



University of Cyprus – MSc Artificial Intelligence

# MAI644 – COMPUTER VISION

## Lecture 16: Stereo Vision

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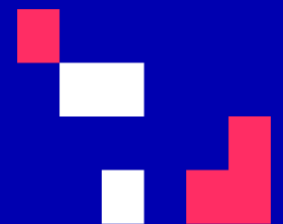
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# Last time

- Full camera model in matrix form
- Camera calibration
- Calibration – Projective camera model
- Calibration – Affine camera model

# Today's Agenda

- Recovery of world position
- Triangulation
- Epipolar Geometry

[material based on Paris Kaimakis]



# Today's Agenda

- Recovery of world position
- Triangulation
- Epipolar Geometry



# Recovery of world position

- Previously we saw that the imaging process can be described as a transformation in homogeneous coordinates.
- If we can invert this transformation, the world coordinates of each pixel in the image can be computed.
- Recovering world coordinates of objects based on the projection on an image is known as **shape recovery** or **depth recovery**.

# Recovery of world position

- Is this possible using a single camera ?
- The camera needs to be calibrated, i.e. we know all its parameters
- Remember the projective camera model:

$$\tilde{\mathbf{w}} = \mathbf{P}\tilde{\mathbf{X}}$$
$$\Leftrightarrow \begin{bmatrix} su \\ sv \\ s \end{bmatrix} = \begin{bmatrix} p_{11} & p_{12} & p_{13} & p_{14} \\ p_{21} & p_{22} & p_{23} & p_{24} \\ p_{31} & p_{32} & p_{33} & p_{34} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

- Unfortunately the transformation described in  $\mathbf{P}$  is not invertible and the world point  $\mathbf{X}$  cannot be uniquely determined

# Recovery of world position

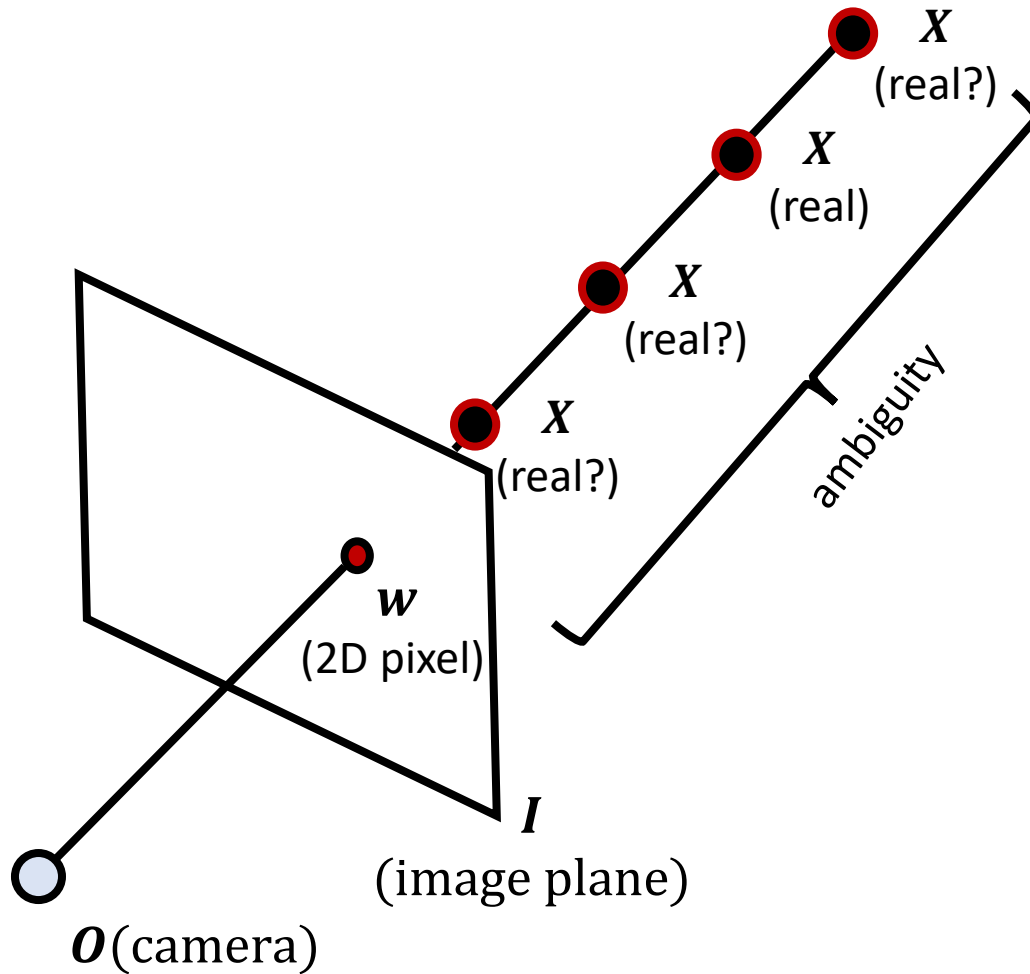
- Depth ambiguity



Courtesy slide S. Lazebnik

# Recovery of world position

- Each observed feature on the image gives 2 equations with 3 unknowns and therefore defines a line (a ray) of solutions for  $X$





# Recovery of world position

- Each observed feature on the image gives 2 equations with 3 unknowns and therefore defines a line (a ray) of solutions for  $\mathbf{X}$
- This system of equations is under-constrained.
- This can be seen by the size of  $\mathbf{P}$ . There are more columns than rows.

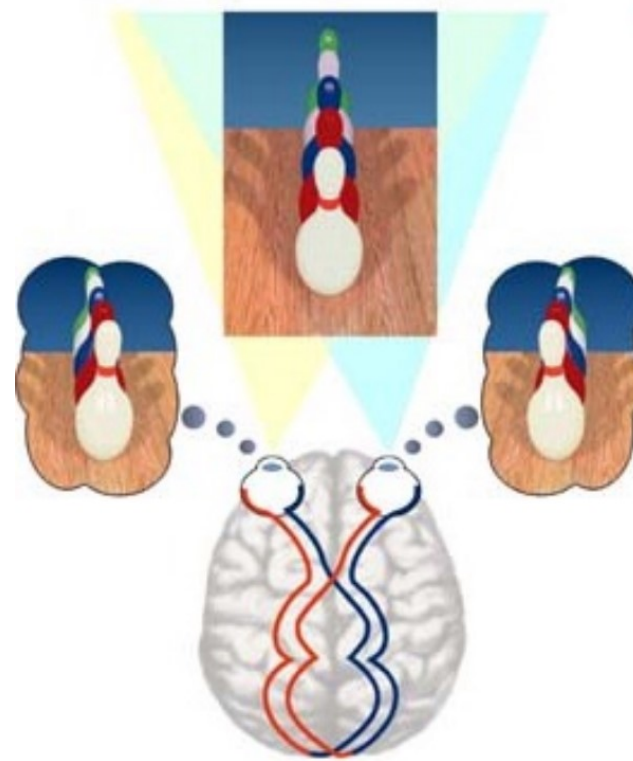
$$\tilde{\mathbf{w}} = \mathbf{P}\tilde{\mathbf{X}}$$
$$\Leftrightarrow \begin{bmatrix} su \\ sv \\ s \end{bmatrix} = \begin{bmatrix} p_{11} & p_{12} & p_{13} & p_{14} \\ p_{21} & p_{22} & p_{23} & p_{24} \\ p_{31} & p_{32} & p_{33} & p_{34} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

# Recovery of world position

- Under-constrained problems never have a unique solution.
- To uniquely recover  $\mathbf{X}$ , additional views must be used, so that the transformation between  $\mathbf{w}$  (pixel coordinates) and  $\mathbf{X}$  (world coordinates) is *forced* to become invertible.
- This is the subject of **stereo vision**.

# Recovery of world position

- Two eyes/cameras help.

