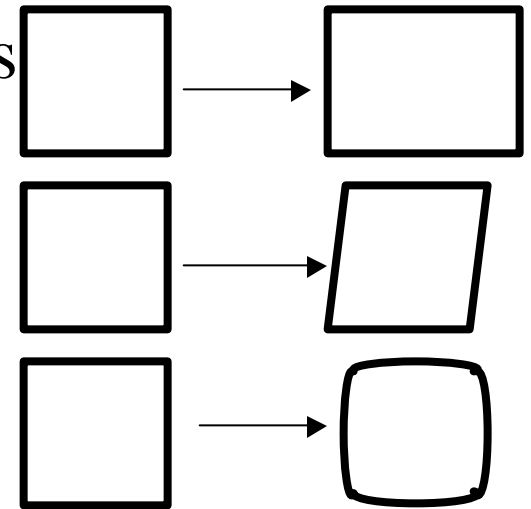


Camera Calibration

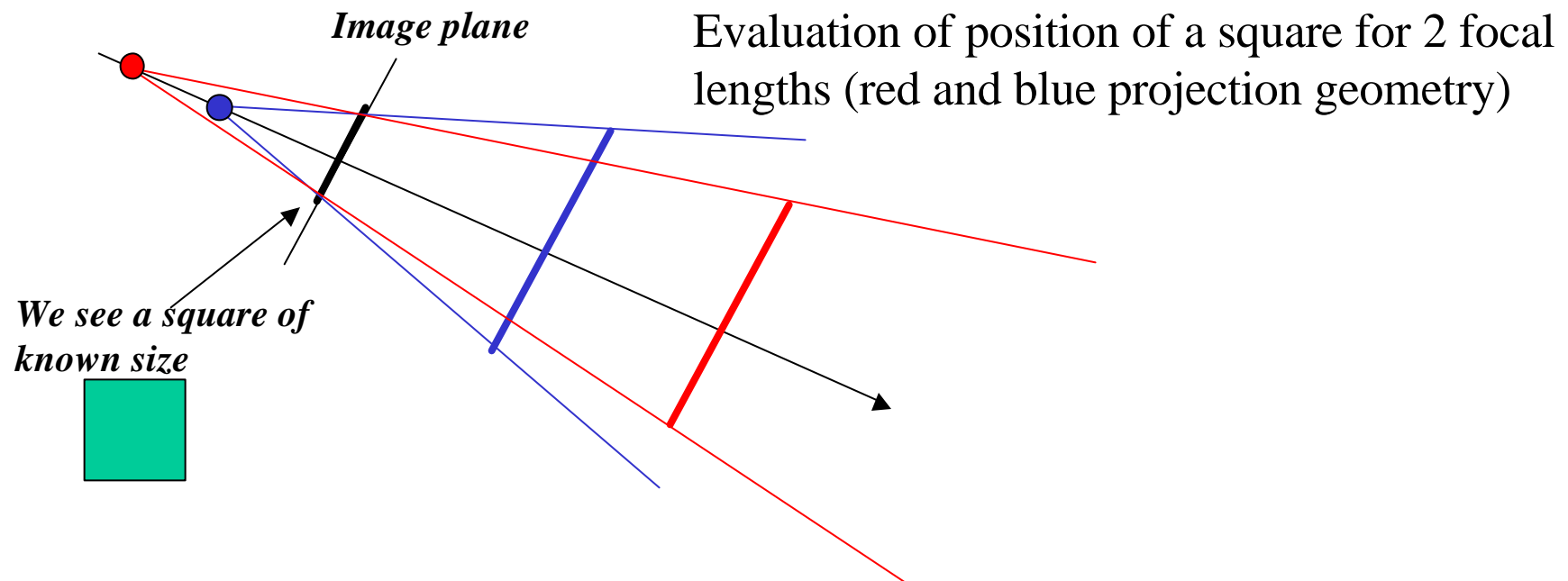
What is Camera Calibration?

- Primarily, finding the quantities internal to the camera that affect the imaging process
 - Position of image center in the image
 - It is typically not at $(\text{width}/2, \text{height}/2)$ of image
 - Focal length
 - Different scaling factors for row pixels and column pixels
 - Skew factor
 - Lens distortion (pin-cushion effect)



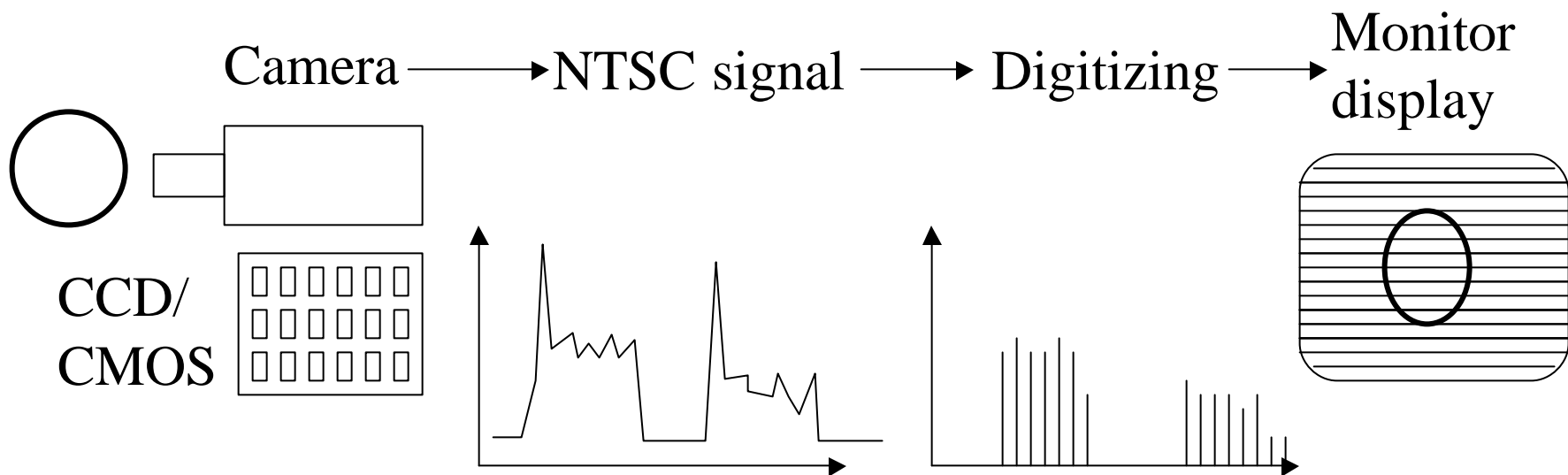
Motivation

- Good calibration is important when we need to
 - Reconstruct a world model: Virtual L.A. project
 - Interact with the world
 - Robot, hand-eye coordination



