

Generating Logical Scenarios from Naturalistic Driving Data for Testing Autonomous Vehicle at Intersections

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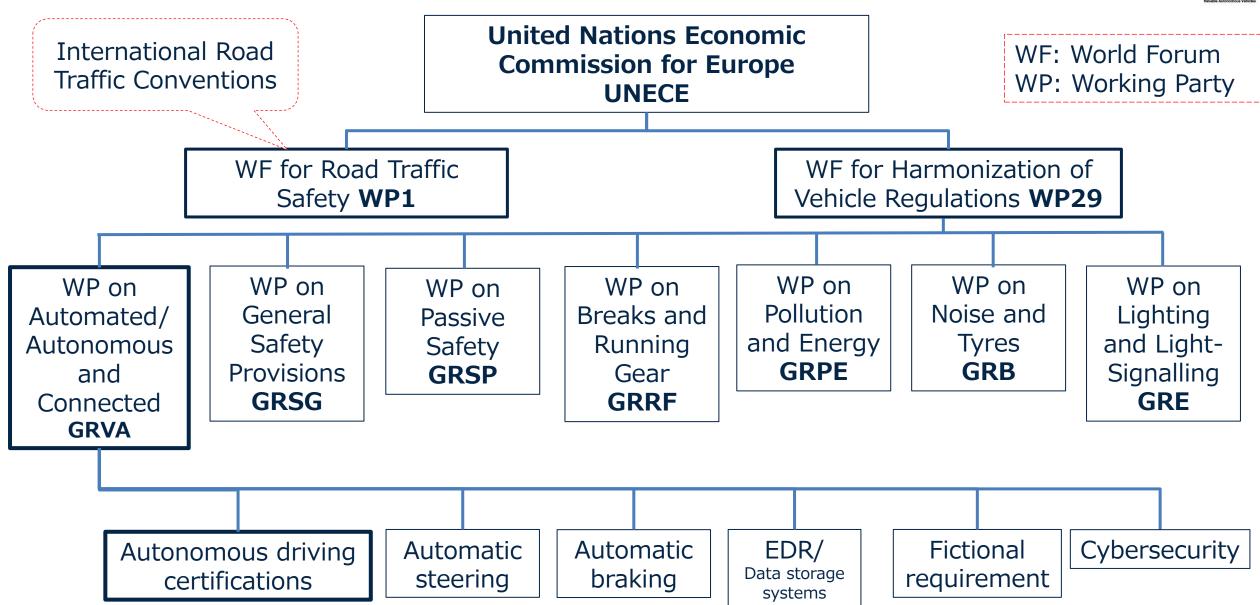


1. Introduction

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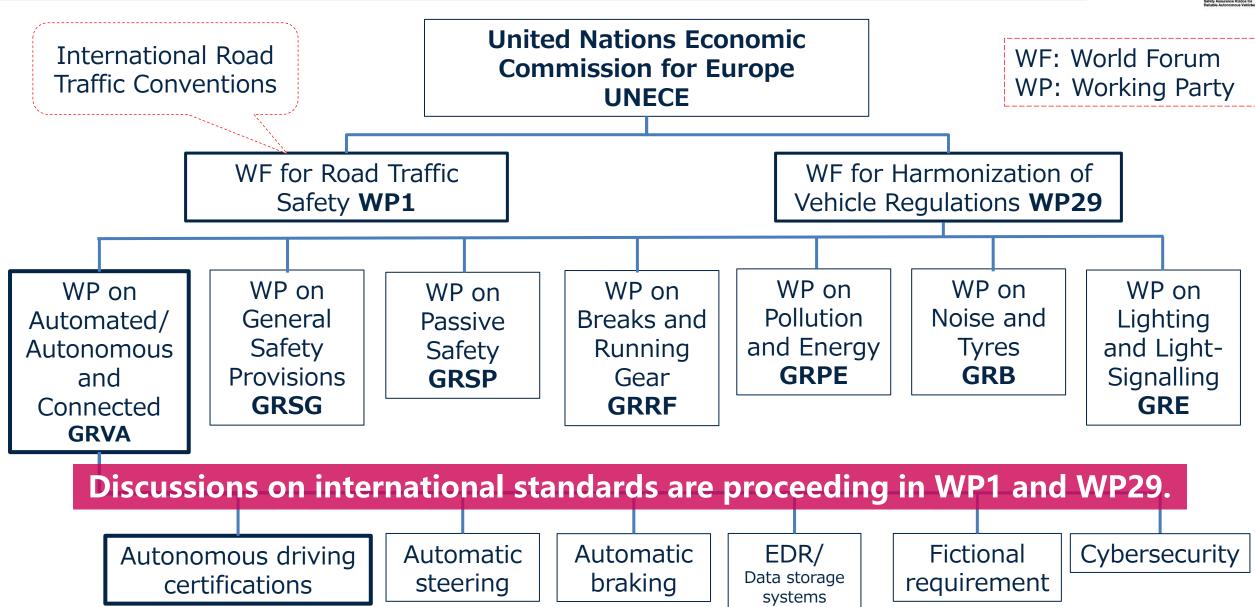
International Standards Framework for Autonomous Vehicle Safety





