

# Lecture 8: Control I

Professor Katie Driggs-Campbell

February 8, 2024

ECE484: Principles of Safe Autonomy



# Administrivia

- Field trips next week!
  - 2/13 Field trip 1 to high bay to see the GEM
  - 2/15 Field trip 2 to see F1 tenth cars
  - 2/20 Simulation project walkthrough
- No office hours next week (Tues. 2/13)
- MP1 released Friday 2/9
- Upcoming due dates:
  - HW0 and MP0 due Friday 2/9
    - “Demos” in lab sections
  - HW1 and MP1 due Friday 2/23



# Field Trip to High Bay

Address: 201 St Mary's Rd (near I-Hotel)

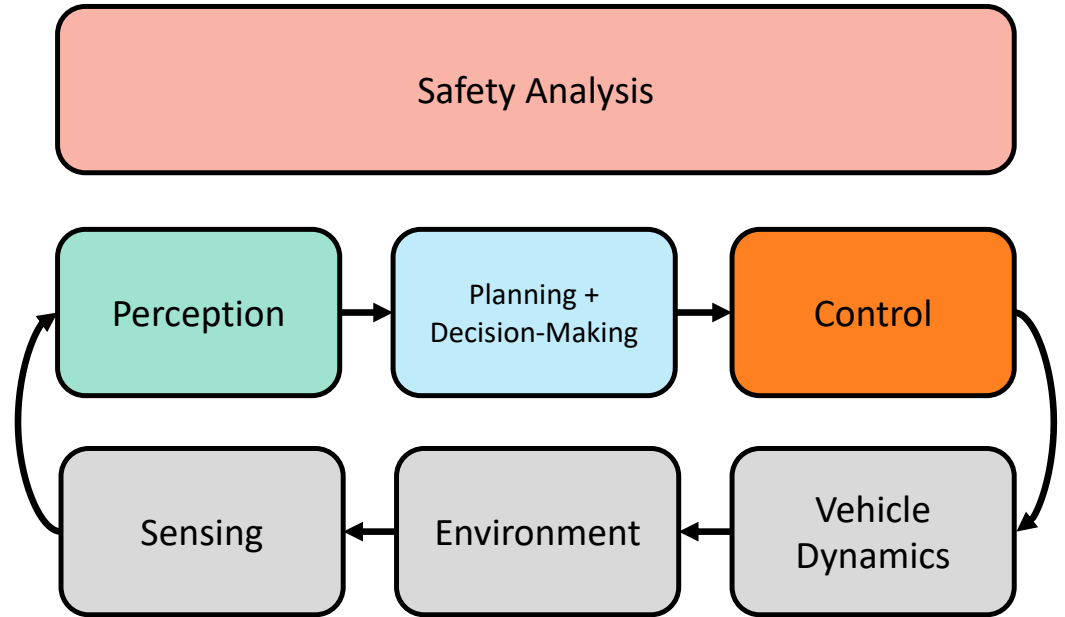
Two sessions:

- 9:30am lab sections AB1 & AB2
- 10:10am lab sections AB4 & AB5
- AB3 can attend either!

Indoor and outdoor tour – please dress appropriately!



# Autonomous GEM Vehicle



# Today's Plan

- What's a model?
- Planning and Control Motivation
  - Open-loop control
- Vehicle Models
  - How to design your model
  - Dubin's Car
  - Advanced Models: bicycle, tire dynamics



# 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



# What is control?

- A means of regulating or limiting something
- Algorithms (or process) for manipulating a system to achieve to desired value



