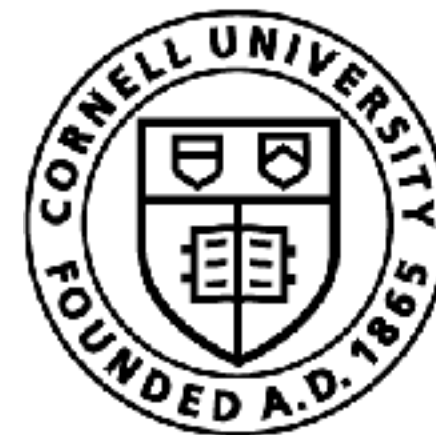


# Imitation Learning: Feedback and Covariate Shift

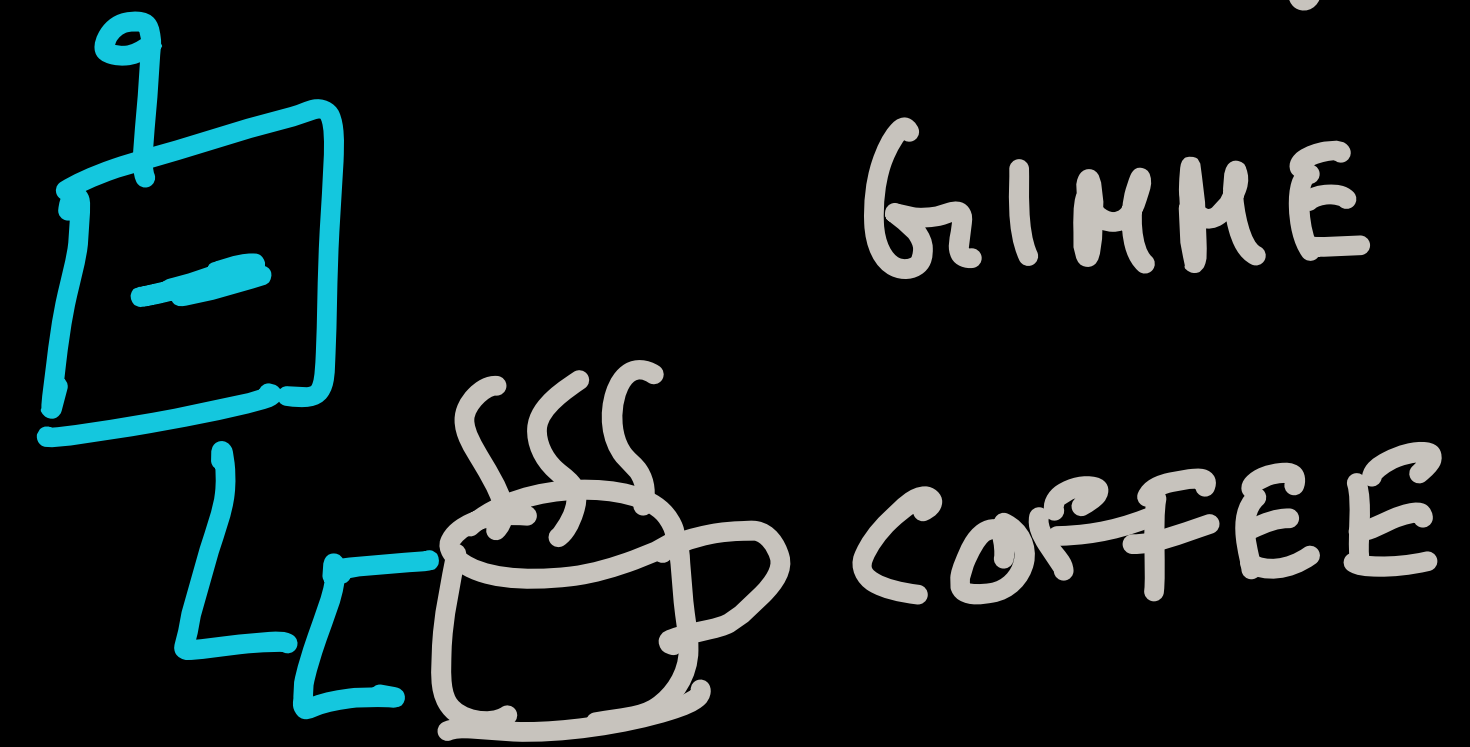
Sanjiban Choudhury



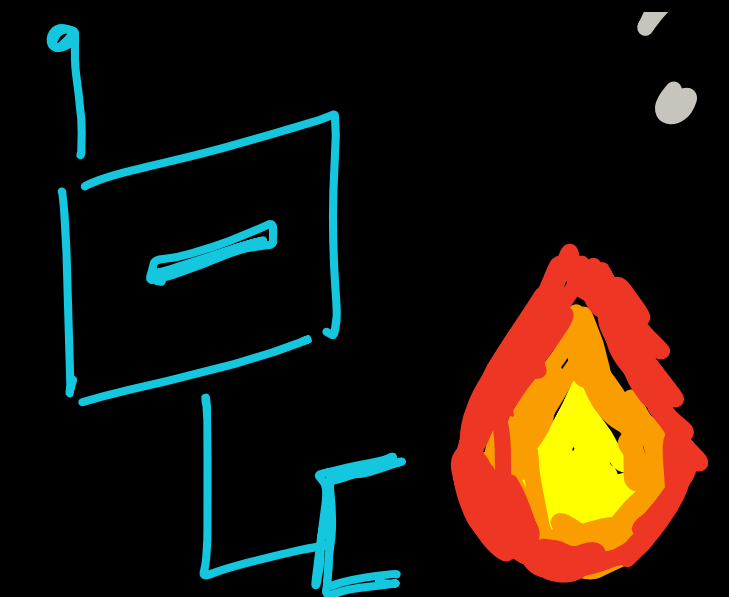
Cornell Bowers CIS  
**Computer Science**

# Programming a task ...

tell the robot to make coffee ..

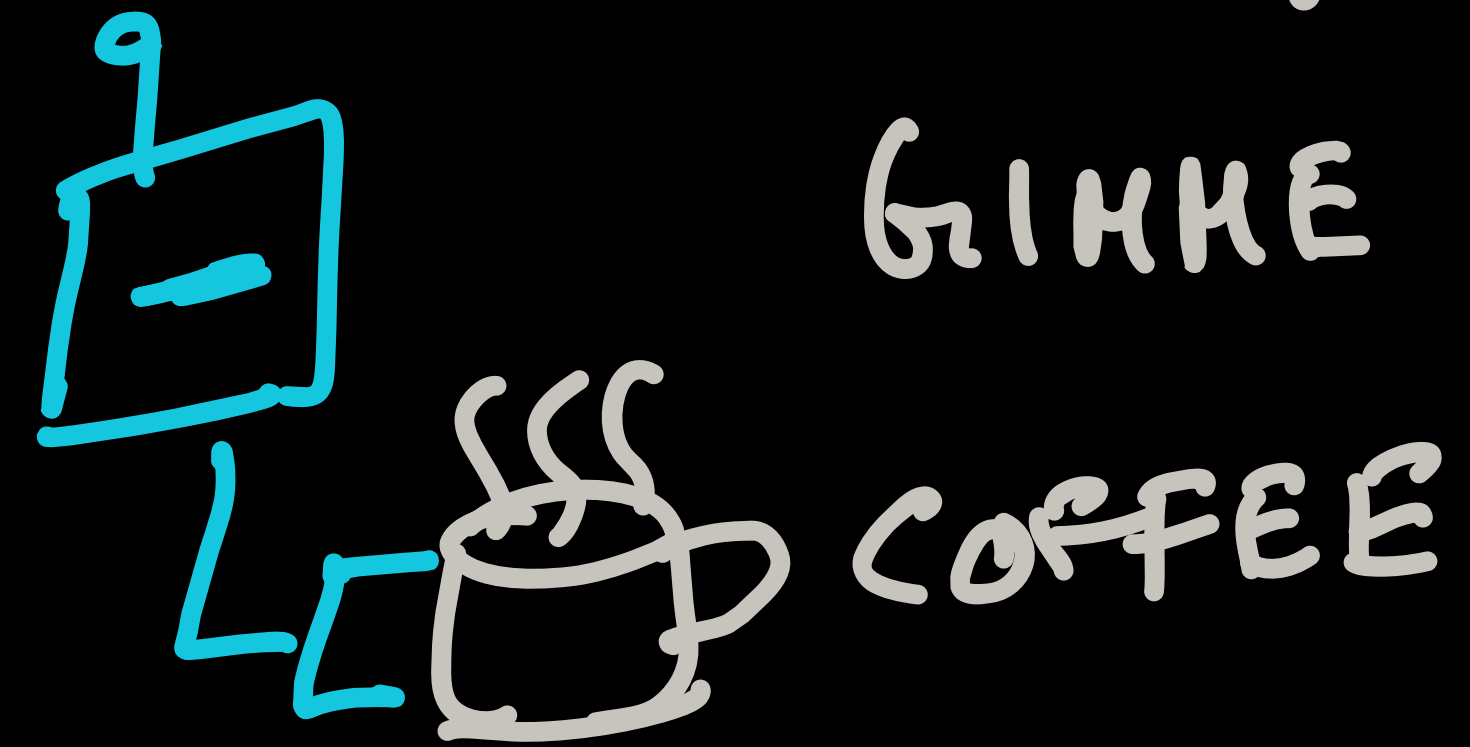


robot burns down  
the house!



# Programming a task ...

tell the robot to make coffee ..



DON'T ...