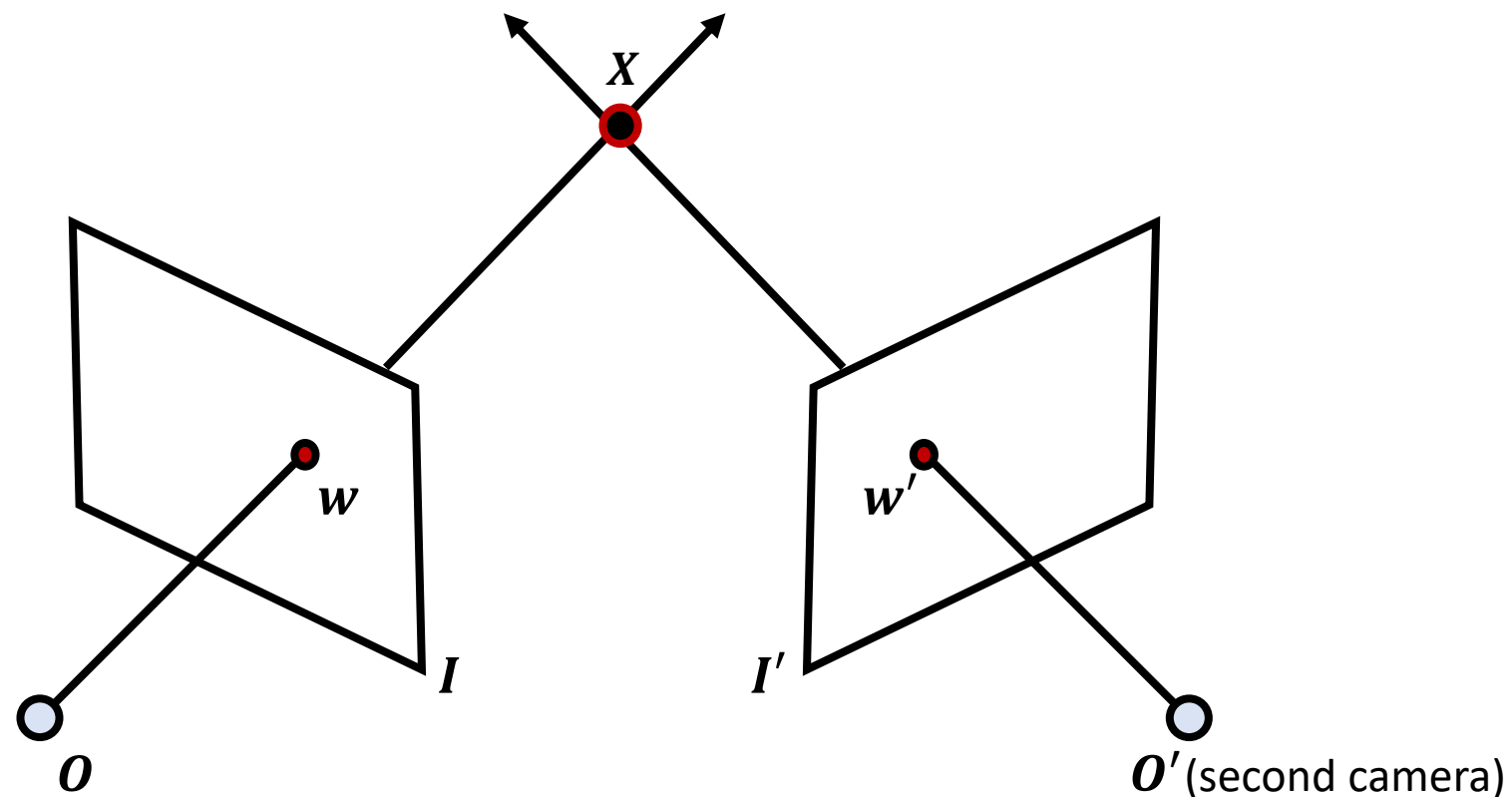


# Today's Agenda

- Recovery of world position
- Triangulation
- Epipolar Geometry

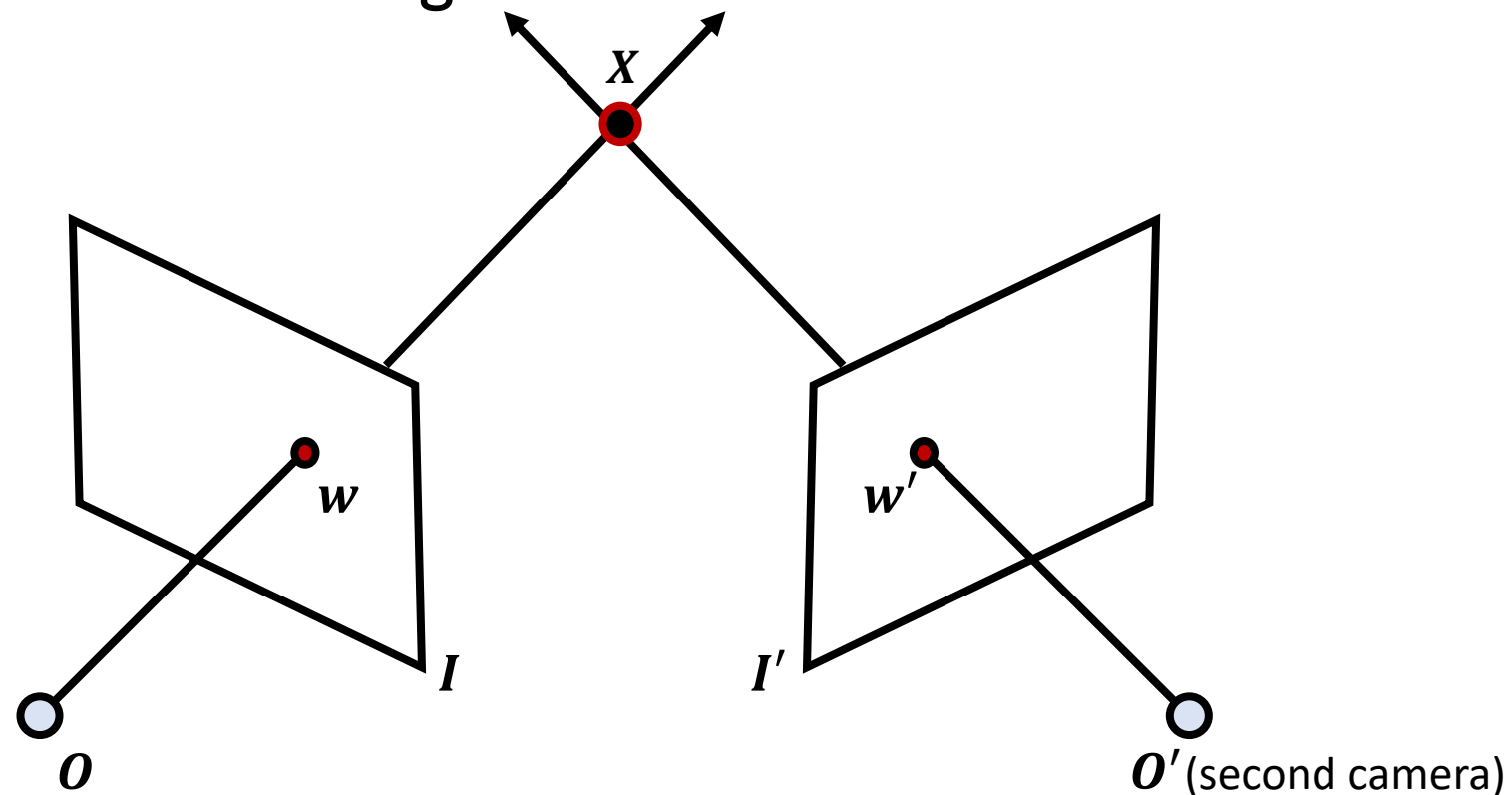
# Triangulation

- In stereo vision at least two cameras are set up to view the  $3D$  scene.
- Each  $3D$  world location  $X$  projects to pixel  $w$  on camera 1 ( $O$ ) and to pixel  $w'$  on camera 2 ( $O'$ ).



# Triangulation

- If both cameras are calibrated, the 3D world location  $X$  projected on the pair of corresponding pixel locations  $w$  and  $w'$  can be estimated via a process known as **triangulation**.





# Triangulation

- Consider the projection of  $\mathbf{X}$  onto  $\mathbf{w}$ :

$$\tilde{\mathbf{w}} = \mathbf{P}\tilde{\mathbf{X}}$$

$$\Leftrightarrow \begin{bmatrix} su \\ sv \\ s \end{bmatrix} = \begin{bmatrix} p_{11} & p_{12} & p_{13} & p_{14} \\ p_{21} & p_{22} & p_{23} & p_{24} \\ p_{31} & p_{32} & p_{33} & p_{34} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

- Compute the equation for  $u$  and rearrange so our unknowns  $X, Y, Z$  are on the left:

$$u = \frac{su}{s} = \frac{p_{11}X + p_{12}Y + p_{13}Z + p_{14}}{p_{31}X + p_{32}Y + p_{33}Z + p_{34}}$$

$$\Rightarrow p_{11}X + p_{12}Y + p_{13}Z + p_{14} = p_{31}uX + p_{32}uY + p_{33}uZ + p_{34}u$$

$$\Rightarrow (p_{11} - p_{31}u)X + (p_{12} - p_{32}u)Y + (p_{13} - p_{33}u)Z = p_{34}u - p_{14}$$



# Triangulation

- Compute the equation for  $u$  and rearrange so our unknowns  $X, Y, Z$  are on the left:

$$u = \frac{su}{s} = \frac{p_{11}X + p_{12}Y + p_{13}Z + p_{14}}{p_{31}X + p_{32}Y + p_{33}Z + p_{34}}$$

$$\Rightarrow p_{11}X + p_{12}Y + p_{13}Z + p_{14} = p_{31}uX + p_{32}uY + p_{33}uZ + p_{34}u$$

$$\Rightarrow (p_{11} - p_{31}u)X + (p_{12} - p_{32}u)Y + (p_{13} - p_{33}u)Z = p_{34}u - p_{14}$$

- Put this in matrix form:

$$\begin{bmatrix} p_{11} - p_{31}u & p_{12} - p_{32}u & p_{13} - p_{33}u \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = [p_{34}u - p_{14}]$$
$$\Leftrightarrow \mathbf{Ax} = \mathbf{b}$$



# Triangulation

- Put this in matrix form:

$$\begin{bmatrix} p_{11} - p_{31}u & p_{12} - p_{32}u & p_{13} - p_{33}u \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = [p_{34}u - p_{14}]$$
$$\Leftrightarrow \mathbf{Ax} = \mathbf{b}$$

- $\mathbf{A}$  is a  $1 \times 3$  matrix
- The resulting system is under-constrained

# Triangulation

- Put this in matrix form:

$$\begin{aligned}
 & [p_{11} - p_{31}u \quad p_{12} - p_{32}u \quad p_{13} - p_{33}u] \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = [p_{34}u - p_{14}] \\
 & [p_{21} - p_{31}v \quad p_{22} - p_{32}v \quad p_{23} - p_{33}v] \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = [p_{34}v - p_{24}] \\
 & \Leftrightarrow \mathbf{Ax} = \mathbf{b}
 \end{aligned}$$

- By computing  $v$  in the same way as  $u$  and rearranging we can add a new row in  $\mathbf{A}$  and in  $\mathbf{b}$ , making it  $2 \times 3$



# Triangulation

- Put this in matrix form:

$$\begin{bmatrix} p'_{11} - p'_{31}u' & p'_{12} - p'_{32}u' & p'_{13} - p'_{33}u' \\ p'_{21} - p'_{31}v' & p'_{22} - p'_{32}v' & p'_{23} - p'_{33}v' \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = [p'_{34}u' - p'_{14}]$$

$$\begin{bmatrix} p'_{21} - p'_{31}v' & p'_{22} - p'_{32}v' & p'_{23} - p'_{33}v' \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = [p'_{34}v' - p'_{24}]$$

$$\Leftrightarrow \mathbf{Ax} = \mathbf{b}$$

- By also considering the projection of  $\mathbf{X}$  onto  $\mathbf{w}'$ , a new pair of rows will be added to  $\mathbf{A}$ , thus forcing it to be  $4 \times 3$ , i.e. over-constrained



