

Reconstruction

Reconstruction

- The forward process:
 - Given
 - 3D shapes
 - Their locations+orientations
 - Their material+paint
 - Light directions+intensity
 - The Camera parameters
 - Produce the image
- Reconstruction: *Reverse* this process
- Next two/three classes: reconstructing geometry

Final perspective projection


Camera extrinsics: where your camera is relative to the world. Changes if you move the camera

$$\vec{\mathbf{x}}_{img} \equiv K [R \quad \mathbf{t}] \vec{\mathbf{x}}_w$$

Camera intrinsics: how your camera handles pixel. Changes if you change your camera

$$\vec{\mathbf{x}}_{img} \equiv P \vec{\mathbf{x}}_w$$

Final perspective projection

$$\vec{\mathbf{x}}_{img} \equiv K \begin{bmatrix} R & \mathbf{t} \end{bmatrix} \vec{\mathbf{x}}_w$$


Camera parameters

$$\vec{\mathbf{x}}_{img} \equiv P \vec{\mathbf{x}}_w$$

Camera calibration

- Goal: find the parameters of the camera

$$\vec{\mathbf{x}}_{img} \equiv K \begin{bmatrix} R & \mathbf{t} \end{bmatrix} \vec{\mathbf{x}}_w$$

- Why?
 - Tells you where the camera is relative to the world/particular objects
 - Equivalently, tells you where objects are relative to the camera
 - Can allow you to "render" new objects into the scene

Camera calibration

$$\vec{\mathbf{x}}_{img} \equiv P\vec{\mathbf{x}}_w$$

- Need to estimate P
- How many parameters does P have?
 - Size of P : 3 x 4
 - But: $\lambda P\vec{\mathbf{x}}_w \equiv P\vec{\mathbf{x}}_w$
 - P can only be known *upto a scale*
 - $3*4 - 1 = 11$ parameters

Camera calibration

$$\vec{\mathbf{x}}_{img} \equiv P \vec{\mathbf{x}}_w$$

- Suppose we know that (X,Y,Z) in the world projects to (x,y) in the image.
- How many equations does this provide?

$$\begin{bmatrix} x \\ y \\ 1 \end{bmatrix} \equiv P \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

Need to convert equivalence into equality.

Camera calibration

$$\vec{\mathbf{x}}_{img} \equiv P\vec{\mathbf{x}}_w$$

- Suppose we know that (X,Y,Z) in the world projects to (x,y) in the image.
- How many equations does this provide?

Note: λ is unknown

$$\begin{bmatrix} \lambda x \\ \lambda y \\ \lambda \end{bmatrix} = P \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

Camera calibration

$$\vec{\mathbf{x}}_{img} \equiv P \vec{\mathbf{x}}_w$$

- Suppose we know that (X,Y,Z) in the world projects to (x,y) in the image.
- How many equations does this provide?

$$\begin{bmatrix} \lambda x \\ \lambda y \\ \lambda \end{bmatrix} = \begin{bmatrix} P_{11} & P_{12} & P_{13} & P_{14} \\ P_{21} & P_{22} & P_{23} & P_{24} \\ P_{31} & P_{32} & P_{33} & P_{34} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

Camera calibration

$$\vec{\mathbf{x}}_{img} \equiv P\vec{\mathbf{x}}_w$$

- Suppose we know that (X,Y,Z) in the world projects to (x,y) in the image.
- How many equations does this provide?

$$\lambda x = P_{11}X + P_{12}Y + P_{13}Z + P_{14}$$

$$\lambda y = P_{21}X + P_{22}Y + P_{23}Z + P_{24}$$

$$\lambda = P_{31}X + P_{32}Y + P_{33}Z + P_{34}$$

Camera calibration

$$\vec{\mathbf{x}}_{img} \equiv P\vec{\mathbf{x}}_w$$

- Suppose we know that (X,Y,Z) in the world projects to (x,y) in the image.
- How many equations does this provide?

$$(P_{31}X + P_{32}Y + P_{33}Z + P_{34})x = P_{11}X + P_{12}Y + P_{13}Z + P_{14}$$

$$(P_{31}X + P_{32}Y + P_{33}Z + P_{34})y = P_{21}X + P_{22}Y + P_{23}Z + P_{24}$$

- 2 equations!
- Are the equations linear in the parameters?
- How many equations do we need?

Camera calibration

$$(P_{31}X + P_{32}Y + P_{33}Z + P_{34})x = P_{11}X + P_{12}Y + P_{13}Z + P_{14}$$

$$XxP_{31} + YxP_{32} + ZxP_{33} + xP_{34} - XP_{11} - YP_{12} - ZP_{13} - P_{14} = 0$$

- In matrix vector form: $\mathbf{A}p = 0$
- 6 points give 12 equations, 12 variables to solve for
- But can only solve upto scale

Camera calibration

- In matrix vector form: $\mathbf{A}\mathbf{p} = 0$
- We want non-trivial solutions
- If \mathbf{p} is a solution, $\alpha\mathbf{p}$ is a solution too
- Let's just search for a solution with unit norm

$$\mathbf{A}\mathbf{p} = 0$$

s.t

$$\|\mathbf{p}\| = 1$$

Camera calibration

- In matrix vector form: $\mathbf{A}\mathbf{p} = 0$
- We want non-trivial solutions
- If \mathbf{p} is a solution, $\alpha\mathbf{p}$ is a solution too
- Alternative form to deal with noisy inputs

$$\min_{\mathbf{p}} \|\mathbf{A}\mathbf{p}\|^2 \equiv \min_{\mathbf{p}} \mathbf{p}^T \mathbf{A}^T \mathbf{A} \mathbf{p}$$

s.t

$$\|\mathbf{p}\| = 1$$

- How do you solve this?
 - *Eigenvector of $\mathbf{A}^T\mathbf{A}$ with smallest eigenvalue!*

Camera calibration

- We need 6 world points for which we know image locations
- Would any 6 points work?
 - What if all 6 points are the same?
- Need at least 6 non-coplanar points!

