

Multiview Geometry



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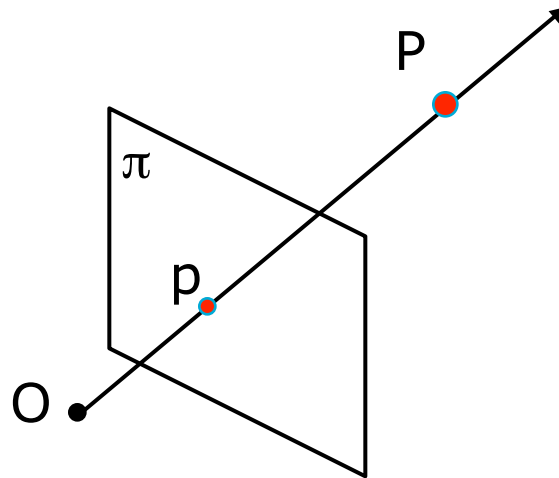
Multiview Analysis

- Observing the same scene point from multiple distinct viewpoints allows the recovery of 3D structure.
- A key component of multiview analysis is finding corresponding scene regions in the different image planes – *the correspondence problem*.
- The relative shift between corresponding projections, *the disparity*, provides 3D structure information.
- Recovery of exact 3D data requires further knowledge about the camera setup.



First Camera

■ Camera 1



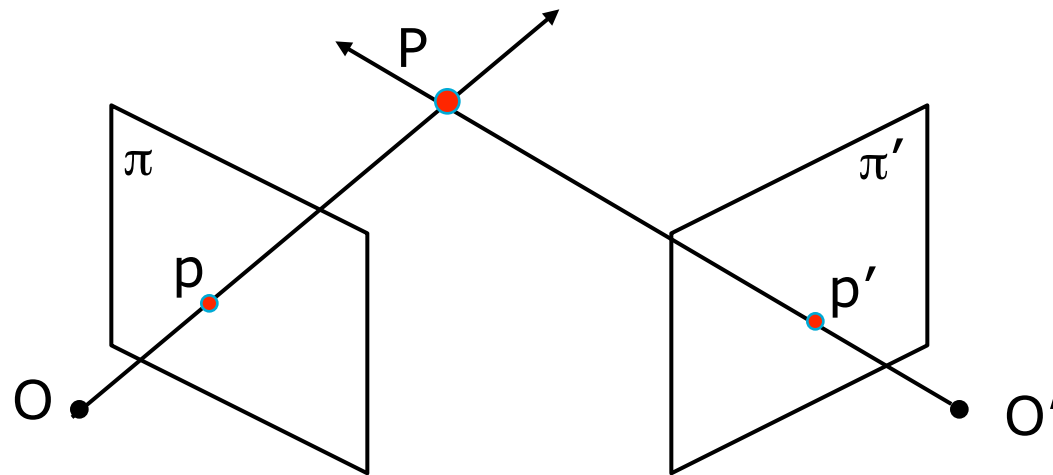
■ Camera 1:

- Center of Projection O
- Image plane π
- Scene point P projects on point p on π .



Second Camera

■ Camera 2



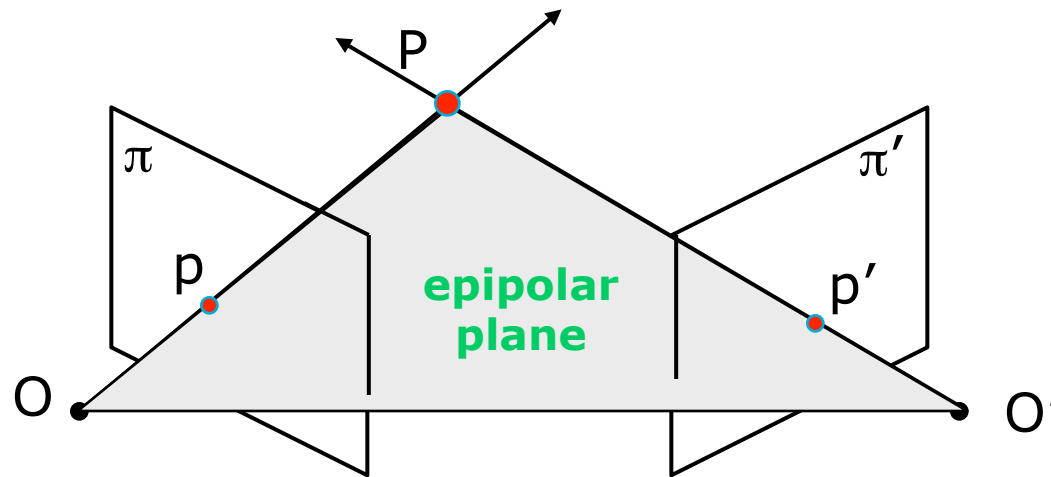
■ Camera 2:

- Center of Projection O'
- Image plane π'
- Scene point P projects on point p' on π' .



Epipolar Plane

- The epipolar plane is defined by the 2 COPs O and O' and a point in the scene P .

