



# Autonomous Robotics

## Spring 2026

Abhishek Gupta, Siddhartha Srinivasa

TAs: Helen Wang, Sidharth Talia, Rohan Baijal, Christopher Tan



Recap

# Computing control laws

We will only control steering angle;  
fixed constant speed  
As a result, no real control for along-track  
error  
Some control laws will only minimize cross-  
track error, others will also minimize heading



$$u = K(e)$$

# Bang-bang control

Simple control law - choose between hard left and hard right



$$u = \begin{cases} u_{max} & \text{if } e_{ct} < 0 \\ -u_{max} & \text{otherwise} \end{cases}$$

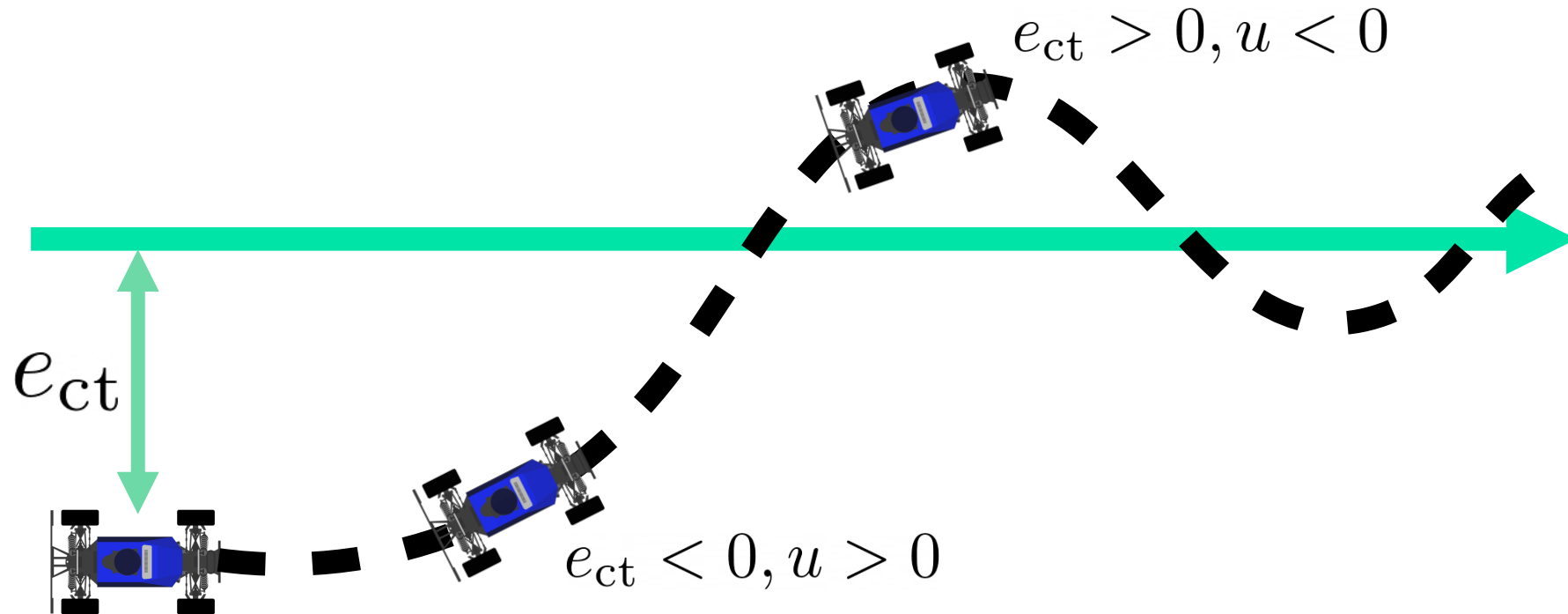
# PID control

Select a control law that tries to drive error to zero (and keep it there)



$$u = - \left( \underbrace{K_p e_{ct}}_{\text{PROPORTIONAL (PRESENT)}} + \underbrace{K_i \int e_{ct} dt}_{\text{INTEGRAL (PAST)}} + \underbrace{K_d \dot{e}_{ct}}_{\text{DERIVATIVE (FUTURE)}} \right)$$

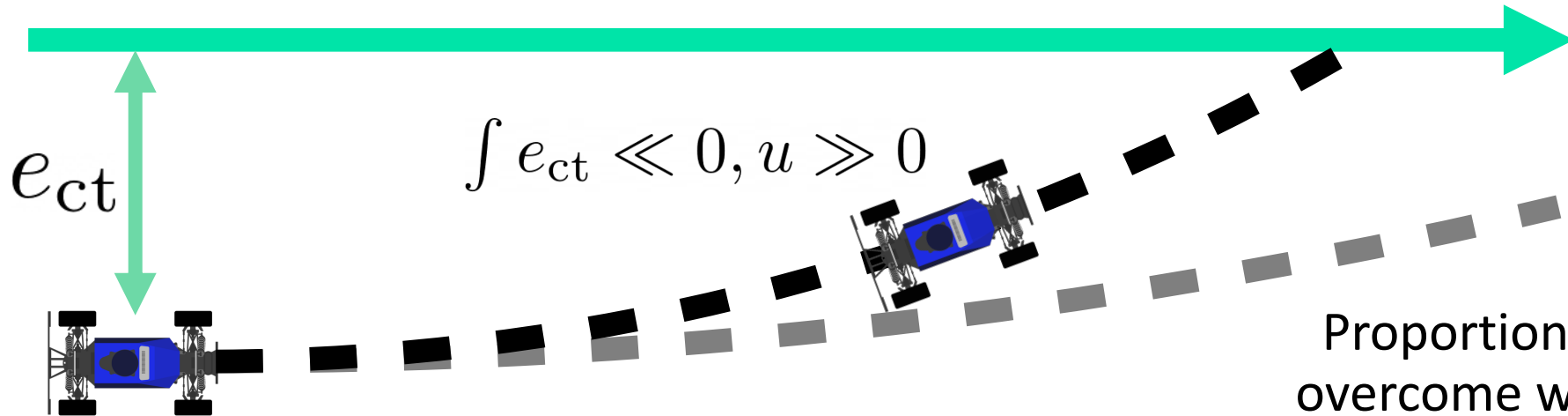
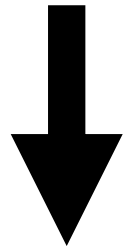
# Proportional Control