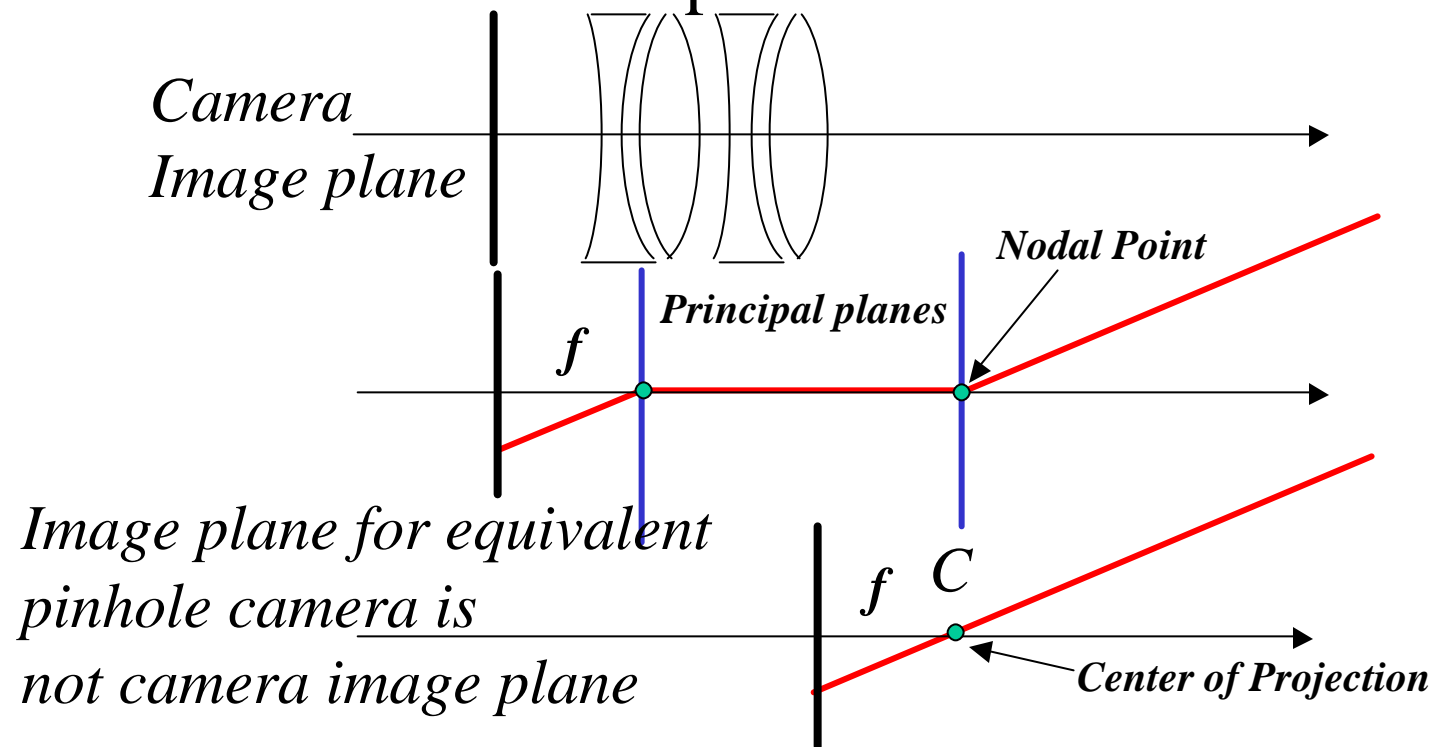
Scaling of Rows and Columns in Image

- Camera pixels are not necessarily square
- Camera output may be analog (NTSC)
- Image may be obtained by digitizing card
 - A/D converter samples NTSC signal



Compound Lens Imaging

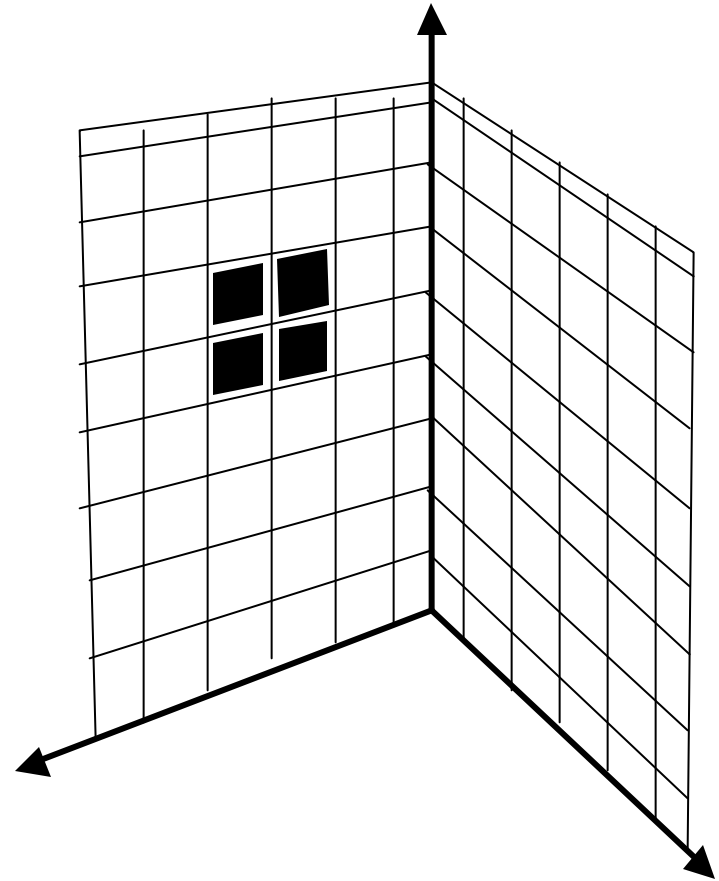
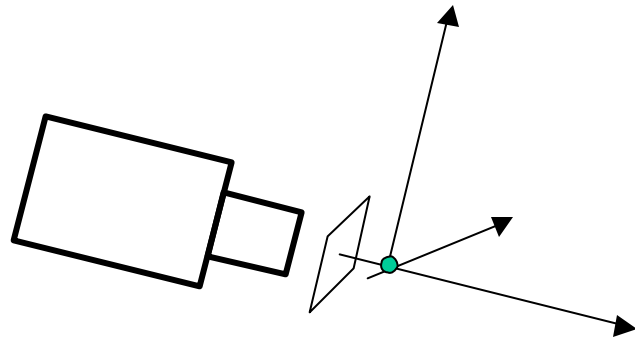
- Inexpensive single lens systems distort image at its periphery
- Compound lenses may be used to reduce chromatic effects and pin-cushion effects