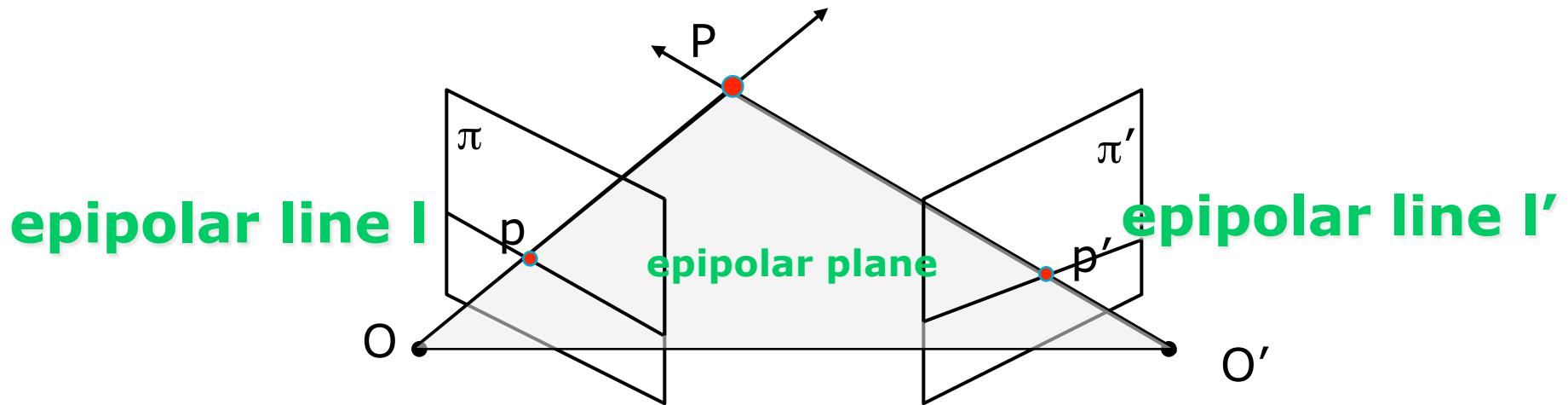


- The lines OP and $O'P$ lie on the epipolar plane Γ .
- Point p lies on the OP line and on the image plane π . It is the intersection of OP and π .
- Point p' lies on the $O'P$ line and on the image plane π' . It is the intersection of $O'P$ and π' .



Epipolar Line

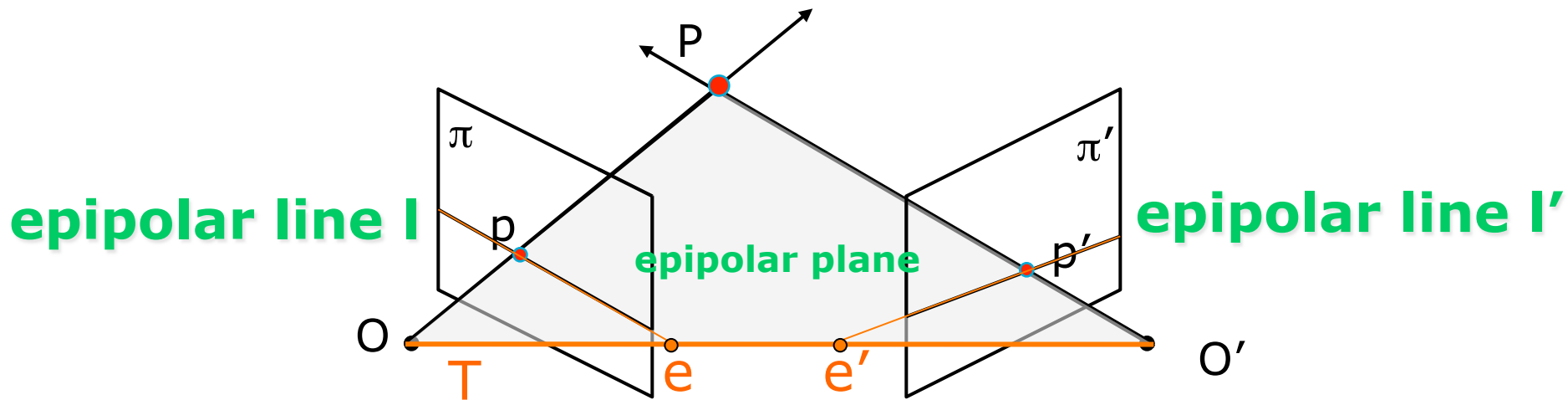
- The epipolar line is the intersection of the epipolar plane with the image plane.



- Since point p' lies on the $O'P$ line and on the image plane π' , it also lies on the intersection of the epipolar plane with the image plane π' , i.e. on the epipolar line l'
- Since point p lies on the OP line and on the image plane π , it also lies on the intersection of the epipolar plane with the image plane π , i.e. on the epipolar line l .



