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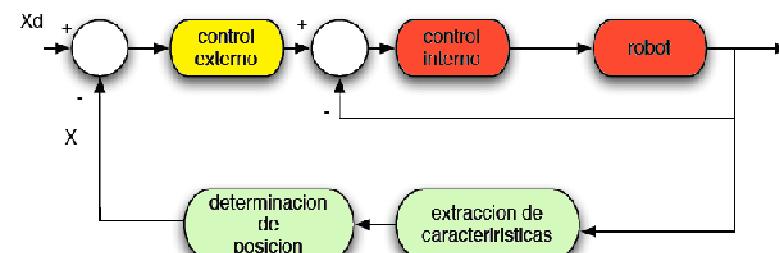
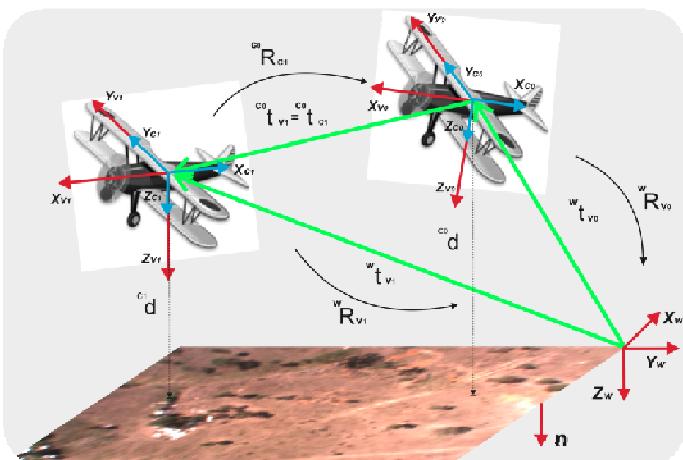
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Curso: Técnicas Avanzadas de Visión por Computador Máster Automática y Robótica

Pose Estimation and Position Based Visual Servoing



Carol Martínez
May 2011



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OUTLINE

- Introduction
- Position Based Visual Servoing PBVS
- Pose estimation techniques
- Results
- Summary
- References



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Introduction

INTRODUCTION

Problem

To track and follow the car with one of the **ROBOTS** we have (ground or aerial)

Visual servo control,

Visual servoing:

The use of computer vision data to control the motion of a robot





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Introduction

INTRODUCTION

Problem

To track and follow the car with one of the **ROBOTS** we have (ground or aerial)



To achieve the task we need: detection, segmentation, tracking, recognition, alignment-visual servoing.



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Introduction

INTRODUCTION

Problem

To track and follow the car with one of the **ROBOTS** we have (ground or aerial)



To achieve the task we need:

- Robust perception
- Robust control



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Introduction



Perceptual Robustness

- Camera configuration static or moving
- Number of cameras
- Calibrations issues
- Image processing techniques

According to

K. Toyama and G. Hager, Incremental focus of attention for robust visual tracking, CVPR 1996

*Robustness is the **ability of a vision-based tracking** system to track accurately and precisely during or after visual **circumstances that are less than ideal**. ... The robust vision-based tracking problem is therefore a vision-based tracking sub-problem – the problem of **coping with a complex environment***



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Introduction



Vision system requirements

1. **Handling** temporal inconsistencies in appearance and **occlusions** of the target object
2. **Handling** situations when the object is **outside of the FOV** (reinitialization)
3. **Adapt** to unpredictable object motion
4. Be insensitive to **lighting conditions** and specular reflections
5. **Detect errors** (in tracking or detection) and **to recover** the tracking afterwards
6. Produce estimates in **Real-Time**
7. Use **minimum a-priori knowledge** about the object



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Introduction

Lack of robustness due to

1- Figure-ground **segmentation**
(detection of the target or
initialization of tracking
sequence)

2- **Matching** across images (in
particular in the presence of large
and varying inter-frame motions)





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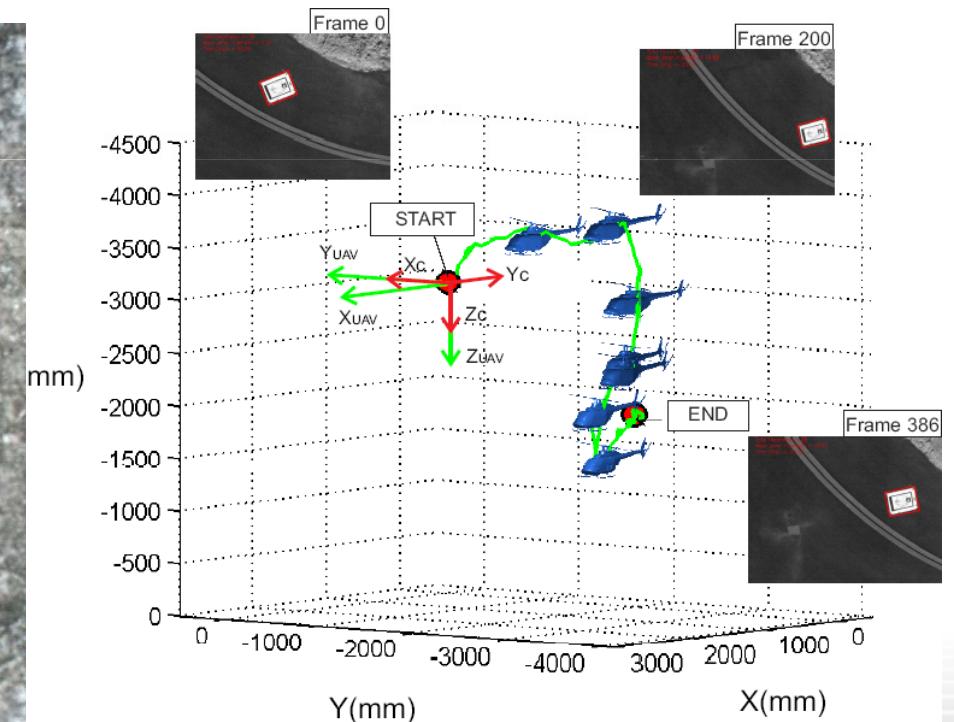
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Introduction

Lack of robustness due to

3- Inadequate modeling of motion (to enable prediction of the target in new images)





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Introduction



Lack of robustness due to

3- **Inadequate modeling** of motion (to enable prediction of the target in new images)



3 parameters (Tx, Ty, Rz)



4 parameters (Tx, Ty, Rz, scale)



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Introduction

However ...

There have been successful works using vision for controlling purpose



On-board cameras: monocular or stereo



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Introduction



However ...

There have been successful works using vision for controlling purpose



Eidgenössische Technische Hochschule Zürich
Swiss Federal Institute of Technology Zurich



Monocular Vision based Autonomous Helicopter
in Unstructured Environments



AUTONOMOUS SYSTEMS LAB August 2009

On-board camera: monocular



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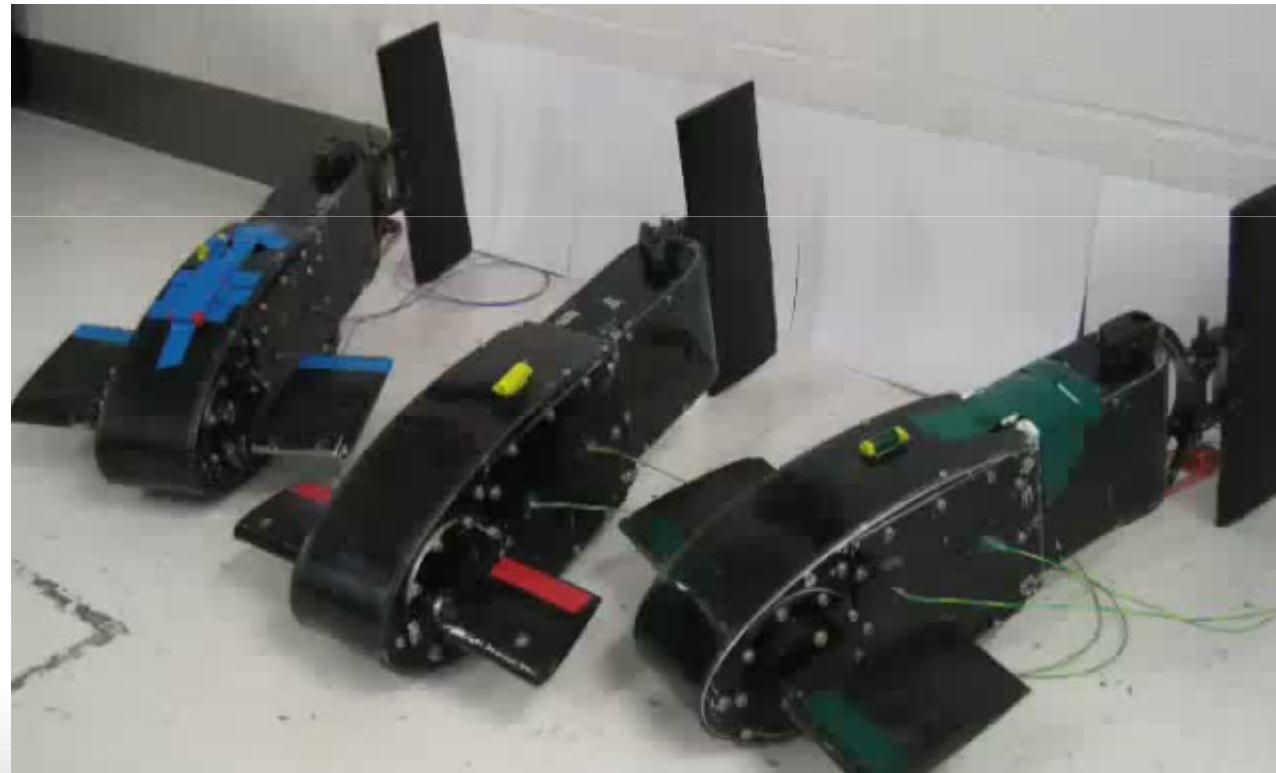
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Introduction

However ...

There have been successful works using vision for controlling purpose



External camera systyem



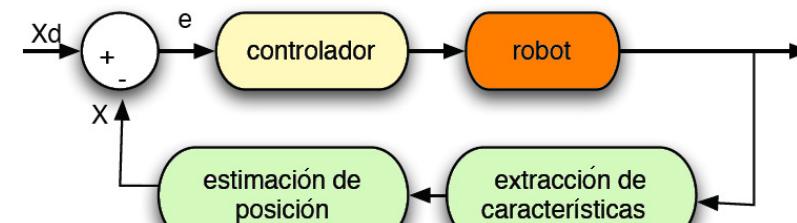
Position Based Visual Servoing PBVS

Classification of visual servoing techniques:

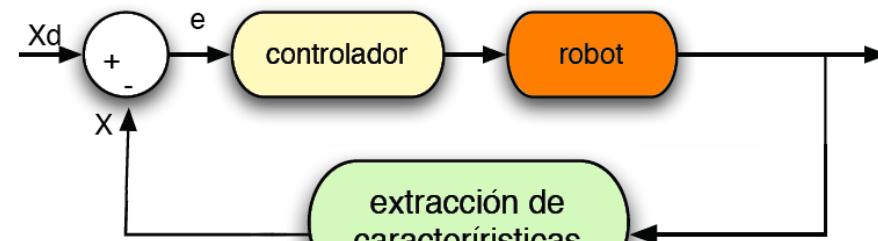
Image-Based (IBVS)

Position-Based (PBVS)

Hybrid approach



PBVS

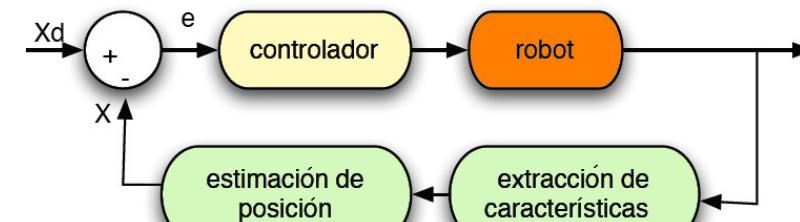


IBVS



Position Based Visual Servoing PBVS

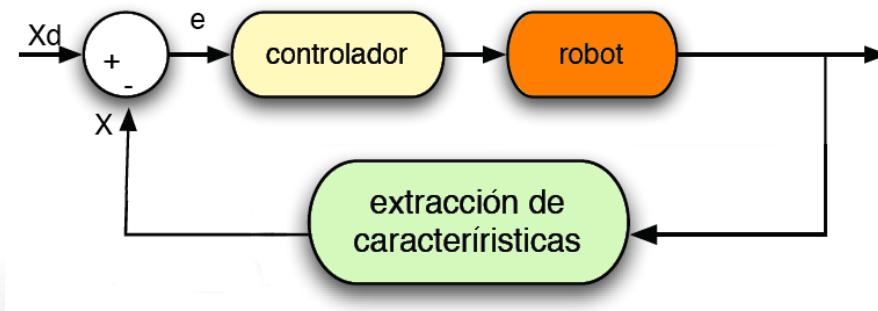
PBVS and **IBVS** are **different** in the nature of **the inputs used** in their respective control schemes



PBVS

Both approaches give **satisfactory results**:

convergence, stability, robust to camera calibration errors, measurements errors.

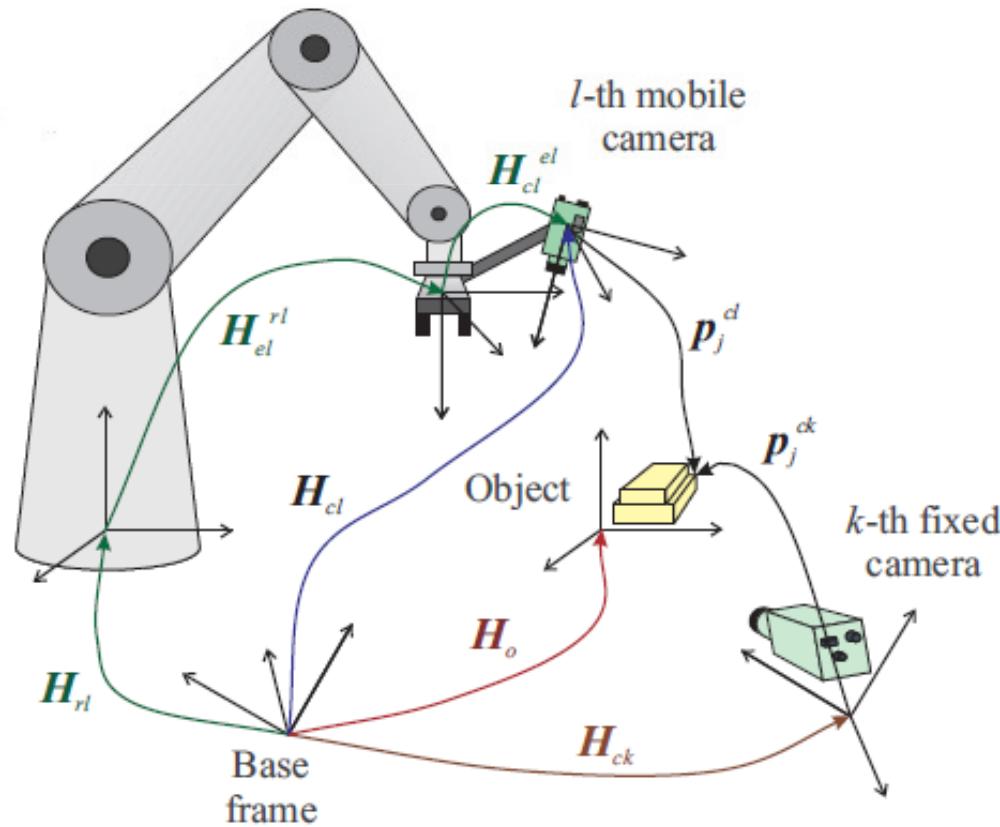


IBVS



Position Based Visual Servoing PBVS

Depending on the camera-robot configuration



- Eye to hand
- Eye in hand
- Hybrid approach



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PBVS



Position Based Visual Servoing PBVS

Depending on the number of cameras

- **Monocular:**

There is a **lost of information** (depth), make the control more complicated.

Positioning tasks look for solving this problem:

- Estimating depth before the tasks, or with metric information of the object.

