CS5670: Computer Vision Noah Snavely

Multi-view stereo



Announcements

• Project 4 due April 22, 2019, by 11:59pm

• Final quiz in-class next Wednesday, April 24

• In-class final exam, May 6

Recommended Reading

• Szeliski Chapter 11.6

- *Multi-View Stereo: A Tutorial*, Furukawa and Hernandez, 2015
 - <u>http://carlos-hernandez.org/papers/fnt_mvs_2015.pdf</u>

Multi-view Stereo

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





Multi-view stereo

Stereo

Multi-view Stereo



Point Grey's Bumblebee XB3



Point Grey's ProFusion 25



CMU's <u>3D Room</u>

Multi-view Stereo

Input: calibrated images from several viewpoints (known intrinsics and extrinsics)

Output: 3D object model



Figures by Carlos Hernandez

Applications







L .J



Renderpeople

E

THE RENDERPEOPLE MISSION

IMPROVING THE QUALITY

https://renderpeople.com/about-us/



Google











https://code.facebook.com/posts/1755691291326688/introducingfacebook-surround-360-an-open-high-quality-3d-360-video-capturesystem?hc_location=ufi

f



reference view

neighbor views



reference view

neighbor views



reference view

neighbor views



In this manner, solve for a depth map over the whole reference view





reference view

neighbor views

Multi-view stereo: advantages

- Can match windows using more than 1 neighbor, giving a **stronger match signal**
- If you have lots of potential neighbors, can choose the best subset of neighbors to match per reference image
- Can reconstruct a depth map for each reference frame, and the merge into a complete 3D model



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.





 I_2





precise due to narrow triangulation



 For larger baselines^z, must search larger area in second image

M. Okutomi and T. Kanade, <u>"A Multiple-Baseline Stereo System,"</u> IEEE Trans. on Pattern Analysis and Machine Intelligence, 15(4):353-363(1993).



Multiple-baseline stereo



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.

Multiple-baseline stereo results



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

Multibaseline Stereo

Basic Approach

- Choose a reference view
- Use your favorite stereo algorithm BUT
 - replace two-view SSD with **SSSD** over all baselines
 - **SSSD**: the SSD values are computed first for each pair of stereo images, and then add all together from multiple stereo pairs.

Limitations

- Only gives a depth map (not an "object model")
- Won't work for widely distributed views.



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)

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

- SAD (Sum of Absolute Difference) $\sum_{x,y} |W_1(x,y) W_2(x,y)|$
- ZNCC (Zero-mean Normalized Cross Correlation)

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

$$\sigma_{W_1} \sigma_{W_2}$$

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

– what advantages might NCC have?

Questions?







- Sweep family of planes parallel to the reference camera image plane
- Reproject neighbors onto each plane (via homography) and compare reprojections





Left neighbor



Reference image



Right neighbor



Left neighbor projected into reference image



Average images on each plane



Right neighbor projected into reference image

Another example



Planar image reprojections swept over depth (averaged)

Cost Volumes -> Depth Maps



Fusing multiple depth maps

- Compute depth map per image
- Fuse the depth maps into a 3D model



Figures by Carlos Hernandez

Questions?