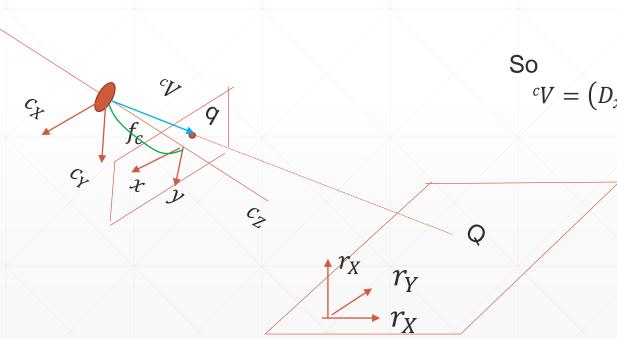
# Visual sensory II

#### Contents.

- 1. Binocular vision
- 2. Continuous Epipolar line constraint

Walking direction (vision guided walking)



For a projection q w.r.t camera frame

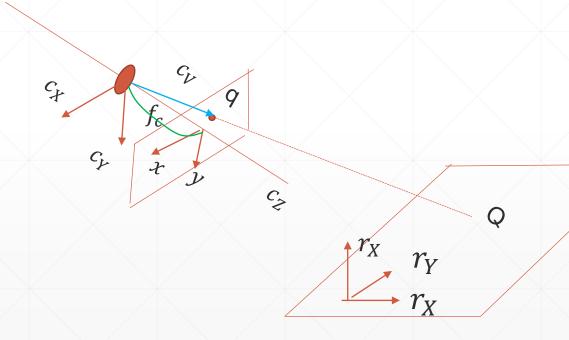
$$^{c}V = (x, y, f_{c})$$

At index coordinates

$$x = D_x(u - u_0)$$
  
$$y = D_y(v - v_0)$$

$$^{c}V = (D_{x}(u - u_{0}), D_{y}(v - v_{0}), f_{c})$$

Walking direction (vision guided walking)



For a projection q w.r.t camera frame  ${}^{c}V = (x, y, f_{c})$ 

At index coordinates

$$x = D_x(u - u_0)$$
$$y = D_y(v - v_0)$$

Normalizing by  $f_c$ 

$${}^{c}V = \left(\frac{D_{x}(u - u_{0})}{f_{c}}, \frac{D_{y}(v - v_{0})}{f_{c}}, 1\right)$$

$$= \left(\frac{(u - u_{0})}{f_{x}}, \frac{(v - v_{0})}{f_{y}}, 1\right)$$

Where

$$f_x = \frac{f_c}{D_x}$$
,  $f_y = \frac{f_c}{D_y}$ 

And also

$${}^{r}V = {}^{r}M_{c}{}^{c}V = ({}^{c}M_{r})^{-1}{}^{c}V$$

### Binocular vision

for a point A

$$s_1 \begin{pmatrix} u \\ v \\ 1 \end{pmatrix} = H \begin{pmatrix} rX \\ rY \\ rZ \\ 1 \end{pmatrix}$$

4 knowns, 3 constraints:

Need one more constraint to solve inverse projective-mapping problem

Way to solve:

If displacement vector is known

$$s_{2} \begin{pmatrix} u + \Delta u \\ v + \Delta v \\ 1 \end{pmatrix} = H \begin{pmatrix} rX + \Delta X \\ rY + \Delta Y \\ rZ + \Delta Z \\ 1 \end{pmatrix}$$

5 knowns, 6 constraints

- => Point A can be uniquely determined
- => Motion stereo, dynamic monocular vision