“Almost used with eye in hand configuration”





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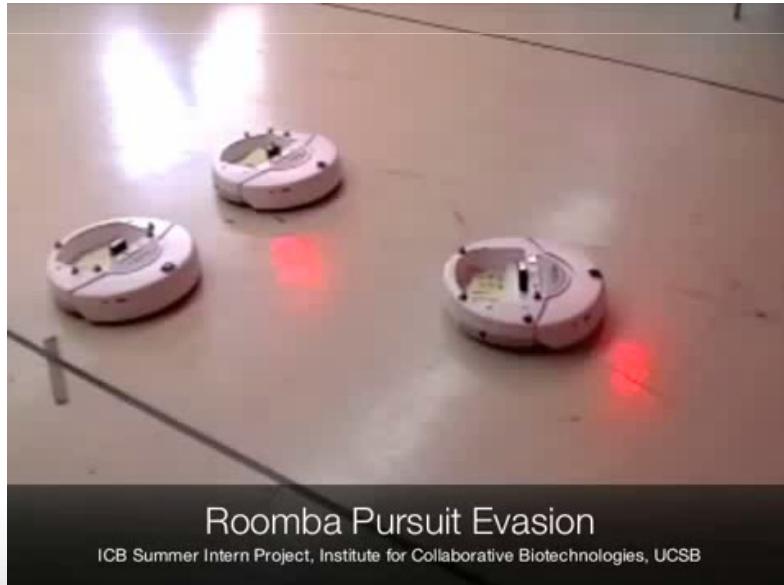
PBVS



Position Based Visual Servoing PBVS

Depending on the number of cameras

Stereo:
3D information can be obtained



Two autonomous robots try to catch a remotely controlled evader

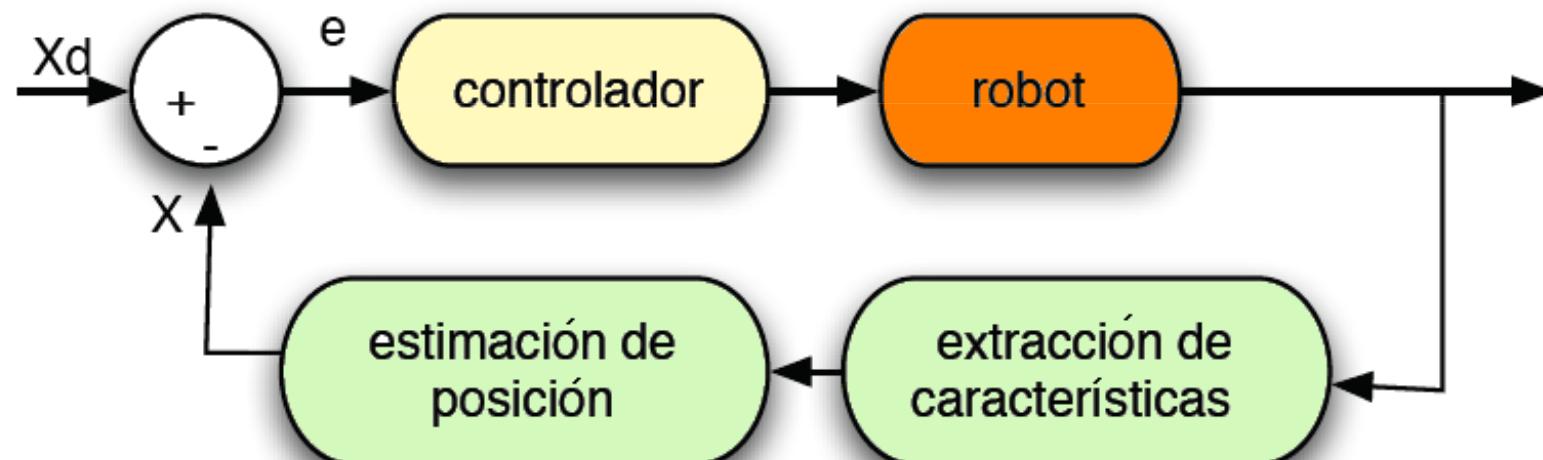


Redundant system:
3D information can be obtained.
Adding robustness.
Processing time increases



Position Based Visual Servoing PBVS

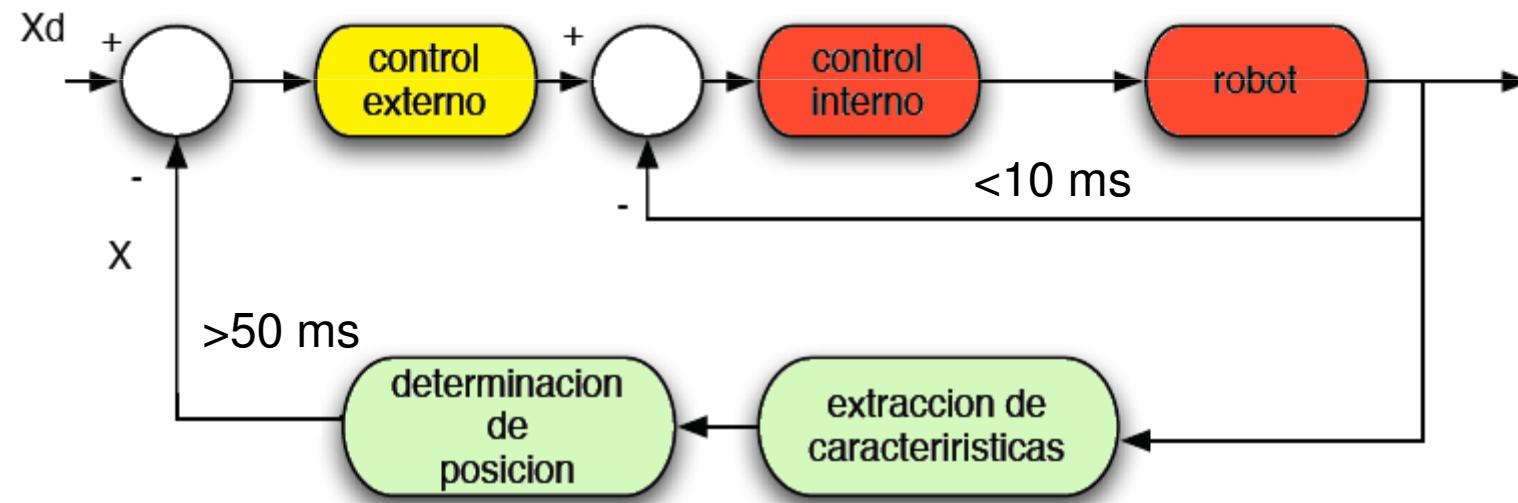
Control structure: direct visual control





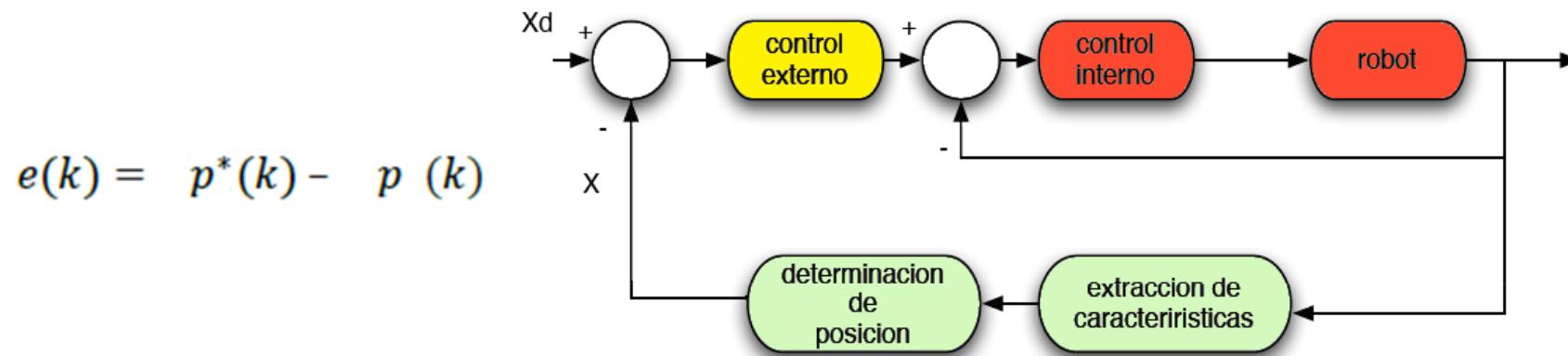
Position Based Visual Servoing PBVS

Control structure: indirect visual servoing, dynamic look and move





Position Based Visual Servoing PBVS

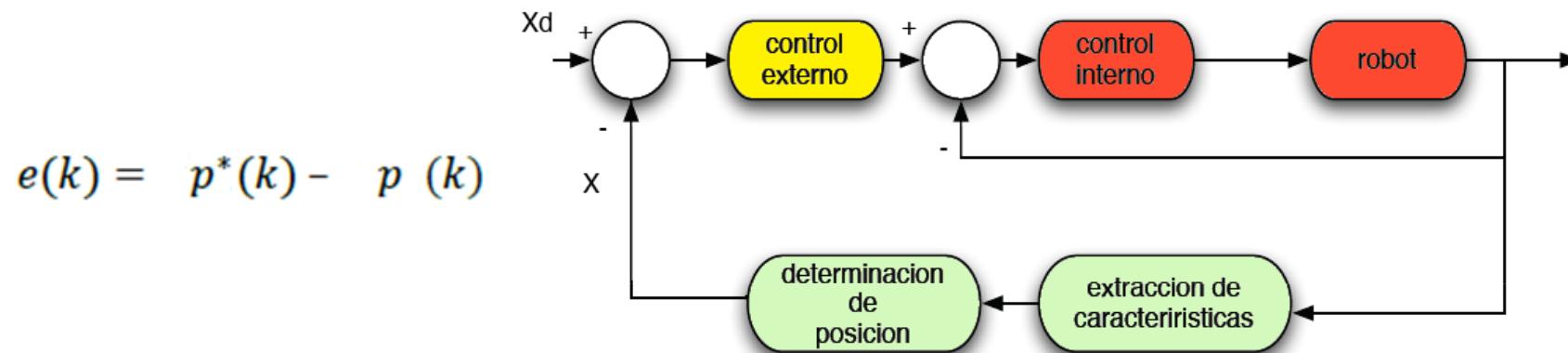


Error function based on the **3D cartesian space**, It is also called pose-based visual servoing.

Image features are extracted as well, but are additionally **used to estimate 3D** information (pose of the object in the cartesian space), hence it is servoing in 3D



Position Based Visual Servoing PBVS



Error function based on the **3D cartesian space**, It is also called pose-based visual servoing.

Image features are extracted as well, but are additionally **used to estimate 3D information** (pose of the object in the cartesian space), hence it is servoing in 3D

Geometric models: **required**

Camera calibration: **required**

Camera robot transformation: **required**



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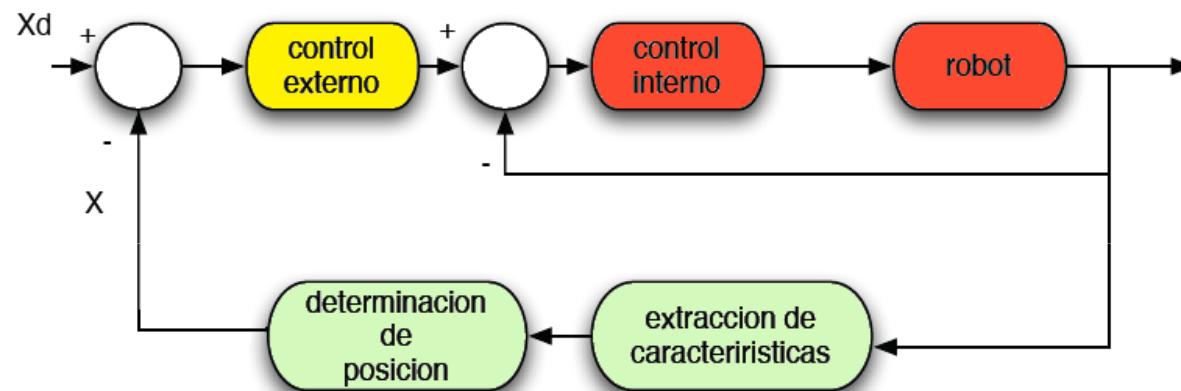
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PBVS



Position Based Visual Servoing PBVS



Because **there is not direct control in the image plane**, the object can **go out the field of view** of the camera during the control task.

Solution: observing the object and the robot.



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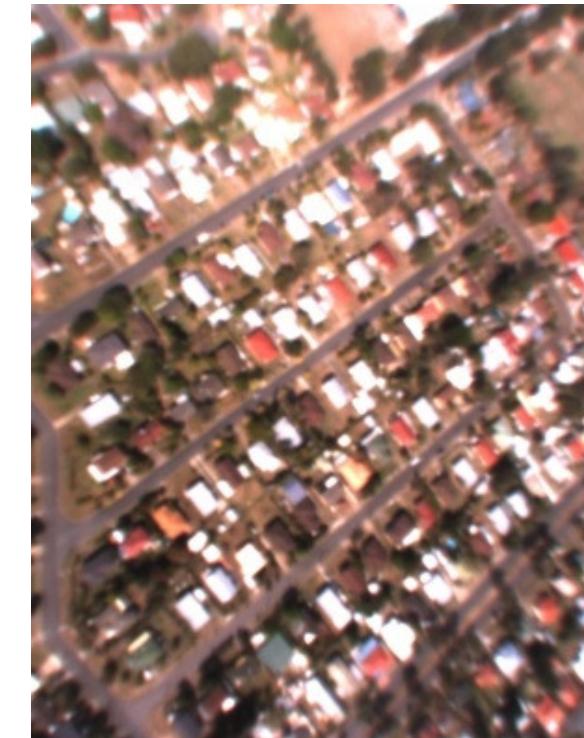


Pose Estimation

Solving the pose estimation problem



How to recover 6DOF?



Pose estimation using an on-board camera. There is not a specific object to follow



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Pose Estimation

Solving the pose estimation problem



Pose estimation using an on-board camera. Following a specific object



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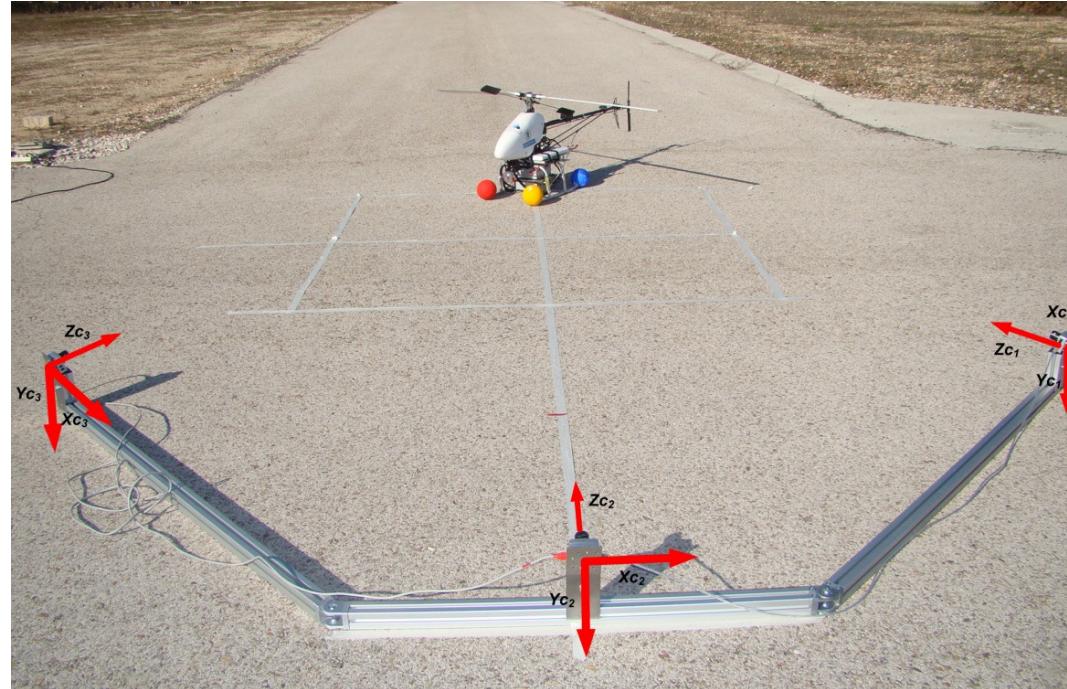
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Pose Estimation

Solving the pose estimation problem



Pose estimation using an external camera system.



