

# Lecture 9: PID Control

Professor Katie Driggs-Campbell

March 2, 2021

ECE484: Principles of Safe Autonomy



# Administrivia

- Safety training information posted on discord
- MP1 due this week
  - Demo due Thursday
  - Report due Friday
- Milestone report due Friday 3/19 by 5pm
  - Rubric posted tonight



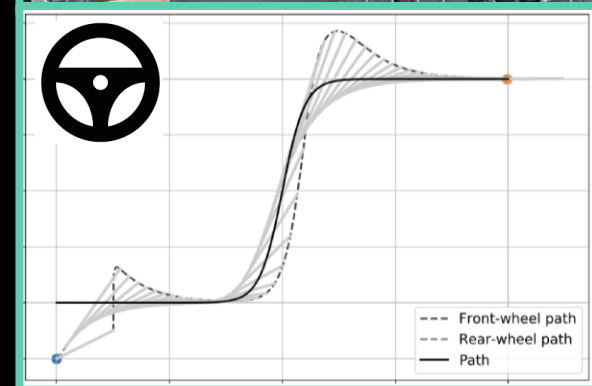
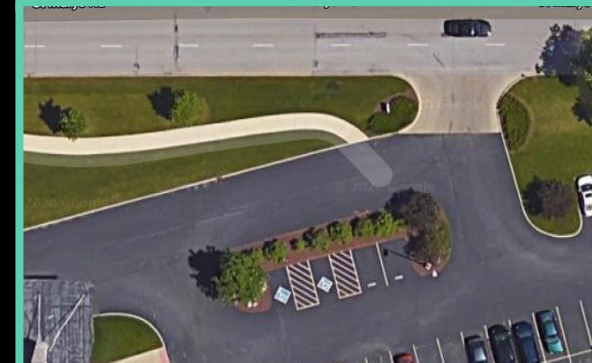
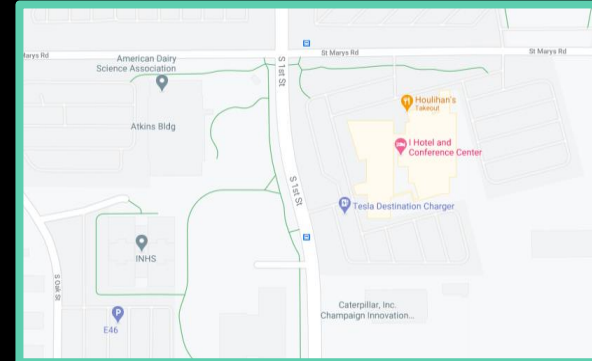
# Today's Plan

- Review some properties of dynamical systems and differential equations
- Take a look at PID controllers
- Build up to waypoint following using the models discussed last week!



# Typical planning and control modules

- Global navigation and planner
  - Find paths from source to destination with static obstacles
  - Algorithms: Graph search, Dijkstra, Sampling-based planning
  - Time scale: Minutes
  - Look ahead: Destination
  - Output: reference center line, semantic commands
- Local planner
  - Dynamically feasible trajectory generation
  - Dynamic planning w.r.t. obstacles
  - Time scales: 10 Hz
  - Look ahead: Seconds
  - Output: Waypoints, high-level actions, directions / velocities
- Controller
  - Waypoint follower using steering, throttle
  - Algorithms: PID control, MPC, Lyapunov-based controller
  - Lateral/longitudinal control
  - Time scale: 100 Hz
  - Look ahead: current state
  - Output: low-level control actions



# Dynamical Systems Model

Describe behavior in terms of instantaneous laws:

$$\frac{dx(t)}{dt} = \dot{x}(t) = f(x(t), u(t)) \quad // \quad x[t+1] = f(x[t], u[t])$$

where  $t \in \mathbb{R}$ ,  $x(t) \in \mathbb{R}^n$ ,  $u(t) \in \mathbb{R}^m$ , and  $f: \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^n$  gives the dynamics / transition function

$$\dot{x} = Ax$$



# Examples

$$\dot{x} = Ax + Bu$$

$$u = -Kx$$

$$\dot{x} = (A - BK)x$$

$$\begin{bmatrix} \dot{x}_2 \\ \dot{x}_1 \end{bmatrix} = \begin{bmatrix} -1/4 & -2/5 \\ 3 & -1/4 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

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$$\begin{bmatrix} \dot{x}_2 \\ \dot{x}_1 \end{bmatrix} = \begin{bmatrix} -1/4 & -2/5 \\ 3 & -1/2 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix}$$

$$\lambda_1 = -0.25 - i1.10$$

$$\lambda_2 = -0.25 + i1.10$$

$$\lambda_1 = +i0.1066$$

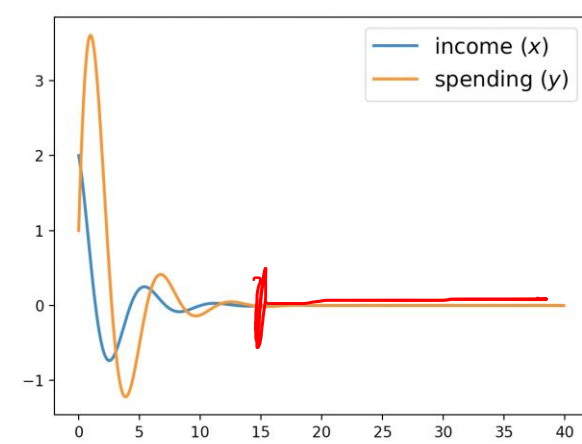
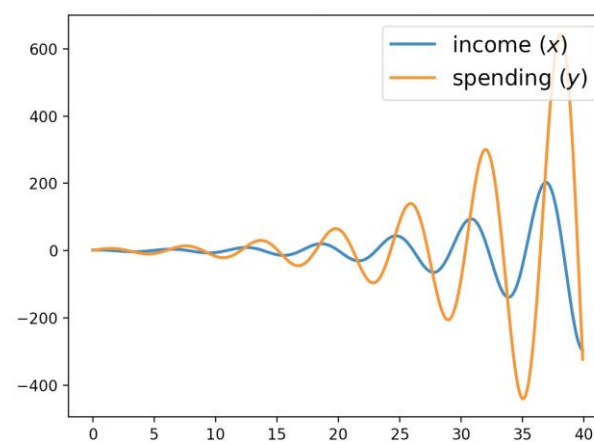
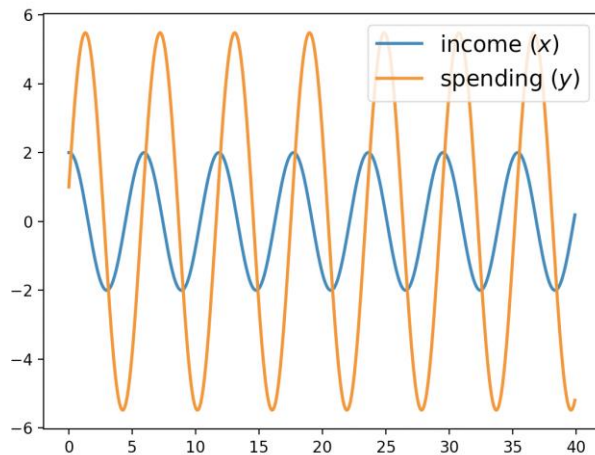
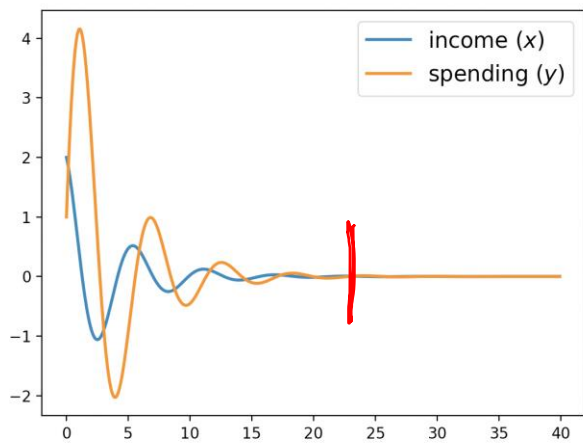
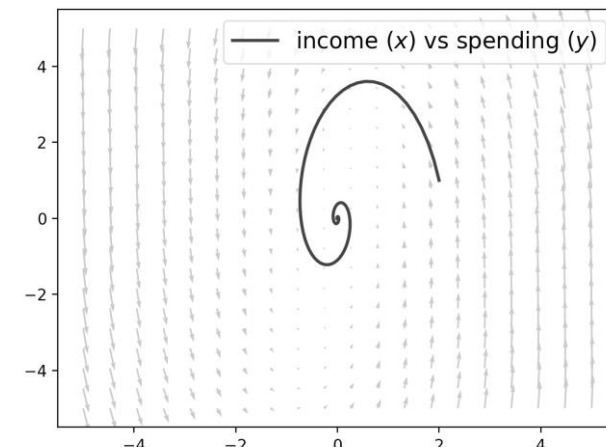
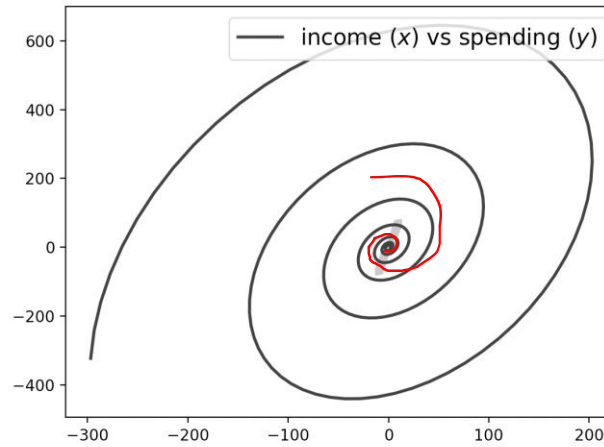
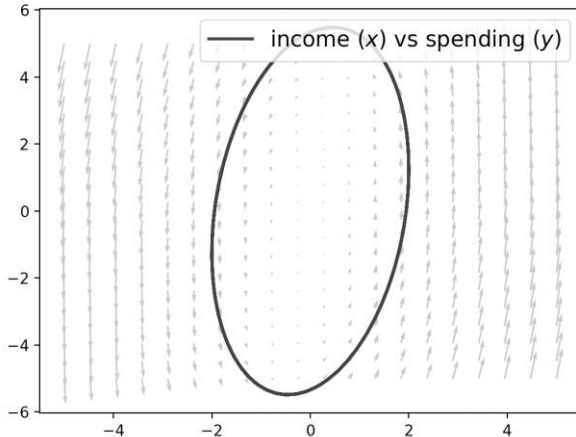
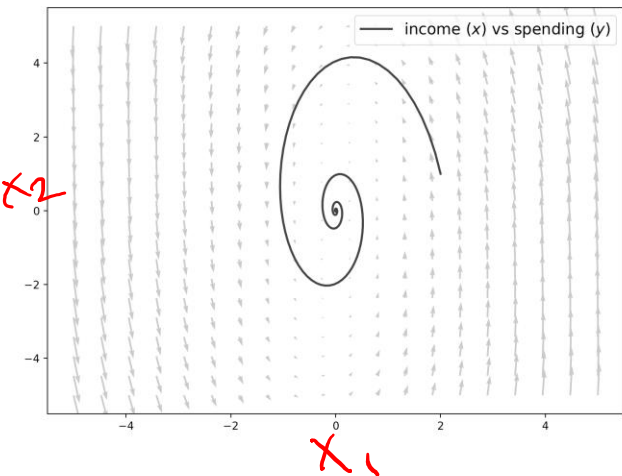
$$\lambda_2 = -i0.1066$$

