

Assistive Robotics

IBVS Image-Based Visual Servoing

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Image-Based Visual Servoing

- Image-Based Visual Servoing
 - The target is defined in image coordinates
 - The task is defined in the image
 - Can reduce computational delay, eliminating the need for image interpretation
 - Non-linear and coupled system

Image-Based Visual Servoing: Introduction

- Image-based visual servoing
 - ☐ The target is defined in image coordinates
 - ☐ The task is defined in the image space
 - Non-linear and coupled system
 - We seek for robustness to poor calibration

- ☐ Related bibliography:
 - ➤ [Chaumette et al. 1991], [Espiau et al. 1992], [Corke & Good 1996], [Ma et al. 1996], [Matsumoto et al. 1996], [Hager 1997], [Christensen et al. 1999], [Drummond & Cipolla 1999], [Kragic et al. 2001], [Conticelli & Allotta 2001], [Graefe & Bischoff 2004], [Adachi & Sato 2004], [Argyros et al. 2005], [Remazeilles & Chaumette 2007]

Image-Based Visual Servoing

- Problem statement
 - Select k image features **s** suitable for controlling m degrees of freedom. $(k \ge m), (m \le 6)$
 - \square Determine the target s^*
 - Regulate the error to zero $\mathbf{e}(\mathbf{t}) = (\mathbf{s} \mathbf{s}^*) \to 0$ if $\mathbf{s}^* = cte \Rightarrow \dot{\mathbf{e}} = \dot{\mathbf{s}}$
- Relation between temporal variation of s with camera velocity

$$\dot{\mathbf{s}} = \mathbf{L} \cdot \mathbf{v}$$
, with $\mathbf{L} \in \Re^{k \times 6}$

$$\mathbf{v} = \begin{pmatrix} v_x & v_y & v_z & \omega_x & \omega_y & \omega_z \end{pmatrix}^T$$

- L: Interaction matrix (image Jacobian)
 - Relation between time variation of error with camera velocity

$$\dot{\mathbf{e}} = \mathbf{L} \cdot \mathbf{v}$$

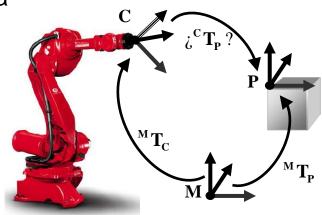
- Deduction of the interaction matrix
 - Location of a point with respect to the camera
 - World reference: M
 - Camera reference: C(t)
 - Point reference: P

$$\mathcal{L}^{\mathbf{C}} \mathbf{T}_{\mathbf{P}}? \longrightarrow^{\mathbf{C}} \mathbf{T}_{\mathbf{P}} = ({}^{\mathbf{M}} \mathbf{T}_{\mathbf{C}})^{-1} \cdot {}^{\mathbf{M}} \mathbf{T}_{\mathbf{P}}$$

$${}^{\mathbf{C}}\mathbf{T}_{\mathbf{P}} = \begin{pmatrix} {}^{\mathbf{C}}\mathbf{R}_{\mathbf{P}}(t) & {}^{\mathbf{C}}\mathbf{t}_{\mathbf{P}}(t) \\ 0 & 1 \end{pmatrix} = \begin{pmatrix} {}^{\mathbf{M}}\mathbf{R}_{\mathbf{C}}(t) & {}^{\mathbf{M}}\mathbf{t}_{\mathbf{C}}(t) \\ 0 & 1 \end{pmatrix}^{-1} \cdot \begin{pmatrix} {}^{\mathbf{M}}\mathbf{R}_{\mathbf{P}} & {}^{\mathbf{M}}\mathbf{t}_{\mathbf{P}} \\ 0 & 1 \end{pmatrix}$$

$$\begin{cases} {^{\mathbf{C}}}\mathbf{R}_{\mathbf{P}}(t) = {^{\mathbf{C}}}\mathbf{R}_{\mathbf{M}}(t) \cdot {^{\mathbf{M}}}\mathbf{R}_{\mathbf{P}} \\ {^{\mathbf{C}}}\mathbf{t}_{\mathbf{P}}(t) = {^{\mathbf{C}}}\mathbf{R}_{\mathbf{M}}(t) \cdot {^{\mathbf{M}}}\mathbf{t}_{\mathbf{P}} - {^{\mathbf{M}}}\mathbf{t}_{\mathbf{C}}(t) \end{cases}$$





