



Assistive Robotics

Hybrid Visual Servoing and Stability Analysis

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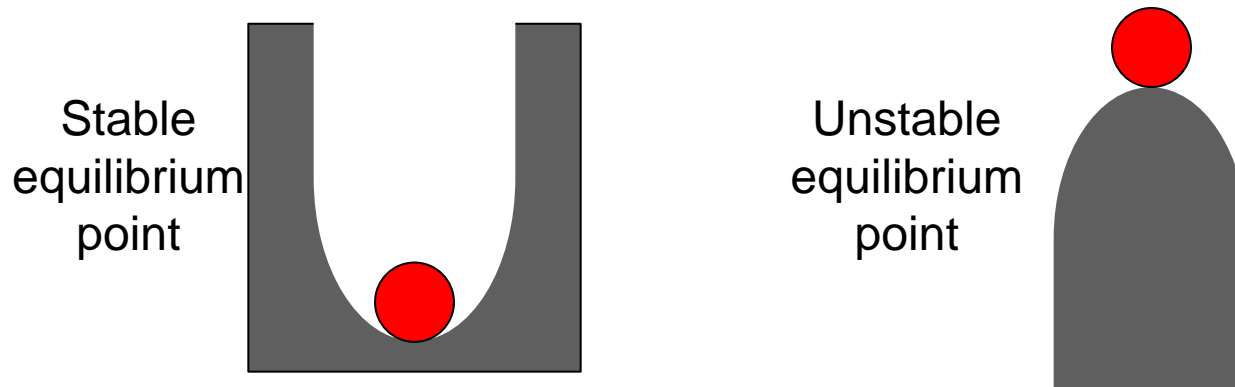
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- Vision systems
- Classification of visual control systems
- Position-Based Visual Servoing
- Image-Based Visual Servoing

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Stability analysis

- It is important that the systems are stable. An unstable system is generally useless and potentially dangerous.
- A system is said to be stable if the system being close to its operating point implies that it will always remain around that point.



- Stability of systems
 - Time invariant linear systems:
 - Multiple techniques and criteria available (Nyquist, Routh, etc.).
 - Non-linear or time-varying systems:
 - Difficult and sometimes impossible
 - Lyapunov method

Stability analysis: Lyapunov

- Lyapunov's methods (1892)
 - For determining the stability of dynamical systems described by ordinary differential equations.
 - First Lyapunov method
 - Applicable when the explicit solution of the differential equations is available.
 - Lyapunov's second method or direct method
 - To analyze the stability without solving the differential equations.
- The basic idea of Lyapunov's direct method is based on the physical fact that if the total "energy" of a mechanical system is continuously dissipated, then, whether the system is linear or nonlinear, it must eventually reach a point of equilibrium (zero energy).
- Therefore, the stability of the system can be analyzed by studying the variation of an energy function associated with the system.



