

Autonomous Valet Parking (AVP)

Theory and Practice

自主代客泊车理论与实践

Lecture 3: Camera Model



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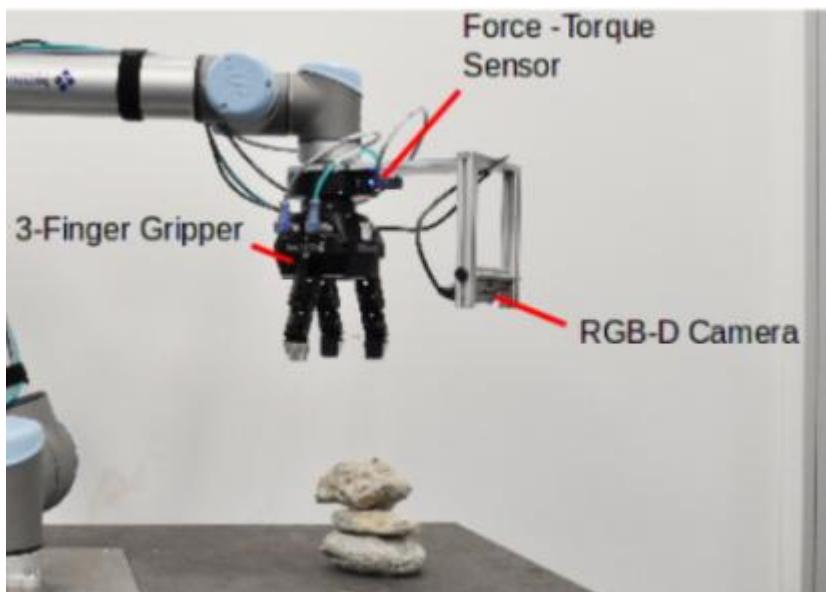
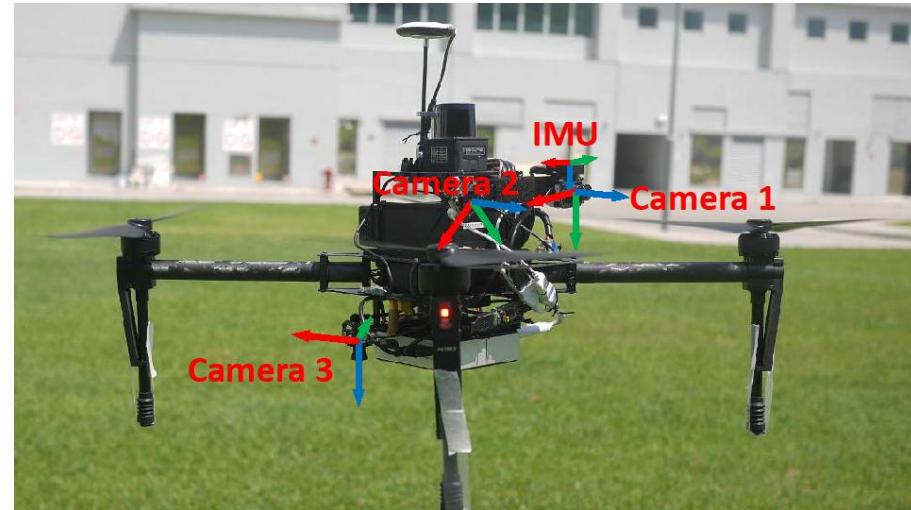




-  1. Camera Background
-  2. Pinhole Camera Model
-  3. Distortion
-  4. Fisheye Camera Model (MEI)
-  5. IPM (Inverse Projective Mapping)
-  6. Assignment



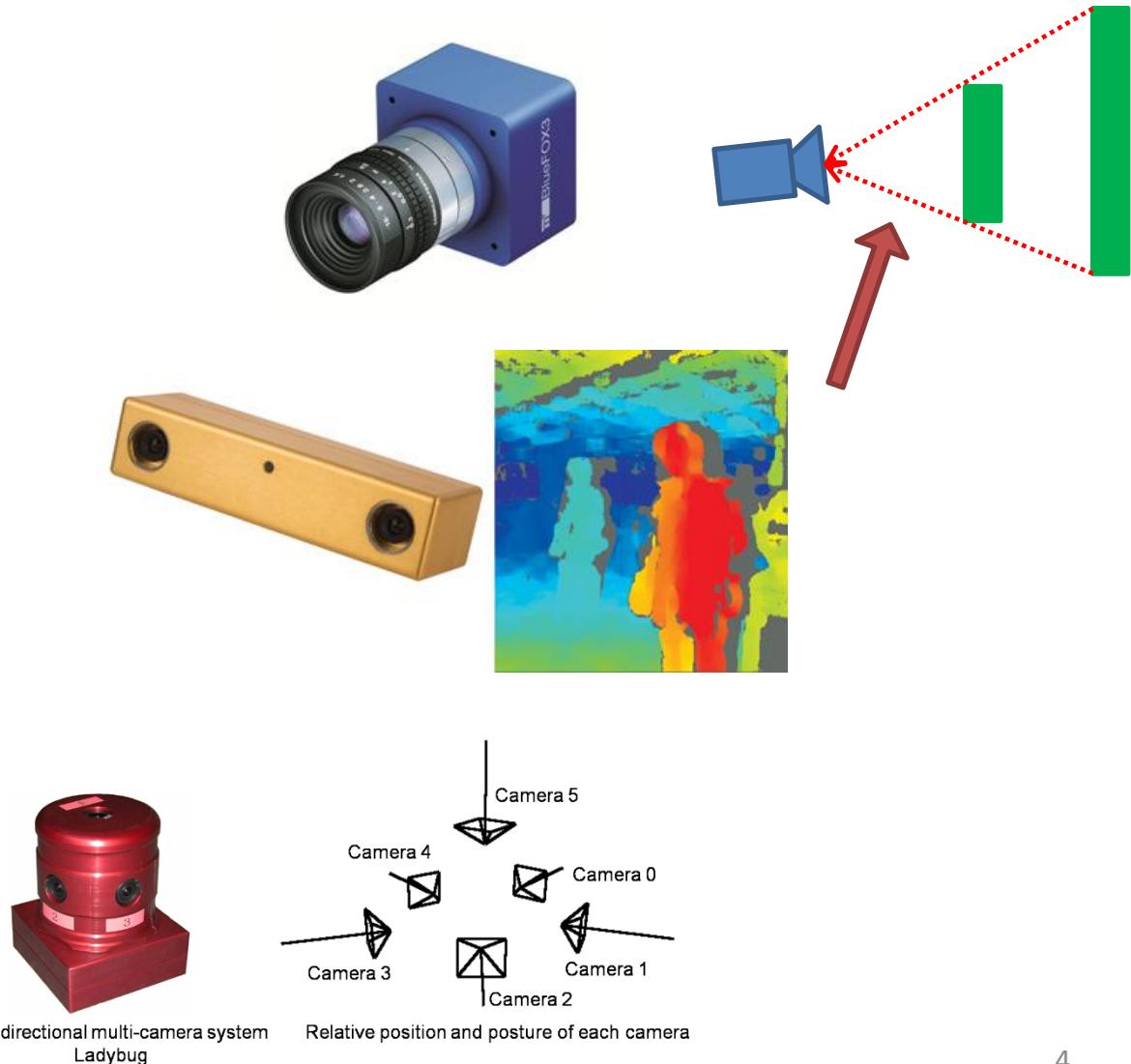
Robot Sees with Cameras





Cameras

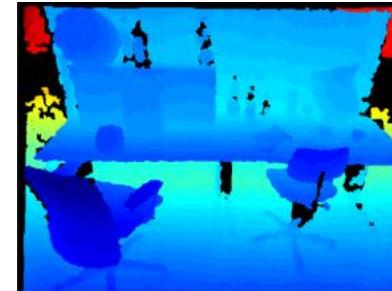
- Monocular
 - Simplest setup
 - Depth unknown
- Stereo
 - Able to compute depth
 - Depth accuracy affected by baseline, resolution, and calibration
- Multi-Camera
 - Overlapping / Non-overlapping field-of-view





Cameras

- RGB-D Sensor
 - Great depth in limited range
 - Does not work outdoors



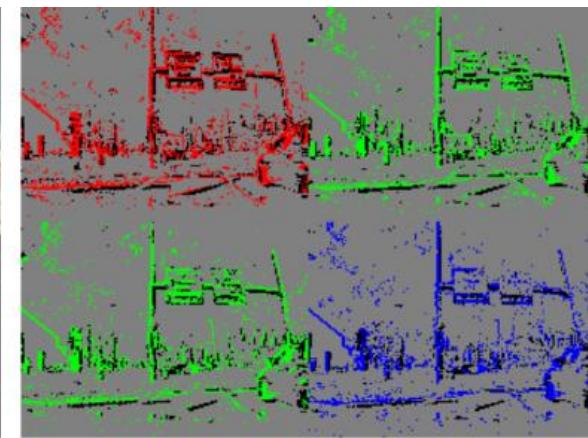
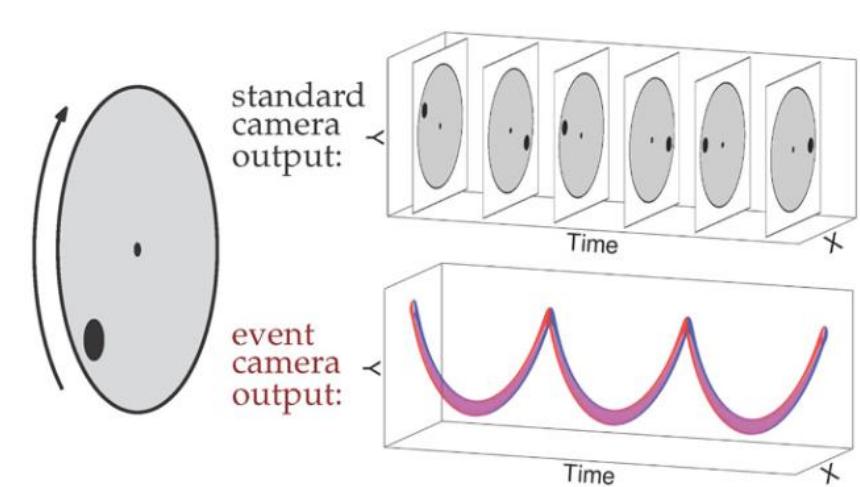
- Omnidirectional camera
 - 360 capture
 - Strong distortion





Cameras

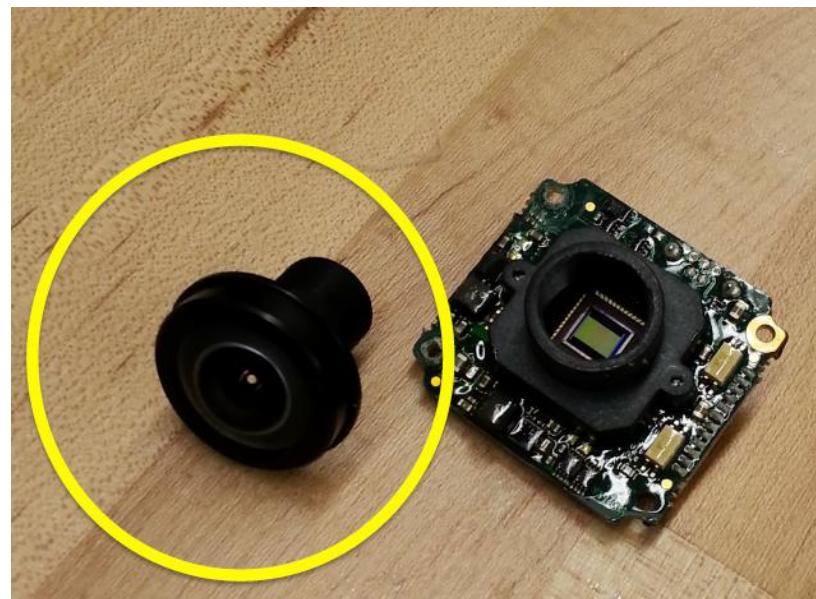
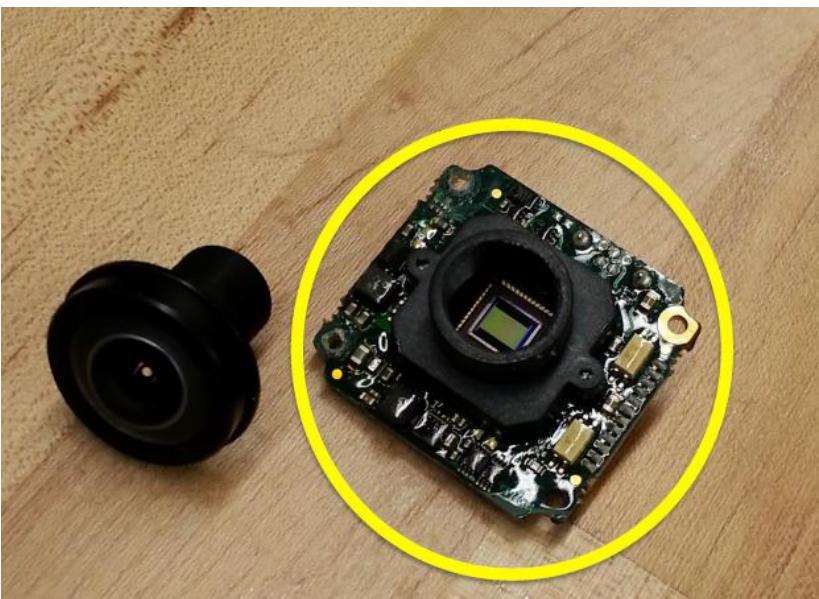
- Event Camera
 - High Temporal Resolution
 - Lack of Rich Visual Information





Cameras

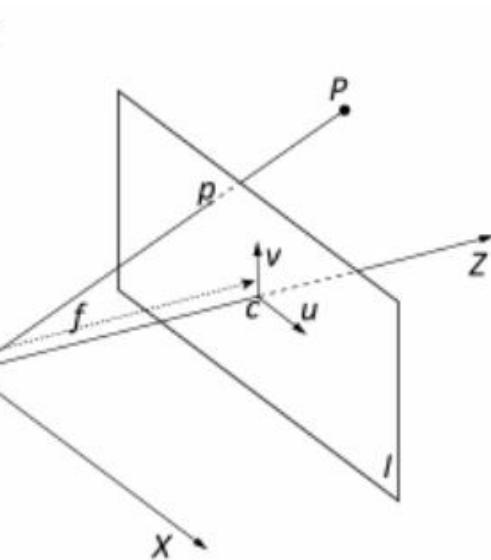
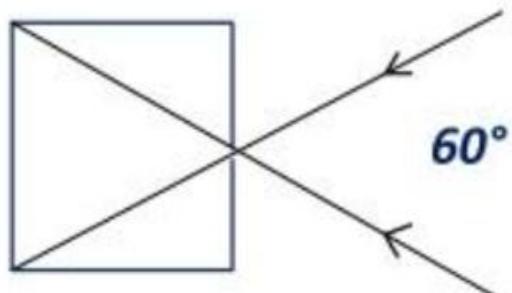
- Sensor
- Lens





