Principles of Safe Autonomy ECE 484 Spring 2022 Lecture 1

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Jan 20, 2022

https://publish.illinois.edu/safe-autonomy/

https://mitras.ece.illinois.edu/

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Welcome from Safe Autonomy team!

Professor

Sayan Mitra (mitras) (CSL 266) TAs

Tianchen Ji (tj12)

Pranav Sriram (psriram2)

Lab assistants

Haoyuan You (hy19) Ninghan Zhong (innghan2)



Sayan Mitra (Instructor)



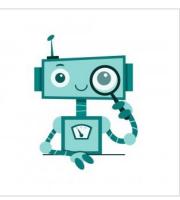
Tianchen Ji (tj12)



Pranav Sriram (psriram2)



Ninghan Zhong (ninghan2)



Haoyuan You (hy19)



Plan for today

- ► What is this course about?
- ► Why *Principles* of Safe Autonomy?
- ► How will this course work?



100 years of progress in safer roads

Traffic infrastructure (e.g., lane markings, traffic signals, etc)
Police enforcement and traffic regulations
Driver training
Passenger safety (e.g., seatbelts, airbags)
Improved vehicle design (e.g., crumple zones)
Rear-view and blind spot sensors (e.g., camera)
Advanced Driver Assistance Systems (e.g., ABS, ACC, etc.)





Autonomous systems will be awesome!

Driverless cars will make us more productive

 Average American drives 13,474 miles (300 hrs) per year

Our cities will be greener

• 40% of city surface is parking

Travel and deliveries will be safer

• 32K+ fatalities and 3M+ injuries every year



PREDICTIONS SCORECARD, 2022 JANUARY 01 by Rodney Brooks http://rodneybrooks.com/predictions-scorecard-2022-january-01/

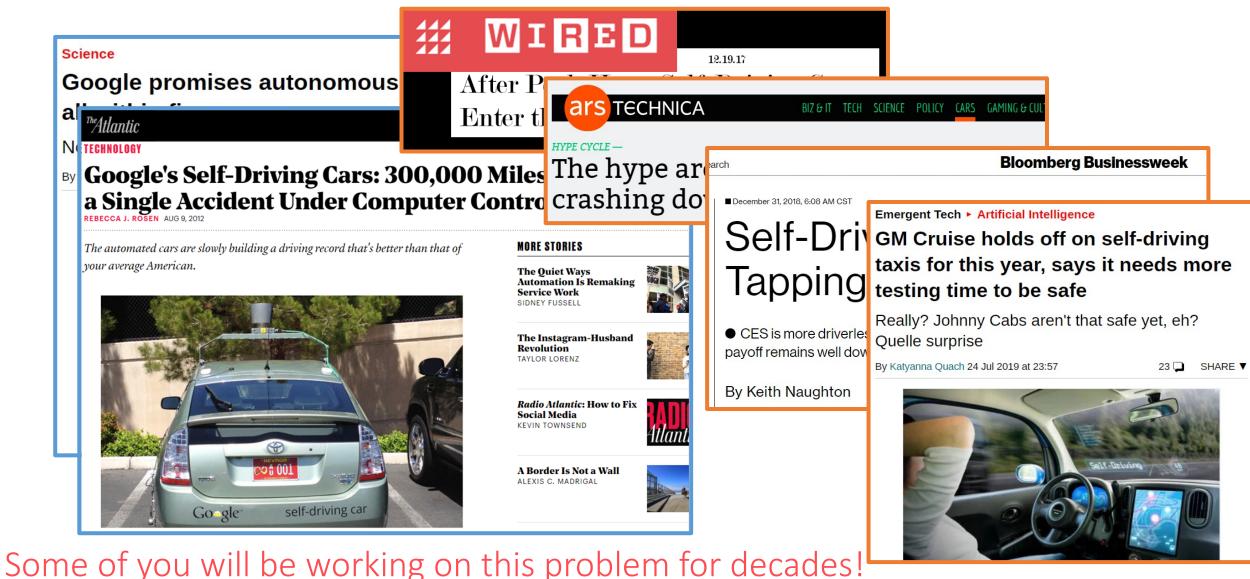
The years in blue indicate when the industry leaders thought these predictions would come to pass. I have highlighted all the dates up through 2021, now numbering 17 of the 23 predictions. Not one of them has happened or is even close to happening.

