# Introduction to Aerial Robotics Lecture 3

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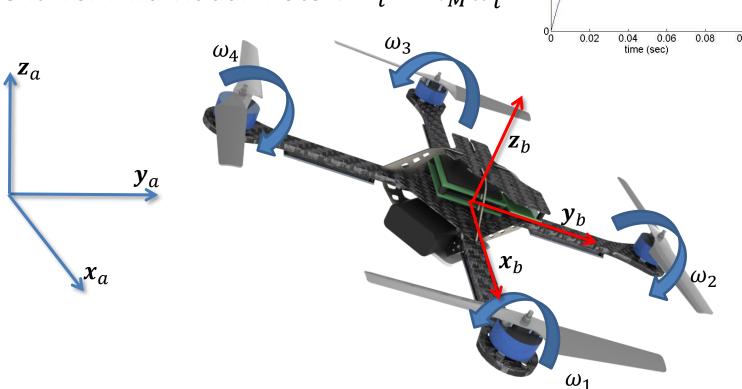
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## Outline

- Quadrotor Dynamics (Review)
- Control Basics
- Quadrotor Control
- Trajectory Generation

# Quadrotor Dynamics (Review)

- 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:  $M_i = k_M \omega_i^2$

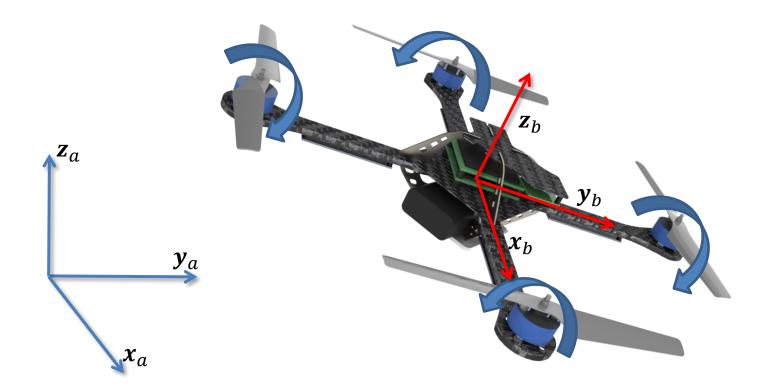


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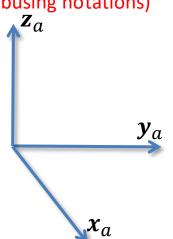
- Z-X-Y Euler Angles:  $\mathbf{\textit{R}}_{ab} = \mathbf{\textit{R}}_{z}(\psi) \cdot \mathbf{\textit{R}}_{x}(\phi) \cdot \mathbf{\textit{R}}_{y}(\theta)$
- Sequence of three rotations about body-fixed axes
- What are the singularities?

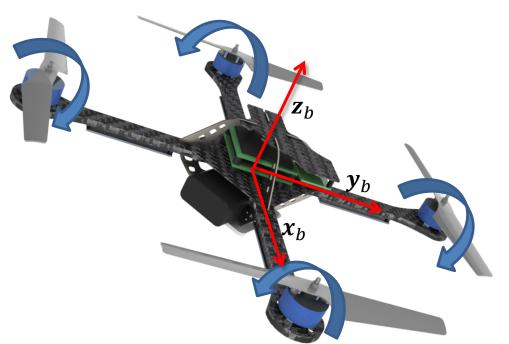


• 
$$\mathbf{R}_{ab} = \begin{bmatrix} c\psi c\theta - s\phi s\psi s\theta & -c\phi s\psi & c\psi s\theta + c\theta s\phi s\psi \\ c\theta s\psi + c\psi s\phi s\theta & c\phi c\psi & s\psi s\theta - c\psi c\theta s\phi \\ -c\phi s\theta & s\phi & c\phi c\theta \end{bmatrix}$$

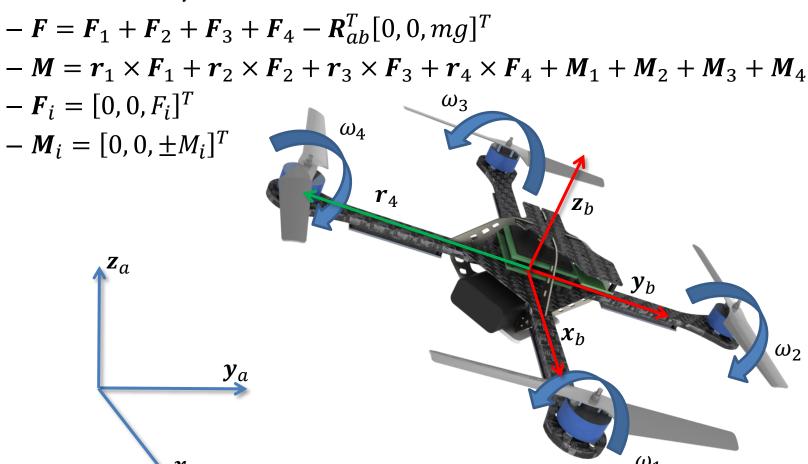
$$\bullet \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} = \begin{bmatrix} c\theta & 0 & -c\phi s\theta \\ 0 & 1 & s\phi \\ s\theta & 0 & c\phi c\theta \end{bmatrix} \begin{bmatrix} \dot{\phi} \\ \dot{\theta} \\ \dot{\psi} \end{bmatrix}$$

Instantaneous body angular velocity viewed in the body frame (sorry for abusing notations)

