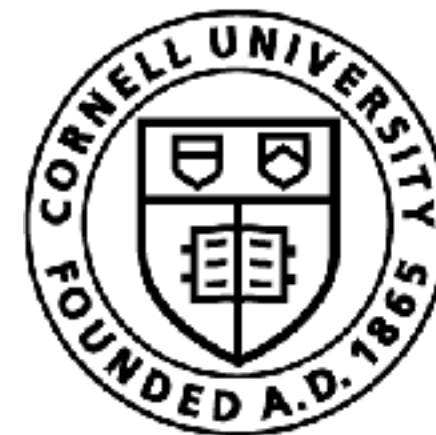


CS 4756/5756: Robot Learning

Sanjiban Choudhury



Cornell Bowers CIS
Computer Science

WHAT A TIME TO BE
T I M E
A L L I V E !

2024 continues to be an exciting year for
Machine Learning

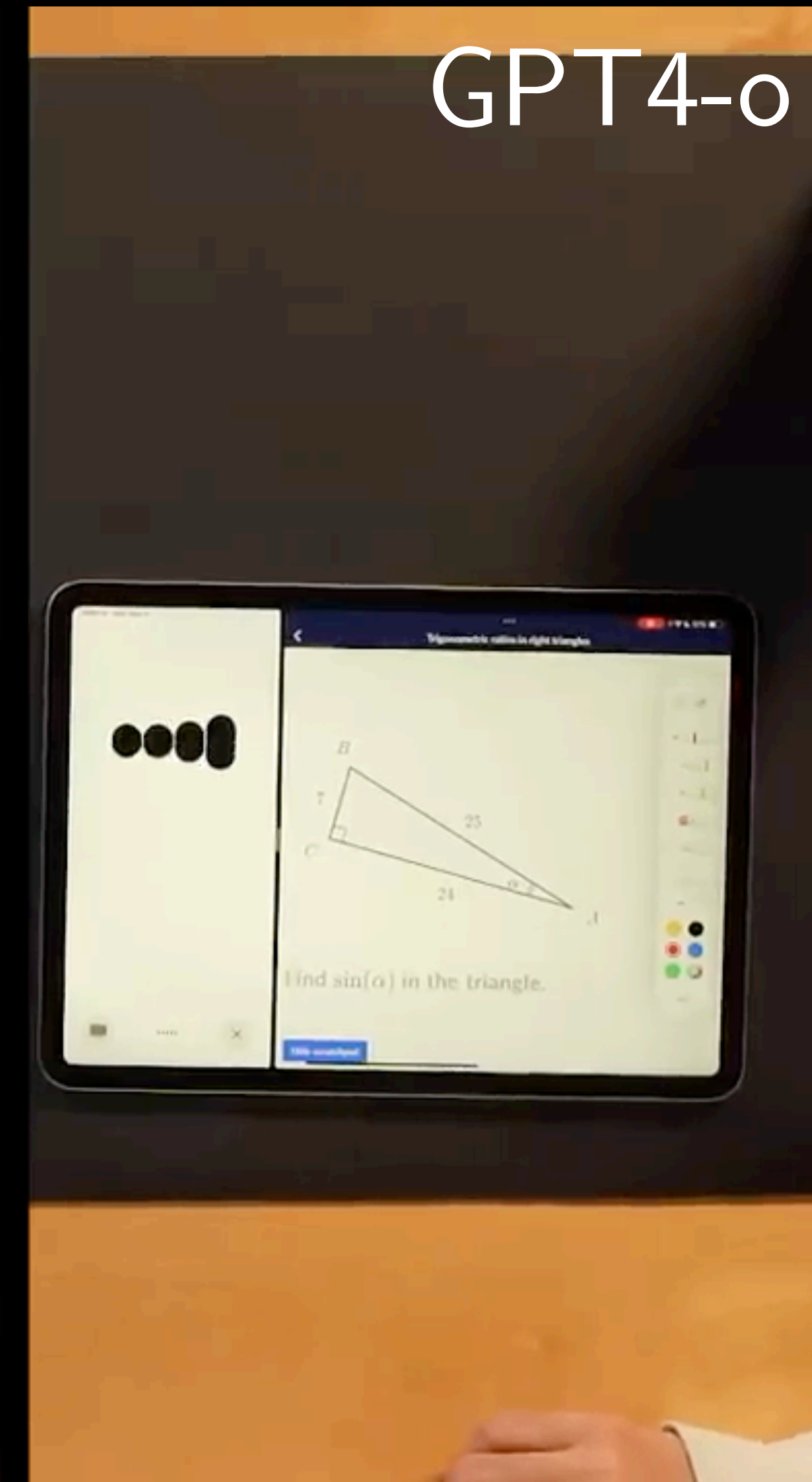
Different things we can do with GPT-4/LLMs/Transformers?



GPT-4

Three advances in the last semester
that I am excited about:

#1: Real-time, multi-modal interactions



#2: Video Generation

SORA

A litter of golden retriever puppies playing in the snow.
their heads pop out of the snow

#3: LLMs that act

SWE-Bench

Problem source Write here

Between releases 3.0.0rc8 and 3.0.0rc9, DateTime fields have started throwing an error when being instantiated as inner fields of container fields like List or Tuple. The snippet below works in <=3.0.0rc8 and throws the error below in >=3.0.0rc9 (and, worryingly, 3.0.0):

```
"""python
from marshmallow import fields, Schema

class MySchema(Schema):
    times = fields.List(fields.DateTime())

s = MySchema()
...

Traceback:

...

Traceback (most recent call last):
  File "test-mm.py", line 8, in <module>
    s = MySchema()
  File "/Users/victor/.pyenv/versions/marshmallow/lib/python3.6/site-packages/marshmallow/schema.py", line 383, in __init__
    self.fields = self._init_fields()
  File "/Users/victor/.pyenv/versions/marshmallow/lib/python3.6/site-packages/marshmallow/schema.py", line 913, in _init_fields
```

Local repo path or GitHub URL <https://github.com/swe-agent-demo/marshmallow>

▶ Run ■ Stop

[GitHub readme](#)

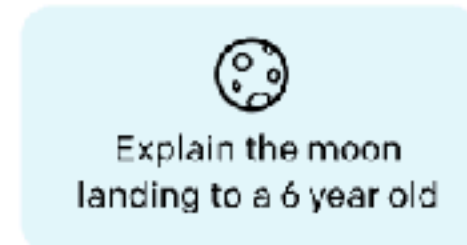
What algorithms are powering these
advancements?

Reinforcement Learning from Human Feedback (RLHF)

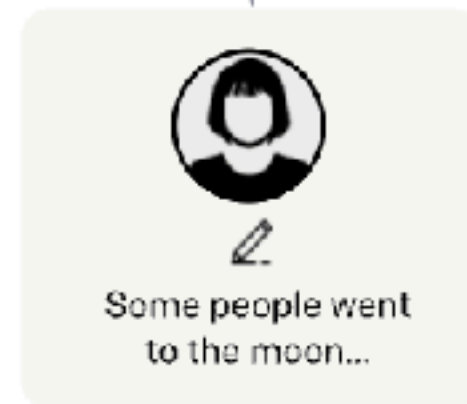
Step 1

Collect demonstration data, and train a supervised policy.

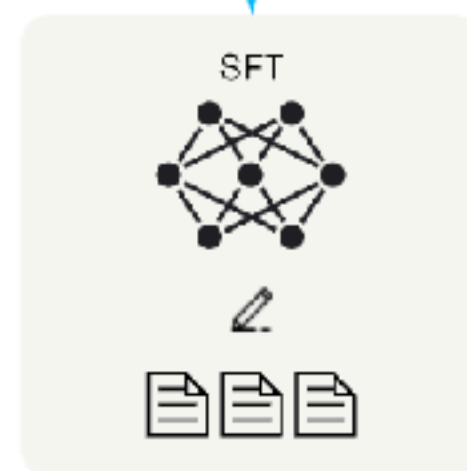
A prompt is sampled from our prompt dataset.



A labeler demonstrates the desired output behavior.



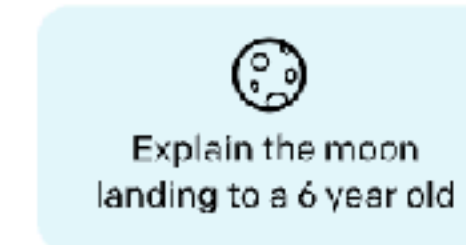
This data is used to fine-tune GPT-3 with supervised learning.



Step 2

Collect comparison data, and train a reward model.

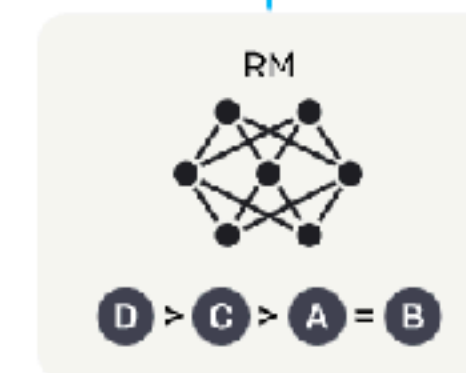
A prompt and several model outputs are sampled.



A labeler ranks the outputs from best to worst.



This data is used to train our reward model.



Step 3

Optimize a policy against the reward model using reinforcement learning.

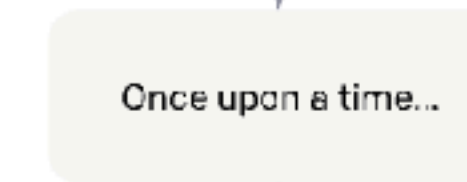
A new prompt is sampled from the dataset.



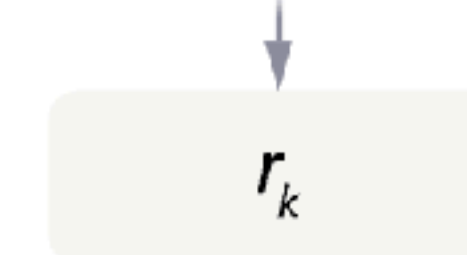
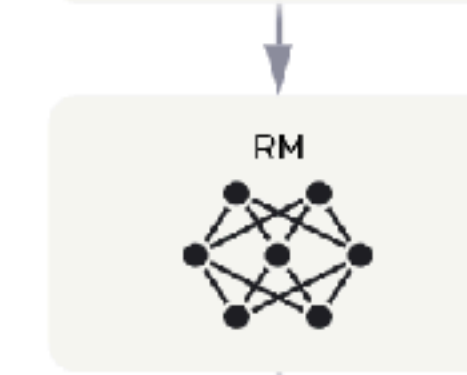
The policy generates an output.



The reward model calculates a reward for the output.



The reward is used to update the policy using PPO.



Open-source fine-tunable models



LLAMA



Alpaca

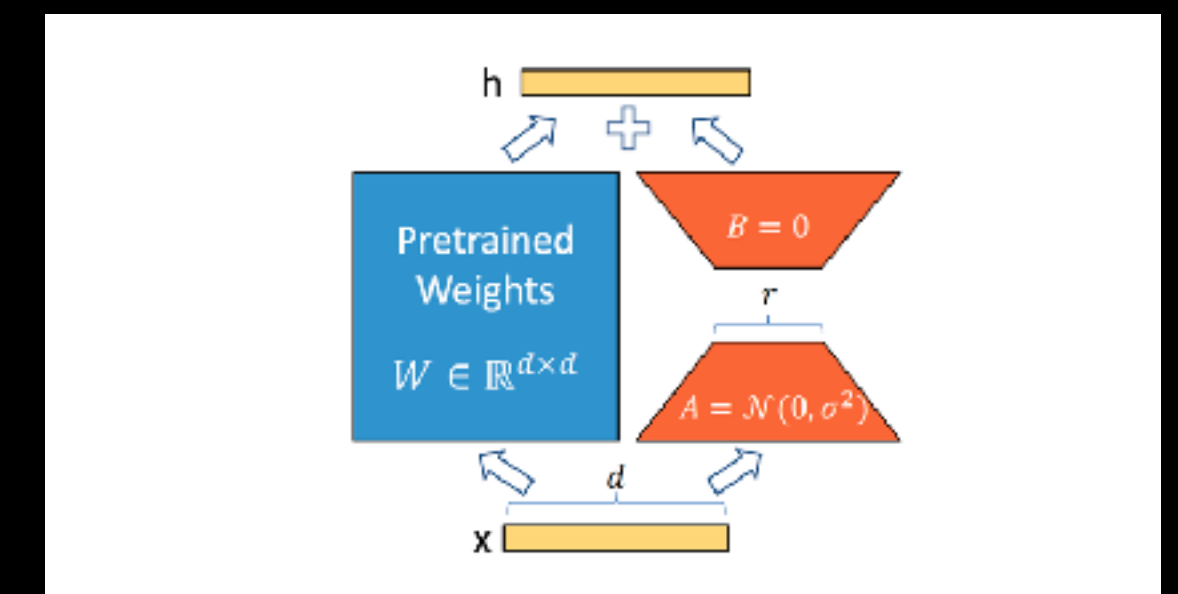


Vicuna

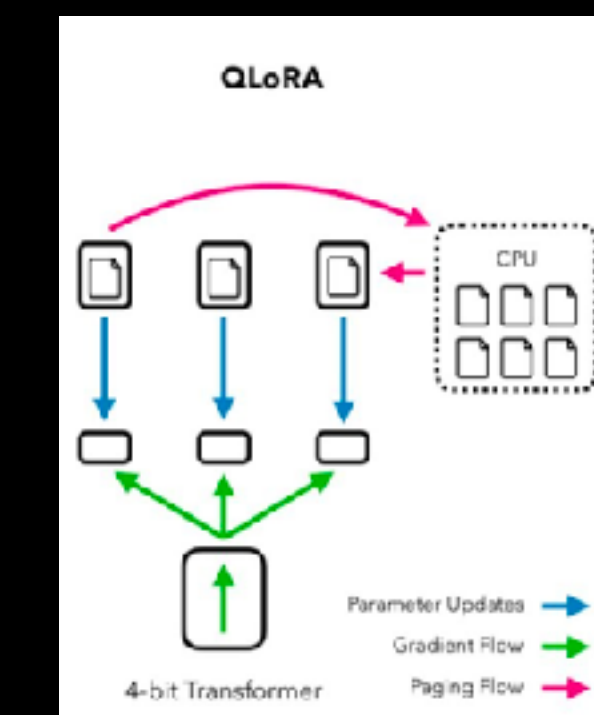


Mistral

Parameter Efficient Fine Tuning (PEFT)

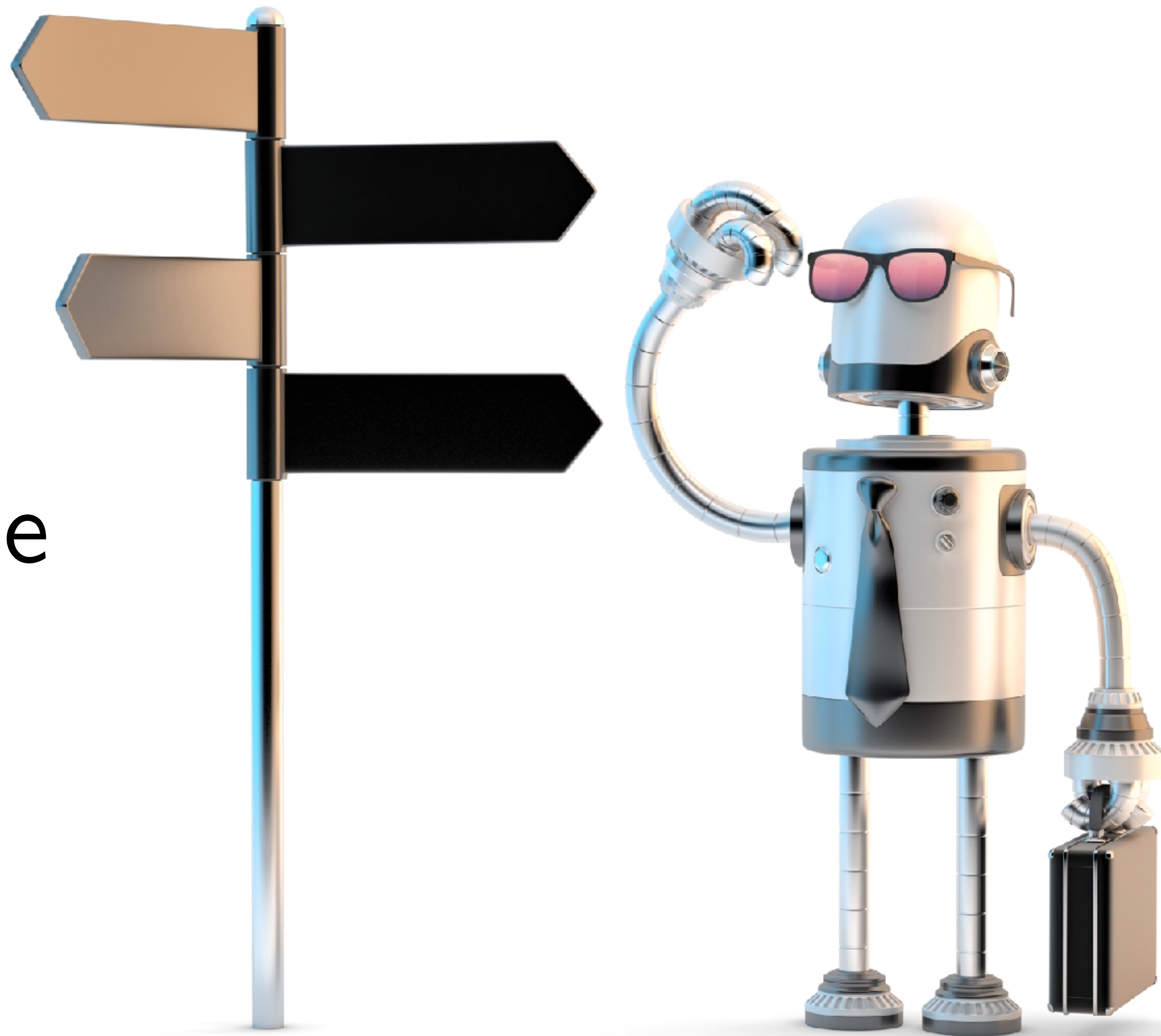


LORA

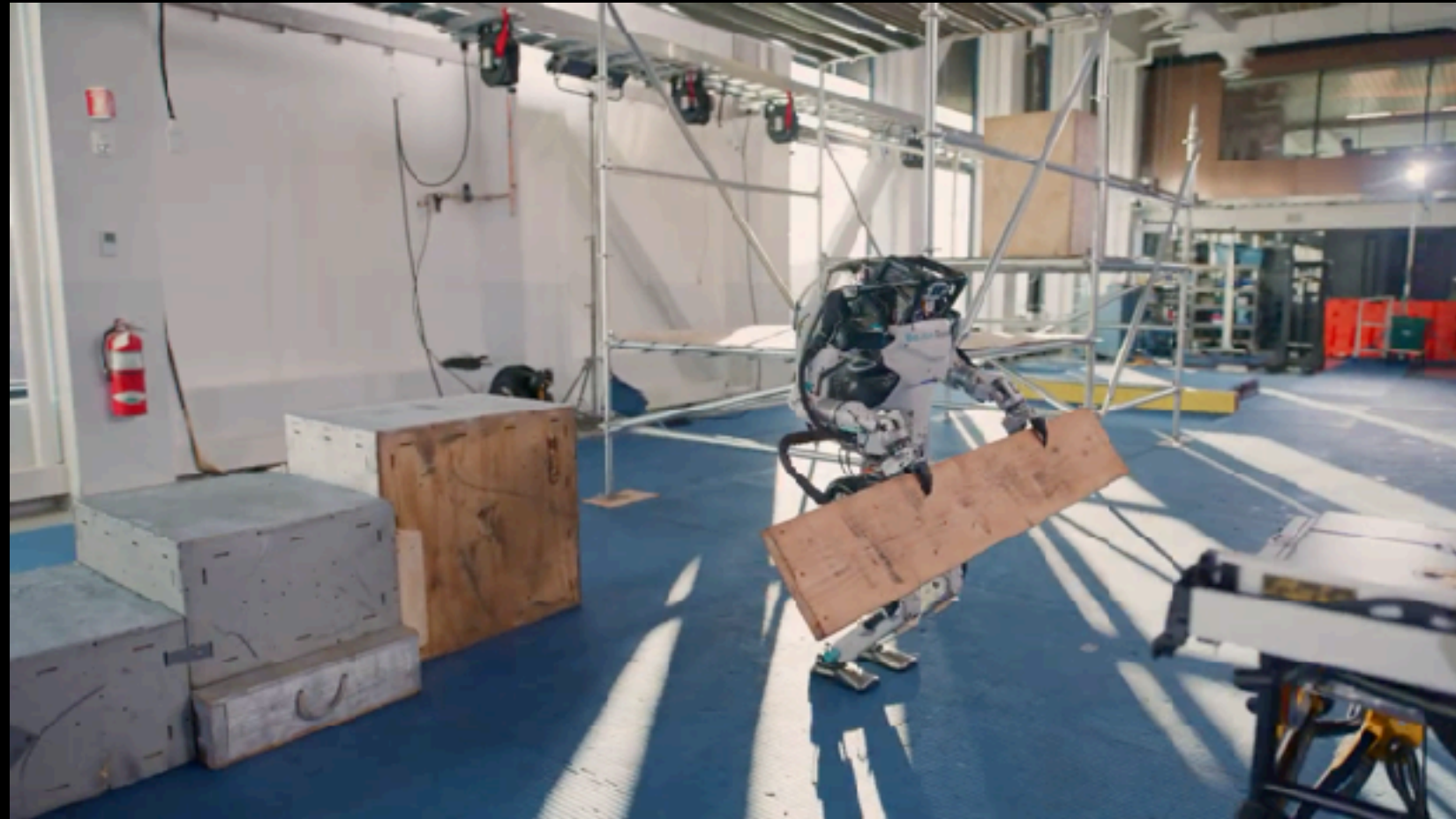


QLORA

Where are the
robots?



