

# SELF-DRIVING CARS

## INTRODUCTION



# **Photogrammetry & Robotics Lab**

## **Introduction to Techniques for Self-Driving Cars**

**Stachniss, Behley, Mersch**

---

# Self-Driving Cars



[Courtesy of S. Thrun, S. Jurvetson, Grendelkhan]

# Aims of this Lecture

- Introduction to self-driving cars
- Getting to know key building blocks
- Gaining practical experience realizing simplified building blocks in Python
- Test your own developments in a car simulator
- Perspective from academia & industry

# **What Do Self-Driving Cars Need to Do?**

# **Basic Steps Executed by a Self-Driving Car**

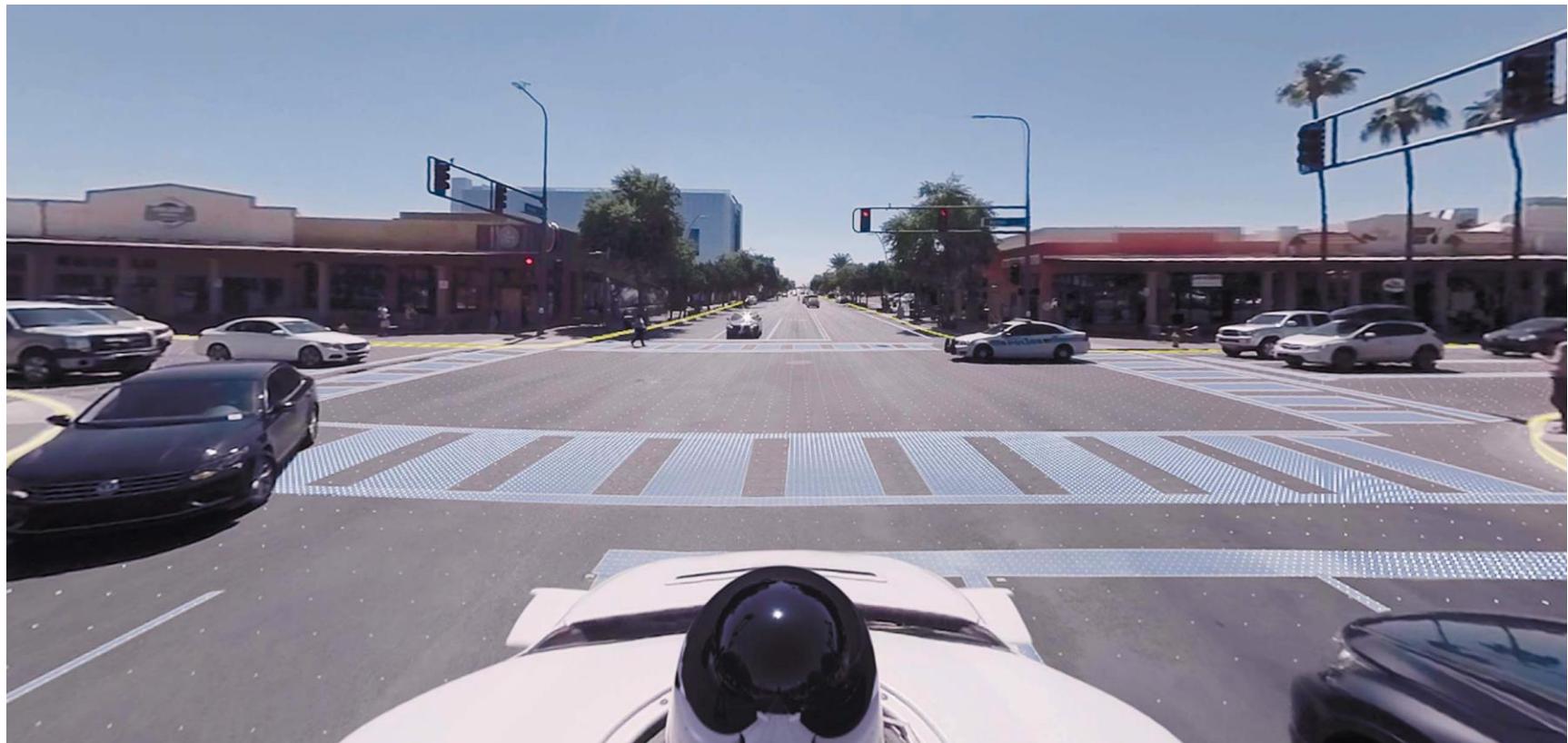
- Must have the right prior knowledge
- Estimate the own state
- Perceive the surroundings
- Predict what is going to happen
- Reason and plan what to do
- Act – execute steering commands

# Basic Steps Executed by a Self-Driving Car

- Must have the right **prior knowledge**
- **Estimate** the own state
- **Perceive** the surroundings
- **Predict** what is going to happen
- Reason and **plan** what to do
- **Act** – execute steering commands

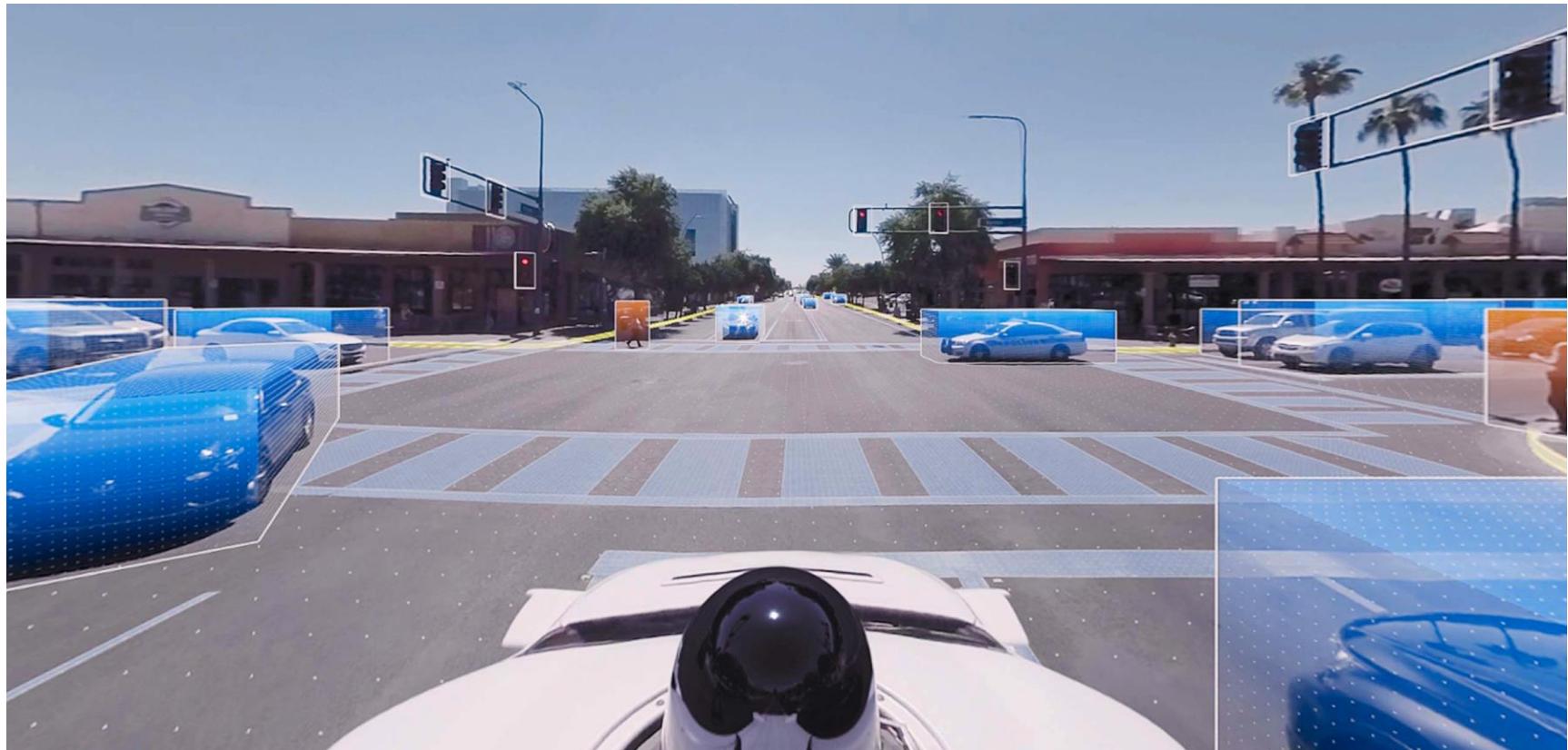
# Prior Knowledge & Own State

E.g., map of the environment & pose



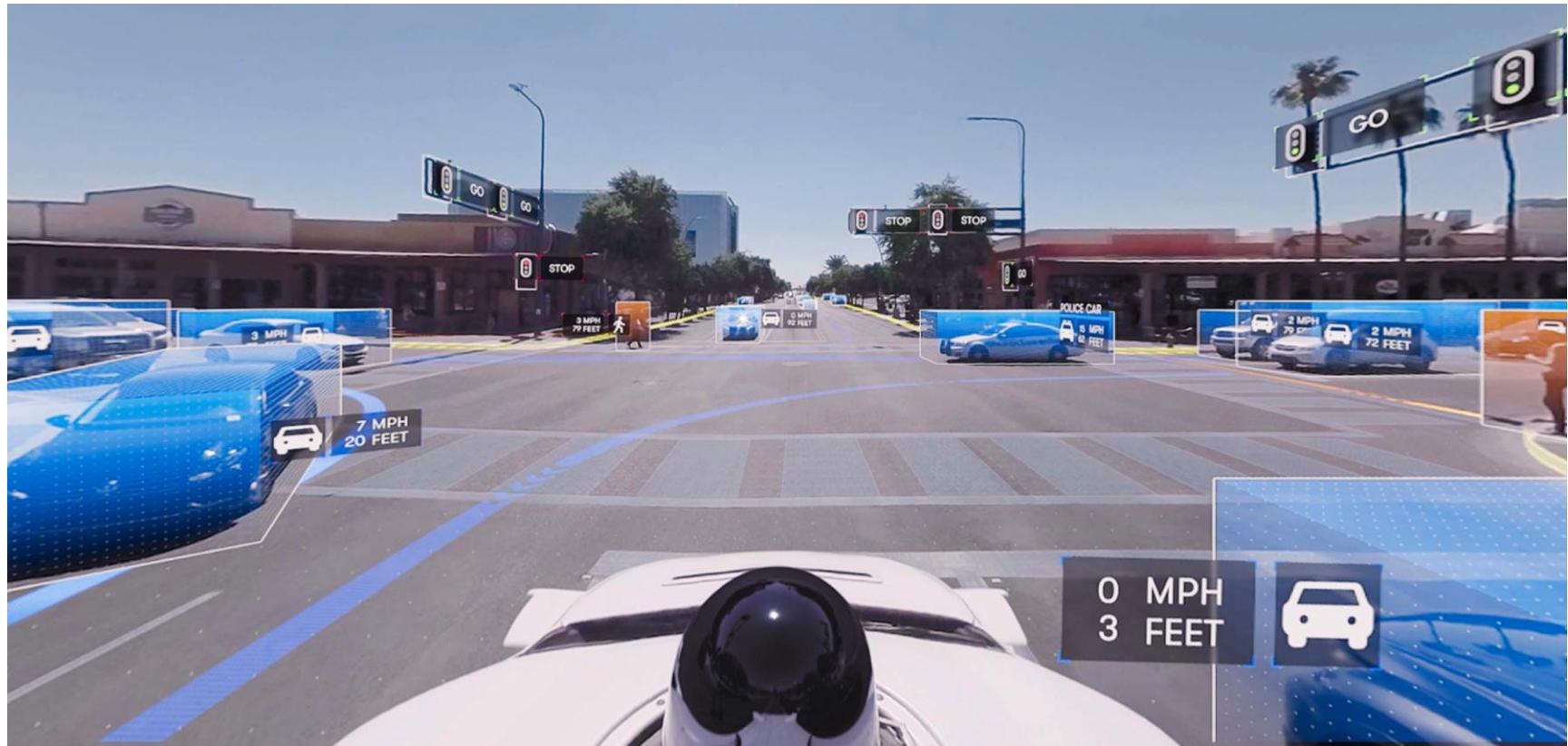
# Perception & State Estimation

E.g., what is around me and where?



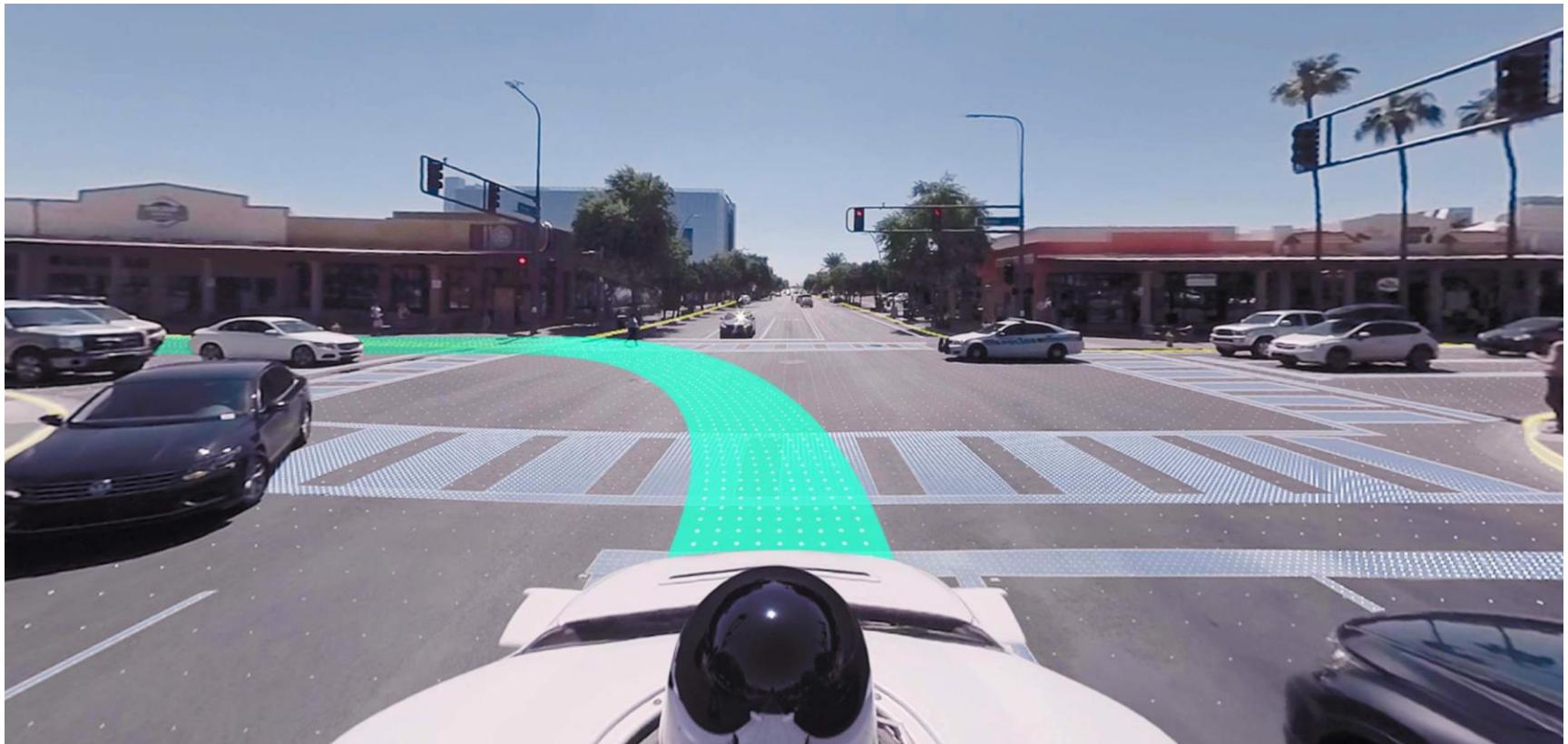
# Prediction

E.g., what will the others do?



# Act

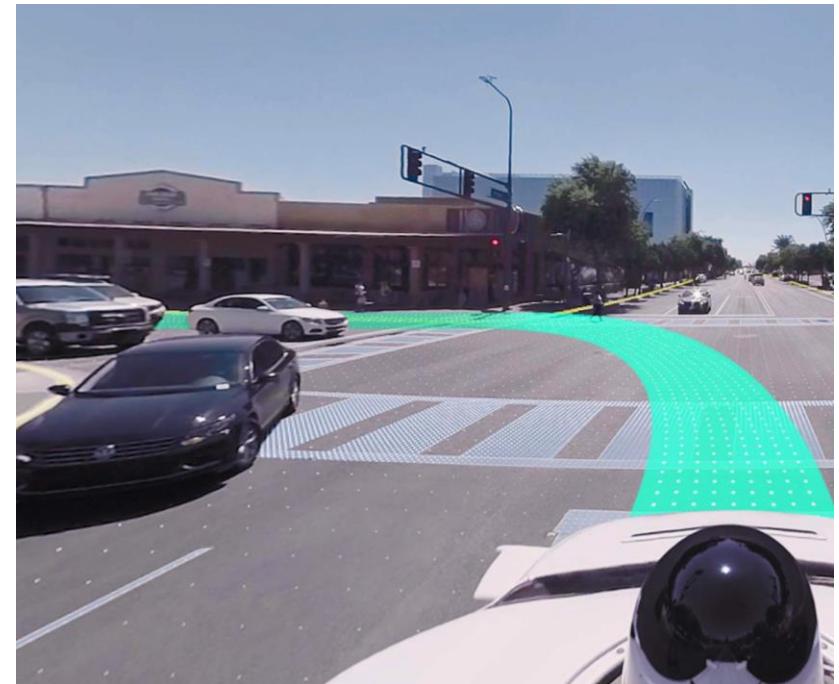
E.g., where should I drive to and how?



