

ME6406 Machine Vision

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Part 3A Camera model and calibration

<http://kmlee.gatech.edu/me6406>

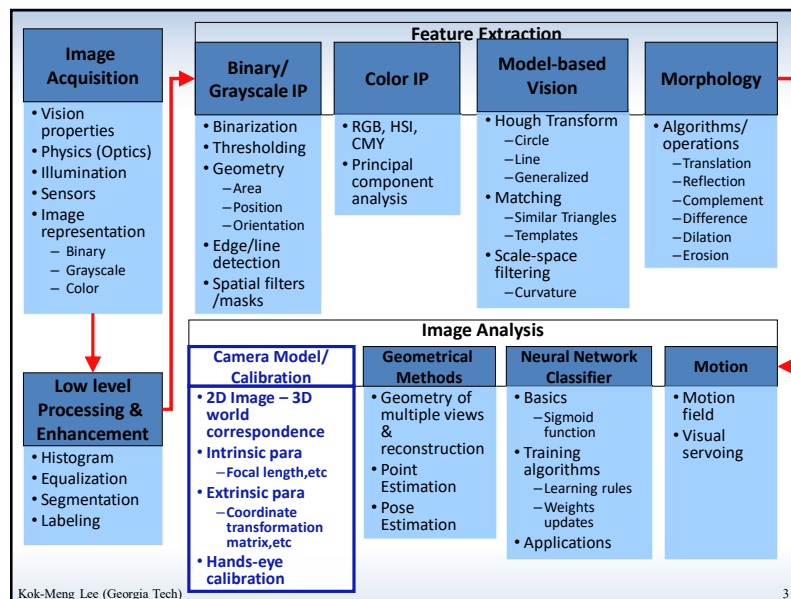
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Course Outline

- Introduction and low-level processing
 - Physics of digital images, histogram equalization, segmentation, edge detection, linear filters.
- Model-based Vision
 - Hough transform, pattern representation, matching
- **Geometric methods**
 - **Camera model, calibration, pose estimation**
- Neural network for machine vision
 - Basics, training algorithms, and applications
- Color images and selected topics
 - Physics, perception, processing and applications

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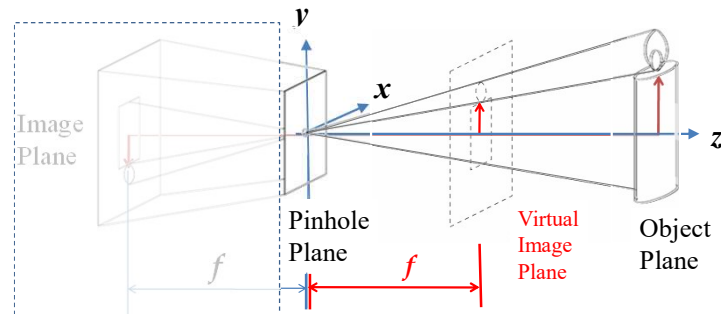
Reading Materials:

5. Tsai, R. "A Versatile Camera Calibration Technique for High-accuracy 3D Machine Vision Metrology using Off-the-shelf TV Cameras and Lenses," IEEE Trans. on Robotics and Automation, Vol. 3, No.4, August 1987, pp: 323- 344

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Pinhole Camera Model



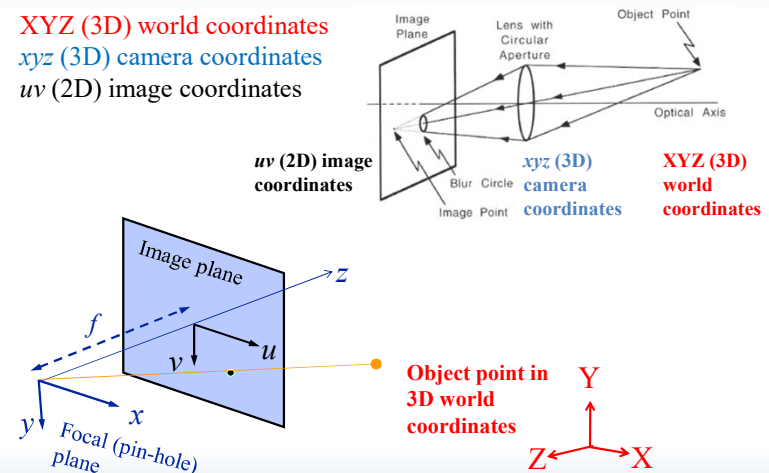
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Camera Model

XYZ (3D) world coordinates
xyz (3D) camera coordinates
uv (2D) image coordinates



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Reasons for Camera Calibration

- ✦ Need to recover **3D quantitative measures** about the observed scene from **2D images**.
- ✦ **Model and predict** the performance or accuracy of any machine vision algorithms
- ✦ Determine the **camera location** relative to the calibration board (or the working plane)
- ✦ **Basis for other calibration**; robot kinematics, hand-eye relationship, geometric calibrations.

Definition

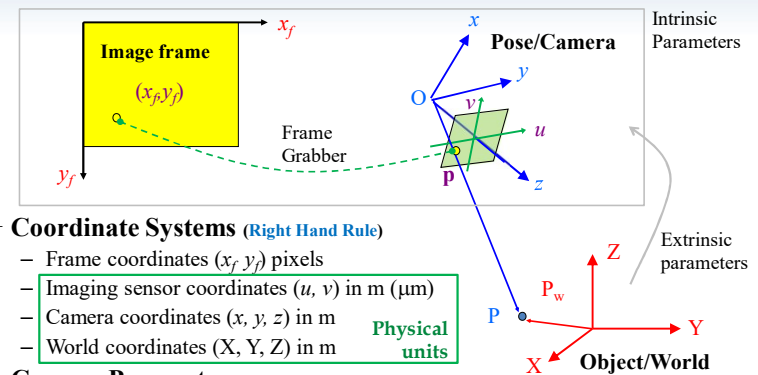
- ✦ The problem of determining the elements that govern the relationship or transformation between the 2D image that a camera sees and the 3D of the observed scene.
- ✦ Two kinds of parameters defining this 2D/3D relationship:
 - Intrinsic
 - Extrinsic

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Camera Model



Coordinate Systems (Right Hand Rule)

- Frame coordinates (x_f, y_f) pixels
- Imaging sensor coordinates (u, v) in m (μm)
- Camera coordinates (x, y, z) in m
- World coordinates (X, Y, Z) in m

Camera Parameters

- Intrinsic Parameters (of the camera and the frame grabber): link the **frame coordinates** of an image point with its corresponding **camera coordinates**.
- Extrinsic parameters: define the relative location and orientation of the **camera coordinate system** with the **world coordinate system**.

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❑ Intrinsic parameters

⊕ Parameters that characterize the inherent geometric properties of the camera and optics:

- Image center
- Image X and Y scale factors
- Lens principal distance (effective **focus length**)
- **Lens distortion coefficients**

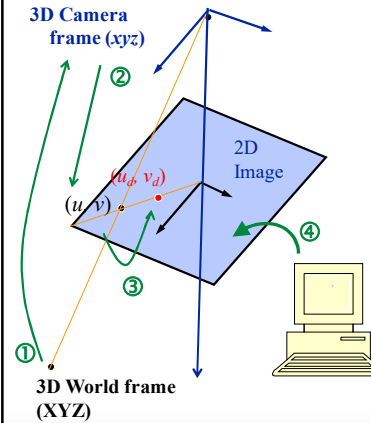
❑ Extrinsic parameters

⊕ Parameters that indicates the position and orientation of the camera with respect to the world coordinate system:

- Translation (T_x , T_y and T_z)
- Rotation about X, Y and Z axes.

