

CV HW #4

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## 1.1 Camera Models

(a) for only one pair  $(x_i, y_i, z_i)$  and  $(u_i, v_i)$ ,

$$k \begin{pmatrix} x_i \\ y_i \\ z_i \end{pmatrix} = \begin{pmatrix} k_{11}x_i + k_{12}y_i + k_{13}z_i \\ k_{21}x_i + k_{22}y_i + k_{23}z_i \\ k_{31}x_i + k_{32}y_i + k_{33}z_i \end{pmatrix} = \begin{pmatrix} u_i \\ v_i \\ 1 \end{pmatrix}$$

$$\therefore k_{11}x_i + k_{12}y_i + k_{13}z_i = k_{33}u_i z_i$$

$$k_{21}x_i + k_{22}y_i + k_{23}z_i = k_{33}v_i z_i$$

$$\begin{bmatrix} x_i & y_i & z_i & 0 & 0 & -u_i z_i \\ 0 & 0 & 0 & y_i & z_i & -v_i z_i \end{bmatrix} \begin{bmatrix} k_{11} \\ k_{12} \\ k_{13} \\ k_{21} \\ k_{22} \\ k_{23} \\ k_{33} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}$$

for multiple pairs, ( $n$  pairs)

$$\begin{bmatrix} x_1 & y_1 & z_1 & 0 & 0 & -u_1 z_1 \\ 0 & 0 & 0 & y_1 & z_1 & -v_1 z_1 \\ \vdots & & & \vdots & & \vdots \\ x_n & y_n & z_n & 0 & 0 & -u_n z_n \\ 0 & 0 & 0 & y_n & z_n & -v_n z_n \end{bmatrix} \begin{bmatrix} k_{11} \\ k_{12} \\ k_{13} \\ k_{21} \\ k_{22} \\ k_{23} \\ k_{33} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix}$$

$2n \times 6 (A)$

$6 \times 1 (X)$

$6 \times 1$

its form of  $Ax = 0$

$$(b) \begin{bmatrix} x_1 & y_1 & z_1 & 0 & 0 \\ 0 & 0 & 0 & y_1 & z_1 \\ \vdots & & & & \\ x_n & y_n & z_n & 0 & 0 \\ 0 & 0 & 0 & y_n & z_n \end{bmatrix} \begin{bmatrix} k_{11} \\ k_{12} \\ k_{13} \\ k_{22} \\ k_{23} \end{bmatrix} = \begin{bmatrix} u_1 z_1 \\ v_1 z_1 \\ \vdots \\ u_n z_n \\ v_n z_n \end{bmatrix}$$

$\uparrow 2n \times 5 \quad 5 \times 1 \quad 2n \times 1$

$\rightarrow$  set this as  $A$   $b$

from least squares,

$$\begin{bmatrix} k_{11} \\ k_{12} \\ k_{13} \\ k_{22} \\ k_{23} \end{bmatrix} = (A^T A)^+ A^T b$$

## 1.2 Epipolar Geometry

$$(a) M = K \begin{pmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{pmatrix} = \begin{pmatrix} K & 0 \\ 0 & 0 \end{pmatrix}$$

$$M' = K' (R | t) = (K' R | K' t)$$

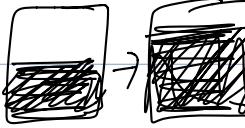
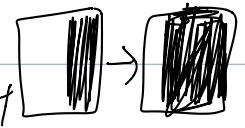
$$(b) e = M \begin{pmatrix} -R^T t \\ 1 \end{pmatrix} = -KR^T t$$

$$e' = M' \cdot \begin{pmatrix} 0 \\ 0 \\ 0 \\ 1 \end{pmatrix} = K' t$$

$$(c) e' \times x' = [e']_x x' = [e']_x (M' \begin{pmatrix} X \\ 1 \end{pmatrix}) = [e']_x (M' \begin{pmatrix} K' x \\ 1 \end{pmatrix})$$
$$= [e']_x (K' R K^{-1} x + K' t) \xrightarrow{\cancel{K' t}} 0$$
$$= [e']_x K' R K^{-1} x$$

(d) vector perpendicular to epipolar plane.

