# Control System Design



### Introduction to Control Theory

- Order of a system
- Linear Time Invariant (LTI) systems
  - Single integrator (kinematic)
  - Double integrator
  - Feedforward, feedback control
- Controller Design
  - Gain tuning
- Linear controller for a quadrotor
  - Planar quadrotor
  - Project 1 Phase 2 introduction



### Control of a simple first-order system

#### **Problem**

State, input

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

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

$$\dot{e} + K_p e = 0$$
  $K_P > 0$  
$$u(t) = \dot{x}^{des}(t) + K_P e(t)$$
 Feedforward

#### Control of a simple second-order system

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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

#### 0.25 0.2 0.15 0.1 0 0.05 -0.05 -0.05 0 0.1 0.2 0.3 0.4 0.5

#### **Strategy**

Find *u* such that

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

$$K_p, K_v > 0$$

$$u(t) = \ddot{x}^{des}(t) + K_V \dot{e}(t) + K_P e(t)$$
Feedforward

Proportional

Derivative

# Control for trajectory tracking in a simple second-order system

#### PD control

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

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

In the presence of disturbances or modeling errors, it is often advantageous to use PID control

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

PID control generates a third-order closed-loop system Integral control makes the steady-state error go to zero



Disadvantages of PID or PD control schemes

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

- performance will depend on the model
- need to tune gains to maximize performance

#### Model based control law

model based 
$$f(t) = m(\ddot{x}_d(t) + k_p e(t) + k_v \dot{e}(t)) + b\dot{x}(t) + kx(t)$$
feedforward + PD feedback model based

Two parts of a model based scheme

- 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

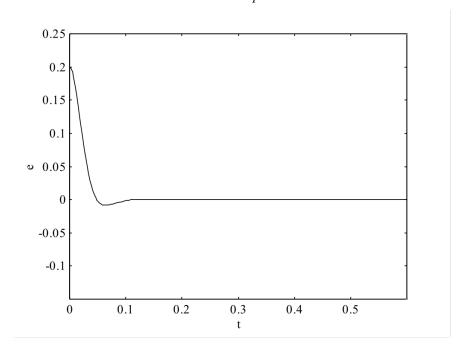
$$m\ddot{x}(t) + b\dot{x}(t) + kx(t) = f(t)$$

Model based control law

model based
$$f(t) = m \left( \ddot{x}_d(t) + k_p e(t) + k_v \dot{e}(t) \right) + b \dot{x}(t) + k x(t)$$
servo

Performance

$$\ddot{e} + k_v \dot{e} + k_p e = 0$$





#### 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

Model based control law (based on estimates of model parameters)

$$f(t) = \hat{m}(\ddot{x}_d(t) + k_p e(t) + k_v \dot{e}(t)) + \hat{b}\dot{x}(t) + \hat{k}x(t)$$

Ideal performance

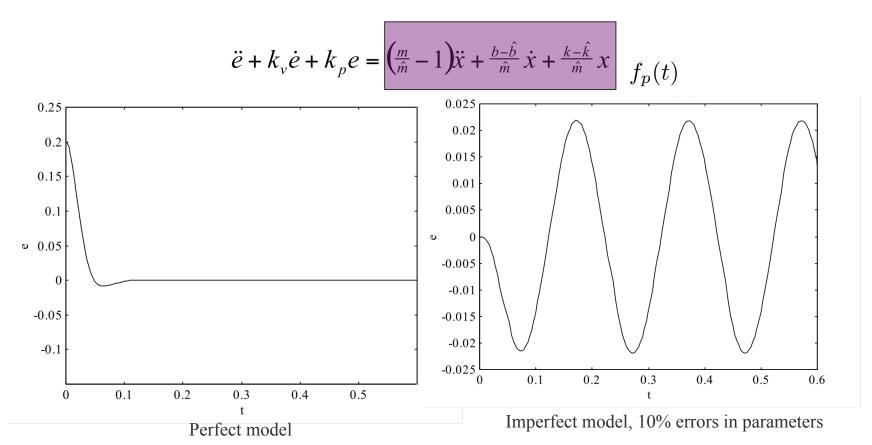
$$\ddot{e} + k_{v}\dot{e} + k_{p}e = 0$$

