

Maritime Informatics & Robotics

SUMMER SCHOOL > MARITIME INFORMATICS & ROBOTICS

PROGRAMME START DATES

18 - 27 June 2025

LOCATION

Syros Island, Greece

LANGUAGE

English

FORMAT

In class

Robotic Motion Control

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In this Lecture

- Introduction
- Motion Models
- Forward Kinematics, Inverse Kinematics
- Trajectory Tracking
- Open-loop vs Closed-loop Control
- Introduction to PID control
- Introduction to Model Predictive Control (MPC)
- Introduction to Marine Craft Model and Control

Introduction

- **What is a robot?**

- First defined by Czech writer Karel Čapek in his play R.U.R. (Rossum's Universal Robots) in 1920

Introduction

- **What is a robot?**

- First defined by Czech writer Karel Čapek in his play R.U.R. (Rossum's Universal Robots) in 1920

- **Many alternative definitions but common features:**

- Should be able to “move” (i.e. do physical work)
- Should be programmable
- “Intelligence” and autonomy are optional but desirable

Motion Planning and Control

Motion Planning and Control

- **Motion planning:**

- Where should I go and what motions should I go through?
- Decides "what path" the robot should take to reach a goal while avoiding obstacles
- ✓ Generates the trajectory (offline/online)

- **Motion control:**

- How do I perform the desired motions?
- Determines "how to follow" that path by computing real-time actuator commands
- ✓ Executes the trajectory accurately (real-time)

What did we assume?

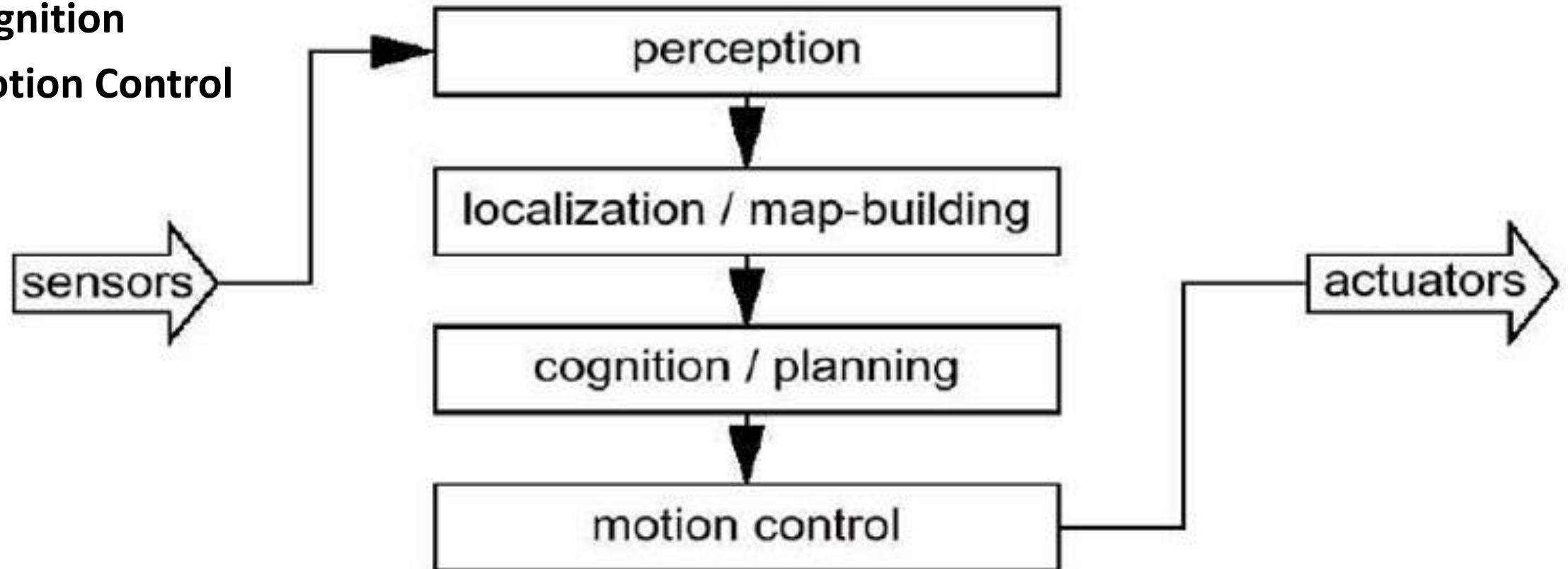
- **Perfect sensors?**
 - What information is available?
 - Uncertainty?
- **Perfect control?**
 - What controls are available?
 - Uncertainty?

Motion Planning and Control

- Through **perception**, a model of the “real” world is captured in memory
 - A goal is given, and a **plan** is generated, assuming the “real” world is not changing
 - Then, the plan is **executed**, one (abstract) operation at a time
-
- What is “interesting” in the “real” world to be captured?
 - At what level of details should we represent the “real” world?
 - The “real” world model must be at all times accurate (consistent and reliable)
 - What if during plan execution, the “real” world changes?
 - Sudden changes in the world may not be reflected instantly in our model

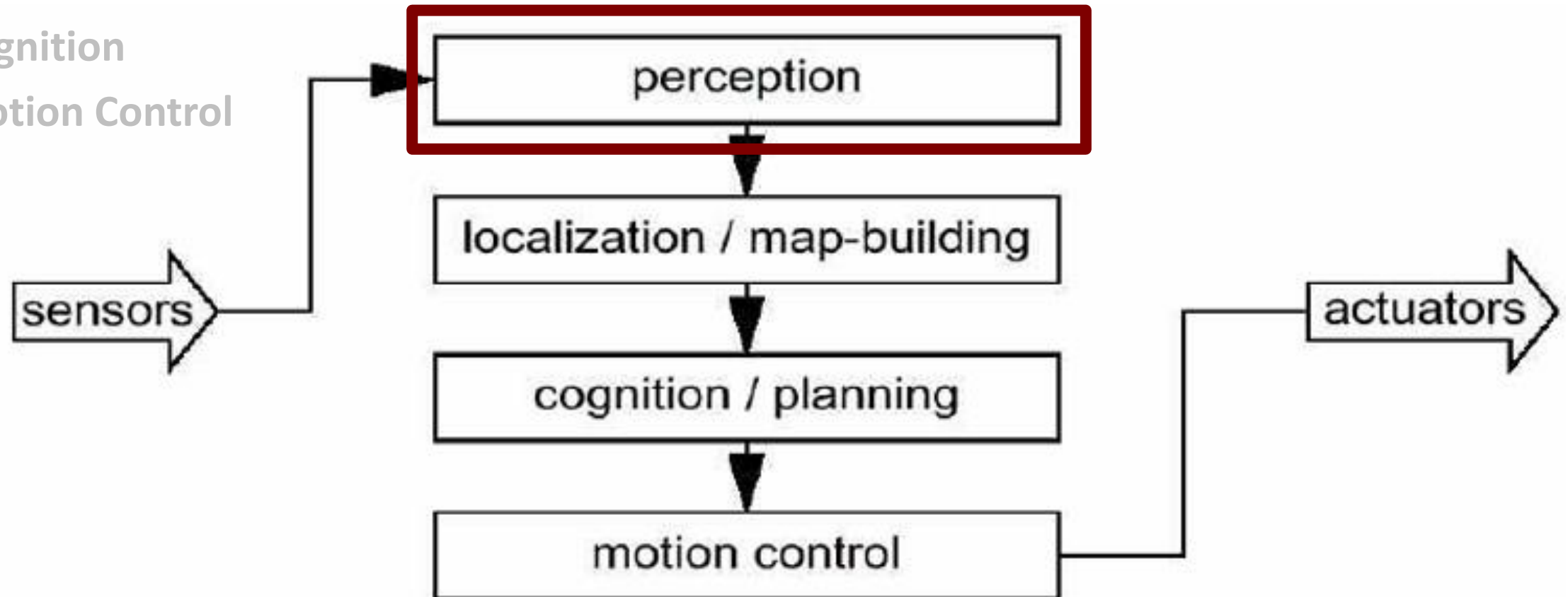
Robot Navigation

- Perception
- Localization
- Cognition
- Motion Control



Robot Navigation

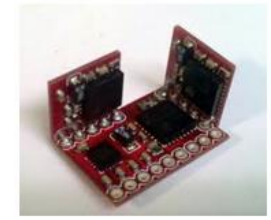
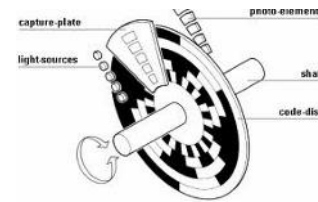
- Perception
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Robot Navigation - Perception

■ Proprioceptive

- Compass
- Encoders
- Accelerometers
- IMU – Inertial Measurement Unit



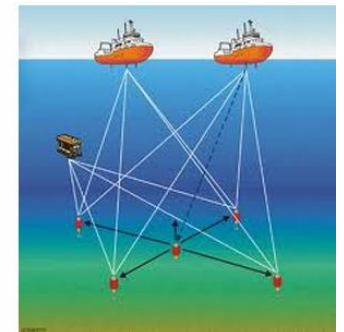
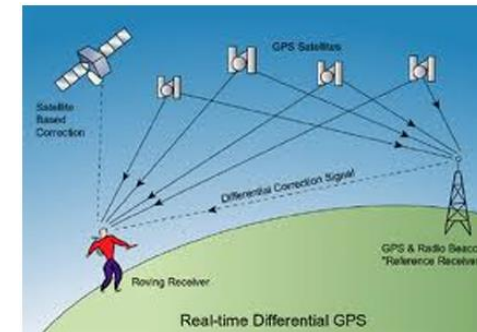
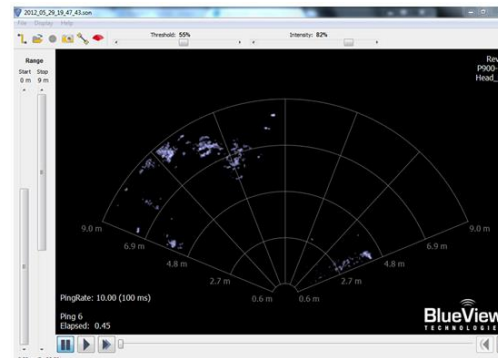
Images from elechouse.com, hightechsa.blogspot.com inmotion.pt, xkyle.com

■ Exteroceptive

- Range Sensors
- Vision Systems
- Sonar
- Positioning Systems (e.g. GPS)



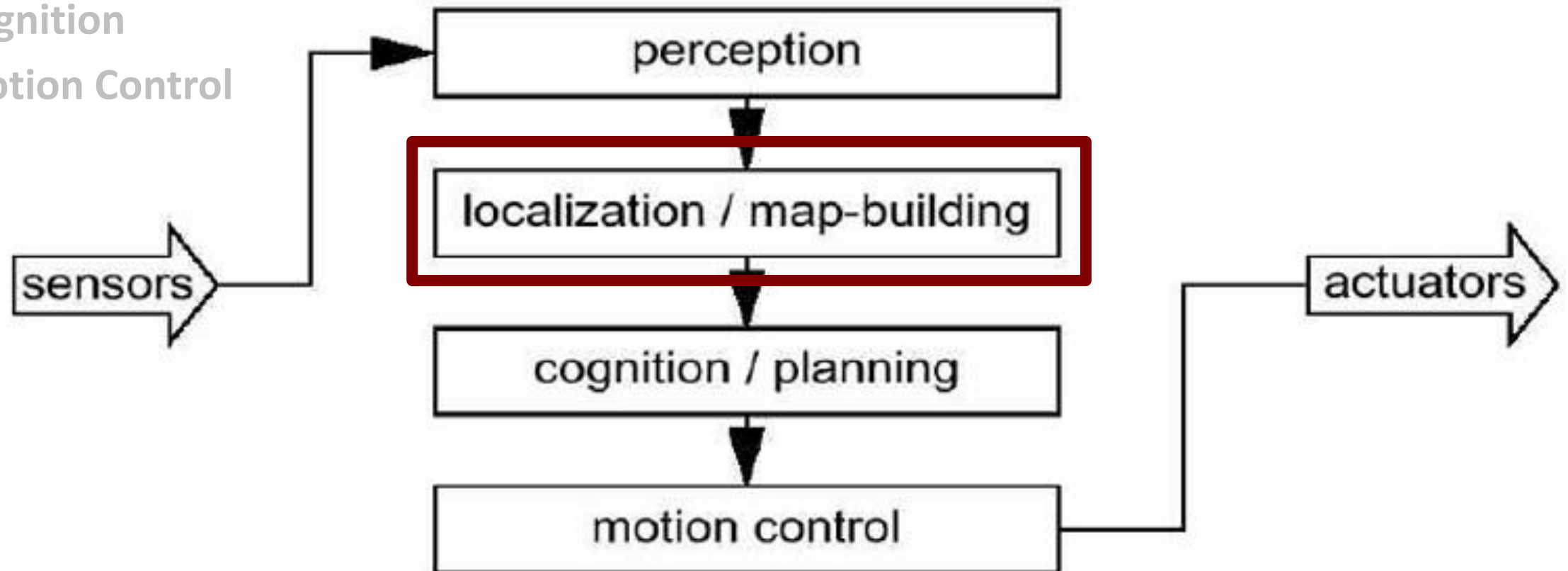
Images from nauticexpo.com, acroname.com, robots.com, www8.garmin.com