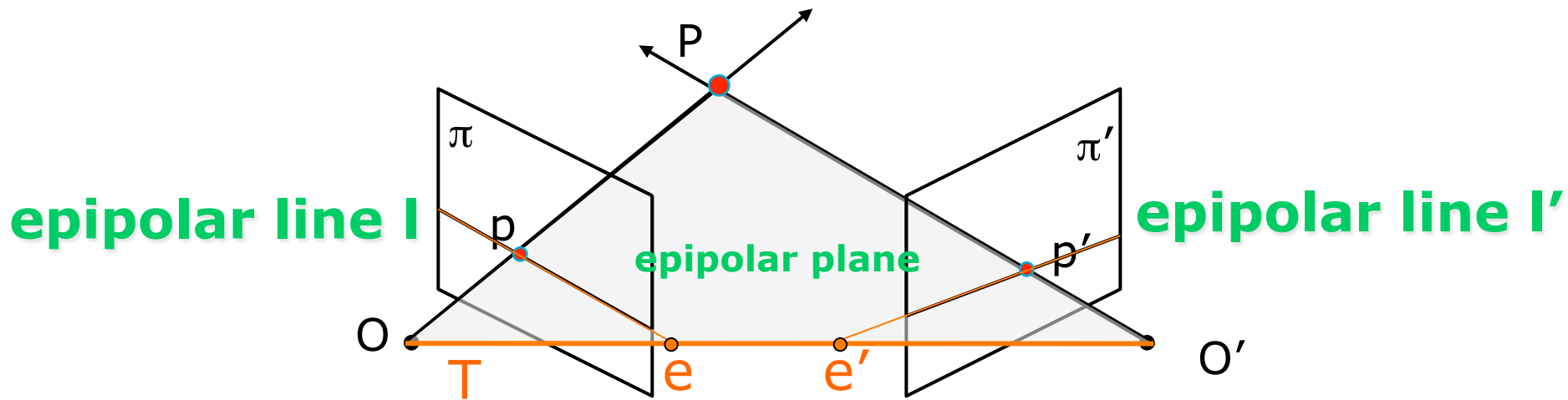
Epipoles



- The baseline T is the line between the 2 COPs O and O' . In verged cameras, this line intersects both plane π and π' .
- The epipole is the intersection of the baseline with the respective image plane.



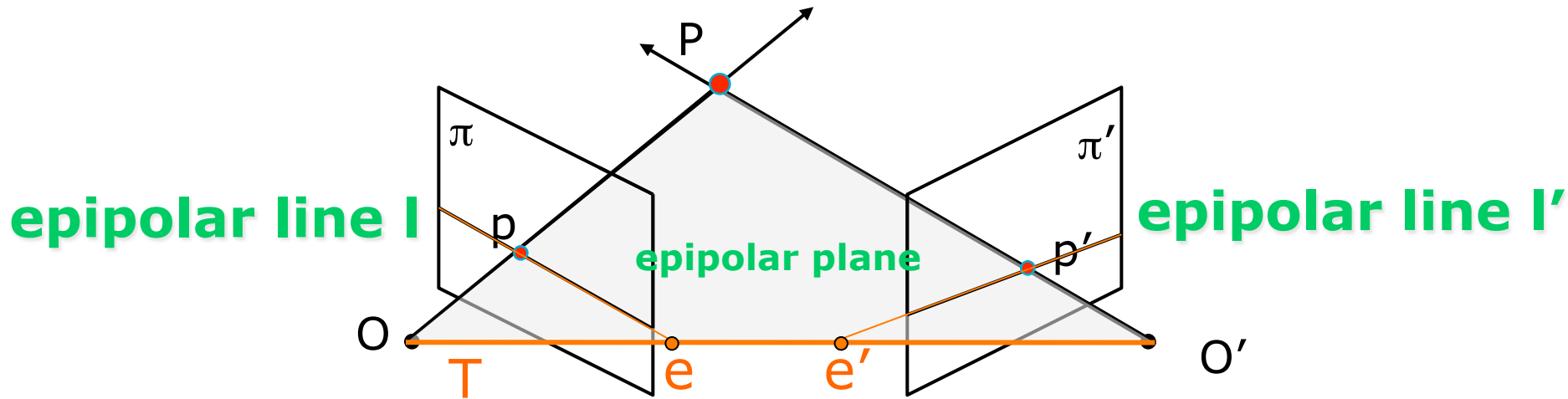
Epipolar Constraint



- The epipolar line l passes through the epipole e .
- The epipolar line l' passes through the epipole e' .
- If both p and p' are projections of the same point P , then p and p' must lie on the same epipolar plane. They must lie on epipolar lines l and l' respectively. This is called the **epipolar constraint**.



Impact of the Epipolar Constraint



- The epipolar constraint has a fundamental role in stereo and motion analysis.
- It reduces the correspondence problem to a 1D search along *conjugate epipolar lines*.
- Given an image point p , one needs to only search in the epipolar line l' for the corresponding point p' .