Actual performance

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

- 1. Error term will not go exponentially to zero
- 2. Right hand side is a forcing function driving the error away from zero





#### Not all is lost however

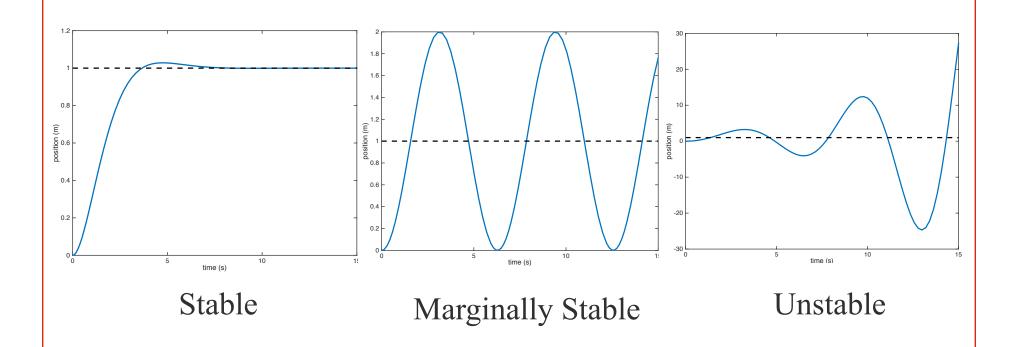
- Treat  $f_p$  as a perturbation or a disturbance force
- -If  $\max_{t} f_p(t) < M$  we can prove that the error e(t) is also bounded