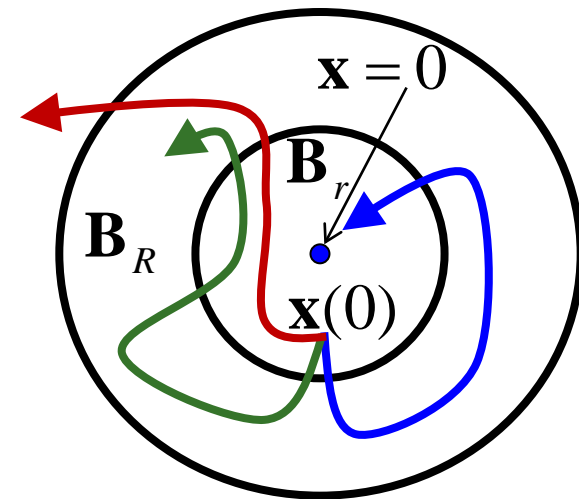
Aleksandr Liapunov
1857-1918

**Russian mathematician
and physicist**

Stability analysis: Lyapunov

■ Lyapunov stability

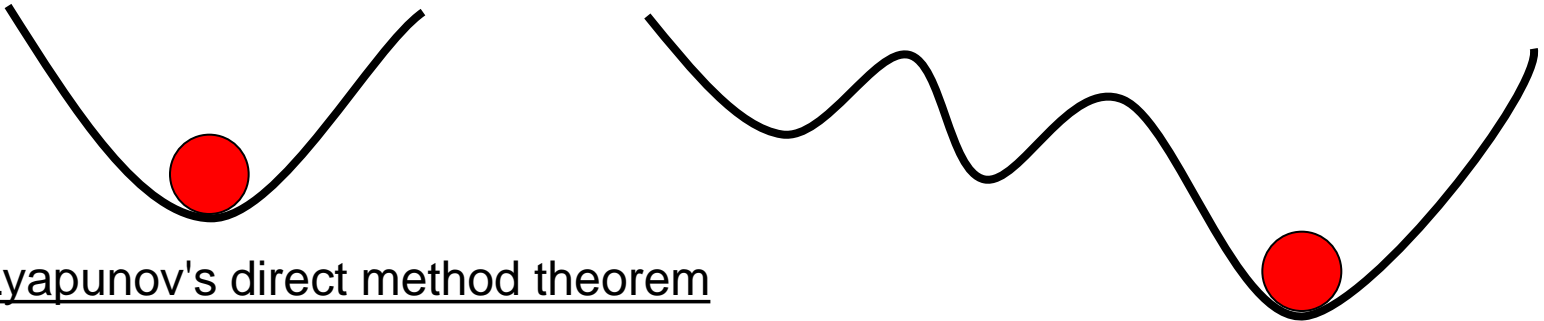
- Let be a system with state vector \mathbf{x} such that: $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, t)$
- We define state or equilibrium point of the system: $\mathbf{f}(\mathbf{x}_e = \mathbf{0}, t) = \mathbf{0}, \forall t$
 - If the system is linear: $\dot{\mathbf{x}} = \mathbf{A} \cdot \mathbf{x}$, if $\det(\mathbf{A}) \neq 0 \Rightarrow \exists$ one single \mathbf{x}_e
 - In nonlinear systems there may be more than one equilibrium point.
- Lyapunov stability:
 $\forall R > 0, \exists r > 0, \mathbf{x}(0) \in \mathbf{B}_r \Rightarrow \forall t \geq 0, \mathbf{x}(t) \in \mathbf{B}_R$
 - An equilibrium point is **stable** if for any $R > 0$ there exists an $r > 0$ such that if $\mathbf{x}(0)$ belongs to the sphere of radius r , then $\mathbf{x}(t)$ will belong to the sphere of radius R for all t .
 - Otherwise the equilibrium point is **unstable**.
- Asymptotic stability:
 - An equilibrium point $\mathbf{x} = \mathbf{0}$ is **asymptotically stable** if it is stable and in addition
 $\exists r > 0, \mathbf{x}(0) \in \mathbf{B}_r \Rightarrow \mathbf{x}(t) \rightarrow \mathbf{0}, t \rightarrow \infty$



Stability analysis: Lyapunov

■ Lyapunov stability

- Global: If asymptotic stability is satisfied for any initial state, the equilibrium point is globally asymptotically stable.
- Local: If asymptotic stability is not satisfied for every initial state, the equilibrium point is locally asymptotically stable.



□ Lyapunov's direct method theorem

Given the system $\dot{\mathbf{x}} = \mathbf{f}(\mathbf{x}, t)$, with $\mathbf{f}(\mathbf{0}, t) = \mathbf{0}$, $\forall t$, if $\exists V(\mathbf{x}, t) \ni V(\mathbf{x}, t) > 0$ (positive definite) and $\dot{V}(\mathbf{x}, t) < 0$ (negative definite)
 \Rightarrow The equilibrium point in the origin is asymptotically stable

- The scalar function $V(\mathbf{x}, t)$ is called Lyapunov function
- A function V is positive definite if $V(\mathbf{x}) > 0, \forall \mathbf{x} \neq \mathbf{0} \wedge V(\mathbf{0}) = 0$

