

Introduction to Aerial Robotics

Lecture 3

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10 March 2020

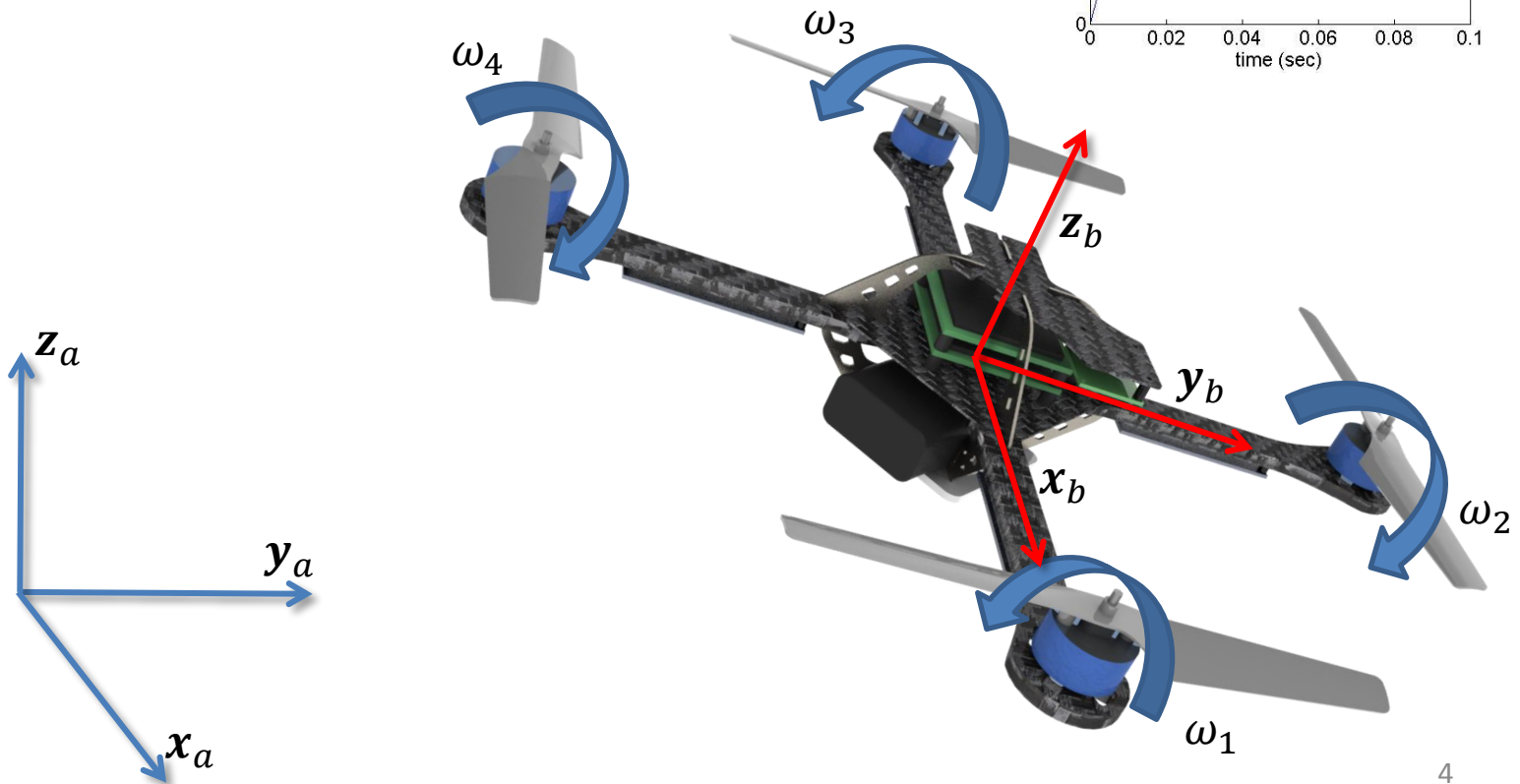
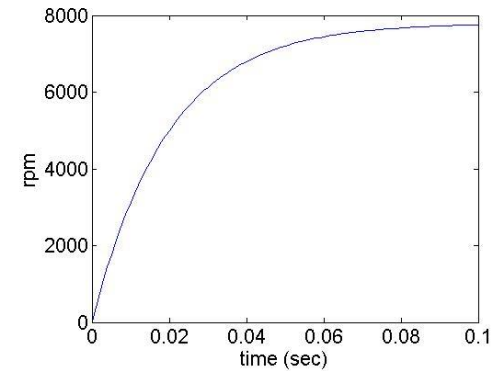
Outline

- Review: Quadrotor Dynamics
- Control Basics
- Quadrotor Control
- Trajectory Generation

Review: Quadrotor Dynamics

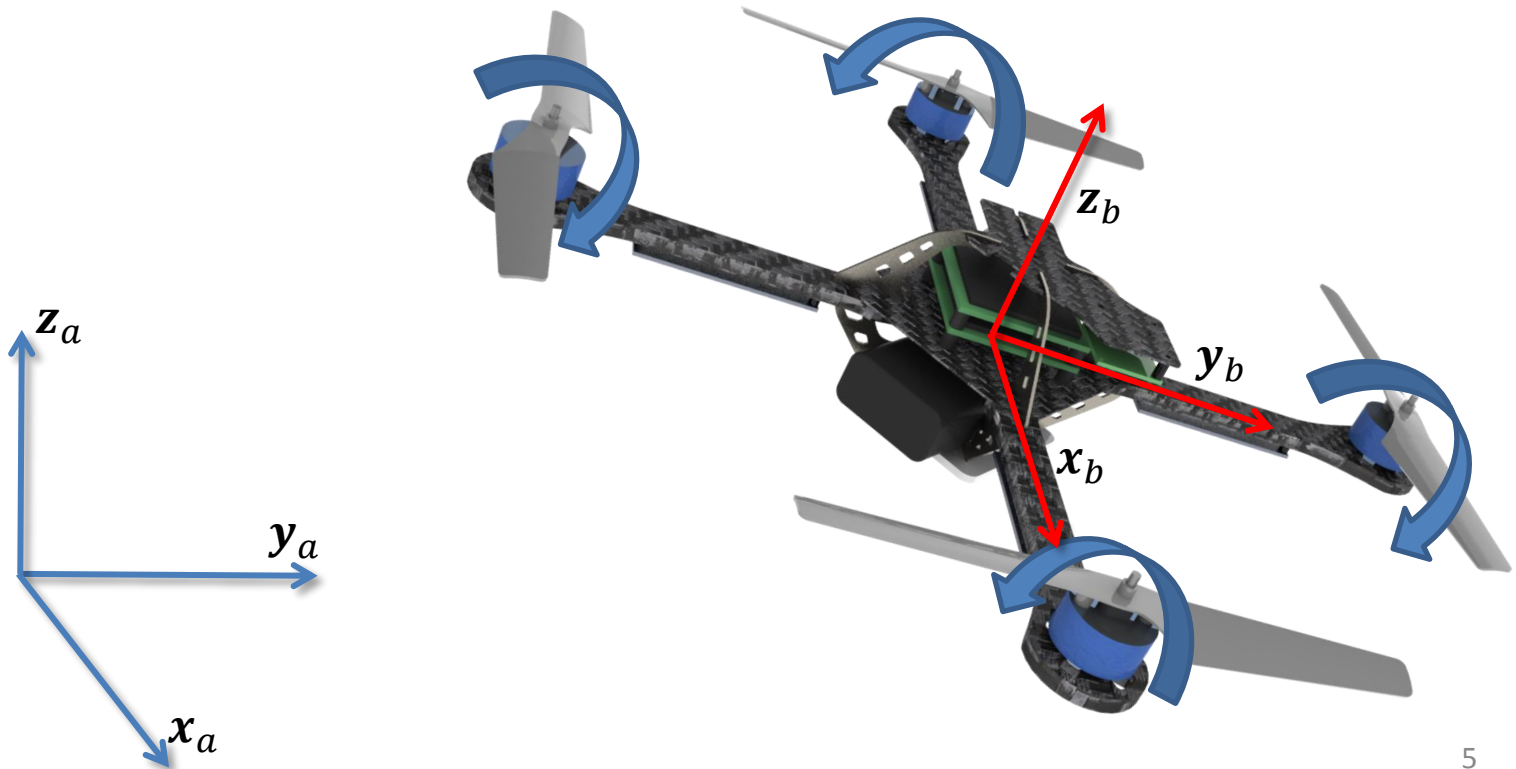
Quadrotor Dynamics

- Motor model: $\dot{\omega}_i = k_m(\omega_i^{des} - \omega_i)$
- Thrust from individual motor: $F_i = k_F \omega_i^2$
- Moment from individual motor: $M_i = k_M \omega_i^2$



Quadrotor Dynamics

- Z-X-Y Euler Angles: $\mathbf{R}_{ab} = \mathbf{R}_z(\psi) \cdot \mathbf{R}_x(\phi) \cdot \mathbf{R}_y(\theta)$
- Sequence of three rotations about body-fixed axes

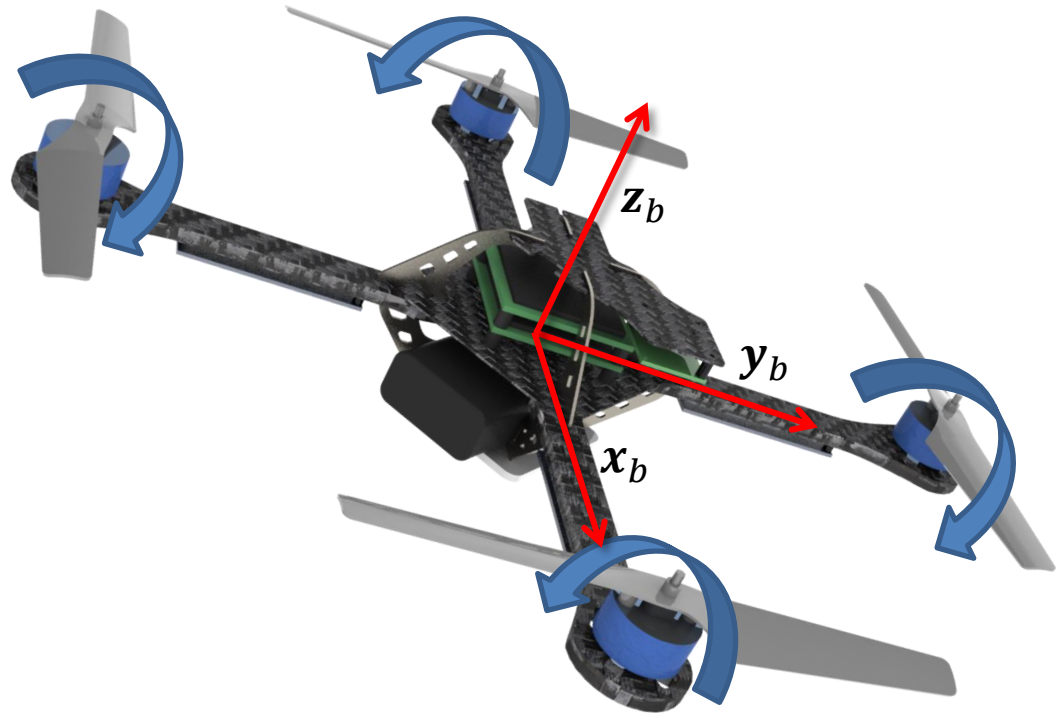
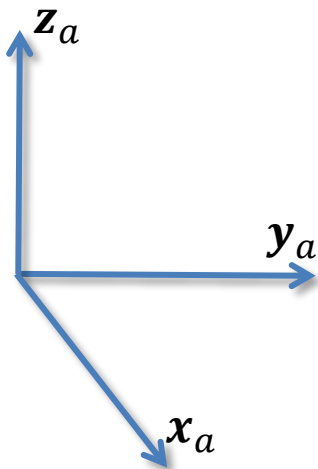


Quadrotor Dynamics

$$\bullet \quad \mathbf{R}_{ab} = \begin{bmatrix} c\psi c\theta - s\phi s\psi s\theta & -c\phi s\psi & c\psi s\theta + c\theta s\phi s\psi \\ c\theta s\psi + c\psi s\phi s\theta & c\phi c\psi & s\psi s\theta - c\psi c\theta s\phi \\ -c\phi s\theta & s\phi & c\phi c\theta \end{bmatrix}$$

