



UNIVERSITY OF
ILLINOIS
URBANA-CHAMPAIGN

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

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Outline

1. Introduction

- 1.1 Background
- 1.2 Motivation

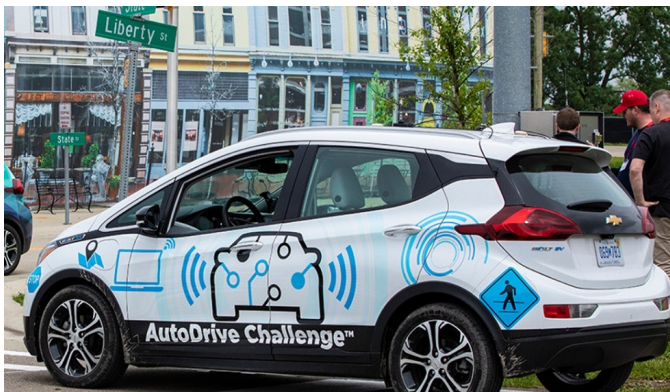
2. Major Findings

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

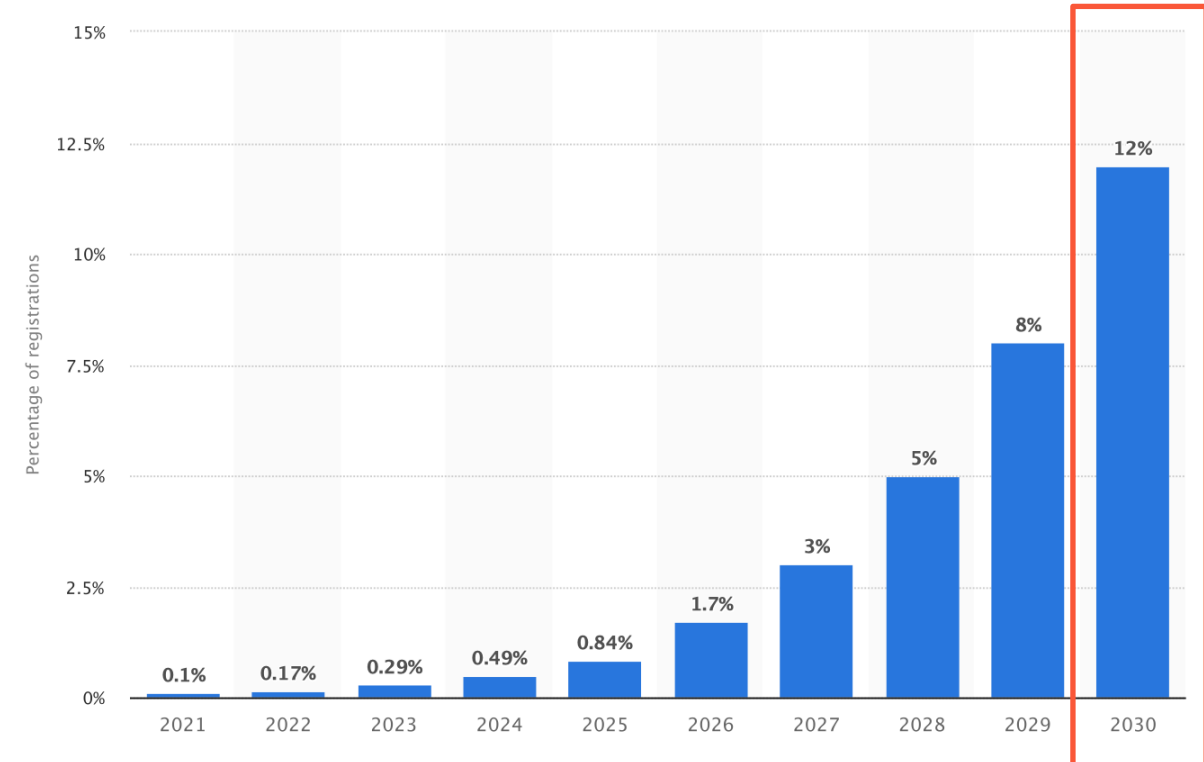
3. Conclusions

1.1 Background

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



