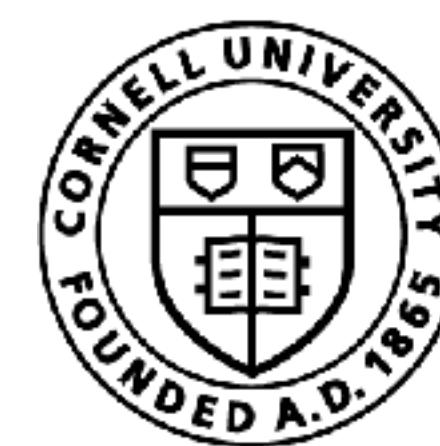


Imitation Learning: Feedback and Covariate Shift

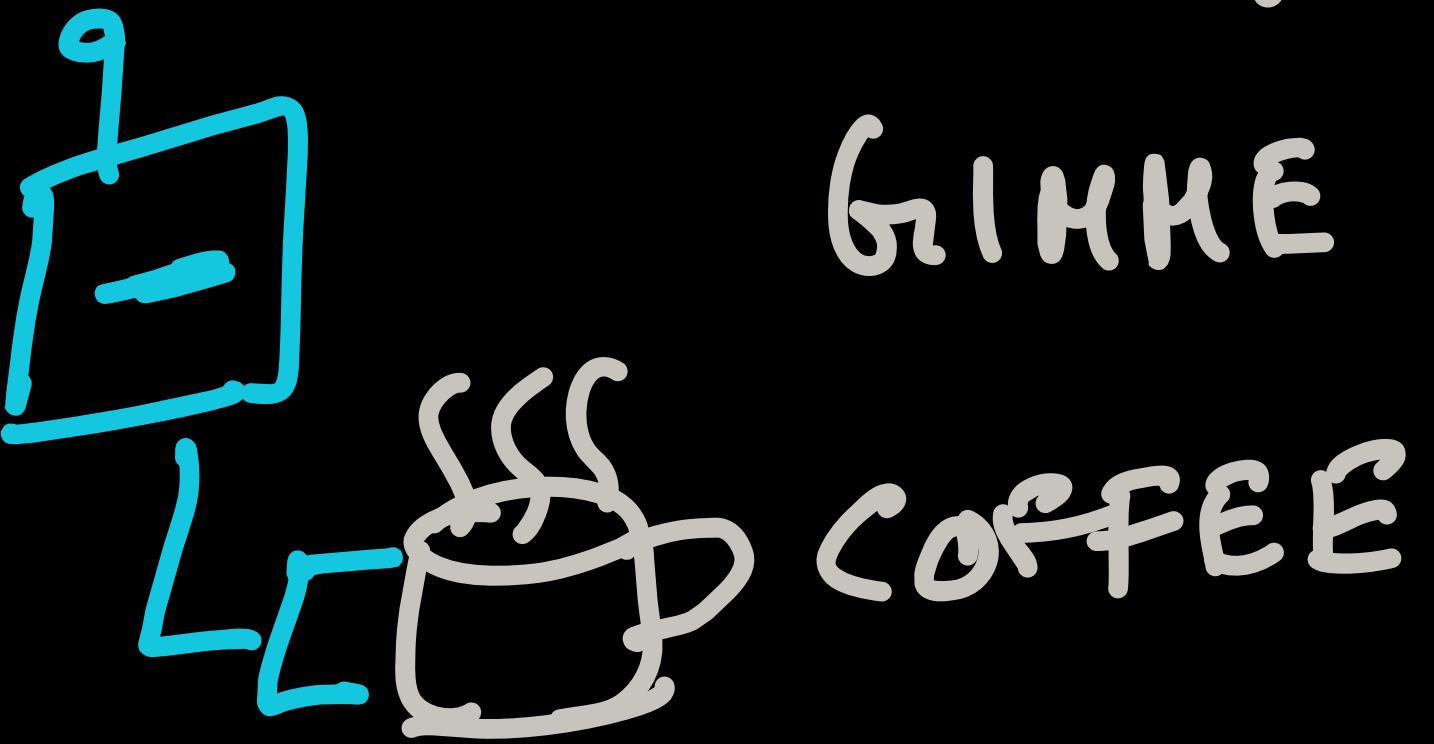
Sanjiban Choudhury



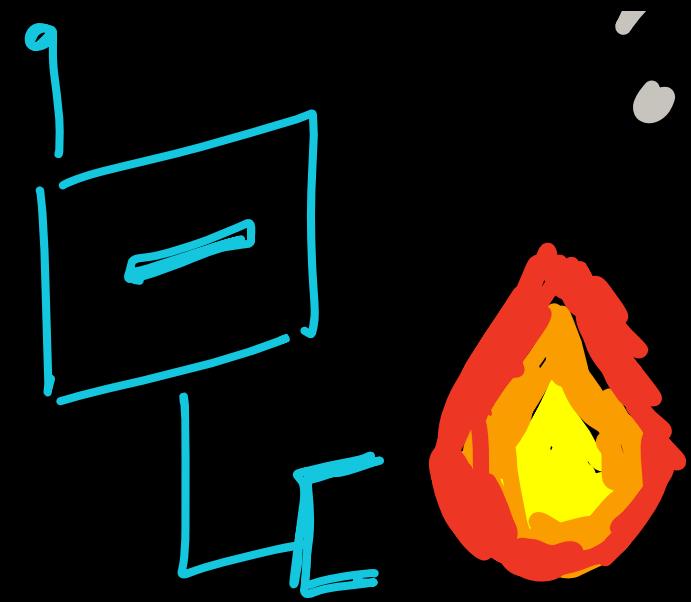
Cornell Bowers CIS
Computer Science

Programming a task ...

tell the robot to make coffee ..

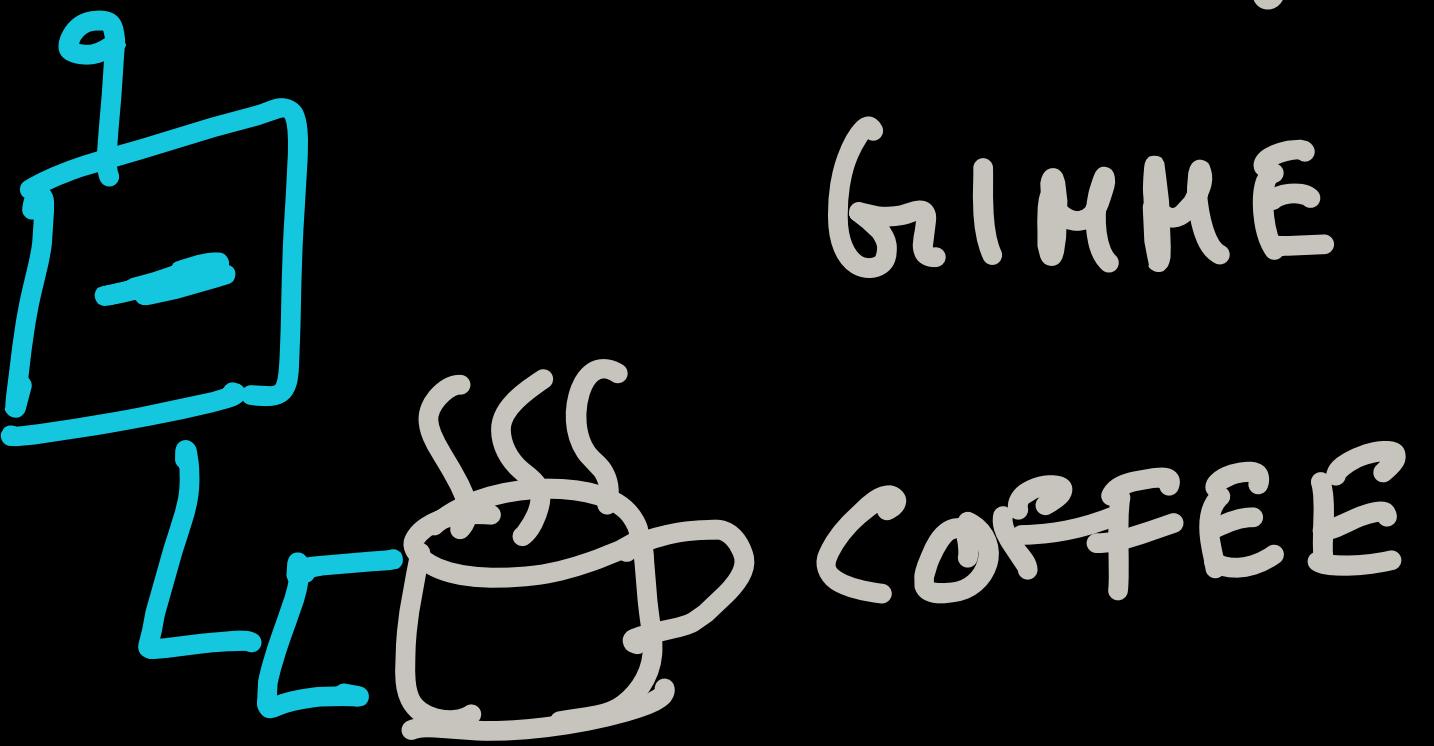


robot burns down
the house!



Programming a task ...

tell the robot to make coffee ..



