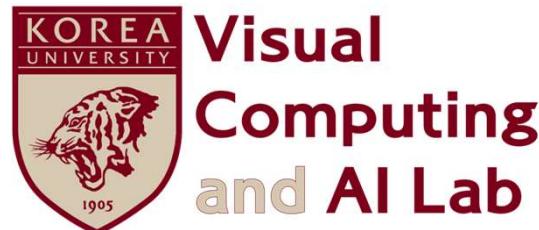


INTRODUCTION TO COMPUTER VISION

Lecture 5 – Epipolar Geometry

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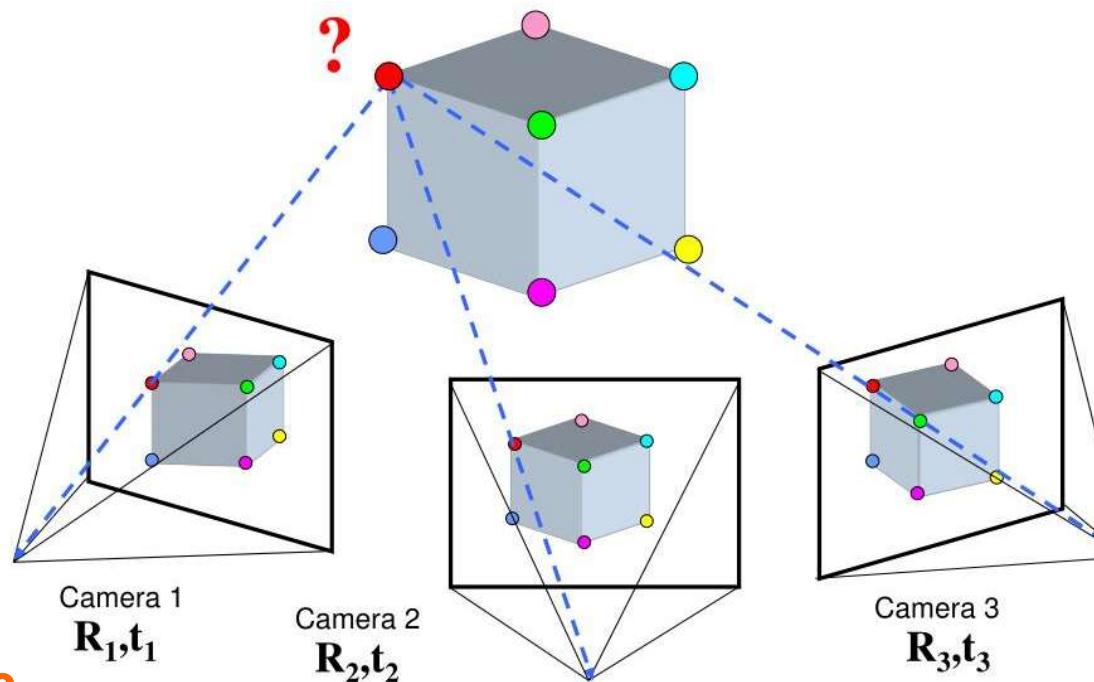


Slides Credit: [Prof. Dr. Davide Scaramuzza](#)

Multi-View Geometry (MVG) Theory

According to MVG (we'll learn this later), if we know

- Camera intrinsic/extrinsic parameters (we learnt what are they in prev. classes)
 - *Matched points across multiple viewpoints (same-colored dots in images)*
- , then, we can lift the multi-view observations to the 3D space



Slide from Lecture 3

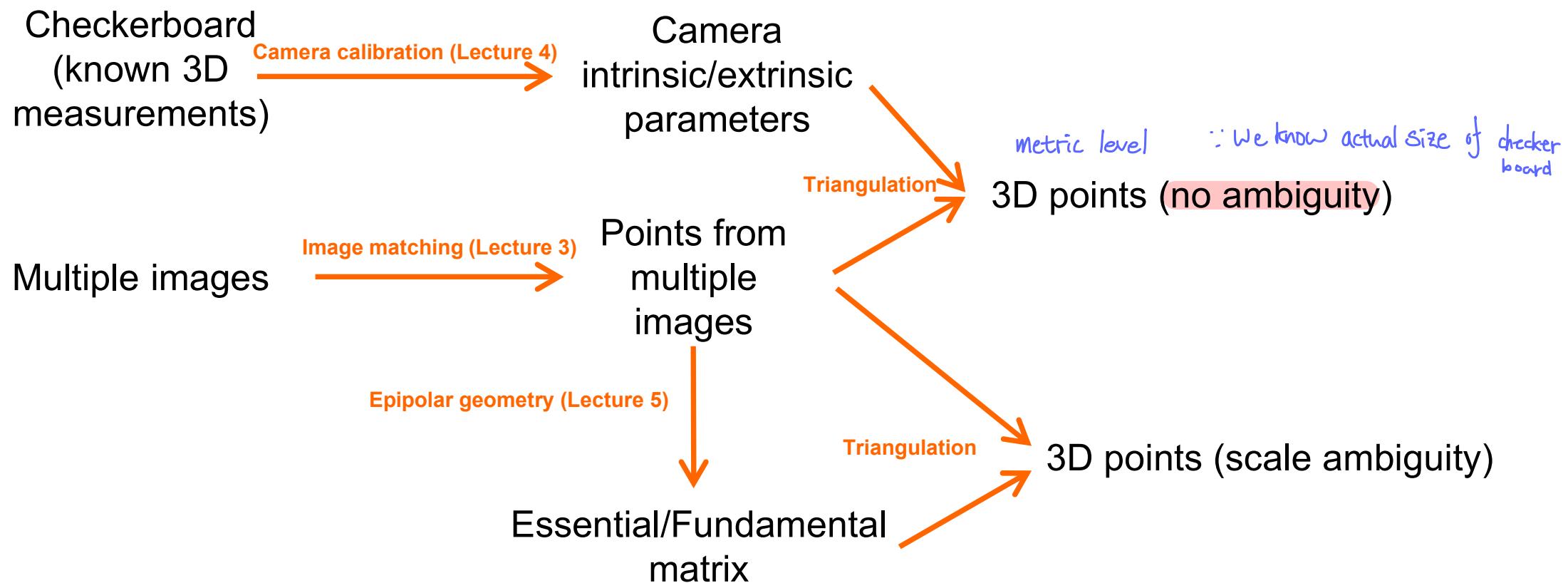
Epipolar Geometry

- We learned both 1) camera calibration to get camera parameters and 2) image matching to get coherent points from multiple images
- Finally, we can lift observations from multiple images to 3D space! *triangulation.*
- But what if we don't want to (or can't) calibrate cameras? *e.g. images from internet.*
 - E.g., taking a video with your mobile phone without camera calibration
- How? -> epipolar geometry
- Epipolar geometry: only two-view images
- Multi-view geometry: multi-view images (# of images ≥ 2)

