



Universidad
de Zaragoza

Control de robots y sistemas multi-robot basado en visión

Ciclo de conferencias
Master y Programa de Doctorado en
“Ingeniería de Sistemas y de Control”

UNED – ETS Ingeniería Informática

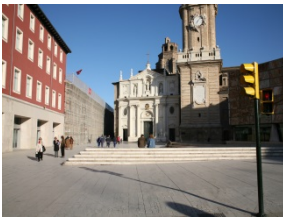
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Colaboradores:

Gonzalo López Nicolás
Héctor Manuel Becerra
Rosario Aragüés
Eduardo Montijano
Miguel Aranda

Carlos Sagues
Universidad de Zaragoza
<http://www.unizar.es/~csagues>

Motivation



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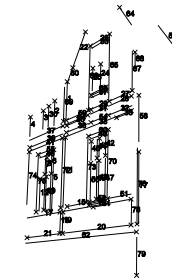
- ◆ Features. FM, H, TT (Fundamental Matriz, Homography and Trifocal Tensor)
- ◆ Visual mobile robot control
 - FM based
 - H based
 - TT based
 - Long term navigation
- ◆ Control of Multi-robot systems
 - Data association
 - Coordinated motion with epipoles
 - Central decision with flying camera on scene - Homography

Features

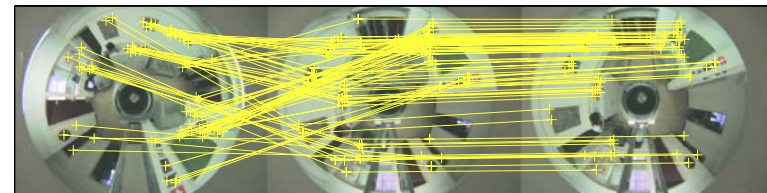
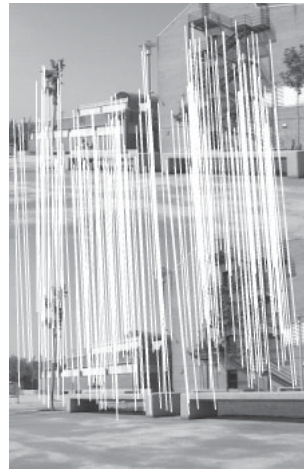
◆ Harris corner extractor



◆ Lines

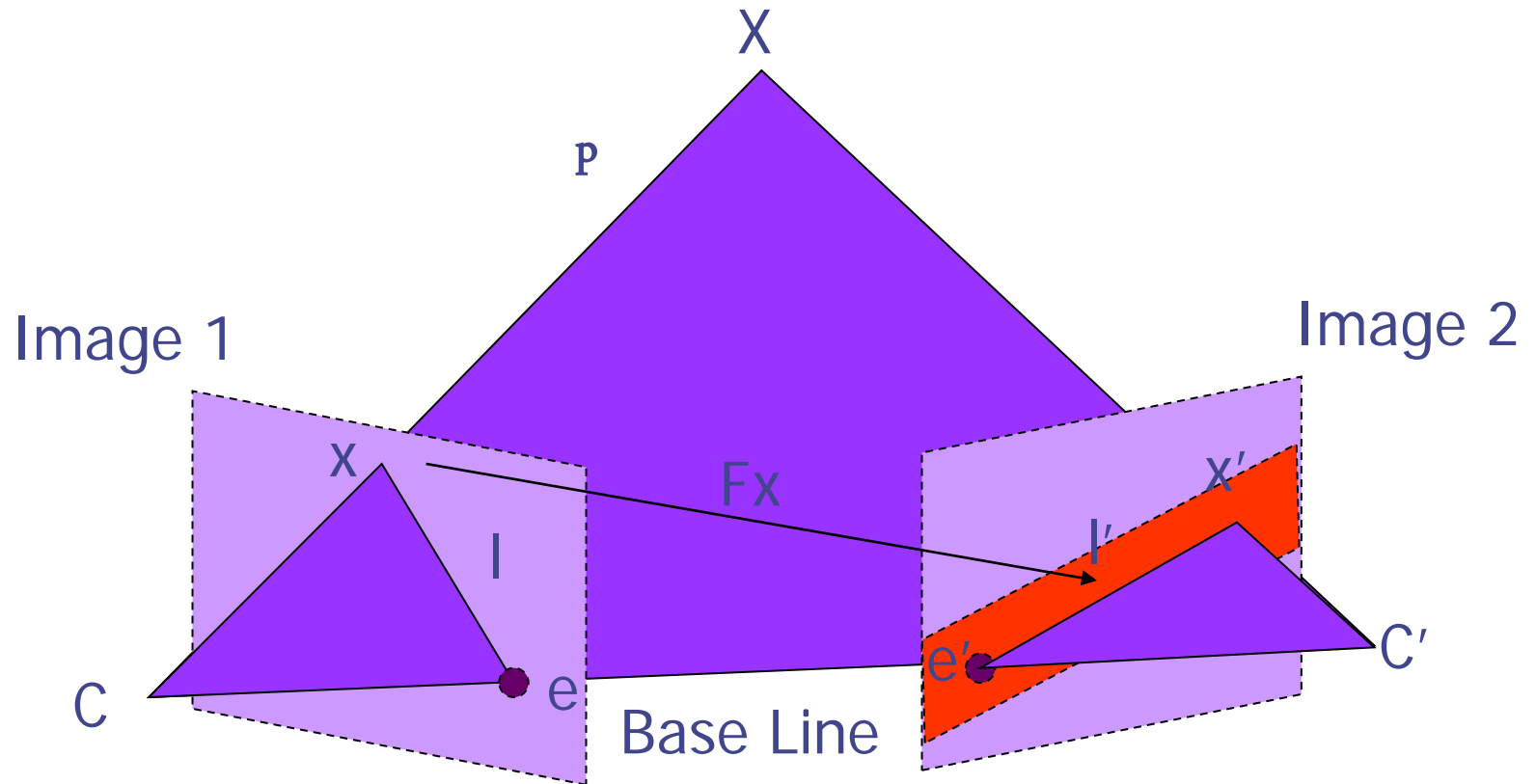


◆ SIFT



◆ SURF

FM: Fundamental Matrix



FM: Matriz Fundamental

◆ Fundamental Matrix

- Matrix 3x3 satisfying: $\mathbf{x}'^T \mathbf{F} \mathbf{x} = 0$
 - ◆ Independent of scene structure

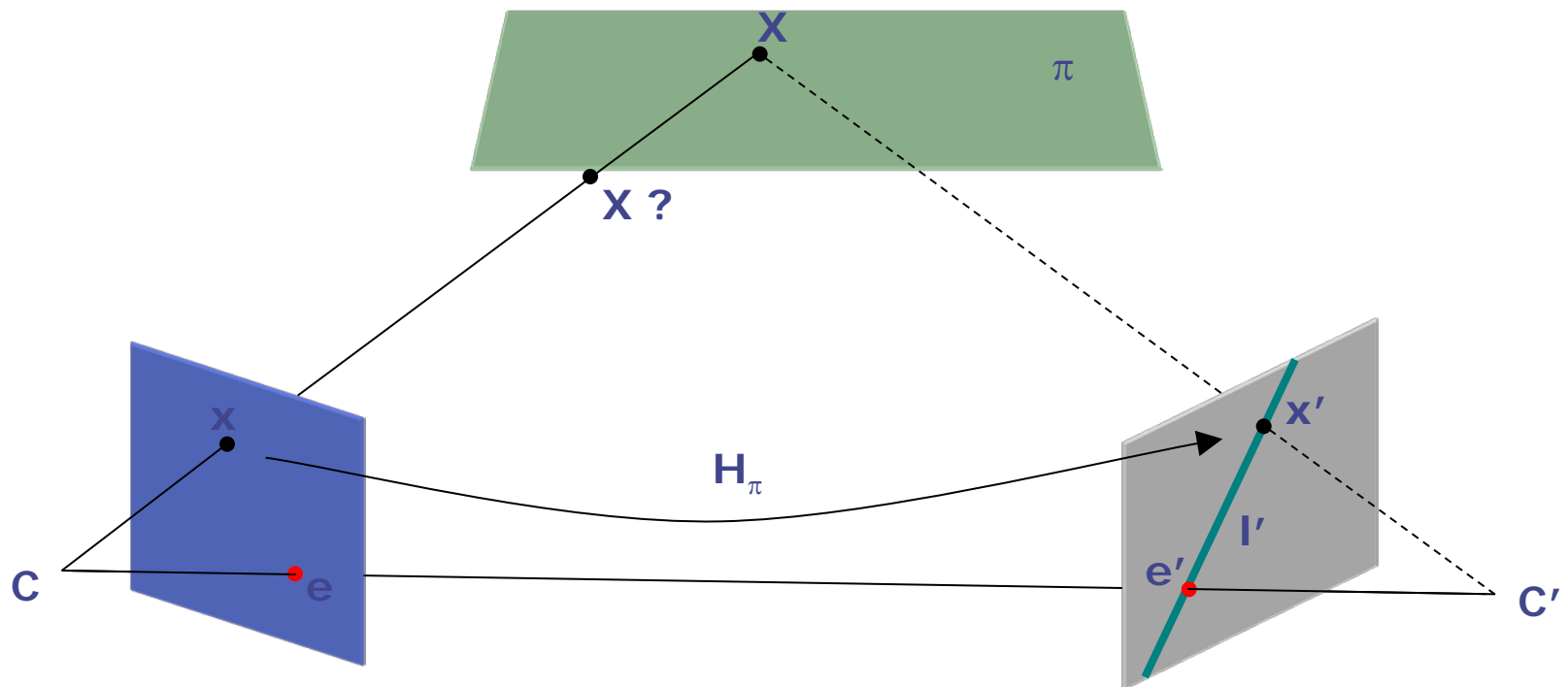
$$\mathbf{F} = \begin{bmatrix} f_{11} & f_{12} & f_{13} \\ f_{21} & f_{22} & f_{23} \\ f_{31} & f_{32} & f_{33} \end{bmatrix}$$

- As a dot product:
- $(x' \cdot x, x' \cdot y, x', y' \cdot x, y' \cdot y, y', x, y, 1) \cdot \mathbf{f} = 0$
- With 8 points we have: $\mathbf{A} \cdot \mathbf{f} = 0$
 - ◆ 8 points \Rightarrow Solution to scale factor
 - ◆ $\text{SVD}(\mathbf{A}) \Rightarrow$ Singular vector of smallest singular value

$$\mathbf{F} = \mathbf{K}_2^{-T} ([\mathbf{t}]_{\times} \mathbf{R}) \mathbf{K}_1^{-1}$$

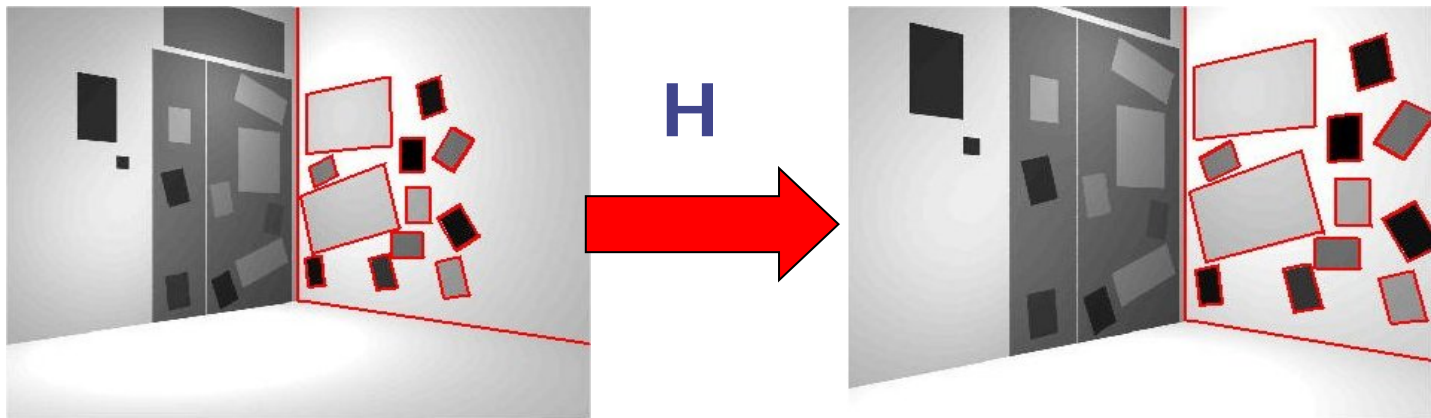
H: Homography

- ◆ Projective transformation between two planes



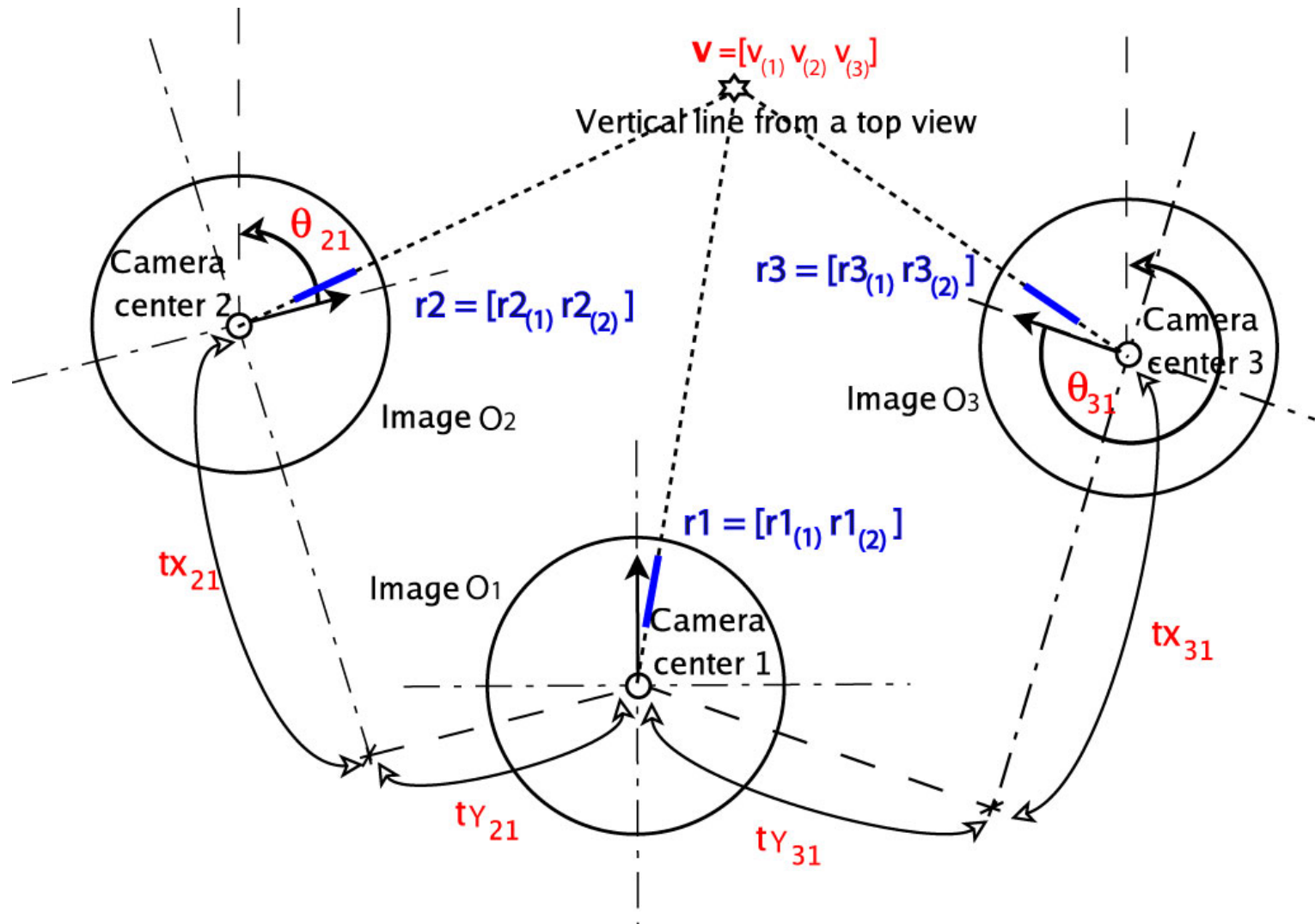
H: Homography

$$\begin{bmatrix} x'_1 \\ x'_2 \\ x'_3 \end{bmatrix} = \begin{pmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{pmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} \Leftrightarrow \mathbf{x}' = \mathbf{H}\mathbf{x}$$



$$\mathbf{H} = \mathbf{K} \left(\mathbf{R} - \mathbf{t} \frac{\mathbf{n}^T}{d} \right) \mathbf{K}^{-1}$$

TT: Trifocal tensor (1D)



TT: Trifocal tensor

$$\lambda_1 \mathbf{r}_1 = \mathbf{P}_1 \mathbf{v}$$

$$\lambda_2 \mathbf{r}_2 = \mathbf{P}_2 \mathbf{v}$$

$$\lambda_3 \mathbf{r}_3 = \mathbf{P}_3 \mathbf{v}$$

$$\begin{bmatrix} \mathbf{P}_1 & \mathbf{r}_1 & 0 & 0 \\ \mathbf{P}_2 & 0 & \mathbf{r}_2 & 0 \\ \mathbf{P}_3 & 0 & 0 & \mathbf{r}_3 \end{bmatrix} [\mathbf{v}, -\lambda_1, -\lambda_2, -\lambda_3]^T = 0 \quad \left| \begin{array}{cccc} \mathbf{P}_1 & \mathbf{r}_1 & 0 & 0 \\ \mathbf{P}_2 & 0 & \mathbf{r}_2 & 0 \\ \mathbf{P}_3 & 0 & 0 & \mathbf{r}_3 \end{array} \right| = 0$$

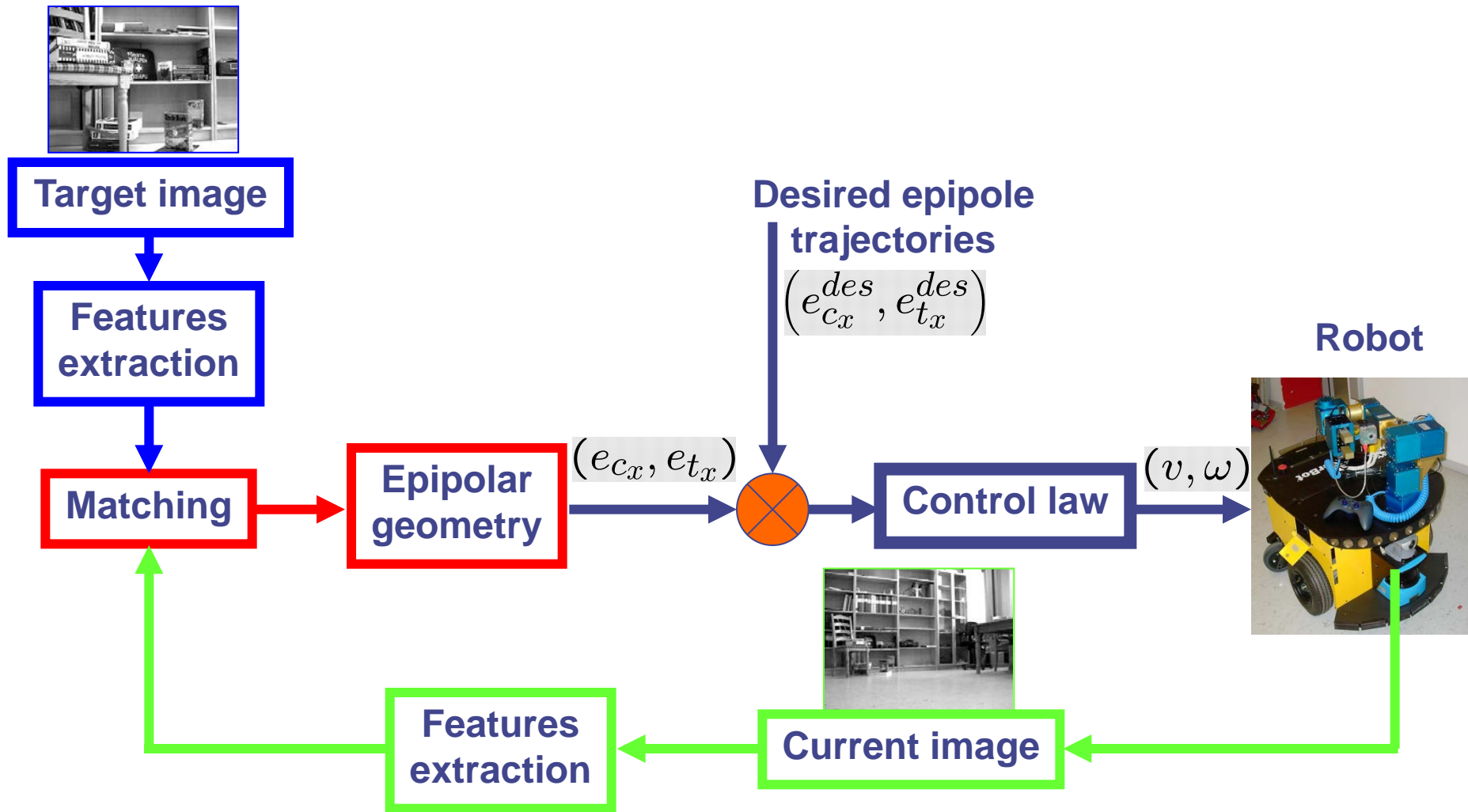
$$\sum_{i=1}^2 \sum_{j=1}^2 \sum_{k=1}^2 T_{ijk} \mathbf{r}_1(i) \mathbf{r}_2(j) \mathbf{r}_3(k) = 0$$

$$\begin{aligned} T_{111} &= t'_z \sin \theta'' - t''_z \sin \theta'; & T_{211} &= -t'_z \cos \theta'' + t''_z \cos \theta' \\ T_{112} &= t'_z \cos \theta'' + t''_x \sin \theta'; & T_{212} &= t'_z \sin \theta'' - t''_x \cos \theta' \\ T_{121} &= -t'_x \sin \theta'' - t''_z \cos \theta'; & T_{221} &= t'_x \cos \theta'' - t''_z \sin \theta' \\ T_{122} &= -t'_x \cos \theta'' + t''_x \cos \theta'; & T_{222} &= -t'_x \sin \theta'' + t''_x \sin \theta'. \end{aligned}$$

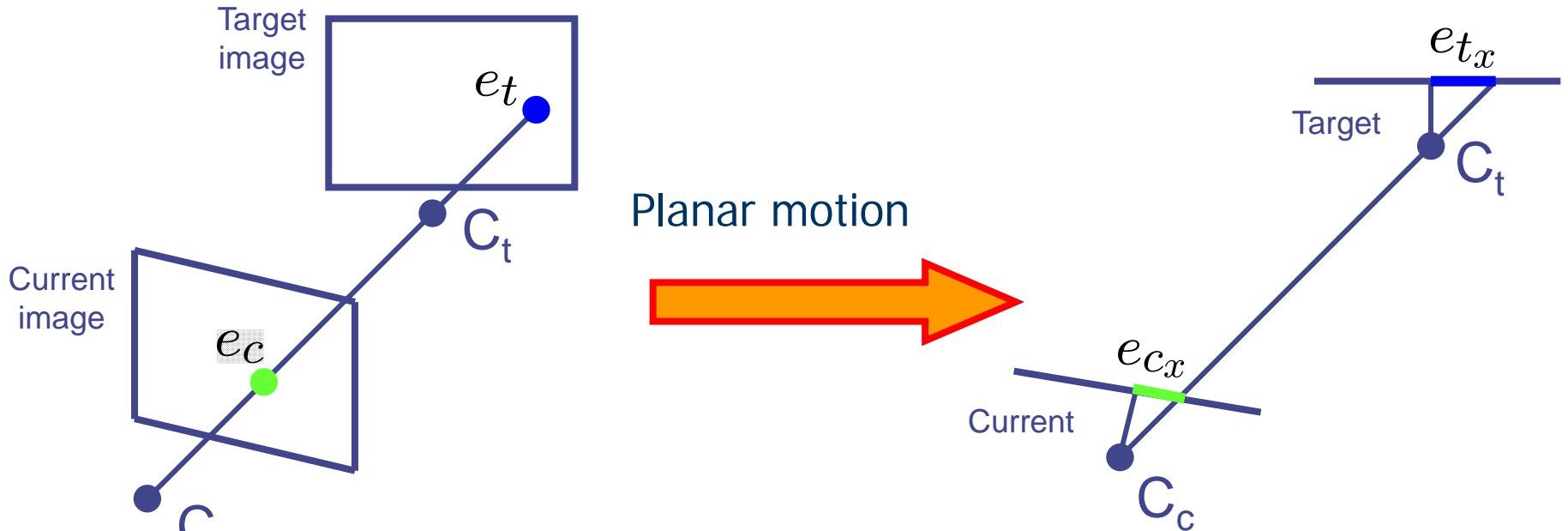
The tensor 1D has 2x2x2 elements, $wl < 3w-3+2l-1$, 5 features needed