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International Standards Framework for Autonomous Vehicle Safety



ADS: Automated Driving System

ODD: Operational Design Domain

WP1 Road Traffic Safety Working Party

- Compliance with Road Traffic Laws
- *As there is currently no internationally unified road traffic law, compliance with each country's road traffic laws is required.

WP29 Harmonization Forum GRVA ADS shall not, while engaged, cause:

- unreasonable risks within their ODD.
- reasonably foreseeable and preventable traffic accidents within their ODD.

ADS, while engaged in their ODD, must be Safe, Predictable & Preventable-Accident-Free

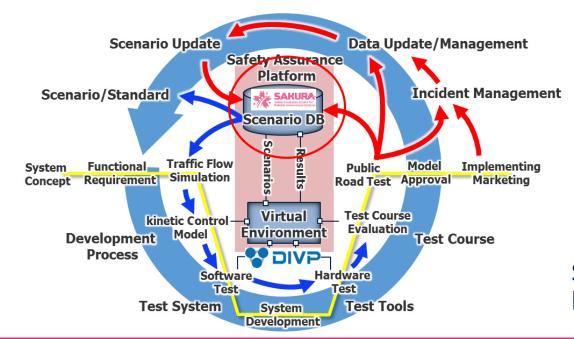
Overview of SAKURA project



- Project aims
 - Establishing a continuous safety evaluation eco-system for safer ADS development
 - Developing scenario-based safety assurance framework and platform

Coordinating standardization activities with JAMA and academia:

ISO3450X





SAKURA project: https://www.sakura-prj.go.jp/

Aims



This study aims to

 evaluate traffic data using the safety evaluation methodology developed by the SAKURA Project,

 demonstrate steps of scenario parameterization (functional, logical, and concrete scenarios)

2. Method

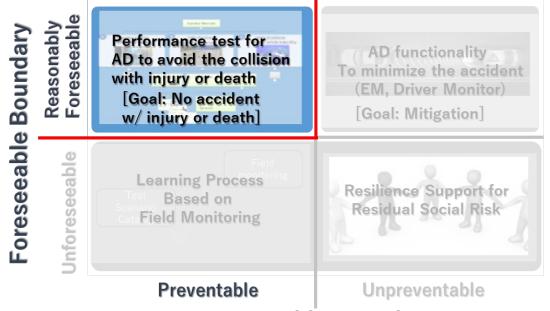
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What safety is required for ADS?



According to UN-ECE WP29 framework document, safety principle is defined as follows

Automated vehicles shall not cause any non-tolerable risk, meaning that, under their operational domain, shall not cause any traffic accidents resulting in injury or death that are reasonably foreseeable and preventable



Preventable Boundary UN/WP29: Framework document on automated/autonomous vehicles(2019) JAMA: Automated Driving Safety Evaluation Framework (Ver.3)(2022)

Requirements: efficiently evaluation the reasonably foreseeable and preventable situation

Scenario based safety evaluation



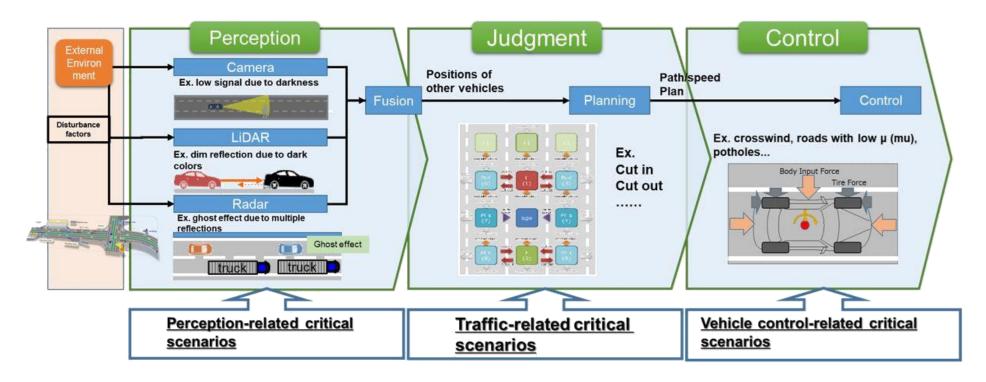
- Comprehensively extract scenes and identify the necessary parameters
- Create the scenario framework
 - Functional Scenario (FS): qualitative, described in natural language
 - Logical Scenario (LS): quantify functional scenarios by parameterizing and using data from real traffic or driving experiments
 - Concrete Scenario (CS): define test parameters using values from the parameter space

