

# Geometry of standard stereo

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

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

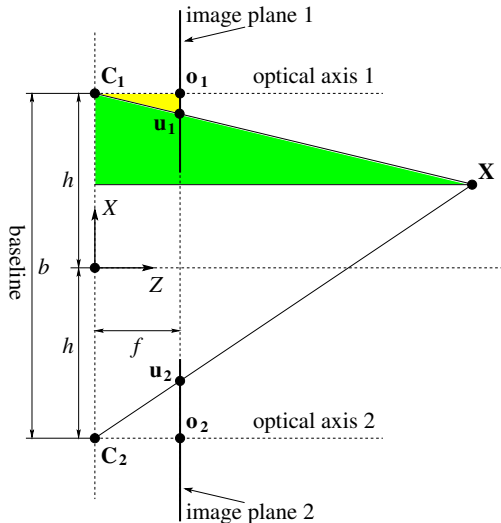
$$v_1 = v_2$$

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

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

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

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



# 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

# 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
  - Retification of stereo images
- 4 3D reconstruction from stereo images
  - Triangulation and metric reconstruction
  - Projective reconstruction
  - Planar Motion
- 5 Summary

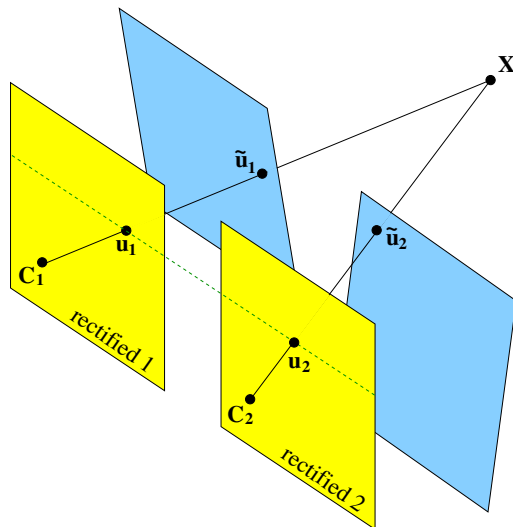
# Goals of rectification

- Input of rectification: non-standard stereo image pair
- Goal of rectification: make stereo matching more accurate
  - After rectification, corresponding pixels are located in the same row
    - standard stereo, 1D search
- Rectification based on epipolar geometry
  - Images are transformed based on epipolar geometry
    - after transformation, corresponding epipolar lines are placed on the same rows
    - epipoles are in the infinity
- For rectification, only the fundamental matrix has to be known
  - Fundamental matrix represents epipolar geometry

# Rectification methods

- Only the general principles are discussed here.
  - Rectification is a complex method.
  - Rectification **is not required**, it has both advantages and disadvantages.
- Rectification can be carried out by homographies.
- It has ambiguity: there are infinite number of rectification transformations for the same image pair.
- The aim is to find a 2D projective transformation that
  - fulfills the requirement for rectification and
  - distorts minimally the images.
- Knowledge of camera intrinsic parameters helps the rectification.

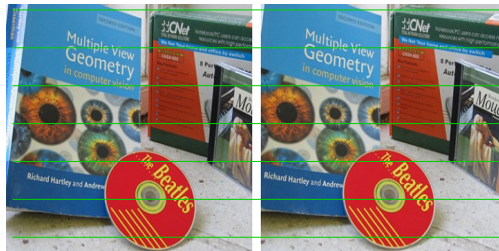
# Geometry of rectification



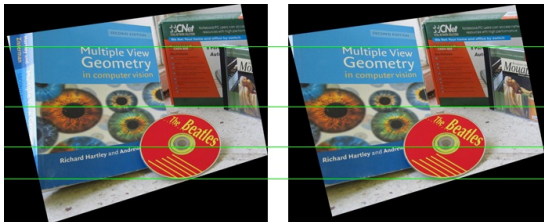
# Rectification: a video video

Epipoles transformed to infinity

# Rectification: an example



before



after



# Benefits of rectifications

- Modify the image in order to get a standard stereo,  
→ then algorithms for standard stereo can be applied.
- The properties of epipolar geometry can be visualized by rectifying the images.
- For practical purposes, the rectification has to be very accurate
  - otherwise there will be a shift between corresponding rows.
  - feature matching more challenging, 1D cannot be run.