# Triangulation

- Here is the resulting system

$$\begin{bmatrix} p_{11} - p_{31}u & p_{12} - p_{32}u & p_{13} - p_{33}u \\ p_{21} - p_{31}v & p_{22} - p_{32}v & p_{23} - p_{33}v \\ p'_{11} - p'_{31}u' & p'_{12} - p'_{32}u' & p'_{13} - p'_{33}u' \\ p'_{21} - p'_{31}v' & p'_{22} - p'_{32}v' & p'_{23} - p'_{33}v' \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = \begin{bmatrix} p_{34}u - p_{14} \\ p_{34}v - p_{24} \\ p'_{34}u' - p'_{14} \\ p'_{34}v' - p'_{24} \end{bmatrix}$$

- Where  $p'_{11}$  etc. are the parameters inside the camera projection matrix for camera 2 ( $\mathbf{O}'$ ), and  $\mathbf{w}' = (u', v')$  are the pixel coordinates of  $\mathbf{X}$  projected on the image plane  $\mathbf{I}'$  of the second camera

# Triangulation

- Using an additional camera has forced  $\mathbf{A}$  to become over-constrained.

$$\begin{bmatrix} p_{11} - p_{31}u & p_{12} - p_{32}u & p_{13} - p_{33}u \\ p_{21} - p_{31}v & p_{22} - p_{32}v & p_{23} - p_{33}v \\ p'_{11} - p'_{31}u' & p'_{12} - p'_{32}u' & p'_{13} - p'_{33}u' \\ p'_{21} - p'_{31}v' & p'_{22} - p'_{32}v' & p'_{23} - p'_{33}v' \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \end{bmatrix} = \begin{bmatrix} p_{34}u - p_{14} \\ p_{34}v - p_{24} \\ p'_{34}u' - p'_{14} \\ p'_{34}v' - p'_{24} \end{bmatrix}$$

- Therefore we can now find a least-squares solution:

$$\begin{aligned} \mathbf{Ax} &= \mathbf{b} \\ \Leftrightarrow \mathbf{x} &= (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{b} \end{aligned}$$



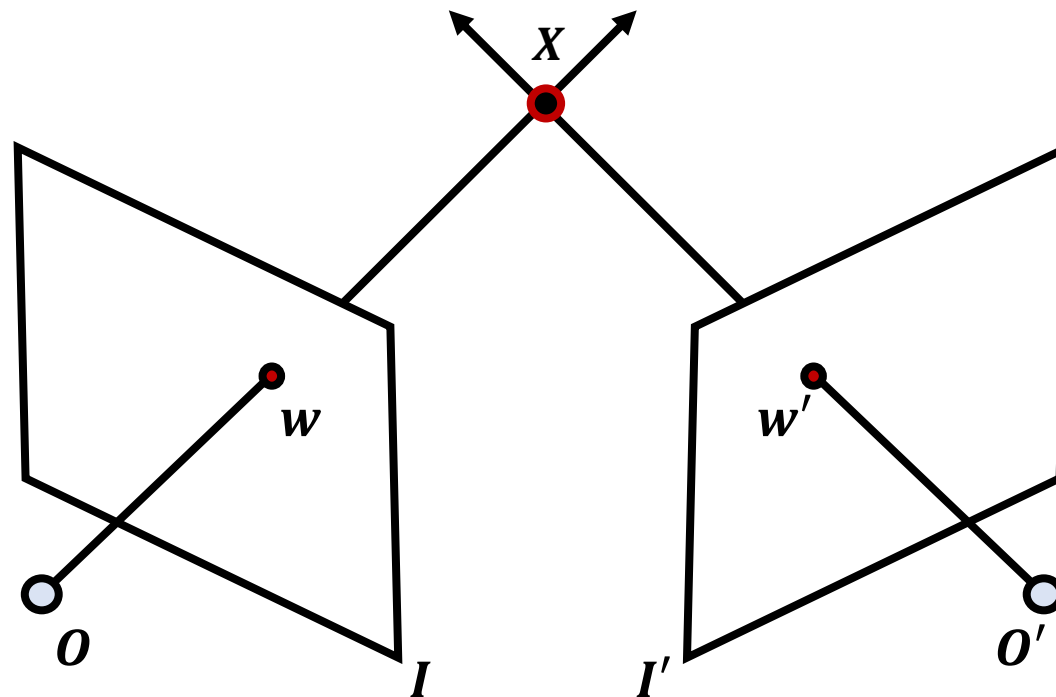
# Today's Agenda

- Recovery of world position
- Triangulation
- Epipolar Geometry



# Beyond triangulation

We have seen the simplest form of stereo vision: Given a pair of *calibrated* cameras observing a single feature at *corresponding* pixel locations  $\mathbf{w} \leftrightarrow \mathbf{w}'$ , the 3D position of the corresponding world location  $\mathbf{X}$  can be estimated via triangulation



# Beyond triangulation

How is the correspondence problem solved if there are several points  $\{\mathbf{w}_i\}_{i=1}^{N_1}$  in image 1 and several points  $\{\mathbf{w}'_j\}_{j=1}^{N_2}$  in image 2 ?





# Beyond triangulation

SIFT will give us a set of proposed correspondences  $\{w_i \leftrightarrow w'_j\}$  but there will be **many** outliers in these proposals

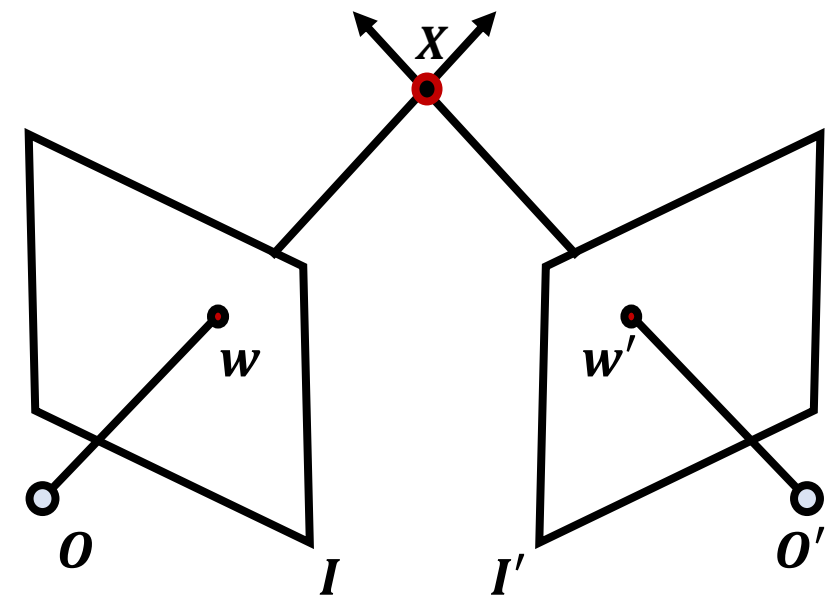
The question is then how can we remove outliers ?

Using the **epipolar constraint**



# Epipolar Geometry

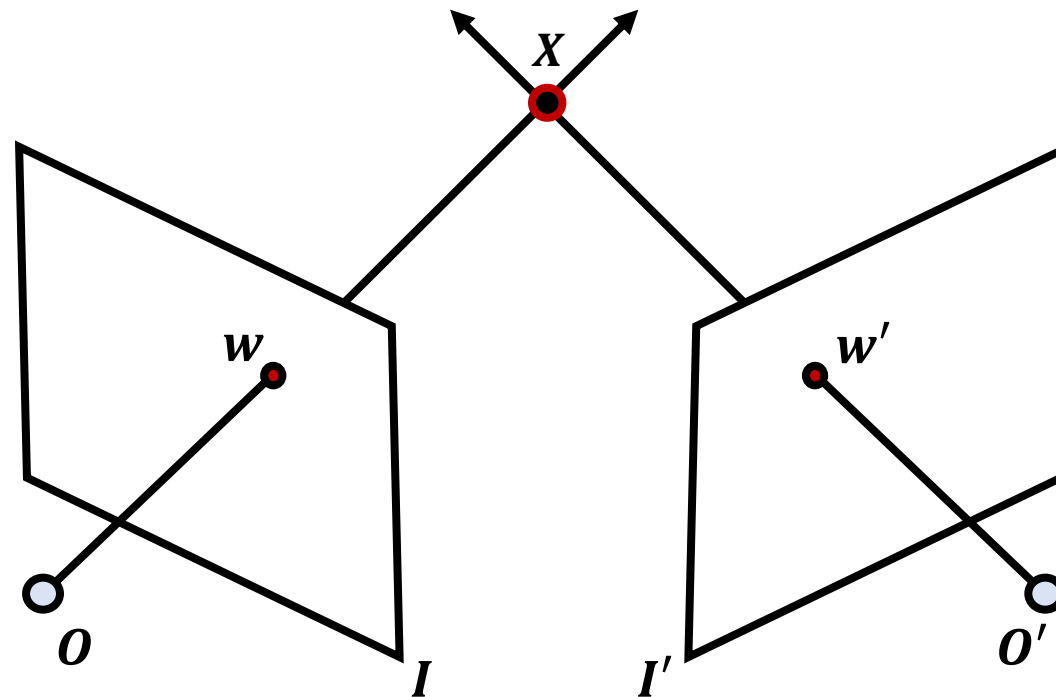
- To understand the **epipolar constraint**, we first need to understand the **geometry** that relates the
  - cameras
  - points in **3D** space
  - and their corresponding observations  $\{w_i \leftrightarrow w'_j\}$