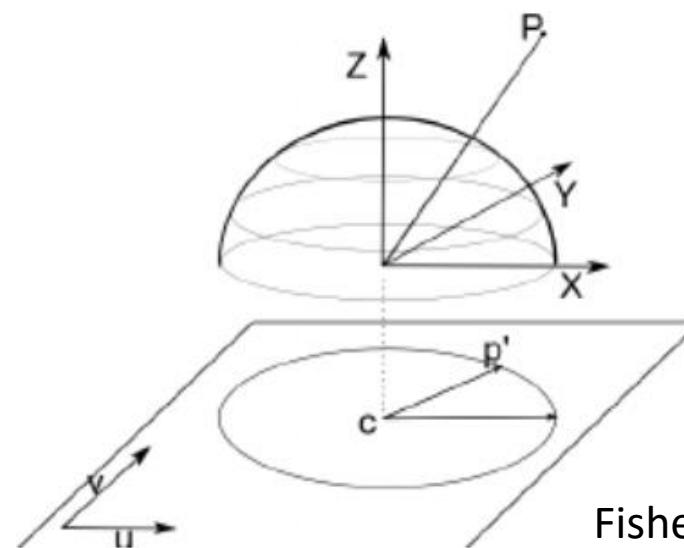
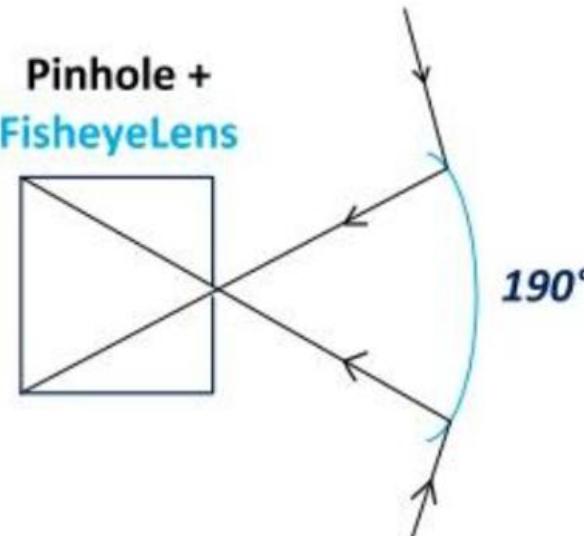
Cameras

Pinhole



Pinhole model

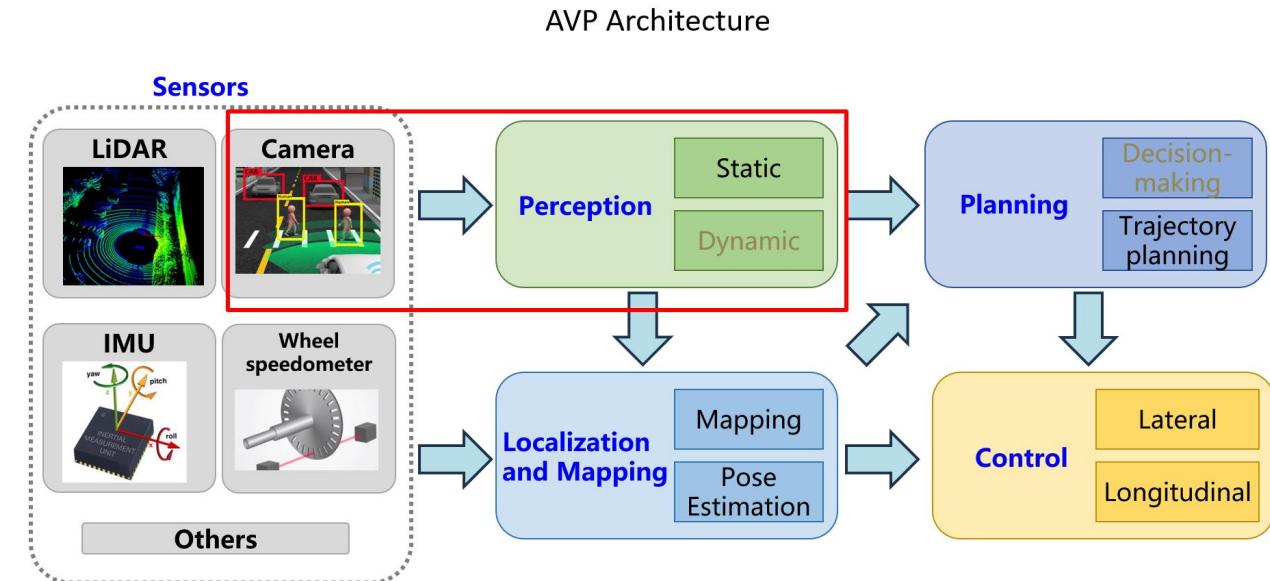
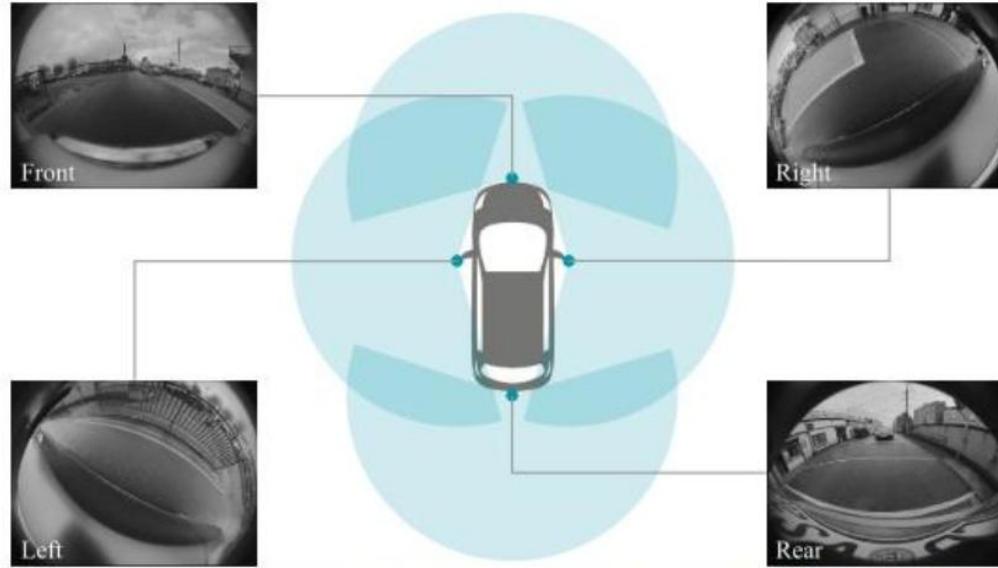
**Pinhole +
FisheyeLens**



Fisheye model Fisheye model 8



Cameras for AVP



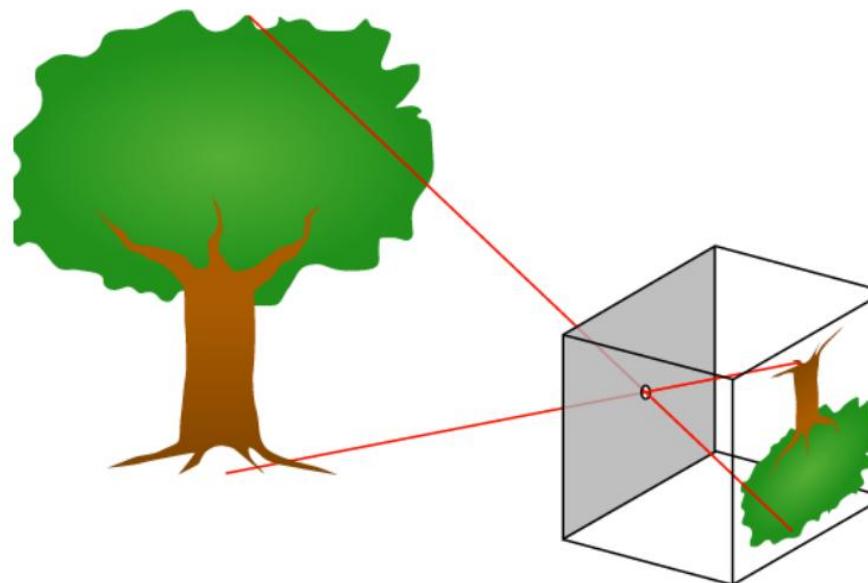


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Pin-hole Camera Model

Pinhole imaging



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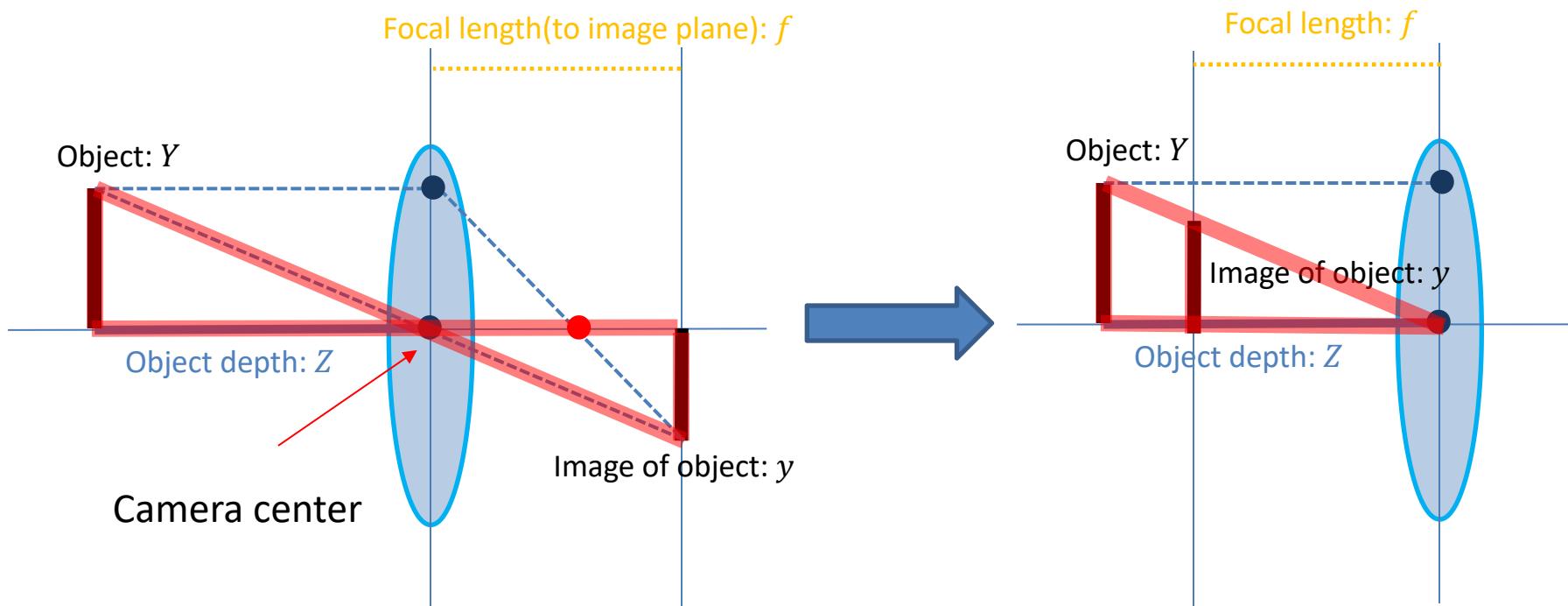
Light from an object passes through a small hole (the pinhole) and projects an inverted image onto a surface or screen on the other side.



Pin-hole Camera Model

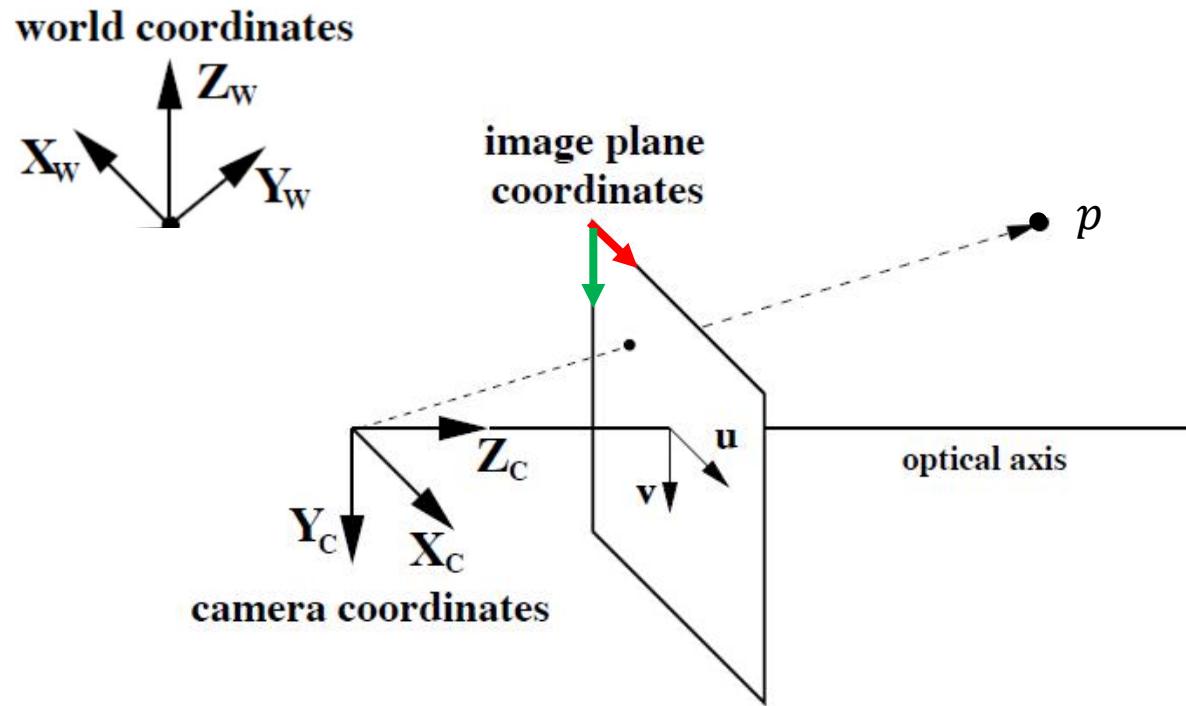
- Assume that the image plane is in front of the lens:

$$-y = f \frac{Y}{Z}$$





Camera Coordinate System



Coordinates:

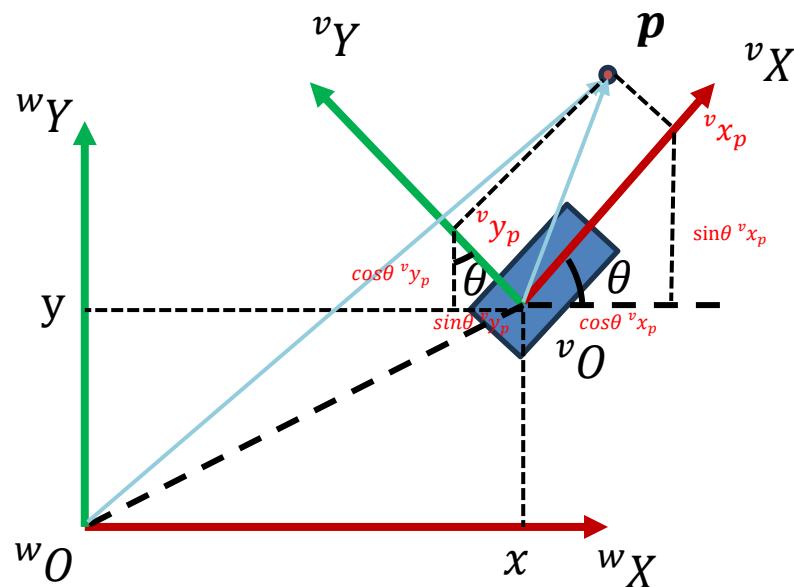
- World ${}^w p = [x_w, y_w, z_w]$
- Camera ${}^c p = [x_c, y_c, z_c]$
- Image plane ${}^I p = [u, v]$

What's the relationship between ${}^w p$, ${}^c p$, ${}^I p$?



