



# COS 495 - Lecture 2

## Autonomous Robot Navigation

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Semester: Fall 2011

# Navigation and Control

1. Control Architectures
2. Navigation Example
3. Basic Tools for AUV Navigation

# Control Architectures

- Today, most robots control systems have a mixture of planning and behavior-based control strategies.
- To implement these strategies, a *control architecture* is used.
- Control architectures should be:
  - Modular
  - Localized

# Control Architectures

## Desired Characteristics

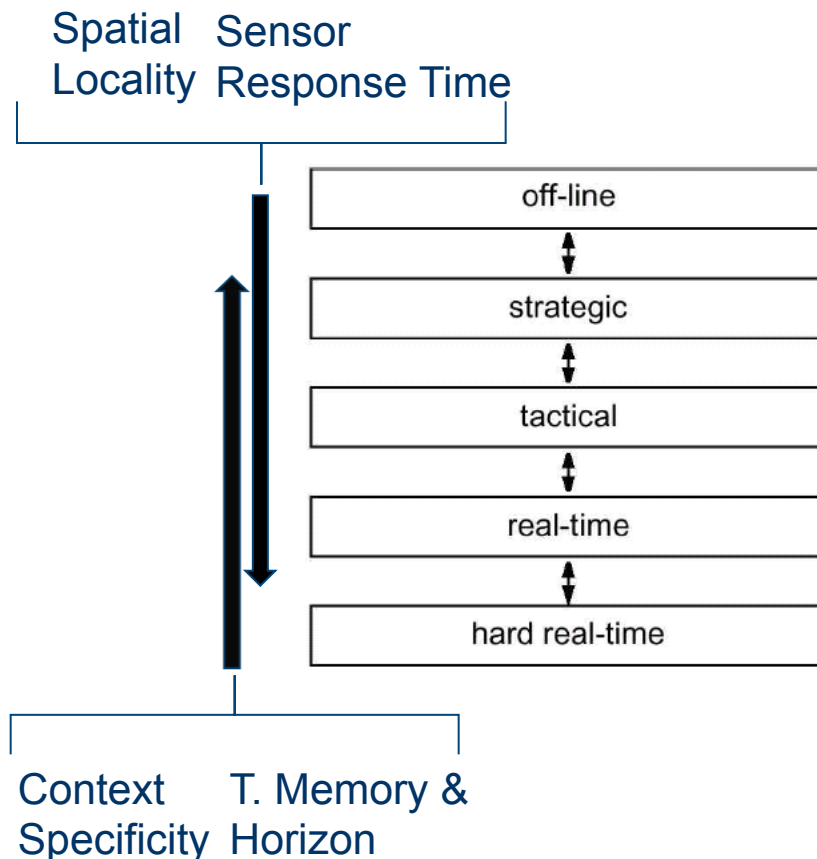
- Code Modularity
  - Allows programmers to interchange environment types sensors, path planners, propulsion, etc.
- Localization
  - Embed specific navigation functions within modules to allow different levels of control (e.g. from task planning to wheel velocity control)

# Control Architectures Decomposition

- Decomposition allows us to modularize our control system based on different axes:
  1. Temporal Decomposition
    - Facilitates varying degrees of real-time processes
  2. Control Decomposition
    - Defines how modules should interact: serial or parallel?

# Control Architectures

## Temporal Decomposition

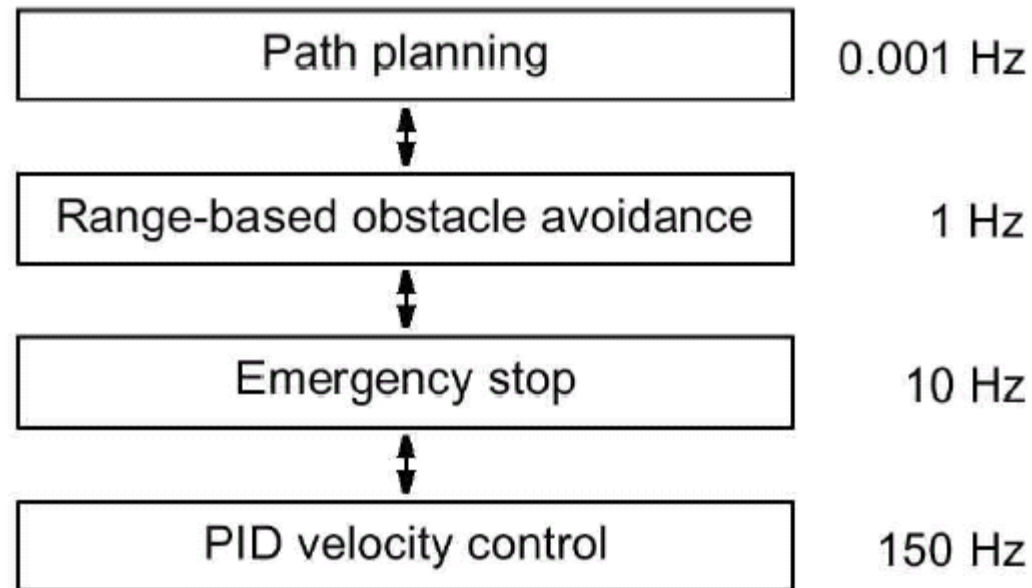


- Factors affecting temporal decomposition:
  - Sensor response time
  - Temporal memory and horizon
  - Spatial Locality
  - Context Specificity

# Control Architectures

## Temporal Decomposition

- Example

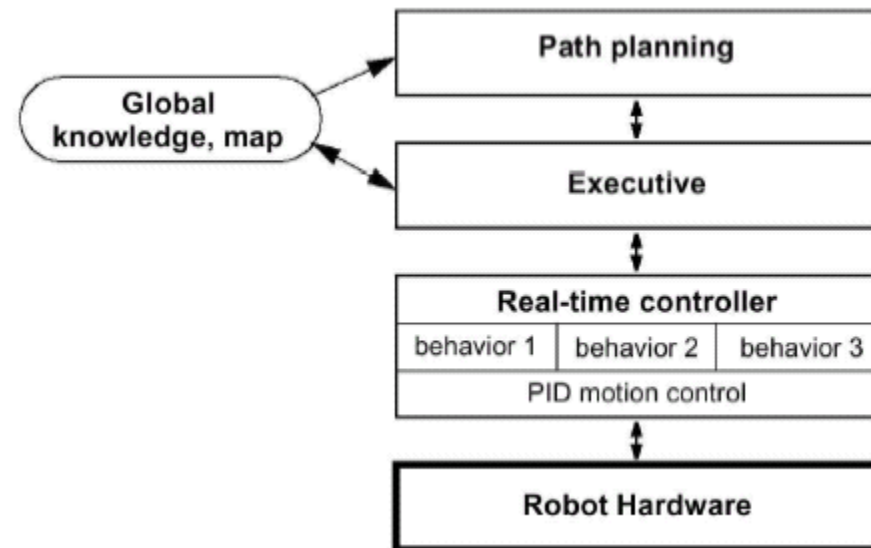


# Control Architectures

## Tiered Architectures

- A general tiered architecture for *episodic* planning
- Role of the Executive is:

- Switch behaviors
- Monitor failures
- Call the planner



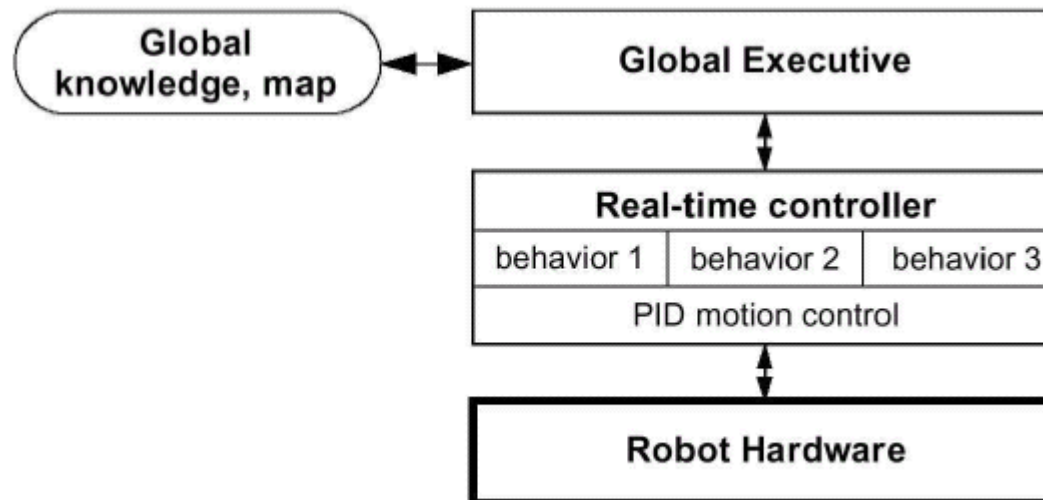
- Planning only when required (e.g. blockage)



# Control Architectures

## Tiered Architectures

- A tiered architecture for *integrated* planning

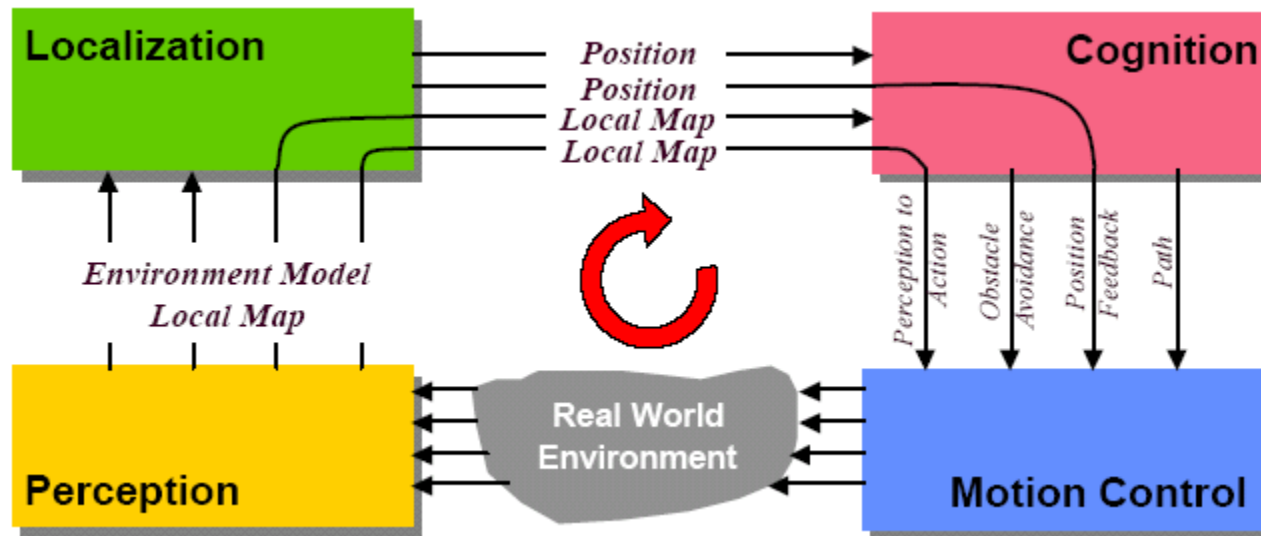


- Planning is fast and is embedded as a behavior.