DON'T ...
burn down the house
steal the neighbors coffee
don't make a mess

⋮



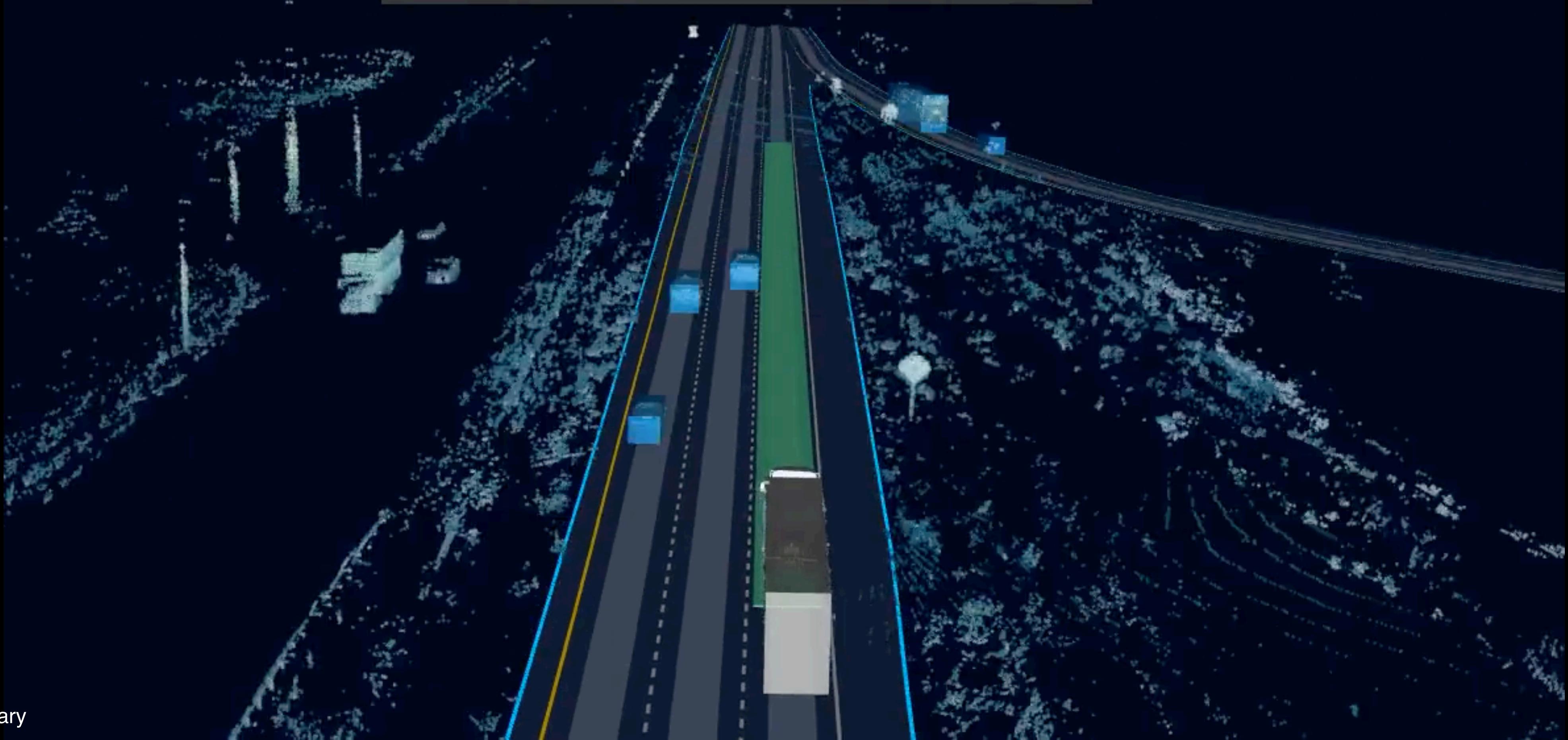
STRAIGHT
2.1 MI

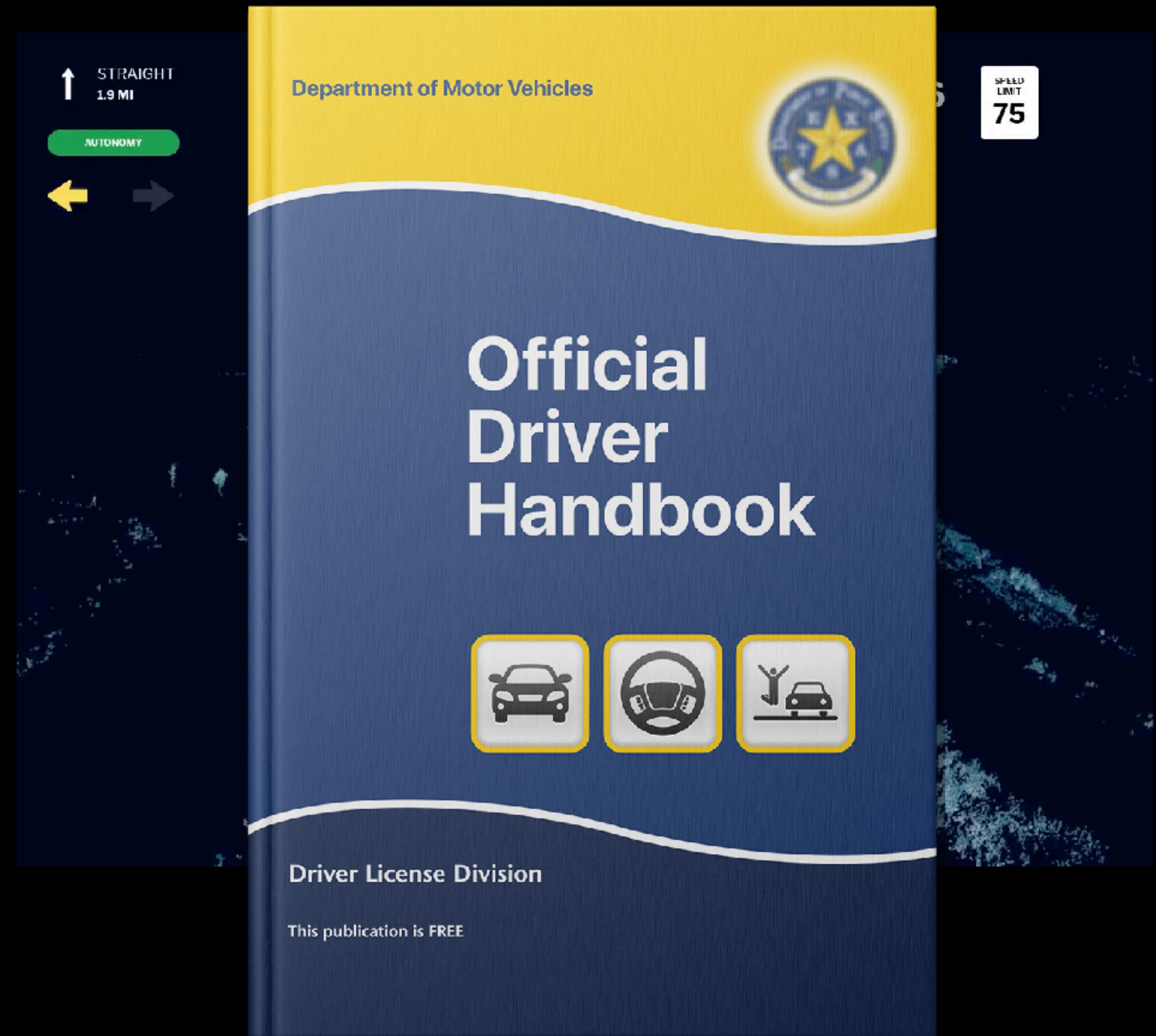
AUTONOMY



65
MPH

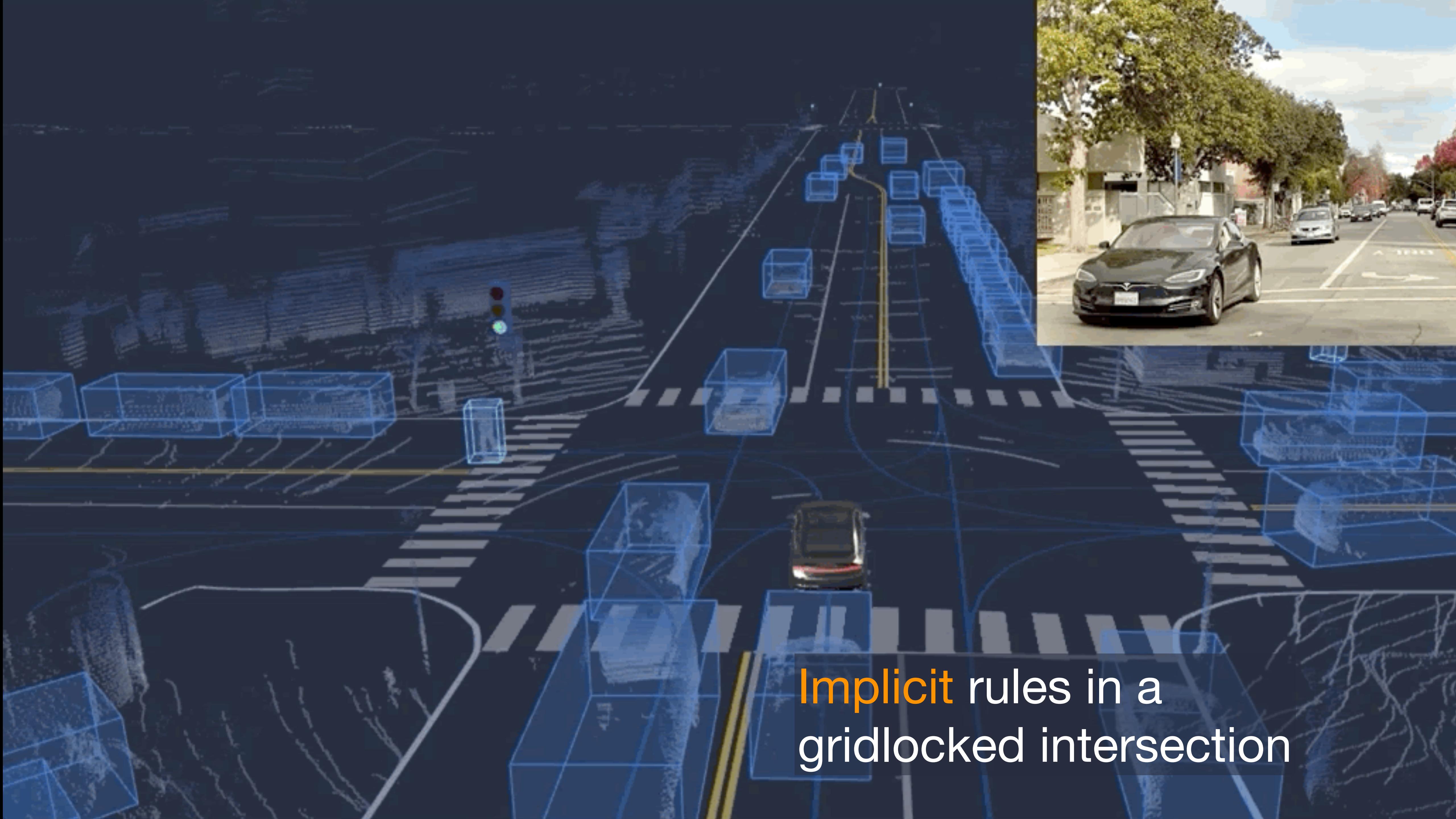
SPEED
LIMIT
75





The implicit rules of human driving





Implicit rules in a
gridlocked intersection



Explicitly programming
rules may be tedious ...

... but rules are implicit
in how we drive everyday!



Implicitly program robots
via
imitation learning

Imitation learning is everywhere

Helicopter Aerobatics



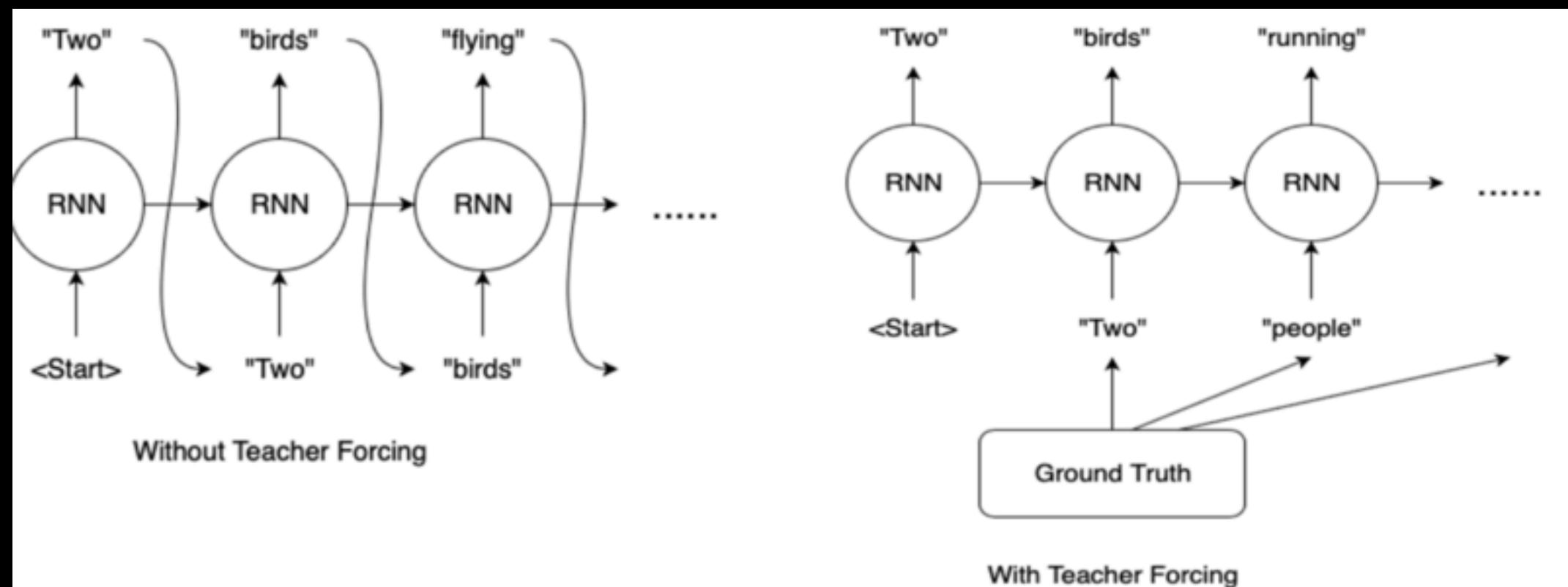
Abbeel et al. 2009



Game AI

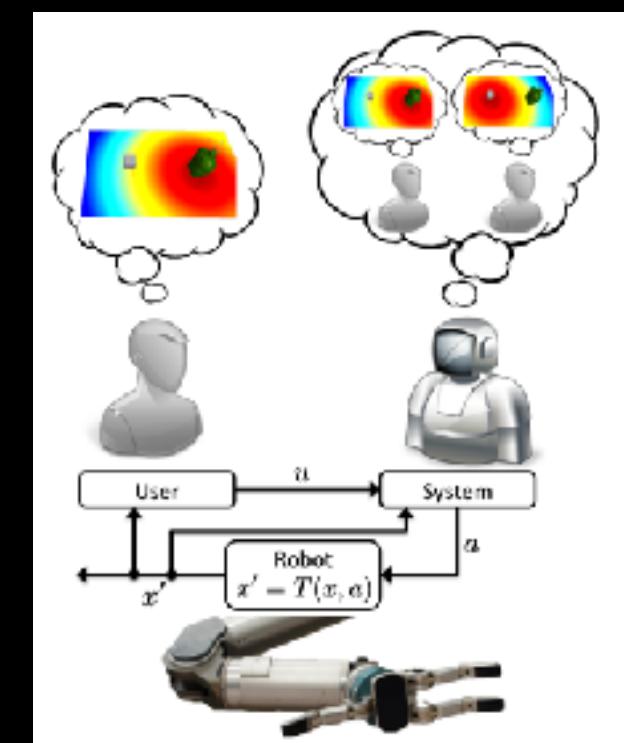
Kozik et al. 2021

Sequence models in NLP



Daume et al. 2009

Shared autonomy



Javdani et al. 2015

Activity!

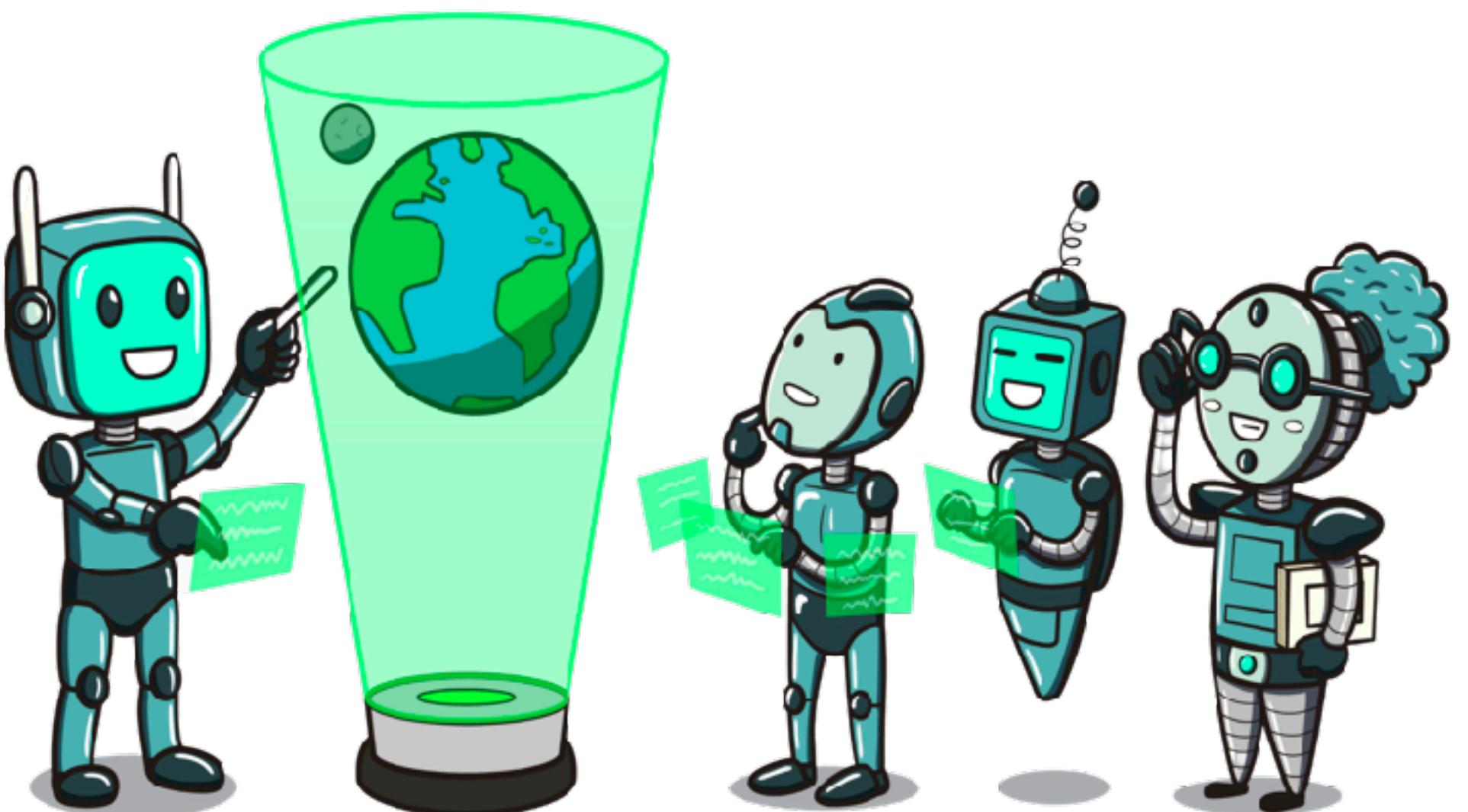


Think-Pair-Share!