$$\lambda_1 = 0.125 + i1.029$$

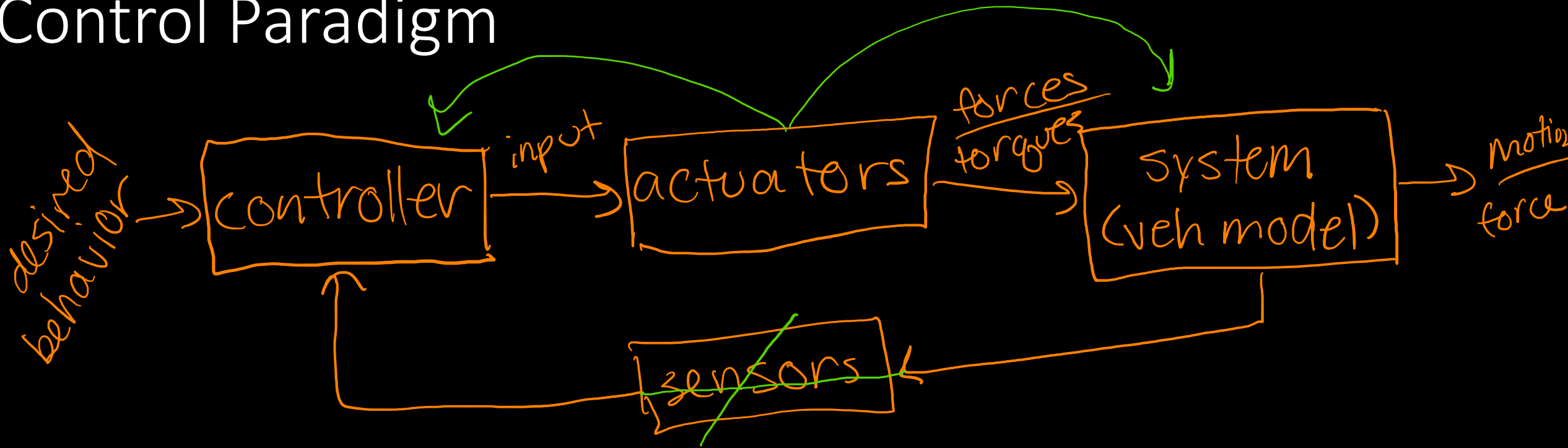
$$\lambda_2 = -0.125 - i1.029$$

$$\lambda_1 = -0.375 - i1.088$$

$$\lambda_2 = -0.375 + i1.088$$



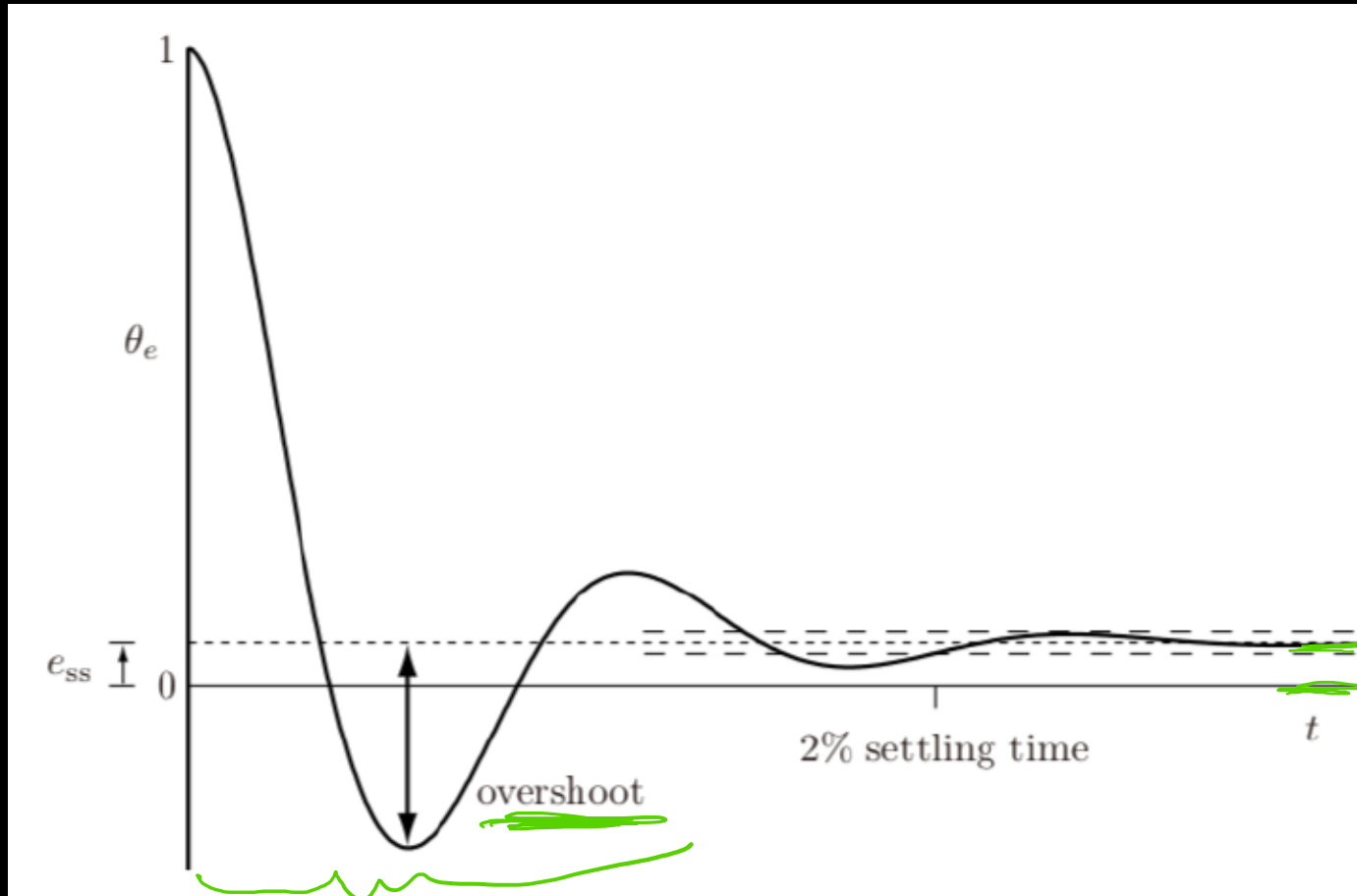
# Control Paradigm



→ consider error dynamics



# Error Dynamics





# Feedback Control



## PID controller

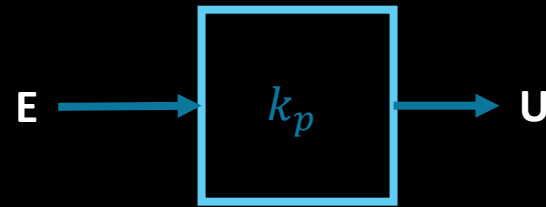
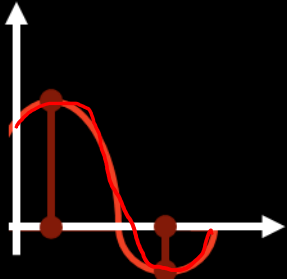
$$u = \underbrace{k_p e}_{\text{reduce pos error}} + \underbrace{k_d \dot{e}}_{\text{dampen vel error}} + \underbrace{k_i \int e(\tau) d\tau}_{\text{remove ss or accumulated errors}}$$



# PID Controllers

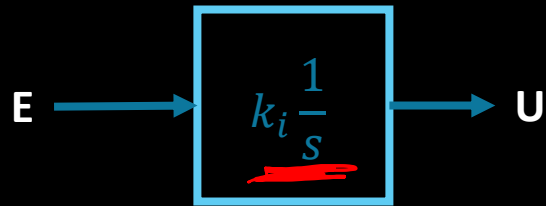
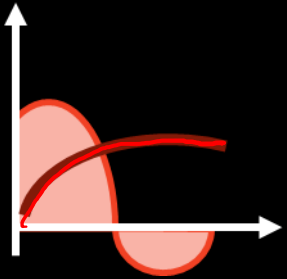
- Proportional

$$u = k_p e$$



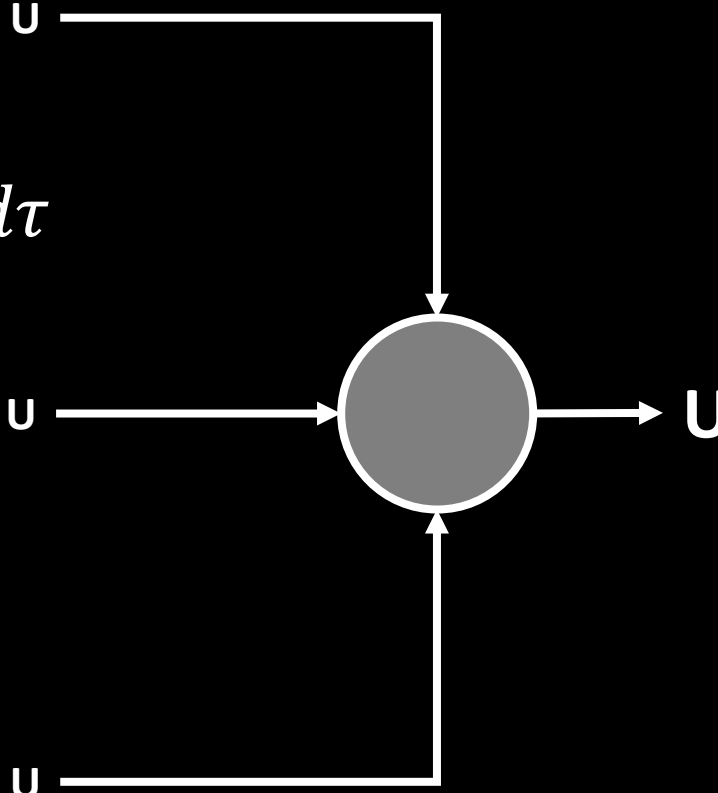
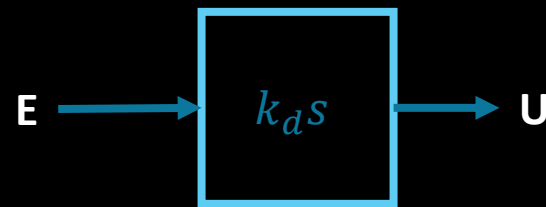
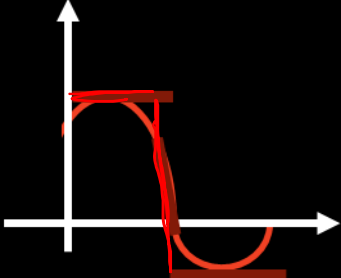
- Integral

$$u = k_i \int e(\tau) d\tau$$



- Derivative

$$u = k_d \dot{e}$$



# Linear Error Dynamics and Stability

$$a_p \underline{e^{(p)}} + a_{p-1} \underline{e^{(p-1)}} + \dots + a_1 \underline{\dot{e}} + a_0 \underline{e} = 0$$

$$\underline{x_1 = e}, \underline{x_2 = \dot{e}}, x_3 = \ddot{e}, \dots$$