Images from 660audio.com, km.kongsberg.com

Robot Navigation

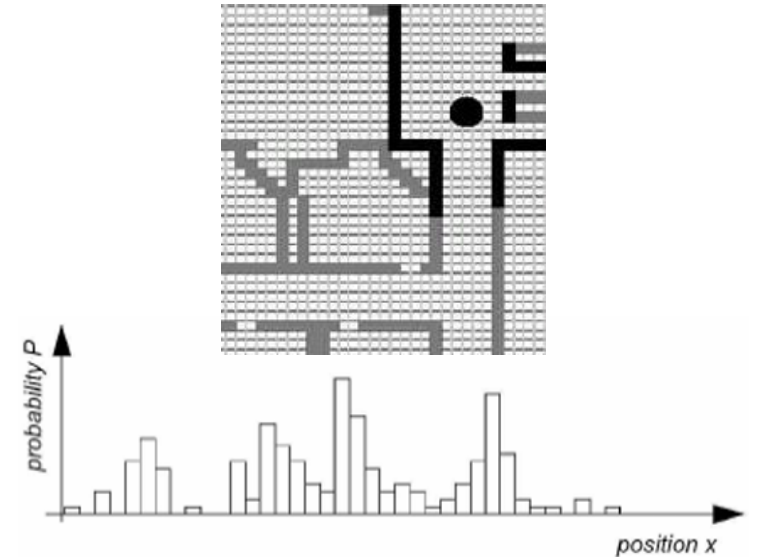
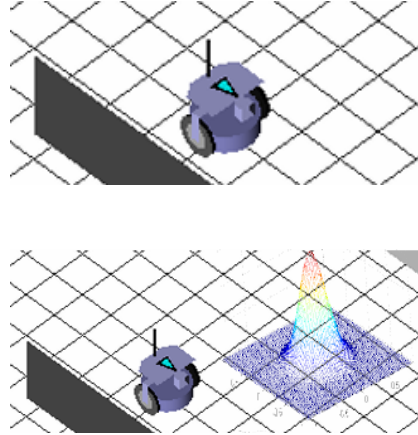
- Perception
- **Localization**
- Cognition
- Motion Control



Robot Navigation - Localization

■ Probabilistic Representations

- Continuous
- Discrete

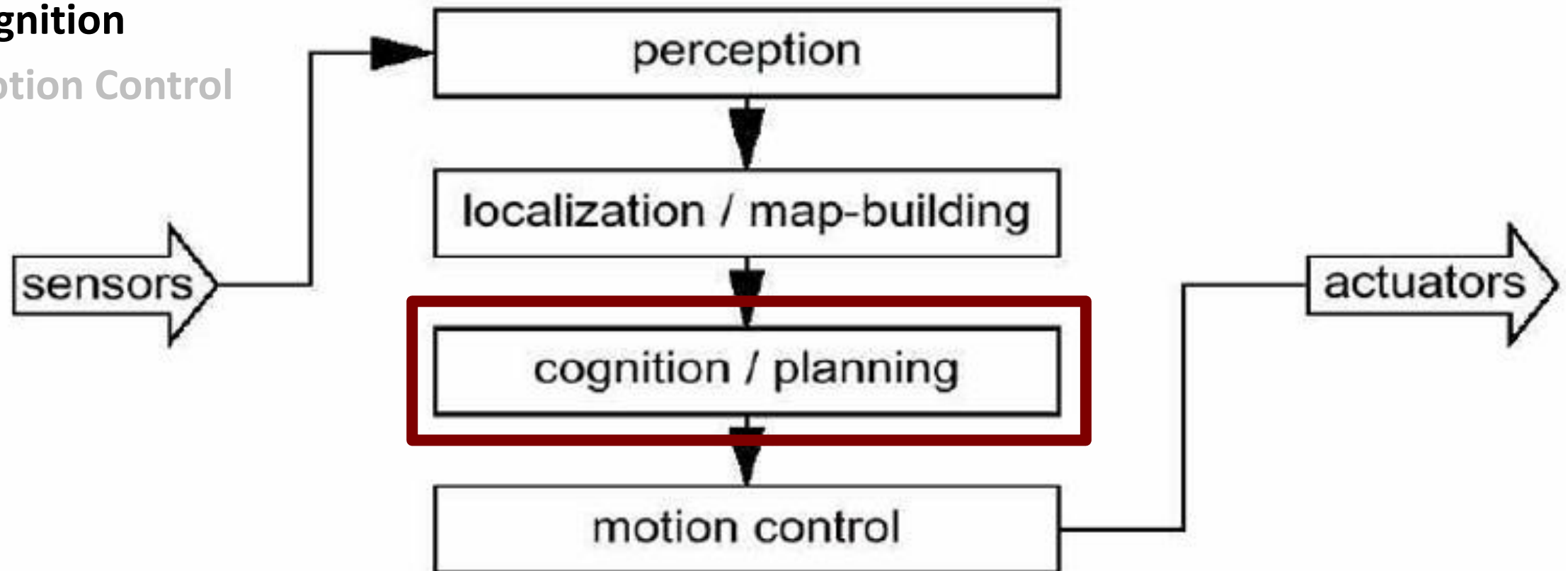


■ Probabilistic Algorithms

- Kalman Filter Based
 - Assumes Gaussian representation of robot state
 - Compact representation good for real time implementation
- Particle Filter Based
 - Uses many particles to represent robot state, each particle is an estimate of the robot position with an associated weight

Robot Navigation

- Perception
- Localization
- **Cognition**
- Motion Control



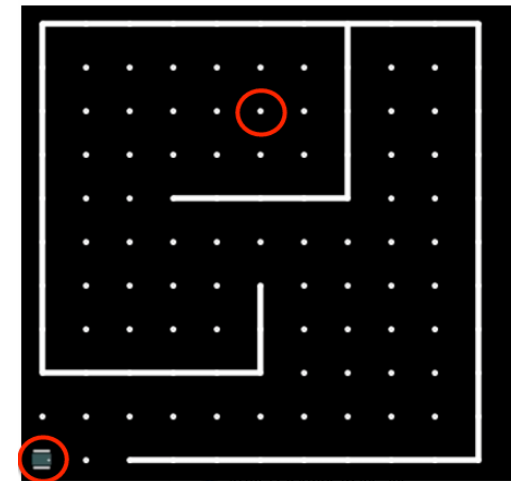
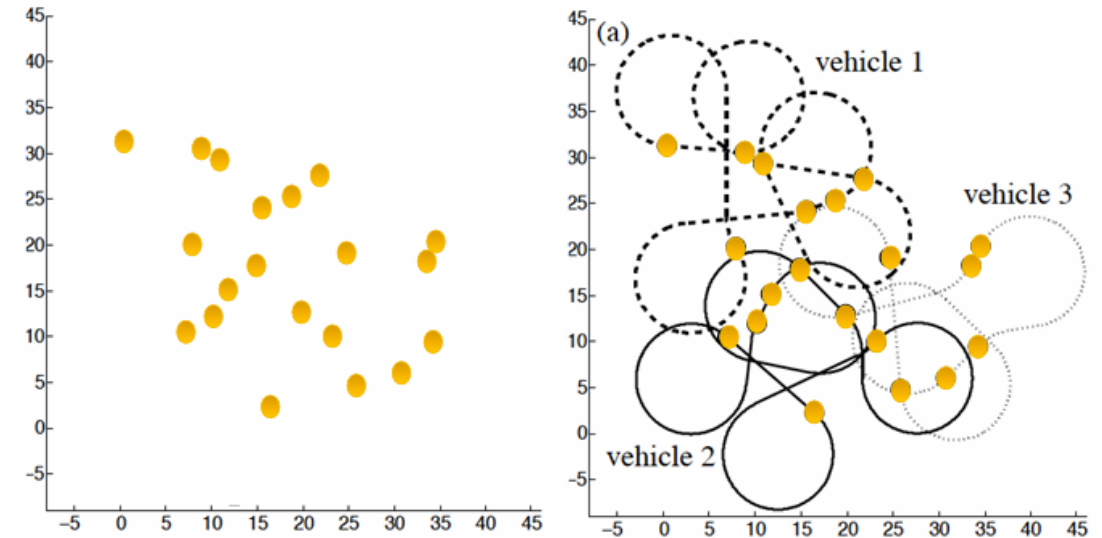
Robot Navigation - Cognition

- **Task Planning**

- Given a set of tasks (e.g. task locations), identify ordering sequence for the tasks

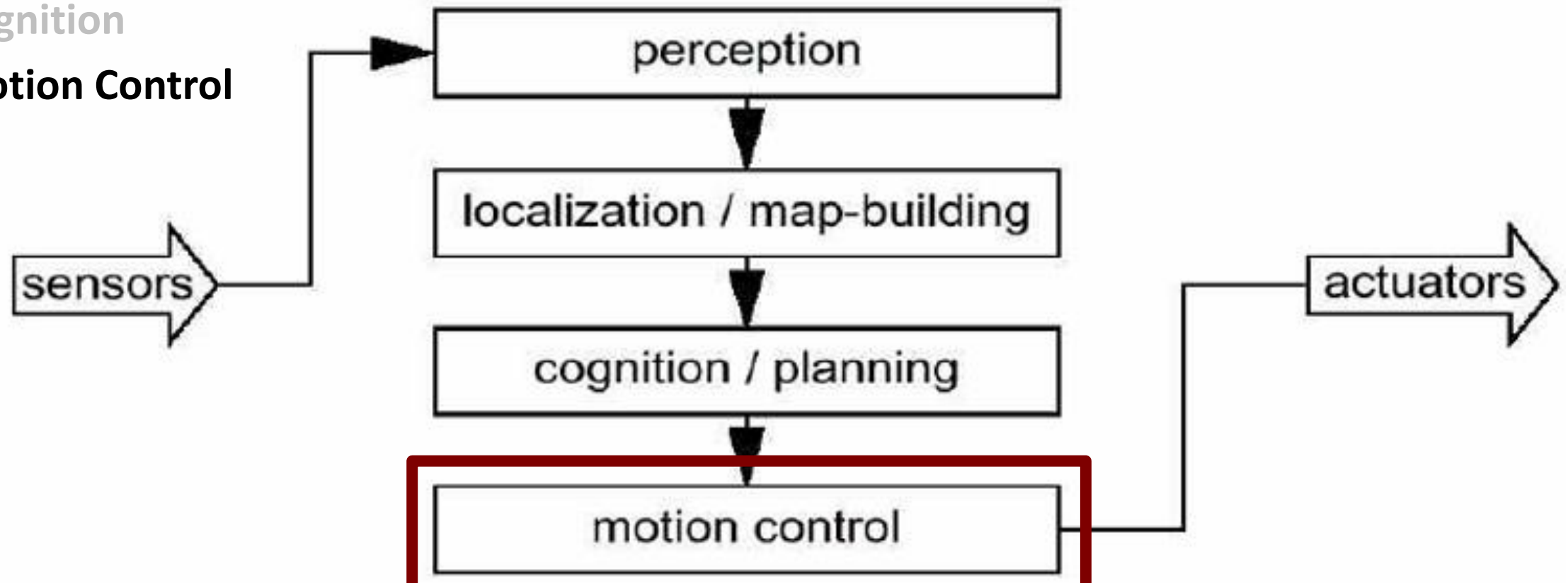
- **Motion Planning**

- Given a robot's Start Configuration and Goal Configuration, construct a collision free trajectory from Start to Goal