# Two Key Topics Primarily Addressed in This Course

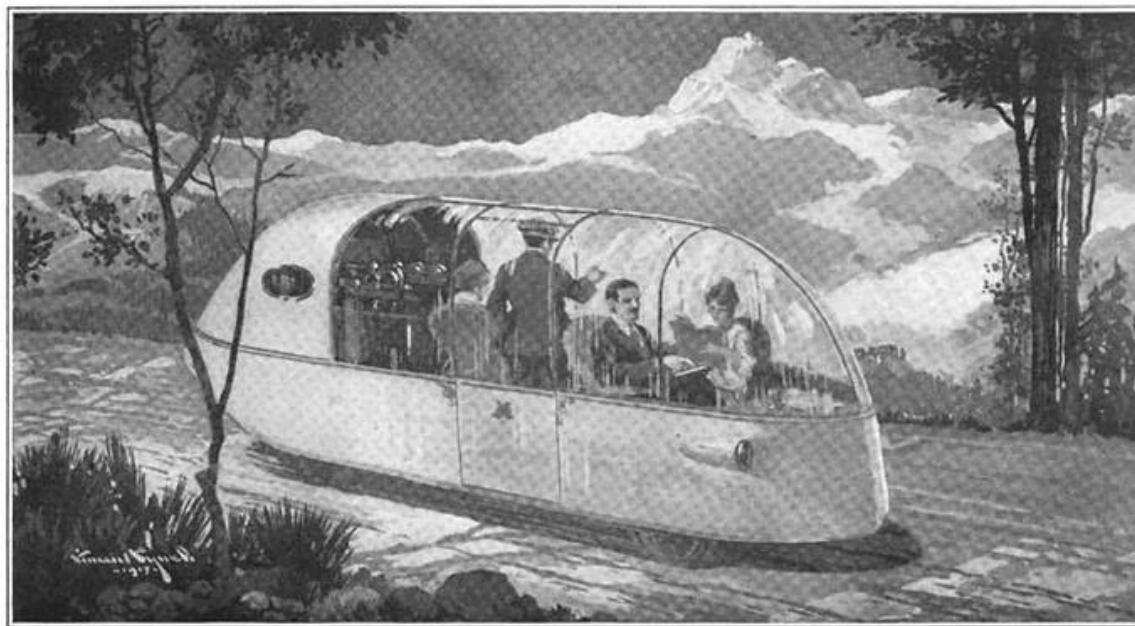


Perception:  
“Making the car see.”



Planning and control:  
“Making the car act.”

# The Dream Started in 1918...



The motorist's dream: a car that is controlled by a set of push buttons

[Courtesy of Scientific American (January 5, 1918)] 13

# History of Self-Driving Cars

- 1918: Scientific American describes “a motorist’s dream: a car that is controlled by a set of push buttons”
- 1921: First radio-controlled car
- 1930-1970/80ies: several predictions
- 1977: First street vehicle (Tsugawa et al. 1977)
- 1985: DARPA Autonomous Land Vehicle (ALV)
- 1987-94: Dickmann’s EUREKA project
- 1993/4: BOSCH/Toyota: line detection and following vehicle
- 1994: PROMETHEUS drives 1,700+km: Munich ↔ Denmark
- 1996: Project ARGO: 2,000km driving (Italy)
- 1997: Driving demonstration on I-15 (San Diego)

# History of Self-Driving Cars

- 1918: Scientific American describes “a motorist’s dream: a car that is controlled by a set of push buttons”
- 1921: First radio-controlled car
- 1930-1980ies: several predictions
- **1977: First street vehicle (Tsugawa et al. 1977)**
- 1985: DARPA Autonomous Land Vehicle (ALV)
- **1987-94: Dickmann's EUREKA project**
- 1993/4: BOSCH/Toyota: line detection and following vehicle
- **1994: PROMETHEUS drives 1,700+km: Munich ⇔ Denmark**
- 1996: Project ARGO: 2,000km driving (Italy)
- 1997: Driving demonstration on I-15 (San Diego)

# History of Self-Driving Cars: Tsukuba Mechanical Eng. 1977



# History of Self-Driving Cars: Dickmann's Achievements



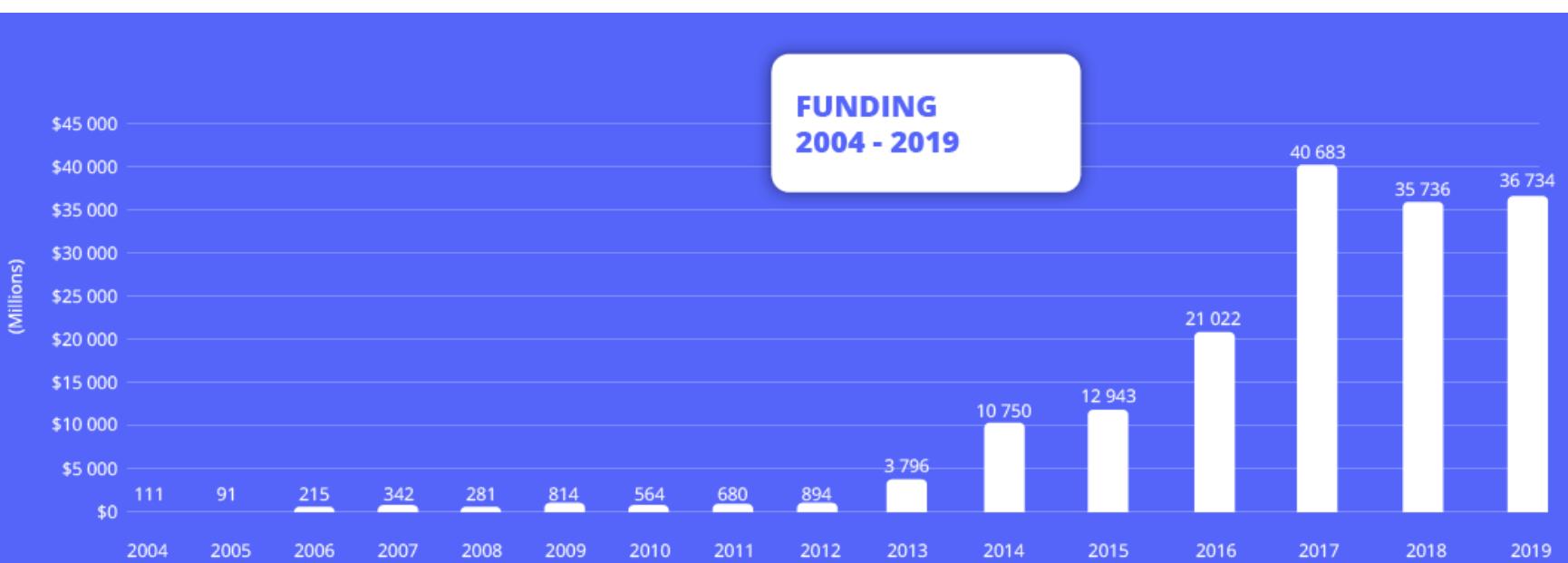
# History of Self-Driving Cars: Dickmann's Achievements



# Self-Driving Cars Since 2004

- 2004: DARPA Gran Challenge, broad interest in self-driving
- 2005: Stanley, who won the Gran Challenge
- 2005: First startups enter the scene
- 2007: DARPA Urban Challenge
- 2009: Google self-driving car project (=> Waymo)
- 2012: Stanford race-car drives up Pikes Peak autonomously
- 2013-2017: Substantial increase in funding for self-driving and several new startups enter the scene
- 2017-2020: Funding stagnates, some consolidation
- 2017-2020: Several large-scale collaborations/joint ventures
- 2020: Companies claim to be close to market but several people have doubts that it will be very close
- 2020/21: Tesla starts Autopilot
- 2022/23: Autonomous taxi services available in selected cities in the US (L4)

# Funding for Self-Driving Companies



# **Some of the Players in the Field**

- Waymo (Google X)
- GM Cruise
- Tesla
- Apple
- Baidu Apollo
- Oxa (Oxbotica)
- Apex.AI
- VW/Cariad
- Toyota/Woven/TRI
- Daimler-Bosch
- Aptiv
- Auroa
- Zoox
- Holomatic
- comma.ai
- Wayve
- Lyft

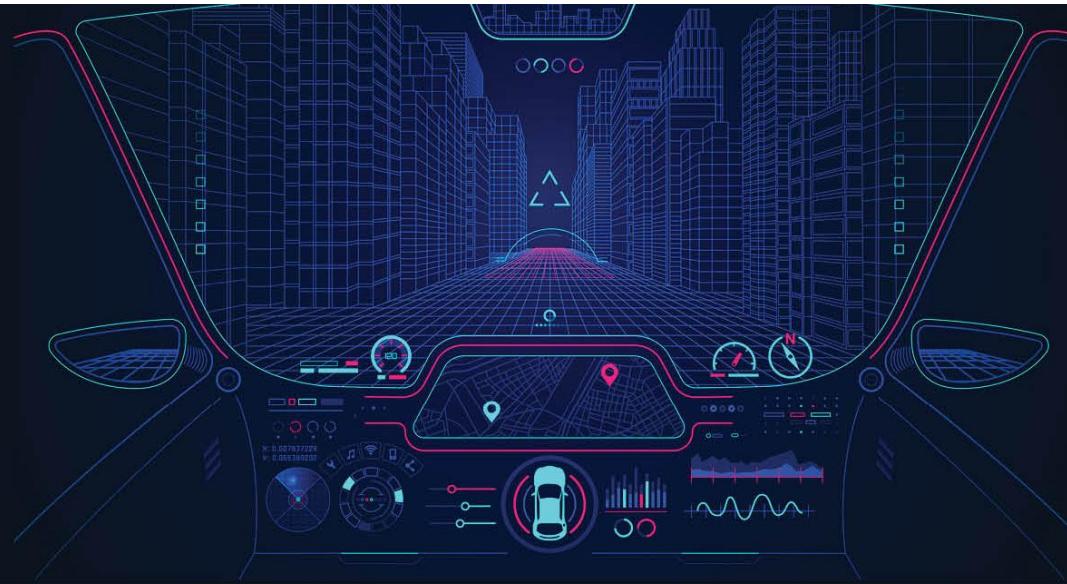
**... and several others ...**

# **What Means “Self-Driving”?**

# Levels of Autonomy

## DEFINING THE 6 LEVELS OF SELF-DRIVING AUTONOMY

BY CARLOS M. GONZÁLEZ



Self-driving cars are not all created equal. While researchers predict 8 million autonomous vehicles will be on the road by 2025, their self-driving capabilities will differ. SAE International updated the SAE J3016 in 2021 to clarify how the automotive industry can define driver support systems and advanced driver assistance systems.



# Levels of Autonomy

THE LEVELS OF AUTOMATION IN VEHICLES BREAKS DOWN INTO SIX DIFFERENT CATEGORIES:

**LEVEL 0** ► No Driving Automation

**LEVEL 1** ► Driver Assistance

**LEVEL 2** ► Partial Driving Automation

**LEVEL 3** ► Conditional Driving Automation

**LEVEL 4** ► High Driving Automation

**LEVEL 5** ► Full Driving Automation

Today's car market offers different autonomous capabilities, and the selection of vehicles is growing. So let us explore the different driving levels and which cars are better equipped to dominate the road.

# Levels of Autonomy

No Driving Automation ..... **LEVEL 0**

At this level, the human driver is entirely in control and performs all the dynamic driving tasks (DDT). Cars with active safety systems that may provide alerts or momentary emergency actions. Such subsystems as blind-spot detection, stability control, forward-collision warning, or automatic emergency braking are considered level 0 because the safety systems do not drive the vehicle.

**VEHICLES:** The majority of cars on the road today are Level 0.



# Levels of Autonomy

Driver Assistance

LEVEL

1

Level 1 is the simplest form of vehicle automation with at least one driver support. The system can perform a sustained and operational design domain (ODD) specific subtask in either the lateral or longitudinal direction as part of a DDT automation. The driver must supervise the driving automation system by performing the object and event detection and response (OEDR) and other dimensions of the vehicle motion. The automation is short-term and meant to assist the driver.

**VEHICLES:** The Audi 2021 lineup has both adaptive cruise control and hands-on lane-centering.



# Levels of Autonomy

Partial Driving Automation ..... **LEVEL 2**

In comparison to Level 1, Level 2 performs DDT automation for longer periods. The ODD performs extended motion in the lateral and longitudinal direction, but the driver is still expected to have their hands on the wheel for any emergency OEDR actions. The SAE J3016 classifies this as automated movement under the driver's supervision.

**VEHICLES:** The 2021 Genesis G70 can perform adaptive cruise control down to a stop.



# Levels of Autonomy

## Conditional Driving Automation ..... ➔ LEVEL

3

The Level 3 of self-driving automation is a significant step up from Level 2 and is not legal yet on U.S. roads. Conditional driving automation introduces sustained and ODD-specific performance by an automated driving system (ADS). Here the driver is not constantly supervising the vehicle but instead is on standby and must be ready to take control in an emergency. However, the car is driven entirely by the ADS and artificial intelligence software to make movement decisions based on its environment. An example of Level 3 driving is a vehicle capable of performing all DDT in a low-speed area or in stop-and-go freeway traffic.

### VEHICLES:

The Honda Legend became the world's first commercially available Level 3 autonomous car in 2021.



# Levels of Autonomy

## High Driving Automation ..... LEVEL 4

High driving automation is a complete ADS and does not require any human interaction to operate the vehicle, even in emergencies, as the car is programmed to stop in the event of a system failure or emergency. These cars are expected to be the future of autonomous taxis and public transportsations. Depending on the vehicle features, certain Level 4 vehicles may require human intervention to finish the route or handle transitions, such as exiting the freeway, and their capabilities may be restricted to a given geographical region. An external dispatcher may also be needed to program the destination of the vehicle.

**VEHICLES:** Waymo, owned by Alphabet Inc., is a Level 4 vehicle, and is currently testing its self-driving taxi service out of Phoenix, Arizona.



# Levels of Autonomy

Full Driving Automation ..... **LEVEL**

**5**

The highest level of automation, Level 5, requires no human interaction whatsoever. The route planning, the vehicle DDT, and the transitions between low and high-speed zones are controlled entirely by the ADS. These vehicles are not bound geographically, nor are they affected by external conditions such as weather or congested traffic environments. The only human action needed is to pick a destination.

**VEHICLES:**

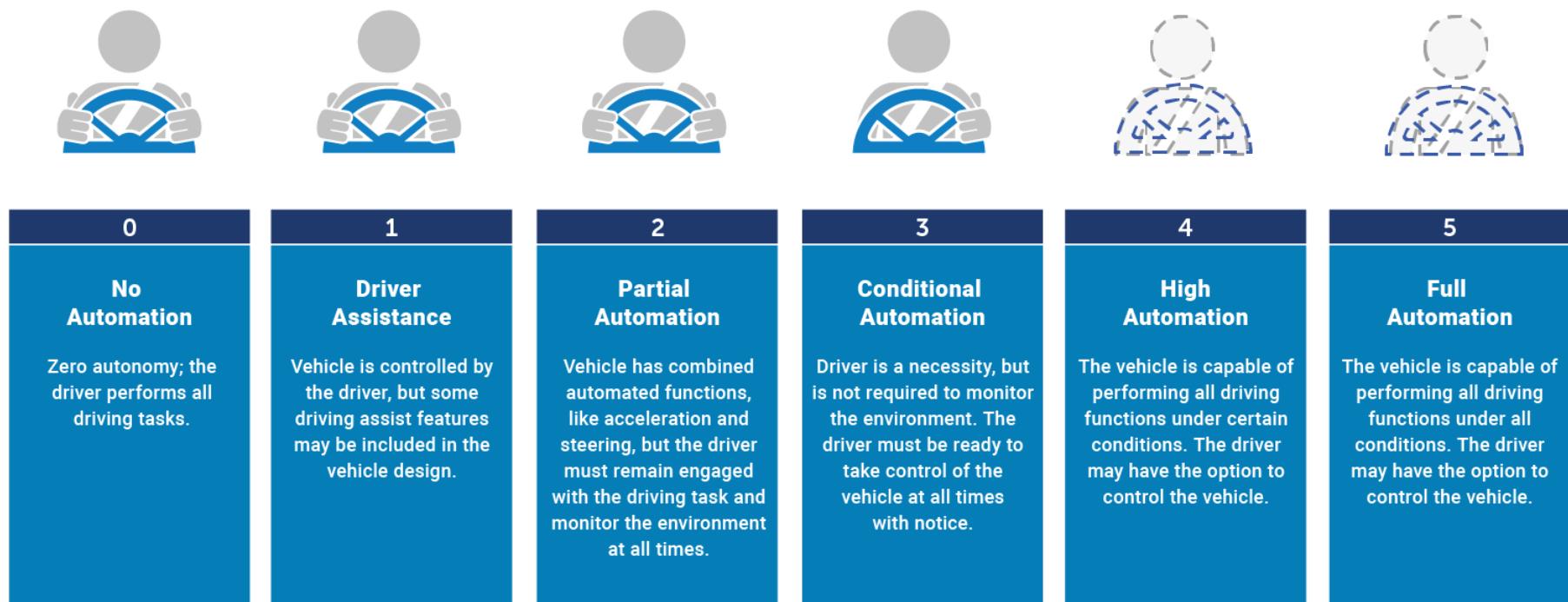
There are currently no Level 5 vehicles on the road today.



