



# Characterizing Human-Automated Vehicle Interactions: An Investigation into Car-following Behavior

2023 TRB Annual Meeting

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01/10/2023

# Outline

## 1. Introduction

- 1.1 Background
- 1.2 Motivation

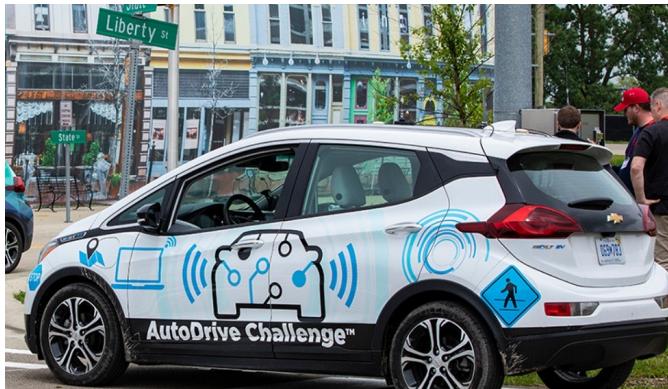
## 2. Major Findings

- 2.1 Data description
- 2.2 Data-driven method
- 2.3 Model-based method

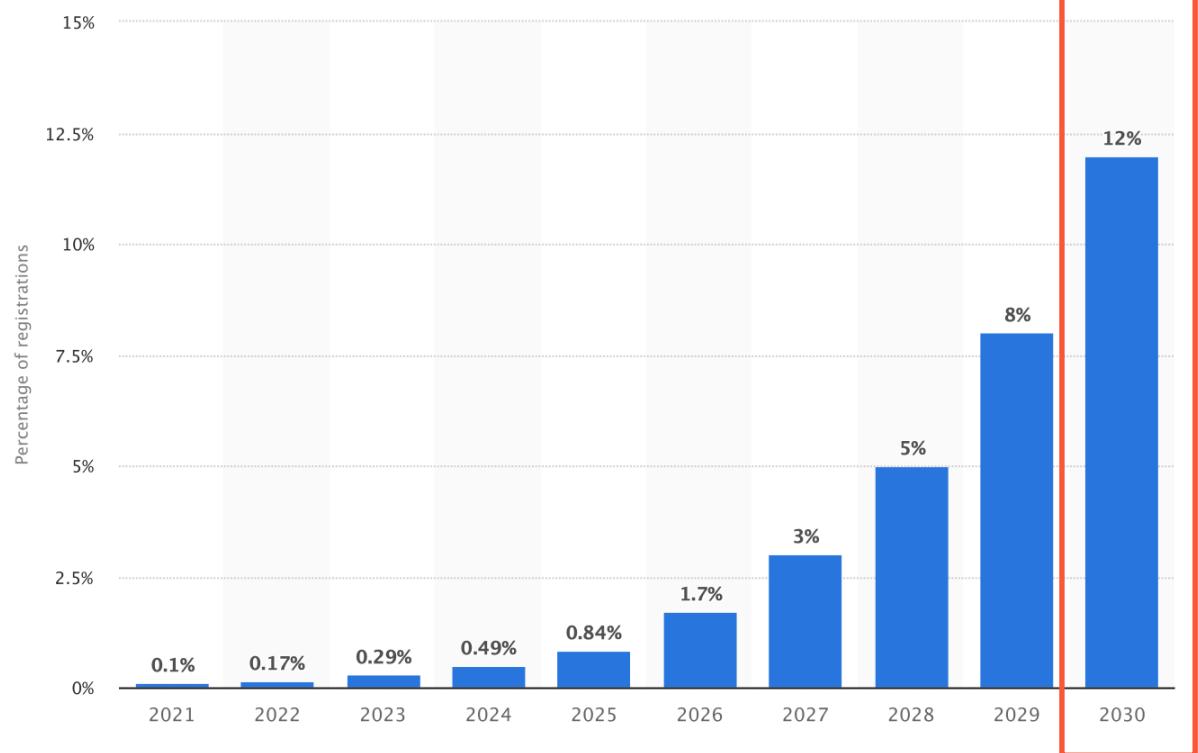
## 3. Conclusions

## 1.1 Background

- Automated vehicles build their presence
- Mixed autonomy traffic will remain for decades



