

Outline

1. Very Brief Introduction
2. The Projective Plane
3. Projective Transformations
4. Recovery of Affine Properties from Images
5. Angles in the Projective Plane
6. Recovery of Metric Properties from Images
7. Organizational Stuff

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Computer Vision 1. Very Brief Introduction

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Topics of the Lecture

1. Simultaneous Localization and Mapping from Video (Visual SLAM)
2. Image Classification and Description

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Computer Vision 1. Very Brief Introduction

Simultaneous Localization and Mapping



[source <https://www.youtube.com/watch?v=bD0nn0-4Nq8>]

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Simultaneous Localization and Mapping from Video










- ▶ SLAM usually employs laser range scanners (lidars).
- ▶ **Visual SLAM**: use video sensors (cameras).
- ▶ main parts required:
 1. Projective Geometry
 2. Point Correspondences
 3. Estimating Camera Positions (Localization)
 4. Triangulation (Mapping)

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Computer Vision 1. Very Brief Introduction

Image Classification and Description

Describes without errors	Describes with minor errors	Somewhat related to the image	Unrelated to the image
 <p>A person riding a motorcycle on a dirt road.</p>	 <p>Two dogs play in the grass.</p>	 <p>A skateboarder does a trick on a ramp.</p>	 <p>A dog is jumping to catch a frisbee.</p>
 <p>A group of young people playing a game of frisbee.</p>	 <p>Two hockey players are fighting over the puck.</p>	 <p>A little girl in a pink hat is blowing bubbles.</p>	 <p>A refrigerator filled with lots of food and drinks.</p>
 <p>A herd of elephants walking across a dry grass field.</p>	 <p>A close up of a cat laying on a couch.</p>	 <p>A red motorcycle parked on the side of the road.</p>	 <p>A yellow school bus parked in a parking lot.</p>

[source: <http://googleresearch.blogspot.de/2014/11/a-picture-is-worth-thousand-coherent.html>]

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Computer Vision 2. The Projective Plane

Motivation

In Euclidean (planar) geometry, there are many exceptions, e.g.,

- ▶ most two lines intersect in exactly one point.
- ▶ but some two lines do not intersect.
 - ▶ parallel lines

Idea:

- ▶ add **ideal points**, one for each set of parallel lines / direction
- ▶ define these points as intersection of any two parallel lines
- ▶ now any two lines intersect in exactly one point
 - ▶ either in a finite or in an ideal point

Homogeneous Coordinates: Points

Inhomogeneous coordinates:

$$x \in \mathbb{R}^2$$

Homogeneous coordinates:

$$x \in \mathbb{P}^2 := \mathbb{R}^3 / \equiv$$

$$x \equiv y : \Longleftrightarrow \exists s \in \mathbb{R} \setminus \{0\} : sx = y, \quad x, y \in \mathbb{R}^3$$

Example:

$$\begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} \equiv \begin{pmatrix} 4 \\ 8 \\ 12 \end{pmatrix} \text{ represent the same point in } \mathbb{P}^2$$

$$\begin{pmatrix} 1 \\ 2 \\ 4 \end{pmatrix} \text{ represent a different point in } \mathbb{P}^2$$

$$\begin{pmatrix} x_1 \\ \vdots \end{pmatrix}$$

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Homogeneous Coordinates: Lines

Inhomogeneous coordinates:

$$a \in \mathbb{R}^3 : \ell_a := \left\{ \begin{pmatrix} x_1 \\ x_2 \end{pmatrix} \mid a_1 x_1 + a_2 x_2 + a_3 = 0 \right\}$$

- ▶ $a_1 \neq 0$ or $a_2 \neq 0$ (or both $a_1, a_2 \neq 0$).
- ▶ $sa = (sa_1, sa_2, sa_3)^T$ encodes the same line as a (any $s \in \mathbb{R}, s \neq 0$).

Homogeneous coordinates:

$$a \in \mathbb{P}^2 : \ell_a := \{x \in \mathbb{P}^2 \mid a^T x = a_1 x_1 + a_2 x_2 + a_3 x_3 = 0\}$$

- ▶ contains all finite points of $a' \in \kappa^{-1}(a)$: $\ell_{\kappa(a')} \supsetneq \iota(\ell_{a'})$
- ▶ and the ideal point $(a_2, -a_1, 0)^T$.
 - ▶ intersection of parallel lines (same a_1, a_2 , different a_3)

Note: $\kappa : \mathbb{R}^3 \rightarrow \mathbb{P}^2, a \mapsto [a] := \{a' \in \mathbb{R}^3 \mid a' \equiv a\}$.

A point on a line

A point x lies on line a iff $x^T a = 0$.

Intersection of two lines

Lines a and b intersect in $a \times b := \begin{pmatrix} a_2 b_3 - a_3 b_2 \\ -a_1 b_3 + a_3 b_1 \\ a_1 b_2 - a_2 b_1 \end{pmatrix}$

Proof:

$$a^T(a \times b) = a_1 a_2 b_3 - a_1 a_3 b_2 - a_2 a_1 b_3 + a_2 a_3 b_1 + a_3 a_1 b_2 - a_3 a_2 b_1 = 0$$

$$b^T(a \times b) = \dots = 0$$

Example:

$$x = 1 : a = (-1, 0, 1)^T$$

$$y = 1 : b = (0, -1, 1)^T$$

$$a \times b = (1, 1, 1)^T$$

Esp. for parallel lines: $b_1 = a_1, b_2 = a_2, b_3 \neq a_3$:

$$a \times b \equiv \begin{pmatrix} a_2 \\ -a_1 \end{pmatrix}$$

Line joining points

The line through x and y is $x \times y$.