Rise of the Humanoids



Boston Dynamics



Tesla



Agility Robotics



Figure AI

Quadrupeds going strong



Boston Dynamics SPOT

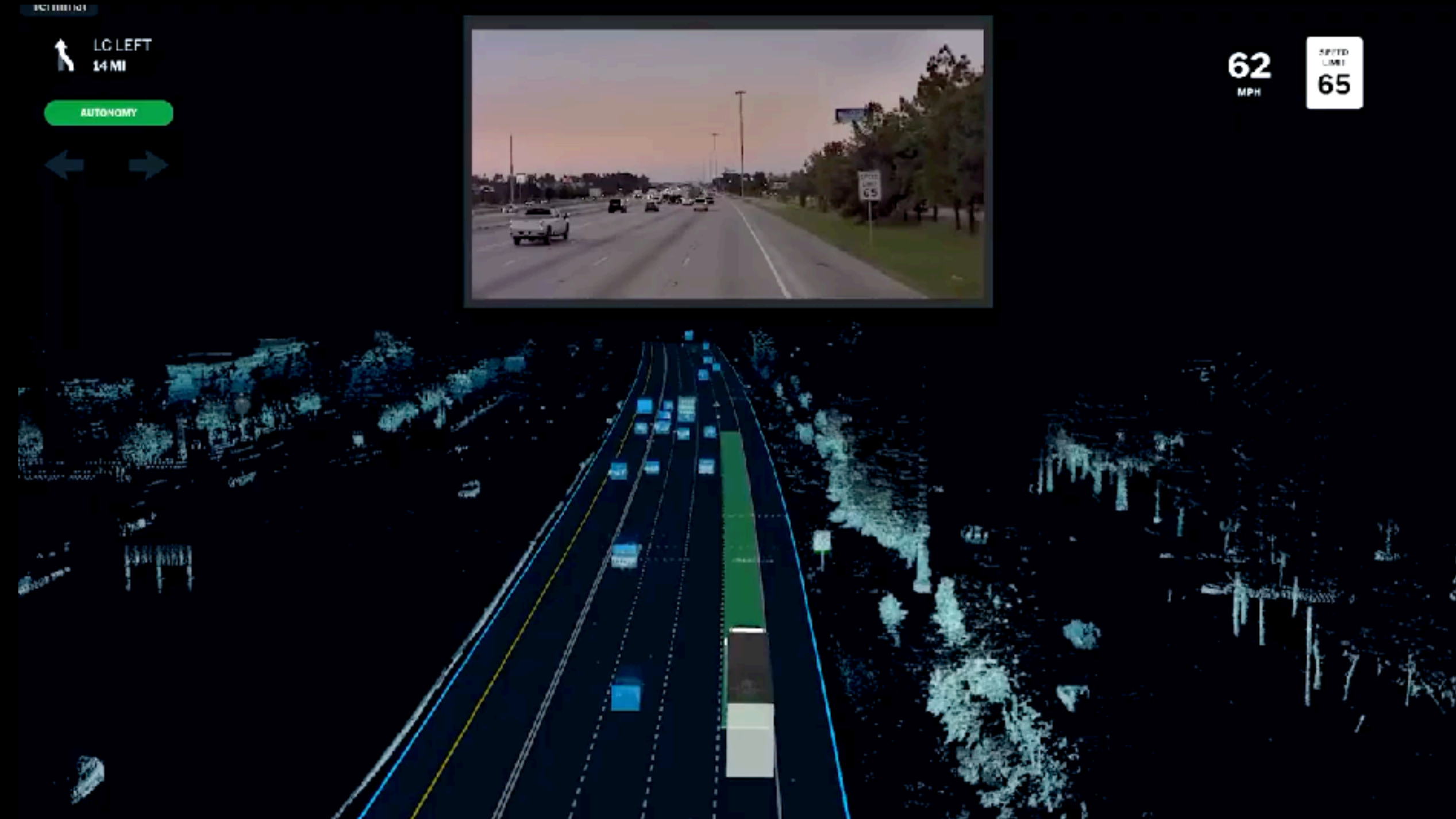
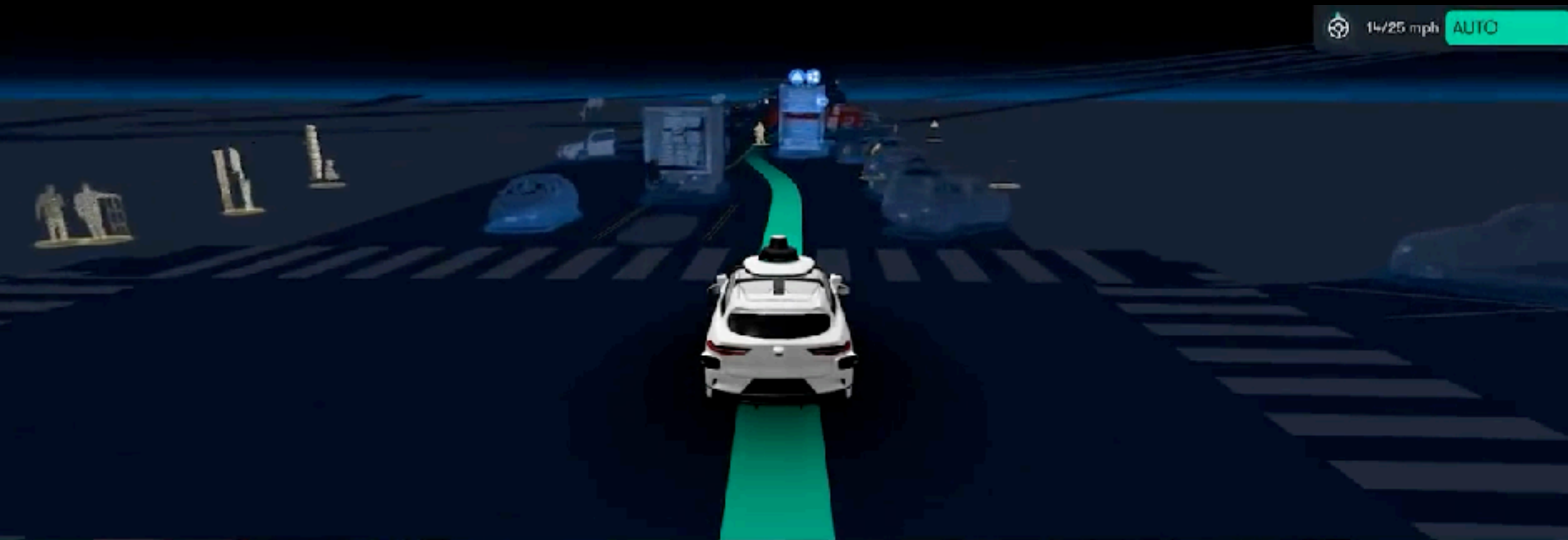


Unitree GO 2



ANYmal

Self-driving continues driverless runs



But ...

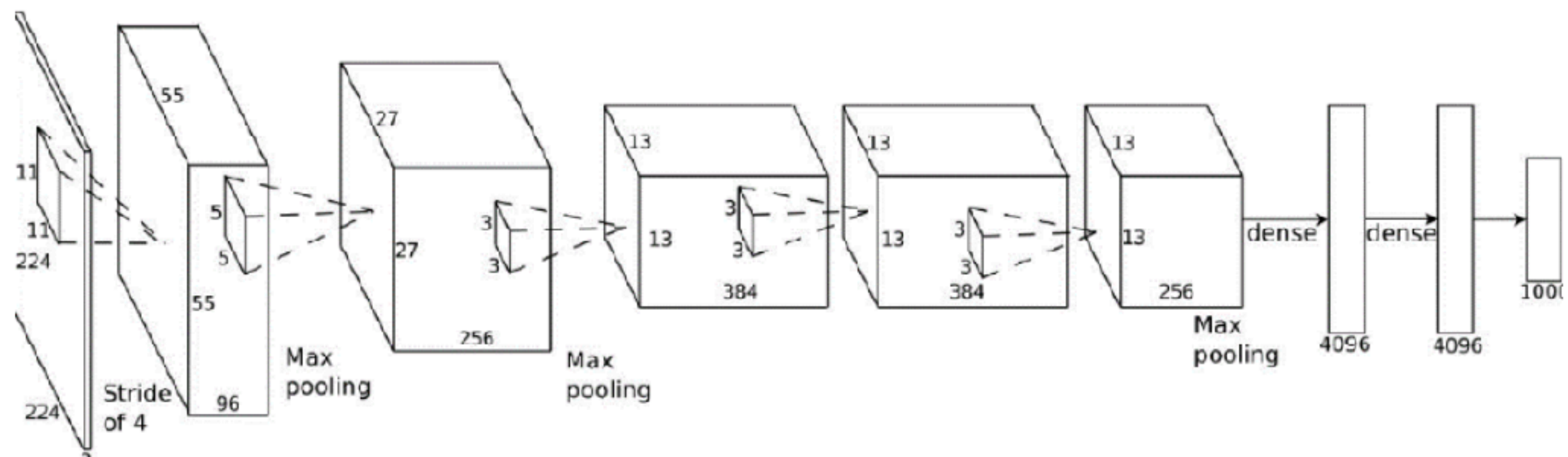
... these robots are not in
millions of homes yet.

Why?

What is so challenging about our homes?



Why can't we throw ML at this problem?

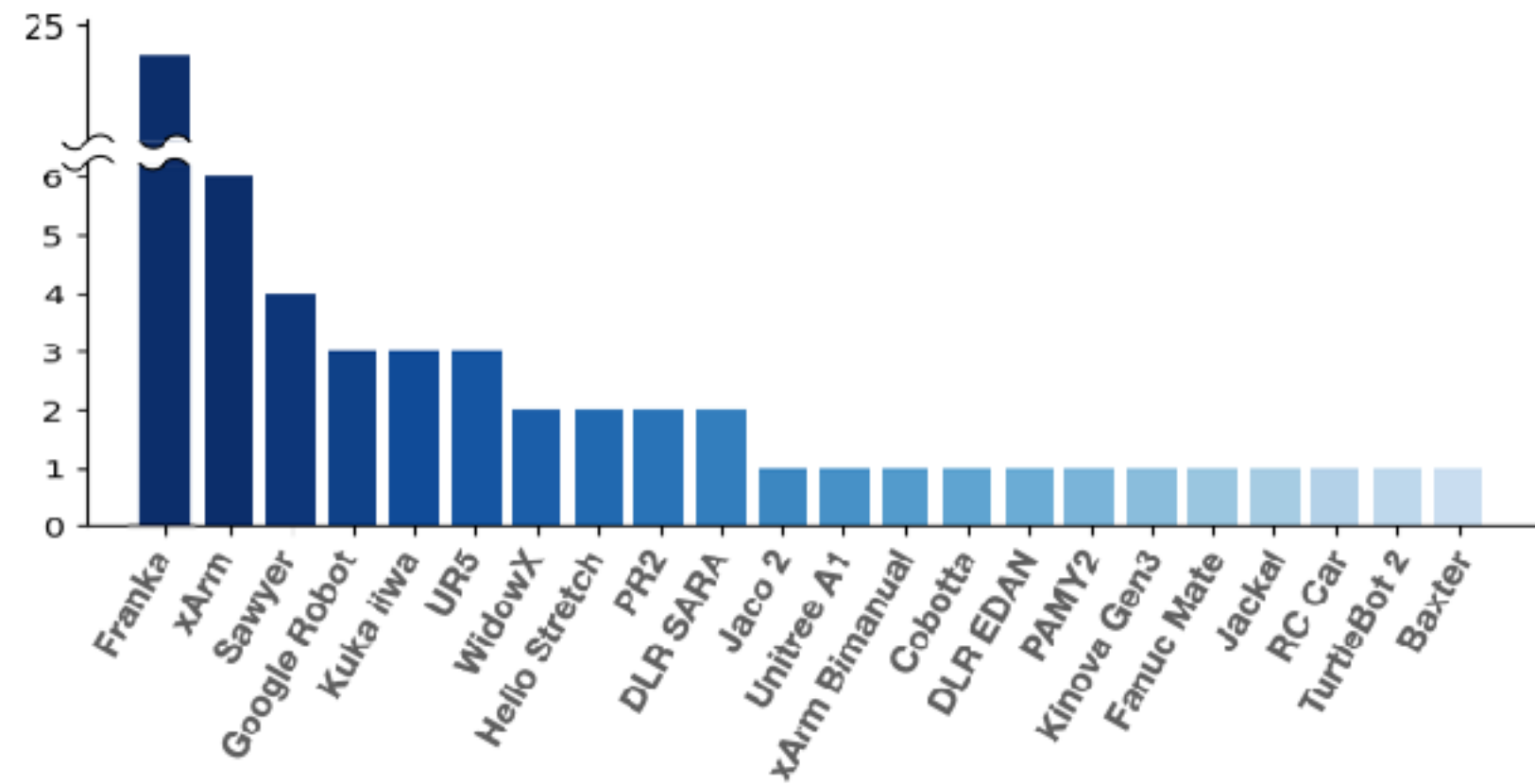


Big Data

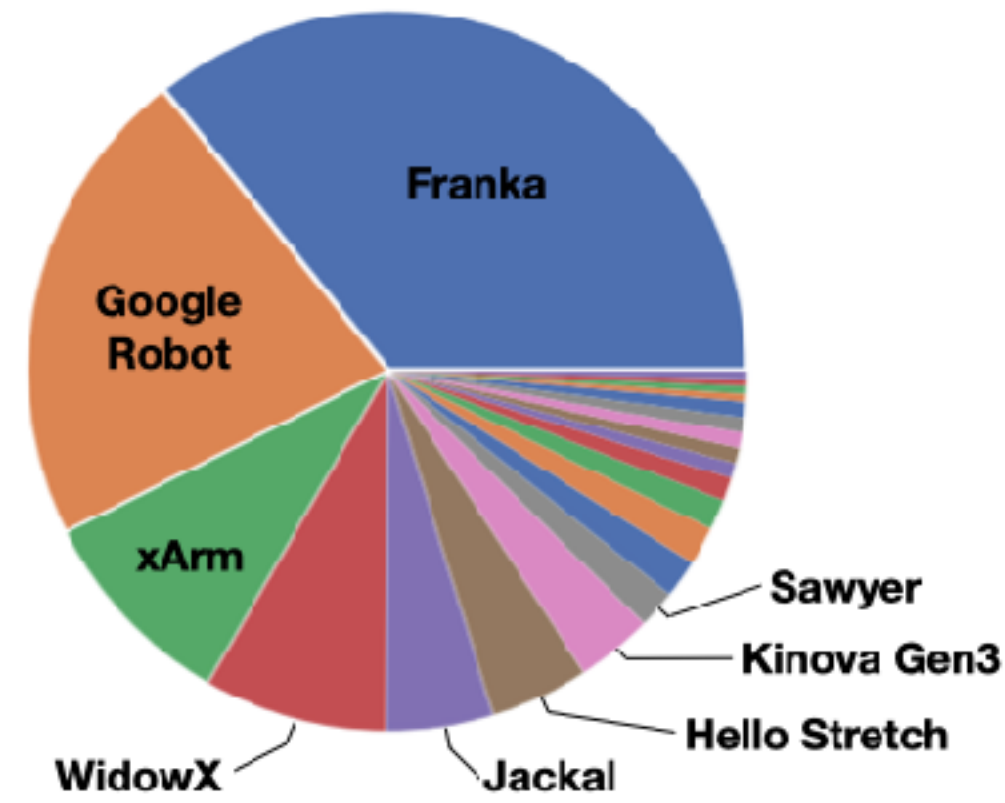
Big Models

Big Data

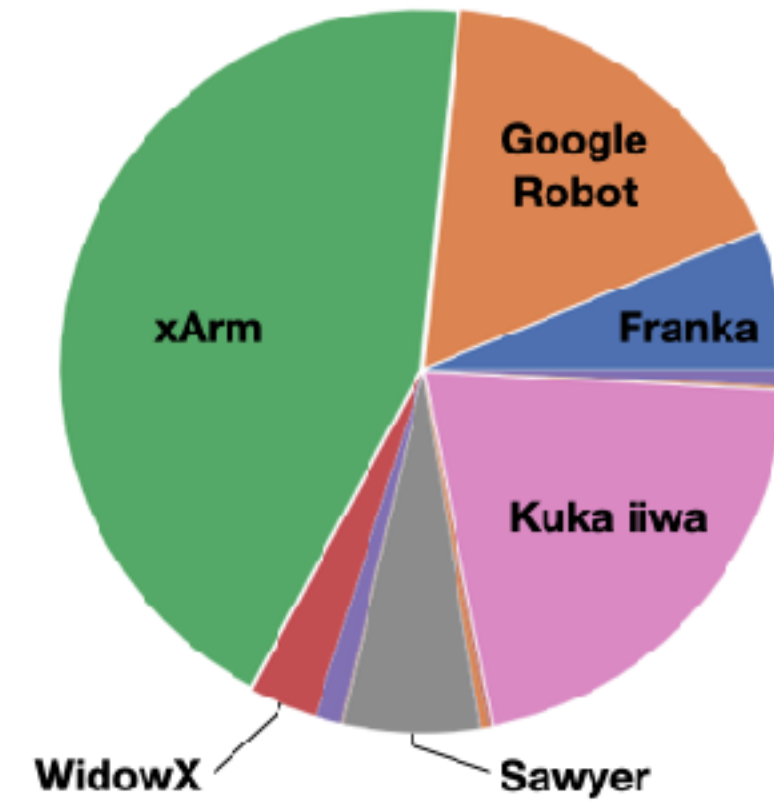
1M trajectories, 22 robots, 21 different institutions