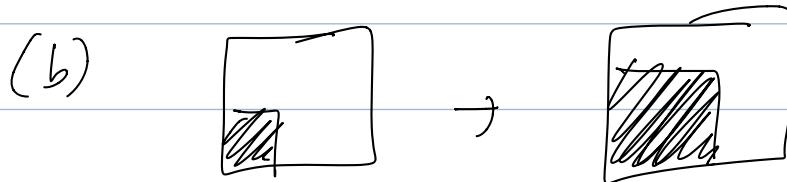
# 1.3 Optical flow

(a)  can't recover horizontal component  can't recover vertical component

$$\begin{bmatrix} \sum I_x^2 & \sum I_x I_y \\ \sum I_y I_x & \sum I_y^2 \end{bmatrix} \begin{bmatrix} u \\ v \end{bmatrix} = \begin{bmatrix} -\sum I_x I_t \\ -\sum I_y I_t \end{bmatrix}$$

2 unknowns  $u, v \Rightarrow$  need 2 indep equations to get  $u, v$ .

if  $\sum I_x^2 \sum I_y^2 - (\sum I_x I_y)^2 = 0$ ,  
cannot recover  $u$  or  $v$ .



Can recover both  $u$  and  $v$   
in this case

$\begin{bmatrix} \sum I_x^2 & \sum I_x I_y \\ \sum I_y I_x & \sum I_y^2 \end{bmatrix}$  is not singular because

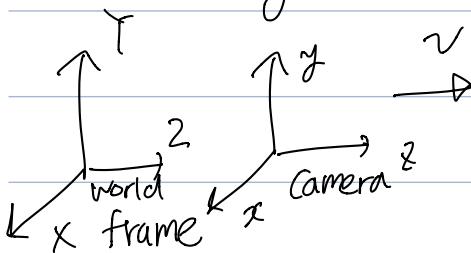
$$\sum I_x^2 \sum I_y^2 - (\sum I_x I_y)^2 \neq 0.$$

$\Rightarrow$  2 indep equations for 2 unknowns  $u, v$

(c) Optical flow equation still holds.

It is relevant to movement of pixel, not real motion.

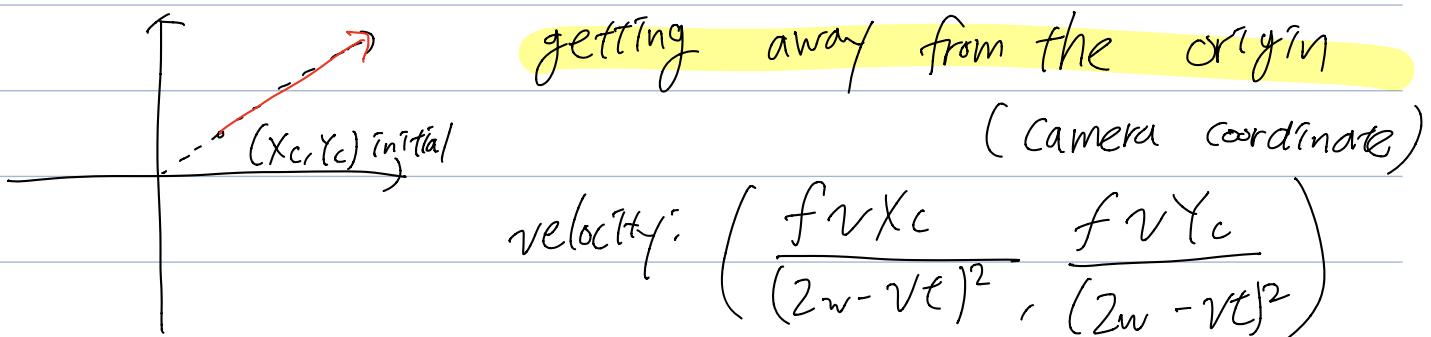
To figure out which scene point the object is moving toward,



- $(X_w, Y_w, Z_w)$  in world coordinate
- $(X_c, Y_c, Z_w - vt)$  in camera coordinate  
 $= (X_c, Y_c, Z_c)$