# Weak points of rectification

- Distortion under rectification hardly depends on baseline width.
- For wide-baseline stereo:
  - Rectification significantly distorts the image.
    - Pixel-based method can be applied for feature matching
    - Correspondence-based methods often fail.
- Size and shape of rectified images differ from original ones.
  - Feature matching is more challenging.
- Many experts do not agree that rectification is necessary.
  - Epipolar lines can be followed if fundamental matrix is given.
  - Matching can be carried out in original frames.
    - Then noise is not distorted by rectifying transformation.

# Outline

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  - Estimation of the fundamental matrix
- 3 Standard stereo and rectification
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  - Triangulation and metric reconstruction
  - Projective reconstruction
  - Planar Motion
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# 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

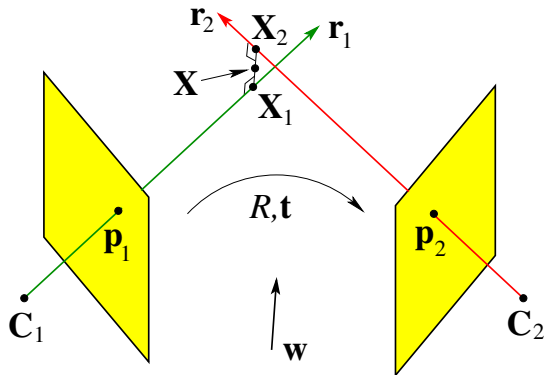
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# Triangulation

## ● Task:

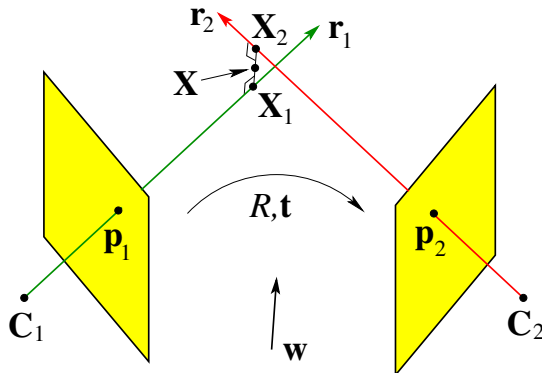
- Two calibrated cameras are given, including both intrinsic and extrinsic parameters, and
  - Locations  $\mathbf{u}_1, \mathbf{u}_2$  of the projection of spatial point  $\mathbf{X}$  are given
  - Goal is to estimate spatial location  $\mathbf{X}$ .
- Two calibration matrices are known, therefore
    - 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



- $\alpha \mathbf{p}_1$  is a point on ray  $\mathbf{r}_1$  ( $\alpha \in \mathbb{R}$ )
- $\mathbf{t} + \beta \mathbf{R}^T \mathbf{p}_2$  a point on other ray  $\mathbf{r}_2$  ( $\beta \in \mathbb{R}$ )  
 → coordinate system fixed to the first camera
- Let  $\mathbf{X}_1 = \alpha_0 \mathbf{p}_1$ ,  $\mathbf{X}_2 = \mathbf{t} + \mathbf{R}^T (\beta_0 \mathbf{p}_2 - \mathbf{t})$



# Linear triangulation: solution

- Task is to determine

- the middle point of the line section  $\mathbf{X}_1\mathbf{X}_2$
- determination of  $\alpha_0$  and  $\beta_0$  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  $\alpha\mathbf{p}_1$  ( $\gamma \in \mathbb{R}$ ).

→  $\alpha_0, \beta_0$  (as well as  $\gamma_0$ ) 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:

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

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

- $\lambda_1$  and  $\lambda_2$  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:

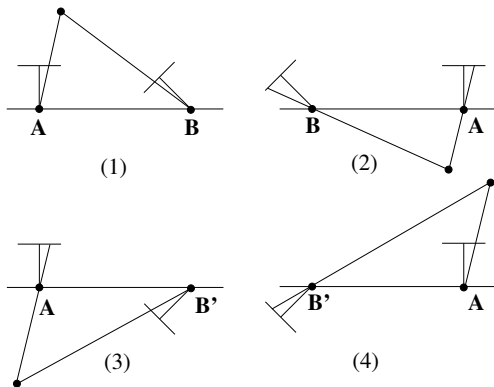
$$\begin{aligned}\mathbf{R}_1 &= \mathbf{UWV}^T & \mathbf{R}_2 &= \mathbf{UW}^T\mathbf{V}^T \\ [\mathbf{t}_1]_{\times} &= \delta\mathbf{UZU}^T & [\mathbf{t}_2]_{\times} &= -\delta\mathbf{UZU}^T\end{aligned}$$

- 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.