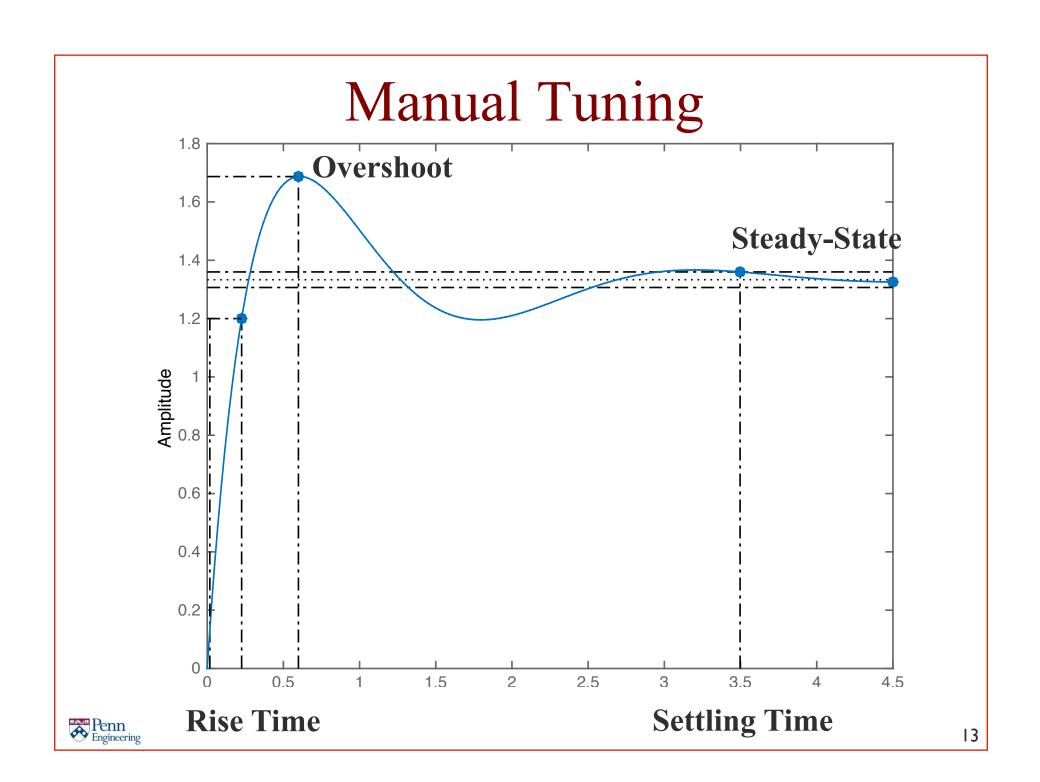
# Gain Tuning



# Manual Tuning

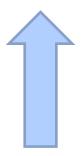






# Manual Tuning

Parameter Increased	K <sub>p</sub>	K <sub>d</sub>	K <sub>i</sub>
Rise Time	Decrease	-	Decrease
Overshoot	Increase	Decrease	Increase
Settling Time	-	Decrease	Increase
Steady-State Error	Decrease	n/a	Eliminate









### 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:

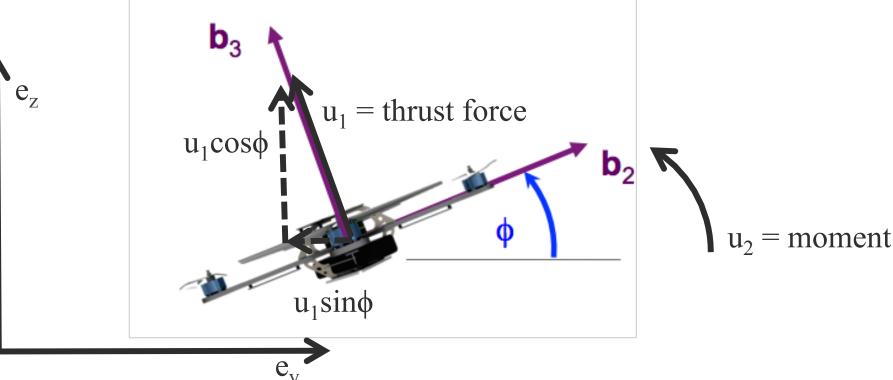
Controller	K <sub>p</sub>	K <sub>d</sub>	K <sub>i</sub>
Р	0.50K <sub>u</sub>	-	-
PD	0.80K <sub>u</sub>	K <sub>p</sub> T <sub>u</sub> /8	-
PID	0.60K <sub>u</sub>	2K <sub>p</sub> /T <sub>u</sub>	K <sub>p</sub> T <sub>u</sub> /8



# Application to Quadrotors



### Planar Quadrotor Model



$$\mathbf{x} = [y \ z \ \phi \ \dot{y} \ \dot{z} \ \dot{\phi}]^T$$

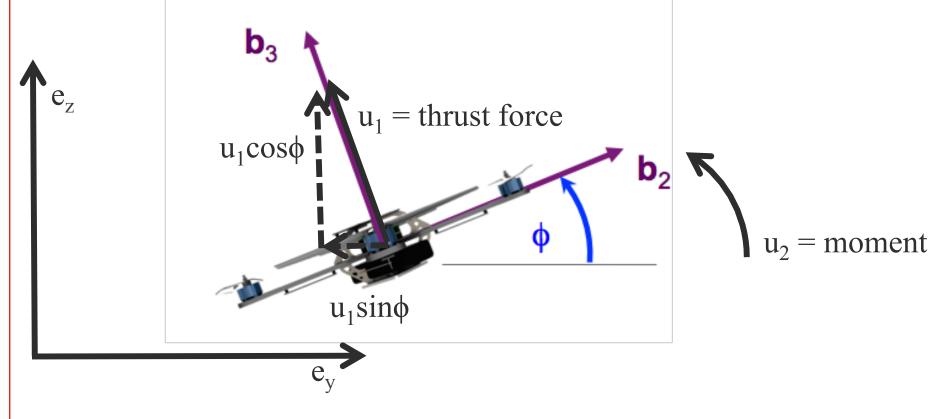
$$\sum \mathbf{F}_y = -u_1 \sin(\phi) = m\ddot{y}$$

