



# SMASH

machine learning for science and humanities postdoctoral program

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MINISTRY OF THE ENVIRONMENT, CLIMATE AND ENERGY  
SLOVENIAN ENVIRONMENT AGENCY

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AI usage is for demonstrational purpose only and does not reflect reality. The image in the title slide is generated with AI



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# AUTONOMOUS DRIVING IN COMPLEX ENVIRONMENTS

**1. 3D PERCEPTION**  
I see everything...  
(even your bad lane change)

**PEDESTRIAN AWARENESS**  
People will do anything.  
We're prepared.

SO MANY VARIABLES...  
SO LITTLE PATIENCE  
(HUMANS!)

आई.टी.ओ  
I.T.O  
आनंद विहार  
Anand Vihar  
गाजियाबाद  
Ghaziabad

IMPORTANT  
CALL...

CROSSING IS  
MY BIRTH RIGHT!

BHAI SAAB,  
THODA SIDE!

METER  
CHALU HAI!

RELAX, BOTS  
I GOT THIS

CHALLENGE  
ACCEPTED!

MOOVING  
AT MY OWN  
PACE!

**SENSOR FUSION**  
LiDAR, camera, radar...  
teaming up like  
Avengers!

**TRAFFIC  
UNDERSTANDING**  
Rules are guidelines.  
Context is everything.

DESTINATION:  
HOME SAFE,  
EGO INTACT!

**SAFE DECISIONS**  
Brake? Wait? Go?  
I choose safety.  
(Always.)

INDICATOR?  
OPTIONAL!









**SUBURBAN  
ROADWAYS**



**URBAN  
ROADWAYS**



**SUBWAYS /  
METROS**



**MOTORWAYS /  
HIGHWAYS**



**LOW-SPEED  
RAILWAYS**



**HIGH SPEED  
RAILWAYS**

# Data Sources & Methodology

Data is synthesized from 2023-2025 reporting periods across the following primary institutional databases:

- **Roadways & Motorways:**

- *International Road Federation (IRF)* World Road Statistics.
- *World Bank Open Data* (Logistics Performance Index & Infrastructure datasets).
- *OECD iLibrary* (Road transport infrastructure exposure).
- *World Health Organization (WHO)*: Global Status Report on Road Safety.
- *European Commission (2026)*: Preliminary Road Safety Statistics (detailing the 53% rural / 38% urban / 8% motorway fatality split).
- *National Highway Traffic Safety Administration (NHTSA)*: Fatality Analysis Reporting System (FARS).

- **Railways (Metro & HSR):**

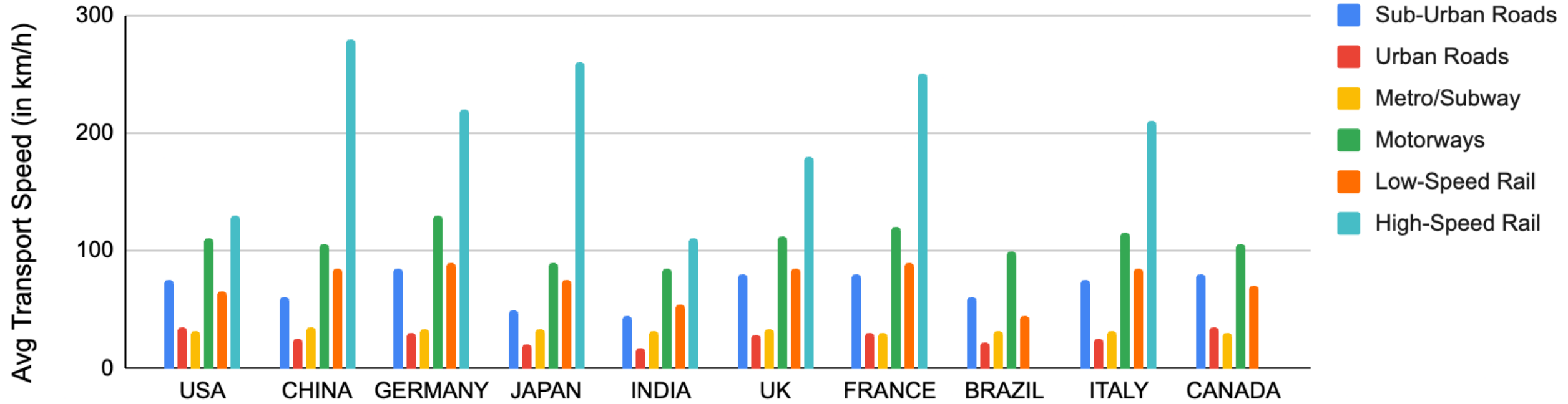
- *International Association of Public Transport (UITP)* Statistics Briefs on World Metro Figures.
- *International Union of Railways (UIC)* High-Speed Rail reports and Atlas.
- *Statista* Rail Industry reports (Passenger volume by country).
- *Eurostat*: Railway safety statistics (accidents to persons by rolling stock).
- *International Union of Railways (UIC)*: Safety Performance Reports.
- *Ministry of Railways (India)*: Annual Statistical Statements (reflecting high numbers due to level crossing and trespassing incidents).

- **Daily Usage & Intensity:**

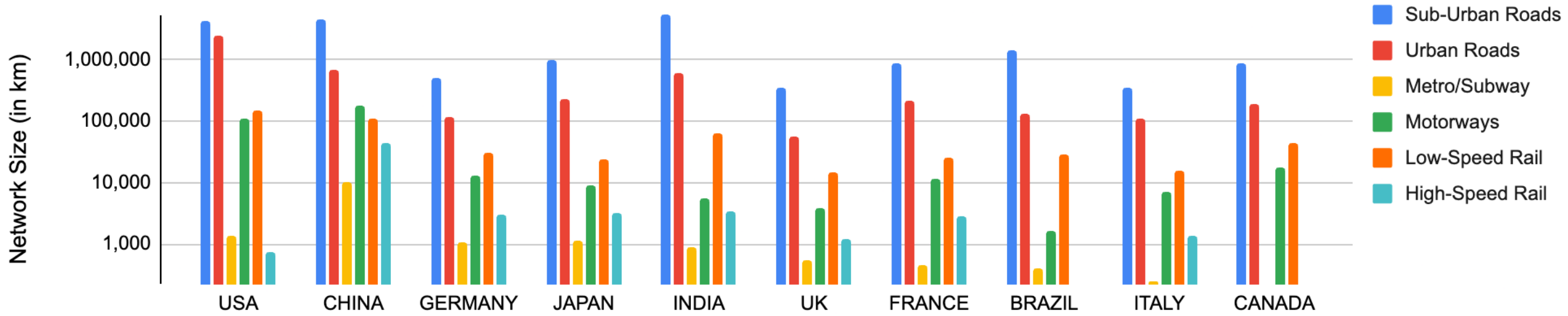
- *National Bureau of Statistics* (various: e.g., US Bureau of Transportation Statistics, China Ministry of Transport).



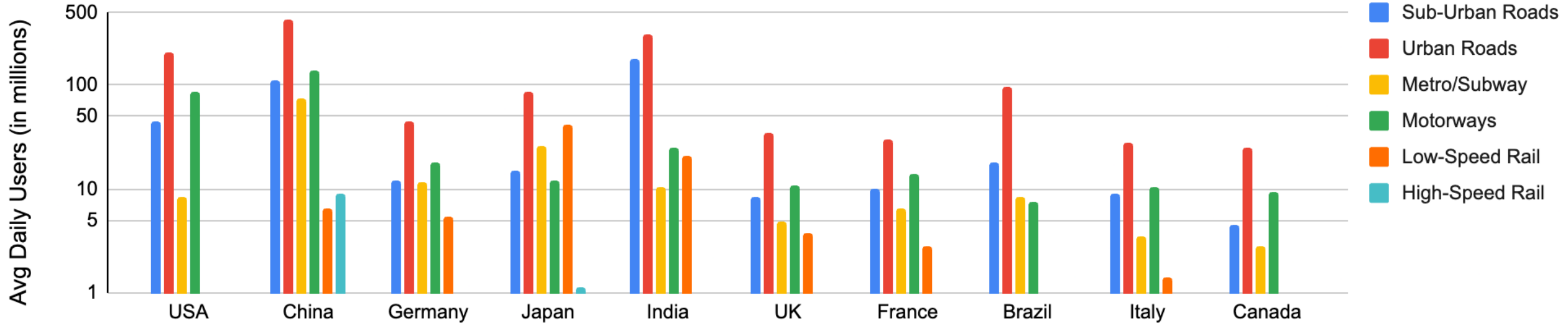
## Average Transport Speed



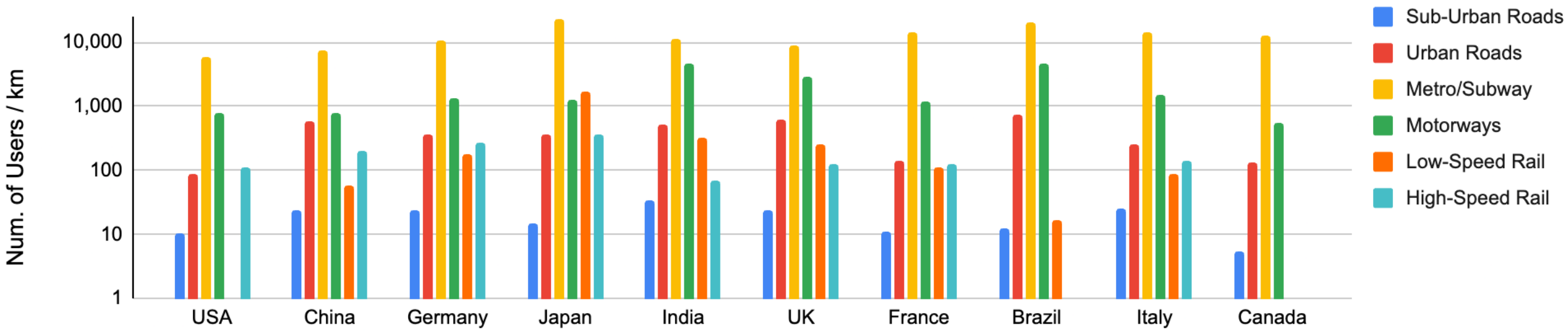
### Transport Network Size (Log Scale)



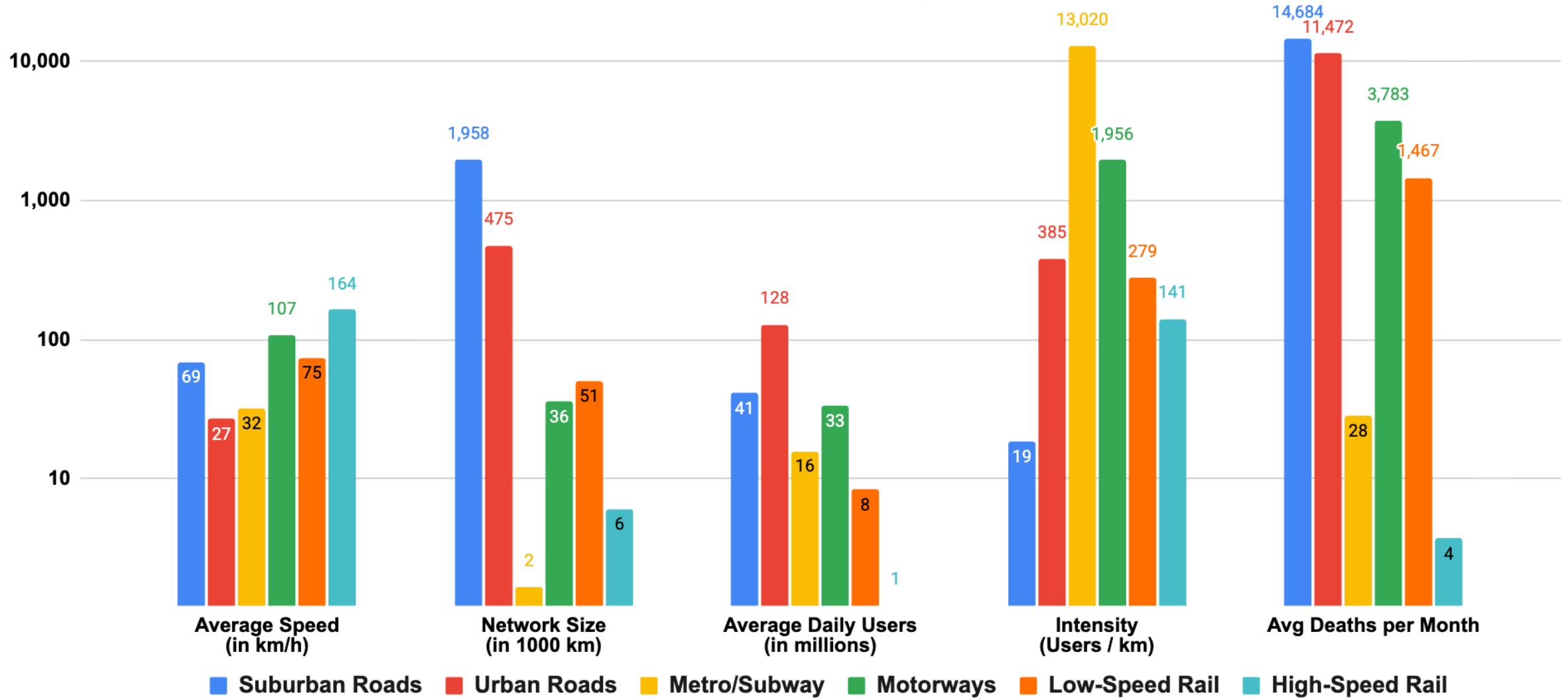
### Average Daily Users (Log Scale)