Variety of Techniques

- VERY large literature on the subject
- Work of Roger Tsai influential
- Linear algebra method described here
 - Can be used as initialization for iterative non linear methods.
- Some interesting methods use vanishing points

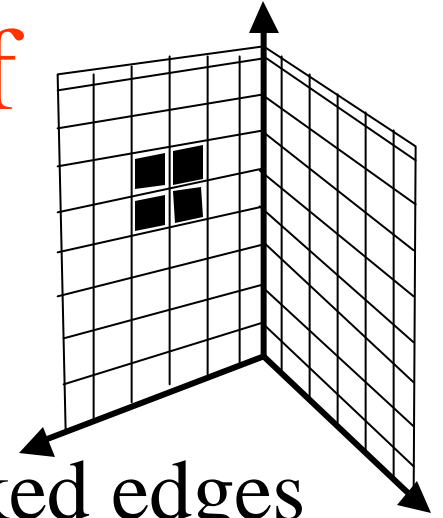
Camera and Calibration Target



Calibration Procedure

- Calibration target: 2 planes at right angle with checkerboard patterns (Tsai grid)
 - We know positions of pattern corners only with respect to a coordinate system of the target
 - We position camera in front of target and find images of corners
 - We obtain equations that describe imaging and contain internal parameters of camera
 - As a side benefit, we find position and orientation of camera with respect to target (camera *pose*)

Image Processing of Image of Target



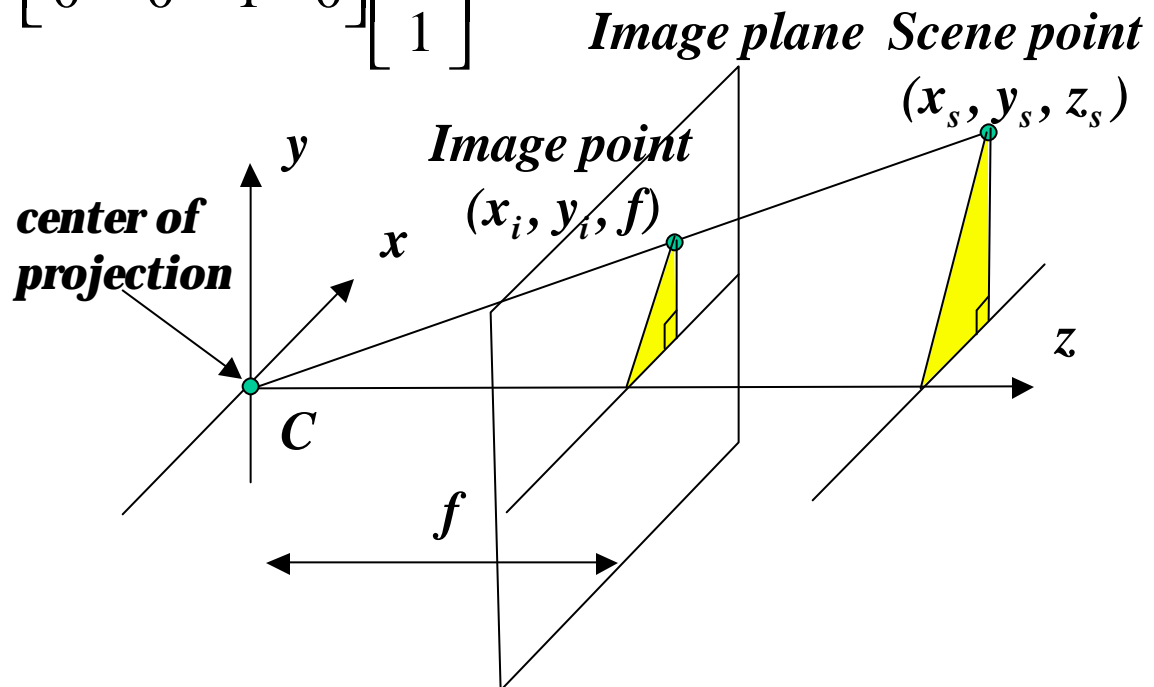
- Canny edge detection
- Straight line fitting to detected linked edges
- Intersecting the lines to obtain the image corners
- Matching image corners and 3D target checkerboard corners
 - By counting if whole target is visible in image
- We get pairs (image point)--(world point)

$$(x_i, y_i) \rightarrow (X_i, Y_i, Z_i)$$

Central Projection

If world and image points are represented by homogeneous vectors, central projection is a linear transformation:

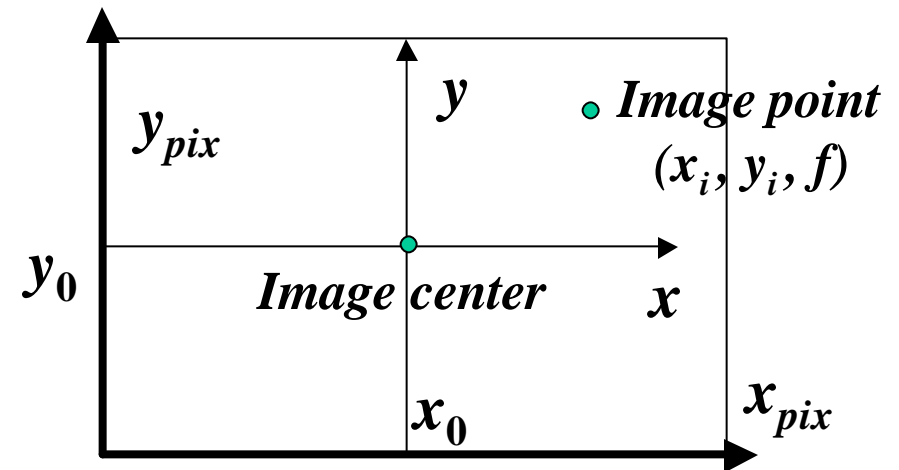
$$x_i = f \frac{x_s}{z_s}$$
$$y_i = f \frac{y_s}{z_s}$$
$$\begin{bmatrix} u \\ v \\ w \end{bmatrix} = \begin{bmatrix} f & 0 & 0 & 0 \\ 0 & f & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_s \\ y_s \\ z_s \\ 1 \end{bmatrix}$$
$$x_i = u / w, \quad y_i = v / w$$



