

Automated driving in microscopic traffic simulation

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Background

- Automated driving systems (ADS) are expected.
- Road authorities concerns about traffic performance.

Research questions

- How to model automated driving in microscopic traffic simulation?
- How will mix traffic affect transportation systems?

Modeling

Impacts

Why mixed traffic?

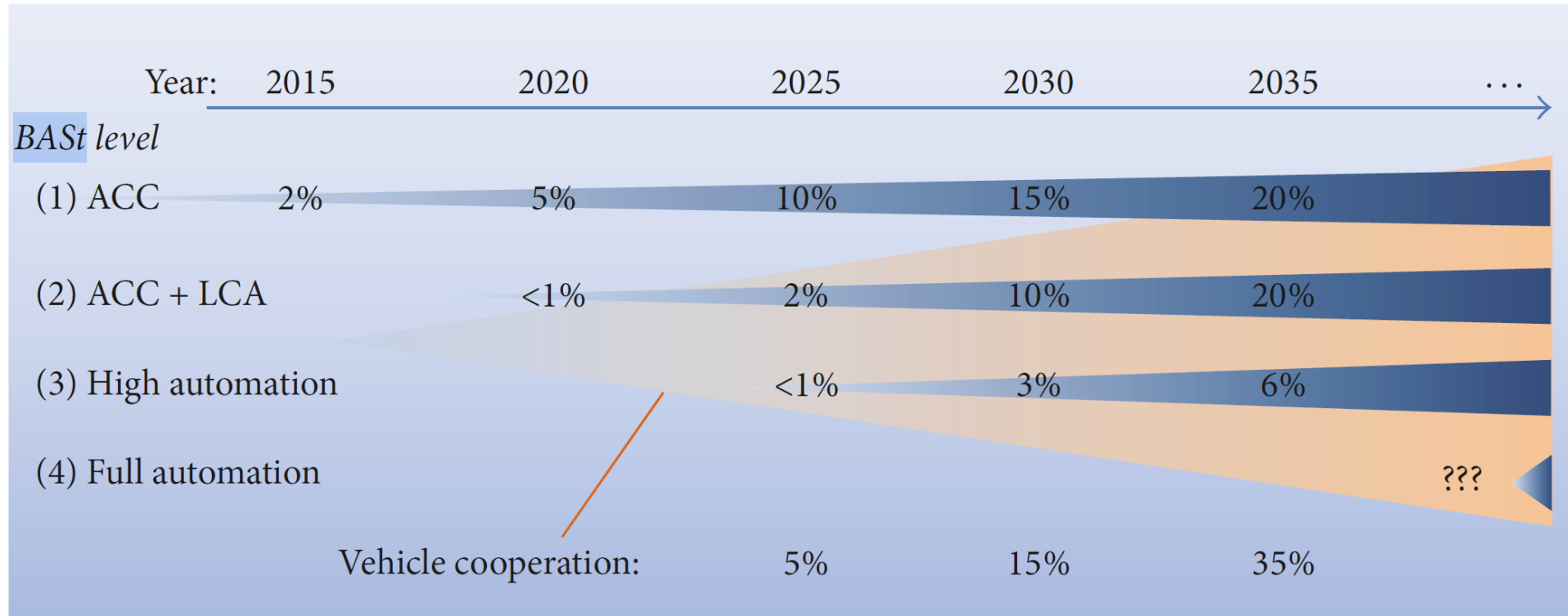
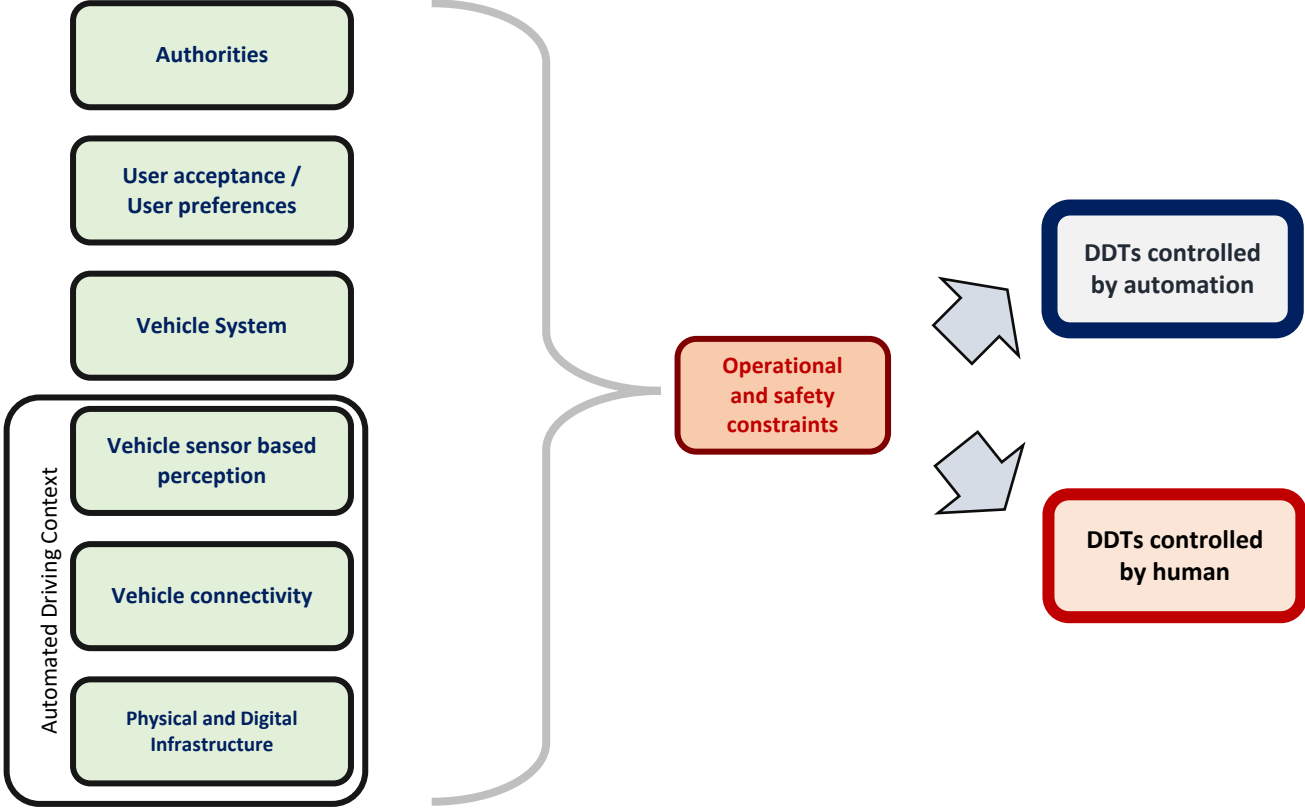


FIGURE 1: Estimated automated vehicle share on roads.