# Control Architectures

## Control Decomposition

- An example of a control decomposition using a mixture of serial and parallel approaches.



# Navigation and Control

1. Control Architectures
2. **Navigation Example**
  1. Motion Modeling
  2. Estimation and Control
  3. Experiments
3. Basic Tools for AUV Navigation

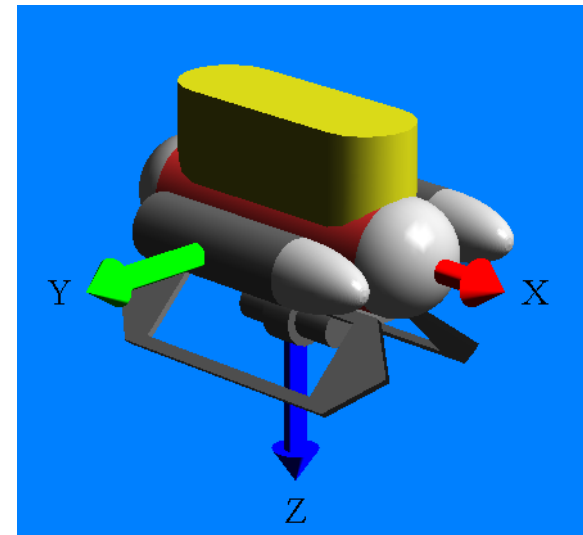
# Navigation Example

- Autonomous Control
  - Goal is to enable autonomous trajectory tracking capabilities.
  - Given individual ROVs have autonomous control, multi-vehicle control research will be facilitated.



# Equations of Motion

- 6 degrees of freedom (DOF):
- State vectors:  
 body-fixed velocity vector:  
 earth-fixed pos. vector:



DOF	Surge	Sway	Heave	Roll	Pitch	Yaw
Velocities	$u$	$v$	$w$	$p$	$q$	$r$
Position & Attitude	$x$	$y$	$z$	$\phi$	$\theta$	$\psi$
Forces & Moments	$X$	$Y$	$Z$	$K$	$M$	$N$

# Equations of Motion

- The 6-DOF nonlinear dynamic equations of motion can be expressed as:

$$M\dot{\boldsymbol{v}} + C(\boldsymbol{v})\boldsymbol{v} + D(\boldsymbol{v})\boldsymbol{v} + \boldsymbol{g}(\boldsymbol{\eta}) = \boldsymbol{\tau}$$

$$\dot{\boldsymbol{\eta}} = J(\boldsymbol{\eta})\boldsymbol{v}$$

where:

inertia matrix:

$$M = M_{RB} + M_A$$

Coriolis & centripetal matrix:

$$C(\boldsymbol{v}) = C_{RB}(\boldsymbol{v}) + C_A(\boldsymbol{v})$$

hydrodynamic damping:

$$D(\boldsymbol{v})$$

restoring forces:

$$\boldsymbol{g}(\boldsymbol{\eta})$$

propulsion forces:

$$\boldsymbol{\tau}$$

# Equations of Motion

- Initial Assumptions:

- The ROV will usually move with low velocity when on mission
- Almost three planes of symmetry;
- Vehicle is assumed to be performing non-coupled motions.

- Horizontal Plane:

$$m_{11}\dot{u} = -m_{22}vr + X_u u + X_{u|u}|u|u| + X$$

$$m_{22}\dot{v} = m_{11}ur + Y_v v + Y_{v|v}|v|v|,$$

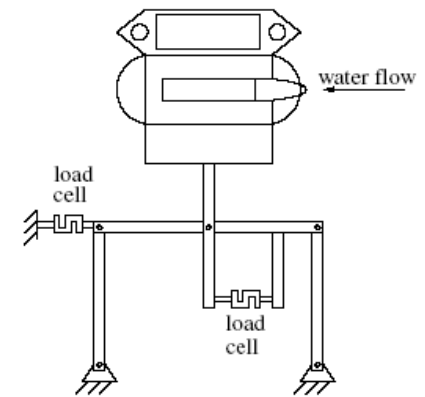
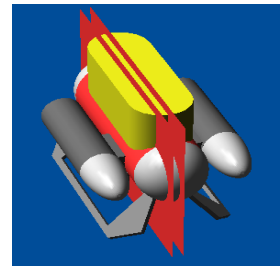
$$I\dot{r} = N_r r + N_{r|r}|r|r| + N,$$

- Vertical Plan:

$$m_{33}\dot{w} = Z_w w + Z_{w|w}|w|w| + Z$$

# Theory vs. Experiment

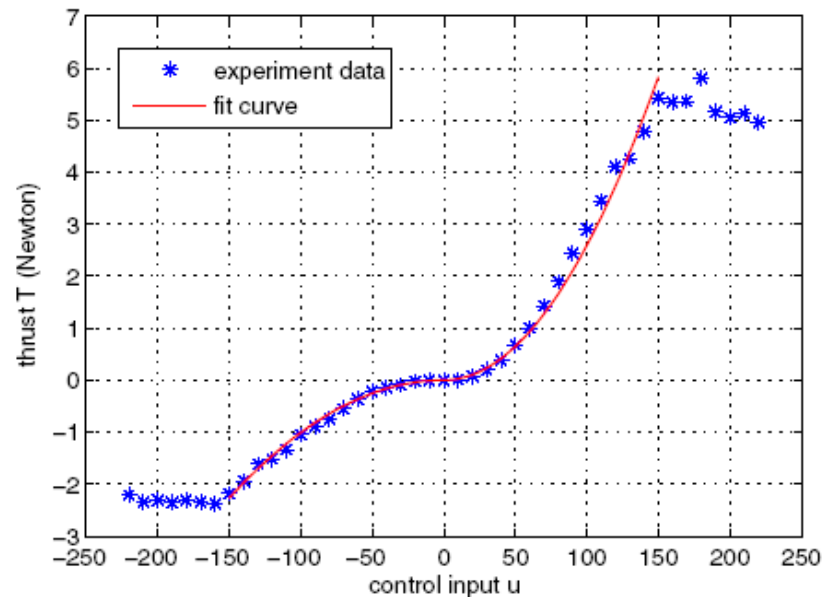
- Coefficients for the dynamic model are pre-calculated using strip theory;
- A series of tests are carried out to validate the hydrodynamic coefficients, including
  - Propeller mapping
  - Added mass coefficients
  - Damping coefficients





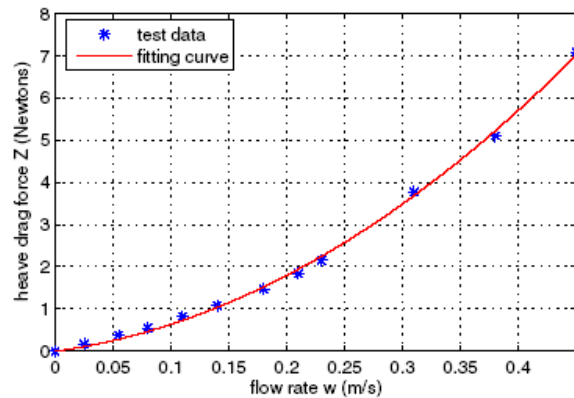
# Propeller Thrust Mapping

- The forward thrust can be represented as:

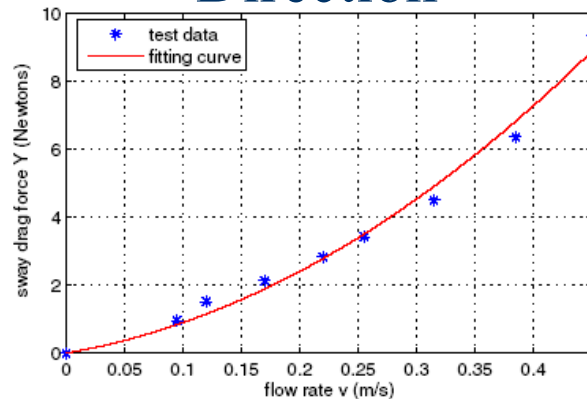


# Direct Drag Forces

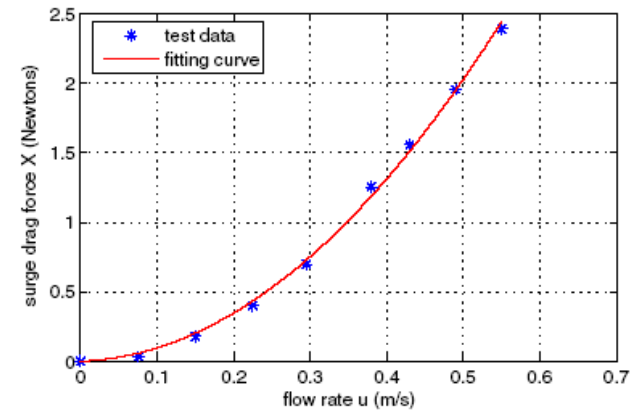
## Drag in Heave Direction



## Drag in Sway Direction



## Drag in Surge Direction



# Estimation & Control

- Sensor Overview

- VideoRay Compass

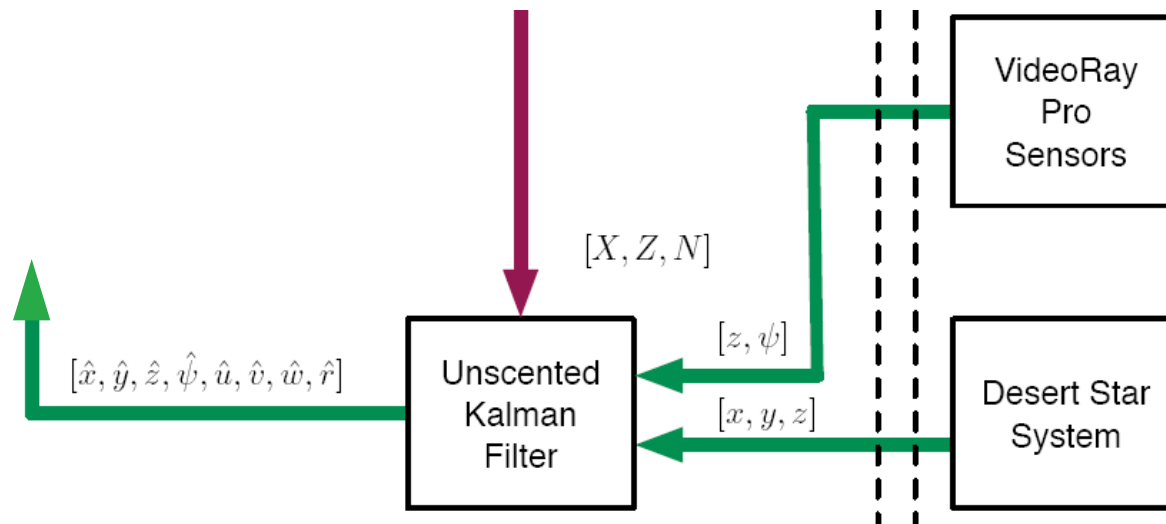
- VideoRay Depth Sensor

- Desert Star Acoustic Positioning System



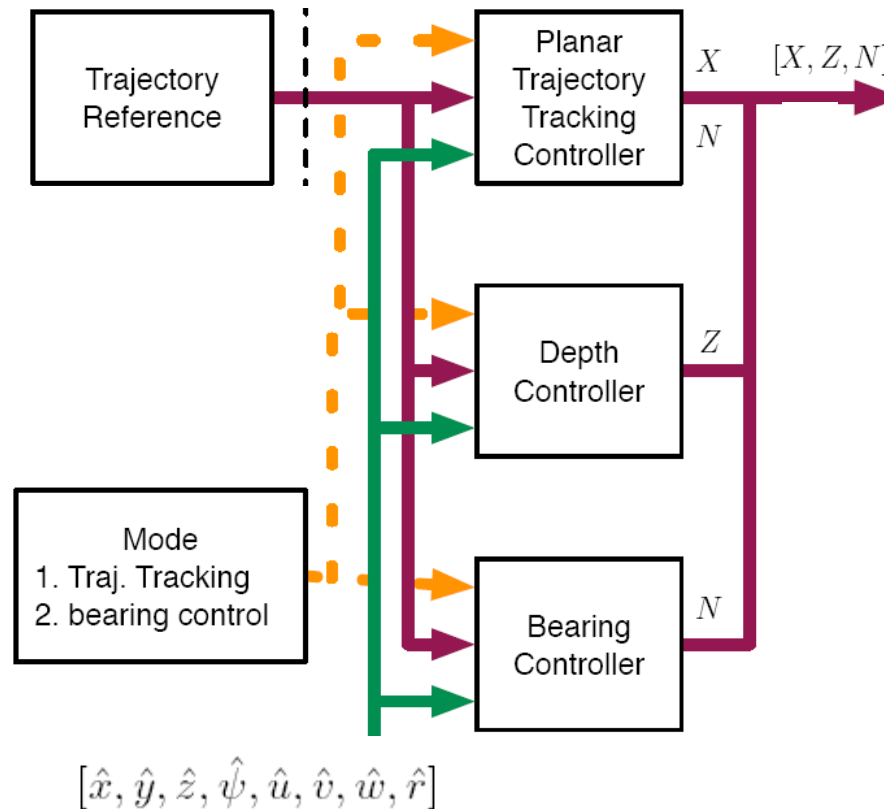
# Estimation & Control

- State Estimation
  - We fuse several sensor measurements using an Unscented Kalman Filter:

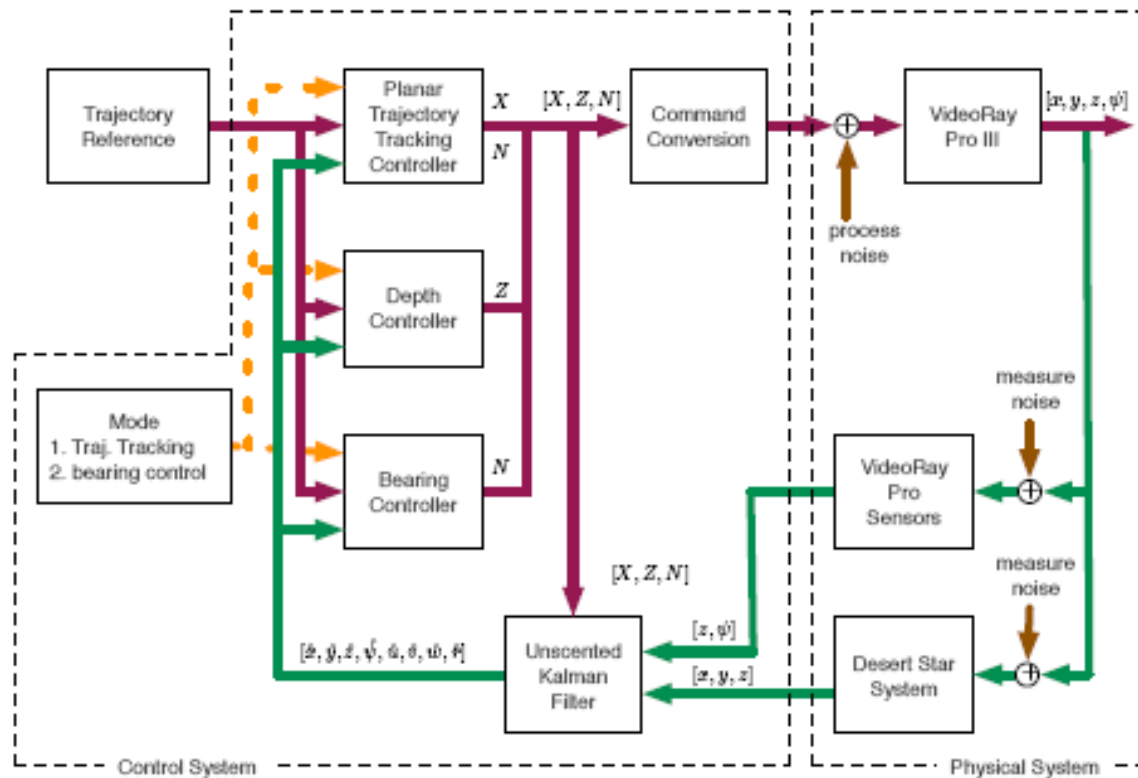


# Estimation & Control

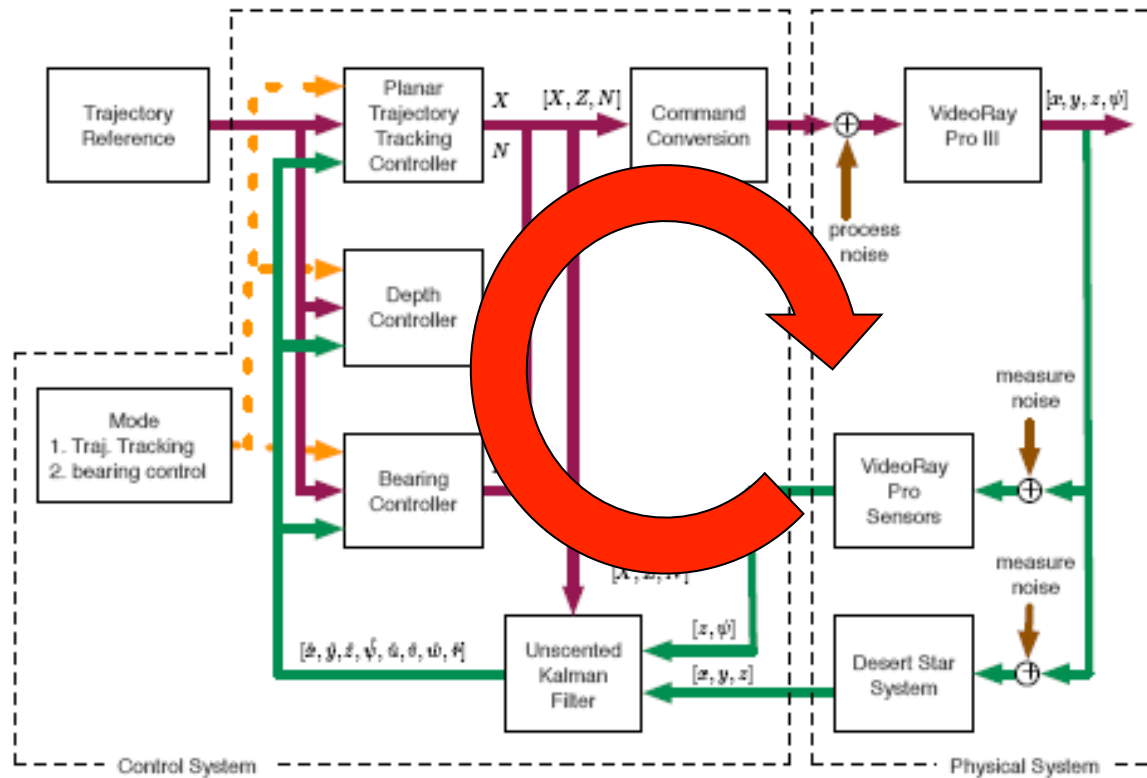
- Trajectory Tracking
  - We use three different controllers:



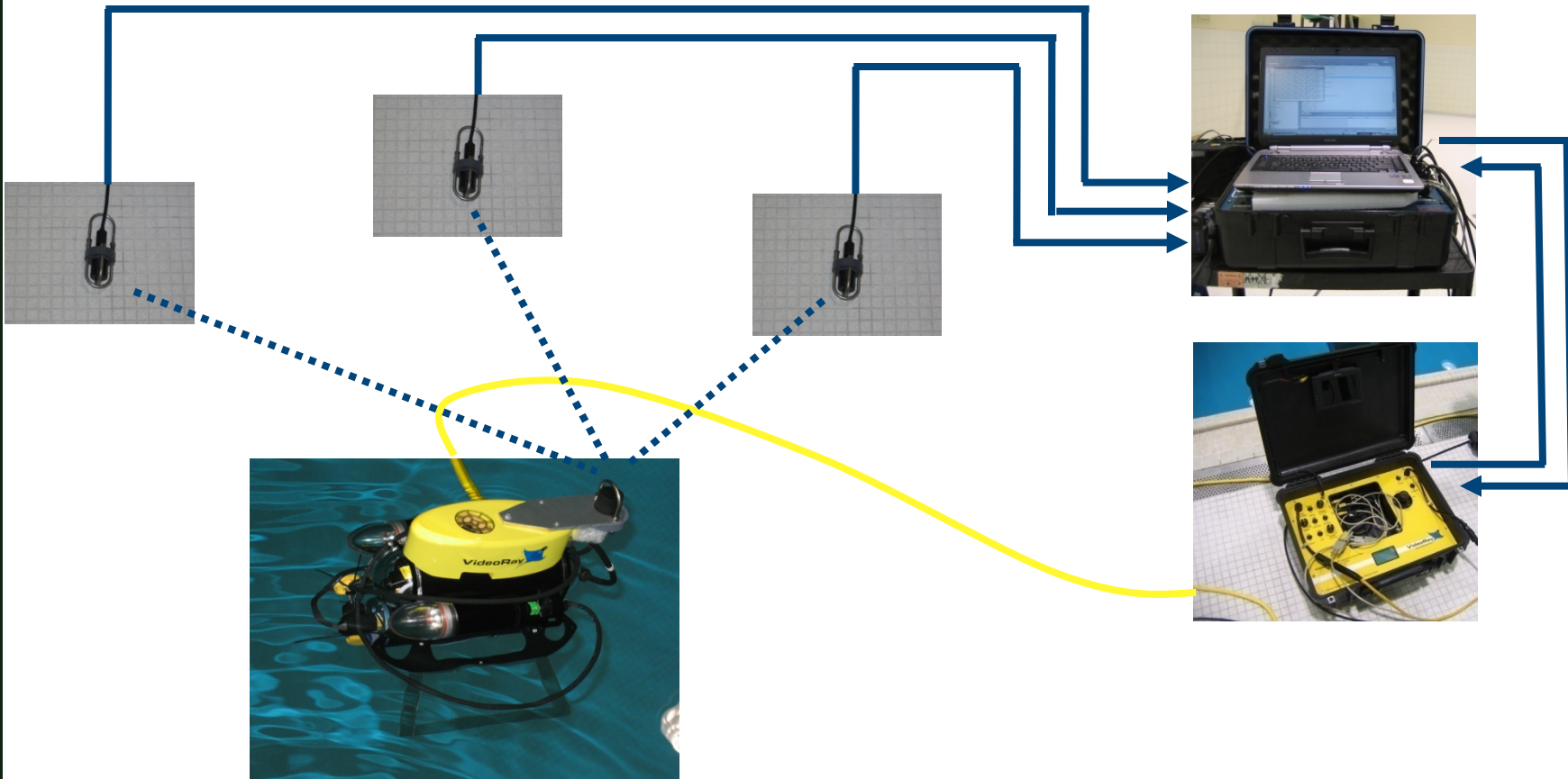
# Estimation & Control



# Estimation & Control



# Autonomous Control



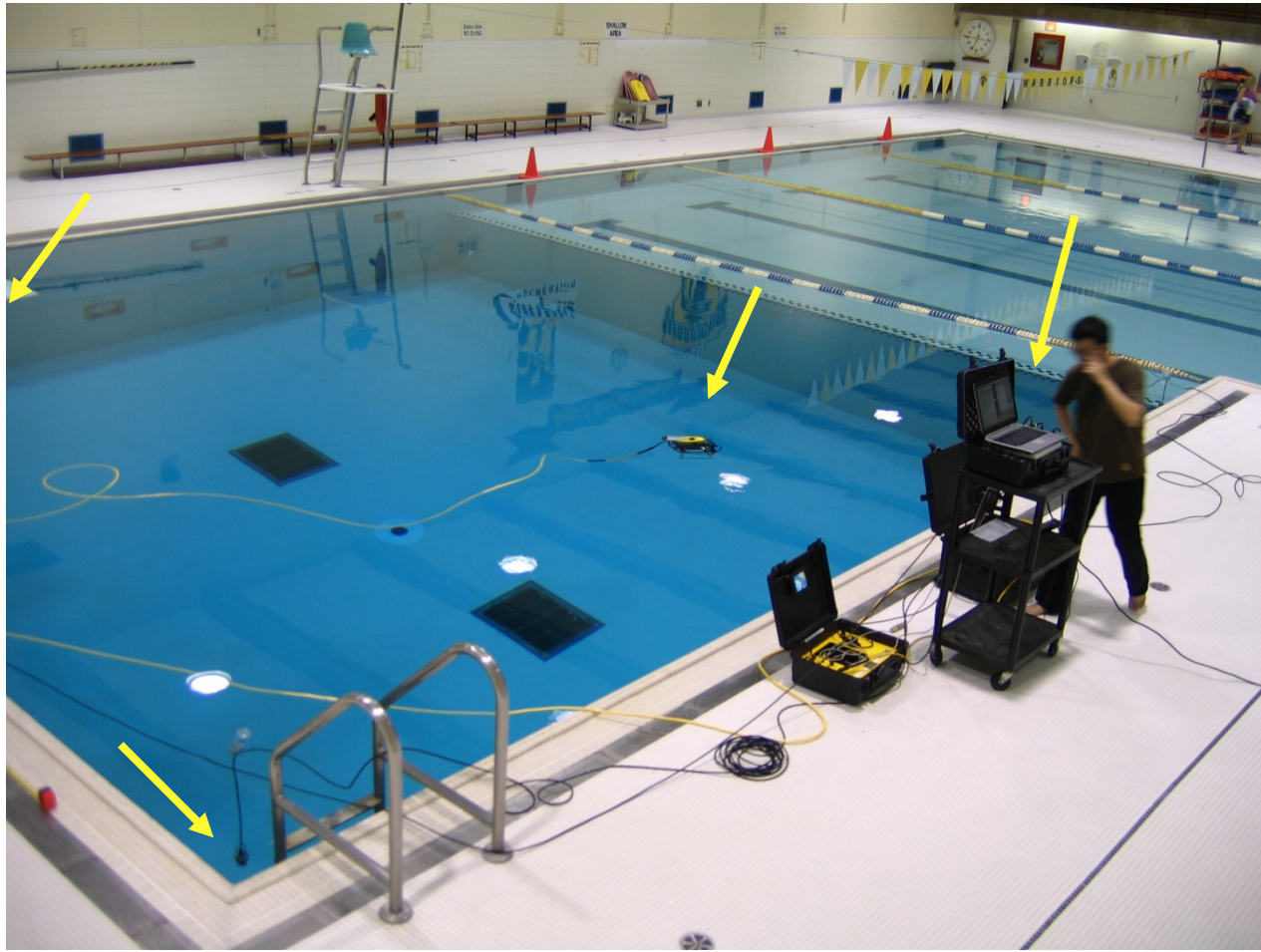


# Autonomous Control

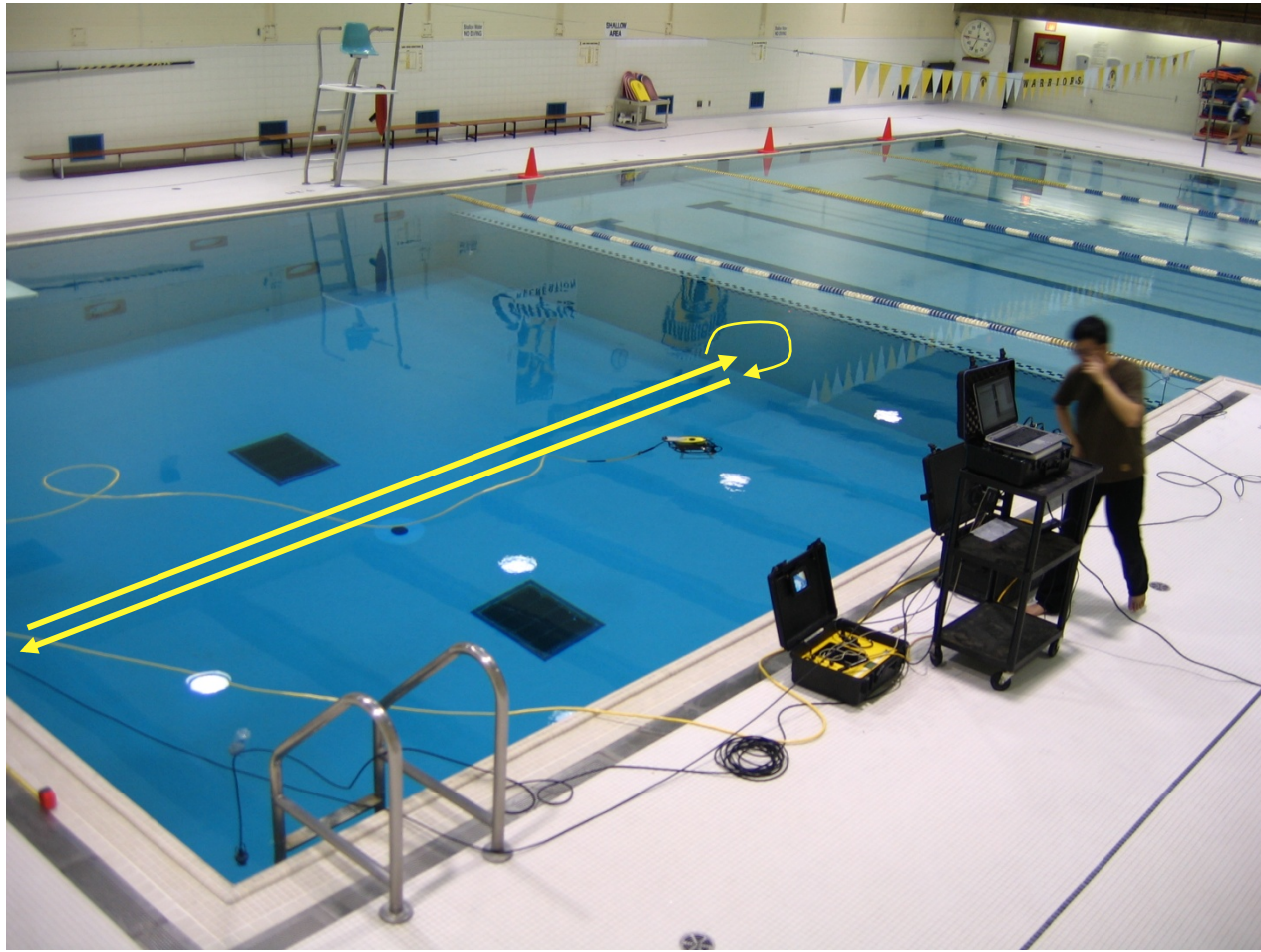


- Hardware Modifications
  - Added transceiver
  - Added bouyancy
  - Shifted weight
  - Extended feet