Alternatives: Placing two cameras at two different locations

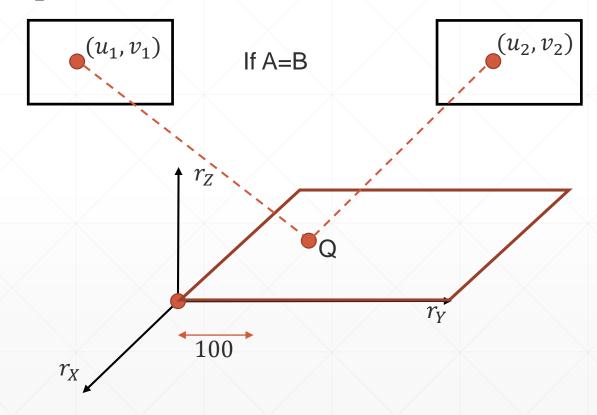
=> Binocular vision

for a point A

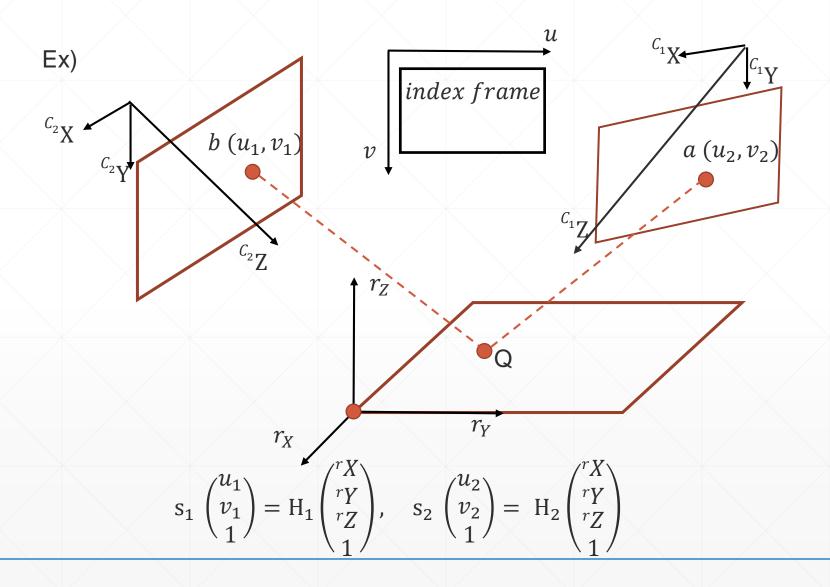
$$\mathbf{s}_1 \begin{pmatrix} u_1 \\ v_1 \\ 1 \end{pmatrix} = \mathbf{H}_1 \begin{pmatrix} {}^r X \\ {}^r Y \\ {}^r Z \\ 1 \end{pmatrix}$$

for a point B

$$s_2 \begin{pmatrix} u_2 \\ v_2 \\ 1 \end{pmatrix} = H_2 \begin{pmatrix} rX \\ rY \\ rZ \\ 1 \end{pmatrix}$$



#### 1) Forward projective-mapping



(Forward projective-mapping)

$$\mathbf{s}_{1} \begin{pmatrix} u_{1} \\ v_{1} \\ 1 \end{pmatrix} = \mathbf{H}_{1} \begin{pmatrix} {}^{r}X \\ {}^{r}Y \\ {}^{r}Z \\ 1 \end{pmatrix} \quad , \quad \mathbf{s}_{2} \begin{pmatrix} u_{2} \\ v_{2} \\ 1 \end{pmatrix} = \mathbf{H}_{2} \begin{pmatrix} {}^{r}X \\ {}^{r}Y \\ {}^{r}Z \\ 1 \end{pmatrix}$$

Given  $H_1$ , and  $H_2$ , if  $\begin{pmatrix} r_X \\ r_Y \\ r_Z \\ 1 \end{pmatrix}$  is known then  $(u_1, v_1)$  and  $(u_2, v_2)$  can be determined.

#### Inverse projective mapping

given 
$$a = (u_1, v_1)$$
,  $b = (u_1, v_1)$ 

$$determine Q = \begin{pmatrix} rX \\ rY \\ rZ \end{pmatrix}$$

$$H_{1} = \begin{pmatrix} i_{11} & i_{12} & i_{13} & i_{14} \\ i_{21} & i_{22} & i_{23} & i_{24} \\ i_{31} & i_{32} & i_{33} & 1 \end{pmatrix} , H_{2} = \begin{pmatrix} j_{11} & j_{12} & j_{13} & j_{14} \\ j_{21} & j_{22} & j_{23} & j_{24} \\ j_{31} & j_{32} & j_{33} & 1 \end{pmatrix}$$

By 
$$s_1 \begin{pmatrix} u_1 \\ v_1 \\ 1 \end{pmatrix} = H_1 \begin{pmatrix} rX \\ rY \\ rZ \\ 1 \end{pmatrix}$$

Thus  $s_{1}u_{1} = i_{11}^{r}X + i_{12}^{r}Y + i_{13}^{r}Z + i_{14}$   $s_{1}v_{1} = i_{21}^{r}X + i_{22}^{r}Y + i_{23}^{r}Z + i_{24}$   $s_{1} = i_{31}^{r}X + i_{32}^{r}Y + i_{33}^{r}Z + 1$ 

Eliminating  $s_1$ 

$$(i_{11} - i_{31}u_1)^r X + (i_{12} - i_{32}u_1)^r Y + (i_{13} - i_{33}u_1)^r Z = u_1 - i_{14}$$
  

$$(i_{21} - i_{31}v_1)^r X + (i_{22} - i_{32}v_1)^r Y + (i_{23} - i_{33}v_1)^r Z = v_1 - i_{24}$$

