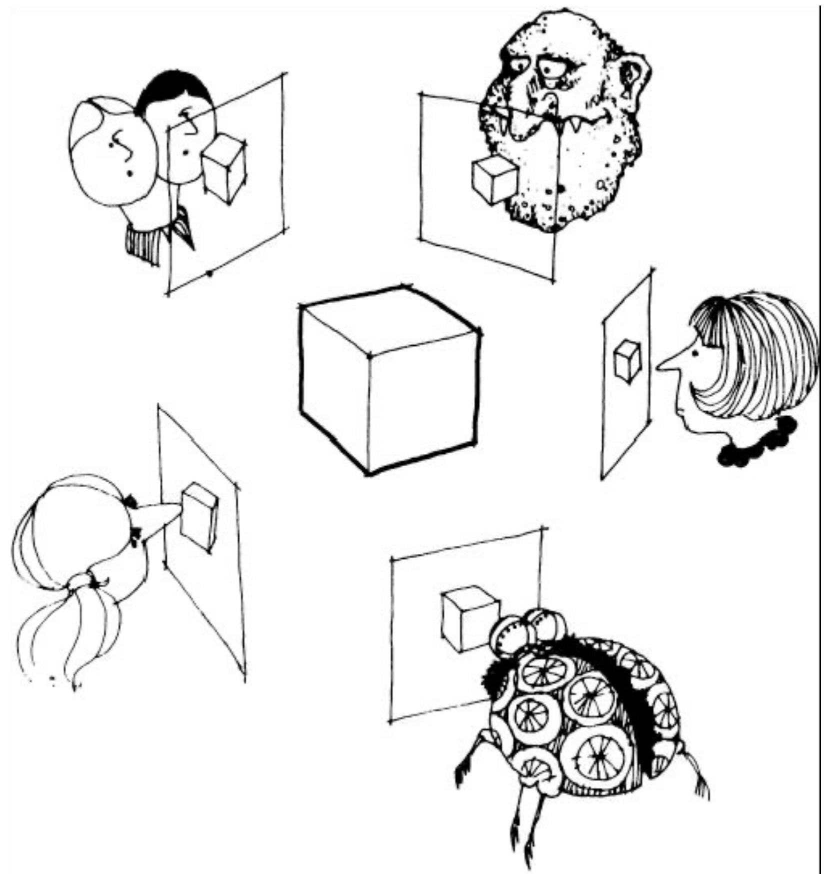
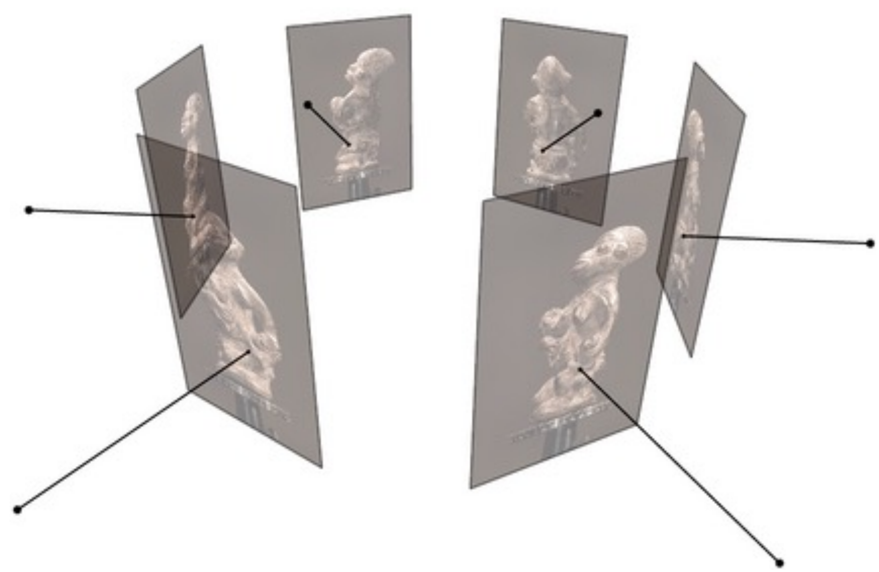
Multi-view stereo



Many slides adapted from S. Seitz, Y. Furukawa, N. Snavely

Multi-view stereo

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



Source: C. Hernandez, N. Snavely

Multi-view stereo

- Goal: 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 (special rig, camera network or video)
  - Calibration may be known or unknown

Multi-view stereo

- Goal: 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 (special rig, camera network or video)
  - Calibration may be known or unknown
- "Representation of 3D shape"
  - Depth maps
  - Meshes
  - Point clouds
  - Patch clouds
  - Volumetric models
  - ….

Outline

- Applications and motivation
- Plane sweep stereo
- Depth map fusion
- Patch-based multi-view stereo (PMVS)
- Stereo from Internet photo collections
- Recent trends

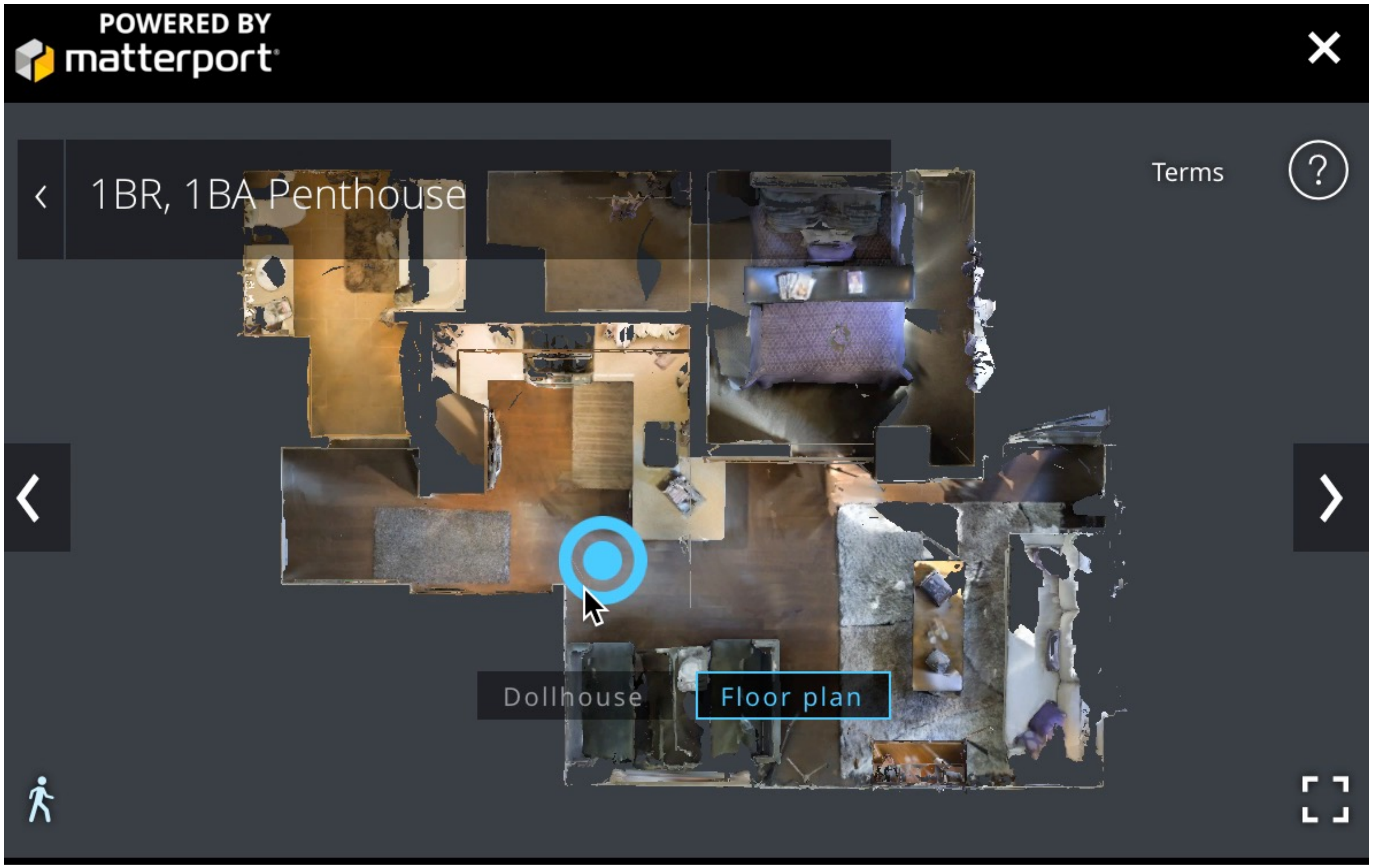
Applications

Source: N. Snavely

Applications

Source: N. Snavely

Applications



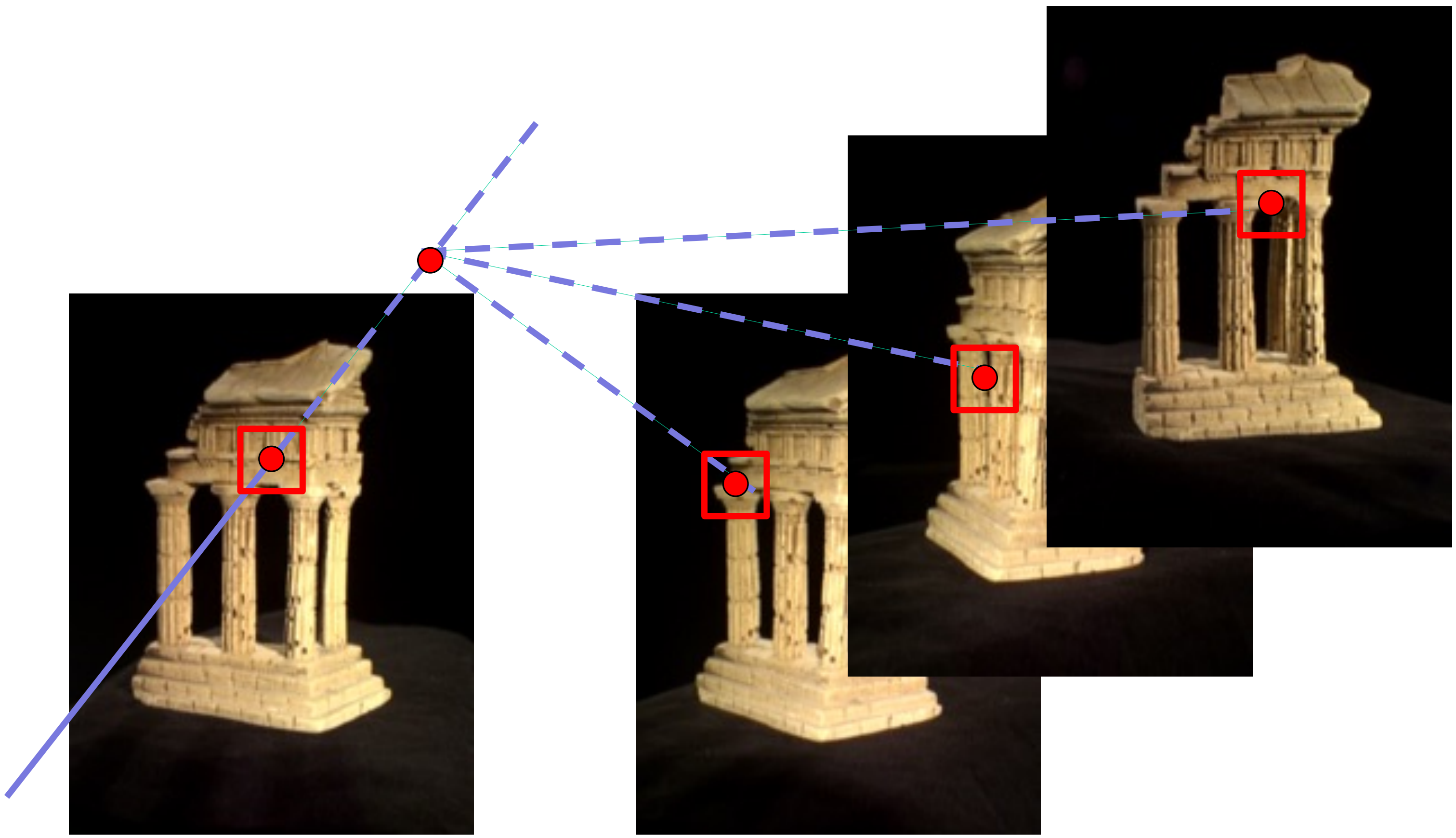
Source: N. Snavely

Applications

- Enable inspection in hard to reach areas with drone photos and 3D reconstruction
- Create 3D model from images
- Provide tools to inspect on images and map interactions to 3D