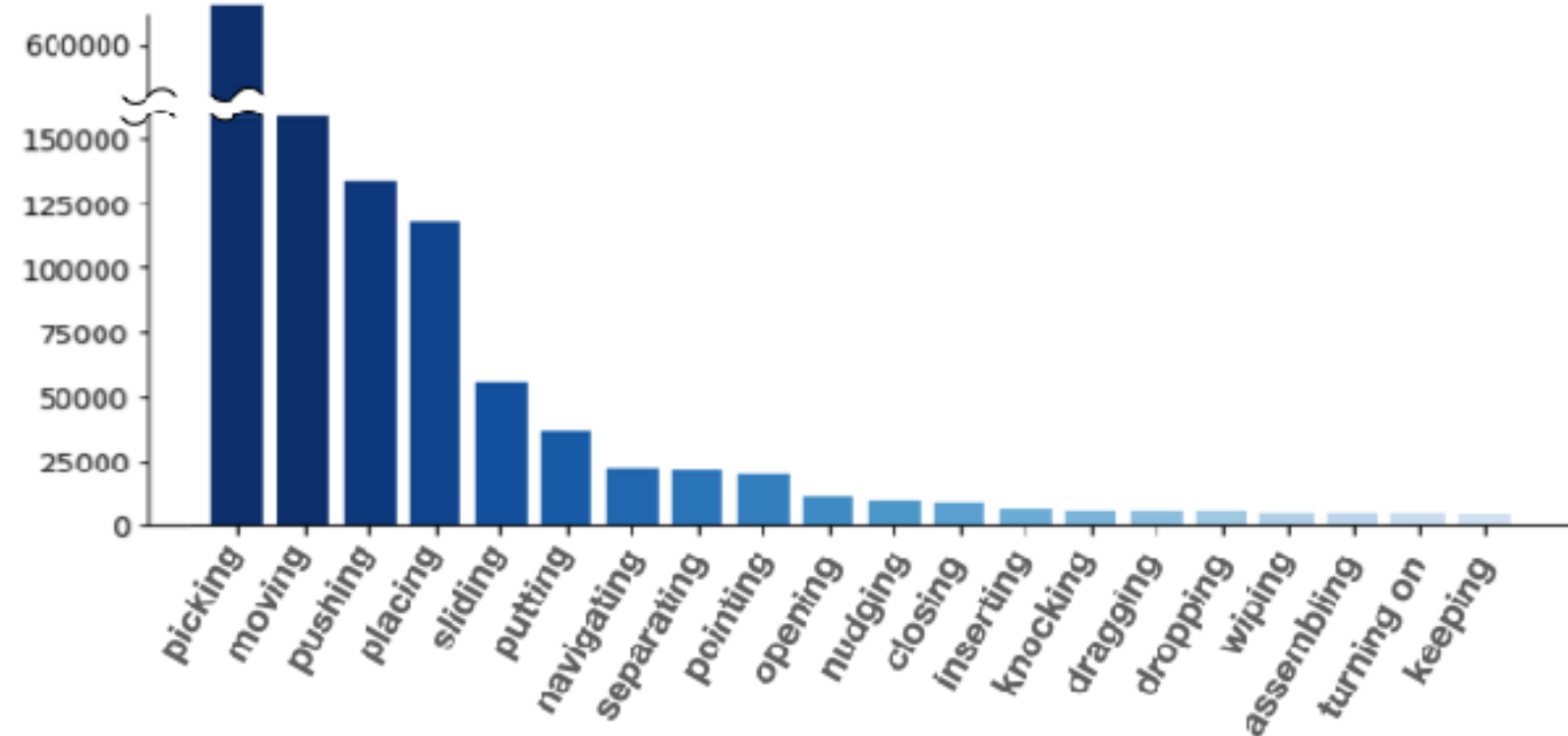
(a) # Datasets per Robot Embodiment



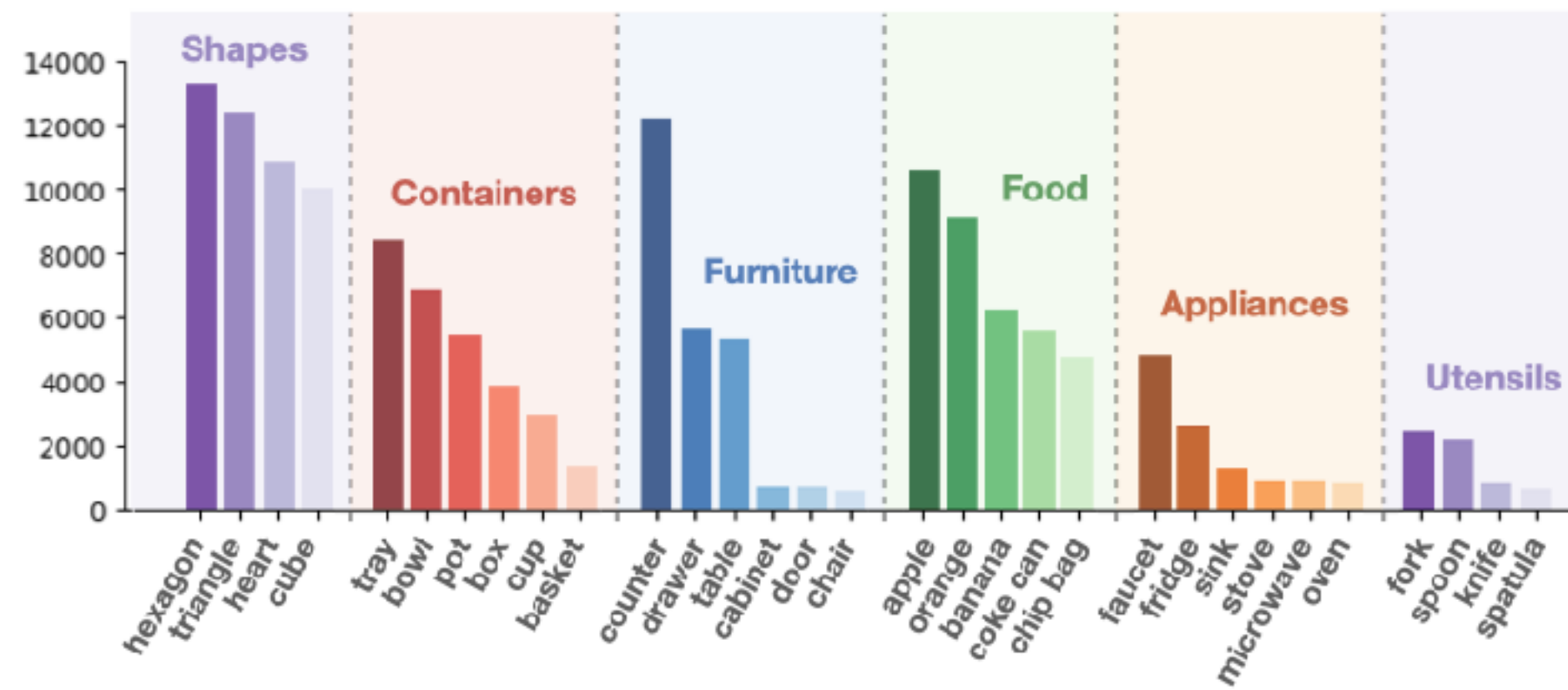
(b) # Scenes per Embodiment



(c) # Trajectories per Embodiment



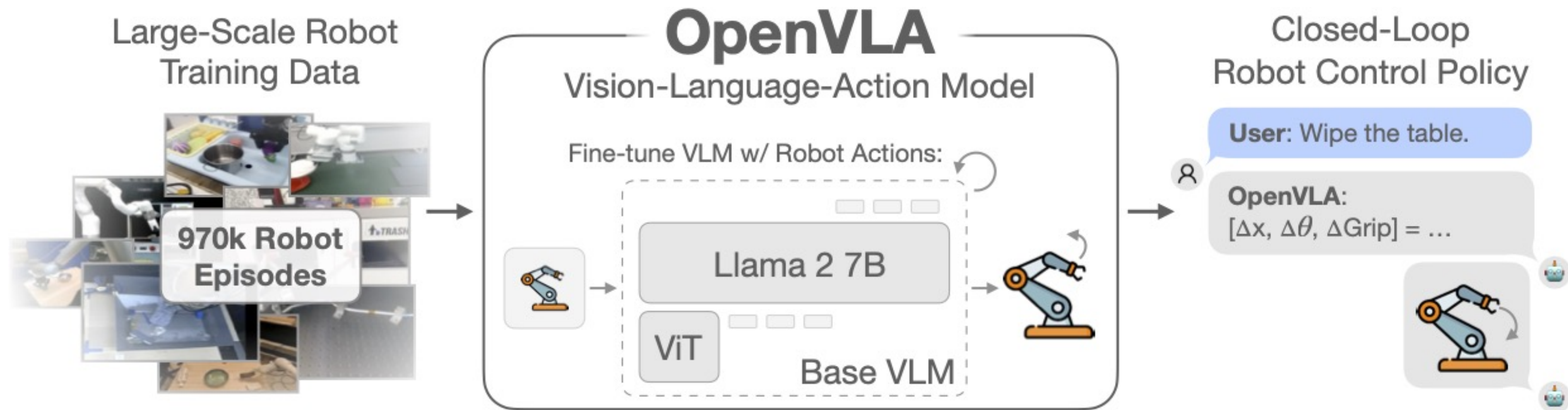
(d) Common Dataset Skills



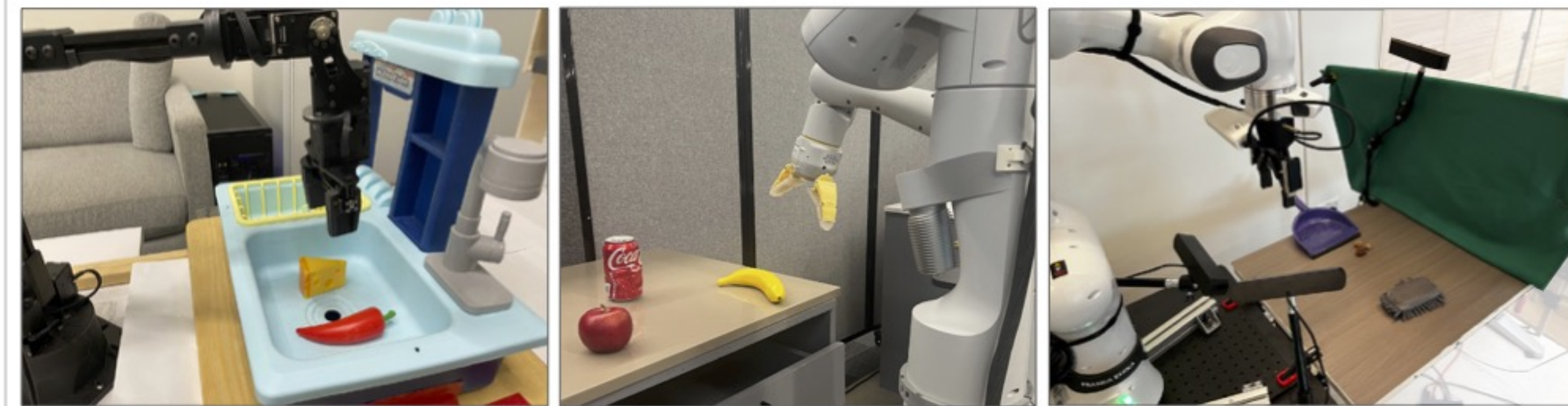
(e) Common Dataset Objects

Open-X Embodiment Dataset




Big Models



Multi-Robot Control & Efficient Fine-Tuning



Fully Open-Source

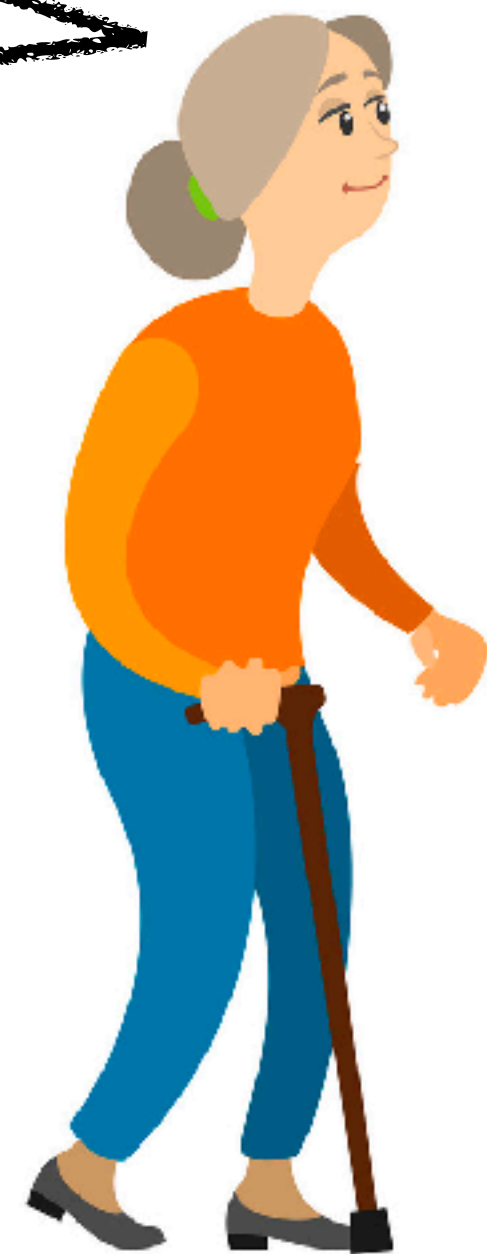
-  Data
-  Weights
-  Code

<https://openvla.github.io/>

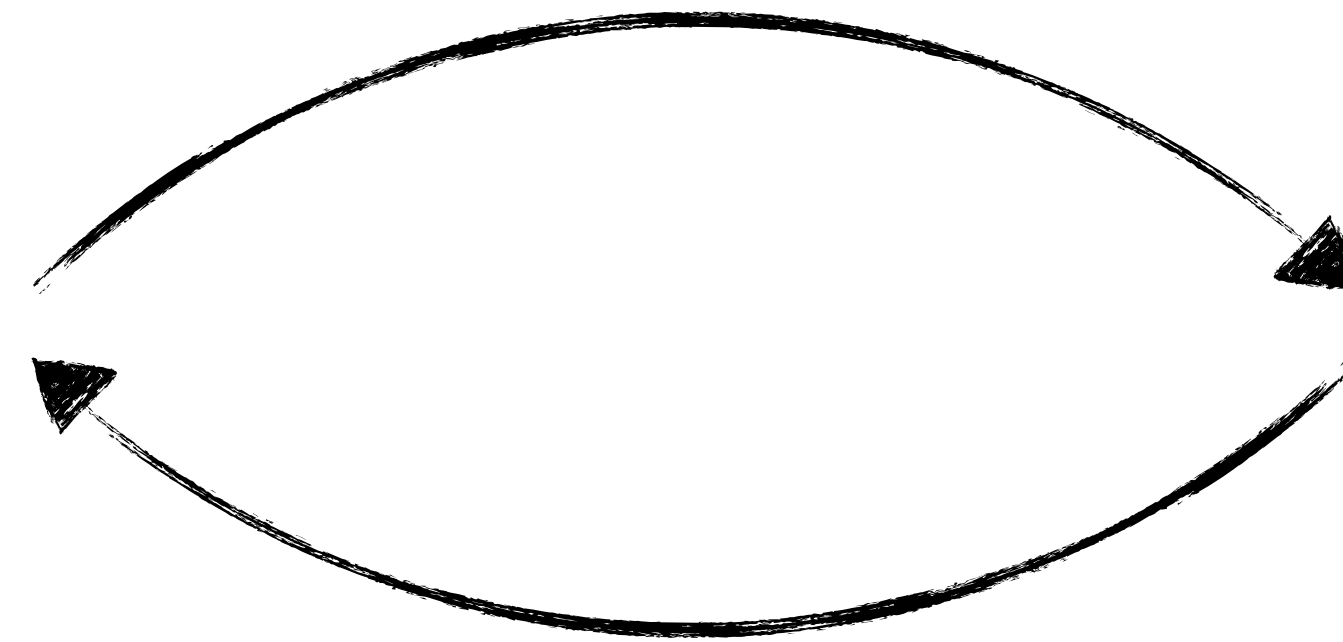
Activity!



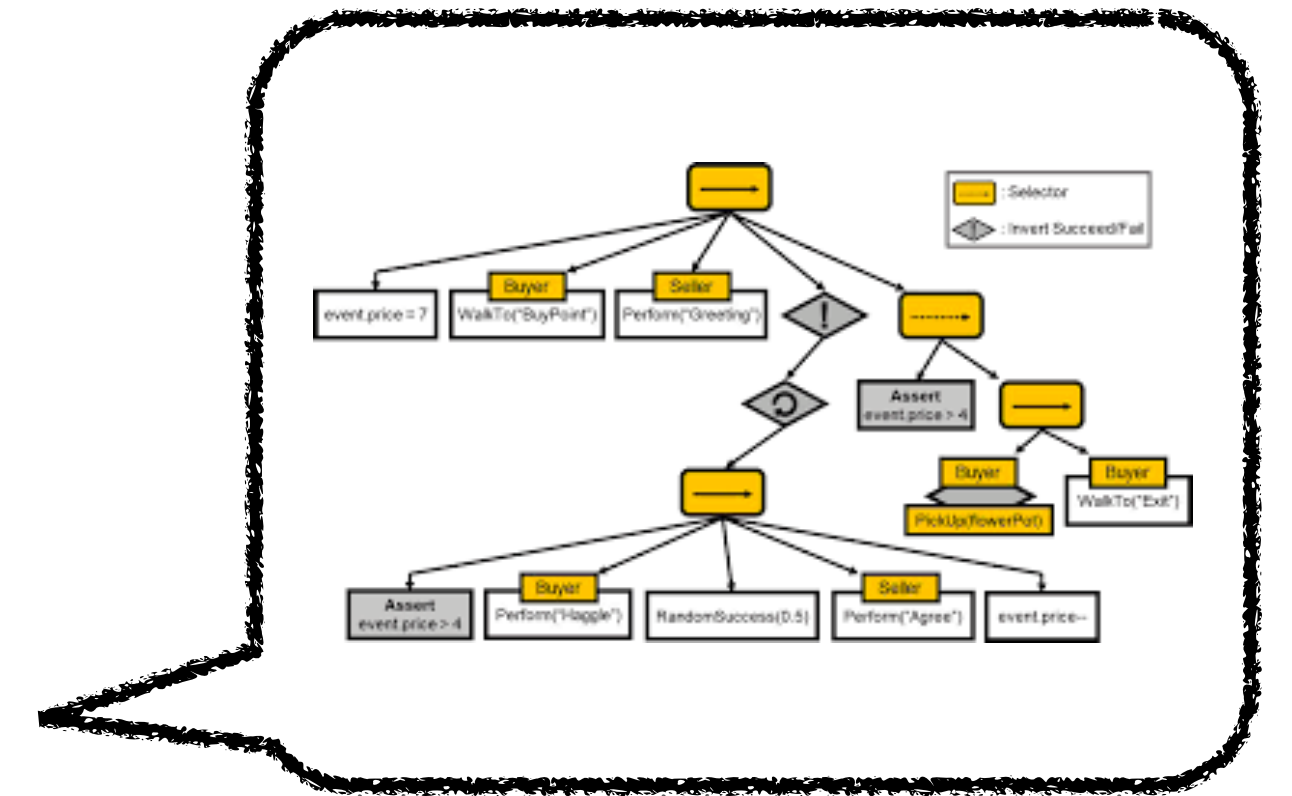
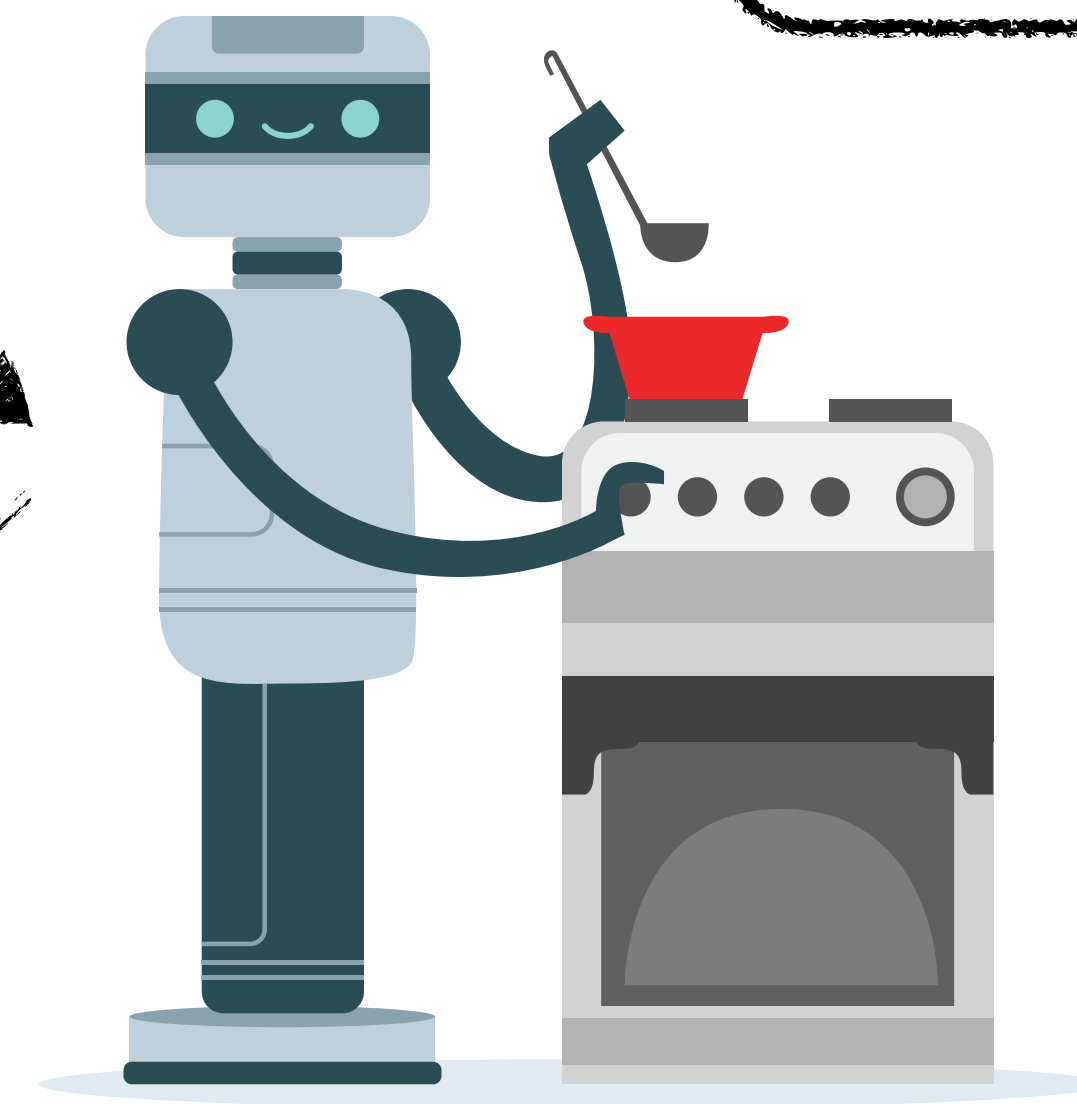
Train home robot apprentice to help grandma!