3D Coordinate Transformation

- Recall: 2D Transformations
 - World (Global) Coordinate
 - Vehicle (Local) Coordinate



Vehicle 2D State: $s = [x, y, \theta]$

$${}^w\mathbf{p} = \underbrace{\begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}}_{{}^wR_v} {}^v\mathbf{p} + \underbrace{\begin{bmatrix} x \\ y \end{bmatrix}}_{{}^w\mathbf{t}_v}$$

$${}^v\mathbf{p} = \underbrace{\begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}^T}_{{}^vR_w} {}^w\mathbf{p} - \underbrace{\begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}^T \begin{bmatrix} x \\ y \end{bmatrix}}_{{}^v\mathbf{t}_w}$$

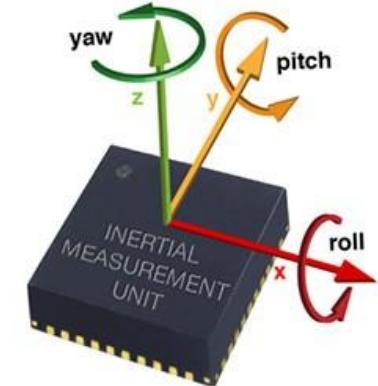
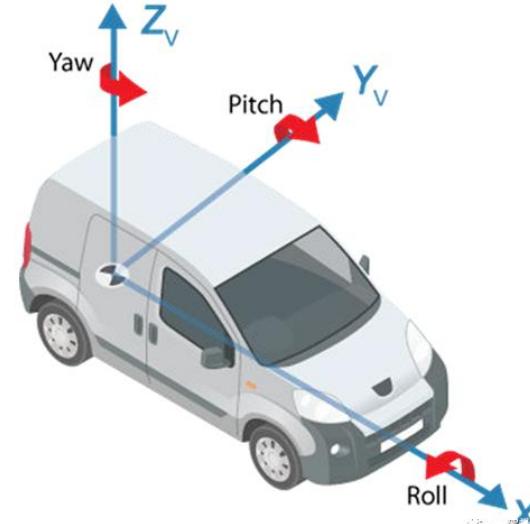
${}^wR_v, {}^w\mathbf{t}_v$: rotation matrix and translation vector from **vehicle frame to world frame**
 ${}^vR_w, {}^v\mathbf{t}_w$: rotation matrix and translation vector from **world frame to vehicle frame**



3D Coordinate Transformation

Rotation Presentation

- Euler Angle
 - Roll(横滚) / Pitch(俯仰) / Yaw(偏航)



- Rotation Matrix

$$R = \begin{bmatrix} a_{00} & a_{01} & a_{02} \\ a_{10} & a_{11} & a_{12} \\ a_{20} & a_{21} & a_{22} \end{bmatrix}_{3 \times 3}$$

- Quaternion

$$\mathbf{q} = [w, x, y, z]^\top$$

Quaternion -> R

$$\mathbf{R} = \begin{bmatrix} q_w^2 + q_x^2 - q_y^2 - q_z^2 & 2(q_x q_y - q_w q_z) & 2(q_x q_z + q_w q_y) \\ 2(q_x q_y + q_w q_z) & q_w^2 - q_x^2 + q_y^2 - q_z^2 & 2(q_y q_z - q_w q_x) \\ 2(q_x q_z - q_w q_y) & 2(q_y q_z + q_w q_x) & q_w^2 - q_x^2 - q_y^2 + q_z^2 \end{bmatrix}$$



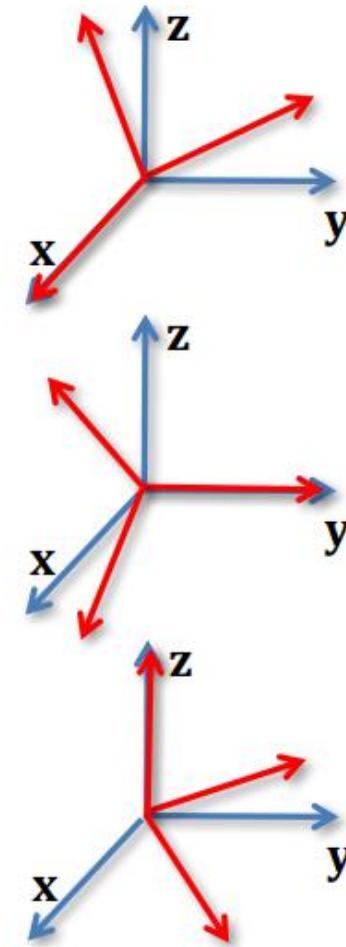
3D Coordinate Transformation

Correspondence: Rotation Matrix - Euler Angles

$$- \mathbf{R}_x(\phi) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \phi & -\sin \phi \\ 0 & \sin \phi & \cos \phi \end{bmatrix}$$

$$- \mathbf{R}_y(\theta) = \begin{bmatrix} \cos \theta & 0 & \sin \theta \\ 0 & 1 & 0 \\ -\sin \theta & 0 & \cos \theta \end{bmatrix}$$

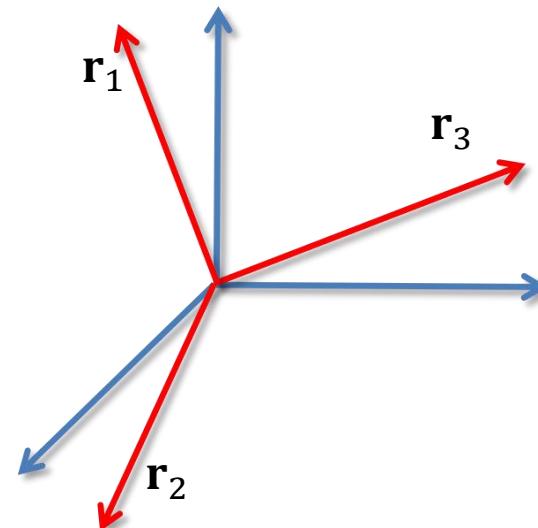
$$- \mathbf{R}_z(\psi) = \begin{bmatrix} \cos \psi & -\sin \psi & 0 \\ \sin \psi & \cos \psi & 0 \\ 0 & 0 & 1 \end{bmatrix}$$





3D Coordinate Transformation

- Let $\mathbf{R} = [\mathbf{r}_1, \mathbf{r}_2, \mathbf{r}_3]$ be a rotation matrix
- Orthogonal:
 - $\mathbf{r}_i^T \cdot \mathbf{r}_j = \begin{cases} 0 & \text{if } i \neq j \\ 1 & \text{if } i = j \end{cases}$
 - $\mathbf{R} \cdot \mathbf{R}^T = \mathbf{I}$
- Special orthogonal:
 - $\det \mathbf{R} = \mathbf{r}_1^T \cdot (\mathbf{r}_2 \times \mathbf{r}_3) = \mathbf{r}_1^T \cdot \mathbf{r}_1 = 1$
- The set of all rotations forms the Special Orthogonal Group
 - Special orthogonal group
 - 3D rotations: $SO(3)$
 - 2D rotations: $SO(2)$
 - $SO(n) = \{\mathbf{R} \in \mathbb{R}^{n \times n} | \mathbf{R} \cdot \mathbf{R}^T = \mathbf{I}, \det \mathbf{R} = 1\}$





3D Coordinate Transformation

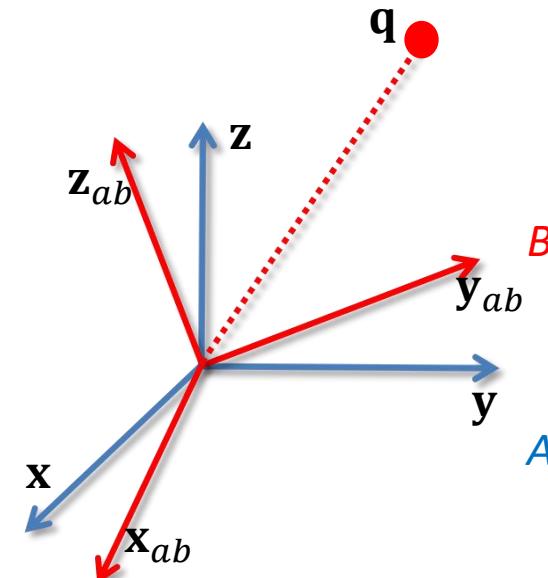
- A transformation that rotates the coordinates of a point from frame B to frame A

- Let $\mathbf{q}_b = [x_b, y_b, z_b]^T \in \mathbb{R}^3$ be coordinate of point \mathbf{q} in frame B

- Let \mathbf{q}_a be coordinate of point \mathbf{q} in frame A

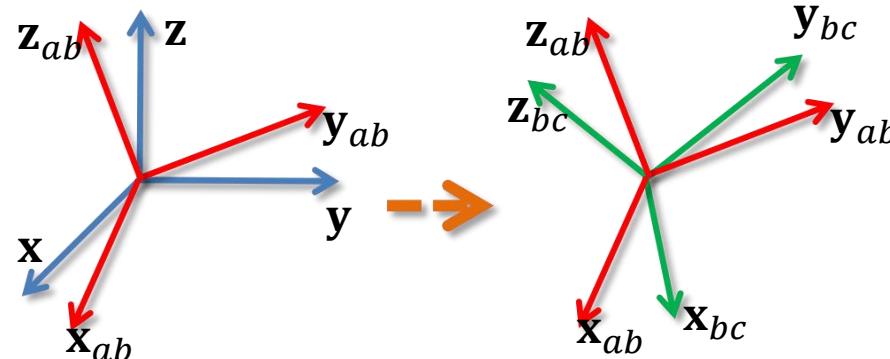
- $$\begin{aligned}\mathbf{q}_a &= x_b \cdot \mathbf{x}_{ab} + y_b \cdot \mathbf{y}_{ab} + z_b \cdot \mathbf{z}_{ab} \\ &= [\mathbf{x}_{ab}, \mathbf{y}_{ab}, \mathbf{z}_{ab}] \begin{bmatrix} x_b \\ y_b \\ z_b \end{bmatrix} = {}^a\mathbf{R}_b \cdot \mathbf{q}_b\end{aligned}$$

Rotation matrix from B to A



- Composition Rule

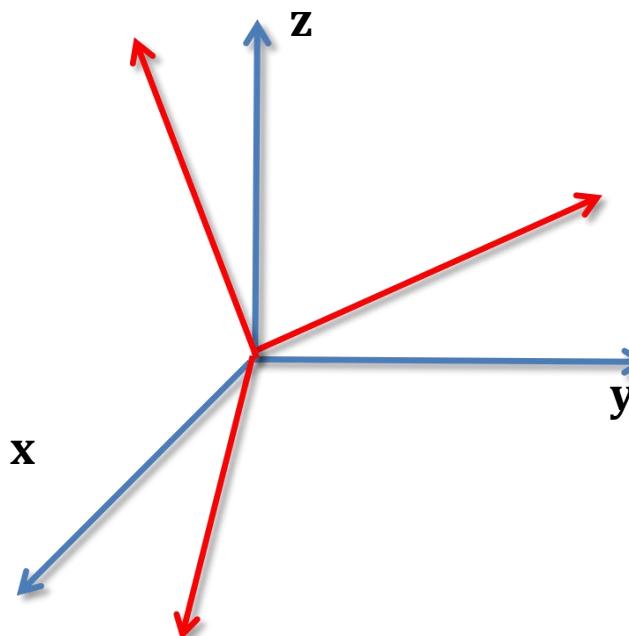
- $${}^a\mathbf{R}_c = {}^a\mathbf{R}_b \cdot {}^b\mathbf{R}_c$$





3D Coordinate Transformation

- Any rotation can be described by three successive rotations about linear independent axes
- However, this is an almost 1-1 transform with singularities:
 - $R_z(\psi) \cdot R_x(\phi) \cdot R_y(\theta) \Rightarrow R$
 - $R_z(\psi) \cdot R_x(\phi) \cdot R_y(\theta) \not\Leftarrow R$





3D Coordinate Transformation

- Different Euler angle conversions:

Proper Euler angles	Tait-Bryan angles
$X_1Z_2X_3 = \begin{bmatrix} c_2 & -c_3s_2 & s_2s_3 \\ c_1s_2 & c_1c_2c_3 - s_1s_3 & -c_3s_1 - c_1c_2s_3 \\ s_1s_2 & c_1s_3 + c_2c_3s_1 & c_1c_3 - c_2s_1s_3 \end{bmatrix}$	$X_1Z_2Y_3 = \begin{bmatrix} c_2c_3 & -s_2 & c_2s_3 \\ s_1s_3 + c_1c_3s_2 & c_1c_2 & c_1s_2s_3 - c_3s_1 \\ c_3s_1s_2 - c_1s_3 & c_2s_1 & c_1c_3 + s_1s_2s_3 \end{bmatrix}$
$X_1Y_2X_3 = \begin{bmatrix} c_2 & s_2s_3 & c_3s_2 \\ s_1s_2 & c_1c_3 - c_2s_1s_3 & -c_1s_3 - c_2c_3s_1 \\ -c_1s_2 & c_3s_1 + c_1c_2s_3 & c_1c_2c_3 - s_1s_3 \end{bmatrix}$	$X_1Y_2Z_3 = \begin{bmatrix} c_2c_3 & -c_2s_3 & s_2 \\ c_1s_3 + c_3s_1s_2 & c_1c_3 - s_1s_2s_3 & -c_2s_1 \\ s_1s_3 - c_1c_3s_2 & c_3s_1 + c_1s_2s_3 & c_1c_2 \end{bmatrix}$
$Y_1X_2Y_3 = \begin{bmatrix} c_1c_3 - c_2s_1s_3 & s_1s_2 & c_1s_3 + c_2c_3s_1 \\ s_2s_3 & c_2 & -c_3s_2 \\ -c_3s_1 - c_1c_2s_3 & c_1s_2 & c_1c_2c_3 - s_1s_3 \end{bmatrix}$	$Y_1X_2Z_3 = \begin{bmatrix} c_1c_3 + s_1s_2s_3 & c_3s_1s_2 - c_1s_3 & c_2s_1 \\ c_2s_3 & c_2c_3 & -s_2 \\ c_1s_2s_3 - c_3s_1 & c_1c_3s_2 + s_1s_3 & c_1c_2 \end{bmatrix}$
$Y_1Z_2Y_3 = \begin{bmatrix} c_1c_2c_3 - s_1s_3 & -c_1s_2 & c_3s_1 + c_1c_2s_3 \\ c_3s_2 & c_2 & s_2s_3 \\ -c_1s_3 - c_2c_3s_1 & s_1s_2 & c_1c_3 - c_2s_1s_3 \end{bmatrix}$	$Y_1Z_2X_3 = \begin{bmatrix} c_1c_2 & s_1s_3 - c_1c_3s_2 & c_3s_1 + c_1s_2s_3 \\ s_2 & c_2c_3 & -c_2s_3 \\ -c_2s_1 & c_1s_3 + c_3s_1s_2 & c_1c_3 - s_1s_2s_3 \end{bmatrix}$
$Z_1Y_2Z_3 = \begin{bmatrix} c_1c_2c_3 - s_1s_3 & -c_3s_1 - c_1c_2s_3 & c_1s_2 \\ c_1s_3 + c_2c_3s_1 & c_1c_3 - c_2s_1s_3 & s_1s_2 \\ -c_3s_2 & s_2s_3 & c_2 \end{bmatrix}$	$Z_1Y_2X_3 = \begin{bmatrix} c_1c_2 & c_1s_2s_3 - c_3s_1 & s_1s_3 + c_1c_3s_2 \\ c_2s_1 & c_1c_3 + s_1s_2s_3 & c_3s_1s_2 - c_1s_3 \\ -s_2 & c_2s_3 & c_2c_3 \end{bmatrix}$
$Z_1X_2Z_3 = \begin{bmatrix} c_1c_3 - c_2s_1s_3 & -c_1s_3 - c_2c_3s_1 & s_1s_2 \\ c_3s_1 + c_1c_2s_3 & c_1c_2c_3 - s_1s_3 & -c_1s_2 \\ s_2s_3 & c_3s_2 & c_2 \end{bmatrix}$	$Z_1X_2Y_3 = \begin{bmatrix} c_1c_3 - s_1s_2s_3 & -c_2s_1 & c_1s_3 + c_3s_1s_2 \\ c_3s_1 + c_1s_2s_3 & c_1c_2 & s_1s_3 - c_1c_3s_2 \\ -c_2s_3 & s_2 & c_2c_3 \end{bmatrix}$

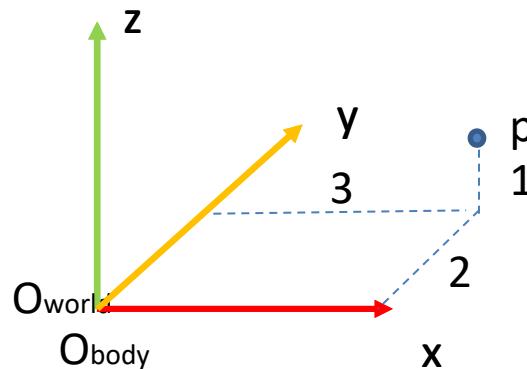


3D Coordinate Transformation

- World frame (fixed)
- Body frame (moving)

P is a fixed point in body frame.

$${}^b p = \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$$



The two coordinate systems coincide.

$${}^w p = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} {}^b p + \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix} \quad {}^w p = \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$$

${}^w R_b$ ${}^w t_b$

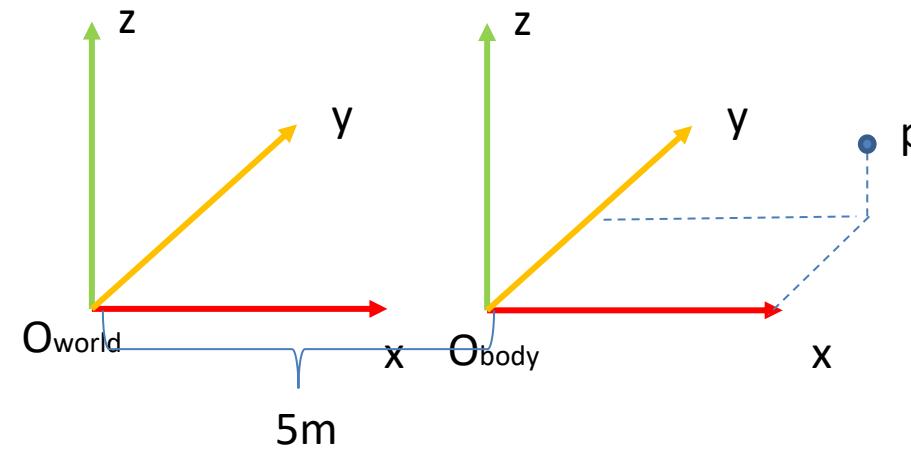


3D Coordinate Transformation

- World frame (fixed)
- Body frame (moving)

P is a fixed point in body frame.

$${}^b p = \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$$



$${}^w p = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} {}^b p + \begin{bmatrix} 5 \\ 0 \\ 0 \end{bmatrix}$$

$${}^w R_b$$

$${}^w t_b$$

$${}^w p = \begin{bmatrix} 8 \\ 2 \\ 1 \end{bmatrix}$$

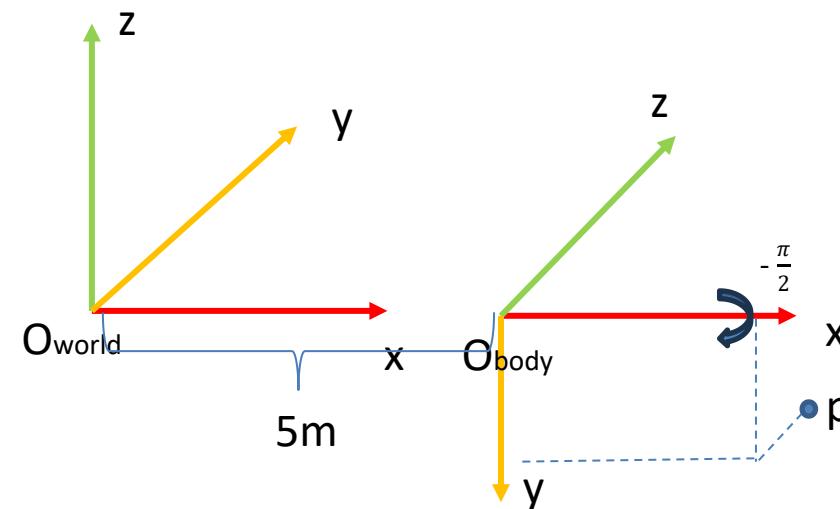


3D Coordinate Transformation

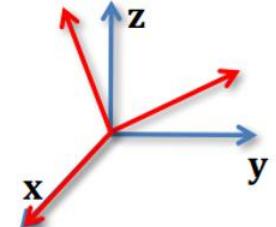
- World frame (fixed)
- Body frame (moving)

P is a fixed point in body frame.

$${}^b p = \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$$



$$R_x(\phi) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \cos \phi & -\sin \phi \\ 0 & \sin \phi & \cos \phi \end{bmatrix}$$



$${}^w p = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -1 & 0 \end{bmatrix} {}^b p + \begin{bmatrix} 5 \\ 0 \\ 0 \end{bmatrix}$$

$${}^w p = \begin{bmatrix} 8 \\ 1 \\ -2 \end{bmatrix}$$

$${}^w R_b$$

$${}^w t_b$$



3D Coordinate Transformation

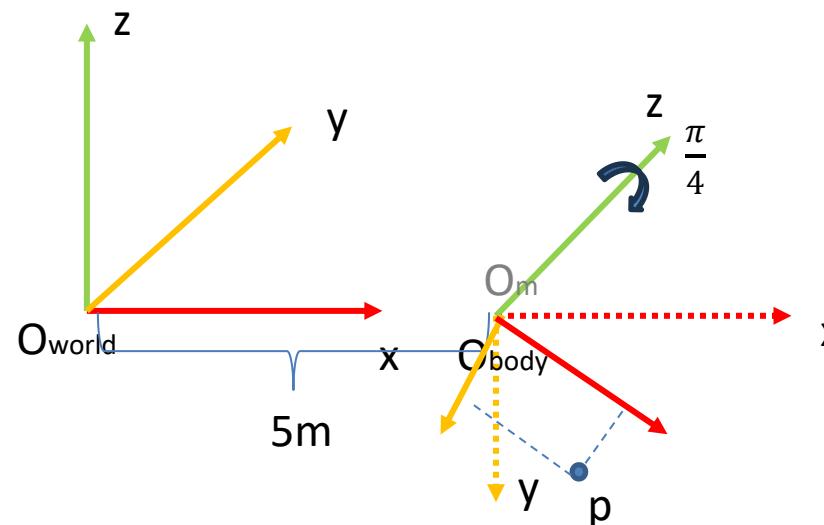
- World frame (fixed)
- Body frame (moving)

P is a fixed point in body frame.

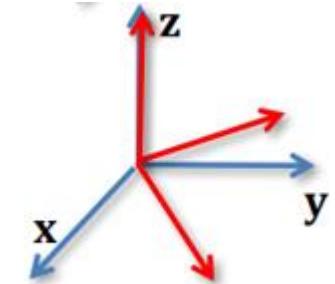
$${}^b p = \begin{bmatrix} 3 \\ 2 \\ 1 \end{bmatrix}$$

$${}^{b_m} p = \begin{bmatrix} \frac{\sqrt{2}}{2} & -\frac{\sqrt{2}}{2} & 0 \\ \frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} & 0 \\ \frac{1}{2} & 0 & 1 \end{bmatrix} {}^b p + \begin{bmatrix} 0 \\ 0 \\ 0 \end{bmatrix}$$

$${}^w p = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -1 & 0 \end{bmatrix} {}^{b_m} p + \begin{bmatrix} 5 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 5 + \frac{\sqrt{2}}{2} \\ 1 \\ -\frac{5\sqrt{2}}{2} \end{bmatrix}$$



$$R_z(\psi) = \begin{bmatrix} \cos \psi & -\sin \psi & 0 \\ \sin \psi & \cos \psi & 0 \\ 0 & 0 & 1 \end{bmatrix}$$



$$\begin{bmatrix} {}^w p \\ 1 \end{bmatrix} = \begin{bmatrix} {}^w R_{b_m} & {}^w t_{b_m} \\ 0 & 1 \end{bmatrix} \begin{bmatrix} {}^{b_m} R_b & {}^{b_m} t_b \\ 0 & 1 \end{bmatrix} \begin{bmatrix} {}^b p \\ 1 \end{bmatrix}$$

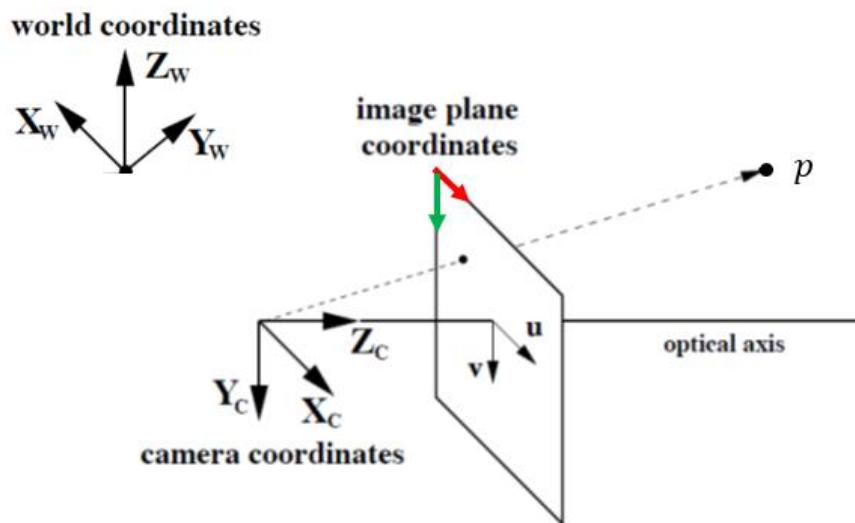


Perspective Projection

- Optical axis is the z-axis
- The image plane (u, v) is perpendicular to the optical axis
- Intersection of the image plane and the optical axis is the image center (u_0, v_0)
- f is the distance of the image plane from the origin (in pixels)
- Point in Camera Coordinate: ${}^c p = [x_c, y_c, z_c]$
- Formulation:

$$- u = \frac{fx_c}{z_c} + u_0$$

$$- v = \frac{fy_c}{z_c} + v_0$$

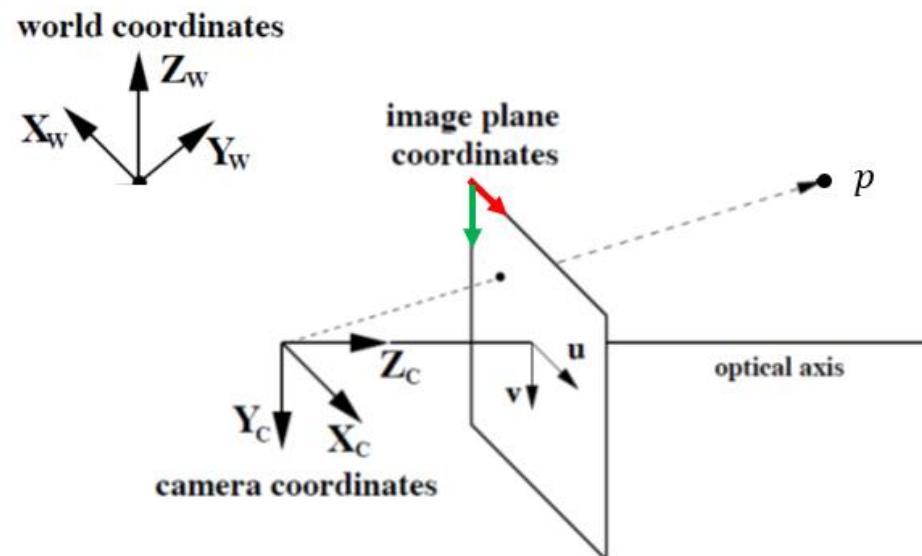




Perspective Projection

- Matrix form:

$$\lambda \begin{pmatrix} u \\ v \\ 1 \end{pmatrix} = \begin{pmatrix} f & 0 & u_o \\ 0 & f & v_o \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_c \\ y_c \\ z_c \end{pmatrix}$$