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

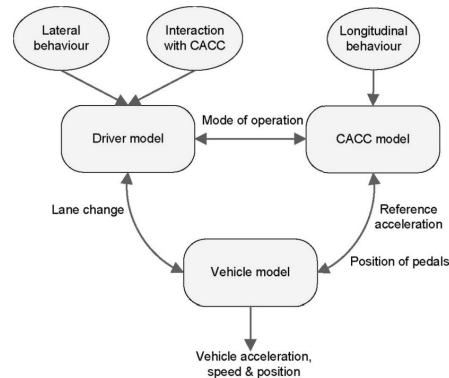


[Data source: Statista 2022]

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

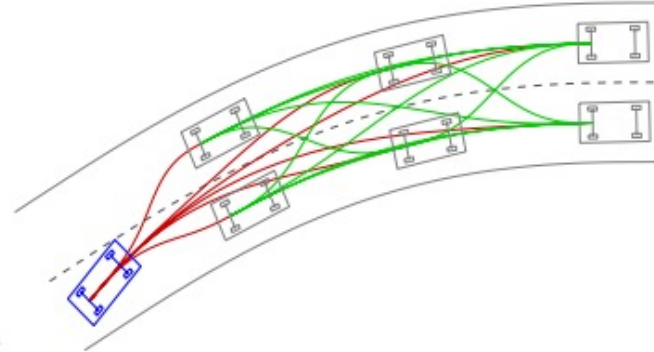
- Research was focusing on AV motion planning and trajectory design

ACC / CACC



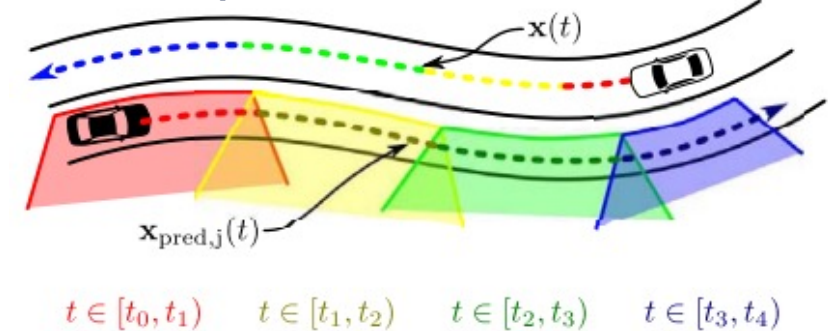
(Van Arem, et al., 2006)

Curve interpolation



(Xu, et al., 2012.)

