

Safety impact of dedicated lanes for connected and autonomous vehicles – a traffic microsimulation study

Event: RSS 2022
Location: Athens
Date: June 2022



levitate



LEVITATE has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement No 824361.

LEVITATE

Societal Level Impact of Connected & Automated Vehicles




€ 6,4 million project funded by the European Commission under the Horizon 2020 research framework programme

Project coordinator: Loughborough University

Start date: 1st of December 2018 – 3.5 years

Partners: 12 – from 10 countries

Scope – Connectivity & Automation related services

Use case	Sub-use cases					
Urban mobility 	Single point to point shuttle	Point to point shuttle across wide area	On demand anywhere to anywhere shuttle	On demand last mile shuttle	On demand e-hailing service	
Passenger cars 	Dedicated lanes for CAVs	Road use pricing	Parking space regulations	Parking Price Policies	Automated ride sharing	Green light optimal speed advisory
Freight and logistics 	Automated urban delivery	Automated freight consolidation	Hub to hub automated transport	Platooning on urban bridges		

Manchester Network

- An arterial – A road (A6) which leads to the centre of Manchester
- A motorway M602 which leads to the centre of Manchester
- An extensive network of small roads in between

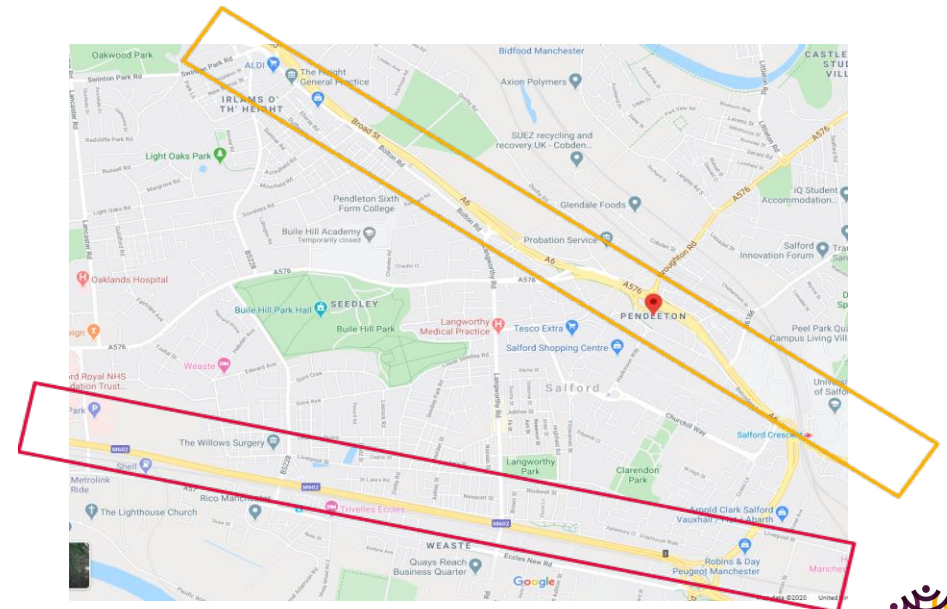
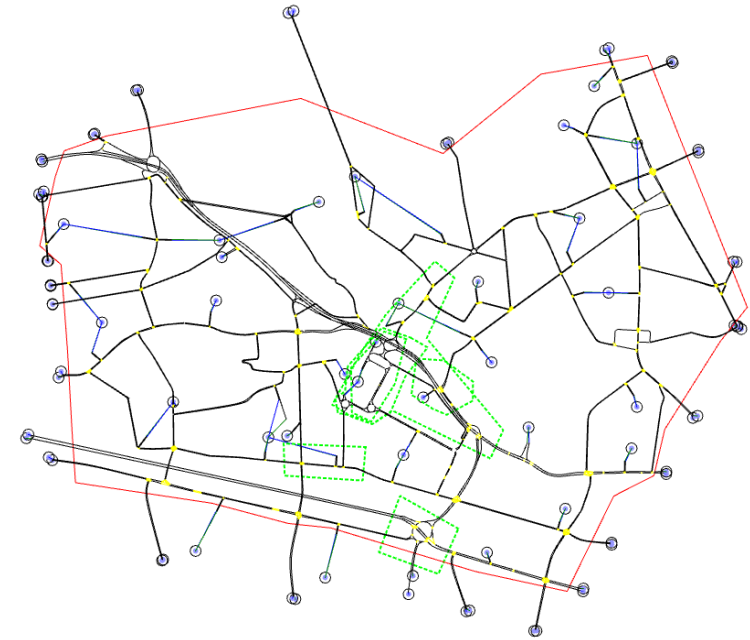
Area: 13 km²

