

ECE 417/598: Camera calibration

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¹See Hartley and Zisserman's Multiple View Geometry for details.

Announcements

- ▶ Midterm next Friday. ✓
- ▶ Will cover everything until today's lecture
- ▶ I will release Sample exam on Monday
- ▶ No programming questions. Expect Linear Algebra question though. Only the concepts that we have touched in class.
- ▶ Grading: compare your answers to the solution.

Rot transform

Rot_C transformation

camera projection

pseudo inverse

stronger

$$u \approx v(\epsilon) \equiv |u - v| \leq \epsilon \min(|u|, |v|)$$

(1)

weaker

$$u \sim v(\epsilon) \equiv |u - v| \leq \epsilon \max(|u|, |v|)$$

(2)

double $d = \frac{\text{R.determinant}(G)}{\text{float or double}} \approx 1$ $\text{abs}(d-1) < \text{precision}$

float
or double

double precision

std::numeric_limit < double::min

Homogeneous representation of lines

$$\underline{l}_i = k \underline{l}_j$$

↑
Scalar

$$ax + by + c = 0$$

$$ax + by + c \cdot 1 = 0$$

$$\underbrace{[a \ b \ c]}_{\underline{l}} \underbrace{\begin{bmatrix} x \\ y \\ 1 \end{bmatrix}}_{\underline{x}} = 0$$

$$\underline{l}^T \underline{x} = 0$$

$$a:b$$

$$b:c$$

$$\underline{x}_i = \lambda \underline{x}_j$$

↑
 \mathbb{R}

$$a(\lambda x) + b(\lambda y) + c(\lambda 1) = 0$$

$$(ka)x + (kb)y + (kc)1 = 0$$

Homogeneous
line
= \underline{l}

$$\underline{l}^T \underline{x} = 0$$

Projective space

 \mathbb{P}^2

$$2D- \quad \mathbb{P}^2 = \mathbb{R}^3 - \{(0, 0, 0)^T\}$$

$$\underline{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \end{pmatrix}$$

$$\underline{x}_i = \lambda \underline{x}_j$$

$$\lambda \neq 0$$

$$\mathbb{P}^3 = \mathbb{R}^4 - \{(0, 0, 0, 0)^T\}$$

$$\underline{x} = \begin{pmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{pmatrix}$$

$$\underline{x}_i = \lambda \underline{x}_j$$

Homogeneous representation of points

$$ax + by + c = 0$$

$$\underline{x} = \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = \begin{bmatrix} \lambda x \\ \lambda y \\ \lambda \end{bmatrix} = \begin{bmatrix} x' \\ y' \\ w' \end{bmatrix}$$

$$\underline{l} = \begin{bmatrix} a \\ b \\ c \end{bmatrix}$$

$$\underline{l}^T \underline{x} = 0$$

Eq of line in Projective space

The point $\mathbf{x} \in \mathbb{P}^2$ lies on a line \mathbf{l} if and only if $\mathbf{x}^T \mathbf{l} = 0$.

$\in \mathbb{P}^2$ $\in \mathbb{R}^3$ $\in \mathbb{P}^2$ $\in \mathbb{R}^3$
Points are rays and lines are planes

$$\underline{u} = KX$$

intrinsic 3×3
camera

$$\underline{u} = \begin{bmatrix} u \\ v \\ 1 \end{bmatrix}$$

$$\underline{X} = \begin{bmatrix} x \\ y \\ z \end{bmatrix}$$

ideal point

$$\underline{u} \in \mathbb{P}^2$$

$$\underline{X} \in \mathbb{R}^3$$

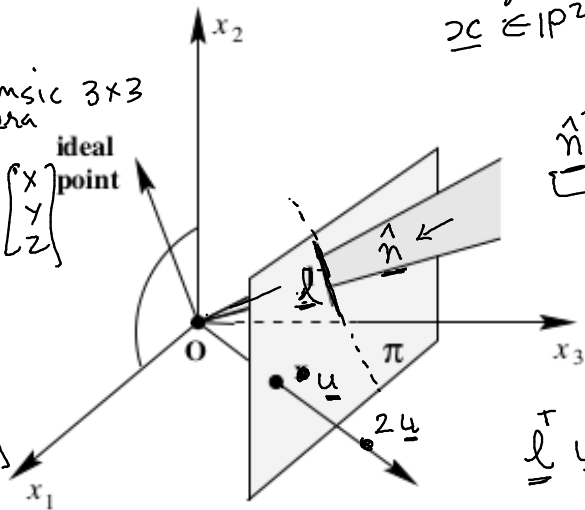
$$\underline{u} = \begin{bmatrix} \lambda u \\ \lambda v \\ \lambda \end{bmatrix}$$