FORECASTS: <u>http://www.driverless-future.com/?page_id=384</u> March 27, 2017

NVIDIA to introduce level-4 enabling system by 2018 (2017) NuTonomy to provide self-driving taxi services in Singapore by 2018, expand to 10 cities around world by 2020 (2016) Delphi and MobilEye to provide off-the-shelf self-driving system by 2019 (2016) Ford CEO announces fully autonomous vehicles for mobility services by 2021 (2016) Volkswagen expects first self driving cars on the market by 2019 (2016) GM: Autonomous cars could be deployed by 2020 or sooner (2016) BMW to launch autonomous iNext in 2021 (2016) Ford's head of product development: autonomous vehicle on the market by 2020 (2016) Baidu's Chief Scientist expects large number of self-driving cars on the road by 2019 (2016) First autonomous Toyota to be available in 2020 (2015) Elon Musk now expects first fully autonomous Tesla by 2018, approved by 2021 (2015) US Sec Trans: Driverless cars will be in use all over the world by 2025 (2015) Uber fleet to be driverless by 2030 (2015) Ford CEO expects fully autonomous cars by 2020 (2015) Next generation Audi A8 capable of fully autonomous driving in 2017 (2014) Jaguar and Land-Rover to provide fully autonomous cars by 2024 says Director of Research and Technology (2014) Fully autonomous vehicles could be ready by 2025, predicts Daimler chairman (2014) Nissan to provide fully autonomous vehicles by 2020 (2013) -Truly autonomous cars to populate roads by 2028-2032 estimates insurance think tank executive (2013) Continental to make fully autonomous driving a reality by 2025 (2012)

Hubris on a mass delusion scale. Audi fully autonomous by 2017? That is Teslan in its delusion level.

Building autonomous cars will be harder than going to the Moon



Plan for today

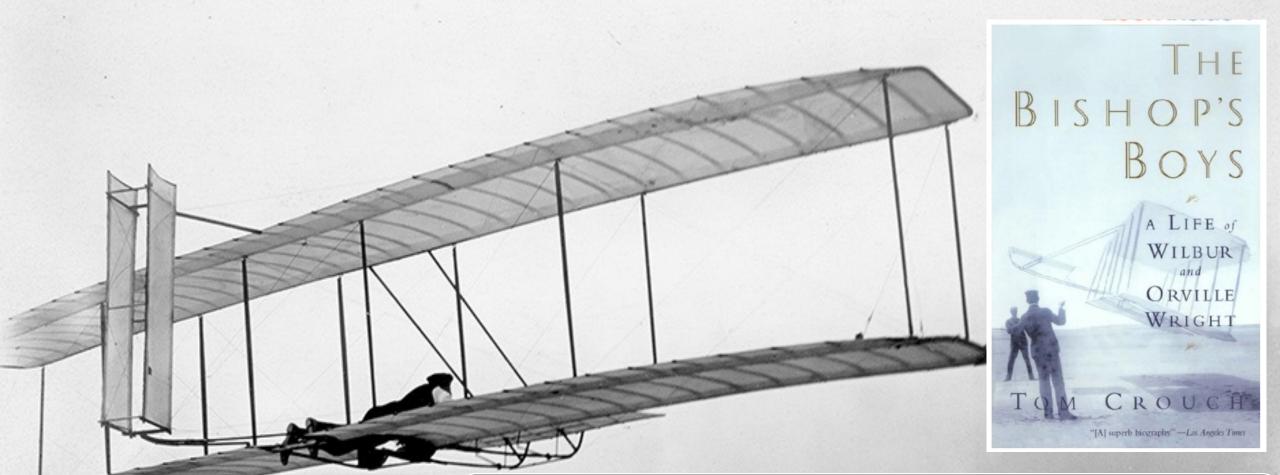
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Another transportation revolution from a century ago

