

# Humans Should not Be Obstacles

Anca Dragan

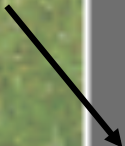
[anca@berkeley.edu](mailto:anca@berkeley.edu)

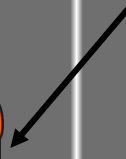
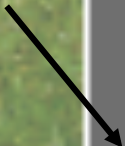


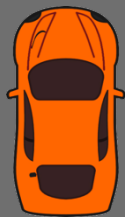
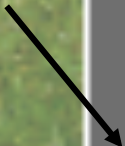
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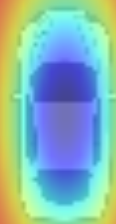






Compatibility = Stay out of the way

$$\max_{\mathbf{u}_R} \int U_R(x, \mathbf{u}_R, \mathbf{u}_H) P(\mathbf{u}_H | x) d\mathbf{u}_H$$





2015/02/06 23:09:54





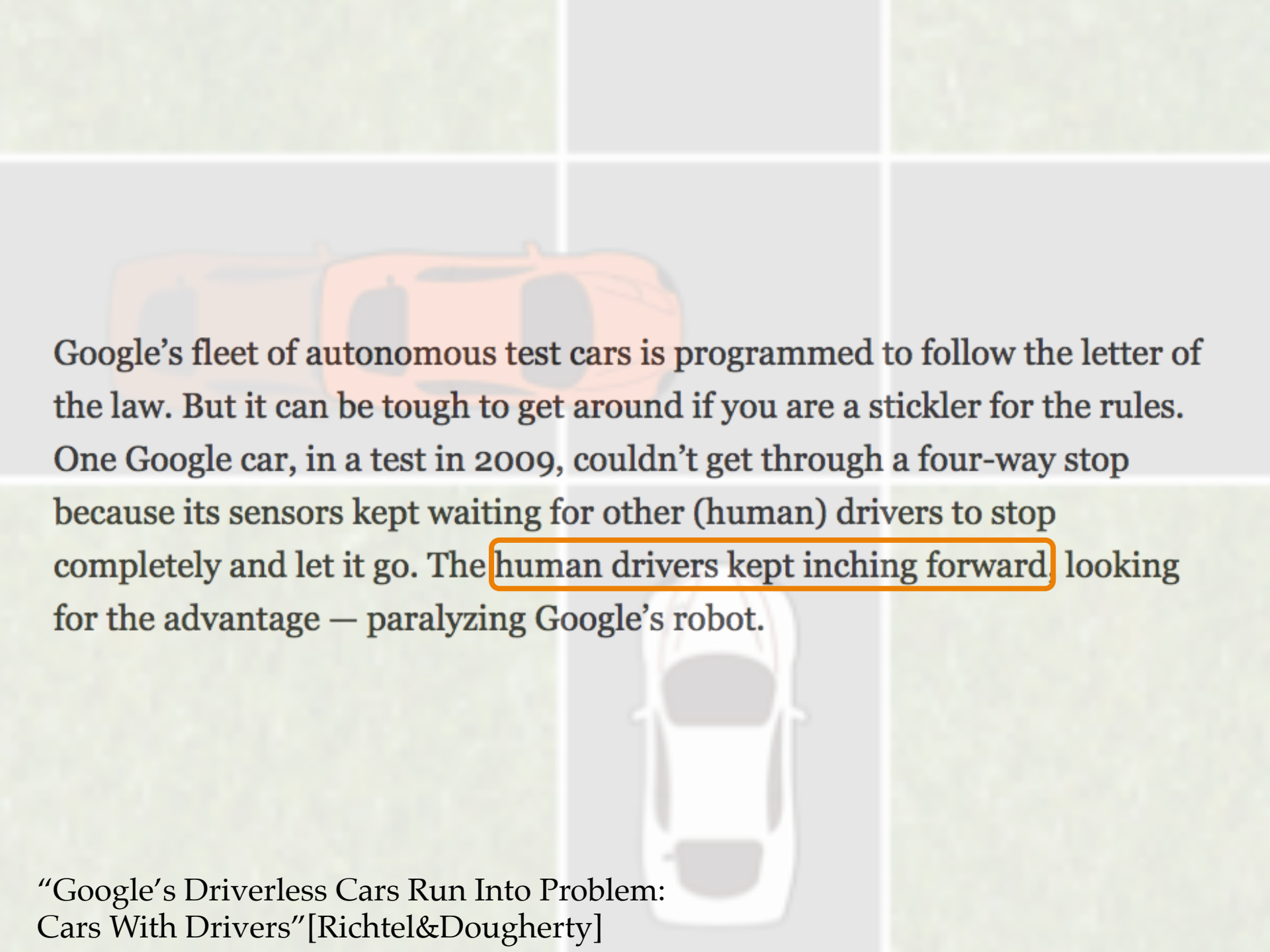
2015/02/06 23:10:06



Modeling people like moving obstacles is modeling them like this truck driver.