Nodes: 308

Sections: 732

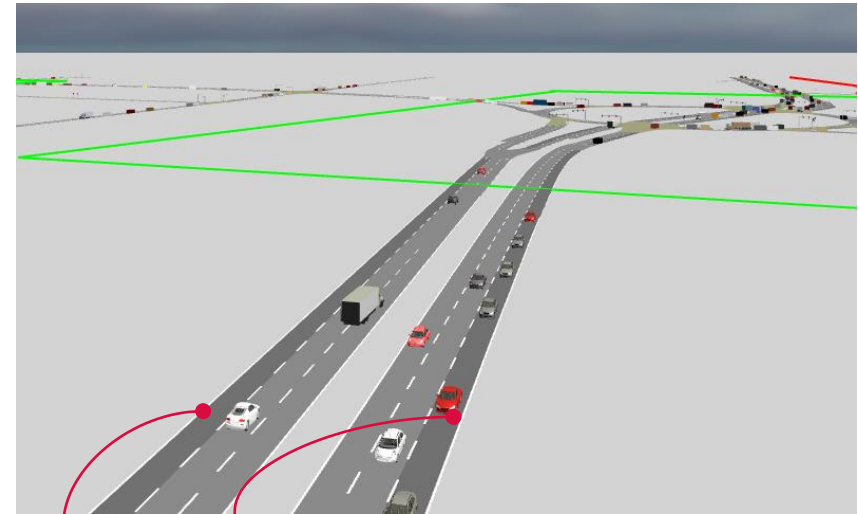
Traffic Characteristics:

- Car: 23226 trips
- LGV: 1867 trips
- HGV: 763 trips



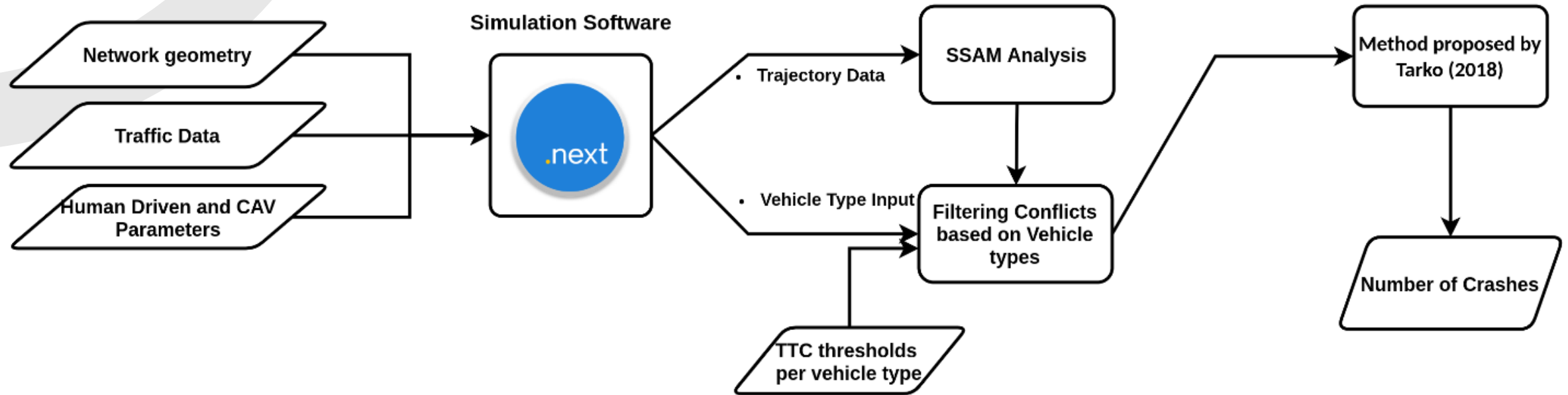
Scenarios

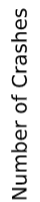
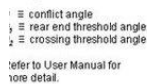
- Baseline scenario – CAV implementation without a dedicated lane
- Motorways only – CAVs use a dedicated (innermost) lane in the motorway
- Motorway & A-roads – CAVs use a dedicated (innermost) lane in the motorway and the A-road
- A-roads left most lane – CAVs use a dedicated (innermost) lane in the A-road
- A-road right most lane – CAVs use a dedicated (outermost) lane in the A-road