Tsai Camera Model

(Radial lens distortion assumption;)



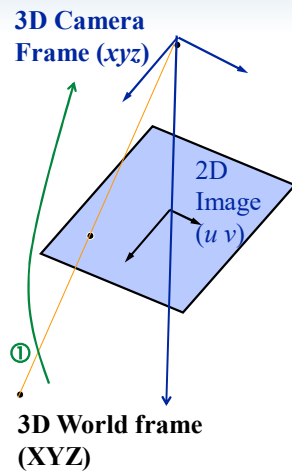
Step 1
Rigid body transformation from
world frame to camera frame
Parameters to be calibrated: $[R], T$

Step 2
Perspective projection with pin hole
geometry
Parameters to be calibrated: f

Step 3
Radial lens distortion correction
Parameters to be calibrated: k_1

Step 4
Relation between physical and
computer coordinates
Parameters to be calibrated:
scaling

Tsai Camera Model

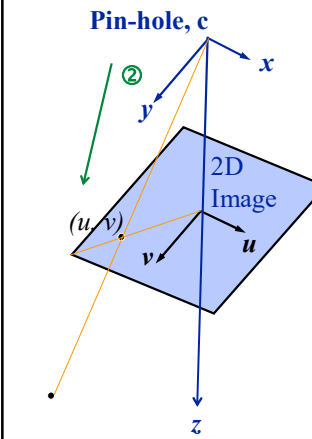


Step 1: Rigid body transformation
from world frame to
camera frame.

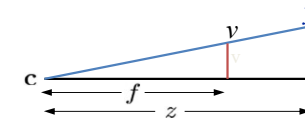
$$\begin{matrix} \text{Intermediate} & & \text{Knowns} \\ \begin{pmatrix} x \\ y \\ z \end{pmatrix} & = & \underbrace{\begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix}}_{[R]} \begin{pmatrix} X \\ Y \\ Z \end{pmatrix} + \underbrace{\begin{pmatrix} T_x \\ T_y \\ T_z \end{pmatrix}}_T \end{matrix}$$

Parameters to be calibrated:
 $[R], T$

Tsai Camera Model



Step 2: Perspective projection
with pin-hole geometry



Similarly, **Intermediate**

$$\begin{matrix} u = f \frac{x}{z} & v = f \frac{y}{z} \end{matrix}$$

Parameter to be calibrated:
 f

Tsai Camera Model

Step 3: Radial lens distortion correction

Assumption 1:
Radial lens distortion

$$\begin{bmatrix} u_d \\ v_d \end{bmatrix} = L(r) \begin{bmatrix} u \\ v \end{bmatrix}$$

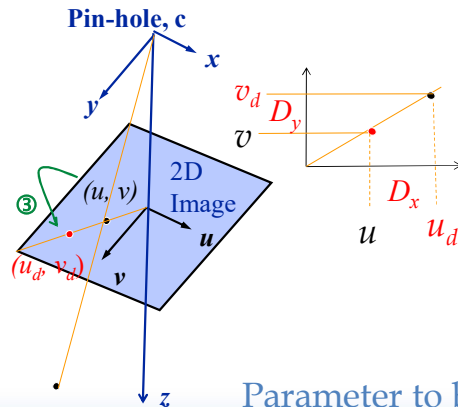
where $L = 1 + k_1 r^2 + \dots$

Actual measurements

$$\begin{aligned} u_d &= u(1 + k_1 r_d^2) \\ v_d &= v(1 + k_1 r_d^2) \end{aligned}$$

where $r_d^2 = u_d^2 + v_d^2$

Parameter to be calibrated: k_1



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Radial Distortion (straight lines curve around the image)



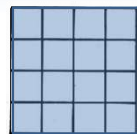
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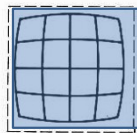
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Radial Distortion (straight lines curve around the image)

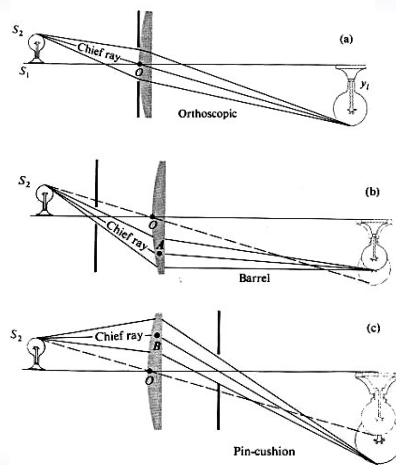
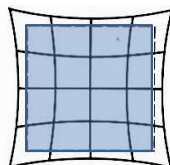
No distortion



Barrel



Pin cushion



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Correcting Radial Distortion (example)

⊕ Radial distortion of the image

- Caused by imperfect lenses
- Deviations are most noticeable for rays that pass through the edge of the lens



Distorted

After correction

courtesy of Shawn Becker



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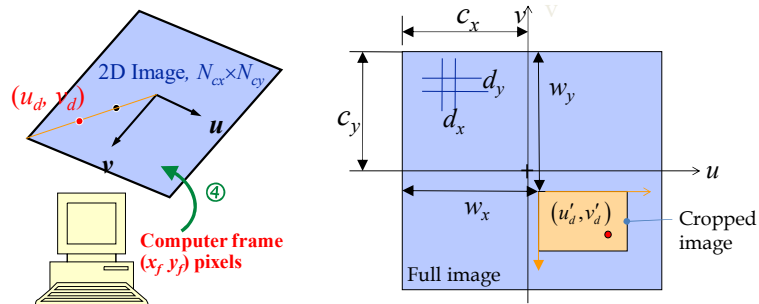
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Tsai's Camera Model

Step 4: Relation between physical and computer coordinates.

Frame (x_f, y_f) pixels
Image (u, v) in m (μm)
Camera (x, y, z) in m
World (X, Y, Z) in m



Given $(c_x, c_y), (w_x, w_y), (dx, dy)$, $\begin{bmatrix} u_d \\ v_d \end{bmatrix} = \begin{bmatrix} (w_x + u'_d - c_x) dx \\ -(w_y + v'_d - c_y) dy \end{bmatrix}$
find (u_d, v_d) from (u'_d, v'_d) .

