

# Stereo view geometry

## Epipolar geometry, triangulation

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# Learning objectives

After this lecture you should be able to:

- Derive and apply the fundamental matrix in computer vision
- Derive and apply the essential matrix in computer vision
- Implement the linear algorithm for triangulation
- Explain the pros and cons of using a linear algorithm

# Presentation topics

Projection of points, lines and planes

Epipolar planes and lines

Essential and fundamental matrices

Triangulation

Problems with linear algorithm

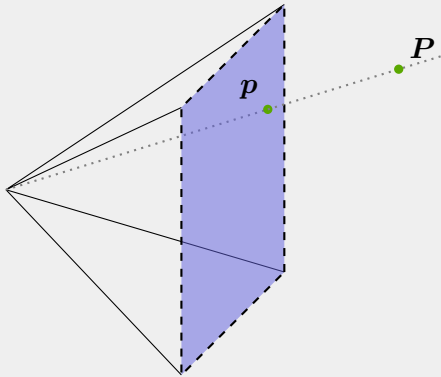
# Projection of points, lines and planes

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# Point projections

Any 3D point  $P$  projects to a 2D point  $p$ :

$$P \xrightarrow{\text{projection}} p.$$

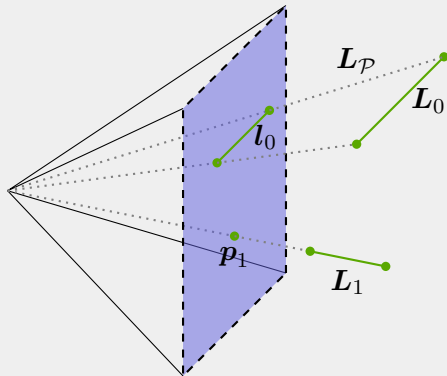


# Line projections

Any 3D line  $L$  projects to a 2D line  $l$ , except if  $L \parallel L_{\mathcal{P}}$ :

$$L_0 \xrightarrow{\text{projection}} l_0 ,$$

$$L_1 \xrightarrow{\text{projection}} p_1 .$$



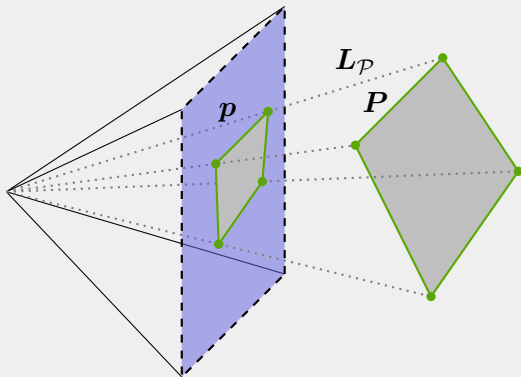
# Plane projections

Any 3D plane  $P$  projects to  
the image plane,

$$P \xrightarrow{\text{projection}} p,$$

except if  $P \parallel L_{\mathcal{P}}$

Where then?

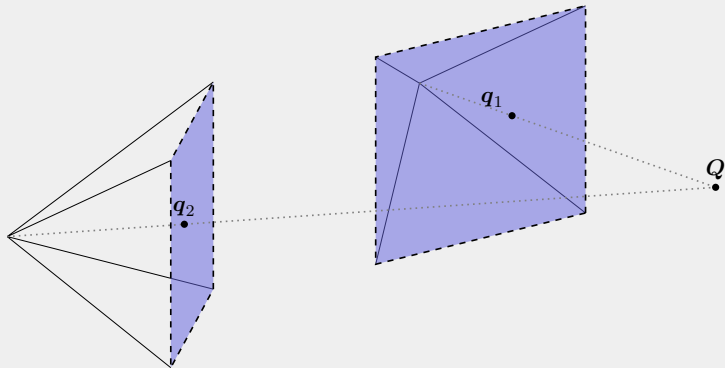


# Epipolar planes and lines

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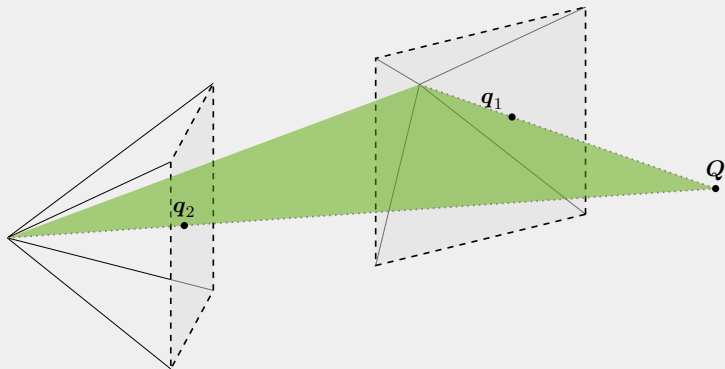


