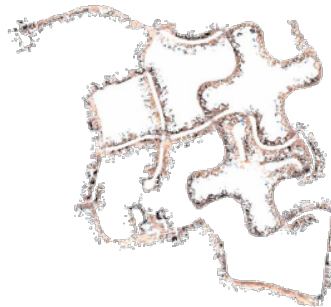


1186 fragments

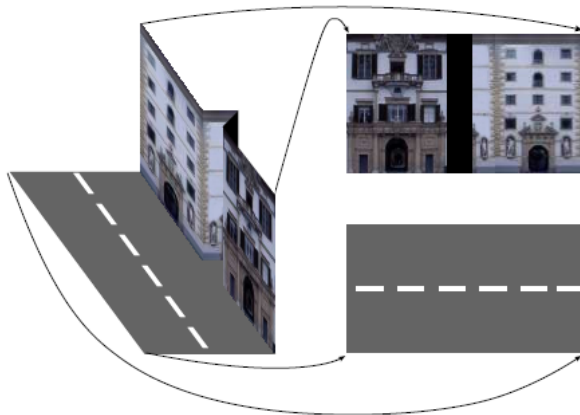
Applications: large scale modelling



[Furukawa10]



[Pollefeys08]



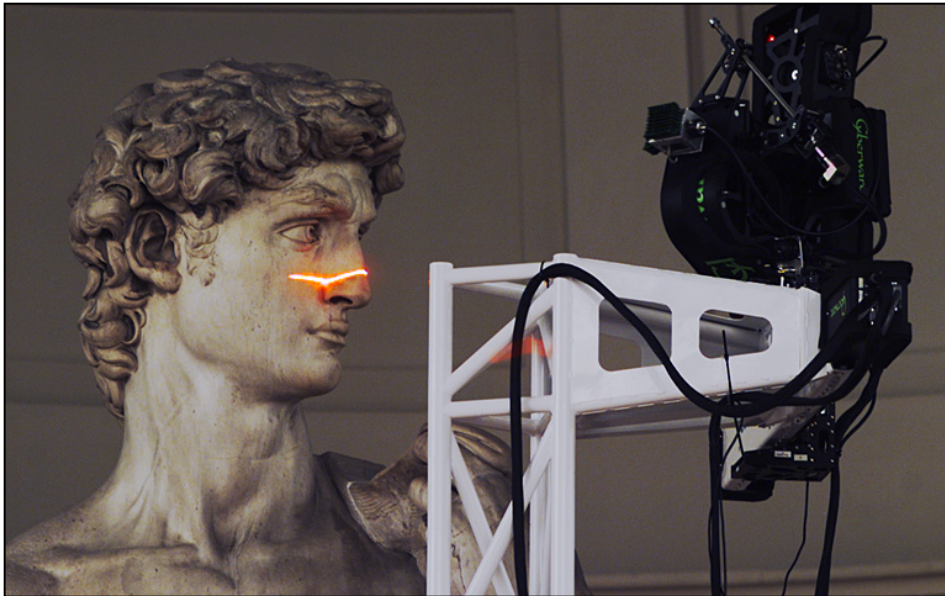
[Cornelis08]



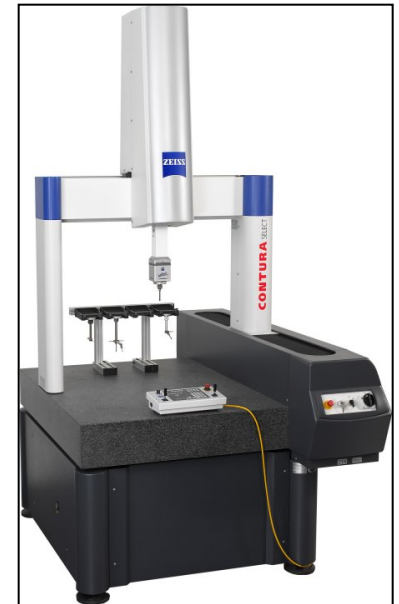
[Goesele07]

Scanning technologies

- Laser scanner, coordinate measuring machine
 - Very accurate
 - Very Expensive
 - Complicated to use

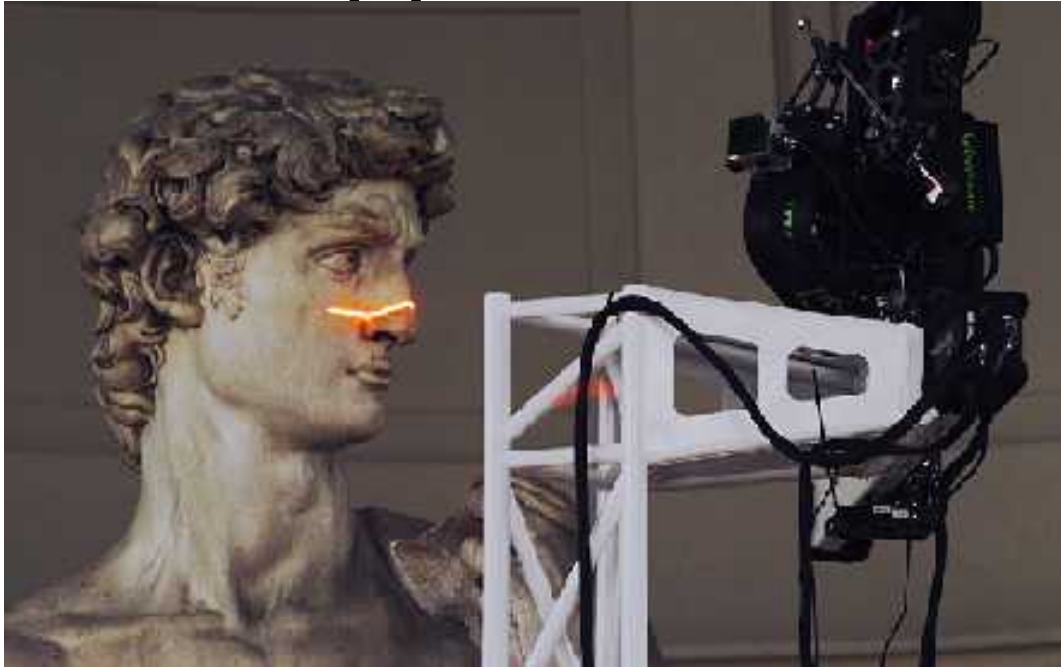


Minolta



Contura CMM

Applications: 3D Scanning



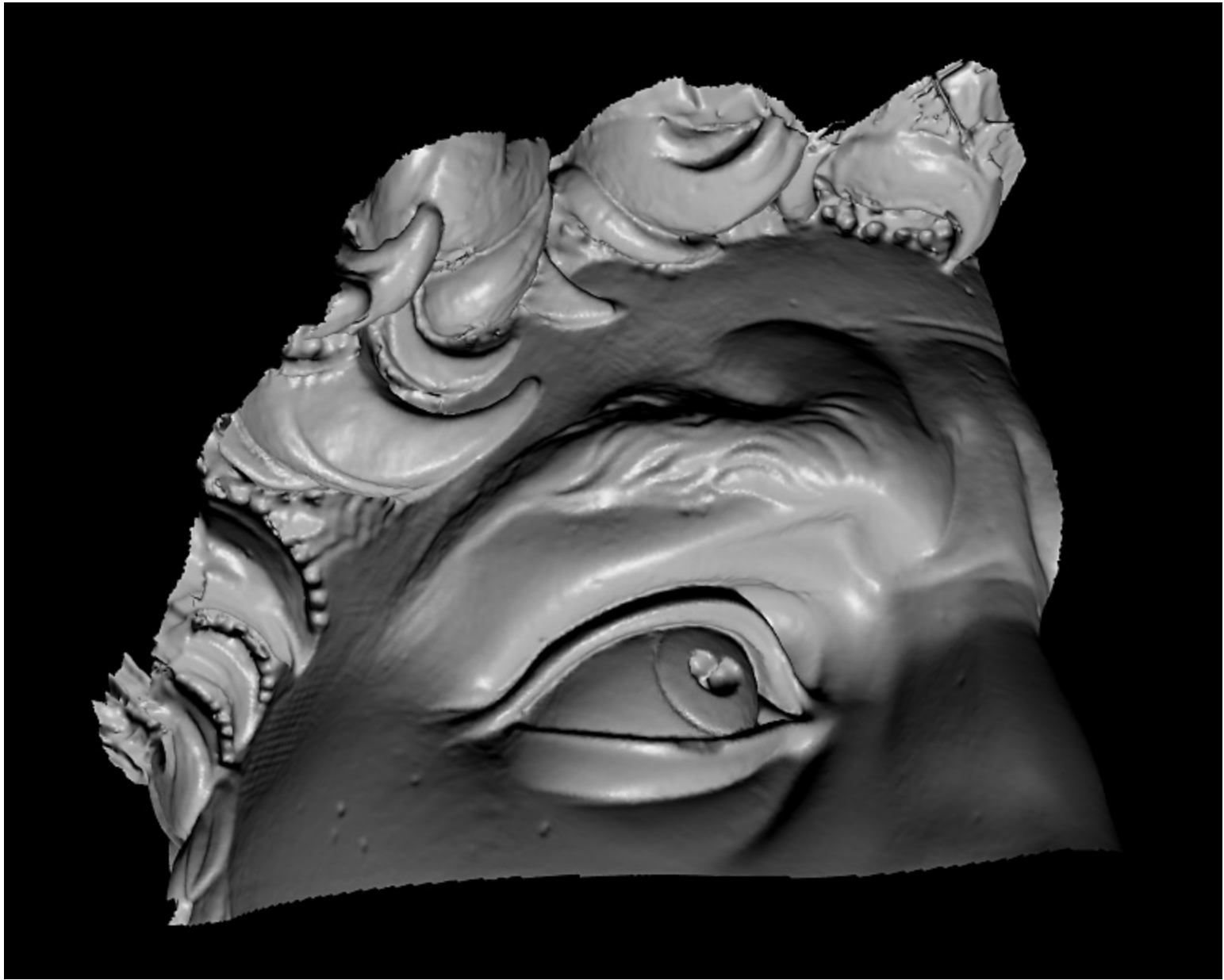
Scanning Michelangelo's "*The David*"

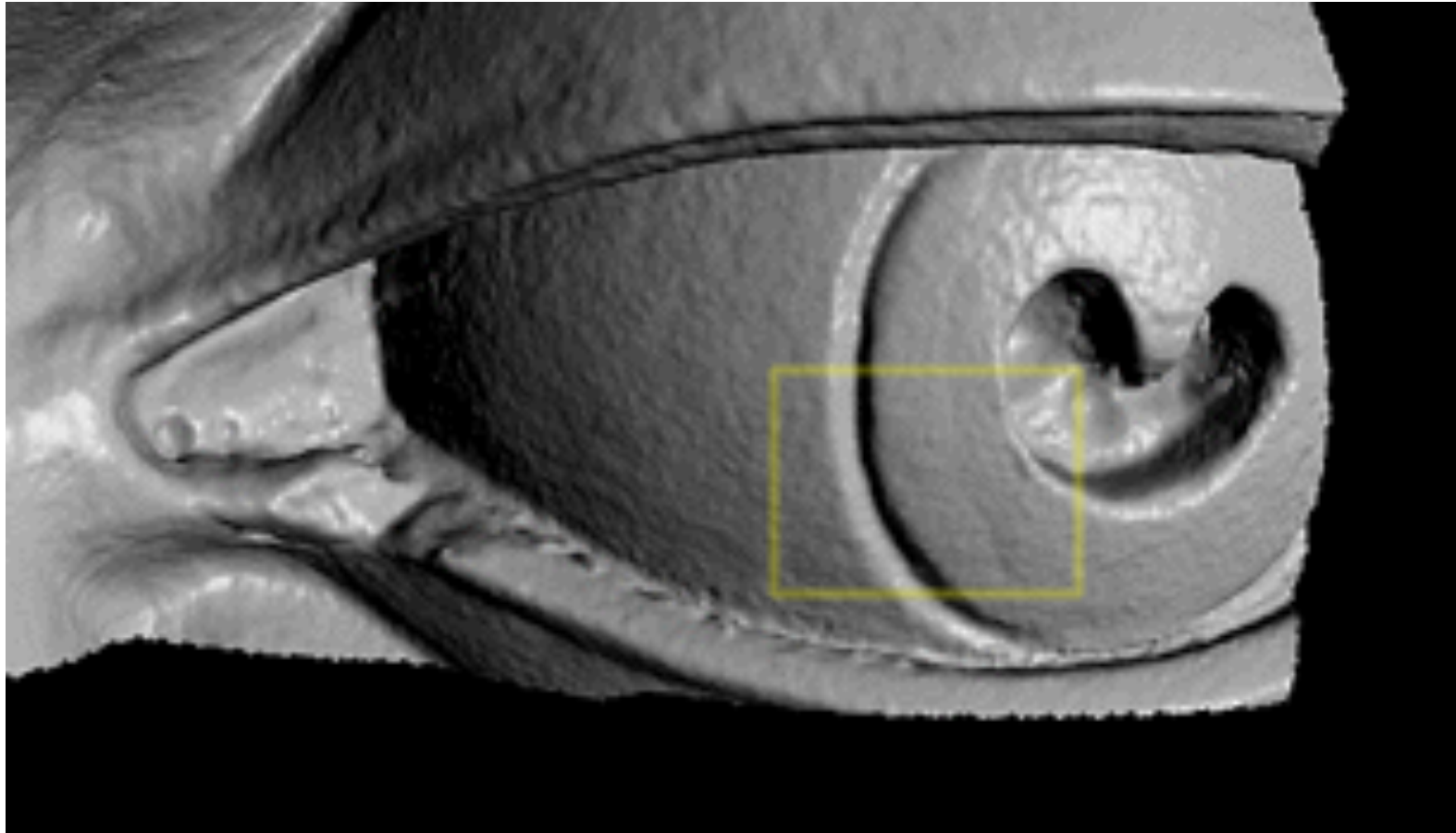
- [The Digital Michelangelo Project](http://graphics.stanford.edu/projects/mich/)
 - <http://graphics.stanford.edu/projects/mich/>
 - 2 BILLION polygons, accuracy to .29mm