# Autonomous Control



# Autonomous Control

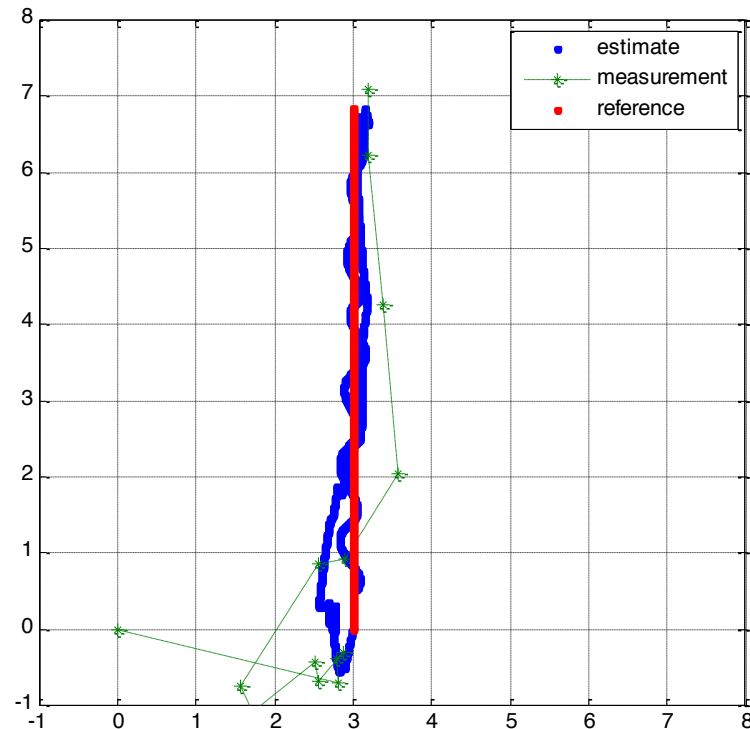


# Autonomous Control



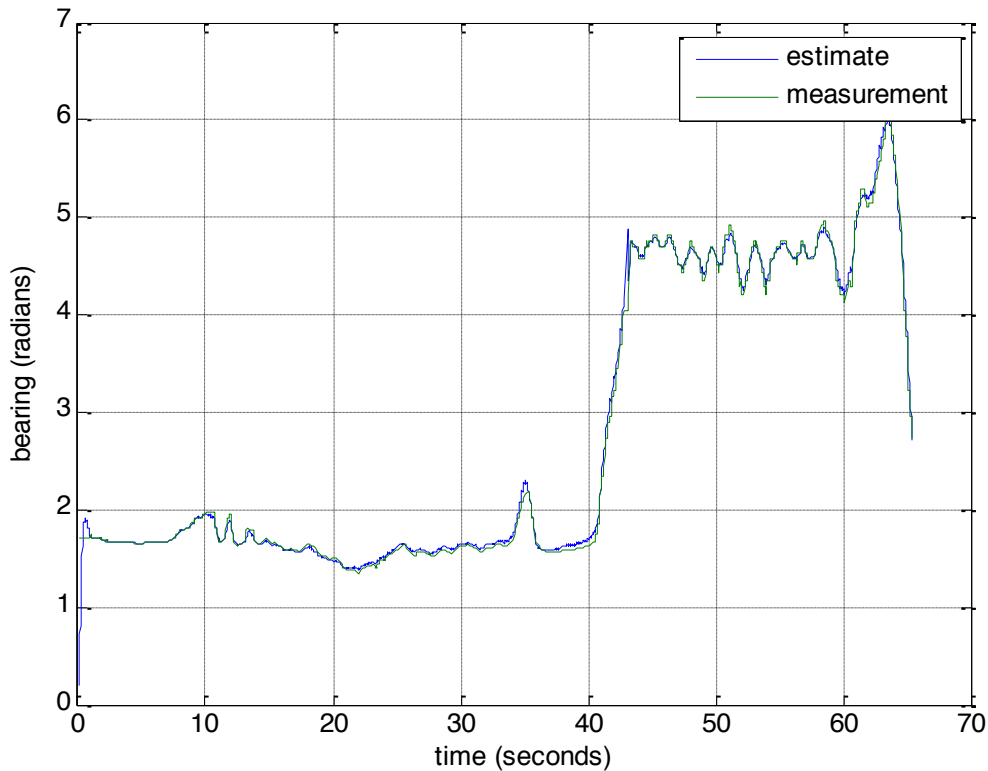
# Swimming Pool Experiments

- Sample run: x, y state estimates



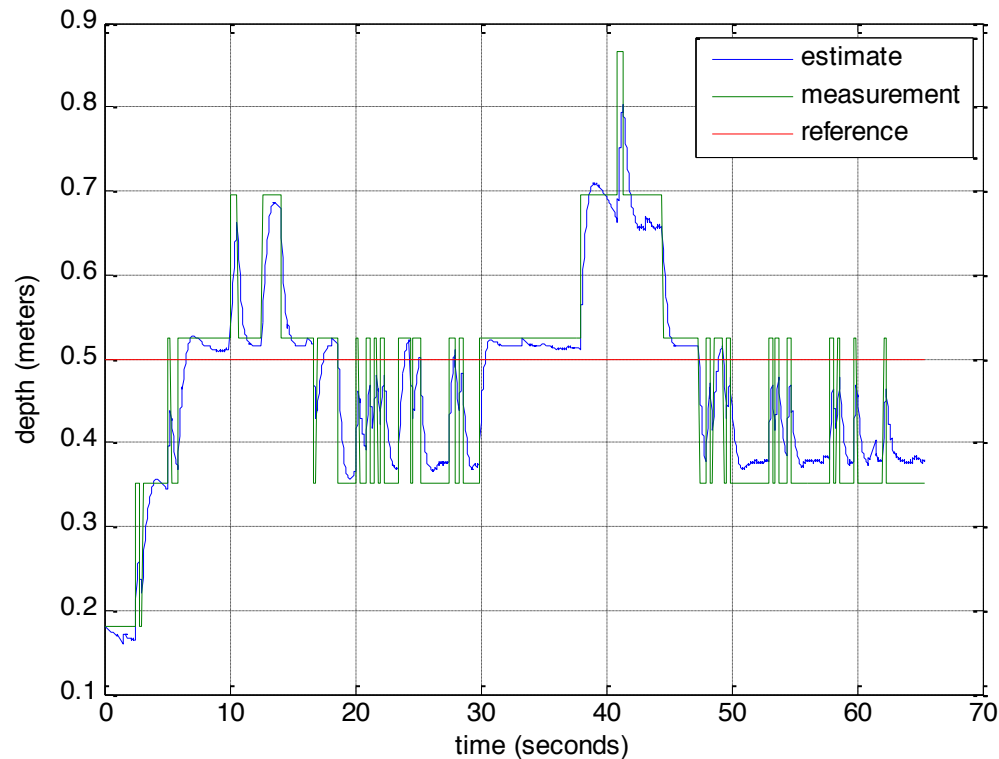
# Swimming Pool Experiments

- Sample run: bearing state estimates



# Swimming Pool Experiments

- Sample run: depth state estimates





# Swimming Pool Experiments

Trial #	Mean of x (m)	Standard Deviation (m)
1	3.186	0.431
2	2.906	0.495
3	3.095	0.129
4	3.040	0.137
5	3.192	0.179
6	2.890	0.265
7	2.966	0.154

*[W. Wang et al., 2006]*

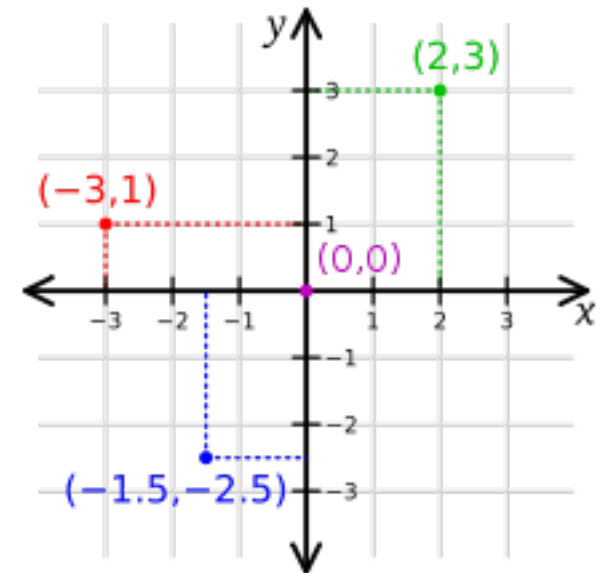


# Navigation and Control

1. Control Architectures
2. Navigation Example 2
3. Basic Tools for AUV Navigation
  1. **Coordinate Frames**
  2. Motion Modeling
  3. P Control

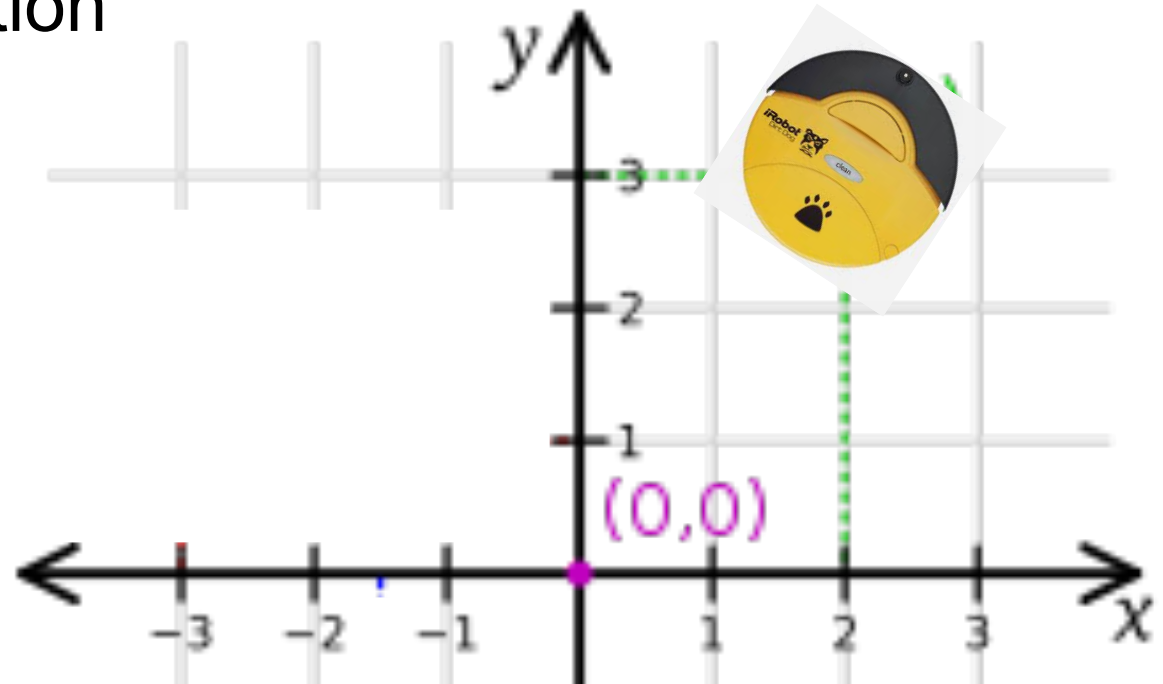
# Cartesian Coordinates

- Describes unique position of points in a plane with respect to the axis
- For each dimension there is 1 axis
- Coordinates are measured in “units” in the direction parallel to the axis
- The origin is fixed to the plane



