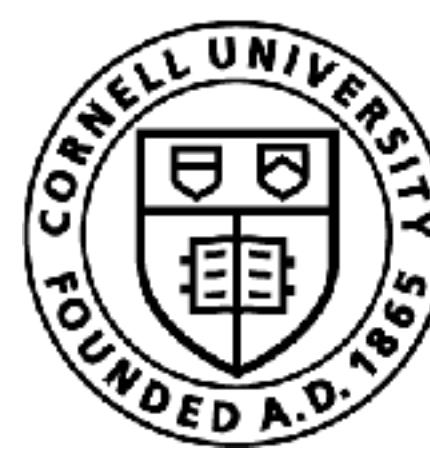


# CS 4756/5756: Robot Learning

Sanjiban Choudhury



Cornell Bowers CIS  
**Computer Science**

**WHAT A  
TIME TO  
ALIVE**

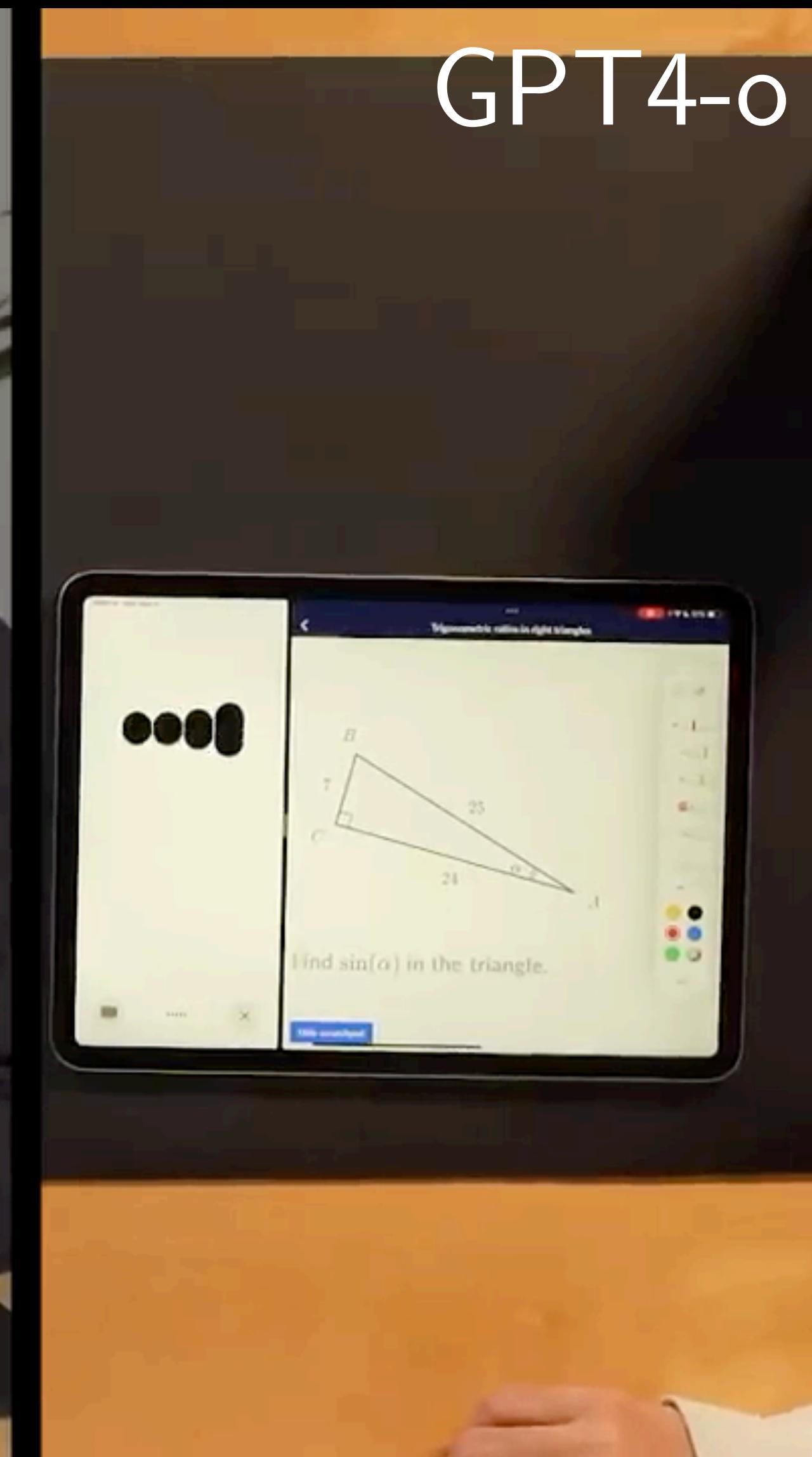
2024 continues to be an exciting year for  
Machine Learning

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



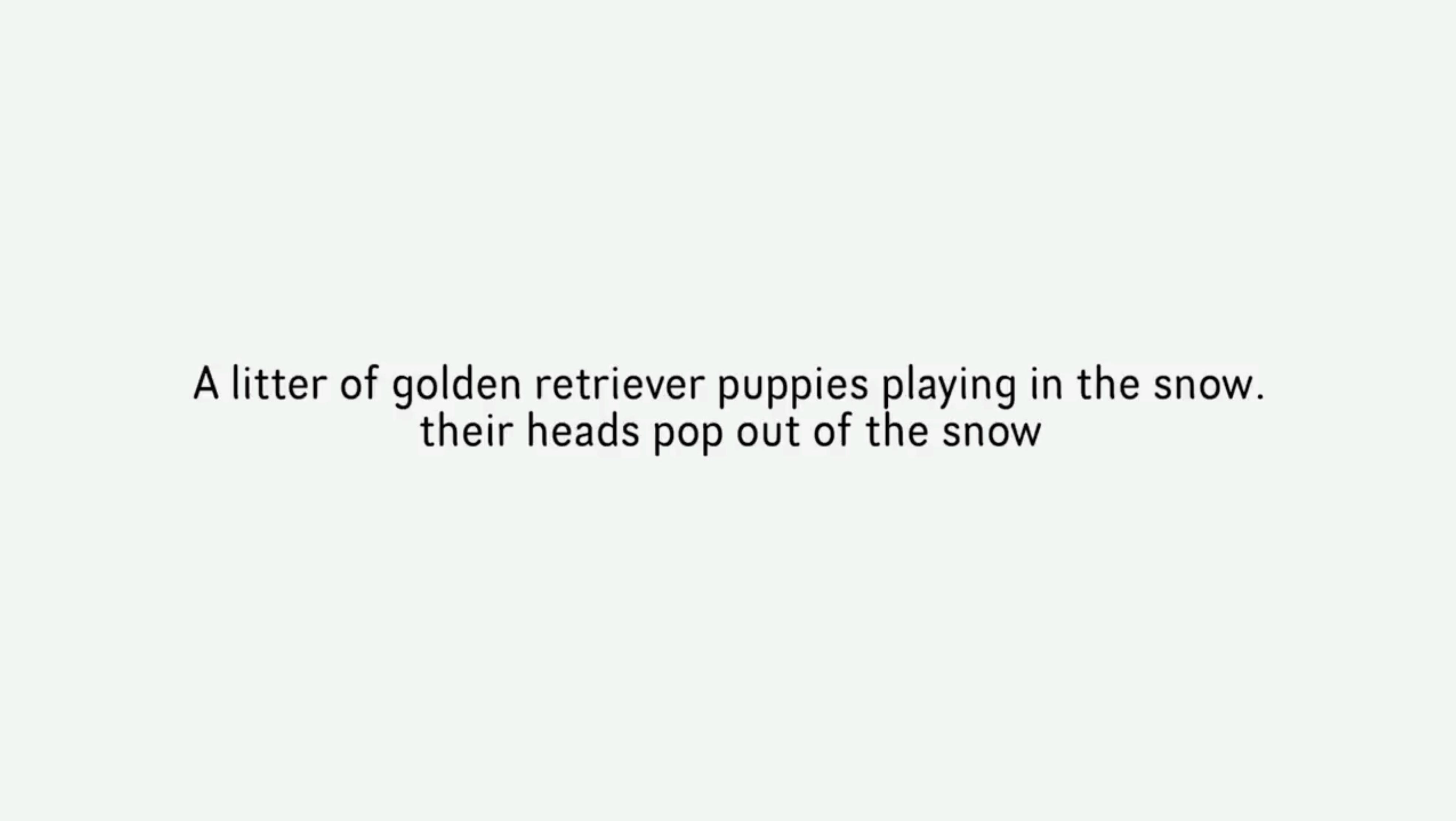
Three advances in the last semester  
that I am excited about:

# #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:

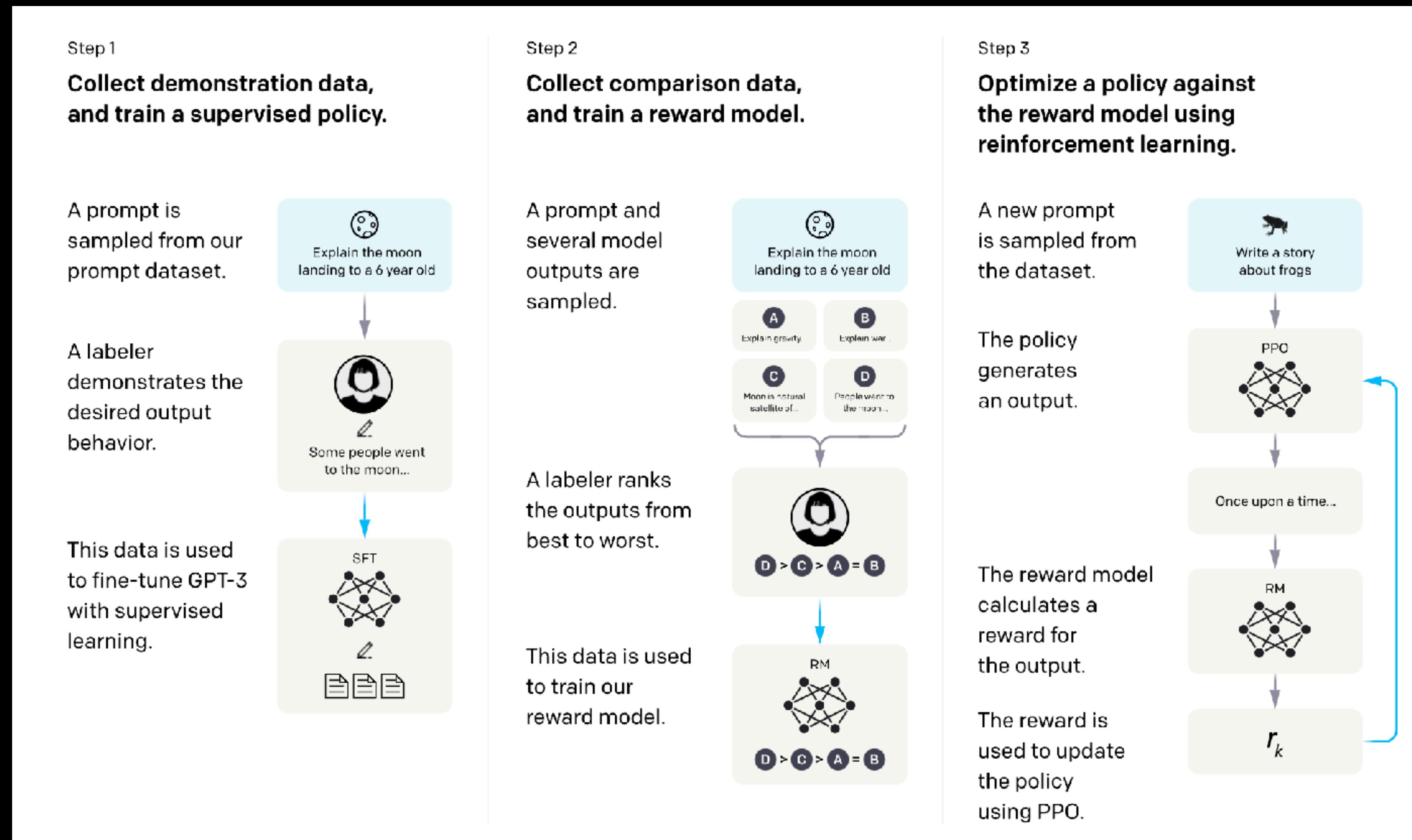
```
...
Traceback (most recent call last):
  File "test-mm.py", line 8, in <module>
    s = MySchema()
  File "/Users/victor/.pyenv/versions/marshmallow/lib/python3.6/site-packages/marshmallow/schema.py", line 383, in __init__
    self.fields = self._init_fields()
  File "/Users/victor/.pyenv/versions/marshmallow/lib/python3.6/site-packages/marshmallow/schema.py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



# Open-source fine-tunable models



LLAMA



Alpaca

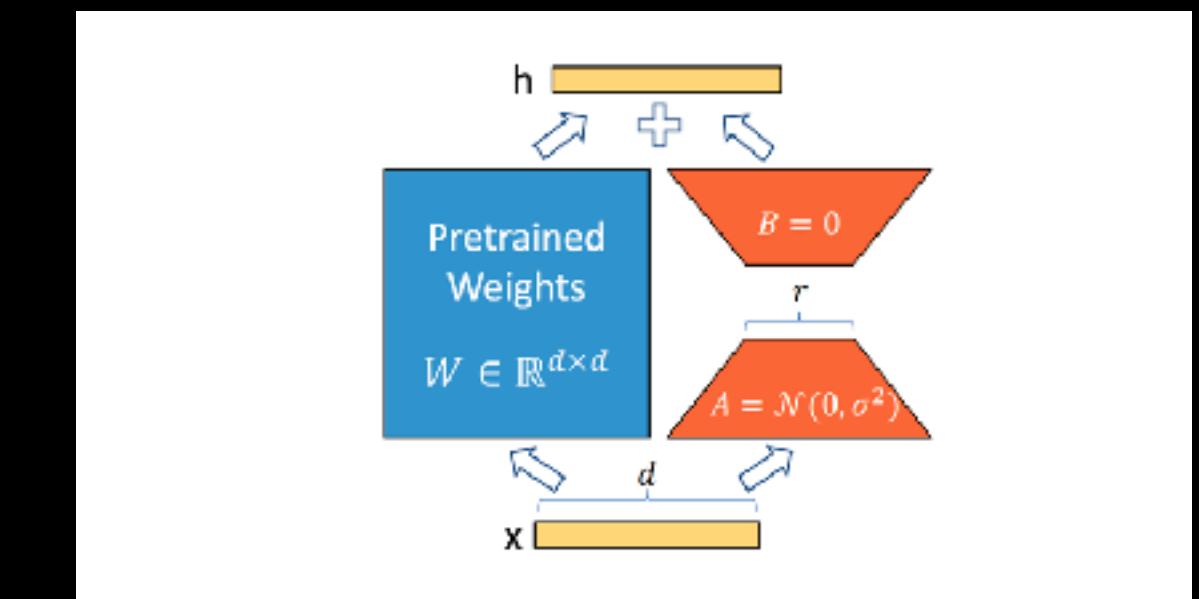


Vicuna

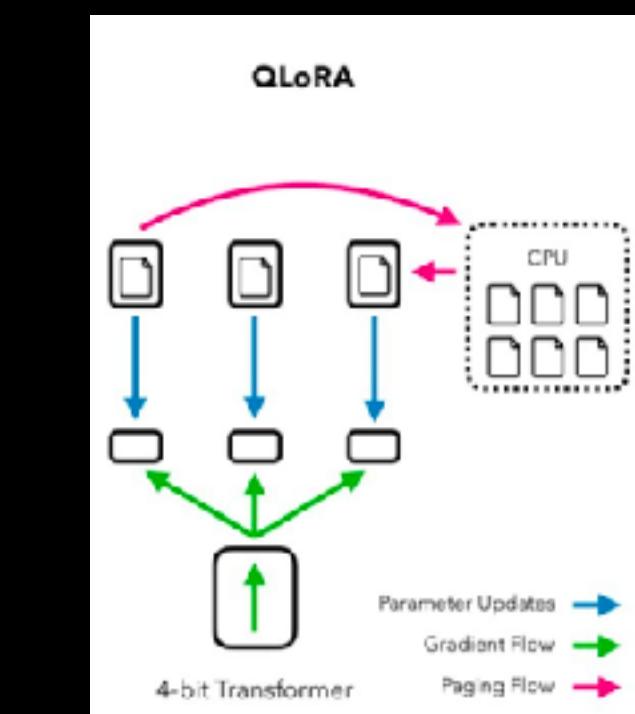


Mistral

Parameter Efficient  
Fine Tuning  
(PEFT)

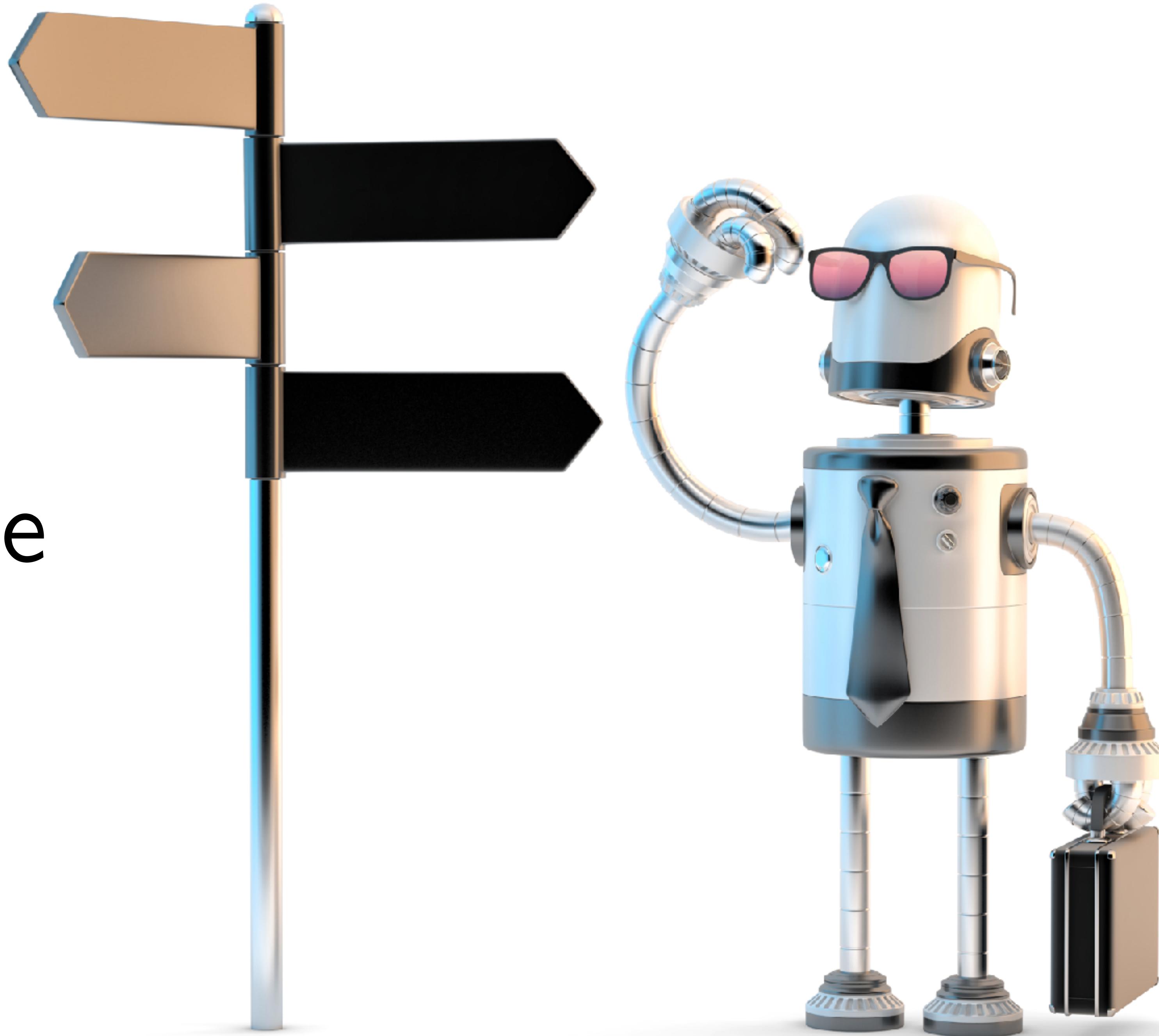


LORA

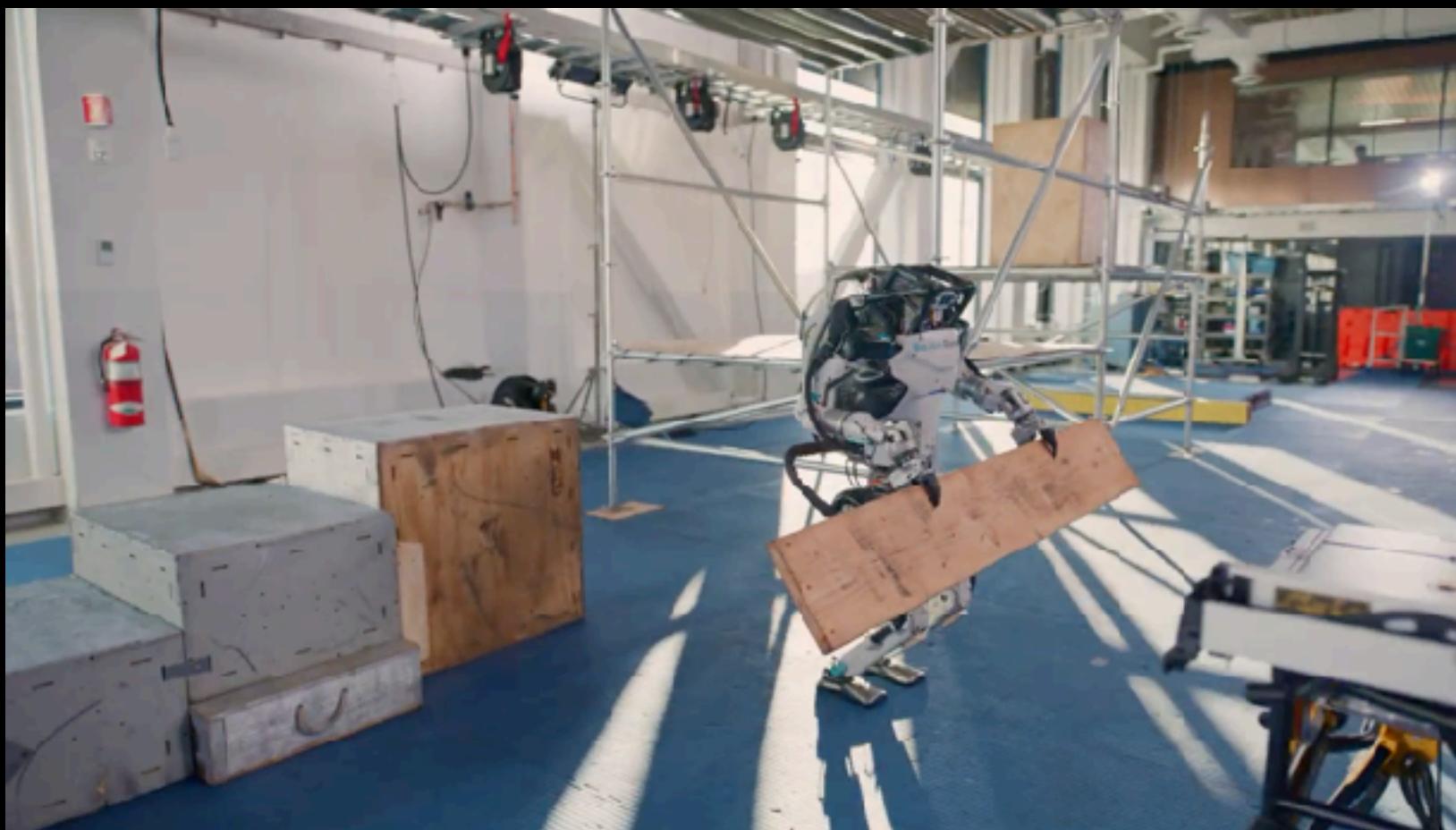


QLORA

Where are the  
robots?