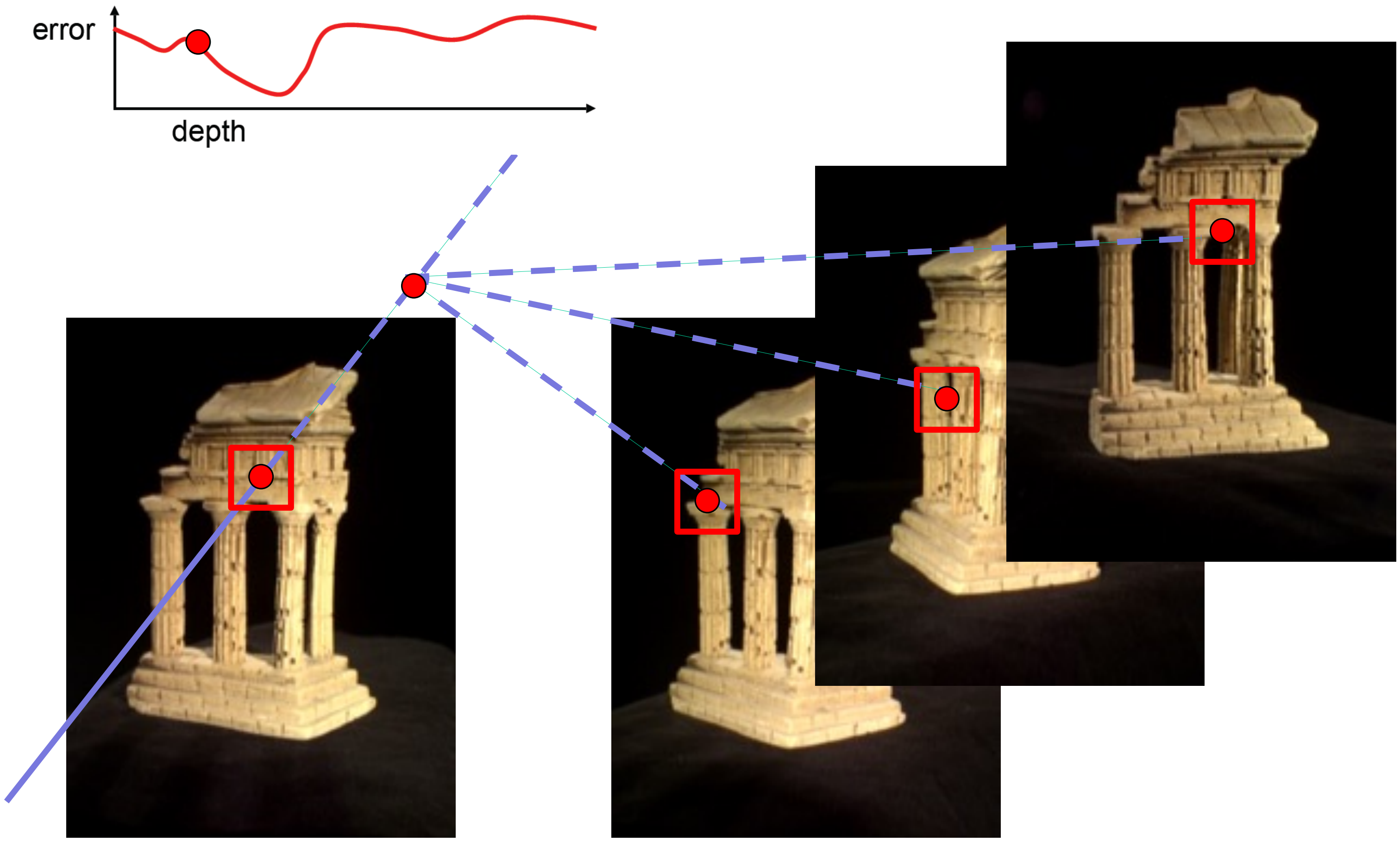
Source: D. Hoiem

Multi-view stereo: Basic idea



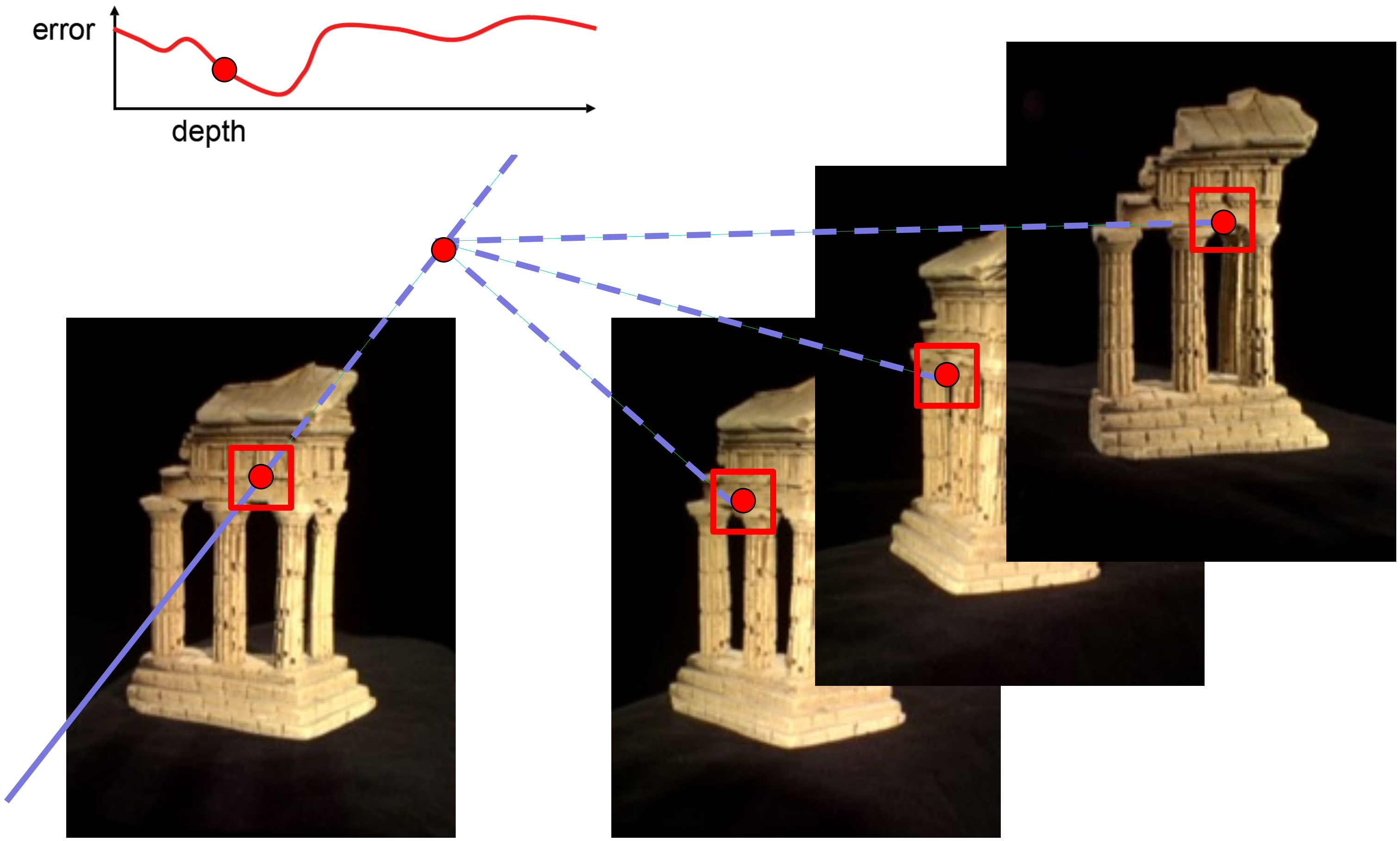
Source: Y. Furukawa

Multi-view stereo: Basic idea



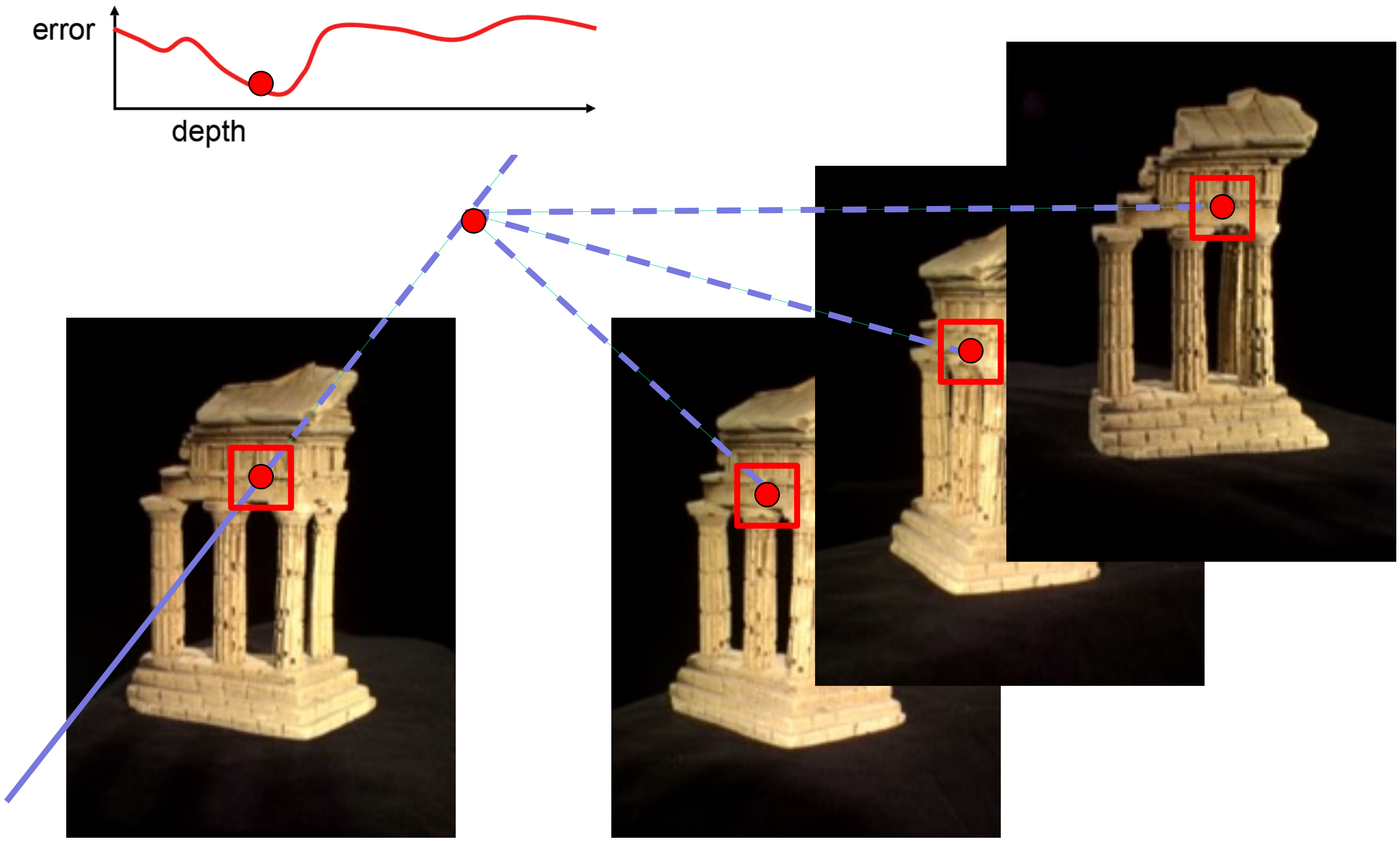
Source: Y. Furukawa

Multi-view stereo: Basic idea



Source: Y. Furukawa

Multi-view stereo: Basic idea

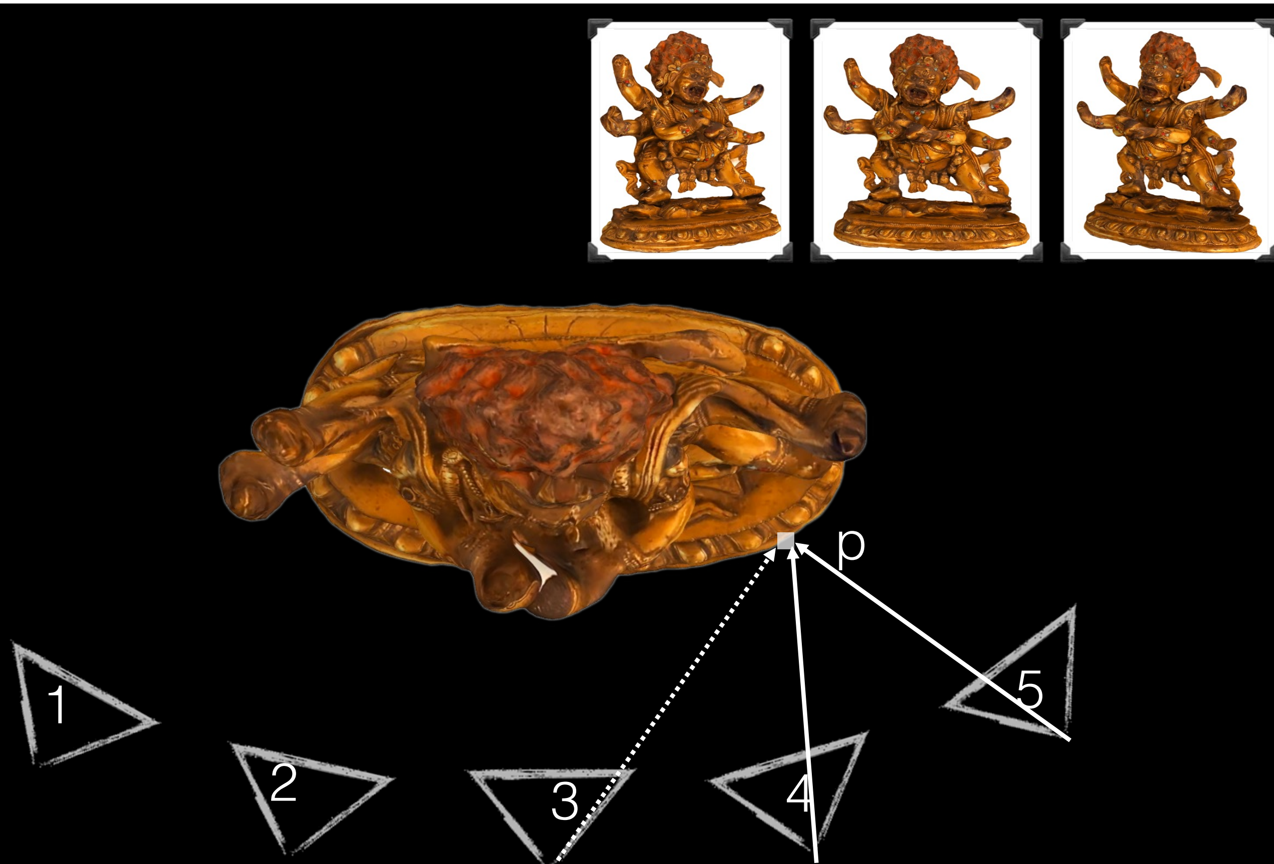


Source: Y. Furukawa

Why MVS?

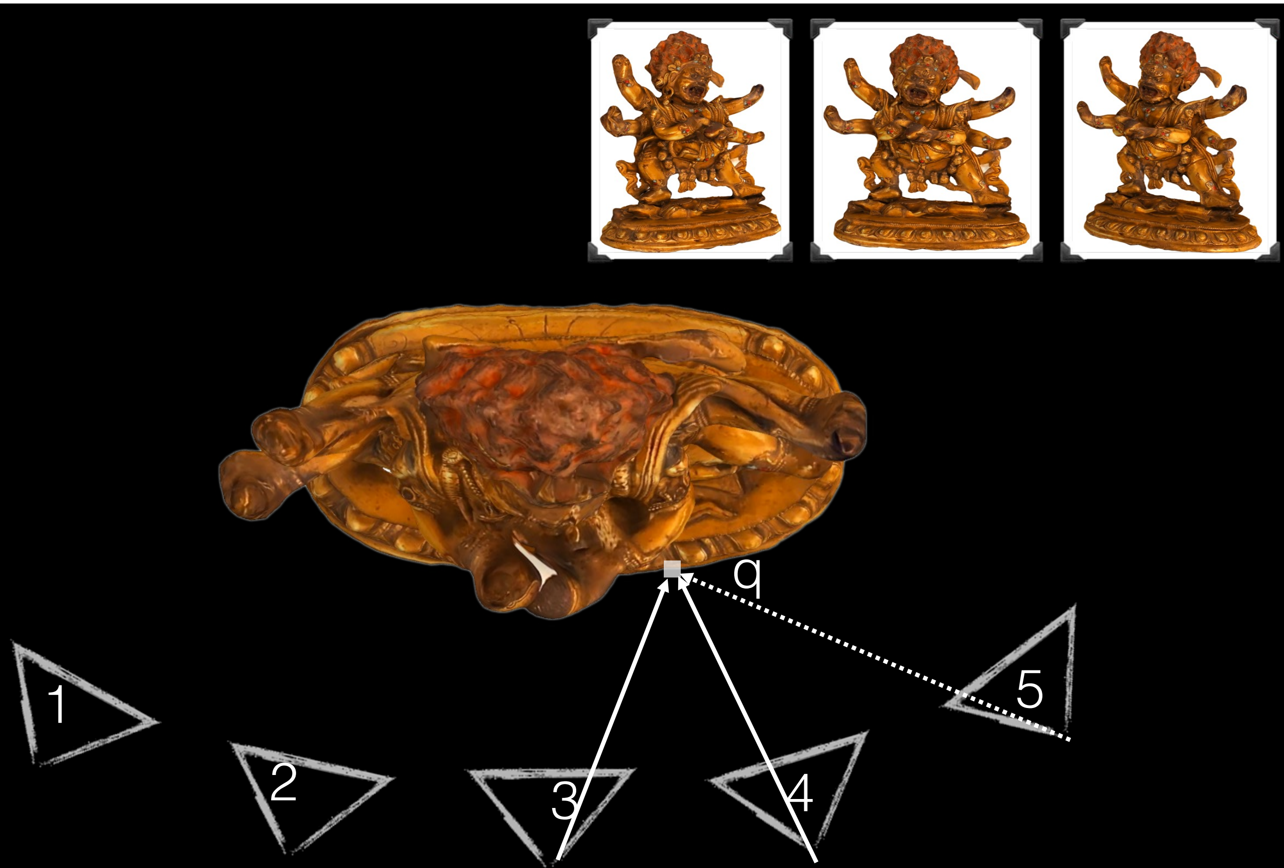
- Different points on the object's surface will be more clearly visible in some subset of cameras
  - Could have high-res closeups of some regions
  - Some surfaces are foreshortened from certain views
  - Some points may be occluded entirely in certain views

Source: N. Snavely



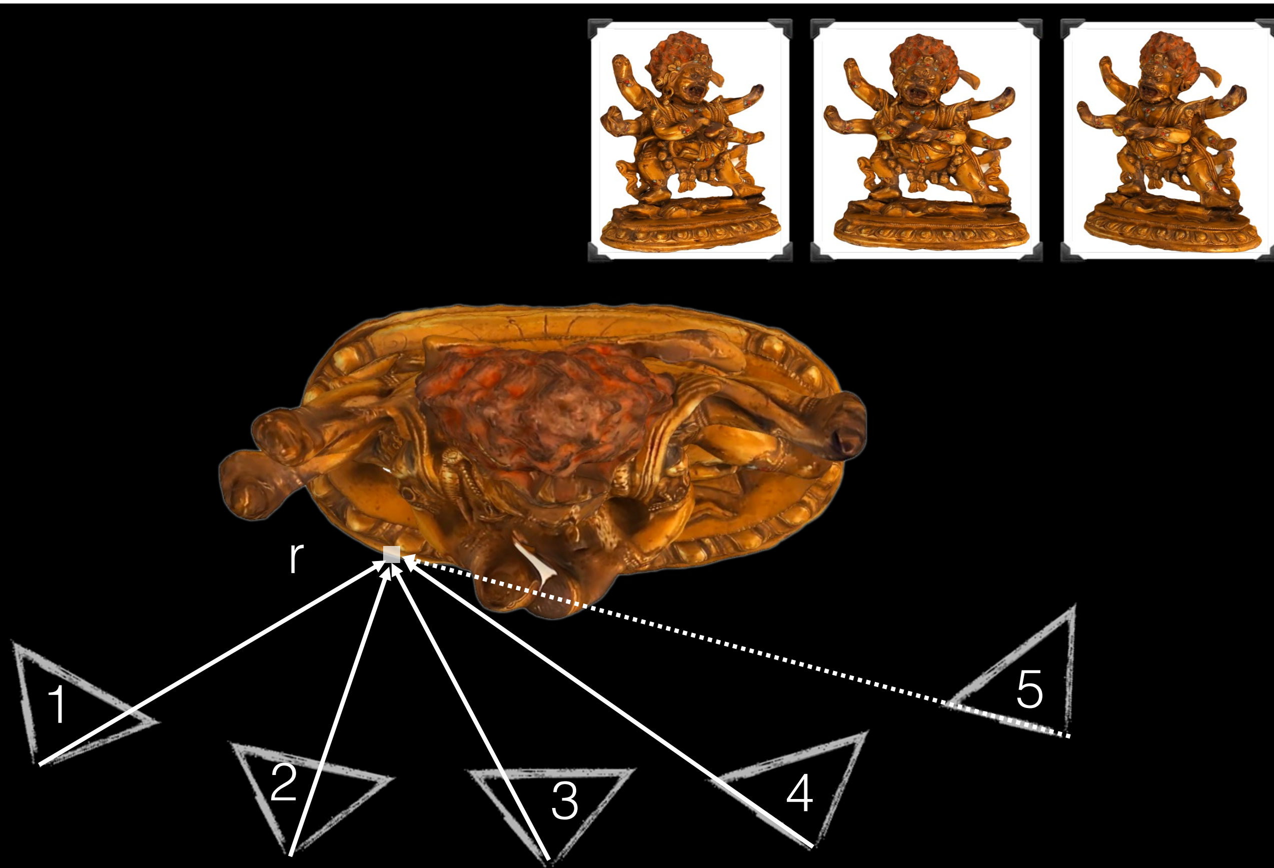
Cameras 4 and 5 can more clearly see point p

