

# Foundation models

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# For autonomous driving

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Vincent Vanhoucke  
Distinguished Engineer  
Waymo



OUR MISSION –

# Be the world's most trusted driver.





A panoramic view of the San Francisco skyline, featuring the Transamerica Pyramid and other high-rise buildings against a clear blue sky. The foreground shows a dense residential area with white buildings.

# San Francisco

A view of the Phoenix skyline, showing various skyscrapers and a large mountain range in the background under a hazy sky. A marina with many sailboats is visible in the foreground.

# phoenix

A view of the Los Angeles skyline, featuring several tall skyscrapers, including the US Bank Tower, and a mix of urban buildings. The foreground shows a parking lot and some lower-level structures.

# Los Angeles

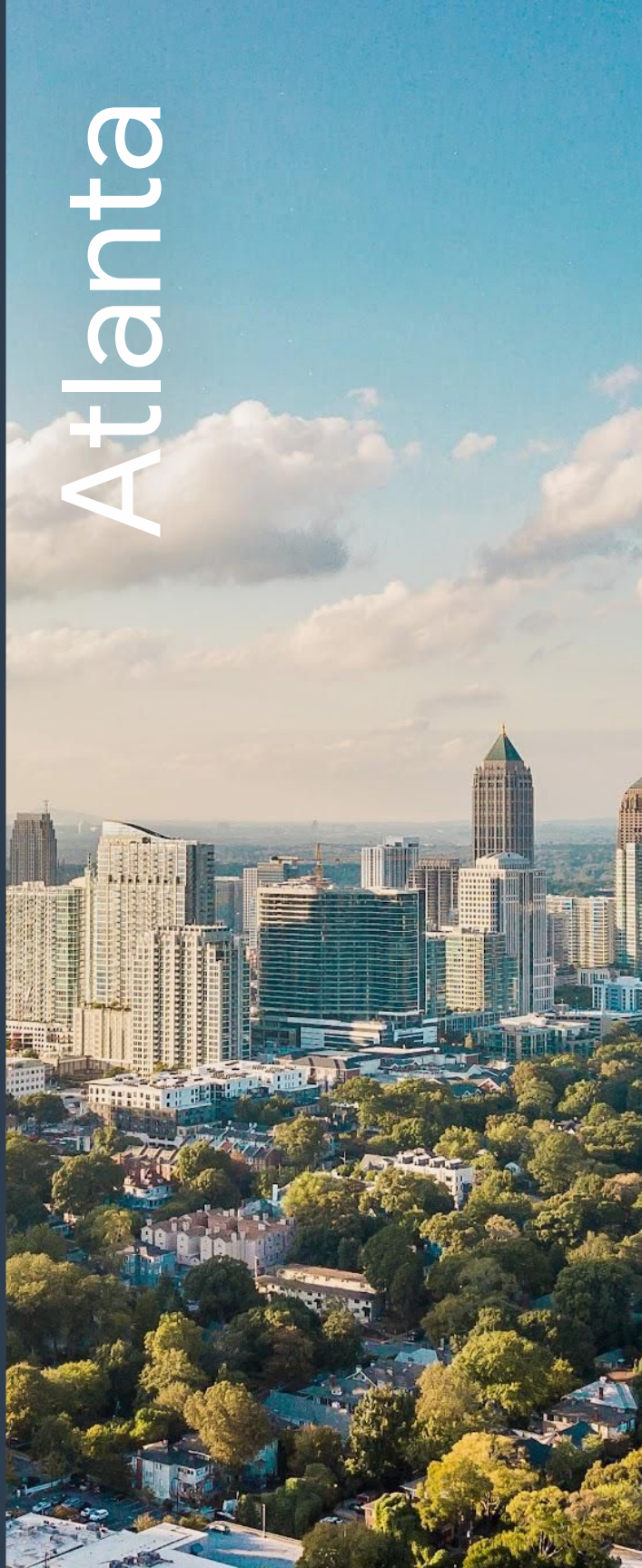
A view of the Austin skyline, showing a mix of modern and older buildings. A large bridge spans a river in the foreground, and a green park area is visible. The sky is blue with some clouds.