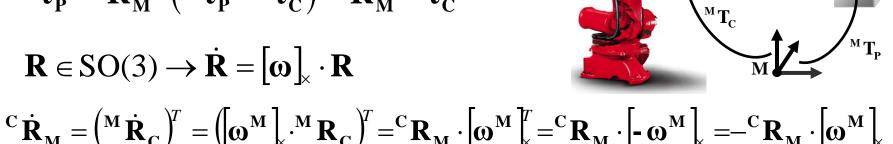
$$\frac{\mathbf{M} \mathbf{t}_{\mathbf{C}}(t)}{1} \cdot \begin{pmatrix} \mathbf{M} \mathbf{R}_{\mathbf{P}} & \mathbf{M} \mathbf{t}_{\mathbf{P}} \\ 0 & 1 \end{pmatrix}$$

- Deduction of the interaction matrix
 - Derivative of point location with respect to camera

$${}^{\mathbf{C}}\mathbf{t}_{\mathbf{P}}(t) = {}^{\mathbf{C}}\mathbf{R}_{\mathbf{M}}(t) \cdot \left({}^{\mathbf{M}}\mathbf{t}_{\mathbf{P}} - {}^{\mathbf{M}}\mathbf{t}_{\mathbf{C}}(t)\right)$$

$${}^{\mathbf{C}}\dot{\mathbf{t}}_{\mathbf{P}} = {}^{\mathbf{C}}\dot{\mathbf{R}}_{\mathbf{M}} \cdot \left({}^{\mathbf{M}}\mathbf{t}_{\mathbf{P}} - {}^{\mathbf{M}}\mathbf{t}_{\mathbf{C}}\right) - {}^{\mathbf{C}}\mathbf{R}_{\mathbf{M}} \cdot {}^{\mathbf{M}}\dot{\mathbf{t}}_{\mathbf{C}}$$

$$\mathbf{R} \in \mathbf{SO}(3) \to \dot{\mathbf{R}} = \left[\boldsymbol{\omega}\right]_{\times} \cdot \mathbf{R}$$



$$\overset{C}{\Rightarrow} \dot{t}_{P} = -\overset{C}{\Rightarrow} \mathbf{R}_{M} \cdot \left[\boldsymbol{\omega}^{M}\right]_{\times} \cdot \left(\overset{M}{\Rightarrow} \mathbf{t}_{P} - \overset{M}{\Rightarrow} \mathbf{t}_{C}\right) - \overset{C}{\Rightarrow} \mathbf{R}_{M} \cdot \overset{M}{\Rightarrow} \dot{t}_{C}$$

$$\overset{C}{\Rightarrow} \dot{t}_{P} = -\overset{C}{\Rightarrow} \mathbf{R}_{M} \cdot \boldsymbol{\omega}^{M} \times \overset{C}{\Rightarrow} \mathbf{R}_{M} \cdot \left(\overset{M}{\Rightarrow} \mathbf{t}_{P} - \overset{M}{\Rightarrow} \mathbf{t}_{C}\right) - \overset{C}{\Rightarrow} \mathbf{R}_{M} \cdot \overset{M}{\Rightarrow} \dot{t}_{C}$$

$$\overset{C}{\Rightarrow} \dot{t}_{P} = -\overset{C}{\Rightarrow} \mathbf{R}_{M} \cdot \boldsymbol{\omega}^{M} \times \overset{C}{\Rightarrow} \mathbf{R}_{M} \cdot \left(\overset{M}{\Rightarrow} \mathbf{t}_{P} - \overset{M}{\Rightarrow} \mathbf{t}_{C}\right) - \overset{C}{\Rightarrow} \mathbf{R}_{M} \cdot \overset{M}{\Rightarrow} \dot{t}_{C}$$

$$\dot{\mathbf{p}} = -\mathbf{\omega} \times \mathbf{p} - \mathbf{v}$$

v: camera velocity

ω: camera angular velocity

Deduction of the interaction matrix

$$\dot{\mathbf{p}} = -\mathbf{\omega} \times \mathbf{p} - \mathbf{v}$$

$$\dot{\mathbf{s}} = \begin{pmatrix} \dot{s}_1 \\ \vdots \\ \dot{s}_n \end{pmatrix} = \frac{\partial \mathbf{s}}{\partial \mathbf{r}} \cdot \dot{\mathbf{r}} = \mathbf{L} \cdot \mathbf{v} = \begin{pmatrix} \frac{\partial \mathbf{s}_1}{\partial r_1} & \dots & \frac{\partial \mathbf{s}_1}{\partial r_k} \\ \vdots & \ddots & \vdots \\ \frac{\partial \mathbf{s}_n}{\partial r_1} & \dots & \frac{\partial \mathbf{s}_n}{\partial r_k} \end{pmatrix} \cdot \begin{pmatrix} v_x \\ v_y \\ v_z \\ \omega_x \\ \omega_y \\ \omega_z \end{pmatrix}$$
Image features: \mathbf{s}