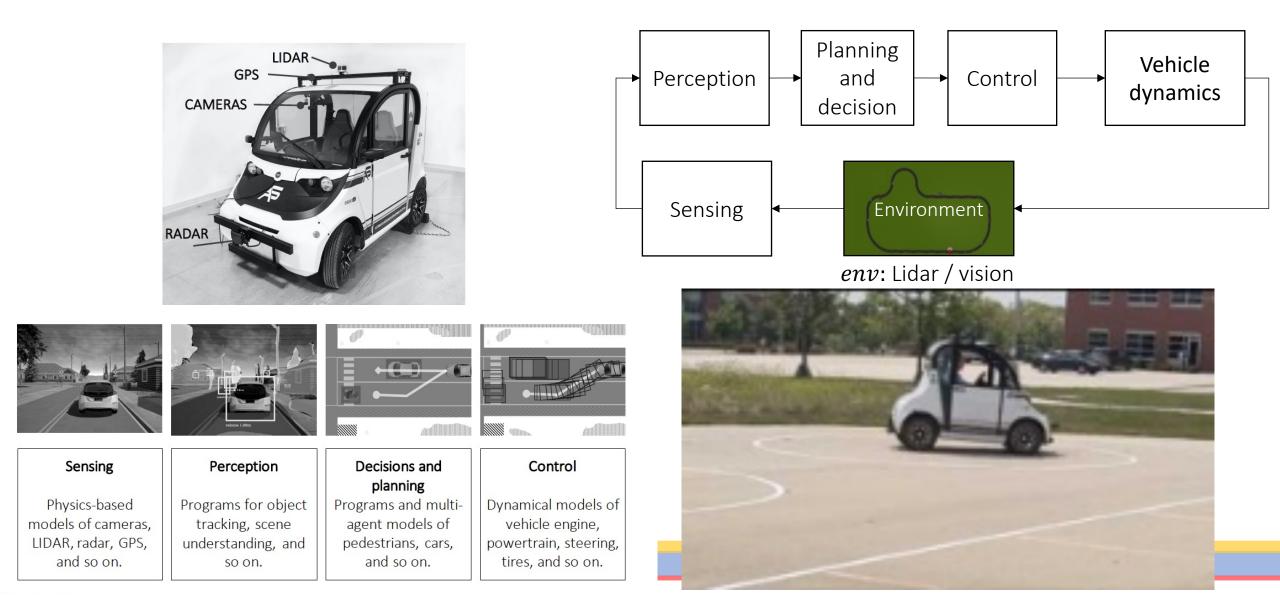






1902: after thousands of tests Oliver and Wilbur Wright had acquired the knowledge and the skill to fly. *They could soar, float, dive, circle and land, all with assurance*. Now they had only to build a motor. - David McCullough

Autonomous GEM vehicle: An example CPS



We need to understand the existing components and discover new principles for building safe autonomy

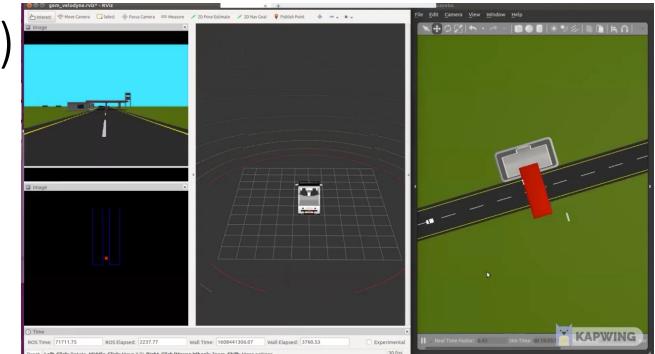


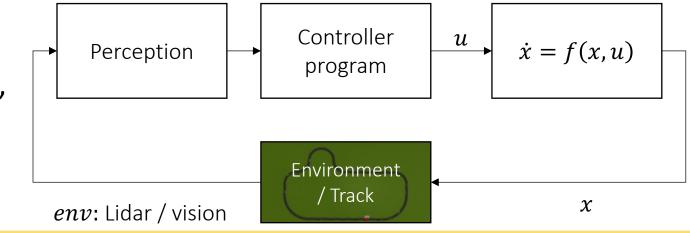
Open problem (project B)

Simulated race car following a track with Lidar-based perception and control.