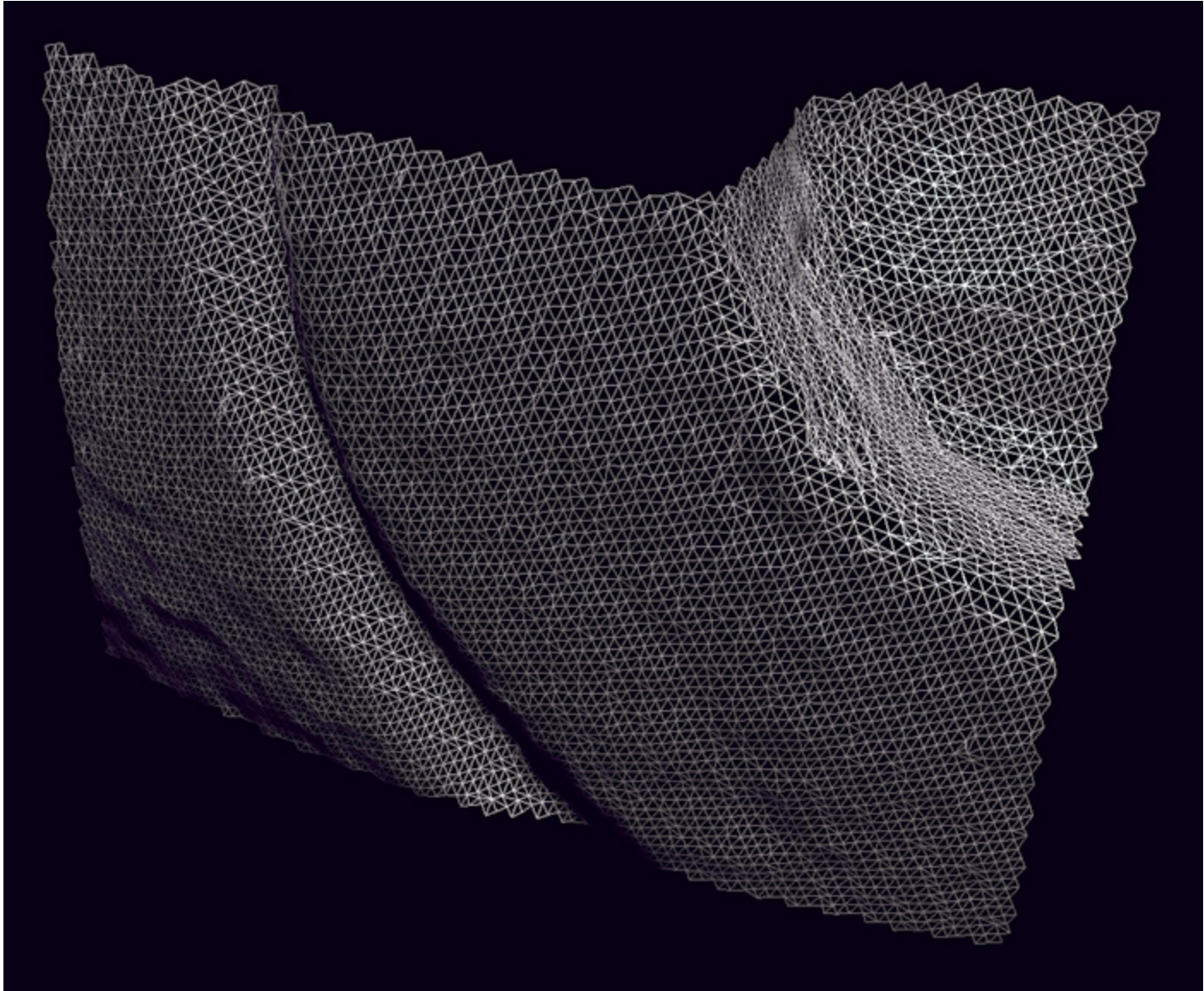


The Digital Michelangelo Project, Levoy et al.



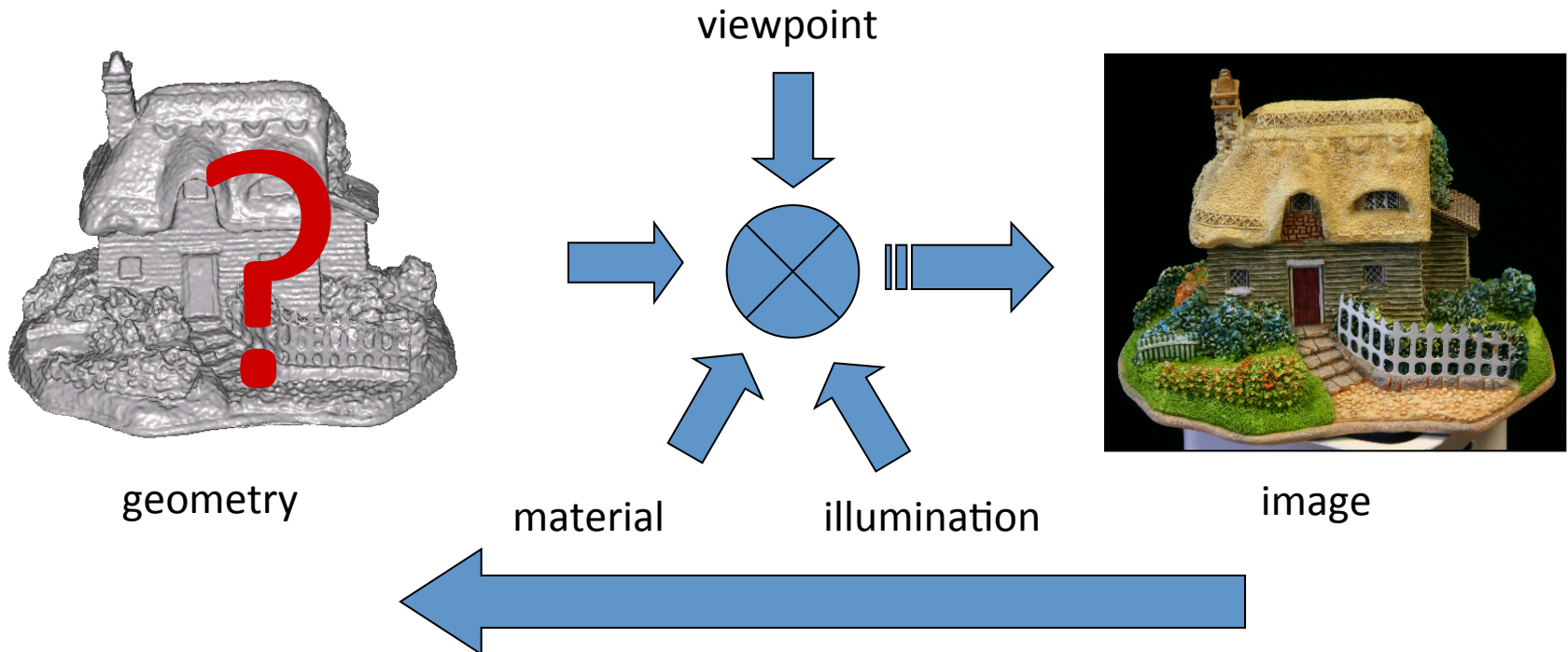






3d shape from photographs

“Estimate a 3d shape that would generate the input photographs given the same material, viewpoints and illumination”



3D shape from photographs

“Estimate a 3d shape that would generate the input photographs given the same material, viewpoints and illumination”

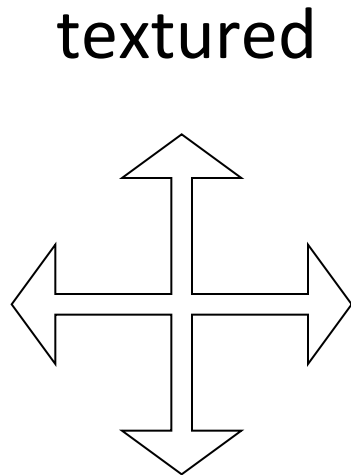


3d shape from photographs

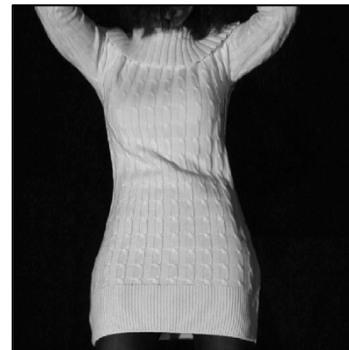
Appearance strongly depends on the material and lighting

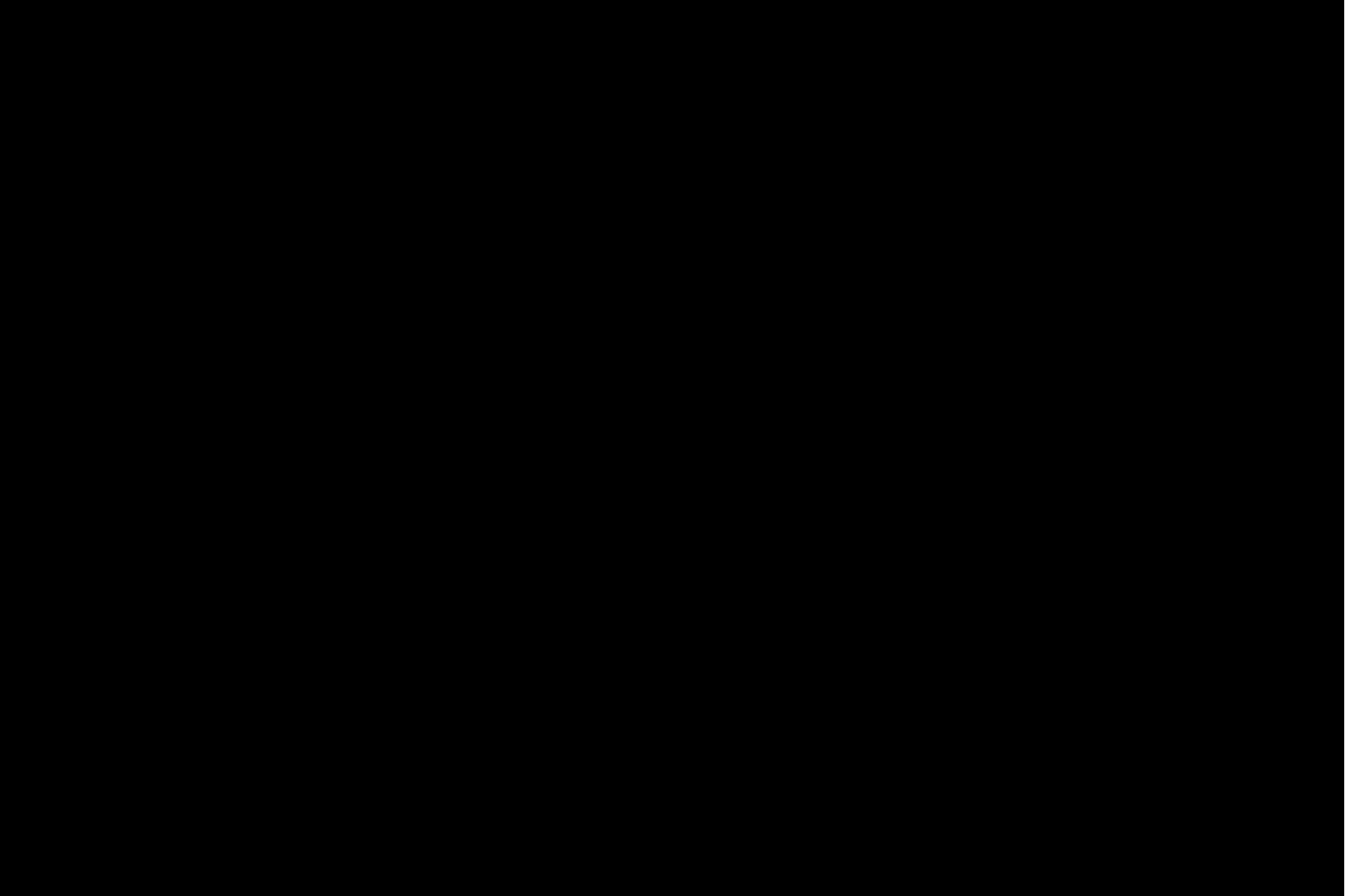


rigid



deforming





3d shape from photographs

Appearance strongly depends on the material and lighting



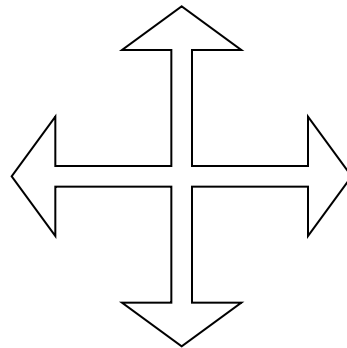
No single algorithm exists dealing with **any** type of scene



rigid



textured



textureless



deforming



3d shape from photographs

Photograph based 3d reconstruction is:

- ✓ practical
- ✓ fast
- ✓ non-intrusive
- ✓ low cost
- ✓ Easily deployable outdoors
- ✗ “low” accuracy
- ✗ Results depend on material properties

Reconstruction

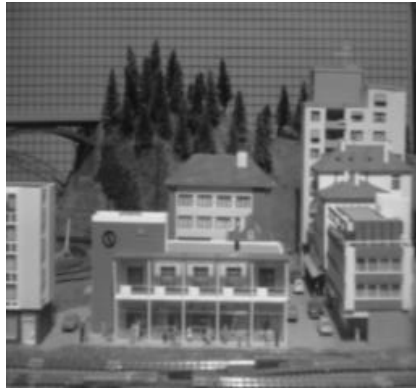
- Generic problem formulation: given several images of the same object or scene, compute a representation of its 3D shape



Reconstruction

- Generic problem formulation: given several images of the same object or scene, compute a representation of its 3D shape
- “Images of the same object or scene”
 - Arbitrary number of images (from two to thousands)
 - Arbitrary camera positions (camera network or video sequence)
 - Calibration may be initially unknown
- “Representation of 3D shape”
 - Depth maps
 - Meshes
 - Point clouds
 - Patch clouds
 - Volumetric models
 - Layered models

Multiple-baseline stereo



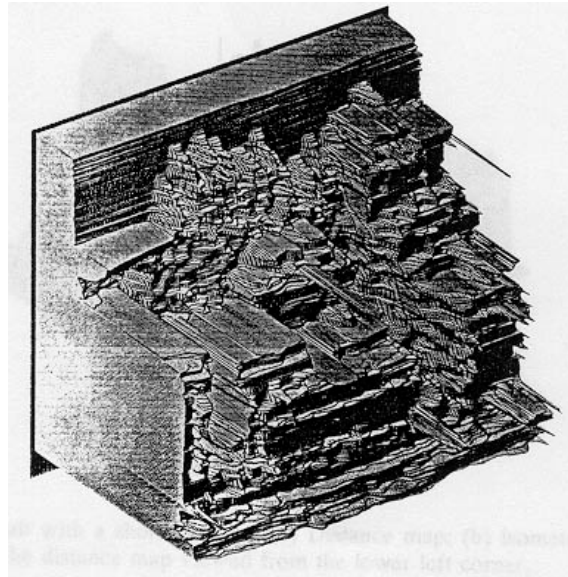
I1



I2



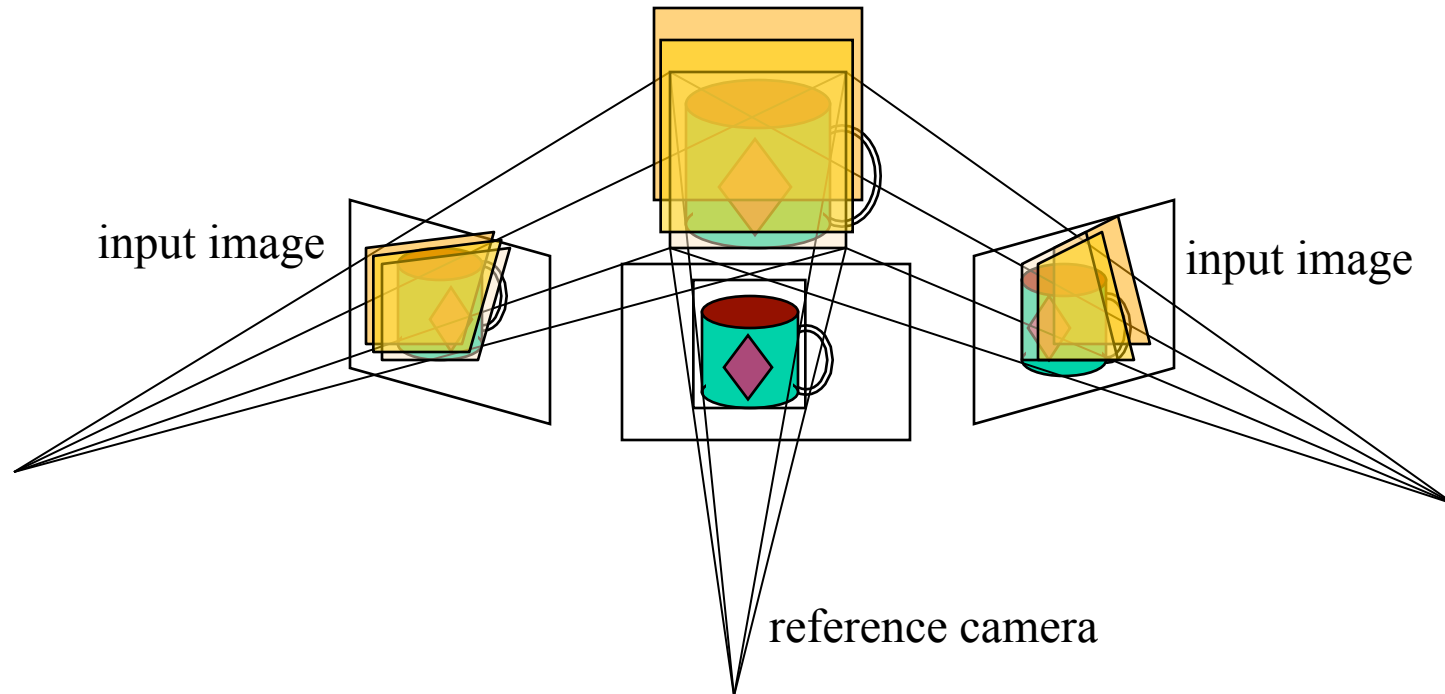
I10



M. Okutomi and T. Kanade, [“A Multiple-Baseline Stereo System,”](#) IEEE Trans. on Pattern Analysis and Machine Intelligence, 15(4):353-363 (1993).

Plane Sweep Stereo

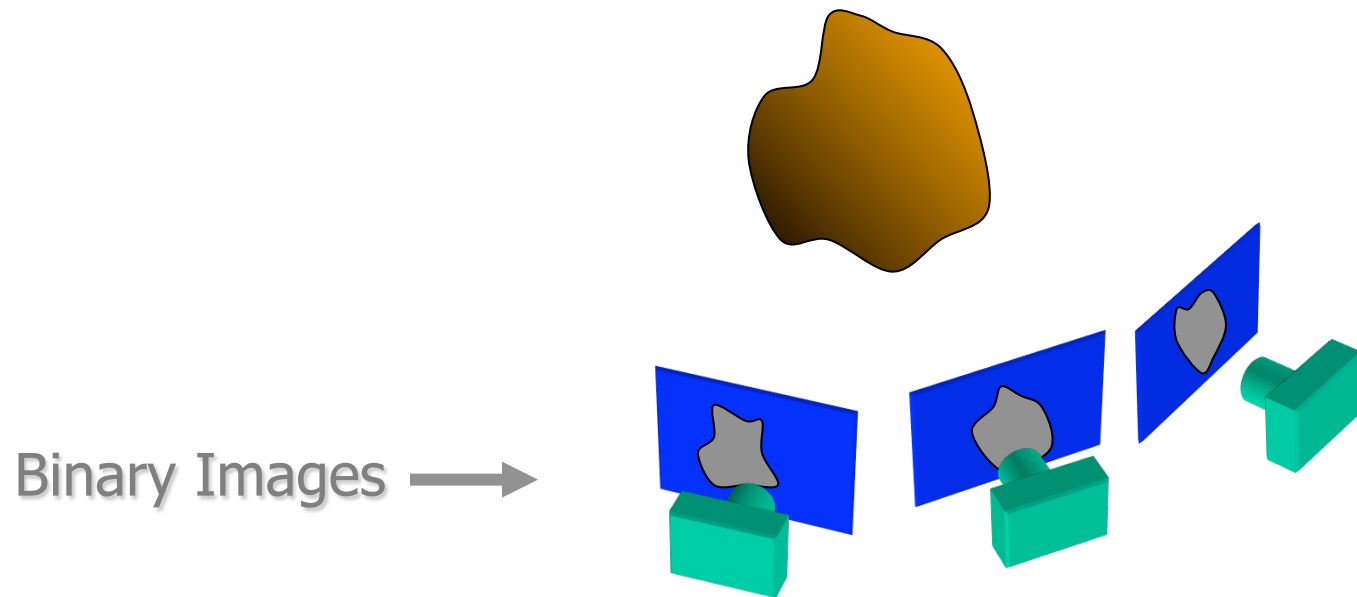
- Choose a reference view
- Sweep family of planes at different depths with respect to the reference camera



Each plane defines a homography warping each input image into the reference view

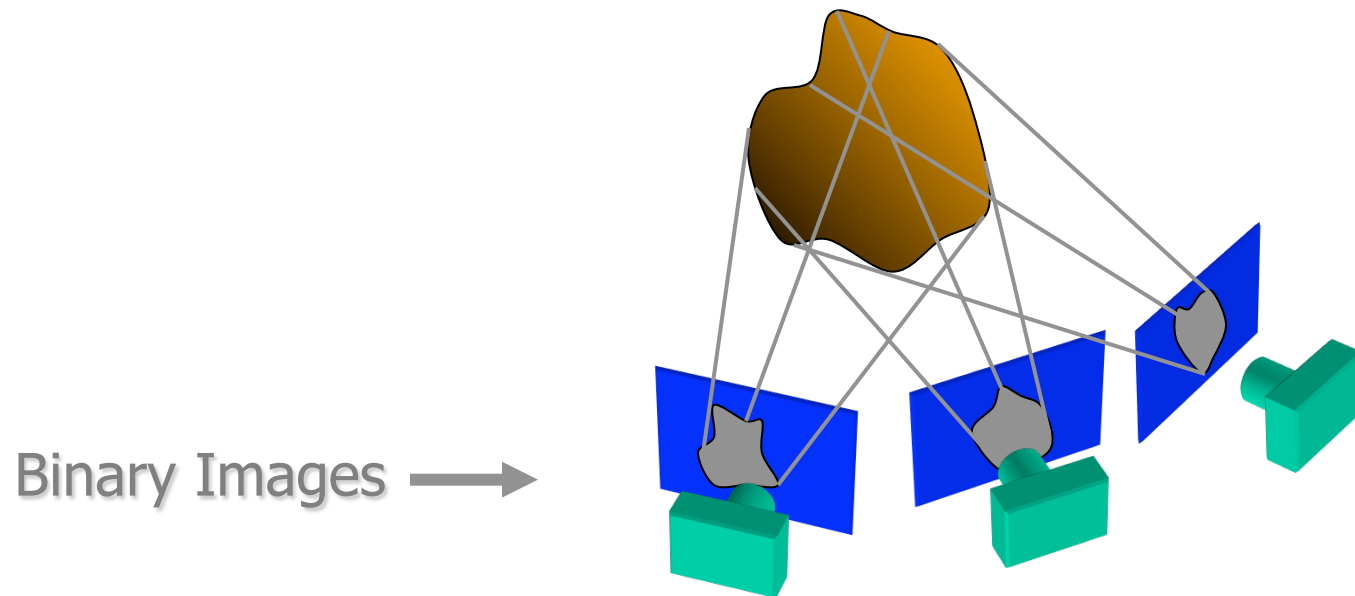
Reconstruction from Silhouettes

- The case of binary images: a voxel is photo-consistent if it lies inside the object's silhouette in all views