Stability analysis: Lyapunov

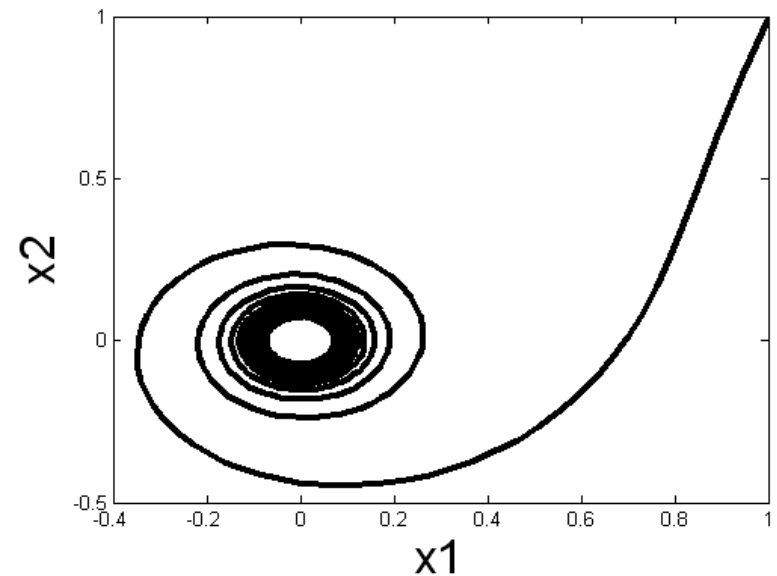
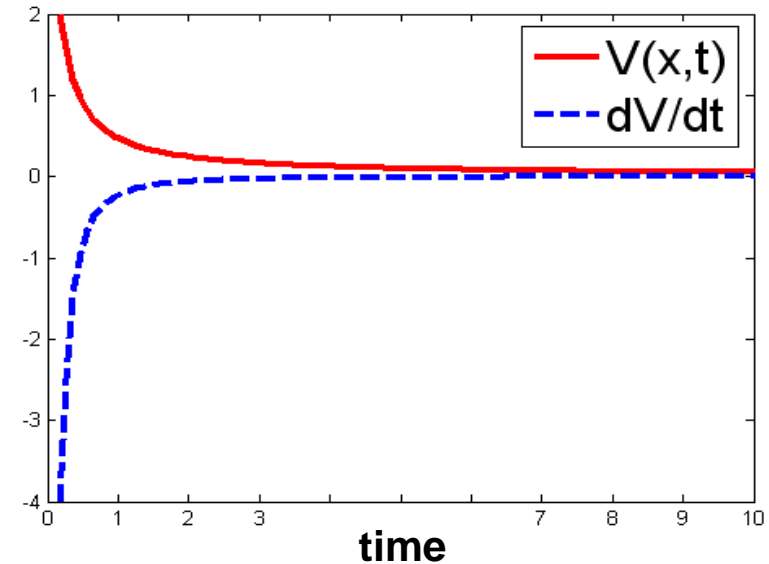
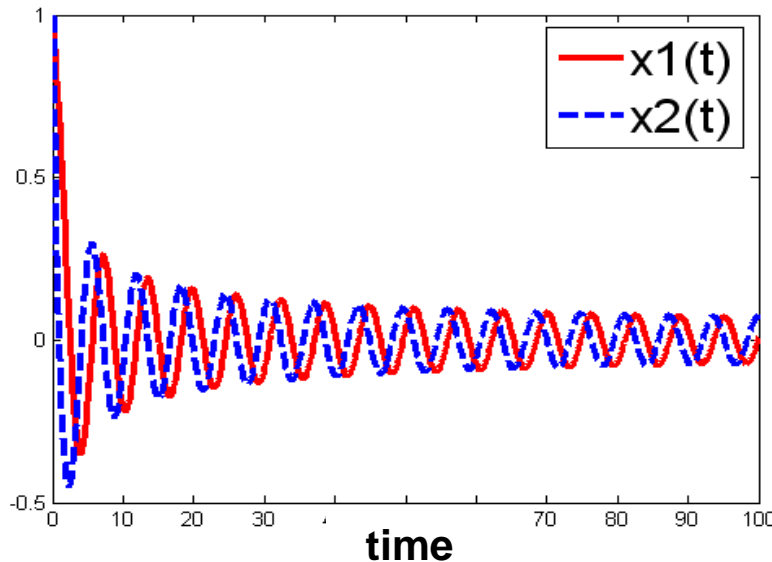
■ Example of Lyapunov stability

$$\begin{cases} \dot{x}_1 = x_2 - x_1(x_1^2 + x_2^2) \\ \dot{x}_2 = -x_1 - x_2(x_1^2 + x_2^2) \end{cases}$$

$$V(\mathbf{x}) = x_1^2 + x_2^2 > 0$$

$$\dot{V}(\mathbf{x}) = 2x_1\dot{x}_1 + 2x_2\dot{x}_2$$

$$= -2(x_1^2 + x_2^2)^2 < 0$$



Stability analysis: Visual servoing

■ Lyapunov stability for image-based visual servoing

- Control law: $\dot{\mathbf{e}} = \mathbf{L} \cdot \mathbf{v}$, with $\mathbf{v} = -\lambda \cdot \hat{\mathbf{L}}^+ \cdot \mathbf{e}$
- We define the Lyapunov function: $V(\mathbf{e}) = \frac{1}{2} \cdot \|\mathbf{e}(t)\|^2 > 0$
- Derivate to obtain the stability condition:

$$\dot{V} = \mathbf{e}^T \cdot \dot{\mathbf{e}} = -\lambda \cdot \mathbf{e}^T \cdot \mathbf{L} \cdot \hat{\mathbf{L}}^+ \cdot \mathbf{e} < 0$$