Problem: For a given track and initial conditions check that the *trajectory* of the car does not collide and stays in lane.

Can we check *efficiently*? Can we *generalize* to *similar* tracks? What should we assume about perception, accuracy of the vehicle model? What should we assume about the execution of the controller?







A disagreeable position about autonomy

"Collect lots of data, build faster computers, and train larger neural networks, and safe autonomous vehicles will follow" --- A founder How to assure safety of an AV ? Run tests In an absolute sense:

"Testing can be used to show the presence of bugs, but never to show their absence!" --- Edsger W. Dijkstra

Because there are infinitely many *executions* and we can only test finitely many of those in any testing algorithm

In a probabilistic sense also, purely using data to gain safety assurance is not practical



Naively collecting test driving miles is also not going to work

Probability of a fatality caused by an accident per one hour of human driving is known to be 10⁻⁶

Assume* that for AV this has to be 10^{-9}

Data required to guarantee a probability of 10^{-9} fatality per hour of driving is proportional to its inverse, 10^{9} hours, 30 billion miles

Multi-agent, open system, with human interactions => cannot be simulated offline to generate data

Any change is software means tests have to be rerun

To learn or extrapolate about all---infinitely many---executions from a finite sampling of executions, we need to make some assumptions about the system. A collection of these assumptions defines a <u>model</u>

Different types of model (and data) for sensing, control, planning, and we need to understand how to analyze and compose them

<u>On a Formal Model of Safe and Scalable Self-driving Cars</u> by Shai Shalev-Shwartz, Shaked Shammah, Amnon Shashua, 2017 (Responsibility Sensitive Safety)

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Why are we here? Course goals





Components of an autonomous system , safety standards, ...

How to use software modules for perception, planning, control, ROS, Yolo, OpenCV, Z3, ...





Code and analyze algorithms for perception clustering, convolution, filtering, edge detection, filtering, localization, planning, formal verification

Plan, propose, organize and execute a team project





Models, algorithms, data, biases, assumptions for building trustworthy autonomous systems Theoretical properties of algorithms and their limitations



Become the Isaac Newton of Autonomy

"To do things right, first you need love, then technique." – Antoni Gaudí

Why are we here? Course goals



MPs, Homework, Project, Participate

Understand

Know

Do

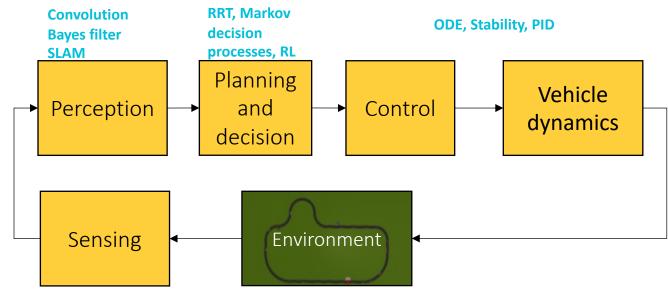
Lectures, MP, Homework, Exams



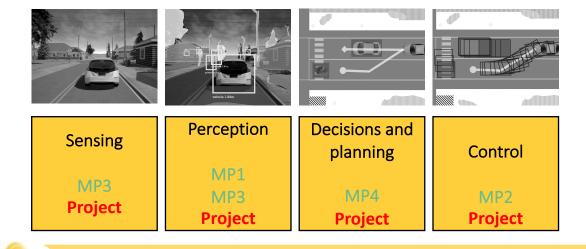
Course structure



Safety, end-to-end testing, simulation, system integration MPO, MP5, Project State machines, model checking, hypothesis testing, ROS



env: Lidar / vision





About the course

Everything starts here: https://publish.illinois.edu/safe-autonomy/

Schedule, lab, resources, papers, homework, MP, code, project, gitlab links

Campuswire for announcements, but no SLA

- Discussions, forming teams, occasional polls, feedback
- Do not expect to get answers to HWs, MPs, and exam related questions in the last minute.