Reconstruction from Silhouettes

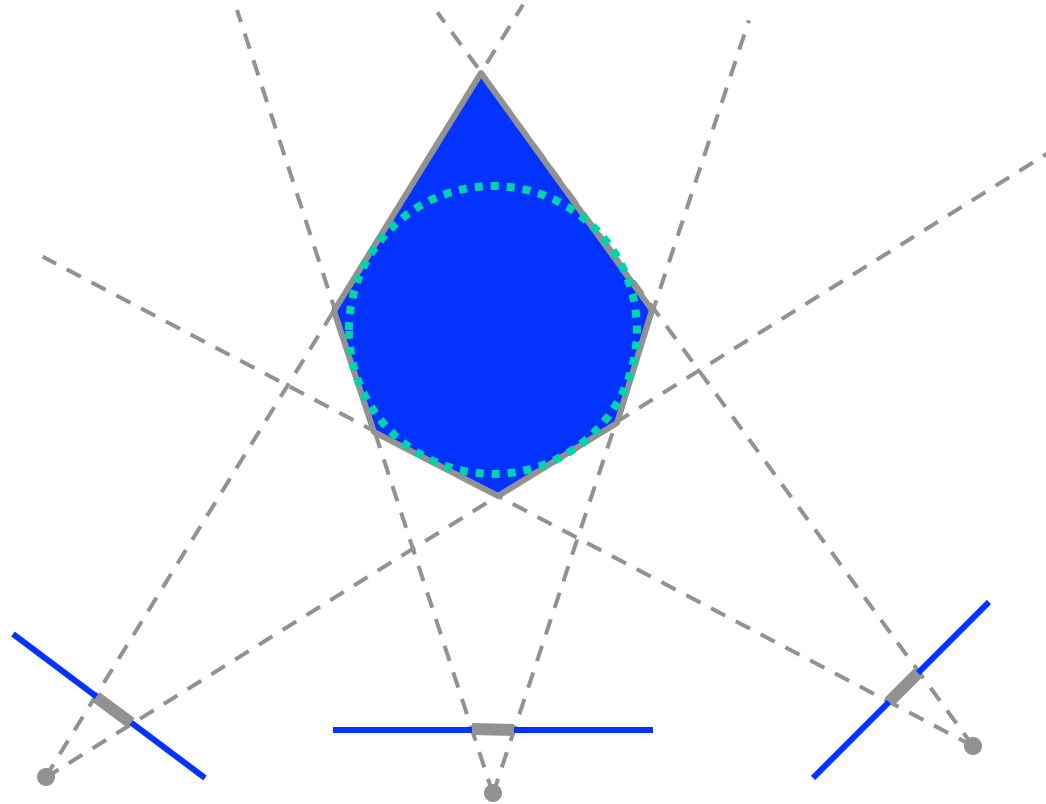
- The case of binary images: a voxel is photo-consistent if it lies inside the object's silhouette in all views



Finding the silhouette-consistent shape (*visual hull*):

- *Backproject* each silhouette
- Intersect backprojected volumes

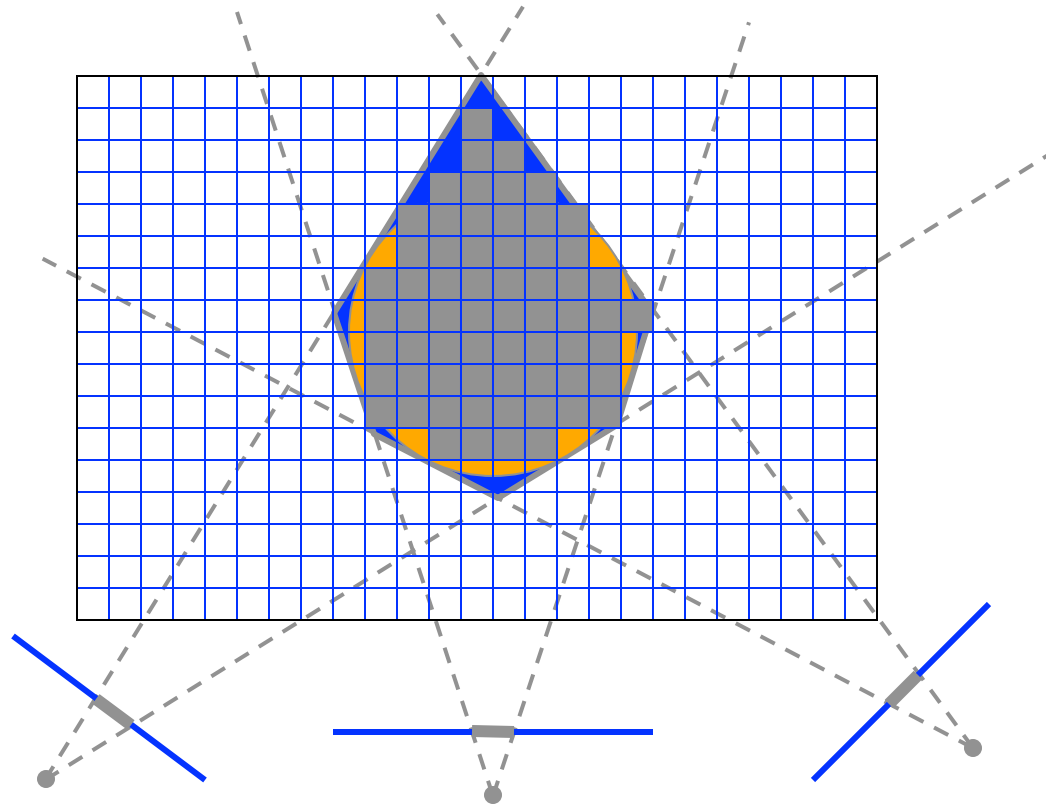
Volume intersection



Reconstruction Contains the True Scene

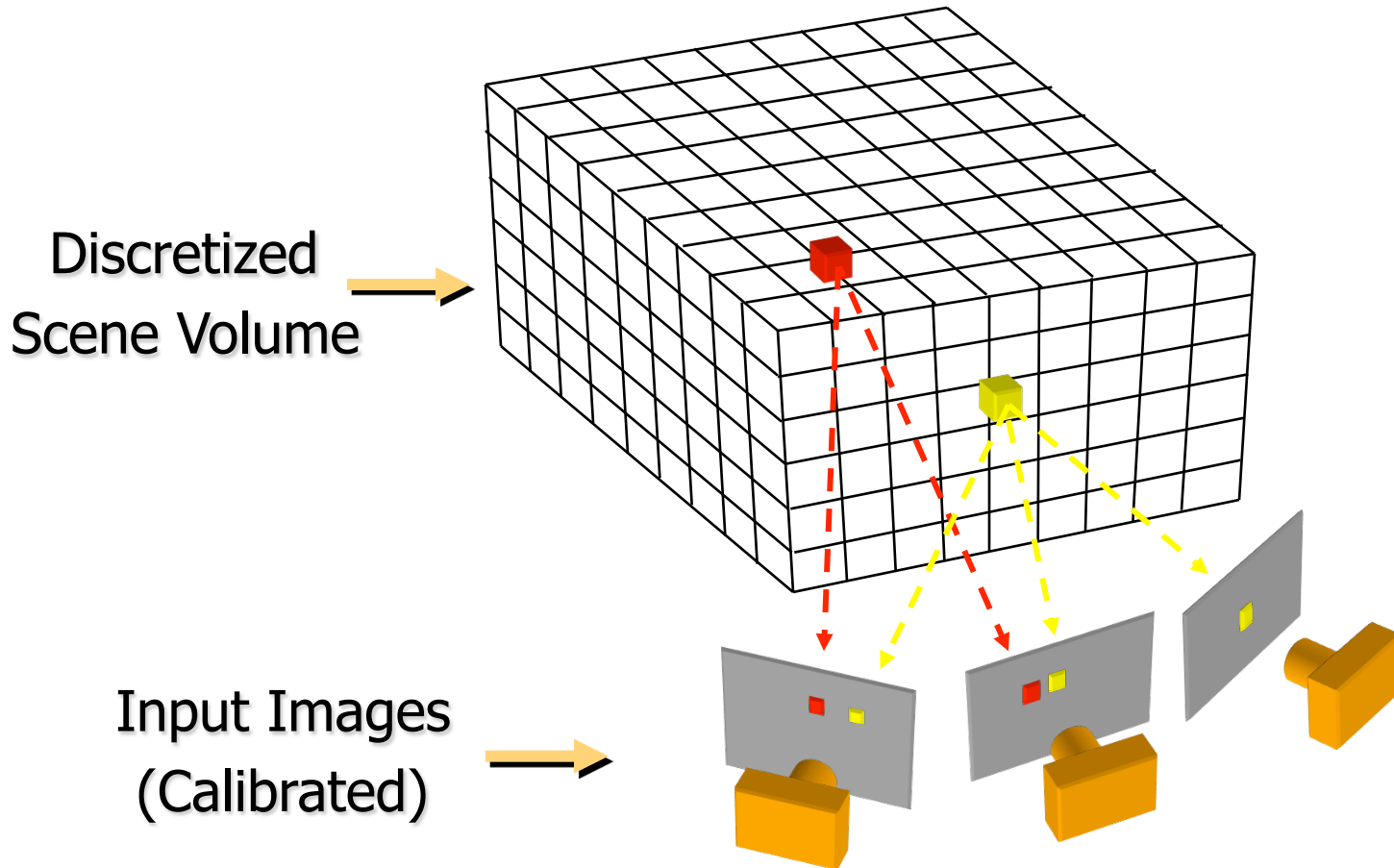
- But is generally not the same
- In the limit (all views) get *visual hull*