Dedicated
lane for AVs

Methodology





iVal = 0

SSAM

[illegible]

Post analysis and conversion conflicts to crashes based on Andrew Tarko's method

CAV parameters and deployment scenarios

Within LEVITATE, two types of CAVs were considered: 1st Generation (Gen) CAVs and 2nd Gen CAVs.

Both types are assumed to be fully automated vehicles with level 5 automation.

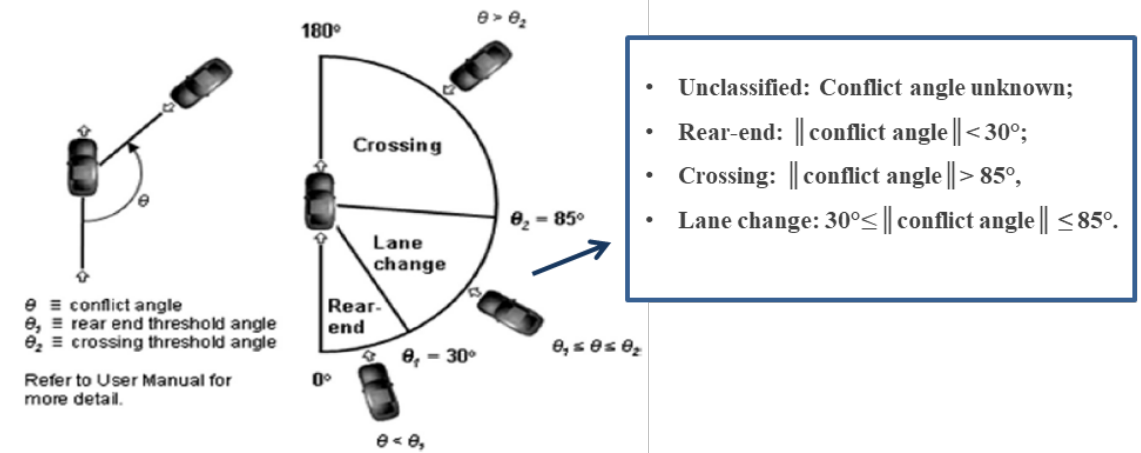
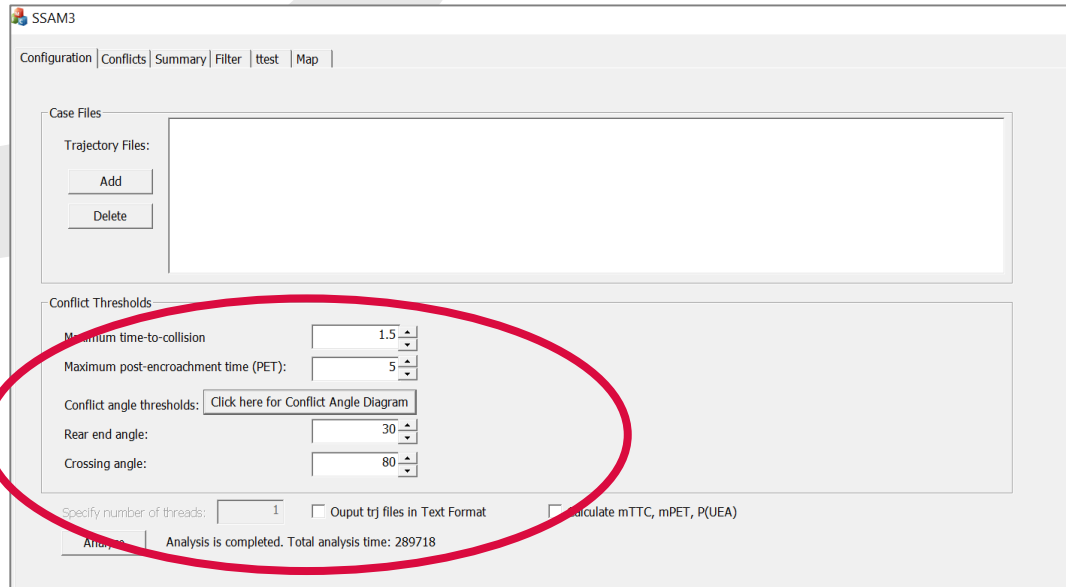
- 1st Gen: limited sensing and cognitive ability, long gaps, earlier anticipation of lane changes than human-driven vehicles and longer time in give way situations.
- 2nd Gen: advanced sensing and cognitive ability, data fusion usage, faster decision making, smaller gaps, earlier anticipation of lane changes than human-driven vehicles and less time in give way situations