Transformation From Lengths to Pixels

Transformation uses:

- image center (x_0, y_0)
- scaling factors k_x and k_y



$$x_i = f \frac{x_s}{z_s}$$

$$x_{pix} = k_x x_i + x_0 = f k_x \frac{x_s + z_s x_0}{z_s}$$

$$y_i = f \frac{y_s}{z_s}$$

$$y_{pix} = k_y y_i + y_0 = f k_y \frac{y_s + z_s y_0}{z_s}$$

$$\begin{bmatrix} u' \\ v' \\ w' \end{bmatrix} = \begin{bmatrix} \mathbf{a}_x & 0 & x_0 & 0 \\ 0 & \mathbf{a}_y & y_0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_s \\ y_s \\ z_s \\ 1 \end{bmatrix} \quad \text{with}$$

$$\mathbf{a}_x = f k_x$$

$$\mathbf{a}_y = f k_y$$

then

$$x_{pix} = u' / w'$$

$$y_{pix} = v' / w'$$

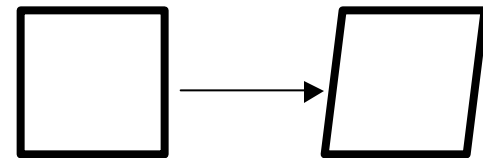
Internal Camera Parameters

$$\begin{bmatrix} u' \\ v' \\ w' \end{bmatrix} = \begin{bmatrix} \mathbf{a}_x & s & x_0 & 0 \\ 0 & \mathbf{a}_y & y_0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x_s \\ y_s \\ z_s \\ 1 \end{bmatrix} \quad \text{with} \quad \begin{aligned} \mathbf{a}_x &= f k_x & x_{pix} &= u' / w' \\ \mathbf{a}_y &= -f k_y & y_{pix} &= v' / w' \end{aligned}$$

$$\begin{bmatrix} \mathbf{a}_x & s & x_0 & 0 \\ 0 & \mathbf{a}_y & y_0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} = \begin{bmatrix} \mathbf{a}_x & s & x_0 \\ 0 & \mathbf{a}_y & y_0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} = \mathbf{K} [\mathbf{I}_3 \mid \mathbf{0}_3]$$

- \mathbf{a}_x and \mathbf{a}_y “focal lengths” in pixels
- x_0 and y_0 coordinates of image center in pixels

