

## Exercise 2

### Epipolar Geometry

a)  $K = R = I$

$$\tilde{E}_1 = [t_1] \times R = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & -1 \\ 0 & 1 & 0 \end{bmatrix}$$

$$\tilde{E}_2 = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ -1 & 0 & 0 \end{bmatrix}$$

$$\tilde{E}_3 = \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix}$$

$t_1$  时

设点为  $\bar{X}_1 = \begin{pmatrix} x_1 \\ y_1 \\ 1 \end{pmatrix}$   $\bar{X}_2 = \begin{pmatrix} x_2 \\ y_2 \\ 1 \end{pmatrix}$

$$\tilde{x}_2 = \tilde{E} \tilde{x}_1 = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & -1 \\ 0 & 1 & 0 \end{bmatrix} \begin{pmatrix} x_1 \\ y_1 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ -1 \\ y_1 \end{pmatrix}$$

$$\tilde{l}_1 = \tilde{E}^T \tilde{x}_2 = \begin{bmatrix} 0 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -1 & 0 \end{bmatrix} \begin{pmatrix} x_2 \\ y_2 \\ 1 \end{pmatrix} = \begin{pmatrix} 0 \\ 1 \\ -y_2 \end{pmatrix}$$

两条线均为水平线

$\therefore$  平行  $\therefore$  极点位于无穷远处  
为 ideal point

$t_2$  时  $\tilde{x}_2 = \tilde{E} \tilde{x}_1 = \begin{bmatrix} 0 & 0 & 1 \\ 0 & 0 & 0 \\ -1 & 0 & 0 \end{bmatrix} \begin{pmatrix} x_1 \\ y_1 \\ 1 \end{pmatrix} = \begin{pmatrix} 1 \\ 0 \\ -x_1 \end{pmatrix}$   $\tilde{l}_1 = \tilde{E}^T \tilde{x}_2 = \begin{bmatrix} 0 & 0 & -1 \\ 0 & 0 & 0 \\ 1 & 0 & 0 \end{bmatrix} \begin{pmatrix} x_2 \\ y_2 \\ 1 \end{pmatrix} = \begin{pmatrix} -1 \\ 0 \\ x_2 \end{pmatrix}$

为垂直线  $\rightarrow$  平行  $\rightarrow$  极点位于无穷远处

$t_3$  时  $\tilde{x}_2 = \tilde{E} \tilde{x}_1 = \begin{bmatrix} 0 & -1 & 0 \\ 1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{pmatrix} x_1 \\ y_1 \\ 1 \end{pmatrix} = \begin{pmatrix} -y_1 \\ x_1 \\ 0 \end{pmatrix}$

$$\tilde{l}_1 = \tilde{E}^T \tilde{x}_2 = \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 0 \end{bmatrix} \begin{pmatrix} x_2 \\ y_2 \\ 1 \end{pmatrix} = \begin{pmatrix} y_2 \\ -x_1 \\ 0 \end{pmatrix}$$

c)

$$K_2^T \tilde{E} K_1^T = \tilde{F} = \tilde{E} \text{ 时}$$

$$\begin{aligned} \text{即要求 } & \begin{cases} \bar{x}_2^T K_2^T = \bar{x}_2^T \\ K_1^T \bar{x}_1 = \bar{x}_1 \end{cases} \\ \Rightarrow & \begin{cases} K_2^T = I & \therefore K_2 = I \\ K_1^T = I & K_1 = I \end{cases} \end{aligned}$$

$$\tilde{x}_2^T \tilde{E} \tilde{x}_1 = 0$$

$$\tilde{x}_i = K_i^T \bar{x}_i$$

↓

$$\bar{x}_2^T K_2^T \tilde{E} K_1^T \bar{x}_1 = \bar{x}_2^T \tilde{F} \bar{x}_1 = 0$$

$$\therefore \begin{aligned} \tilde{x}_2 &= \bar{x}_2 & \therefore K_2 &= I \\ \tilde{x}_1 &= \bar{x}_1 & K_1 &= I \end{aligned}$$