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

⋮



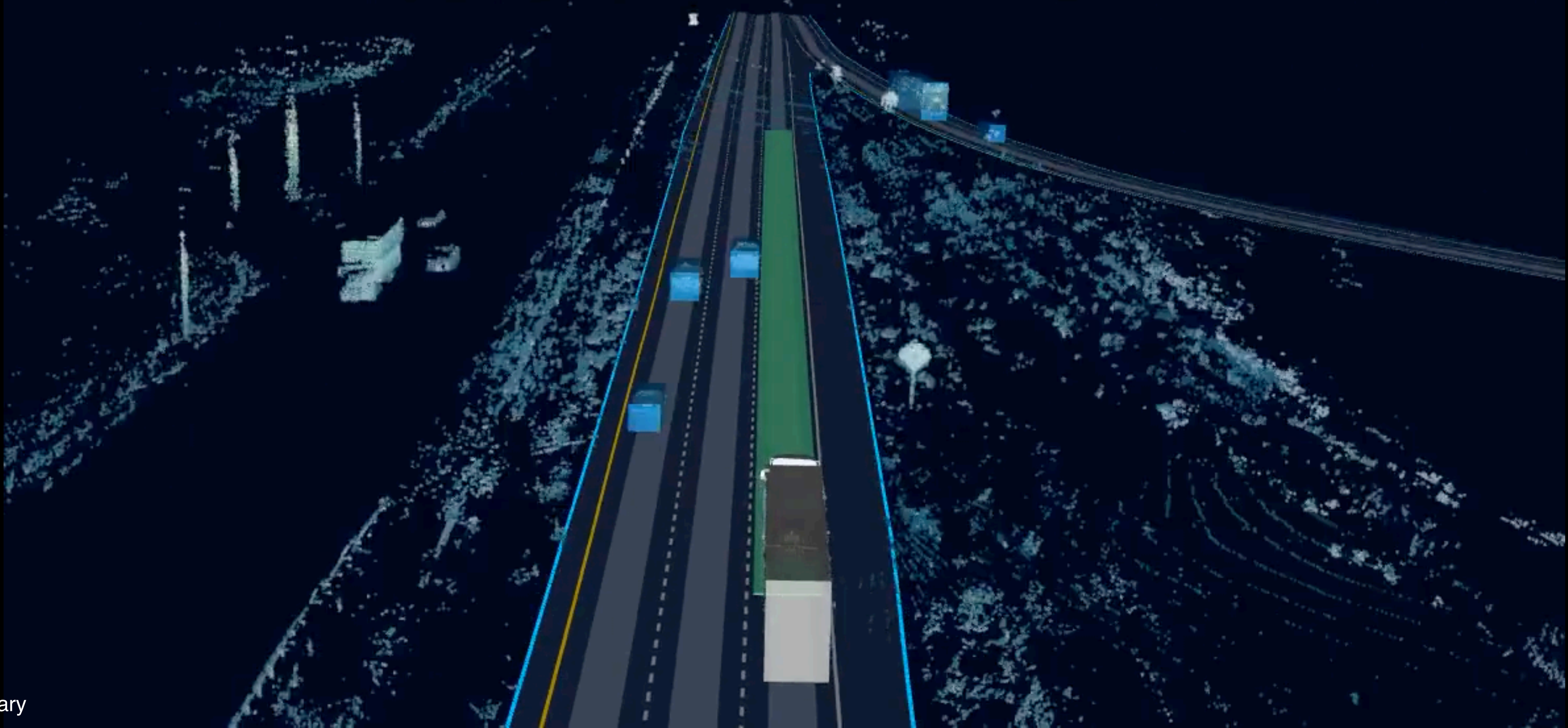
↑ STRAIGHT  
2.1 MI

AUTONOMY



65  
MPH

SPEED  
LIMIT  
75





↑ STRAIGHT  
1.9 MI

AUTONOMY



Department of Motor Vehicles



SPEED  
LIMIT  
**75**

# Official Driver Handbook



Driver License Division

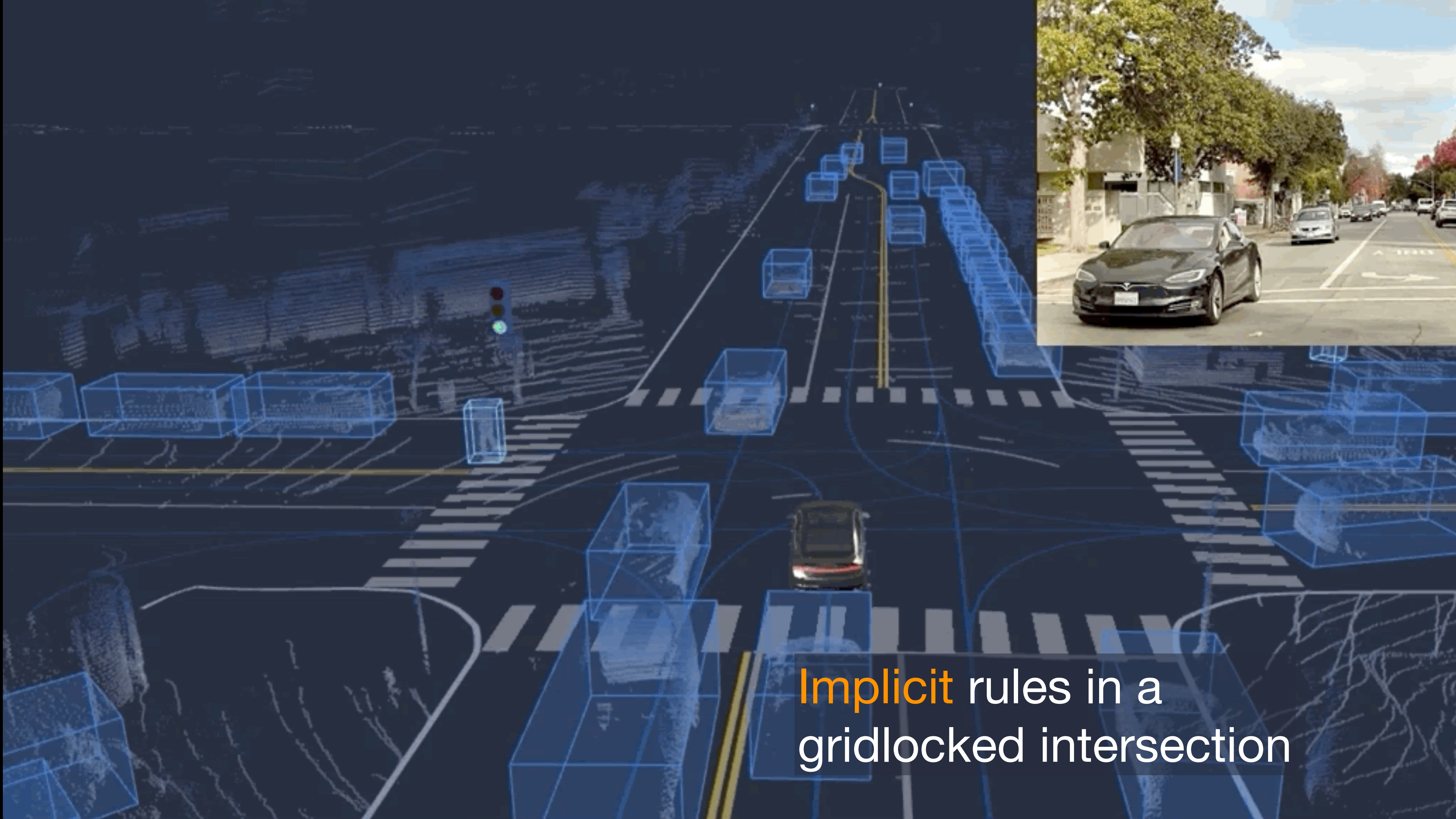
This publication is FREE



# 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*

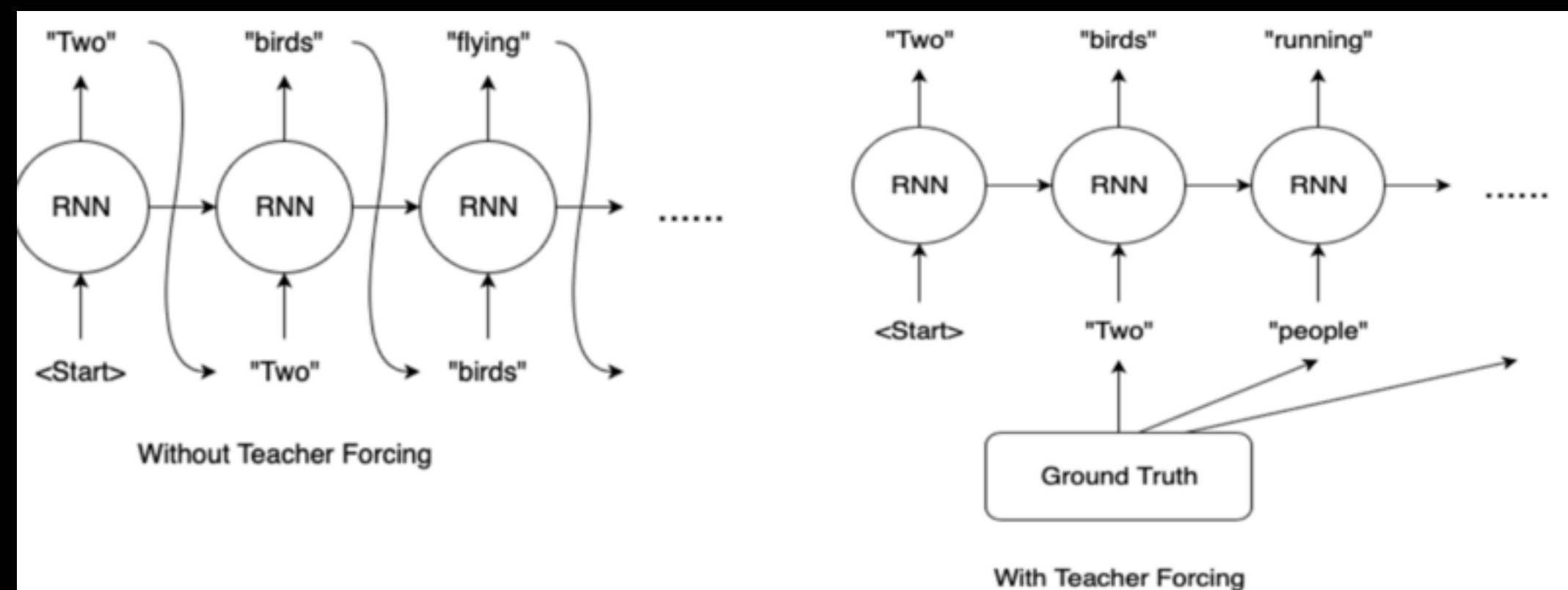


## Game AI

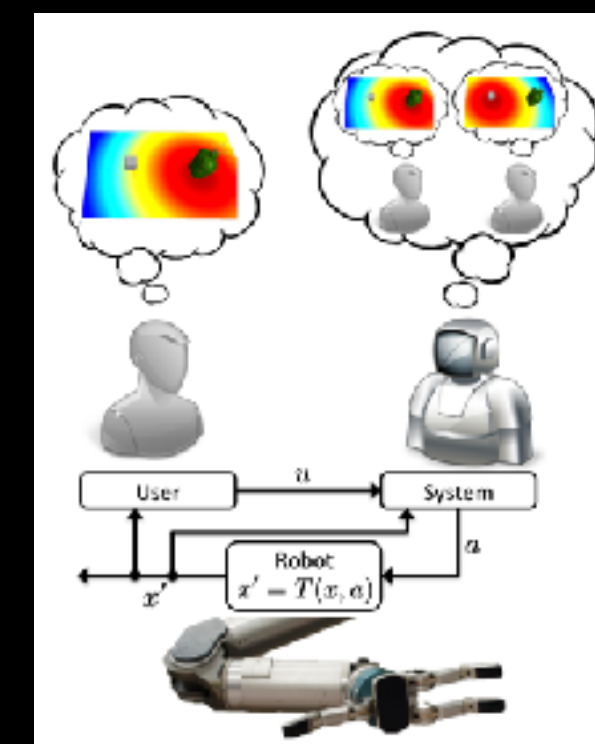
*Kozik et al. 2021*

## Sequence models in NLP

## Shared autonomy



*Daume et al. 2009*



*Javdani et al. 2015*



Activity!