Proof: exactly the same as previous slide.

Example:

$$\begin{aligned}x &= (-1, 0, 1)^T \\ y &= (0, -1, 1)^T \\ x \times y &= (1, 1, 1)^T\end{aligned}$$

Line at infinity

All ideal points form a line:

$$l_{\infty} := (0, 0, 1)^T \quad \text{line at infinity}$$

Proof:

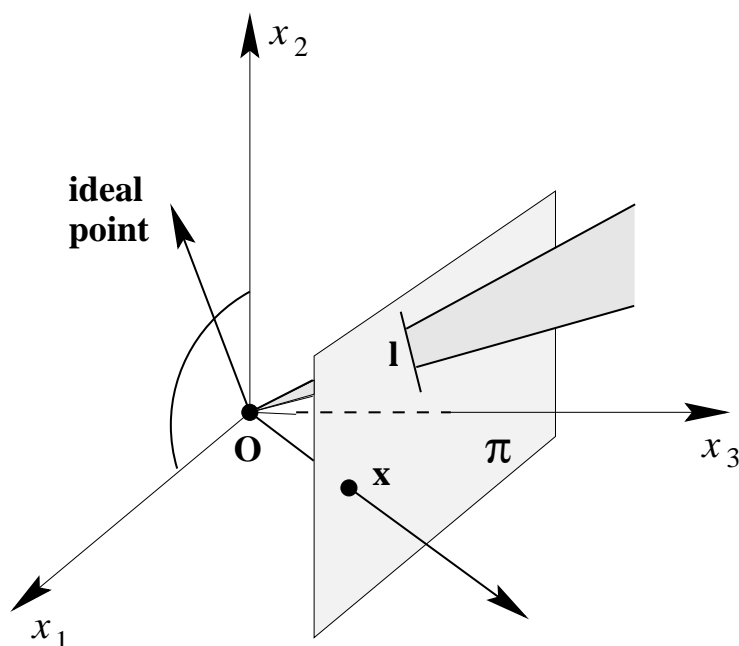
for any ideal point $x = (x_1, x_2, 0)^T$: $x^T l_{\infty} = 0$.

for any finite (real-valued) point $x = (x_1, x_2, 1)^T$: $x^T l_{\infty} = 1 \neq 0$.

Furthermore:

- This is the only line in \mathbb{P}^2 **not** corresponding to an Euclidean line.
- Two parallel lines meet at the line at infinity.

A model for the projective plane



- points correspond to rays (lines through the origin)
- lines correspond to planes through the origin. [HZ04, p. 29]

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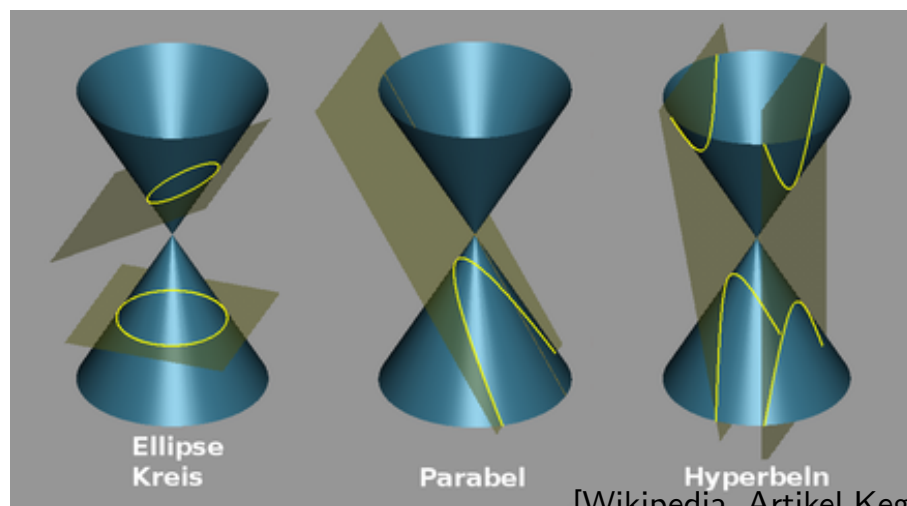
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Computer Vision 2. The Projective Plane

Conics

- A **conic section** (or just **conic**) is a curve one gets as intersection of a cone and a plane
 - ellipsis, parabola, hyperbola
- Corresponds to a curve of degree 2:
Heterogeneous coordinates:

$$a \in \mathbb{R}^6 : \mathbf{C}_a := \{x \in \mathbb{R}^2 \mid a_1 x_1^2 + a_2 x_1 x_2 + a_3 x_2^2 + a_4 x_1 + a_5 x_2 + a_6 = 0\}$$



[Wikipedia, Artikel Kegelschnitt]

Homogeneous coordinates:

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A conic joining 5 points

- ▶ Let $x^1, \dots, x^5 \in \mathbb{P}^2$ be 5 points
 - ▶ in general position (i.e., never more than 2 on the same line)
- ▶ Conic parameters a have to fulfil the following system of linear equations:

$$\begin{pmatrix} x_1^1 x_1^1 & x_1^1 x_2^1 & x_2^1 x_2^1 & x_1^1 x_3^1 & x_2^1 x_3^1 & x_3^1 x_3^1 \\ x_1^2 x_1^2 & x_1^2 x_2^2 & x_2^2 x_2^2 & x_1^2 x_3^2 & x_2^2 x_3^2 & x_3^2 x_3^2 \\ x_1^3 x_1^3 & x_1^3 x_2^3 & x_2^3 x_2^3 & x_1^3 x_3^3 & x_2^3 x_3^3 & x_3^3 x_3^3 \\ x_1^4 x_1^4 & x_1^4 x_2^4 & x_2^4 x_2^4 & x_1^4 x_3^4 & x_2^4 x_3^4 & x_3^4 x_3^4 \\ x_1^5 x_1^5 & x_1^5 x_2^5 & x_2^5 x_2^5 & x_1^5 x_3^5 & x_2^5 x_3^5 & x_3^5 x_3^5 \end{pmatrix} a = 0$$

Degenerate Conics

Conic C **degenerate**: C does not have full rank.

Example: two lines $C := ab^T + ba^T$ (rank 2).

- ▶ contains lines a and b .

proof: for points x on line a : $x^T a = 0$.

$\leadsto x$ also on C : $x^T C x = x^T a b^T x + x^T b a^T x = 0$.

Conic tangent lines

The tangent line to a conic C at a point x is Cx .

Proof:

x lies on Cx : $x^T Cx = 0$.

If there is another common point y : $y^T Cy = 0$ and $y^T Cx = 0$.

$\leadsto x + \alpha y$ is common for all α , i.e., the whole line.

$\leadsto C$ is degenerate (or there is no such y).

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Projectivity

A map $h : \mathbb{P}^2 \rightarrow \mathbb{P}^2$ is called **projectivity**, if

1. it is invertible and
2. it preserves lines,
i.e., whenever x, y, z are on a line, so are $h(x), h(y), h(z)$.

Equivalently, $h(x) := Hx$ for a non-singular $H \in \mathbb{P}^{3 \times 3}$.

Proof:

Any map $h(x) := Hx$ is a projectivity:

Let x be a point on line a : $a^T x = 0$.

Then point Hx is on line $H^{-T}a$: $(H^{-1}a)^T Hx = a^T H^{-1}Hx = a^T x = 0$.

Any projectivity h is of type $h(x) = Hx$: more difficult to show.

Note: $H^{-T} := (H^{-1})^T$.

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Computer Vision 3. Projective Transformations

Transformation of Lines and Conics

The image of a line a under projectivity H is the line $H^{-T}a$:

$$H(l_a) = l_{H^{-T}a}$$

Proof:

Let x be a point on line a : $a^T x = 0$.

Then point Hx is on line $H^{-1}a$: $(H^{-T}a)^T Hx = a^T H^{-1}Hx = a^T x = 0$.

The image of a conic C under projectivity H is the conic $H^{-T}CH^{-1}$:

$$H(\mathbf{C}_C) = \mathbf{C}_{H^{-T}CH^{-1}}$$

Proof:

Let x be a point on conic C : $x^T Cx = 0$.

Then point Hx is on conic $H^{-T}CH^{-1}$: $x^T H^T H^{-T}CH^{-1}H^{-1}x = 0$

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A Hierarchy of Transformations

- ▶ The projective transformations form a group (**projective linear group**):

$$\text{PL}_n := \text{GL}_n / \equiv = \{H \in \mathbb{P}^{3 \times 3} \mid H \text{ invertible}\}$$

- ▶ There are several subgroups:
 - ▶ **affine group**: last row is $(0, 0, 1)$
 - ▶ **Euclidean group**: additionally $H_{1:2,1:2}$ orthogonal
 - ▶ **oriented Euclidean group**: additionally $\det H = 1$
- ▶ These subgroups can be described two ways:
 - ▶ structurally (as above)
 - ▶ by **invariants**: objects or sets of objects mapped to themselves

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Computer Vision 3. Projective Transformations

Isometries

$$\begin{pmatrix} x'_1 \\ x'_2 \\ 1 \end{pmatrix} = \begin{pmatrix} \epsilon \cos \theta & -\sin \theta & t_1 \\ \epsilon \sin \theta & \cos \theta & t_2 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ 1 \end{pmatrix} = \begin{pmatrix} R & t \\ 0^T & 1 \end{pmatrix} x$$

- ▶ **rotation matrix R** : $R^T R = R R^T = I$
- ▶ **translation vector t** .
- ▶ **orientation preserving** if $\epsilon = +1$ (equivalent to $\det R = +1$)
($\epsilon \in \{+1, -1\}$)

Invariants:

- ▶ length, angle, area
- ▶ line at infinity l_∞

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Similarity Transformations

$$\begin{pmatrix} x'_1 \\ x'_2 \\ 1 \end{pmatrix} = \begin{pmatrix} s \cos \theta & -s \sin \theta & t_1 \\ s \sin \theta & s \cos \theta & t_2 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ 1 \end{pmatrix} = \begin{pmatrix} sR & t \\ 0^T & 1 \end{pmatrix} x$$

- **isotropic scaling** s .