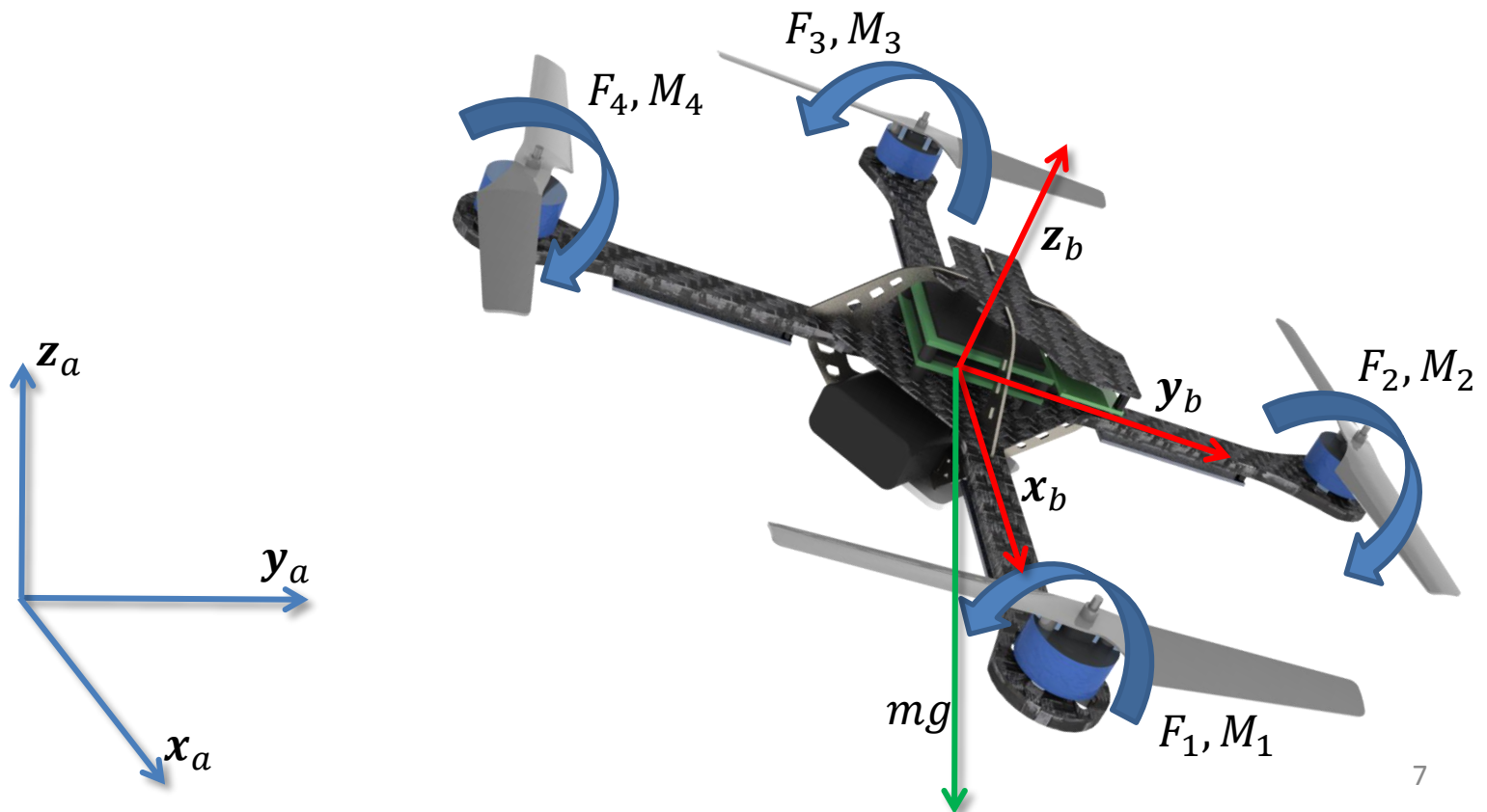
$$\bullet \quad \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} = \begin{bmatrix} c\theta & 0 & -c\phi s\theta \\ 0 & 1 & s\phi \\ s\theta & 0 & c\phi c\theta \end{bmatrix} \begin{bmatrix} \dot{\phi} \\ \dot{\theta} \\ \dot{\psi} \end{bmatrix}$$

Instantaneous body
angular velocity viewed
in the body frame.



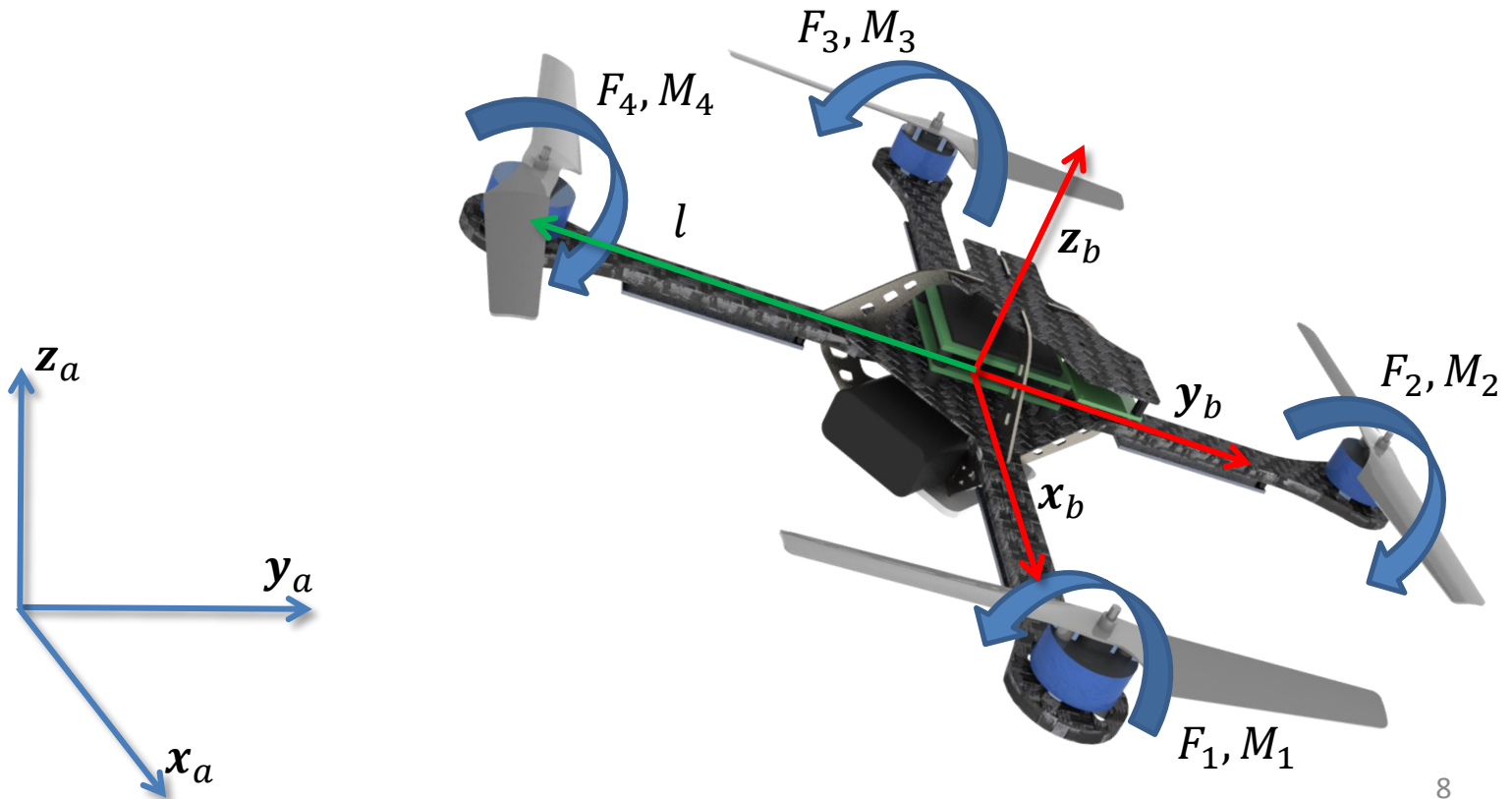
Newton-Euler Equations

- Newton Equation: $m\ddot{\mathbf{p}}^a = \begin{bmatrix} 0 \\ 0 \\ -mg \end{bmatrix} + \mathbf{R}_{ab} \begin{bmatrix} 0 \\ 0 \\ F_1 + F_2 + F_3 + F_4 \end{bmatrix}$



Newton-Euler Equations

- Euler Equation:
$$\mathbf{I}^b \cdot \begin{bmatrix} \dot{\omega}_x \\ \dot{\omega}_y \\ \dot{\omega}_z \end{bmatrix} + \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} \times \mathbf{I}^b \cdot \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} = \begin{bmatrix} l(F_2 - F_4) \\ l(F_3 - F_1) \\ M_1 - M_2 + M_3 - M_4 \end{bmatrix}$$

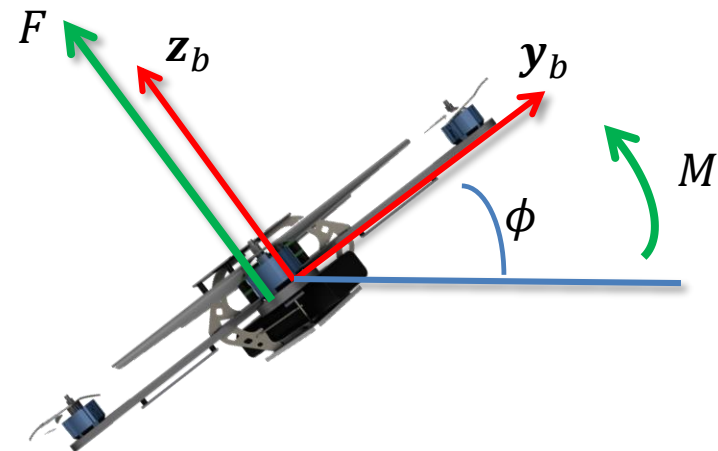
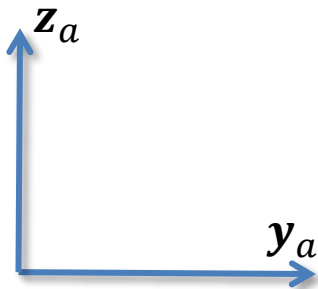


Quadrotor Dynamics

- Motor model: $\dot{\omega}_i = k_m(\omega_i^{des} - \omega_i)$
- Thrust from individual motor: $F_i = k_F \omega_i^2$
- Moment from individual motor: $M_i = k_M \omega_i^2$
- Euler Equation: $\mathbf{I}^b \cdot \begin{bmatrix} \dot{\omega}_x \\ \dot{\omega}_y \\ \dot{\omega}_z \end{bmatrix} + \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} \times \mathbf{I}^b \cdot \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} = \begin{bmatrix} l(F_2 - F_4) \\ l(F_3 - F_1) \\ M_1 - M_2 + M_3 - M_4 \end{bmatrix}$
- Angular velocities to derivatives of Euler angles: $\begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} = \begin{bmatrix} c\theta & 0 & -c\phi s\theta \\ 0 & 1 & s\phi \\ s\theta & 0 & c\phi c\theta \end{bmatrix} \begin{bmatrix} \dot{\phi} \\ \dot{\theta} \\ \dot{\psi} \end{bmatrix}$
- Rotation matrix: $\mathbf{R}_{ab} = \begin{bmatrix} c\psi c\theta - s\phi s\psi s\theta & -c\phi s\psi & c\psi s\theta + c\theta s\phi s\psi \\ c\theta s\psi + c\psi s\phi s\theta & c\phi c\psi & s\psi s\theta - c\psi c\theta s\phi \\ -c\phi s\theta & s\phi & c\phi c\theta \end{bmatrix}$
- Newton Equation: $m\ddot{\mathbf{p}}^a = \begin{bmatrix} 0 \\ 0 \\ -mg \end{bmatrix} + \mathbf{R}_{ab} \begin{bmatrix} 0 \\ 0 \\ F_1 + F_2 + F_3 + F_4 \end{bmatrix}$

A Planar Quadrotor

$$\bullet \begin{bmatrix} \ddot{y} \\ \ddot{z} \\ \ddot{\phi} \end{bmatrix} = \begin{bmatrix} 0 \\ -g \\ 0 \end{bmatrix} + \begin{bmatrix} -\frac{1}{m} \sin \phi & 0 \\ \frac{1}{m} \cos \phi & 0 \\ 0 & \frac{1}{I_{xx}} \end{bmatrix} \begin{bmatrix} F \\ M \end{bmatrix}$$

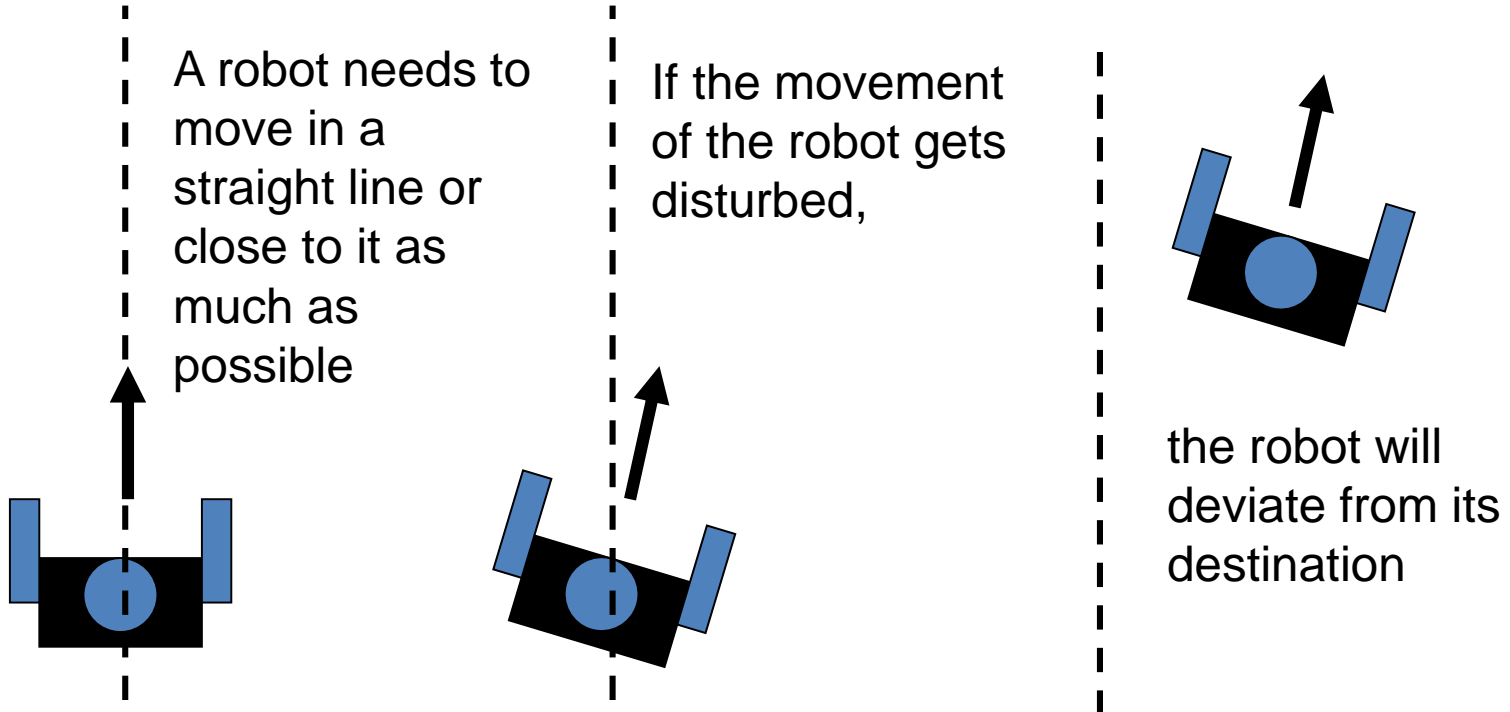


Control System Design

Control System Design

- Introduction of control systems
- Linear Time Invariant (LTI) systems
 - Simple first-order system
 - Simple second-order system
- Controller design
 - Gain tuning
 - Model-based control

