AA 274 Principles of Robotic Autonomy

Stereo vision and structure from motion





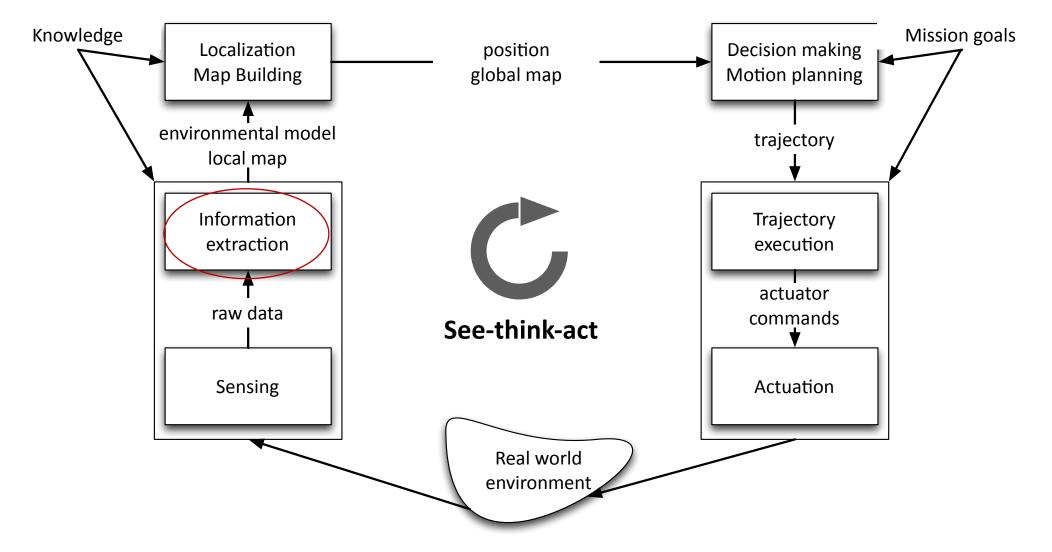
Logistics

- It's the final (project) stretch!
 - All sections are open office hours for project discussion with TAs
 - Final project check-in was due Tuesday
 - Final project demos: Thursday, December 15th, 3:30 6:30pm
 - Sign up for the correct slots with correct group number

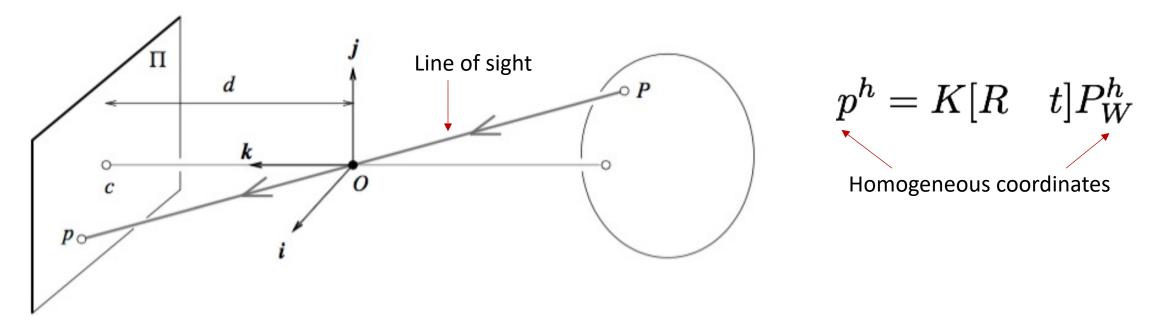
Today's lecture

- Aim
 - Learn fundamental geometric concepts needed for 3D reconstruction
 - Learn basic techniques to recover scene structure, chiefly stereo and structure from motion
- Readings
 - SNS: 4.2.5 4.2.7
 - D. A. Forsyth and J. Ponce [FP]. Computer Vision: A Modern Approach (2nd Edition). Prentice Hall, 2011. Sections 7.1 and 7.2.

The see-think-act cycle



Measuring depth



Once the camera is calibrated, can we measure the location of a point *P* in 3D given its known observation *p*?

• No: one can only say that *P* is located *somewhere* along the line joining *p* and *O*!