Three kind of scenarios

Concept of safety evaluation scenario



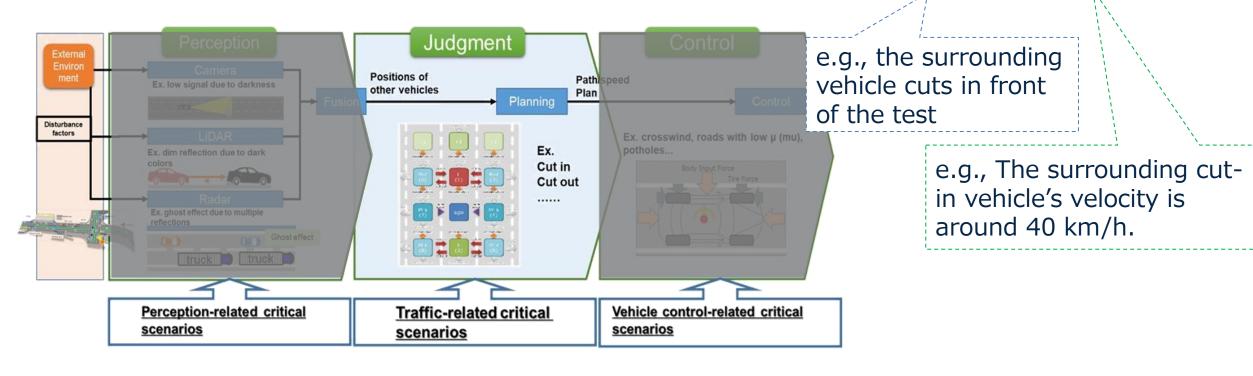
- Divide driving tasks into 3 independent tasks
- ◆ Define critical situations (scenario source) for each task
- Combination of 3 tasks are defined as "scenario" to be evaluated
- ◆ This framework is reflected in ISO 34502



Focus of this study



In this study, we focus on **judgment tasks to make FS and LS**,



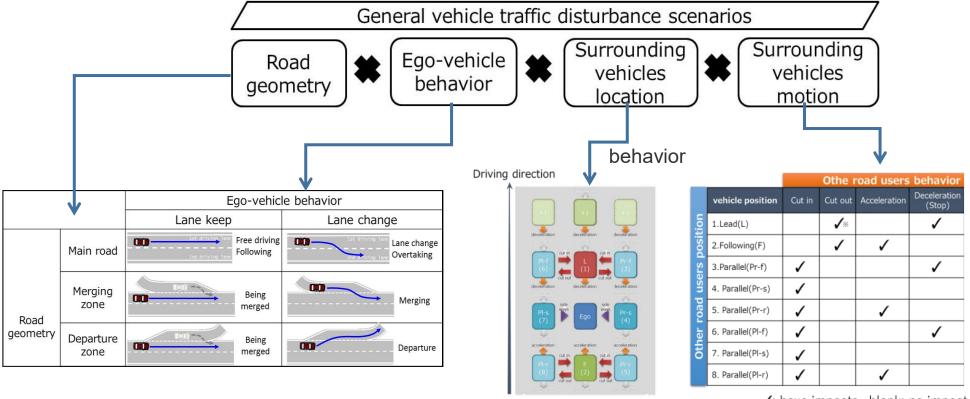
- **♦** Traffic disturbance scenarios
- Combining physical factors related to safety

Generating FS for traffic disturbance scenarios



Traffic scenarios combine physical factors related to safety.