Voxel algorithm for volume intersection



Color voxel black if on silhouette in every image

Volumetric Stereo / Voxel Coloring



Goal: Assign RGB values to voxels in V
photo-consistent with images

Photo-consistency of a 3d point

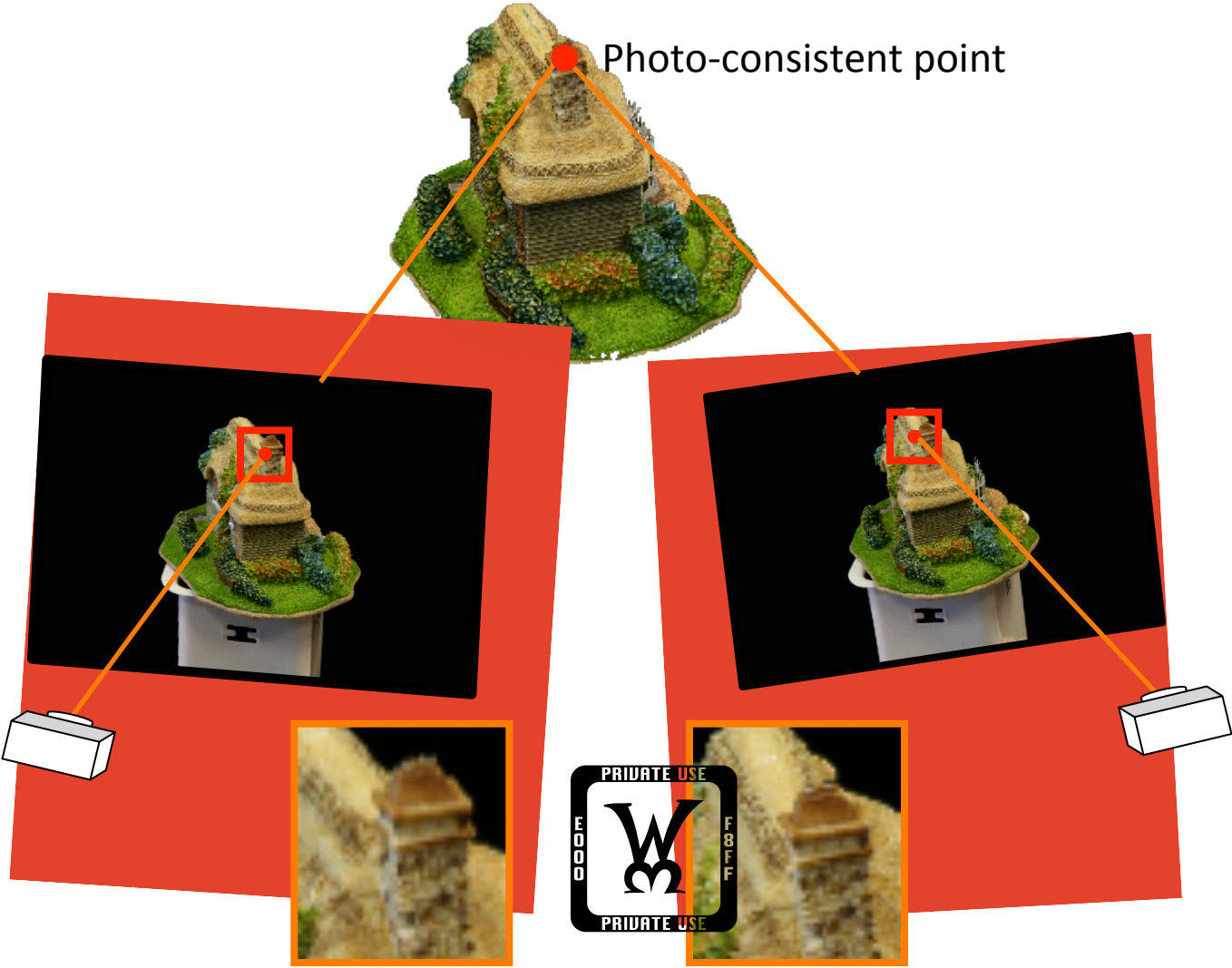


Photo-consistency of a 3d point

Non photo-consistent point

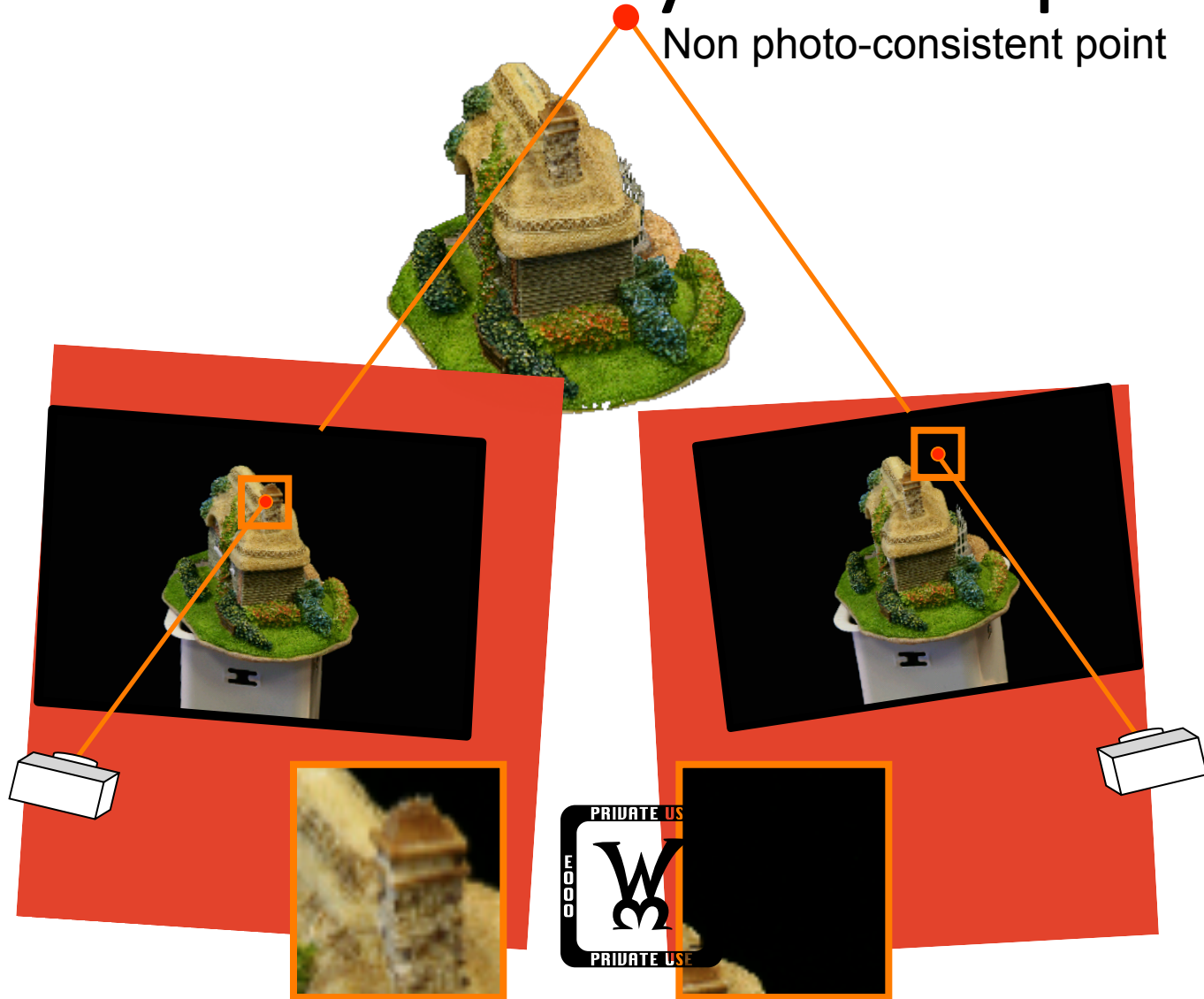


Photo-consistency of a 3d patch



Measuring photo-consistency

- Equivalent statements
 - voxel v is photo-consistent
 - image content is (nearly) identical for all projections of v
 - any pair of projections of v matches well
- Examples:

$$\Phi(v) = f \left(\frac{1}{K} \sum_{j=1}^K (c_j - c_{mean})^2 \right)$$

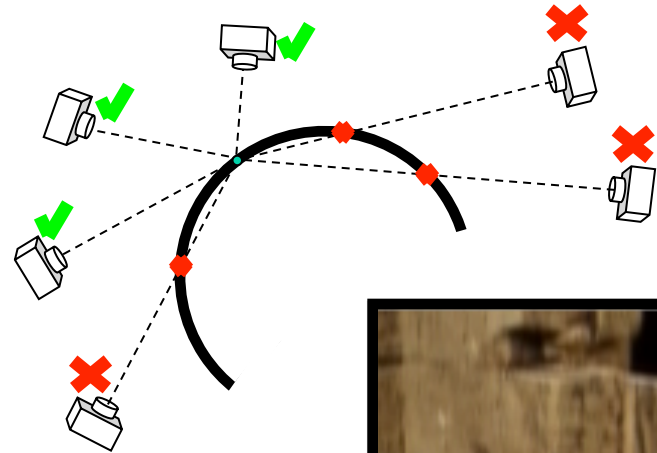
variance of average colour c_j over all K visible images [Seitz&Kutulakos]

$$\Phi(v) = \frac{2}{K(K+1)} \sum_{i=1}^K \sum_{j=i+1}^K \text{NCC}(p_i, p_j)$$

average normalised cross-correlation over all pairs of visible images [Vogiatzis et al.]

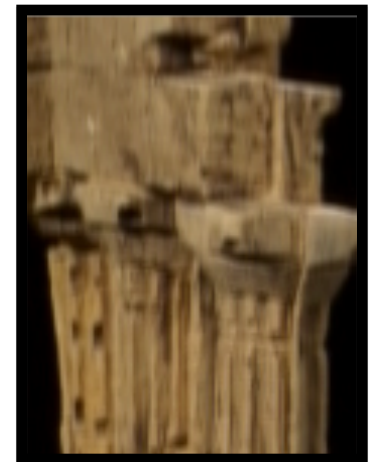
Challenges of photo-consistency

- Camera visibility



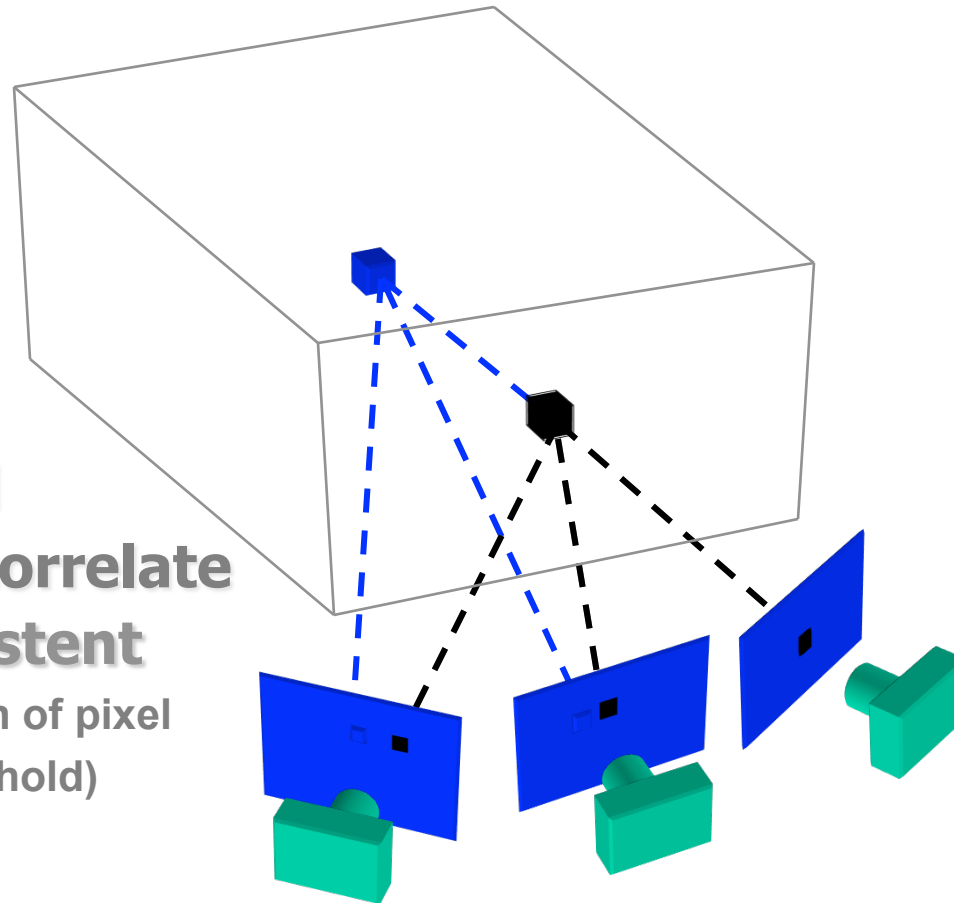
- Failure of comparison metric

- repeated texture
- lack of texture
- specularities



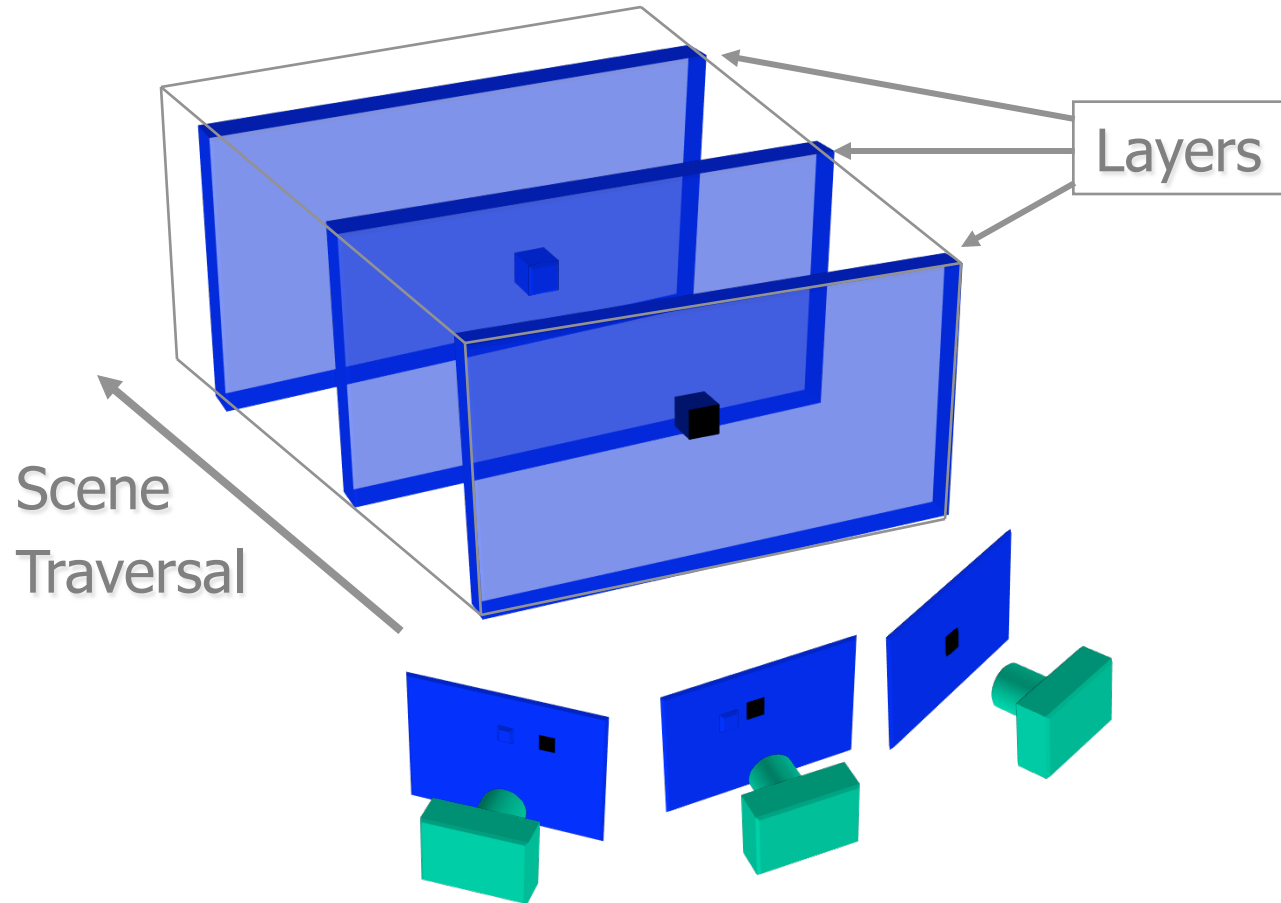
Voxel Coloring Approach

- 1. Choose voxel**
- 2. Project and correlate**
- 3. Color if consistent**
(standard deviation of pixel colors below threshold)



Visibility Problem: in which images is each voxel visible?