Canvass for MP release and grades



Schedule: https://publish.illinois.edu/safe-autonomy/schedule-spring-2022/

- Simple safety (MPO, 1 week)
- Perception (MP1 lane detection, 2 weeks)
- Modeling and control (MP2 vehicle control, 2 weeks)
- Filtering: localization particle filtering (MP3 localization, 2 weeks)
- March 1: Midterm 1
- Planning (MP4 planning, 2 weeks)
- Spring Break
- Decision making RL (MP5 racing, end-to-end safety)
- End-to-end system and safety analysis
- April 14: Midterm 2

- Group formation
- Lab safety training
- Labs, MPs

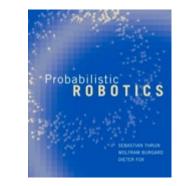
- Spring Break
- Project pitch
- Intermediate checking
- Practice presentation
- Final presentation
- Report

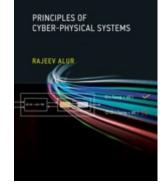
Course materials

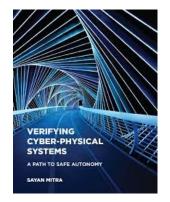
Lecture notes, slides, code, video lectures, lab manuals created and curated from recent research publications

Reference books:

- Probabilistic robotics, By Sebastian Thrun, Wolfram Burgard and Dieter Fox, 2005
- Principles of Cyber-Physical Systems, Rajeev Alur, MIT Press, 2015
- Verifying Cyber-physical Systems, Sayan Mitra, MIT Press 2021* (if you are interested in safety verification)









Course: components and (tentative) weights

- ► 5-6 programming assignments or MPs 45% (group)
 - ROS + Python, Ubuntu, VM BYOD or use lab workstations
 - labs (Friday 4-8 pm starting tomorrow)
 - ► Office hours
- Homework assignments 10% (individual)
 - math, analysis, critical reasoning; preparation for midterms
- Midterms x2 20% (individual)
- Mini project 20% (group): more on this later, 2 tracks:
 - A. Dev and test concepts on GEM
 - ► B. <u>GRAIC autonomous racing competition / testing</u>
- Participation 5%
 - Lab and class attendance and participation, early feedback on notes, solve exercise problems in class notes, class participation, MP beta testers

Tentative grade boundaries		
A	>	90
В	>	80
С	>	65
D	>	55

Homework, participation, & exam: Individual work 35%

- Testing principles, concepts
- Read course notes and slides routinely; exercises are provided
- Homework sets (synchronized with MPs)
- > 2 in-class midterms March 1, April 14
- ► No final exam



Teamwork: MP, labs, and mini project

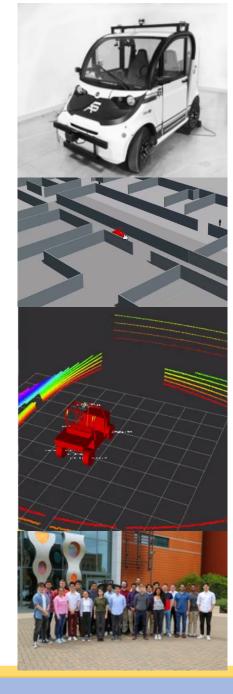
- In groups: Form your group of 3-4 now! <u>Make your group</u> (form), make new friends
 - ▶ If you do not have a group by Sunday (12 midnight AoE) we will assign you a group
- ► Each MP will build a significant component of an autonomous system over **2 weeks**
- Use our VM or your own computer with Ubuntu 16.04 or broadband internet connection to remote login to lab computers
- ▶ TAs and LAs will run live labs in ECEB 5072 (first one on zoom)
 - ▶ MP walkthough, setup, bridge the lecture and the assignments
- MP0+HW0 will be release this Friday (1/21), labs starts this Friday
 - > Your entire group has to attend 1 lab after each new MP is released,
 - And 1 lab after the MP is due (to demo your work). For the other weeks, at least one representative from the group should discuss progress with MPs and projects