Robot Navigation

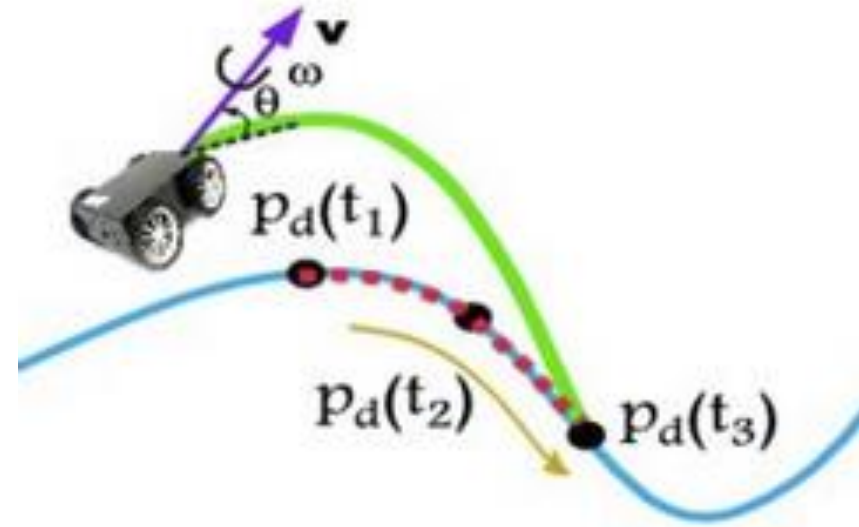
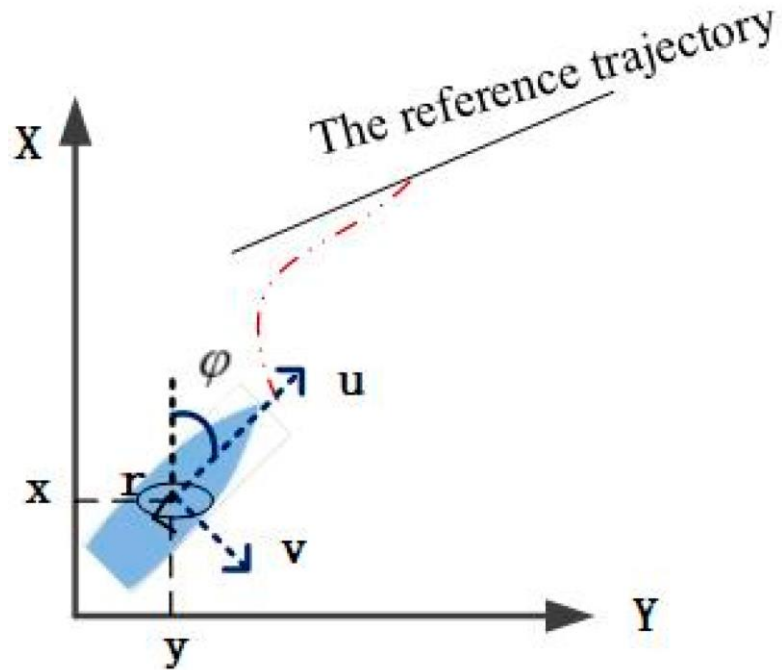
- Perception
- Localization
- Cognition
- **Motion Control**



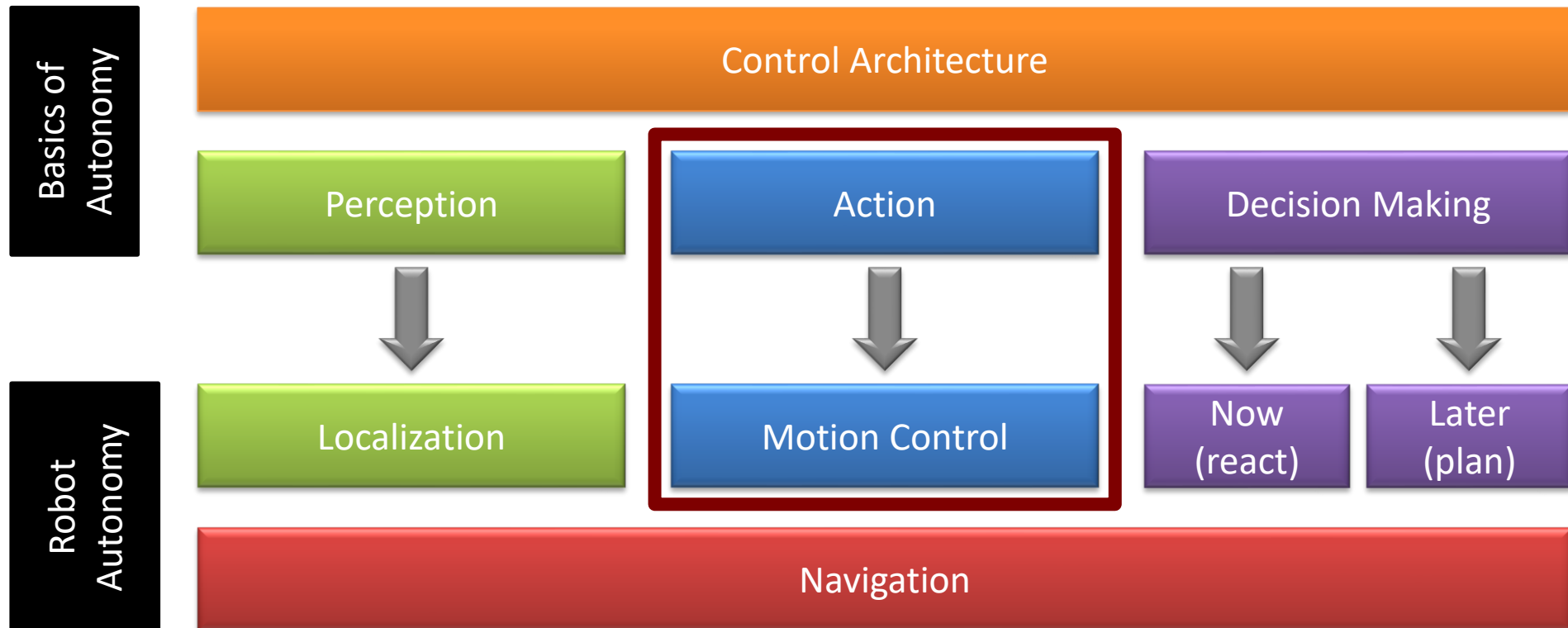
Robot Navigation - Motion Control

- **Trajectory Tracking Control**

- Given a trajectory, determine the control signals sent to actuators that guarantee the robot will follow the trajectory

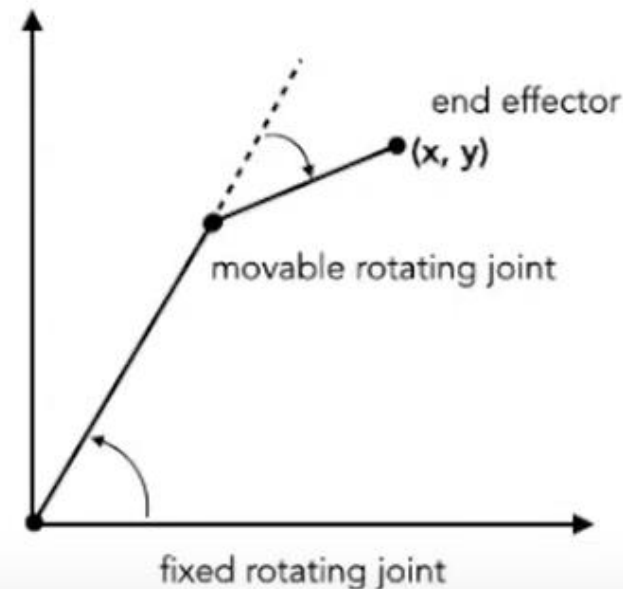
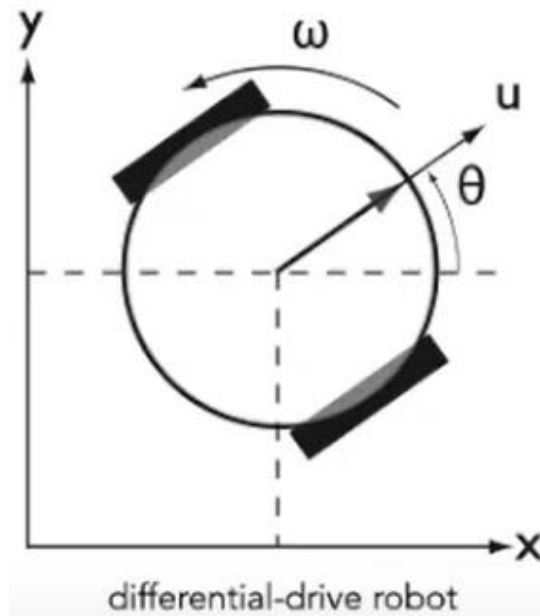


Control Architecture



Degrees of Freedom

- Most actuators control a single **degree of freedom** (DOF)
- Every robot has a specific number of DOF
- If there is an actuator for every DOF, then all SOF are controllable

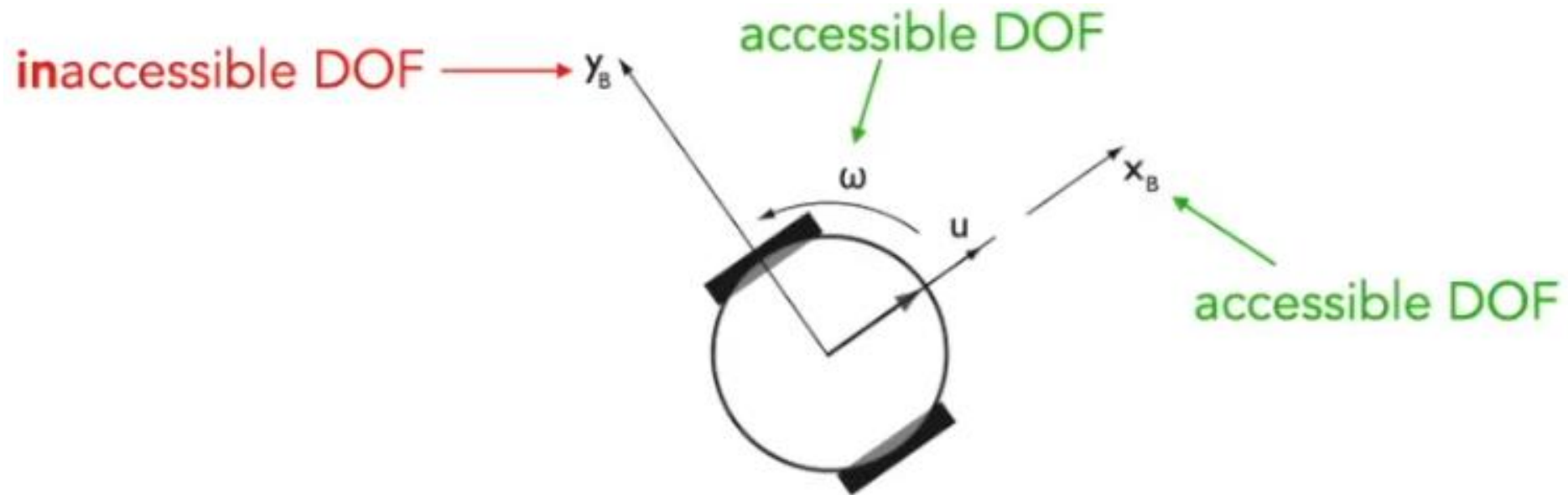


Holonomic Motion

- **Degree of mobility: DOM (differentiable DOF)**
 - Number of DOF that can be **directly accessed** by the actuators
 - A robot in the plane has at most 3 DOMs (position and heading)
- **Holonomic motion:**
 - **Holonomic robot:**
 - When the number of DOF is equal to robot's DOM
 - **Non-holonomic robot:**
 - When the number of DOF is greater than robot's DOM
- When a robot's DOM is larger than its DOF, the robot has “**redundant**” actuation

Differential-Drive Robot

- Differential-drive robots can actuate left and right wheels (independently)



- DOF = 3, but DOM = 2: differential-drive robots are **non-holonomic**
- Impact of non-holonomicity: **motion constraint affect motion planning**

Actuators

- Different purposes
 - Locomotion: e.g., wheeled, legged
 - Other motion: e.g., manipulation
 - Other type of actuation: e.g., heating, sound emission
- Examples of electrical-to-mechanical actuators:
 - DC motors, stepper motors, servo
 - **Control input** example: A driver can steer and accelerate/decelerate, so there are 2 control inputs
- Uncertainty/disturbance/noise:
 - Wheel slip, slack in mechanism, environmental factors (wind, friction, etc.)