$$u = -K_p e_{ct}$$

# Proportional Integral (PI) Control

WIND

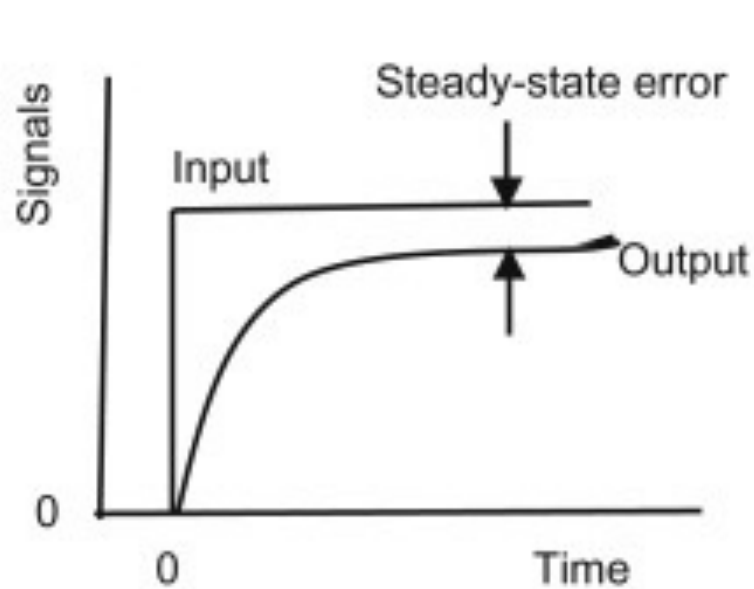


$$u = - \left( K_p e_{ct} + K_i \int e_{ct} dt \right)$$

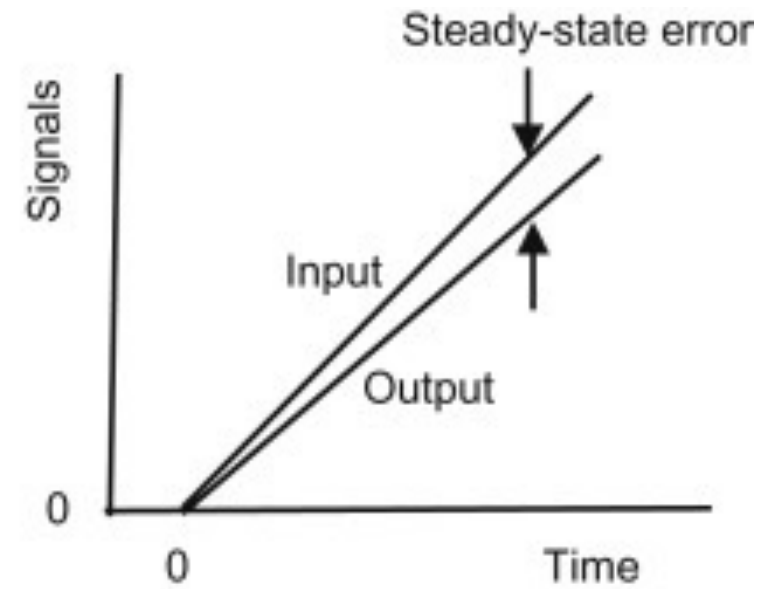
# Proportional Integral (PI) Control

$$u = - \left( K_p e_{ct} + K_i \int e_{ct} dt \right)$$

Integral control gets rid of this term since the integral keeps growing



(a)



(b)

# Lecture Outline

---

Recap



PID Control Wrap-up



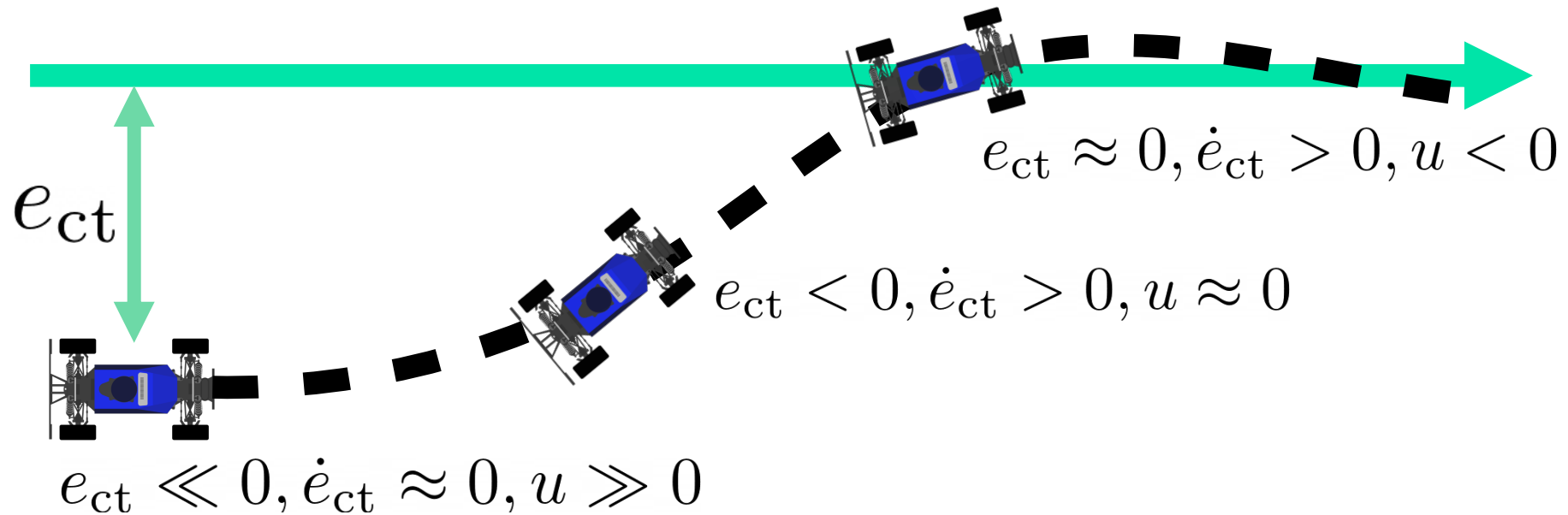
Pure Pursuit



Optimal Control

# Proportional Derivative (PD) Control

Apply the brakes when moving too fast! → converge to the steady state



$$u = - (K_p e_{ct} + K_d \dot{e}_{ct})$$

# How do you evaluate the derivative term?

**Terrible way:** Numerically differentiate error. Why is this a bad idea?

**Smart way:** Analytically compute the derivative of the cross track error

$$e_{ct} = -\sin(\theta_{ref})(x - x_{ref}) + \cos(\theta_{ref})(y - y_{ref})$$

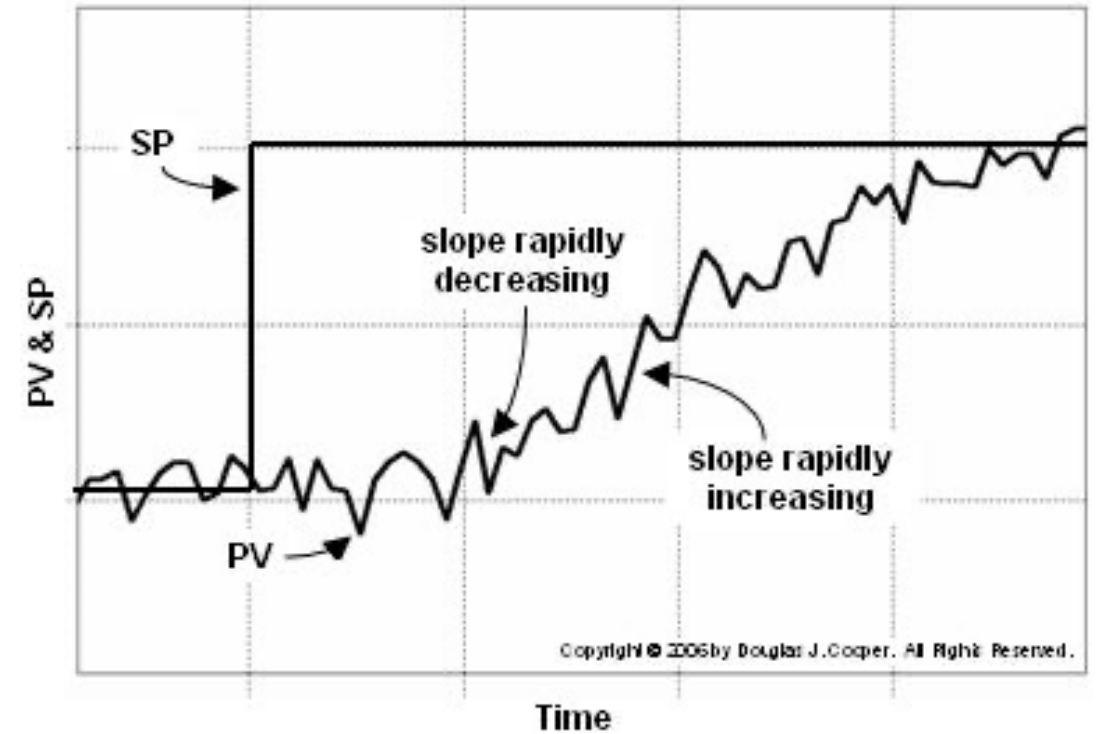
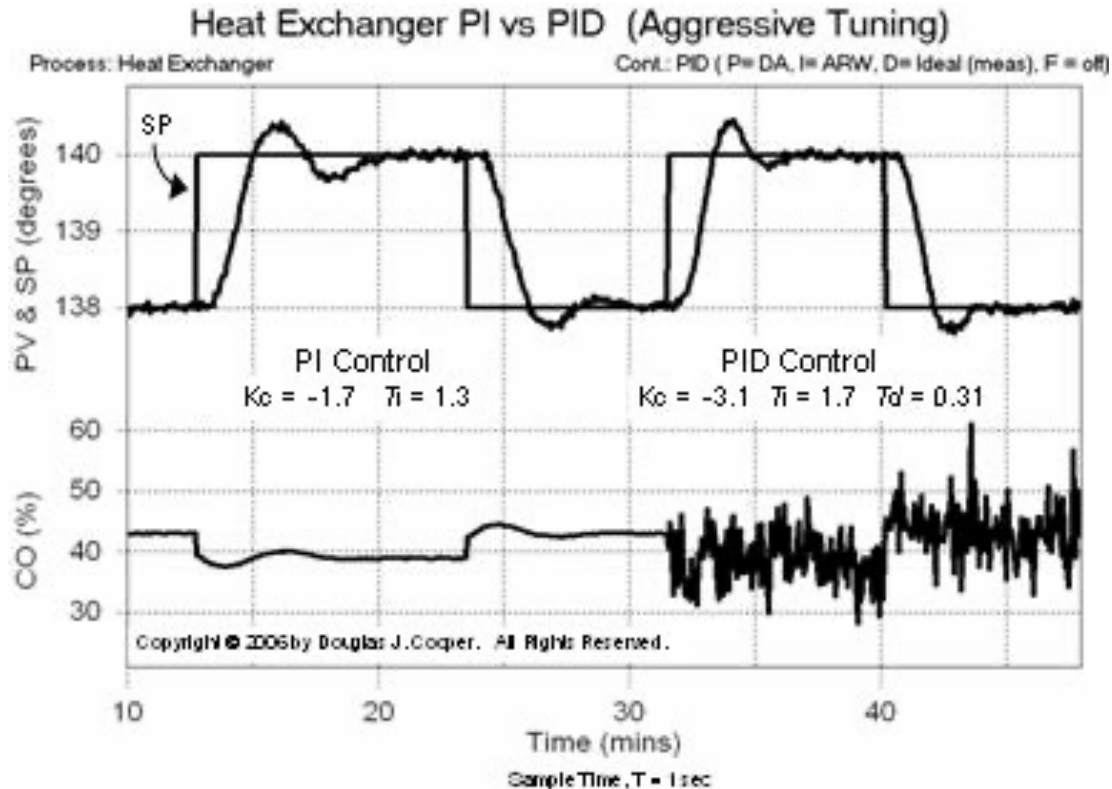
$$\begin{aligned}\dot{e}_{ct} &= -\sin(\theta_{ref})\dot{x} + \cos(\theta_{ref})\dot{y} \\ &= -\sin(\theta_{ref})V \cos(\theta) + \cos(\theta_{ref})V \sin(\theta) \\ &= V \sin(\theta - \theta_{ref}) = V \sin(\theta_e)\end{aligned}$$