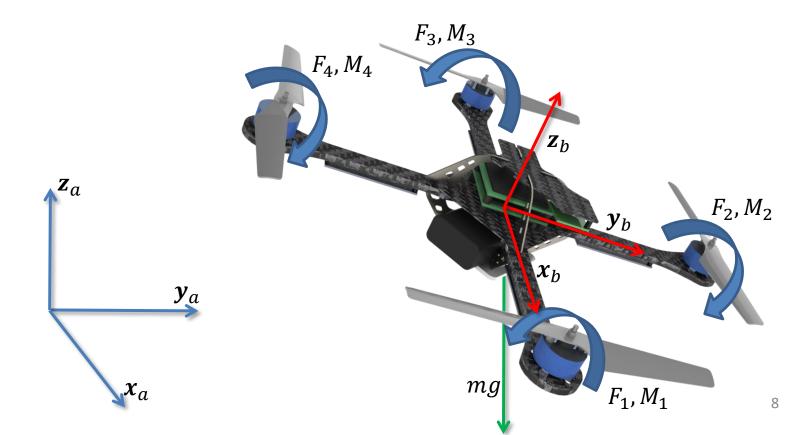




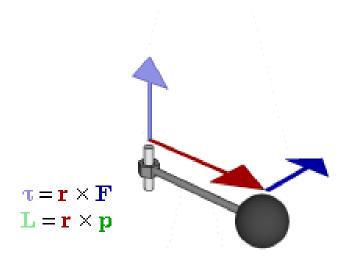
Consider body frame



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





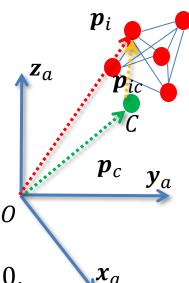


Relationship between force (F), torque/moment of force( $\tau$ ), momentum (p), and angular momentum (L) vectors in a rotating system. r is the position vector.

- The rigid body as a collection of particles
  - Center of mass (CoM):  $p_c$
  - Position of the i-th particle to CoM:  $m{p}_{ic} = m{p}_i m{p}_c$
  - Velocity of the i-th particle to CoM:  $m{v}_{ic} = m{\dot{p}}_i m{\dot{p}}_c$   $= m{v}_i m{v}_c$
  - Angular momentum of the i-th particle:

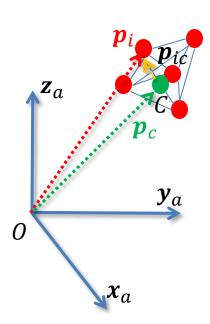
$$\boldsymbol{H}_i = \boldsymbol{p}_{ic} \times m_i \boldsymbol{v}_i$$

- Angular momentum of the rigid body:
  - $\boldsymbol{H} = \sum \boldsymbol{p}_{ic} \times m_i \boldsymbol{v}_i$
  - Since:  $\sum m_i \boldsymbol{p}_{ic} = \sum m_i (\boldsymbol{p}_i \boldsymbol{p}_c) = \sum m_i \boldsymbol{p}_i \boldsymbol{p}_c \sum m_i = 0$ ,
  - We have:  $\sum \boldsymbol{p}_{ic} \times m_i \boldsymbol{v}_c = (\sum m_i \boldsymbol{p}_{ic}) \times \boldsymbol{v}_c = 0$
  - Therefore:  $\mathbf{H} = \sum \mathbf{p}_{ic} \times m_i \mathbf{v}_i \sum \mathbf{p}_{ic} \times m_i \mathbf{v}_c = \sum \mathbf{p}_{ic} \times m_i \mathbf{v}_{ic}$
  - Since:  $v_{ic} = \boldsymbol{\omega} \times \boldsymbol{p}_{ic}$ ,
  - We have:  $\mathbf{H} = \sum \mathbf{p}_{ic} \times (\boldsymbol{\omega} \times m_i \mathbf{p}_{ic}) = -\sum \mathbf{p}_{ic} \times (m_i \mathbf{p}_{ic} \times \boldsymbol{\omega})$



#### Rotational dynamics

- Angular momentum:  $m{H} = \sum m{p}_{ic} imes m_i m{v}_i$
- Take the derivative:  $\dot{\boldsymbol{H}} = \sum \dot{\boldsymbol{p}}_{ic} \times m_i \boldsymbol{v}_i + \sum \boldsymbol{p}_{ic} \times m_i \dot{\boldsymbol{v}}_i$
- Since  $\sum \dot{\boldsymbol{p}}_{ic} \times m_i \boldsymbol{v}_i = \sum \boldsymbol{v}_i \times m_i \boldsymbol{v}_i \boldsymbol{v}_c \times m_i \boldsymbol{v}_i = \sum -\boldsymbol{v}_c \times m_i \boldsymbol{v}_i = -\boldsymbol{v}_c \times \frac{d}{dt} \sum m_i \boldsymbol{p}_i = -\boldsymbol{v}_c \times \frac{d}{dt} \boldsymbol{p}_c \sum m_i = 0$
- We have  $\dot{\boldsymbol{H}} = \sum \boldsymbol{p}_{ic} \times m_i \dot{\boldsymbol{v}}_i$
- Referring to Newton's second law:  ${m F}_i + \sum_{i \neq j} {m F}_{ij} = m_i \dot{{m v}}_i$
- $\dot{\mathbf{H}} = \sum \mathbf{p}_{ic} \times m_i \dot{\mathbf{v}}_i = \sum \mathbf{p}_{ic} \times (\mathbf{F}_i + \sum_{i \neq j} \mathbf{F}_{ij}) = \sum \mathbf{p}_{ic} \times \mathbf{F}_i$
- We also know that the external moment:  $m{M} = \sum m{p}_{ic} imes m{F}_i$
- We have the rotational dynamics: M = H



- Finishing the work on rotational dynamics
  - Given:  $\boldsymbol{H} = -\sum \boldsymbol{p}_{ic} \times (m_i \boldsymbol{p}_{ic} \times \boldsymbol{\omega})$
  - And using the fact:  $(\mathbf{R}\mathbf{a}) \times (\mathbf{R}\mathbf{b}) = \mathbf{R}(\mathbf{a} \times \mathbf{b})$ 
    - **R**: rotation matrix
    - o *a*, *b*: vectors
  - We can transform the representation of the angular momentum to the body frame with constant inertiametrix:

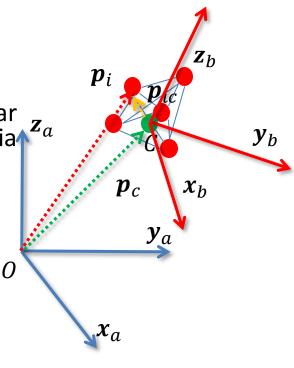
$$H = -\sum \mathbf{p}_{ic} \times (m_i \mathbf{p}_{ic} \times \boldsymbol{\omega})$$

$$= -\sum \mathbf{R}_{ab} \mathbf{p}_{ic}^b \times (m_i \mathbf{R}_{ab} \mathbf{p}_{ic}^b \times \mathbf{R}_{ab} \boldsymbol{\omega}^b)$$