Demonstrations,
Language



Interactive
feedback



Home Robots in our lab!

I am preparing vegetables for the soup. Can you pour some salt after stirring?

R1 Subtasks

Stir

Pour salt

R1 Current Code

```
pick_up("ladle")  
move_arm_to("pot")  
→ stir("pot")
```

R2 Subtasks

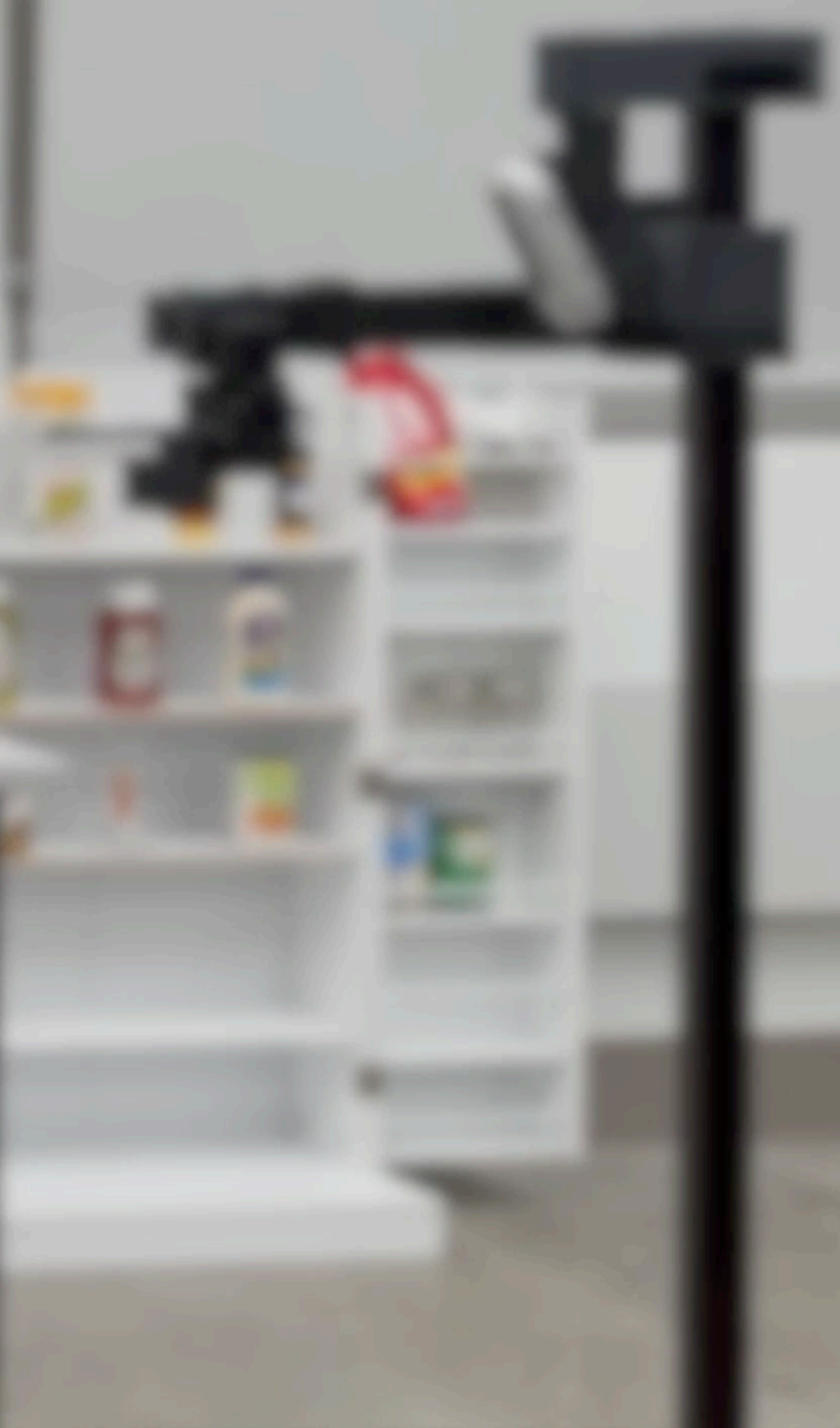
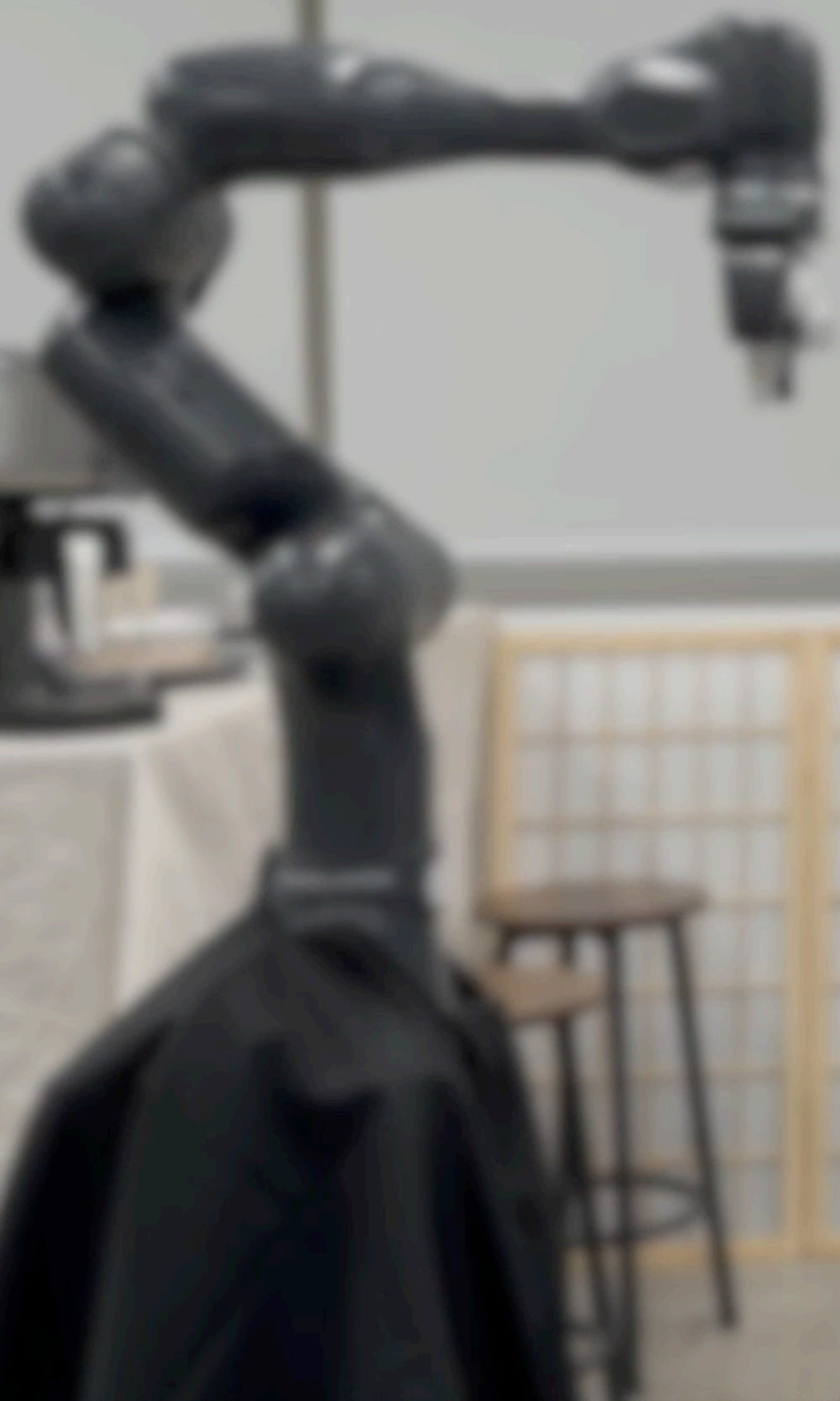
Fetch Salt

R2 Current Code

```
go_to("pantry")  
pick_up("salt")  
go_to("table")  
→ place("table")
```

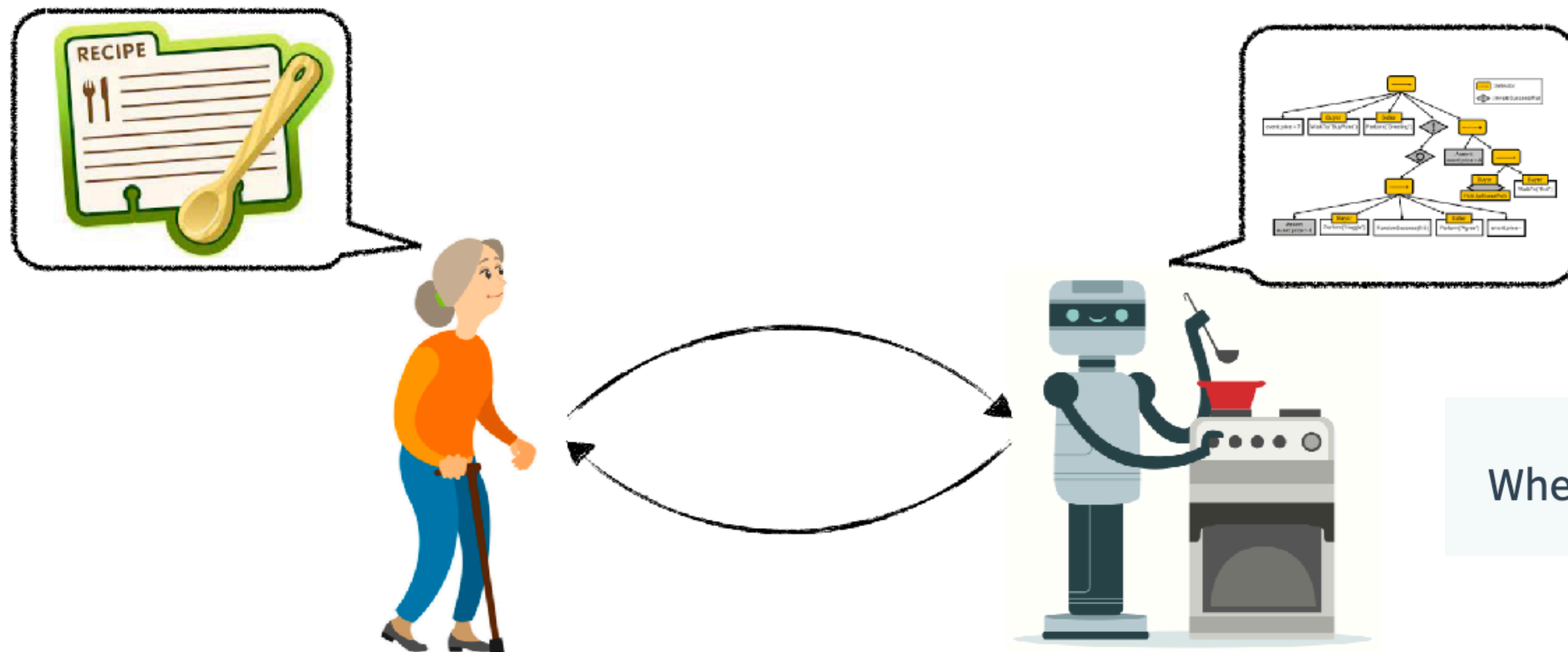


Hey MOSAIC! It's so cold today! Can you suggest some soup recipe?



Question

What is main challenge in apply machine learning to home robots?



When poll is active respond at PollEv.com/sc2582

What is special about robot learning?

$$\min_{\theta} \mathbb{E}_{x,y} \ell(y, \theta(x))$$

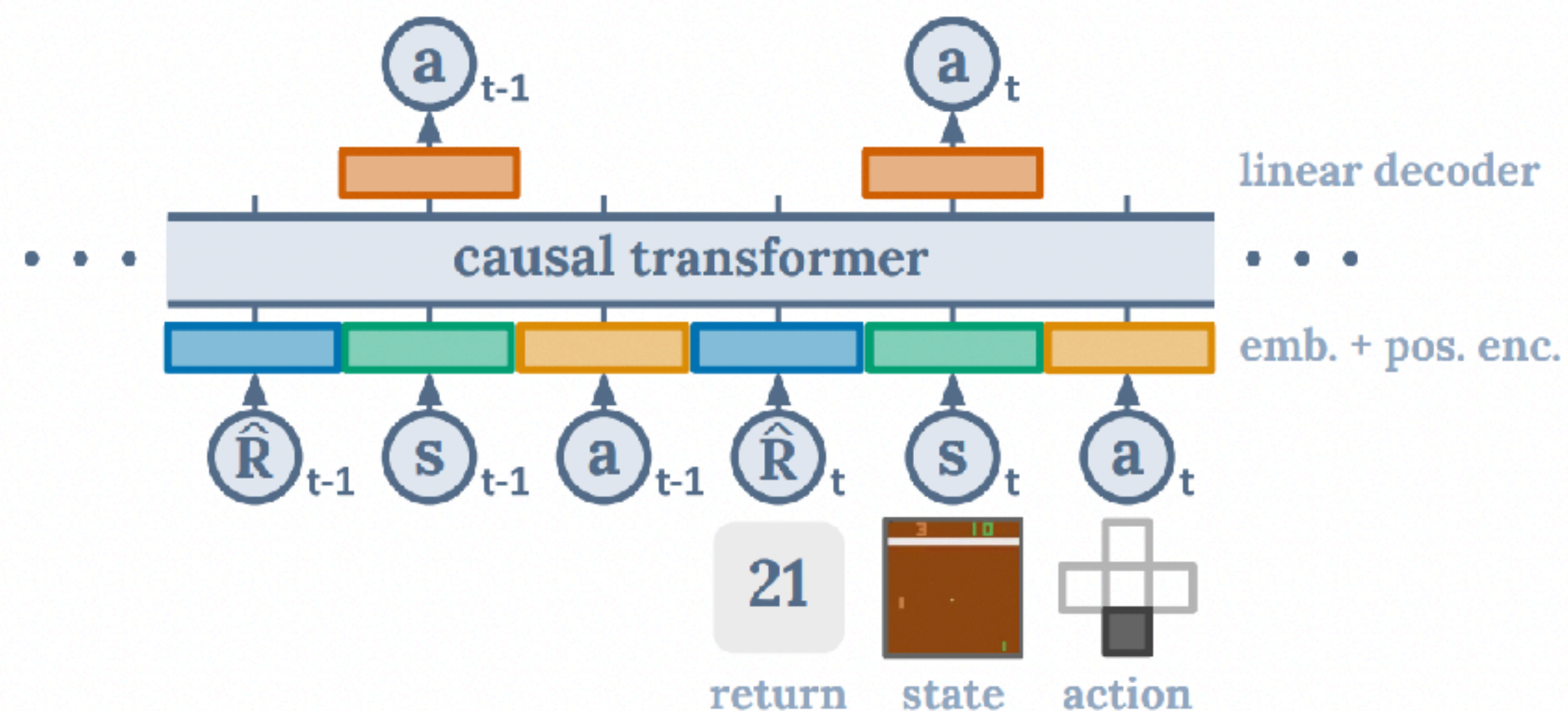
x is a sequence of inputs, y is a sequence of outputs, θ is a model

What is special about robot learning?

$$\min_{\theta} \mathbb{E}_{x,y} \ell(y, \theta(x))$$

x is a sequence of inputs, y is a sequence of outputs, θ is a **model**

Transformers are
pretty standard choice
for the **model**



What is special about robot learning?

$$\min_{\theta} \mathbb{E}_{x,y} \ell(y, \theta(x))$$

x is a sequence of inputs, y is a sequence of outputs, θ is a model

Problem 1: How do we gather the right **data**?

What is special about robot learning?

$$\min_{\theta} \mathbb{E}_{x,y} \ell(y, \theta(x))$$

x is a sequence of inputs, y is a sequence of outputs, θ is a model

Problem 2: How do we choose the right **loss**?

WHY this course?



Formulate as a Markov Decision Problem (MDP)



Solve MDPs using an all-purpose toolkit
(Imitation/Reinforcement learning, Model based/free)



Deploy learners in real-world
(Safety, distribution shift, value alignment)

Take *any* robot application

Build robots that can *learn from humans!*



“Sanjiban”

He / Him

Office hours:
Tues 11:30 – 1:30pm
Gates 413B

Undergrad



PhD



Research Engineer

PostDoc

We are PoRTaL

(People and Robots, Teaching and Learning)



PORTAL

<https://portal.cs.cornell.edu/>

Belonging



Some news!

We are expecting a baby this fall (October 22nd!)

We are super excited (and nervous!)

Some news!

We are expecting a baby this fall (October 22nd!)

Since I don't get an official parental leave, I will need YOUR help

I will teach actively up until the due date.

After, I will have my brilliant colleagues come in for guest lectures.

These are amazing researchers from different ML fields.

I need YOUR help to make these classes engaging and give a good impression.

Let's get started!

Self-driving



A brief history of self-driving

One of the first self-driving car drove from Pittsburgh to Sandiego with 2800 miles of autonomy. Which year did this happen?

