



ZJU-UIUC Institute

Zhejiang University / University of Illinois at Urbana-Champaign Institute



ECE 470: Introduction to Robotics

Lecture 27

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Camera-Based Pose Estimation

ECE 470 Introduction to Robotics

Schedule Check

• Lecture


O.	Overview	
	• Science & Engineering in Robotics	
I.	Spatial Representation & Transformation	Fundamentals
	• Coordinate Systems; Pose Representations; Homogeneous Transformations	Week 1-4
II.	Kinematics	
	• Multi-body frame assignment; D-H Convention; Joint-space; Work-space; Forward/Inverse Kinematics	Revision/ Quiz on Week 5
III.	Velocity Kinematics and Static Forces	
	• Translational/Rotational Velocity; Joint torque; Generalized Force Coordinates; Jacobian; Singularity	
IV.	Dynamics	Essentials
	• Acceleration of Body; Newton-Euler Equations of Motion; Lagrangian Formulation	
V.	Control	Week 6-9
	• Closed-Loop Control and Feedback, Control of 2 nd order system, Independent Joint Control, Force Control	
VI.	Planning	Revision/ Quiz on Week 10
	• Joint-Based Scheme; Cartesian-Based Scheme; Collision Free Path Planning	
	VII. Robot Vision (Perception)	Applied
	• Image Formation; Image Processing; Visual Tracking & Pose Estimation; Vision-based Control & Image-guided robotics	Week 11-14
		Reading Wk/ Exam on Week 15-16

Image Mapping

- Relating different viewpoints of on a common scene
- Image registration and geometric transformation
- How?
 - Recall the techniques learn so far: detect, describe, match...
 - Then, solve for transformation (homography) based on a specific model: translation, rigid, similarity, affine, projective

Image Mapping

- Homography
 - Relationship (i.e. transformation; mapping) between two (planar) images
- Geometric image transformation

Image Mapping

- Geometric image transformation: Types, Representations, DOF, Attributes

Name	Matrix	# D.O.F.	Preserves:
translation	$\begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} t_x \\ t_y \end{bmatrix}$	2	orientation + ...
rigid (Euclidean)	$\begin{bmatrix} R & t \\ 0 & 1 \end{bmatrix}$	3	lengths + ...
similarity	$\begin{bmatrix} sR & t \\ 0 & 1 \end{bmatrix}$	4	angles + ...
affine	$\begin{bmatrix} A & t \\ 0 & 1 \end{bmatrix}$	6	parallelism + ...
projective	$\begin{bmatrix} H \\ 0 & 1 \end{bmatrix}$	8	straight lines

Image Mapping: Application Examples

Constructing a map for the environment

Visualization and Navigation

Integrated Self-Contained Navigation

2D Topological Map 3D Navigation Map
3D Placental Vasculature Image Mapping

Visual Sensing and Perception

- The Physics
 - Principles of image projection (camera model)
- The Mathematics
 - Projective geometry and spatial representation
- The Techniques
 - Camera-based pose estimation, camera calibration and robot-camera calibration

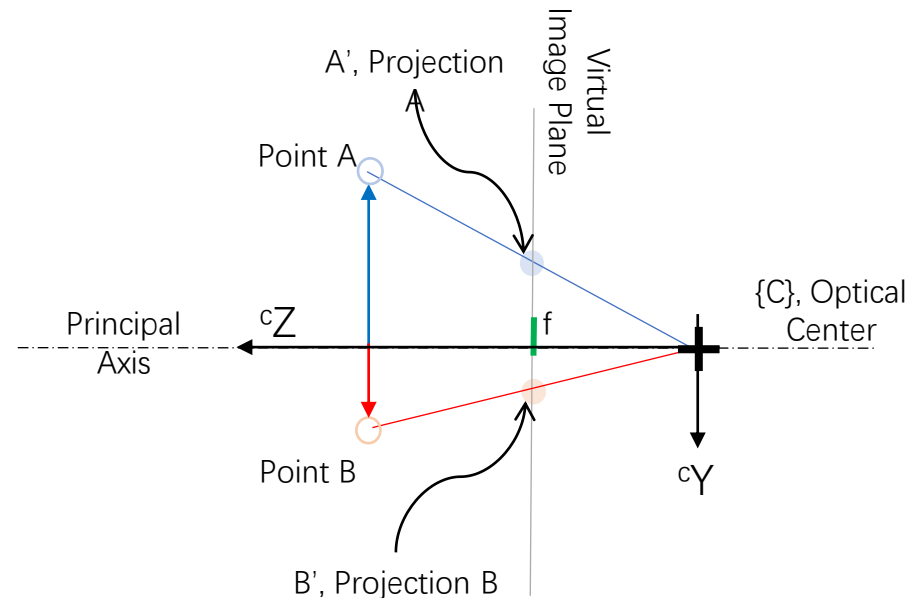


Camera Model

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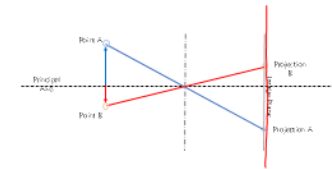
Recall: Camera Model

- Central Projection Camera Model
 - A simplified model for camera geometry



Recall: Camera Model

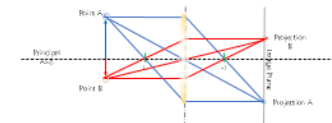
- Pin-hole Camera Model
 - describes the relationship between points and their projections on an image plane via a small opening



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Recall: Camera Model

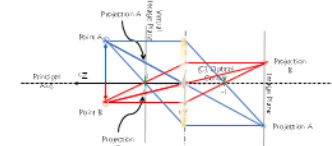
- Convex Focusing Lens with Pin-hole Effect
 - same image formation geometry but light passing not restrict



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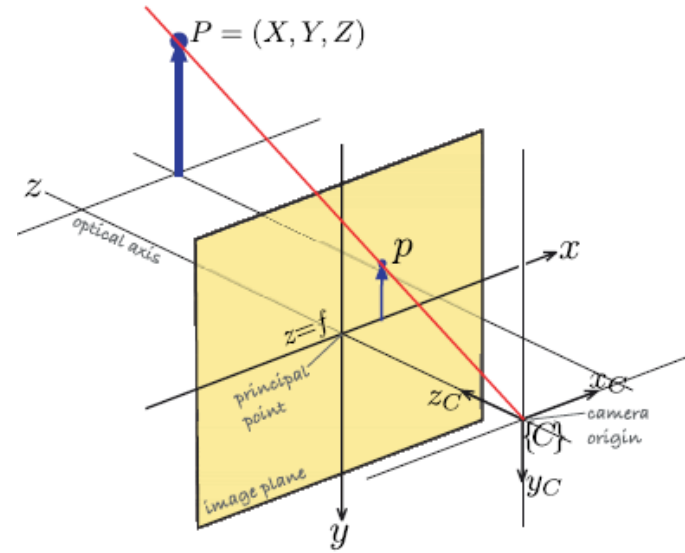
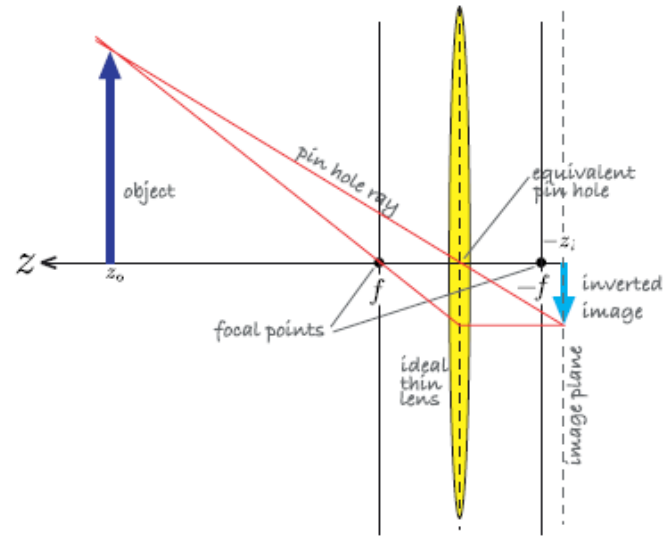
Recall: Camera Model

- Central Projection Camera Model
 - A simplified model for camera geometry



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Camera Model



$$x = f_x \frac{X}{Z}$$

$$y = f_y \frac{Y}{Z}$$

Corke, Peter. *Robotics, vision and control: fundamental algorithms in MATLAB.*

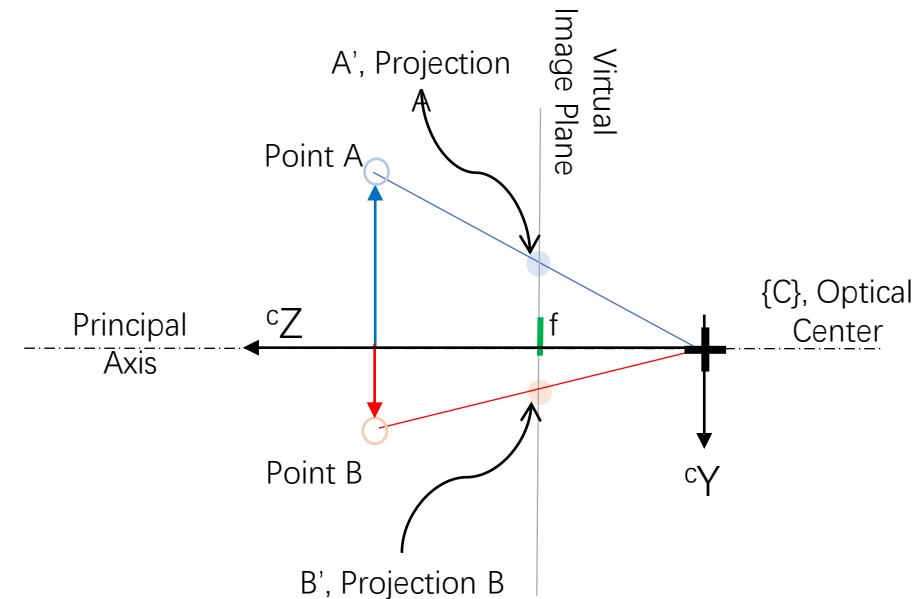
Coordinate Systems

- Examples of Coordinate Systems:
 - Cartesian \$(x, y, z)\$
 - Spherical \$(r, \theta, \phi)\$
 - Cylindrical \$(\rho, \theta, z)\$
- Homogenous Coordinate System
- Projective coordinates