$$x_p = \frac{-a_0}{a_p} x_1 - \frac{a_1}{a_p} x_2 - \dots$$

$$\dot{x} = Ax, \quad x = [x_1 \dots x_p]^T$$

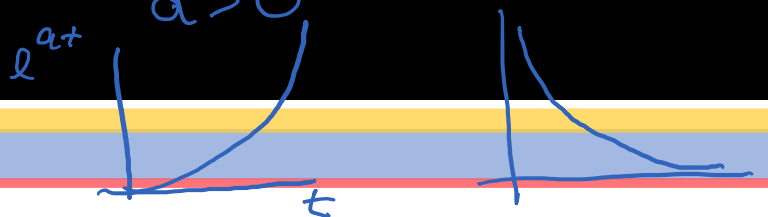
$$\begin{bmatrix} \dot{x}_1 \\ \vdots \\ \dot{x}_p \end{bmatrix} = \begin{bmatrix} 0 & 1 & 0 & \dots & 0 \\ 0 & 0 & 1 & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ -\frac{a_0}{a_p} & \dots & \dots & \dots & -\frac{a_{p-1}}{a_p} \end{bmatrix} \begin{bmatrix} x_1 \\ \vdots \\ x_p \end{bmatrix}$$

recall:  $\dot{x} = Ax \rightsquigarrow x(t) = \underline{\underline{e^{At}}} x_0$

sys is stable if the real parts of the eigenvalues are neg.  
 → unstable if  $\exists x_0$

s.t.  $\lim_{t \rightarrow \infty} \|x(t)\| = \infty$

$\int e^{at} dt$  where  $d = a + bi \in \mathbb{C}$   
 $\int e^{at} (\cos bt + i \sin bt)$   
 $a > 0$   
 $a < 0$



# Viewing as a Second Order System

- The second order system is:  $\ddot{e} + \underline{c_1}\dot{e} + \underline{c_2}e = 0$
- In standard form, we write:

$$\ddot{e}(t) + \underline{2\xi\omega_n}\dot{e}(t) + \underline{\omega_n^2}e(t) = 0$$

where  $\xi$  is the *damping ratio* and  $\omega_n$  is the *natural frequency*

- The eigenvalues are given as:

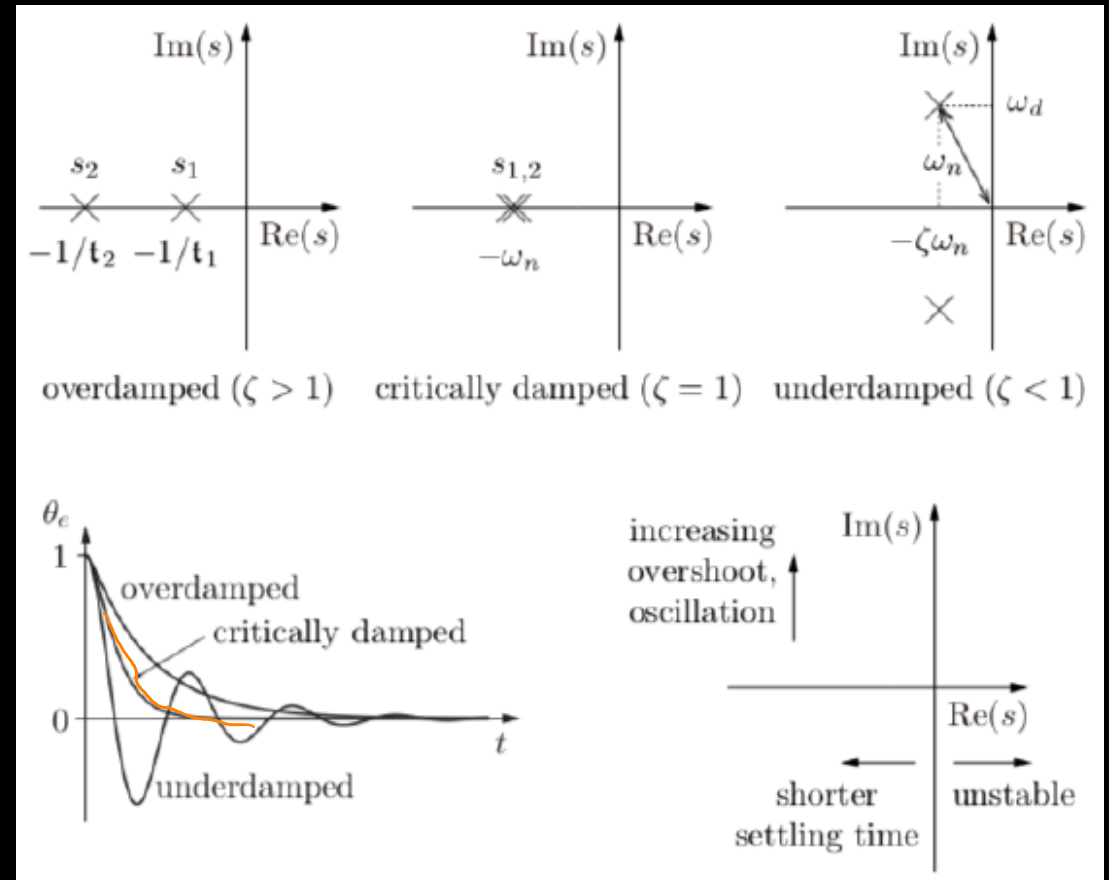
$$\lambda_{1,2} = -\xi\omega_n \pm \omega_n\sqrt{\xi^2 - 1}$$

- Note that the system is stable iff  $\omega_n$  and  $\xi$  are positive

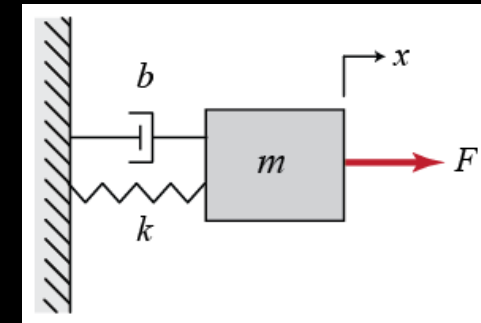


# Second Order Dynamics: Cases

- Overdamped:  $\zeta > 1$ 
  - Roots  $s_1$  and  $s_2$  are distinct
  - $\theta_e(t) = c_1 e^{s_1 t} + c_2 e^{s_2 t}$
  - Time constant is the less negative root
- Critically damped:  $\zeta = 1$ 
  - Roots  $s_1$  and  $s_2$  are equal and real
  - $\theta_e(t) = (c_1 + c_2 t) e^{-\omega_n t}$
  - Time constant is given by  $1/\omega_n$
- Underdamped:  $\zeta < 1$ 
  - Roots are complex conjugates:
 
$$s_{1,2} = -\zeta\omega_n \pm j\omega_n\sqrt{1-\zeta^2}$$
  - $\theta_e(t) = (c_1 \cos \omega_d t + c_2 \sin \omega_d t) e^{-\zeta\omega_n t}$



# Simple Damped Spring System



$$m \ddot{x} + b \dot{x} + kx = F$$

$$\ddot{x} + \frac{b}{m} \dot{x} + \frac{k}{m} x = u$$

$$\ddot{x} + 2\xi\omega_0 \dot{x} + \omega_0^2 x = u$$

$\xi$  damping ratio

$\omega_0$  natural frequency

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

$$\mathcal{L}\{m\ddot{x} + b\dot{x} + kx\} =$$

$$ms^2X(s) + bsX(s) + kX(s)$$

Transfer Function:

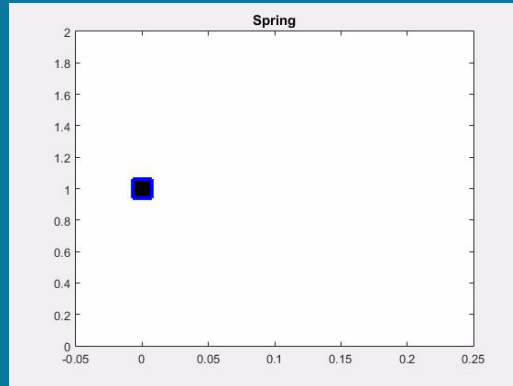
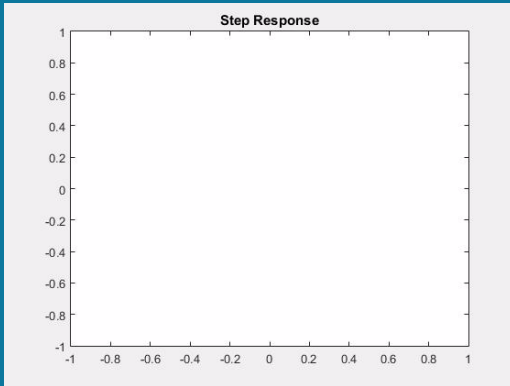
$$\frac{X(s)}{U(s)} = \frac{1}{ms^2 + bs + k}$$

Poles:

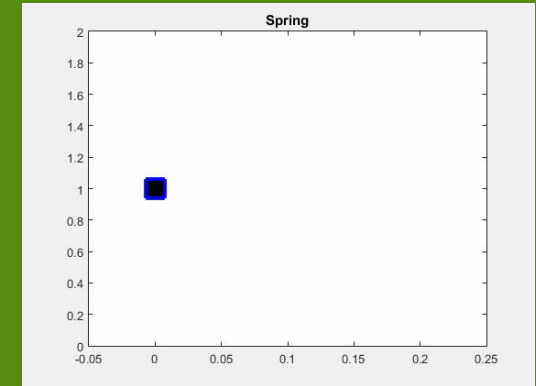
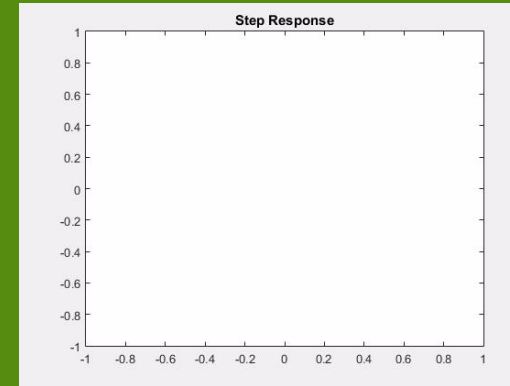
$$s = \frac{-b \pm \sqrt{b^2 - 4mk}}{2m}$$



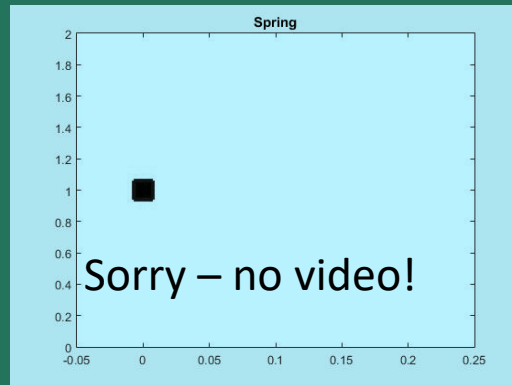
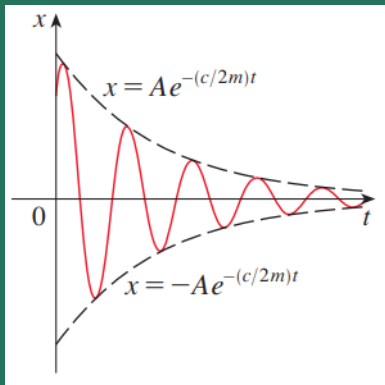
## Undamped Case: $b = 0$