Think (30 sec): What are the various ways to give feedback to a robot to teach it a new task?

Pair: Find a partner

Share (45 sec): Partners exchange ideas



Myths about Imitation Learning

- ✗ Imitation learning: Do exactly what the human will do
- ✗ Imitation learning requires humans to demonstrate actions
- ✗ Imitation learning is a way to warm start reinforcement learning
- ✗ Imitation learning means you can't do better than the human

Two Core Ideas

Data

*“What is the distribution
of states?”*

Loss

*“What is the
metric to match to
human?”*

Two Core Ideas

Data

*“What is the distribution
of states?”*

Loss

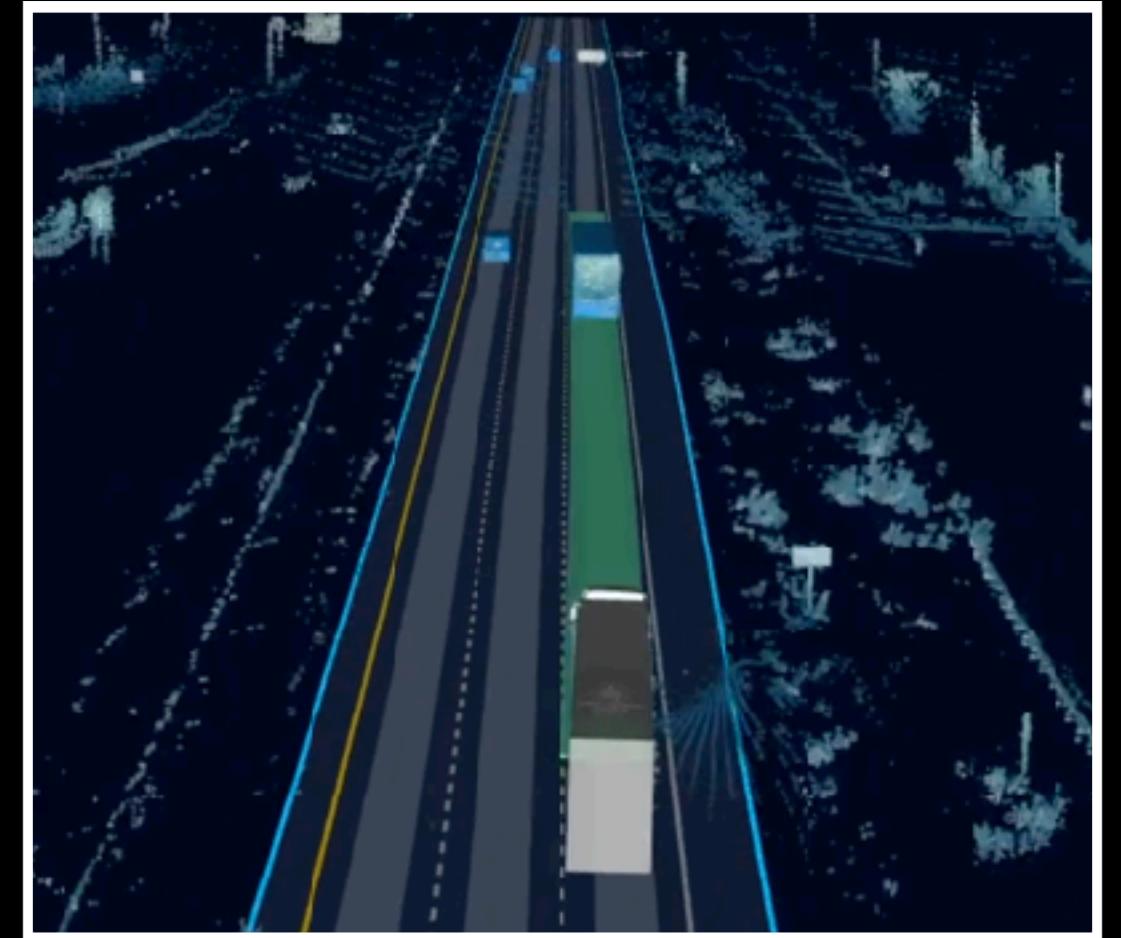
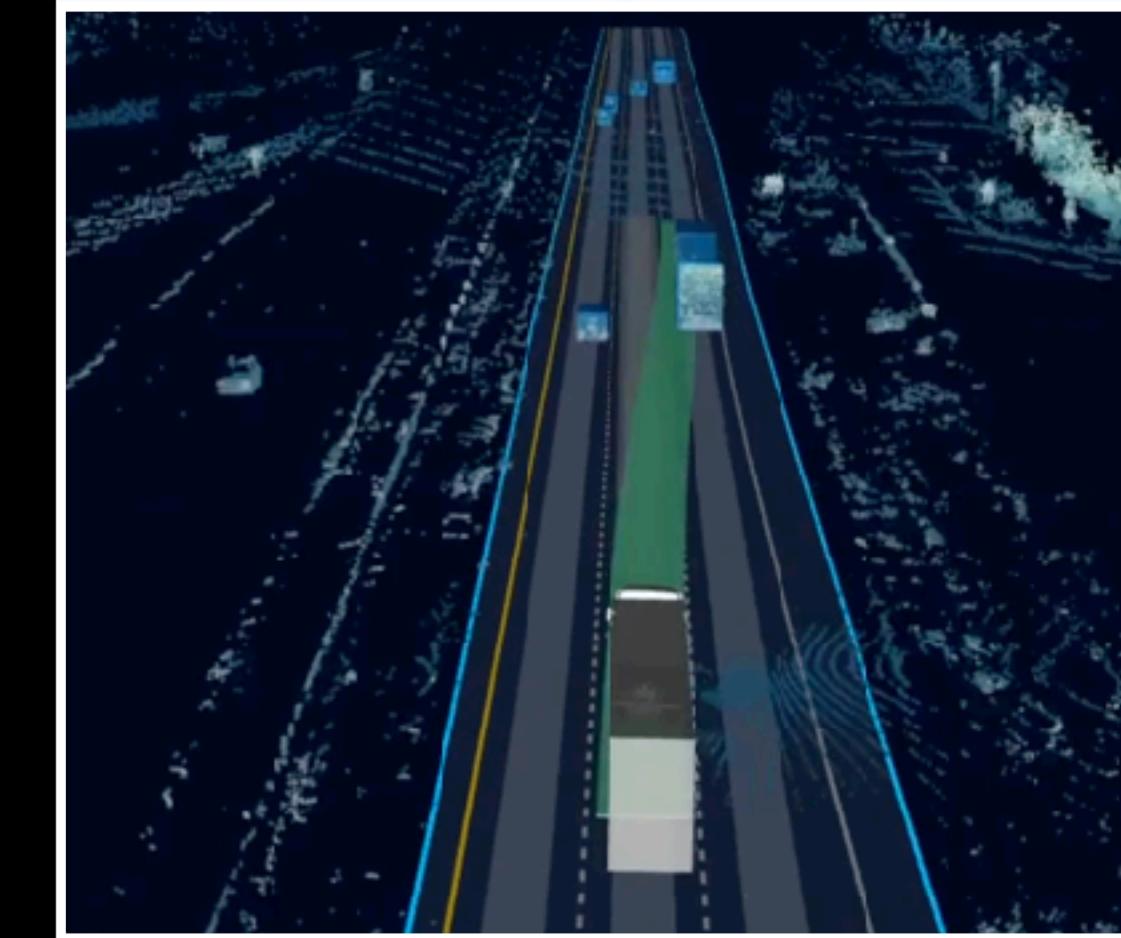
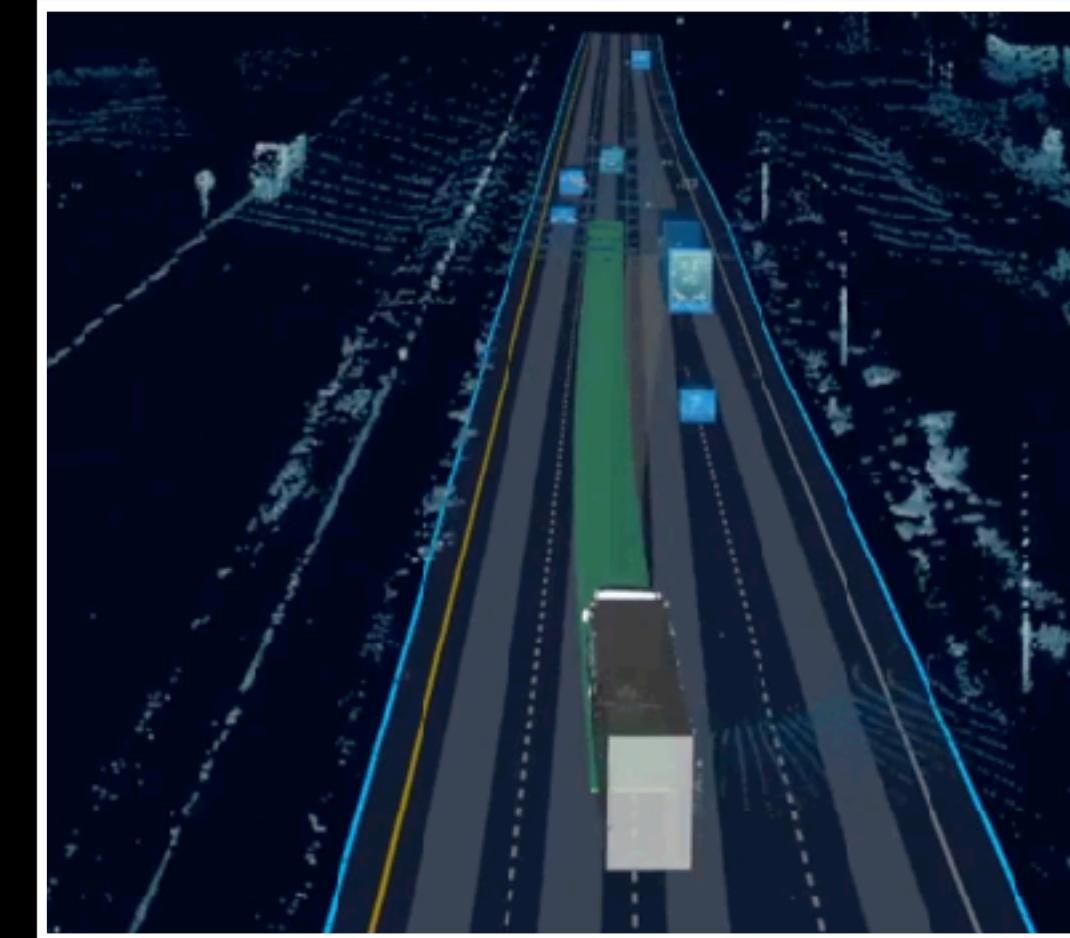
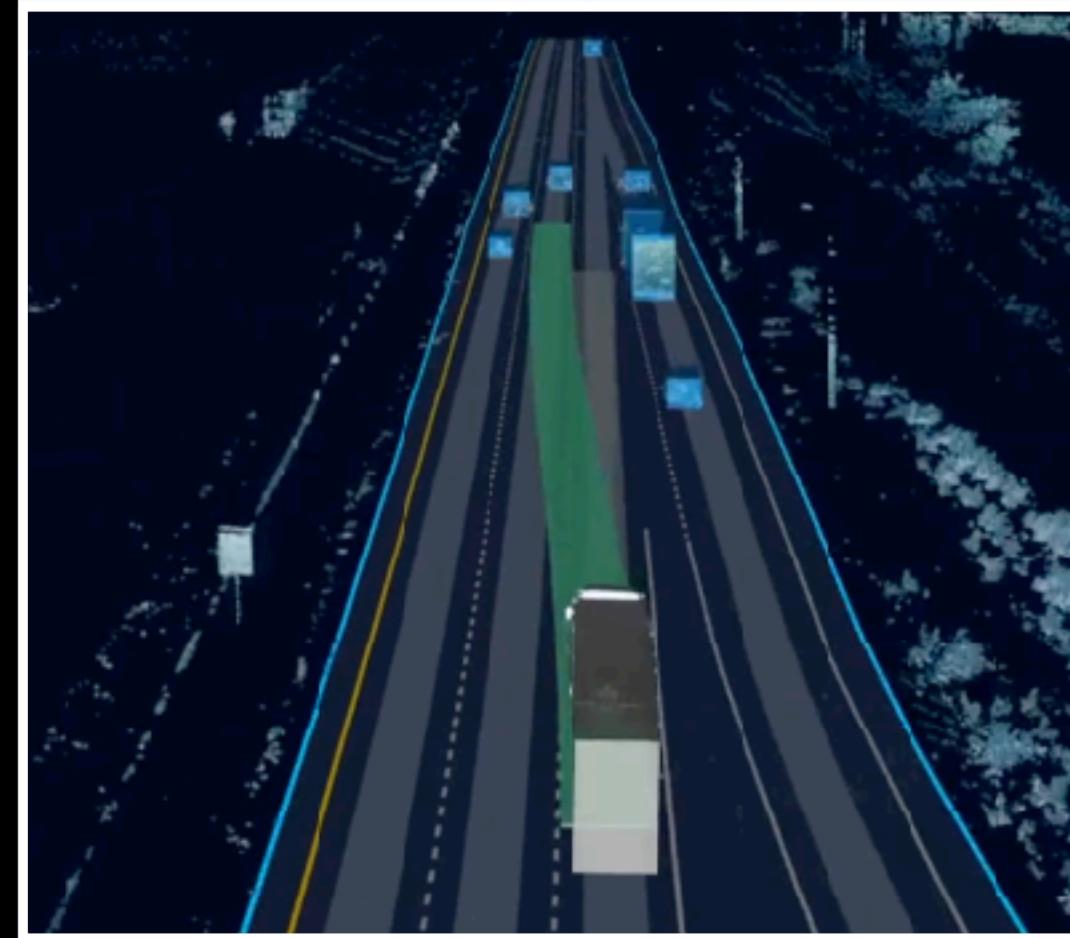
*“What is the
metric to match to
human?”*

