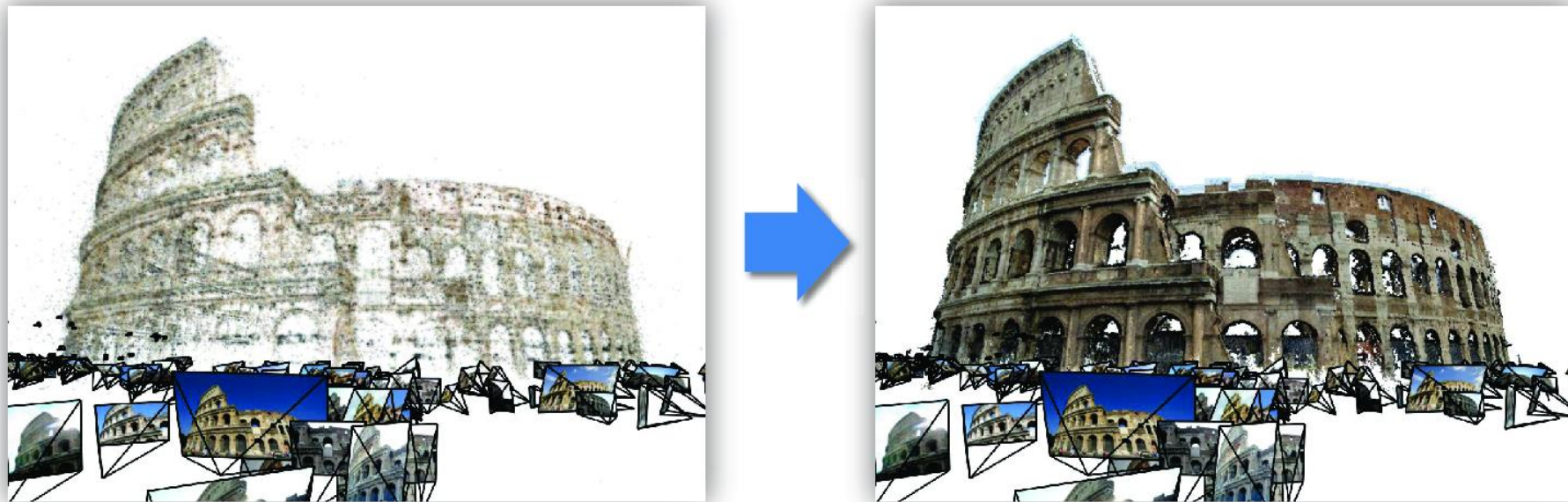


Multiview stereo

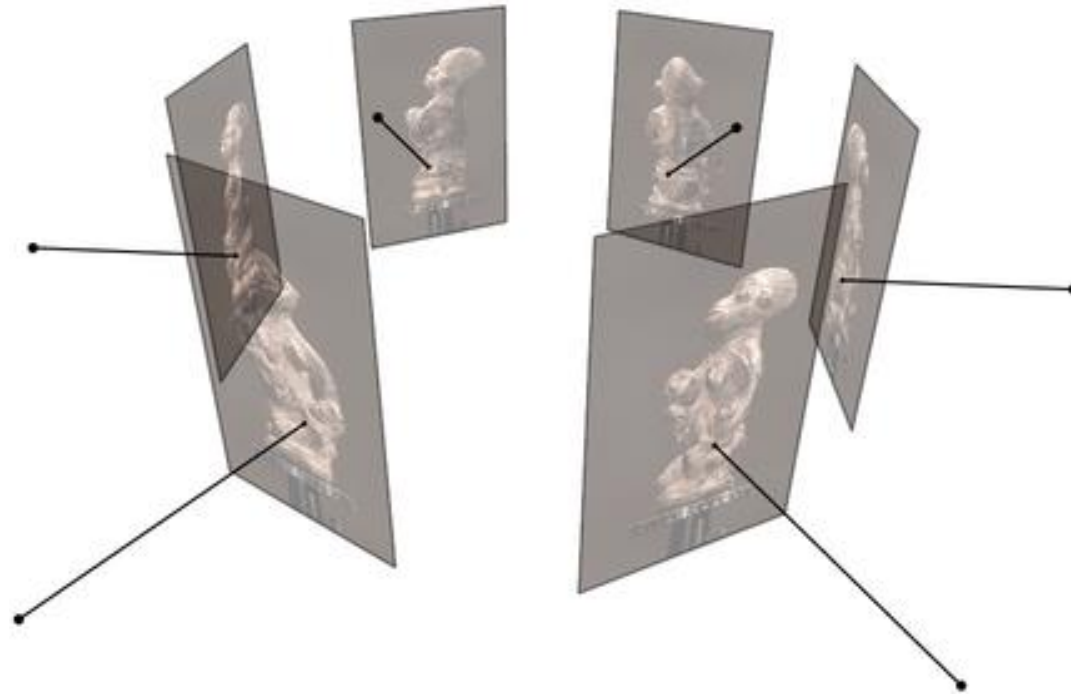


Slides largely from

- <http://luthuli.cs.uiuc.edu/~daf/courses/CV23/planned.html>
- https://www.cs.cornell.edu/courses/cs5670/2018sp/lectures/lec16_mvs.pdf

Multi-view stereo

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



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- “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

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 - Arbitrary number of images (from two to thousands)
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 - Calibration may be known or unknown
- “Representation of 3D shape”
 - Depth maps
 - Meshes
 - Point clouds
 - Volumetric models
 -

Outline

- Applications
- Two view stereo
- Multi-view stereo
 - Plane sweep stereo
 - Patch-based multi-view stereo
- Demonstration

Applications

Whistle in the Form of Female Figure *600 AD - 900 AD*



☰ Details Los Angeles County Museum of Art



Los Angeles County Museum of Art



Sculpture



Mexico

Share

Compare

Saved 0

Discover

Google

Applications



Data SIO, NOAA, U.S. Navy, NGA, GEBCO

Google earth

Source: N. Snively

Applications



Outline

- Applications
- **Two view stereo**
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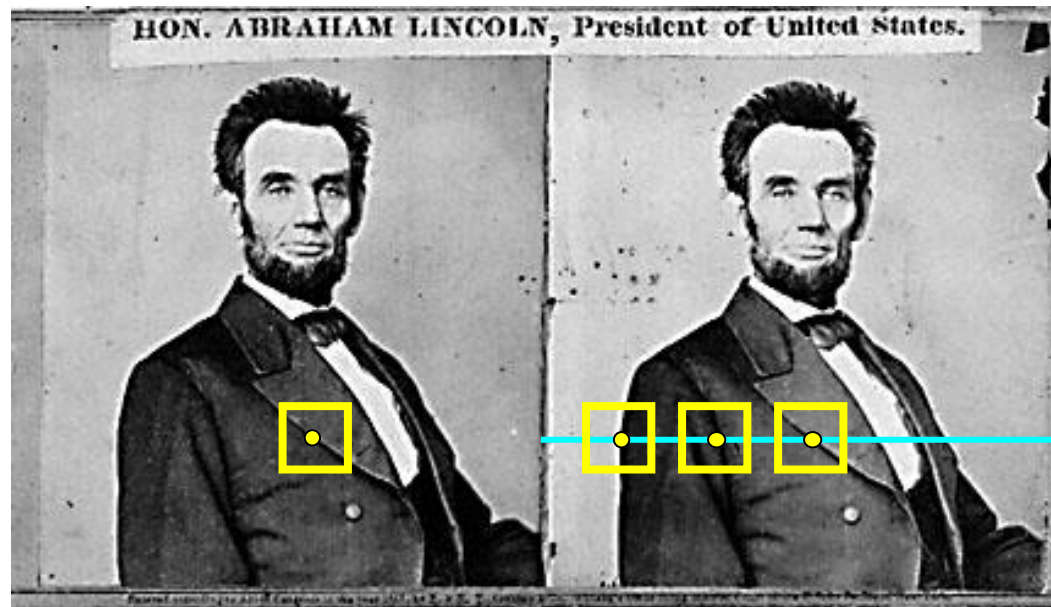


Problem formulation

- **Given:** stereo pair (assumed calibrated)
- **Wanted:** dense depth map



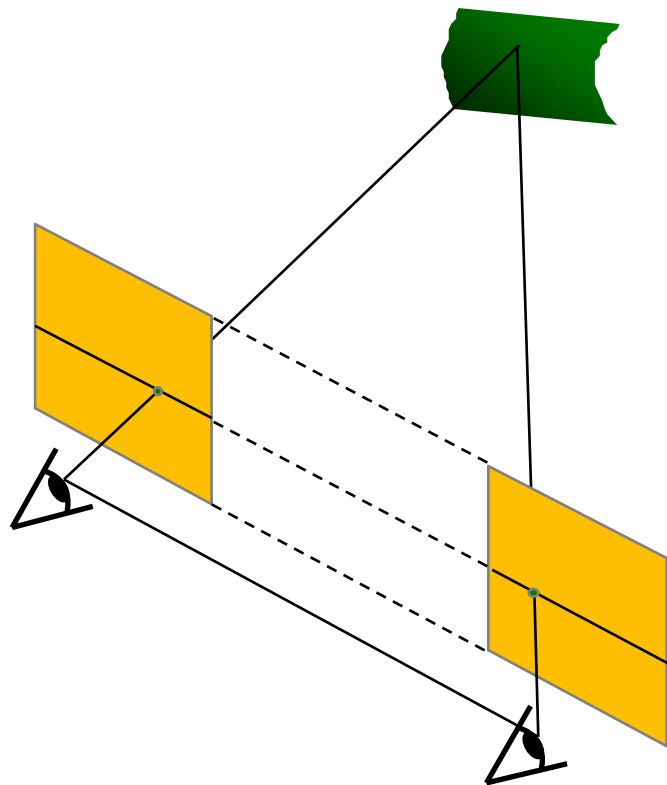
Basic stereo matching algorithm



- For each pixel in the first image
 - Find corresponding epipolar line in the right image
 - Examine all pixels on the epipolar line and pick the best match
 - Triangulate the matches to get depth information
- Simplest case: epipolar lines are corresponding scanlines
 - When does this happen?



Parallel images

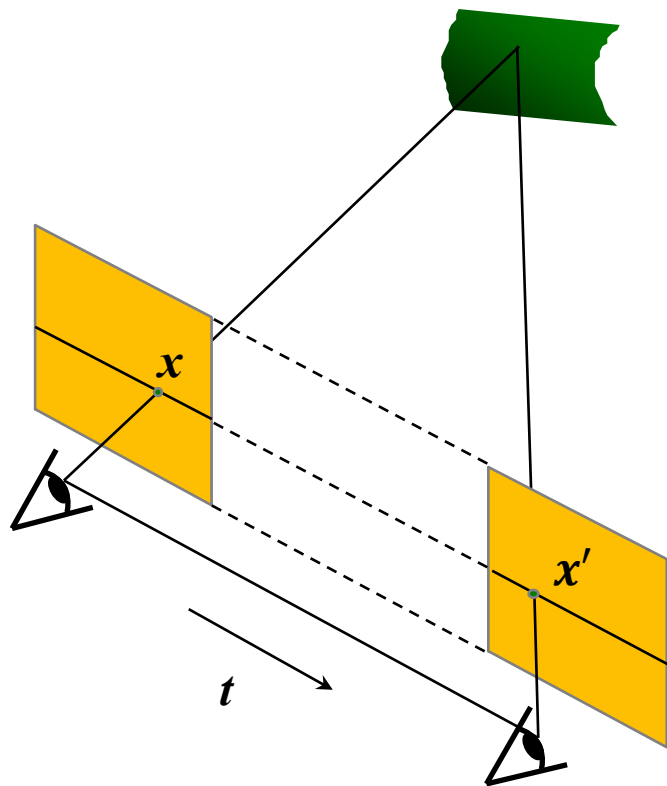


- Image planes of cameras are parallel to each other and to the baseline
- Camera centers are at the same height
- Focal lengths are the same
- Then epipolar lines fall along horizontal scan lines of the images

Can you mathematically prove this?



Essential matrix for parallel images



Epipolar constraint:

$$x'^T E x = 0, \quad E = [t_{\times}] R$$

$$R = I \quad t = (t, 0, 0)$$

$$E = [t_{\times}] R = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & -t \\ 0 & t & 0 \end{bmatrix}$$

$$(u' \ v' \ 1) \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & -t \\ 0 & t & 0 \end{bmatrix} \begin{pmatrix} u \\ v \\ 1 \end{pmatrix} = 0$$

$$(u' \ v' \ 1) \begin{pmatrix} 0 \\ -t \\ tv \end{pmatrix} = 0$$

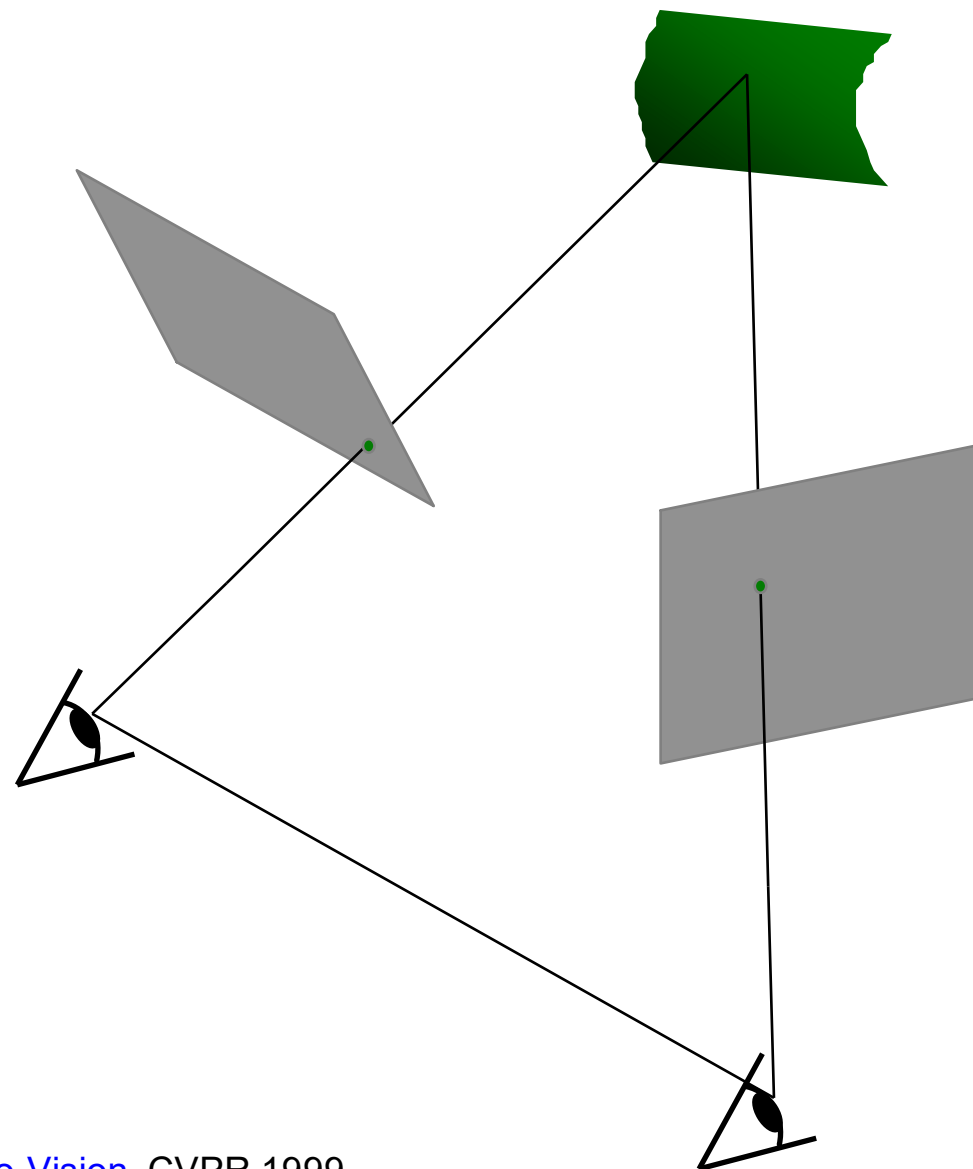
$$-tv + tv' = 0$$

$$v = v'$$

The y-coordinates of corresponding points are the same!

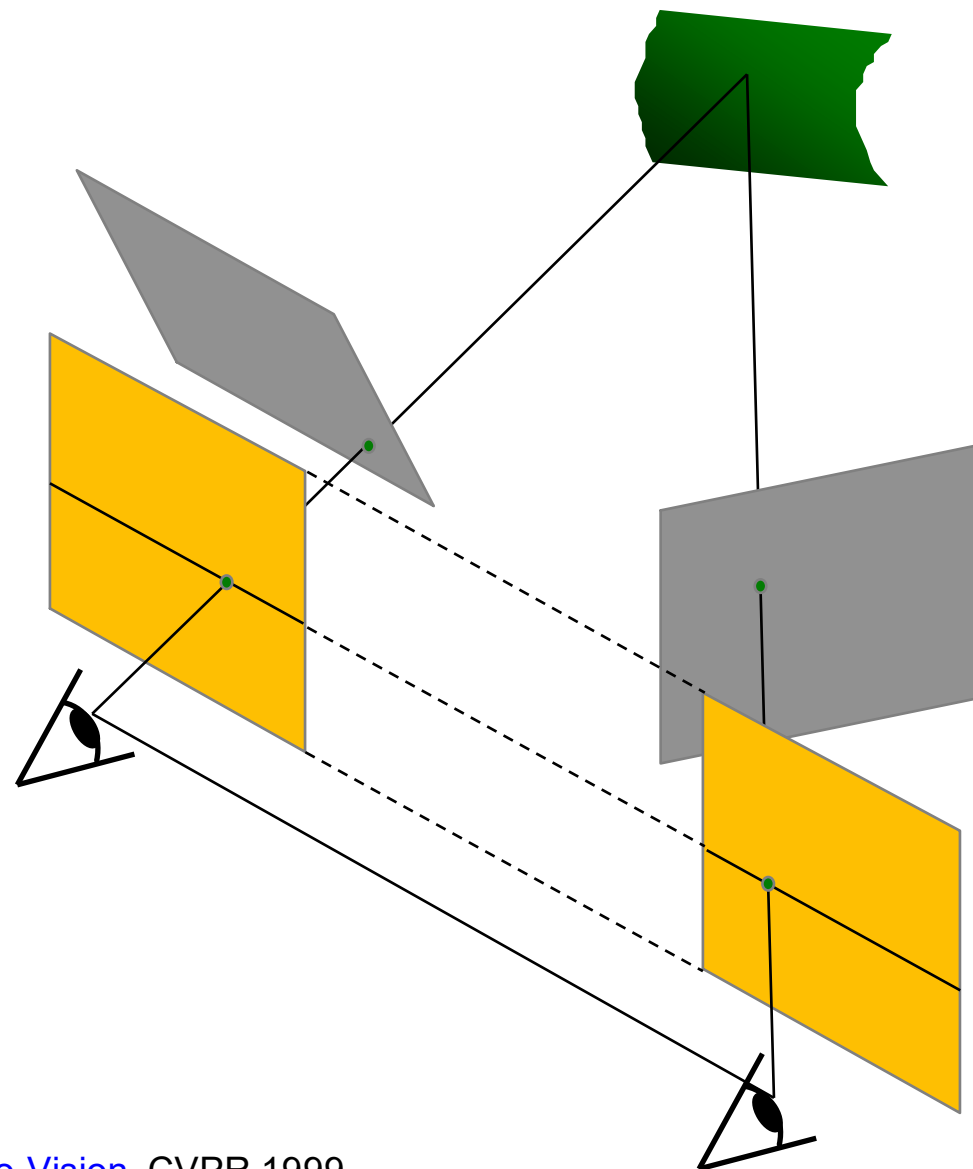
Stereo image rectification

- If the image planes are not parallel, we can find homographies to project each view onto a common plane parallel to the baseline