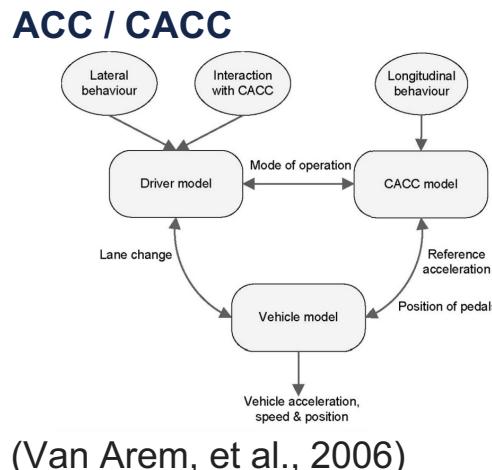
[Figures: [waymo.com](http://waymo.com), [sae.org](http://sae.org)]



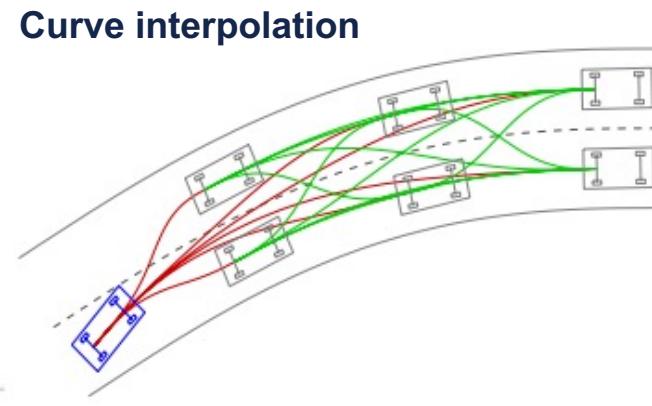
[Data source: Statista 2022]

## 1.2 Motivation: human's behaviors in HV-AV interactions are essential

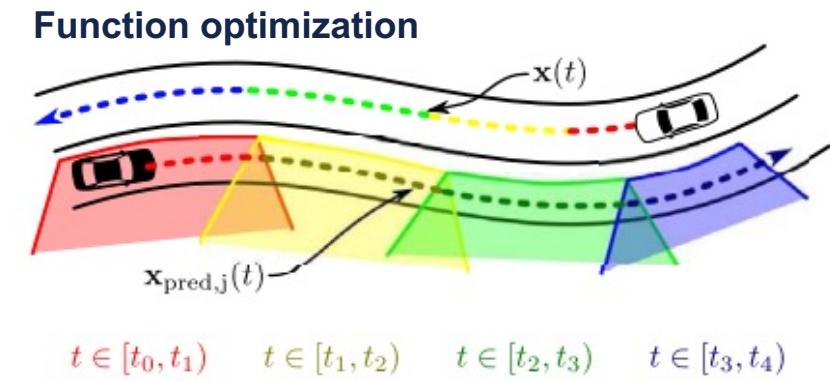
- Research was focusing on AV motion planning and trajectory design



(Van Arem, et al., 2006)



(Xu, et al., 2012.)



(Ziegler, et al., 2014)

