# AA 274 <br> 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 $p^{\prime}$
- Find corresponding scene point by intersecting the two rays $\overline{O p}$ and $\overline{O^{\prime} p^{\prime}}$ (process known as triangulation)


## Approximate triangulation



- Due to noise, triangulation problem is often solved as finding the point $Q$ with images $q$ and $q^{\prime}$ that minimizes

$$
\underbrace{d^{2}(p, q)+d^{2}\left(p^{\prime}, q^{\prime}\right)}_{\text {Re-projection error }}
$$

## 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


## Epipolar geometry



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


## Epipolar constraint



- Search for matches can be restricted to the epipolar line instead of the whole image! $\rightarrow$ one dimensional search


## Epipolar constraint: derivation



- Epipolar constraint: $\overline{O p}, \overline{O^{\prime} p^{\prime}}$, and $\overline{O O^{\prime}}$ must be coplanar, or

$$
\overline{O p} \cdot\left[\overline{O O^{\prime}} \times \overline{O^{\prime} p^{\prime}}\right]=0
$$

## Aside: matrix notation for cross product

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

$$
a \times b=\underbrace{\left[\begin{array}{ccc}
0 & -a_{3} & a_{2} \\
a_{3} & 0 & -a_{1} \\
-a_{2} & a_{1} & 0
\end{array}\right]}_{:=[a]_{\times}}\left[\begin{array}{l}
b_{1} \\
b_{2} \\
b_{3}
\end{array}\right]=[a]_{\times} b
$$

## 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]

$$
p^{T} F p^{\prime}=0
$$

where: $F=K^{-T}[t]_{\times} R K^{\prime-1}$

## Key facts

- $F$ is referred to as the fundamental matrix
- $l=F p^{\prime}$ (resp. $l^{\prime}=F^{T} p$ ) represents the epipolar line corresponding to the point $p^{\prime}$ (resp. $p$ ) in the first (resp. second) image. This exploits the homogenous notation for lines.
- $F^{T} e=F e^{\prime}=0 \rightarrow F$ is also singular (as $t$ is parallel to the coordinate vectors of the epipoles)
- $F$ has 7 DoF ( 9 elements - common scaling $-\operatorname{det}(F)=0$ )


## 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!


## Estimating the fundamental matrix

- 8-point algorithm

$$
\begin{aligned}
p=[u, v, 1]^{T}, \quad p^{\prime}=\left[u^{\prime}, v^{\prime}, 1\right]^{T}
\end{aligned} \Rightarrow \quad[u, v, 1]\left[\begin{array}{lll}
F_{11} & F_{12} & F_{13} \\
F_{21} & F_{22} & F_{23} \\
F_{31} & F_{32} & F_{33}
\end{array}\right]\left[\begin{array}{c}
u^{\prime} \\
v^{\prime} \\
1
\end{array}\right]=00
$$

- Given $n \geq 8$ correspondences, one then solves

$$
\min _{f \in R^{9}}\|W f\|^{2} \Rightarrow \tilde{F}
$$

$$
\text { subject to }\|f\|^{2}=1
$$

## 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)

$$
\begin{aligned}
\text { Find } F \text { that minimizes } & \|F-\tilde{F}\|^{2} \longleftarrow \text { Frobenius norm } \\
\text { subject to } & \operatorname{det}(F)=0
\end{aligned}
$$

- 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



## Triangulation under rectified images

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


From similar triangles:

$$
z=\frac{b f}{p_{u}-p_{u}^{\prime}}
$$

Large baseline: Object might be visible from one camera, but not the other
Small baseline: large depth error

## Disparity map

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

Stereo pair



Disparity map


## 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 $P_{j}$ (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!