Camera calibration

$$\vec{\mathbf{x}}_{img} \equiv P \vec{\mathbf{x}}_w$$

$$\vec{\mathbf{x}}_{img} \equiv K \begin{bmatrix} R & \mathbf{t} \end{bmatrix} \vec{\mathbf{x}}_w$$

- How do we get K, R and t from P?
- Need to make some assumptions about K
- What if K is upper triangular?

$$K = \begin{bmatrix} s_x & \alpha & t_u \\ 0 & s_y & t_v \\ 0 & 0 & 1 \end{bmatrix}$$

Added skew if image x and y axes are not perpendicular

Camera calibration

- How do we get K , R and t from P ?
- Need to make some assumptions about K
- What if K is upper triangular?

$$K = \begin{bmatrix} s_x & \alpha & t_u \\ 0 & s_y & t_v \\ 0 & 0 & 1 \end{bmatrix}$$

- $P = K [R \ t]$
- First 3 x 3 matrix of P is KR
- “RQ” decomposition: decomposes an $n \times n$ matrix into product of upper triangular and rotation matrix

Camera calibration

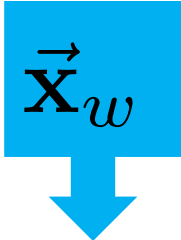
- How do we get K , R and t from P ?
- Need to make some assumptions about K
- What if K is upper triangular?
- $P = K [R \ t]$
- First 3×3 matrix of P is KR
- “RQ” decomposition: decomposes an $n \times n$ matrix into product of upper triangular and rotation matrix
- $t = K^{-1}P[:,2] \leftarrow$ last column of P

Camera calibration and pose estimation



- Where camera is relative to car equivalent to where car is relative to camera

Reconstructing world points given camera

$$\vec{\mathbf{x}}_{img} \equiv K \begin{bmatrix} R & \mathbf{t} \end{bmatrix} \vec{\mathbf{x}}_w$$


Can we recover this
from just a single
equation?

$$\vec{\mathbf{x}}_{img} \equiv P \vec{\mathbf{x}}_w$$

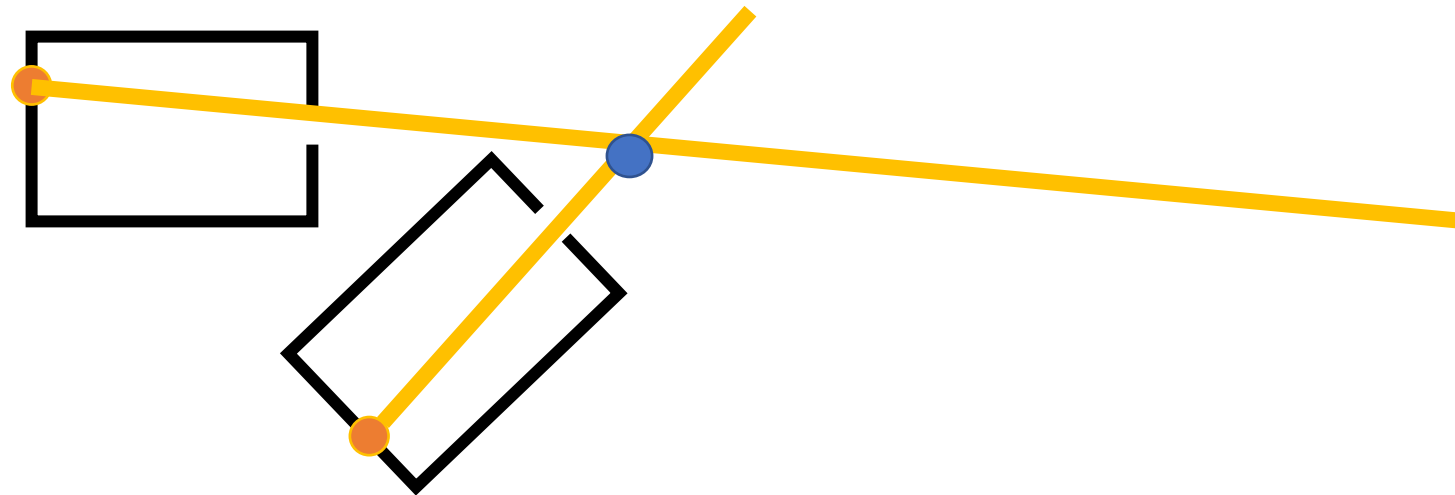
Ambiguity

- A pixel corresponds to an entire ray
 - 2 linear equations in 3D space
- Need additional constraints!



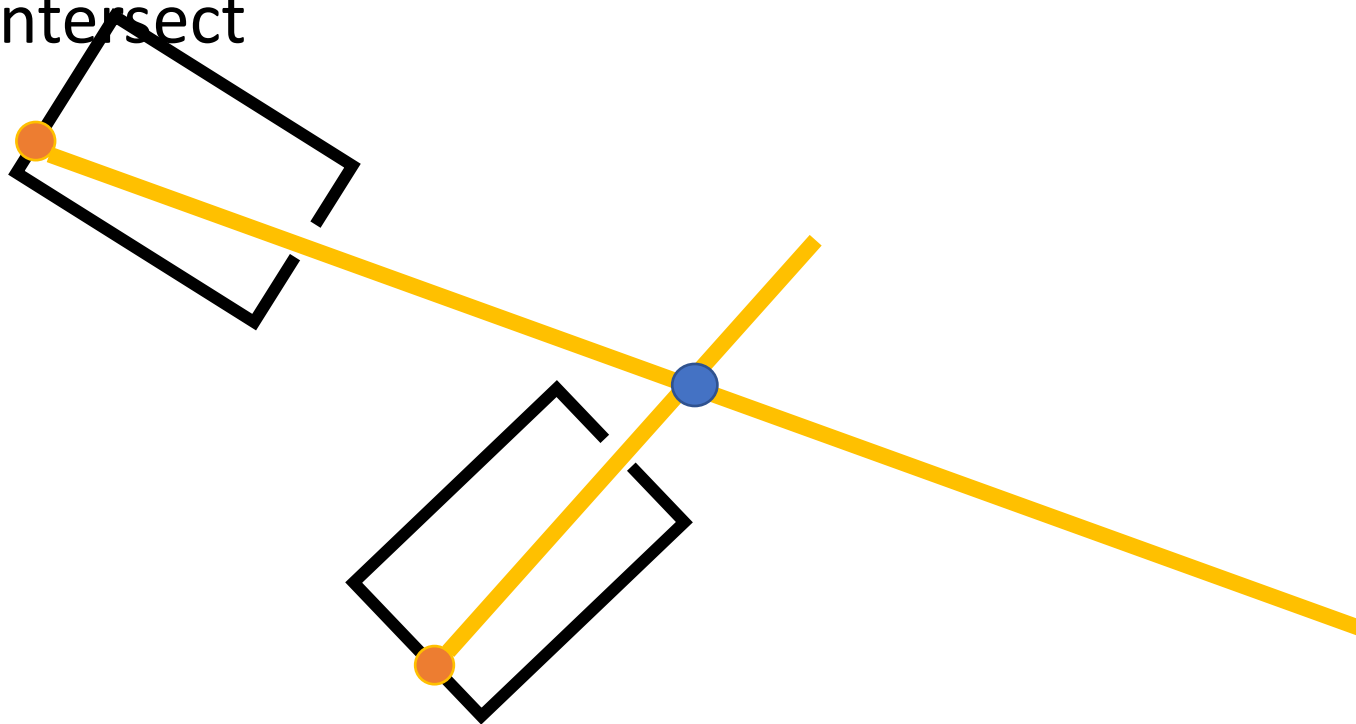
Triangulation

- A pixel corresponds to an entire ray
 - 2 linear equations in 3D space
- If we have corresponding pixel from another view, can intersect rays!
 - 4 equations