Camera Calibration vs. Epipolar Geometry

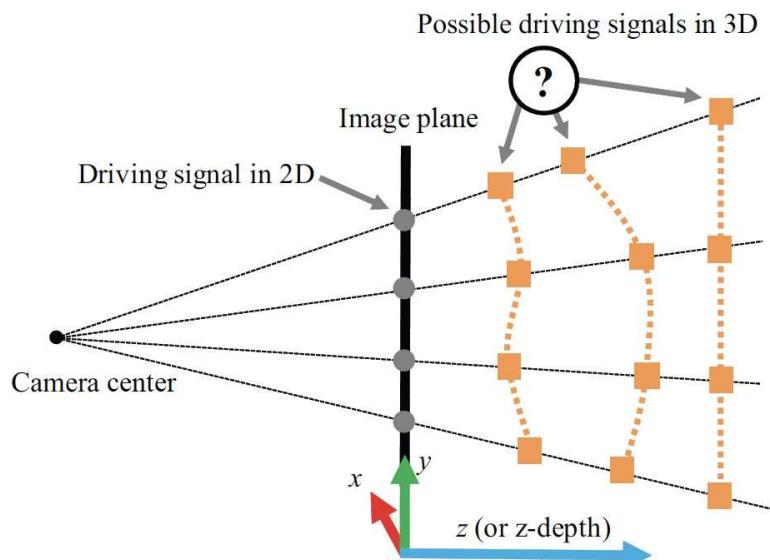
- If we **know extrinsics** with camera calibrations and checkerboard
 - E.g., you calibrate cameras with checkerboards
 - Skip all things we'll learn today
 - Just triangulate (we'll learn this later) 2D points to the 3D space
- If we **do not know extrinsics** (today's focus)
 - E.g., taking a video with your mobile phone without any calibrations/checkerboards
 - **Epipolar geometry!**
 - We get essential/fundamental matrices
 - We do not have checkerboard, which provides actual 3D measurements -> **we get extrinsics up to scale**

- All the entire pipeline: multi-view geometry theory



Recovering 3D Geometry

- Camera calibration from a single view
- Recover 3D geometry from a single view?
 - No: due to ambiguity of 3D \rightarrow 2D mapping



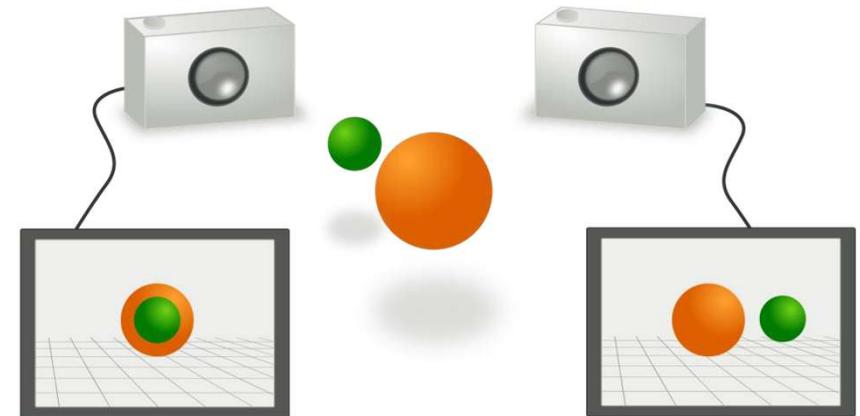
depth ambiguity



Epipolar Geometry

monocular – 1 view

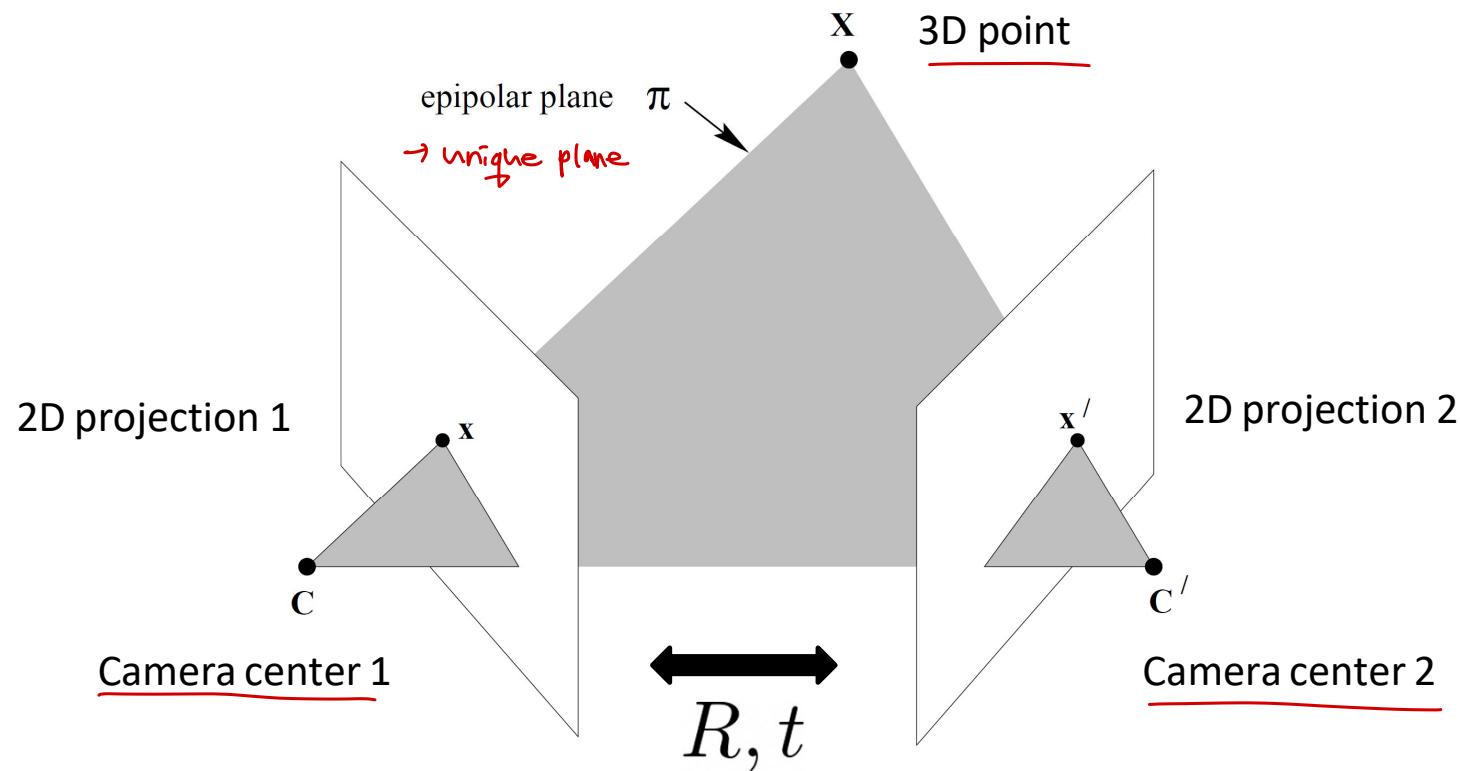
- The geometry of stereo vision
 - Given 2D images of two views
 - What is the relationship between pixels of the images?
 - Can we recover the 3D structure of the world from the 2D images?