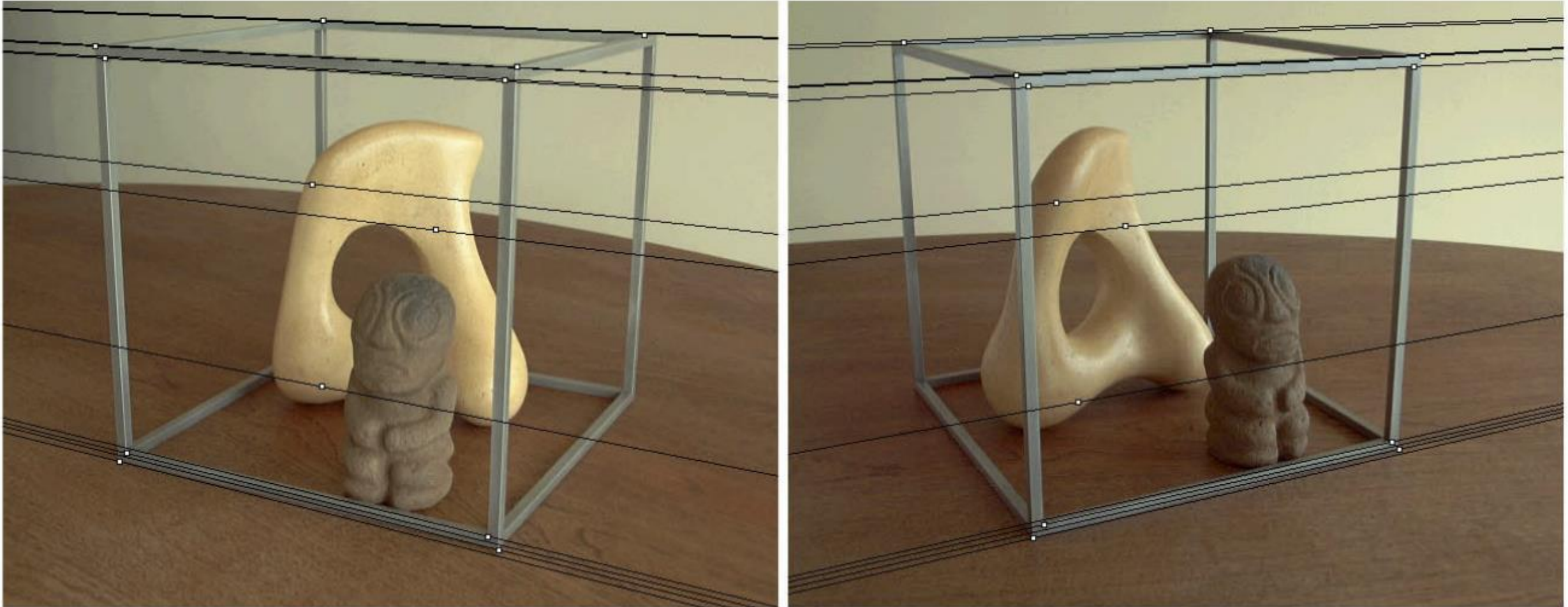
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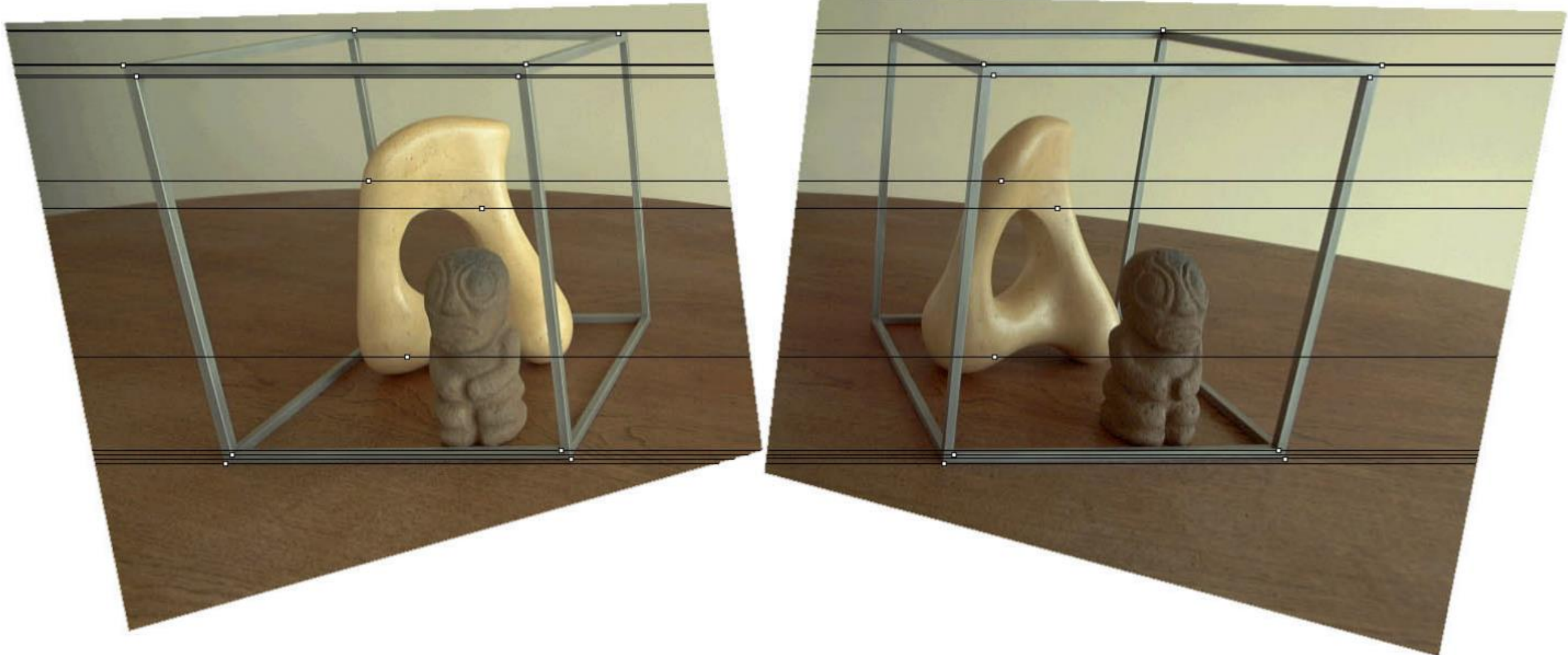
Stereo image rectification

- Before rectification:

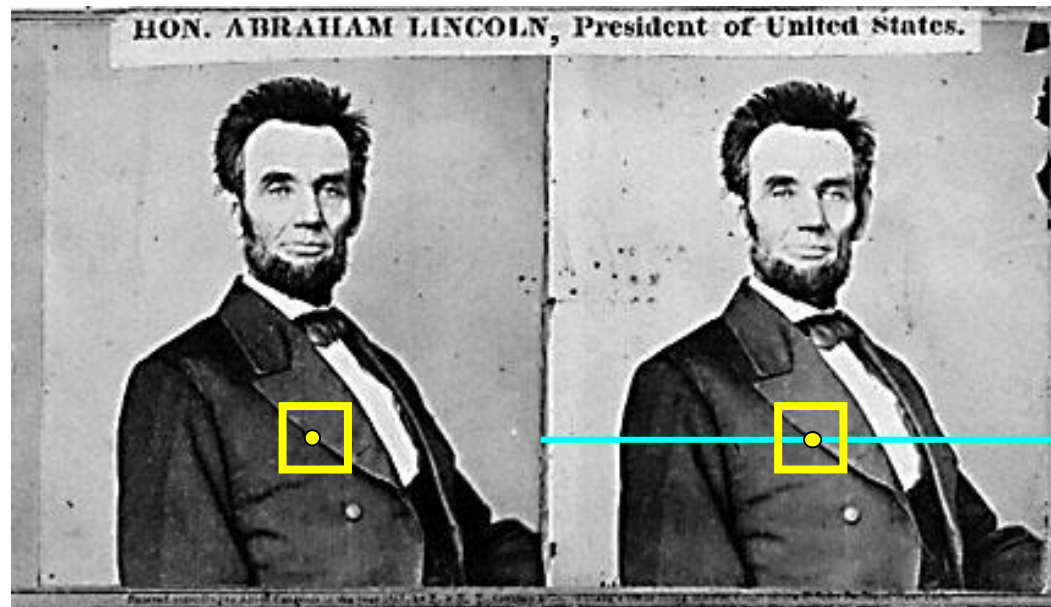


Stereo image rectification

- After rectification:

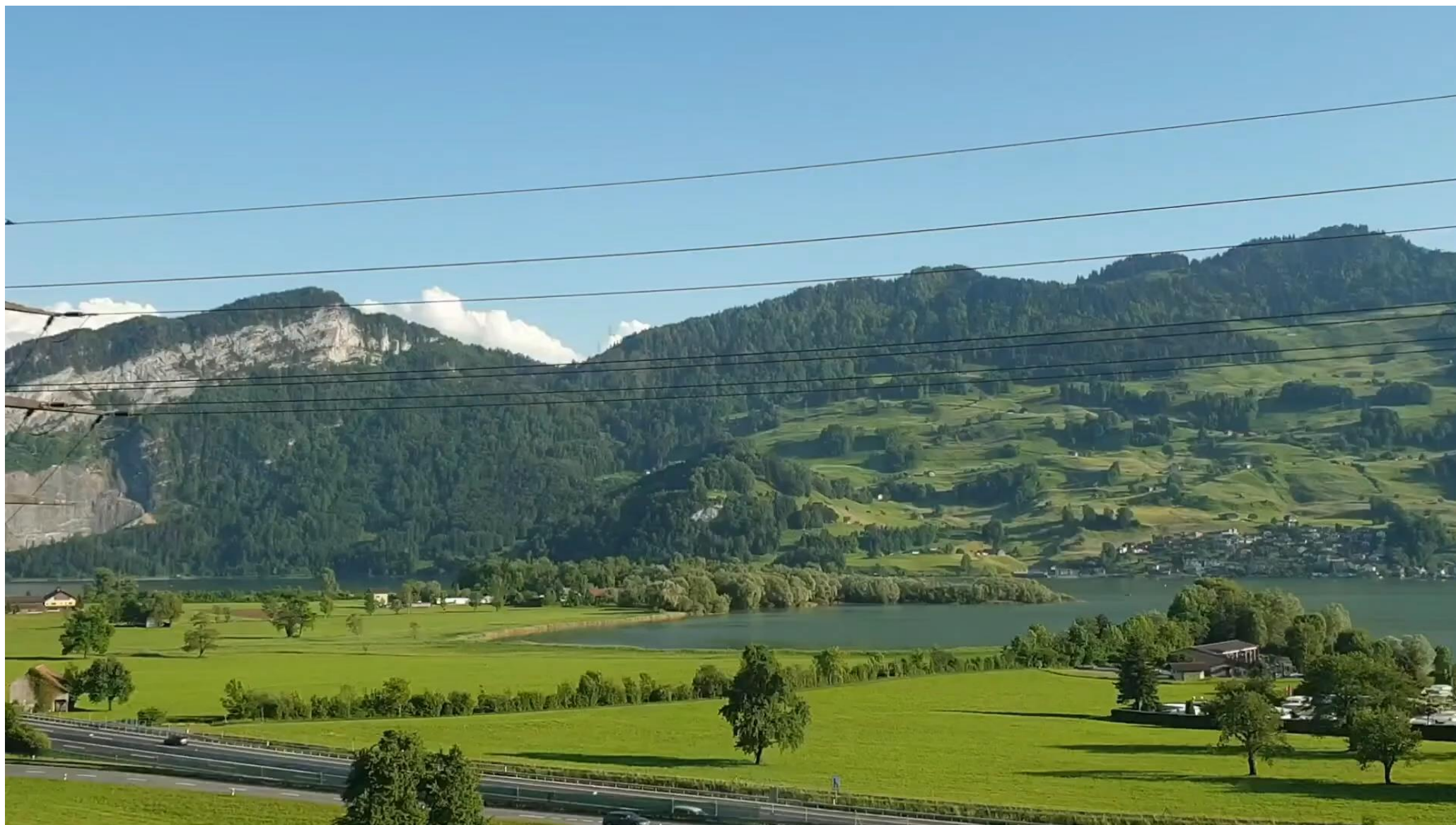


Basic stereo matching algorithm

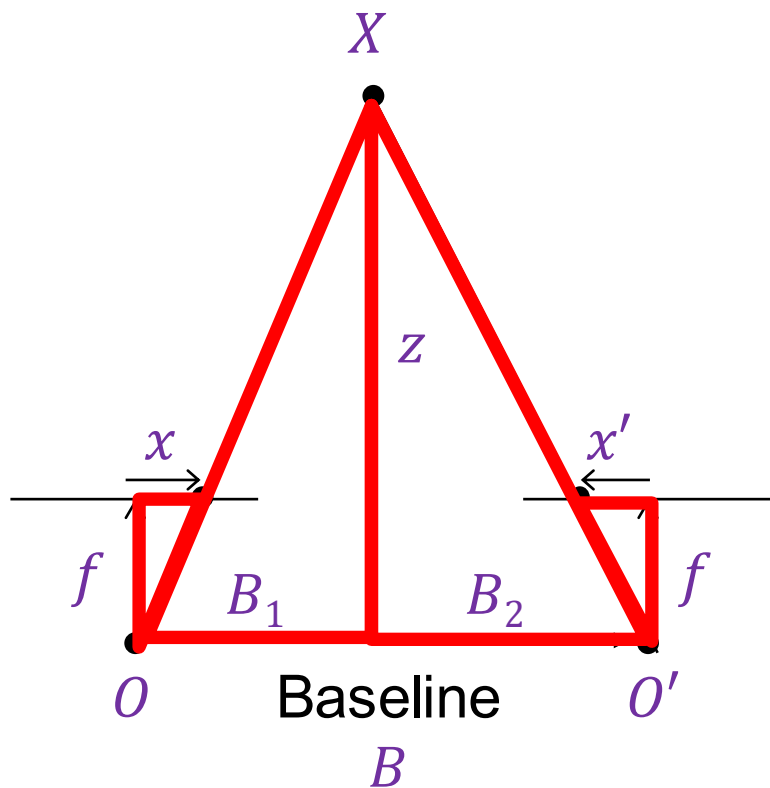


- If necessary, rectify the two stereo images to transform epipolar lines into scanlines
- For each pixel x in the first image
 - Find corresponding epipolar scanline in the right image
 - Examine all pixels on the scanline and pick the best match x'
- Triangulate the matches to get depth information

Depth from disparity



Depth from disparity



$$\frac{x}{f} = \frac{B_1}{z} \quad \frac{-x'}{f} = \frac{B_2}{z}$$

$$\frac{x - x'}{f} = \frac{B_1 + B_2}{z}$$

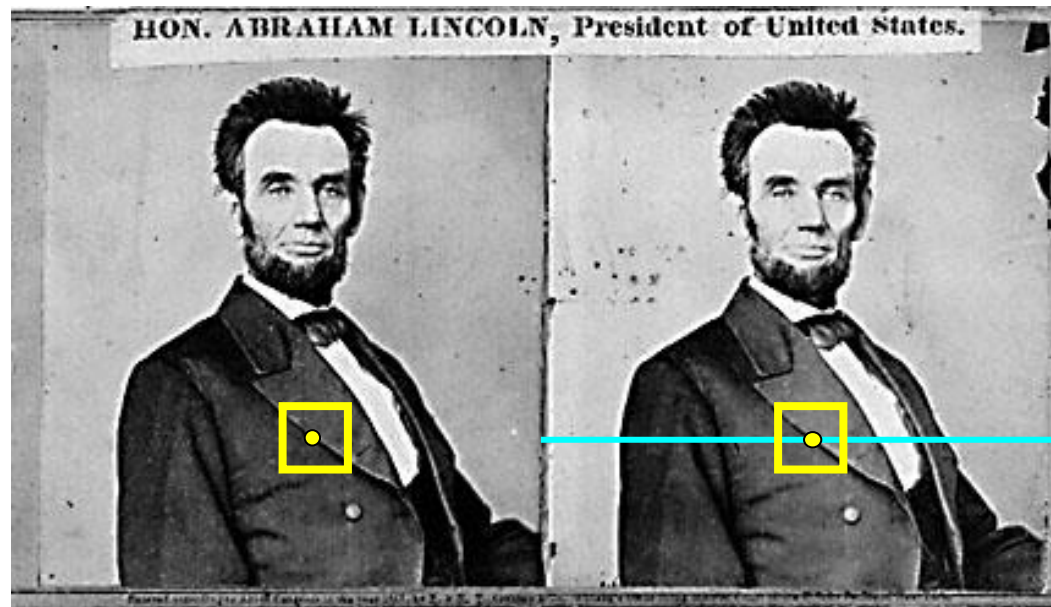
$$\underbrace{x - x'} = \frac{fB}{z}$$

$$z = \frac{fB}{x - x'}$$

Disparity is
inversely
proportional to
depth!

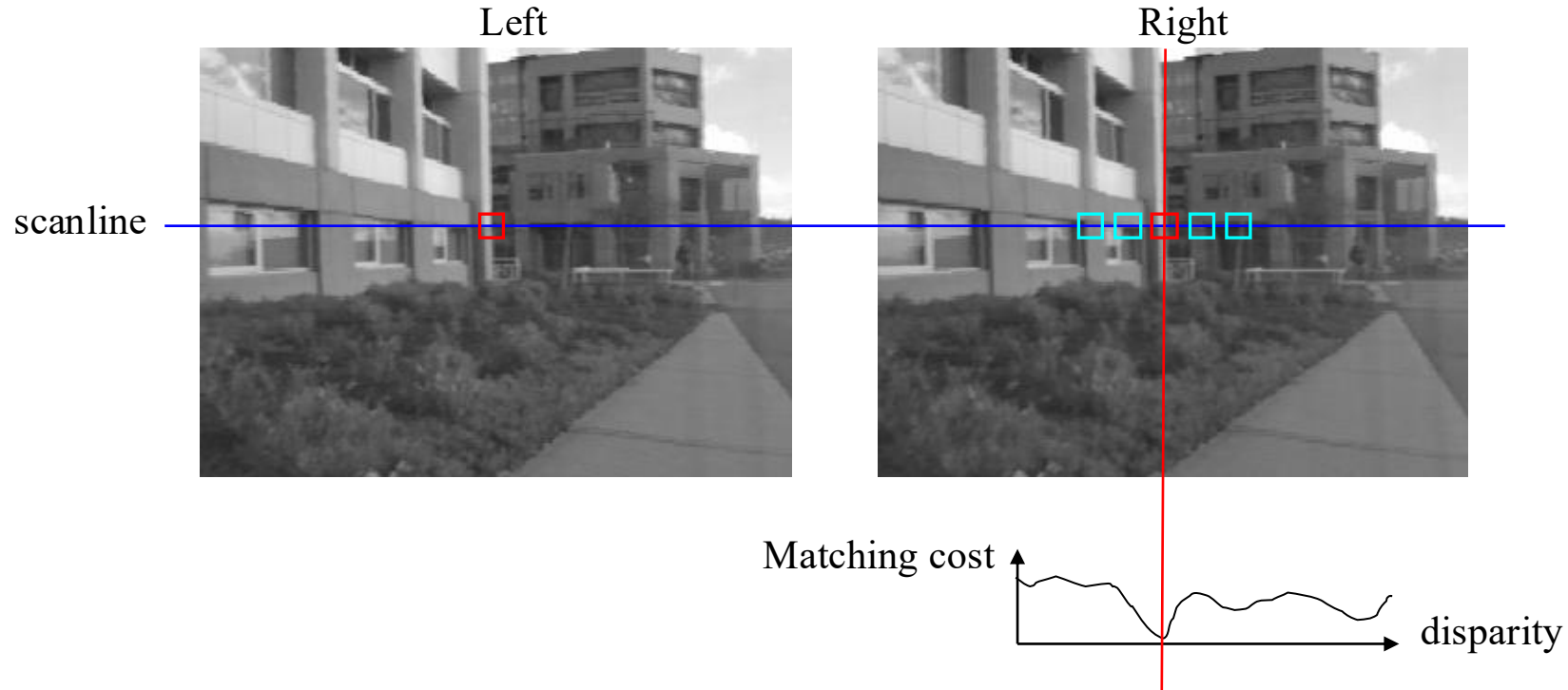


Basic stereo matching algorithm



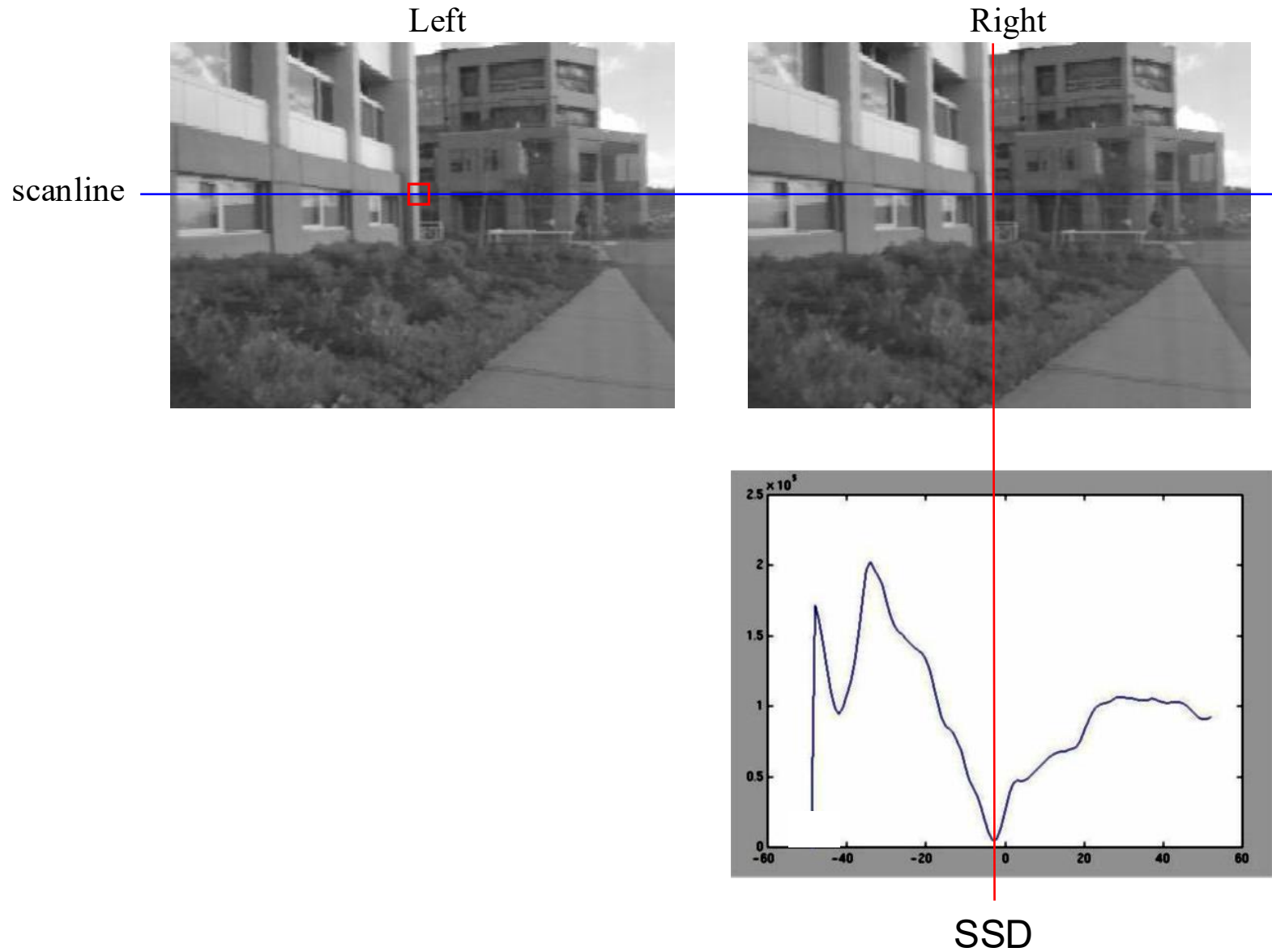
- If necessary, rectify the two stereo images to transform epipolar lines into scanlines
- For each pixel x in the first image
 - Find corresponding epipolar scanline in the right image
 - Examine all pixels on the scanline and pick the best match x'
 - Compute disparity $x - x'$ and set $\text{depth}(x) = Bf / (x - x')$