The 2D tensor has 3x3x3 elements

Nonholonomic Epipolar Visual Servoing – FM based



Nonholonomic Epipolar Visual Servoing – FM based



$$e_{tx} = \alpha_x \frac{x}{z}$$

$$e_{cx} = \alpha_x \frac{x \cos \theta - z \sin \theta}{z \cos \theta + x \sin \theta}$$

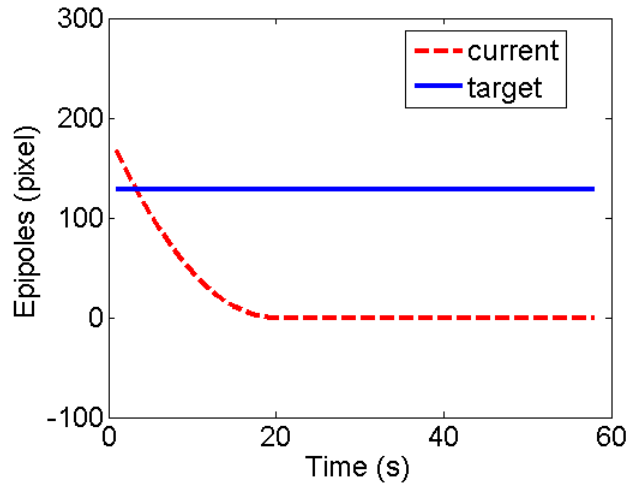
$$\dot{e}_{cx} = -\frac{\alpha_x \cos(\theta + \psi)}{d \sin^2(\theta + \psi)} v - \frac{\alpha_x}{\sin^2(\theta + \psi)} \omega$$

$$\dot{e}_{tx} = -\frac{\alpha_x \cos(\theta + \psi)}{d \sin^2(\psi)} v$$

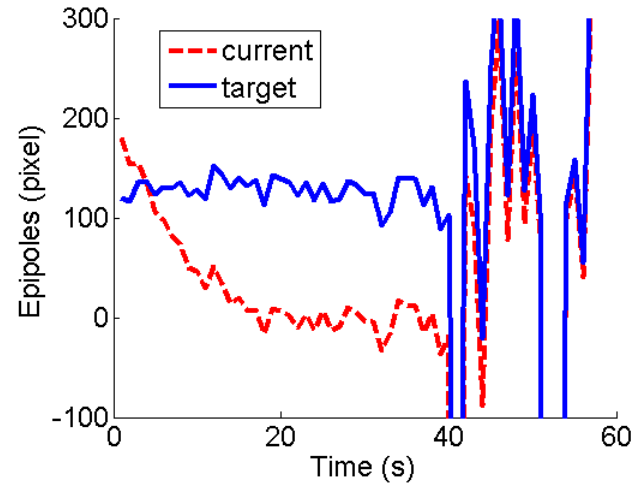
$$\begin{pmatrix} v \\ \omega \end{pmatrix} = L^{-1} \begin{pmatrix} v_c \\ v_t \end{pmatrix} \quad \text{with} \quad L = \begin{bmatrix} -\frac{\alpha_x \cos(\theta + \psi)}{d \sin^2(\theta + \psi)} & -\frac{\alpha_x}{\sin^2(\theta + \psi)} \\ -\frac{\alpha_x \cos(\theta + \psi)}{d \sin^2(\psi)} & 0 \end{bmatrix}$$

Nonholonomic Epipolar Visual Servoing – FM based

Desired epipole trajectories



Epipoles evolution



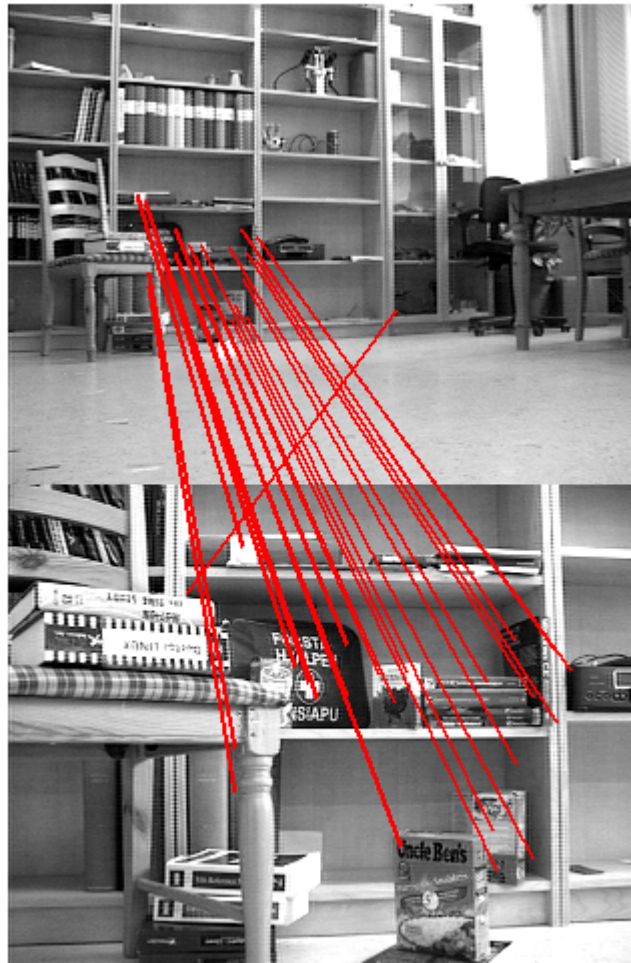
Invertible if $\det \neq 0$

$$\det(L) = -\alpha^2_x \cos(\theta + \psi) / d \sin^2(\psi) \sin^2(\theta + \psi)$$

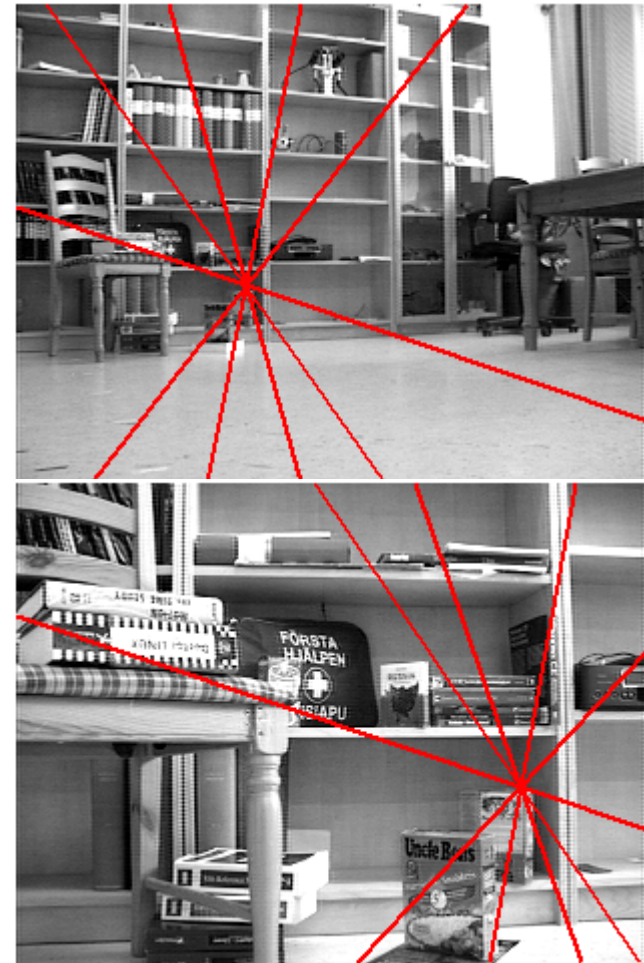
Singularidad $e_{cx} = 0$

$$(\theta + \psi) = 90^\circ$$

Nonholonomic Epipolar Visual Servoing – FM based



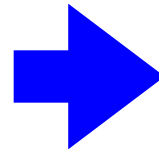
Matches



Epipoles and epipolar lines

Nonholonomic Epipolar Visual Servoing – FM based

Target position



Target image



Current image



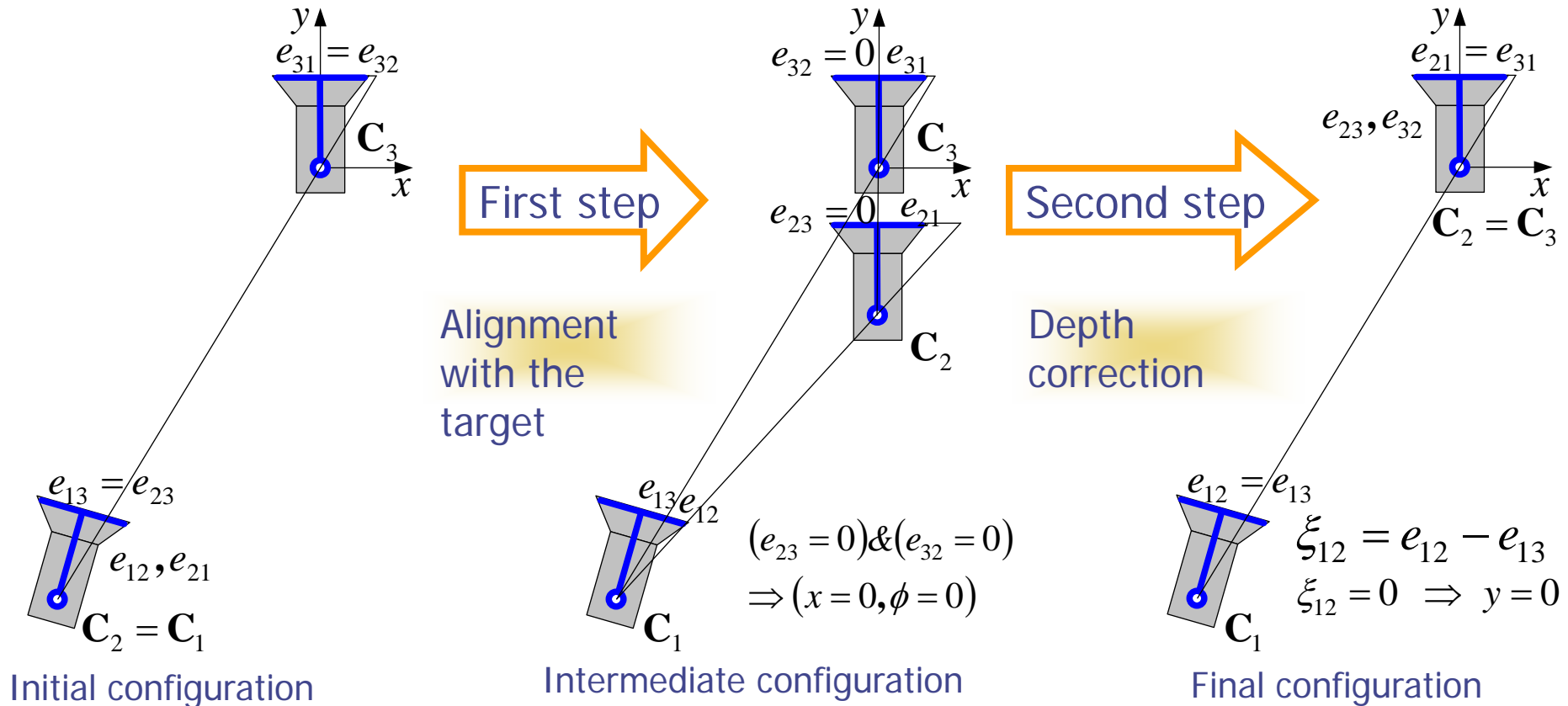
Nonholonomic Epipolar Visual Servoing – FM based



Nonholonomic Epipolar Visual Servoing – FM based

Sliding mode control to avoid singularity

- The control task is carried out in two steps:



Nonholonomic Epipolar Visual Servoing – FM based

- ◆ **Control goal of the step** – Solve the stabilization problem in the following error system, where $\xi_{23} = e_{23} - e_{23}^d(t)$, $\xi_{32} = e_{32} - e_{32}^d(t)$.

$$\begin{bmatrix} \dot{\xi}_{23} \\ \dot{\xi}_{32} \end{bmatrix} = \begin{bmatrix} -\frac{\alpha_x \sin(\phi_2 - \psi_2)}{d_{23} \cos^2(\phi_2 - \psi_2)} & \frac{\alpha_x}{\cos^2(\phi_2 - \psi_2)} \\ -\frac{\alpha_x \sin(\phi_2 - \psi_2)}{d_{23} \cos^2(\psi_2)} & 0 \end{bmatrix} \begin{bmatrix} v \\ \omega \end{bmatrix} - \begin{bmatrix} \dot{e}_{23}^d \\ \dot{e}_{32}^d \end{bmatrix} = \mathbf{L}(\phi_2, \psi_2) \mathbf{u} - \dot{\mathbf{e}}^d$$

Desired trajectories

$$e_{23}^d(t) = \sigma \frac{e_{23}(0)}{2} \left(1 + \cos\left(\frac{\pi}{\tau} t\right) \right)$$

$$e_{32}^d(t) = \frac{e_{32}(0)}{2} \left(1 + \cos\left(\frac{\pi}{\tau} t\right) \right)$$

where $\mathbf{L}(\phi, \psi)$ is the so-called decoupling matrix.

- Sliding mode control with sliding surfaces

$$\mathbf{s} = \begin{bmatrix} s_c \\ s_t \end{bmatrix} = \begin{bmatrix} \xi_{23} \\ \xi_{32} \end{bmatrix} = \begin{bmatrix} e_{23} - e_{23}^d \\ e_{32} - e_{32}^d \end{bmatrix} = \mathbf{0}.$$

- Decoupling-based controller.

$$\mathbf{u}_{db} = \begin{bmatrix} v_{db} \\ \omega_{db} \end{bmatrix} = \frac{1}{\alpha_{xe}} \begin{bmatrix} 0 & -\frac{d_{23e} \cos^2(\psi_2)}{\sin(\phi_2 - \psi_2)} \\ \cos^2(\phi_2 - \psi_2) & -\cos^2(\psi_2) \end{bmatrix} \begin{bmatrix} u_c \\ u_t \end{bmatrix}$$

Singularity if
 $\phi_2 - \psi_2 = n\pi$

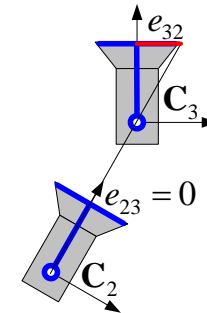
where $u_c = \dot{e}_{23}^d - \lambda_c s_c - \kappa_c \text{sign}(s_c)$,

$u_t = \dot{e}_{32}^d - \lambda_t s_t - \kappa_t \text{sign}(s_t)$

Nonholonomic Epipolar Visual Servoing – FM based

- ◆ A singular pose is shown in the figure

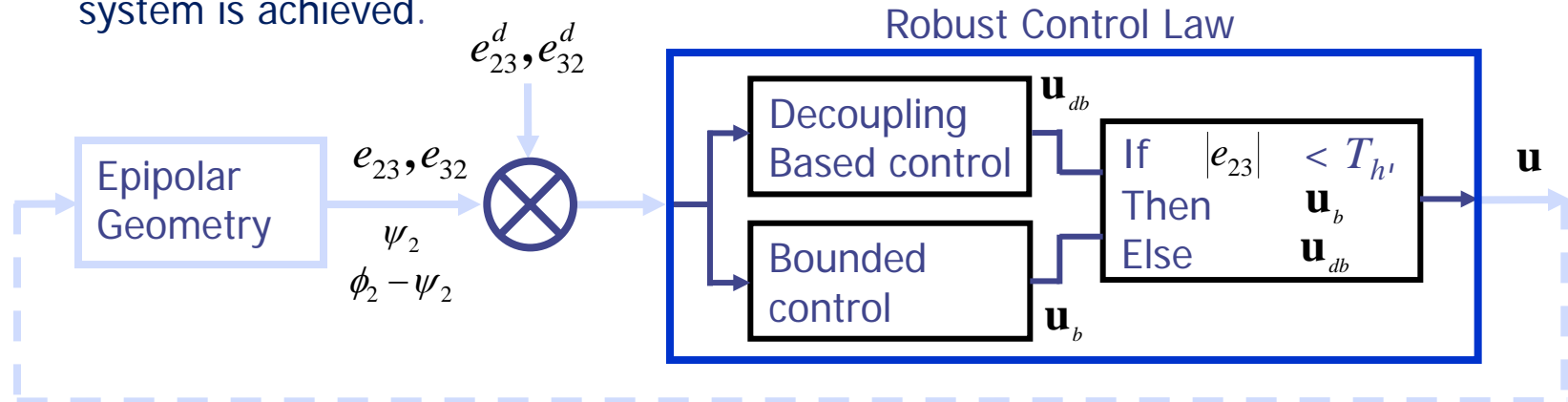
$$\phi - \psi = \arctan(e_{23} / \alpha_x) = 0$$



- ◆ **Bounded controller.** These inputs don't use the decoupling matrix

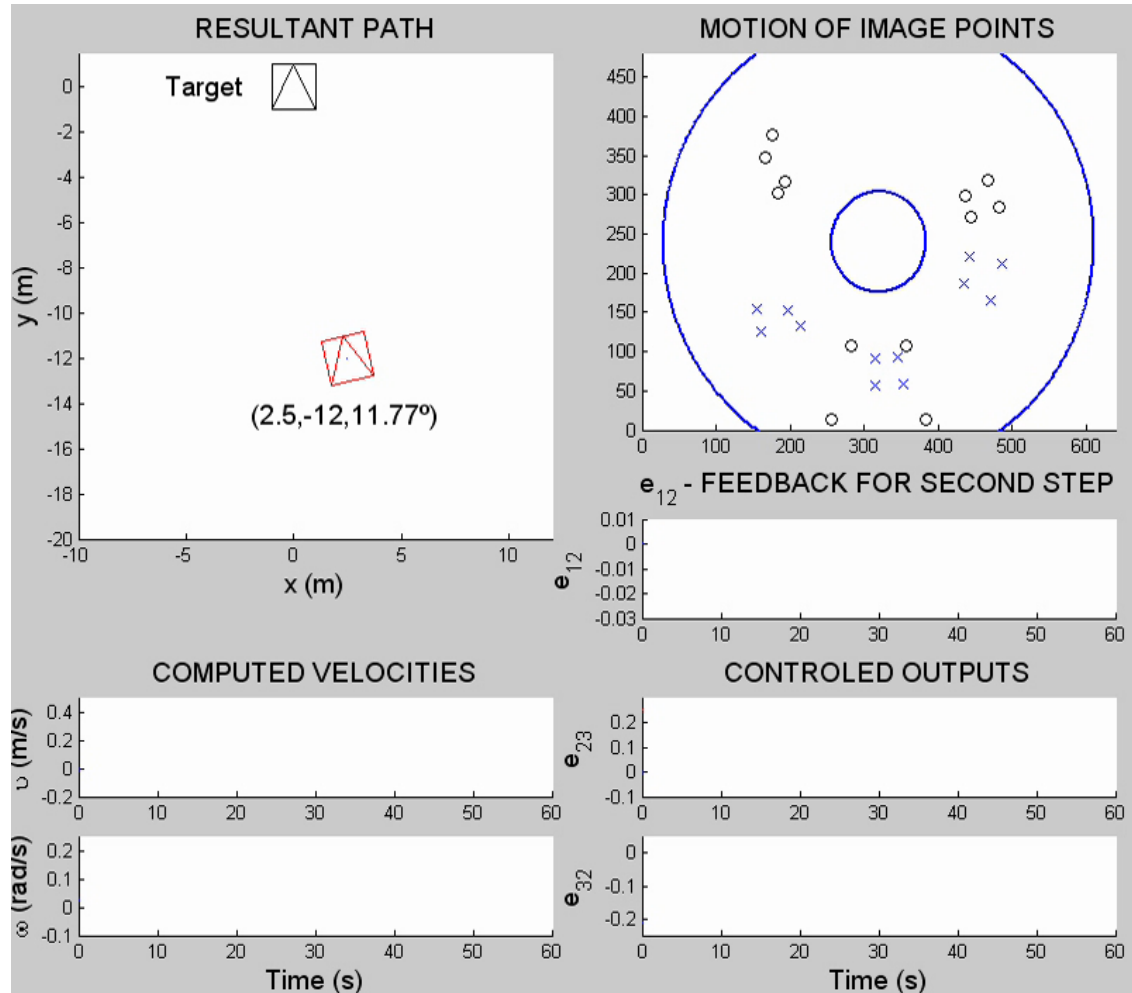
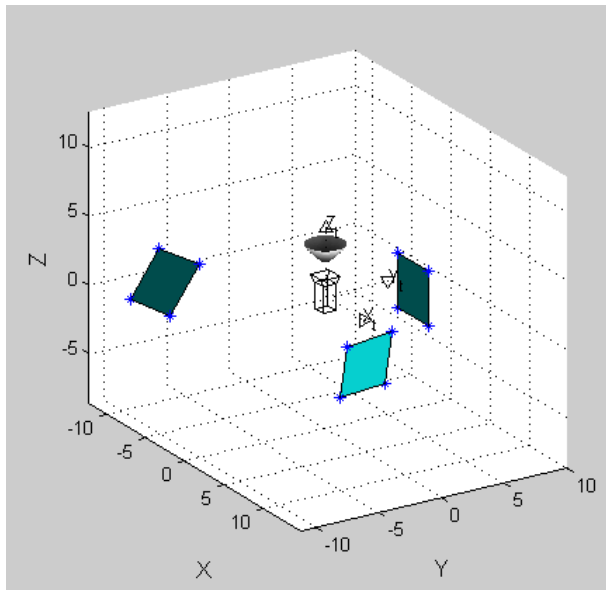
$$\mathbf{u}_b = \begin{bmatrix} v_b \\ \omega_b \end{bmatrix} = \begin{bmatrix} k_v \text{sign}(s_t \sin(\phi_2 - \psi_2)) \\ -k_\omega \text{sign}(s_c) \end{bmatrix}.$$

- ◆ This is a local control law for the error system.
- By switching between controllers accordingly, robust global stabilization of the error system is achieved.