Invariants:

- angle
- ratio of lengths, ratio of areas
- line at infinity l_∞

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Computer Vision 3. Projective Transformations

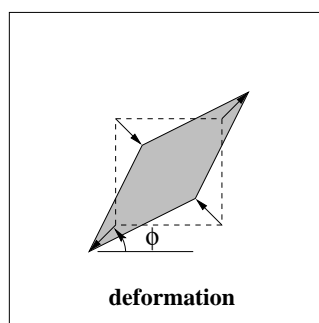
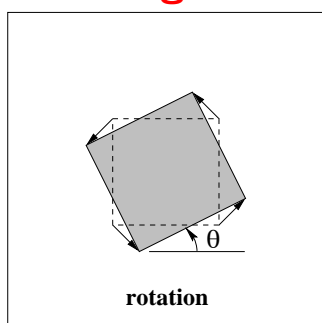
Affine Transformations

$$\begin{pmatrix} x'_1 \\ x'_2 \\ 1 \end{pmatrix} = \begin{pmatrix} a_{1,1} & a_{1,2} & t_1 \\ a_{2,1} & a_{2,2} & t_2 \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ 1 \end{pmatrix} = \begin{pmatrix} A & t \\ 0^T & 1 \end{pmatrix} x$$

- A non-singular, decompose via SVD:

$$A = R(\theta)R(-\phi) \begin{pmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{pmatrix} R(\phi)$$

- **non-isotropic scaling** with axis ϕ



[HZ04, p. 40]

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Projective Transformations

$$\begin{pmatrix} x'_1 \\ x'_2 \\ x'_3 \end{pmatrix} = \begin{pmatrix} a_{1,1} & a_{1,2} & t_1 \\ a_{2,1} & a_{2,2} & t_2 \\ v_1 & v_2 & v_3 \end{pmatrix} \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix} = \begin{pmatrix} A & t \\ v^T & v_3 \end{pmatrix} x$$

- v moves the line at infinity l_∞

Invariants:

- ratio of ratios of lengths of parallel line segments (**cross ratio**)

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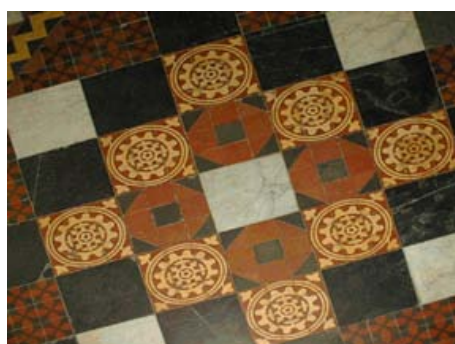
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Computer Vision 3. Projective Transformations

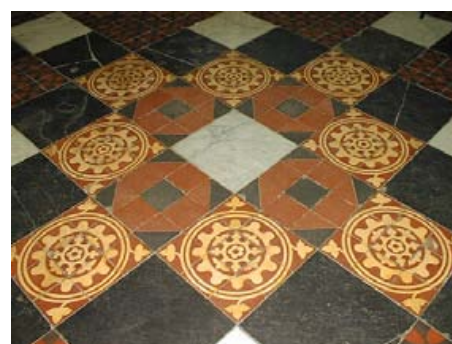
Similarity, Affine & Projective Transformations / Example



a



b



c

	a) similarity	b) affine	c) projective
circles	circles	ellipses	conic
squares	squares	diamond	quadrangle
parallel lines	parallel	parallel	converging
orthogonal lines	orthogonal	non-orthogonal	non-orthogonal

[HZ04, p. 37]

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Projective Transformations / Decomposition

$$\begin{pmatrix} A & t \\ v^T & v_3 \end{pmatrix} = \begin{pmatrix} sR & t \\ 0^T & 1 \end{pmatrix} \begin{pmatrix} K & 0 \\ 0^T & 1 \end{pmatrix} \begin{pmatrix} I & 0 \\ v^T & v_3 \end{pmatrix}$$

$$A = sRK + tv^T$$


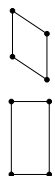
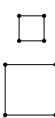
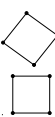
- ▶ K upper triangular matrix with $\det K = 1$
- ▶ valid for $v_3 \neq 0$
- ▶ unique if s is chosen $s > 0$

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Computer Vision 3. Projective Transformations

Summary of Projective Transformations

Group	Matrix	Distortion	Invariant properties
Projective 8 dof	$\begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix}$		Concurrency, collinearity, order of contact : intersection (1 pt contact); tangency (2 pt contact); inflections (3 pt contact with line); tangent discontinuities and cusps. cross ratio (ratio of ratio of lengths).
Affine 6 dof	$\begin{bmatrix} a_{11} & a_{12} & t_x \\ a_{21} & a_{22} & t_y \\ 0 & 0 & 1 \end{bmatrix}$		Parallelism, ratio of areas, ratio of lengths on collinear or parallel lines (e.g. midpoints), linear combinations of vectors (e.g. centroids). The line at infinity, l_∞ .
Similarity 4 dof	$\begin{bmatrix} sr_{11} & sr_{12} & t_x \\ sr_{21} & sr_{22} & t_y \\ 0 & 0 & 1 \end{bmatrix}$		Ratio of lengths, angle. The circular points, I, J (see section 2.7.3).
Euclidean 3 dof	$\begin{bmatrix} r_{11} & r_{12} & t_x \\ r_{21} & r_{22} & t_y \\ 0 & 0 & 1 \end{bmatrix}$		Length, area