Function optimization

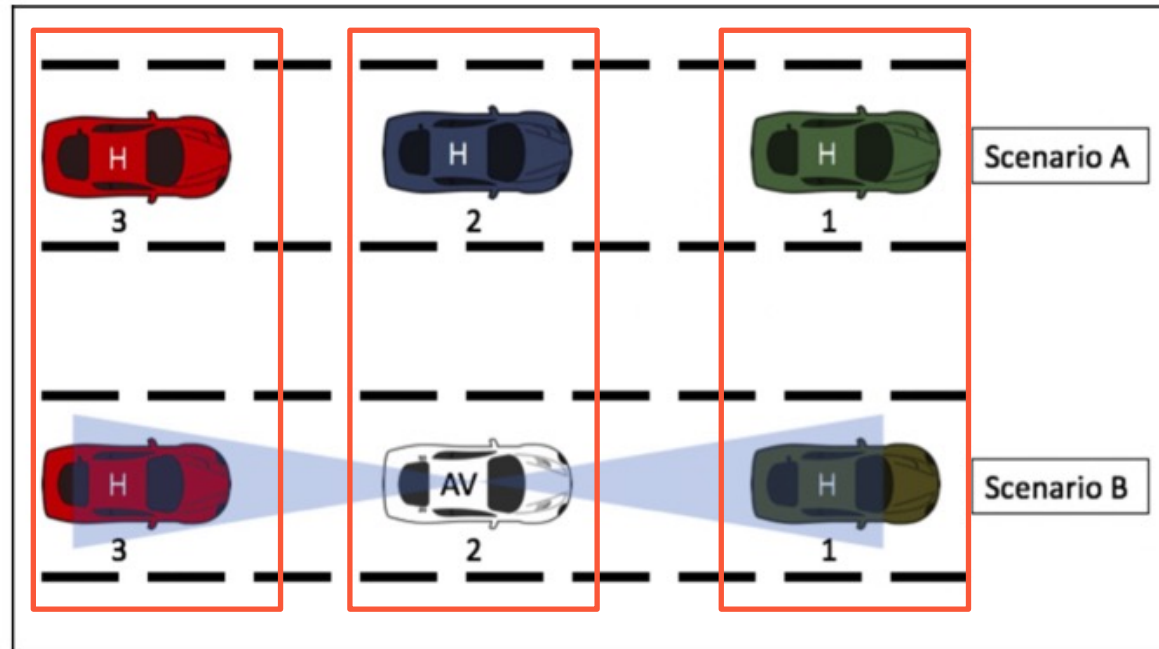


(Ziegler, et al., 2014)

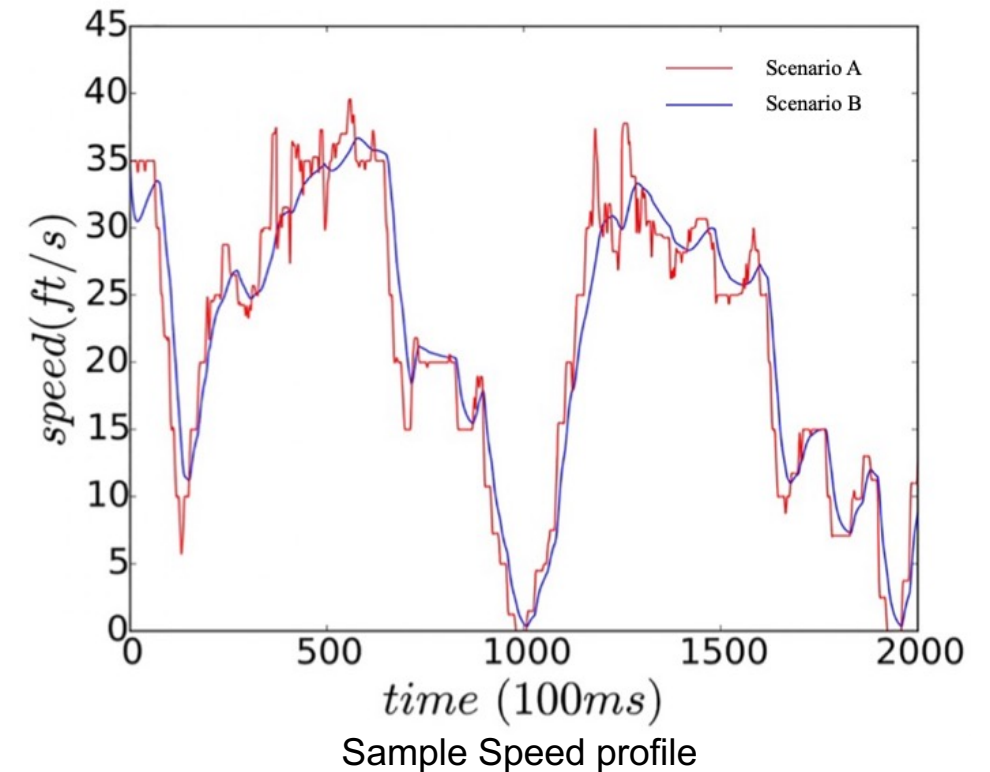
- Evidence revealed that ACC-equipped vehicles lead to string instability** (Ciuffo, et al., 2021)
- Significant difference was observed on human drivers' behavior on conventional cars when following an AV** (Rahmati, et al., 2019)

Which parameter / model can best capture such behavioral changes?

