



From Automated to Manual - Modeling Control Transitions with SUMO



Leonhard Lücken, Evangelos Mintsis,
Kallirroi Porfyri, Robert Alms,
Yun-Pang Flötteröd, Dimitris Koutras
leonhard.luecken@dlr.de

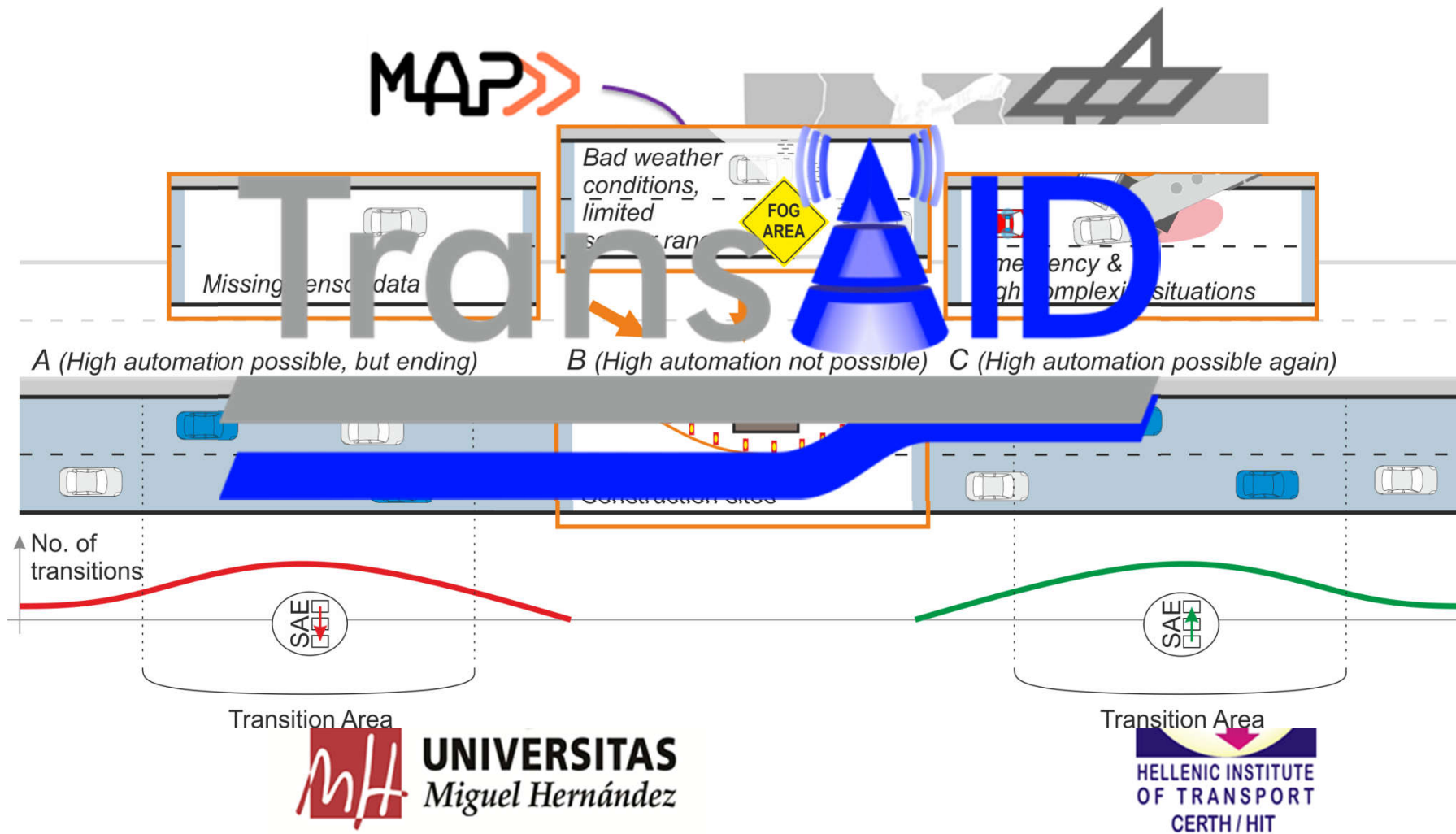


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TransAID - Transition Areas for Infrastructure-Assisted Driving



Outline

- A Model for Automated Vehicles
- Transitions of Control and a Model for human driving
- Traffic Management in Transition Areas – Two Use Cases

Models for automated vehicles

- ACC Car-Following Model [Milanés et al., 2014]
 - i. **Speed control mode:** is designed to maintain the by the driver chosen desired speed,
 - ii. **Gap control mode:** aims to maintain a constant time gap between the controlled vehicle and its predecessor,
 - iii. **Gap-closing control mode:** enables the smooth transition from speed control mode to gap control mode,
 - iv. **Collision avoidance mode:** prevents rear-end collisions.