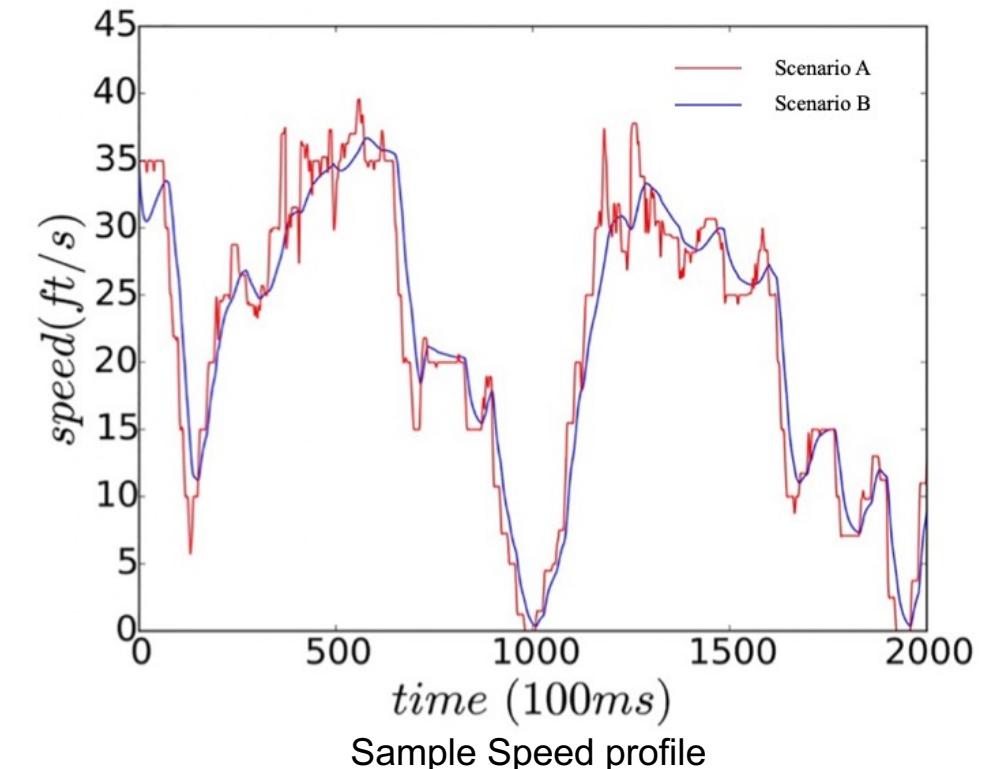
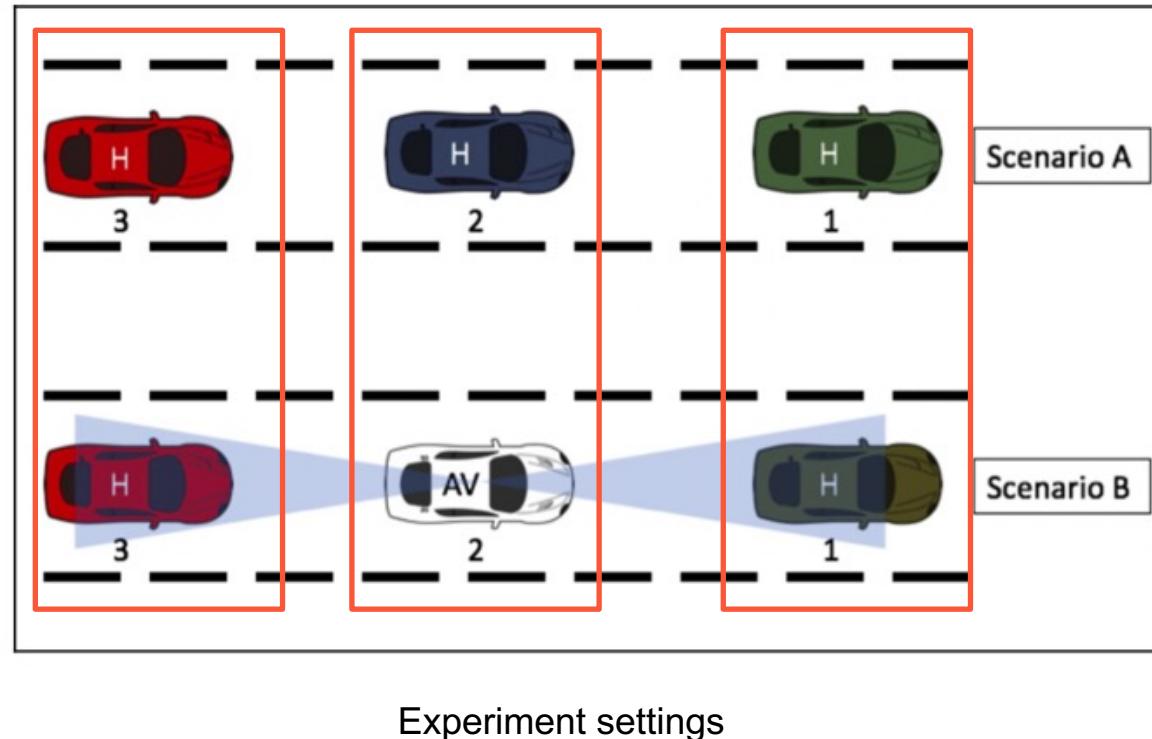
- Evidence revealed that ACC-equipped vehicles lead to string instability (Ciuffo, et al., 2021)
  - Significant difference was observed on human drivers' behavior on conventional cars when following an AV (Rahmati, et al., 2019)