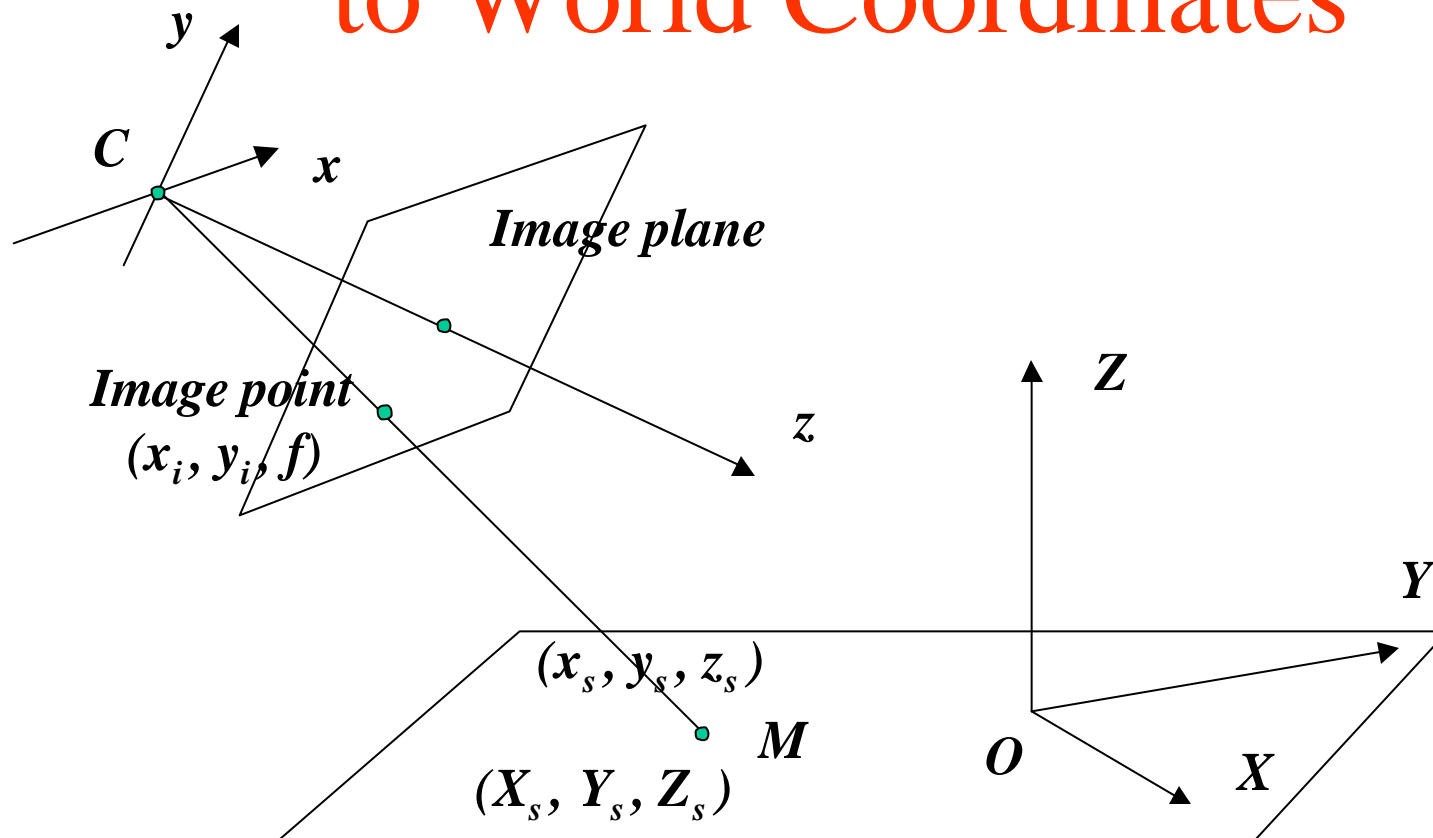
- Added parameter S is skew parameter



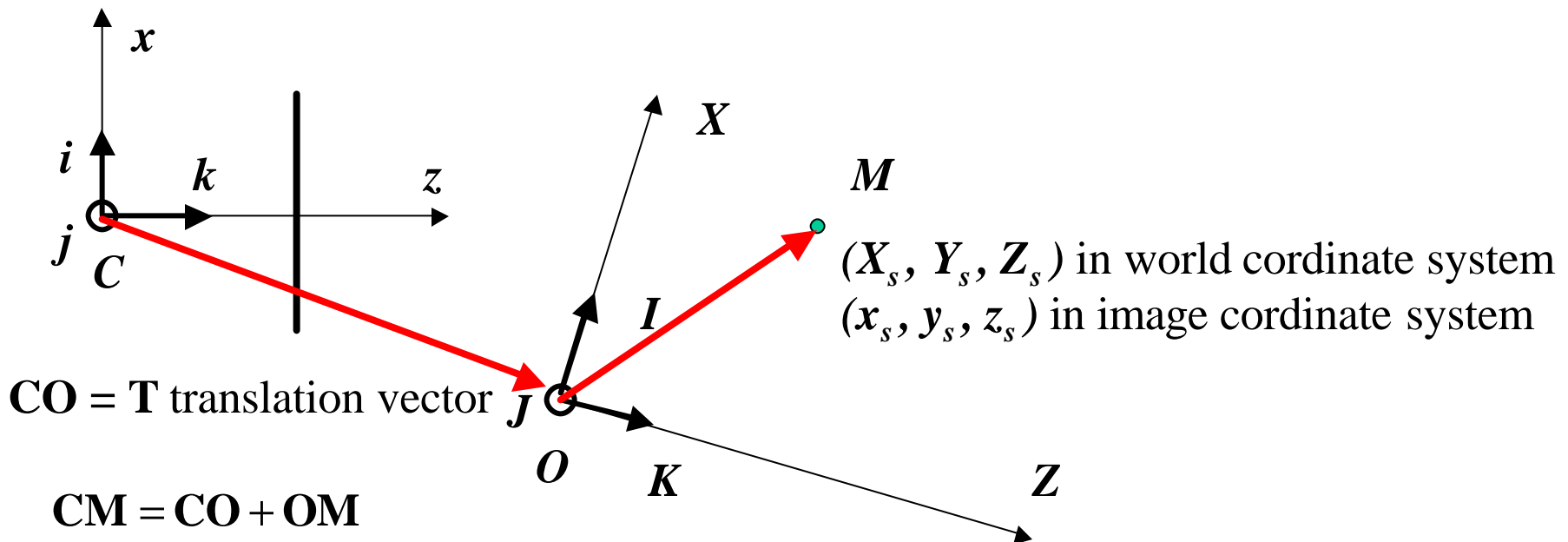
- \mathbf{K} is called *calibration matrix*. **Five degrees of freedom.**

- \mathbf{K} is a 3x3 upper triangular matrix

From Camera Coordinates to World Coordinates



From Camera Coordinates to World Coordinates 2



$$\mathbf{CM} = \mathbf{CO} + \mathbf{OM}$$

$$x_s \mathbf{i} + y_s \mathbf{j} + z_s \mathbf{k} = T_x \mathbf{i} + T_y \mathbf{j} + T_z \mathbf{k} + X_s \mathbf{I} + Y_s \mathbf{J} + Z_s \mathbf{K}$$

$$x_s = T_x + X_s \mathbf{I} \cdot \mathbf{i} + Y_s \mathbf{J} \cdot \mathbf{i} + Z_s \mathbf{K} \cdot \mathbf{i}$$

$$\begin{bmatrix} x_s \\ y_s \\ z_s \end{bmatrix} = \begin{bmatrix} T_x \\ T_y \\ T_z \end{bmatrix} + \begin{bmatrix} \mathbf{I} \cdot \mathbf{i} & \mathbf{J} \cdot \mathbf{i} & \mathbf{K} \cdot \mathbf{i} \\ \mathbf{I} \cdot \mathbf{j} & \mathbf{J} \cdot \mathbf{j} & \mathbf{K} \cdot \mathbf{j} \\ \mathbf{I} \cdot \mathbf{k} & \mathbf{J} \cdot \mathbf{k} & \mathbf{K} \cdot \mathbf{k} \end{bmatrix} \begin{bmatrix} X_s \\ Y_s \\ Z_s \end{bmatrix}$$