Extract exhaustive and finite scenarios



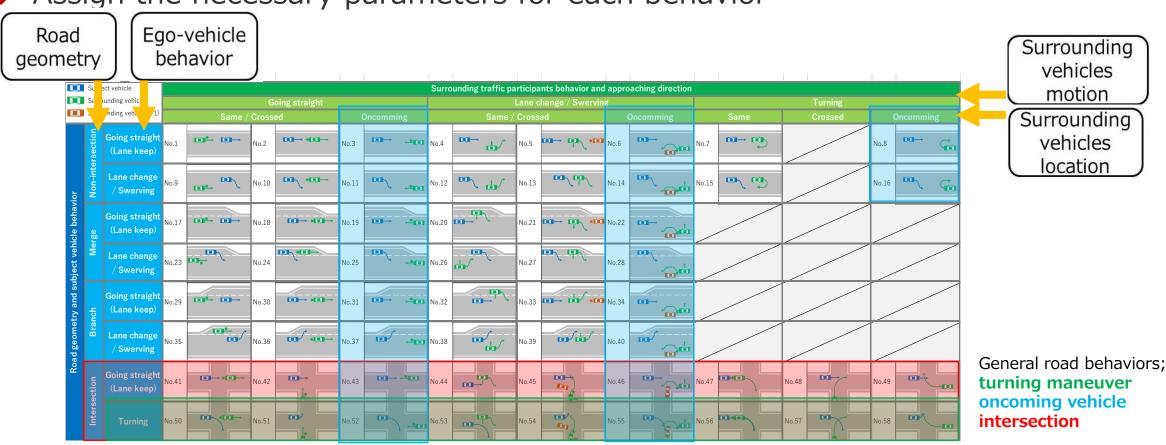
✓: have impacts, blank: no impact

Vehicle specific traffic-related critical scenario structure schematic

FS in urban environment

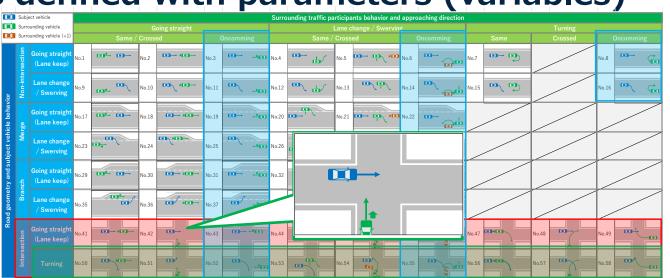


- ◆ All patterns on the public roads are covered
 - e.g. roundabout is considered as sub-categories of merging or branch
- Assign the necessary parameters for each behavior



