

Geometry of standard stereo

$$\frac{u_1}{f} = \frac{h - X}{Z}$$

$$-\frac{u_2}{f} = \frac{h + X}{Z}$$

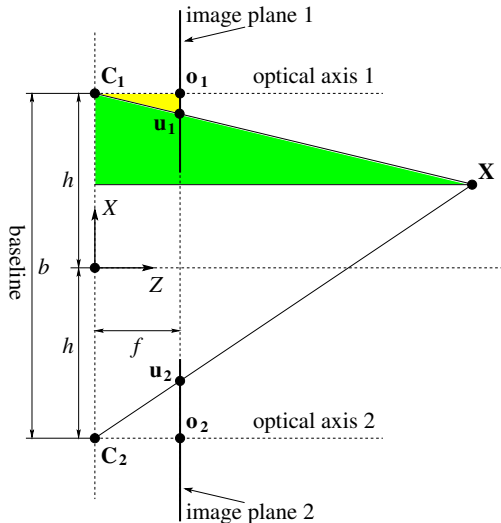
$$v_1 = v_2$$

$$Z = \frac{2hf}{u_1 - u_2} = \frac{bf}{d}$$

$$X = -\frac{b(u_1 + u_2)}{2d}$$

$$Y = \frac{bv_1}{d} = \frac{bv_2}{d}$$

$$d \doteq u_1 - u_2 \text{ disparity}$$



Precision of depth estimation

- If $d \rightarrow 0$, and $Z \rightarrow \infty$
 - Disparity of distant points are small.
- Relation between disparity and precision of depth estimation

$$\frac{|\Delta Z|}{Z} = \frac{|\Delta d|}{|d|}$$

- larger the disparity, smaller the relative depth error
→ precision is increasing
- Influence of base length

$$d = \frac{bf}{Z}$$

- For larger b , same depth value yields larger disparity
→ Precision of depth estimation increasing
→ more pixels → precision of disparity increasing

Types of stereo reconstruction

- **Fully calibrated** reconstruction

- Known **intrinsic and extrinsic** camera parameters
- reconstruction by triangulation
- known baseline \rightarrow known scale

- **Metric (Euclidean)** reconstruction

- known **intrinsic** camera parameters, $n \geq 8$ point correspondences given
 - Extrinsic camera parameters obtained from **essential matrix**
 - Reconstruction up to a similarity transformation
- \rightarrow up to a scale

- **Projective reconstruction**

- **unknown** camera parameters, $n \geq 8$ point correspondences are given
- Composition of projective matrices from a **fundamental matrix**
- reconstruction can be computed up to a projective transformation

Overview

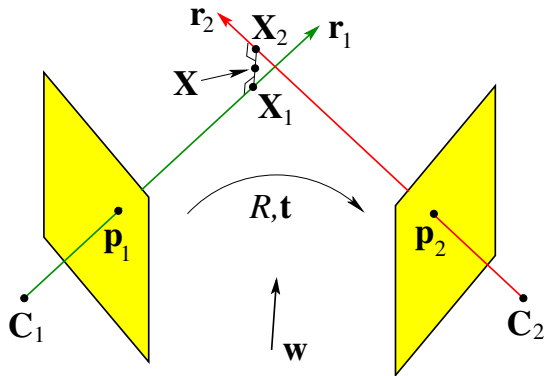
- 1 Image-based 3D reconstruction
- 2 Geometry of stereo vision
 - Epipolar geometry
 - Essential and fundamental matrices
 - Estimation of the fundamental matrix
- 3 Standard stereo and rectification
 - Triangulation for standard stereo
 - Rectification of stereo images
- 4 3D reconstruction from stereo images
 - Triangulation and metric reconstruction
 - Projective reconstruction
 - Planar Motion
- 5 Summary

Triangulation

● Task:

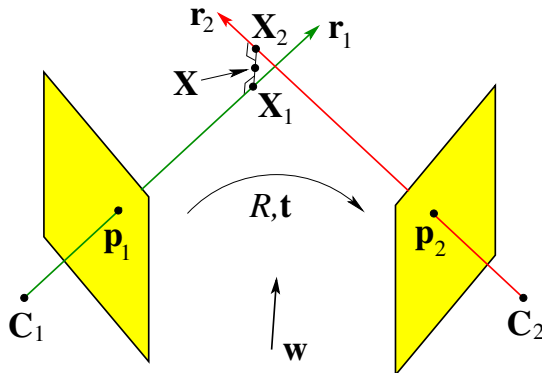
- Two calibrated cameras are given, including both intrinsic and extrinsic parameters, and
 - Locations $\mathbf{u}_1, \mathbf{u}_2$ of the projection of spatial point \mathbf{X} are given
 - Goal is to estimate spatial location \mathbf{X} .
- Two calibration matrices are known, therefore
 - for a projection matrix: $\mathbf{K}^{-1}\mathbf{P} = [\mathbf{R} | -\mathbf{t}]$ and
 - for calibrated (aka. normalized) coordinates: $\mathbf{p} = \mathbf{K}^{-1}\mathbf{u}$.
 - For the sake of simplicity, the first camera gives the world coordinate system
 - **non-homogeneous** coordinates are used
 - $\mathbf{p}_2 = \mathbf{R}(\mathbf{p}_1 - \mathbf{t}), \mathbf{p}_1 = \mathbf{t} + \mathbf{R}^T\mathbf{p}_2$
 - Image points are back-projected to 3D space
 - two rays obtained, they usually do not intersect each other due to noise/calibration error
 - task is to give an estimate for spatial point \mathbf{X} .

Linear triangulation: geometry



- Line $\mathbf{X}_1\mathbf{X}_2$ perpendicular to both \mathbf{r}_1 and \mathbf{r}_2 .
- Estimate \mathbf{X} is the middle point of section $\mathbf{X}_1\mathbf{X}_2$
- Vector \mathbf{w} is parallel to $\mathbf{X}_1\mathbf{X}_2$.

Linear triangulation: notations



- αp_1 is a point on ray r_1 ($\alpha \in \mathbb{R}$)
- $t + \beta R^T p_2$ a point on other ray r_2 ($\beta \in \mathbb{R}$)
 → coordinate system fixed to the first camera
- Let $X_1 = \alpha_0 p_1$, $X_2 = t + R^T(\beta_0 p_2 - t)$

Linear triangulation: solution

- Task is to determine
 - the middle point of the line section $\mathbf{X}_1\mathbf{X}_2$
→ determination of α_0 and β_0 required
 - Remark that
 - Vector $\mathbf{w} = \mathbf{p}_1 \times \mathbf{R}^T(\mathbf{p}_2 - \mathbf{t})$ perpendicular to both \mathbf{r}_1 and \mathbf{r}_2 .
 - Line $\alpha\mathbf{p}_1 + \gamma\mathbf{w}$ parallel to \mathbf{w} and contain the point $\alpha\mathbf{p}_1$ ($\gamma \in \mathbb{R}$).
- α_0, β_0 (as well as γ_0) are given by the solution of the following linear system: :

$$\alpha\mathbf{p}_1 + \mathbf{t} + \beta\mathbf{R}^T(\mathbf{p}_2 - \mathbf{t}) + \gamma[\mathbf{p}_1 \times \mathbf{R}^T(\mathbf{p}_2 - \mathbf{t})] = 0 \quad (7)$$

- Triangulated point is obtained, e.g by $\alpha_0\mathbf{p}_1$
- There is no solution if \mathbf{r}_1 and \mathbf{r}_2 are parallel

Linear triangulation: an algebraic solution

- Two projected locations of spatial point \mathbf{X} are given:

$$\lambda_1 \mathbf{u}_1 = \mathbf{P}_1 \mathbf{X}$$

$$\lambda_2 \mathbf{u}_2 = \mathbf{P}_2 \mathbf{X}$$

- λ_1 and λ_2 can be eliminated. 2 + 2 equations are obtained:

$$u \mathbf{p}_3^T \mathbf{X} = \mathbf{p}_1^T \mathbf{X}$$

$$v \mathbf{p}_3^T \mathbf{X} = \mathbf{p}_2^T \mathbf{X}$$

- where \mathbf{p}_i^T is the i -th row of projection matrix \mathbf{P} .
- Both projections yield 2 equations. Only vector \mathbf{X} is unknown.
- Solution for \mathbf{X} is calculated by solving the homogeneous linear system of equations.
- Important remark: solution is obtained in homogeneous coordinates.