# Austin



Up next

WAYMO ONE x Uber



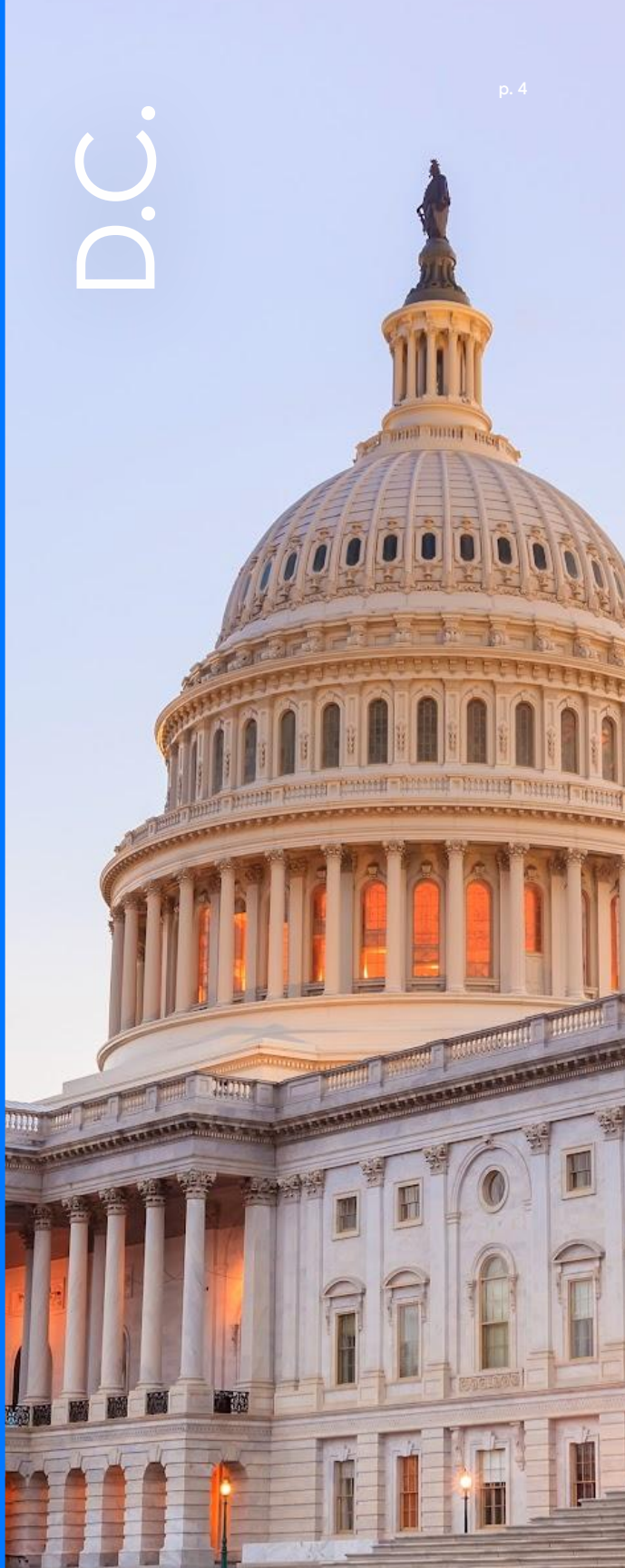
Atlanta

WAYMO ONE



Miami

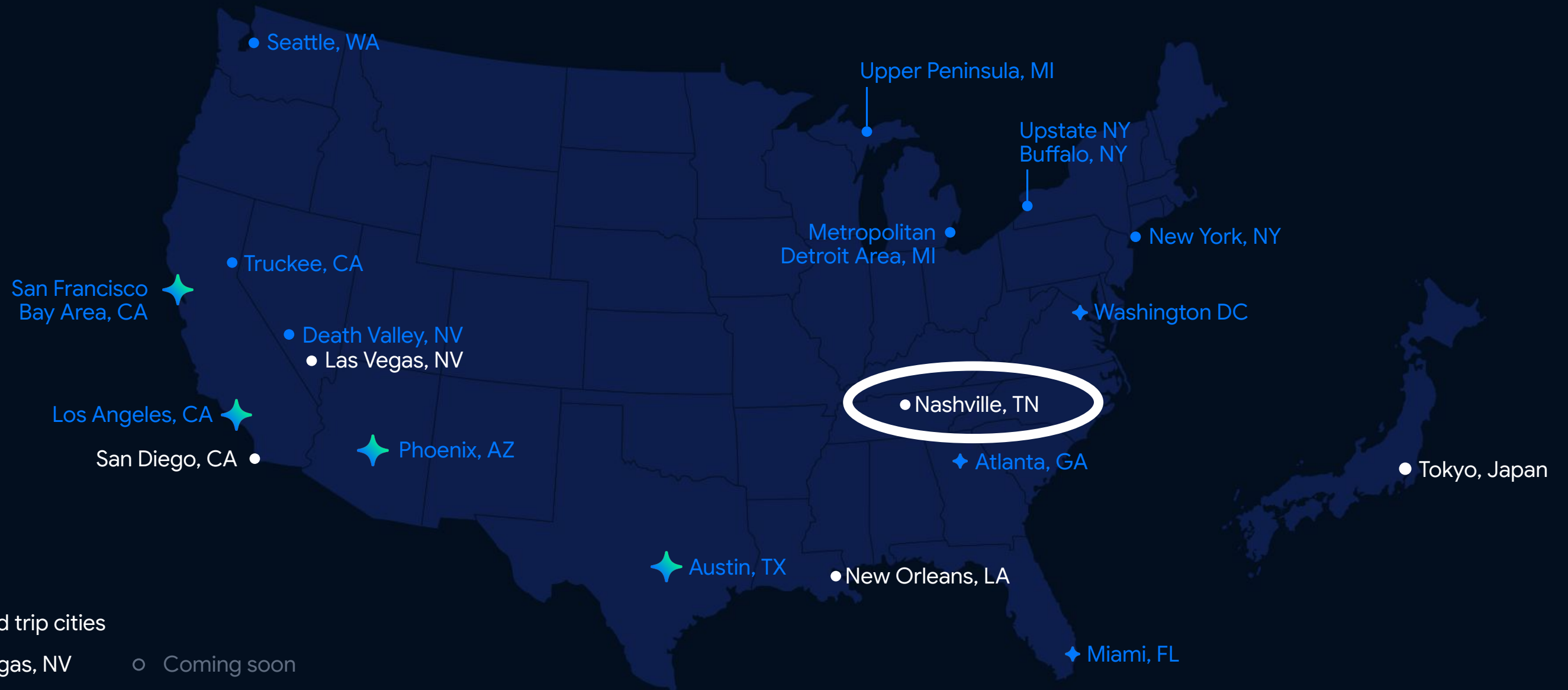
WAYMO ONE



D.C.



✦ Waymo operating cities   ✦ Operation coming soon   ● Waymo tested cities



#### 2025 Road trip cities

- |                   |               |
|-------------------|---------------|
| ● Las Vegas, NV   | ○ Coming soon |
| ● San, Diego, CA  | ○ Coming soon |
| ● New Orleans, LA | ○ Coming soon |
| ● Nashville, TN   | ○ Coming soon |
| ○ Coming soon     | ○ Coming soon |

Road trip ✦



WAYMO ONE

Now  
serving over

250,000

paid trips  
per week





# Protecting Vulnerable Road Users

Compared to human drivers over **56.7 million** miles in our operational cities, Waymo Driver had:



92%

Fewer crashes with injuries to pedestrians



82%

Fewer crashes with injuries to cyclists



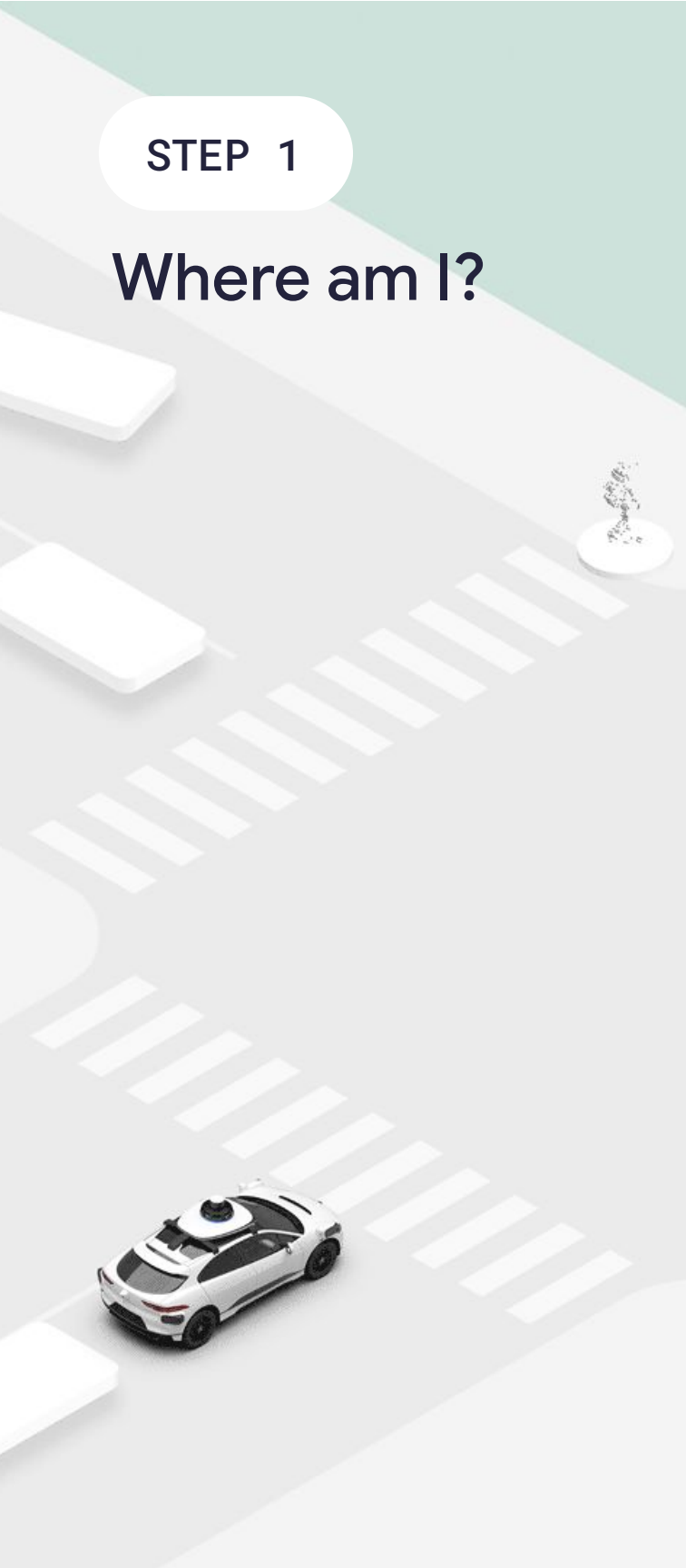
82%

Fewer crashes with injuries to motorcyclists



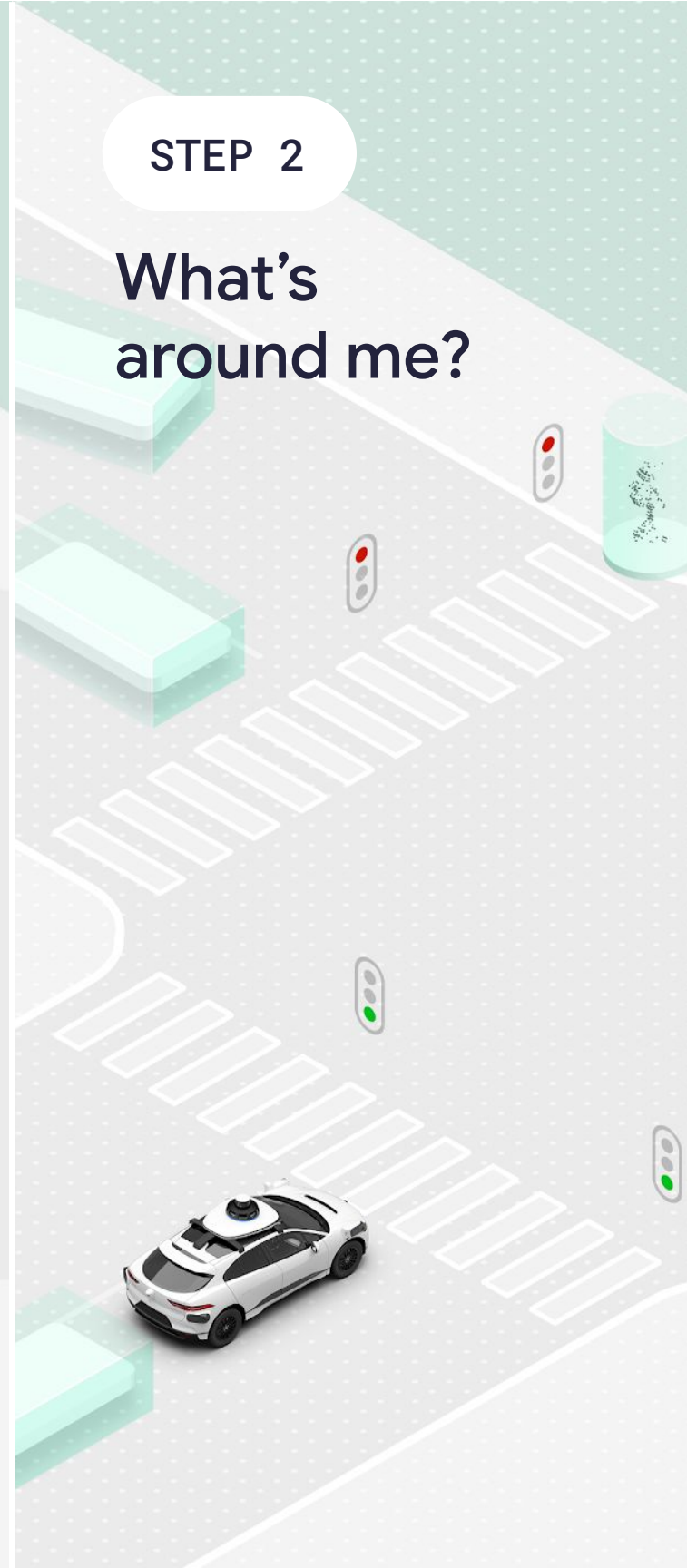
## STEP 1

Where am I?



## STEP 2

What's around me?



## STEP 3

What will happen next?



## STEP 4

What should I do?









