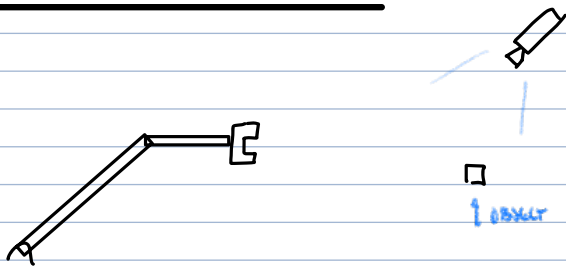


20 VISUAL SERVING



"EYE-TO-HAND"

EYE-IN-HAND

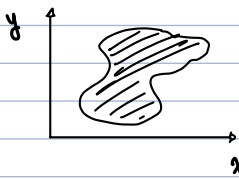
CAMERA ON ROBOTIC HAND

EYE-IN-HAND		VS	EYE-TO-HAND	
PROS	CONS		PROS	CONS
• NO OCCLUSIONS	• VARIABILITY IN ACCURACY		• FIELD OF VIEW + ACCURACY DON'T CHANGE	• OCCLUSIONS FROM ROBOT
	• CAN'T OBSERVE END EFFECTOR			

IMAGE PROCESSING

1. IMAGE SEGMENTATION: RECOGNIZING OBJECT

A. REGION BASED - E.G. BINARY SEGMENTATION



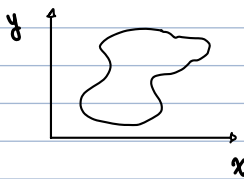
GRAYSCALE IMAGE

↓
APPLY THRESHOLD

↓
BINARY IMAGE

$$b(x,y) = \begin{cases} 1 & \text{OBJECT } g \geq g_0 \\ 0 & \text{BACKGROUND } g < g_0 \end{cases}$$

B. BOUNDARY-BASED (E.G. EDGE DETECTION)

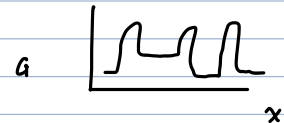


COMPUTE GRADIENT

$$\text{x-GRADIENT } \Delta_1 = g(x+1, y) - g(x, y)$$

$$\text{y-GRADIENT } \Delta_2 = g(x, y+1) - g(x, y)$$

$$\text{MAGNITUDE: } G(x,y) = \sqrt{\Delta_1^2 + \Delta_2^2}$$



$$b(x,y) = \begin{cases} 1 & \text{EDGE } G \geq G_0 \\ 0 & \text{NO EDGE } G < G_0 \end{cases}$$

2. IMAGE INTERPRETATION: COMPUTE FEATURE PARAMETERS

SIMPLE GEOMETRIC FEATURES:

$$\text{• AREA } A = \sum_x \sum_y b(x,y)$$

$$\text{• CENTROID (1ST ORDER MOMENT)}$$

$$x_c = \frac{\sum_x \sum_y x b(x,y)}{A}$$

2ND MOMENT OF INERTIA

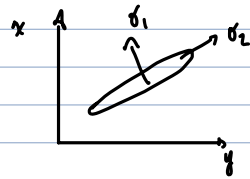
$$y_c = \frac{\sum_x \sum_y y b(x, y)}{A}$$

$$I_o = \begin{bmatrix} I_{xx} & I_{xy} \\ I_{xy} & I_{yy} \end{bmatrix}$$

$$I_{xx} = \sum_x \sum_y (y - y_c)^2 b(x, y)$$

$$I_{yy} = \sum_x \sum_y (x - x_c)^2 b(x, y)$$

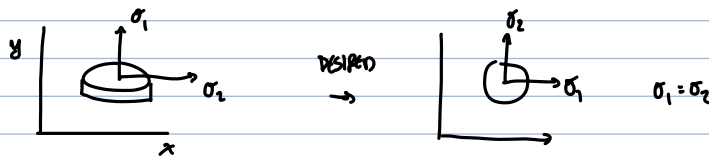
$$I_{xy} = \sum_x \sum_y (x - x_c)(y - y_c) b(x, y)$$



σ_1 σ_2 PRINCIPLE MOMENTS
EIGENVALUES OF I_o

$$\alpha = \frac{1}{2} \tan^{-1} \left(\frac{2 I_{xy}}{I_{xx} - I_{yy}} \right)$$