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Tsai's Camera Model (Step 4 example)

Table 1 PULNIX Camera Properties

Imager	2/3 inch progressive scanning interline transfer CCD
No. Pixels	768 (H) x 484 (V)
Cell Size	11.6 μm x 13.6 μm progressive scanning
Scanning	525 lines, 30 Hz or 60 Hz 2:1 interlace
Sync	Internal/External autoswitch HD/VD 4.0 \forall p-p impedance 4.7 k Ω VD=interlace/non-interlace, HD=15.734 kHz \pm 5%
Dataclock Output	14.31818 MHz
TV Resolution	470 (H) x 484 (V) analog 760(H) x 484 (V) digital sampling
S/N Ratio	50 dB min. (AGC=off)
Min. Illumination	10.0 lux, f=1.4 (no shutter)
Size (WxHxL)	46 x 51 x 171.7 mm 1.81 x 2.0 x 6.766 inches
Weight	225 grams (4.3 oz)
Power Requirement	12 V DC 500 mA
Lens Mount	C Mount
Gamma	0.45 or 1.0 (0.45) std
Operating Temperature	-10 $^{\circ}$ C to 50 $^{\circ}$ C

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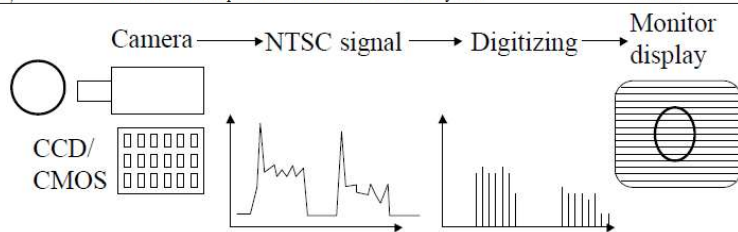
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Tsai Camera Model (Step 4 example)

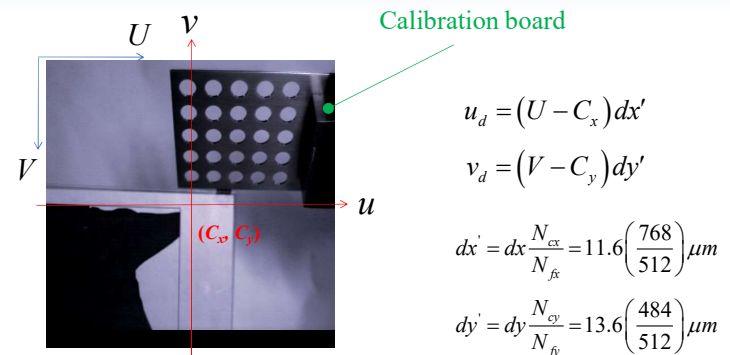
Table 2 Camera Sensor and Variables, Definitions, and Values

Variable	Definition	Value
dx	Center to center distance between adjacent sensor elements in the x scanline	11.6 μm
dy	Center to center distance between adjacent sensor elements in the y scanline	13.6 μm
N_{cx}	Number of sensor elements in the x direction	768 pixels
N_{cy}	Number of sensor elements in y direction	484 pixels
N_{fx}	Number of pixels in a line as sampled by the computer	512 pixels
N_{fy}	Number of rows (sensor elements plus blank rows) in y direction	512 pixels
C_x	Camera center x-coordinate taken to be the center of the camera sensor	768/2 = 384 pixels
C_y	Camera center y-coordinates taken to be the center of the camera sensor	484/2 = 242 pixels
w_x	X-coordinate of top-left window element defined by user	varies
w_y	Y-coordinate of top-left window element defined by user	varies



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Example 1: Use Camera Model Step 4 to convert from pixels to physical units



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Example 2 (one-one digital): Use Camera Model Step 4 to convert from pixels to physical units.

Calibration board

Cropped image

$u_d = (w_x + u'_d - C_x) dx$
 $v_d = -(w_y + v'_d - C_y) dy$

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Two-stage approach

Stage 1: Radial Alignment Constraint

1a) Solve for five parameters:

$$r_{11}/T_y, r_{21}/T_y, r_{12}/T_y, r_{22}/T_y, T_x/T_y$$

1b) Find T_y

1c) Solve for $[R], T_x, T_y$

Planar parameters

Out of plane parameters

Note: Only three of nine r_{ij} are independent.

$$[R]^T = [R]^{-1}$$

Stage 2: Perspective Constraint

Given $[R], T_x, T_y$ solve for f, k, T_z

Depth related parameters

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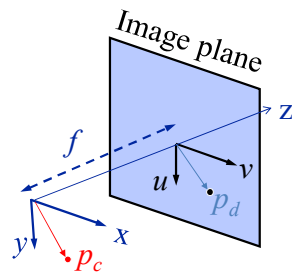
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Two-stage approach (Stage 1a)

1a) Solve for five parameters:

$$r_{11}/T_y, r_{21}/T_y, r_{12}/T_y, r_{22}/T_y, T_x/T_y$$

Assumption 2: xy plane $// uv$ plane,



$$\mathbf{p}_c \times \mathbf{p}_d = 0$$

$$\begin{bmatrix} i & j & k \\ x & y & 0 \\ u & v & 0 \end{bmatrix} = 0$$

$$xv - yu = 0 \quad (1)$$

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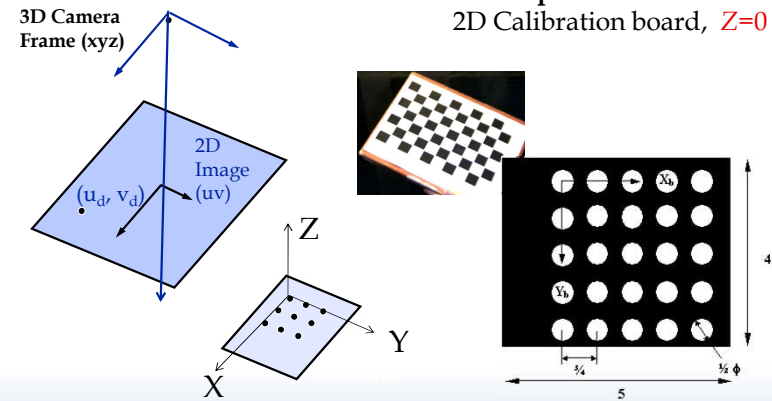
Two-stage approach (Stage 1a)

1a) Solve for five parameters:

$$r_{11}/T_y, r_{21}/T_y, r_{12}/T_y, r_{22}/T_y, T_x/T_y$$

Assumption 3:

2D Calibration board, $Z=0$



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Two-stage approach (Stage 1a)

1a) Solve for five parameters:

$$r_{11}/T_y, r_{21}/T_y, r_{12}/T_y, r_{22}/T_y, T_x/T_y$$

Assumption 2: xy plane // uv plane,

$$xv_d - yu_d = 0 \quad (1)$$

From Camera Model Step 1:

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \begin{pmatrix} X \\ Y \\ Z \end{pmatrix} + \begin{pmatrix} T_x \\ T_y \\ T_z \end{pmatrix} \quad (2)$$

Substituting (2) into (1):

$$(r_{11}X + r_{12}Y + r_{13}Z + T_x)v_d - (r_{21}X + r_{22}Y + r_{23}Z + T_y)u_d = 0$$

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Two-stage approach (Stage 1a)

1a) Solve for five parameters:

$$r_{11}/T_y, r_{21}/T_y, r_{12}/T_y, r_{22}/T_y, T_x/T_y$$

XYZ (world coordinate system)
Located on the calibration board;
also called $(x_b, y_b, z_b = 0)$.

