# Principles of Robot Autonomy I

Camera models and camera calibration





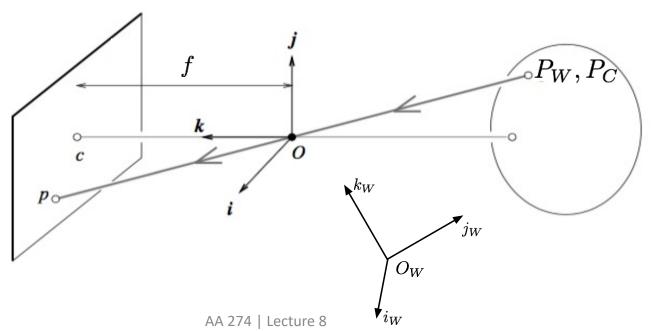
### Camera models and camera calibration

• Aim

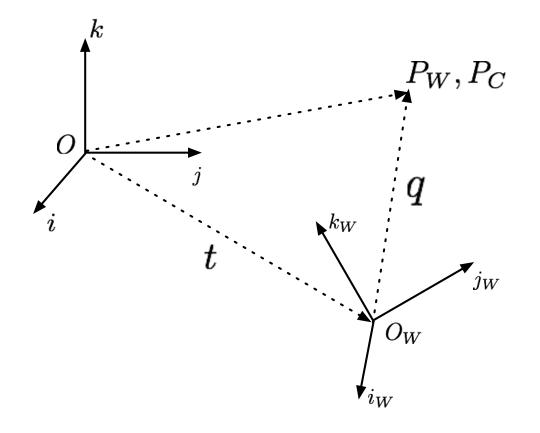
- Learn how to calibrate a camera
- Learn about 3D reconstruction
- Readings
  - SNS: 4.2.3
  - D. A. Forsyth and J. Ponce [FP]. Computer Vision: A Modern Approach (2nd Edition). Prentice Hall, 2011. Chapter 1.
  - R. Hartley and A. Zisserman [HZ]. Multiple View Geometry in Computer Vision. Academic Press, 2002. Chapter 6.1.
  - Z. Zhang. A Flexible New Technique for Camera Calibration. IEEE Transactions on Pattern Analysis and Machine Intelligence, 2000.

## Step 3

- In previous lecture, we have derived a mapping between a point *P* in the 3D camera reference frame to a point *p* in the 2D image plane
- Last step is to include in our mapping an additional transformation to account for the difference between the world frame and the 3D camera reference frame



### **Rigid transformations**



$$P_C = t + q$$
$$q = R P_W$$

where *R* is the rotation matrix relating camera and world frames

$$R = egin{bmatrix} i_W \cdot i & j_W \cdot i & k_W \cdot i \ i_W \cdot j & j_W \cdot j & k_W \cdot j \ i_W \cdot k & j_W \cdot k & k_W \cdot k \end{bmatrix}$$

$$\Rightarrow P_C = t + R P_W$$

Rigid transformations in homogeneous coordinates

 $\begin{pmatrix} P_C \\ 1 \end{pmatrix} = \begin{bmatrix} R & t \\ 0_{1 \times 3} & 1 \end{bmatrix} \begin{pmatrix} P_W \\ 1 \end{pmatrix}$ 

Point  $P_c$  in homogeneous coordinates

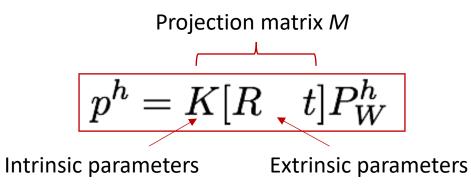
Point  $P_w$  in homogeneous coordinates

### Perspective projection equation

• Collecting all results

$$p^{h} = \begin{bmatrix} K & 0_{3\times 1} \end{bmatrix} P_{C}^{h} = K \begin{bmatrix} I_{3\times 3} & 0_{3\times 1} \end{bmatrix} \begin{bmatrix} R & t \\ 0_{1\times 3} & 1 \end{bmatrix} P_{W}^{h}$$

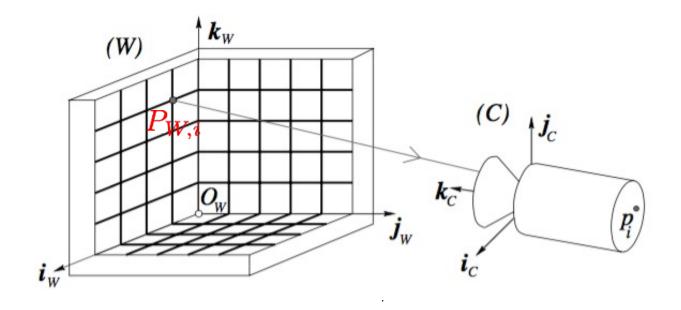
• Hence



• Degrees of freedom: 4 for K (or 5 if we also include skewness), 3 for R, and 3 for t. Total is 10 (or 11 if we include skewness)