# Rise of the Humanoids



Boston Dynamics



Tesla



Agility Robotics



Figure AI

# Quadrupeds going strong



Boston Dynamics SPOT

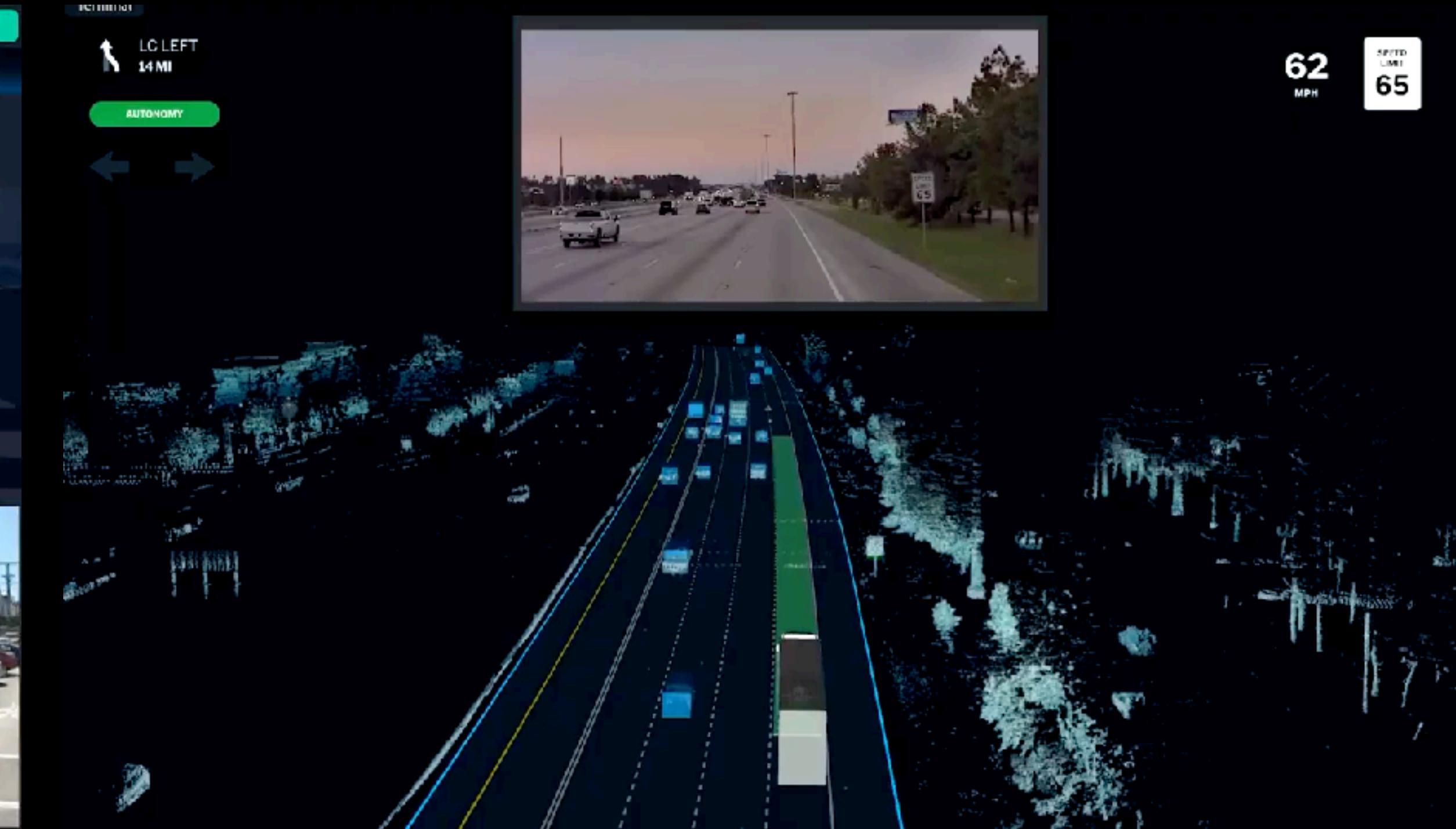


Unitree GO 2



ANYmal

# Self-driving continues driverless runs



But ...

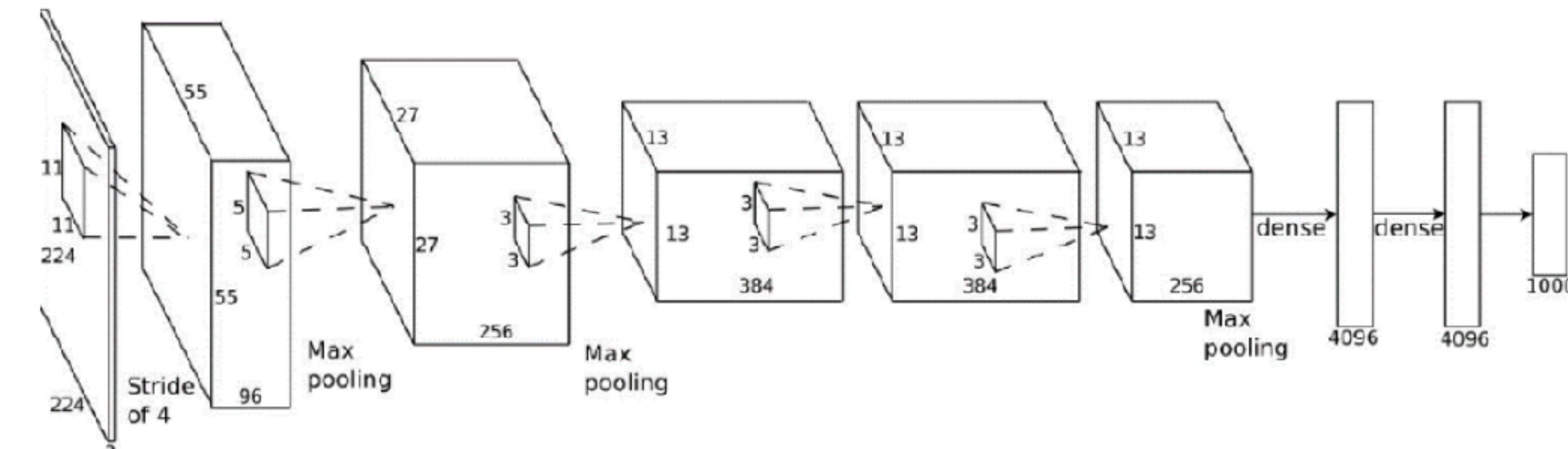
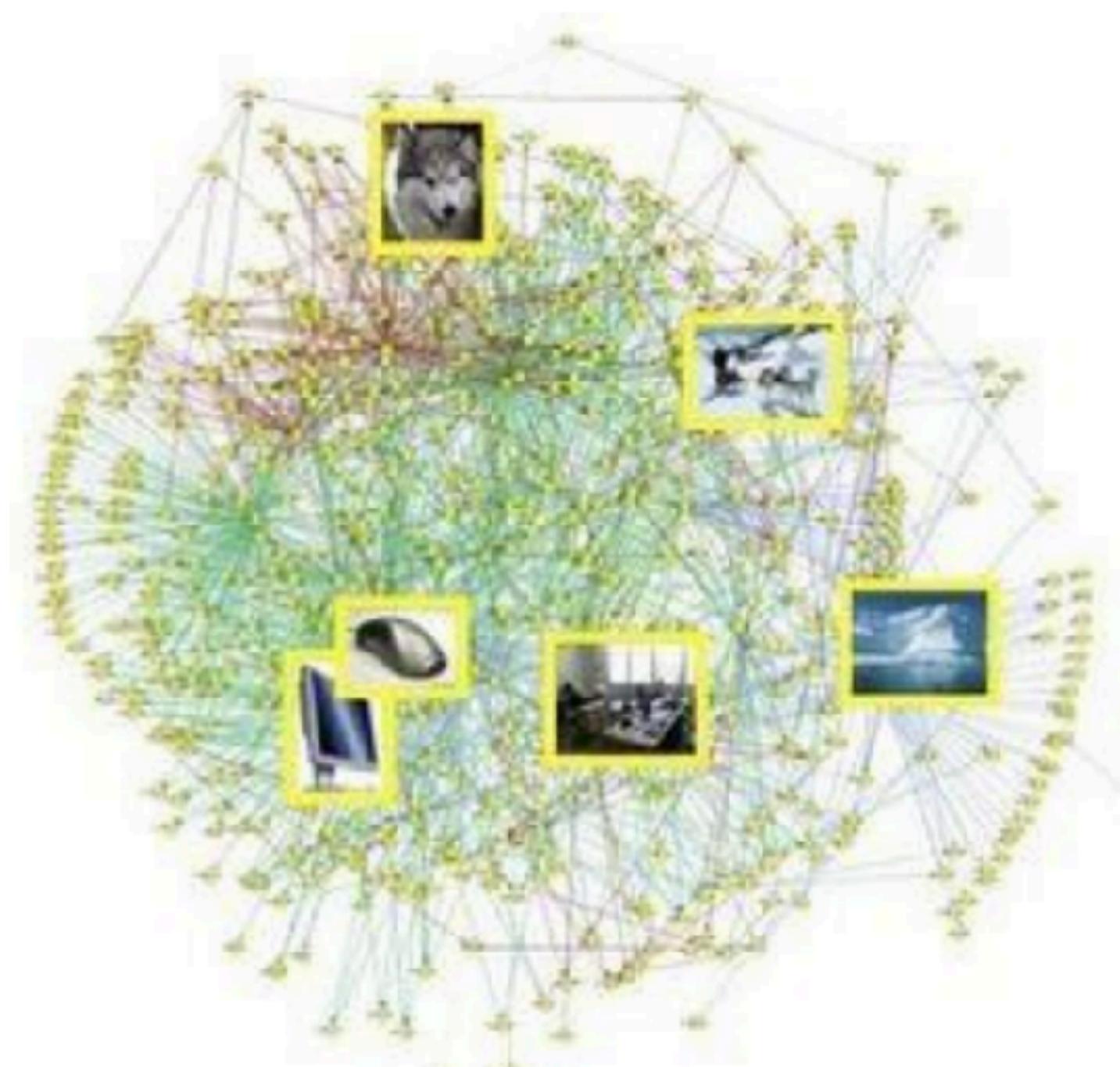
... these robots are not in  
millions of homes yet.

Why?

# What is so challenging about our homes?