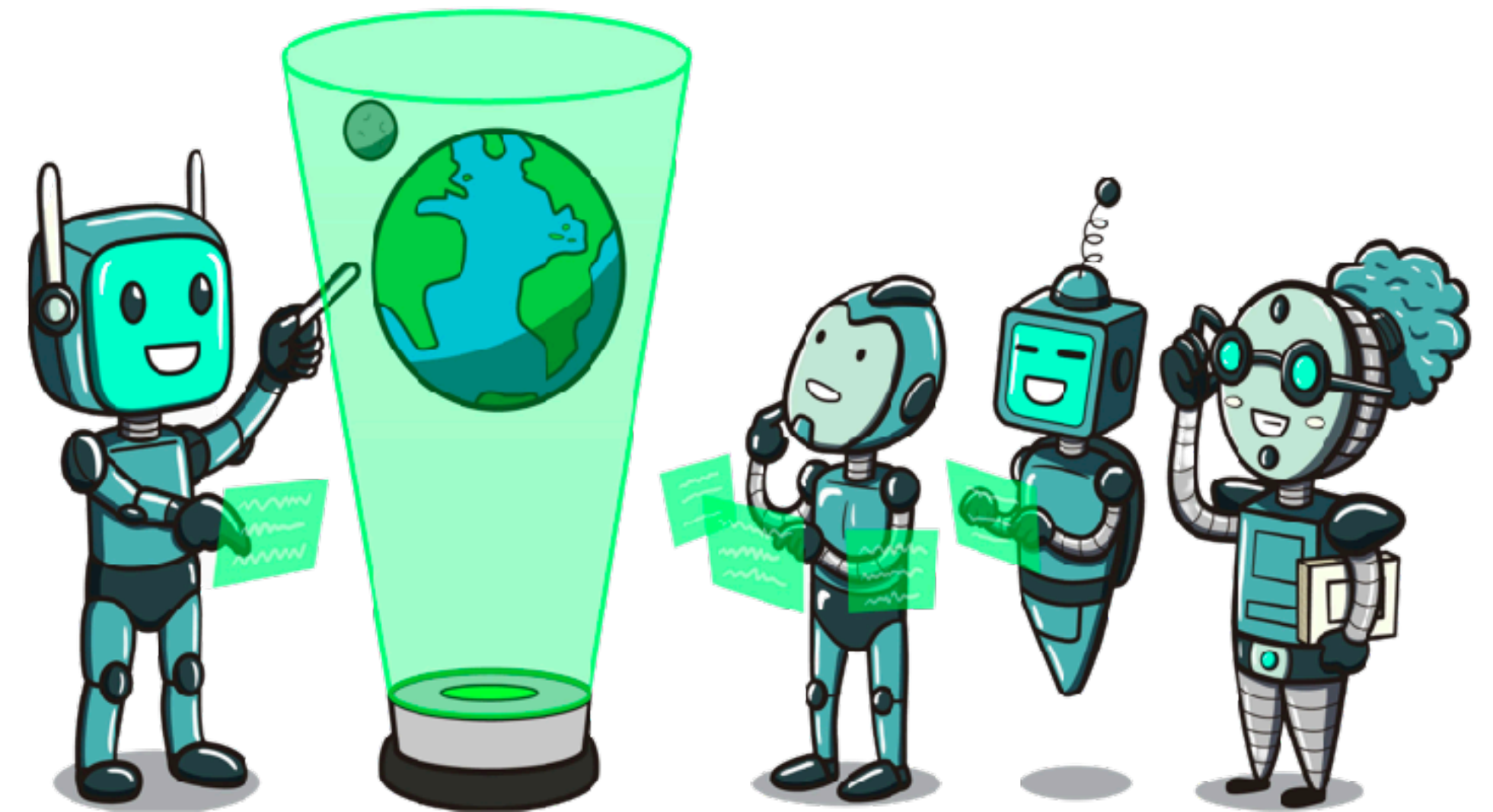


# Think-Pair-Share!





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

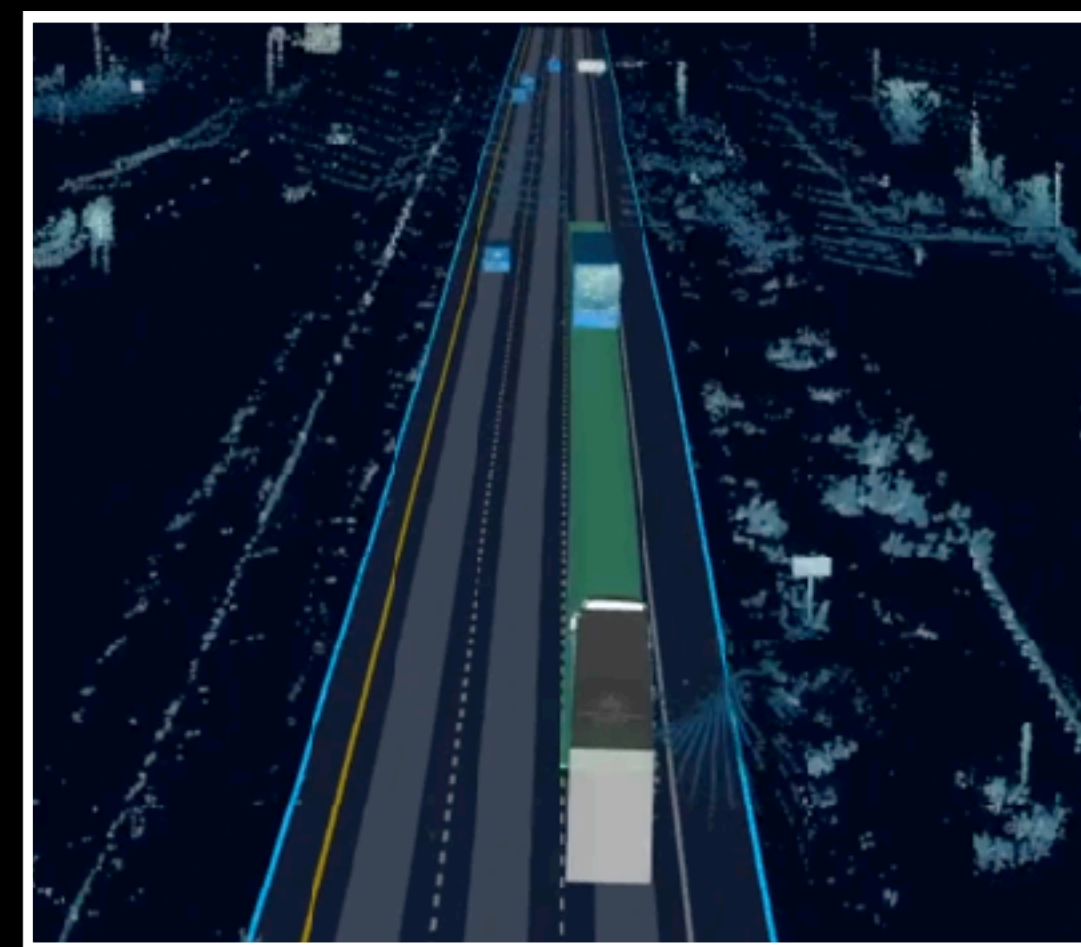
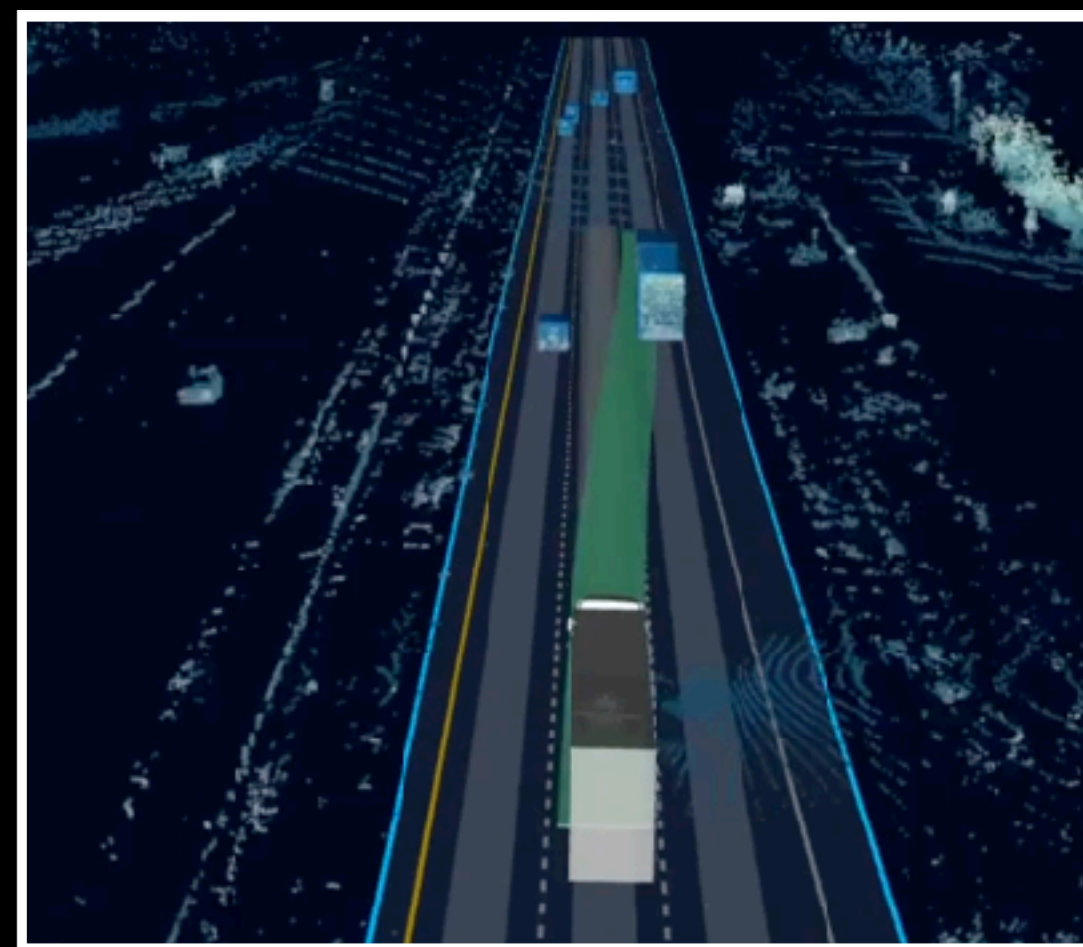
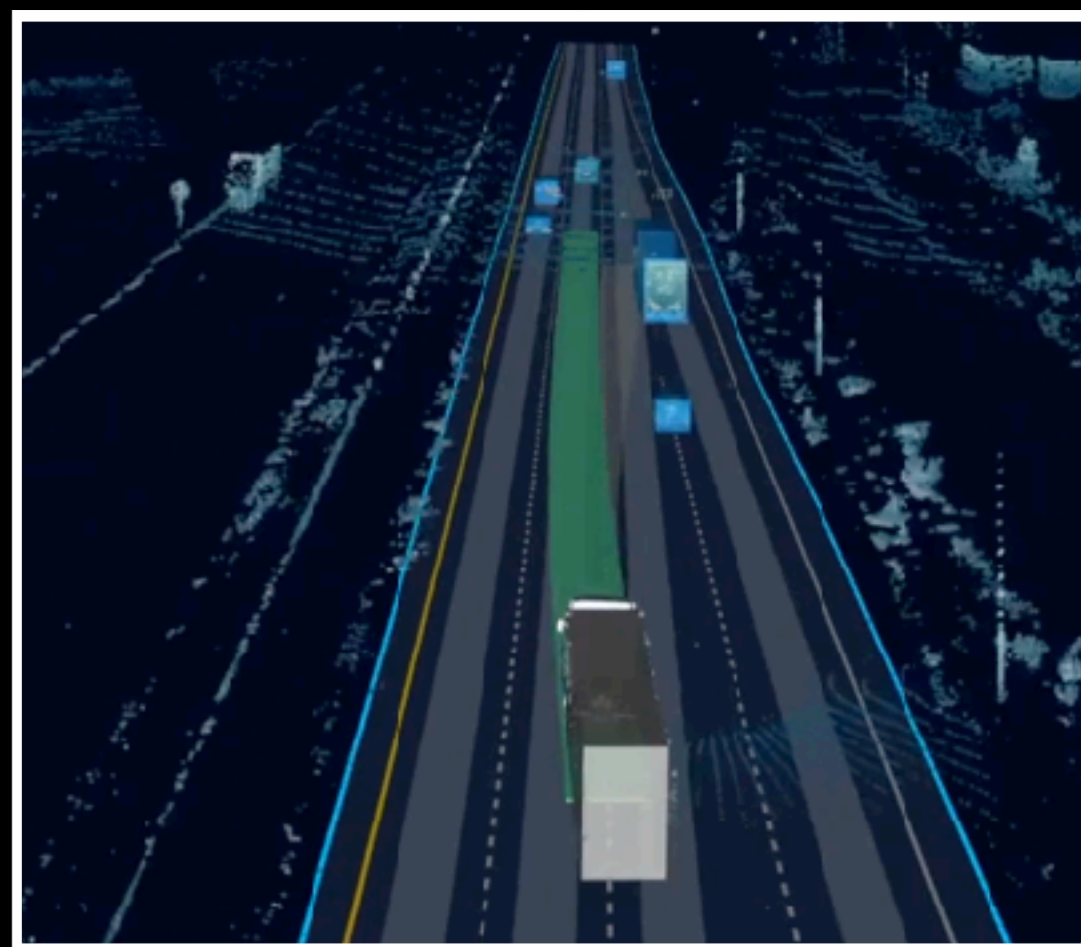
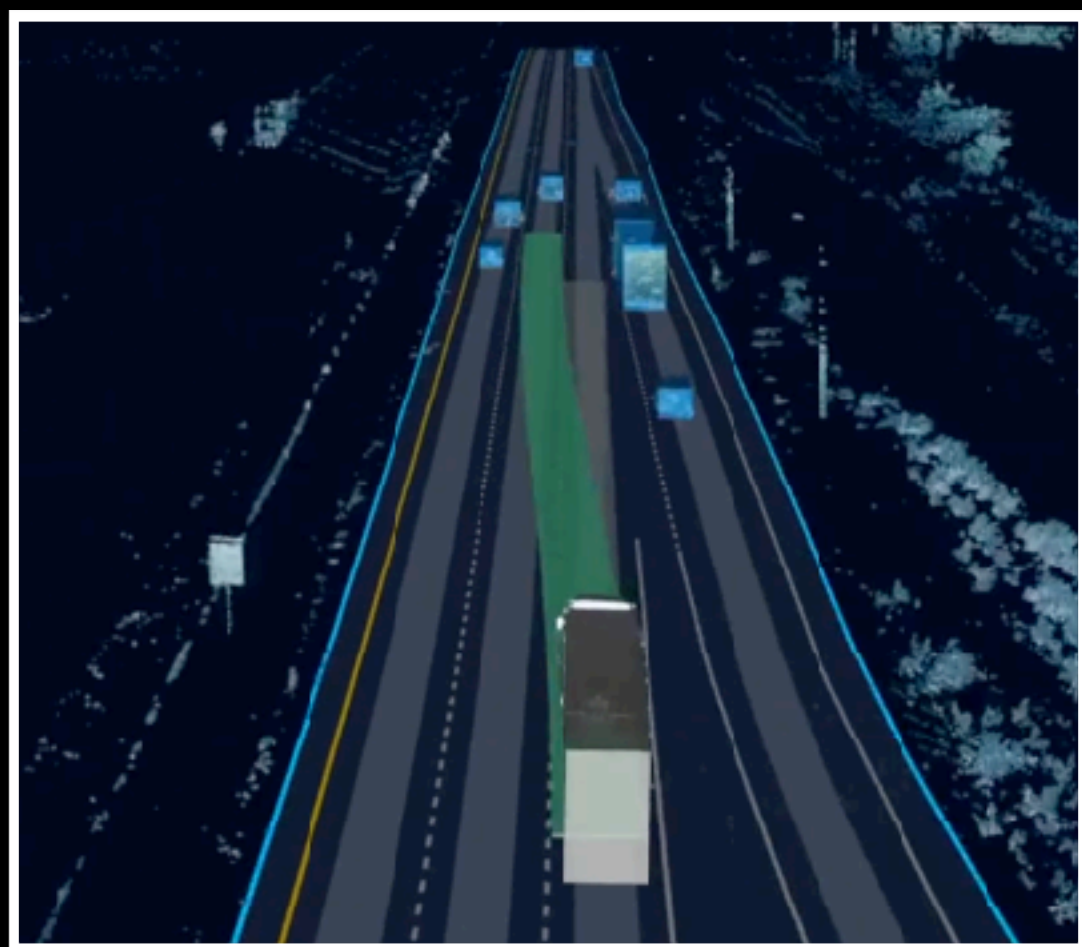
Data