## Camera calibration: direct linear transformation method

• Goal: find the intrinsic and extrinsic parameters of the camera



Strategy: given known correspondences  $p_i \leftrightarrow P_{W,i}$ , compute unknown parameters *K*, *R*, *t* by applying perspective projection

 $P_{W,1}, P_{W,2}, \dots, P_{W,n}$  with known positions in world frame  $p_1, p_2, \dots, p_n$  with known positions in image frame

### Step 1

• First consider combined parameters

$$p_i^h = M P_{W,i}^h, \ i = 1, \dots, n,$$
 where  $M = K[R \ t] = \begin{bmatrix} m_1 \\ m_2 \\ m_3 \end{bmatrix}$ 

• This gives rise to 2n component-wise equations, for i = 1, ..., n

$$u_{i} = \frac{m_{1} \cdot P_{W,i}^{h}}{m_{3} \cdot P_{W,i}^{h}} \qquad \text{or} \qquad u_{i} \left(m_{3} \cdot P_{W,i}^{h}\right) - m_{1} \cdot P_{W,i}^{h} = 0$$
$$v_{i} = \frac{m_{2} \cdot P_{W,i}^{h}}{m_{3} \cdot P_{W,i}^{h}} \qquad \text{or} \qquad v_{i} \left(m_{3} \cdot P_{W,i}^{h}\right) - m_{2} \cdot P_{W,i}^{h} = 0$$

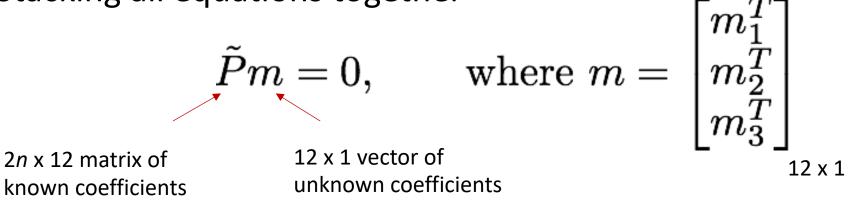
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 $1 \times 4$  vector

 $[m_1]$ 

### Calibration problem

• Stacking all equations together



- $\tilde{P}$  contains in block form the known coefficients stemming from the given correspondences
- To estimate 11 coefficients, we need at least 6 correspondences

### Solution

• To find non-zero solution

$$\min_{\substack{m \in R^{12} \\ \text{subject to}}} \|\tilde{P}m\|^2$$

- Solution: select eigenvector of  $\tilde{P}^T \tilde{P}$  with the smallest eigenvalue
- Readily computed via SVD (singular value decomposition)

### Step 2

• Next, we need to extract the camera parameters, i.e., we want to factorize *M* as

$$M = \begin{bmatrix} KR & Kt \end{bmatrix}$$

- This can be done efficiently (indeed, explicitly) by using RQ factorization, whereby the submatrix  $M_{1:3,1:3}$  is decomposed into the product of an upper triangular matrix *K* and a rotation matrix *R*
- Calibration will be investigated in Problem 1 in HW3

### Radial distortion

- So far, we have assumed that a linear model is an accurate model of the imaging process
- For real (non-pinhole) lenses this assumption will not hold







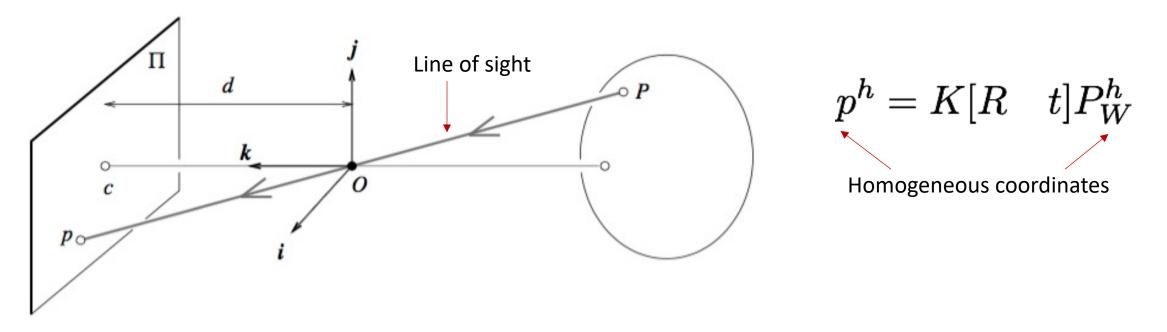
Credit: SNS

#### No distortion

#### Barrel distortion

#### Pincushion distortion

### Measuring depth



Once the camera is calibrated, can we measure the location of a point *P* in 3D given its known observation *p*?

