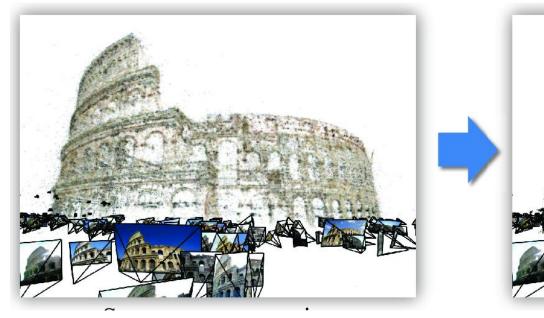
CS5670: Computer Vision

Noah Snavely

Multi-view stereo





Announcements

 Project 3 ("Autostitch") due Monday 4/17 by 11:59pm

Recommended Reading

Szeliski Chapter 11.6

Multi-View Stereo: A Tutorial

Furukawa and Hernandez, 2015

http://www.cse.wustl.edu/~furukawa/papers/fnt mvs.pdf



Stereo



Multi-view stereo



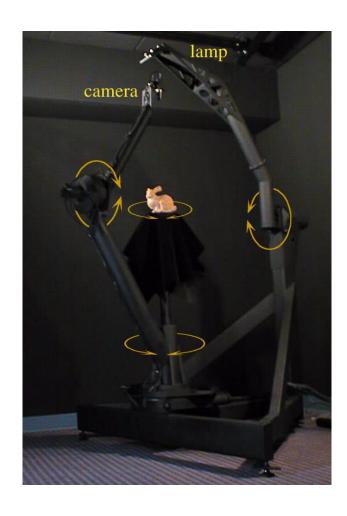
Point Grey's Bumblebee XB3

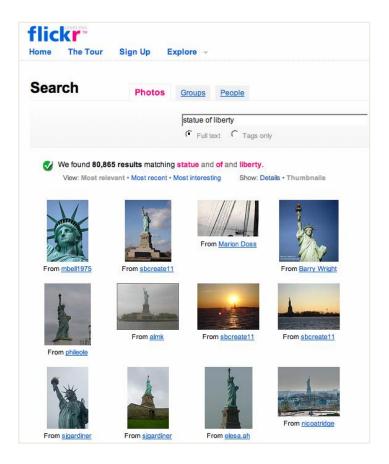


Point Grey's ProFusion 25



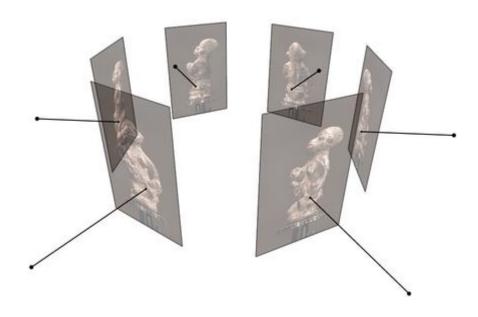
CMU's 3D Room





Input: calibrated images from several viewpoints

Output: 3D object model



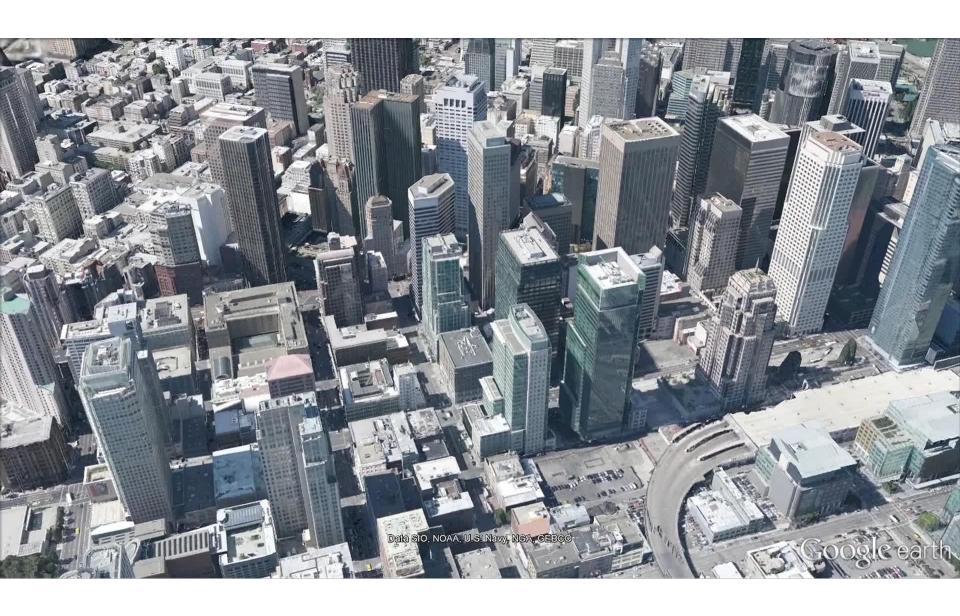
Figures by Carlos Hernandez



Goesele et al.