Image Credit: Justas Galaburda and Vincent Mokuenko

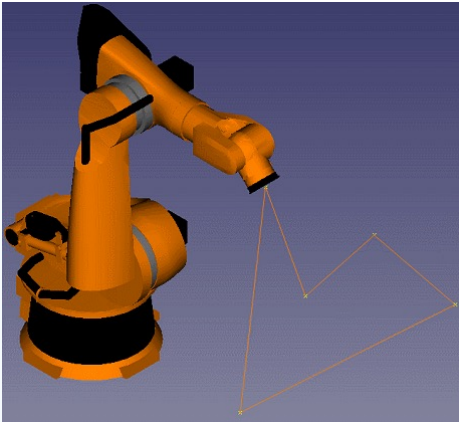
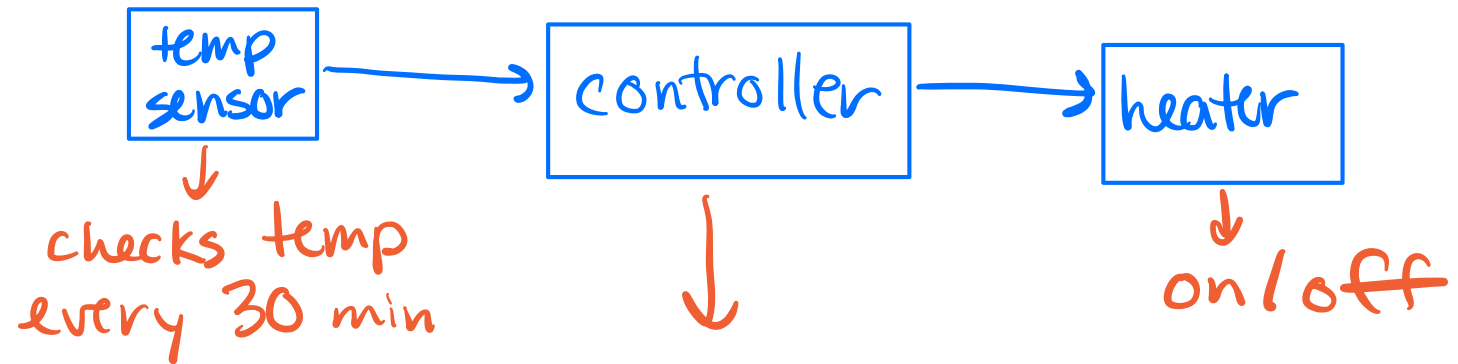


Image Credit: FreeCAD





# Open-Loop Example



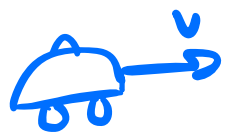
if temp  $\leq 70$  then  
heat on for T min  
if temp  $> 75$  then  
heat off (for T min)







# Closed-Loop Example



goal: maintain  $v_d$

Basic intuition:

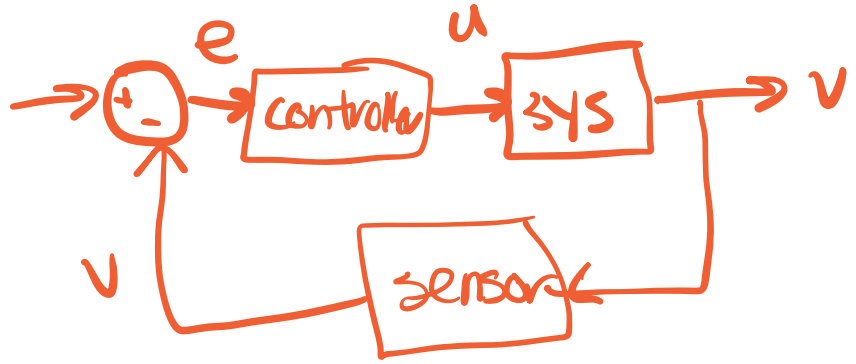
if  $v > v_d$ , want input  $\downarrow$

