

# Image Analysis - Lecture 13

## Computer Vision

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October 11, 2011

# Lecture 13

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- ▶ Pinhole camera
- ▶ Two View Geometry

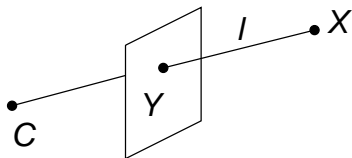
# What is Computer Vision?

*Can be seen as an attempt at mimicking human (biological) vision.*

Some examples of research areas:

- ▶ Recognition
- ▶ Motion estimation
- ▶ 3D reconstruction from 2D images

# Pinhole Camera



All points on the line  $l$  are mapped to the same point  $Y$ .  
 $c$  is called **camera center**:

The transformation  $P$ :  $X$  to  $Y$  is called **perspective transformation**.

*Note:  $P$  is not invertible. (The depth is unknown)*

Special case: Planar object -  $P$  invertible.

Special case:  $c$  infinitely far away.  $P$  parallel projection (affine transformation)

a) Affine transformations (changes of coordinates)  $\mathbb{R}^3 \rightarrow \mathbb{R}^3$   
can be written

$$y = Ax + b ,$$

where

$$y = \begin{bmatrix} y_1 \\ y_2 \\ y_3 \end{bmatrix}, x = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix}, A = \begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{bmatrix}, b = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \end{bmatrix} .$$

This can also be written

$$y = [A \quad b] \begin{bmatrix} x \\ 1 \end{bmatrix} \Leftrightarrow \begin{bmatrix} y \\ 1 \end{bmatrix} = \begin{bmatrix} A & b \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ 1 \end{bmatrix}$$

# Euclidean transformations

b) **Euclidean transformations** can be written

$$\begin{bmatrix} y \\ 1 \end{bmatrix} = \begin{bmatrix} R & t \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ 1 \end{bmatrix} ,$$

where  $R$  is an orthogonal matrix, i.e.

$$RR^T = R^T R = I .$$

# Projective transformations

c) **Projective transformations** are defined as

$$\lambda \begin{bmatrix} y \\ 1 \end{bmatrix} = \begin{bmatrix} A & b \\ c & d \end{bmatrix} \begin{bmatrix} x \\ 1 \end{bmatrix} .$$

$$\lambda(y_1, y_2, y_3, 1) = (\lambda y_1, \lambda y_2, \lambda y_3, \lambda)$$

are called **homogeneous coordinates** for

$$(y_1, y_2, y_3) .$$

In the **projective space**,  $\mathcal{P}^3$ , one identifies

$$(y_1, y_2, y_3, 1) \quad \text{and} \quad (\lambda y_1, \lambda y_2, \lambda y_3, \lambda).$$

These are considered as the same point.

# Homogeneous coordinates

An image point  $(x, y)$  is represented in homogeneous coordinate as

$$\begin{pmatrix} x \\ y \\ 1 \end{pmatrix}$$

or multiples of this vector, e.g.

$$\begin{pmatrix} 5x \\ 5y \\ 5 \end{pmatrix}$$

What homogenous coordinates does the image point  $(127, 321)$  have? Answer:  $(127 \ 321 \ 1)^T$ .

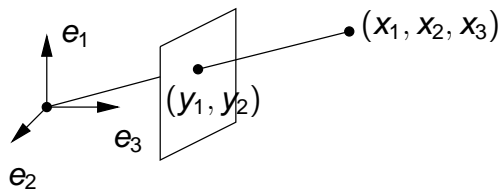


Which image point does the homogeneous coordinates  $(1 \ 2 \ 3)^T$  correspond to? Answer: Image point:  $(1/3, 2/3)$ . In the same way, a scene point  $(x, y, z)$  can be represented in homogeneous coordinates by

$$\begin{pmatrix} x \\ y \\ z \\ 1 \end{pmatrix}$$

(Note the similarity with line representations:  $(a \ b \ c)^T$ . This is the homogeneous coordinate representation of the line  $ax + by + c = 0$ .)

# Perspective transformation



$$y_1 = \frac{x_1}{x_3}, \quad y_2 = \frac{x_2}{x_3}$$

This is a *non linear expression*, which can be written

$$\lambda \begin{bmatrix} y_1 \\ y_2 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix},$$

with  $\lambda = x_3$ .

$\lambda$  is called the **depth**.

# External calibration

Using homogeneous coordinates

$$\lambda \begin{bmatrix} y_1 \\ y_2 \\ 1 \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = [I \quad 0] \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ 1 \end{bmatrix}$$

Euclidean change of coordinates in the scene gives:

$$\lambda \begin{bmatrix} y_1 \\ y_2 \\ 1 \end{bmatrix} = [I \quad 0] \begin{bmatrix} R & t \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x'_1 \\ x'_2 \\ x'_3 \\ 1 \end{bmatrix} = [R \quad t] \begin{bmatrix} x'_1 \\ x'_2 \\ x'_3 \\ 1 \end{bmatrix}$$

$R$  and  $t$  are called **external calibration**.

