3D photography (II)

Digital Visual Effects, Spring 2005

Yung-Yu Chuang

2005/5/25

with slides by Szymon Rusinkiewicz, Richard Szeliski, Steve Seitz and Brian Curless

Announcements



- Final project will be online tomorrow
- Proposal presentation on next Wednesday
- I will send out your current grades by next Wednesday
- Scribe (SIGGRAPH 2005, CVPR 2005, readings)
- Schedule for the next few weeks
 - 6/1 proposal
 - 6/8 making face/human
 - 6/15 random topics
 - 6/28 final project presentation

Outline

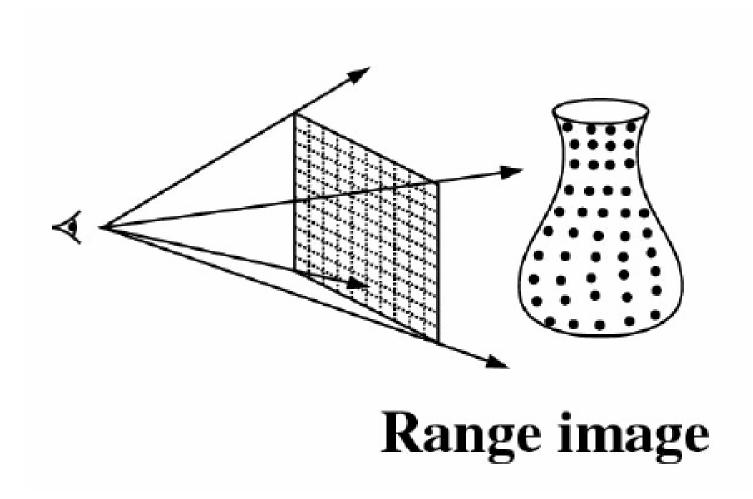


- Range acquisition techniques
- Full model reconstruction
 - ICP
 - Volumetric reconstruction
- Systems, projects and applications
- Final project