if  $v < v_d$ , want input  $\uparrow$

if  $v \approx v_d$ , want small input

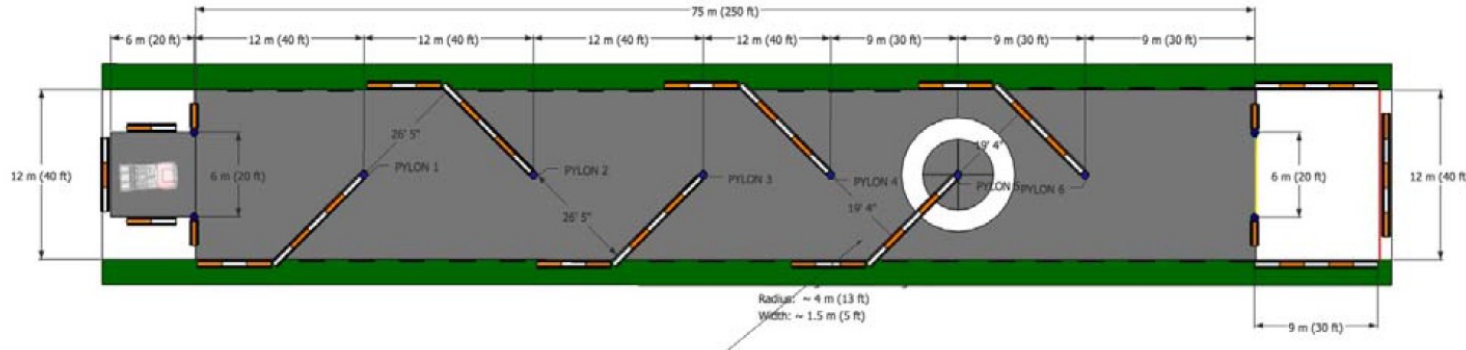
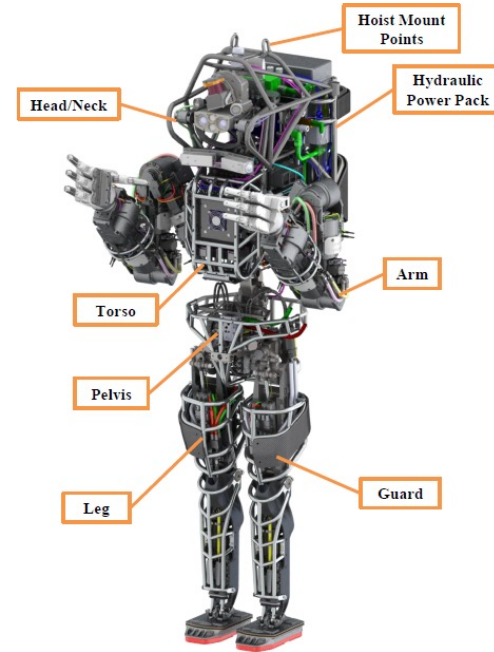
→ look at errors!

$$e = v_d - v$$



# Complex control tasks: DARPA Robotics Challenge

- Robot drives the vehicle through the course
- Robot gets out of the vehicle and travels dismounted out of the end zone





# First: Come up with a model!

For some common AV tasks, what are the desired behaviors, requirements of the system, actions/ inputs?

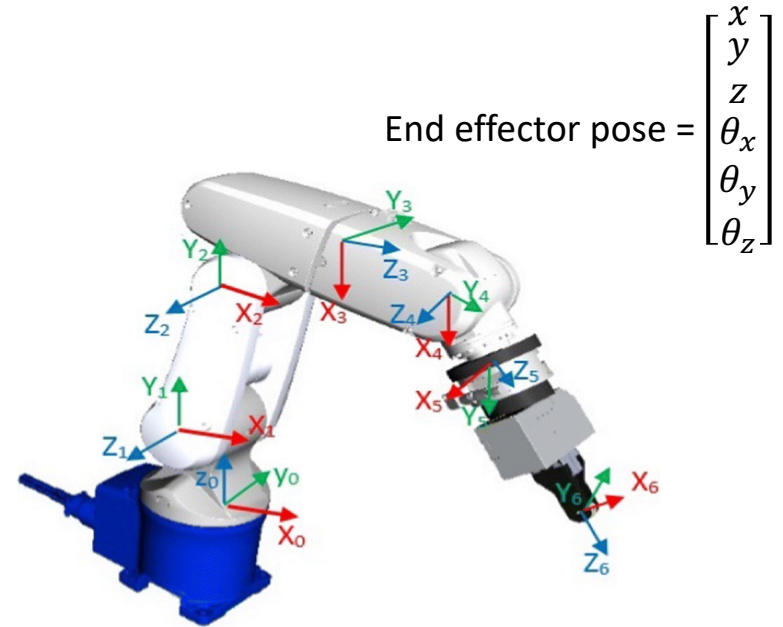
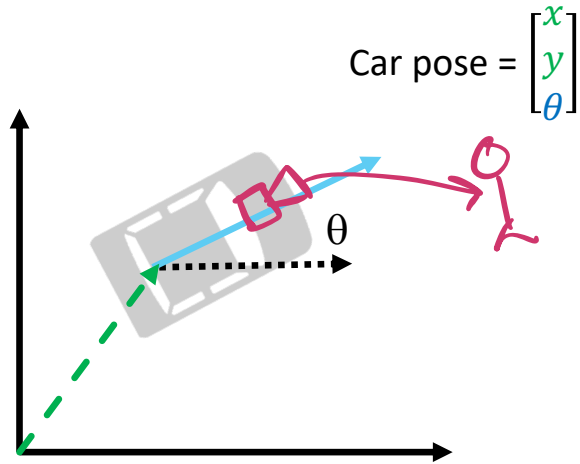
MPO has a very simple model. How can it be improved?