CAV Deployment scenarios in LEVITATE project

Type of Vehicle	CAV Deployment Scenarios							
	A	B	C	D	E	F	G	H
Human-Driven Vehicle - passenger vehicle	100%	80%	60%	40%	20%	0%	0%	0%
1 st Gen (Cautious) CAV - passenger vehicle	0%	20%	40%	40%	40%	40%	20%	0%
2 nd Gen (Aggressive) CAV - passenger vehicle	0%	0%	0%	20%	40%	60%	80%	100%
Human-Driven LGV	100%	80%	40%	0%	0%	0%	0%	0%
LGV-CAV	0%	20%	60%	100%	100%	100%	100%	100%
Human-Driven HGV	100%	80%	40%	0%	0%	0%	0%	0%
HGV-CAV	0%	20%	60%	100%	100%	100%	100%	100%

Identifying traffic Conflicts

In SSAM, conflicts are identified based on the specific thresholds for TTC and PET and the conflict angle.



Within LEVITATE, the TTC threshold was set to 1.5s for human-driven vehicles, 1.0s for 1st Gen CAVs, and 0.5s for 2nd Gen CAVs based on existing studies (Morando et al., 2018; Sinha et al., 2020; Viridi et al., 2019).

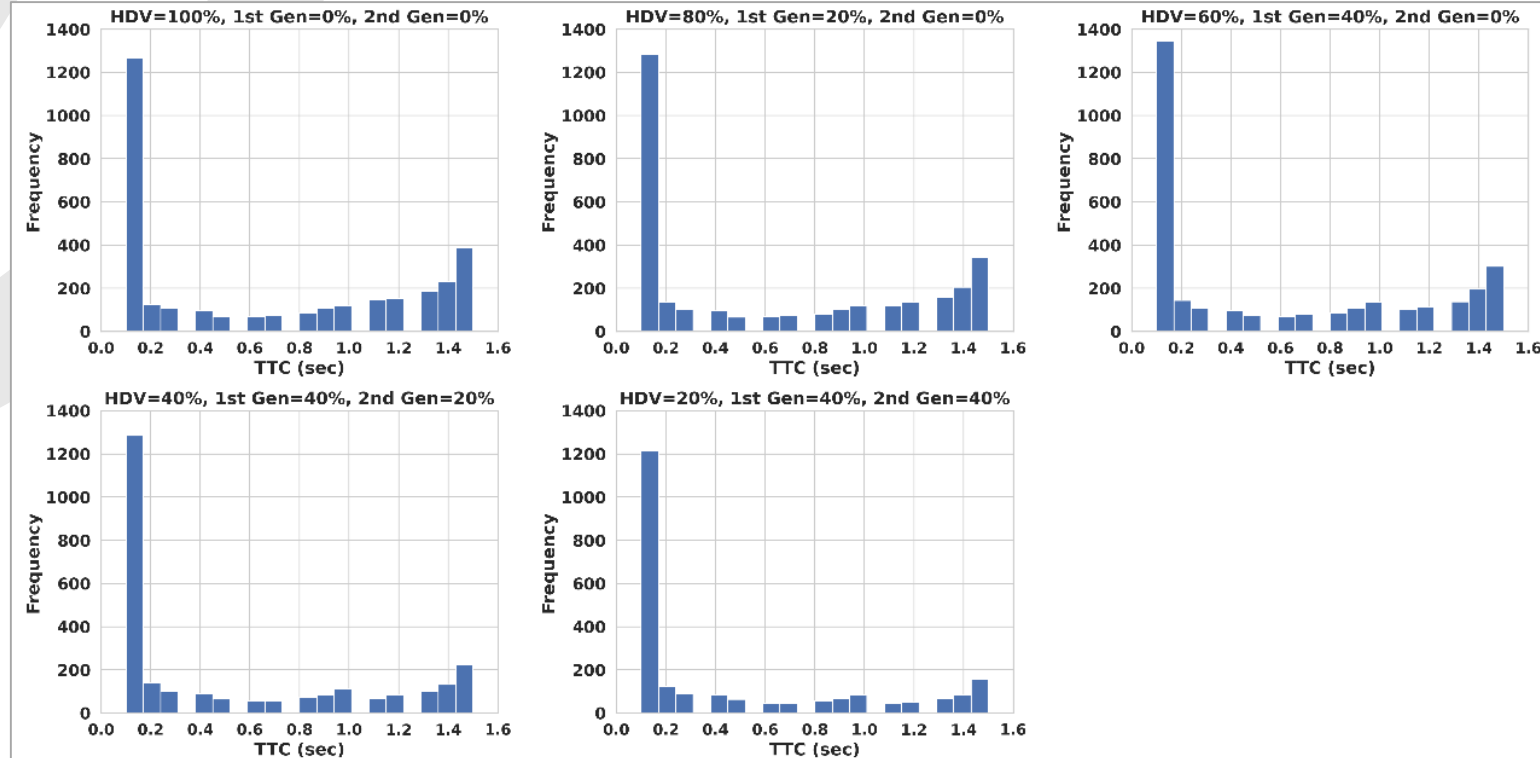
Converting the number of conflicts to number of crashes –Tarko method

Within LEVITATE, the number of conflicts was converted to the number of crashes by applying a probabilistic method proposed by Tarko (2018).

$$F(x) = \begin{cases} 1 - (1 + \theta x)^k & \text{if } \theta > 0 \\ 1 - e^{-k\theta x} & \text{if } \theta = 0 \end{cases}$$

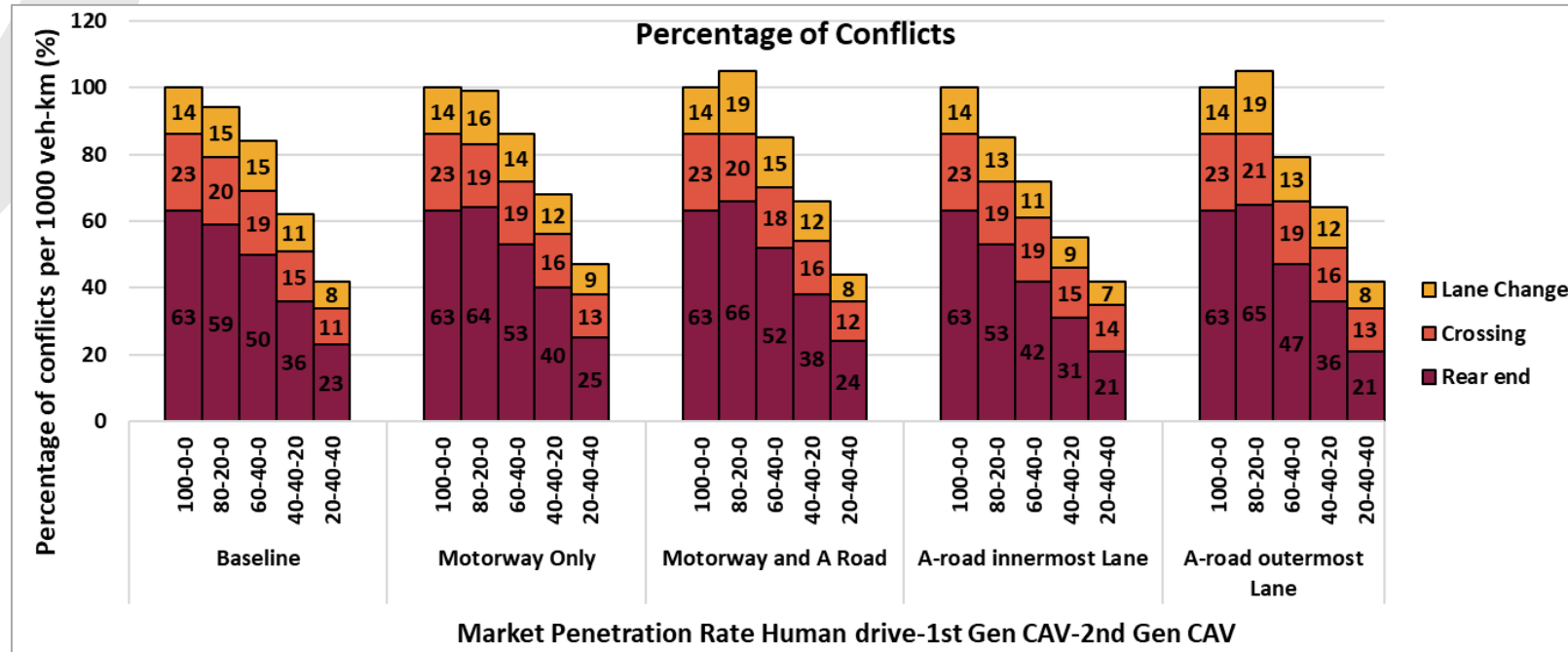
- There is unavailability of suitable empirical crash data involving automated vehicles.
- It provides a theoretical and numerical basis for justifying the Lomax distribution.
- It relies on TTC distribution and does not require crash data.