Example of Epipolar Lines

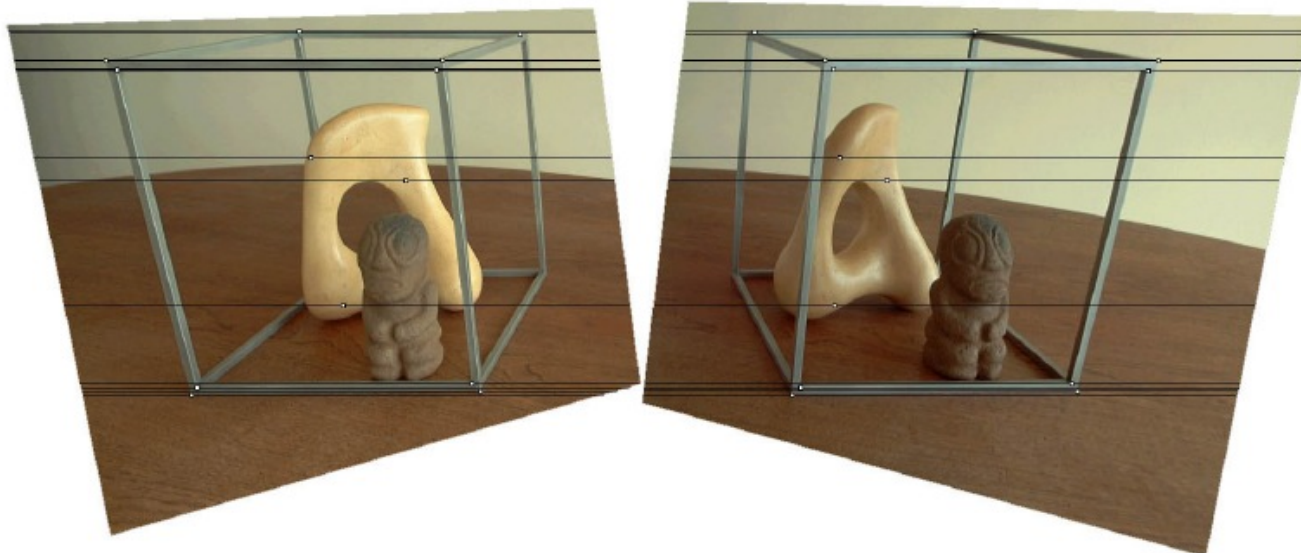
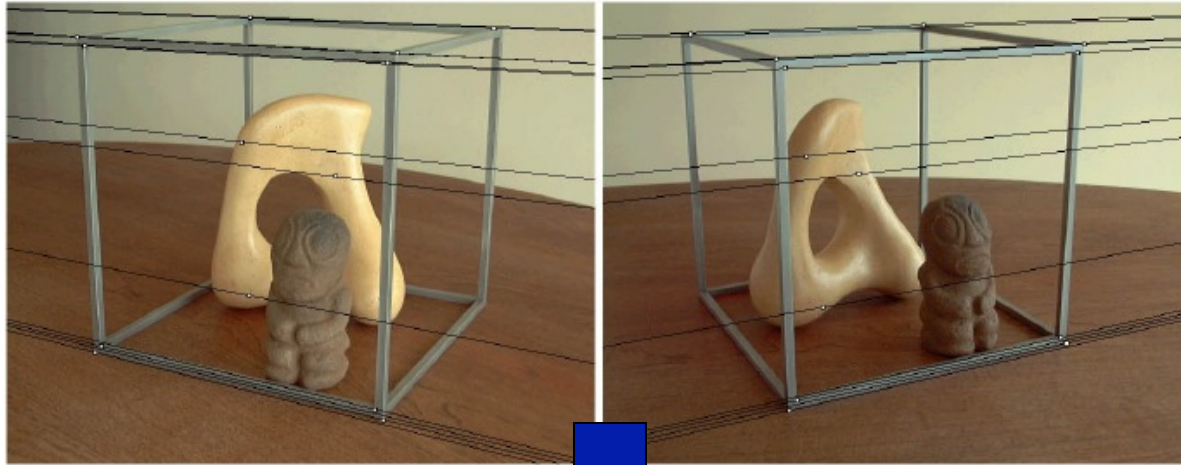


Figure courtesy of wikipedia,

<http://de.m.wikipedia.org/wiki/Datei:Epipolar-geometry-church-epipolar-lines.png>



Stereo Rectification Example





Required Knowledge

- In order to know the epipolar geometry, we need:
 - The location of the two COPs
 - The location of the two image planes
 - The orientation of the image planes
- We need to know the intrinsic and extrinsic camera characteristics.
- Intrinsic camera characteristics
 - Pixel size
 - Focal length
 - Principal point
- Extrinsic camera characteristics
 - The relative position of the 2 optical centers
 - The relative orientation of the two image planes

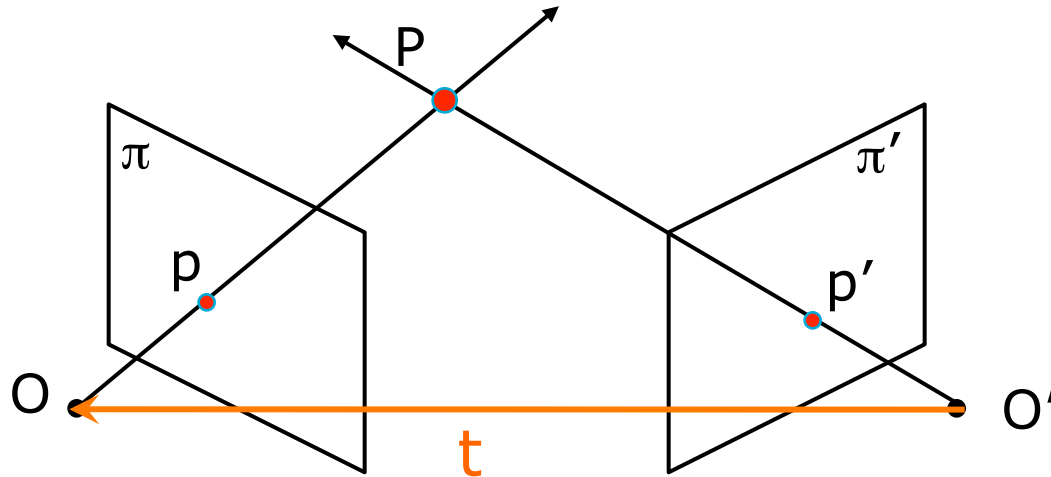


Epipolar Constraint – Calibrated Case

- Assume that the intrinsic parameters of each of the cameras are known, i.e. the mapping from the image coordinate system to a metric camera coordinate system.
- Goal: Express algebraically the epipolar constraint, so that it can be incorporated in our correspondence, stereo and motion algorithms.