task: staying in lane  
actions: steering angle  
assumption: const vel

task: parking  
actions: steering, acc



# Coordinate Systems and Configurations



# Dynamical Systems Model

Describe behavior in terms of instantaneous laws:

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

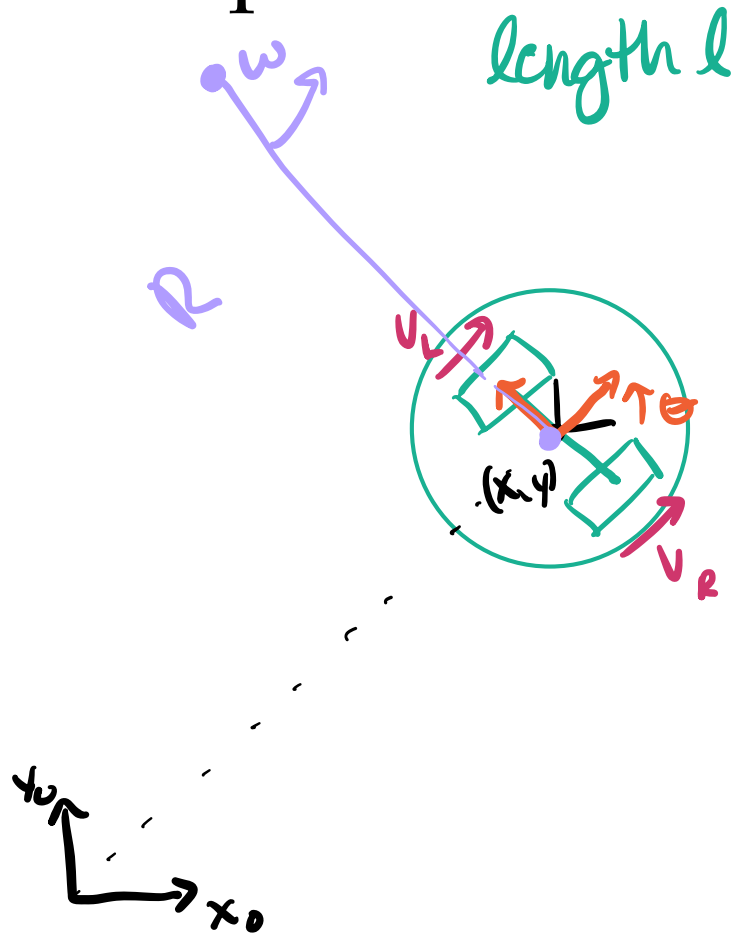
where  $t \in \mathbb{R}$ ,  $x(t) \in \mathbb{R}^n$ ,  $u(t) \in \mathbb{R}^m$ , and  $f: \mathbb{R}^n \times \mathbb{R}^m \rightarrow \mathbb{R}^n$  gives the dynamics / transition function

$$x_1 = \Theta, \quad x_2 = \dot{\Theta} = \dot{x}_1$$

$$\dot{x} = \begin{bmatrix} \dot{x}_1 \\ \dot{x}_2 \end{bmatrix} = \begin{bmatrix} x_2 \\ \underline{\quad} \end{bmatrix}$$



# Example: Differential Drive Robot



Instantaneous Center of Curvature

$$\begin{bmatrix} I_{CC_x} \\ I_{CC_y} \end{bmatrix} = \begin{bmatrix} x - R \sin \theta \\ y - R \cos \theta \end{bmatrix}$$

$$v_R = \omega (R + l/2)$$

$$v_L = \omega (R - l/2)$$

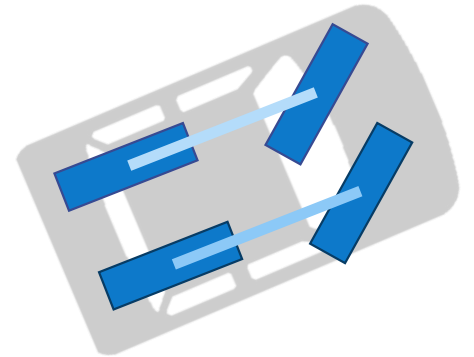
$$R = l/2 \cdot \frac{(v_R + v_L)}{(v_R - v_L)}$$