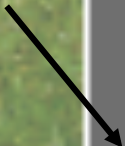


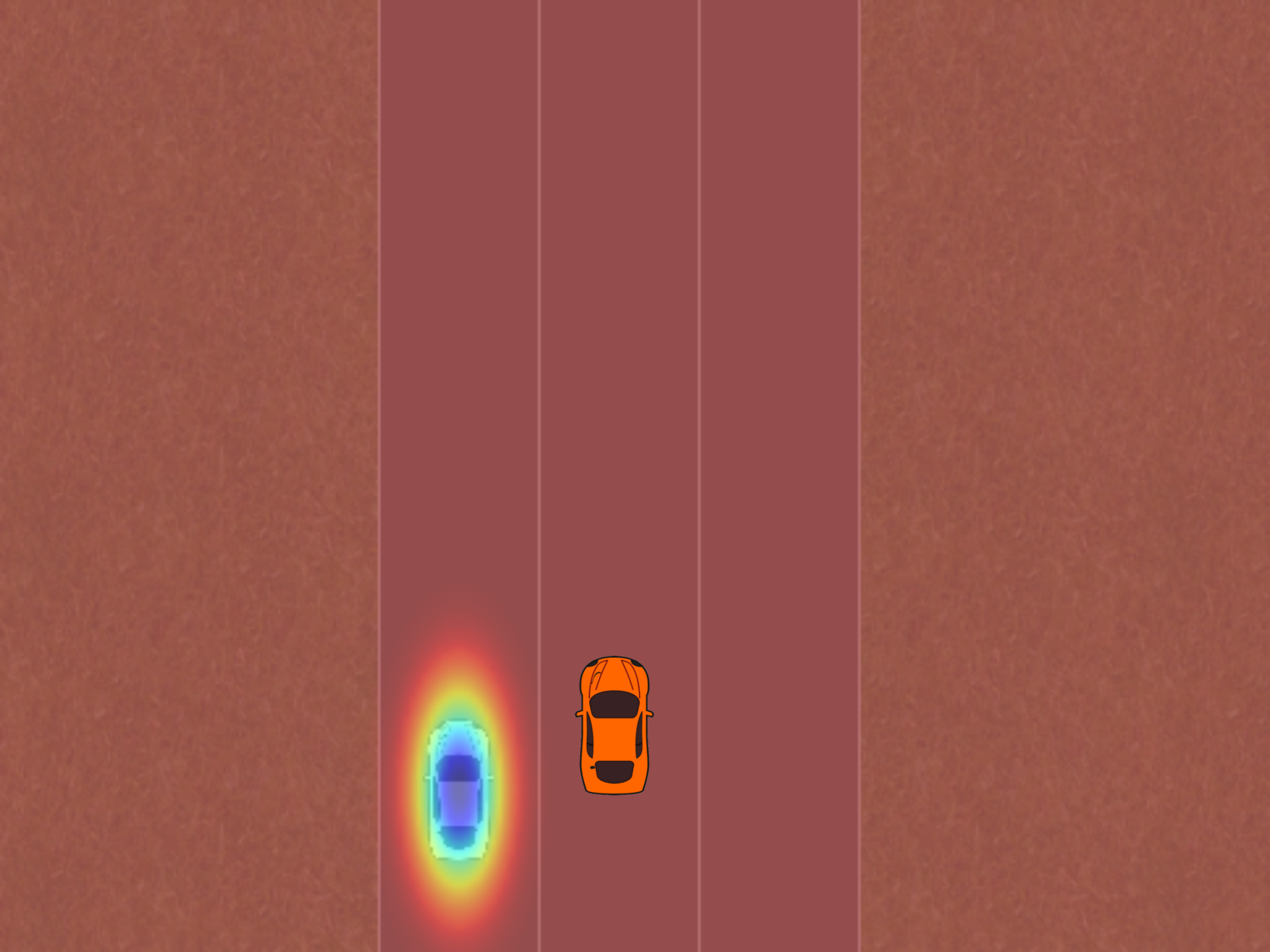
2015/02/06 23:10:08



Google's fleet of autonomous test cars is programmed to follow the letter of the law. But it can be tough to get around if you are a stickler for the rules. One Google car, in a test in 2009, couldn't get through a four-way stop because its sensors kept waiting for other (human) drivers to stop completely and let it go. The **human drivers kept inching forward** looking for the advantage — paralyzing Google's robot.

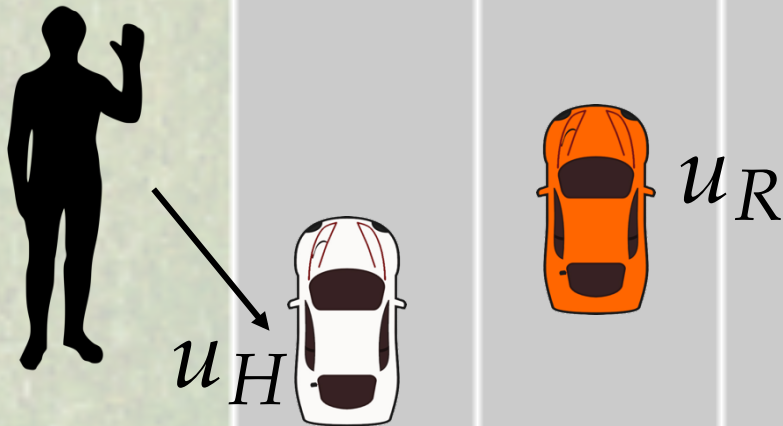
“Google's Driverless Cars Run Into Problem:  
Cars With Drivers” [Richtel&Dougherty]