# Why can't we throw ML at this problem?

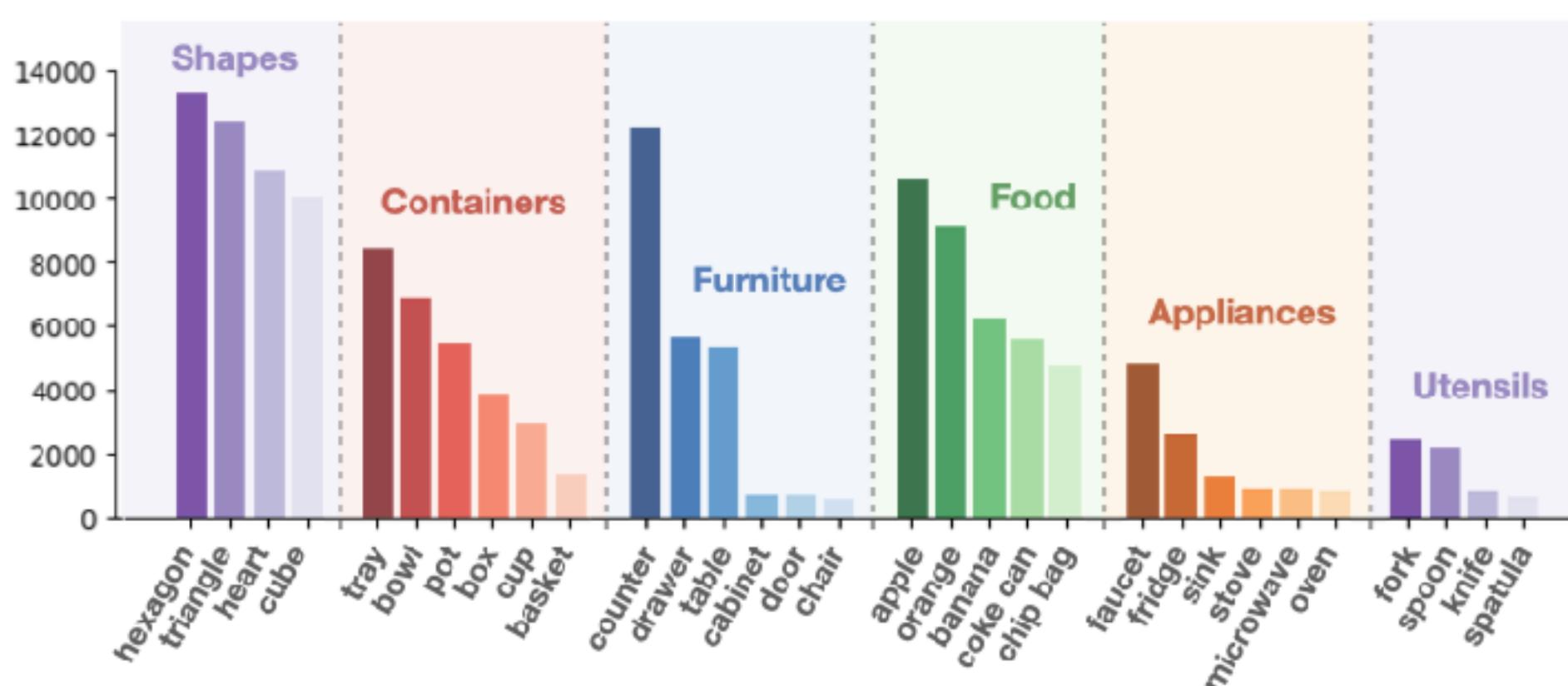
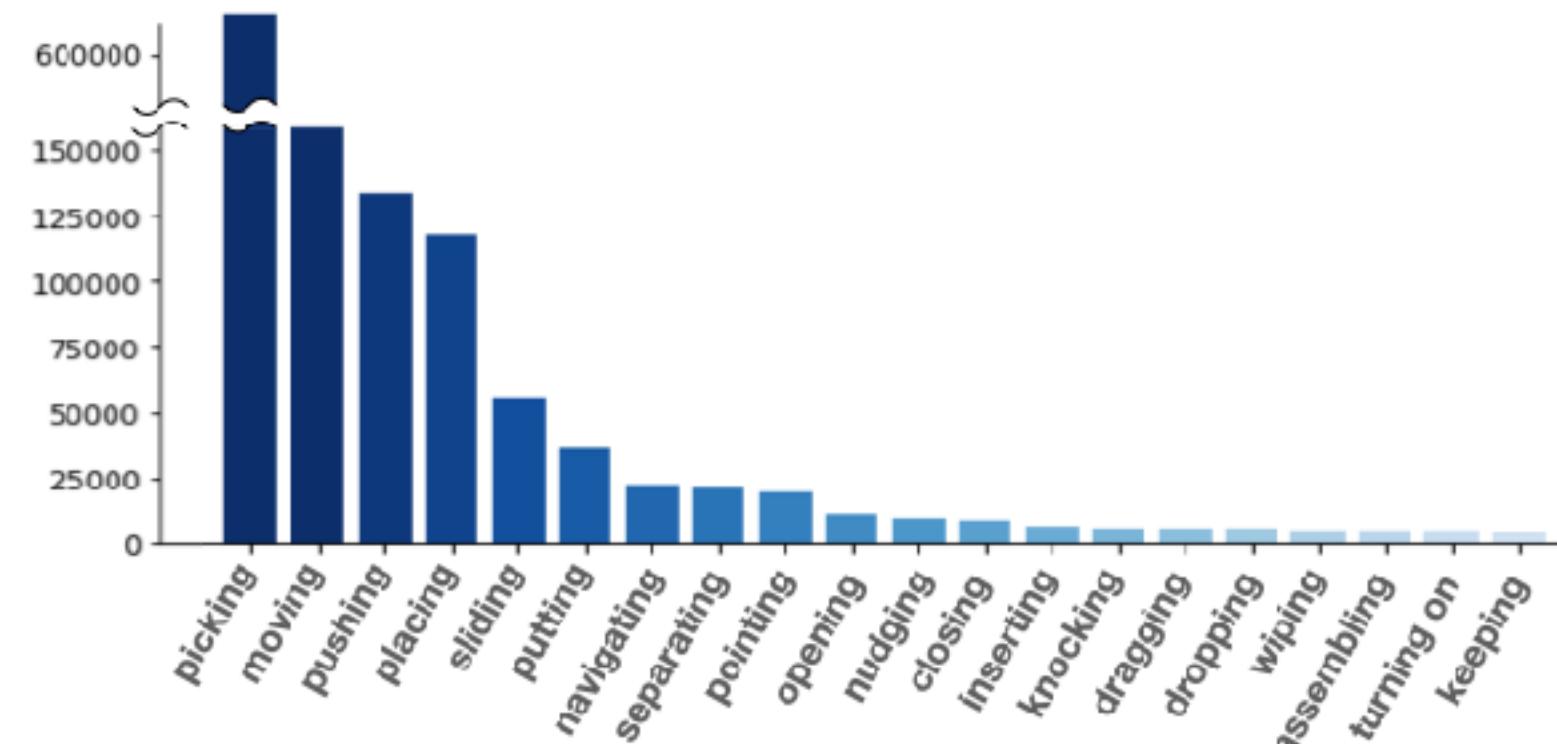
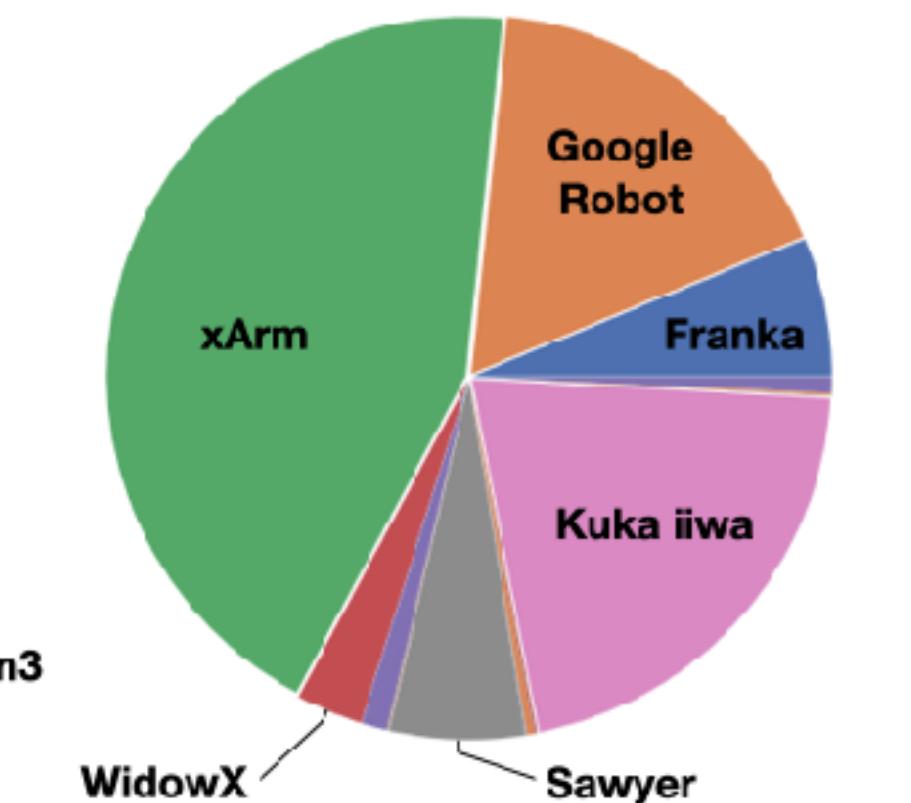
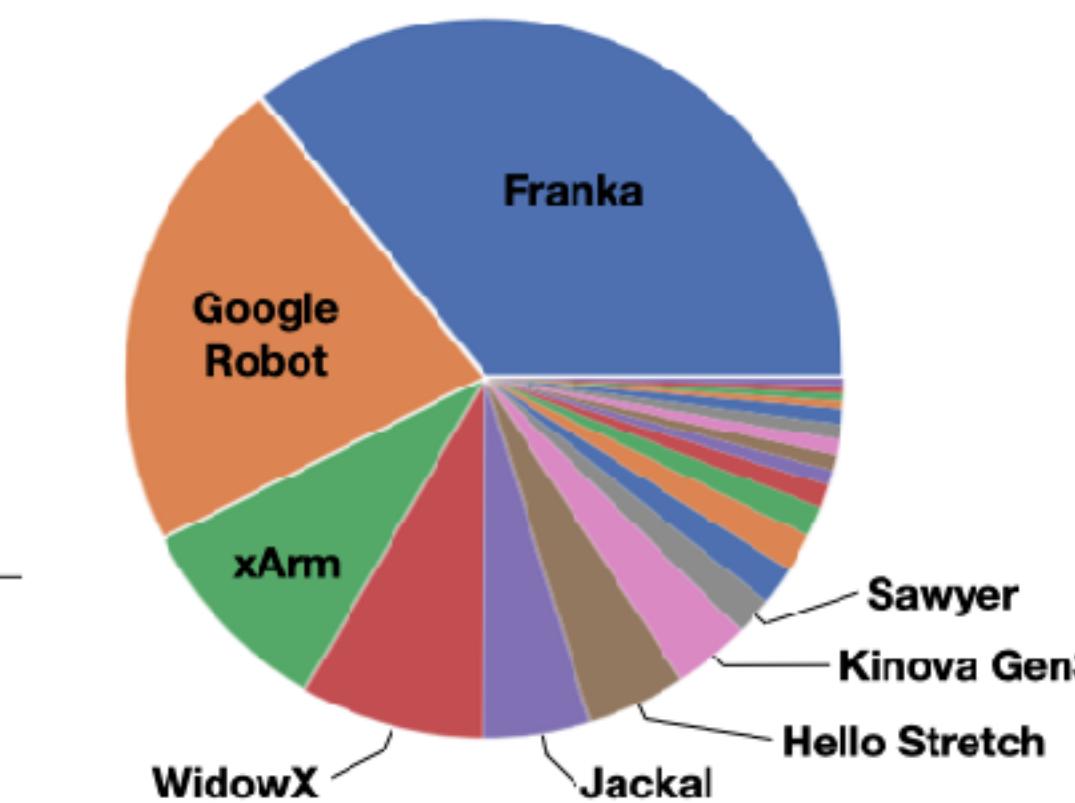
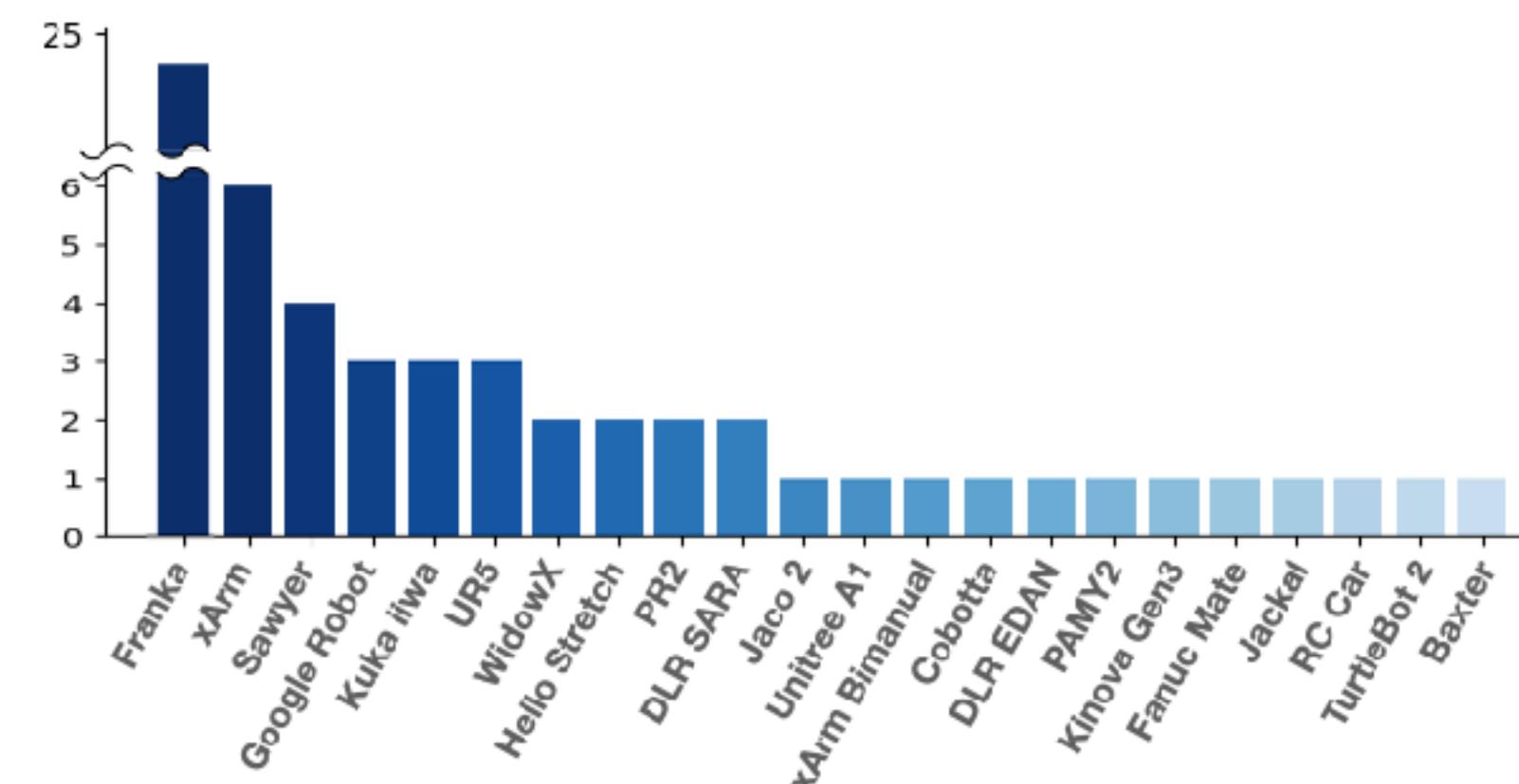


Big Data

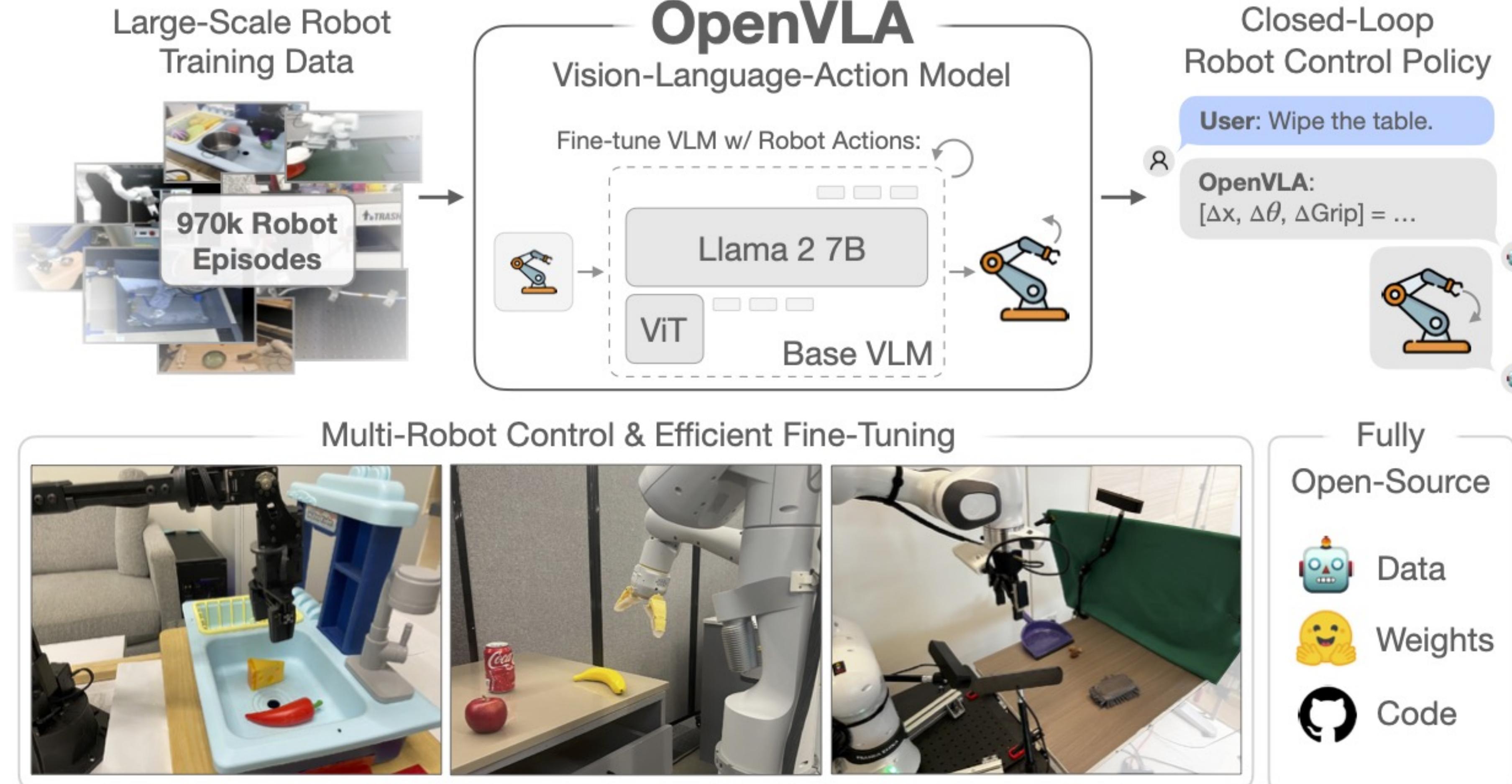
Big Models

# Big Data

1M trajectories, 22 robots, 21 different institutions



# Big Models



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