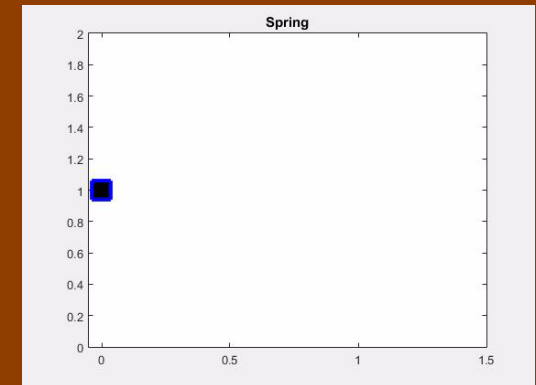
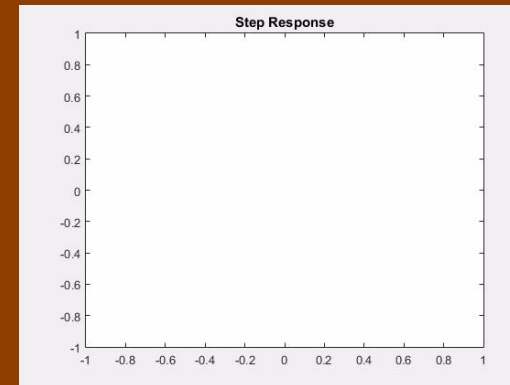
## Overdamped Case: $b^2 - 4mk > 0$



## Underdamped Case: $b^2 - 4mk < 0$



## With Feedback Control



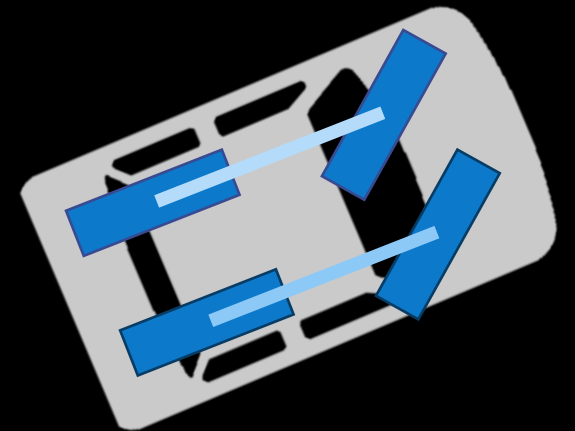
Break!





# Simple vehicle model: Dubin's car

- Key assumptions
  - Front and rear wheel in the plane in a stationary coordinate system
  - Steering input, front wheel steering angle  $\delta$
  - No slip: wheels move only in the direction of the plane they reside in
- Zeroing out the accelerations perpendicular to the plane in which the wheels reside, we can derive simple equations



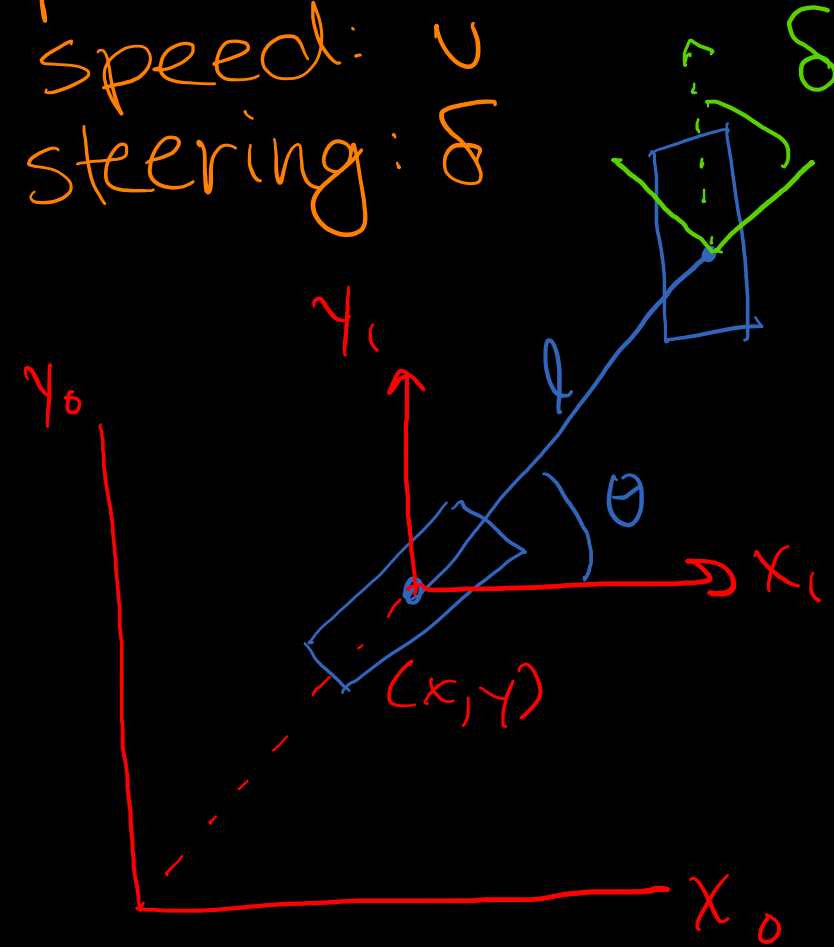
**Reference:** Paden, Brian, Michal Cap, Sze Zheng Yong, Dmitry S. Yershov, and Emilio Frazzoli. 2016. A survey of motion planning and control techniques for self-driving urban vehicles. IEEE Transactions on Intelligent Vehicles 1 (1): 33–55.



# Dubin's Model

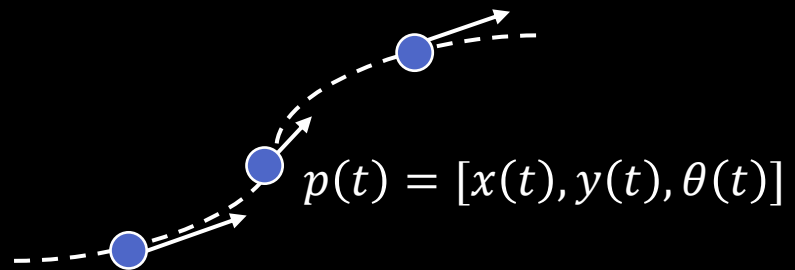
$$\begin{aligned}\dot{x} &= v \cos \theta \\ \dot{y} &= v \sin \theta \\ \dot{\theta} &= \frac{v}{l} \tan \delta\end{aligned}$$

length  $l$   
pose:  $[x, y, \theta]^T$   
speed:  $v$   
steering:  $\delta$



# Path following control

- The path followed by a robot can be represented by a *trajectory or path* parameterized by time
  - from a higher-level planner
- Defines the desired instantaneous pose  $p(t)$



# Open-loop waypoint following

- We can write an **open-loop controller** for a robot that is naturally controlled via angular velocity, such as a differential-drive robot:

$$u_{\omega,OL}(t) = \begin{bmatrix} v(t) \\ \omega(t) \end{bmatrix} = \begin{bmatrix} \sqrt{\dot{x}(t)^2 + \dot{y}(t)^2} \\ \dot{\theta}(t) \end{bmatrix}$$

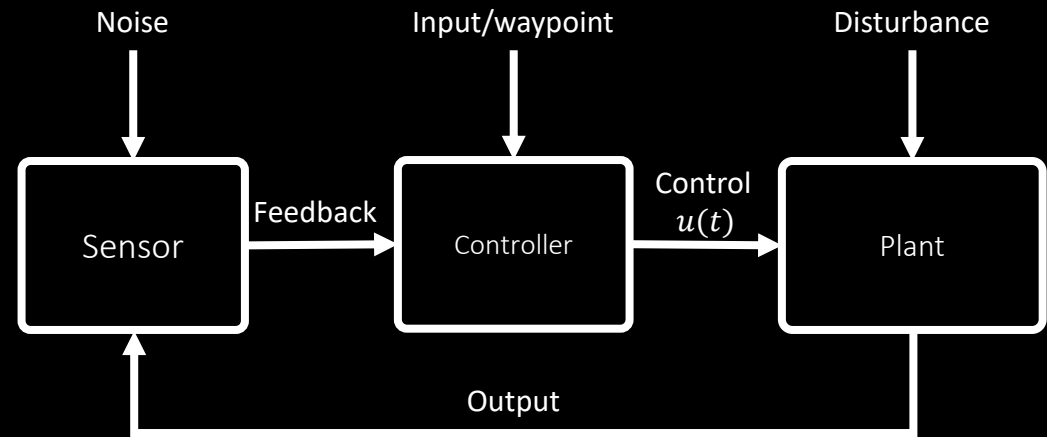
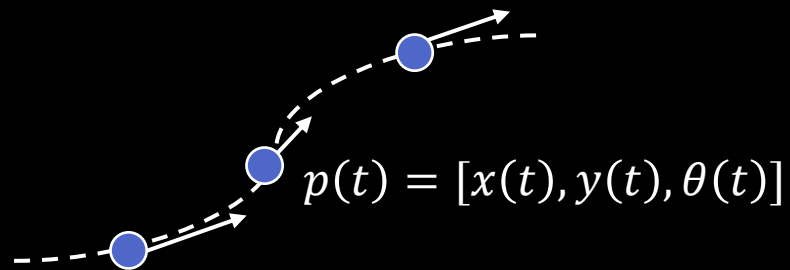
- We can write an **open-loop controller** for a robot with car-like steering:

$$u_{\kappa,OL}(t) = \begin{bmatrix} v(t) \\ \kappa(t) \end{bmatrix} = \begin{bmatrix} \sqrt{\dot{x}(t)^2 + \dot{y}(t)^2} \\ \dot{\theta}(t) \\ \sqrt{\dot{x}(t)^2 + \dot{y}(t)^2} \end{bmatrix}$$