Binocular stereo

- Single
- If we know where cameras are, we can shoot rays from corresponding pixels and intersect



Triangulation

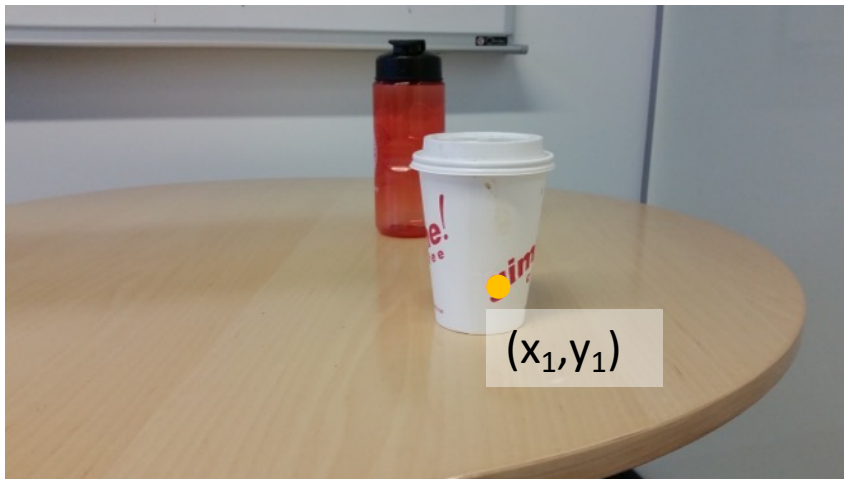
- Suppose we have two cameras
 - Calibrated: parameters known
- And a pair of corresponding pixels
- Find 3D location of point!



Triangulation

- Suppose we have two cameras
 - Calibrated: parameters known
- And a pair of corresponding pixels
- Find 3D location of point!

$P^{(1)}$



$P^{(2)}$



Triangulation

$$\begin{array}{c} \begin{bmatrix} x_1 \\ y_1 \\ 1 \end{bmatrix} \\ \\ \begin{bmatrix} x_2 \\ y_2 \\ 1 \end{bmatrix} \end{array} \leftarrow \begin{array}{c} \vec{\mathbf{x}}_{img}^{(1)} \\ \\ \vec{\mathbf{x}}_{img}^{(2)} \end{array} \equiv \begin{array}{c} P^{(1)} \vec{\mathbf{x}}_w \\ \\ P^{(2)} \vec{\mathbf{x}}_w \end{array} \rightarrow \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

The diagram illustrates the triangulation process. On the left, two image points are shown as column vectors: $\begin{bmatrix} x_1 \\ y_1 \\ 1 \end{bmatrix}$ and $\begin{bmatrix} x_2 \\ y_2 \\ 1 \end{bmatrix}$. Arrows point from these image points to a central set of equations. The first equation is $\vec{\mathbf{x}}_{img}^{(1)} \equiv P^{(1)} \vec{\mathbf{x}}_w$ and the second is $\vec{\mathbf{x}}_{img}^{(2)} \equiv P^{(2)} \vec{\mathbf{x}}_w$. From the right-hand side of these equations, two arrows point to a single column vector representing the world point: $\begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$.

Triangulation

$$\vec{\mathbf{X}}_{img}^{(1)} \equiv P^{(1)} \vec{\mathbf{X}}_w$$

$$\lambda x_1 = P_{11}^{(1)} X + P_{12}^{(1)} Y + P_{13}^{(1)} Z + P_{14}^{(1)}$$

$$\lambda y_1 = P_{21}^{(1)} X + P_{22}^{(1)} Y + P_{23}^{(1)} Z + P_{24}^{(1)}$$

$$\lambda = P_{31}^{(1)} X + P_{32}^{(1)} Y + P_{33}^{(1)} Z + P_{34}^{(1)}$$

$$(P_{31}^{(1)} X + P_{32}^{(1)} Y + P_{33}^{(1)} Z + P_{34}^{(1)}) x_1 = P_{11}^{(1)} X + P_{12}^{(1)} Y + P_{13}^{(1)} Z + P_{14}^{(1)}$$
$$X(P_{31}^{(1)} x_1 - P_{11}^{(1)}) + Y(P_{32}^{(1)} x_1 - P_{12}^{(1)}) + Z(P_{33}^{(1)} x_1 - P_{13}^{(1)}) + (P_{34}^{(1)} x_1 - P_{14}^{(1)}) = 0$$

Triangulation

$$\vec{\mathbf{x}}_{img}^{(1)} \equiv P^{(1)} \vec{\mathbf{x}}_w$$

$$X(P_{31}^{(1)}x_1 - P_{11}^{(1)}) + Y(P_{32}^{(1)}x_1 - P_{12}^{(1)}) + Z(P_{33}^{(1)}x_1 - P_{13}^{(1)}) + (P_{34}^{(1)}x_1 - P_{14}^{(1)}) = 0$$

$$X(P_{31}^{(1)}y_1 - P_{21}^{(1)}) + Y(P_{32}^{(1)}y_1 - P_{22}^{(1)}) + Z(P_{33}^{(1)}y_1 - P_{23}^{(1)}) + (P_{34}^{(1)}y_1 - P_{24}^{(1)}) = 0$$

- 1 image gives 2 equations
- Need 2 images!
- Solve linear equations to get 3D point location