Also, for camera 2

$$(j_{11} - j_{31}u_2)^r X + (j_{12} - j_{32}u_2)^r Y + (j_{13} - j_{33}u_2)^r Z = u_2 - j_{14}$$
  
$$(j_{21} - j_{31}v_2)^r X + (j_{22} - j_{32}v_2)^r Y + (j_{23} - j_{33}v_2)^r Z = v_2 - j_{24}$$

Let 
$$A = \begin{pmatrix} i_{11} - i_{31}u_1 & i_{12} - i_{32}u_1 & i_{13} - i_{33}u_1 \\ i_{21} - i_{31}v_1 & i_{22} - i_{32}v_1 & i_{23} - i_{33}v_1 \\ j_{11} - j_{31}u_2 & j_{12} - j_{32}u_2 & j_{13} - j_{33}u_2 \\ j_{21} - j_{31}v_2 & j_{22} - j_{32}v_2 & j_{23} - j_{33}v_2 \end{pmatrix}$$

Then

$$A \binom{rX}{rY}_{rZ} = \binom{u_1 - i_{14}}{v_1 - i_{24}}_{u_2 - j_{14}}$$

$$v_2 - j_{24}$$

*Therefore* 

$$X = \begin{pmatrix} {}^{r}X \\ {}^{r}Y \\ {}^{r}Z \end{pmatrix} = (A^{T}A)^{-1}A^{T}B \quad with \quad B = \begin{pmatrix} u_{1} - l_{14} \\ v_{1} - l_{24} \\ u_{2} - j_{14} \\ v_{2} - j_{24} \end{pmatrix}$$

E.g. Given two projections, a, and b  $a = (300,200) = (u_1, v_1), \quad b = (200,200) = (u_2, v_2)$ And  $H_1 = \begin{pmatrix} 2.74 & 1.89 & -0.33 & -18.45 \\ 0 & 1.410 & -3.03 & 748.91 \\ 0 & 0.007 & -0.0013 & 1 \end{pmatrix}, H_2 = \begin{pmatrix} 2.74 & 1.89 & -0.33 & -18.45 \\ 0 & 1.410 & -3.03 & 748.91 \\ 0 & 0.007 & -0.0013 & 1 \end{pmatrix}$ Then  $\begin{pmatrix} 2.74 & -0.325 & 0.057 \end{pmatrix} \begin{pmatrix} 318.4 \end{pmatrix}$ 

$$A = \begin{pmatrix} 2.74 & -0.325 & 0.057 \\ 0 & -0.062 & -2.79 \\ 2.74 & 0.414 & -0.073 \\ 0 & -0.062 & -2.776 \end{pmatrix} , B = \begin{pmatrix} 318.4 \\ -548.9 \\ 438.0 \\ -548.91 \end{pmatrix}$$

$$X = \begin{pmatrix} {}^{r}X \\ {}^{r}Y \\ {}^{r}Z \end{pmatrix} = (A^{T}A)^{-1}A^{T}B = \begin{pmatrix} 135.2 \\ 195.8 \\ 193.3 \end{pmatrix}$$

#### Issues for Binocular vision

- a. Search for its corresponding point in other image
  - => Verify all possible locations in other image
- b. Different parameters for each camera must be calibrated in advance: intrinsic and extrinsic parameters

- -Limiting the search for the correspondence to an interval along a line
- -Reduction of search space

Let the index coordinates of two feature point be

$$a = (u_1, v_1)$$
  
 $b = (u_2, v_2)$ 

Walking direction in camera 1:

Projection line passing through point a:

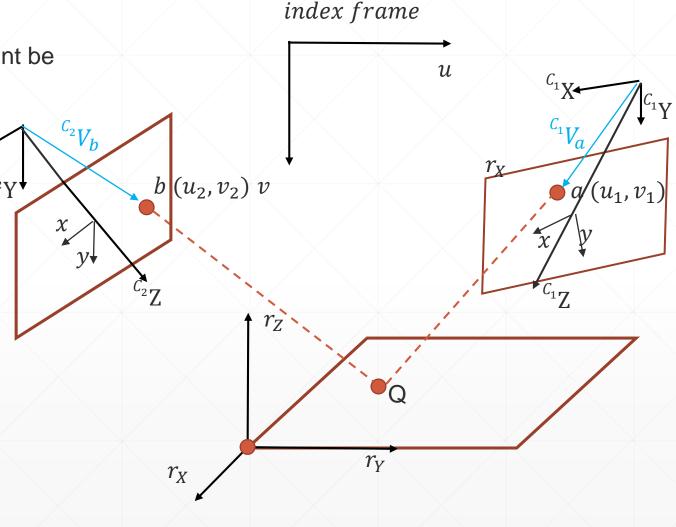
$$^{C_1}V_a = \left(\frac{u_1 - u_{1,0}}{f_{1,x}}, \frac{v_1 - v_{1,0}}{f_{1,y}}, 1\right)$$