Because of Assumption 3:
2D Calibration board, $Z=0$

$$(r_{11}X_i + r_{12}Y_i + r_{13}Z_i + T_x)v_d - (r_{21}X_i + r_{22}Y_i + r_{23}Z_i + T_y)u_d = 0$$

6 unknowns: $r_{11}, r_{12}, r_{21}, r_{22}, T_x, T_y$

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Two-stage approach (Stage 1a)

1a) Solve for five parameters:

$$r_{11}/T_y, r_{21}/T_y, r_{12}/T_y, r_{22}/T_y, T_x/T_y$$

$$(r_{11}X_i + r_{12}Y_i + T_x)v_{di} - (r_{21}X_i + r_{22}Y_i + T_y)u_{di} = 0$$

$$\underbrace{\begin{bmatrix} r_{11} \\ r_{12} \\ r_{21} \\ r_{22} \\ T_x \\ T_y \end{bmatrix}}_{\substack{[A] \\ n \times 6}} = 0$$

6 unknowns: $r_{11}, r_{12}, r_{21}, r_{22}, T_x, T_y$

Homogeneous equation: $[A]\mathbf{x}^* = 0$

For calibration, over-determined system
 $n \geq 6$ (trivial solutions, $\mathbf{x}^* = 0$)

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Two-stage approach (Stage 1a)

1a) Solve for five parameters using least square (pseudo-inverse):

$$\mu_1 = \mu_{11} = r_{11}/T_y; \mu_2 = \mu_{12} = r_{12}/T_y; \mu_3 = \mu_{21} = r_{21}/T_y; \mu_4 = \mu_{22} = r_{22}/T_y; \mu_5 = T_x/T_y$$

$$\left[X_i \left(\frac{r_{11}}{T_y} \right) + Y_i \left(\frac{r_{12}}{T_y} \right) + \left(\frac{T_x}{T_y} \right) \right] v_{di} - \left[X_i \left(\frac{r_{21}}{T_y} \right) + Y_i \left(\frac{r_{22}}{T_y} \right) + 1 \right] u_{di} = 0$$

$$\underbrace{\begin{bmatrix} X_1 v_{d1} & Y_1 v_{d1} & -X_1 u_{d1} & -Y_1 u_{d1} & v_{d1} \\ X_2 v_{d2} & Y_2 v_{d2} & -X_2 u_{d2} & -Y_2 u_{d2} & v_{d2} \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ X_n v_{dn} & Y_n v_{dn} & -X_n u_{dn} & -Y_n u_{dn} & v_{dn} \end{bmatrix}}_{\substack{[A] \\ n \times 5}} \underbrace{\begin{bmatrix} r_{11}/T_y \\ r_{12}/T_y \\ r_{21}/T_y \\ r_{22}/T_y \\ T_x/T_y \end{bmatrix}}_{\substack{\boldsymbol{\mu} \\ 5 \times 1}} = \underbrace{\begin{bmatrix} u_{d1} \\ u_{d2} \\ \vdots \\ u_{dn} \end{bmatrix}}_{\substack{\mathbf{b} \\ n \times 1}} \quad [A]\boldsymbol{\mu} = \mathbf{b} \quad \boldsymbol{\mu} = \mathbf{A}^+ \mathbf{b}$$

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Two-stage approach (Stage 1b, 1c)

Given $\mu_{ij} = r_{ij}/T_y$ ($i, j=1,2$) and $\mu_5 = r_5/T_y$

From pseudo-inverse solutions to Stage 1a):
 $\mu_1 = \mu_{11y}; \mu_2 = \mu_{12y};$
 $\mu_3 = \mu_{21y}; \mu_4 = \mu_{22y}$

$$[\mathbf{R}] = \begin{bmatrix} \mu_{11}T_y & \mu_{12}T_y & r_{13} \\ \mu_{21}T_y & \mu_{22}T_y & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} = \begin{bmatrix} \alpha^T \\ \beta^T \end{bmatrix} \quad \alpha = \begin{bmatrix} \mu_{11}T_y \\ \mu_{12}T_y \\ r_{13} \end{bmatrix}, \beta = \begin{bmatrix} \mu_{21}T_y \\ \mu_{22}T_y \\ r_{23} \end{bmatrix}$$

Find T_y using the orthogonality of $[\mathbf{R}]$

$$\alpha \cdot \beta = 0$$

$$\alpha \cdot \alpha = 1 \Rightarrow r_{13}^2 = 1 - T_y^2 (\mu_{11}^2 + \mu_{12}^2)$$

$$\beta \cdot \beta = 1 \Rightarrow r_{23}^2 = 1 - T_y^2 (\mu_{21}^2 + \mu_{22}^2)$$

$$\begin{bmatrix} r_{31} \\ r_{32} \\ r_{33} \end{bmatrix} = \begin{bmatrix} r_{11} \\ r_{12} \\ r_{13} \end{bmatrix} \times \begin{bmatrix} r_{21} \\ r_{22} \\ r_{23} \end{bmatrix}$$

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Two-stage approach (Stage 1(b, c))

Find T_y using the orthogonality of $[\mathbf{R}]$

$$\alpha \cdot \beta = 0$$

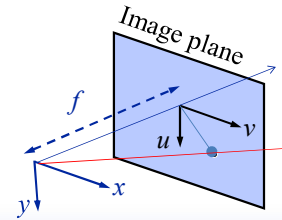
Step 1

Solve for T_y^2

$$\alpha \cdot \alpha = 1 \Rightarrow r_{13}^2 = 1 - T_y^2 (\mu_{11}^2 + \mu_{12}^2)$$

$$\beta \cdot \beta = 1 \Rightarrow r_{23}^2 = 1 - T_y^2 (\mu_{21}^2 + \mu_{22}^2)$$

Solve for r_{13}^2 and r_{23}^2



Multiple “±” solutions exist !

T_y may be positive or negative but

- 1) the image and object are in the same quadrant; and
- 2) f and T_z are positive (based on the definition and the fact that object is in front of the camera).

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Two-stage approach (Stage 1b, 1c)

Find T_y using the orthogonality of $[\mathbf{R}]$

$$\alpha = \begin{bmatrix} \mu_{11}T_y \\ \mu_{12}T_y \\ r_{13} \end{bmatrix}, \beta = \begin{bmatrix} \mu_{21}T_y \\ \mu_{22}T_y \\ r_{23} \end{bmatrix}$$

$$\alpha \cdot \beta = 0$$

$$T_y^2 \mu_{11} \mu_{21} + T_y^2 \mu_{12} \mu_{22} + r_{13} r_{23} = 0$$

To avoid square roots,

$$T_y^4 [\mu_{11} \mu_{21} + \mu_{12} \mu_{22}]^2 = [-r_{13} r_{23}]^2$$

$$\alpha \cdot \alpha = 1 \Rightarrow r_{13}^2 = 1 - T_y^2 (\mu_{11}^2 + \mu_{12}^2)$$

$$\beta \cdot \beta = 1 \Rightarrow r_{23}^2 = 1 - T_y^2 (\mu_{21}^2 + \mu_{22}^2)$$

$$T_y^4 \left[(\cancel{\mu_{11} \mu_{21}})^2 + 2(\mu_{11} \mu_{21})(\mu_{12} \mu_{22}) + (\cancel{\mu_{12} \mu_{22}})^2 \right] = [1 - T_y^2 (\mu_{11}^2 + \mu_{12}^2)] [1 - T_y^2 (\mu_{21}^2 + \mu_{22}^2)]$$

$$= 1 - T_y^2 (\mu_{11}^2 + \mu_{12}^2 + \mu_{21}^2 + \mu_{22}^2) + T_y^4 (\mu_{11}^2 + \mu_{12}^2)(\mu_{21}^2 + \mu_{22}^2)$$

$$= 1 - T_y^2 U + T_y^4 (\mu_{11}^2 + \mu_{12}^2)(\mu_{21}^2 + \mu_{22}^2)$$

$$= 1 - T_y^2 U + T_y^4 (\mu_{11}^2 \mu_{21}^2 + \mu_{11}^2 \mu_{22}^2 + \mu_{12}^2 \mu_{21}^2 + \mu_{12}^2 \mu_{22}^2)$$

$$\text{Let } U = \mu_{11}^2 + \mu_{12}^2 + \mu_{21}^2 + \mu_{22}^2 = \sum_{j=1}^4 \mu_j^2$$