Parametrized Lane Change Model

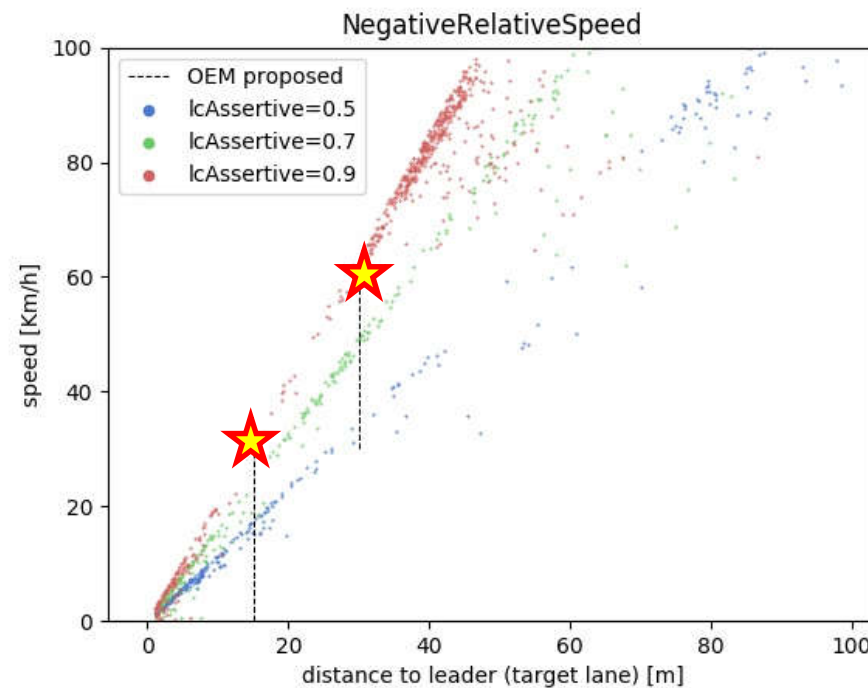
i. Variance based sensitivity analysis

→ Influential lane change calibration parameters

Speed Range [0, 100] (km/h)			
Parameter	Leader gap (ego lane)	Leader gap (target lane)	Follower gap (target lane)
Sensitivity Index	S_i [%]	ST_i [%]	
<i>lcStrategic</i>	0.39	0.62	
<i>lcKeepRight</i>	1.08	0.83	
<i>lcSpeedGain</i>	0.90	8.12	
<i>lcAssertive</i>	59.15	77.03	

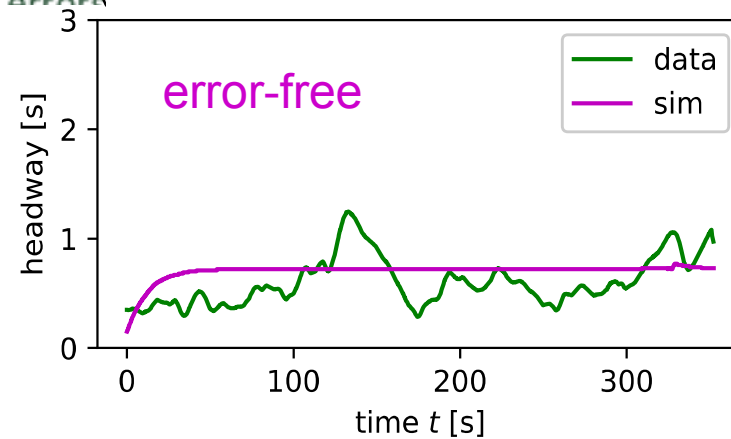
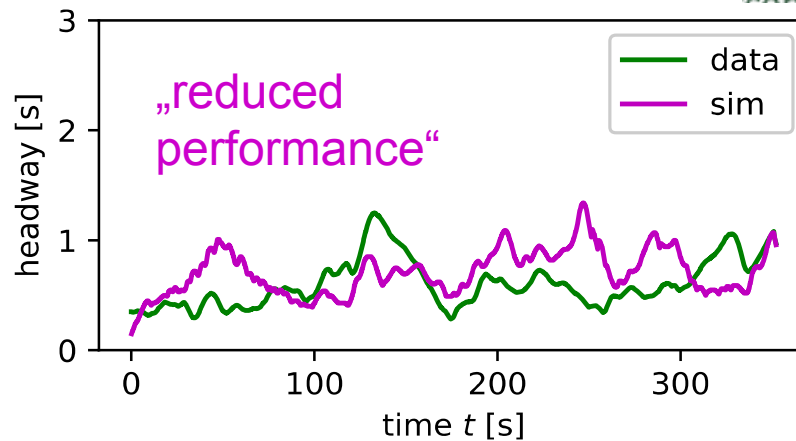
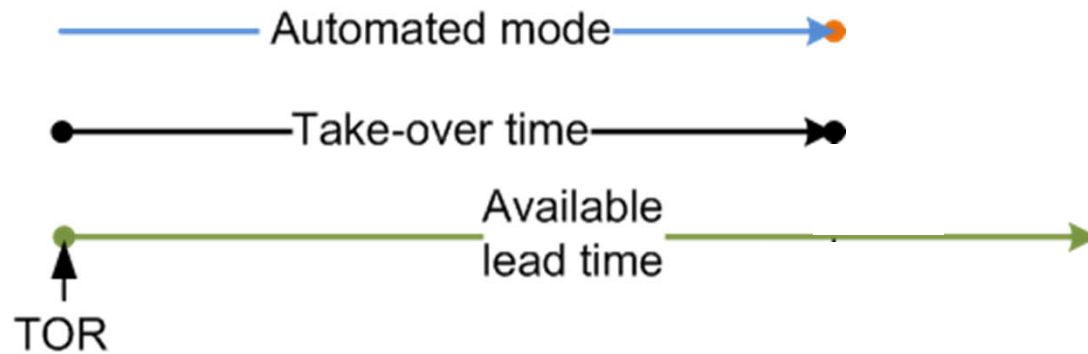
ii. SUMO lane change output vs HMETC lane change data