Mini projects: explore, inspire, and impress

- GEM Track. Build on existing SW to live demo at Highbay
 - > Reliable parallel parking, lane following and pedestrian avoidance outdoor track, biker intent estimation and reaction
 - COVID related uncertainty about highbay access; Frontload the work
- GRAIC Track. Participate in an open simulation-based autonomous racing competition
 - Build your on driver software, test it in a number of environments
 - Test, verify, find bugs in other people's sdriver software
 - Automatically create complex scenarios
- <u>Outcomes</u>: Write research papers, jumpstart grad research, career in autonomy, incubate startup ideas, sharpen presentation skills
- We provide: Polaris GEM vehicle (camera, LIDAR, RADAR, IMU, GPS, and drive-by-wire system) modules for pedestrian detection, lane tracking, and vehicle control, a vehicle simulator, and testing facility (highbay) with indoor positioning system. GRAIC autonomy software stack
- Expertise (TA, lab and office hours, TBD)
- ► Timeline: <u>Get started, be a member of IRL from this link</u>
 - High-bay virtual site visit and training (in next 2 weeks)
 - Project pitch after Spring break
 - Public presentation, demo, awards (End of April or May 6)

Spring 2020 projects Fall 2020 projects



Summary

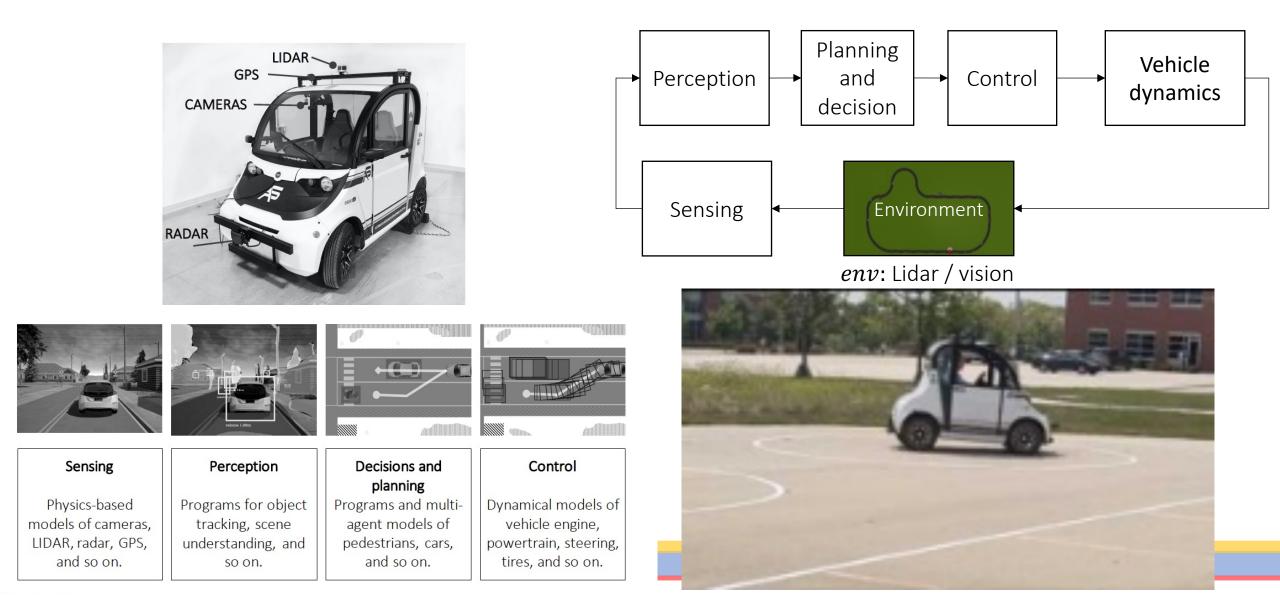
- ▶ Form your team. Decide track. Sign-up to be member of IRL
- Careful modeling and reasoning can expose flawed assumptions, bad design bugs, make the system explainable
- Discrete time model: states, initial states, transition function
- Requirements, e.g., safety



Absolute safety analysis



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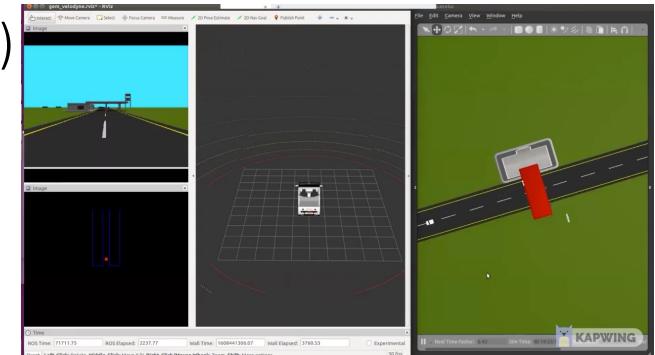