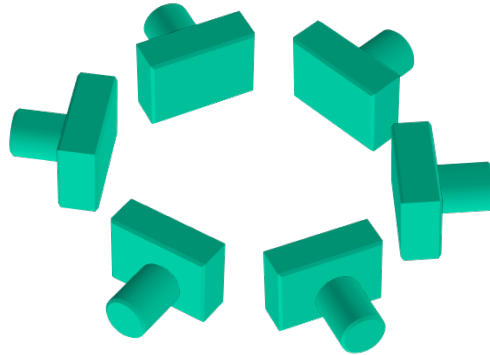
Depth Ordering: visit occluders first!



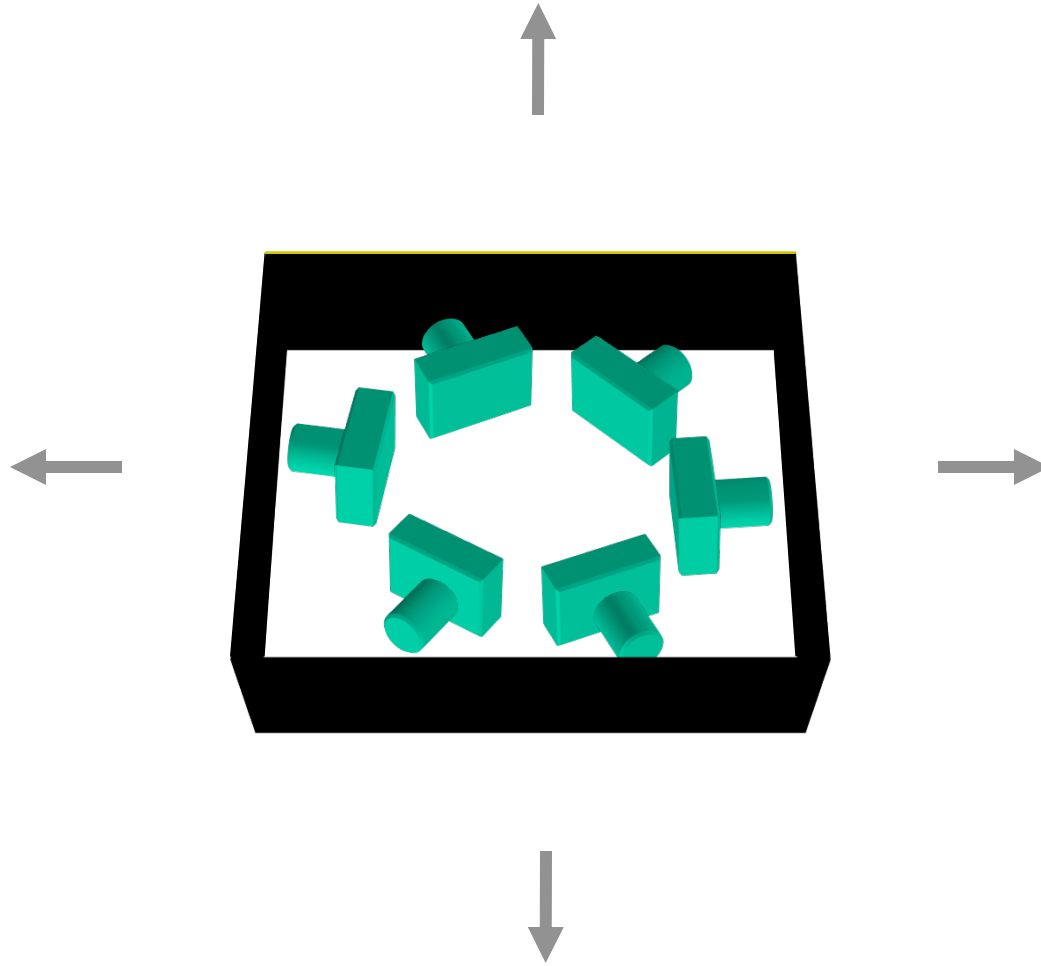
Condition: depth order is the *same for all input views*

Panoramic Depth Ordering

- Cameras oriented in many different directions
- Planar depth ordering does not apply

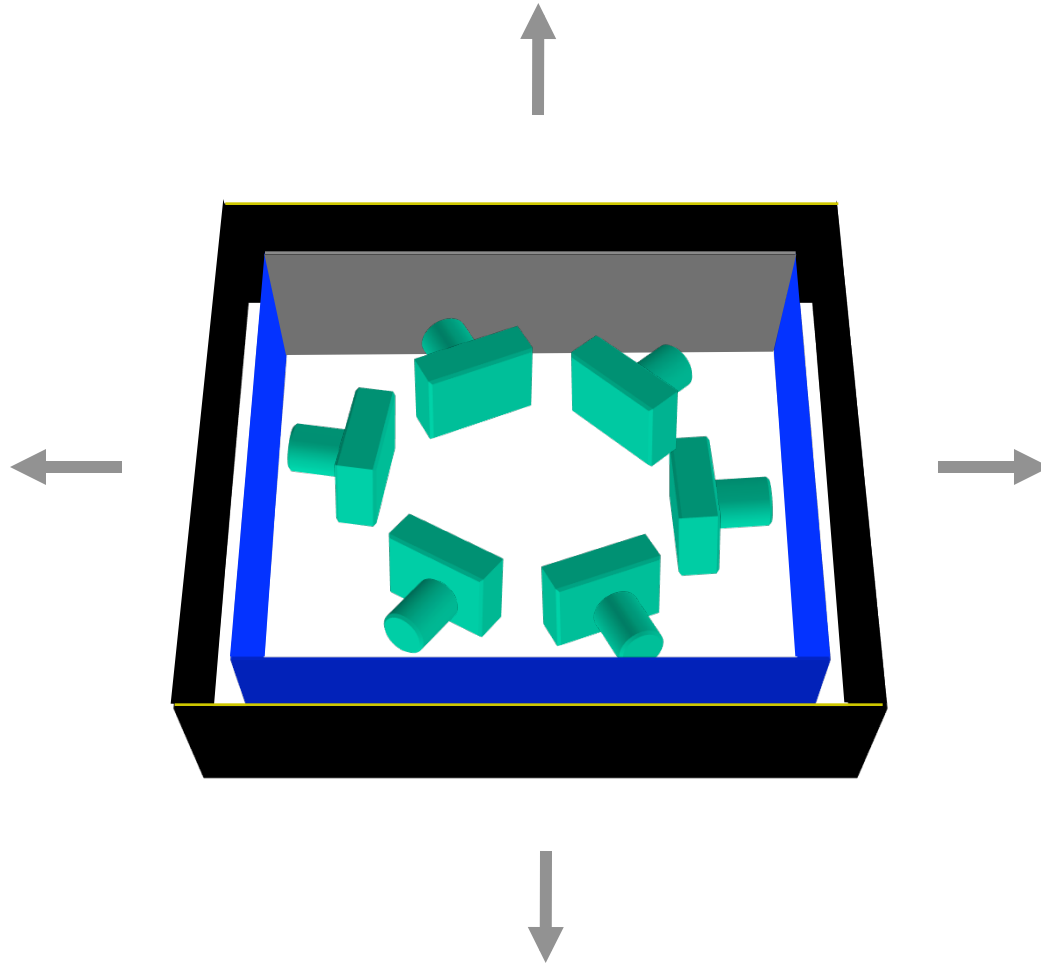


Panoramic Depth Ordering



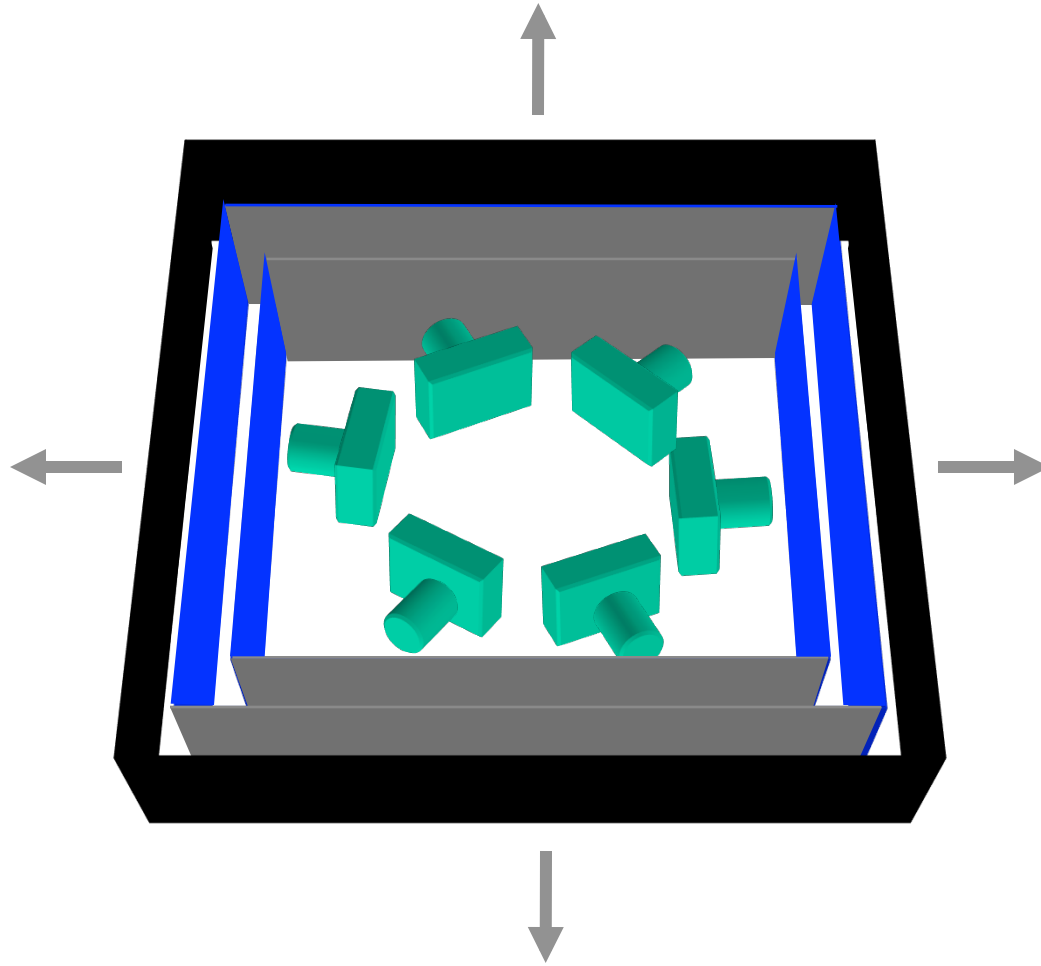
Layers radiate outwards from cameras

Panoramic Layering



Layers radiate outwards from cameras

Panoramic Layering

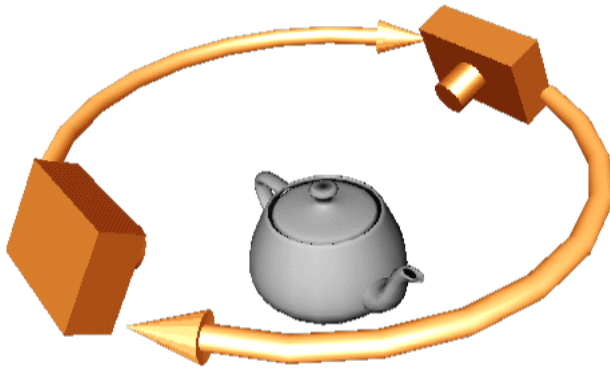


Layers radiate outwards from cameras

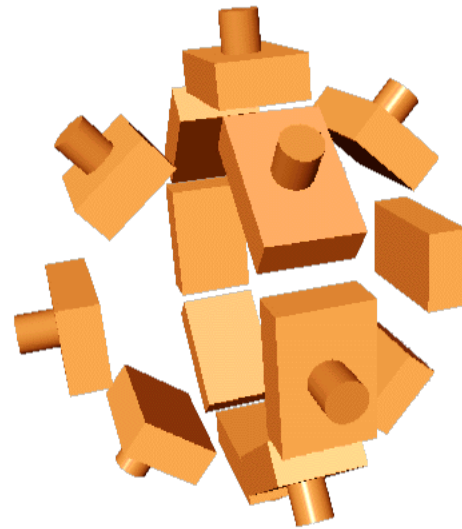
Compatible Camera Configurations

Depth-Order Constraint

- Scene outside convex hull of camera centers



*Inward-Looking
cameras above scene*



*Outward-Looking
cameras inside scene*

Calibrated Image Acquisition



Calibrated Turntable



Selected Dinosaur Images



Selected Flower Images

Voxel Coloring Results (Video)