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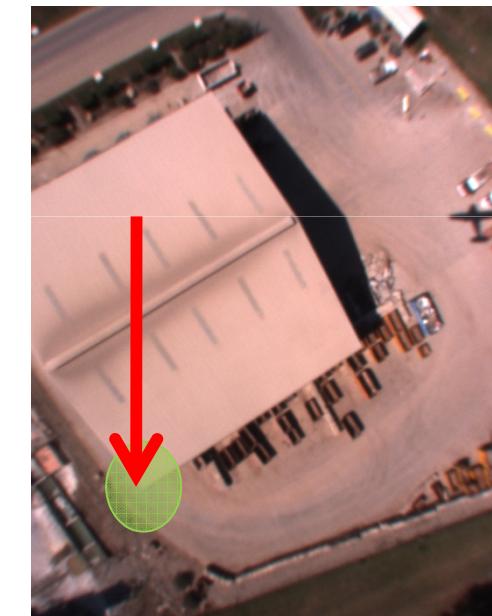
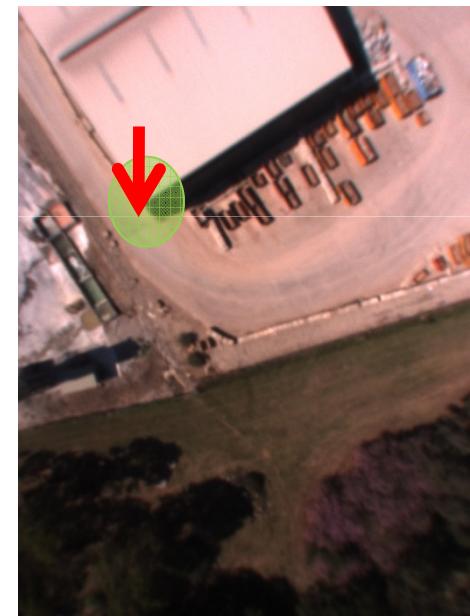
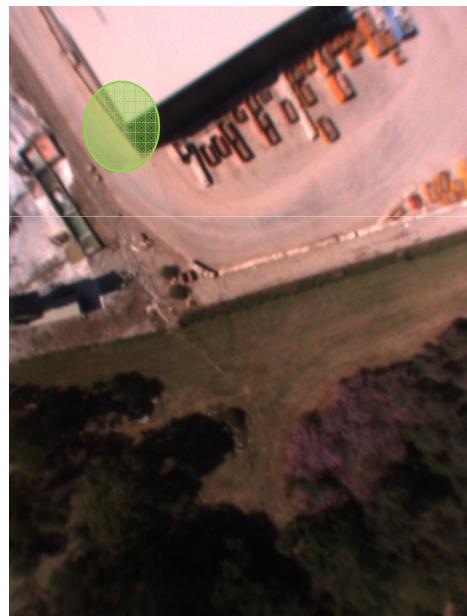
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Pose Estimation

Pose estimation Problem





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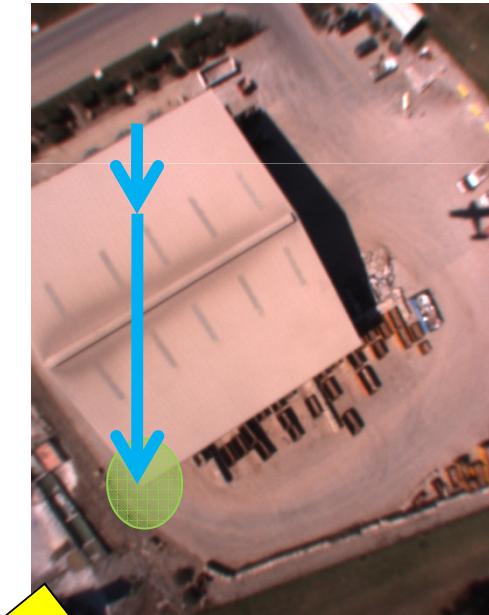
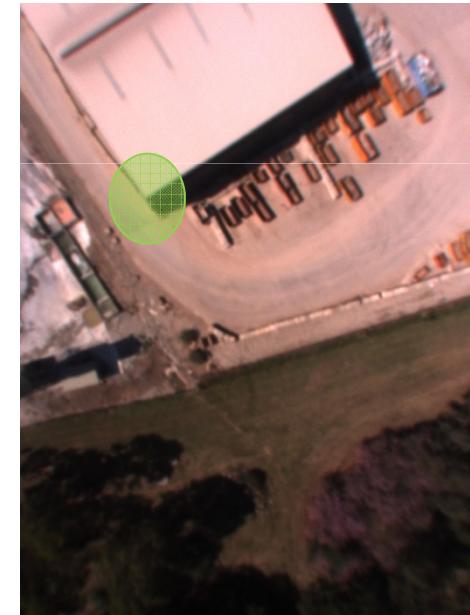
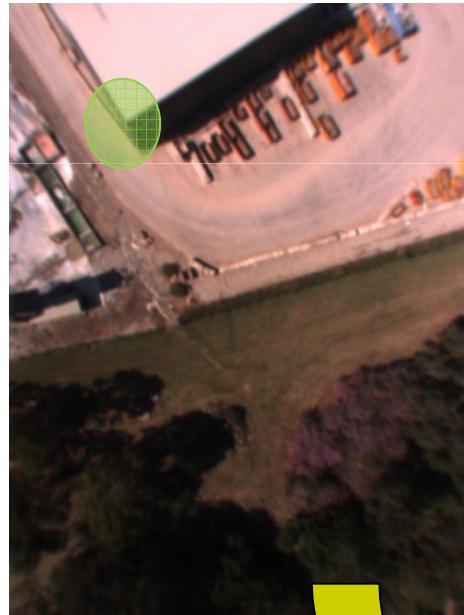
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Pose Estimation

Pose estimation Problem

Assuming flat terrain, dominant movement is due to vehicle movement



$$R_{tot}/t_{tot}$$



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Pose Estimation

Tracking of features



Frame to Frame Motion

Feature-based

Direct methods

Recovering different motion models:

- Translation
- Rotation
- Scale
- Homography



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Pose Estimation

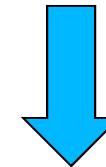
Pose estimation problem



Frame to Frame Motion

Homography

$$\mathbf{x}' = \begin{bmatrix} 1 + p_1 & p_2 & p_3 \\ p_4 & 1 + p_5 & p_6 \\ p_7 & p_8 & 1 \end{bmatrix} \mathbf{x}$$



$$\mathbf{H}_e = c_2 \mathbf{R}_{c_1} + \frac{1}{d} c_2 \mathbf{t}_{c_1} \mathbf{n}^T$$



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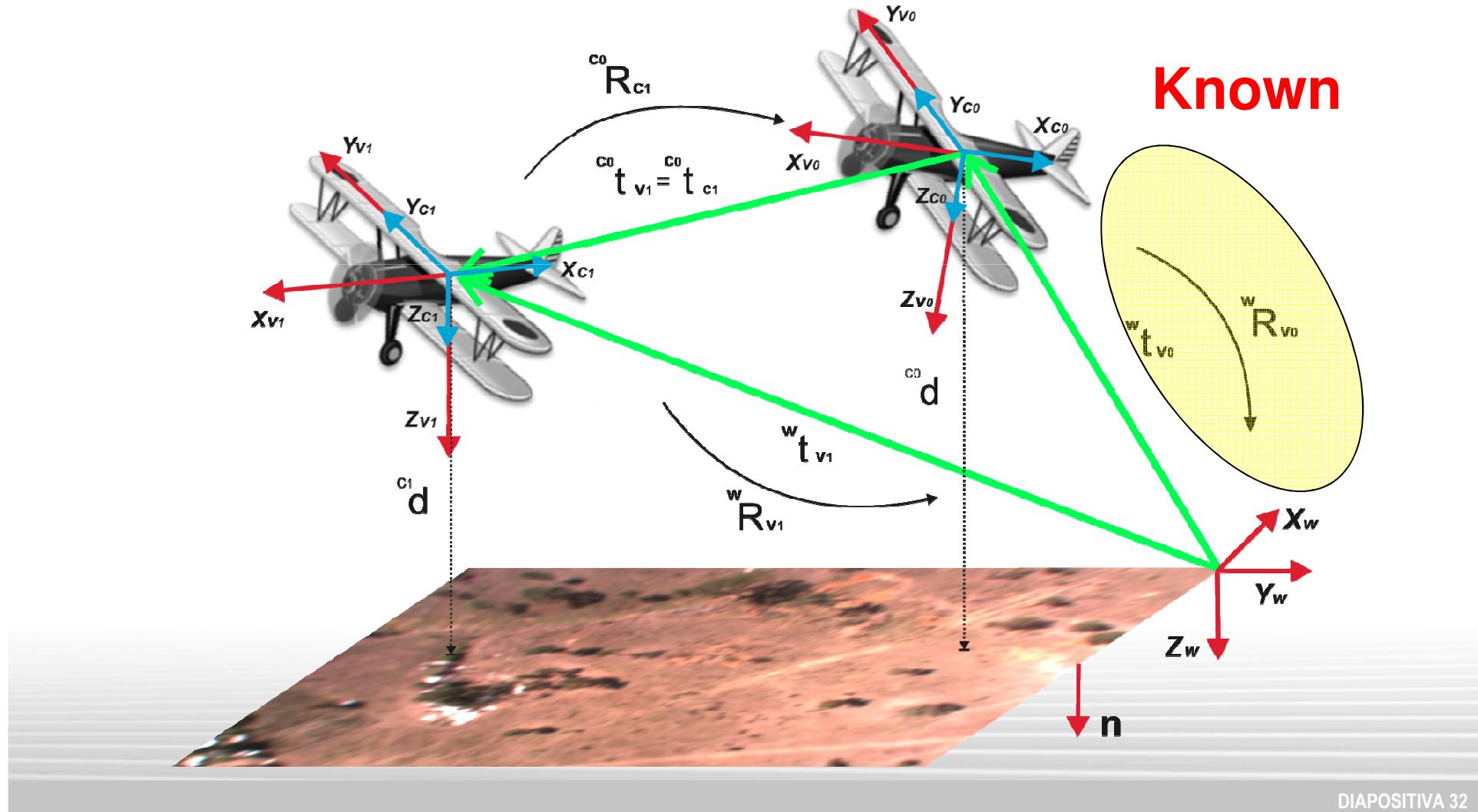
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Pose Estimation

Pose estimation problem





Pose estimation problem

$$\lambda \ i_x = M^w x$$

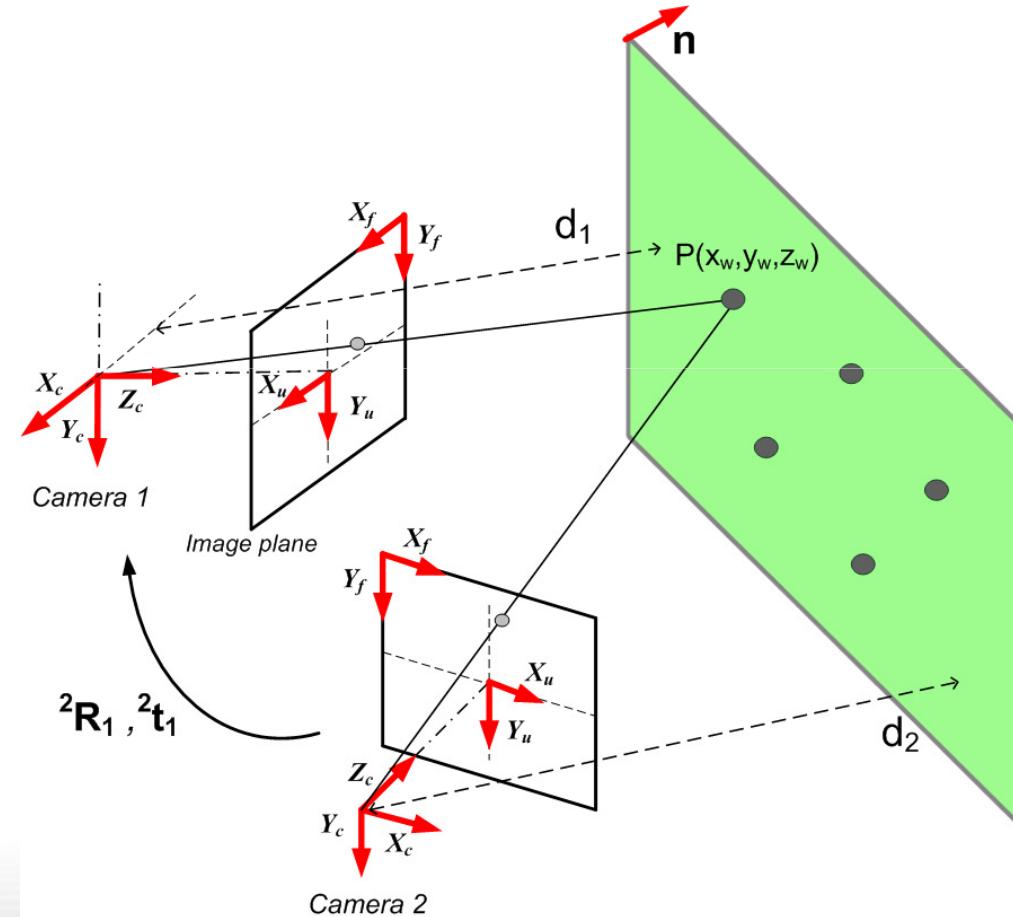
Scale factor $\rightarrow \lambda = \frac{Z}{f}$

Rigid transformation

$$c_2 x = c_2 R_{c_1} c_1 x + c_2 t_{c_1}$$

$$c_2 x = \left(c_2 R_{c_1} + \frac{1}{d} c_2 t_{c_1} n^T \right) c_1 x$$

$$H_e = c_2 R_{c_1} + \frac{1}{d} c_2 t_{c_1} n^T$$



Frame to frame estimation



Pose estimation problem: H decomposition

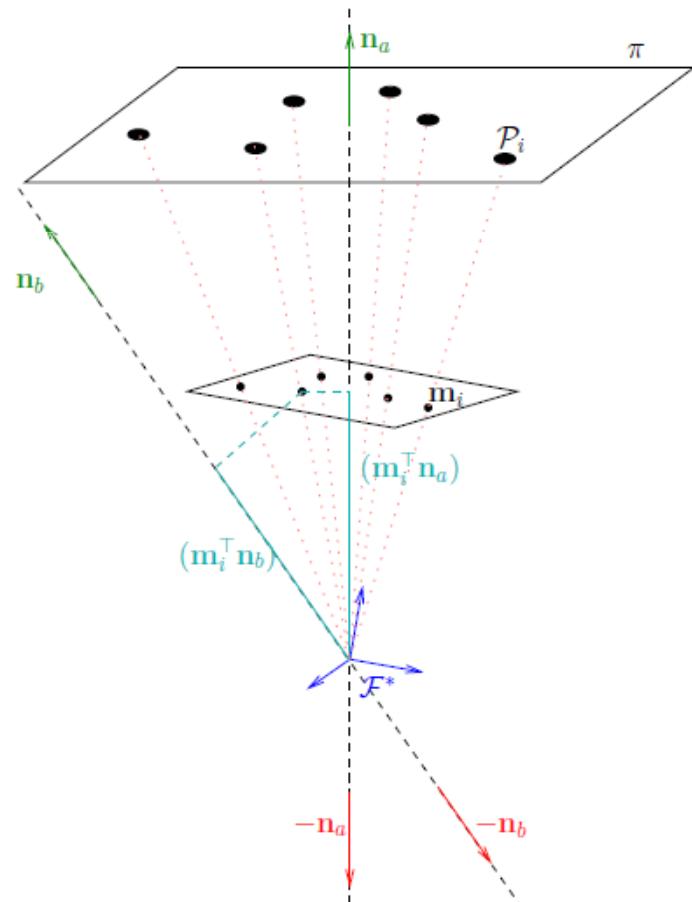
1- H decomposition: Method in book → “An invitation to 3D vision”

Solution 1	$R_1 = W_1 U_1^T$ $N_1 = \hat{v}_2 u_1$ $\frac{1}{d} T_1 = (H - R_1) N_1$	Solution 3	$R_3 = R_1$ $N_3 = -N_1$ $\frac{1}{d} T_3 = -\frac{1}{d} T_1$
Solution 2	$R_2 = W_2 U_2^T$ $N_2 = \hat{v}_2 u_2$ $\frac{1}{d} T_2 = (H - R_2) N_2$	Solution 4	$R_4 = R_2$ $N_4 = -N_2$ $\frac{1}{d} T_4 = -\frac{1}{d} T_2$