- Which is satisfied if the matrix $(\mathbf{L} \cdot \hat{\mathbf{L}}^+)$ is positive definite:

$$\mathbf{L} \cdot \hat{\mathbf{L}}^+ > 0 \Rightarrow \text{Asymptotically stable}$$

- Note: A symmetric matrix is positive definite if for every non-zero vector its quadratic form is positive

$$\mathbf{A} \in \mathfrak{R}^{n \times n}, \mathbf{A} = \mathbf{A}^T$$

$$\mathbf{A} > 0 \Leftrightarrow \forall \mathbf{b} \neq \mathbf{0}, \mathbf{b}^T \cdot \mathbf{A} \cdot \mathbf{b} > 0$$

$$\text{e.g. } \mathbf{A} = \begin{pmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{pmatrix}, \mathbf{A} > 0 \Leftrightarrow |a_{11}| > 0, \begin{vmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{vmatrix} > 0, \begin{vmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \end{vmatrix} > 0$$

Stability analysis: Visual servoing

$$\mathbf{L} \cdot \hat{\mathbf{L}}^+ > 0 \Rightarrow \text{Stable} \quad \mathbf{L} \in \mathbb{R}^{k \times m} \quad \left\{ \begin{array}{l} m = \text{number of degrees of freedom (in general } m=6) \\ k = \text{number of features (e.g. } k=2n \text{ with } n \text{ points)} \\ \text{Assuming maximum rank} \end{array} \right.$$

$$k = m, \text{rank}(\mathbf{L}) = m$$

$$k < m, \text{rank}(\mathbf{L}) = k$$

$$k > m, \text{rank}(\mathbf{L}) = m$$

- Number of features equal to the number of degrees of freedom of the camera

$$(\mathbf{L} \cdot \hat{\mathbf{L}}^+) \in \mathbb{R}^{m \times m} \rightarrow \mathbf{L} \cdot \hat{\mathbf{L}}^+ > 0$$

- Ideally one would have:

$$\mathbf{L} \cdot \hat{\mathbf{L}}^+ = \mathbf{L} \cdot \hat{\mathbf{L}}^{-1} = \mathbf{I} > 0$$

- The estimation is assumed to be reasonably accurate

Stability analysis: Visual servoing

$$\mathbf{L} \cdot \hat{\mathbf{L}}^+ > 0 \Rightarrow \text{Stable} \quad \mathbf{L} \in \mathbb{R}^{k \times m} \quad \left\{ \begin{array}{l} m = \text{number of degrees of freedom (in general } m=6) \\ k = \text{number of features (e.g. } k=2n \text{ with } n \text{ points)} \\ \text{Assuming maximum rank} \end{array} \right.$$

$$k = m, \text{rank}(\mathbf{L}) = m$$

$$k < m, \text{rank}(\mathbf{L}) = k$$

$$k > m, \text{rank}(\mathbf{L}) = m$$

- Number of features less than the degrees of freedom of the camera.

- ☐ For example, for a point ($k=2$) and a camera with 6 degrees of freedom ($m=6$):

$$\mathbf{L} \in \mathbb{R}^{2 \times 6} \rightarrow \mathbf{L}^+ = \mathbf{L}^T \cdot (\mathbf{L} \cdot \mathbf{L}^T)^{-1} \rightarrow (\mathbf{L} \cdot \hat{\mathbf{L}}^+) \in \mathbb{R}^{2 \times 2}$$

- The null space (kernel) of the interaction matrix is defined:

$$\ker(\mathbf{L}) = \{ \mathbf{v} \in \mathbb{R}^6 : \mathbf{L} \cdot \mathbf{v} = \mathbf{0} \}$$

- ☐ The kernel is a subspace of dimension ($m-k$). Example: $6-2=4$

- Not all camera movements will be observable

$$\mathbf{v} \in \ker(\mathbf{L})$$

- ☐ For example, movement of the camera along the projection ray of the point, or rotation of the camera along the projection ray of the point

- Hybrid schemes: Some degrees of freedom are controlled with image-based control and the rest with other techniques