# Levels of Autonomy (by SAE)

SOCIETY OF AUTOMOTIVE ENGINEERS (SAE) AUTOMATION LEVELS

Full Automation



Society of Automotive Engineers (SAE)

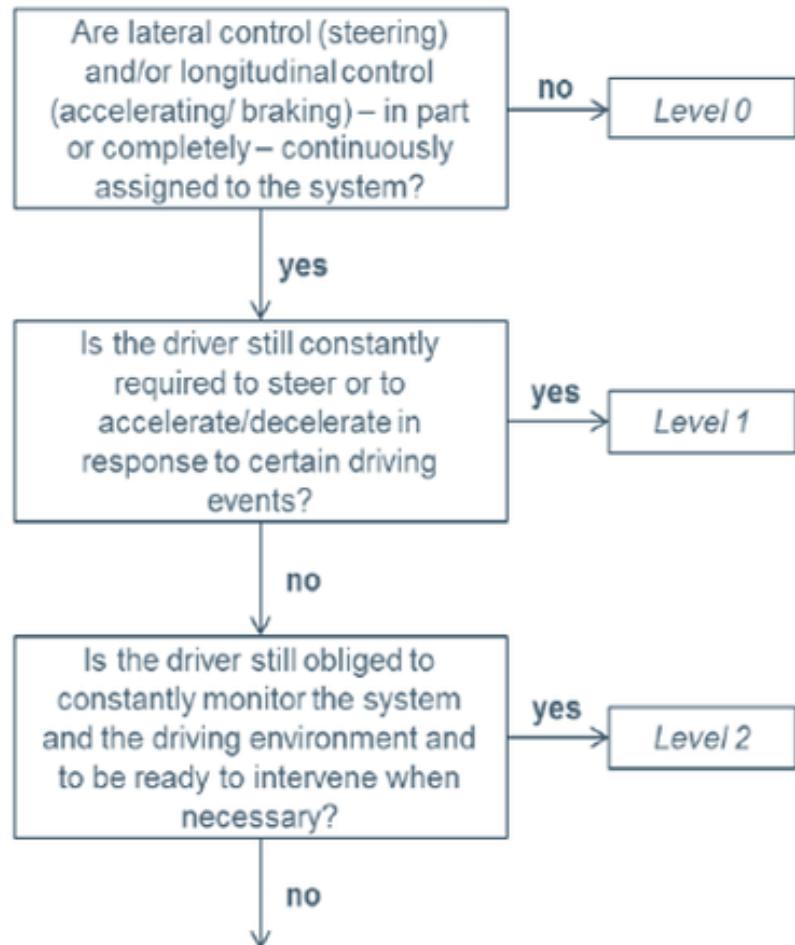
# Levels of Autonomy: L0-L2

- **L0: No** automation
- **L1: Small steering or acceleration tasks** without human intervention, but everything else is fully under human control
- **L2: Car controls** itself (steering and acceleration) and can automatically take safety actions but the **driver needs to stay alert and monitor the environment.**

# Levels of Autonomy: L3-L5

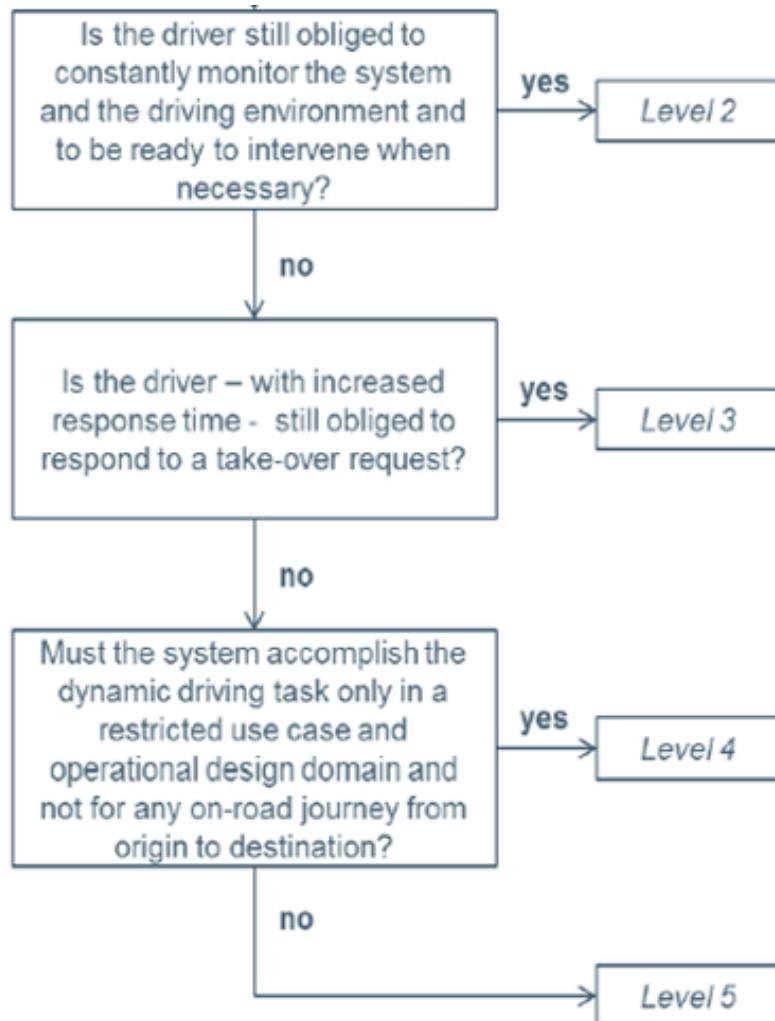
- **L3: some safety-critical functions** performed by the vehicle, under certain conditions. The **human is still required**
- **L4: Autonomous driving almost all the time** without human input. Driving may be limited to mapped areas and may not work under all weather conditions or similar. No human interaction in regular operation
- **L5: Full automation in all conditions**

# Decision Tree for Determining the SAE Level



Source: U. Eberle, "AdaptIVe"  
Final Event, 2017  
(Opel Automobile)

# Decision Tree for Determining the SAE Level

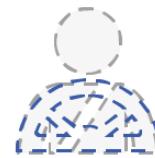


Source: U. Eberle, "AdaptIVe" Final Event, 2017 (Opel Automobile)

[Image Courtesy of Eberle] 35

# Current Target: Level 2 or 4

- Most players head for L2 or L4
- L2 as intelligent cruise control but human in charge (advancement of current cars)
- L4 as a fully “self-driving car”



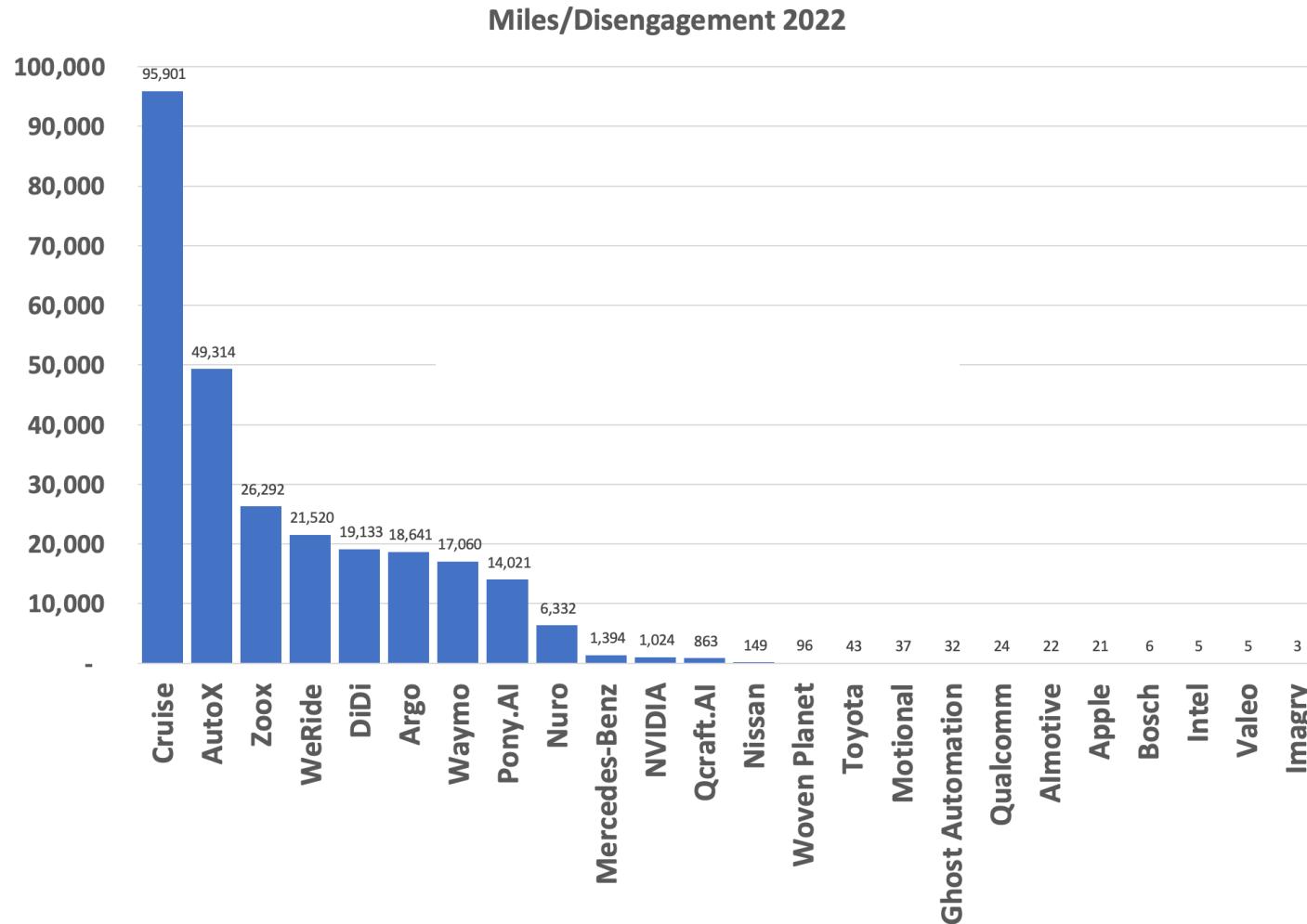
**2**  
**Partial Automation**  
Vehicle has combined automated functions, like acceleration and steering, but the driver must remain engaged with the driving task and monitor the environment at all times.

**4**  
**High Automation**  
The vehicle is capable of performing all driving functions under certain conditions. The driver may have the option to control the vehicle.

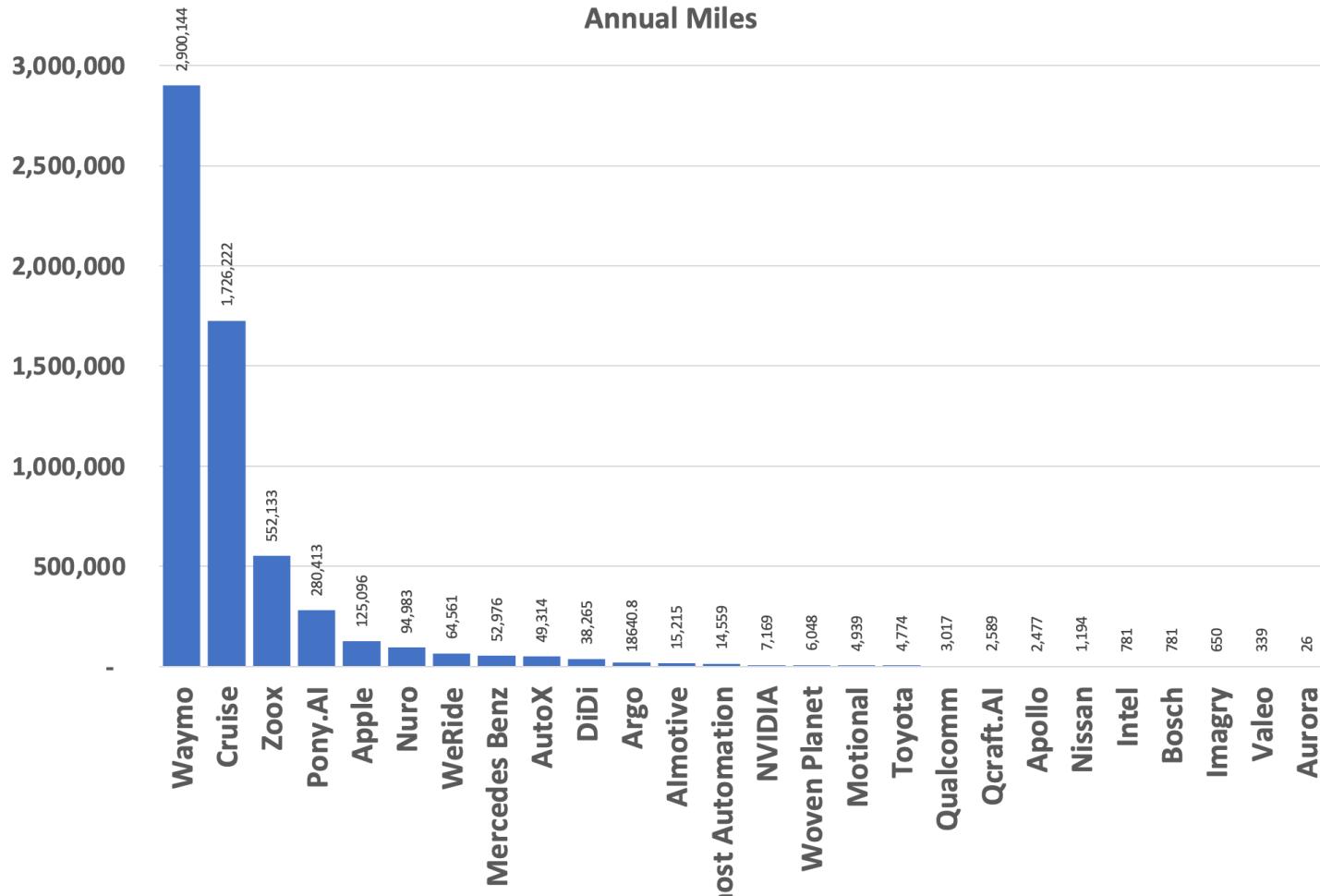
# Miles per Disengagement Driven in California in 2019



# Miles per Disengagement Driven in California in 2022



# Annual Miles Driven in California in 2022



[Courtesy of thelastdriverlicenseholder] 39

# No-Driver-in-the-Car Miles Driven in California in 2022

Companies with driverless licensees in CA: Apollo, AutoX, Cruise, Nuro, Waymo, WeRide, and Zoox

	Miles	Kilometers	Vehicles
Apollo (Baidu)	21,774	34,838	4
Cruise	546,492	874,388	222
Nuro	924	1,479	9
Waymo	51,639	82,623	317
WeRide	2,859	4,575	3
	<b>623,689</b>	<b>997,903</b>	<b>555</b>