Kinematics

- **Forward kinematics:**

- Given the control parameters (e.g., wheel velocities), and the time of movement t , **find the pose** (x, y, ϑ) reached by the robots

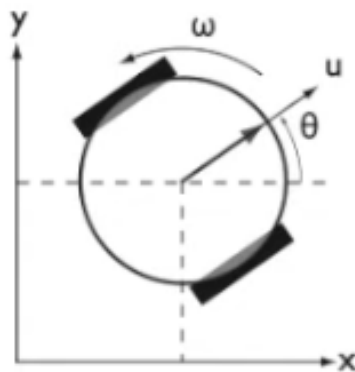
- **Inverse kinematics:**

- Given the final desired pose (x, y, θ) , **find the control parameters** to move the robot there at a given time t



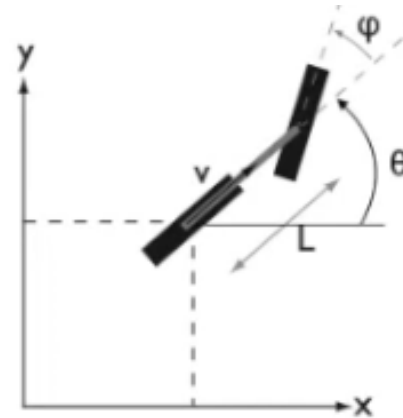
Forward Kinematics

- Differential equations describe robot motion
- How does robot state change over time as a function of control inputs?



$$\begin{cases} \dot{x} = u \cdot \cos \theta \\ \dot{y} = u \cdot \sin \theta \\ \dot{\theta} = \omega \end{cases}$$

differential-drive model
3 DOF (2 controllable)



$$\begin{cases} \dot{x} = v \cdot \cos \theta \\ \dot{y} = v \cdot \sin \theta \\ \dot{\theta} = v \cdot \frac{\tan \phi}{L} \end{cases}$$

bicycle model
3 DOF (2 controllable)

Forward Kinematics (body frame)

Actuators of differential-drive:

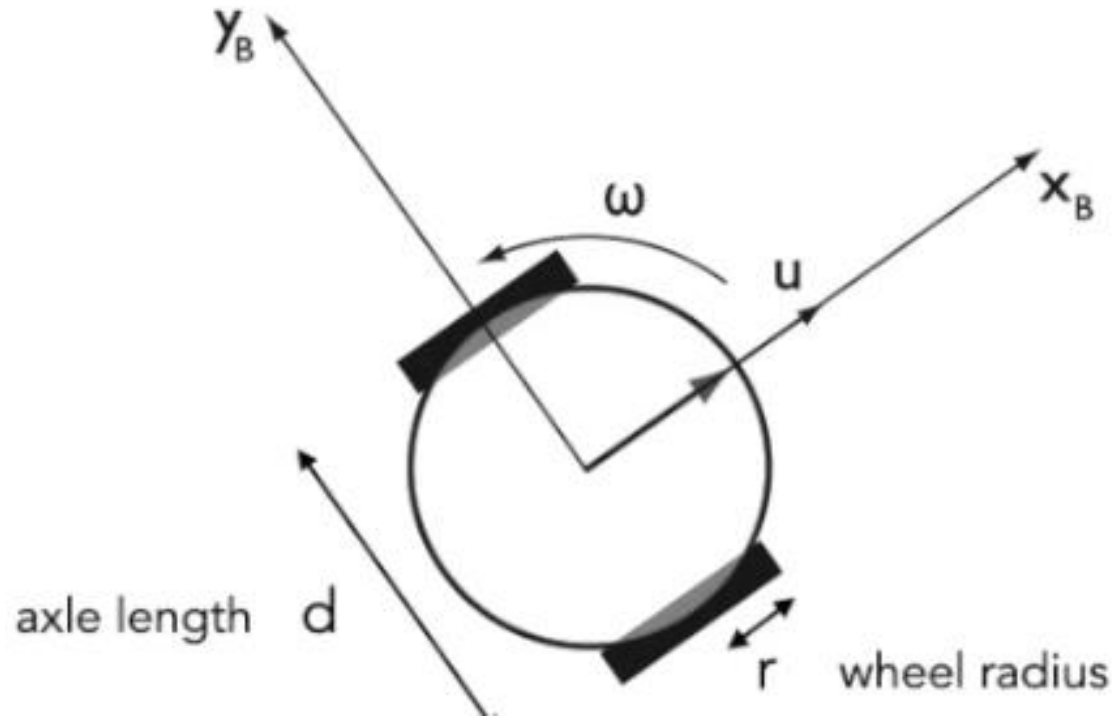
- Left wheel speed $\dot{\phi}_l$
- Right wheel speed $\dot{\phi}_r$

Forward velocity:

$$u = \frac{r\dot{\phi}_r}{2} + \frac{r\dot{\phi}_l}{2}$$

Rotational velocity:

$$\omega = \frac{r\dot{\phi}_r}{d} - \frac{r\dot{\phi}_l}{d}$$



Motion:

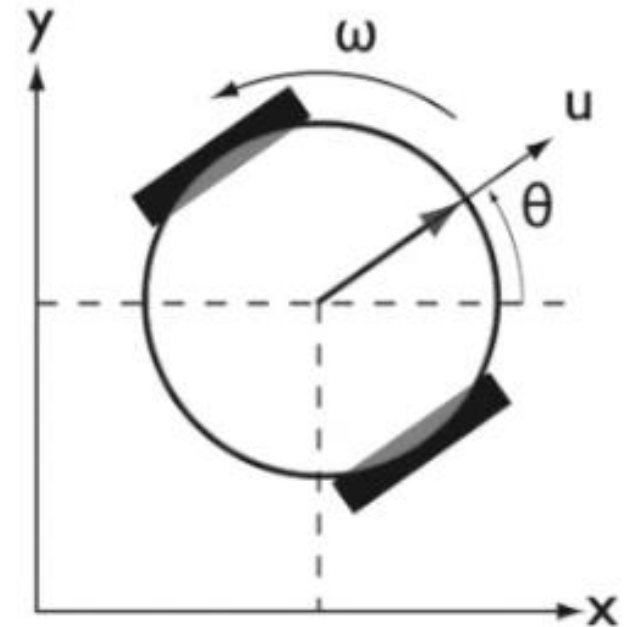
$$\begin{aligned}\dot{x}_B &= u \\ \dot{y}_B &= 0 \\ \dot{\theta}_B &= \omega\end{aligned}$$

Forward Kinematics (world frame)

- Given known control inputs, how does the robot move w.r.t. a **global coordinate system**?
- Use a **rotation matrix**:
 - From body to world frames, the axes rotate by θ

$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix} = \underbrace{\begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix}}_{T(\theta)} \begin{bmatrix} \dot{x}_B \\ \dot{y}_B \\ \dot{\theta}_B \end{bmatrix}$$

$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta & 0 \\ \sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} u \\ 0 \\ \omega \end{bmatrix} = \begin{bmatrix} u \cos \theta \\ u \sin \theta \\ \omega \end{bmatrix}$$



Inverse Kinematics

- We would like to control the robot motion in the world frame: $\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix}$
- We **invert** the previous equations to **find control inputs**:

$$\begin{bmatrix} u \\ 0 \\ \omega \end{bmatrix} = T^{-1}(\theta) \begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix} = \begin{bmatrix} \cos \theta & \sin \theta & 0 \\ -\sin \theta & \cos \theta & 0 \\ 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix}$$

- yielding $u = \dot{x} \cos \theta + \dot{y} \sin \theta$

$$\omega = \dot{\theta}$$

- and finally

$$\begin{aligned} \dot{\phi}_l &= u - \frac{\omega d}{2r} \\ \dot{\phi}_r &= u + \frac{\omega d}{2r} \end{aligned} \Rightarrow \begin{aligned} \dot{\phi}_l &= \dot{x} \cos \theta + \dot{y} \sin \theta - \frac{\dot{\theta} d}{2r} \\ \dot{\phi}_r &= \dot{x} \cos \theta + \dot{y} \sin \theta + \frac{\dot{\theta} d}{2r} \end{aligned}$$

we can now control the wheel speeds

- BUT: under the **constraint** (because this robot is non-holonomic)

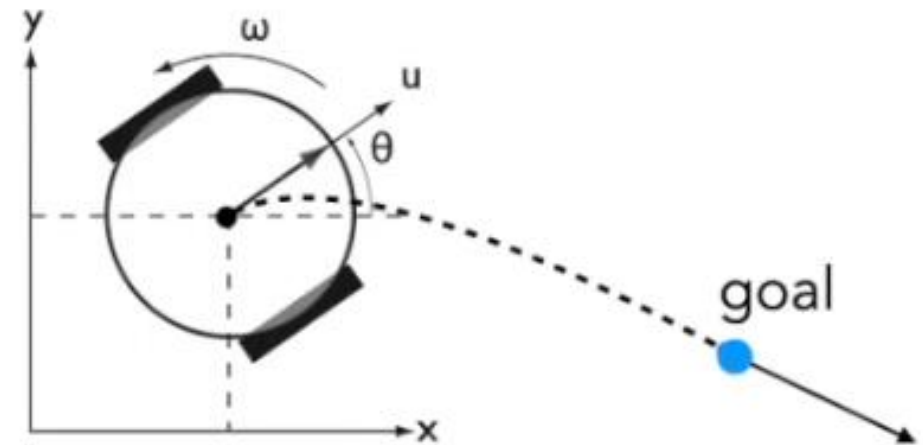
$$\dot{x} \sin \theta = \dot{y} \cos \theta$$

Inverse Kinematics

- We would like to control the robot to reach a goal pose: $\begin{bmatrix} x_G \\ y_G \\ \theta_G \end{bmatrix}$
- Ideally (if the robot would be holonomic), we would set:

$$\begin{bmatrix} \dot{x} \\ \dot{y} \\ \dot{\theta} \end{bmatrix} = K \begin{bmatrix} x_G - x \\ y_G - y \\ \theta_G - \theta \end{bmatrix}$$

control gain

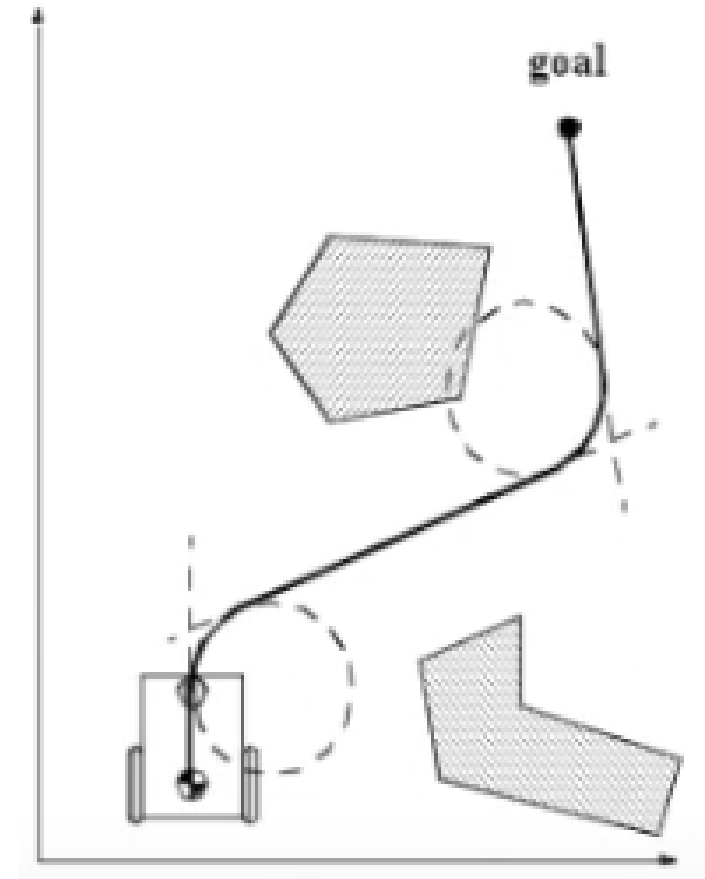


- However, we need to satisfy the non-holonomic constraint:

$$\dot{x} \sin \theta = \dot{y} \cos \theta$$

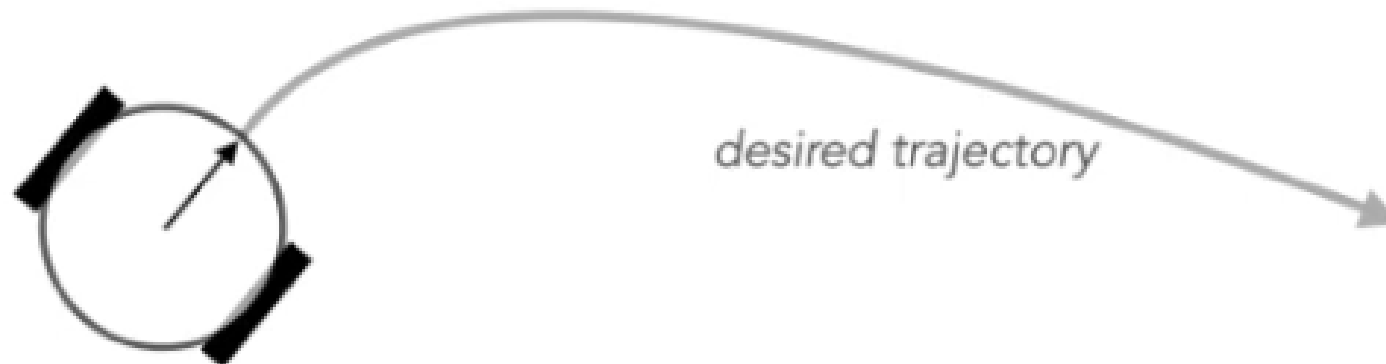
Trajectory Tracking

- Trajectory Tracking:
 - Pre-compute a smooth trajectory
 - Follow trajectory (in open-loop or closed-loop)
- Challenges
 - Feasibility of trajectory given motion constraints
 - Adaptation of trajectory in dynamical environments
 - Must guarantee smoothness of resulting trajectories (kinematic / dynamic feasibility): e.g., continuity of 1st derivative for 1st order control