Where

$$f_{1,x} = \frac{f_{c1}}{D_{1,x}}$$
  $f_{1,y} = \frac{f_{c1}}{D_{1,y}}$ 

Or

$$C_{1}V_{a} = \begin{pmatrix} \frac{1}{f_{1,x}} & 0 & -\frac{u_{1,0}}{f_{1,x}} \\ 0 & \frac{1}{f_{1,y}} & -\frac{v_{1,0}}{f_{1,y}} \end{pmatrix} \begin{pmatrix} u_{1} \\ v_{1} \\ 1 \end{pmatrix}$$



#### Continuous Epipolar line constraint

Walking direction in camera 2: Projection line passing through point b:

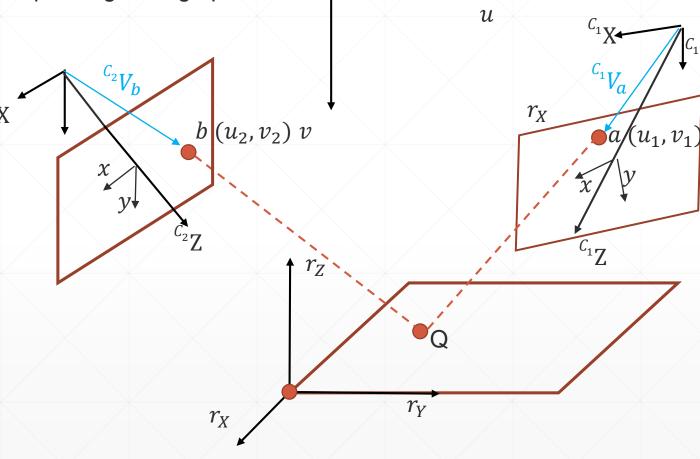
$$C_2V_b = \left(\frac{u_2 - u_{2,0}}{f_{2,x}}, \frac{v_2 - v_{2,0}}{f_{2,y}}, 1\right)$$

Where

$$f_{2,x} = \frac{f_{c2}}{D_{2,x}}$$
  $f_{2,y} = \frac{f_{c2}}{D_{2,y}}$ 

Or

index frame



If a, b is projection of Q,  ${}^{C_1}V_a$  and  ${}^{C_2}V_b$  will intersect. Transformation between left camera and right camera

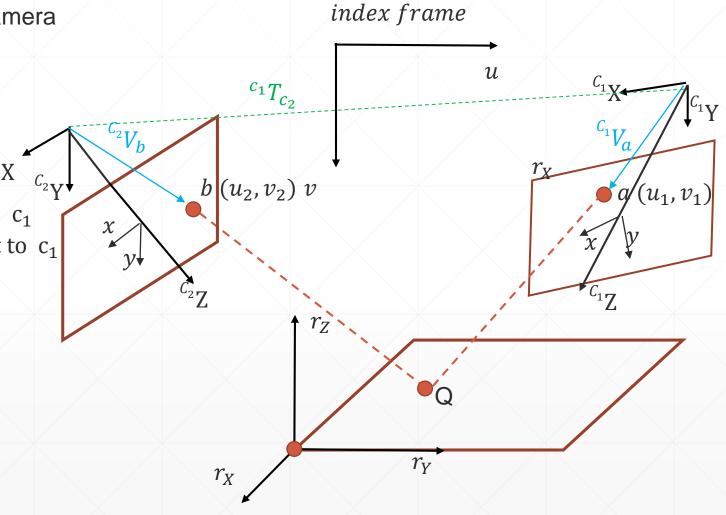
$${}^{c_1}M_{c_2} = \begin{bmatrix} {}^{c_1}R_{c_2} & {}^{c_1}T_{c_2} \\ 000 & 1 \end{bmatrix}$$