TTC distribution



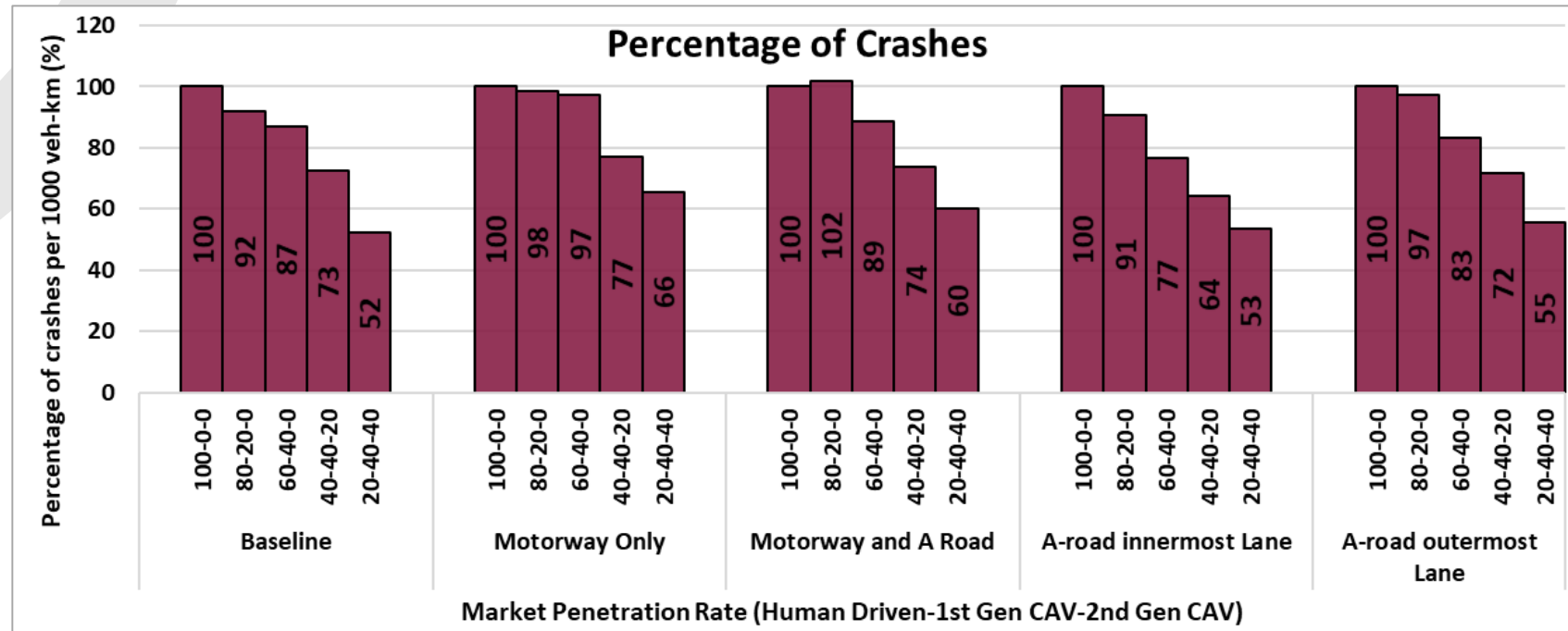
- A significant number of events are falling at very low TTC values i.e., at 0.1s.
- SSAM is likely to mark even safe interactions involving CAVs as conflicts due to shorter headways.
- Thus, the number of events with very low values of TTC (0.1) can be considered as noise/systematic bias which could be either from AIMSUN Next or/and SSAM.
- It is decided to remove the noise in the conflicts data with very low TTC values.