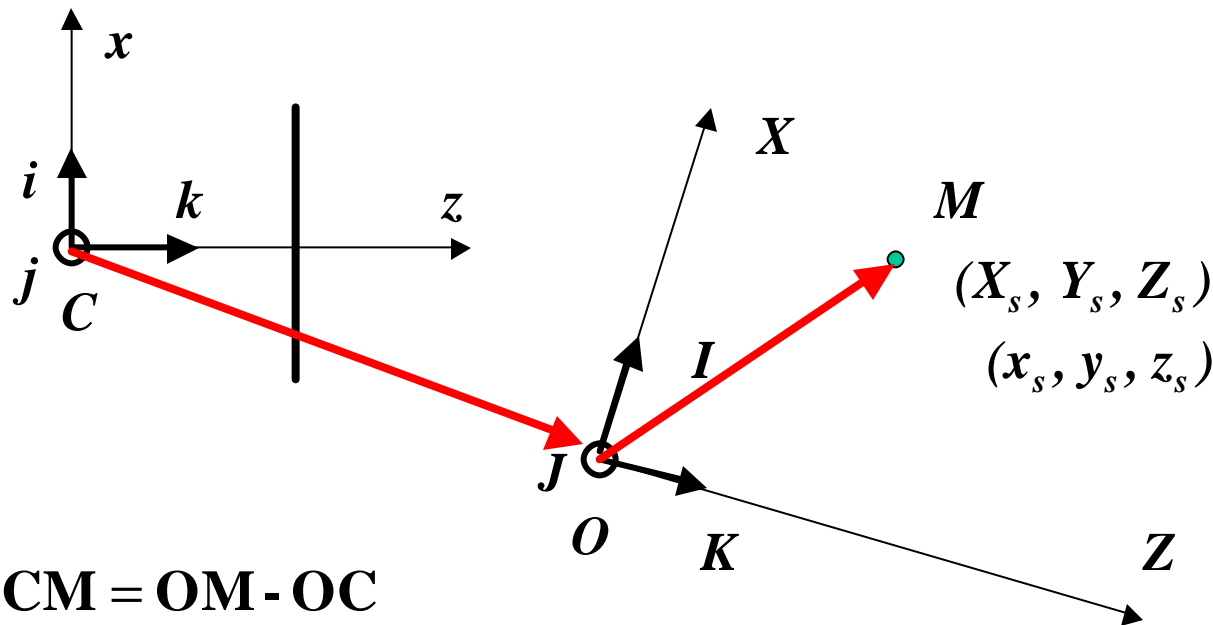
Homogeneous Coordinates

$$\begin{bmatrix} x_S \\ y_S \\ z_S \end{bmatrix} = \begin{bmatrix} T_x \\ T_y \\ T_z \end{bmatrix} + \begin{bmatrix} \mathbf{I.i} & \mathbf{J.i} & \mathbf{K.i} \\ \mathbf{I.j} & \mathbf{J.j} & \mathbf{K.j} \\ \mathbf{I.k} & \mathbf{J.k} & \mathbf{K.k} \end{bmatrix} \begin{bmatrix} X_S \\ Y_S \\ Z_S \end{bmatrix}$$

$$\begin{bmatrix} x_S \\ y_S \\ z_S \\ 1 \end{bmatrix} = \begin{bmatrix} \mathbf{I.i} & \mathbf{J.i} & \mathbf{K.i} & T_x \\ \mathbf{I.j} & \mathbf{J.j} & \mathbf{K.j} & T_y \\ \mathbf{I.k} & \mathbf{J.k} & \mathbf{K.k} & T_z \\ \mathbf{0} & \mathbf{0} & \mathbf{0} & 1 \end{bmatrix} \begin{bmatrix} X_S \\ Y_S \\ Z_S \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} x_S \\ y_S \\ z_S \\ 1 \end{bmatrix} = \begin{bmatrix} \mathbf{R} & \mathbf{T} \\ \mathbf{0}_3^T & 1 \end{bmatrix} \begin{bmatrix} X_S \\ Y_S \\ Z_S \\ 1 \end{bmatrix}$$

From Camera Coordinates to World Coordinates 3



$$\mathbf{CM} = \mathbf{OM} - \mathbf{OC}$$

$$x_s \mathbf{i} + y_s \mathbf{j} + z_s \mathbf{k} = (X_s - X_c) \mathbf{I} + (Y_s - Y_c) \mathbf{J} + (Z_s - Z_c) \mathbf{K}$$

$$x_s = (X_s - X_c) \mathbf{I} \cdot \mathbf{i} + (Y_s - Y_c) \mathbf{J} \cdot \mathbf{i} + (Z_s - Z_c) \mathbf{K} \cdot \mathbf{i}$$

$$\mathbf{x}_{\text{cam}} = \mathbf{R}(\mathbf{X} - \tilde{\mathbf{C}}) \quad (\mathbf{T} = -\mathbf{R}\tilde{\mathbf{C}})$$

$\tilde{\mathbf{C}}$ is vector \mathbf{OC} expressed in world coordinate system

Homogeneous Coordinates 2

- Here we use $-\mathbf{R}\tilde{\mathbf{C}}$ instead of \mathbf{T}

$$\begin{bmatrix} x_s \\ y_s \\ z_s \\ 1 \end{bmatrix} = \begin{bmatrix} \mathbf{R} & -\mathbf{R}\tilde{\mathbf{C}} \\ \mathbf{0}_3^T & 1 \end{bmatrix} \begin{bmatrix} X_s \\ Y_s \\ Z_s \\ 1 \end{bmatrix}$$

Linear Transformation from World Coordinates to Pixels

- Combine camera projection and coordinate transformation matrices into a single matrix \mathbf{P}

$$\begin{aligned}
 \begin{bmatrix} u' \\ v' \\ w' \end{bmatrix} &= \mathbf{K} \begin{bmatrix} \mathbf{I}_3 & | & \mathbf{0}_3 \end{bmatrix} \begin{bmatrix} x_s \\ y_s \\ z_s \\ 1 \end{bmatrix} \\
 \begin{bmatrix} x_s \\ y_s \\ z_s \\ 1 \end{bmatrix} &= \begin{bmatrix} \mathbf{R} & -\mathbf{R}\tilde{\mathbf{C}} \\ \mathbf{0}_3^T & 1 \end{bmatrix} \begin{bmatrix} X_s \\ Y_s \\ Z_s \\ 1 \end{bmatrix} \\
 \Rightarrow \begin{bmatrix} u' \\ v' \\ w' \end{bmatrix} &= \mathbf{K} \begin{bmatrix} \mathbf{I}_3 & | & \mathbf{0}_3 \end{bmatrix} \begin{bmatrix} \mathbf{R} & -\mathbf{R}\tilde{\mathbf{C}} \\ \mathbf{0}_3^T & 1 \end{bmatrix} \begin{bmatrix} X_s \\ Y_s \\ Z_s \\ 1 \end{bmatrix} \\
 \Rightarrow \begin{bmatrix} u' \\ v' \\ w' \end{bmatrix} &= \mathbf{P} \begin{bmatrix} X_s \\ Y_s \\ Z_s \\ 1 \end{bmatrix} \qquad \mathbf{x} = \mathbf{P}\mathbf{X}
 \end{aligned}$$

Properties of Matrix \mathbf{P}

- Further simplification of \mathbf{P} :

$$\mathbf{x} = \mathbf{K} \left[\mathbf{I}_3 \quad | \quad \mathbf{0}_3 \right] \begin{bmatrix} \mathbf{R} & -\mathbf{R}\tilde{\mathbf{C}} \\ \mathbf{0}_3^T & 1 \end{bmatrix} \mathbf{X}$$

$$\left[\mathbf{I}_3 \quad | \quad \mathbf{0}_3 \right] \begin{bmatrix} \mathbf{R} & -\mathbf{R}\tilde{\mathbf{C}} \\ \mathbf{0}_3^T & 1 \end{bmatrix} = \left[\mathbf{R} \quad -\mathbf{R}\tilde{\mathbf{C}} \right] = \mathbf{R} \left[\mathbf{I}_3 \quad | \quad -\tilde{\mathbf{C}} \right]$$

$$\mathbf{x} = \mathbf{K} \mathbf{R} \left[\mathbf{I}_3 \quad | \quad -\tilde{\mathbf{C}} \right] \mathbf{X}$$

$$\mathbf{P} = \mathbf{K} \mathbf{R} \left[\mathbf{I}_3 \quad | \quad -\tilde{\mathbf{C}} \right]$$