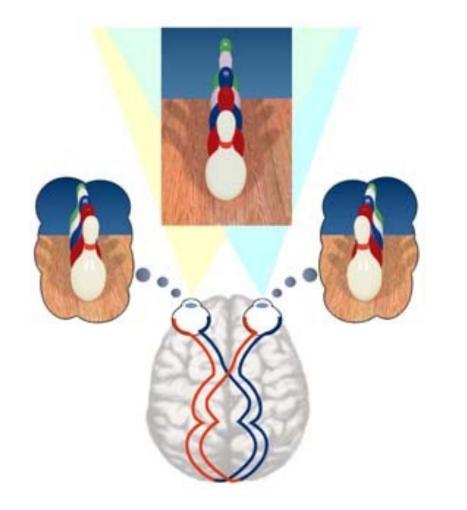
Recovering structure

• Structure: 3D scene to be reconstructed by having access to 2D images

Common methods

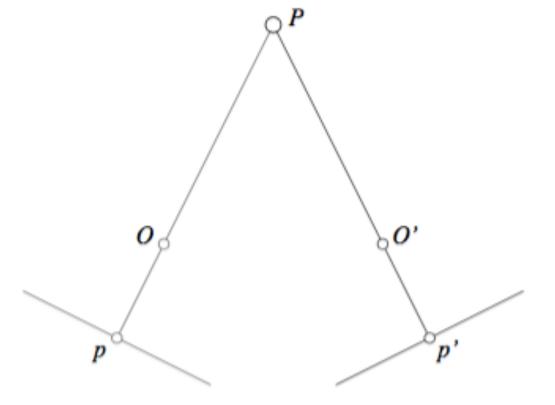
- 1. Through recognition of landmarks (e.g., orthogonal walls)
- 2. Depth from focus: determines distance to one point by taking multiple images with better and better focus
- 3. Stereo vision: processes two distinct images taken at the *same time* and assumes that the relative pose between the two cameras is *known*
- 4. Structure from motion: processes two images taken with the same or different cameras at *different times* and from different *unknown* positions

Stereopsis (why we have two eyes)



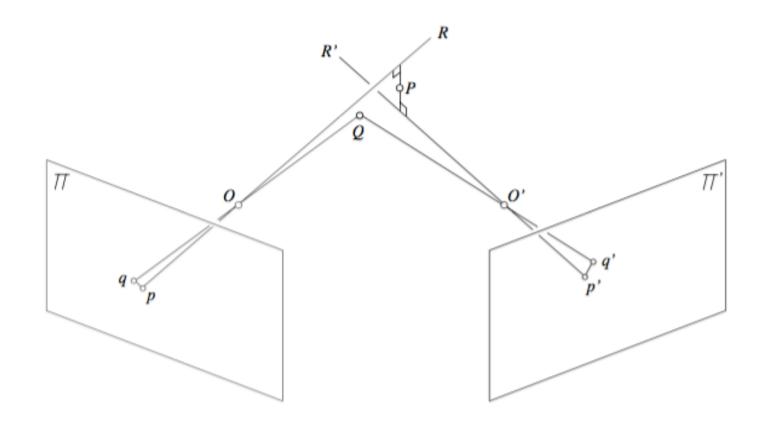


Binocular reconstruction



- Given: calibrated stereo rig and two image matching points p and p'
- Find corresponding scene point by intersecting the two rays Op and O'p' (process known as triangulation)

Approximate triangulation



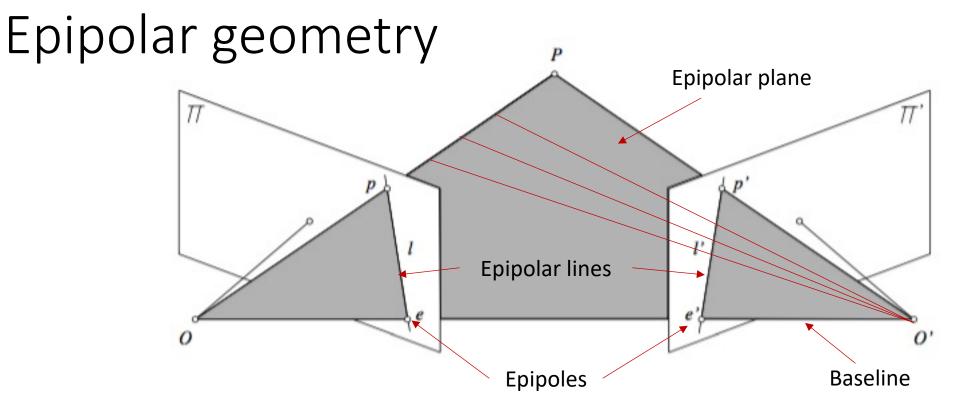
 Due to noise, triangulation problem is often solved as finding the point Q with images q and q' that minimizes