Pose estimation problem: H decomposition

2- From 4 solutions to 2: applying visibility constraint



All points seen by the camera must lie in front of it

$$\mathbf{m}^* = \mathbf{K}^{-1} \mathbf{p}^*$$

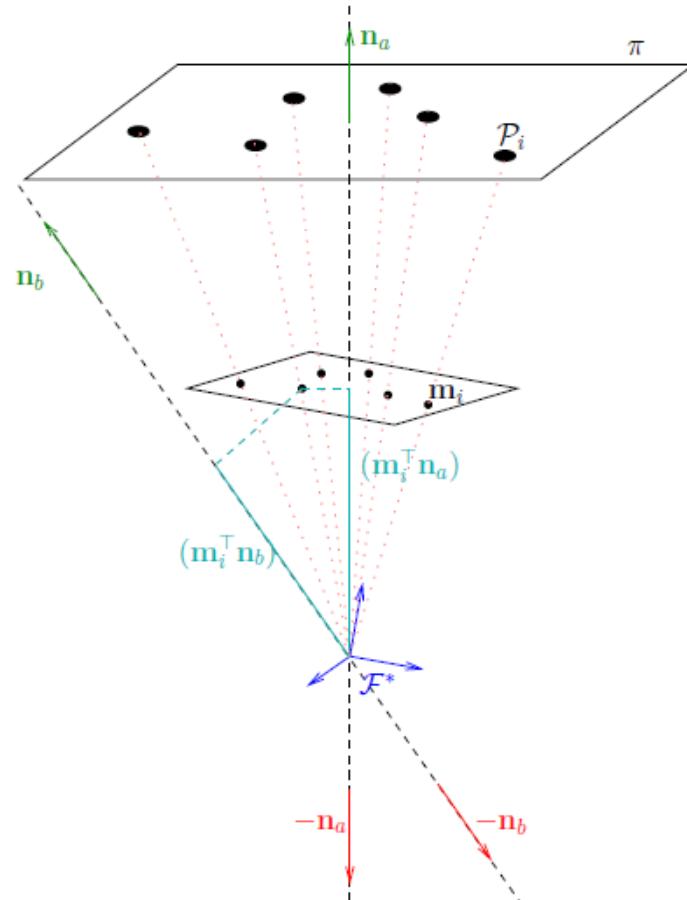
$$\mathbf{m}^{*\top} \mathbf{n}^* > 0$$

TWO SOLUTIONS



Pose estimation problem: H decomposition

3- From 2 solutions to 1: assuming flat terrain



$$n=[0, 0, 1]$$

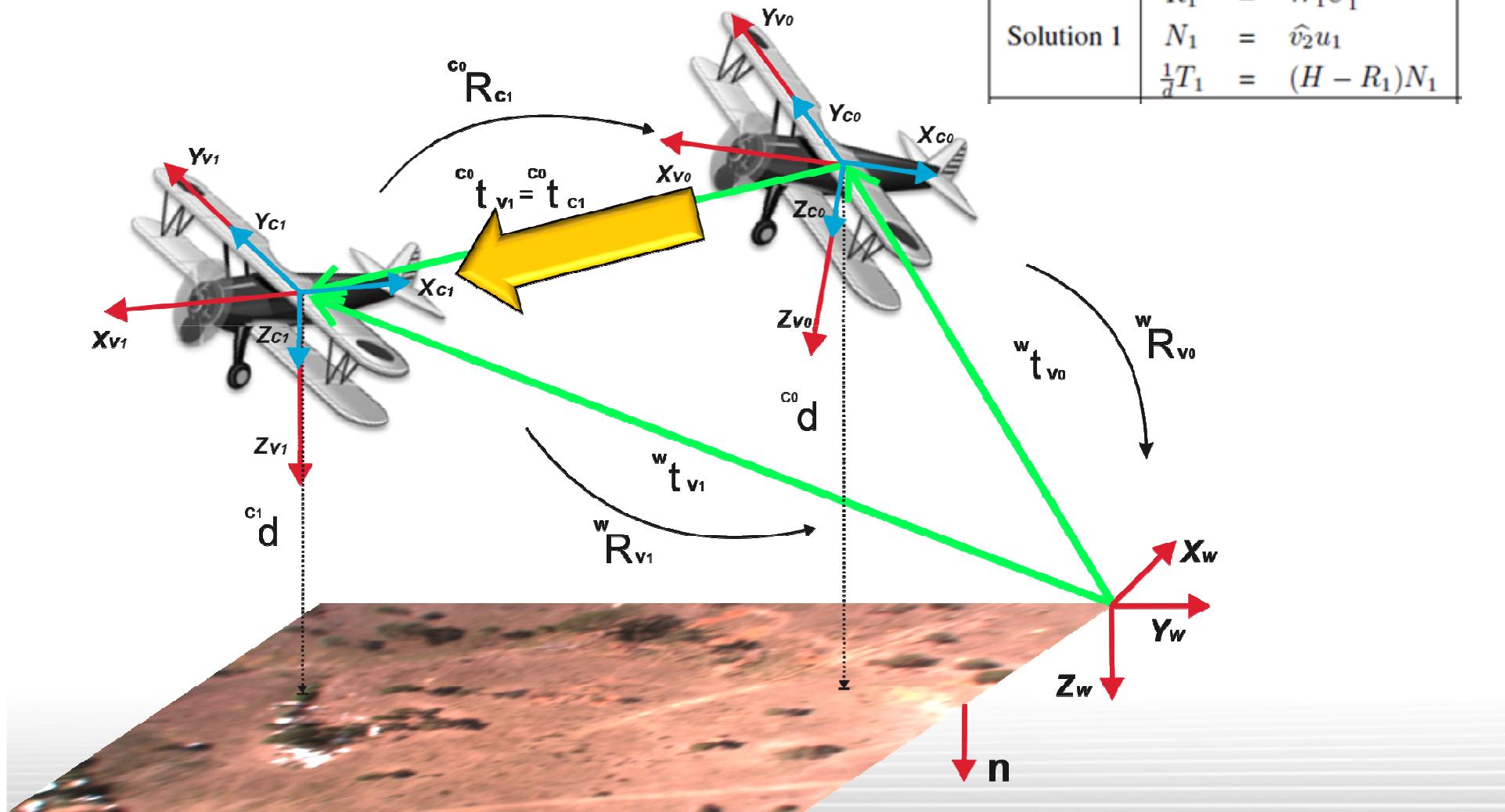
one SOLUTION

Solution 1	$R_1 = W_1 U_1^T$ $N_1 = \hat{v}_2 u_1$ $\frac{1}{d} T_1 = (H - R_1) N_1$	Solution 3	$R_3 = R_1$ $N_3 = -N_1$ $\frac{1}{d} T_3 = -\frac{1}{d} T_1$
Solution 2	$R_2 = W_2 U_2^T$ $N_2 = \hat{v}_2 u_2$ $\frac{1}{d} T_2 = (H - R_2) N_2$	Solution 4	$R_4 = R_2$ $N_4 = -N_2$ $\frac{1}{d} T_4 = -\frac{1}{d} T_2$



Pose Estimation

Solution 1	$R_1 = W_1 U_1^T$
	$N_1 = \hat{v}_2 u_1$
	$\frac{1}{d} T_1 = (H - R_1) N_1$

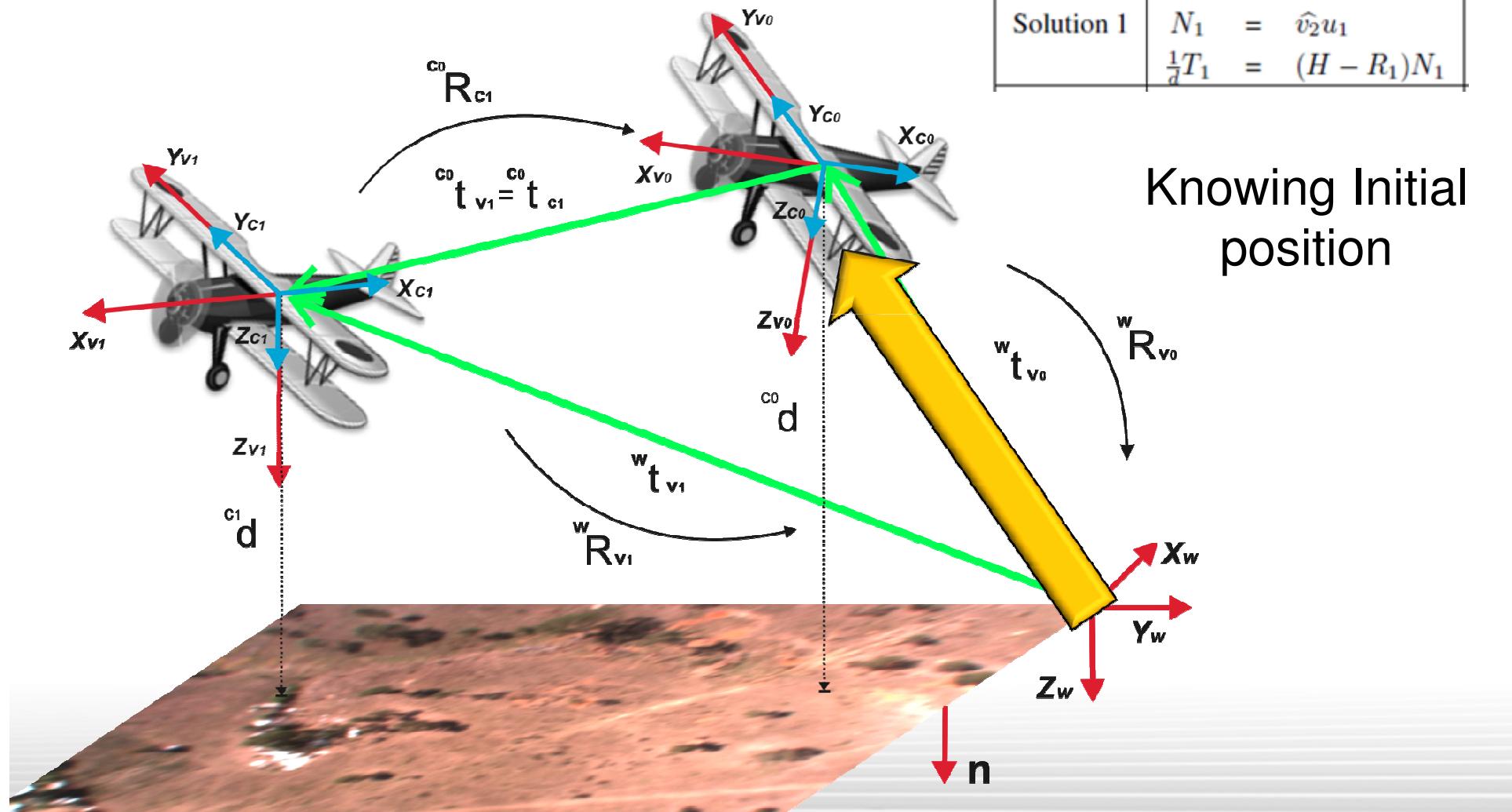




Pose Estimation

Solution 1	$R_1 = W_1 U_1^T$
	$N_1 = \hat{v}_2 u_1$
	$\frac{1}{d} T_1 = (H - R_1) N_1$

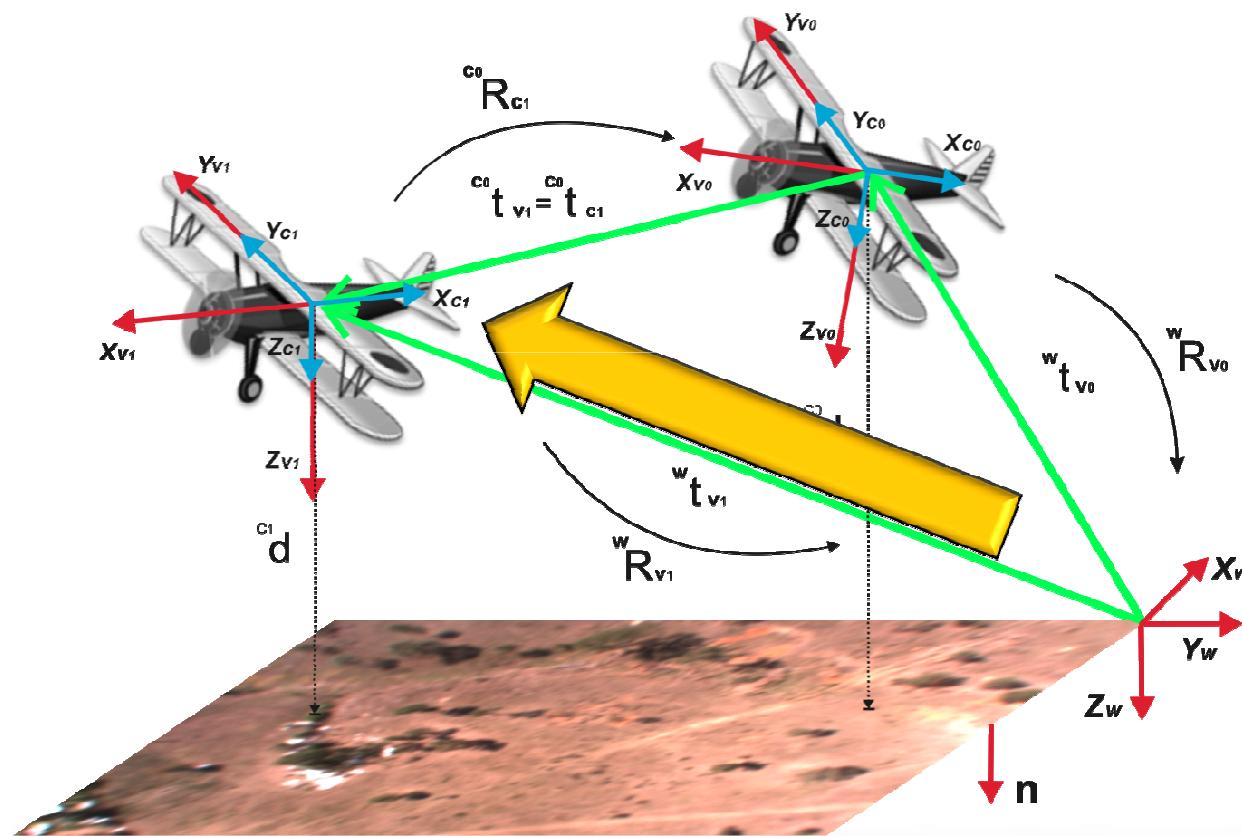
Knowing Initial position



Pose Estimation

Solution 1	$R_1 = W_1 U_1^T$
	$N_1 = \hat{v}_2 u_1$
	$\frac{1}{d} T_1 = (H - R_1) N_1$

6 DOF are recovered



$${}^w\mathbf{R}_{c_1} = {}^w\mathbf{R}_{c_0} \, {}^{c_0}\mathbf{R}_{c_1}$$

$${}^w\mathbf{t}_{c_1} = {}^w\mathbf{R}_{c_0} \, {}^{c_0}\mathbf{t}_{c_1} + {}^w\mathbf{t}_{c_0}$$



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Pose Estimation Results



- This strategy has been **used for pose estimation** of aerial vehicles using **frame to frame motion** estimation.