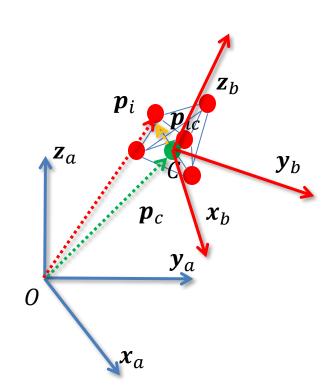
$$= -\mathbf{R}_{ab} \sum \mathbf{p}_{ic}^b \times (m_i \mathbf{p}_{ic}^b \times \boldsymbol{\omega}^b)$$

$$= -\mathbf{R}_{ab} \sum m_i \cdot \mathbf{p}_{ic}^b \times (\widehat{\mathbf{p}}_{ic}^b \cdot \boldsymbol{\omega}^b)$$

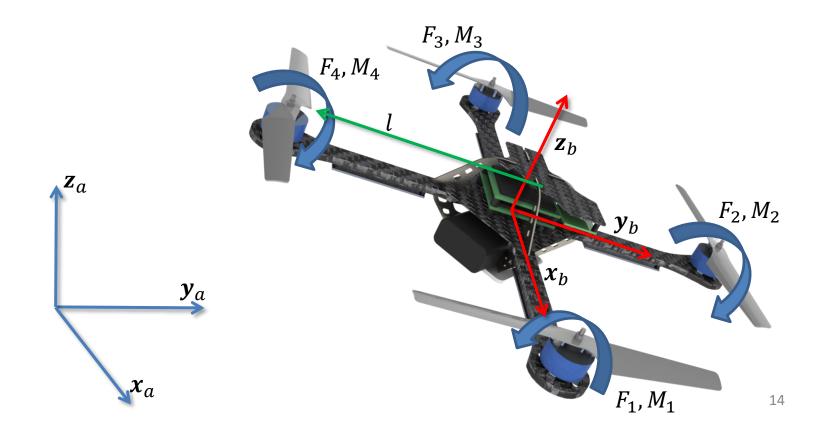
$$= \mathbf{R}_{ab} (-\sum m_i \cdot \widehat{\mathbf{p}}_{ic}^b \cdot \widehat{\mathbf{p}}_{ic}^b) \cdot \boldsymbol{\omega}^b = \mathbf{R}_{ab} (\mathbf{I}^b \boldsymbol{\omega}^b)$$



- Finishing the work on rotational dynamics
  - Given  $\boldsymbol{H} = \boldsymbol{R}_{ab}(\boldsymbol{I}^b \boldsymbol{\omega}^b)$
  - Take the derivative:  $\dot{\mathbf{H}} = \dot{\mathbf{R}}_{ab} \mathbf{I}^b \boldsymbol{\omega}^b + \mathbf{R}_{ab} \mathbf{I}^b \dot{\boldsymbol{\omega}}^b = \mathbf{R}_{ab} \widehat{\boldsymbol{\omega}}^b \mathbf{I}^b \boldsymbol{\omega}^b + \mathbf{R}_{ab} \mathbf{I}^b \dot{\boldsymbol{\omega}}^b = \mathbf{R}_{ab} (\boldsymbol{\omega}^b \times (\mathbf{I}^b \boldsymbol{\omega}^b) + \mathbf{I}^b \dot{\boldsymbol{\omega}}^b)$
  - Also transform the moment into body frame:  $\mathbf{M} = \mathbf{R}_{ab}\mathbf{M}^b$
  - Finally:  $\mathbf{M}^b = \boldsymbol{\omega}^b \times (\mathbf{I}^b \boldsymbol{\omega}^b) + \mathbf{I}^b \dot{\boldsymbol{\omega}}^b$



• Euler Equation: 
$$I \cdot \begin{bmatrix} \dot{\omega}_{x} \\ \dot{\omega}_{y} \\ \dot{\omega}_{z} \end{bmatrix} + \begin{bmatrix} \omega_{x} \\ \omega_{y} \\ \omega_{z} \end{bmatrix} \times I \cdot \begin{bmatrix} \omega_{x} \\ \omega_{y} \\ \omega_{z} \end{bmatrix} = \begin{bmatrix} l(F_{2} - F_{4}) \\ l(F_{3} - F_{1}) \\ M_{1} - M_{2} + M_{3} - M_{4} \end{bmatrix}$$



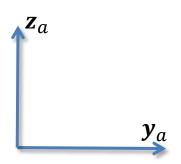
- Motor model:  $\dot{\omega}_i = k_m(\omega_i^{des} \omega_i)$
- Thrust from individual motor:  $F_i = k_F \omega_i^2$
- Moment from individual motor:  $M_i = k_M \omega_i^2$

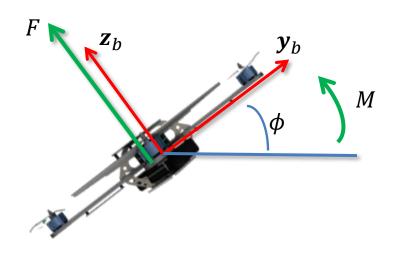
• Newton Equation: 
$$m\ddot{\pmb{p}}=\begin{bmatrix}0\\0\\-mg\end{bmatrix}+\pmb{R}\begin{bmatrix}0\\F_1+F_2+F_3+F_4\end{bmatrix}$$