Linear vs non-linear optimization

$$\lambda x_1 = P_{11}^{(1)} X + P_{12}^{(1)} Y + P_{13}^{(1)} Z + P_{14}^{(1)}$$

$$\lambda y_1 = P_{21}^{(1)} X + P_{22}^{(1)} Y + P_{23}^{(1)} Z + P_{24}^{(1)}$$

$$\lambda = P_{31}^{(1)} X + P_{32}^{(1)} Y + P_{33}^{(1)} Z + P_{34}^{(1)}$$

$$x_1 = \frac{P_{11}^{(1)} X + P_{12}^{(1)} Y + P_{13}^{(1)} Z + P_{14}^{(1)}}{P_{31}^{(1)} X + P_{32}^{(1)} Y + P_{33}^{(1)} Z + P_{34}^{(1)}}$$

$$y_1 = \frac{P_{21}^{(1)} X + P_{22}^{(1)} Y + P_{23}^{(1)} Z + P_{24}^{(1)}}{P_{31}^{(1)} X + P_{32}^{(1)} Y + P_{33}^{(1)} Z + P_{34}^{(1)}}$$

Linear vs non-linear optimization

$$x_1 = \frac{P_{11}^{(1)} X + P_{12}^{(1)} Y + P_{13}^{(1)} Z + P_{14}^{(1)}}{P_{31}^{(1)} X + P_{32}^{(1)} Y + P_{33}^{(1)} Z + P_{34}^{(1)}}$$

$$y_1 = \frac{P_{21}^{(1)} X + P_{22}^{(1)} Y + P_{23}^{(1)} Z + P_{24}^{(1)}}{P_{31}^{(1)} X + P_{32}^{(1)} Y + P_{33}^{(1)} Z + P_{34}^{(1)}}$$

$$\left(x_1 - \frac{P_{11}^{(1)} X + P_{12}^{(1)} Y + P_{13}^{(1)} Z + P_{14}^{(1)}}{P_{31}^{(1)} X + P_{32}^{(1)} Y + P_{33}^{(1)} Z + P_{34}^{(1)}}\right)^2 + \left(y_1 - \frac{P_{21}^{(1)} X + P_{22}^{(1)} Y + P_{23}^{(1)} Z + P_{24}^{(1)}}{P_{31}^{(1)} X + P_{32}^{(1)} Y + P_{33}^{(1)} Z + P_{34}^{(1)}}\right)^2$$

Reprojection error

Linear vs non-linear optimization

$$\left(x_1 - \frac{P_{11}^{(1)} X + P_{12}^{(1)} Y + P_{13}^{(1)} Z + P_{14}^{(1)}}{P_{31}^{(1)} X + P_{32}^{(1)} Y + P_{33}^{(1)} Z + P_{34}^{(1)}}\right)^2$$
$$+ \left(y_1 - \frac{P_{21}^{(1)} X + P_{22}^{(1)} Y + P_{23}^{(1)} Z + P_{24}^{(1)}}{P_{31}^{(1)} X + P_{32}^{(1)} Y + P_{33}^{(1)} Z + P_{34}^{(1)}}\right)^2$$

Reprojection error

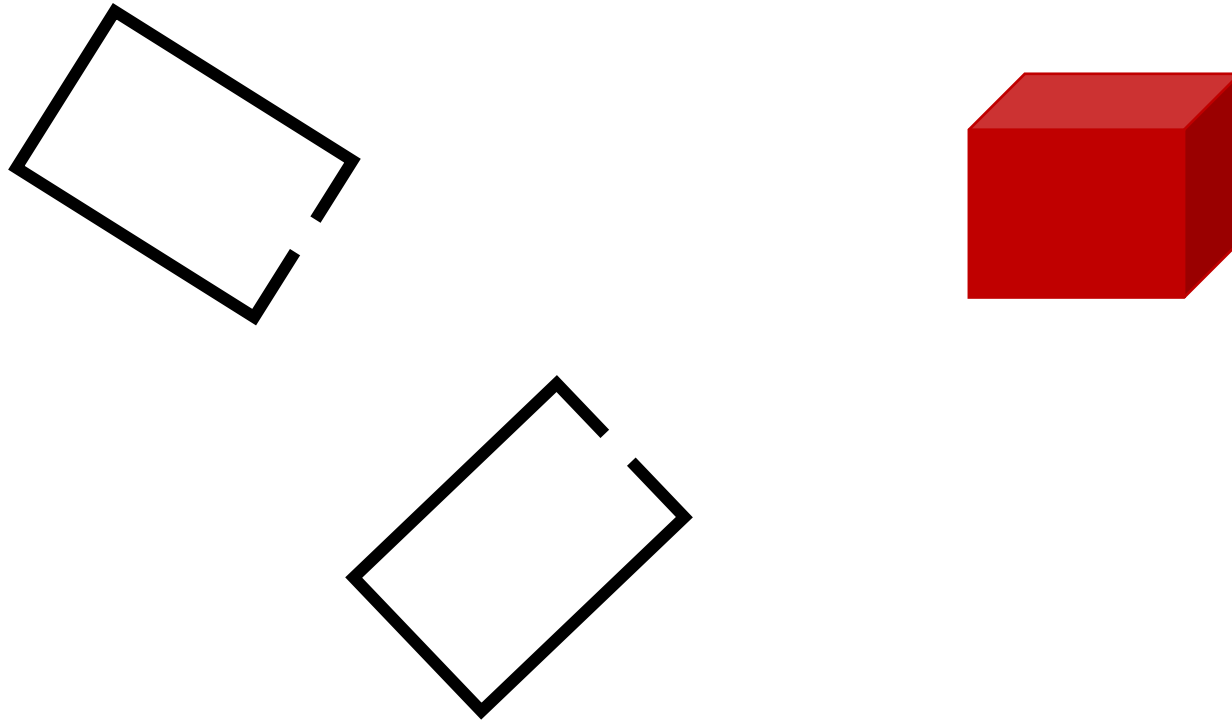
- Reprojection error is the squared error between the true image coordinates of a point and the projected coordinates of hypothesized 3D point
- Actual error we care about
- Minimize total sum of reprojection error across all images
- *Non-linear optimization*

Binocular stereo

- Given two *calibrated* cameras
 - Find pairs of corresponding pixels
 - Use corresponding image locations to set up equations on world coordinates
 - Solve!