[HZ04, p. 44]

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Computer Vision 4. Recovery of Affine Properties from Images

Recovery of Affine and Metric Properties

Decomposition of general projective transformation:

$$\begin{pmatrix} A & t \\ v^T & v_3 \end{pmatrix} = \begin{pmatrix} sR & t \\ 0^T & 1 \end{pmatrix} \begin{pmatrix} K & 0 \\ 0^T & 1 \end{pmatrix} \begin{pmatrix} I & 0 \\ v^T & v_3 \end{pmatrix}$$

1. undo proper projective transformation (**affine rectification**):
 - ▶ then original and image differ only by an affine transformation
 - ▶ \rightsquigarrow measure **affine properties** of the original in the image
(= properties invariant under affine transformations)
 - ▶ parallel lines, ratio of lengths on parallel lines
2. undo proper affine transformation (**metric rectification**):
 - ▶ then original and image differ only by a similarity transformation
 - ▶ \rightsquigarrow measure **metric properties** of the original in the image
(= properties invariant under similarity transformations)
 - ▶ angles, ratio of lengths

Recovery of Affine Properties

Undo proper projective transformation:

$$\begin{pmatrix} I & 0 \\ v^T & v_3 \end{pmatrix} : \begin{pmatrix} x_1 \\ x_2 \\ 0 \end{pmatrix} \mapsto \begin{pmatrix} x_1 \\ x_2 \\ v_1 x_1 + v_2 x_2 \end{pmatrix}$$

$$l_\infty := \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix} \mapsto \begin{pmatrix} -v/v_3 \\ 1/v_3 \end{pmatrix} = \frac{1}{v_3} \begin{pmatrix} v_1 \\ v_2 \\ 1 \end{pmatrix}$$

- ▶ maps line at infinity to finite line $(v_1, v_2, 1)^T$
- ▶ to undo:
 - ▶ locate image $(v_1, v_2, 1)^T$ of line at infinity
 - ▶ undo by applying the inverse $H^{-1} = \begin{pmatrix} I & 0 \\ -v^T/v_3 & 1/v_3 \end{pmatrix}$

Note: Lines transform by H^{-T} : $\begin{pmatrix} I & 0 \\ v^T & v_3 \end{pmatrix}^{-T} = \begin{pmatrix} I & -v/v_3 \\ 0 & 1/v_3 \end{pmatrix}$.

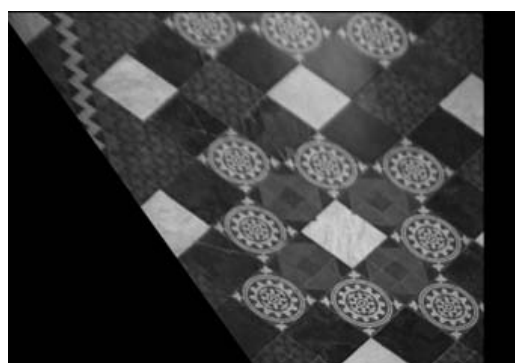
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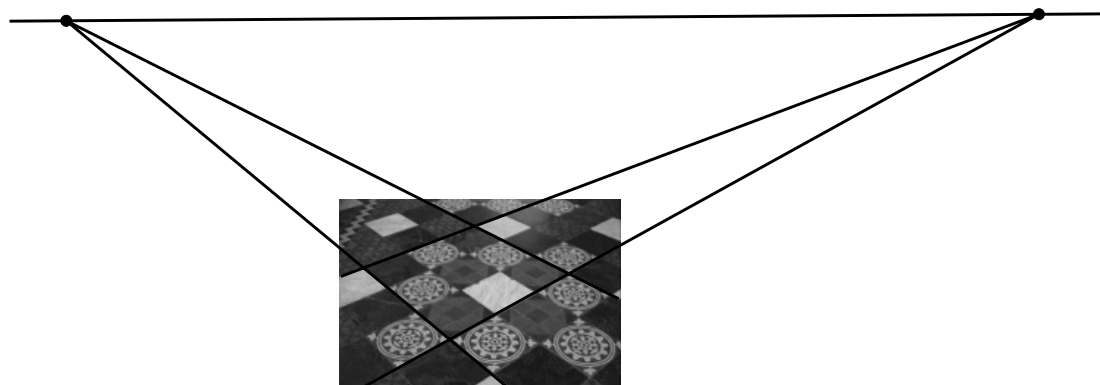
Recovery of Affine Properties / Example



a



b



Now we can measure area ratios !