Applications

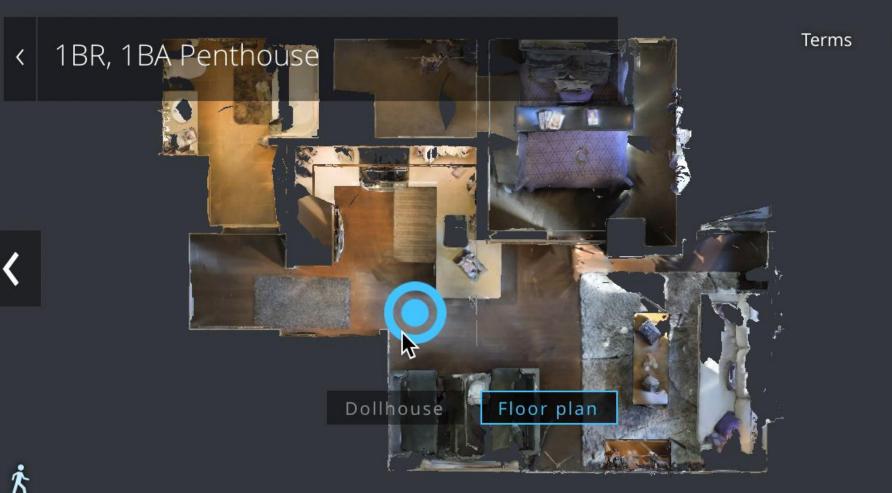


















Whistle in the Form of Female Figure 600 AD - 500 AD



Los Angeles County Museum of Art





Los Angeles County Museum of Art



Sculpture



Mexico

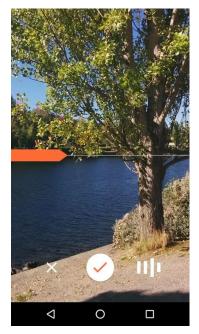




Google





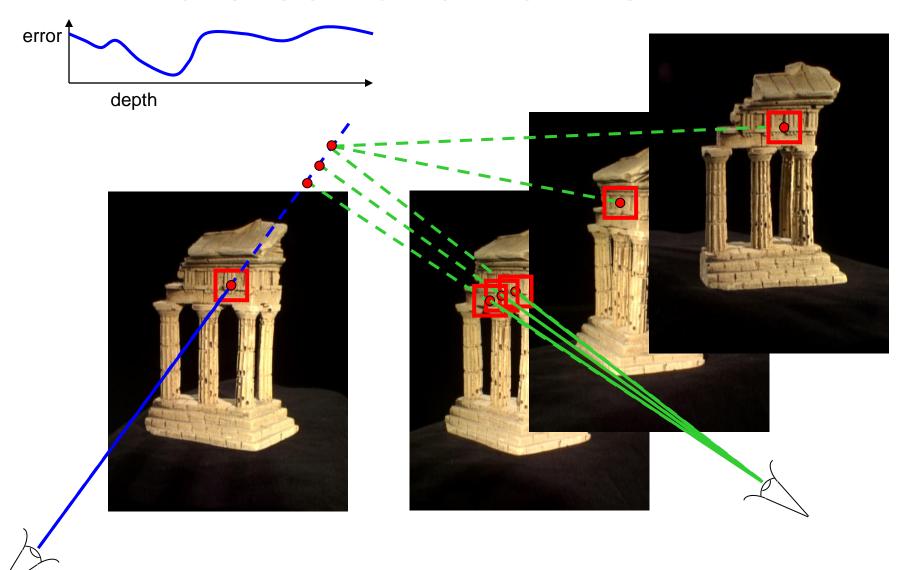








Stereo: another view



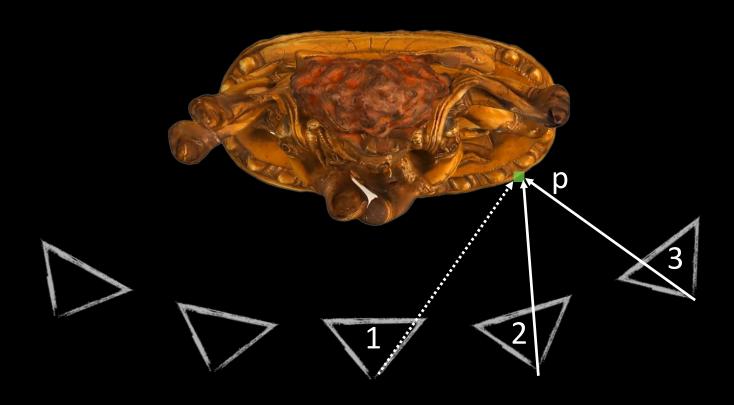




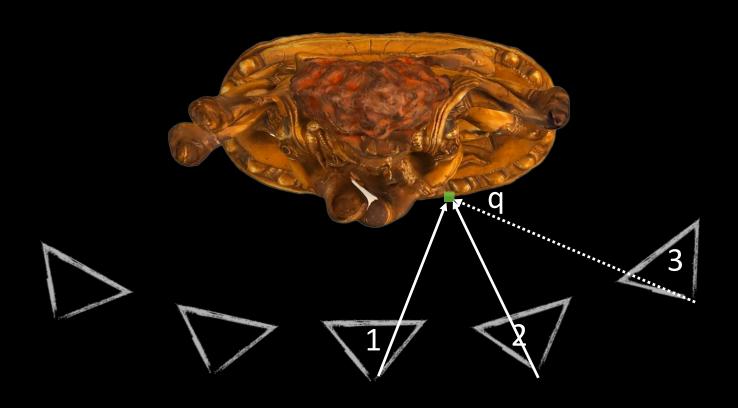


Why MVS?

- Different points on the object's surface will be more clearly visible in some subset of cameras
 - Could have high-res closeups of some regions
 - Some surfaces are foreshortened from certain views



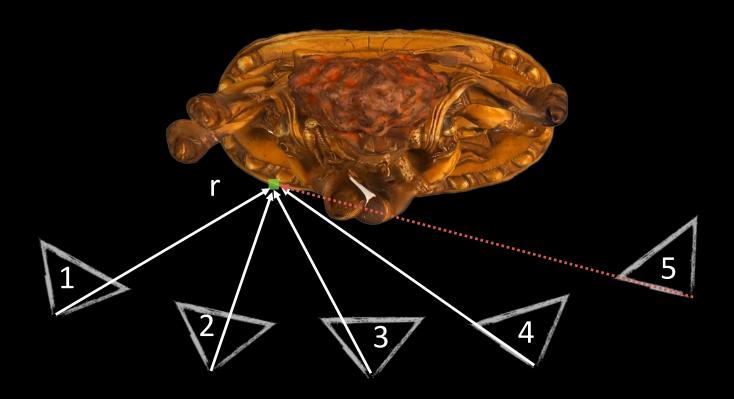
Cameras 2 and 3 can more clearly see point p.



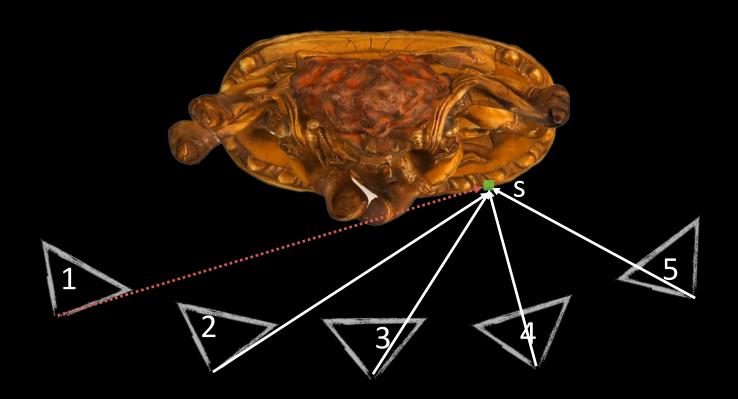
Cameras 1 and 2 can more clearly see point q.

Why MVS?

- Different points on the object's surface will be more clearly visible in some subset of cameras
 - Could have high res closeups of some regions
 - Some surfaces are foreshortened from certain views
- Some points may be occluded entirely in certain views



Camera 5 can't see point r.



Camera 1 can't see point s.

Why MVS?

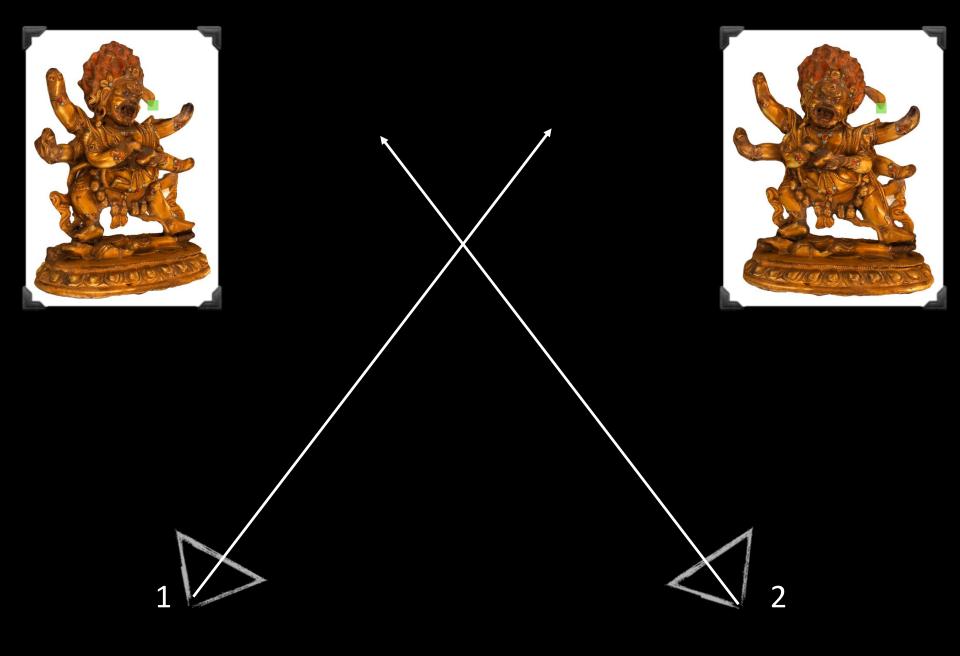
- Different points on the object's surface will be more clearly visible in some subset of cameras
 - Could have high res closeups of some regions
 - Some surfaces are foreshortened from certain views
- Some points may be occluded entirely in certain views
- More measurements per point can reduce error

