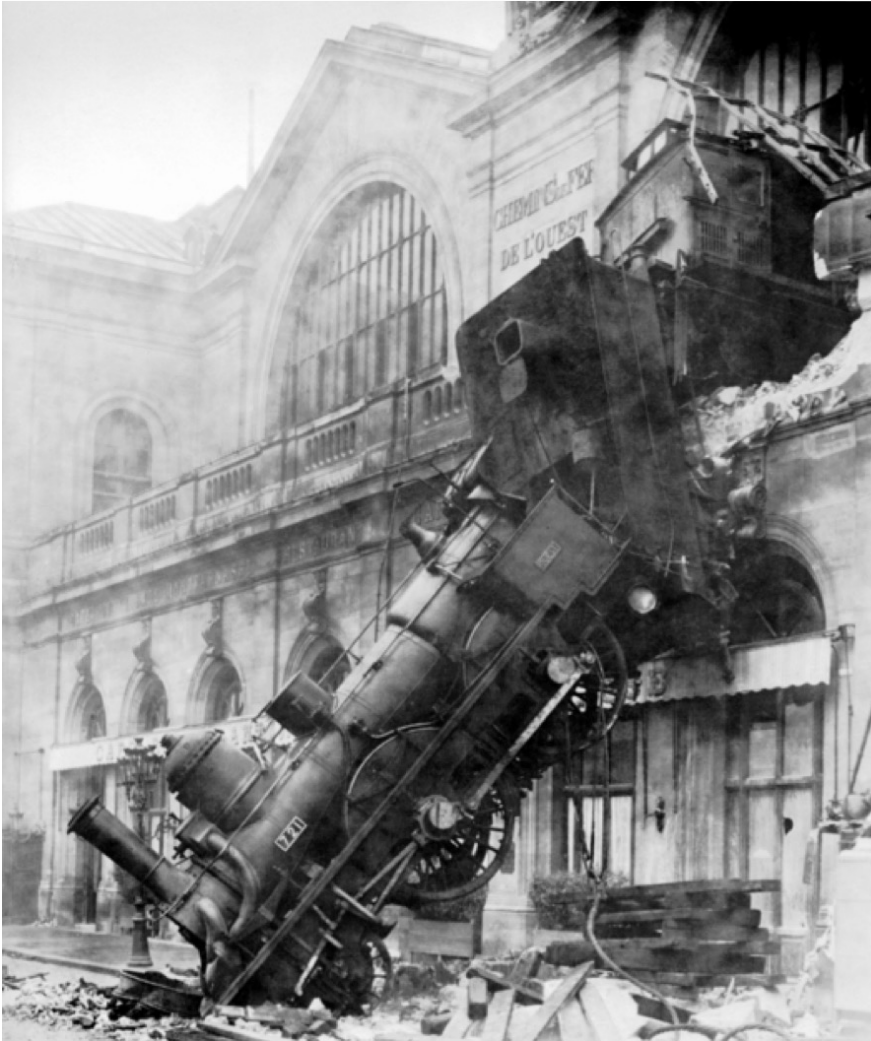


Review - Computer Vision

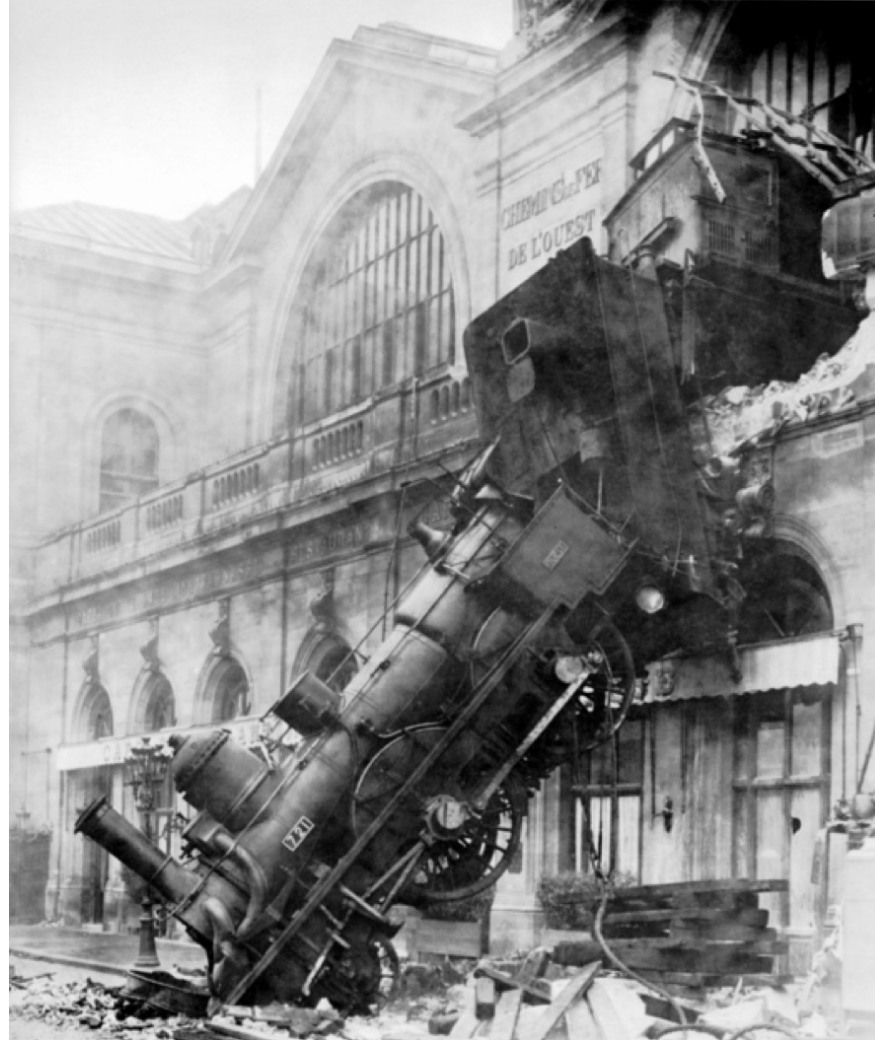
Saurabh Gupta

The goal(s) or computer vision

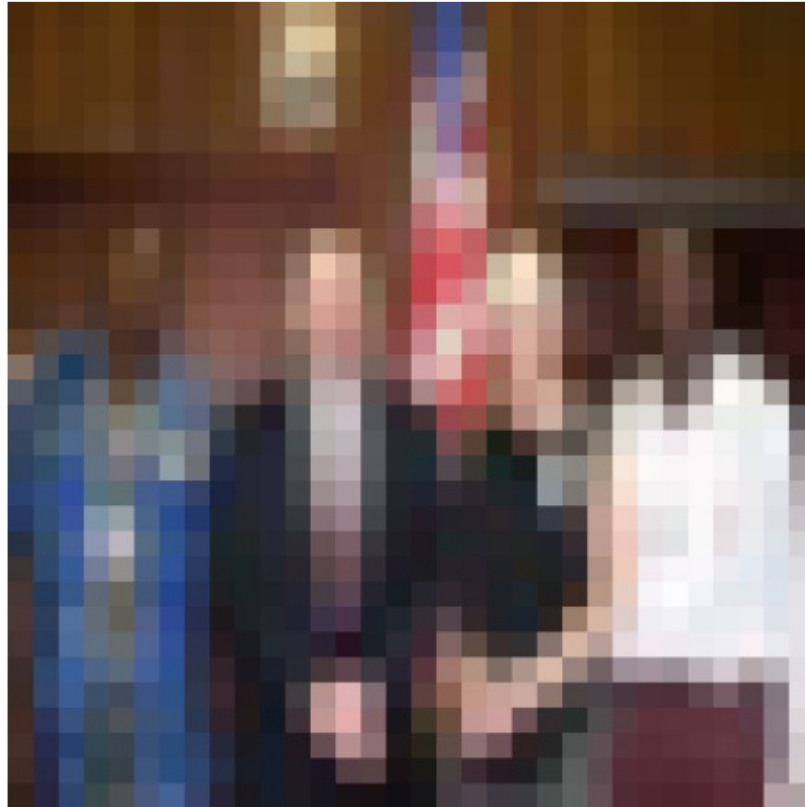


- What is the image about?
- What objects are in the image?
- Where are they?
- How are they oriented?
- What is the layout of the scene in 3D?
- What is the shape of each object?