Source: N. Snavely



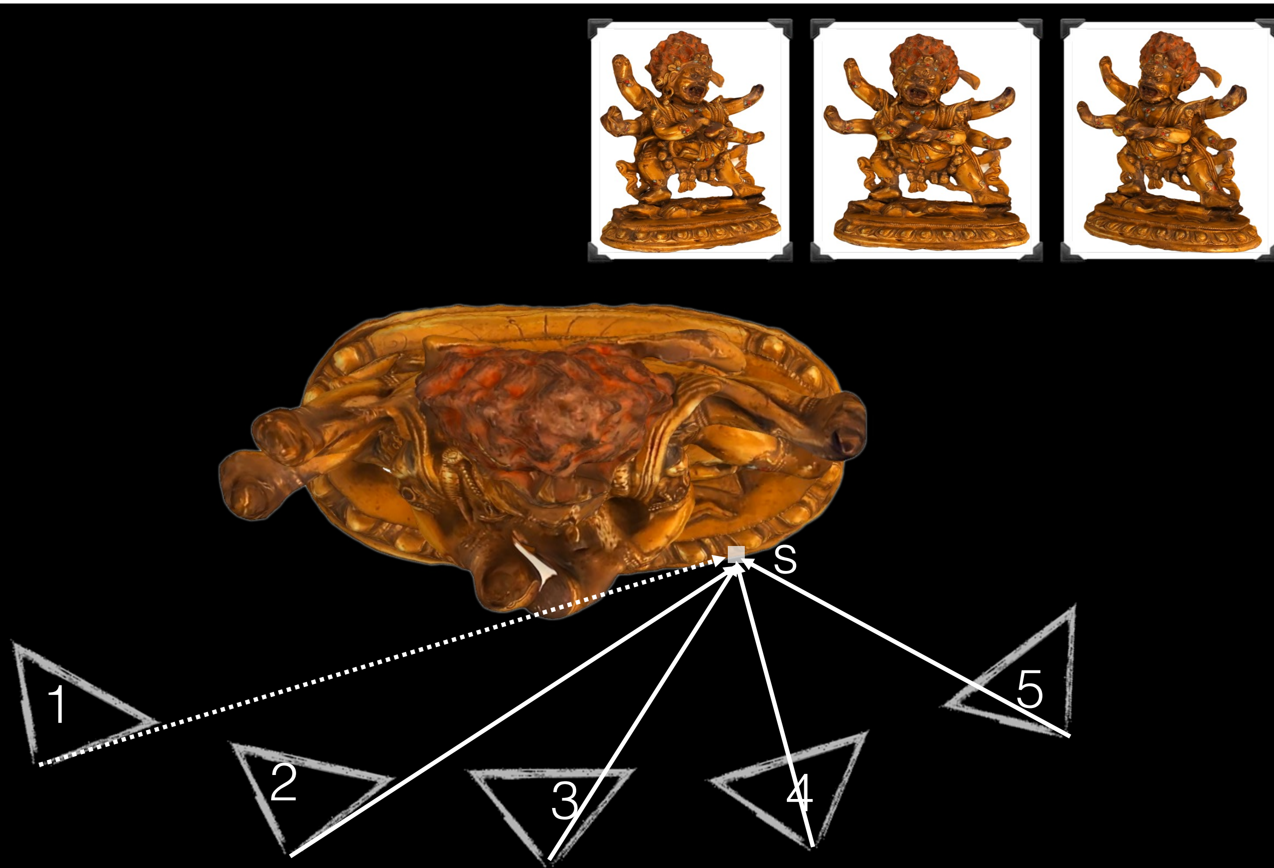
Cameras 3 and 4 can more clearly see point q

Source: N. Snavely



Camera 5 can't see point r

Source: N. Snavely



Camera 1 can't see point s

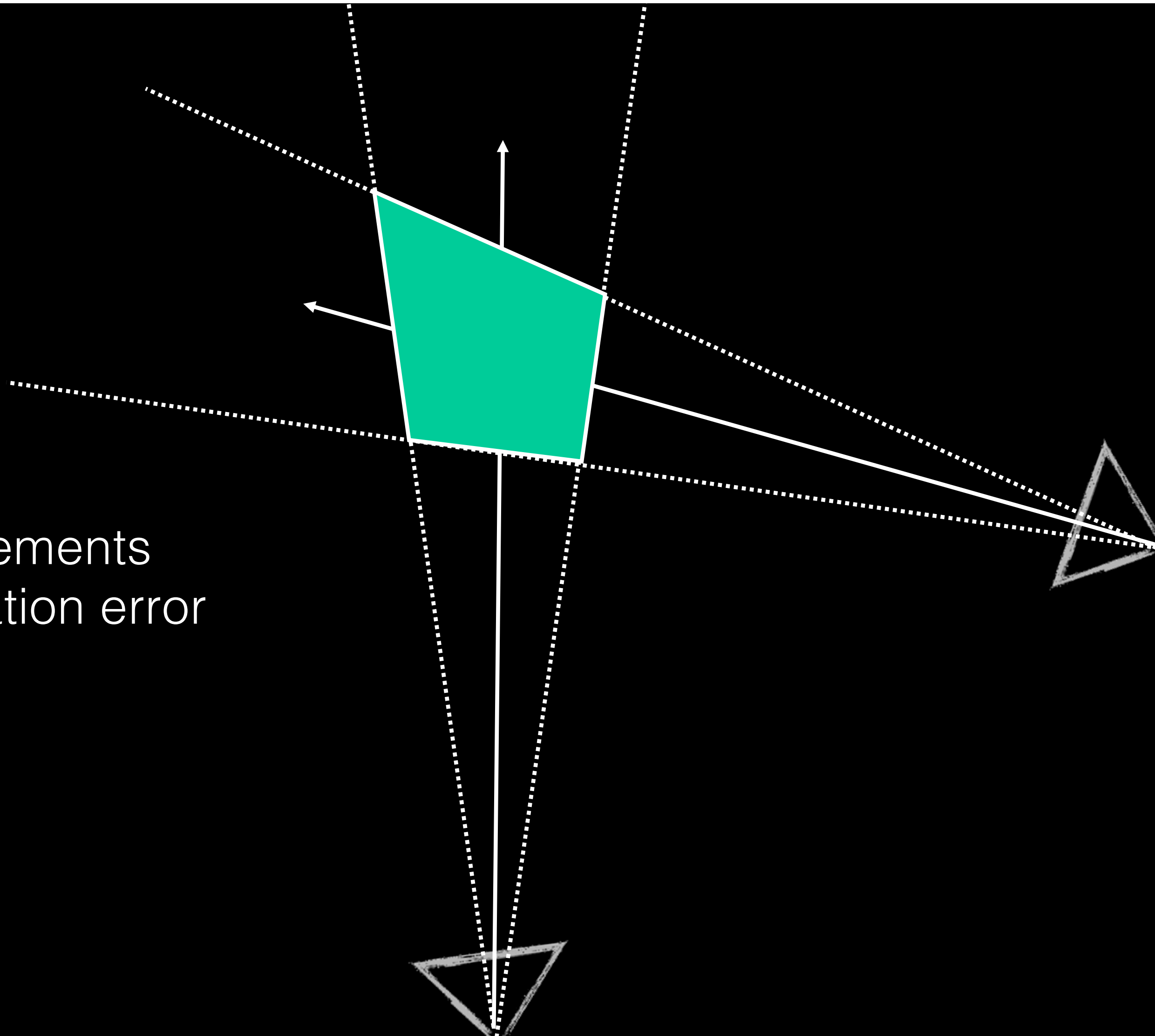
Source: N. Snavely

Why MVS?

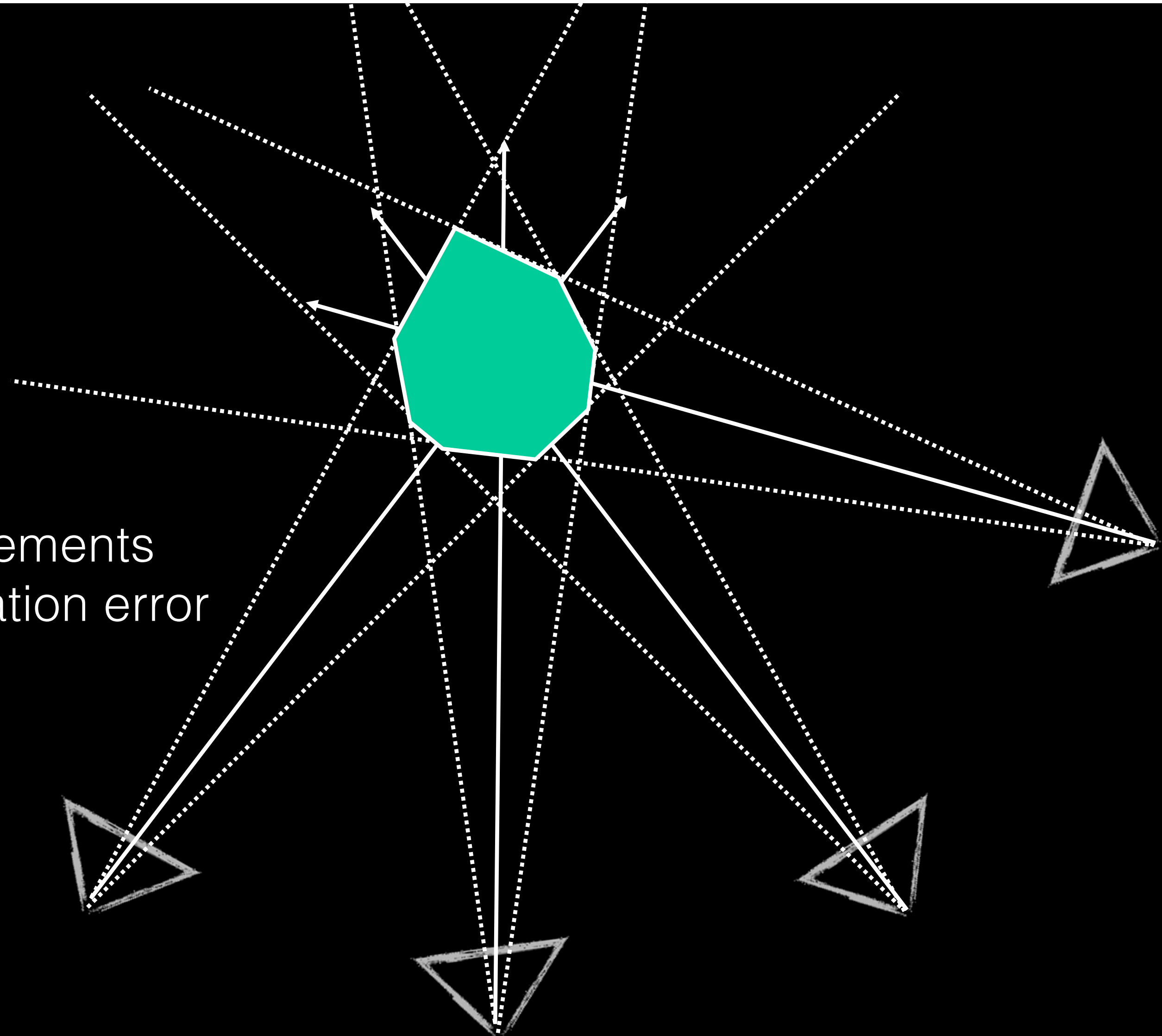
- Different points on the object's surface will be more clearly visible in some subset of cameras
  - Could have high-res closeups of some regions
  - Some surfaces are foreshortened from certain views
  - Some points may be occluded entirely in certain views
- More measurements per point can reduce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

Source: N. Snavely

More measurements reduce triangulation error



Source: N. Snavely



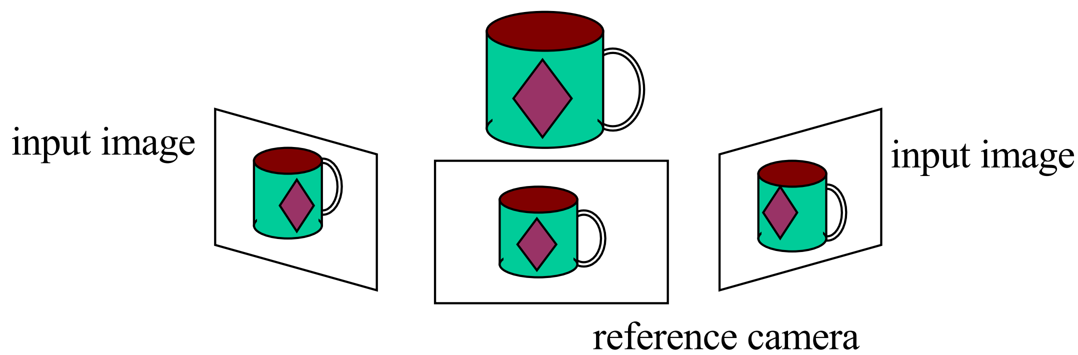
More measurements
reduce triangulation error

Source: N. Snavely

Outline

- Applications and motivation
- Plane sweep stereo
- Depth map fusion

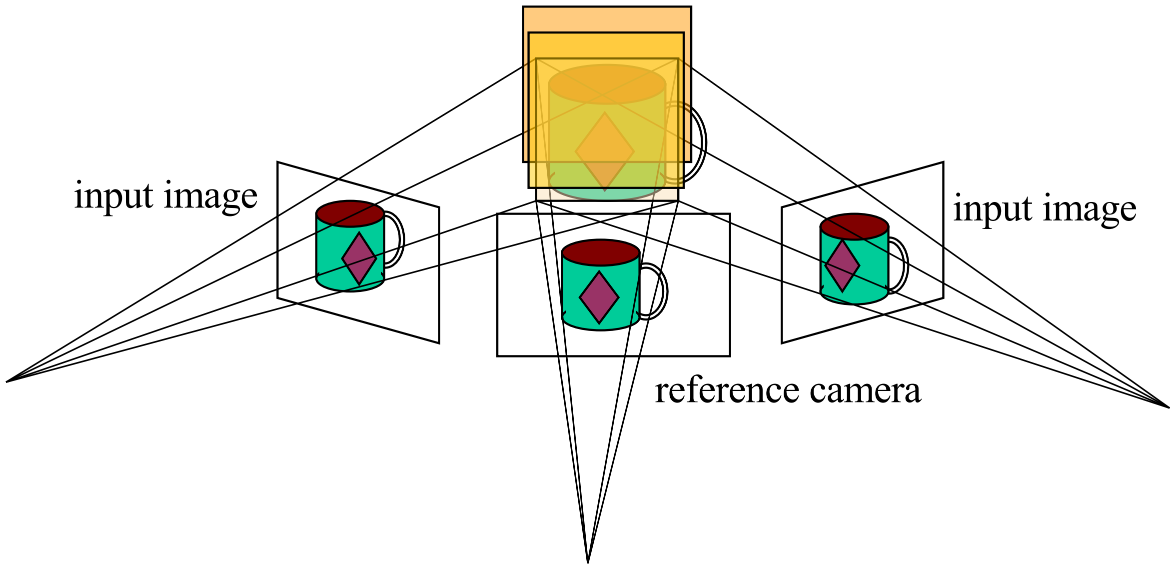
Plane sweep stereo



- Sweep plane across a range of depths w.r.t. a reference camera
- For each depth, project each input image onto that plane (homography) and compare the resulting stack of images