Epipolar Plane Constraint



- The vectors Op , $O'p'$ and $O'O$ are all co-planar, i.e. they must satisfy the following equation:

$$\overrightarrow{Op} \cdot (\overrightarrow{O'O} \times \overrightarrow{O'p'}) = 0$$

- The vector Op is perpendicular to the vector resulting from the cross-product of $O'O$ and $O'p'$.

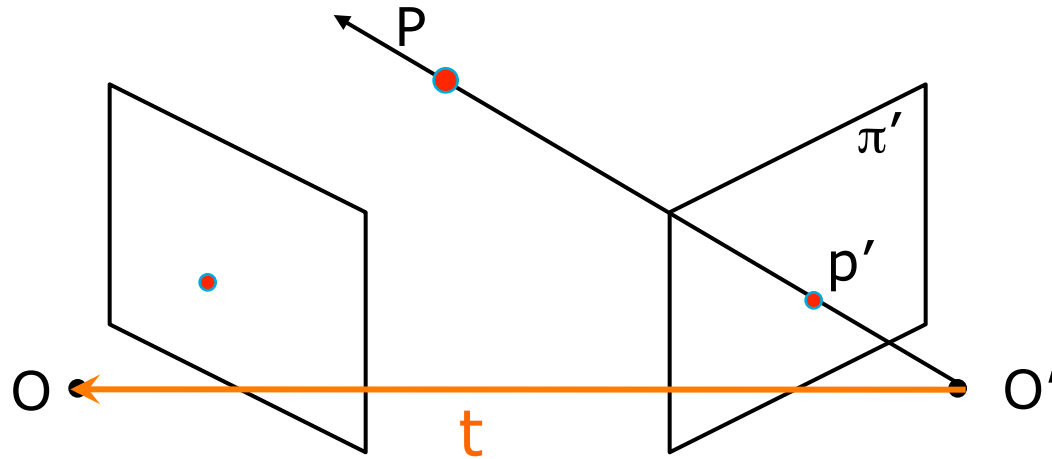


Relating the 2 Camera Coord. Systems

- Each image is unaware of the other camera.
- Point p is specified in the local coordinate system of the camera with COP O .
- Similarly point p' is specified in the local coordinate system of the camera with COP O' .
- We need to express everything in terms of a single coordinate system.
- Without loss of generality we choose as the reference coordinate system the one of the camera with COP O .



Translation

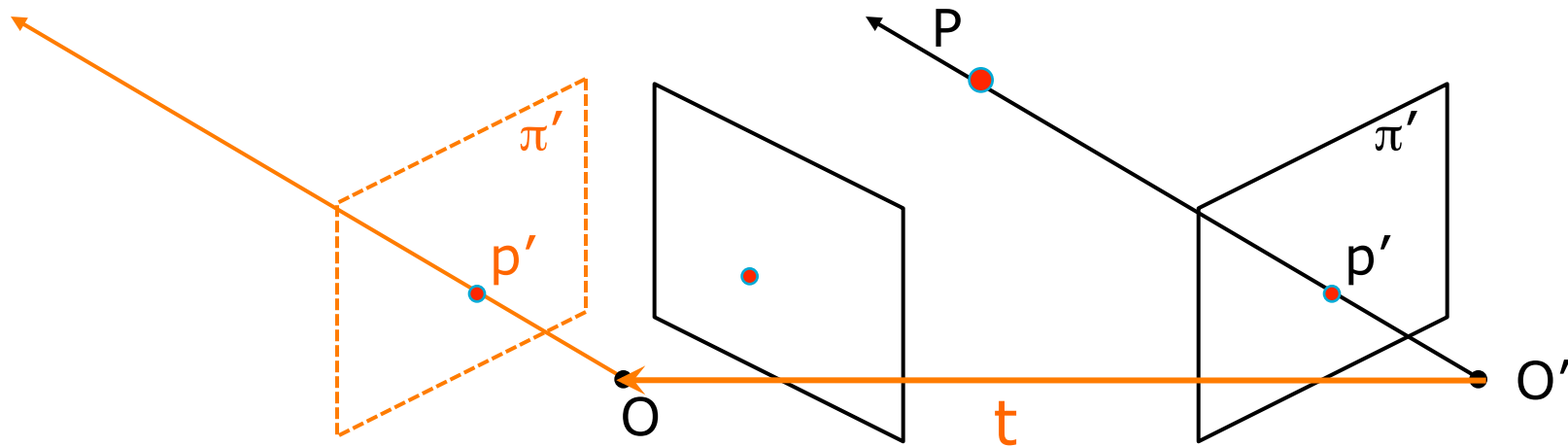


- There is a translation vector t , (the baseline T to be precise) that shows you how one can move COP O' to COP O .