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

## A Planar Quadrotor

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





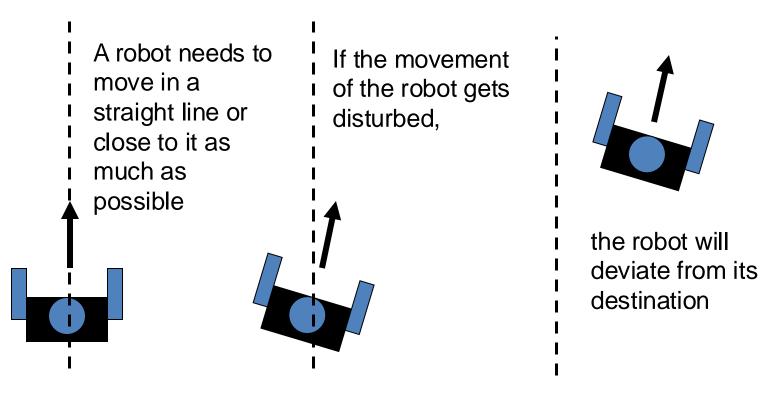
# Control System Design



## 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





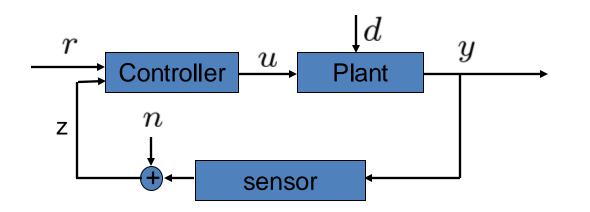
Therefore we need to have a controller to control its movement in real time based on its movement and the destination



- 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



 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

z: state feedback

u: controller output

d: plant disturbance

y: output

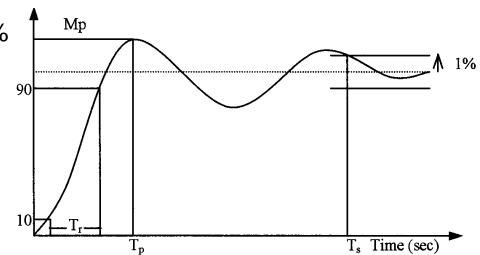
n: communication noise

 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"



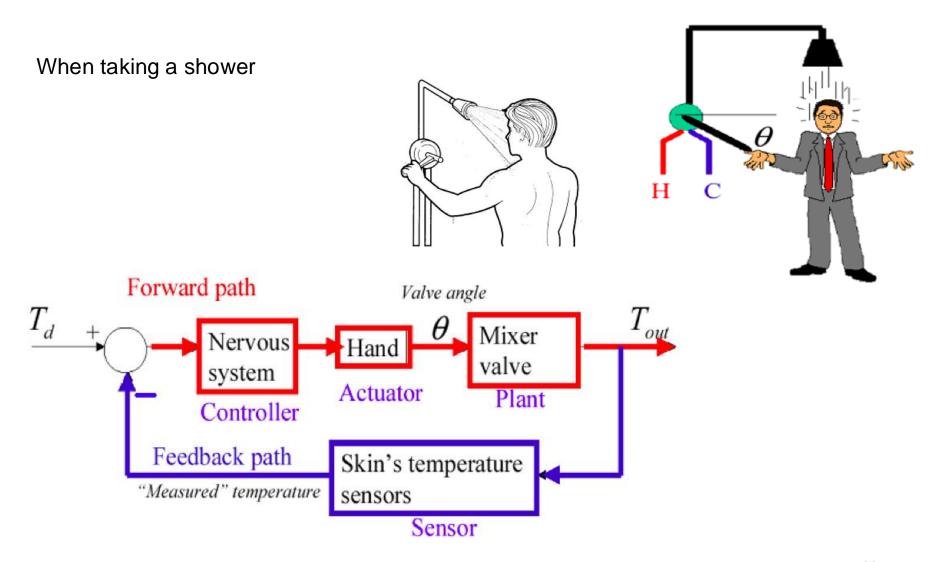
- Rise time:
  - Time it takes from 10% to 90%
- Steady-state error
- Overshoot
  - Percentage by which peak exceeds final value



- Settling time
  - Time it takes to reach 1% of final value
- A good control system has small rise time, overshoot, settling time and steadystate error

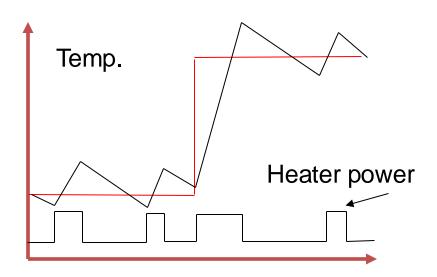
Response







- 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

$$x, u \in R$$

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

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

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

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

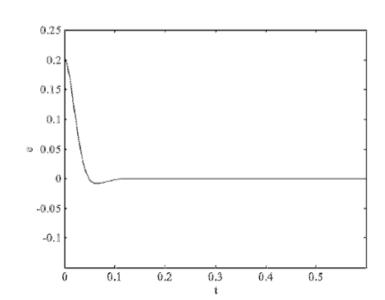
Define error, 
$$e(t) = x^{des}(t) - x(t)$$

Want e(t) to converge exponentially to zero



Find u such that

$$\ddot{e} + K_d \dot{e} + K_p e = 0 \qquad K_d, K_p > 0$$
 
$$u(t) = \ddot{x}^{des}(t) + K_d \dot{e}(t) + K_p e(t)$$
 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

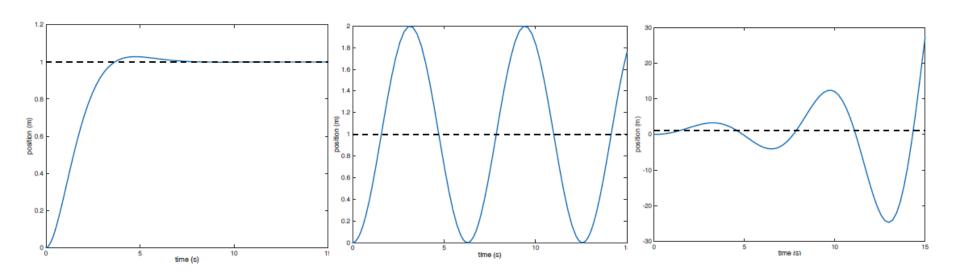
Large derivative gain makes the system overdamped and the system converges slowly