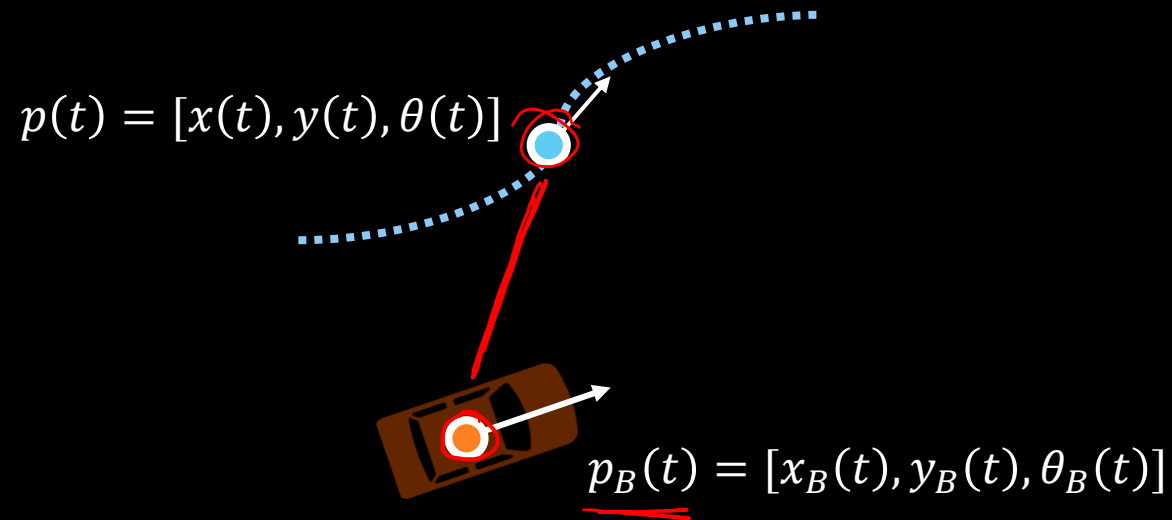
# Path following control

- The path followed by a robot can be represented by a *trajectory or path* parameterized by time
  - from a higher-level planner
- Defines the desired instantaneous pose  $p(t)$



# Path following control

- Desired instantaneous pose  $p(t)$
- How to define error between actual pose  $p_B(t)$  and desired pose  $p(t)$  in the form of  $y_d(t) - y(t)$ ?



# Path following control

The error vector measured vehicle coordinates

$$e(t) = [\delta_s(t), \delta_n(t), \delta_\theta(t), \delta_v(t)]$$

$[\delta_s, \delta_n]$  define the coordinate errors in the vehicle's reference frame:  
along track error and cross track error

- **Along track error:** distance ahead or behind the target in the instantaneous direction of motion.

$$\delta_s = \cos(\theta_B(t)) (x(t) - x_B(t)) + \sin(\theta_B(t)) (y(t) - y_B(t))$$

- **Cross track error:** portion of the position error orthogonal to the intended direction of motion

$$\delta_n = -\sin(\theta_B(t)) (x(t) - x_B(t)) + \cos(\theta_B(t)) (y(t) - y_B(t))$$

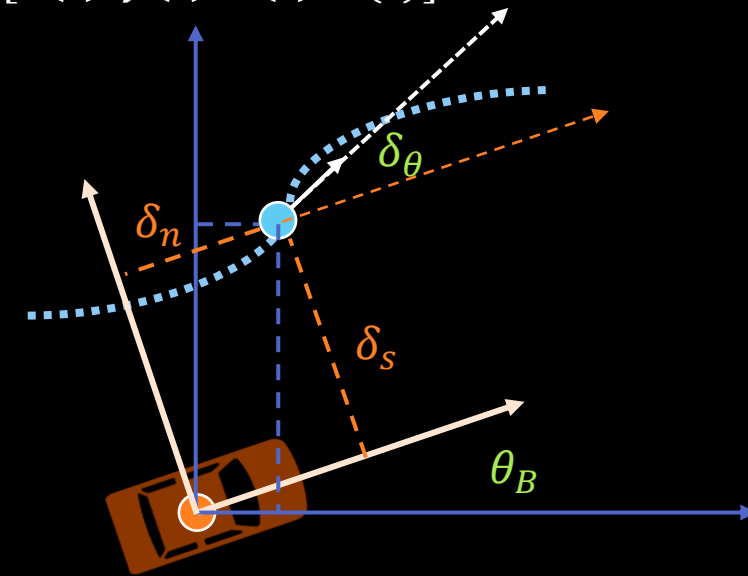
- **Heading error:** difference between desired and actual orientation and direction

$$\delta_\theta = \theta(t) - \theta_B(t)$$

$$\delta_v = v(t) - v_B(t)$$

→ Each of these errors match the form  $y_d(t) - y(t)$

$$p(t) = [x(t), y(t), \theta(t), v(t)]$$



$$p_B(t) = [x_B(t), y_B(t), \theta_B(t), v_B(t)]$$



# A simple P-controller example

- Given a simple system:  $\dot{y}(t) = u(t) + d(t)$
- Using proportional (P) controller:

$$u(t) = -K_P e(t) = -K_P (y(t) - y_d(t))$$

$$\dot{y}(t) = -K_P y(t) + K_P y_d(t) + d(t)$$

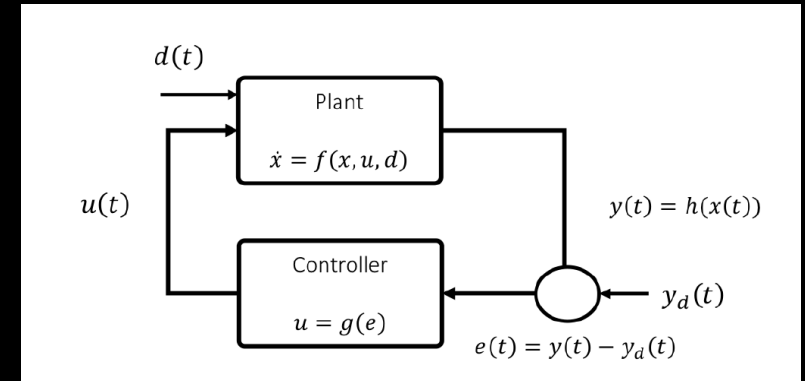
- Consider constant setpoint  $y_0$  and disturbance  $d_{ss}$

$$\dot{y}(t) = -K_P y(t) + K_P y_0 + d_{ss}$$

- What is the steady state output?

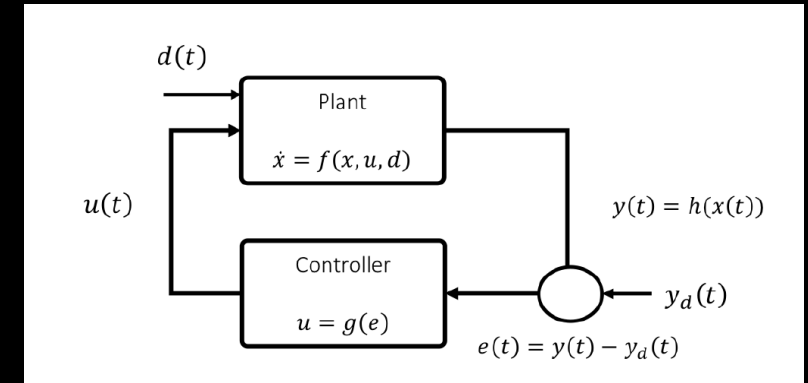
- Set:  $-K_P y(t) + K_P y_0 + d_{ss} = 0$

- Solve for  $y_{ss}$ :  $y(t) = \frac{d_{ss}}{K_P} + y_0$





# A simple P-controller example



- Given a simple system:  $\dot{y}(t) = u(t) + d(t)$
- Consider constant setpoint  $y_0$  and disturbance  $d_{ss}$

$$\dot{y}(t) = \underline{-K_P y(t)} + \underline{K_P y_0} + \underline{d_{ss}} \leftarrow$$