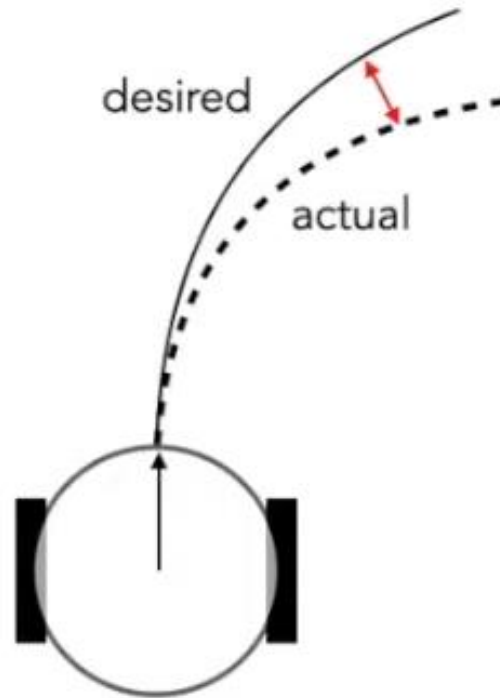
Open-Loop vs Closed-Loop

- Once we have a trajectory that enables the robot to reach its goal, we need to follow that trajectory
- There are two ways of doing this:
 - **Open-loop control:** Robot follows path blindly by applying the pre-computed control input
 - **Closed-loop control:** Robot can follow path for a small duration, then observe if anything changed in the world, recompute a new adapted path (repeatedly)



Open-Loop Controller

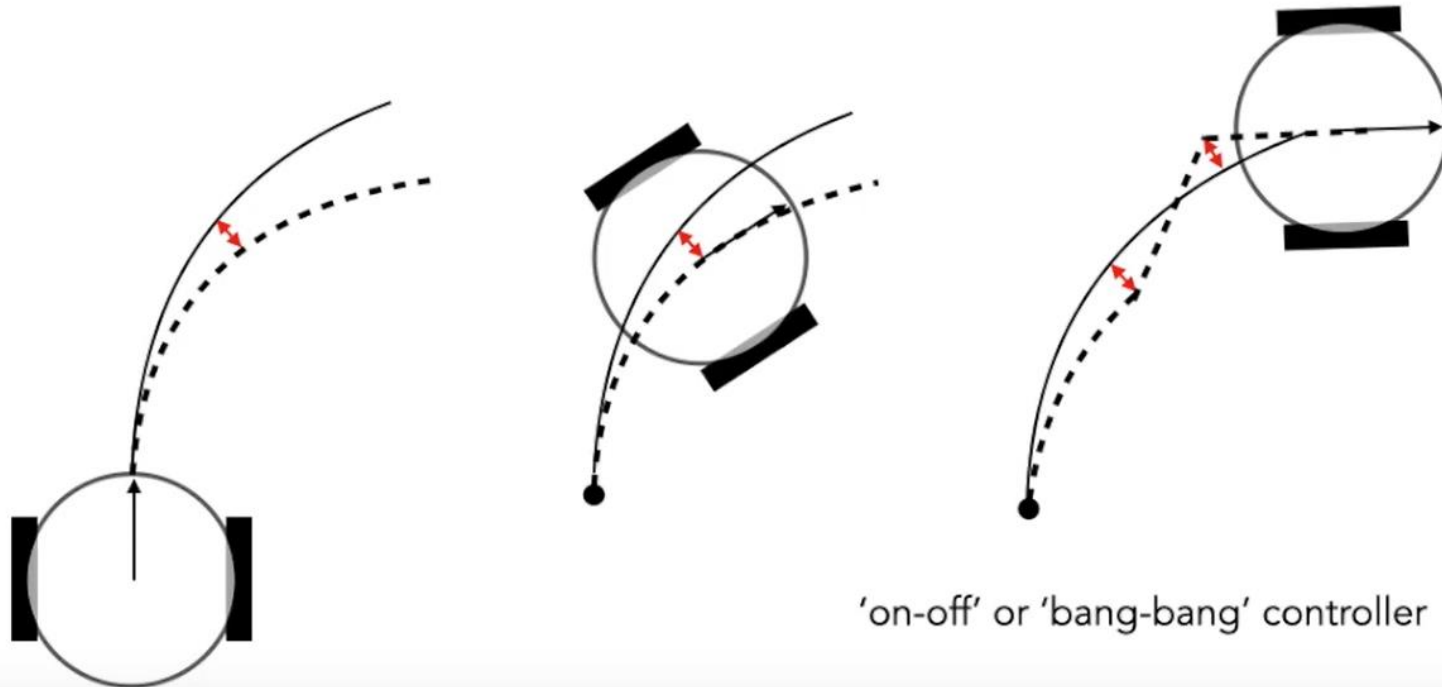
- Example: trajectory tracking
- In open-loop, the robot executes predefined control inputs



- Under imperfect conditions, the robot deviates from desired behavior

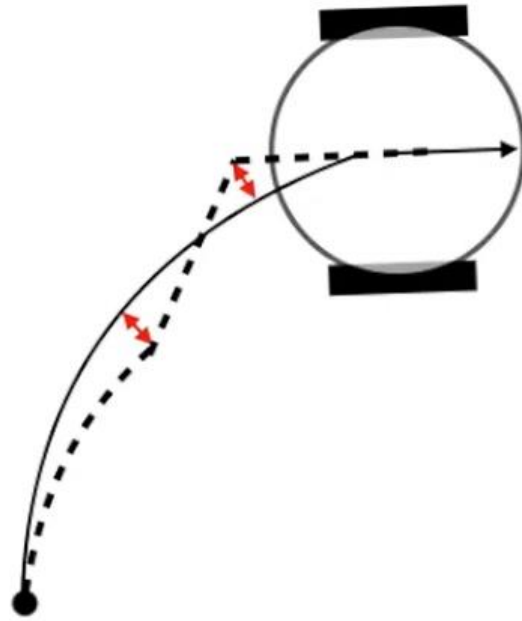
Closed-Loop Controller

- Example: trajectory tracking
- The robot uses **feedback** to maintain a desired set-point
- Assumption: robot receives **feedback** on distance to desired trajectory

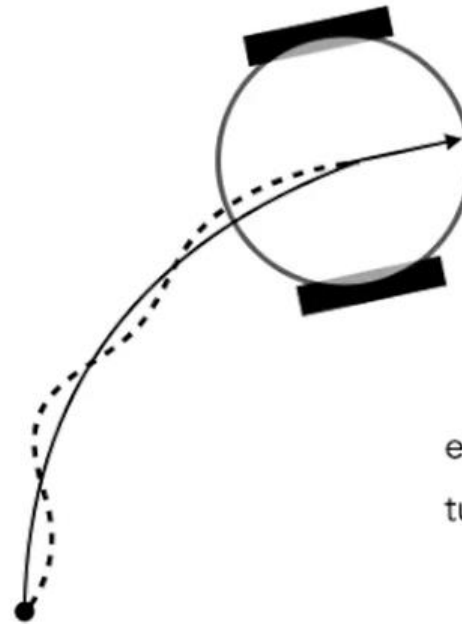


Proportional Control (P-Control)

- Example: trajectory tracking
- The robot uses **feedback** to maintain a desired set-point
- Robot computes error, and adjust control **as a function of error**



previous slide: oscillatory behavior



adjustment is proportional to error!

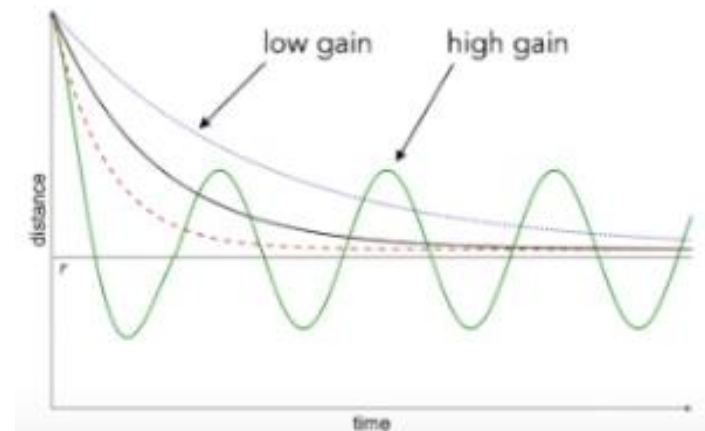
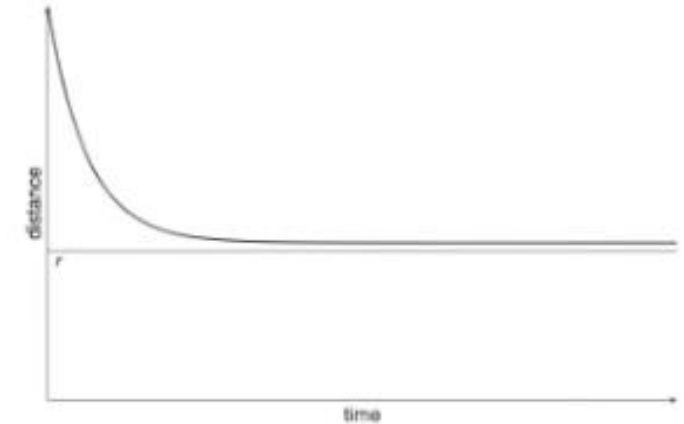
error = distance-to-trajectory
turning-control = $K * \text{error}$

Proportional Control (P-Control)

- Behavior of P-controller:
 - Adapt control proportionally to your perceived error to set-point

$$u(t) = \kappa_p e(t)$$

- Behavior for varying gain values
- High gains not desirable! We call this an **unstable** controller



PID Control

- PI-controller:

- takes into account **accumulated error** over time

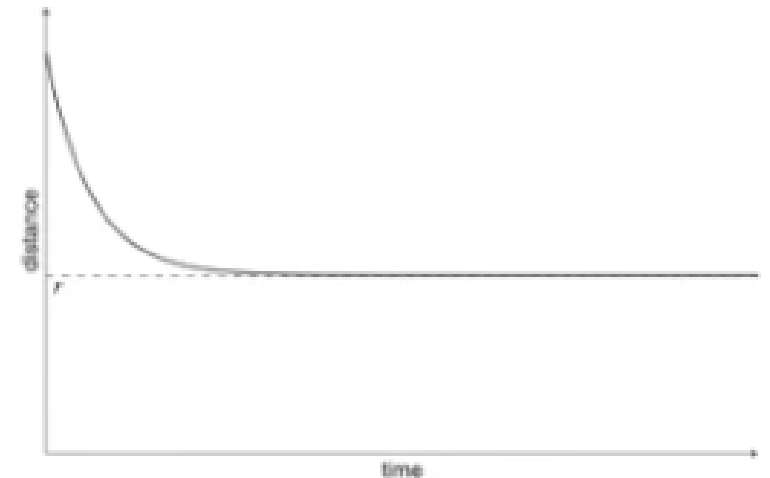
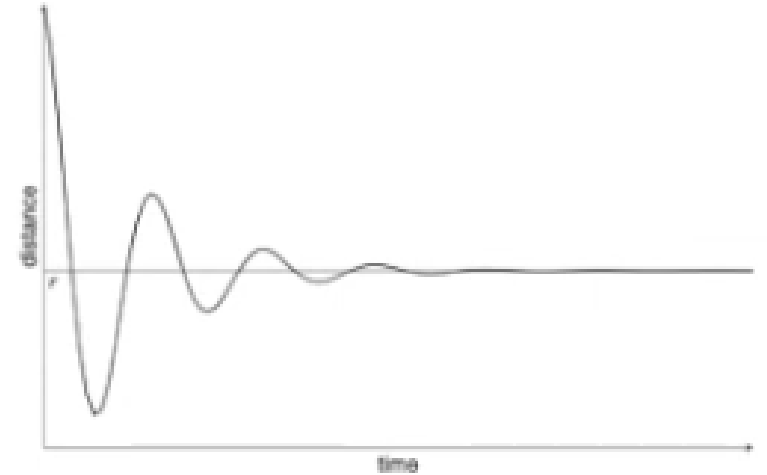
$$u(t) = \kappa_p e(t) + \kappa_i \int_0^t e(\tau) d\tau$$