*“What is the distribution of states?”*

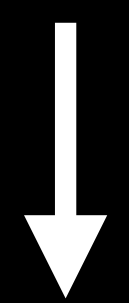
Loss

*“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

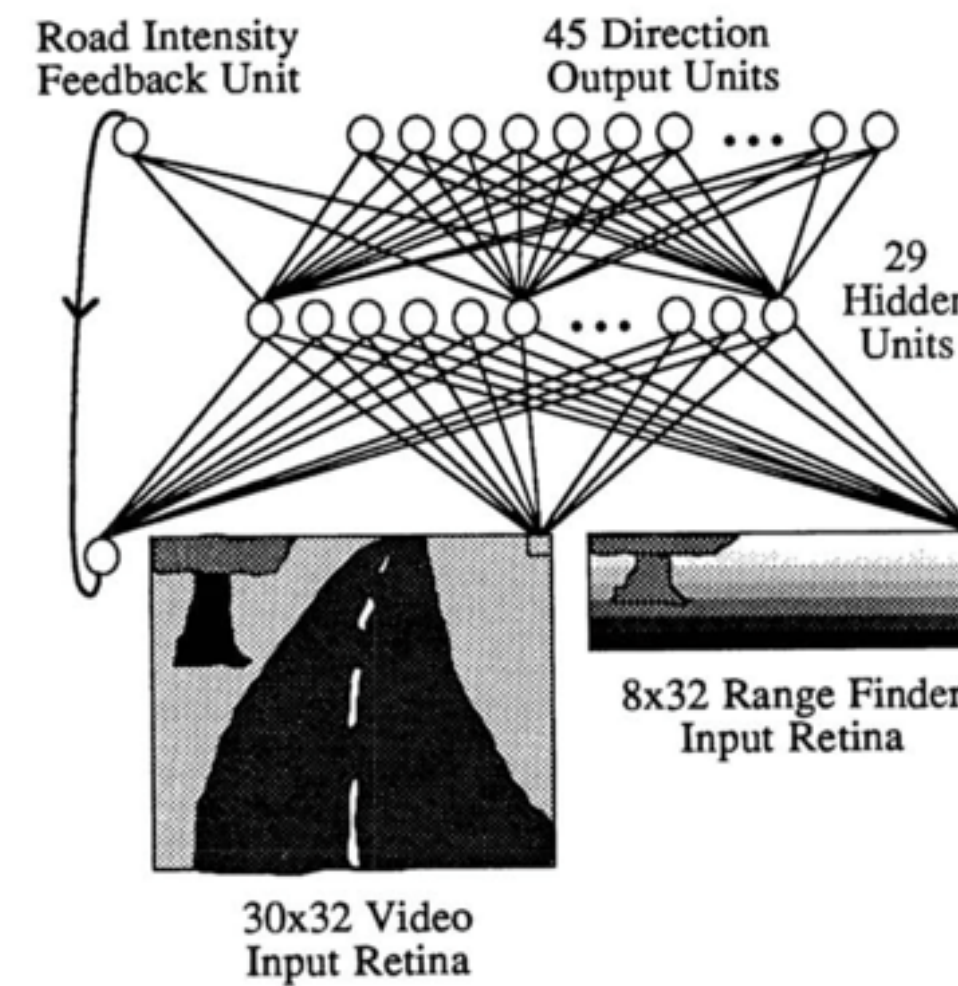


Figure 1: ALVINN Architecture

“...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 ...”

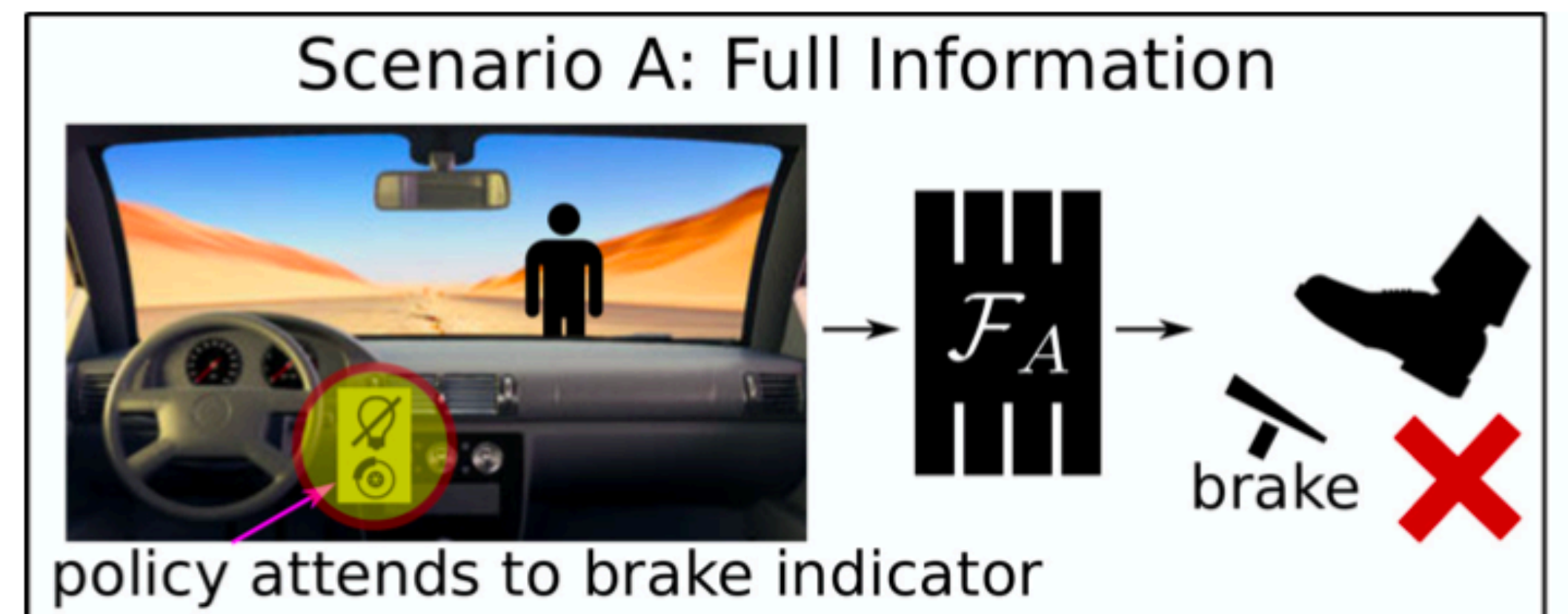
“Exploring the Limitations of Behavior Cloning for Autonomous Driving.”  
F. Codevilla, E. Santana, A. M. Lopez, A. Gaidon. ICCV 2019