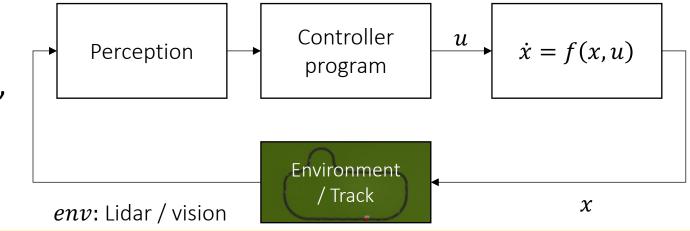
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Today: How can we use a simple mode to get absolute safety guarantees

- ► A simple class of models: automata
- What are executions of automata?
- What are safety requirements?
- Reachable states, Invariants for safety guarantees



A "simple" safety scenario

A car moving down a straight road has to detect any pedestrian (or another car) in front of it and stop before it collides.

Automatic Emergency Braking

Not a trivial requirement

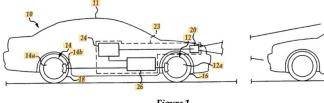
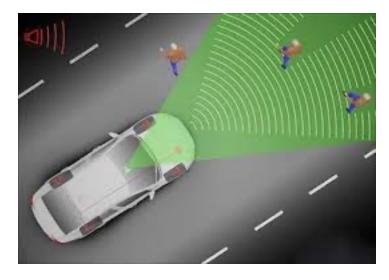


Figure 1

Today: There is no enforced standard for testing AEB



www.google.com > patents

US20110168504A1 - Emergency braking system - Google ...

Jump to Patent citations (18) - US4053026A * 1975-12-09 1977-10-11 Nissan Motor Co., Ltd, Logic circuit for an automatic braking system for a motor .

www.google.com > patents

US5170858A - Automatic braking apparatus with ultrasonic ...

An automatic braking apparatus includes: an ultrasonic wave emitter provided in a ... Info: Patent citations (13); Cited by (7); Legal events; Similar documents; Priority and ... US6523912B1 2003-02-25 Autonomous emergency braking system.

www.google.com > patents

DE102004030994A1 - Brake assistant for motor vehicles

B60T7/22 Brake-action initiating means for automatic initiation: for initiation not ... Info: Patent citations (3); Cited by (9); Legal events; Similar documents ... data from the environment sensor and then automatically initiates emergency braking.

www.google.com.pg > patents

Braking control system for vehicle - Google Patents

An automatic emergency braking system for a vehicle includes a forward viewing camera and a control. At least in part responsive to processing of captured .

www.automotiveworld.com > news-releases > toyota-ip... *

Toyota IP Solutions and IUPUI issue first commercial license ... Jul 22, 2020 - ... and validation of automotive automatic emergency braking (AEB) ... and Director of Patent Licensing for Toyota Motor North America. "We are ...

insurancenewsnet.com > oarticle > patent-application-tit...

Patent Application Titled "Multiple-Stage Collision Avoidance ...

Apr 3, 2019 - No assignee for this patent application has been made. ... Automatic emergency braking systems will similarly, also, soon be required for tractor .