## Which parameter / model can best capture such behavioral changes?

## 2. Major Findings



### 2.1 Data description: a car-following experiment



Rahmati, Y., Khajeh Hosseini, M., Talebpour, A., Swain, B., & Nelson, C. (2019). Influence of autonomous vehicles on car-following behavior of human drivers. *Transportation research record*, 2673(12), 367-379.

## 2. Major Findings



### 2.2 Data-driven methods: DTW analysis

- Which parameter can best capture such behavioral changes?

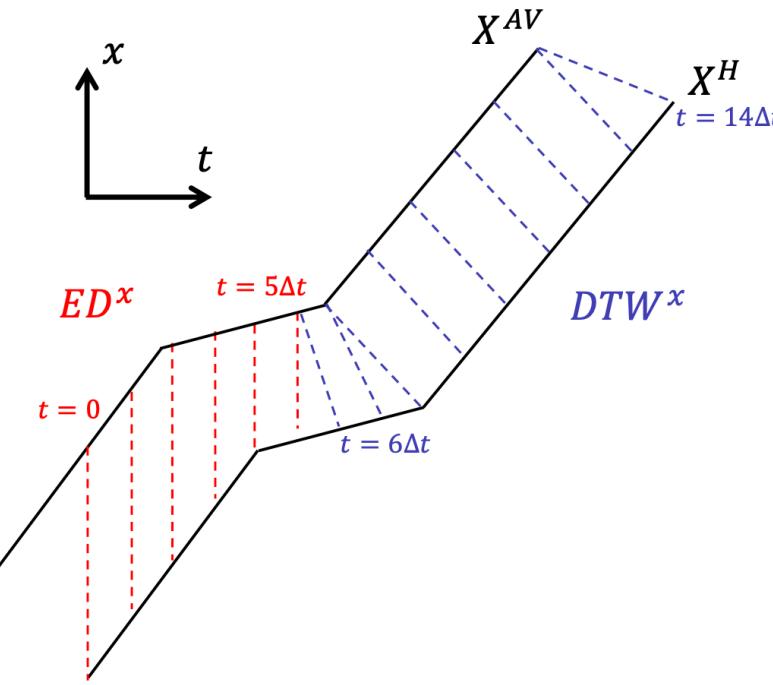
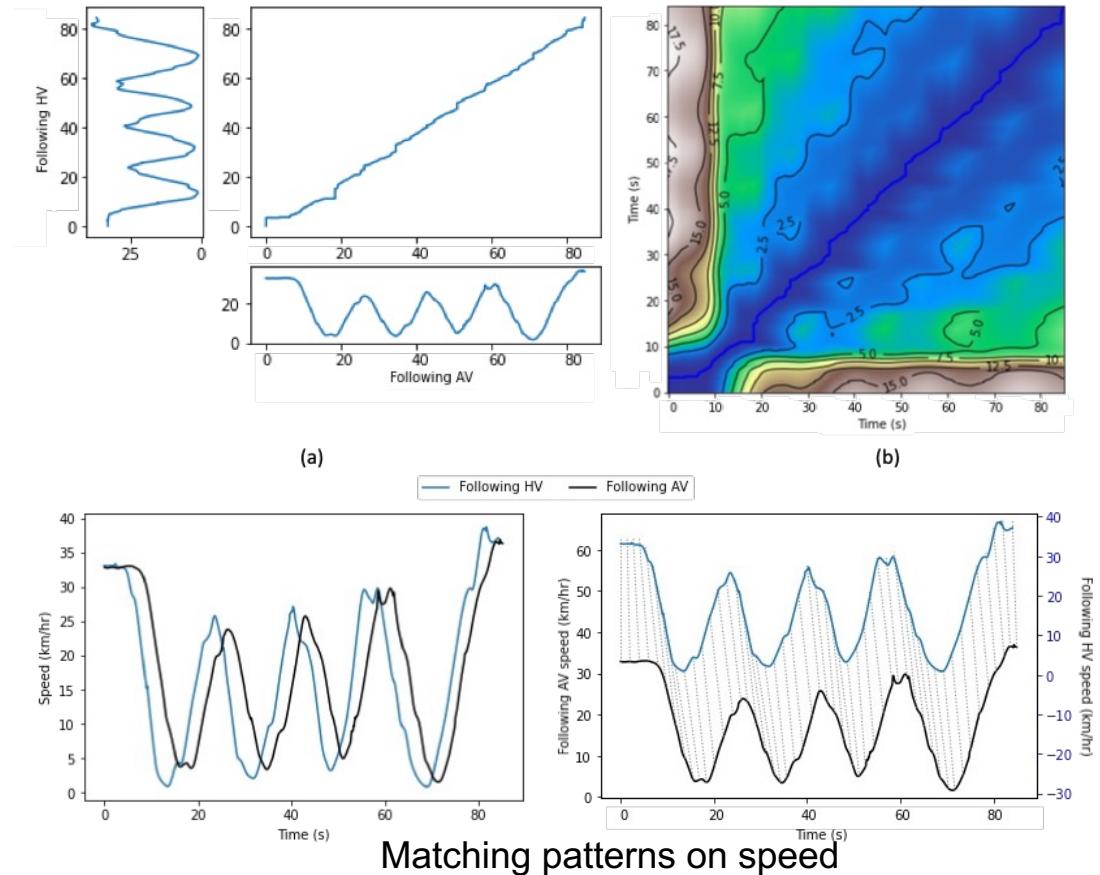
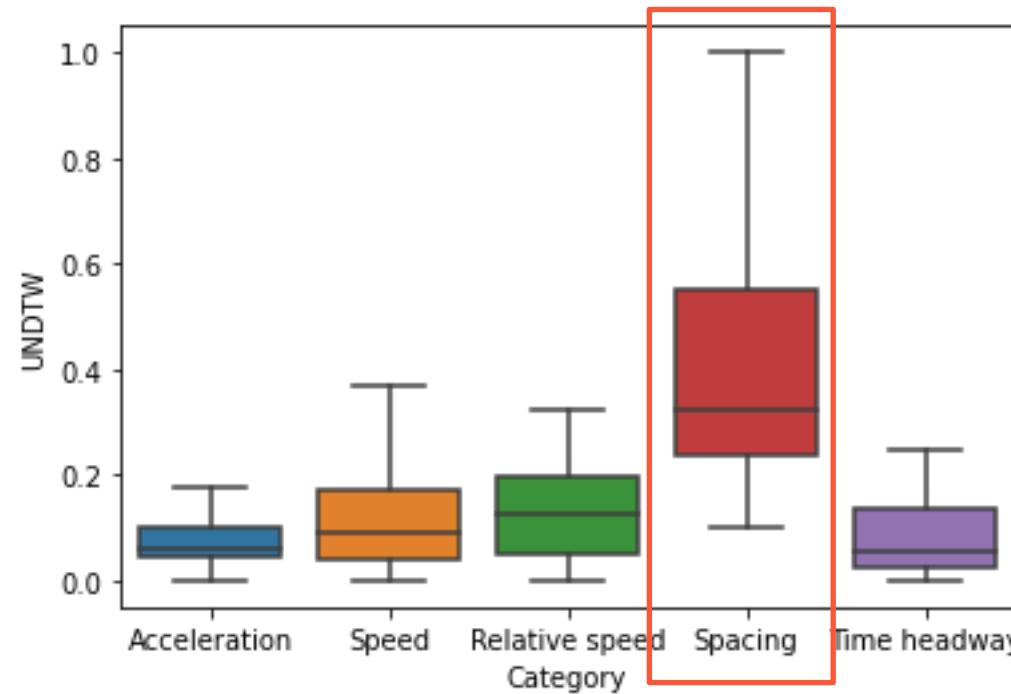


Illustration of dynamic time warping



### 2.2 Data-driven methods: DTW analysis

- Which parameter can best capture such behavioral changes?