- \mathbf{P} has 11 degrees of freedom:
 - 5 from triangular calibration matrix \mathbf{K} , 3 from \mathbf{R} and 3 from $\tilde{\mathbf{C}}$
- \mathbf{P} is a fairly general 3 x 4 matrix
 - left 3x3 submatrix $\mathbf{K}\mathbf{R}$ is non-singular

Calibration

- 1. Estimate matrix \mathbf{P} using scene points and their images
- 2. Estimate the interior parameters and the exterior parameters

$$\mathbf{P} = \mathbf{K} \mathbf{R} \left[\mathbf{I}_3 \quad | \quad -\tilde{\mathbf{C}} \right]$$

- Left 3x3 submatrix of \mathbf{P} is product of upper-triangular matrix and orthogonal matrix

Finding Camera Translation

- Find homogeneous coordinates of C in the scene
- C is the null vector of matrix \mathbf{P}

- $\mathbf{P} \mathbf{C} = 0$:

$$\mathbf{P} = \mathbf{K} \mathbf{R} \left[\mathbf{I}_3 \quad | \quad -\tilde{\mathbf{C}} \right]$$

$$\begin{bmatrix} 1 & 0 & 0 & X_c \\ 0 & 1 & 0 & Y_c \\ 0 & 0 & 1 & Z_c \end{bmatrix} \begin{bmatrix} X_c \\ Y_c \\ Z_c \\ 1 \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

- Find null vector \mathbf{C} of \mathbf{P} using SVD
 - \mathbf{C} is the unit singular vector of \mathbf{P} corresponding to the smallest singular value (the last column of \mathbf{V} , where $\mathbf{P} = \mathbf{U} \mathbf{D} \mathbf{V}^T$ is the SVD of \mathbf{P})

Finding Camera Orientation and Internal Parameters

- Left 3x3 submatrix \mathbf{M} of \mathbf{P} is of form $\mathbf{M}=\mathbf{K} \mathbf{R}$
 - \mathbf{K} is an upper triangular matrix
 - \mathbf{R} is an orthogonal matrix
- Any non-singular square matrix \mathbf{M} can be decomposed into the product of an upper-triangular matrix \mathbf{K} and an orthogonal matrix \mathbf{R} using the RQ factorization
 - Similar to QR factorization but order of 2 matrices is reversed

RQ Factorization of M

$$\mathbf{R}_x = \begin{bmatrix} 1 & 0 & 0 \\ 0 & c & -s \\ 0 & s & c \end{bmatrix}, \quad \mathbf{R}_y = \begin{bmatrix} c' & 0 & s' \\ 0 & 1 & 0 \\ -s' & 0 & c' \end{bmatrix}, \quad \mathbf{R}_z = \begin{bmatrix} c'' & -s'' & 0 \\ s'' & c'' & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

- Compute $c = -\frac{m_{33}}{(m_{32}^2 + m_{33}^2)^{1/2}}$, $s = \frac{m_{32}}{(m_{32}^2 + m_{33}^2)^{1/2}}$
- Multiply \mathbf{M} by \mathbf{R}_x . The resulting term at (3,2) is zero because of the values selected for c and s
- Multiply the resulting matrix by \mathbf{R}_y , after selecting c' and s' so that the resulting term at position (3,1) is set to zero
- Multiply the resulting matrix by \mathbf{R}_z , after selecting c'' and s'' so that the resulting term at position (2,1) is set to zero

$$\mathbf{M} \mathbf{R}_x \mathbf{R}_y \mathbf{R}_z = \mathbf{K} \Rightarrow \mathbf{M} = \mathbf{K} \mathbf{R}_z^T \mathbf{R}_y^T \mathbf{R}_x^T = \mathbf{K} \mathbf{R}$$

Computing Matrix \mathbf{P}

- Use corresponding image and scene points
 - 3D points \mathbf{X}_i in world coordinate system
 - Images \mathbf{x}_i of \mathbf{X}_i in image
- Write $\mathbf{x}_i = \mathbf{P} \mathbf{X}_i$ for all i
- Similar problem to finding projectivity matrix \mathbf{H} (i.e. homography) in homework

Improved Computation of P

- $\mathbf{x}_i = \mathbf{P} \mathbf{X}_i$ involves homogeneous coordinates, thus \mathbf{x}_i and $\mathbf{P} \mathbf{X}_i$ just have to be proportional: $\mathbf{x}_i \times \mathbf{P} \mathbf{X}_i = 0$
- Let $\mathbf{p}_1^T, \mathbf{p}_2^T, \mathbf{p}_3^T$ be the 3 row vectors of \mathbf{P}

$$\mathbf{P} \mathbf{X}_i = \begin{bmatrix} \mathbf{p}_1^T \mathbf{X}_i \\ \mathbf{p}_2^T \mathbf{X}_i \\ \mathbf{p}_3^T \mathbf{X}_i \end{bmatrix} \quad \mathbf{x}_i \times \mathbf{P} \mathbf{X}_i = \begin{bmatrix} v'_i \mathbf{p}_3^T \mathbf{X}_i - w'_i \mathbf{p}_2^T \mathbf{X}_i \\ w'_i \mathbf{p}_1^T \mathbf{X}_i - u'_i \mathbf{p}_3^T \mathbf{X}_i \\ u'_i \mathbf{p}_2^T \mathbf{X}_i - v'_i \mathbf{p}_1^T \mathbf{X}_i \end{bmatrix}$$

$$\Rightarrow \begin{bmatrix} \mathbf{0}_4^T & -w'_i \mathbf{X}_i^T & v'_i \mathbf{X}_i^T \\ w'_i \mathbf{X}_i^T & \mathbf{0}_4^T & -u'_i \mathbf{X}_i^T \\ -v'_i \mathbf{X}_i^T & u'_i \mathbf{X}_i^T & \mathbf{0}_4^T \end{bmatrix} \begin{bmatrix} \mathbf{p}_1 \\ \mathbf{p}_2 \\ \mathbf{p}_3 \end{bmatrix} = 0 \quad \begin{bmatrix} \mathbf{p}_1 \\ \mathbf{p}_2 \\ \mathbf{p}_3 \end{bmatrix} \text{ is a } 12 \times 1 \text{ vector}$$