- e.g., in presence of friction, error will be integrated causing higher motor setting to overcome remaining delta

- PID-controller:

- take into account **future error** by computing rate of error's change
- act as a “dampener” on control effort

$$u(t) = \kappa_p e(t) + \kappa_i \int_0^t e(\tau) d\tau + \kappa_d \frac{de(t)}{dt}$$



Open-Loop vs Closed-Loop

- **Closed-loop** is much more robust to external perturbation:
 - Noisy sensors: wrong estimate of the goal position, wrong estimate of the robot position
 - Noisy actuation: robot does not move precisely
 - Unforeseen events, dynamic obstacles
- **Open-loop** is only useful when feedback is not possible:
 - Sensors cannot operate in certain circumstances
 - Limited bandwidth
 - Limited computational resources



Model Predictive Control

Model Predictive Control

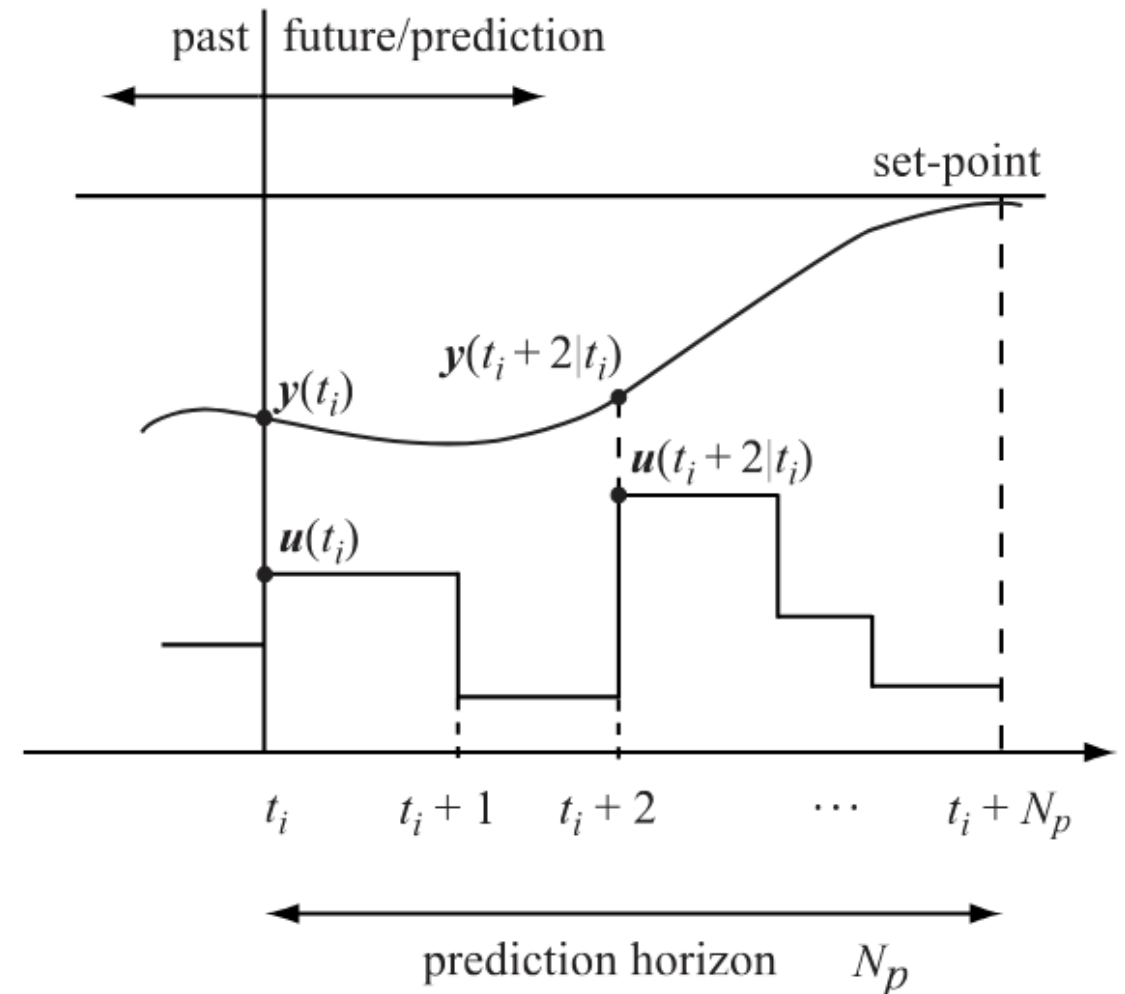
- **Model Predictive Control (MPC)** is a control scheme where a model is used for predicting the future behavior of the system over finite time window, the horizon
- Based on these predictions and the current measured/estimated state of the system, the optimal control inputs with respect to a defined control objective and subject to system constraints is computed
- After a certain time interval, the measurement, estimation and computation process is repeated with a shifted horizon

Model Predictive Control

- In the MPC approach, the current control action is computed on-line rather than using a pre-computed, off-line, control law
- A model predictive controller uses, at each sampling instant, the plant's current input and output measurements, the plant's current state, and the plant's model to
 - calculate, over a finite horizon, a future control sequence that optimizes a given performance index and satisfies constraints on the control action
 - use the first control in the sequence as the plant's input

Model Predictive Control

- N_p is the prediction horizon
- $u(t + k|t)$ is the predicted control action at $(t + k)$ given $u(t)$
- $y(t + k|t)$ is the predicted output at $(t + k)$ given $y(t)$



Model Predictive Control

- Major advantages of MPC in comparison to traditional reactive control approaches, e.g. PID, etc. are
 - **Proactive control action:** The controller is anticipating future disturbances, set-points etc.
 - **Non-linear control:** MPC can explicitly consider non-linear systems without linearization
 - **Constrained formulation:** Explicitly consider physical, safety or operational system constraints

Introduction to Marine Craft Model

Introduction to Marine Craft Model

- **Fossen's marine craft model** provides a mathematical framework for expressing the nonlinear equations of motion of marine craft in a compact matrix–vector form
- **The matrix-vector representation** is particularly valuable for **designing Guidance, Navigation, and Control (GNC) systems**. It is widely applied to marine craft, including ships, floating offshore structures, submarines, autonomous underwater vehicles (AUVs), and **uncrewed surface vehicles (USVs)**.

Introduction to Marine Craft Model

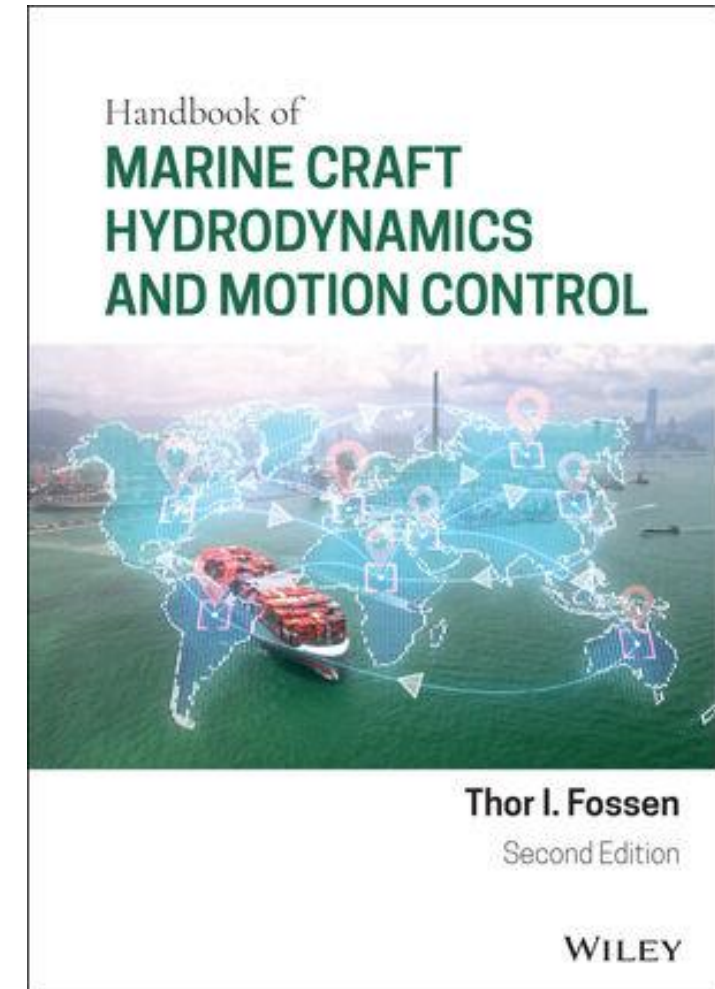
**Handbook of Marine Craft Hydrodynamics and Motion Control
(2nd ed.) Wiley**

Fossen, T. I. (2021)

Textbook supplements: <https://wiley.fossen.biz>

MSS Toolbox: <https://github.com/cybergalactic/MSS>

Python Vehicle Simulator: <https://github.com/cybergalactic/PythonVehicleSimulator>



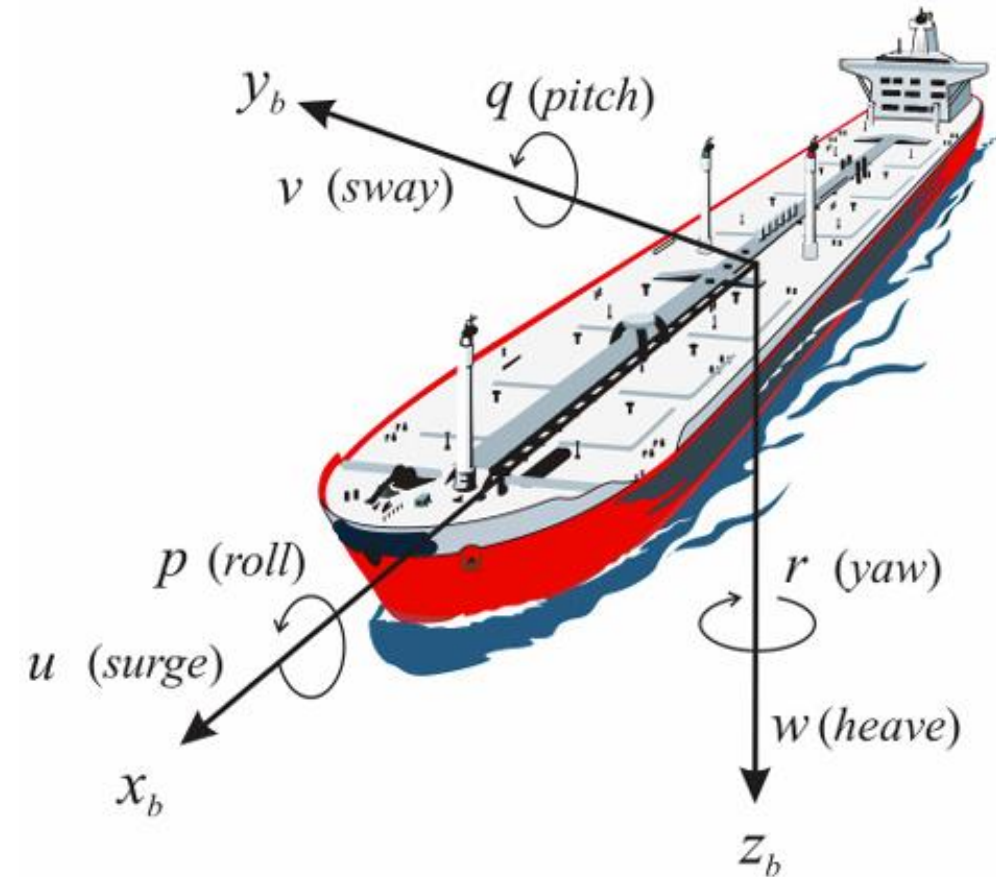
Marine Craft

In Encyclopedia Britannica, a ship and a boat are distinguished by their size through the following definition:

- **Ship:** any large floating vessel capable of crossing open waters, as opposed to a boat, which is generally a smaller craft
- **Submarine:** "any naval vessel that is capable of propelling itself beneath the water as well as on the water's surface"
- **Underwater Vehicle:** "small vehicle that is capable of propelling itself beneath the water surface as well as on the water's surface"
 - This includes unmanned underwater vehicles (UUV), remotely operated vehicles (ROV) and autonomous underwater vehicles (AUV)

Degrees of Freedom (DOFs)

- For a marine craft, DOF is the set of **independent displacements and rotations** that completely specify the **displaced position and orientation** of the craft
- A craft that can move freely in the 3-D space has maximum **6 DOFs—three translational and three rotational** components
- Consequently, a fully actuated marine craft operating in 6 DOFs must be equipped with **actuators that can produce independent forces and moments in all directions**

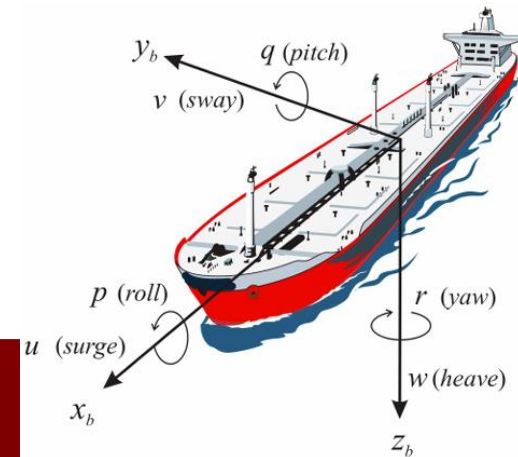


Degrees of Freedom (DOFs)

When designing **feedback control systems for marine craft**, reduced-order models are often used since most vehicles do not have actuation in all DOF

This is usually done by decoupling the motions of the vessel:

- **1-DOF** models can be used to design forward speed controllers (**surge**), heading autopilots (**yaw**) and roll damping systems (**roll**)

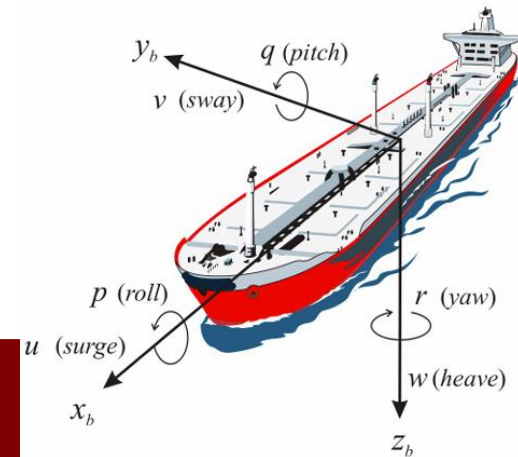


Degrees of Freedom (DOFs)

- **3-DOF** models are usually horizontal-plane models (**surge, sway and yaw**) for ships, semi-submersibles and underwater vehicles, trajectory-tracking control systems and path-following systems

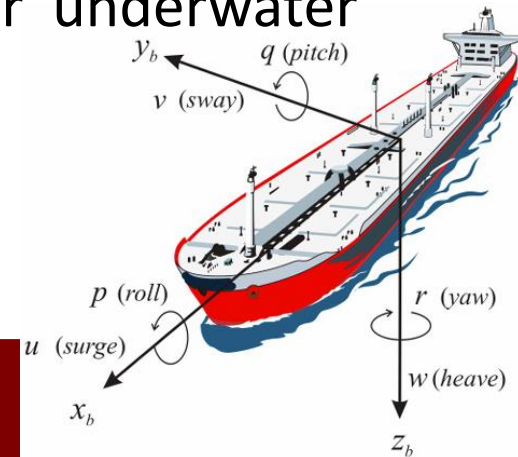
For slender bodies such as torpedo-shaped AUVs and submarines, it is also common to assume that the motions can be decoupled into longitudinal and lateral motions

- **Longitudinal models (surge, heave and pitch)** for forward speed, diving and pitch control
- **Lateral model (sway, roll and yaw)** for turning and heading control



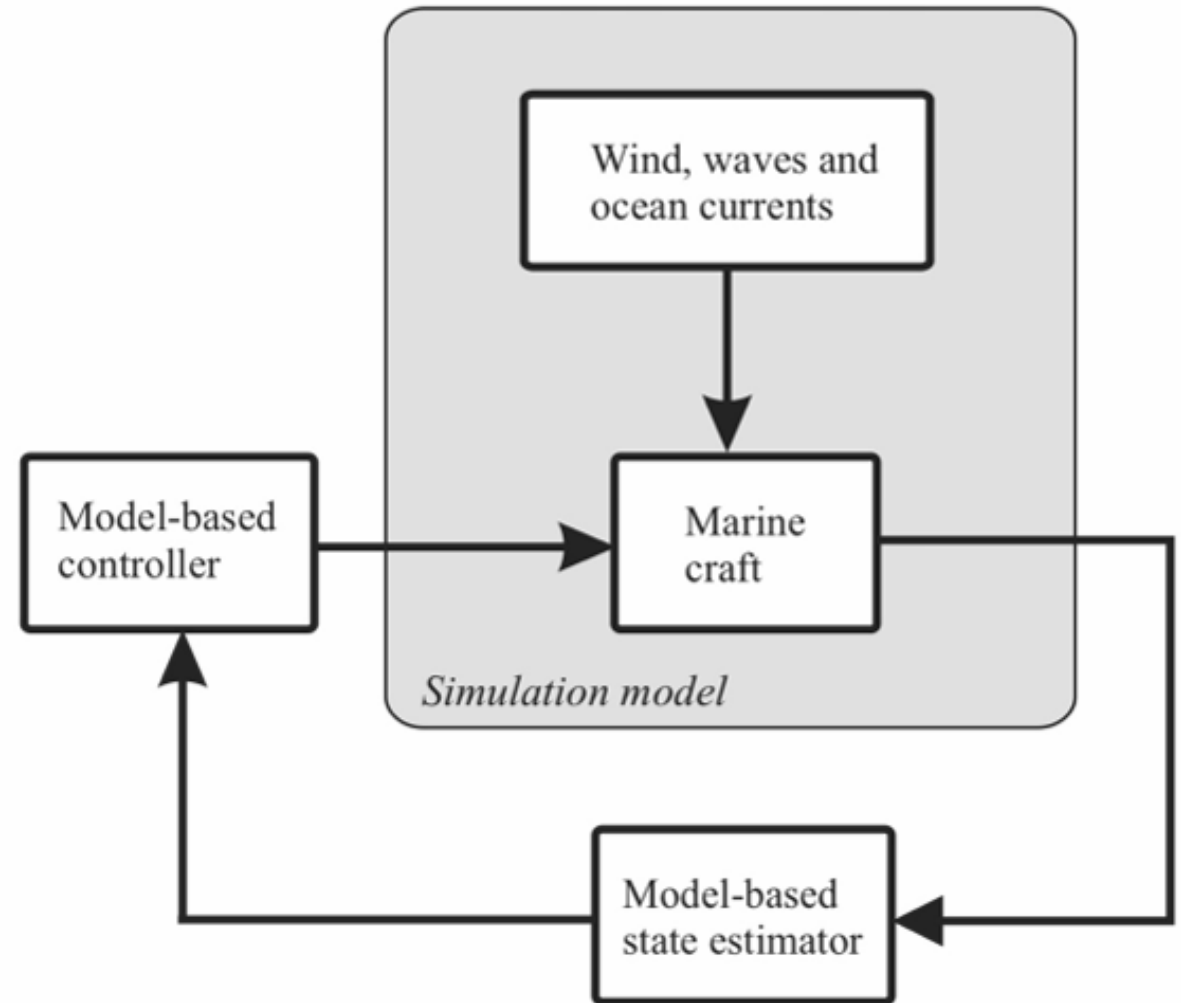
Degrees of Freedom (DOFs)

- **4-DOF models (surge, sway, roll and yaw)** are usually formed by adding the roll equation to the 3-DOF horizontal-plane model
 - These models are used in maneuvering situations where the purpose is to reduce roll by active control of fins, rudders or stabilizing liquid tanks
- **6-DOF models (surge, sway, heave, roll, pitch and yaw)** are fully coupled equations of motion used for simulation and prediction of coupled vessel motions
 - These models can also be used in advanced control systems for underwater vehicles, which are actuated in all DOFs



Classification of Models

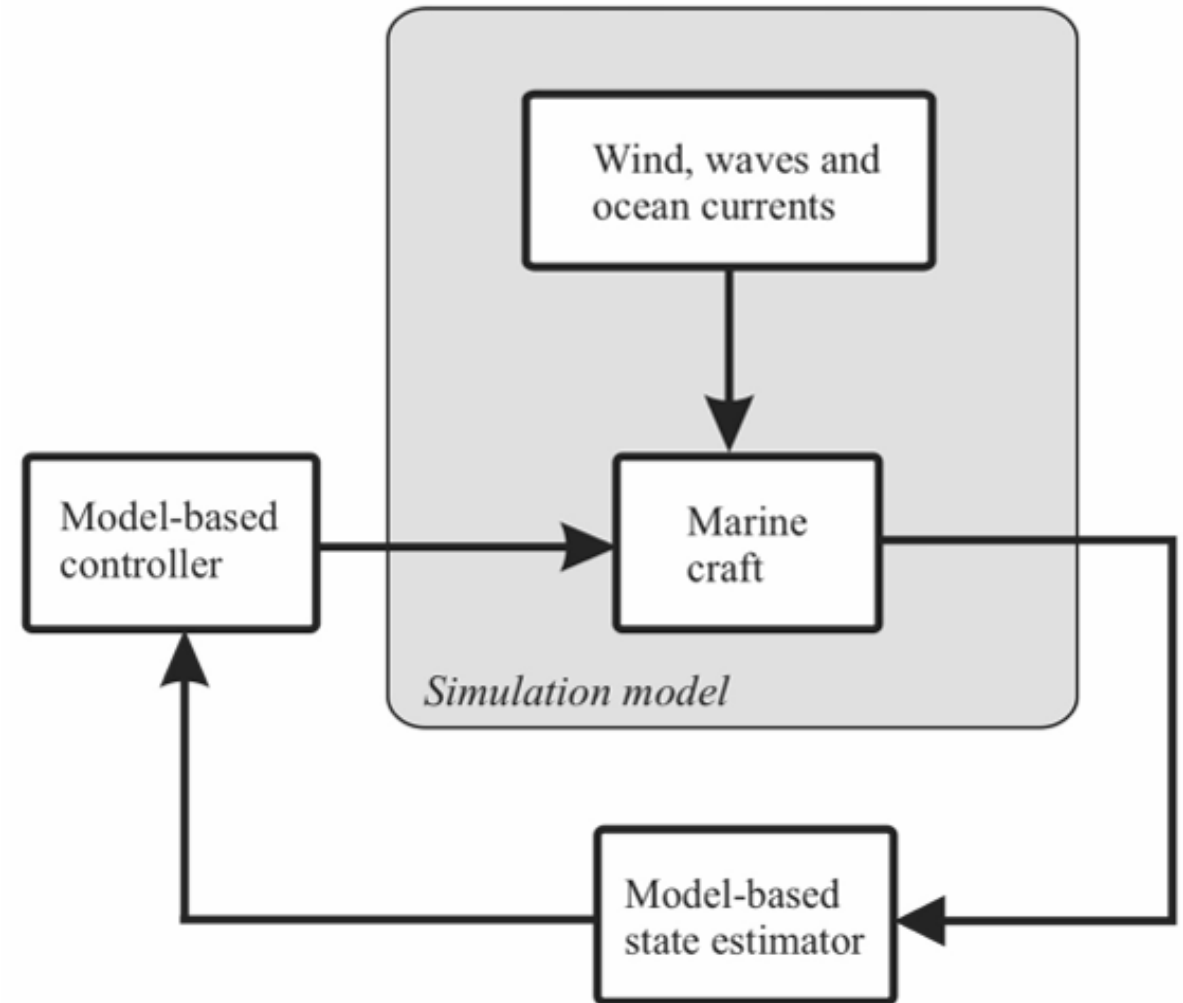
- **Simulation Model:** This model is the most accurate description of a system, for instance a 6-DOF high-fidelity model for simulation of coupled motions in the time domain



Classification of Models

- **Control Design Model:** The controller model is a reduced-order or simplified simulation model that is used to design the motion control system