Triangulation

$$a) \quad \tilde{x}_i^{(s)} = \tilde{p}_i \tilde{x}_w$$

$$\tilde{x}_1^{(s)} = \tilde{p}_1 \tilde{x}_w = \begin{pmatrix} \frac{1}{6} \\ \frac{1}{2} \\ 1 \end{pmatrix}$$

$$\tilde{x}_2^{(s)} = \tilde{p}_2 \tilde{x}_w = \begin{pmatrix} -\frac{1}{3} \\ \frac{1}{2} \\ 1 \end{pmatrix}$$

$$P_1 = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix}$$

$$P_2 = \begin{bmatrix} 2 & 0 & 1 \\ 0 & 2 & 1 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} -1 & 0 & 0 & -2 \\ 0 & -1 & 0 & 0 \\ 0 & 0 & 1 & 1 \end{bmatrix} = \begin{bmatrix} -2 & 0 & 1 & -3 \\ 0 & -2 & 1 & 1 \\ 0 & 0 & 1 & 1 \end{bmatrix}$$

$$\underline{P = K[R|t]}$$

$$\text{由 } \begin{bmatrix} x_i^s & \tilde{p}_{i3}^T & -\tilde{p}_{i1}^T \\ y_i^s & \tilde{p}_{i3}^T & -\tilde{p}_{i2}^T \end{bmatrix} \tilde{x}_w = 0 \quad \text{得} \quad A = \begin{bmatrix} -1 & 0 & 0.25 & 0 \\ 0 & -1 & 0.5 & 0 \\ 2 & 0 & -1.2 & 2.8 \\ 0 & 2 & 0.8 & -0.8 \end{bmatrix}$$

$$\text{解得 } \tilde{x}_w = \begin{bmatrix} 1 \\ 2 \\ 4 \\ 1 \end{bmatrix}$$

## Stereo Vision

a)

$$z(d) = \frac{fb}{d}$$

$$z(d+\Delta d) = \frac{fb}{d+\Delta d}$$

$$\text{error} = z(d+\Delta d) - z(d)$$

对  $d$  导数  $-\frac{fb}{d^2} = -\frac{z^2}{fb}$

$$\frac{z(d+\Delta d) - z(d)}{\Delta d} = -\frac{z^2}{fb}$$

$$\therefore \text{error} = -\Delta d \cdot \frac{z^2}{fb}$$

error 随  $z^2$  增长

b)

增加  $b$  的大小

基线太长可能导致测量不准

## Block Matching

a)

$ZNCC$  :

$$NCC(x, y, d) = \frac{(w_L(x, y) - \bar{w}_L(x, y))^T}{\|w_L(x, y) - \bar{w}_L(x, y)\|_2} \cdot \frac{(w_R(x-d, y) - \bar{w}_R(x-d, y))}{\|w_R(x-d, y) - \bar{w}_R(x-d, y)\|_2}$$

$$w'_i = \alpha_i w_i + 1 \beta_i \quad \text{线性变化}$$

$$\bar{w}'_i = \alpha_i \bar{w}_i + 1 \beta_i$$

$$w'_i = \alpha_i w_i + 1 \beta_i$$

$$\therefore NCC'(x, y, d) = \frac{(w'_L(x, y) - \bar{w}'_L(x, y))^T}{\|w'_L(x, y) - \bar{w}'_L(x, y)\|_2} \cdot \frac{(w'_R(x-d, y) - \bar{w}'_R(x-d, y))}{\|w'_R(x-d, y) - \bar{w}'_R(x-d, y)\|_2}$$

$$= \frac{\alpha_1 (w_L(x, y) - \bar{w}_L(x, y))^T}{\alpha_1 \|w_L(x, y) - \bar{w}_L(x, y)\|_2} \cdot \frac{\alpha_2 (w_R(x-d, y) - \bar{w}_R(x-d, y))}{\alpha_2 \|w_R(x-d, y) - \bar{w}_R(x-d, y)\|_2}$$

$$= NCC(x, y, d)$$

b)