→ Reconciliation



ToC / MRM Model

(a) Successful ToC



Imperfect Driving

General CF Model:

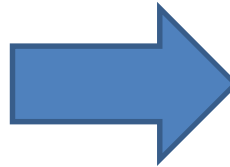
$$\dot{x}(t) = v(t)$$

$$\dot{v}(t) = a(\Delta x(t), \Delta v(t))$$

Perceived quantities:

$$\Delta \tilde{x} = \Delta x + \eta_x$$

$$\Delta \tilde{v} = v + \eta_v$$

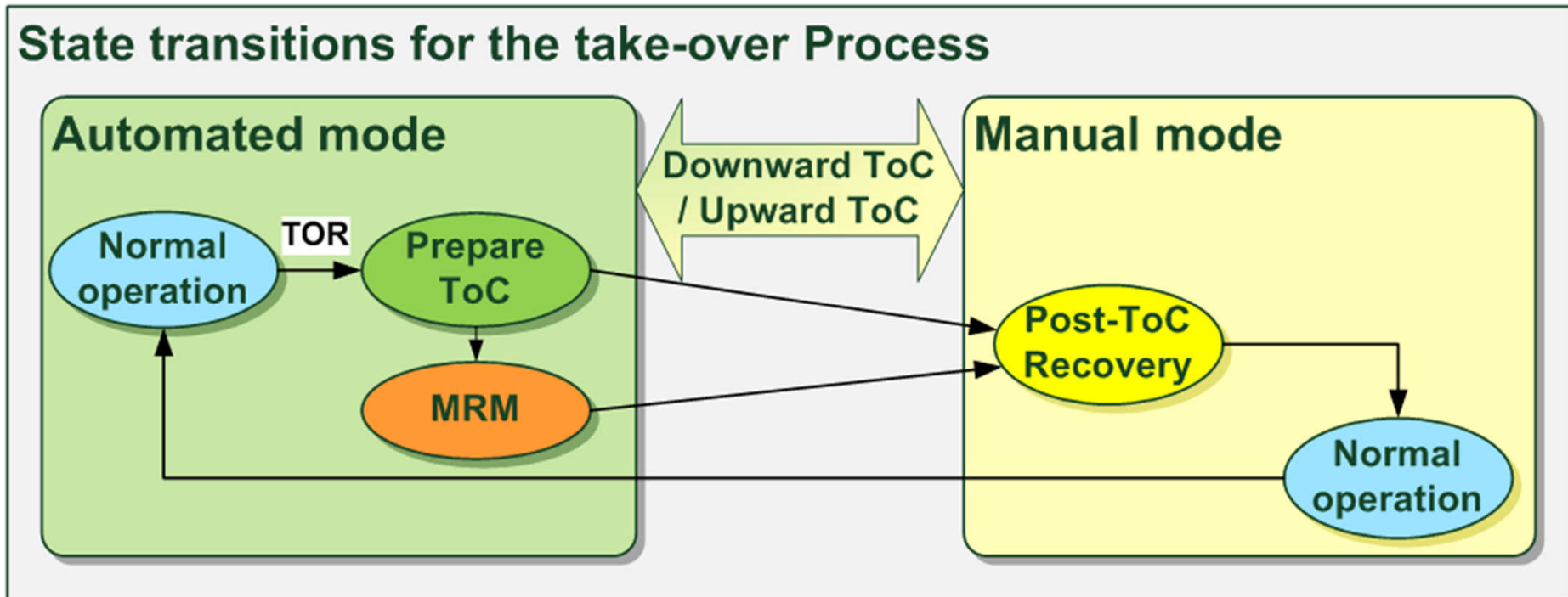


Erroneous CF Model:

$$\dot{x}(t) = v(t)$$

$$\dot{v}(t) = a(\Delta \tilde{x}(t), \Delta \tilde{v}(t))$$

ToC / MRM Model

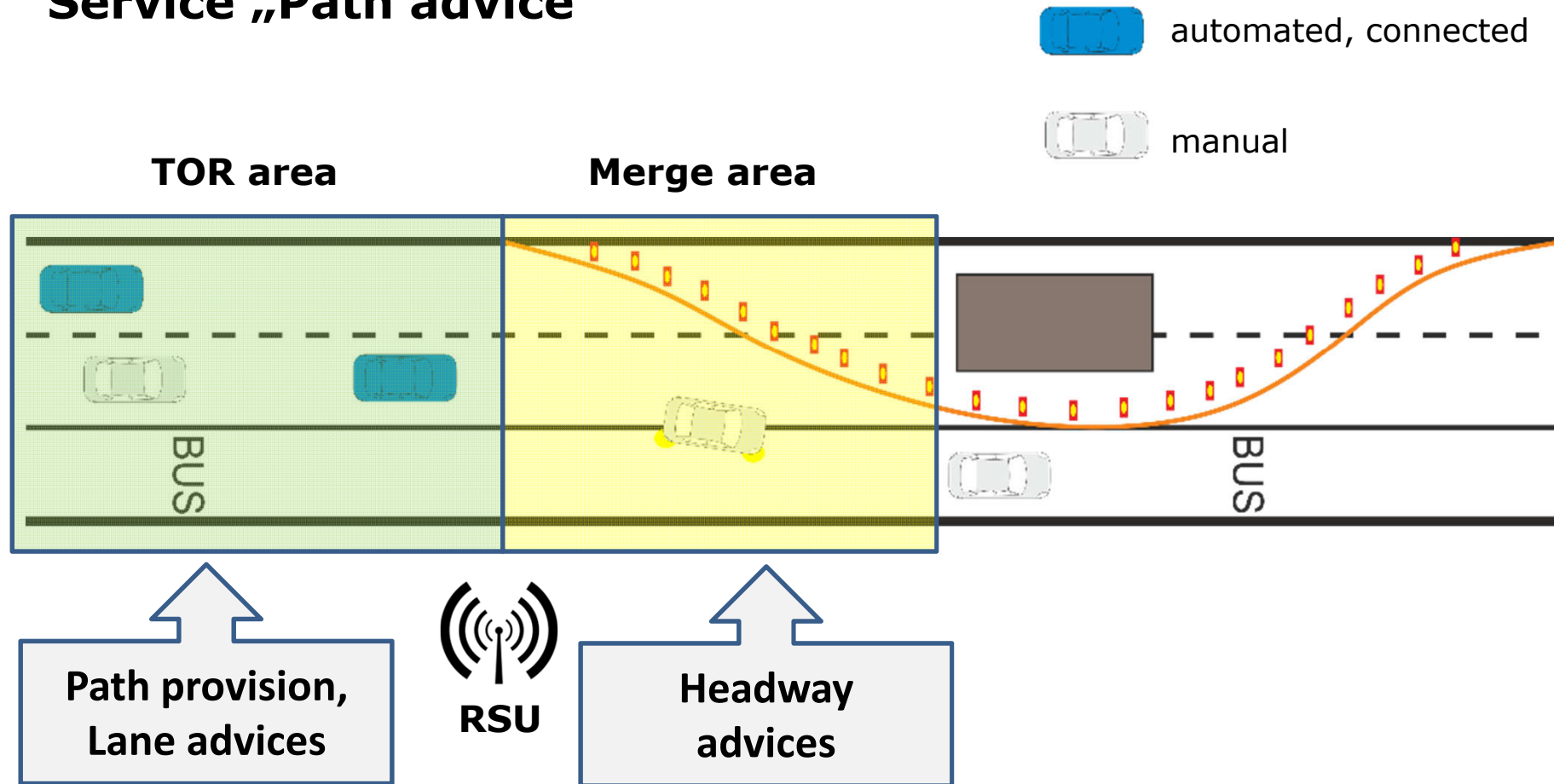


- <https://sumo.dlr.de/wiki/Car-Following-Models/ACC>
- [https://sumo.dlr.de/wiki/ToC Device](https://sumo.dlr.de/wiki/ToC_Device)
- [https://sumo.dlr.de/wiki/Driver State](https://sumo.dlr.de/wiki/Driver_State)

Traffic management in Transition Areas

Scenario 1

Service „Path advice“



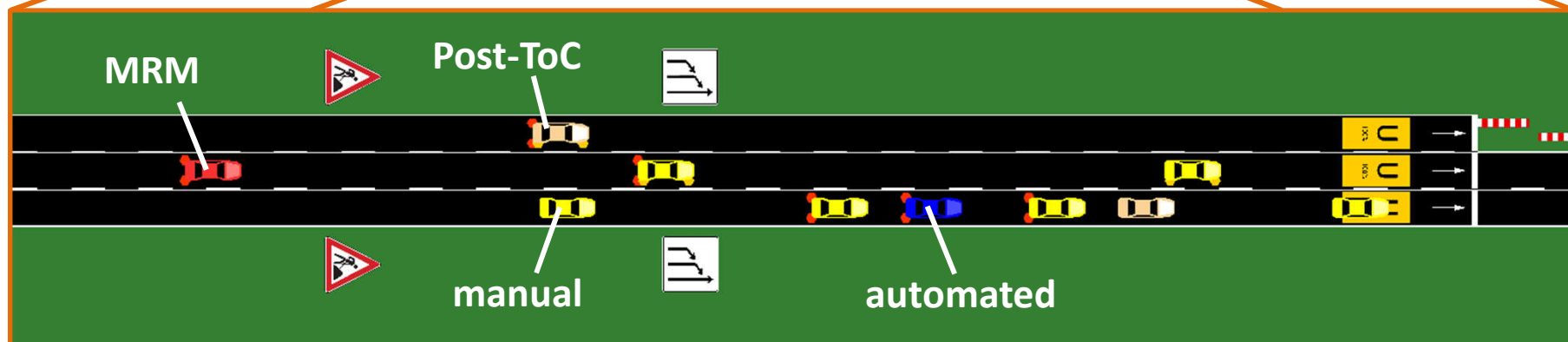
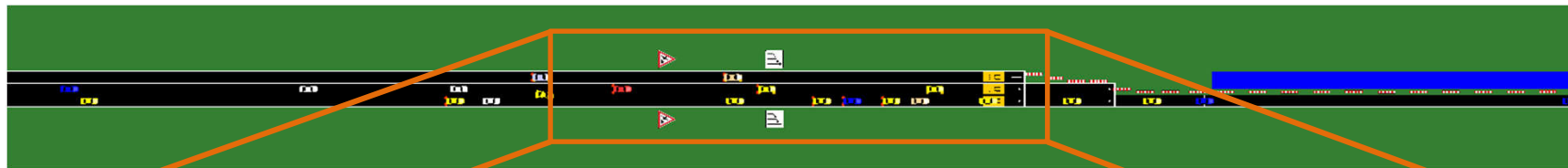
Traffic management in Transition Areas