$R$  describes camera orientation.

$-R^{-1}t$  is called the camera center.

# Internal calibration

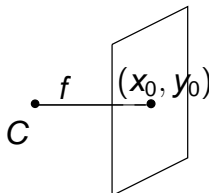
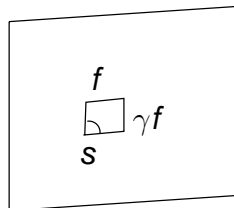
Modeling of the camera gives:

$$\lambda \begin{bmatrix} y'_1 \\ y'_2 \\ 1 \end{bmatrix} = K [R \quad t] \begin{bmatrix} x'_1 \\ x'_2 \\ x'_3 \\ 1 \end{bmatrix}, \quad K = \begin{bmatrix} f & sf & x_0 \\ 0 & \gamma f & y_0 \\ 0 & 0 & 1 \end{bmatrix}$$

The parameters in  $K$  are called **internal parameters**:

- ▶  $f$ : **focal length**
- ▶  $(x_0, y_0)$ : **principal point** (orthogonal projection of the camera center in the image plane.)
- ▶  $\gamma$ : **aspect ratio** (rectangular light sensitive elements instead of quadratic)
- ▶  $s$ : **skew** (skewed light sensitive elements)

# Calibrated vs uncalibrated camera



- ▶ Calibrated camera =  $K$  known
- ▶ Uncalibrated camera =  $K$  unknown
- ▶ External orientation =  $R$  and  $t$
- ▶ Internal orientation =  $K$

If both  $K$ ,  $R$  and  $t$  are unknown, then you can rewrite  $\lambda \mathbf{y} = P\mathbf{x}$ , where  $P$  is a  $3 \times 4$  matrix. Such a matrix  $P$  is called a **camera matrix**. This can be written as (in homogeneous coordinates)  $\mathbf{y} \sim P\mathbf{x}$ .

# Camera matrix

The camera matrix  $P$  is a  $3 \times 4$  matrix, that describes camera position, orientation, zoom, etc.

Assume that the camera matrix is

$$P = \begin{bmatrix} 5 & -14 & 2 & 17 \\ -10 & -5 & -10 & 50 \\ 10 & 2 & -11 & 19 \end{bmatrix}$$

and that a 3D-point with coordinates  $(0, 3, 2)$  are given. What is the image point?

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Answer: The homogeneous coordinates of the 3D point is

$X = [0 \ 3 \ 2 \ 1]^T$ . The product  $PX$  is  $PX = [-21 \ 15 \ 3]^T$ . This corresponds to the image point  $(-21/3, 15/3) = (-7, 5)$ .

# Camera center

The homogeneous coordinates of the camera center (focal point)  $C$  can be calculated by

$$PC = 0.$$

What is the 3D-coordinates of the camera center for the camera matrix above?



# Camera center

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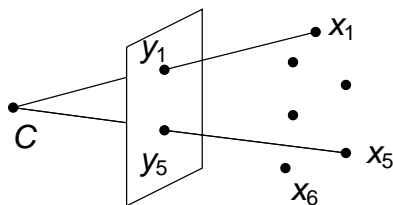
What is the 3D-coordinates of the camera center for the camera matrix above?

Answer: The null space of  $P$  is given by  $C = t [2 \ 4 \ 6 \ 2]^T$ .  
The scene coordinates for the camera center is  $(1, 2, 3)$ .

# Calibration

## Example

Assume that we have measured 6 3D points,  $x_i$ . By taking image of these points we obtain 6 image points,  $y_i$ . How can we calculate  $K$ ?



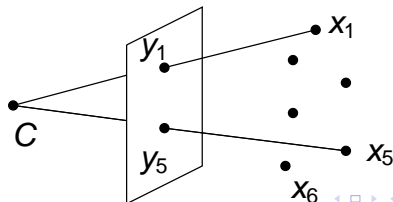
# Calibration

## Example

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$$\lambda_i \mathbf{y}_i = P \mathbf{x}_i$$

gives 3 equations for every point (linear in  $\lambda_i$  and  $p_{ij}$ ). We get 18 equations in 18 unknown parameters (12 for  $P$  and one for every point). Solve these equations and factorize  $P$  as  $P = K[R | t]$ . ■



# Internal Calibration

Use:

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

for 6 points and calculate  $P = [A | b]$ .