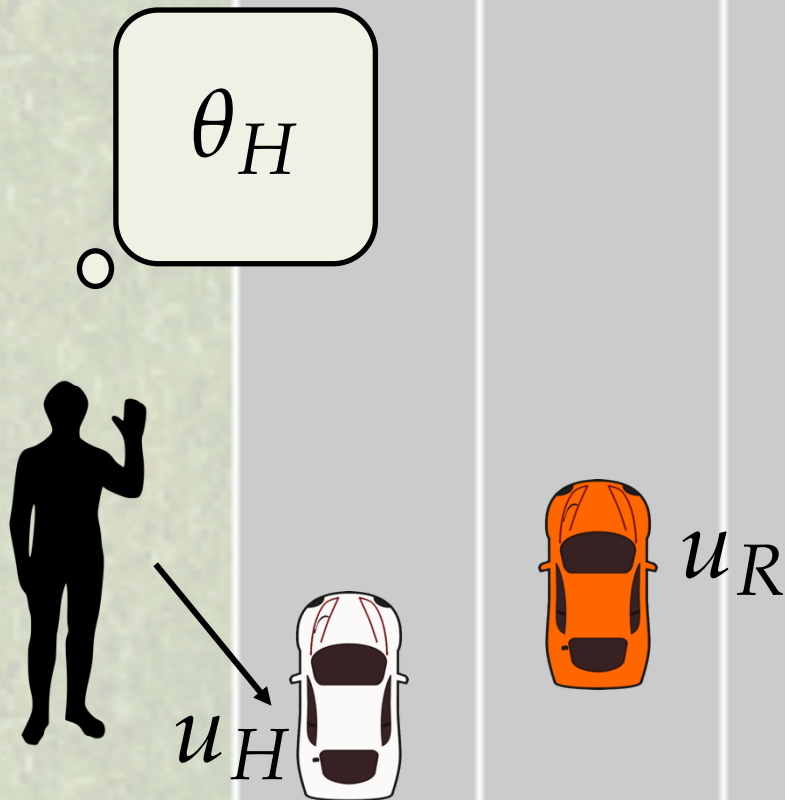
# A Human-Robot Game

$$U_H(x, u_R, u_H) \quad U_R(x, u_R, u_H)$$



# A Partial Information Human-Robot Game

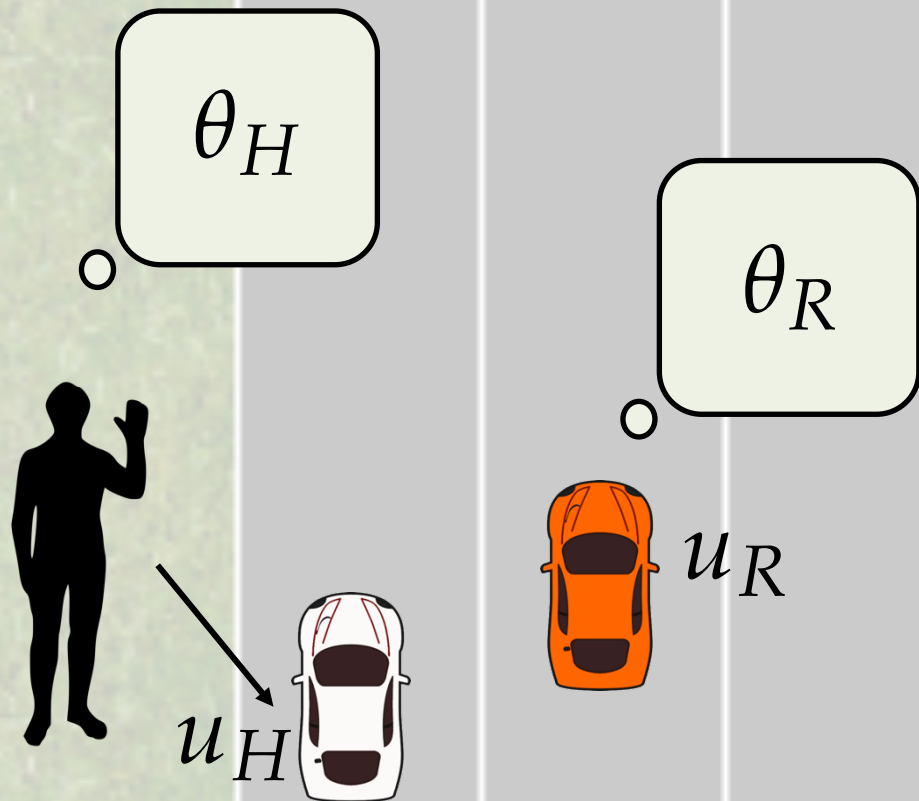
$$U_H(x, u_R, u_H; \theta_H) \quad U_R(x, u_R, u_H)$$





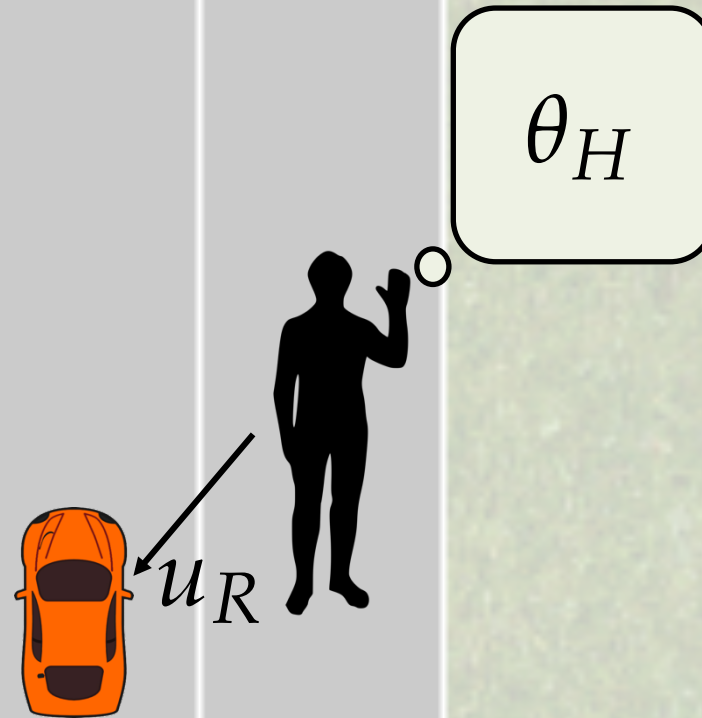
# A Partial Information Human-Robot Game

$$U_H(x, u_R, u_H; \theta_H) \quad U_R(x, u_R, u_H; \theta_R)$$



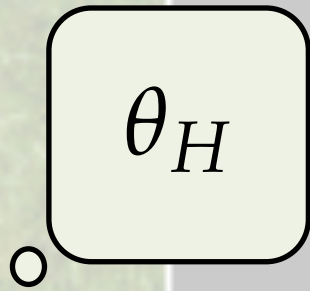
# A Partial Information Human-Robot Game

$$U(x, u_R, u_H; \theta_H)$$

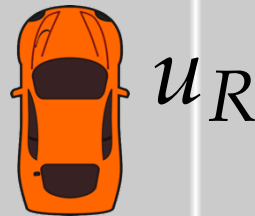
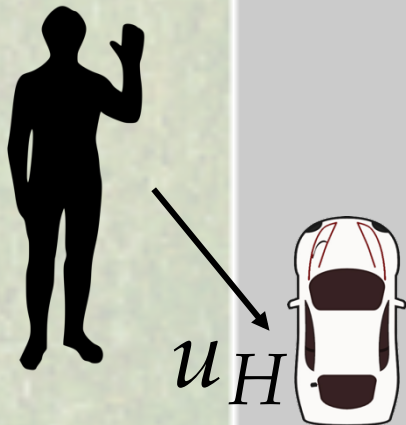