• No: one can only say that *P* is located *somewhere* along the line joining *p* and *O*!

### Issues with recovering structure



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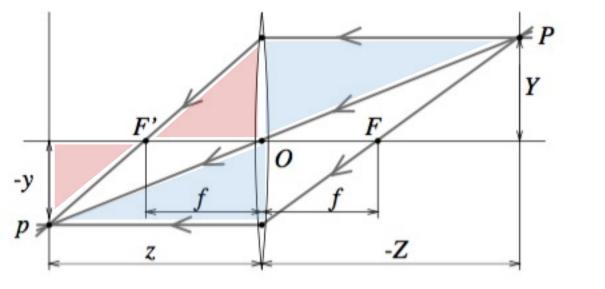
### Recovering structure

• Structure: 3D scene to be reconstructed by having access to 2D images

### Common methods

- 1. Through recognition of landmarks (e.g., orthogonal walls)
- 2. Depth from focus: determines distance to one point by taking multiple images with better and better focus
- 3. Stereo vision: processes two distinct images taken at the *same time* and assumes that the relative pose between the two cameras is *known*
- 4. Structure from motion: processes two images taken with the same or different cameras at *different times* and from different *unknown* positions

### Depth from focus



Credit: FP Chapter 1

• Take several images until the projection of point *P* is in focus; let *z* denote the distance at which the image is in focus

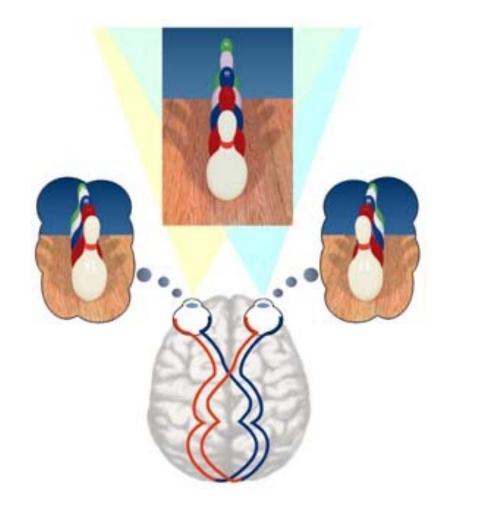
 $\Rightarrow \frac{\mathbf{1}}{z} + \frac{1}{Z} = \frac{1}{\mathbf{f}}$ 

• Since we know z and f, through the thin lens equation we obtain Z

Thin lens

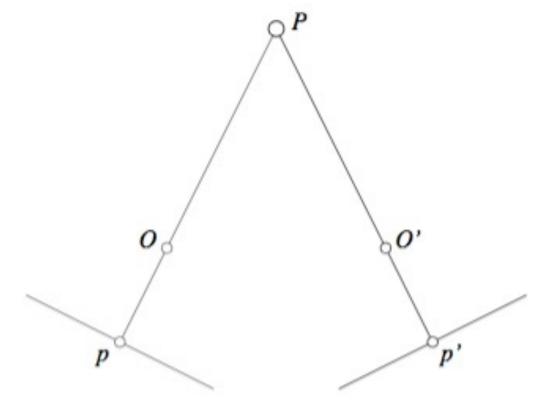
equation

### Stereopsis, or why we have two eyes



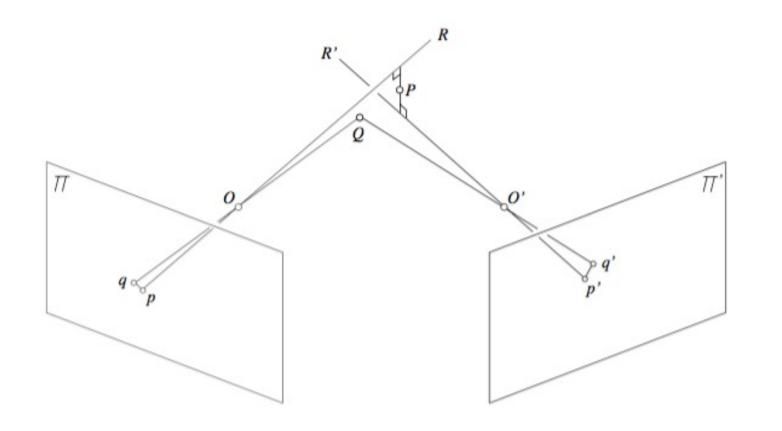


### Binocular reconstruction



- Given: calibrated stereo rig and two image matching points p and p'
- Find corresponding scene point by intersecting the two rays Op and O'p' (process known as triangulation)

### Approximate triangulation



 Due to noise, triangulation problem is often solved as finding the point Q with images q and q' that minimizes

$$d^2(p,q) + d^2(p',q')$$

**Re-projection error** 

### Next time: image processing, feature detection & description