Long tail of driving scenarios presents significant challenges

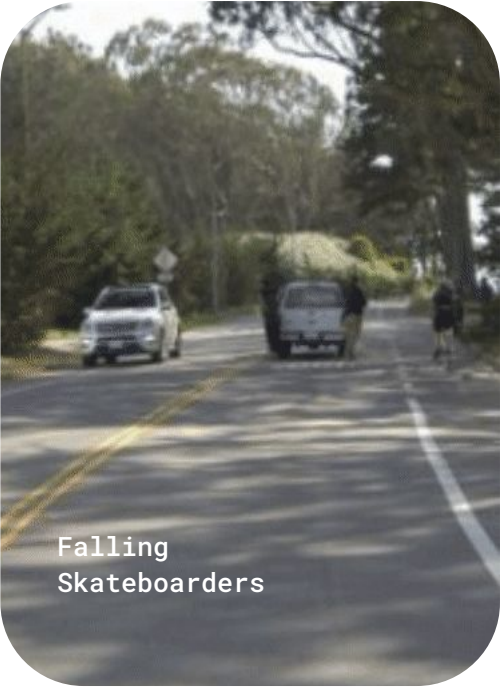
Unusual Behaviors



Stray Cyclists



Toddlers on the Loose



Falling Skateboarders

Foreign Objects on the Road



A BBQ Grill Falling Off Truck

Extreme Weather



Falling Trees



Flooding

Unique Interactions



Road Blockages

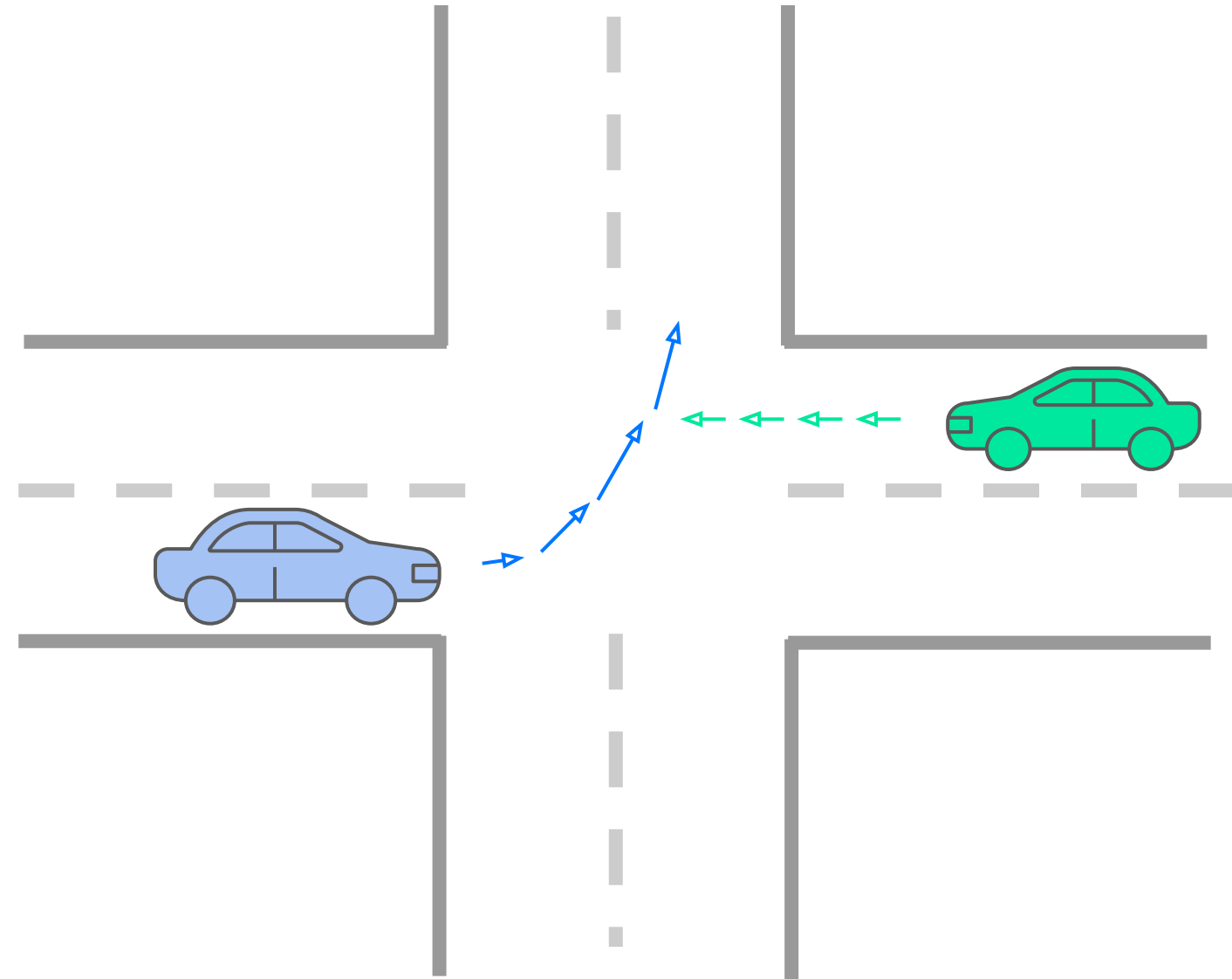


Last Minute Lane Changes

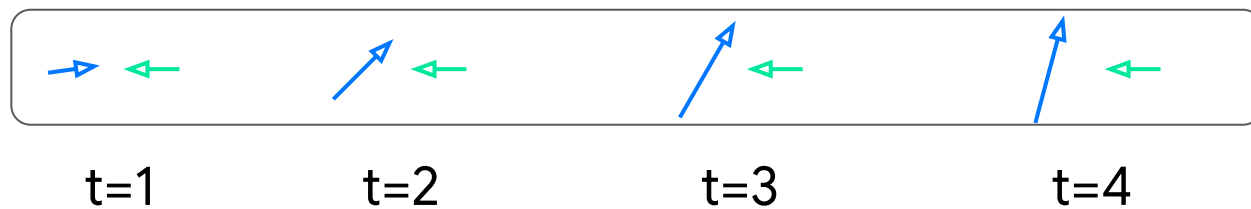


# Driving as a Conversation

- Trajectories as sentences in a **new language**
- Vocabulary consists of **state / motion words** (vectors)
- Like language, trajectories have **local continuity and global context**
- Model architecture is **similar to** that of a **LLM**



Motion token sequence:



## MotionLM: Multi-Agent Motion Forecasting as Language Modeling

Ari Seff, Brian Cera, Dian Chen, Mason Ng, Aurick Zhou, Nigamaa Nayakanti, Khaled S. Refaat, Rami Al-Rfou, Benjamin Sapp

ICCV 2023



# Examples

MARGINAL PREDICTION



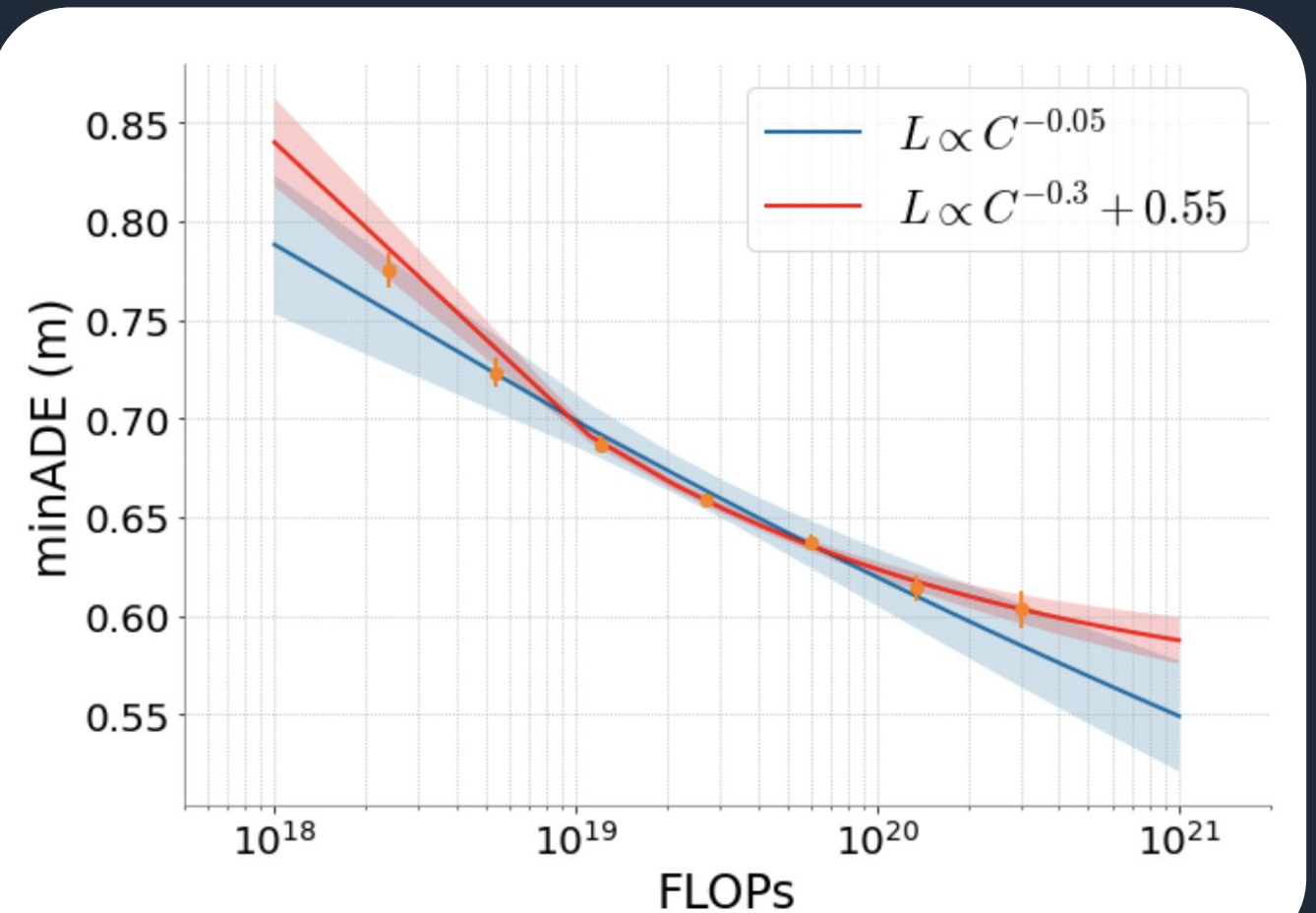
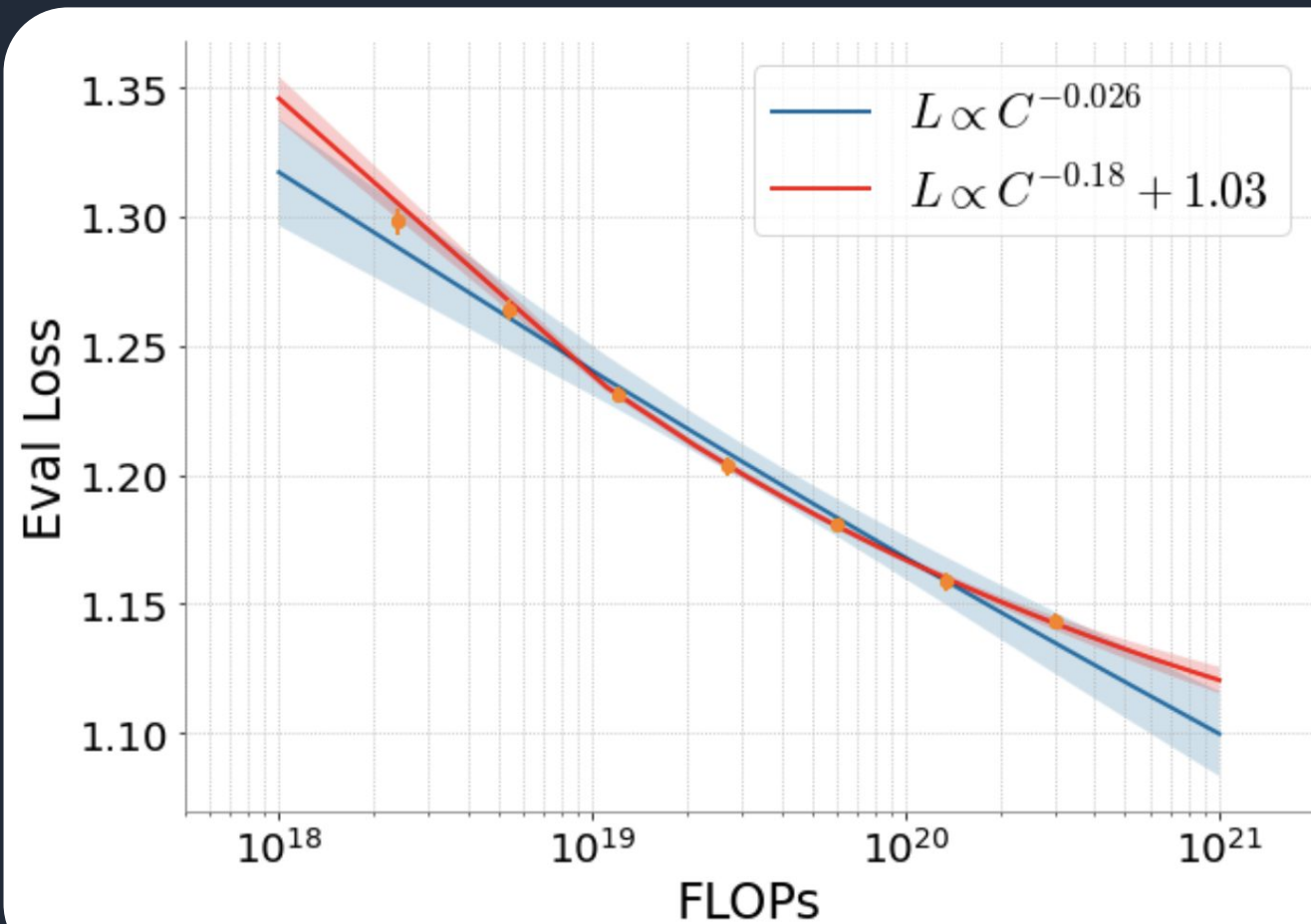
CAUSAL JOINT "DIALOGUE"