$$d^2(p,q) + d^2(p',q')$$

Re-projection error

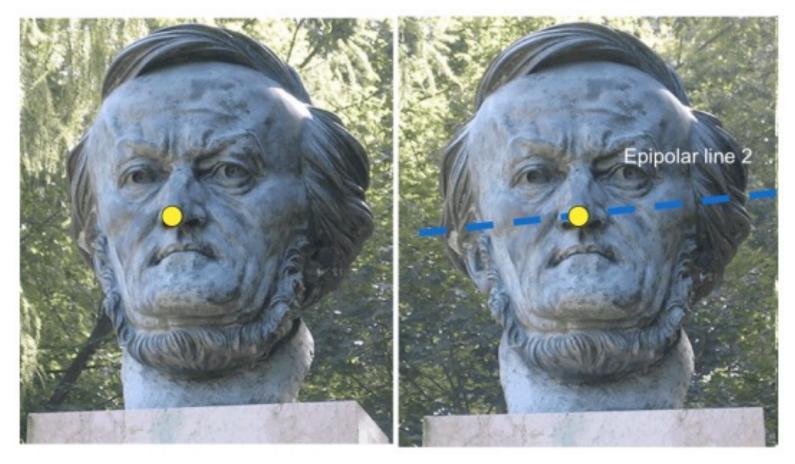
Stereo vision process

- Stereo vision consists of two steps:
 - 1. *fusion* of features observed by two (or more) cameras -> correspondence
 - 2. reconstruction of their three-dimensional preimages -> triangulation
- Step 2 is relatively easy; Step 1 requires you to establish correct correspondences and avoid erroneous depth measurements
- Several constraints can be leveraged to simplify Step 1 (e.g., similarity constraint, continuity constraints, etc.); most important: epipolar constraint



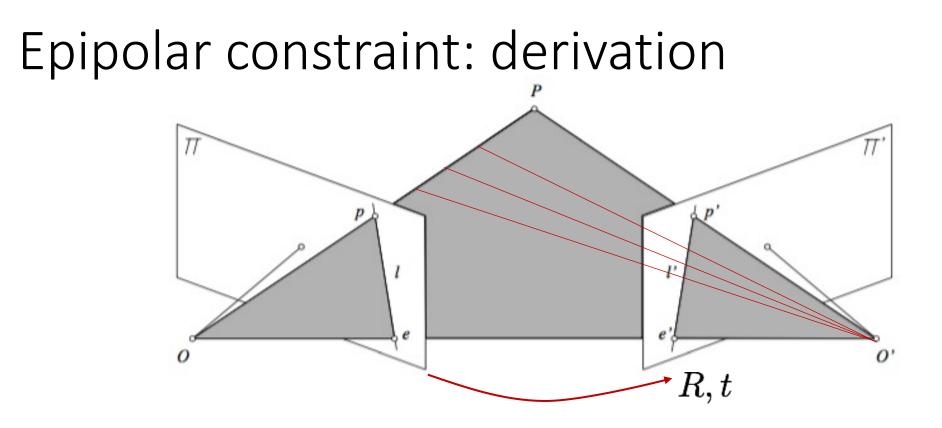
- Consider images p and p' of a point P observed by two cameras from O, O'
- These five points all belong to the *epipolar plane* defined by p, O, O', or equivalently, p', O, O'
- Epipolar constraint: potential matches for p must lie on epipolar line l' (and vice-versa)

Epipolar constraint



 Search for matches can be restricted to the epipolar line instead of the whole image! → one dimensional search

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• Epipolar constraint: \overline{Op} , $\overline{O'p'}$, and $\overline{OO'}$ must be coplanar, or

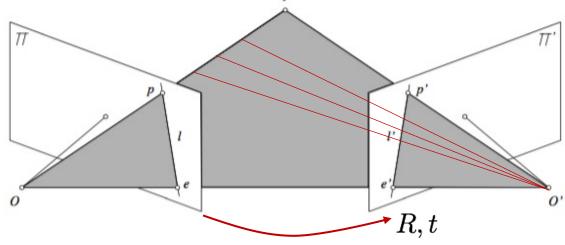
$$\overline{Op} \cdot [\overline{OO'} \times \overline{O'p'}] = 0$$