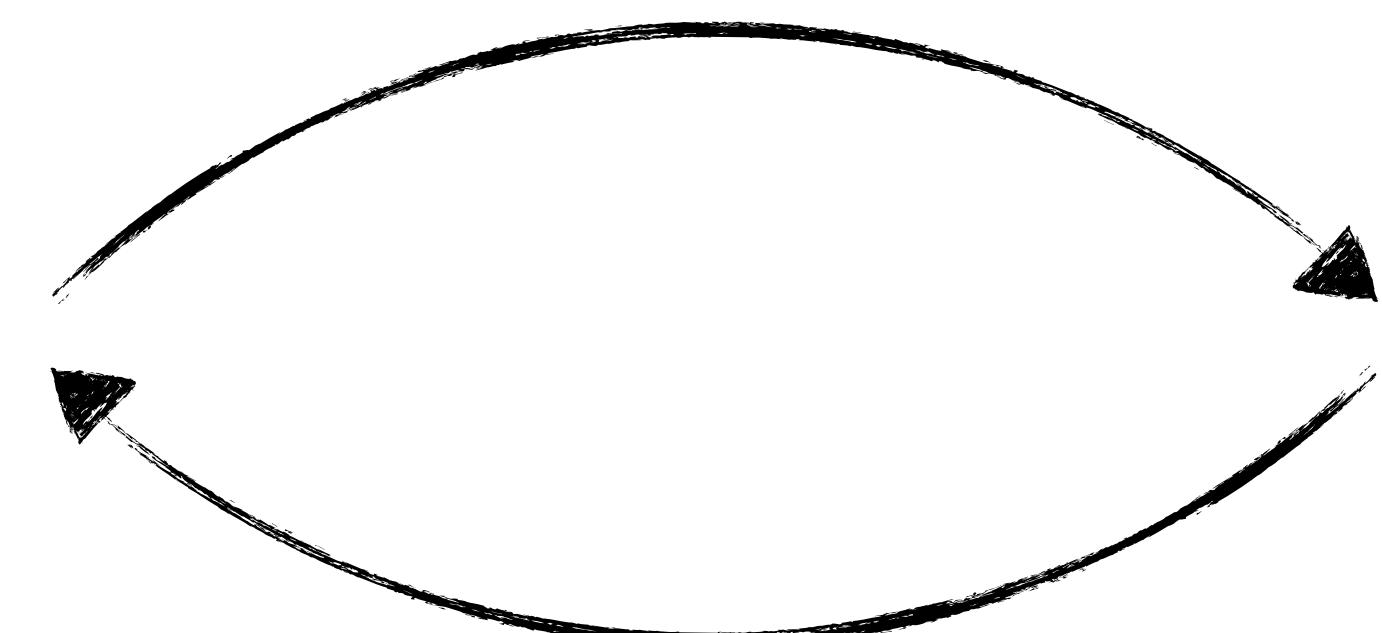
# Activity!



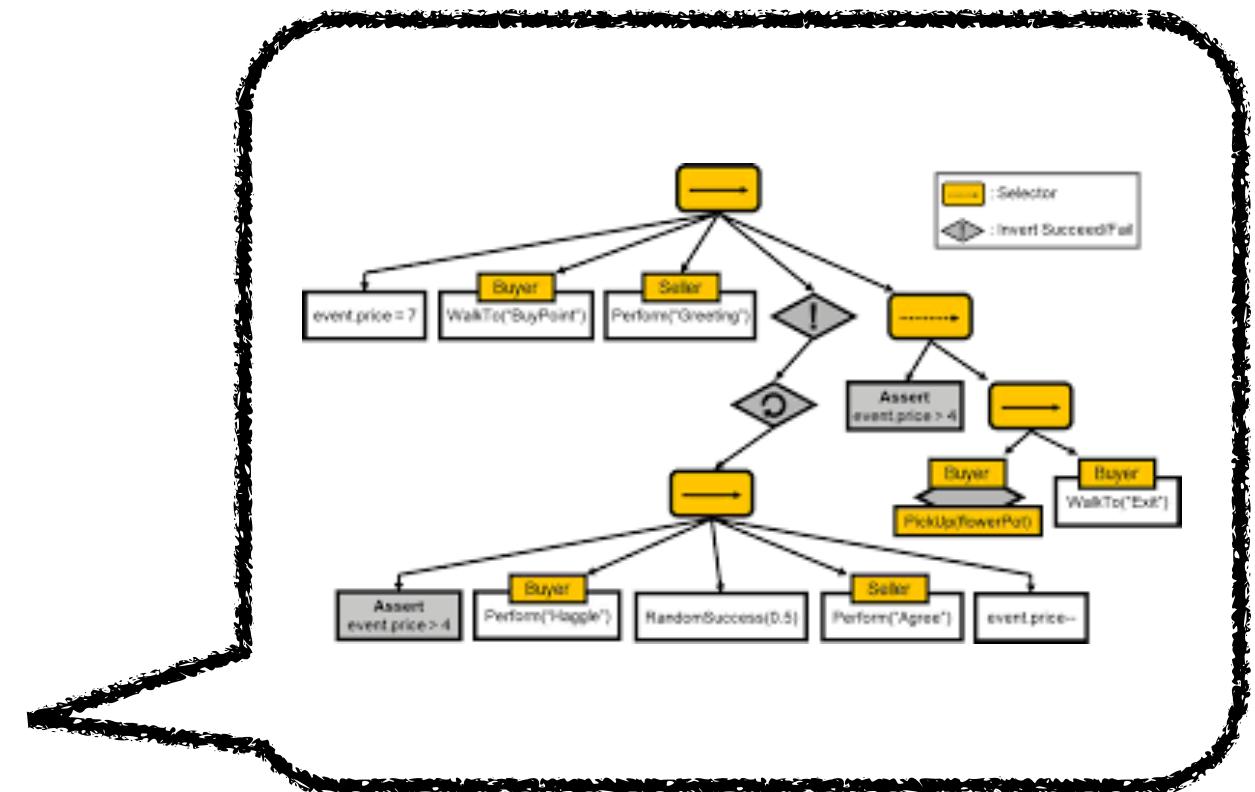
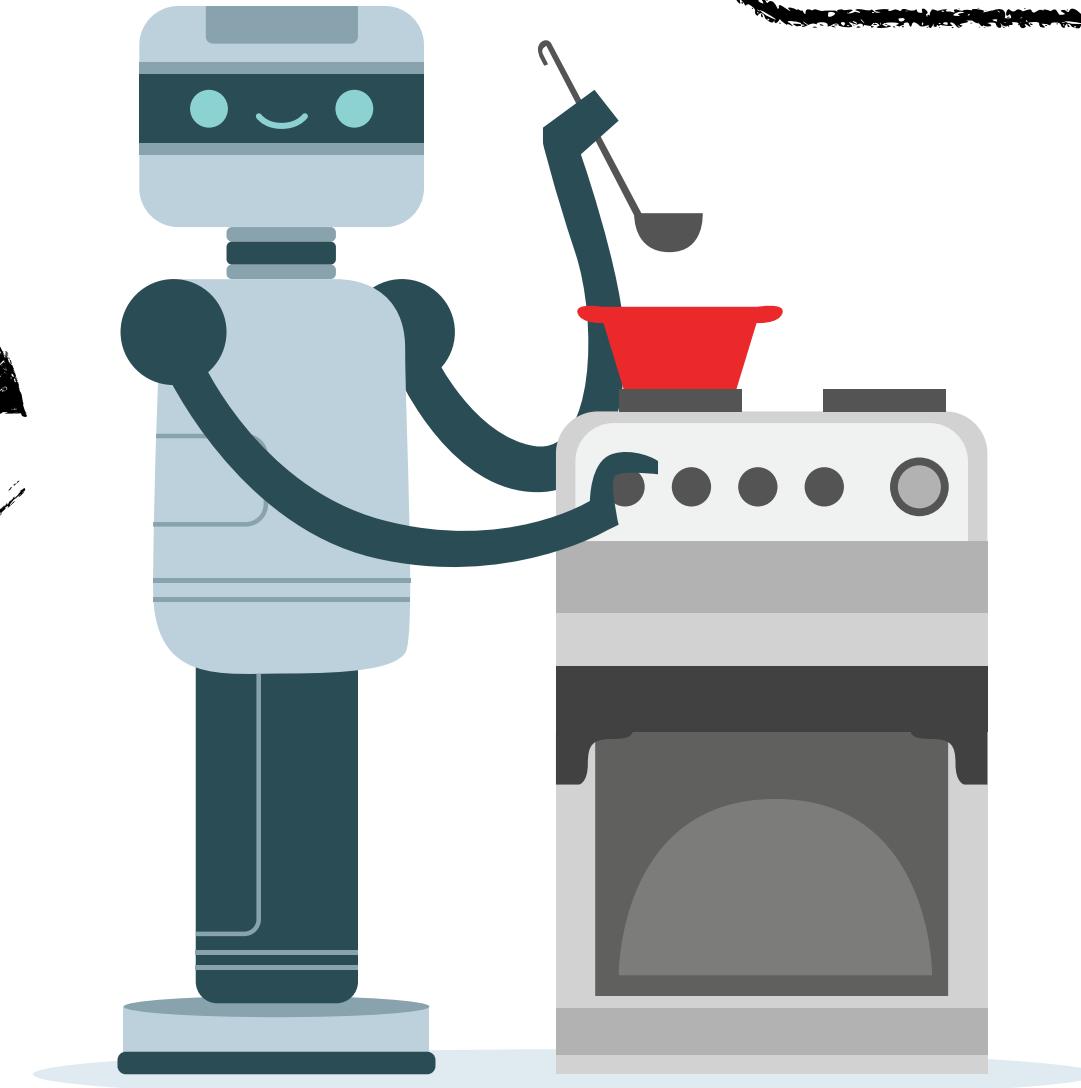
# Train home robot apprentice to help grandma!



Demonstrations,  
Language



Interactive  
feedback



# Home Robots in our lab!

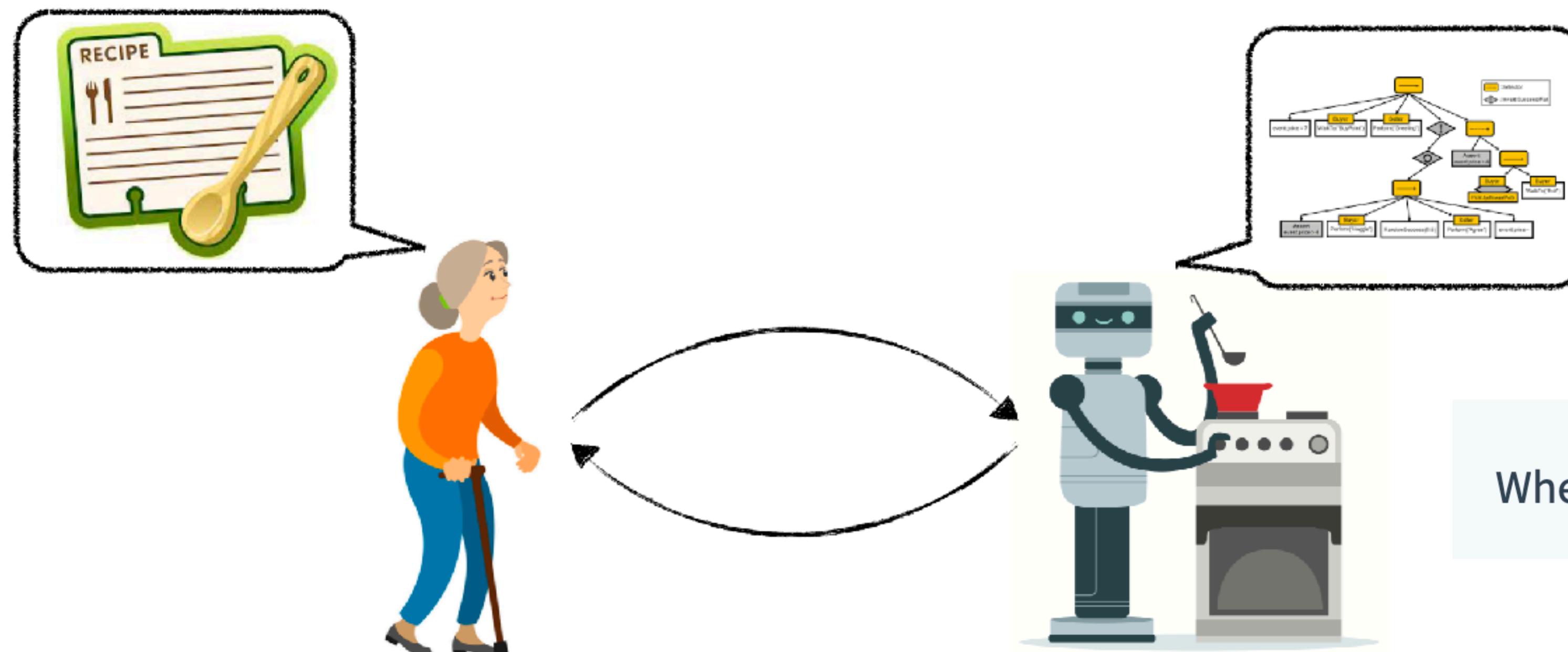


Hey MOSAIC! It's so cold today! Can you suggest some soup recipe?



# Question

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



When poll is active respond at [PollEv.com/sc2582](https://PollEv.com/sc2582)



# What is special about robot learning?

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

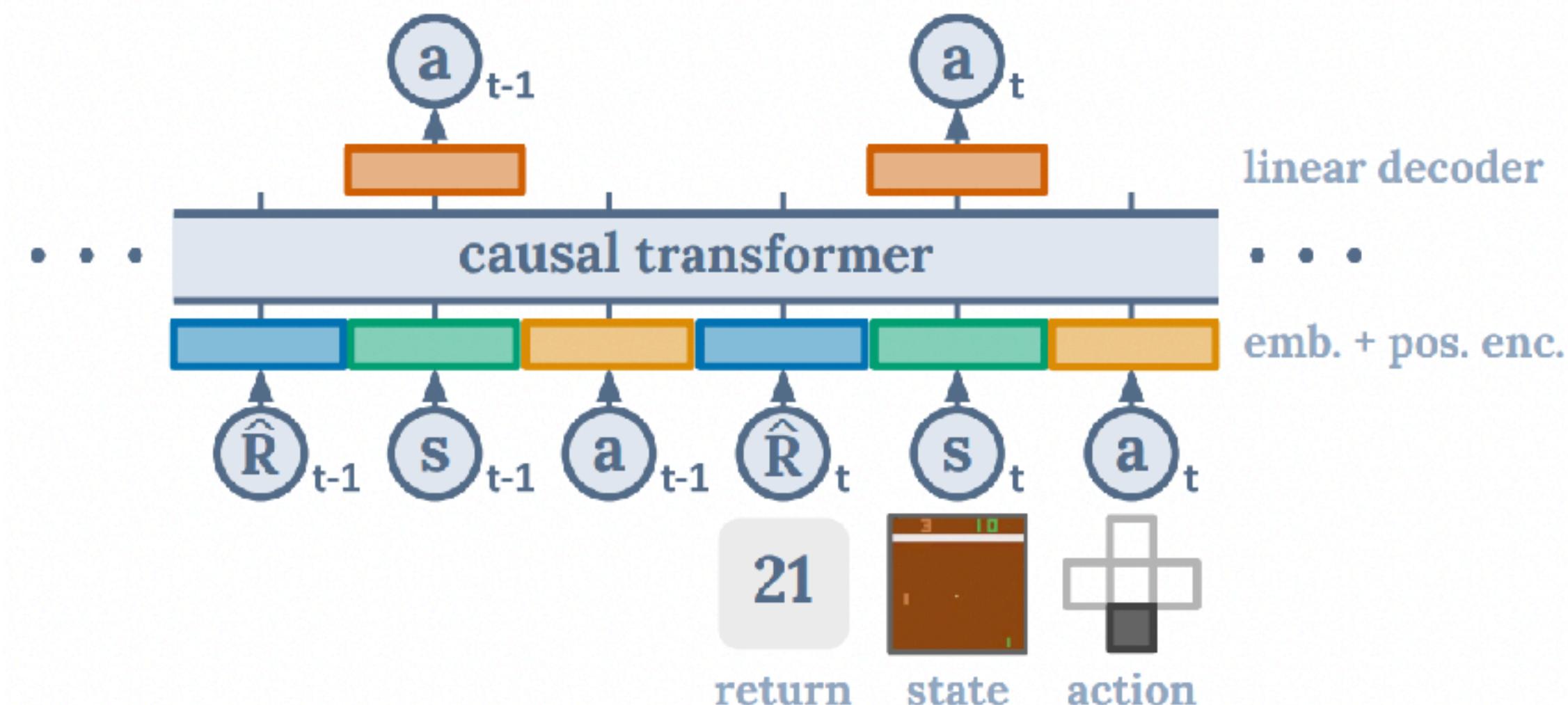
$x$  is a sequence of inputs,  $y$  is a sequence of outputs,  $\theta$  is a model

# 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,  $\theta$  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,  $\theta$  is a model

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

# What is special about robot learning?

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

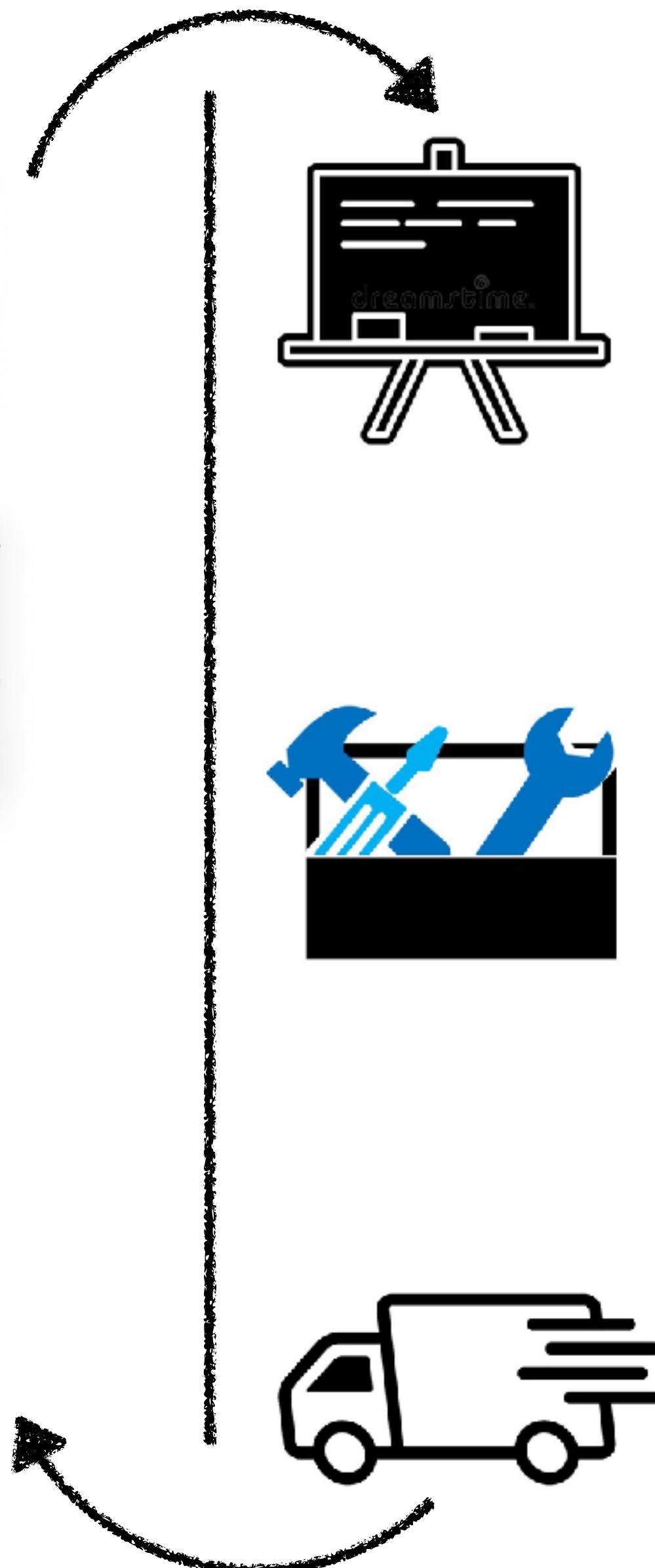
$x$  is a sequence of inputs,  $y$  is a sequence of outputs,  $\theta$  is a model

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