Dinosaur Reconstruction

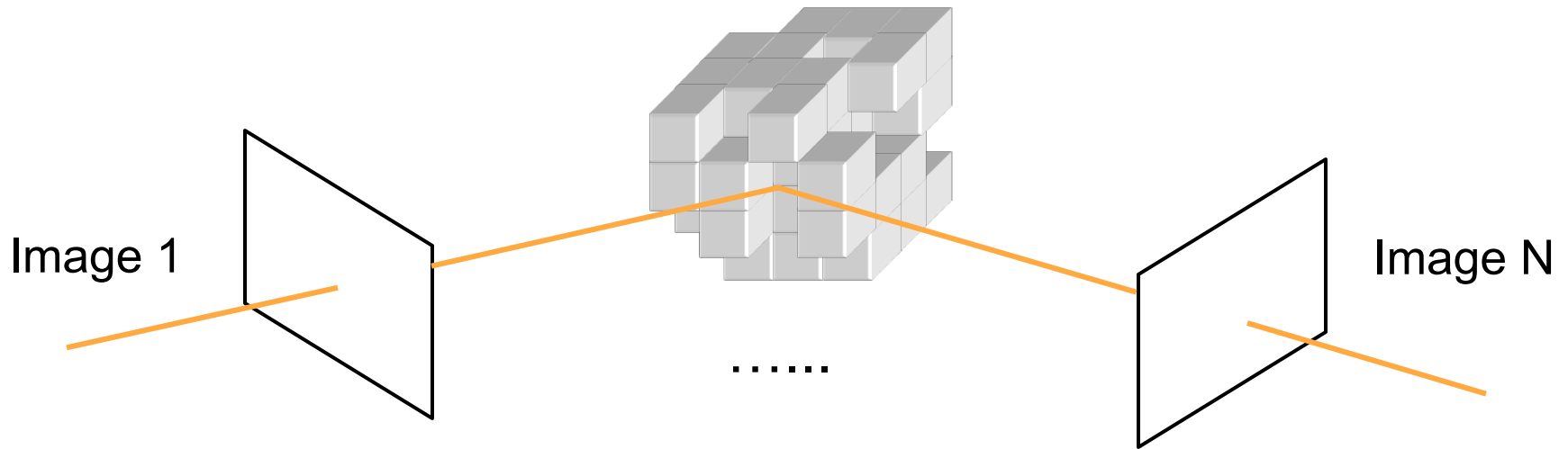
72 K voxels colored
7.6 M voxels tested
7 min. to compute
on a 250MHz SGI



Flower Reconstruction

70 K voxels colored
7.6 M voxels tested
7 min. to compute
on a 250MHz SGI

Space Carving



Space Carving Algorithm

- Initialize to a volume V containing the true scene
- Choose a voxel on the outside of the volume
- Project to visible input images
- Carve if not photo-consistent
- Repeat until convergence

Space Carving Results: African Violet



Input Image (1 of 45)



Reconstruction



Reconstruction



Reconstruction

Space Carving Results: Hand



**Input Image
(1 of 100)**



Views of Reconstruction

Carved visual hulls

- The visual hull is a good starting point for optimizing photo-consistency
 - Easy to compute
 - Tight outer boundary of the object
 - Parts of the visual hull (rims) already lie on the surface and are already photo-consistent

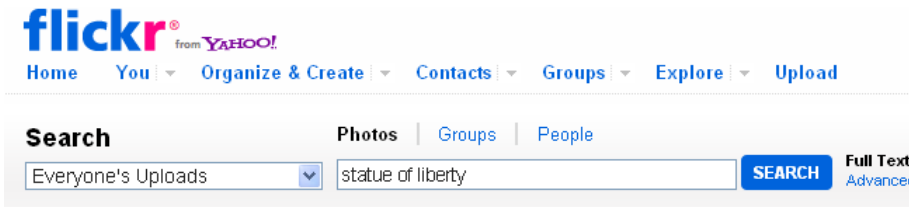
Carved visual hulls

1. Compute visual hull
2. Use dynamic programming to find rims (photo-consistent parts of visual hull)
3. Carve the visual hull to optimize photo-consistency keeping the rims fixed



Stereo from community photo collections

- Up to now, we've always assumed that camera calibration is known
- For photos taken from the Internet, we need *structure from motion* techniques to reconstruct both camera positions and 3D points



Sort: **Relevant** | [Recent](#) | [Interesting](#) View: **Small** | [Medium](#) | [Detail](#) | [Slideshow](#)



From EdZa



From micbaun



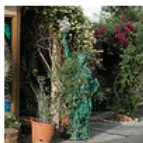
From rafaj



From Iepublicme



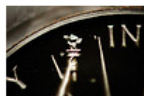
From Jesus...



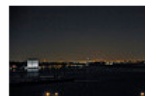
From Julio...



From StephiGra...



From alabs



From BigMs.Take



From laurenbou...



From laurenbou...



From StephiGra...



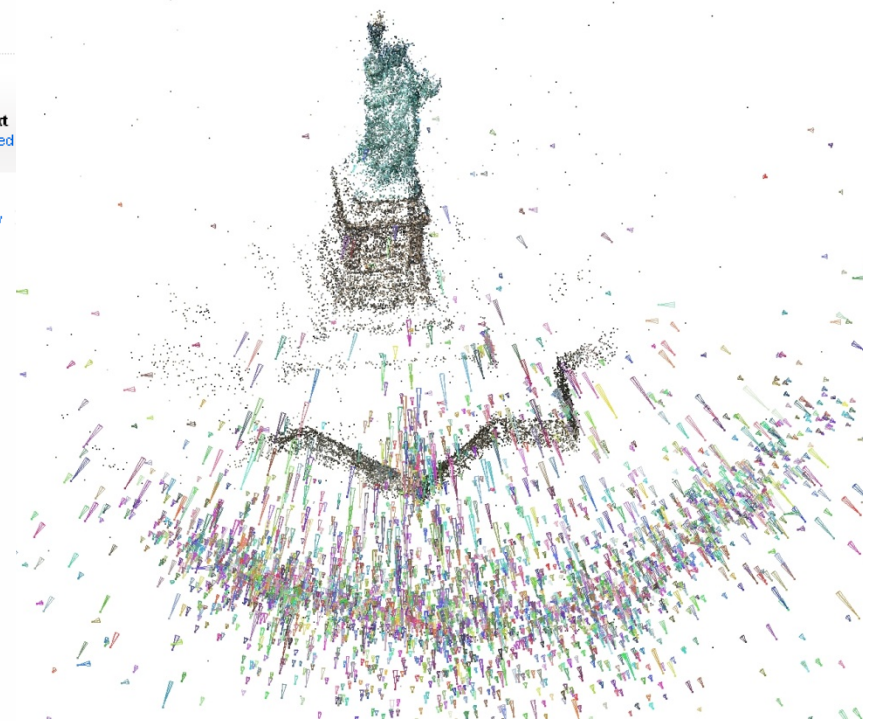
From dmp0309

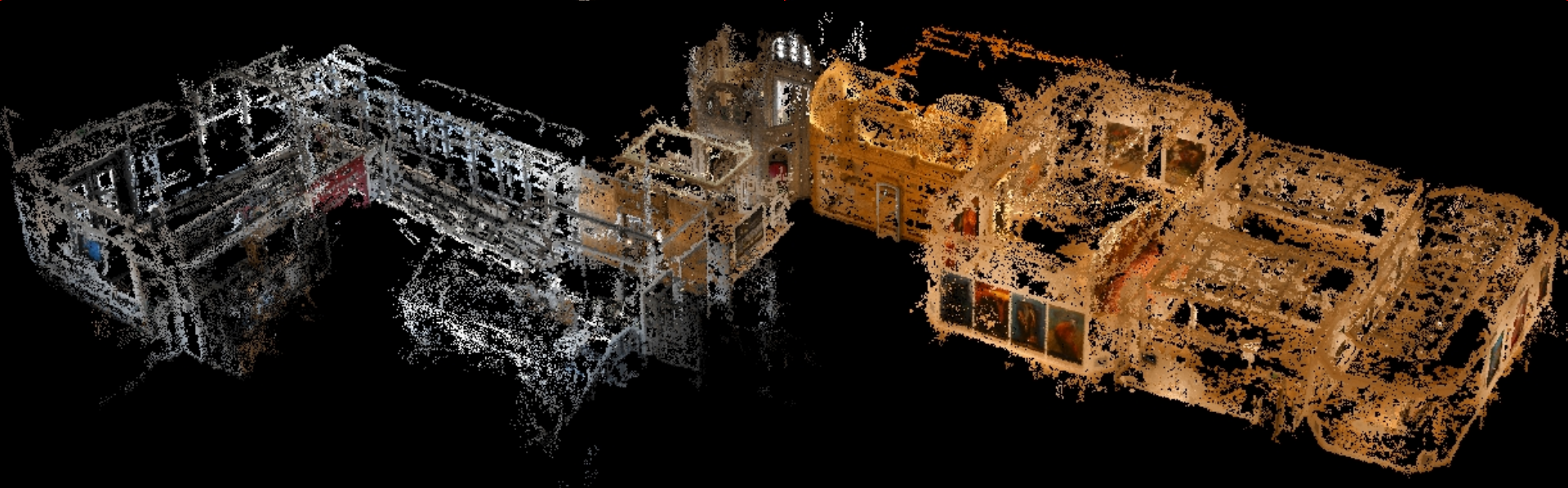


From laverrue

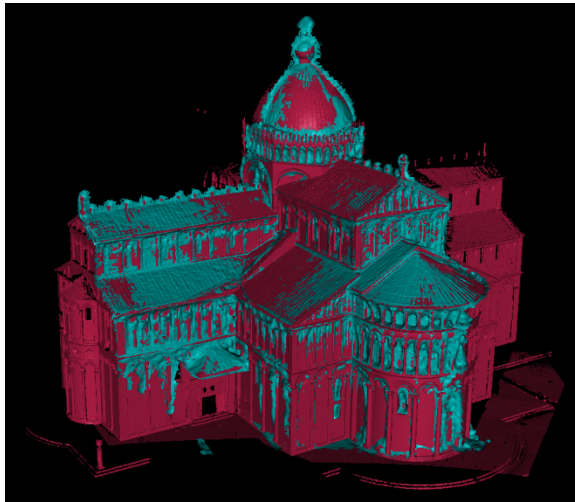
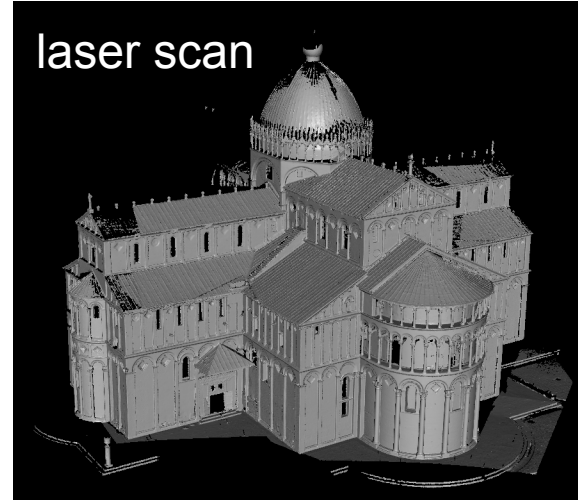
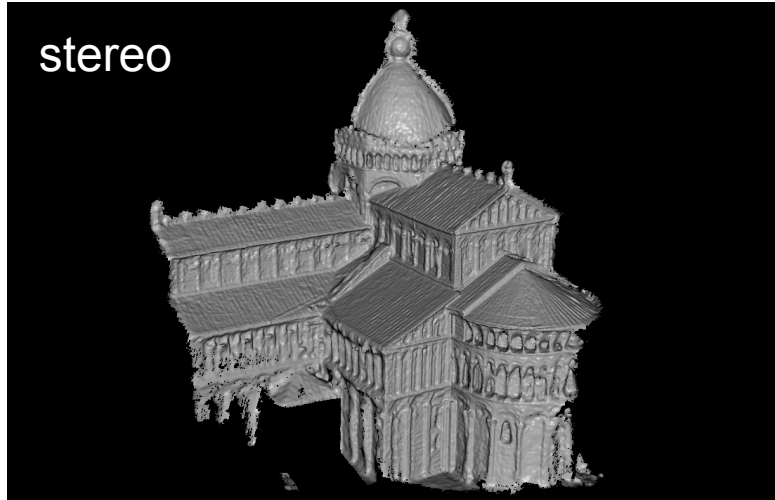


From Mojumbo22...