[HZ04, p. 50]

Recovery of Affine Properties / Algorithm

```

1: procedure RECTIFY-AFFINE-TWO-PARALLELS( $a^1, a^2, b^1, b^2 \in \mathbb{P}^2$ )
2:    $s^1 := a^1 \times a^2$  ▷ compute intersection of parallels  $a^1, a^2$ 
3:    $s^2 := b^1 \times b^2$  ▷ compute intersection of parallels  $b^1, b^2$ 
4:    $l_\infty := s^1 \times s^2$  ▷ compute image of line at infinity
5:    $H^{-1} := \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -l_{\infty,1}/l_{\infty,3} & -l_{\infty,2}/l_{\infty,3} & 1/l_{\infty,3} \end{pmatrix}$  ▷ compute inverse
6:   return  $H^{-1}$ 
  
```

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Computer Vision 5. Angles in the Projective Plane

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Circular Points

A conic

$$C := \begin{pmatrix} a_1 & a_2/2 & a_4/2 \\ a_2/2 & a_3 & a_5/2 \\ a_4/2 & a_5/2 & a_6 \end{pmatrix} = \begin{pmatrix} a_1 & 0 & a_4/2 \\ 0 & a_1 & a_5/2 \\ a_4/2 & a_5/2 & a_6 \end{pmatrix}$$

is a circle if $a_1 = a_3$ and $a_2 = 0$.

Ideal points $x = (x_1, x_2, 0)^T$ on a circle:

$$x^T C x = a_1 x_1^2 + a_1 x_2^2 = 0$$

are exactly the **circular points**:

$$I := \begin{pmatrix} 1 \\ i \\ 0 \end{pmatrix}, \quad J := \begin{pmatrix} 1 \\ -i \\ 0 \end{pmatrix}$$

Line Conics

$C \in \text{Sym}(\mathbb{P}^{3 \times 3})$ defines a **point conic** via

$$\mathbf{C}_C := \{x \in \mathbb{P}^2 \mid x^T C x = 0\}$$

It also can be used to define a **line conic** / **dual conic**:

$$\mathbf{C}_C^* := \{a \in \mathbb{P}^2 \mid a^T C a = 0\}$$

(where a denotes a line)

Adjugate of a Matrix

For a square matrix $A \in \mathbb{R}^{n \times n}$,

$$A^* \in \mathbb{R}^{n \times n} \text{ with } A_{i,j}^* := (-1)^{i+j} \det A_{-j,-i}$$

is called its **adjugate** A^* .

It holds:

- ▶ for any A : $A^* A = A A^* = (\det A) I$
- ▶ A^* is continuous in A .
- ▶ if A is invertible, the adjoint is the scaled inverse: $A^* = (\det A) A^{-1}$
- ▶ if A is not invertible, the adjoint nullifies A : $A^* A = A A^* = 0$
- ▶ the adjugate is the transposed of the cofactor matrix.

Note: $A_{-j,-i}$ denotes the matrix A with row j and column i removed.

The adjugate is also called **adjoint**.

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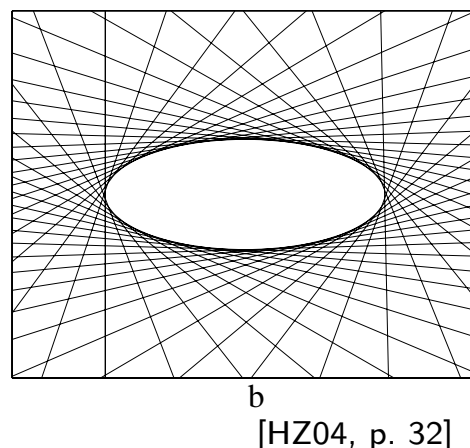
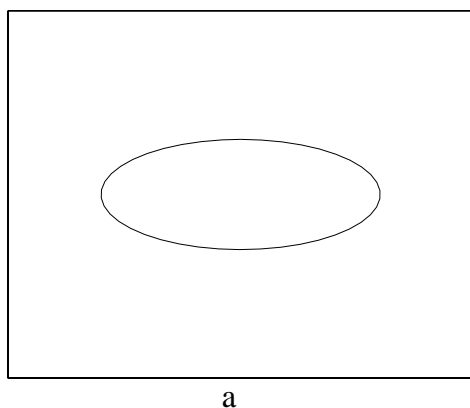
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Dual Conic

For any point conic $C \in \text{Sym}(\mathbb{P}^{3 \times 3})$, the set of tangent lines

- ▶ forms a line conic,
- ▶ parametrized by the adjugate C^* :

$$\{a \in \mathbb{P}^2 \mid a \text{ tangent to } C\} = \mathbf{C}_C^*$$



[HZ04, p. 32]

Dual Conic to the Circular Points

Dual conic to the circular points (degenerate):

$$C_{\infty}^* := IJ^T + JI^T = \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

- ▶ contains exactly all lines through the circular points I or J .
- ▶ transforms as HC^*H^T : $H(C_{\infty}^*) = C_{HC^*H^T}^*$.
- ▶ fixed under projectivity H iff H is a similarity.
- ▶ 4 dof (general C has 5, minus 1 due to $\det C = 0$)
- ▶ I_{∞} is the null vector of C_{∞}^* .