Spacing has the best performance in measuring the changes in human drivers' behavior

### 2.3 Model-based method: GA calibration

- Which model can substantiate behavioral changes?

**Intelligent driver model (IDM)** (Treiber et al., 2000)

$$a_n(t) = a_{max} \left[ 1 - \left( \frac{v_n(t)}{v_{des}} \right)^4 - \left( \frac{s_{min} + v_n(t)T_{des} + \frac{v_n(t)\Delta v_n(t)}{2\sqrt{a_{max}b_{des}}}}{x_{n+1}(t) - l_{n+1}(t) - x_n(t)} \right)^2 \right]$$

	Model 1: IDM			AV following			K-S test p-value
	Human following			mean	sd	obs.	
v_des	85.2503	27.7736	42	92.6560	22.6403	43	0.3432
t_des	1.5205	0.3932	42	1.5280	0.3678	43	0.9821
s_min	3.0963	1.2173	42	2.9840	1.3073	43	0.7993
a_max	1.0013	0.0010	42	1.0015	0.0013	43	0.6094
b_des	2.3426	0.3331	42	2.4161	0.2571	43	0.2823

**Stochastic intelligent driver model**

(Treiber & Kesting, 2017)

$$a \sim N(a|a_{IDM}, \sigma_{IDM}^2)$$

$$v_{t+1} \sim N(v_t + a_{IDM}\Delta t | \Delta t^2 \sigma_{IDM}^2)$$

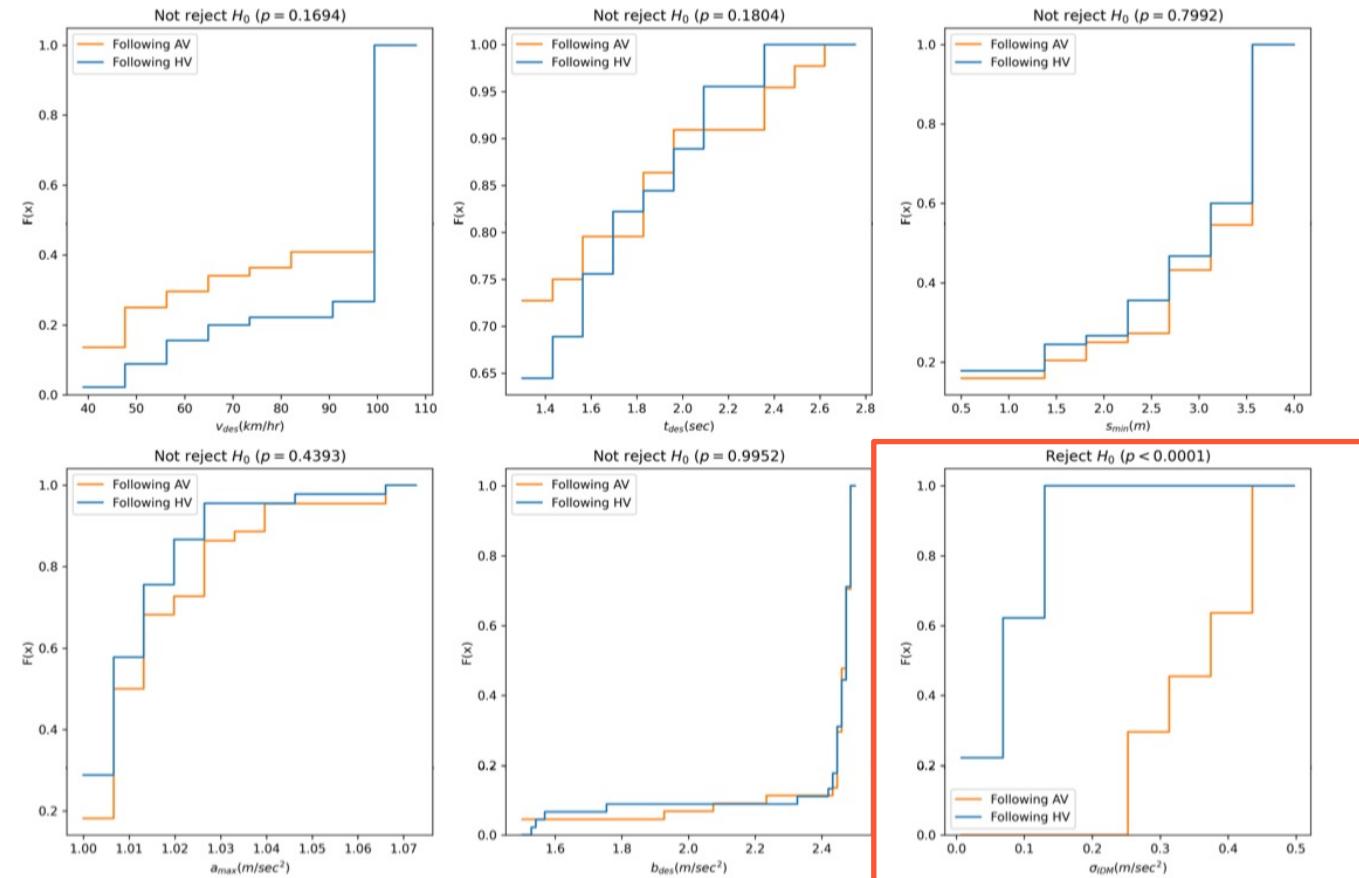
$$v_{t+1} = v_t + a_t \Delta t$$

## 2. Major Findings

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### 2.3 Which model can substantiate behavioral changes?

- Which model can substantiate behavioral changes?



- ✓ Evidence from experiments reveals that human drivers' behaviors will change when following an AV
- ✓ The spacing parameter can best substantiate such phenomenon
- ✓ A stochastic extension to IDM can better model the car-following behavior in mixed autonomy traffic, and following an AV leads to less uncertainty for a human driver



Thank You!  
Any Questions?