$$(x_c, y_c, z_w - vt)$$

$$\equiv \left( \frac{fx_c}{z_w - vt}, \frac{fy_c}{z_w - vt}, f \right)$$

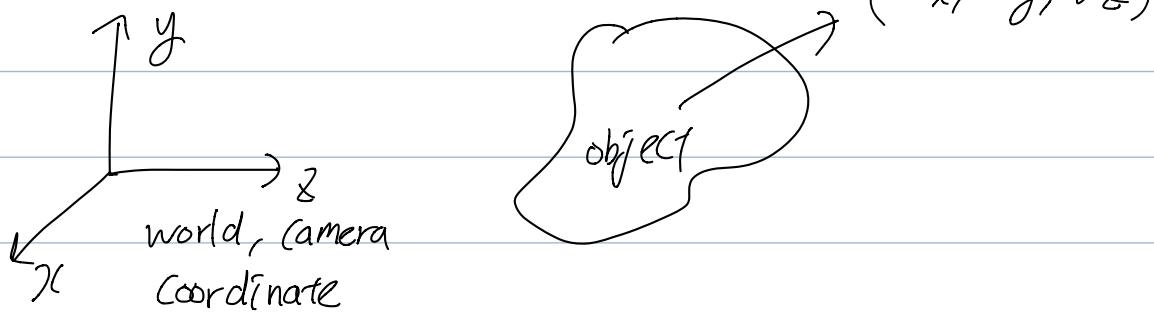


$$\text{velocity: } \left( \frac{fvx_c}{(z_w - vt)^2}, \frac{fvy_c}{(z_w - vt)^2} \right)$$

for image coordinate  $(u, v)$

origin corresponds to  $K \begin{pmatrix} 0 \\ 0 \\ 1 \end{pmatrix}$ , where  $K$  is intrinsic matrix.

(d) like (c)



let initial position of some point of the object

is  $(x, y, z) \equiv (fx/2, fy/2, f) := (x_i, y_i, f)$

at time  $t$ , its position is  $(x + v_x t, y + v_y t, z + v_z t)$

$$\equiv \left( \frac{(x + v_x t)f}{z + v_z t}, \frac{(y + v_y t)f}{z + v_z t}, f \right) = (x_c, y_c, f)$$

$$\frac{dx_c}{dt} = \frac{v_x f (z + v_z t) - (x + v_x t) v_z f}{(z + v_z t)^2} = \frac{(v_x z - v_z x) f}{(z + v_z t)^2}$$

$$\frac{dy_c}{dt} = \frac{(v_y z - v_z y)f}{(z + v_z t)^2}$$

Show that  $\begin{pmatrix} fv_x/v_z \\ fv_y/v_z \\ f \end{pmatrix} := \begin{pmatrix} x_2 \\ y_2 \\ f \end{pmatrix}$  in camera coordinate is the focus of expansion

$$\frac{dx_c}{dt} = \frac{(x_2 - x_1)v_z z}{(z + v_z t)^2} \quad \frac{dy_c}{dt} = \frac{(y_2 - y_1)v_z z}{(z + v_z t)^2}$$

$$\frac{dx}{dt} - \frac{dy}{dt} = x_2 - x_1 - y_2 - y_1$$

If  $x_2 = x_1$  and  $y_2 = y_1$ , its optical flow is 0.  
in image plane,

the focus of expansion corresponds to

$$K \begin{pmatrix} x_2 \\ y_2 \\ f \end{pmatrix}, \text{ where } K \text{ is intrinsic matrix}$$

## 2.1 dominant Motion Estimation

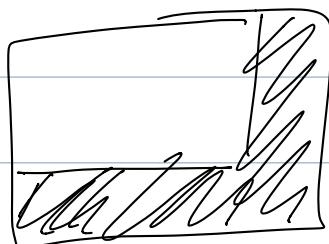
### Lukas-Kanade Method

$$\Delta p = H^{-1} \sum_x \left[ \nabla I \frac{\partial w}{\partial p} \right]^T [T(x) - I(w(x:p))]$$

where

$$\text{hessian } H = \sum_x \left[ \nabla I \frac{\partial w}{\partial p} \right] \left[ \nabla I \frac{\partial w}{\partial p} \right]^T$$

for element in sigma, only consider



this part of the image, where cars never appear.

(want  $\Delta p$  of the background, not cars)

and when calculating  $dI/dx$   $dI/dy$ ,  
we should divide  $G_x, G_y$  by 128,  
because when applying  $5 \times 5$  sobel filter

to

$$\begin{matrix} 1 & 2 & 3 & 4 & 5 \\ 1 & 2 & 3 & 4 & 5 \\ 1 & 2 & 3 & 4 & 5 \\ 1 & 2 & 3 & 4 & 5 \\ 1 & 2 & 3 & 4 & 5 \end{matrix}$$

we get 128, not 1  
(it should be 1)

## 2.2 Moving Object Detection

we update global  $P$  by

$$P(t + \Delta t) = P(t) + \Delta P$$

and check absolute value of pixel value difference between  $\text{img2}$  and warped  $\text{img1}$  (warped by  $P$ ).

thresholds:

high:  $0.2 * 256$

low :  $0.15 * 256$