

Calibrating Car-Following Models using SUMO-in-the-loop and Vehicle Trajectories from Roadside Radar

Maxwell Schrader, Arya Karnik, Alexander Hainen, Joshua Bittle





Can we design a pipeline to calibrate SUMO car-following models using roadside radar?





Considered CF-Models

Intelligent Driver Model

- $a = \text{maximum acceleration } [m/s^2]$
- $b = \max \max \det [m/s^2]$
- $\tau = \text{time headway} [s]$
- $v_0 = \text{desired speed } [m/s]$

Krauss Driver Model

- $a = \text{maximum acceleration } [m/s^2]$
- $b = maximum deceleration [m/s^2]$
- $\tau = \text{time headway} [s]$
- $v_0 = \text{desired speed } [m/s]$

W99 Model

$$\dot{v}_f(v_f, s, \Delta v) = a \left[1 - \left(\frac{v_f}{v_0} \right)^{\beta} - \left(\frac{s^*(v_f, \Delta v)}{s} \right)^2 \right]$$
$$s^*(v_f, \Delta v) = s_0 + \tau + \frac{v_f \Delta v}{2\sqrt{a}}$$

$$\begin{aligned} v_{\text{des}}(t) &= \min\left[v_{\text{safe}}(t), v_f(t) + a, v_0\right] \\ v_{\text{safe}}(t) &= v_l + \frac{g(t) - v_l(t) \cdot \tau}{\frac{v_f}{b \cdot v_f} + \tau} \end{aligned}$$





Car-following Models and Calibration Challenges

- Car-following models (CF-models) are essential components of traffic microsimulation
- Acceleration, deceleration, and speed are highly influential in vehicle-level emissions and fuel consumption^[1]
- Calibration is required for model reliability
- Calibration based on aggregate measures (travel time, queue length, etc..) has a non-unique solution for CF parameters^[2]
- Trajectory-based calibration requires detailed trajectories (NGSIM, etc.) and is computationally expensive
- But, shown to be necessary [3]
- 1. Jie, Li, et al. "Calibration of a microscopic simulation model for emission calculation." *Transportation Research Part C: Emerging Technologies* 31 (2013): 172-184.
- 2. Asamer, Johannes, Henk J. van Zuylen, and Bernhard Heilmann. "Calibrating VISSIM to adverse weather conditions." 2nd International Conference on Models and Technologies for Intelligent Transportation Systems. 2011.
- 3. Schrader, M., Al Abdraboh, M., & Bittle, J. (2023, June). Comparing Measured Driver Behavior Distributions to Results from Car-Following Models using SUMO and Real-World Vehicle Trajectories from Radar: SUMO Default vs. Radar-Measured CF model Parameters. In SUMO Conference Proceedings (Vol. 4, pp. 41-



HE CAPSTONE



Radar Data Fusion

- 6 radars, partially overlapping FOVs
- Vehicle position & velocity recorded every 100ms
- IMM Filtering/Fusion occur in **Frenet Frame** for road-context aware predictions
- Tracklet level fusion







Roadside Radar Fusion, Cont.

- 26 hour period
- 70,000+ Trajectories
 - Contain numerous scenarios
 - Lane-change events, signal queuing
 - Open-sourcing the dataset w/ paper
- Calibration-worthy trajectories identified
 - Trajectories should be composed of many regimes^[1]
 - 2000+ leader-follower pairs

HER EDUCATION



1. Sharma, A., Zheng, Z., & Bhaskar, A. (2019). Is more always better? The impact of vehicular trajectory completeness on car-following model calibration and validation. *Transportation research part B: methodological*, 120, 49-75





Trajectory Calibration

- Calibration handled via Facebook's Nevergrad Meta-optimizer
 - Each trajectory takes ~40s
- Parallelized at trajectory level using Ray



Radar Data

73,048 vehicles

(03/12/2023)



SUMO User Conference 2024

Trajectory

Identification &

Processing

[i] = 2,397 pairs

Radar Data

Fusion,

Transformation,

and Filtering

Trajectory Calibration Results

- Calibration significantly reduces error
- IDM model achieves the best performance

SUMO

- Best-fit model for 81% of vehicles
- Default IDM and Krauss models favor shorter time headways
- W99 performs the best of the default models
 - Calibration without *a* leads to jerky acceleration
- Trade-off between spacing accuracy and acceleration







Trajectory Calibration Results, cont.





Do trajectory calibrated parameter sets create realistic traffic flow in aggregate?

Does it matter in the context of fuel consumption estimation?

Preview of ext. to paper in SUMO Proceedings



Aggregate Calibration / Assessment





SUMO

Aggregate Calibration

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Does it matter?

Fuel Consumption Sensitive to

- 1. Volume
- 2. Fleet Composition
- 3. CF Model & Parameters
- 4. Signal Control Method

Absolute Quantity Uncertain.

Is relative quantity?

- Actuated Coordinated vs. Free
 - Calibrated says ~2.5% decrease in total fuel
 - Krauss says ~6% decrease in total fuel







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Conference Paper

• Maxwell Schrader, Arya Karnik, Alexander Hainen, and Joshua Bittle. (2024) Calibrating Car-Following Models using SUMO-in-the-loop and Vehicle Trajectories from Roadside Radar. Presented at SUMO User Conference.





Questions?

Trajectory Calibration: github.com/UnivOfAlabama-BittleResearchGroup/sumo-cf-calibration

Aggregate Calibration: github.com/UnivOfAlabama-BittleResearchGroup/traffic-simulation-calibration

Parallelized SUMO Pipelines: <u>https://github.com/mschrader15/sumo-pipelines</u>