Introduction of control system



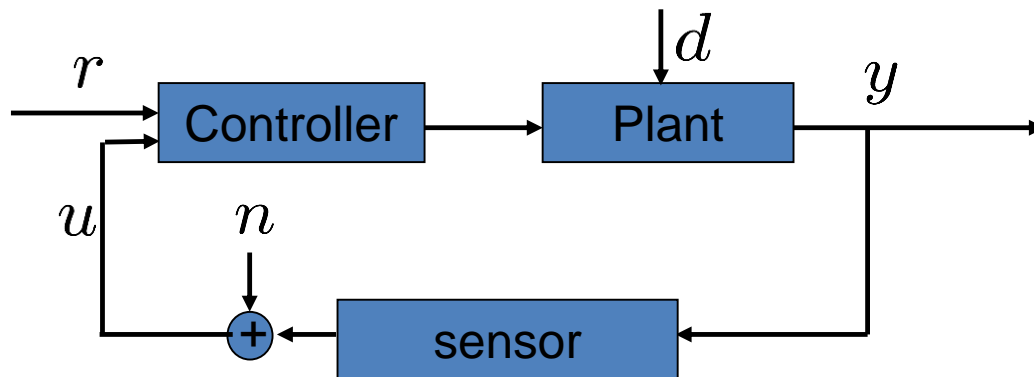
- ❖ Therefore we need to have a controller to control its movement in real time based on its movement and the destination

Introduction of control system

- Open loop control
 - Move the robot in a pre-determined way
 - Example: walking with your eyes closed
- Closed loop (feedback) control
 - Use the output (i.e. the location of the robot) to adjust the input (i.e. the direction and may be speed) to the movement of the robot
 - We also call it feedback control, since we make the control decision based on the output feedback
 - Example: walking with your eyes open
- We want to stabilize a system with closed loop control

Introduction of control system

- One objective of control is to make the plant stable and track a given reference signal as precise and swift as possible



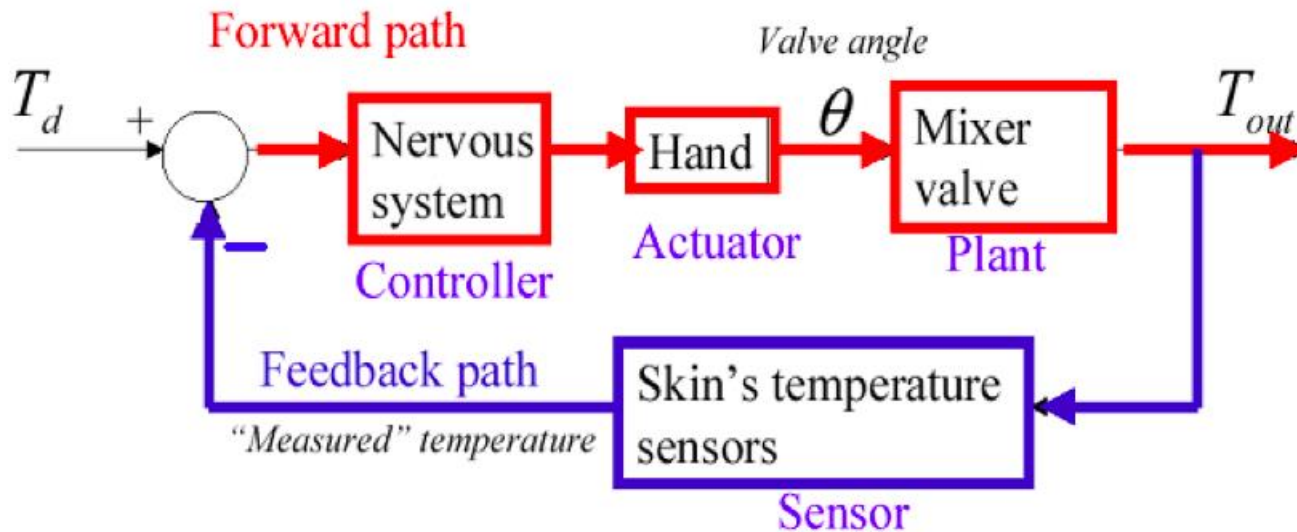
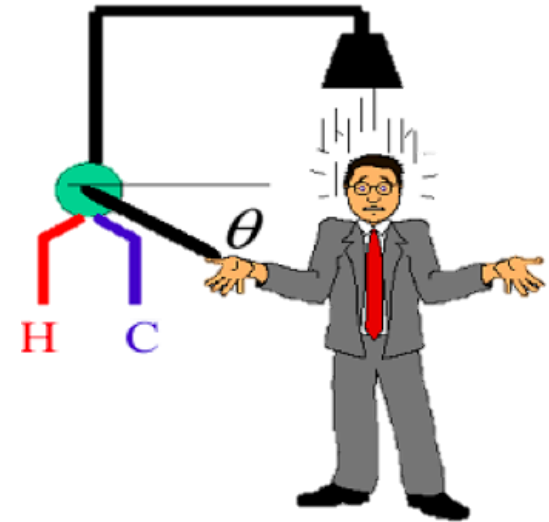
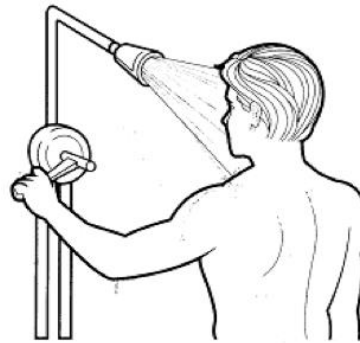
r : reference input
 u : controller input
 d : plant disturbance
 y : output
 n : communication noise

- A controller is simply a computation unit that computes the “optimal” or “desired” input to the plant

“Feedback is a method of controlling a system by inserting into it the result of its past performance”

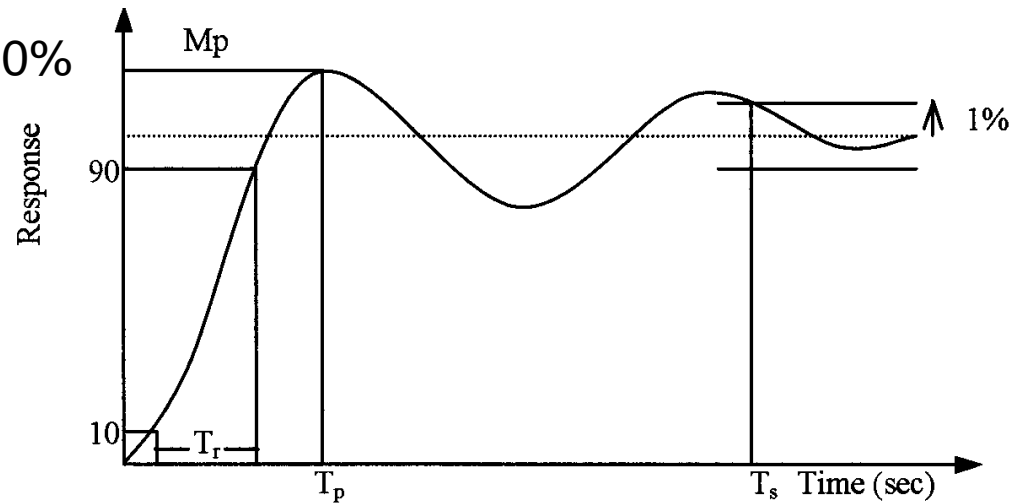
Introduction of control system

When taking a shower



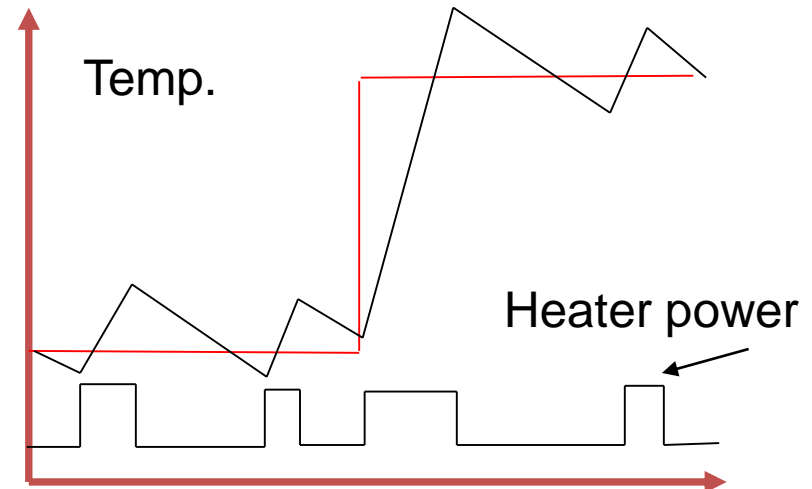
Introduction of control system

- Rise time:
 - Time it takes from 10% to 90%
- Steady-state error
- Overshoot
 - Percentage by which peak exceeds final value
- Settling time
 - Time it takes to reach 1% of final value
- A good control system has small rise time, overshoot, settling time and steady-state error



Introduction of control system

- Example: shower water temperature control
 - Turn the heater on if T_{water} is below certain value
 - Turn the heater off if T_{water} is above certain value
- Simple
- Transition is not smooth



Control of a simple first-order system

- Problem

State, input

$$x, u \in \mathbf{R}$$

Kinematic plant model

$$\dot{x} = u$$

Want x to follow trajectory $x^{des}(t)$

- General Approach

Define error, $e(t) = x^{des}(t) - x(t)$

Want $e(t)$ to converge exponentially to zero

- Strategy

Find u such that

$$\dot{e} + K_p e = 0 \quad K_p > 0$$

$$u(t) = \dot{x}^{des}(t) + K_p e(t)$$



Feed forward Proportional

Control of a simple second-order system

- Problem

State, input

$$x, u \in \mathbf{R}$$

Kinematic plant model

$$\ddot{x} = u$$

Want x to follow trajectory $x^{des}(t)$

- General Approach

Define error, $e(t) = x^{des}(t) - x(t)$

Want $e(t)$ to converge exponentially to zero

- Strategy

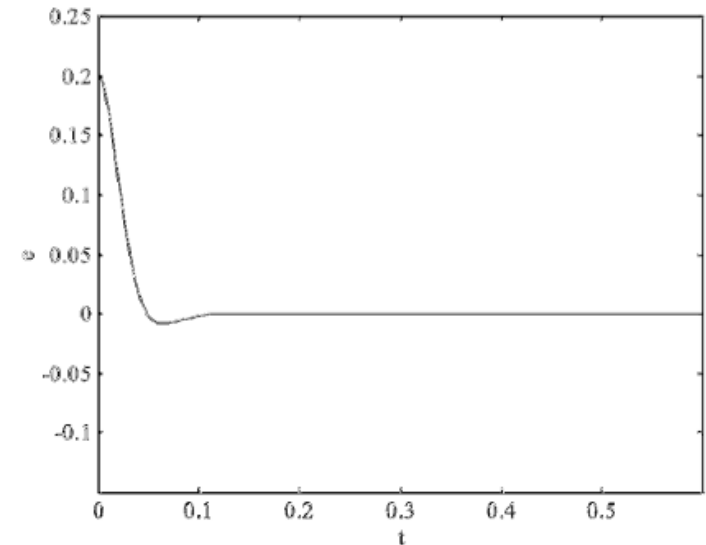
Find u such that

$$\ddot{e} + K_d \dot{e} + K_p e = 0 \quad K_d, K_p > 0$$

$$u(t) = \ddot{x}^{des}(t) + K_d \dot{e}(t) + K_p e(t)$$



Feed forward Derivative Proportional



PD control and PID control

- PD control

$$u(t) = \ddot{x}^{des}(t) + K_d \dot{e}(t) + K_p e(t)$$

Proportional control acts like a spring (capacitance) response

Derivative control is a viscous dashpot (resistance) response

Large derivative gain makes the system overdamped and the system converges slowly

- PID control

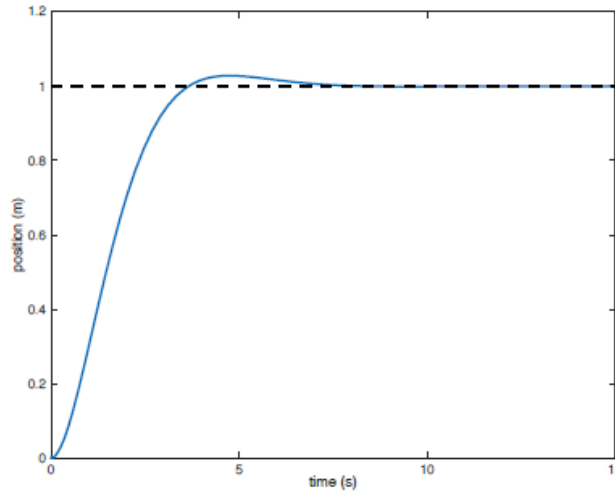
$$u(t) = \ddot{x}^{des}(t) + K_d \dot{e}(t) + K_p e(t) + K_I \int_0^t e(\tau) d\tau$$