$$\omega = \frac{v_R - v_L}{l}$$



# 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.



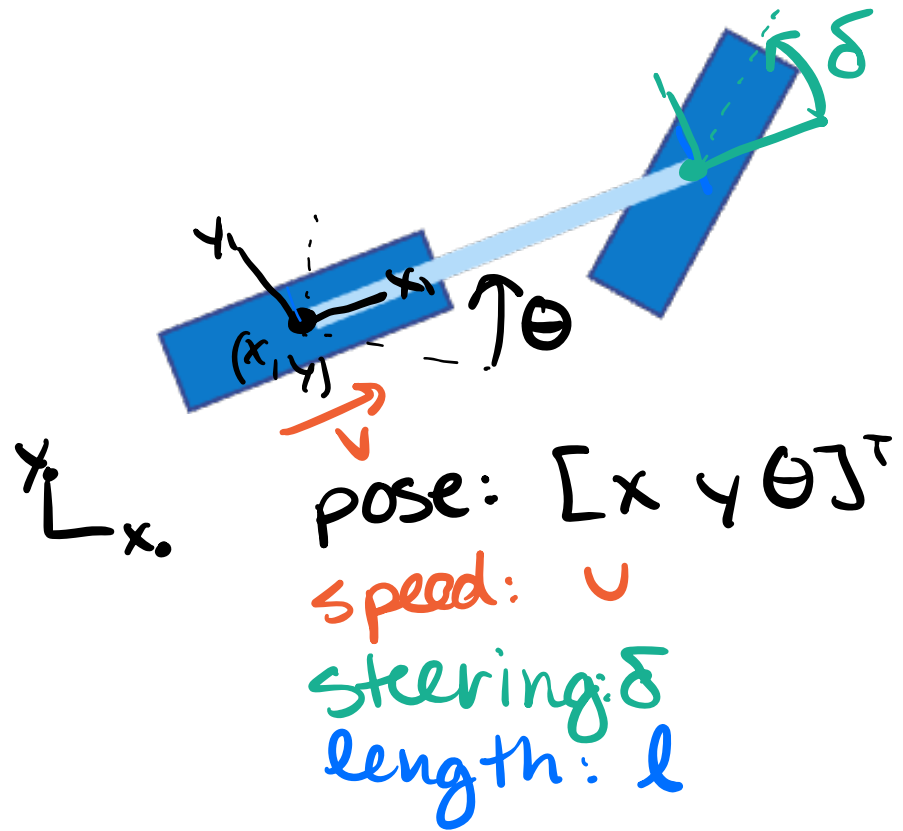


# Dubin's Car

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

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

$$\dot{\theta} = \frac{v}{l} \tan \delta$$



# Many more advanced models...

## [Kinematic] Bicycle Model

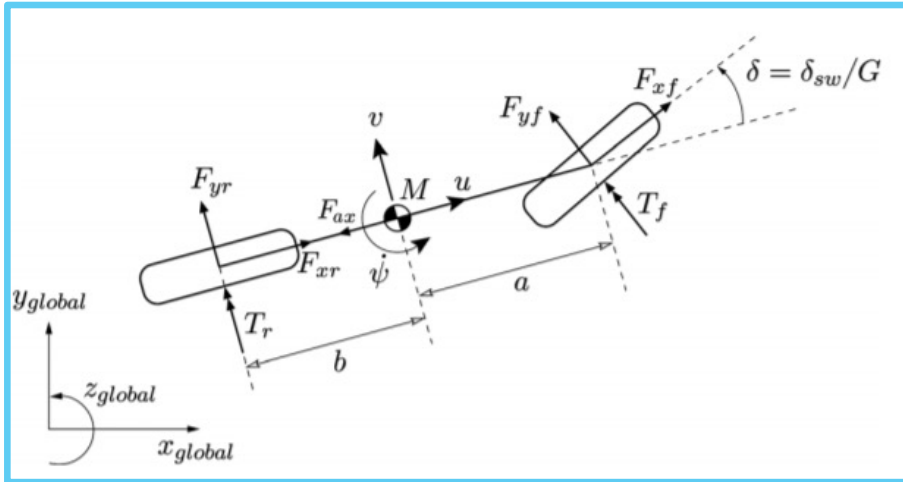


Image Credit and Reference: J.P. Timings and D.J. Cole. "Minimum maneuver time calculation using convex optimization." *Journal of Dynamic Systems, Measurement, and Control* 135.3 (2013).