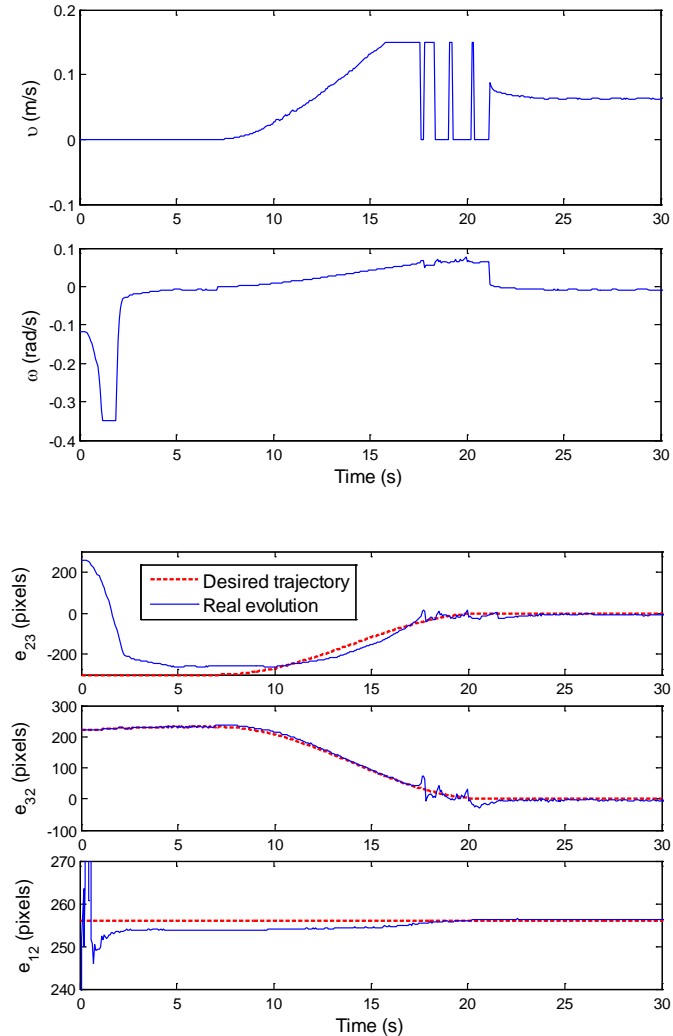
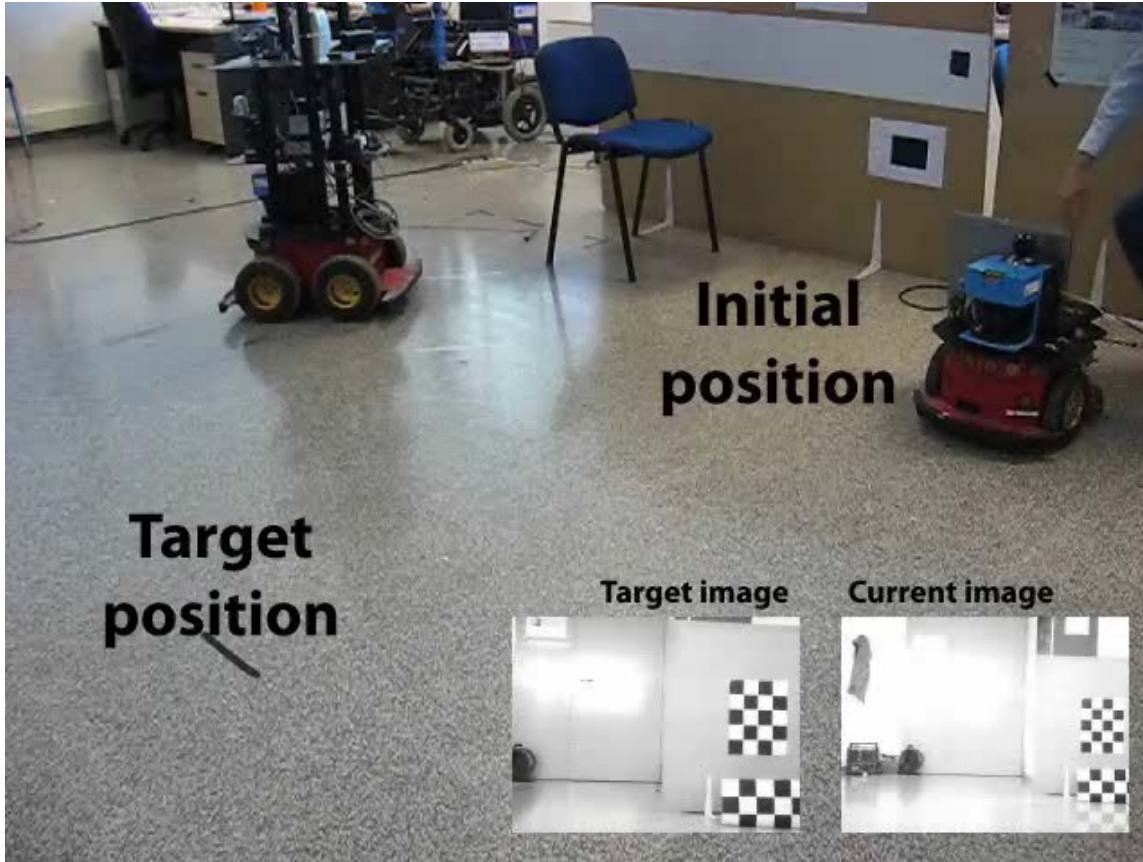


Nonholonomic Epipolar Visual Servoing – FM based

- The epipoles are computed from synthetic images of size 640x480 pixels.
- Target location is $(0,0,0^\circ)$.
- Virtual scene:



Nonholonomic Epipolar Visual Servoing – FM based

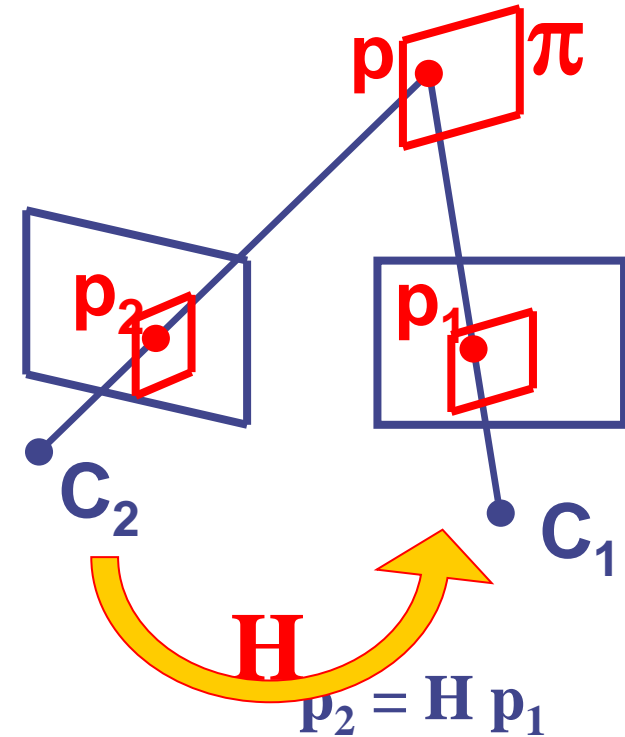


Nonholonomic Homography based – H based

- ◆ Two images can be geometrically linked by a homography
- ◆ The homography is generated by a plane of the scene
- ◆ The homography can be computed from point matches

$$\mathbf{H} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ h_{31} & h_{32} & h_{33} \end{bmatrix}$$

- ◆ Goal: $\mathbf{H} = \mathbf{I}$



Nonholonomic Homography based – H based

◆ The homography is related to camera motion:

$$\mathbf{H} = \mathbf{K} \left(\mathbf{R} - \mathbf{t} \frac{\mathbf{n}^T}{d} \right) \mathbf{K}^{-1}$$

◆ Planar motion:

$$\mathbf{H} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ 0 & 1 & 0 \\ h_{31} & h_{32} & h_{33} \end{bmatrix} \quad \text{with:} \quad \begin{cases} h_{11} = \cos \phi + (x \cos \phi + z \sin \phi) \frac{n_x}{d} \\ h_{12} = \frac{\alpha_x}{\alpha_y} (x \cos \phi + z \sin \phi) \frac{n_y}{d} \\ h_{13} = \alpha_x \left(\sin \phi + (x \cos \phi + z \sin \phi) \frac{n_z}{d} \right) \\ h_{31} = \frac{1}{\alpha_x} \left(-\sin \phi + (-x \sin \phi + z \cos \phi) \frac{n_x}{d} \right) \\ h_{32} = \frac{1}{\alpha_y} \left(-x \sin \phi + z \cos \phi \right) \frac{n_y}{d} \\ h_{33} = \cos \phi + (-x \sin \phi + z \cos \phi) \frac{n_z}{d} \end{cases}$$

◆ Non-linear relation of H with state system:

$$(x, z, \phi)^T \longleftrightarrow h_{ij}$$

Nonholonomic Homography based VS

- ◆ 2 dof system → Two elements of the homography are enough to define the control
- ◆ Derivatives of the output functions:

$$\begin{cases} \dot{h}_{13} = \alpha_x h_{33} \omega \\ \dot{h}_{33} = \frac{n_z}{d} v - \frac{h_{13}}{\alpha_x} \omega \end{cases}$$

- ◆ State space form

$$\begin{cases} \dot{\mathbf{x}} = f(\mathbf{x}, \mathbf{u}) \\ \mathbf{y} = h(\mathbf{x}) \end{cases} \quad \text{with: } \begin{cases} \text{State vector: } \mathbf{x} = (x, z, \phi)^T \\ \text{Input vector: } \mathbf{u} = (v, \omega)^T \\ \text{Output vector: } \mathbf{y} = (h_{13}, h_{33})^T \end{cases}$$

- ◆ Linear relation between the input and output

$$\begin{pmatrix} v \\ \omega \end{pmatrix} = \mathbf{L}^{-1} \begin{pmatrix} \nu_{13} \\ \nu_{33} \end{pmatrix} \quad \text{with: } \mathbf{L} = \begin{bmatrix} 0 & \alpha_x h_{33} \\ \frac{n_z}{d} & -\frac{h_{13}}{\alpha_x} \end{bmatrix}$$

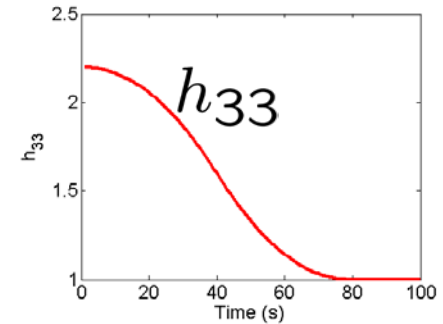
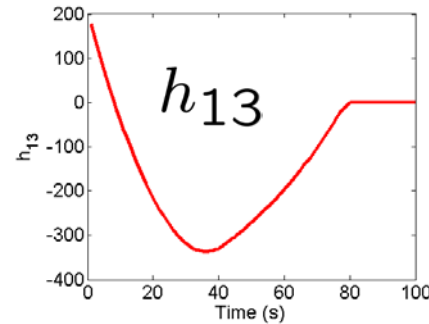
Nonholonomic Homography based – H based

Tracking of the desired trajectories of the homography elements

Input of the control:

Exponentially stable error dynamics

$$\begin{pmatrix} \nu_{13} \\ \nu_{33} \end{pmatrix} = \begin{pmatrix} \dot{h}_{13}^d - k_{13}(h_{13} - h_{13}^d) \\ \dot{h}_{33}^d - k_{33}(h_{33} - h_{33}^d) \end{pmatrix}$$

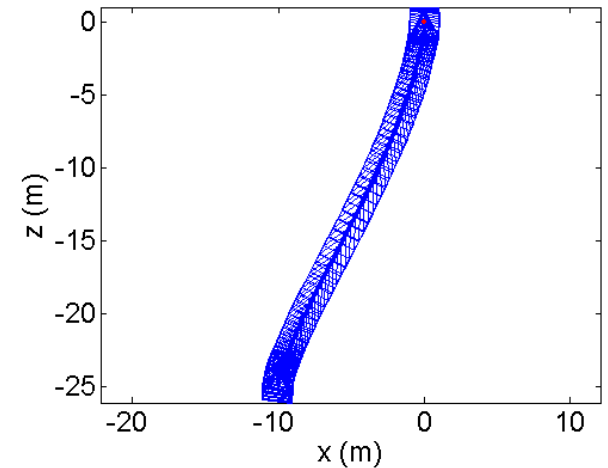


Desired trajectories:

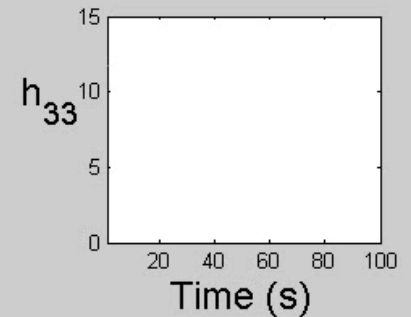
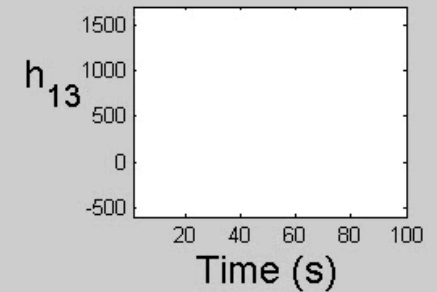
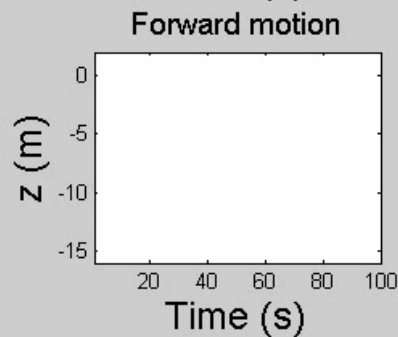
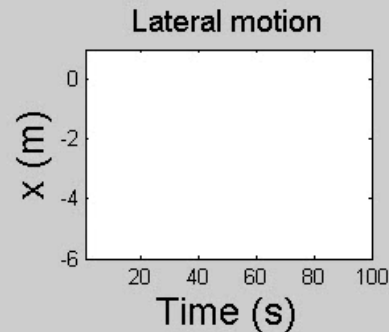
$$0 \leq t \leq T_1 \begin{cases} h_{13}^d(t) = (h_{13}(0) - gt) \left(\frac{t^2}{T_1^2} - 2\frac{t}{T_1} + 1 \right) + gt \\ h_{33}^d(t) = \left(\frac{1-h_{33}(0)}{2} \right) \left(\frac{t^2}{T_1^2} + 1 \right) + (3h_{33}(0) - 1)/2 \end{cases}$$

$$T_1 < t \leq T_2 \begin{cases} h_{13}^d(t) = h_{13}(T_1) \frac{\phi_t(t)}{\phi_t(T_1)} \\ h_{33}^d(t) = \left(\frac{h_{33}(0)-1}{2} \right) \left(\frac{(t-T_1)^2}{(T_2-T_1)^2} - 2\frac{t-T_1}{T_2-T_1} + 1 \right) + 1 \end{cases}$$

$$t > T_2 \begin{cases} h_{13}^d(t) = 0 \\ h_{33}^d(t) = 1 \end{cases}$$

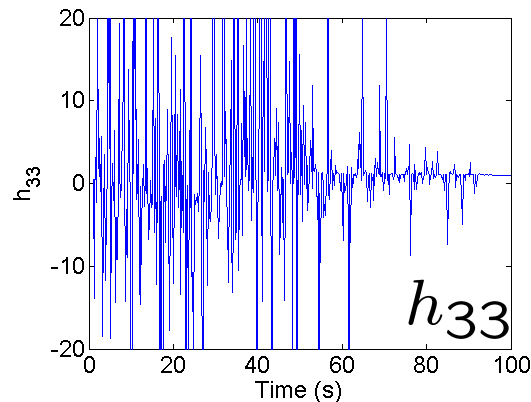
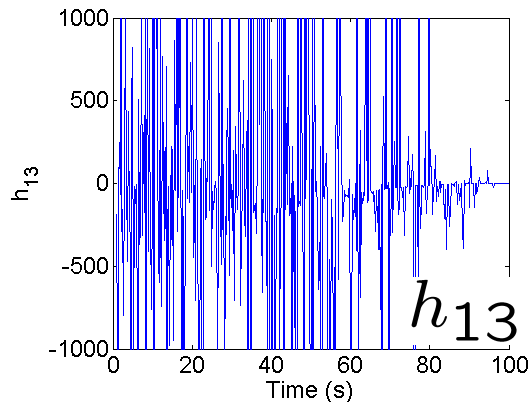
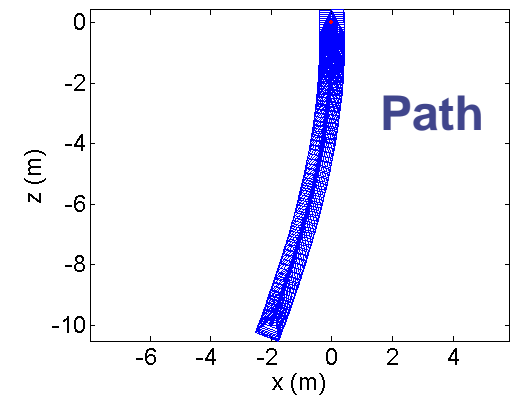
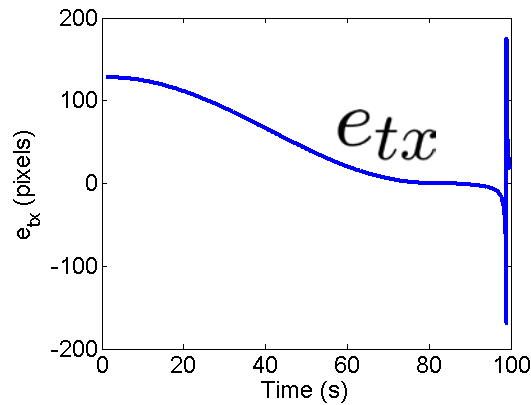
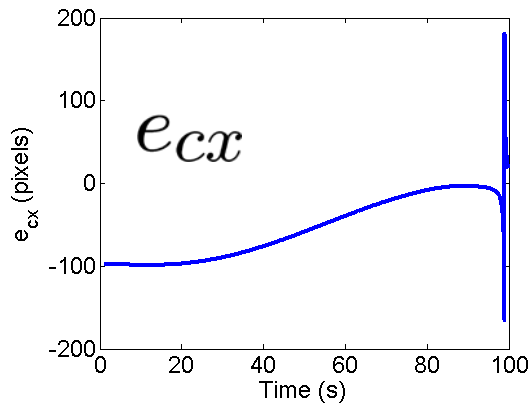


Nonholonomic Homography based – H based





Combination of Epipoles/Homographies for VS



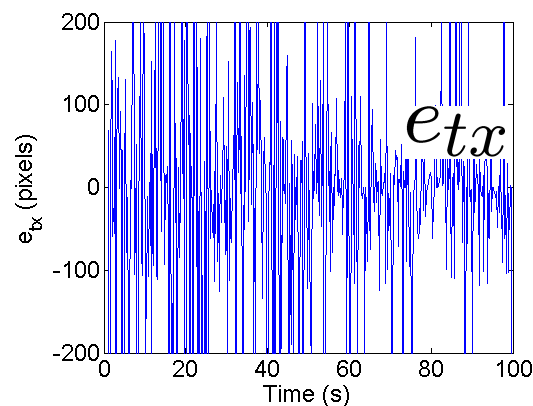
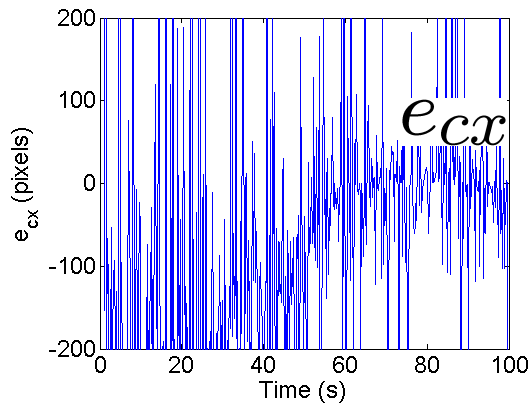
✓ Epipoles

✗ Homography

Epipolar-based control:

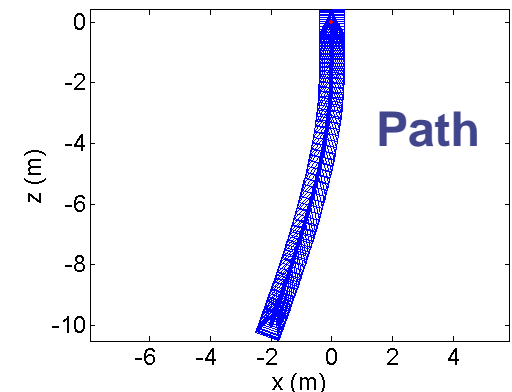
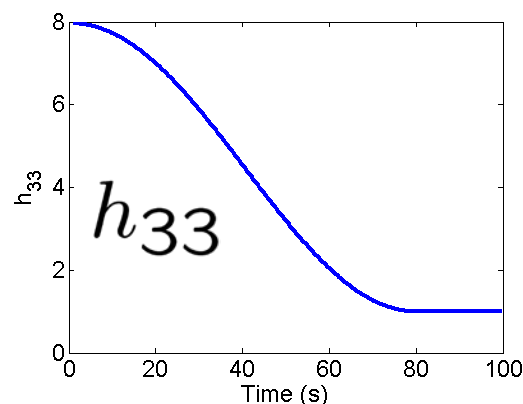
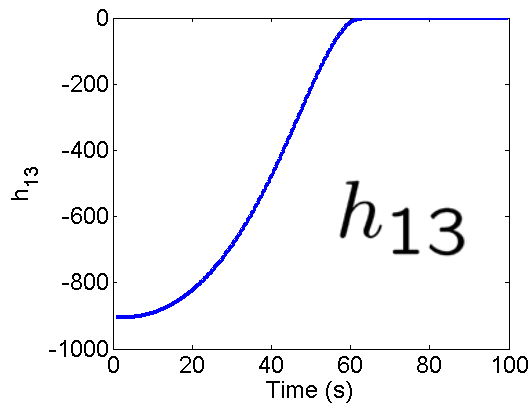
$$\begin{pmatrix} v_F \\ \omega_F \end{pmatrix} = \frac{1}{\alpha_x} \begin{bmatrix} 0 \\ \cos^2(\phi - \psi) \end{bmatrix} - \frac{d \cos^2(\psi)}{\sin(\phi - \psi)} \begin{bmatrix} - \\ \cos^2(\psi) \end{bmatrix} \begin{pmatrix} v_c \\ v_t \end{pmatrix}$$