1995

CMU Navlab Minivan



Pittsburgh -> San Diego,
2800 miles of autonomy

(... but really only lane-keeping)

2005

Stanford's Stanley



Wins the first
DARPA Grand Challenge
beating both of CMU's cars

*(Tested full, driverless
autonomy ... but all in a desert)*

2007

CMU's BOSS

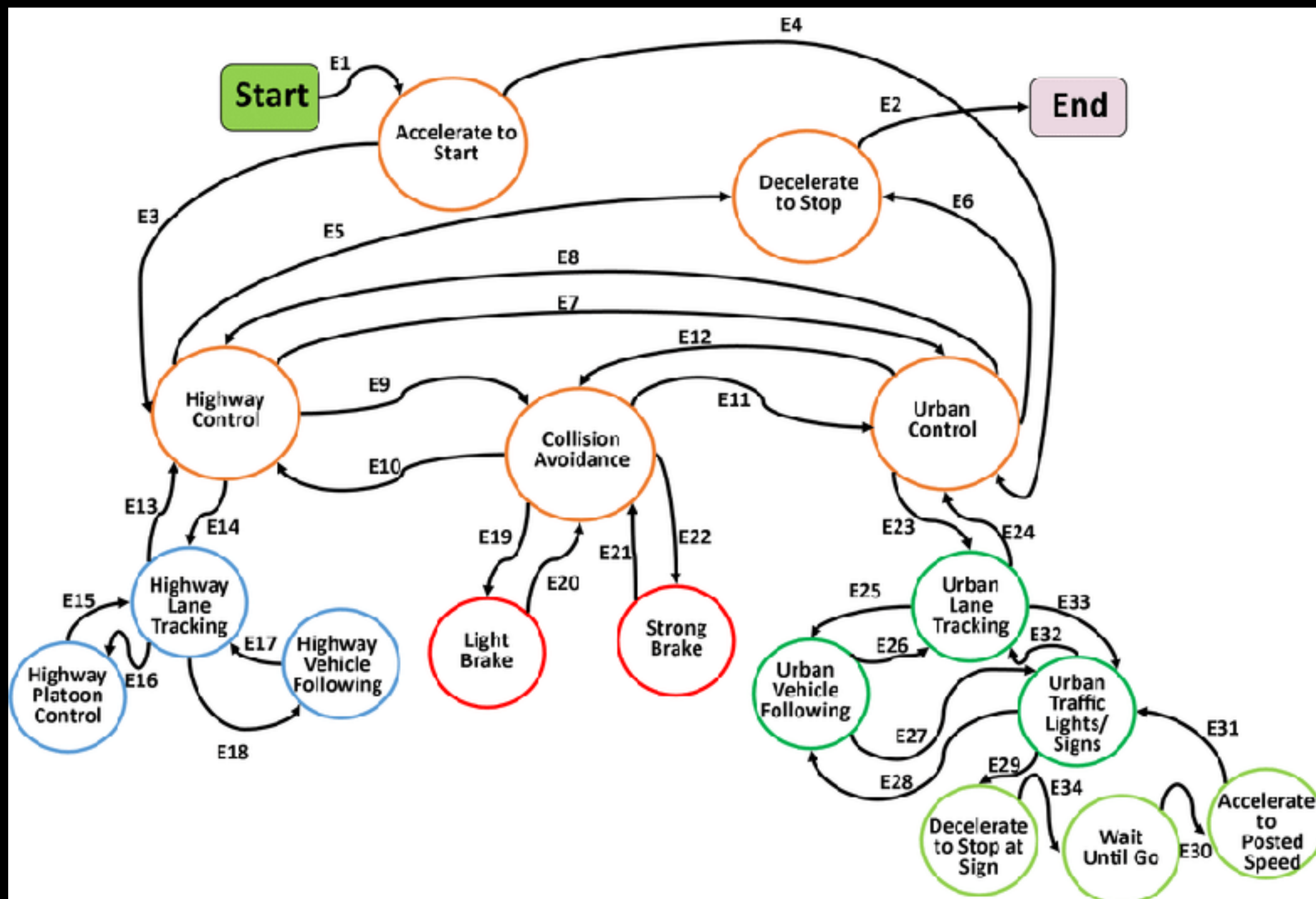


Wins the DARPA
Urban Challenge

*(Urban setting, interaction with
other cars, traffic rules)*

Self-driving 1.0

Hand-engineered rules of driving



Limited use of machine learning

Software that fundamentally could not scale over time

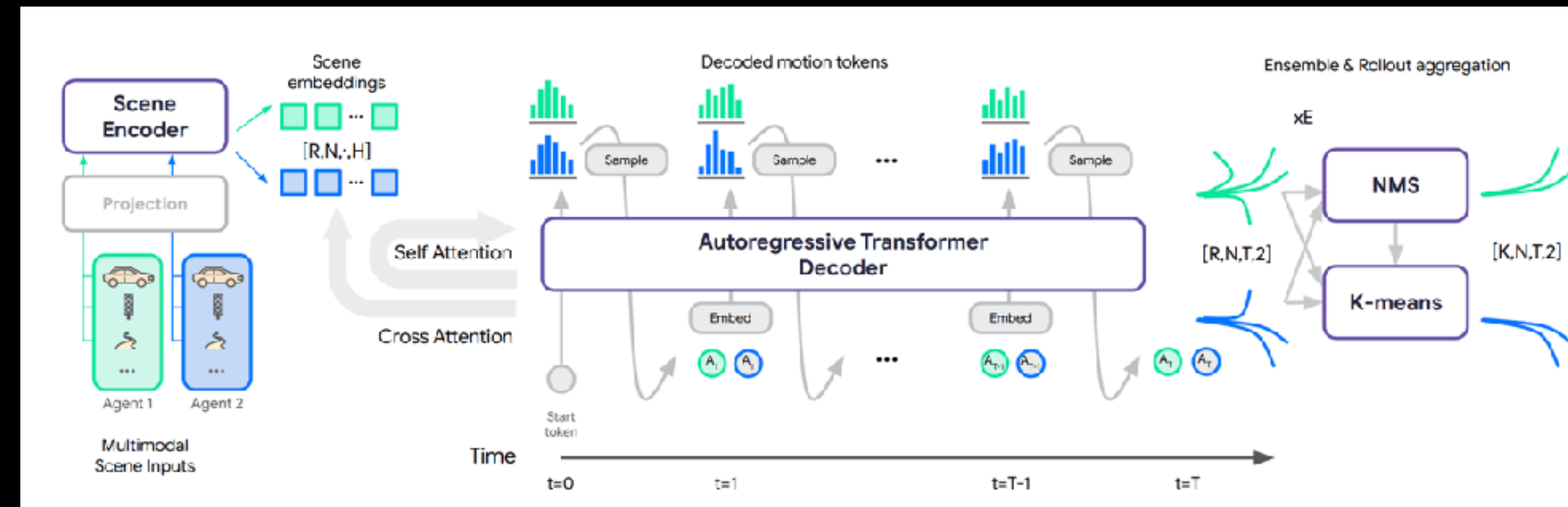
Self-driving 2.0

Software 2.0



Andrej Karpathy · Follow

9 min read · Nov 11, 2017



Design software from the ground-up to be learnable

Scalable pipelines that turn data into tests

Learning not just for perception, but also decision making

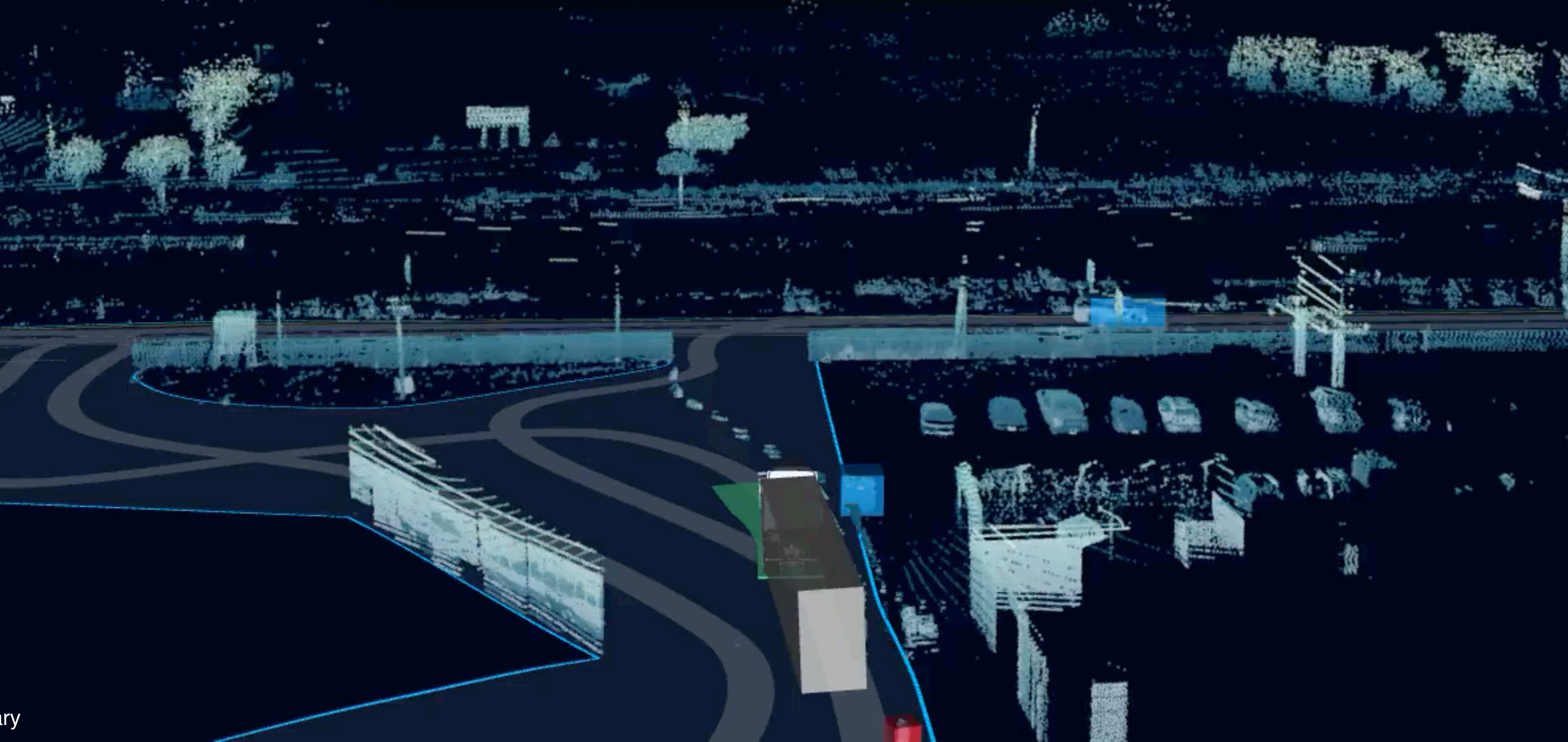
RIGHT TURN
-10 FT

READY



0
MPH

SPEED
LIMIT
5



Aurora

A grim state of affairs

Tesla Recalls Autopilot Software in 2 Million Vehicles

Federal regulators pressed the automaker to make updates to ensure drivers are paying attention while using Autopilot, a system that can steer, accelerate and brake on its own.

Cruise Stops All Driverless Taxi Operations in the United States

The move comes just two days after California regulators told the company to take its autonomously driven cars off the road.

While machine learning is very powerful, getting it to do the right thing in all possible situations has been hard

Even when it makes a mistake, it's hard to know why

The BILLION dollar question

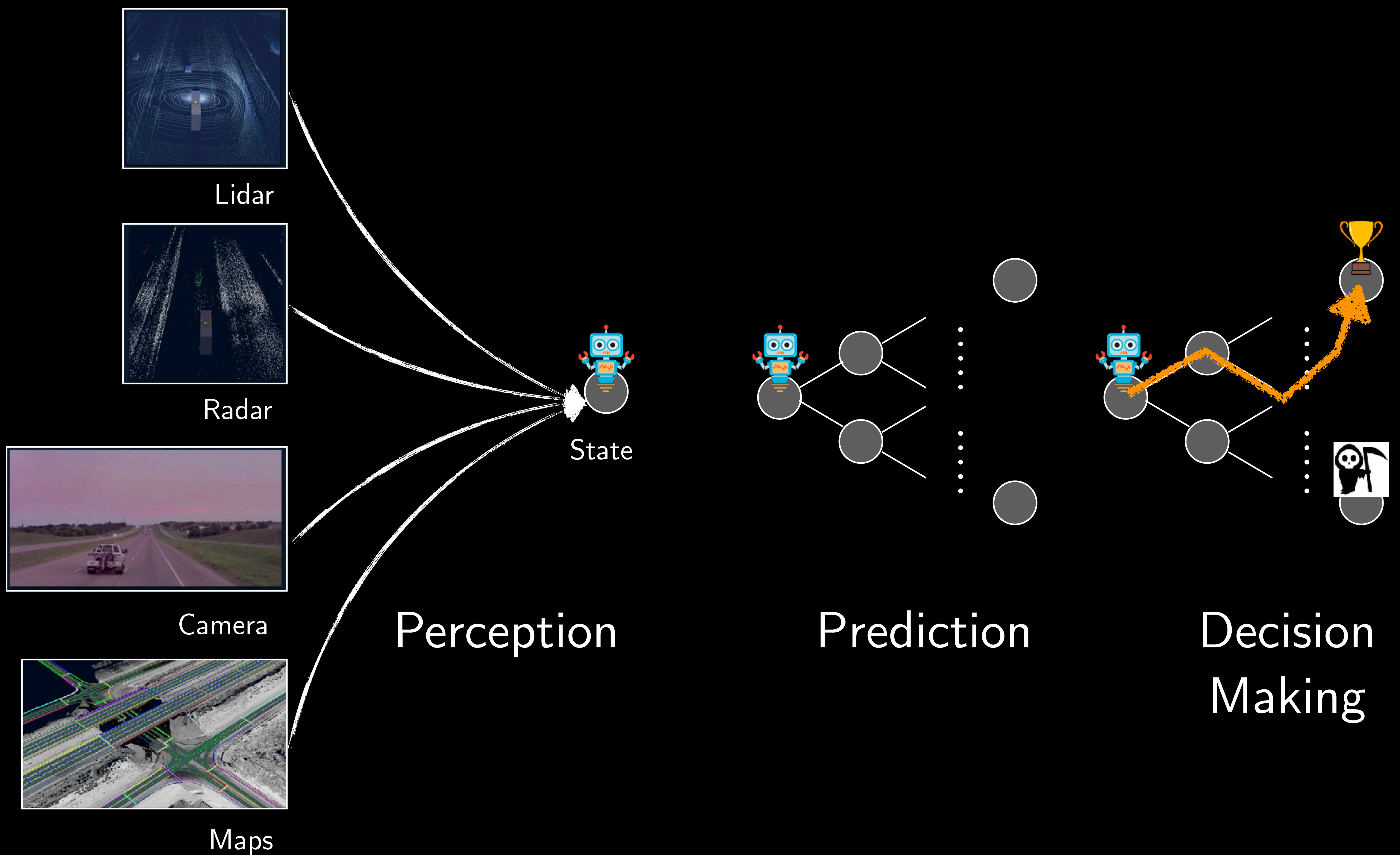
Is this a scaling issue? Should we 10x our data, have bigger models?

Do we need new ways to teach our self-driving cars?
(Think of language models before and after RLHF)

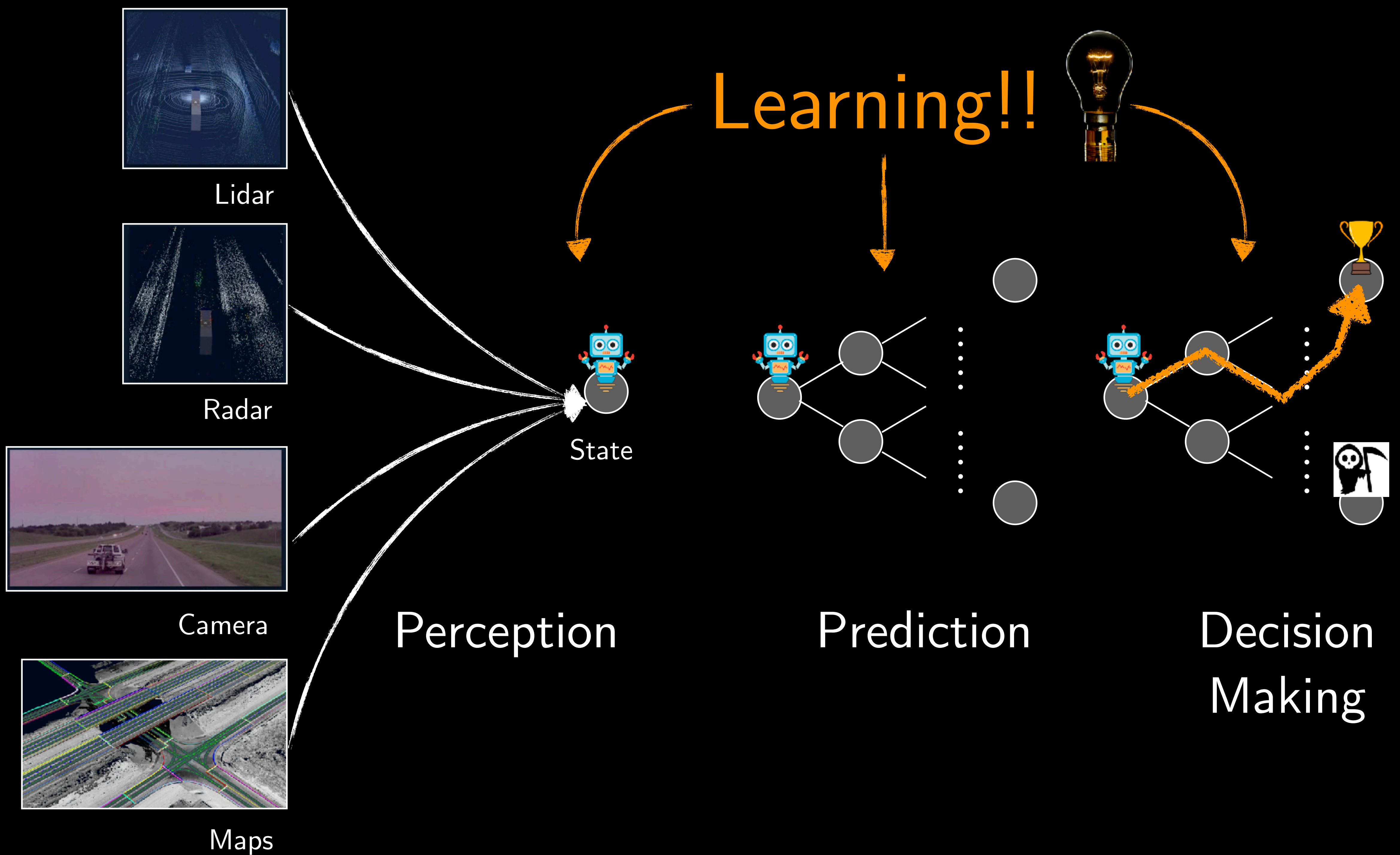
Do we need more powerful simulators and have self-driving cars evolve via natural selection?