# WHY this course?



Take any  
robot application



**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)



“Sanjiban”

He / Him

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

*Build robots that can learn from humans!*

Undergrad



PhD



Research Engineer



PostDoc

# We are PoRTaL (People and Robots, Teaching and Learning)



# 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

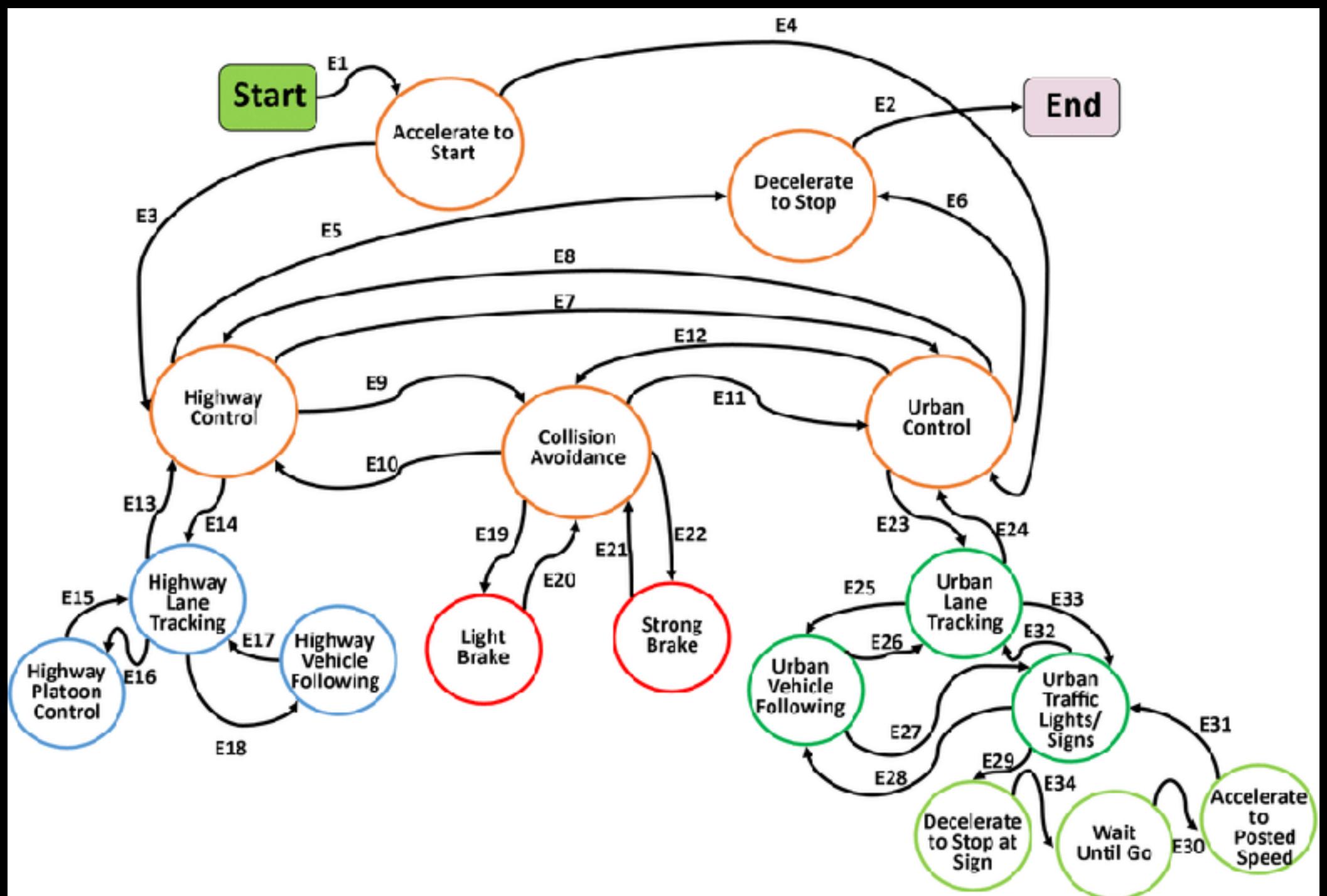


Wins the DARPA  
Urban Challenge

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

# Self-driving 1.0

Hand-engineered rules of driving



Limited use of machine learning

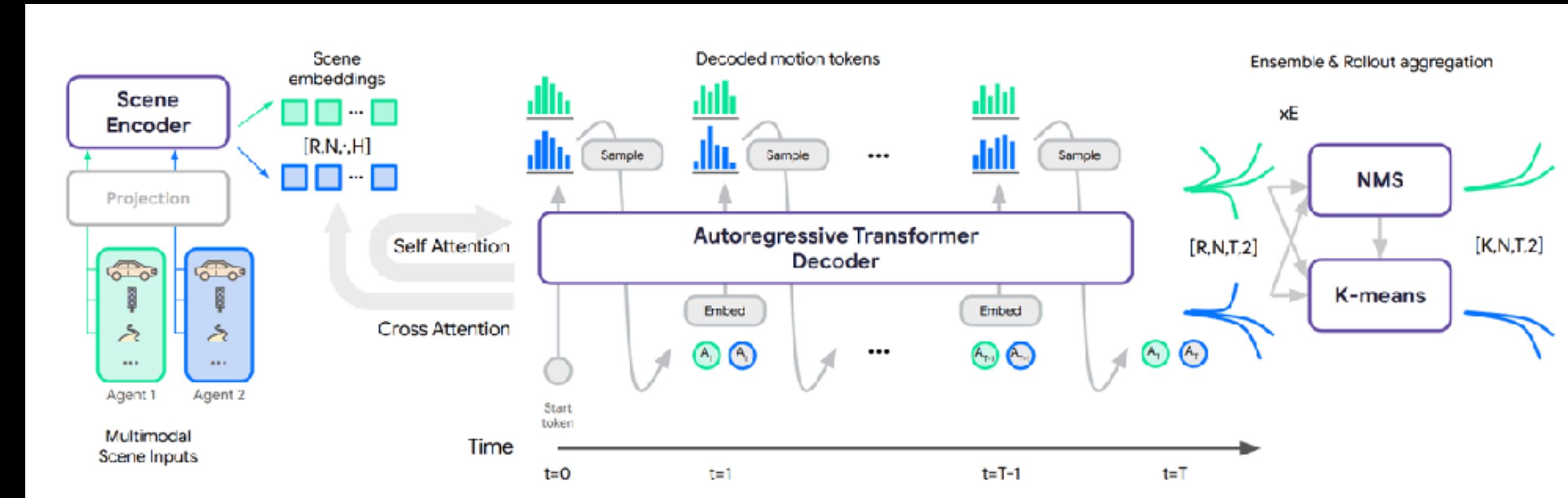
Software that fundamentally  
could not scale over time

# 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

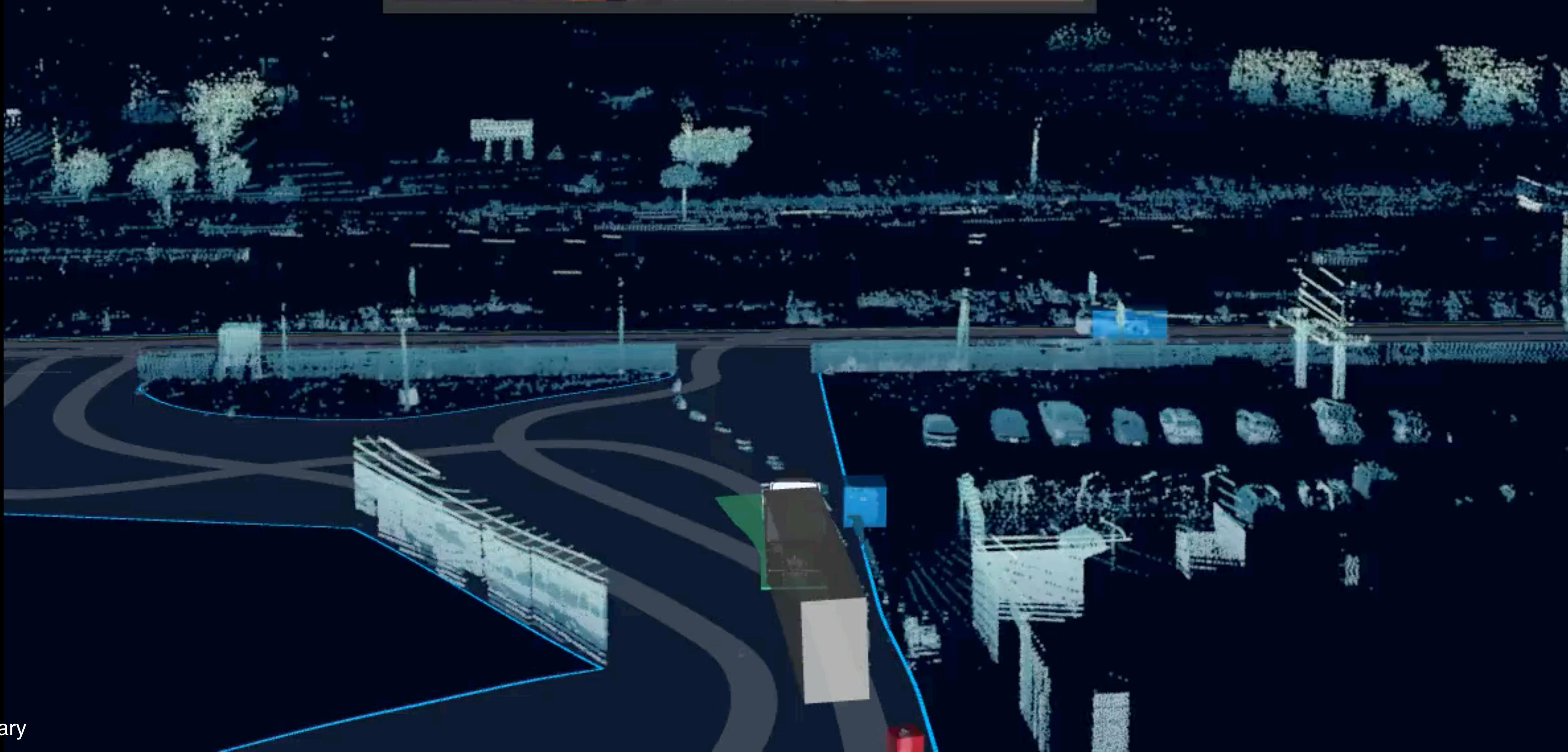