Scenario 1

Color ~ speed



Color ~ ToC state

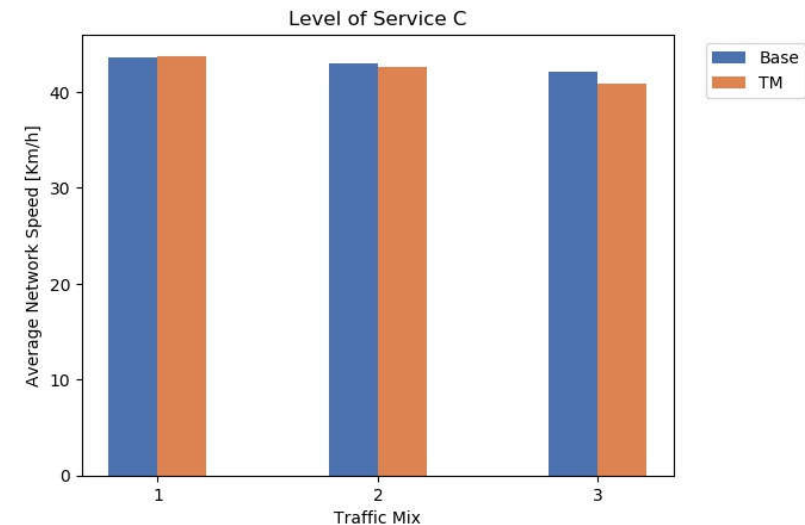
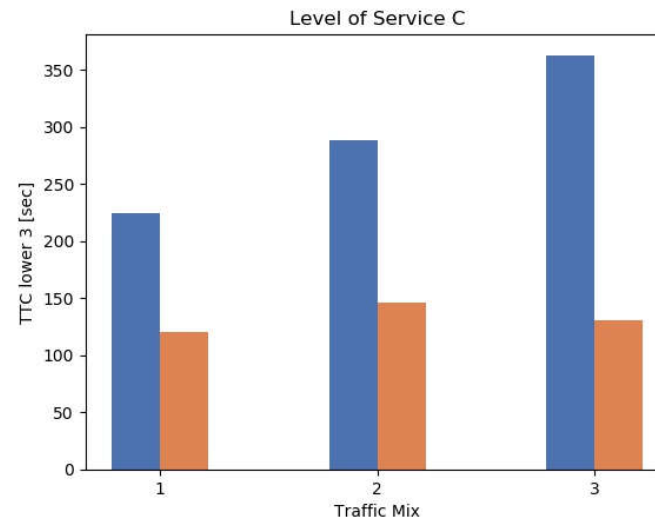
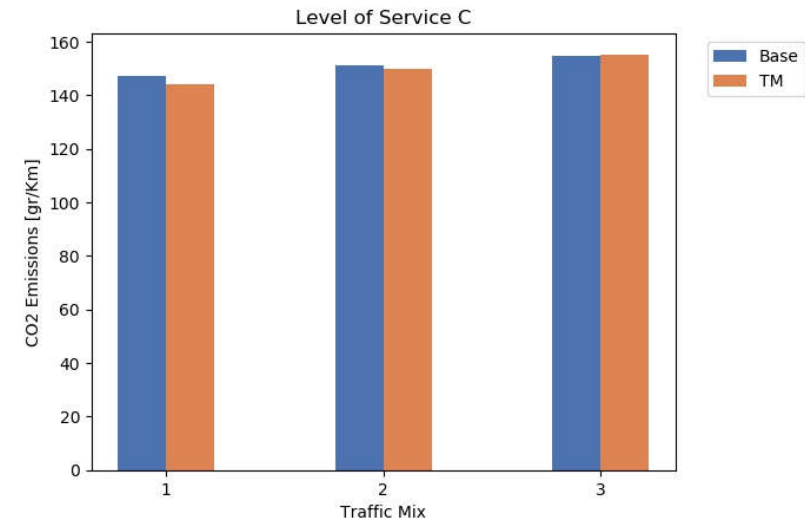