Range acquisition

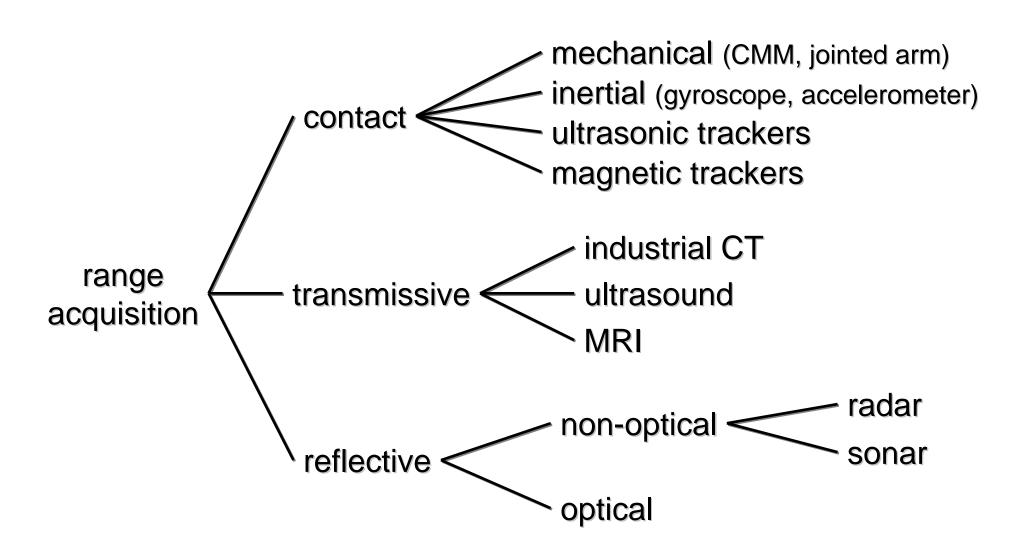
Range acquisition





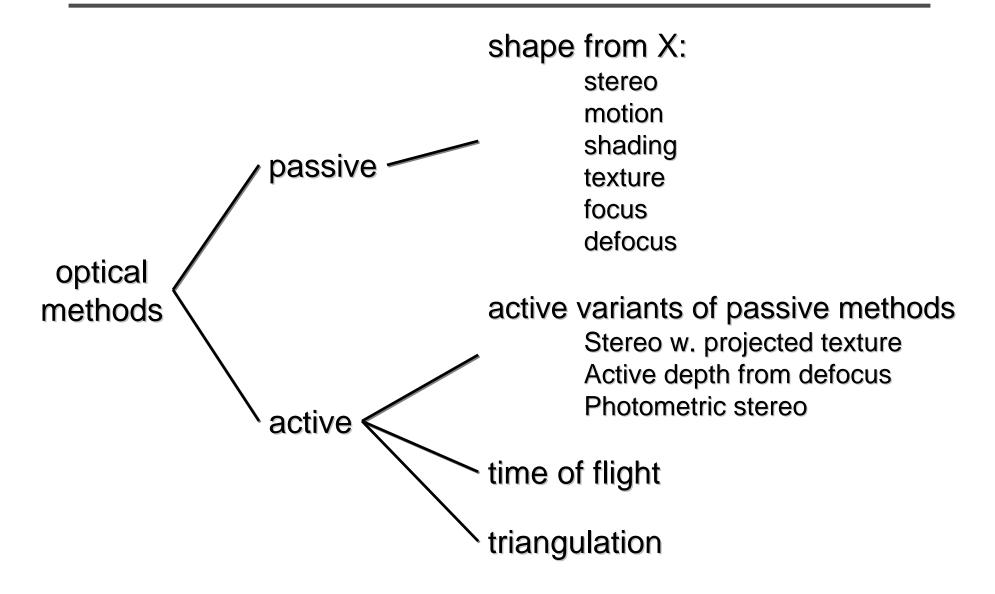


Range acquisition taxonomy





Range acquisition taxonomy



Passive approaches





stereo



space carving

Active approaches





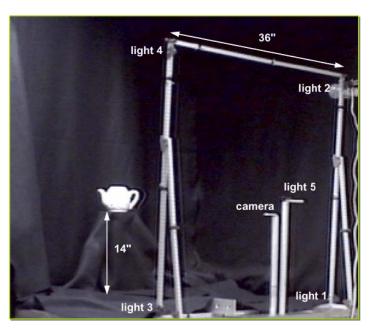


Cyberware whole body scanner

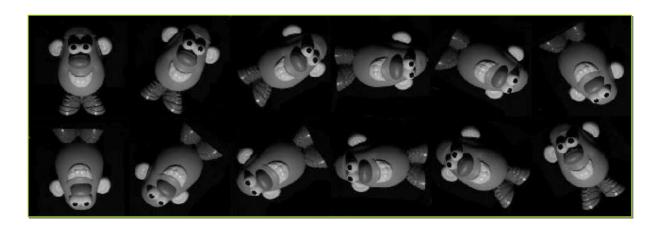
shadow scanning

Active variants





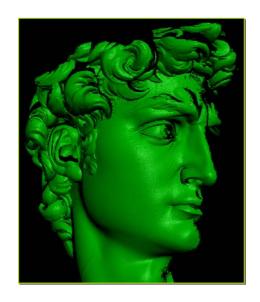




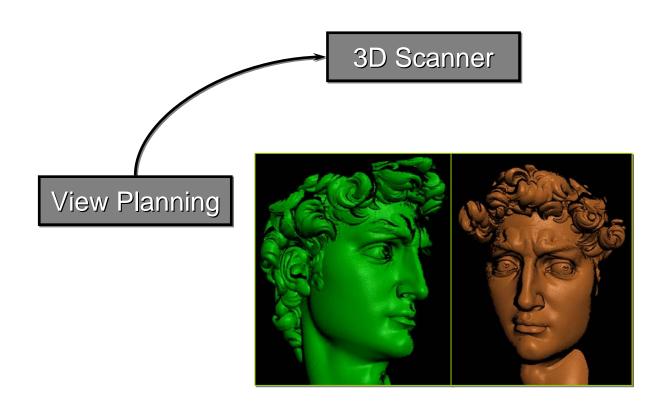
Full model reconstruction



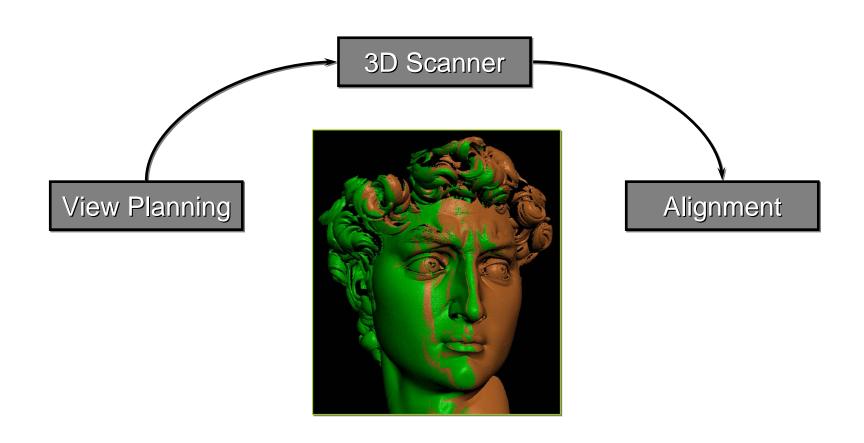
3D Scanner



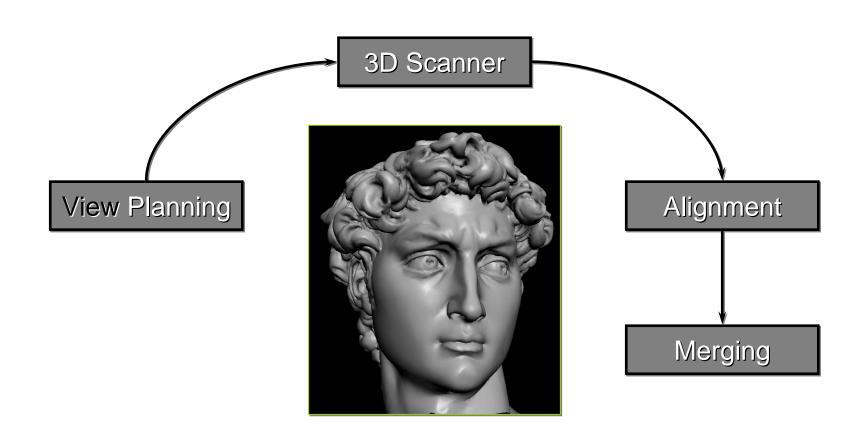








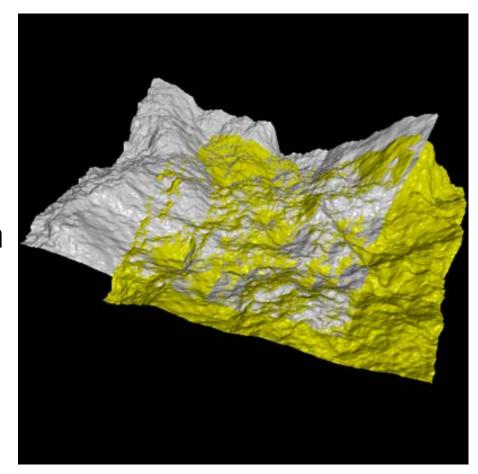




Problem



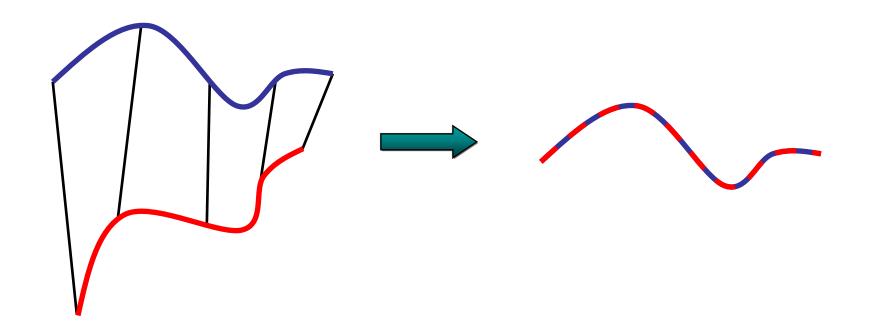
 Align two partiallyoverlapping meshes given initial guess for relative transform





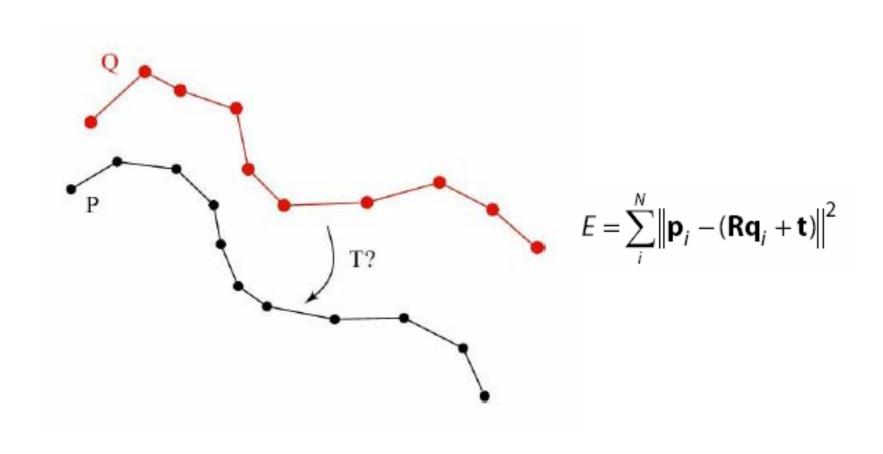
Aligning 3D data

 If correct correspondences are known, it is possible to find correct relative rotation/translation





Find the optimal transformation





Find the optimal transformation

To solve, we first compute the centroid of each point set:

$$\overline{\mathbf{p}} = \sum_{i}^{N} \mathbf{p}_{i} \quad \overline{\mathbf{q}} = \sum_{i}^{N} \mathbf{q}_{i}$$

Horn showed that the best rotation satisfies:

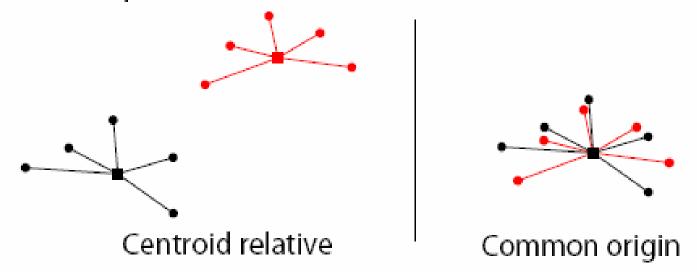
$$\underset{\mathbf{R}}{\operatorname{argmax}} \sum_{i}^{N} (\mathbf{p}_{i} - \overline{\mathbf{p}})^{T} \mathbf{R} (\mathbf{q}_{i} - \overline{\mathbf{q}})$$



Find the optimal transformation

In other words:

- Convert the points into vectors relative to their centroids.
- Find a rotation that makes corresponding vectors have dot products as close to 1 as possible.





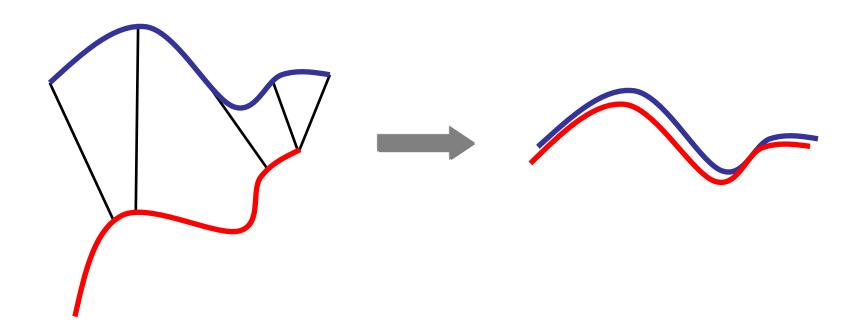
Aligning 3D data

- How to find corresponding points?
- Previous systems based on user input, feature matching, surface signatures, etc.





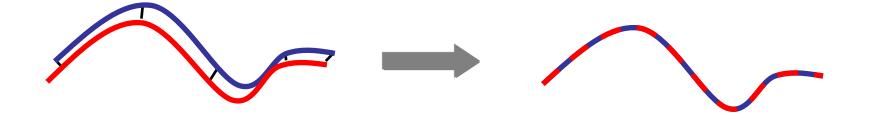
• Alternative: assume closest points correspond to each other, compute the best transform...