# A Partial Information Human-Robot Game

$$U_H(x, u_R, u_H; \theta_H) \quad U_R(x, u_R, u_H)$$



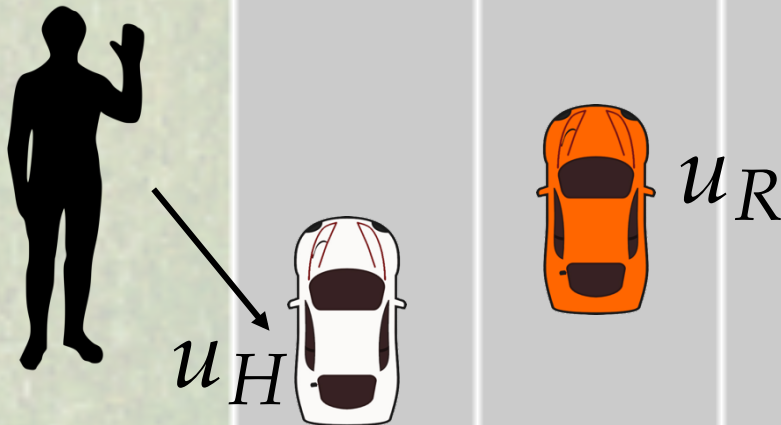
1. tractability, e.g. [Bernstein'02]
2. it's not how people work, e.g. [Hedden'02]



# Approximation as Underactuated System

$$U_H(x, u_R, u_H; \theta_H) \quad U_R(x, u_R, u_H)$$

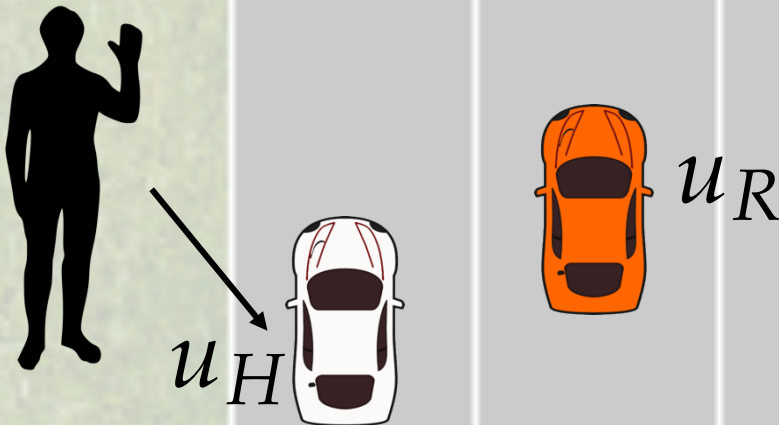
$$\theta_H \quad \mathbf{u}_H^* = \arg \max_{\mathbf{u}_H} U_H(x, \hat{\mathbf{u}}_R, \mathbf{u}_H; \theta_H)$$



# Approximation as Underactuated System

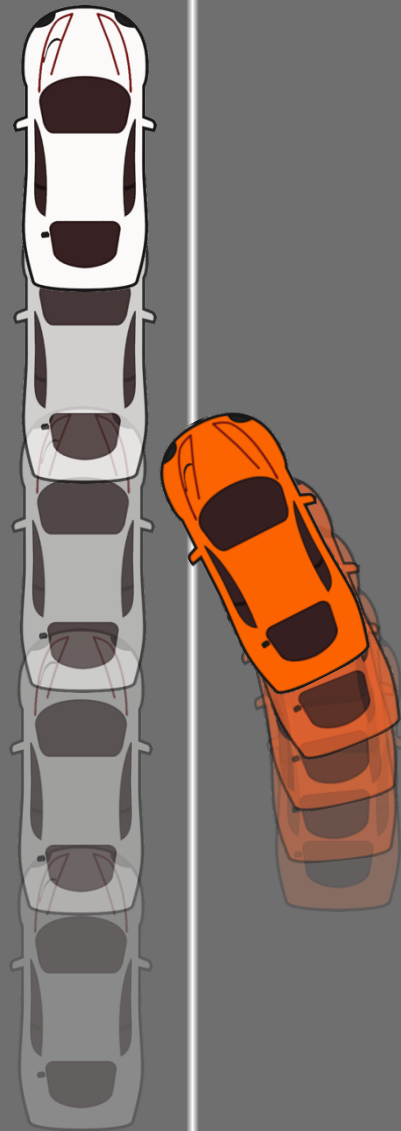
$$U_H(x, u_R, u_H; \theta_H) \quad U_R(x, u_R, u_H)$$

$$\theta_H \quad \mathbf{u}_H^* = \arg \max_{\mathbf{u}_H} U_H(x, \mathbf{u}_R, \mathbf{u}_H; \theta_H)$$