Identified traffic conflicts



- The overall decreasing trend for all tested scenarios in conflicts with higher MPR of CAVs can be observed.
- A reduction between 53% to 58% in conflicts can be achieved for all scenarios.
- The rear-end type conflicts contribute to the majority of the conflicts across all scenarios, this is consistent with the finding from previous studies that rear-end conflicts happen more often with autonomous vehicles in both simulation experiments and in the real world.

Estimation of crashes from conflicts



- The percentage reduction in crashes at full market penetration rate of CAVs is estimated between 34% and 48% for all tested scenarios.

Key Findings

- The results showed a decrease in crashes with increasing MPR of CAVs under all tested configurations of CAVs;
- With respect to the baseline (without dedicated lanes for CAVs), only the A road innermost scenario showed a reduction in crashes.
- Dedicated lanes are expected to increase the number of crashes per km travelled compared to the baseline scenario (no dedicated lane) at low and high MPR levels.
- When CAV penetration rates are very low ($\leq 20\%$), a dedicated lane is predicted to result in up to 17% more crashes than the baseline scenario.
- When the vehicle fleets are more equally split, a small benefit can be seen of dedicated lanes when implemented on A-level roads.

References

- Fagnant, D. J., and Kockelman, K. Preparing a nation for autonomous vehicles: opportunities, barriers and policy recommendations. *Transportation Research Part A: Policy and Practice*, 2015, 77, 167-181.
- Calvert, S. C., and van Arem, B. A generic multi-level framework for microscopic traffic simulation with automated vehicles in mixed traffic. *Transportation Research Part C: Emerging Technologies*, 2020. 110, 291-311.
- Morando, M. M., Tian, Q., Truong, L. T., and Vu, H. L. Studying the Safety Impact of Autonomous Vehicles Using Simulation-Based Surrogate Safety Measures. *Journal of Advanced Transportation*, 2018, 1--11.
- Sinha, A., Chand, S., Wijayaratna, K. P., Viridi, N., and Dixit, V. Comprehensive safety assessment in mixed fleets with connected and automated vehicles: A crash severity and rate evaluation of conventional vehicles. *Accident Analysis & Prevention*, 2020. 142(March), 105567. doi:10.1016/j.aap.2020.105567
- Viridi, N., Grzybowska, H., Waller, S. T., and Dixit, V. A safety assessment of mixed fleets with Connected and Autonomous Vehicles using the Surrogate Safety Assessment Module. *Accident Analysis & Prevention*, 2019. 131(June), 95--111. doi:10.1016/j.aap.2019.06.001
- Tarko, A. P., Estimating the expected number of crashes with traffic conflicts and the Lomax Distribution – A theoretical and numerical exploration. *Accident Analysis Prevention*, Vol. 113, No. January, 2018, pp. 63–73.
- Pu, L., Joshi, R., and Energy, S. Surrogate Safety Assessment Model (SSAM)--software user manual (No. FHWA-HRT-08-050). Turner-Fairbank Highway Research Center. 2008.