Source: Calvert et al. (2017) – Will automated vehicles negatively impact traffic flow?

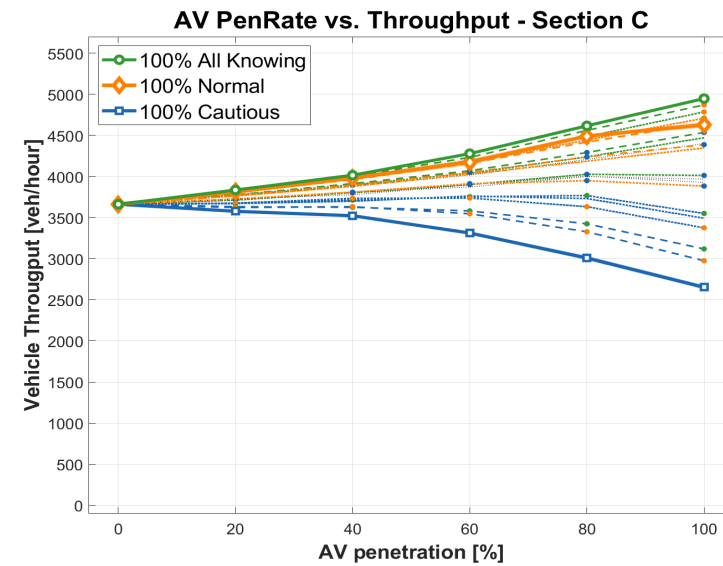
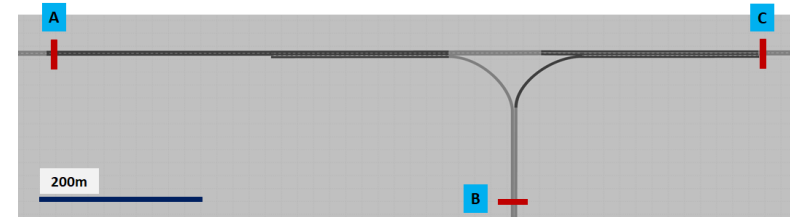
In: *Journal of Advanced Transportation*

State of the practice - Modeling

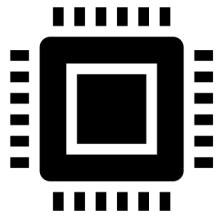


State of the practice - Impacts

- PTV Vissim
- Wiedemann 99 car-following model

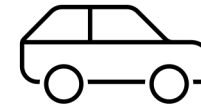


Microscopic driving models



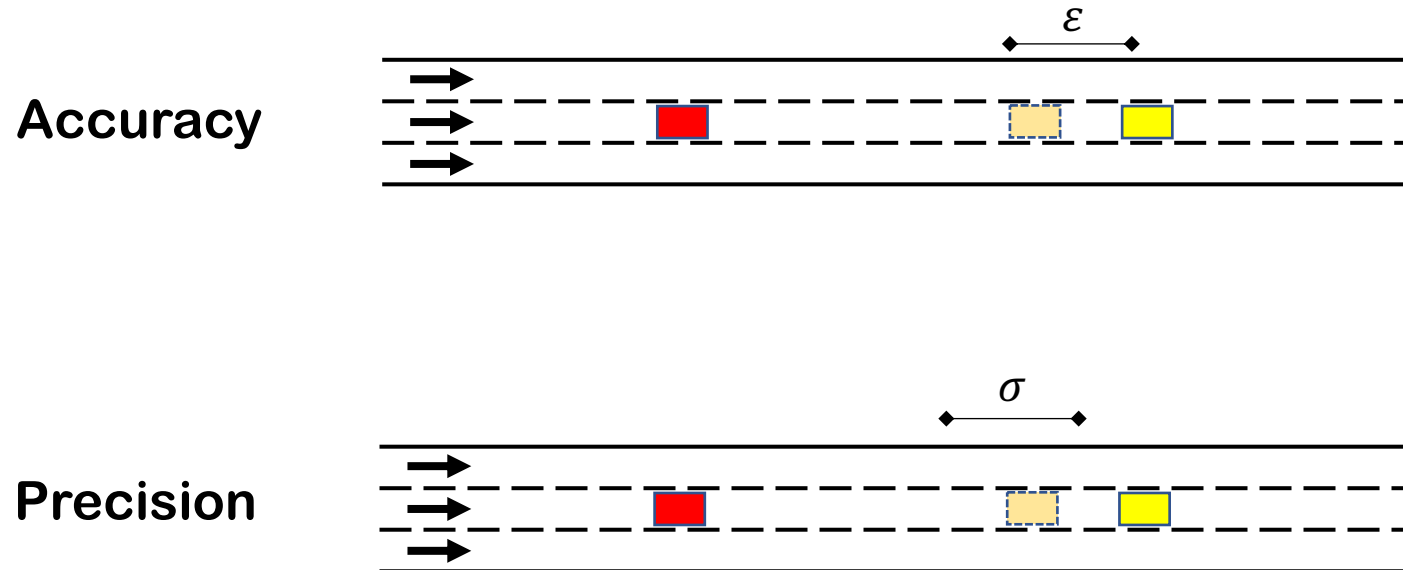
Car-following

Lane-changing



Perception errors

$$f_{err}(d, v) = \boxed{\varepsilon(d)} + \boxed{W_{trans} * (\sigma(d, v))}$$



Postigo et al. (2023) – Modeling perception performance in microscopic simulation of traffic flows including automated vehicles
2023 IEEE 26th International Conference on Intelligent Transportation Systems

Simulation experiment - IDM

	P1 – Human [1]	P2 – ADS [2]
Desired acceleration – a	1.0 m/s ²	1.0 m/s ²
Desired deceleration – b	2.75 m/s ²	2.75 m/s ²
Desired time gap – T	1.2 s	1.2 s
Free accel exponent – delta	4	25
Min. gap – S _o	2.0 m	2.0 m
Desired speed - V _o	25 m/s	25 m/s
Error Correlation – T _w	20s	500s