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Ciuffo, B., Mattas, K., Makridis, M., Albano, G., Anesiadou, A., He, Y., ... & Szalay, Z. (2021). Requiem on the positive effects of commercial adaptive cruise control on motorway traffic and recommendations for future automated driving systems. *Transportation research part C: emerging technologies*, 130, 103305.

Rahmati, Y., Khajeh Hosseini, M., Talebpour, A., Swain, B., & Nelson, C. (2019). Influence of autonomous vehicles on car-following behavior of human drivers. *Transportation research record*, 2673(12), 367-379.

Treiber, M., Hennecke, A., & Helbing, D. (2000). Congested traffic states in empirical observations and microscopic simulations. *Physical review E*, 62(2), 1805.

Treiber, M., & Kesting, A. (2017). The intelligent driver model with stochasticity-new insights into traffic flow oscillations. *Transportation research procedia*, 23, 174-187.

Van Arem, B., Van Driel, C. J., & Visser, R. (2006). The impact of cooperative adaptive cruise control on traffic-flow characteristics. *IEEE Transactions on intelligent transportation systems*, 7(4), 429-436.

Xu, W., Wei, J., Dolan, J. M., Zhao, H., & Zha, H. (2012, May). A real-time motion planner with trajectory optimization for autonomous vehicles. In *2012 IEEE International Conference on Robotics and Automation* (pp. 2061-2067). IEEE.

Ziegler, J., Bender, P., Dang, T., & Stiller, C. (2014, June). Trajectory planning for Bertha—A local, continuous method. In *2014 IEEE intelligent vehicles symposium proceedings* (pp. 450-457). IEEE.

# Backup