Correspondence search

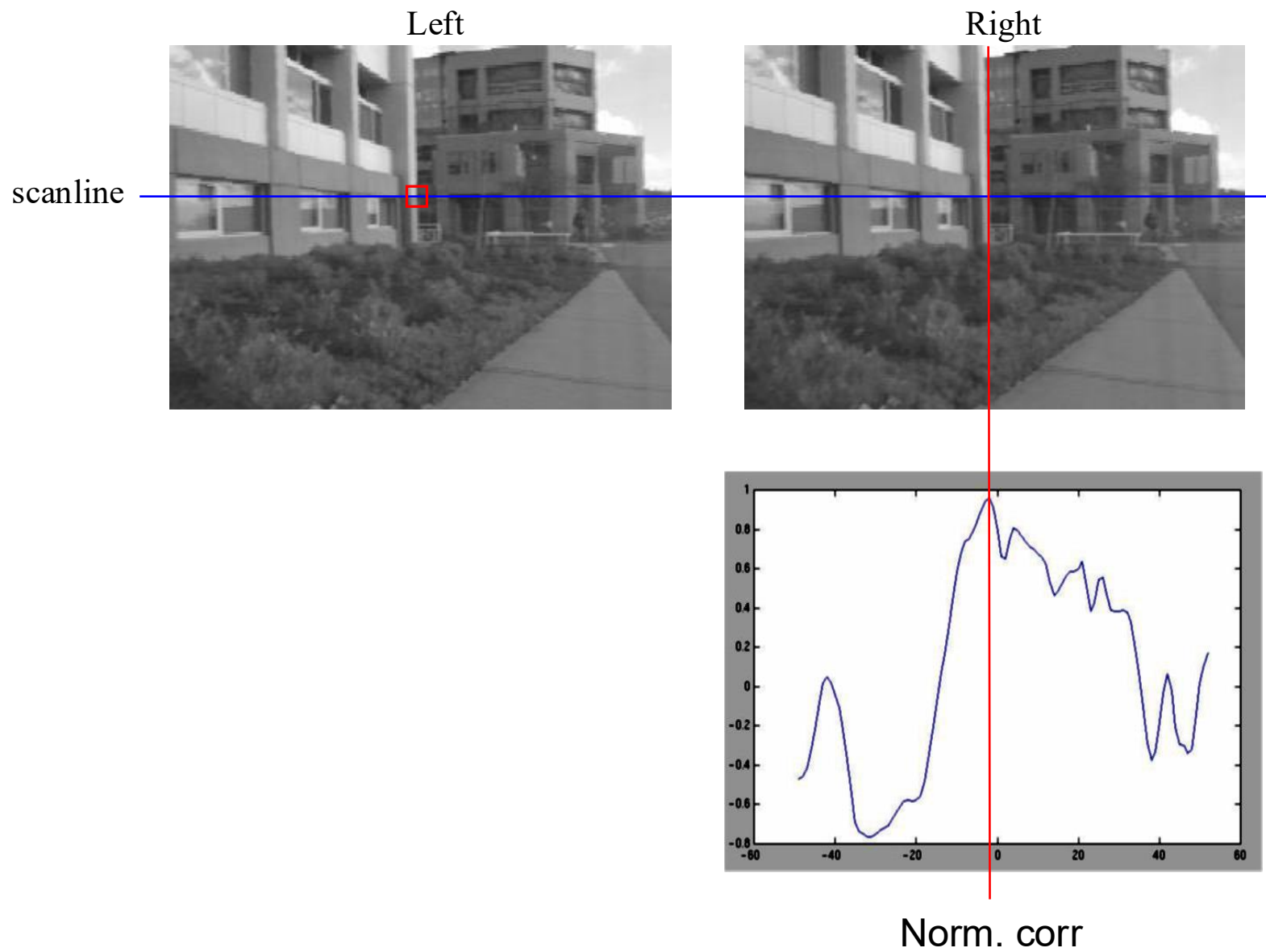


- Slide a window along the right scanline and compare contents of that window with the reference window in the left image
- Matching cost: SSD or normalized correlation

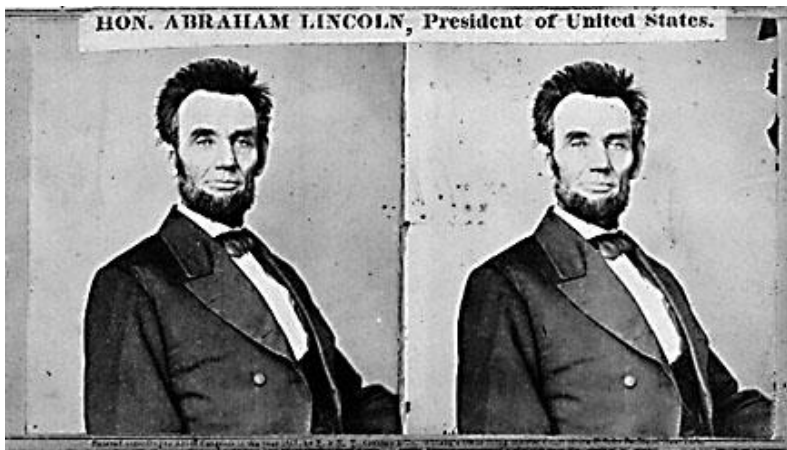
Correspondence search



Correspondence search



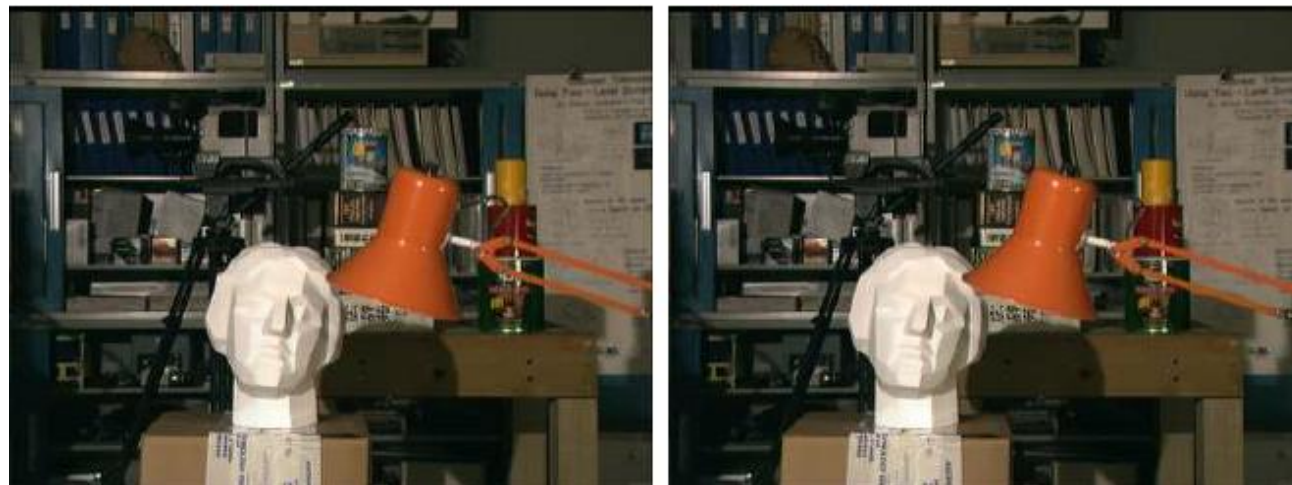
Where will basic window search fail?



Textureless surfaces

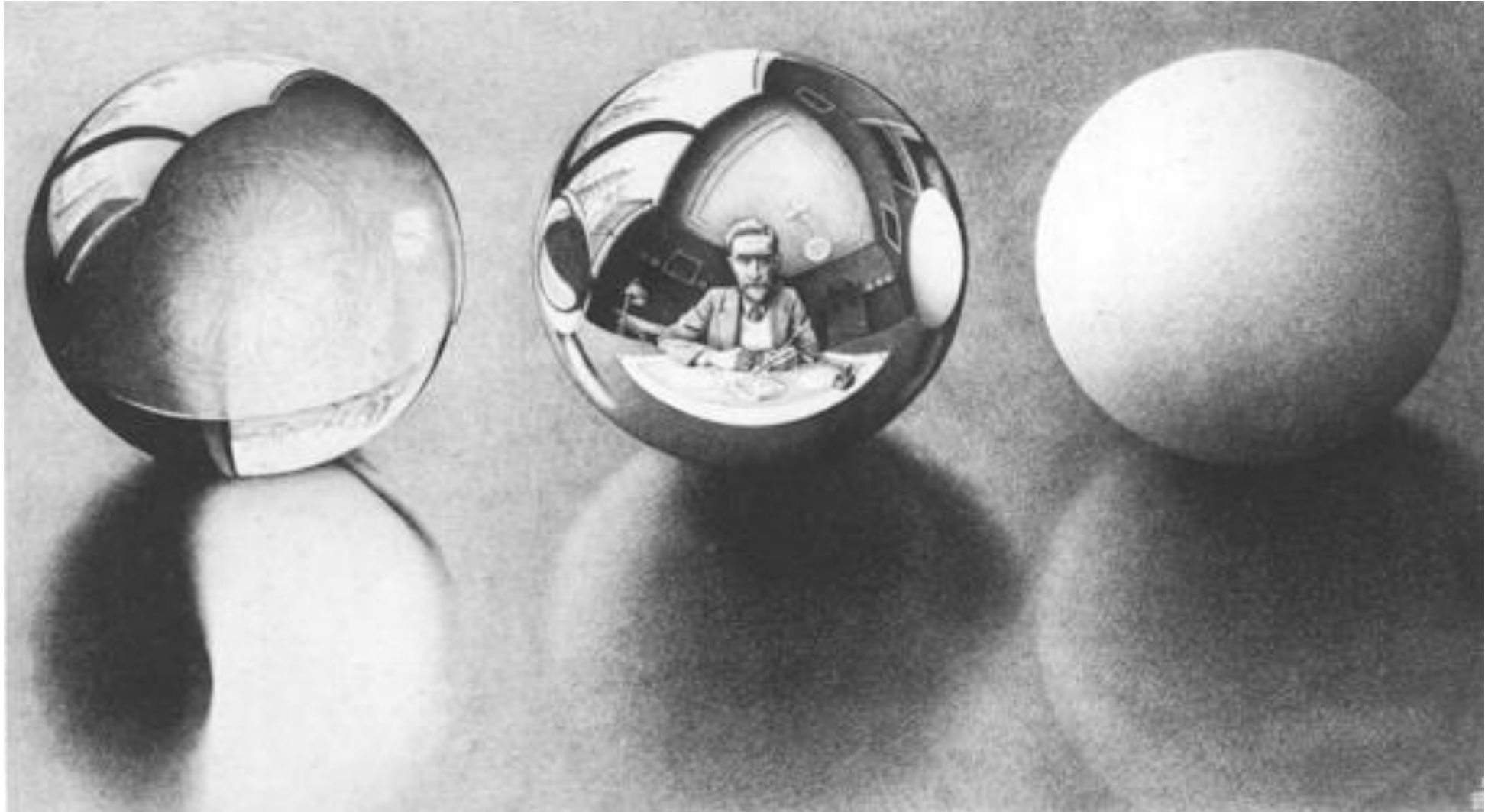


Occlusions, repetition



Non-Lambertian surfaces, specularities

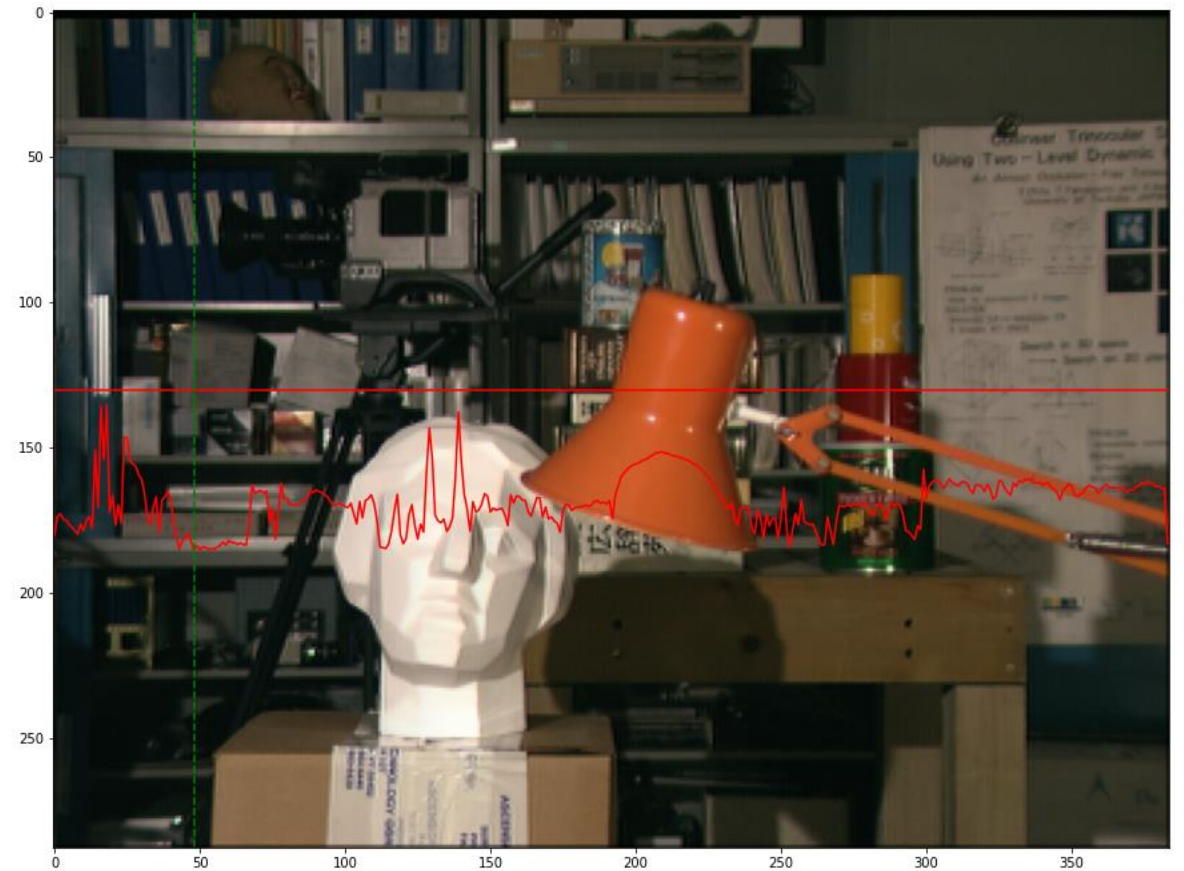
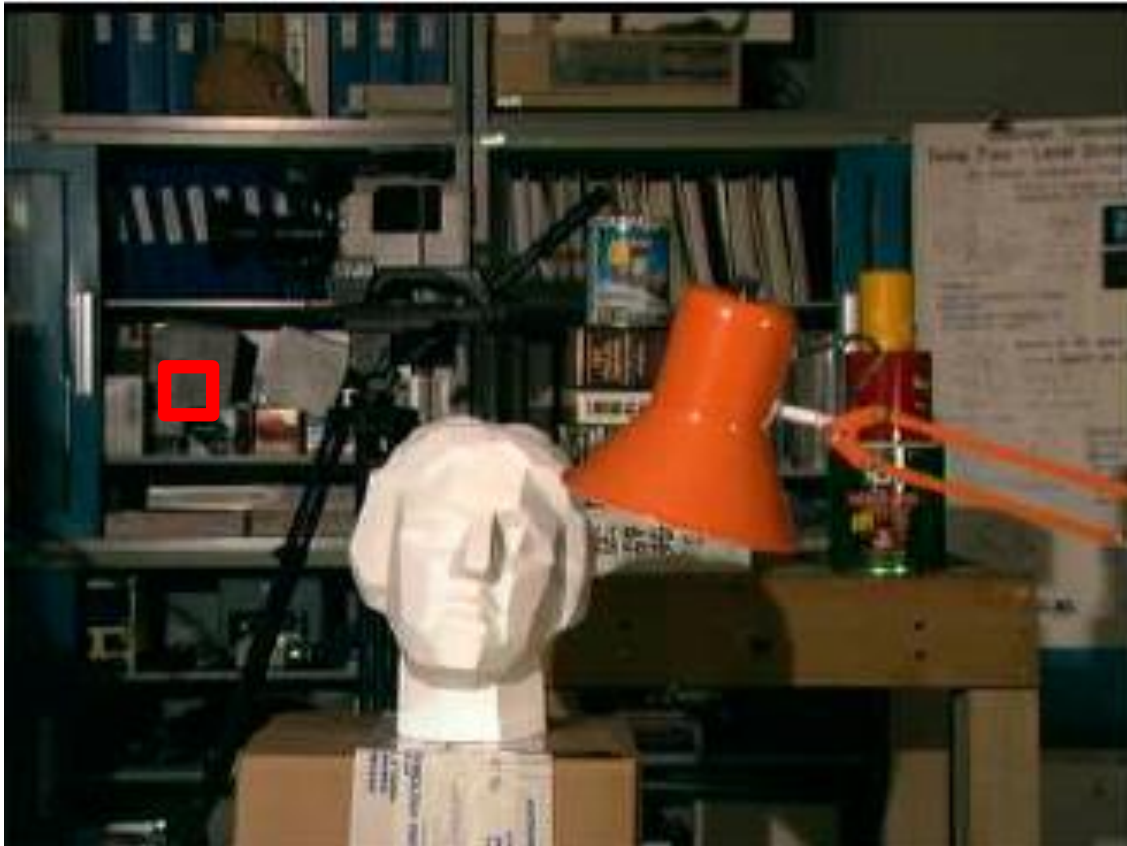
Non-Lambertian surfaces