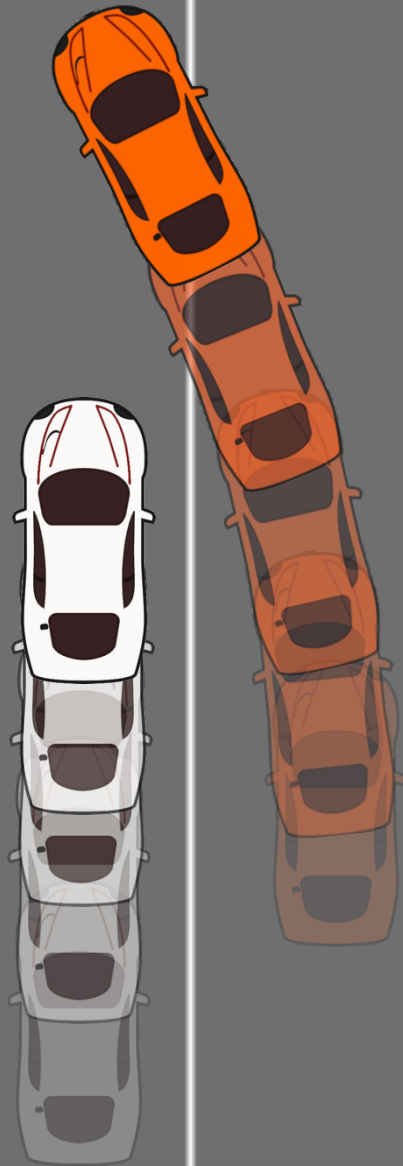




# Human as Obstacle



# Underactuated System, $U_H$ Learned Offline

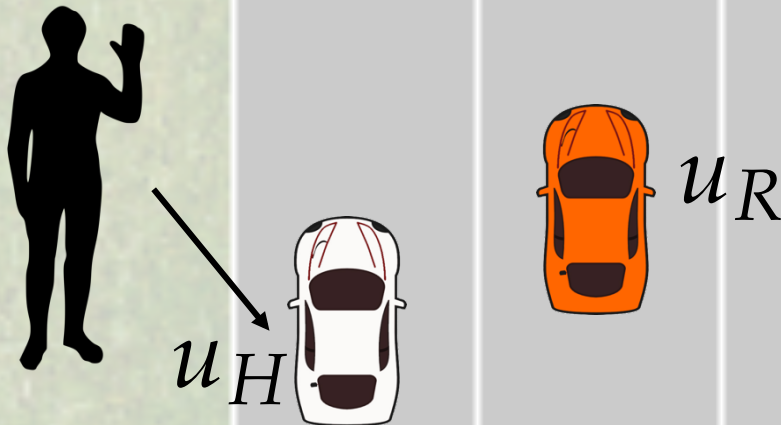




# Approximation as Underactuated System

$$U_H(x, u_R, u_H; \theta_H) \quad U_R(x, u_R, u_H)$$

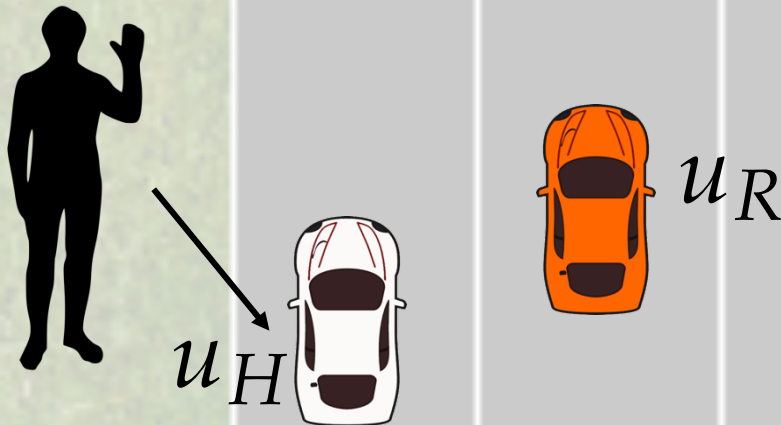
$$\theta_H \quad \mathbf{u}_H^* = \arg \max_{\mathbf{u}_H} U_H(x, \mathbf{u}_R, \mathbf{u}_H; \theta_H)$$



Also useful in collaboration, when H is myopic!

$$U(x, u_R, u_H; \theta_H)$$

$$\theta_H \quad u_H^* = \arg \max U_H(x, u_R, u_H; \theta_H)$$

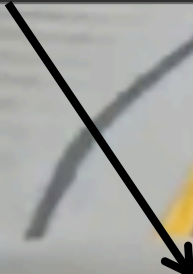


# Guide People to Better Plans in Collaboration

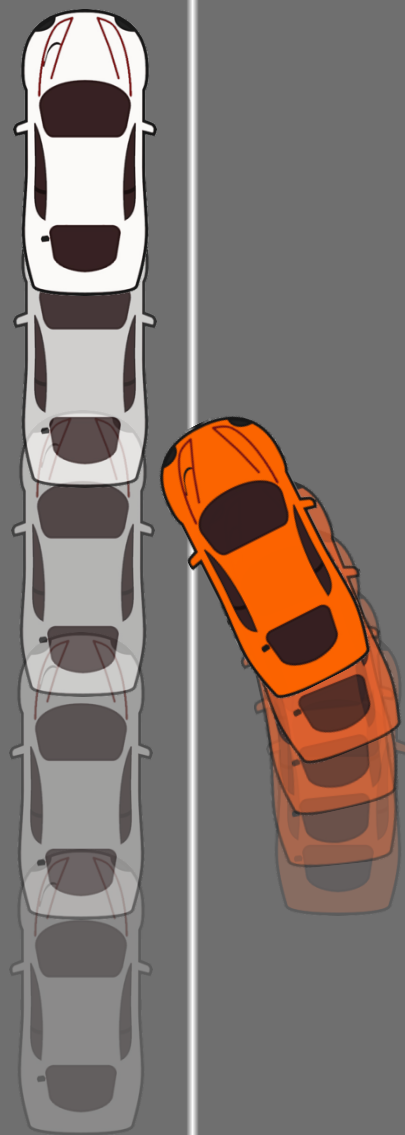
*Actively Assisting Humans to  
Use Good Grasps [ISER'16]*



