

### COL 780 - Assignment 3

#### Introduction:

A 2D point is denoted by  $m = [u, v]^T$ . A 3D point is denoted by  $M = [X, Y, Z]^T$ . We use  $x$  to denote the augmented vector by adding 1 as the last element:  $m_e = [u, v, 1]^T$  and  $M_e = [X, Y, Z, 1]^T$ . A camera is modeled by the usual pinhole: the relationship between a 3D point  $M$  and its image projection  $m$  is given by

$$s\tilde{m} = A \begin{bmatrix} R & t \end{bmatrix} \tilde{M}, \quad (1)$$

where  $s$  is an arbitrary scale factor,  $(R, t)$ , called the extrinsic parameters, is the rotation and translation which relates the world coordinate system to the camera coordinate system, and  $A$ , called the camera intrinsic matrix, is given by

$$A = \begin{bmatrix} \alpha & \gamma & u_0 \\ 0 & \beta & v_0 \\ 0 & 0 & 1 \end{bmatrix}$$

with  $(u_0, v_0)$  the coordinates of the principal point,  $\alpha$  and  $\beta$  the scale factors in image  $u$  and  $v$  axes, and  $\gamma$  the parameter describing the skewness of the two image axes.

Given an image of the model plane, an homography can be estimated. Let's denote it by  $H = [h_1 \ h_2 \ h_3]$ . From (2), we have

$$\begin{bmatrix} h_1 & h_2 & h_3 \end{bmatrix} = \lambda A \begin{bmatrix} r_1 & r_2 & t \end{bmatrix},$$

where  $\lambda$  is an arbitrary scalar. Using the knowledge that  $r_1$  and  $r_2$  are orthonormal, we have

$$h_1^T A^{-T} A^{-1} h_2 = 0 \quad (3)$$

$$h_1^T A^{-T} A^{-1} h_1 = h_2^T A^{-T} A^{-1} h_2. \quad (4)$$

These are the two basic constraints on the intrinsic parameters, given one homography. Because a homography has 8 degrees of freedom and there are 6 extrinsic parameters (3 for rotation and 3 for translation), we can only obtain 2 constraints on the intrinsic parameters.

### Camera Calibration:

We start with an analytical solution, followed by a nonlinear optimization technique based on the maximum likelihood criterion. Finally, we take into account lens distortion, giving both analytical and nonlinear solutions.

Let,

$$\mathbf{B} = \mathbf{A}^{-T} \mathbf{A}^{-1} \equiv \begin{bmatrix} B_{11} & B_{12} & B_{13} \\ B_{12} & B_{22} & B_{23} \\ B_{13} & B_{23} & B_{33} \end{bmatrix}$$

$$= \begin{bmatrix} \frac{1}{\alpha^2} & -\frac{\gamma}{\alpha^2\beta} & \frac{v_0\gamma - u_0\beta}{\alpha^2\beta} \\ -\frac{\gamma}{\alpha^2\beta} & \frac{\gamma^2}{\alpha^2\beta^2} + \frac{1}{\beta^2} & -\frac{\gamma(v_0\gamma - u_0\beta)}{\alpha^2\beta^2} - \frac{v_0}{\beta^2} \\ \frac{v_0\gamma - u_0\beta}{\alpha^2\beta} & -\frac{\gamma(v_0\gamma - u_0\beta)}{\alpha^2\beta^2} - \frac{v_0}{\beta^2} & \frac{(v_0\gamma - u_0\beta)^2}{\alpha^2\beta^2} + \frac{v_0^2}{\beta^2} + 1 \end{bmatrix}. \quad (5)$$

Here, B is symmetric, defined by a 6D vector.

$$\mathbf{b} = [B_{11}, B_{12}, B_{22}, B_{13}, B_{23}, B_{33}]^T. \quad (6)$$

Let the  $i$  th column vector of H be  $\mathbf{h}_i = [h_{i1}, h_{i2}, h_{i3}]^T$ . Then, we have

$$\mathbf{h}_i^T \mathbf{B} \mathbf{h}_j = \mathbf{v}_{ij}^T \mathbf{b} \quad (7)$$

$$\mathbf{v}_{ij} = [h_{i1}h_{j1}, h_{i1}h_{j2} + h_{i2}h_{j1}, h_{i2}h_{j2}, \\ h_{i3}h_{j1} + h_{i1}h_{j3}, h_{i3}h_{j2} + h_{i2}h_{j3}, h_{i3}h_{j3}]^T.$$

Therefore, the two fundamental constraints (3) and (4), from a given homography, can be rewritten as 2 homogeneous equations in  $\mathbf{b}$ :

$$\begin{bmatrix} \mathbf{v}_{12}^T \\ (\mathbf{v}_{11} - \mathbf{v}_{22})^T \end{bmatrix} \mathbf{b} = \mathbf{0} . \quad (8)$$

If  $n$  images of the model plane are observed, by stacking  $n$  such equations as (8) we have,

$$\mathbf{V}\mathbf{b} = \mathbf{0} \quad (9)$$

where  $\mathbf{V}$  is a  $2n \times 6$  matrix.

Once  $\mathbf{b}$  is estimated, we can compute all camera intrinsic matrix  $\mathbf{A}$ .