Stability analysis: Visual servoing

$$\mathbf{L} \cdot \hat{\mathbf{L}}^+ > 0 \Rightarrow \text{Stable} \quad \mathbf{L} \in \mathbb{R}^{k \times m} \quad \left\{ \begin{array}{l} m = \text{number of degrees of freedom (in general } m=6) \\ k = \text{number of features (e.g. } k=2n \text{ with } n \text{ points)} \\ \text{Assuming maximum rank} \end{array} \right.$$

$$k = m, \text{rank}(\mathbf{L}) = m$$

$$k < m, \text{rank}(\mathbf{L}) = k$$

$$k > m, \text{rank}(\mathbf{L}) = m$$

- Number of features greater than the camera's degrees of freedom

$$\mathbf{L} \in \mathbb{R}^{k \times m} \rightarrow \mathbf{L}^+ = (\mathbf{L}^T \cdot \mathbf{L})^{-1} \cdot \mathbf{L}^T \rightarrow (\mathbf{L} \cdot \hat{\mathbf{L}}^+) \in \mathbb{R}^{k \times k}$$

- The following null kernel is defined

$$\ker(\mathbf{L}^+) = \{\mathbf{e} \in \mathbb{R}^k : \mathbf{L}^+ \cdot \mathbf{e} = \mathbf{0}\}$$

- ☐ The kernel is a subspace of dimension (k-m)
- There are configurations of points, different than the target, that produce zero velocity: $\mathbf{e} \in \ker(\mathbf{L}^+)$
- The system converges exponentially to a solution different than the desired one: **local minimum**
- Therefore only asymptotic local stability can be demonstrated.
 - ☐ Example: $k-m=8-6=2$. The null space has dimension 2. However this does not necessarily imply that there are 2 local minima, since they must be physically coherent.

Stability analysis: Visual servoing

$$\mathbf{L} \cdot \hat{\mathbf{L}}^+ > 0 \Rightarrow \text{Stable} \quad \mathbf{L} \in \mathbb{R}^{k \times m} \quad \left\{ \begin{array}{l} m = \text{number of degrees of freedom (in general } m=6) \\ k = \text{number of features (e.g. } k=2n \text{ with } n \text{ points)} \\ \text{Assuming maximum rank} \end{array} \right.$$

$$k = m, \text{rank}(\mathbf{L}) = m$$

$$k < m, \text{rank}(\mathbf{L}) = k$$

$$k > m, \text{rank}(\mathbf{L}) = m$$

- With $k > m$ the stability condition cannot be fulfilled.

$$k > m \Rightarrow (\mathbf{L} \cdot \hat{\mathbf{L}}^+) \in \mathbb{R}^{k \times k} \Rightarrow \text{rank}(\mathbf{L} \cdot \hat{\mathbf{L}}^+) \leq m \Rightarrow \mathbf{L} \cdot \hat{\mathbf{L}}^+ \leq 0$$

- Asymptotic local stability

- We define a new error: $\mathbf{e}' = \hat{\mathbf{L}}^+ \cdot \mathbf{e}$

- Derivative of the error:

$$\dot{\mathbf{e}}' = \hat{\mathbf{L}}^+ \cdot \dot{\mathbf{e}} + \dot{\hat{\mathbf{L}}^+} \cdot \mathbf{e} = \hat{\mathbf{L}}^+ \cdot \mathbf{L} \cdot \mathbf{v} + \mathbf{O} \cdot \mathbf{v} = (\hat{\mathbf{L}}^+ \cdot \mathbf{L} + \mathbf{O}) \cdot \mathbf{v}$$

$$\text{with } \mathbf{O} \in \mathbb{R}^{6 \times 6} \ni \mathbf{e} \rightarrow \mathbf{0} \Rightarrow \mathbf{O} \rightarrow \mathbf{0}$$

- Closed-loop behavior with the control law

$$\dot{\mathbf{e}}' = -\lambda \cdot (\hat{\mathbf{L}}^+ \cdot \mathbf{L} + \mathbf{O}) \cdot \hat{\mathbf{L}}^+ \cdot \mathbf{e} = -\lambda \cdot (\hat{\mathbf{L}}^+ \cdot \mathbf{L} + \mathbf{O}) \cdot \mathbf{e}'$$

- It is locally asymptotically stable if

$$\hat{\mathbf{L}}^+ \cdot \mathbf{L} > 0 \quad \text{around } \mathbf{e} = \mathbf{0}$$

- Provided the estimation is reasonably accurate, this condition will be satisfied if $\text{rank}(\mathbf{L}) = \text{rank}(\hat{\mathbf{L}}^+) = 6$