$$\vec{t} = \overrightarrow{O'O}$$



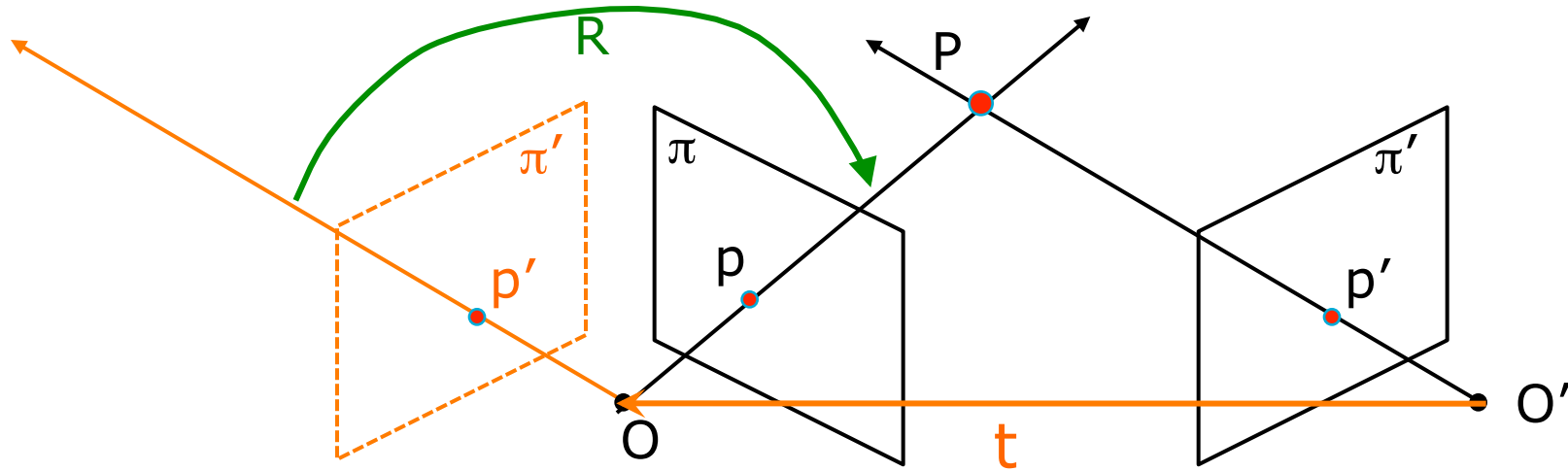
Need for Rotation



- If we apply this translation t to every point p' of the camera with COP O' then we will move the coordinate system with COP O' so that both camera coordinates are pinned to the same origin O .



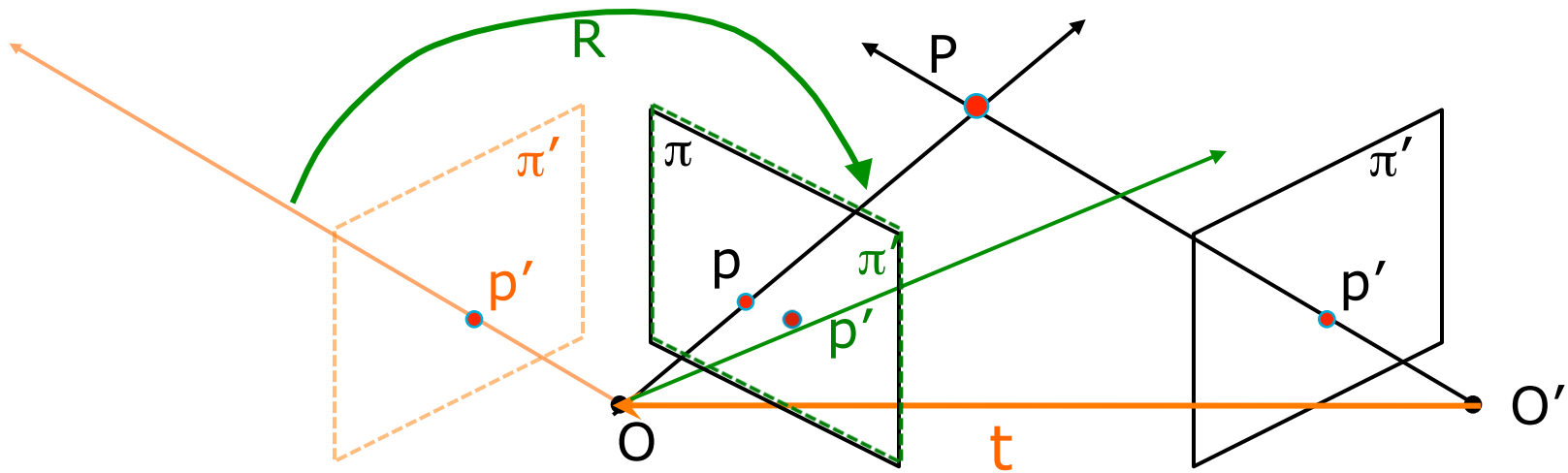
Rotation



- Still the two coordinate systems can differ by a rotation. Let R be the rotation matrix that aligns the corresponding axes of the two camera coordinates.