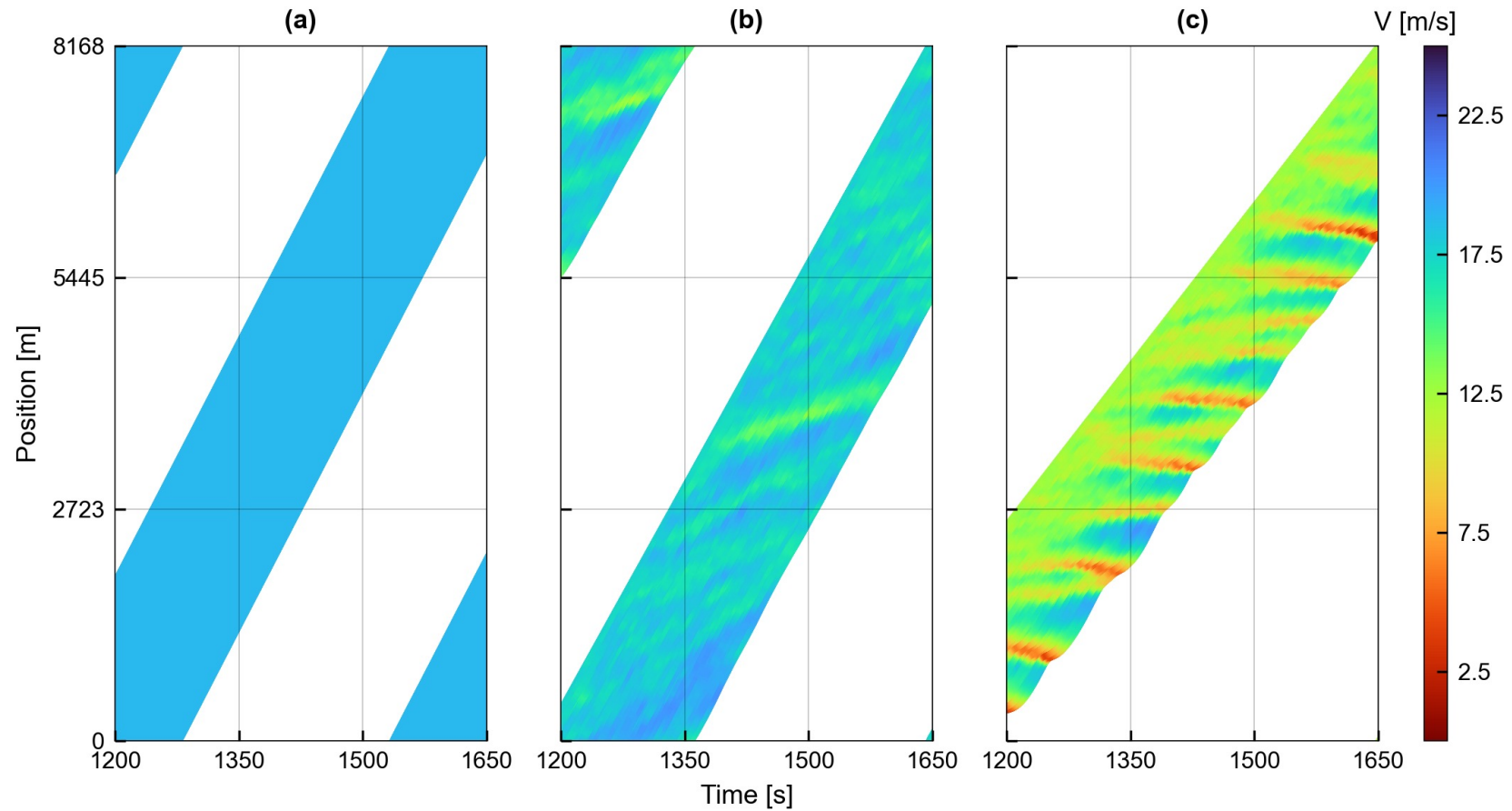
$$\dot{v} = a * \left(1 - \left(\frac{v}{V_o} \right)^\delta - \left(\frac{S^*}{S} \right)^2 \right)$$

$$S^* = S_o + \max \left\{ \left(0, vT + \frac{v\Delta v}{2\sqrt{ab}} \right) \right\}$$

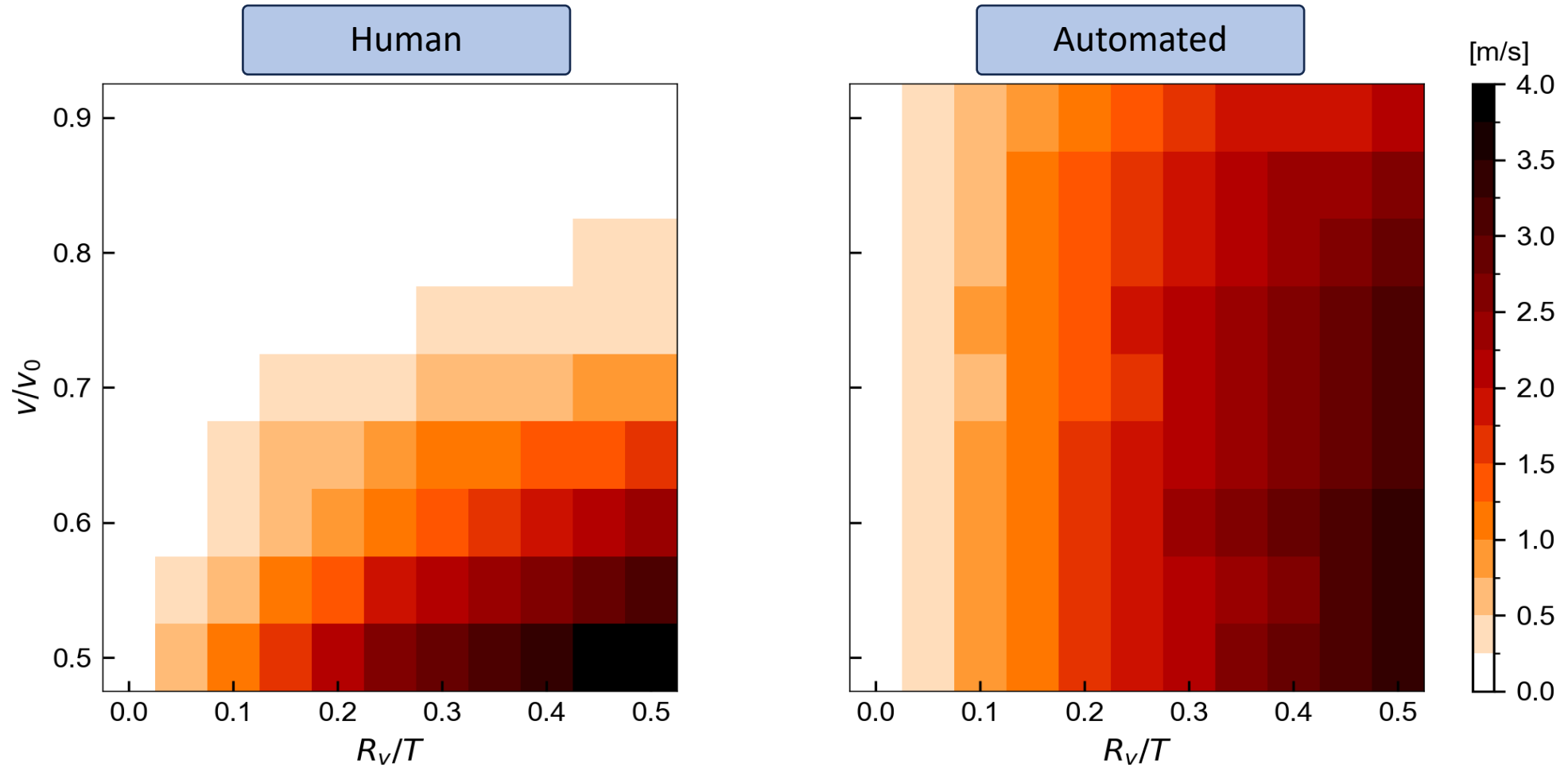
References:

- [1] : Zhu et al. 2018, Pourabdollah et al. 2017, Treiber et al. 2000,
[2] : De Souza et al. 2020, Gunter et al. 2019

Effects on traffic flow dynamics (I)

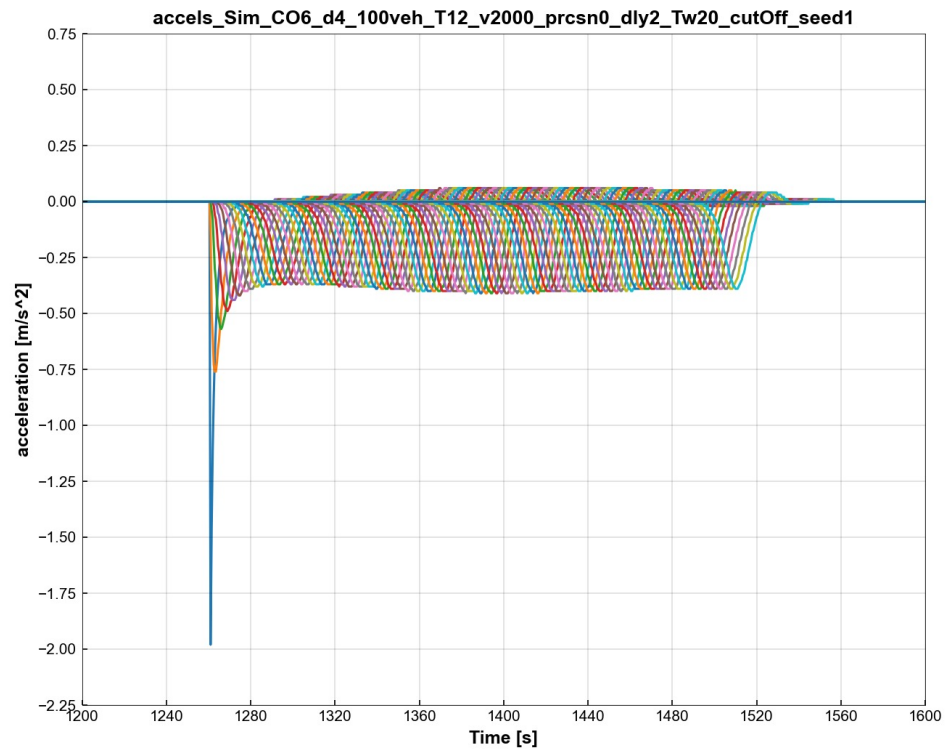


Effects on traffic flow dynamics – Free cruising

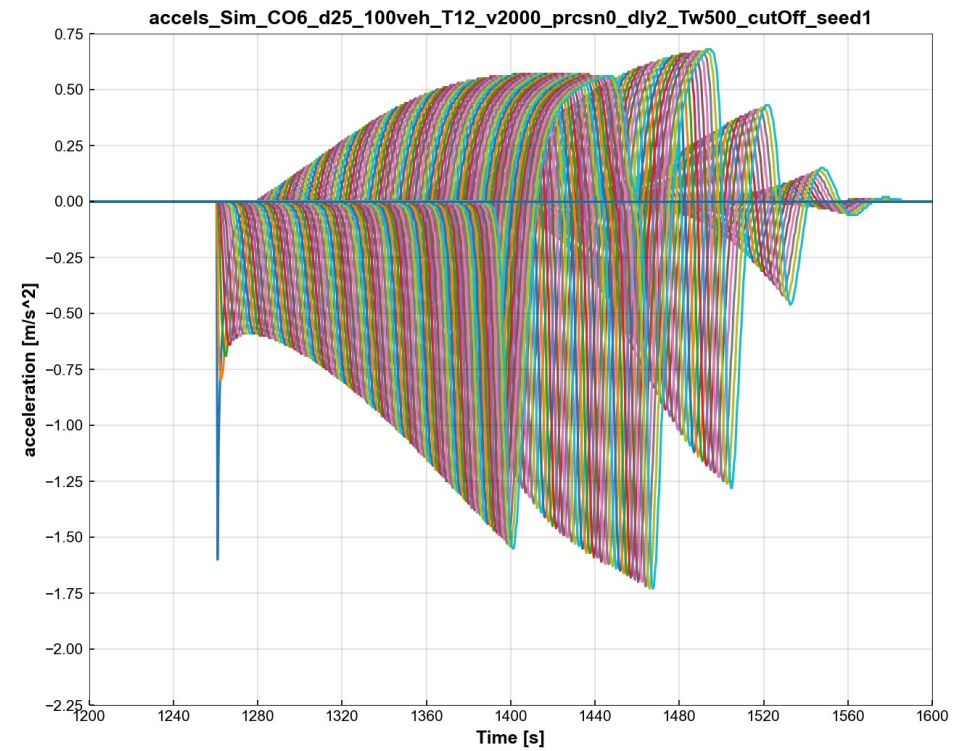


Effects on traffic flow dynamics – Cut in (I)