R. Collins, [A space-sweep approach to true multi-image matching](), CVPR 1996

Plane sweep stereo



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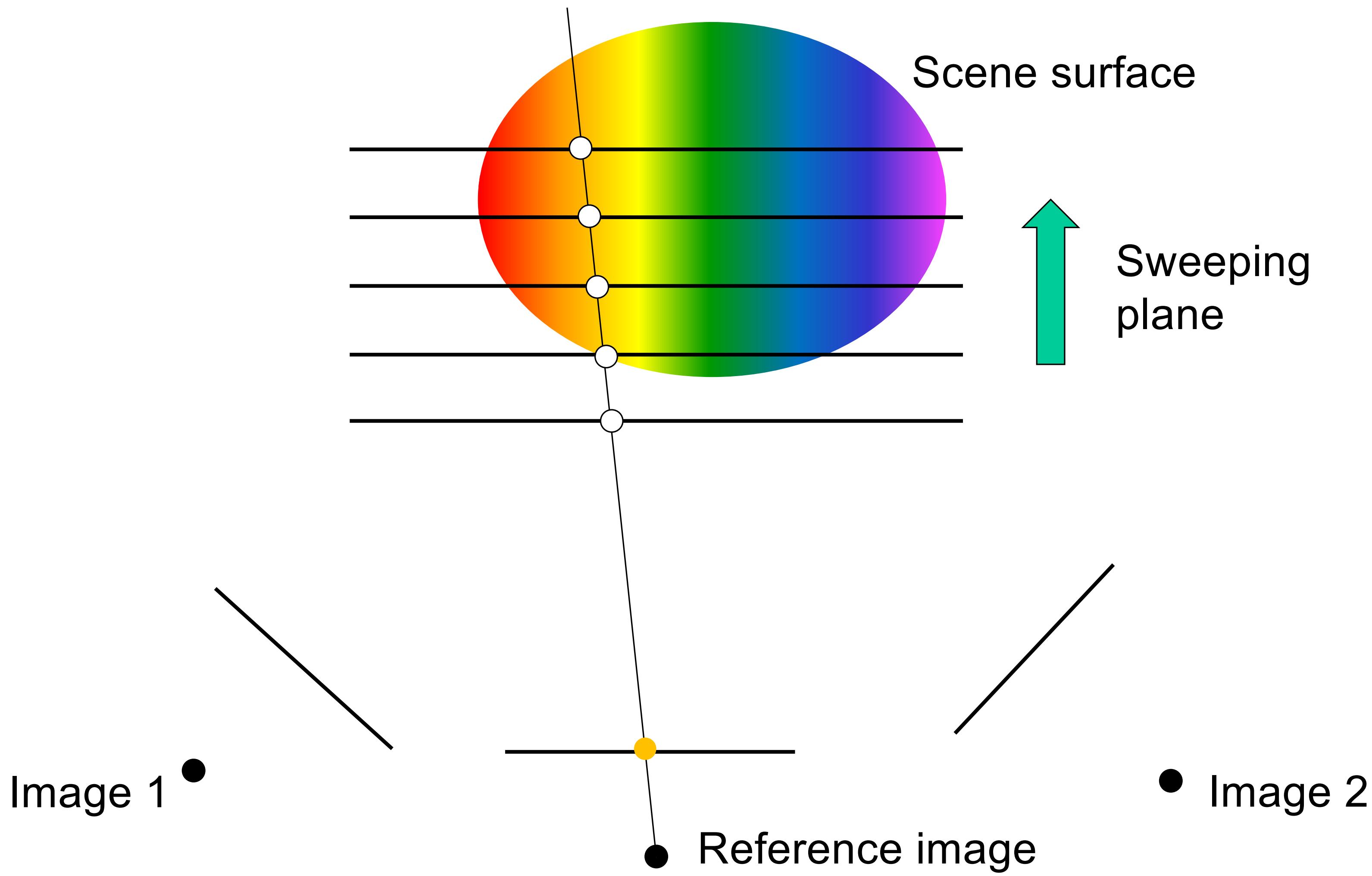
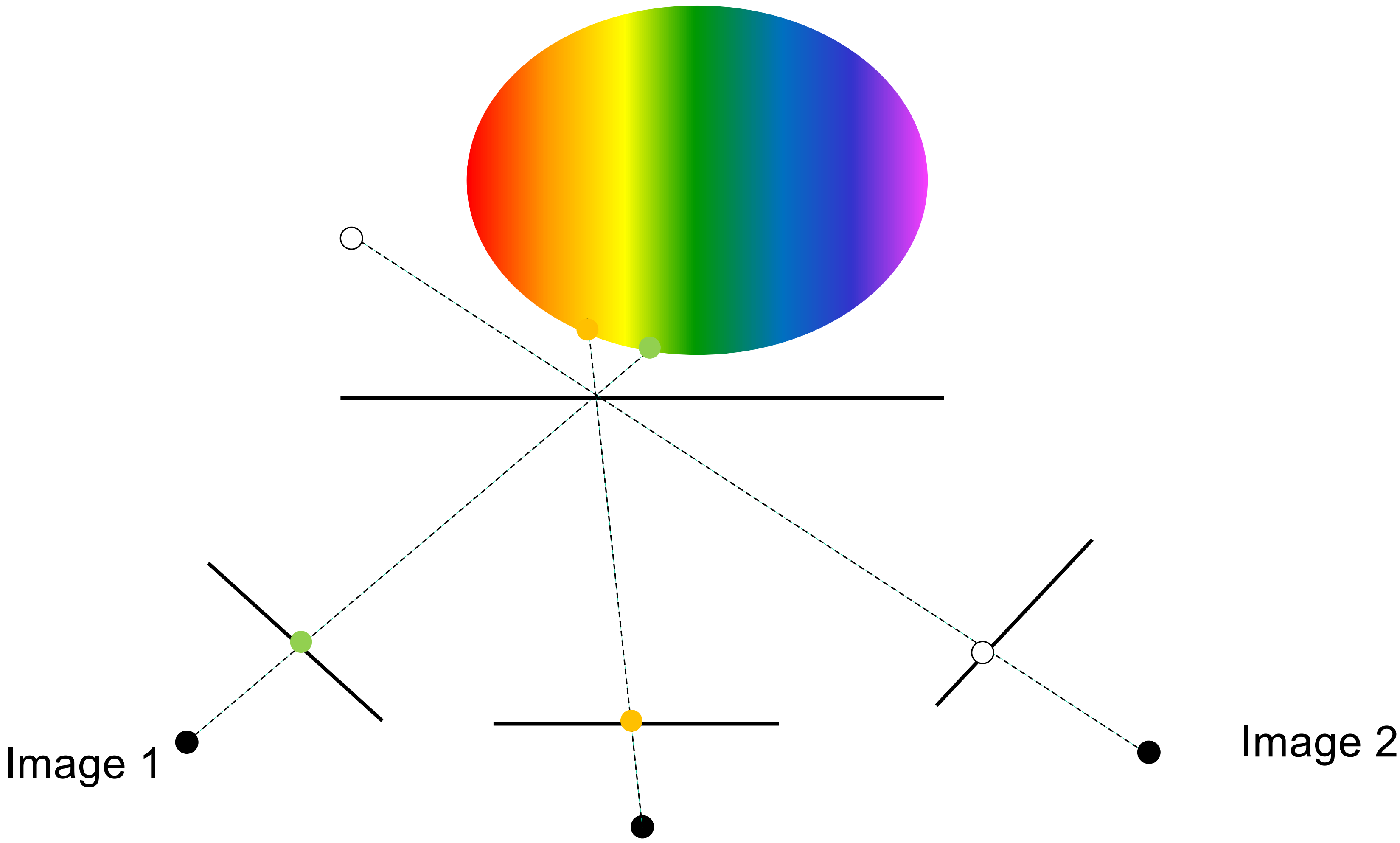
R. Collins, A space-sweep approach to true multi-image matching, CVPR 1996