↑ Integral

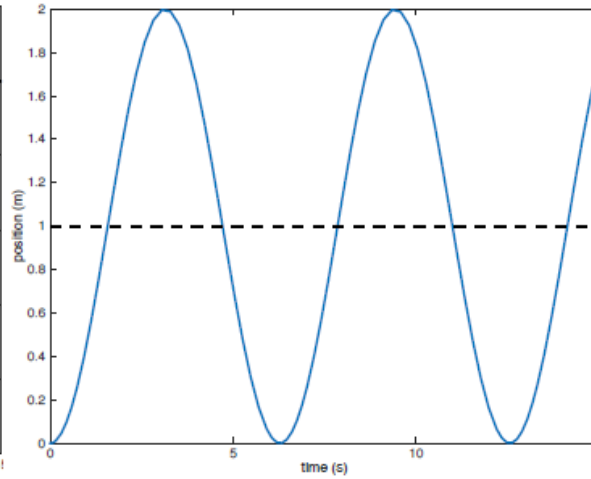
PID control generates a third-order closed-loop system

Integral control makes the steady-state error go to zero

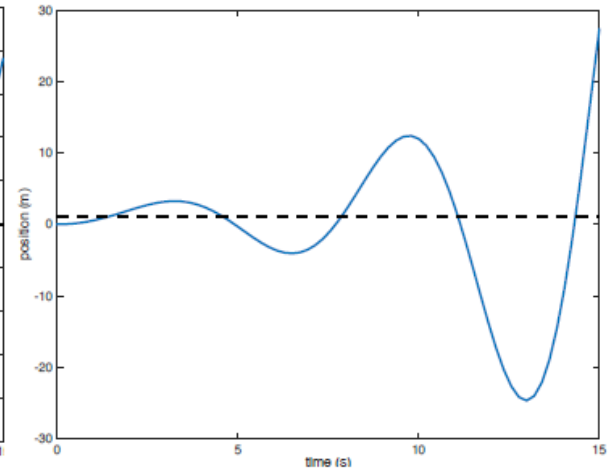
Gain Tuning



Stable
(converge)

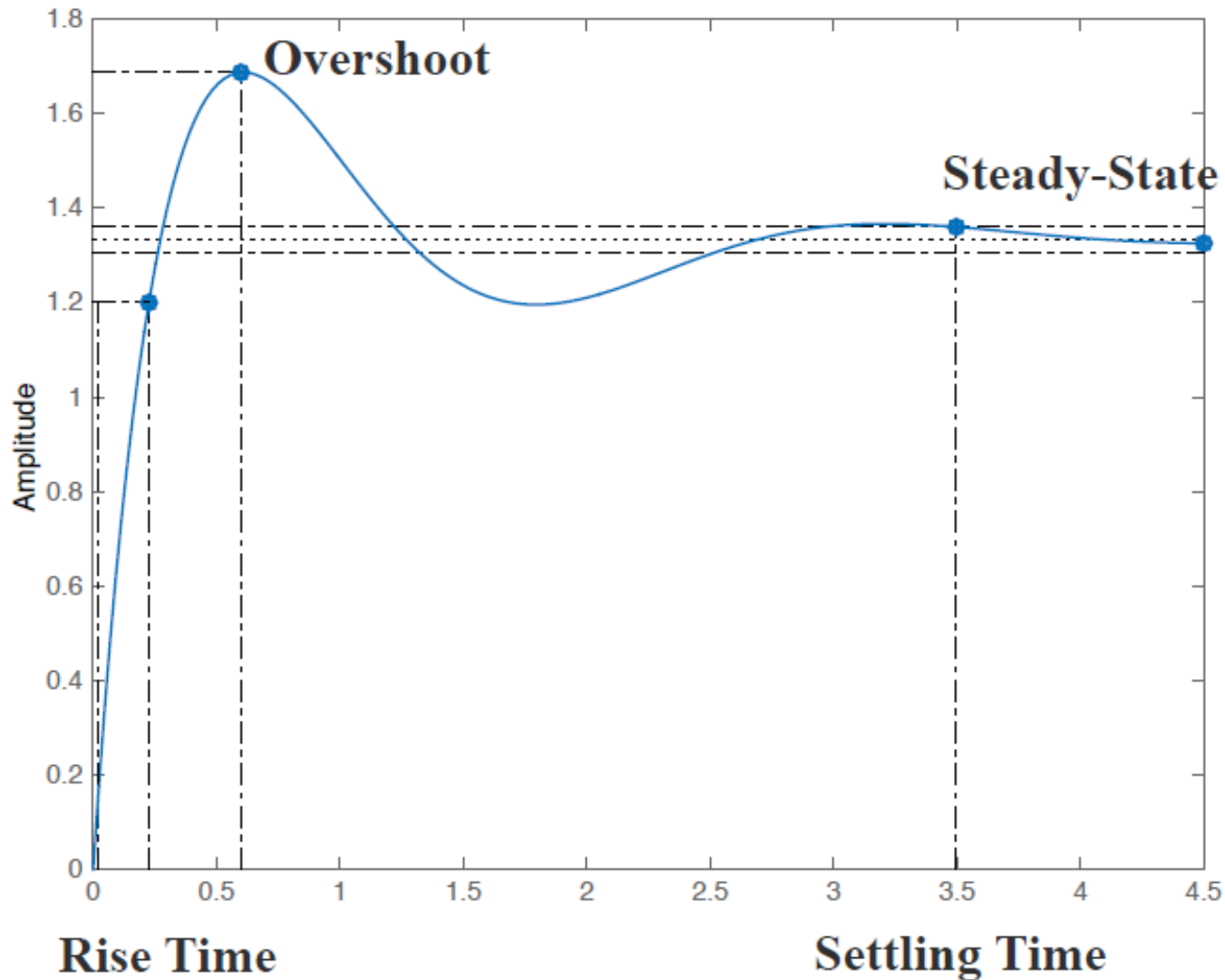


Marginally Stable
(oscillate)



Unstable
(diverge)

Manual Tuning

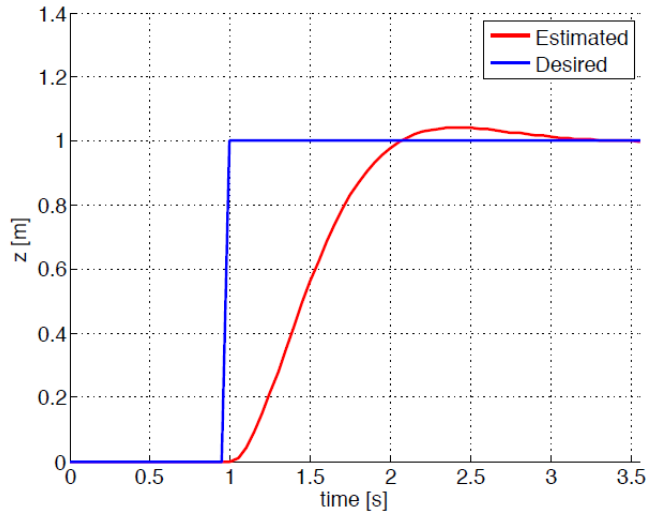


Manual Tuning

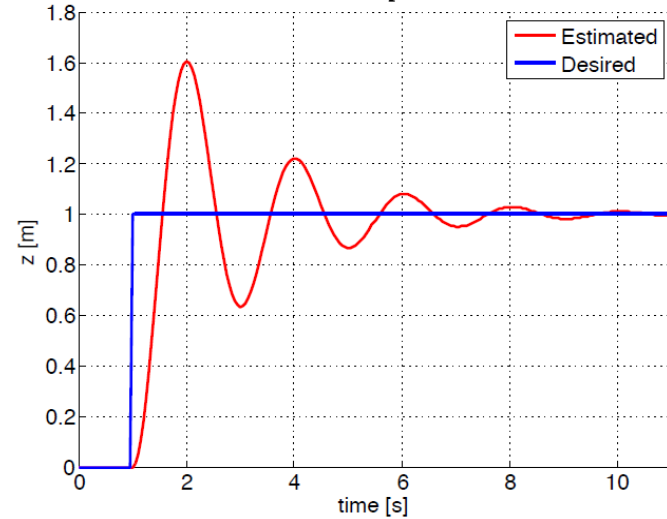
Parameter Increased	$K_p \uparrow$	$K_d \uparrow$	$K_i \uparrow$
Rise Time	Decrease	-	Decrease
Overshoot	Increase	Decrease	Increase
Setting Time	-	Decrease	Increase
Steady-State Error	Decrease	-	Eliminate

Manual Tuning

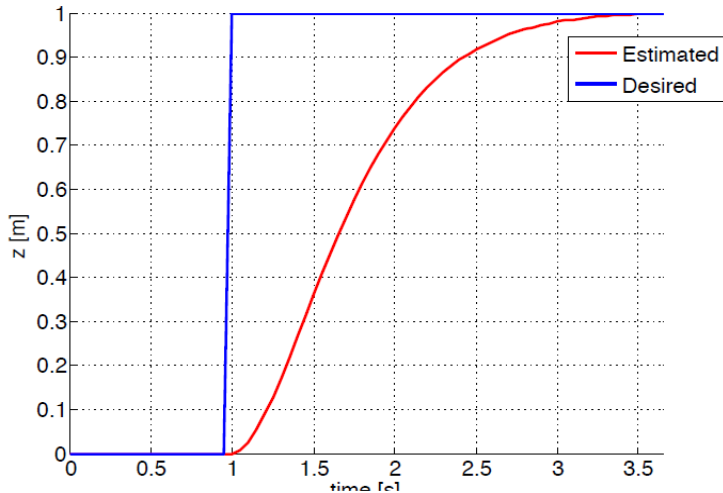
PD controller



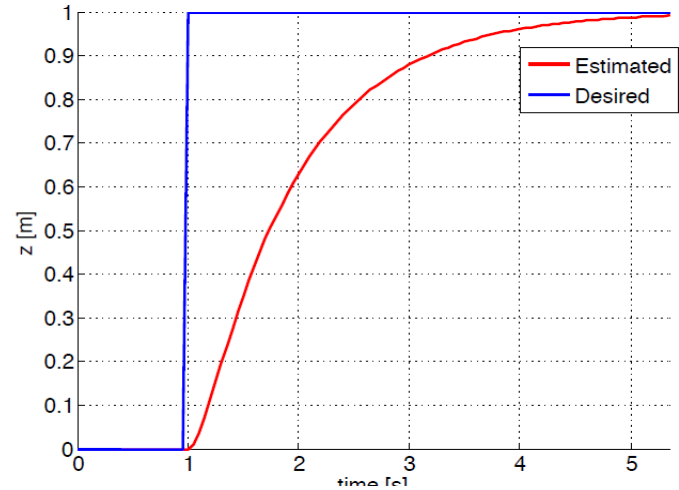
High K_p (overshoot)



Low K_p (soft response)



High K_d (overdamped)



Ziegler-Nichols Method

- Heuristic for tuning gains
 1. Set $K_i = K_d = 0$
 2. Increase K_p until ultimate gain, K_u , when output starts to oscillate
 3. Find the oscillation period T_u at K_u
 4. Set gains according to:

Controller	K_p	K_d	K_i
P	$0.5K_u$	-	-
PD	$0.8K_u$	$K_p T_u / 8$	-
PID	$0.6K_u$	$K_p T_u / 8$	$2K_p / T_u$

Model-based control

- Consider a general second-order model

$$m\ddot{x}(t) + b\dot{x}(t) + kx(t) = u(t)$$

- Disadvantages of PID or PD control schemes

$$u(t) = \ddot{x}^{des}(t) + K_d\dot{e}(t) + K_p e(t)$$

- Performance will depend on the model
- Need to tune gains to maximize performance

- Model based control law

Model based

$$u(t) = \underbrace{\hat{m}(\ddot{x}^{des}(t) + K_d\dot{e}(t) + K_p e(t))}_{\text{Servo: feedforward + PD feedback}} + \hat{b}\dot{x}(t) + \hat{k}x(t)$$

- Model based part
 - Cancel the dynamics of the system
 - Specific to the model
- Servo based part
 - Use PID or PD with feedforward to drive errors to zero
 - Independent of the model of the system

Model-based control

- Model based control law

$$u(t) = \underbrace{\hat{m}(\ddot{x}^{des}(t) + K_d \dot{e}(t) + K_p e(t))}_{\text{Servo: feedforward + PD feedback}} + \underbrace{\hat{b}\dot{x}(t) + \hat{k}x(t)}_{\text{Model based (estimates)}}$$

- Advantage

- Decomposes the control law into
 - Model-dependent part (depends on the knowledge of the model)
 - Model-independent part (servo control, gains are independent of the model)

- Disadvantage

- Based on estimates of model parameters

- Ideal performance

$$\ddot{e} + K_d \dot{e} + K_p e = 0$$

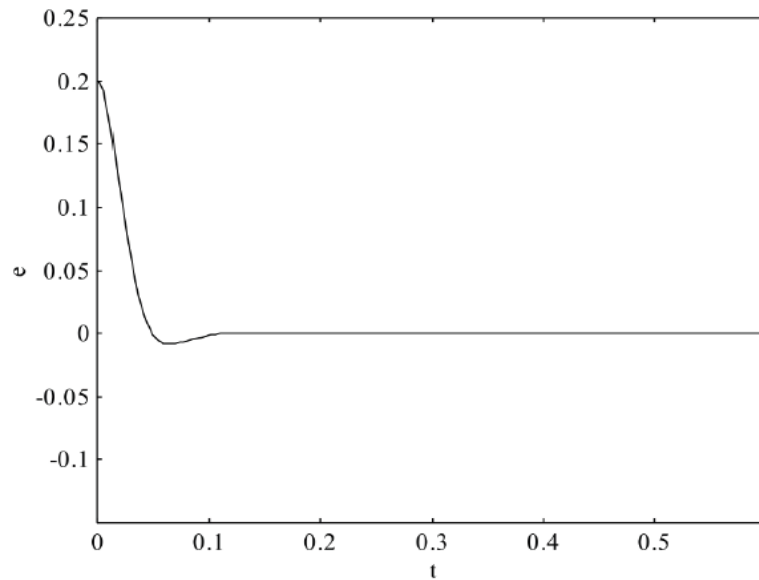
- Actual performance

$$\ddot{e} + K_d \dot{e} + K_p e = \left(\frac{m}{\hat{m}} - 1\right) \ddot{x} + \frac{(b - \hat{b})}{\hat{m}} \dot{x} + \frac{(k - \hat{k})}{\hat{m}} x$$