Improved Computation of P, cont'd

- Third row can be obtained from sum of u'_i times first row - v'_i times second row

$$\begin{bmatrix} \mathbf{0}_4^T & -w'_i \mathbf{X}_i^T & v'_i \mathbf{X}_i^T \\ w'_i \mathbf{X}_i^T & \mathbf{0}_4^T & -u'_i \mathbf{X}_i^T \\ -v'_i \mathbf{X}_i^T & u'_i \mathbf{X}_i^T & \mathbf{0}_4^T \end{bmatrix} \begin{bmatrix} \mathbf{p}_1 \\ \mathbf{p}_2 \\ \mathbf{p}_3 \end{bmatrix} = 0$$

- So we get 2 independent equations in 11 unknowns (ignoring scale)
- With 6 point correspondences, we get enough equations to compute matrix \mathbf{P}

$$\mathbf{A} \mathbf{p} = 0$$

Solving $\mathbf{A} \mathbf{p} = 0$

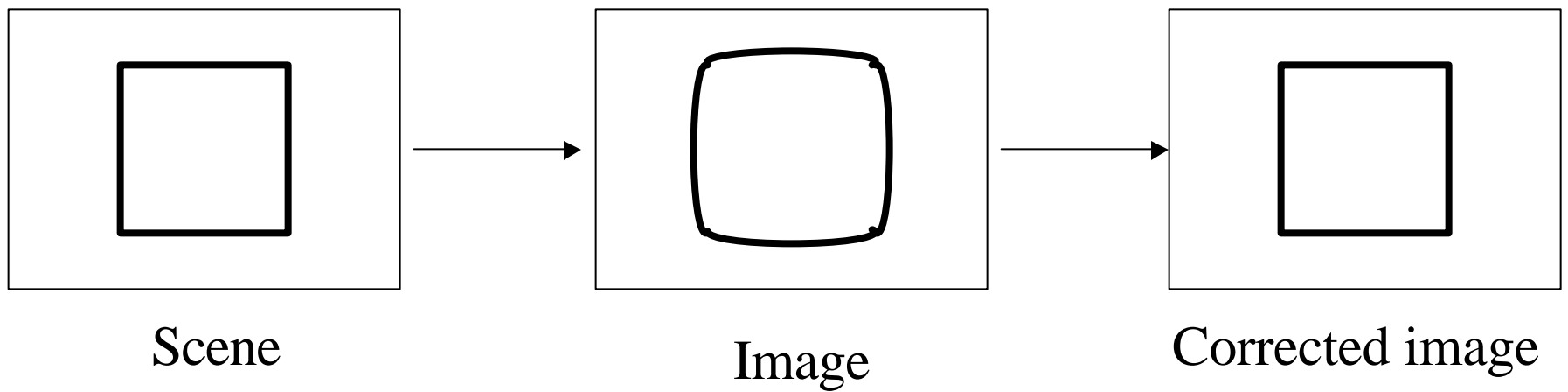
- Linear system $\mathbf{A} \mathbf{p} = 0$
- When possible, have at least 5 times as many equations as unknowns (28 points)
- Minimize $\| \mathbf{A} \mathbf{p} \|$ with the constraint $\| \mathbf{p} \| = 1$
 - \mathbf{p} is the unit singular vector of \mathbf{A} corresponding to the smallest singular value (the last column of \mathbf{V} , where $\mathbf{A} = \mathbf{U} \mathbf{D} \mathbf{V}^T$ is the SVD of \mathbf{A})
- Called Direct Linear Transformation (DLT)

Improving \mathbf{P} Solution with Nonlinear Minimization

- Find \mathbf{p} using DLT
- Use as initialization for nonlinear minimization of $\sum_i d(\mathbf{x}_i, \mathbf{P}\mathbf{X}_i)^2$
 - Use Levenberg-Marquardt iterative minimization

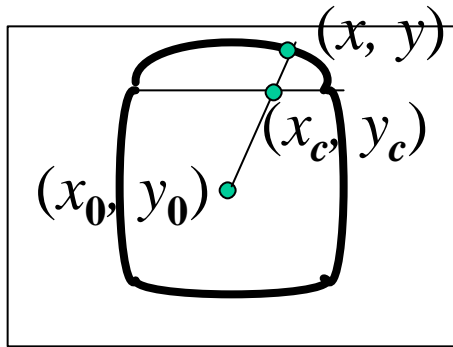
Radial Distortion

- We have assumed that lines are imaged as lines
- Not quite true for real lenses
 - Significant error for cheap optics and for short focal lengths



Radial Distortion Modeling

- In pixel coordinates the correction is written



$$x_c - x_0 = L(r)(x - x_0)$$

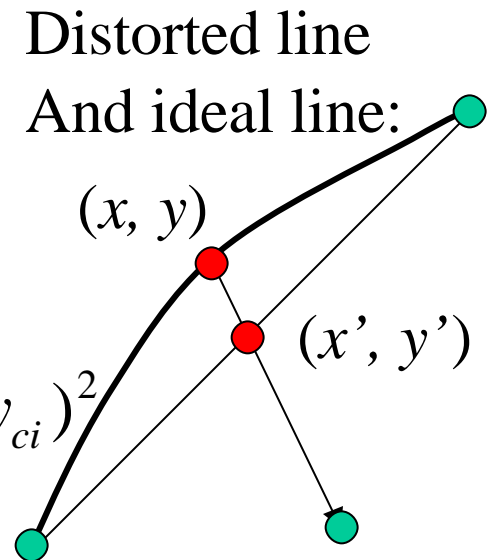
$$y_c - y_0 = L(r)(y - y_0)$$

with

$$r^2 = (x - x_0)^2 + (y - y_0)^2$$

$$L(r) = 1 + \mathbf{k}_1 r + \mathbf{k}_2 r^2 + \dots$$

- Minimize $f(\mathbf{k}_1, \mathbf{k}_2) = \sum (x'_i - x_{ci})^2 + (y'_i - y_{ci})^2$
using lines known toⁱ be straight
(x', y') is radial projection of (x, y) on straight line



References

- Multiple View Geometry in Computer Vision, R. Hartley and A. Zisserman, Cambridge University Press, 2000, pp. 138-183
- Three-Dimensional Computer Vision: A Geometric Approach, O. Faugeras, MIT Press, 1996, pp. 33-68
- “A Versatile Camera Calibration Technique for 3D Machine Vision”, R. Y. Tsai, IEEE J. Robotics & Automation, RA-3, No. 4, August 1987, pp. 323-344