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Two-stage approach (Stage 1b)

Find T_y using the orthogonality of $[\mathbf{R}]$

$$T_y^4 [2(\mu_{11} \mu_{21})(\mu_{12} \mu_{22})] = 1 - T_y^2 U + T_y^4 (\mu_{11}^2 \mu_{22}^2 + \mu_{12}^2 \mu_{21}^2)$$

$$T_y^4 \left[\underbrace{\mu_{11}^2 \mu_{22}^2 - 2(\mu_{11} \mu_{21})(\mu_{12} \mu_{22}) + \mu_{12}^2 \mu_{21}^2}_{(\mu_{11} \mu_{22} - \mu_{12} \mu_{21})^2} \right] - UT_y^2 + 1 = 0$$

$$(\mu_{11} \mu_{22} - \mu_{12} \mu_{21})^2 T_y^4 - UT_y^2 + 1 = 0$$

$$T_y^2 = \begin{cases} \frac{1}{U} & \text{if } \mu_1 \mu_4 = \mu_2 \mu_3 \\ U - \left[U^2 - 4(\mu_1 \mu_4 - \mu_2 \mu_3)^2 \right]^{1/2} & \text{if } \mu_1 \mu_4 \neq \mu_2 \mu_3 \end{cases}$$

where $U = \sum_{j=1}^4 \mu_j^2$

From pseudo-inverse solutions to Stage 1a):

$$\mu_1 = \mu_{11y}; \mu_2 = \mu_{12y};$$

$$\mu_3 = \mu_{21y}; \mu_4 = \mu_{22y}$$

Only the negative sign is relevant. Proof is given in Tsai's paper (Appendix)

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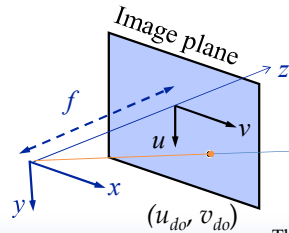
Two-stage approach (Stage 1b)

Let $T_y = (T_y^2)^{1/2}$ then $r_{ij} = \mu_{ij} T_y$ ($i, j=1,2$) and $T_x = \mu_5 T_y$

To determine the sign of T_y , select one object point $P(X_o, Y_o, 0)$:

$$\begin{bmatrix} \xi_x \\ \xi_y \end{bmatrix} = \begin{bmatrix} r_{11}X_o + r_{12}Y_o + T_x \\ r_{21}X_o + r_{22}Y_o + T_y \end{bmatrix}$$

If (ξ_x, ξ_y) have the sign as (u_{do}, v_{do}) , then T_y has the correct sign. Otherwise, negate it.



Select test point $P(X_o, Y_o, 0)$

The image and object are in the same quadrant;

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Two-stage approach (Stage 1c)

1c) Solve for $[R] T_x T_y$

The unknown signs s_1 and s_2 are determined from the orthogonal property of $[R]$.

$$[R] = \begin{bmatrix} r_{11} & r_{12} & s_1 \sqrt{1-r_{11}^2-r_{12}^2} \\ r_{21} & r_{22} & s_2 \sqrt{1-r_{21}^2-r_{22}^2} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \quad (\alpha \bullet \beta) = \alpha^T \beta = 0$$

$$\Rightarrow [r_{11} \ r_{12} \ r_{13}] \begin{bmatrix} r_{21} \\ r_{22} \\ r_{23} \end{bmatrix} = 0$$

$$\Rightarrow \underbrace{(r_{11}r_{21} + r_{12}r_{22})}_a + s_1 s_2 \underbrace{\sqrt{1-r_{11}^2-r_{12}^2} \sqrt{1-r_{21}^2-r_{22}^2}}_b = 0$$

The two factors, a and b , must be equal and opposite.

Thus, choose $s_1 = s_2 = +1$, check $\text{sgn}(r_{11}r_{21} + r_{12}r_{22}) = \text{negative}$

If yes, keep the signs; otherwise $s_2 = -1$.

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Summary of Stage 1

1a) Calculate T_y^2

1b) Let $T_y = (T_y^2)^{1/2}$ Determine the sign of T_y

$$\begin{bmatrix} \xi_x \\ \xi_y \end{bmatrix} = \begin{bmatrix} r_{11}X_o + r_{12}Y_o + T_x \\ r_{21}X_o + r_{22}Y_o + T_y \end{bmatrix}$$

If (ξ_x, ξ_y) have the sign as (u_{do}, v_{do}) , then T_y has the correct sign. Otherwise, negate it.

1c) Check $\text{sgn}(r_{11}r_{21} + r_{12}r_{22}) = \text{negative}$

Choose $s_1 = s_2 = +1$: If yes, keep the signs; otherwise $s_2 = -1$.

$$r_{13} = s_1 \sqrt{1 - T_y^2 (\mu_{11}^2 + \mu_{12}^2)} \quad r_{23} = s_2 \sqrt{1 - T_y^2 (\mu_{21}^2 + \mu_{22}^2)}$$

Note: The signs of (s_1, s_2) and all associated signs may need to be adjusted after computing f and T_z , which must be positive; two other possibilities:

$$s_1 = s_2 = -1 \text{ or } s_1 = -s_2 = -1$$

Calculate $\gamma = \alpha \times \beta$ where $\gamma = [r_{31} \ r_{32} \ r_{33}]^T$

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Two-stage approach (Stage 2)

Stage 2: Perspective Constraint

Given $[R] T_x T_y$ solve for f, k, T_z

Recall Camera Model Step 1 and Steps 2 and 3:

$$\begin{pmatrix} x \\ y \\ z \end{pmatrix} = \begin{bmatrix} r_{11} & r_{12} & r_{13} \\ r_{21} & r_{22} & r_{23} \\ r_{31} & r_{32} & r_{33} \end{bmatrix} \begin{pmatrix} X \\ Y \\ Z \end{pmatrix} + \begin{pmatrix} T_x \\ T_y \\ T_z \end{pmatrix} \Rightarrow \frac{x_i}{z_i} = \frac{r_{11}X_i + r_{12}Y_i + T_x}{r_{31}X_i + r_{32}Y_i + T_z}$$

$Z=0$ (2D board)

$$\left. \begin{aligned} u &= f \frac{x}{z} \\ u &= \frac{u_d}{(1 + k_1 r_d^2)} \end{aligned} \right\} \text{Eliminating } u, \quad \frac{x}{z} = \frac{u_d}{f(1 + k_1 r_d^2)}$$

$$u_d = f(1 + k_1 r_d^2) \frac{r_{11}X_i + r_{12}Y_i + T_x}{r_{31}X_i + r_{32}Y_i + T_z}$$

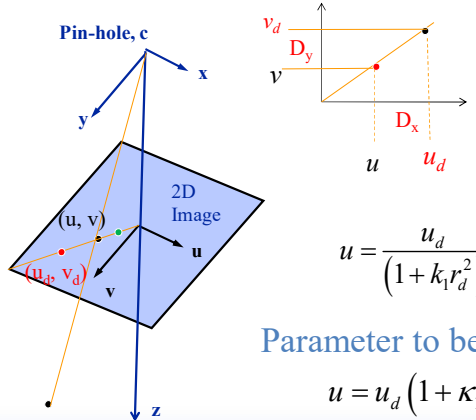
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Notes on Tsai's Camera Model

Recall Step 3: Radial lens distortion correction



where $r_d^2 = u_d^2 + v_d^2$

$u = \frac{u_d}{(1 + k_1 r_d^2)}$ (Modified)

Parameter to be calibrated: k_1

$u = u_d (1 + \kappa_1 r_d^2)$ (Tsai's paper)

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Notes on Tsai's Radial lens distortion

If Tsai's model were used,

Recall Steps 1 and 2

$$x_i = r_{11}X_i + r_{12}Y_i + T_x = \text{known}$$

$$z_i = r_{31}X_i + r_{32}Y_i + T_z$$

$$u_i = f \frac{x_i}{z_i}$$

$$u_i = u_{di} (1 + \kappa_1 r_{di}^2)$$

$$\frac{x_i}{z_i} = \frac{u_{di}}{f} (1 + \kappa_1 r_{di}^2)$$

$$\frac{u_{di}}{f} (1 + \kappa_1 r_{di}^2) = \frac{x_i}{r_{31}X_i + r_{32}Y_i + T_z}$$

$$u_{di} (r_{31}X_i + r_{32}Y_i + T_z) (1 + \kappa_1 r_{di}^2) = f x_i$$

Tsai's model for solving T_z , k_1 and f are non-linear!