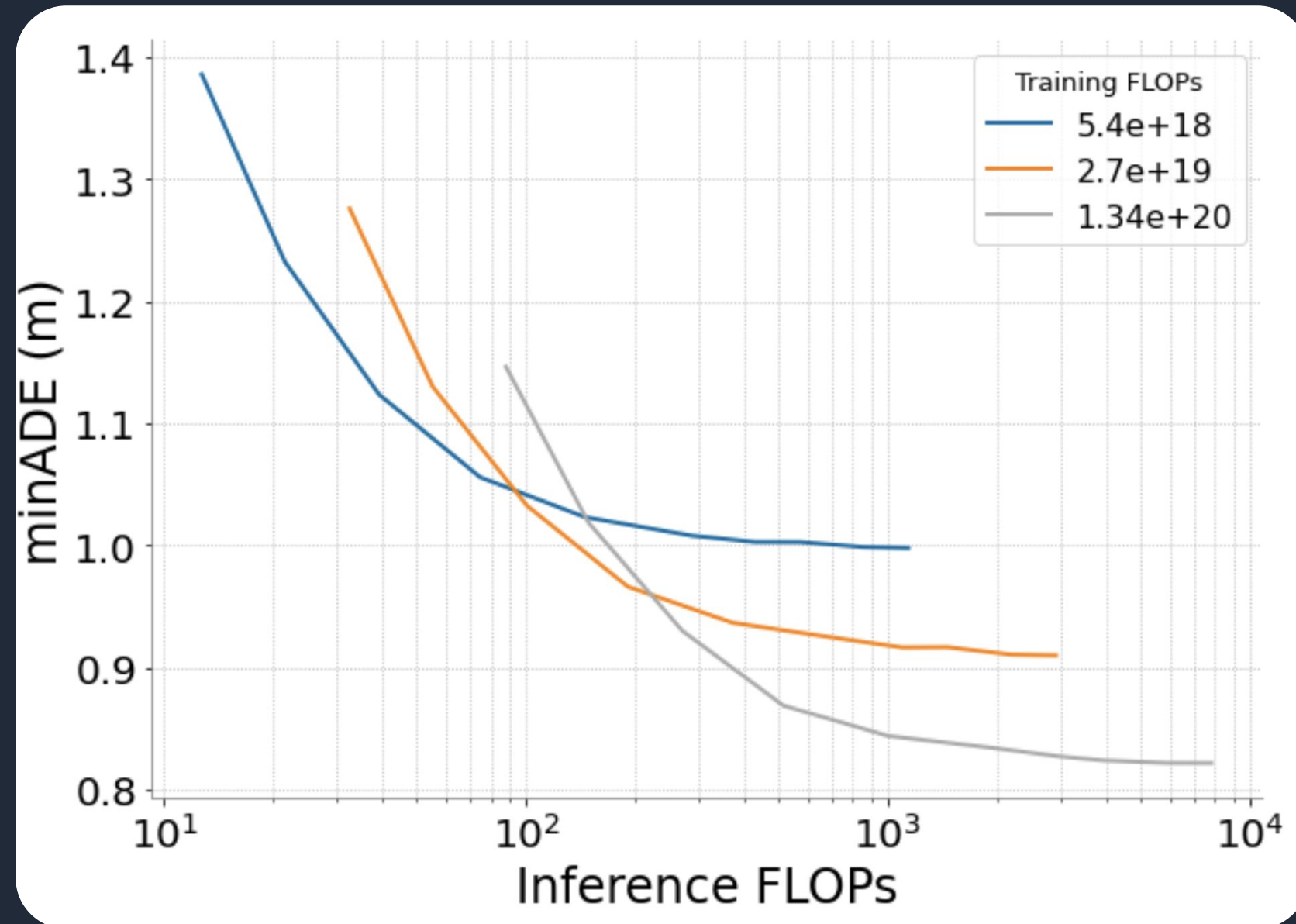
## Scaling Laws

Once you have a good architecture, performance scales with model size and data



## Inference Scaling Laws

Performance scales with the amount of compute used at runtime



For more on motion scaling laws, don't miss **Ben Sapp's** talk tomorrow at the Workshop on Distillation of Foundation Models for Autonomous Driving!

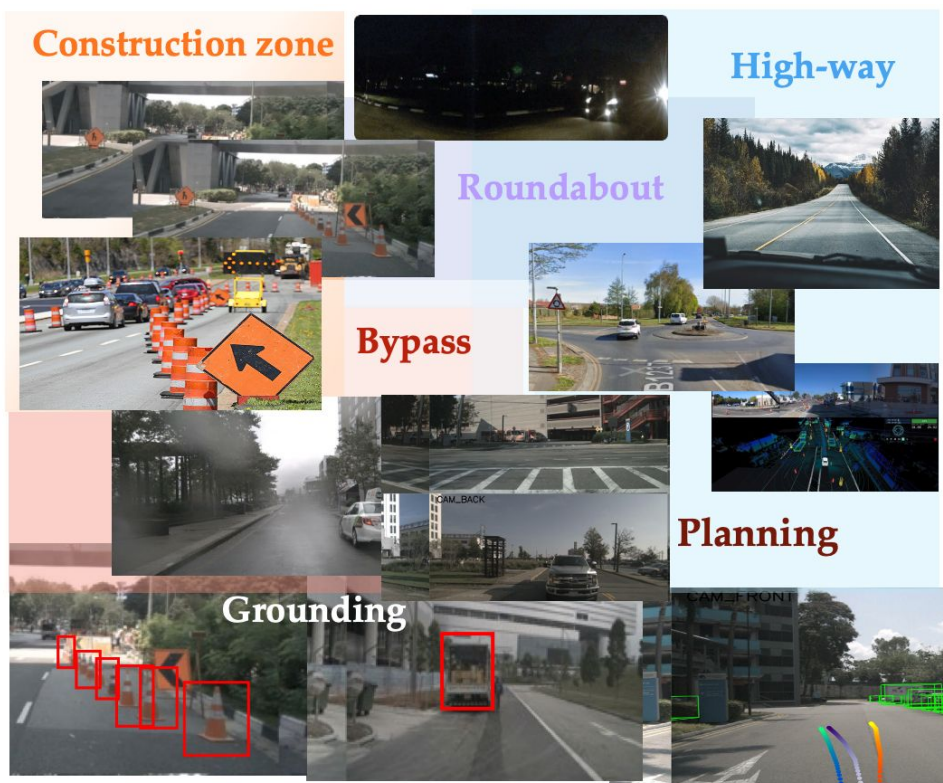
(NEW!) Now on ArXiv






# Post-training Preference Alignment

Large scale driving demonstration data



Let's remember what the expert did and copy them!