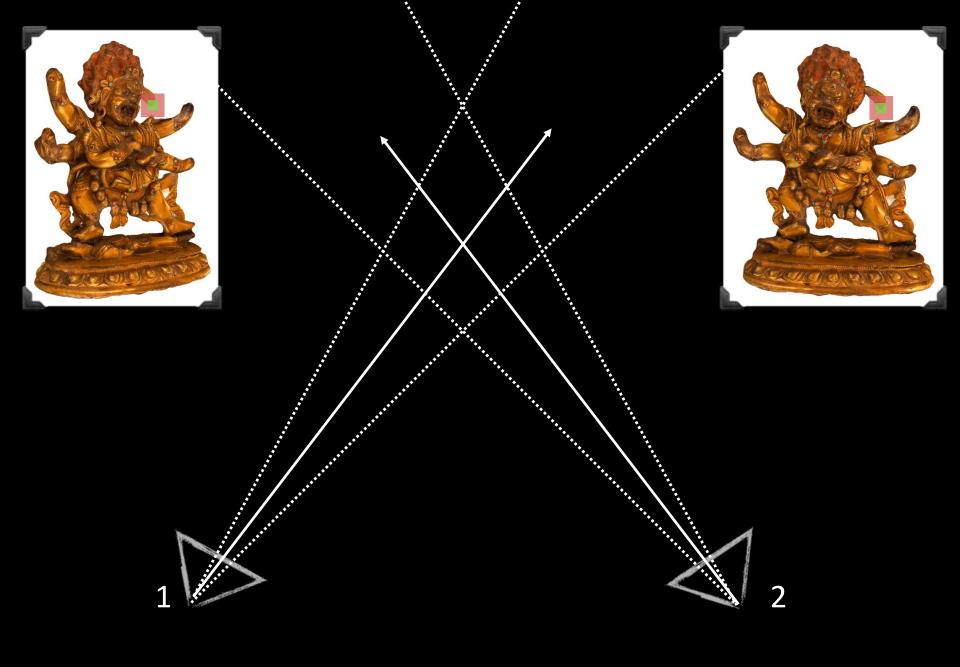




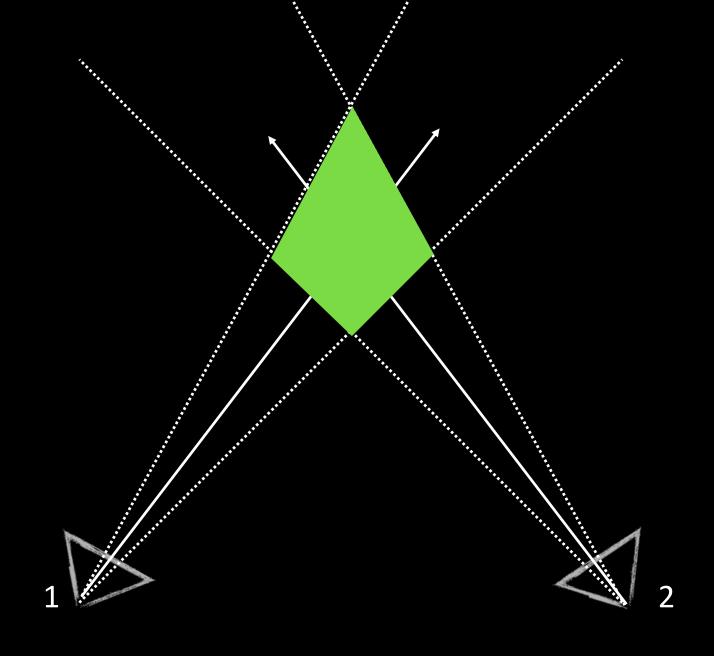




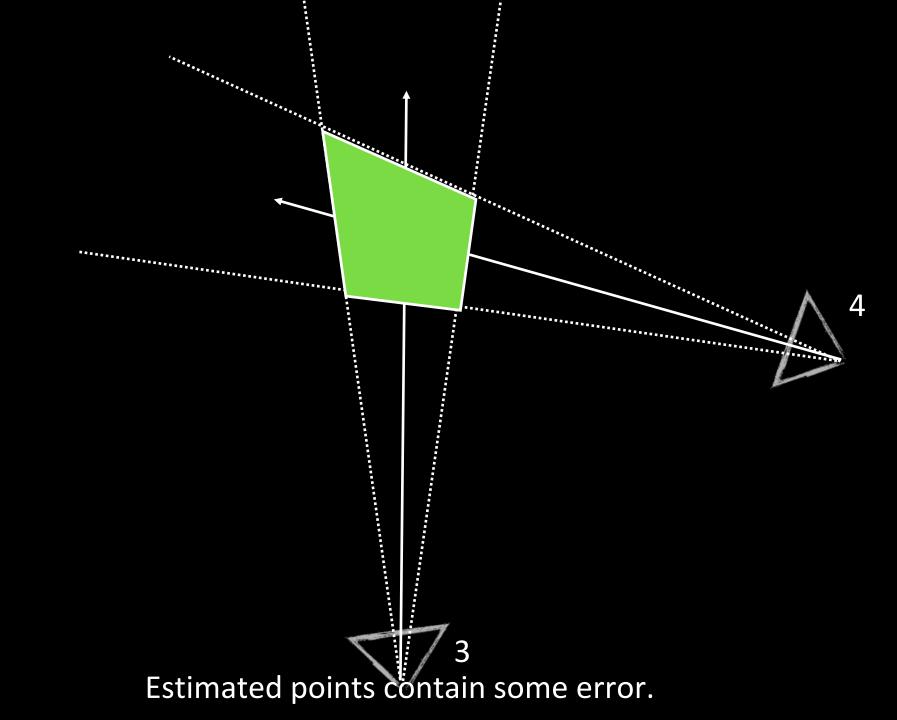


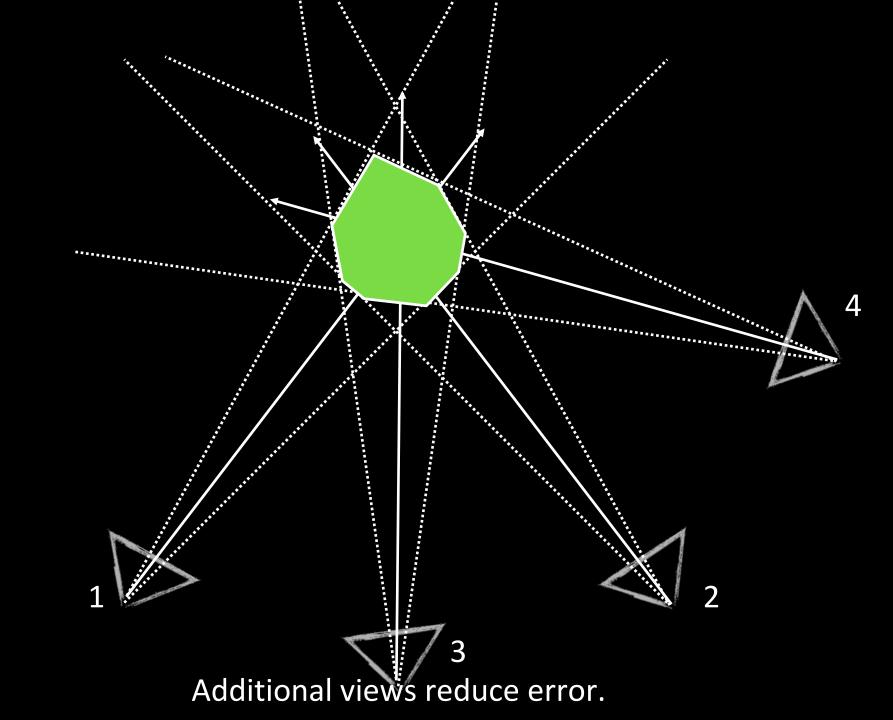


Estimated points contain some error.

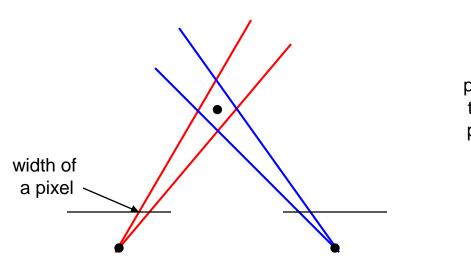


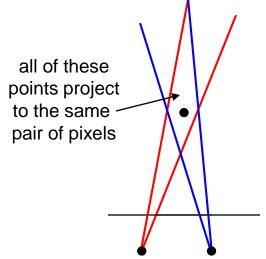
Estimated points contain some error.





Choosing the stereo baseline





Large Baseline

Small Baseline

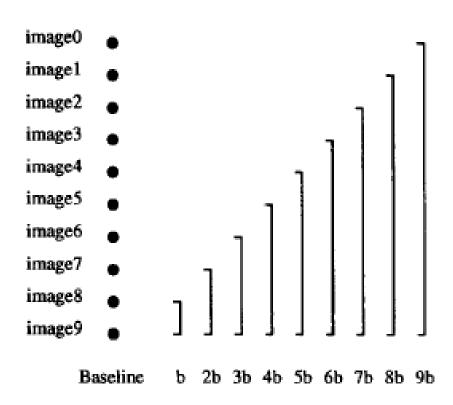
What's the optimal baseline?

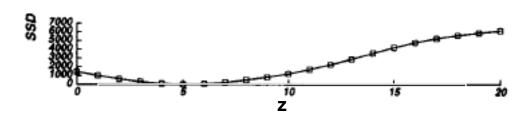
- Too small: large depth error
- Too large: difficult search problem

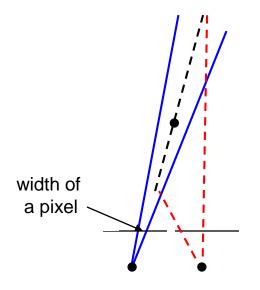
The Effect of Baseline on Depth Estimation



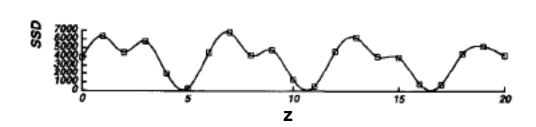
Figure 2: An example scene. The grid pattern in the background has ambiguity of matching.