#### 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**

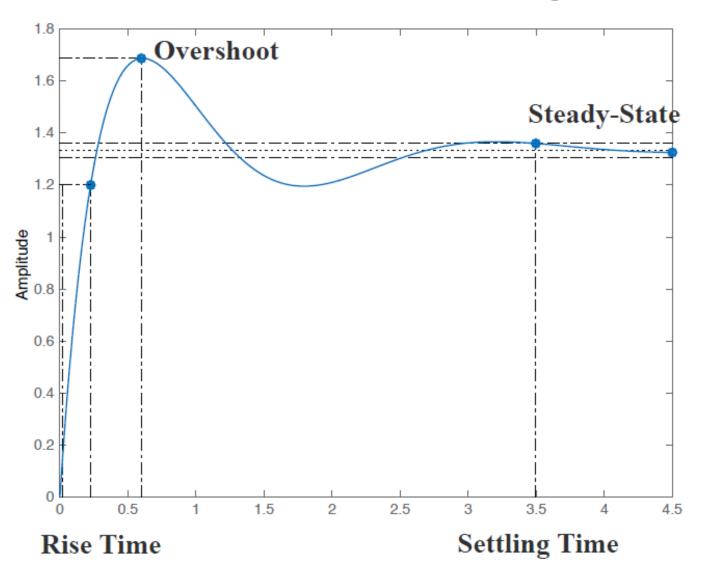


Stable (converge)

Marginally Stable (oscillate)

Unstable (diverge)

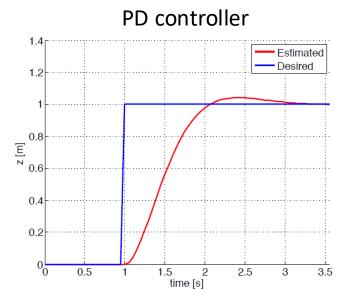
## **Manual Tuning**



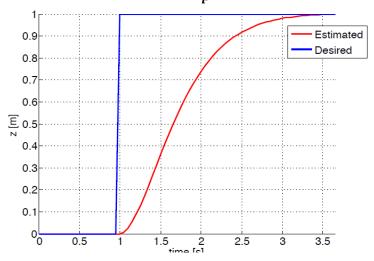
# **Manual Tuning**

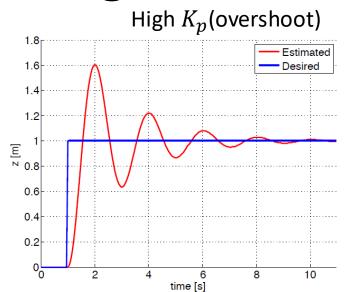
Parameter Increased	$K_p \uparrow$	$K_d \uparrow$	$K_i \uparrow$
Rise Time	Decrease	-	Decrease
Overshoot	Increase	Decrease	Increase
Settling Time	-	Decrease	Increase
Steady-State Error	Decrease	-	Eliminate

## **Manual Tuning**

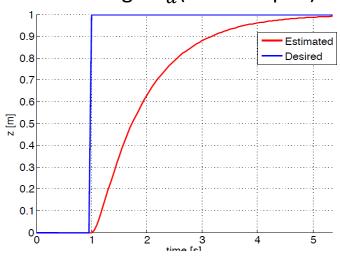


Low  $K_p$  (soft response)





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	$K_p$	$K_d$	$K_i$
Р	0.5 <i>K<sub>u</sub></i>	-	-
PD	$0.8K_u$	$K_pT_u/8$	-
PID	0.6 <i>K</i> <sub>u</sub>	$K_pT_u/8$	$2K_p/T_u$

#### 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

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

Servo: feedforward + PD feedback

- 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

Model based (estimates)

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

Advantage

Servo: feedforward + PD feedback

- 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

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

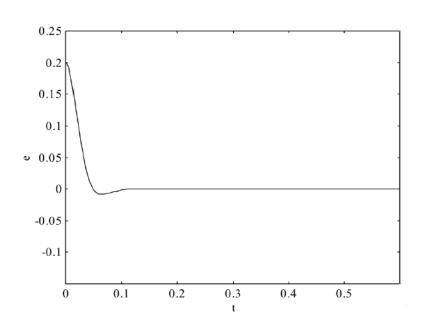
Actual performance

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

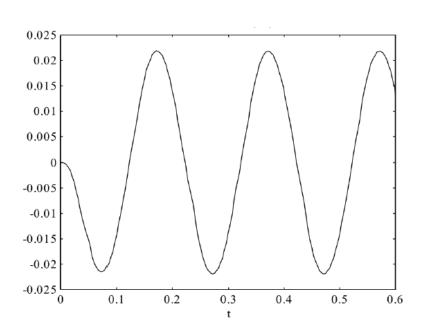
#### Model-based control

#### Performance

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



Perfect model

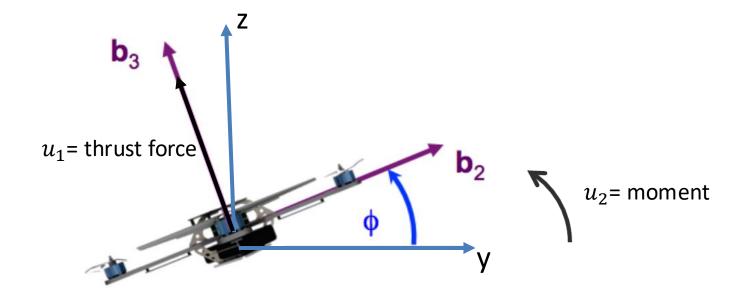


Imperfect model, 10% errors in parameters

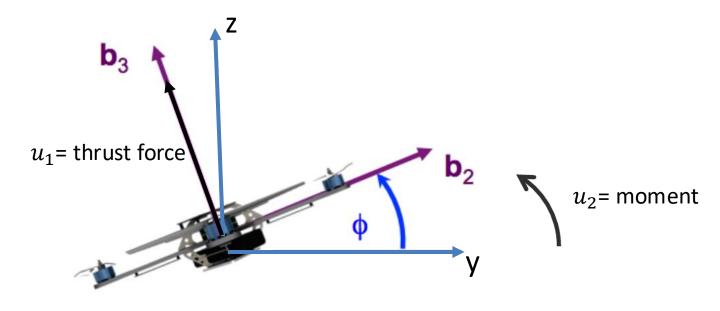
## **Quadrotor Control**



# **Application to Quadrotors**



### Planar Quadrotor Model



$$\sum_{x} F_{y} = -u_{1} \sin(\phi) = m\ddot{y}$$

$$\sum_{x} F_{z} = -mg + u_{1} \cos(\phi) = m\ddot{z}$$

$$M = u_{2} = I_{xx}\ddot{\phi}$$

$$\begin{bmatrix} \ddot{y} \\ \ddot{z} \\ \ddot{\phi} \end{bmatrix} = \begin{bmatrix} 0 \\ -g \\ 0 \end{bmatrix} + \begin{bmatrix} -\frac{1}{m} \sin\phi & 0 \\ \frac{1}{m} \cos\phi & 0 \\ 0 & \frac{1}{I_{xx}} \end{bmatrix} \begin{bmatrix} u_{1} \\ u_{2} \end{bmatrix}$$