- Image features:
- Camera position:
- Camera velocity: $\mathbf{v} = \dot{\mathbf{r}}$
- Interaction matrix: **I**

 $i\dot{\mathbf{p}}\leftrightarrow\dot{\mathbf{s}}\leftrightarrow\mathbf{v}?$

Deduction of the interaction matrix

$$\dot{\mathbf{p}} = -\mathbf{\omega} \times \mathbf{p} - \mathbf{v}$$

$$\mathbf{p} = \begin{pmatrix} \mathbf{X} \\ \mathbf{Y} \\ \mathbf{Z} \end{pmatrix} \rightarrow \dot{\mathbf{p}} = \begin{pmatrix} \dot{\mathbf{X}} \\ \dot{\mathbf{Y}} \\ \dot{\mathbf{Z}} \end{pmatrix} = -\begin{pmatrix} \omega_{\mathbf{x}} \\ \omega_{\mathbf{y}} \\ \omega_{\mathbf{z}} \end{pmatrix} \times \begin{pmatrix} \mathbf{X} \\ \mathbf{Y} \\ \mathbf{Z} \end{pmatrix} - \begin{pmatrix} v_{\mathbf{x}} \\ v_{\mathbf{y}} \\ v_{\mathbf{z}} \end{pmatrix} = \begin{pmatrix} Y \cdot \omega_{\mathbf{z}} - Z \cdot \omega_{\mathbf{y}} - v_{\mathbf{x}} \\ Z \cdot \omega_{\mathbf{x}} - X \cdot \omega_{\mathbf{z}} - v_{\mathbf{y}} \\ X \cdot \omega_{\mathbf{y}} - Y \cdot \omega_{\mathbf{x}} - v_{\mathbf{z}} \end{pmatrix}$$

 \square Points in the image (Projection of **p** in **x**):

$$\mathbf{x} = (x, y)^{T} = (X/Z, Y/Z)^{T}$$

$$\dot{\mathbf{x}} = \frac{\dot{\mathbf{X}} \cdot \mathbf{Z} - \mathbf{X} \cdot \dot{\mathbf{Z}}}{\mathbf{Z}^{2}} = \frac{\dot{\mathbf{X}} - \mathbf{x} \cdot \dot{\mathbf{Z}}}{\mathbf{Z}}$$

$$\dot{\mathbf{y}} = \frac{\dot{\mathbf{Y}} \cdot \mathbf{Z} - \mathbf{Y} \cdot \dot{\mathbf{Z}}}{\mathbf{Z}^{2}} = \frac{\dot{\mathbf{Y}} - \mathbf{y} \cdot \dot{\mathbf{Z}}}{\mathbf{Z}}$$

$$\dot{\mathbf{z}} = \frac{\dot{\mathbf{Y}} - \dot{\mathbf{Y}} \cdot \dot{\mathbf{Z}}}{\mathbf{Z}^{2}} = \frac{\dot{\mathbf{Y}} - \dot{\mathbf{Y}} \cdot \dot{\mathbf{Z}}}{\mathbf{Z}}$$

Deduction of the interaction matrix
$$\mathbf{L}$$
: $\dot{\mathbf{s}} = \mathbf{L} \cdot \mathbf{v}$

$$\begin{pmatrix} \dot{\mathbf{x}} \\ \dot{\mathbf{y}} \end{pmatrix} = \begin{pmatrix} -1/\mathbf{Z} & 0 & \mathbf{x}/\mathbf{Z} & \mathbf{x} \cdot \mathbf{y} & -(1+\mathbf{x}^2) & \mathbf{y} \\ 0 & -1/\mathbf{Z} & \mathbf{y}/\mathbf{Z} & (1+\mathbf{y}^2) & -\mathbf{x} \cdot \mathbf{y} & -\mathbf{x} \end{pmatrix} \cdot \mathbf{v}$$

$$\mathbf{v} = \begin{pmatrix} v_x & v_y & v_z & \omega_x & \omega_y & \omega_z \end{pmatrix}^T$$