# Stereo view



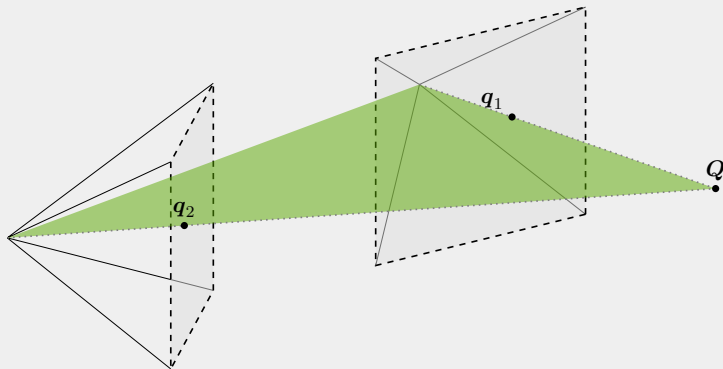
Assume stereo view

# Epipolar planes



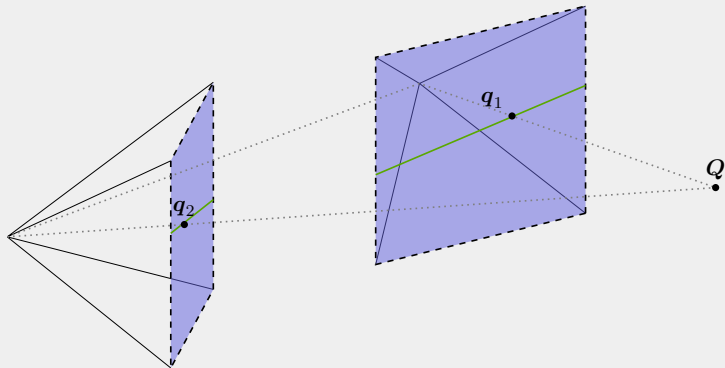
The camera centers and the 3D point  $Q$  form an **epipolar plane**!

# Epipolar planes



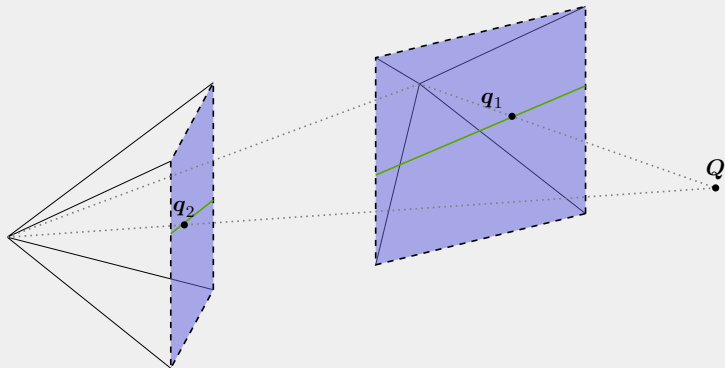
Which line in space is always part of the epipolar planes?

# Epipolar lines



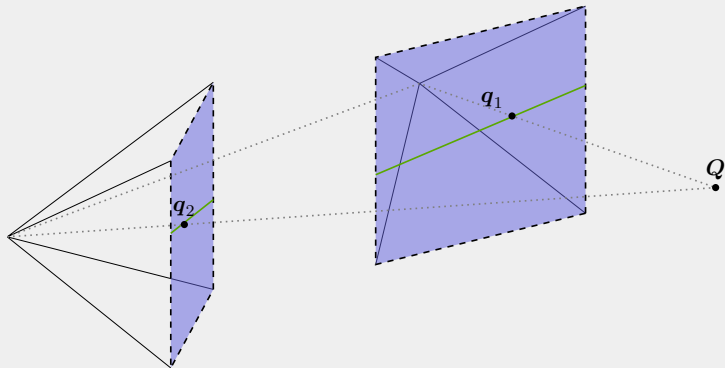
The **epipolar lines** intersect the epipolar plane and images!