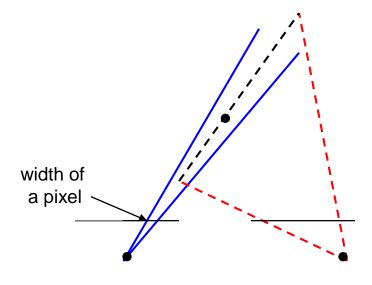






pixel matching score





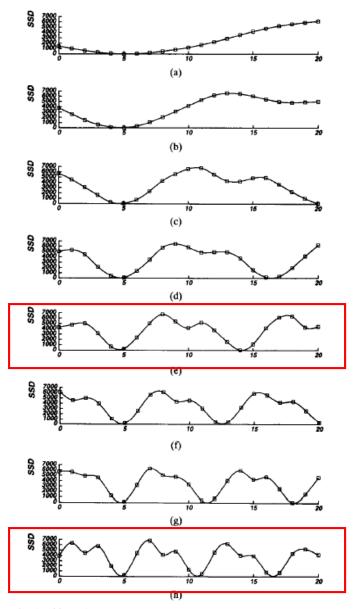


Fig. 5. SSD values versus inverse distance: (a) B=b; (b) B=2b; (c) B=3b; (d) B=4b; (e) B=5b; (f) B=6b; (g) B=7b; (h) B=8b. The horizontal axis is normalized such that 8bF=1.

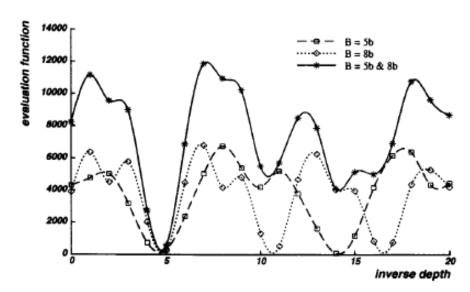


Fig. 6. Combining two stereo pairs with different baselines.

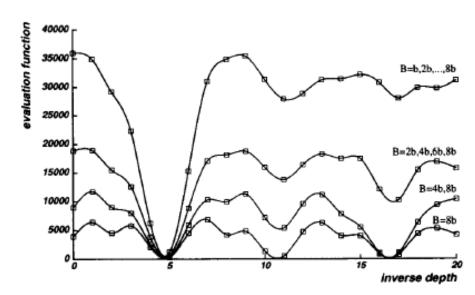


Fig. 7. Combining multiple baseline stereo pairs.

Multibaseline Stereo

Basic Approach

- Choose a reference view
- Use your favorite stereo algorithm BUT
 - replace two-view SSD with SSSD over all baselines

Limitations



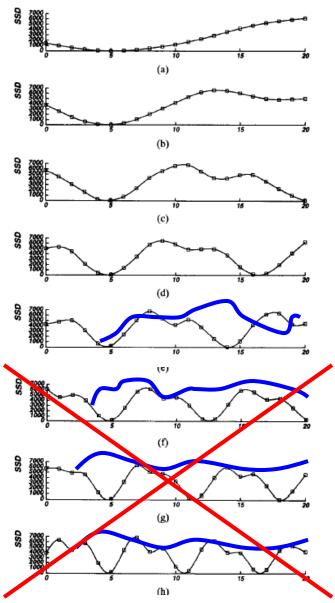


Fig. 5. SSD values versus inverse distance: (a) B=b; (b) B=2b; (c) B=3b; (d) B=4b; (e) B=5b; (f) B=6b; (g) B=7b; (h) B=8b. The horizontal axis is normalized such that 8bF=1.

Problem: visibility

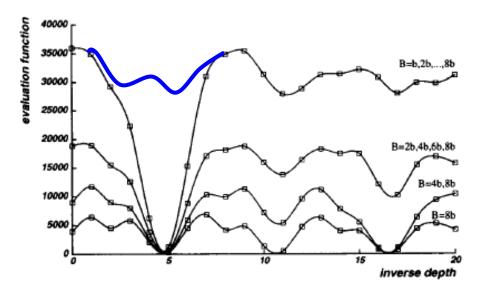


Fig. 7. Combining multiple baseline stereo pairs.

Some Solutions

- Match only nearby photos [Narayanan 98]
- Use NCC instead of SSD,
 Ignore NCC values > threshold
 [Hernandez & Schmitt 03]

Popular matching scores

SSD (Sum Squared Distance)

$$\sum_{x,y} |W_1(x,y) - W_2(x,y)|^2$$

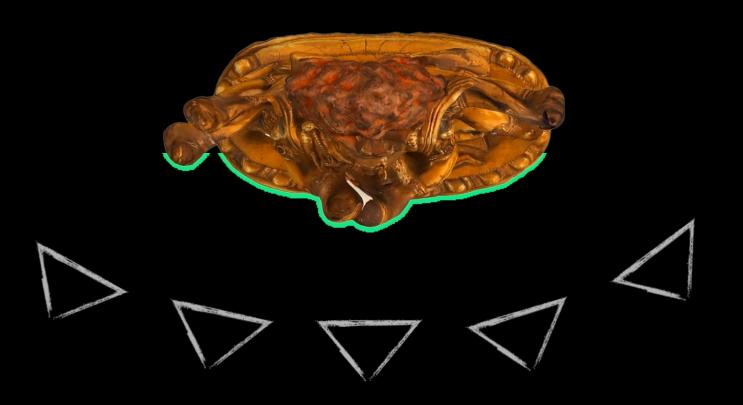
NCC (Normalized Cross Correlation)

$$\frac{\sum_{x,y} (W_1(x,y) - \overline{W_1})(W_2(x,y) - \overline{W_2})}{\sigma_{W_1} \sigma_{W_2}}$$

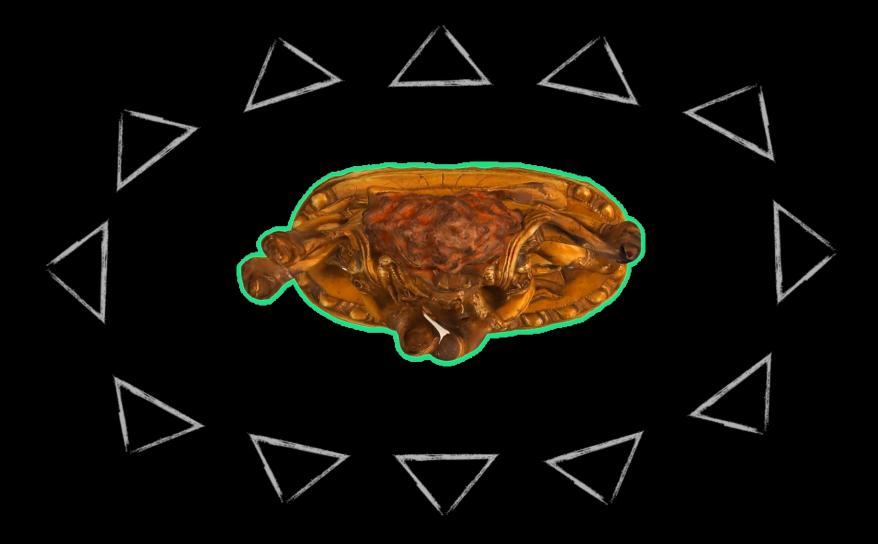
$$- \ \text{where} \frac{\overline{W_i}}{n} = \frac{1}{n} \sum_{x,y} W_i \qquad \sigma_{W_i} = \sqrt{\frac{1}{n} \sum_{x,y} (W_i - \overline{W_i})^2}$$

– what advantages might NCC have?