# Example Vehicles



[Courtesy of Waymo, Pony.ai, Baidu, Cruise] 41

# Example Vehicles

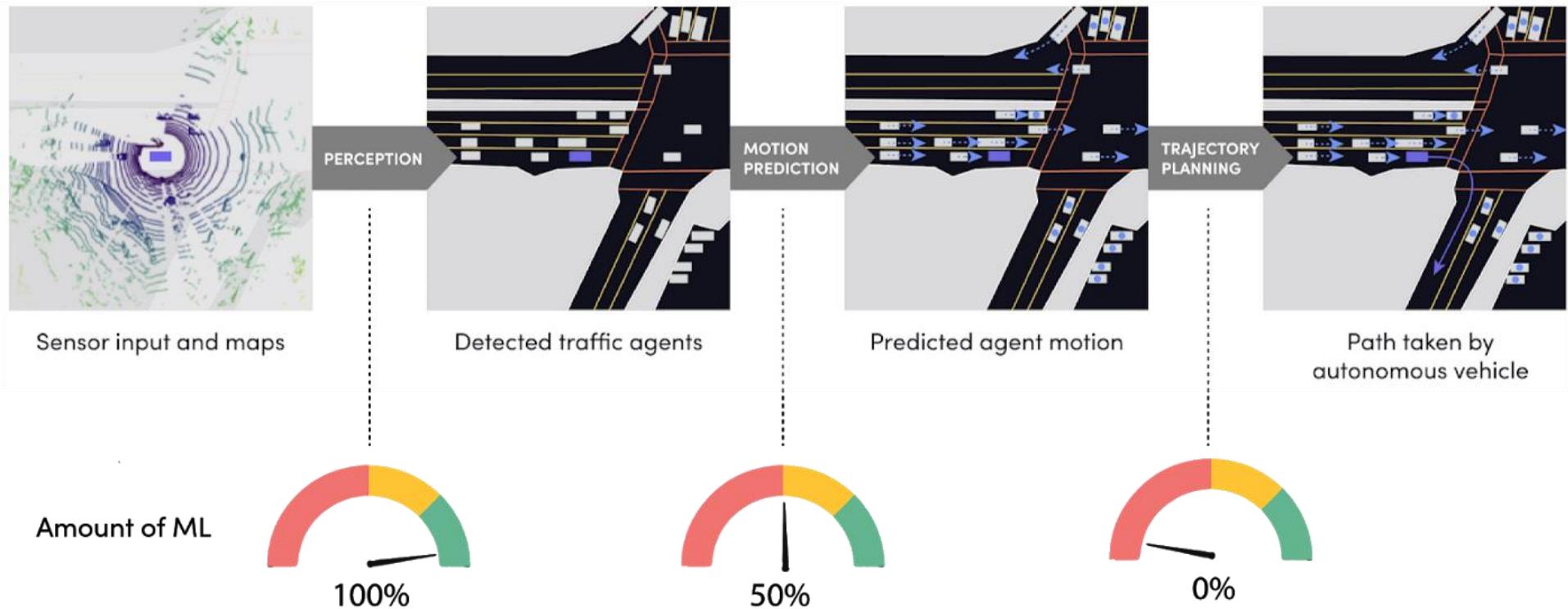


[Courtesy of Uber, Nuro.ai, Tesla, AID] 42

# Technology in Self-Driving Cars: Perceive – Estimate – Predict – Act

- **Online state estimation:**  
e.g., localization, SLAM, sensor fusion, ...
- **Machine learning/deep learning:**  
e.g., detection, semantics, behavior prediction, imitation learning, ...
- **Planning & control:**  
e.g., path planning, high-level behavior, planning under uncertainty, control, ...
- ...

# Learning in Self-Driving Cars



[Courtesy: Ondruska, Lyft] 44

# End-to-End Approaches

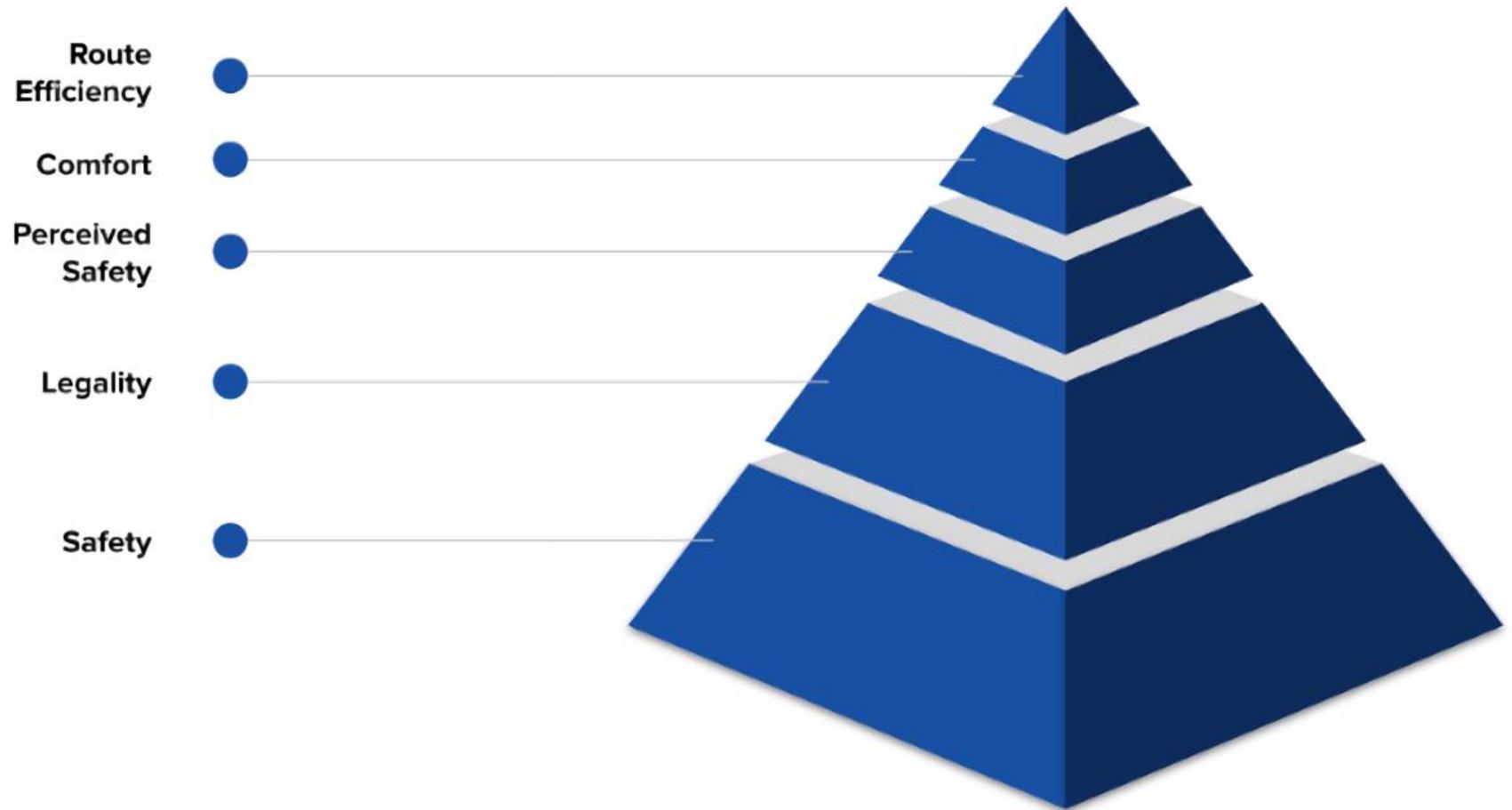
Directly learn mapping from camera data to steering commands



# First End-to-End Driving in London in Summer 2021



# What Is Most Important?



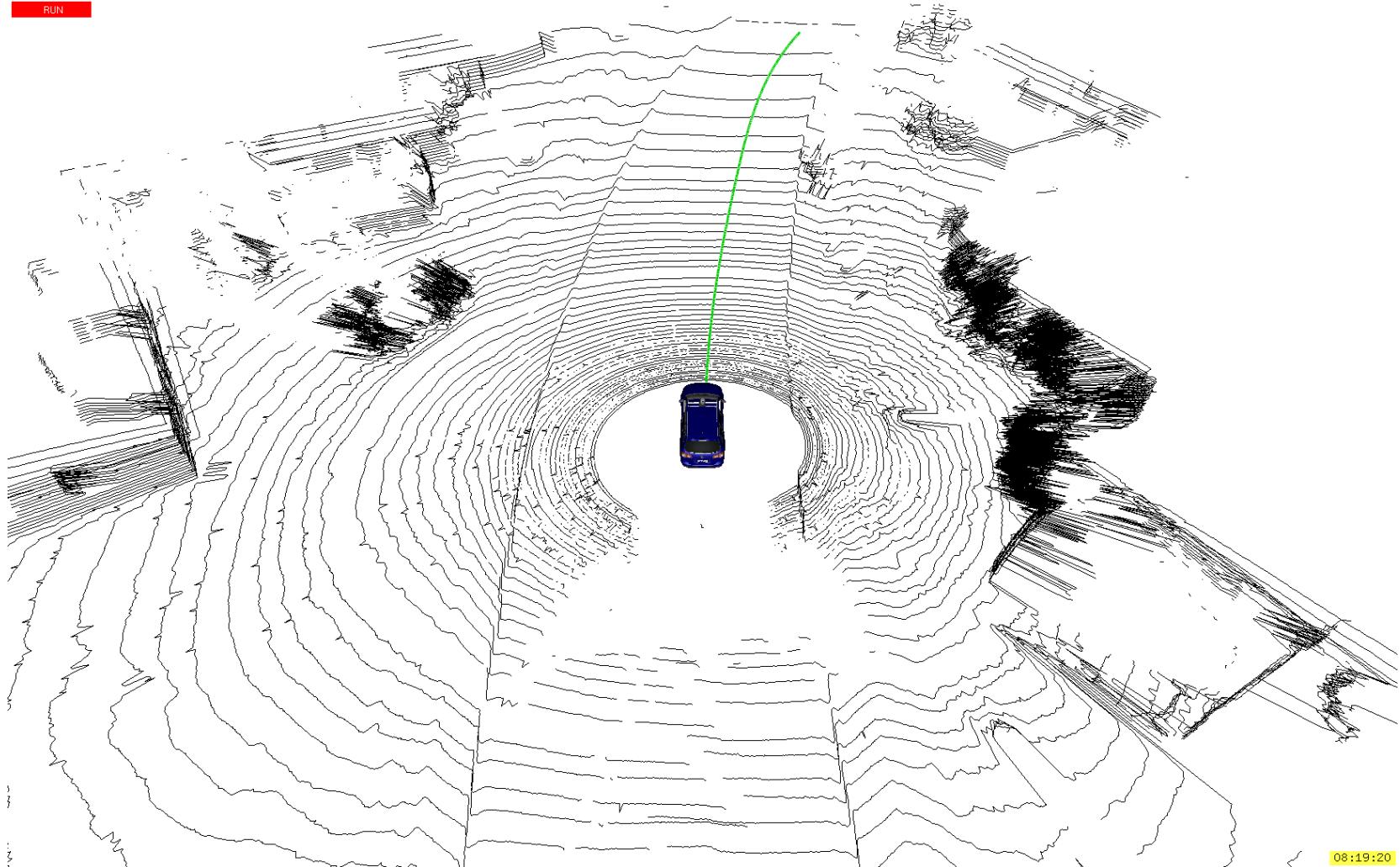
# **What Do Cars See?**

# Google Car (~2009)



[Courtesy: Google] 49

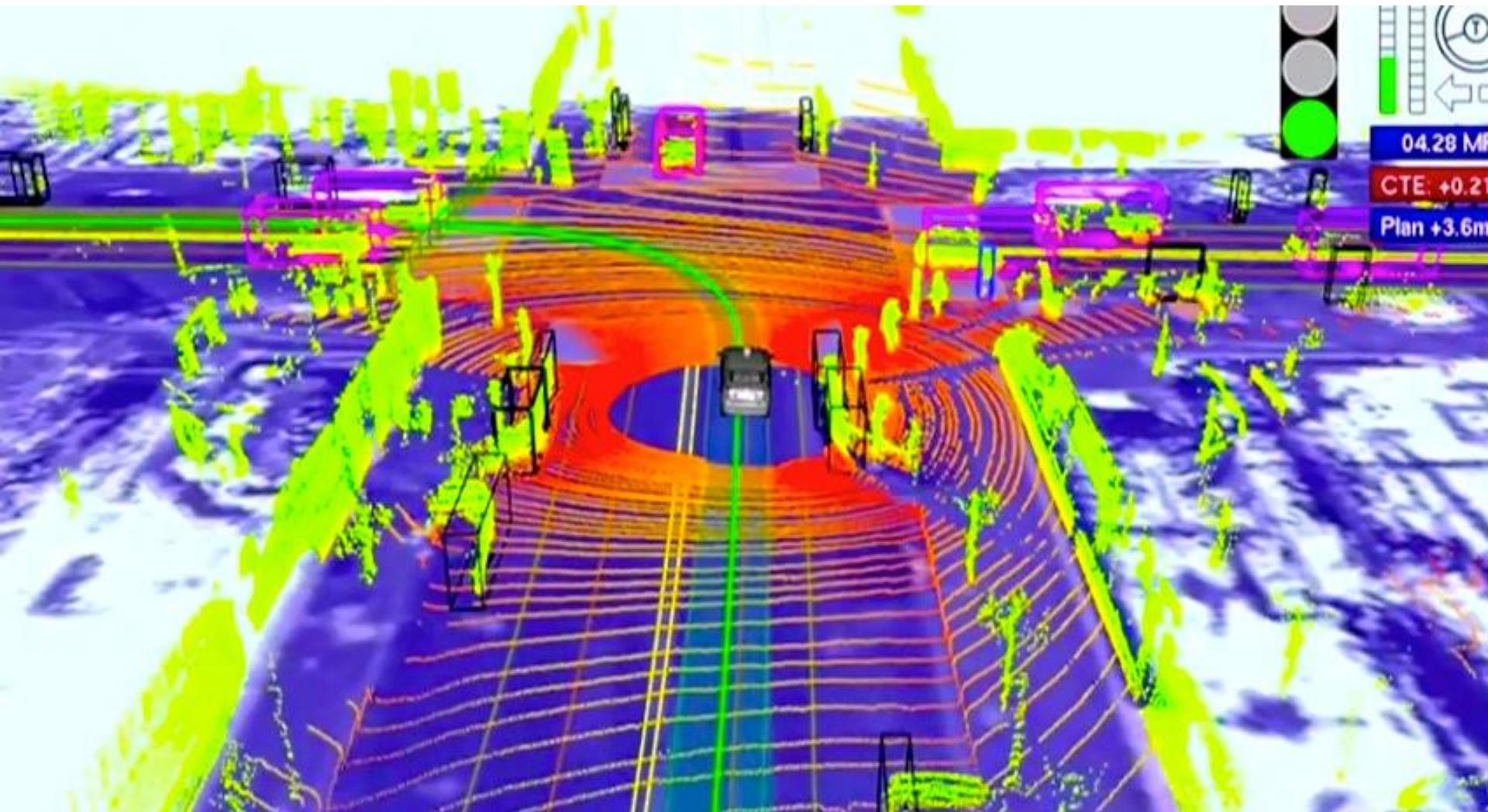
# What Did the Car See? (~2009)



08:19:20

[Courtesy: Google] 50

# What Did the Car See? (~2009)



[Courtesy: Google] 51

# First Public Demonstration



[Courtesy: Google] 52

# Driving in Urban Environments



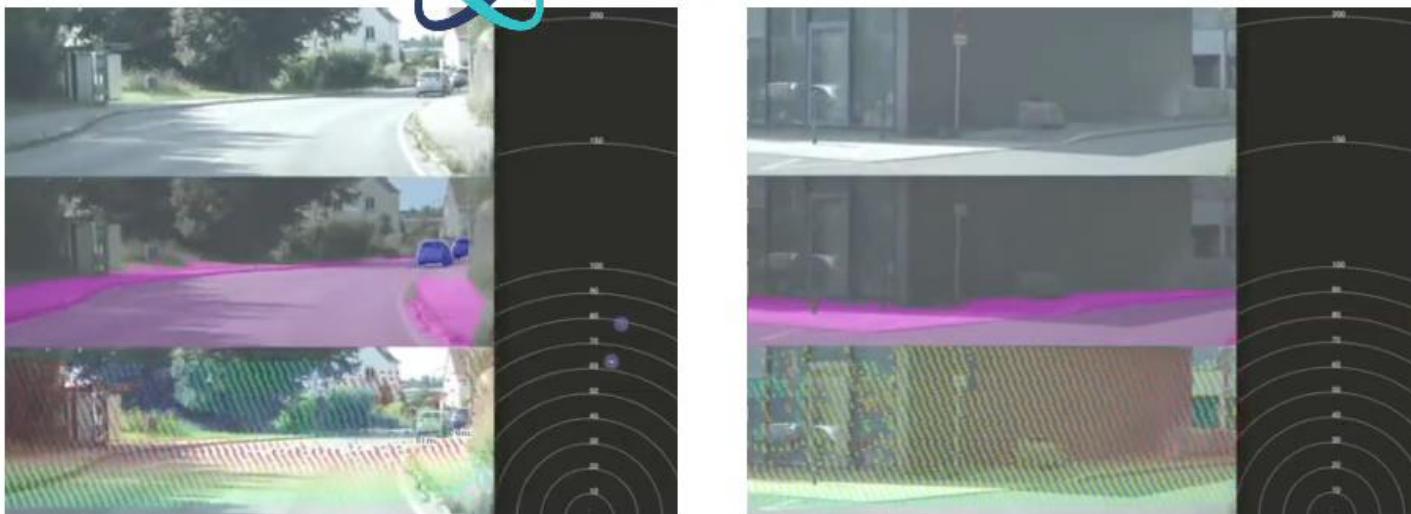
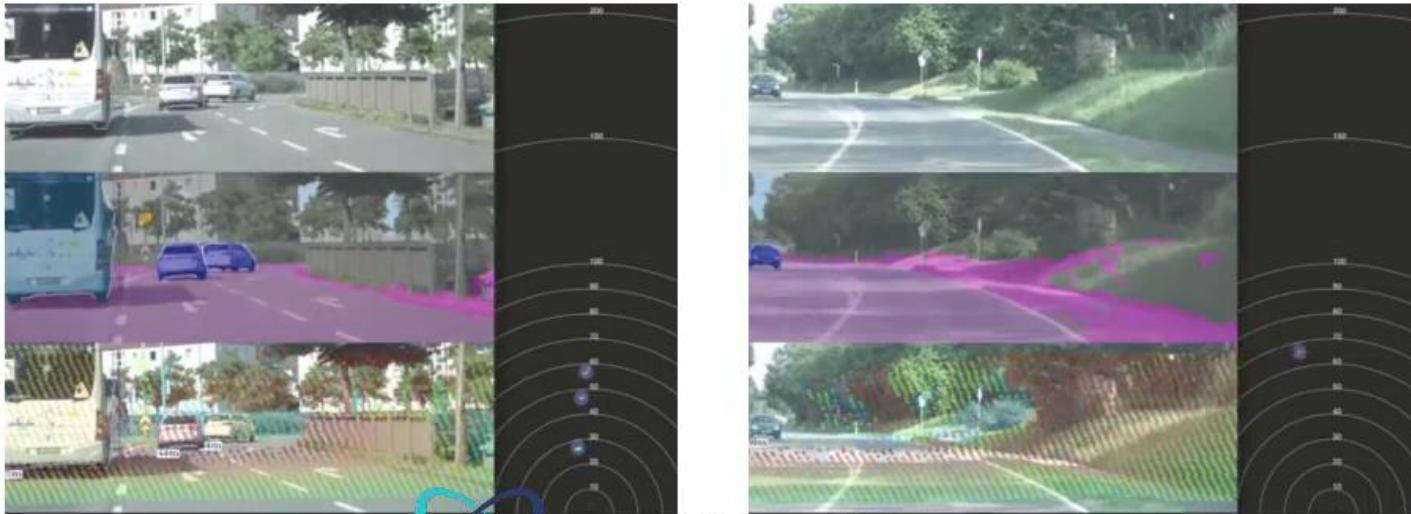
[Courtesy: Oxbotica P. Newman/B. Upcroft] 53