Vision is easy for humans



Vision is easy for humans



Vision is easy for humans

Attneave's Cat

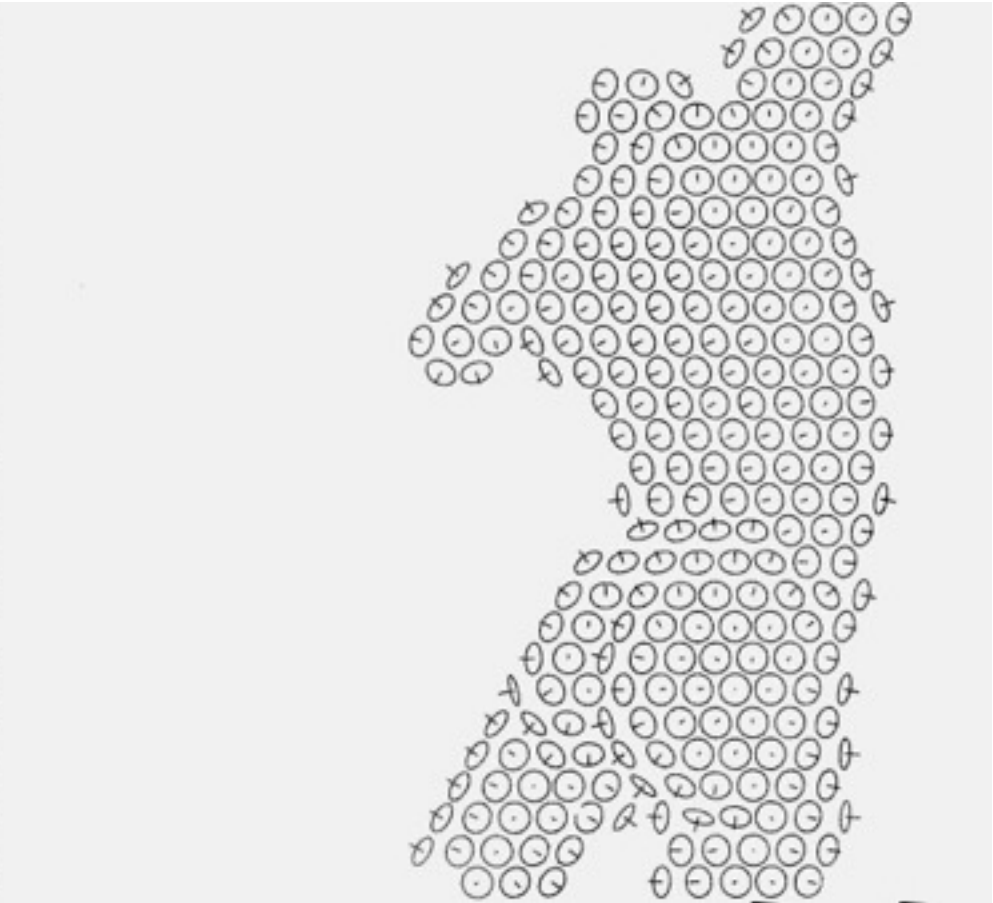
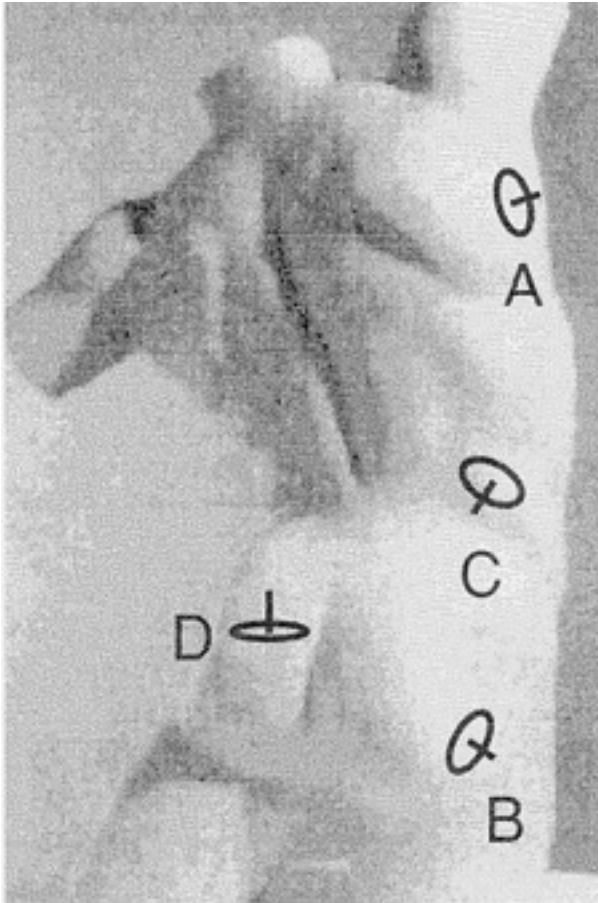


Vision is easy for humans

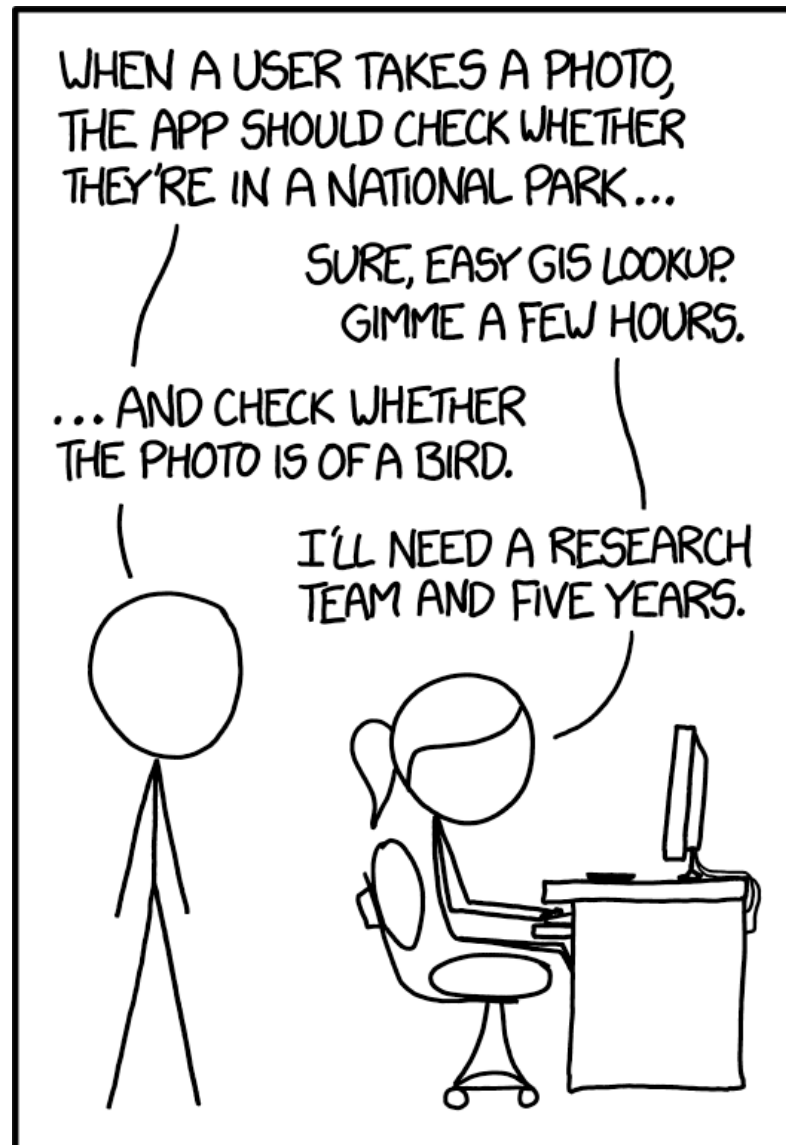
Mooney Faces



Vision is easy for humans



Remarkably Hard for Computers



Vision is hard: Objects Blend Together



Vision is hard: Objects Blend Together



Vision is hard: Intra-class Variation



Viewpoint variation



Illumination



Scale

Vision is hard: Intra-class Variation



Shape variation



Occlusion



Background clutter

Source: B. Hariharan

Vision is hard: Intra-class Variation



Vision is hard: Concepts are subtle



Tennessee Warbler



Orange Crowned Warbler

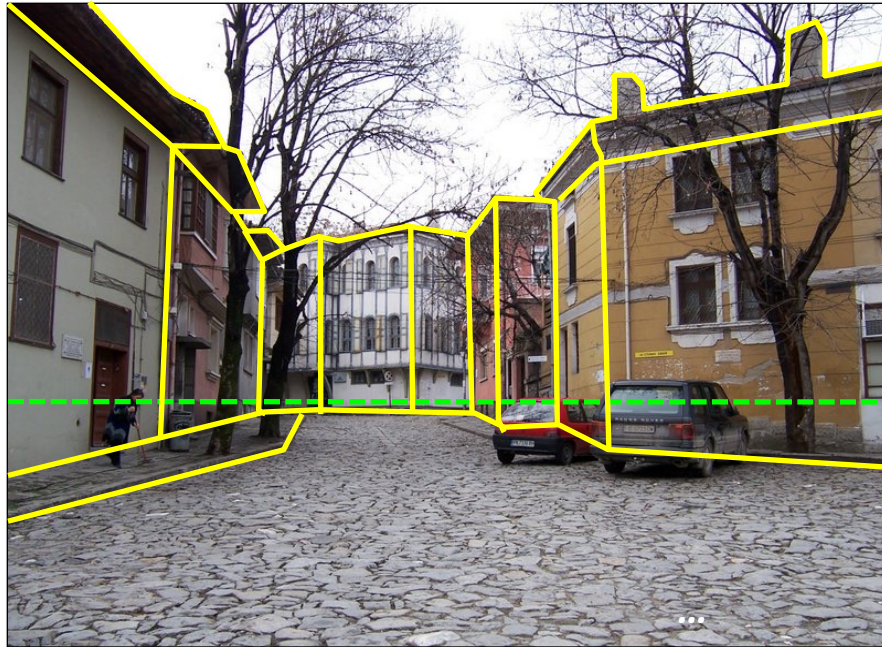
Vision is hard: Images are ambiguous



What kind of information can be extracted from an image?

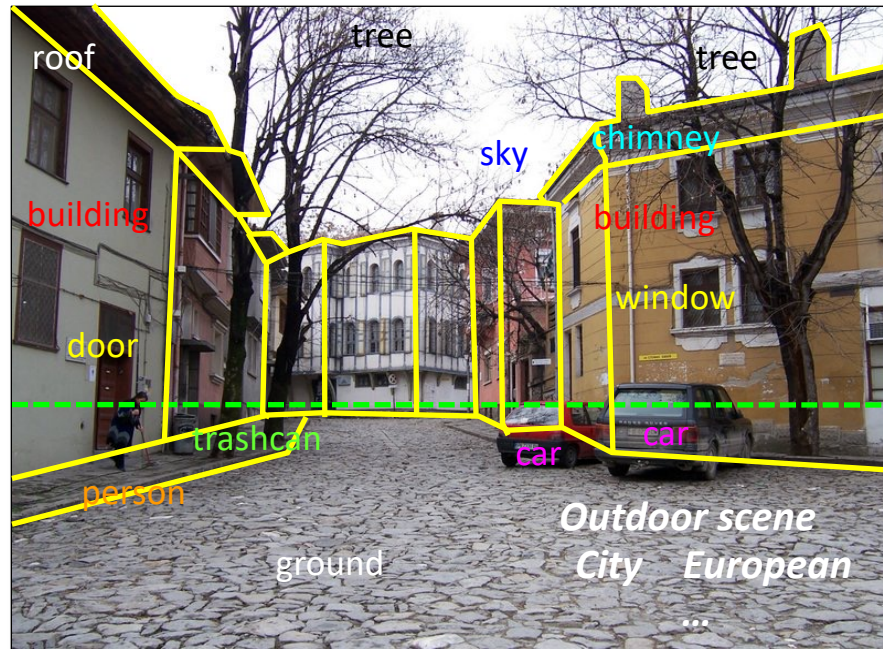


What kind of information can be extracted from an image?



Geometric information

What kind of information can be extracted from an image?