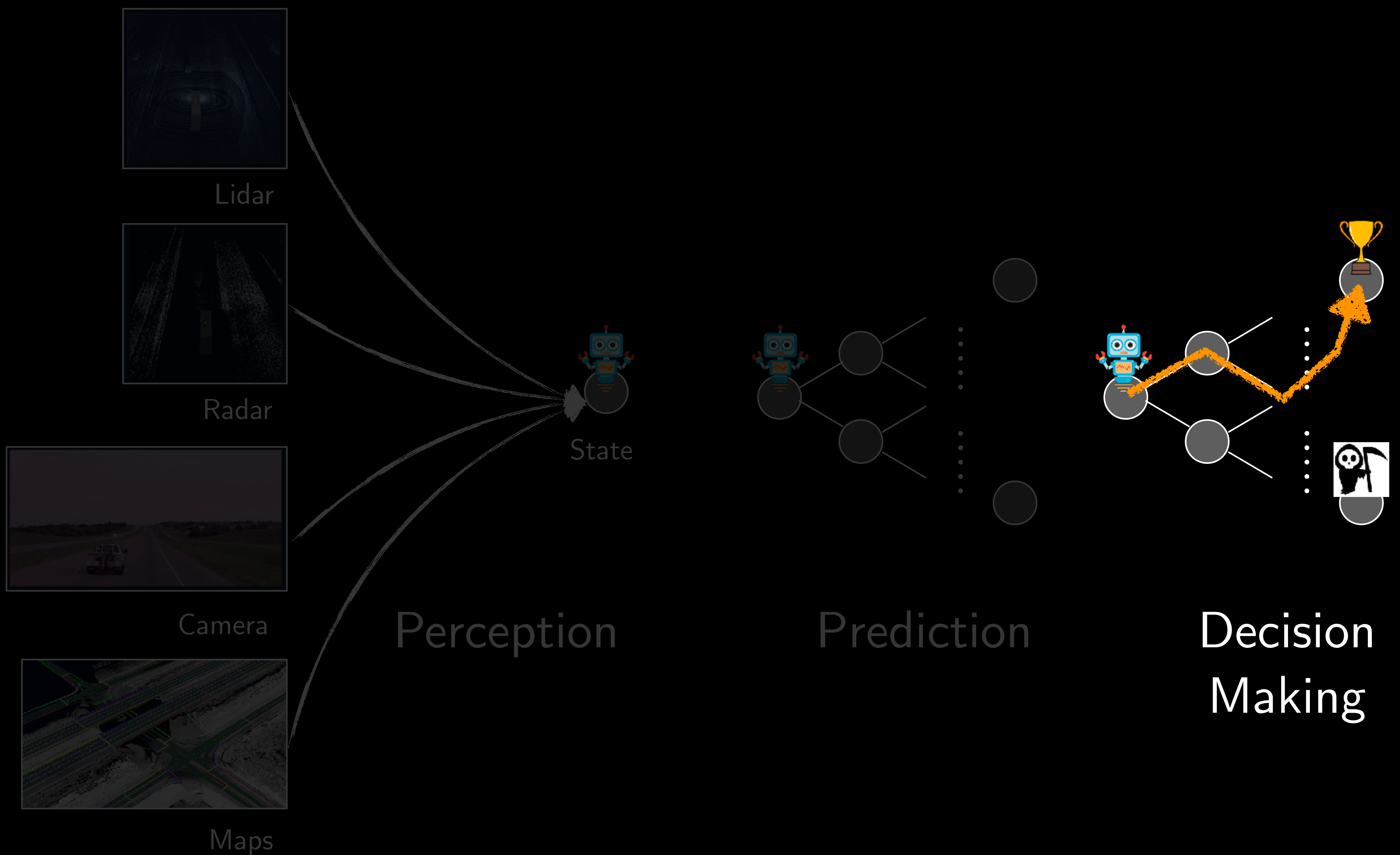
Do we need new policies for safety and interpretability?

Let's dive a bit deeper

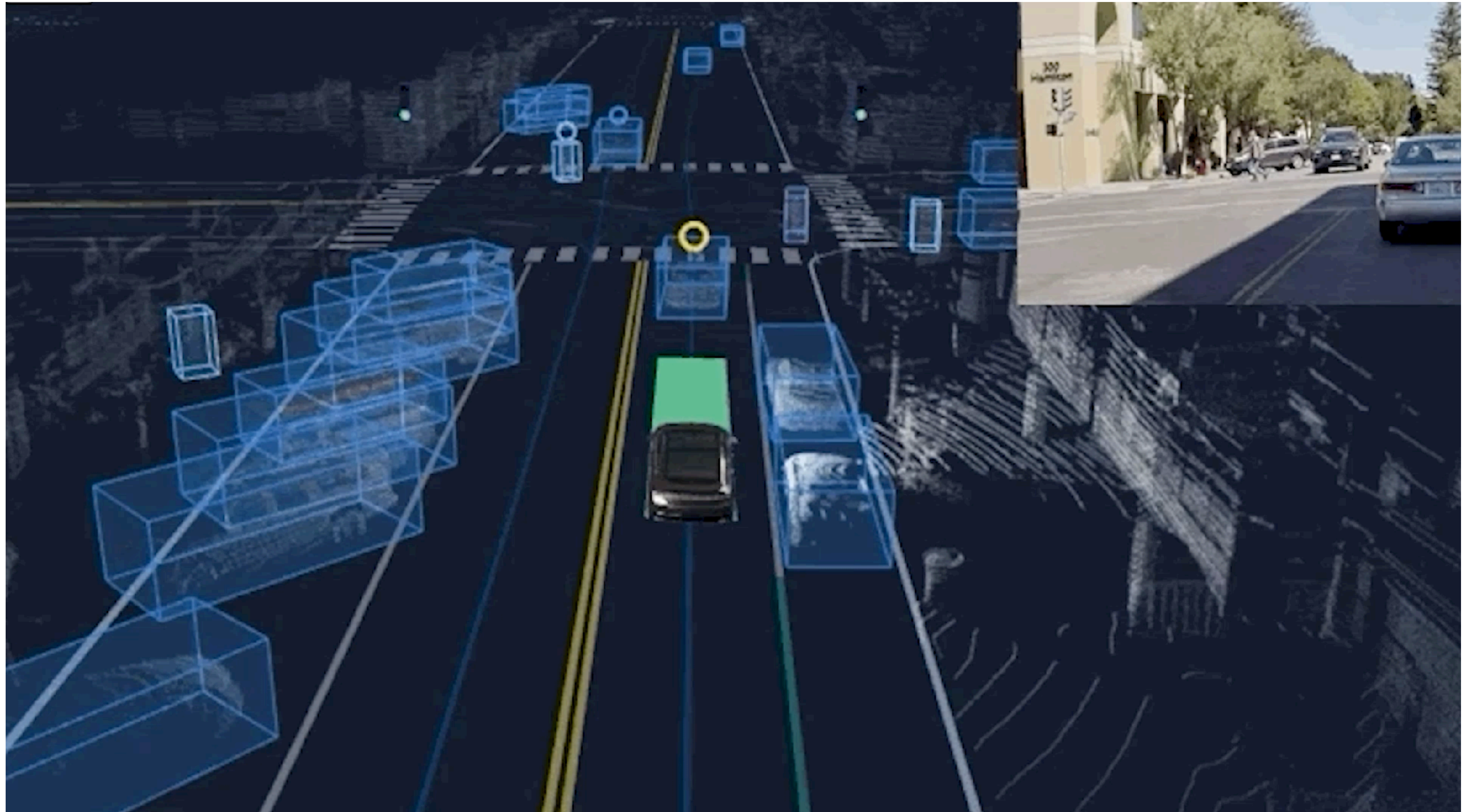


Learning!!





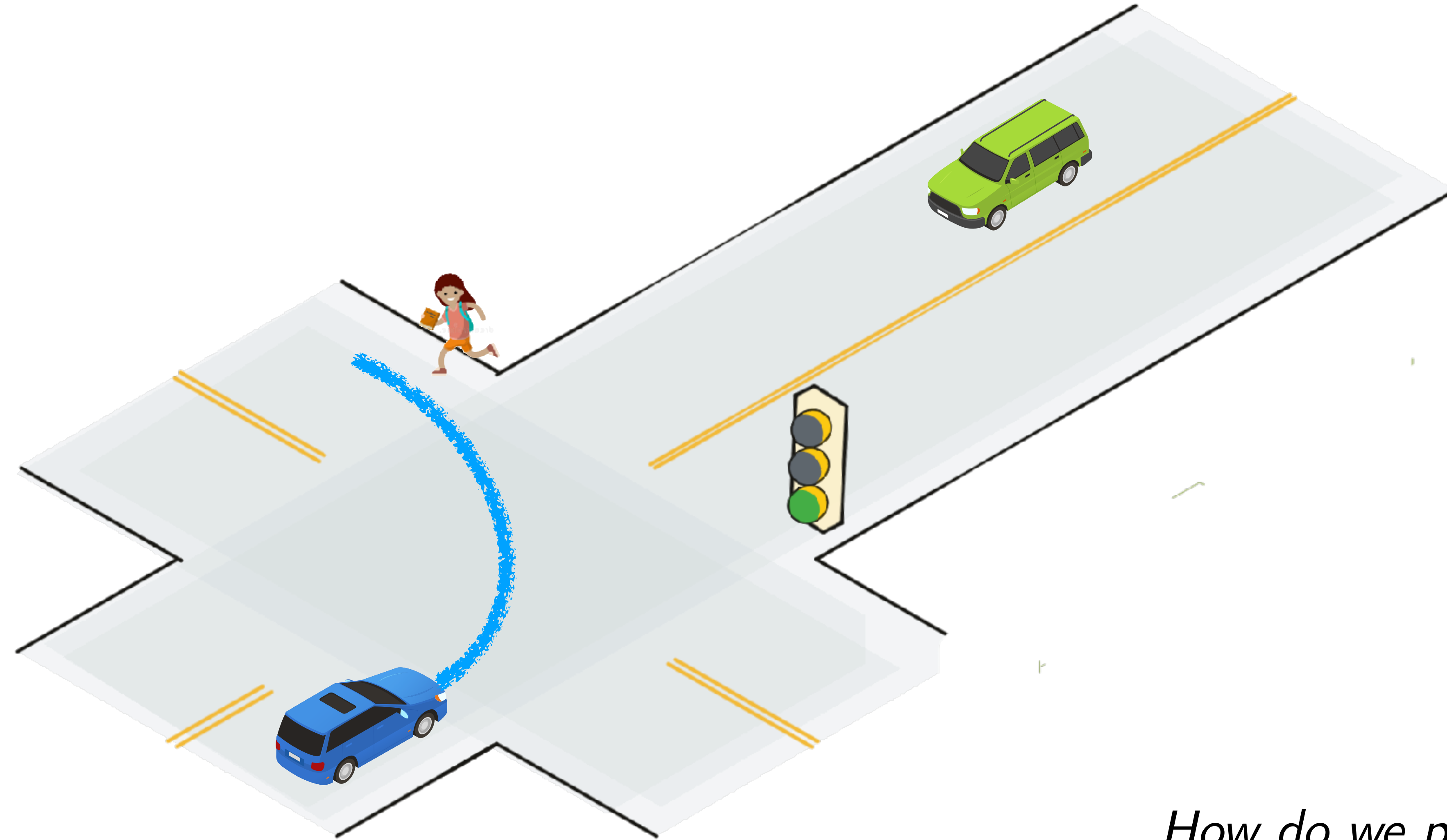
Activity: What is “good” behavior in a left turn?



Activity!



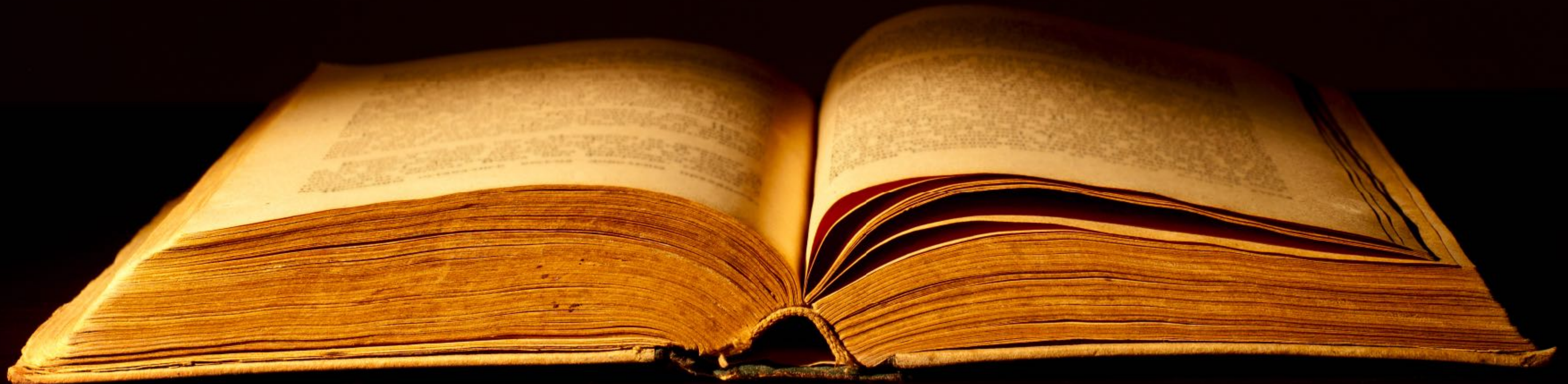
Activity: What is “good” behavior in a left turn?

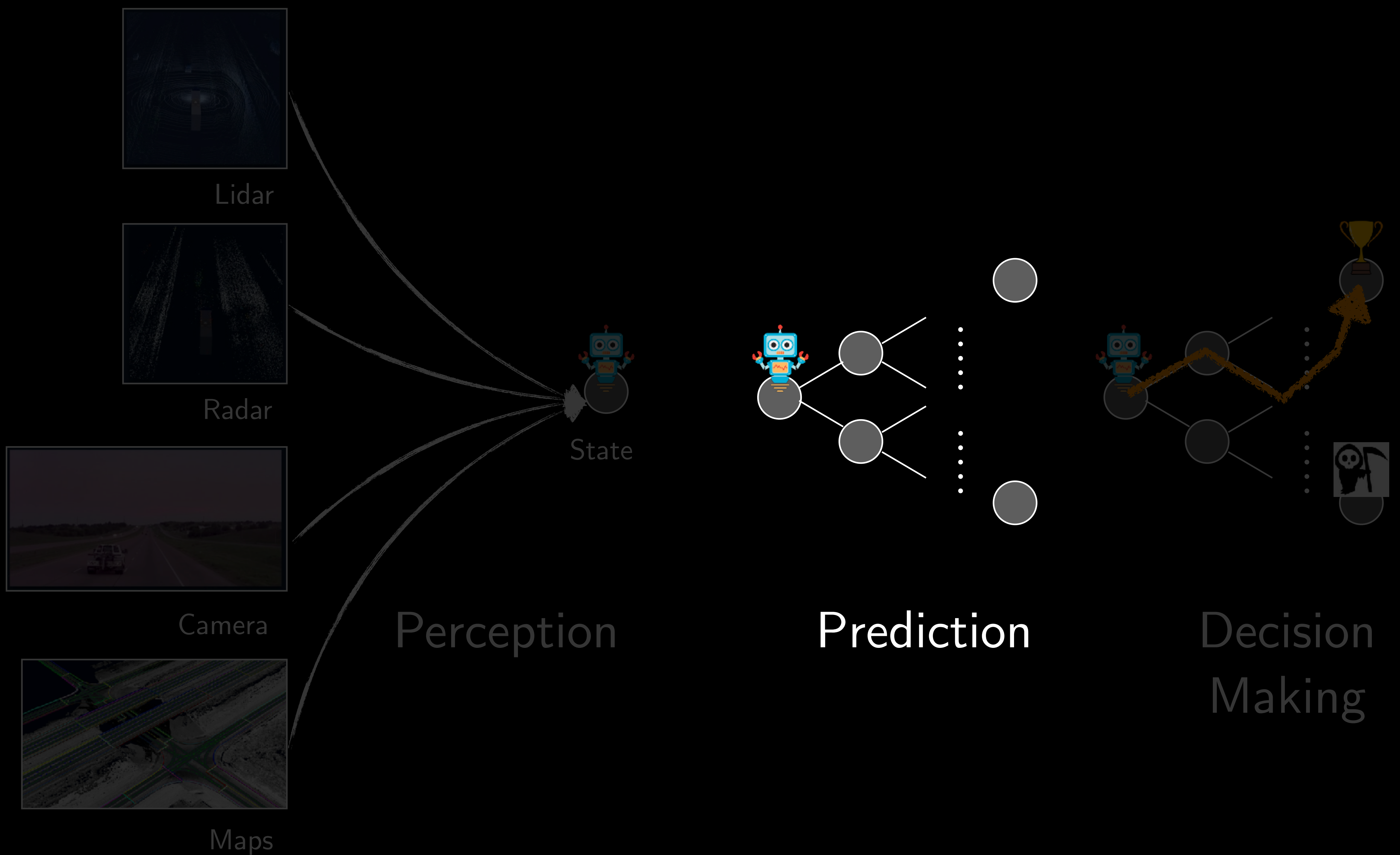


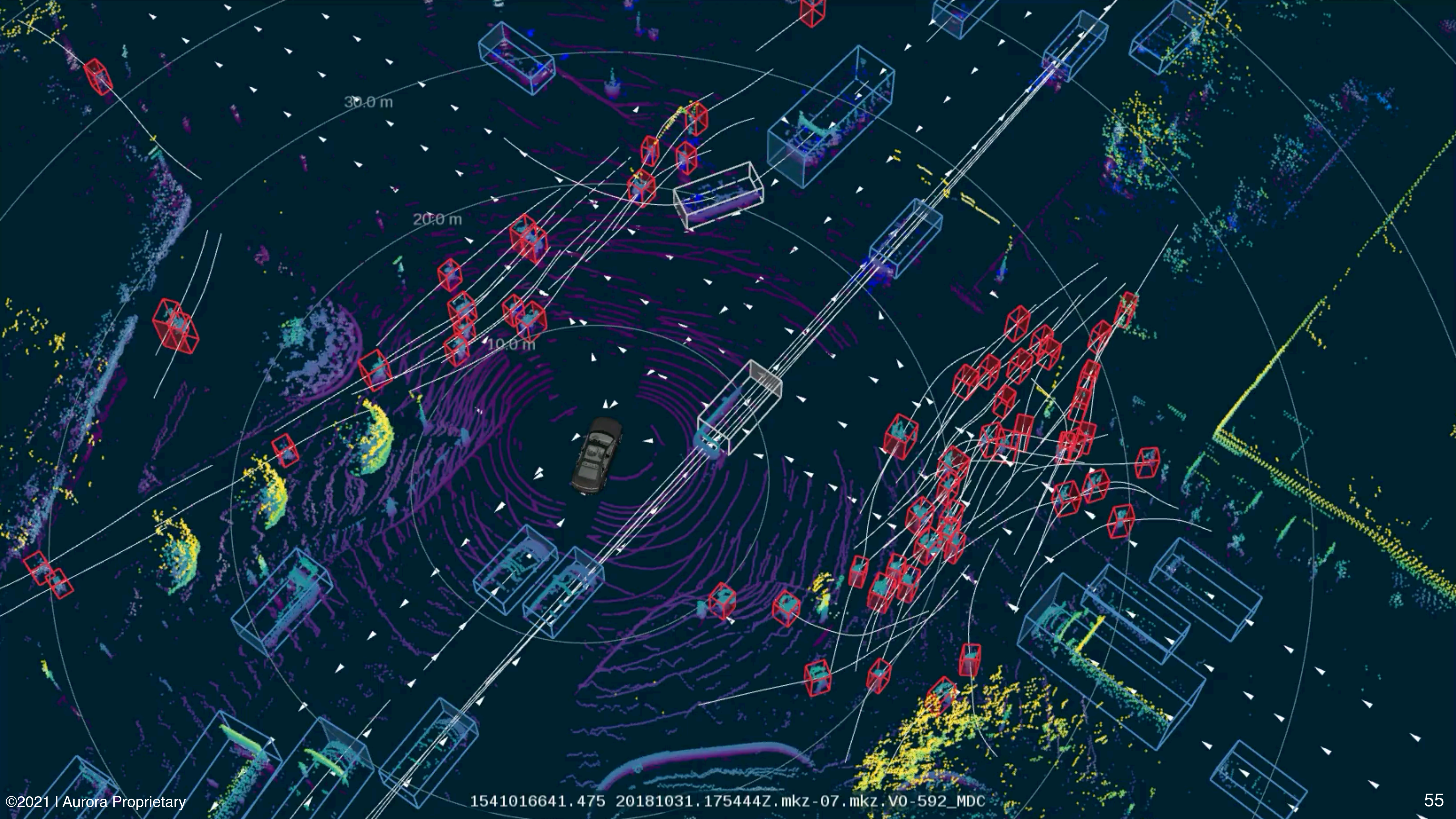
How do we program in these values?