#### Where

 $^{c_1}R_{c_2}$  : rotation of camera frame  $c_2$  with respect to  $\,c_1$ 

 $c_1 T_{c_2}$ : translation of camera  $c_2$  origin with respect to  $c_1$ 

$${}^{c_1}V_b = {}^{c_1}T_{c_2} + {}^{c_1}R_{c_2} {}^{c_2}V_b$$
  
$$\triangleq {}^{c_1}T_{c_2} + {}^{c_1}\hat{V}_b$$



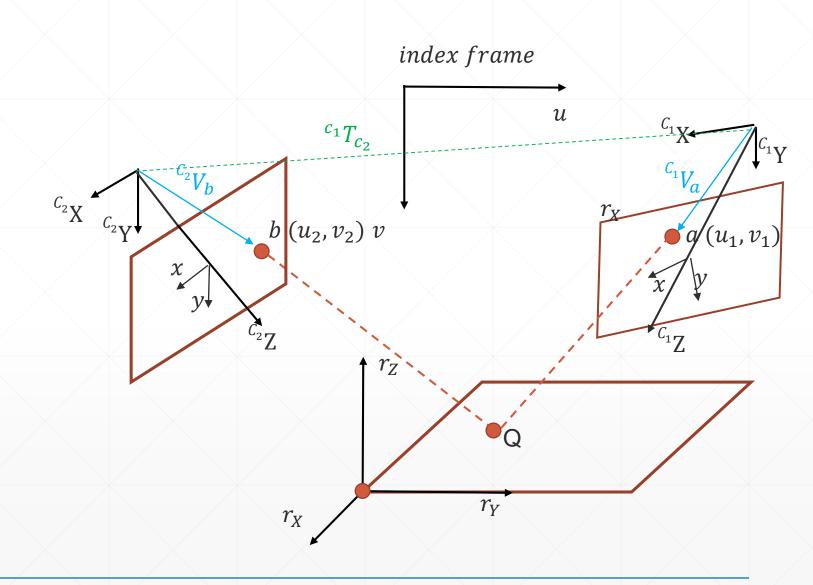
$$if , c_1 T_{c_2} = 0 (ideal case)$$

a and b will be superimposed So

$${}^{c_1}V_a = {}^{c_1}V_b = {}^{c_1}R_{c_2} {}^{c_2}V_b$$

*Therefore* 

$$c_1 V_a = \begin{pmatrix} \frac{1}{f_{1,x}} & 0 & -\frac{u_{1,0}}{f_{1,x}} \\ 0 & \frac{1}{f_{1,y}} & -\frac{v_{1,0}}{f_{1,y}} \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} u_1 \\ v_1 \\ 1 \end{pmatrix}$$



$${}^{c_1}V_a = {}^{c_1}V_b = {}^{c_1}R_{c_2} {}^{c_2}V_b$$

**Therefore** 

Finally

$$\begin{pmatrix} u_1 \\ v_1 \\ 1 \end{pmatrix} = M_{3\times 3} \begin{pmatrix} u_2 \\ v_2 \\ 1 \end{pmatrix}$$

Where

$$M_{3\times3} = \begin{pmatrix} \frac{1}{f_{1,x}} & 0 & -\frac{u_{1,0}}{f_{1,x}} \\ 0 & \frac{1}{f_{1,y}} & -\frac{v_{1,0}}{f_{1,y}} \\ 0 & 0 & 1 \end{pmatrix}^{-1} \begin{pmatrix} \frac{1}{f_{2,x}} & 0 & -\frac{u_{2,0}}{f_{2,x}} \\ 0 & \frac{1}{f_{2,y}} & -\frac{v_{2,0}}{f_{2,y}} \\ 0 & 0 & 1 \end{pmatrix}$$

$$practically \\ {}^{c_1}T_{c_2} \neq 0$$

if 
$${}^{c_1}V_a$$
,  ${}^{c_1}\hat{V}_b$ ,  ${}^{c_1}T_{c_2}$  are coplanar  
 $\Rightarrow$  have intersection between ( ${}^{c_1}V_a$ ,  ${}^{c_2}V_b$ )

normal vector of plane N

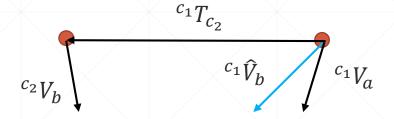
$$c_1 N = {}^{c_1} T_{c_2} \times {}^{c_1} V_a$$

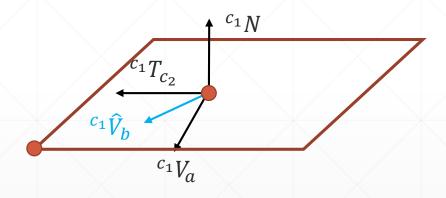
$${}^{c_1} N \perp {}^{c_1} \hat{V}_b$$

$$\Rightarrow ({}^{c_1} N)^T {}^{c_1} \hat{V}_b = 0$$

$$({}^{c_1} T_{c_2} \times {}^{c_1} V_a)^T {}^{c_1} \hat{V}_b = 0$$

$$let c_1 T_{c_2} = \left( t_x \ t_y \ t_z \right)^T$$



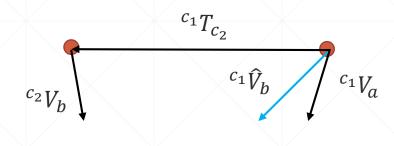


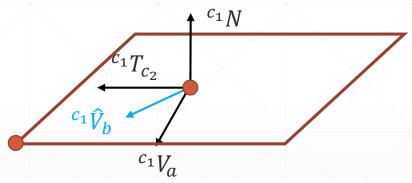
By skew symmetric property

$$S_{T} = S(c_{1}T_{c_{2}}) = \begin{pmatrix} 0 & -t_{z} & t_{y} \\ t_{z} & 0 & -t_{x} \\ -t_{y} & t_{x} & 0 \end{pmatrix}$$
$$(c_{1}T_{c_{2}} \times c_{1}V_{a})^{T} c_{1}\hat{V}_{b} = 0 \Rightarrow (S(c_{1}T_{c_{2}}) c_{1}V_{a})^{T} c_{1}\hat{V}_{b} = 0$$