RIGHT TURN  
-10 FT

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

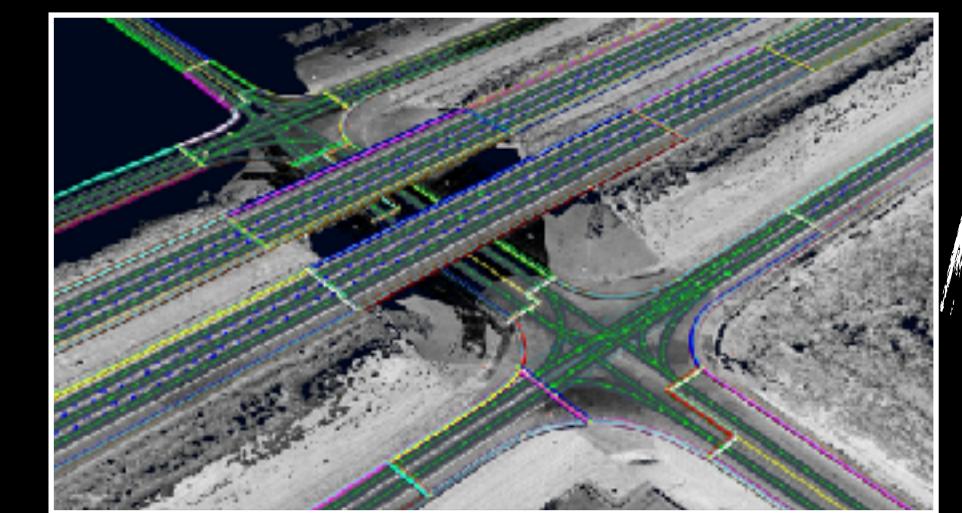
Is this a scaling issue? Should we 10x our data, have bigger models?

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?

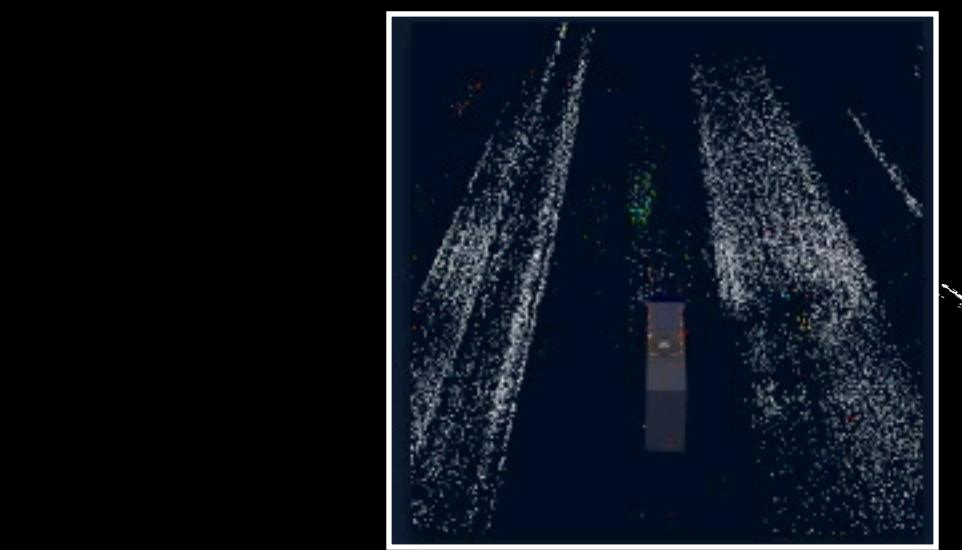
Let's dive a bit deeper



Maps



Camera

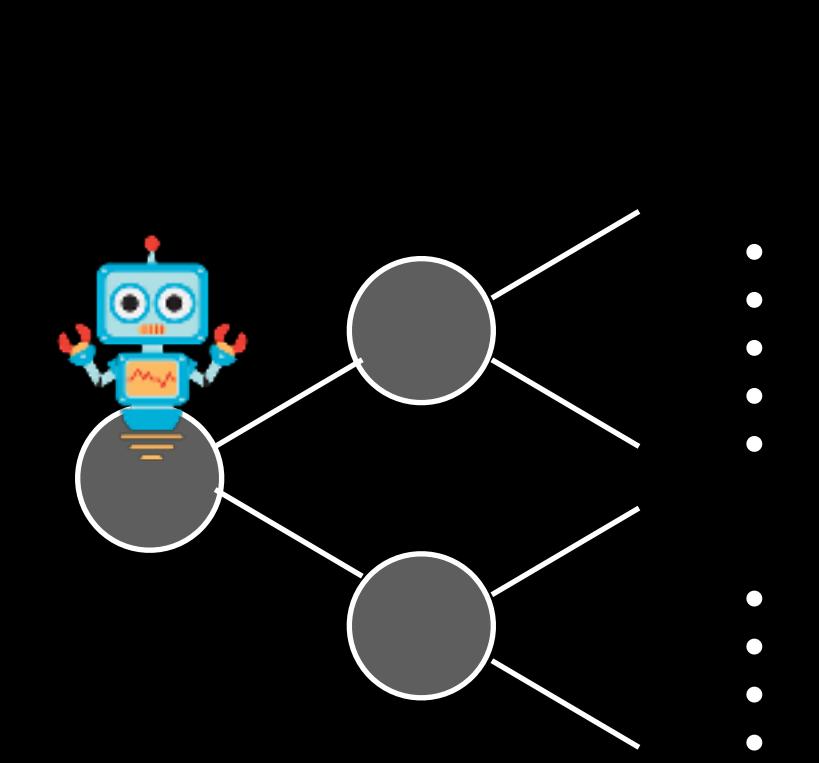
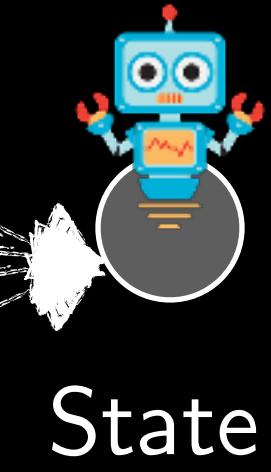


Radar

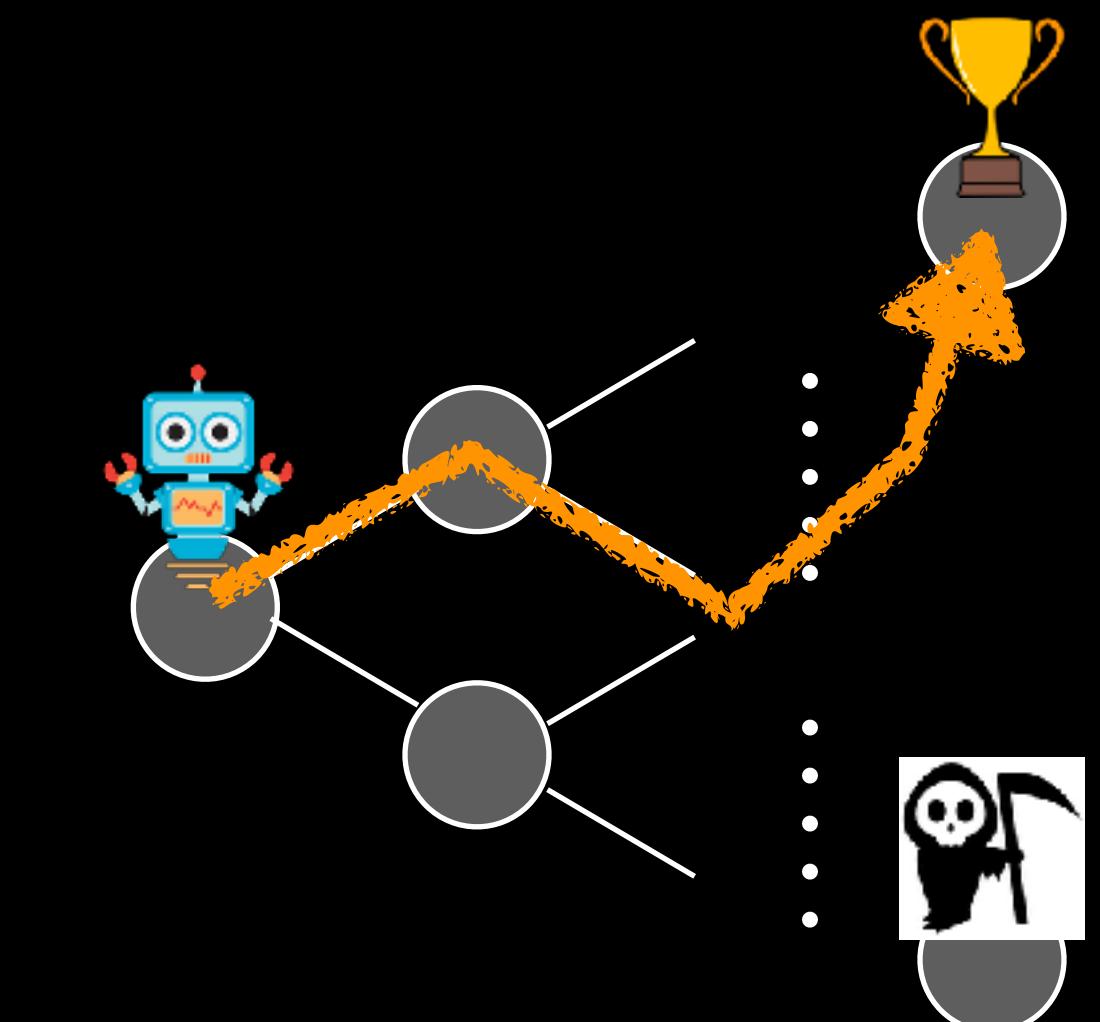


Lidar

## Perception

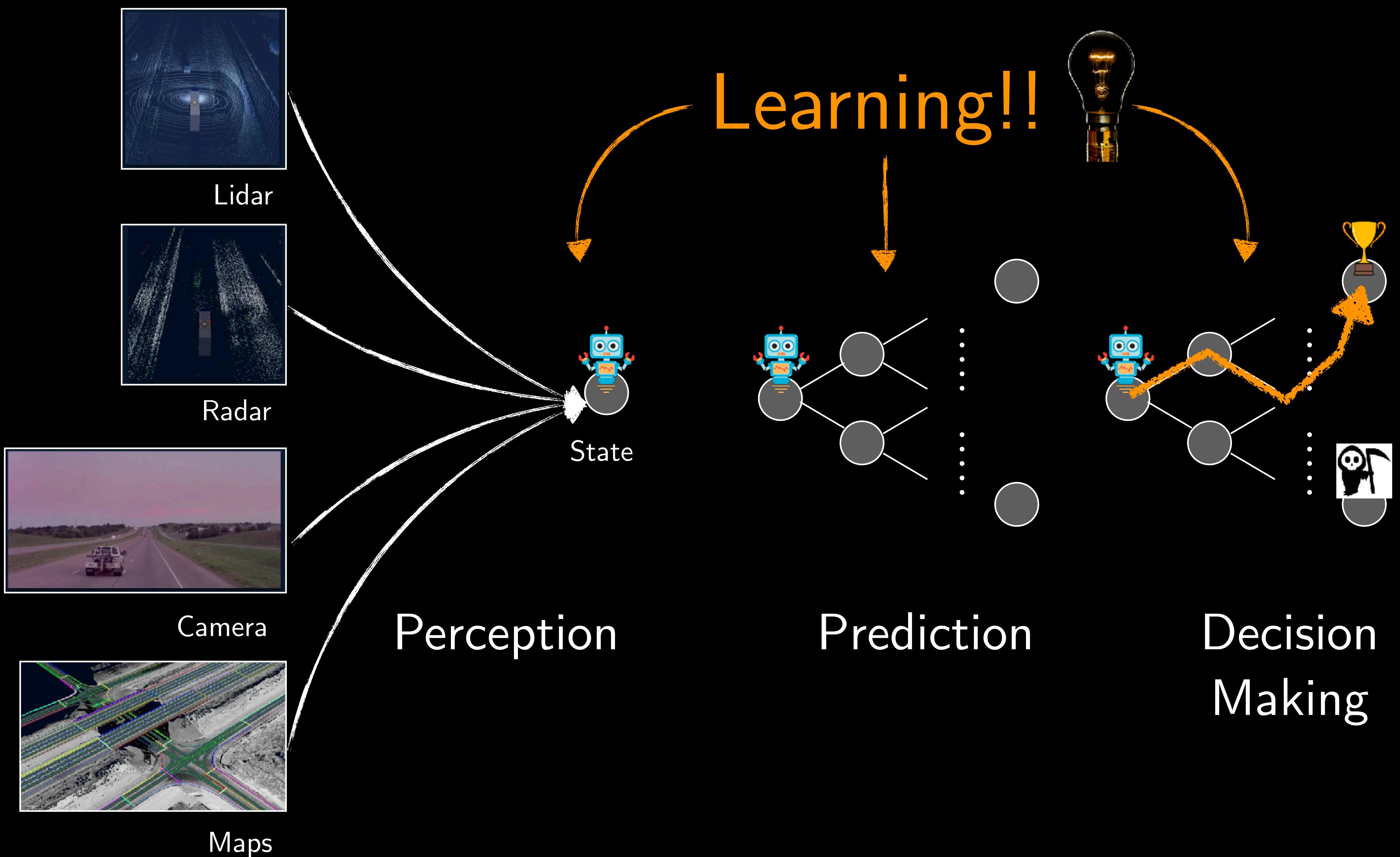


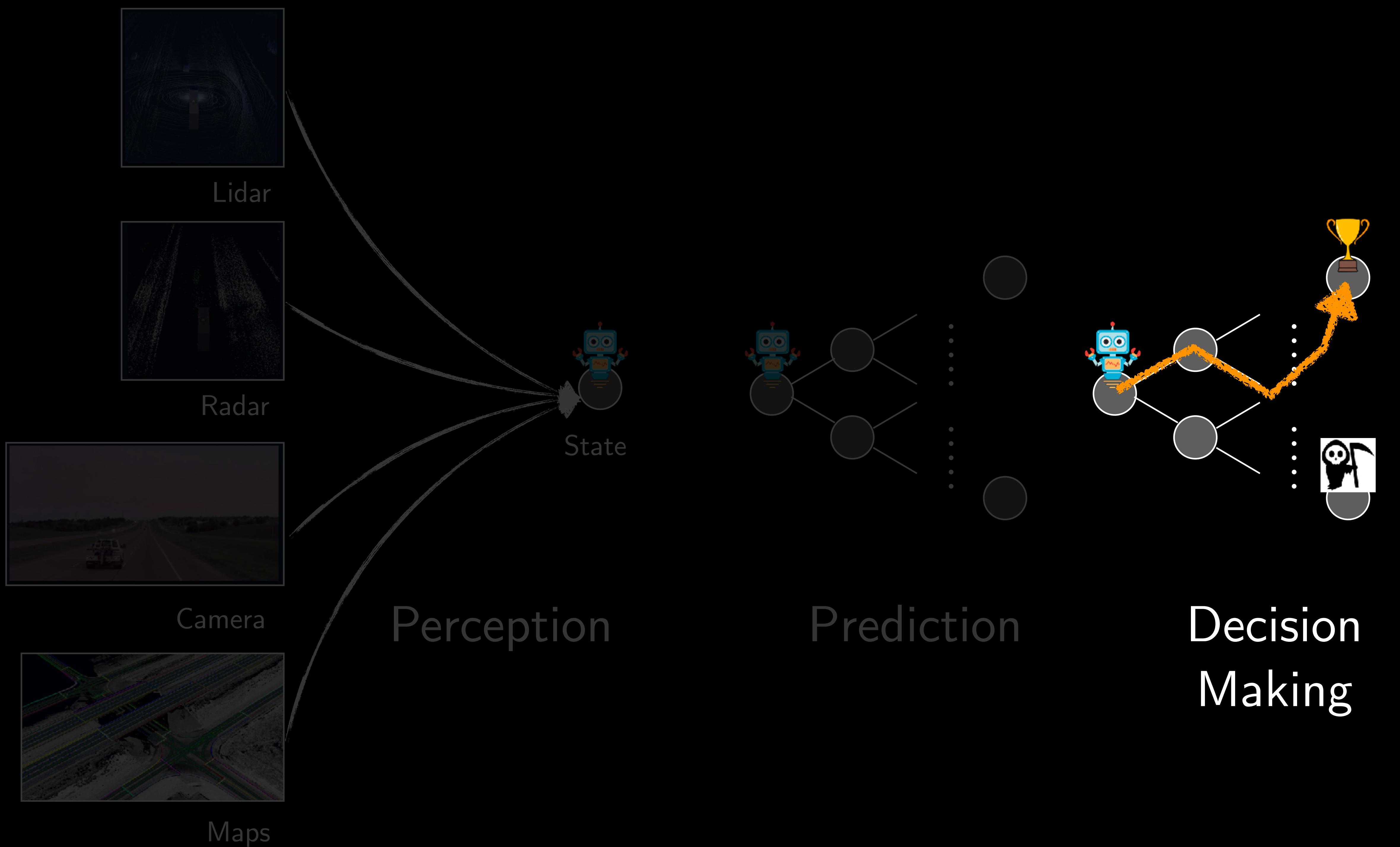
## Prediction



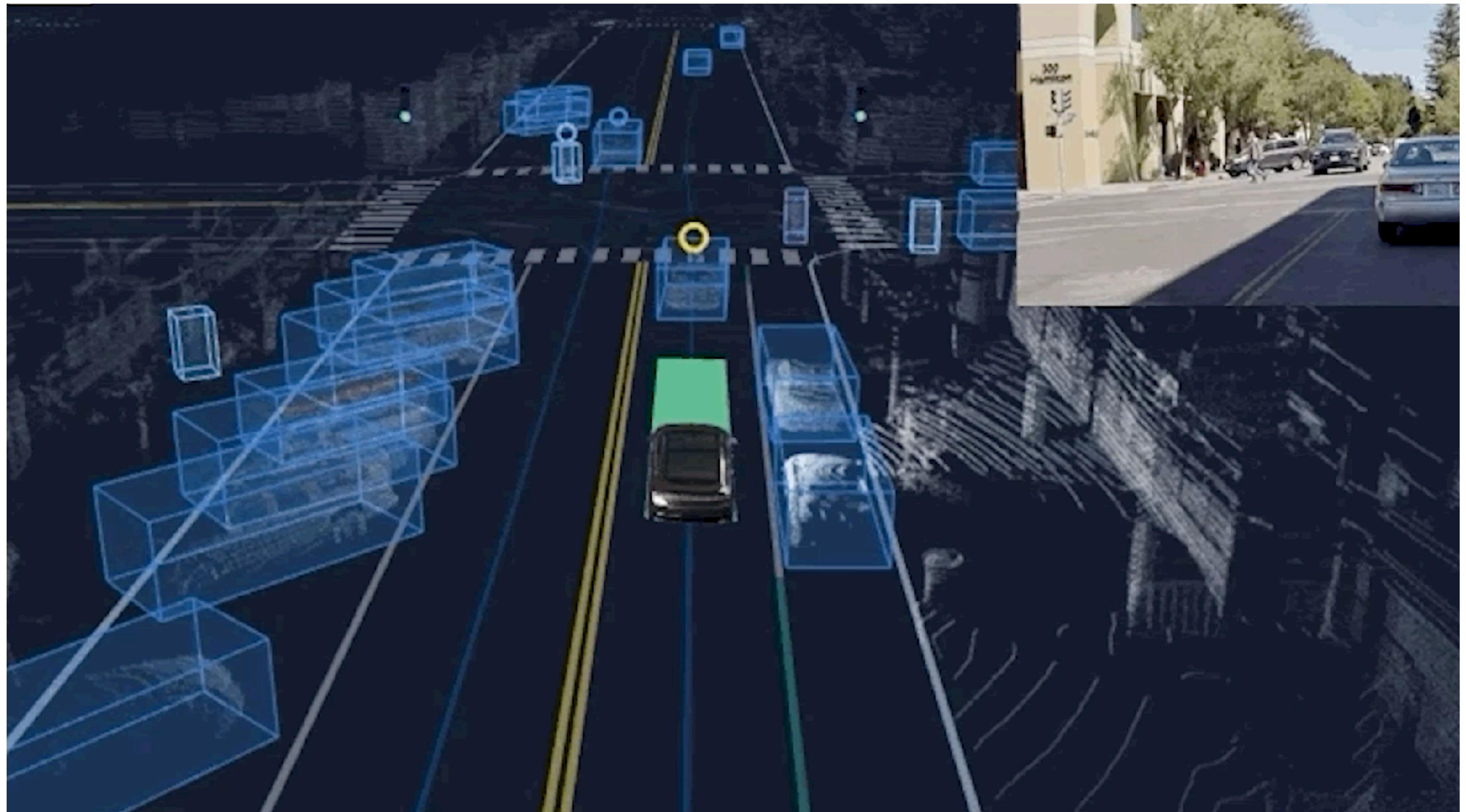
## Decision Making

# Learning!!





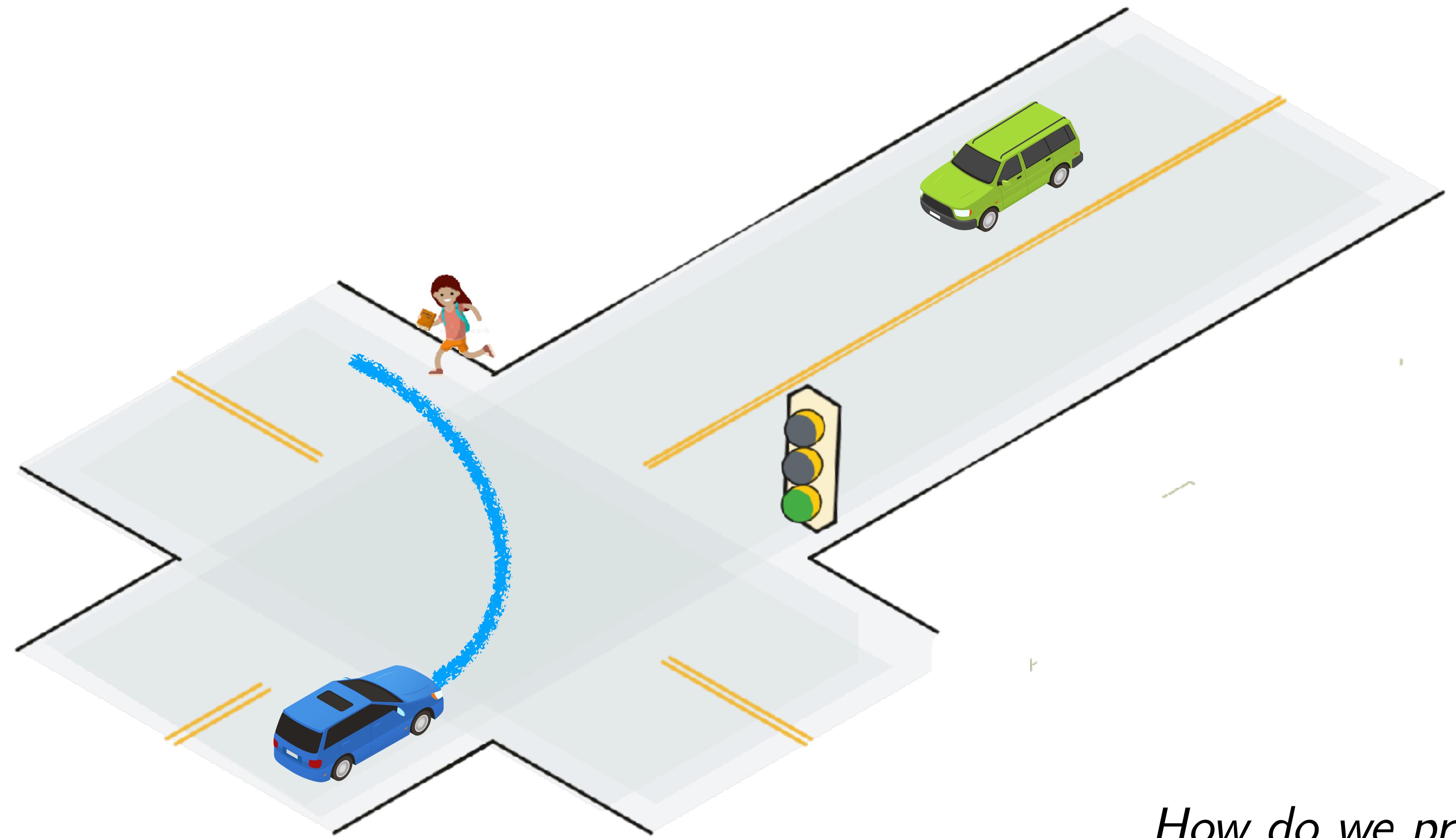
# Activity: What is “good” behavior in a left turn?