- This type of geometry is referred to as the **epipolar geometry**

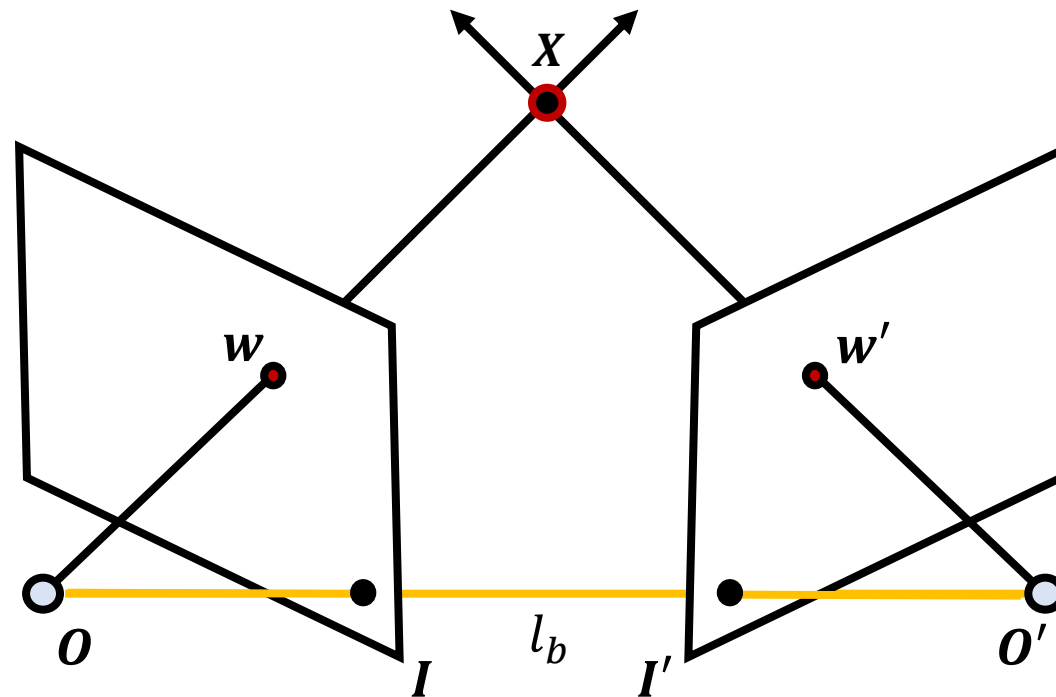
# Epipolar Geometry - fundamentals

Lets revisit our stereo pair



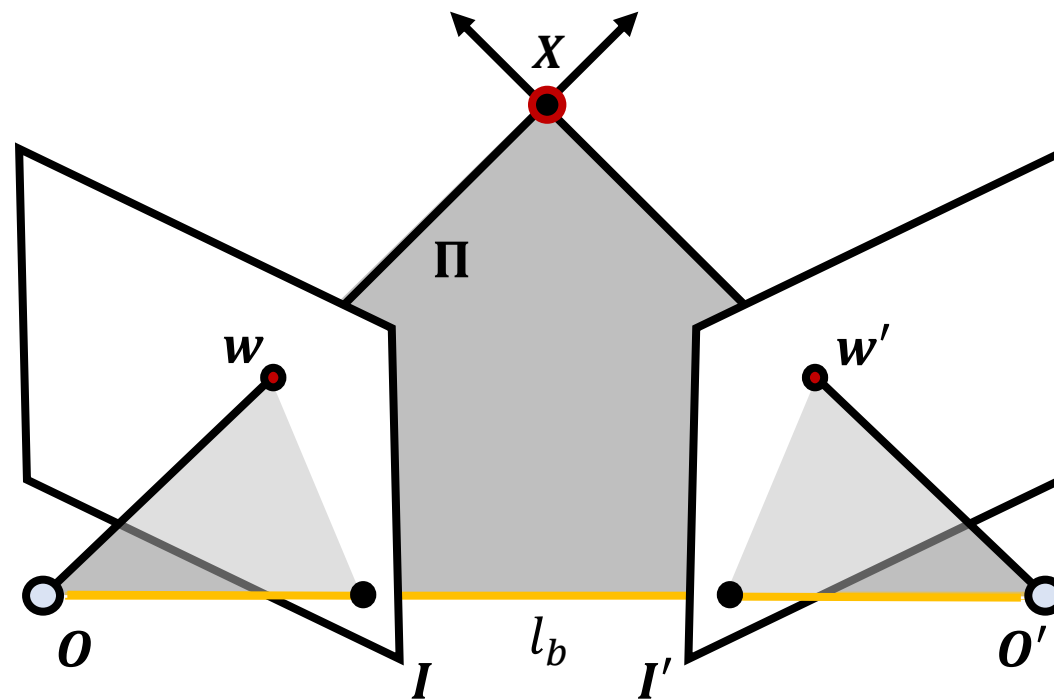
# Epipolar Geometry - fundamentals

The **baseline**  $l_b$  is the line joining the two optical centers.



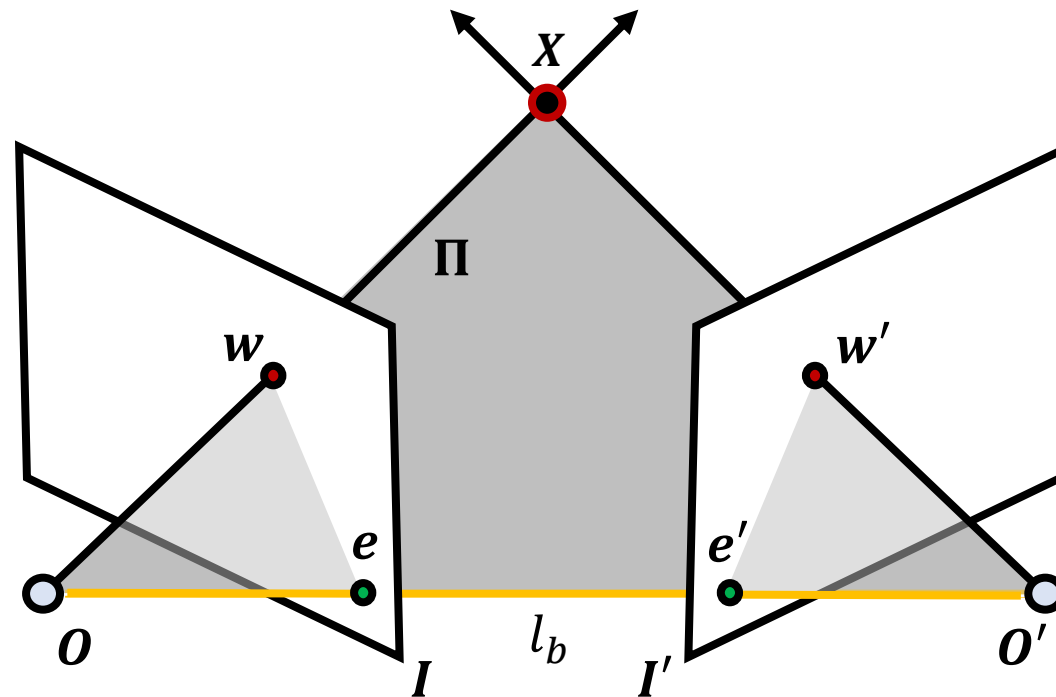
# Epipolar Geometry - fundamentals

The epipolar plane  $\Pi$  is the plane defined by the 3D point  $X$  and the optical centers of the cameras.



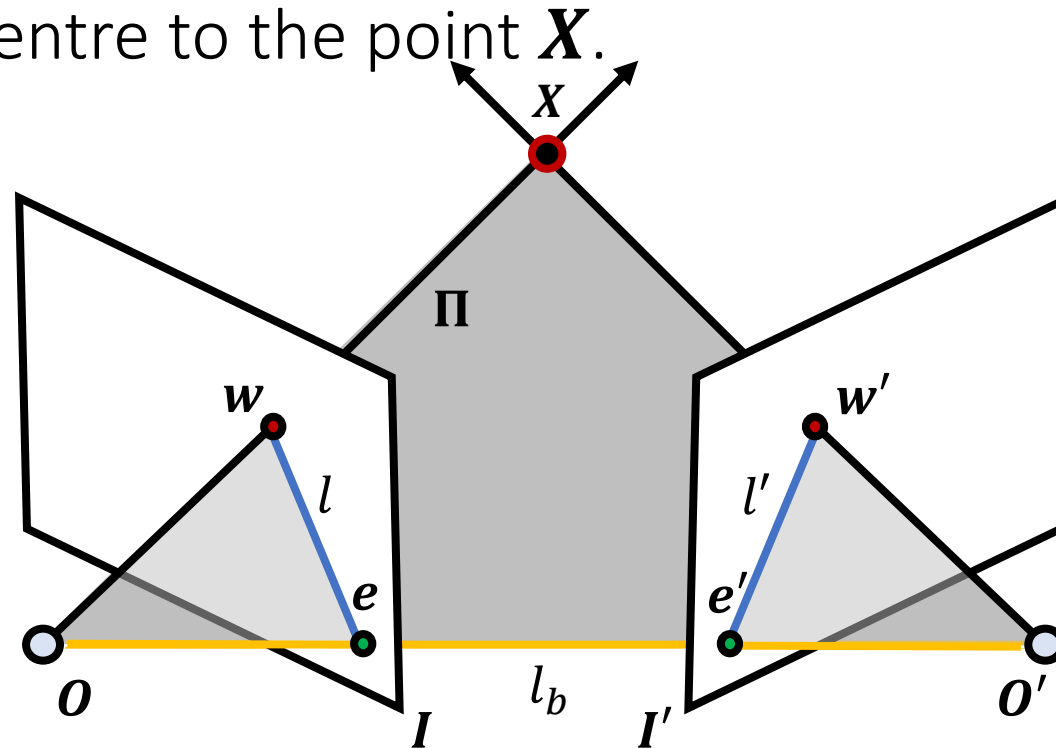
# Epipolar Geometry - fundamentals

An **epipole** is the point of intersection of the **baseline** with the image plane. There are two epipoles  $e$  and  $e'$ , one for each image.