hence, 
$$({}^{c_1}V_a)^T S_T^{\ T\ c_1} \hat{V}_b = 0$$

skew symmetric property  $a \times p = S(a)p$ 

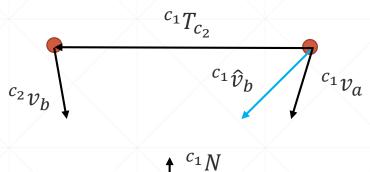


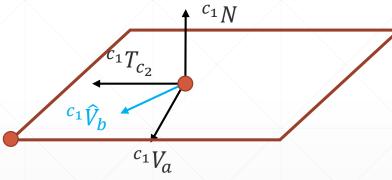


$$({}^{c_1}V_a)^T S_T^{\ T\ c_1} \hat{V}_b = 0$$

$$\begin{bmatrix} \left(\frac{1}{f_{1,x}} & 0 & -\frac{u_{1,0}}{f_{1,x}} \\ 0 & \frac{1}{f_{1,y}} & -\frac{v_{1,0}}{f_{1,y}} \\ 0 & 0 & 1 \end{bmatrix} \begin{pmatrix} u_1 \\ v_1 \\ 1 \end{bmatrix} \end{bmatrix}^T (S_T)^{T} c_1 R_{c_2} c_2 V_b = 0$$

$$\begin{pmatrix} \frac{1}{f_{2,x}} & 0 & -\frac{u_{2,o}}{f_{1,x}} \\ 0 & \frac{1}{f_{2,y}} & -\frac{v_{2,0}}{f_{2,y}} \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} u_2 \\ v_2 \\ 1 \end{pmatrix}$$





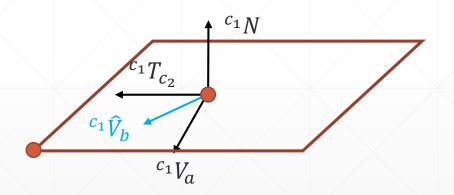
$$(u_1 \quad v_1 \quad 1) \begin{bmatrix} \frac{1}{f_{1,x}} & 0 & -\frac{u_{1,0}}{f_{1,x}} \\ 0 & \frac{1}{f_{1,y}} & -\frac{v_{1,0}}{f_{1,y}} \\ 0 & 0 & 1 \end{bmatrix}^T (S_T)^{T} c_1 R_{c_2} \begin{pmatrix} \frac{1}{f_{2,x}} & 0 & -\frac{u_{2,0}}{f_{2,x}} \\ 0 & \frac{1}{f_{2,y}} & -\frac{v_{2,0}}{f_{2,y}} \\ 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} u_2 \\ v_2 \\ 1 \end{pmatrix} = 0$$

$$c_1 v_a$$

 $F_{3\times 3}$ : Fundamental matrix

$$(u_1 \quad v_1 \quad 1)F_{3\times 3} \begin{pmatrix} u_2 \\ v_2 \\ 1 \end{pmatrix} = 0 \qquad \Rightarrow a u_2 + b v_2 + c = 0$$

Epipolar line constraint



E.g.

$$left\ cam: (f_{1,x}, f_{1,y}, u_{1,0}, v_{1,0}) = (365.6, 365.6, 256, 256)$$

$$c_1 M_r = \begin{pmatrix} 0.9848 & -0.17 & 0.03 & -121.61 \\ 0 & -0.173 & -0.984 & 179.59 \\ 0.173 & 0.969 & -0.171 & 113.82 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

right cam : 
$$(f_{2,x}, f_{2,y}, u_{2,0}, v_{2,0}) = (365.6, 365.6, 256, 256)$$

$$c_2 M_r = \begin{pmatrix} 0.9848 & 0.17 & -0.03 & -153.8 \\ 0 & -0.173 & -0.984 & 189.44 \\ -0.173 & 0.969 & -0.171 & 164.15 \\ 0 & 0 & 0 & 1 \end{pmatrix}$$