- Steady state output  $y_{ss} = \frac{d_{ss}}{K_P} + y_0$

- Transient behavior:

$$y(t) = \underline{y_0 e^{-t/T}} + \underline{y_{ss} (1 - e^{-t/T})}, T = \underline{1/K_P}$$

- To make steady state error small, we can increase  $\underline{K_P}$  at the expense of longer transients

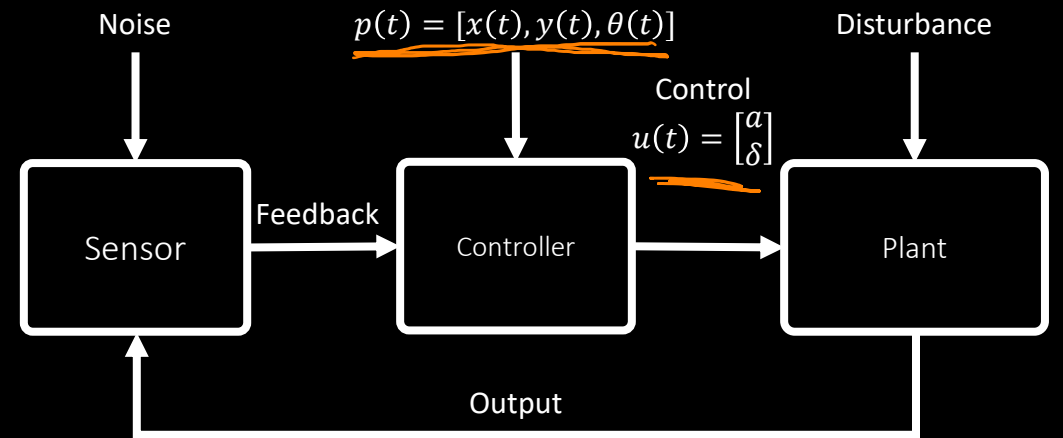


# Control Law

Control input is given by  $u = [a, \delta]^T$

where  $a$  is the acceleration and  $\delta$  is the steering angle

$$u = K \begin{bmatrix} \delta_s \\ \delta_n \\ \delta_\theta \\ \delta_v \end{bmatrix}$$
$$K = \begin{bmatrix} K_s & 0 & 0 & K_v \\ 0 & K_n & K_\theta & 0 \end{bmatrix}$$

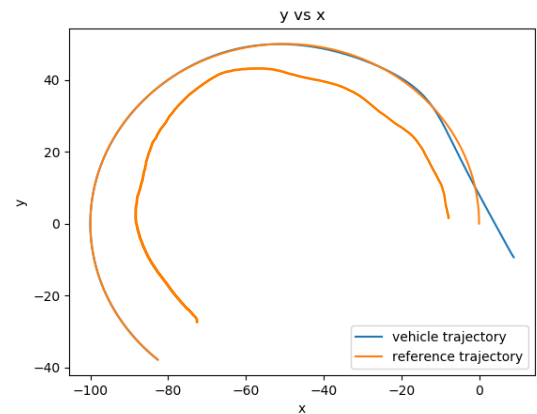
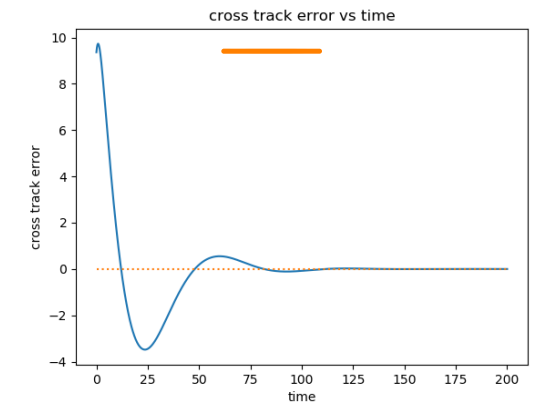
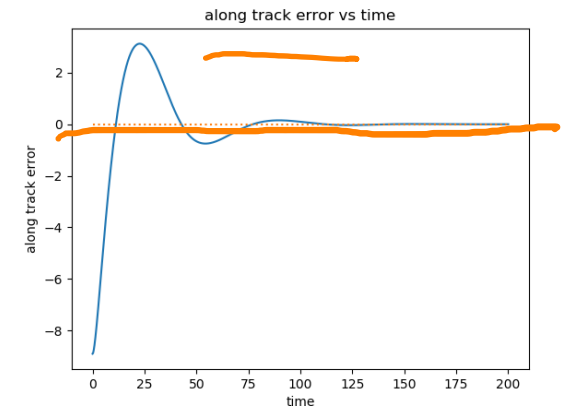


# Control Law

$$K = \begin{bmatrix} \underline{K_s} & 0 & 0 & \underline{K_v} \\ 0 & \underline{K_n} & \underline{K_\theta} & 0 \end{bmatrix}$$

The **pure-pursuit controller** produced by this gain matrix performs a PD-control. It uses a PD-controller to correct **along-track error**.

The control on curvature is also a PD-controller for **cross-track error** because  $\delta_\theta$  is related to the derivative of  $\delta_n$ .



# Summary

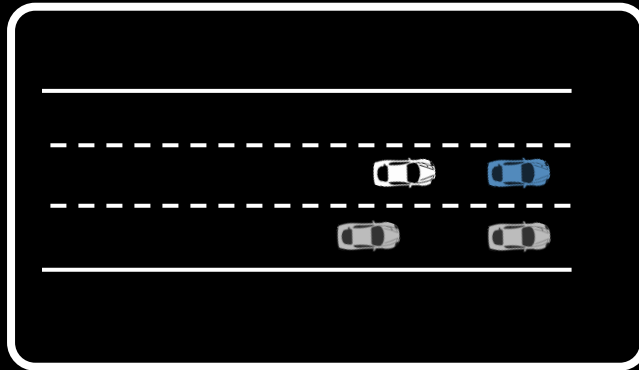
- Reviewed linear systems and stability of differential equations
- Looked at PID controllers as a way to regulate systems using state feedback
- Derived a waypoint following error dynamics
  - This will be MP2! Tutorial on 3/11.
- *Next time: Advanced Control Topics!*



# Extra Slides

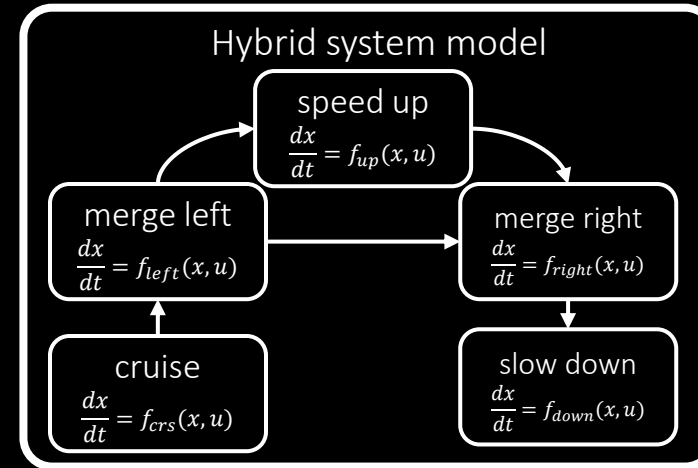
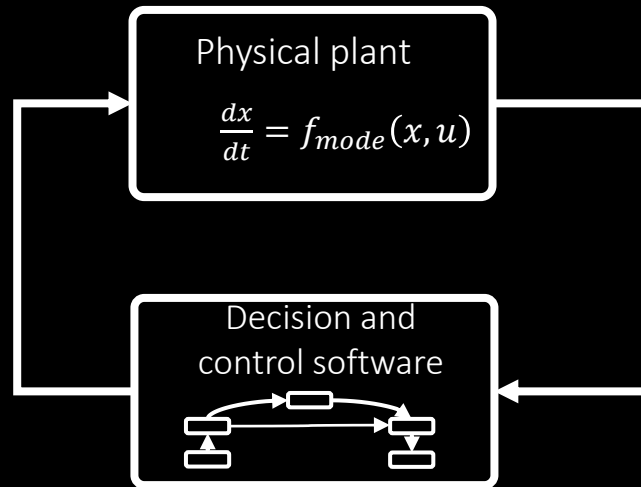