Matrix Representation: Intrinsic Matrix

- Intrinsic Matrix, $[K]$
 - focal length: $(f_x \ f_y)^T$
 - principal point: $(i_c \ j_c)^T$
 - skew coefficient: a

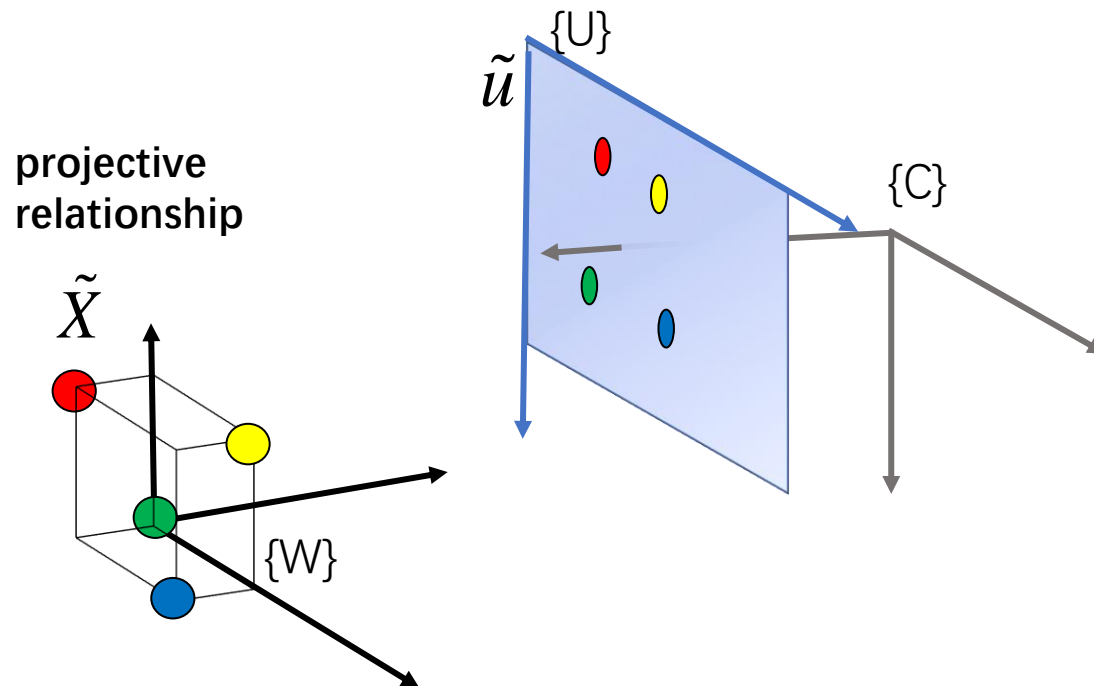
$$K = \begin{bmatrix} f_x & a & i_c \\ 0 & f_y & j_c \\ 0 & 0 & 1 \end{bmatrix}$$



What about a moving camera?

Matrix Representation: Extrinsic Matrix

- Extrinsic Matrix, ${}^c[R \mid t]$
 - R : Orientation of world reference frame w.r.t. camera coord.
 - t : Position offset of world reference frame w.r.t. camera coord.



Camera Matrix

- Camera Matrix, M
 - Relates world with image coord. System
 - 2 Components:
 - Extrinsic Matrix
 - Intrinsic Matrix

Camera Matrix, M

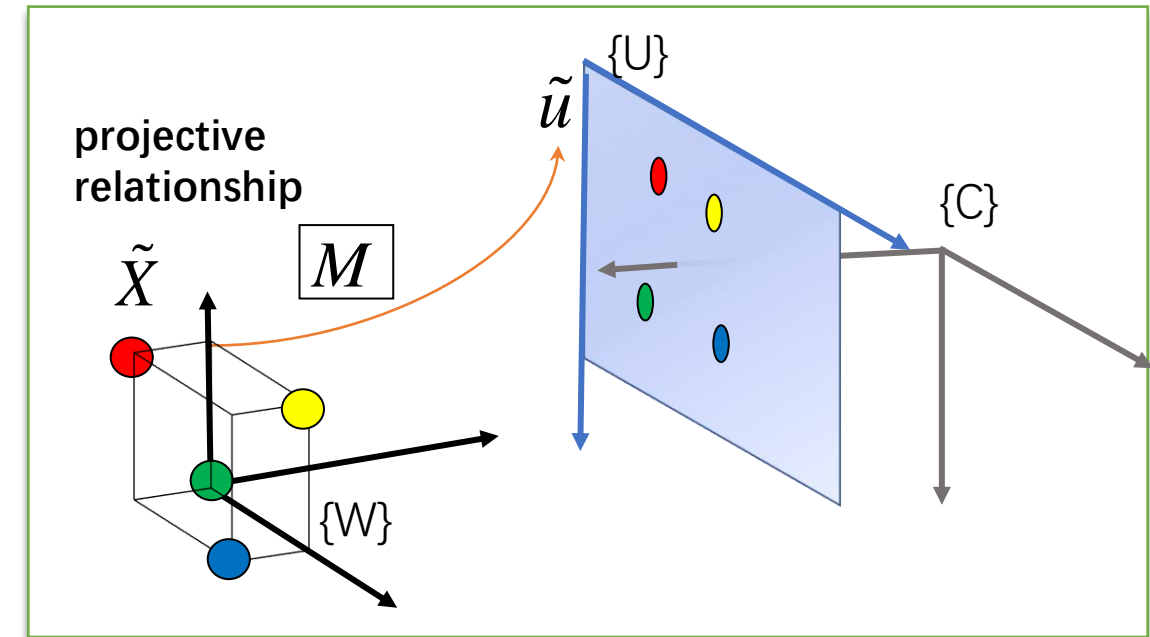
For a given set of points

${}^w\tilde{X}$ in 3D,

the projected set of points can be expressed as

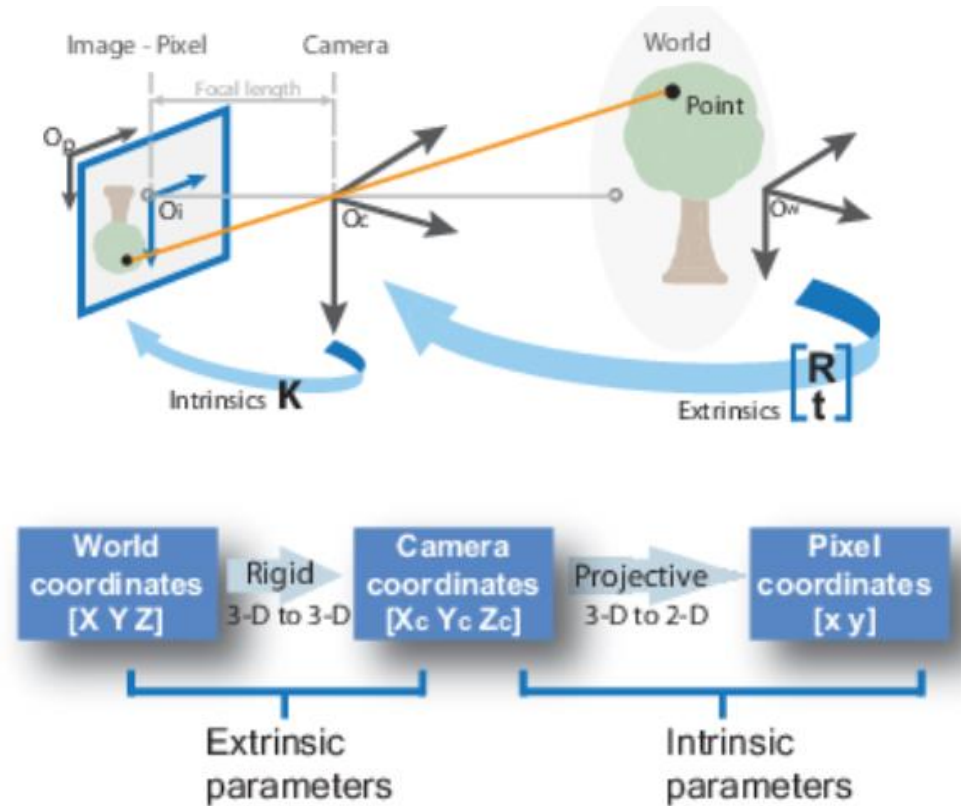
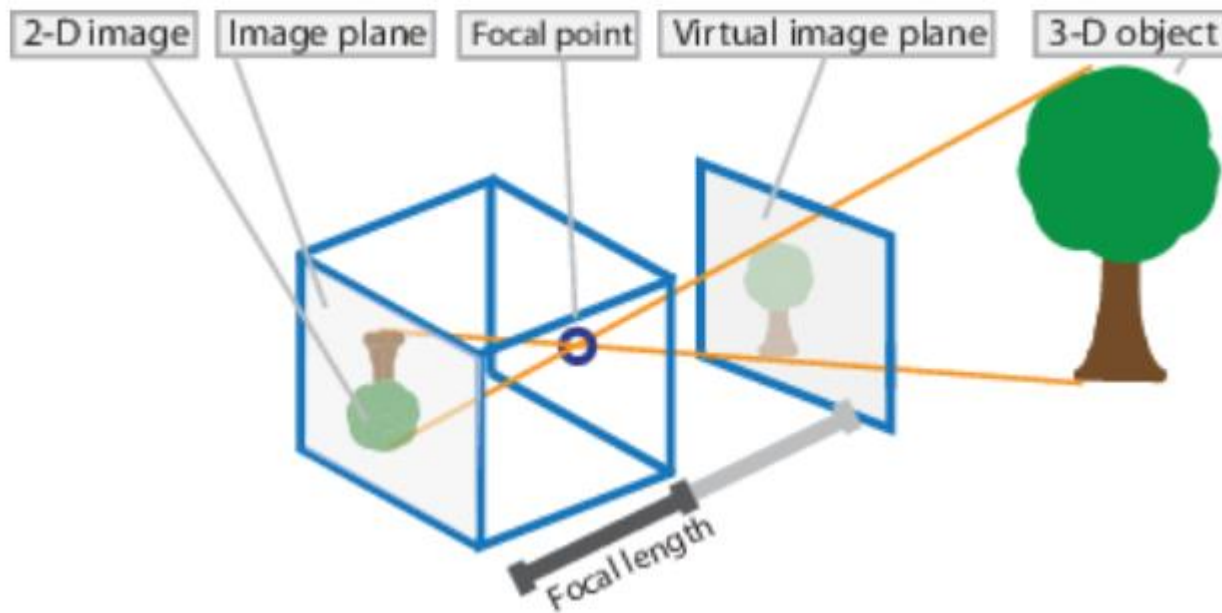
$$s\tilde{u} = M {}^w\tilde{X}, \quad \text{where } M = K {}^c[R|t]_w,$$