Eg

$$c_1 M_{c_2} = {}^{c_1} M_r {}^r M_{c_2} = {}^{c_1} M_r ({}^{c_2} M_r)^{-1} = \begin{pmatrix} {}^{c_1} R_{c_2} & {}^{c_1} T_{c_2} \\ \hline 000 & 1 \end{pmatrix}$$

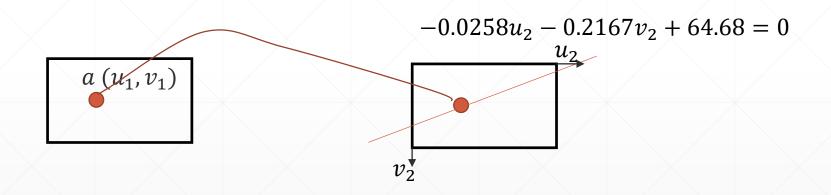
$$F_{3\times3} = \begin{pmatrix} \frac{1}{f_{1,x}} & 0 & -\frac{u_{1,0}}{f_{1,x}} \\ 0 & \frac{1}{f_{1,y}} & -\frac{v_{1,0}}{-f_{1,y}} \\ 0 & 0 & 1 \end{pmatrix}^{T} (S_{T})^{T} c_{1} R_{c_{2}} \begin{pmatrix} \frac{1}{f_{2,x}} & 0 & -\frac{u_{2,o}}{f_{1,x}} \\ 0 & \frac{1}{f_{2,y}} & -\frac{v_{2,0}}{-f_{2,y}} \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} 0 & 0.0001 & -0.0045 \\ 0.0001 & 0 & 0.1848 \\ -0.061 & -0.239 & 19.06 \end{pmatrix}$$

$$(u_1 \quad v_1 \quad 1) \ F_{3\times 3} \begin{pmatrix} u_2 \\ v_2 \\ 1 \end{pmatrix} = 0$$

$$if(u_1, v_1) = (252, 253) \Rightarrow -0.0258u_2 - 0.2167v_2 + 64.68 = 0$$

Eg

$$(u_1 \quad v_1 \quad 1) F_{3\times 3} \begin{pmatrix} u_2 \\ v_2 \\ 1 \end{pmatrix} = 0$$
  
 $if(u_1, v_1) = (252, 253) \Rightarrow -0.0258u_2 - 0.2167v_2 + 64.68 = 0$ 



#### **Correspondence problems:**

- 1. Issues
  - i) Select candidate match
  - ii) Determine the goodness of the match
- 2. Correspondence algorithm
  - i) Correlation based
    - -By matching image intensity over a window of pixels
  - ii) Feature based
    - -By matching a sparse sets of image features such as edges, lines, etc
- 3. Correlation based algorithm Inputs:
  - a) Left camera image intensity:  $I_l$  Right camera image intensity:  $I_r$
  - b) Width of sub window : 2w + 1
  - c) Search region in the right image  $R(P_l)$  associated with a pixel  $p_l$  in left image

#### **Correspondence problems:**

For each pixel  $P_l(i,j)$  in the left image

- a) for each displacement  $d = (d_1, d_2) \in R(P_l)$  compute cross-correlation C(d)
- $C(d) = \sum_{k=-w}^{w} \sum_{l=-w}^{w} I_l (i+k,j+l) I_r (i+k-d_1,j+l-d_2)$
- b) select  $\bar{d} = (\bar{d}_1, \bar{d}_2)$  that maximizes C(d) over  $R(P_l)$

Other forms of C(d)

$$C(d) = \sum_{k=-w}^{w} \sum_{l=-w}^{w} [I_l(i+k,j+l) - I_r(i+k-d_1,j+l-d_2)]^2$$
or
$$C(d) = \sum_{k=-w}^{w} \sum_{l=-w}^{w} |I_l(i+k,j+l) - I_r(i+k-d_1,j+l-d_2)|$$

- 4. Feature based method
  - a) select for a feature in an image that matches a feature in the other image
  - b) Typical features
    - -edge points
    - -line segments
    - -corners

#### **Correspondence problems:**

- 4. Feature based method
- e.g. Line feature descriptor
  - -length *l*
  - -orientation θ
  - -mid point m
  - -average intensity along line i

#### Similarity function:

$$S = \frac{1}{w_0 (l_l - l_r)^2 + w_1(\theta_l - \theta_r)^2 + w_2(m_l - m_r)^2 + w_3(i_l - i_r)^2}$$
   
  $w_0$ ,  $w_1$ ,  $w_2$ ,  $w_3$ : weighting factors Select maximum value of S to find a feature.