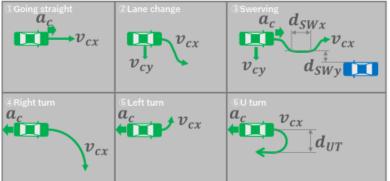
Urban 58 Scenarios





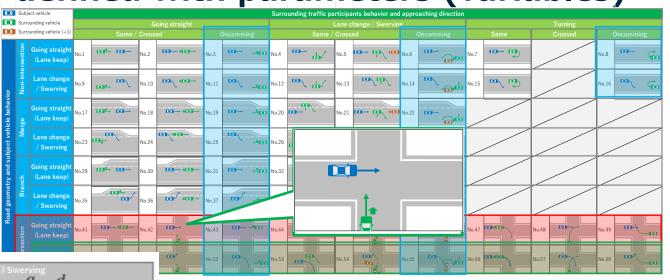


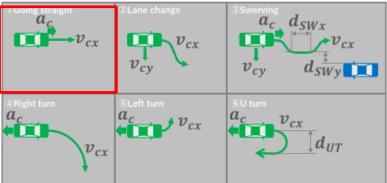




Movement of other vehicles



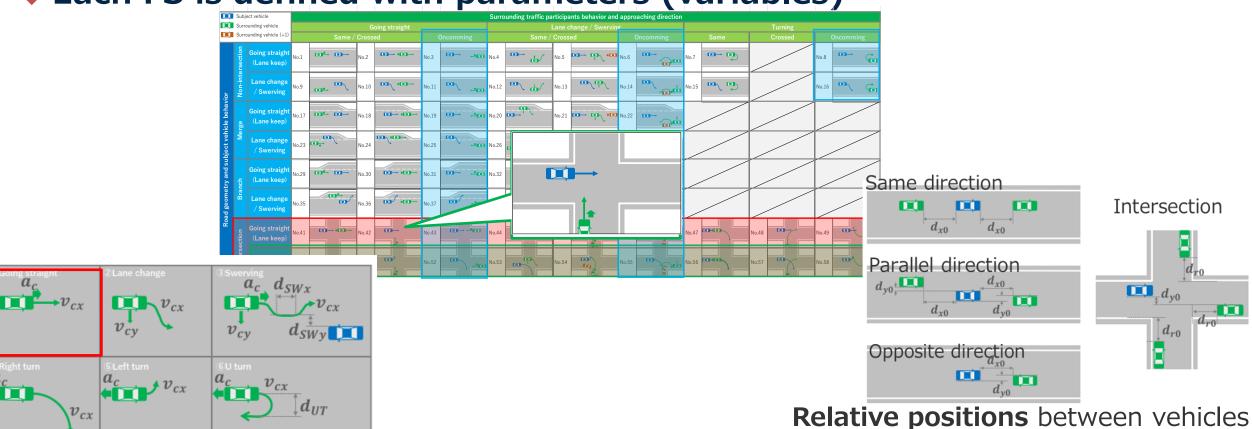




Movement of other vehicles



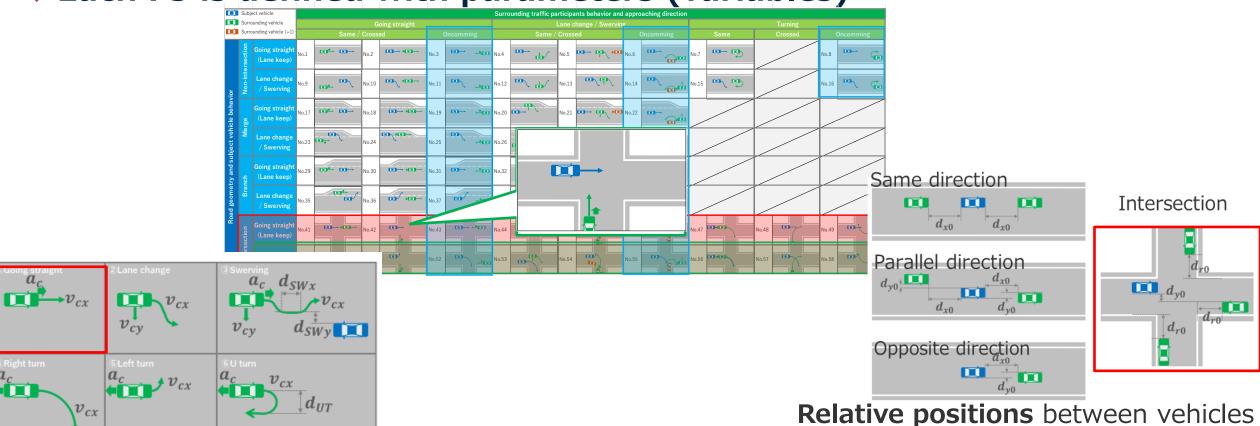
Each FS is defined with parameters (variables)