## [Dynamic] Tire Models

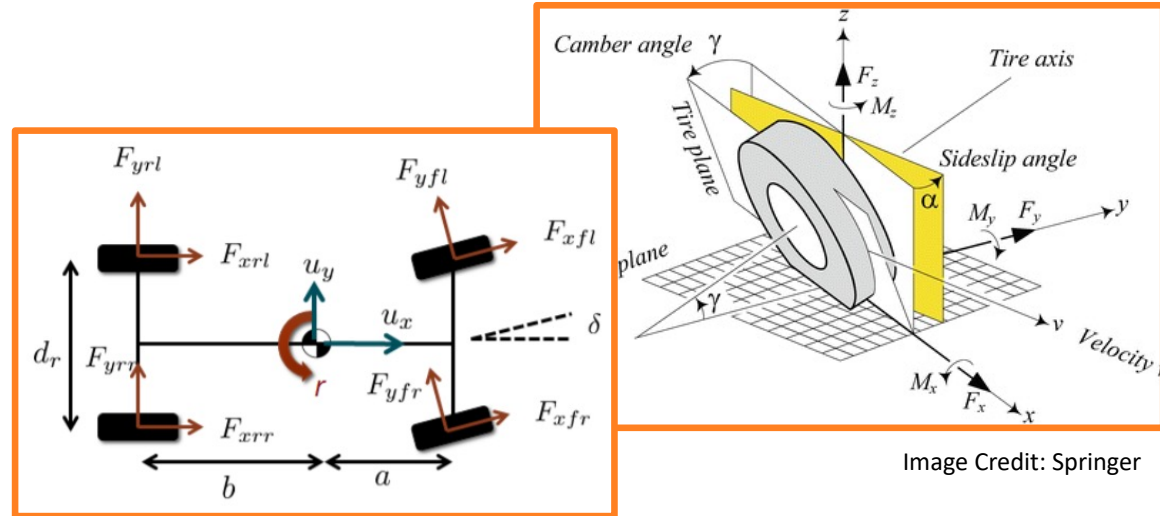
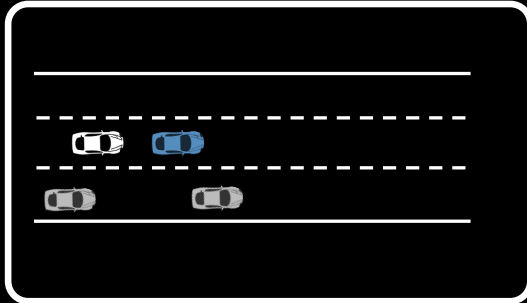


Image Credit: Springer

Image Credit and Reference: J.K. Subosits and J.C. Gerdes. "Impacts of Model Fidelity on Trajectory Optimization for Autonomous Vehicles in Extreme Maneuvers." *IEEE Transactions on Intelligent Vehicles*, 2021.



# Dynamical system models



Nonlinear dynamics

Generally, nonlinear ODEs do not have closed form solutions!

## Dubin's car model

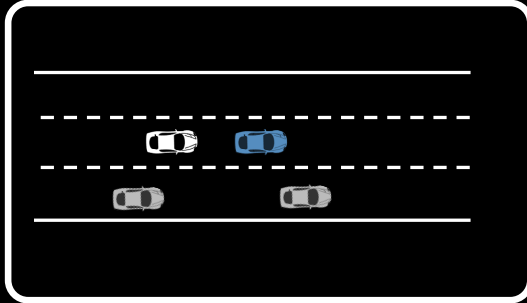
$\dot{v} = a$	Speed
$\frac{ds_x}{dt} = v \cos(\psi)$	Horizontal position
$\frac{ds_y}{dt} = v \sin(\psi)$	Vertical position
$\frac{d\delta}{dt} = v_\delta$	Steering angle
$\frac{d\psi}{dt} = \frac{v}{l} \tan(\delta)$	Heading angle

## Physical plant

$\frac{dx}{dt} = f(x, u)$	System dynamics
$x[t + 1] = f(x[t], u[t])$	
$x = [v, s_x, s_y, \delta, \psi]$	State variables
$u = [a, v_\delta]$	Control inputs