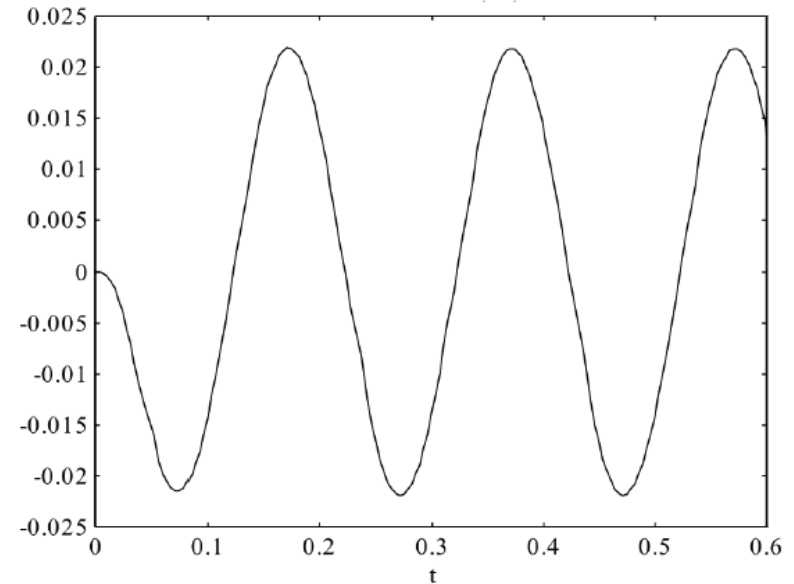
Model-based control

- Performance

$$u(t) = \hat{m}(\ddot{x}^{des}(t) + K_d \dot{e}(t) + K_p e(t)) + \hat{b}\dot{x}(t) + \hat{k}x(t)$$



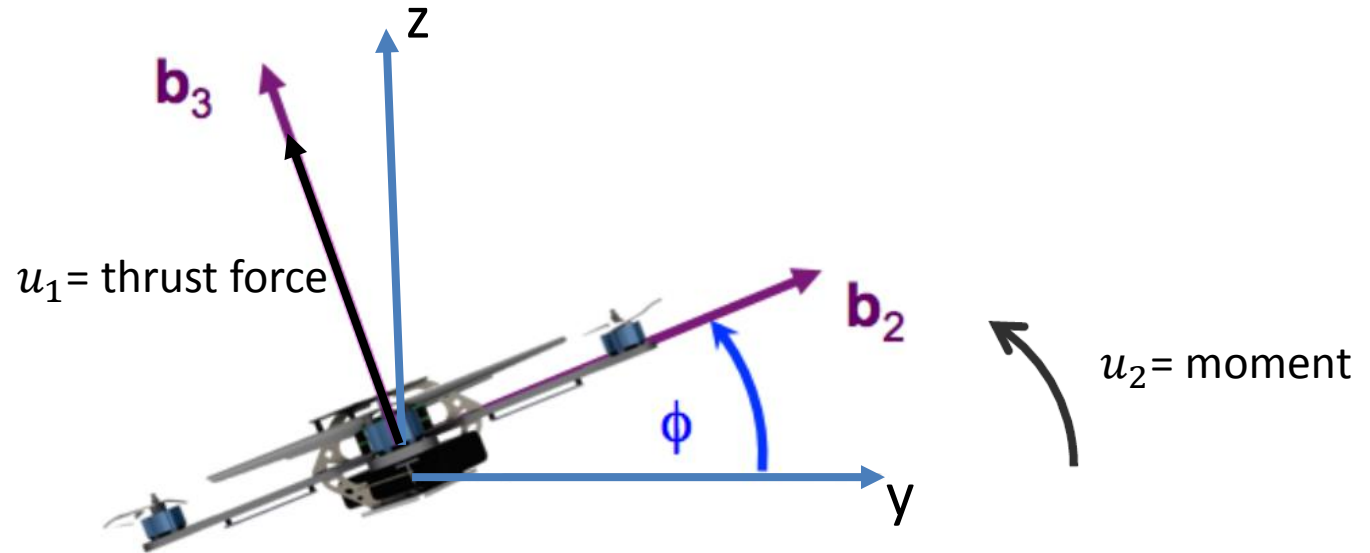
Perfect model



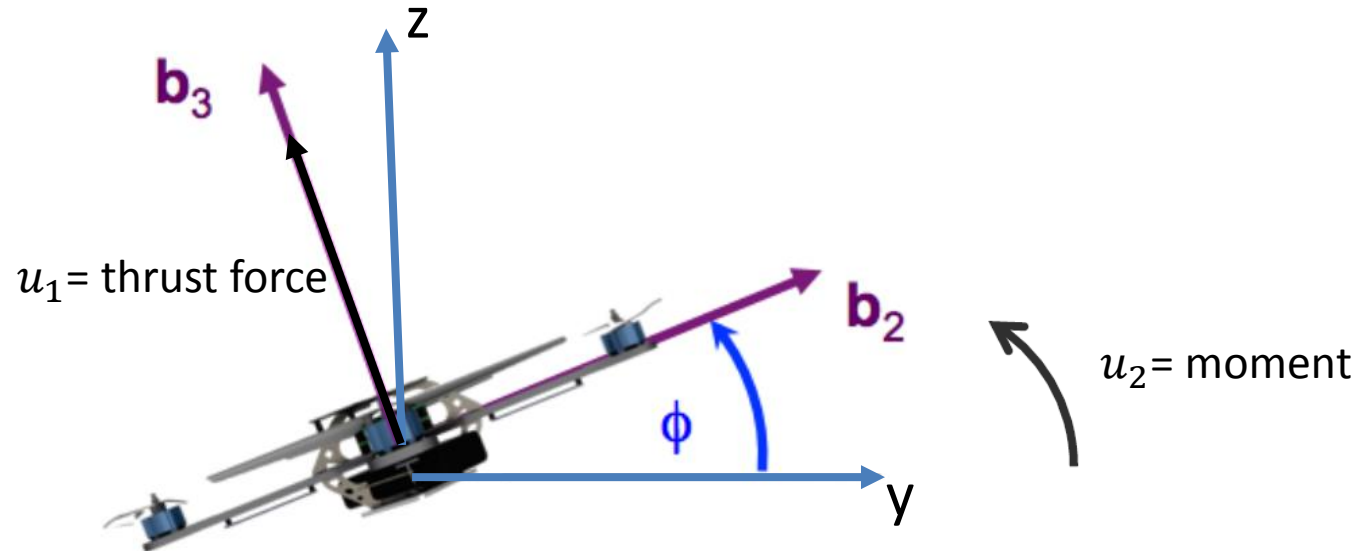
Imperfect model, 10% errors in parameters

Quadrotor Control

Application to Quadrotors



Planar Quadrotor Model



$$\begin{aligned}
 \sum F_y &= -u_1 \sin(\phi) = m\ddot{y} \\
 \sum F_z &= -mg + u_1 \cos(\phi) = m\ddot{z} \\
 M &= u_2 = I_{xx}\ddot{\phi}
 \end{aligned}
 \quad \Rightarrow \quad
 \begin{bmatrix} \ddot{y} \\ \ddot{z} \\ \ddot{\phi} \end{bmatrix} = \begin{bmatrix} 0 \\ -g \\ 0 \end{bmatrix} + \begin{bmatrix} -\frac{1}{m}\sin\phi & 0 \\ \frac{1}{m}\cos\phi & 0 \\ 0 & \frac{1}{I_{xx}} \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$

Linearized Dynamic Model

- Nonlinear dynamics

$$\begin{aligned}\ddot{y} &= -\frac{u_1}{m} \sin(\phi) \\ \ddot{z} &= -g + \frac{u_1}{m} \cos(\phi) \\ \ddot{\phi} &= \frac{u_2}{I_{xx}}\end{aligned}$$

- Equilibrium hover configuration

$$y_0, z_0, \phi_0 = 0, u_{1,0} = mg, u_{2,0} = 0$$

- Linearized dynamics

$$\ddot{y} = -g\phi$$

Cascaded second order system

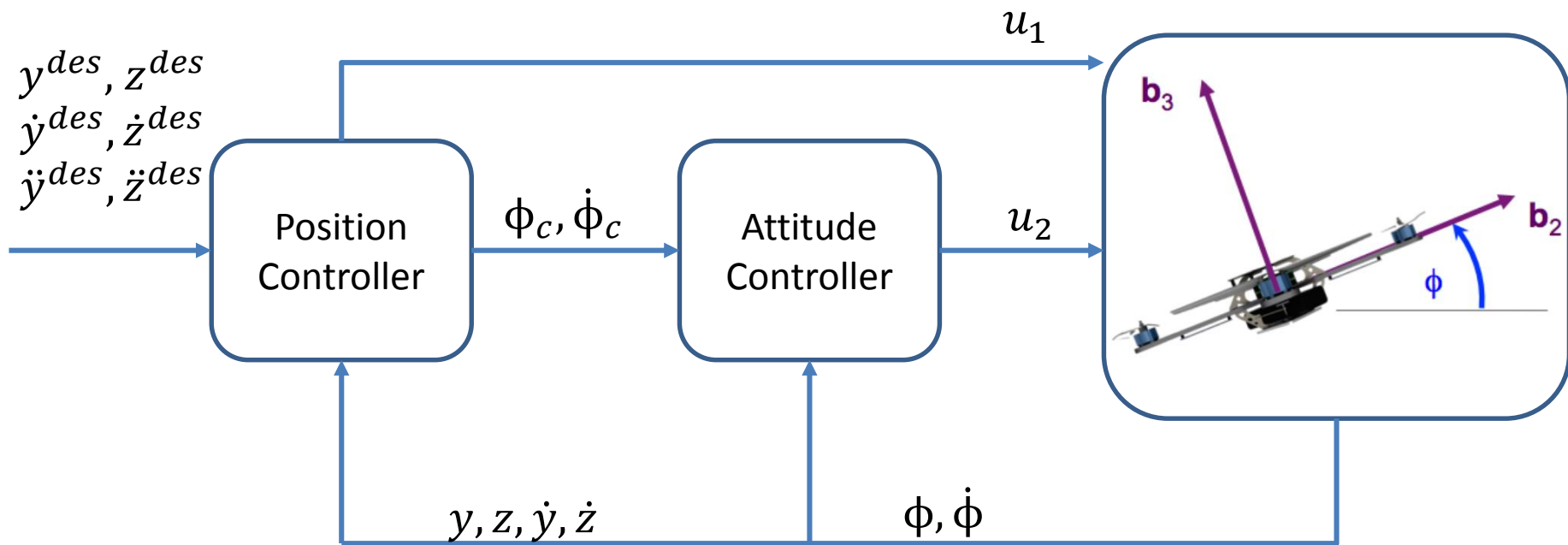
$$\ddot{z} = -g + \frac{u_1}{m}$$

$$\ddot{\phi} = \frac{u_2}{I_{xx}}$$

A simple second order system



Control Structure



Control Equations

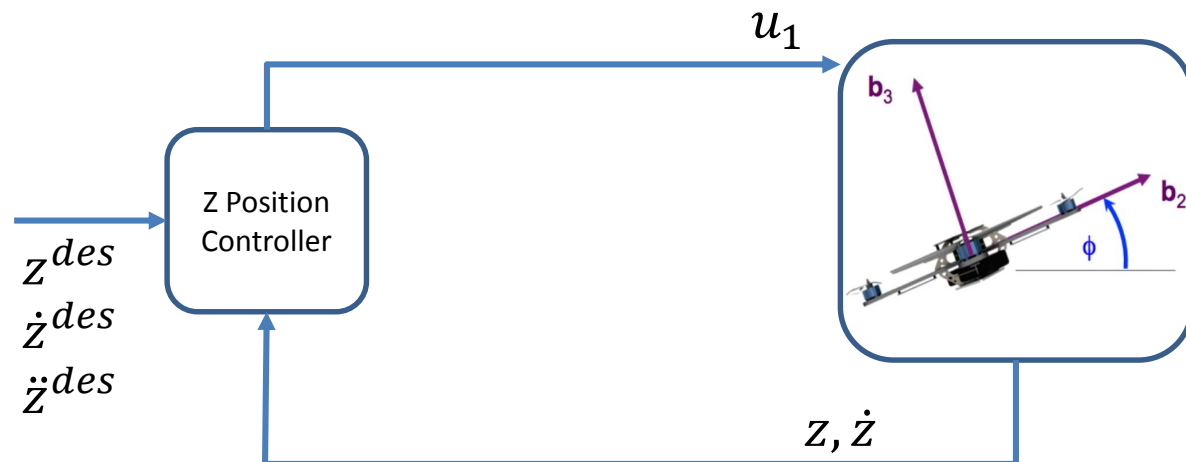
- Z-position control

PD: $\ddot{z}_c = \ddot{z}^{des} + K_{d,z}(\dot{z}^{des} - \dot{z}) + K_{p,z}(z^{des} - z)$

Model: $\ddot{z} = -g + \frac{u_1}{m}$



$$u_1 = m(g + \ddot{z}^{des} + K_{d,z}(\dot{z}^{des} - \dot{z}) + K_{p,z}(z^{des} - z))$$



Linearized Dynamic Model

- Y-position control

PD: $\ddot{y}_c = \ddot{y}^{des} + K_{d,y}(\dot{y}^{des} - \dot{y}) + K_{p,y}(y^{des} - y)$