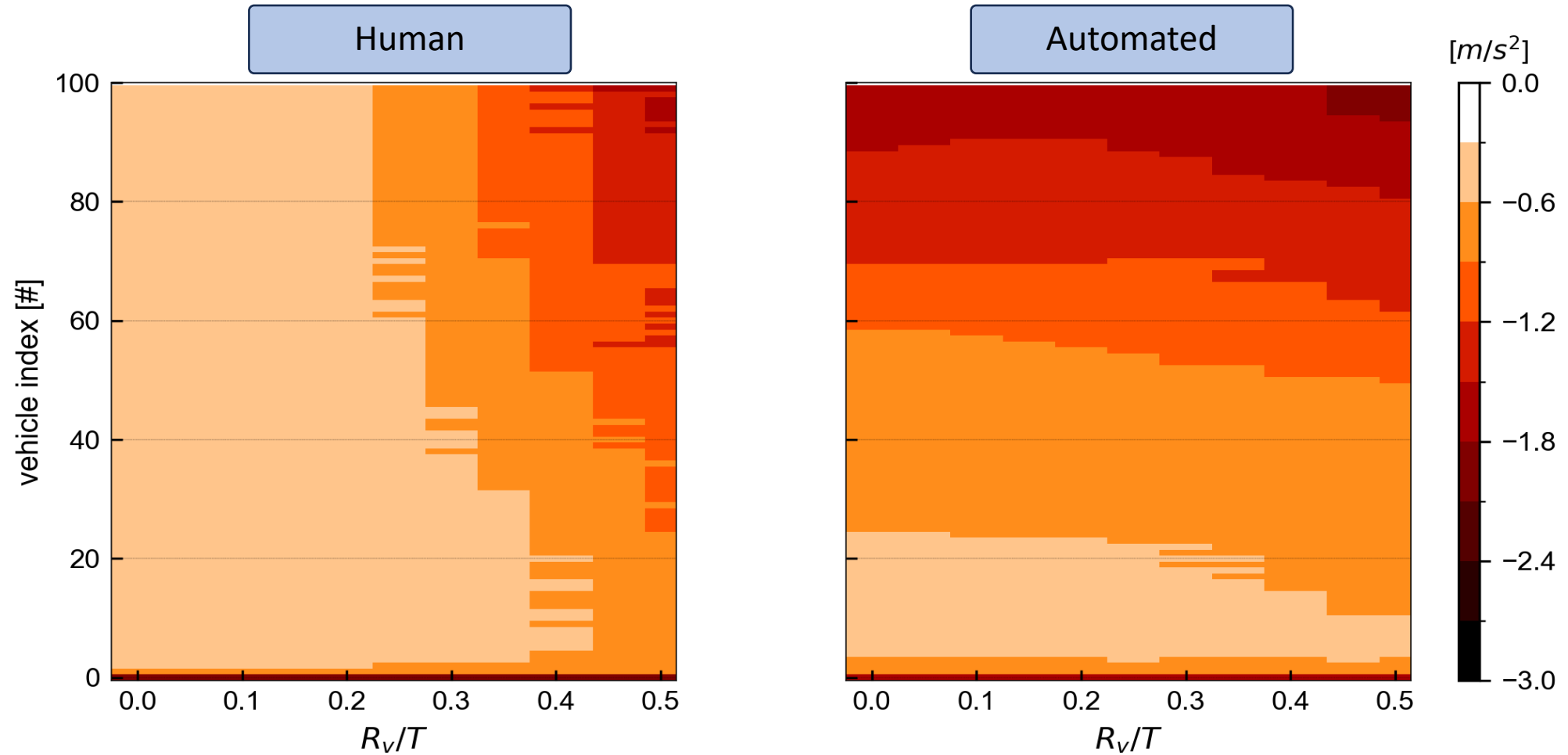
Human



Automated



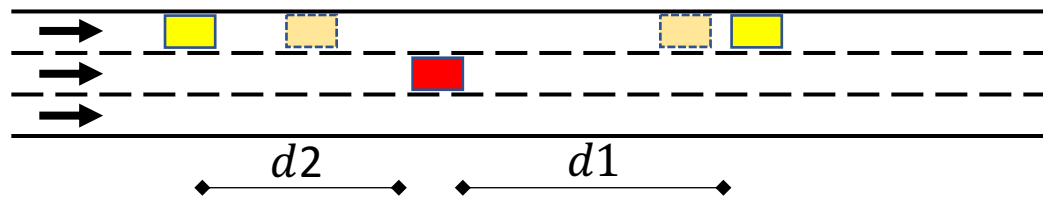
Effects on traffic flow dynamics – Cut in (II)