Traffic management in Transition Areas

Scenario 1

Results

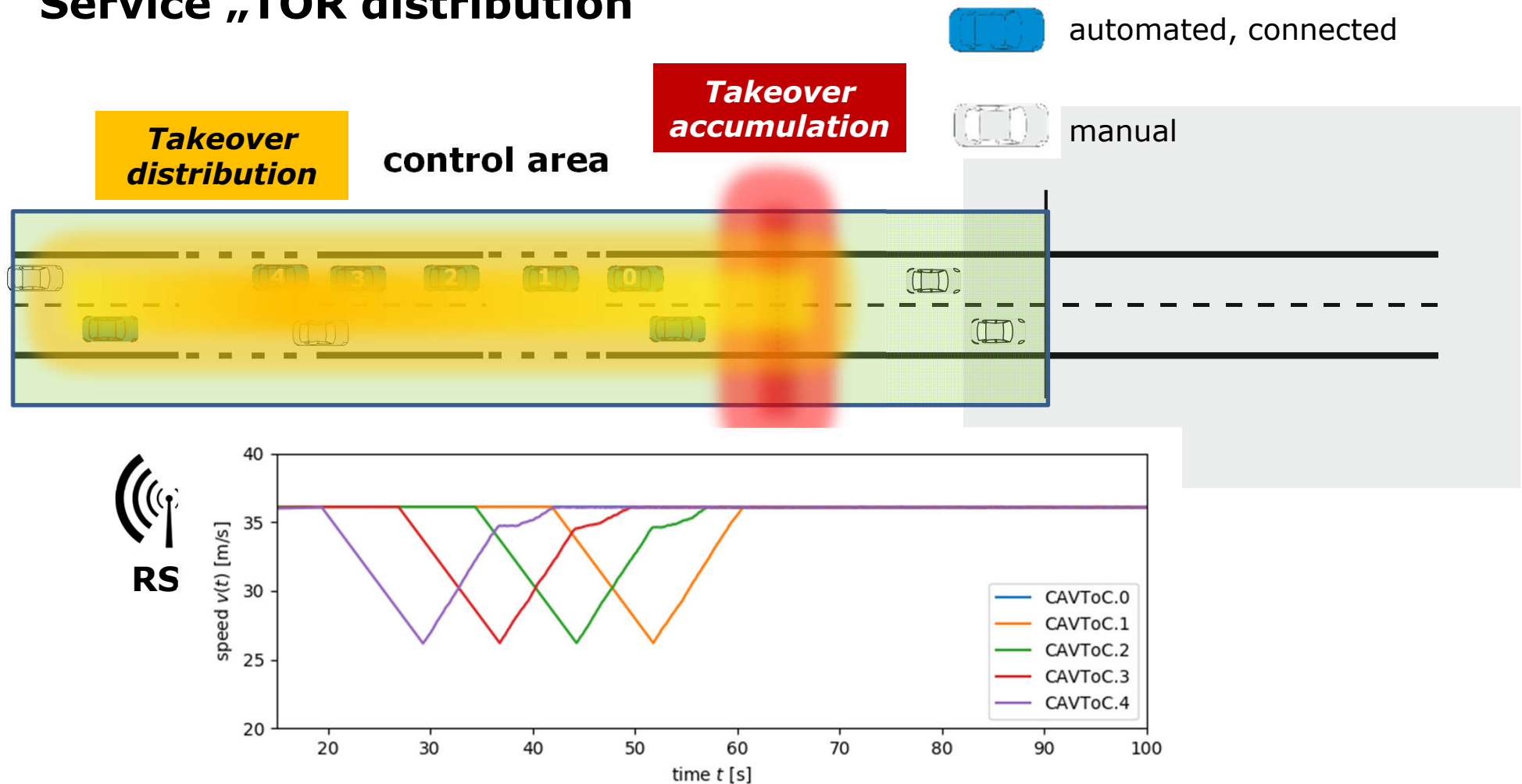
- 1h random vehicle flow (LoS C \sim 1155 veh/h)
- Fleet mixes (MV-AV):
 - mix 1: 70-30
 - mix 2: 50-50
 - mix 3: 20-80