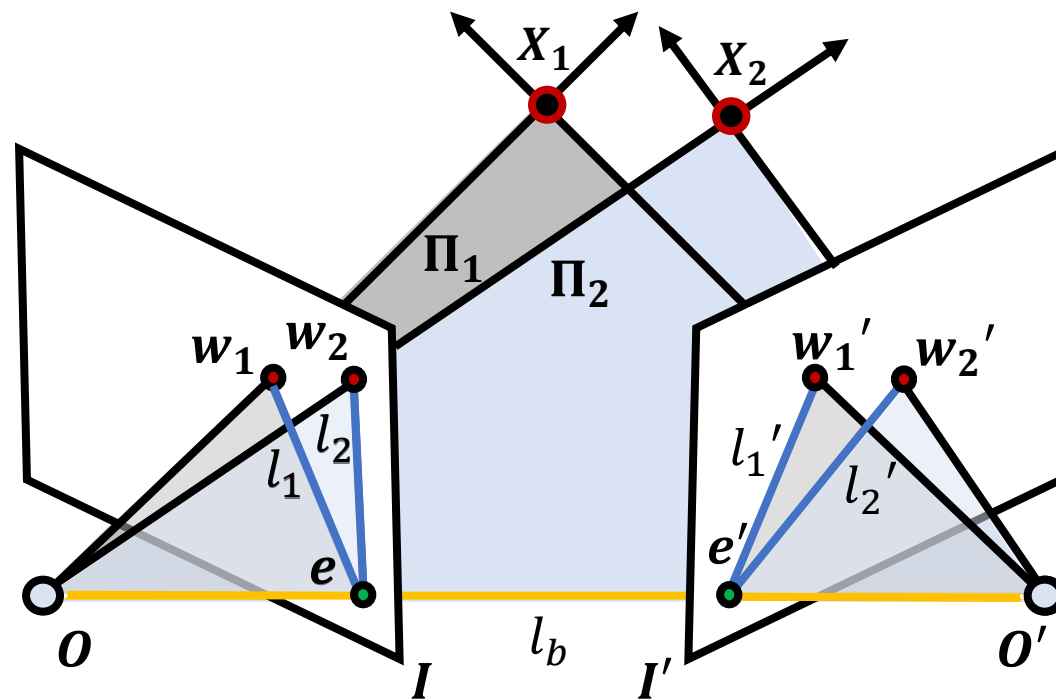
# Epipolar Geometry - fundamentals

An **epipolar line** is a line of intersection of the epipolar plane with an image plane. It is the image, in one camera, of the ray from the other camera's optical centre to the point  $X$ .



# Epipolar Geometry - fundamentals

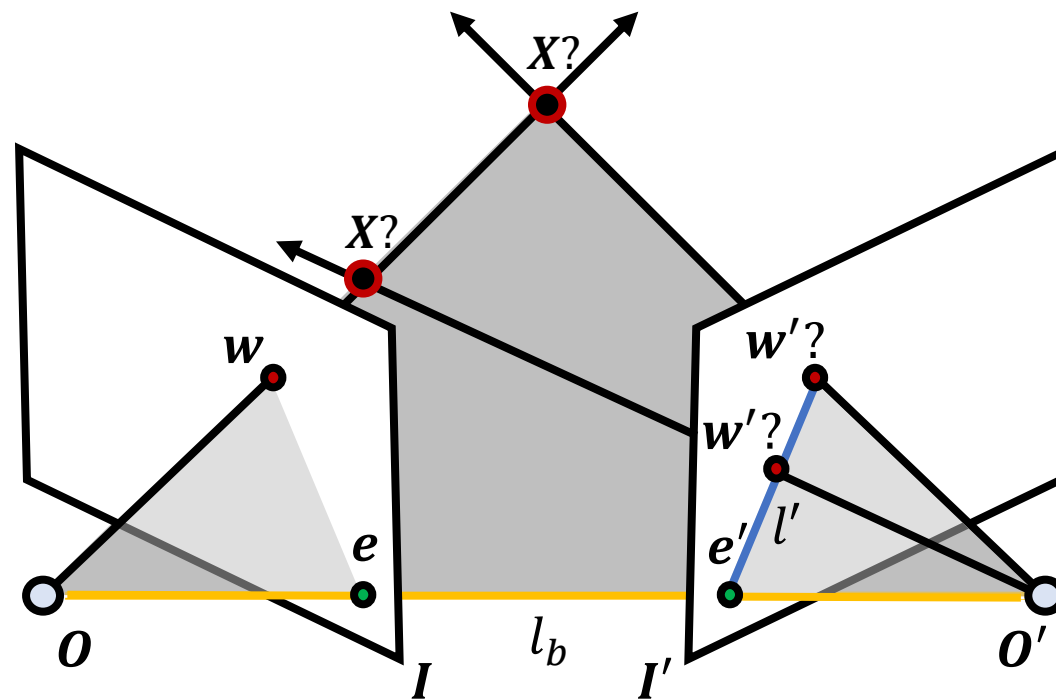
For different world points  $X$ , the epipolar plane rotates about the baseline. All epipolar lines intersect at their corresponding epipole.





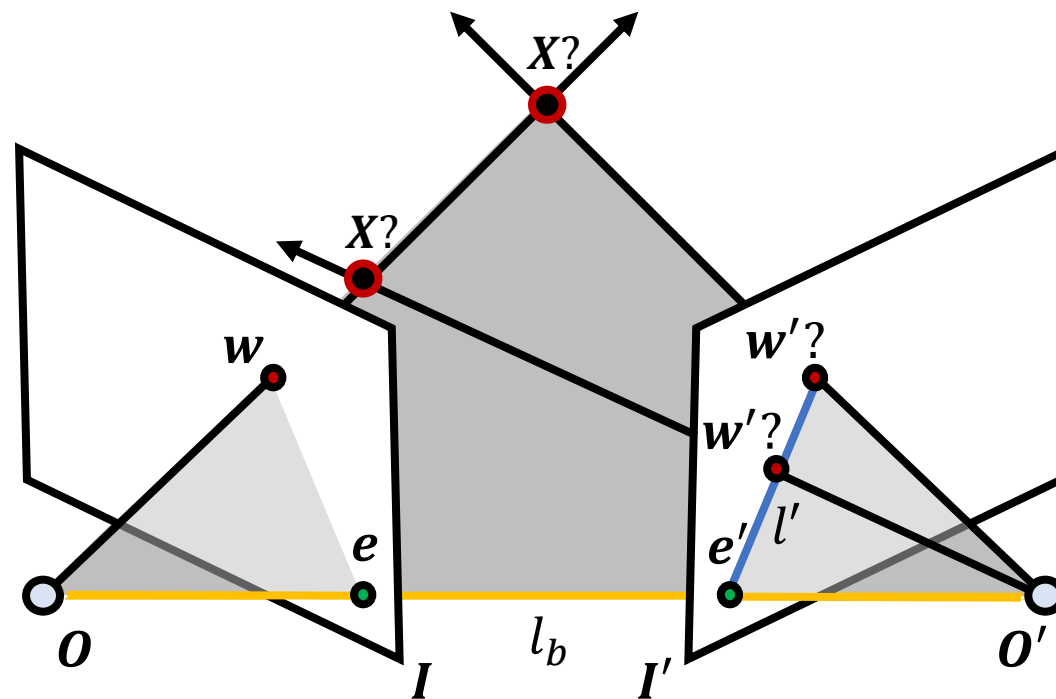
# Epipolar Geometry - fundamentals

The **epipolar constraint** limits the search for correspondences, from the region of the whole image, to only the pixels spanned by the epipolar line.



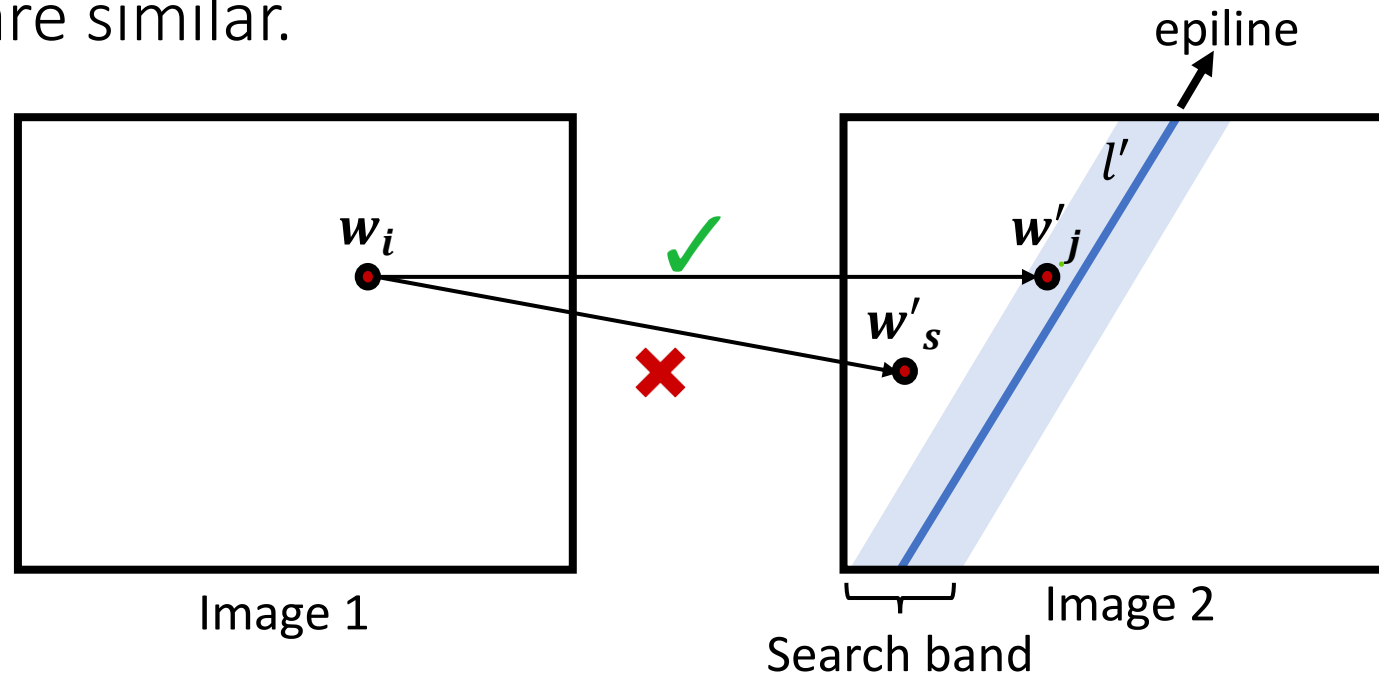
# Epipolar Geometry - fundamentals

If a point feature  $w$  is observed in one image, then its location  $w'$  in the other image must lie on its corresponding epipolar line  $l'$