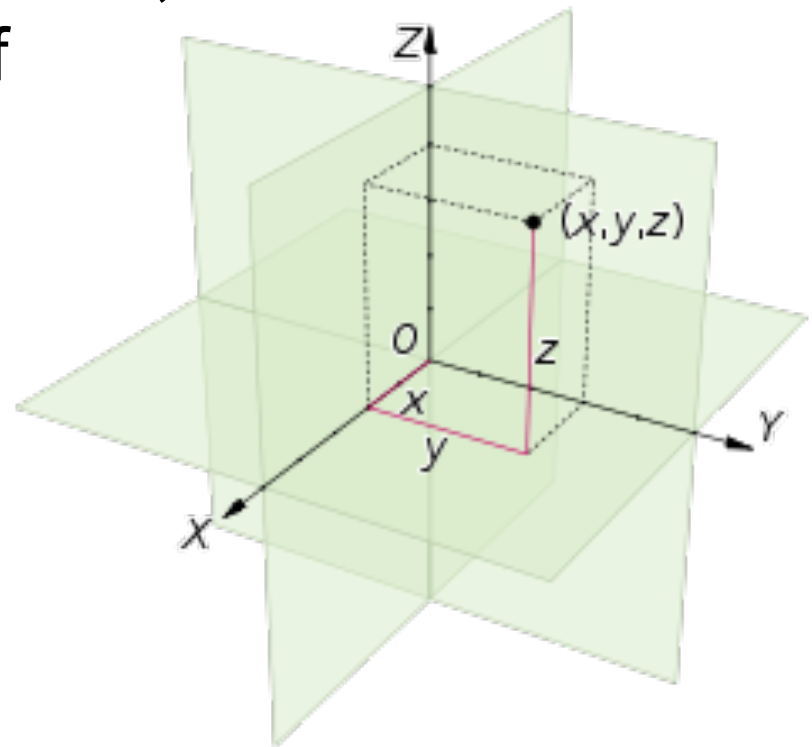
# Cartesian Coordinates

- One can use cartesian coordinates to describe a robot's position



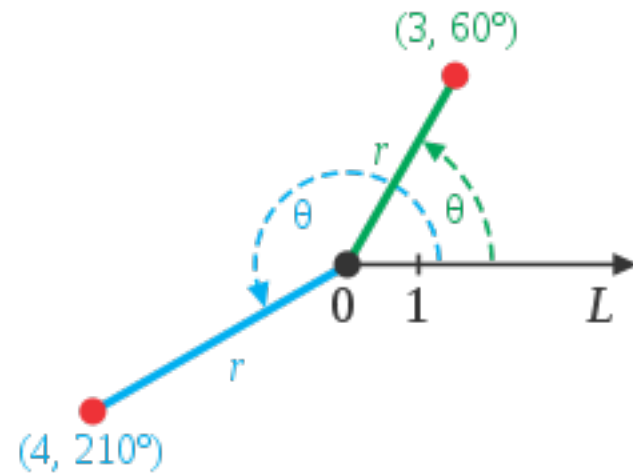
# Cartesian Coordinates

- For our underwater robots, we need 3 degrees of freedom to express position



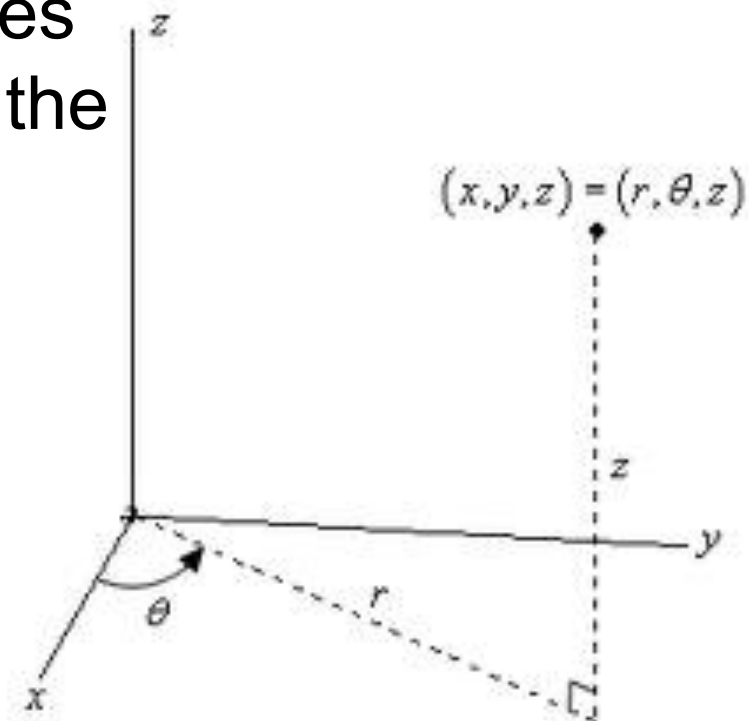
# Polar Coordinates

- In polar coordinates, we specify points on a 2D plane using the length of a radius arm and an angle



# Cylindrical Coordinates

- For specifying point locations in 3D, cylindrical coordinates can be used by specifying the length of a radius arm, an angle, and a height.

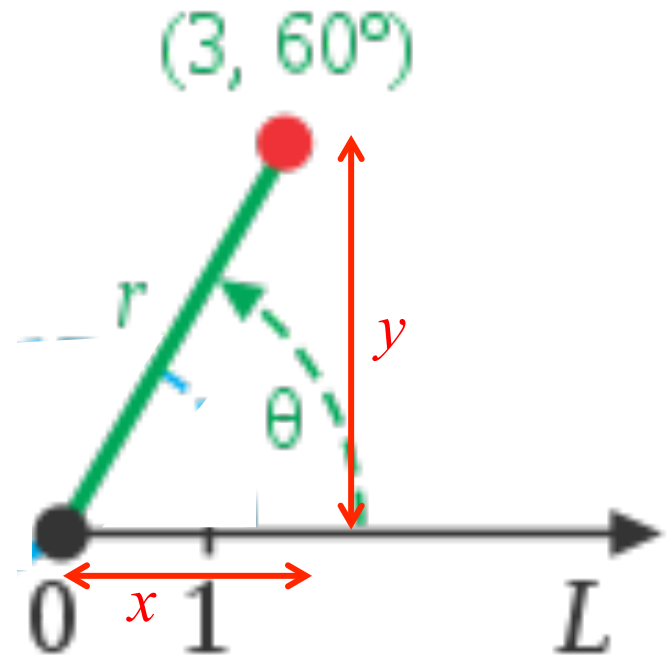


# Polar to Cartesian

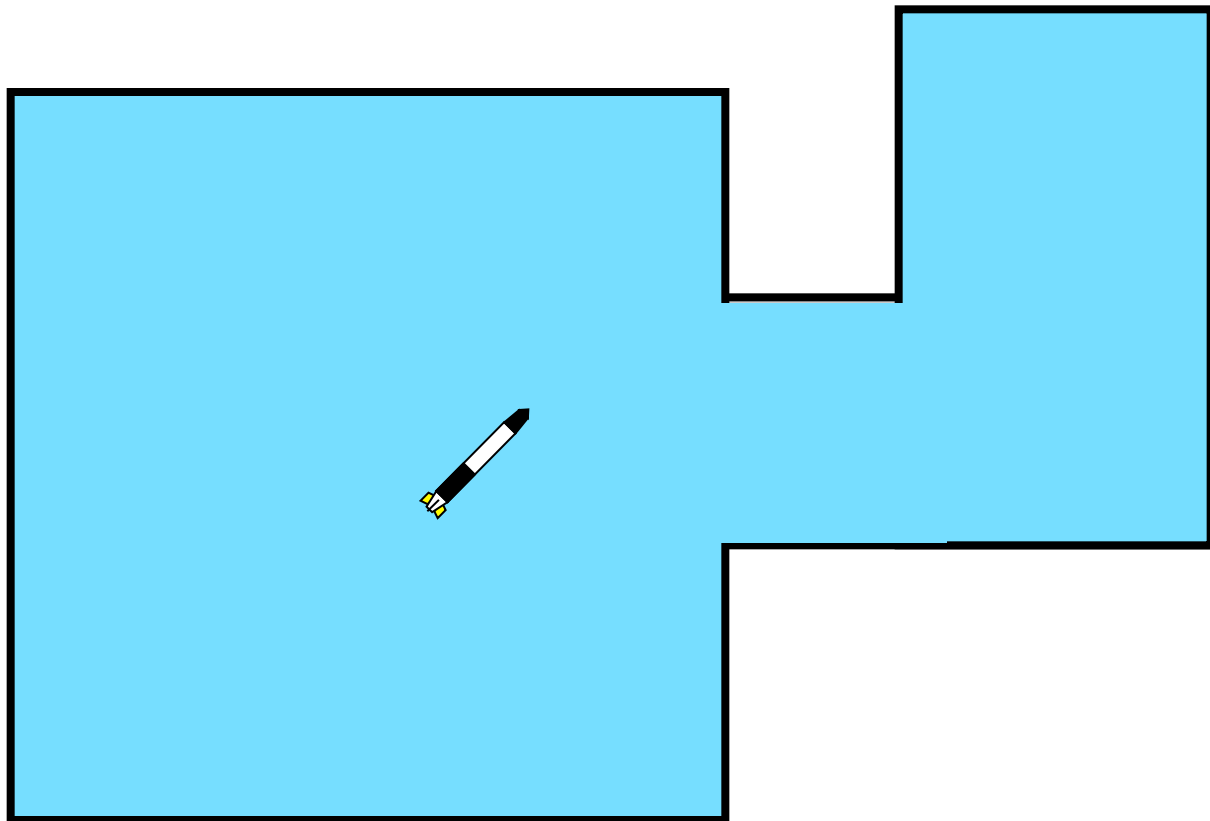
- How do we convert from polar coordinates to Cartesian coordinates?

$$x = r \cos(\theta)$$

$$y = r \sin(\theta)$$



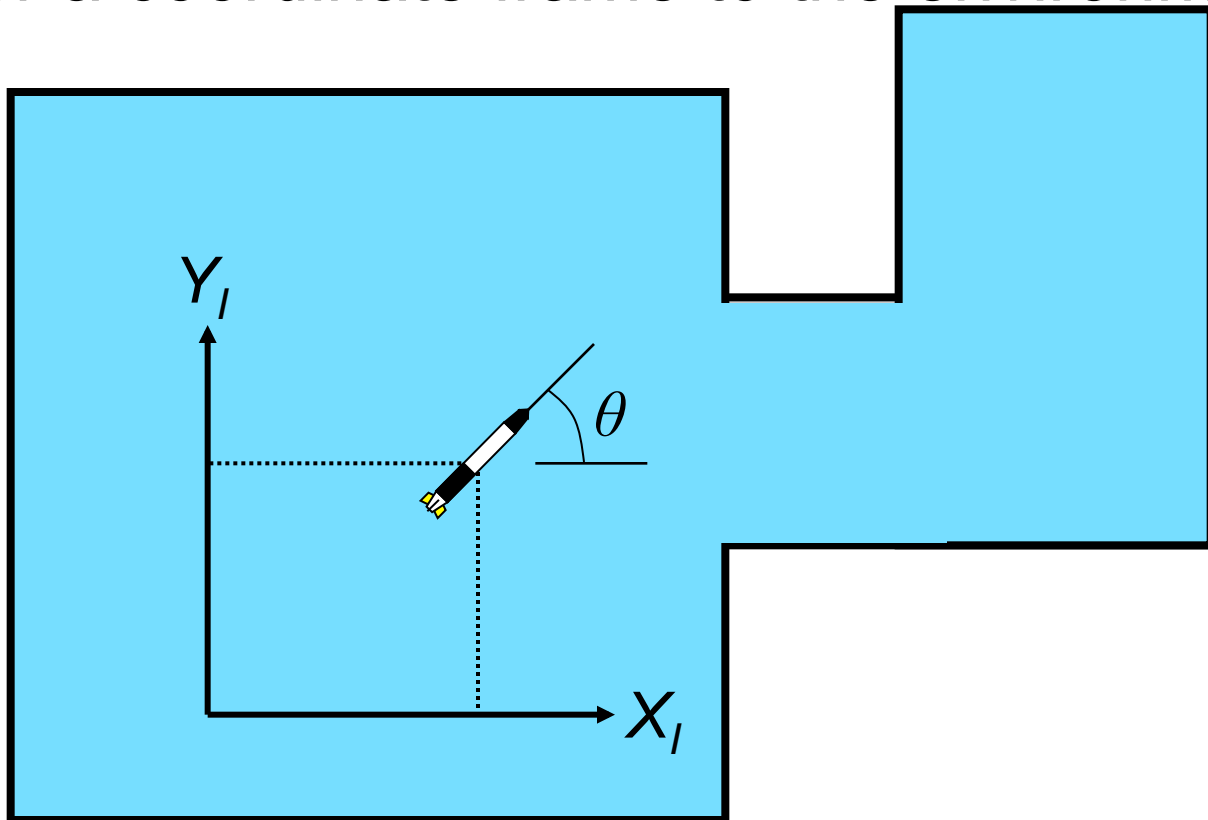
# Global (Inertial) Coordinate frame





# Global (Inertial) Coordinate frame

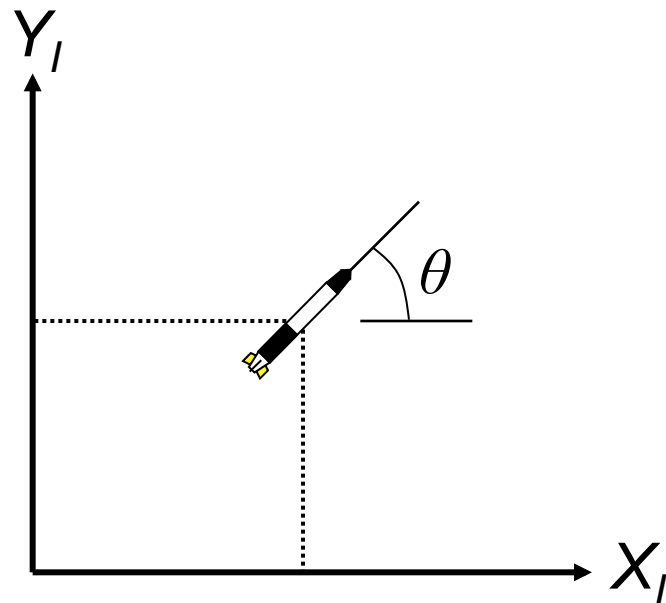
- Anchor a coordinate frame to the environment