- ... and iterate to find alignment
 - Iterated Closest Points (ICP) [Besl & McKay 92]
- Converges if starting position "close enough"



ICP variants



- Variants on the following stages of ICP have been proposed:
 - 1. Selecting source points (from one or both meshes)
 - 2. Matching to points in the other mesh
 - 3. Weighting the correspondences
 - 4. Rejecting certain (outlier) point pairs
 - 5. Assigning an error metric to the current transform
 - 6. Minimizing the error metric

ICP variants



- 1. Selecting source points (from one or both meshes)
- 2. Matching to points in the other mesh
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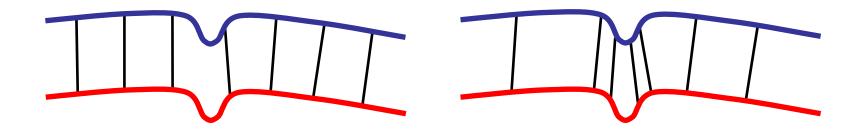




- Use all points
- Uniform subsampling
- Random sampling
- Normal-space sampling
 - Ensure that samples have normals distributed as uniformly as possible

Normal-space sampling





Uniform Sampling

Normal-Space Sampling

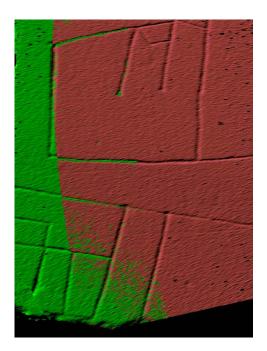


Normal-space sampling

 Conclusion: normal-space sampling better for mostly-smooth areas with sparse features



Random sampling



Normal-space sampling

ICP variants



- 1. Selecting source points (from one or both meshes)
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Matching

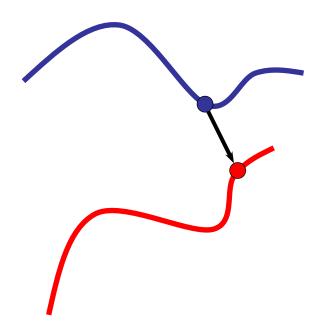


- Matching strategy has greatest effect on convergence and speed
- Closest point
- Normal shooting
- Closest compatible point
- Projection



Closest-point matching

Find closest point in other mesh

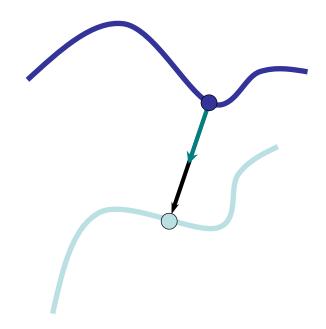


Closest-point matching generally stable, but slow and requires preprocessing



Normal shooting

Project along normal, intersect other mesh



Slightly better than closest point for smooth meshes, worse for noisy or complex meshes



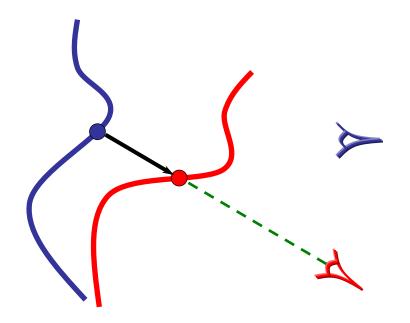
Closest compatible point

- Can improve effectiveness of both of the previous variants by only matching to compatible points
- Compatibility based on normals, colors, etc.
- At limit, degenerates to feature matching



Projection to find correspondences

- Finding closest point is most expensive stage of the ICP algorithm
- Idea: use a simpler algorithm to find correspondences
- For range images, can simply project point [Blais 95]





Projection-based matching

- Slightly worse performance per iteration
- Each iteration is one to two orders of magnitude faster than closest-point

ICP variants



- 1. Selecting source points (from one or both meshes)
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ICP variants

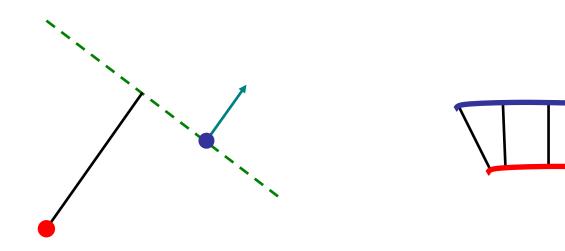


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Point-to-plane error metric

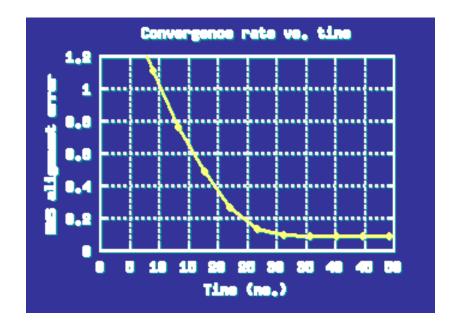
 Using point-to-plane distance instead of pointto-point lets flat regions slide along each other [Chen & Medioni 91]





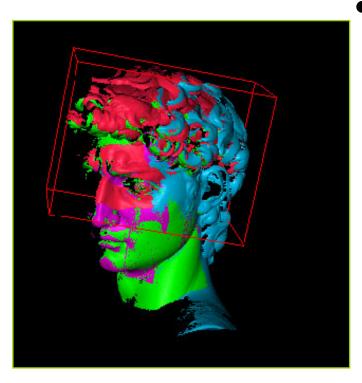
High-speed ICP algorithm

 ICP algorithm with projection-based correspondences, point-to-plane matching can align meshes in a few tens of ms. (cf. over 1 sec. with closest-point)







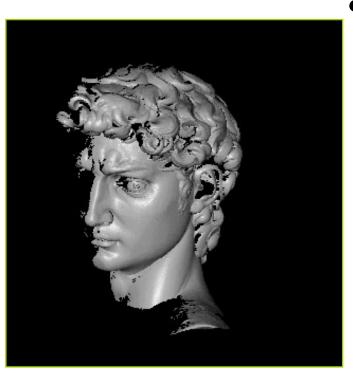


Steps

- 1. manual initial alignment
- 2. ICP to one existing scan
- 3. automatic ICP of all overlapping pairs
- 4. global relaxation to spread out error
- 5. merging using volumetric method





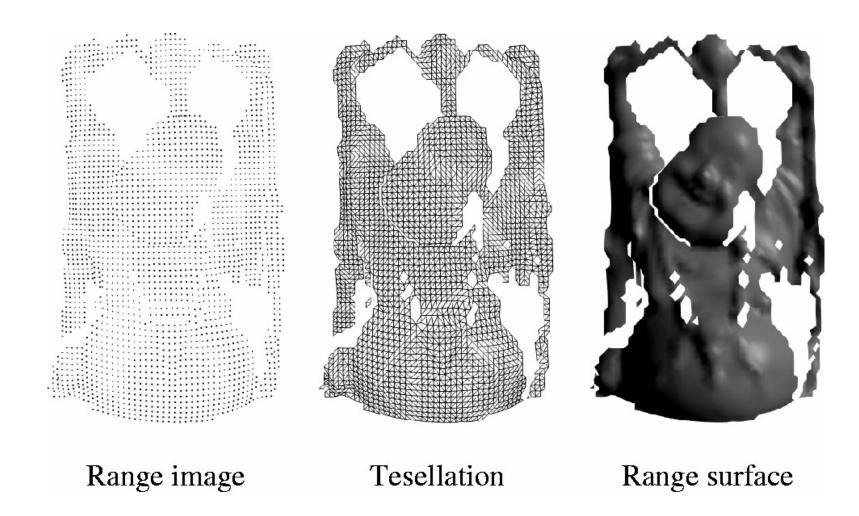


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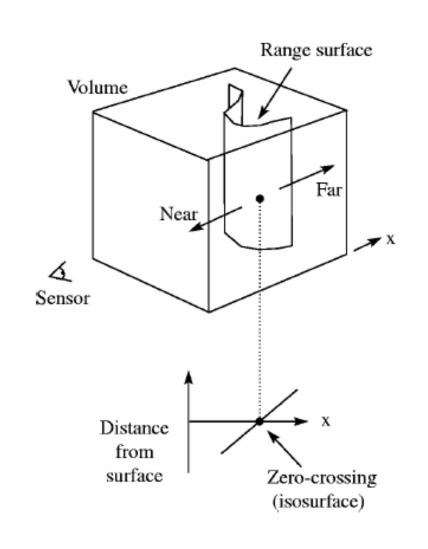
Volumetric reconstruction