Model: $\ddot{y} = -g\phi$

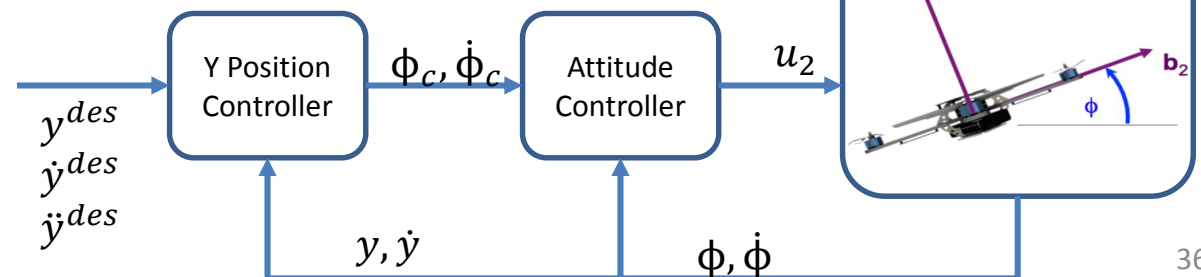
$$\phi_c = -\frac{1}{g}(\ddot{y}^{des} + K_{d,y}(\dot{y}^{des} - \dot{y}) + K_{p,y}(y^{des} - y))$$

- Attitude control

PD: $\ddot{\phi}_c = K_{d,\phi}(\dot{\phi}_c - \dot{\phi}) + K_{p,\phi}(\phi_c - \phi)$

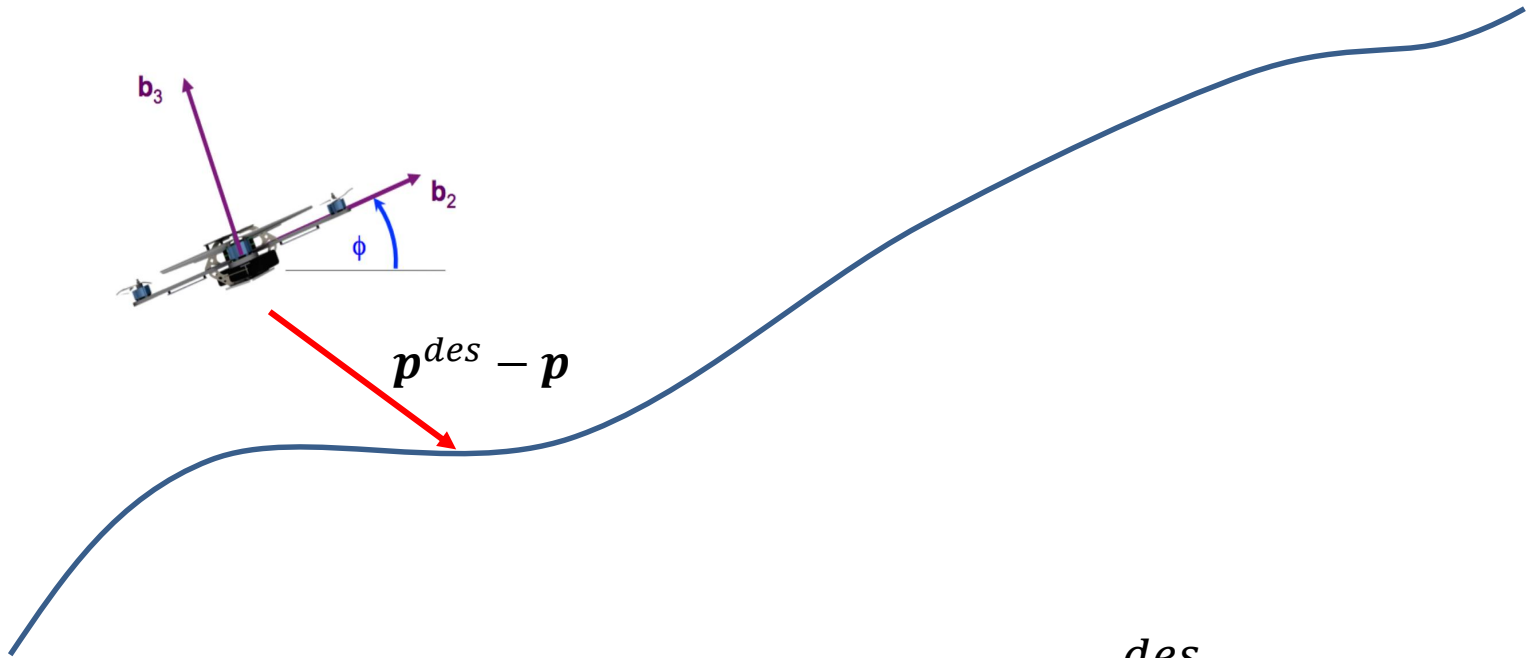
Model: $\ddot{\phi} = \frac{u_2}{I_{xx}}$

$$u_2 = I_{xx}(K_{d,\phi}(\dot{\phi}_c - \dot{\phi}) + K_{p,\phi}(\phi_c - \phi))$$



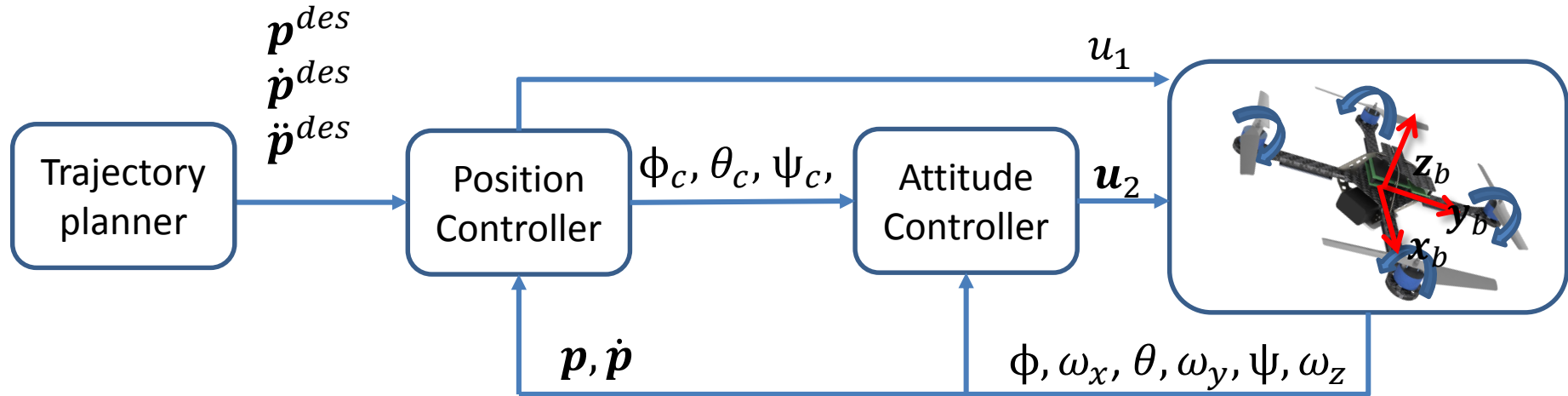
Trajectory Tracking

Given \mathbf{p}^{des} , $\dot{\mathbf{p}}^{des}$, $\ddot{\mathbf{p}}^{des}$



$$\begin{aligned}
 \mathbf{e}_p &= \mathbf{p}^{des} - \mathbf{p} \\
 \mathbf{e}_v &= \dot{\mathbf{p}}^{des} - \dot{\mathbf{p}} \\
 \ddot{\mathbf{p}}_c &= \ddot{\mathbf{p}}^{des} + \mathbf{K}_d \mathbf{e}_v + \mathbf{K}_p \mathbf{e}_p
 \end{aligned}$$

3-D Quadrotor



- Nonlinear dynamics

Newton Equation:
$$m\ddot{\mathbf{p}} = \begin{bmatrix} 0 \\ 0 \\ -mg \end{bmatrix} + \mathbf{R} \begin{bmatrix} 0 \\ 0 \\ \boxed{F_1 + F_2 + F_3 + F_4} \end{bmatrix} \mathbf{u}_1$$

Euler Equation:
$$\mathbf{I} \cdot \begin{bmatrix} \dot{\omega}_x \\ \dot{\omega}_y \\ \dot{\omega}_z \end{bmatrix} + \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} \times \mathbf{I} \cdot \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} = \begin{bmatrix} l(F_2 - F_4) \\ l(F_3 - F_1) \\ \boxed{M_1 - M_2 + M_3 - M_4} \end{bmatrix} \mathbf{u}_2$$

3-D Quadrotor

- Linearization

Equilibrium hover ($\phi_0 \sim 0, \theta_0 \sim 0, u_{1,0} \sim mg$)

Newton equation

$$m\ddot{\mathbf{p}} = \begin{bmatrix} 0 \\ 0 \\ -mg \end{bmatrix} + \mathbf{R} \begin{bmatrix} 0 \\ 0 \\ F_1 + F_2 + F_3 + F_4 \end{bmatrix} \quad \Rightarrow \quad \begin{aligned} \ddot{\mathbf{p}}_1 &= \ddot{x} = g(\theta \cos \psi + \phi \sin \psi) \\ \ddot{\mathbf{p}}_2 &= \ddot{y} = g(\theta \sin \psi - \phi \cos \psi) \\ \ddot{\mathbf{p}}_3 &= \ddot{z} = -g + \frac{u_1}{m} \end{aligned}$$

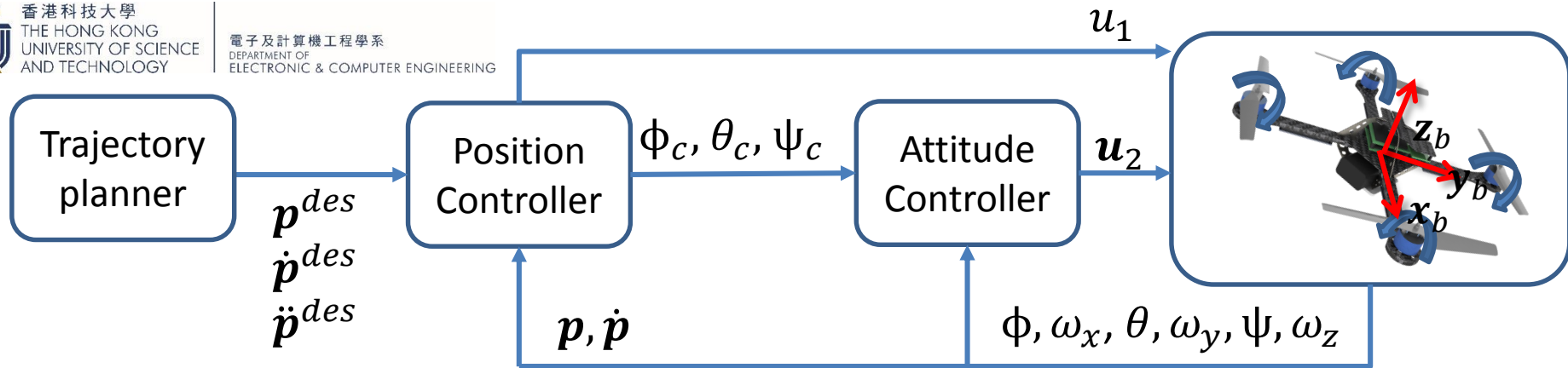
$$\mathbf{R} = \begin{bmatrix} c\psi c\theta - s\phi s\psi s\theta & -c\phi s\psi & c\psi s\theta + c\theta s\phi s\psi \\ c\theta s\psi + c\psi s\phi s\theta & c\phi c\psi & s\psi s\theta - c\psi c\theta s\phi \\ -c\phi s\theta & s\phi & c\phi c\theta \end{bmatrix}$$

Euler angle derivative

$$\begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} = \begin{bmatrix} c\theta & 0 & -c\phi s\theta \\ 0 & 1 & s\phi \\ s\theta & 0 & c\phi c\theta \end{bmatrix} \begin{bmatrix} \dot{\phi} \\ \dot{\theta} \\ \dot{\psi} \end{bmatrix} \quad \Rightarrow \quad \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} = \begin{bmatrix} \dot{\phi} \\ \dot{\theta} \\ \dot{\psi} \end{bmatrix}$$

Euler Equation:

$$\mathbf{I} \cdot \begin{bmatrix} \ddot{\phi} \\ \ddot{\theta} \\ \ddot{\psi} \end{bmatrix} + \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} \times \mathbf{I} \cdot \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} = \begin{bmatrix} l(F_2 - F_4) \\ l(F_3 - F_1) \\ M_1 - M_2 + M_3 - M_4 \end{bmatrix}$$



- Position control

PID: $\ddot{\mathbf{p}}_{i,c} = \ddot{\mathbf{p}}_i^{des} + K_{d,i}(\dot{\mathbf{p}}_i^{des} - \dot{\mathbf{p}}_i) + K_{p,i}(\mathbf{p}_i^{des} - \mathbf{p}_i)$

Model: $u_1 = m(g + \ddot{\mathbf{p}}_{3,c})$ (Newton Equation)

$$\phi_c = \frac{1}{g} (\ddot{\mathbf{p}}_{1,c} \sin \psi - \ddot{\mathbf{p}}_{2,c} \cos \psi) \quad \theta_c = \frac{1}{g} (\ddot{\mathbf{p}}_{1,c} \cos \psi + \ddot{\mathbf{p}}_{2,c} \sin \psi)$$