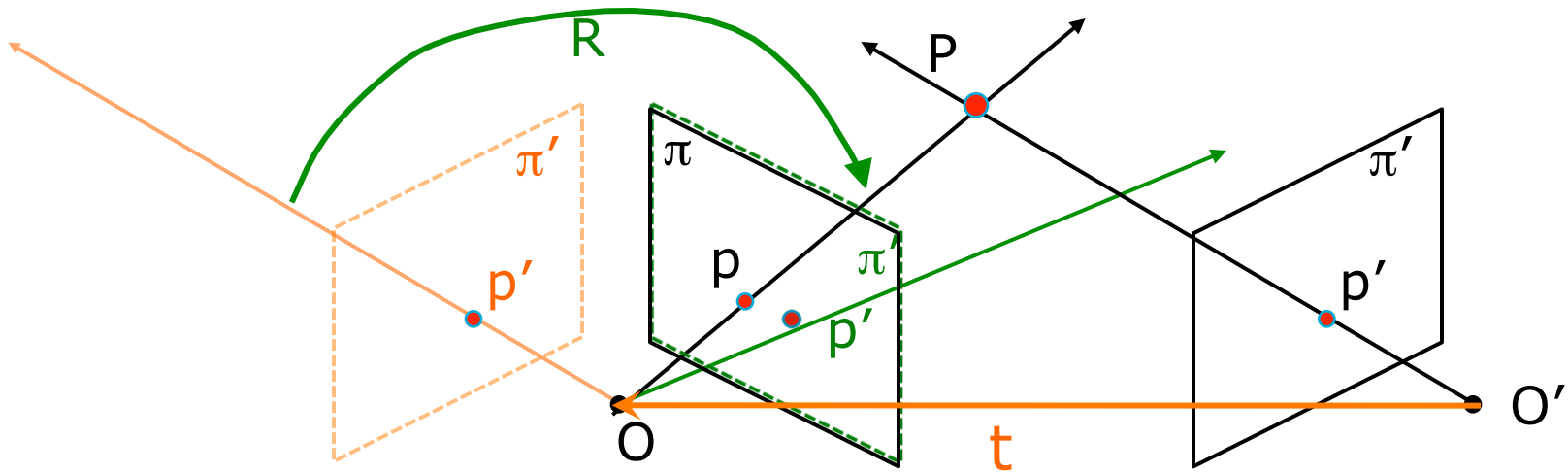
Translation and Rotation



- Each point p' after the translation from camera O' to camera O , is rotated by R .
- The two camera coordinate systems are now aligned.
- Everything can be expressed in terms of the coordinate system of camera O .



Epipolar Constraint Revisited



- Recall that vectors Op , $O'p'$ and $O'O$ are co-planar:

$$\overrightarrow{Op} \cdot (\overrightarrow{O'O} \times \overrightarrow{O'p'}) = 0$$

- Rewritten in the coordinate frame of camera O :

$$\vec{p} \cdot (\vec{t} \times (R\vec{p}')) = 0$$



Epipolar Constraint – Matrix Form

- The epipolar equation can be rewritten as a series of matrix multiplications:

$$\mathbf{p}^T (\mathbf{t} \times \mathbf{R}) \mathbf{p}' = 0$$

- This is often represented more compactly as:

$$\mathbf{p}^T \mathbf{E} \mathbf{p}' = 0$$

where \mathbf{E} is a 3x3 matrix of the form: $\mathbf{E} = [\mathbf{t}_\times] \mathbf{R}$
and it is known as the *essential matrix*.

$[\mathbf{t}_\times]$ is a skew-symmetric matrix such that $[\mathbf{t}_\times] \mathbf{b} = \mathbf{t} \times \mathbf{b}$

$[\mathbf{t}_\times]$ is the matrix representation of the cross product with \mathbf{t} .

$$\text{if } \mathbf{t} = \begin{bmatrix} t_x \\ t_y \\ t_z \end{bmatrix} \quad \text{then } [\mathbf{t}_\times] = \begin{bmatrix} 0 & -t_z & t_y \\ t_z & 0 & -t_x \\ -t_y & t_x & 0 \end{bmatrix}$$



Epipolar Constraint Equations

- The equation $\mathbf{p}^T \mathbf{E} \mathbf{p}' = 0$ is the algebraic representation of epipolar constraint.
- The vector that corresponds to the epipolar line l that is associated with point p' is $\mathbf{l} = \mathbf{E} \mathbf{p}'$.
- Similarly, the vector that corresponds to the epipolar line l' that is associated with point p is $\mathbf{l}' = \mathbf{E}^T \mathbf{p}$.
- Thus, once the essential matrix \mathbf{E} is recovered, one can reduce the search space for finding the corresponding points to a 1D space.



Epipolar Constraint –Uncalibrated case

- For uncalibrated cases, the matrices (rotation \mathbf{R} and translation \mathbf{t}) that express point p' in terms of the coordinate system of camera O must also incorporate the intrinsic camera parameters.
- Instead of $\mathbf{p}^T \mathbf{E} \mathbf{p}' = 0$ we have:

$$\mathbf{p}^T \mathbf{K}^{-T} \mathbf{E} \mathbf{K}'^{-1} \mathbf{p}' = 0$$

$$\mathbf{p}^T \mathbf{F} \mathbf{p}' = 0$$

where $\mathbf{F} = \mathbf{K}^{-T} \mathbf{E} \mathbf{K}'^{-1}$ and \mathbf{K} and \mathbf{K}' are the intrinsic parameter matrices of cameras O and O' accordingly

- \mathbf{F} is called the *fundamental matrix*.



Multiple Views

- For binocular setups the epipolar constraint can be represented in a 3×3 matrix form, called the *fundamental matrix*.
- When we have 3 images the epipolar constraint is represented by a $3 \times 3 \times 3$ structure, called the *trifocal tensor*.
- When we have 4 images the epipolar constraint is represented by a $3 \times 3 \times 3 \times 3$ structure, called the *quadrifocal tensor*.