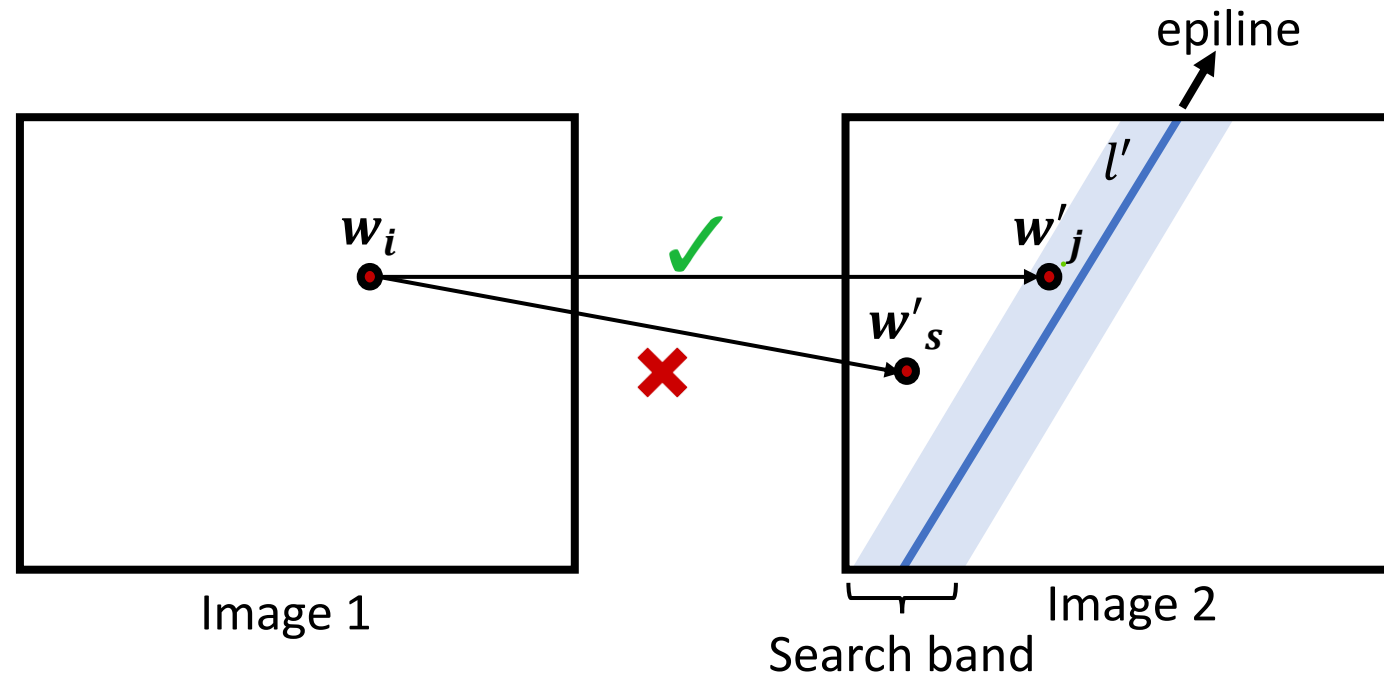
# Epipolar Geometry - fundamentals

So a simple algorithm for determining correspondences is to match each feature  $w_i$  from camera 1 to a feature  $w'_j$  from camera 2 which is close to the epipolar line of camera 2, provided that the SIFT descriptors for  $w_i$  and  $w'_j$  are similar.



# Epipolar Geometry - fundamentals

Therefore, we must calculate the equation of the epipolar lines.



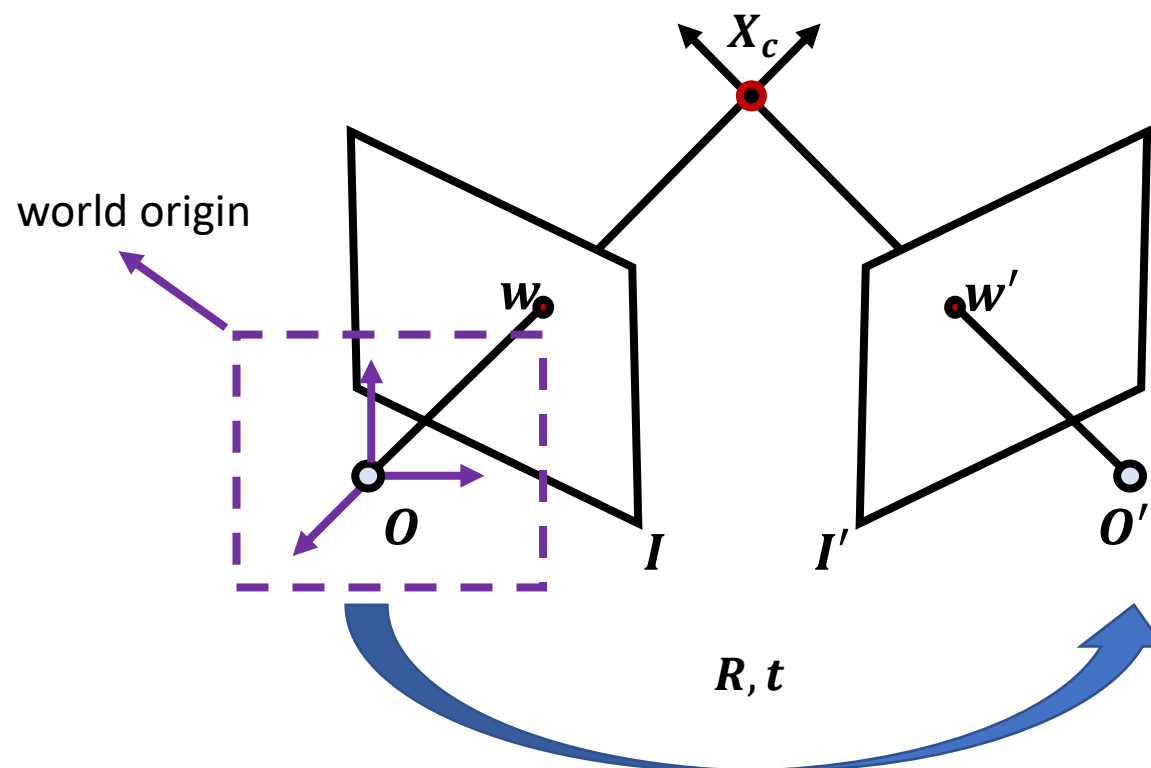
# Essential Matrix

- Firstly, let's assume that the 1<sup>st</sup> camera is located at the **world origin  $O$**
- This means that every point  $X$  is expressed in the camera coordinates of the 1<sup>st</sup> camera, since its optical center coincides with the world origin  $\Rightarrow X \rightarrow X_c$



# Essential Matrix

- If this is the case, the 2<sup>nd</sup> camera is at a location  $\mathbf{O}'$  in world coordinates, which can be expressed by a **translation  $\mathbf{t}$**  and **rotation  $\mathbf{R}$** , w.r.t. 1<sup>st</sup> camera, i.e., the world origin



# Essential Matrix

- Every  $3D$  point  $\mathbf{X}$  can be expressed as  $\mathbf{X}_c$ , which is the camera-centered coordinate system of the 1<sup>st</sup> camera, and as  $\mathbf{X}'_c$ , which is the camera-centered coordinate system of the 2<sup>nd</sup> camera.
- Since  $\mathbf{X} = \mathbf{X}_c$ , we can relate  $\mathbf{X} \leftrightarrow \mathbf{X}'_c$  or  $\mathbf{X}_c \leftrightarrow \mathbf{X}'_c$  using a **Euclidean transformation** composed by the **translation  $\mathbf{t}$**  and **rotation  $\mathbf{R}$**  of the 2<sup>nd</sup> camera, w.r.t. the 1<sup>st</sup> camera

# Essential Matrix

Here is how to find an expression for the epipolar line:

$$\begin{aligned}
 \widetilde{X}'_c &= P_e \widetilde{X}_c \\
 \Leftrightarrow X'_c &= RX_c + t \\
 \Leftrightarrow t \times X'_c &= t \times RX_c + \cancel{t \times t}^0 \\
 \Leftrightarrow X'_c \cdot (t \times X'_c) &= X'_c \cdot (t \times RX_c) \\
 \Leftrightarrow \mathbf{0} &= X'_c \cdot (t \times RX_c)
 \end{aligned}$$

- ← This is in homogeneous coordinates
- ← Express it in cartesian coordinates
- ← Apply cross product with  $t$  to both sides
- ← Apply dot product with  $X'_c$  to both sides



# Essential Matrix

This can be rewritten in matrix form:

$$\begin{aligned} \mathbf{X}'_c \cdot (\mathbf{t} \times \mathbf{R}\mathbf{X}_c) &= \mathbf{0} \\ \Leftrightarrow \mathbf{X}'_c{}^T \mathbf{E} \mathbf{X}_c &= \mathbf{0} \end{aligned}$$

Where  $\mathbf{E} = \mathbf{T}_\times \mathbf{R}$  is the **essential matrix**, and

$$\mathbf{T}_\times = \begin{bmatrix} 0 & -t_z & t_y \\ t_z & 0 & -t_x \\ -t_y & t_x & 0 \end{bmatrix}$$

is a matrix representing the cross product with  $\mathbf{t}$  such that  $\mathbf{t} \times \mathbf{v} = \mathbf{T}_\times \mathbf{v}$

# Fundamental Matrix

Recall that for  $\mathbf{X}_c \leftrightarrow \mathbf{w}$ :  $\tilde{\mathbf{w}} = \mathbf{K}\mathbf{X}_c \Leftrightarrow \mathbf{X}_c = \mathbf{K}^{-1}\tilde{\mathbf{w}}$

and similarly, for  $\mathbf{X}'_c \leftrightarrow \mathbf{w}'$ :  $\tilde{\mathbf{w}}' = \mathbf{K}'\mathbf{X}'_c \Leftrightarrow \mathbf{X}'_c = \mathbf{K}'^{-1}\tilde{\mathbf{w}}'$

Combining the two equations yields the equation of **the two epipolar lines in pixel coordinates**:

$$\begin{aligned} \mathbf{X}'^T \mathbf{E} \mathbf{X}_c &= 0 \\ \Rightarrow (\mathbf{K}'^{-1}\tilde{\mathbf{w}}')^T \mathbf{E} (\mathbf{K}^{-1}\tilde{\mathbf{w}}) &= 0 \\ \Rightarrow \tilde{\mathbf{w}}'^T (\mathbf{K}'^{-T} \mathbf{E} \mathbf{K}^{-1}) \tilde{\mathbf{w}} &= 0 \\ \Rightarrow \tilde{\mathbf{w}}'^T \mathbf{F} \tilde{\mathbf{w}} &= 0 \end{aligned}$$

where  $\mathbf{F} = \mathbf{K}'^{-T} \mathbf{E} \mathbf{K}^{-1}$  is the fundamental matrix.