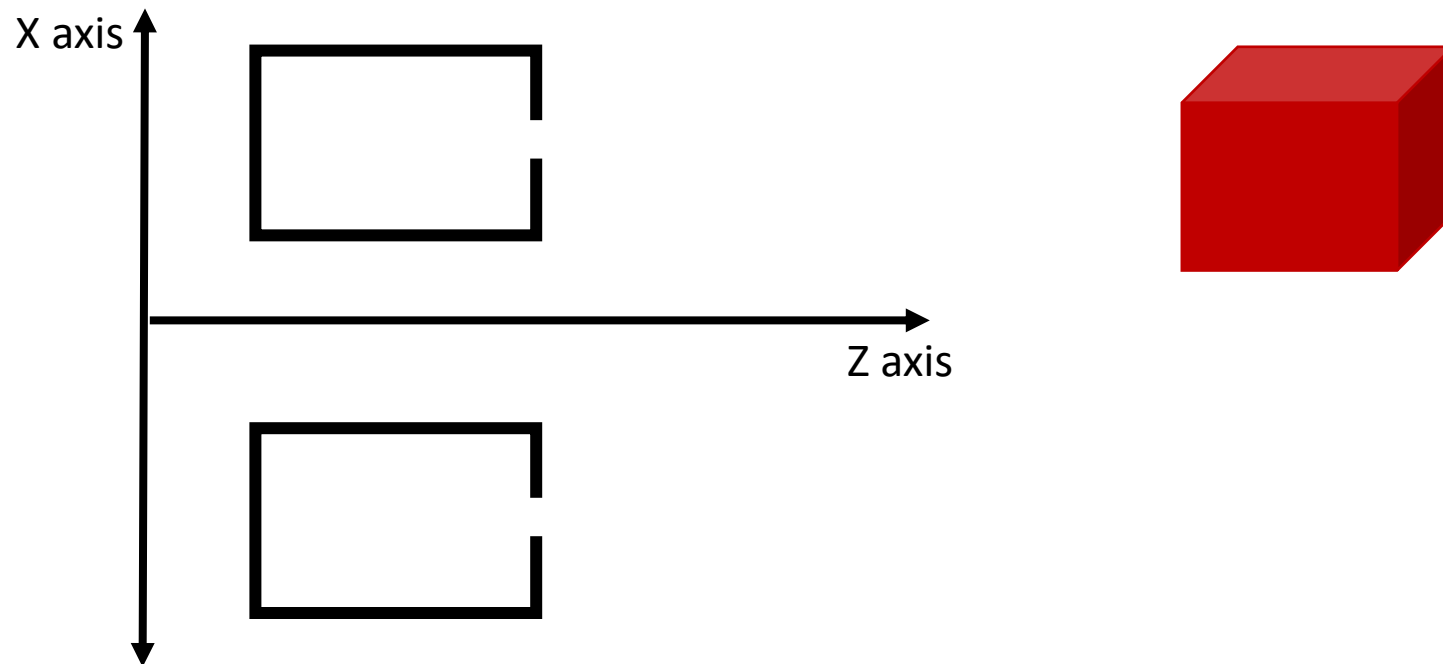
Binocular stereo

- General case: cameras can be arbitrary locations and orientations



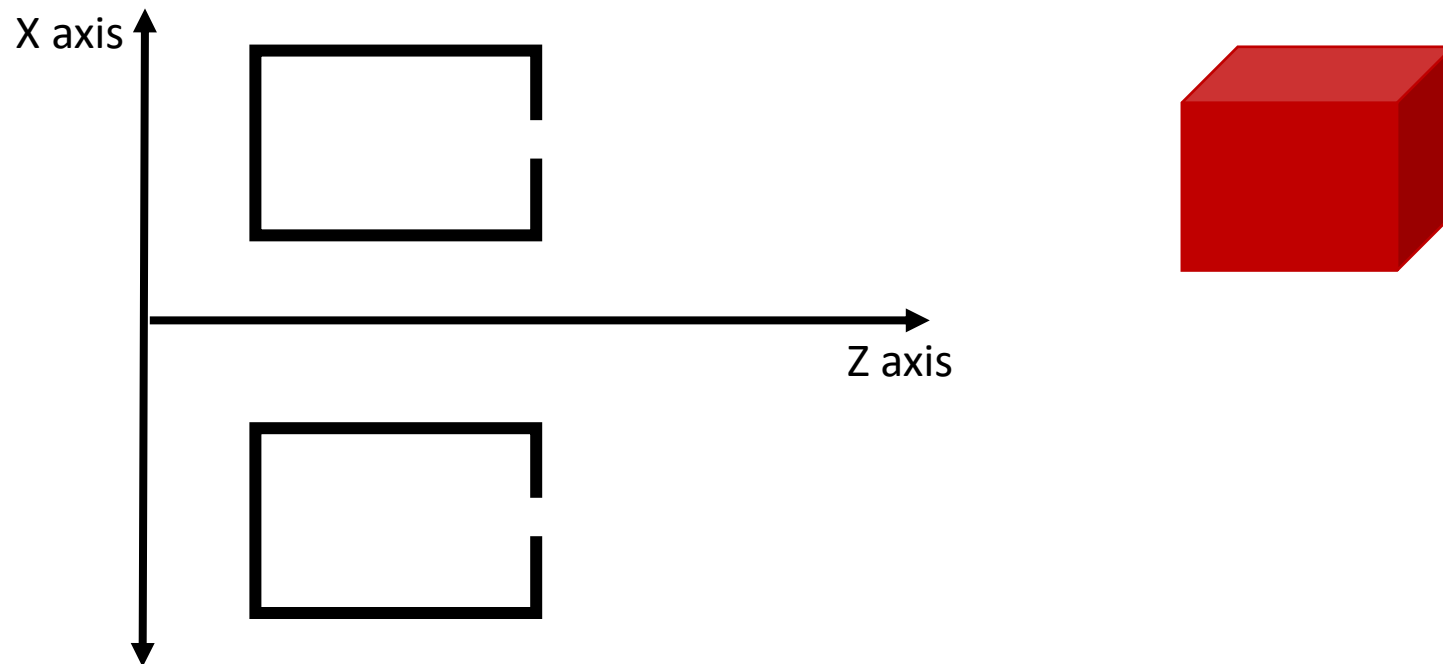
Binocular stereo

- Special case: cameras are parallel to each other and translated along X axis

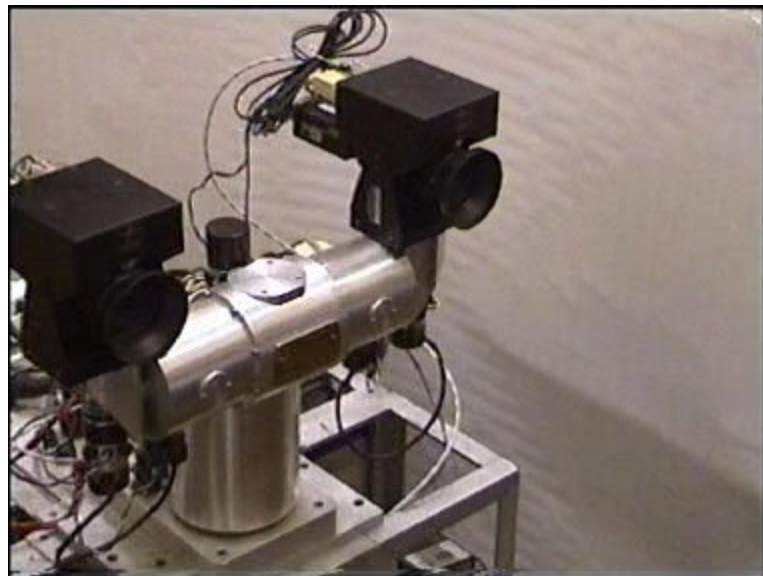


Stereo with *rectified* cameras

- Special case: cameras are parallel to each other and translated along X axis



Stereo head



Kinect / depth cameras



Stereo with “rectified cameras”



Perspective projection in rectified cameras

- Without loss of generality, assume origin is at pinhole of 1st camera

$$\vec{\mathbf{x}}_{img}^{(1)} \equiv \begin{bmatrix} I & \mathbf{0} \end{bmatrix} \vec{\mathbf{x}}_w$$

$$\vec{\mathbf{x}}_{img}^{(2)} \equiv \begin{bmatrix} I & \mathbf{t} \end{bmatrix} \vec{\mathbf{x}}_w$$

$$\mathbf{t} = \begin{bmatrix} t_x \\ 0 \\ 0 \end{bmatrix}$$

Perspective projection in rectified cameras

- Without loss of generality, assume origin is at pinhole of 1st camera

$$\vec{\mathbf{x}}_{img}^{(1)} \equiv \begin{bmatrix} I & \mathbf{0} \end{bmatrix} \vec{\mathbf{x}}_w$$

$$\vec{\mathbf{x}}_{img}^{(2)} \equiv \begin{bmatrix} I & \mathbf{t} \end{bmatrix} \vec{\mathbf{x}}_w$$

$$\mathbf{t} = \begin{bmatrix} t_x \\ 0 \\ 0 \end{bmatrix} \quad \vec{\mathbf{x}}_w = \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix} = \begin{bmatrix} \mathbf{x}_w \\ 1 \end{bmatrix}$$

Perspective projection in rectified cameras

- Without loss of generality, assume origin is at pinhole of 1st camera

$$\vec{\mathbf{x}}_{img}^{(1)} \equiv [I \quad \mathbf{0}] \begin{bmatrix} \mathbf{x}_w \\ 1 \end{bmatrix} = \mathbf{x}_w = \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$