# Linearized Dynamic Model

Nonlinear dynamics

$$\ddot{y} = -\frac{u_1}{m} sin(\phi)$$

$$\ddot{z} = -g + \frac{u_1}{m} cos(\phi)$$

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

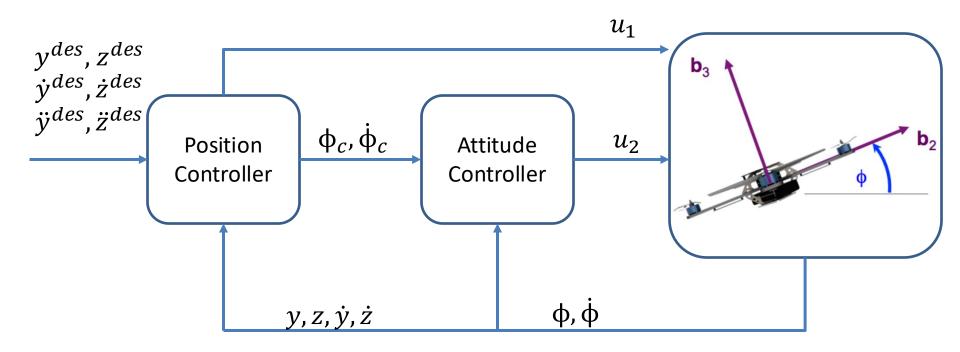
Equilibrium hover configuration

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

Linearized dynamics

$$\ddot{y} = -g \varphi$$
 Cascaded second order system  $\ddot{z} = -g + \frac{u_1}{m}$   $\ddot{\varphi} = \frac{u_2}{I_{xx}}$  A simple second order system

### **Control Structure**



# **Control Equations**

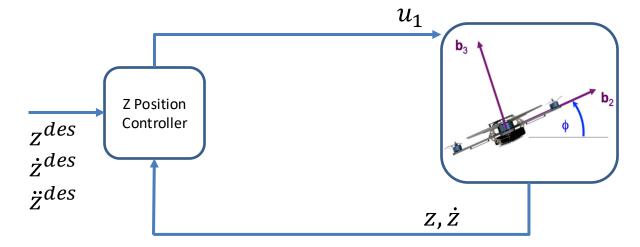
Z-position control

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

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



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



# 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:  $\ddot{y} = -g\varphi$   
 $\varphi_{c} = -\frac{1}{g}(\ddot{y}^{des} + K_{d,y}(\dot{y}^{des} - \dot{y}) + K_{p,y}(y^{des} - y))$ 

Attitude control

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

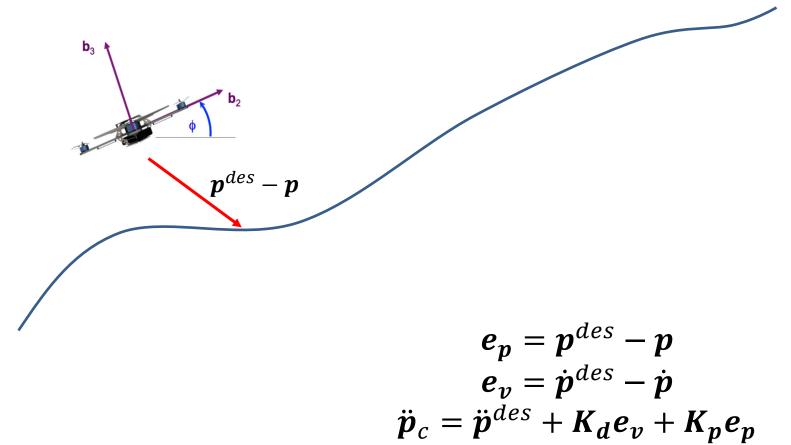
Model:  $\ddot{\phi} = \frac{u_2}{I_{xx}}$ 

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

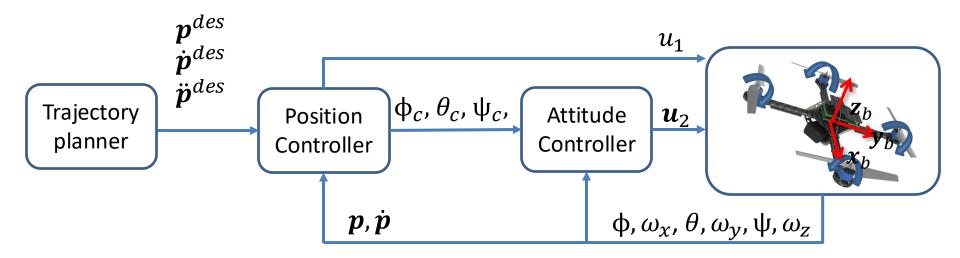
Y Position Controller  $\dot{y}_{des}$ 
 $\dot{y}_{des}$ 

# **Trajectory Tracking**

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



### 3-D Quadrotor



#### Nonlinear dynamics

Newton Equation: 
$$m\ddot{p} = \begin{bmatrix} 0\\0\\-mg \end{bmatrix} + R \begin{bmatrix} 0\\0\\F_1 + F_2 + F_3 + F_4 \end{bmatrix} \underbrace{u_1}$$