### Traffic Intensity (Log Scale)



# Transport Network Summary



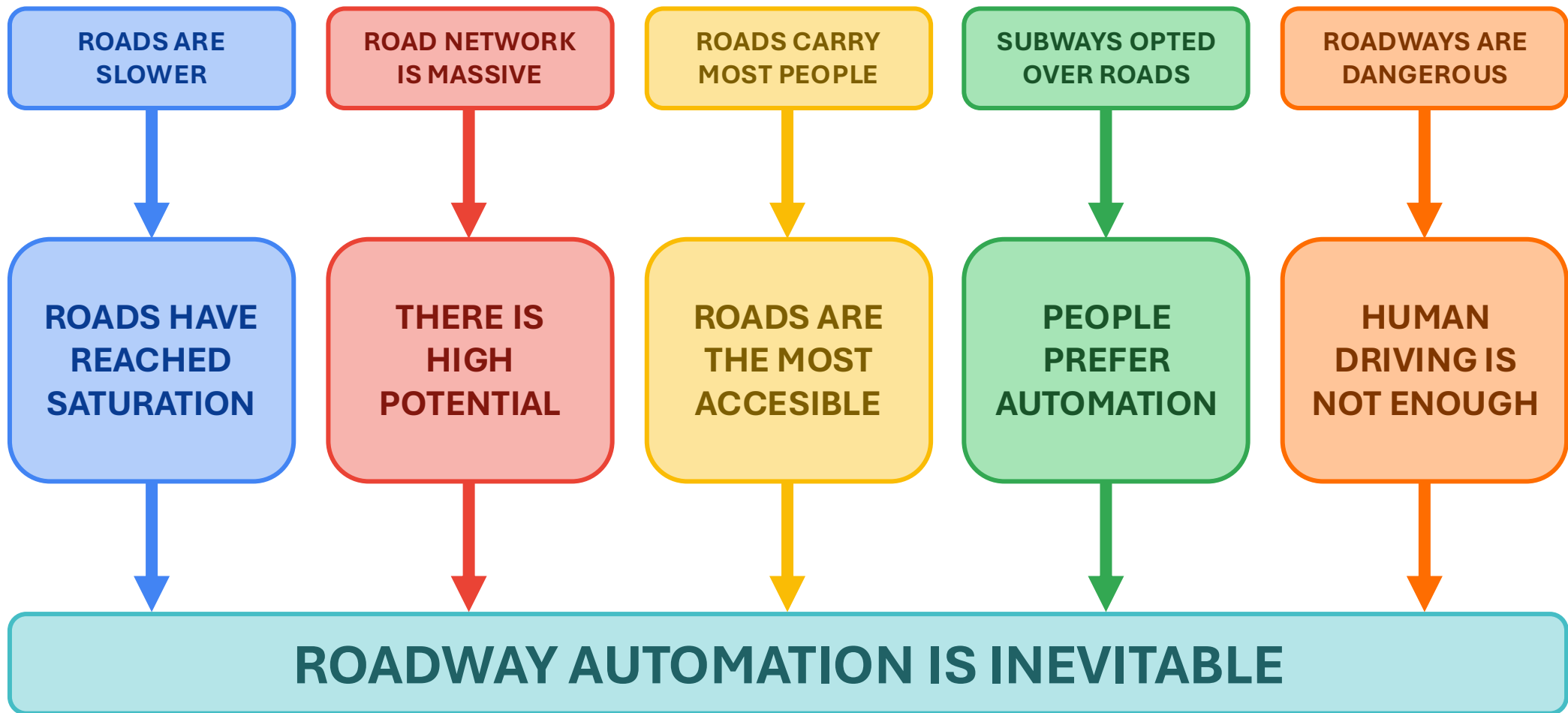
**ROADS ARE SLOWER**

**ROAD NETWORK IS MASSIVE**

**ROADS CARRY MOST PEOPLE**

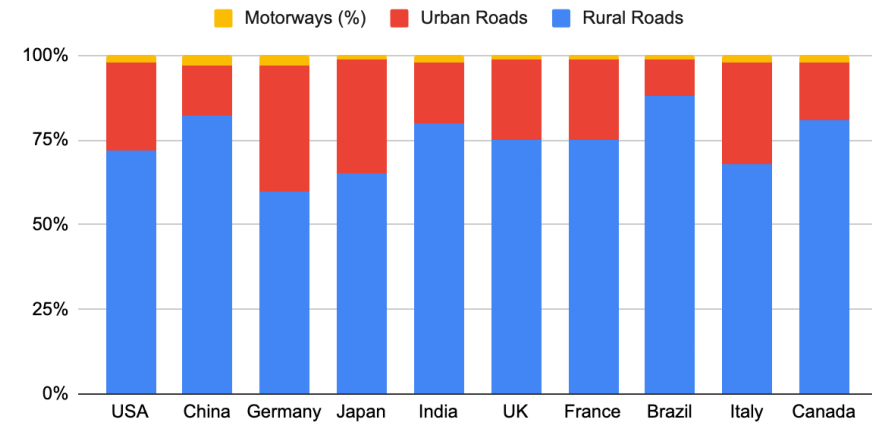
**SUBWAYS OPTED OVER ROADS**

**ROADWAYS ARE DANGEROUS**

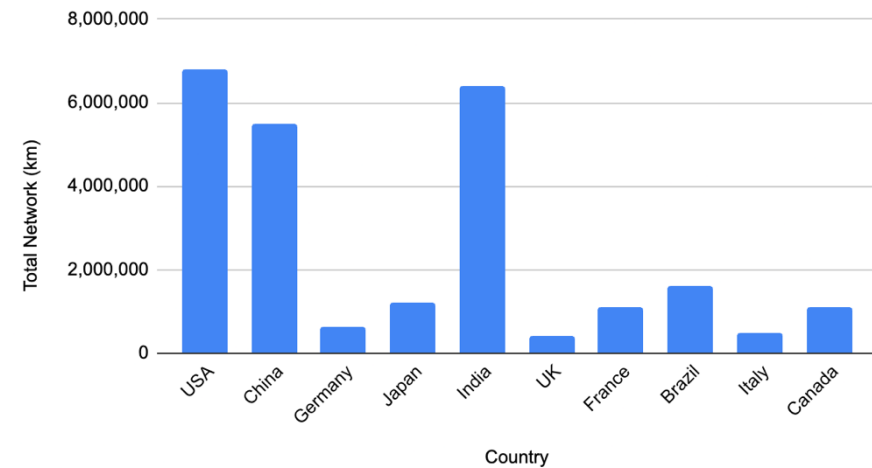


- **The Rural Dominance**
  - from "lane-keeping" to handling unmapped road
  - lack of infrastructure (V2X)
  - and high-speed edge cases.
- **The "Urban" Intensity:**
  - **Interactive complexity** (pedestrians, cyclists, and traffic signals).
- **Infrastructure Gaps:**
  - Low motorway percentages.
  - Infrastructure-agnostic navigation systems
- **The Scalability Challenge:**
  - Human driven traffic monitoring and regulation is a major challenge

Rural Roads, Urban Roads and Motorways (%)



Total Network (km) vs. Country



# The Society of Automotive Engineers

AE International (formerly the Society of Automotive Engineers) is a global association of more than 128,000 engineers and related technical experts in the aerospace, automotive, and commercial-vehicle industries.

## Core Mission & Influence

- **Standards Development:** Maintains over 10,000 active standards that ensure safety, quality, and interoperability across global transport sectors.
- **Professional Development:** Provides lifelong learning resources, certifications, and technical publications for mobility professionals.
- **STEM Outreach:** Encourages future talent through programs like the Collegiate Design Series (e.g., Formula SAE).

# J3016 - Taxonomy and Definitions for Terms Related to Driving Automation Systems for On-Road Motor Vehicles

## Standardization

Provides a functional terminology for driving automation to eliminate confusion among policymakers, engineers, and the public.

## Operational Scope

Applies specifically to "on-road" motor vehicles (highways and surface streets).

## Levels of Automation

Describes six levels of driving automation ranging from "No Driving Automation" (Level 0) to "Full Driving Automation" (Level 5)

# J3016 - Taxonomy and Definitions for Terms Related to Driving Automation Systems for On-Road Motor Vehicles

**DDT** (Dynamic Driving Task) ✓



The DDT is generally split into two main categories of functions:

## Operational Functions (Physical Control):

- **Lateral Motion Control:** Steering the vehicle.
- **Longitudinal Motion Control:** Accelerating and braking (managing speed).

## Tactical Functions (Planning and Response):

- **OEDR (Object and Event Detection and Response):** Monitoring the environment (detecting pedestrians, other cars, traffic lights) and reacting to them.
- **Maneuver Planning:** Deciding to change lanes, turn at an intersection, or signal a turn.
- **Enhancing Conspicuity:** Using lights, signals, or the horn to communicate with other road users.

**Trip Scheduling:** Deciding what time to leave.

**Destination Selection:** Deciding where to go.

**Waypoint Selection:** Choosing which route to take (e.g., "Take the highway instead of the scenic route").

# Level 0: No Driving Automation

## Description

- The driver is always in full control of the vehicle.
- Car may provide warnings or momentary assistance (like emergency braking),
- No sustained control of steering or acceleration.

## Example Features:

Forward collision warning, lane departure warning, and blind-spot monitoring.

**Real-World Examples:** Most older vehicle models and basic modern economy cars without "active" cruise control.



AI Usage

# Level 1: Driver Assistance

## Description:

- The vehicle can assist with either steering or acceleration/braking, but not both simultaneously.
- The driver must remain fully engaged and keep their hands on the wheel.

## Example Features:

Adaptive Cruise Control (maintains distance) or Lane Keep Assist (corrects steering).

## Real-World Examples:

Standard Adaptive Cruise Control found in many modern Toyotas, Hondas, and Fords.



AI Usage

# Level 2: Partial Driving Automation

## Description

- The vehicle can control both steering and acceleration/braking at the same time.
- Driver is still "driving" and must supervise the system constantly,
- Often requiring hands on the wheel or eyes on the road via camera monitoring.

## Features:

Cruise on a highway. Both speed & steering control

## Real-World Examples:

Tesla Autopilot, Volvo Pilot Assist, Hyundai Highway Driving Assist, GM Super Cruise, Ford BlueCruise.



AI Usage

# Level 3: Conditional Driving Automation

## Description

Under very specific conditions (such as clear weather on divided highways at lower speeds), the vehicle can take full control. The driver can take their eyes off the road but must be ready to take over immediately when the system requests it.

## Example Features:

Environmental monitoring and traffic jam pilot systems. stop and go motions, cruising on emptier roads

## Real-World Examples:

Mercedes-Benz DRIVE PILOT (available in select regions like Germany and Nevada/California) and Honda Legend Sensing Elite (Japan).



AI Usage

# Level 4: High Driving Automation

## Description

- Vehicle can perform all driving tasks and monitor the environment in specific geographic areas (geofencing) or under certain conditions.
- No driver intervention is required within these "Operational Design Domains."
- In many cases, the vehicle might not even have pedals or a steering wheel.

## Example Features:

- Autonomous urban taxi services performing door to door service.

## Real-World Examples:

- Waymo One (operating in Phoenix, San Francisco, and Los Angeles) and Cruise (robotaxi services).



AI Usage

# Level 5: Full Driving Automation

## Description

Vehicle can drive itself anywhere a human can drive, under any conditions (weather, off-road, etc.). No human interaction is required, and the "driver" is effectively just a passenger.

## Example Features:

A car with no steering wheel that can navigate a blizzard on an unmarked dirt road just as well as a paved highway.

**Real-World Examples:** There are **no** commercially available Level 5 vehicles currently deployed. This remains a goal for the future of automotive technology.



AI Usage

	SAE LEVEL 0™	SAE LEVEL 1™	SAE LEVEL 2™	SAE LEVEL 3™	SAE LEVEL 4™	SAE LEVEL 5™
What does the human in the driver's seat have to do?	You <b>are driving</b> whenever these driver support features are engaged – even if your feet are off the pedals and you are not steering			You <b>are not driving</b> when these automated driving features are engaged – even if you are seated in “the driver’s seat”		
	You <b>must constantly supervise</b> these support features; you must steer, brake or accelerate as needed to maintain safety			When the feature requests, you must drive	These automated driving features will not require you to take over driving	

Copyright © 2021 SAE International.

	These are driver support features			These are automated driving features		
What do these features do?	These features are limited to providing warnings and momentary assistance	These features provide steering <b>OR</b> brake/acceleration support to the driver	These features provide steering <b>AND</b> brake/acceleration support to the driver	These features can drive the vehicle under limited conditions and will not operate unless all required conditions are met	This feature can drive the vehicle under all conditions	
Example Features	<ul style="list-style-type: none"> <li>• automatic emergency braking</li> <li>• blind spot warning</li> <li>• lane departure warning</li> </ul>	<ul style="list-style-type: none"> <li>• lane centering <b>OR</b></li> <li>• adaptive cruise control</li> </ul>	<ul style="list-style-type: none"> <li>• lane centering <b>AND</b></li> <li>• adaptive cruise control at the same time</li> </ul>	<ul style="list-style-type: none"> <li>• traffic jam chauffeur</li> </ul>	<ul style="list-style-type: none"> <li>• local driverless taxi</li> <li>• pedals/steering wheel may or may not be installed</li> </ul>	<ul style="list-style-type: none"> <li>• same as level 4, but feature can drive everywhere in all conditions</li> </ul>

# DISCLAIMER

**THE NEXT FEW SLIDES ARE A CASE STUDY OF THE WAYMO AUTONOMOUS DRIVING SYSTEM**

I am not an authorised spokesperson of Waymo  
I am only showcasing information provided by Waymo publicly in their websites and blogs



Co-funded by  
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# WAYMO

## 20M+ rides served

And counting - with a 93% satisfaction rate from our riders.\*

### 2009

- The Google self-driving car project begins
- 10 uninterrupted 100-mile routes in a Toyota Prius

### 2015

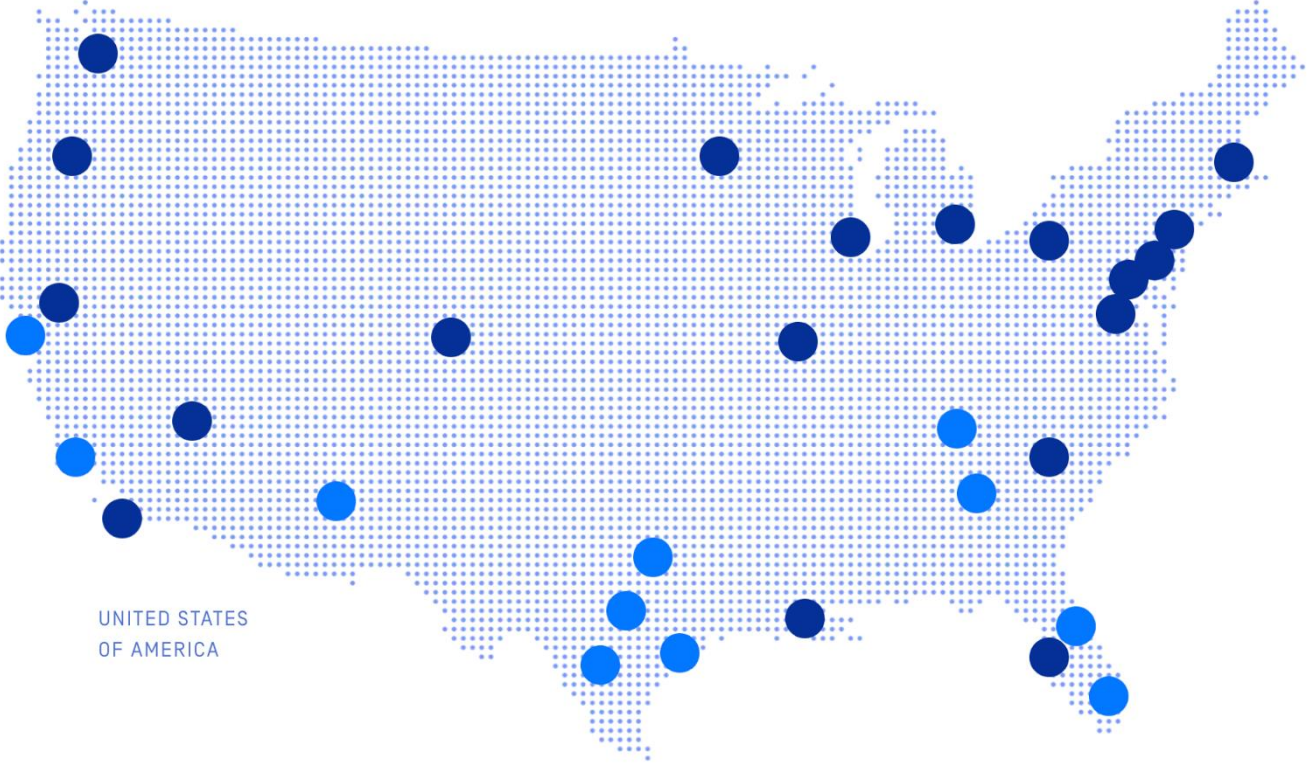
- First fully autonomous ride on public roads.
- Steven Mahan took the world's first fully self-driving ride on public roads in Austin, TX. Steve is legally blind.