K = intrinsic matrix
 $[R|t]$ = extrinsic matrix



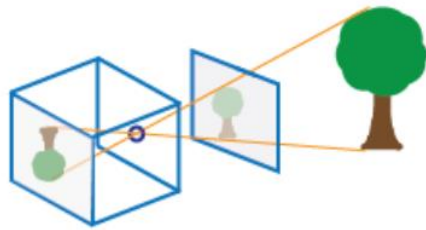
Camera Model

- Pin-hole Projective Camera Model

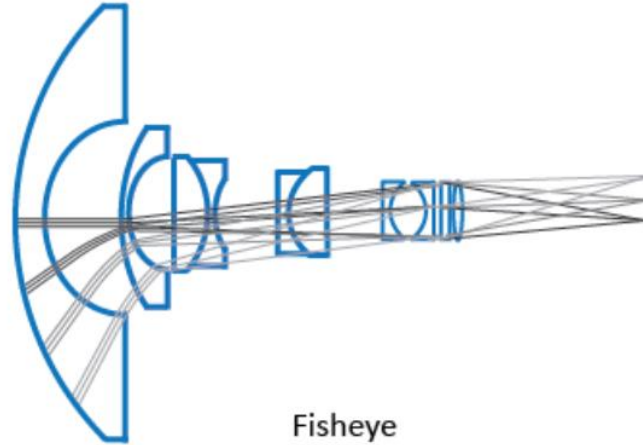


Camera Model

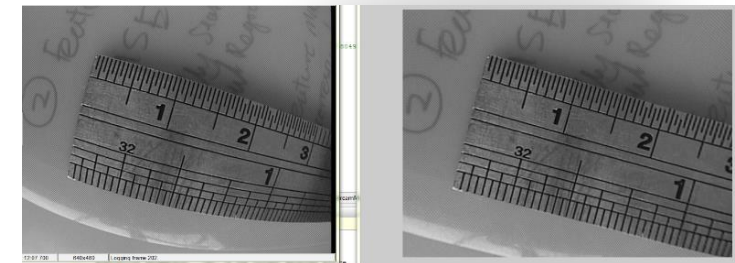
- Pin-hole Projective Camera Model
- Lens Distortion Model



Pinhole



Fisheye



Camera Calibration

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Camera Calibration

- Aim
 - To obtain a mapping between 3D pts in the world to their pixel coord. on the image plane
- Application
 - Robot & automation; object measurement; vision-based control; navigation; 3D reconstruction of scene...

Camera Calibration

- Two approach to solve for Camera Matrix, P
 1. Direct Linear Transformation (DLT)
 - Obtain several corresponding 3D world pts. \tilde{X} and 2D image pts. \tilde{u}
 - (Qn: how many pairs needed?)
 - (Qn: what happen if all points of \tilde{X} lie on a plane?)
 - Estimate a matrix \hat{P} that minimized the square error
 2. Solve K & $[R|t]$ separately (aka camera resection)
 - Perform geometric calibration to obtain K
 - Use perspective-n-point method to solve $[R|t]$
 - (Qn: how many points needed?)

Camera Calibration (DLT)

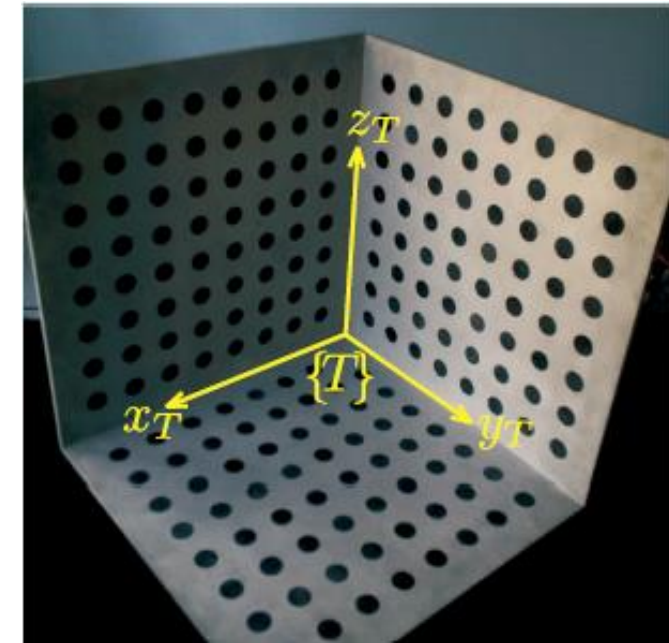
- Solve entire camera matrix

For a given point ${}^w\tilde{X}$ in 3D world, its projection on the image plane can be expressed as

$$s\tilde{u} = M {}^w\tilde{X}$$

$$s \begin{bmatrix} u \\ v \\ 1 \end{bmatrix} = \begin{bmatrix} M(1,1:4) \\ M(2,1:4) \\ M(3,1:4) \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$$

Qn: What is the last row of equation for?



Corke, Peter. *Robotics, vision and control: fundamental algorithms in MATLAB*.

Camera Calibration (DLT)

$$\begin{bmatrix} M(1,1:4) \\ M(2,1:4) \end{bmatrix} \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix} - s \begin{bmatrix} u \\ v \end{bmatrix} = \tilde{0}$$

where $s = [M(3,1:4)] \begin{bmatrix} X \\ Y \\ Z \\ 1 \end{bmatrix}$

$$M(1,1)X + M(1,2)Y + M(1,3)Z + M(1,4) - s\tilde{u} = 0$$

$$M(2,1)X + M(2,2)Y + M(2,3)Z + M(2,4) - s\tilde{v} = 0$$

Substitute s ,

$$M(1,1)X + M(1,2)Y + M(1,3)Z + M(1,4) - M(3,1)Xu - M(3,2)Yu - M(3,3)Zu - M(3,4)u = 0$$

$$M(2,1)X + M(2,2)Y + M(2,3)Z + M(2,4) - M(3,1)Xv - M(3,2)Yv - M(3,3)Zv - M(3,4)v = 0$$

Solve for M in the form of

$$\mathbf{AM} = \mathbf{0}$$

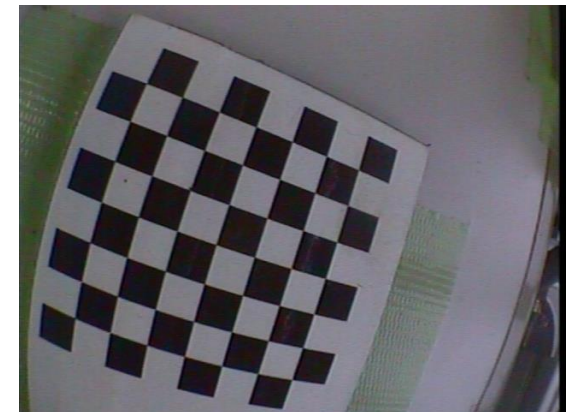
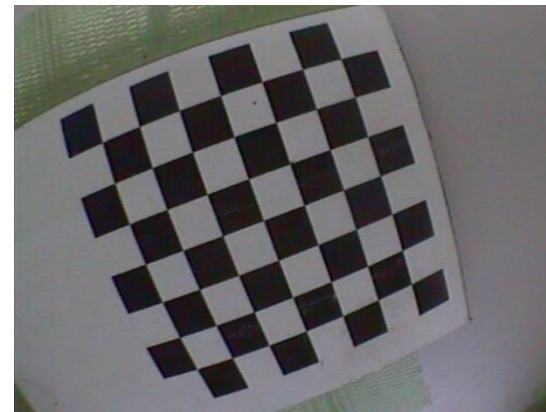
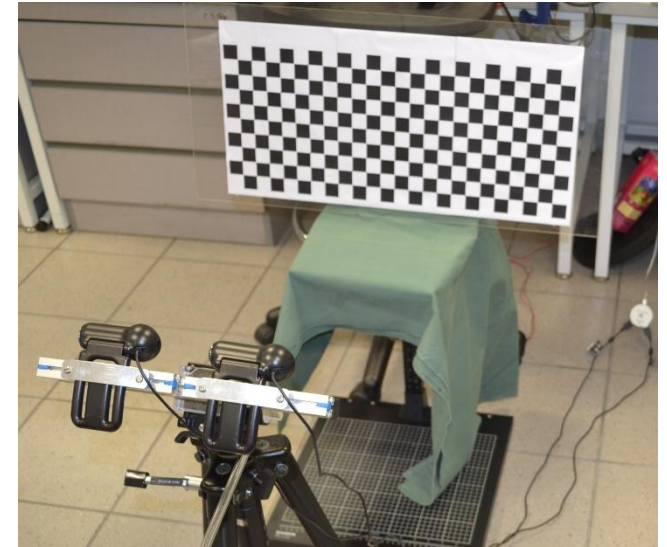
Don't worry, matlab can deal with it, not so impt...

What if they are planar?