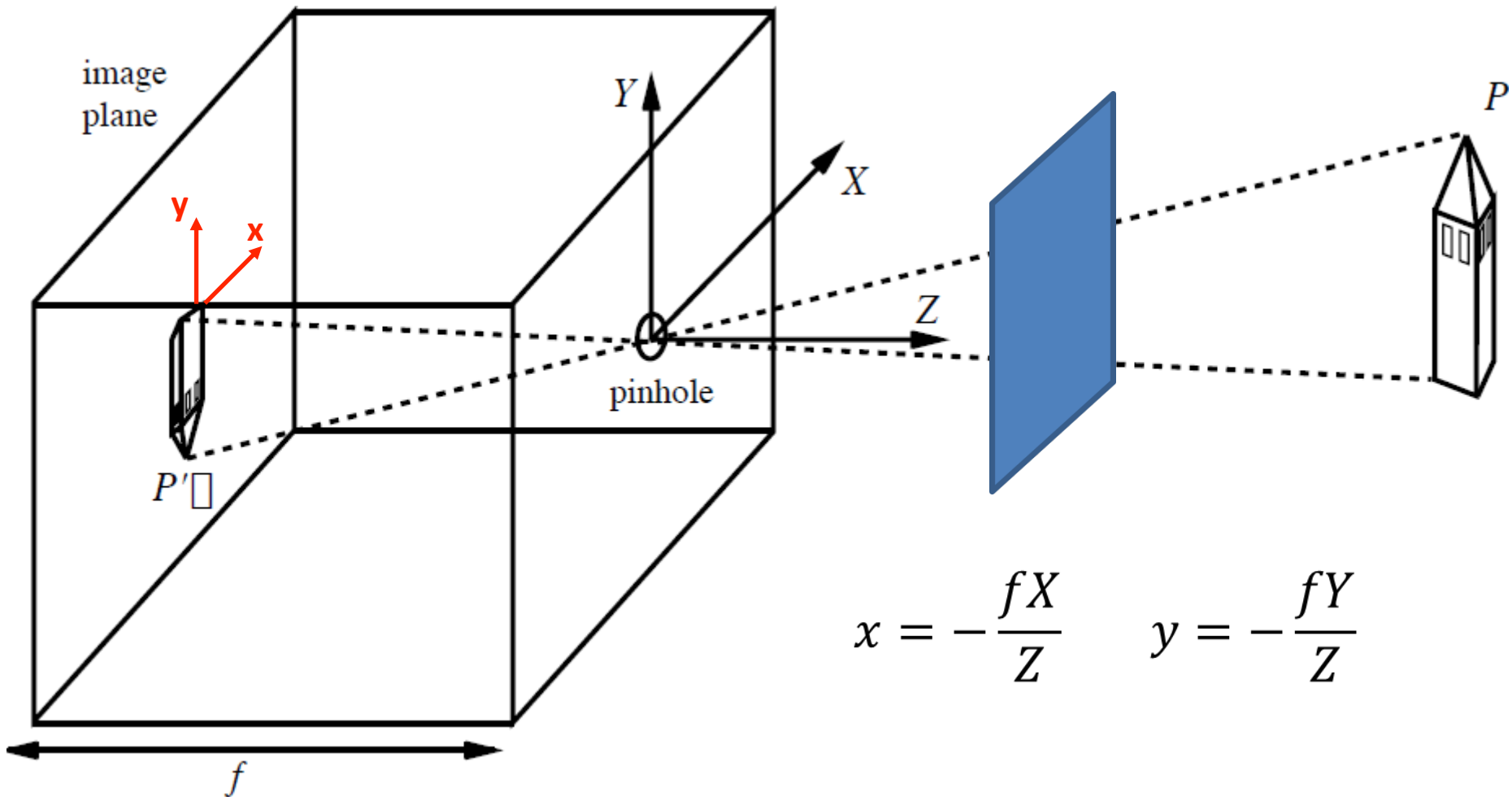


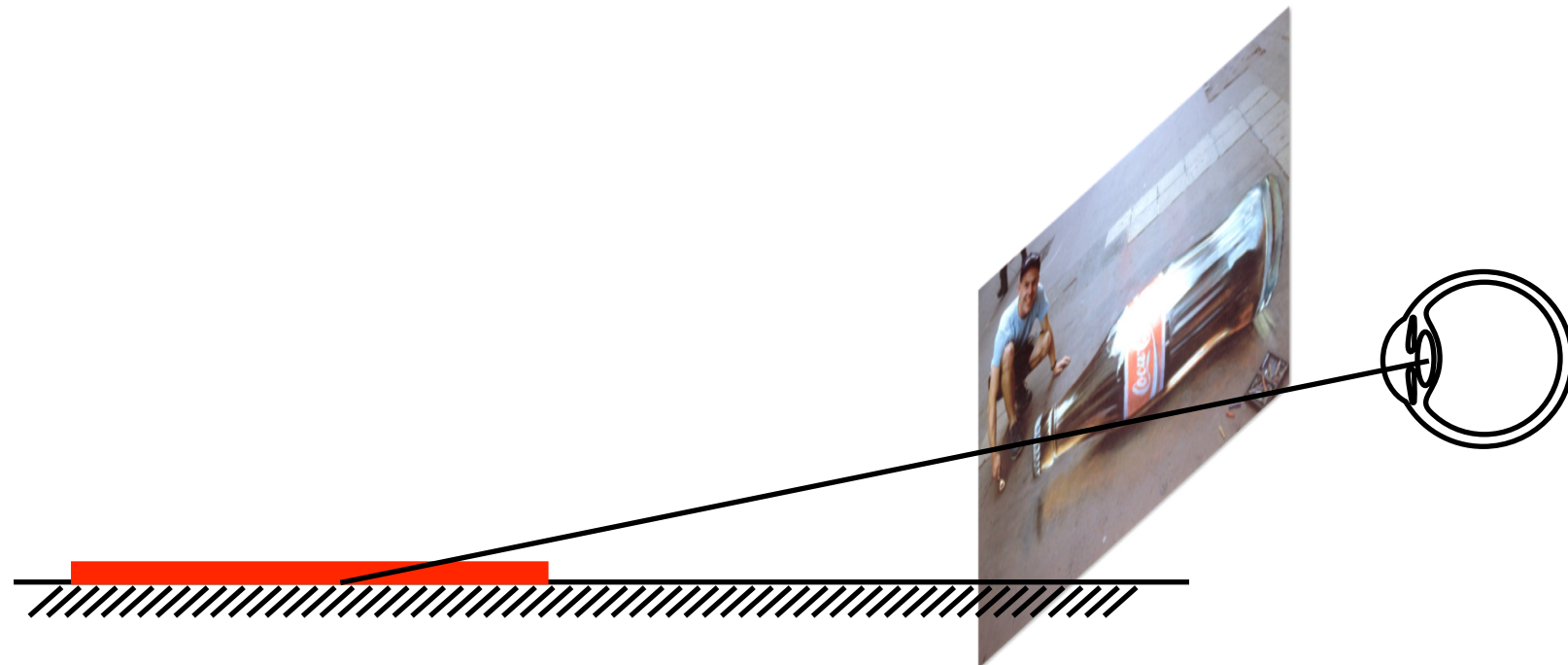
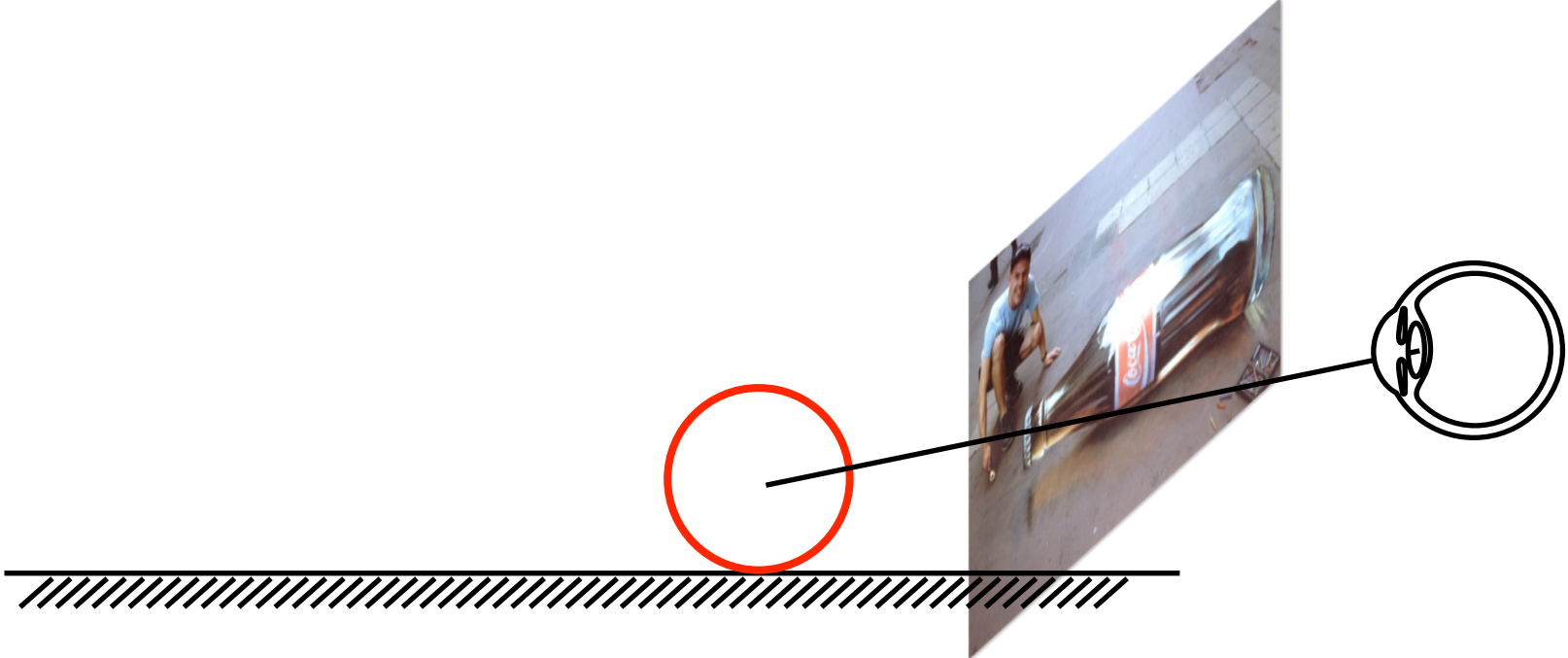
Geometric information
Semantic information