### 2016

- Waymo becomes independent



# Where Waymo is driving



## Mapping out every intersection, sign, and signal

Before our Waymo Driver begins operating in a new area, we first map the territory with incredible detail, from lane markers to stop signs to curbs and crosswalks. Then, instead of relying solely on external data such as GPS which can lose signal strength, the Waymo Driver uses these highly detailed custom maps, matched with real-time sensor data and artificial intelligence (AI) to determine its exact road location at all times.

How it works

## Keeping an eye on everything, all at once

The Waymo Driver's perception system takes complex data gathered from its advanced suite of car sensors, and deciphers what's around it using AI - from pedestrians to cyclists, vehicles to construction, and more. The Waymo Driver also responds to signs and signals, like traffic light colors and temporary stop signs.

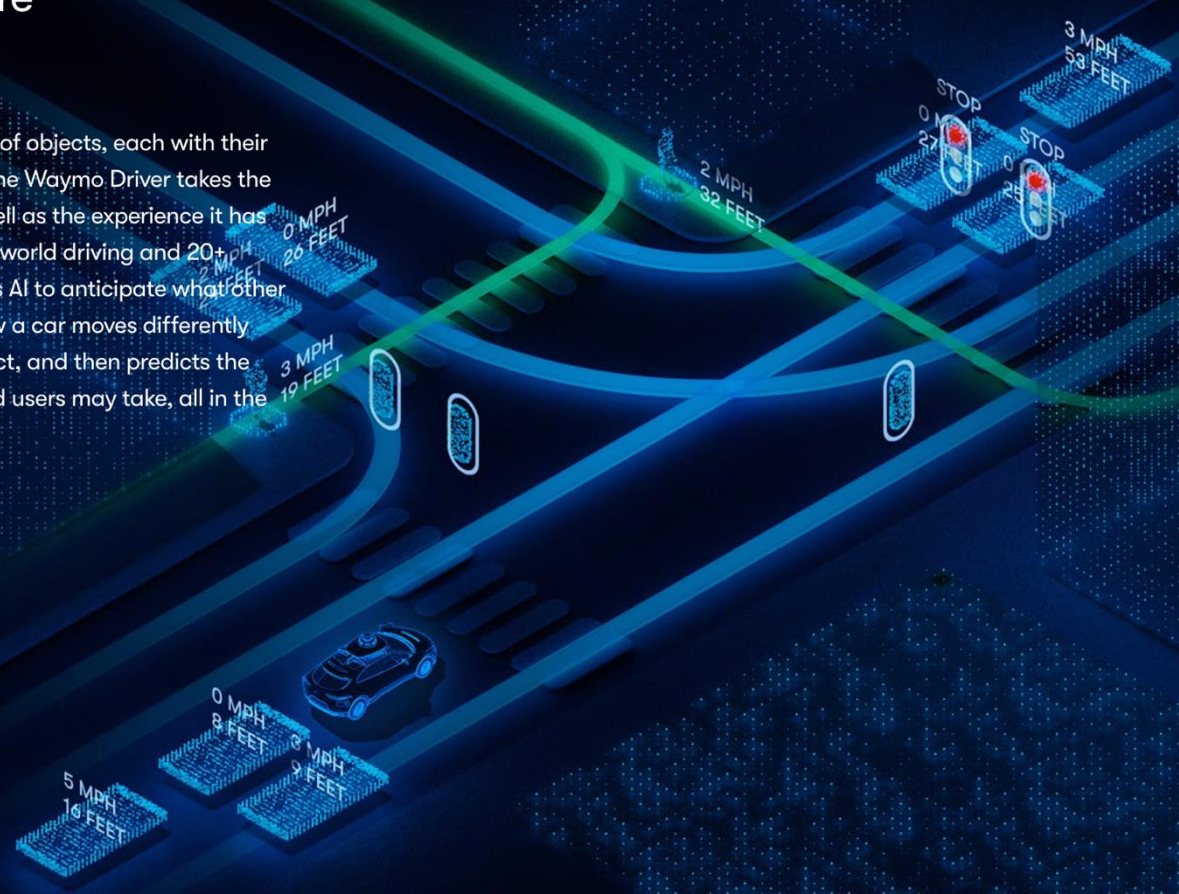
How it works



## Predicting things before they happen

Driving situations can involve hundreds of objects, each with their own unique behaviors and intentions. The Waymo Driver takes the information it gathers in real time, as well as the experience it has built up over its 20+ million miles of real world driving and 20+ billion miles in simulation, and leverages AI to anticipate what other road users might do. It understands how a car moves differently than a cyclist, pedestrian, or other object, and then predicts the many possible paths that the other road users may take, all in the blink of an eye.

How it works



## Planning for the safest outcome

The Waymo Driver takes all of this information – from its highly-detailed maps, to what objects are around and where they might go – and uses AI to plan the best action or route to take. It instantly determines the exact trajectory, speed, lane, and steering maneuvers needed to behave **safely** throughout its journey.

How it works

Long Range Camera + Radar

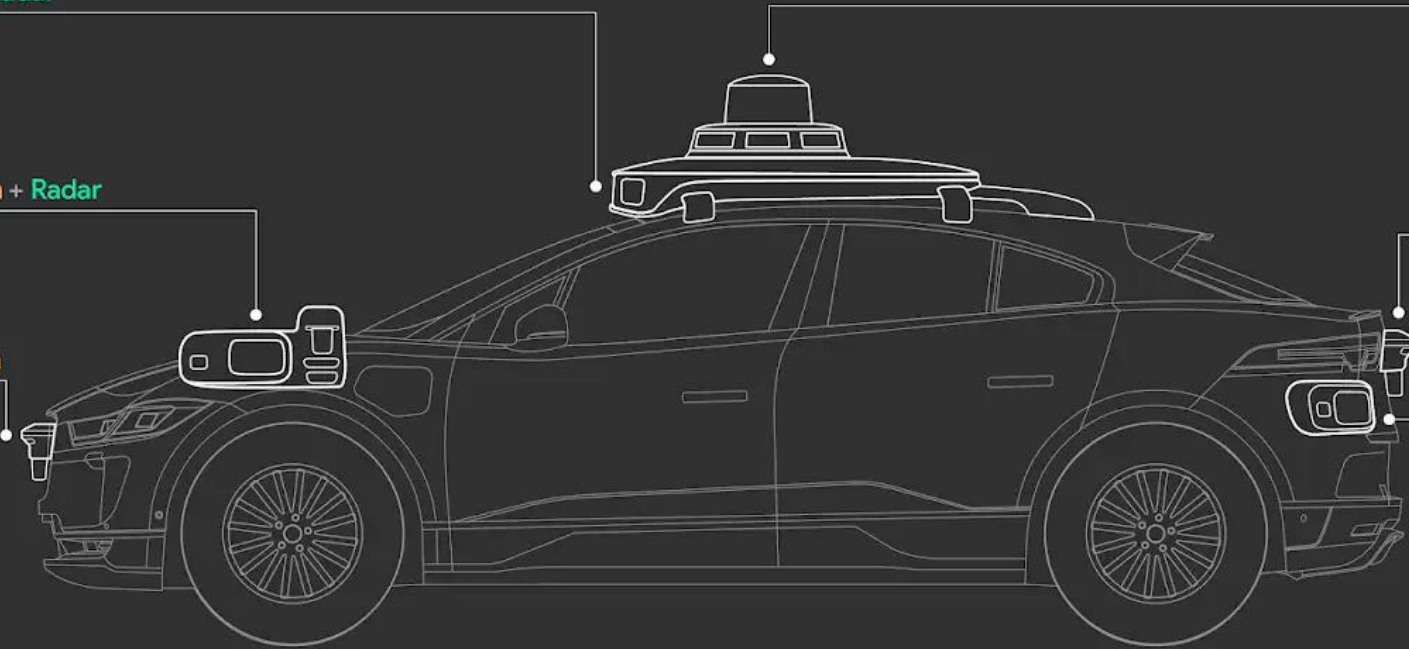
360 Lidar + 360 Vision System

Perimeter Lidar +  
Peripheral Vision System + Radar

Perimeter Lidar +  
Perimeter Vision System

Perimeter Lidar +  
Perimeter Vision System

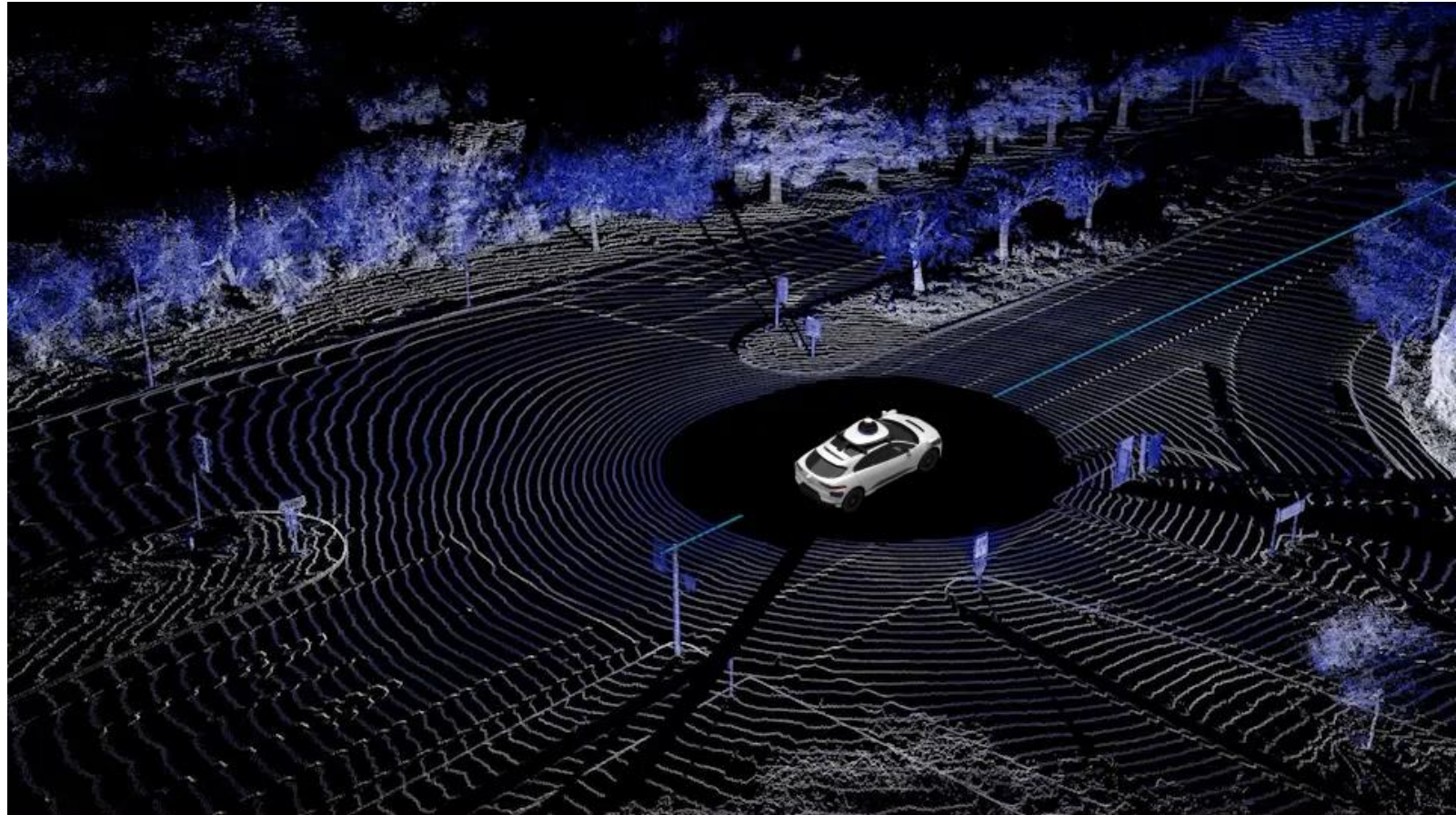
Peripheral Vision System  
+ Radar



# LiDAR: The 3D Architect

(Light Detection and Ranging)

- **Primary Function:** High-precision 3D geometry and depth mapping.
- **360° Vision:** Top-mounted and perimeter sensors eliminate all blind spots.
- **Spatial Accuracy:** Centimeter-level precision for object shape and distance.
- **Lighting Independent:** Consistent performance in total darkness.