Wikipedia

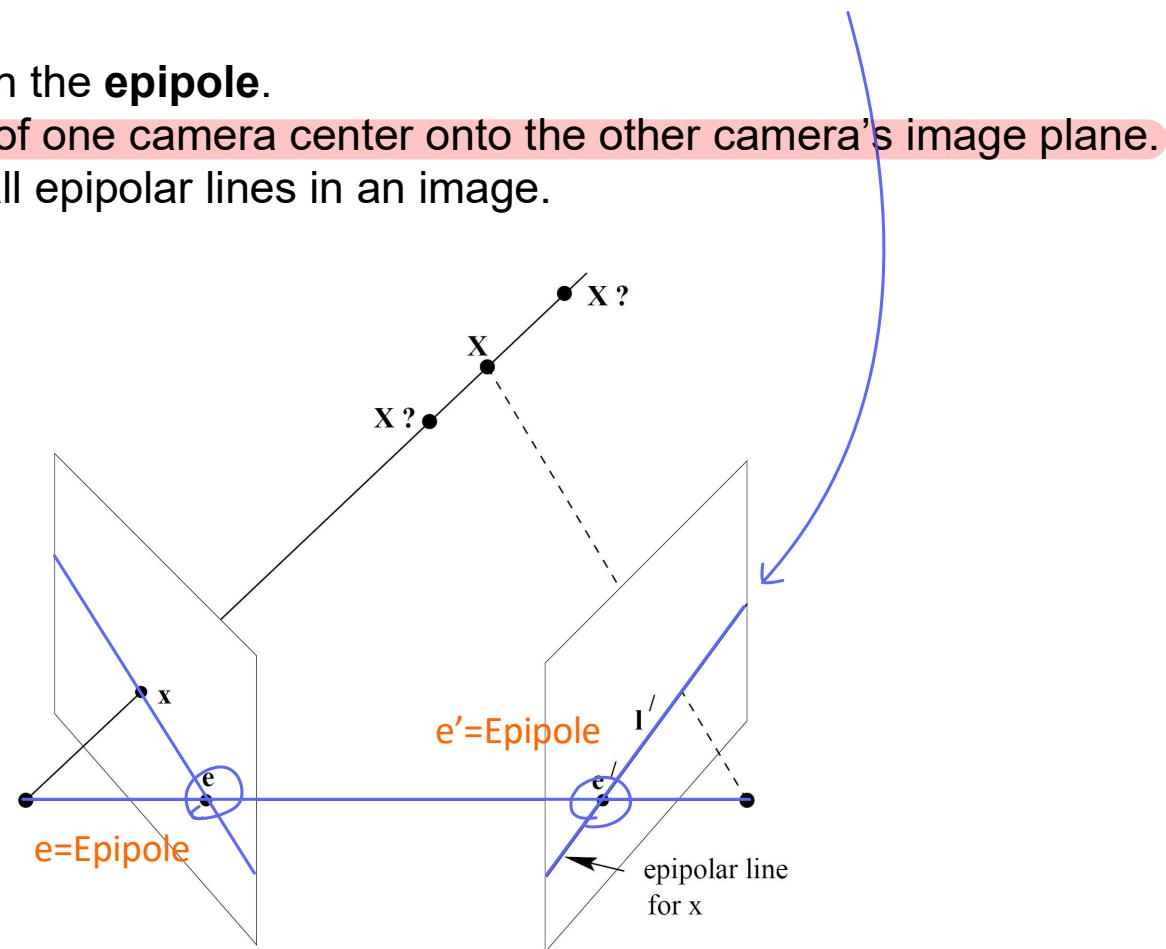
Epipolar Geometry

- The epipolar plane is defined by two camera centers and a 3D point.
uniquely



Epipolar Geometry

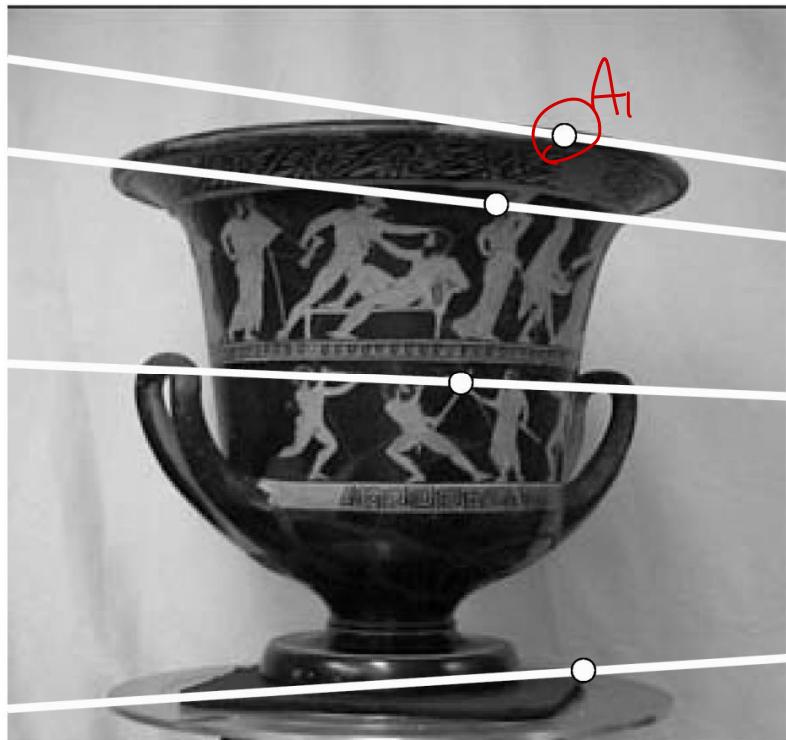
- The **epipolar plane** is defined by two camera centers and a 3D point.
- Its intersection with an image plane gives the corresponding **epipolar line**.
- All epipolar lines pass through the **epipole**.
- The epipole is the **projection of one camera center onto the other camera's image plane**.
- It is the intersection point of all epipolar lines in an image.



Epipolar Geometry

- If we know 1) the intrinsic and extrinsic parameters of two cameras, and 2) the 2D image position of a point in one camera, then the corresponding epipolar line in the other camera defines the locus of all potential 2D positions where the point could appear. *지 line 위에서만 찾으면 됨*

location ↗



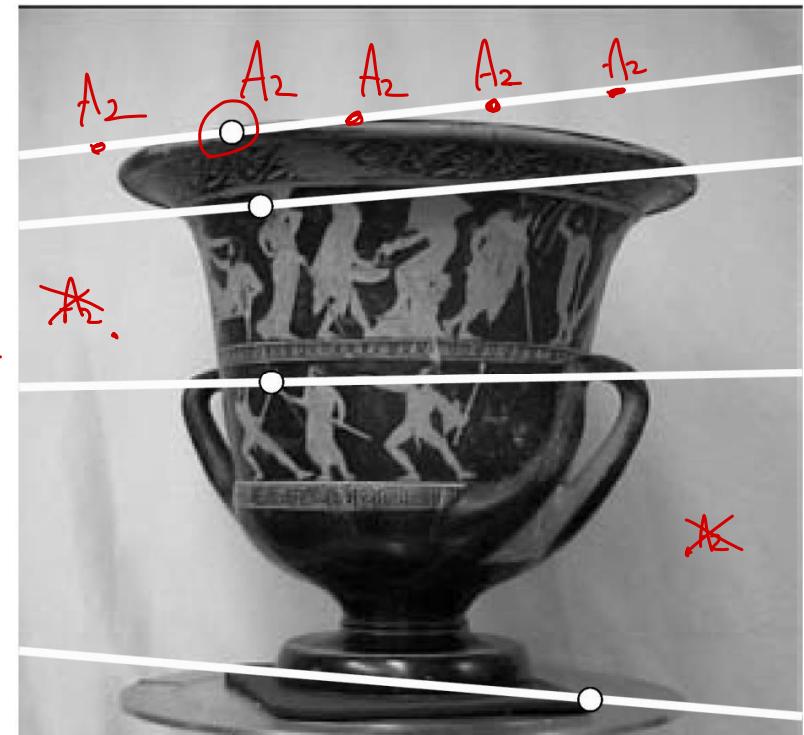
I_1

Epipolar lines

A_1 의 corresponding
점은 A_2 는 Epipolar
line 위에만 있을 수 있음.