$$\vec{\mathbf{x}}_{img}^{(2)} \equiv [I \quad \mathbf{t}] \begin{bmatrix} \mathbf{x}_w \\ 1 \end{bmatrix} = \mathbf{x}_w + \mathbf{t} = \begin{bmatrix} X + t_x \\ Y \\ Z \end{bmatrix}$$

Perspective projection in rectified cameras

- Without loss of generality, assume origin is at pinhole of 1st camera

$$\vec{\mathbf{x}}_{img}^{(1)} \equiv \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$

$$\vec{\mathbf{x}}_{img}^{(2)} \equiv \begin{bmatrix} X + t_x \\ Y \\ Z \end{bmatrix}$$

Perspective projection in rectified cameras

- Without loss of generality, assume origin is at pinhole of 1st camera

$$\begin{bmatrix} x_1 \\ y_1 \\ 1 \end{bmatrix} \equiv \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$

$$\begin{bmatrix} x_2 \\ y_2 \\ 1 \end{bmatrix} \equiv \begin{bmatrix} X + t_x \\ Y \\ Z \end{bmatrix}$$

Perspective projection in rectified cameras

- Without loss of generality, assume origin is at pinhole of 1st camera

$$\begin{bmatrix} x_1 \\ y_1 \\ 1 \end{bmatrix} \equiv \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$

$$\begin{bmatrix} x_2 \\ y_2 \\ 1 \end{bmatrix} \equiv \begin{bmatrix} X + t_x \\ Y \\ Z \end{bmatrix}$$

Perspective projection in rectified cameras

- Without loss of generality, assume origin is at pinhole of 1st camera

$$\begin{bmatrix} \lambda x_1 \\ \lambda y_1 \\ \lambda \end{bmatrix} = \begin{bmatrix} X \\ Y \\ Z \end{bmatrix}$$

$$\begin{bmatrix} \lambda x_2 \\ \lambda y_2 \\ \lambda \end{bmatrix} = \begin{bmatrix} X + t_x \\ Y \\ Z \end{bmatrix}$$

Perspective projection in rectified cameras

- Without loss of generality, assume origin is at pinhole of 1st camera

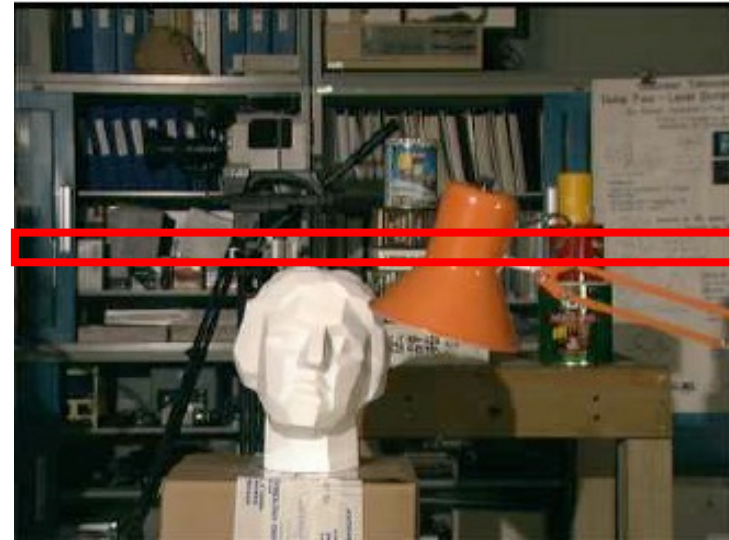
X coordinate differs by t_x/Z

$$x_1 = \frac{X}{Z} \qquad x_2 = \frac{X + t_x}{Z}$$

$$y_1 = \frac{Y}{Z} \qquad y_2 = \frac{Y}{Z}$$

Y coordinate is the same!