# Typical system models



Nonlinear *hybrid* dynamics

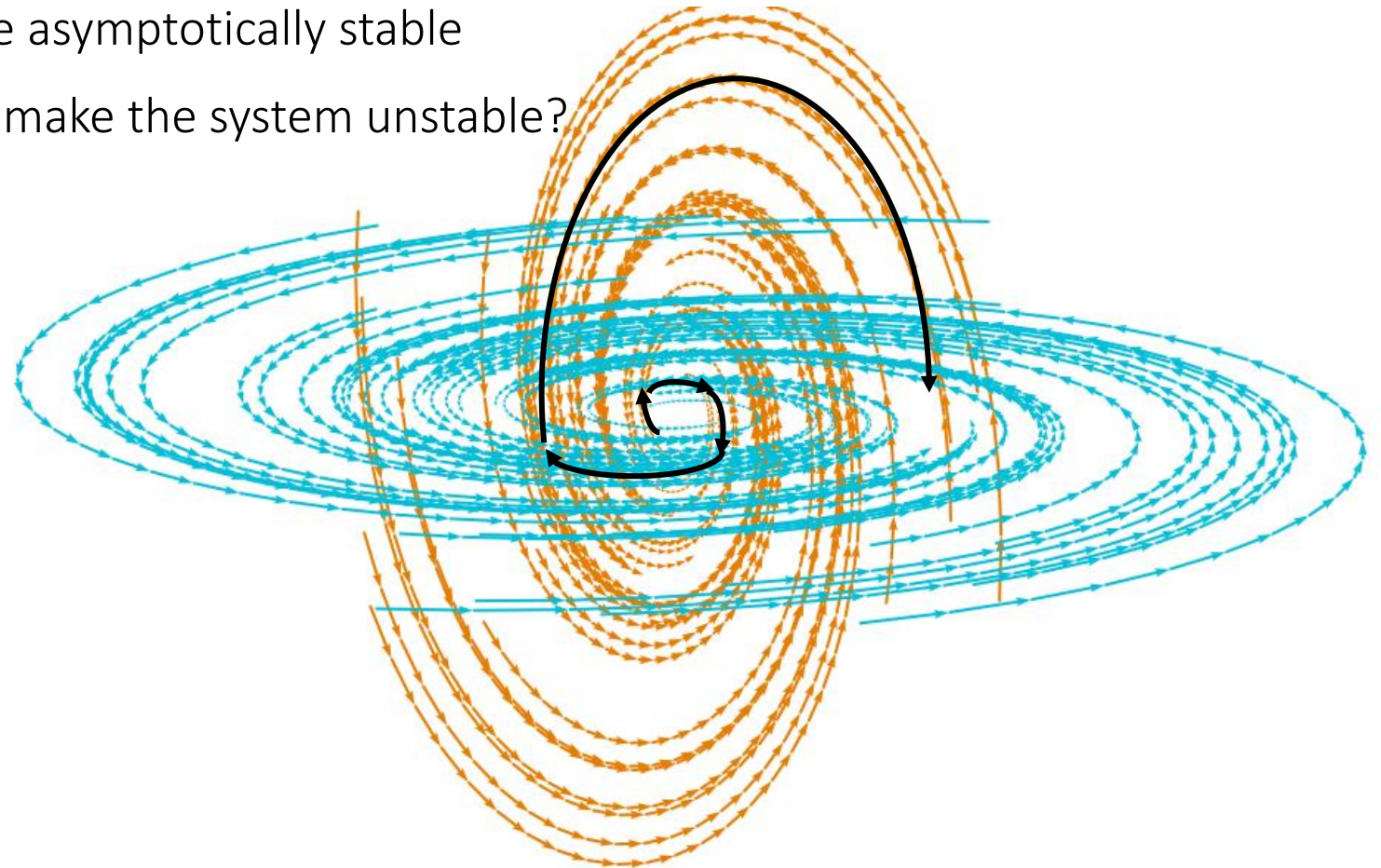
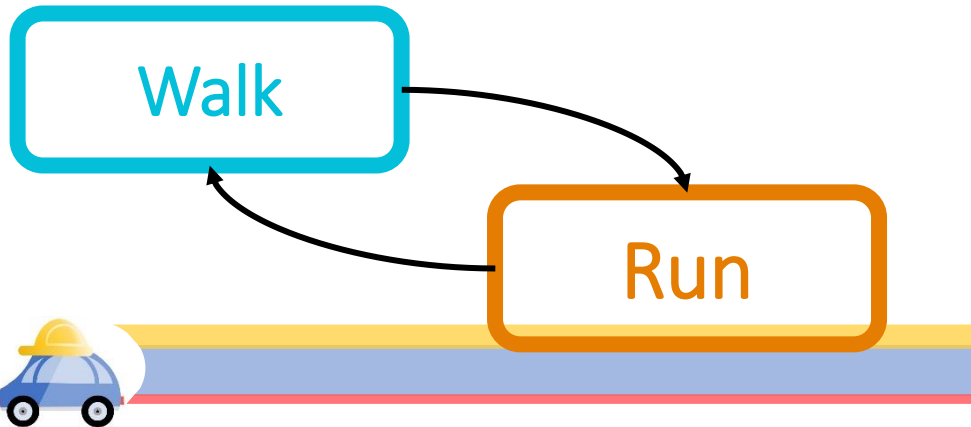
Interaction between computation and physics can lead to unexpected behaviors



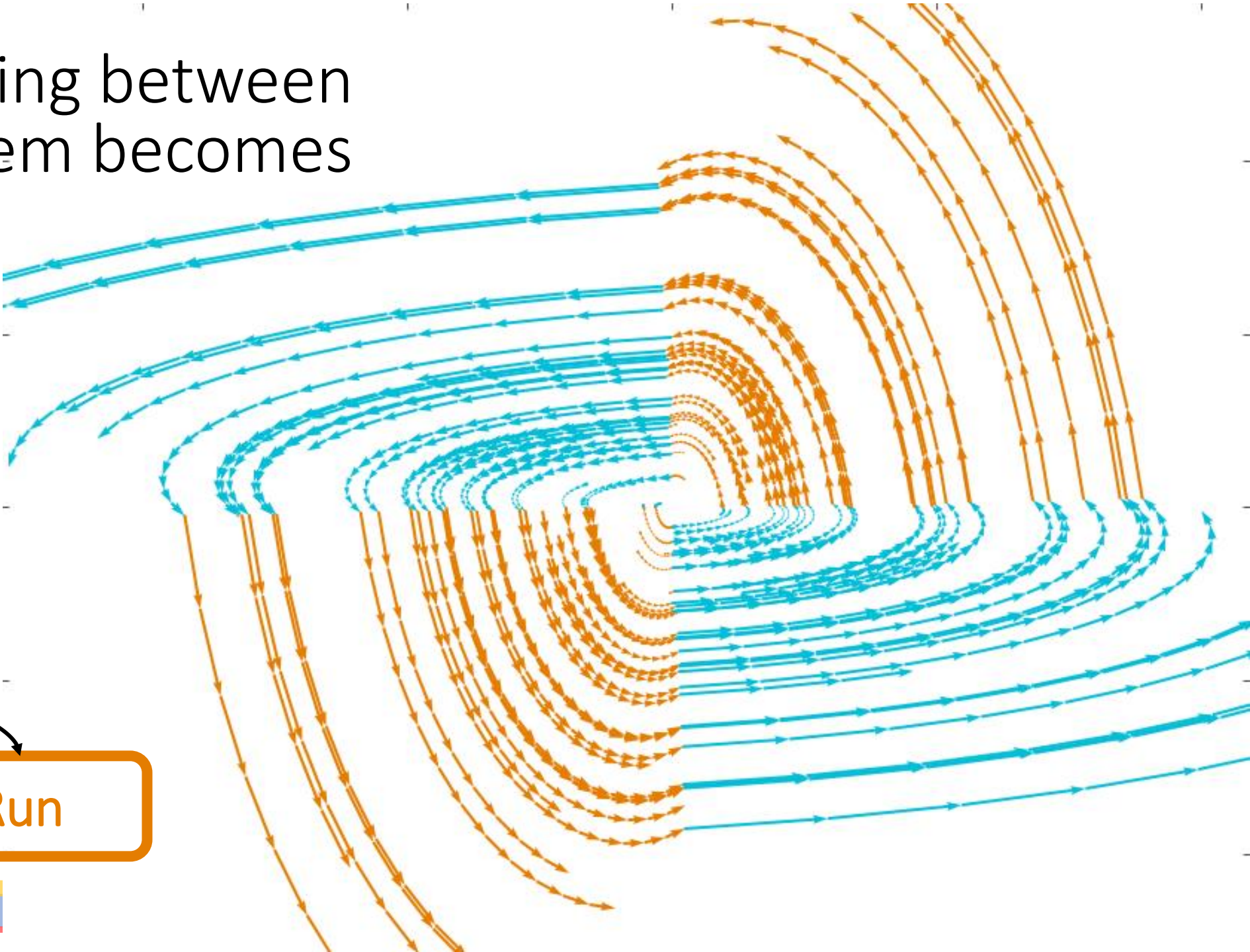
# Hybrid Instability: Switching between two stable linear models

Each of the modes of a walking robot are asymptotically stable

Is it possible to switch between them to make the system unstable?



Yes! By switching between them the system becomes unstable



Walk

Run