# Global (Inertial) Coordinate frame

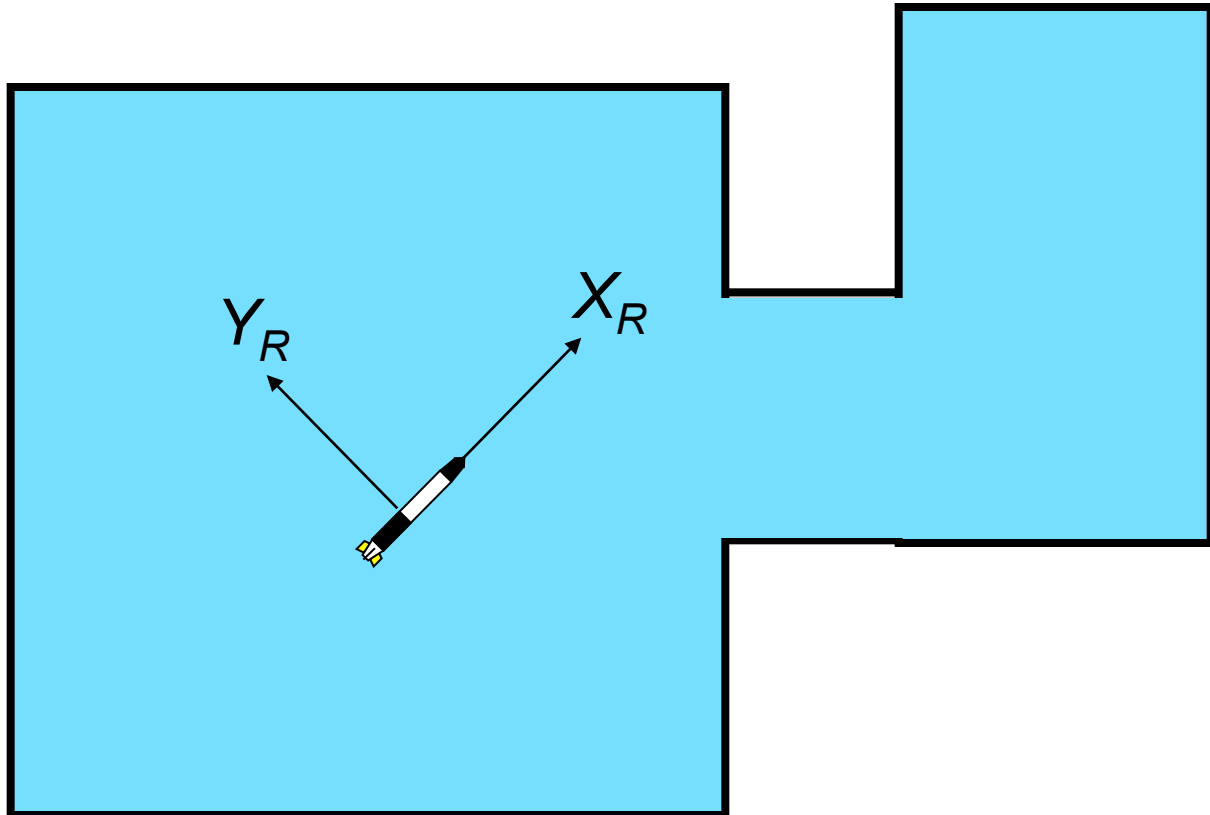
- With this coordinate frame, we describe the robot state as:

$$X_I = [x \ y \ \theta]_I$$



# Local Coordinate frame

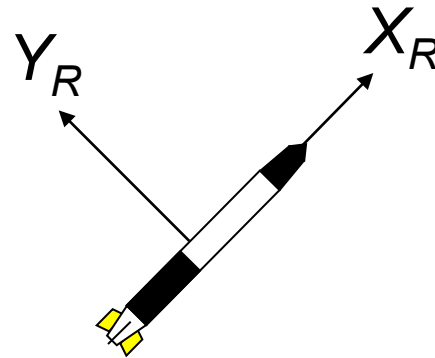
- Anchor a coordinate frame to the robot



# Local Coordinate frame

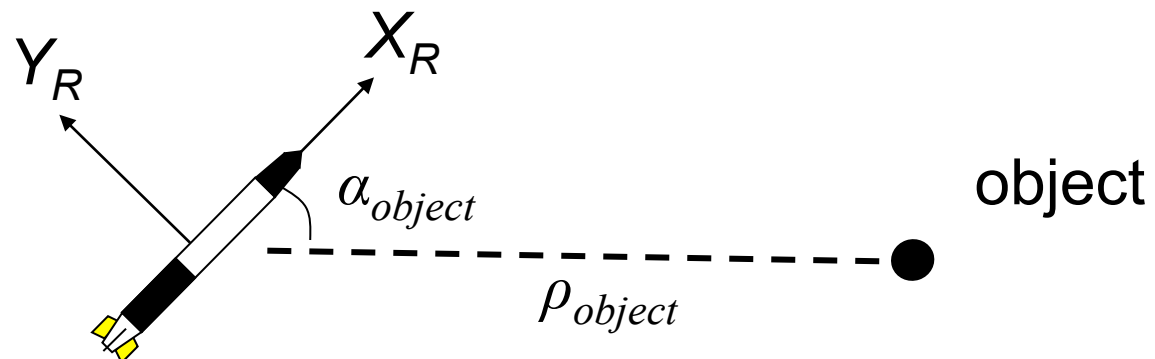
- With this coordinate frame, we describe the robot state as:

$$X_R = [x \ y \ \theta]_R = [0 \ 0 \ 0]$$



# Local Coordinate frame

- The local frame is useful when considering taking measurements of environment objects.
  - Consider the detection of an wall using a range finder:

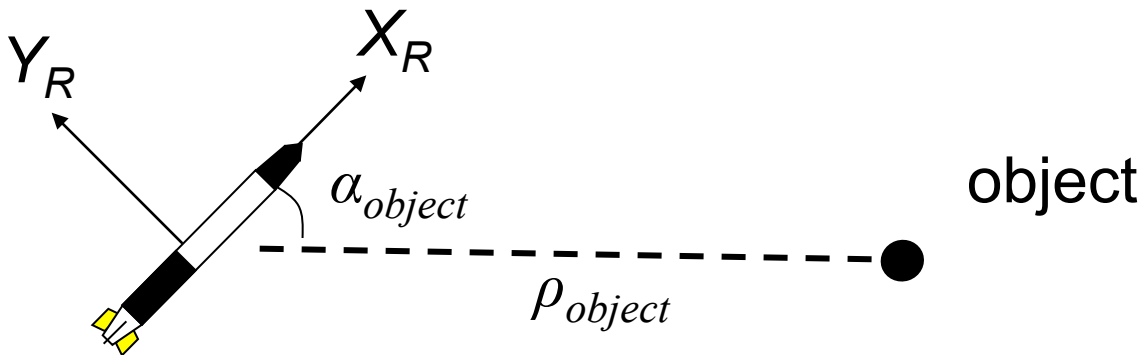


# Local Coordinate frame

- The measurement is taken relative to the robot's local coordinate frame  $(\rho_{object}, \alpha_{object})$
- We can calculate the position of the measurement in local coordinate frames:

$$x_{object, R} = \rho_{object} \cos(\alpha_{object})$$

$$y_{object, R} = \rho_{object} \sin(\alpha_{object})$$



# Local Coordinate frame

- One can calculate the position of the object in the global coordinate frame  $(x_{object,I}, y_{object,I})$

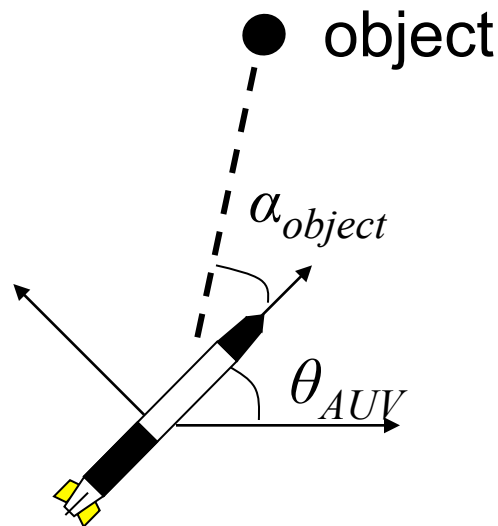
$$x_{object,I} = x_{AUV,I} + (x_{object,R} \cos(\theta_{AUV,I})) + (y_{object,R} \sin(\theta_{AUV,I}))$$

$$y_{object,I} = y_{AUV,I} + (x_{object,R} \sin(\theta_{AUV,I})) + (y_{object,R} \cos(\theta_{AUV,I}))$$

# Local Coordinate frame

- One can calculate  $\alpha_{object}$  if the positions are known

$$\theta_{AUV, I} + \alpha_{object} = \text{atan2}(y_{object} - y_{AUV}, x_{object} - x_{AUV})$$



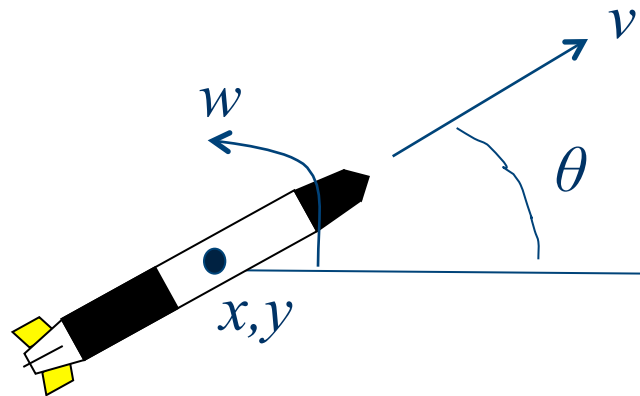


# Navigation and Control

1. Control Architectures
2. Navigation Example 2
3. Basic Tools for AUV Navigation
  1. Coordinate Frames
  2. **Motion Modeling**
  3. P Control

# A (simple) motion model

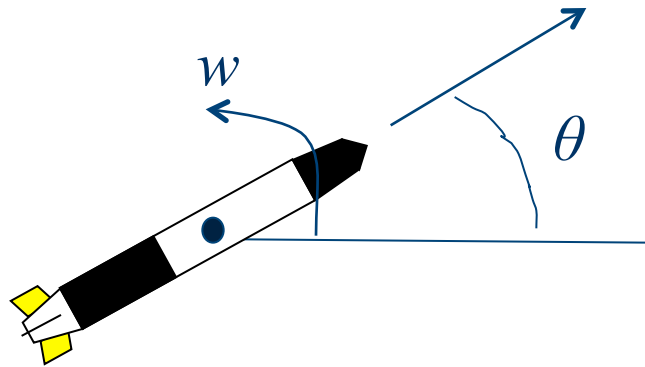
- Consider a robot moving from position  $x, y$  in direction  $\theta$  radians with forward velocity  $v$  m/s and rotational velocity  $w$  rad/s.