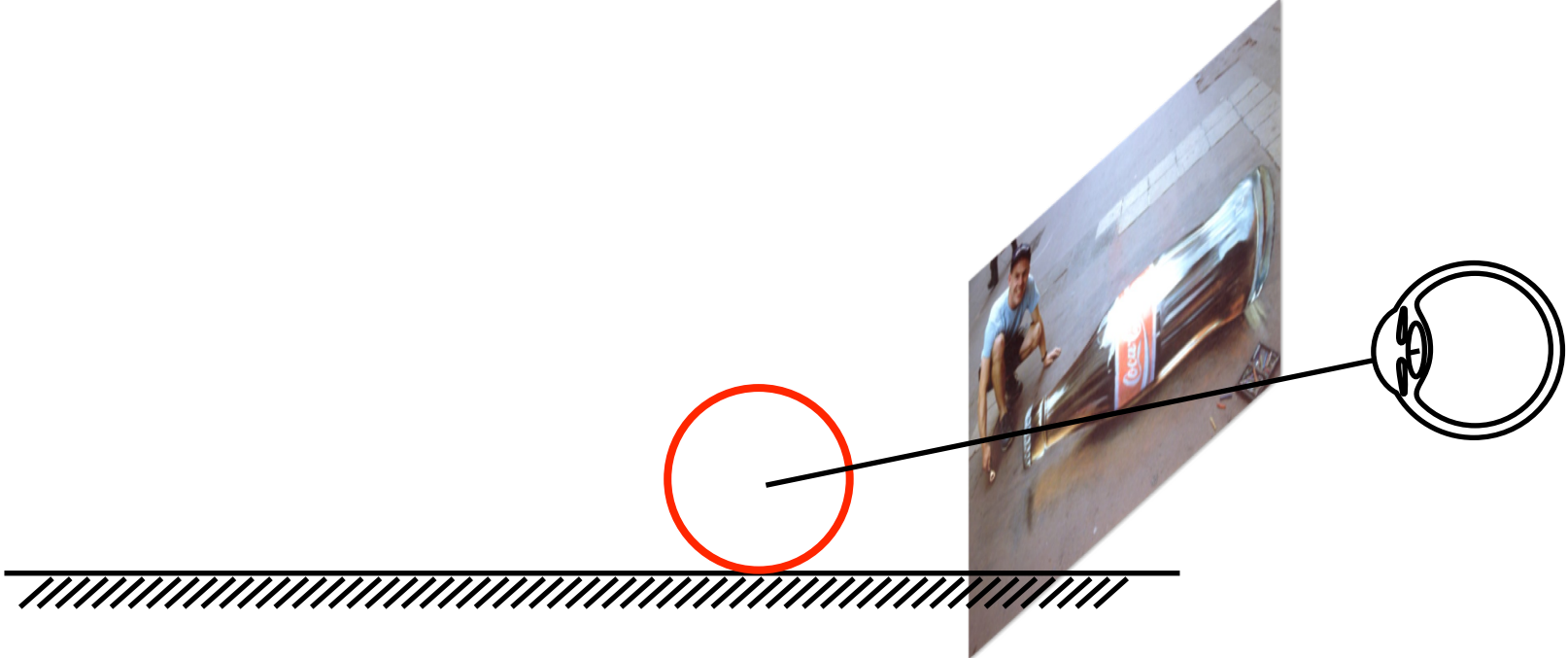
Vision is hard: Images are ambiguous



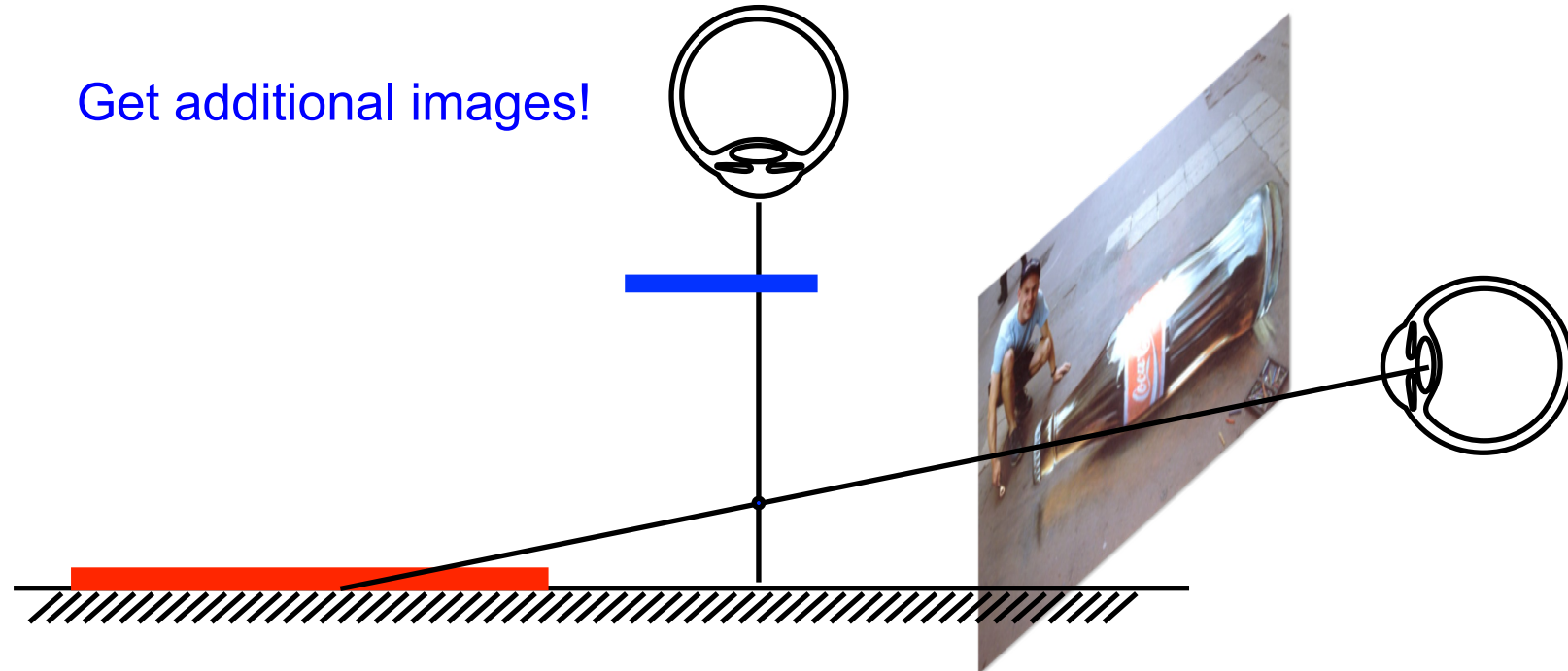
The Pinhole Camera



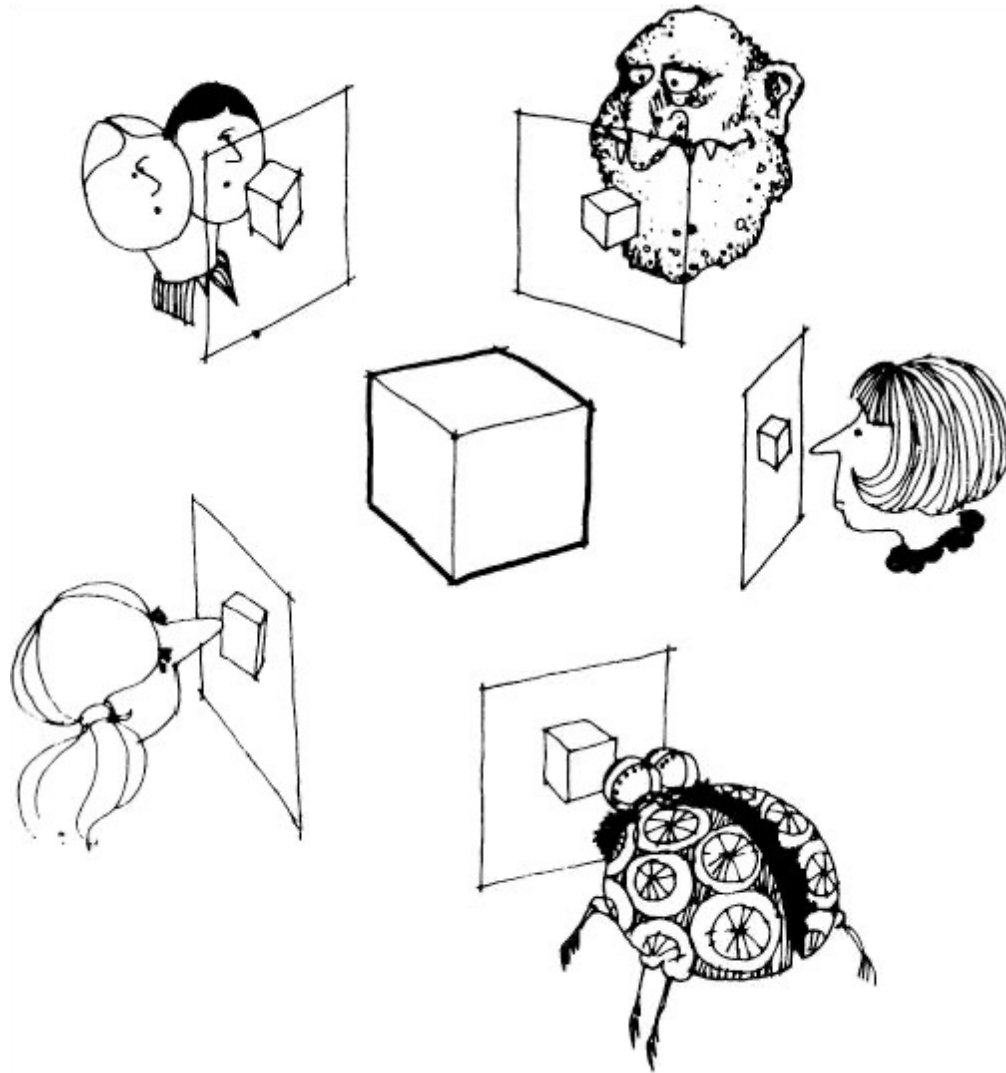




Get additional images!