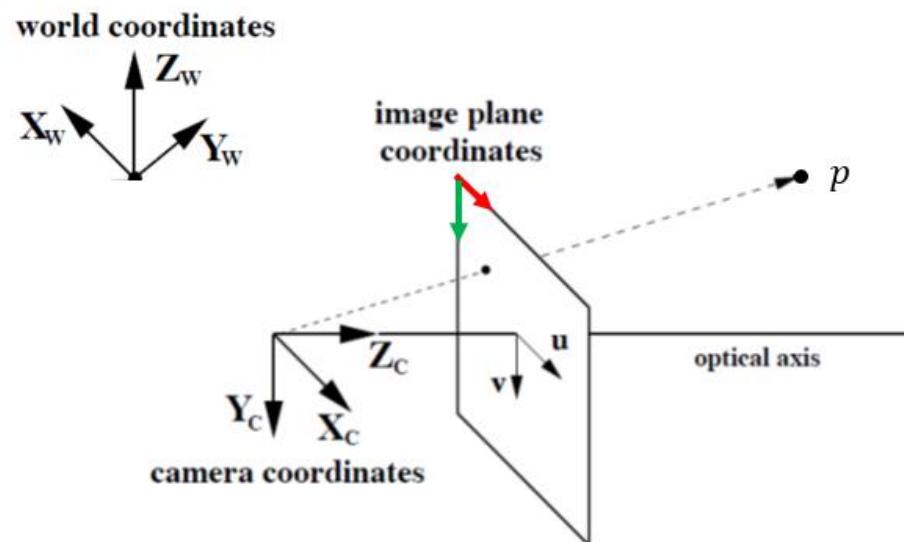




Perspective Projection

- From camera to world:

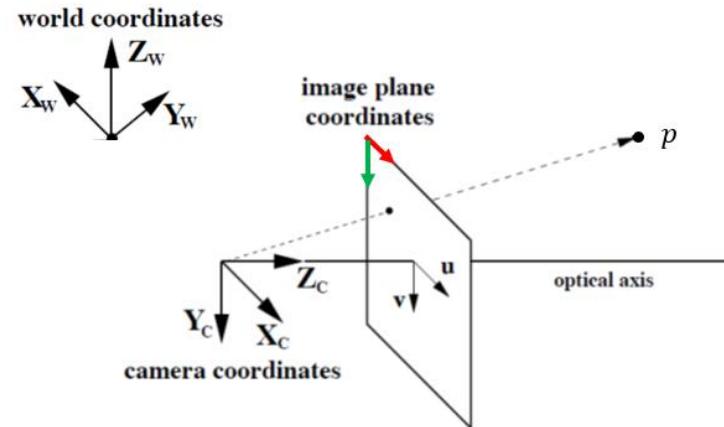
$$\begin{pmatrix} x_c \\ y_c \\ z_c \\ 1 \end{pmatrix} = \begin{pmatrix} {}^c\mathbf{R}_w & {}^c\mathbf{t}_w \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x_w \\ y_w \\ z_w \\ 1 \end{pmatrix}$$





Perspective Projection

$$\lambda \begin{pmatrix} u \\ v \\ 1 \end{pmatrix} = \begin{pmatrix} f & 0 & u_o \\ 0 & f & v_o \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} x_c \\ y_c \\ z_c \end{pmatrix}$$

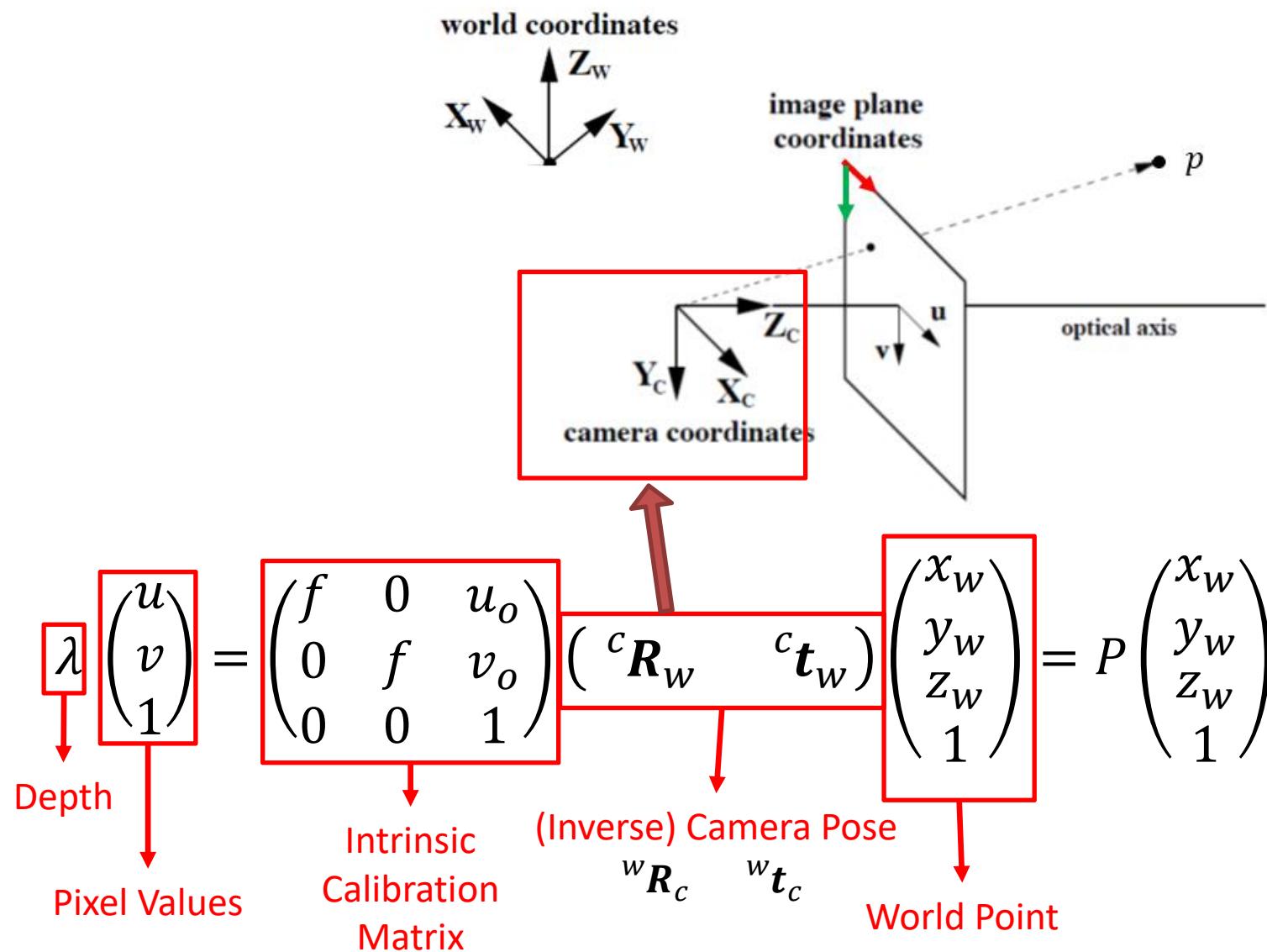


- From camera to world:

$$\begin{pmatrix} x_c \\ y_c \\ z_c \\ 1 \end{pmatrix} = \begin{pmatrix} {}^c\boldsymbol{R}_w & {}^c\boldsymbol{t}_w \\ 0 & 1 \end{pmatrix} \begin{pmatrix} x_w \\ y_w \\ z_w \\ 1 \end{pmatrix}$$



Pin-hole Camera Model



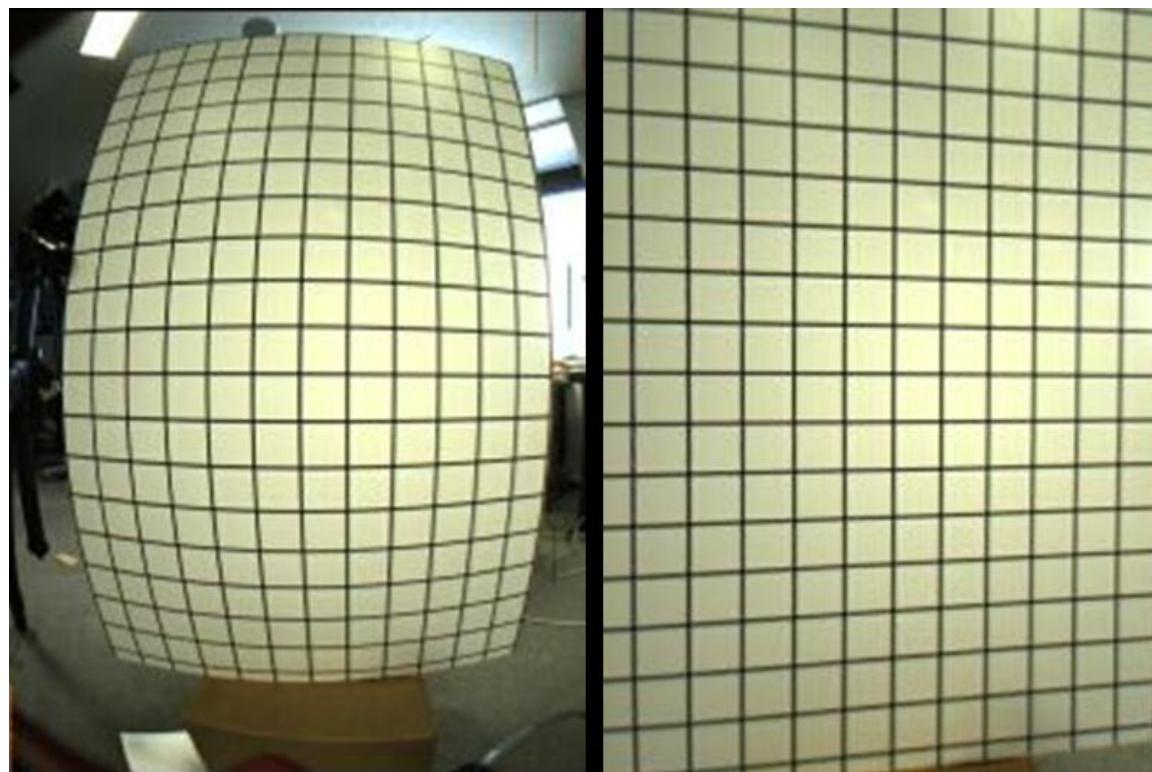


1. Camera Background
2. Pinhole Camera Model
3. Distortion
4. Fisheye Camera Model (MEI)
5. IPM (Inverse Projective Mapping)
6. Assignment



Camera Calibration

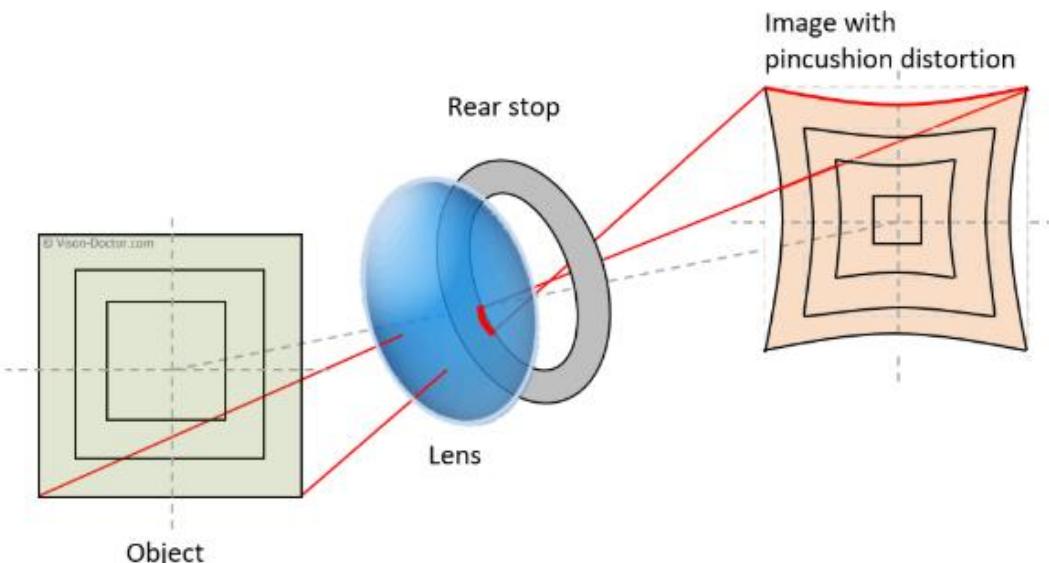
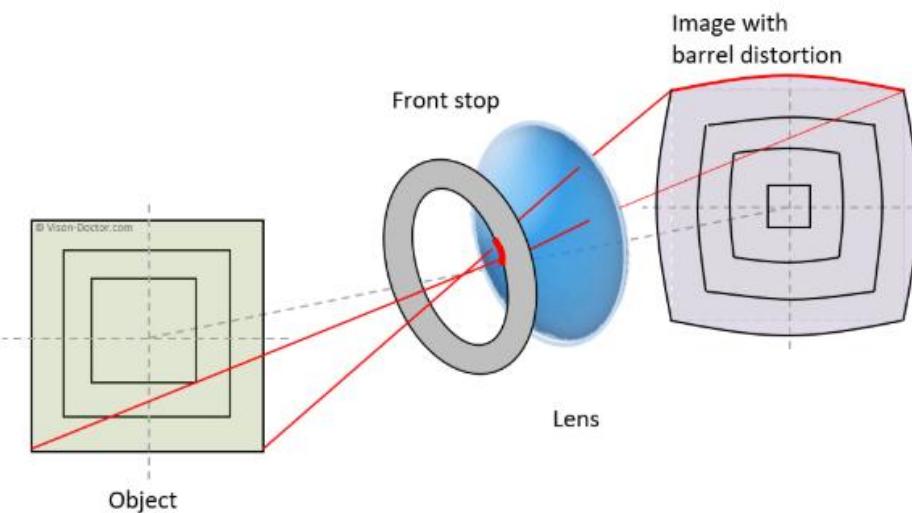
- Straight lines should be straight