Combination of Epipoles/Homographies for VS



✗ Epipoles

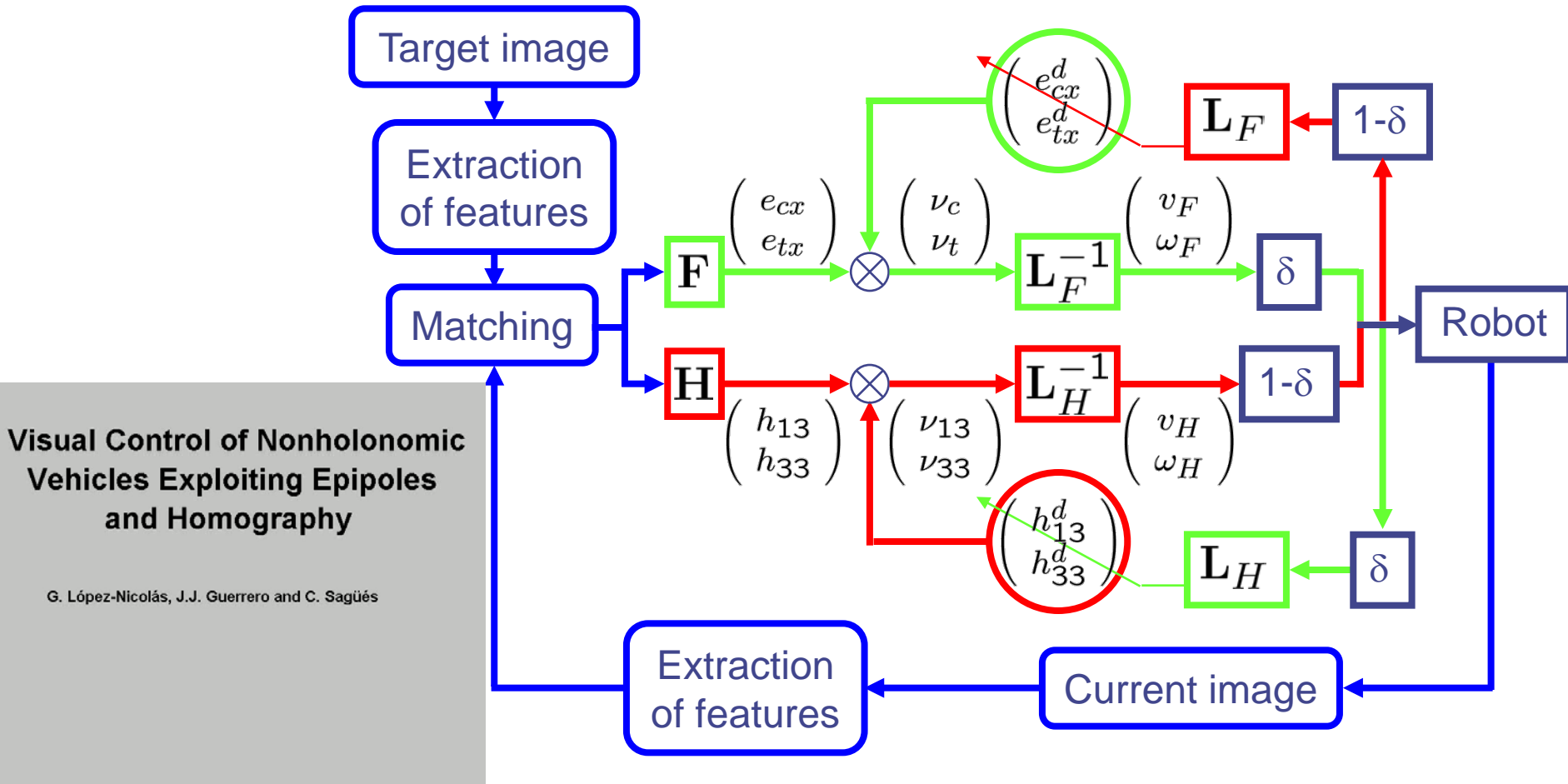
✓ Homography



Homography-based control:

$$\begin{pmatrix} v_H \\ \omega_H \end{pmatrix} = \begin{bmatrix} \frac{h_{13}}{\alpha_x^2 h_{33}} \frac{d\pi}{n_z} \\ \frac{1}{\alpha_x h_{33}} \end{bmatrix} \begin{pmatrix} v_{13} \\ v_{33} \end{pmatrix}$$

Combination of Epipoles/Homographies for VS



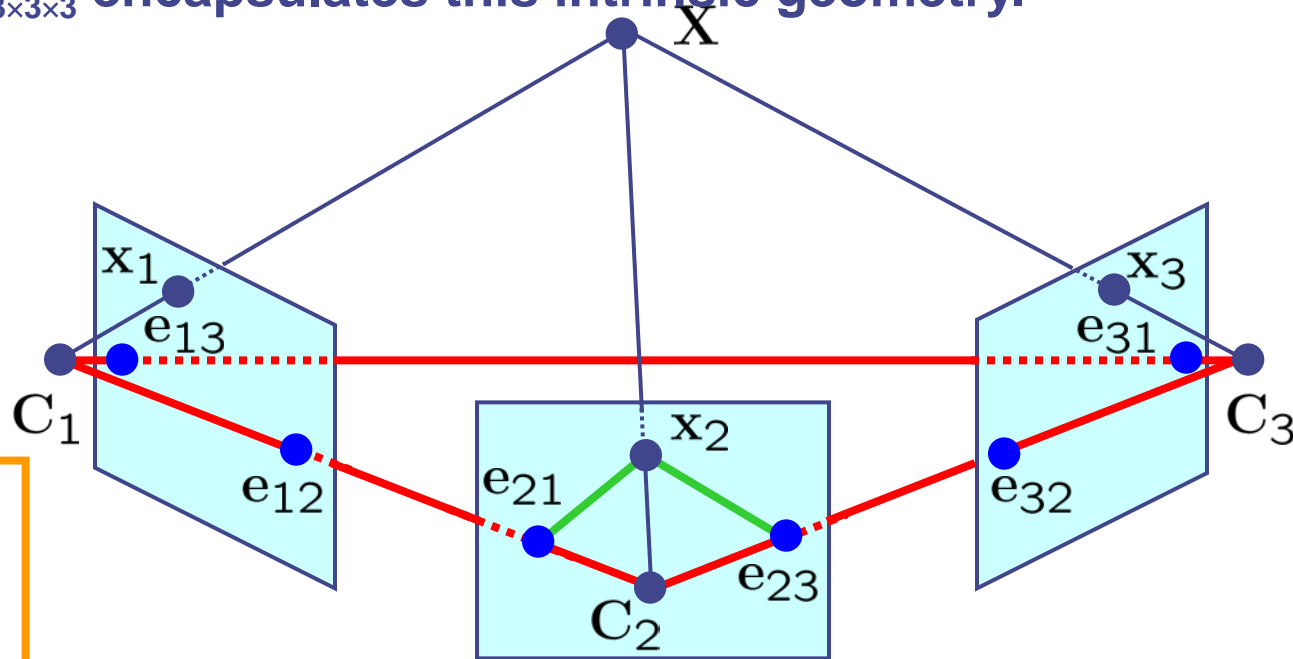
$$\begin{pmatrix} v \\ \omega \end{pmatrix} = \begin{bmatrix} 0 & \frac{-\delta d \cos^2(\psi)}{\alpha_x \sin(\phi - \psi)} & \frac{(1-\delta)h_{13}d\pi}{\alpha_x^2 h_{33} n_z} & \frac{(1-\delta)d\pi}{n_z} \\ \frac{\delta \cos^2(\phi - \psi)}{\alpha_x} & \frac{-\delta \cos^2(\psi)}{\alpha_x} & \frac{1-\delta}{\alpha_x h_{33}} & 0 \end{bmatrix} \begin{pmatrix} \nu_c \\ \nu_t \\ \nu_{13} \\ \nu_{33} \end{pmatrix}$$

Combination of Epipoles/Homographies for VS



Visual control – TT based

- The trifocal tensor is the intrinsic geometry between three views.
- It only depends on the camera internal parameters and relative pose.
- The trifocal tensor $T_{3 \times 3 \times 3}$ encapsulates this intrinsic geometry.



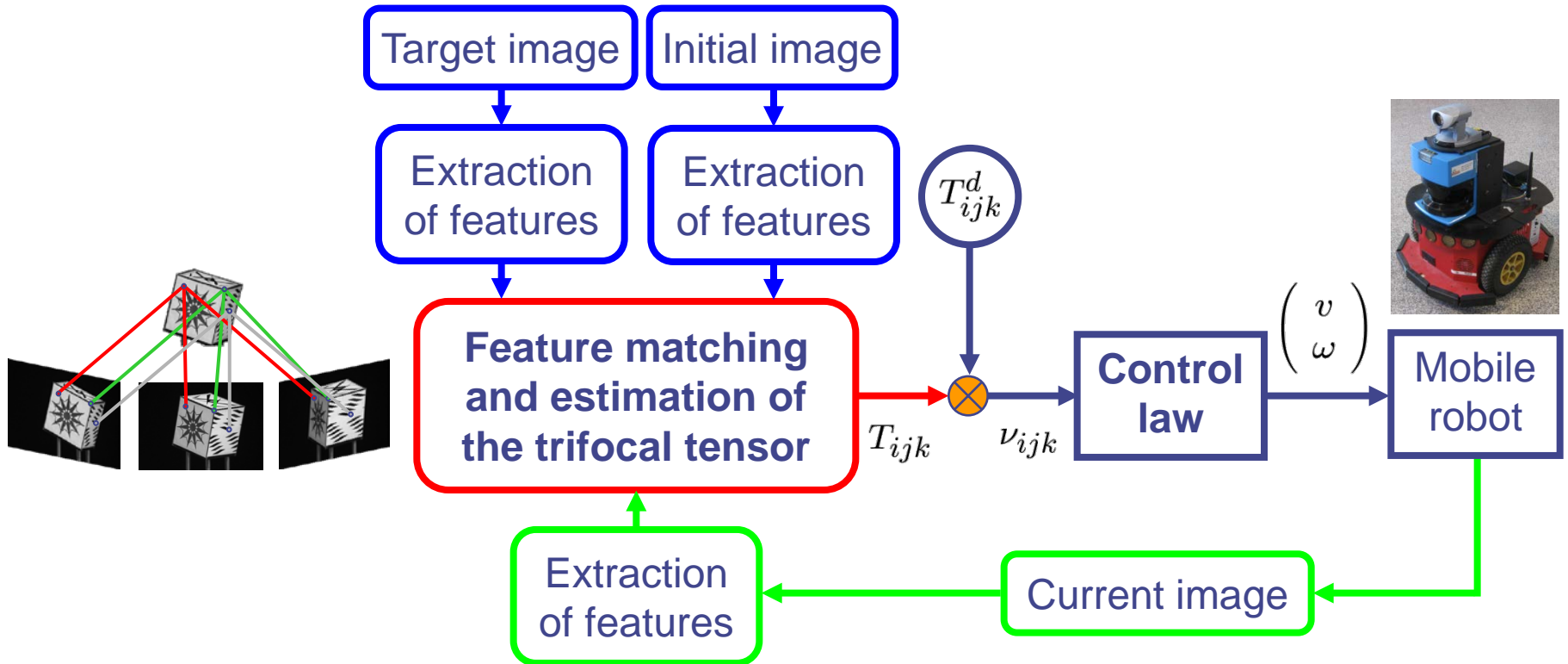
Matrix notation

$$T = [T_1, T_2, T_3]$$

$$[x_2]_{\times} \left(\sum_i x_1^i T_i \right) [x_3]_{\times} = 0_{3 \times 3}$$

Seven correspondences needed

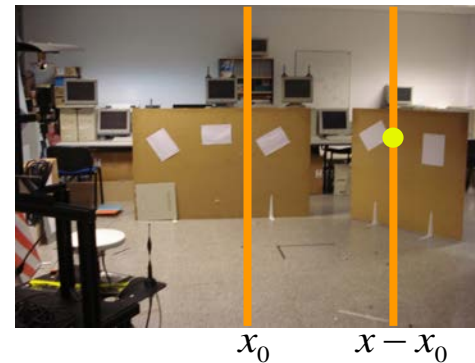
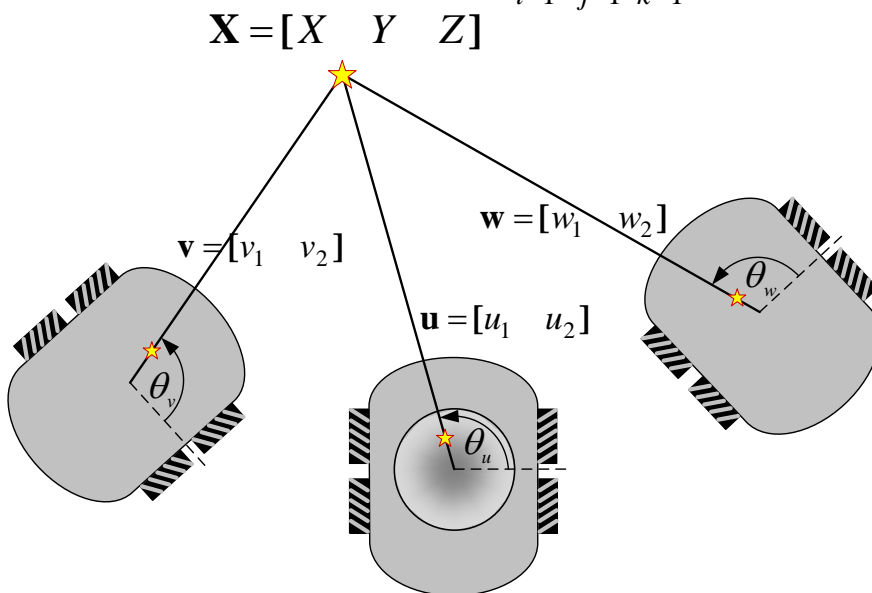
Visual control – TT based



Visual control – TT based

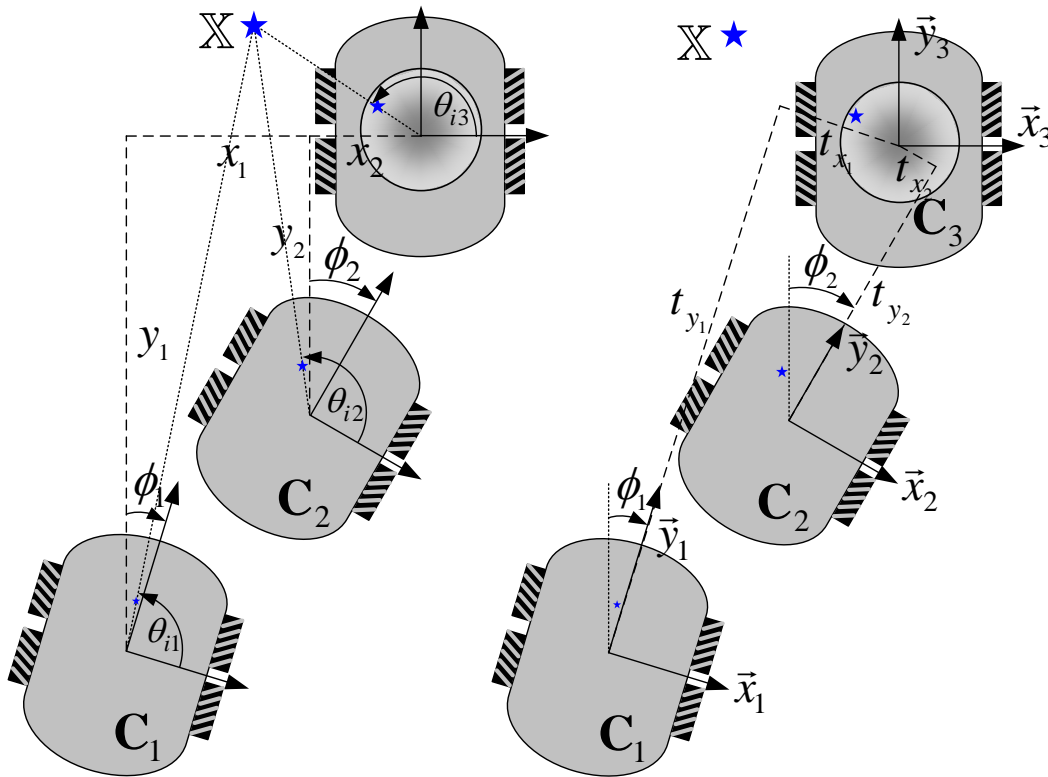
- Particularly the 1D trifocal tensor allows:
 - Exploit the bearing information.
 - Reduce the camera calibration parameters required for control (center of projection and vertical alignment).
- The trifocal tensor is a more general geometric constraint than epipolar geometry.
- Epipolar geometry is ill-conditioned with short baseline and with planar scenes.
- Five corresponding points

$$\sum_{i=1}^2 \sum_{j=1}^2 \sum_{k=1}^2 T_{ijk} u_i v_j w_k = 0$$



Visual control – TT based

- ◆ Initial location $C_1 = (x_1, y_1, \phi_1)$.
- ◆ Target location $C_3 = (0,0,0)$.
- ◆ Current location (moving camera) $C_2 = (x_2, y_2, \phi_2)$.



8 elements of the tensor:

$$T_{ijk}^m = \begin{bmatrix} T_{111}^m \\ T_{112}^m \\ T_{121}^m \\ T_{122}^m \\ T_{211}^m \\ T_{212}^m \\ T_{221}^m \\ T_{222}^m \end{bmatrix} = \begin{bmatrix} t_{y_1} \sin \phi_2 - t_{y_2} \sin \phi_1 \\ -t_{y_1} \cos \phi_2 + t_{y_2} \cos \phi_1 \\ t_{y_1} \cos \phi_2 + t_{x_2} \sin \phi_1 \\ t_{y_1} \sin \phi_2 - t_{x_2} \cos \phi_1 \\ -t_{x_1} \sin \phi_2 - t_{y_2} \cos \phi_1 \\ t_{x_1} \cos \phi_2 - t_{y_2} \sin \phi_1 \\ -t_{x_1} \cos \phi_2 + t_{x_2} \cos \phi_1 \\ -t_{x_1} \sin \phi_2 + t_{x_2} \sin \phi_1 \end{bmatrix}$$

where the relative locations between cameras are given as

$$\begin{bmatrix} t_{x_i} \\ t_{y_i} \end{bmatrix} = - \begin{bmatrix} \cos \phi_i & \sin \phi_i \\ -\sin \phi_i & \cos \phi_i \end{bmatrix} \cdot \begin{bmatrix} x_i \\ y_i \end{bmatrix}$$

for $i = 1, 2$.

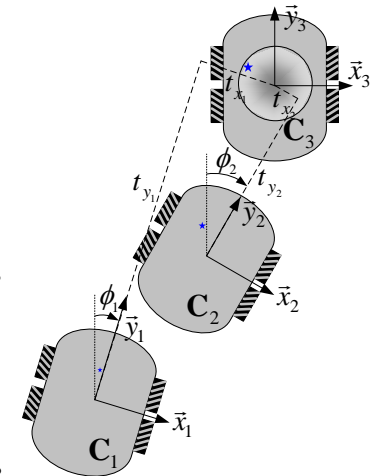
This is an over-constrained measurement

Visual control – TT based

Values of the trifocal tensor in particular locations

◆ When $C_2=C_1$ ($t_{x_2} = t_{x_1}, t_{y_2} = t_{y_1}$) \longrightarrow $T_{111} = 0, T_{112} = 0, T_{121} + T_{211} = 0,$
 $T_{221} = 0, T_{222} = 0, T_{122} + T_{212} = 0.$

◆ When $C_2=C_3$ ($t_{x_2} = 0, t_{y_2} = 0$) \longrightarrow $T_{111} = 0, T_{122} = 0, T_{112} + T_{121} = 0,$
 $T_{211} = 0, T_{222} = 0, T_{212} + T_{221} = 0.$



Time-derivatives of the elements of the tensor

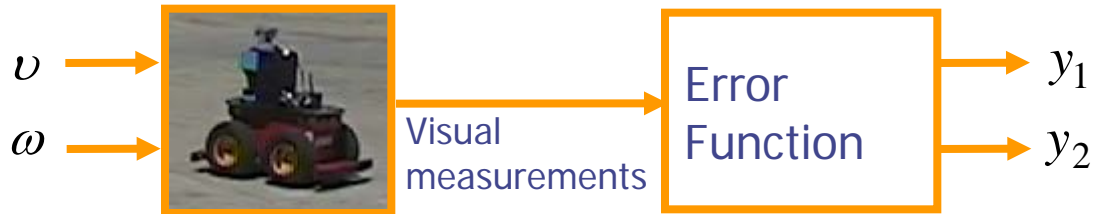
$$\begin{aligned} \dot{T}_{111} &= \frac{\sin \phi_1}{T_N^m} \nu + T_{121} \omega, & \dot{T}_{211} &= \frac{\cos \phi_1}{T_N^m} \nu + T_{221} \omega, \\ \dot{T}_{112} &= -\frac{\cos \phi_1}{T_N^m} \nu + T_{122} \omega, & \dot{T}_{212} &= \frac{\sin \phi_1}{T_N^m} \nu + T_{222} \omega, \end{aligned}$$

$$\begin{aligned} \dot{T}_{121} &= -T_{111} \omega, & \dot{T}_{221} &= -T_{211} \omega, \\ \dot{T}_{122} &= -T_{112} \omega, & \dot{T}_{222} &= -T_{212} \omega. \end{aligned}$$

Useful only for orientation control

Visual control – TT based

- ◆ Three variables to desired values but we choose to make a Square control system.



- ◆ By using two outputs, the tensor provides three possibilities:

	First part of the control	Second part	
	Correcting	DOF	Drawback
1	Orientation and depth (ϕ, y)	Lateral error (x)	Non-holonomic constraint does not allow to correct the remainder lateral error.
2	Orientation and lateral error (ϕ, x)	Depth (y)	Unknown final values of the tensor elements to define the control objective.
3	Lateral error and depth (x, y)	Orientation (ϕ)	Differential-drive allows to correct the remainder orientation error.

Visual control – TT based

Position correction with two selected outputs:

$$\begin{aligned}\xi_1 &= T_{112} + T_{121}, \\ \xi_2 &= T_{212} + T_{221}.\end{aligned}$$

◆ When $\xi_1 \equiv 0$, $\xi_2 \equiv 0$

$$\begin{bmatrix} T_{112} + T_{121} \\ T_{212} + T_{221} \end{bmatrix} = \begin{bmatrix} \sin \phi_1 & \cos \phi_1 \\ \cos \phi_1 & -\sin \phi_1 \end{bmatrix} \begin{bmatrix} t_{x_2} \\ t_{y_2} \end{bmatrix} = \begin{bmatrix} 0 \\ 0 \end{bmatrix}.$$

◆ Zero dynamics:

$$\begin{aligned}Z^* &= \left\{ [x_2 \quad y_2 \quad \phi_2]^T \mid \xi_1 \equiv 0, \xi_2 \equiv 0 \right\} \\ &= \left\{ [0 \quad 0 \quad \phi_2]^T, \phi_2 \in R \right\}.\end{aligned}$$

◆ **Control goal of the step** – Stabilize the following error system, where $e_1 = \xi_1 - \xi_1^d$ and $e_2 = \xi_2 - \xi_2^d$

$$\begin{bmatrix} \dot{e}_1 \\ \dot{e}_2 \end{bmatrix} = \begin{bmatrix} -\frac{\cos \phi_1}{T_N^m} & T_{122} - T_{111} \\ -\frac{\sin \phi_1}{T_N^m} & T_{222} - T_{211} \end{bmatrix} \begin{bmatrix} v \\ \omega \end{bmatrix} - \begin{bmatrix} \dot{\xi}_1^d \\ \dot{\xi}_2^d \end{bmatrix} = \mathbf{M}(\mathbf{T}, \phi_1) \mathbf{u} - \dot{\xi}^d.$$

Desired trajectories

$$\begin{aligned}\xi_1^d &= \frac{T_{112}^{ini} + T_{121}^{ini}}{2} \left(1 + \cos \left(\frac{\pi}{\tau} t \right) \right) \\ \xi_2^d &= \frac{T_{212}^{ini} + T_{221}^{ini}}{2} \left(1 + \cos \left(\frac{\pi}{\tau} t \right) \right).\end{aligned}$$

◆ The initial orientation ϕ_1 introduces uncertainty in this system and a robust control law is required.

Visual control – TT based

Position correction: It is carried out by two controllers, because the first one has a singularity problem when the robot is reaching the target location.