# Discrete Epiploar line constraint

-one coordinate varies within a fixed range

Assume Z from the object varies within  $[Z_{min}, Z_{max}]$ 

$$Z_i = Z_{min} + i \Delta Z, i \in [0, n]$$

 $\Delta Z$ : accuracy

By Binocular vision

$$s_1 {v_1 \choose 1} = H_1 {r_X \choose r_Y \choose Z_i = Z_{min} + i \Delta Z}, \qquad i = 0,1,2, \qquad \dots n$$

Also

$$s_2 {u_2 \choose v_2} = H_2 {x_1 \choose x_i} = H_2 {x_1 \choose x_i} = Z_{min} + i \Delta Z, \qquad i = 0,1,2, \qquad \dots n$$

**Procedures** 

Step1: Specify the range and accuracy of one coordinate

Step 2: Setup binocular vision equation

# Discrete Epiploar line constraint

**Procedures** 

Step 3: From

$$s_1 \begin{pmatrix} u_1 \\ v_1 \\ 1 \end{pmatrix} = H_1 \begin{pmatrix} r_X \\ r_Y \\ Z_i = Z_{min} + i \Delta Z \end{pmatrix}, i = 0,1,2, \dots n$$

Compute the predicted coordinates  $({}^{r}X_{i}, {}^{r}Y_{i}, {}^{r}Z_{i})$  by  $(u_{1}, v_{1})$  and given  ${}^{r}Z_{i}$ 

Step 4: Given  $({}^{r}X_{i}, {}^{r}Y_{i}, {}^{r}Z_{i})$ , compute  $(u_{2i}, v_{2i})$  from

$$s_2 \begin{pmatrix} u_{2i} \\ v_{2i} \\ 1 \end{pmatrix} = H_2 \begin{pmatrix} rX_i \\ rY_i \\ rZ_i \\ 1 \end{pmatrix}, \quad i = 0,1,2, \dots n$$

Step 5: Compute the intensity difference between at  $(u_1 \ v_1)$  and  $(u_{2i} \ v_{2i})$   $\Delta I_i = I(u_1 \ v_1) - I(u_{2i} \ v_{2i})$ : Dissimilarity

Step 6: Repeat steps 3,4,5 for all i

Step 7: Choose the location  $(u_{2j} \ v_{2j})$  where the dissimilarity is minimum

$$\left[ \left( u_{2j} \ v_{2j} \right) | \ \min_{i} \Delta I_{i} \right]$$

# Discrete Epiploar line constraint

$$\begin{bmatrix} (u_{2j} \ v_{2j}) | \ \min_i \Delta I_i \end{bmatrix}$$
 
$$(u_{2j} \ v_{2j}) \text{ is the correspondence}$$
 
$$(u_{2j} \ v_{2j}) \text{ exists on an Epipolar line}$$

 $\Rightarrow$  ( ${}^{r}X_{j}$ ,  ${}^{r}Y_{j}$ ,  ${}^{r}Z_{j}$ ) is the 3D geometry of the binocular vision's inverse projective-mapping

$$\Rightarrow \text{No need to use } X = \begin{pmatrix} rX \\ rY \\ rZ \end{pmatrix} = (A^T A)^{-1} A^T B$$

$$\begin{pmatrix} i_{11} - i_{31} u_1 & i_{12} - i_{32} u_1 & i_{13} - i_{33} u_1 \\ i_{21} - i_{31} v_1 & i_{22} - i_{32} v_1 & i_{23} - i_{33} v_1 \end{pmatrix} \qquad \begin{pmatrix} u_1 - i_{14} \\ v_1 - i_{24} \end{pmatrix}$$

$$A = \begin{pmatrix} i_{11} - i_{31}u_1 & i_{12} - i_{32}u_1 & i_{13} - i_{33}u_1 \\ i_{21} - i_{31}v_1 & i_{22} - i_{32}v_1 & i_{23} - i_{33}v_1 \\ j_{11} - j_{31}u_2 & j_{12} - j_{32}u_2 & j_{13} - j_{33}u_2 \\ j_{21} - j_{31}v_2 & j_{22} - j_{32}v_2 & j_{23} - j_{33}v_2 \end{pmatrix}, B = \begin{pmatrix} u_1 - i_{14} \\ v_1 - i_{24} \\ u_2 - j_{14} \\ v_2 - j_{24} \end{pmatrix}$$

Calibration matrices

$$H_{1} = \begin{pmatrix} i_{11} & i_{12} & i_{13} & i_{14} \\ i_{21} & i_{22} & i_{23} & i_{24} \\ i_{31} & i_{32} & i_{33} & 1 \end{pmatrix} , H_{2} = \begin{pmatrix} j_{11} & j_{12} & j_{13} & j_{14} \\ j_{21} & j_{22} & j_{23} & j_{24} \\ j_{31} & j_{32} & j_{33} & 1 \end{pmatrix}$$