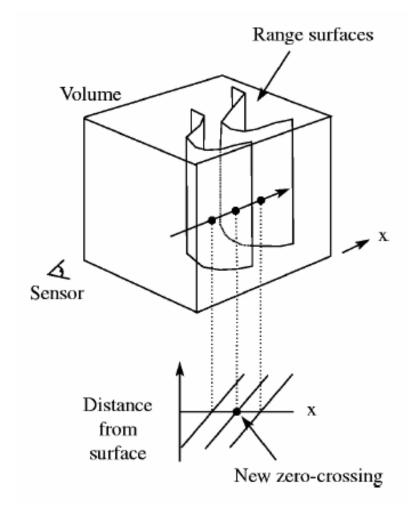




Signed distance function

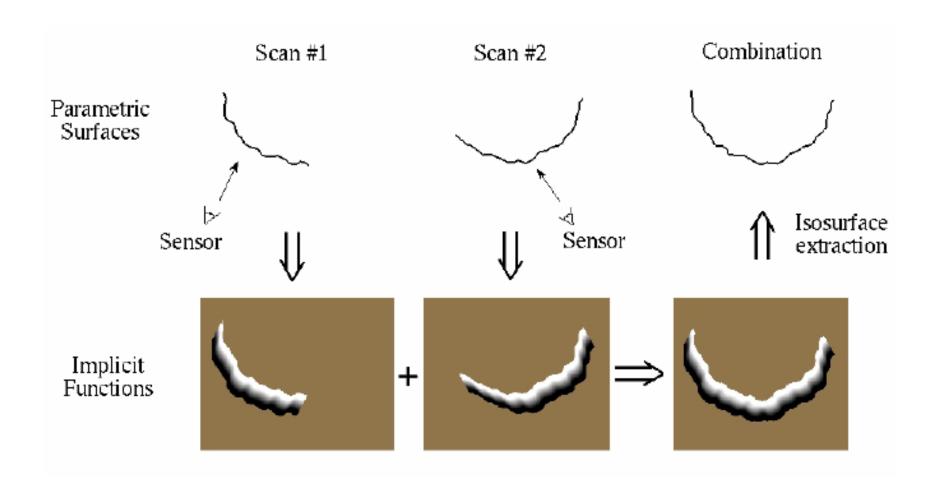






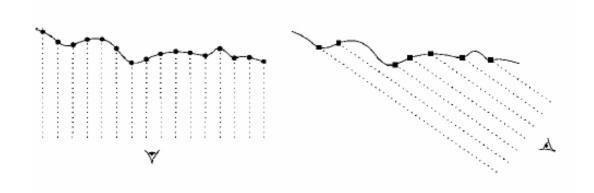
Overview



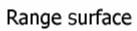


Weighting









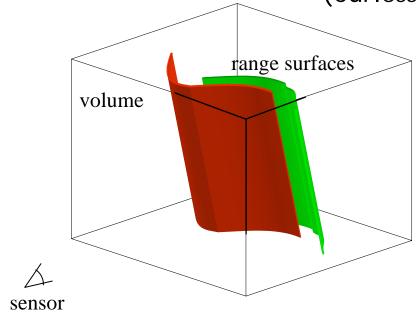


Confidence rendering

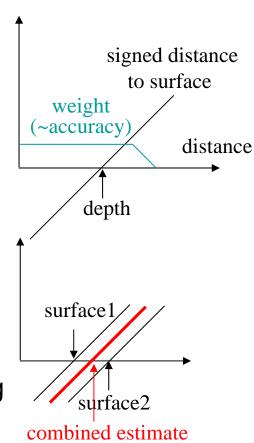
Volumetric integration



(Curless and Levoy, Siggraph 96)



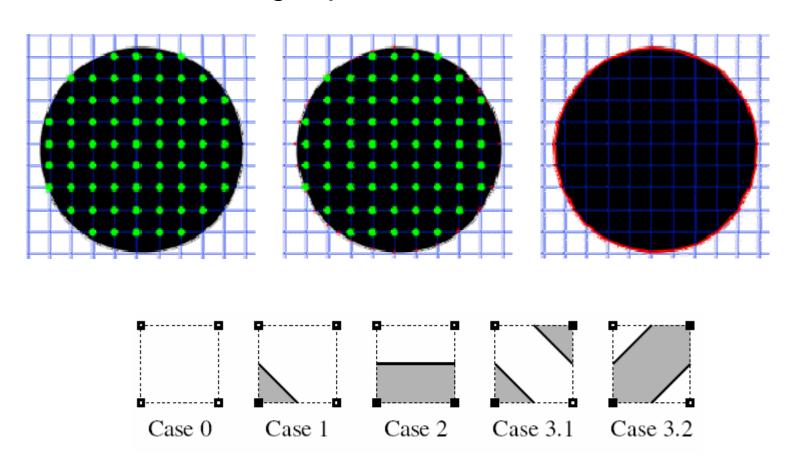
- use voxel space
- new surface as zero-crossing (find using marching cubes)
- least-squares estimate (zero derivative=minimum)