- ◆ **Sliding mode control** with sliding surfaces:

$$\mathbf{s} = \begin{bmatrix} s_1 \\ s_2 \end{bmatrix} = \begin{bmatrix} e_1 \\ e_2 \end{bmatrix} = \begin{bmatrix} \xi_1 - \xi_1^d \\ \xi_1 - \xi_2^d \end{bmatrix} = \mathbf{0}.$$

- ◆ **Decoupling-based controller**

$$\mathbf{u}_{db} = \begin{bmatrix} v_{db} \\ \omega_{db} \end{bmatrix} = \frac{1}{\det(\mathbf{M})} \begin{bmatrix} T_{222} - T_{211} & T_{111} - T_{122} \\ \sin \phi_1 & -\cos \phi_1 \end{bmatrix} \frac{1}{T_N^m} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$

Singularity if
 $|\det(\mathbf{M})| = 0.$

At the final location

where $\det(\mathbf{M}) = \frac{1}{T_N^m} [(T_{122} - T_{111}) \sin \phi_1 + (T_{211} - T_{222}) \cos \phi_1]$, $T_N^m = T_{121}^m$
 $u_1 = \dot{\xi}_1^d - \lambda_1 s_1 - \kappa_1 \text{sign}(s_1)$, $u_2 = \dot{\xi}_2^d - \lambda_2 s_2 - \kappa_2 \text{sign}(s_2).$

- ◆ **Bounded controller**

$$\mathbf{u}_b = \begin{bmatrix} v_b \\ \omega_b \end{bmatrix} = \begin{bmatrix} k_v \text{sign}(s_1) \\ -k_\omega \text{sign}(s_2 (T_{222} - T_{211})) \end{bmatrix}.$$

- Robust global stabilization of the error system is achieved by commuting from the decoupling controller to the bounded one if $|\det(\mathbf{M})| < T_h$.

Visual control – TT based

- ◆ Correction orientation: We can use any single tensor element whose dynamics depends on ω and its final value being zero.

- ◆ **Control goal of the step** – Stabilization of the following dynamics

$$\dot{T}_{122} = -T_{112}\omega.$$

- ◆ A suitable input ω that yields T_{122} **exponentially stable** is

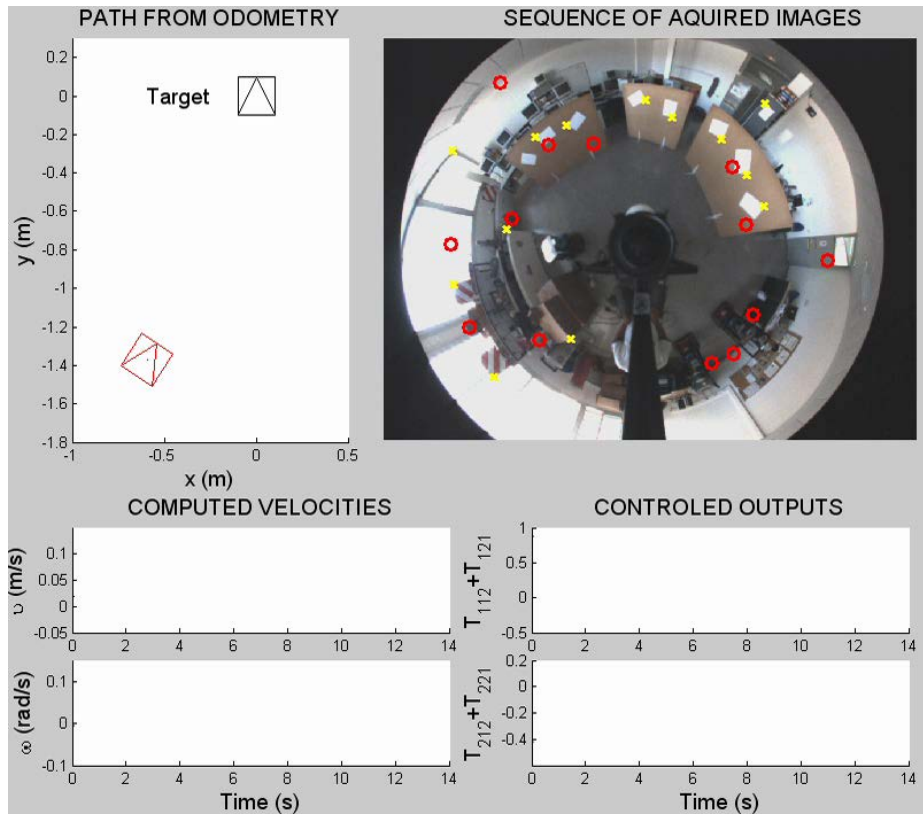
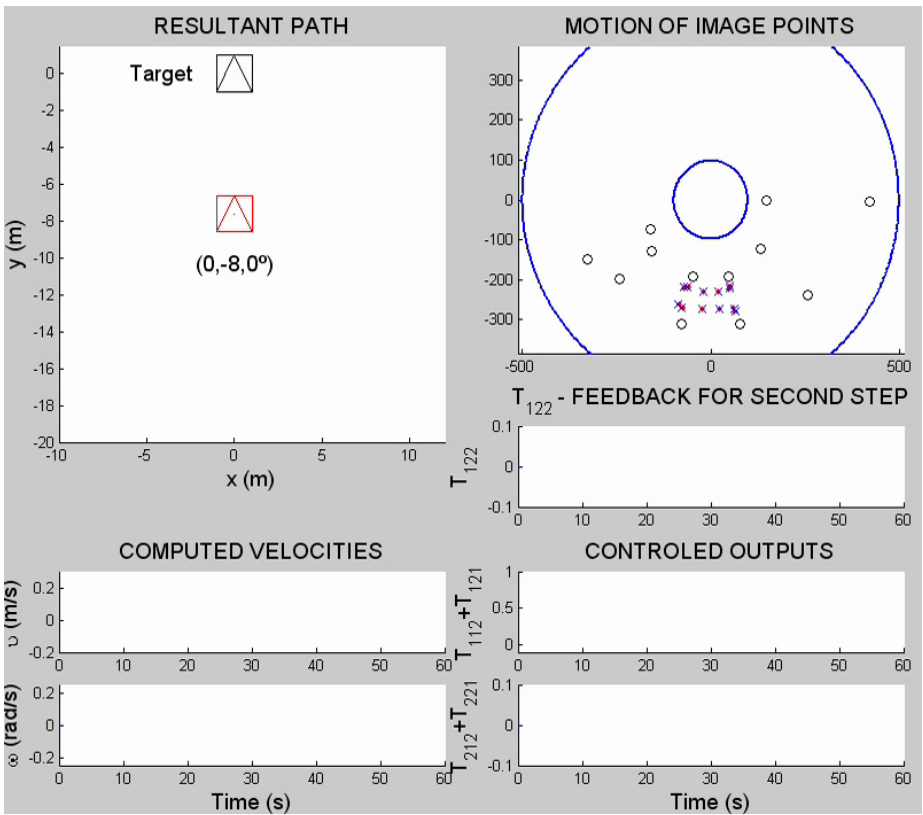
$$\omega = k_{\omega} \frac{T_{122}}{T_{112}}, \quad t > \tau$$

- ◆ When position correction has been reached $T_{122} = t_{y_1} \cos \phi_2$, and consequently, if $T_{122} = 0$ then $\phi_2 = n\pi$ with $n \in \mathbf{Z}$, and the orientation is corrected.

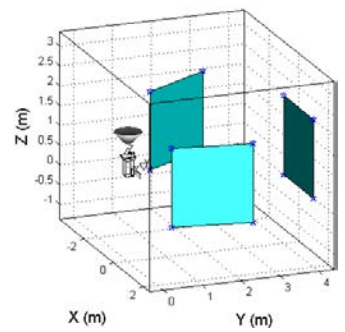
- ◆ Although only a rotation is needed, the same bounded translational velocity is used to maintain the longitudinal position under closed loop control.

$$v = k_v \text{sign}(s_1).$$

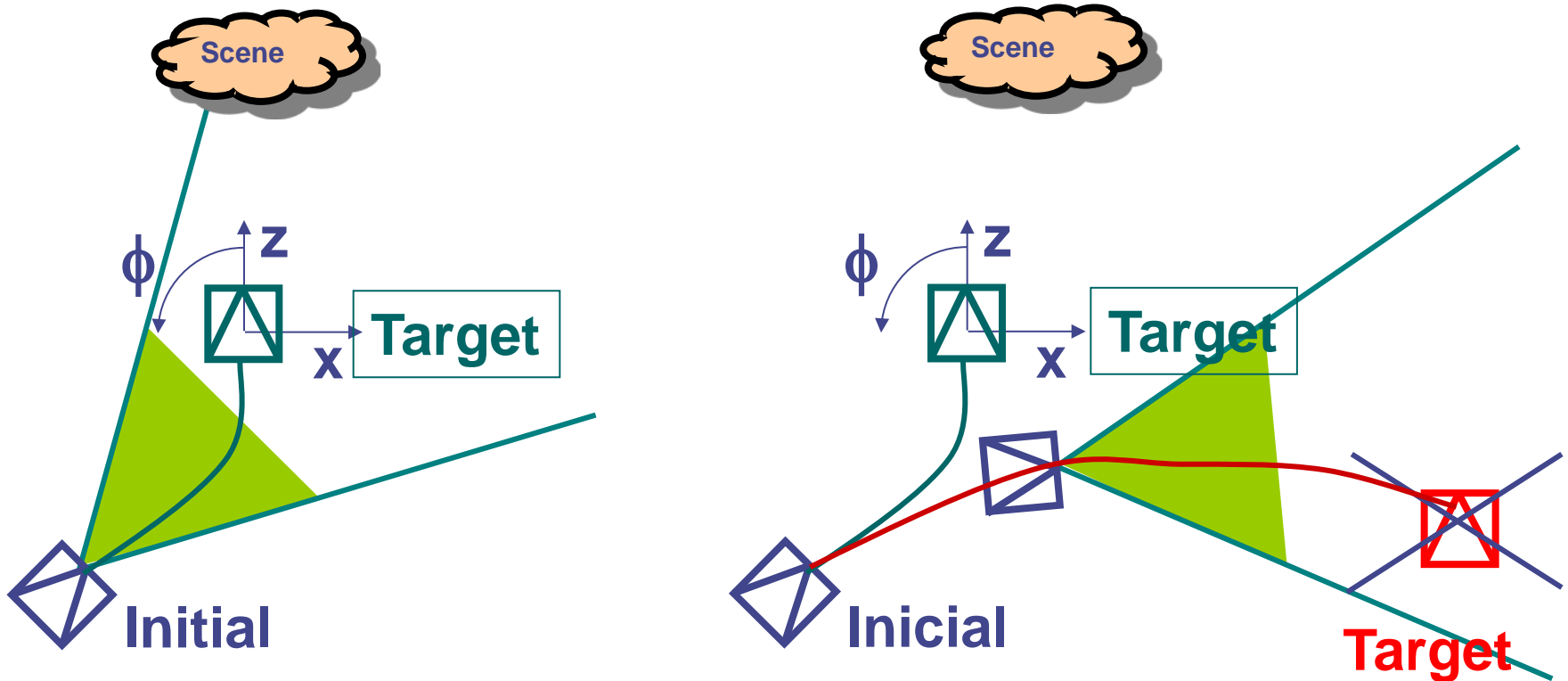
Visual control – TT based



- The 1D-TT is computed from synthetic images of size 1024x768 pixels.
- The desired pose is (0,0,0°).
- Virtual scene:

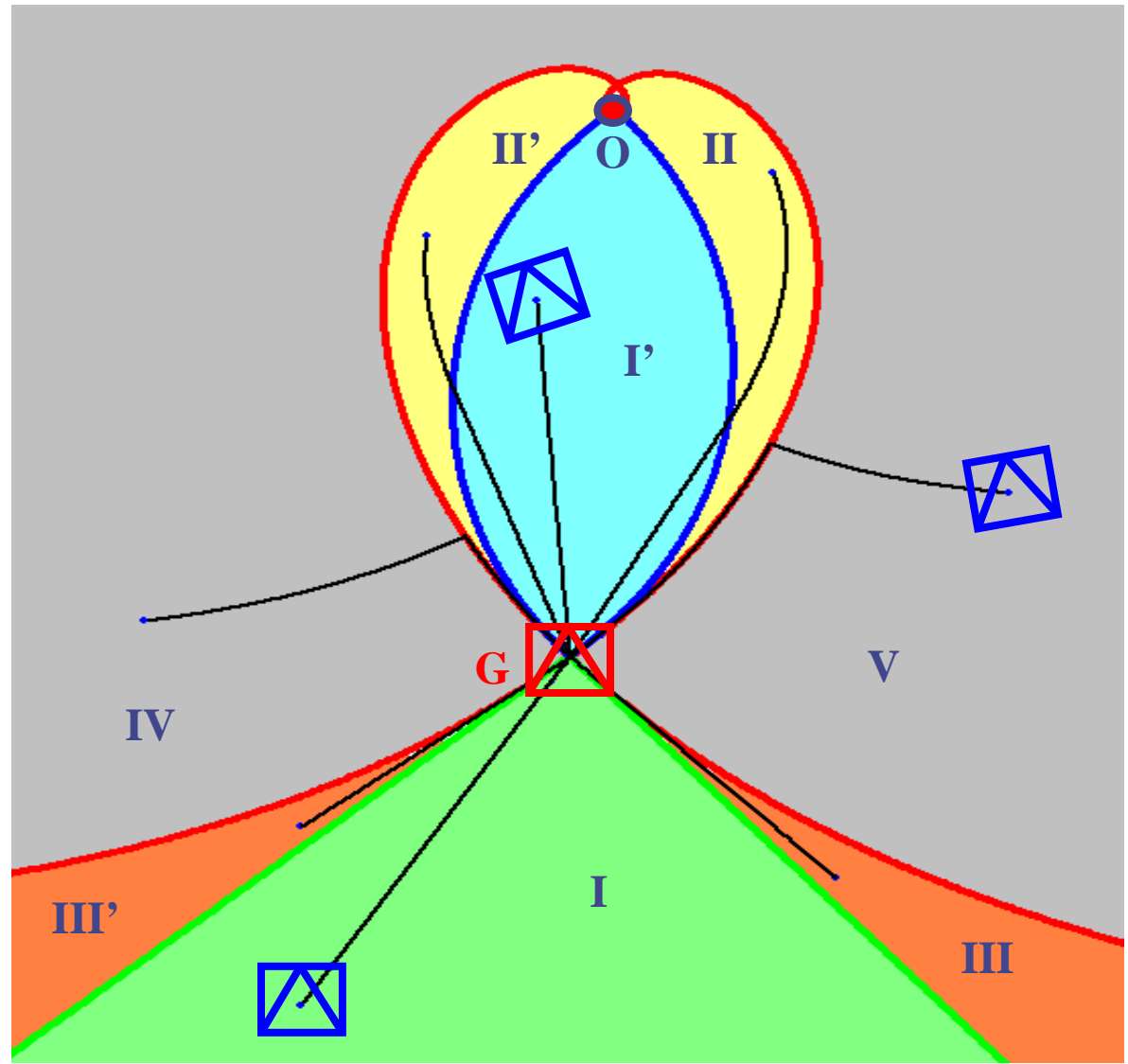


Visual control with FoV constraints



Visual control with FoV constraints

- Observed target
- Initial positions
- Goal



Visual control with FoV constraints

- ◆ The homography between two views is related to camera motion:

$$\mathbf{H} = \mathbf{K} \left(\mathbf{R} - \mathbf{t} \frac{\mathbf{n}^T}{d} \right) \mathbf{K}^{-1}$$

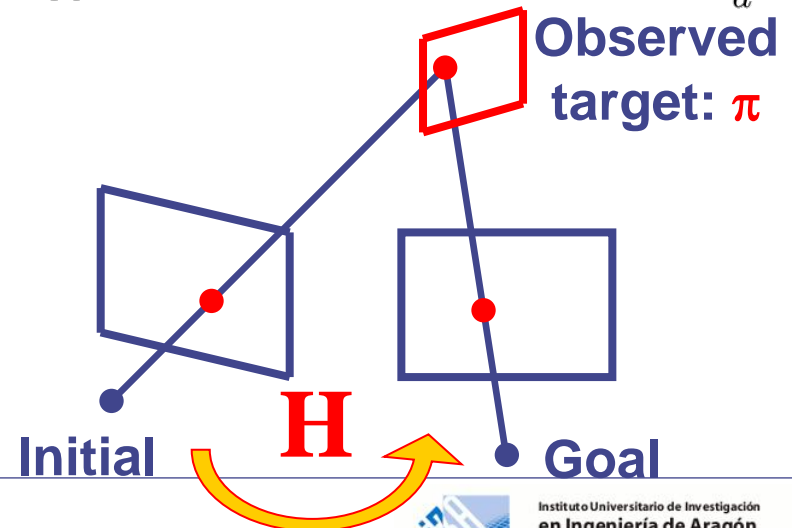
- ◆ Planar motion:

$$\mathbf{H} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ 0 & 1 & 0 \\ h_{31} & h_{32} & h_{33} \end{bmatrix}$$

With:

$$\begin{cases} h_{11} = \cos \phi + (x \cos \phi + z \sin \phi) \frac{n_x}{d} \\ h_{12} = \frac{\alpha_x}{\alpha_y} (x \cos \phi + z \sin \phi) \frac{n_y}{d} \\ h_{13} = \alpha_x \left(\sin \phi + (x \cos \phi + z \sin \phi) \frac{n_z}{d} \right) \\ h_{31} = \frac{1}{\alpha_x} \left(-\sin \phi + (-x \sin \phi + z \cos \phi) \frac{n_x}{d} \right) \\ h_{32} = \frac{1}{\alpha_y} (-x \sin \phi + z \cos \phi) \frac{n_y}{d} \\ h_{33} = \cos \phi + (-x \sin \phi + z \cos \phi) \frac{n_z}{d} \end{cases}$$

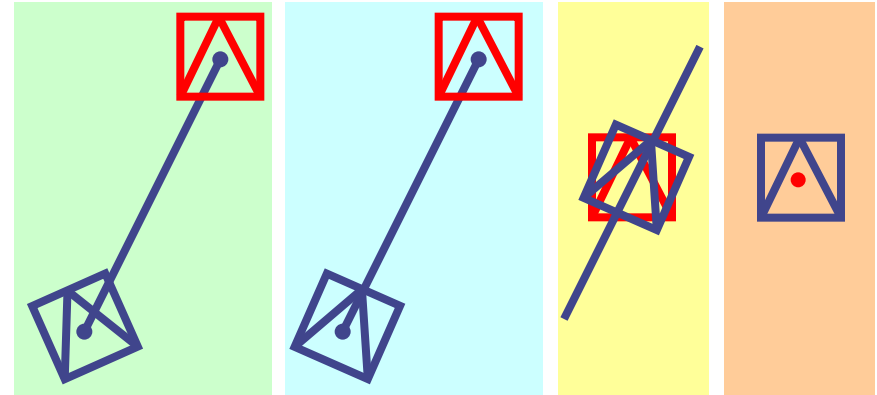
- ◆ Target: Plane of the scene
- ◆ Goal: $\mathbf{H} = \mathbf{I}$
- ◆ Subgoals: $\mathbf{H} = \dots$



Visual control with FoV constraints

Particular homographies in particular positions

$$\mathbf{H}_{(x,z,\phi)} = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ 0 & 1 & 0 \\ h_{31} & h_{32} & h_{33} \end{bmatrix}$$



$$\mathbf{H}_{(x,z,\phi_t)} = \begin{bmatrix} \cos \phi_t & 0 & \alpha_x \sin \phi_t \\ 0 & 1 & 0 \\ -\frac{\sin \phi_t}{\alpha_x} + \frac{zn_x/d}{\alpha_x \cos \phi_t} & \frac{zn_y/d}{\alpha_y \cos \phi_t} & \frac{\cos^2 \phi_t + zn_z/d}{\cos \phi_t} \end{bmatrix}$$

$$\mathbf{H}_{(0,0,\phi_t)} = \begin{bmatrix} \cos \phi_t & 0 & \alpha_x \sin \phi_t \\ 0 & 1 & 0 \\ -\frac{\sin \phi_t}{\alpha_x} & 0 & \cos \phi_t \end{bmatrix}$$

$$\mathbf{H}_{(0,0,0)} = \mathbf{I}$$

Visual control with FoV constraints

◆ **Switched control:**
Three sequential steps

Step 1:

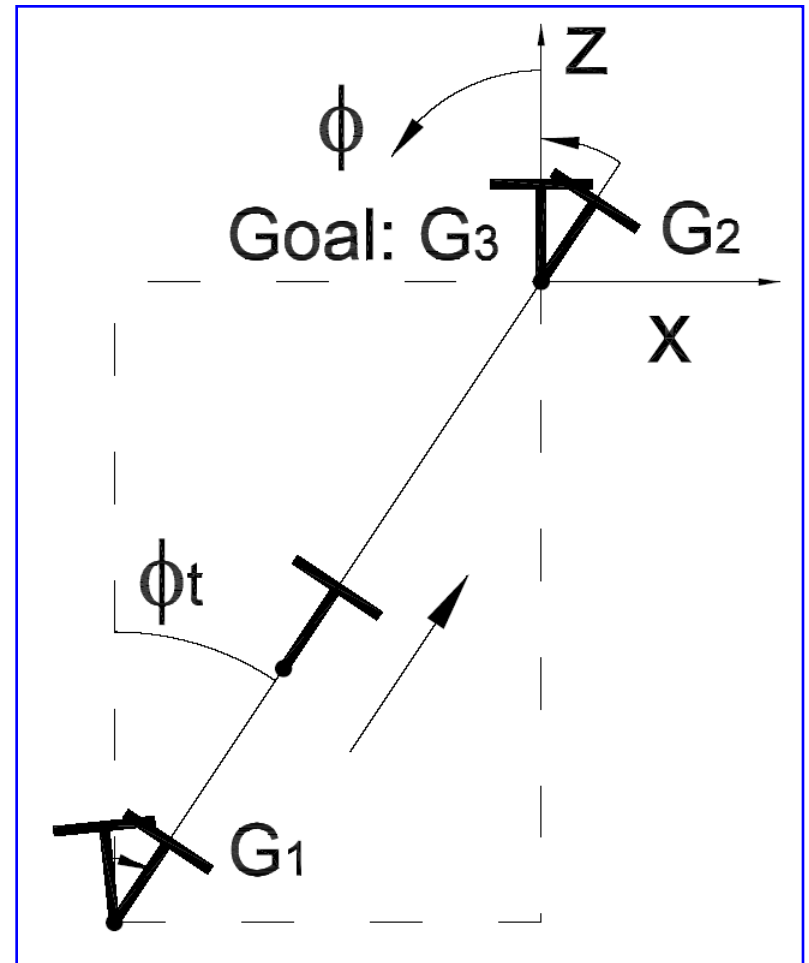
$$\begin{pmatrix} v_1 \\ \omega_1 \end{pmatrix} = \begin{pmatrix} 0 \\ -k_\omega (h_{11}^2 + h_{13}^2 / \alpha_x^2 - 1) \end{pmatrix}$$

Step 2:

$$\begin{pmatrix} v_2 \\ \omega_2 \end{pmatrix} = \begin{pmatrix} -k_v (h_{11} - h_{33}) \\ -k_\omega (h_{11}^2 + h_{13}^2 / \alpha_x^2 - 1) \end{pmatrix}$$