New control law! Penalize error in cross track **and in heading**

$$u = - (K_p e_{ct} + K_d V \sin \theta_e)$$

# Challenges with using the derivative term

Noise can lead to wildly changing derivatives – leading to huge control variations



# PID Intuition

$$u = - \left( \underbrace{K_p e_{ct}}_{\substack{\text{PROPORTIONAL} \\ \text{(PRESENT)}}} + \underbrace{K_i \int e_{ct} dt}_{\substack{\text{INTEGRAL} \\ \text{(PAST)}}} + \underbrace{K_d \dot{e}_{ct}}_{\substack{\text{DERIVATIVE} \\ \text{(FUTURE)}}} \right)$$

Proportional: minimize the current error!

Integral: if I'm accumulating error, try harder!

Derivative: if I'm going to overshoot, slow down!

# Tuning PID controllers

$$u = - \left( K_p e_{ct} + K_i \int e_{ct} dt + K_d \dot{e}_{ct} \right)$$

**PROPORTIONAL**  
(PRESENT)

**INTEGRAL**  
(PAST)

**DERIVATIVE**  
(FUTURE)

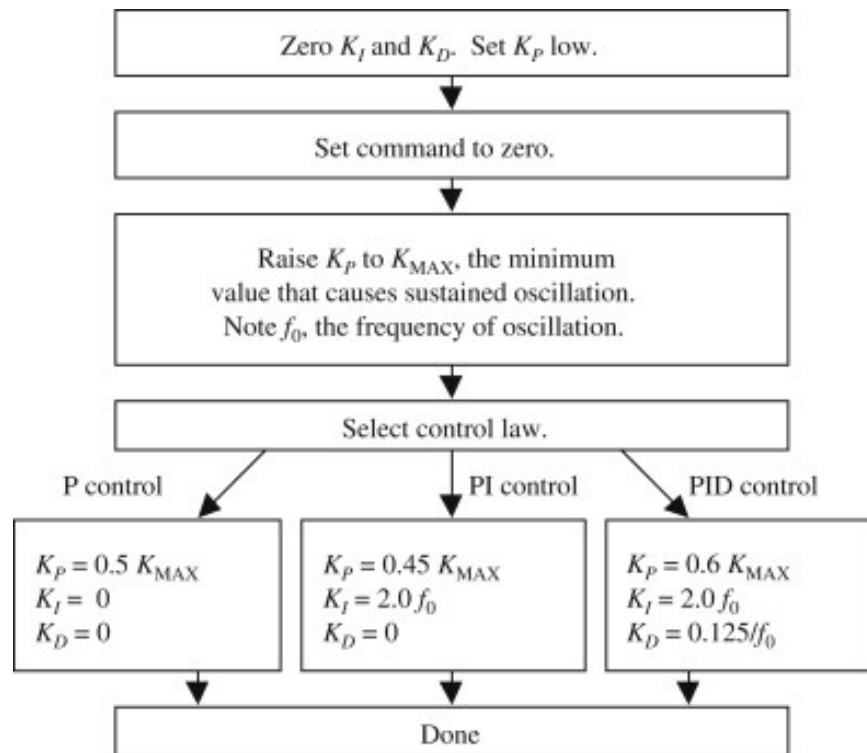
The diagram shows the PID controller equation  $u = - (K_p e_{ct} + K_i \int e_{ct} dt + K_d \dot{e}_{ct})$ . The three terms are highlighted in colored boxes:  $K_p e_{ct}$  is in a yellow box,  $K_i \int e_{ct} dt$  is in a grey box, and  $K_d \dot{e}_{ct}$  is in a yellow box. Below each box is a label: 'PROPORTIONAL (PRESENT)' for the first, 'INTEGRAL (PAST)' for the second, and 'DERIVATIVE (FUTURE)' for the third. Three arrows point from the labels to their respective terms in the equation.

How do you set the  $K_p$ ,  $K_i$ ,  $K_d$  constants for a particular system?