# Fundamental Matrix

This is the equation of the epipolar line in either camera:

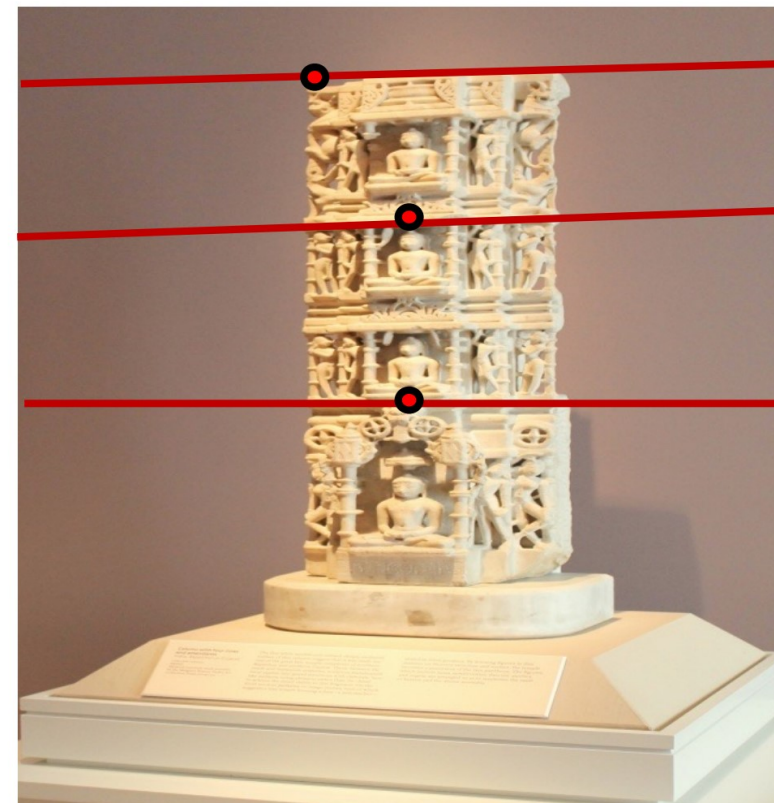
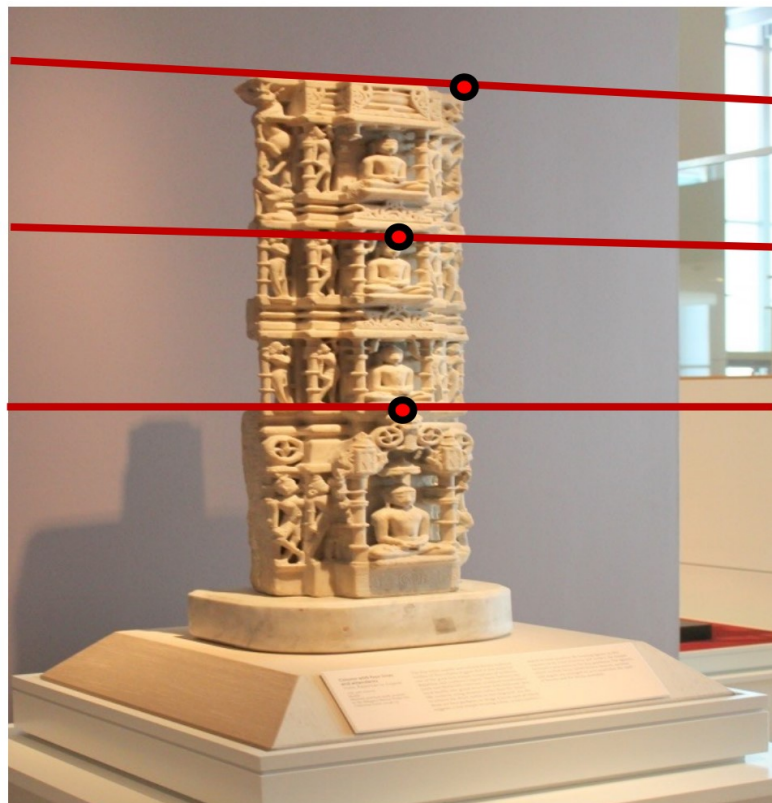
$$\tilde{\mathbf{w}}'^T \mathbf{F} \tilde{\mathbf{w}} = 0$$

Assuming we know the fundamental matrix, for every point  $\mathbf{w}$  in image 1, this expression gives us the line in image 2 on which the corresponding  $\mathbf{w}'$  must lie, and *vice versa*.

- $l' = \mathbf{F}\tilde{\mathbf{w}}$  is the epipolar line in the 2<sup>nd</sup> image, associated with  $\mathbf{w}$
- $l = \mathbf{F}^T \tilde{\mathbf{w}}'$  is the epipolar line in the 1<sup>st</sup> image, associated with  $\mathbf{w}'$
- $l_i: ax + by + c = 0$

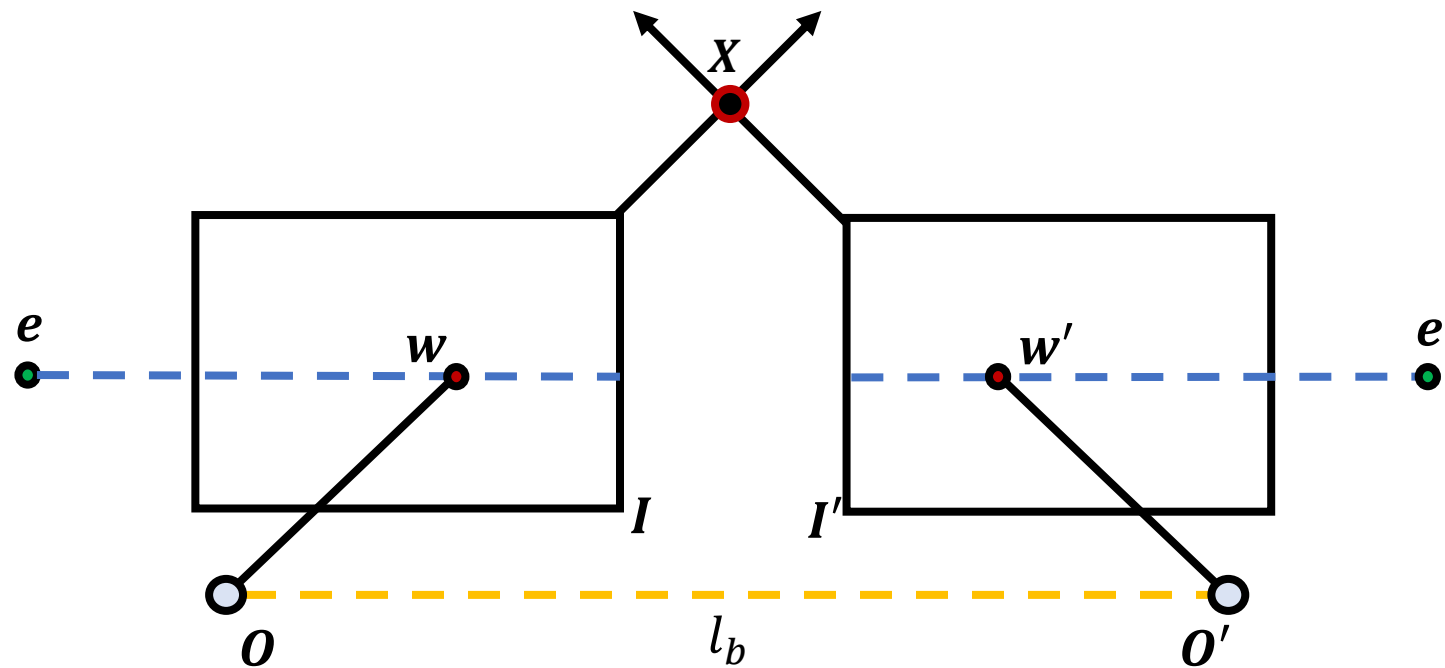
# Examples

Here are a few examples of the epipolar constraint



# Examples

Epipolar constraint examples: Parallel image planes



- **Baseline** intersects the image planes at infinity
- **Epipoles** are at infinity
- **Epipolar lines** are parallel to the  $u$ -axis of each image plane