“... 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

“... 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



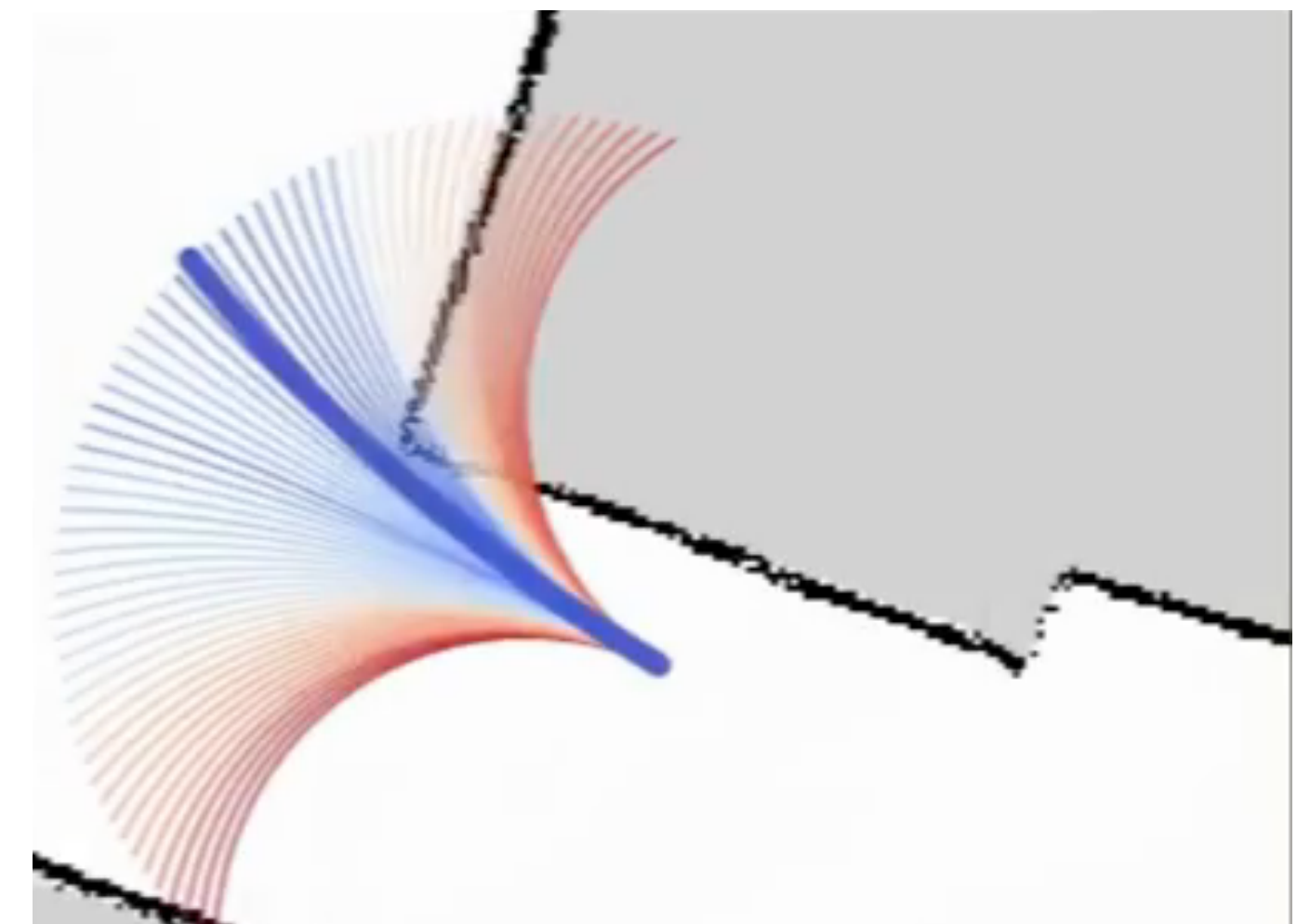
“Causal Confusion in Imitation Learning”.  
P. de Haan, D. Jayaraman, S. Levine, NeurIPS '19



# Feedback is an old adversary!



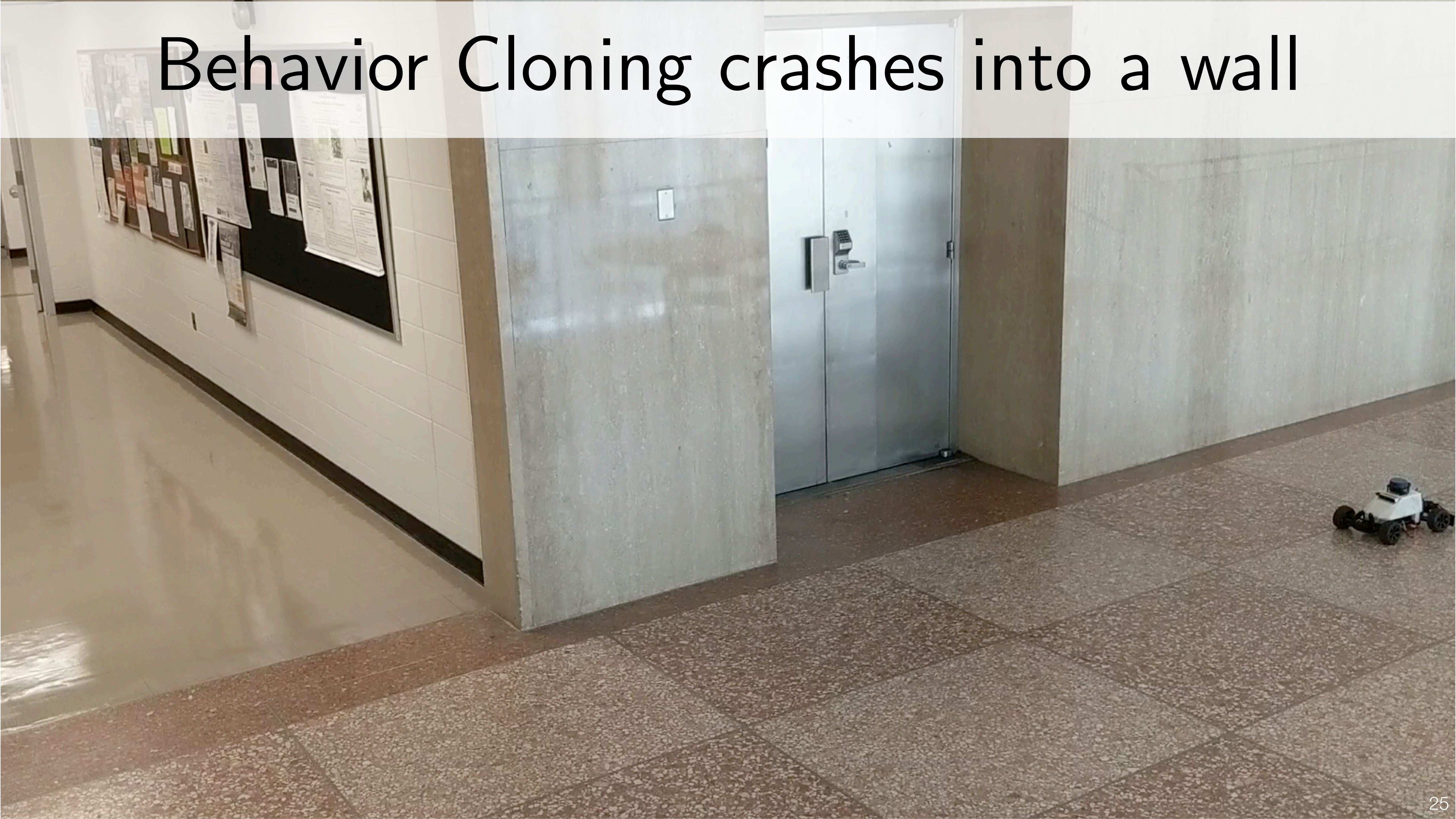
[SCB+ RSS'20]



Learnt policy

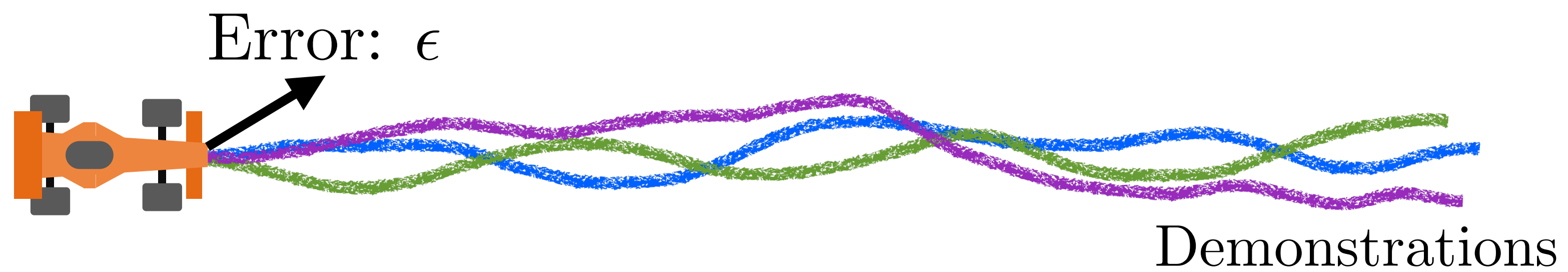


# Behavior Cloning crashes into a wall





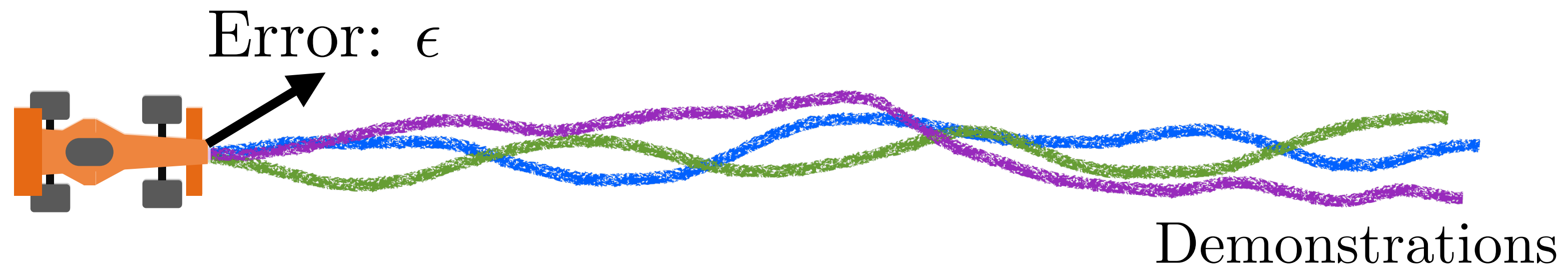
# Why did the robot crash?