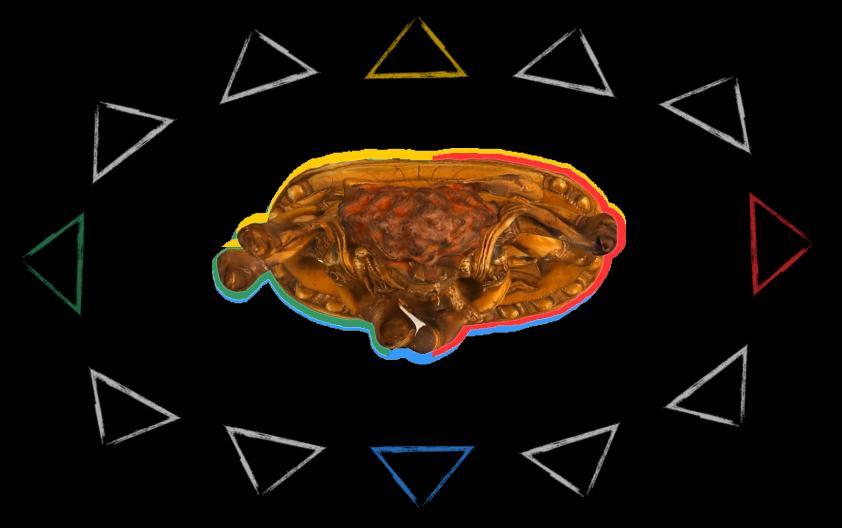
Single depth map often isn't enough



Really want full coverage

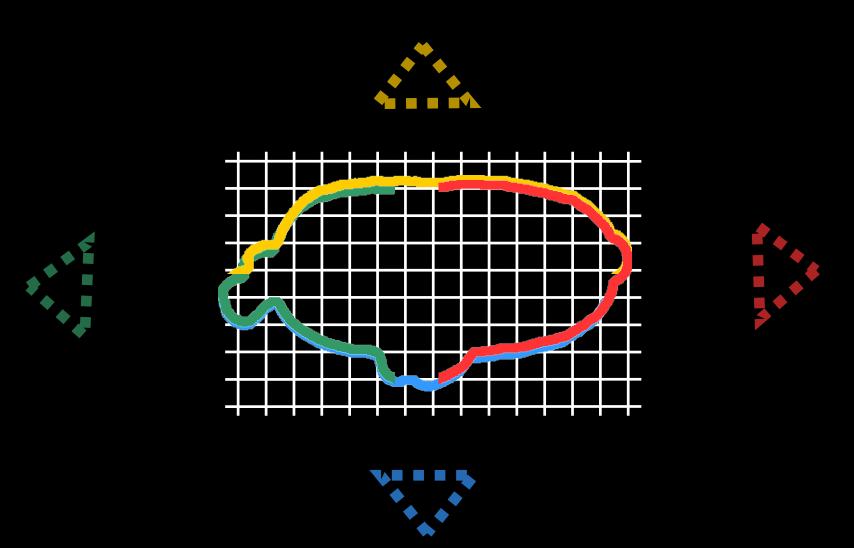


Idea: Combine many depth maps



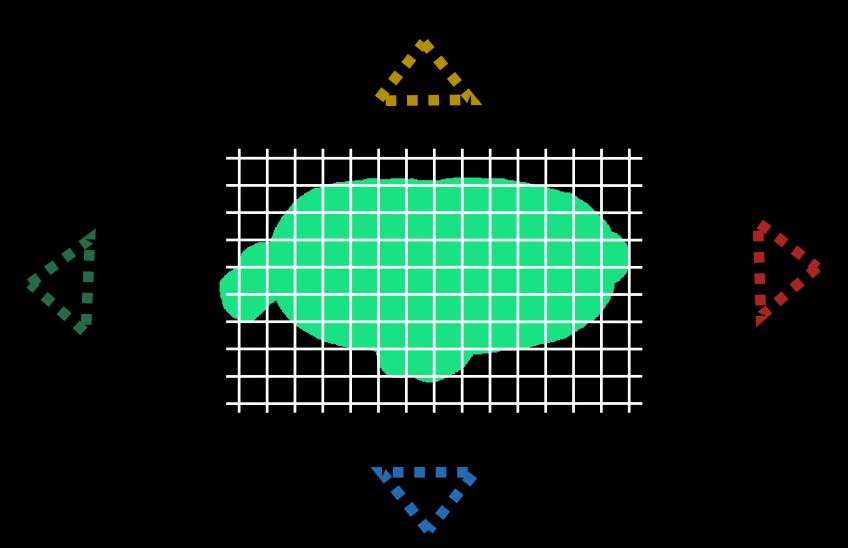
Many depth maps, each with error. How can we fuse these?

Volumetric fusion

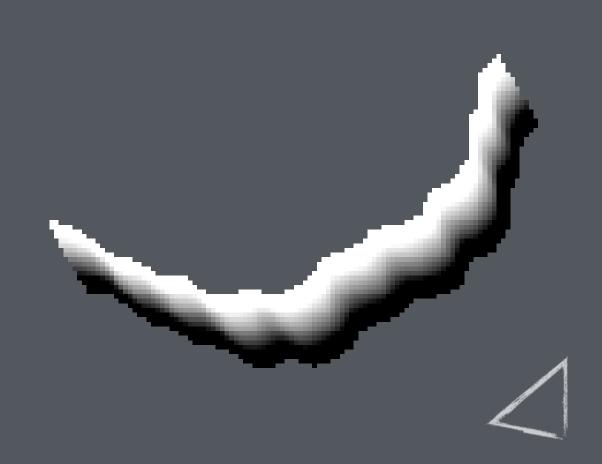


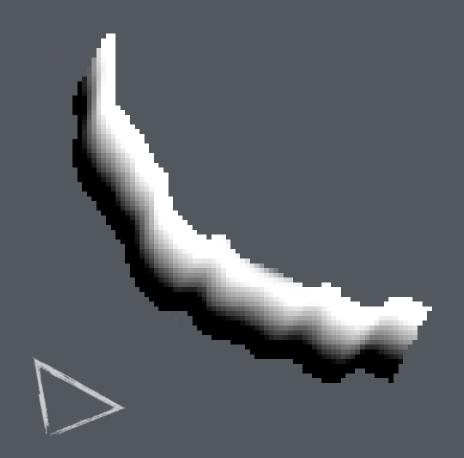
A common world-space coordinate system.

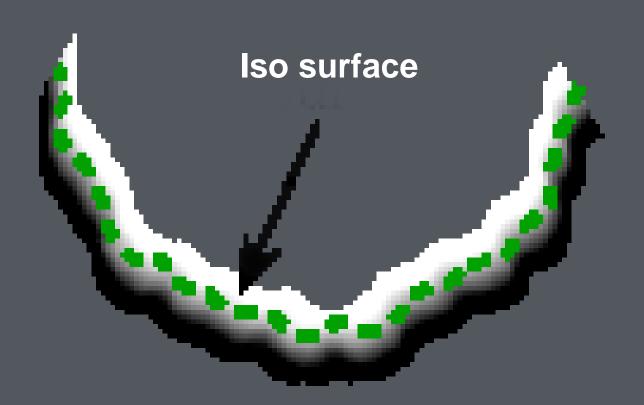
Volumetric fusion



A common world-space coordinate system.