Camera Calibration (Geometric Calibration)

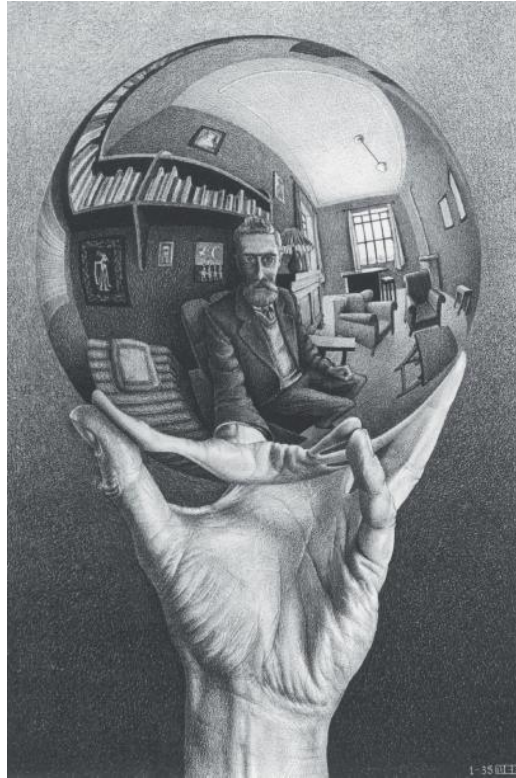
- Intrinsic parameters
 - (i_o, j_o) : Principal Point (optical center)
 - (f_x, f_y) : Focal Length
 - a : Skew Coefficient

$$\mathbf{K} = \begin{pmatrix} f_x & a & i_o \\ 0 & f_y & j_o \\ 0 & 0 & 1 \end{pmatrix}$$

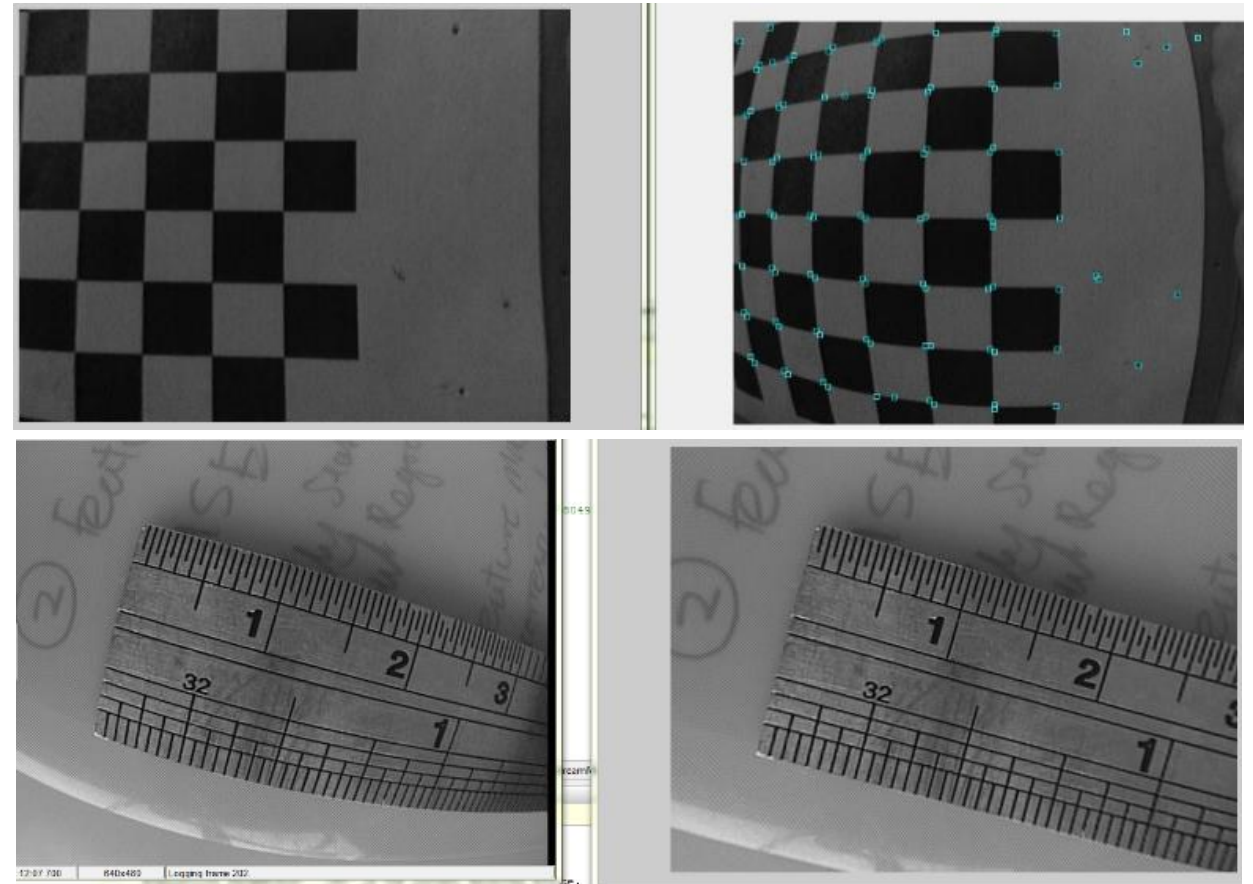


Camera Calibration (Geometric Calibration)

- Lens Distortion

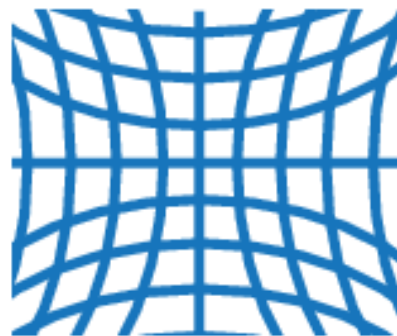


Hand with Reflecting Sphere, M. C. Escher, 1935

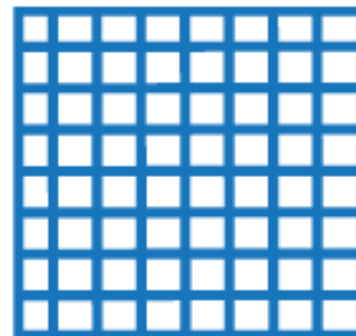


Camera Calibration (Geometric Calibration)

- Lens Distortion
 - Radial
 - Tangential



Negative radial distortion
"pincushion"



No distortion

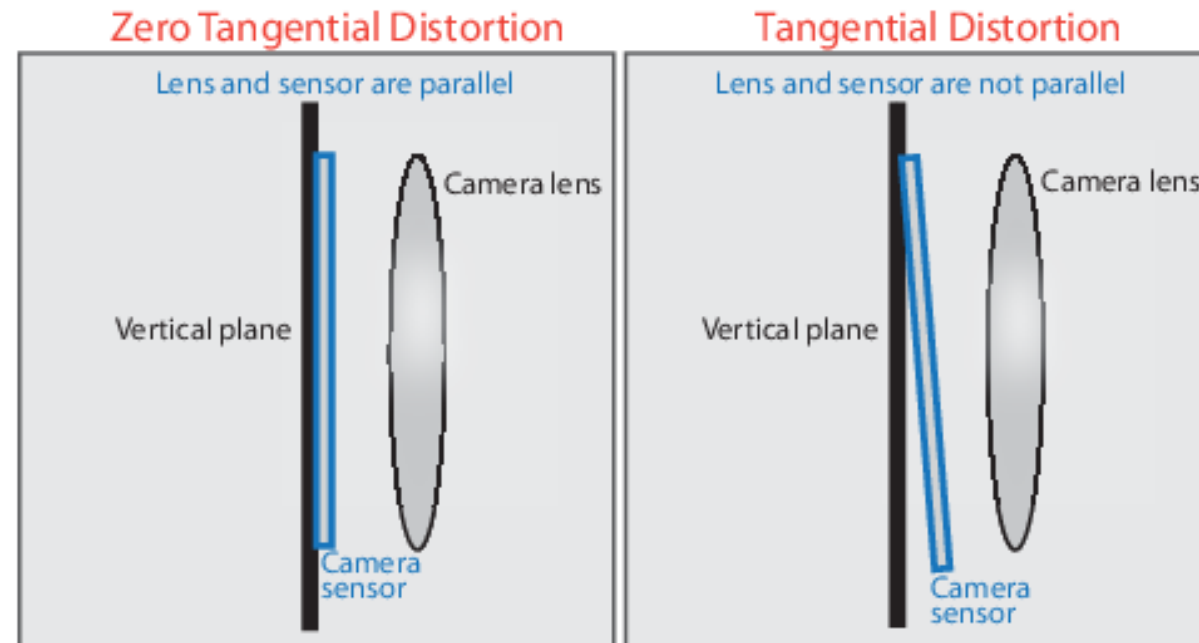


Positive radial distortion
"barrel"

"What Is Camera Calibration?", Mathworks Inc.,
<http://www.mathworks.com/help/vision/ug/camera-calibration.html>

Camera Calibration (Geometric Calibration)