Behavior Cloning

State



↓



Action



Behavior Cloning

1. Collect data from a human demonstrator

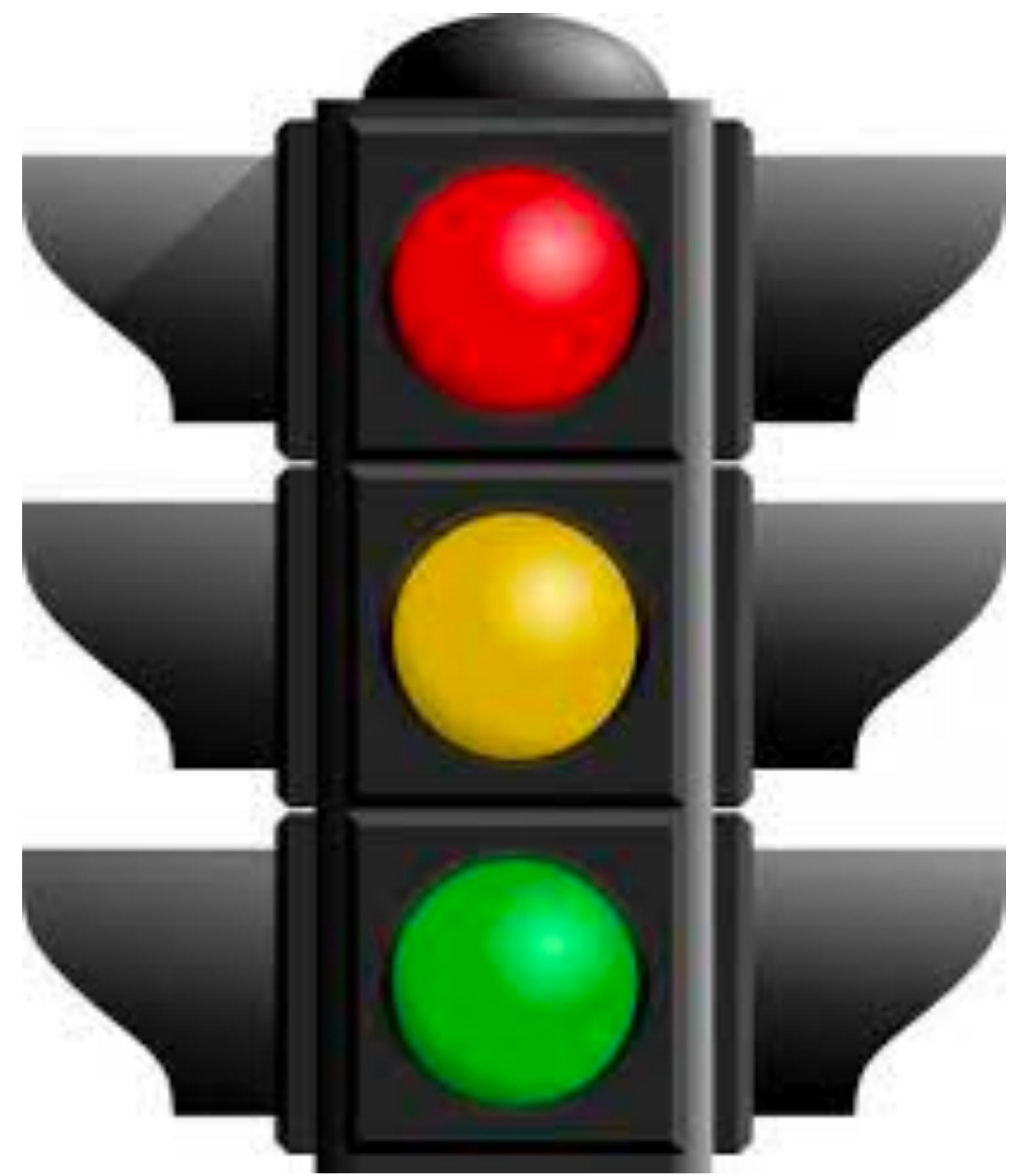
$$s_1, a_1^*, s_2, a_2^*, s_3, a_3^*, \dots$$

2. Train a policy $\pi : s_t \rightarrow a_t$

3. Validate on held out dataset

What could possibly go wrong?







Feedback drives
covariate shift



An old problem

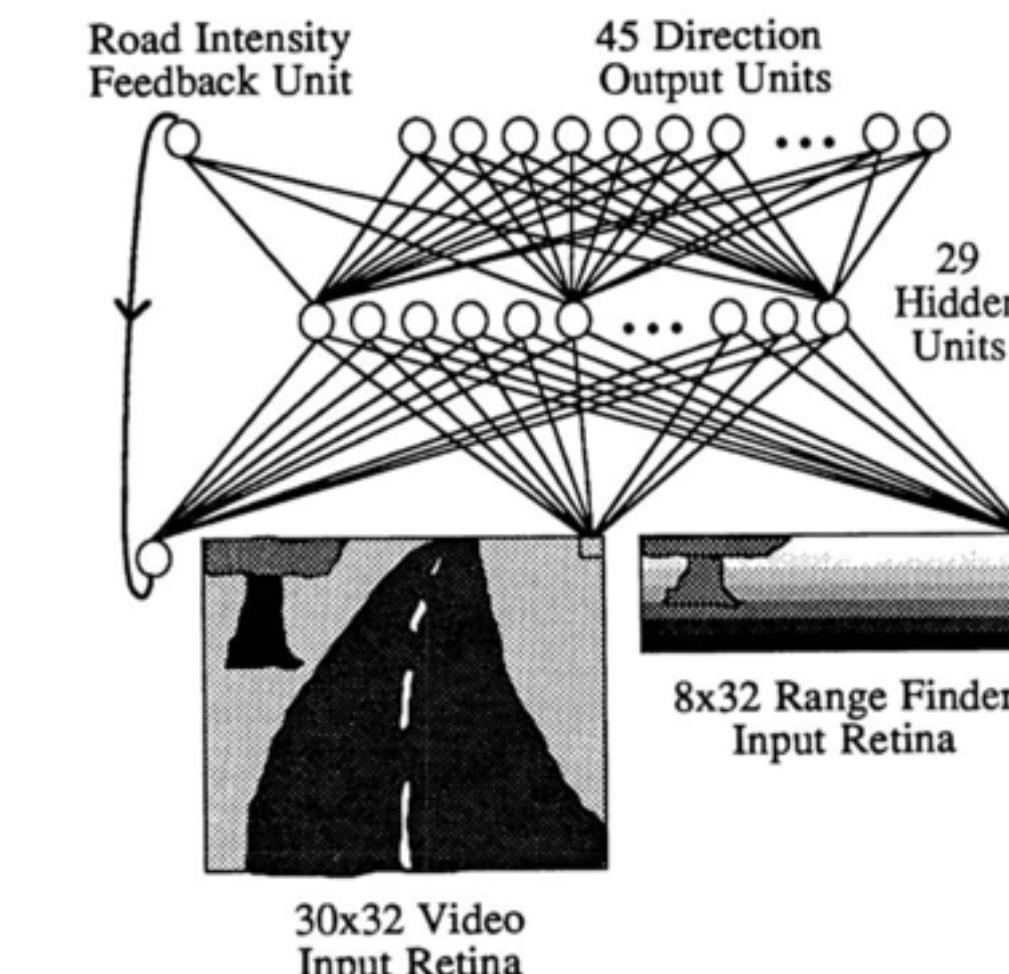


Figure 1: ALVINN Architecture

“...the network must not solely be shown examples of accurate driving, **but also how to recover** (i.e. return to the road center) once a mistake has been made.”

D. Pomerleau

ALVINN: An Autonomous Land Vehicle In A Neural Network, NeurIPS'89

Also observed by [LeCun'05]

Feedback is a pervasive problem in self-driving

“... the inertia problem. When the ego vehicle is stopped (e.g., at a red traffic light), the probability it stays static is indeed overwhelming in the training data. This creates a spurious correlation between low speed and no acceleration, inducing excessive stopping and difficult restarting in the imitative policy ...”

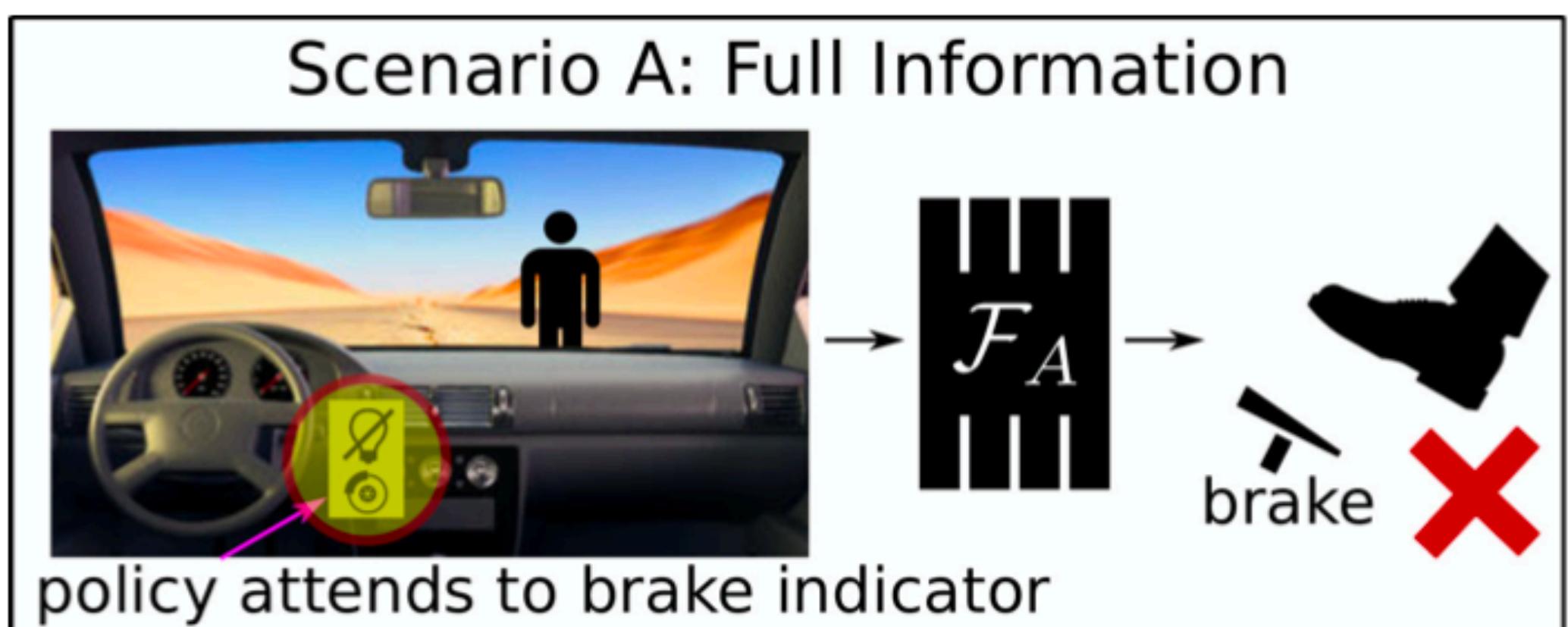
“Exploring the Limitations of Behavior Cloning for Autonomous Driving.”
F. Codevilla, E. Santana, A. M. Lopez, A. Gaidon. ICCV 2019

“... small errors in action predictions to compound over time, eventually leading to states that human drivers infrequently visit and are not adequately covered by the training data. Poorer predictions can cause a feedback cycle known as cascading errors ...”

“Imitating Driver Behavior with Generative Adversarial Networks”.
A. Kuefler, J. Morton, T. Wheeler, M. Kochenderfer, IV 2017

“... During closed-loop inference, this breaks down because the past history is from the net’s own past predictions. For example, such a trained net may learn to only stop for a stop sign if it sees a deceleration in the past history, and will therefore never stop for a stop sign during closed-loop inference ...”