# Tuning PID controllers: Ziegler-Nichols

Heuristic/empirical method for computing  $K_p$ ,  $K_i$ ,  $K_d$

$$u = - \left( K_p e_{ct} + K_i \int e_{ct} dt + K_d \dot{e}_{ct} \right)$$



See how the system responds to proportional gain

Adjust integral and proportional accordingly

# Lecture Outline

---

Recap



**PID Control Wrap-up**

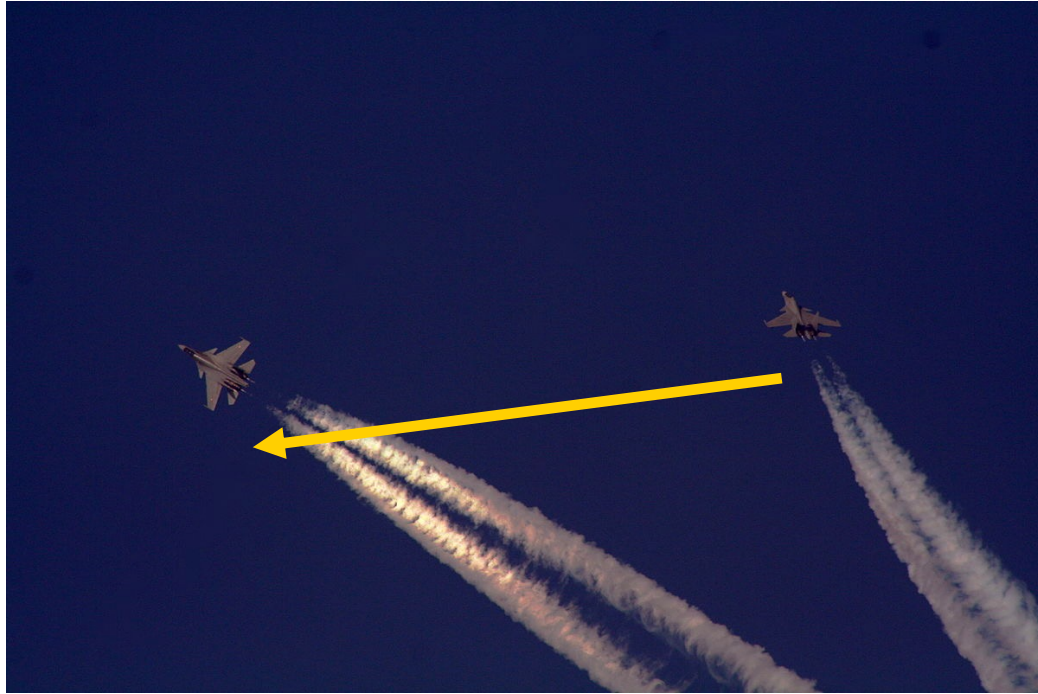


Pure Pursuit



Optimal Control

# Pure Pursuit Control



Aerial combat in which aircraft **pursues** another aircraft by pointing its nose directly towards it



Similar to  
carrot on a stick!

# Rationale: Controller should leverage model!

---

$$\dot{x} = v \cos \theta$$

$$\dot{y} = v \sin \theta$$

$$\dot{\theta} = \omega = \frac{v}{R} = \frac{v \tan \delta}{L}$$

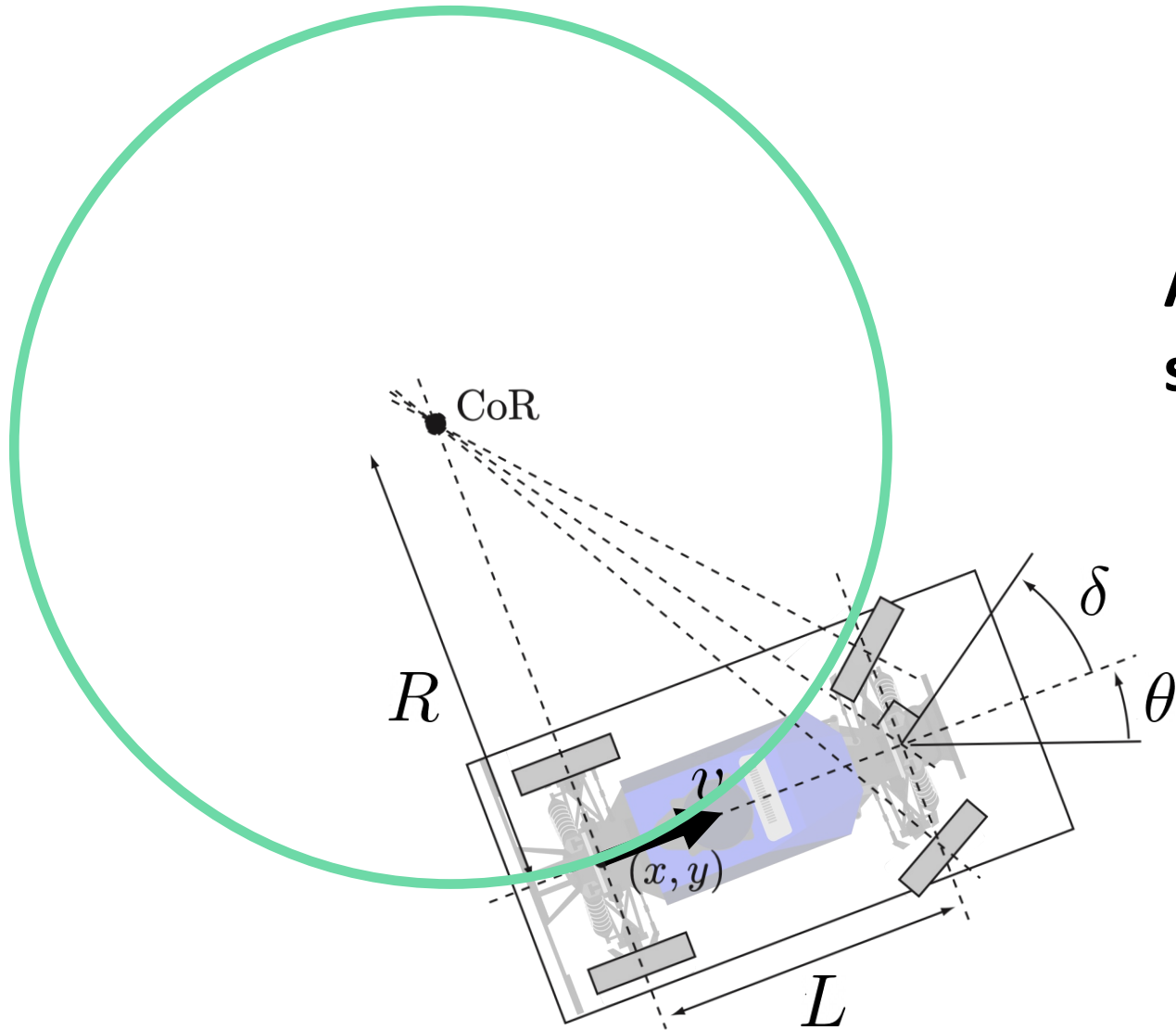
PID control doesn't directly utilize the fact that we know the kinematic car model

---

Key Idea:

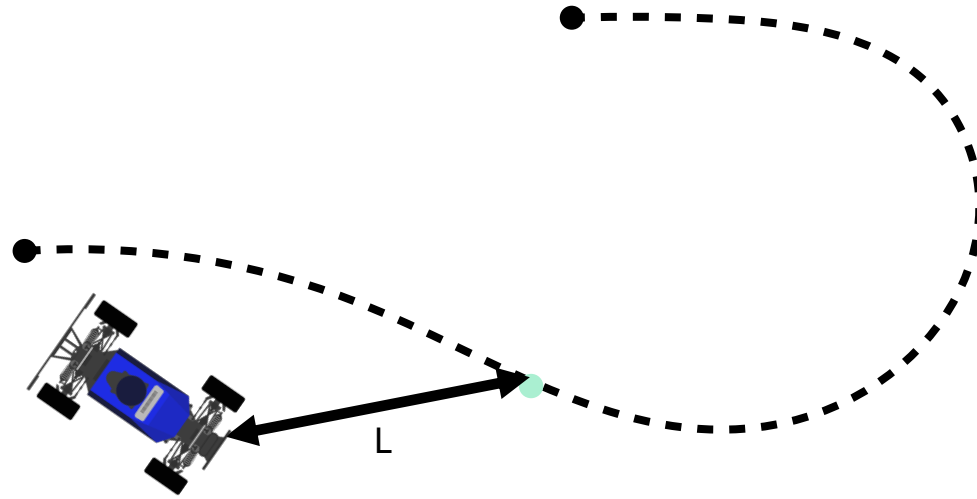
The car is *always* moving  
in a circular arc