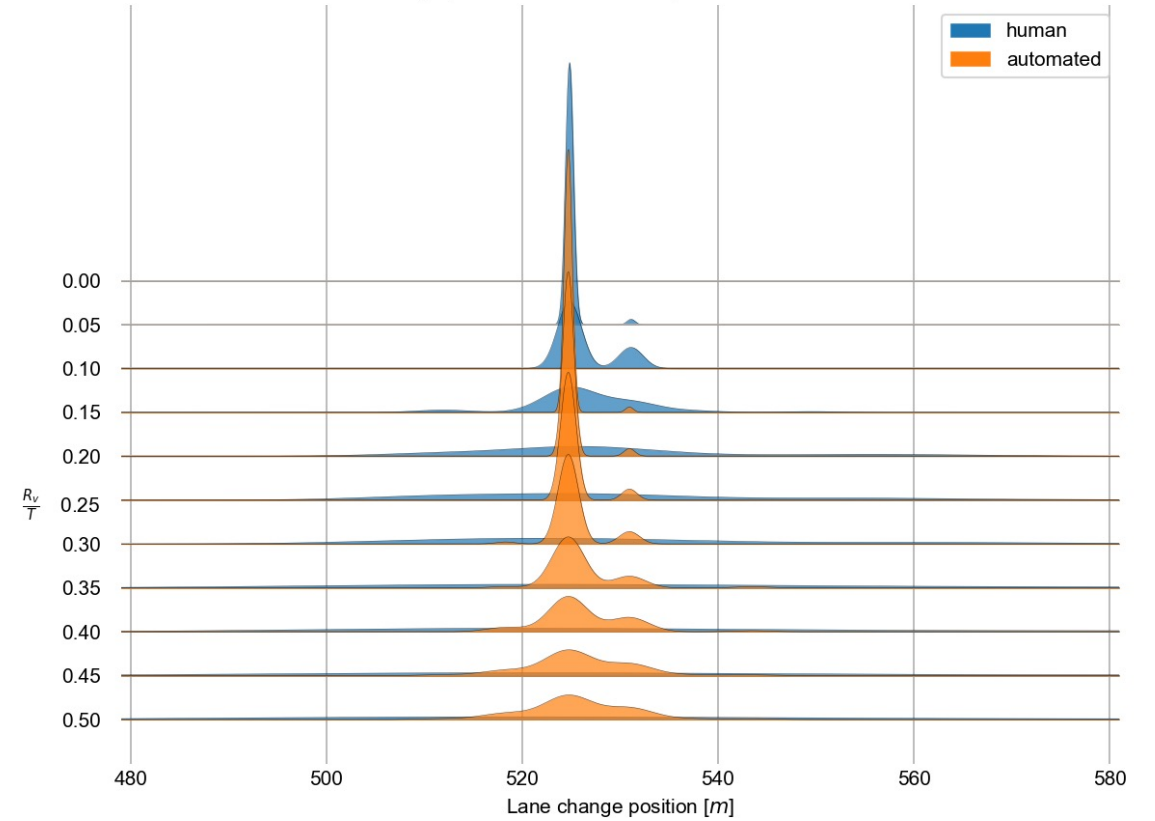
Lane changing



1000m



Lane change position for different precision levels - $\tau = 0.2$



Conclusions and next steps

Modeling

- The perception performance is a key point of difference between automated and human driving.
- The simulation experiment shows that the common assumption of perfect perception misses potential drawbacks such as a reduced road capacity or reduced traffic safety.
- The explicit modeling of the perception enables a wider range of assumptions to study mixed traffic in microscopic traffic simulation.

Impacts

- Next is to study the impacts of mixed traffic using the proposed modeling approach.
- Heterogeneity of human and automated vehicles to be included in a motorway environment.

Thanks for your attention!

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Accuracy (i)

$$f^{\Omega}(d, v) = \boxed{\varepsilon^{\Omega}(d)} + W_{trans} * (\sigma^{\Omega}(d) + \sigma^{\Omega}(v))$$

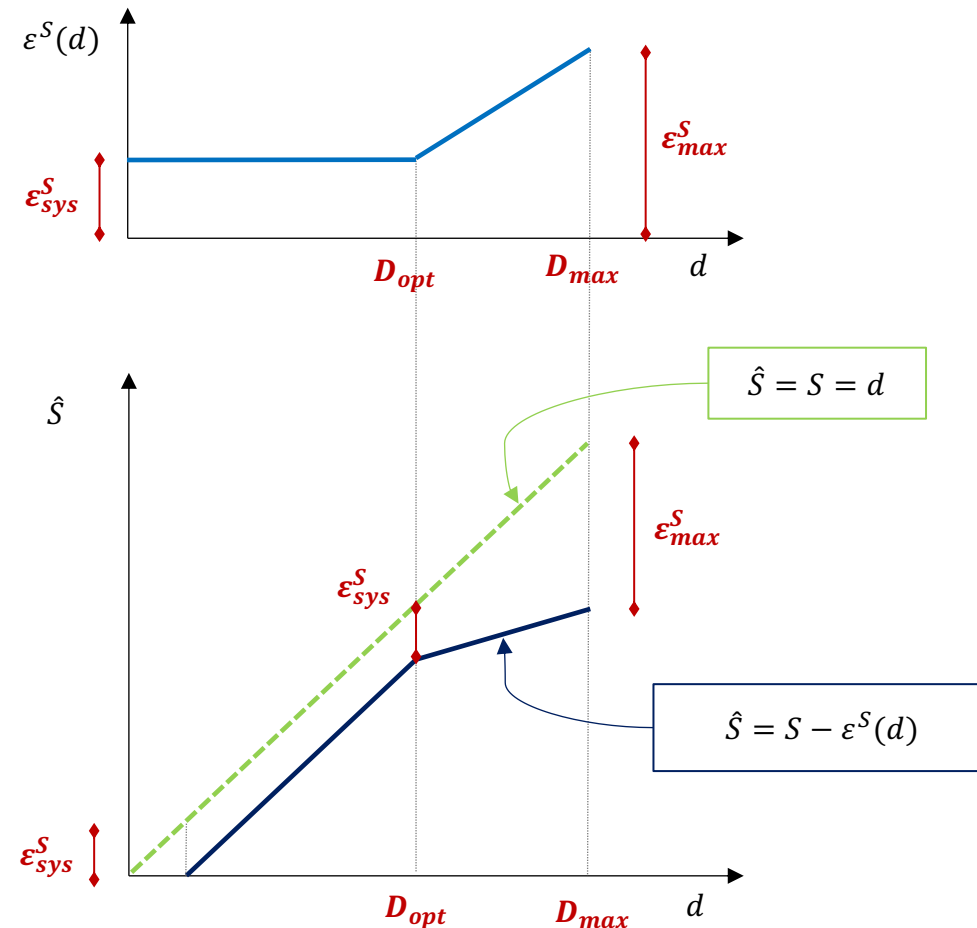
$\varepsilon^{\Omega}(d)$ - Parameters :

$\varepsilon_{sys}^{\Omega}$: Systematic, persistent or minimum error

$\varepsilon_{max}^{\Omega}$: Error at maximum detection range

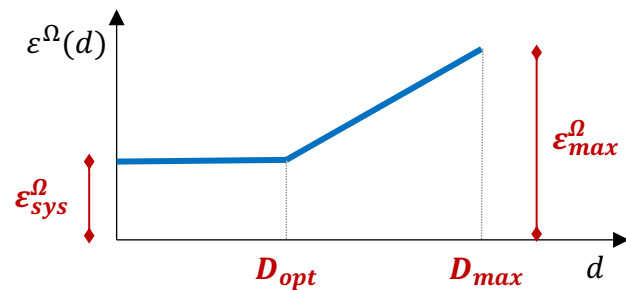
D_{opt} : Optimal operational range

D_{max} : Maximum detection range

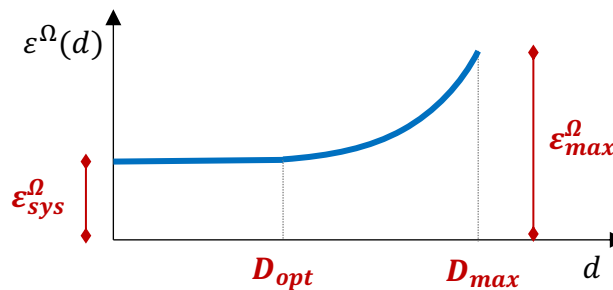


Accuracy (ii)