# Why did the robot crash?



??  No training data  
Error: 1.0



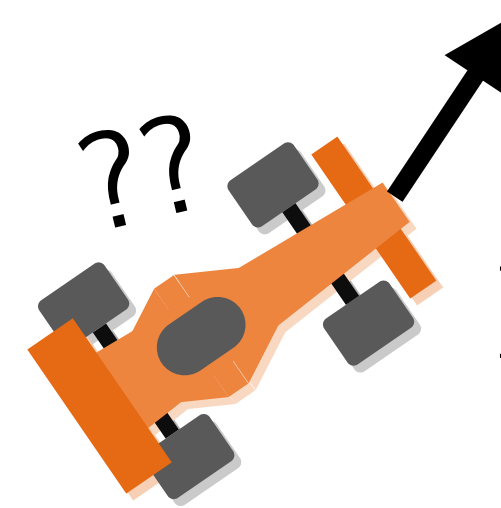


# Why did the robot crash?



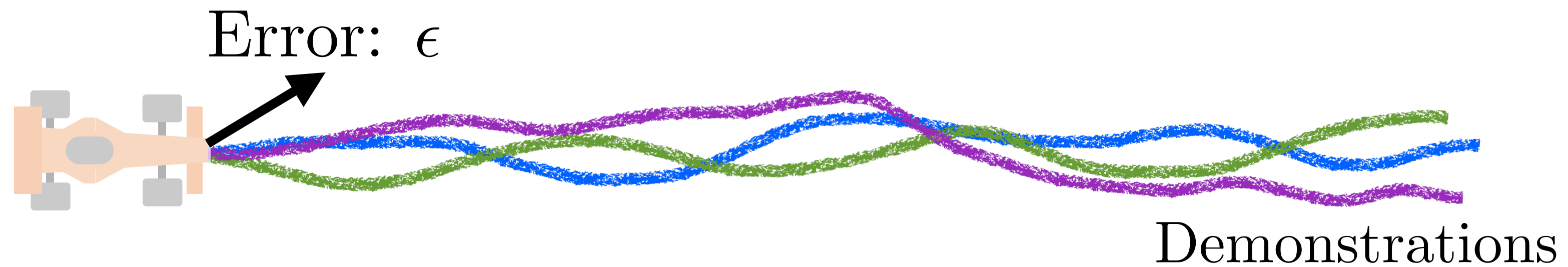
No training data

Error: 1.0



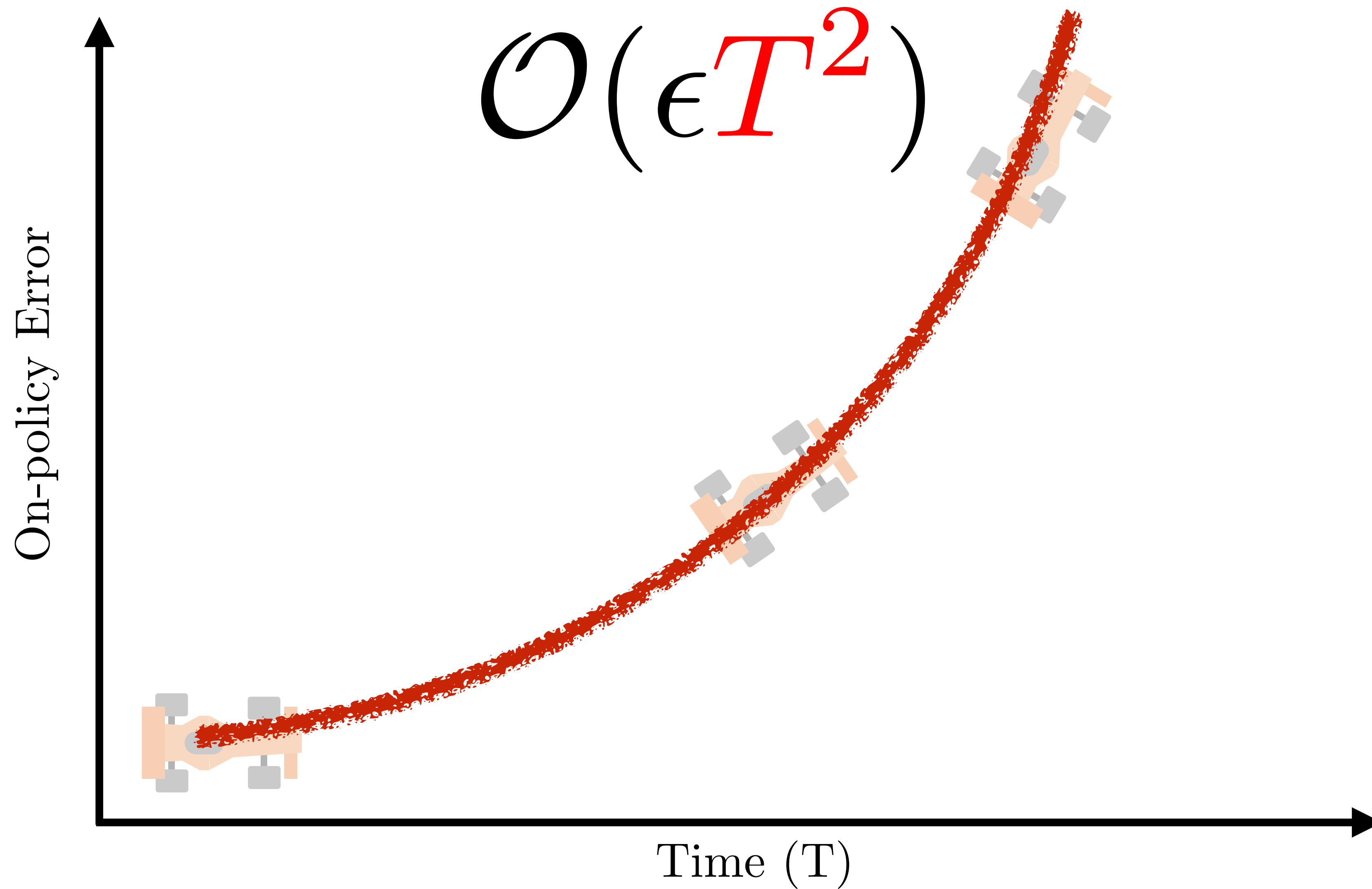
No training data

Error: 1.0



# Errors feedback and compound

[Ross and Bagnell'10]





Prove it!







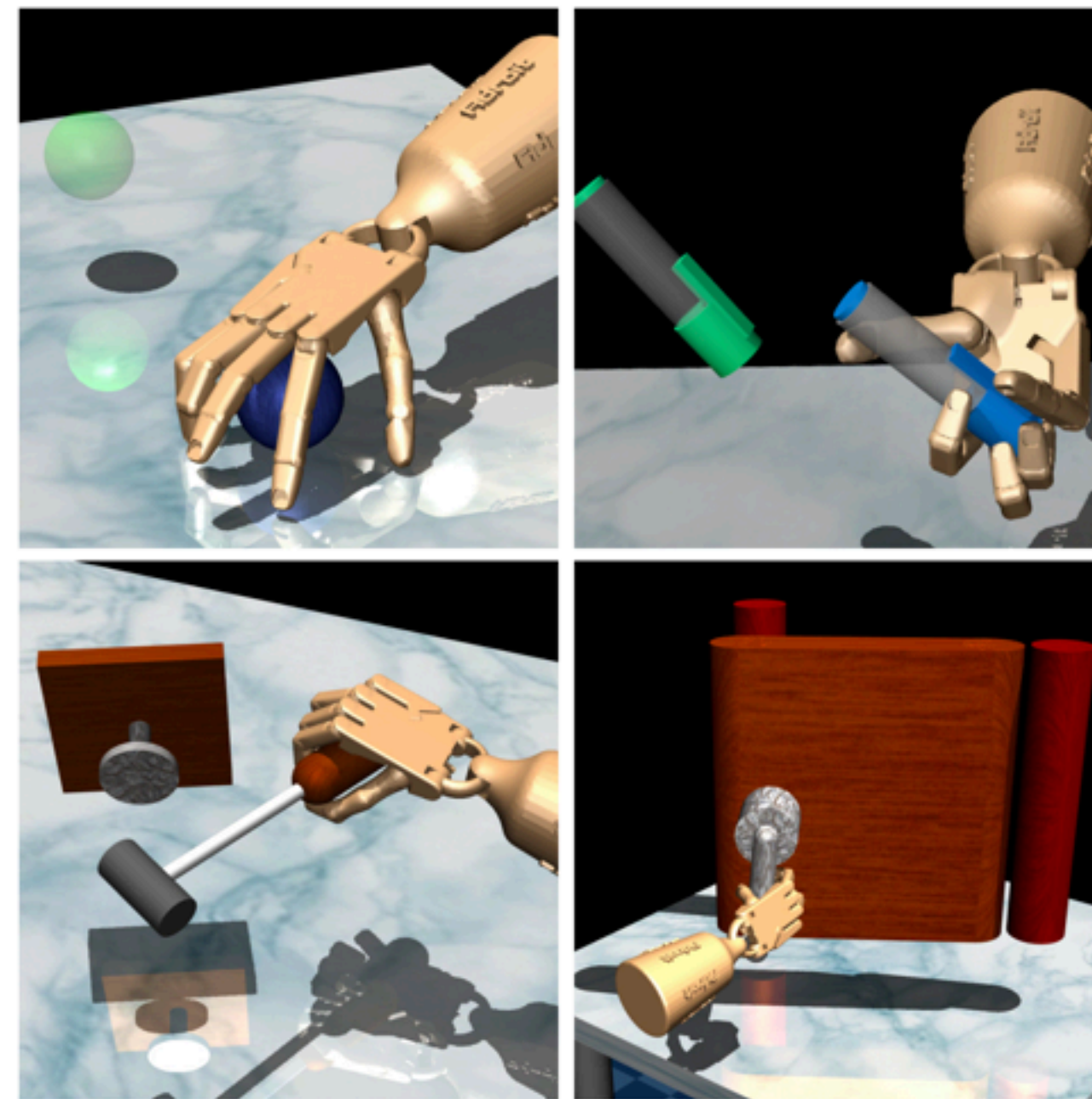
Feedback drives  
covariate shift



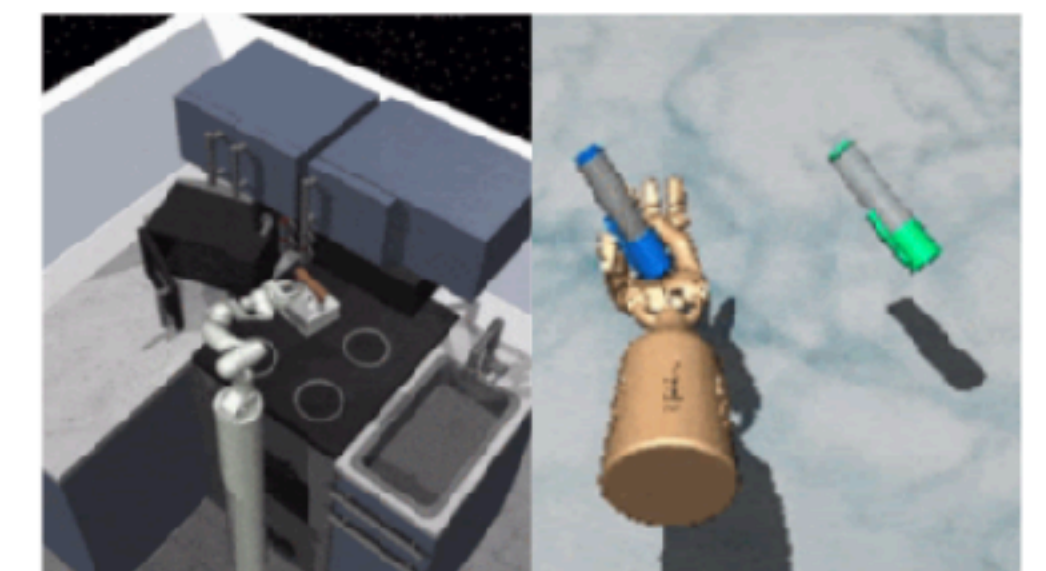
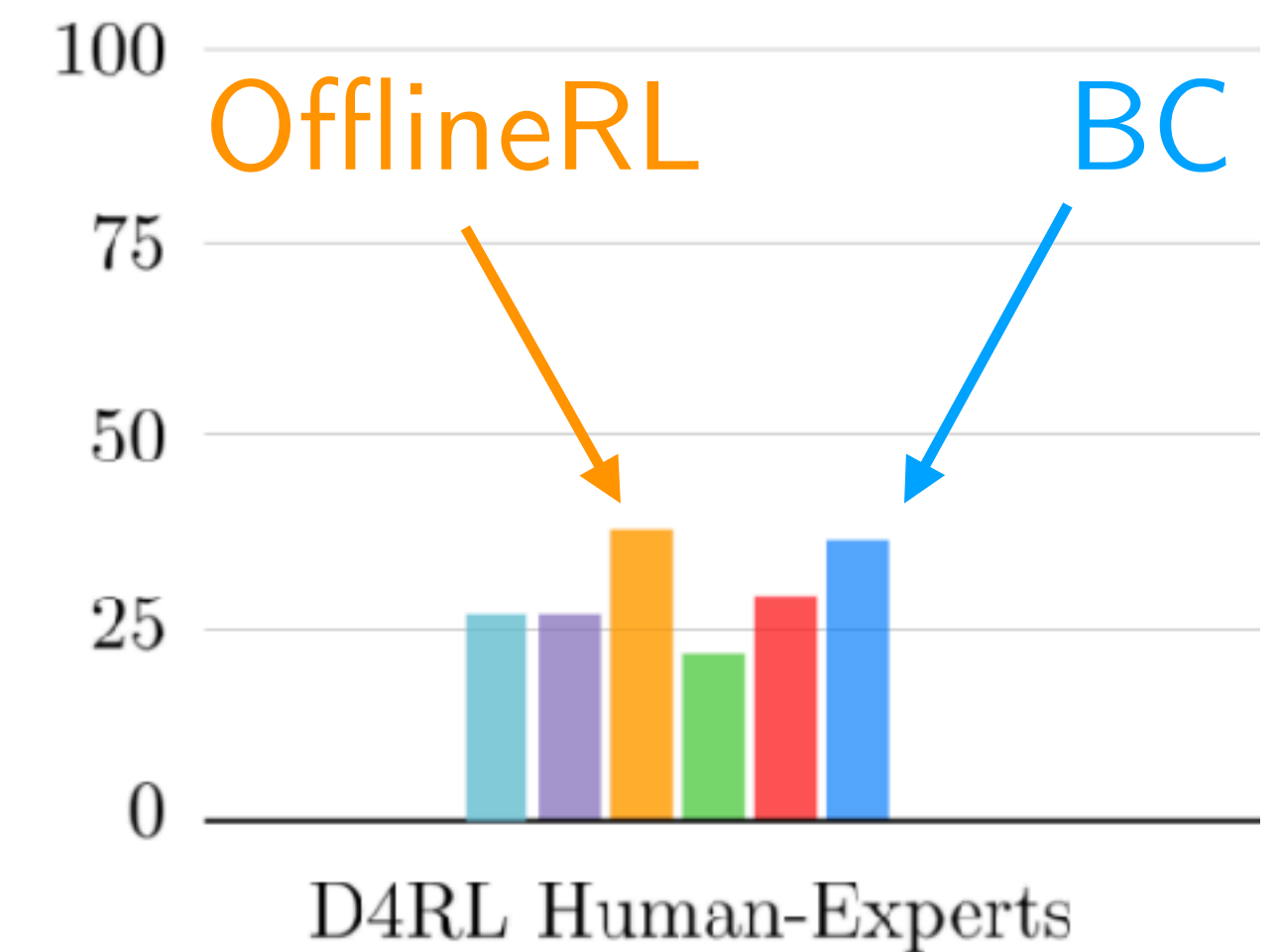
# But ... Behavior Cloning works just fine on benchmark datasets!