# Nonlinear hybrid dynamics



Physical plant

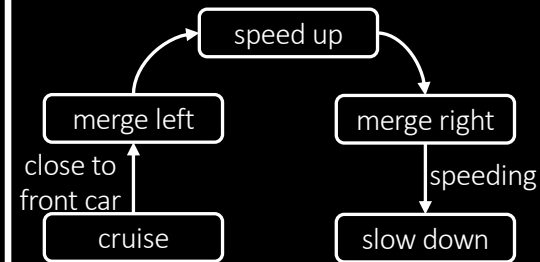
$$\frac{dx}{dt} = f(x, u) \quad \text{System dynamics}$$

$$x[t + 1] = f(x[t], u[t])$$

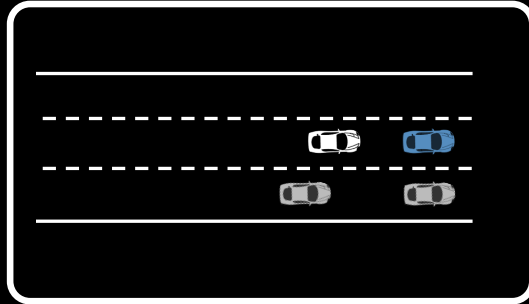
$$x = [v, s_x, s_y, \delta, \psi] \quad \text{State variables}$$

$$u = [a, v_\delta] \quad \text{Control inputs}$$

Decision and control software

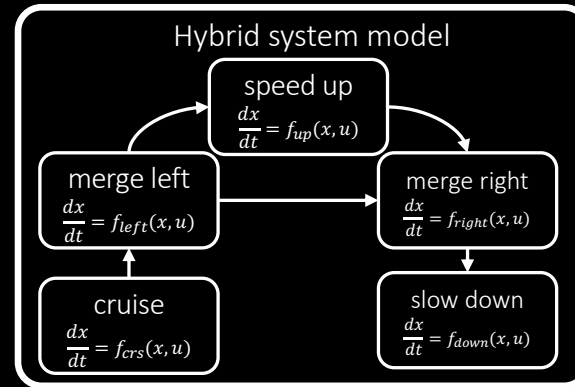
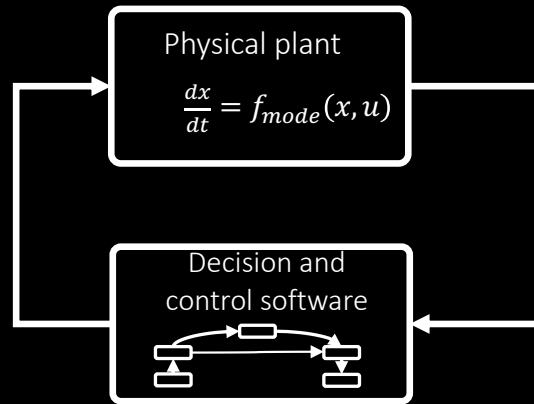


# Typical system models



Nonlinear *hybrid* dynamics

Interaction between computation and physics can lead to unexpected behaviors



# Summary

- Dynamical systems models allow us to reason about low-level behaviors of systems and determine what is and is not feasible
  - Typically required to design controllers!
- Discussed a few types of models from simple to complex
- *Next time*: Look at simple PID control design for trajectory following



# Extra Slides



# An aside: Coordinate transformations

## Rotation matrix

The following matrix rotates a vector  $[x, y]$  counter-clockwise by an angle of  $\theta$

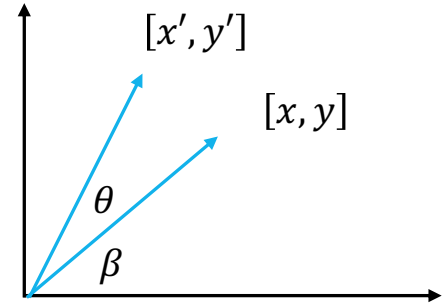
$$R(\theta) = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix}$$

That is:

$$\begin{bmatrix} x' \\ y' \end{bmatrix} = \begin{bmatrix} \cos \theta & -\sin \theta \\ \sin \theta & \cos \theta \end{bmatrix} \begin{bmatrix} x \\ y \end{bmatrix}$$

Derivation

$$\begin{aligned} x' &= r \cos(\beta + \theta) = r(\cos \beta \cos \theta - \sin \theta \sin \beta) \\ &= r \cos \beta \cos \theta - r \sin \theta \sin \beta \\ &= x \cos \theta - y \sin \theta \end{aligned}$$





# 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  $p(t)$