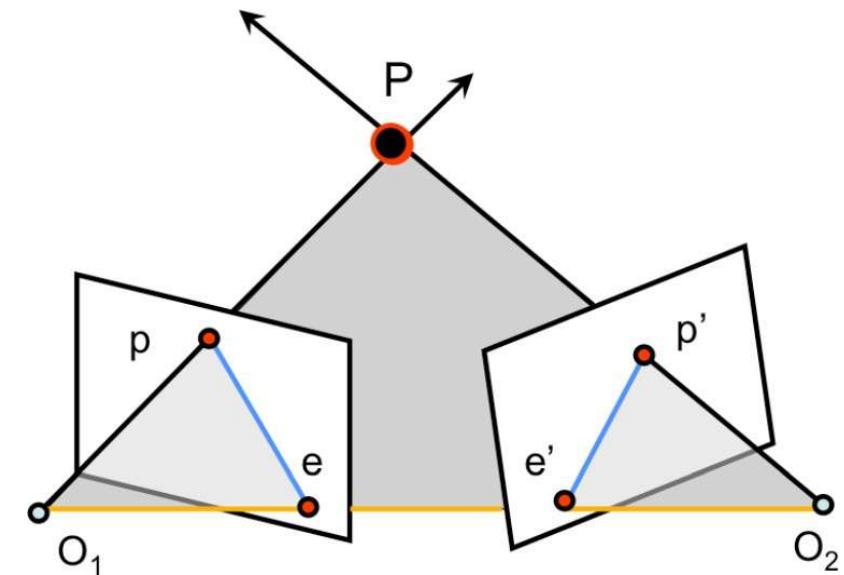
Rotation and Translation
between two views



I_2

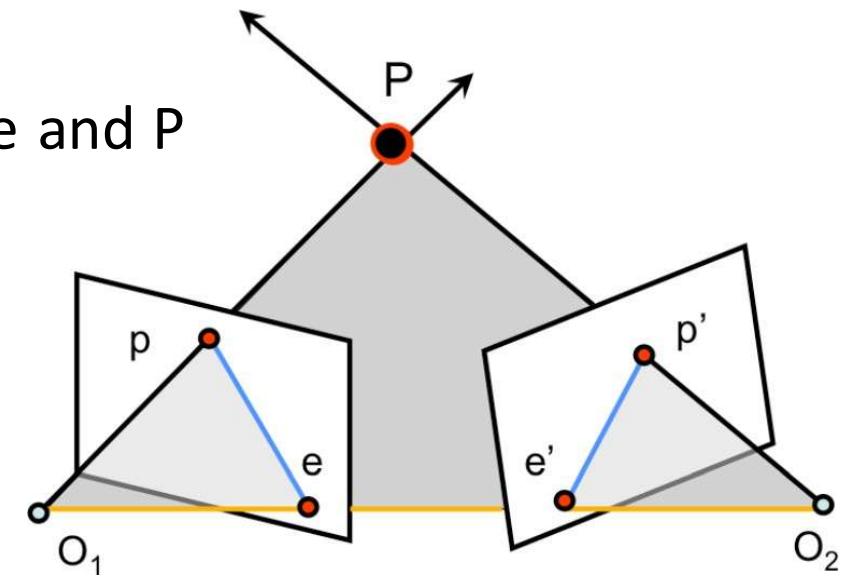
Epipolar Geometry

- Baseline (Yellow line)
 - The line between the two camera centers O_1 and O_2



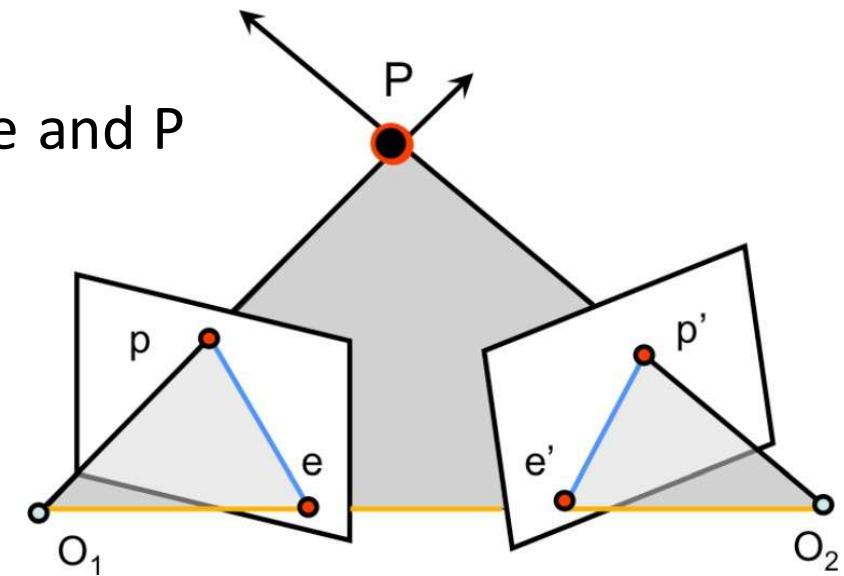
Epipolar Geometry

- Baseline (Yellow line)
 - The line between the two camera centers O_1 and O_2
- Epipolar plane (gray plane)
 - Defined by P , O_1 , and O_2 ; contains baseline and P



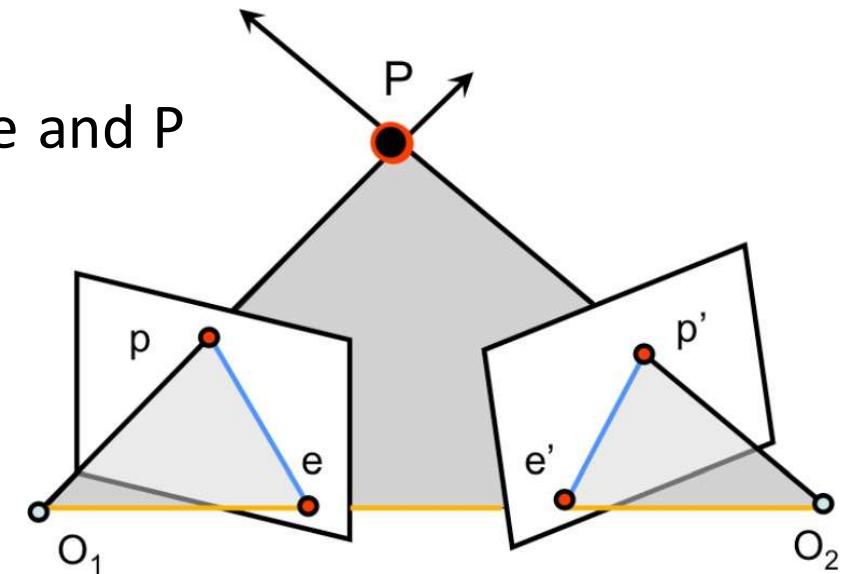
Epipolar Geometry

- Baseline (Yellow line)
 - The line between the two camera centers O_1 and O_2
- Epipolar plane (gray plane)
 - Defined by P , O_1 , and O_2 ; contains baseline and P
- Epipoles (e and e')
 - \cap of baseline and image plane: e and e'
 - Projection of the other camera center



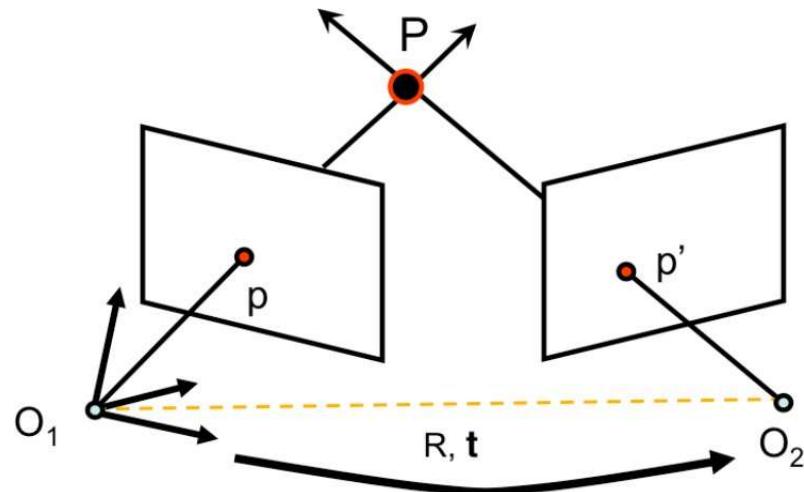
Epipolar Geometry

- Baseline (Yellow line)
 - The line between the two camera centers O_1 and O_2
- Epipolar plane (gray plane)
 - Defined by P , O_1 , and O_2 ; contains baseline and P
- Epipoles (e and e')
 - \cap of baseline and image plane
 - Projection of the other camera center
- Epipolar lines (Blue lines)
 - \cap of epipolar plane with the image plane