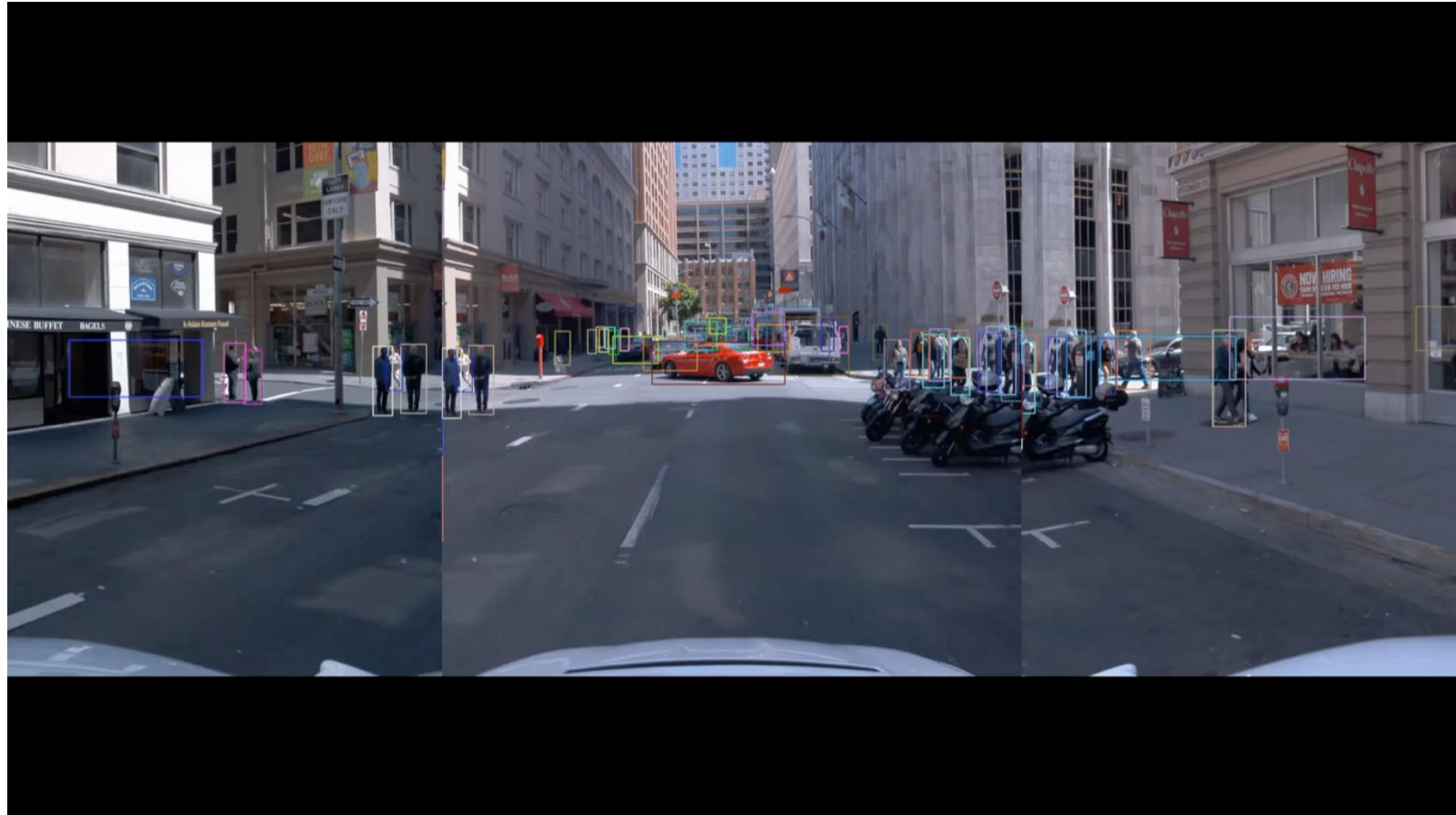
# Cameras: The Semantic Eye

**Primary Function:** Classification and long-range visual context.

**Object Identification:** Reads traffic lights, road signs, and emergency signals.

**Extreme Range:** Detects small objects and obstacles over 500m away.

**Multi-Lens Array:** 13 overlapping cameras for a seamless visual field.



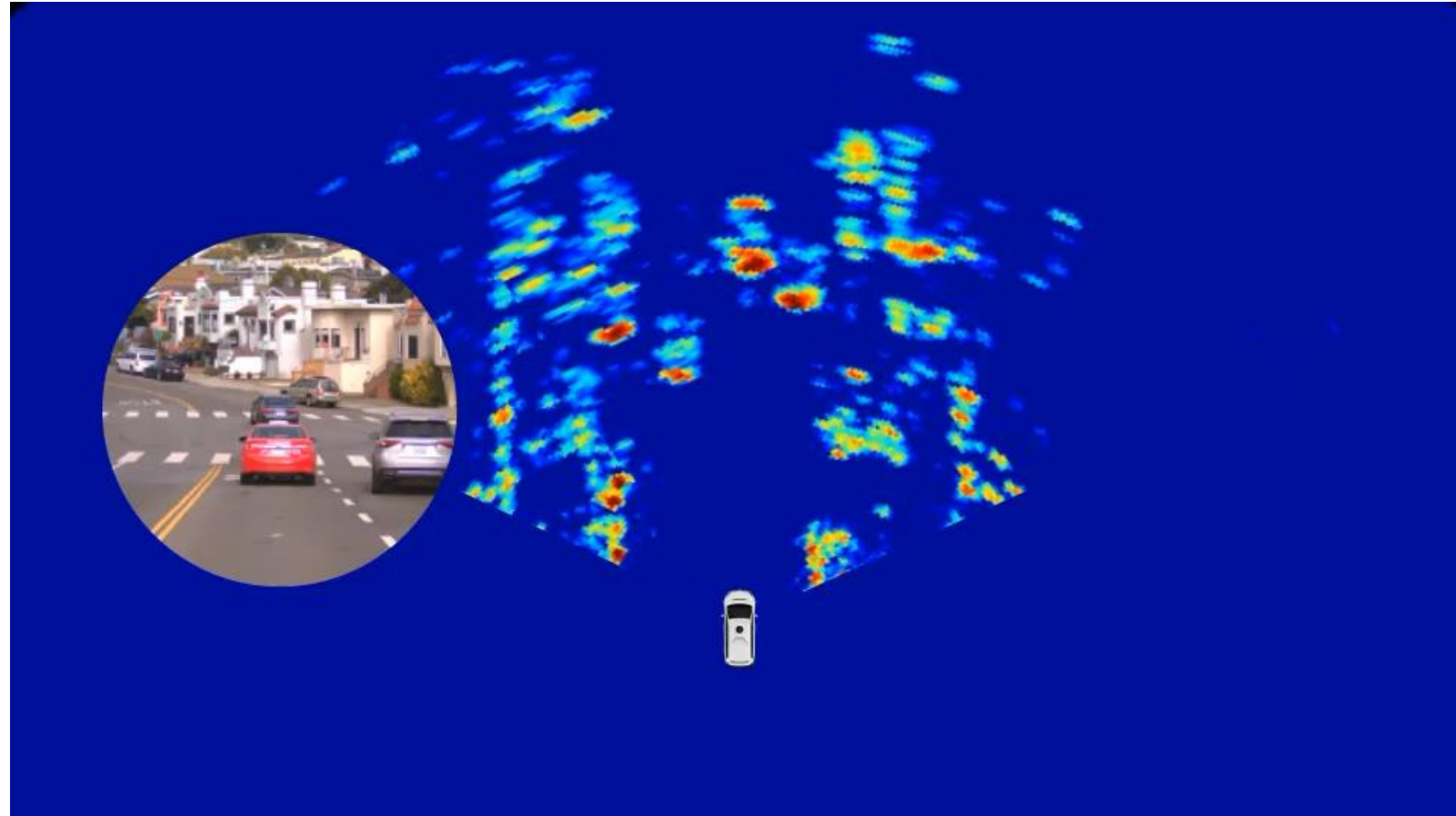
# RADAR: The Weather Specialist

**Primary Function:** Velocity tracking and all-weather reliability.

**Doppler Effect:** Instant measurement of object speed and direction.

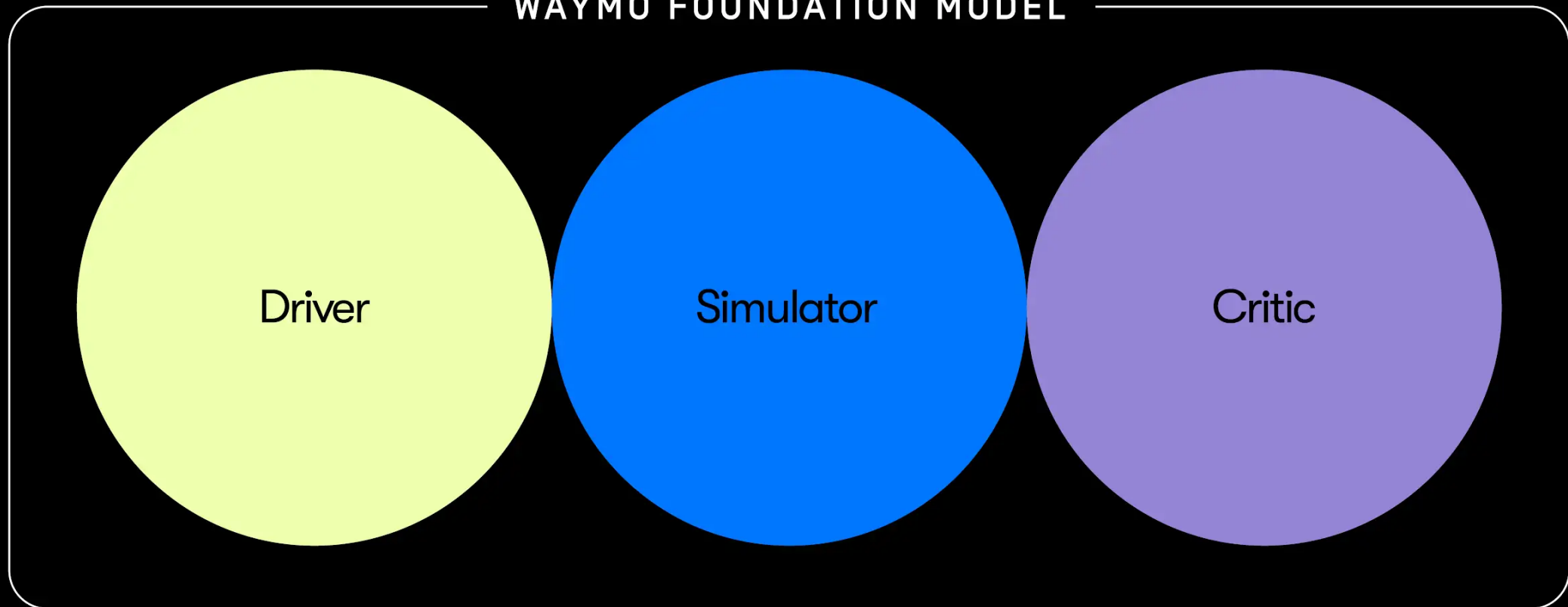
**Environmental Immunity:** Operates through heavy fog, rain, and snow.