Distortion model

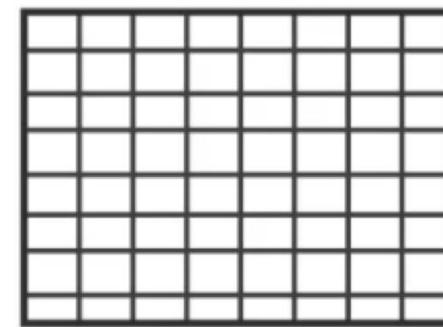
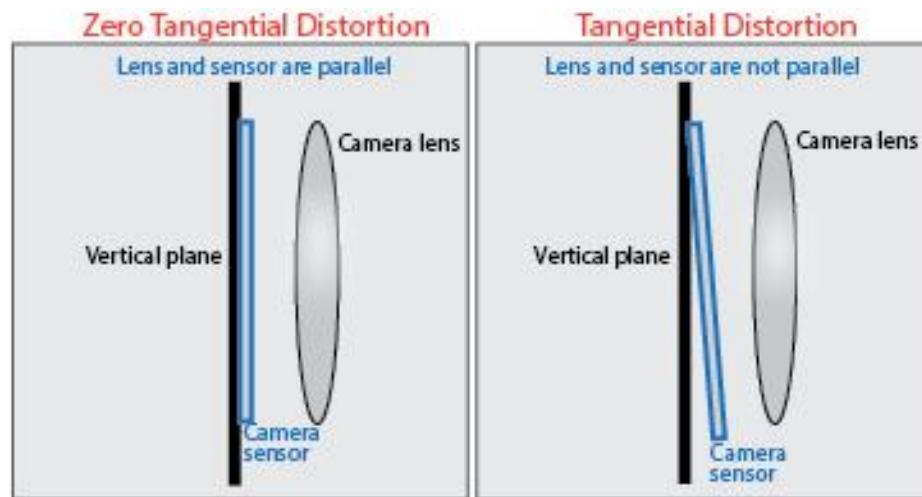
- Tangential and radial distortion model
 - Radial distortion



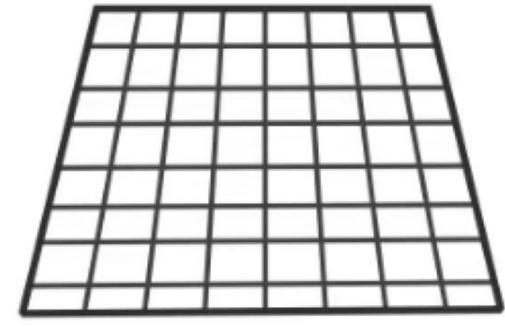


Distortion model

- Tangential and radial distortion model
 - Tangential distortion



Original Image

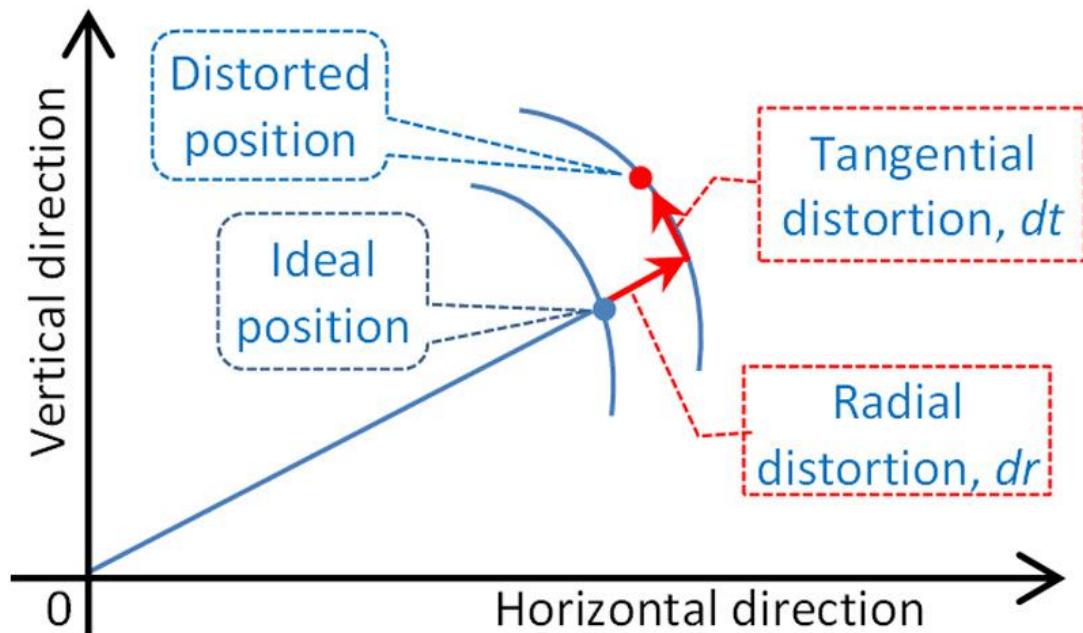


Tangential Distortion



Camera Calibration

- Tangential and radial distortion model

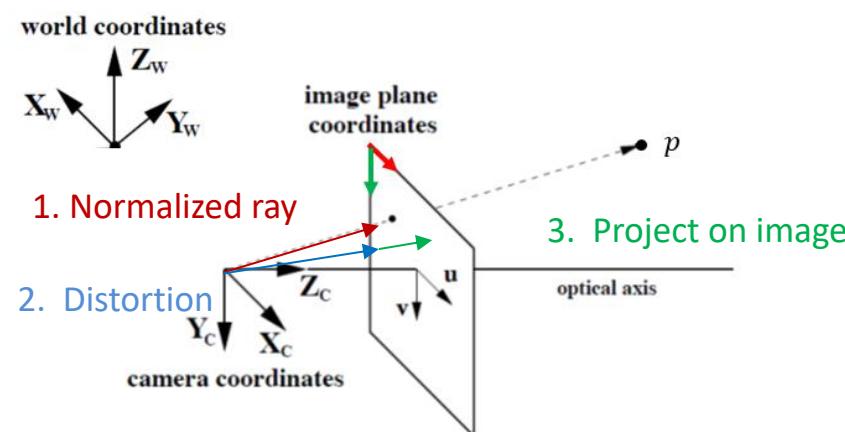




Distortion model

- Tangential and radial distortion model

- $k_1, k_2, [k_3, [k_4, [k_5, [k_6]]]]$ are the radial coefficients
- p_1 and p_2 are the tangential distortion coefficients



1. Project to unit plane ($z = 1$)

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} x_c/z_c \\ y_c/z_c \end{bmatrix}$$

2. Distortion:
(Distortion coefficient D) $r^2 = x'^2 + y'^2$

$$\begin{bmatrix} x'' \\ y'' \end{bmatrix} = \begin{bmatrix} x' \frac{1+k_1r^2+k_2r^4+k_3r^6}{1+k_4r^2+k_5r^4+k_6r^6} + 2p_1x'y' + p_2(r^2 + 2x'^2) \\ y' \frac{1+k_1r^2+k_2r^4+k_3r^6}{1+k_4r^2+k_5r^4+k_6r^6} + p_1(r^2 + 2y'^2) + 2p_2x'y' \end{bmatrix}$$

3. Image Projection:
(Camera matrix K)

$$\begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} f_x x'' + c_x \\ f_y y'' + c_y \end{bmatrix}$$



Projection Example

- Pinhole + Tangential and radial Distortion

$$^c p = [6, 4, 2]$$

```
1 %YAML:1.0
2 ---
3 calibration_Time: "2024年12月12日 星期四 16时32分21秒"
4 nrOfFrames: 57
5 image_Width: 640
6 image_Height: 480
7 board_Width: 7
8 board_Height: 5
9 square_Size: 1.0000000149011612e-01
10 Camera_Matrix: !!opencv-matrix
11   rows: 3
12   cols: 3
13   dt: d
14   data: [ 5.6865194941315929e+02, 0., 3.2513762614210782e+02, 0.,
15     7.6375465300688415e+02, 2.3177702561676526e+02, 0., 0., 1. ]
16 Distortion_Coefficients: !!opencv-matrix
17   rows: 1
18   cols: 5
19   dt: d
20   data: [ 1.1726086573014034e-02, -5.2021897536855632e-01,
21     -4.8681362994895585e-03, 9.6693728067371231e-04,
22     2.2217257704057318e+00 ]
```

$$(k_1 \quad k_2 \quad p_1 \quad p_2 \quad k_3)$$



$$\begin{bmatrix} f_x & 0 & c_x \\ 0 & f_y & c_y \\ 0 & 0 & 1 \end{bmatrix}$$

$$\begin{bmatrix} x'' \\ y'' \end{bmatrix} = \begin{bmatrix} x' \frac{1+k_1r^2+k_2r^4+k_3r^6}{1+k_4r^2+k_5r^4+k_6r^6} + 2p_1x'y' + p_2(r^2 + 2x'^2) \\ y' \frac{1+k_1r^2+k_2r^4+k_3r^6}{1+k_4r^2+k_5r^4+k_6r^6} + p_1(r^2 + 2y'^2) + 2p_2x'y' \end{bmatrix}$$

1. Project to unit plane : $\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} x_c/z_c \\ y_c/z_c \end{bmatrix} = \begin{bmatrix} 3 \\ 2 \end{bmatrix}$

2. Distortion: $r^2 = 9 + 4 = 13$

$$\begin{bmatrix} x'' \\ y'' \end{bmatrix} = \begin{bmatrix} x'(1 + k_1r^2 + k_2r^4 + k_3r^6) + 2p_1x'y' + p_2(r^2 + 2x'^2) \\ y'(1 + k_1r^2 + k_2r^4 + k_3r^6) + p_1(r^2 + 2y'^2) + 2p_2x'y' \end{bmatrix}$$

3. Image Projection:

$$\begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} f_x x'' + c_x \\ f_y y'' + c_y \end{bmatrix} = \begin{bmatrix} 568.65 x'' + 325.13 \\ 763.75 y'' + 231.77 \end{bmatrix}$$