Structure from Motion



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

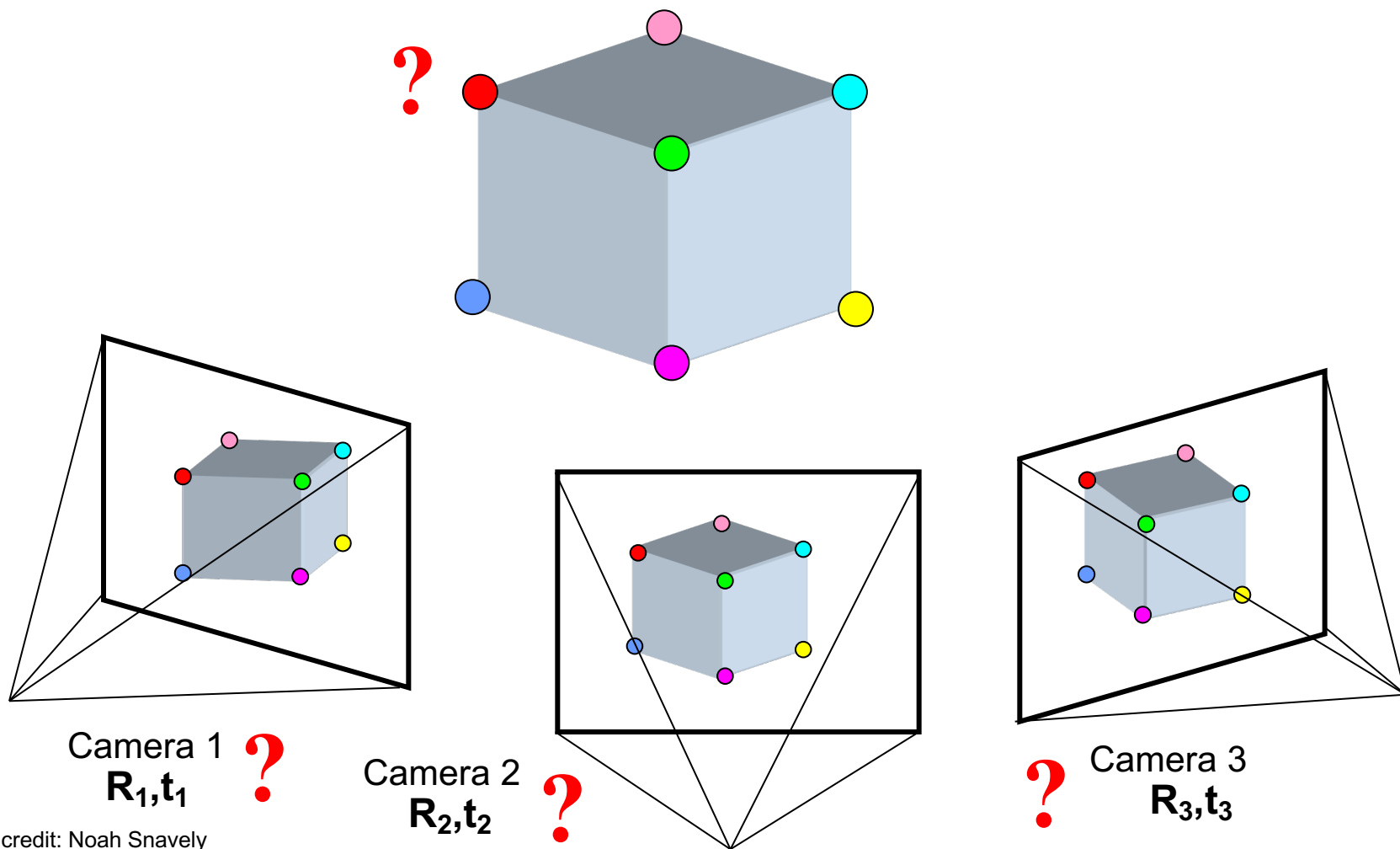
Structure from motion

- Generic problem formulation: 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 sequence)
 - Camera parameters may be known or unknown



Structure from motion

- Given a set of corresponding points in two or more images, compute the camera parameters and the 3D point coordinates

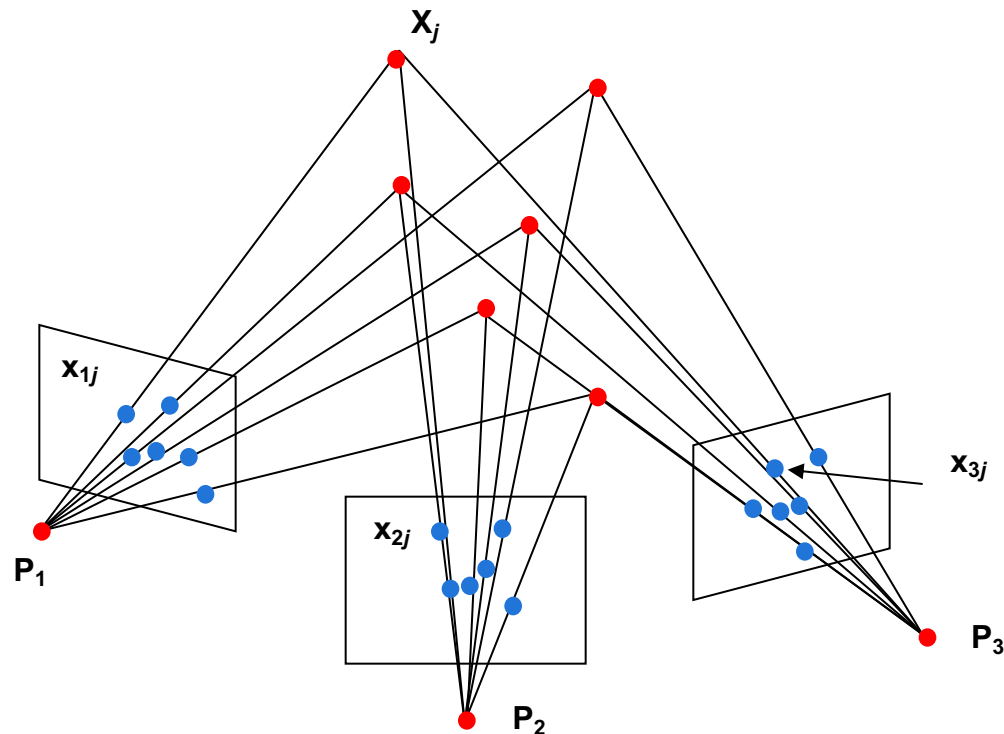


Structure from motion

- Given: m images of n fixed 3D points

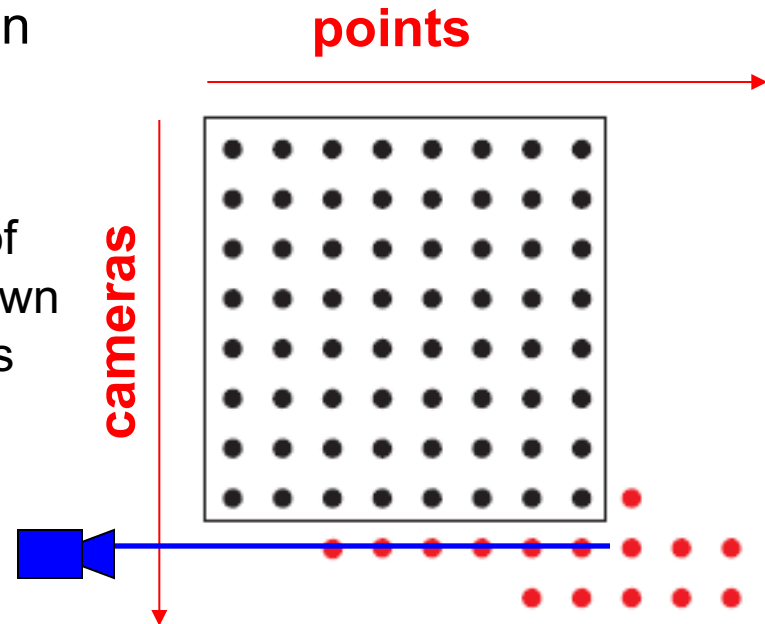
$$\lambda_{ij} \mathbf{x}_{ij} = \mathbf{P}_i \mathbf{X}_j, \quad i = 1, \dots, m, \quad j = 1, \dots, n$$

- Problem: estimate m projection matrices \mathbf{P}_i and n 3D points \mathbf{X}_j from the mn correspondences \mathbf{x}_{ij}



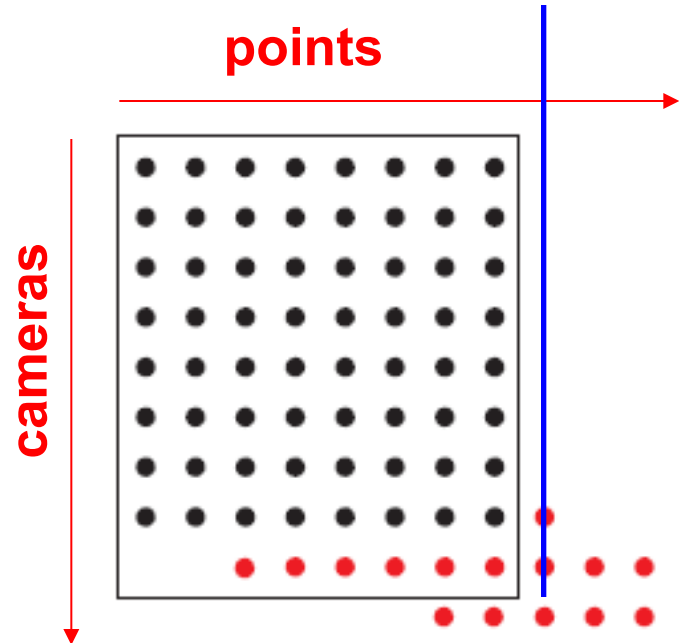
Incremental structure from motion

- Initialize motion from two images using fundamental matrix
- Initialize structure by triangulation
- For each additional view:
 - Determine projection matrix of new camera using all the known 3D points that are visible in its image – *calibration*



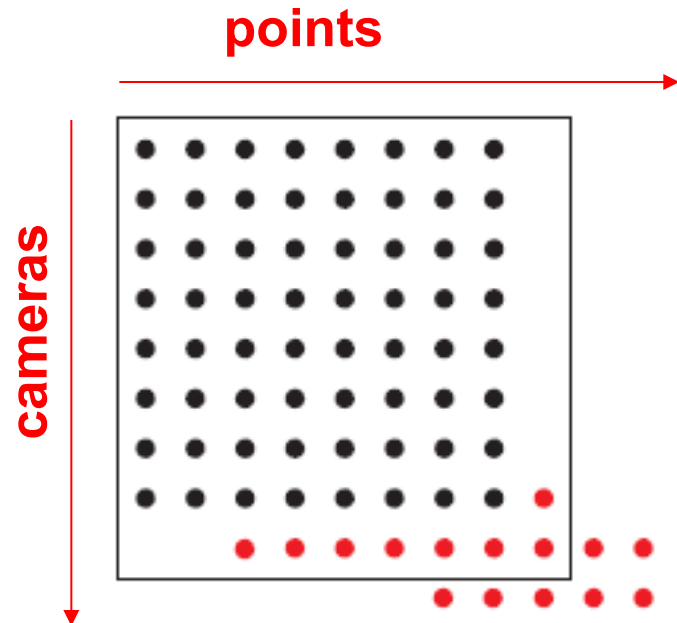
Incremental structure from motion

- Initialize motion from two images using fundamental matrix
- Initialize structure by triangulation
- For each additional view:
 - Determine projection matrix of new camera using all the known 3D points that are visible in its image – *calibration*
 - Refine and extend structure: compute new 3D points, re-optimize existing points that are also seen by this camera – *triangulation*