Stability analysis: Visual servoing

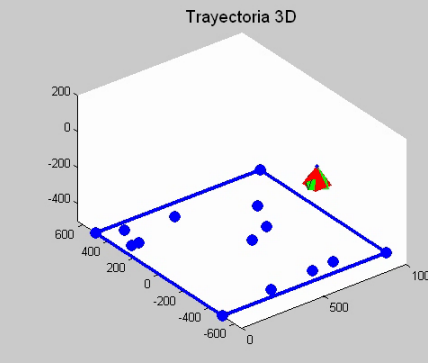
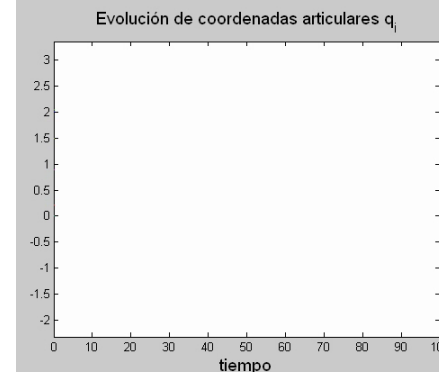
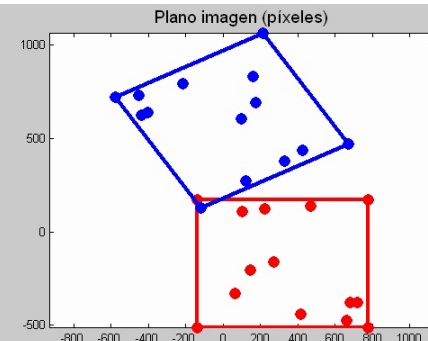
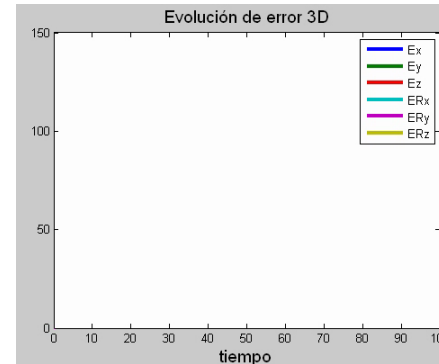
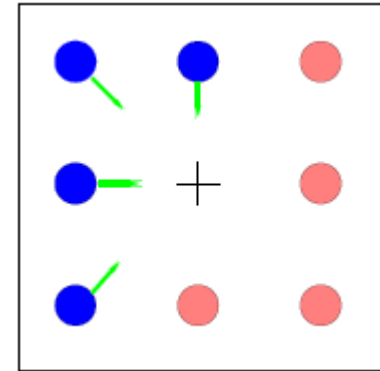
■ Singularity issues $\hat{\mathbf{L}} \rightarrow \hat{\mathbf{L}}^+$

□ The interaction matrix can be singular:

- If it is defined by three collinear points
- If the points and the optical center belong to a cylinder
- The use of more points reduces the problem but does not guarantee non-singularity

□ Therefore, the use of points can cause the interaction matrix to reach a singular configuration during visual control

□ Example of singularity: Consider a desired camera movement of 180 degrees around the Z-axis. The camera moves towards infinity and the interaction matrix reduces its range from 6 to 2.



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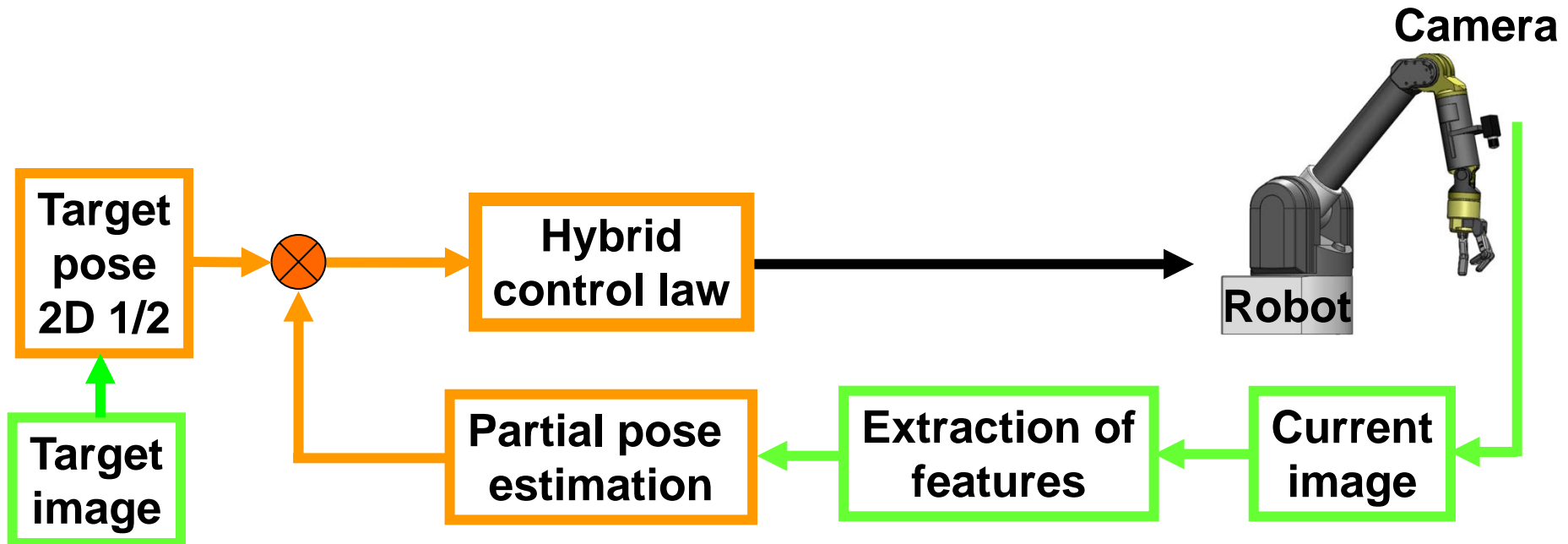
- **Hybrid Visual Servoing**
 - Hybrid visual servoing 2-1/2D