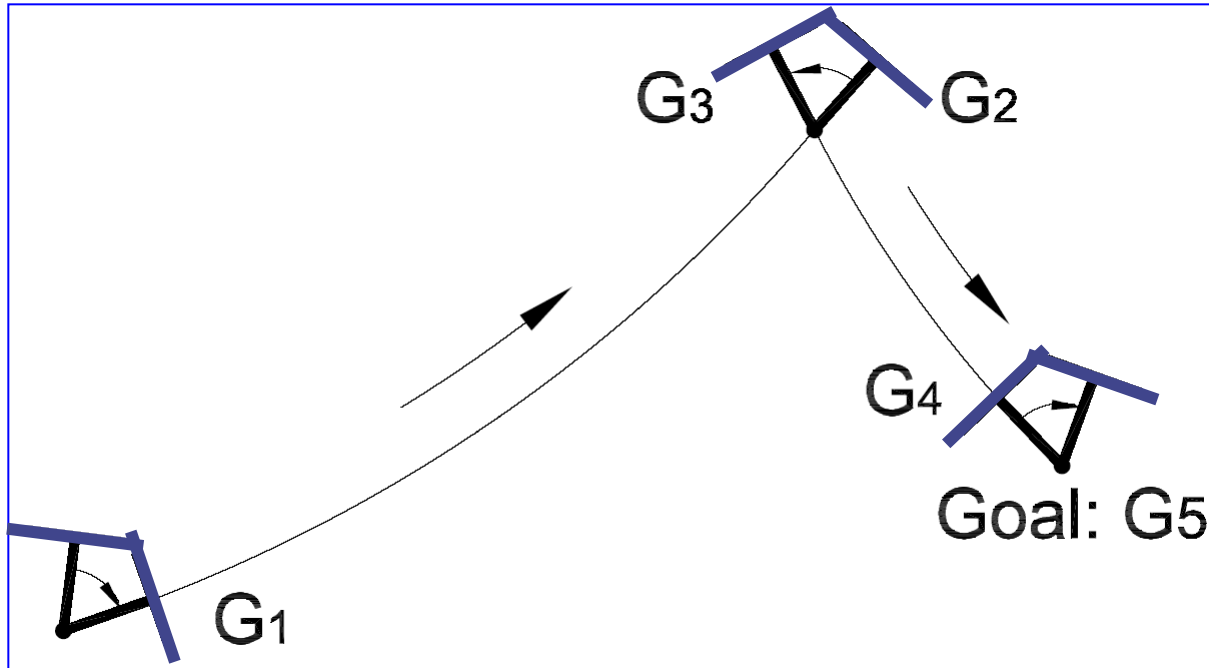
Step 3:

$$\begin{pmatrix} v_3 \\ \omega_3 \end{pmatrix} = \begin{pmatrix} 0 \\ -k_\omega h_{13} \end{pmatrix}$$



Visual control with FoV constraints

◆ Switched control: Five sequential steps



•Subgoals

- G_1 : Pure rotation until reaching the first T-curve
- G_2 : Follow the first T-curve forward
- G_3 : Pure rotation until reaching the second T-curve
- G_4 : Follow the second T-curve backward
- G_5 : Pure rotation until reaching desired Goal

Visual control with FoV constraints

◆ Switched control: Five sequential steps

$$\text{Step 1: } \begin{pmatrix} v_1 \\ \omega_1 \end{pmatrix} = \begin{pmatrix} 0 \\ -k_\omega(h_{13} - h_{13}^{G_1}) \end{pmatrix} \quad \text{Step 4: } \begin{pmatrix} v_4 \\ \omega_4 \end{pmatrix} = \begin{pmatrix} -k_v(h_{33} - h_{11}) \\ -k_\omega(h_{13} - h_{13}^{G_4}) \end{pmatrix}$$

$$\text{Step 2: } \begin{pmatrix} v_2 \\ \omega_2 \end{pmatrix} = \begin{pmatrix} -k_v(h_{33} - h_{33}^{G_2}) \\ -k_\omega(h_{13} - h_{13}^{G_2}) \end{pmatrix} \quad \text{Step 5: } \begin{pmatrix} v_5 \\ \omega_5 \end{pmatrix} = \begin{pmatrix} 0 \\ -k_\omega h_{13} \end{pmatrix}$$

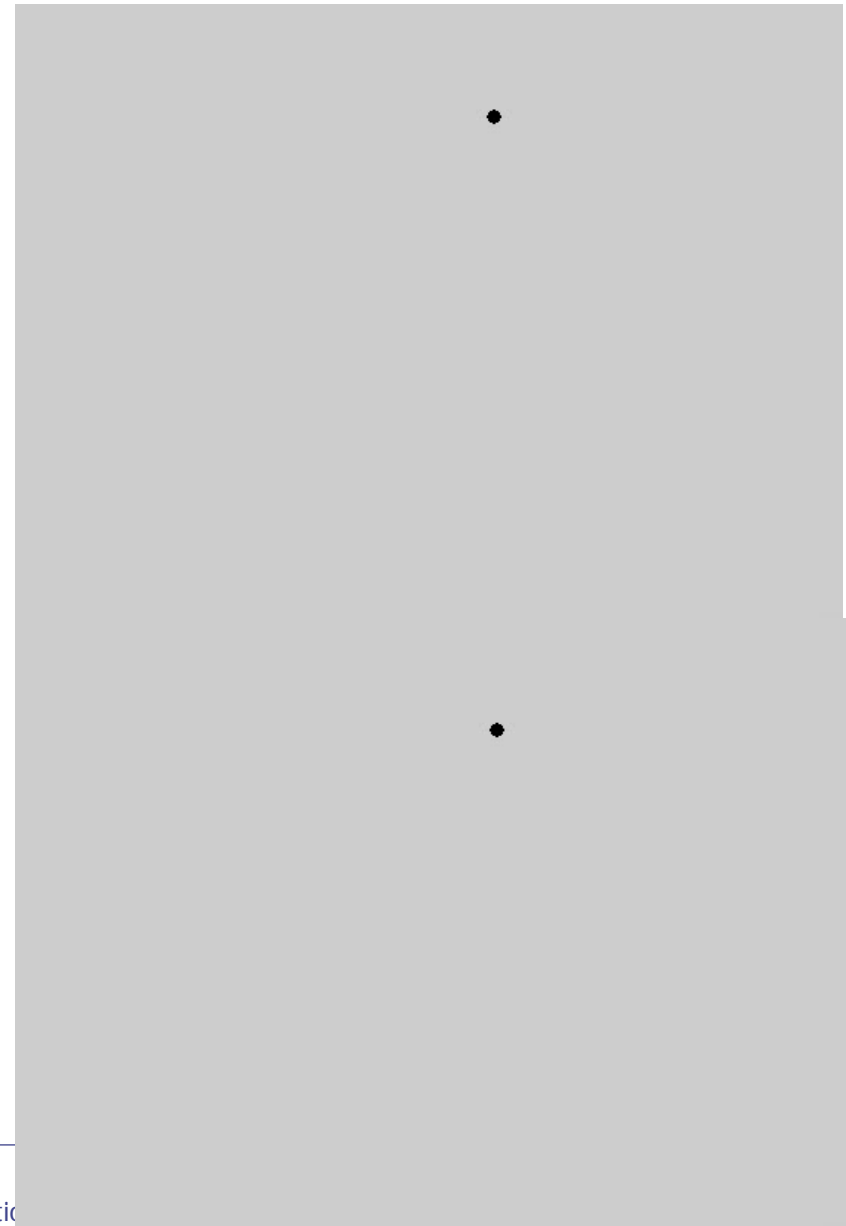
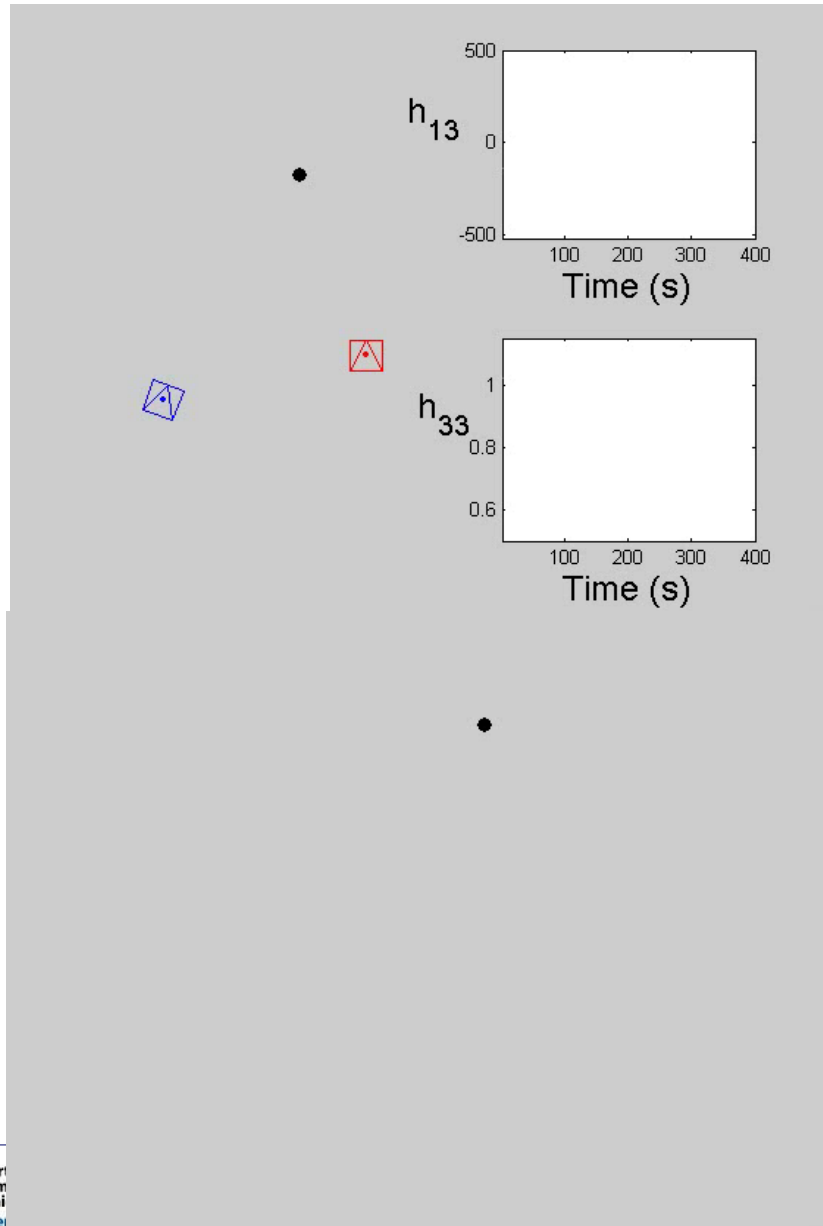
$$\text{Step 3: } \begin{pmatrix} v_3 \\ \omega_3 \end{pmatrix} = \begin{pmatrix} 0 \\ -k_\omega(h_{13} - h_{13}^{G_3}) \end{pmatrix}$$

◆ Subgoals:

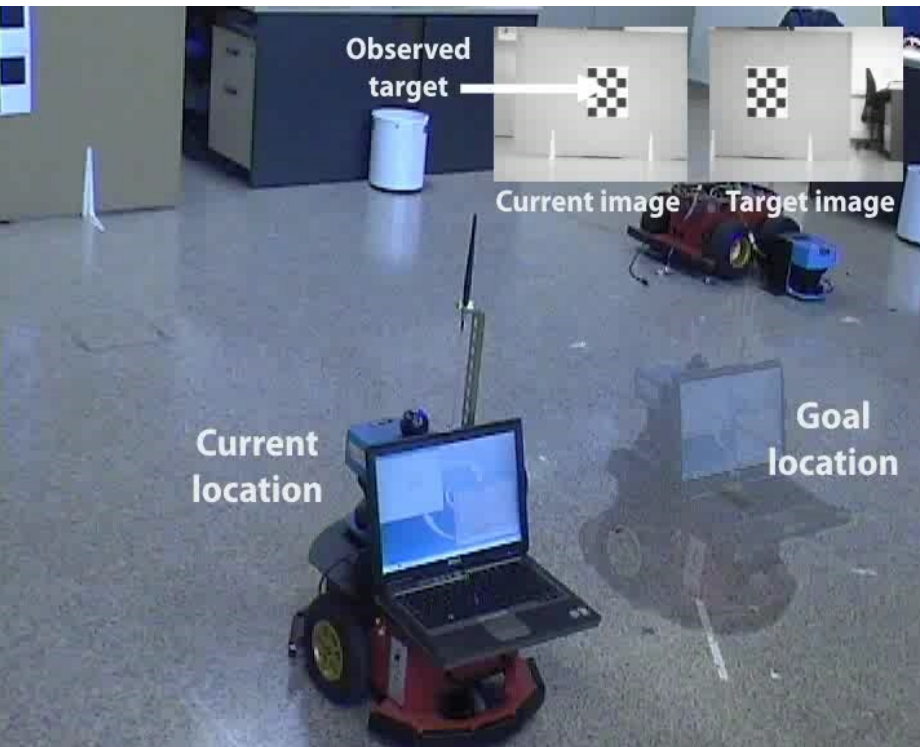
- Defined in terms of homography parameters
- Decomposition of the homography

$$G_i \quad (i = 1..5) \quad \begin{cases} h_{13}^{G_i} = \frac{(\frac{h_{13}}{\alpha_x} - \sin \phi)(\rho^{G_i} \cos \phi^{G_i} + \sin \phi^{G_i})}{(\rho \cos \phi + \sin \phi)\rho_z / \alpha_x} + \alpha_x \sin \phi^{G_i} \\ h_{33}^{G_i} = \frac{(h_{33} - \cos \phi)(-\rho^{G_i} \sin \phi^{G_i} + \cos \phi^{G_i})}{(-\rho \sin \phi + \cos \phi)\rho_z} + \cos \phi^{G_i} \end{cases}$$

Visual control with FoV constraints



Visual control with FoV constraints

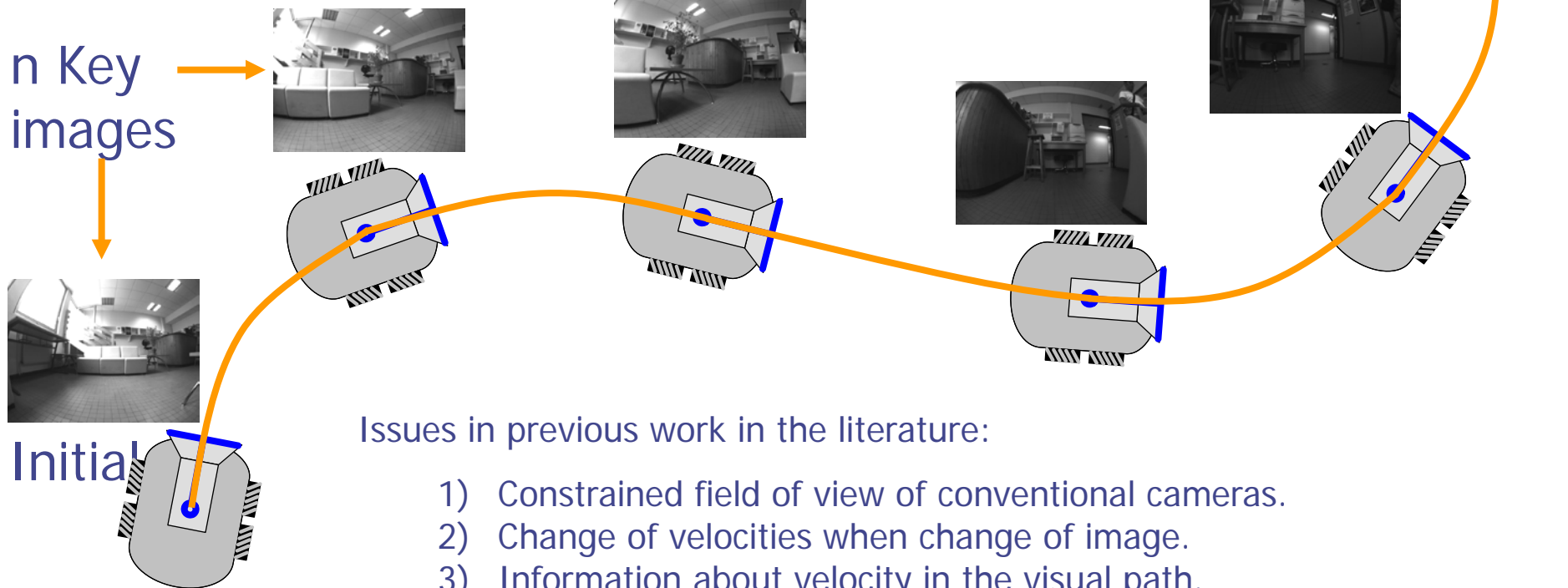


An Optimal Homography-Based Control Scheme for Mobile Robots with Nonholonomic and Field-of-View Constraints

G. López-Nicolás, N. Gans, S. Bhattacharya,
C. Sagüés, J.J. Guerrero and S. Hutchinson

Long term navigation

- Task: reach a desired position associated with a target image, which belongs to a visual memory acquired in a teaching phase.
- A visual path of n key images is extracted from the visual memory, which must be followed autonomously in order to reach the target.

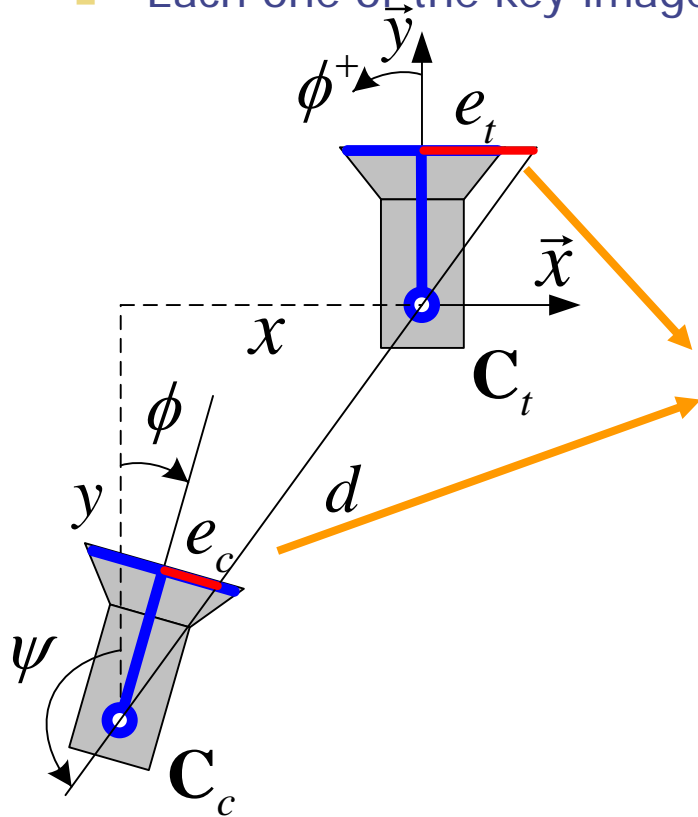


Issues in previous work in the literature:

- 1) Constrained field of view of conventional cameras.
- 2) Change of velocities when change of image.
- 3) Information about velocity in the visual path.

Long term navigation

- The omnidirectional cameras can be virtually represented as conventional cameras when working with points on the sphere.
- Each one of the key images is used as target image accordingly.



Target location

$$\Rightarrow \mathbf{C}_t = (0,0,0)$$

Current location

$$\Rightarrow \mathbf{C}_c = (x, y, \phi)$$

Epipoles

$$\left\{ \begin{array}{l} e_c = \alpha_x \frac{x \cos \phi + y \sin \phi}{y \cos \phi - x \sin \phi}, \\ e_t = \alpha_x \frac{x}{y}. \end{array} \right.$$

- Interaction with the robot velocities:

$$\dot{e}_c = -\frac{\alpha_x \sin(\phi - \psi)}{d \cos^2(\phi - \psi)} v + \frac{\alpha_x}{\cos^2(\phi - \psi)} \omega,$$

$$\dot{e}_t = -\frac{\alpha_x \sin(\phi - \psi)}{d \cos^2(\psi)} v$$

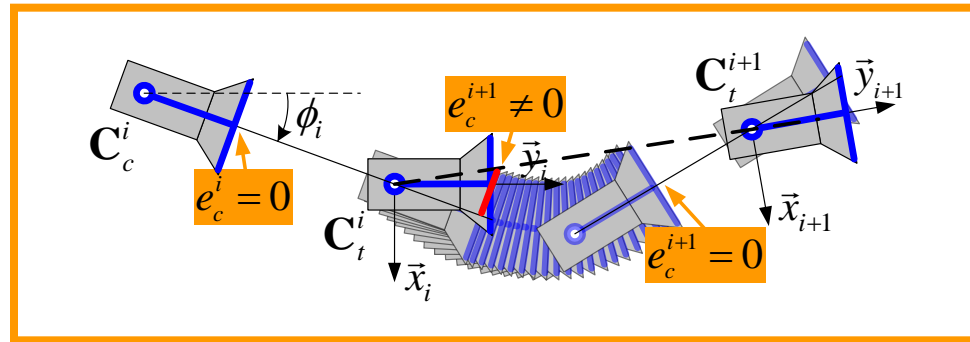
Long term navigation

- The current epipole gives information of the translation direction and it is directly related to the required robot rotation to be aligned with the target.
- Use of the x -coordinate of the current epipole as feedback information to control the robot heading and so, to correct the lateral deviation.

Non-null translational velocity $v \neq 0$

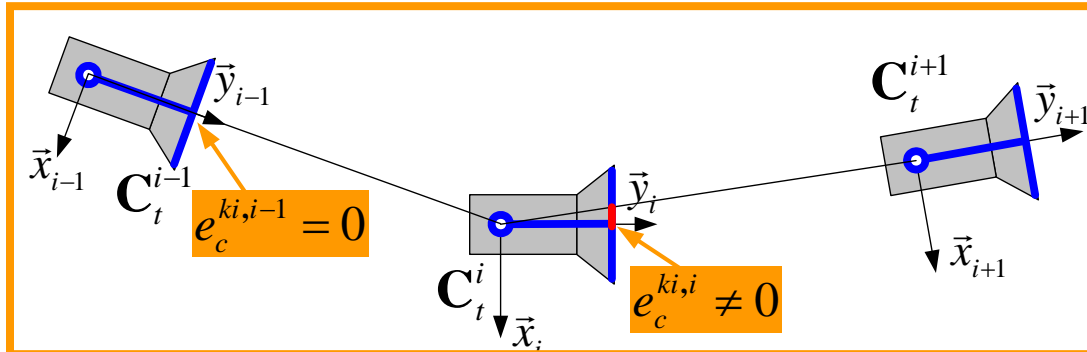
$$\omega^{ce} = k_t \omega_{rt}^{ce} + \bar{\omega}^{ce}.$$

First component of the rotational velocity



$$\omega \Rightarrow f(e_c)$$

Second component of the rotational velocity



$$\omega \Rightarrow f(e_c^{ki})$$

Long term navigation

- Let us define a tracking error to drive the epipole smoothly to zero for every segment between key images

$$\zeta_{ce} = e_c - e_c^d(t) = 0.$$

where $e_c^d(t) = \frac{e_c(0)}{2} \left(1 + \cos\left(\frac{\pi}{\tau} t\right) \right), 0 \leq t \leq \tau$ with $\tau = \frac{d_{\min}}{v}$.

$$e_c^d(t) = 0, \quad t > \tau$$

- Control goal – Stabilization of the error system:

$$\dot{\zeta}_{ce} = -\frac{\alpha_x \sin(\phi - \psi)}{d \cos^2(\phi - \psi)} v + \frac{\alpha_x}{\cos^2(\phi - \psi)} \omega_{rt}^{ce} - \dot{e}_c^d.$$

- Considering that the translational velocity is known, the following rotational velocity, referred as **reference tracking (RT) control**, stabilizes the error system

$$\omega_{rt}^{ce} = \frac{\sin(\phi - \psi)}{d} v + \frac{\cos^2(\phi - \psi)}{\alpha_x} (\dot{e}_c^d - k_c \zeta_{ce}).$$

with $k_c > 0$.