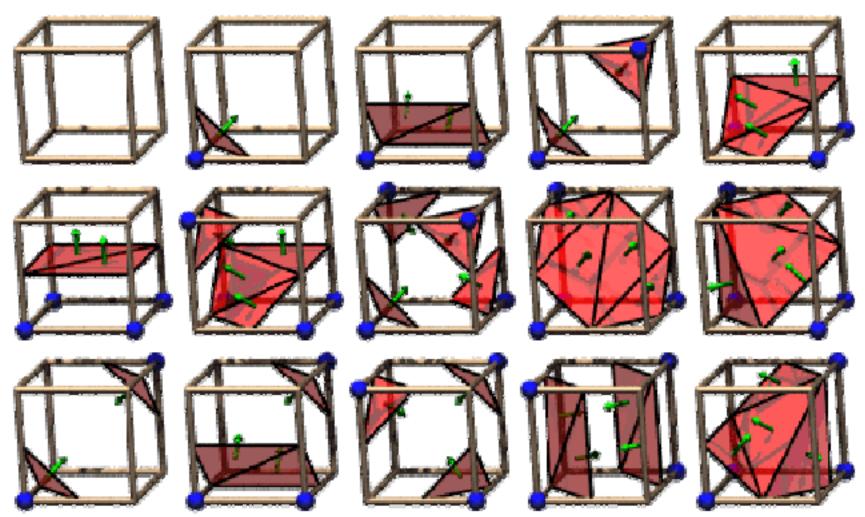
Isosurfacing: marching cubes

First 2D, marching squares



Marching cubes

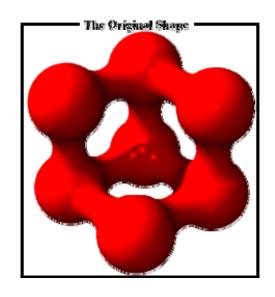


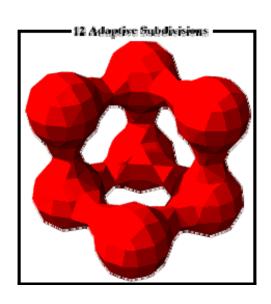


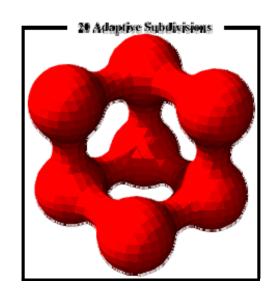
The 15 Cube Combinations

Marching cubes









Results





Photograph of original model

Photograph of painted original

Range surface from one scan

Reconstruction before hole–filling

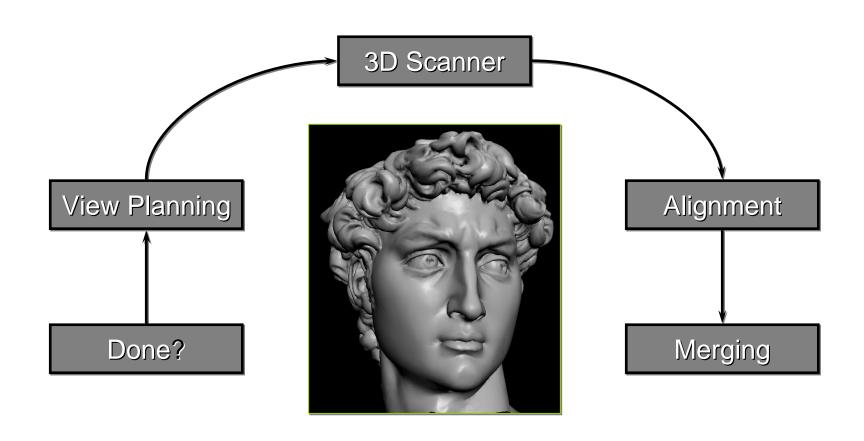
Reconstruction after hole–filling

Hardcopy

Systems, projects and applications

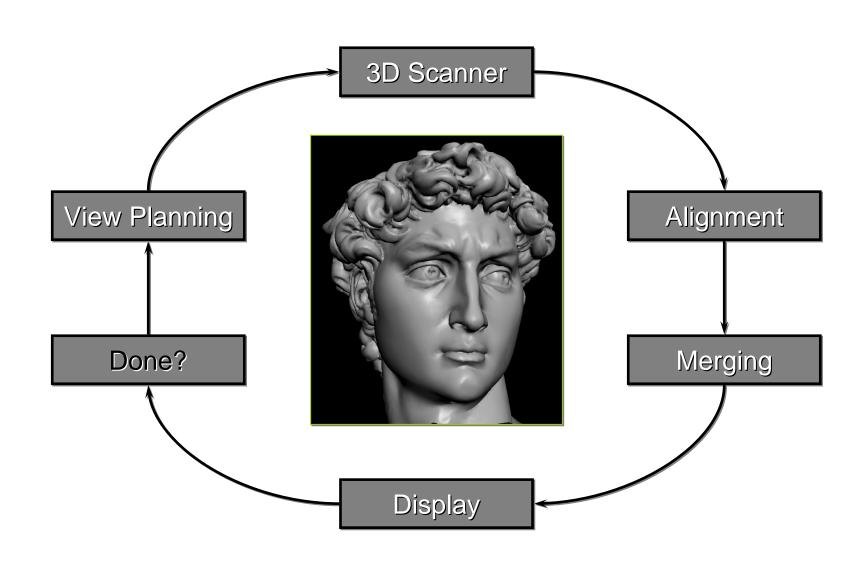


3D model acquisition pipeline



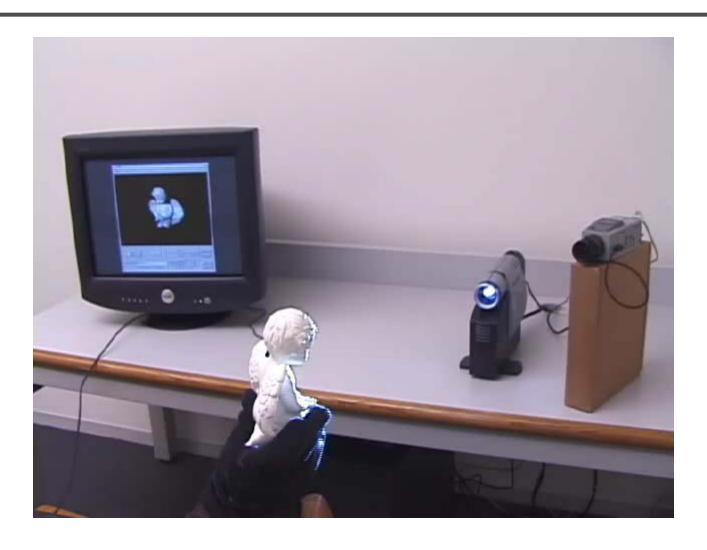


3D model acquisition pipeline



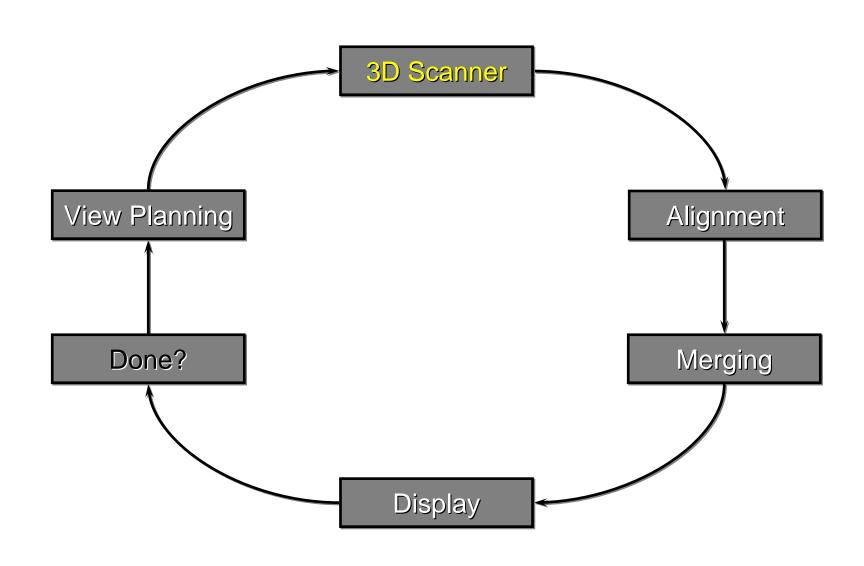


Real-time 3D model acquisition



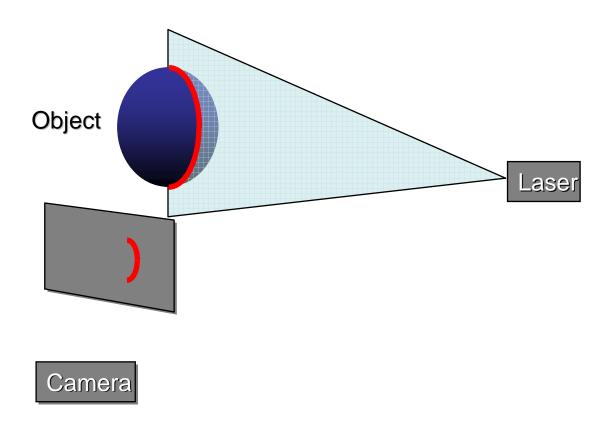


Real-time 3D model acquisition



Triangulation

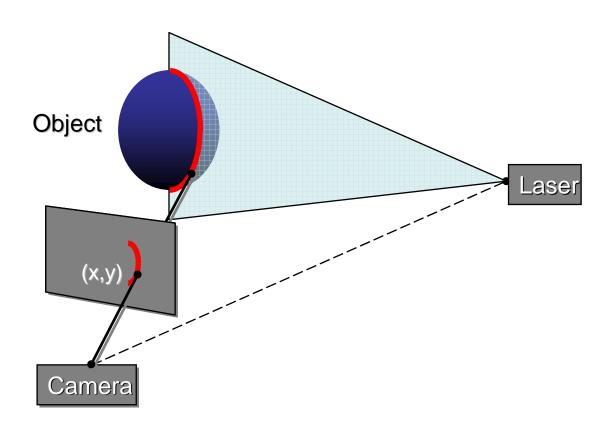




Project laser stripe onto object

Triangulation





• Depth from ray-plane triangulation

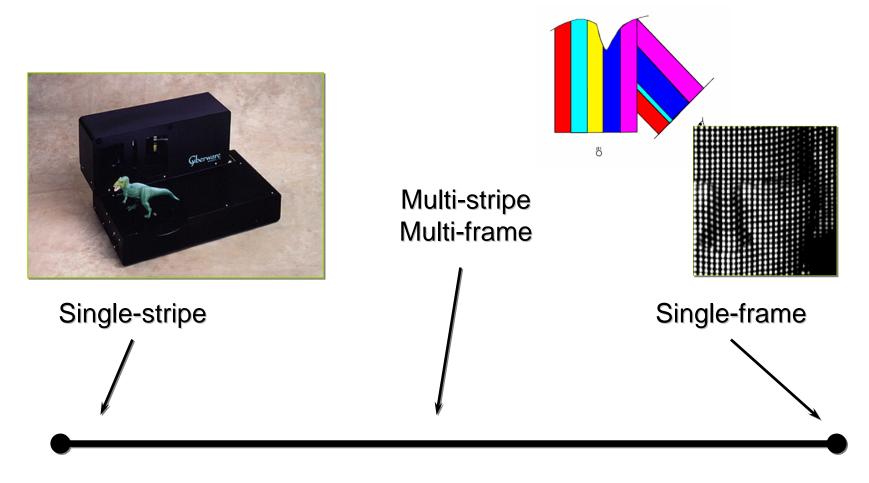


Triangulation

- Faster acquisition: project multiple stripes
- Correspondence problem: which stripe is which?