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Computer Vision 5. Angles in the Projective Plane

Angles in the Projective Plane

Angles are defined as:

$$\cos \theta(a, b) := \frac{a^T C_{\infty}^* b}{\sqrt{(a^T C_{\infty}^* a)(b^T C_{\infty}^* b)}}, \quad a, b \in \mathbb{P}^2$$

- ▶ for the canonical C_{∞}^* , coincides with the Euclidean definition:

$$\cos \theta(a, b) := \frac{a^T b}{\sqrt{(a^T a)(b^T b)}}, \quad a, b \in \mathbb{R}^2$$

- ▶ stays invariant under projective transformation:

$$\begin{aligned} a' &= H^{-T} a, & b' &= H^{-T} b, & C_{\infty}^{*'} &= HC_{\infty}^* H^T \\ a'^T C_{\infty}^{*'} b' &= a^T H^{-1} H C_{\infty}^* H^T H^{-T} b = a^T C_{\infty}^* b \end{aligned}$$

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Outline

1. Very Brief Introduction
2. The Projective Plane
3. Projective Transformations
4. Recovery of Affine Properties from Images
5. Angles in the Projective Plane
6. Recovery of Metric Properties from Images
7. Organizational Stuff

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Computer Vision 6. Recovery of Metric Properties from Images

Recovery of Metric Properties

- ▶ assume there is no pure projective transformation (i.e., affine rectification already done).
- ▶ need only to find pure affine transformation:

$$H_a := \begin{pmatrix} K & 0 \\ 0^T & 1 \end{pmatrix}, \quad \text{with } K \text{ upper triangular}$$

- ▶ under H_a we get $C_\infty^{* '}$ as

$$C_\infty^{* ' := H_a C_\infty^* H_a^T = \begin{pmatrix} KK^T & 0 \\ 0^T & 0 \end{pmatrix}$$

1. find symmetric matrix $S := KK^T$
2. find K via Cholesky decomposition of S

Recovery of Metric Properties (2/2)

- ▶ for two lines a', b' that are orthogonal in the original:

$$\begin{aligned}
 0 &= a'^T C_{\infty}^* b' = a'_{1:2}^T S b_{1:2} \\
 &= a'_1 S_{1,1} b'_1 + a'_1 S_{1,2} b'_2 + a'_2 S_{2,1} b'_1 + a'_2 S_{2,2} b'_2 \\
 &= a'_1 b'_1 S_{1,1} + (a'_1 b'_2 + a'_2 b'_1) S_{1,2} + a'_2 b'_2 S_{2,2} \\
 &= (a'_1 b'_1, a'_1 b'_2 + a'_2 b'_1, a'_2 b'_2) (S_{1,1}, S_{1,2}, S_{2,2})^T
 \end{aligned}$$

we get 1 linear constraint in $s := (S_{1,1}, S_{1,2}, S_{2,2})^T$.

- ▶ for two pairs of lines that are orthogonal in the original we get 2 linear constraints for 3 variables

$$\begin{pmatrix} a'_1 b'_1 & a'_1 b'_2 + a'_2 b'_1 & a'_2 b'_2 \\ c'_1 d'_1 & c'_1 d'_2 + c'_2 d'_1 & c'_2 d'_2 \end{pmatrix} s$$

where $s \neq 0$ has to be identified only up to a factor.

Recovery of Metric Properties / Algorithm

1: **procedure**

RECTIFY-METRIC-TWO-ORTHOGONALS($a^1, a^2, b^1, b^2 \in \mathbb{P}^2$)

$$2: \quad A := \begin{pmatrix} a_1^1 a_1^2 & a_1^1 a_2^2 + a_2^1 a_1^2 & a_2^1 a_2^2 \\ b_1^1 b_1^2 & b_1^1 b_2^2 + b_2^1 b_1^2 & b_2^1 b_2^2 \end{pmatrix}$$

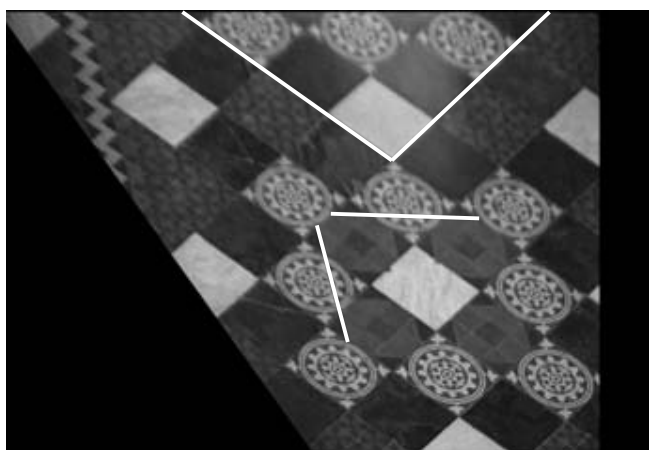
$$3: \quad \text{find } s \neq 0 : As = 0 \quad \triangleright \text{find } C_{\infty}^* := \begin{pmatrix} s_1 & s_2 & 0 \\ s_2 & s_3 & 0 \\ 0 & 0 & 0 \end{pmatrix}$$

$$4: \quad K := \text{cholesky}\left(\begin{pmatrix} s_1 & s_2 \\ s_2 & s_3 \end{pmatrix}\right) \quad \triangleright \text{find } H := \begin{pmatrix} K & 0 \\ 0^T & 1 \end{pmatrix}$$

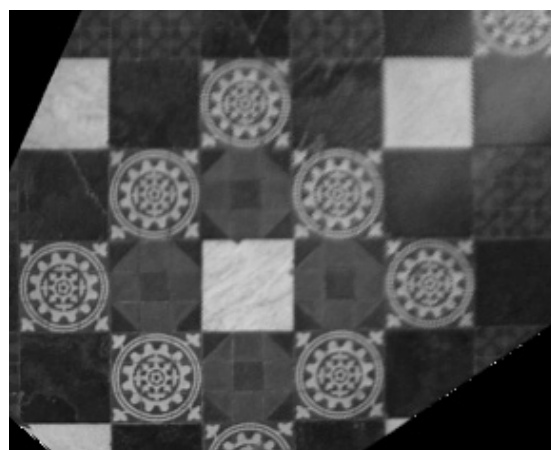
$$5: \quad H^{-1} := \begin{pmatrix} 1/K_{1,1} & -1/(K_{1,2}K_{2,2}) & 0 \\ 0 & 1/K_{2,2} & 0 \\ 0 & 0 & 1 \end{pmatrix} \quad \triangleright \text{compute inverse}$$

6: **return** H^{-1}

Recovery of Metric Properties / Example



a



b

a) affine rectified image

b) metric rectified image

Now we can measure angles and length ratios !