The same object look different from different angles

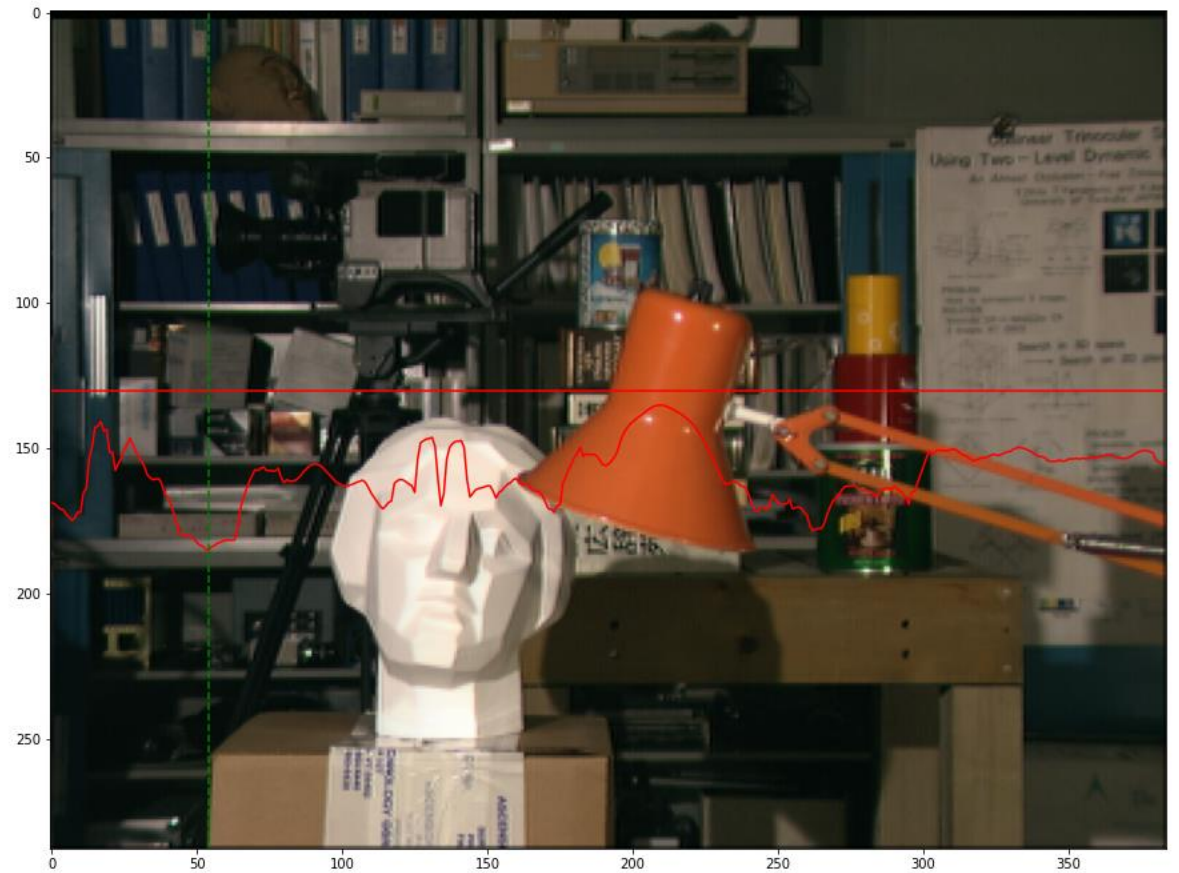
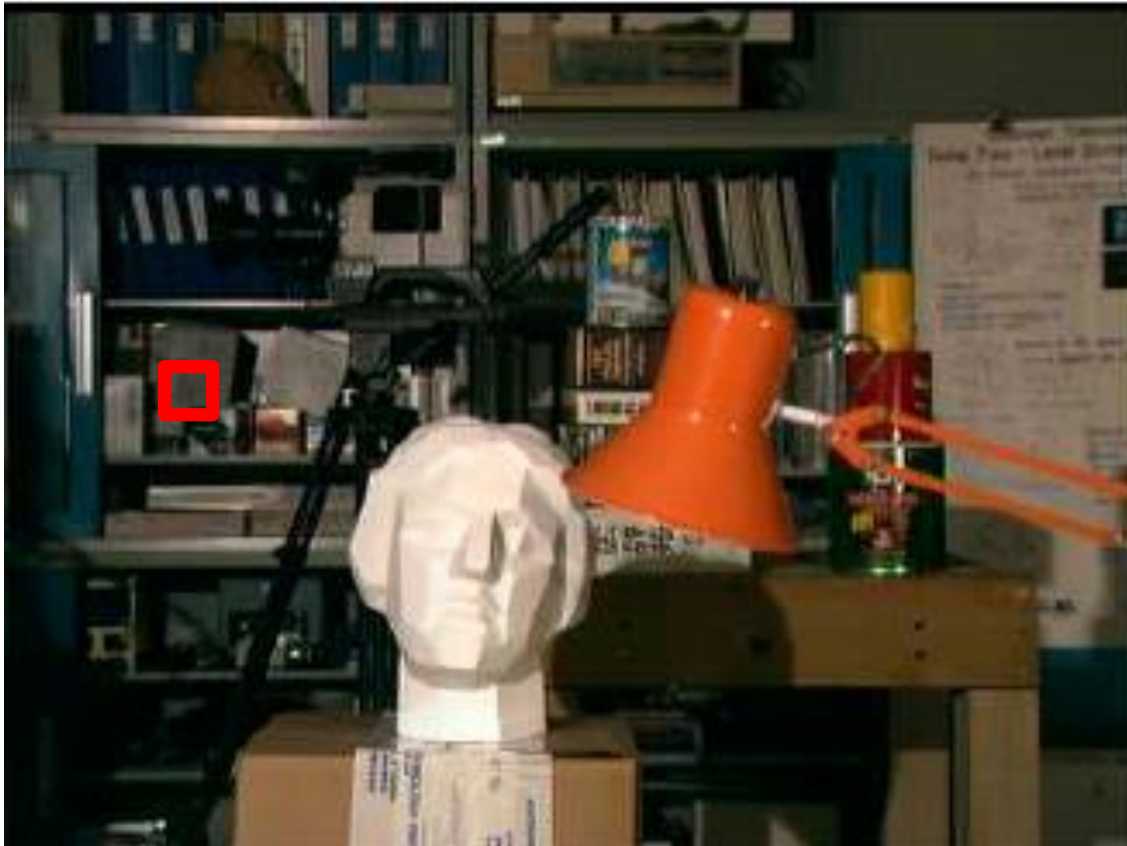
Example: Textured neighborhood

Window size: 1 pixel



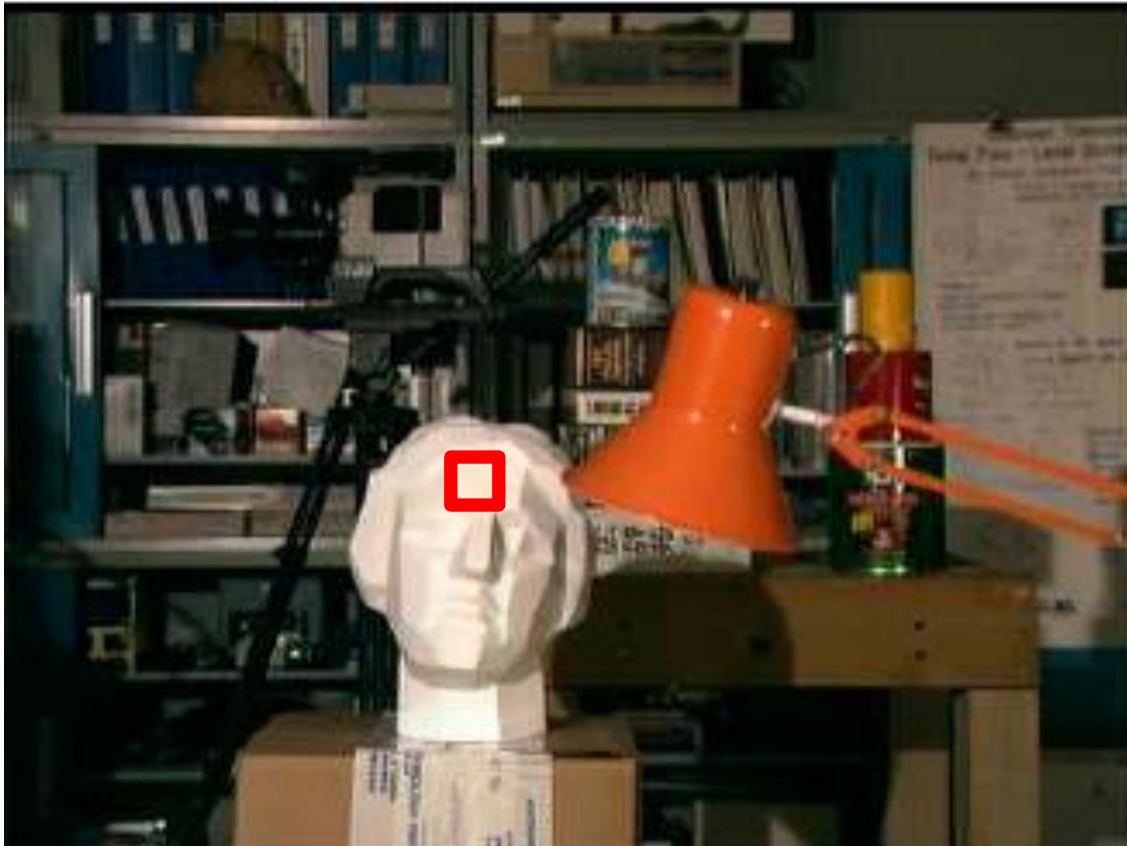
Example: Textured neighborhood

Window size: 7 pixels



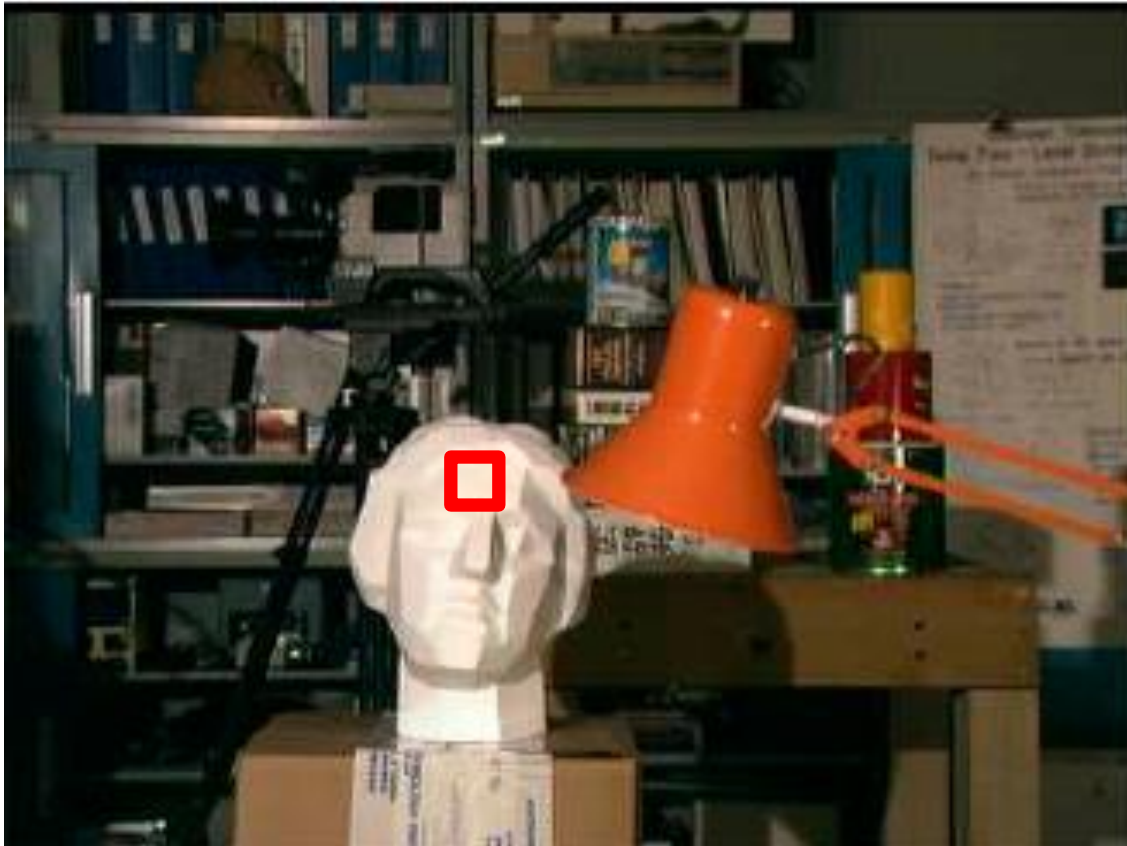
Example: Smooth neighborhood

Window size: 1 pixel



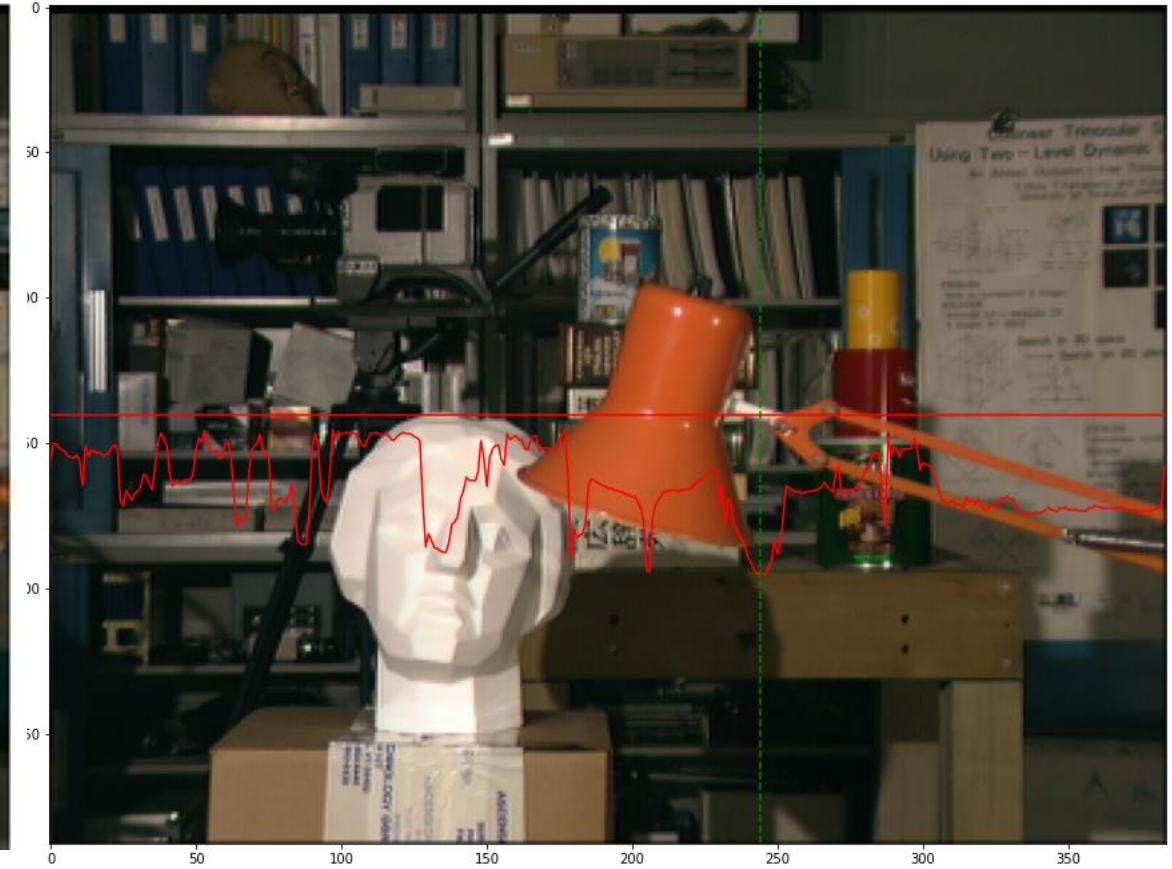
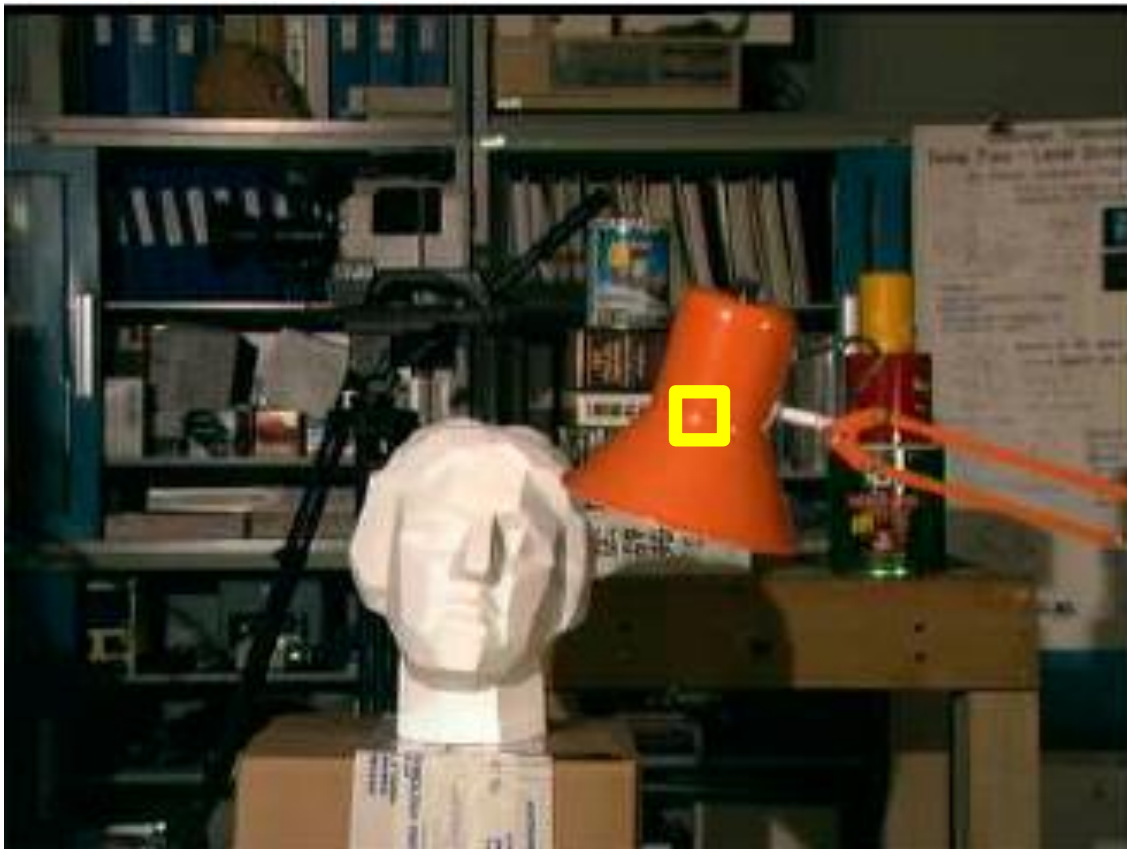
Example: Smooth neighborhood