Aside: matrix notation for cross product

 Cross product can be expressed as the product of a skew-symmetric matrix and a vector

$$a \times b = \begin{bmatrix} 0 & -a_3 & a_2 \\ a_3 & 0 & -a_1 \\ -a_2 & a_1 & 0 \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} = [a]_{\times} b$$
$$:= [a]_{\times}$$

Epipolar constraint: derivation. More details: CS231a



- Assume that the world reference system is co-located with camera 1
- After some algebra, epipolar constraint becomes [FP, Section 7.1]

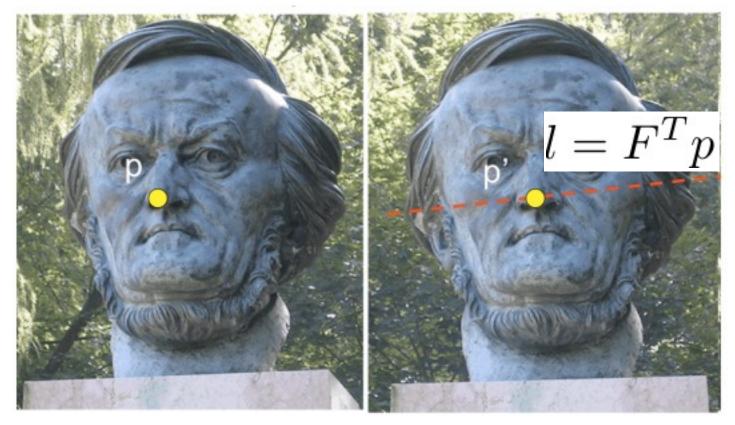
$$p^T F p' = 0$$

where: $F = K^{-T} [t]_{\times} R K'^{-1}$

Key facts

- *F* is referred to as the fundamental matrix
- l = Fp' (resp. $l' = F^Tp$) represents the epipolar line corresponding to the point p' (resp. p) in the first (resp. second) image. This exploits the homogenous notation for lines.
- $F^T e = F e' = 0 \rightarrow F$ is also singular (as t is parallel to the coordinate vectors of the epipoles)
- F has 7 DoF (9 elements common scaling det(F)=0)

Usefulness of fundamental matrix



- Assume *F* is given
- Given a point in image 1, one can compute the corresponding epipolar line in image 2 without any additional information needed!

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Estimating the fundamental matrix

8-point algorithm

 $p = [u, v, 1]^{T}, \quad p' = [u', v', 1]^{T} \implies [u, v, 1] \begin{bmatrix} F_{11} & F_{12} & F_{13} \\ F_{21} & F_{22} & F_{23} \\ F_{31} & F_{32} & F_{33} \end{bmatrix} \begin{bmatrix} u' \\ v' \\ 1 \end{bmatrix} = 0$ $\implies [uu', uv', u, vu', vv', v, u', v', 1] \begin{bmatrix} F_{11} \\ F_{12} \\ F_{13} \\ F_{21} \\ F_{22} \\ F_{23} \\ F_{31} \\ F_{32} \\ F_{32} \\ F_{32} \end{bmatrix} = 0 \implies Wf = 0$ $\xrightarrow{n \times 9 \text{ matrix of known coefficients}} f = 0$ F_{33} $\min_{f \in R^9} \|Wf\|^2 \Longrightarrow \tilde{F}$ • Given $n \ge 8$ correspondences, one then solves subject to $||f||^2 = 1$ 12/08/22 AA 274 | Lecture 18

Enforcing the rank constraint

- \tilde{F} satisfies the epipolar constraints, but is not necessarily singular (hence, is not necessarily a proper fundamental matrix)
- Enforce rank constraint (again, via SVD decomposition)

Find F that minimizes $\|F - \tilde{F}\|^2$ \leftarrow Frobenius norm

subject to det(F) = 0

- 8-point algorithm
 - 1. Use linear least squares to compute \tilde{F}
 - 2. Enforce rank-2 constraint via SVD

Parallel image planes

- Assume image planes are parallel
- Epipolar lines are horizontal
- v coordinates are equal
 - Easier triangulation
 - Easier correspondence problem
- Is it possible to warp images to simulate a parallel image plane?