# Epipolar lines



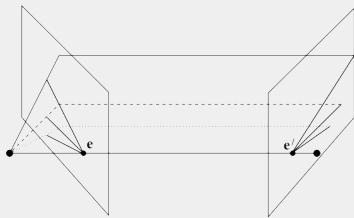
The relation between the line  $q_i Q$  and the epipolar lines?

# Epipolar lines



All epipolar lines intersect in the epipoles.

# Epipoles



Where are the epipoles?

**Stereo view is an epipolar geometry:**  
**Each  $Q$  in 3D has an epipolar plane.**  
**Pixels correspond if and only if**  
**they lie in the same epipolar plane**



# Essential and fundamental matrices

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## Setting the scene

Let  $\mathbf{q}$  refer to a 2D point in pixels and  $\mathbf{p}$  refer to the same point in 3D in the reference frame of the camera.

These are related as follows:

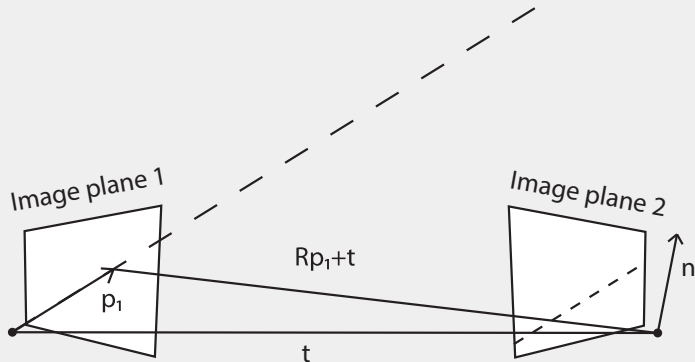
$$\mathbf{q} = \mathbf{K}\mathbf{p}$$

$$\mathbf{p} = \mathbf{K}^{-1}\mathbf{q}$$

$\mathbf{R}, \mathbf{t}$  maps from the reference frame of camera one to the reference frame of camera two (their relative transformation)

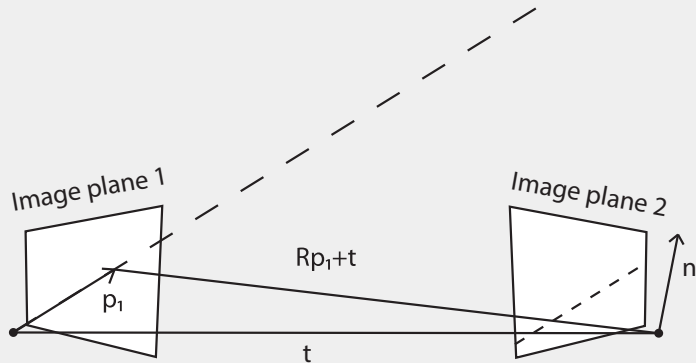
# The essential matrix

Consider the epipolar plane given by  $p_1$



# The essential matrix

Relative to camera 2, what is the normal of the epipolar plane?



The normal is orthogonal to  $t$  and  $Rp_1 + t$

# The essential matrix

The normal is then

$$\begin{aligned}n &= t \times (Rp_1 + t) \\&= t \times (Rp_1) \\&= \underbrace{[t]_{\times}}_E Rp_1.\end{aligned}$$

$E$  is called the **essential matrix**!

# The essential matrix

Dot product of orthogonal vectors are zero.

For the corresponding  $\mathbf{p}_2$  in the second camera

$$\begin{aligned} 0 &= \mathbf{p}_2^\top \mathbf{n} \\ &= \mathbf{p}_2^\top \mathbf{E} \mathbf{p}_1 \end{aligned}$$

The essential matrix imposes a constraint on corresponding points.

## How to interpret this?

- $p = K^{-1}q$ .
- Are  $p_1$  and  $p_2$  in homogeneous or inhomogeneous coordinates?

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- $p = K^{-1}q$ .
- Are  $p_1$  and  $p_2$  in homogeneous or inhomogeneous coordinates?
  - Yes!...
- There are two interpretations:
  - $p_1$  and  $p_2$  are 3D points and  $n$  is a vector in 3D. We use this in our derivations.
  - $p_1$  and  $p_2$  are 2D points and  $n = Ep_1$  is the epipolar line, both are in homogeneous coordinates.