Movement of other vehicles

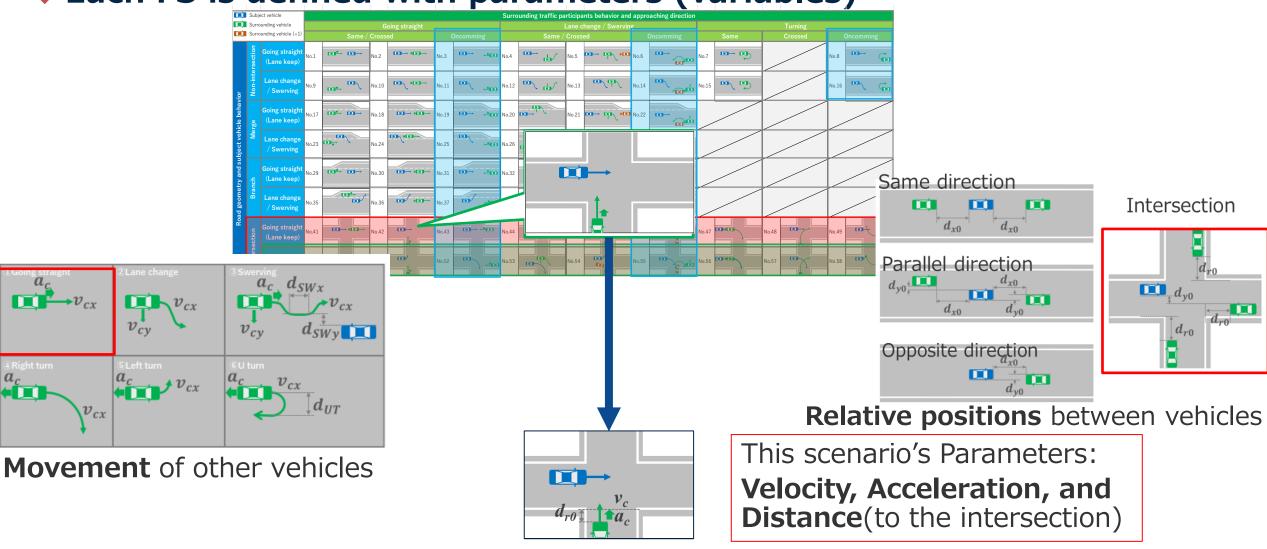


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Movement of other vehicles

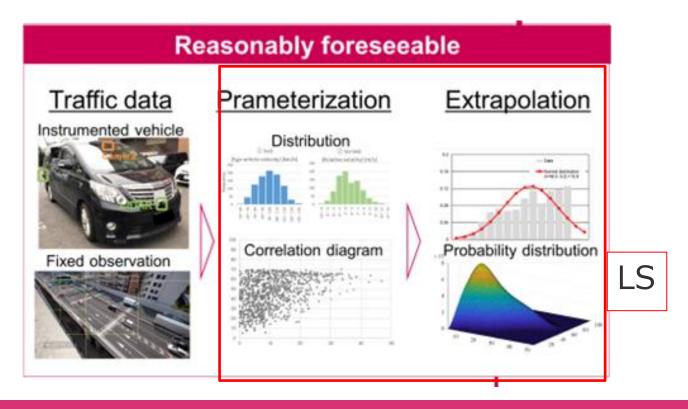




Parameter Ranges



- Evaluating parameterized scenarios by using traffic data collected by:
 - Instrumented vehicle data
 - Fixed camera observation data
 - Unmanned Aerial vehicles
 - Rule and Lows

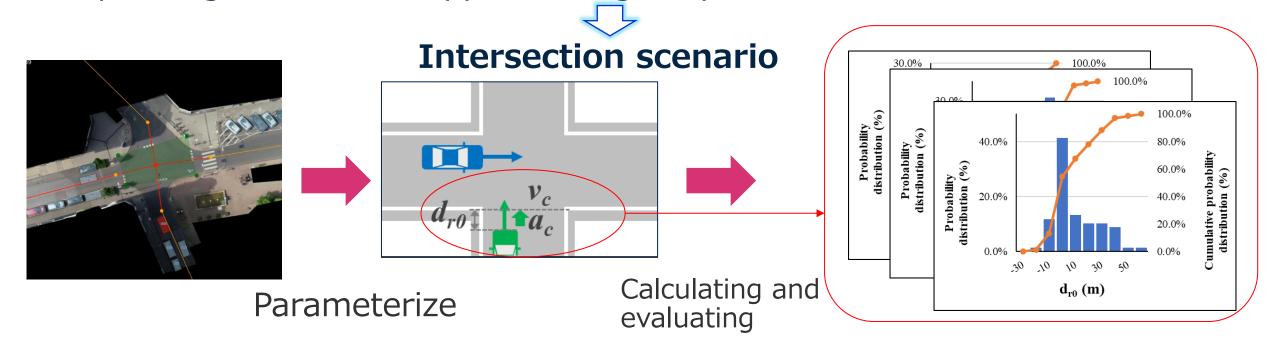


Overview of real traffic-based data analysis



Creating **LS** based on naturalistic driving data(NDD) following the **scenario-based safety evaluation approach**

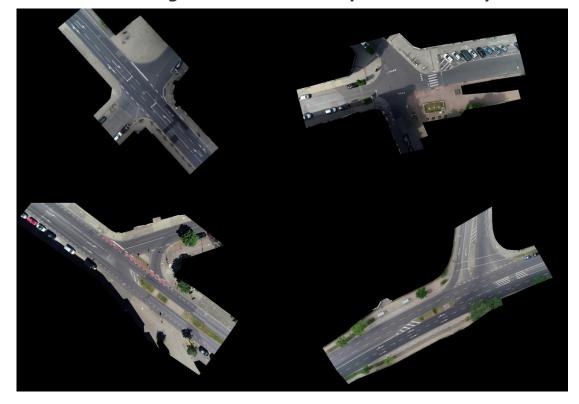
- Focusing on intersection scenario
 - high-risk areas (25% of crashes are side-impact)
- ◆ Expanding the method applied in highway scenarios (Nakamura et. al 2022)



Dataset



- ◆ Intersection Drone Dataset (inD), RWTH Aachen University
- ◆ 4 unsignalized intersections (2 crossroads, 2 T-junctions)
- ◆ Speed limit: 50 km/h
- High-resolution vehicle trajectories captured by drones

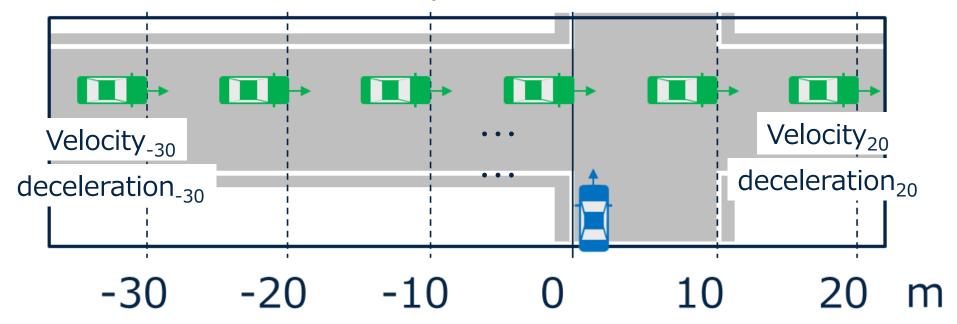


Using German urban road data

Scenario Extraction



- Two vehicles entering an intersection simultaneously
- ◆ Test vehicle vs. conflicting vehicle (focus on accelerating cases)
- ◆ 212 valid interaction cases analyzed



Extract the interaction between two vehicles at an intersection.