Continuum of triangulation methods



Slow, robust

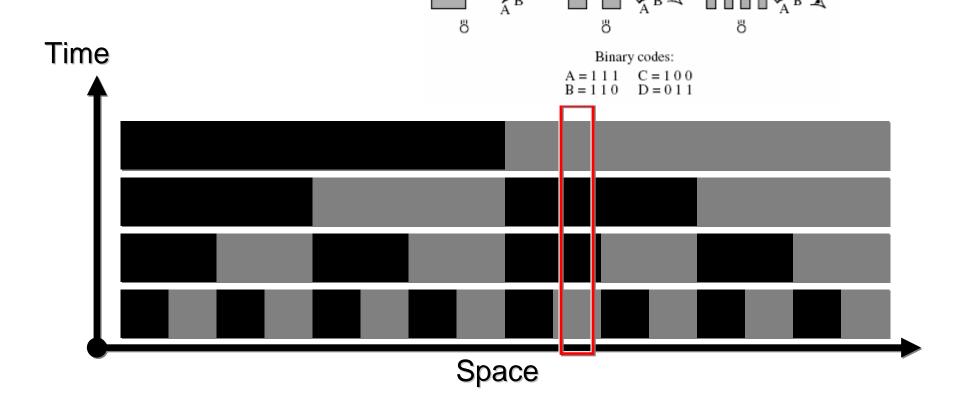
Fast, fragile



Time-coded light patterns

• Assign each stripe a unique illumination code

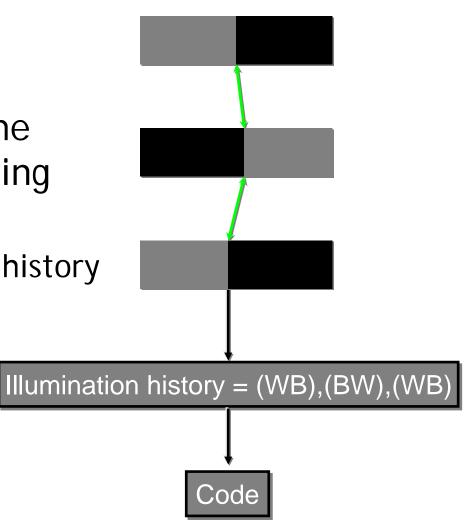
over time





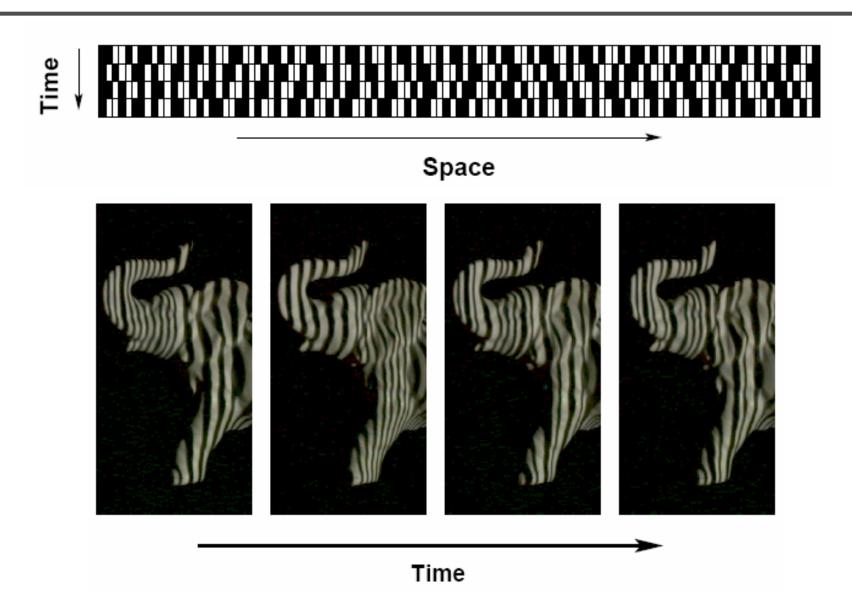


- Assign time codes to stripe boundaries
- Perform frame-to-frame tracking of corresponding boundaries
 - Propagate illumination history



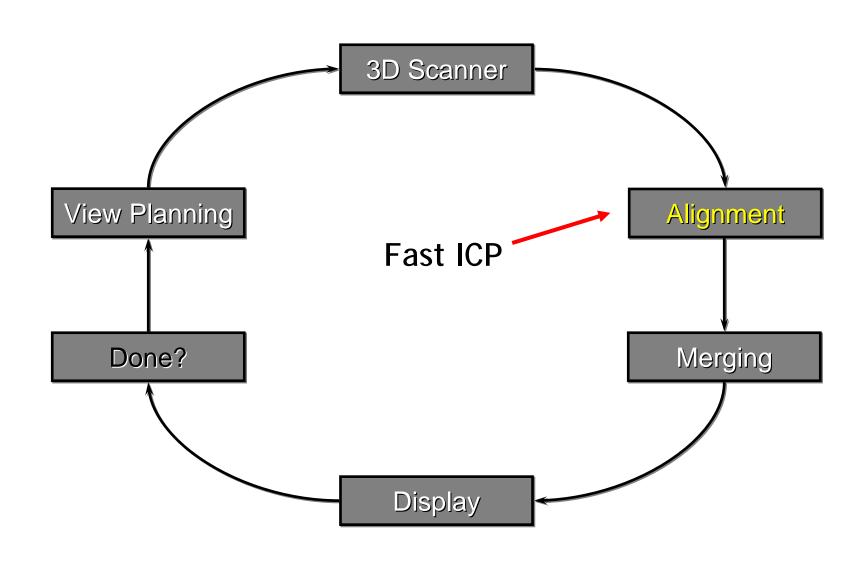


Codes for moving scenes



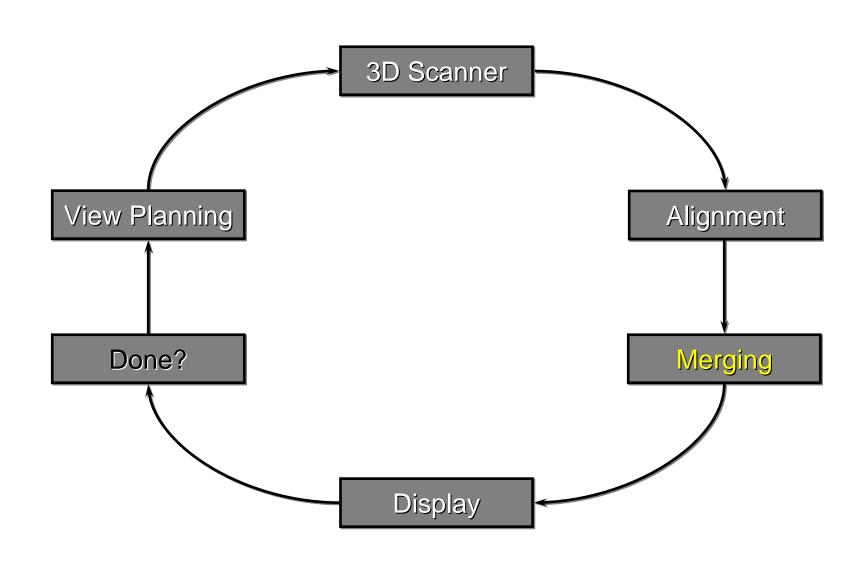


Real-time 3D model acquisition





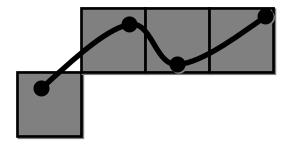
Real-time 3D model acquisition





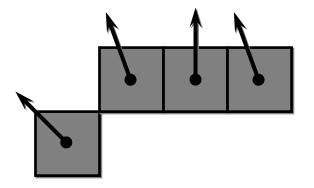




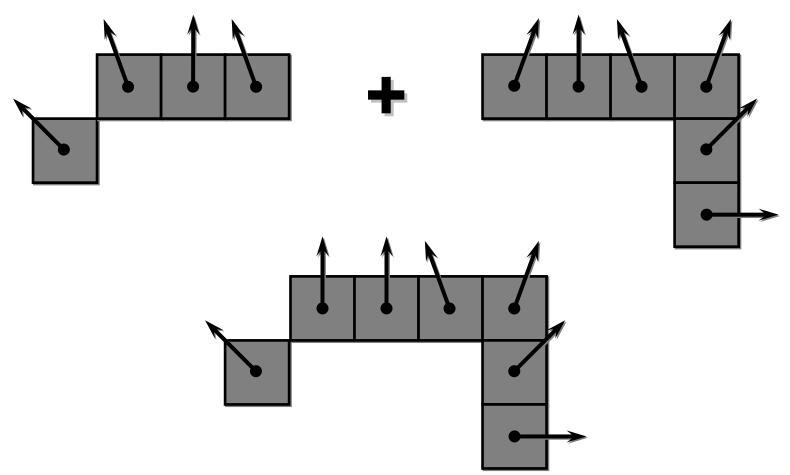












Point rendering, using accumulated normals for lighting



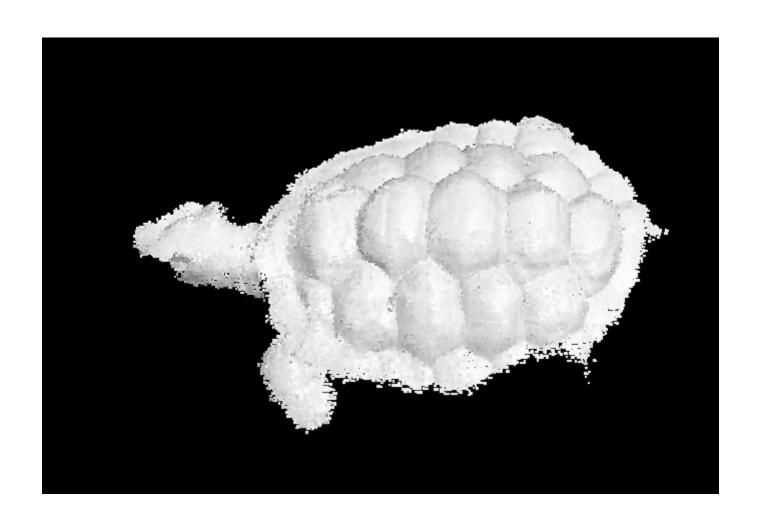
Example: photograph



18 cm.