Window size: 7 pixels



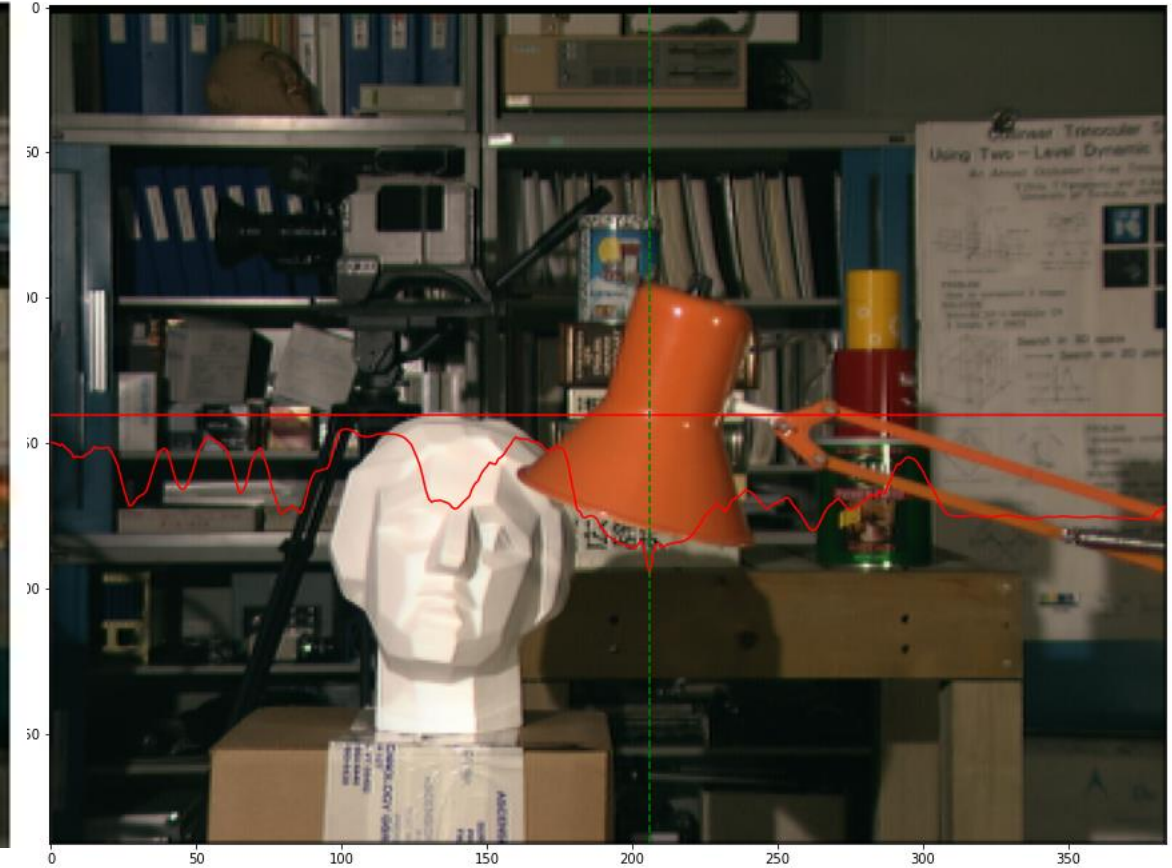
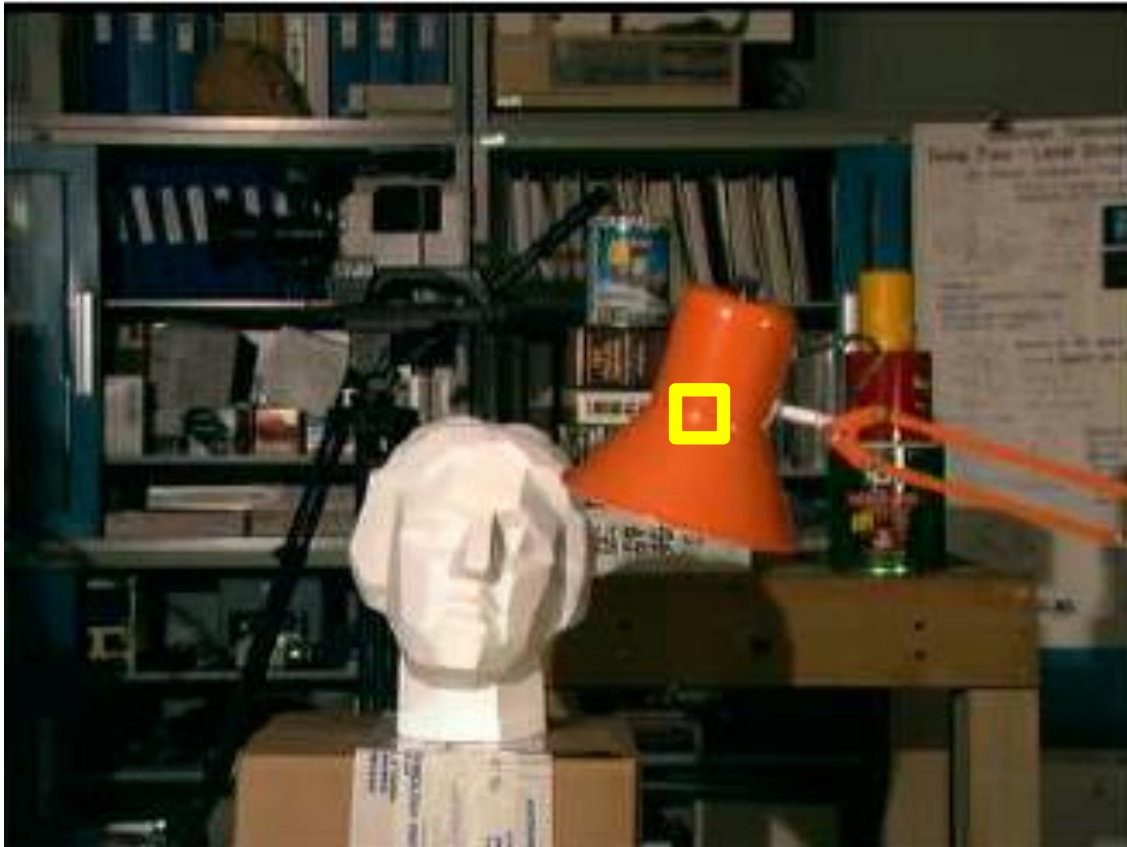
Example: Specular highlight

Window size: 1 pixel

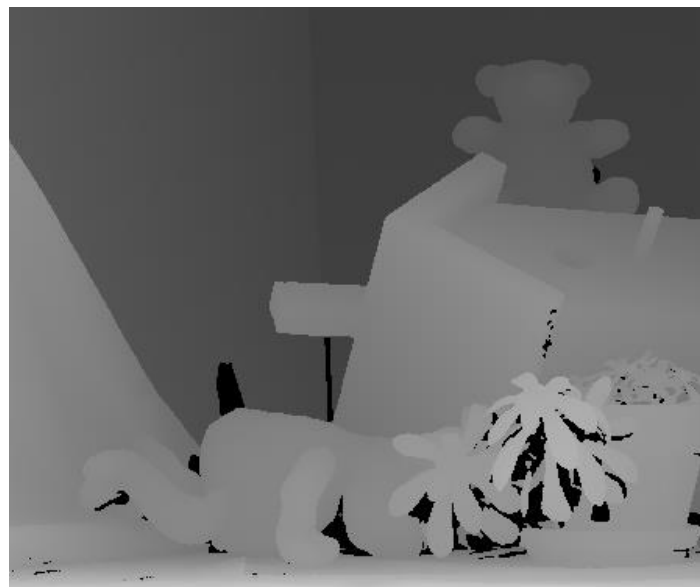
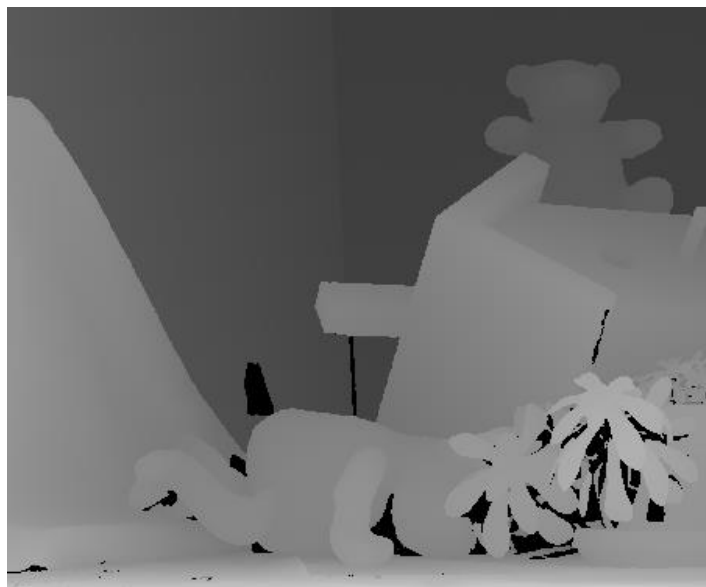


Example: Specular highlight

Window size: 7 pixels



Resulting depth maps



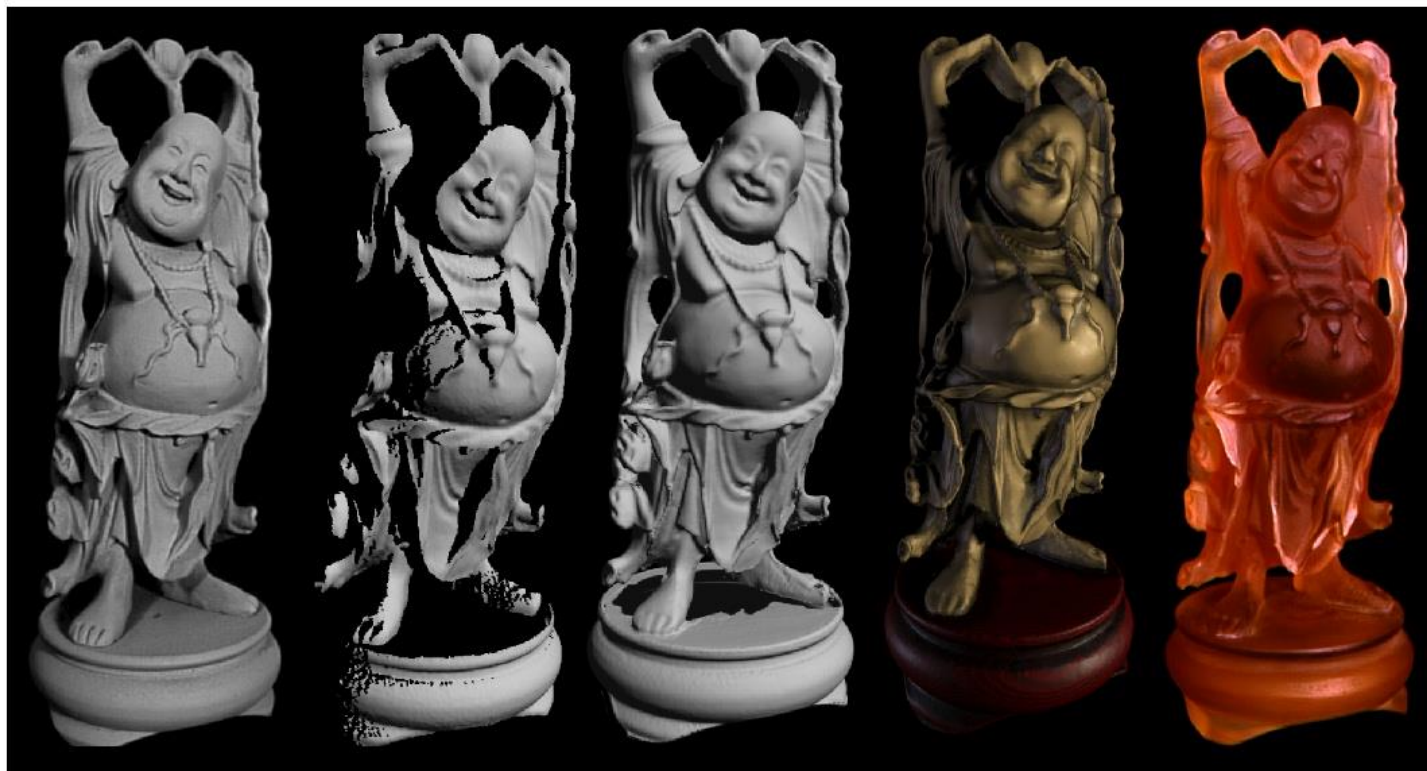
Kinect: Structured infrared light



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

Aligning range images

- A single range scan is not sufficient to capture a complex surface
- Need techniques to register multiple range images



B. Curless and M. Levoy, [A Volumetric Method for Building Complex Models from Range Images](#),
SIGGRAPH 1996

Aligning range images

- A single range scan is not sufficient to capture a complex surface
- Need techniques to register multiple range images

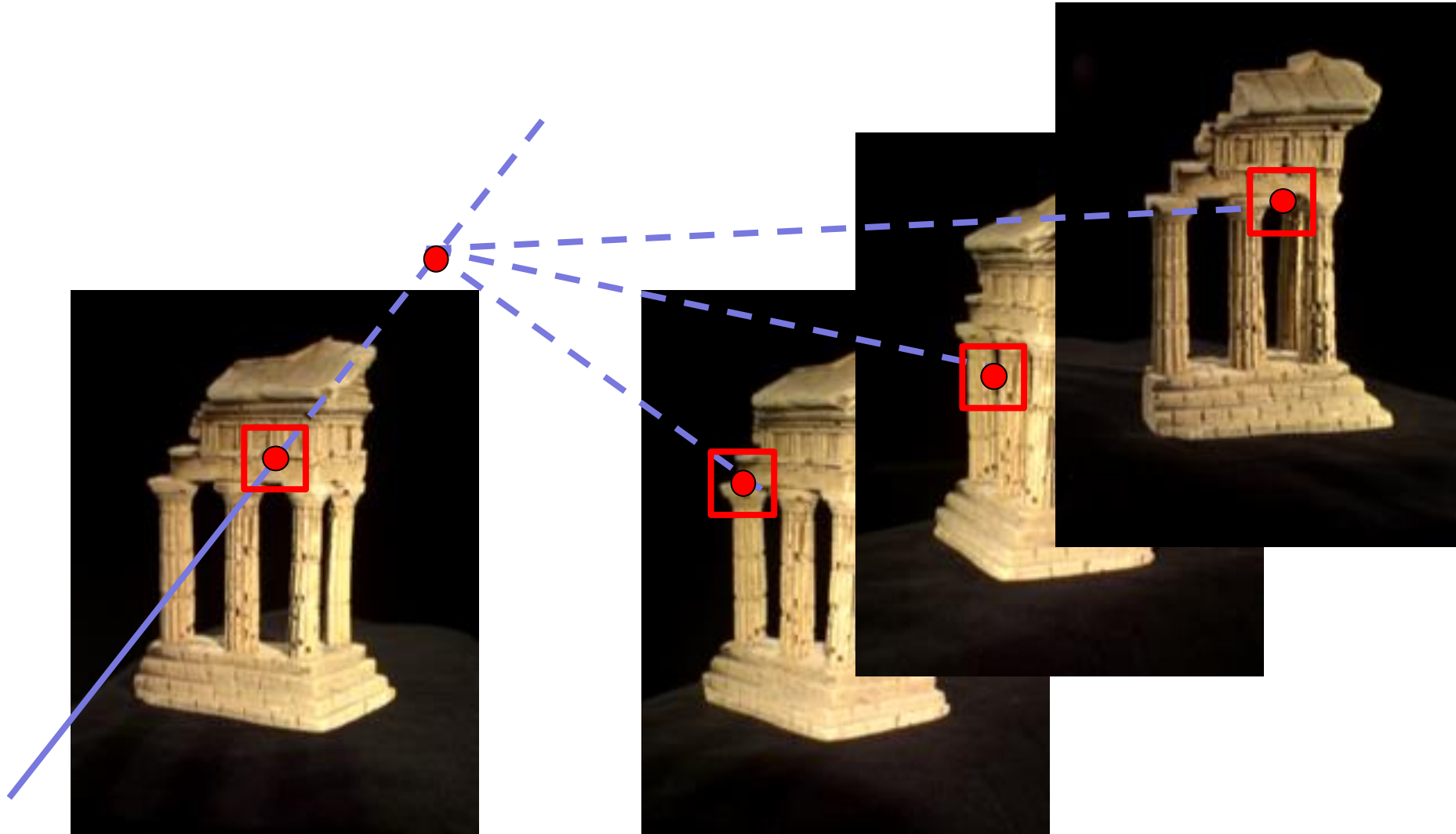
... which brings us to *multi-view stereo*

Outline

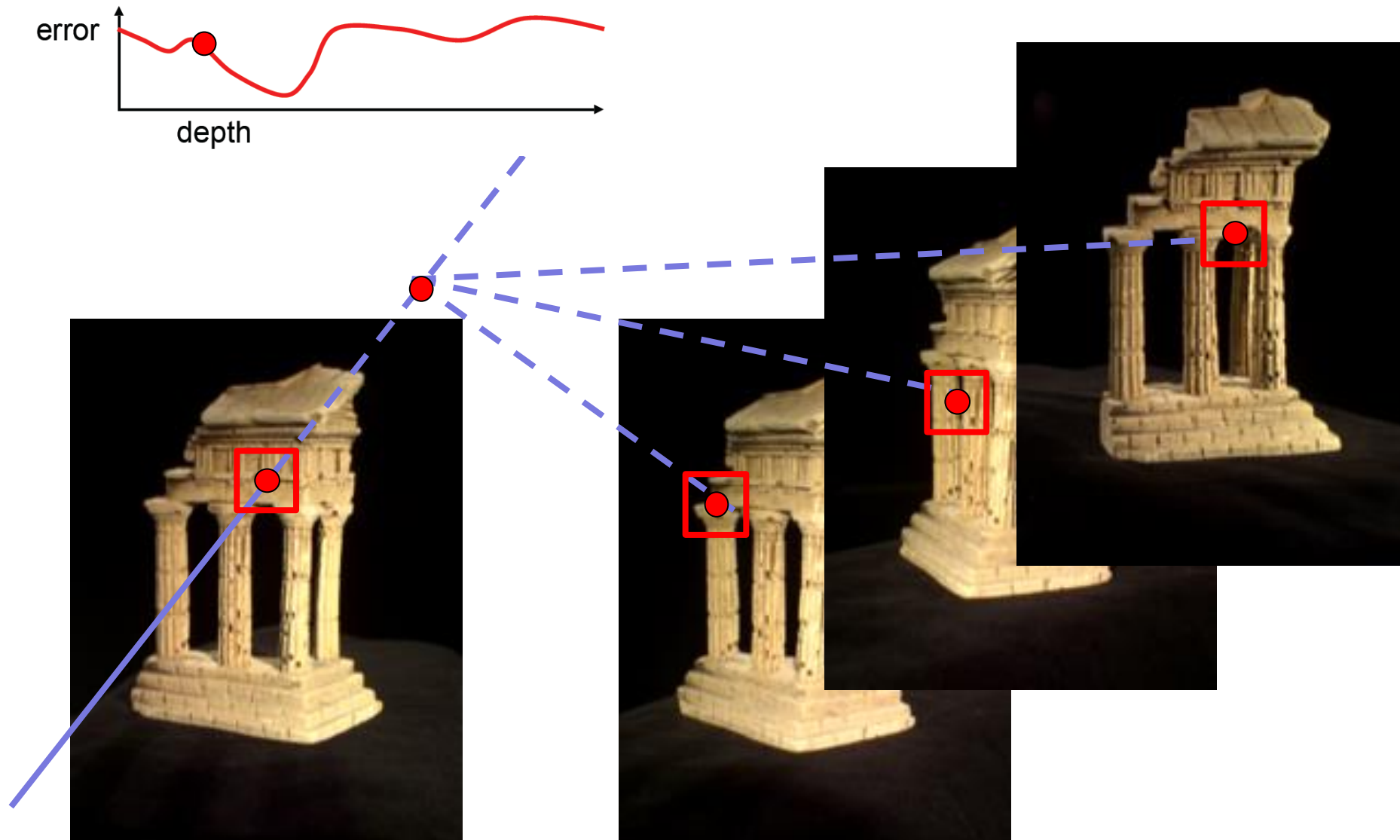
- Applications
- Two view stereo
- **Multi-view stereo**
 - Plane sweep stereo
 - Patch-based multi-view stereo
 - Depth map fusion
- Demonstration