Long term navigation

- A varying translational velocity according to the shape of the path can be computed depending on the epipoles between key images.

$$v^{ce} = v_{\max} + v_{\min} + \frac{v_{\max} - v_{\min}}{2} \tanh\left(1 - \frac{|e_c^{ki} / d_{\min}|}{\sigma}\right).$$

- We propose the following nominal rotational velocity, which is computed from the epipoles between key images:

$$\bar{\omega}^{ce} = \frac{k_m v^{ce}}{d_{\min}} e_c^{ki}.$$

- So that, the complete rotational velocity (RT+ control) is given as:

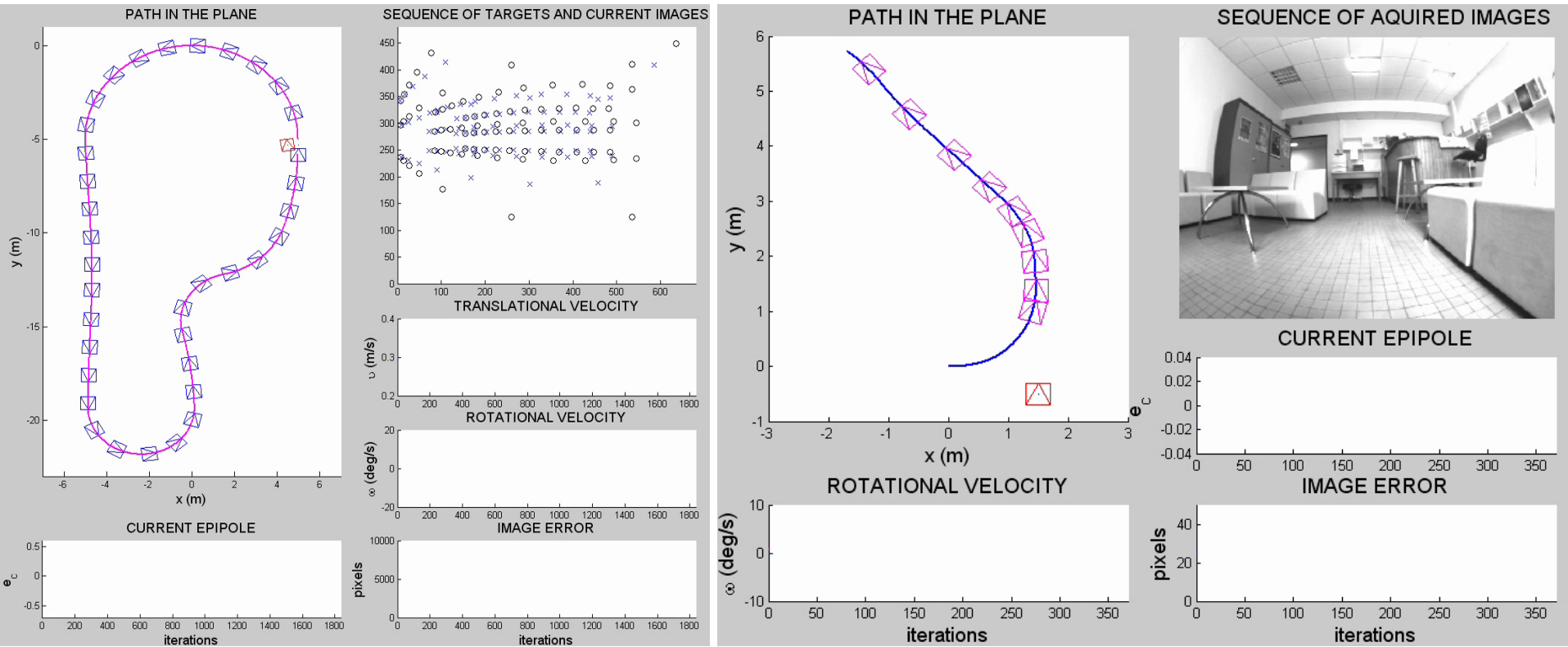
$$\omega^{ce} = k_t \omega_{rt}^{ce} + \bar{\omega}^{ce}.$$

Switching and stop condition

- The switching condition to the next key image or to stop the task is given when the image error starts to increase, which is defined as follows:

$$\varepsilon = \frac{1}{r} \sum_{j=1}^r \|\mathbf{p}_j - \mathbf{p}_{i,j}\|.$$

Long term navigation



Index

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 - Data association
 - Coordinated motion with epipoles
 - Central decision with flying camera on scene - Homography

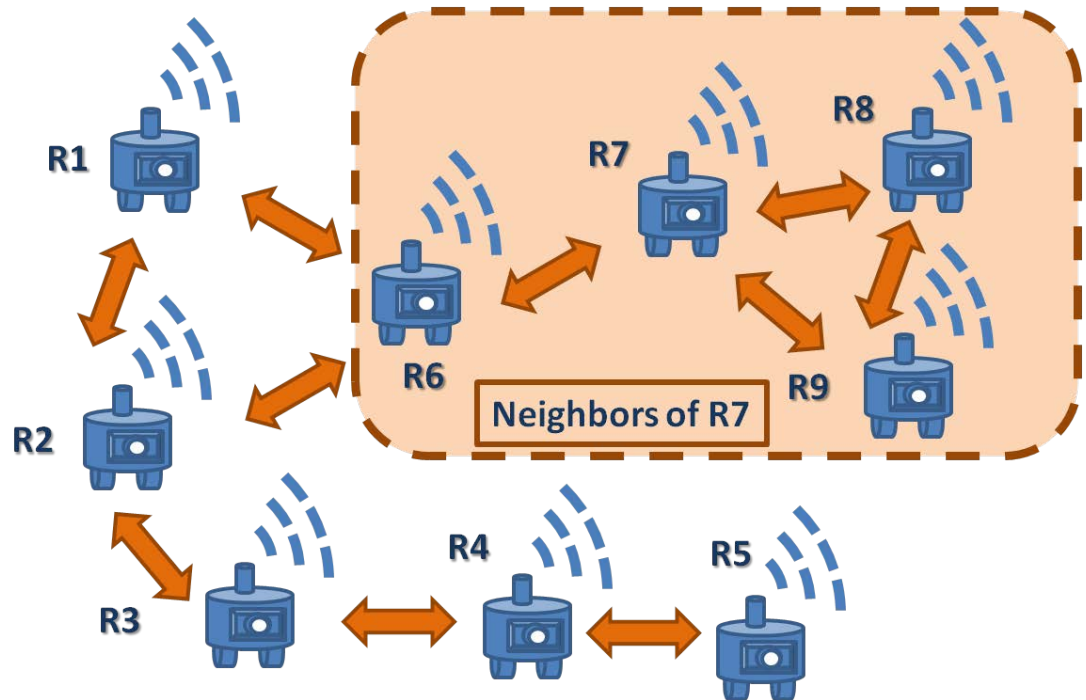
Multi-Robot Systems

- ❑ Robots communication is limited

- Wireless network.
- Range-limited.
- Visibility (Comm.)

- ❑ Communication graphs

- Nodes: the robots
- Edges: link between robots that can exchange data



- ❑ Each robot exchange data with its one-hop neighbors

- ❑ Robots are moving: new edges may appear / previous links disappear

- Communication graphs with switching topology

Distributed Data Association

- ◆ Each robot $i \in \{1, \dots, n\}$ in the team has a set $\mathcal{S}_i = \{f_1^i, \dots, f_{m_i}^i\}$ of m_i features .
- ◆ It has executed a local association method F to match its features \mathcal{S}_i and its neighbors' ones \mathcal{S}_j , for $j \in \mathcal{N}_i$

$$F(\mathcal{S}_i, \mathcal{S}_j) = \mathbf{A}_{ij} = \mathbf{A}_{ji}^T = (F(\mathcal{S}_j, \mathcal{S}_i))^T \quad F(\mathcal{S}_i, \mathcal{S}_i) = \mathbf{A}_{ii} = \mathbf{I}$$

$$[\mathbf{A}_{ij}]_{r,s} = \begin{cases} 1 & \text{if } f_r^i \text{ and } f_s^j \text{ are associated,} \\ 0 & \text{otherwise,} \end{cases}$$

$r = 1, \dots, m_i$ and $s = 1, \dots, m_j$.

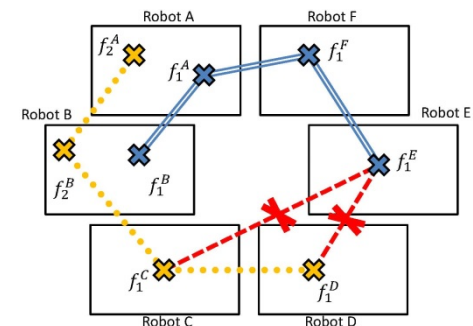
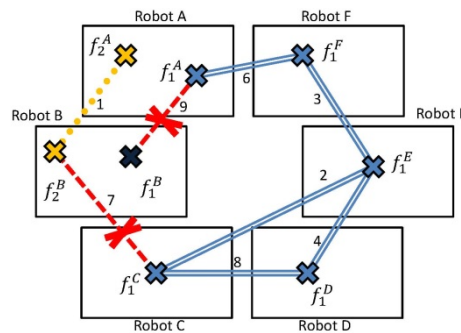
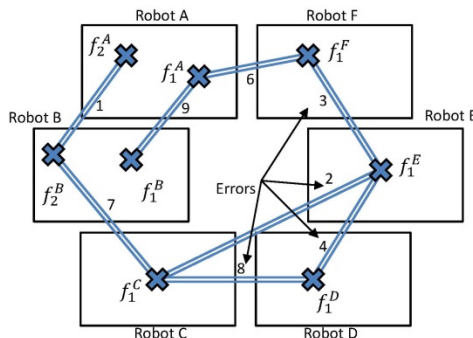
- ◆ This information can be represented with graph, where the nodes are the features of all the robots, and there is a link between two features if they have been locally matched by F .
 - ◆ The adjacency matrix of this graph is with
- $$\mathbf{A}_{ij} = \begin{cases} F(\mathcal{S}_i, \mathcal{S}_j) & \text{if } j \in \{\mathcal{N}_i \cup i\}, \\ 0 & \text{otherwise.} \end{cases} \quad \mathbf{A} = \begin{bmatrix} \mathbf{A}_{11} & \dots & \mathbf{A}_{1n} \\ \vdots & \ddots & \vdots \\ \mathbf{A}_{n1} & \dots & \mathbf{A}_{nn} \end{bmatrix}$$

Distributed Data Association

- ◆ **Goal** (robot i). Discover for each the features f_r^i , all the other features which are connected to f_r^i through a path.
- ◆ **Idea**. If there is a link between features f_r^i and f_s^j , then the features connected to f_r^i and to f_s^j through a path are the same.
- ◆ **Formal**. Distributed computation of the powers of the adjacency matrix,
 - Each robot maintain \mathbf{A}^t the rows of the adjacency matrix power associated to its own features, and updates them using data from its neighbors
 - For each of this features f_r^i , each robot i obtains all f_s^j connected to f_r^i through a path, and detects the **inconsistent** ones.

Distributed Data Association

- ◆ **Idea:** break local associations so that there are no two features from the same robot related by a path.
 - Note that each inconsistency is motivated by, at least, one spurious local link (false positives).
- ◆ All local links are equal \implies Resol. algorithm based on Trees
 - For each conflictive feature belonging to the same robot, use it as root of its tree and incrementally add features linked to it.
 - If a feature already belongs to a tree, or receives requests from more than a tree, it selects one of the trees and erases links to the others.
- ◆ Links with quality information \implies Maximum Error Cut
 - For each pair of inconsistent features belonging to same robot, select and erase the link with the largest error that breaks the inconsistency.

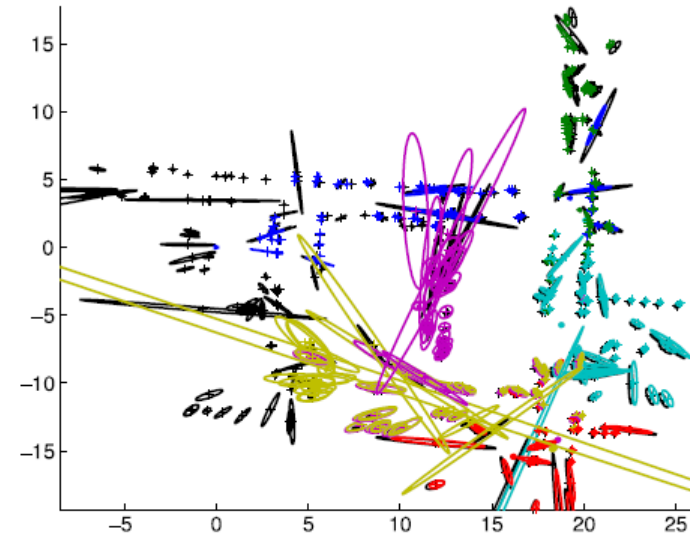


Distributed Data Association

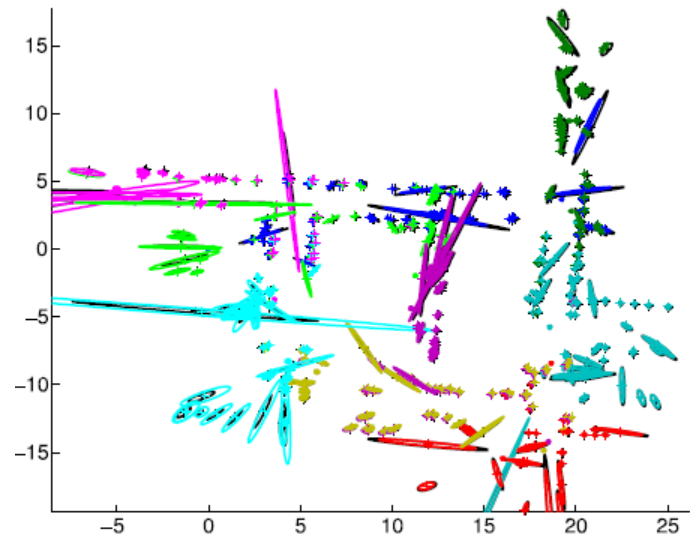


- ◆ Compute the robot positions in a common reference frame
- ◆ Each robot measures the relative position of its neighbors
- ◆ Distributed map merging scenario
 - Local maps aligned before merging
 - It only needs to be computed once

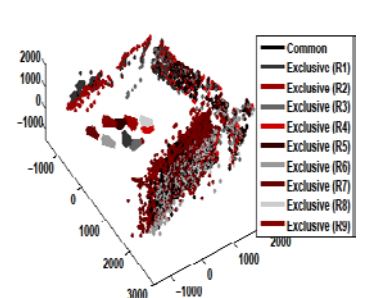
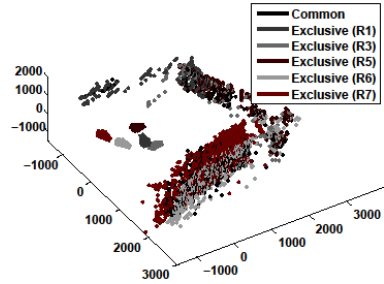
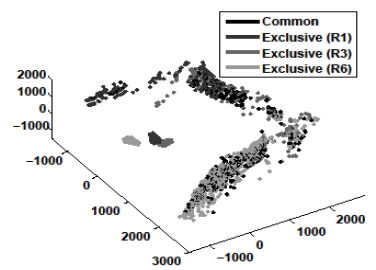
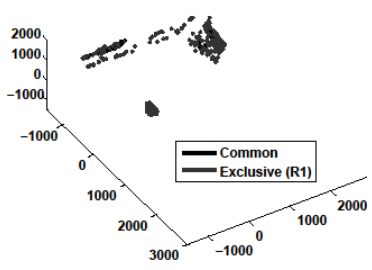
(a) Merged map after 5 iterations



(b) Merged map after 20 iterations

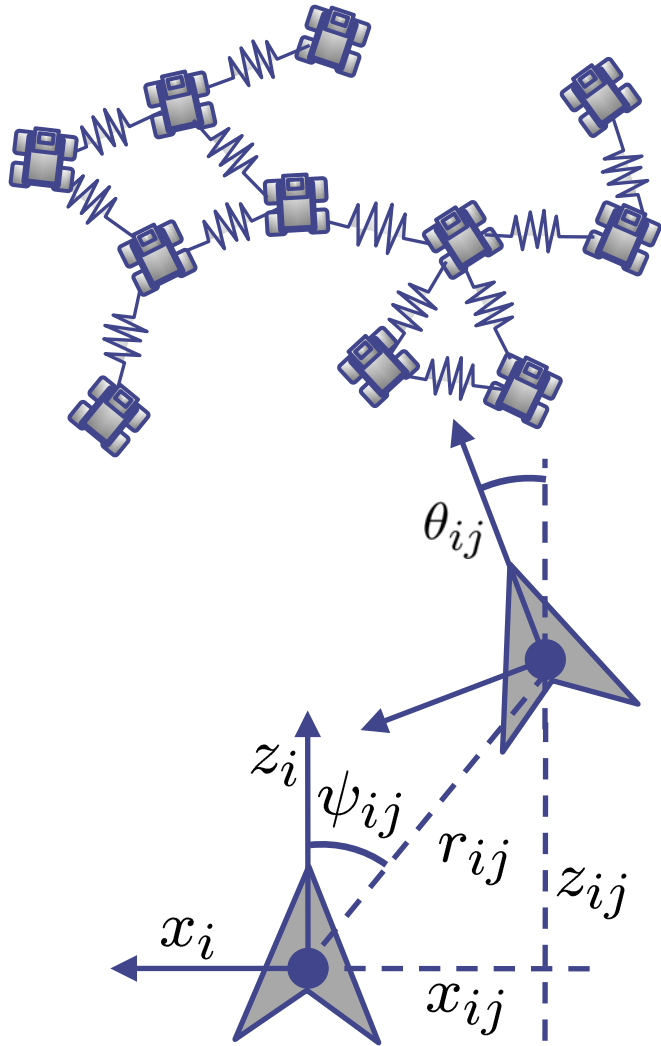


Distributed Data Association



Multi robot control based on epipoles

◆ Coordinated control for attitude sincronization



Modeled with an undirected graph

$$\mathcal{G} = (\mathcal{V}, \mathcal{E})$$

$$\mathcal{N}_i = \{j \in \mathcal{V} \mid (i, j) \in \mathcal{E}\}$$

Non holonomic motion on the plane

$$\begin{bmatrix} \dot{x}_i \\ \dot{z}_i \\ \dot{\theta}_i \end{bmatrix} = \begin{bmatrix} \sin(\theta_i) & 0 \\ \cos(\theta_i) & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} v_i \\ w_i \end{bmatrix}$$

Polar coordinates

$$r_{ij} = \sqrt{x_{ij}^2 + z_{ij}^2} \in \mathbb{R}_{\geq 0},$$

$$\psi_{ij} = \arctan(x_{ij}/z_{ij}) \in (-\pi/2, \pi/2],$$

$$\theta_{ij} = \theta_j - \theta_i \in (-\pi, \pi],$$

Multi robot control based on epipoles

The robots exchange the visual features

Correspondences satisfy the epipolar constraint

$$\mathbf{p}_i^T \mathbf{F}_{ij} \mathbf{p}_j = 0$$

The epipoles are the null space of \mathbf{F}_{ij} and \mathbf{F}_{ij}^T

$$e_{ij} = \alpha \tan(\psi_{ij})$$

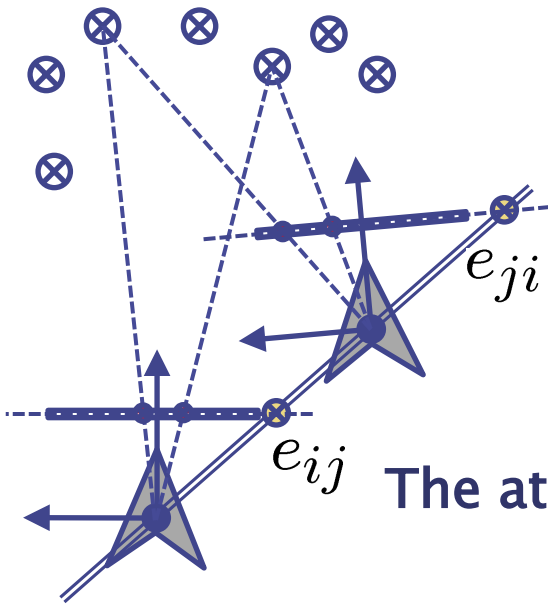
$$e_{ji} = \alpha \tan(\psi_{ij} - \theta_{ij})$$

The attitude consensus implies the epipoles to be equal

$$\theta_{ij} = 0 \Rightarrow e_{ij} = e_{ji}$$

Note that the opposite is not necessarily true

$$\theta_{ij} = \pi \Rightarrow e_{ij} = e_{ji}$$



Multi robot control based on epipoles

Define

$$d_{ij} = \arctan\left(\frac{e_{ij}}{\beta}\right) - \arctan\left(\frac{e_{ji}}{\beta}\right) \in (-\pi, \pi], \beta > 0$$

The “geodesic” in the epipole domain

$$w_{ij} = \begin{cases} d_{ij} & \text{if } |d_{ij}| \leq \frac{\pi}{2} \\ -\text{sign}(d_{ij})(\pi - |d_{ij}|) & \text{otherwise} \end{cases},$$

If the calibration is known, then choosing $\beta = \alpha$
the exact relative orientation can be computed and we
have a standard consensus problem

Multi robot control based on epipoles

The distributed controller used by the robots is

$$w_i = K \sum_{j \in \mathcal{N}_i} w_{ij}, \quad K > 0$$

Properties of the controller

$$w_{ij} = -w_{ji}$$

$$\sum_{i \in \mathcal{V}} w_i = 0$$

$$\text{sign}(e_{ij}) = \text{sign}(e_{ji}) \Rightarrow |d_{ij}| < \pi/2$$

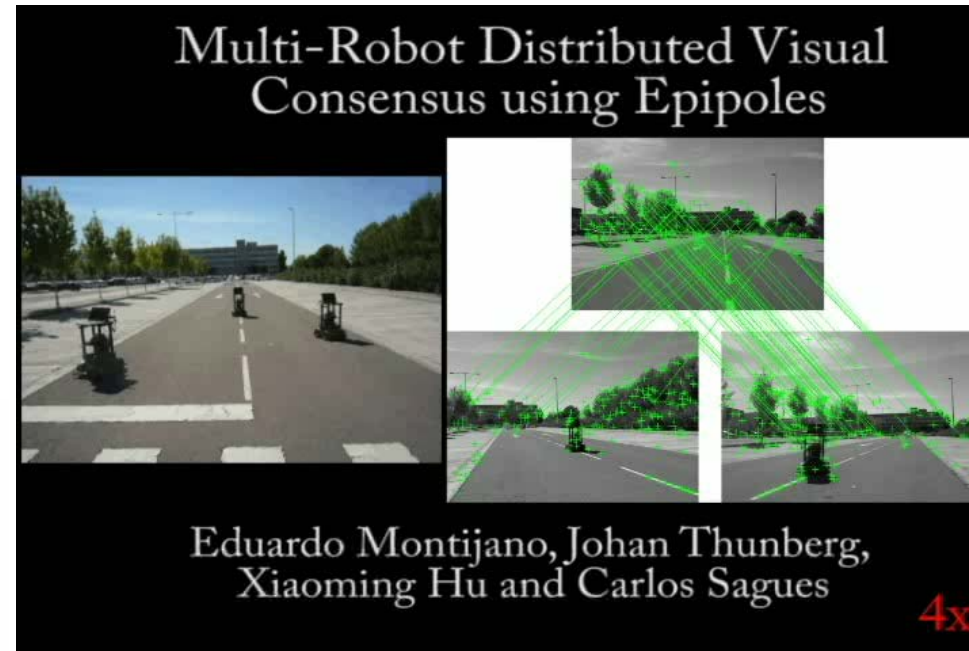
Multi robot control based on epipoles

Multi-Robot Distributed Visual Coordination using Epipoles

Eduardo Montijano, Johan Thunberg,
Xiaoming Hu and Carlos Sagues

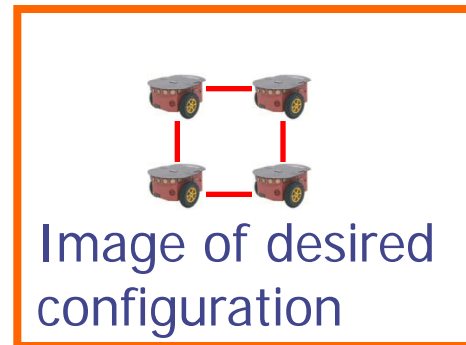
Multi-Robot Distributed Visual Coordination using Epipoles