**Imaging RADAR:** High-resolution detection of stationary and moving targets.

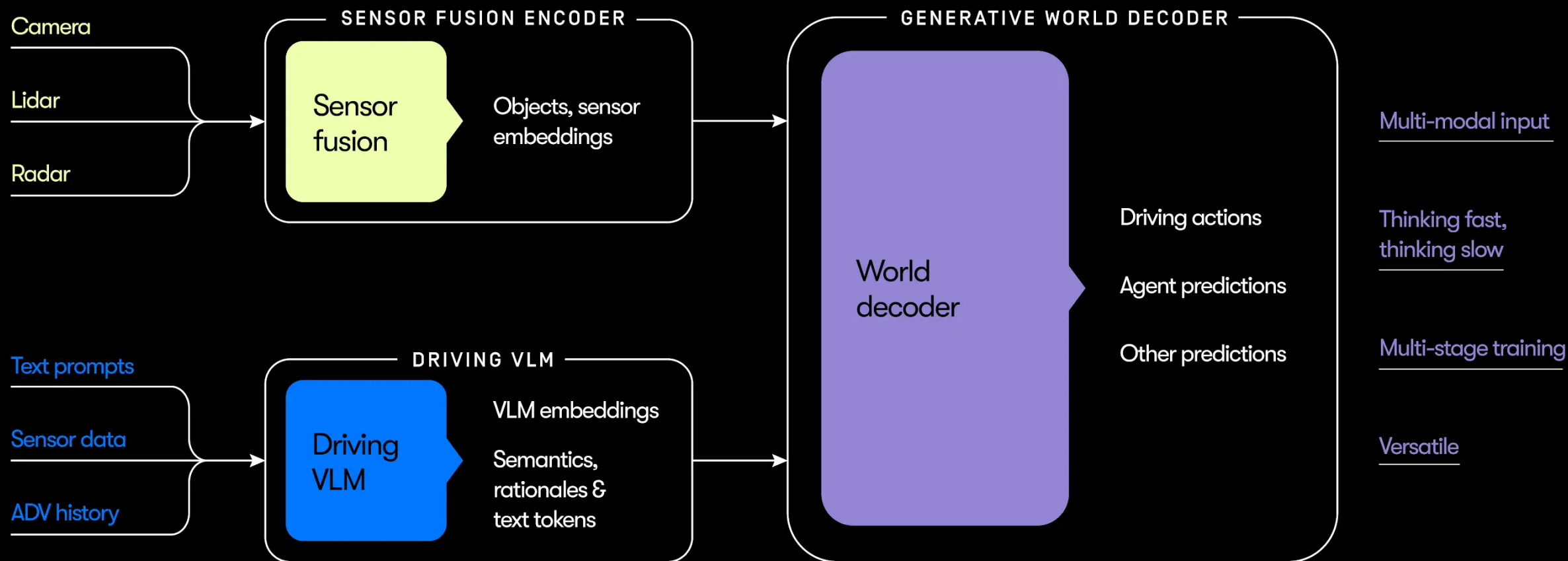


# Waymo's Holistic Approach to Demonstrably Safe AI

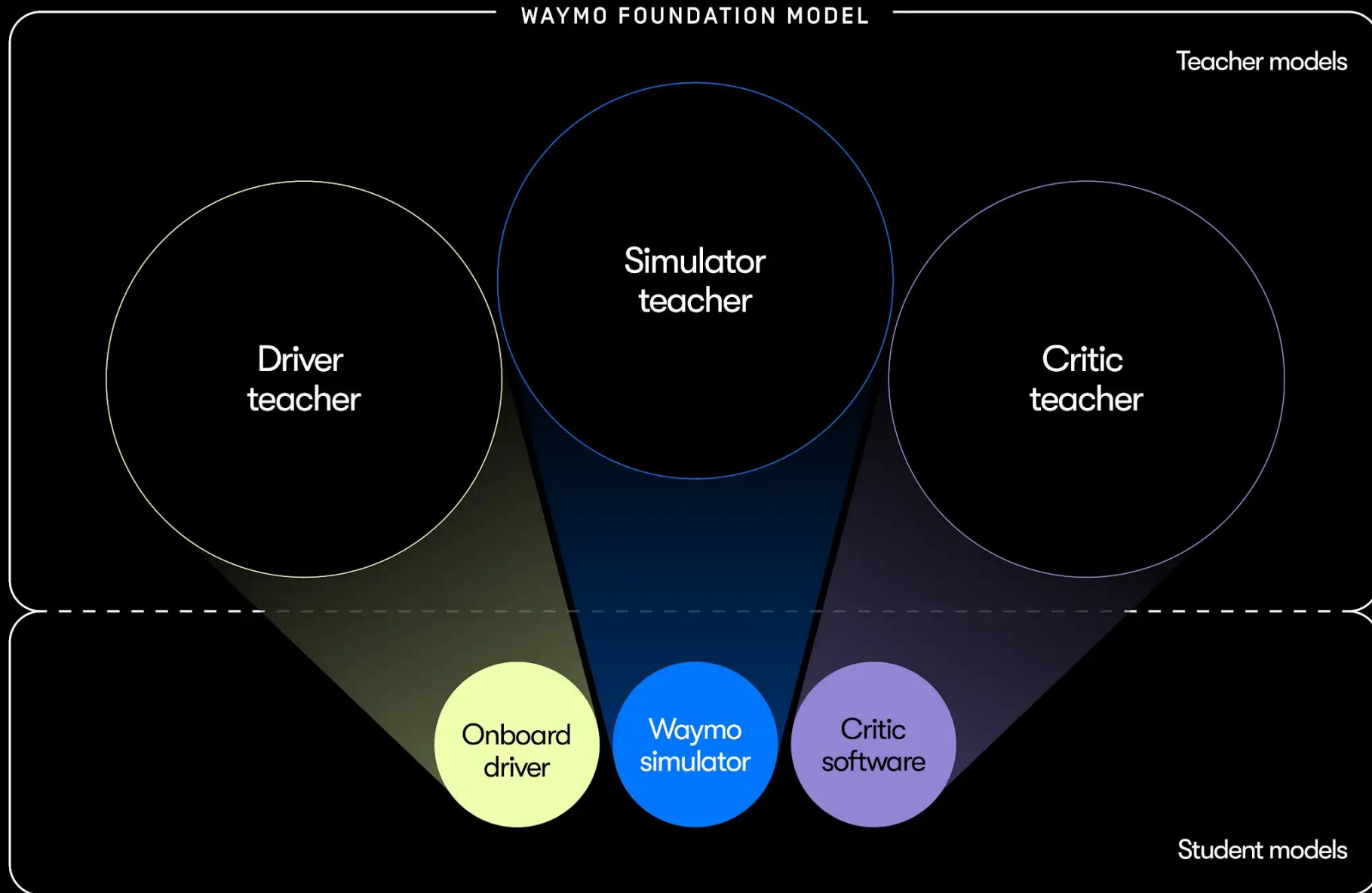
## WAYMO FOUNDATION MODEL



# The Waymo Foundation Model

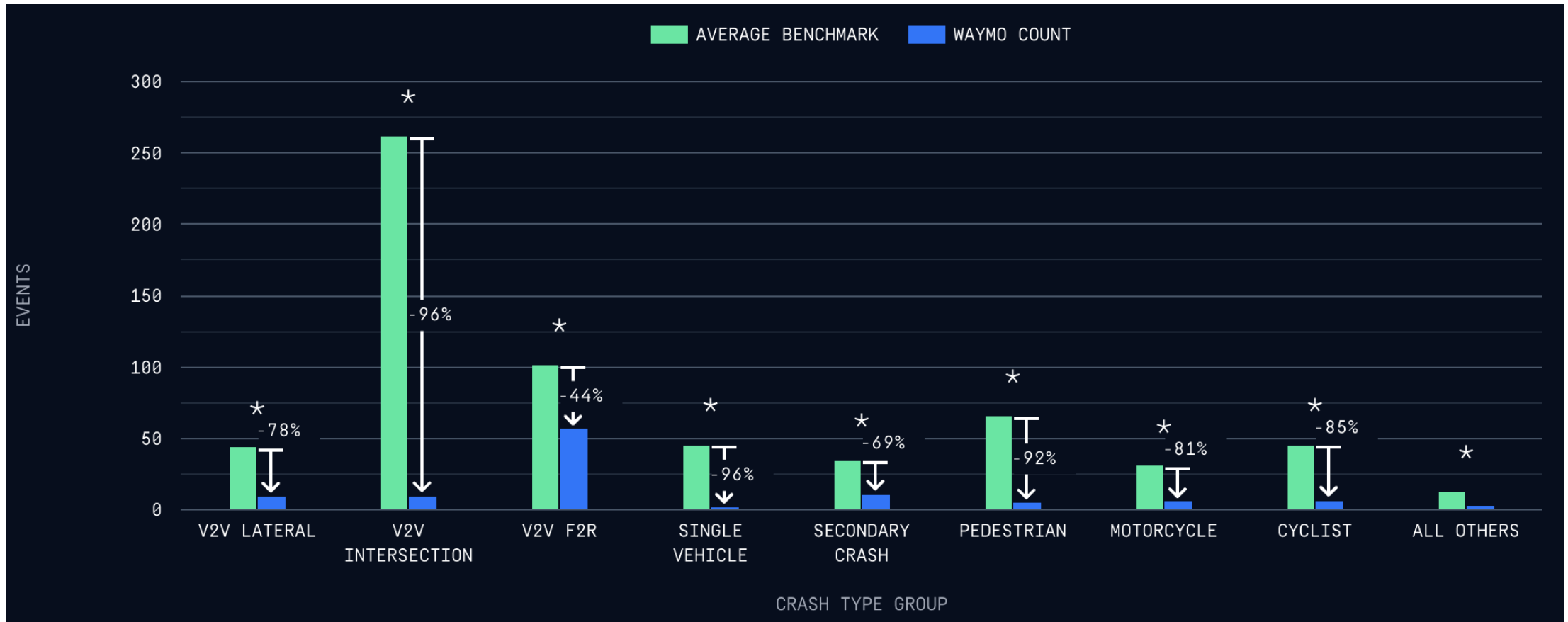


# Waymo's AI ecosystem

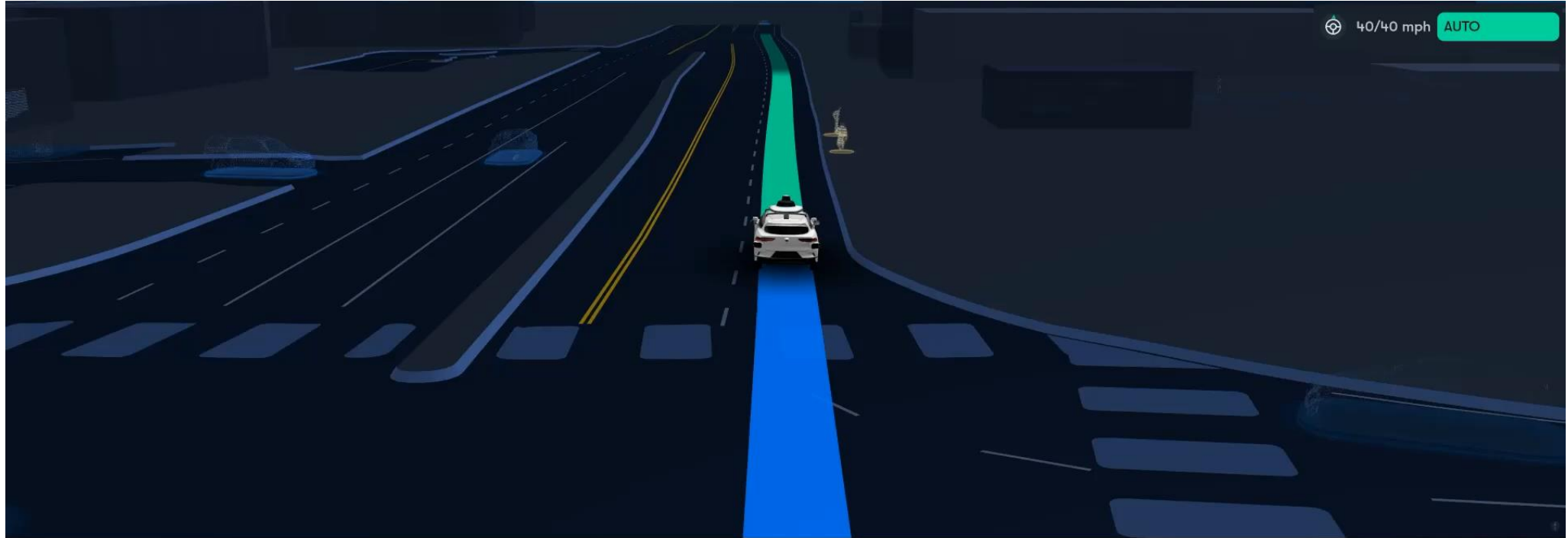




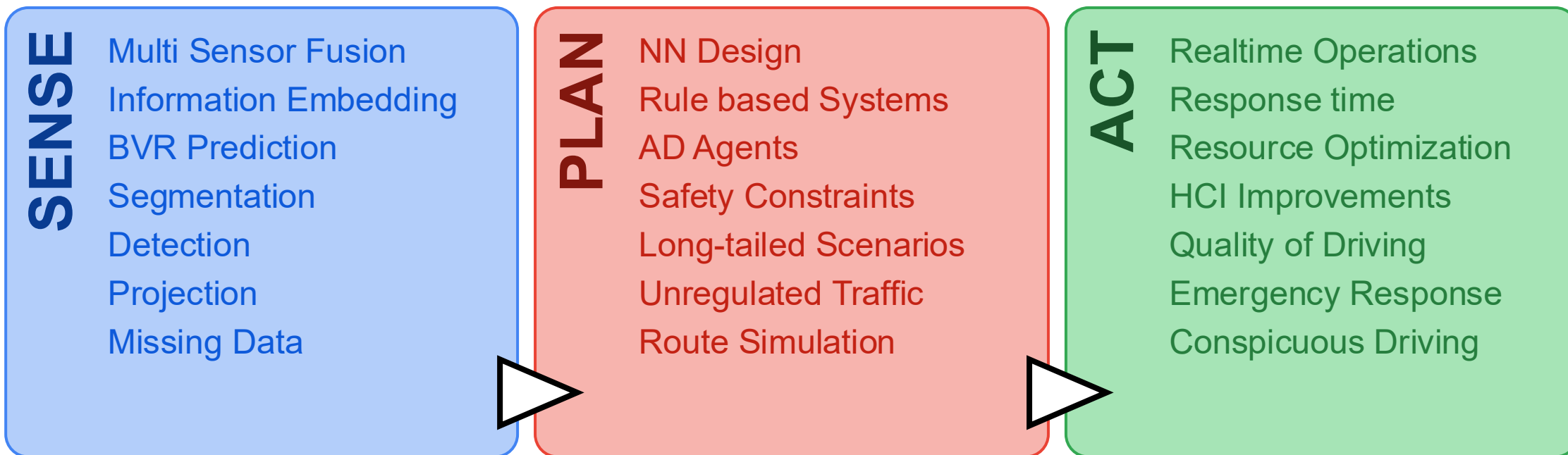
# Accident Reduction



# Accident Reduction



# SCOPE FOR RESEARCH



# GAPS



## TECHNICAL

- Long-tailed Events
- Extreme weathers
- Entropy in Traffic
- Data Scarcity
- Reaction Time
- Communication



## COGNITIVE

- Ability of Computer Vision
- Semantic Intelligence
- Social Norms
- Emotional Aspects of Traffic



## INFRASTRUCTURE

- Road Qualities
- Infrastructure is built for Humans
- V2X communication
- VIoT infrastructure
- 5G and connectivity
- Monitoring



## REGULATORY

- AI Education
- New Policies
- Uniformity of Traffic Laws
- AI Certification

# OPPURTUNITIES

## SERVICE

- Mobility as a Service
- From Ownership to Subscription on demand
- Ride Sharing - Reduced number of vehicles, parking spaces
- Better mobility solutions for physically impaired or elder individuals

## LOGISTICS

- AV supply chains
- AV trucks can operate 20+ hours a day
- AV Platoons can operate more optimally and consume lesser fuel
- Personnel Costs will be lowered
- Reduced journey times due to V2X capabilities

## SAFETY

- AVs will have lesser accident rates
- AVs have a more diverse set of sensors
- AVs are not emotional, always rational
- No loss of focus due to fatigue
- V2X communication can improve holistic traffic quality.

## MARKET

- Biggest Consumer Market in Densest Areas
- Most complex traffic, most frustration, most loss of time and fuel in big cities
- Human drivers in dense traffic can cause chaos.
- Shift from personal cars to shared AVs.
- Optimized fuel usage, No Honking.

# THREATS

## SECURITY

- Adversarial Attacks
- Denial of Service
- Hijacking
- Jamming



## TRUST

- AI Mistakes won't be forgiven
- Public reactions
- Lack of AI knowledge will prevent trust building



## LABOR

- AVs will eventually replace drivers one day.
- AVs will need personnel with new skill sets
- Labor Unions and Politics
- Transition will be messy



## ETHICS

- Choosing the lesser evil
- Moral Frameworks (Human vs AI)
- Passenger vs Pedestrian
- Who will be accused / defended ?

# SOME EXISTING COMPUTATIONAL INFRASTRUCTURE

Category	Key Solutions & Technologies	Primary Functions & Features
<b>Software Stacks &amp; AI</b>	<b><i>NVIDIA Alpamayo 1</i></b>	Open-source VLA model for reasoning-based autonomy without HD maps.
	<b><i>Autoware</i></b>	Open-source stack with native Internet of Cameras (NATIX) support.
	<b><i>Wayve AI Driver</i></b>	Mapless "Embodied AI" platform using generative world models (GAIA).
	<b><i>Apollo (Baidu)</i></b>	Mature open-source platform for L4 robotaxi and urban NOA fleets.
<b>Hardware &amp; Compute</b>	<b><i>NVIDIA DRIVE Thor</i></b>	Centralized car computer delivering up to 2,000 TOPS for AI workloads.
	<b><i>Qualcomm Dragonwing</i></b>	SoC designed for Physical AI and cross-domain cockpit/ADAS integration.
	<b><i>MediaTek Dimensity Auto</i></b>	High-end AI processors (S1/P1 Ultra) with integrated NVIDIA GPUs.
	<b><i>Horizon Robotics</i></b>	Specialized SoCs for scaled highway and urban navigation (NOA).
<b>Simulation &amp; Infrastructure</b>	<b><i>NVIDIA Omniverse</i></b>	High-fidelity world models for physics-based, closed-loop sensor testing.
	<b><i>CARLA</i></b>	Industry-standard simulator for autonomous driving research and RL training.
	<b><i>IPG nxtAIM</i></b>	Generative trajectory planning and synthetic dataset validation.
	<b><i>C-V2X / Smart City</i></b>	Cooperative perception and Roadside Units (RSUs) for edge intelligence.
<b>Middleware &amp; Dev-Ops</b>	<b><i>ROS 2</i></b>	Standardized robotic communication layer and middleware.
	<b><i>BlackBerry QNX</i></b>	Safety-certified real-time operating system (RTOS) for automotive.
	<b><i>NVIDIA Halos</i></b>	Comprehensive safety architecture for ASIL-D compliant systems.
	<b><i>Cloud Data Factories</i></b>	Managed pipelines (AWS/Azure) for curating and training on massive datasets.