# Driving in Urban Environments



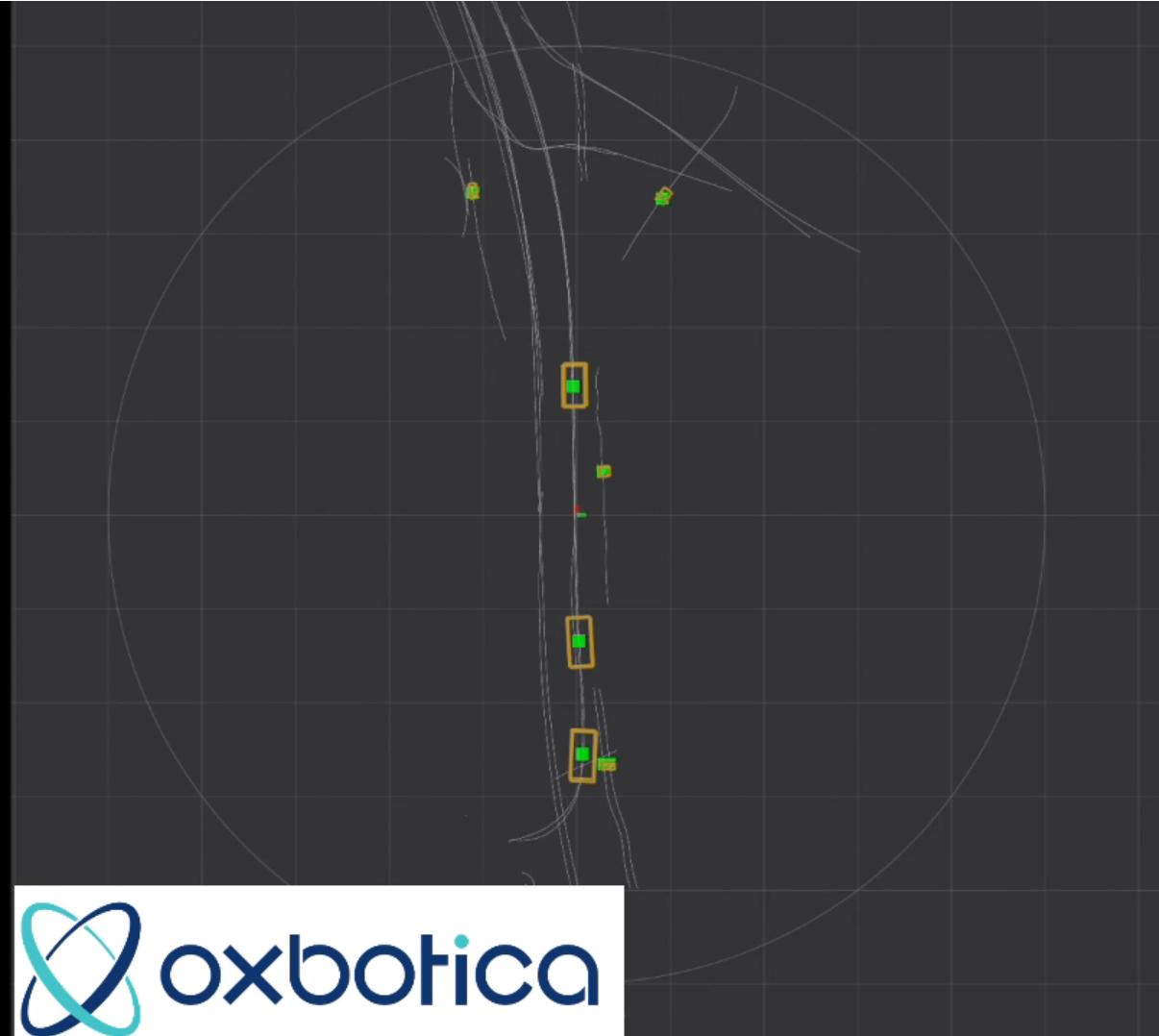
[Courtesy: Oxbotica P. Newman/B. Upcroft] 54

# Learning-based Perception



[Courtesy: Oxbotica P. Newman/B. Upcroft] 55

# Estimating What Happens



[Courtesy: Oxbotica P. Newman/B. Upcroft] 56

# Current State (Waymo, ~2019)

# **How to Test and Evaluate?**

# How to Test the Developments

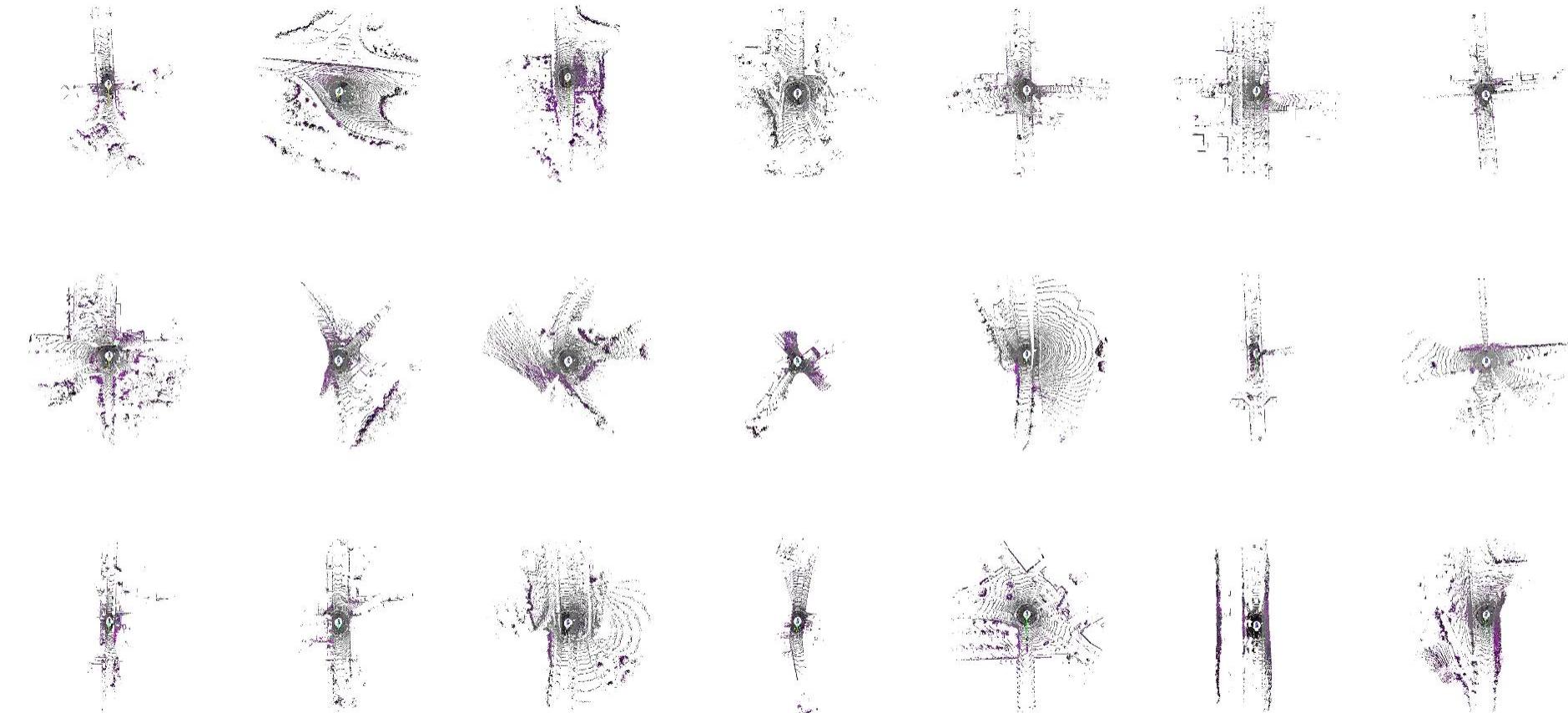
- Real testing is slow and expensive
- How to test at large scale?
- How to evaluate different approaches objectively?
- How to test behaviors in dangerous situations?
- What happens after changes in the code base? Run all real tests again?
- How to get ground truth data?

# Datasets

- Datasets recorded from vehicles are a key building block in development
- More and more datasets have been released within the past 2 years
- More vehicles, more variations, more locations, more drivers, ...



# Example: LiDAR SLAM on KITTI

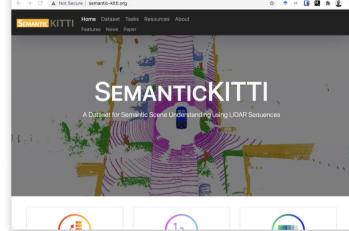


# KITTI Dataset 2012

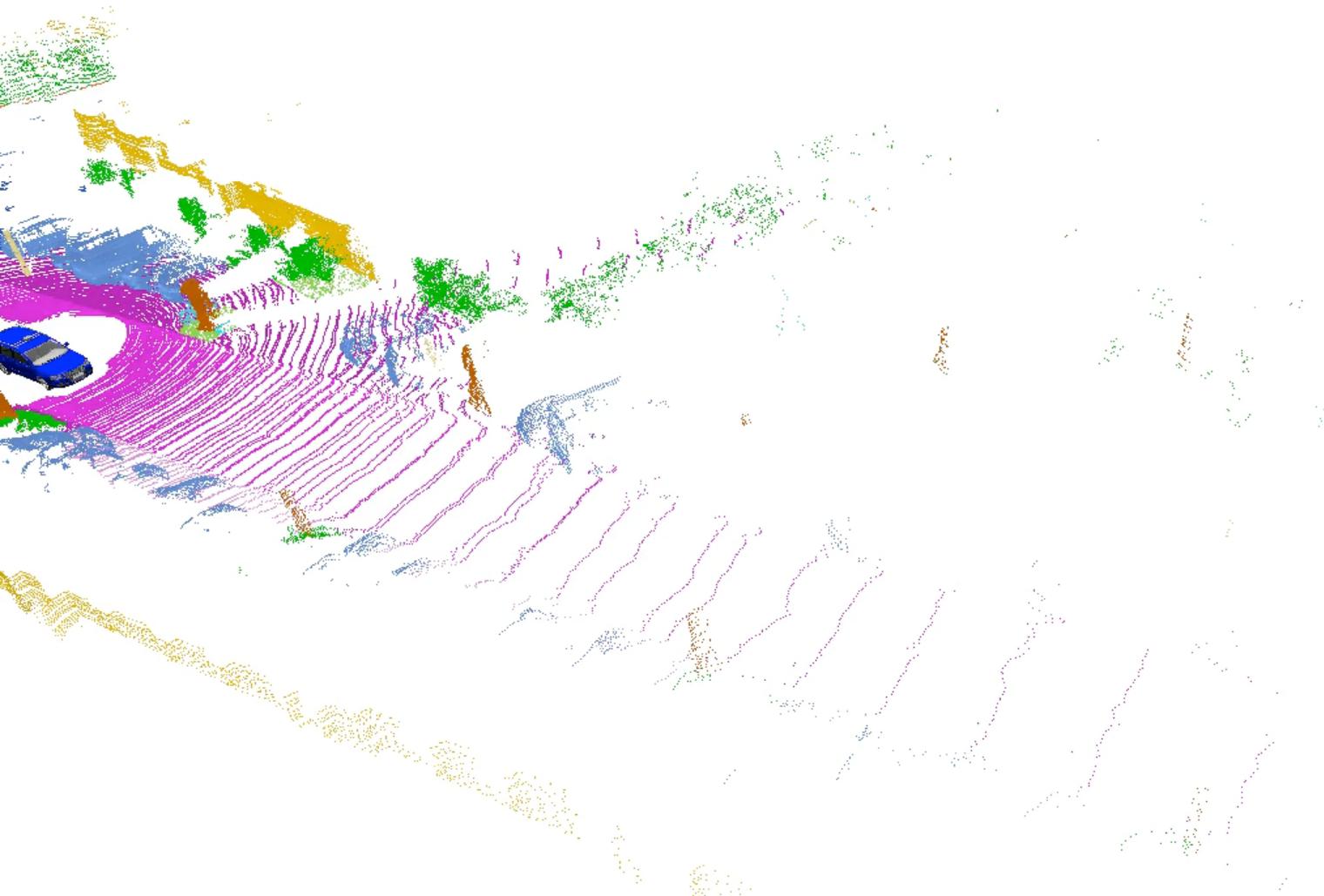
- Are we ready for Autonomous Driving?  
The KITTI Vision Benchmark Suite
- Heavily used for state estimation tasks  
(VO, SLAM, depth from vision,...)
- Comes with own benchmarks



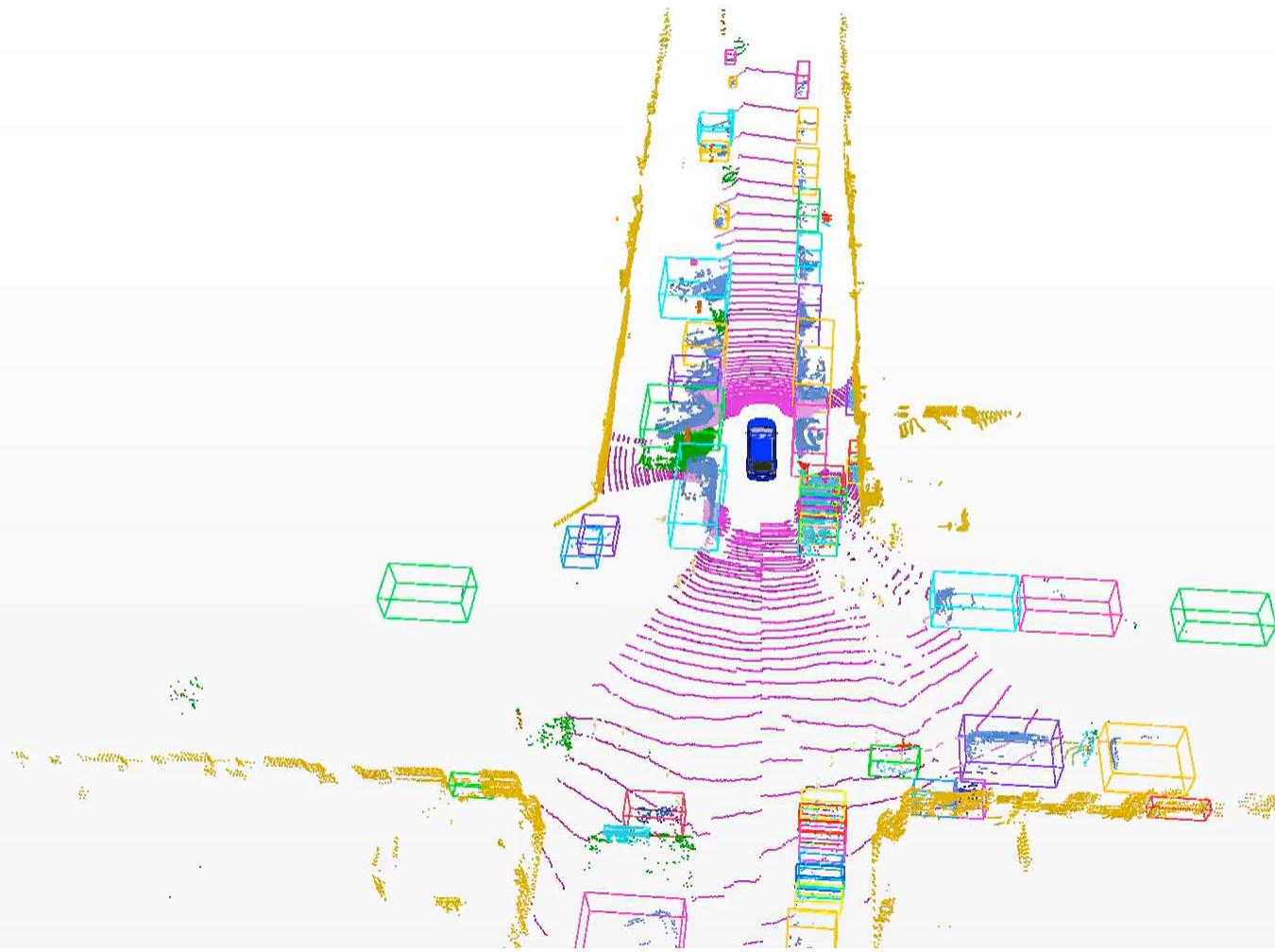
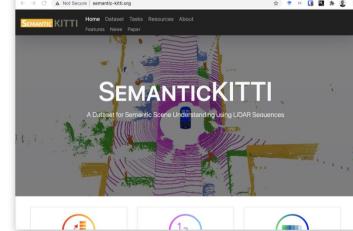
# KITTI & Semantic KITTI



"A semantic label for every 3D point."