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Pose Estimation Results

Results: take-off

Results:

- Similar Behavior
- Low drift, only based on visual information

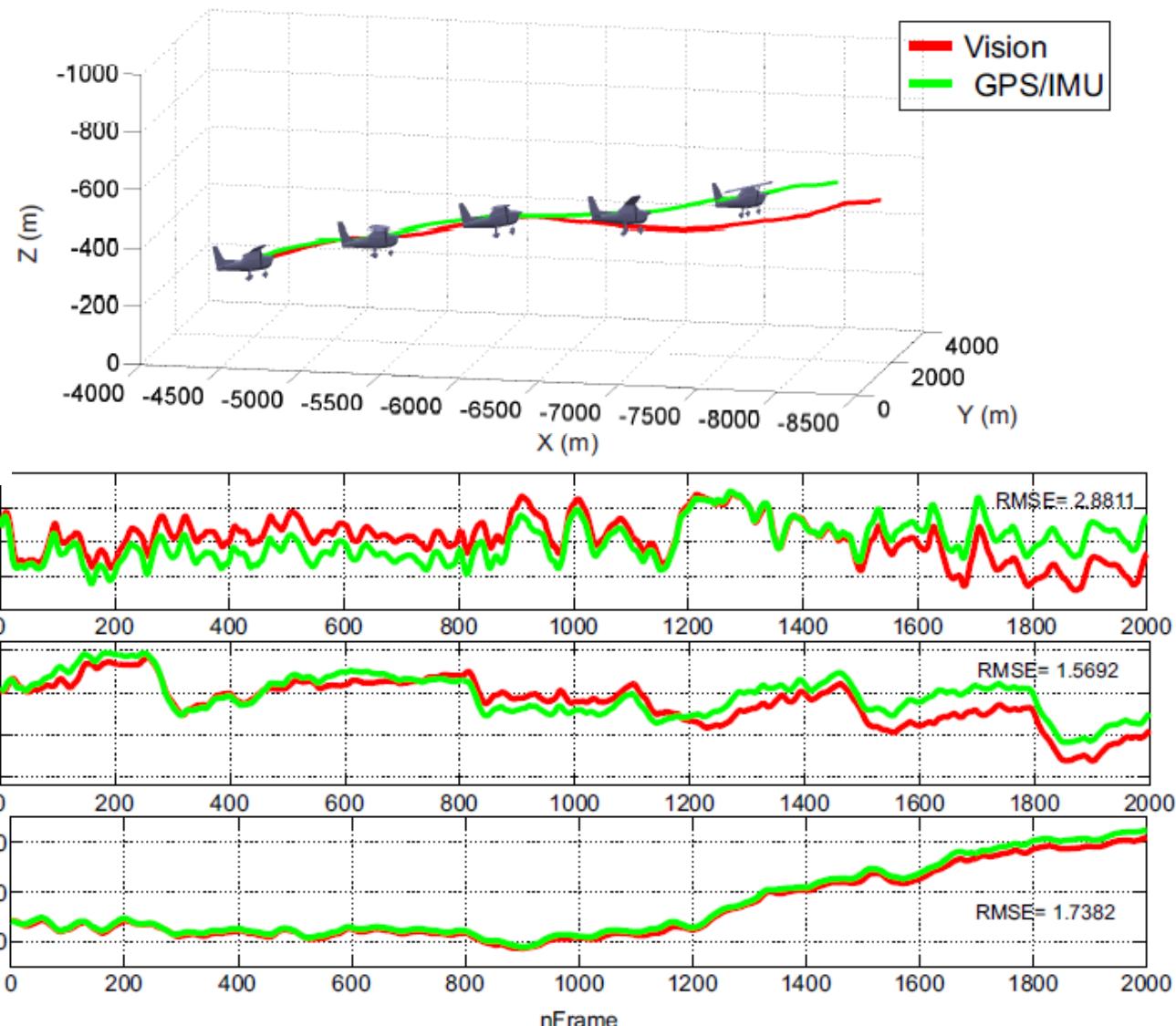
1 minute flight

MAPE x,y,z

[13.74%, 3.32%, 7.26%]

RMSE roll, pitch, yaw

[2.8, 1.5, 1.7] deg





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Pose Estimation Results



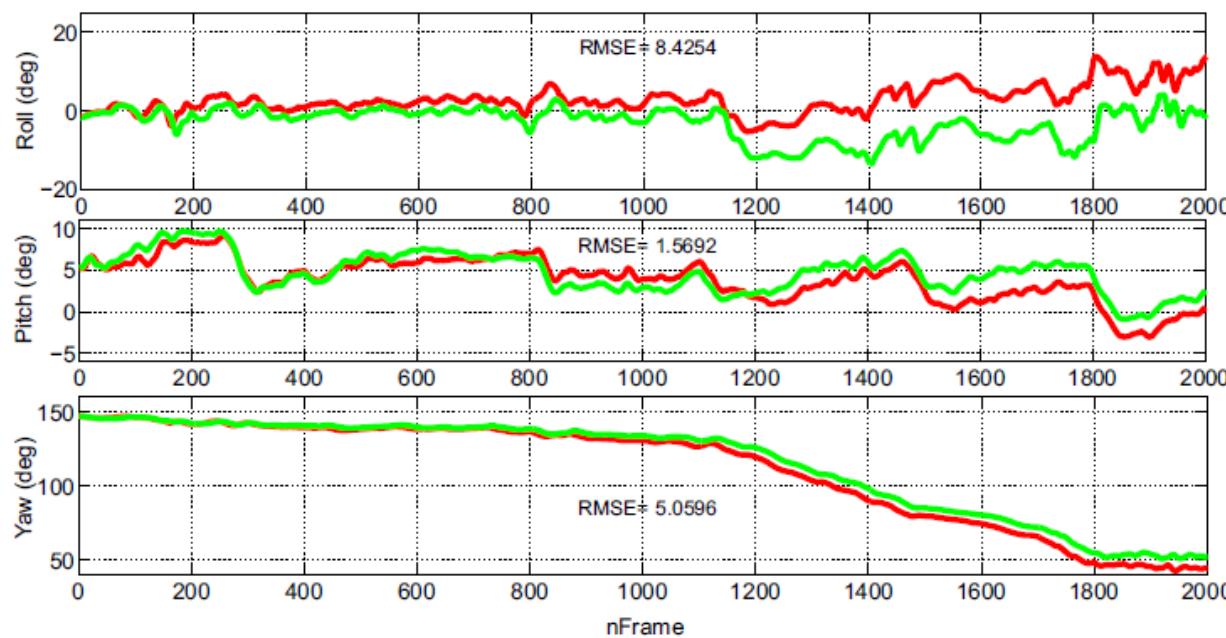
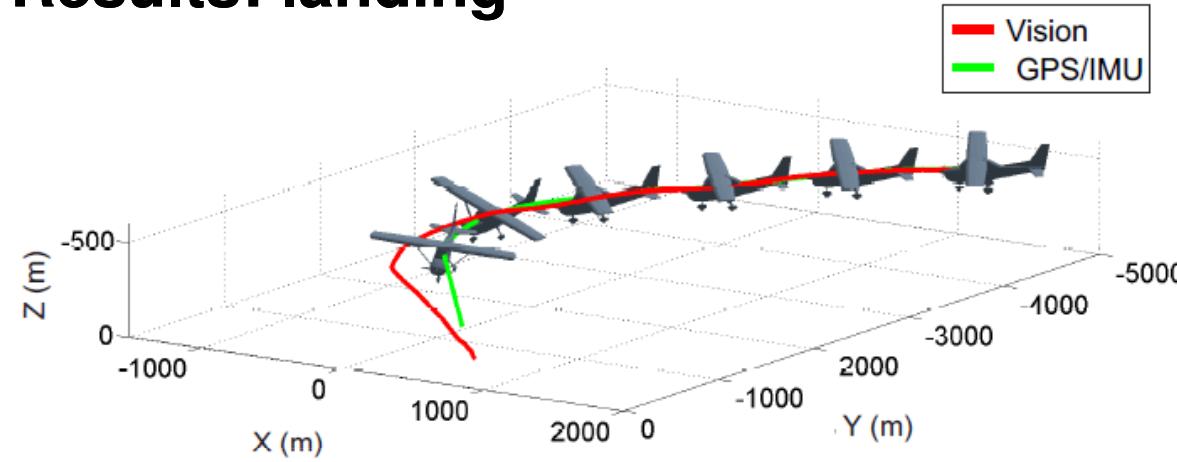
Results: cruise

Cruise



Pose Estimation Results

Results: landing



Results:

- Similar Behavior
- Low drift, only based on visual information

MAPE x,y,z

[8.12%, 15.44%, 3.70%]

RMSE roll, pitch, yaw

[8.4, 1.5, 5] deg



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Pose Estimation



Problems

- Planar assumption
- Drift due to integration of the data.
- What if there is a frame to frame error, it is integrated



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Pose Estimation

Using a external camera system

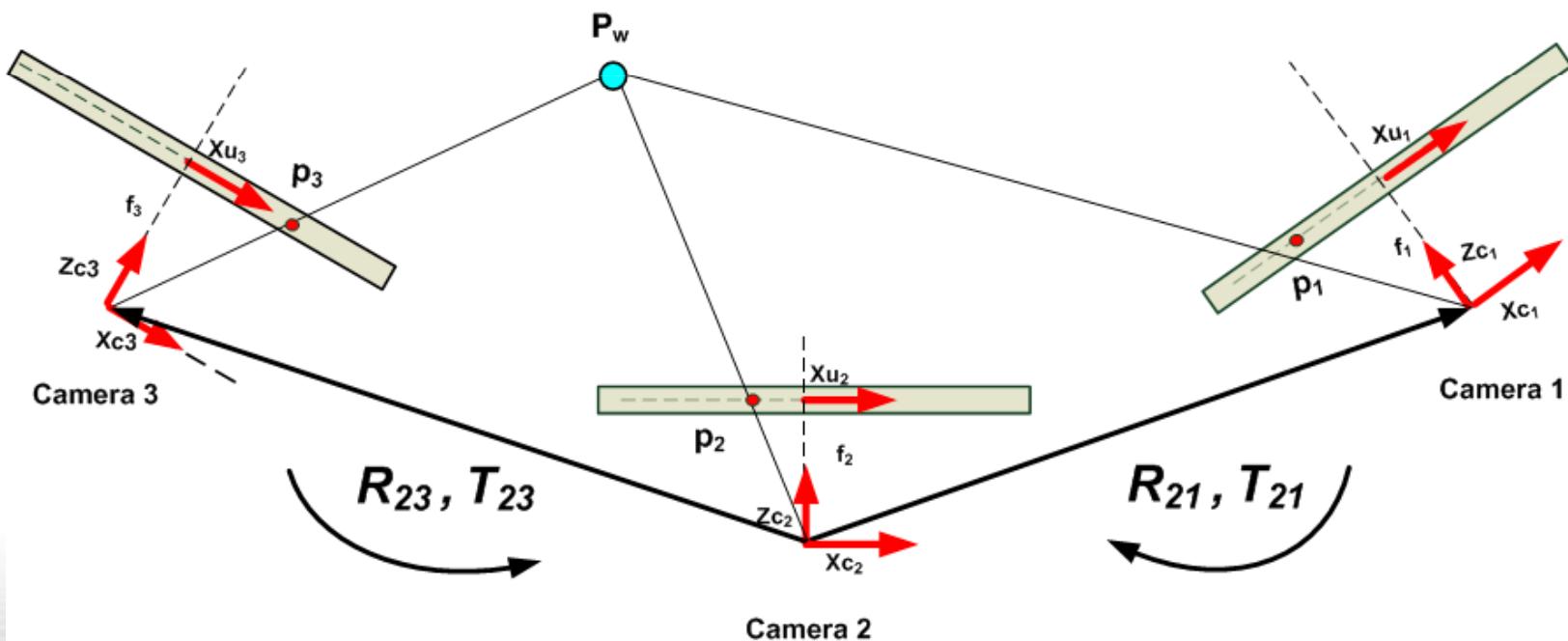
Define 3D position by detecting the coordinates of the object in each image





Using a external camera system

Define 3D position by detecting the coordinates of the object in each image





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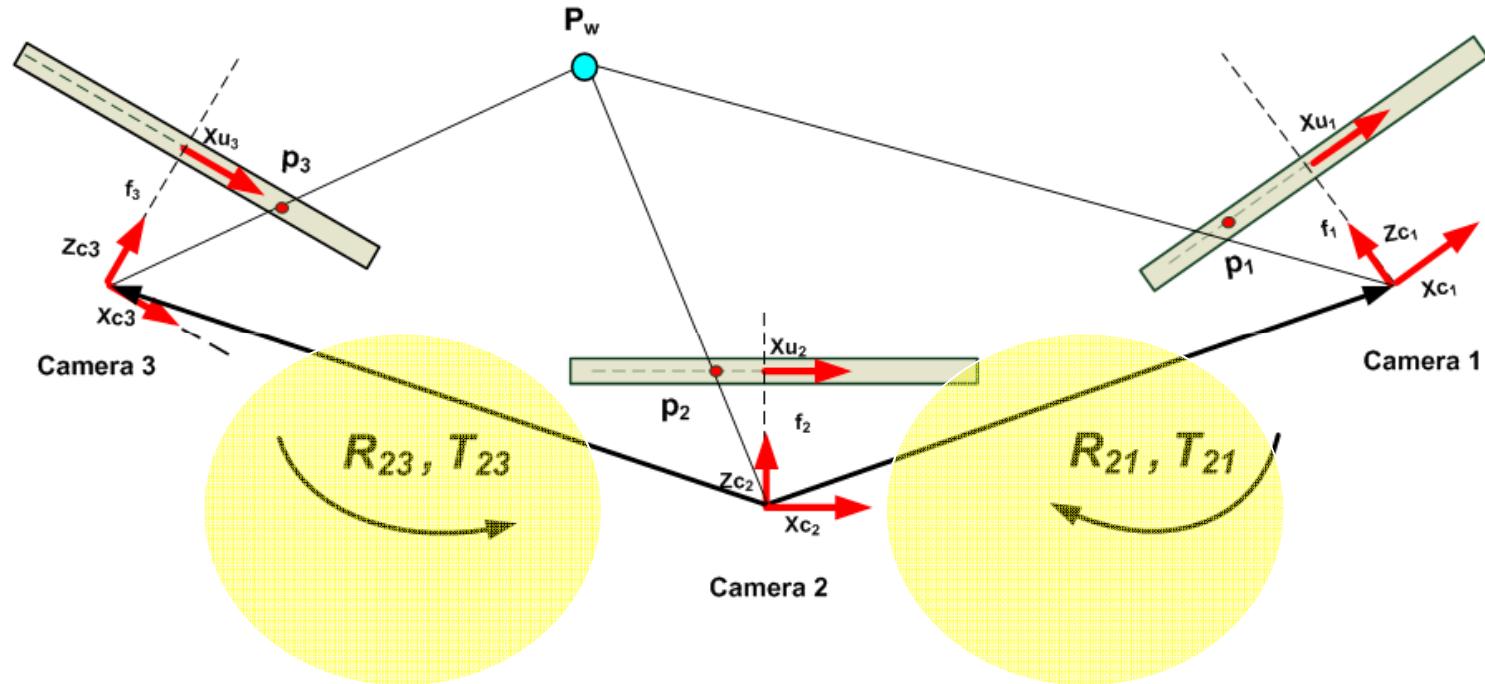
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Pose Estimation