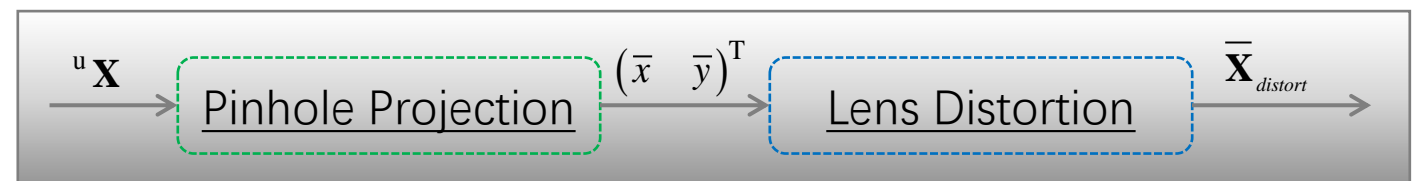
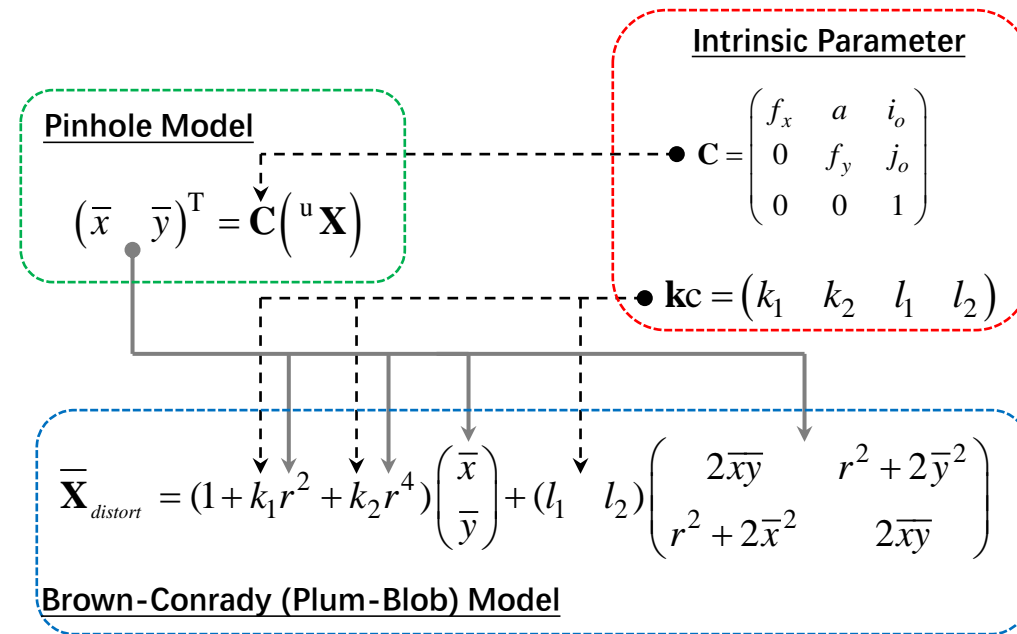
- Lens Distortion
 - Radial
 - Tangential



“What Is Camera Calibration?”, Mathworks Inc.,
<http://www.mathworks.com/help/vision/ug/camera-calibration.html>

Camera Model

- Lens Distortion



Camera Calibration (Geometric Calibration)

- Lens Distortion
 - Brown-Conrady Model (FYI)

Brown-Conrady (Plum-Blob) Model

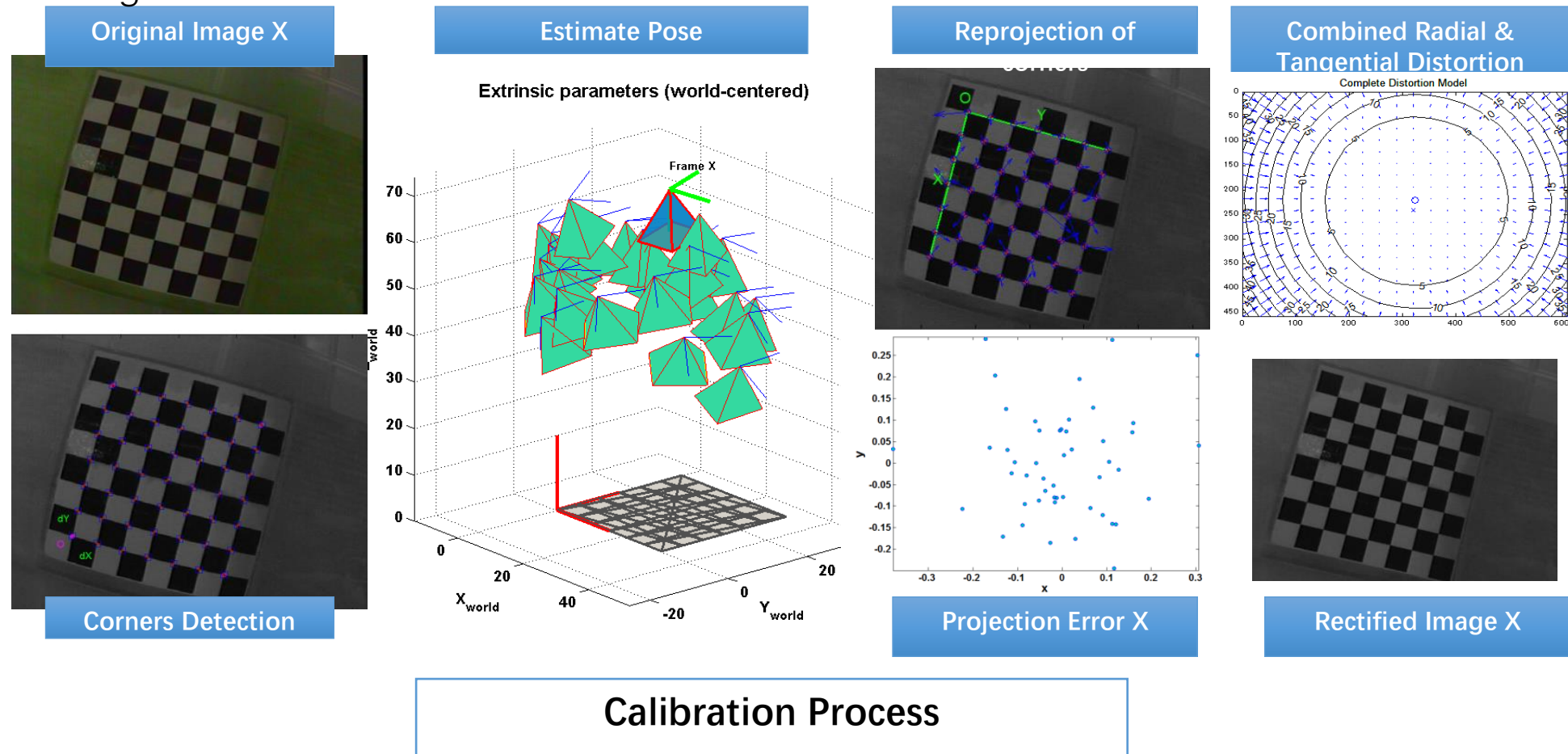
Radial

Tangential

$$\bar{\mathbf{X}}_{\text{distort}} = \underbrace{(1 + k_1 r^2 + k_2 r^4)}_{\text{Radial}} \begin{pmatrix} \bar{x} \\ \bar{y} \end{pmatrix} + \underbrace{(l_1 \quad l_2)}_{\text{Tangential}} \begin{pmatrix} 2\bar{x}\bar{y} & r^2 + 2\bar{y}^2 \\ r^2 + 2\bar{x}^2 & 2\bar{x}\bar{y} \end{pmatrix}$$

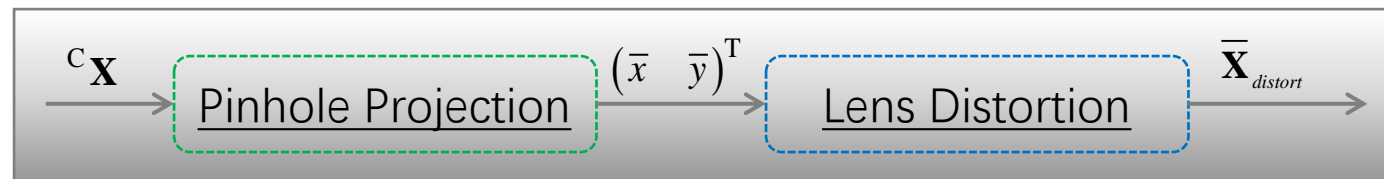
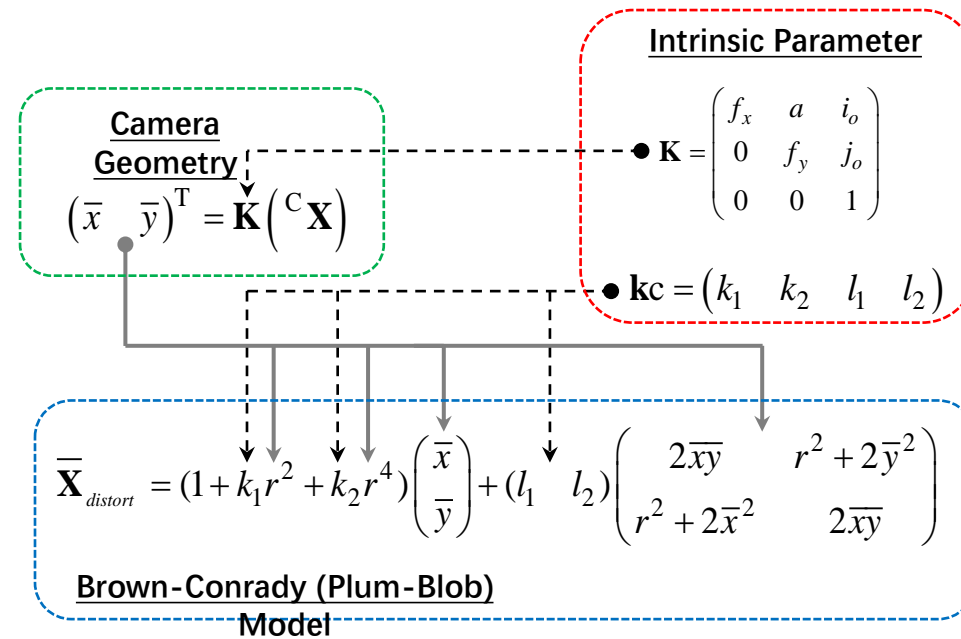
Camera Calibration (Geometric Calibration)

Example of calibration process using Matlab Camera Calibration Toolbox developed by Bouget



Camera Calibration (Geometric Calibration)