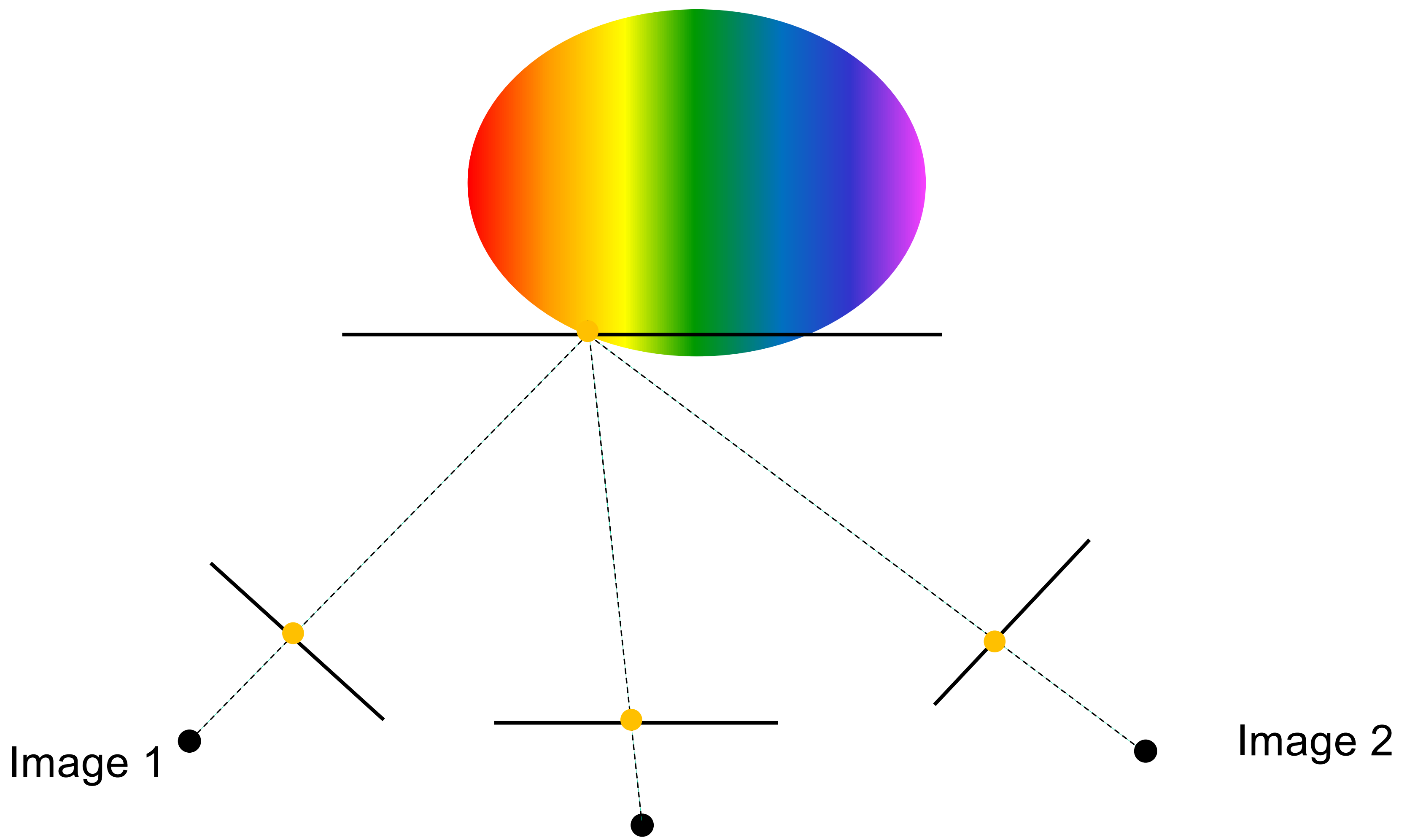
Plane sweep stereo: Key idea



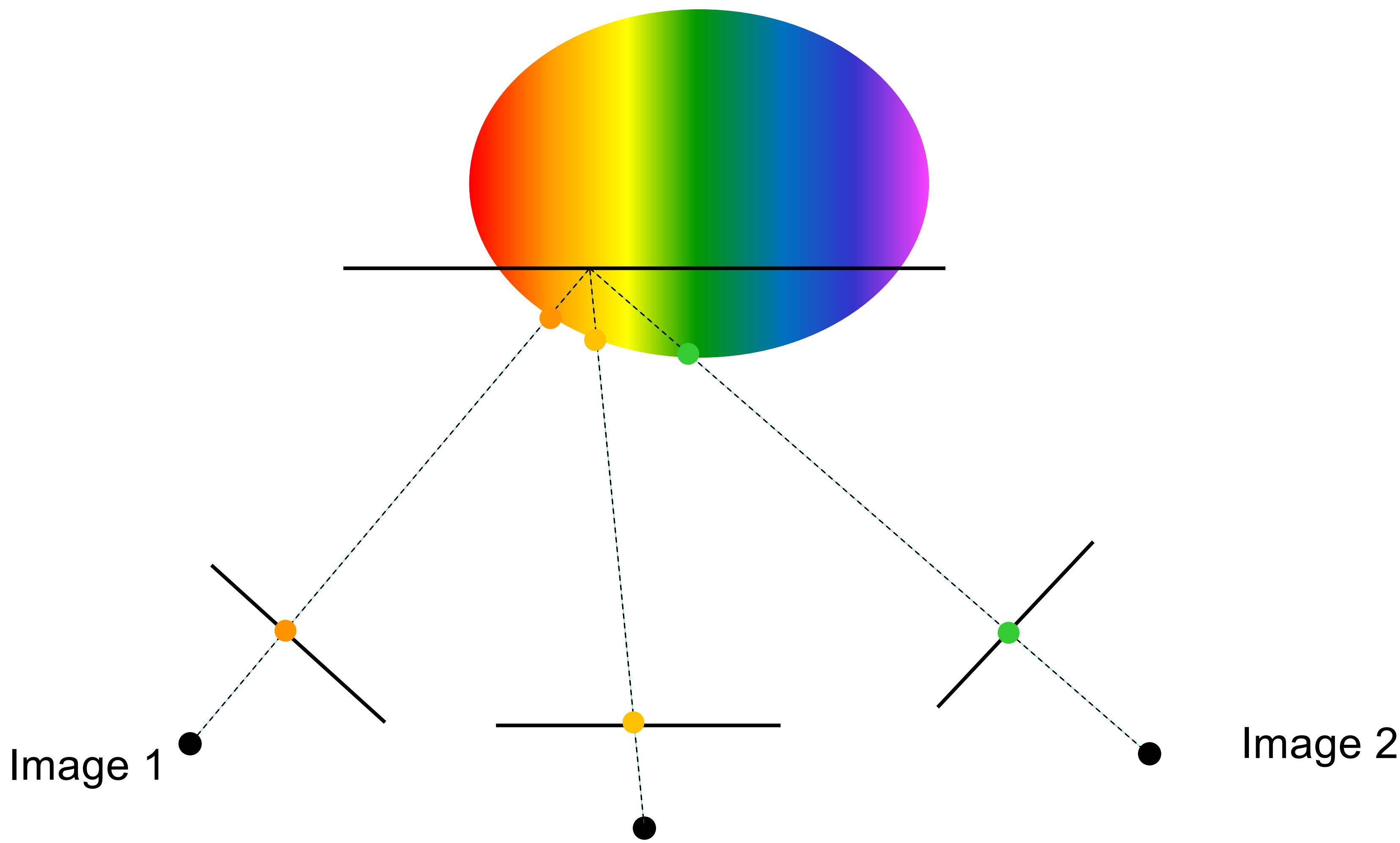
Plane sweep stereo: Key idea



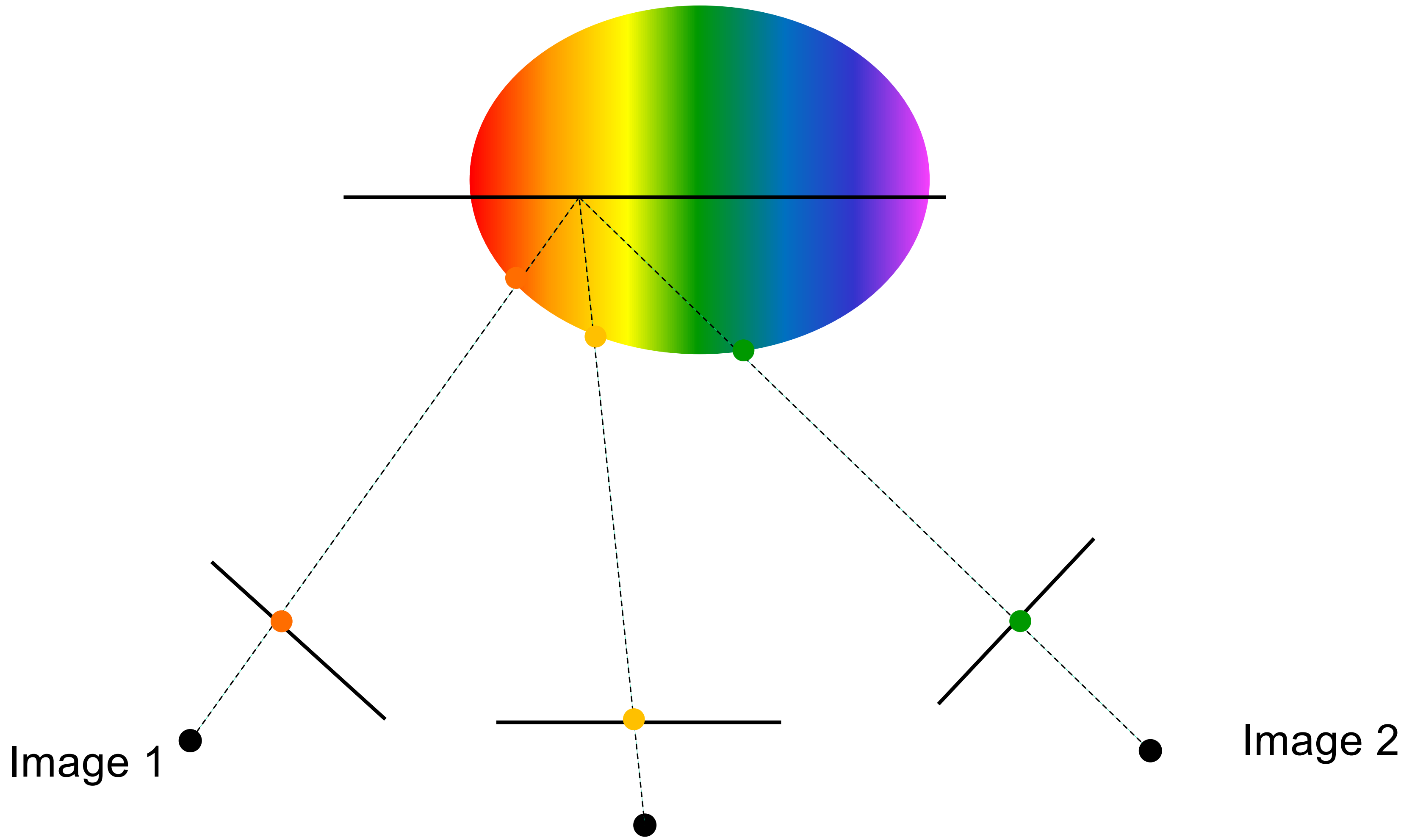
Plane sweep stereo: Key idea



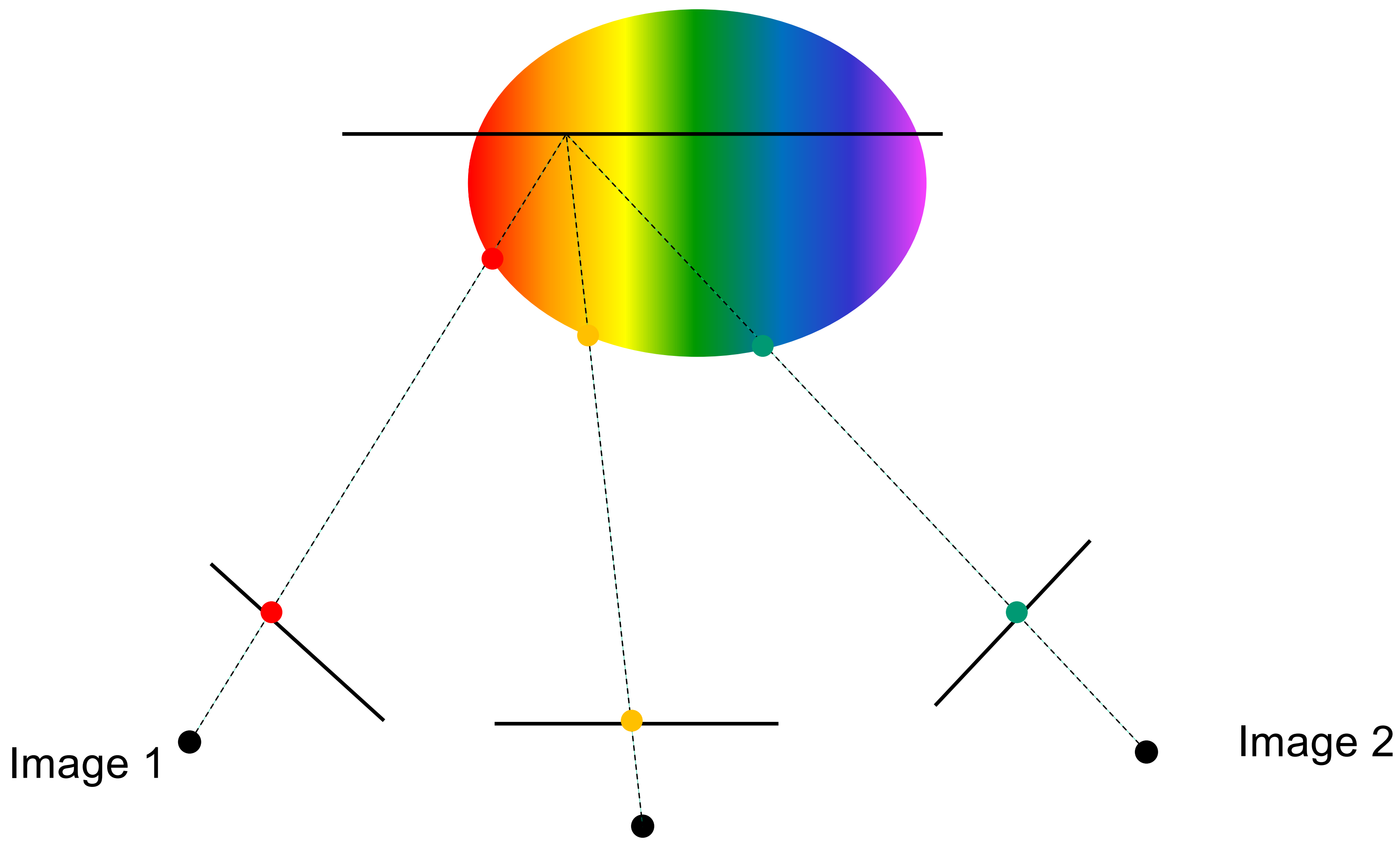
Plane sweep stereo: Key idea



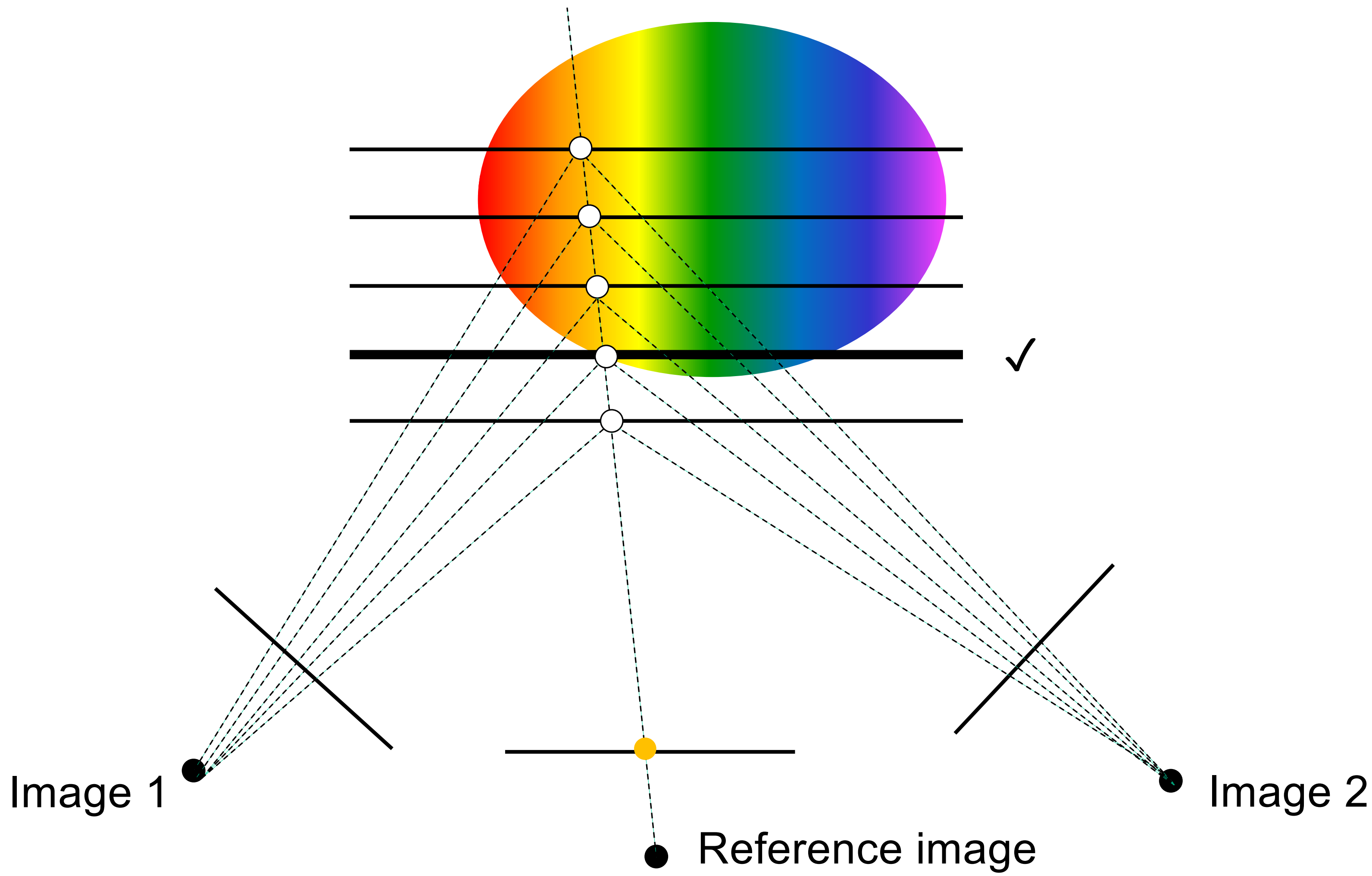
Plane sweep stereo: Key idea



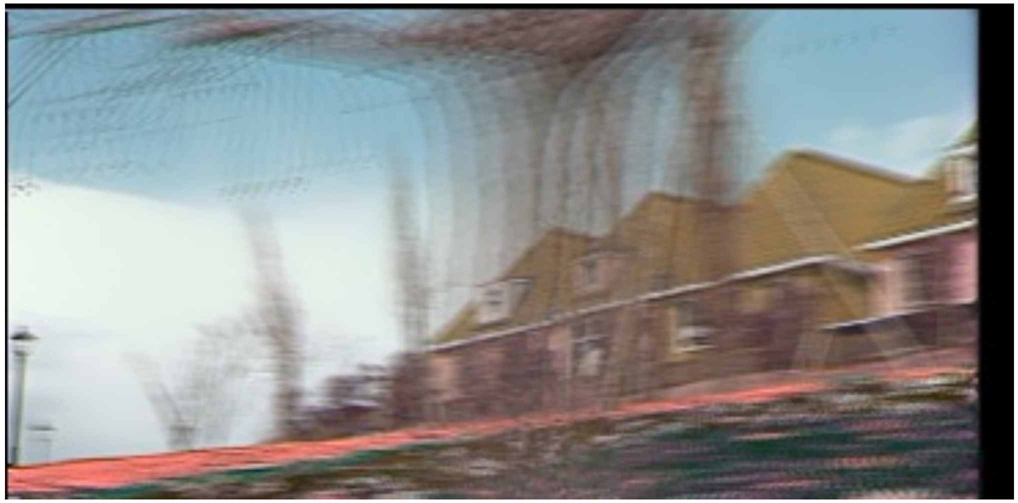
Plane sweep stereo: Key idea



Plane sweep stereo: Key idea



Plane sweep stereo: Fast implementation

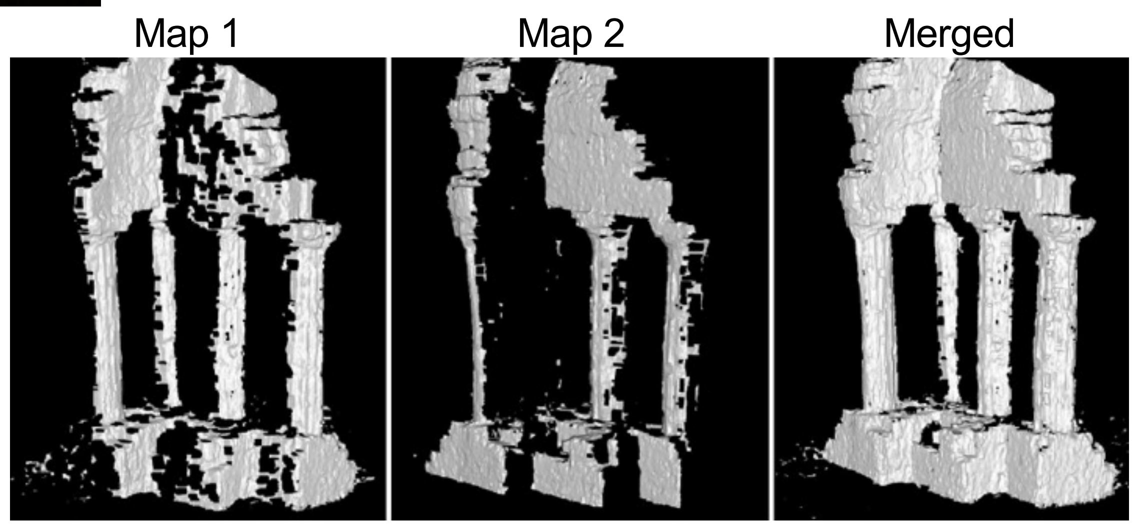


- For each depth plane
  - Compute homographies projecting each image onto that depth plane
  - For each pixel in the composite image stack, compute the variance
- For each pixel, select the depth that gives the lowest variance

R. Yang and M. Pollefeys, Multi-Resolution Real-Time Stereo on Commodity Graphics Hardware, CVPR 2003

Merging depth maps

- Given a group of images, compute a depth map using each view as a reference
- Merge multiple depth maps into a volume or a mesh (see, e.g., [Curless and Levoy, 1996](#))



Volumetric fusion, I

Depths from cameras read into a voxel space yield likely labels for SOME voxels (blue – empty; pink – occupied)

Q: what about other voxels?

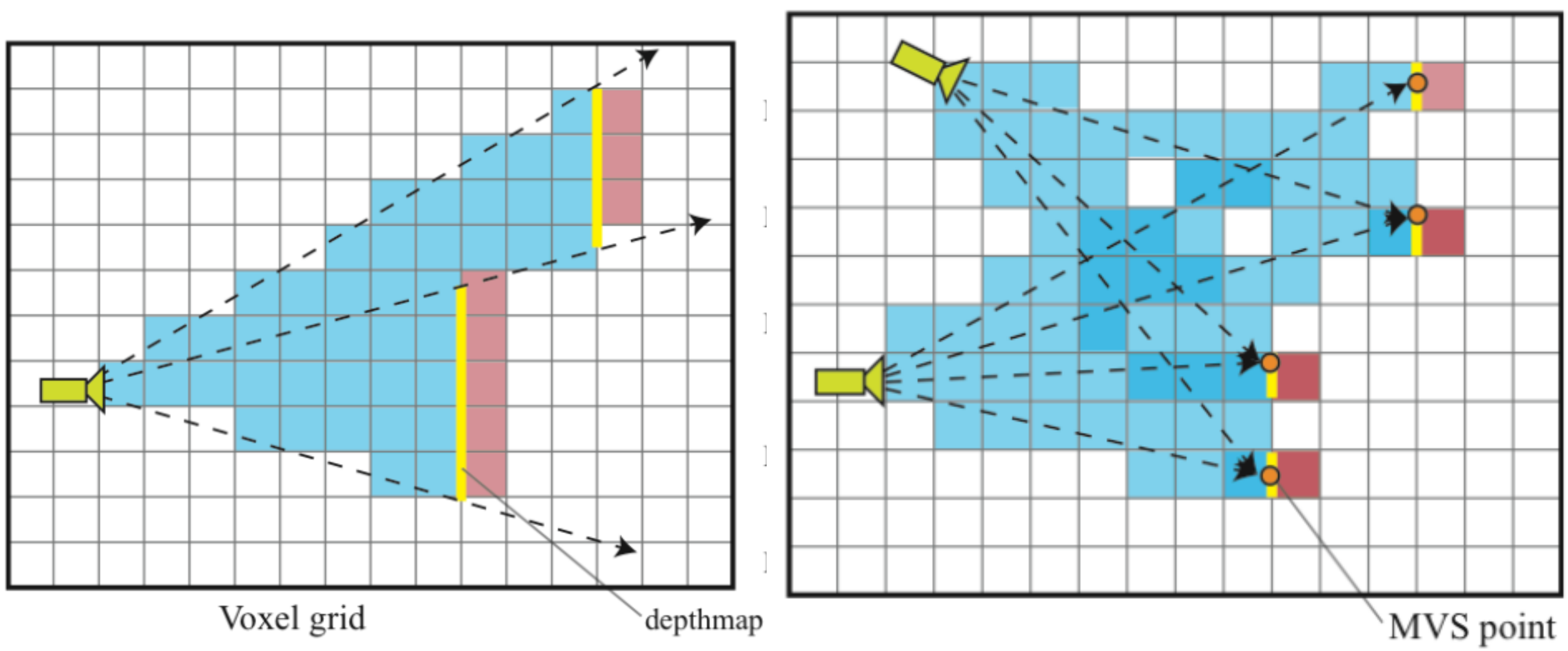


Figure 3.21: An example of how 3D MRF cost function should be set from a single depthmap.