EXAMPLE FEATURE SET

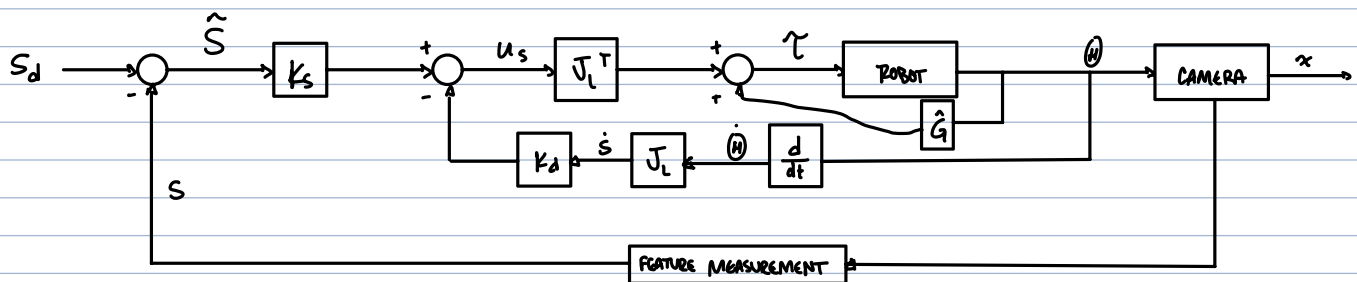


FEATURE SET $S = \begin{bmatrix} x_c \\ y_c \\ \frac{1}{2}(\sigma_1 + \sigma_2) \\ \frac{1}{2}(\sigma_1 - \sigma_2) \end{bmatrix}$

← AVERAGE SIZE OF OBJECT IN IMAGE

← DISTRIBUTION OF IMAGE

IMAGE-BASED VISUAL SERVOING



WE HAVE NO CONTROL OVER THIS

$$\dot{S} = J_S^c V_{c,0} = J_S^c \cancel{V_o} + L_S^c V_c$$

INTERACTION MATRIX

V_o = VELOCITY OF OBJECT
WRT BASE FRAME

V_c = VELOCITY OF CAMERA WRT BASE
FRAME

${}^c V_{c,0}$ = VELOCITY OF OBJECT RELATIVE
TO CAMERA (IN CAMERA FRAME)

$$L_s = J_s^{-1} \Gamma({}^c x_{c,o})$$

ACCOUNTS FOR FACT THAT
CAMERA ROTATION CAUSES A
FEATURE TRANSLATION

L_s MUST BE DERIVED ON A CASE-BY-CASE BASIS

• DEPENDS ON CHOICE IN FEATURES

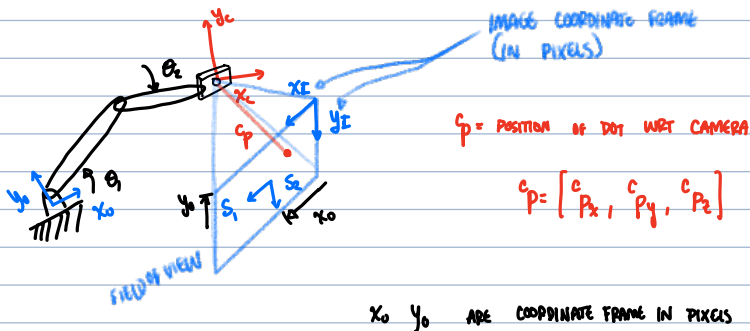
o FRAME = BASE FRAME

c FRAME = CAMERA FRAME

$$\dot{s} = L_s {}^c v_c = L_s {}^o P_c {}^o v_c = \underbrace{L_s {}^c P_o}_{J_L} J(\theta) \dot{\theta}$$

$$\therefore J_L^T = J^T(\theta) {}^o Z_c L_s^T$$

EXAMPLE: 2-DOF ROBOT TRACKING A DOT



$$\begin{bmatrix} s_1 \\ s_2 \end{bmatrix} = \begin{bmatrix} x_c - x_o \\ y_c - y_o \end{bmatrix} = \begin{bmatrix} -\alpha {}^c p_x \\ -\alpha {}^c p_y \end{bmatrix}$$

(PIXELS) (METERS)

α = SCALING FACTOR (PIXELS/M)
EXPERIMENTALLY CALIBRATED

CAN SHOW

$$\begin{bmatrix} \dot{s}_1 \\ \dot{s}_2 \end{bmatrix} = \underbrace{\begin{bmatrix} \alpha & 0 & s_1 \\ 0 & \alpha & s_2 \end{bmatrix}}_{L_s} \begin{bmatrix} \dot{x}_c \\ \dot{y}_c \\ \dot{w}_2 \end{bmatrix}$$

$[2 \times 3]$

$$\dot{\theta}_2 = \dot{\theta}_1 + \dot{\theta}_2$$

$${}^o P_c = \begin{bmatrix} c\phi_2 & -s\phi_2 & 0 \\ s\phi_2 & c\phi_2 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

$$J_L = L_s {}^o P_c J(\theta)$$

$$[2 \times 2] = [2 \times 3] [3 \times 3] [3 \times 2]$$

↑ DIFFERENT JACOBIAN

• WERE FACTORING IN FIRST, SECOND, & 6TH ROWS OF FULL VELOCITY JACOBIAN (w_2)