# Pure Pursuit Controller



**Assume the car is moving with fixed steering angle**

# Consider a reference at a lookahead distance

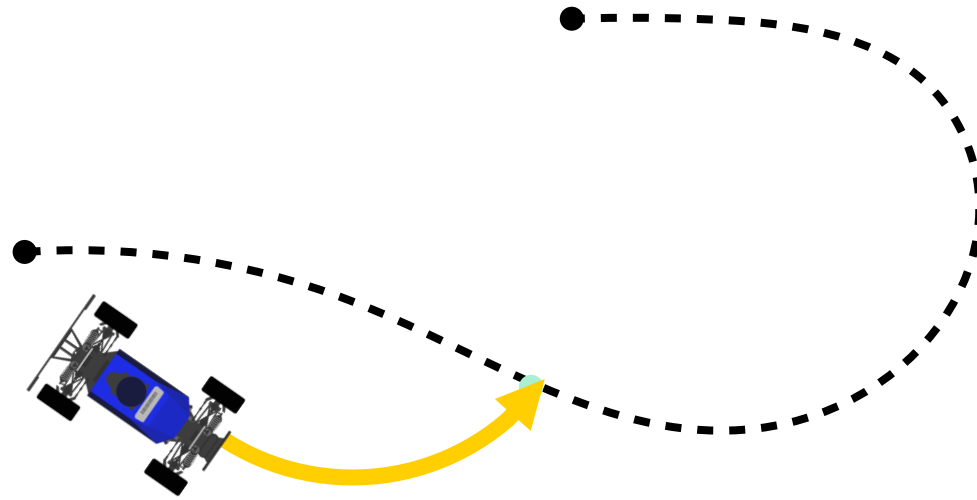


$$\left\| \begin{bmatrix} x \\ y \end{bmatrix} - \begin{bmatrix} x_{ref} \\ y_{ref} \end{bmatrix} \right\| = L$$