# The fundamental matrix

Recall that

$$\mathbf{p} = \mathbf{K}^{-1}\mathbf{q}.$$

Then

$$\begin{aligned} \mathbf{p}_2^\top \mathbf{E} \mathbf{p}_1 &= 0 \\ (\mathbf{K}_2^{-1} \mathbf{q}_2)^\top \mathbf{E} (\mathbf{K}_1^{-1} \mathbf{q}_1) &= 0 \\ \mathbf{q}_2^\top \underbrace{\mathbf{K}_2^{-\top} \mathbf{E} \mathbf{K}_1^{-1}}_{\mathbf{F}} \mathbf{q}_1 &= 0 \end{aligned}$$

where  $\mathbf{F}$  is the **fundamental matrix**!

The essential and fundamental matrices  
form requirements for pixel  
correspondence:

$$p_2^\top E p_1 = 0$$

$$q_2^\top F q_1 = 0$$

## Note on $R, t$

What if  $R, t$  is not given, but you only know the pose of each camera in world coordinates  $(R_1, t_1, R_2, t_2)$ ?

You can compute the relative transformation (which you will do in the exercise today)

# Fundamental/essential matrix vs homography

What is the difference?

# Fundamental/essential matrix vs homography

What is the difference?

- The fundamental and essential matrices yield epipolar lines:
  - $0 = \mathbf{p}_2^\top \mathbf{E} \mathbf{p}_1$
  - $0 = \mathbf{q}_2^\top \mathbf{F} \mathbf{q}_1$
  - Corresponding points must lie on the epipolar line, no matter what is in the image.
- The homography establishes one-to-one correspondences:
  - $\mathbf{q}_2 = \mathbf{H} \mathbf{q}_1$
  - Only valid for planes.

# Degrees of freedom of $F$

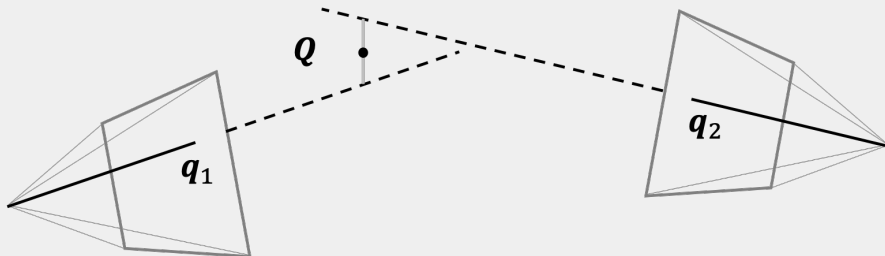
- $F$  has 9 numbers, and is scale invariant.
- $[t]_{\times}$  has rank 2, and thus  $F$  has as well.
  - In other terms:  $\det(F) = 0$ .
- You will estimate  $F$  in week 10.

# Triangulation

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# Triangulation



We've seen the same point in many (known) cameras and want to find the point in 3D.

Because of noise, there is not always a solution without error.

# Triangulation problem

Consider a projection matrix

$$\mathcal{P}_i = \begin{bmatrix} p_i^{(1)} \\ p_i^{(2)} \\ p_i^{(3)} \end{bmatrix}$$

# Triangulation equations

Projection gives the pixels

$$\mathbf{q}_i = \begin{bmatrix} s_i x_i \\ s_i y_i \\ s_i \end{bmatrix} = \mathcal{P}_i \mathbf{Q} = \begin{bmatrix} p_i^{(1)} Q \\ p_i^{(2)} Q \\ p_i^{(3)} Q \end{bmatrix}$$

This is two constraints  $(x_i, y_i)$  in three equations.

