

Camera calibration and the search for infinity

Richard I. Hartley

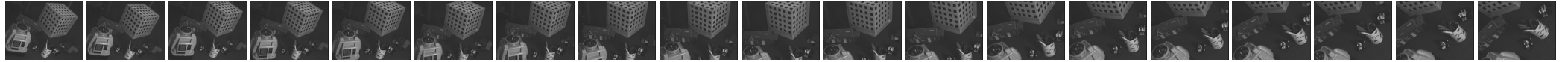
G.E. Corporate Research and Development

hartley@crd.ge.com

Eric Hayman, Lourdes de Agapito and Ian D. Reid

University of Oxford

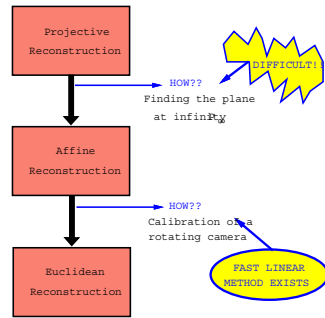
{hayman|lourdes|ian}@robots.ox.ac.uk



The objective

To self-calibrate a zooming camera undergoing general motion using a stratified approach.

- From an initial projective reconstruction of the scene points obtain an affine reconstruction and finally upgrade it to euclidean.



The problem

- Main difficulty: affine reconstruction, i.e. locating the plane at infinity (Π_∞) in the projective frame (an inherently non-linear problem).
- Once we have located Π_∞ the remainder of the problem is equivalent to the self-calibration of a rotating camera with varying intrinsics.
- Recently a couple of methods to achieve this goal have emerged, including a fast linear algorithm (de Agapito *et al.*, CVPR '99).

Our solution: to search for Π_∞

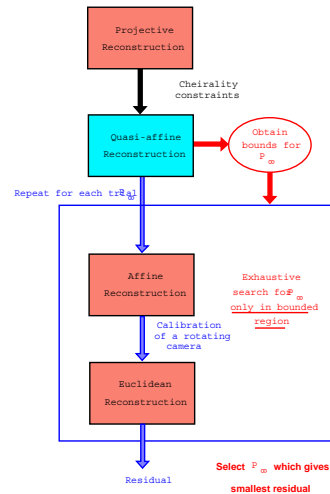
Perform an exhaustive search for the best plane at infinity:

- Hypothesize a candidate location for the plane at infinity.
- Obtain the corresponding affine reconstruction.
- Upgrade to euclidean using the method of calibration of a rotating camera.
- Use the residual from this method as the cost function to select the best Π_∞ .

Key point: use *cheirality* to narrow the search for the plane at infinity

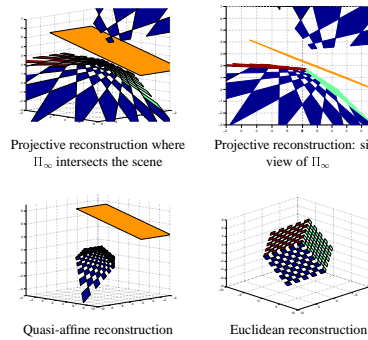
- First use cheirality to obtain an intermediate quasi-affine reconstruction (which preserves the convex hull of the set of points and camera centres).
- By translating the scene to the origin, since Π_∞ cannot pass through the origin, the fourth coordinate of Π_∞ may be set to unity.
- Cheirality is also used to set bounds on the remaining three coordinates of Π_∞ .
- Thus we may restrict the exhaustive search for Π_∞ to this bounded region.

The algorithm



What is a quasi-affine reconstruction?

- A projective reconstruction in which Π_∞ does not split the reconstructed scene.



- How is one obtained? By ensuring that all points lie in front of all cameras in which they are visible by solving the *cheirality inequalities* (Hartley IJCV '98).