Epipolar Constraint

- The relationship between corresponding image points
 - The world reference system aligned with the left camera
 - The right camera has orientation R and offset t



Camera projection matrices

Left camera

$$M = K[I \ 0]$$

$$\mathbf{p} = M\mathbf{P} = \begin{bmatrix} u \\ v \\ 1 \end{bmatrix}$$

Right camera

$$M' = K'[R \ t]$$

$$\mathbf{p}' = M'\mathbf{P} = \begin{bmatrix} u' \\ v' \\ 1 \end{bmatrix}$$

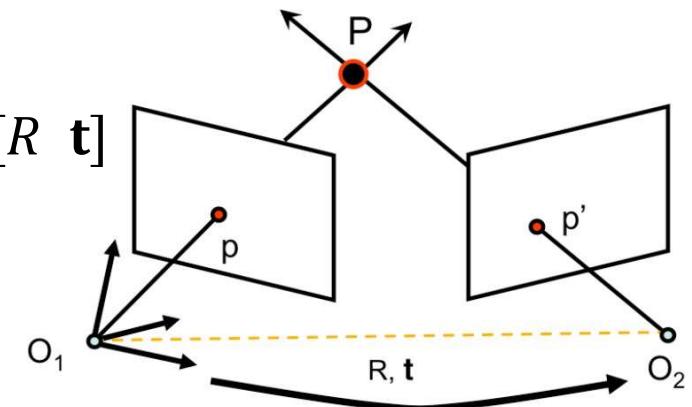
extrinsic

Epipolar Constraint

- The relationship between corresponding image points
 - Canonical cameras ($K = K' = I$)

$$M = K[I \ 0] \rightarrow M = [I \ 0] \quad M' = K'[R \ t] \rightarrow M' = [R \ t]$$

p' in world coordinate system



$$p' = RX + t \quad X: p' \text{ in world}$$

$$R^{-1}(p' - t) = X$$

Epipolar Constraint

$R \circ I$ orthogonal \Rightarrow

$$\|R_x\| = \|x\|$$

$$R^T R = R R^T = I$$

$$R^T = R^{-1}$$

$$(R_x) \cdot (R_y) = x \cdot y$$

- The relationship between corresponding image points

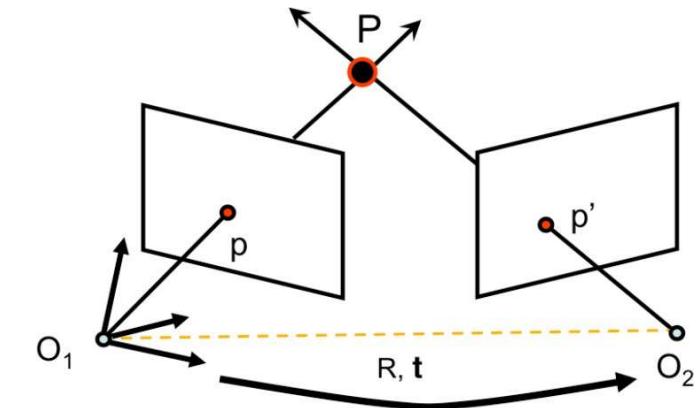
- Canonical cameras ($K = K' = I$)

$$M = [I \ 0] \quad M' = [R \ t]$$

p' in world coordinate system

$$R^T(p' - t)$$

O_2 in world coordinate system



Epipolar Constraint

- The relationship between corresponding image points
 - Canonical cameras ($K = K' = I$)

$$M = [I \ 0] \quad M' = [R \ t]$$

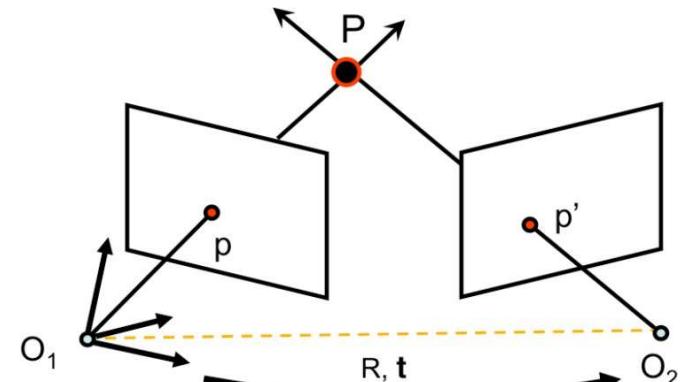
p' in world coordinate system

$$R^T(\mathbf{p}' - \mathbf{t})$$

\mathbf{O}_2 in world coordinate system

$$R^T(\mathbf{O}_2 - \mathbf{t}) = -R^T\mathbf{t}$$

Normal of the epipolar plane



Epipolar Constraint

- The relationship between corresponding image points
 - Canonical cameras ($K = K' = I$)

$$M = [I \ 0] \quad M' = [R \ t]$$

p' in world coordinate system

$$R^T(\mathbf{p}' - \mathbf{t})$$

\mathbf{O}_2 in world coordinate system

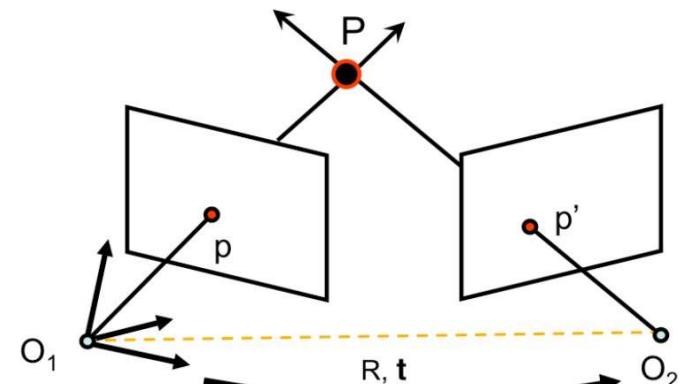
$$R^T(\mathbf{O}_2 - \mathbf{t}) = -R^T\mathbf{t}$$

Normal of the epipolar plane

$$R^T\mathbf{t} \times [R^T(\mathbf{p}' - \mathbf{t})] = R^T(\mathbf{t} \times \mathbf{p}')$$

의미

$\mathbf{O}_1\mathbf{p}$ lies in the epipolar plane: so the dot product of the normal of the Epipolar plane and $\mathbf{O}_1\mathbf{p}$ is 0



Epipolar Constraint

- The relationship between corresponding image points
 - Canonical cameras ($K = K' = I$)

$$M = [I \ 0] \quad M' = [R \ t]$$

p' in world coordinate system

$$R^T(\mathbf{p}' - \mathbf{t})$$

\mathbf{O}_2 in world coordinate system

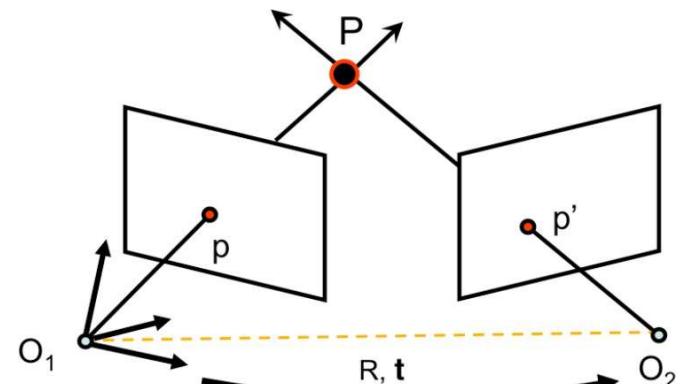
$$R^T(\mathbf{O}_2 - \mathbf{t}) = -R^T\mathbf{t}$$