Image model

$$\underline{c} \in \mathbb{P}^2$$

$$\hat{n}^T X = 0$$

plane



$$l^T u = 0$$

Intersection of lines

$$\underline{l}_1 = \begin{bmatrix} a_1 \\ b_1 \\ c_1 \end{bmatrix}$$

$$\underline{l}_2 = \begin{bmatrix} a_2 \\ b_2 \\ c_2 \end{bmatrix}$$

$$a_1 x + b_1 y + c_1 = 0$$

$$a_2 x + b_2 y + c_2 = 0$$

Find a point \underline{x} that lies on both \underline{l}_1 and \underline{l}_2

$$\begin{bmatrix} a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \end{bmatrix} \begin{bmatrix} x \\ y \\ 1 \end{bmatrix} = 0$$

$$\begin{bmatrix} a_1 & b_1 \\ a_2 & b_2 \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} -c_1 \\ -c_2 \end{bmatrix}$$

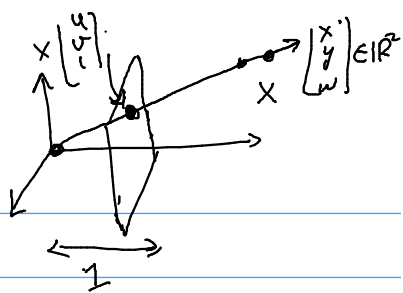
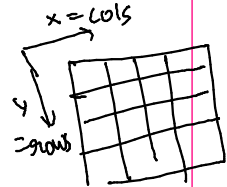
$$\begin{bmatrix} \underline{l}_1^T \underline{x} = 0 \\ \underline{l}_2^T \underline{x} = 0 \end{bmatrix}$$

$$\underline{x} = \underline{l}_1 \times \underline{l}_2$$

\underline{l}_2^T

old way





$$\underline{X} = \underline{l}_1 \times \underline{l}_2$$

$$\begin{bmatrix} x \\ y \\ w \end{bmatrix} = \begin{vmatrix} i & j & k \\ a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \end{vmatrix}$$

$$\begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \frac{1}{w} \begin{bmatrix} x \\ y \\ w \end{bmatrix} = \begin{vmatrix} i & j & k \\ a_1 & b_1 & c_1 \\ a_2 & b_2 & c_2 \end{vmatrix}$$

$$\underline{u} = K \underline{X}$$

Intersection of parallel lines

Find the intersection of parallel lines $\frac{l_1}{ax + by + c = 0}$ and $ax + by + c' = 0$.

$$\underline{x} = \frac{l_1 \times l_2}{\begin{vmatrix} a & b & c \\ a & b & c' \end{vmatrix}}$$

$$= \begin{bmatrix} bc' - bc \\ -(ac' - ac) \\ ab - ab \end{bmatrix} = \begin{bmatrix} b \\ -a \\ 0 \\ c \end{bmatrix} (c' - c) =$$

$\underline{x} = \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \in \mathbb{P}^2$ with $x_3 = 0$ then it is a point at ∞ \underline{x}_∞

Defn. Ideal point:

$$\underline{x} \in \mathbb{P}^2 : x_3 = 0$$

$$\underline{x} = \begin{bmatrix} x_1 \\ x_2 \\ 0 \end{bmatrix} \quad \underline{l}^T \underline{x} = 0 \quad \forall (x_1, x_2)$$

$$[l_1 \ l_2 \ l_3] \begin{bmatrix} x_1 \\ x_2 \\ 0 \end{bmatrix} = 0$$

$$l_1 x_1 + l_2 x_2 + l_3 \cdot 0 = 0$$

$\forall x_1, x_2$

$$l_1 = 0 \quad l_2 = 0$$

$$l_3 \neq 0$$

$$l_1 = l_2 = l_3 = 0 \\ \in \mathbb{R}^2$$

$$l_{-\infty} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$$

line at infinity

all ideal points $\underline{x} = \begin{bmatrix} x_1 \\ x_2 \\ 0 \end{bmatrix}$

lie on $l_{-\infty} = \begin{bmatrix} 0 \\ 0 \\ 1 \end{bmatrix}$

Numerical example

$$\begin{array}{l} \text{line} \\ \text{line} \end{array} \quad \begin{array}{l} x=1 \\ x-1=0 \end{array}$$

Find the intersection of $x=1$ and $y=1$ using perspective geometry.