Refinement by minimizing the reprojection error

- Linear algorithm yield points \mathbf{X}_i , $i = 1, 2, \dots, n$ if n point pairs are given
- The solution should be refined
 - minimization of **reprojection error** yields **more accurate** estimate
- For minimizing the reprojection error, the following parameters have to be refined:
 - Spatial points \mathbf{X}_i
 - Rotation matrix \mathbf{R} and baseline vector \mathbf{t}
 - intrinsic camera parameters are usually fixed as cameras are pre-calibrated
- Initial values for numerical optimization
 - Spatial points \mathbf{X}_i from linear triangulation
 - Initial rotation matrix \mathbf{R} and baseline vector \mathbf{t} by decomposing the essential matrix

Metric reconstruction by decomposing the essential matrix

- Intrinsic camera matrices \mathbf{K}_1 and \mathbf{K}_2 given, fundamental matrix computed from $n \geq 8$ point correspondences
 - \mathbf{E} can be retrieved from \mathbf{F} , \mathbf{K}_1 and \mathbf{K}_2 .
 - from \mathbf{E} , extrinsic parameters can be obtained by decomposition
- Unknown baseline \longrightarrow unknown scale
 - baseline normalized to 1
 - \longrightarrow Euclidean reconstruction possible up to a similarity transformation
- It is assumed that world coordinate is fixed to the first camera
 - \longrightarrow Therefore, $P_1 = [I|0]$, where I is the identity matrix
- Position of second camera computed from essential matrix \mathbf{E} by SVD.
 - Four solutions obtained,
 - only one is correct.

Camera pose estimation by SVD

- The Singular Value Decomposition of \mathbf{E} is $\mathbf{E} = \mathbf{UDV}^T$, where $\mathbf{D} = \text{diag}(\delta, \delta, 0)$
 → \mathbf{E} has two equal singular values
- Four solutions can be obtained as follows:

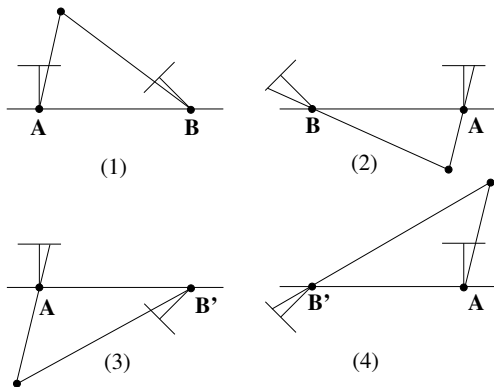
$$\begin{aligned}\mathbf{R}_1 &= \mathbf{UWV}^T & \mathbf{R}_2 &= \mathbf{UW}^T\mathbf{V}^T \\ [\mathbf{t}_1]_{\times} &= \delta\mathbf{UZU}^T & [\mathbf{t}_2]_{\times} &= -\delta\mathbf{UZU}^T\end{aligned}$$

- where

$$\mathbf{W} \doteq \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix} \quad \mathbf{Z} \doteq \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

- Combination of 2-2 candidates for translation and rotation yield 4 solutions.
- Determinants of \mathbf{R}_1 and \mathbf{R}_2 have to be positive, otherwise matrices should be multiplied by -1 .

Visualization of the four solutions



- Left and right: camera locations replaces
- Top and bottom: mirror to base lane
- 3D point is in front of the cameras only in the top-left case.