- Intrinsic Properties



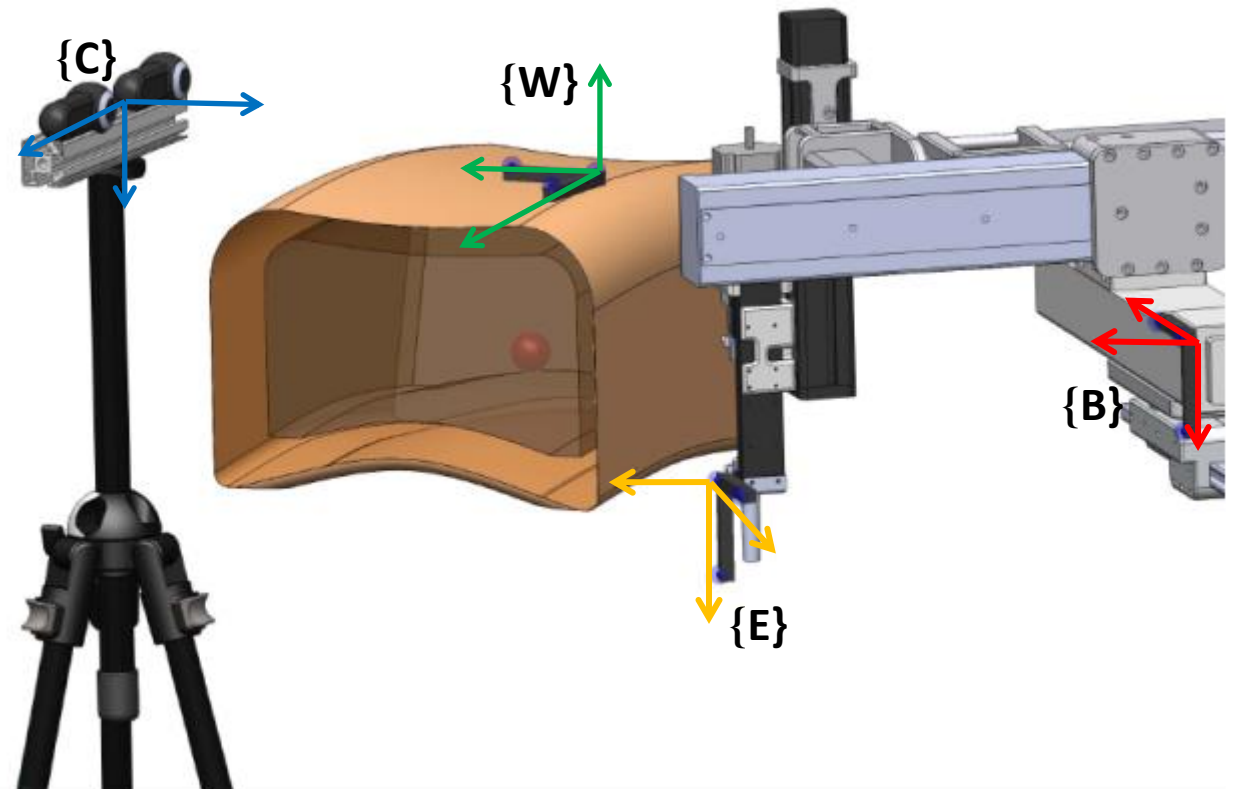
Camera Calibration Summary

- Transformations from world (Cartesian) to image (pixel) coordinates



Case Problem 1

- Given K , ${}^c[R \mid t]_w$, wT_B , BP_E
- Find the image coordinates of point E

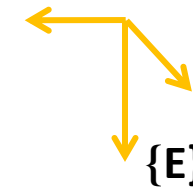
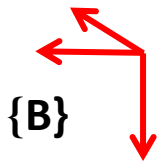
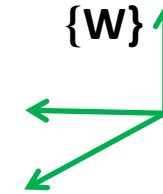
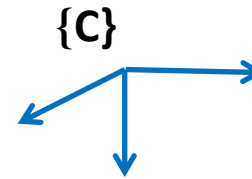


Case Problem 1

Using relationship, $s\mathbf{u} = \mathbf{K}^c [\mathbf{R} \mid \mathbf{t}]_W \mathbf{X}$

$$s\mathbf{u} = \mathbf{K}^c [\mathbf{R} \mid \mathbf{t}]_W {}^W\mathbf{T}_B {}^B\mathbf{P}_E$$
$$s[u \ v \ 1]^T = \mathbf{K}^c [\mathbf{R} \mid \mathbf{t}]_W {}^W\mathbf{T}_B {}^B\mathbf{P}_E$$

There will be 3 rows
using the last row, solve for s
Obtain image coord, \mathbf{u}



After calibration, how do we estimate the pose of the camera?



Robot-Camera Calibration

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Robot-Camera Calibration

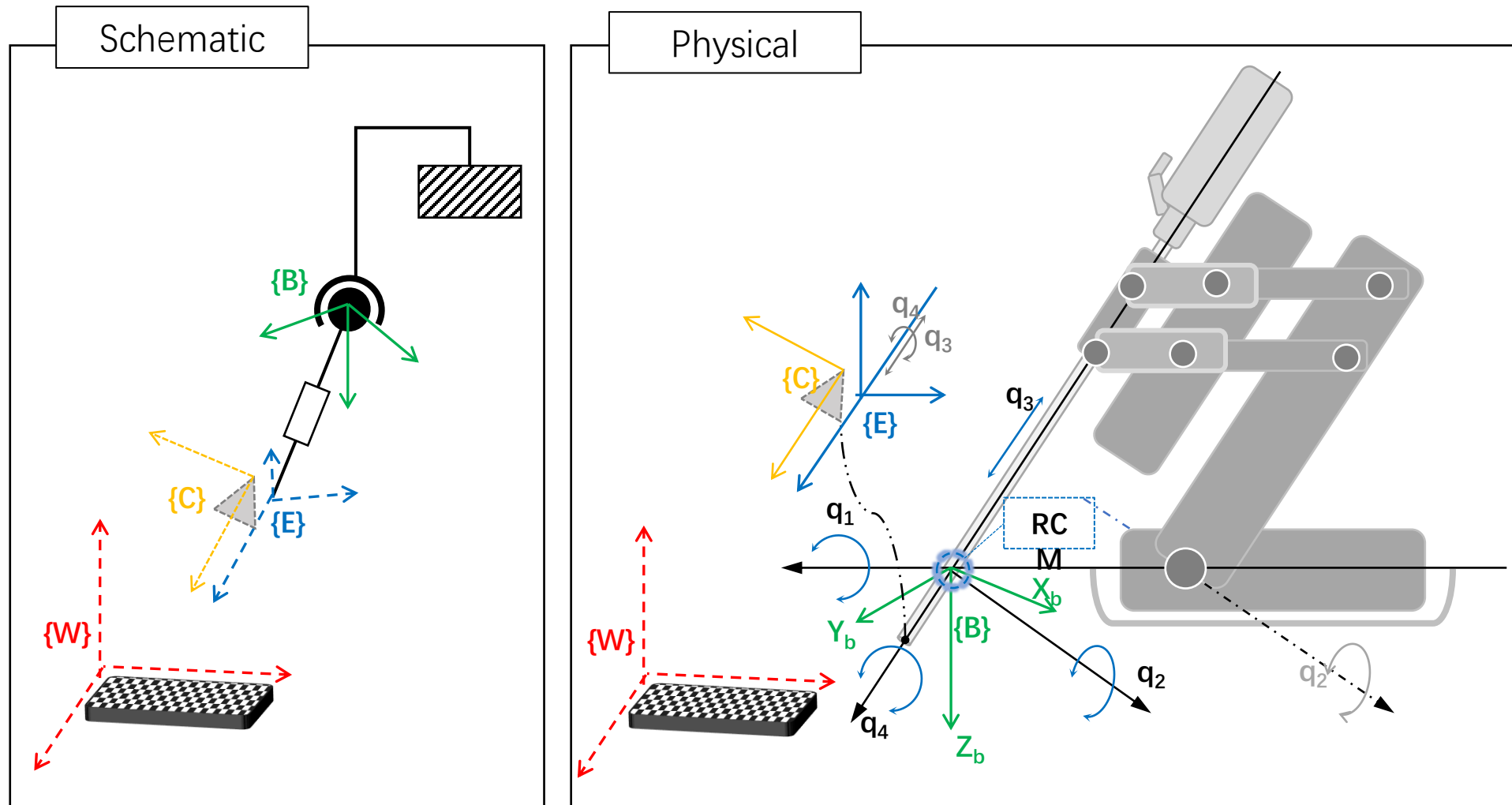
- For robotics, we are interested in **pose of robot** end-effector
- **Camera pose** and camera calibration is not enough
- **Robot-camera calibration** is required

Robot-Camera Calibration

- Also known as Tracker-camera; Hand-eye calibration
- To obtain the end-effector pose w.r.t. the camera, cT_C
- Use camera-acquired data for robot pose estimation



Robot-Camera Calibration: Illustration



Case Problem 2

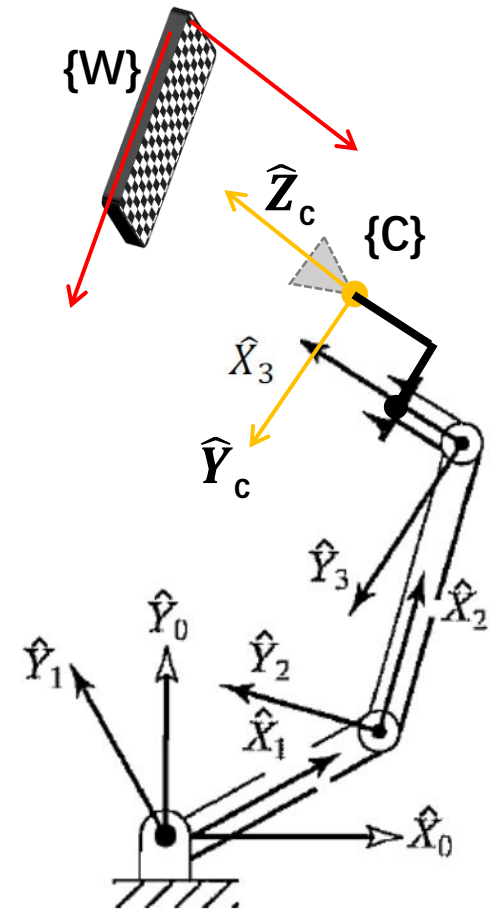
Find the camera matrix M that maps a 3D point (x,y,z) in world ref. frame $\{W\}$ to its projection on image coordinates (u,v) given the following:

${}^W\mathbf{T}_0$: Robot base in world ref. frame

${}^3\mathbf{T}_c$: Camera in Link 3 ref. frame

\mathbf{K} : Intrinsic matrix of camera

$(\theta_1, \theta_2, \theta_3)$: Joint Variables



Case Problem 2

Using relationship, $\mathbf{M} = \mathbf{K} {}^c[\mathbf{R}|\mathbf{t}]_w$
where ${}^c[\mathbf{R}|\mathbf{t}]_w = {}^c\mathbf{T}_w$

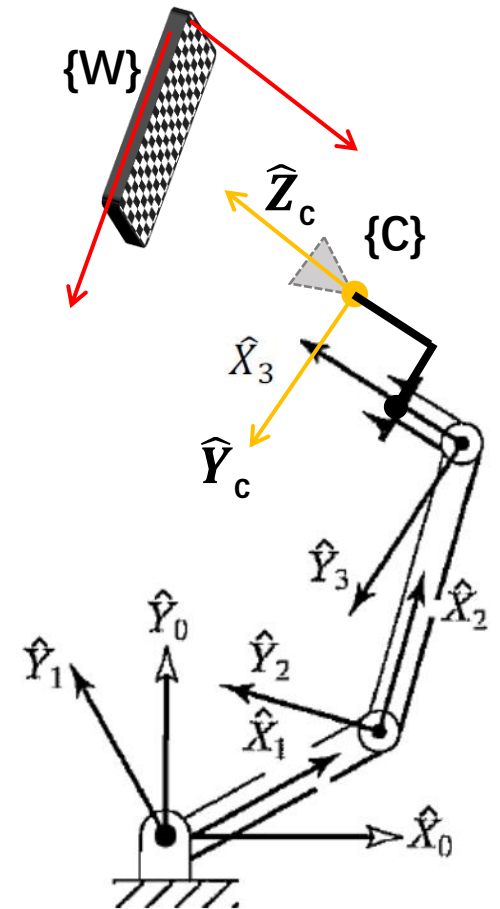
We can express \mathbf{M} as $\mathbf{M} = \mathbf{K} [{}^w\mathbf{T}_0 {}^0\mathbf{T}_3 {}^3\mathbf{T}_c]^{-1}$
by substituting ${}^c\mathbf{T}_w = [{}^w\mathbf{T}_c]^{-1}$
(Qn: what is the physical meaning of ${}^w\mathbf{T}_c$ or ${}^c\mathbf{T}_w$?)

Do forward kinematics,

$${}^0\mathbf{T}_3 = {}^0\mathbf{T}_1 {}^1\mathbf{T}_2 {}^2\mathbf{T}_3$$

$$\mathbf{M} = \mathbf{K} [{}^w\mathbf{T}_0 {}^0\mathbf{T}_1 {}^1\mathbf{T}_2 {}^2\mathbf{T}_3 {}^3\mathbf{T}_c]^{-1}$$

Qn: Can we do camera calibration with this setup?
The robot is planar, you need many views in 3D for calibration



Camera Pose Estimation

- Solve extrinsic parameters
 - Perspective-n-Point Problem
 - Given point correspondence and known \mathbf{K} (calibrated), obtain $[\mathbf{R}|\mathbf{t}]$
 - To localized the camera with 6 dof in 3D space
 - ≥ 3 non-collinear points needed

In Matlab,

Calibration Toolbox: `compute_extrinsic()`

Robotics Toolbox: `estpose`

In OpenCV,

`solvePnP`

**What if there is no known structure/model (like the checkerboard)
in the scene?**

Camera Pose Estimation

- EPnP Algorithm (FYI)

$${}^k\mathbf{z}_l \cdot \begin{pmatrix} {}^k(i_l & j_l & 1)^T \end{pmatrix} = K \begin{pmatrix} R & | & t \end{pmatrix}$$

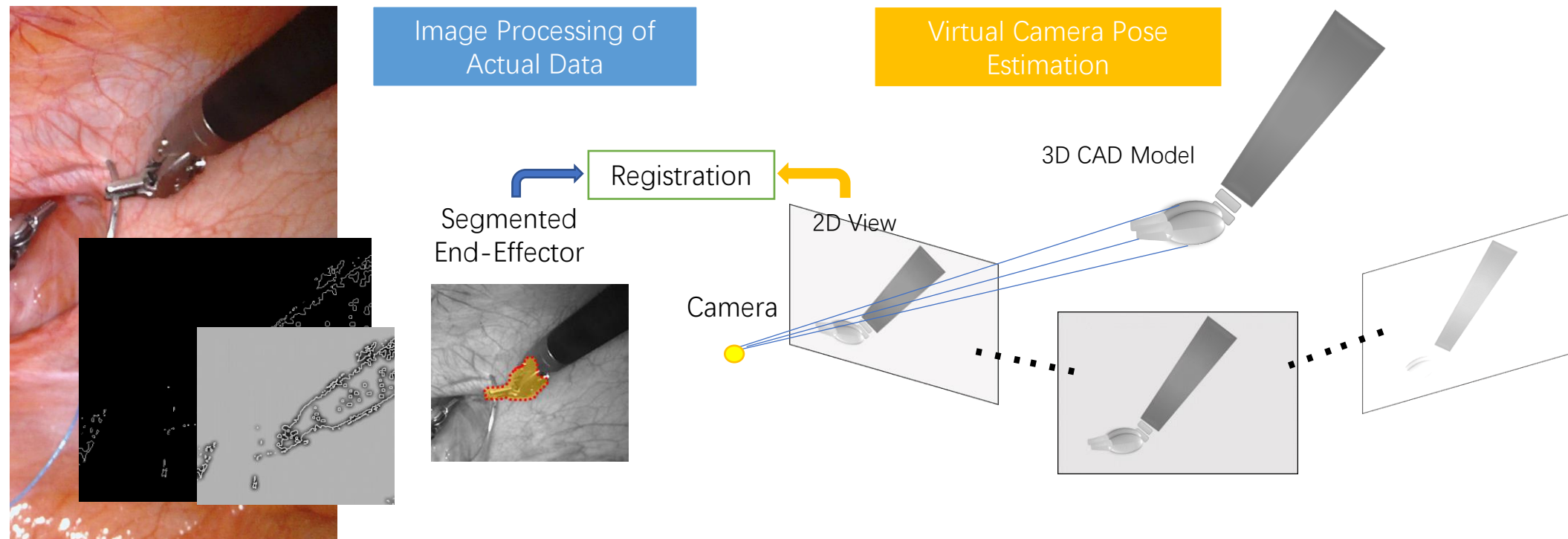
For M virtual control points $\mathbf{q} = \{q_1 \dots q_m \dots q_M\}$,

$${}^k\mathbf{z}_l \cdot \begin{pmatrix} {}^k(i_l & j_l & 1)^T \end{pmatrix} = K \sum_m^M \lambda_{lm}^c q_m$$

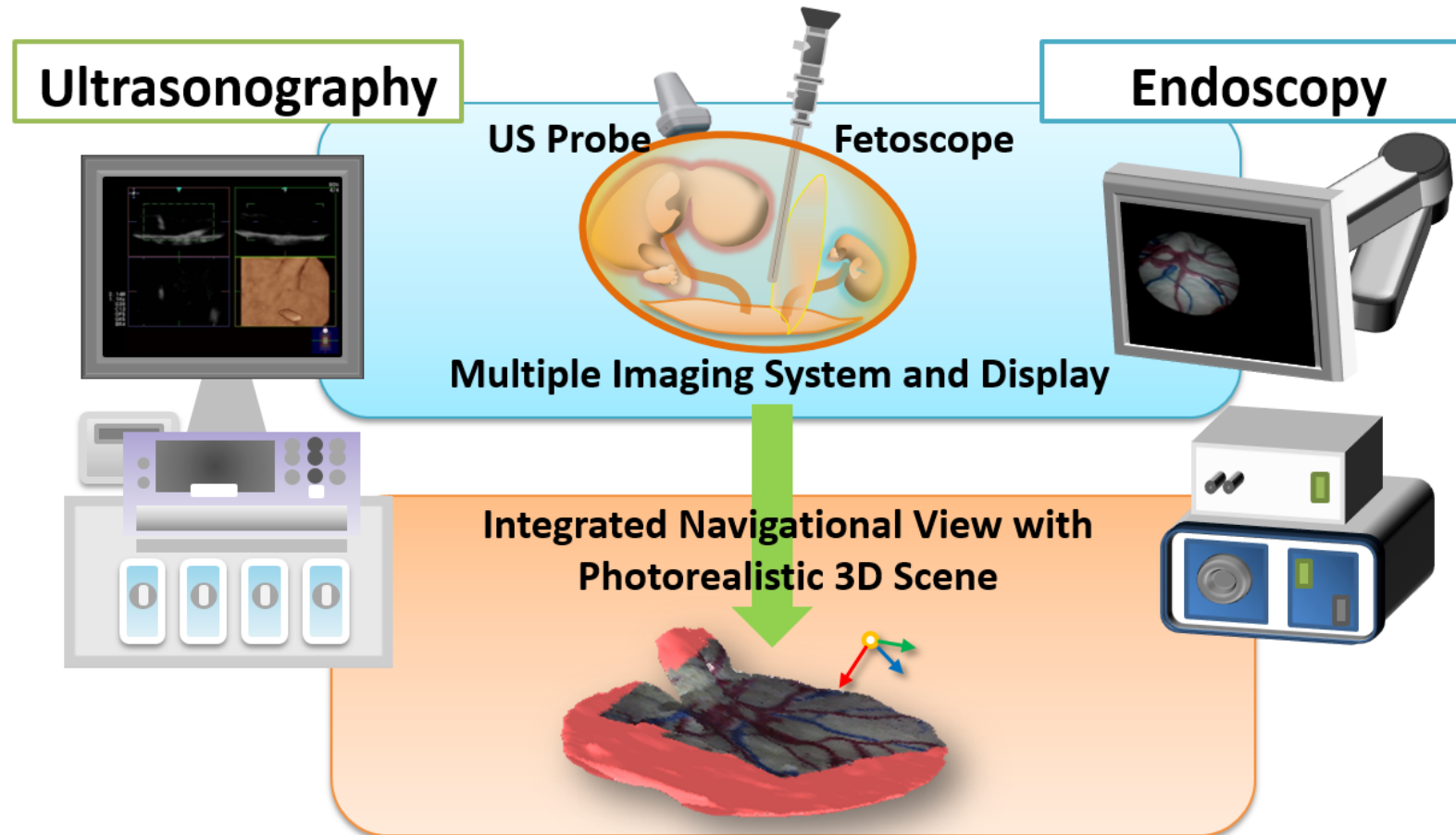
λ_{lm} : homogeneous Barycentric coordinates