Normal of the epipolar plane

$$R^T\mathbf{t} \times [R^T(\mathbf{p}' - \mathbf{t})] = R^T(\mathbf{t} \times \mathbf{p}')$$

$\mathbf{O}_1\mathbf{p}$ lies in the epipolar plane

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



Epipolar Constraint

- The relationship between corresponding image points
 - Canonical cameras ($K = K' = I$)

$$M = [I \ 0] \quad M' = [R \ t]$$

p' in world coordinate system

$$R^T(\mathbf{p}' - \mathbf{t})$$

\mathbf{O}_2 in world coordinate system

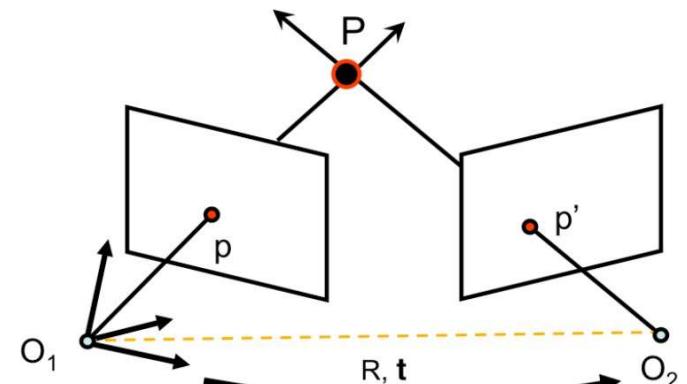
$$R^T(\mathbf{O}_2 - \mathbf{t}) = -R^T\mathbf{t}$$

Normal of the epipolar plane

$$R^T\mathbf{t} \times [R^T(\mathbf{p}' - \mathbf{t})] = R^T(\mathbf{t} \times \mathbf{p}')$$

$\mathbf{O}_1\mathbf{p}$ lies in the epipolar plane

$$[R^T(\mathbf{t} \times \mathbf{p}')]^T\mathbf{p} = 0 \xrightarrow{\text{blue arrow}} (\mathbf{t} \times \mathbf{p}')^T R \mathbf{p} = 0$$



Epipolar Constraint

- The relationship between corresponding image points

– Canonical cameras ($K = K' = I$)

$$M = [I \ 0] \quad M' = [R \ t]$$

\mathbf{p}' in world coordinate system

$$R^T(\mathbf{p}' - \mathbf{t})$$

\mathbf{O}_2 in world coordinate system

$$R^T(\mathbf{O}_2 - \mathbf{t}) = -R^T\mathbf{t}$$

Normal of the epipolar plane

$$R^T\mathbf{t} \times [R^T(\mathbf{p}' - \mathbf{t})] = R^T(\mathbf{t} \times \mathbf{p}')$$

$\mathbf{O}_1\mathbf{p}$ lies in the epipolar plane

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

Cross product as matrix-vector multiplication

$$\mathbf{a} \times \mathbf{b} = \begin{bmatrix} 0 & -a_z & a_y \\ a_z & 0 & -a_x \\ -a_y & a_x & 0 \end{bmatrix} \begin{bmatrix} b_x \\ b_y \\ b_z \end{bmatrix} = [\mathbf{a}]^T \mathbf{b}$$

$[\mathbf{a}]^T = -[\mathbf{a}]^T$
Skew-symmetric matrix (rank=2)

$$\mathbf{a}^T = -\mathbf{a}$$

↑
rank 2

Epipolar Constraint

- The relationship between corresponding image points
 - Canonical cameras ($K = K' = I$)

$$M = [I \ 0] \quad M' = [R \ t]$$

p' in world coordinate system

$$R^T(p' - t)$$

O_2 in world coordinate system

$$R^T(O_2 - t) = -R^Tt$$

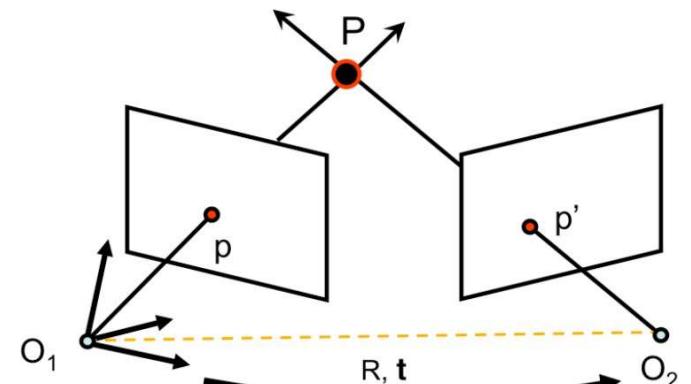
Normal of the epipolar plane

$$R^Tt \times [R^T(p' - t)] = R^T(t \times p')$$

O_1p lies in the epipolar plane

$$[R^T(t \times p')]^T p = 0 \xrightarrow{\text{blue arrow}} (t \times p')^T R p = 0 \xrightarrow{\text{blue arrow}} ([t \times p'])^T R p = 0$$

$$\xrightarrow{\text{blue arrow}} p'^T [t \times] R p = 0$$



Epipolar Constraint

$$\lambda \cdot p'^T E p = 0$$

$$p'^T (\lambda E) p = 0 \quad \lambda = 1, 2, \dots ?$$

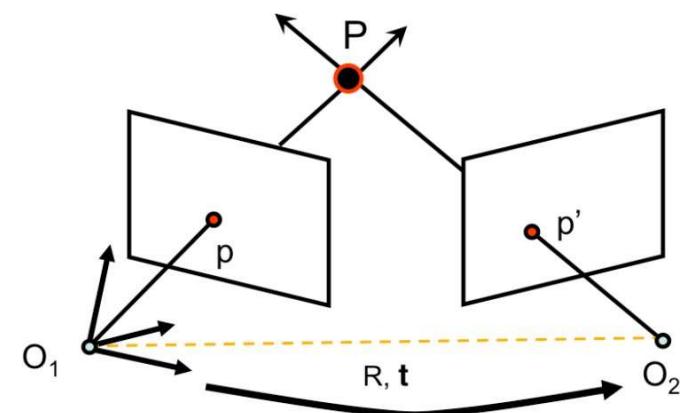
- Essential matrix
 - Establish constraints between matching image points
 - Determine relative position and orientation of two cameras
 - 5 degrees of freedom (R : 3, t : 3, but scale is not known)

$$p'^T \begin{bmatrix} t \\ x \end{bmatrix} R p = 0 \quad b-l=5$$

$$\downarrow \quad E = \begin{bmatrix} t \\ x \end{bmatrix} R$$

$$p'^T E p = 0$$

Essential matrix



Epipolar Constraint

- How to generalize Essential matrix?
 - Canonical cameras → general cameras

$$\begin{array}{c} M = K[I \ 0] \\ M' = K'[R \ t] \end{array} \xrightarrow{\text{Essential}} \begin{array}{c} K = K' = I \\ M = [I \ 0] \\ M' = [R \ t] \end{array} \xrightarrow{\quad} \begin{array}{c} p = MP =^2 [I \ 0]P \\ p' = M'P = [R \ t]P \end{array} \xrightarrow{\quad} \begin{array}{c} p'^T E p = 0 \\ E = [\mathbf{t}_x] R \end{array}$$

Fundamental

$K \neq I, K' \neq I$

?