3x3 窗口

$$SSD(x, y, d) = \|W_L(x, y) - W_R(x-d, y)\|_2^2$$

$$d=0$$

$$SSD(x, y, 0) = 3 \times (10-5)^2 + 3 \times (5-6)^2 + 3 \times (6-7)^2 \\ = 3 \times (25 + 1 + 1) = 3 \times 27 = 81$$

$$d=1$$

$$SSD(x, y, 1) = 3 \times ((10-4)^2 + (5-5)^2 + (6-6)^2) \\ = 3 \times 36 = 108$$

$$d=2$$

$$SSD(x, y, 2) = 3 \times ((10-1)^2 + (5-6)^2 + (6-5)^2) \\ = 3 \times 2 = 6$$

$\therefore d=2$  为 best match

c)

left  
Cyan 坐标为 (2, 6),  $d=1 \therefore \text{right} \Rightarrow (2, 6-1) = (2, 5)$ ,

right (2, 5),  $d=1 \therefore \text{left} \Rightarrow (2, 5+1) = \underline{(2, 6)}$  - 取

Green, left (3, 4),  $d=2 \rightarrow \text{right} (3, 4-2) = (3, 2)$

right (3, 2),  $d=2 \therefore \text{left} \rightarrow (3, 2+2) = \underline{(3, 4)}$  - 取

Red, left (5, 3),  $d=2 \therefore \text{right} (5, 3-2) = (5, 1)$

right (5, 1),  $d=2 \therefore \text{left} (5, 1+2) = (5, 3)$  - 取

Test 没有成功 不能定错误估计

right 最左边一列出错了

## Learned Stereo and End-to-End Models

a)

$$N = \frac{(W - F + 2P)}{S} + 1$$

W: 输入大小  
F: kernel 大小  
P: 补1  
S: 步长.

$$\text{params} = \text{channels}_{\text{out}} \times (\text{kernel}_w \times \text{kernel}_h \times \text{channels}_{\text{in}}) + \text{channels}_{\text{out}}$$

$$\text{param\_Memory} = \text{params} \times 4 / 1024 / 1024$$

$$\text{activate\_Memory} = \text{out}_w \times \text{out}_h \times \text{out}_{\text{channel}} \times 4$$

Layer	Input shape	Output shape	#Trainable Parameters	Memory
Conv2D (32, 64, 3)	(32, 128, 128)	(64, 128, 128)	$64 \times (3 \times 3 \times 32) + 64$	
Conv2D (64, 128, 3)	(64, 128, 128)	(128, 128, 128)	$128 \times (3 \times 3 \times 64) + 128$	12.352 MB
Conv3D (1, 64, 3)	(1, 32, 128, 128)	(64, 32, 128, 128)	$64 \times 3 \times 3 \times 3 \times 1 + 64$	
Conv3D (64, 128, 3)	(64, 32, 128, 128)	(128, 32, 128, 128)	$128 \times 3 \times 3 \times 3 \times 64 + 128$	391.10 MB

$$2D: \quad (64 \times 128 \times 128 + 128 \times 128 \times 128) \times 4 = 12952320 \text{ Byte.}$$

$$+ \text{param\_Memory} = 12.3522 \text{ MB}$$

$$3D: \quad (64 \times 32 \times 128 \times 128 + 128 \times 32 \times 128 \times 128) \times 4 = 410099200 \text{ Byte}$$

$$+ (64 \times 3^3 + 64 + 128 \times 3^3 \times 64 + 128) \times 4 = 391.10 \text{ MB}$$

c)

对于  $P_1$ ,

$$\sigma(-C_\theta) = [0.44, 0.06, 0.00, 0.06, 0.44]$$

$$E(d) = 0 \times 0.44 + 1 \times 0.06 + 2 \times 0 + 3 \times 0.06 + 4 \times 0.44 = 2$$

对于  $P_2$

$$\sigma(-C_\theta) = [0, 0.21, 0.58, 0.21, 0]$$

$$E(d) = 0 \times 0 + 1 \times 0.21 + 2 \times 0.58 + 3 \times 0.21 + 4 \times 0 = 2$$