Using a external camera system

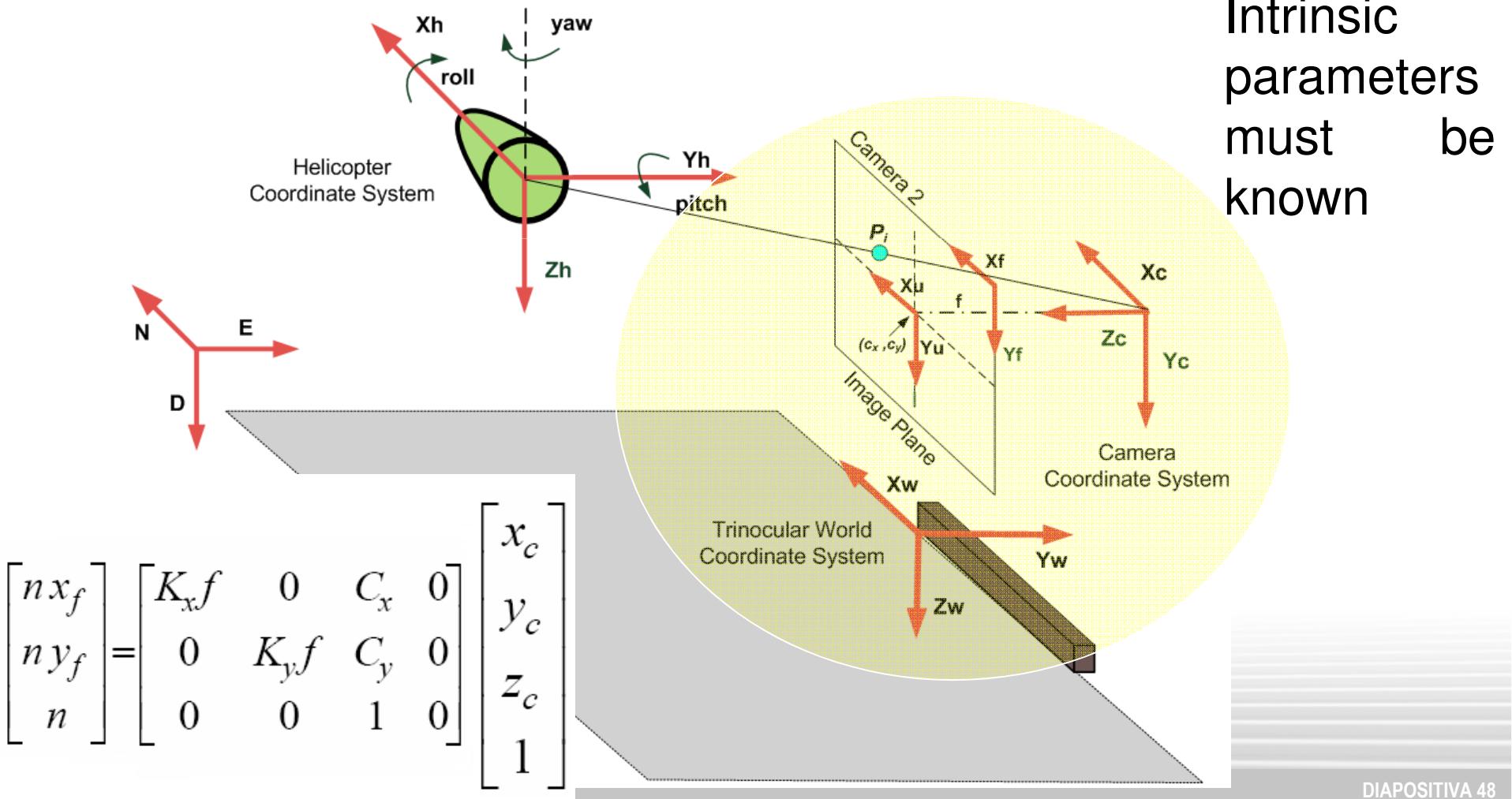


Extrinsic parameters must be known



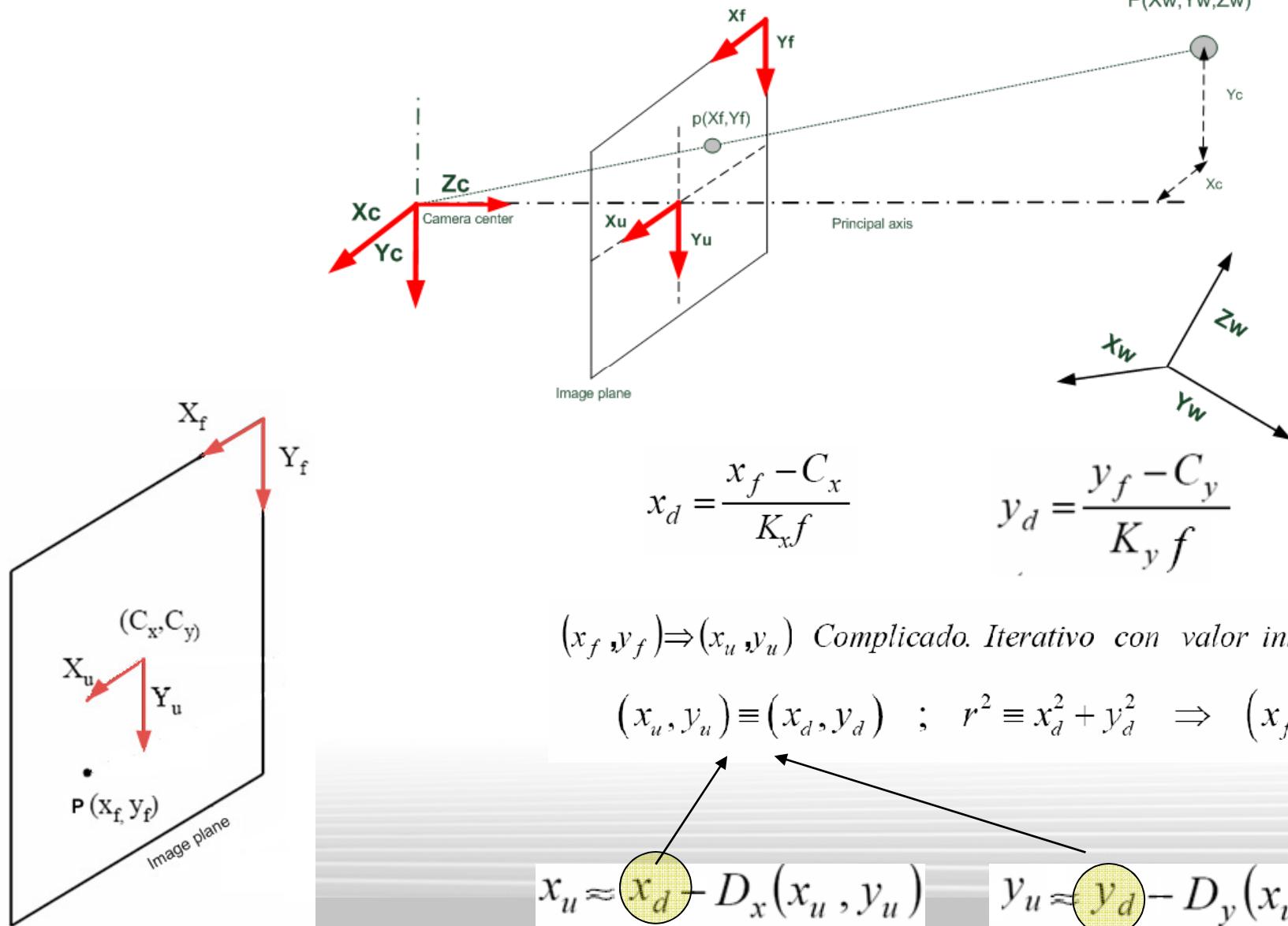
Pose Estimation

Using a external camera system



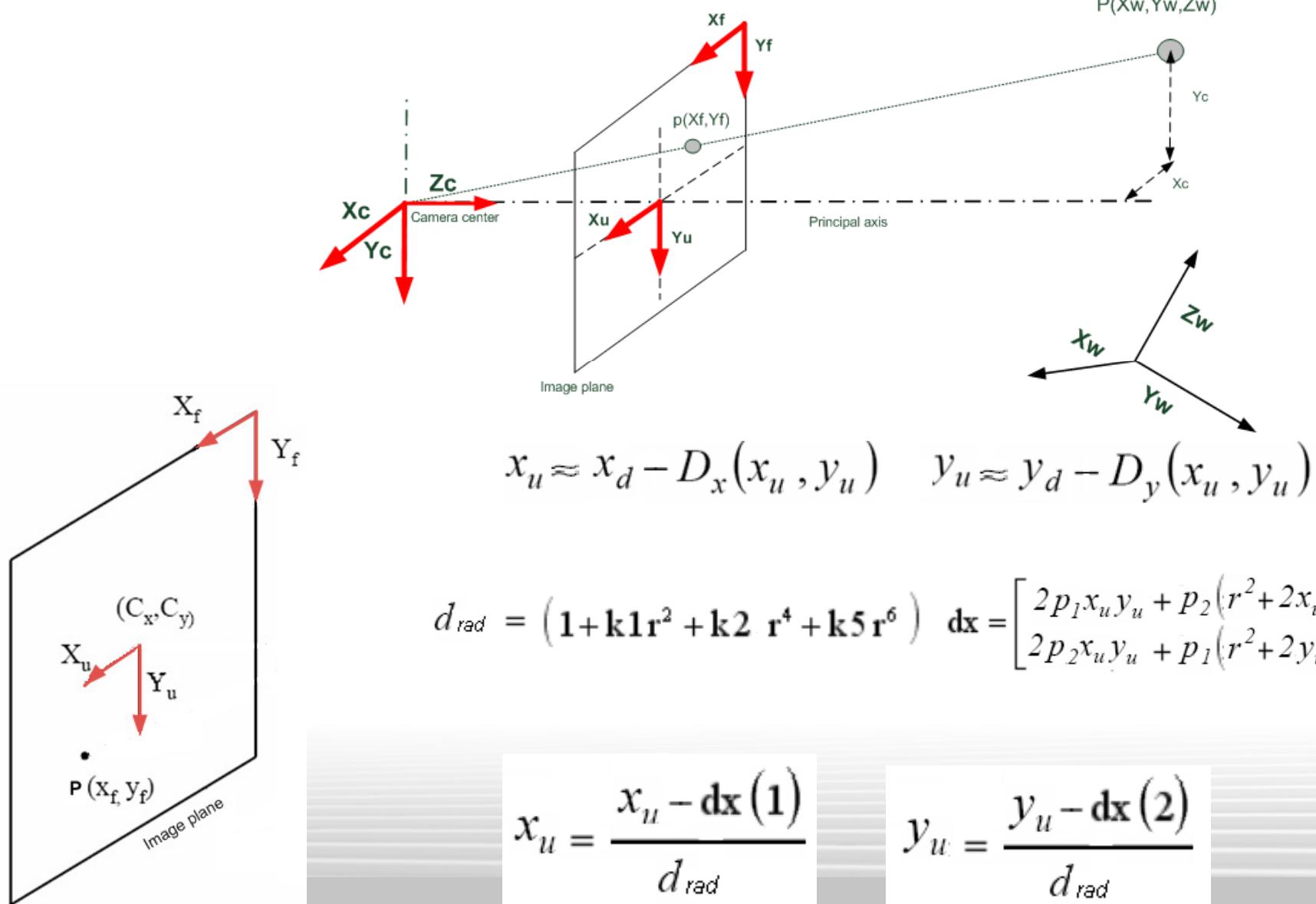


Pose Estimation



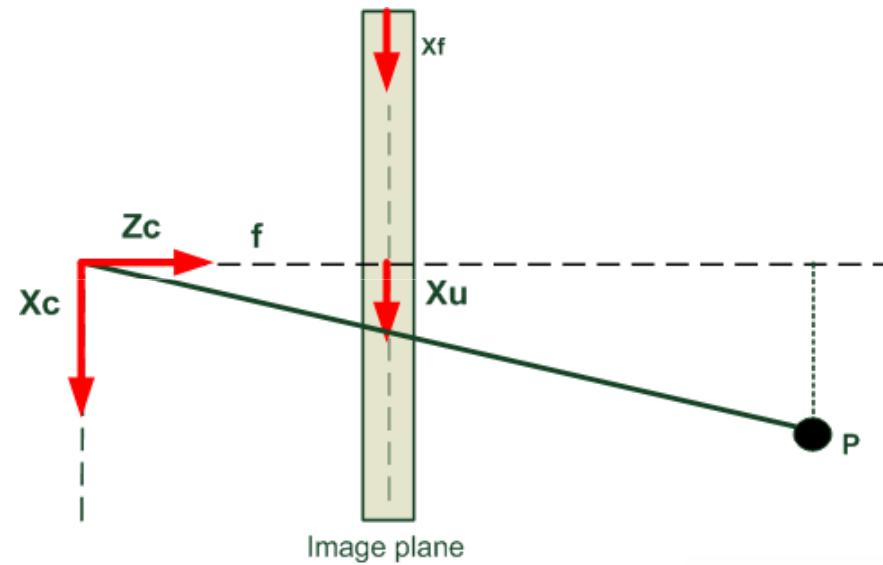


Pose Estimation





Pose Estimation



$$\frac{x_u}{f} = \frac{x_c}{z_c}$$

$$\frac{y_u}{f} = \frac{y_c}{z_c}$$

$$\begin{bmatrix} nx_u \\ ny_u \\ n \end{bmatrix} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & \frac{1}{f} & 0 \end{bmatrix} \begin{bmatrix} x_c \\ y_c \\ z_c \\ 1 \end{bmatrix}$$



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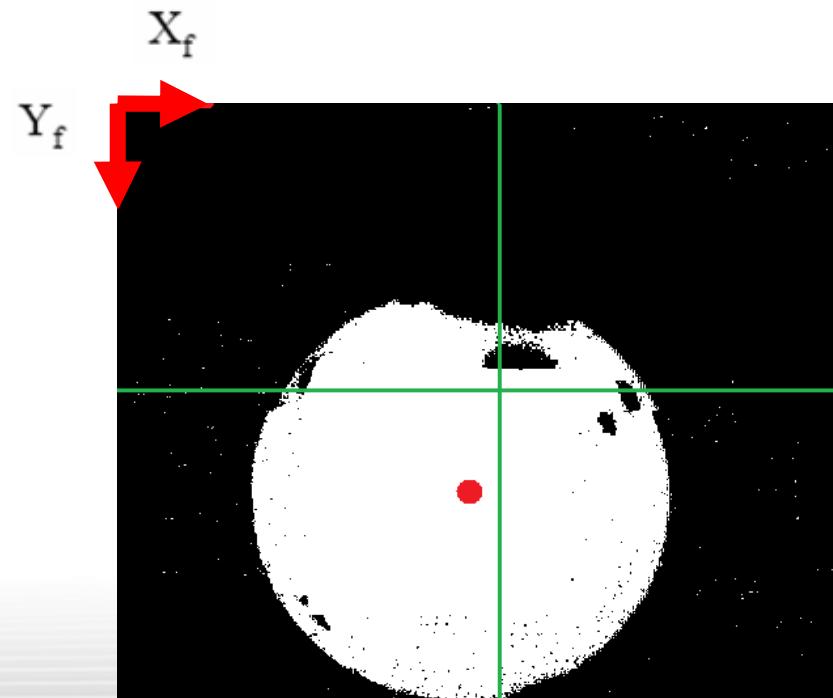
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Pose Estimation



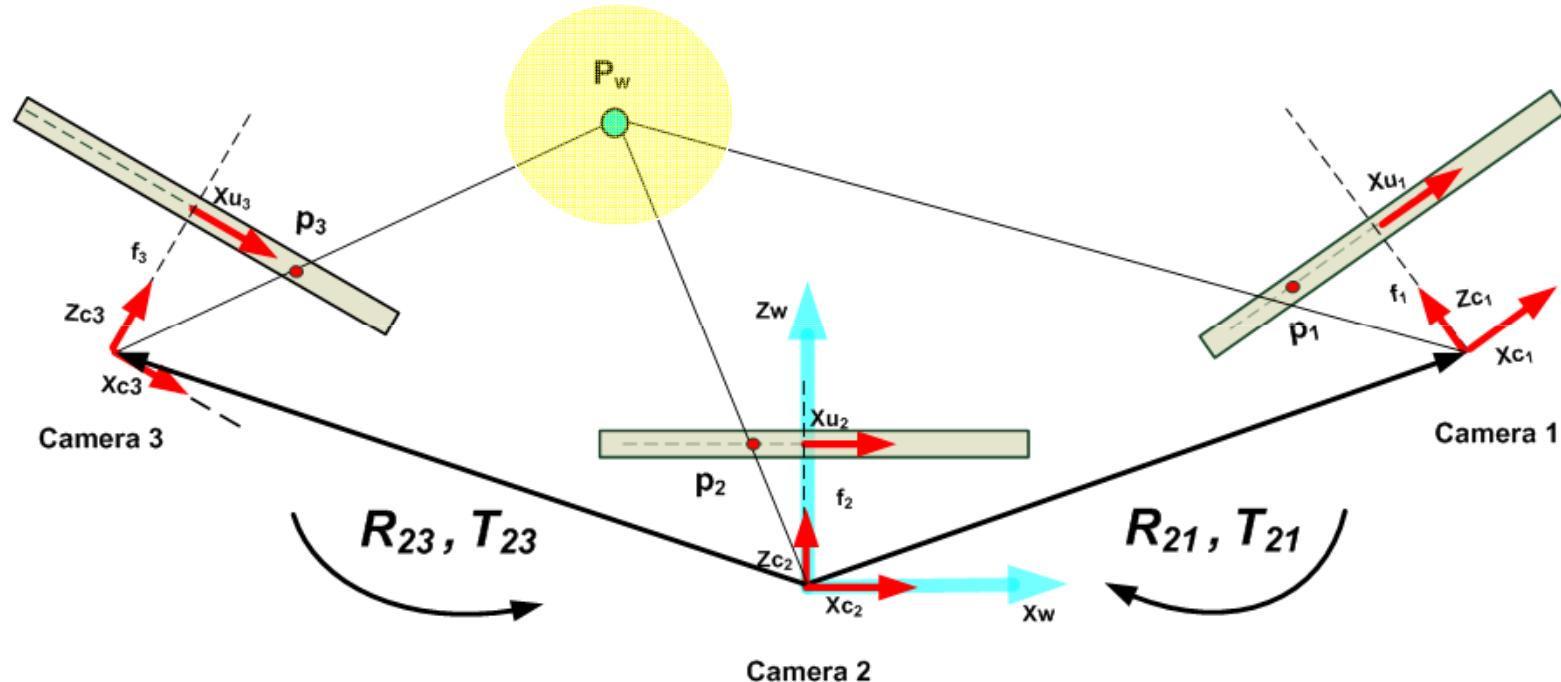
Feature extraction and tracking

- Color information
- Feature: center of gravity





Using a external camera system



$$x_{u1i} = f \frac{r_{11}^1 X_w + r_{12}^1 Y_w + r_{13}^1 Z_w + t_x^1}{r_{31}^1 X_w + r_{32}^1 Y_w + r_{33}^1 Z_w + t_z^1}$$