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Notes on Tsai's Radial lens distortion

Recall Steps 1 and 2

$$x_i = r_{11}X_i + r_{12}Y_i + T_x = \text{known}$$

$$z_i = r_{31}X_i + r_{32}Y_i + T_z$$

Modified: $u_i = \frac{u_{di}}{(1 + \kappa_1 r_{di}^2)}$

$$\frac{u_{di}}{f (1 + \kappa_1 r_{di}^2)} = \frac{x_i}{r_{31}X_i + r_{32}Y_i + T_z}$$

$$\Rightarrow u_{di} (r_{31}X_i + r_{32}Y_i + T_z) = x_i f (1 + \kappa_1 r_{di}^2)$$

$$x_i f + x_i r_{di}^2 f \kappa_1 - u_{di} T_z = u_{di} (r_{31}X_i + r_{32}Y_i)$$

known

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Two-stage approach (Stage 2)

Stage 2: Perspective Constraint

Given $[R]$ T_x , T_y solve for f , k_1 , T_z

$$\begin{bmatrix} x_1 & r_{d1}^2 x_1 & -u_{d1} \\ x_2 & r_{d2}^2 x_2 & -u_{d2} \\ \vdots & \vdots & \vdots \\ x_n & r_{dn}^2 x_n & -u_{dn} \end{bmatrix} \begin{bmatrix} f \\ f k_1 \\ T_z \end{bmatrix} = \begin{bmatrix} (r_{31}X_1 + r_{32}Y_1)u_{d1} \\ (r_{31}X_2 + r_{32}Y_2)u_{d2} \\ \vdots \\ (r_{31}X_n + r_{32}Y_n)u_{dn} \end{bmatrix}$$

where $x_i = r_{11}X_i + r_{12}Y_i + T_x$

$$[A']x' = b' \quad x' = A'^+ b'$$

Note: The signs of (s_j, s_2) and all associated signs may need to be adjusted after computing f and T_z , which must be positive.

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Numerical Example

- ⊕ The following Table gives 5 point-correspondences input to the calibration system.
- ⊕ The units for both the world coordinate system and the u - v image coordinate system are centimeters.
- ⊕ Assume that this is no lens distortion, compute $(f, [\mathbf{R}], \mathbf{T})$

i	Object points			Image points	
	X_i	Y_i	Z_i	u_i	v_i
1	0.00	5.00	0.00	-0.58	0.00
2	10.00	7.50	0.00	1.73	1.00
3	10.00	5.00	0.00	1.73	0.00
4	5.00	10.00	0.00	0.00	1.00
5	5.00	0.00	0.00	0.00	-1.00

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Numerical Example

Stage 1 $[\mathbf{A}]\boldsymbol{\mu} = \mathbf{b}$

$$\boldsymbol{\mu} = \begin{bmatrix} \frac{r_{11}}{T_y} & \frac{r_{12}}{T_y} & \frac{r_{21}}{T_y} & \frac{r_{22}}{T_y} & \frac{T_x}{T_y} \end{bmatrix}^T$$

points	Object			Image	
i	X_i	Y_i	Z_i	u_i	v_i
1	0.00	5.00	0.00	-0.58	0.00
2	10.00	7.50	0.00	1.73	1.00
3	10.00	5.00	0.00	1.73	0.00
4	5.00	10.00	0.00	0.00	1.00
5	5.00	0.00	0.00	0.00	-1.00

$$\mathbf{A} = \begin{bmatrix} v_1 X_i & v_1 Y_i & -u_1 X_i & -u_1 Y_i & v_1 \\ 0 & 0 & 0 & 2.9 & 0 \\ 10 & 7.5 & -17.3 & -12.9 & 1 \\ 0 & 0 & -17.3 & -8.65 & 0 \\ 5 & 10 & 0 & 0 & 1 \\ -5 & 0 & 0 & 0 & -1 \end{bmatrix} \quad \mathbf{b} = \begin{bmatrix} u_i \\ -0.58 \\ 1.73 \\ 1.73 \\ 0 \\ 0 \end{bmatrix} \quad \boldsymbol{\mu} = \mathbf{A}^+ \mathbf{b} = \begin{bmatrix} -0.173 \\ 0 \\ 0 \\ -0.2 \\ 0.865 \end{bmatrix}$$

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Numerical Example (cont.)

Stage 1: Calculate T_y

$$U = \sum_{j=1}^4 \mu_j^2 = 0.0699$$

$$T_y^2 = \frac{U - [U^2 - 4(\mu_1 \mu_4 - \mu_2 \mu_3)]^{1/2}}{2(\mu_1 \mu_4 - \mu_2 \mu_3)^2} = 25$$

Try $T_y = +5$

$$r_{11} = -0.865; r_{12} = r_{21} = 0; r_{22} = -1; T_x = 4.325$$

$$\text{Check Point 2: } \xi_x = r_{11}X + r_{12}Y + T_x = -4.325$$

$$\xi_y = r_{21}X + r_{22}Y + T_y = -2.5$$

Wrong sign $\Rightarrow T_y = -5$.

$$\boldsymbol{\mu} = \begin{bmatrix} r_{11}/T_y \\ r_{12}/T_y \\ r_{21}/T_y \\ r_{22}/T_y \\ T_x/T_y \end{bmatrix} = \begin{bmatrix} -0.173 \\ 0 \\ 0 \\ -0.2 \\ 0.865 \end{bmatrix}$$

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Numerical Example (cont.)

$$T_y = -5 \quad \text{Recalculate} \quad r_{11} = 0.865; r_{12} = r_{21} = 0; r_{22} = 1; T_x = -4.325$$

$$\text{Try } s_1 = s_2 = +1 \quad r_{13} = s_1 \sqrt{1 - r_{11}^2 - r_{12}^2} = 0.5018 \quad r_{23} = s_2 \sqrt{1 - r_{21}^2 - r_{22}^2} = 0$$

$$\text{sign}(r_{11}r_{21} + r_{12}r_{22}) = 0 \quad \text{non-positive, OK}$$

$$\begin{bmatrix} \gamma \\ r_{31} \\ r_{32} \\ r_{33} \end{bmatrix} = \begin{bmatrix} \alpha \\ r_{11} \\ r_{12} \\ r_{13} \end{bmatrix} \times \begin{bmatrix} \beta \\ r_{21} \\ r_{22} \\ r_{23} \end{bmatrix} \quad \begin{aligned} r_{31} &= r_{12}r_{23} - r_{13}r_{22} = -0.5018 \\ r_{32} &= r_{13}r_{21} - r_{11}r_{23} = 0 \\ r_{33} &= r_{11}r_{22} - r_{12}r_{21} = 0.8650 \end{aligned}$$

$$\Rightarrow \mathbf{R} = \begin{bmatrix} 0.865 & 0 & 0.5018 \\ 0 & 1 & 0 \\ -0.5018 & 0 & 0.865 \end{bmatrix}$$

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Numerical Example (cont.)