Camera Pose Estimation

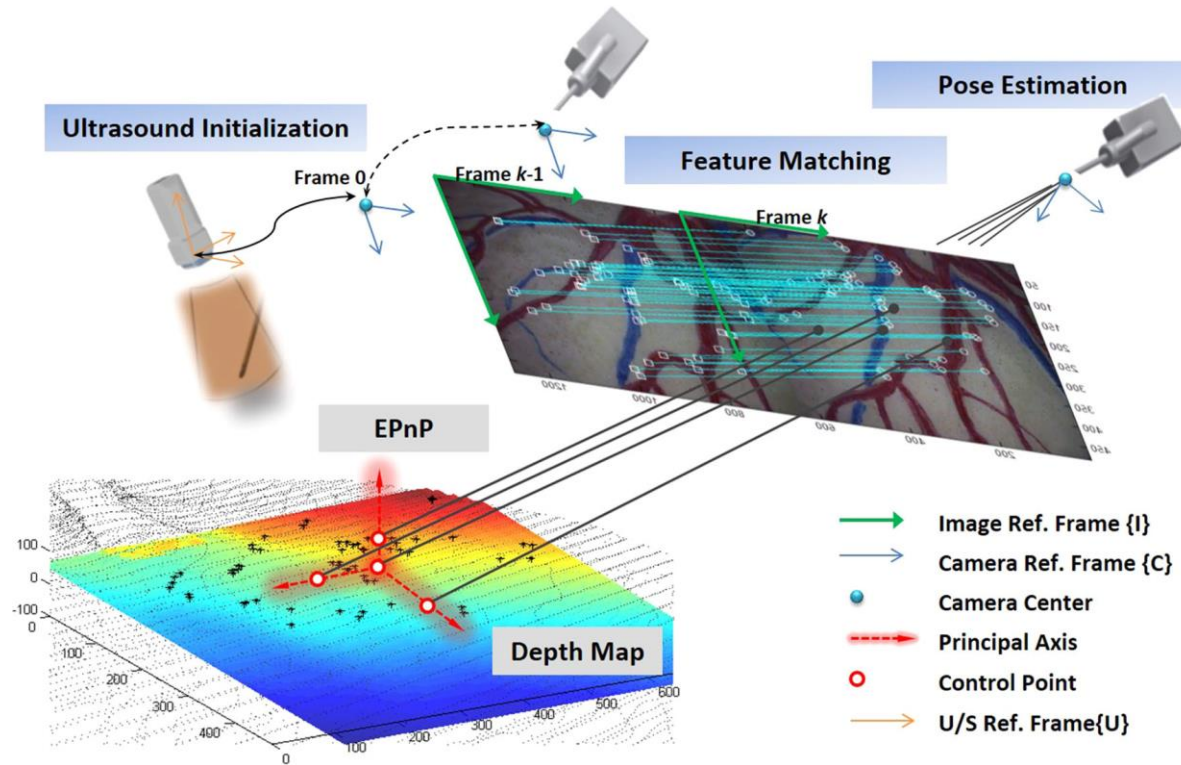
- Solve extrinsic parameters
 - Other model based approach



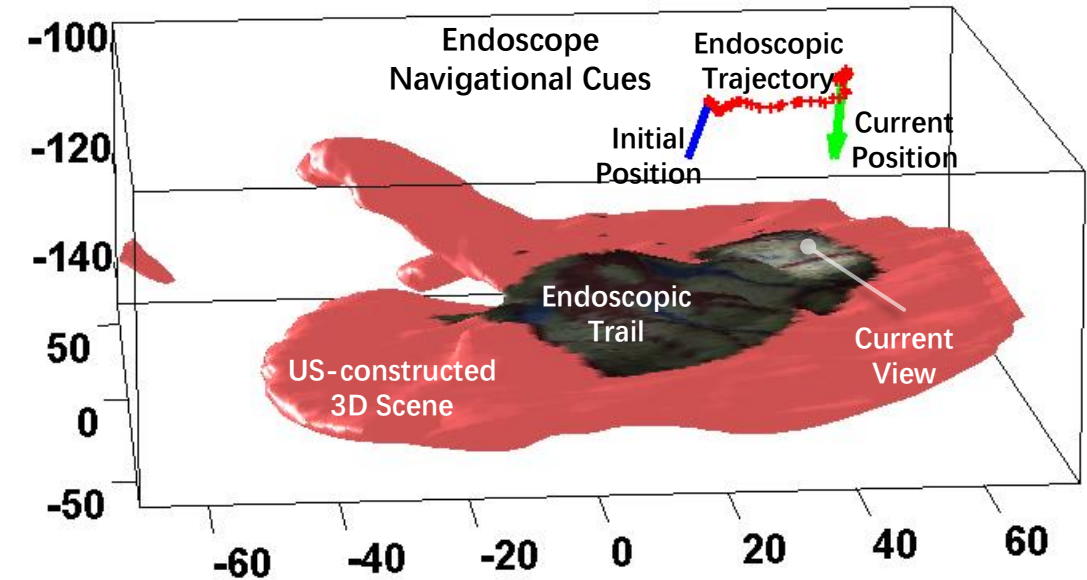
Pose Estimation: Fusion of Data



Pose Estimation: Fusion of Data

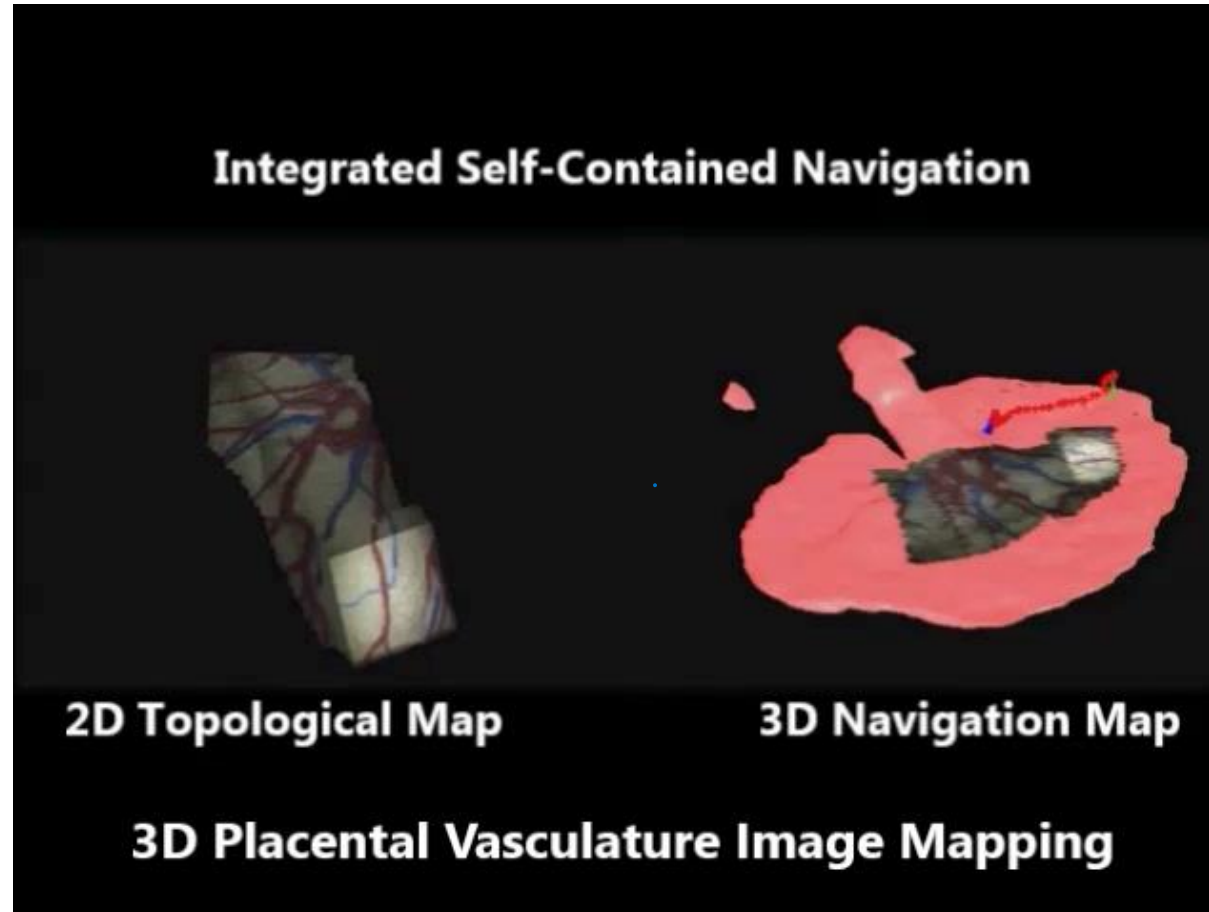


3 D Navigation Map



“Self-contained image mapping of placental vasculature in 3D ultrasound-guided fetoscopy”, Yang et al., 2015

Pose Estimation: Fusion of Data



After calibration, how do we estimate the pose of the camera?
Or the pose of our robot?
How about spatial representation of the surrounding?

Describing the pose of camera is mathematically equivalent to
describing the surrounding spatial information

A Problem of Spatial Representation and 3D Vision



3D Vision

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