# Activity!



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

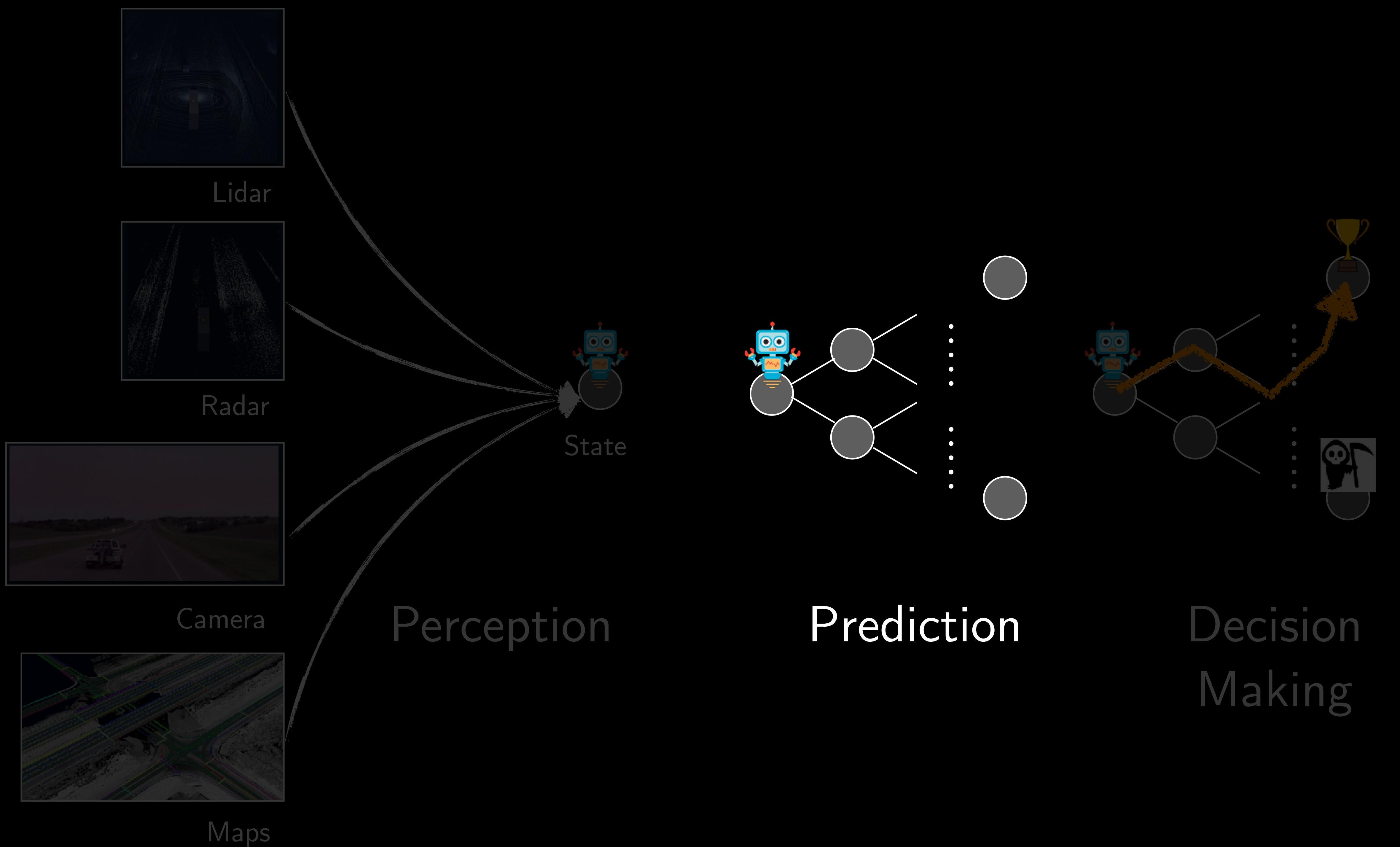


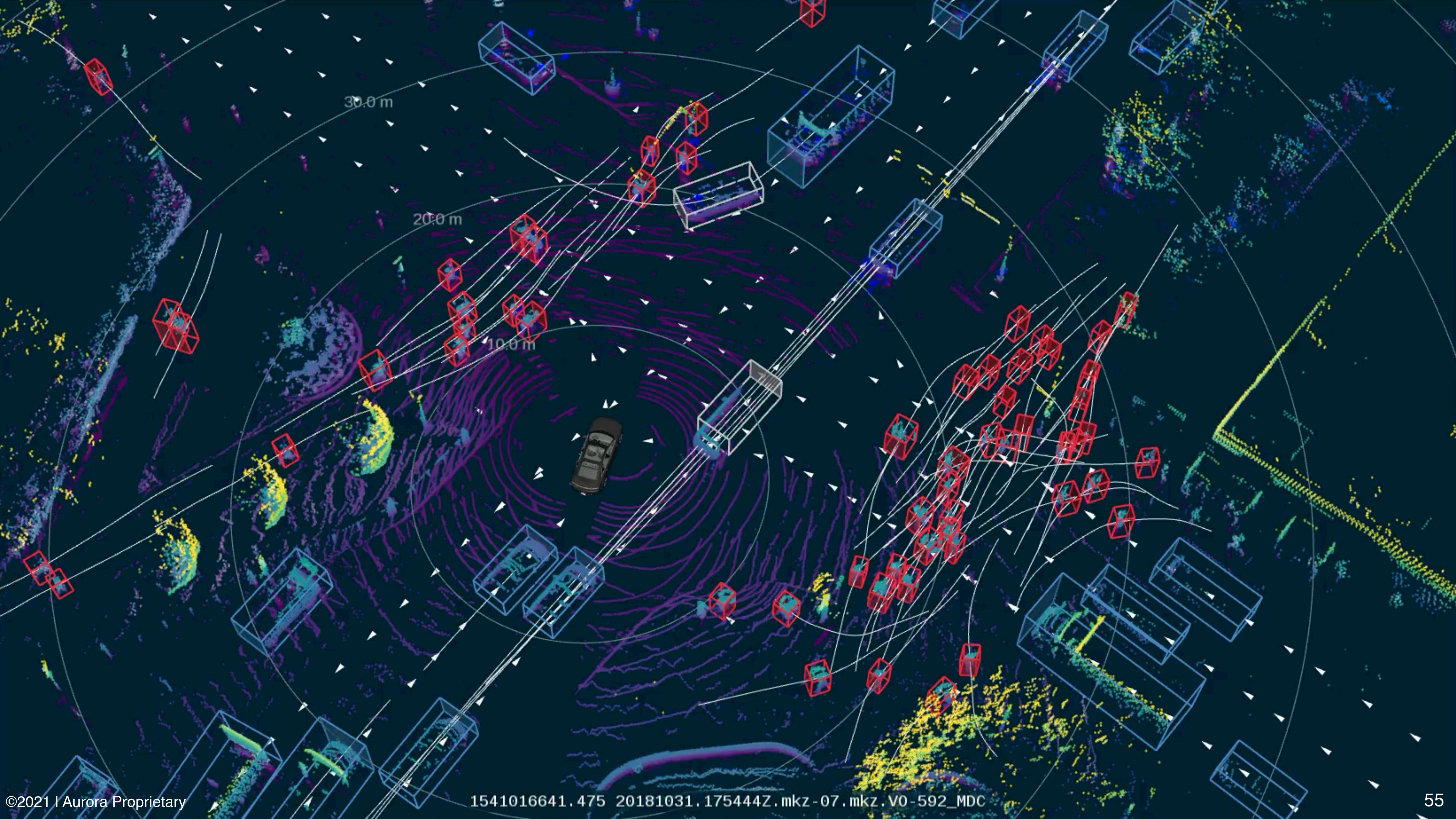
*How do we program in these values?*

# Lesson #1

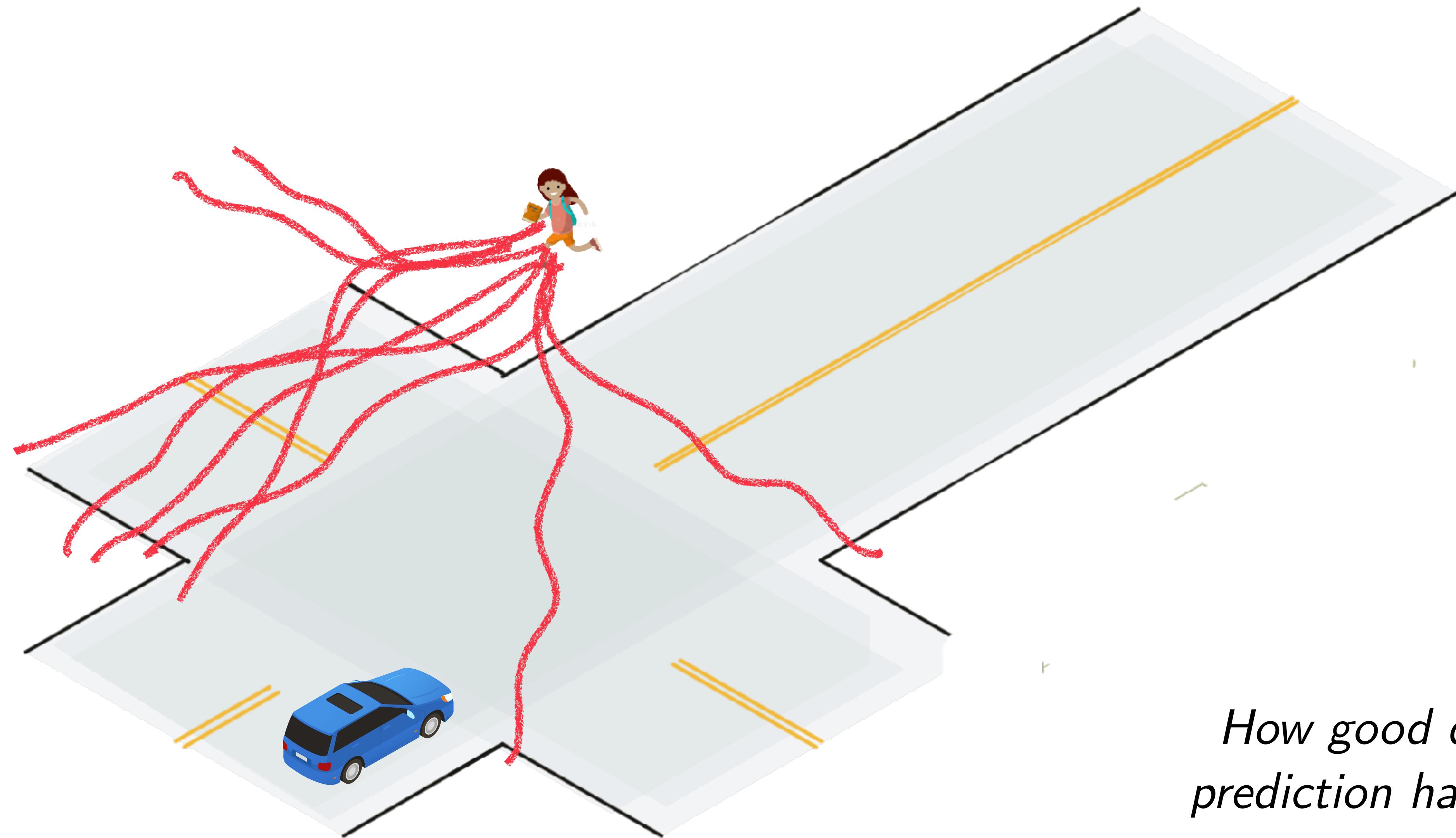
Values are implicit in human driving!







# Activity: How can we predict pedestrian motion?



*How good does the prediction have to be?*

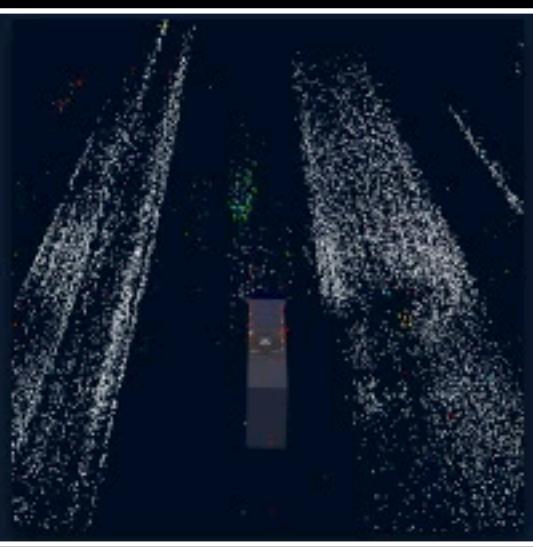
# Lesson #2

Models are useful fictions





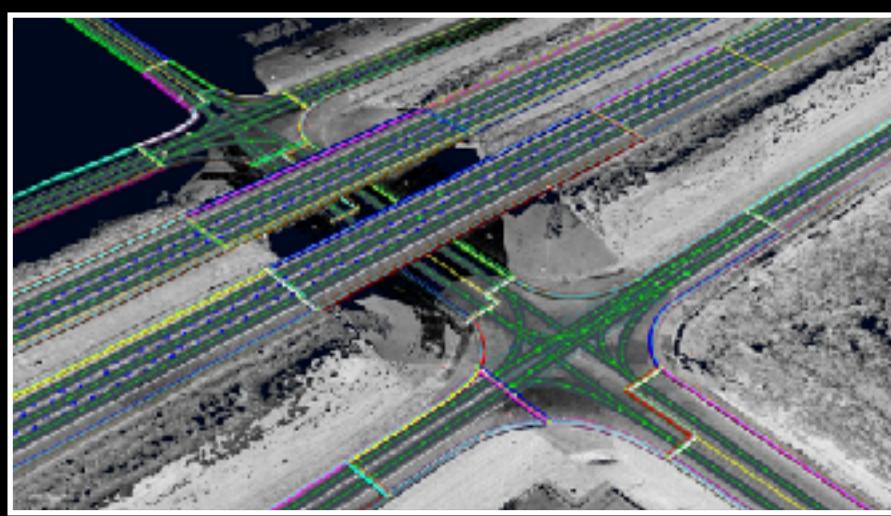
Lidar



Radar

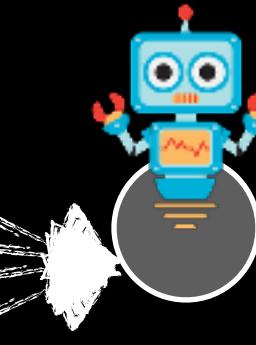


Camera

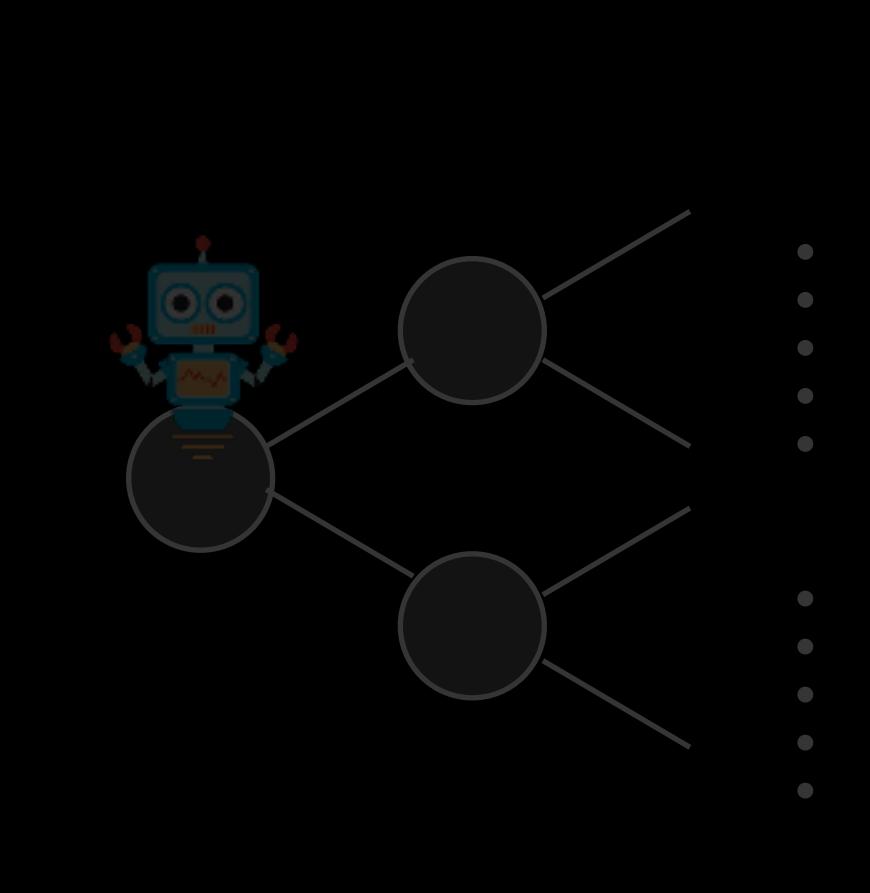


Maps

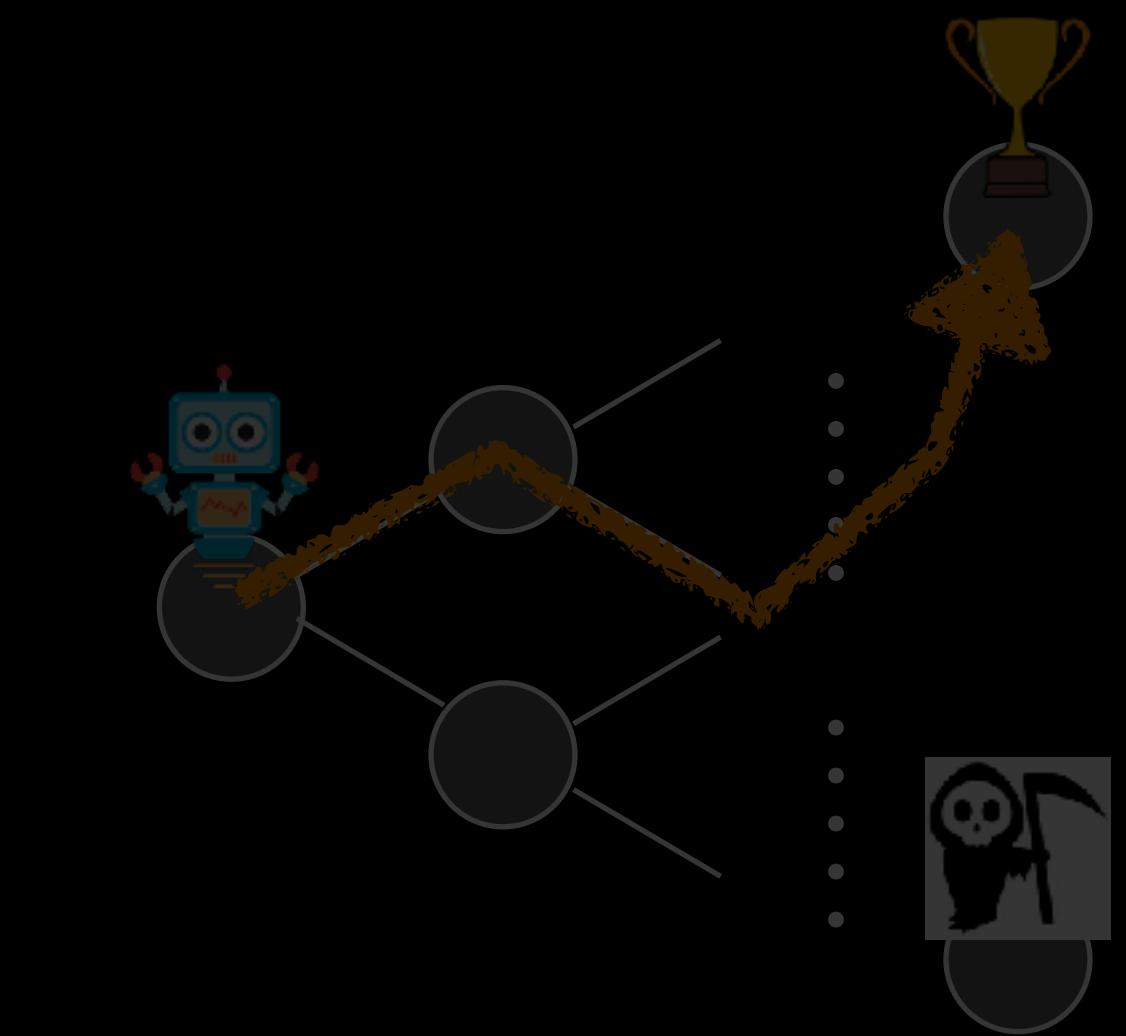
## Perception



State



## Prediction



## Decision Making

