Projection Models and Homogeneous Coordinates

Calibration - Part 1

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Topics

Calibration - Simplified Case

Definition
Calibration Patterns
Calibration of X-ray Systems

Summary

Take Home Messages Further Readings





Definition

The estimation of the projection parameters is called *calibration*.

Let us first consider a *simplified case* for calibration, where we consider the following problems in detail:

- Design a calibration pattern whose 3-D geometry is known exactly.
- Capture an X-ray image of the calibration pattern.
- Compute the correspondence of 2-D to 3-D points.
- Compute the focal length f using a least square estimator.





Calibration Patterns



Figure 1: Calibration pattern used for C-arm calibration (image courtesy of Siemens Healthcare)



Figure 2: Observed 2-D point features (image courtesy of Siemens Healthcare)





PDS2 Calibration Phantom

Example

Let us assume we have a so-called calibration pattern that is manufactured precisely, and N centroids of the 3-D metal spheres are known in 3-D world coordinates (x_0, x_1, x_2) :

$$X = \{ \mathbf{x}_i = (x_{i,0}, x_{i,1}, x_{i,2})^T \mid i = 1, \dots, N \}.$$

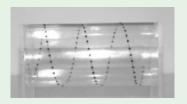


Figure 3: Calibration pattern used for the computation of the projection geometry of a C-arm





PDS2 Calibration Phantom

Example

Properties of the PDS2 calibration pattern used for today's C-arm systems:

- 108 steel spheres,
- · large spheres represent a logical one,
- small spheres represent a logical zero,
- 8 Bit binary encoding,
- position of spheres is known in the world coordinate system,
- the world coordinate system is attached to the phantom.





Let $(x_{i,0}, x_{i,1}, x_{i,2})^T$, i = 1, ..., N, be the set of 3-D points of the calibration pattern, and let $(y_{i,0}, y_{i,1})^T$ be the corresponding set of 2-D points.

The projection is defined by:

$$\left(\begin{array}{cccc} f & 0 & 0 & 0 \\ 0 & f & 0 & 0 \\ 0 & 0 & 1 & 0 \end{array}\right) \left(\begin{array}{c} x_{i,0} \\ x_{i,1} \\ x_{i,2} \\ 1 \end{array}\right) \cong \left(\begin{array}{c} y_{i,0} \\ y_{i,1} \\ 1 \end{array}\right).$$





For each pair of points 2-D/3-D points we get the pair of equations:

$$\frac{fx_{i,0}}{x_{i,2}} = y_{i,0},\tag{1}$$

$$\frac{fx_{i,1}}{x_{i,2}} = y_{i,1}. (2)$$

Due to image noise and segmentation errors, the equations have to be replaced by the following minimization problem:

$$\hat{f} = \underset{f}{\operatorname{arg\,min}} \sum_{i=1}^{N} \sum_{k=0}^{1} \left(\frac{f x_{i,k}}{x_{i,2}} - y_{i,k} \right)^{2}.$$
 (OP 1)





Solution: We compute the zero crossings of the partial derivatives w. r. t. the unknown parameter f and get the following estimator for f:

$$\hat{f} = \frac{\sum\limits_{i=1}^{N}\sum\limits_{k=0}^{1}\frac{y_{i,k}x_{i,k}}{x_{i,2}}}{\sum\limits_{i=1}^{N}\sum\limits_{k=0}^{1}\left(\frac{x_{i,k}}{x_{i,2}}\right)^{2}}.$$





Alternative Objective Functions

We can multiply (1) and (2) by $x_{i,2}$ and get:

$$fx_{i,0} = x_{i,2}y_{i,0},$$

 $fx_{i,1} = x_{i,2}y_{i,1}.$

The resulting objective function is:

$$\hat{f} = \underset{f}{\arg\min} \sum_{i=1}^{N} \sum_{k=0}^{1} (f x_{i,k} - x_{i,2} y_{i,k})^{2}.$$
 (OP 2)





There is yet another option to rewrite (1) and (2):

$$f = \frac{x_{i,2}y_{i,0}}{x_{i,0}},$$

$$f = \frac{x_{i,2}y_{i,1}}{x_{i,1}}.$$

Using these equations, the resulting objective function is:

$$\hat{f} = \underset{f}{\operatorname{arg\,min}} \sum_{i=1}^{N} \sum_{k=0}^{1} \left(f - x_{i,2} \frac{y_{i,k}}{x_{i,k}} \right)^{2}. \tag{OP 3}$$





Question: Which objective function is the best one? Should we optimize (OP 1), (OP 2) or (OP 3)?





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Exercise: Compute the estimators for (OP 1), (OP 2), and (OP 3) and analyze the sensitivity of estimates by adding noise to the input data.





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Take Home Messages

- In a real imaging system the detector has to be calibrated.
- Calibration can be performed by use of calibration phantoms that contain several markers whose pattern is exactly known.
- We have studied how a projection parameter like the focal length can be estimated by solving an optimization problem with measured 2-D points and known 3-D positions.





Further Readings

For further details on geometric aspects of imaging see:

- Richard Hartley and Andrew Zisserman. Multiple View Geometry in Computer Vision. 2nd ed. Cambridge: Cambridge University Press, 2004. DOI: 10.1017/CB09780511811685
- 2. Olivier Faugeras. *Three-Dimensional Computer Vision: A Geometric Viewpoint.* MIT Press, Nov. 1993