# A (simple) motion model

- How much will it rotate in  $t$  seconds?

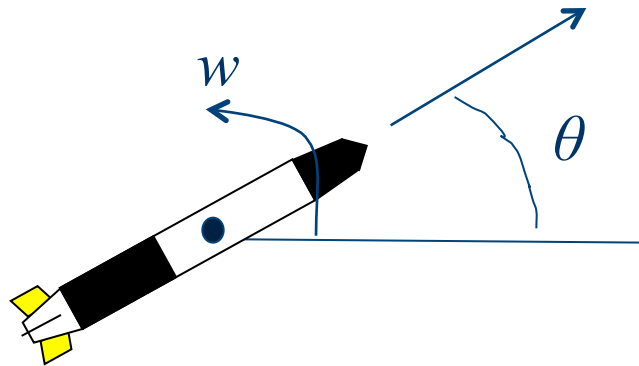
It will rotate a distance of  $\omega t$  radians.



# A (simple) motion model

- So, from time step  $k$  to time step  $k+1$ , the angle changes to:

$$\theta_{k+1} = \theta_k + \omega t$$



# A (simple) motion model

- So, from time step  $k$  to time step  $k+1$ , the position changes to:

$$x_{k+1} = x_k + vt \cos(\theta_k)$$

$$y_{k+1} = y_k + vt \sin(\theta_k)$$

# Navigation and Control

1. Control Architectures
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# P Control

- Proportional Feedback Control – P Control – uses the error between the desired and measured state to determine the control signal.
- If  $x_{desired}$  is the desired state, and  $x$  is the actual state, we define the error as:

$$e = x_{desired} - x$$

# P Control

- The control signal  $u$  is calculated as

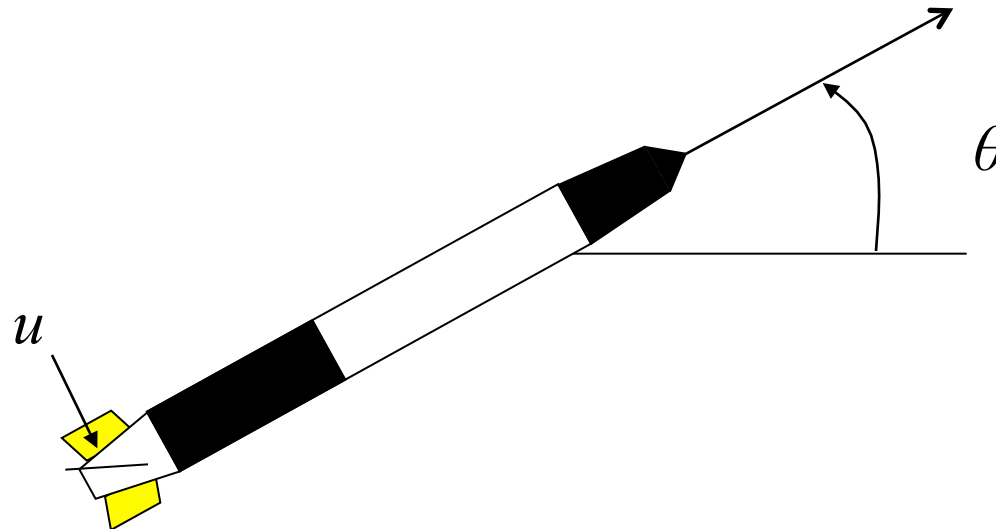
$$u = K_P e$$

where  $K_P$  is called the proportional gain.



# P Control

- Example:
  - Consider the orientation control of an AUV. Assume the orientation is completely controlled by the rear rudder fins.



# P Control

- Example cont':
  - The control signal  $u$  is calculated as

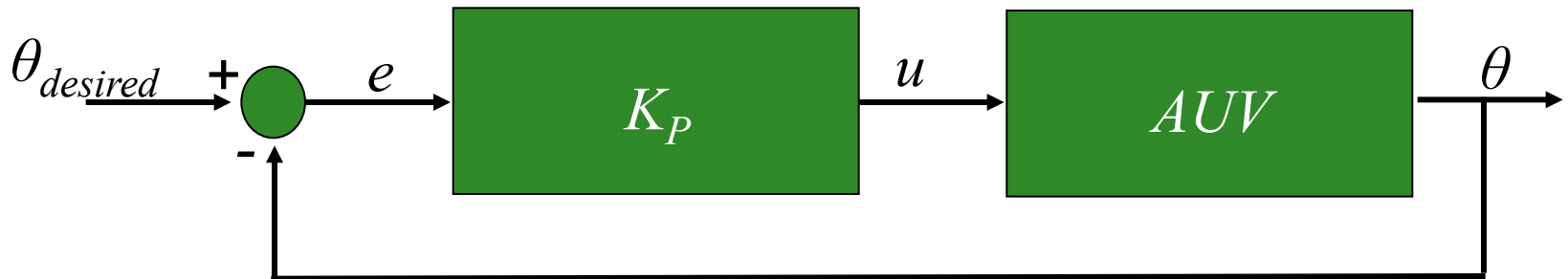
$$u = K_P(\theta_{desired} - \theta)$$

- Notes:
  - If  $\theta_{desired} = \theta$ , the control signal is 0.
  - If  $\theta_{desired} < \theta$ , the control signal is negative, resulting in an decrease in  $\theta$ .
  - If  $\theta_{desired} > \theta$ , the control signal is positive, resulting in an increase in  $\theta$ .
  - The magnitude of the increase/decrease depends on  $K_p$

# P Control

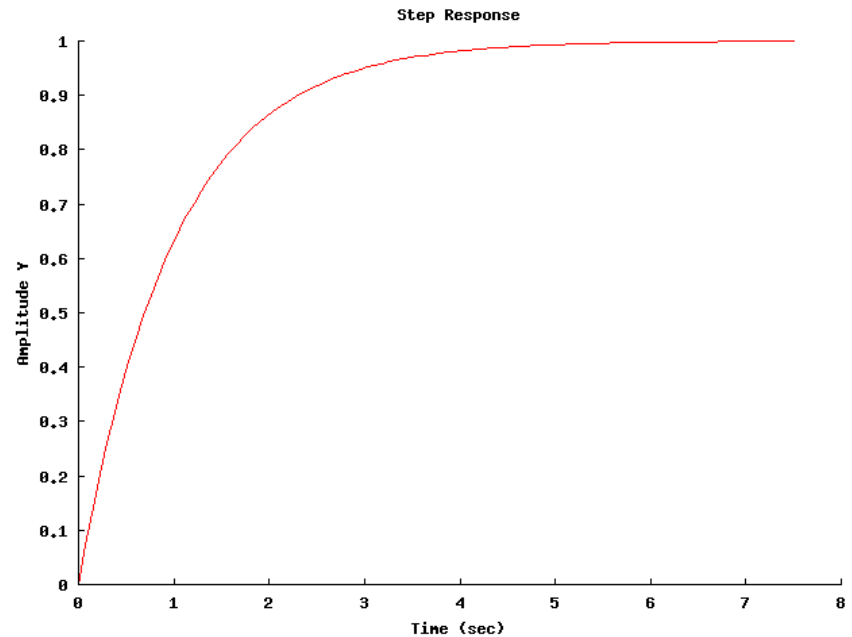
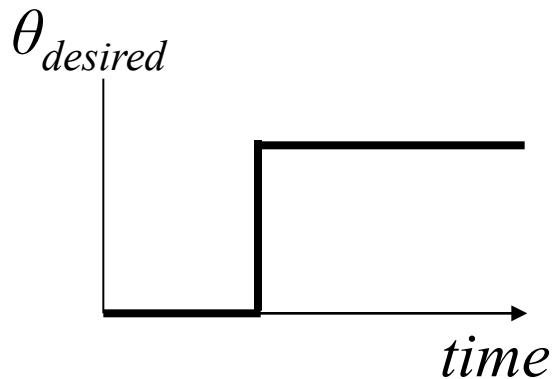
- Block Diagram:

$$u = K_P(\theta_{desired} - \theta)$$



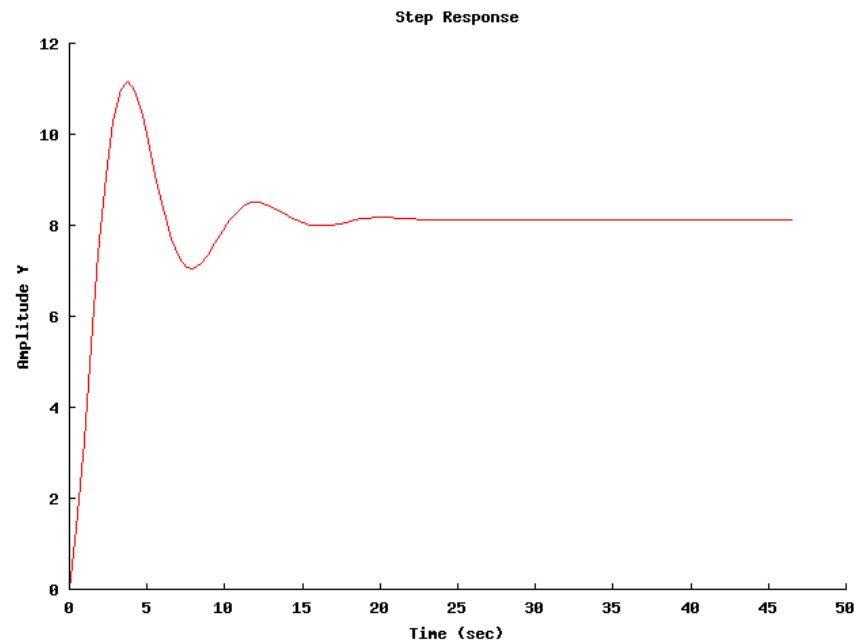
# P Control

- Time Domain Response of step response
  - Step from  $\theta_{desired} = 0$  to  $\theta_{desired} = 1$ .



# P Control

- Time Domain Response:
  - Step from  $\theta_{desired} = 0$  to  $\theta_{desired} = 8$ .
  - Different dynamics in this example... overshoot!





# Lab

- Form a group of Three
  - Email instructor one list of names along with a group number
- Start Lab 0
  - Optional for those with MVS and C# experience
- Start Lab 1
  - Code can be downloaded from the internet
- Read Lab 3
  - Can brainstorm in your groups