# Triangulation equations

As  $s_i = p_i^{(3)}Q$ , we have

$$\left(p_i^{(3)}Q\right) \begin{bmatrix} x_i \\ y_i \end{bmatrix} = \begin{bmatrix} p_i^{(1)}Q \\ p_i^{(2)}Q \end{bmatrix}$$

rearranged into

$$\begin{aligned} 0 &= \begin{bmatrix} p_i^{(3)}x_i - p_i^{(1)} \\ p_i^{(3)}y_i - p_i^{(2)} \end{bmatrix} Q \\ &= B^{(i)}Q \end{aligned}$$

# Triangulation solution

Define  $B$

$$B = \begin{bmatrix} B^{(1)} \\ B^{(2)} \\ \vdots \\ B^{(n)} \end{bmatrix} = \begin{bmatrix} \boldsymbol{p}_1^{(3)} x_1 - \boldsymbol{p}_1^{(1)} \\ \boldsymbol{p}_1^{(3)} y_1 - \boldsymbol{p}_1^{(2)} \\ \boldsymbol{p}_2^{(3)} x_2 - \boldsymbol{p}_2^{(1)} \\ \boldsymbol{p}_2^{(3)} y_2 - \boldsymbol{p}_2^{(2)} \\ \vdots \end{bmatrix}.$$

Use SVD to find  $\arg \min_Q \|BQ\|_2$ , s.t.  $\|Q\|_2 = 1$ .

# Linear algorithm, hmm

- $\arg \min_Q \|BQ\|_2, \quad \text{s.t. } \|Q\|_2 = 1.$
- This is a linear algorithm used to solve the problem.
- Which error would we actually like to minimize?

# Linear algorithm, hmm

- $\arg \min_Q \|BQ\|_2, \quad \text{s.t. } \|Q\|_2 = 1.$
- This is a linear algorithm used to solve the problem.
- Which error would we actually like to minimize?
  - Depends why we believe that there are errors
  - Errors in the observed pixel location

# Triangulation: ideal error

- Let  $\begin{bmatrix} \tilde{x}_i & \tilde{y}_i \end{bmatrix}$  and  $\begin{bmatrix} x_i & y_i \end{bmatrix}$  refer to the observed and projected pixel coordinates, respectively.
- Which error would we like to minimize?

$$e_{\text{ideal}} = \sum_i \left\| \begin{bmatrix} \tilde{x}_i \\ \tilde{y}_i \end{bmatrix} - \begin{bmatrix} x_i \\ y_i \end{bmatrix} \right\|_2^2$$



# Triangulation uh oh!

- Which error are actually we minimizing?

$$\begin{aligned}e_{\text{algebraic}} &= \sum_i \| \mathbf{B}_i \mathbf{Q} \|_2^2 \\&= \sum_i \left\| \begin{bmatrix} \mathbf{p}_i^{(3)} \tilde{x}_i - \mathbf{p}_i^{(1)} \\ \mathbf{p}_i^{(3)} \tilde{x}_i - \mathbf{p}_i^{(2)} \end{bmatrix} \mathbf{Q} \right\|_2^2 \\&= \sum_i \left\| \underbrace{\mathbf{p}_i^{(3)} \mathbf{Q}}_{s_i} \begin{bmatrix} \tilde{x}_i \\ \tilde{y}_i \end{bmatrix} - \begin{bmatrix} \mathbf{p}_i^{(1)} \\ \mathbf{p}_i^{(2)} \end{bmatrix} \mathbf{Q} \right\|_2^2 \\&= \sum_i \left\| s_i \begin{bmatrix} \tilde{x}_i \\ \tilde{y}_i \end{bmatrix} - s_i \begin{bmatrix} x_i \\ y_i \end{bmatrix} \right\|_2^2\end{aligned}$$

# Error terms compared

- We can rewrite the terms slightly again

$$e_{\text{ideal}} = \sum_i (\tilde{x}_i - x_i)^2 + (\tilde{y}_i - y_i)^2$$

$$e_{\text{algebraic}} = \sum_i s_i^2 (\tilde{x}_i - x_i)^2 + s_i^2 (\tilde{y}_i - y_i)^2$$

- $s_i$  is larger for cameras that are further from  $Q$

# Linear vs non-linear algorithms

- $e_{\text{ideal}}$  can be minimized using non-linear optimization
- Linear algorithms are very fast.
  - Only minimizes an algebraic error.
- We can estimate many things using linear algorithms
  - Triangulation, homography, fundamental matrix, projection matrix
  - They all have the problem that they don't minimize the exact error we desire
- Linear algorithms are acceptable in most cases.
- When high accuracy is desired initialize the non-linear optimization with the linear solution.

# Learning objectives

After this lecture you should be able to:

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- Implement the linear algorithm for triangulation
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**Exercise time!**