Furukawa + Hernandez, 15, Multi-View Stereo: A tutorial

Volumetric fusion, II

Other voxels:

- ideally, agree with original estimates
  - agree with neighbors
- This yields a cost function that can be minimized (rather like in stereo above)

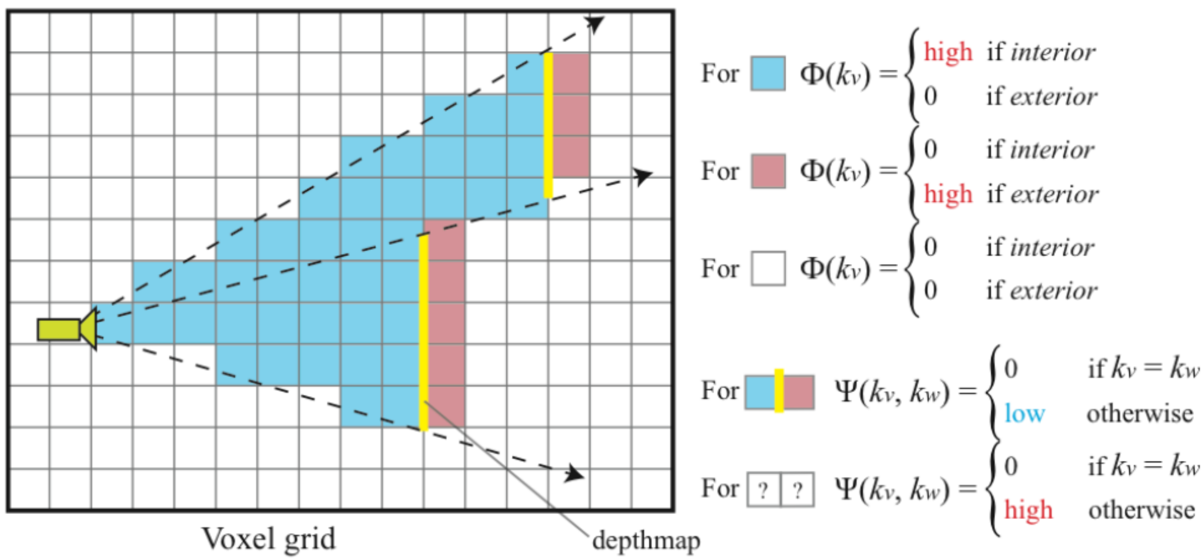


Figure 3.21: An example of how 3D MRF cost function should be set from a single depthmap.

Furukawa + Hernandez, 15, Multi-View Stereo: A tutorial

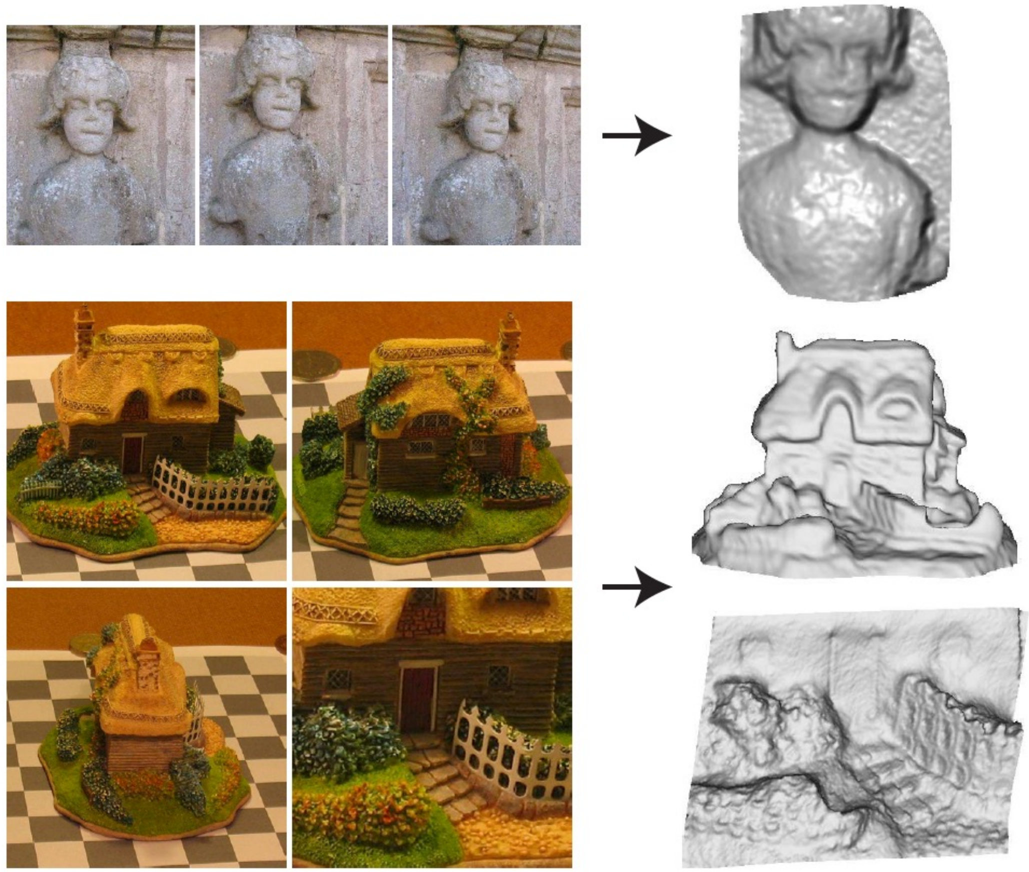
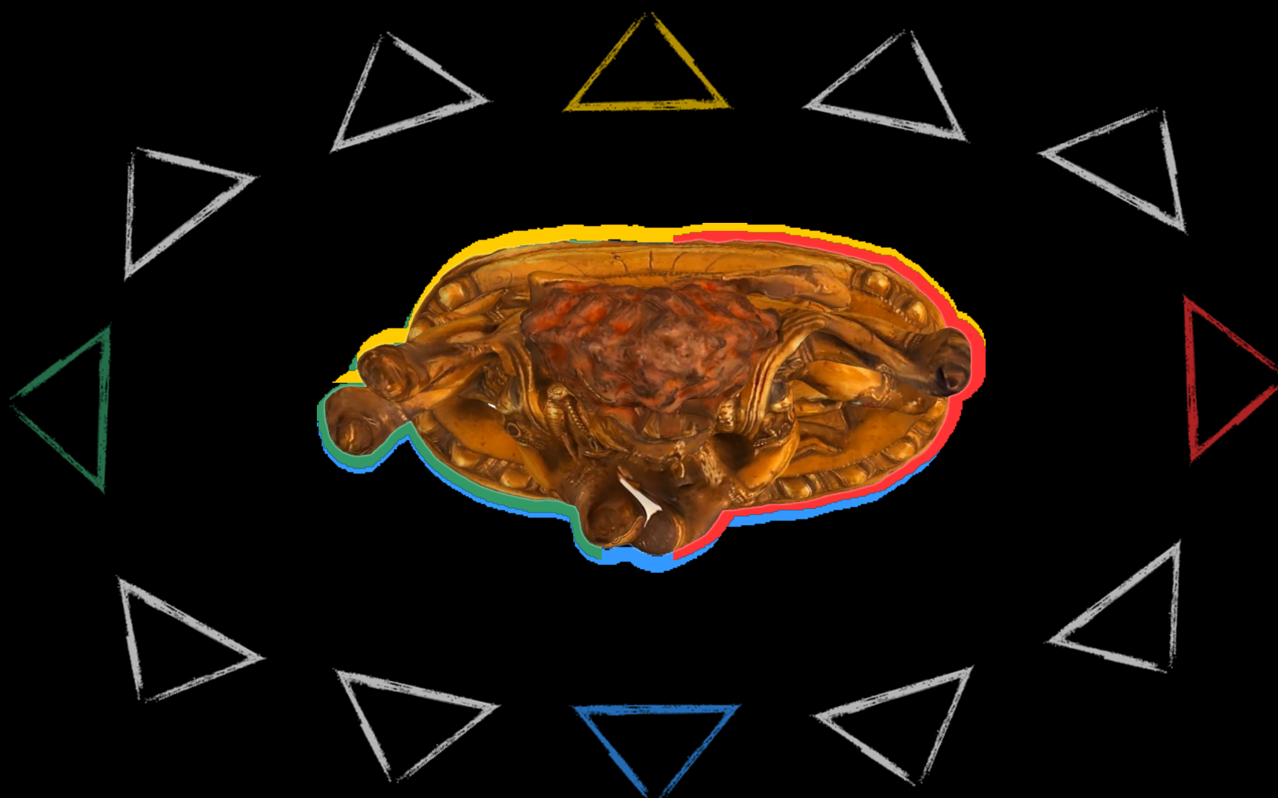


Figure 3.23: One of the earliest volume fusion techniques based on the volumetric graph-cuts by Vogiatzis, Torr and Cipolla [191]. (Figure courtesy of Vogiatzis et al.)

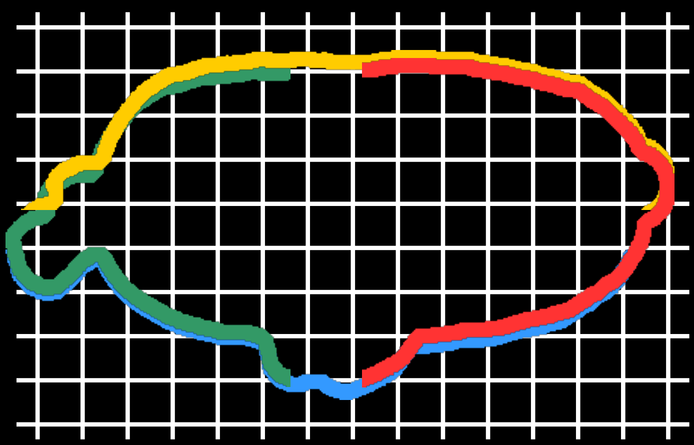
Furukawa + Hernandez, 15, Multi-View Stereo: A tutorial

Volumetric fusion



Source: N. Snavely

Volumetric fusion



Source: N. Snavely

Fast depth map fusion using height maps

- Start with a cluster of registered views (from SFM on Internet photo collections)

J.-M. Frahm et al., Building Rome on a Cloudless Day, ECCV 2010

D. Gallup et al. 3D Reconstruction using an n-Layer Heightmap. DAGM 2010

Fast depth map fusion using height maps

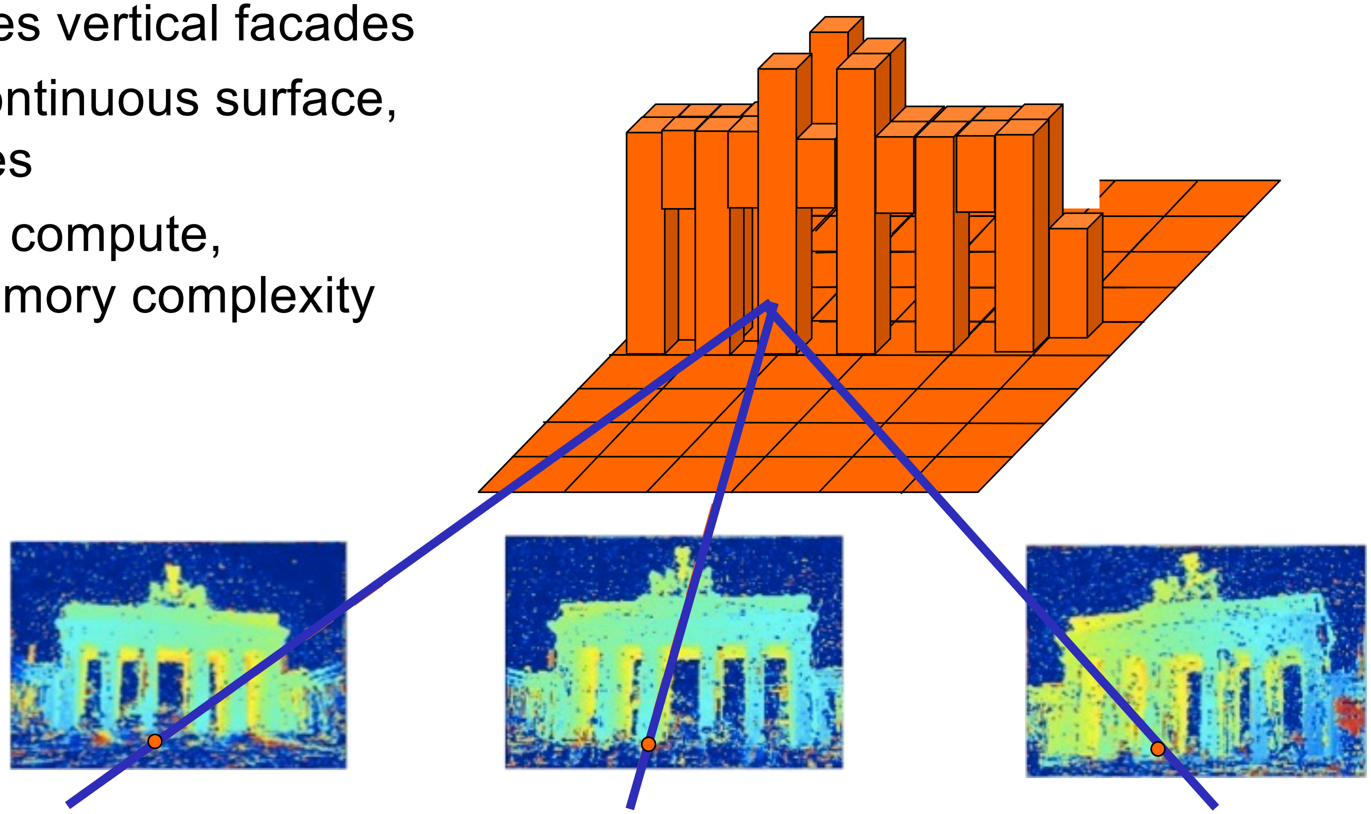
- Obtain a (noisy) depth map for every view using plane sweeping stereo with normalized cross-correlation

J.-M. Frahm et al., Building Rome on a Cloudless Day, ECCV 2010

D. Gallup et al. 3D Reconstruction using an n-Layer Heightmap. DAGM 2010

Fast depth map fusion using height maps

- Enforces vertical facades
- One continuous surface, no holes
- Fast to compute, low memory complexity



J.-M. Frahm et al., Building Rome on a Cloudless Day, ECCV 2010

D. Gallup et al. 3D Reconstruction using an n-Layer Heightmap. DAGM 2010

Fast depth map fusion using height maps

[YouTube Video]

J.-M. Frahm et al., Building Rome on a Cloudless Day, ECCV 2010

Outline

- Applications and motivation
- Plane sweep stereo
- Depth map fusion
- Patch-based multi-view stereo (PMVS)
- Stereo from Internet photo collections

Patch-based multi-view stereo (PMVS)

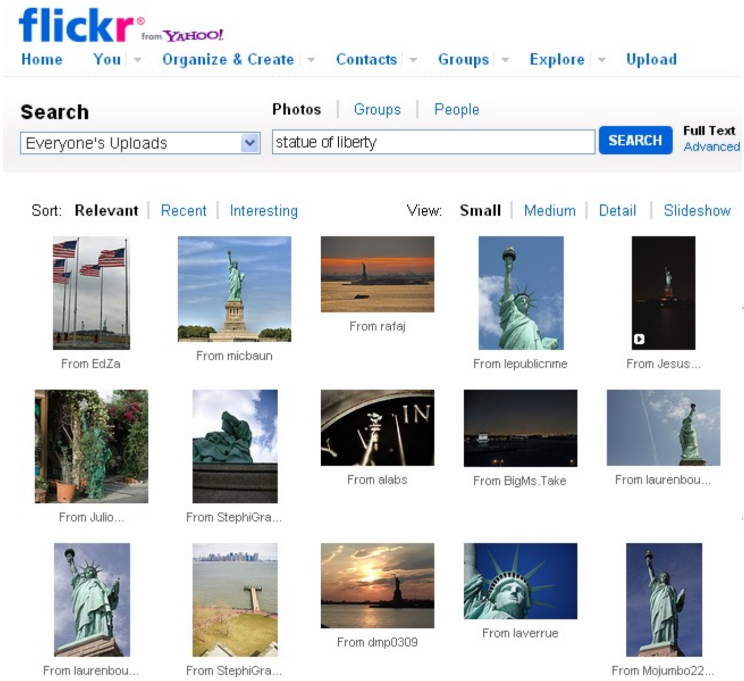
1. Detect keypoints
2. Triangulate a sparse set of initial matches
3. Iteratively expand matches to nearby locations
4. Use visibility constraints to filter out false matches
5. Perform surface reconstruction