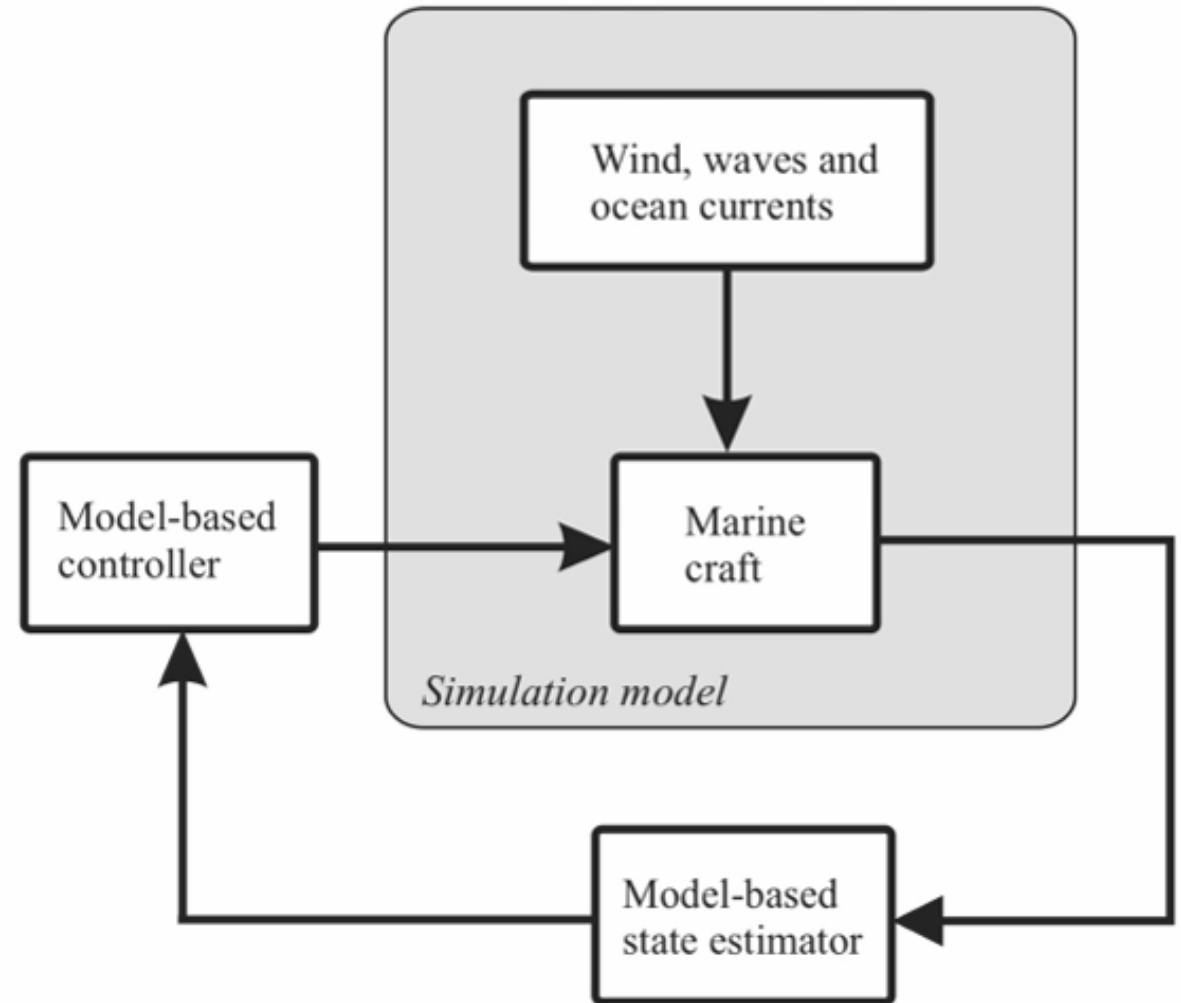
In its simplest form, this model is used to compute a set of constant gains for a PID controller



Classification of Models

- **State Estimator Design Model:**

Stochastic state estimators (Kalman filters) and deterministic state observers are both designed using **mathematical models** which are different from the models used in the simulator and **the controller** since the purpose is to capture the additional dynamics associated with the sensors and navigation system as well as disturbances



Guidance, Navigation and Control

Guidance is the action or the system that continuously computes the reference (desired) position, velocity and attitude of a marine craft to be used by the motion control system. These data are usually provided to the human operator and the navigation system.

How do I get there?
Actions required to reach
the destination

Guidance, Navigation and Control

Navigation is the science of directing a craft by determining its position/attitude, course and distance traveled. In some cases velocity and acceleration are determined as well.

Where am I?

Where do I want to go?

Which way should I go?

Planning a route

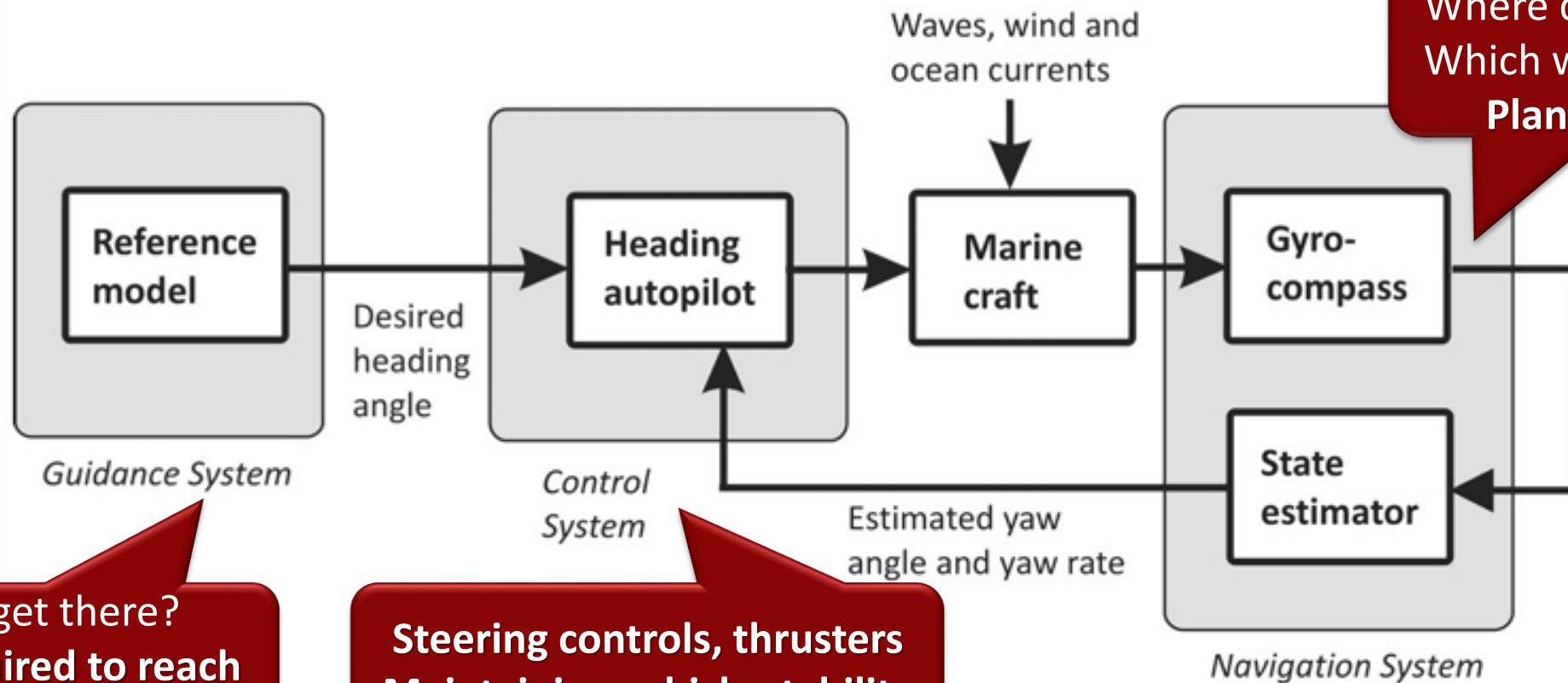
Guidance, Navigation and Control

Control, or more specifically motion control and control allocation, is the action of determining the necessary control forces and moments to be provided by the craft in order to satisfy a certain control objective.

Steering controls, thrusters
Maintaining vehicle stability

Open-Loop Guidance System

- GNC blocks with an open-loop guidance system represented by a reference model



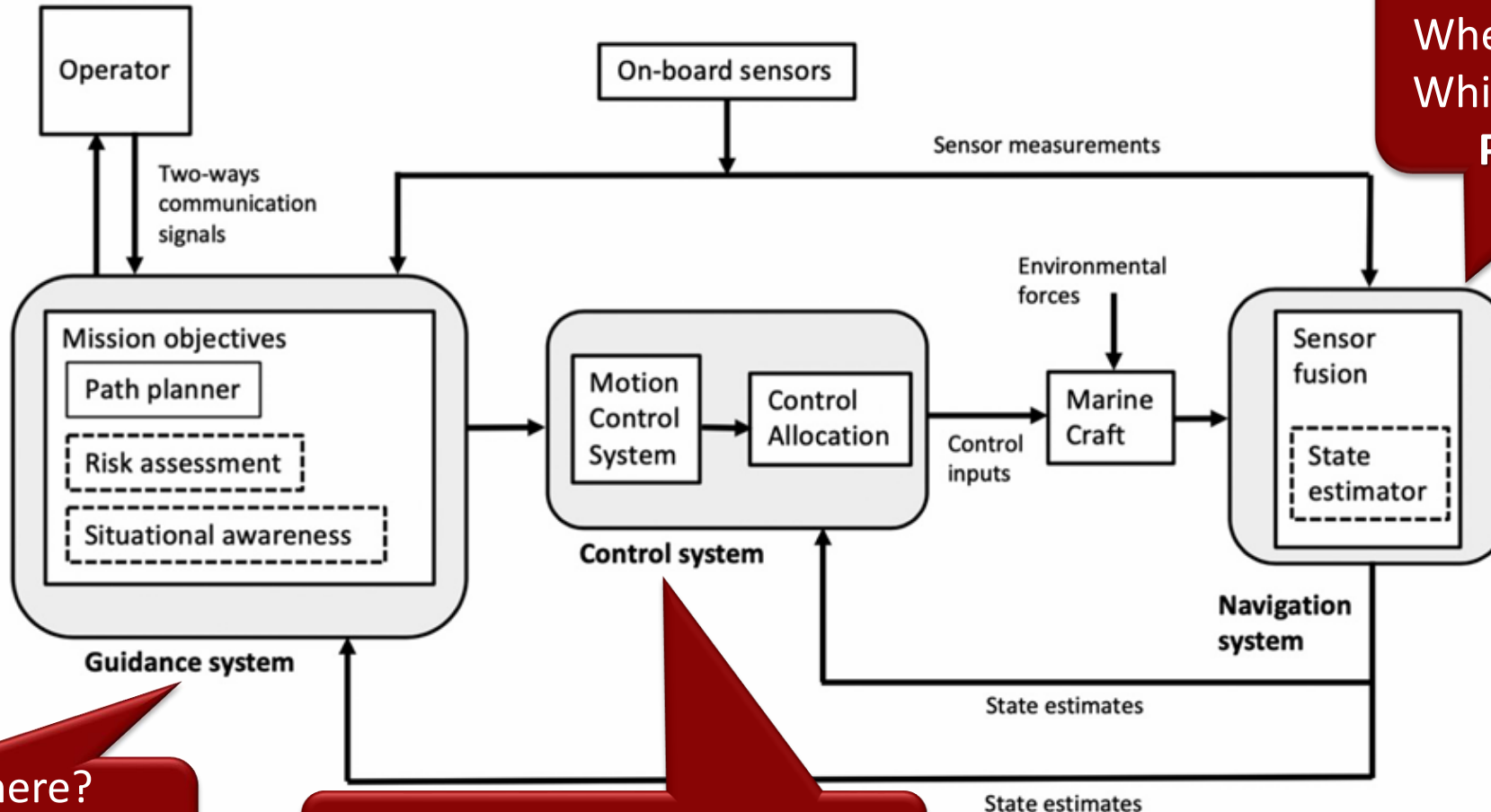
Where am I?
Where do I want to go?
Which way should I go?
Planning a route

How do I get there?
Actions required to reach
the destination

Steering controls, thrusters
Maintaining vehicle stability

Closed-Loop Guidance System

- GNC blocks where the guidance system makes use of the estimated alternatively measured positions and velocities



Where am I?
Where do I want to go?
Which way should I go?
Planning a route

How do I get there?
Actions required to reach
the destination

Steering controls, thrusters
Maintaining vehicle stability

Thank you very much for your attention!