**Problem:** Can we solve for a steering angle that guarantees that the car will pass through the reference?

# Solution: Compute a circular arc

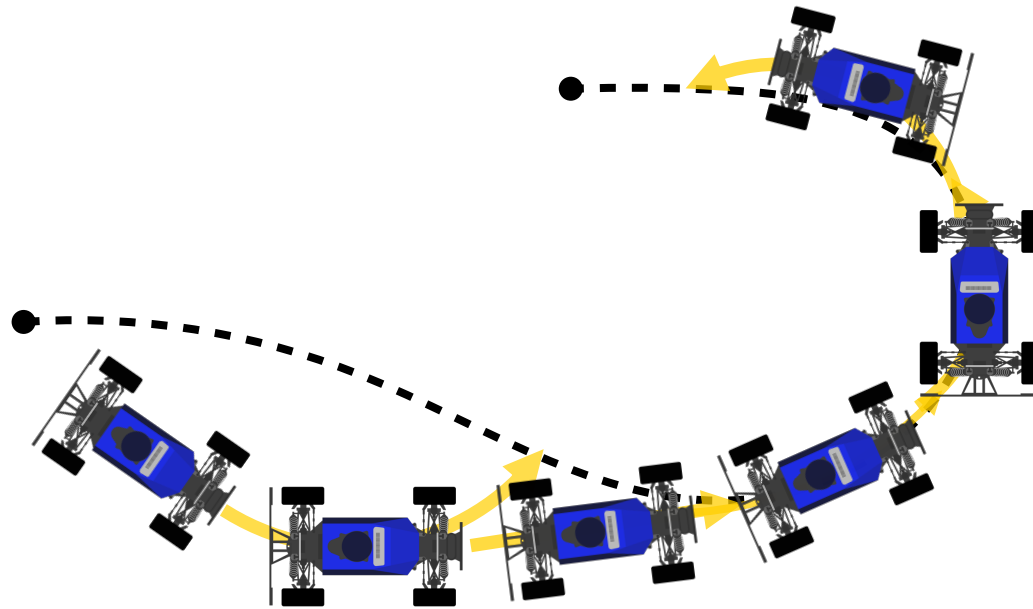
---



We can always solve for a arc that passes through a lookahead point

**Note:** As the car moves forward, the point keeps moving

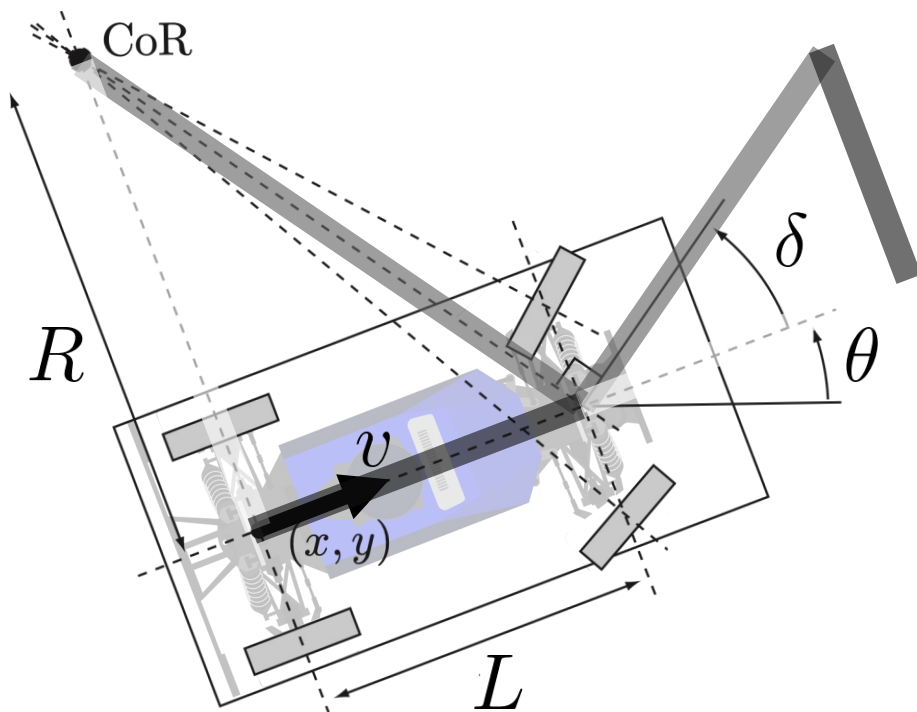
# Pure pursuit: Keep chasing lookahead



1. Find a lookahead and compute arc
2. Move along the arc
3. Go to step 1

# Equations of Motion

RECALL



$$\dot{x} = v \cos \theta$$

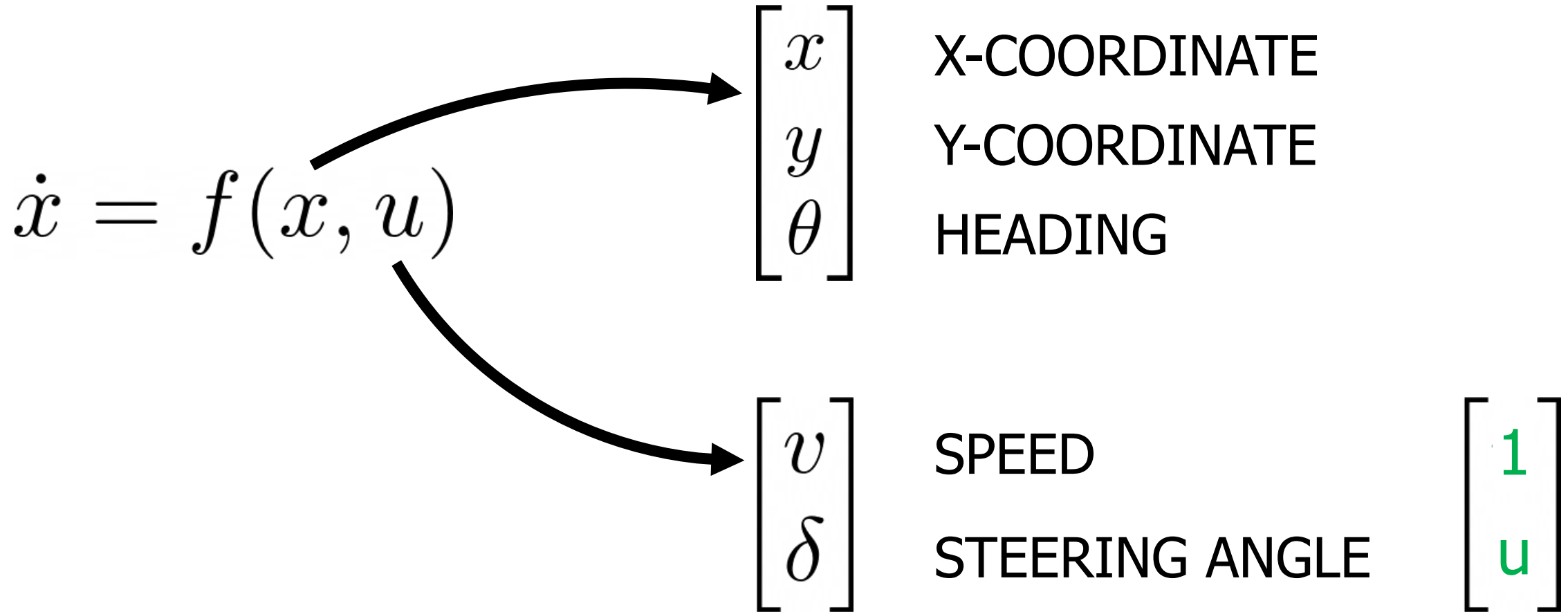
$$\dot{y} = v \sin \theta$$

$$\dot{\theta} = \omega = \frac{v}{R} = \frac{v \tan \delta}{L}$$

$$\tan \delta = \frac{L}{R} \rightarrow R = \frac{L}{\tan \delta}$$

# Kinematic Car Model

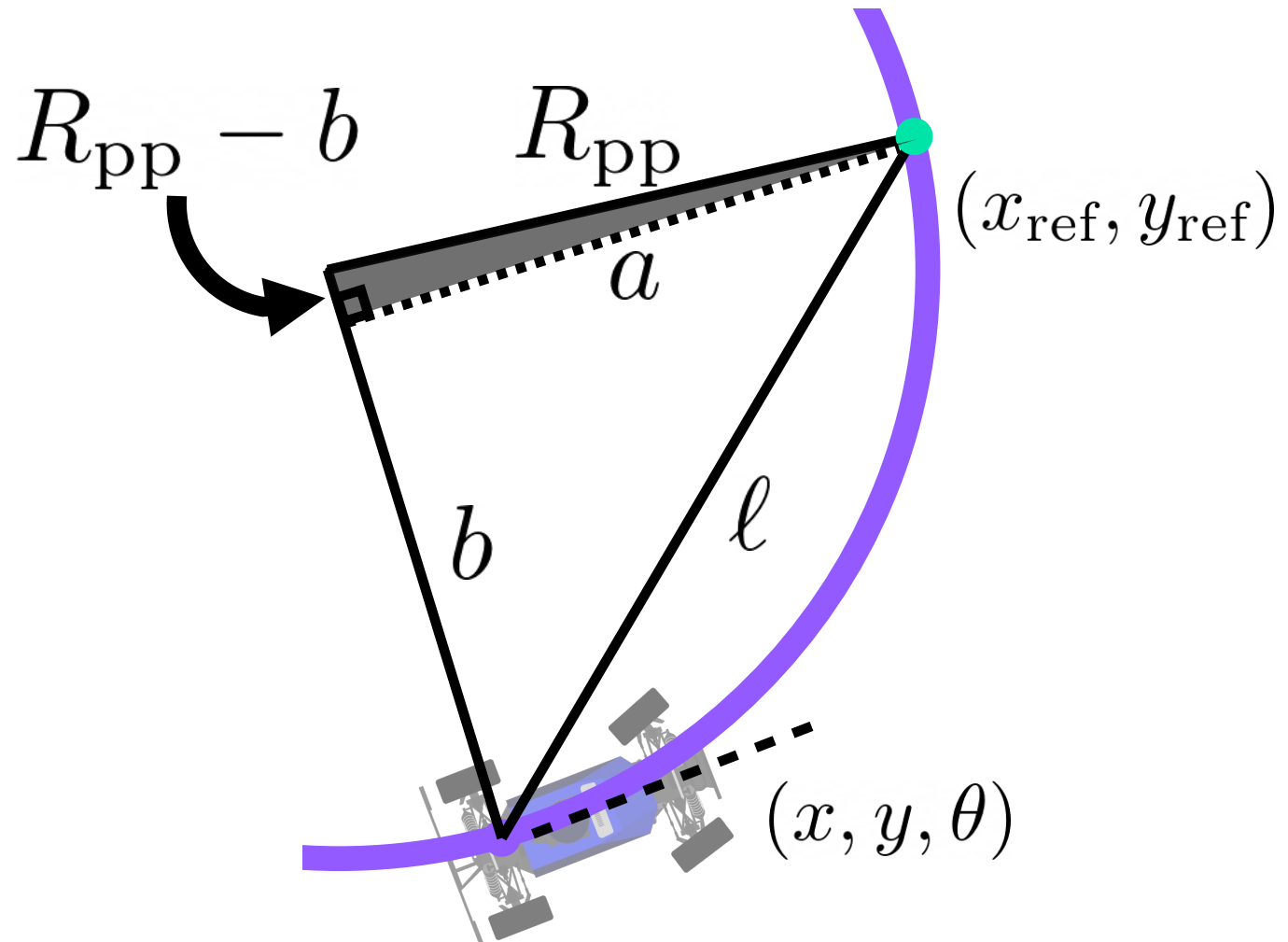
RECALL



# Computing the Arc Radius

$$(R_{pp} - b)^2 + a^2 = R_{pp}^2$$

$$R_{pp} = \frac{a^2 + b^2}{2b}$$

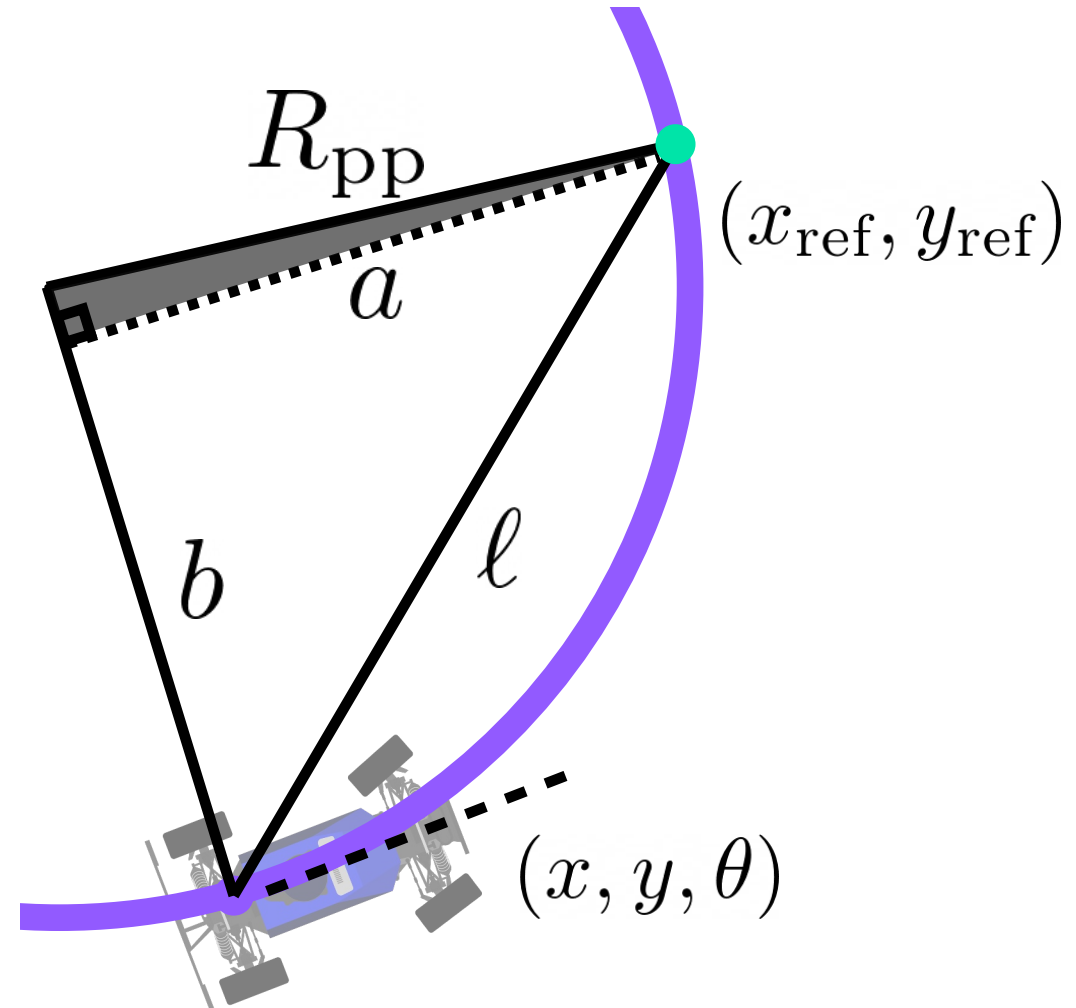


# Computing the Arc Radius

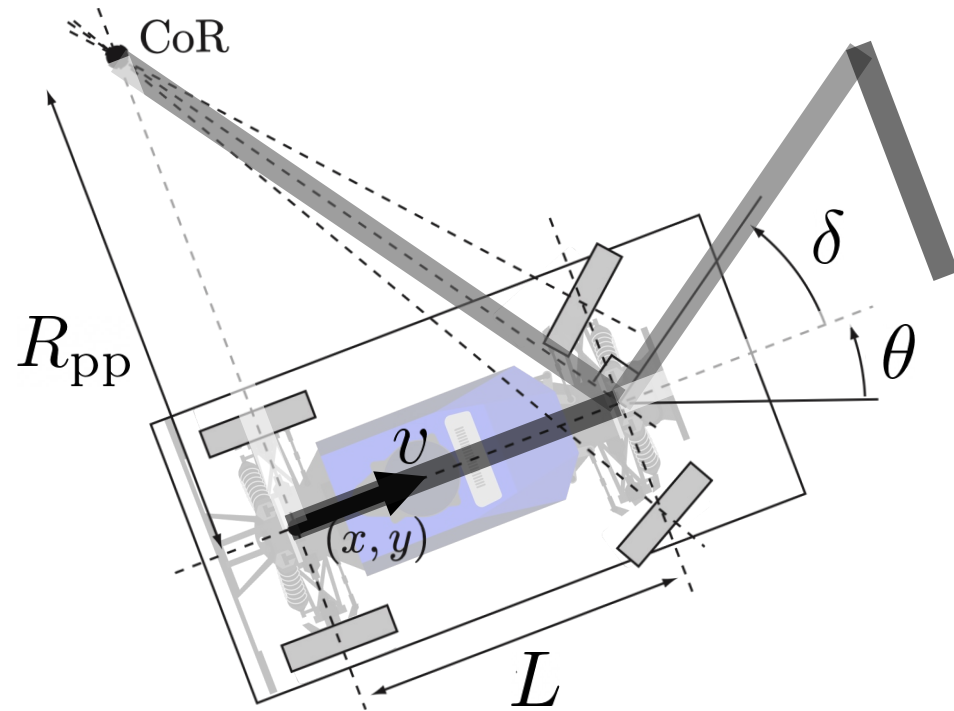
$$R_{pp} = \frac{a^2 + b^2}{2b}$$