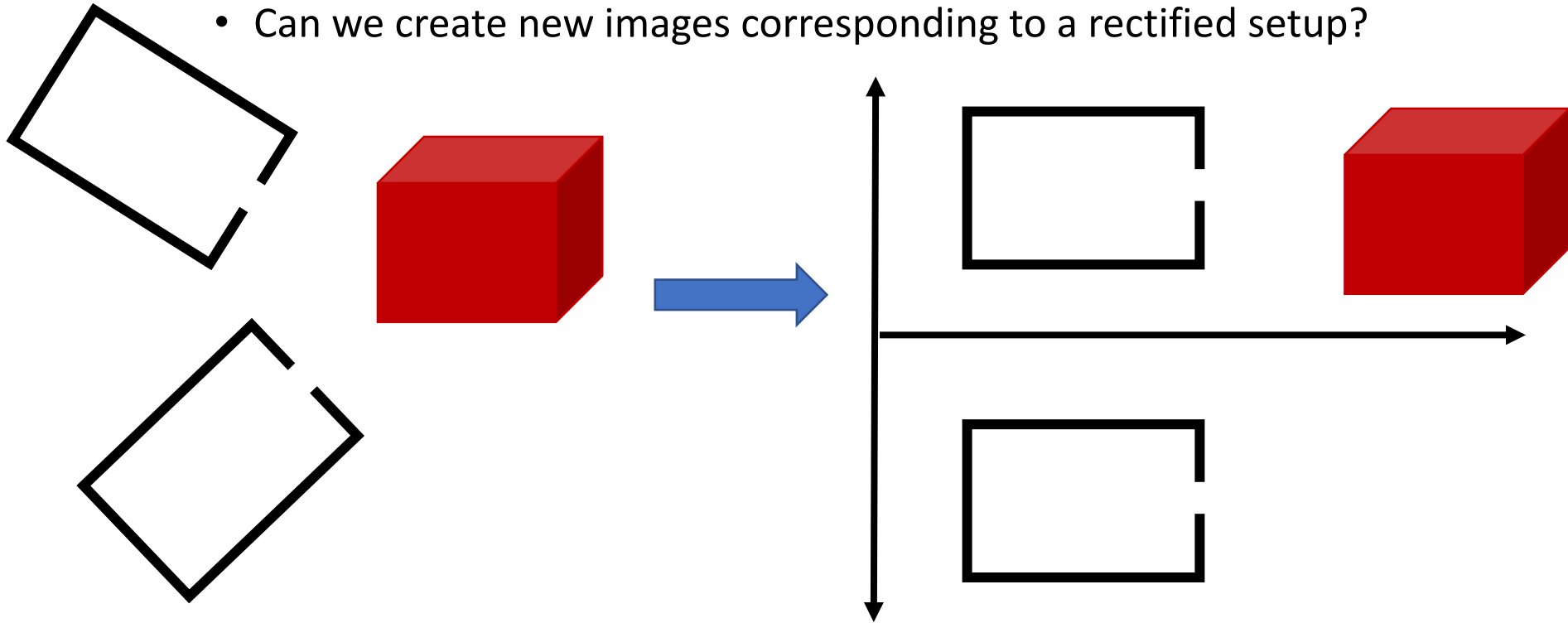
Perspective projection in rectified cameras



- For rectified cameras, correspondence problem is easier
- Only requires searching along a particular *row*.

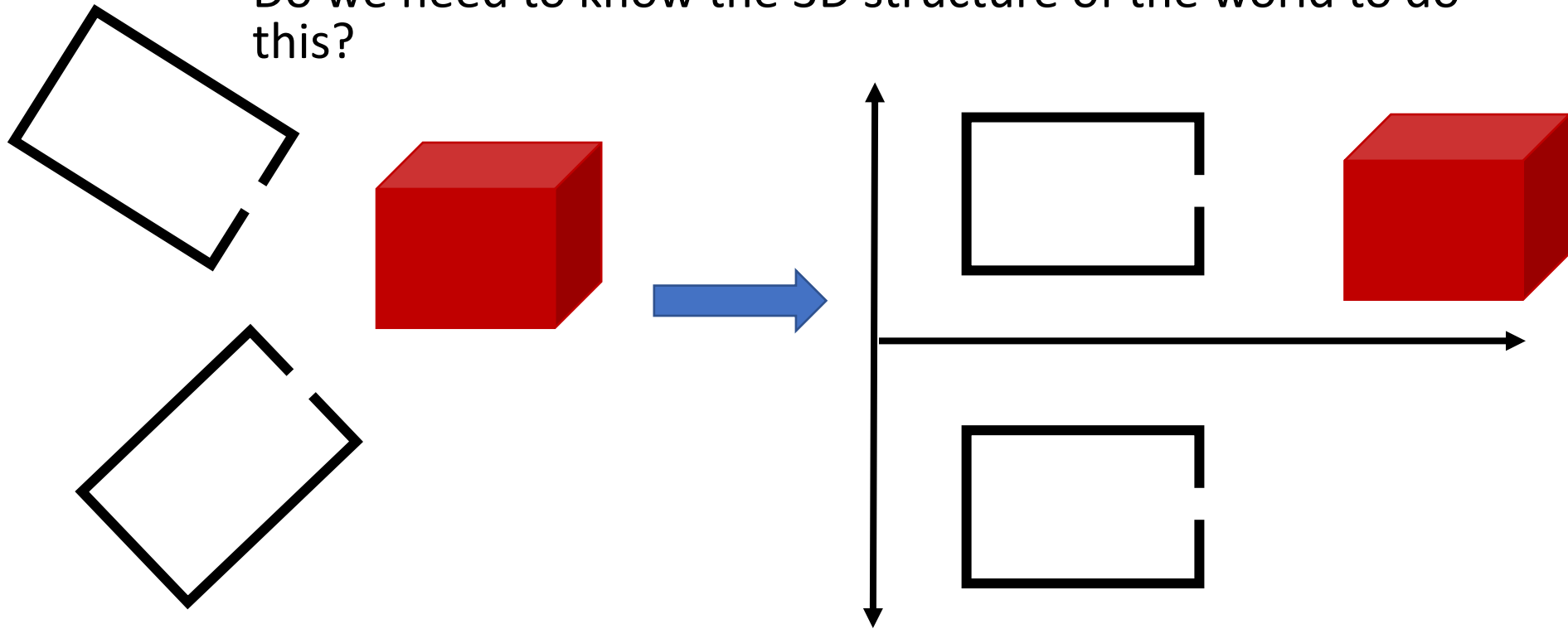
Rectifying cameras

- Given two images from two cameras with known P , can we rectify them?
 - Can we create new images corresponding to a rectified setup?