$$\max_{\theta} \mathbb{E}_{(\xi, c) \sim \mathcal{D}} [\Pi_{t=0}^T \mathbb{P}_{\theta}(\mathbf{a}_t^* | \mathbf{a}_{<t}^*; c)]$$

Misalignment: by optimizing an incomplete or mis-specified objective, these models lead to **undesirable** behaviors at best and **safety hazards** at worst!

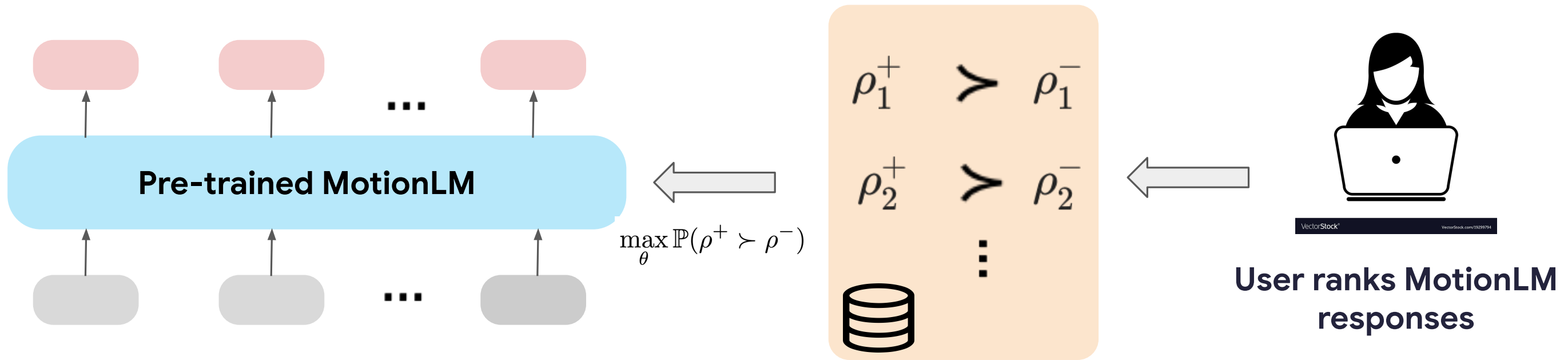
Direct Post-Training Preference Alignment for Multi-Agent Motion Generation Model

Using Implicit Feedback from Pre-training Demonstrations

Thomas (Ran) Tian, Kratarth Goel

ICLR 2025, Spotlight

# Post-training Preference Alignment



Reconcile the disparity between the next-token prediction objective and human preferences.

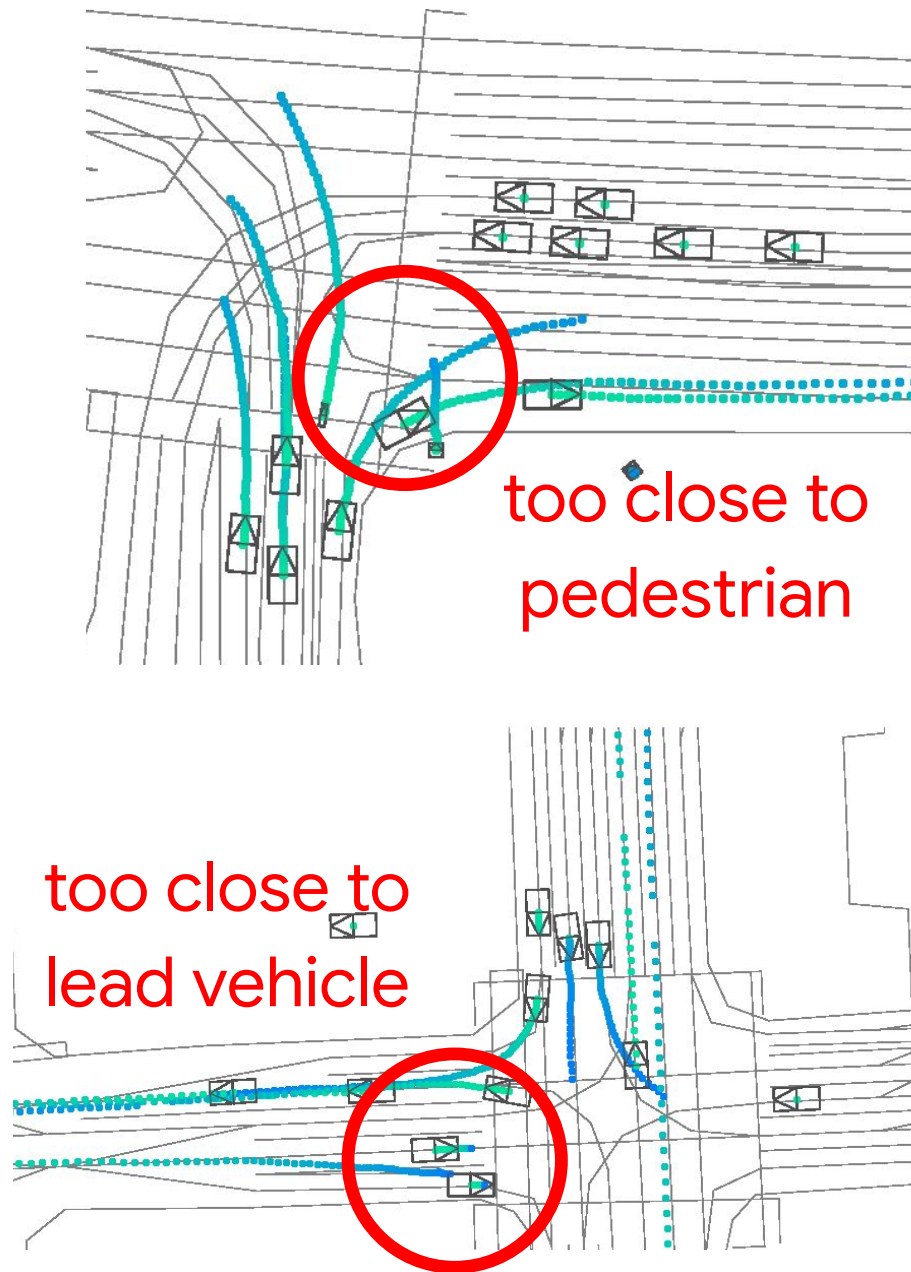
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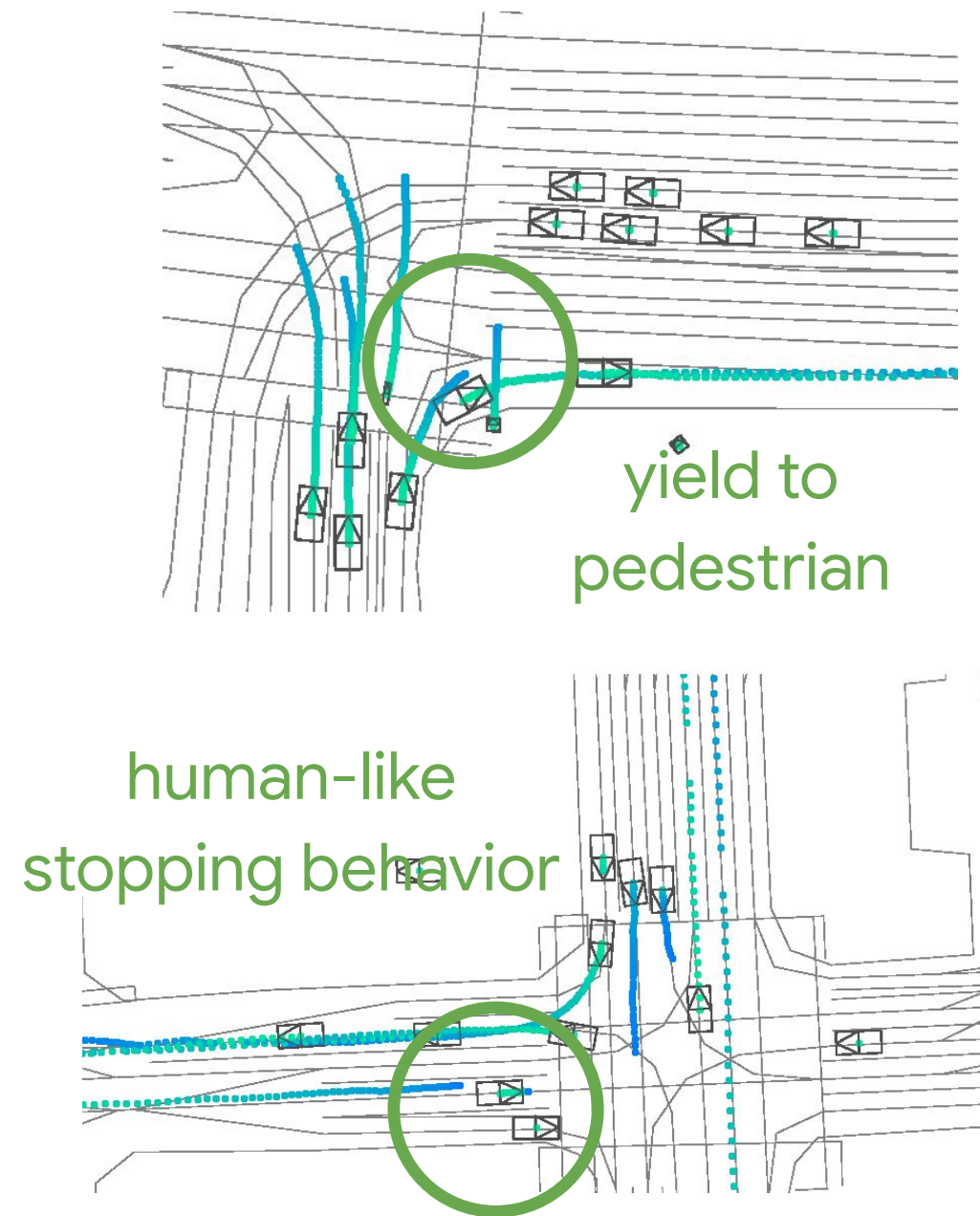
ICLR 2025, Spotlight