Traffic management in Transition Areas

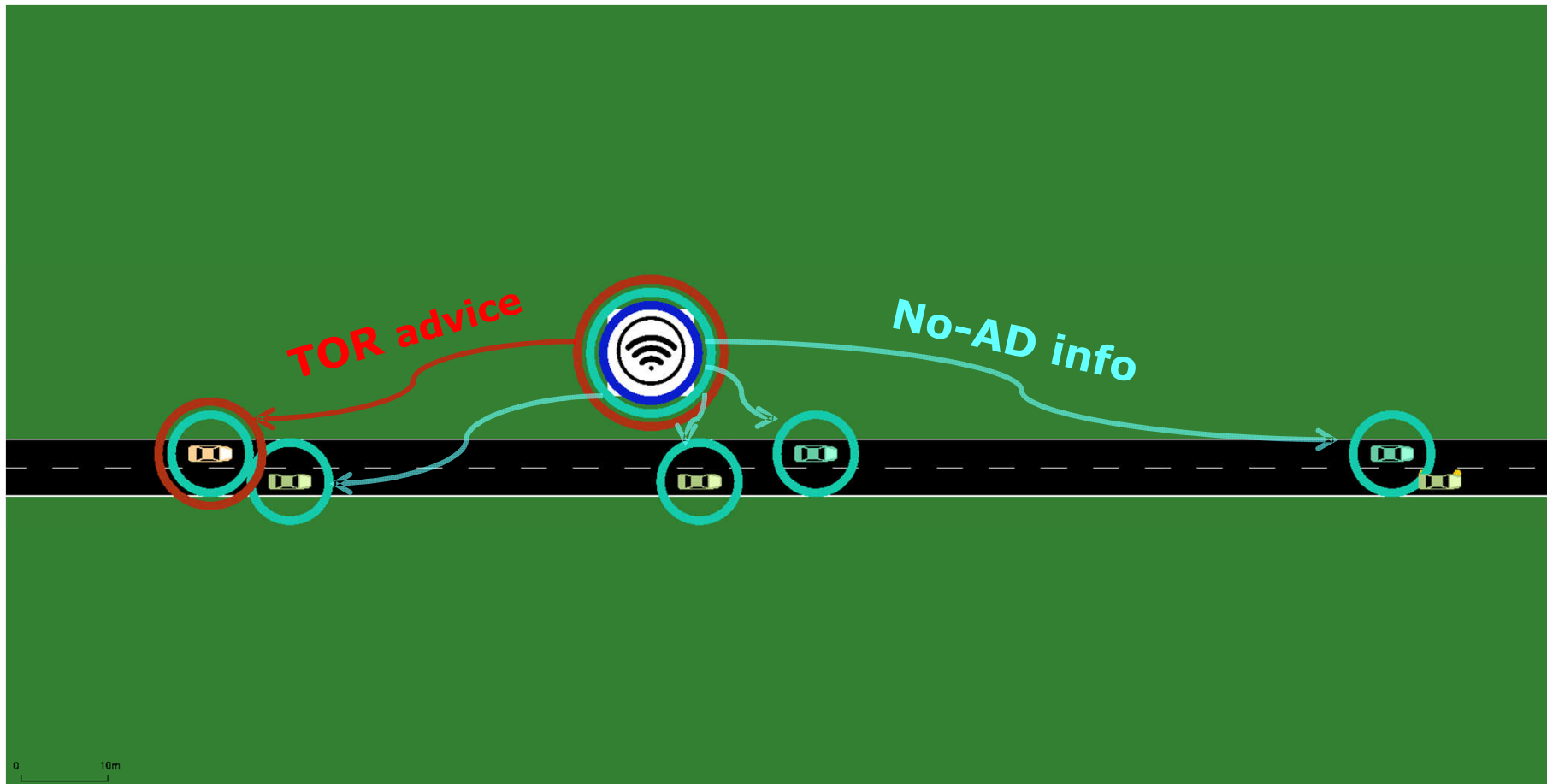
Scenario 2

Service „TOR distribution“



Traffic management in Transition Areas

Scenario 2



https://sumo.dlr.de/wiki/TraCI/Change_Vehicle_State
https://sumo.dlr.de/wiki/TraCI/Change_PoI_State
https://sumo.dlr.de/wiki/TraCI/Change_Polygon_State

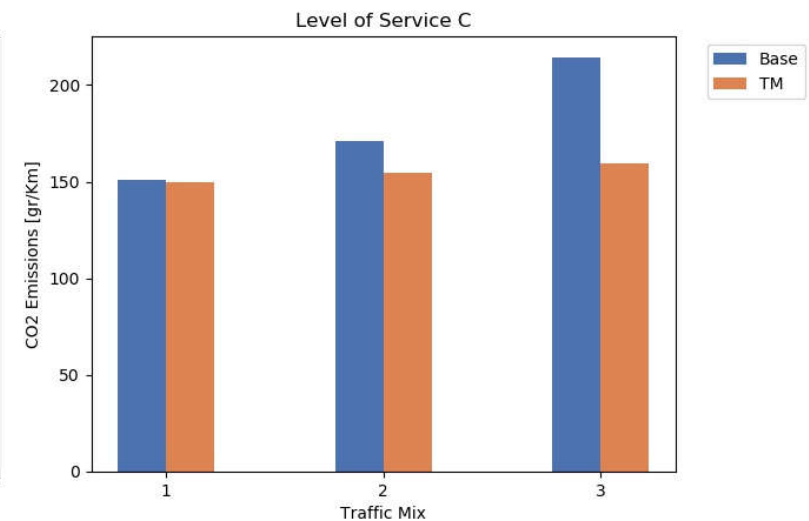
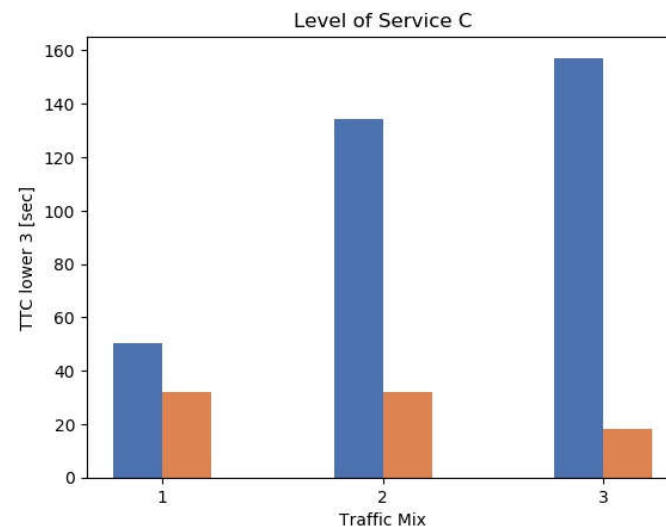
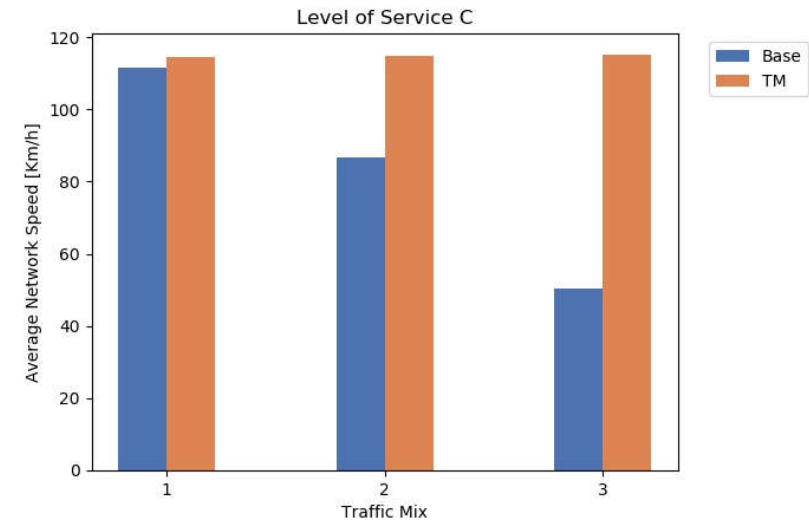
Highlighting
& polygon
dynamics