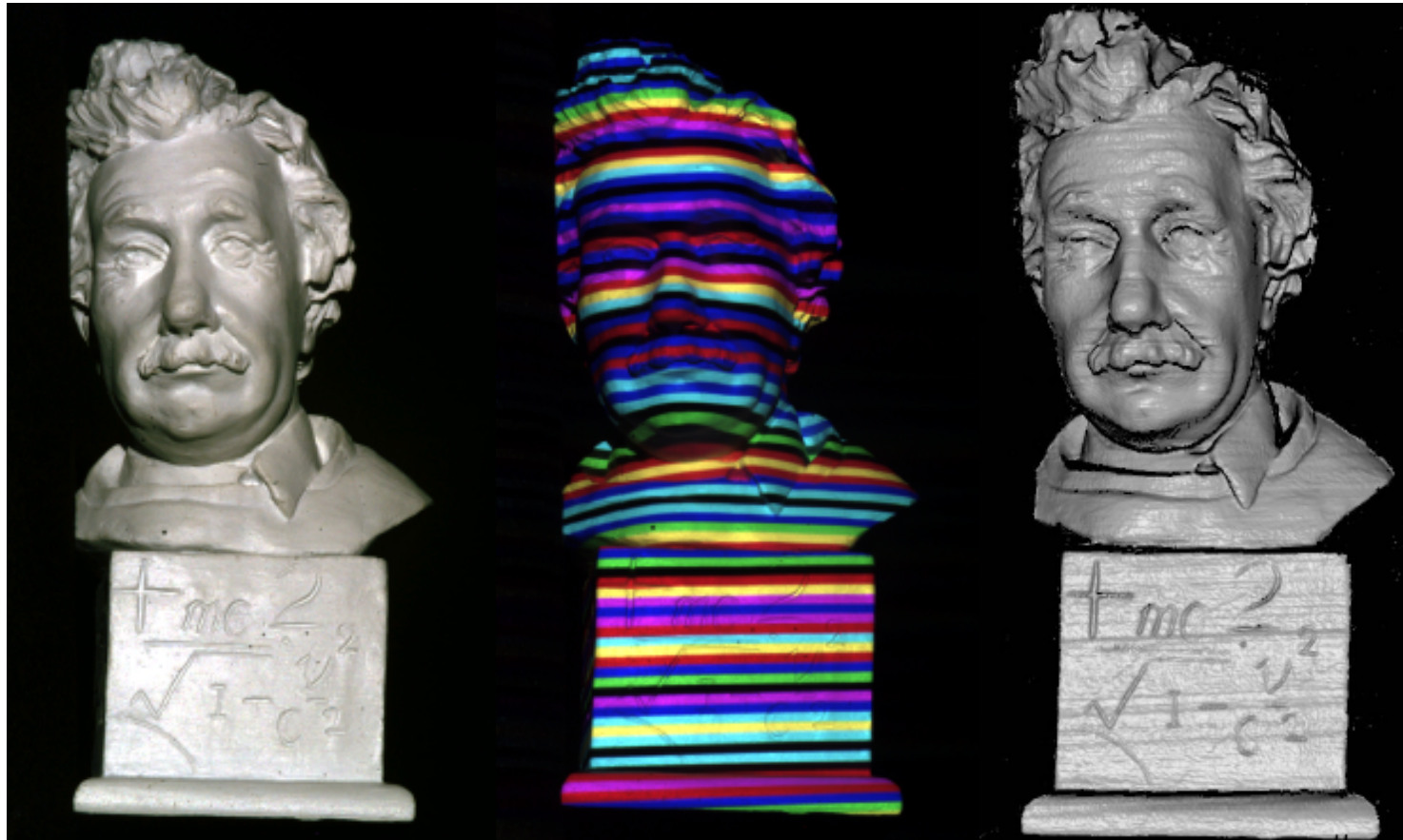
Stereo from community photo collections



Comparison: 90% of points
within 0.128 m of laser scan
(building height 51m)

Scanning technologies

Structured light

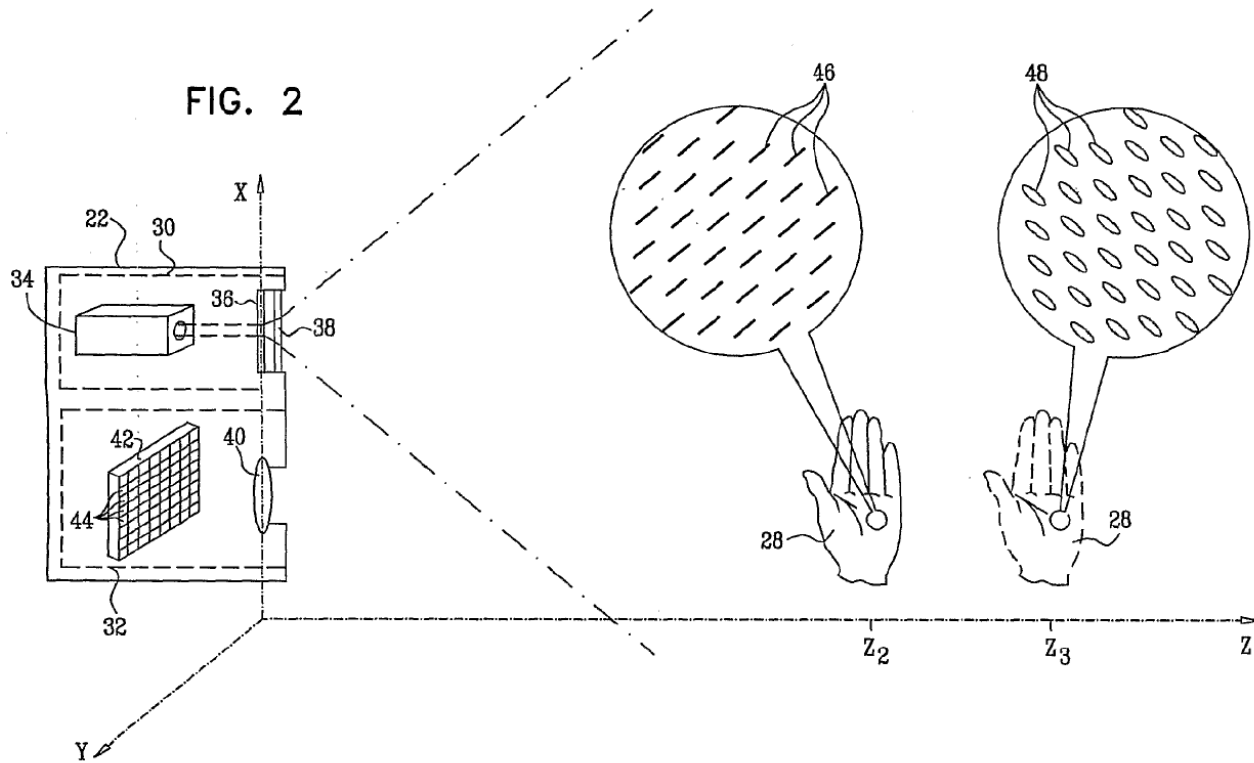


Kinect: Structured infrared light

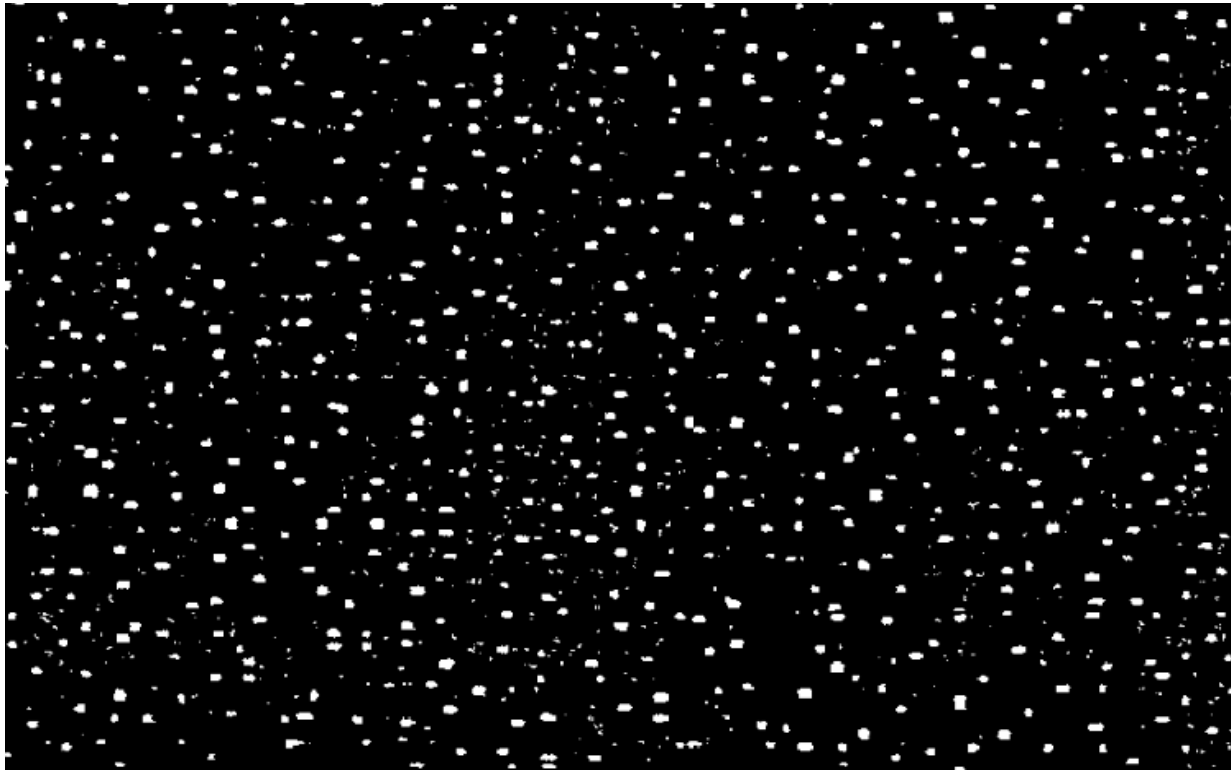


<http://bbzipo.wordpress.com/2010/11/28/kinect-in-infrared/>

PrimeSense Patents



The Kinect uses infrared laser light,
with a speckle pattern



Shpunt et al, PrimeSense patent application
US 2008/0106746

KinectFusion: Real-time 3D Reconstruction and Interaction Using a Moving Depth Camera*

SIGGRAPH Talks 2011

KinectFusion:

**Real-Time Dynamic 3D Surface
Reconstruction and Interaction**

**Shahram Izadi 1, Richard Newcombe 2, David Kim 1,3, Otmar Hilliges 1,
David Molyneaux 1,4, Pushmeet Kohli 1, Jamie Shotton 1,
Steve Hodges 1, Dustin Freeman 5, Andrew Davison 2, Andrew Fitzgibbon 1**

1 Microsoft Research Cambridge 2 Imperial College London

3 Newcastle University 4 Lancaster University

5 University of Toronto