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

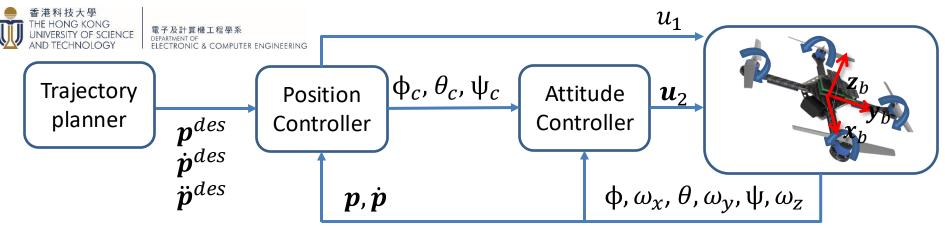
### 3-D Quadrotor

#### Linearization

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

Euler angle derivative 
$$\begin{bmatrix} \omega_{\chi} \\ \omega_{y} \\ \omega_{z} \end{bmatrix} = \begin{bmatrix} c\theta & 0 & -c\phi s\theta \\ 0 & 1 & s\phi \\ s\theta & 0 & c\phi c\theta \end{bmatrix} \begin{bmatrix} \dot{\phi} \\ \dot{\theta} \\ \dot{\psi} \end{bmatrix} \longrightarrow \begin{bmatrix} \omega_{\chi} \\ \omega_{y} \\ \omega_{z} \end{bmatrix} = \begin{bmatrix} \dot{\phi} \\ \dot{\theta} \\ \dot{\psi} \end{bmatrix}$$

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



Position control

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

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

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

Attitude control

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

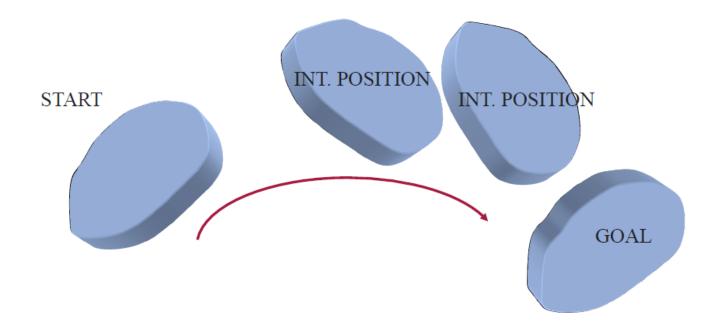
Model: 
$$\mathbf{u}_2 = \mathbf{I} \cdot \begin{bmatrix} \Phi_c \\ \ddot{\theta}_c \\ \ddot{\psi}_z \end{bmatrix} + \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix} \times \mathbf{I} \cdot \begin{bmatrix} \omega_x \\ \omega_y \\ \omega_z \end{bmatrix}$$
 (Euler Equation)

# **Trajectory Generation**



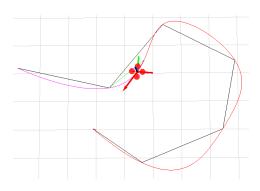
## **Smooth 3D Trajectories**

- 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?



- 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:
    - $\boldsymbol{\sigma} = [x, y, z, \psi]^T$
  - Trajectory in the space of flat outputs:
    - $\sigma(t) = [T_0, T_M] \rightarrow \mathbb{R}^3 \times SO(2)$

Body angular velocity

viewed in the body frame

- 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:

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

$$\boldsymbol{\omega}_{B} = \begin{bmatrix} \omega_{x} \\ \omega_{y} \\ \omega_{z} \end{bmatrix}, \quad \boldsymbol{\omega}_{B} = \boldsymbol{I}^{-1} \begin{bmatrix} -\boldsymbol{\omega}_{B} \times \boldsymbol{I} \cdot \boldsymbol{\omega}_{B} + \begin{bmatrix} u_{2} \\ u_{3} \\ u_{4} \end{bmatrix} \end{bmatrix}$$

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

#### 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:

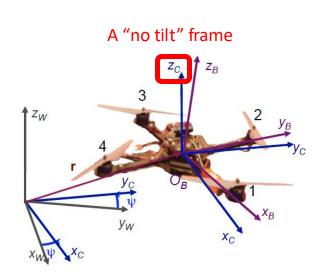
$$\mathbf{z}_B = \frac{\mathbf{t}}{\|\mathbf{t}\|}, \mathbf{t} = [\ddot{\boldsymbol{\sigma}}_1, \ddot{\boldsymbol{\sigma}}_2, \ddot{\boldsymbol{\sigma}}_3 + g]^T$$

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

$$\mathbf{x}_C = [cos\boldsymbol{\sigma}_4, sin\boldsymbol{\sigma}_4, 0]^T$$

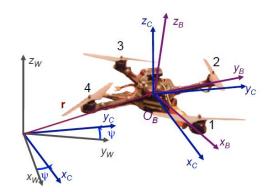
Orientation can be expressed in terms of flat outputs

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



- Angular velocity
  - Quadrotor state:

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



Take the derivative of the equation of motion

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

— Quadrotors only have vertical thrust:

Body angular velocity viewed in the world frame

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

– We have:

$$\mathbf{h}_{\omega} = \boldsymbol{\omega}_{BW} \times \mathbf{z}_{B} = \frac{m}{u_{1}} (\dot{\boldsymbol{a}} - (\mathbf{z}_{B} \cdot \dot{\boldsymbol{a}}) \mathbf{z}_{B}).$$

- 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:

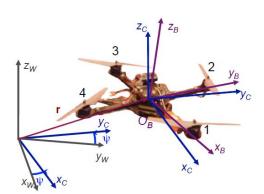
$$\mathbf{h}_{\omega} = \boldsymbol{\omega}_{BW} \times \mathbf{z}_{B} = \frac{m}{u_{1}} (\dot{\boldsymbol{a}} - (\mathbf{z}_{B} \cdot \dot{\boldsymbol{a}}) \mathbf{z}_{B}).$$

- This is the projection of  $\frac{m}{u_1}\dot{a}$  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 A "no tilt" frame

- Angular velocity
  - Quadrotor state:

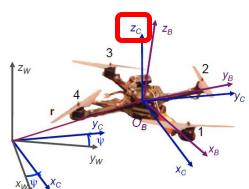
$$\mathbf{X} = [x, y, z, \phi, \theta, \psi, \dot{x}, \dot{y}, \dot{z}, \omega_{x}, \omega_{y}, \omega_{z}]^{T}$$



$$\mathbf{h}_{\omega} = \boldsymbol{\omega}_{BW} \times \mathbf{z}_{B} = \frac{m}{u_{1}} (\dot{\boldsymbol{a}} - (\mathbf{z}_{B} \cdot \dot{\boldsymbol{a}}) \mathbf{z}_{B}).$$

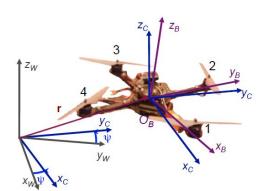
- This is the projection of  $\frac{m}{y_A}\dot{a}$  onto the  $x_B-y_B$  plane
- Since  $\omega_{BW} = \omega_{BC} + \omega_{CW}$ , where  $\omega_{BC}$  has no  $\mathbf{z}_{B}$ component:

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



- 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:

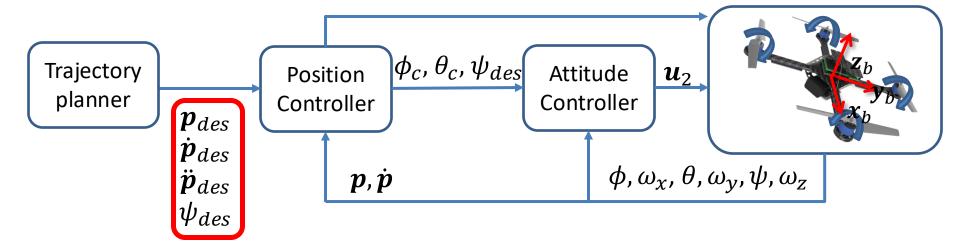
How about Force and Moment Input  $(u_1, u_2)$ ?

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

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



#### Nonlinear dynamics

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

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:

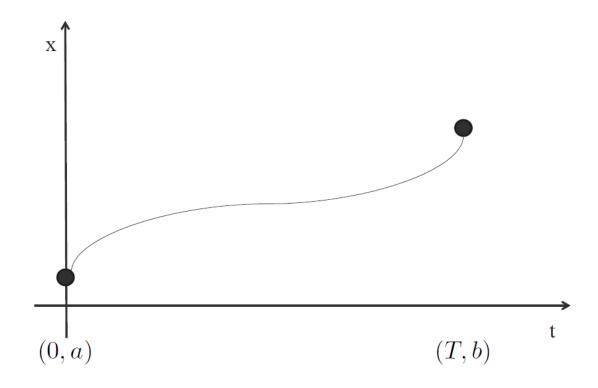
$$-\boldsymbol{\sigma} = [x, y, z, \psi]^T$$

Trajectory in the space of flat outputs:

$$-\boldsymbol{\sigma}(t) = [T_0, T_M] \rightarrow \mathbb{R}^3 \times SO(2)$$

- 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

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



• 5<sup>th</sup> order polynomial trajectory:

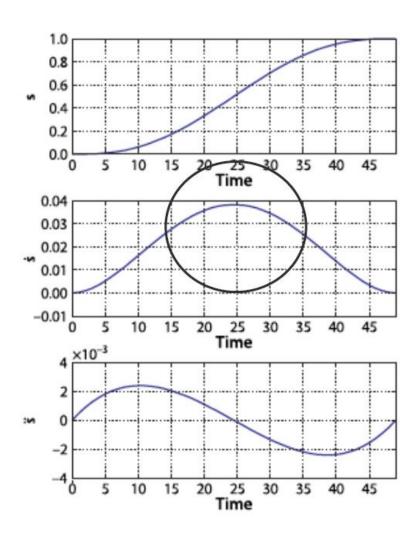
$$-x(t) = c_5 t^5 + c_4 t^4 + c_3 t^3 + c_2 t^2 + c_1 t + c_0$$

Boundary conditions

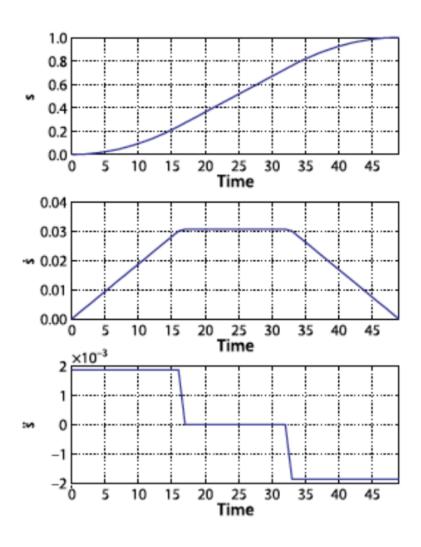
	Position	Velocity	Acceleration
t = 0	a	0	0
t = T	b	0	0

Solve:

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



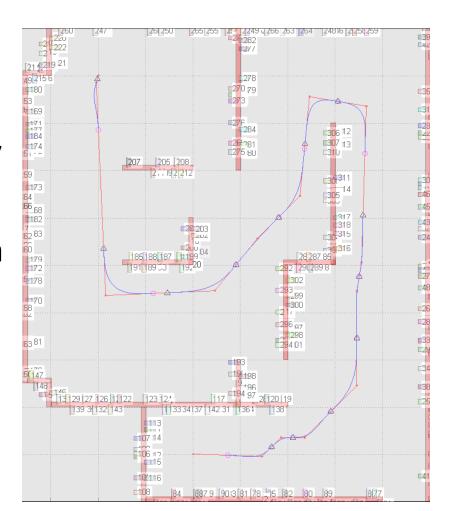
## **Bang-Bang Trajectory**





## 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



• Generate each  $5^{th}$  order polynomial independently:

$$-x(t) = c_5 t^5 + c_4 t^4 + c_3 t^3 + c_2 t^2 + c_1 t + c_0$$

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 (02/18)
  - Due on 02/28. Early submission is encouraged.
  - Notes: Cite the paper, GitHub repo, or code url if you use or reference the code online. Please keep <u>academic integrity</u>; plagiarism is not tolerated in this course.