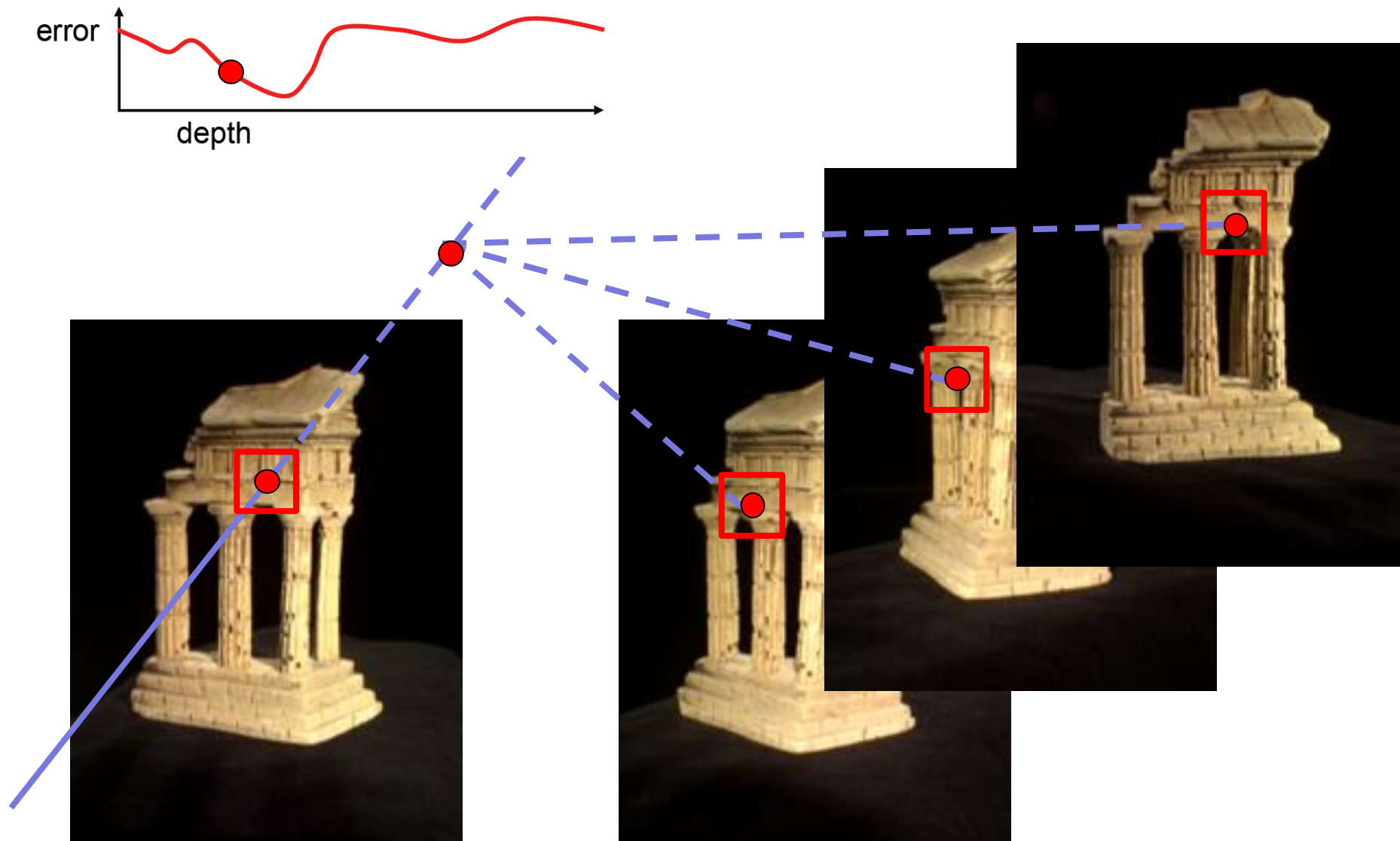
Multi-view stereo: Basic idea



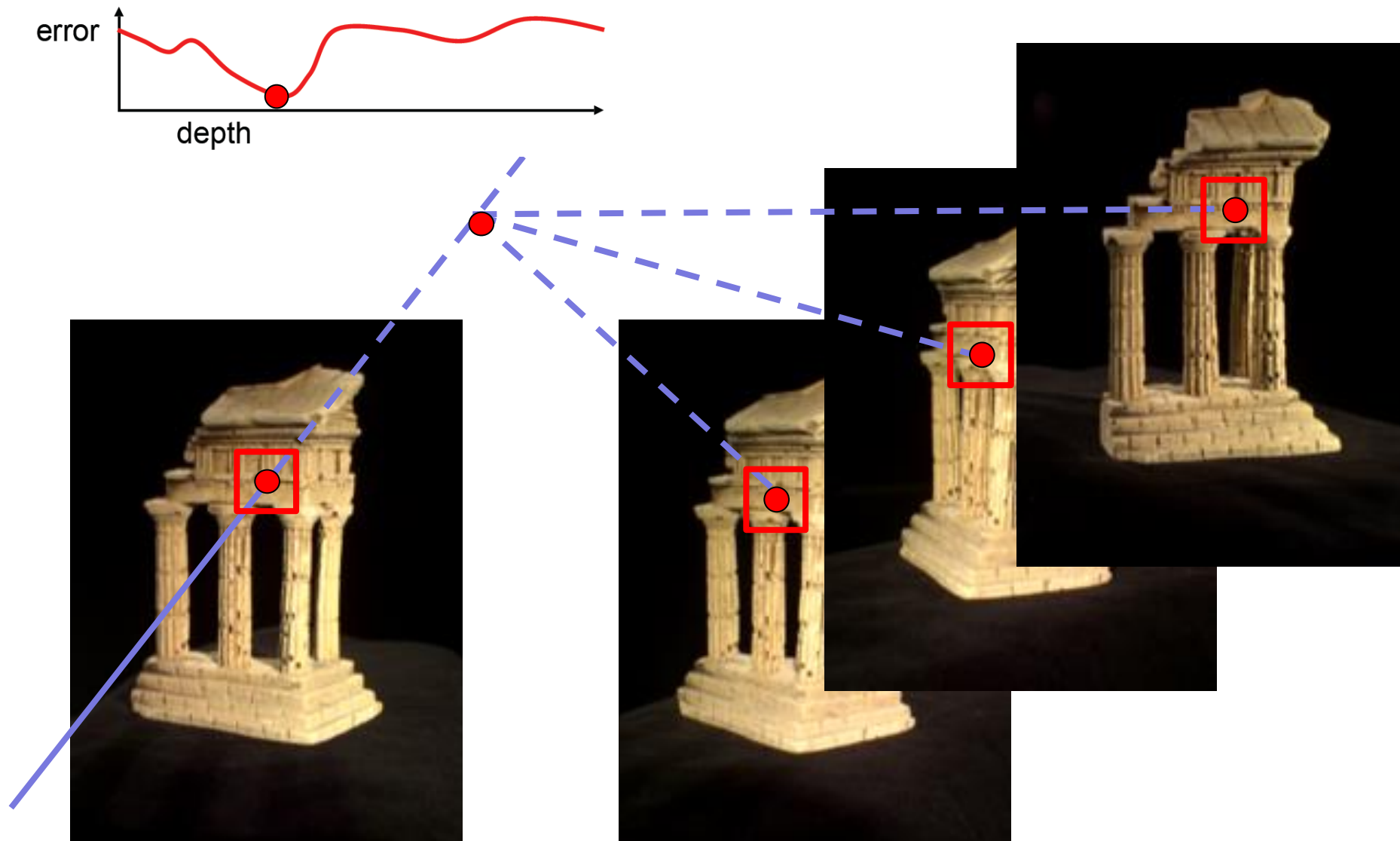
Multi-view stereo: Basic idea



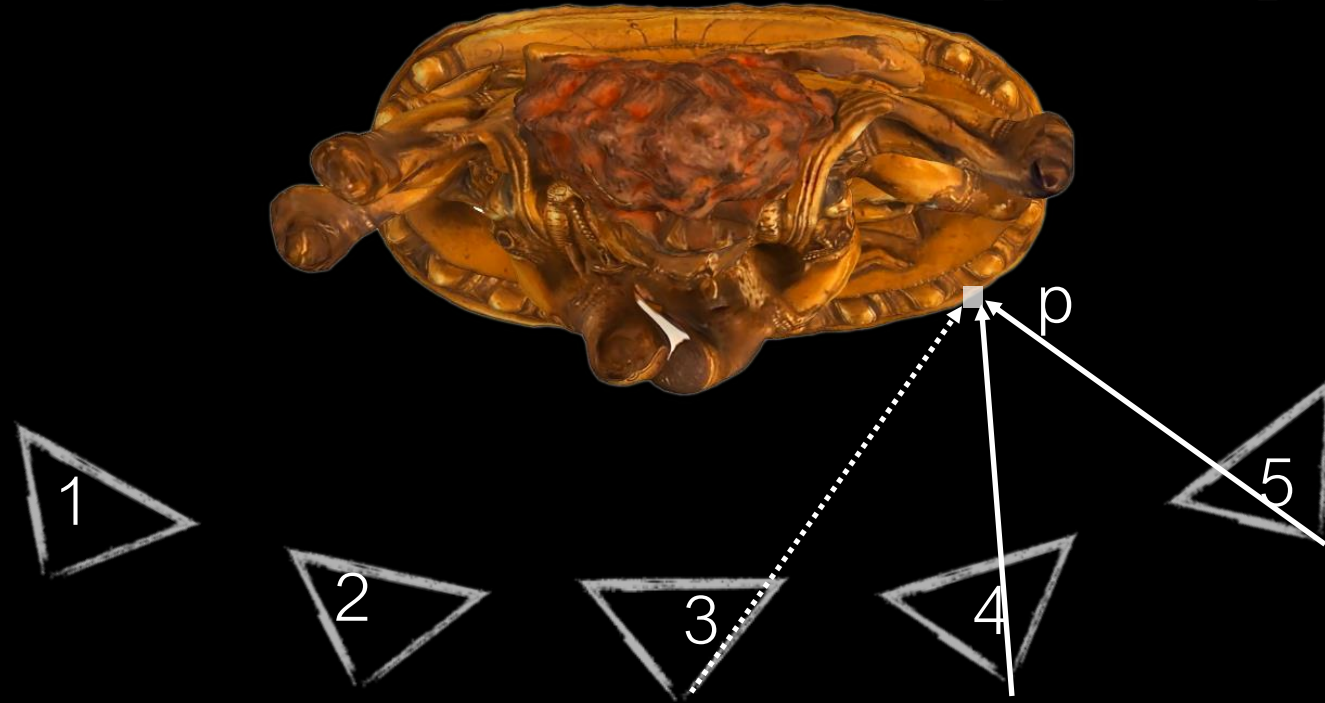
Multi-view stereo: Basic idea



Multi-view stereo: Basic idea

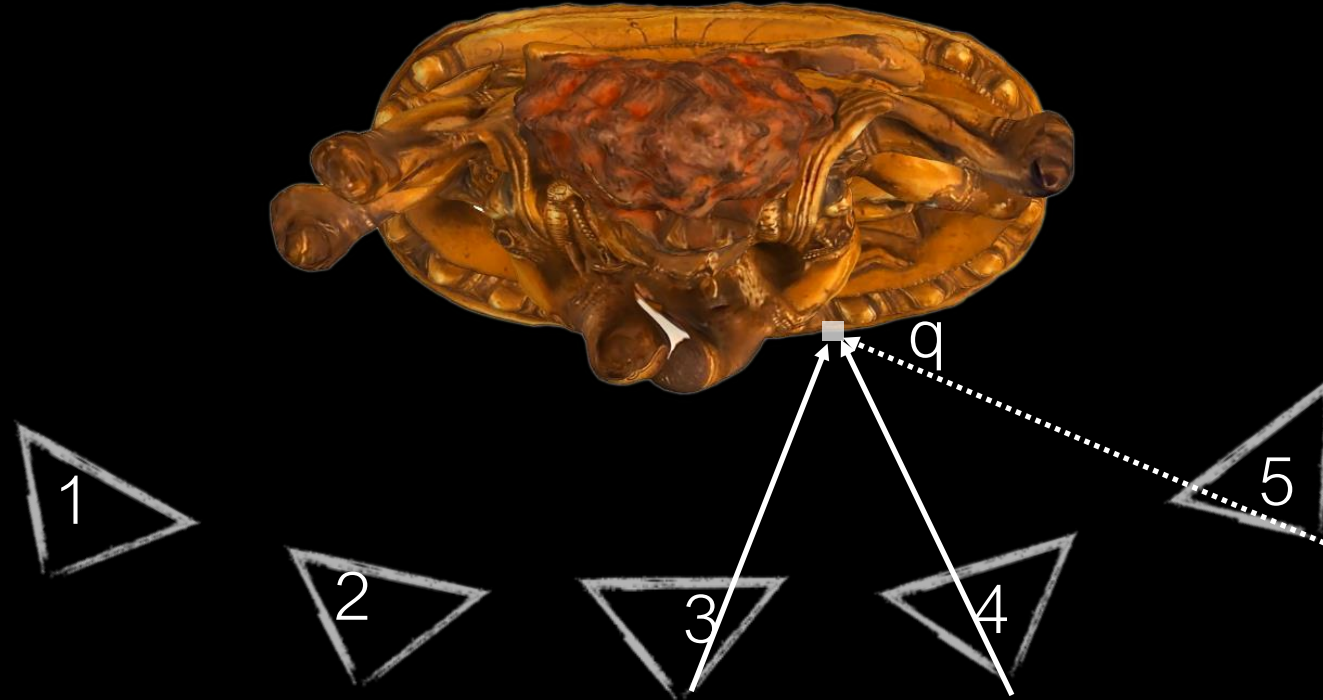


Some points occluded entirely in certain views



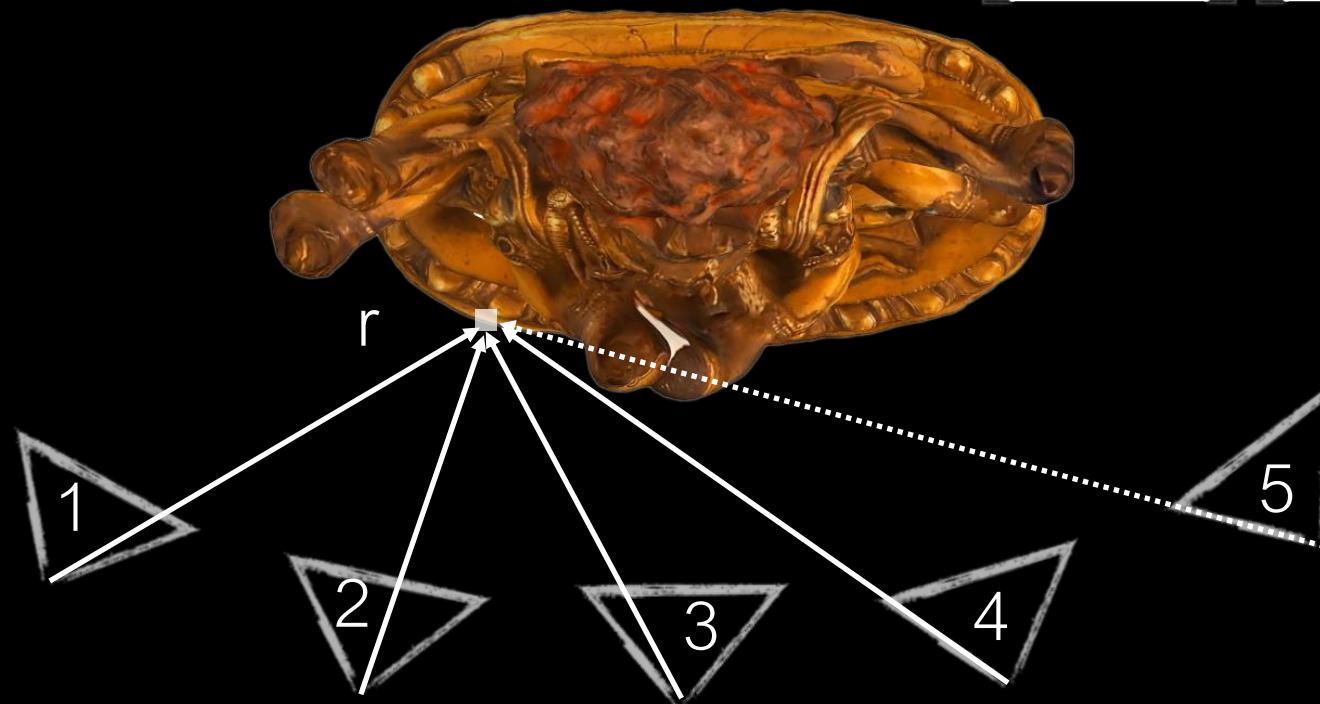
Cameras 4 and 5 can more clearly see point p

Some points occluded entirely in certain views



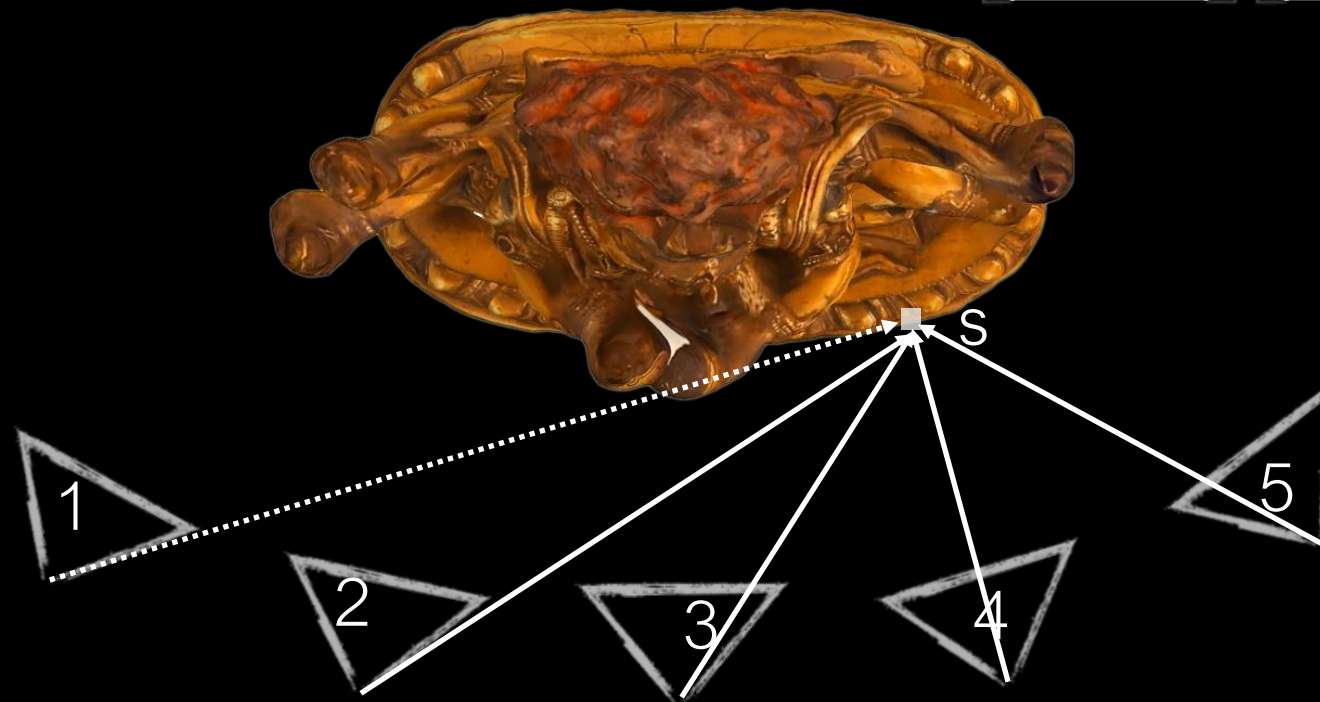
Cameras 3 and 4 can more clearly see point q