"simple"≠ Easy





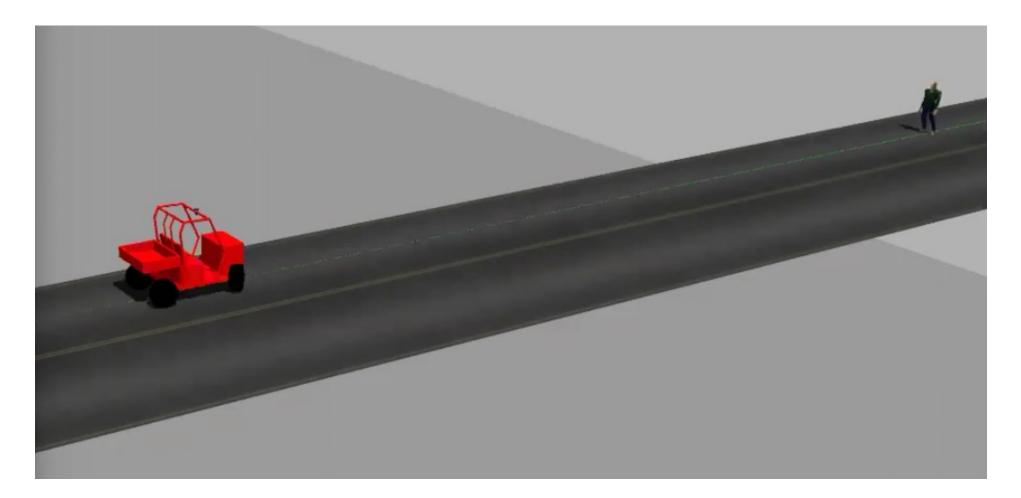
Modeling the scenario

- ► What is a model of a system?
- ► A *mathematical model* describes how a system behaves.
 - What are the key parameters and states?
 - ► How are the parameters selected by nature?
 - ► What are the initial conditions of the state?
 - ▶ How do the state change over time? ...
 - ▶ What parts of the model are available for observation/analysis?

Models include the implicit and explicit assumptions (biases) we are making about the system



Model (switch to notes)

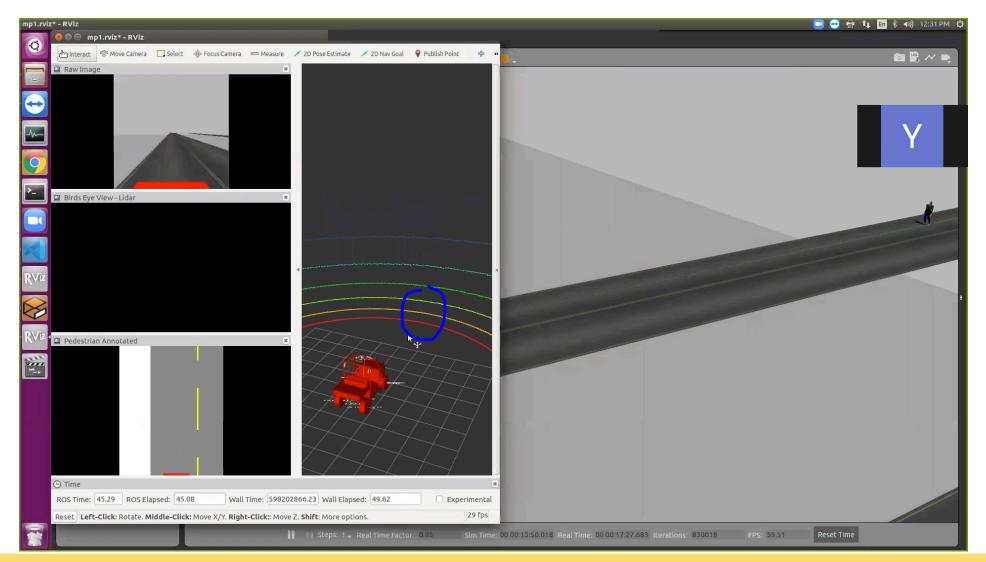




"All models are wrong, some are useful."



MPO: Simulate model for testing





An honest scientific approach

- 1. Create detailed *mathema*tical *models* of the autonomous systems and its environment
- 2. Enumerate the precise *requirements* of the system and the conditions on the environment under which it is supposed to work
- 3. Analyze the system to either
 - prove that all behaviors meet the requirement (perhaps with high probability)
 - find counter-examples, corner cases, etc., debug and repeat
- Currently there are fundamental flaws in making this work for autonomous systems
- Why study this approach?
 - Careful reasoning can expose flawed assumptions, bad design choices
 - > The approach has been successful in other industries: microprocessors, aviation, cloud computing, nuclear, ..
 - Big strides in the last few years
 - ▶ Working deliberately towards a more perfect understanding is a worthwhile intellectual struggle



Plan for today

- ► What is this course about?
- ▶ Why is this approach to autonomous systems important?
- How will this course work? (Administrivia)

