

CS-565 Computer Vision

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12. Camera Calibration

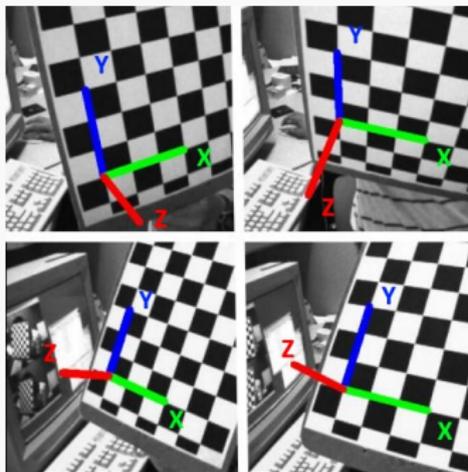
Factorization of the Camera Matrix

- ▶ Let K be the 3×3 product of the intrinsic and projection matrices after ignoring the last column of zeros.
- ▶ Extrinsic matrix performs a 3×3 rotation R followed by 3×1 translation \mathbf{t} .
- ▶ We can represent these operations as $[R|\mathbf{t}]$. So $[R|\mathbf{t}]M = RM + \mathbf{t}$.
- ▶ Then $P = K[R|\mathbf{t}]$.
- ▶ Camera center C is the null-vector of P .

$$PC = 0 \implies K[R|\mathbf{t}]C = 0 \implies KRC + K\mathbf{t} = 0 \implies C = -R^T\mathbf{t}$$

Camera Calibration Using 2D Checkerboard

- ▶ Camera calibration is the process of finding the 12 values of P .
- ▶ We can use a checkerboard of known size and structure.
- ▶ Fix one corner as origin of world coordinate system.
- ▶ Then all points on the checkerboard lie in the XY -plane ($Z = 0$).
- ▶ Z -axis is orthogonal to the checkerboard.



Camera Calibration Using 2D Checkerboard

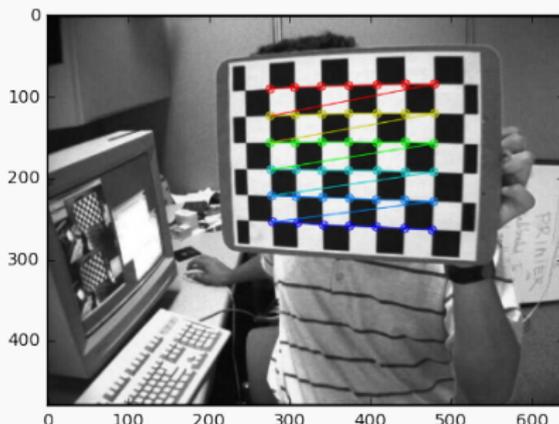
- ▶ For any point $\tilde{\mathbf{M}}$ on the checkerboard, its image $\tilde{\mathbf{m}}$ is obtained as

$$\begin{pmatrix} u \\ v \\ w \end{pmatrix} = (K \quad \mathbf{0}_{3 \times 1}) \begin{pmatrix} R & \mathbf{t} \\ \mathbf{0}_{1 \times 3} & 1 \end{pmatrix} \begin{pmatrix} X \\ Y \\ Z=0 \\ 1 \end{pmatrix} = \underbrace{K (\mathbf{r}_1 \quad \mathbf{r}_2 \quad \mathbf{t})}_{\text{Homography } H} \begin{pmatrix} X \\ Y \\ 1 \end{pmatrix}$$

- ▶ That is $\tilde{\mathbf{m}}_i = H\tilde{\mathbf{M}}_i$.
- ▶ Homography for checkerboard image j can be computed via DLT using at least 4 correspondences $\tilde{\mathbf{m}}_i \longleftrightarrow \tilde{\mathbf{M}}_i$.

Camera Calibration Using 2D Checkerboard

- ▶ The \mathbf{m}_i can be found using corner detection.



- ▶ The corresponding \mathbf{M}_i can be computed since the size and structure of the checkerboard are known (squares of measured dimensions).

Camera Calibration Using 2D Checkerboard

- ▶ For the estimated homography $H = [\mathbf{h}_1 \quad \mathbf{h}_2 \quad \mathbf{h}_3]$, we can write

$$[\mathbf{h}_1 \quad \mathbf{h}_2 \quad \mathbf{h}_3] = K [\mathbf{r}_1 \quad \mathbf{r}_2 \quad \mathbf{t}]$$

- ▶ So $\mathbf{r}_i \equiv K^{-1}\mathbf{h}_i$.
- ▶ Since \mathbf{r}_1 and \mathbf{r}_2 are columns of a rotation matrix

$$\mathbf{r}_1^T \mathbf{r}_2 = 0 \implies \mathbf{h}_1^T K^{-T} K^{-1} \mathbf{h}_2 = 0$$

$$\mathbf{r}_1^T \mathbf{r}_1 = 1 \implies \mathbf{h}_1^T K^{-T} K^{-1} \mathbf{h}_1 = 1$$

$$\mathbf{r}_2^T \mathbf{r}_2 = 1 \implies \mathbf{h}_2^T K^{-T} K^{-1} \mathbf{h}_2 = 1$$

- ▶ The term $K^{-T}K^{-1}$ is a symmetric and positive definite matrix that we denote by B .

$$B = \begin{pmatrix} b_{11} & b_{12} & b_{13} \\ b_{12} & b_{22} & b_{23} \\ b_{13} & b_{23} & b_{33} \end{pmatrix}$$

Camera Calibration Using 2D Checkerboard

- ▶ So for each homography (*i.e.*, checkerboard image), we have two constraints

$$\mathbf{h}_1^T B \mathbf{h}_2 = 0$$

$$\mathbf{h}_1^T B \mathbf{h}_1 - \mathbf{h}_2^T B \mathbf{h}_2 = 0$$

- ▶ For all images, these constraints can be written as a linear system $V\mathbf{b} = 0$ from which $\mathbf{b} = (b_{11} \ b_{12} \ b_{22} \ b_{13} \ b_{23} \ b_{33})^T$ can be found via SVD.
- ▶ Matrix K can be recovered through the Cholsky decomposition of B .

$$\text{chol}(B) = AA^T$$

which means that $K = A^{-T}$.

Camera Calibration Using 2D Checkerboard

- ▶ Once K is found, the extrinsic parameters can be found easily.

$$\mathbf{r}_1 \equiv K^{-1}\mathbf{h}_1$$

$$\mathbf{r}_2 \equiv K^{-1}\mathbf{h}_2$$

$$\mathbf{t} \equiv K^{-1}\mathbf{h}_3$$

and after normalising \mathbf{r}_1 and \mathbf{r}_2 , we can find the third column of R as $\mathbf{r}_3 \equiv \mathbf{r}_1 \times \mathbf{r}_2$.

- ▶ Notice that K is computed from B which is computed using the constraint $\mathbf{h}_1^T B \mathbf{h}_1 = \mathbf{h}_2^T B \mathbf{h}_2$.
- ▶ This means that \mathbf{r}_1 and \mathbf{r}_2 will have the same magnitude m which represents the homogeneous scale of matrix K .

Camera Calibration Using 2D Checkerboard

- ▶ So the actual rotation and translation vectors are

$$\mathbf{r}_1 \leftarrow \frac{1}{m} \mathbf{r}_1$$

$$\mathbf{r}_2 \leftarrow \frac{1}{m} \mathbf{r}_2$$

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

$$\mathbf{t} \leftarrow \frac{1}{m} \mathbf{t}$$