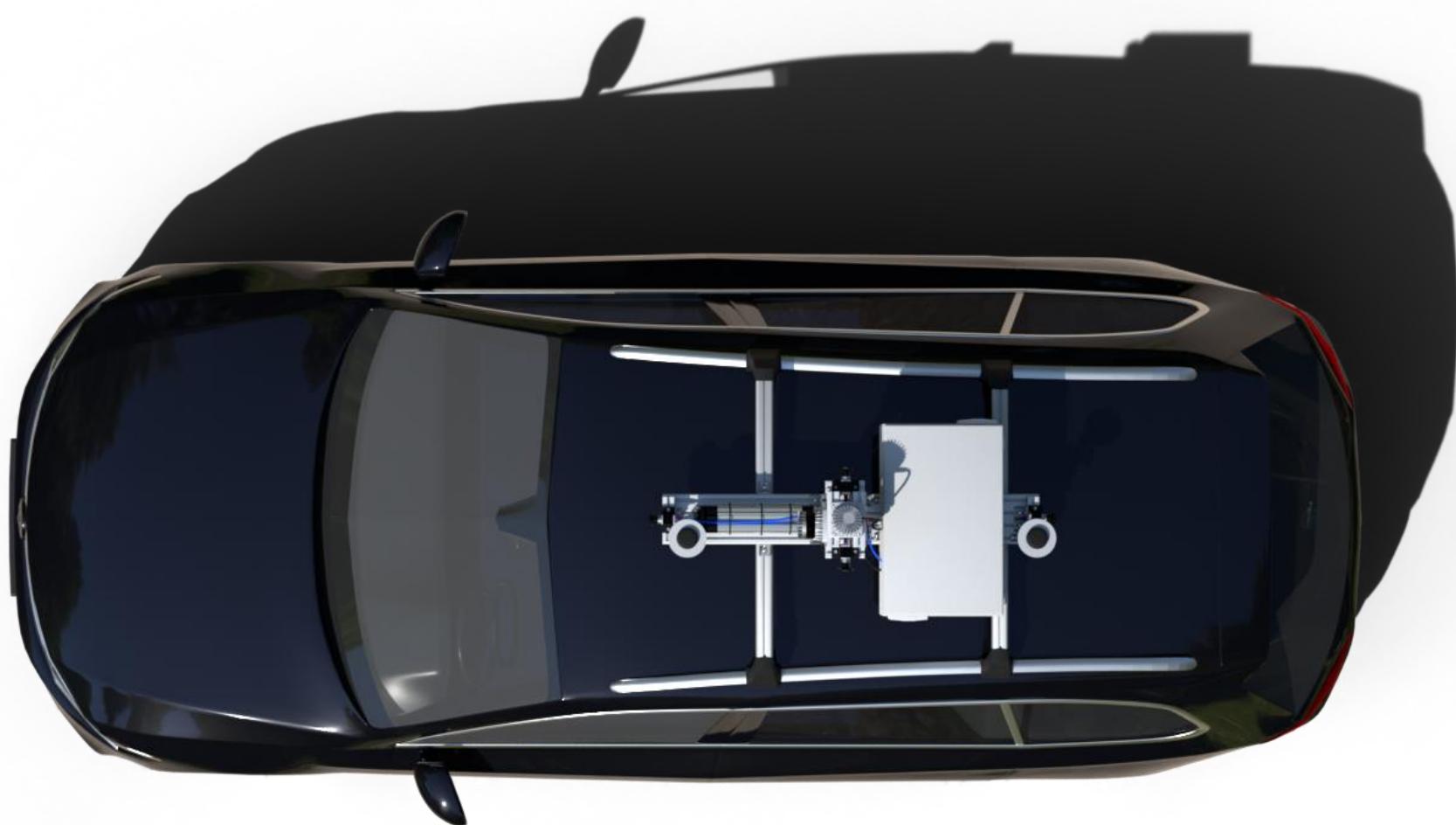
# Semantic KITTI Instances



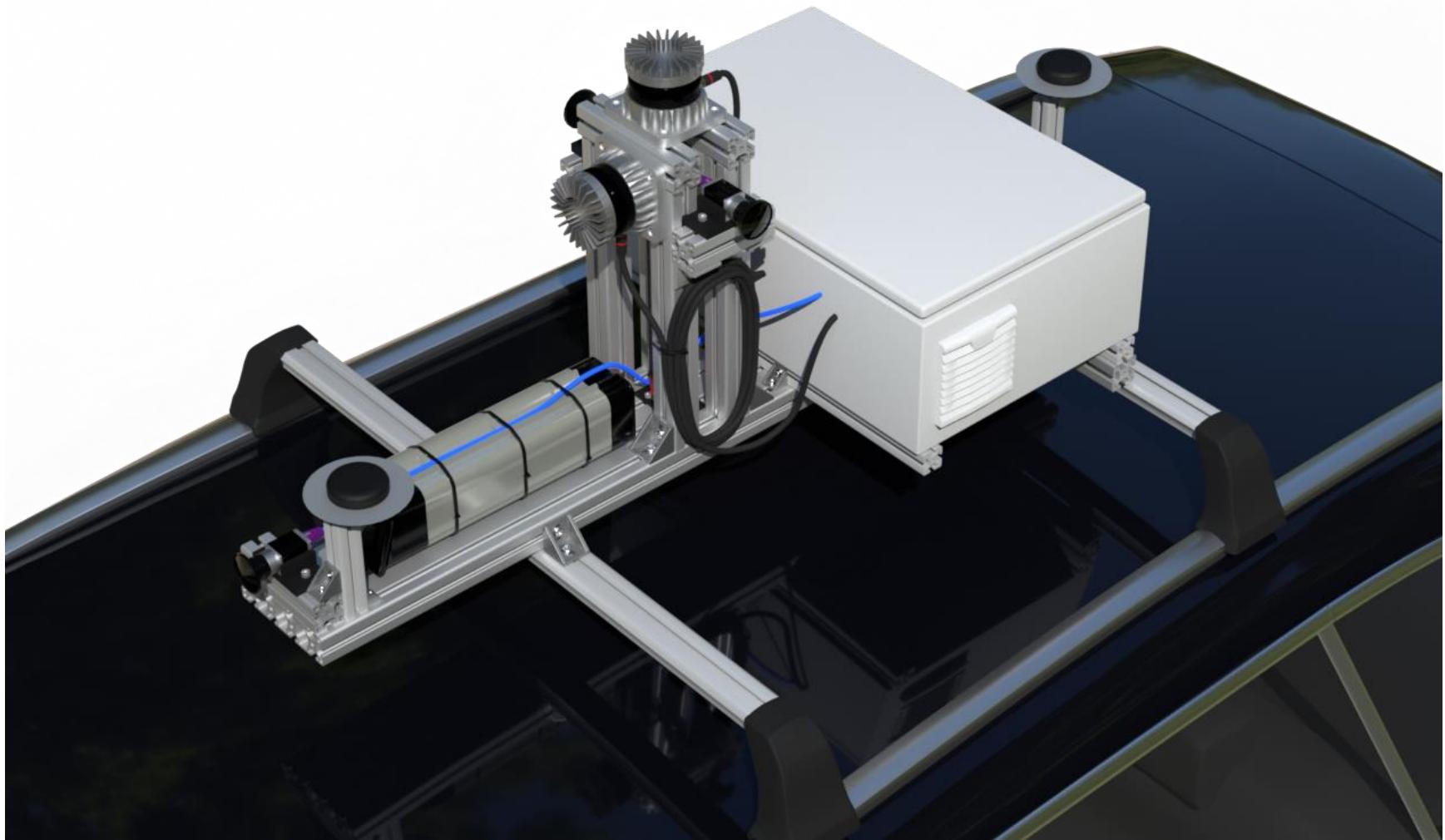
# IPB Car



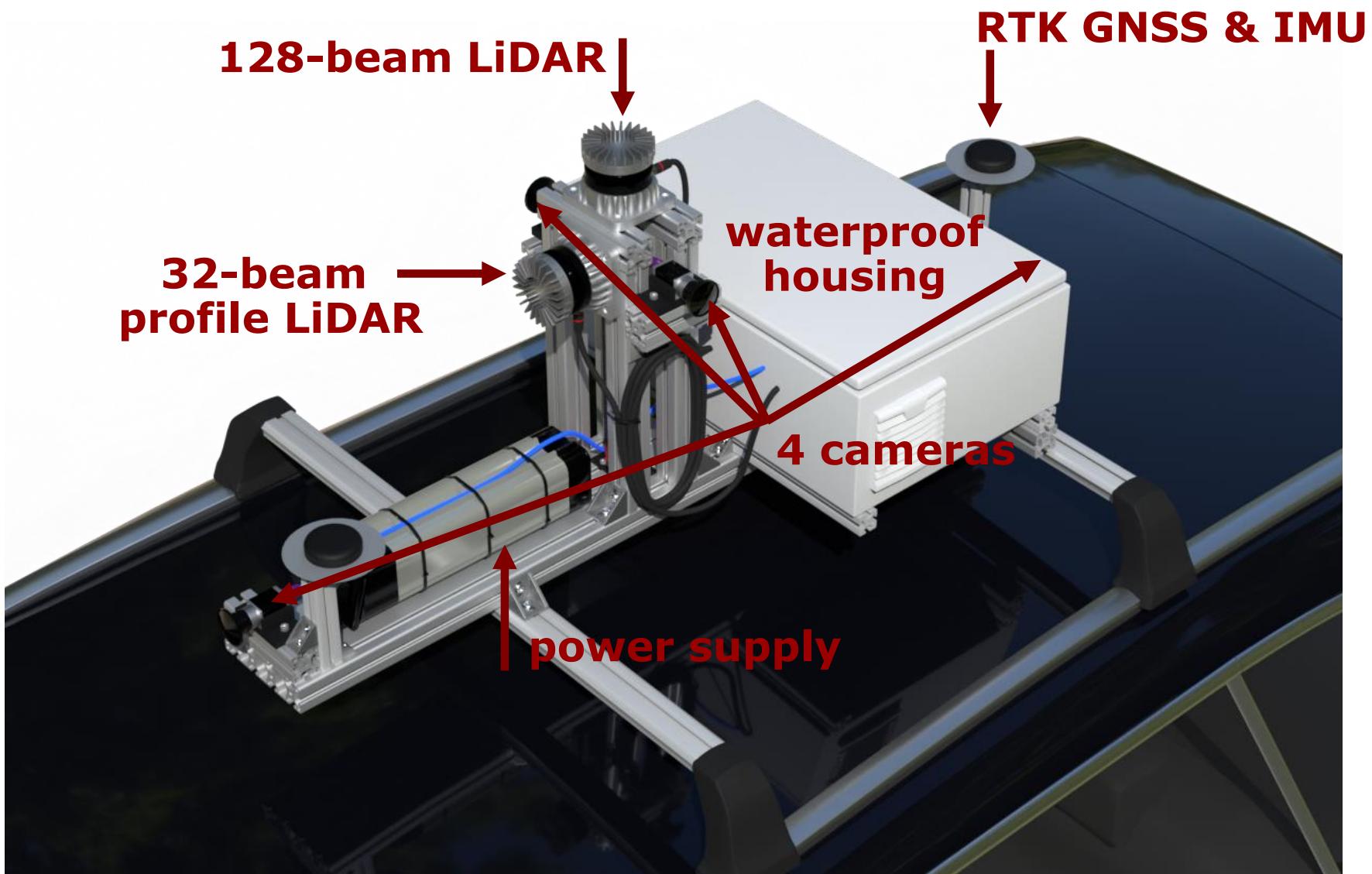
# IPB Car



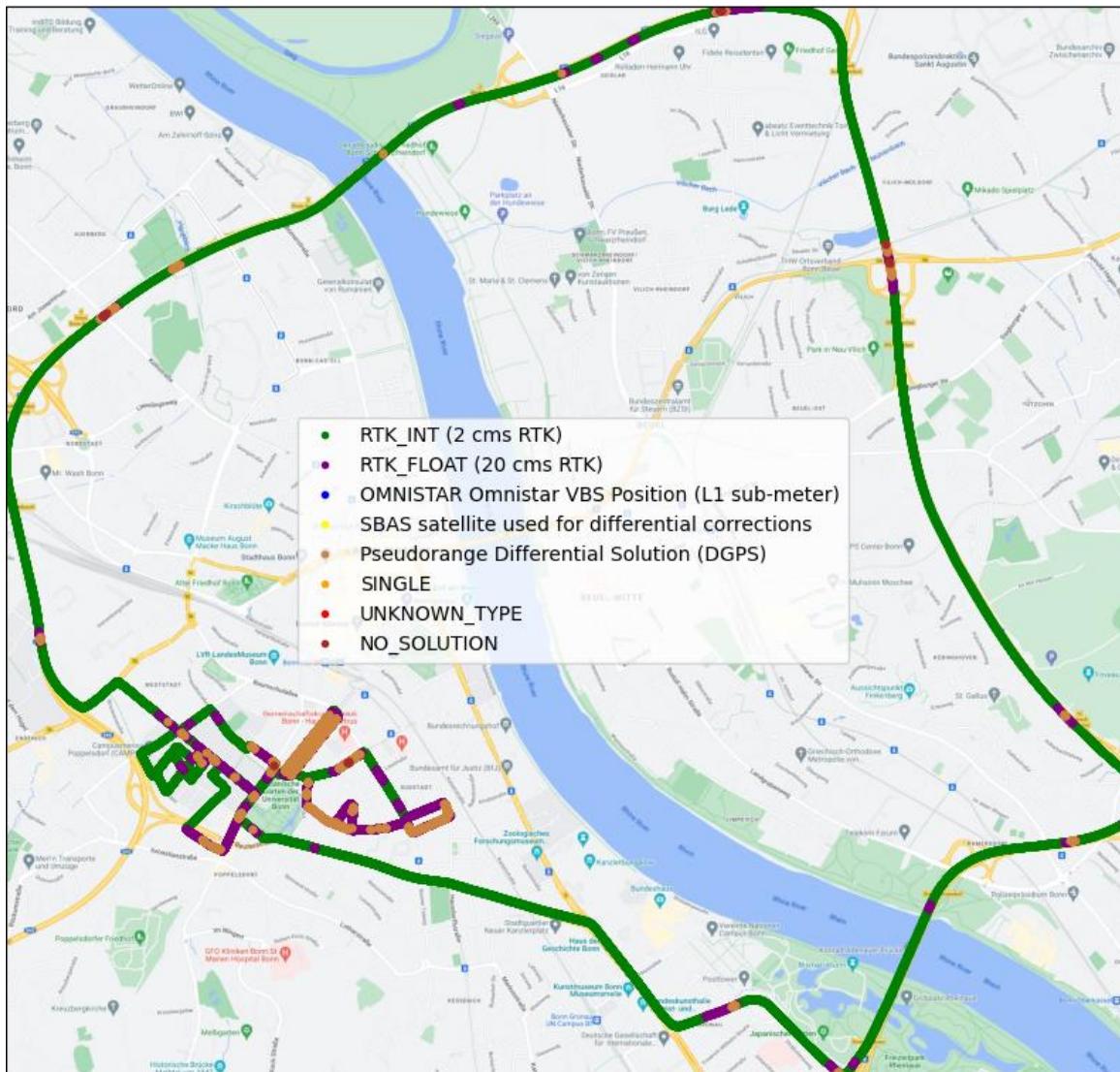
# IPB Car Sensor Setup



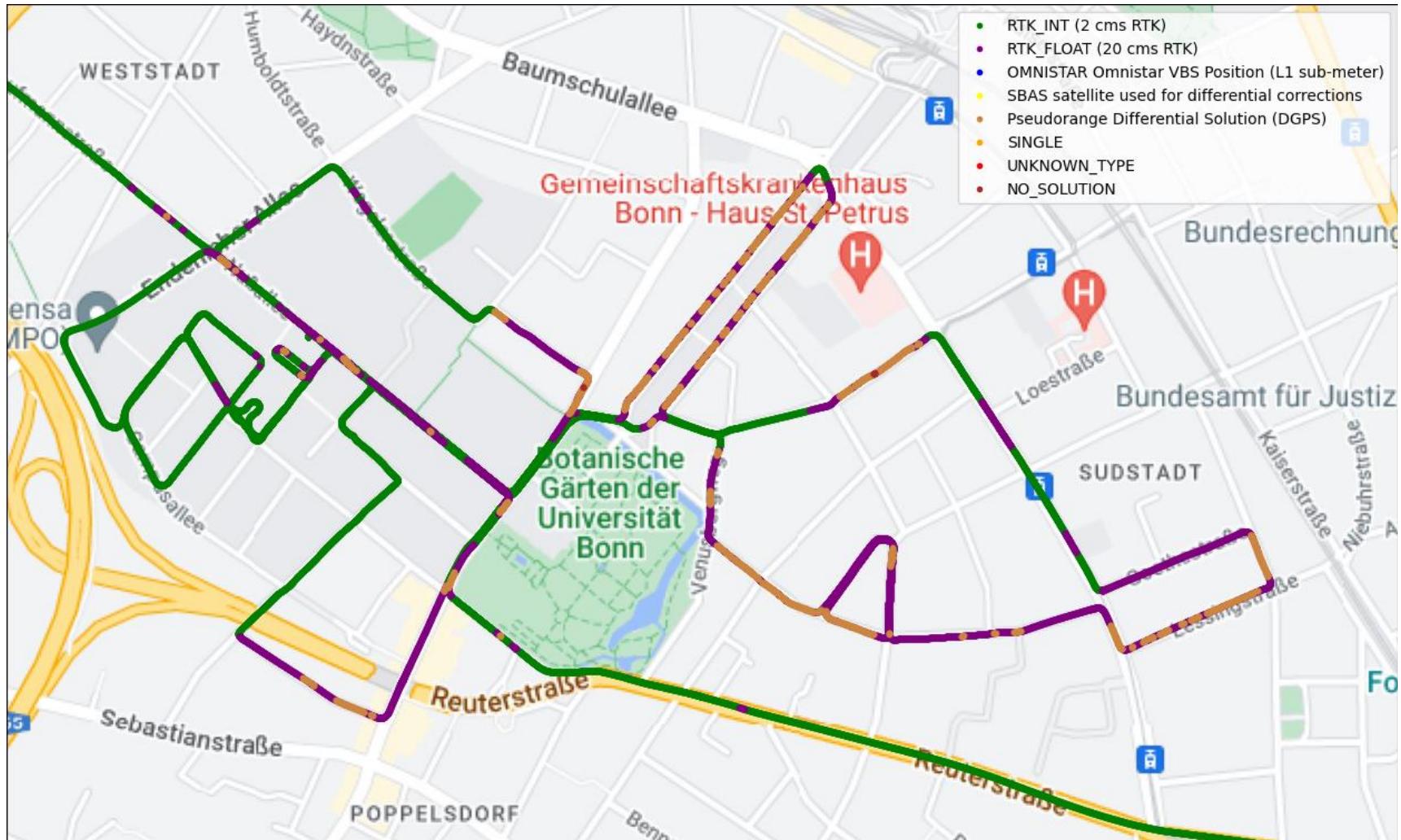
# IPB Car Sensor Setup



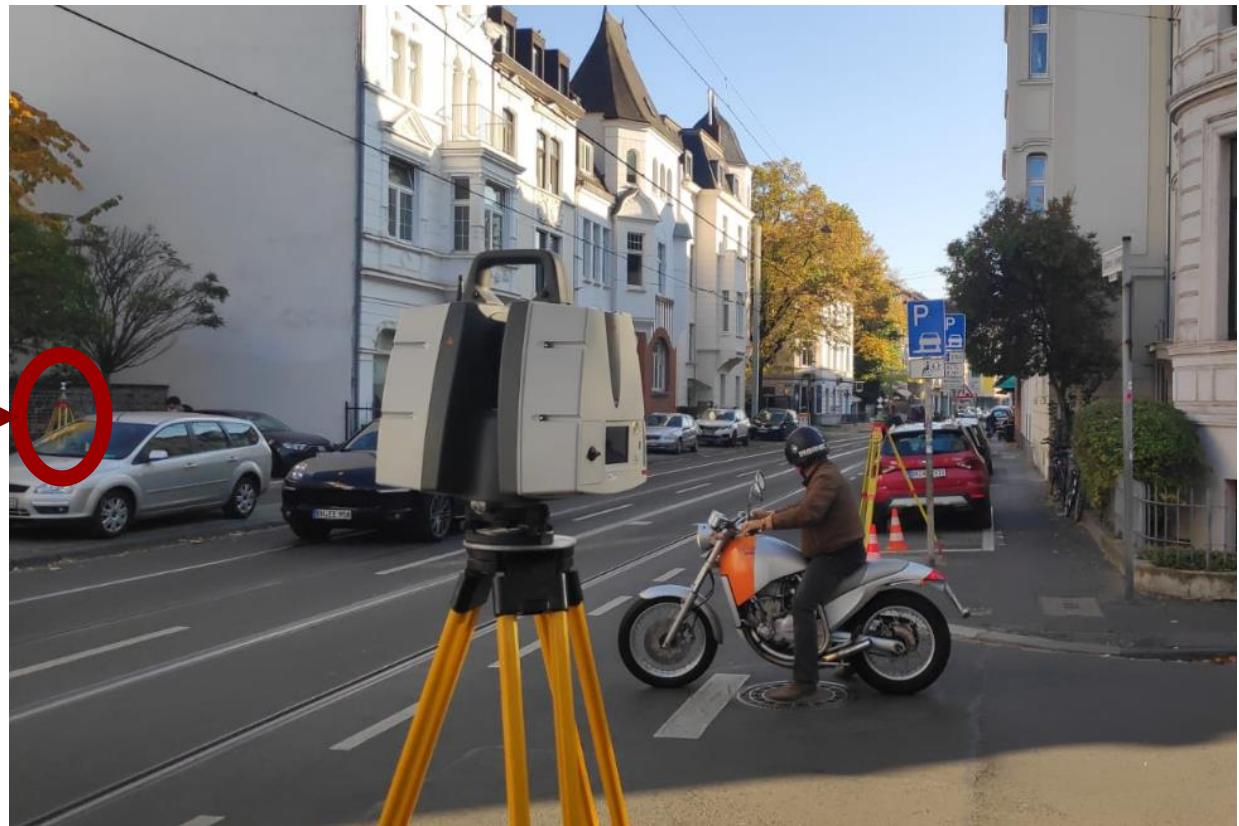