2.1 Data description: a car-following experiment



Experiment settings



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.

2.2 Data-driven methods: DTW analysis

- Which parameter can best capture such behavioral changes?

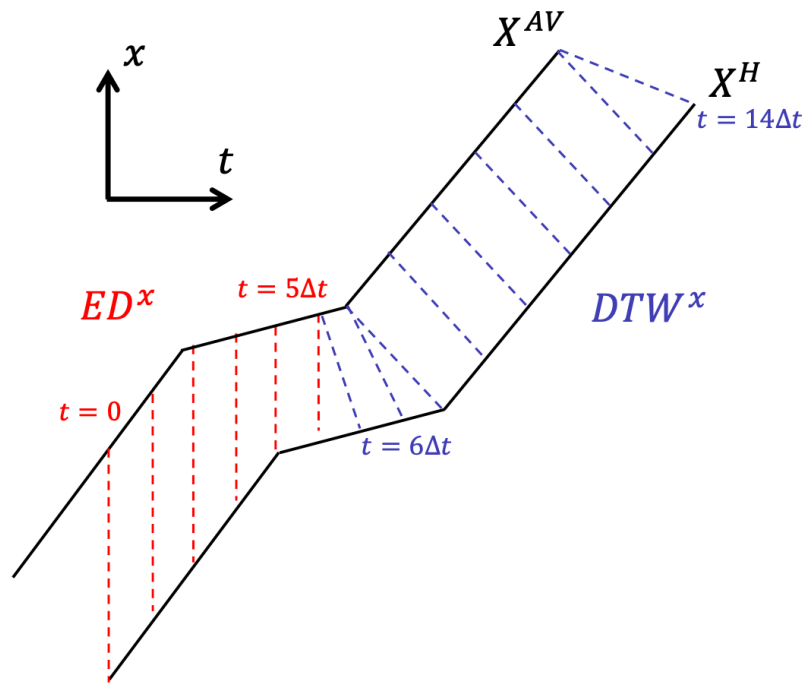
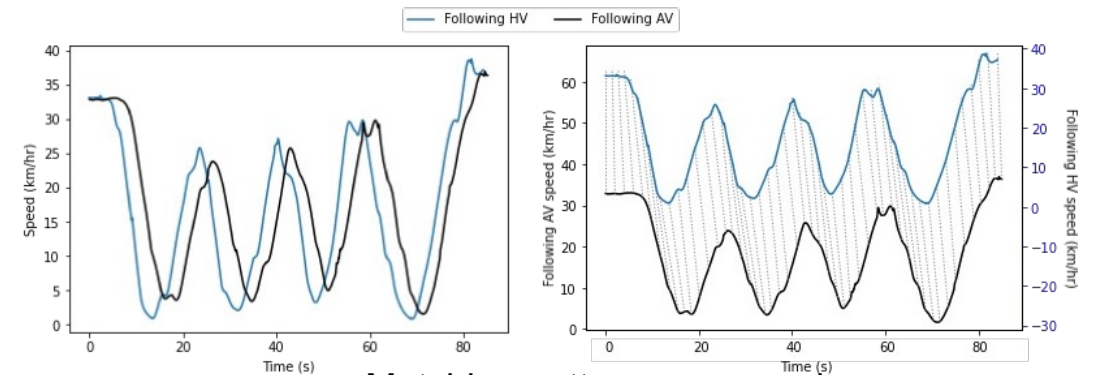
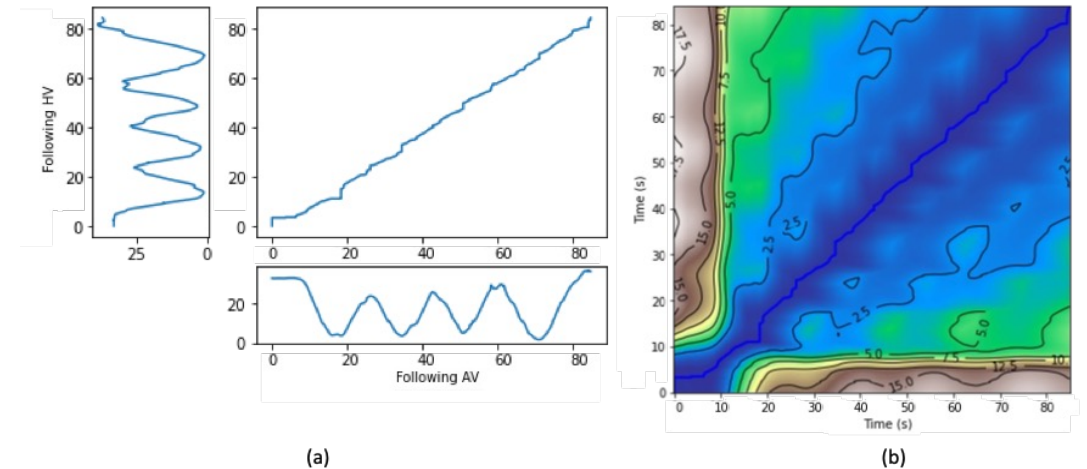