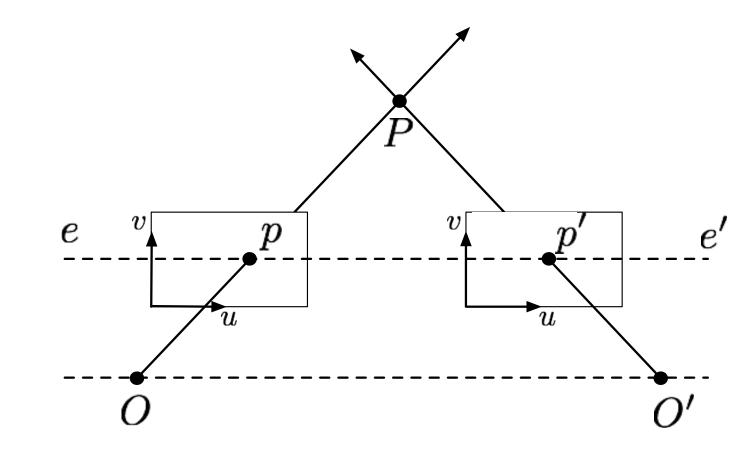
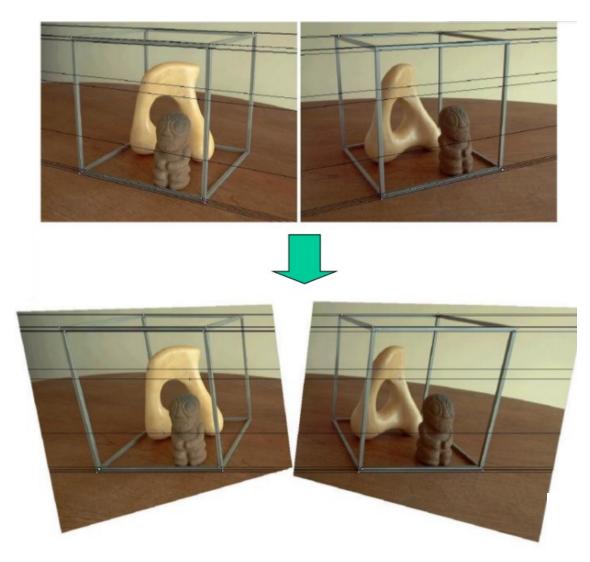


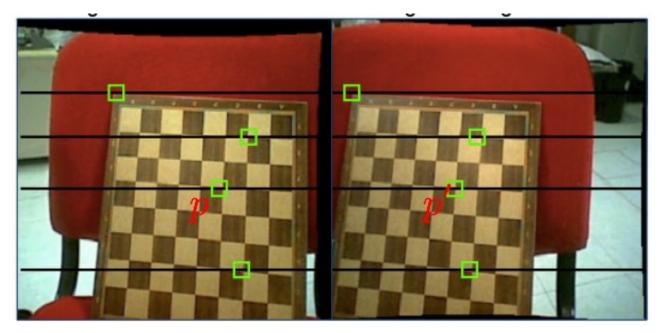
Image rectification



- Achieved by applying an appropriate projective transformation
- Several algorithms exist
- From now on, we assume rectified image pairs

Back to stereo vision process

- Recall that stereo vision consists of two steps:
 - *1. fusion* of features observed by two (or more) cameras (correspondence)
 - 2. reconstruction of their three-dimensional preimages (triangulation)
- Correspondence problem

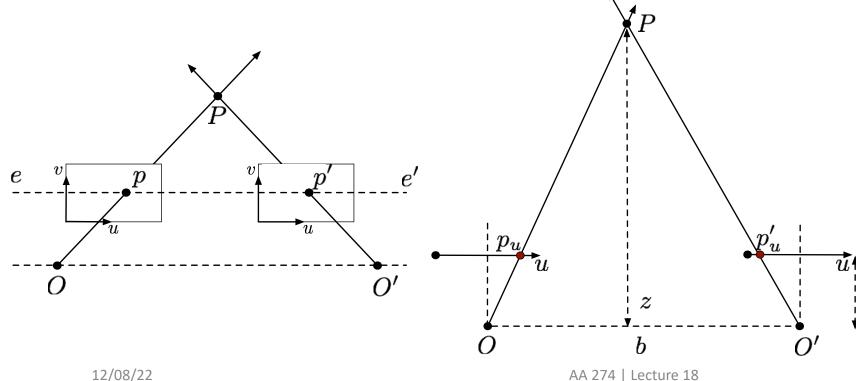


Goal: find corresponding observations p and p' Exploits epipolar constraints Two classes of algos: *area-based* and *feature-based* Hard problem: occlusions, repetitive patterns, etc.; more on this later