Incremental structure from motion

- Initialize motion from two images using fundamental matrix
- Initialize structure by triangulation
- For each additional view:
 - Determine projection matrix of new camera using all the known 3D points that are visible in its image – *calibration*
 - Refine and extend structure: compute new 3D points, re-optimize existing points that are also seen by this camera – *triangulation*
- Refine structure and motion: bundle adjustment

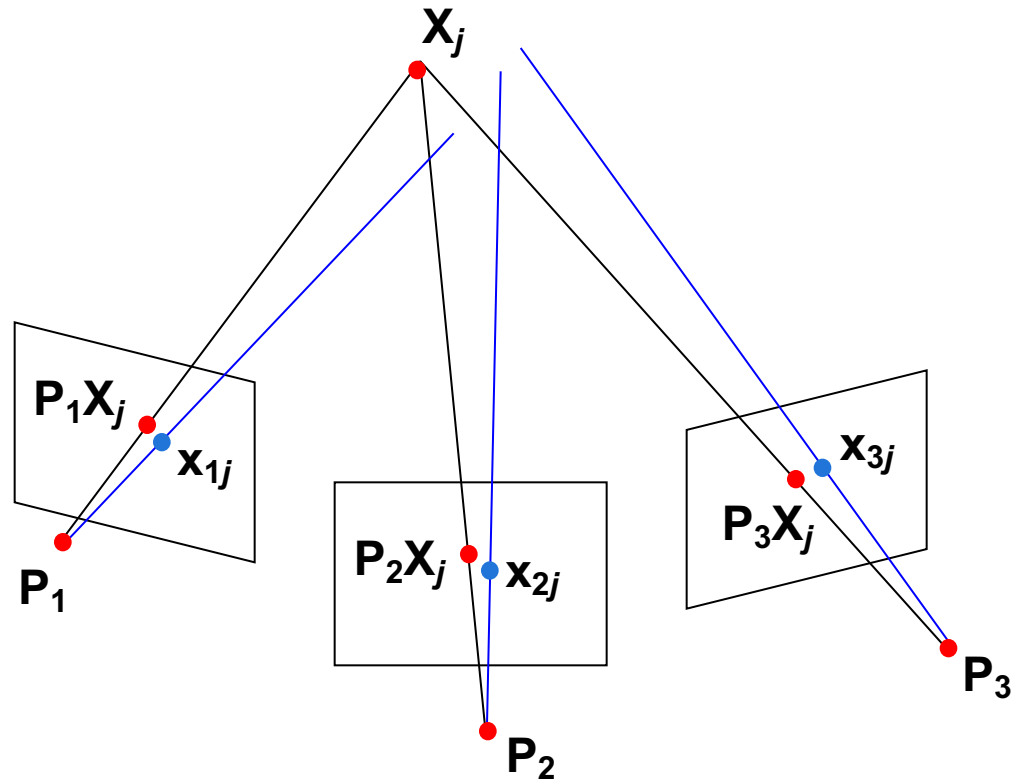


Bundle adjustment

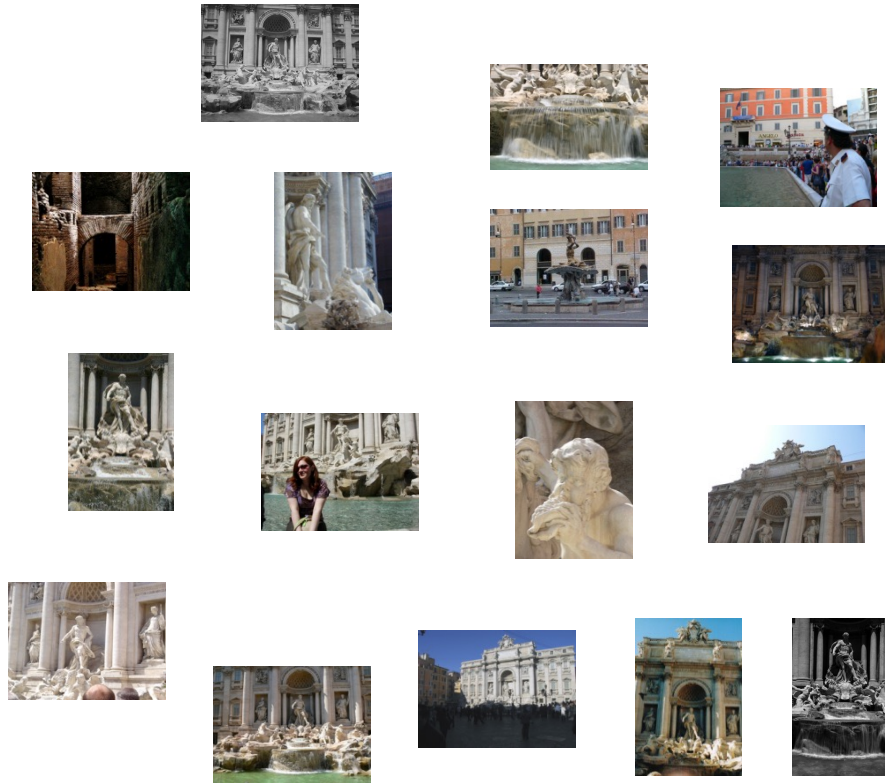
- Non-linear method for refining structure and motion
- Minimize reprojection error

$$\sum_{i=1}^m \sum_{j=1}^n w_{ij} \left\| \mathbf{x}_{ij} - \frac{1}{\lambda_{ij}} \mathbf{P}_i \mathbf{X}_j \right\|^2$$

visibility
flag: is point
j visible in
view i?