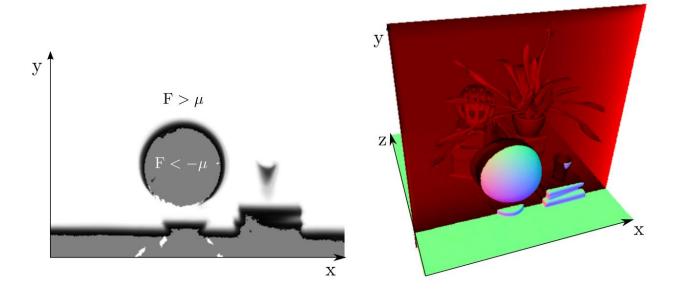
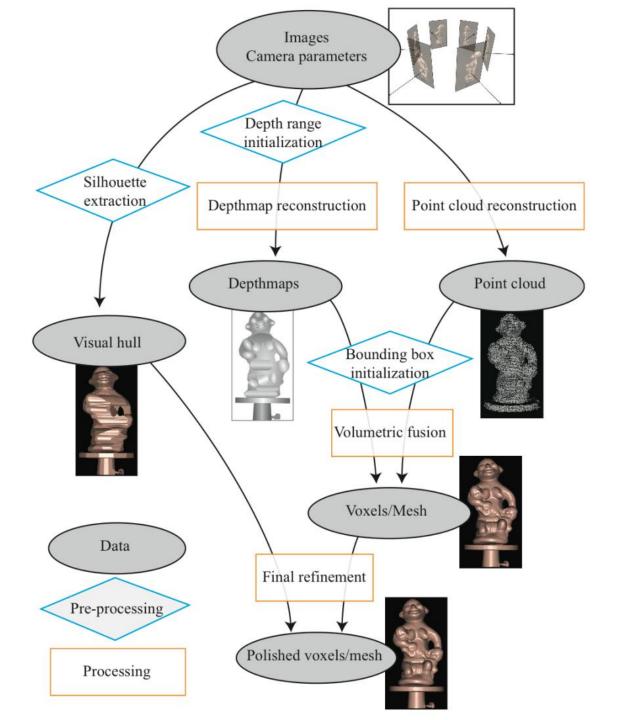
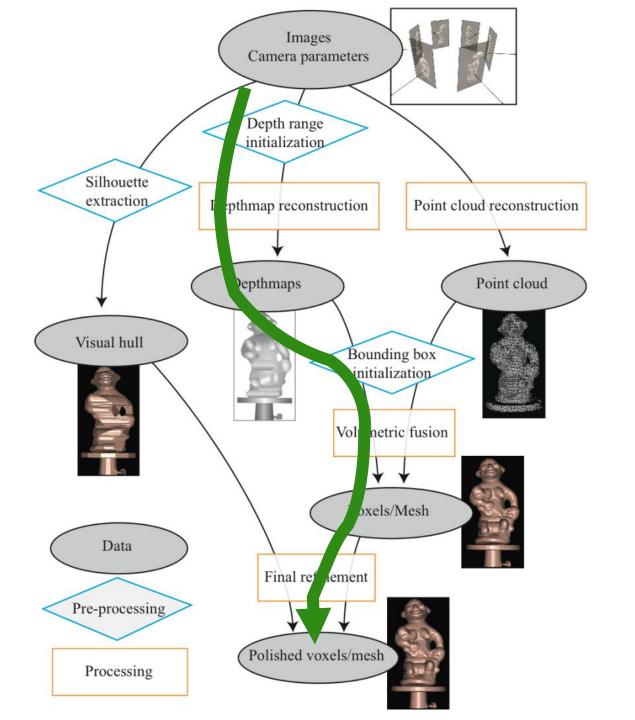
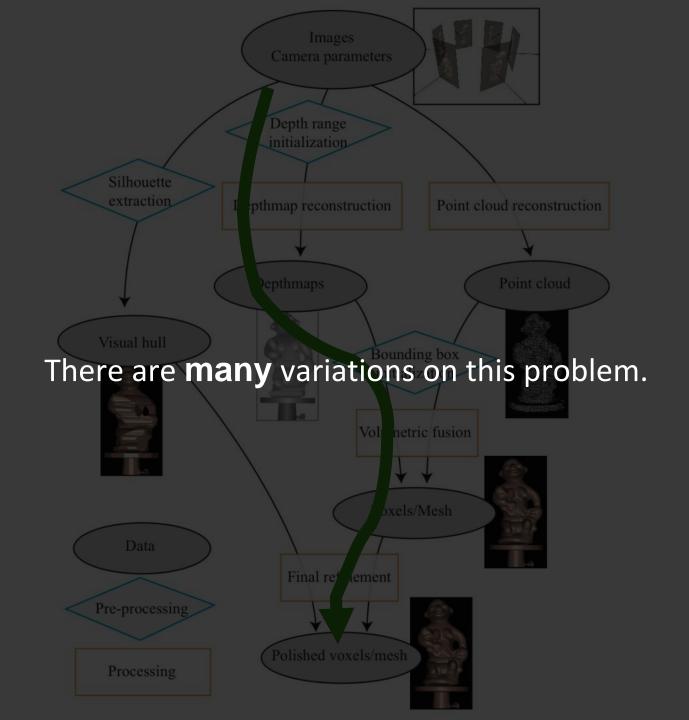


Figure 4: A slice through the truncated signed distance volume showing the truncated function $F > \mu$ (white), the smooth distance field around the surface interface F = 0 and voxels that have not yet had a valid measurement(grey) as detailed in eqn. 9.







Questions?

Are depth maps enough?



Are depth maps enough?



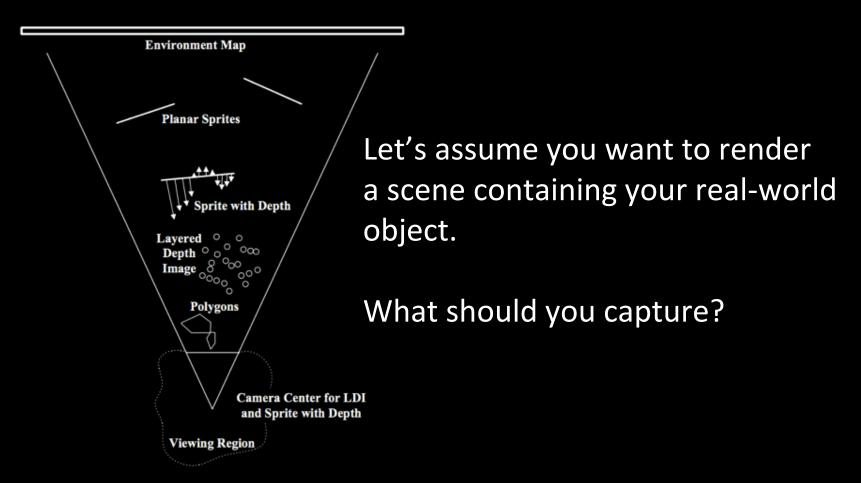
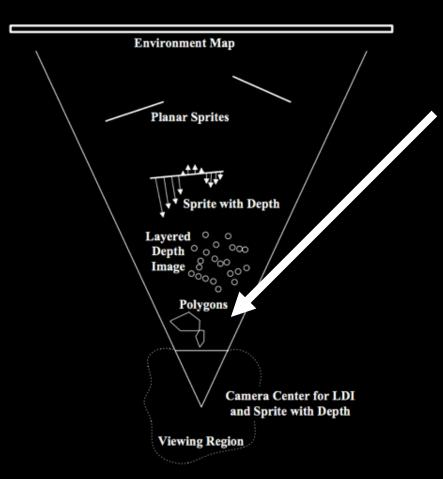


Figure 1 Different image based primitives can serve well depending on distance from the camera



The object is right in front of you and you want to look all around it.

Figure 1 Different image based primitives can serve well depending on distance from the camera

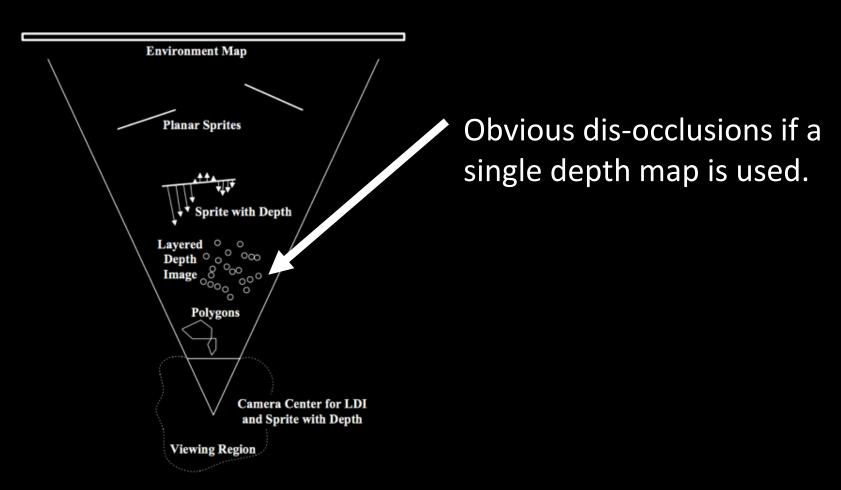
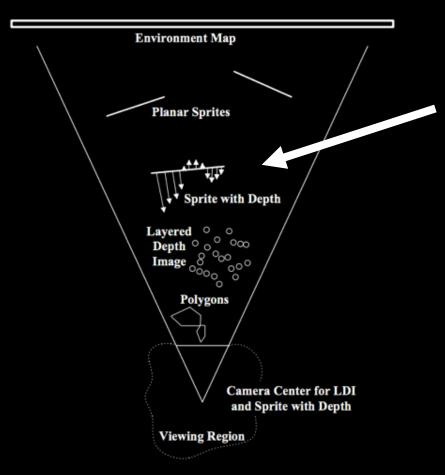


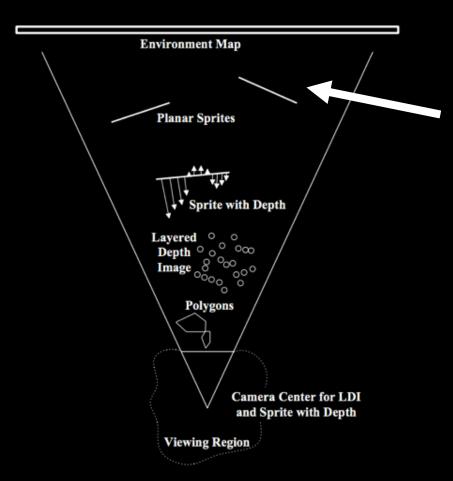
Figure 1 Different image based primitives can serve well depending on distance from the camera





Close enough that you can tell it shouldn't be flat, but **dis-occlusions** are minimal.