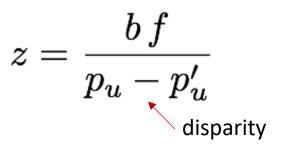
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Triangulation under rectified images

- We already saw how to triangulate correspondences in the general case
- Triangulation problem under rectified images:



From similar triangles:



Large baseline: Object might be visible from one camera, but not the other

Small baseline: large depth error 23

Disparity map

- Disparity: pixel displacement between corresponding points
- Disparity map: holds the disparity values for every pixel
- Nearby objects experience largest disparity

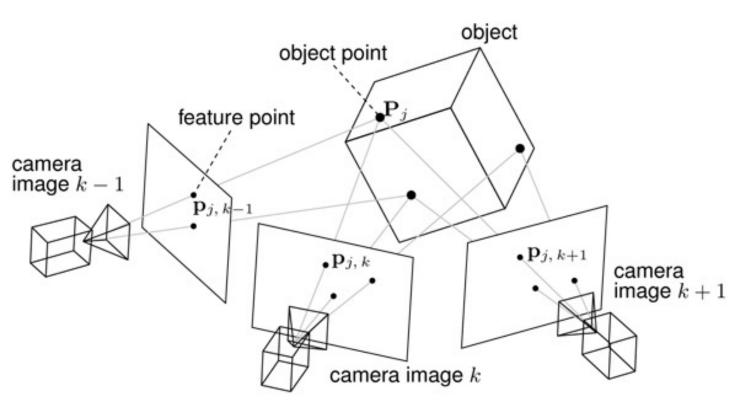


Disparity map



Stereo pair

Method #3: structure from motion (SFM)



Given *m* images of *n* fixed 3D points

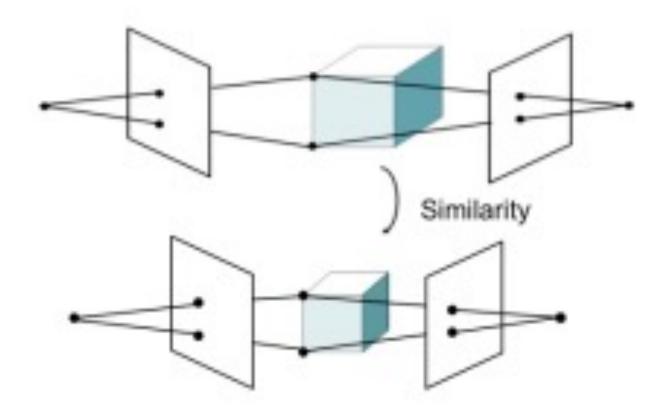
 $p_{j,k}^h = M_k P_j^h$

Find:

- *m* projection matrices *M_k* (motion)
- *n* 3D points *P_i* (structure)

SFM ambiguity

• It is not possible to recover the absolute scale of the observed scene

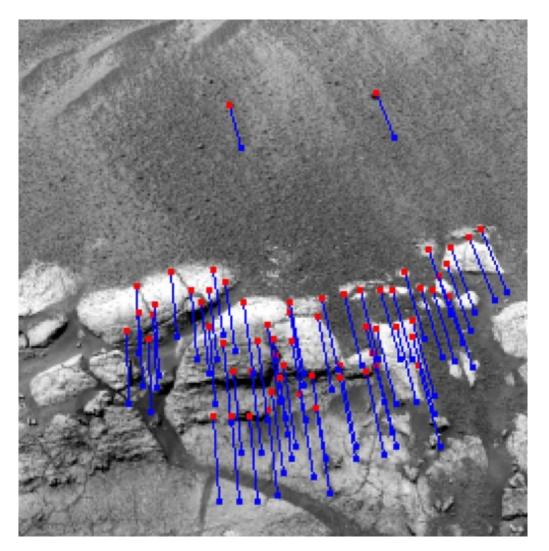


Solution to SFM problem (high-level)

- Several approaches available:
 - Algebraic approach (by fundamental matrix)
 - Bundle adjustment
- Algebraic approach (2-views)
 - 1. Compute fundamental matrix *F* (e.g., via 8-point algorithm)
 - 2. Use *F* to estimate projection camera matrices
 - 3. Use projection camera matrices for triangulation

Application of SFM: visual odometry

- Visual odometry: estimate the motion of the robot by using visual input (and possibly additional information)
 - Single camera: absolute scale must be estimated in other ways
 - Stereo camera: measurements are directly provided in absolute scale



Thanks for a great quarter!