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

[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

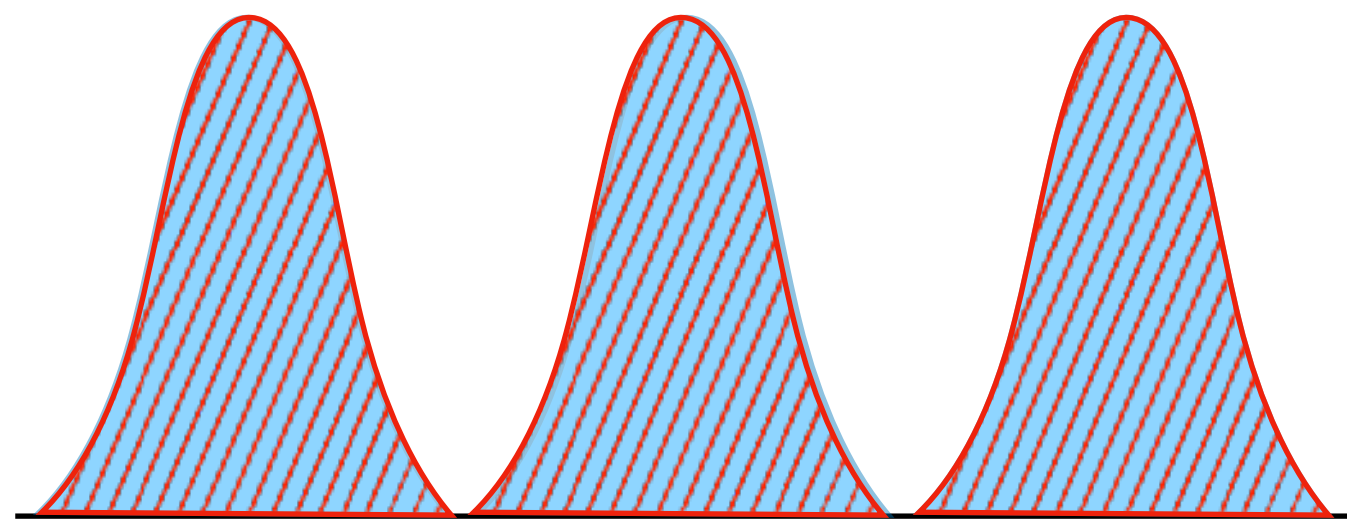


$$\text{Expert } \rho^{\pi^E}(s) \approx \text{Learner } \rho^{\pi}(s)$$

Expert is **realizable**

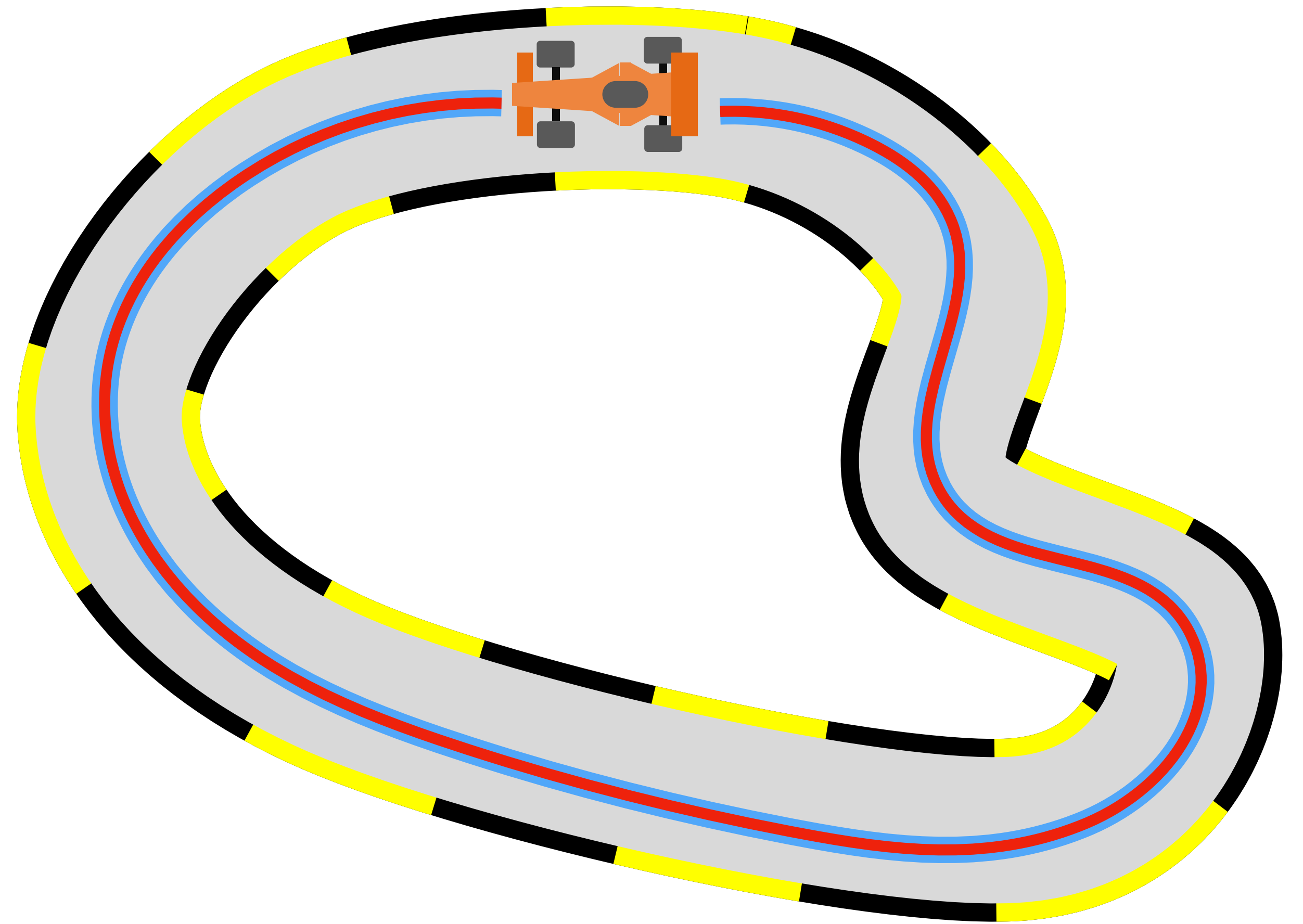
$$\pi^E \in \Pi$$

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



Nothing special.

Collect lots of data and  
do Behavior Cloning





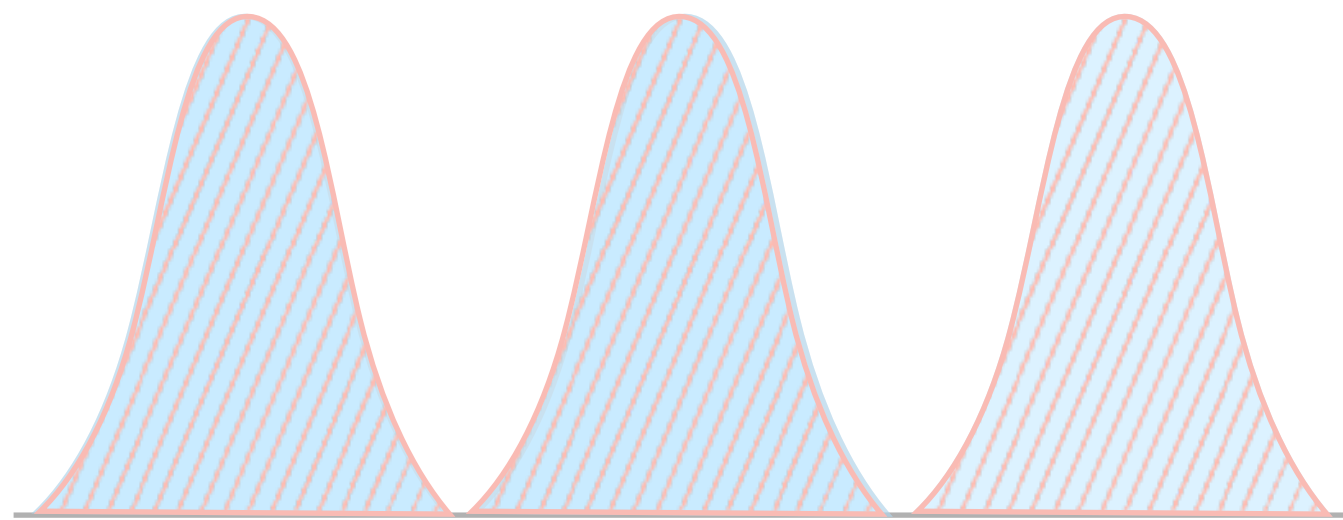
Easy



Expert is **realizable**

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

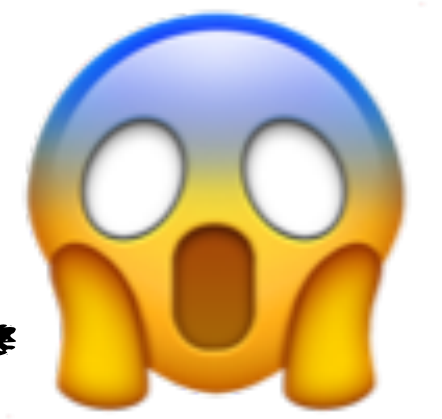
As  $N \rightarrow \infty$ , drive down  
 $\epsilon = 0$  (or Bayes error)



Nothing special.

