

Calibration by a spatial object

- Given m point correspondence between 3D scene and images plane $\mathbf{X}_i \rightarrow \mathbf{u}_i: \mathbf{u}_i \sim P\mathbf{X}_i, \quad i = 1, \dots, m$
- Task: estimation of $P = KR[I] - t$.
 - At least 6 correspondences required
 - Over-determined system
- Wrong correspondence \rightarrow robust methods
 - Many correspondences \rightarrow outlier detection

Calibration by Cartesian coordinates

$$\alpha \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} P_{11} & P_{12} & P_{13} & P_{14} \\ P_{21} & P_{22} & P_{23} & P_{24} \\ P_{31} & P_{32} & P_{33} & P_{34} \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix},$$

where $\alpha \neq 0$ is an arbitrary scale factor.

Equations can be rewritten as

$$u = \frac{P_{11}X + P_{12}Y + P_{13}Z + P_{14}}{P_{31}X + P_{32}Y + P_{33}Z + P_{34}} = \frac{\mathbf{p}_1^T \mathbf{X}}{\mathbf{p}_3^T \mathbf{X}}, \quad (22)$$

$$v = \frac{P_{21}X + P_{22}Y + P_{23}Z + P_{24}}{P_{31}X + P_{32}Y + P_{33}Z + P_{34}} = \frac{\mathbf{p}_2^T \mathbf{X}}{\mathbf{p}_3^T \mathbf{X}}, \quad (23)$$

where \mathbf{p}_i is the i -th row of projection matrix P .

Linear estimation of a projection matrix 1/2

Equations are multiplied by the common denominator:

$$(P_{31}X + P_{32}Y + P_{33}Z + P_{34})u = P_{11}X + P_{12}Y + P_{13}Z + P_{14} \quad (24)$$

$$(P_{31}X + P_{32}Y + P_{33}Z + P_{34})v = P_{21}X + P_{22}Y + P_{23}Z + P_{24} \quad (25)$$

For the i -th point, $A_i \mathbf{p} = 0$, where

$$A_i = \begin{bmatrix} X_i & Y_i & Z_i & 1 & 0 & 0 & 0 & 0 & -u_i X_i & -u_i Y_i & -u_i Z_i & -u_i \\ 0 & 0 & 0 & 0 & X_i & Y_i & Z_i & 1 & -v_i X_i & -v_i Y_i & -v_i Z_i & -v_i \end{bmatrix} \quad (26)$$

$$\mathbf{p} = [P_{11}, P_{12}, P_{13}, P_{14}, P_{21}, P_{22}, P_{23}, P_{24}, P_{31}, P_{32}, P_{33}, P_{34}]^T \quad (27)$$

For all the points, a homogeneous linear system of equation obtained in the form $A\mathbf{p} = \mathbf{0}$ where

$$A = [A_1, A_2, \dots, A_m]^T$$

Linear estimation of a projection matrix 2/2

- $\mathbf{p} = \mathbf{0}$ trivial solution omitted.
 - estimation obtained up to a scale
 - norm is fixed as $\|\mathbf{p}\| = 1$
- For noiseless case
 - rank of \mathbf{A} is 11, perfect solution is obtained
- For over-determined and noisy case,
 - only estimation can be computed
 - minimization of $\|\mathbf{A}\mathbf{p}\|$ subject to: $\|\mathbf{p}\| = 1$.
 - optimal solution if the eigenvector of $\mathbf{A}^T \mathbf{A}$ corresponding to the least eigenvalue.
 - solution can be obtained by Singular Value Decomposition (SVD) as well.

Decomposition of a projection matrix

- Structure of a projection matrix:

$$\mathbf{P} = \mathbf{KR} [\mathbf{I} | -\mathbf{t}] \quad (28)$$

- First three columns of matrix \mathbf{P} : $\mathbf{P}_{3 \times 3} = \mathbf{KR}$
 - Decomposition can be obtained by RQ - decomposition
 - It decomposes \mathbf{P} into product of an upper triangular and an orthonormal matrices
- Last column of matrix \mathbf{P} :

$$\mathbf{p}_4 = -\mathbf{KRt} \quad (29)$$

- Thus,

$$\mathbf{t} = -\mathbf{R}^T \mathbf{K}^{-1} \mathbf{p}_4 \quad (30)$$

Data normalization

- Point coordinates can be normalized similarly to homography estimation
- Original transformation: $[u, v, 1]^T \sim \mathbf{P}[X, Y, Z, 1]^T$
- Normalizing transformations: $[u', v', 1]^T = \mathbf{T}_{2D}[u, v, 1]^T$ and $[X', Y', Z', 1]^T = \mathbf{T}_{3D}[X, Y, Z, 1]^T$
 - \mathbf{T}_{2D} 2D transformation(s) (size: 3×3)
 - \mathbf{T}_{3D} 3D transformation(s) (size: 4×4)
- Projection by normalized coordinates:
 $[u', v', 1]^T \sim \mathbf{P}'[X', Y', Z', 1]^T$
- Solution applied normalized coordinates:
 - $\mathbf{P} = \mathbf{T}_{2D}^{-1} \mathbf{P}' \mathbf{T}_{3D}$ or $\mathbf{P}' = \mathbf{T}_{2D} \mathbf{P} \mathbf{T}_{3D}^{-1}$.