$$ax + by + c = 0$$
$$1x + 0y - 1 = 0$$

$$\underline{l_1} = \begin{bmatrix} 1 \\ 0 \\ -1 \end{bmatrix}$$

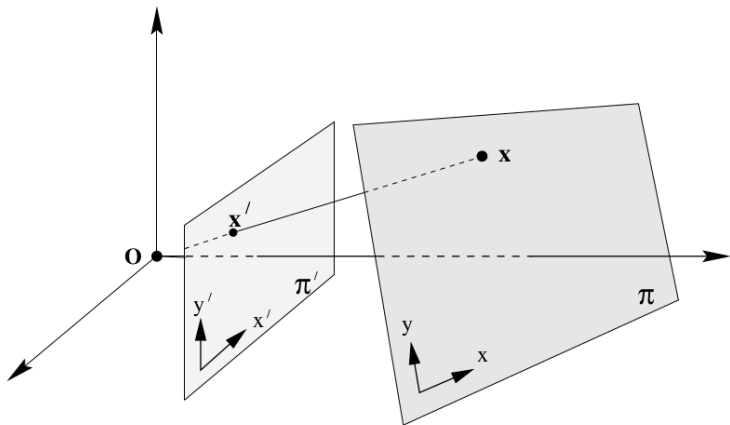
$$y=1$$
$$0x + 1y - 1 = 0$$

$$\underline{l_2} = \begin{bmatrix} 0 \\ 1 \\ -1 \end{bmatrix}$$

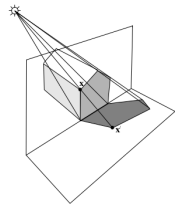
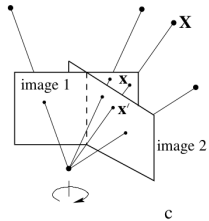
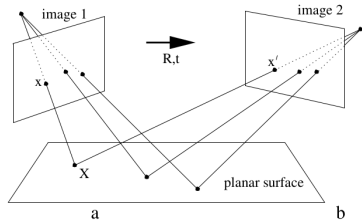
$$\underline{x} = \underline{l_1} \times \underline{l_2}$$

Line joining points

Homography



Examples of Homography



Computing Homography





2D homography

Given a set of points $\mathbf{x}_i \in \mathbb{P}^2$ and a corresponding set of points $\mathbf{x}'_i \in \mathbb{P}^2$, compute the projective transformation that takes each \mathbf{x}_i to \mathbf{x}'_i . In a practical situation, the points \mathbf{x}_i and \mathbf{x}'_i are points in two images (or the same image), each image being considered as a projective plane \mathbb{P}^2 .

Direct Linear Transformation (DLT) algorithm

Objective

Given $n \geq 4$ 2D to 2D point correspondences $\{\mathbf{x}_i \leftrightarrow \mathbf{x}'_i\}$, determine the 2D homography matrix \mathbf{H} such that $\mathbf{x}'_i = \mathbf{H}\mathbf{x}_i$.

Algorithm

- (i) For each correspondence $\mathbf{x}_i \leftrightarrow \mathbf{x}'_i$ compute the matrix \mathbf{A}_i from (4.1). Only the first two rows need be used in general.
- (ii) Assemble the $n \times 9$ matrices \mathbf{A}_i into a single $2n \times 9$ matrix \mathbf{A} .
- (iii) Obtain the SVD of \mathbf{A} (section A4.4(p585)). The unit singular vector corresponding to the smallest singular value is the solution \mathbf{h} . Specifically, if $\mathbf{A} = \mathbf{U}\mathbf{D}\mathbf{V}^T$ with \mathbf{D} diagonal with positive diagonal entries, arranged in descending order down the diagonal, then \mathbf{h} is the last column of \mathbf{V} .
- (iv) The matrix \mathbf{H} is determined from \mathbf{h} as in (4.2).

3D to 2D camera projection matrix estimation

Given a set of points \mathbf{X}_i in 3D space, and a set of corresponding points \mathbf{x}_i in an image, find the 3D to 2D projective \mathbf{P} mapping that maps \mathbf{X}_i to $\mathbf{x}_i = \mathbf{P}\mathbf{X}_i$.

Eigenvalues and Eigenvectors

For a square matrix A , the λ_i and \mathbf{x}_i that satisfy the following equation are called eigenvalues and eigenvectors respectively.

$$A\mathbf{x} = \lambda\mathbf{x} \text{ or } (A - \lambda I)\mathbf{x} = 0 \quad (3)$$

λ is chosen to ensure that $A - \lambda I$ has null space, hence, characteristic equation

$$\det(A - \lambda I) = 0 \quad (4)$$

For symmetric matrix $A = A^T$, eigenvalues are real, and eigenvectors are orthonormal,

$$A[\mathbf{x}_1, \dots, \mathbf{x}_n] = [\mathbf{x}_1, \dots, \mathbf{x}_n] \begin{bmatrix} \lambda_1 & \dots & 0 \\ \vdots & \ddots & \vdots \\ 0 & \dots & \lambda_n \end{bmatrix} \quad (5)$$

$$AS = SA \quad (6)$$

$$\text{if } A = A^T \text{ then } A = SAS^T \quad (7)$$

Singular Value Decomposition (SVD)

$$A = U \begin{bmatrix} \Sigma & 0 \\ 0 & 0 \end{bmatrix} V^T \quad (8)$$

$$A^T A = V \Sigma^2 V^{-1} \quad (9)$$

$$A^T A \mathbf{v}_i = \lambda_i \mathbf{v}_i \quad \lambda_i = \sigma_i^2 \quad (10)$$

$$AV = U \begin{bmatrix} \Sigma & 0 \\ 0 & 0 \end{bmatrix} \quad (11)$$

$$U^+ = \Sigma^{-1} AV^+ \quad (12)$$