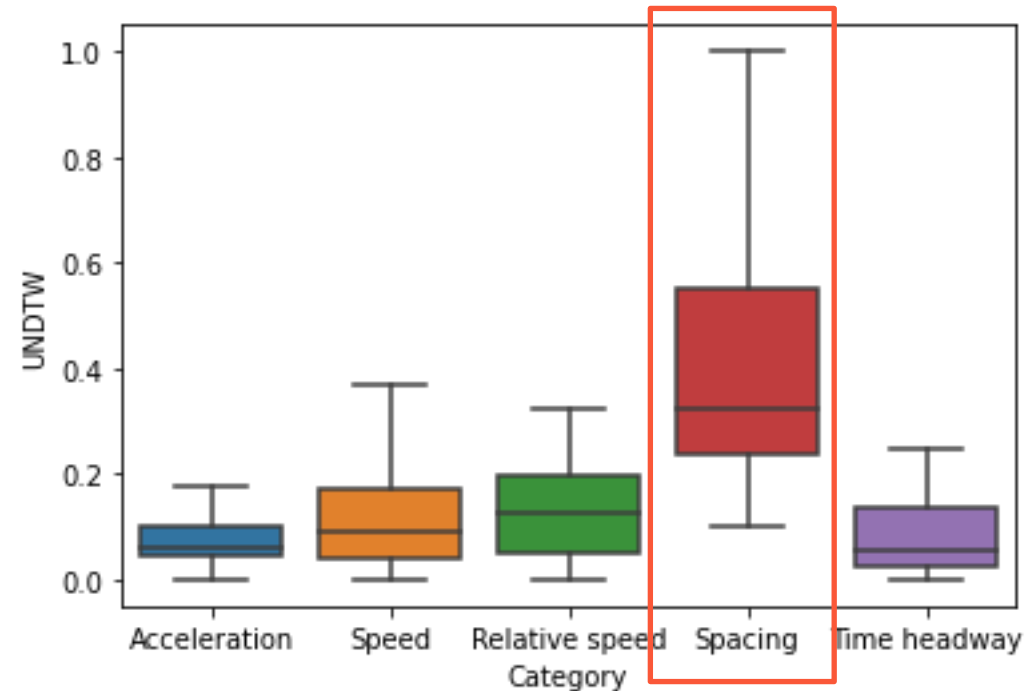
Illustration of dynamic time warping



Matching patterns on speed

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)

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

Stochastic intelligent driver model

(Treiber & Kesting, 2017)