Figure 1 Different image based primitives can serve well depending on distance from the camera



Far enough away that a plane can approximate it.

Figure 1 Different image based primitives can serve well depending on distance from the camera

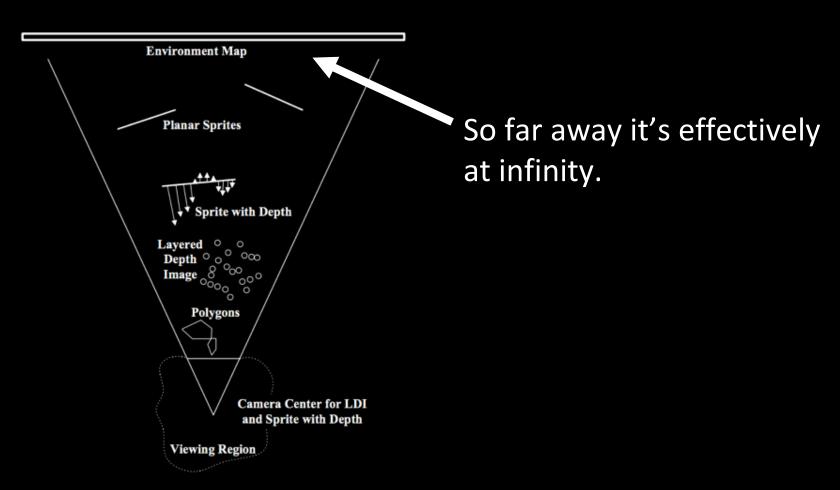
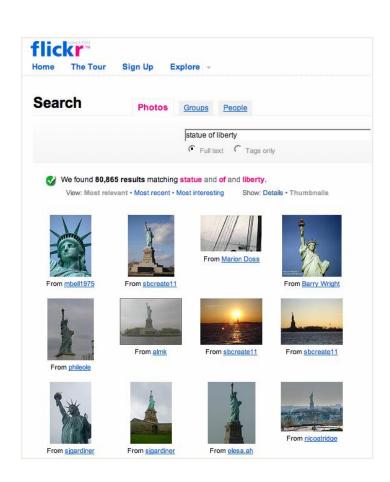


Figure 1 Different image based primitives can serve well depending on distance from the camera

Questions?

Multi-view stereo from Internet Collections

[Goesele, Snavely, Curless, Hoppe, Seitz, ICCV 2007]







Challenges

appearance variation









resolution





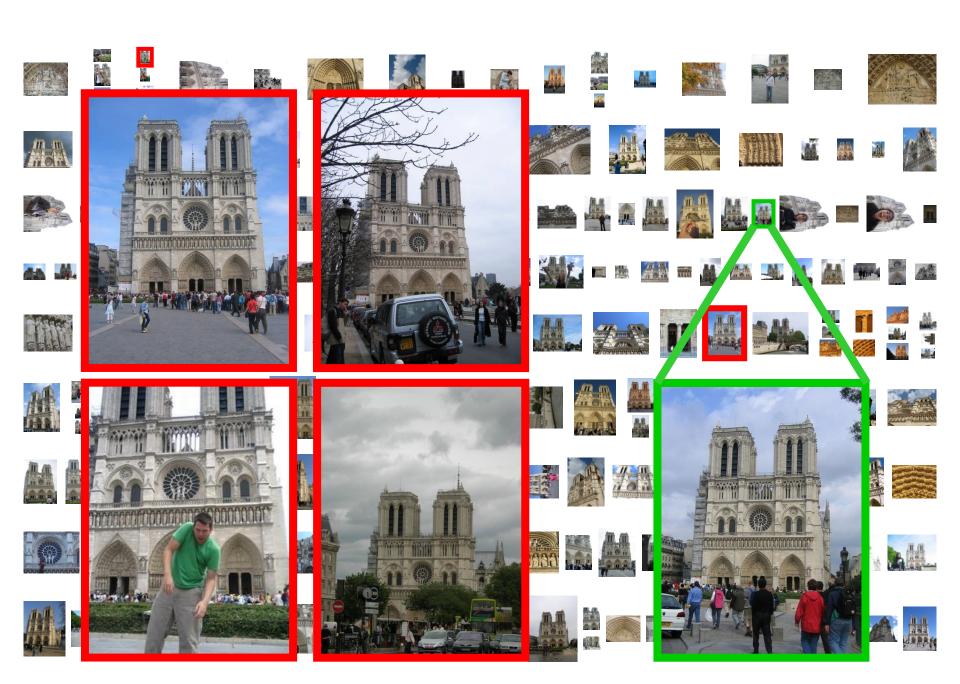






massive collections

82,754 results for photos matching notre and dame and paris.























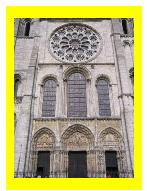




reference view

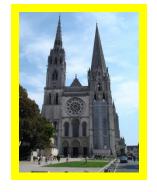
- Automatically select neighboring views for each point in the image
- Desiderata: good matches AND good baselines























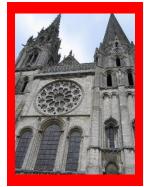


reference view

- Automatically select neighboring views for each point in the image
- Desiderata: good matches AND good baselines





















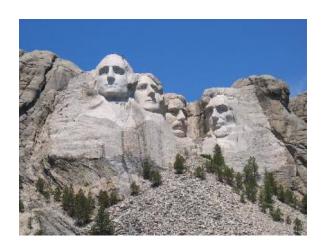


reference view

Local view selection

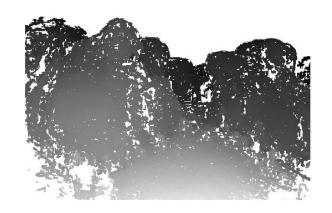
- Automatically select neighboring views for each point in the image
- Desiderata: good matches AND good baselines