$$\begin{bmatrix} a \\ b \end{bmatrix} = R(-\theta) \left( \begin{bmatrix} x_{\text{ref}} \\ y_{\text{ref}} \end{bmatrix} - \begin{bmatrix} x \\ y \end{bmatrix} \right)$$

**Different than cross-track error  
(this is ref. position in robot frame;  
vice versa for cross-track error)**



# Computing the Steering Angle

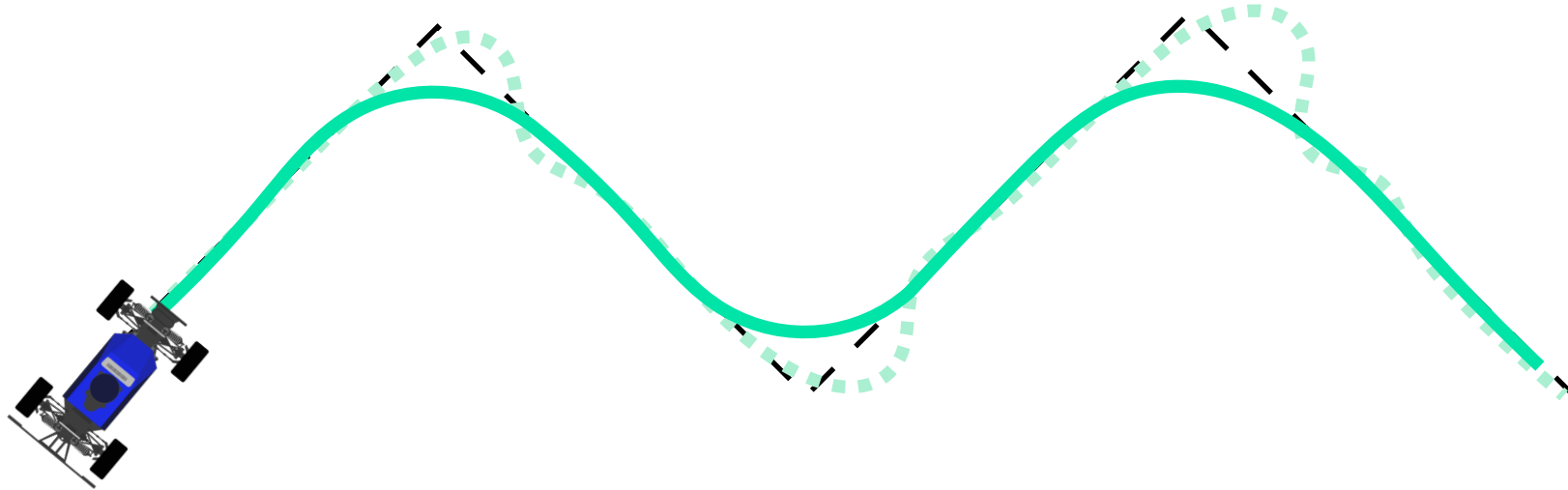


$$R_{pp} = \frac{a^2 + b^2}{2b}$$

$$\tan \delta = \frac{L}{R_{pp}}$$

# Question: How do I choose L?

---



# Controller Design Decisions

---

1. Get a reference path/trajectory to track
2. Pick a reference state from the reference path/trajectory
3. Compute error to reference state
4. Compute control law to minimize error



Option 1:

Bang-bang control



Option 2:

PID control



Option 3:

Pure-pursuit control

Are we done?

# Lecture Outline

---

**Recap**



**PID Control Wrap-up**

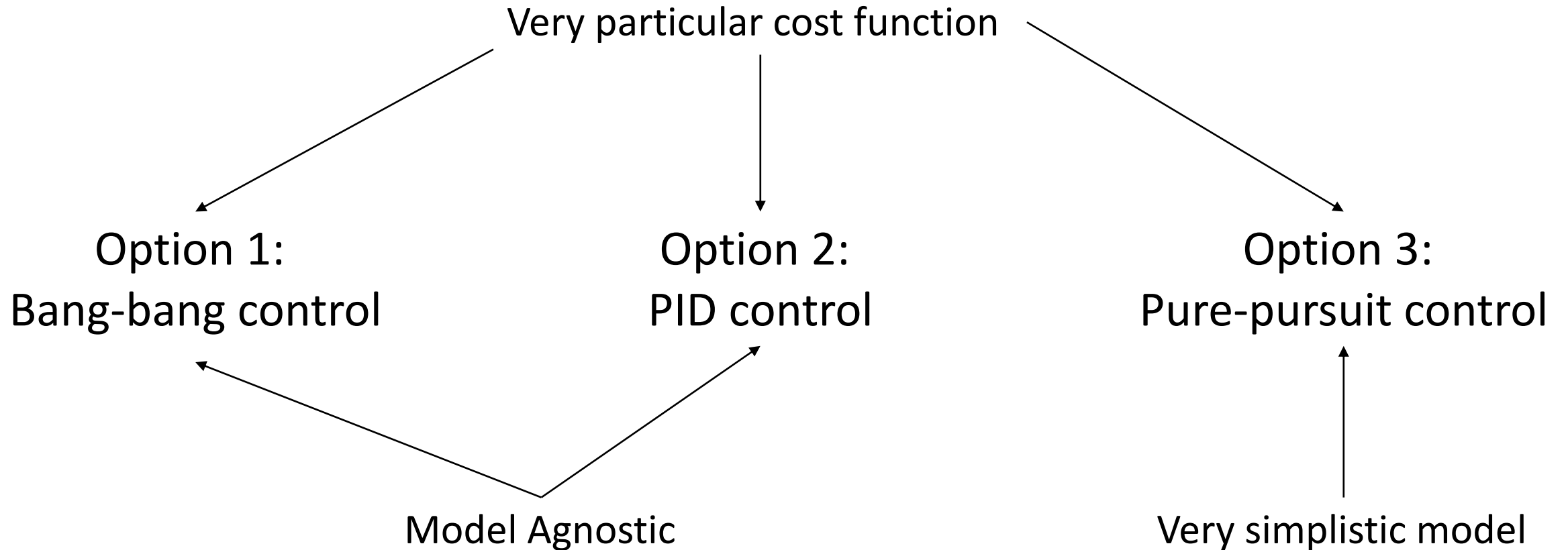


**Pure Pursuit**



Optimal Control

# Controller Design Decisions



# Control as an Optimization Problem

---

For a sequence of  $H$  control actions

1. Use model to predict consequence of actions (i.e.,  $H$  future states)
2. Evaluate the cost function

Compute optimal sequence of  $H$  control actions (minimizes cost)