The cheirality inequalities

- We assume a projective reconstruction $\{P_j, X_i\}$ with camera centres C^{P_j} .
- We seek a non-singular transformation, G with fourth row V^T such that each point has positive depth with respect to each camera. The necessary conditions are:

$$X_i^T V > 0 \text{ for all points } X_i$$

$$\epsilon C^{P_j T} V > 0 \text{ for all cameras } P_j$$

- The *signs* of X and C^{P_j} matter (see the paper for details).
- The inequalities may be solved by linear programming.
- There may be two oppositely orientated solutions for $\epsilon = \pm 1$. G is chosen such that $\text{sign}(\det G) = \epsilon$.
- If V is Π_∞ we obtain an affine reconstruction. In this case the infinite homographies H_j are obtained as the left 3×3 blocks of $P_j G^{-1}$.

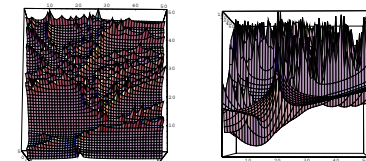
Locating the plane at infinity

- The cheirality inequalities place upper and lower bounds on the first 3 coordinates of Π_∞ .
- We perform an exhaustive search in this rectangular region in parameter space with the following steps for each trial.

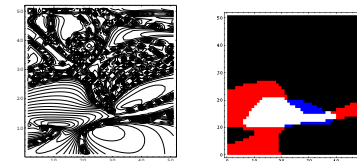
Locating the plane at infinity

- Cheirality test : The cheirality inequalities must be satisfied.
- If they are, obtain the infinite homographies, H_i , for this trial.
- Find camera intrinsics from an algorithm for self-calibration of a non-translating camera
- IAC test : If the images of the absolute conic are not positive definite then reject this trial.
- Otherwise, return a cost value (or vector) associated with the computed calibration.

- The minimum of the search may be used to initialize a non-linear optimization in the three parameters, using the same cost function.
- The error surface is well-behaved close to the minimum, but highly indented around it. A non-linear minimization initialized at a random point would therefore be sure to fail.



Logarithmic plot of the cost function at a cross-section through the search cube.



Contour plot of the cost function.

Blue: Cheirality test satisfied
Red: IAC test satisfied
White: Both satisfied
Black: Neither satisfied

Linear algorithm for self-calibration of non-translating cameras

The matrix of intrinsic parameters is:

$$K = \begin{pmatrix} \alpha_x & s & x_0 \\ & \alpha_y & y_0 \\ & & 1 \end{pmatrix}$$

The IAC in the i -th image, $\omega_i = K_i^T K_i^{-1}$, relates to ω_0 via the infinite homographies, H_i :

$$\omega_i = H_i^T \omega_0 H_i^{-1}$$

Under the assumption of zero skew, $s = 0$

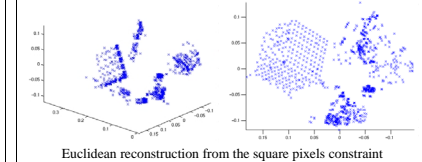
$$\omega = \begin{pmatrix} 1/\alpha_x^2 & 0 & -x_0/\alpha_x^2 \\ 0 & 1/\alpha_y^2 & -y_0/\alpha_y^2 \\ -x_0/\alpha_x^2 & -y_0/\alpha_y^2 & 1 + x_0^2/\alpha_x^2 + y_0^2/\alpha_y^2 \end{pmatrix}$$

We may then impose the following constraints:

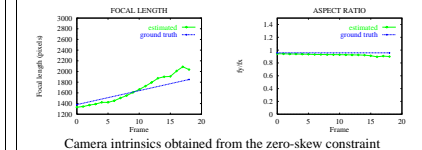
- Zero-skew** : If $s = K_{12} = 0$, then $\omega_{12} = 0$.
- Square-pixels** : If $s = 0$ and $\alpha_x = \alpha_y$, then $\omega_{11} - \omega_{22} = 0$.
- Known principal point** : If $s = 0$ and $x_0 = 0$, then $\omega_{13} = 0$. Similarly if $y_0 = 0$ then $\omega_{23} = 0$.

Together with the homography constraint this yields linear equations in the entries of ω_0 .

Experimental results



Euclidean reconstruction from the square pixels constraint



Camera intrinsics obtained from the zero-skew constraint