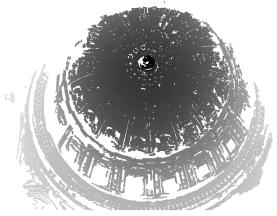
Results









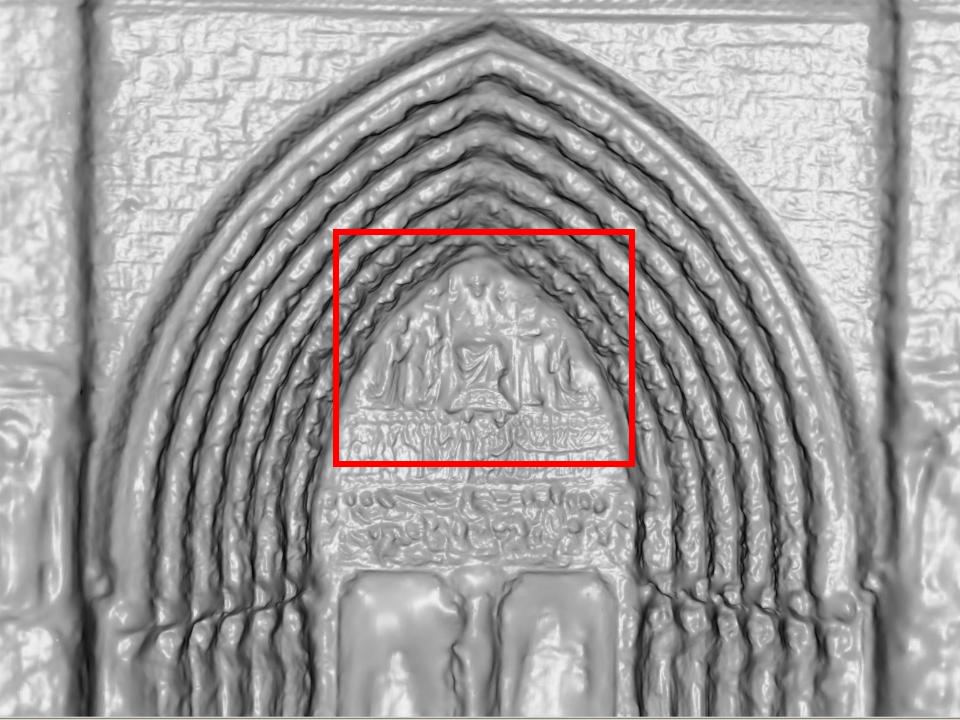


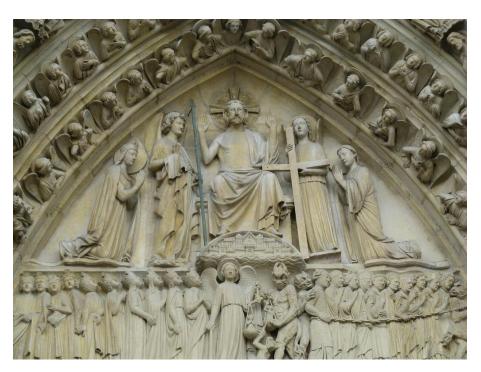


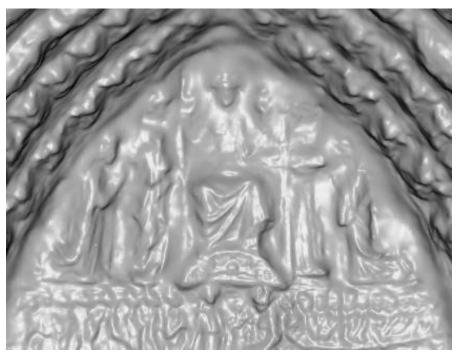
Notre Dame de Paris

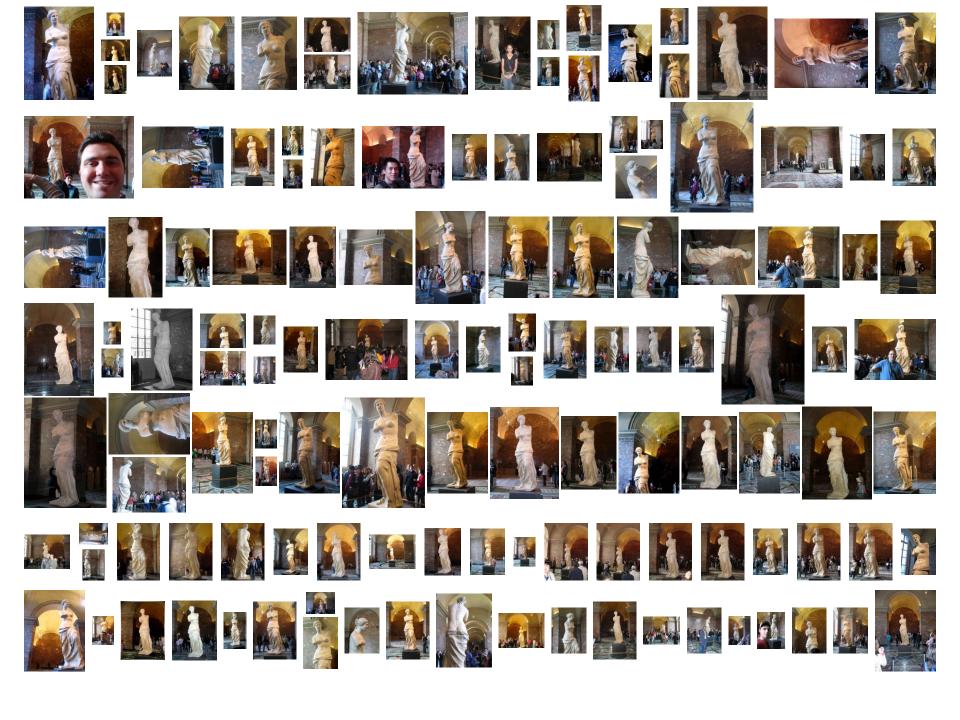
653 images 313 photographers





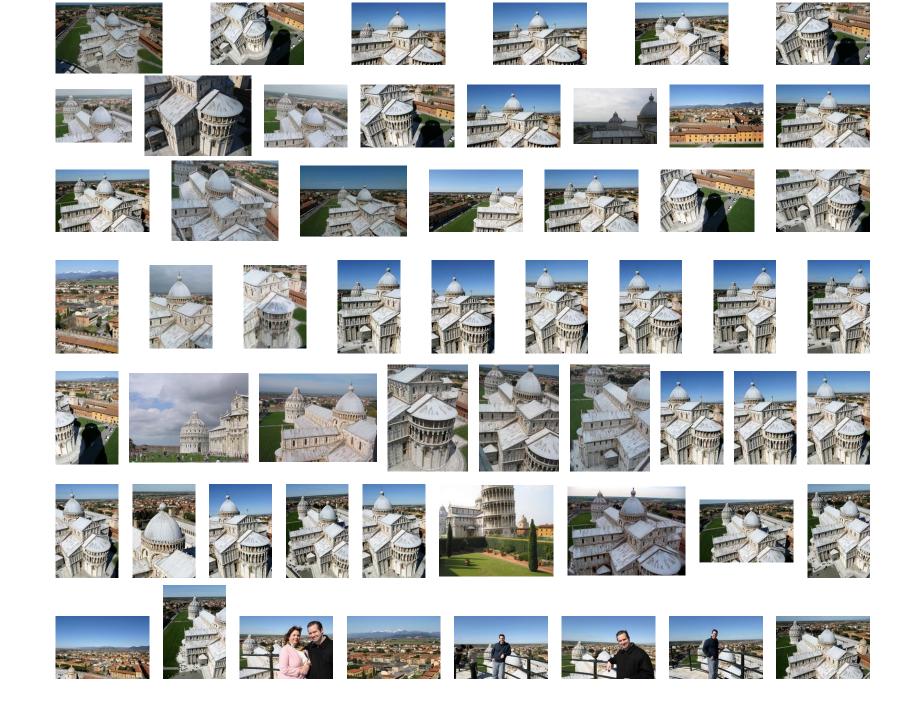




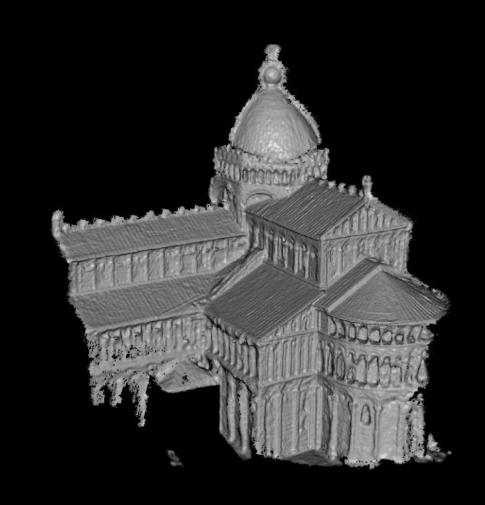




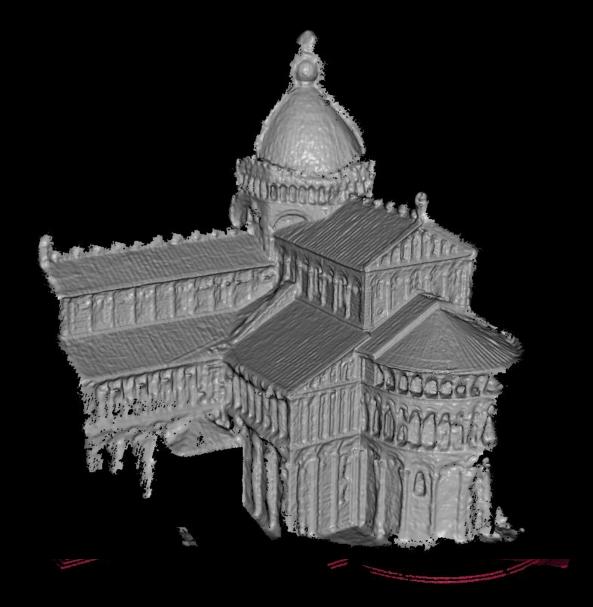
merged model of Venus de Milo







merged model of Pisa Cathedral



Accuracy compared to laser scanned model: 90% of points within 0.25% of ground truth