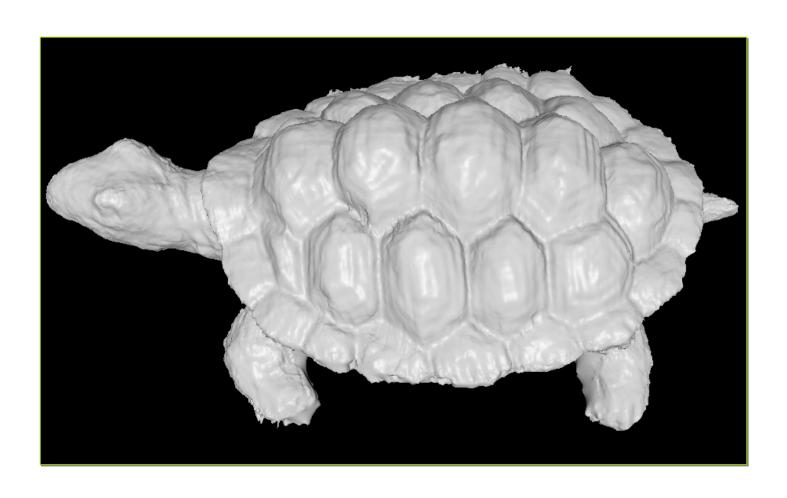
Result





Postprocessed model





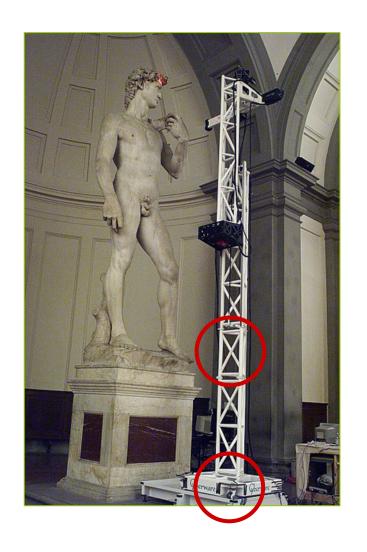


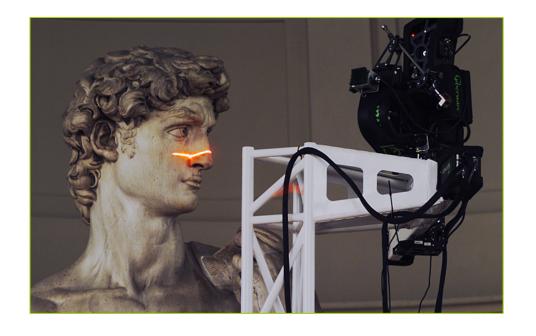
The Digital Michelangelo Project

- Goal: scan 10 sculptures by Michelangelo
- High-resolution ("quarter-millimeter") geometry
- Stanford University, led by Marc Levoy

Scanning the David





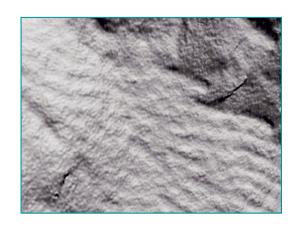


height of gantry: 7.5 meters

weight of gantry: 800 kilograms

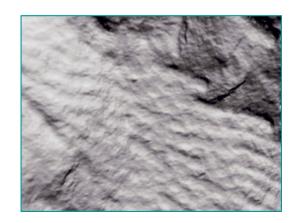


Range processing pipeline





- 1. manual initial alignment
- 2. ICP to one existing scan
- 3. automatic ICP of all overlapping pairs
- 4. global relaxation to spread out error
- 5. merging using volumetric method





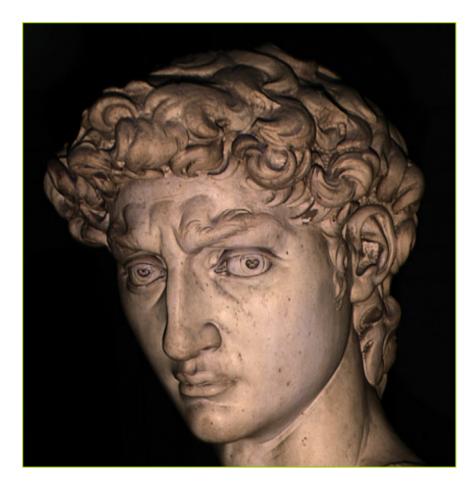




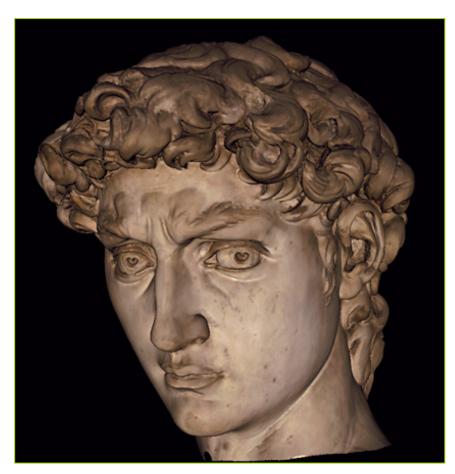
- 480 individually aimed scans
- 2 billion polygons
- 7,000 color images
- 32 gigabytes
- 30 nights of scanning
- 22 people

Comparison

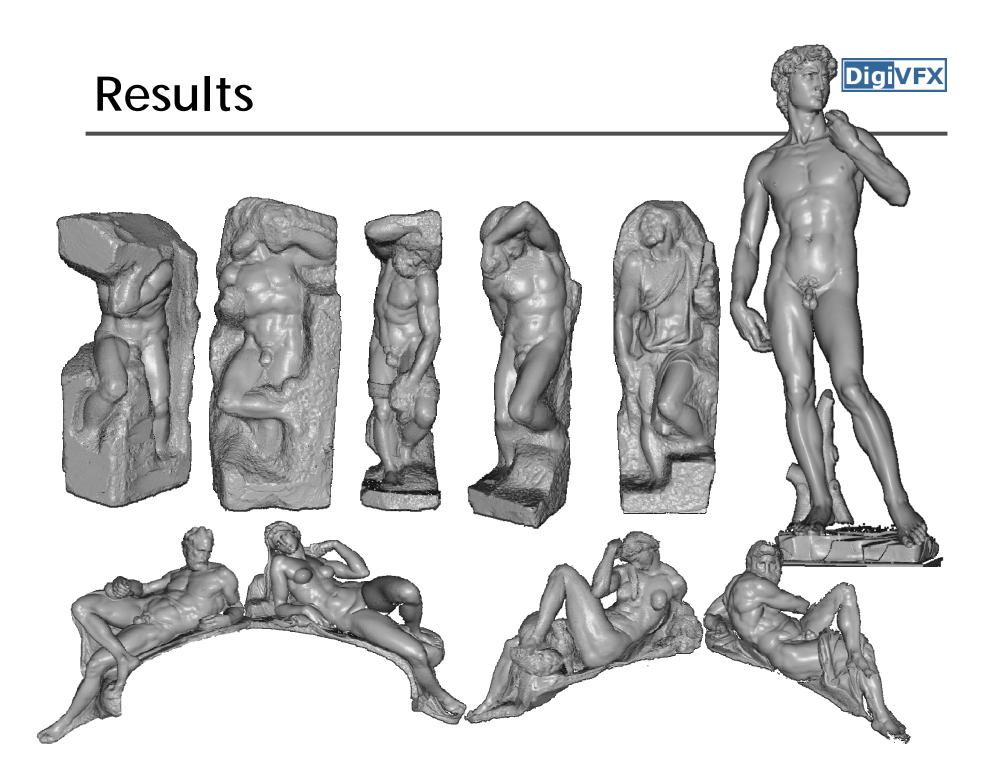




photograph



1.0 mm computer model





The Great Buddha Project

- Great Buddha of Kamakura
- Original made of wood, completed 1243
- Covered in bronze and gold leaf, 1267
- Approx. 15 m tall
- Goal: preservation of cultural heritage
- Institute of Industrial Science, University of Tokyo, led by Katsushi Ikeuchi