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Multi robot control with flying camera (H)

- ◆ What? Visual control of mobile robots
 - Desired configuration defined by an image
 - Task: Navigate to the desired configuration



Initial configuration

Desired configuration

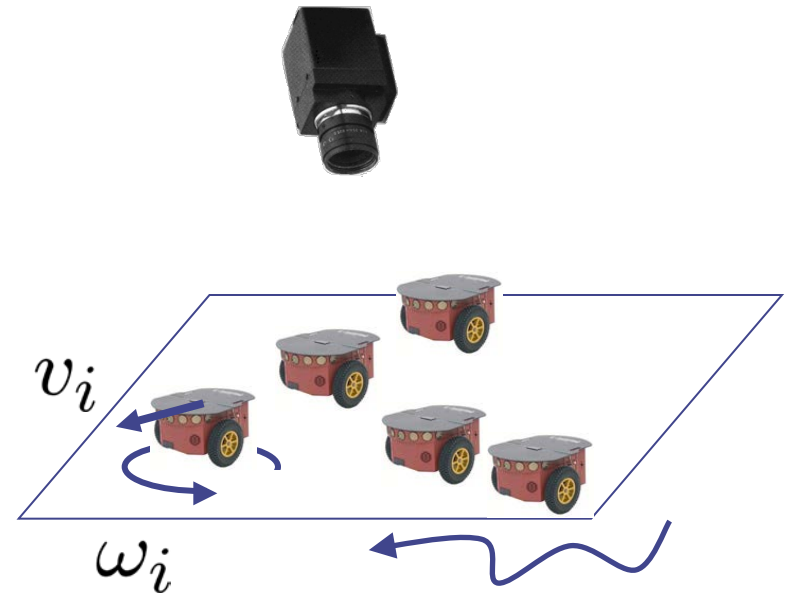
Multi robot control with flying camera (H)

- ◆ What? Visual control of mobile robots
- ◆ Who? Set of nonholonomic vehicles
 - Nonholonomic kinematics
 - ◆ Cartesian coordinates

$$\begin{aligned}\dot{x} &= -v \sin \phi \\ \dot{y} &= v \cos \phi \\ \dot{\phi} &= \omega\end{aligned}$$

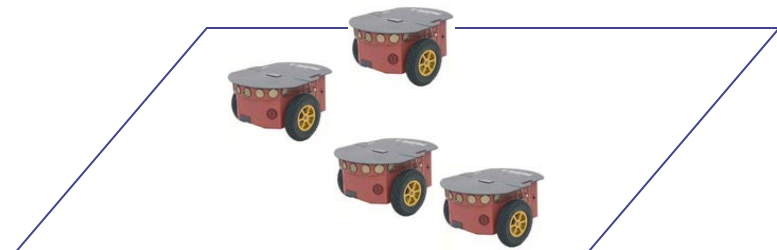
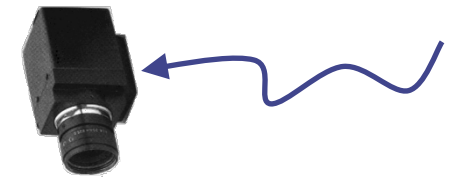
- ◆ Polar coordinates

$$\begin{aligned}\dot{\rho} &= v \cos \alpha \\ \dot{\alpha} &= \omega - \frac{v}{\rho} \sin \alpha \\ \dot{\phi} &= \omega\end{aligned}$$



Multi robot control with flying camera (H)

- ◆ What? Visual control of mobile robots
- ◆ Who? Set of nonholonomic vehicles
- ◆ How? Flying camera
 - ◆ Flying camera looking downward
 - ◆ Camera motion unknown
 - ◆ Intrinsic camera parameters known
 - ◆ Homography: Only visual information

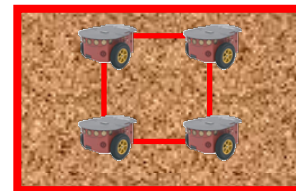


Multi robot control with flying camera (H)

- ◆ What? Visual control of mobile robots
- ◆ Who? Set of nonholonomic vehicles
- ◆ How? Flying camera
- ◆ Where? Motion occurs in a planar floor
 - This gives additional constraints on the homography
 - Only the set of robots may remain common in the scene



Image of desired configuration:



Desired configuration



Actual configuration



Multi robot control with flying camera (H)

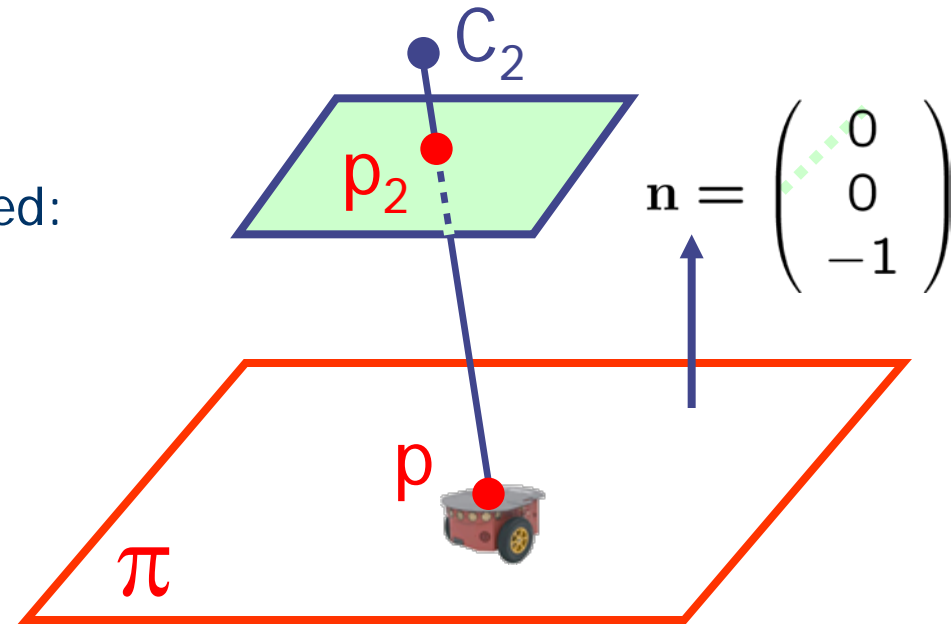
◆ The homography in our framework:

- Multi-robot motion in a planar floor
- Points = Robots => Homography
- Camera flies parallel to the floor

◆ Then, the homography is constrained:

$$H = \begin{bmatrix} h_{11} & h_{12} & h_{13} \\ h_{21} & h_{22} & h_{23} \\ 0 & 0 & 1 \end{bmatrix}$$

$$H = \begin{bmatrix} \cos \phi & \sin \phi & -t_x/d \\ -\sin \phi & \cos \phi & -t_y/d \\ 0 & 0 & 1 \end{bmatrix}$$



◆ This homography can be computed from a minimal set of two points/robots

Multi robot control with flying camera (H)

H_{rigid}

◆ If the robots are in the desired configuration:

- The homography is conjugate to a planar Euclidean transformation
- The homography is not the identity matrix

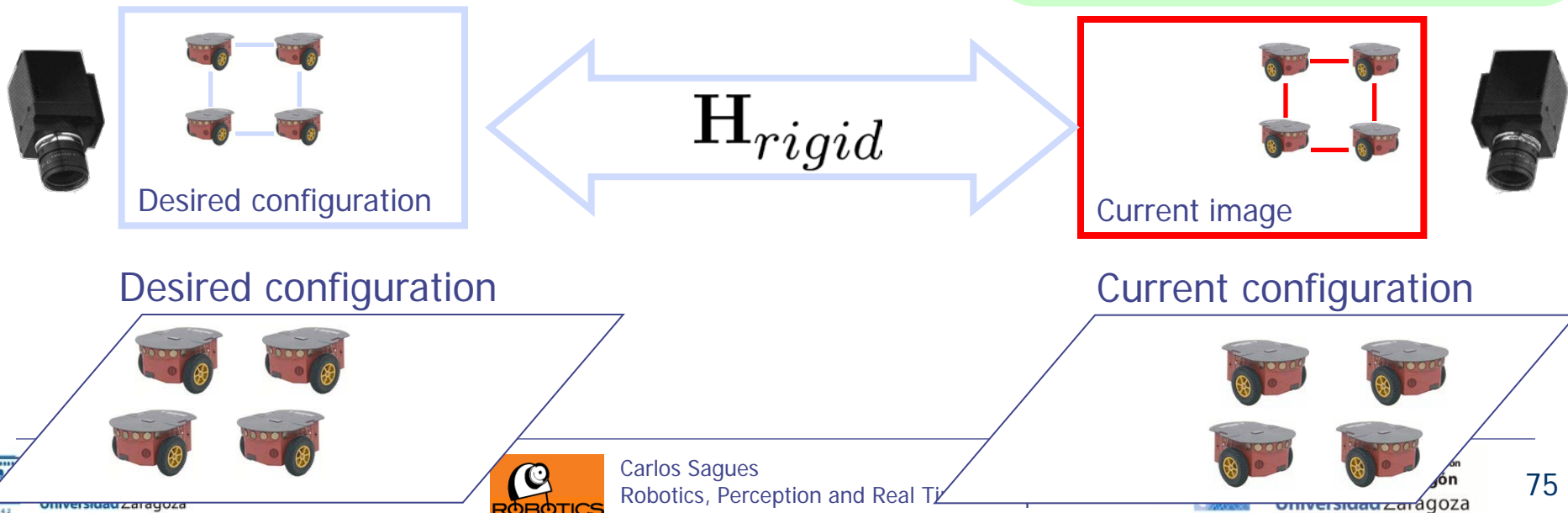
$$H_{rigid} = \begin{bmatrix} \cos \phi & \sin \phi & h_{13} \\ -\sin \phi & \cos \phi & h_{23} \\ 0 & 0 & 1 \end{bmatrix}$$

Extracting Motion parameters

$$\mathbf{n} = (0, 0, -1)^T$$

$$\mathbf{x} = (x, y, 0)^T$$

Which is coherent with a rigid motion. So, the robots are in the desired formation



Multi robot control with flying camera (H)

$H_{nonrigid}$

◆ If the robots are **NOT** in the desired configuration:

- The homography is a similarity transformation with isotropic scaling s
- The H computation with the 2-point method

$$H_{nonrigid} = \begin{bmatrix} s \cos \phi & s \sin \phi & h_{13} \\ -s \sin \phi & s \cos \phi & h_{23} \\ 0 & 0 & 1 \end{bmatrix}$$

Extracting Motion parameters

$$\mathbf{n} = (0, 0, -1)^T$$

$$\mathbf{x} = (x, y, (s-1)d^2)^T$$

Which is **NOT** coherent with a rigid motion. So, the robots are not in formation



Desired configuration



Current configuration



Multi robot control with flying camera (H)



- ◆ We have
 - Robots not in formation
 - Nonrigid homography
 - Each pair of robots induces a different Homography, valid but not coherent

$$\mathbf{H}_{nonrigid} = \begin{bmatrix} s \cos \phi & s \sin \phi & h_{13} \\ -s \sin \phi & s \cos \phi & h_{23} \\ 0 & 0 & 1 \end{bmatrix}$$

$$\mathbf{p}' = \mathbf{H}_{nonrigid} \mathbf{p}$$

- ◆ We want
 - Robots in formation
 - Rigid homography
 - Every pair of robots induce the same Homography

$$\mathbf{H}_{rigid} = \begin{bmatrix} \cos \phi & \sin \phi & h_{13} \\ -\sin \phi & \cos \phi & h_{23} \\ 0 & 0 & 1 \end{bmatrix}$$

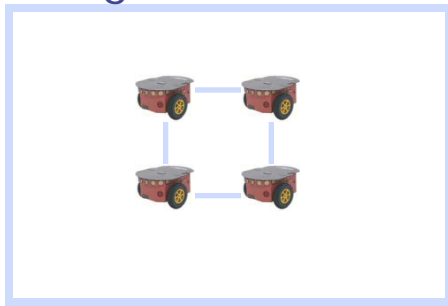
- ◆ We define a desired homography
 - Like the nonrigid homography but being induced by keeping the motion constraints
 - The task is to drive the robots to the desired homography
 - The desired homography is not constant and depends on the robots and camera motion

$$\mathbf{H}^d = \mathbf{H}_{nonrigid} \begin{bmatrix} 1/s & 0 & 0 \\ 0 & 1/s & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

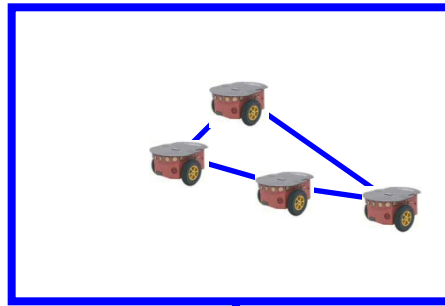
$$\mathbf{p}^d = (\mathbf{H}^d)^{-1} \mathbf{p}'$$

Multi robot control with flying camera (H)

Image of desired configuration



Current image



Flying camera



Target homography

Control law

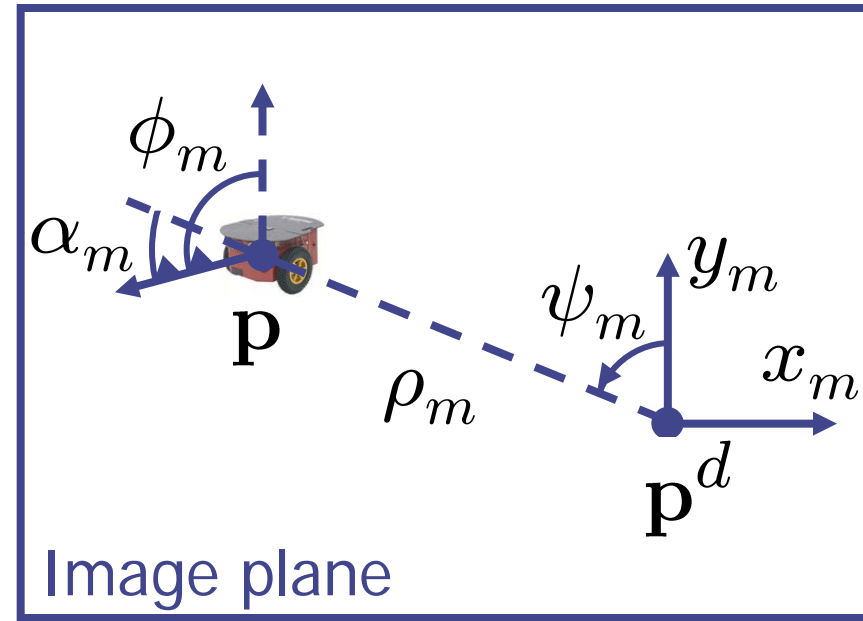
Set of robots:

$$\begin{pmatrix} v_i \\ \omega_i \end{pmatrix}$$



Multi robot control with flying camera (H)

- ◆ Image-based control law
- ◆ Control error:
 - Current state of the robots on the image vs desired states given by the desired homography
- ◆ Switched control consisting of three sequential steps:



$$\text{Step 1} \begin{cases} v = 0 \\ \omega = \dot{\psi}_c - k_\omega (\alpha_m - \pi) \end{cases}$$

$$\text{Step 2} \begin{cases} v = \dot{\rho}_d - k_v \rho_m \\ \omega = \dot{\psi}_c - k_\omega (\alpha_m - \pi) \end{cases}$$

$$\text{Step 3} \begin{cases} v = 0 \\ \omega = -k_\omega \left((\phi_m - \psi_{Fm}) - (\phi_m^0 - \psi_{Fm}^0) \right) \end{cases}$$

$$\rho_m = \sqrt{(p_x - p_x^d)^2 + (p_y - p_y^d)^2}$$

$$\psi_m = \text{atan2} \left(-(p_x - p_x^d), (p_y - p_y^d) \right)$$

$$\psi_{Fm} = \text{atan2} \left(-(p_x^i - p_x^j), (p_y^i - p_y^j) \right)$$

$$\mathbf{x}^d(t) = (x^d, y^d, \phi^d)^T$$

$$\dot{\rho}_d = \partial \rho_c / \partial \mathbf{x}^d$$

Multi robot control with flying camera (H)

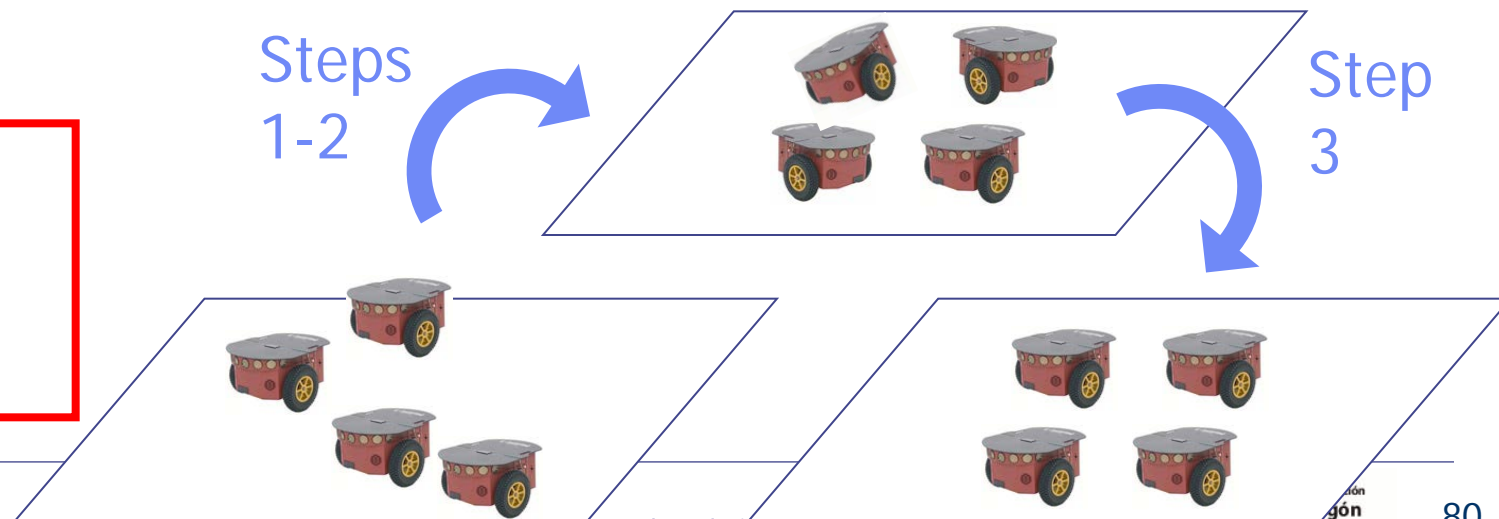
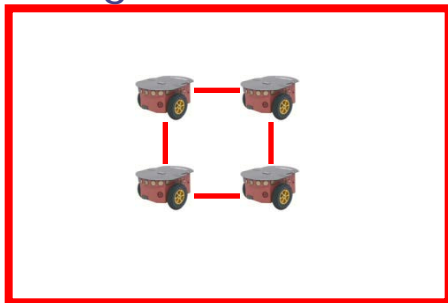
- Steps 1-2 orientate and drive the robots toward their target locations. In practice, they are carried out simultaneously:

$$\text{Step 1 and 2} \begin{cases} v = \dot{\rho}_d - k_v \rho_m \\ \omega = \dot{\psi}_c - k_\omega (\alpha_m - \pi) \end{cases}$$

- Step 3 rotates the robots until they are in the required relative orientation within the formation

$$\text{Step 3} \begin{cases} v = 0 \\ \omega = -k_\omega ((\phi_m - \psi_{Fm}) - (\phi_m^0 - \psi_{Fm}^0)) \end{cases}$$

Image of desired configuration



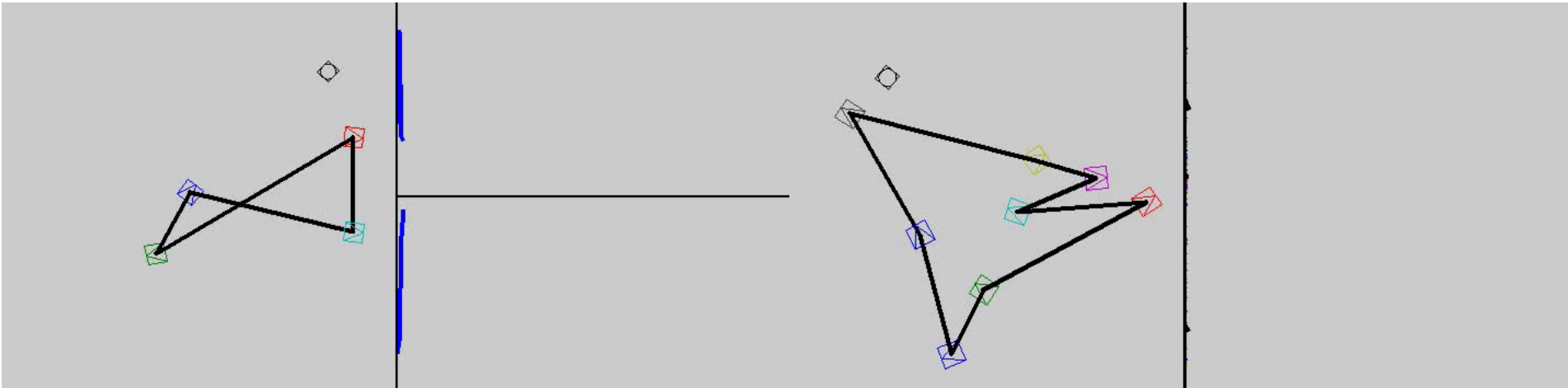
Multi robot control with flying camera (H)

Top view

Linear velocity: v

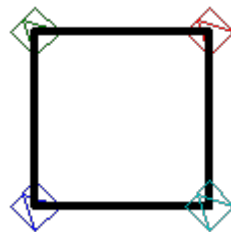
Top view

Homography entries

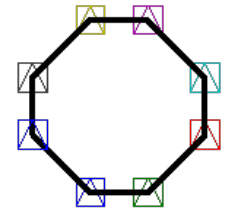


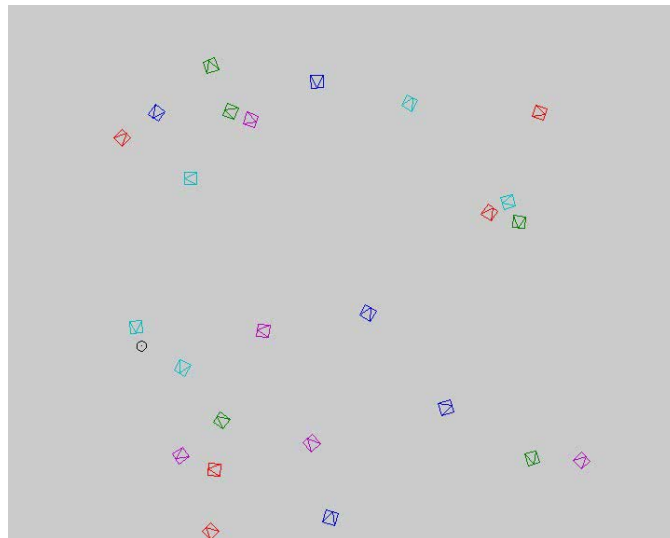
Angular velocity: ω

Desired configuration:



Desired configuration:





Control de robots y sistemas multi-robot basado en visión

Ciclo de conferencias
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UNED – ETS Ingeniería Informática

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Colaboradores:

Gonzalo López Nicolás
Héctor Manuel Becerra
Rosario Aragüés
Eduardo Montijano
Miguel Aranda

Carlos Sagues
Universidad de Zaragoza
<http://www.unizar.es/~csagues>