# SOME EXISTING PHYSICAL INFRASTRUCTURE



Facility	Location	Core Infrastructure & Capabilities
ZalaZone	Hungary	250-hectare site featuring a <b>Smart City zone</b> , high-speed handling, and specific V2X (Vehicle-to-Everything) communication testing modules. Integrated with AVL's simulation toolchain.
AstaZero	Sweden	World's first full-scale independent test environment. Features a unique <b>Multilane Road</b> and <b>Rural Road</b> setup designed for complex edge-case scenarios and 5G-enabled autonomous drone/vehicle coordination.
MIRA Technology Park	UK	A leading mobility R&D cluster with over 35 global OEMs. It offers dedicated <b>Cybersecurity testing</b> and a specialized human-centered design center for passenger trust in autonomy.
American Center for Mobility (ACM)	USA (Michigan)	A 500-acre campus with a <b>2.5-mile high-speed highway loop</b> , tunnels, and urban intersections. It acts as a primary node for the US National V2X Deployment Plan.
K-City / Gwangju City	South Korea	Beyond the K-City track, the government recently designated the <b>entire city of Gwangju</b> as a regulatory sandbox, deploying 200 software-centric vehicles for 24/7 real-road data collection.
Transpolis	France	A "living lab" that recreates a 1:1 scale urban environment with reconfigurable buildings, enabling testing of vehicle interaction with pedestrians and cyclists in a controlled setting.

# SOME EXISTING PHYSICAL INFRASTRUCTURE



Facility

Location

Core Infrastructure & Capabilities

ZalaZone

Hungary

250-hectare site featuring a **Smart City zone**, high-speed handling, and specific V2X (Vehicle-to-Everything) communication testing modules. Integrated with AVL's simulation toolchain.



AstaZero



MIRA Technology Park



American Center for Mobility (ACM)



K-City / Gwangju City



Transpolis



the **Multilane** and 5G-

designed for

and urban environment Plan.

the city of cycles for 24/7

supportable cyclists in a

# EXISTING PHYSICAL INFRASTRUCTURE



Location  
Hungary

## Core Infrastructure & Capabilities

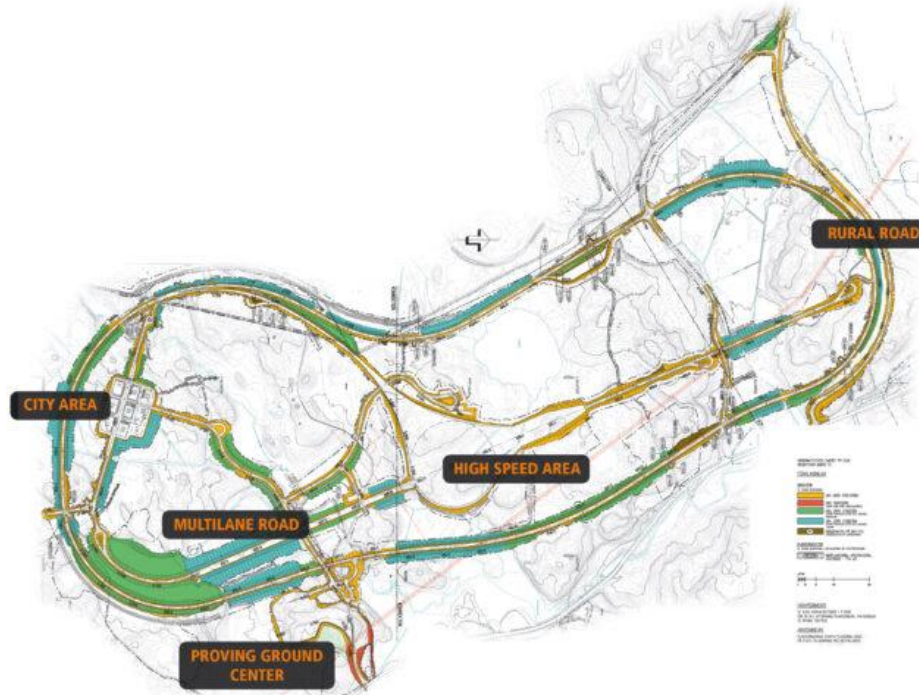
250-hectare site featuring a **Smart City zone**, high-speed handling, and specific V2X (Vehicle-to-Everything) communication testing modules. Integrated with AVL's simulation toolchain.



AstaZero

Sweden

World's first full-scale independent test environment. Features a unique **Multilane Road** and **Rural Road** setup designed for complex edge-case scenarios and 5G-enabled autonomous drone/vehicle coordination.



Partner with over 35 global OEMs. It offers dedicated specialized human-centered design center for

100-mile high-speed highway loop, tunnels, and urban primary node for the US National V2X Deployment Plan.

US government recently designated the **entire city of Fox**, deploying 200 software-centric vehicles for 24/7

1:1 scale urban environment with reconfigurable vehicle interaction with pedestrians and cyclists in a



# 5G PHYSICAL INFRASTRUCTURE

## Infrastructure & Capabilities

Test track site featuring a **Smart City zone**, high-speed handling, and specific V2X (Vehicle-to-Everything) communication testing modules. Integrated with AVL's test toolchain.

World's first full-scale independent test environment. Features a unique **Multilane** and **Rural Road** setup designed for complex edge-case scenarios and 5G-enabled autonomous drone/vehicle coordination.



<b>MIRA Technology Park</b>	UK	A leading mobility R&D cluster with over 35 global OEMs. It offers dedicated <b>Cybersecurity testing</b> and a specialized human-centered design center for passenger trust in autonomy.
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<b>American Center for Mobility (ACM)</b>	USA (Michigan)	A 500-acre campus with a <b>2.5-mile high-speed highway loop</b> , tunnels, and urban intersections. It acts as a primary node for the US National V2X Deployment Plan.
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<b>K-City / Gwangju City</b>	South Korea	Beyond the K-City track, the government recently designated the <b>entire city of Gwangju</b> as a regulatory sandbox, deploying 200 software-centric vehicles for 24/7 real-road data collection.
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<b>Transpolis</b>	France	A "living lab" that recreates a 1:1 scale urban environment with reconfigurable buildings, enabling testing of vehicle interaction with pedestrians and cyclists in a controlled setting.
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# SOME EXISTING PHYSICAL



Facility

Location

Core Infrastructure &

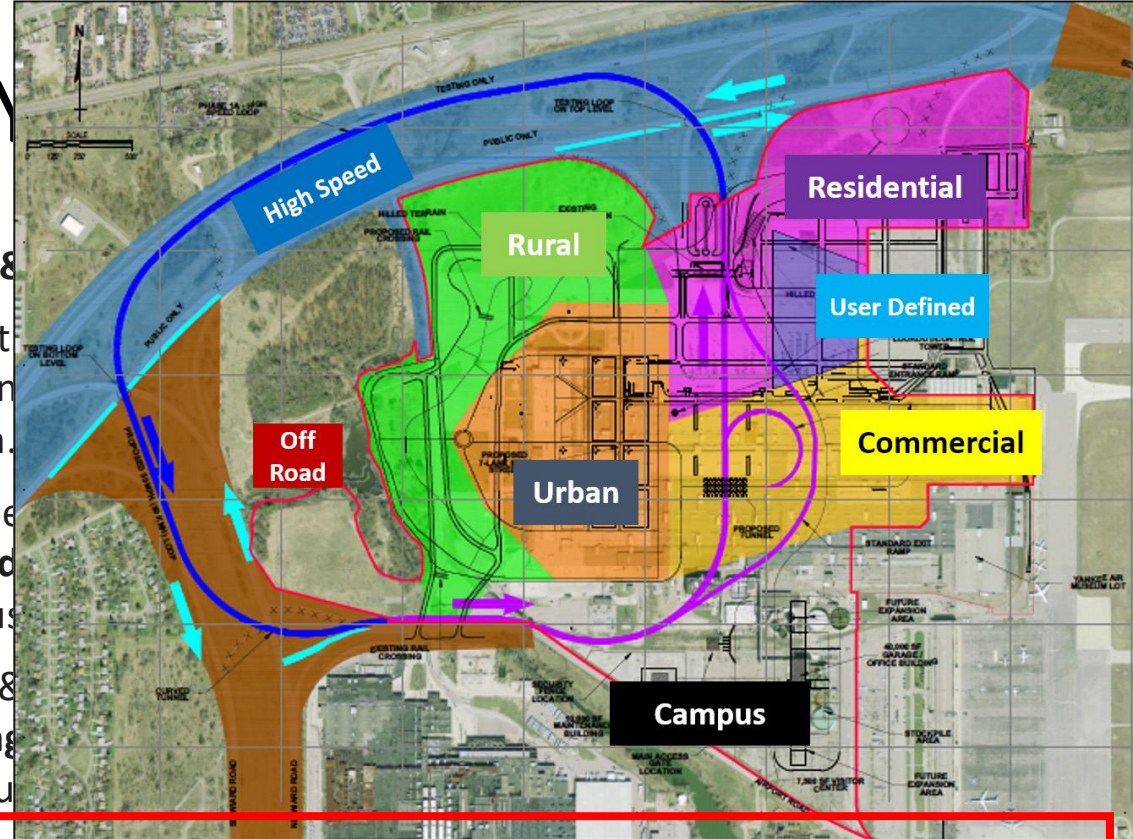
ZalaZone

Hungary

250-hectare site featuring a full-scale simulation toolchain.

World's first full-scale **Road and Rural Road** enabled autonomous driving.

A leading mobility R&D center for **Cybersecurity testing** and passenger trust in autonomous driving.



**American Center for Mobility (ACM)**

USA (Michigan)

A 500-acre campus with a **2.5-mile high-speed highway loop**, tunnels, and urban intersections. It acts as a primary node for the US National V2X Deployment Plan.

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**Transpolis**

France

A "living lab" that recreates a 1:1 scale urban environment with reconfigurable buildings, enabling testing of vehicle interaction with pedestrians and cyclists in a controlled setting.



# SOME EXISTING

Facility	Location	Core Info
ZalaZone	Hungary	250-hectare (Vehicle simulation) World's first Road enabled A leading Cyberse passenger A 500-acre intersec



**K-City / Gwangju City** South Korea Beyond the K-City track, the government recently designated the **entire city of Gwangju** as a regulatory sandbox, deploying 200 software-centric vehicles for 24/7 real-road data collection.



Transpolis	France	A "living lab" that recreates a 1:1 scale urban environment with reconfigurable buildings, enabling testing of vehicle interaction with pedestrians and cyclists in a controlled setting.
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# SOME EXIS



Facility

Location



ZalaZone

Hungary



AstaZero

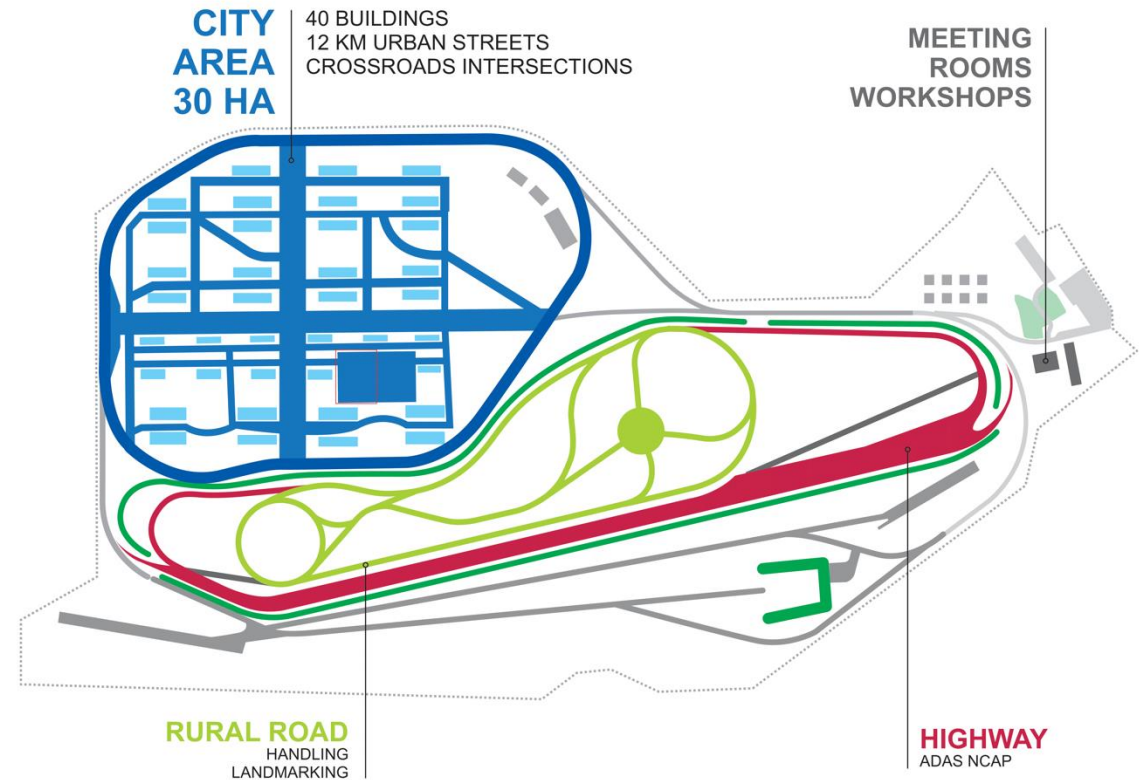
Sweden



Transpolis

France

A "living lab" that recreates a 1:1 scale urban environment with reconfigurable buildings, enabling testing of vehicle interaction with pedestrians and cyclists in a controlled setting.