Scanner



- Cyrax range scanner by Cyra Technologies
- Laser pulse time-of-flight
- Accuracy: 4 mm
- Range: 100 m



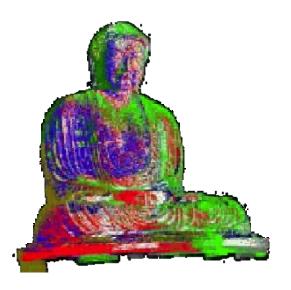






- 20 range images (a few million points)
- Simultaneous all-to-all ICP
- Variant of volumetric merging (parallelized)







Results





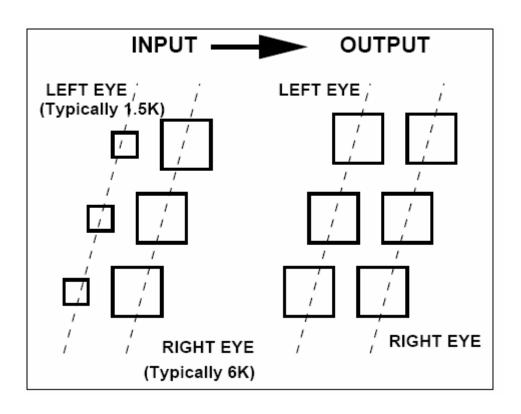
IMAX 3D



- 6K resolution, 42 linear bits per pixel
- For CG, it typically takes 6 hours for a frame
- 45-minute IMAX 3D CG film requires a 100-CPU rendering farm full-time for about a year just for rendering
- For live-action, camera is bulky (like a refrigerator)

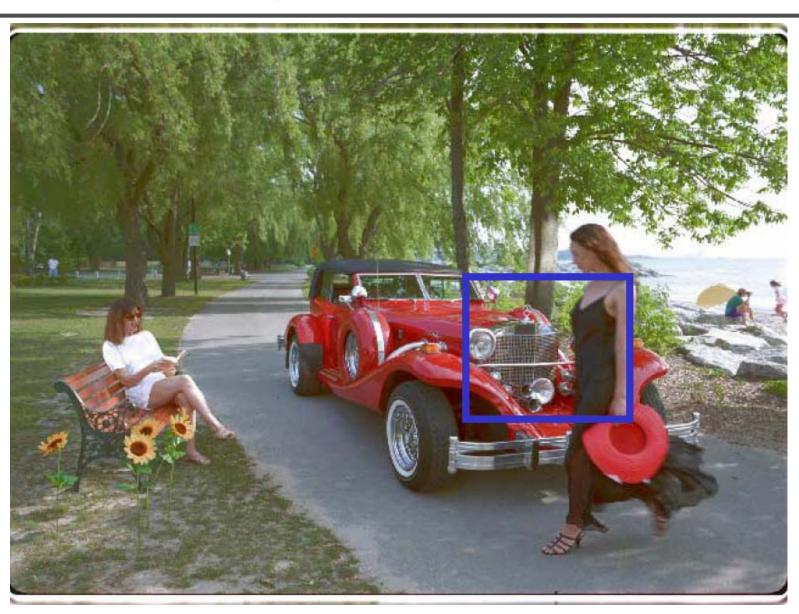


Hybrid stereo camera





Live-action sequence



Hybrid input





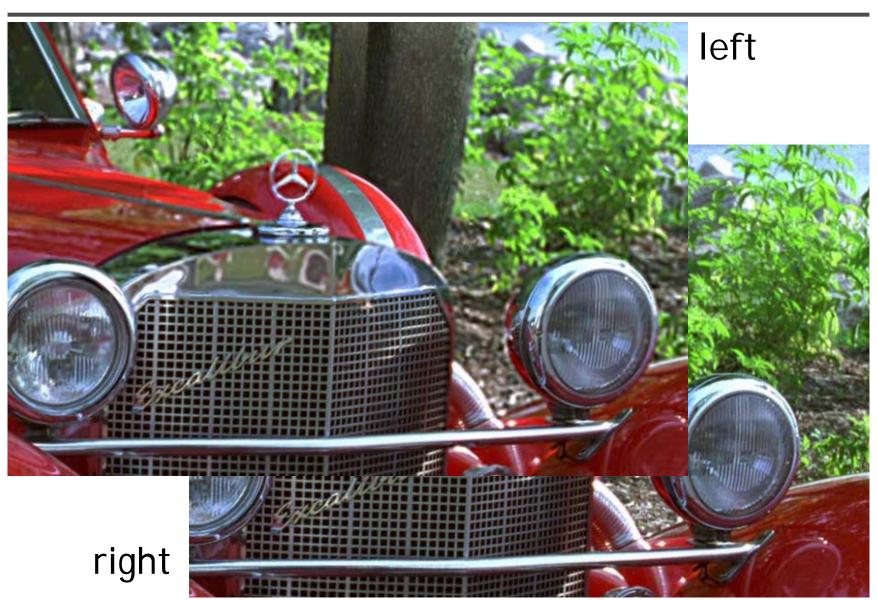
left



right

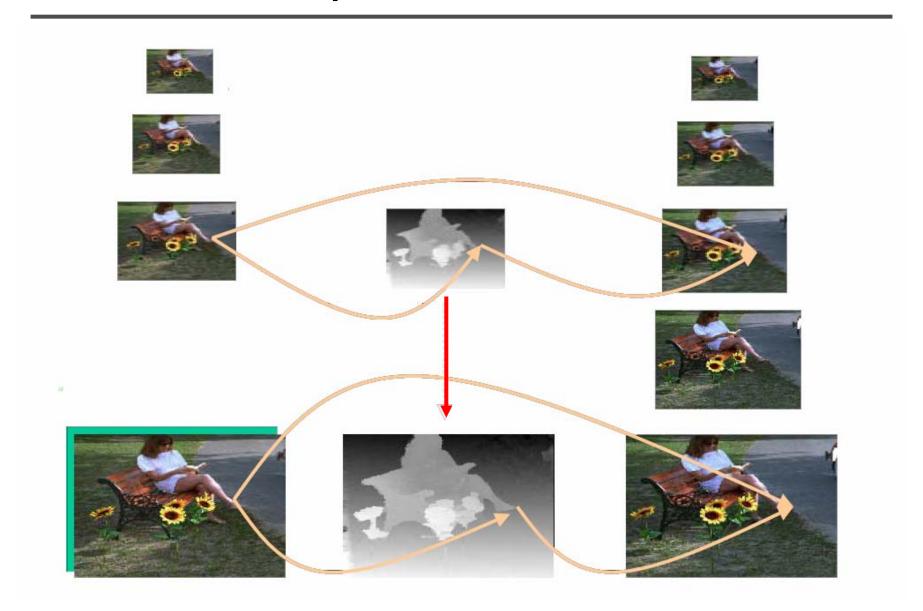








Combine multiple hires to lores





View interpolation





Bullet time video







High-quality video view interpolation

Final project

Final project



- Assigned: today
- Due: 6/28 Tuesday
- Final presentation: 6/28 Tuesday, 1:30pm?
- Proposal presentation next week, send me slides by 1:00pm next Wednesday if you want to use my laptop for your presentation
- Send me the topic and team members by next Tuesday

Final project



- Research (1-2 people)
- System (1-3 people)
- Film (3-4 people)

Research



- Define a problem and try to solve it
- You don't need to solve it all, but have to make a reasonable progress, for example, solve a simplified version.
- Find inspirations from SIGGRAPH/CVPR/ICCV papers

System



 Implement existing algorithm into a useful system

Film



- It must be an "effect" film.
- You can use any tools as you want. But, I guess that you have to write some on your own.
- Find inspirations from

Gatech's vfx course

http://www.cc.gatech.edu/classes/AY2004/cs4480_spring/

independent film makers

http://www.peerlessproductions.com/

Submit two videos, final and making-of.

Reference



- S. Rusinkiewicz and M. Levoy. <u>Efficient Variants of the ICP Algorithm</u>, 3DIM 2001.
- B. Curless and M. Levoy. <u>A Volumetric Method for Building Complex</u> Models from Range Images, SIGGRAPH 1996.
- S. Rusinkiewicz and O. Hall-Holt and M. Levoy. Real-Time 3D Model Acquisition, SIGGRAPH 2001.
- H. Sawhney, Y. Guo, K. Hanna, R. Kumar, S. Adkins and S. Zhou.
 <u>Hybrid Stereo Camera: An IBR Approach for Synthesis of Very High</u>
 Resolution Stereoscopic Image Sequences, SIGGRAPH 2001.
- C. L. Zitnick, S. B. Kang, M. Uyttendaele, S. Winder and R. Szeliski.
 <u>High-quality video view interpolation using a layered representation</u>,
 SIGGRAPH 2004.