$P$  is written as

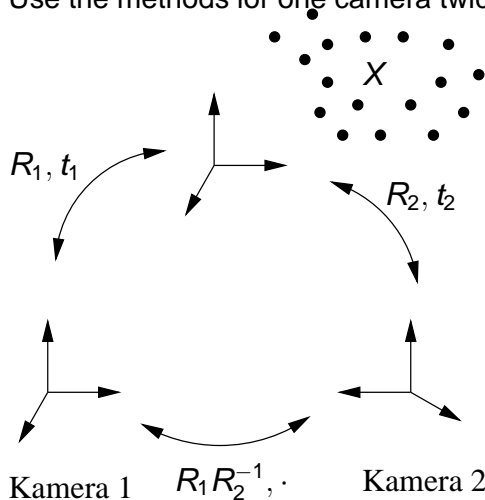
$$P = K[Q | t] = [KQ | Kt]$$

$K$  is obtained by  $RQ$ -factorization of  $A$ .

$$A = \begin{bmatrix} \times & \times & \times \\ & \times & \times \\ \circ & & \times \end{bmatrix} \times \begin{bmatrix} \text{ORT.} \end{bmatrix}$$

# Two views: Absolute orientation

Determine the camera matrix using a known object.  
Use the methods for one camera twice!



# Relative Orientation

*Determine the cameras relative motion. Not using a known scene.*

Note: Corresponding points in the two images are still needed.

## Problem

*Relative orientation. Assume that only image correspondences  $(x_1, x_2)$  are given in the two images. Use the equations:*

$$\lambda_1 x_1 = P_1 X$$

$$\lambda_2 x_2 = P_2 X$$

*to solve for camera matrices  $P_1$ ,  $P_2$  and for 3D coordinates of the point  $X$ .*

# Fundamental (Essential) matrix

Two projections of the same point:

$$\lambda_1 x_1 = P_1 X$$

$$\lambda_2 x_2 = P_2 X$$

Rewrite as one coordinate system:

$$\underbrace{\begin{pmatrix} P_1 & x_1 & 0 \\ P_2 & 0 & x_2 \end{pmatrix}}_M \begin{pmatrix} X \\ -\lambda_1 \\ -\lambda_2 \end{pmatrix} = 0$$

The matrix  $M$  has  $\det(M) = 0$ . Discussion: Why?

Expand the determinant (It is linear in each column) in  $x_1$  and  $x_2$  to obtain  $\det(M) = x_1^T F x_2 = 0$  for some matrix  $F$ .

# Camera matrix $\rightarrow$ fundamental matrix

The elements of  $F$  depends on the camera matrices  $(P_1, P_2)$ .  
By expanding the determinant  $\det(M)$  and identifying with  $x_1^T F x_2 = 0$ , one can show that the elements of  $F$  can be written as determinants of matrices that contains some of the rows of the two camera matrices

As an example, element

$$F_{11} = \det \begin{pmatrix} P_{1,21} & P_{1,22} & P_{1,23} & P_{1,24} \\ P_{1,31} & P_{1,32} & P_{1,33} & P_{1,34} \\ P_{2,21} & P_{2,22} & P_{2,23} & P_{2,24} \\ P_{2,31} & P_{2,32} & P_{2,33} & P_{2,34} \end{pmatrix}$$

The mapping  $(P_1, P_2) \mapsto F$  is thus relatively simple.



# Fundamental matrix $\rightarrow$ camera matrix

There are 24 elements in the camera matrices  $(P_1, P_2)$ .

The scale is arbitrary (-2)

There is freedom in choosing a projective coordinate system (-15)

Thus there remains 7 degrees of freedom

There are 9 elements in the fundamental matrix  $F$ .

The scale is arbitrary (-1).

There is an extra constraint  $\det(F) = 0$  (-1).

Thus there remains 7 degrees of freedom in  $F$ .

It turns out that one can compute the relative motion  $(P_1, P_2)$  from  $F$ . The mapping is 'almost' unique.

The mapping  $F \mapsto (P_1, P_2)$  is thus known.

# Solution to the relative orientation problem

Each point correspondence  $(x_1, x_2)$  gives a linear constraint on  $F$ .

$$x_1^T F x_2 = 0$$

Given at least 7 such correspondences one can calculate  $F$ .

>From  $F$  one can calculate relative orientation  $(P_1, P_2)$ .

Using image points  $(x_1, x_2)$  and camera matrices  $(P_1, P_2)$  one can calculate  $X$  (and  $(\lambda_1, \lambda_2)$ ).

Thus one has solved for all unknowns in the equations

$$\lambda_1 x_1 = P_1 X$$

$$\lambda_2 x_2 = P_2 X$$

# Essential matrix

A particular case of the fundamental matrix is the case of camera matrices that have been corrected for internal calibration.