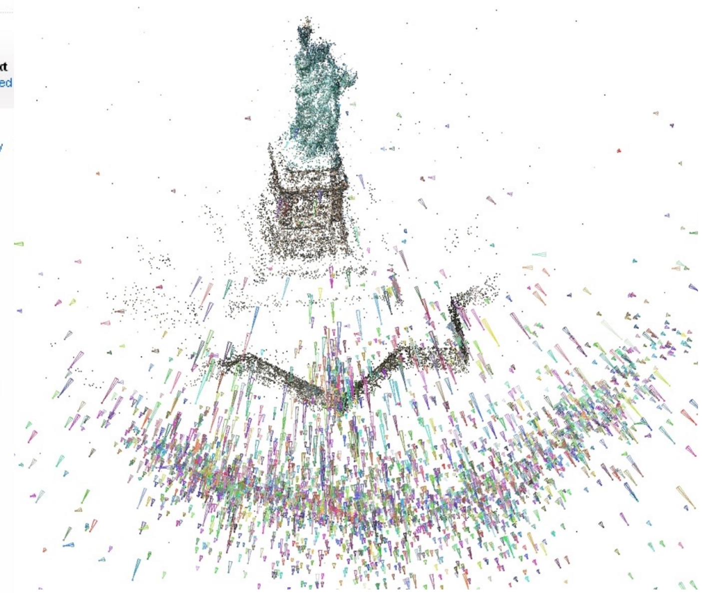
Y. Furukawa and J. Ponce, Accurate, Dense, and Robust Multi-View Stereopsis, CVPR 2007.
PMVS software

Patch-based multi-view stereo (PMVS)

Y. Furukawa and J. Ponce, Accurate, Dense, and Robust Multi-View Stereopsis, CVPR 2007.
PMVS software

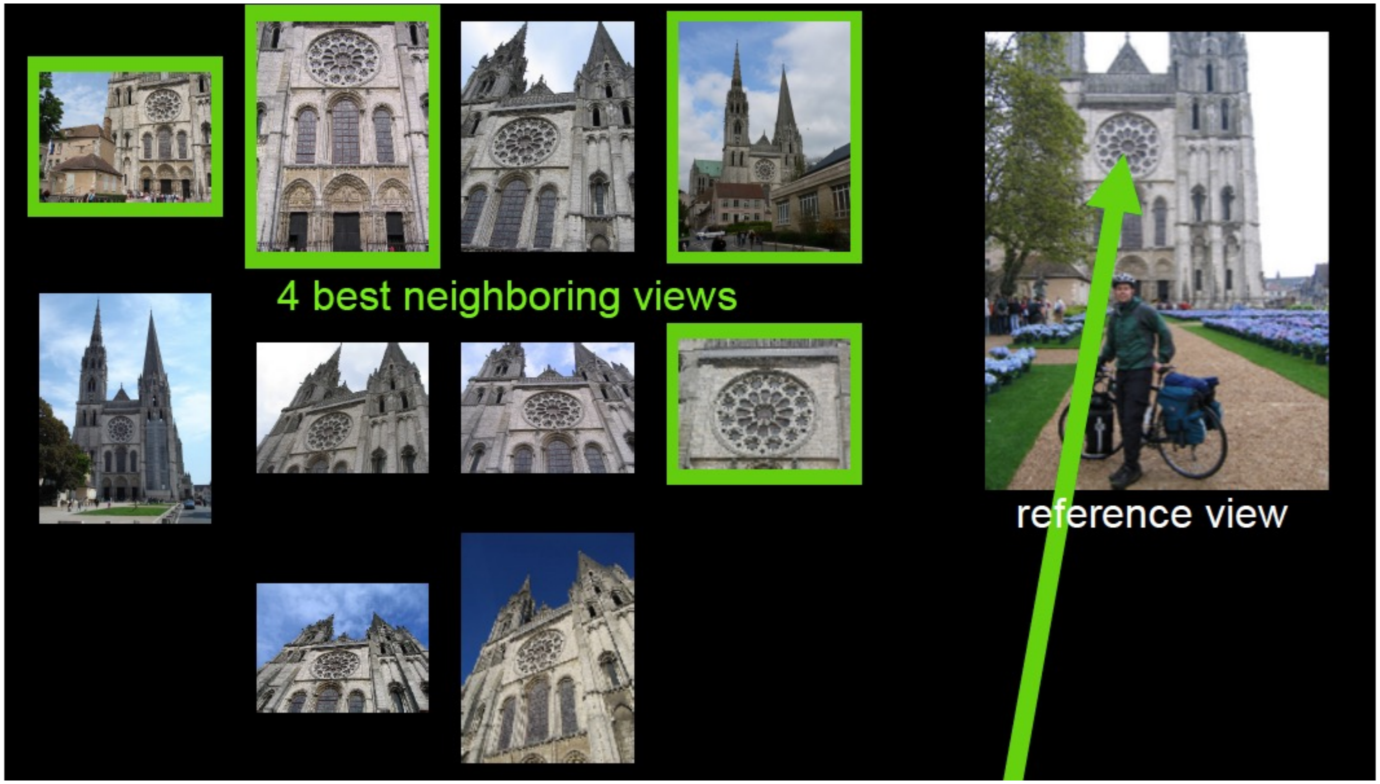
Stereo from community photo collections





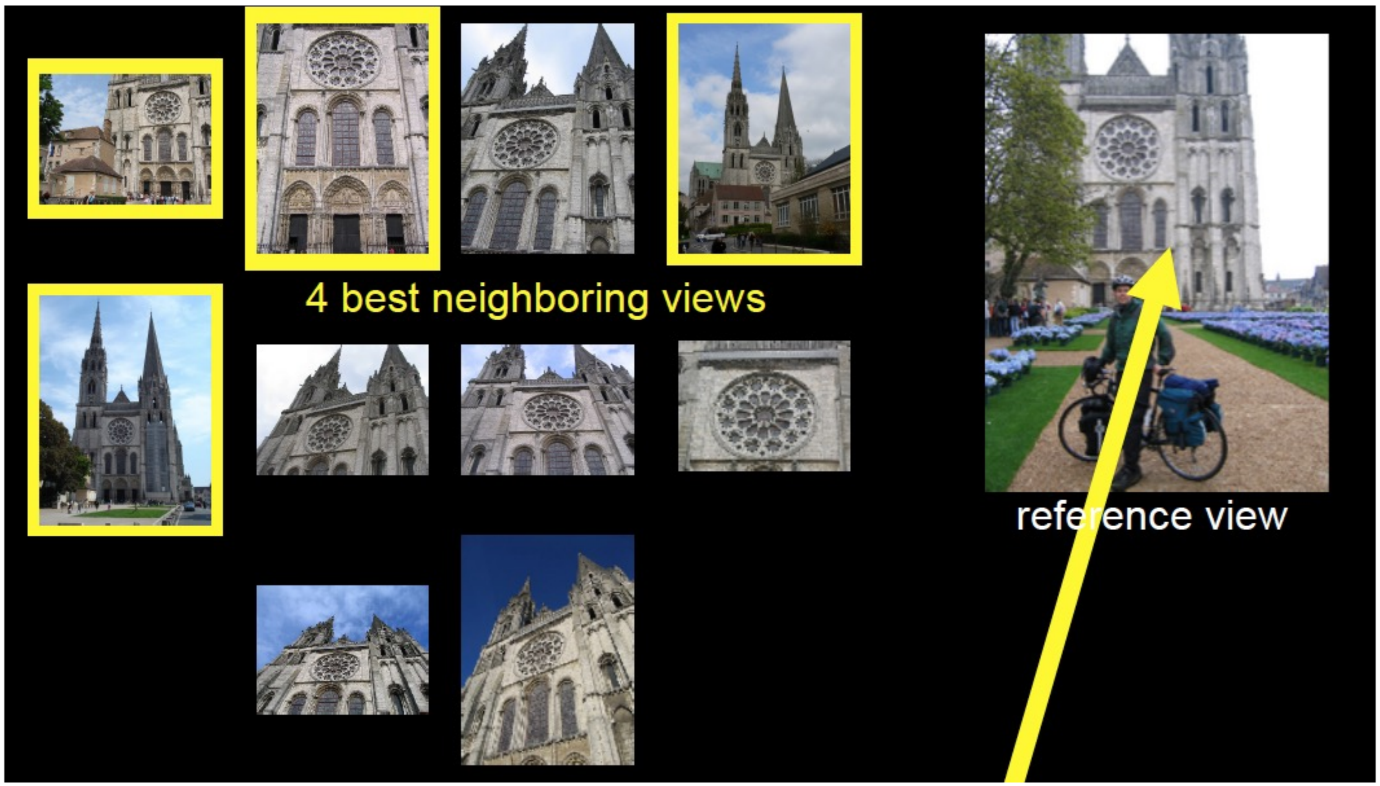
- Need *structure from motion* to recover unknown camera parameters
- Need *view selection* to find good groups of images on which to run dense stereo

Local view selection



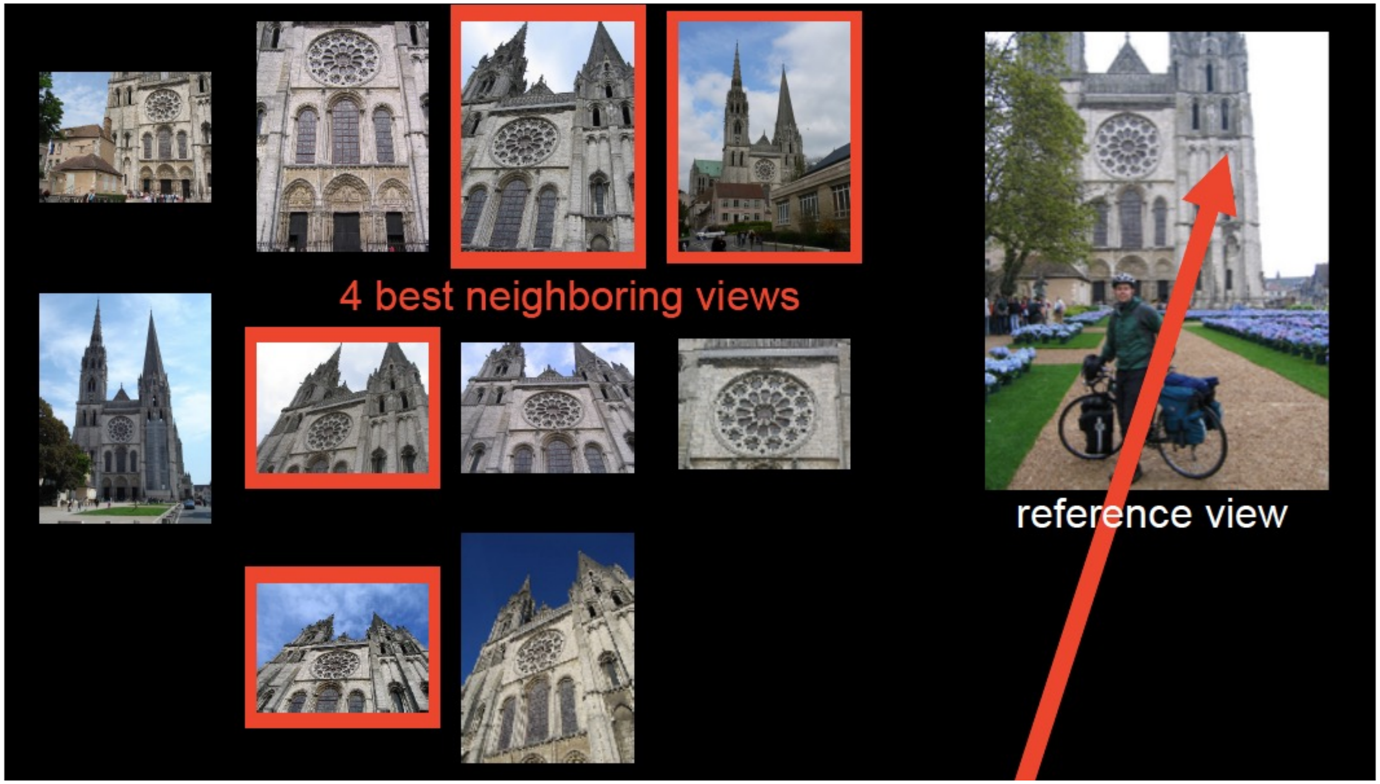
M. Goesele et al., [Multi-View Stereo for Community Photo Collections](), ICCV 2007