# No one sensor tells the whole story!



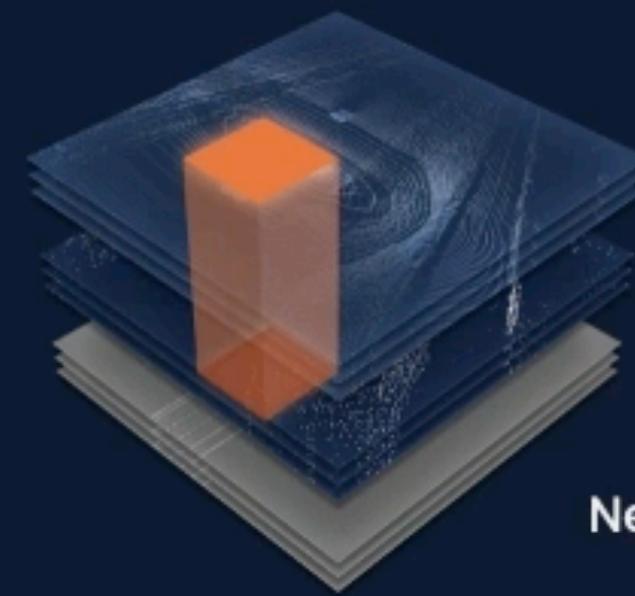
LiDAR



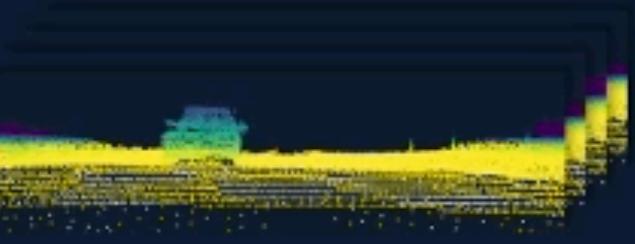
Radar



Camera



Neural Convolution Engine

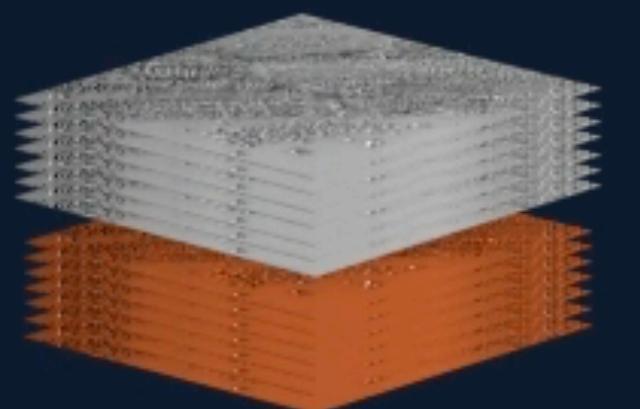


Range Conv. Engine

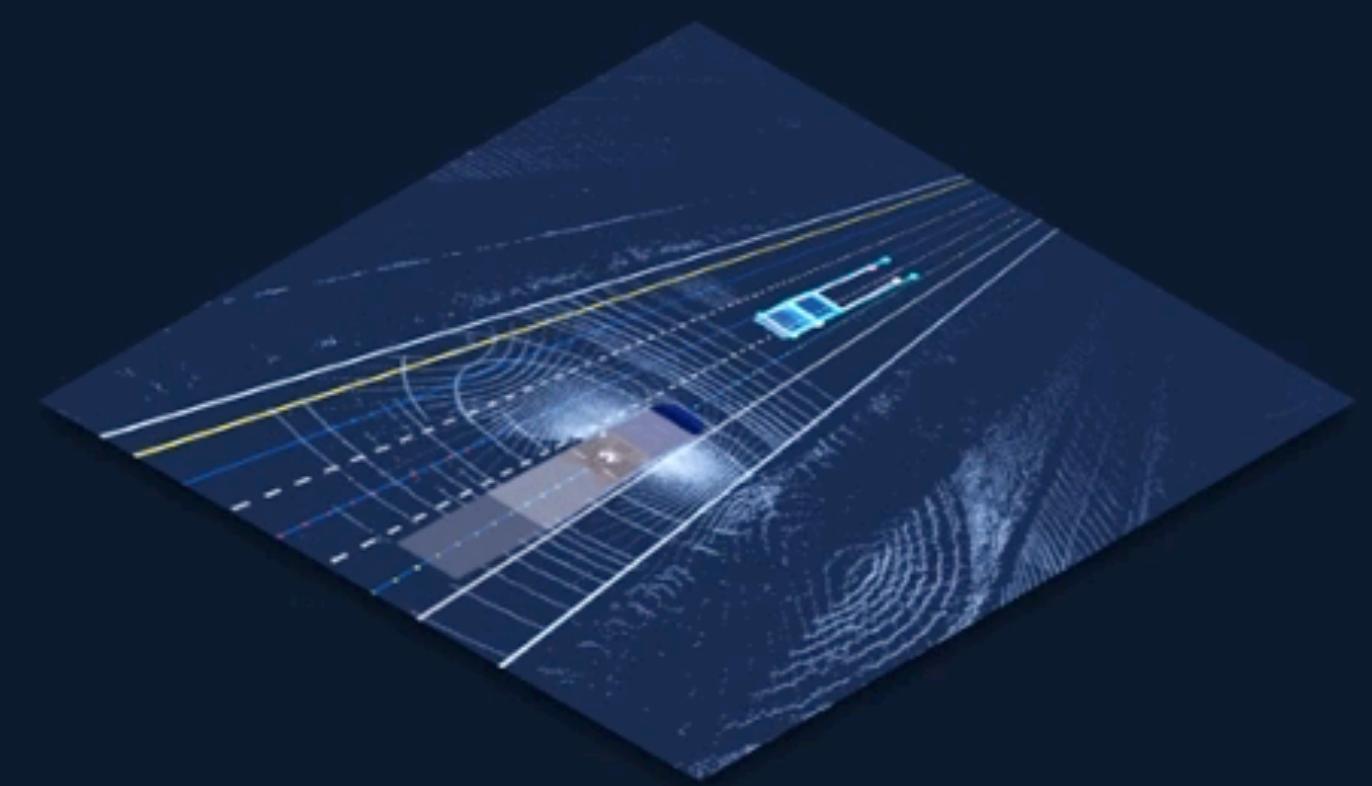


Image Fusion Engine

Euclidian Ray  
Scatter Engine

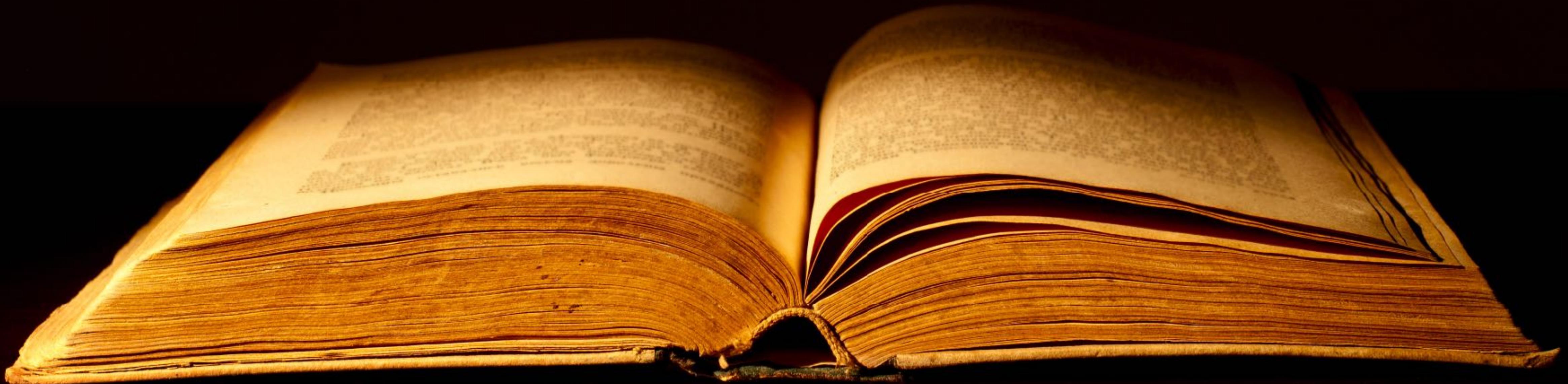


Neural Convolution  
Engine



# Lesson #3

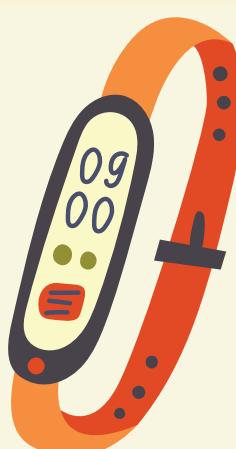
Solve for the state  
that explains all observations



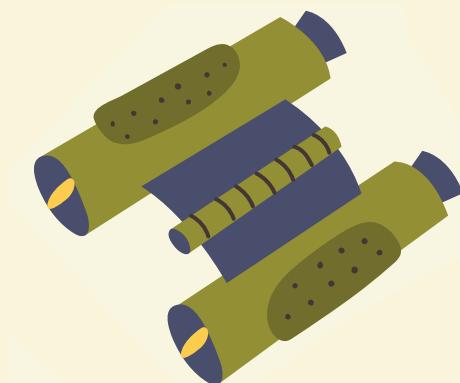
The  
journey  
ahead!



Frontiers

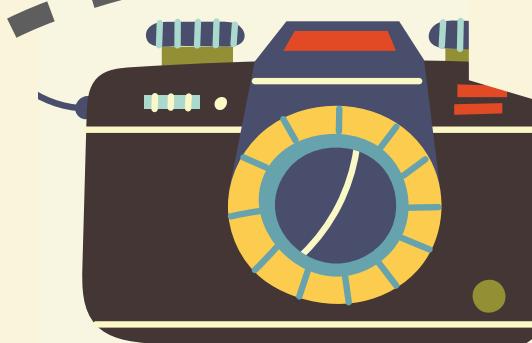


World Models  
& Forecasting

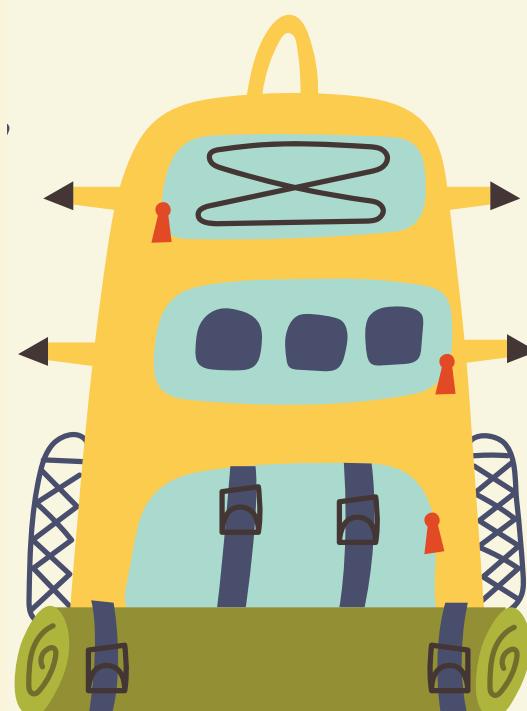


Multi-modal  
Models

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		<a href="#">The Bitter Lesson</a>
08/29/24	Robots as Markov Decision Problems	<a href="#">MACRL Ch. 1</a>	Dan Klein <a href="#">slides I</a>

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.



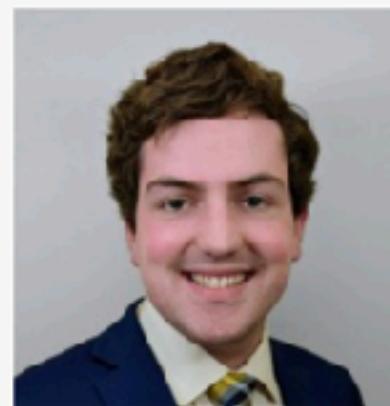
# The Crew



## Sanjiban Choudhury

Instructor

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[Zach Garcia](#)

Teaching Assistant

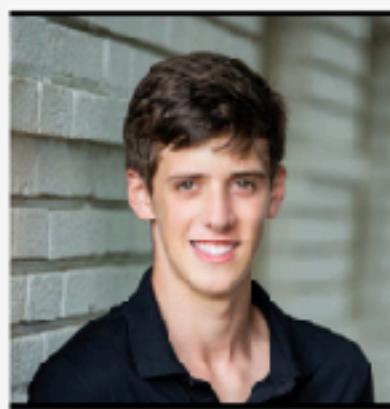
[zag7@cornell.edu](mailto:zag7@cornell.edu)



[Ved Sriraman](#)

Teaching Assistant

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[Zubin Bhaumik](#)

Teaching Assistant

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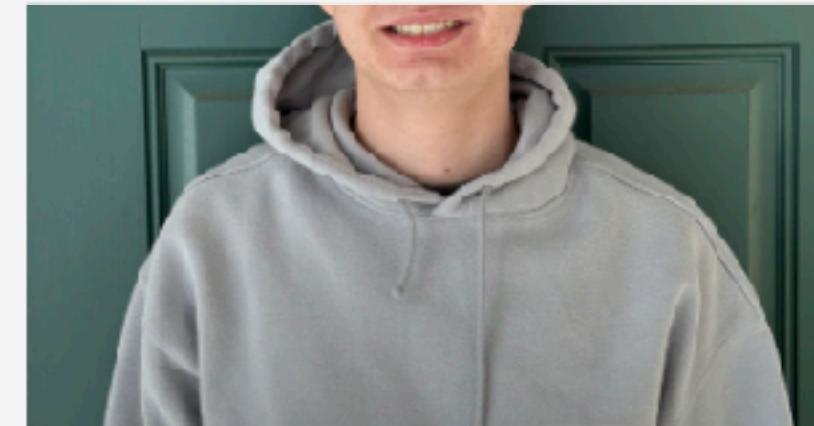
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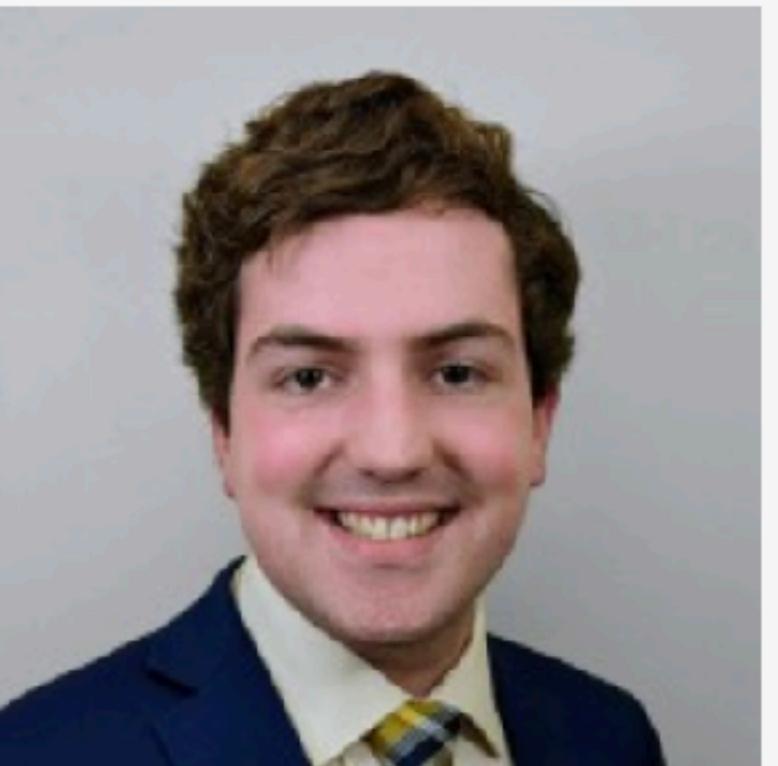


## Qian Meng

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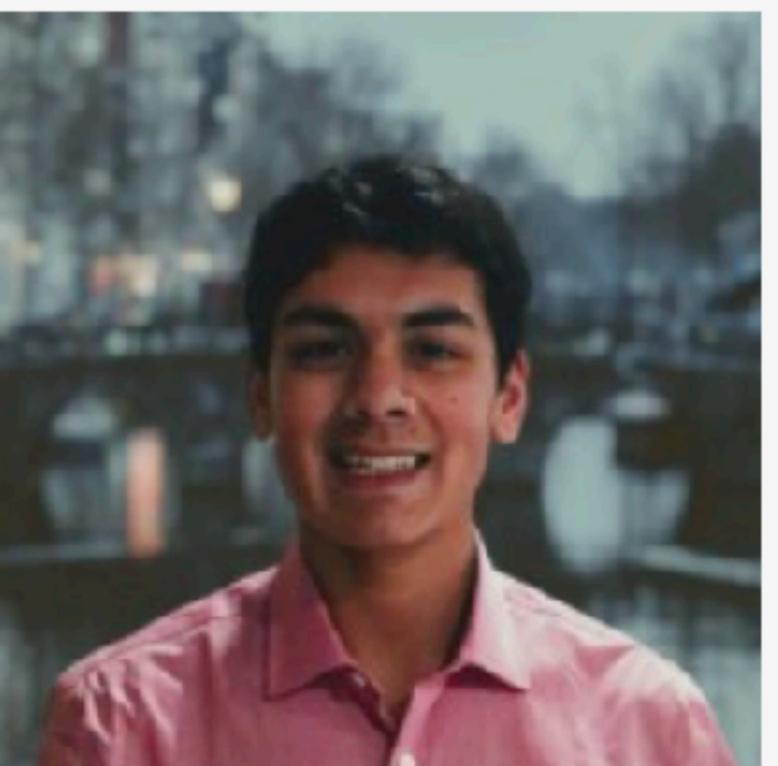




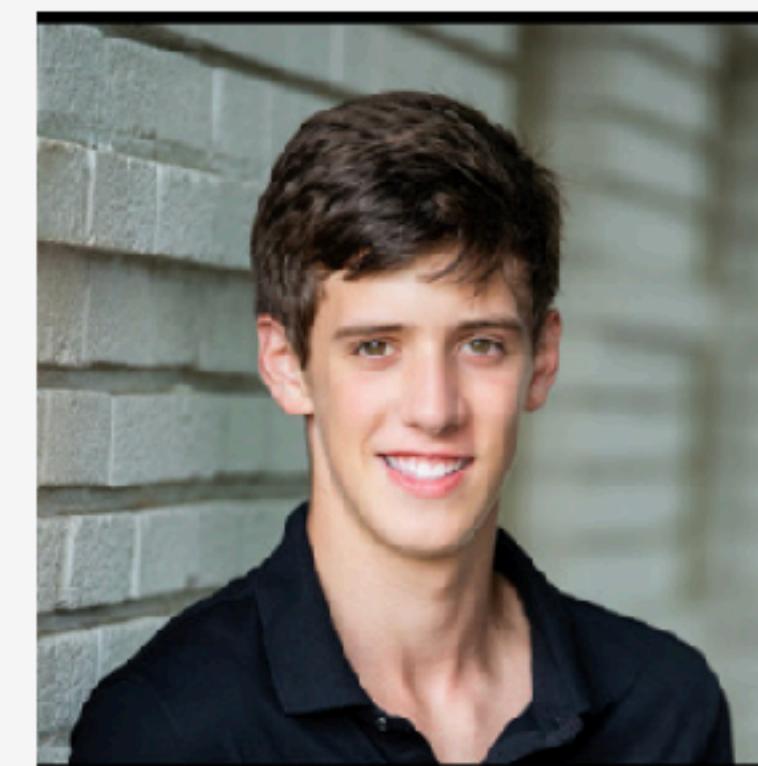
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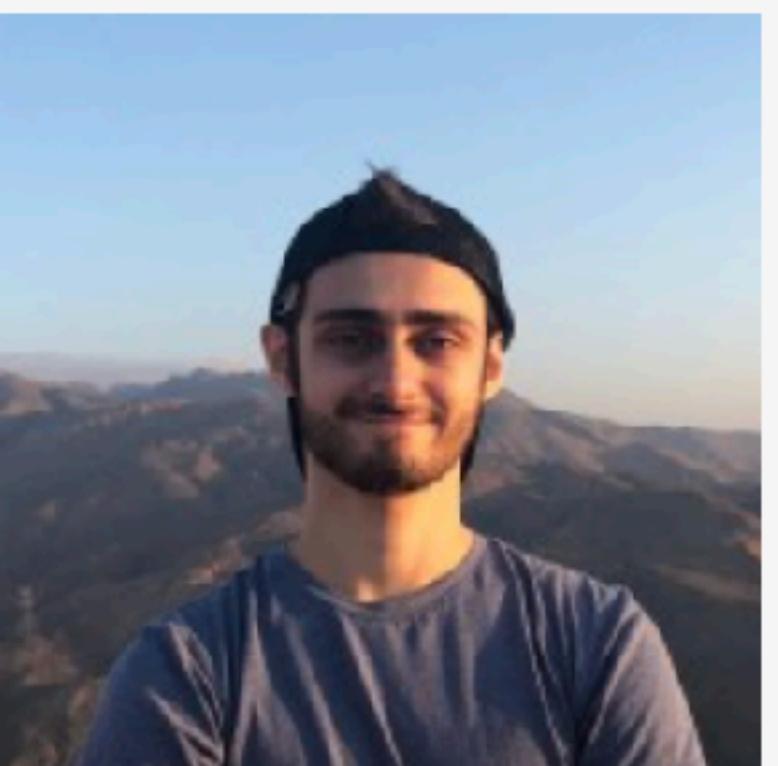
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# Wait list info

- 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!