Epipolar Constraint

- How to generalize Essential matrix?
 - Canonical cameras → general cameras

Canonical cameras: the image points in homogeneous coordinates are actually the 3D point expressed in the camera coordinate system

Epipolar Constraint

- Essential matrix vs. Fundamental matrix
 - Similarity
 - Both relate the matching image points
 - – Encode epipolar geometry of two views & camera parameters
 - Differences
 - *E* (essential matrix) encodes only the camera extrinsic parameter
 - *F* (fundamental matrix) also encodes the intrinsic parameters

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

$$E = [\mathbf{t}_\times]R$$

Essential matrix

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

$$F = K'^{-T} [\mathbf{t}_\times] R K^{-1}$$

Fundamental matrix

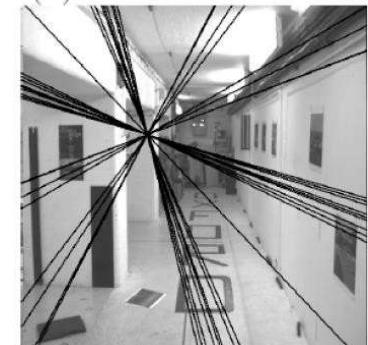
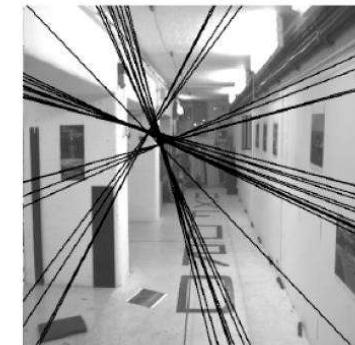
Epipolar Constraint

- Properties of the Fundamental matrix

- 3 by 3
- homogeneous (has scale ambiguity)
- $\text{rank}(F) = 2$
 - The potential matching point is located on a line
- F has 7 degrees of freedom ($3 \times 3 - 1$ (rank 2) – 1 (scale ambiguity)) = 7

$$\mathbf{p}'^T F \mathbf{p} = 0 \quad F = K'^{-T} [\mathbf{t}]_x R K^{-1}$$

Fundamental matrix has rank 2 : $\det(F) = 0$.



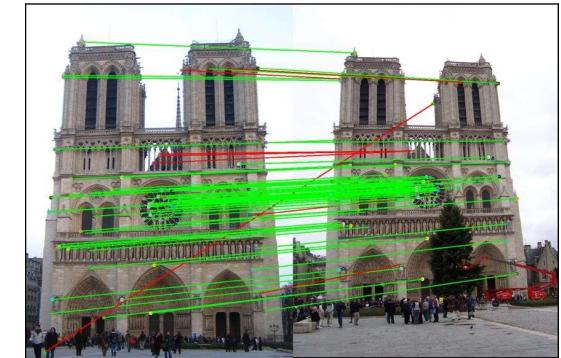
Left: Uncorrected F – epipolar lines are not coincident.

Right: Epipolar lines from corrected F .

Epipolar Constraint

- How can we use the fundamental matrix?

- Given a 2D point in one image and fundamental matrix, we can get epipolar line from the other image (candidate positions of matched points)
- Without knowing camera intrinsic and extrinsic parameters
- If we know p , epipolar line is $l = Fp$
- If we know p' , epipolar line is $l' = F^T p'$
- With the epipolar line, we can reduce the matching space
 - Not entire image, only search points on the epipolar line
 - Find matched points and lift them to the 3D space with triangulation



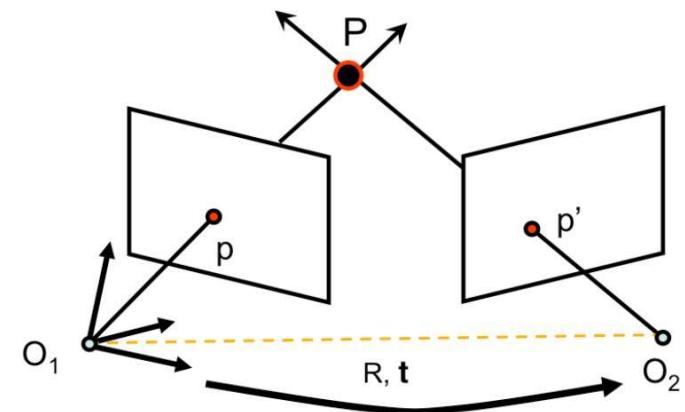
$$p'^T F p = 0 \quad F = K'^{-T} [\begin{matrix} \mathbf{t}_x \\ \mathbf{t}_y \end{matrix}] R K^{-1}$$

Recovering Fundamental Matrix

- How to recover F ?
 - From image correspondences

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

$$F = K'^{-T} [\mathbf{t}_x] R K^{-1}$$

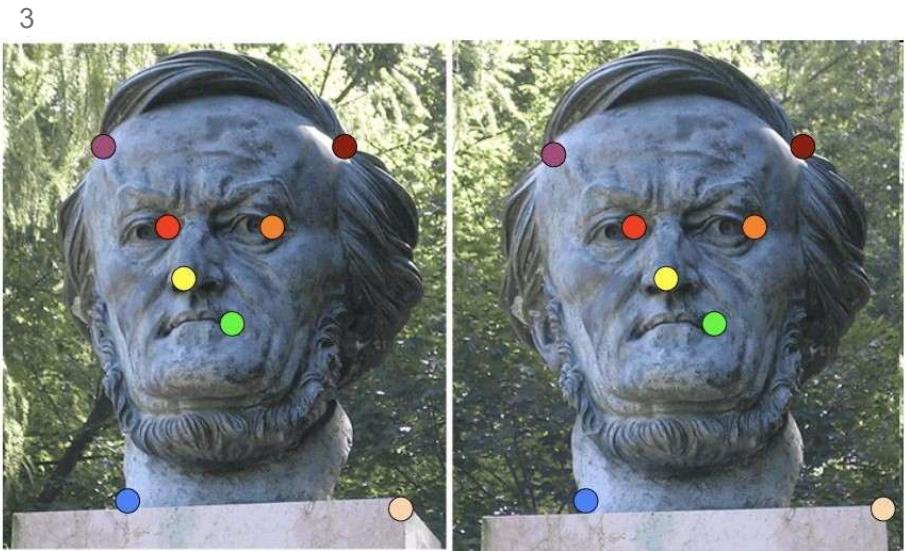


Recovering Fundamental Matrix

- How to recover F ?
 - From image correspondences
 - How many point pairs needed?



$$\mathbf{p}'^T F \mathbf{p} = 0 \quad F = K'^{-T} [\mathbf{t}_x] R K^{-1}$$



Recovering Fundamental Matrix

- How to recover F ?
 - From image correspondences
 - 8-point pairs required
 - Each point pair gives one equation
 - F is known up to scale
 - The linear system is homogeneous

$$\begin{cases} \mathbf{p}_i = (u_i, v_i, 1) \\ \mathbf{p}'_i = (u'_i, v'_i, 1) \end{cases}$$

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

$$\begin{bmatrix} u_i u'_i & v_i u'_i & u'_i & u_i v'_i & v_i v'_i & v'_i & u_i & v_i & 1 \end{bmatrix} \begin{bmatrix} F_{11} \\ F_{12} \\ F_{13} \\ F_{21} \\ F_{22} \\ F_{23} \\ F_{31} \\ F_{32} \\ F_{33} \end{bmatrix} = 0$$

↓

Recovering Fundamental Matrix

- How to recover F ?
 - From image correspondences
 - 8-point pairs required
 - Each point pair gives one equation
 - F is known up to scale
 - The linear system is homogeneous

$$\begin{cases} \mathbf{p}_i = (u_i, v_i, 1) \\ \mathbf{p}'_i = (u'_i, v'_i, 1) \end{cases}$$

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

$$\begin{bmatrix} u_i u'_i & v_i u'_i & u'_i & u_i v'_i & v_i v'_i & v'_i & u_i & v_i & 1 \end{bmatrix} \begin{bmatrix} F_{11} \\ F_{12} \\ F_{13} \\ F_{21} \\ F_{22} \\ F_{23} \\ F_{31} \\ F_{32} \\ F_{33} \end{bmatrix} = 0$$



Recovering Fundamental Matrix

F has 7 degrees of freedom Are 7-point pairs sufficient?