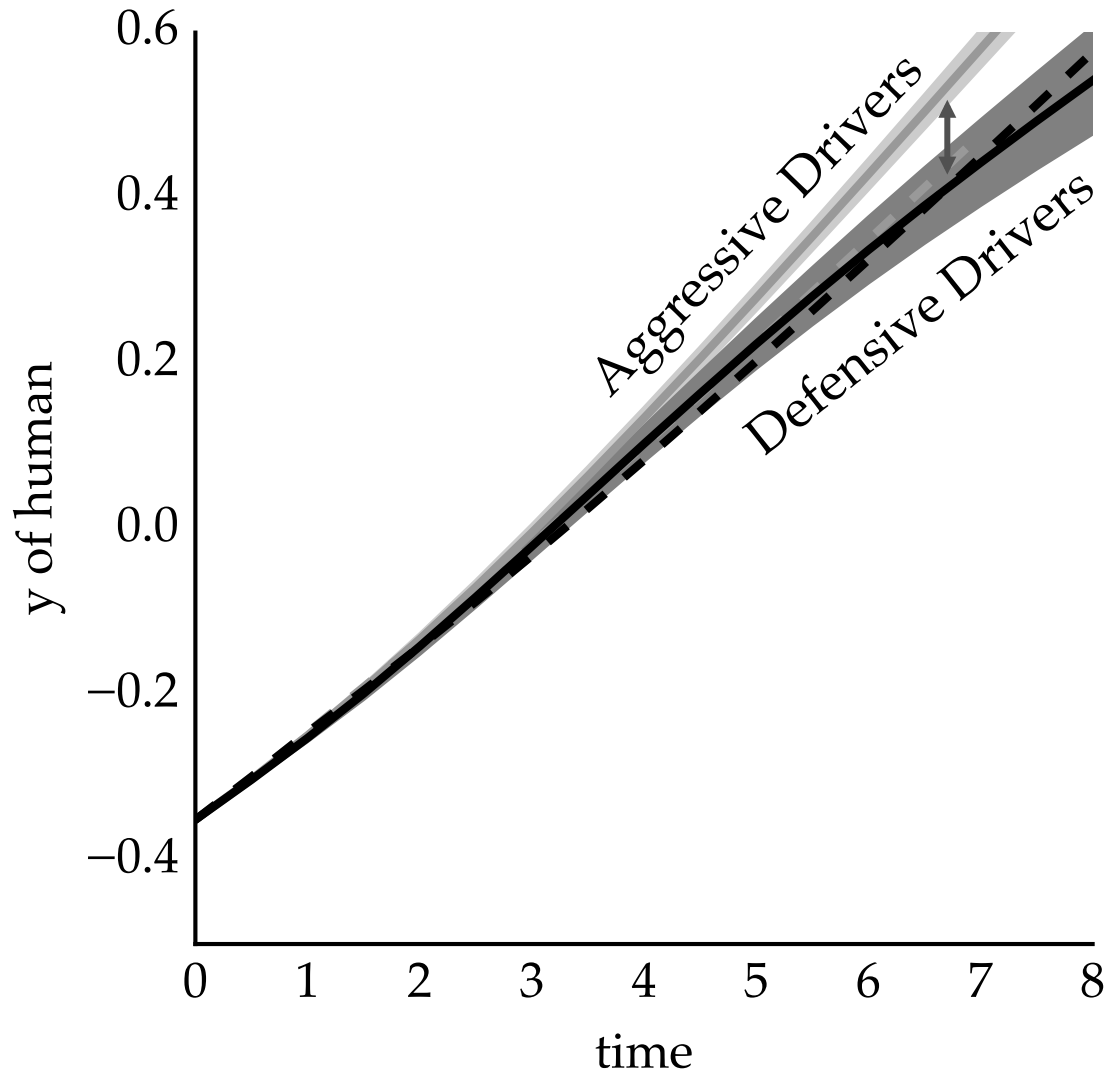
Big Limitation



# Underactuated System, Online $U_H$



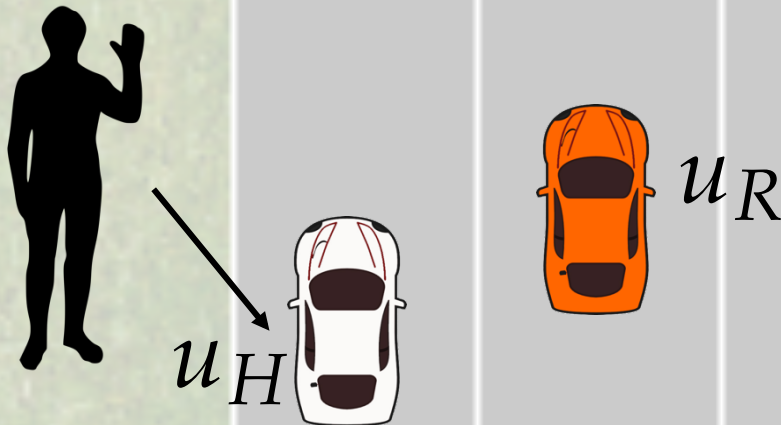
# All Users React in Almost the Same Way