“ChauffeurNet: Learning to Drive by Imitating the Best and Synthesizing the Worst”. M. Bansal, A. Krizhevsky, A. Ogale, Waymo 2018



“Causal Confusion in Imitation Learning”.
P. de Haan, D. Jayaraman, S. Levine, NeurIPS ‘19



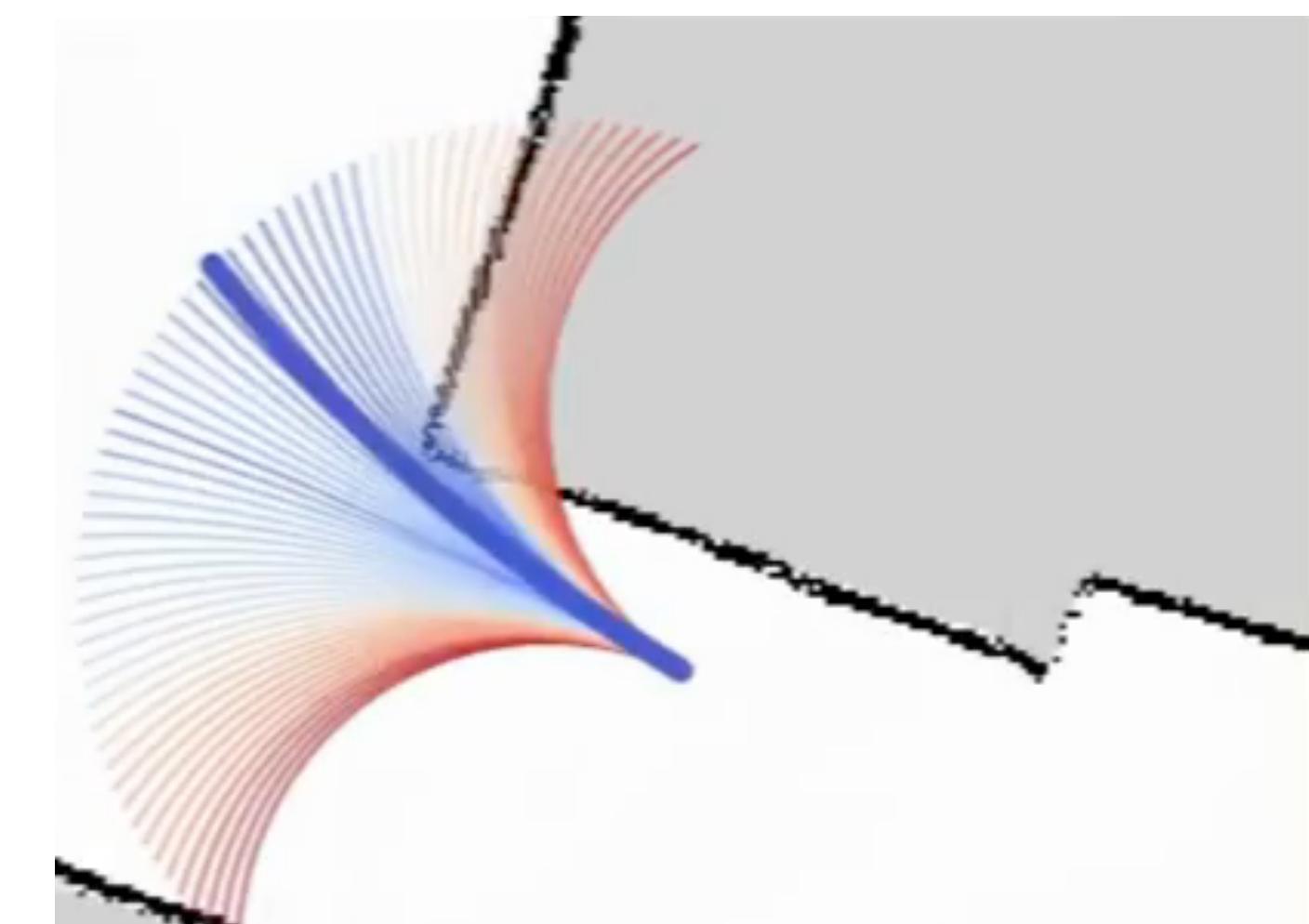
Feedback is an old adversary!



[SCB+ RSS'20]



Demonstration

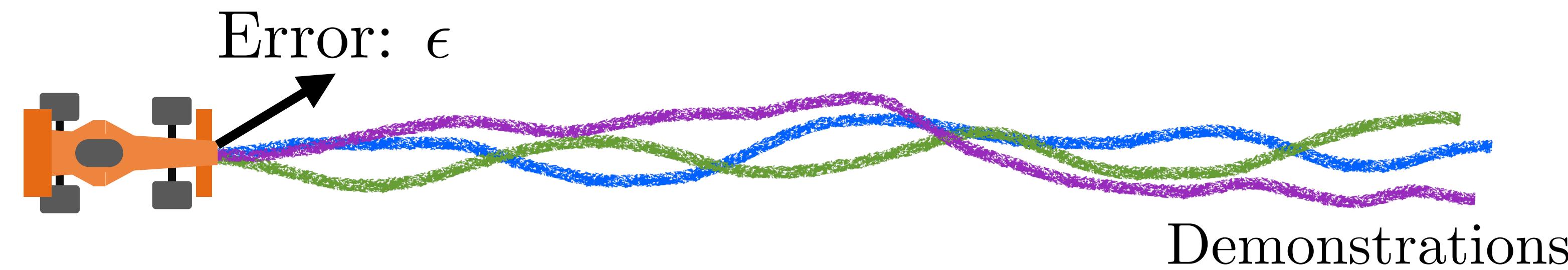


Learnt policy

Behavior Cloning crashes into a wall

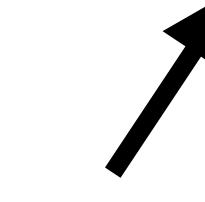


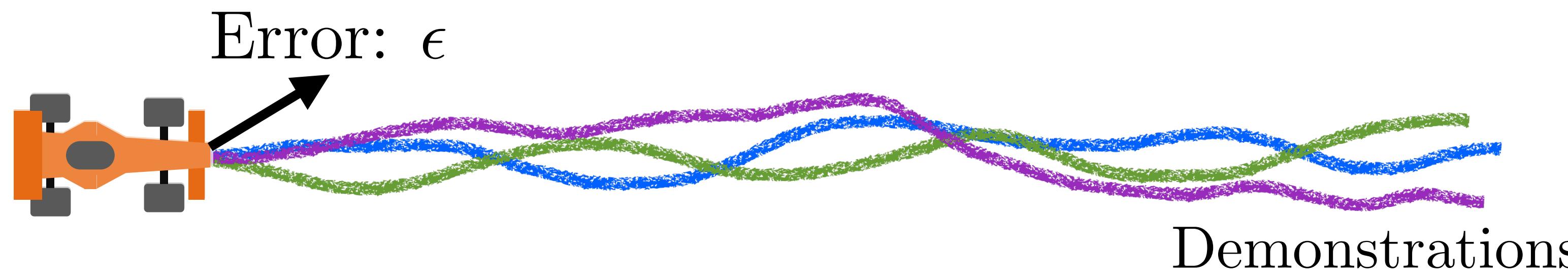
Why did the robot crash?



Why did the robot crash?



??  No training data
Error: 1.0



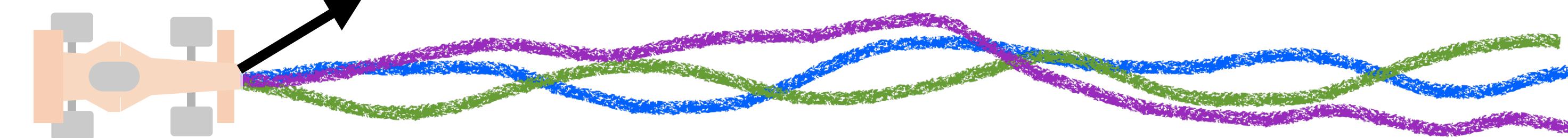
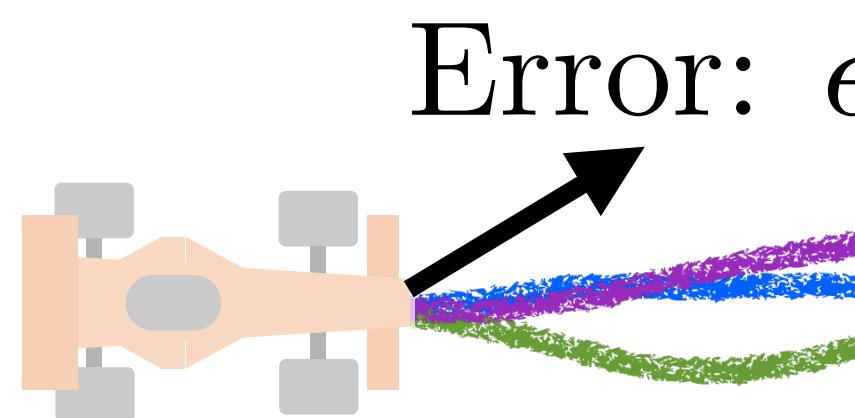
Why did the robot crash?



No training data
Error: 1.0



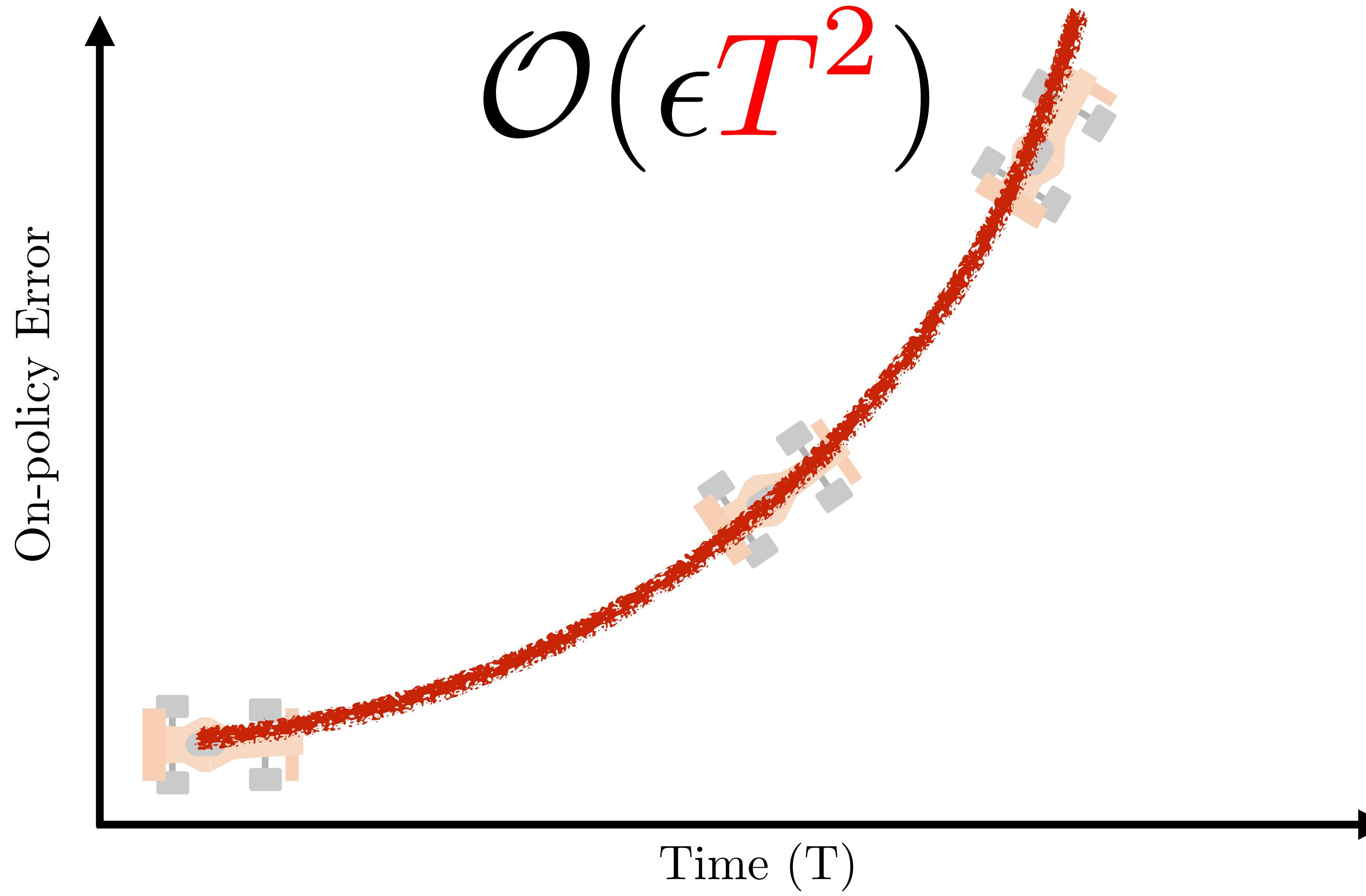
No training data
Error: 1.0



Demonstrations

Errors feedback and compound

[Ross and Bagnell'10]



Prove it!