Stage 2

points	Object			Image	
i	X_i	Y_i	Z_i	u_i	v_i
1	0.00	5.00	0.00	-0.58	0.00
2	10.00	7.50	0.00	1.73	1.00
3	10.00	5.00	0.00	1.73	0.00
4	5.00	10.00	0.00	0.00	1.00
5	5.00	0.00	0.00	0.00	-1.00

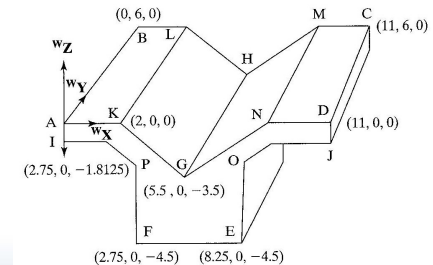
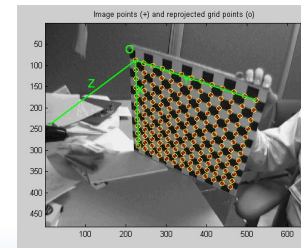
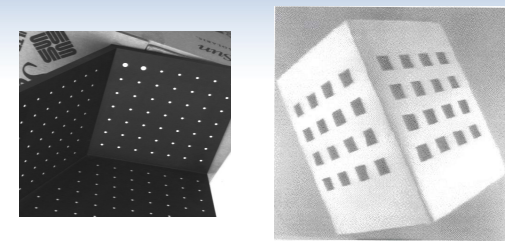
Solving for the unknown \mathbf{x}' ,
$$[\mathbf{A}'] \begin{bmatrix} f \\ T_z \end{bmatrix} = \mathbf{b}'$$

$$\mathbf{x}' = \mathbf{A}^+ \mathbf{b}' \quad \text{Hence, } f = 1.0123 \quad \mathbf{T} = \begin{bmatrix} -4.325 \\ -5 \\ 7.5484 \end{bmatrix}$$

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Example calibration boards



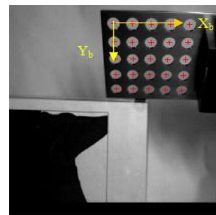
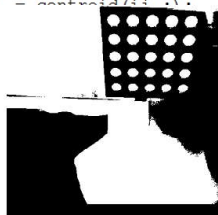
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Example

Getting feature points from image (using regionprops):

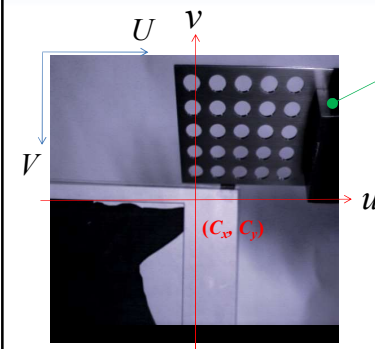
```
img_lab = bwlabel(img_bin);
props_struct = regionprops(img_lab, 'centroid','area'); %obtaining the
structure datatype with info
centroid = cat(1, centr_v.Centroid); %converting desired structured data
into numerical
count = 0;
%using area to filter features
for ii = 1:length(areas)
    areav = areas(ii);
    if (areav > 200) & (areav < 800)
        count = count+1;
        area_t(count) = areav;
        centr_t(count,:) = centr_v(ii,:);
    end
end
```



Binarized image and calibration board image with 25 feature points labeled with '+'.
 Y_i X_i

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Use Camera Model Step 4 to convert from pixels to physical units



Calibration board

$$u_d = (U - c_x) dx'$$

$$v_d = (V - c_y) dy'$$

$$dx' = dx \frac{N_{cx}}{N_{fx}} = 11.6 \left(\frac{768}{512} \right) \mu m$$

$$dy' = dy \frac{N_{cy}}{N_{fy}} = 13.6 \left(\frac{484}{512} \right) \mu m$$

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Use Camera Model Step 4 to convert from pixels to physical units

Cropped image

Calibration board

$u_d = (w_x + u'_d - c_x) dx$

$v_d = (w_y + v'_d - c_y) dy$

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Example (cont., Results)

Camera Model Step 4:

***NOTE: IN THIS CASE, $c_{fx} = 256$, $c_{fy} = c_{fx} = 256$

```
Xd_phy = (Xd_pix(:,1) - Cfx)*dyp; %use computer centers!!!
Yd_phy = (Xd_pix(:,2) - Cfy)*dyp; %use computer centers!!!
```

Stage I:

R - Rotation matrix			T - Translation mm (in)	
0.9979	0.0412	0.0492	T_x	-1.8719 (-0.0737)
0.0147	0.5992	-0.8004	T_y	-63.5523 (-2.5011)
-0.0624	0.7995	0.5974		

Stage II:

f, κ_1, T_z - Stage 2 parameters

$f = 14.1 \text{ mm}$ (0.5568 in) $\kappa_1 = 1228.8689 \text{ mm}^{-2}$ (0.7766 in^{-2}) $T_z = 336.5 \text{ mm}$ (13.2457 in)

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Example (cont., Results)

Solving for parameters using distortion model 1 for intrinsic parameters.

R - Rotation matrix			T - Translation mm (in)	
0.9979	0.0412	0.0492	T_x	-1.8719 (-0.0737)
0.0147	0.5992	-0.8004	T_y	-63.5523 (-2.5011)
-0.0624	0.7995	0.5974	T_z	340.6396 (13.4057)

Intrinsic parameters

$f = 14.3 \text{ mm}$ (0.5622 in) $\kappa_1 = -879.7752 \text{ mm}^{-2}$ (-0.5560 in^{-2})

In terms of Euler angles yaw θ , pitch ϕ , and tilt ψ for rotation,

$$R = \begin{bmatrix} \cos \psi \cos \theta & \sin \psi \cos \theta & -\sin \theta \\ -\sin \psi \cos \phi + \cos \psi \sin \theta \cos \phi & \cos \psi \cos \phi + \sin \psi \sin \theta \sin \phi & \cos \theta \sin \phi \\ \sin \psi \sin \phi + \cos \psi \sin \theta \cos \phi & -\cos \psi \sin \phi + \sin \psi \sin \theta \cos \phi & \cos \theta \cos \phi \end{bmatrix}$$

θ (x-axis rotation)	ϕ (y-axis rotation)	ψ (z-axis rotation)
$\theta = \tan^{-1}(r_{23} / r_{33})$	$\phi = -\sin^{-1}(r_{13})$	$\psi = \tan^{-1}(r_{12} / r_{11})$
-53°	-3°	2°