$\mathbf{B} = \lambda \mathbf{A}^{-T} \mathbf{A}$  with  $\lambda$  an arbitrary scale. Without difficulty, we can uniquely extract the intrinsic parameters from matrix  $\mathbf{B}$ .

$$\begin{aligned} v_0 &= (B_{12}B_{13} - B_{11}B_{23}) / (B_{11}B_{22} - B_{12}^2) \\ \lambda &= B_{33} - [B_{13}^2 + v_0(B_{12}B_{13} - B_{11}B_{23})] / B_{11} \\ \alpha &= \sqrt{\lambda / B_{11}} \\ \beta &= \sqrt{\lambda B_{11} / (B_{11}B_{22} - B_{12}^2)} \\ \gamma &= -B_{12}\alpha^2\beta / \lambda \\ u_0 &= \gamma v_0 / \beta - B_{13}\alpha^2 / \lambda . \end{aligned}$$

Once  $\mathbf{A}$  is known, the extrinsic parameters for each image is readily computed.

$$\mathbf{r}_1 = \lambda \mathbf{A}^{-1} \mathbf{h}_1$$

$$\mathbf{r}_2 = \lambda \mathbf{A}^{-1} \mathbf{h}_2$$

$$\mathbf{r}_3 = \mathbf{r}_1 \times \mathbf{r}_2$$

$$\mathbf{t} = \lambda \mathbf{A}^{-1} \mathbf{h}_3$$

with,

$$\lambda = 1/\|\mathbf{A}^{-1} \mathbf{h}_1\| = 1/\|\mathbf{A}^{-1} \mathbf{h}_2\|.$$

### Distortion Coefficients:

Let  $(u, v)$  be the ideal (non observable distortion-free) pixel image coordinates, and  $(\tilde{u}, \tilde{v})$  the corresponding real observed image coordinates. The ideal points are the projection of the model points according to the pinhole model. Similarly,  $(x, y)$  and  $(\tilde{x}, \tilde{y})$  are the ideal (distortion-free) and real (distorted) normalized image coordinates.

$$\begin{aligned} \tilde{x} &= x + x[k_1(x^2 + y^2) + k_2(x^2 + y^2)^2] \\ \tilde{y} &= y + y[k_1(x^2 + y^2) + k_2(x^2 + y^2)^2] , \end{aligned} \quad (10,11)$$

where  $k_1$  and  $k_2$  are the coefficients of the radial distortion. The center of the radial distortion is the same as the principal point.

Now,

$$\begin{bmatrix} (u-u_0)(x^2+y^2) & (u-u_0)(x^2+y^2)^2 \\ (v-v_0)(x^2+y^2) & (v-v_0)(x^2+y^2)^2 \end{bmatrix} \begin{bmatrix} k_1 \\ k_2 \end{bmatrix} = \begin{bmatrix} \tilde{u}-u \\ \tilde{v}-v \end{bmatrix} .$$

As ,  $\tilde{u} = u_0 + \alpha \tilde{x} + \gamma \tilde{y}$  and  $\tilde{v} = v_0 + \beta \tilde{x}$  and assuming  $\gamma = 0$ ,

Now , Given  $m$  points in  $n$  images, we can stack all equations together to obtain in total  $2mn$  equations, or in matrix form as  $D\mathbf{k} = \mathbf{d}$ , where  $\mathbf{k} = [k_1, k_2]^T$  . The linear least-squares solution is given by:

$$\mathbf{k} = (\mathbf{D}^T \mathbf{D})^{-1} \mathbf{D}^T \mathbf{d} . \quad (12)$$

## REFINING:

Now , refine the estimate of the other parameters by solving the equation

$$\sum_{i=1}^n \sum_{j=1}^m \|\mathbf{m}_{ij} - \check{\mathbf{m}}(\mathbf{A}, k_1, k_2, \mathbf{R}_i, \mathbf{t}_i, \mathbf{M}_j)\|^2 , \quad (13)$$

where  $\check{\mathbf{m}}(\mathbf{A}, k_1, k_2, \mathbf{R}_i, \mathbf{t}_i, \mathbf{M}_j)$  is the projection of point  $\mathbf{M}_j$  in image  $i$  according to equation (2), followed by distortion according to (11) and (12).

An initial guess of  $k_1$  and  $k_2$  can be obtained simply by setting them to 0.

Thus in summary ,

- Print a pattern and attach it to a planar surface.
- Take a few images of the model plane under different orientations by moving either the plane or the camera.
- Detect the feature points in the images.
- Estimate the five intrinsic parameters and all the extrinsic parameters using the closed-form solution.
- Estimate the coefficients of the radial distortion by solving the linear least-squares (12).
- Refine all parameters by minimizing (13)

Now , we need to set up a 3D coordinate system on a 2D image plane.We do this by :

- Defining an objp2 array of object points in 3D space
- Using the rotation and translational matrix i.e the extrinsic parameters.for the detected corners (corners2) in the undistorted image.
- Finally, we call the cv.projectPoints() function to project the 3D points onto the 2D image plane (imgpts2 and imgpts3).
- We then define a draw function, draw2() , that uses the detected corners and projected 3D points to draw a coordinate system and a cube in the image, respectively.
- Return: The undistorted image with the drawn coordinate system and cube is returned at the end.

**OUTPUT IMAGES LINK**— [output\\_assign3](#)