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Fisheye Camera

- Dioptric camera

折射

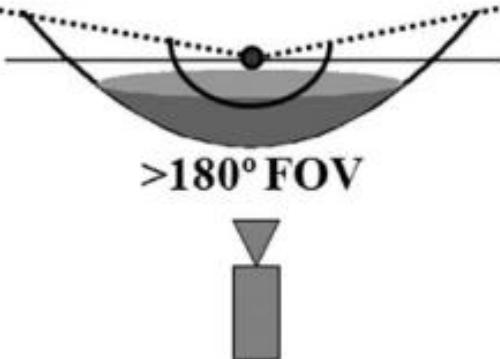


~180° FOV



- Catadioptric camera

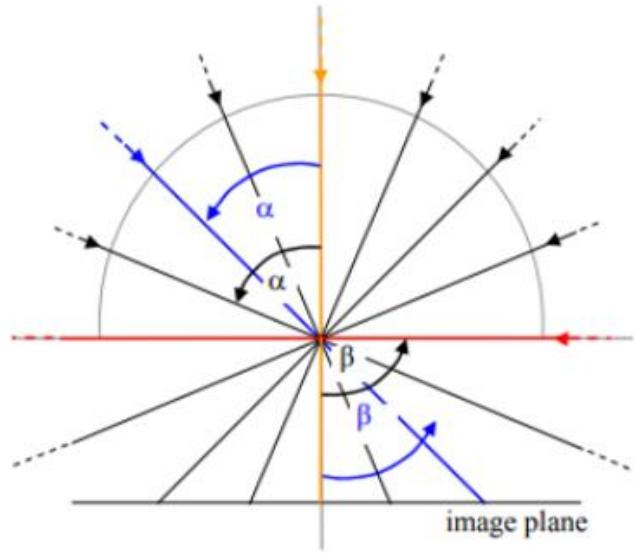
反射折射



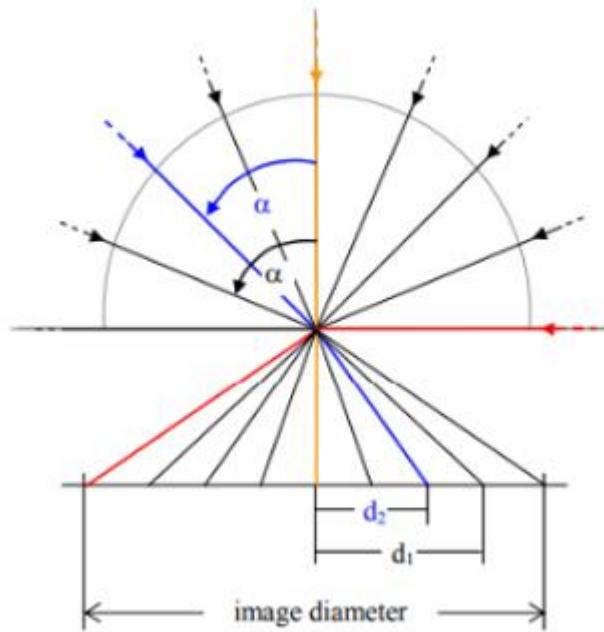
>180° FOV



Fisheye Camera Model



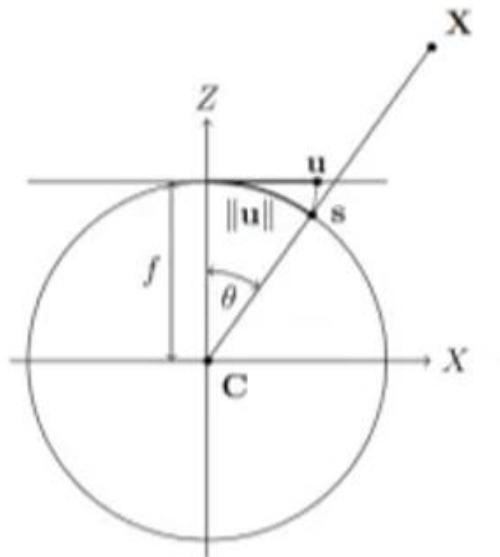
Pinhole perspective model



Fisheye perspective model

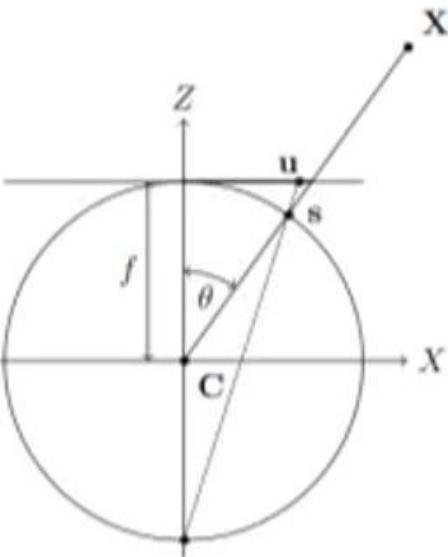


Classical Fisheye Camera Models



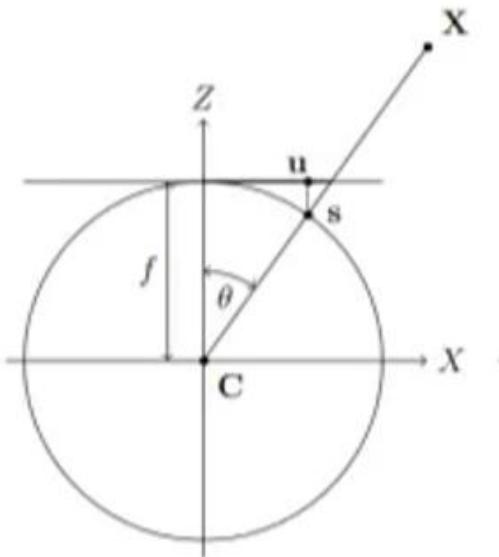
(a) Equidistant
等距

$$u = f\theta$$



(b) Stereographic
体视

$$u = 2f \tan \frac{\theta}{2}$$



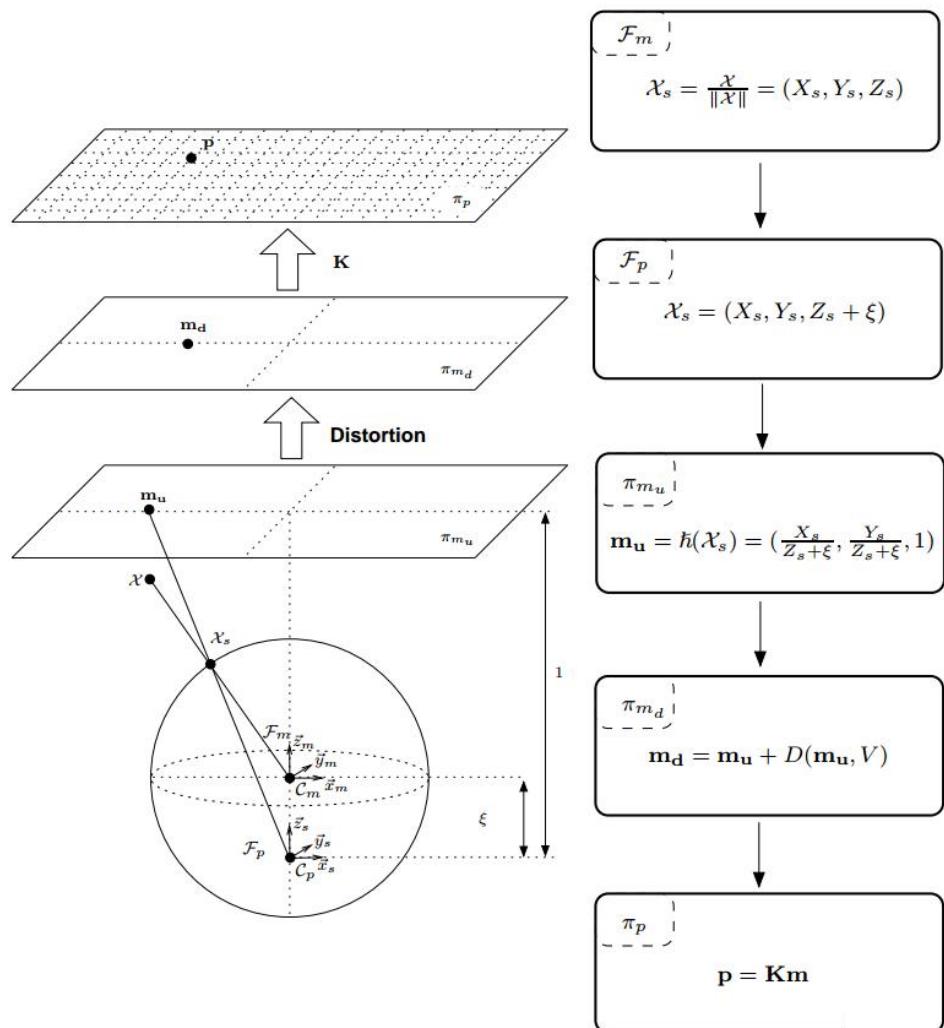
(c) Orthographic
正交

$$u = f \sin \theta$$

....



MEI [1] Camera Model



1. 在 mirror 坐标系上的世界点映射到单位圆上:

$$(\mathcal{X})_{\mathcal{F}_m} \rightarrow (\mathcal{X}_s)_{\mathcal{F}_m} = \frac{\mathcal{X}}{\|\mathcal{X}\|} = (X_s, Y_s, Z_s)$$

2. 将这些点转换为以 $C_p = (0, 0, \xi)$ 为中心的新参考系上:

$$(\mathcal{X}_s)_{\mathcal{F}_m} \rightarrow (\mathcal{X}_s)_{\mathcal{F}_p} = (X_s, Y_s, Z_s + \xi)$$

3. 将它们投影到归一化图像平面上:

$$\mathbf{m}_u = \left(\frac{X_s}{Z_s + \xi}, \frac{Y_s}{Z_s + \xi}, 1 \right) = \hbar(\mathcal{X}_s)$$

4. 添加上径向和切向畸变:

$$\mathbf{m}_d = \mathbf{m}_u + D(\mathbf{m}_u, V)$$

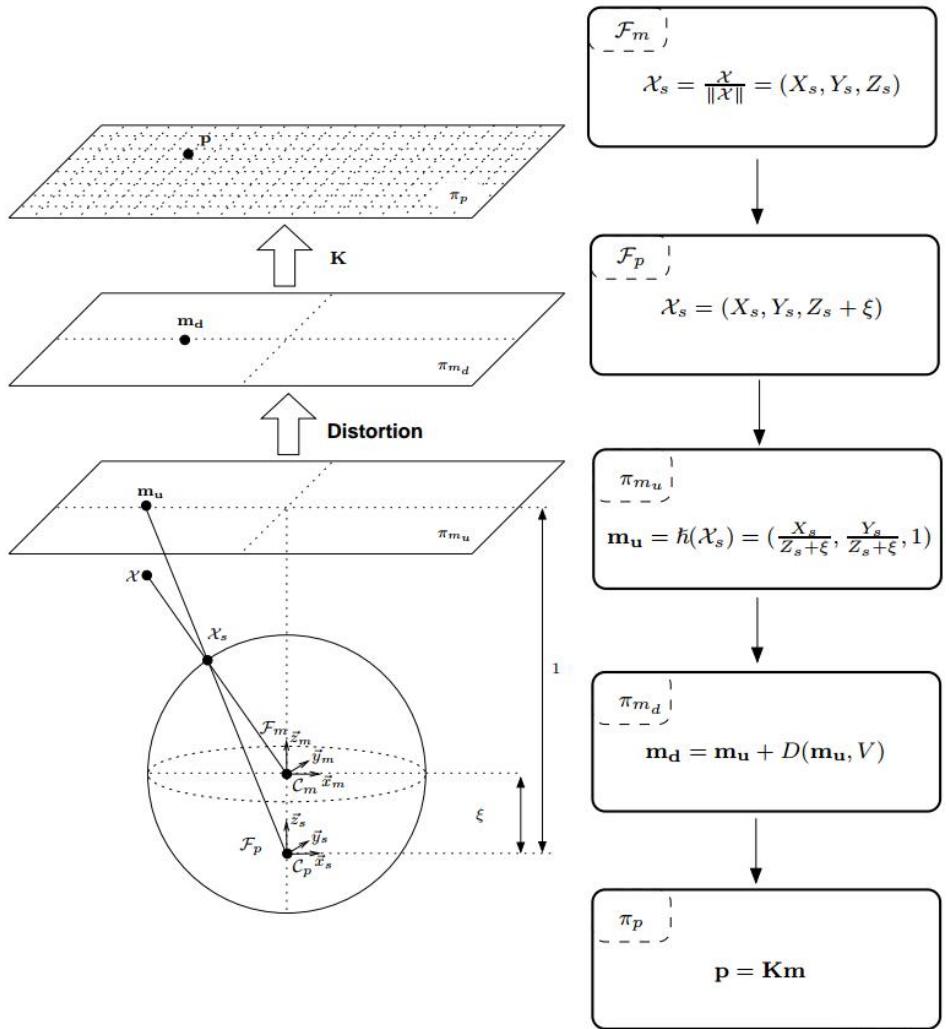
5. 最终投影涉及一个广义相机投影矩阵 \mathbf{K} (其中 γ 为广义焦距, (u_0, v_0) 为主点, s 斜度):

$$\mathbf{p} = \mathbf{K}\mathbf{m} = \begin{bmatrix} \gamma & \gamma s & u_0 \\ 0 & \gamma r & v_0 \\ 0 & 0 & 1 \end{bmatrix} \mathbf{m} = k(\mathbf{m})$$

[1] C. Mei and P. Rives, "Single View Point Omnidirectional Camera Calibration from Planar Grids," *Proceedings 2007 IEEE International Conference on Robotics and Automation*, Rome, Italy, 2007, pp. 3945-3950, doi: 10.1109/ROBOT.2007.364084.



Example: MEI Camera Model



$$^c p = [2, 2, 1]$$

```

1 %YAML:1.0
2 ---
3 model_type: MEI
4 camera_name: camera
5 image_width: 1920
6 image_height: 1080
7 mirror_parameters:
8   xi: 1.0
9 distortion_parameters:
10  k1: 0.0
11  k2: 0.0
12  p1: 0.0
13  p2: 0.0
14 projection_parameters:
15  gamma1: 935.307436087
16  gamma2: 935.307436087
17  u0: 960
18  v0: 540

```

1. Normalized ray vector:

$$\begin{bmatrix} x' \\ y' \\ z' \end{bmatrix} = \begin{bmatrix} x_c / \|p\| \\ y_c / \|p\| \\ z_c / \|p\| \end{bmatrix} = \begin{bmatrix} 0.67 \\ 0.67 \\ 0.33 \end{bmatrix}$$

2. Change to C_p coordinate:

$$\begin{bmatrix} x'' \\ y'' \\ z'' \end{bmatrix} = \begin{bmatrix} x' \\ y' \\ z' + \xi \end{bmatrix} = \begin{bmatrix} 0.67 \\ 0.67 \\ 1.33 \end{bmatrix}$$

3. Project to unit plane:

$$\begin{bmatrix} x''' \\ y''' \end{bmatrix} = \begin{bmatrix} x''/z'' \\ y''/z'' \end{bmatrix} = \begin{bmatrix} 0.5 \\ 0.5 \end{bmatrix}$$

4. Distortion:

5. Image projection:

$$\begin{bmatrix} u \\ y \end{bmatrix} = \begin{bmatrix} f_x x''' + c_x \\ f_y y''' + c_y \end{bmatrix} = \begin{bmatrix} 935.3 x''' + 960 \\ 935.3 y''' + 540 \end{bmatrix} = \begin{bmatrix} 1427.66 \\ 1007.66 \end{bmatrix}$$



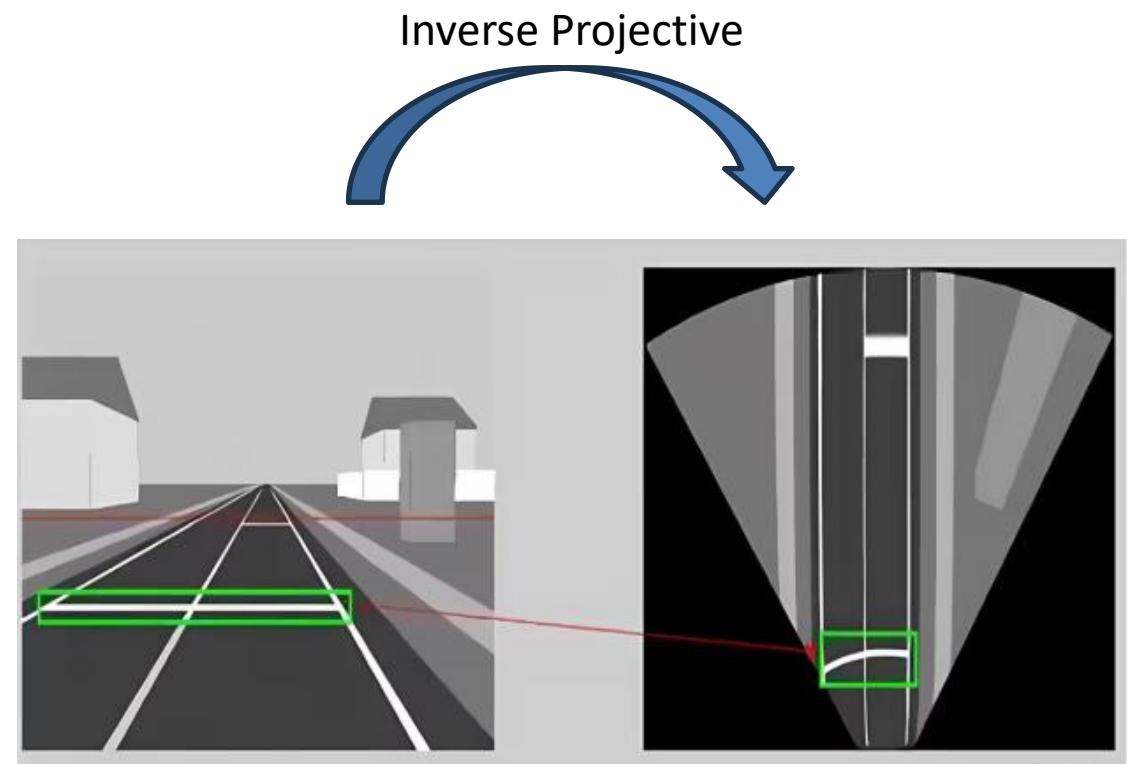
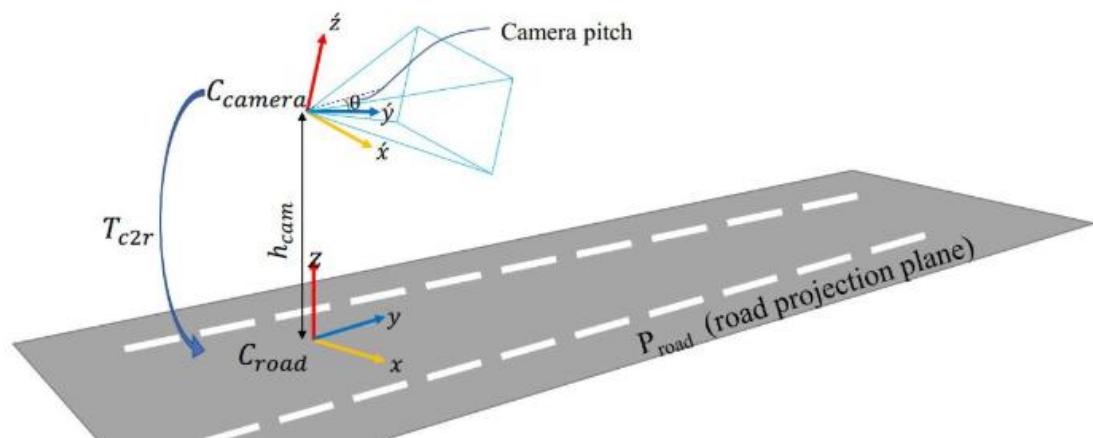
Content