$$\sum \mathbf{F}_z = -mg + u_1 \cos(\phi) = m\ddot{z}$$

$$\mathbf{M} = u_2 = I_{xx}\ddot{\phi}$$



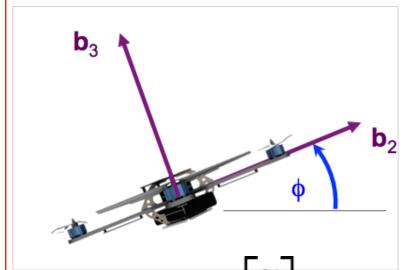




$$\begin{bmatrix} \ddot{y} \\ \ddot{z} \\ \ddot{\phi} \end{bmatrix} = \begin{bmatrix} 0 \\ -g \\ 0 \end{bmatrix} + \begin{bmatrix} -\frac{1}{m}\sin\phi & 0 \\ \frac{1}{m}\cos\phi & 0 \\ 0 & \frac{1}{I_{xx}} \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$



### Planar Quadrotor Model



$$\begin{bmatrix} \ddot{y} \\ \ddot{z} \\ \ddot{\phi} \end{bmatrix} = \begin{bmatrix} 0 \\ -g \\ 0 \end{bmatrix} + \begin{bmatrix} -\frac{1}{m}\sin\phi & 0 \\ \frac{1}{m}\cos\phi & 0 \\ 0 & \frac{1}{I_{xx}} \end{bmatrix} \begin{bmatrix} u_1 \\ u_2 \end{bmatrix}$$

$$x = egin{bmatrix} x_1 \ x_2 \end{bmatrix} = egin{bmatrix} \dot{z} \ \dot{\phi} \ \dot{\dot{z}} \ \dot{\dot{\phi}} \end{bmatrix}$$

### Linearized Dynamic Model

Nonlinear dynamics

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

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

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

Equilibrium hover configuration

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

Linearized dynamics

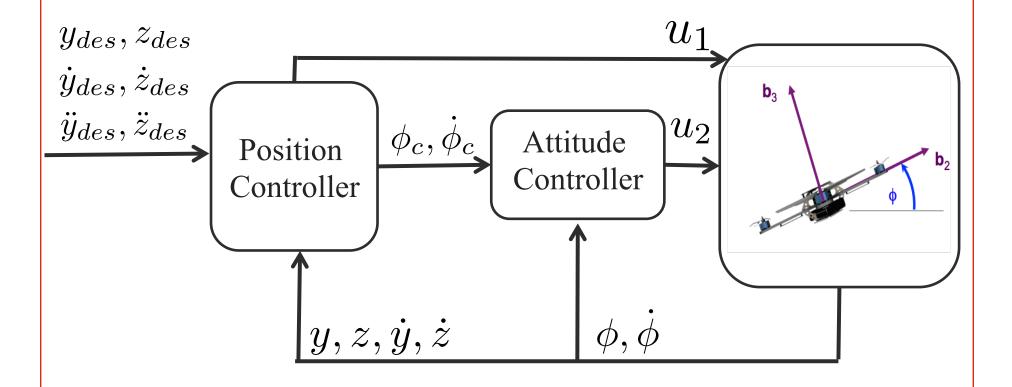
$$\ddot{y} = -g\phi$$

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

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



#### Nested Control Structure





Lateral dynamics

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

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

Attitude control

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

Position control - determining  $\phi_c, \dot{\phi}_c$ 

$$\phi_c = -\frac{\ddot{y}_c}{g}$$

$$\dot{\phi}_c = 0$$



Vertical dynamics

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

Z-position control

$$u_1 = m(g + \ddot{z}_c)$$

Determining  $\ddot{y}_c, \ddot{z}_c$ 

$$\mathbf{x} = \begin{bmatrix} y & z \end{bmatrix}^T$$

$$\ddot{\mathbf{e}} + k_{d,x}\dot{\mathbf{e}}_{des} + k_{p,x}\mathbf{e} = 0$$



Determining  $\ddot{y}_c, \ddot{z}_c$ 

$$\mathbf{x} = [y \ z]^T$$

$$(\ddot{\mathbf{x}}_{des} - \ddot{\mathbf{x}}_c) + k_{d,x}(\dot{\mathbf{x}}_{des} - \dot{\mathbf{x}}) + k_{p,x}(\mathbf{x}_{des} - \mathbf{x}) = 0$$
Actual states