These are current yaw, not the commanded yaw

- Attitude control

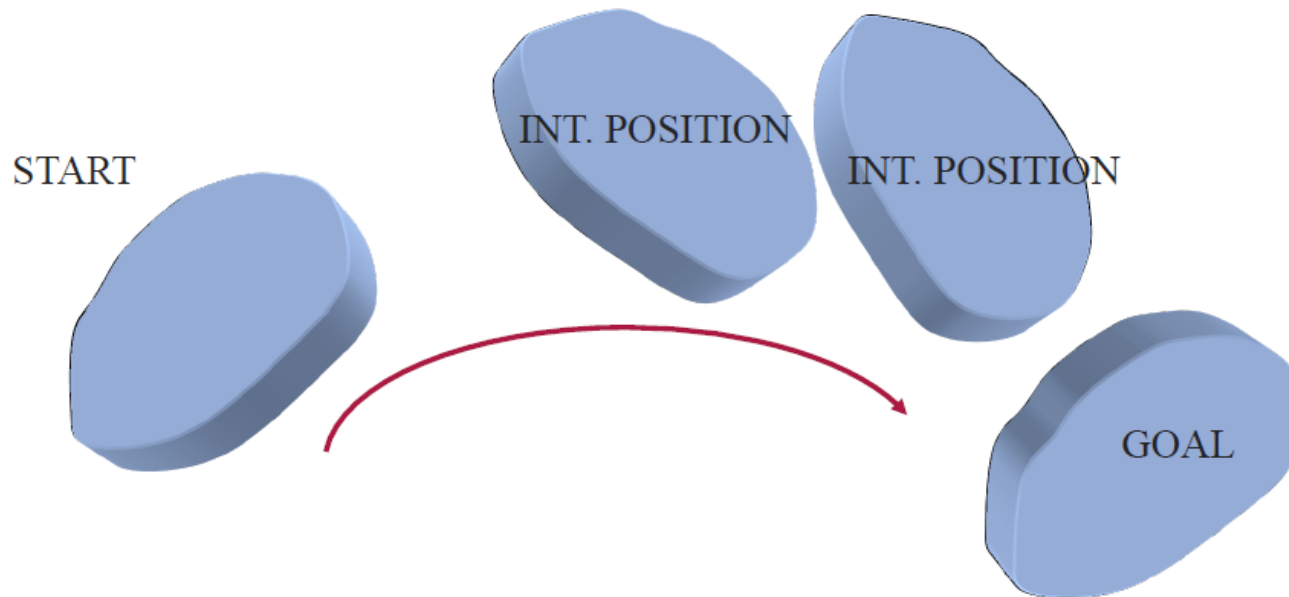
PID:
$$\begin{bmatrix} \ddot{\phi}_c \\ \ddot{\theta}_c \\ \ddot{\psi}_c \end{bmatrix} = \begin{bmatrix} K_{p,\phi}(\phi_c - \phi) + K_{d,\phi}(\dot{\phi}_c - \dot{\phi}) \\ K_{p,\theta}(\theta_c - \theta) + K_{d,\theta}(\dot{\theta}_c - \dot{\theta}) \\ K_{p,\psi}(\psi_c - \psi) + K_{d,\psi}(\dot{\psi}_c - \dot{\psi}) \end{bmatrix}$$

Model:
$$\mathbf{u}_2 = \mathbf{I} \cdot \begin{bmatrix} \ddot{\phi}_c \\ \ddot{\theta}_c \\ \ddot{\psi}_c \end{bmatrix} + \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} \times \mathbf{I} \cdot \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} \text{ (Euler Equation)}$$

Trajectory Generation

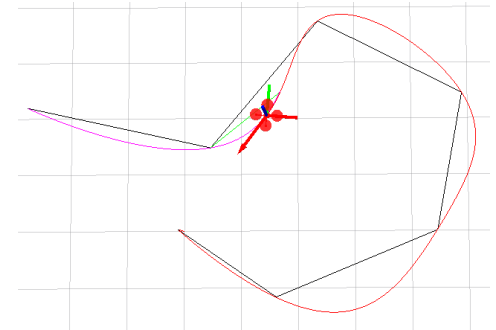
Smooth 3D Trajectories

- Smooth trajectory is beneficial for autonomous flight
 - Smooth trajectories respect the continuous nature of aerial robots
 - The robot should not stop at turns



Smooth 3D Trajectories

- General setup
 - Start, goal positions (orientations)
 - Waypoint positions (orientations)
 - Waypoints can be found by path planning (A^* , RRT^* , etc)
 - To be covered in the next lecture
 - Smoothness criterion
 - Generally translates into minimizing rate of change of “input”
- Question: How to make sure that a trajectory can be tracked by the quadrotor?



Differential Flatness

- The states and the inputs of a quadrotor can be written as algebraic functions of four carefully selected flat outputs and their derivatives
 - Enables automated generation of trajectories
 - Any smooth trajectory in the space of flat outputs (with reasonably bounded derivatives) can be followed by the under-actuated quadrotor
 - A possible choice:
 - $\sigma = [x, y, z, \psi]^T$
 - Trajectory in the space of flat outputs:
 - $\sigma(t) = [T_0, T_M] \rightarrow \mathbb{R}^3 \times SO(2)$

Differential Flatness

- Quadrotor states

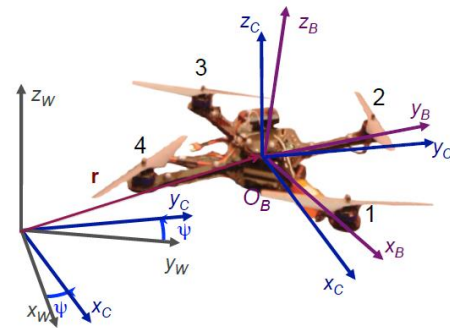
- Position, orientation, linear velocity, angular velocity

$$\mathbf{X} = [x, y, z, \phi, \theta, \psi, \dot{x}, \dot{y}, \dot{z}, \omega_x, \omega_y, \omega_z]^T$$

- Equation of motions:

$$m\ddot{\mathbf{p}} = -mg\mathbf{z}_W + u_1\mathbf{z}_B.$$

Body angular velocity
viewed in the body frame



$$\omega_B = \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix}, \quad \dot{\omega}_B = I^{-1} \left[-\omega_B \times I \cdot \omega_B + \begin{bmatrix} u_2 \\ u_3 \\ u_4 \end{bmatrix} \right]$$

- Position, velocity, and acceleration are simply derivatives of the flat outputs

Differential Flatness

- Orientation

- Quadrotor state:

$$\mathbf{X} = [x, y, z, \phi, \theta, \psi, \dot{x}, \dot{y}, \dot{z}, \omega_x, \omega_y, \omega_z]^T$$

- From the equation of motion:

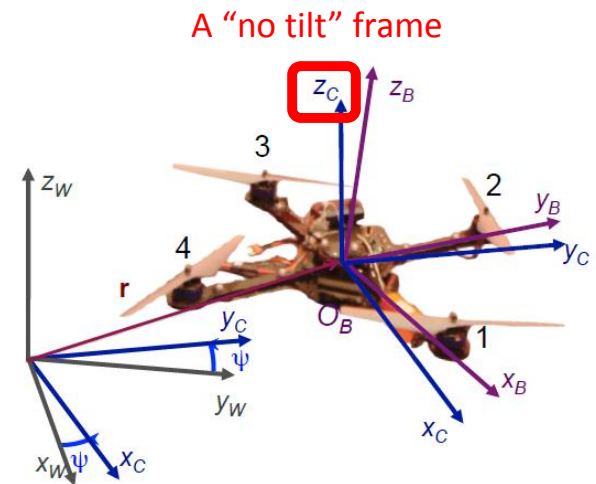
$$\mathbf{z}_B = \frac{\mathbf{t}}{\|\mathbf{t}\|}, \mathbf{t} = [\ddot{\sigma}_1, \ddot{\sigma}_2, \ddot{\sigma}_3 + g]^T$$

- Define the yaw vector (Z-X-Y Euler):

$$\mathbf{x}_C = [\cos \sigma_4, \sin \sigma_4, 0]^T$$

- Orientation can be expressed in terms of flat outputs

$$\mathbf{y}_B = \frac{\mathbf{z}_B \times \mathbf{x}_C}{\|\mathbf{z}_B \times \mathbf{x}_C\|}, \quad \mathbf{x}_B = \mathbf{y}_B \times \mathbf{z}_B \quad \mathbf{R}_B = [\mathbf{x}_B \quad \mathbf{y}_B \quad \mathbf{z}_B]$$



Differential Flatness

- Angular velocity

- Quadrotor state:

$$\mathbf{X} = [x, y, z, \phi, \theta, \psi, \dot{x}, \dot{y}, \dot{z}, \omega_x, \omega_y, \omega_z]^T$$

- Take the derivative of the equation of motion

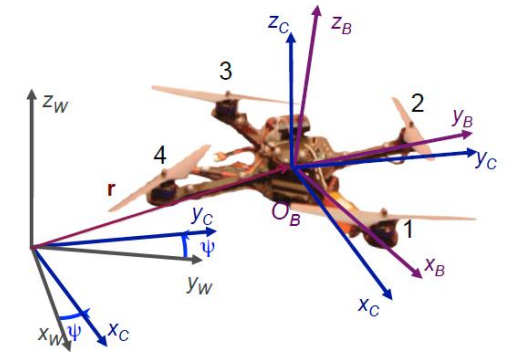
$$m\ddot{\mathbf{p}} = -mg\mathbf{z}_W + u_1\mathbf{z}_B. \quad \longrightarrow \quad m\dot{\mathbf{a}} = \dot{u}_1\mathbf{z}_B + \boldsymbol{\omega}_{BW} \times u_1\mathbf{z}_B$$

- Quadrotors only have vertical thrust:

$$\dot{u}_1 = \mathbf{z}_B \cdot m\dot{\mathbf{a}}$$

- We have:

$$\mathbf{h}_\omega = \boldsymbol{\omega}_{BW} \times \mathbf{z}_B = \frac{m}{u_1} (\dot{\mathbf{a}} - (\mathbf{z}_B \cdot \dot{\mathbf{a}})\mathbf{z}_B).$$



Body angular velocity
viewed in the world frame

Differential Flatness

- Angular velocity

- Quadrotor state:

$$\mathbf{X} = [x, y, z, \phi, \theta, \psi, \dot{x}, \dot{y}, \dot{z}, \omega_x, \omega_y, \omega_z]^T$$

- We have:

$$\mathbf{h}_\omega = \boldsymbol{\omega}_{BW} \times \mathbf{z}_B = \frac{m}{u_1} (\dot{\mathbf{a}} - (\mathbf{z}_B \cdot \dot{\mathbf{a}}) \mathbf{z}_B).$$

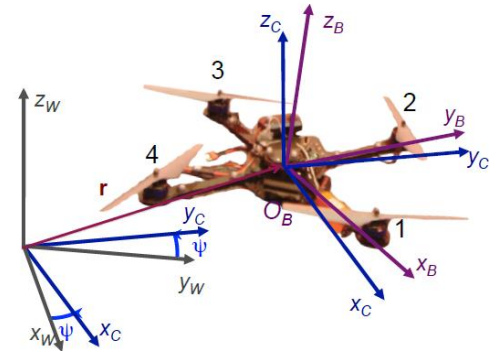
- This is the projection of $\frac{m}{u_1} \dot{\mathbf{a}}$ onto the $x_B - y_B$ plane

- We know that:

$$\boldsymbol{\omega}_{BW} = \omega_x \mathbf{x}_B + \omega_y \mathbf{y}_B + \omega_z \mathbf{z}_B$$

- Angular velocities along x_B and y_B directions can be found as:

$$\omega_x = -\mathbf{h}_\omega \cdot \mathbf{y}_B, \quad \omega_y = \mathbf{h}_\omega \cdot \mathbf{x}_B$$



Differential Flatness

- Angular velocity

- Quadrotor state:

$$\mathbf{X} = [x, y, z, \phi, \theta, \psi, \dot{x}, \dot{y}, \dot{z}, \omega_x, \omega_y, \omega_z]^T$$

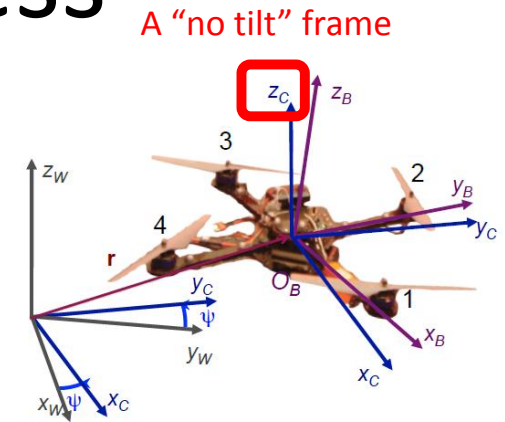
- We have:

$$\mathbf{h}_\omega = \boldsymbol{\omega}_{BW} \times \mathbf{z}_B = \frac{m}{u_1} (\dot{\mathbf{a}} - (\mathbf{z}_B \cdot \dot{\mathbf{a}}) \mathbf{z}_B).$$

- This is the projection of $\frac{m}{u_1} \dot{\mathbf{a}}$ onto the $x_B - y_B$ plane

- Since $\boldsymbol{\omega}_{BW} = \boldsymbol{\omega}_{BC} + \boldsymbol{\omega}_{CW}$, where $\boldsymbol{\omega}_{BC}$ has no \mathbf{z}_B component:

$$\omega_z = \boldsymbol{\omega}_{BW} \cdot \mathbf{z}_B = \boldsymbol{\omega}_{CW} \cdot \mathbf{z}_B = \dot{\psi} \mathbf{z}_W \cdot \mathbf{z}_B.$$



Differential Flatness

- Summary

- Quadrotor state:

$$\mathbf{X} = [x, y, z, \phi, \theta, \psi, \dot{x}, \dot{y}, \dot{z}, \omega_x, \omega_y, \omega_z]^T$$

- Flat outputs:

- $\sigma = [x, y, z, \psi]^T$

- Position, velocity, acceleration

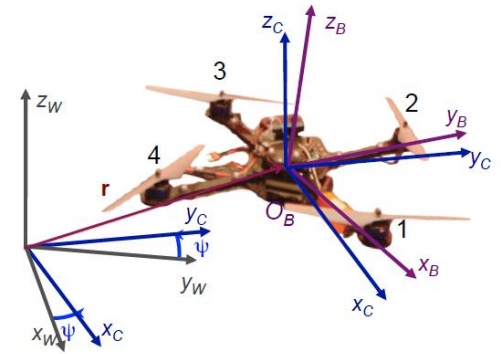
- Derivatives of flat outputs

- Orientation

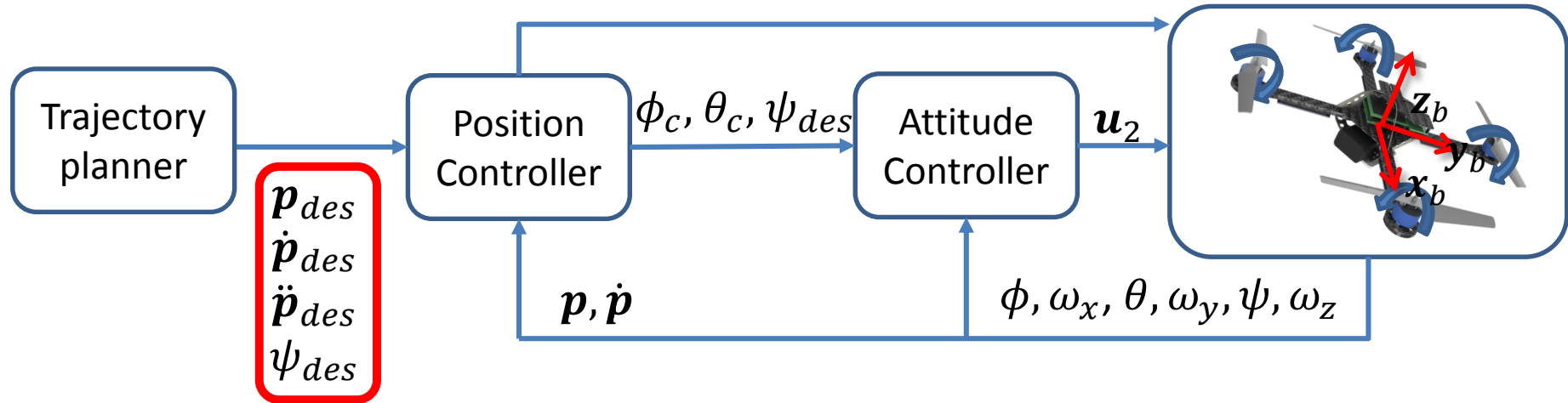
$$\mathbf{x}_C = [\cos\sigma_4, \sin\sigma_4, 0]^T \longrightarrow \mathbf{R}_B = [\mathbf{x}_B \ \mathbf{y}_B \ \mathbf{z}_B]$$

- Angular velocity

$$\omega_x = -\mathbf{h}_\omega \cdot \mathbf{y}_B, \quad \omega_y = \mathbf{h}_\omega \cdot \mathbf{x}_B, \quad \omega_z = \dot{\psi} \mathbf{z}_W \cdot \mathbf{z}_B$$



How about Force and Moment Input (u_1, u_2)?



Nonlinear dynamics

Newton Equation:
$$m\ddot{\mathbf{p}} = \begin{bmatrix} 0 \\ 0 \\ -mg \end{bmatrix} + \mathbf{R} \begin{bmatrix} 0 \\ 0 \\ F_1 + F_2 + F_3 + F_4 \end{bmatrix}$$

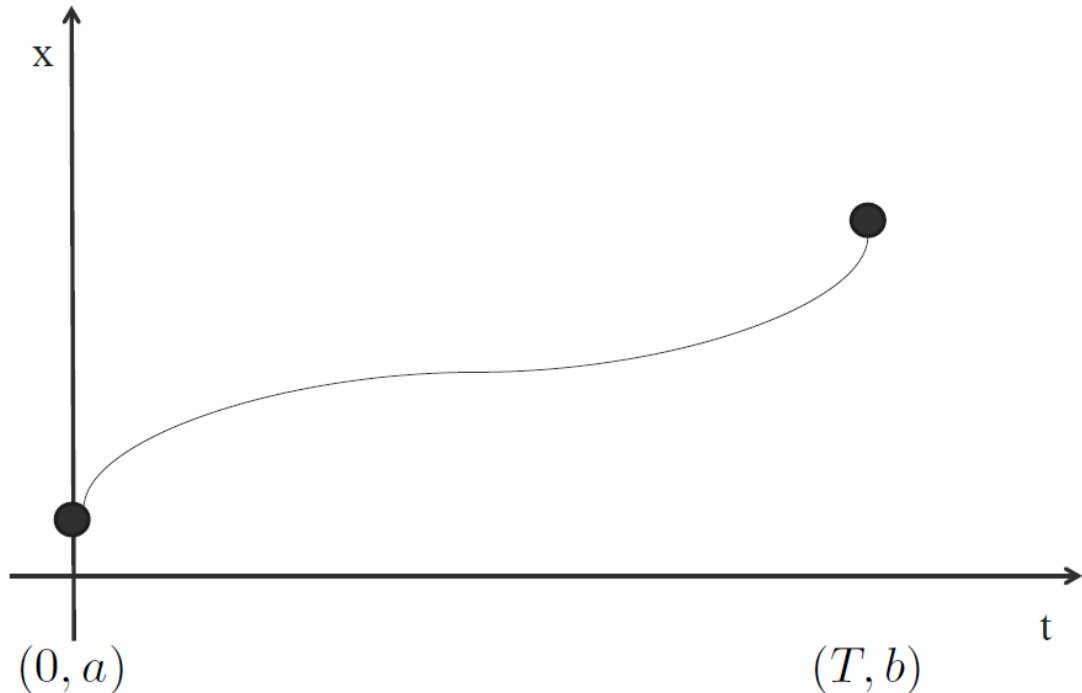
Euler Equation:
$$\mathbf{I} \cdot \begin{bmatrix} \dot{\omega}_x \\ \dot{\omega}_y \\ \dot{\omega}_z \end{bmatrix} + \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} \times \mathbf{I} \cdot \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} = \begin{bmatrix} l(F_2 - F_4) \\ l(F_3 - F_1) \\ M_1 - M_2 + M_3 - M_4 \end{bmatrix}$$

Polynomial Trajectories

- Flat outputs:
 - $\sigma = [x, y, z, \psi]^T$
- Trajectory in the space of flat outputs:
 - $\sigma(t) = [T_0, T_M] \rightarrow \mathbb{R}^3 \times SO(2)$
- Polynomial functions can be used to specify trajectories in the space of flat outputs
 - Easy determination of smoothness criterion with polynomial orders
 - Easy and closed form calculation of derivatives
 - Decoupled trajectory generation in three dimensions

Smooth 1D Trajectory

- Design a trajectory $x(t)$ such that:
 - $x(0) = a$
 - $x(T) = b$



Smooth 1D Trajectory

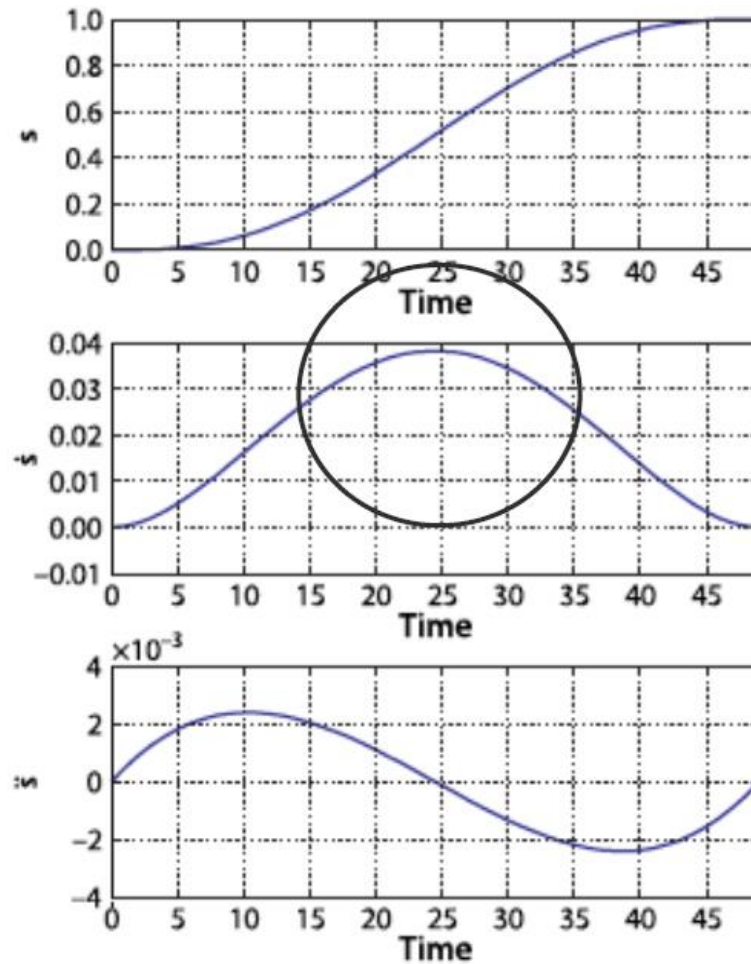
- 5th order polynomial trajectory:
 - $x(t) = c_5t^5 + c_4t^4 + c_3t^3 + c_2t^2 + c_1t + c_0$
- Boundary conditions

	Position	Velocity	Acceleration
t = 0	a	0	0
t = T	b	0	0

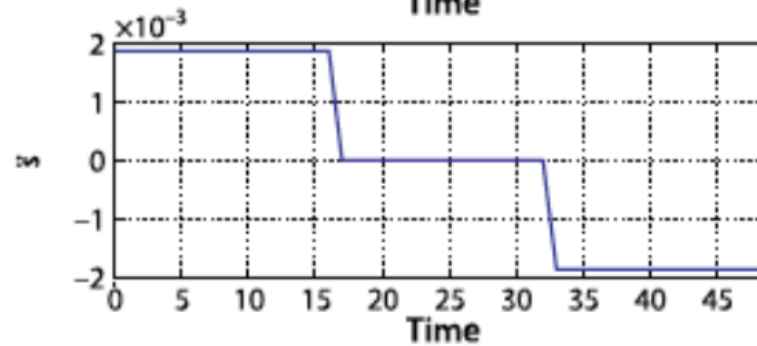
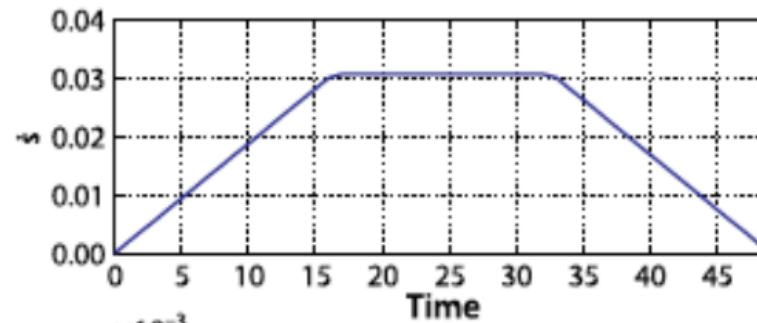
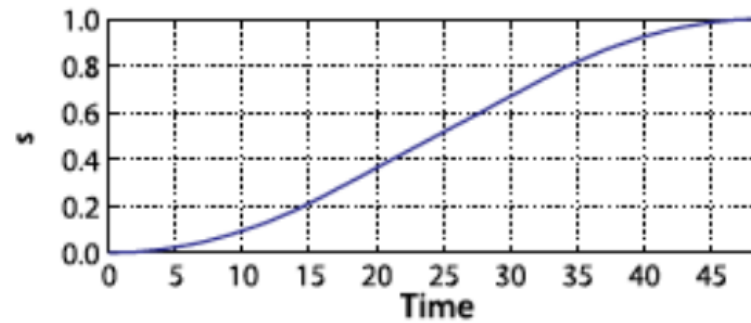
- Solve:

$$\begin{bmatrix} a \\ b \\ 0 \\ 0 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 1 \\ T^5 & T^4 & T^3 & T^2 & T & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 5T^4 & 4T^3 & 3T^2 & 2T & 1 & 0 \\ 0 & 0 & 0 & 2 & 0 & 0 \\ 20T^3 & 12T^2 & 6T & 2 & 0 & 0 \end{bmatrix} \begin{bmatrix} c_5 \\ c_4 \\ c_3 \\ c_2 \\ c_1 \\ c_0 \end{bmatrix}$$

Smooth 1D Trajectory



Bang-Bang Trajectory



Smooth Multi-Segment Trajectory

- Smooth the corners of straight line segments
- Preferred constant velocity motion at v
- Preferred zero acceleration
- Requires special handling of short segments



Smooth 1D Trajectory

- Generate each 5th order polynomial independently:
 - $x(t) = c_5t^5 + c_4t^4 + c_3t^3 + c_2t^2 + c_1t + c_0$
- Boundary conditions

	Position	Velocity	Acceleration
t = 0	a	v_0	0
t = T	b	v_T	0

- Solve:

$$\begin{bmatrix} a \\ b \\ v_0 \\ v_T \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 & 0 & 0 & 0 & 0 & 1 \\ T^5 & T^4 & T^3 & T^2 & T & 1 \\ 0 & 0 & 0 & 0 & 1 & 0 \\ 5T^4 & 4T^3 & 3T^2 & 2T & 1 & 0 \\ 0 & 0 & 0 & 2 & 0 & 0 \\ 20T^3 & 12T^2 & 6T & 2 & 0 & 0 \end{bmatrix} \begin{bmatrix} c_5 \\ c_4 \\ c_3 \\ c_2 \\ c_1 \\ c_0 \end{bmatrix}$$

Next Lecture...

- Continuation on trajectory generation
- Path planning

Logistics

- Project 1, phase 1 is released (03/10)
 - Due in one week: 3/17