## Pre-trained traffic simulation model



## After post-training alignment





# Leveraging language understanding and general reasoning capabilities of LLM/VLMs



It's Wednesday at 4pm. Can I park at this spot right now?  
Tell me in 1 line.

VLM: Yes, you can park for up to 1 hour starting at 4pm.

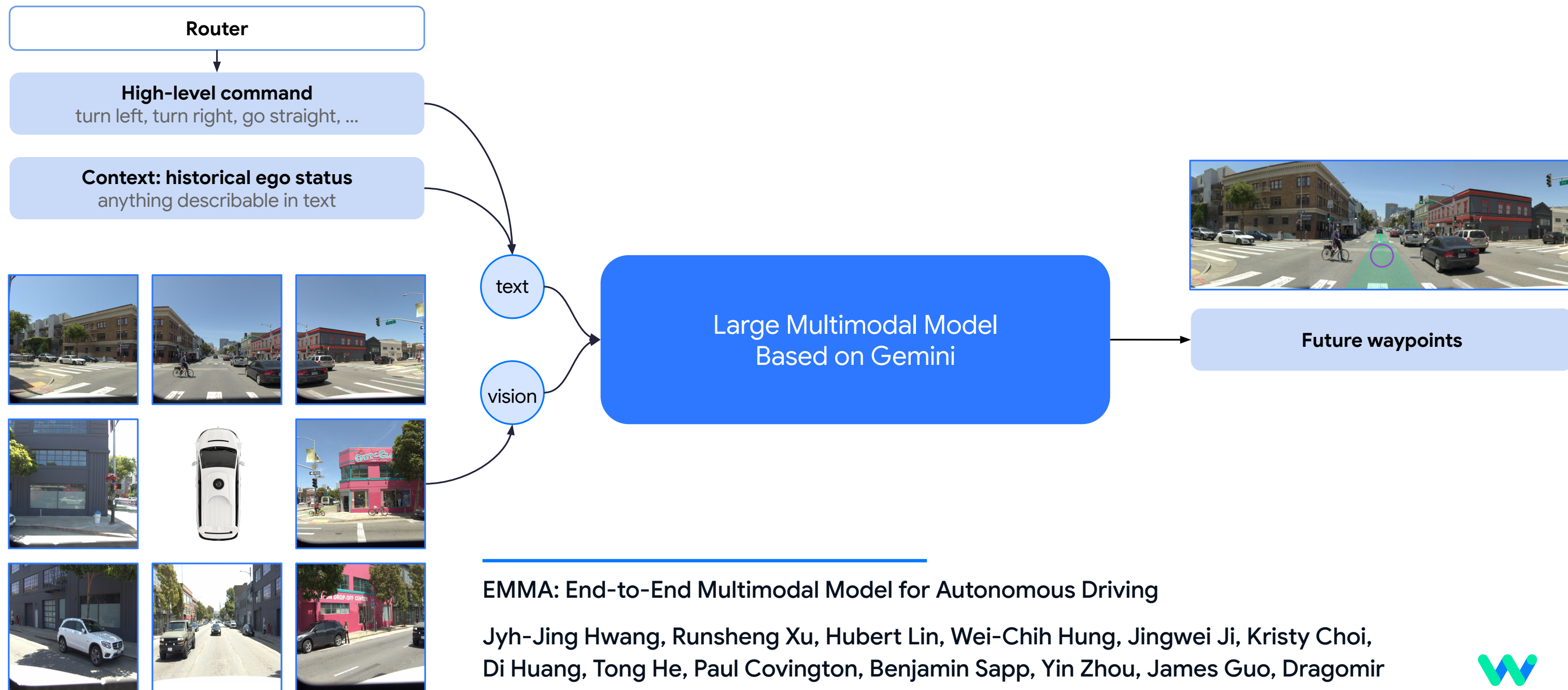


I am driving on the street, this is what I see from the front window. Tell me what's happening and what I should pay special attention to. And tell me what I should do and specify the reasons.

VLM: There's been a serious accident! A car has flipped onto its roof and is blocking the streetcar tracks. Emergency responders are on the scene. Pay close attention to the emergency personnel and any instructions they may give. You should slow down significantly, proceed with extreme caution, and be prepared to stop completely if necessary. The most important thing is to avoid hindering the emergency response and to ensure your own safety and the safety of others around you. If possible, consider an alternate route to avoid the area entirely.