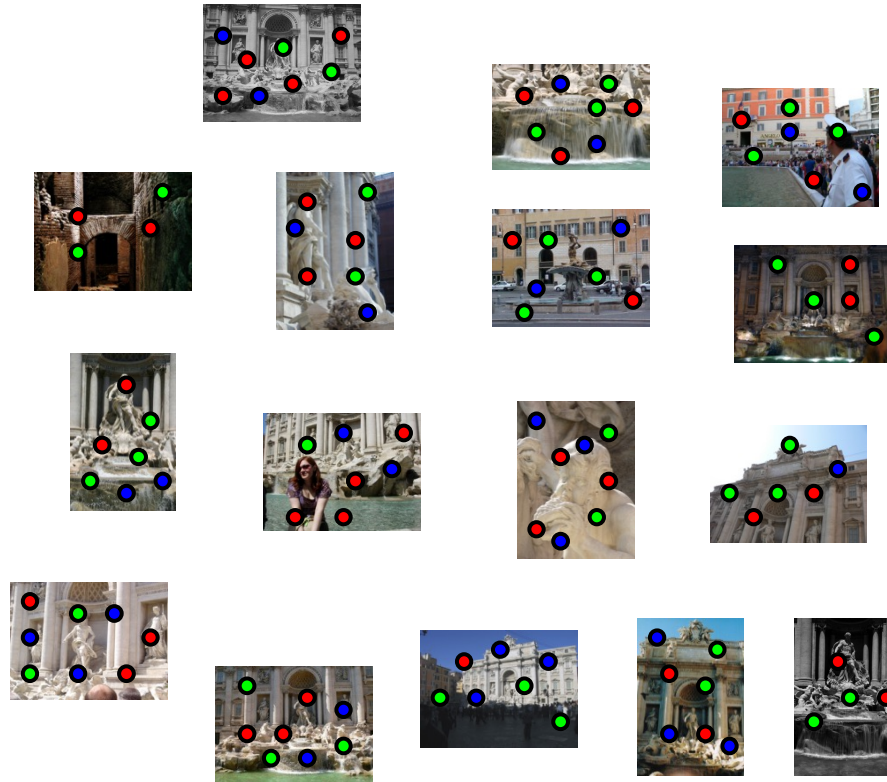


Feature detection



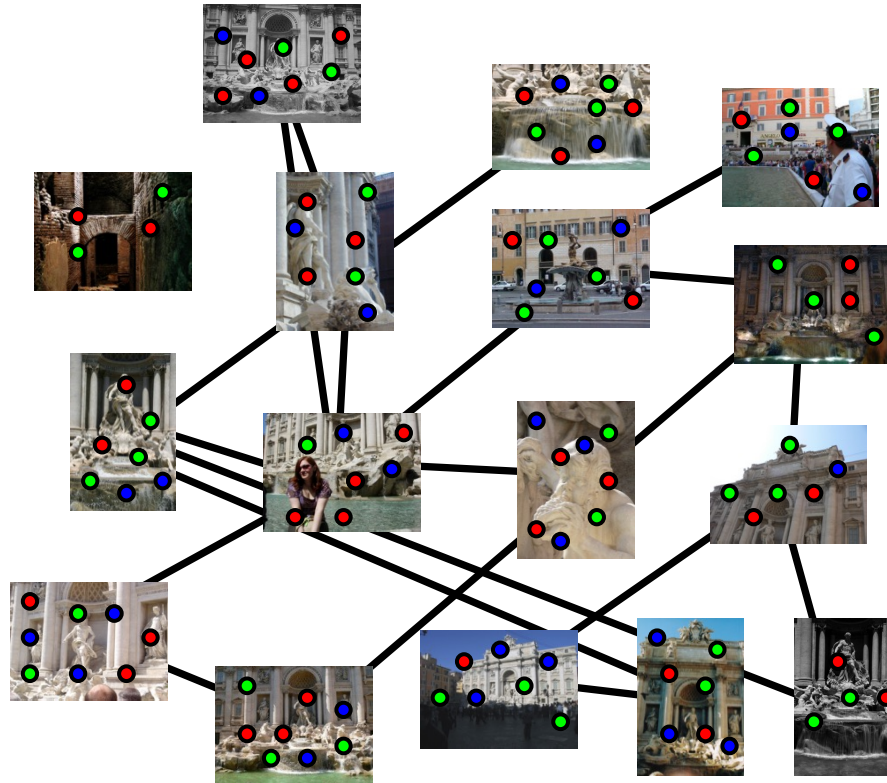
Feature detection

Detect SIFT features



Feature matching

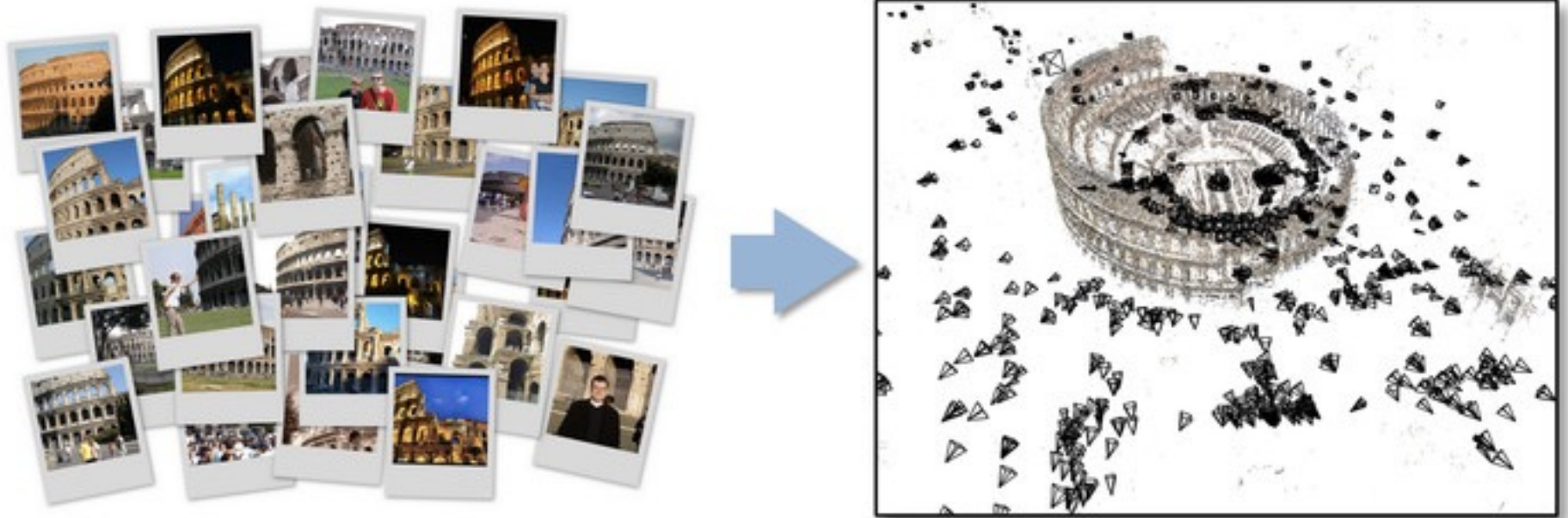
Match features between each pair of images



The devil is in the details

- Handling ambiguities
- Handling degenerate configurations (e.g., homographies)
- Eliminating outliers
- Dealing with repetitions and symmetries

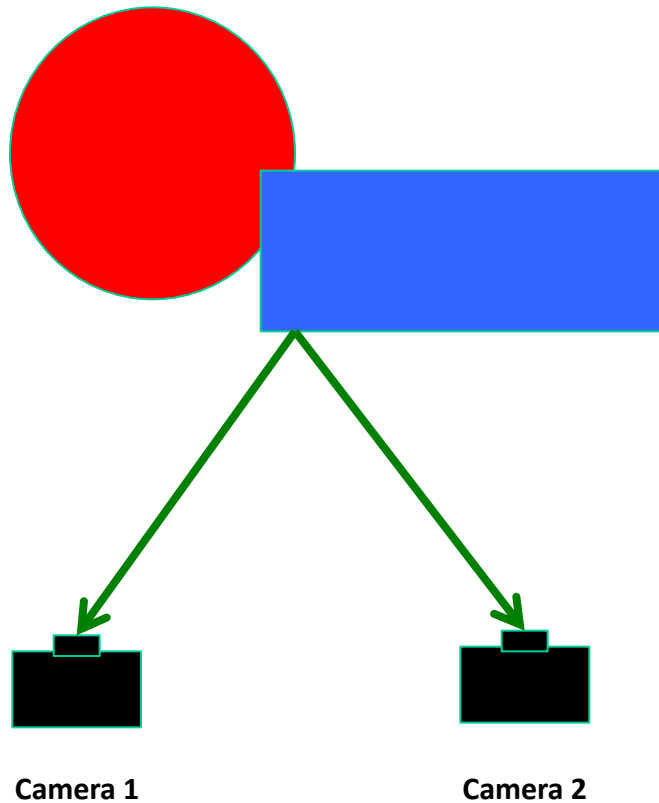
Photo Tourism



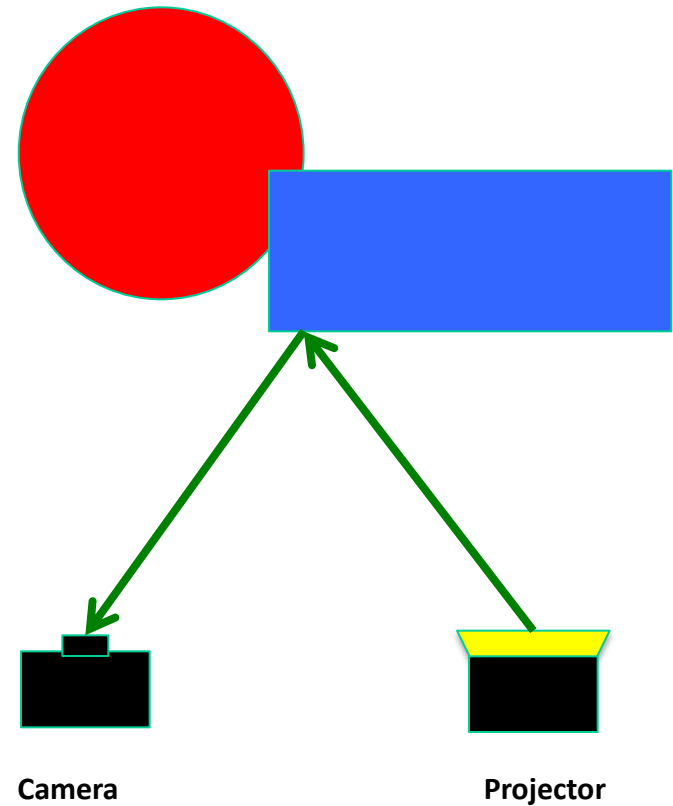
N. Snavely, S. Seitz, and R. Szeliski, Photo tourism: Exploring photo collections in 3D, SIGGRAPH 2006.

<http://phototour.cs.washington.edu/>, <http://grail.cs.washington.edu/projects/rome/>

Depth from Triangulation



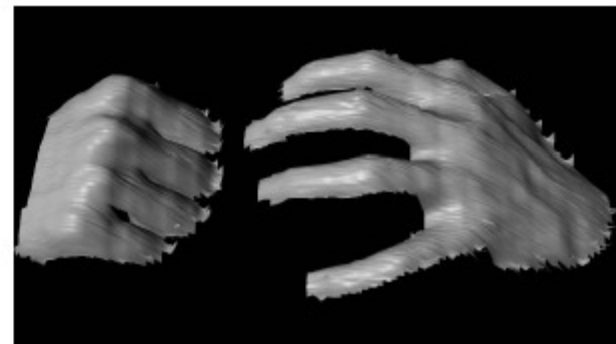
Passive Stereopsis



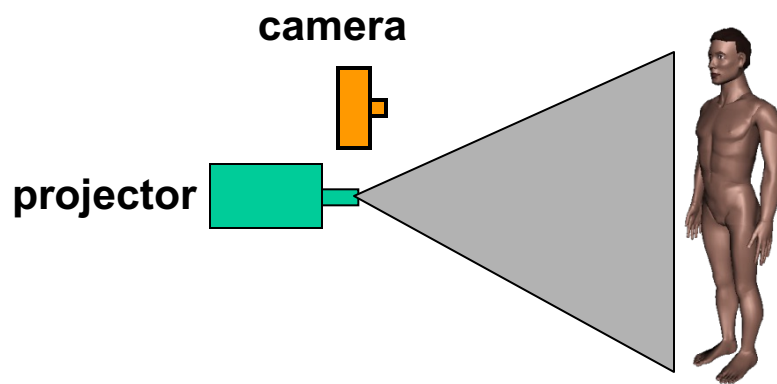
Active Stereopsis

Active sensing simplifies the problem of estimating point correspondences

Active stereo with structured light



- Project “structured” light patterns onto the object
 - Simplifies the correspondence problem
 - Allows us to use only one camera



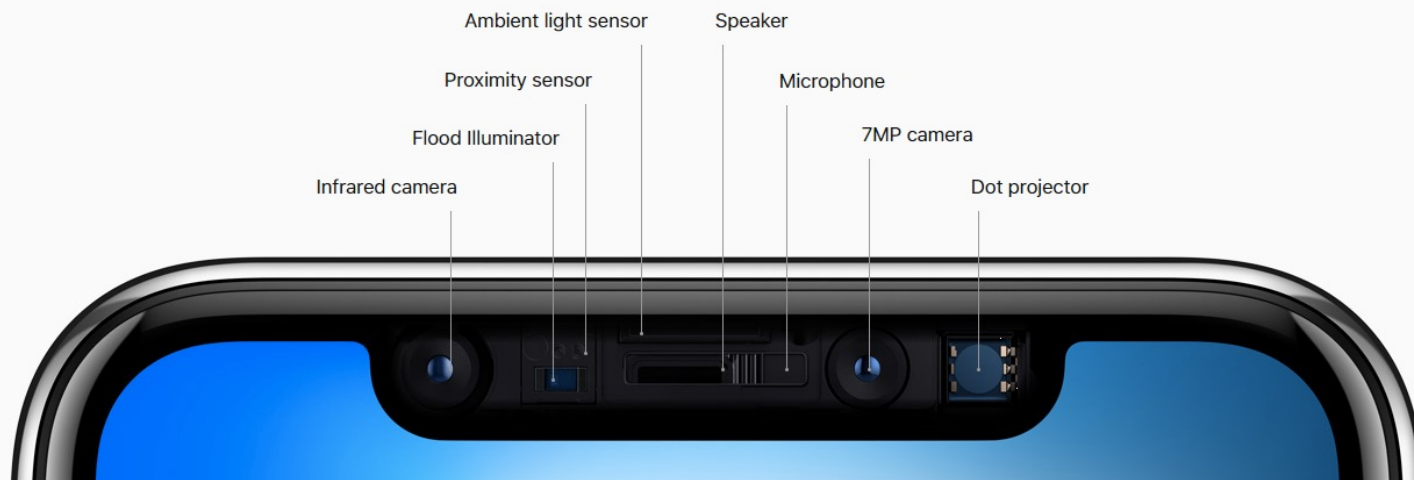
L. Zhang, B. Curless, and S. M. Seitz. Rapid Shape Acquisition Using Color Structured Light and Multi-pass Dynamic Programming. *3DPVT 2002*