$$y_{u1i} = f \frac{r_{21}^1 X_w + r_{22}^1 Y_w + r_{23}^1 Z_w + t_y^1}{r_{31}^1 X_w + r_{32}^1 Y_w + r_{33}^1 Z_w + t_z^1}$$

$$A_i L_i = b_i,$$



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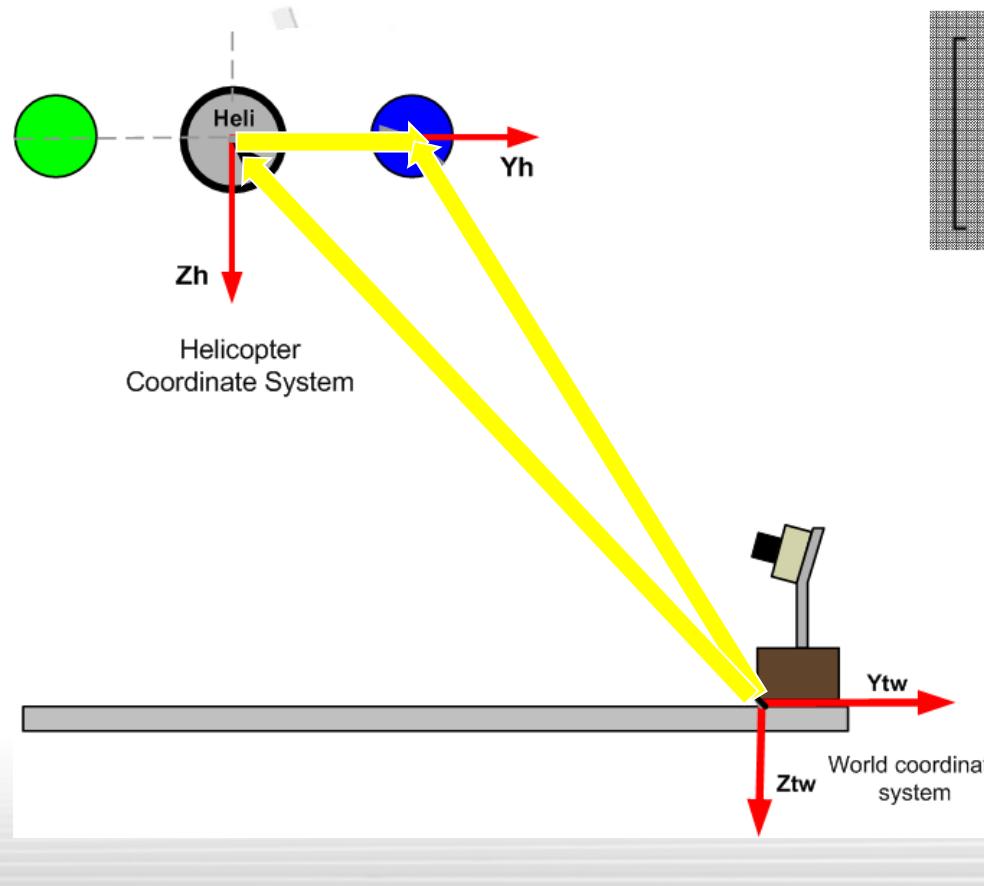
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Pose Estimation

Using a external camera system



$$\begin{bmatrix} r_{tw} \\ y_{tw} \\ z_{tw} \\ 1 \end{bmatrix} = \begin{bmatrix} \cos(\theta) & -\sin(\theta) & 0 & t_x \\ \sin(\theta) & \cos(\theta) & 0 & t_y \\ 0 & 0 & 1 & t_z \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x_h \\ y_h \\ z_h \\ 1 \end{bmatrix}$$

$$\begin{bmatrix} x_{tw}^1 \\ y_{tw}^1 \\ z_{tw}^1 - z_h^1 \\ \vdots \\ z_{tw}^4 - z_h^4 \end{bmatrix} = \begin{bmatrix} x_h^1 & -y_h^1 & 1 & 0 & 0 \\ y_h^1 & x_h^1 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ y_h^4 & x_h^4 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} A \\ B \\ C \\ D \\ E \end{bmatrix}$$



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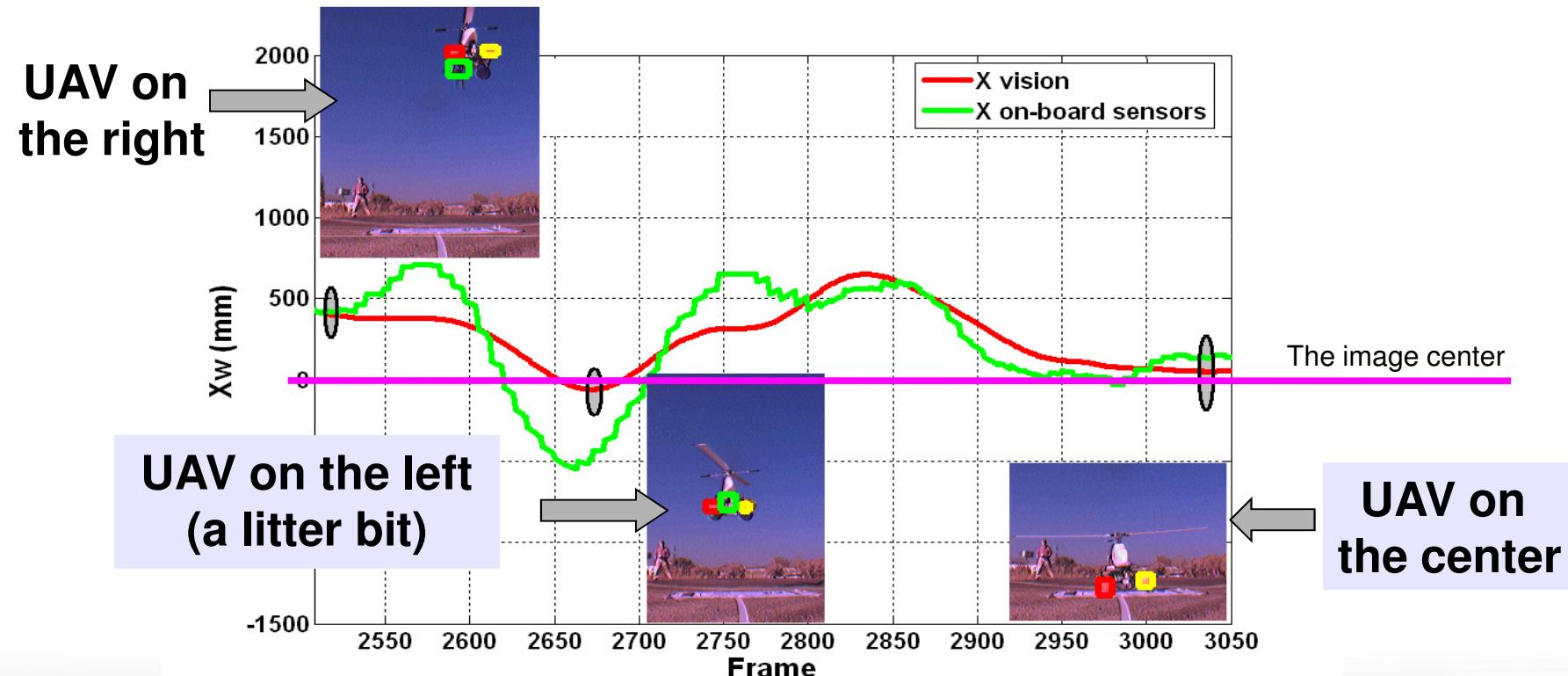
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Pose Estimation

Pose Estimation --> TRINOCULAR SYSTEM

Position estimation during a **landing** task in **manual mode** (RC)



* Visual estimation **corresponds** with **real UAV position**



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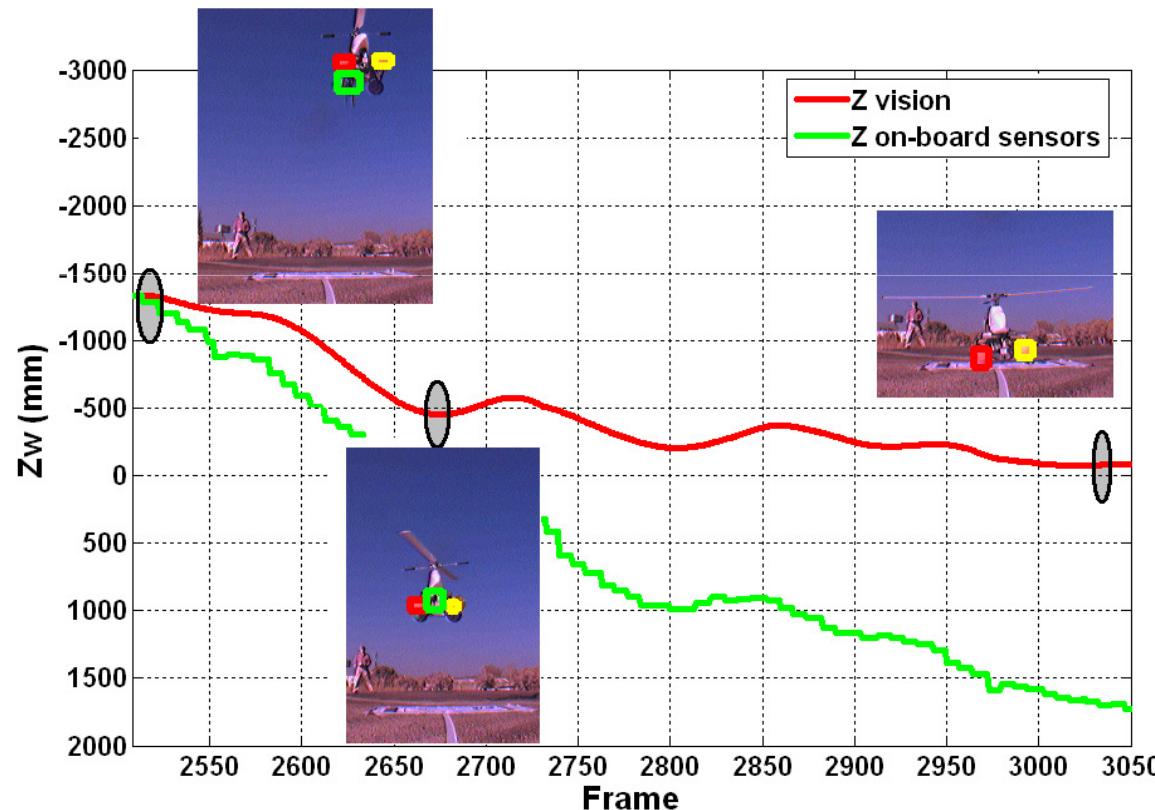
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Pose Estimation

Pose Estimation --> TRINOCULAR SYSTEM

Position estimation during a **landing** task in **manual mode** (RC)



- * Visual estimation **corresponds** with **real UAV position**
- * **Improvement** of the UAV's position estimation



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Pose Estimation

Results

**UAV's YAW ANGLE
ESTIMATION**

**USING AN EXTERNAL
TRINOCULAR SYSTEM**

Computer Vision Group
www.vision4uav.com

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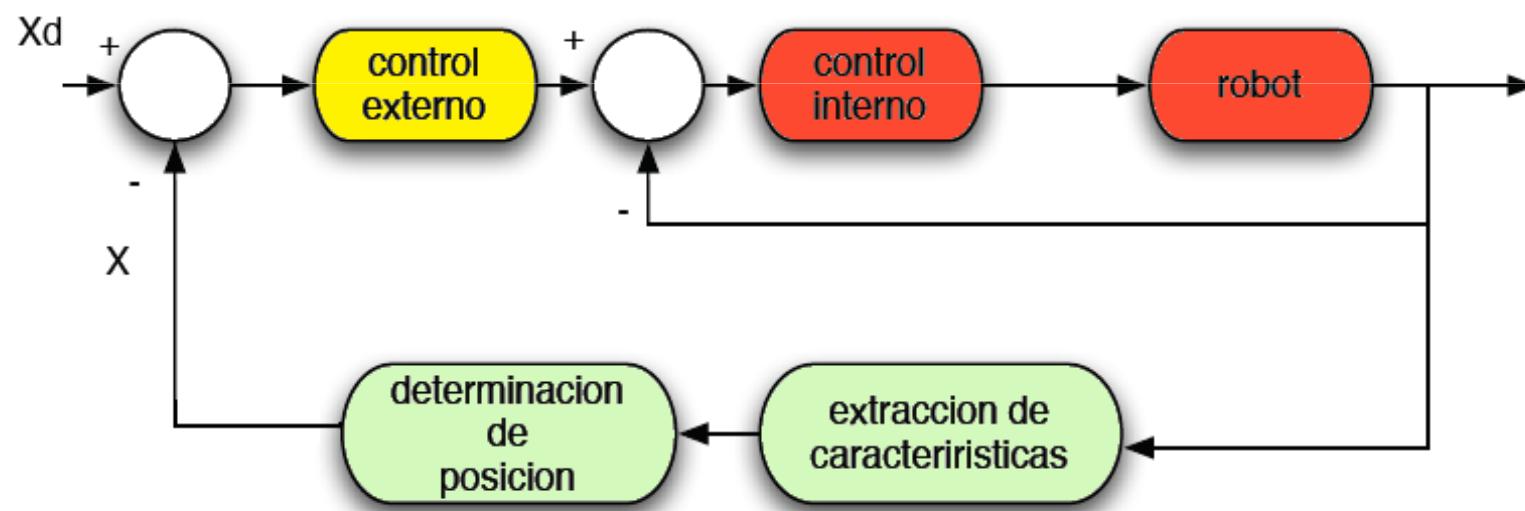
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PBVS results