# Data Recorded in Bonn



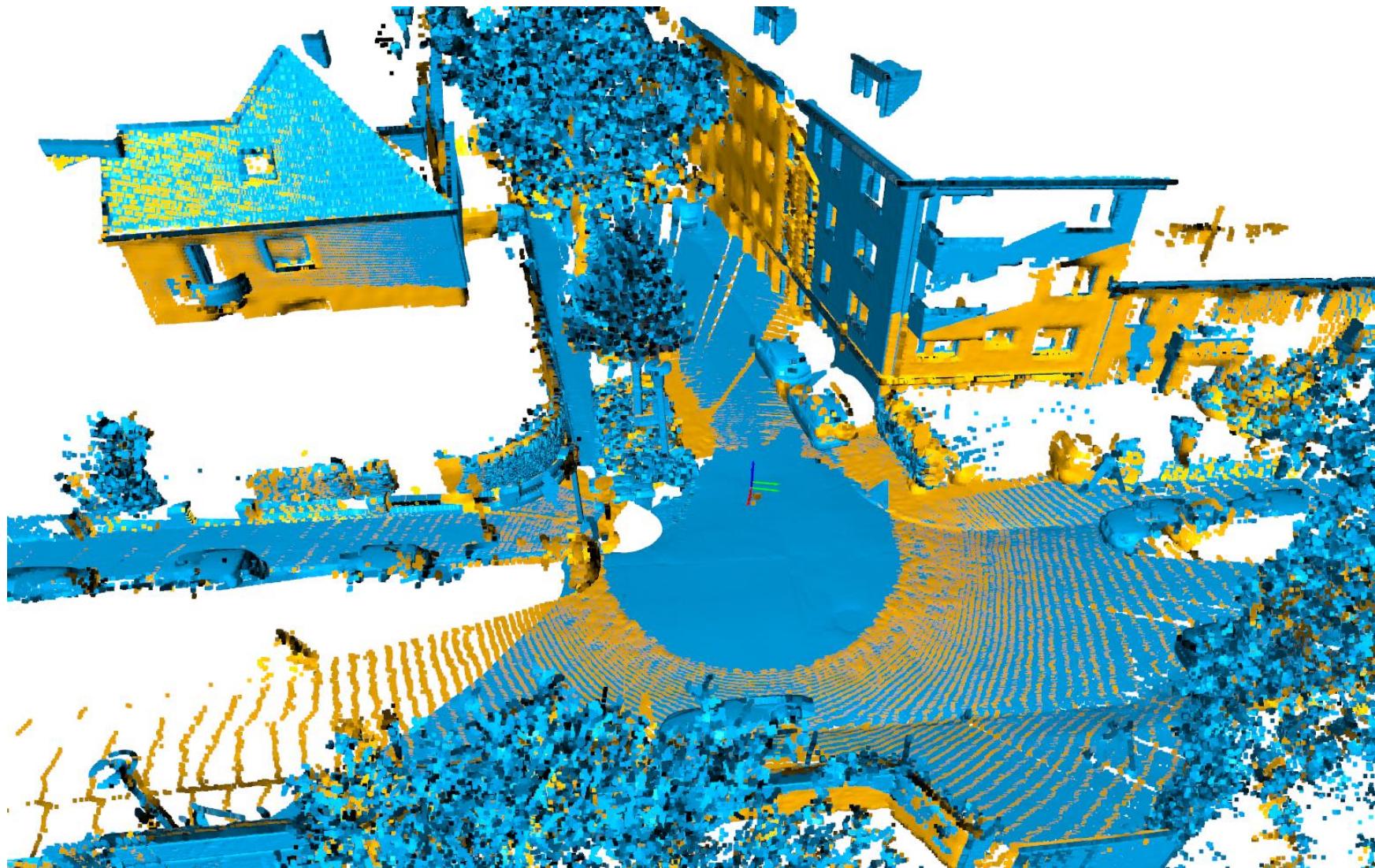
# Data Recorded in Bonn



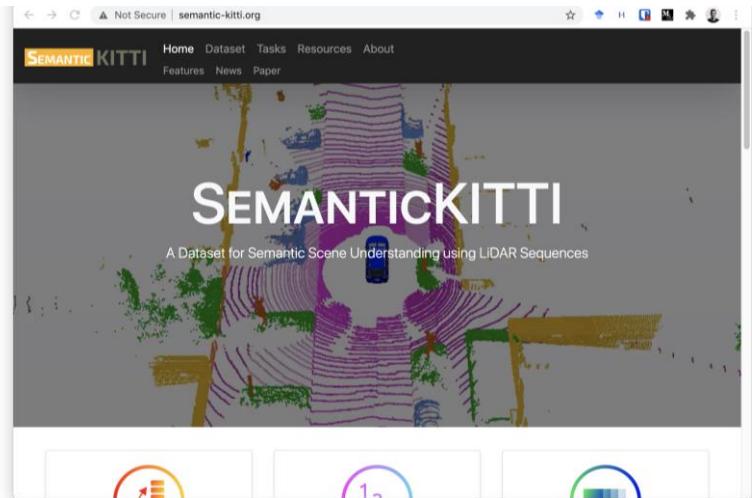
# Reference Data



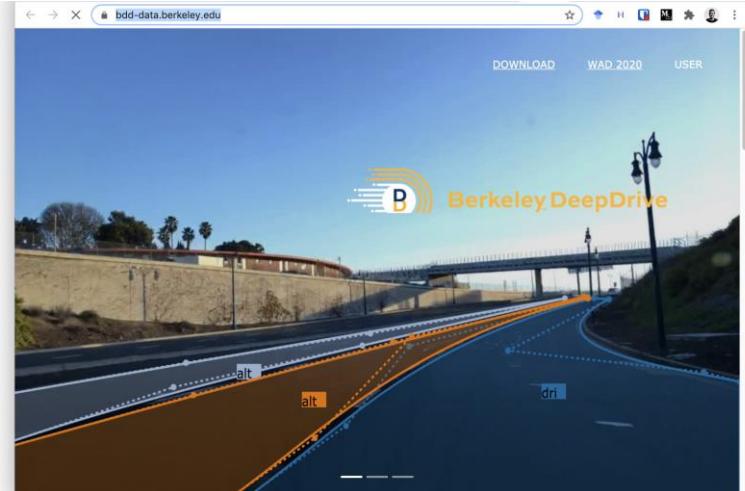
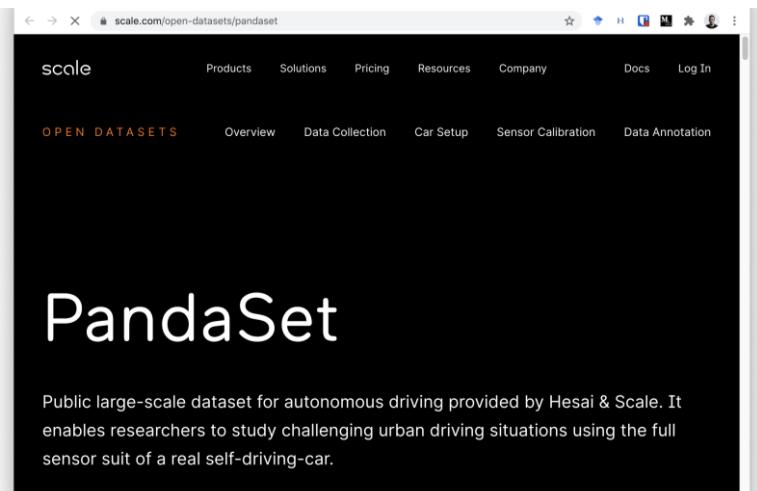
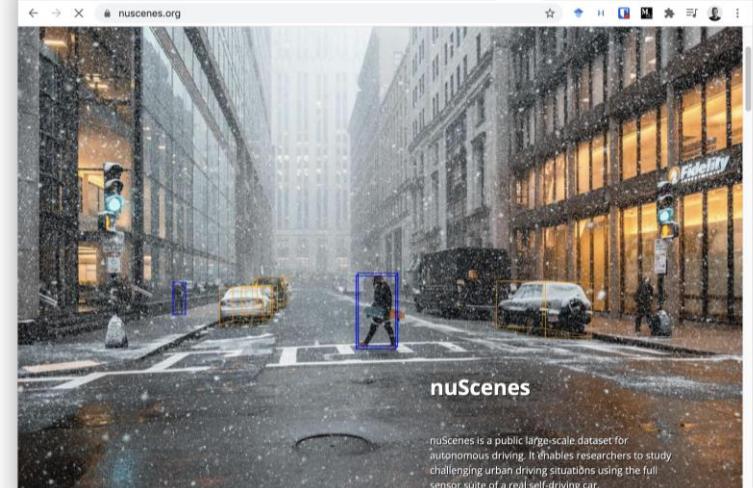
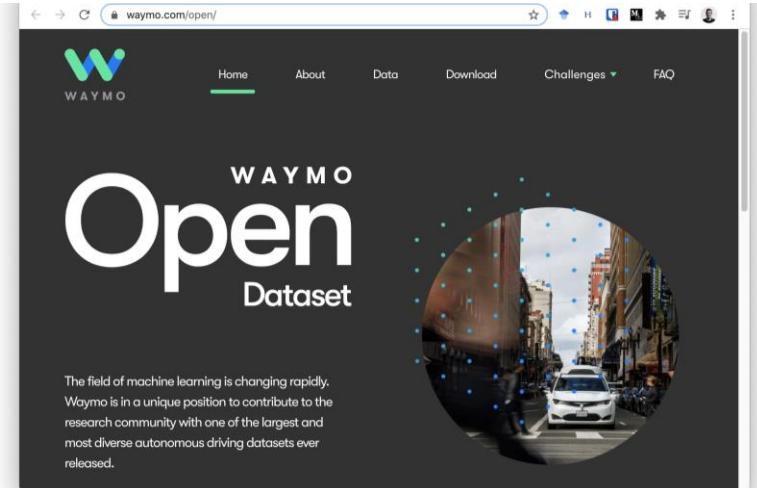
# Reference Data