Kinect: Structured infrared light



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

Apple TrueDepth



<https://www.cnet.com/news/apple-face-id-truedepth-how-it-works/>



SfM software

- [Bundler](#)
- [OpenSfM](#)
- [OpenMVG](#)
- [VisualSfM](#)
- [Colmap](#)
- See also [Wikipedia's list of toolboxes](#)

Basis for SLAM

- Specialized sensors
- Approximately know camera location
- Need dense reconstructions for path-planning
- Needs to be fast

KinectFusion: Real-time 3D Reconstruction and Interaction Using a Moving Depth Camera*

Shahram Izadi¹, David Kim^{1,3}, Otmar Hilliges¹, David Molyneaux^{1,4}, Richard Newcombe², Pushmeet Kohli¹, Jamie Shotton¹, Steve Hodges¹, Dustin Freeman^{1,5}, Andrew Davison², Andrew Fitzgibbon¹

¹Microsoft Research Cambridge, UK ²Imperial College London, UK

³Newcastle University, UK

⁴Lancaster University, UK

⁵University of Toronto, Canada

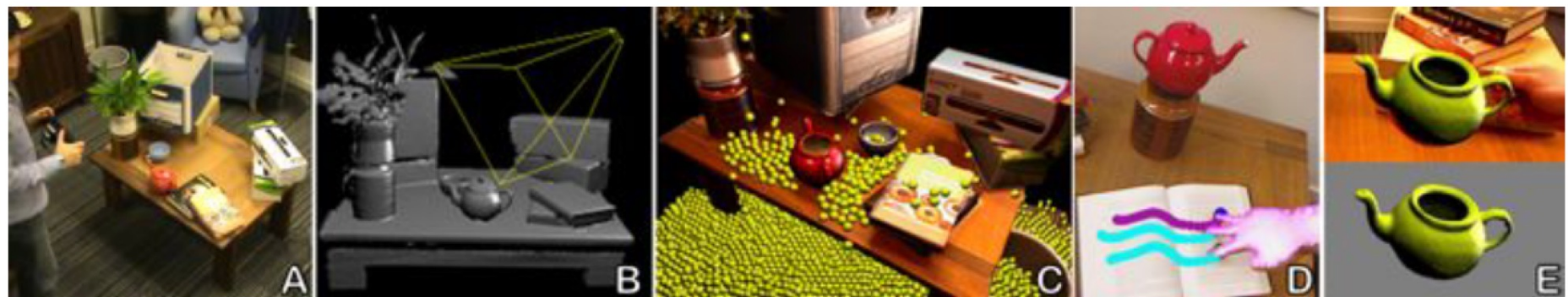


Figure 1: KinectFusion enables real-time detailed 3D reconstructions of indoor scenes using only the depth data from a standard Kinect camera. A) user points Kinect at coffee table scene. B) Phong shaded reconstructed 3D model (the wireframe frustum shows current tracked 3D pose of Kinect). C) 3D model texture mapped using Kinect RGB data with real-time particles simulated on the 3D model as reconstruction occurs. D) Multi-touch interactions performed on any reconstructed surface. E) Real-time segmentation and 3D tracking of a physical object.

[Paper link](#) (ACM Symposium on User Interface Software and Technology, October 2011)

[YouTube Video](#)

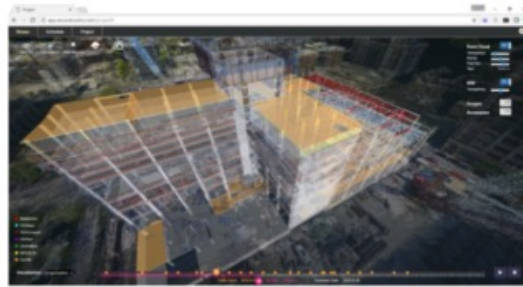
Reconstruction in construction industry

RECONSTRUCT INTEGRATES REALITY AND PLAN



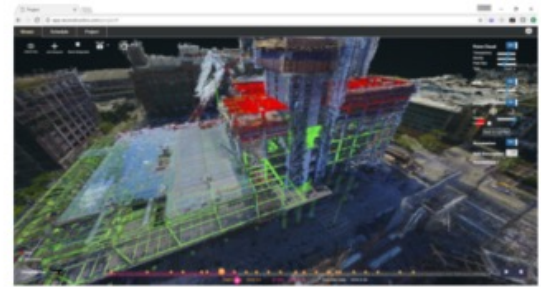
Visual Asset Management

Reconstruct 4D point clouds and organize images and videos from smartphones, time-lapse cameras, and drones around the project schedule. View, annotate, and share anywhere with a web interface.



4D Visual Production Models

Integrate 4D point clouds with 4D BIM, review "who does what work at what location" on a daily basis and improve coordination and communication among project teams.

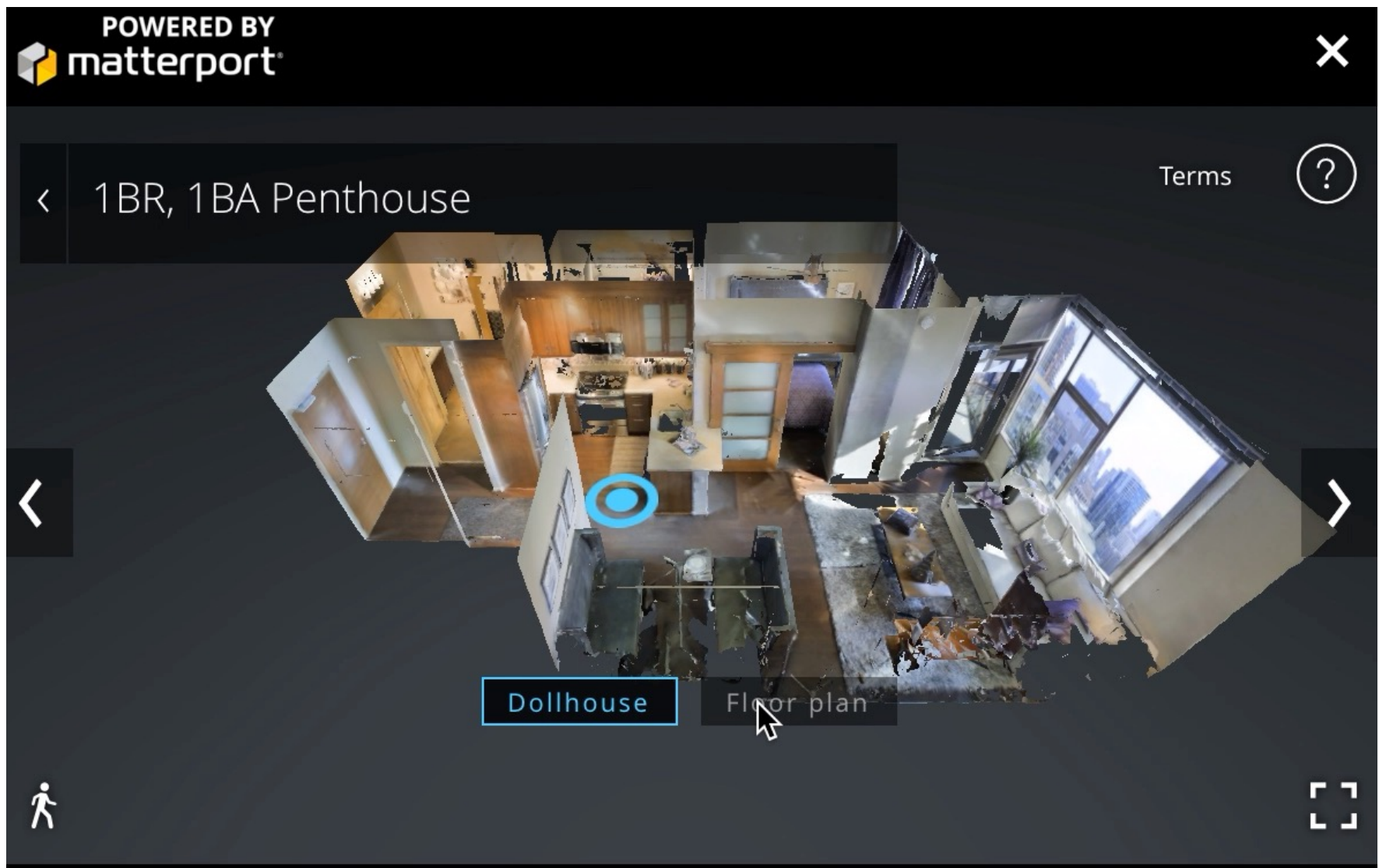


Predictive Visual Data Analytics

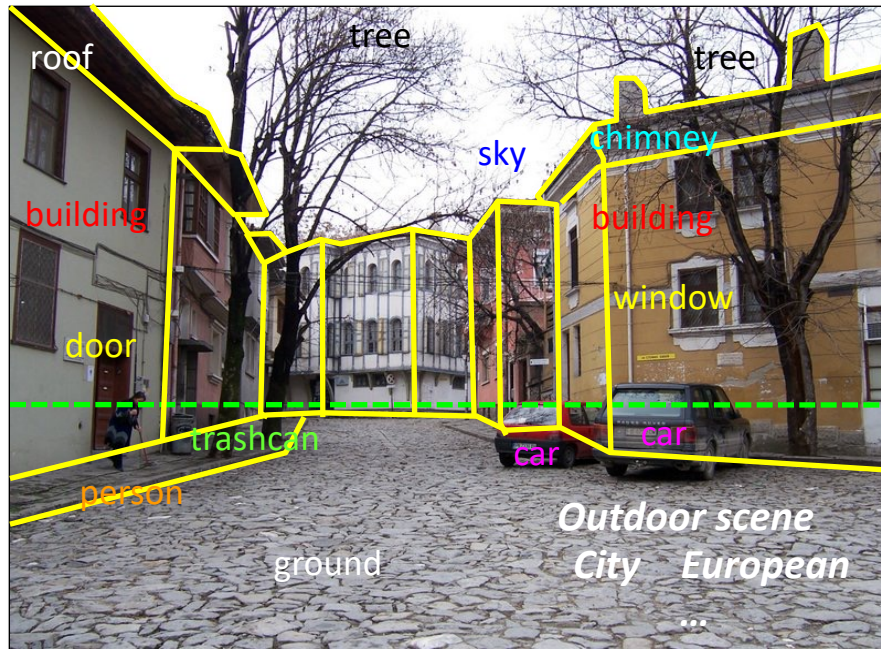
Analyze actual progress deviations by comparing Reality and Plan and predict risk with respect to the execution of the look-ahead schedule for each project location, to offer your project team with an opportunity to tap off potential delays before they surface on your jobsite.

reconstructinc.com

Applications



What kind of information can be extracted from an image?



Geometric information
Semantic information