Lesson #1

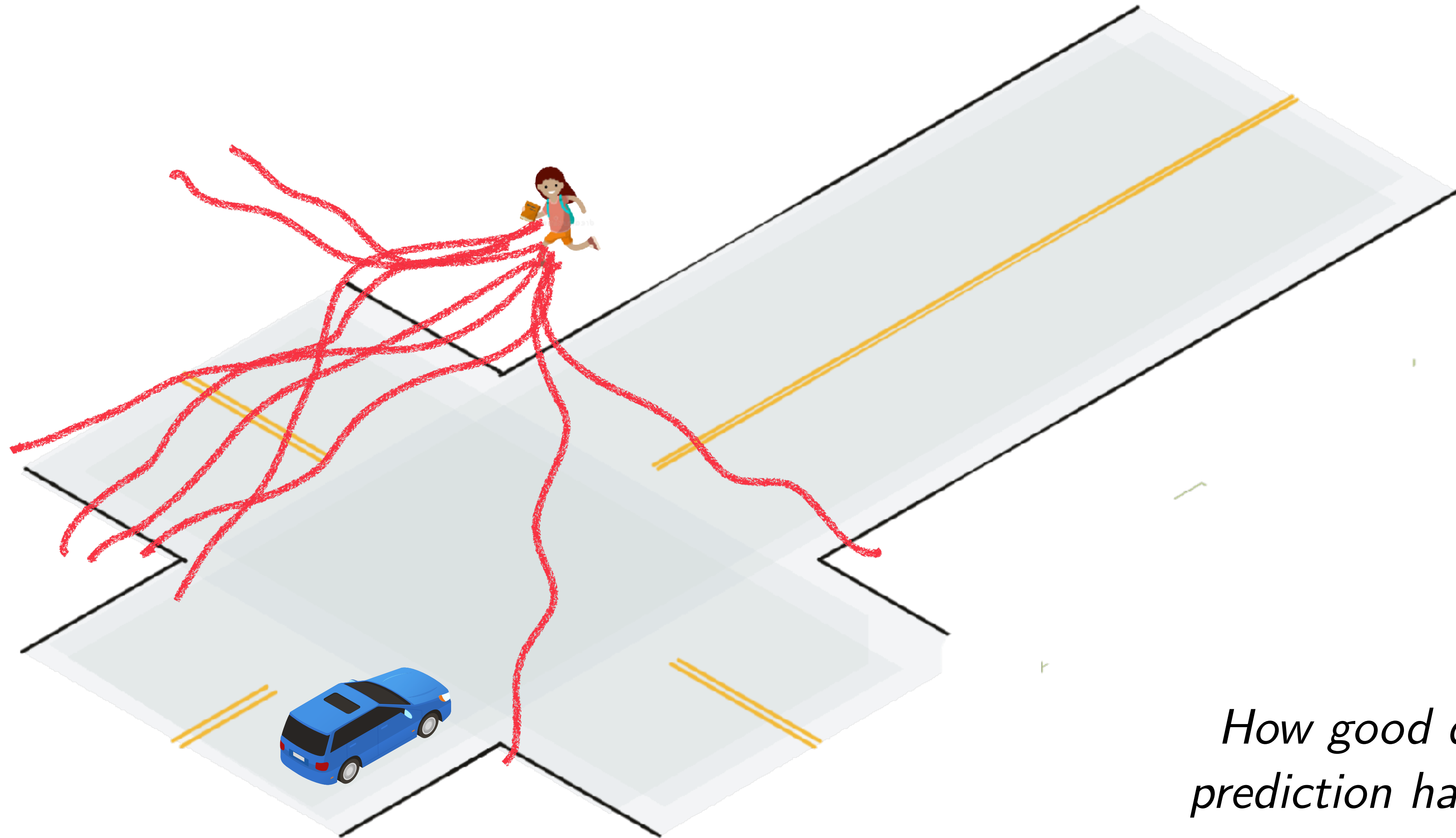
Values are **implicit** in human driving!







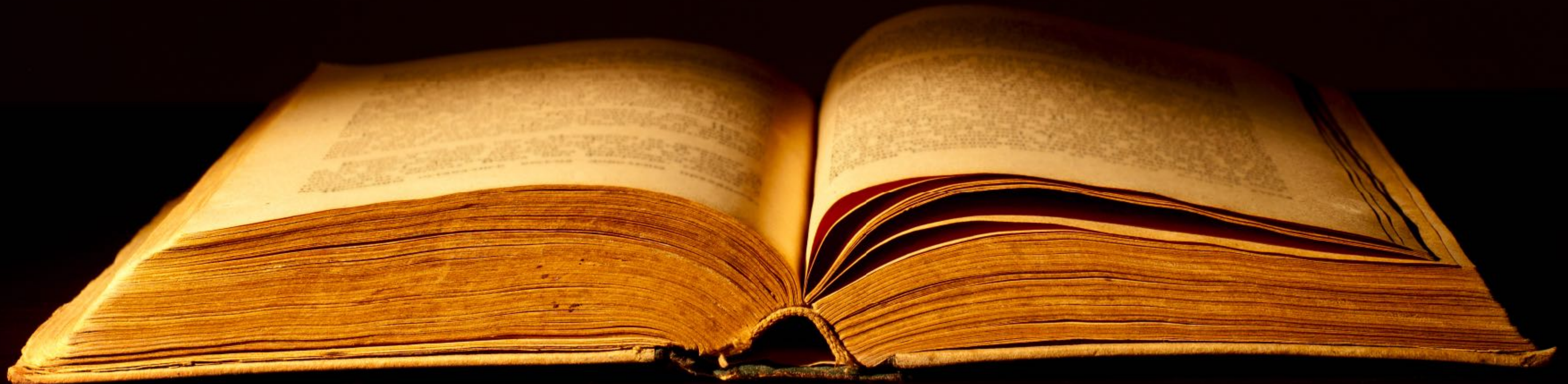
Activity: How can we predict pedestrian motion?



How good does the prediction have to be?

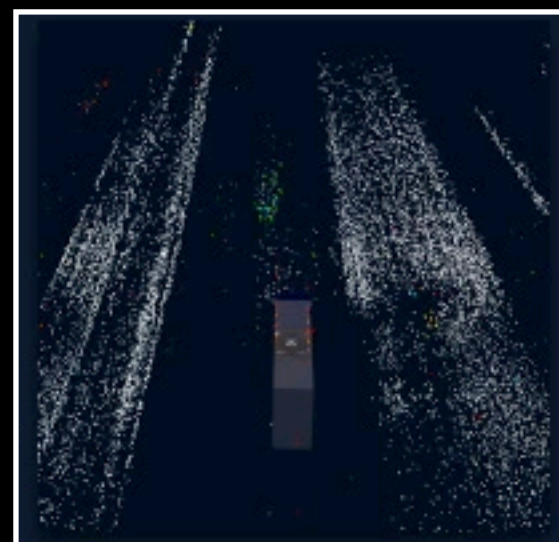
Lesson #2

Models are **useful fictions**





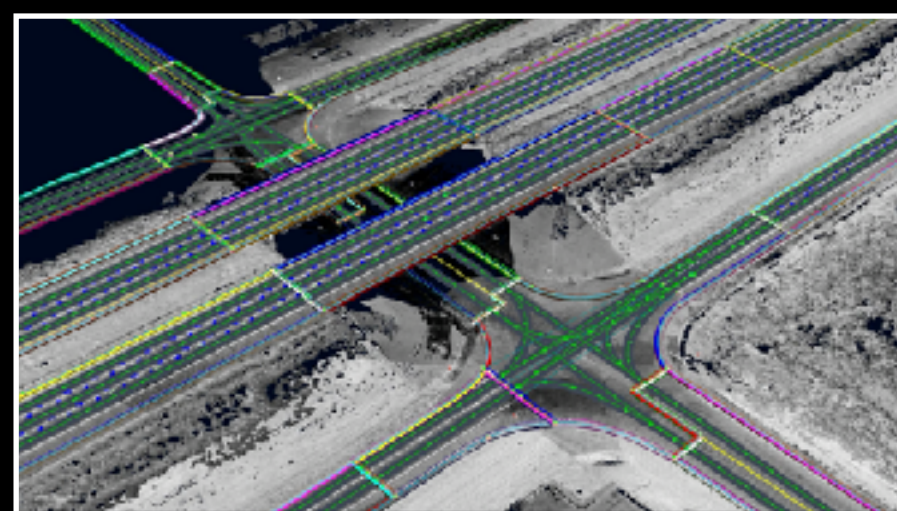
Lidar



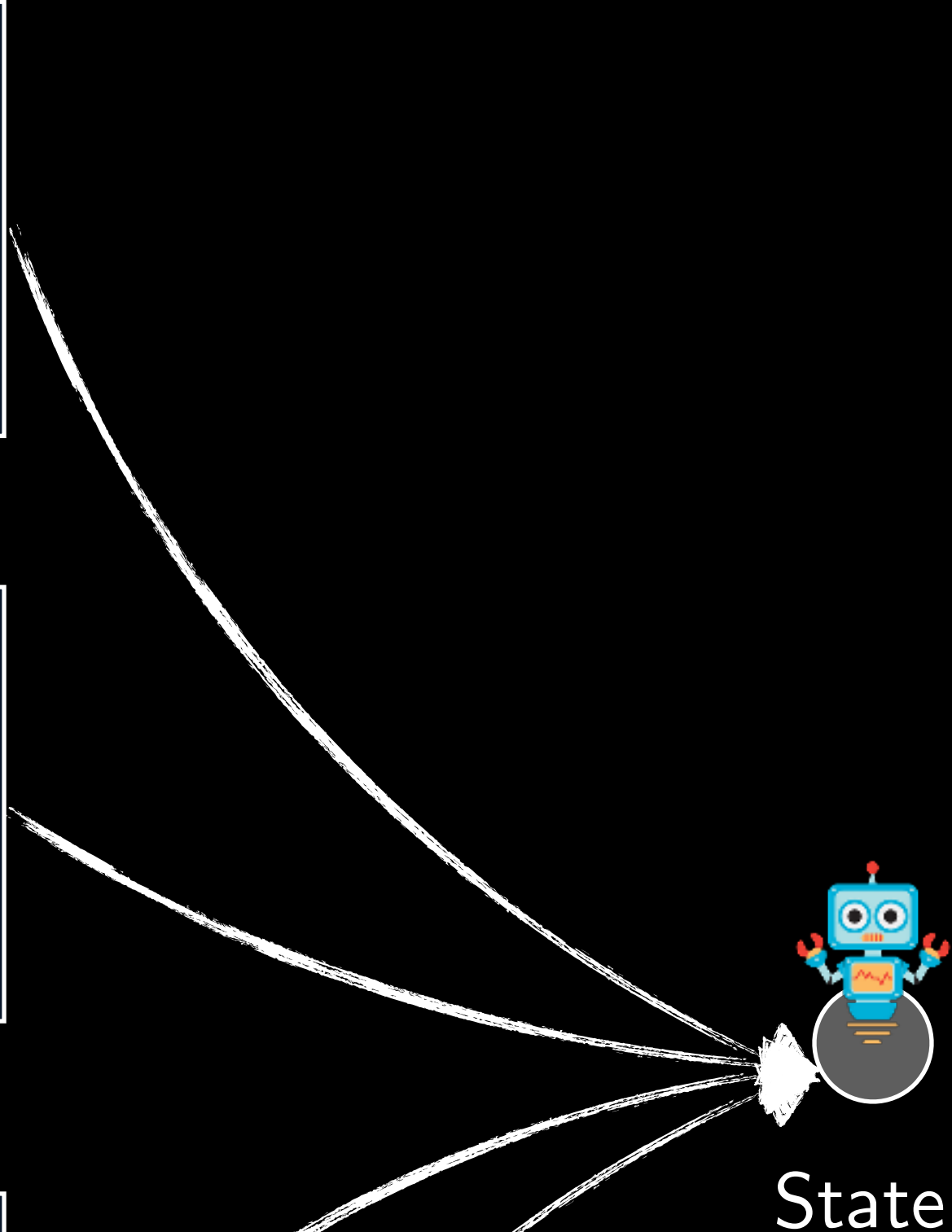
Radar



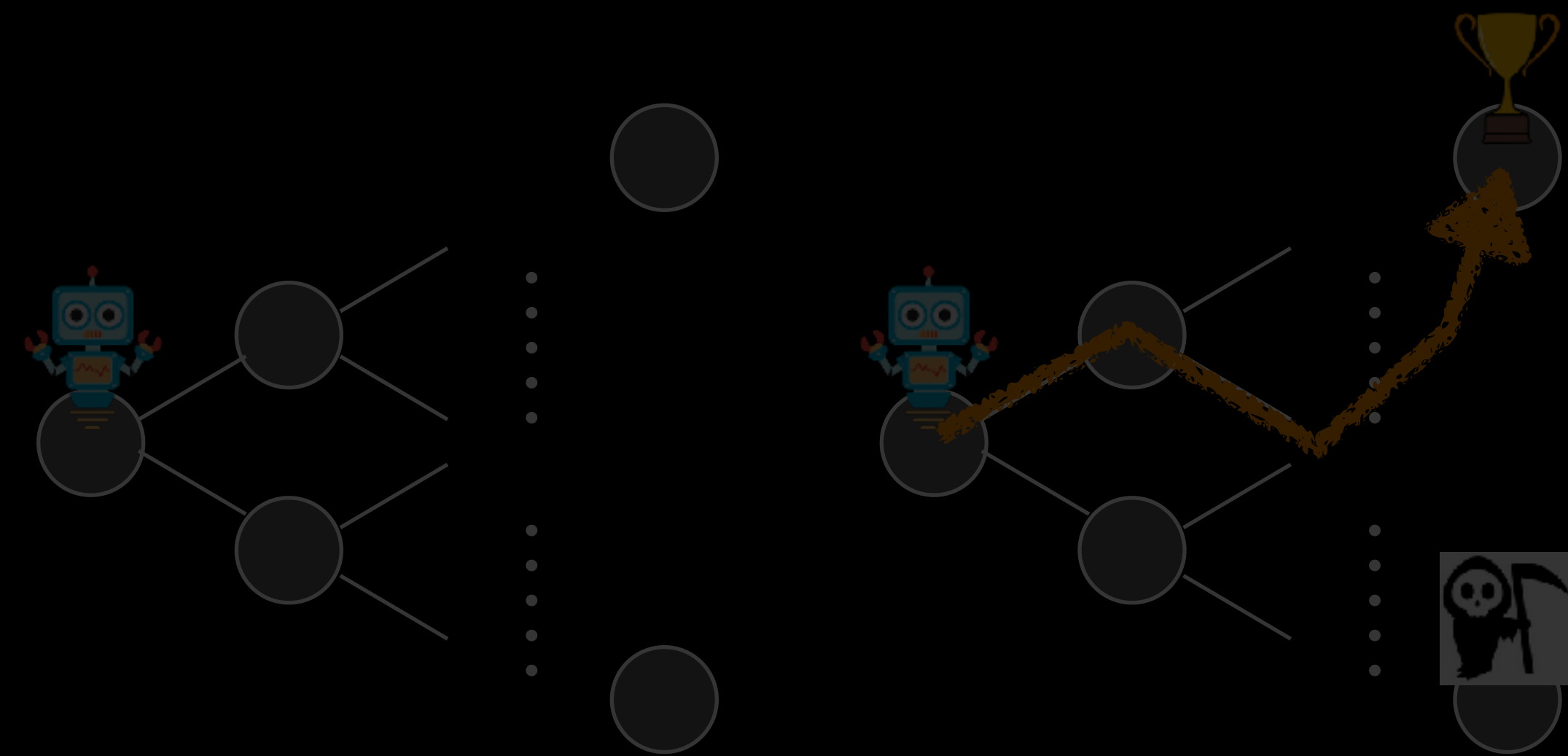
Camera



Maps



Perception



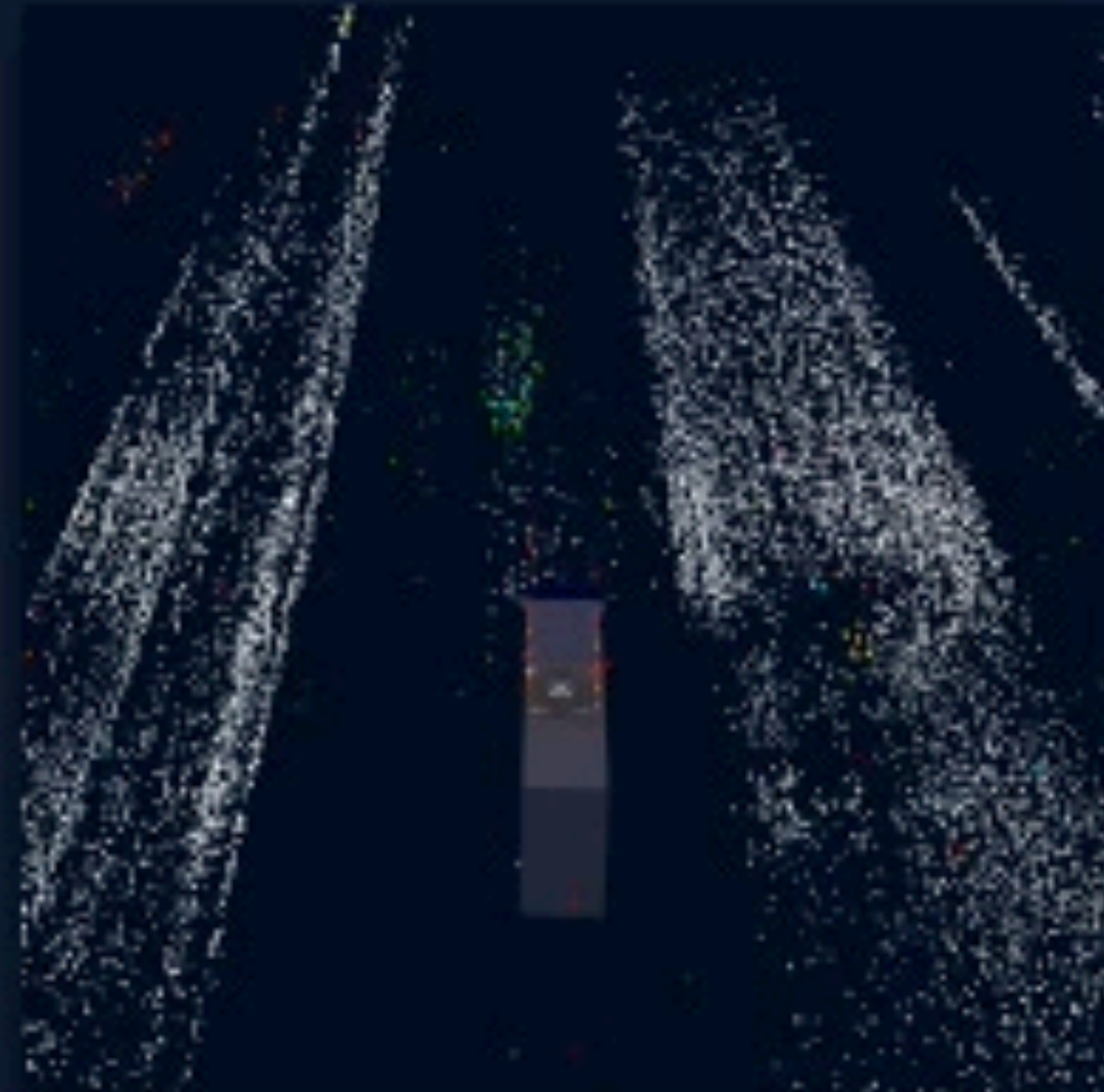
Prediction

Decision Making

No one sensor tells the whole story!