3. Result

J: Nesdi

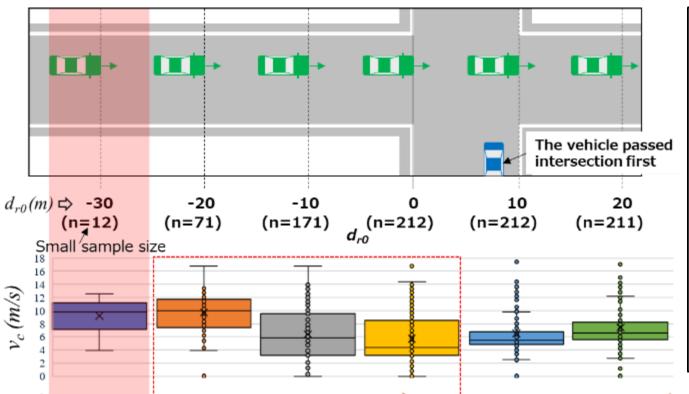
Parameter ranges: Velocity

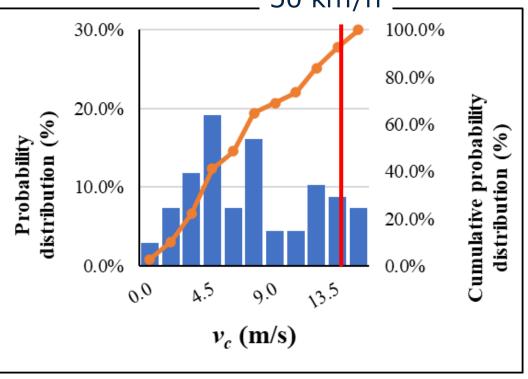


Velocity: up to ~48.8 km/h (95th percentile)

- ◆ Vehicles approach intersections close to the 50 km/h speed limit
- **◆**Decrease from 30 m to 0 m before the intersection

◆The use of high-speed edge cases in scenario generation 50 km/h



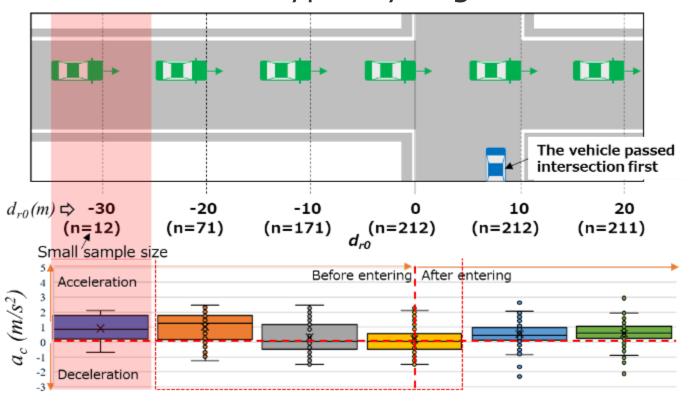


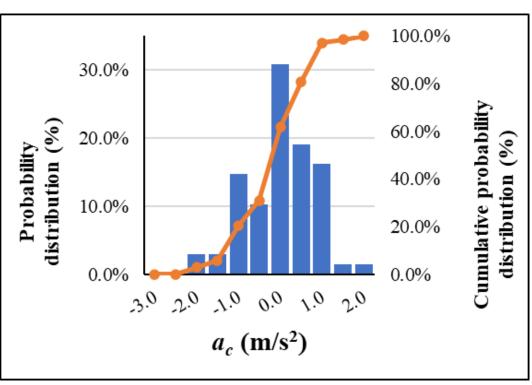
Parameter ranges: Deceleration



Deceleration: up to 2.19 m/s² (95th percentile)

- **◆**Centered near 0 m/s², with most values between -2.0 and 2.0 m/s²
- ◆ Deceleration typically begins around 20 meters before the intersection