Traffic management in Transition Areas

Scenario 2

Results

- 1h random vehicle flow (LoS C ~ 3234 veh/h)
- Fleet mixes (MV-AV):
 - mix 1: 70-30
 - mix 2: 50-50
 - mix 3: 20-80

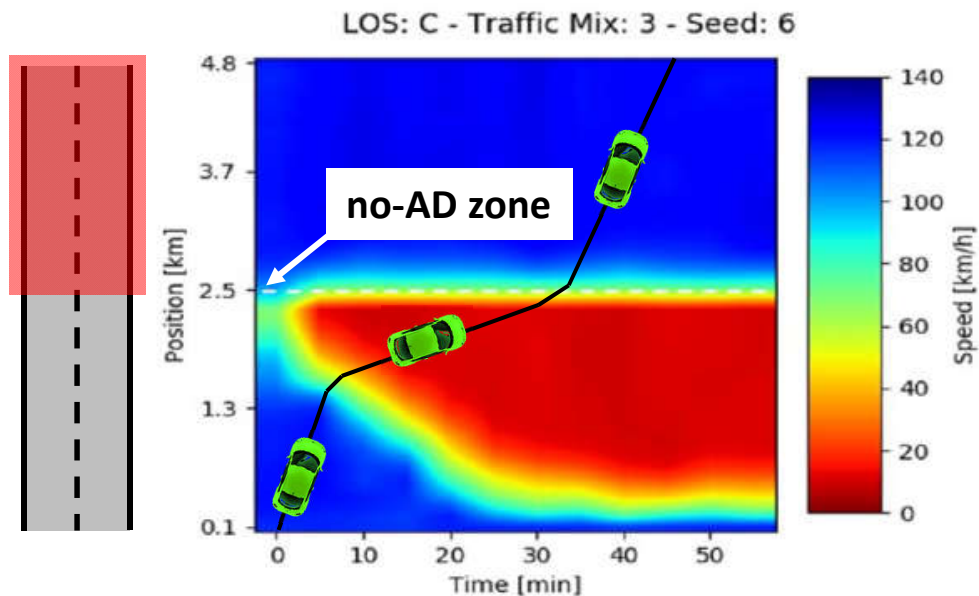


Traffic management in Transition Areas

Scenario 2

Results

Without traffic management



Summary

- Models:
 - New models for automated vehicles (CFModels ACC + CACC)
 - New model for simulation of control transitions
 - Driver State model
- Assessment of TM procedures:
 - Safety improvements for smoother flows at lane drops
 - Reducing perturbances by distribution of ToCs
- Upcoming:
 - Realistic simulation of communications
 - Combination of TransAID Services
 - Real world feasibility assessment

Thank you!

See also:

- Mintsis et al. 2018,
TransAID Deliverable 3.1
- Maerivoet et al. 2018,
TransAID Deliverable 4.2

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EU H2020, GNo 723390



Imperfect Driving 1

General CF Model:

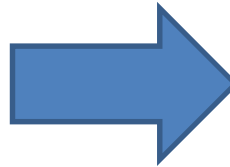
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Erroneous CF Model:

$$\dot{x}(t) = v(t)$$

$$\dot{v}(t) = a(\Delta \tilde{x}(t), \Delta \tilde{v}(t))$$

Imperfect Driving 2

Perception errors:

$$\eta_x(t) = c_x \cdot \Delta x(t) \cdot H_t$$

$$\eta_v(t) = c_v \cdot \Delta x(t) \cdot H_t$$

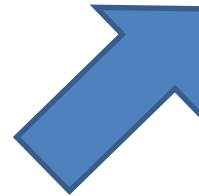
Error base process:

$$dH_t = -\theta_t \cdot H_t \cdot dt + \sigma_t \cdot dW_t$$

Base process coefficients:

$$\theta_t = c_\theta \cdot A(t)$$

$$\sigma_t = c_\sigma \cdot (1 - A(t)) \quad A(t) = \text{„awareness“}$$



Erroneous CF Model:

$$\dot{x}(t) = v(t)$$

$$\dot{v}(t) = a(\Delta \tilde{x}(t), \Delta \tilde{v}(t))$$

Imperfect Driving 3