There is then 2 additional constraints on the fundamental matrix. Such a matrix is often called essential matrix.

## Theorem

*An essential matrix  $E$  can be written in the form*

$$E = R^T T_{\bar{r}_0}$$

*and has singular value decomposition  $E = U\Sigma V$ , with*

$$\Sigma = \begin{bmatrix} a & 0 & 0 \\ 0 & a & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

# Epipoles

## Definition

The fundamental matrix  $F$  defines the **epipolar geometry**. ■

## Definition

The image point  $e_l = (e_{lx}, e_{ly}, 1)$  with

$$e_l^T F = 0$$

is called the **left epipole**. Similarly the point  $e_r = (e_{rx}, e_{ry}, 1)$  with

$$F e_r = 0$$

is called the **right epipole**. ■

## Proposition

*The left epipole is the projection of the camera center of the right camera and vice versa.*

# Epipolar lines

Given a point  $(x'_l, y'_l, 1)$ , in the left image and the epipolar matrix  $F$ .

The epipolar constraint

$$\begin{bmatrix} x'_l & y'_l & 1 \end{bmatrix} F \begin{bmatrix} x'_r \\ y'_r \\ 1 \end{bmatrix} = l^T \begin{bmatrix} x'_r \\ y'_r \\ 1 \end{bmatrix} = 0$$

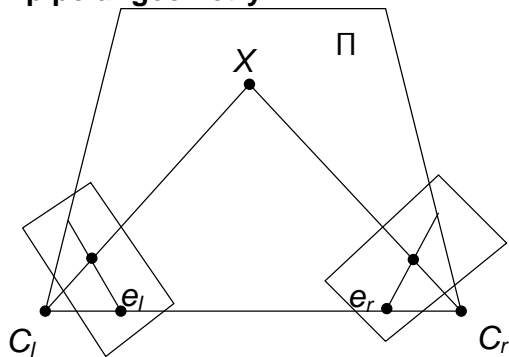
gives that the corresponding point in the right image must lie on a line.

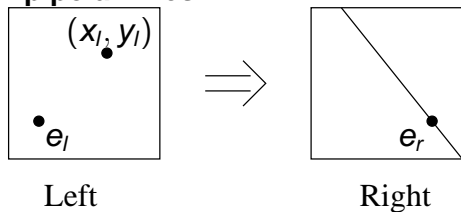
## Definition

This line is called the **epipolar line** to the point  $(x'_l, y'_l, 1)$ . ■

# Illustrations

Epipolar geometry:



**Epipolar lines:**



# Fundamental matrix

Assume that we have two images and that their relative orientation is given by the fundamental matrix

$$F = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}$$

Assume that a point in image 1 has coordinates  $(2, -3)$ . What coordinates can corresponding point have in image 2? Can the point  $(-2, 1)$  be the corresponding point in image 2? Where are the left and right epipoles?

# Fundamental matrix

Assume that we have two images and that their relative orientation is given by the fundamental matrix

$$F = \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix}$$

Assume that a point in image 1 has coordinates  $(2, -3)$ . What coordinates can corresponding point have in image 2? Can the point  $(-2, 1)$  be the corresponding point in image 2? Where are the left and right epipoles?

Answer: Homogeneous coordinates for the points are  $x_1 = (2 \ -3 \ 1)^T$  and  $x_2 = (-2 \ 1 \ 1)^T$ . Since the points fulfill the epipolar constraint  $x_1^T F x_2 = 0$  they could be corresponding points.

In fact all points in image 2 that fulfill

$$(2 \quad -3 \quad 1) \begin{pmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \\ 7 & 8 & 9 \end{pmatrix} \begin{pmatrix} x \\ y \\ 1 \end{pmatrix} = -3x - 3y - 3 = 0$$

i.e. that lie on the line  $x + y + 1 = 0$  could be corresponding points.

The epipoles are given by  $F\mathbf{e}_2 = 0$  and  $\mathbf{e}_1^T F = 0$ . Calculations give  $\mathbf{e}_1 = (1 \quad -2 \quad 1)^T$  and  $\mathbf{e}_2 = (1 \quad -2 \quad 1)^T$

# Masters thesis suggestion of the day:

The general problem of calculating 3D shape from images can be simplified considerably e.g. if the images are taken of an object on a rotating plane. The goal of the masters thesis is to see to what extent this process can be automated and what degree of model complexity can be obtained from such a system.

Image sequence -> VRML model

# Review - Lecture 11

- ▶ Euclidean, affine and projective transformations
- ▶ External orientation
- ▶ Internal orientation
- ▶ Absolute orientation
- ▶ Relative orientation
- ▶ Fundamental matrix (epipoles, epipolar lines)
- ▶ Essential matrix