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IPM Image

- IPM (Inverse Projective Mapping)





IPM Image

Method I:

r, g, b

(u, v)

Camera model
Inverse projection

$\lambda(x_c, y_c, 1)$

${}^v R_c, {}^v t_c$

$(x_v, y_v, 0)$

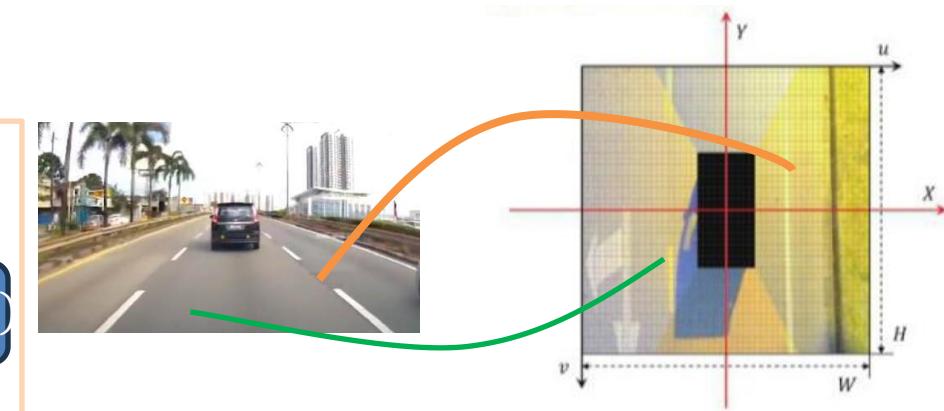
(u_{ipm}, v_{ipm})

Raw image plane

Camera coordinate

Vehicle coordinate

IPM plane



Method II:

(u_{ipm}, v_{ipm})

$(x_v, y_v, 0)$

${}^c R_v, {}^c t_v$

(x_c, y_c, z_c)

Camera model
projection

r, g, b

IPM plane

Vehicle coordinate

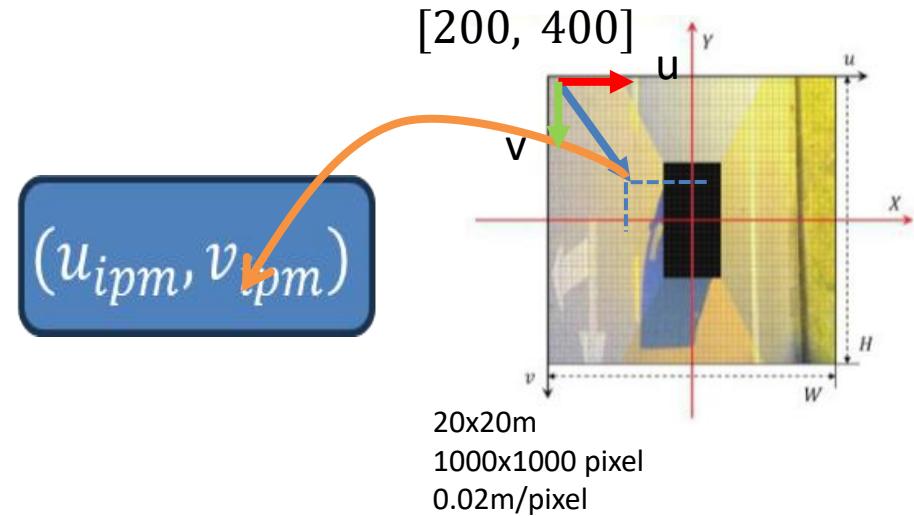
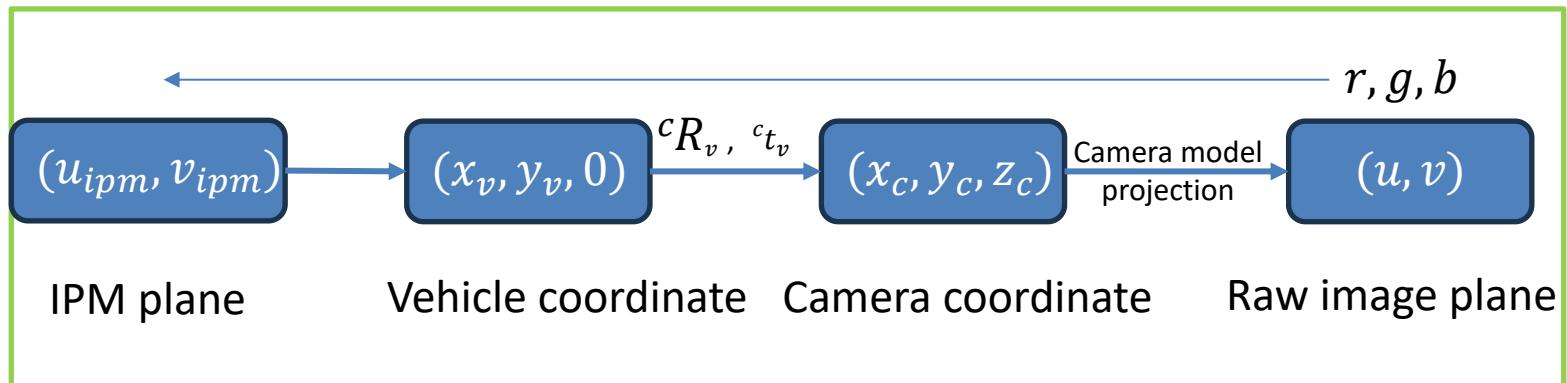
Camera coordinate

Raw image plane



IPM Image

Method II:



1. Choose one point on IPM plane, for example: $\begin{bmatrix} u_{ipm} \\ v_{ipm} \end{bmatrix} = \begin{bmatrix} 200 \\ 400 \end{bmatrix}$

$$2. \text{ Vehicle coordinate: } \begin{bmatrix} x_v \\ y_v \\ z_v \end{bmatrix} = \begin{bmatrix} -(500 - u_{ipm}) * 0.02 \\ (500 - v_{ipm}) * 0.02 \\ 0 \end{bmatrix} = \begin{bmatrix} -6 \\ 2 \\ 0 \end{bmatrix}$$

$$3. \text{ Camera coordinate: } \begin{bmatrix} x_c \\ y_c \\ z_c \\ 1 \end{bmatrix} = \begin{bmatrix} {}^cR_v & {}^c t_v \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_v \\ y_v \\ z_v \\ 1 \end{bmatrix} = \begin{bmatrix} {}^vR_c & {}^v t_c \\ 0 & 1 \end{bmatrix}^{-1} \begin{bmatrix} x_v \\ y_v \\ z_v \\ 1 \end{bmatrix}$$

$$4. \text{ Project to image plane: } \begin{bmatrix} u \\ v \end{bmatrix} = \pi(p_c, K, D)$$



```

1 %YAML:1.0
2 ---
3 frame_id: cam0
4 transform:
5   translation:
6     x: 0.0
7     y: 2.5
8     z: 1.05
9   rotation:
10    x: -0.866
11    y: 0.0
12    z: 0.0
13    w: 0.5
  
```

${}^v t_c$

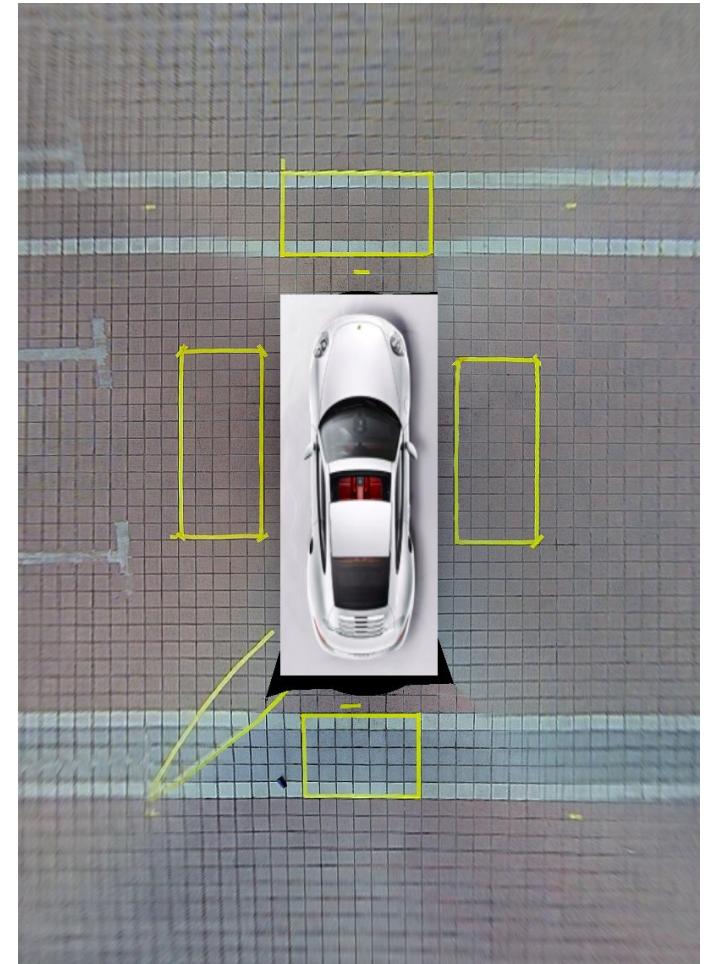
${}^v R_c$



IPM Image Stitching

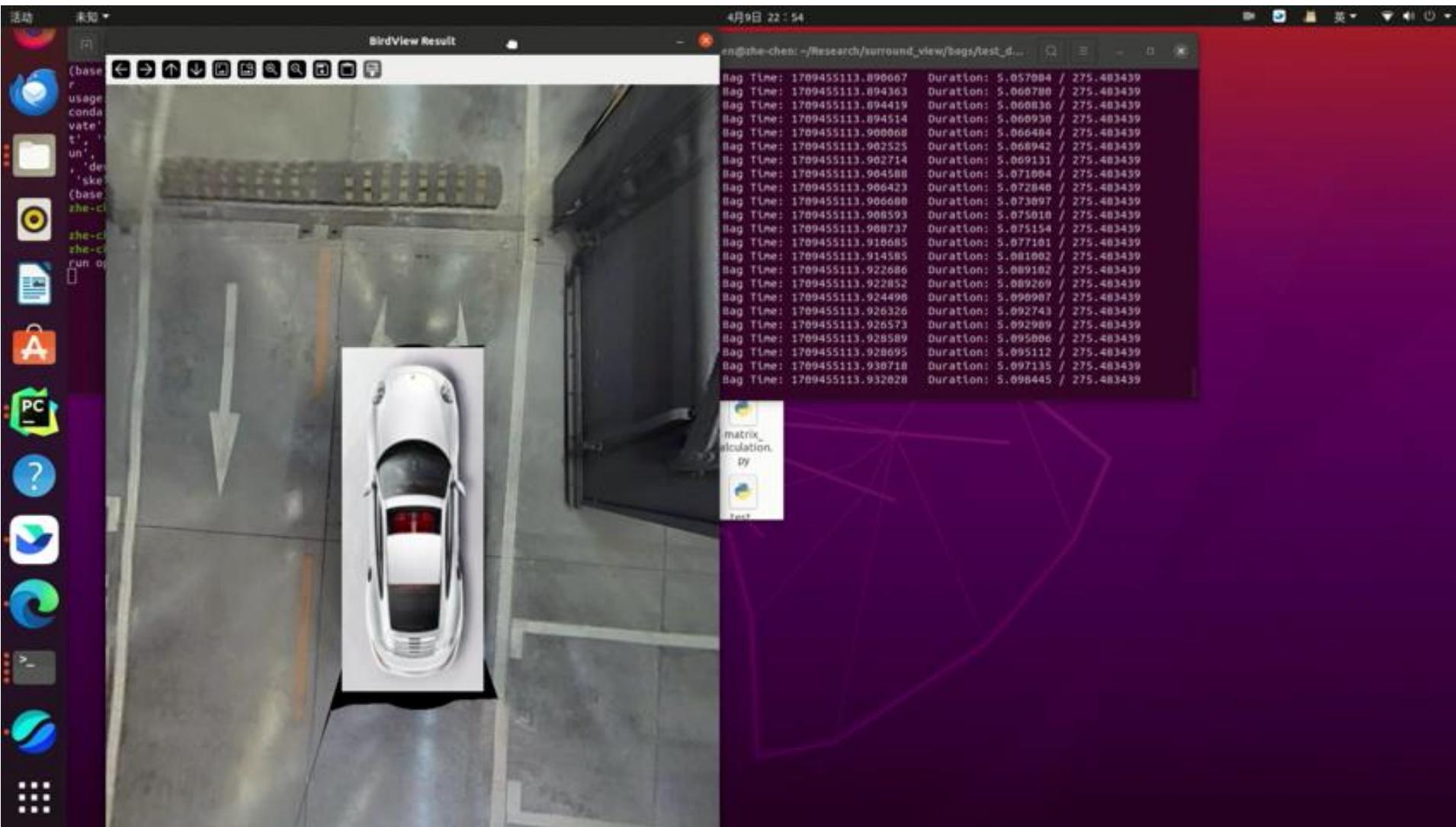


Weight
fusion





IPM Image Stitching





-  1. Camera Background
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-  5. IPM (Inverse Projective Mapping)
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Assignment

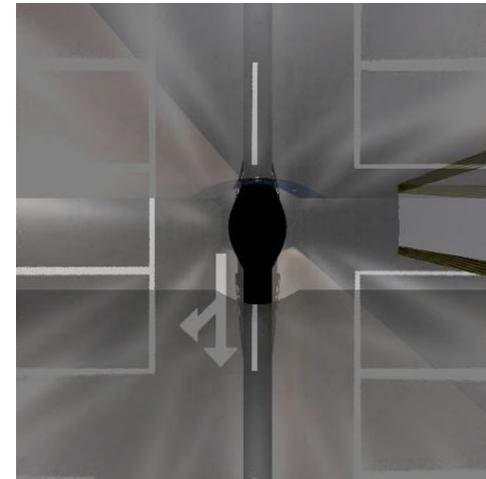
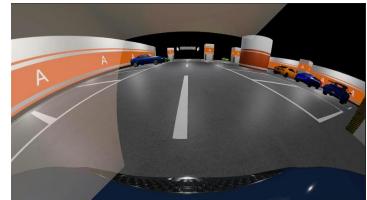
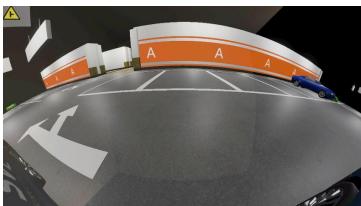
Implement the IPM generation:



- Complete the function *GenerateIPMImage()*

```
84  
85 cv::Mat IPM::GenerateIPMImage(const std::vector<cv::Mat>& images) const {
```

- Expected result:





感谢聆听 !

Thanks for Listening !