Specified by trajectory

Note:

$$u_1 = m(g + \ddot{z}_{des} + k_{d,z}(\dot{z}_{des} - \dot{z}) + k_{p,z}(z_{des} - z))$$

$$\phi_c = -\frac{1}{g}(\ddot{y}_{des} + k_{d,y}(\dot{y}_{des} - \dot{y}) + k_{p,y}(y_{des} - y))$$
Model-based Feedforward Feedback

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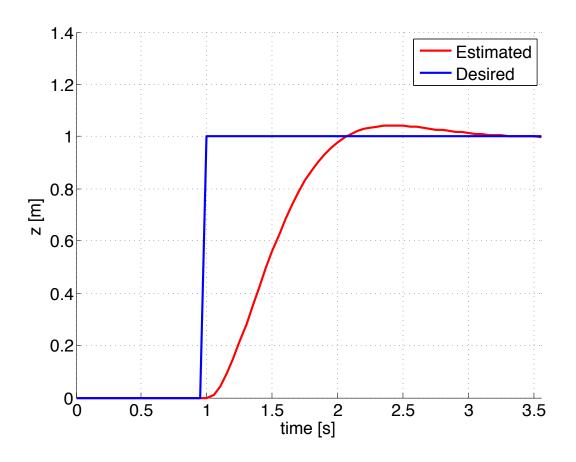
$$u_{1} = m(g + \ddot{z}_{des} + (k_{d,z})(\dot{z}_{des} - \dot{z}) + (k_{p,z})(z_{des} - z))$$

$$u_{2} = (k_{p,\phi})(\phi_{c} - \phi) + (k_{d,\phi})(\dot{\phi}_{c} - \dot{\phi})$$

$$\phi_{c} = -\frac{1}{g}(\ddot{y}_{des} + (k_{d,y})(\dot{y}_{des} - \dot{y}) + (k_{p,y})(y_{des} - y))$$