[HZ04, p. 57]

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Computer Vision 7. Organizational Stuff

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Outline

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7. Organizational Stuff

Exercises and Tutorials

- ▶ There will be a weekly sheet with 4 exercises handed out **each Tuesday** in the lecture.
1st sheet will be handed out Thu. 23.4. in the tutorial.
- ▶ Solutions to the exercises can be submitted until **next Tuesday noon**
1st sheet is due Tue. 28.4.
- ▶ Exercises will be corrected.
- ▶ Tutorials **each Thursday 2pm–4pm**,
1st tutorial at Thur. 23.4.
- ▶ Successful participation in the tutorial gives up to 10% bonus points for the exam.

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Computer Vision 7. Organizational Stuff

Exam and Credit Points

- ▶ There will be a written exam at end of term (2h, 4 problems).
- ▶ The course gives 6 ECTS (2+2 SWS).
- ▶ The course can be used in
 - ▶ IMIT MSc. / Informatik / Gebiet KI & ML
 - ▶ Wirtschaftsinformatik MSc / Informatik / Gebiet KI & ML
 - ▶ as well as in both BSc programs.

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Some Text Books

- ▶ Simon J. D. Prince (2012):
Computer Vision: Models, Learning, and Inference,
Cambridge University Press.
- ▶ Richard Szeliski (2011):
Computer Vision, Algorithms and Applications,
Springer.
- ▶ David A. Forsyth, Jean Ponce (²2012, 2007):
Computer Vision, A Modern Approach,
Prentice Hall.
- ▶ Richard Hartley, Andrew Zisserman (2004):
Multiple View Geometry in Computer Vision,
Cambridge University Press.

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Computer Vision 7. Organizational Stuff

Some First Computer Vision Software

- ▶ Open Computer Vision Library (OpenCV)
 - ▶ C++ library
 - ▶ has wrappers for Python & Octave
 - ▶ originally developed by Intel
 - ▶ v3.0 beta, 11/2014; <http://opencv.org>

Public data sets:

- ▶ ...

Summary (1/3)

- ▶ The **projective plane** \mathbb{P}^2 is an extension of the Euclidean plane with **ideal points**.
- ▶ Points and lines in \mathbb{P}^2 are parametrized by **homogeneous coordinates**.
- ▶ Each two parallels intersect in an ideal point, all ideal points form the **line at infinity** l_∞ .
- ▶ Each circle contains two ideal points, the **circular points**, all lines through the circular points form the **dual conic to the circular points** C_∞^* .
- ▶ **Conics** are curves of order 2 (hyperbolas, parabolas, ellipses), parametrized by a symmetric matrix C containing all points x with $x^T C x = 0$.

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Computer Vision 7. Organizational Stuff

Summary (2/3)

- ▶ **Projectivities** H are invertible mappings of \mathbb{P}^2 onto \mathbb{P}^2 that preserve lines.
- ▶ Lines a transform via $H^{-T} a$, conics C via $H^{-T} C H^{-1}$.
- ▶ There exist several subgroups of the group of projectivities:
 - ▶ **Isometries** **rotate** and **translate** figures.
 - ▶ preserving lengths
 - ▶ **Similarities** additionally **(isotropic) scale** figures.
 - ▶ preserving ratio of lengths, angle
 - ▶ **Affine transforms** additionally **non-isotropic scale** figures.
 - ▶ preserving ratio of lengths on parallel lines, parallel lines
 - ▶ **Projectivities** additionally **move the line at infinity**.
 - ▶ preserving cross ratio
- ▶ Any projectivity can be decomposed into a chain of a pure projective, a pure affine transform and a similarity.

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Summary (3/3)

- ▶ Images distorted by an projective transformation can be **rectified** (i.e., undoes the projective transformation).
- ▶ **Affine rectification**
 - ▶ undoes a proper projective transformation
 - ▶ moves the **line at infinity** back to its **canonical position**.
 - ▶ allows to measure **affine properties**:
 - ▶ ratio of lengths on parallel lines, parallel lines
 - ▶ requires, e.g., **two pairs of parallel lines**.
- ▶ **Metric rectification**
 - ▶ undoes a proper affine transformation
 - ▶ moves the **dual conic to the circular points** back to its canonical position.
 - ▶ allows to measure **metric properties**:
 - ▶ angles, ratio of lengths
 - ▶ requires, e.g., two pairs of orthogonal lines.

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Computer Vision

Further Readings

- ▶ [HZ04, ch. 1 and 2].

References



Richard Hartley and Andrew Zisserman.

Multiple view geometry in computer vision.

Cambridge university press, 2004.