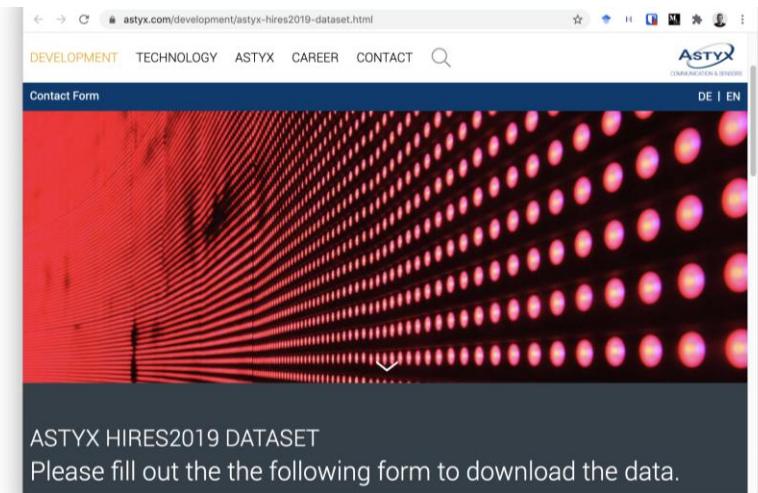
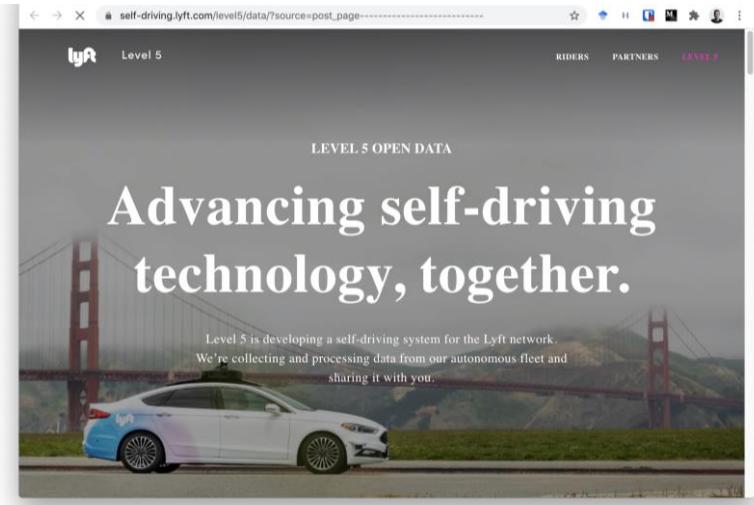
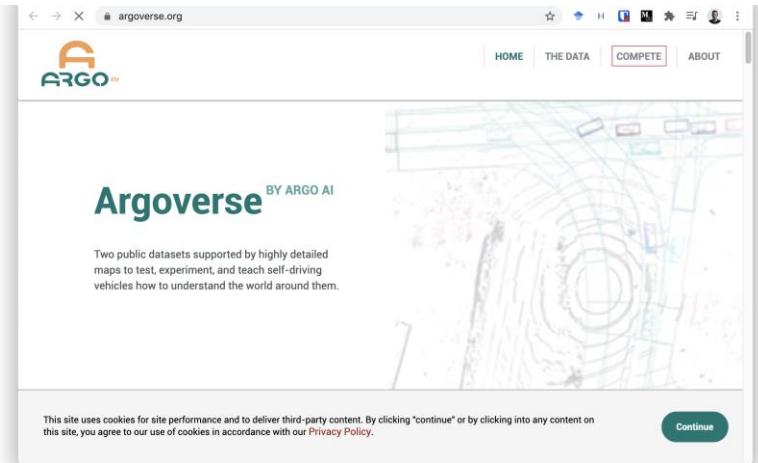
# Popular Datasets (1)



# Popular Datasets (2)



# Popular Datasets (3)



# Links to Popular Datasets

- <http://www.cvlibs.net/datasets/kitti/>
- <http://semantic-kitti.org/>
- <https://waymo.com/open/>
- <https://www.argoverse.org/>
- <https://www.nuscenes.org/>
- <https://bdd-data.berkeley.edu/>
- <https://self-driving.lyft.com/level5/data/>
- <https://robotcar-dataset.robots.ox.ac.uk/>
- <https://scale.com/open-datasets/pandaset>
- <https://www.highd-dataset.com/>
- <https://www.astyx.com/development/astyx-hires2019-dataset.html>

# Datasets

- Real-world data
- Used to train learning models
- Regression testing

## But:

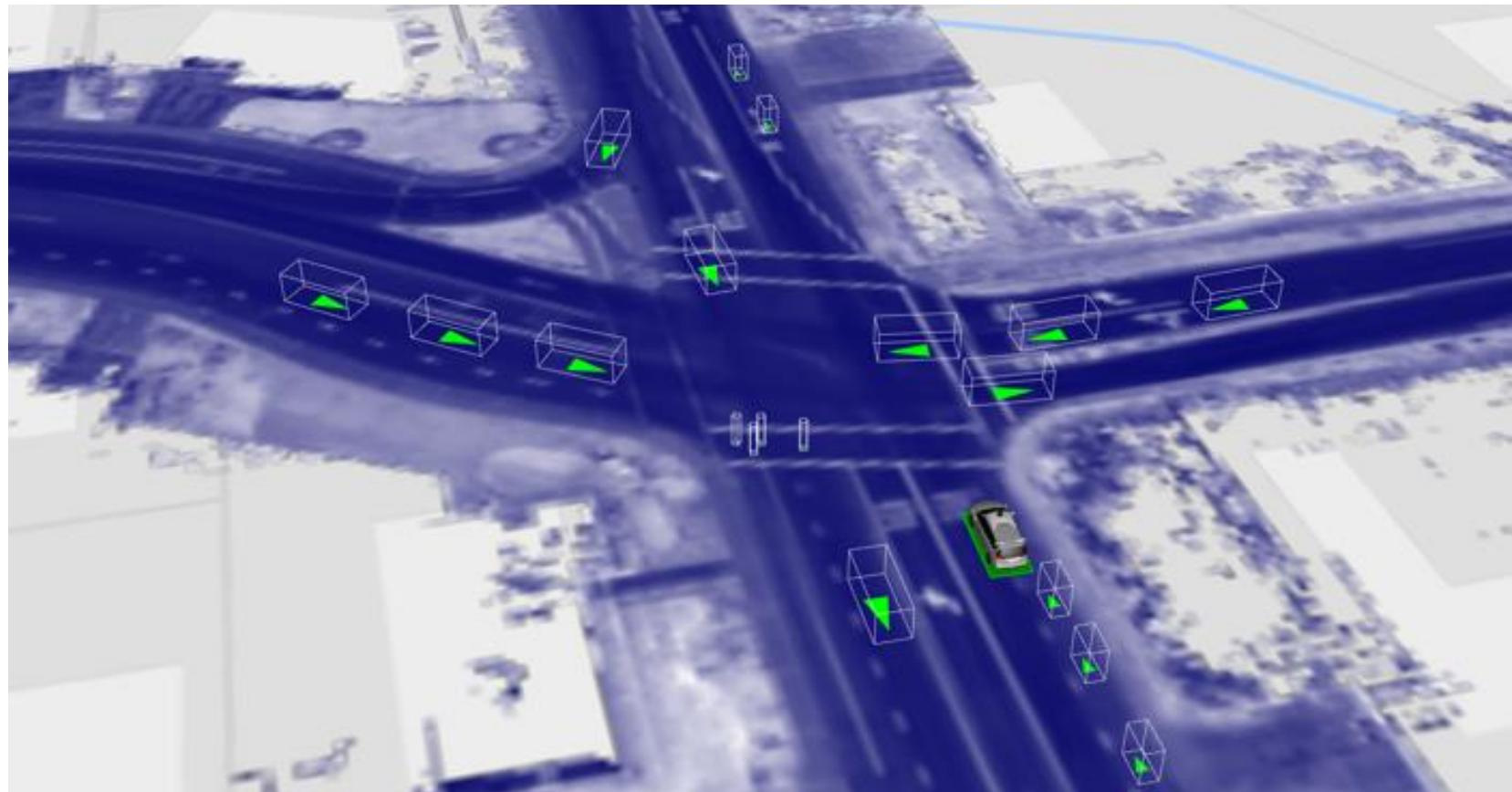
- Datasets are passive: How to test a planning algorithm on a dataset?
- Limited view of the world
- What if the sensor setup changes?

# Simulations Play a Key Role

- Simulators became a critical infrastructure for autonomous driving
- Dmitri Dolgov (Waymo CTO):  
“We’ve driven more than 10 million miles in the real world, and over 10 billion miles in simulation.”

# Waymo's Carcraft Simulator

25,000 virtual cars collectively drive about 10 million miles every day (2019)



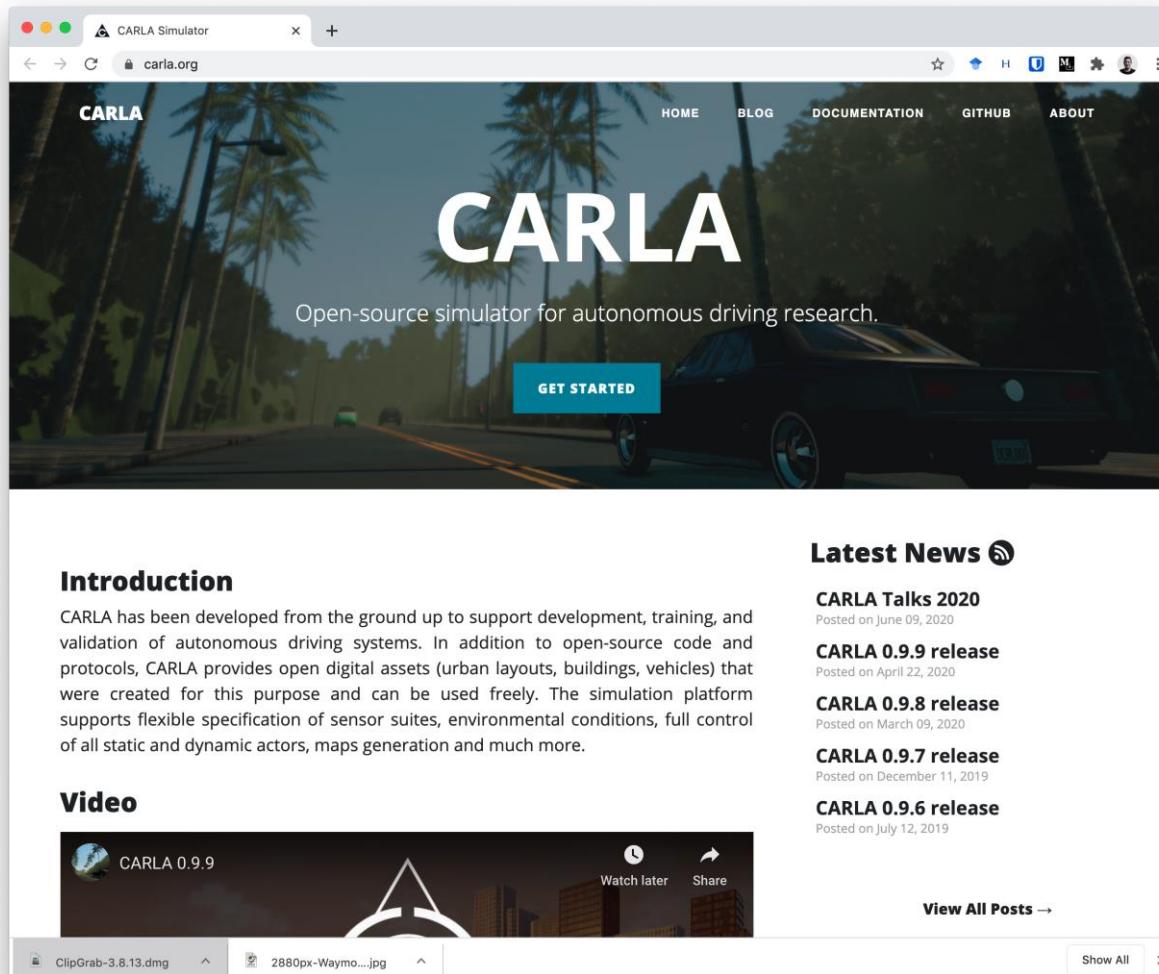
# Great Things About Simulation

- Allow for automated and parallel testing (e.g., test after every commit)
- Can generate identical situations over and over again
- Allow for focusing on tricky situations
- Allow for testing dangerous situations
- Can provide ground truth data for learning algorithms

# Issues with Simulations

- They are only a abstracted model of the real world
- Realistic perceptions/noise/outliers?
- Physics simulation vs. “looks good”?
- Do other traffic participants behave as they would in the real world?
- Are rare cases simulated appropriately?
- Has everything been considered?

# CARLA Simulator

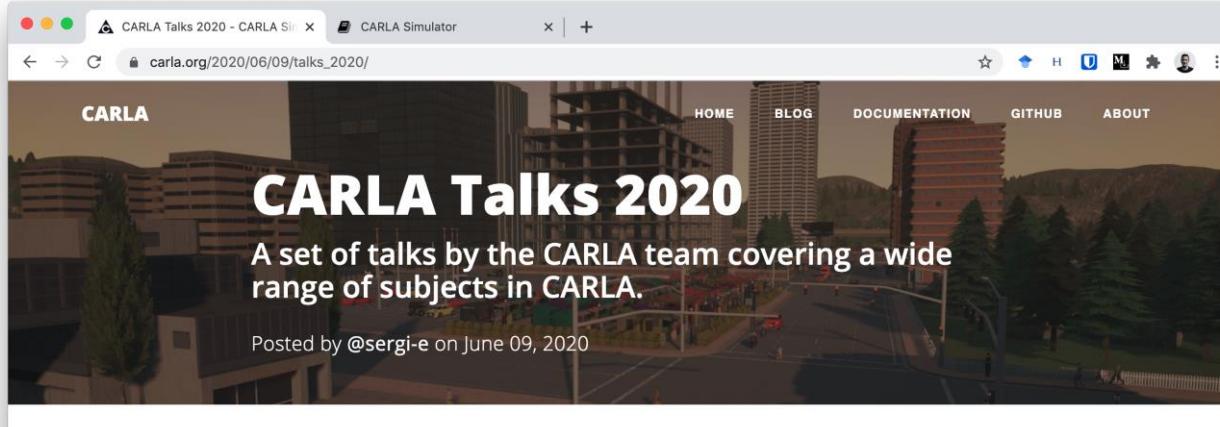


<http://carla.org>

# CARLA Simulator

- Use own worlds
- Autonomous driving sensor suite
- Simulation for planning and control
- Traffic scenarios simulation
- Autonomous Driving baselines
- ROS integration
- Flexible API
- Server – multi-client architecture

# CARLA Video Talks



The screenshot shows a web browser window for 'CARLA Talks 2020 - CARLA Simulator'. The URL in the address bar is 'carla.org/2020/06/09/talks\_2020/'. The page features a large banner image of a CARLA simulation showing a city street with buildings and trees. Overlaid on the banner is the text 'CARLA Talks 2020' in large white letters, followed by a subtitle 'A set of talks by the CARLA team covering a wide range of subjects in CARLA.' Below the banner, it says 'Posted by @sergi-e on June 09, 2020'. The main content area starts with a short paragraph: 'The CARLA team is here, and it has a lot to say!'. It then describes the purpose of the talks: 'In order to make it easier to keep up with all the changes in CARLA, the team has prepared a set of talks which cover a wide range of subjects. The talks have been uploaded to YouTube, and the slides are available to everybody for download.' Below this, there's a section titled 'General' with a bulleted list of topics:

- **Art improvements: environment and rendering** — video | slides  
Learn more about the evolution of graphics in CARLA, the process to achieve this, and the new parametrization of weather and lights.
- **Core implementations: synchrony, snapshots and landmarks** — video | slides  
Review of some important core elements of the API: the synchronous mode, simulation snapshots and everything about the new landmarks to access OpenDRIVE signals.
- **Data ingestion** — video | slides  
Follow an in-depth explanation of the automatic ingestion process for maps, and other assets, depending on how are you running CARLA (from a package or a build from source).
- **Pedestrians and their implementation** — video | slides  
Grasp everything regarding pedestrians, from navigation to animation, and the great upgrade to come.
- **Sensors in CARLA** — video | slides

[https://carla.org/2020/06/09/talks\\_2020/](https://carla.org/2020/06/09/talks_2020/)

# Summary

- Self-driving cars have a high potential to change the transportation industry
- Overview of self-driving car ecosystem
- Cars require solutions to several robotics and perception problems
- Key tasks:
  - Perceive and estimate
  - Predict
  - Plan & act

# People Who Contributed to the Development of the Course



**Thank you for your attention**