Local view selection



M. Goesele et al., [Multi-View Stereo for Community Photo Collections](), ICCV 2007

Local view selection

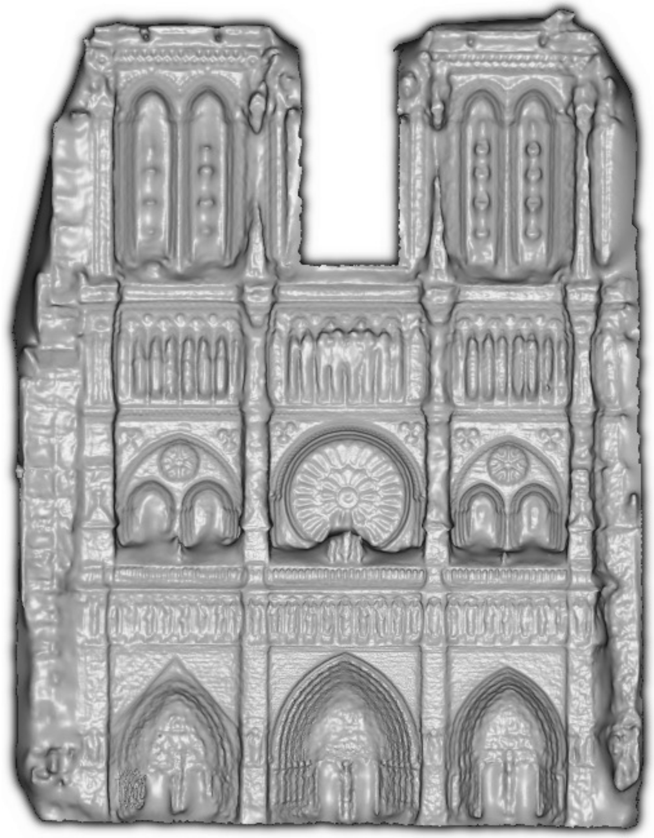


M. Goesele et al., Multi-View Stereo for Community Photo Collections, ICCV 2007

Local view selection

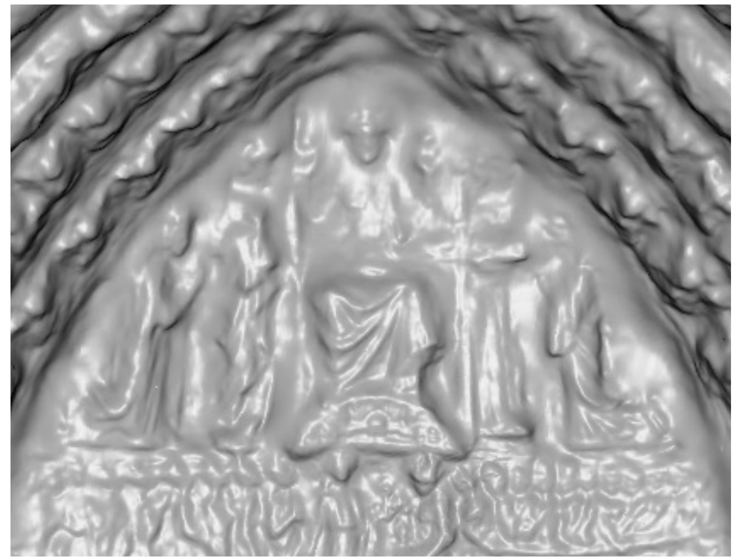
Notre Dame de Paris

653 images
313 photographers



M. Goesele et al., Multi-View Stereo for Community Photo Collections, ICCV 2007

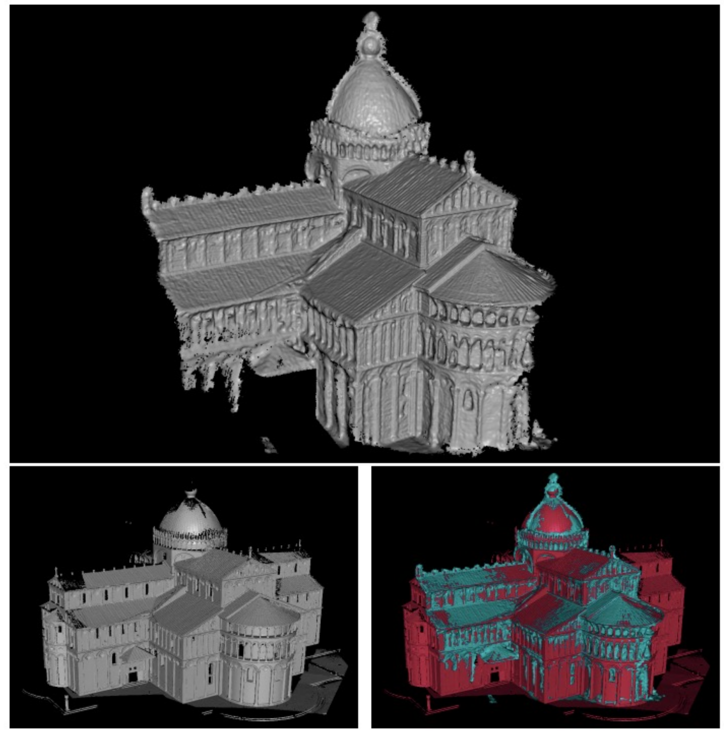
Local view selection



M. Goesele et al., Multi-View Stereo for Community Photo Collections, ICCV 2007

Local view selection

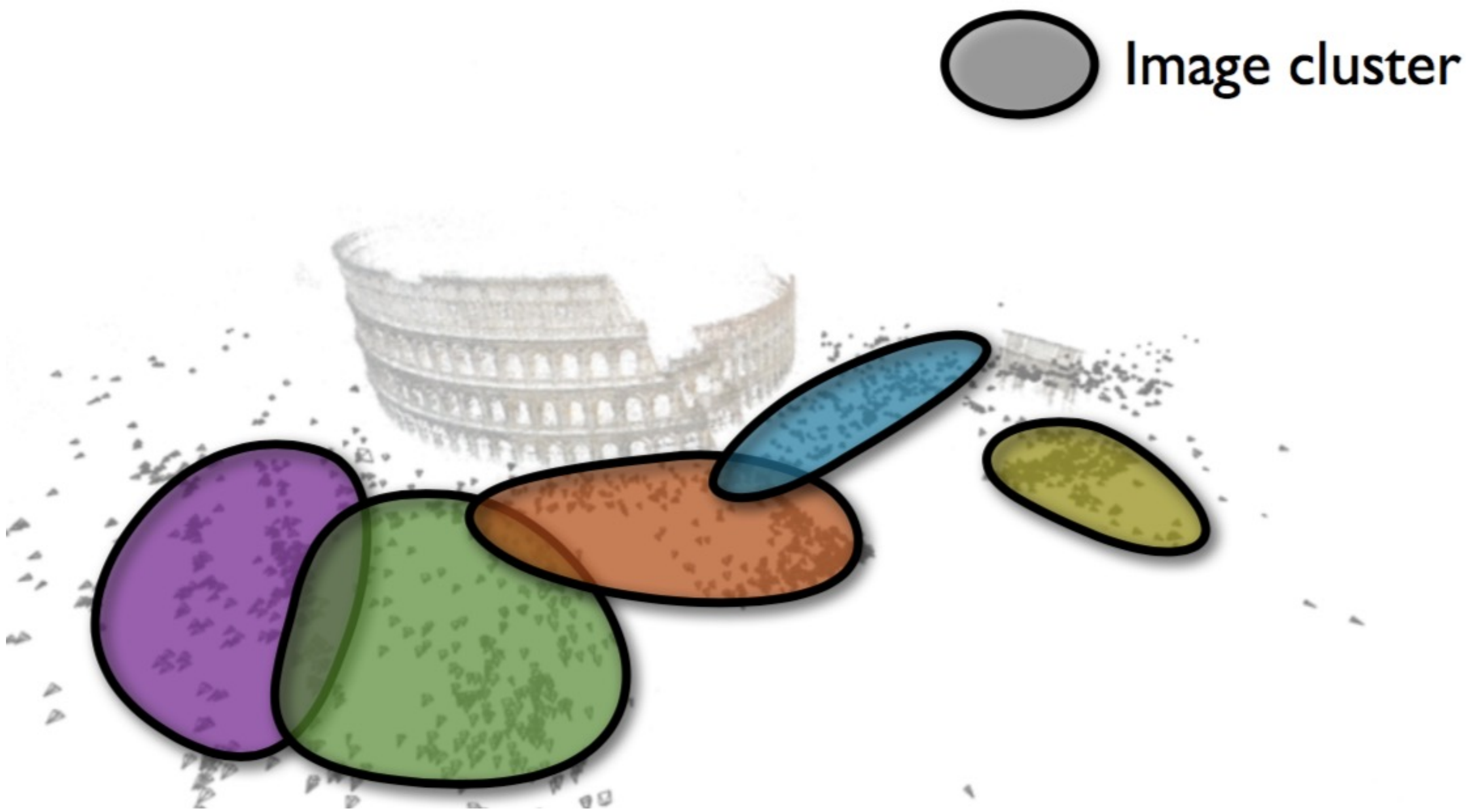
Model merged from 72 depth maps



Model from 56 depth maps with laser scan overlaid
(90% of points within *0.25%* of ground truth)

M. Goesele et al., [Multi-View Stereo for Community Photo Collections](), ICCV 2007

Towards Internet-scale multi-view stereo



Y. Furukawa, B. Curless, S. Seitz and R. Szeliski, Towards Internet-scale Multi-view Stereo, CVPR 2010

Towards Internet-scale multi-view stereo

YouTube video, CMVS software

Y. Furukawa, B. Curless, S. Seitz and R. Szeliski, Towards Internet-scale Multi-view Stereo, CVPR 2010

The Visual Turing Test for scene reconstruction

Rendered Images (Right) vs. Ground Truth Images (Left)

Q. Shan, R. Adams, B. Curless, Y. Furukawa, and S. Seitz, The Visual Turing Test for Scene Reconstruction, 3DV 2013. YouTube video

COLMAP MVS

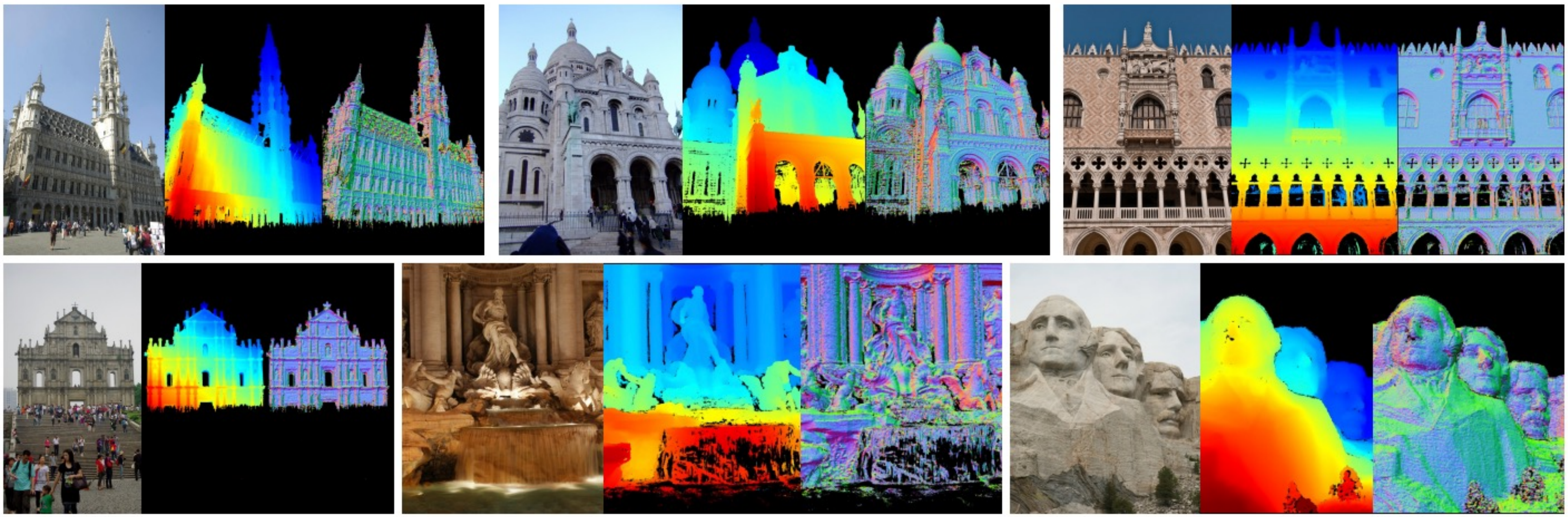


Fig. 6. Reference image with filtered depths and normals for crowd-sourced images.

J. Schonberger et al. [Pixelwise View Selection for Unstructured Multi-View Stereo](). ECCV 2016

[Results video]()

Outline

- Applications and motivation
- Plane sweep stereo
- Patch-based multi-view stereo (PMVS)
- Stereo from Internet photo collections
- Recent trends

Ongoing research directions

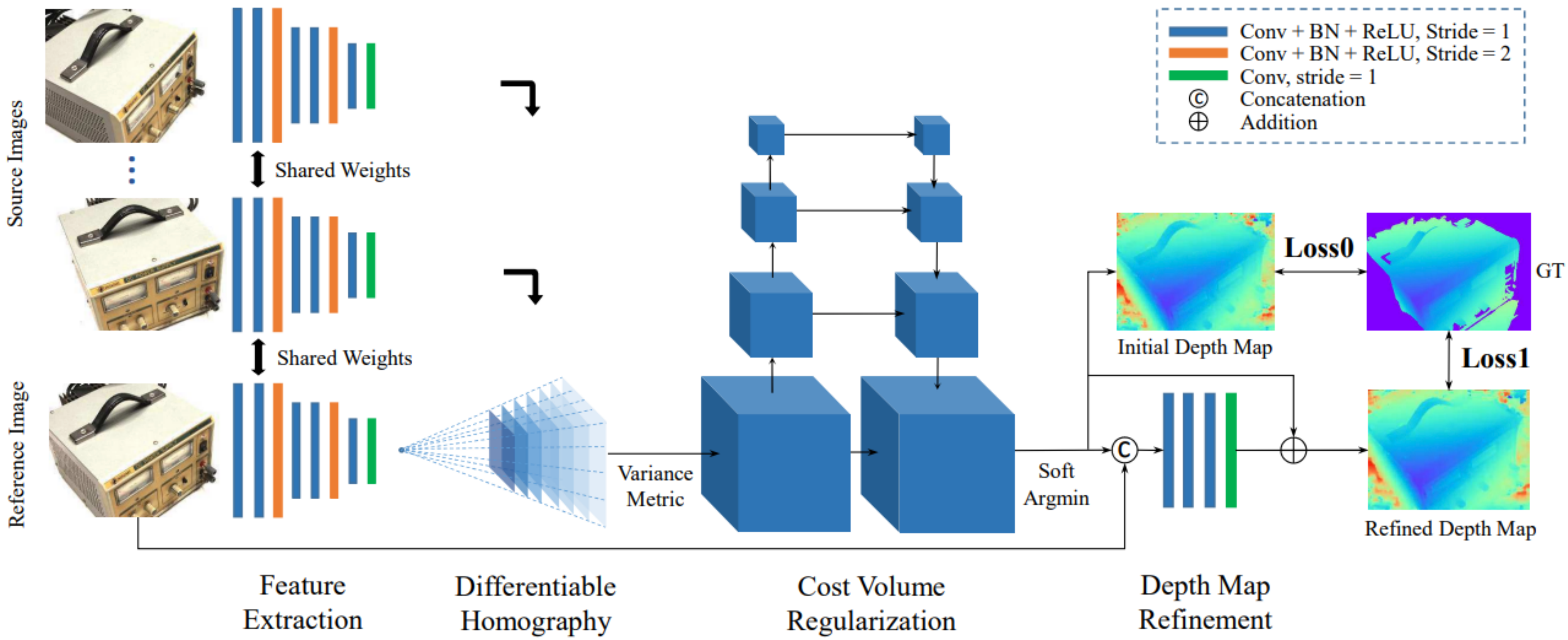
Challenging lighting conditions

Ground/aerial

Indoor modeling

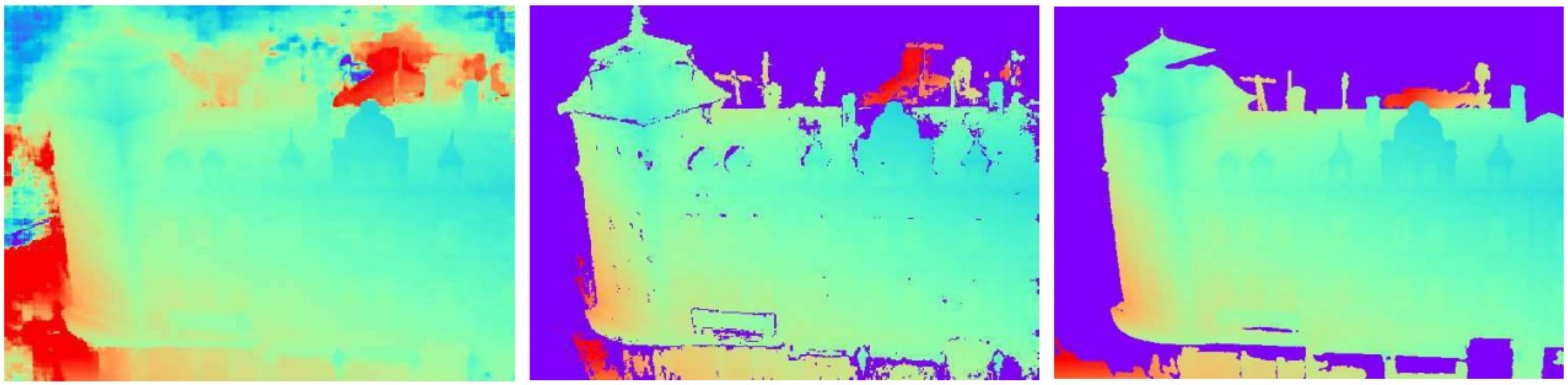
Dynamic reconstruction

Deep learning for MVS



Y. Yao et al. MVSNet: Depth Inference for Unstructured Multi-view Stereo. ECCV 2018

Deep learning for MVS

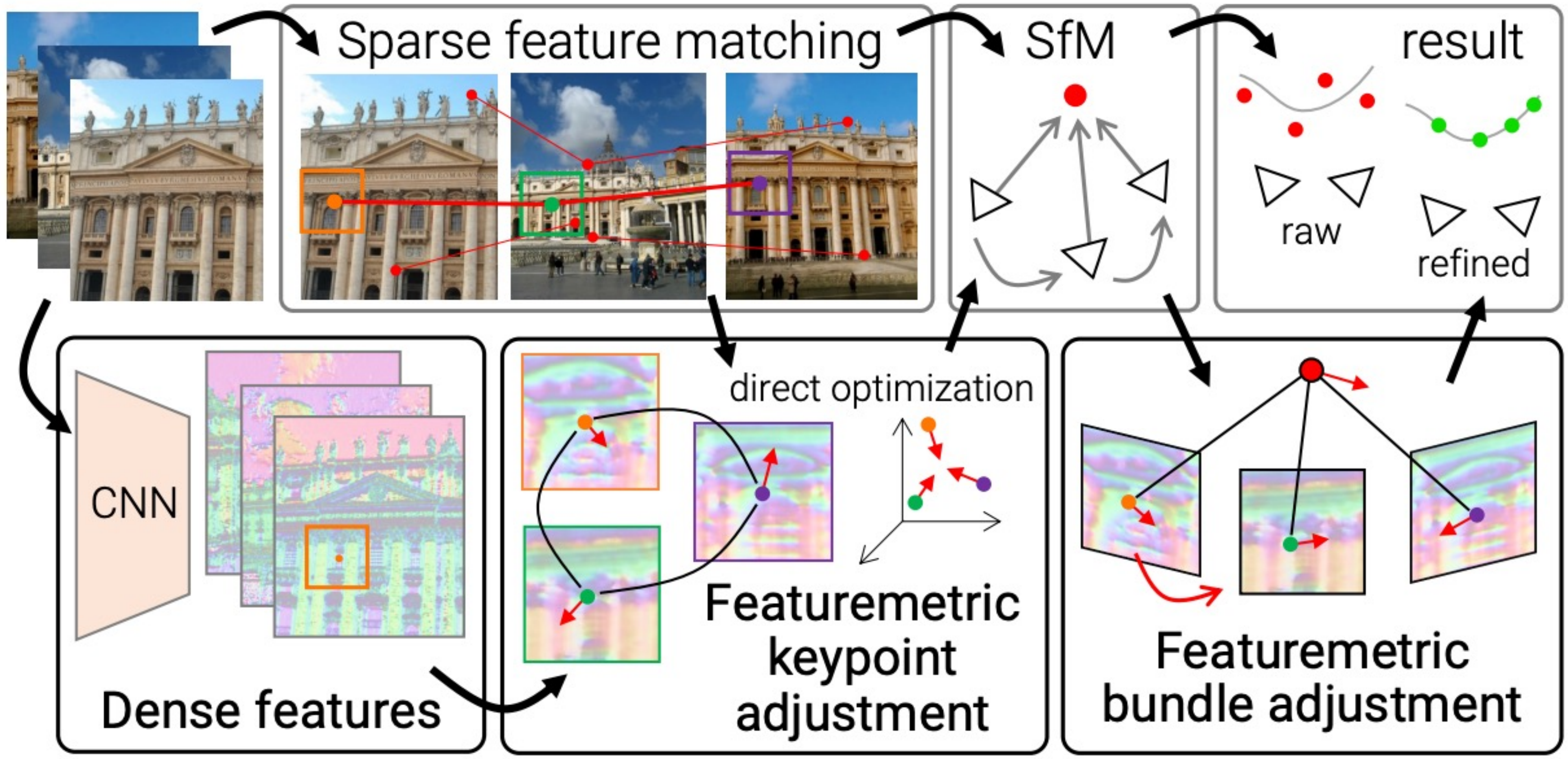


(a) Inferred depth map (b) Filtered depth map (c) GT depth map

(d) Reference image (e) Fused point cloud (f) GT point cloud

Y. Yao et al. MVSNet: Depth Inference for Unstructured Multi-view Stereo. ECCV 2018

Deep learning for improving SFM



P. Lindenberger et al. [Pixel-Perfect Structure-from-Motion with Featuremetric Refinement](). ICCV 2021