The results satisfy $[\mathbf{R}]^T = [\mathbf{R}]^{-1}$

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Example

(validating results)

$$\mathbf{x}_i = [\mathbf{R}]\mathbf{X}_i + \mathbf{T}$$

$$\text{RHS} = [\mathbf{R}]\mathbf{X}_i + \mathbf{T}$$

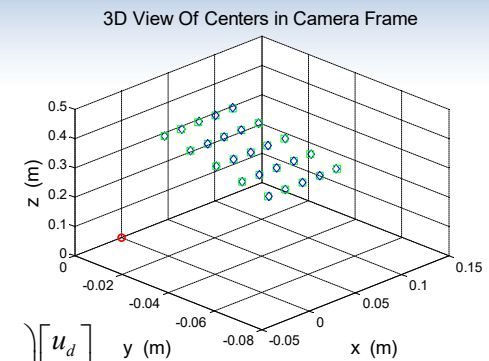
$$\text{LHS: } \begin{bmatrix} x_i \\ y_i \end{bmatrix} = \frac{z_i}{f} \left(\frac{1}{1 + \kappa_1 r_d^2} \right) \begin{bmatrix} u_d \\ v_d \end{bmatrix}$$

3 sample points from the figure above

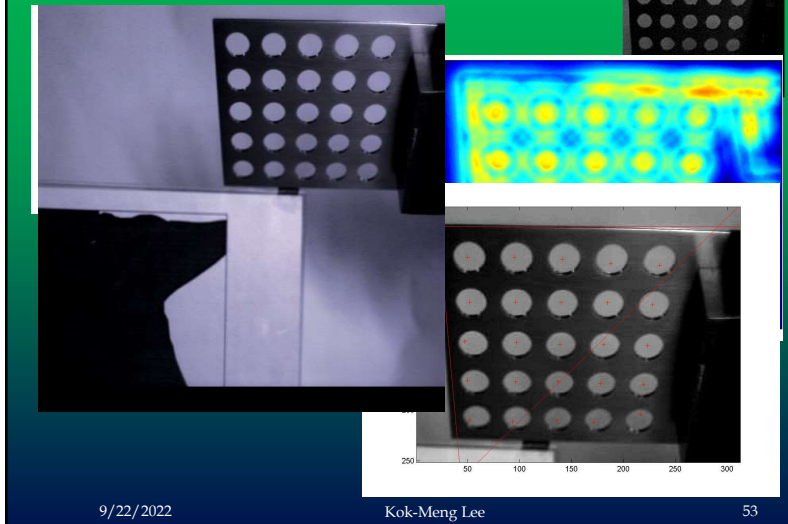
x coordinates (m)		y coordinates (m)		z coordinates (m)	
LHS	RHS	LHS	RHS	LHS	RHS
-0.0018444	-0.0018719	-0.0637924	-0.0635523	0.3388329	0.3388329
-0.0011130	-0.0010866	-0.0521947	-0.0521322	0.3540779	0.3540779
-0.0003378	-0.0003012	-0.0407762	-0.0407121	0.3693227	0.3693227

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Example



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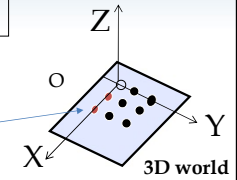
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Simplified for Stage 1a)

1a) Solve for two parameters: $r_{11}/T_y, r_{21}/T_y$

- Appropriately position the camera relative to the calibration such that $T_y > 0$ and $T_x/T_y = \rho$.
- Select points on the x-axis of the 2D calibration board, $Y=Z=0$.



$$(r_{11}X_i + r_{12}Y_i + r_{13}Z_i + T_x)v_{di} - (r_{21}X_i + r_{22}Y_i + r_{23}Z_i + T_y)u_{di} = 0$$

$$X_i v_{di} \underbrace{\left(\frac{r_{11}}{T_y} \right)}_{\mu_1} + v_{di} \underbrace{\left(\frac{T_x}{T_y} \right)}_{\text{known } \rho} - X_i u_{di} \underbrace{\left(\frac{r_{21}}{T_y} \right)}_{\mu_3} - u_{di} = 0$$

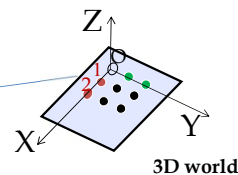
$$\begin{bmatrix} X_i v_{di} & -X_i u_{di} \end{bmatrix} \begin{bmatrix} \mu_1 \\ \mu_3 \end{bmatrix} = u_{di} - \rho v_{di}$$

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Simplified for Stage 1a)

1a) Similarly, solve for $r_{12}/T_y, r_{22}/T_y$

- Appropriately position the camera relative to the calibration such that $T_y > 0$ and $T_x/T_y = \rho$.
- Select points on the x-axis of the 2D calibration board, $X=Z=0$.



Let $a_{i1} = X_i v_{di}$; $a_{i2} = X_i u_{di}$; and $b_i = u_{di} - \rho v_{di}$

$$\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix} \begin{bmatrix} \mu_1 \\ \mu_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \end{bmatrix} \Rightarrow \begin{bmatrix} \mu_1 \\ \mu_3 \end{bmatrix} = \frac{1}{\det[\mathbf{A}]} \begin{bmatrix} a_{22} & -a_{12} \\ -a_{21} & a_{11} \end{bmatrix} \begin{bmatrix} b_1 \\ b_2 \end{bmatrix}$$

where $\det[\mathbf{A}] = a_{11}a_{22} - a_{12}a_{21}$

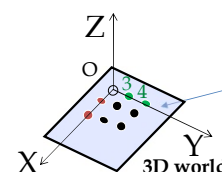
$$\begin{bmatrix} \mu_1 \\ \mu_3 \end{bmatrix} = \frac{1}{X_1 v_{d1} X_2 u_{d2} - X_1 u_{d1} X_2 v_{d2}} \begin{bmatrix} X_2 u_{d2} & -X_1 u_{d1} \\ -X_2 v_{d2} & X_1 v_{d1} \end{bmatrix} \begin{bmatrix} u_{d1} - \rho v_{d1} \\ u_{d2} - \rho v_{d2} \end{bmatrix}$$

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Simplified for Stage 1a)

1a) Similarly, solve for $r_{12}/T_y, r_{22}/T_y$

- Appropriately position the camera relative to the calibration such that $T_y > 0$ and $T_x/T_y = \rho$.
- Select points on the x-axis of the 2D calibration board, $X=Z=0$.



$$(r_{11}X_j + r_{12}Y_j + r_{13}Z_j + T_x)v_{dj} - (r_{21}X_j + r_{22}Y_j + r_{23}Z_j + T_y)u_{dj} = 0$$

$$Y_j v_{dj} \underbrace{\left(\frac{r_{12}}{T_y} \right)}_{\mu_2} + v_{dj} \underbrace{\left(\frac{T_x}{T_y} \right)}_{\text{known } \rho} - Y_j u_{dj} \underbrace{\left(\frac{r_{22}}{T_y} \right)}_{\mu_4} - u_{dj} = 0 \quad \begin{bmatrix} Y_j v_{dj} & -Y_j u_{dj} \end{bmatrix} \begin{bmatrix} \mu_2 \\ \mu_4 \end{bmatrix} = u_{dj} - \rho v_{dj}$$

$$\begin{bmatrix} \mu_2 \\ \mu_4 \end{bmatrix} = \frac{1}{Y_3 v_{d3} Y_4 u_{d4} - Y_3 u_{d3} Y_4 v_{d4}} \begin{bmatrix} Y_4 u_{d4} & -Y_3 u_{d3} \\ -Y_4 v_{d4} & Y_3 v_{d3} \end{bmatrix} \begin{bmatrix} u_{d3} - \rho v_{d3} \\ u_{d4} - \rho v_{d4} \end{bmatrix}$$

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