

Camera Calibration

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Camera calibration

- Issues:
 - what are intrinsic parameters of the camera?
 - what is the camera matrix? (intrinsic+extrinsic)
- General strategy:
 - view calibration object; identify image points in image
 - obtain camera matrix by minimizing error
 - obtain intrinsic parameters from camera matrix

Optimization problem: Notation

The optimization problem is relatively straightforward to formulate. Notation is the main issue. We have N reference points $\mathbf{s}_i = [s_{x,i}, s_{y,i}, s_{z,i}]$ with known position in some reference coordinate system in 3D. The measured location in the image for the i 'th such point is $\hat{\mathbf{t}}_i = [\hat{t}_{x,i}, \hat{t}_{y,i}]$. There may be measurement errors, so the $\hat{\mathbf{t}}_i = \mathbf{t}_i + \xi_i$, where ξ_i is an error vector and \mathbf{t} is the unknown true position. We will assume the magnitude of error does not depend on direction in the image plane (it is *isotropic*), so it is natural to minimize the squared magnitude of the error

$$\sum_i \xi_i^T \xi_i.$$

The main issue here is writing out expressions for ξ_i in the appropriate coordinates. Write \mathcal{T}_i for the intrinsic matrix whose u, v 'th component will be i_{uv} ; \mathcal{T}_e for the extrinsic transformation, whose u, v 'th component will be e_{uv} . Recalling that \mathcal{T}_i is lower triangular, and engaging in some manipulation, we obtain

$$\sum_i \xi_i^T \xi_i = \sum_i (t_{x,i} - p_{x,i})^2 + (t_{y,i} - p_{y,i})^2$$

where

$$p_{x,i} = \frac{i_{11}g_{x,i} + i_{12}g_{y,i} + i_{13}g_{i,3}}{g_{i,3}}$$

$$p_{y,i} = \frac{i_{22}g_{x,i} + i_{23}g_{i,3}}{g_{i,3}}$$

and

$$g_{x,i} = e_{11}s_{x,i} + e_{12}s_{y,i} + e_{13}s_{z,i} + e_{14}$$

$$g_{y,i} = e_{21}s_{x,i} + e_{22}s_{y,i} + e_{23}s_{z,i} + e_{24}$$

$$g_{z,i} = e_{31}s_{x,i} + e_{32}s_{y,i} + e_{33}s_{z,i} + e_{34}$$

(which you should check as an exercise). This is a constrained optimization problem because \mathcal{T}_e is a Euclidean transformation. The constraints here are

$$1 - \sum_v e_{j,1v}^2 = 0 \text{ and } 1 - \sum_v e_{j,2v}^2 = 0 \text{ and } 1 - \sum_v e_{j,3v}^2 = 0$$

$$\sum_v e_{j,1v}e_{j,2v} = 0 \text{ and } \sum_v e_{j,1v}e_{j,3v} = 0 \text{ and } 1 - \sum_v e_{j,2v}e_{j,3v} = 0 \quad .$$

Optimizing

- Chuck in constrained optimizer and run
 - doesn't work well - need a start point

A Start Point

Write \mathbf{C}_j^T for the j 'th row of the camera matrix, and $\mathbf{S}_i = [s_{x,i}, s_{y,i}, s_{z,i}, 1]^T$ for homogeneous coordinates representing the i 'th point in 3D. Then, assuming no errors in measurement, we have

$$\hat{t}_{x,i} = \frac{\mathbf{C}_1^T \mathbf{S}_i}{\mathbf{C}_3^T \mathbf{S}_i} \text{ and } \hat{t}_{y,i} = \frac{\mathbf{C}_2^T \mathbf{S}_i}{\mathbf{C}_3^T \mathbf{S}_i},$$

which we can rewrite as

$$\mathbf{C}_3^T \mathbf{S}_i \hat{t}_{x,i} - \mathbf{C}_1^T \mathbf{S}_i = 0 \text{ and } \mathbf{C}_3^T \mathbf{S}_i \hat{t}_{y,i} - \mathbf{C}_2^T \mathbf{S}_i = 0.$$

2 Homogeneous equations per point - so with enough, can solve for C

RECALL: The camera matrix - II

- Turn previous expression into HC's
 - HC's for 3D point are (X,Y,Z,T)
 - HC's for point in image are (U,V,W)

Transforms points from object coordinates into world coordinates most likely a rotation and translation

$$\begin{pmatrix} U \\ V \\ W \end{pmatrix} = \mathcal{C} \begin{pmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{pmatrix} \mathcal{W} \begin{pmatrix} X \\ Y \\ Z \\ T \end{pmatrix}$$

Transforms camera coordinates
(f is hidden in here)

Camera matrix and focal point

- Trick:
 - camera matrix maps focal point to $(0, 0, 0)$
 - which means given a camera matrix, it is easy to extract focal point
 - which means that we can extract translation from camera matrix

Remember this: *Given a 3×4 camera matrix \mathcal{P} , the homogeneous coordinates of the focal point of that camera are given by \mathbf{X} , where $\mathcal{P}\mathbf{X} = [0, 0, 0]^T$*

Camera matrix and focal point

- Recall that extrinsic transformation is a rotation+translation

Remember this: Assume camera matrix \mathcal{P} has null space $\lambda \mathbf{u} = \lambda [\mathbf{f}^T, 1]^T$. Then we must have $\mathcal{T}_e \mathbf{u} = [0, 0, 0, 1]^T$, so we must have

$$\mathcal{T}_e = \begin{bmatrix} \mathcal{R} & -\mathcal{R}\mathbf{f} \\ \mathbf{0}^T & 1 \end{bmatrix}$$

This means that, if we know \mathcal{R} , we can recover the translation from the focal point. We must now recover the intrinsic transformation and \mathcal{R} from what we know.

$$\lambda \mathcal{P} = \mathcal{T}_i \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} \mathcal{R} & -\mathcal{R}\mathbf{f} \\ \mathbf{0}^T & 1 \end{bmatrix} = \begin{bmatrix} \mathcal{T}_i \mathcal{R} & -\mathcal{T}_i \mathcal{R}\mathbf{f} \end{bmatrix}$$

We do not know λ , but we do know \mathcal{P} . Now write \mathcal{P}_l for the left 3×3 block of \mathcal{P} , and recall that \mathcal{T}_i is upper triangular and \mathcal{R} orthonormal. The first question is the sign of λ . We expect $\text{Det}(\mathcal{R}) = 1$, and $\text{Det}(\mathcal{T}_i) > 0$, so $\text{Det}(\mathcal{P}_l)$ should be positive. This yields the sign of λ – choose a sign $s \in \{-1, 1\}$ so that $\text{Det}(s\mathcal{P}_l)$ is positive.

We can now factor $s\mathcal{P}_l$ into an upper triangular matrix \mathcal{T} and an orthonormal matrix \mathcal{Q} . This is an RQ factorization (Section 25.2). Recall we could not distinguish between scaling caused by the focal length and scaling caused by pixel scale, so that

$$\mathcal{T}_i = \begin{bmatrix} as & k & c_x \\ 0 & s & c_y \\ 0 & 0 & 1 \end{bmatrix}$$

In turn, we have $\lambda = s(1/t_{33})$, $c_y = (t_{23}/t_{33})$, $s = (t_{22}/t_{33})$, $c_x = (t_{13}/t_{33})$, $k = (t_{12}/t_{33})$, and $a = (t_{11}/t_{22})$.