- lacksquare For a point: $s\in \mathfrak{R}^2, L\in \mathfrak{R}^{2 imes 6}$
- $\hfill \square$ For a set of n points: $s\in\Re^{2n}, L\in\Re^{2n\times 6}$

$$\mathbf{s} = \begin{pmatrix} \mathbf{x}_1 & \mathbf{y}_1 & \dots & \mathbf{x}_n & \mathbf{y}_n \end{pmatrix}^T$$

$$\mathbf{L} = \begin{pmatrix} \mathbf{L}_{1} \\ \vdots \\ \mathbf{L}_{n} \end{pmatrix} = \begin{pmatrix} -1/Z_{1} & 0 & x_{1}/Z_{1} & x_{1} \cdot y_{1} & -(1+x_{1}^{2}) & y_{1} \\ 0 & -1/Z_{1} & y_{1}/Z_{1} & (1+y_{1}^{2}) & -x_{1} \cdot y_{1} & -x_{1} \\ \vdots & & \vdots & & \vdots \\ -1/Z_{n} & 0 & x_{n}/Z_{n} & x_{n} \cdot y_{n} & -(1+x_{n}^{2}) & y_{n} \\ 0 & -1/Z_{n} & y_{n}/Z_{n} & (1+y_{n}^{2}) & -x_{n} \cdot y_{n} & -x_{n} \end{pmatrix}$$

- Deduction of the interaction matrix in cylindrical coordinates
 - □ Considering the image points in cylindrical coordinates instead of Cartesian coordinates

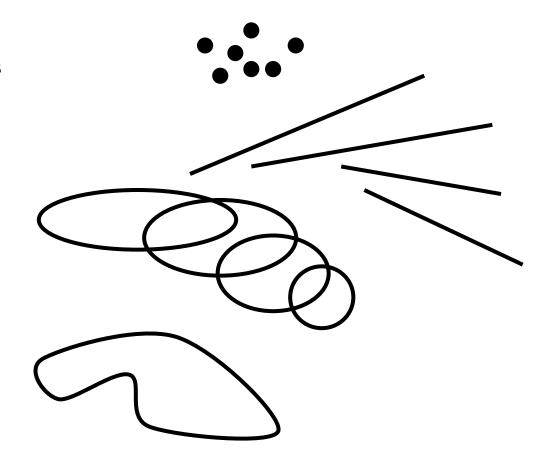
$$\dot{\mathbf{s}} = \mathbf{L} \cdot \mathbf{v} \quad \begin{cases} \rho = \sqrt{\mathbf{x}^2 + \mathbf{y}^2} & \dot{\rho} = \frac{\mathbf{x} \cdot \dot{\mathbf{x}} - \mathbf{y} \cdot \dot{\mathbf{y}}}{\rho} \\ \theta = \arctan \frac{\mathbf{y}}{\mathbf{x}} \to \operatorname{atan2}(\mathbf{y}, \mathbf{x}) & \dot{\theta} = \frac{\mathbf{x} \cdot \dot{\mathbf{y}} - \dot{\mathbf{x}} \cdot \mathbf{y}}{\rho^2} \\ \mathbf{s} = (\rho_1 \quad \theta_1 \quad \dots \quad \rho_n \quad \theta_n)^T \end{cases}$$

$$\mathbf{L} = \begin{pmatrix} \mathbf{L}_{\rho} \\ \mathbf{L}_{\theta} \end{pmatrix} = \begin{pmatrix} \frac{-\cos\theta}{Z} & \frac{-\sin\theta}{Z} & \frac{\rho}{Z} & (1+\rho^{2})\sin\theta & -(1+\rho^{2})\cos\theta & 0 \\ \frac{\sin\theta}{\rho \cdot Z} & \frac{-\cos\theta}{\rho \cdot Z} & 0 & \frac{\cos\theta}{\rho} & \frac{\sin\theta}{\rho} & -1 \end{pmatrix}$$

$$\mathbf{v} = \begin{pmatrix} v_{x} & v_{y} & v_{z} & \omega_{x} & \omega_{y} & \omega_{z} \end{pmatrix}^{T}$$

□ Produces decoupling between columns 3 and 6 of the interaction matrix.