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

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

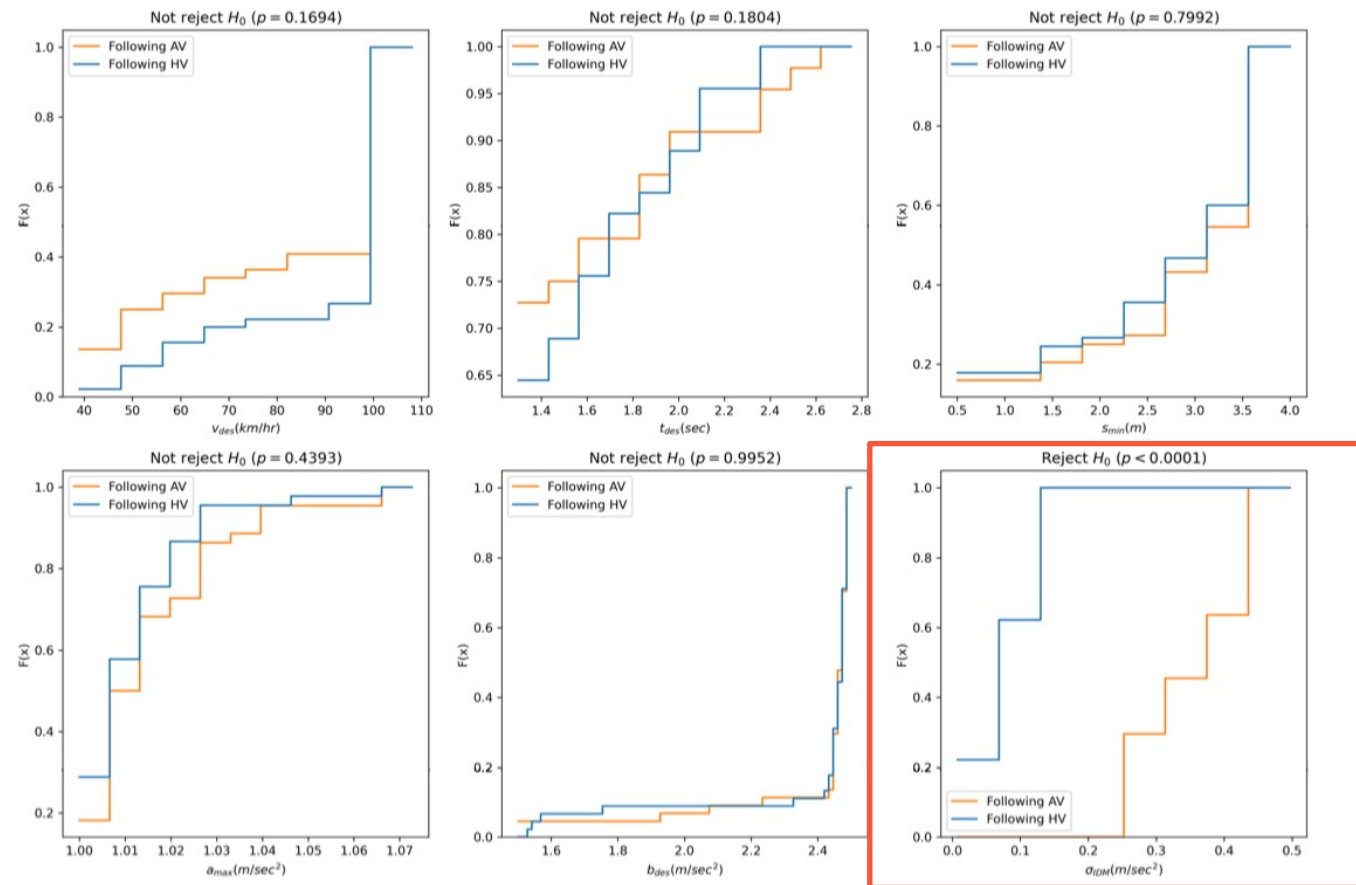
$$v_{t+1} = v_t + a_t \Delta t$$

Model 1: IDM

	Human following			AV following			K-S test p-value
	mean	sd	obs.	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

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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