# Multimodal Models for AV motion prediction



## EMMA: End-to-End Multimodal Model for Autonomous Driving

Jyh-Jing Hwang, Runsheng Xu, Hubert Lin, Wei-Chih Hung, Jingwei Ji, Kristy Choi, Di Huang, Tong He, Paul Covington, Benjamin Sapp, Yin Zhou, James Guo, Dragomir Anguelov, Mingxing Tan

# Competitive Academic Benchmark Results

## WAYMO OPEN MOTION DATASET

Method	ADE 1s	ADE 3s	ADE 5s	ADE 8s
MotionLM <a href="#">[Seff et al., 2023]</a>	0.045	0.251	0.694	1.766
Wayformer* <a href="#">[Nayakanti et al., 2023]</a>	0.044	0.250	0.640	<b>1.517</b>
EMMA	0.032	0.248	0.681	1.718
EMMA+	<b>0.030</b>	<b>0.225</b>	<b>0.610</b>	1.553

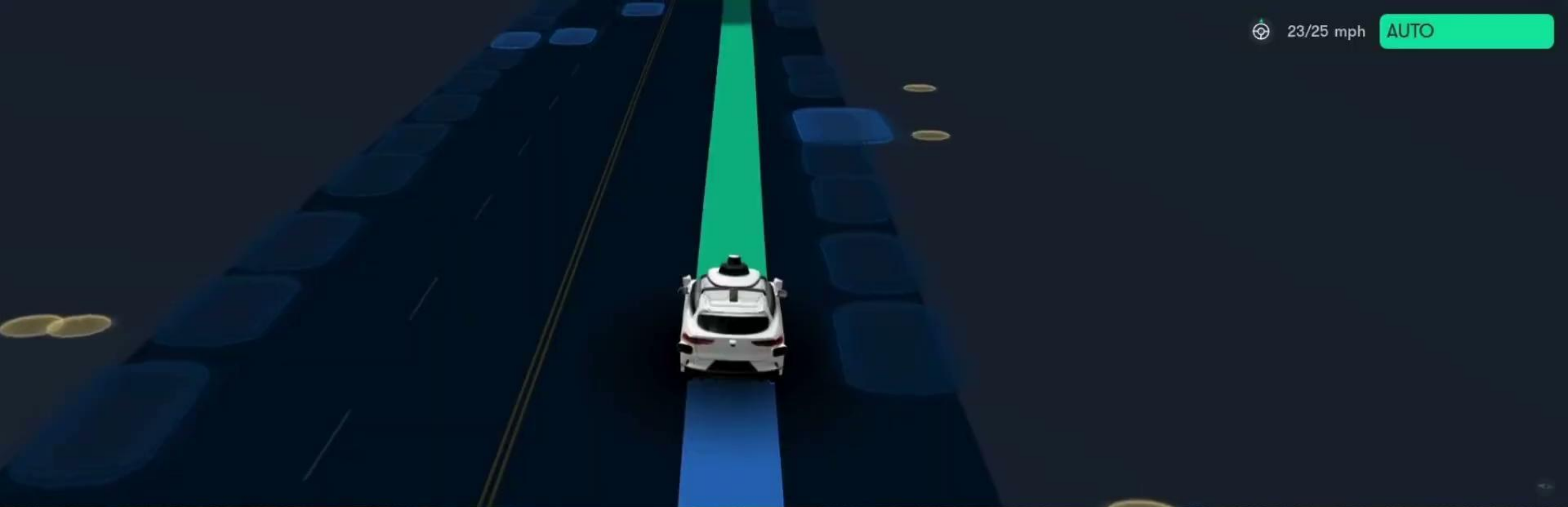
## NUSCENES DATASET

Method	self-supervised?	L2 (m) 1s	L2 (m) 2s	L2 (m) 3s	Avg L2 (m)
UniAD <a href="#">[Hu et al., 2023]</a>	✗	0.42	0.64	0.91	0.66
DriveVLM <a href="#">[Tian et al., 2024]</a>	✗	0.18	0.34	0.68	0.40
VAD <a href="#">[Jiang et al., 2023]</a>	✗	0.17	0.34	0.60	0.37
OmniDrive <a href="#">[Wang et al., 2024a]</a>	✗	0.14	0.29	0.55	0.33
DriveVLM-Dual <a href="#">[Tian et al., 2024]</a>	✗	0.15	0.29	0.48	0.31
Ego-MLP* <a href="#">[Zhai et al., 2023]</a>	✓	0.15	0.32	0.59	0.35
BEV-Planner <a href="#">[Li et al., 2024]</a>	✓	0.16	0.32	0.57	0.35
EMMA (random init)	✓	0.15	0.33	0.63	0.37
EMMA	✓	0.14	0.29	0.54	0.32
EMMA+	✓	<b>0.13</b>	<b>0.27</b>	<b>0.48</b>	<b>0.29</b>



Foundation models open up new, scalable avenues for deep semantic understanding of the world and of human behavior.

Ultimately, these advances are in service of building a product that can delight riders and earn their trust every day.



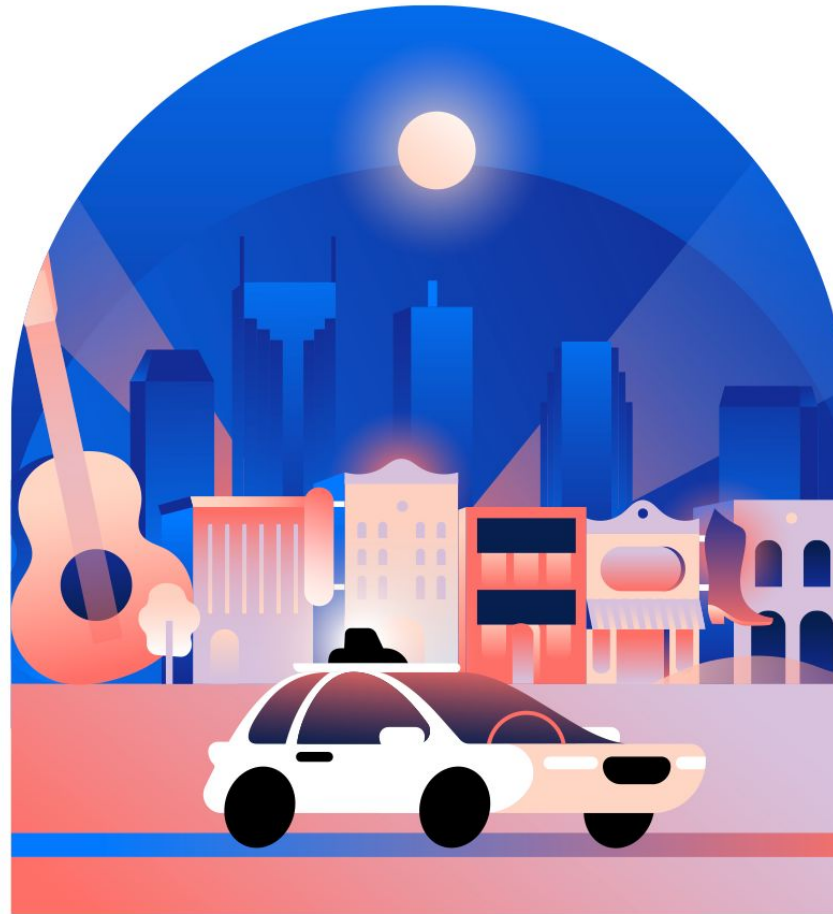


If you're interested in building **foundation models** for robots operating in the real world **today**.

Join us on this ride!



[vanhoucke@waymo.com](mailto:vanhoucke@waymo.com)



CVPR 2025

Come chat with  
me at **Booth 1223!**

June 11-15, 2025