Position Based Visual Servoing PBVS Results





Robo-Tenis

Monocular eye in hand, dynamic look and move strategy

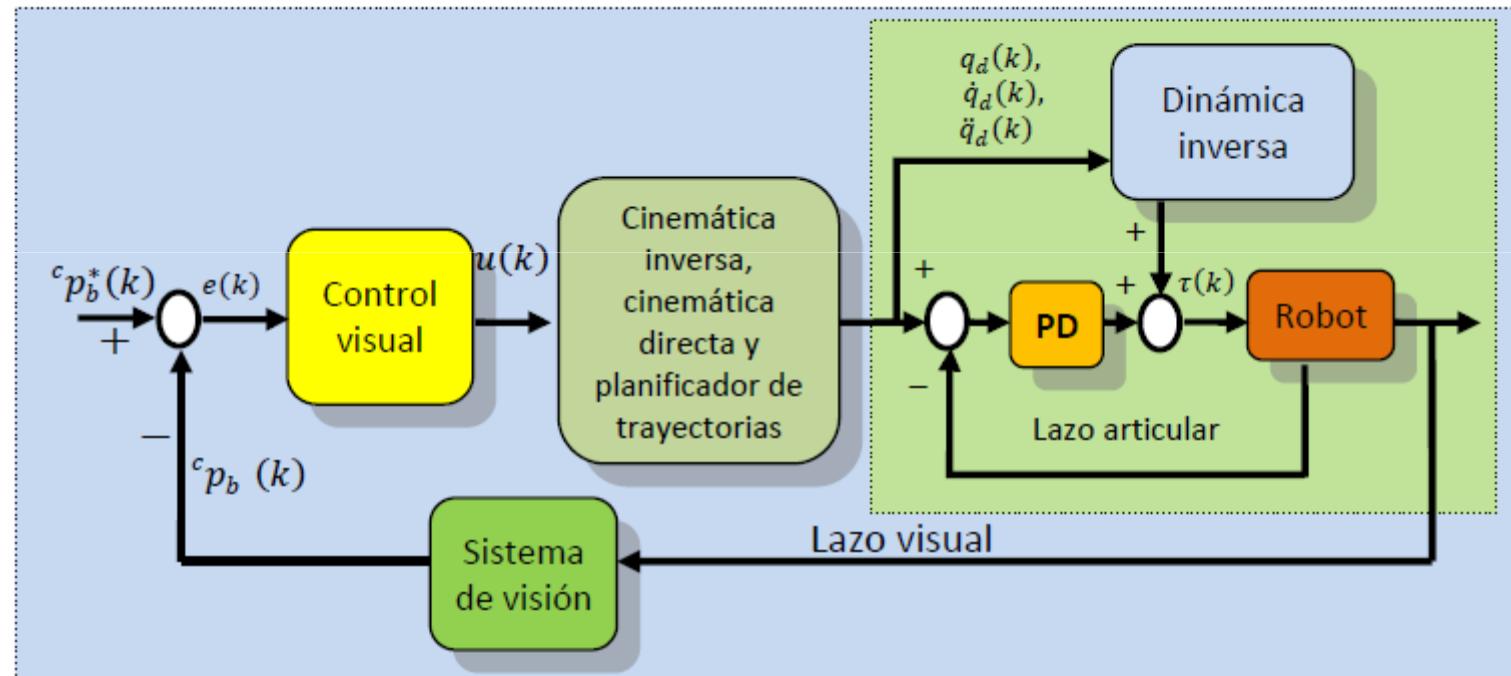


Fig. 6.2 Esquema básico de control del sistema Robotenis.

$$e(k) = {}^c p_b^*(k) - {}^c p_b(k)$$

$$e(k) = {}^c p_b^*(k) - {}^c R_w \left({}^w p_b(k) - {}^w p_c(k) \right)$$



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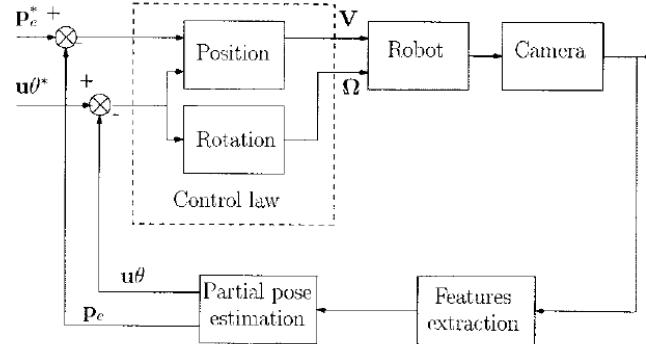
PBVS results



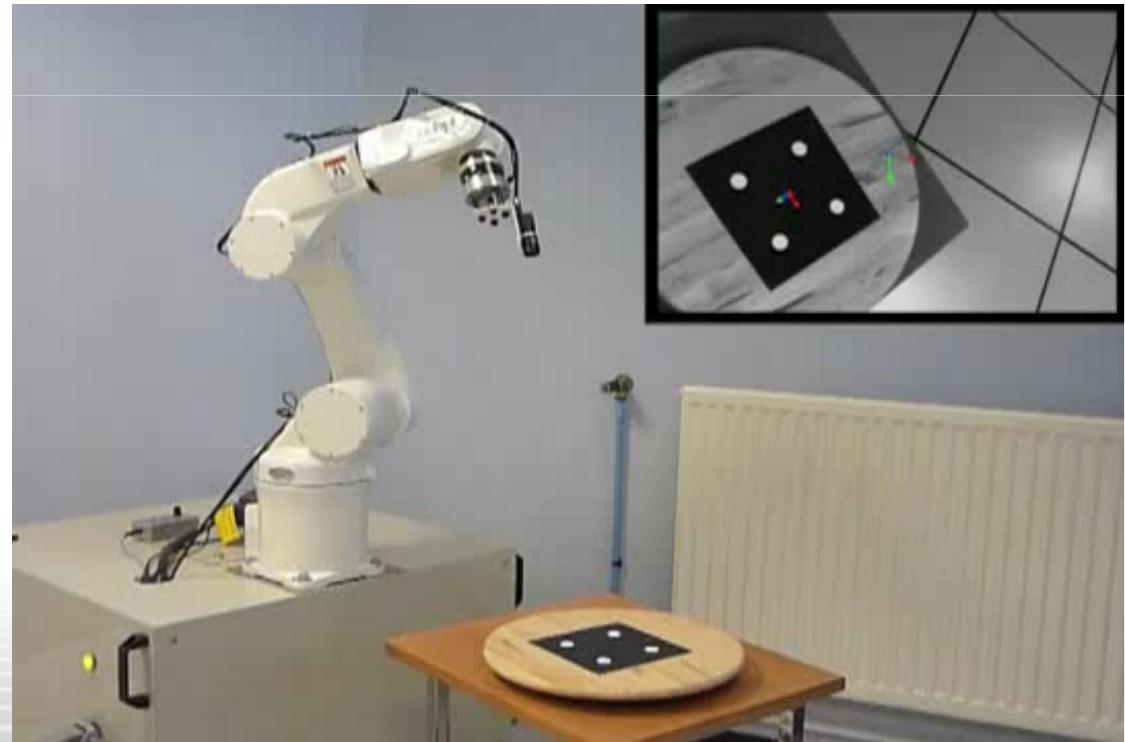
Robo-Tenis

Monocular eye in hand, dynamic look and move strategy





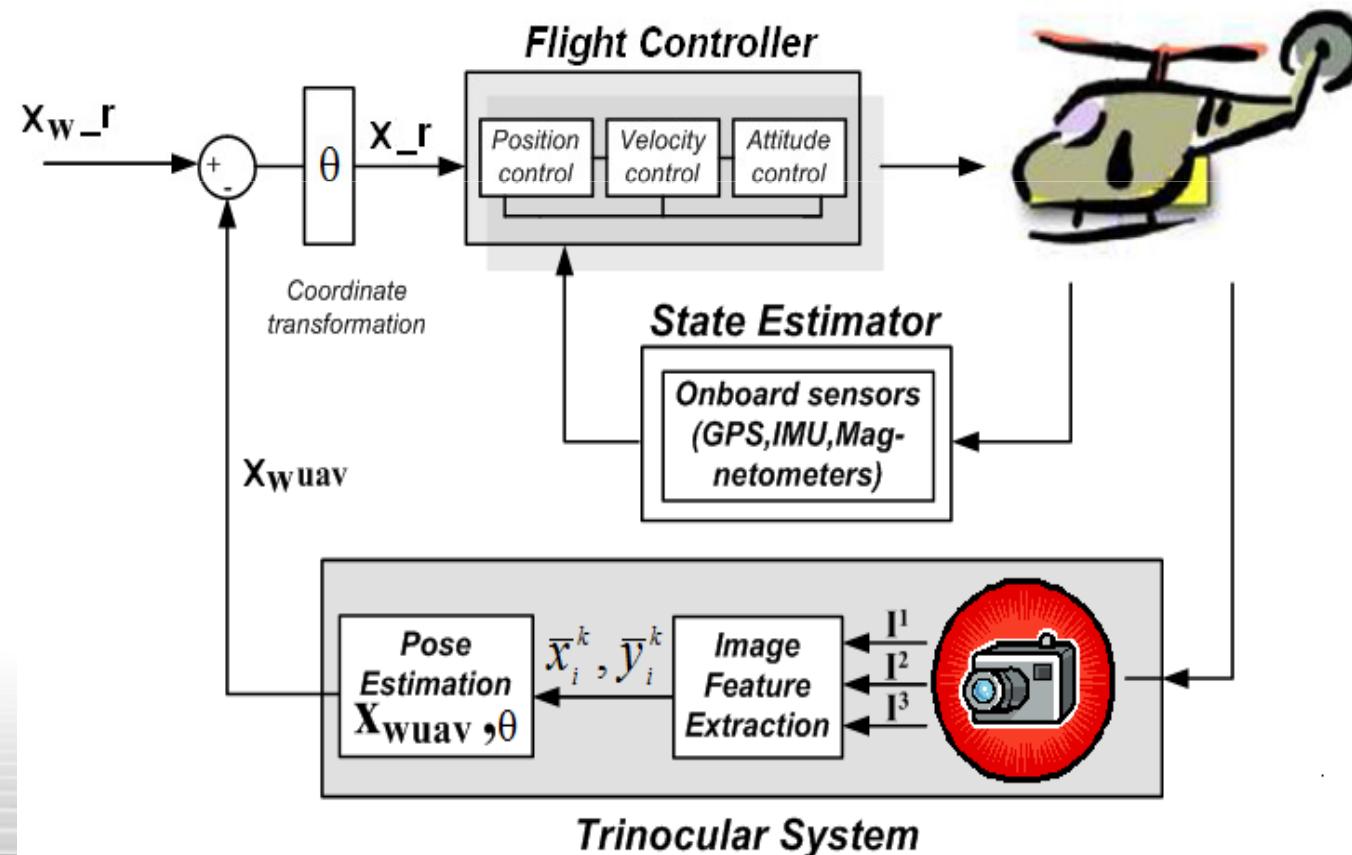
Hybrid approach





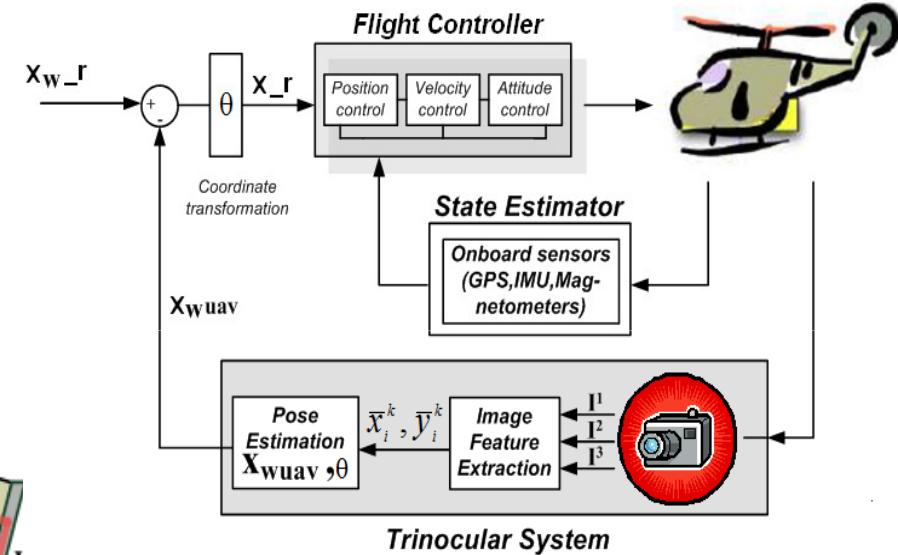
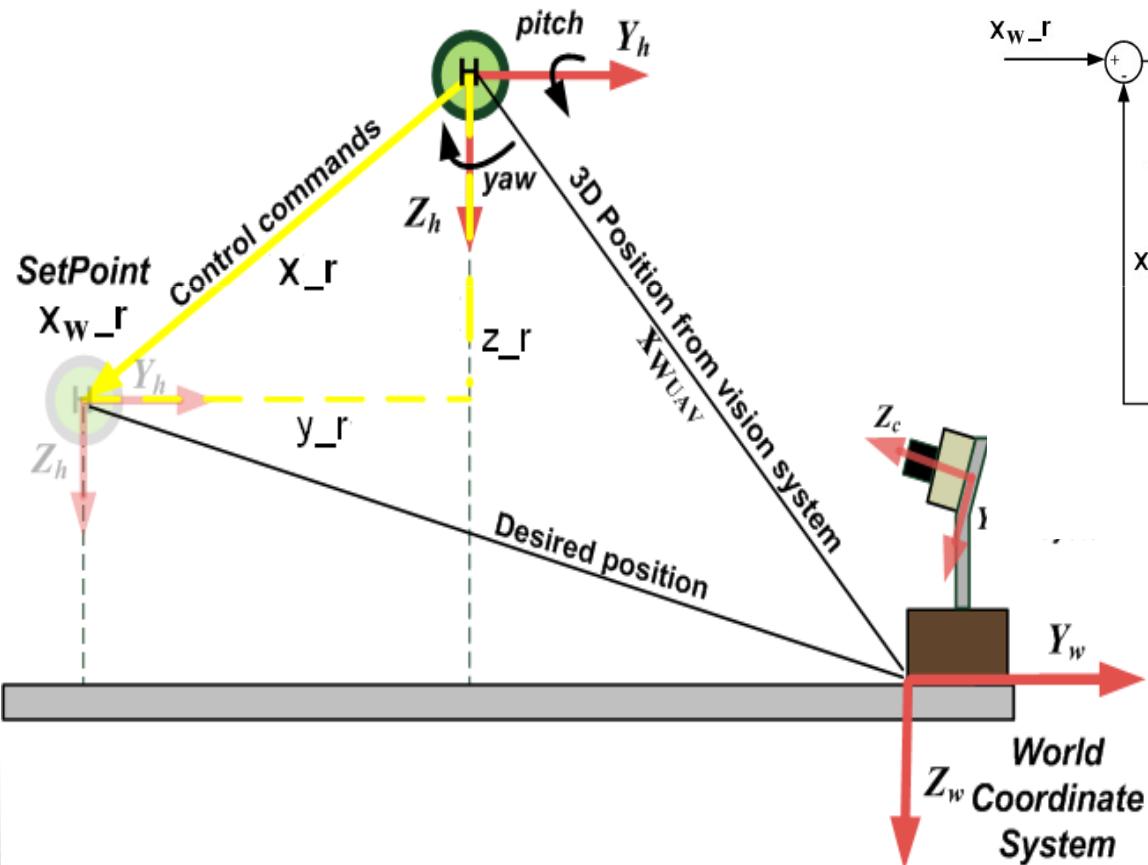
External camera system

Trinocular eye to hand, dynamic look and move strategy





External camera system





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PBVS results



External camera system

Controlling Z axis

Vision-based landing task

vision4uav.com

VISION-BASED LANDING

UAV's HEIGHT CONTROL
USING AN EXTERNAL
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PBVS results



External camera system

Vicon system: <http://www.vicon.com/>

**Precise Aggressive Maneuvers
for Autonomous Quadrotors**

**Daniel Mellinger, Nathan Michael, Vijay Kumar
GRASP Lab, University of Pennsylvania**



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Summary



Summarizing ...

-Different visual control strategies depending on the error function:

- PBVS
- IBVS
- Hybrid



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Summary



Summarizing ...

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- IBVS
- Hybrid

-Position based visual servoing **depends on the pose estimation algorithm**



Summarizing ...

-Different visual control strategies depending on the error function:

- PBVS
- IBVS
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-Position based visual servoing **depends on the pose estimation algorithm**

-Pose estimation algorithms (**depending on the number of cameras**):

- Monocular: require **additional information** to solve the depth
- Multi-camera systems: by **triangulation**, problem speed.



Summarizing ...

-Different visual control strategies depending on the error function:

- PBVS
- IBVS
- Hybrid

-Position based visual servoing **depends on the pose estimation algorithm**

-Pose estimation algorithms (**depending on the number of cameras**):

- Monocular: require **additional information** to solve the depth
 - Multi-camera systems: by **triangulation**, problem speed.
-
- Depending on the references: PBVS with **velocity commands or position commands**.



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References



References

Slides based on:

Thesis:

Pablo Lizardo Pari

Control Visual Basado en Características de un Sistema Articulado. Estimación del Jacobiano de la Imagen Utilizando Múltiples Vistas

Thesis

Alberto Traslocheros Michel

Desarrollo, Implementación y Evaluación de Estrategias de Control Servo Visual para Robots Paralelos: Aplicación a la Plataforma Robotenis

Thesis

Luis Mejías Alvarez

Control Visual de un Vehículo Aéreo Autónomo Usando Detección y Seguimiento de Características en Espacios Exteriores



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¡Thanks!

The Flying Machine Arena Quadrocopter Ball Juggling



Eidgenössische Technische Hochschule Zürich
Swiss Federal Institute of Technology Zurich

Pose Estimation and Position Based Visual Servoing
Carol Martínez
May 2011