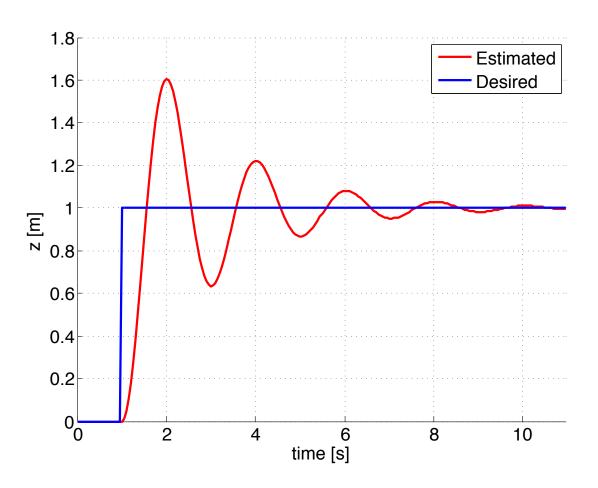


### PD Controller



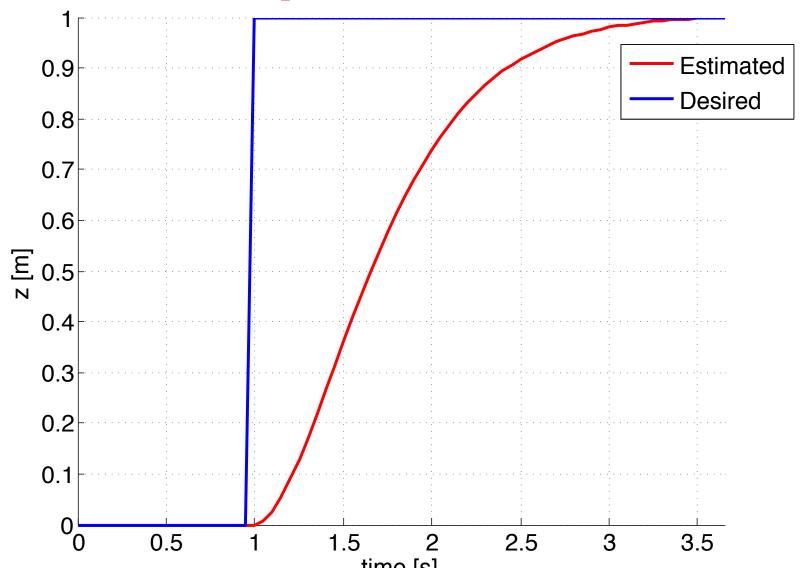


# High K<sub>p</sub>



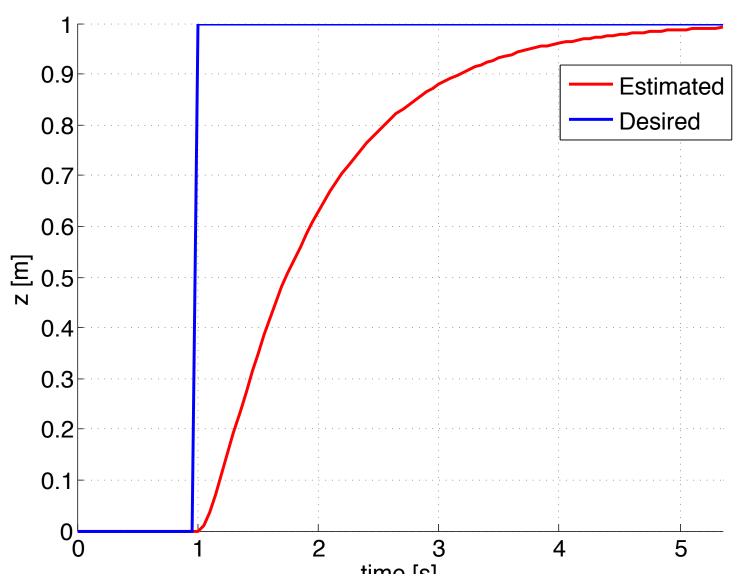


# Low K<sub>p</sub> (soft response)





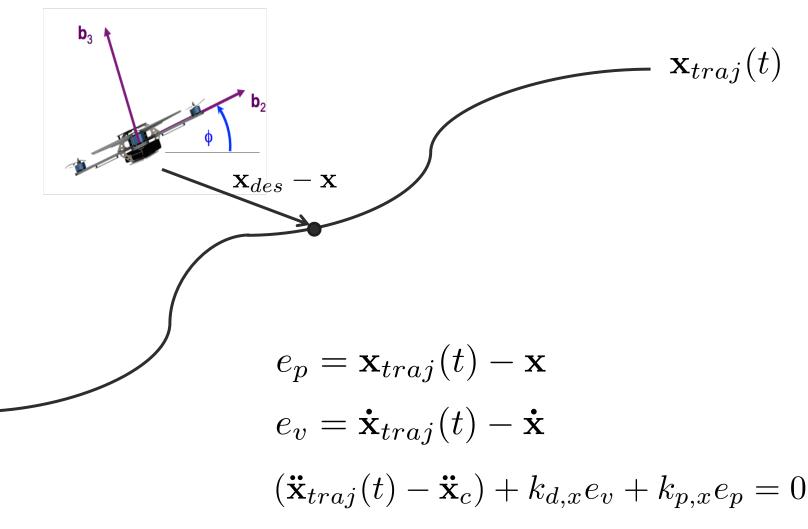
# High K<sub>d</sub> (overdamped)