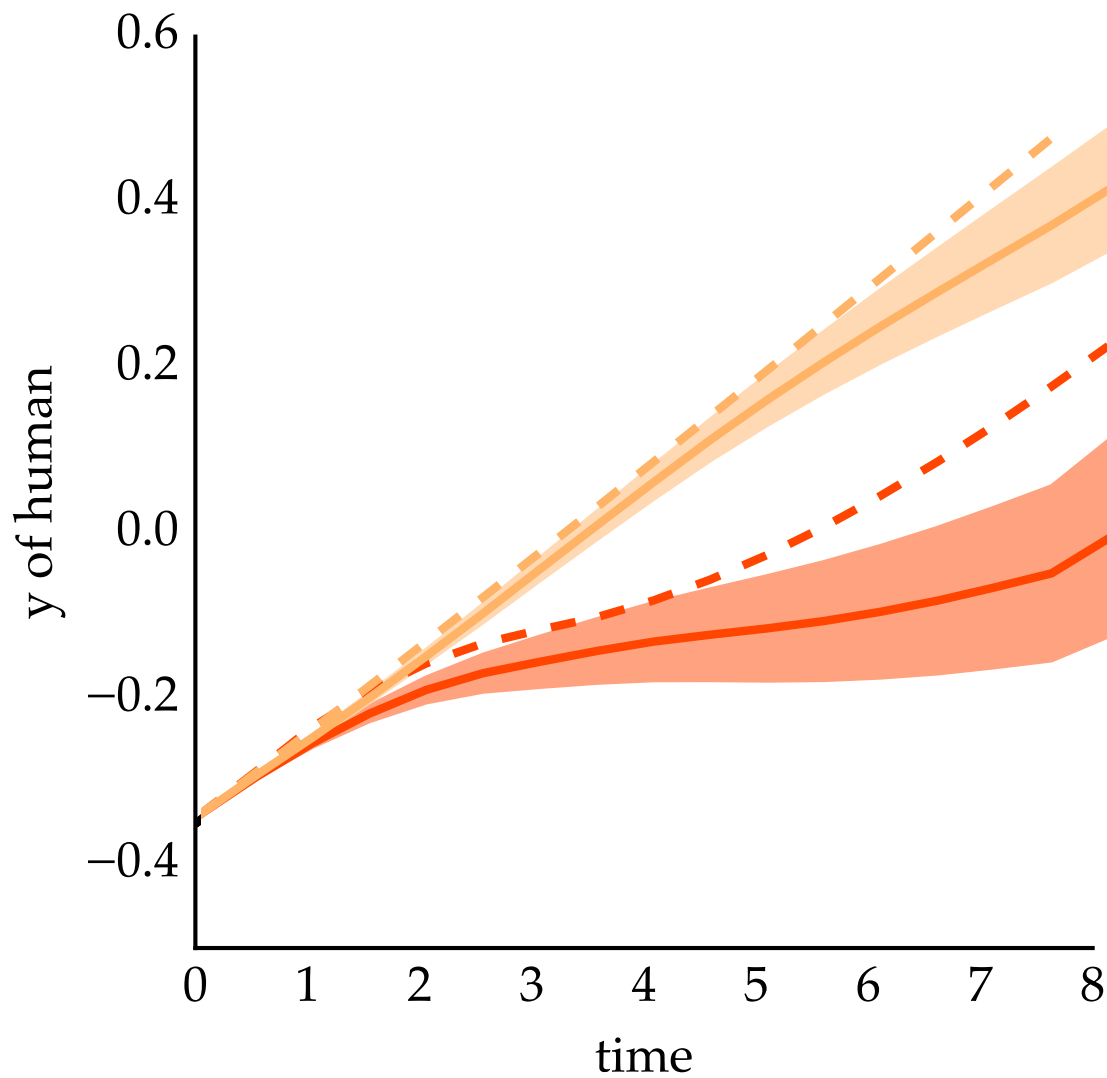
# Approximation as Underactuated System

$$U_H(x, u_R, u_H; \theta_H) \quad U_R(x, u_R, u_H)$$

$$\theta_H \quad \mathbf{u}_H^* = \arg \max_{\mathbf{u}_H} U_H(x, \mathbf{u}_R, \mathbf{u}_H; \theta_H)$$

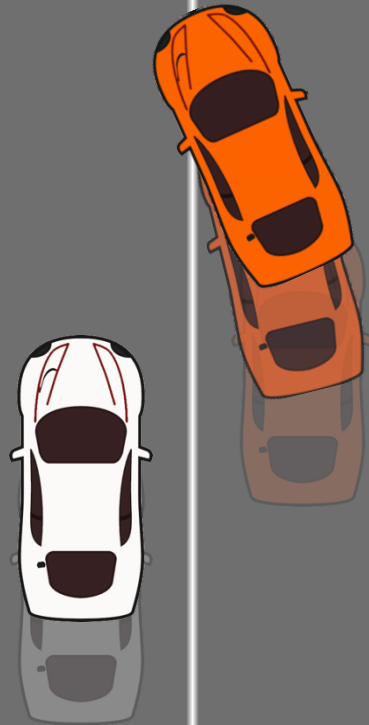


Idea: Leverage the robot's actions!