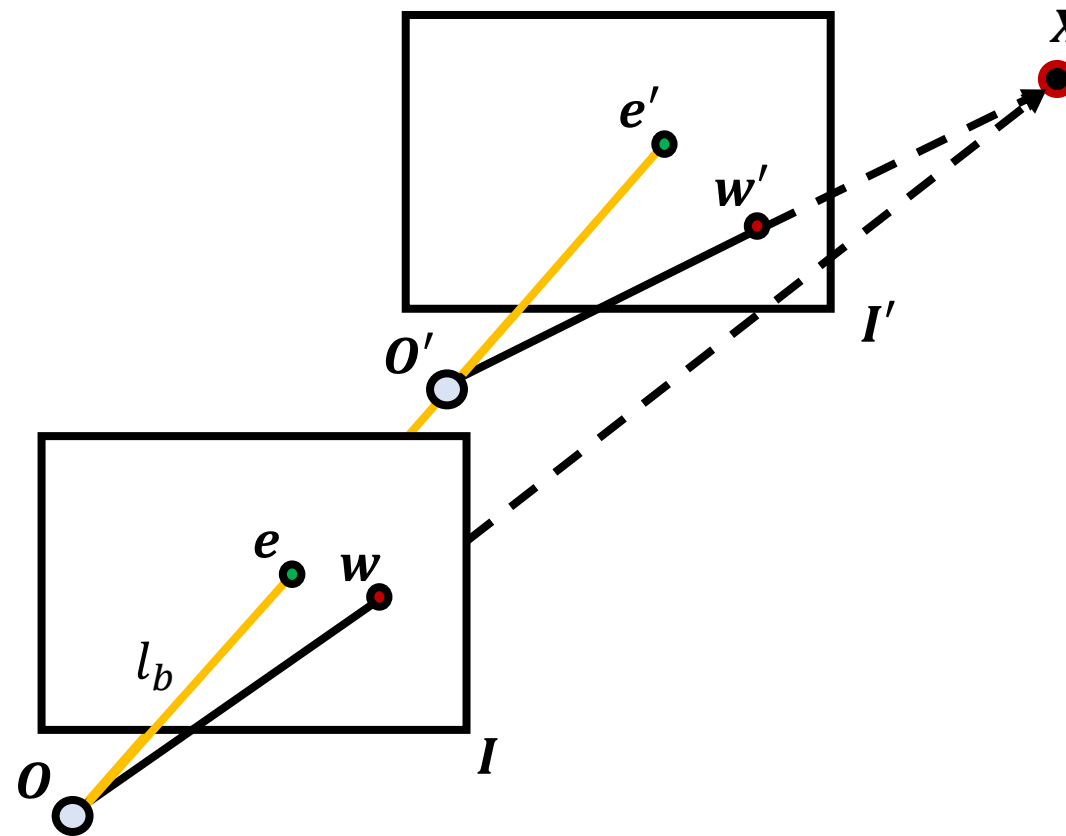
# Examples

Epipolar constraint examples: **Parallel image planes**



# Examples

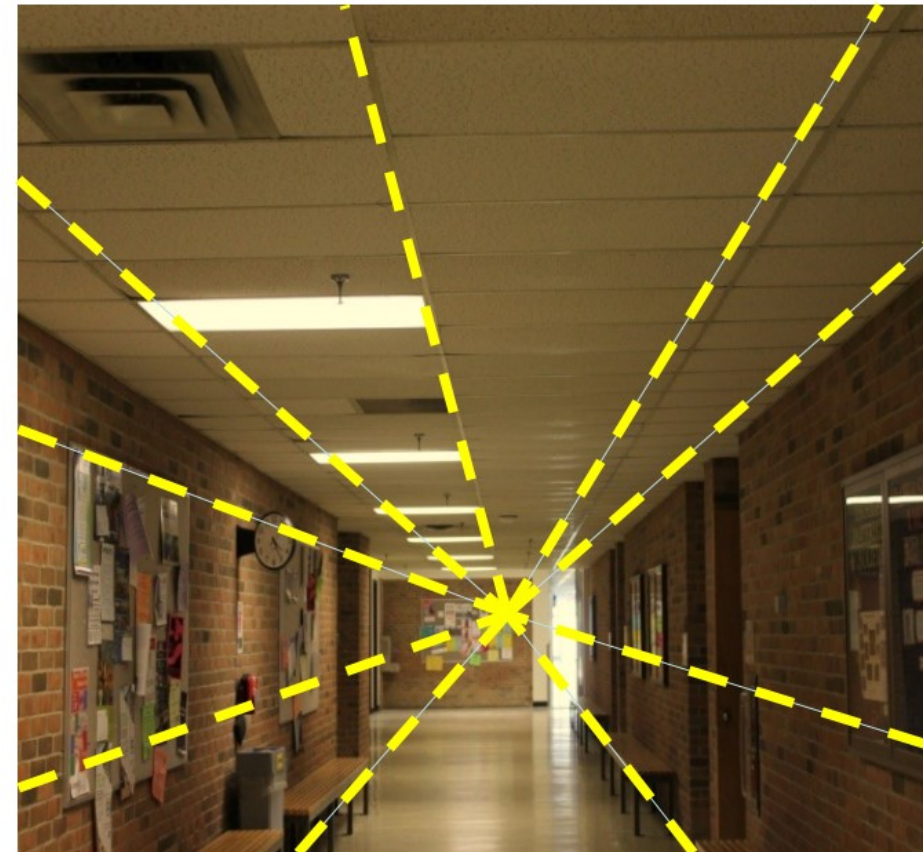
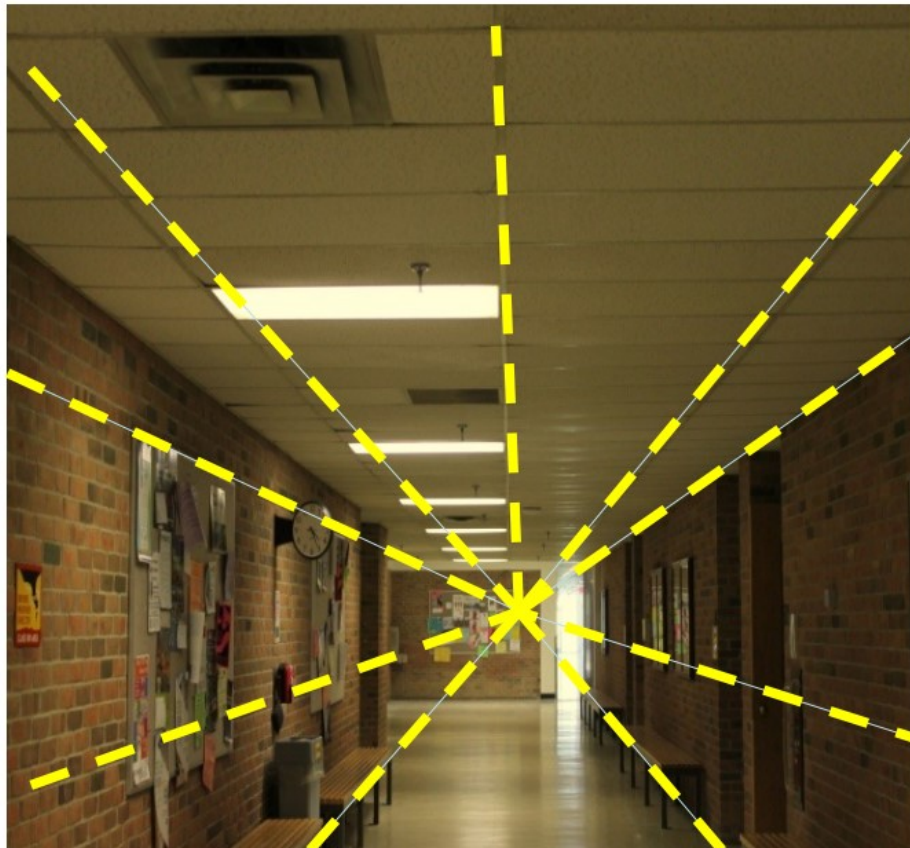
## Epipolar constraint examples: Forward translation



- The **epipoles** have the same position in both images

# Examples

## Epipolar constraint examples: Forward translation





# Fundamental Matrix

In order to apply the epipolar constraint we need to know the fundamental matrix:

$$\begin{aligned}\mathbf{F} &= \mathbf{K}'^{-T} \mathbf{E} \mathbf{K}^{-1} \\ &= \mathbf{K}'^{-T} \mathbf{T}_{\times} \mathbf{R} \mathbf{K}^{-1}\end{aligned}$$

All the parameters of  $\mathbf{F}$  come from the calibration of the two cameras.

Remember that the perspective camera model  $\mathbf{P}_{ps} = \mathbf{K}[\mathbf{R}|\mathbf{T}]$  contains this information.

If, however, these are not available (e.g., because we used a projective camera model to calibrate our cameras),  $\mathbf{F}$  must be estimated using known image correspondences.

# Estimating the fundamental matrix

To estimate the fundamental matrix, we follow a similar approach as in camera calibration

$$\tilde{\mathbf{w}}'^T \mathbf{F} \tilde{\mathbf{w}} = 0 \Rightarrow [u' \ v' \ 1] \begin{bmatrix} f_{11} & f_{12} & f_{13} \\ f_{21} & f_{22} & f_{23} \\ f_{31} & f_{32} & f_{33} \end{bmatrix} \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = 0$$

Each point correspondence  $\mathbf{w} \leftrightarrow \mathbf{w}'$  generates a single equation for estimating the parameters inside  $\mathbf{F}$ .

# Estimating the fundamental matrix

Here are  $N$  such correspondences:

$$\begin{bmatrix} u_1 u'_1 & v_1 u'_1 & u'_1 & u_1 v'_1 & v_1 v'_1 & v'_1 & u_1 & v_1 \\ & & \vdots & & & & & \\ u_N u'_N & v_N u'_N & u'_N & u_N v'_N & v_N v'_N & v'_N & u_N & v_N \end{bmatrix} \begin{bmatrix} f_{11} \\ f_{12} \\ f_{13} \\ f_{21} \\ f_{22} \\ f_{23} \\ f_{31} \\ f_{32} \end{bmatrix} = \begin{bmatrix} -1 \\ \vdots \\ -1 \end{bmatrix}$$

The minimal number for  $N$  is 8, because  $\mathbf{F}$  has 8 degrees of freedom (we can set  $f_{33} = 1$ )

# Estimating the fundamental matrix

- In practise, 8 clearly indicated correspondences are never available
- What is available is a set of correspondences proposed by SIFT, for a large set of features, in which there are **always** errors

# Estimating the fundamental matrix

- We can use RANSAC to solve this problem
  1. Obtain 8 random SIFT correspondences. Use them to estimate  $\mathbf{F}$ .
  2. For every feature  $\mathbf{w}_i$  in image 1 calculate the epipolar line in image 2. Check if the corresponding feature  $\mathbf{w}'_j$  in image 2 proposed by SIFT falls “close” to the epipolar line. Count the number  $S$  of such “inliers”.
  3. If  $S \geq T$  where  $T$  is a threshold, then there is consensus with the random sample taken in the first step. Calculate  $\mathbf{F}$  for all inliers and terminate here.
  4. If  $S < T$  then no consensus is reached. Repeat from step 1.
  5. If after  $N$  iterations no consensus is reached, select the model that gave the highest  $S$ , calculate  $\mathbf{F}$  using all inliers in  $S$  and terminate

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# Thank you.