### Trajectory Tracking

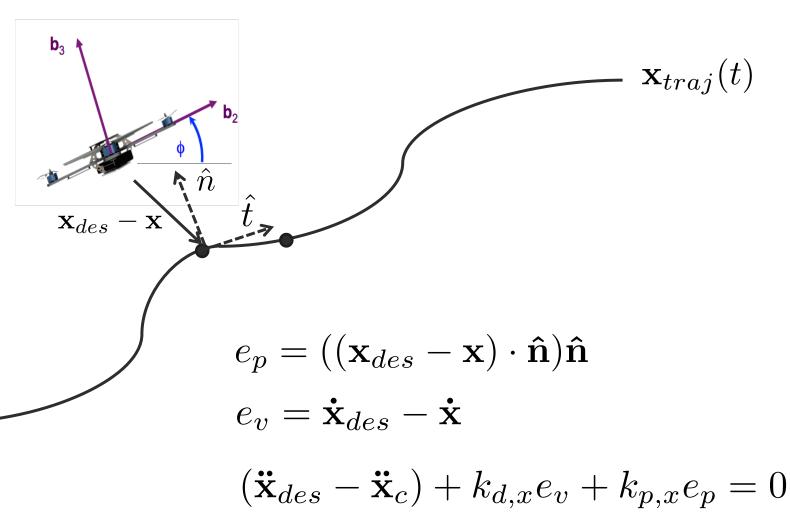
Given  $\mathbf{x}_{traj}(t), \mathbf{\dot{x}}_{traj}(t), \mathbf{\ddot{x}}_{traj}(t)$ 





# Trajectory Tracking

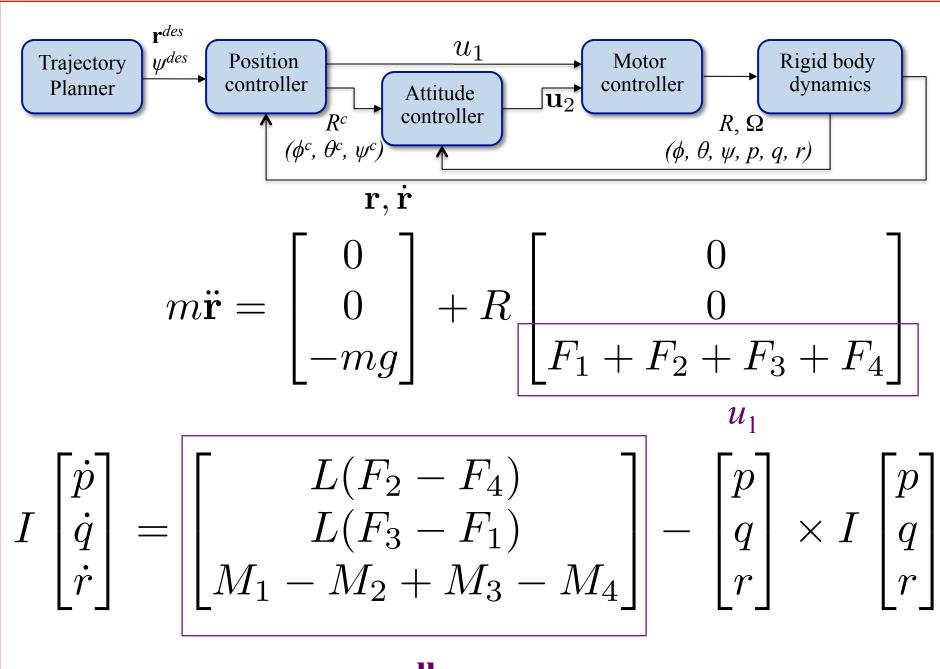
Given  $\mathbf{x}_{traj}(t), \mathbf{\dot{x}}_{traj}(t), \mathbf{\ddot{x}}_{traj}(t)$ 





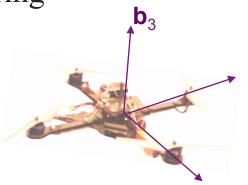
# 3-D Quadrotor





Penn Engineering

#### Control for Hovering



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

Linearization

$$(u_1 \sim mg, \theta \sim 0, \phi \sim 0, \psi \sim \psi_0)$$
  
 $\ddot{r}_1 = \ddot{x} = g(\theta \cos \psi + \phi \sin \psi)$   
 $\ddot{r}_2 = \ddot{y} = g(\theta \sin \psi - \phi \cos \psi)$ 