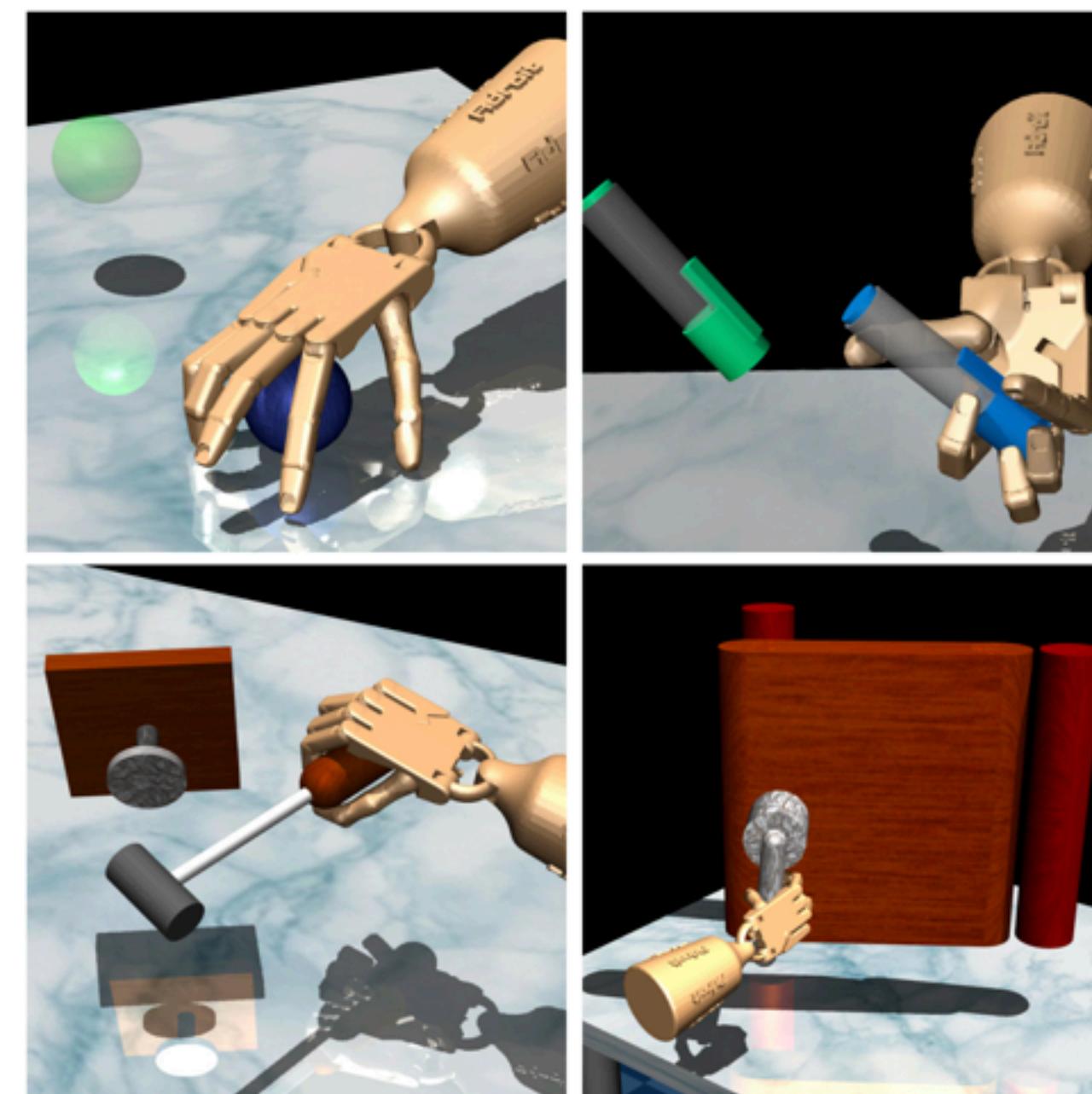


Feedback drives
covariate shift

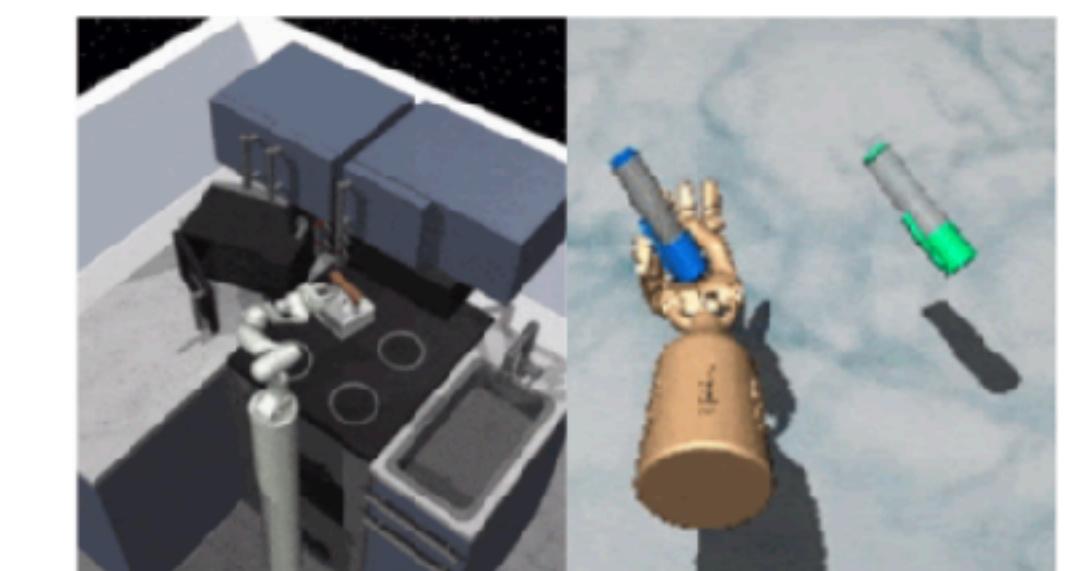
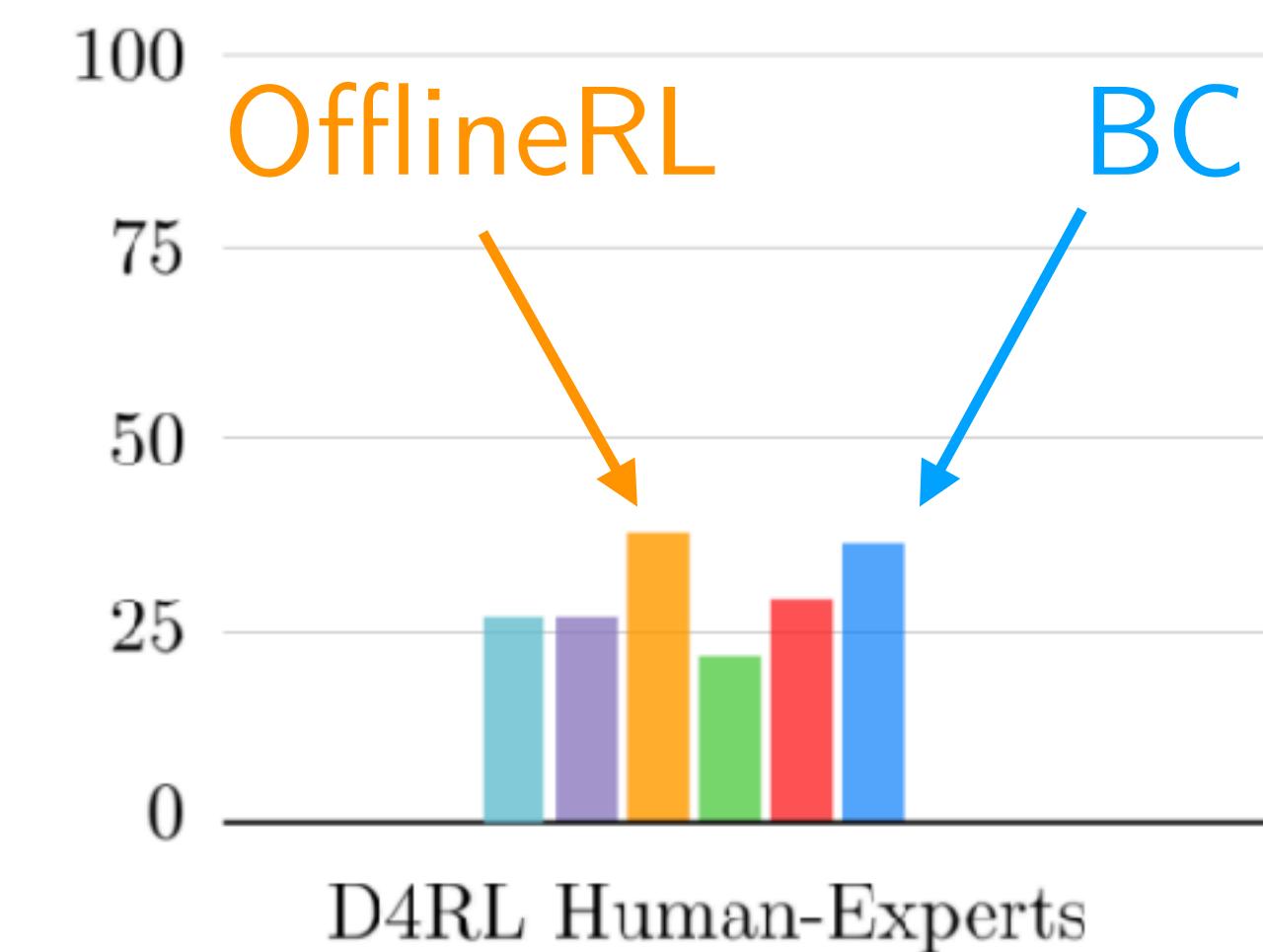
But ... Behavior Cloning works just fine on benchmark datasets!

Environment	Expert	BC
CartPole	500 ± 0	500 ± 0
Acrobot	-71.7 ± 11.5	-78.4 ± 14.2
MountainCar	-99.6 ± 10.9	-107.8 ± 16.4
Hopper	3554 ± 216	3258 ± 396
Walker2d	5496 ± 89	5349 ± 634
HalfCheetah	4487 ± 164	4605 ± 143
Ant	4186 ± 1081	3353 ± 1801

[SCV+ arXiv '21]



[Rajeswaran et al. '17]



[Florence et al. '21]



What explains this mismatch?

Real-world self-driving

vs

Benchmark datasets

*Feedback drives
covariate shift,
Behavior Cloning
compounds in error*

*Behavior Cloning
does just fine!*

Let's travel to the INFINITE data limit!

The Three Regimes of Covariate Shift



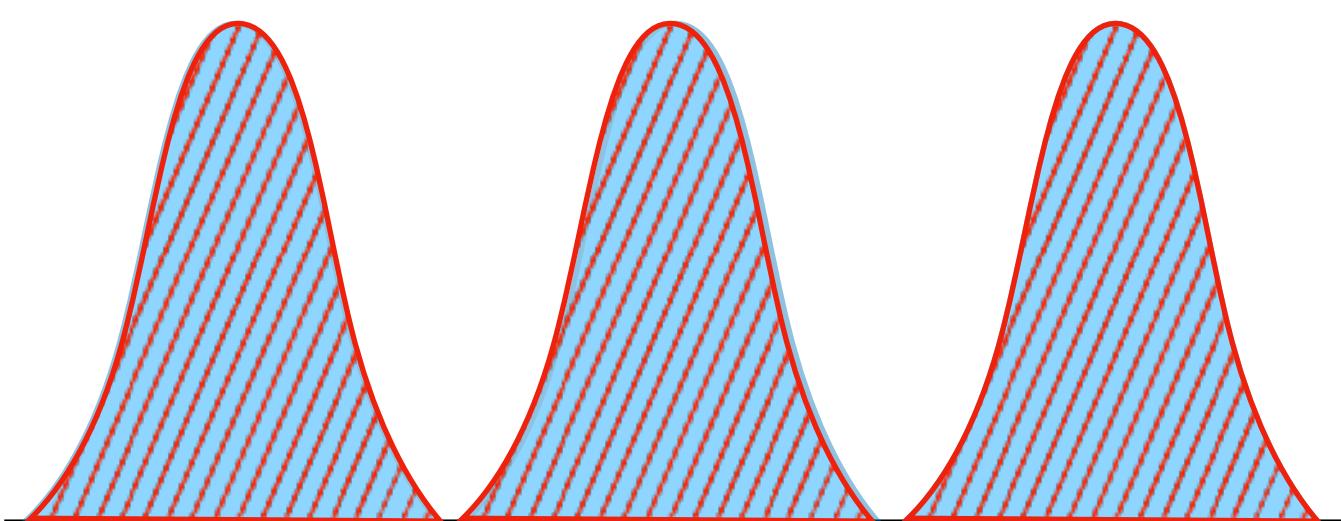
Easy



Expert is **realizable**

$$\pi^E \in \Pi$$

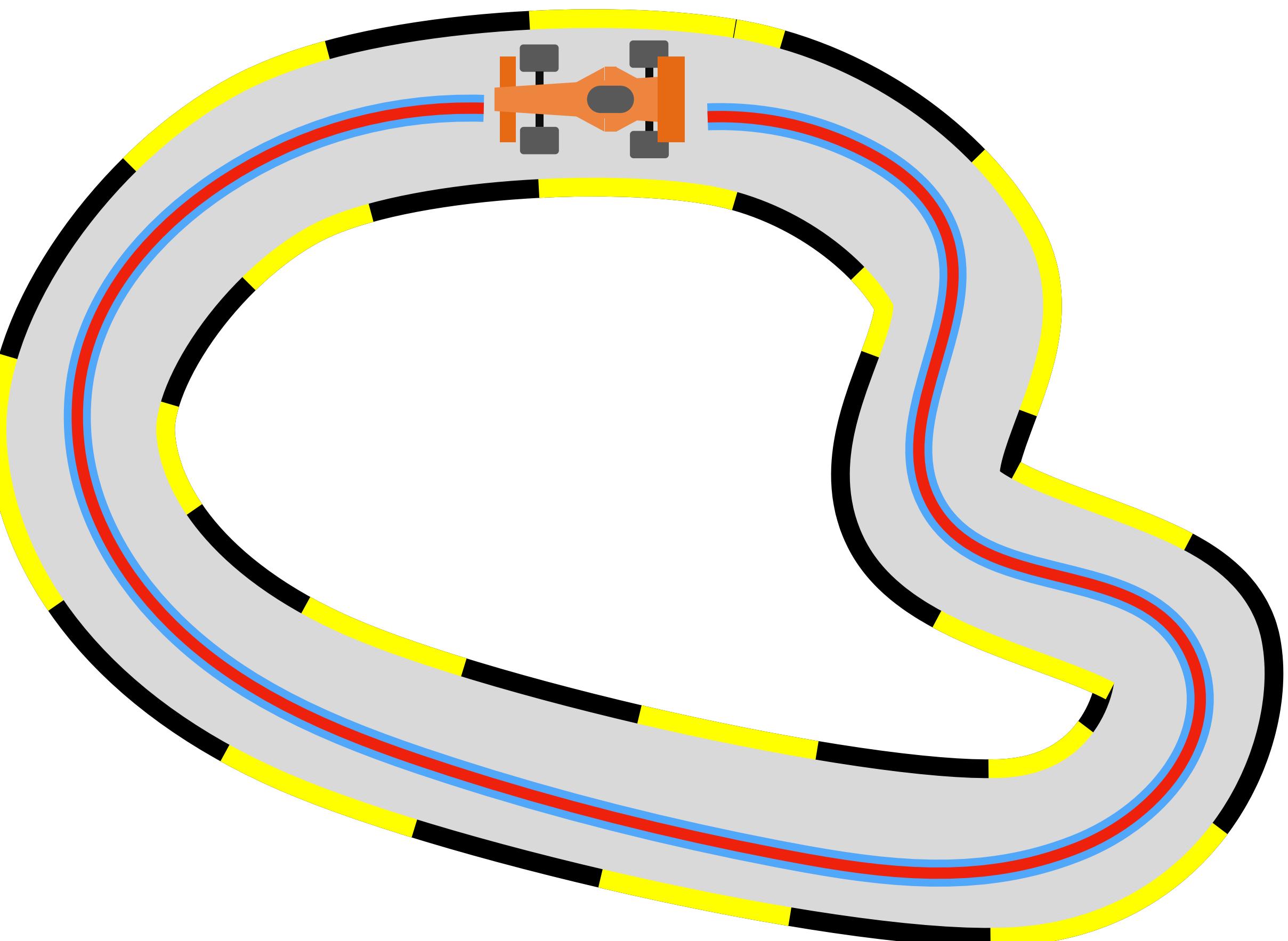
As $N \rightarrow \infty$, drive down
 $\epsilon = 0$ (or Bayes error)



Nothing special.