- Yes in theory
- We need to use $\det(F)=0$ as an additional constraint to use only 7 points
- $\det(F)=0$ gives us cubic equation, so there could be more than one solution
 ↳ non-linear
- Unstable, so we simply use 8 points

Recovering Fundamental Matrix

- 8-point algorithm

$$[u_i u'_i \ v_i u'_i \ u'_i \ u_i v'_i \ v_i v'_i \ v'_i \ u_i \ v_i \ 1]$$

$$\begin{bmatrix} F_{11} \\ F_{12} \\ F_{13} \\ F_{21} \\ F_{22} \\ F_{23} \\ F_{31} \\ F_{32} \\ F_{33} \end{bmatrix}$$

$$= 0$$

$$\boxed{\begin{bmatrix} u_1 u'_1 & v_1 u'_1 & u'_1 & u_1 v'_1 & v_1 v'_1 & v'_1 & u_1 & v_1 & 1 \\ u_2 u'_2 & v_2 u'_2 & u'_2 & u_2 v'_2 & v_2 v'_2 & v'_2 & u_2 & v_2 & 1 \\ u_3 u'_3 & v_3 u'_3 & u'_3 & u_3 v'_3 & v_3 v'_3 & v'_3 & u_3 & v_3 & 1 \\ u_4 u'_4 & v_4 u'_4 & u'_4 & u_4 v'_4 & v_4 v'_4 & v'_4 & u_4 & v_4 & 1 \\ u_5 u'_5 & v_5 u'_5 & u'_5 & u_5 v'_5 & v_5 v'_5 & v'_5 & u_5 & v_5 & 1 \\ u_6 u'_6 & v_6 u'_6 & u'_6 & u_6 v'_6 & v_6 v'_6 & v'_6 & u_6 & v_6 & 1 \\ u_7 u'_7 & v_7 u'_7 & u'_7 & u_7 v'_7 & v_7 v'_7 & v'_7 & u_7 & v_7 & 1 \\ u_8 u'_8 & v_8 u'_8 & u'_8 & u_8 v'_8 & v_8 v'_8 & v'_8 & u_8 & v_8 & 1 \end{bmatrix} \begin{bmatrix} F_{11} \\ F_{12} \\ F_{13} \\ F_{21} \\ F_{22} \\ F_{23} \\ F_{31} \\ F_{32} \\ F_{33} \end{bmatrix}} = 0$$

Recovering Fundamental Matrix

- 8-point algorithm
 - Construct linear system using corresponding image points

$$W\mathbf{f} = 0$$



How to solve it?

$$\begin{bmatrix} u_1u'_1 & v_1u'_1 & u'_1 & u_1v'_1 & v_1v'_1 & v'_1 & u_1 & v_1 & 1 \\ u_2u'_2 & v_2u'_2 & u'_2 & u_2v'_2 & v_2v'_2 & v'_2 & u_2 & v_2 & 1 \\ u_3u'_3 & v_3u'_3 & u'_3 & u_3v'_3 & v_3v'_3 & v'_3 & u_3 & v_3 & 1 \\ u_4u'_4 & v_4u'_4 & u'_4 & u_4v'_4 & v_4v'_4 & v'_4 & u_4 & v_4 & 1 \\ u_5u'_5 & v_5u'_5 & u'_5 & u_5v'_5 & v_5v'_5 & v'_5 & u_5 & v_5 & 1 \\ u_6u'_6 & v_6u'_6 & u'_6 & u_6v'_6 & v_6v'_6 & v'_6 & u_6 & v_6 & 1 \\ u_7u'_7 & v_7u'_7 & u'_7 & u_7v'_7 & v_7v'_7 & v'_7 & u_7 & v_7 & 1 \\ u_8u'_8 & v_8u'_8 & u'_8 & u_8v'_8 & v_8v'_8 & v'_8 & u_8 & v_8 & 1 \end{bmatrix} \begin{bmatrix} F_{11} \\ F_{12} \\ F_{13} \\ F_{21} \\ F_{22} \\ F_{23} \\ F_{31} \\ F_{32} \\ F_{33} \end{bmatrix} = 0$$

Recovering Fundamental Matrix

- 8-point algorithm
 - Construct linear system using corresponding image points
 - Solve for \mathbf{f} using SVD

$$W\mathbf{f} = 0$$

$$W = USV^T$$

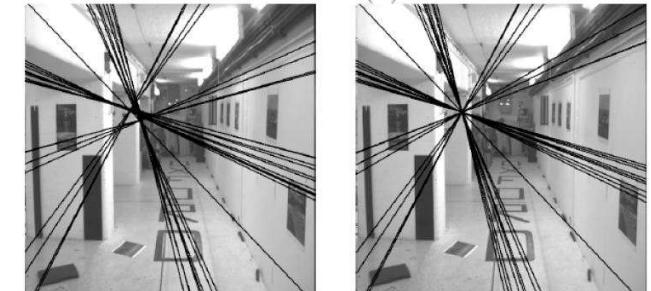
Last column of V gives \mathbf{f}

$$\begin{bmatrix} u_1u'_1 & v_1u'_1 & u'_1 & u_1v'_1 & v_1v'_1 & v'_1 & u_1 & v_1 & 1 \\ u_2u'_2 & v_2u'_2 & u'_2 & u_2v'_2 & v_2v'_2 & v'_2 & u_2 & v_2 & 1 \\ u_3u'_3 & v_3u'_3 & u'_3 & u_3v'_3 & v_3v'_3 & v'_3 & u_3 & v_3 & 1 \\ u_4u'_4 & v_4u'_4 & u'_4 & u_4v'_4 & v_4v'_4 & v'_4 & u_4 & v_4 & 1 \\ u_5u'_5 & v_5u'_5 & u'_5 & u_5v'_5 & v_5v'_5 & v'_5 & u_5 & v_5 & 1 \\ u_6u'_6 & v_6u'_6 & u'_6 & u_6v'_6 & v_6v'_6 & v'_6 & u_6 & v_6 & 1 \\ u_7u'_7 & v_7u'_7 & u'_7 & u_7v'_7 & v_7v'_7 & v'_7 & u_7 & v_7 & 1 \\ u_8u'_8 & v_8u'_8 & u'_8 & u_8v'_8 & v_8v'_8 & v'_8 & u_8 & v_8 & 1 \end{bmatrix} \begin{bmatrix} F_{11} \\ F_{12} \\ F_{13} \\ F_{21} \\ F_{22} \\ F_{23} \\ F_{31} \\ F_{32} \\ F_{33} \end{bmatrix} = 0$$

Recovering Fundamental Matrix

- 8-point algorithm
 - Construct linear system using corresponding image points
 - Solve for \mathbf{f} using SVD
 - Constraint enforcement (**essential step**)
 - $\text{rank}(F) = 2$

Fundamental matrix has rank 2 : $\det(\mathbf{F}) = 0$.



Left: Uncorrected \mathbf{F} – epipolar lines are not coincident.

Right: Epipolar lines from corrected \mathbf{F} .

Left: without constraint rank=2

Right: with constraint rank=2

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$$\hat{F} = UDV^T \quad D = \begin{bmatrix} d_1 & 0 & 0 \\ 0 & d_2 & 0 \\ 0 & 0 & d_3 \end{bmatrix} \quad \xrightarrow{\text{blue arrow}} \quad F = U \begin{bmatrix} d_1 & 0 & 0 \\ 0 & d_2 & 0 \\ 0 & 0 & 0 \end{bmatrix} V^T$$