Key Points of Epipolar Geometry

- For each pair of corresponding points p and p' in camera coordinates (Cartesian metric coordinate system), the following relationship holds:

$$\mathbf{p}^T \mathbf{E} \mathbf{p}' = 0$$

E is the essential matrix

- For each pair of corresponding points q and q' in pixel (image) coordinates the following relationship holds:

$$\mathbf{q}^T \mathbf{F} \mathbf{q}' = 0$$

F is the fundamental matrix



Key Points of Epipolar Geometry 2

- The epipolar line l' that corresponds to the point q has the form $l'_1x + l'_2y + l'_3z = 0$, where $\mathbf{l}' = (l'_1, l'_2, l'_3)$ and is given by:

$$\mathbf{l}' = \mathbf{F}^T \mathbf{q}$$

where x, y, z are in the local coordinate system of camera O' .

- The epipolar line l that corresponds to the point q' has the form $l_1x + l_2y + l_3z = 0$, where $\mathbf{l} = (l_1, l_2, l_3)$ and is given by:

$$\mathbf{l} = \mathbf{F} \mathbf{q}'$$

where x, y, z are in the local coordinate system of camera O .



The Essential Matrix in Practice

- What does the epipolar plane depend on? A point P in the scene and the camera COPs O and O' . It varies from point to point.
- What does the matrix \mathbf{E} (similarly \mathbf{F}) depend on? The rotation \mathbf{R} and the translation \mathbf{t} between the two camera coordinate systems. No dependence on the scene.
- So... recover \mathbf{E} (or \mathbf{F}) once, keep the camera setup stable and then reuse it for every scene point.
- How do we recover \mathbf{E} (or \mathbf{F})?



Estimation of the Fundamental Matrix.

- Assume known correspondences of n points between the two images.

- You have n equations of the form:

$$\mathbf{p}_i^T \mathbf{F} \mathbf{p}_i' = 0, \quad i = 1 \dots n$$

- \mathbf{F} is a 3x3 matrix \Rightarrow 9 unknowns.
- If you have 8 well spread correspondences, you can determine \mathbf{F} .
- Why 8? The n equations are homogeneous linear equations, i.e. all equations have a zero as a constant in the right hand side. So the solution is unique up to a scaling factor.



Over-determined System

- If $n > 8$, then we have an over-determined system. Use SVD (Singular Value Decomposition).
- How? Build a $n \times 9$ matrix \mathbf{A} which contains the coefficients of the n equations: $\mathbf{p}_i^T \mathbf{F} \mathbf{p}_i = 0$, $i = 1 \dots n$
- Run SVD on \mathbf{A} . It decomposes \mathbf{A} to: $\mathbf{A} = \mathbf{U} \mathbf{D} \mathbf{V}^T$
 - \mathbf{D} diagonal matrix; its elements are called singular values.
 - \mathbf{U} is an $n \times n$ orthogonal matrix
 - \mathbf{D} is an $n \times 9$ diagonal matrix
 - \mathbf{V} is a 9×9 orthogonal matrix
- In theory, the solution to \mathbf{F} (the value of its 9 unknowns) is the column of \mathbf{V} that corresponds to the only *null* singular value of \mathbf{A} , i.e. the only zero value on the diagonal.



Estimating \mathbf{F} in Practice

- In reality, due to noise, quantization, numerical errors, inaccuracies in the n correspondences, there is usually no null singular value.
- Thus, in practice we use the *minimum* singular value and its corresponding column in \mathbf{V} .

$$\mathbf{F} = \mathbf{V}(\text{Col}_m)$$

where s_m was the minimum diagonal value in \mathbf{D} and was located in column m in \mathbf{D} .



Estimating \mathbf{F} in Practice - continued

- However, this whole process had inaccuracies. The resulting \mathbf{F} may not be singular. So, run SVD again, this time on \mathbf{F} .

$$\mathbf{F} = \mathbf{U}_F \mathbf{D}_F \mathbf{V}_F^T$$

- Then build the matrix \mathbf{D}' from \mathbf{D}_F where with the minimum singular value s_m of \mathbf{D}_F is replaced by 0.
- Compute a new fundamental matrix which is singular:

$$\mathbf{F}' = \mathbf{U}_F \mathbf{D}' \mathbf{V}_F^T$$

- \mathbf{F}' is a good estimate of the fundamental matrix.



Longuet-Higgins Eight-Point Algorithm

1. Let \mathbf{A} be an $n \times 9$ matrix of the coefficients of the n eqs.:

$$\mathbf{p}_i^T \mathbf{F} \mathbf{p}_i = 0, \quad i = 1 \dots n$$

2. Apply SVD on \mathbf{A} and find matrices \mathbf{U} , \mathbf{D} , \mathbf{V} such that

$$\mathbf{A} = \mathbf{U} \mathbf{D} \mathbf{V}^T$$

3. The entries of \mathbf{F} are the components of the column of \mathbf{V} corresponding to the least singular value of \mathbf{A} .

4. Enforce the singularity constraint by applying SVD on \mathbf{F}

$$\mathbf{F} = \mathbf{U}_F \mathbf{D}_F \mathbf{V}_F^T$$

5. and creating $\mathbf{D}' = \mathbf{D}_F$ with the smallest singular value of \mathbf{D}_F replaced by 0.

6. Get new estimate of \mathbf{F} , call it \mathbf{F}' , such that

$$\mathbf{F}' = \mathbf{U}_F \mathbf{D}' \mathbf{V}_F^T$$

Fundamental Matrix Video



The video is courtesy of Daniel Wedge. You can view it at the following web-site:

<http://danielwedge.com/fmatrix/>