Collect lots of data and
do Behavior Cloning

$$\text{Expert } \rho^{\pi^E}(s) \approx \text{Learner } \rho^\pi(s)$$



Easy



Hard



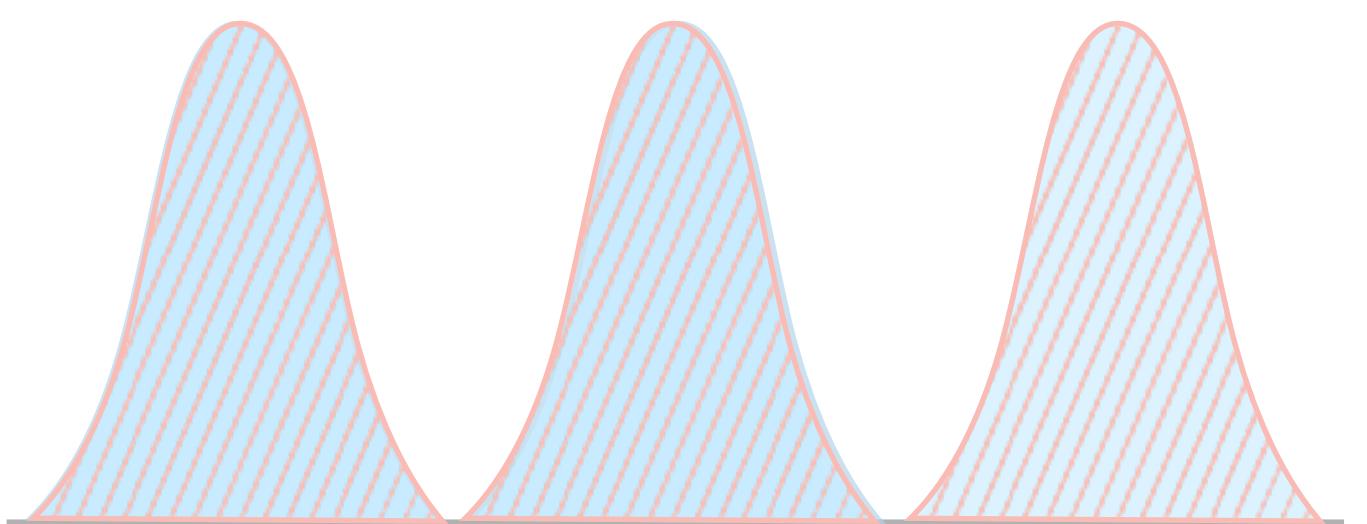
Expert is **realizable**

$$\pi^E \in \Pi$$

Non-realizable expert +
limited expert support

As $N \rightarrow \infty$, drive down

$\epsilon = 0$ (or Bayes error)



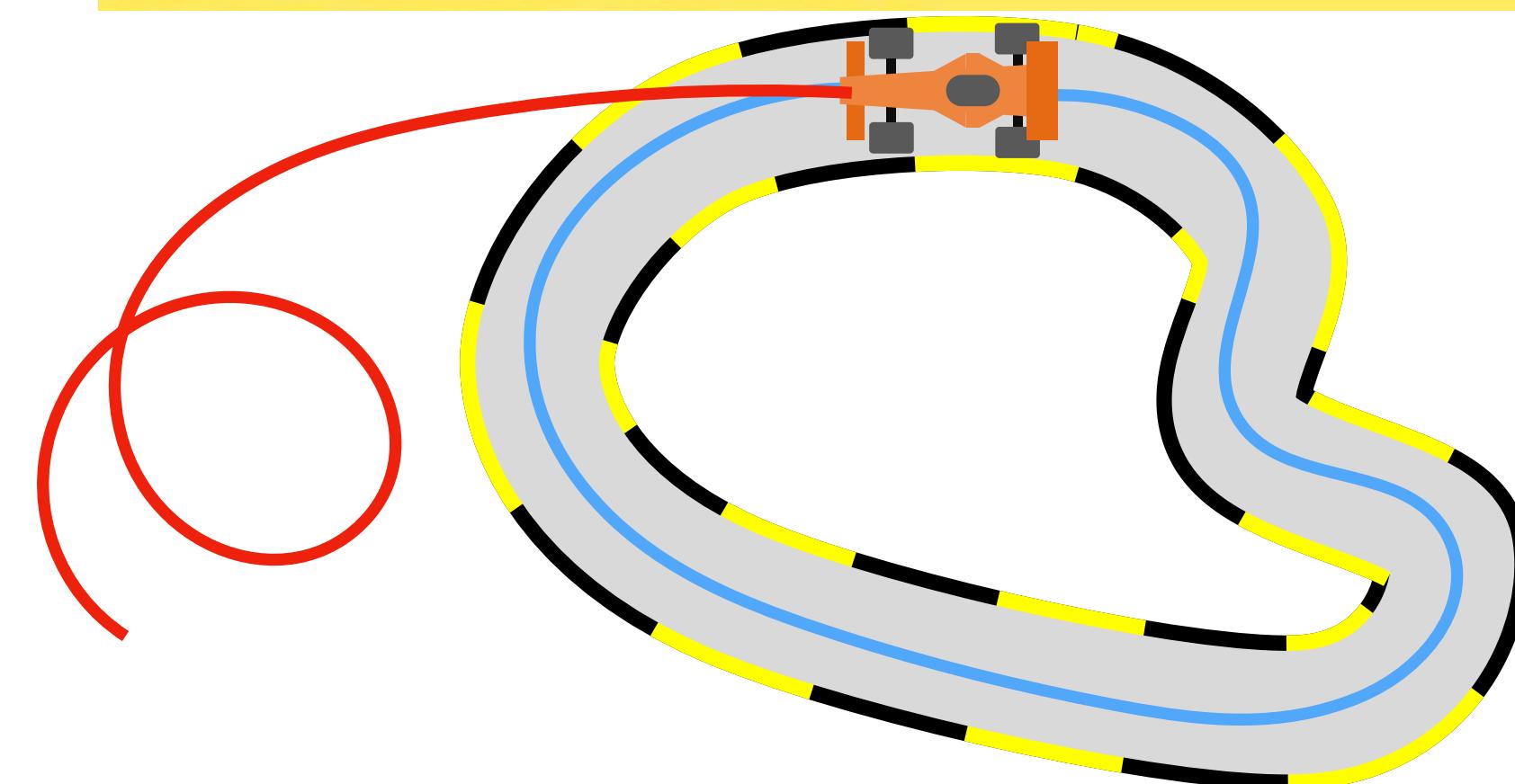
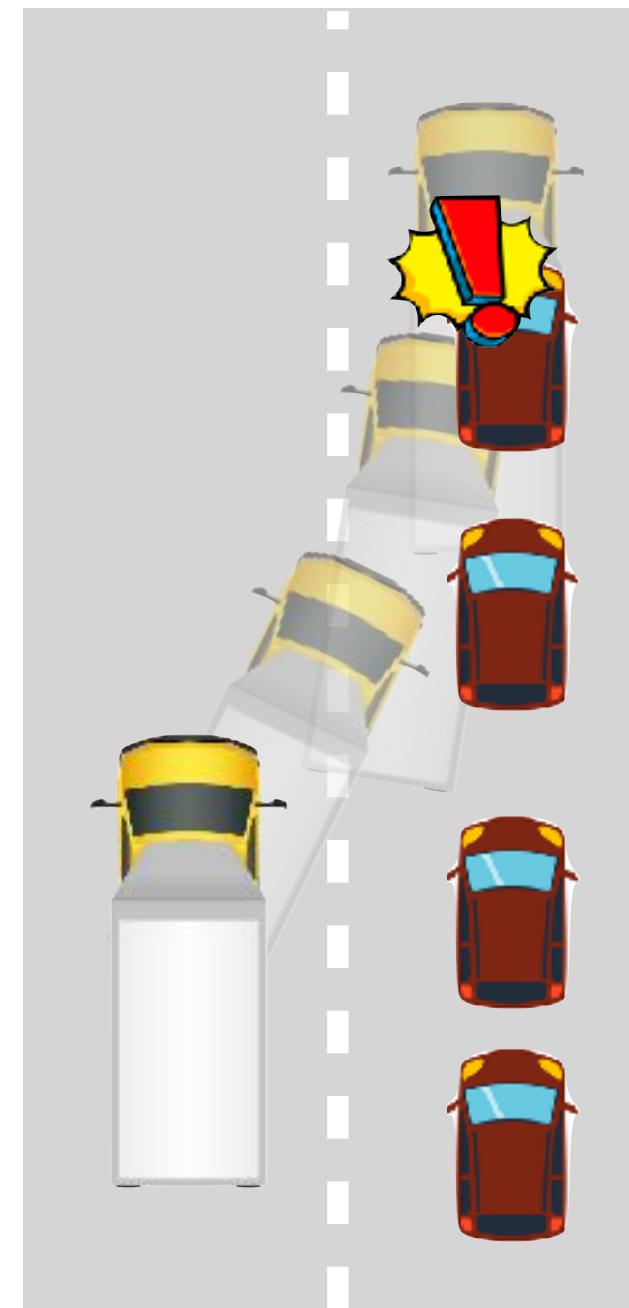
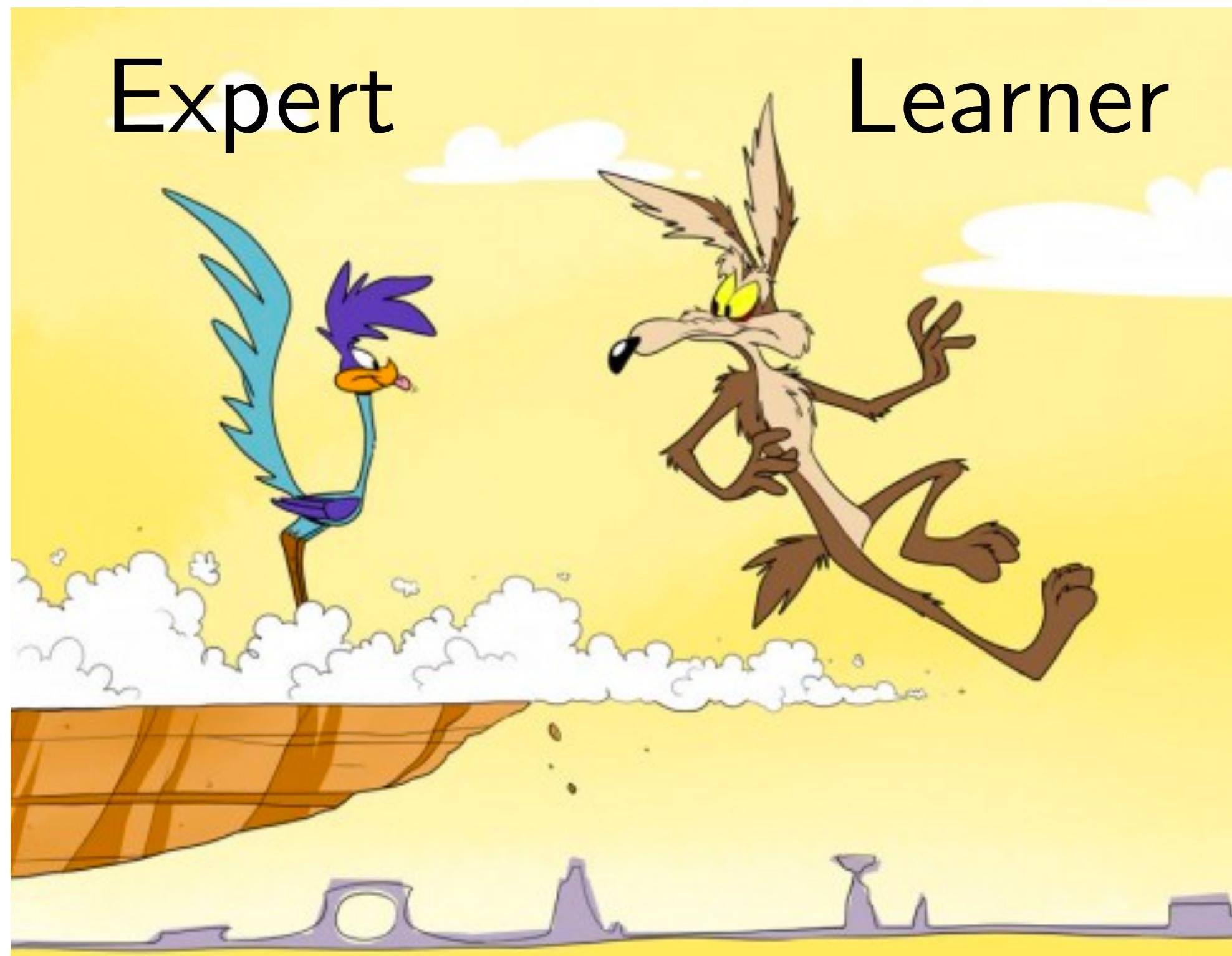
Nothing special.