# Generalized Problem: Optimal Control

- Minimize sum of costs, subject to dynamics and other constraints

$$\begin{aligned} \min_{u_{1:T}} \sum_{t=1}^T c(x_t, u_t) \\ \text{s.t. } x_{t+1} = f(x_t, u_t) \end{aligned}$$

Can be costs like smoothness, preferences, speed

Can be constraints like velocity/acceleration bounds

# Linear Quadratic Regulator

---

Linear system (model)

Quadratic cost function to minimize

$$x_{t+1} = Ax_t + Bu_t$$
$$\sum_t x_t^\top Q x_t + u_t^\top R u_t$$

# Linear System

Linear system (model)

Quadratic cost function to minimize

$$x_{t+1} = Ax_t + Bu_t$$
$$\sum_t x_t^\top Q x_t + u_t^\top R u_t$$

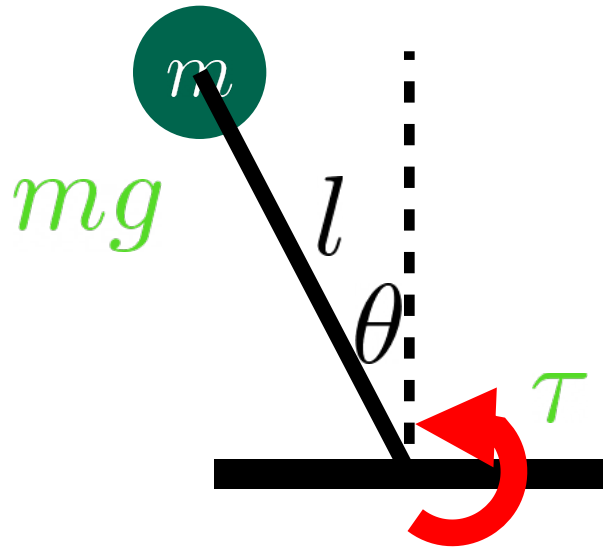
$$x_{t+1} = A x_t + B u_t$$

(N x 1)      (N x N)(N x 1)      (N x M)(M x 1)

**STATE** → **NEXT STATE**

**CONTROL** → **NEXT STATE**

# Example: Inverted Pendulum (Linear System)



$$mgl \sin \theta + \tau = ml^2 \ddot{\theta}$$

$$\ddot{\theta} = \frac{g}{l} \sin \theta + \frac{\tau}{ml^2} \approx \frac{g}{l} \theta + \frac{\tau}{ml^2}$$

$$\begin{bmatrix} \theta_{t+1} \\ \dot{\theta}_{t+1} \end{bmatrix} = \begin{bmatrix} 1 & \Delta t \\ \frac{g}{l} \Delta t & 1 \end{bmatrix} \begin{bmatrix} \theta_t \\ \dot{\theta}_t \end{bmatrix} + \begin{bmatrix} 0 \\ \Delta t \end{bmatrix} \frac{\tau}{ml^2}$$

$x_{t+1} \qquad A \qquad x_t \qquad B \qquad u_t$

# Quadratic Cost Function

Linear system (model)  
Quadratic cost function to  
minimize

$$x_{t+1} = Ax_t + Bu_t$$
$$\sum_t x_t^\top Q x_t + u_t^\top R u_t$$

$$x_t^\top Q x_t$$

$$(1 \times N)(N \times N)(N \times 1)$$

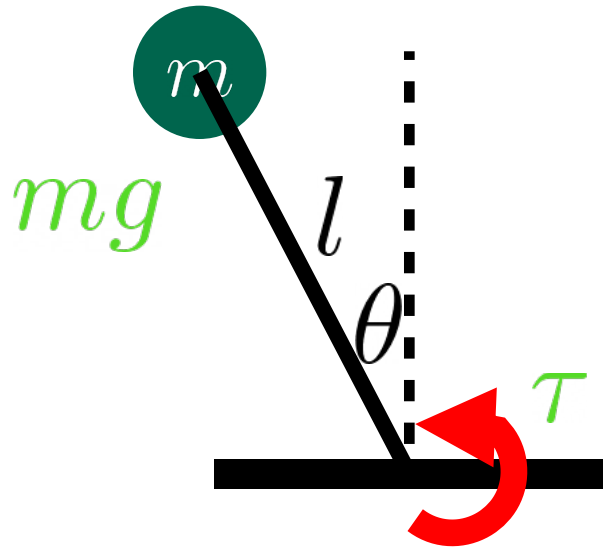
**STATE COST**

$$u_t^\top R u_t$$

$$(1 \times M)(M \times M)(M \times 1)$$

**CONTROL COST**

# Example: Inverted Pendulum (State Cost)



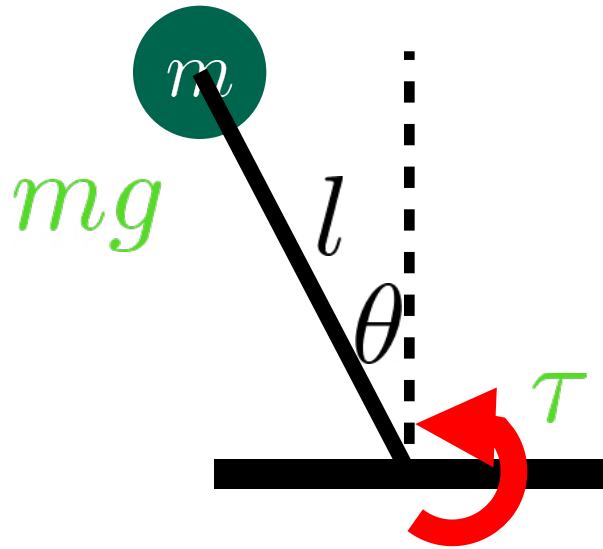
$$x_t^\top Q x_t \quad (\text{QUADRATIC FORM})$$

$$= \begin{bmatrix} \theta_t \\ \dot{\theta}_t \end{bmatrix}^\top \begin{bmatrix} Q_{\theta\theta} & Q_{\theta\dot{\theta}} \\ Q_{\dot{\theta}\theta} & Q_{\dot{\theta}\dot{\theta}} \end{bmatrix} \begin{bmatrix} \theta_t \\ \dot{\theta}_t \end{bmatrix}$$

$$= Q_{\theta\theta}\theta_t^2 + 2Q_{\theta\dot{\theta}}\theta_t\dot{\theta}_t + Q_{\dot{\theta}\dot{\theta}}\dot{\theta}_t^2$$

$$Q \succ 0 \Leftrightarrow z^\top Q z > 0, \forall z \neq 0$$

# Example: Inverted Pendulum (Control Cost)



$$u_t^\top R u_t \quad (\text{QUADRATIC FORM})$$

$$= \frac{\tau_t}{ml^2} [R_{\tau\tau}] \frac{\tau_t}{ml^2}$$

$$= R_{\tau\tau} \left( \frac{\tau_t}{ml^2} \right)^2$$

$$R \succ 0 \Leftrightarrow z^\top R z > 0, \forall z \neq 0$$

# Example: Inverted Pendulum