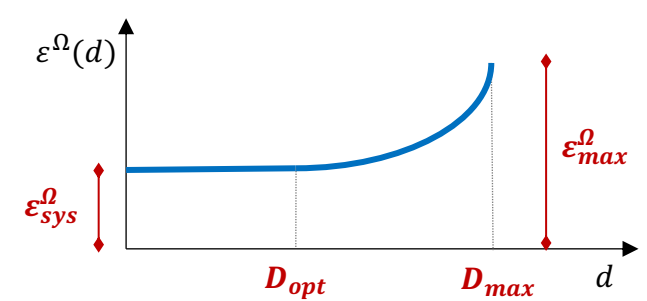
$$f^\Omega(d, v) = \boxed{\varepsilon^\Omega(d)} + W_{trans} * (\sigma^\Omega(d) + \sigma^\Omega(v))$$



linear



quadratic



ellipse

Precision (i)

$$f^{\Omega}(d, v) = \varepsilon^{\Omega}(d) + W_{trans} * (\sigma^{\Omega}(d) + \sigma^{\Omega}(v))$$

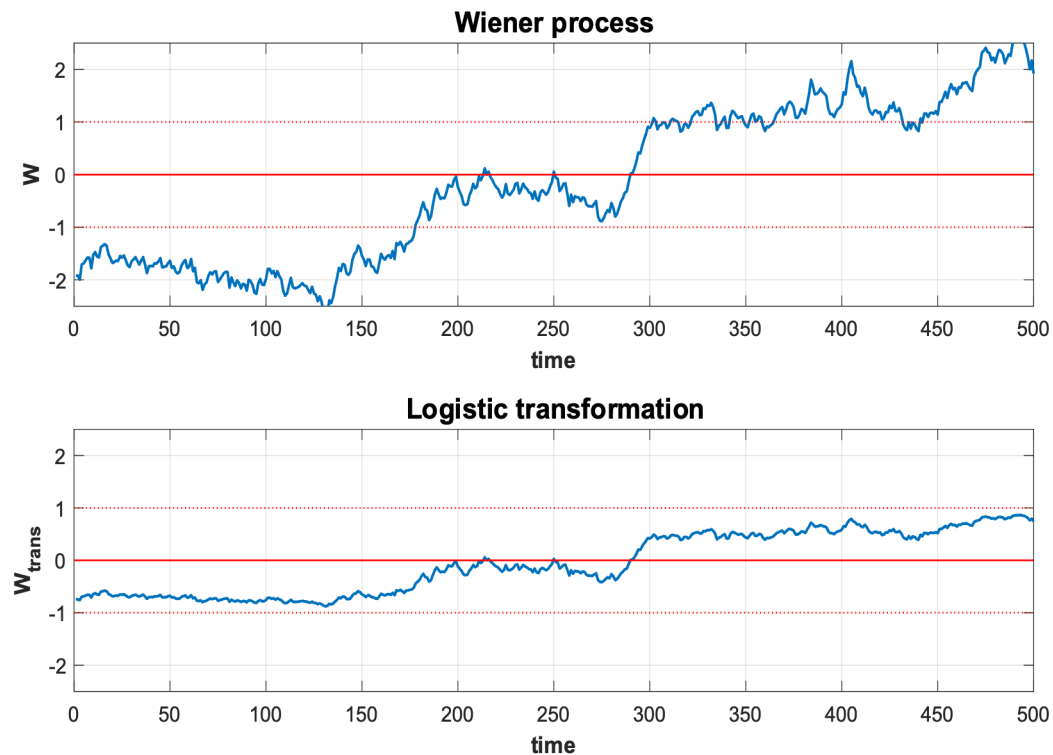
$$W_{trans} \in [-1, 1]$$

$$W_{trans} = \frac{2}{1 + \exp(-W)} - 1$$

$$W(t + \Delta t) = \begin{cases} \eta, & \text{initial} \\ \exp\left(-\frac{\Delta t}{\tau}\right) * W(t) + \eta \sqrt{\frac{2\Delta t}{\tau}}, & \text{otherwise} \end{cases}$$

$$\eta \in N(0,1)$$

τ : Time-window correlation



Precision (ii)

$$f^{\Omega}(d, v) = \varepsilon^{\Omega}(d) + W_{trans} * (\sigma^{\Omega}(d) + \sigma^{\Omega}(v))$$

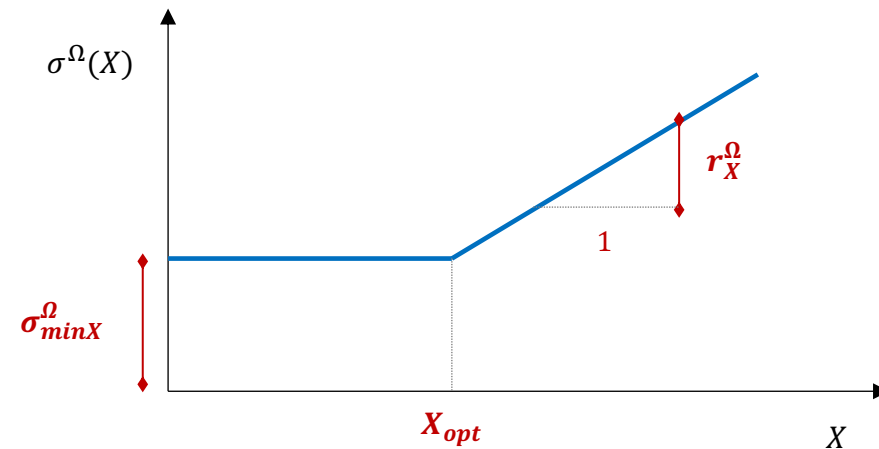
$$X = \{d, v\}$$

$\sigma^{\Omega}(X)$ - Parameters :

σ_{minX}^{Ω} : Minimum variation or noise

r_X^{Ω} : variation increase rate

X_{opt} : Optimal operational range



Accuracy and precision

Parameters :