Collect lots of data and
do Behavior Cloning

Setting

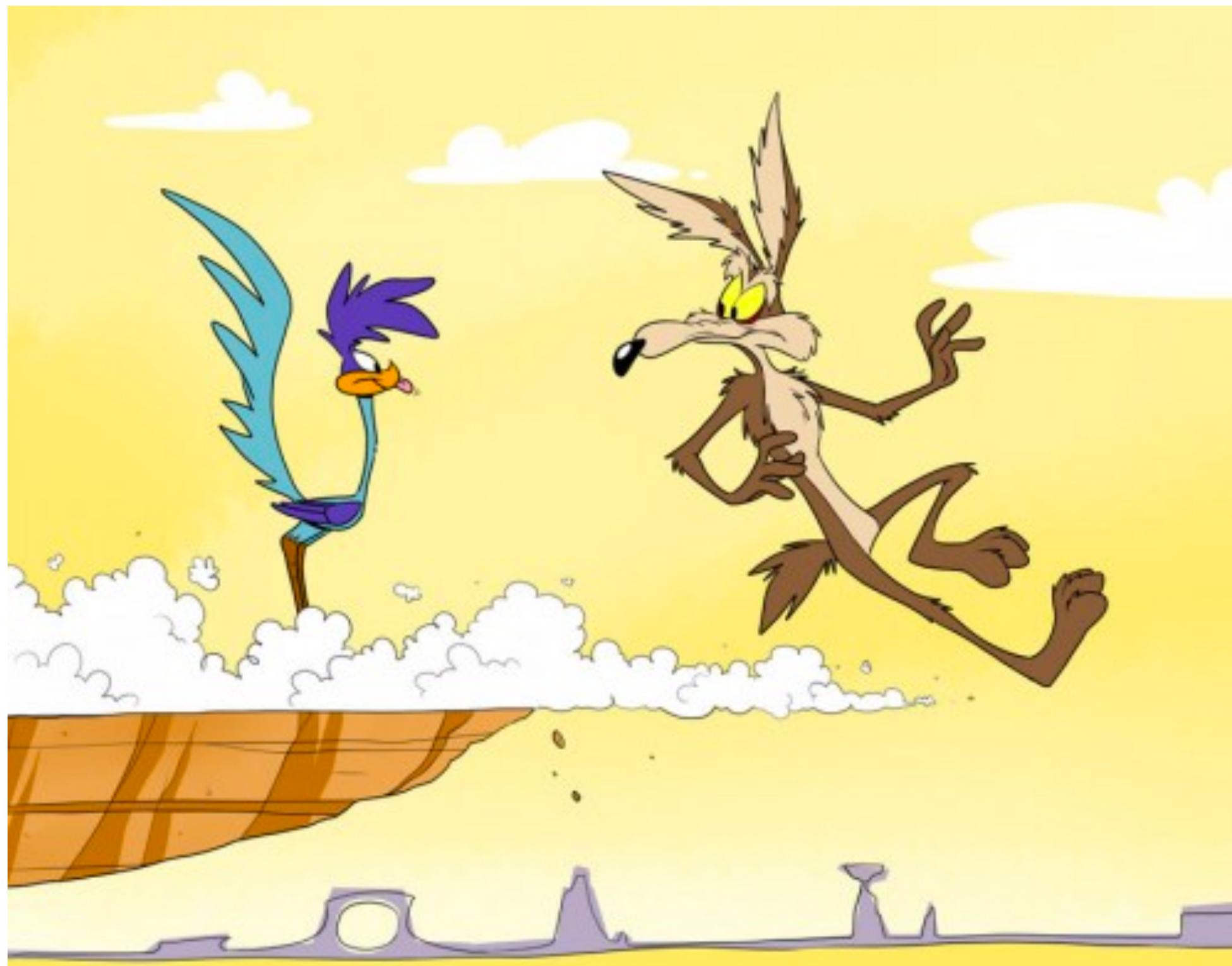
Solution

Non-realizable expert + limited support?

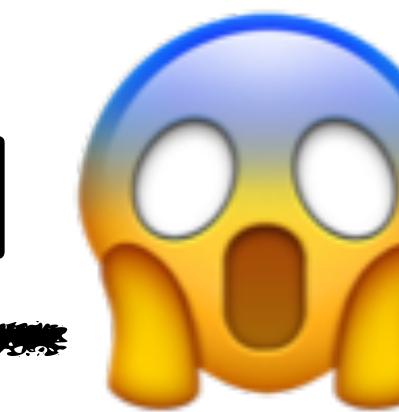


No label for what to do
in this state!

Non-realizable expert + limited support?

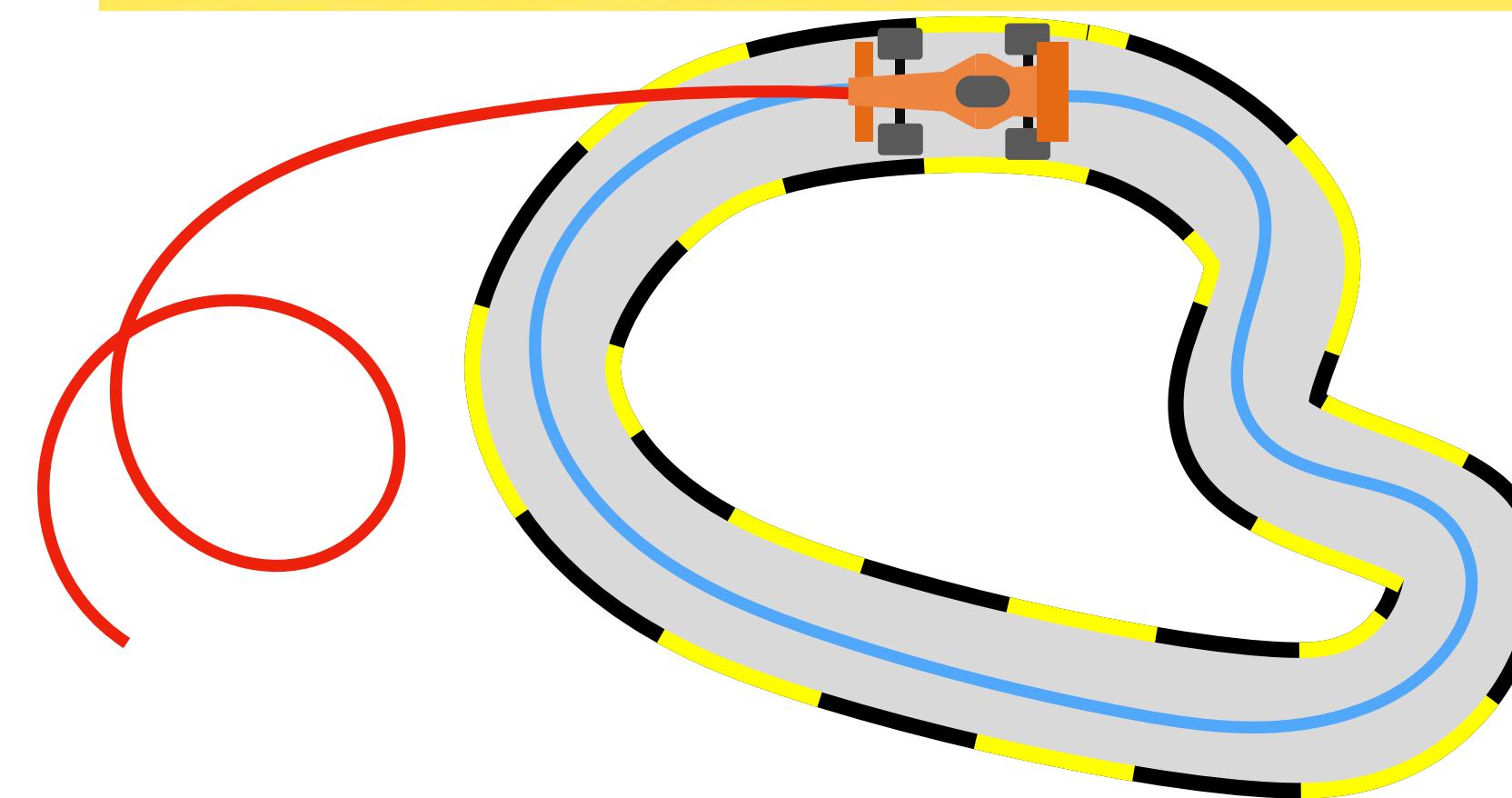


Hard



Behavior Cloning
compounds in error $O(\epsilon T^2)$

[Ross & Bagnell '10]



Easy



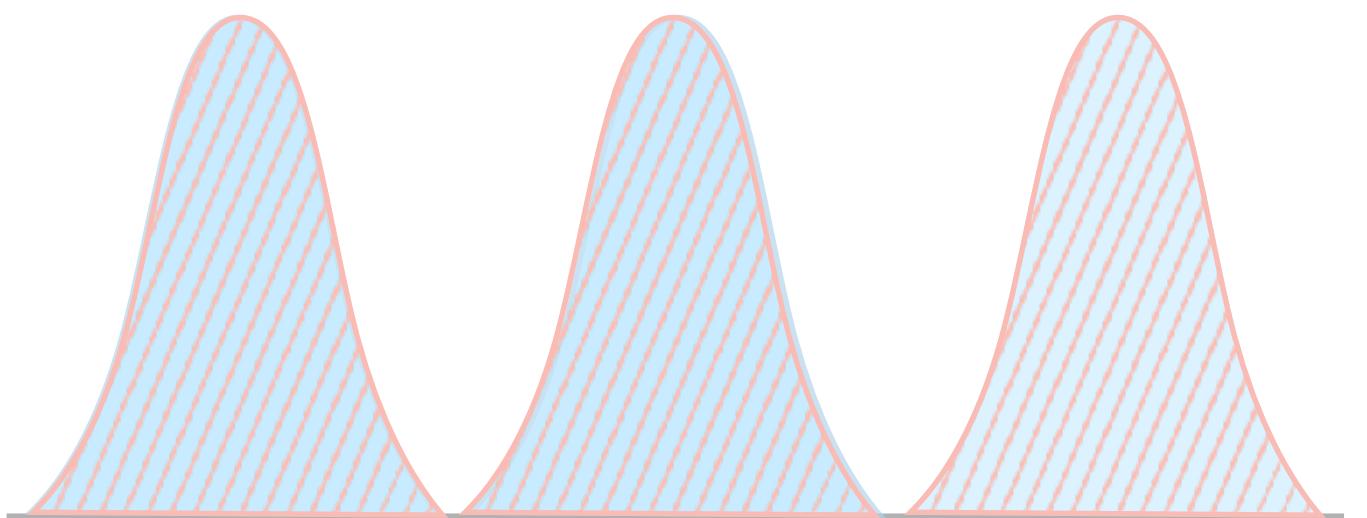
Hard



Expert is **realizable**

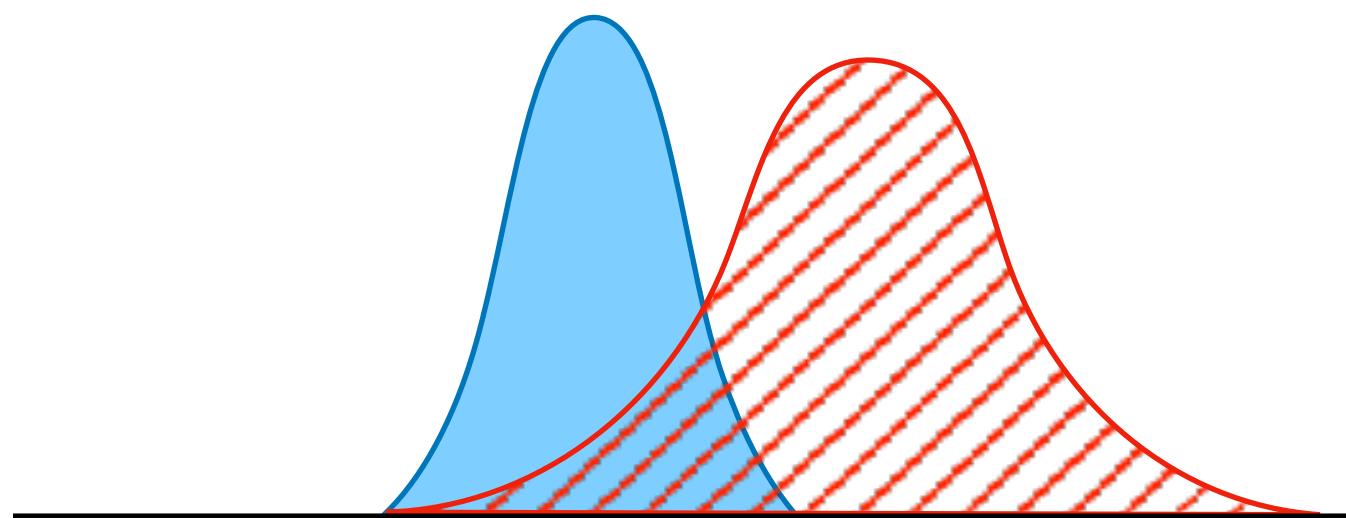
$$\pi^E \in \Pi$$

As $N \rightarrow \infty$, drive down
 $\epsilon = 0$ (or Bayes error)



Non-realizable expert +
limited expert support

Even as $N \rightarrow \infty$,
behavior cloning $O(\epsilon T^2)$



Nothing special.

Collect lots of data and
do Behavior Cloning

?