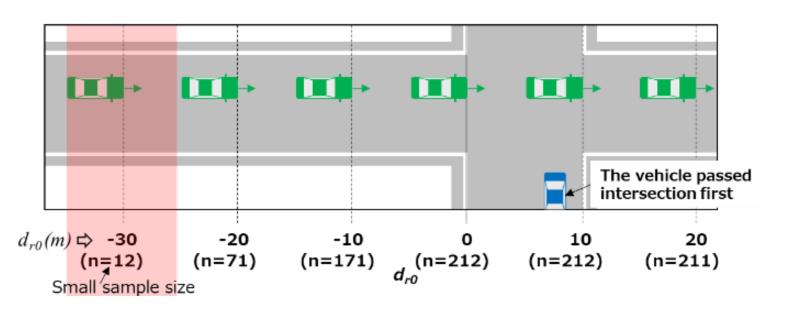


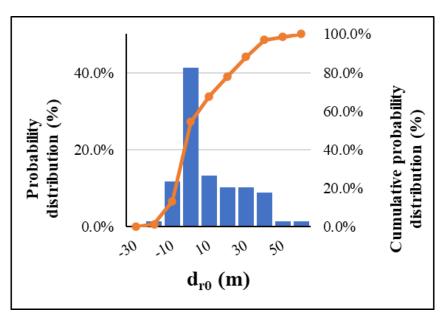


Parameter ranges: Distance



- ◆ Most vehicles were located within ±10 m of the intersection entrance
- Most vehicle interactions occur within this narrow band





Summary of Results



Velocity

- Approaching speeds often near the limit
- 95th percentile ≈
 48.8 km/h

Acceleration

- Mostly light deceleration
- Some cases of acceleration
- 95th percentile ≈ +2.19 m/s²

Distance

- Most conflicts
 within ±10 m
- Critical zone for ADS evaluation

"These results provide data-driven parameter ranges for constructing realistic and safety-relevant Logical Scenarios."

4. Discussion and Conclusion

4. Discussion and conclusion

Discussion



- The critical percentile values of the distribution functions are crucial for safety assessment.
- Most vehicle conflicts occurred within ±10 meters of the intersection entry point
- →ADS be validated in this specific range

Interaction zones

	d _{r0} =-30		d _{r0} =-20		d _{r0} =-10		$d_{r0} = 0$		$d_{r0} = 10$		$d_{r0}=20$	
%ile	n=12		n=71		n=171		n=212		n=212		n=211	
	$v_{\rm c}$	a_{c}	$v_{\rm c}$	a_{c}	v_c	a_c	v_c	a_c	v_c	a_{c}	v_{c}	a_{c}
100	12.55	2.11	16.77	2.46	16.77	2.46	16.77	2.46	17.42	2.62	17.02	2.94
95	12.42	2.03	13.33	2.19	12.80	1.99	12.63	1.97	12.97	1.76	12.93	1.47
75	11.16	1.64	11.60	1.77	9.53	1.14	8.44	0.53	6.79	0.97	8.19	1.06
50	9.84	0.84	9.99	1.29	6.08	0.08	4.33	0.02	5.56	0.45	6.58	0.60
25	7.26	0.23	8.00	0.21	3.15	-0.47	3.15	-0.50	4.82	0.12	5.58	0.26
5	4.74	-0.31	5.59	-0.54	2.07	-1.41	1.25	-1.41	3.44	-0.64	4.24	-0.59
0	3.93	-0.68	3.93	-1.24	0.29	-1.50	0.00	-1.50	0.20	-2.31	1.14	-2.14

Discussion



- Majority of vehicles slightly decelerated before entering the intersection.
- ◆ The 95th percentile acceleration reached 2.19 m/s²
- → Variability in acceleration is critical for ADS validation."

Aggressively behavior

	d _{r0} =-30		d _{r0} =-20		<i>=</i> -10		d _{r0} =0		d _{r0} =10		d _{r0} =20	
%ile	n=12		n=71		n=171		n=212		n=212		n=211	
	$\mathbf{v}_{\mathbf{c}}$	a_c	v_c	a	v_c	ac	v_c	ac	v_c	a_c	$\mathbf{v}_{\mathbf{c}}$	a _c
100	12.55	2.11	16.77		16.77	2.46	16.77	2.46	17.42	2.62	17.02	2.94
95	12.42	2.03	13.33	2.19	12.80	1.99	12.63	1.97	12.97	1.76	12.93	1.47
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0	3.93	-0.68	3.93	-1.24	0.29	-1.50	0.00	-1.50	0.20	-2.31	1.14	-2.14

Conclusion



- ◆ The proposed methodology identified critical parameter ranges for conflicting vehicles at unsignalized intersections using naturalistic driving data.
- ◆ The extracted parameter distributions provide realistic and highrisk behavior boundaries essential for scenario-based testing of ADS.
- ◆ The implicit yielding behavior observed highlights the need to model interdependent vehicle dynamics in logical scenarios.
- ◆ This data-driven framework enables the generation of physically consistent, safety-relevant test conditions for ADS validation.



Thank you for your attention

Any questions?

Acknowledgement
The Ministry of Economy, Trade, and Industry of Japan funded
this research through the SAKURA project.

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SAVE THE DATE

5 March 2026
Tokyo Japan