$\varepsilon_{sys}^{\Omega}$: Systematic, persistent or minimum error

$\varepsilon_{max}^{\Omega}$: Error at maximum detection range

D_{opt} : Optimal operational range

D_{max} : Maximum detection range

σ_{minD}^{Ω} : Minimum distance variation or noise

r_d^{Ω} : Distance variation increase rate

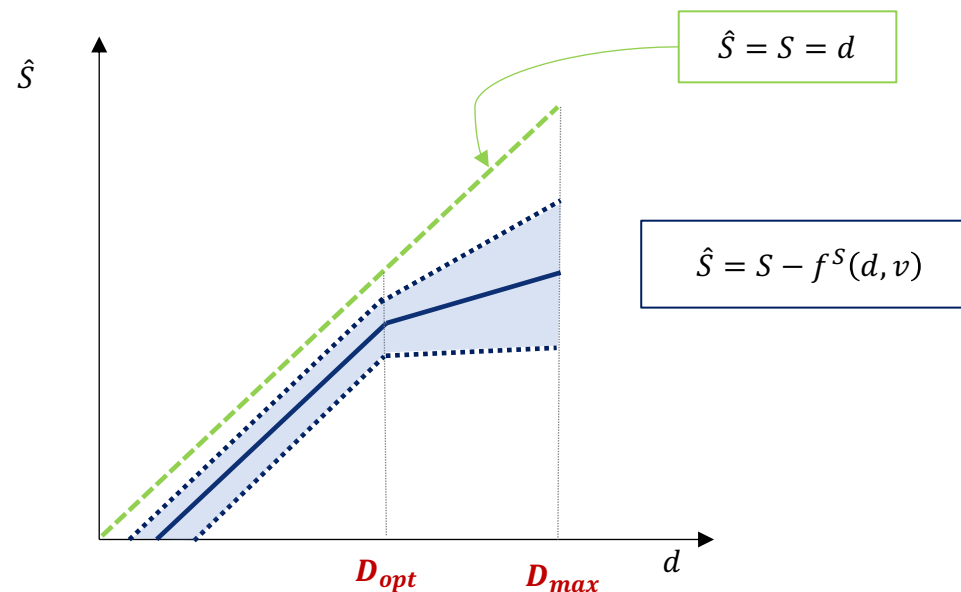
σ_{minV}^{Ω} : Minimum speed variation or noise

r_v^{Ω} : Speed variation increase rate

V_{opt} : Optimal operational speed

τ : Time-window variation correlation

$$f^{\Omega}(d, v) = \varepsilon^{\Omega}(d) + W_{trans} * (\sigma^{\Omega}(d) + \sigma^{\Omega}(v))$$



Intelligent driver model (IDM) sensibility

$$\dot{v} = a \cdot \left(1 - \left(\frac{v}{V_o} \right)^\delta - \left(\frac{S^*}{S} \right)^2 \right)$$

$$S^* = S_o + \max \left\{ \left(0, vT + \frac{v\Delta v}{2\sqrt{ab}} \right) \right\}$$

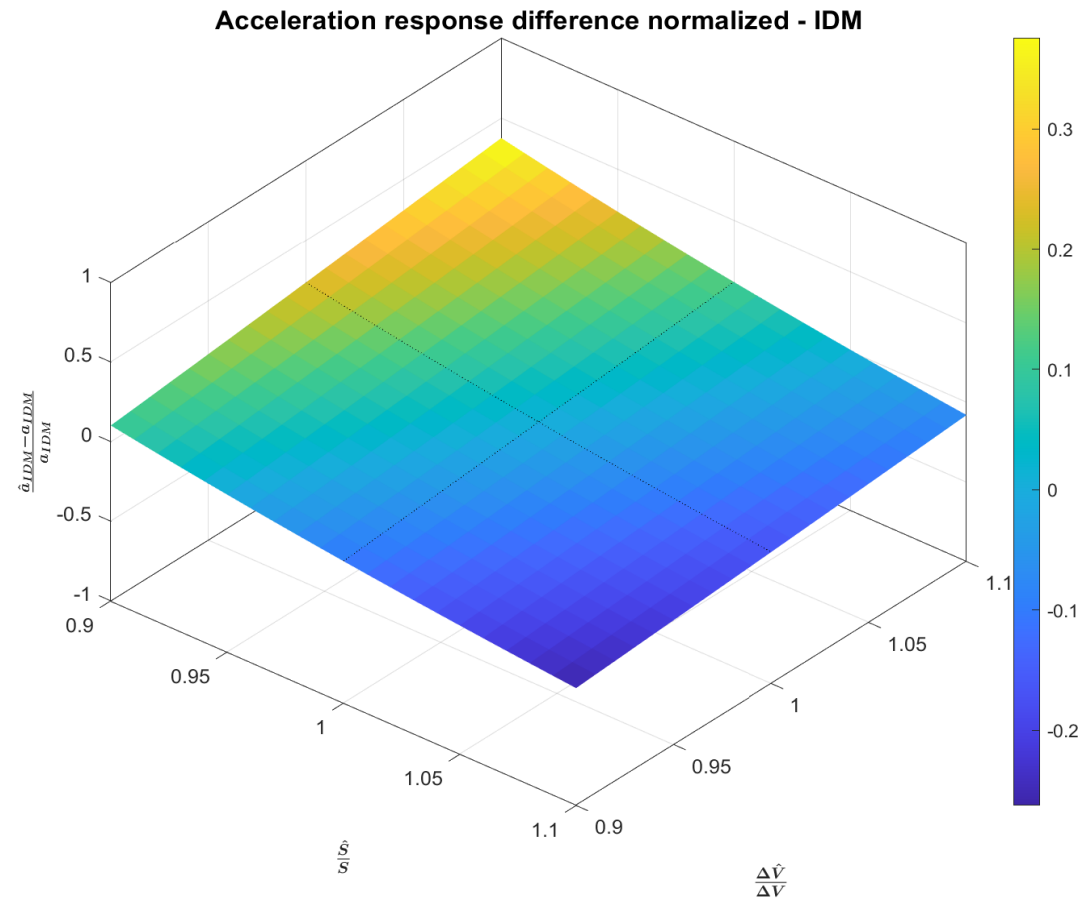
$$a = 1 \text{ m/s}^2$$

$$S = 65 \text{ m}$$

$$v = 25 \text{ m/s}$$

$$V_o = 25 \text{ m/s}$$

$$\Delta V = 5.55 \text{ m/s}$$



Change in fundamental diagram (IDM)

$$V_0 = 19.45 \text{ m/s}$$

$$S = Se(v) = \frac{so + vT}{\sqrt{1 - \left(\frac{v}{v_0}\right)^\delta}}$$

$$\rho = \frac{1}{S}$$

$$Q = \rho * V$$