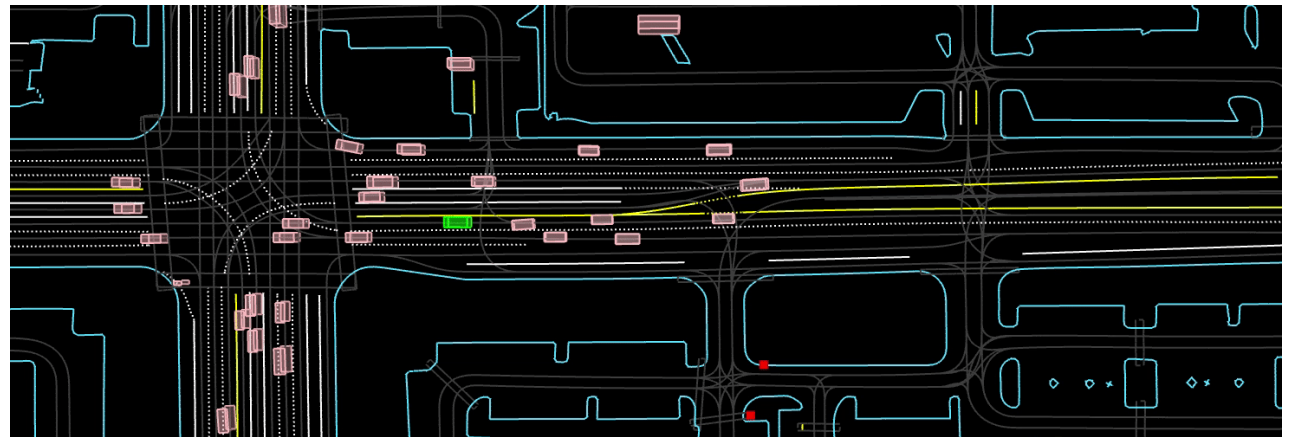
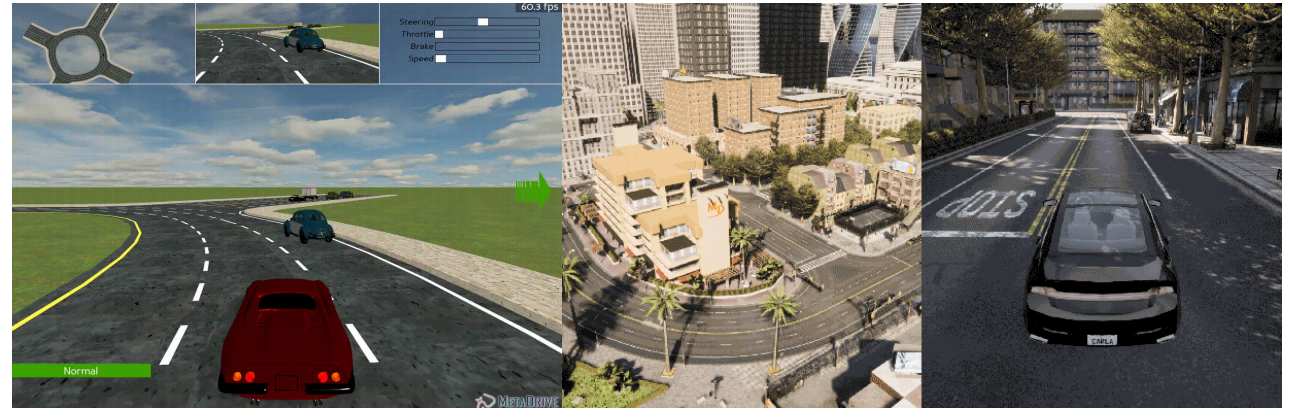
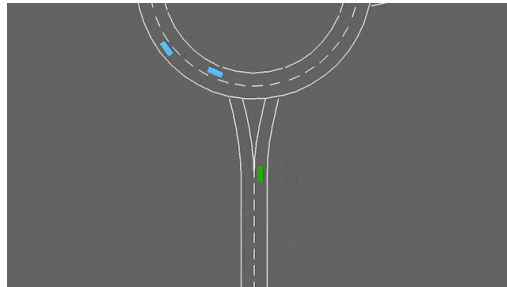
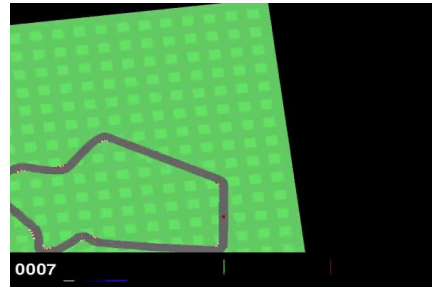


UK

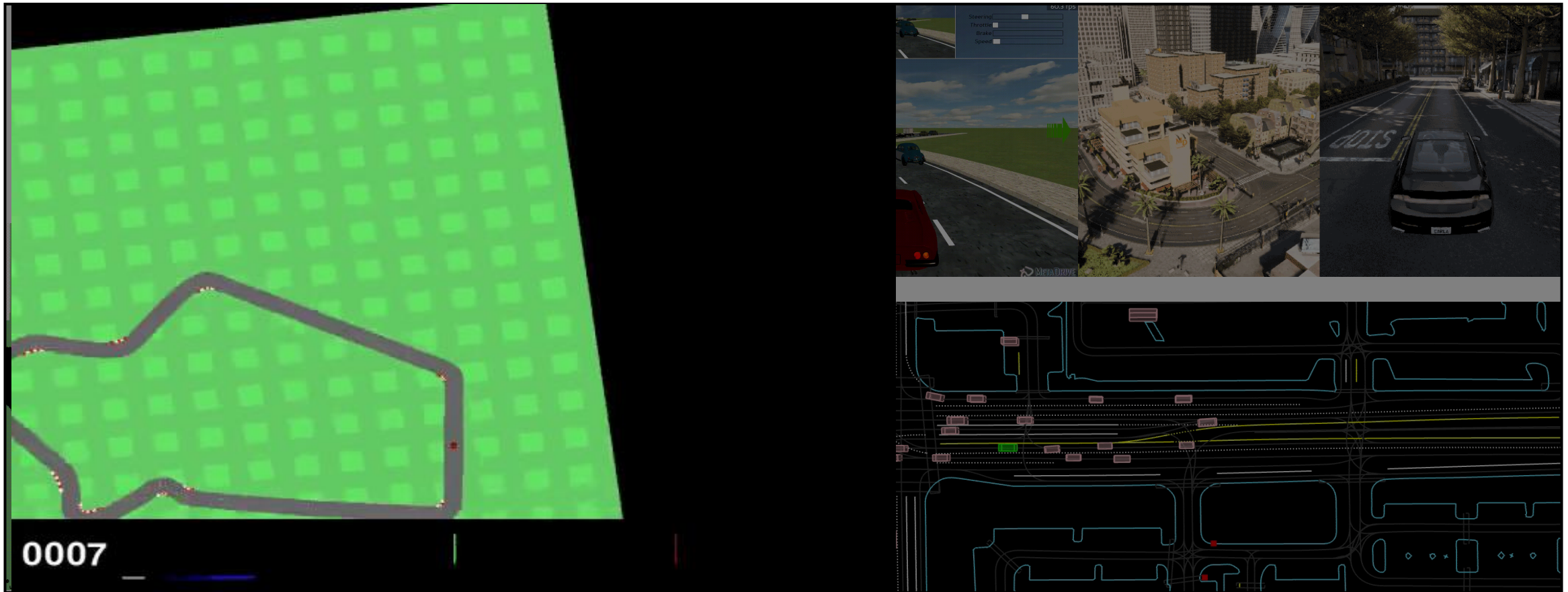
USA  
) (Michigan)