- Conclusion and bibliography

Hybrid visual servoing

■ Hybrid visual servoing

- Partially based on position and image
- Partial decoupling in the interaction matrix
- Presents the advantages of both position-based and image-based methods while overcoming their drawbacks
- Combines 2D and 3D information



Hybrid visual servoing 2-1/2D

■ 2-1/2D visual servoing

- Method that avoids the drawbacks of classical methods:
 - Unlike position-based methods it does not require a 3D model of the object.
 - Unlike image-based methods, convergence is guaranteed over the entire workspace.
- This control decouples rotational and translational motions by appropriate selection of visual features partially defined in 2D and 3D.
 - Partial estimation of camera motion
 - Robustness against calibration errors
 - Asymptotic stability guarantees
- E. Malis, F. Chaumette, and S. Boudet, “2-1/2D visual servoing,” *IEEE Trans. Robot. Automat.*, vol. 15, pp. 238-250, Apr. 1999

Hybrid visual servoing 2-1/2D

■ 2-1/2D Visual servoing: Image features $\dot{\mathbf{s}} = \mathbf{L}_h \cdot \begin{pmatrix} \mathbf{v} \\ \boldsymbol{\omega} \end{pmatrix}$

- We select the following image features

$$\mathbf{s} = (x \quad y \quad \log Z \quad \theta \cdot \mathbf{u}), \text{ with } \mathbf{u} = (u_x \quad u_y \quad u_z)^T$$

$$\mathbf{s}^* = (x^* \quad y^* \quad \log Z^* \quad \mathbf{0})$$

- Where θ and \mathbf{u} are the angle and axis defining the rotation, so that the orientation control is based on position.
- We define the error vector as

$$\mathbf{e} = \mathbf{s} - \mathbf{s}^* = (x - x^* \quad y - y^* \quad \log \rho_z \quad \theta \cdot \mathbf{u})^T, \text{ with } \rho_z = (Z / Z^*)$$

- We want an exponential decrease of the error

$$\dot{\mathbf{e}} = -\lambda \cdot \mathbf{e} = \dot{\mathbf{s}}$$

$$-\lambda \cdot \mathbf{e} = \dot{\mathbf{s}} = \mathbf{L}_h \cdot \begin{pmatrix} \mathbf{v} \\ \boldsymbol{\omega} \end{pmatrix} \Rightarrow \begin{pmatrix} \mathbf{v} \\ \boldsymbol{\omega} \end{pmatrix} = -\lambda \cdot \mathbf{L}_h^+ \cdot \mathbf{e}$$

Hybrid visual servoing 2-1/2D

■ 2-1/2D Visual servoing: Interaction Matrix

□ The hybrid interaction matrix is defined:

$$\dot{\mathbf{s}} = \mathbf{L}_h \cdot \begin{pmatrix} \mathbf{v} \\ \boldsymbol{\omega} \end{pmatrix} = \begin{pmatrix} \mathbf{L}_v & \mathbf{L}_\omega \\ \mathbf{0} & \mathbf{L}_{\theta u} \end{pmatrix} \cdot \begin{pmatrix} \mathbf{v} \\ \boldsymbol{\omega} \end{pmatrix}$$