Rectifying cameras

- Can we rotate / translate cameras?
 - Do we need to know the 3D structure of the world to do this?



Rotating cameras

$$\vec{\mathbf{x}}_{img} \equiv K \begin{bmatrix} R & \mathbf{t} \end{bmatrix} \vec{\mathbf{x}}_w$$

- Assume K is identity
- Assume coordinate system at camera pinhole

$$\begin{aligned} \vec{\mathbf{x}}_{img} &\equiv \begin{bmatrix} I & \mathbf{0} \end{bmatrix} \vec{\mathbf{x}}_w \\ &\equiv \begin{bmatrix} I & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbf{x}_w \\ 1 \end{bmatrix} \\ &\equiv \mathbf{x}_w \end{aligned}$$

Rotating cameras

$$\vec{\mathbf{x}}_{img} \equiv K \begin{bmatrix} R & \mathbf{t} \end{bmatrix} \vec{\mathbf{x}}_w$$

- Assume K is identity
- Assume coordinate system at camera pinhole

$$\begin{aligned} \vec{\mathbf{x}}_{img} &\equiv \begin{bmatrix} I & \mathbf{0} \end{bmatrix} \vec{\mathbf{x}}_w \\ &\equiv \begin{bmatrix} I & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbf{x}_w \\ 1 \end{bmatrix} \\ &\equiv \mathbf{x}_w \end{aligned}$$

Rotating cameras

$$\vec{\mathbf{x}}_{img} \equiv \begin{bmatrix} I & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbf{x}_w \\ 1 \end{bmatrix}$$

$$\vec{\mathbf{x}}_{img} \equiv \mathbf{x}_w$$

- What happens if the camera is rotated?

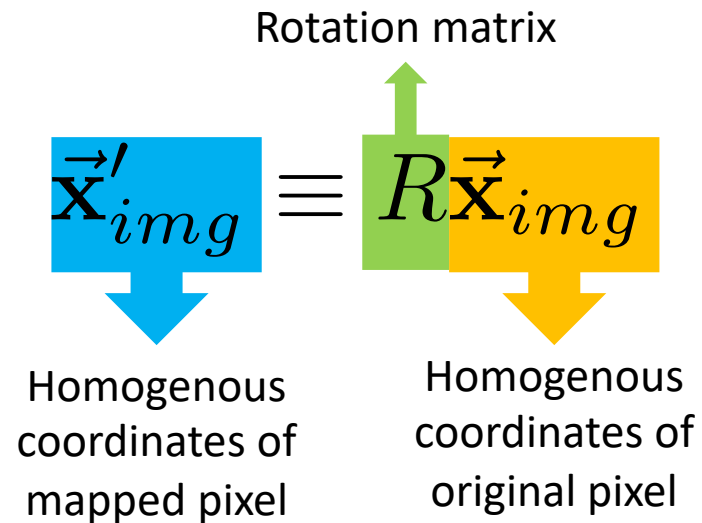
$$\vec{\mathbf{x}}'_{img} \equiv \begin{bmatrix} R & \mathbf{0} \end{bmatrix} \begin{bmatrix} \mathbf{x}_w \\ 1 \end{bmatrix}$$

$$\equiv R\mathbf{x}_w$$

$$\equiv R\vec{\mathbf{x}}_{img}$$

Rotating cameras

- What happens if the camera is rotated?

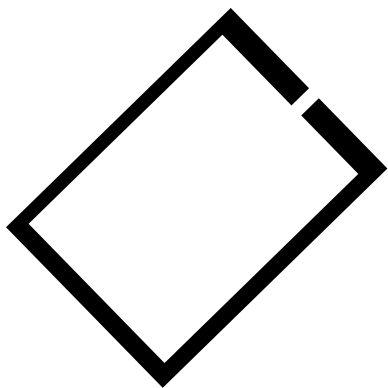
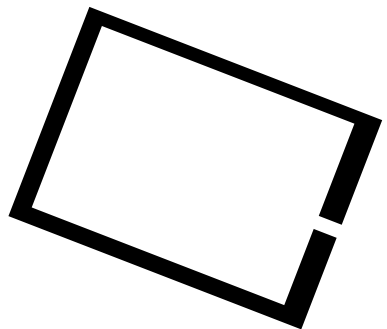


- No need to know the 3D structure

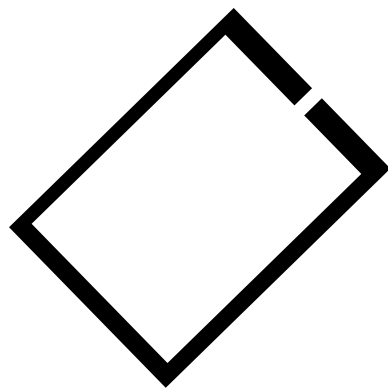
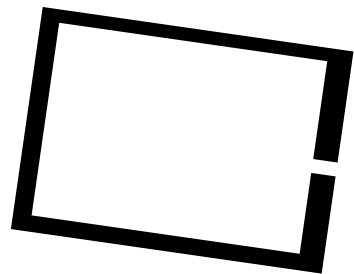
Rotating cameras



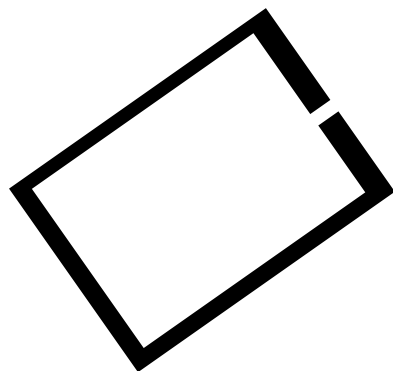
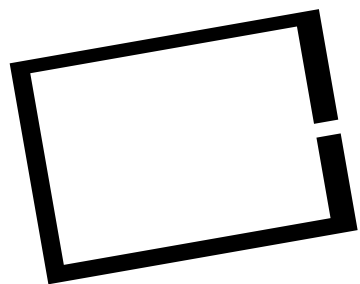
Rectifying cameras



Rectifying cameras



Rectifying cameras



Rectifying cameras