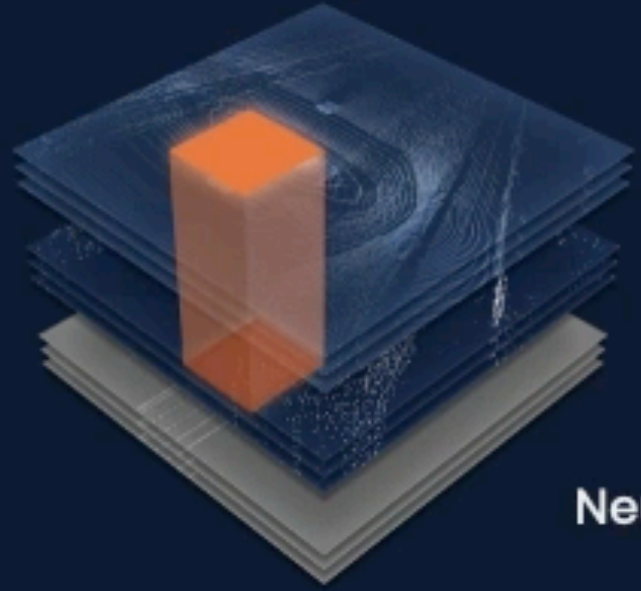
LiDAR



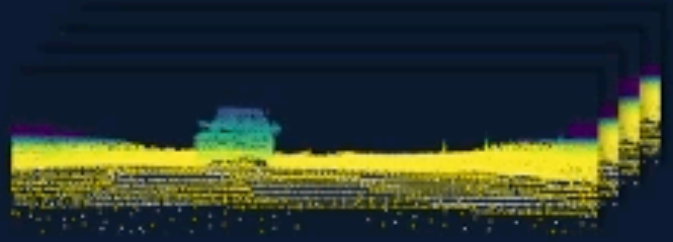
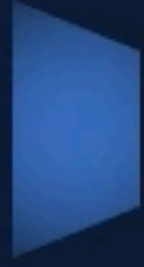
Radar



Camera



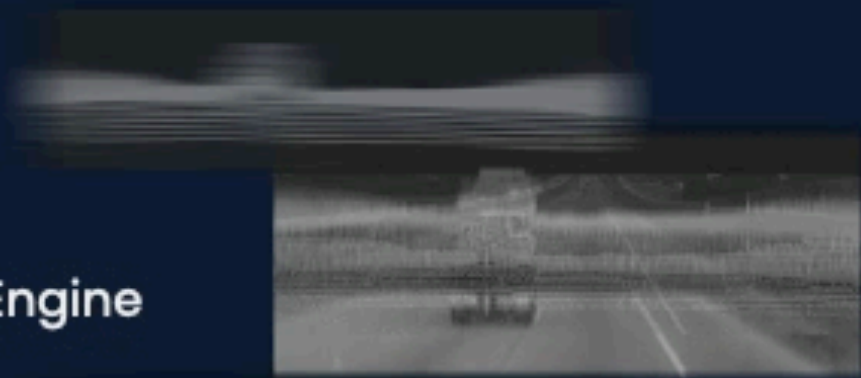
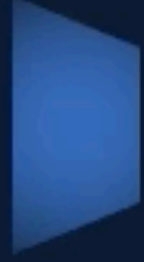
Neural Convolution Engine



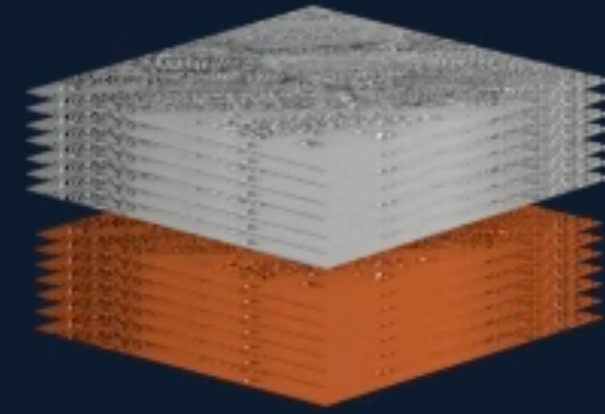
Range Conv. Engine



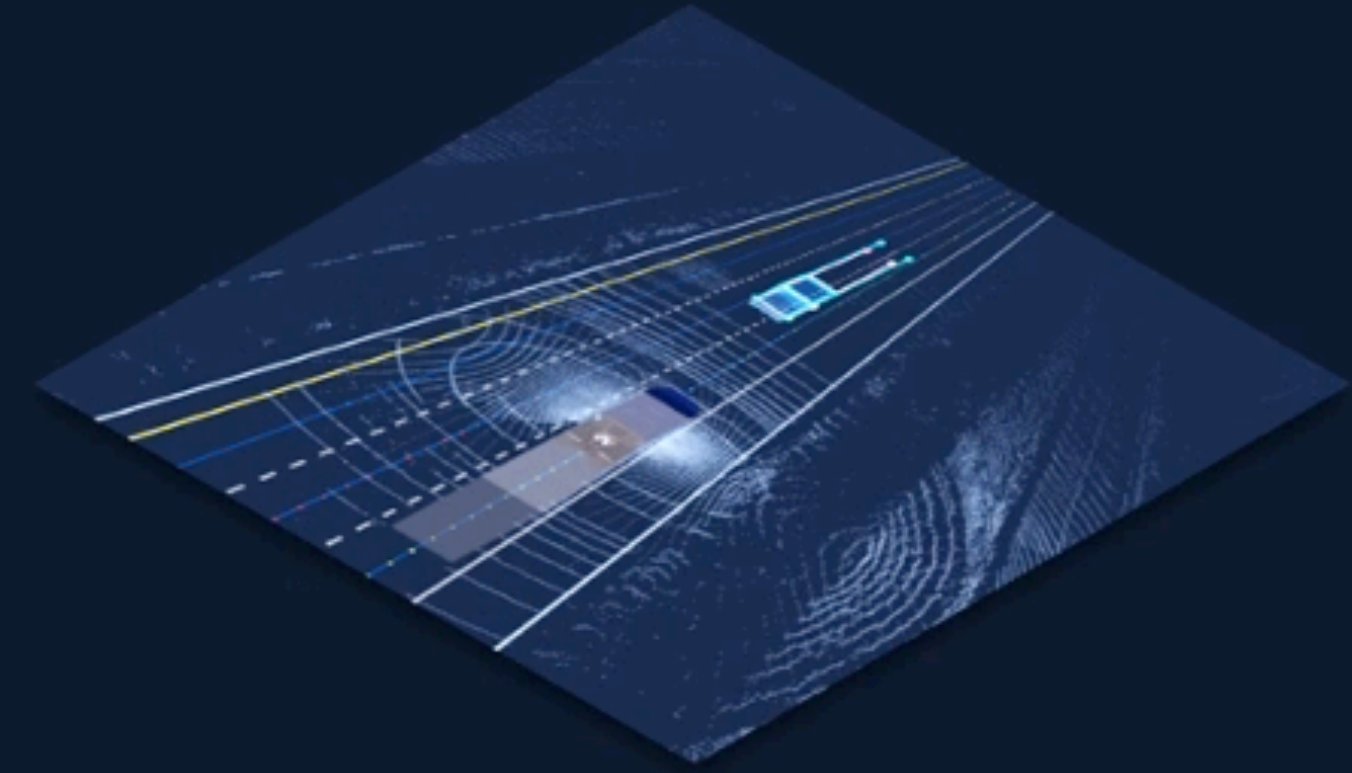
Image Fusion Engine



Euclidian Ray Scatter Engine



Neural Convolution Engine



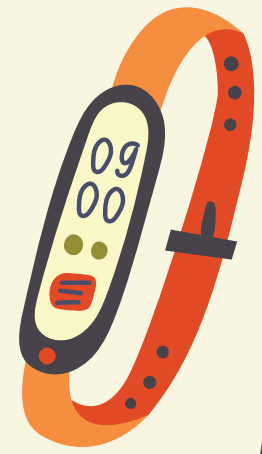
Lesson #3

Solve for the state

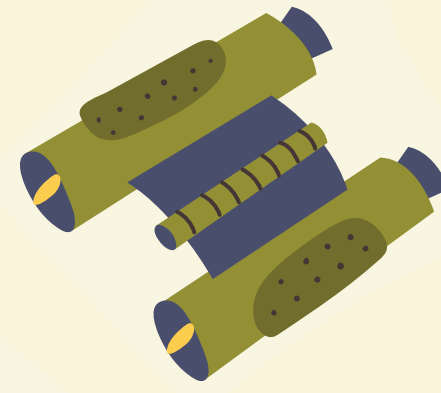
that explains all observations



Frontiers



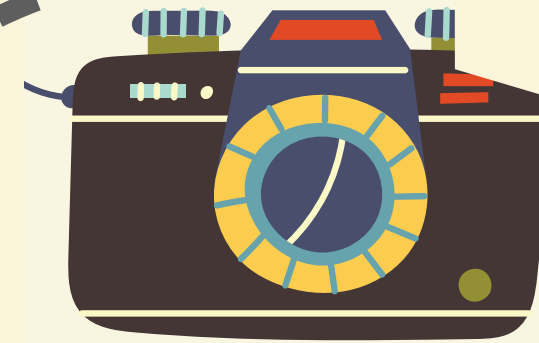
World Models
& Forecasting



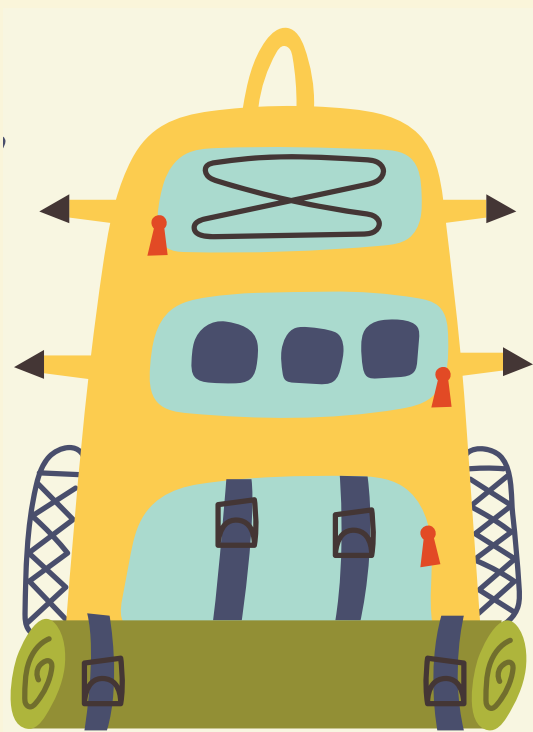
Multi-modal
Models



Visual
Representation



Fundamentals



Planning
& Control



Imitation
Learning



Reinforcement
Learning



Logistics

Website is the
ONE true hub



<https://www.cs.cornell.edu/courses/cs4756/2024fa/>

Course Book

Modern Adaptive Control and Reinforcement Learning (MACRL)

Drew Bagnell, Byron Boots, Sanjiban Choudhury

<https://macrl-book.github.io/>

Pre-reading and Resources

Date	Lecture	Preread	Resources
	Fundamentals		
08/27/24	Introduction to Robot Learning		The Bitter Lesson
08/29/24	Robots as Markov Decision Problems	MACRL Ch. 1	Dan Klein slides I

Please look at the pre-reading before coming to lecture!

Resources are for *after* the lecture if you want to go deeper into a concept.

6 Assignments [50%]

A0: Intro assignment

A1, A3: Written assignment

A2, A4, A5: Programming assignment

Assignments will be based off of concepts / exercises from class!

In-class Prelim [20%]

Use written assignments as a reference

Use course book (pre-reading chapters) as a reference

Final Project [20%]

This is your chance to be creative and apply concepts to solve some robot learning problems!

See this doc for ideas.

We, unfortunately, do not have GPUs to offer, so choose projects wisely that you can run on your machines. Talk to TAs!

The best projects are simple ideas that convey insight!

Participation [10%]

Participate in class polls and exercises!

Graduate Version (CS5756)

If you are enrolled in CS 5756, every assignment has an **extra question** that you must solve.

Course Policies

All policies are posted on the Website!

Course Website: 3 TOTAL late days. Any assignment turned in late will incur a reduction in score by 33% for each late day

Academic Integrity: Any work presented as your own must be your own, with no exceptions tolerated. Submitting work created by ChatGPT, or copied from a bot or a website, as your own work violates academic integrity.

Generative AI

The work you do consists of writing code and natural language descriptions.

To some extent, the new crop of “generative AI” (GAI) tools can do both of these things for you.

However, **we require that the vast majority of the intellectual work must be originated by you**, not by GAI. You may use GAI to look up helper functions, or to proofread your text, but clearly document how you used it.

Generative AI

In this class, for every assignment and final project, you can choose between two options:

Option 1: Avoid all GAI tools. Disable GitHub Copilot in your editor, do not ask chatbots any questions related to the assignment, etc. If you choose this option, you have nothing more to do.

Option 2: Use GAI tools with caution and include a one-paragraph description of everything you used them for along with your writeup. This paragraph must:

1. Link to exactly which tools you used and describe how you used each of them, for which parts of the work.
2. Give at least one concrete example (e.g., generated code or Q&A output) that you think is particularly illustrative of the “help” you got from the tool.
3. Describe any times when the tool was unhelpful, especially if it was wrong in a particularly hilarious way.
4. Conclude with your current opinion about the strengths and weaknesses of the tools you used for real-world compiler implementation.

Remember that you can pick whether to use GAI tools for every assignment, so using them on one set of tasks doesn't mean you have to keep using them forever.



The Crew



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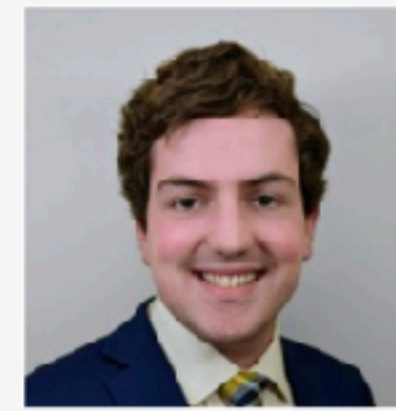
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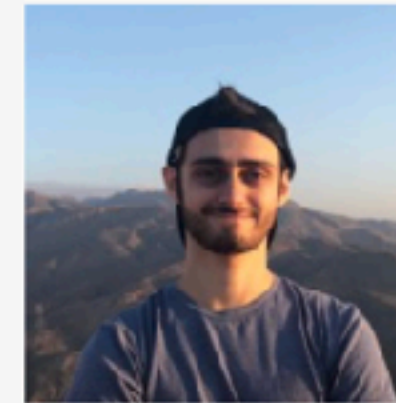
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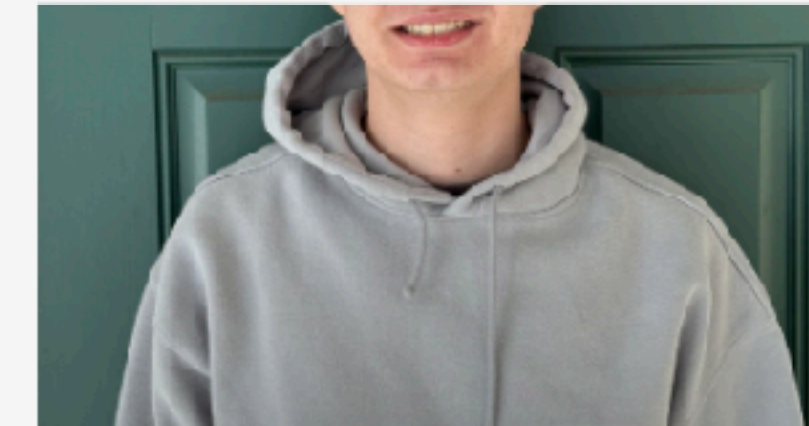
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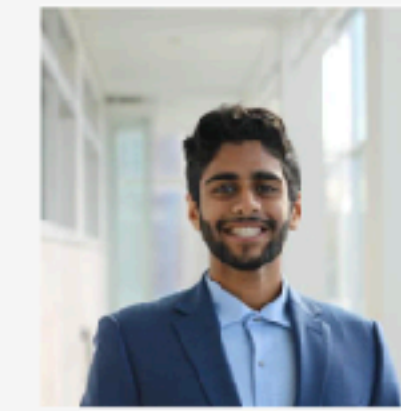
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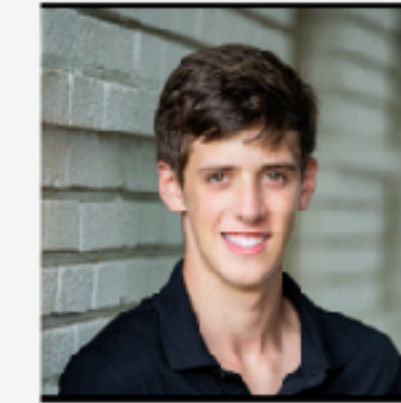
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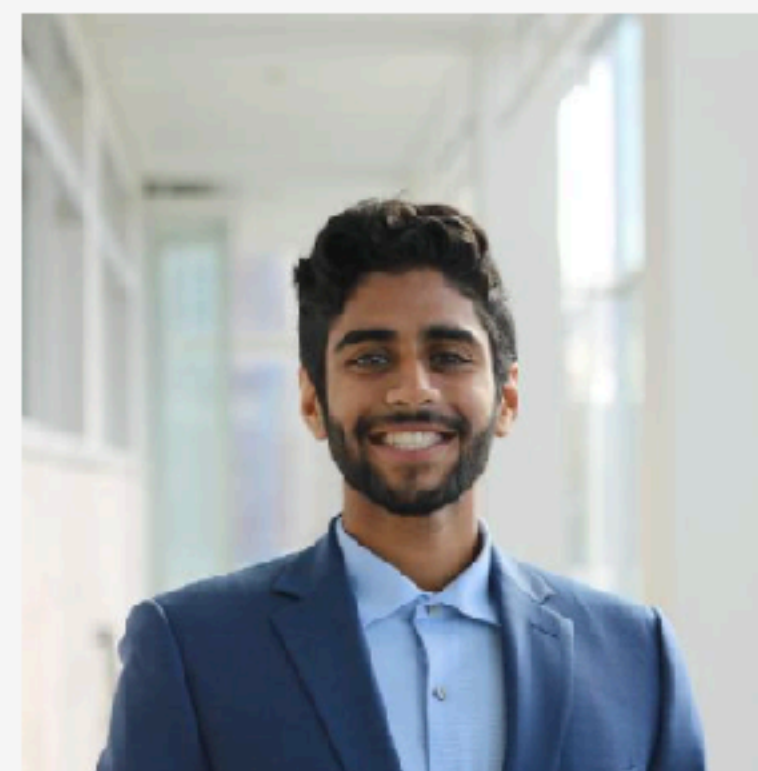
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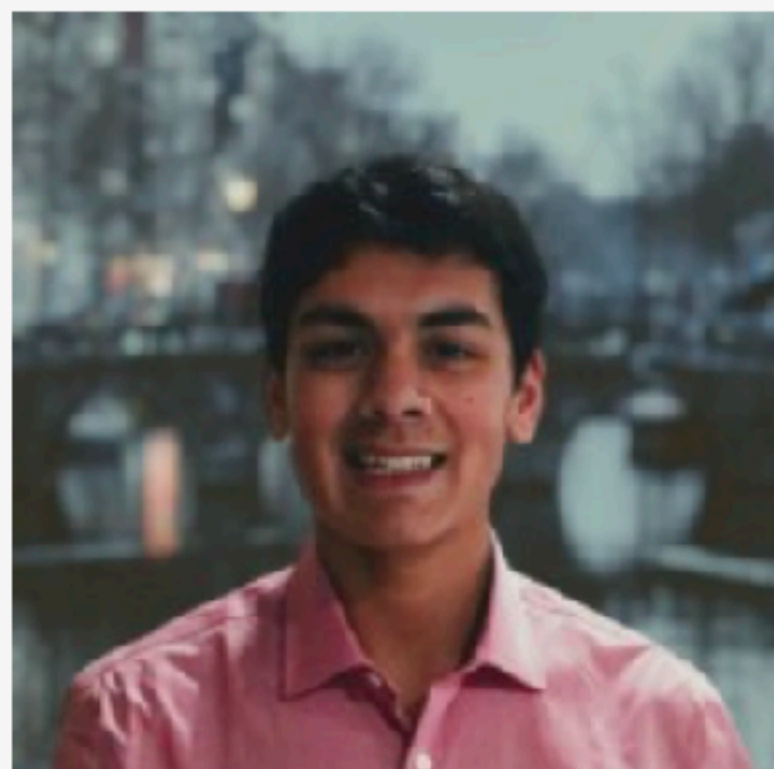
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Wait list info

- Sanjiban cannot do anything about the waitlist / getting into the course
- If you are on the waitlist, you will eventually receive a PIN. Keep checking your emails.
- If you are not on the waitlist, please wait for the waitlist to be cleared before adding yourself on.
- If you are unable to add yourself to the waitlist, reach out to the registrar's office
- Historically, everyone has gotten off of the waitlist. *We hope* this is the case this semester!

TLDR

Checkout course website for all details:

<https://www.cs.cornell.edu/courses/cs4756/2024fa/>

Checkout pre-reading for next lecture!