Some points occluded entirely in certain views

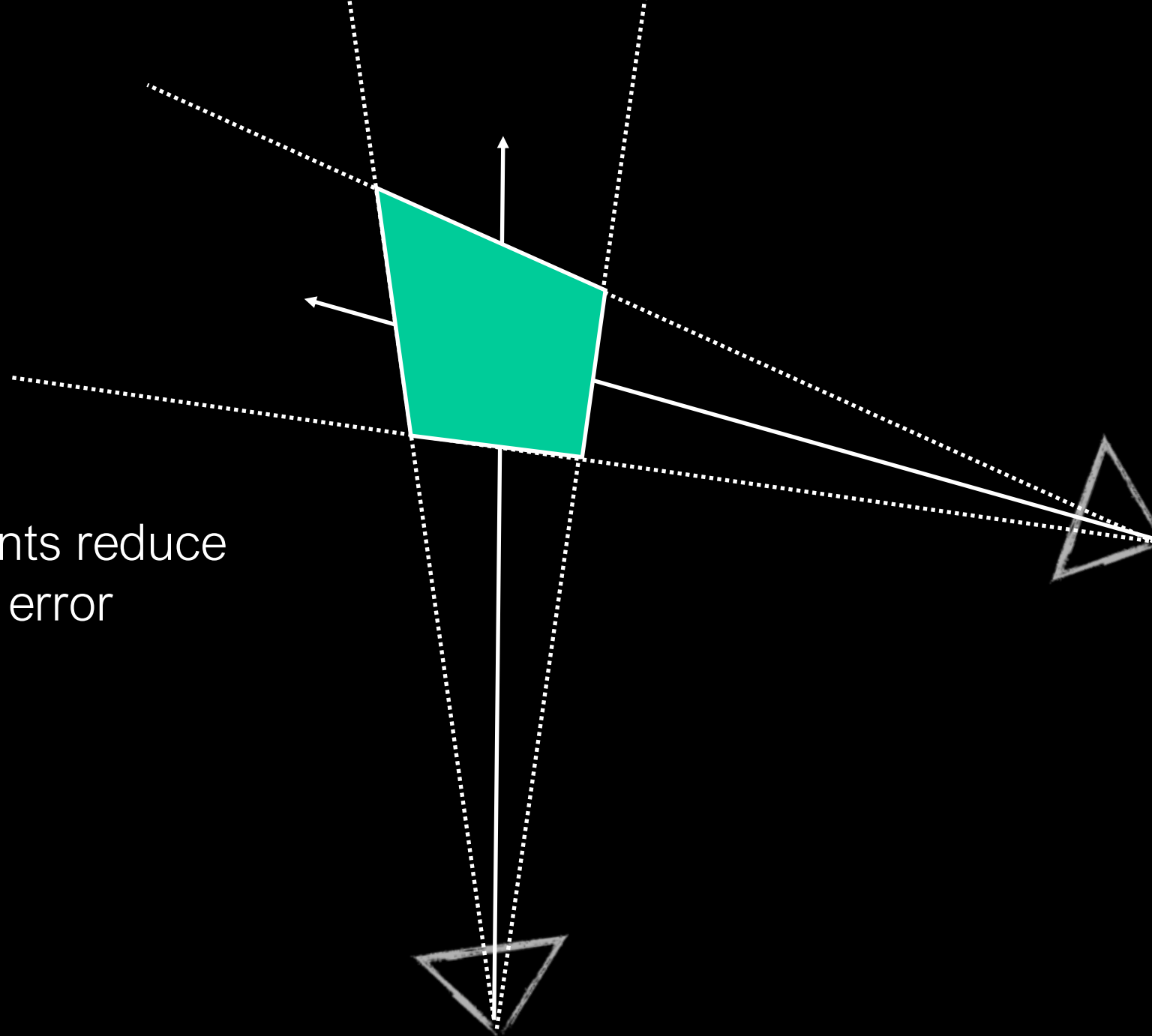


Camera 5 can't see point r

Some points occluded entirely in certain views

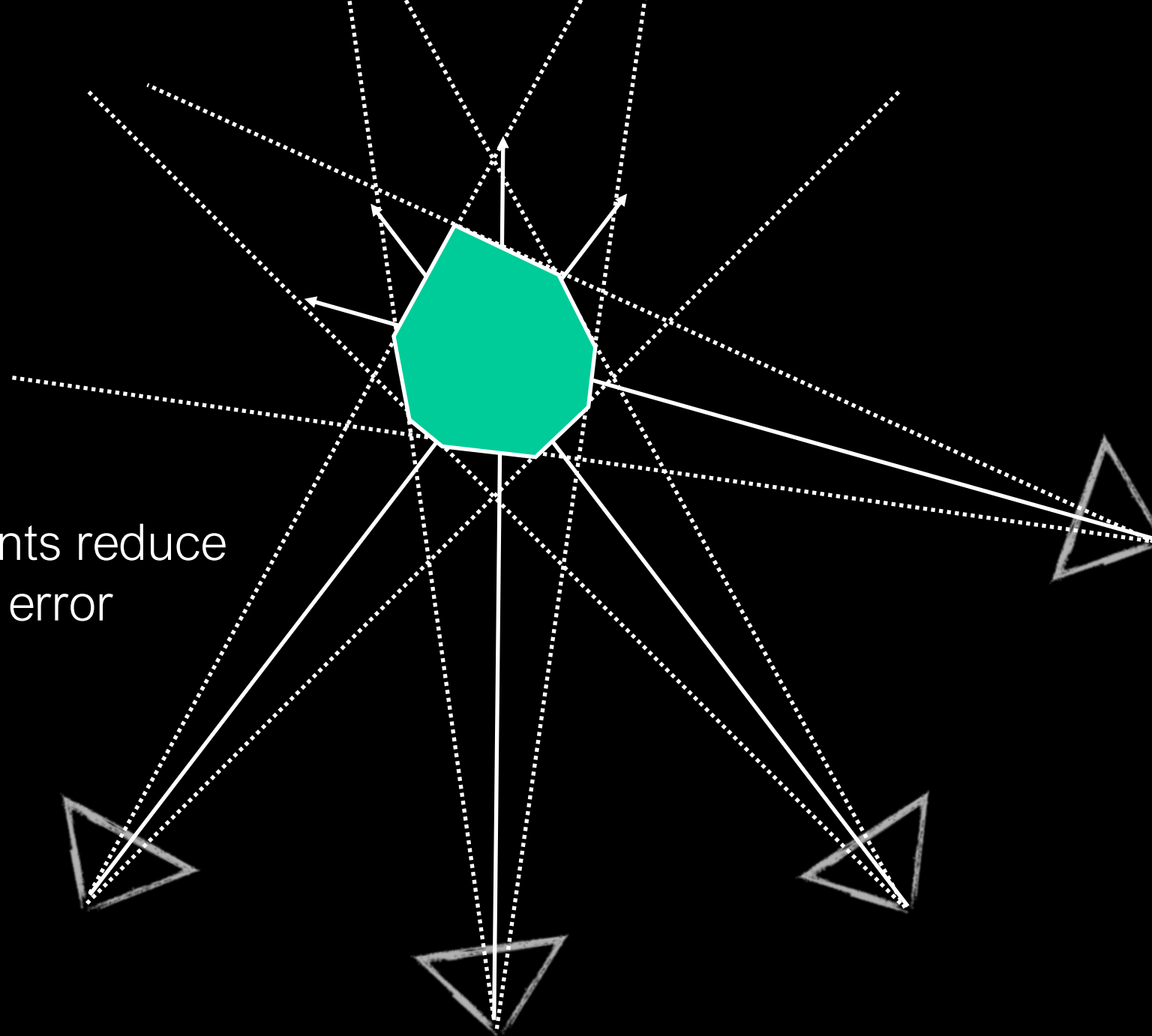


Camera 1 can't see point s



More measurements reduce
triangulation error

More measurements reduce
triangulation error



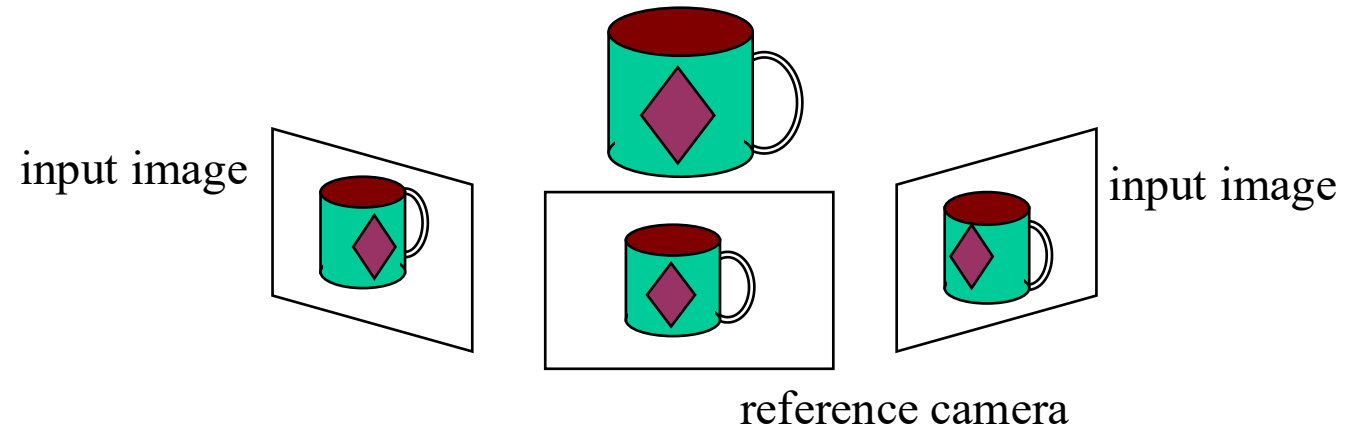
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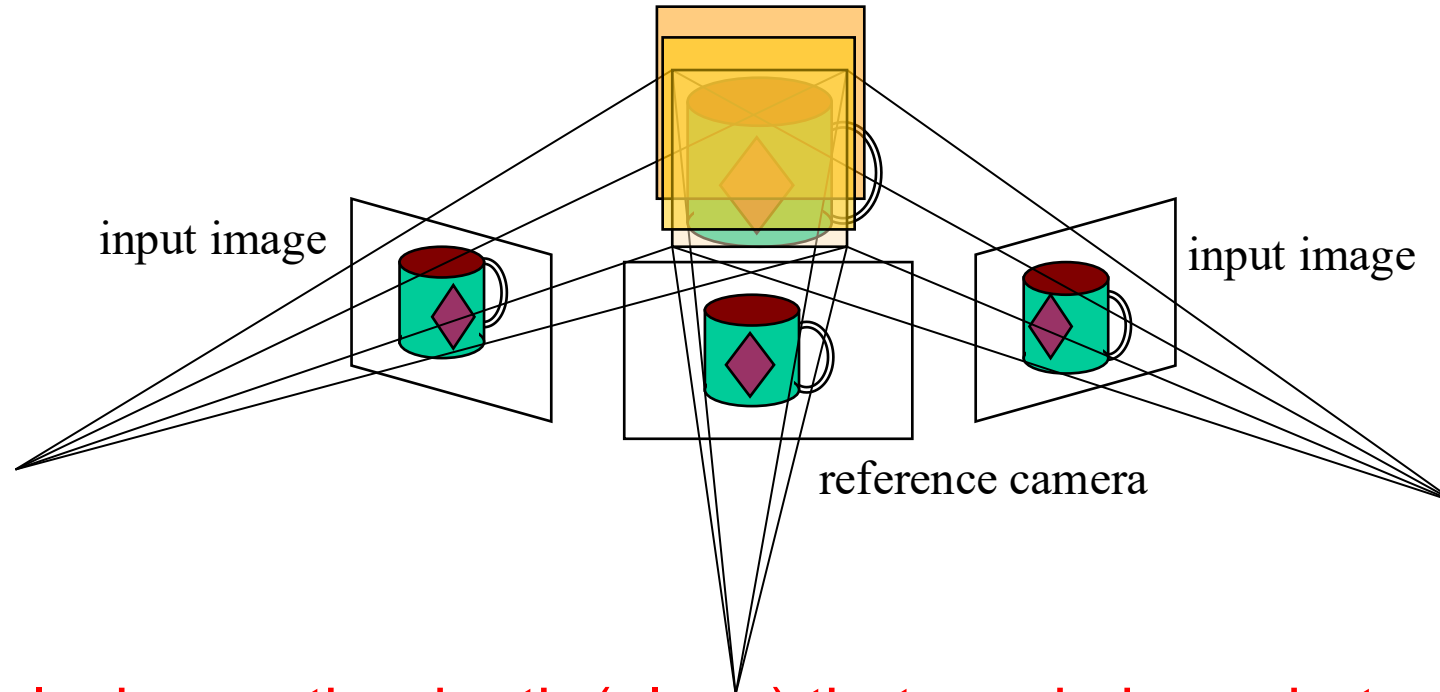
Plane sweep stereo

Goal: recover a **dense depth map** (per-pixel depth) rather than sparse feature points



Plane sweep stereo

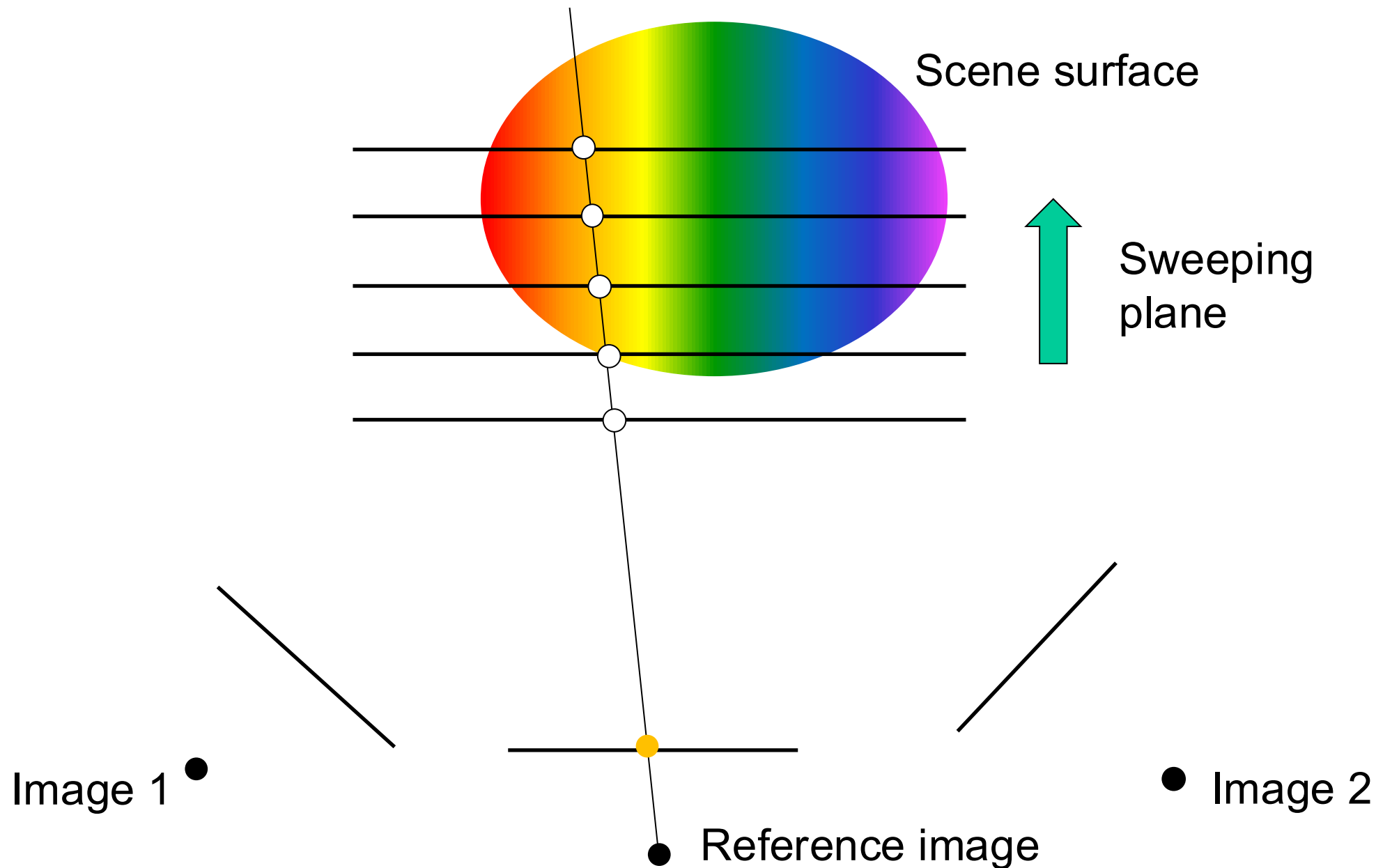
Idea: convert correspondence finding into structured search over depth hypotheses



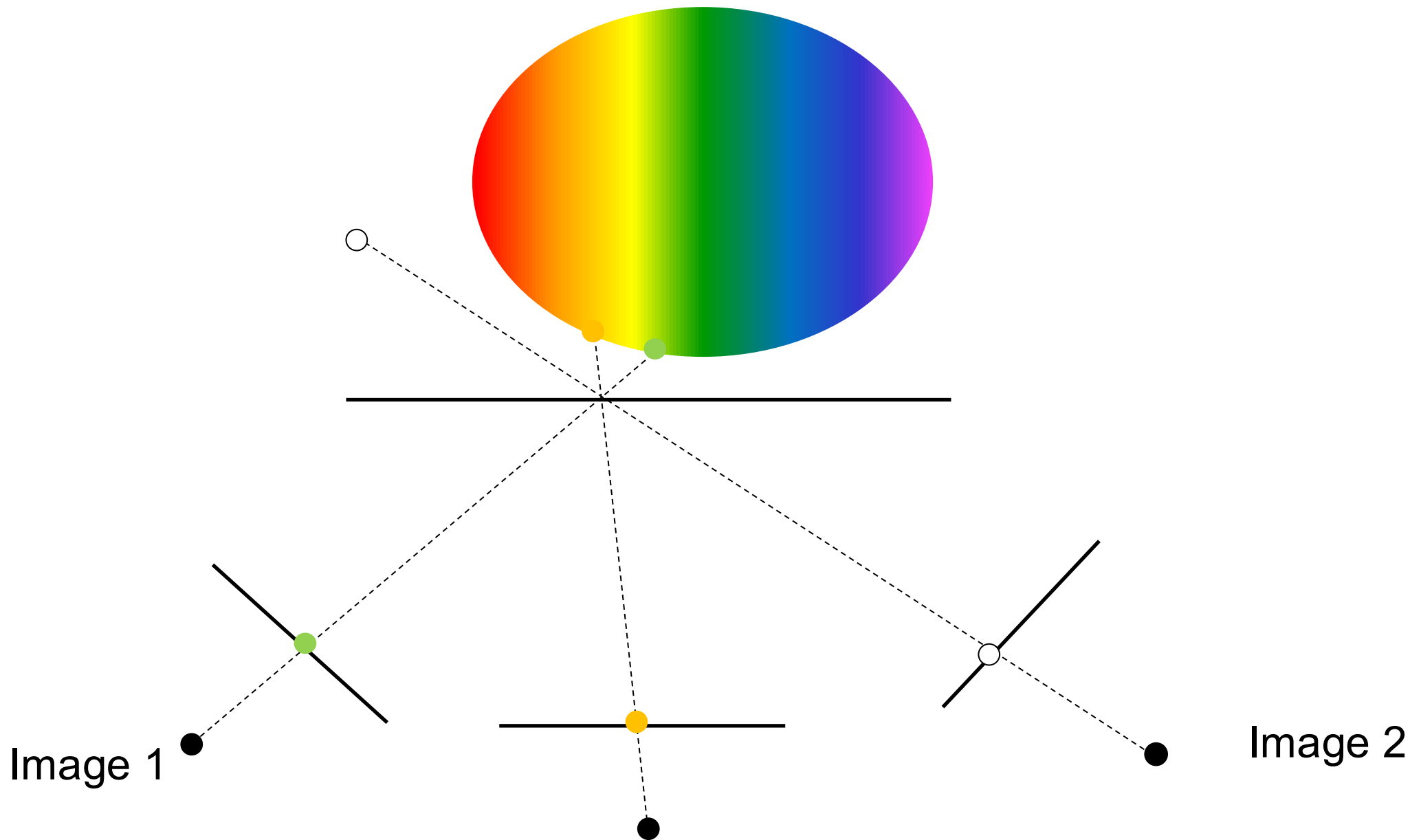
For each pixel, choose the depth (plane) that maximizes photo consistency