Collect lots of data and  
do Behavior Cloning

Hard



Non-realizable expert +  
limited expert support

Setting

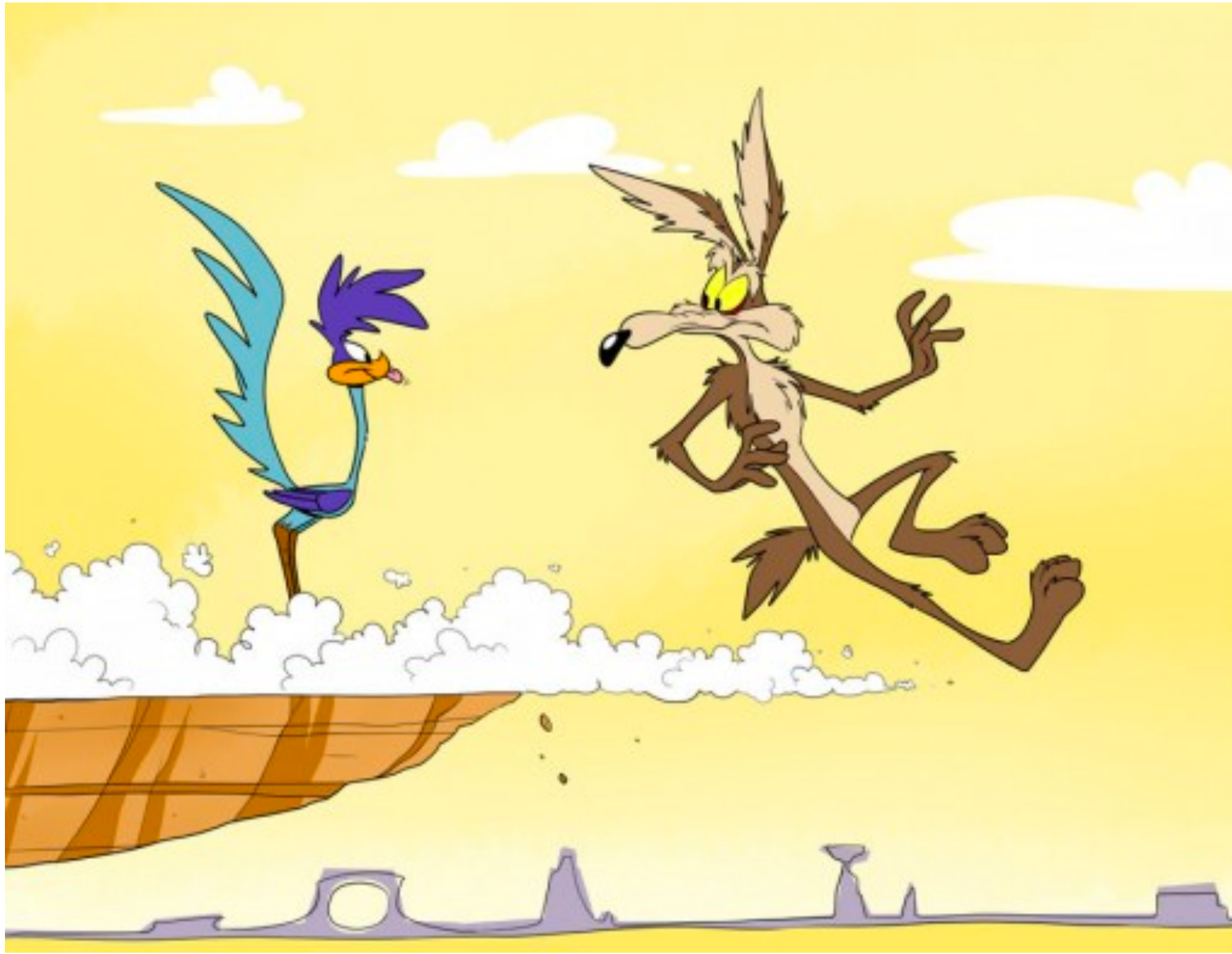
Solution







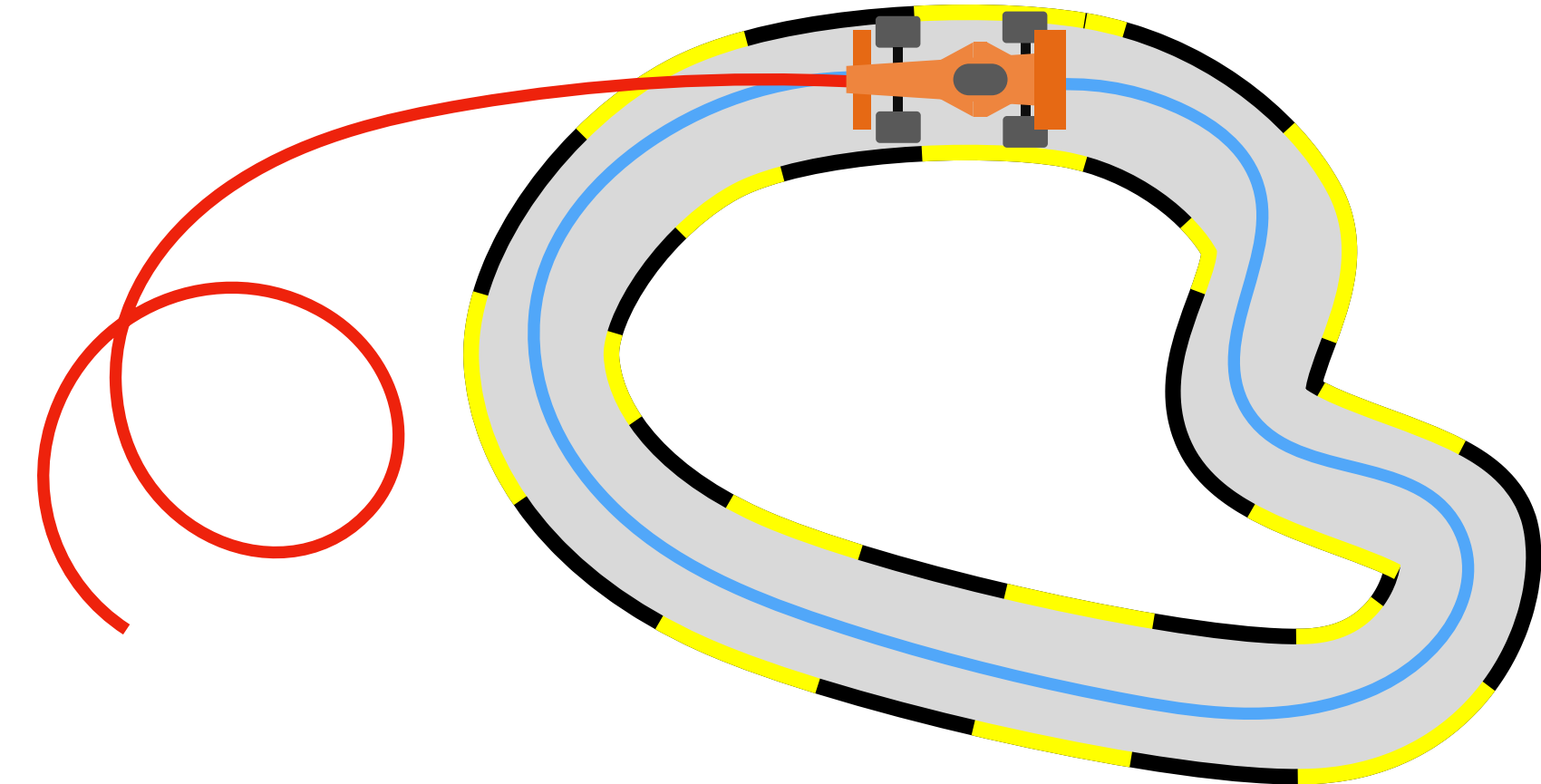
# Non-realizable expert + limited support?



Hard 🤪

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

[Ross & Bagnell '10]





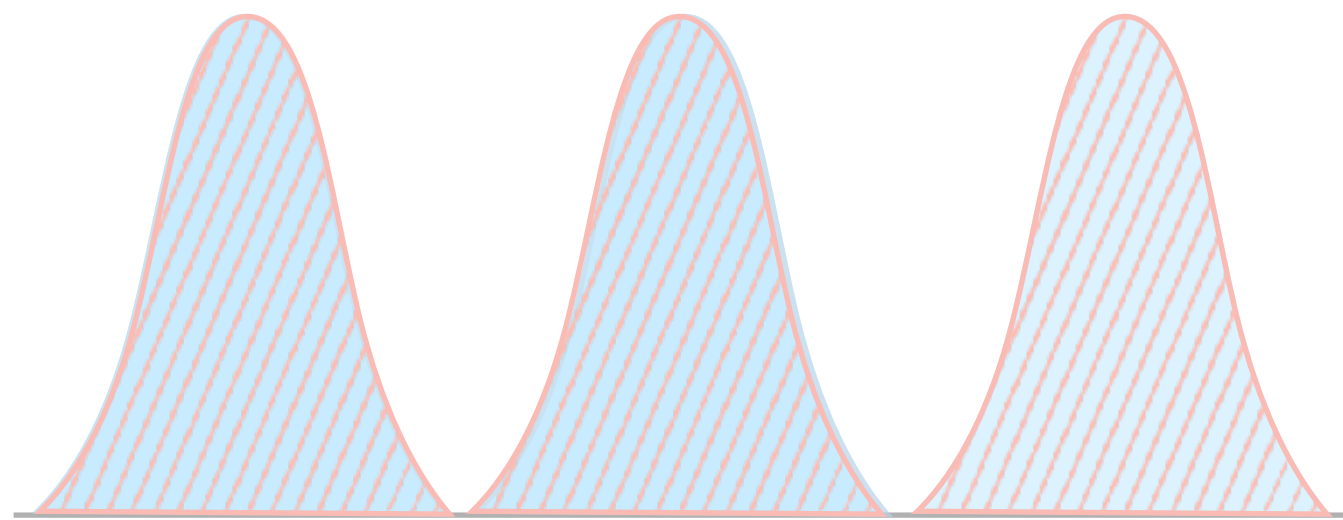
Easy



Expert is **realizable**

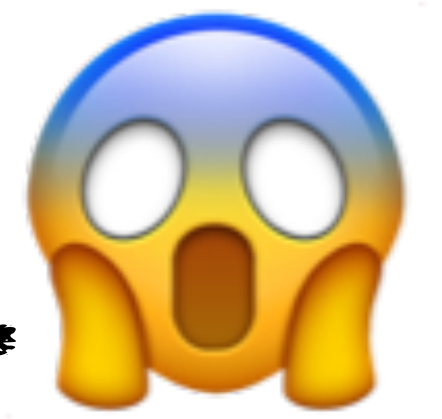
$$\pi^E \in \Pi$$

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



Nothing special.  
Collect lots of data and  
do Behavior Cloning

Hard



Non-realizable expert +  
limited expert support

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