$$\mathbf{L}_v = \frac{1}{Z^* \cdot \rho_z} \begin{pmatrix} -1 & 0 & x \\ 0 & -1 & y \\ 0 & 0 & -1 \end{pmatrix} \quad \mathbf{L}_\omega = \begin{pmatrix} x \cdot y & -(1+x^2) & y \\ 1+y^2 & -x \cdot y & -x \\ -y & x & 0 \end{pmatrix}$$

$$\mathbf{L}_{\theta u} = \mathbf{I}_3 - \frac{\theta}{2} \cdot [\mathbf{u}]_{\times} + \left(1 - \frac{\text{sinc}(\theta)}{\text{sinc}^2(\theta/2)} \right) \cdot [\mathbf{u}]_{\times}^2$$

- Z^* must be estimated.
- By selecting a point in the image (x, y) we have that the matrix \mathbf{L}_v is triangular and always invertible
- The ratio Z/Z^* can be calculated using the homography

Hybrid visual servoing 2-1/2D

■ 2-1/2D visual servoing: Control law

□ Computation of control velocities

$$-\lambda \cdot \mathbf{e} = \dot{\mathbf{s}} = \begin{pmatrix} \mathbf{L}_v & \mathbf{L}_\omega \\ \mathbf{0} & \mathbf{L}_{\theta u} \end{pmatrix} \cdot \begin{pmatrix} \mathbf{v} \\ \boldsymbol{\omega} \end{pmatrix} \Rightarrow \begin{pmatrix} \mathbf{v} \\ \boldsymbol{\omega} \end{pmatrix} = -\lambda \cdot \begin{pmatrix} \mathbf{L}_v^+ & -\mathbf{L}_v^+ \cdot \mathbf{L}_\omega \cdot \mathbf{L}_{\theta u}^{-1} \\ \mathbf{0} & \mathbf{L}_{\theta u}^{-1} \end{pmatrix} \cdot \mathbf{e}$$

$$\mathbf{e} = (\mathbf{e}_t \quad \mathbf{e}_\omega)^T = (x - x^* \quad y - y^* \quad \log \rho_z \quad \theta \cdot \mathbf{u})^T \rightarrow \mathbf{e}_\omega = \theta \cdot \mathbf{u}$$

□ Rotation velocities are based on position:

$$\boldsymbol{\omega} = -\lambda \cdot \mathbf{L}_{\theta u}^{-1} \cdot \mathbf{e}_\omega = -\lambda \cdot \mathbf{L}_{\theta u}^{-1} \cdot \theta \cdot \mathbf{u}, \text{ with } \mathbf{L}_{\theta u} \cdot \theta \cdot \mathbf{u} = \mathbf{L}_{\theta u}^{-1} \cdot \theta \cdot \mathbf{u} = \theta \cdot \mathbf{u}$$

$$\boldsymbol{\omega} = -\lambda \cdot \theta \cdot \mathbf{u}$$

□ Translation velocities are image-based:

$$\mathbf{v} = -\lambda \cdot \mathbf{L}_v^+ \cdot (\mathbf{e}_t - \mathbf{L}_\omega \cdot \mathbf{L}_{\theta u}^{-1} \cdot \mathbf{e}_\omega) = -\mathbf{L}_v^+ \cdot (\lambda \cdot \mathbf{e}_t - \lambda \cdot \mathbf{L}_\omega \cdot \theta \cdot \mathbf{u})$$

$$\mathbf{v} = -\mathbf{L}_v^+ \cdot (\lambda \cdot \mathbf{e}_t + \mathbf{L}_\omega \cdot \boldsymbol{\omega})$$

- The computed velocity depends on the translation part of the error and the angular velocity of the control

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Conclusion

- Visual servoing / Visual control
 - Controlling the position of a robot using vision information allows the positioning of the robot terminal element in an unstructured environment.

- Position-based visual servoing
 - Allows task planning in Cartesian space
 - 3D reconstruction is required
 - Sensitive to calibration errors
 - No control in the image. Object of interest may leave the field of view

- Image-based visual servoing
 - No explicit reconstruction needed. Eliminates geometrical model errors
 - Robustness with respect to camera and robot calibration errors
 - Local minima or singularities of the image Jacobian
 - No direct control in Cartesian space. Can produce elegant trajectories in the image but twisted in Cartesian space

- Hybrid visual servoing
 - Decoupling of degrees of freedom by combining control techniques

Conclusion

- Visual control challenges
 - One feature for each robot degree of freedom
 - Perfect decoupling between image features and robot degrees of freedom
 - Global stability: No singularities or local minima
 - Transformation to a linear system control

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