- 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 reprojections

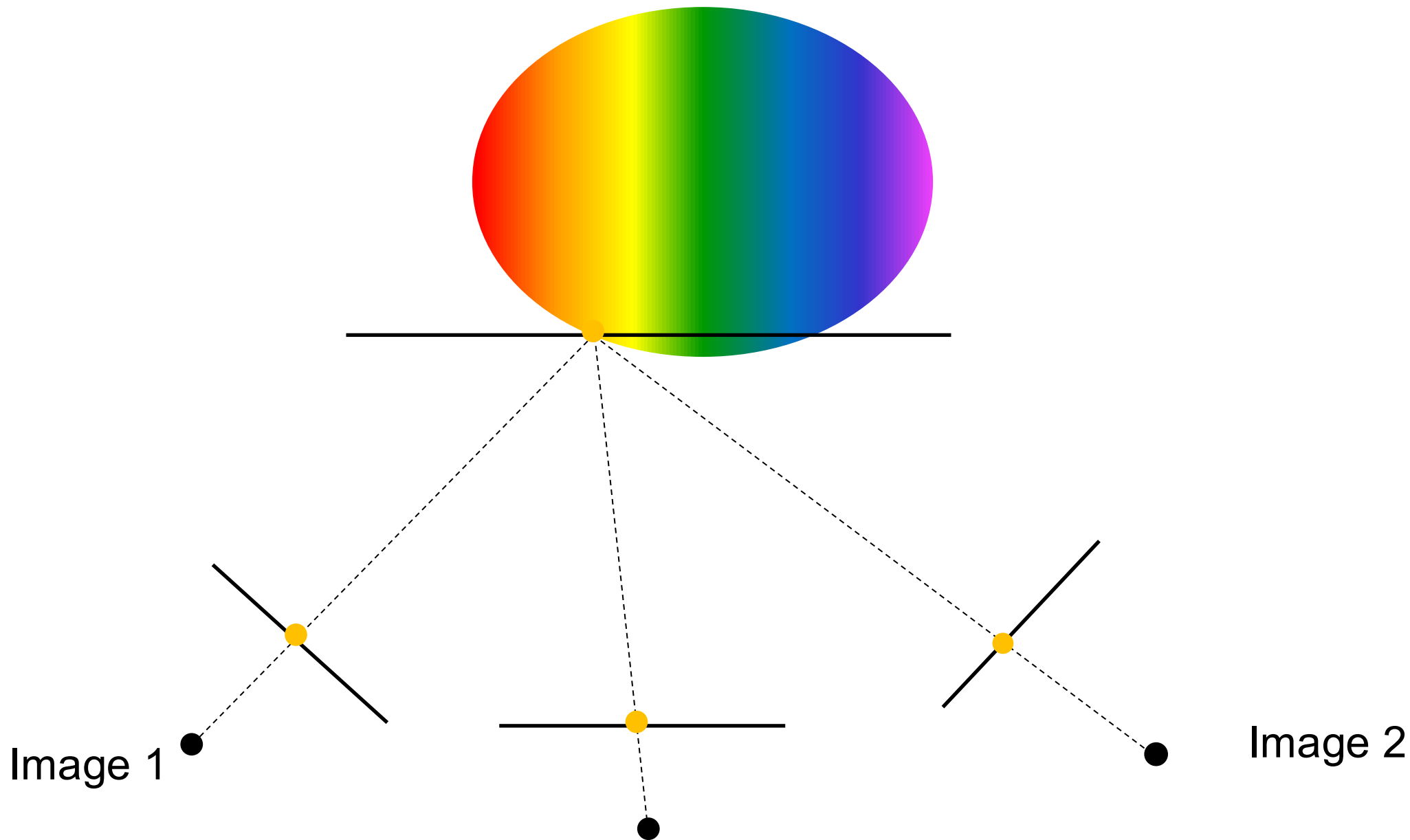
Plane sweep stereo: Key idea



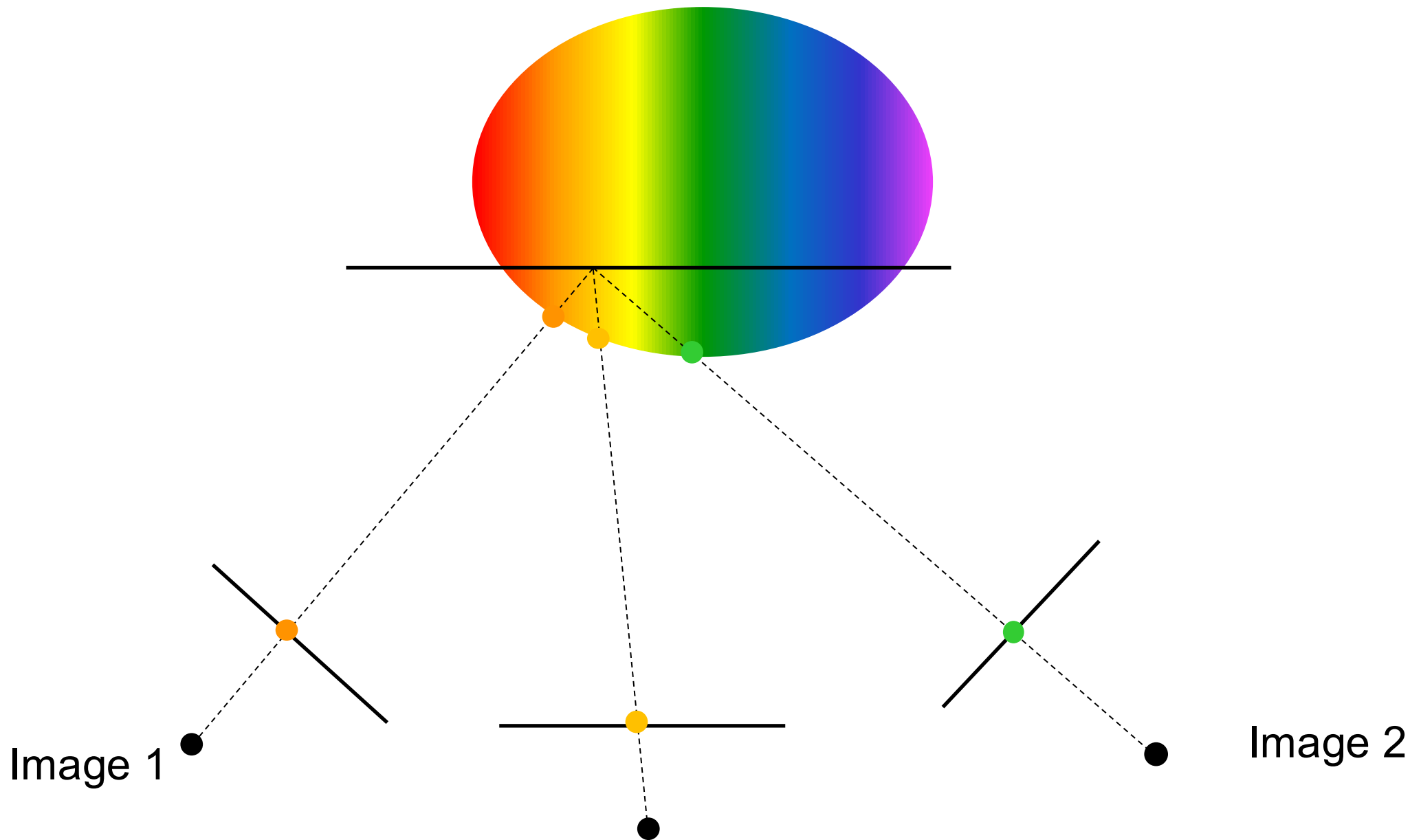
Plane sweep stereo: Key idea



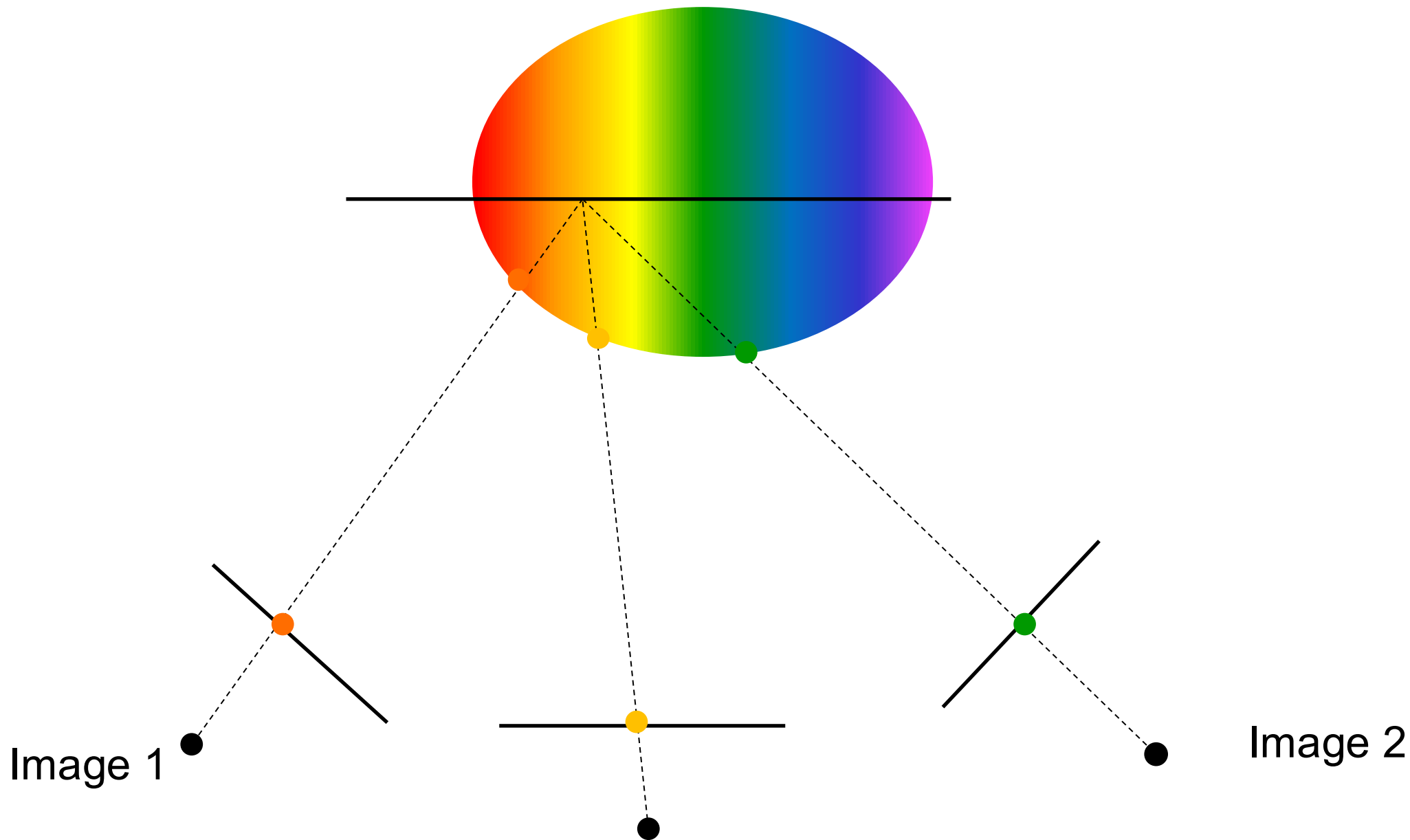
Plane sweep stereo: Key idea



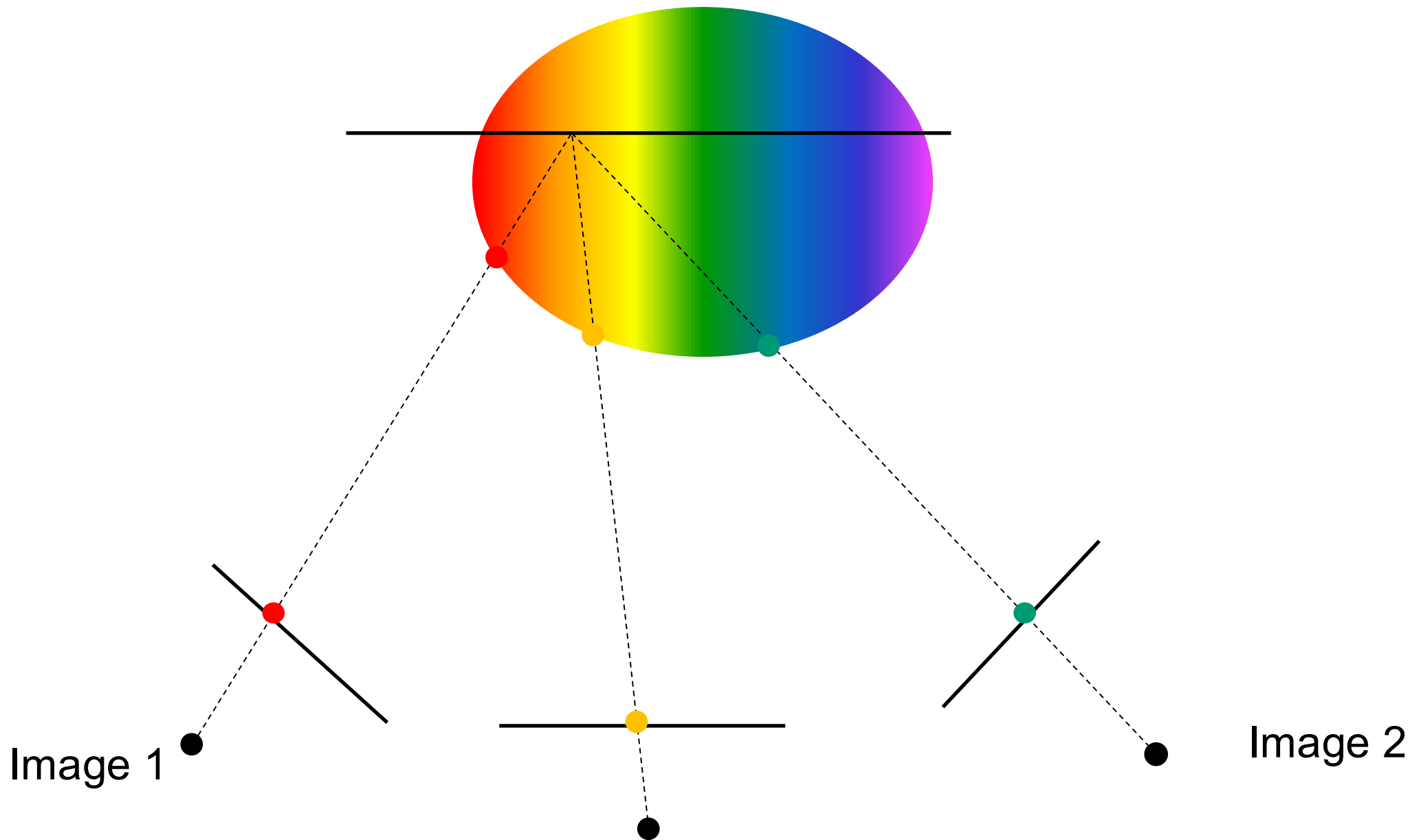
Plane sweep stereo: Key idea



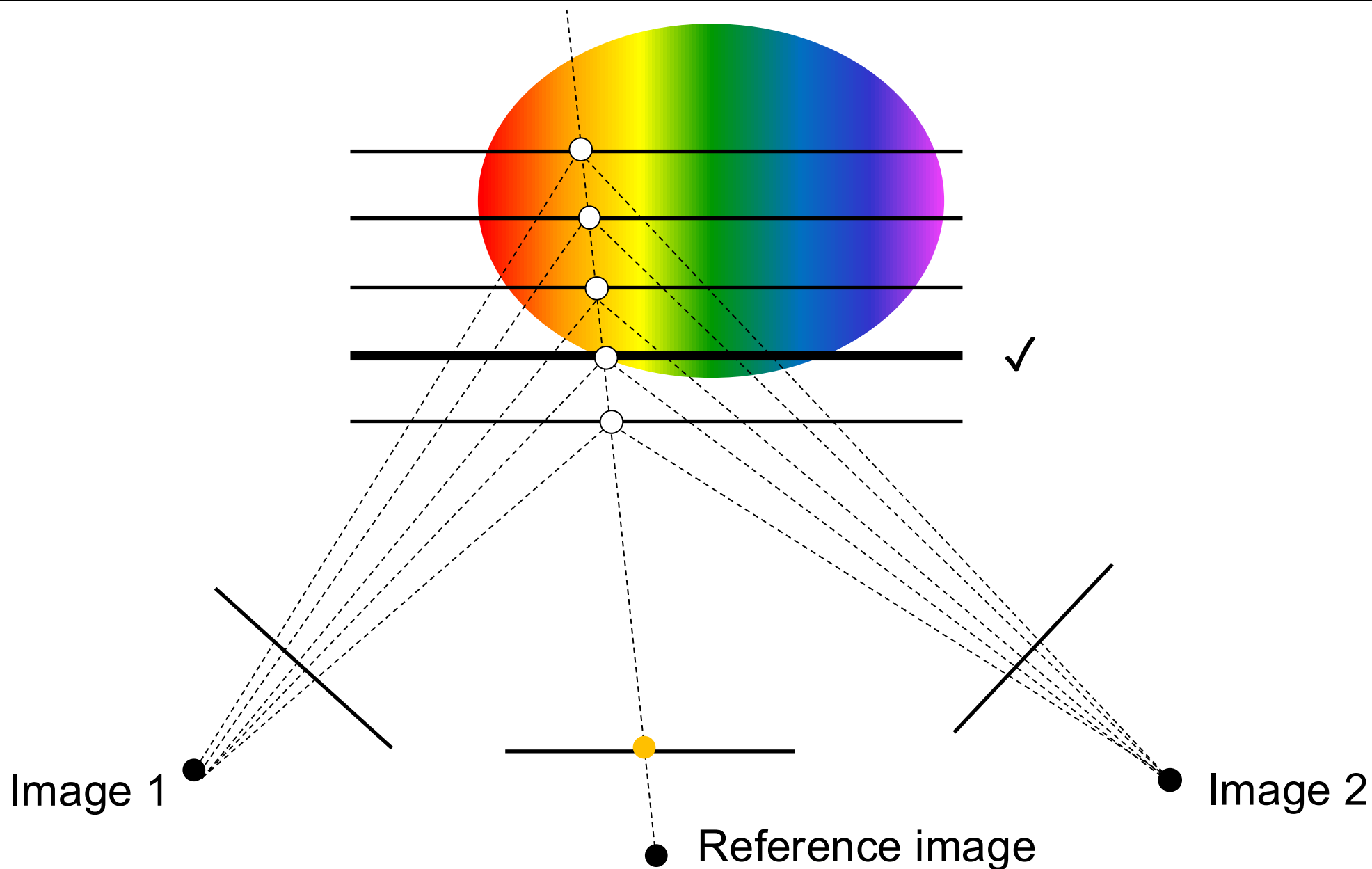
Plane sweep stereo: Key idea



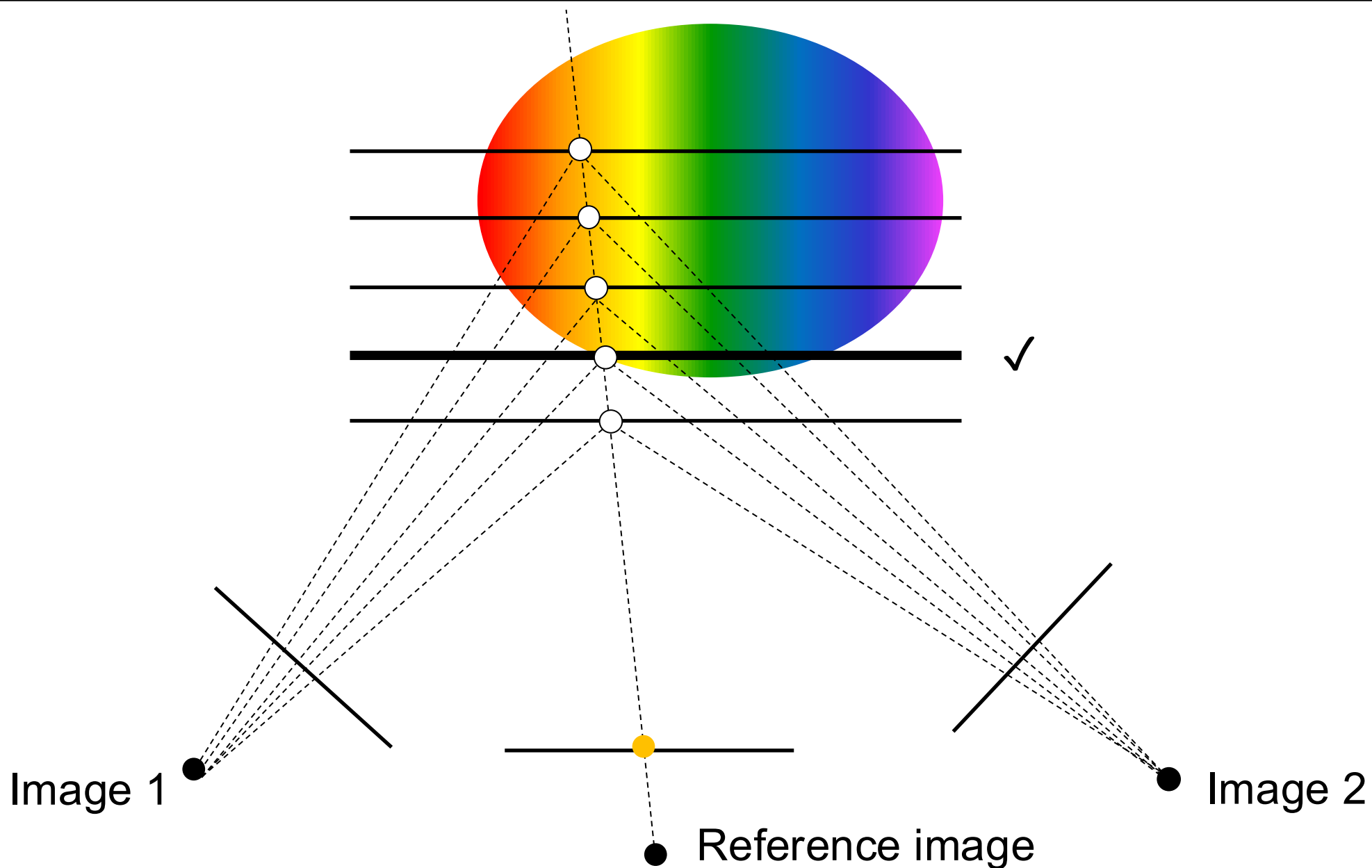
Plane sweep stereo: Key idea



Plane sweep stereo: Key idea



Does this always work?



Plane sweep stereo: Fast implementation



depth 1
depth 2
depth 3
depth 4
depth 5
depth 6

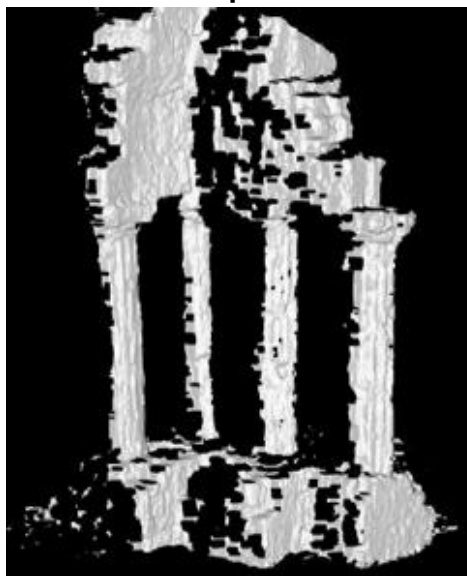
- For each depth plane
 - Compute homographies projecting each image onto that depth plane
 - For each pixel in the composite image stack, compute the variance
- For each pixel, select the depth that gives the lowest variance

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](#))

Map 1



Map 2



Merged



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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](#)

Demonstration



Y. Furukawa and J. Ponce, [Accurate, Dense, and Robust Multi-View Stereopsis](#), CVPR 2007.
[PMVS software](#)

Towards Internet-scale multi-view stereo



[YouTube video](#), [CMVS software](#)

Mesh model reconstructed from point cloud



[YouTube video](#)

More technical details: <https://www.youtube.com/watch?v=OpZs7kfjFPA>

Next lecture: Surface Reconstruction