South Korea

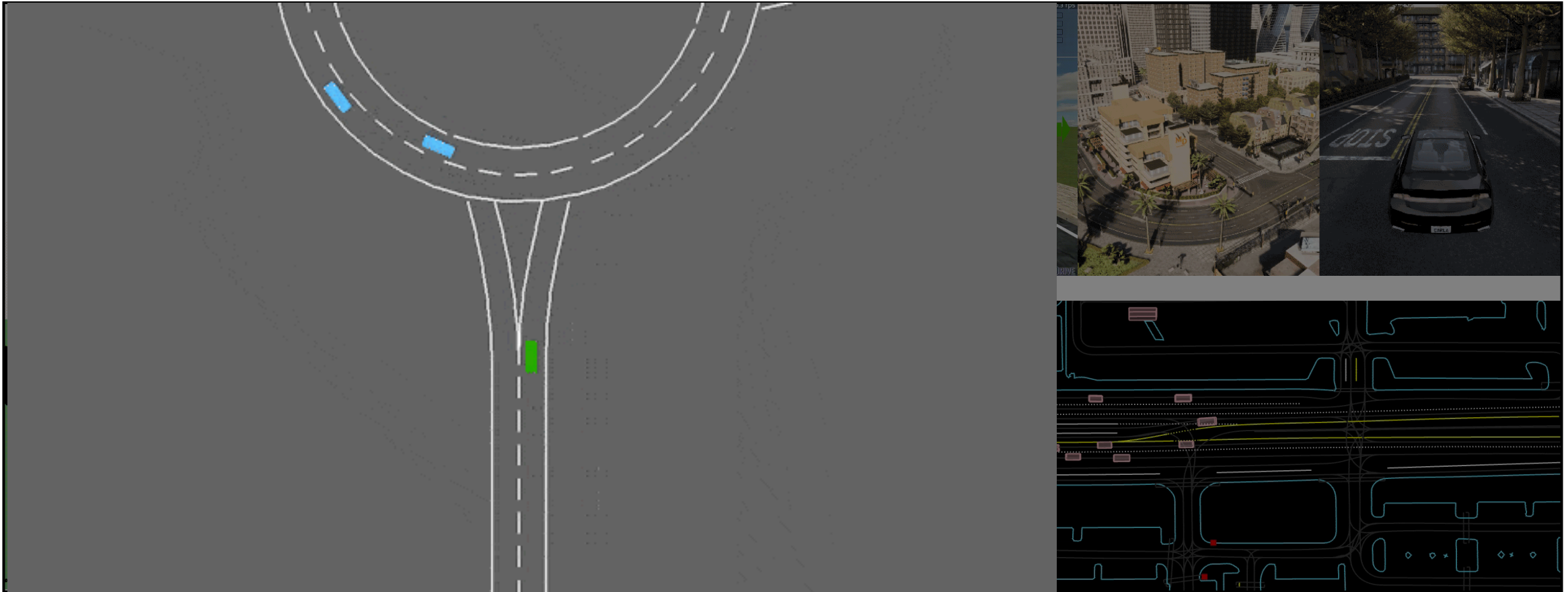
# SOME EXISTING SIMULATION PLATFORMS



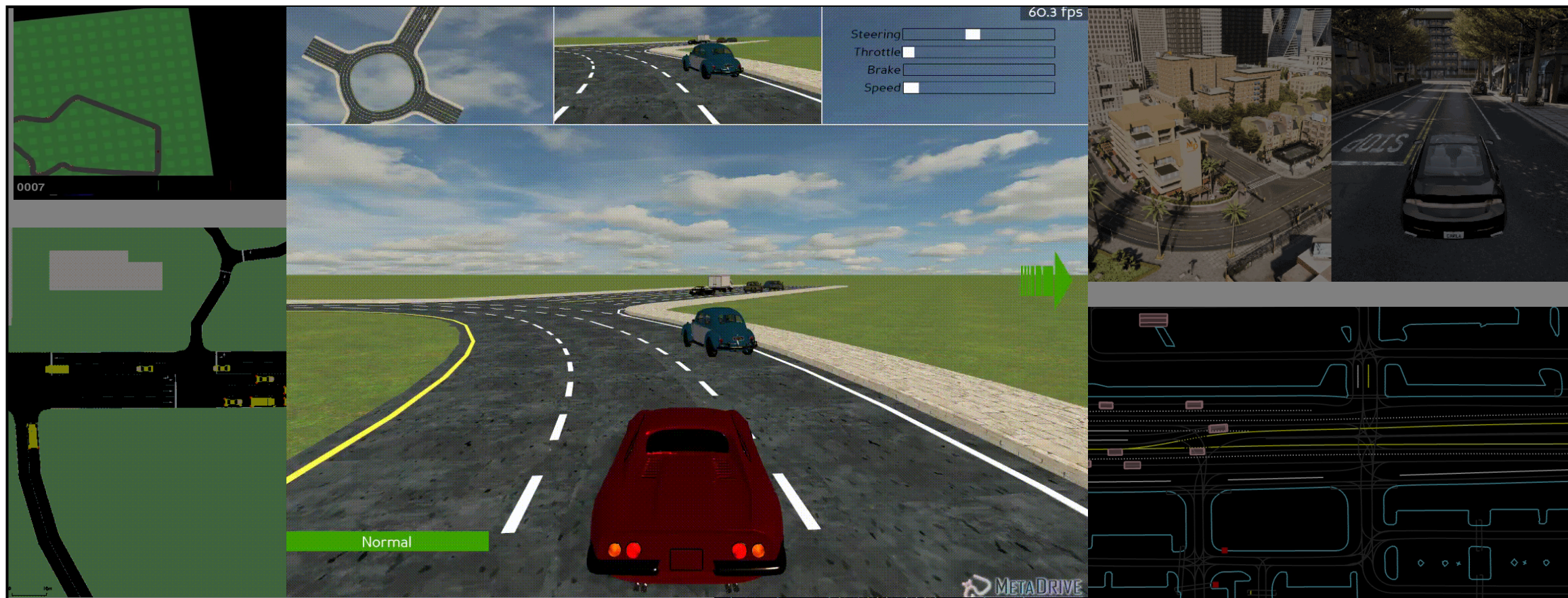
# GYMNASIUM – CAR RACING



# GYMNASIUM - HIGHWAY



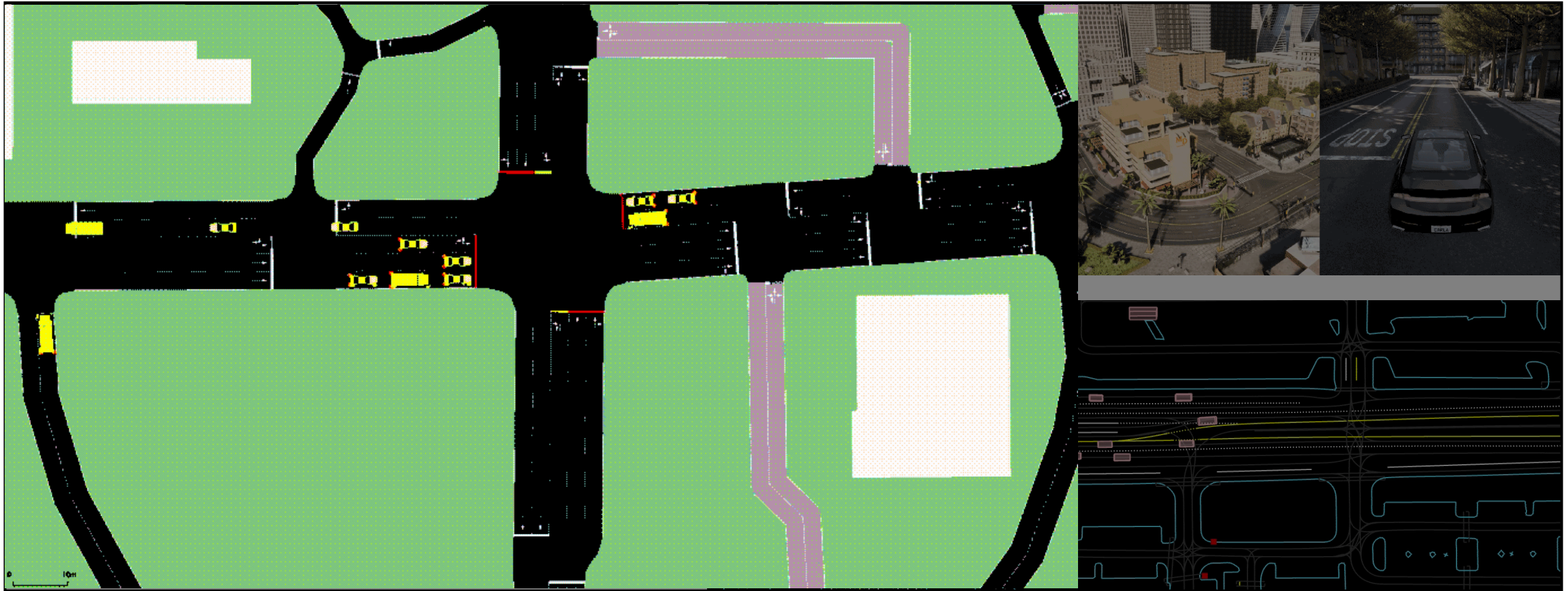
# METADRIVERSE



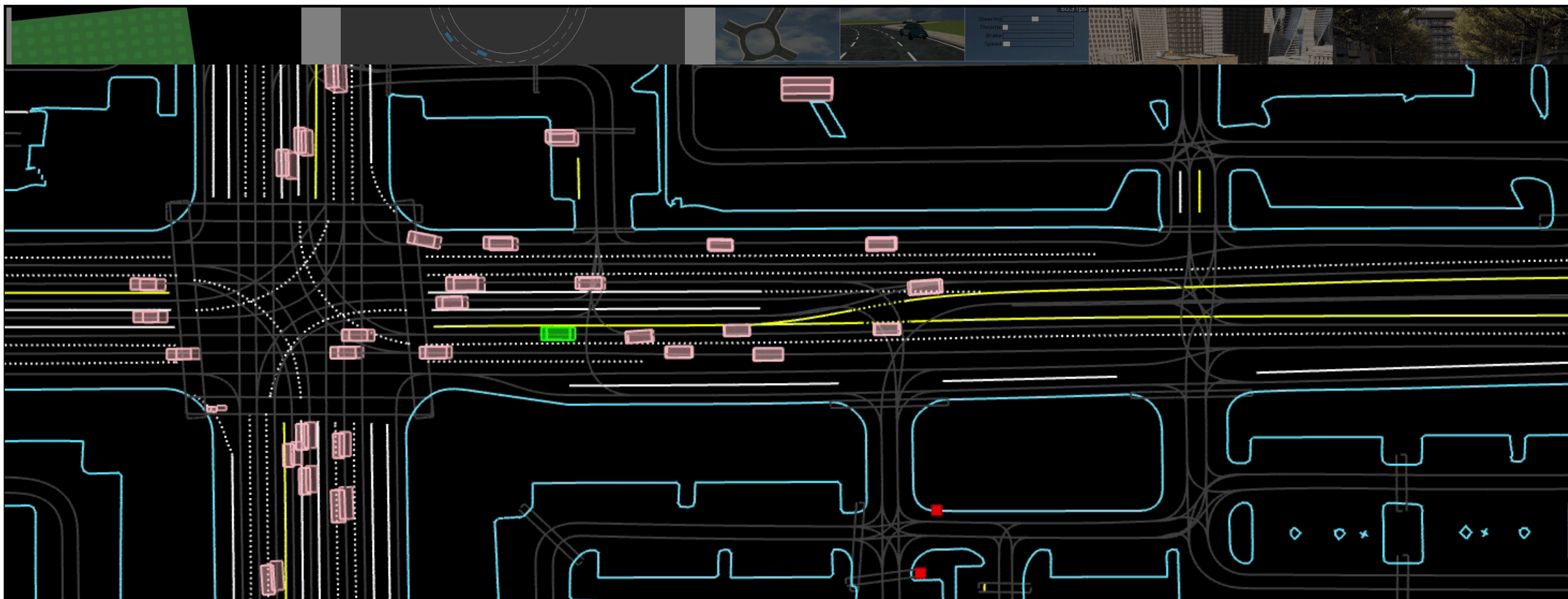
# CARLA



# SUMO



# WAYMAX



## LANCER : LEARNING AUTONOMOUS NAVIGATION IN COMPLEX ENVIRONMENT USING REINFORCEMENT

- Develop reinforcement learning systems for training autonomous driving agents
- Particularly focusing on complex environments with high entropy
- Variable environment conditions, secondary agents, and dynamics of traffic.

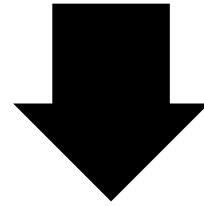
# Why RL for Autonomous Driving

Traditional robotics relies on **explicit programming**, but driving is an **implicit negotiation**.

- **Handling the "Long Tail" of Edge Cases:**
  - Classic algorithms struggle with rare, unpredictable scenarios (e.g., a distracted pedestrian, erratic lane changes).
  - RL generalizes better by learning *policies* rather than hard-coded rules.
- **Adaptive Decision Making in Dynamic Traffic:**
  - Driving isn't just about lane keeping; it's about interacting with other "agents."
  - RL is naturally suited for multi-agent environments where the ego-vehicle must predict and react to the intent of others.
- **Optimization of Conflicting Objectives:**
  - Drivers constantly balance **Safety**, **Efficiency**, and **Comfort**.
  - Through the **Reward Function**, RL can find a mathematical "sweet spot" between these trade-offs that is difficult to tune manually.
- **Reduction of Engineering Complexity:**
  - Replaces thousands of "if-else" statements with a unified learning framework.
  - Reduces the need for hand-engineered heuristics in motion planning

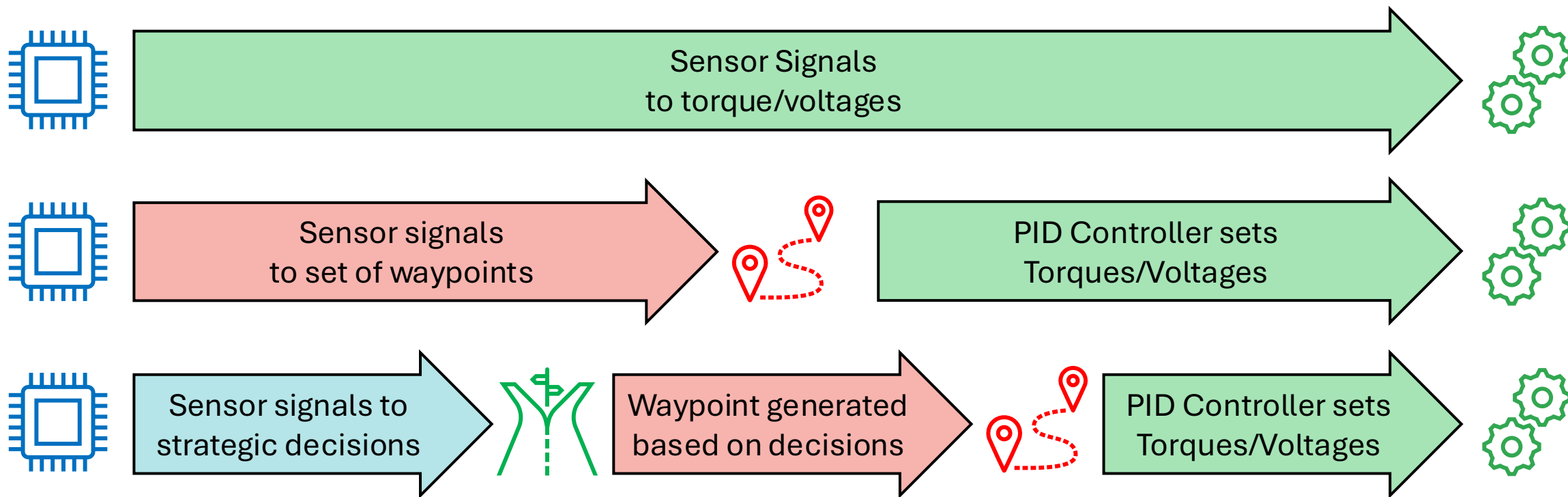
# Why RL for Autonomous Driving

We don't want to tell the car *how* to drive



We want to tell it what a *good* driving looks like

# Control Hierarchies in RL for Driving



# OBSERVATION SPACE

## Raw Sensor Information

- RGB Sensors
- LIDARs
- RADARs
- IMU/GNSS/GPS etc

## Derived Information

- Road Markings
- Pedestrian Position & Vectors
- Vehicle Position & Vectors
- Self Position & Vectors
- Anticipated Future Positions
- Drivable vs Non-Drivable Area
- Traffic Signals and Signs



# ACTION SPACE

## The Control Hierarchy

The action space determines how the agent interacts with the environment. In AV research, we generally categorize these into two levels:

### Continuous Action Space (The Standard)

- **Actions:** Throttle  $[0, 1]$ , Brake  $[0, 1]$ , and Steering  $[-1, 1]$ .
- **Pros:** Allows for high-precision, smooth driving manoeuvres.
- **Cons:** Harder to explore during training; the agent might "vibrate" the steering wheel initially as it learns.

### Discrete Action Space (The Simplified)

- **Actions:** Preset modes: [Stay in Lane, Lane Change Left, Lane Change Right, Accelerate, Brake].
- **Pros:** Simplifies the learning problem significantly; great for high-level tactical decision-making.
- **Cons:** Results in "jerky" movements; cannot handle fine-grained emergency steering.

# ACTION SPACE

## Low-Level vs. Mid-Level Control

- **Direct Control (Low - Level):** The RL agent outputs raw torque/voltage commands. This requires the agent to learn basic physics (like friction) alongside traffic rules.
- **Waypoint Tracking (Mid - Level):** The RL agent outputs a **Target Waypoint** or a **Target Speed**, which is then fed into a traditional controller (like a **PID** or **MPC**).
- **Strategic Decision Making (High - Level):** Agent chooses from a set of commands [Stay in Lane, Lane Change Left, Lane Change Right, Accelerate, Brake]

## Safety Constraints

- **Action Clipping:** Ensuring the agent doesn't "snap" the steering wheel 90 degrees at 100 km/h.
- **Temporal Consistency:** Penalizing the agent for rapid, oscillating changes in action to ensure passenger comfort.

# Reward Modelling

**Episodic Rewards**

Reach Destination

**Episodic Penalties**

Collision

# Reward Modelling

<b>Episodic Rewards</b> Reach Destination	<b>Episodic Penalties</b> Collision
<b>Immediate Rewards</b> Progress to Destination Compliance with Rules Withing Contraints Good Behaviour Smooth Driving	<b>Immediate Penalties</b> Missing Waypoints Breaking Rules Overspeeding Unnecessary Behaviours Rough Jerky Movements

# Reward Shaping – Positive Rewards

Reward Category	Specific Signal	Description	Example Reward Calculation
Mission Success	Reaching Destination	The primary goal; delivered only upon completing the entire planned route.	$R_{final} = +1000$
Progress	Reaching Waypoints	Interim bonuses for staying on the intended path.	$R_{waypoint} = +50$ (upon passing a waypoint)
Progress	Reduced Distance to Target	A continuous signal based on Euclidean or road distance reduction.	$R_{distance} = \alpha \times (d_{t-1} - d_t)$ (if $d_t < d_{t-1}$ )
Compliance	Responding to Signals	Correct behavioral response to dynamic traffic elements.	$R_{compliance} = +10$ (for each step stopped at a red light)
Constraint	Maintaining Target Speed	Rewards keeping pace with the flow of traffic within legal limits.	$R_{speed} = \beta \times e^{-(v_{target} - v_{current})^2}$
Behavioral	Maintaining Lane Center	Promotes stable, predictable, and safe lateral positioning.	$R_{lane} = +1.0$ if $( car_{center} - lane_{center}  < t)$
Comfort	Lower G-Forces	Encourages smooth, non-aggressive acceleration, braking, and steering.	$R_{comfort} = \gamma \times (G_{safe} - G_{current})$

# Reward Shaping – Negative Rewards or Penalties

Penalty Category	Specific Signal	Description	Example Penalty Calculation
Efficiency	Time Consumed	A continuous penalty to encourage the agent to find the quickest route.	$P_{time} = -0.1$ (per unit time)
Safety	Collision	A high negative reward for if vehicle collides with another object	$P_{collision} = -1000$
Navigation	Missing Waypoints	A major penalty for straying off the designated route or missing a turn.	$P_{route} = -500$
Navigation	Increased Distance from Target	A continuous penalty when the agent moves away from its goal.	$P_{distance} = \delta \times (d_t - d_{t-1})$
Compliance	Failing Traffic Signals or Crossing Speed Limit	A significant safety penalty for running red lights, ignoring stop signs, speeding etc.	$P_{violation} = -200$ or $P_{overspeed} = \epsilon \times (speed_{limit} - speed_{current})$
Behavioral	Non-mandatory Lane-Changing	Minimizes unnecessary lane-weaving, which increases collision risk.	$P_{lane\_change} =$ $-50$ (per illegal lane change) or $= -2$ (legal but non – mandatory)
Comfort	Higher G-Forces or Jerks	Penalizes aggressive inputs that degrade passenger comfort.	$R_{comfort} = \gamma \times (G_{safe} - G_{current})$

# The Role of High-Fidelity Simulation

- **Safety (Zero-Risk Training):** Training an RL agent involves thousands of crashes. In a simulator, a "collision" is just a reset button; in reality, it is a catastrophic liability.
- **Cost Efficiency:** Maintaining a fleet of sensor-heavy vehicles is expensive. Running 100 parallel instances of a simulator on a server cluster costs a fraction of the price.
- **Time Scaling:** Simulators can run **faster than real-time**. While a human driver is limited by the 24-hour day, an agent can "experience" years of driving data in a matter of weeks.
- **Edge Case Generation:** It is nearly impossible to find a specific triple-lane-change accident in the real world on demand. In simulation, you can script "The Long Tail" of rare, dangerous events repeatedly until the agent masters them.
- **Curriculum Learning:** Start the agent in a very simple environment (e.g., a straight, empty road with the goal only 5 meters away). As the agent masters the simple task, gradually increase the distance and complexity (traffic, curves, intersections). This ensures the agent "hits" the reward frequently enough to start learning.

# Core Requirements for High Fidelity

To ensure **Sim-to-Real Transfer**, where a model trained in simulation actually works on a real car, the environment must meet three criteria:

Requirement	Description
<b>Realistic Physics</b>	Must accurately model tire-road friction, mass distribution, aerodynamics, and inertia. If the "math" of the car is wrong, the RL agent will learn "cheats" that don't work in reality.
<b>Photorealistic Rendering</b>	Crucial for agents using <b>Camera/Vision sensors</b> . The lighting, shadows, and textures must mimic the real-world so the Neural Network generalizes to actual sunlight and rain.
<b>Programmable Traffic</b>	The environment cannot be static. It requires dynamic "Actors" (pedestrians, cyclists, and other cars) with varied behaviours to test the agent's tactical decision-making.

# OUR GOAL

## SIMULATION

- A customizable traffic simulation platform
- Variable entropy
- Edge Case Events

## INTEGRATION

- Multi-agent Interactions
- Competition and Cooperation
- Adversarial Agents

## SAFETY

- Safe reinforcement learning agents
- Parameterized safety in reward models
- Threat Prediction

## INCLUSIVITY

- Static to Adaptive Agents
- Entropy Modelling
- Entropy guided parameter estimation

I would love some spicy discussions

# THANK YOU FOR THE OPPURTUNITY

*DHONYOBAD - SHUKRIYA - OBRIGADO - HVALA*



**SMASH**  
machine learning for science and humanities postdoctoral program



I FEEL  
SLOVENIA



Co-funded by  
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V LJUBLJANI



**Jožef Stefan  
Institute**

**IZUM**  
Institute of Information Science, Maribor



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MINISTRY OF THE ENVIRONMENT, CLIMATE AND ENERGY  
SLOVENIAN ENVIRONMENT AGENCY

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