- Deduction of the interaction matrix
 - Some possible features to use:
 - Points
 - Cartesian coordinates
 - Cylindrical coordinates
 - Segments
 - Straight lines
 - Circles
 - Cylinders
 - Contours of objects
 - Moments of shapes
 - Center of gravity
 - Object orientation
 - etc.



- lacksquare Computation of the interaction matrix $\dot{f s}={f L}\cdot{f v} o{f v}={f -}\lambda\cdot\hat{f L}^{\!\!+}\cdot{f e}$
 - Estimation in each control loop.
 - Computation using the analytical expression as a function of image features and depth estimation.

$$\hat{\mathbf{L}}(t) = \mathbf{L}(\mathbf{s}, \hat{Z})$$

- Possibility of local minima and singularities
- Computation of the interaction matrix at the target position. We use a constant Jacobian expression computed a priori using the features and depths at the target position.

$$\hat{\mathbf{L}} = \mathbf{L}(\mathbf{s}^*, \hat{Z}^*) = cte$$

- Convergence is only ensured in the neighborhood of the target position.
- Direct estimation
 - Estimation without taking into account its analytical expression.
 - Measure variations of s due to known camera motions

$$\Delta \mathbf{s} = \mathbf{L} \cdot \Delta \mathbf{v}$$

❖ We have k equations and (kx6) unknowns. By performing N independent movements of the camera (N>6) it is possible to solve the system

Image-Based Visual Servoing: Control law

$$\dot{e} = L \cdot v$$

To achieve exponential error reduction we define:

$$\mathbf{v} = -\lambda \cdot \mathbf{L}^+ \cdot \mathbf{e}$$
 , $\lambda > 0$

 $lue{}$ Where the Moore-Penrose pseudo inverse is used when ${f L}$ has rank 6.

$$\mathbf{L}^{+} = \left(\mathbf{L}^{T} \cdot \mathbf{L}\right)^{-1} \cdot \mathbf{L}^{T}$$

 \square If k=6 and the determinant of $\mathbf L$ is nonzero

$$\mathbf{v} = -\lambda \cdot \mathbf{L}^{-1} \cdot \mathbf{e}$$

- In practice it is impossible to know perfectly the interaction matrix. It must be estimated: $\hat{\mathbf{L}}$
 - lacksquare The control law remains in practice as: $\mathbf{v} = -\lambda \cdot \hat{\mathbf{L}}^+ \cdot \mathbf{e}$
- Closed-loop behavior

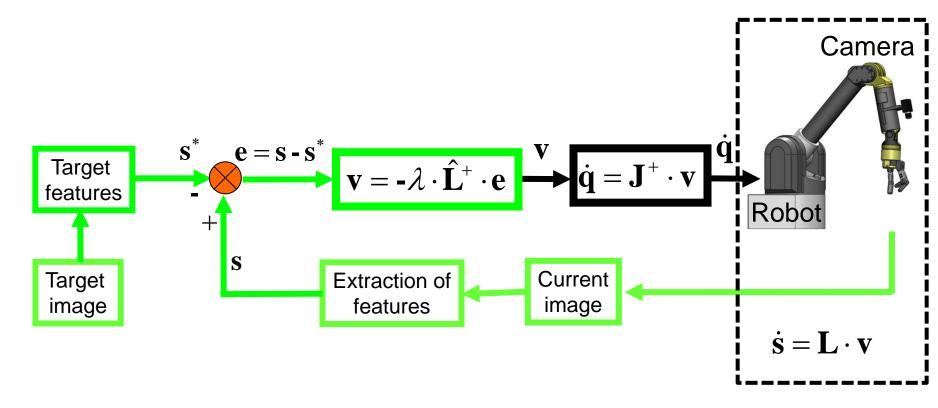
$$\dot{\mathbf{e}} = -\lambda \cdot \mathbf{L} \cdot \hat{\mathbf{L}}^{\scriptscriptstyle +} \cdot \mathbf{e}$$

- $oxedsymbol{\square}$ Exponential error reduction if $\mathbf{L}\cdot\hat{\mathbf{L}}^{\scriptscriptstyle +}=\mathbf{I}_6\Rightarrow\dot{\mathbf{e}}=\mathbf{-}\lambda\cdot\mathbf{e}$
- ☐ In any other case: Stability analysis...

$$\mathbf{L} \cdot \hat{\mathbf{L}}^+ \neq \mathbf{I_6} \Rightarrow i Stability?$$

Image-Based Visual Servoing: Control law

Image-based visual control loop



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Stability analysis

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