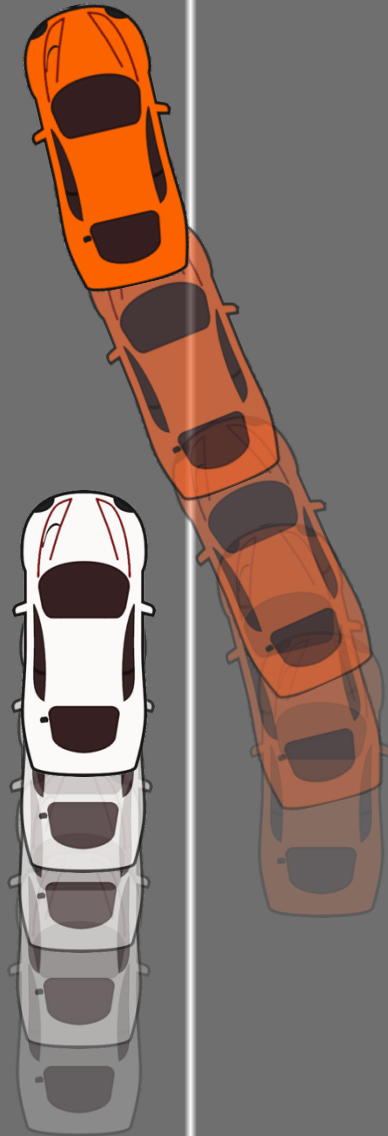




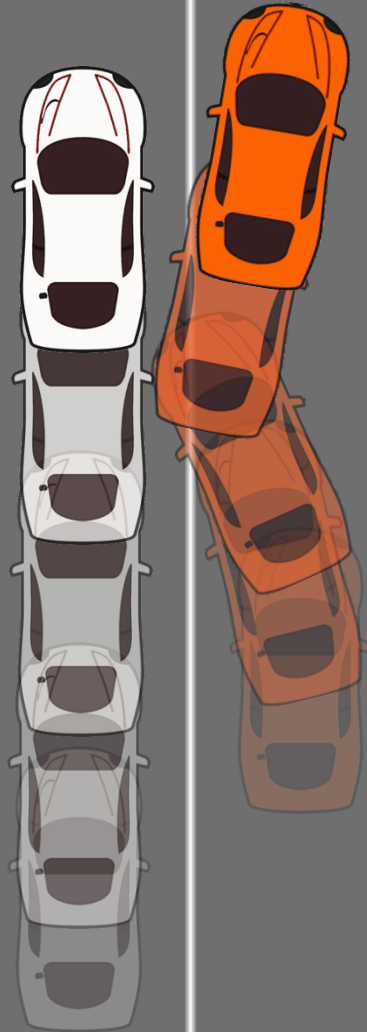
# Underactuated System, Active $U_H$



# Underactuated System, Active $U_H$ - *Defensive*

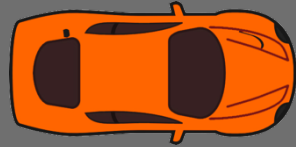


# Underactuated System, Active $U_H$ - *Aggressive*

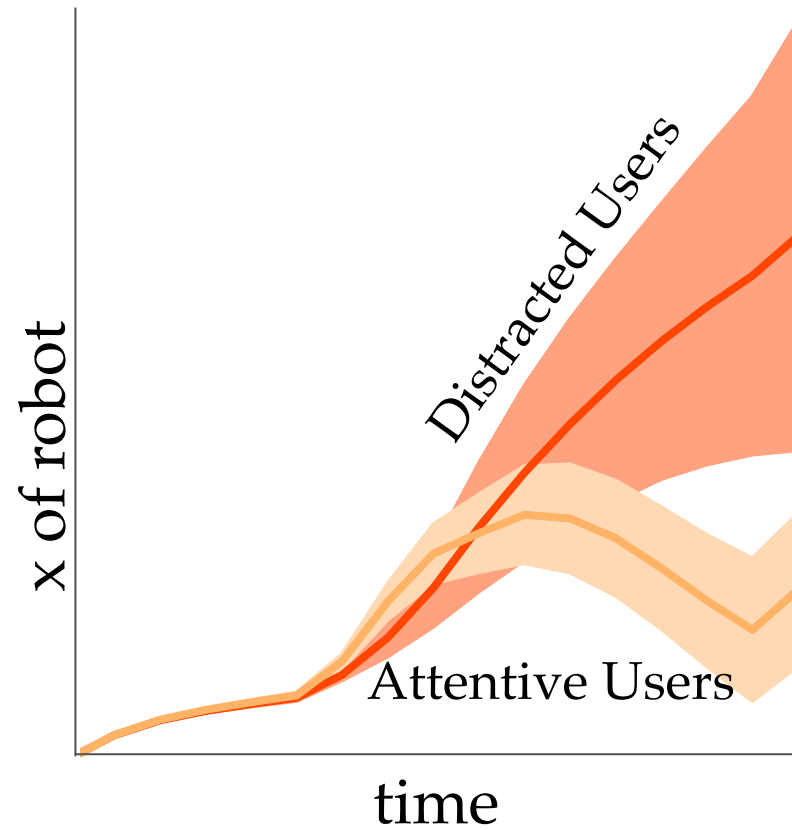


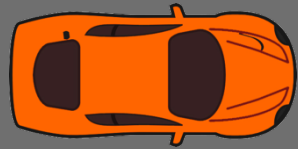


# Coordination at 4-Way Stops



# Robot Trajectories

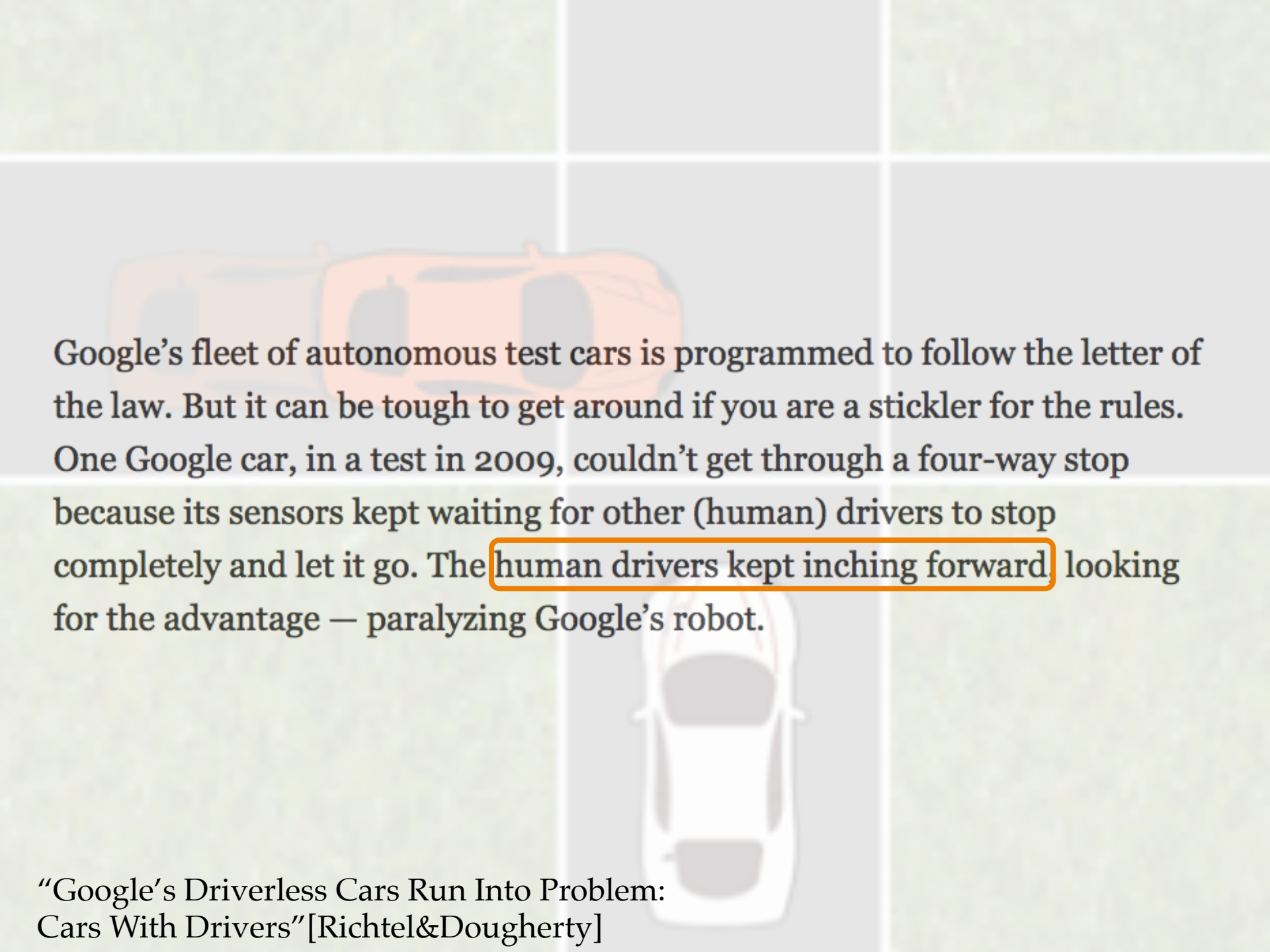




# Inch Forward



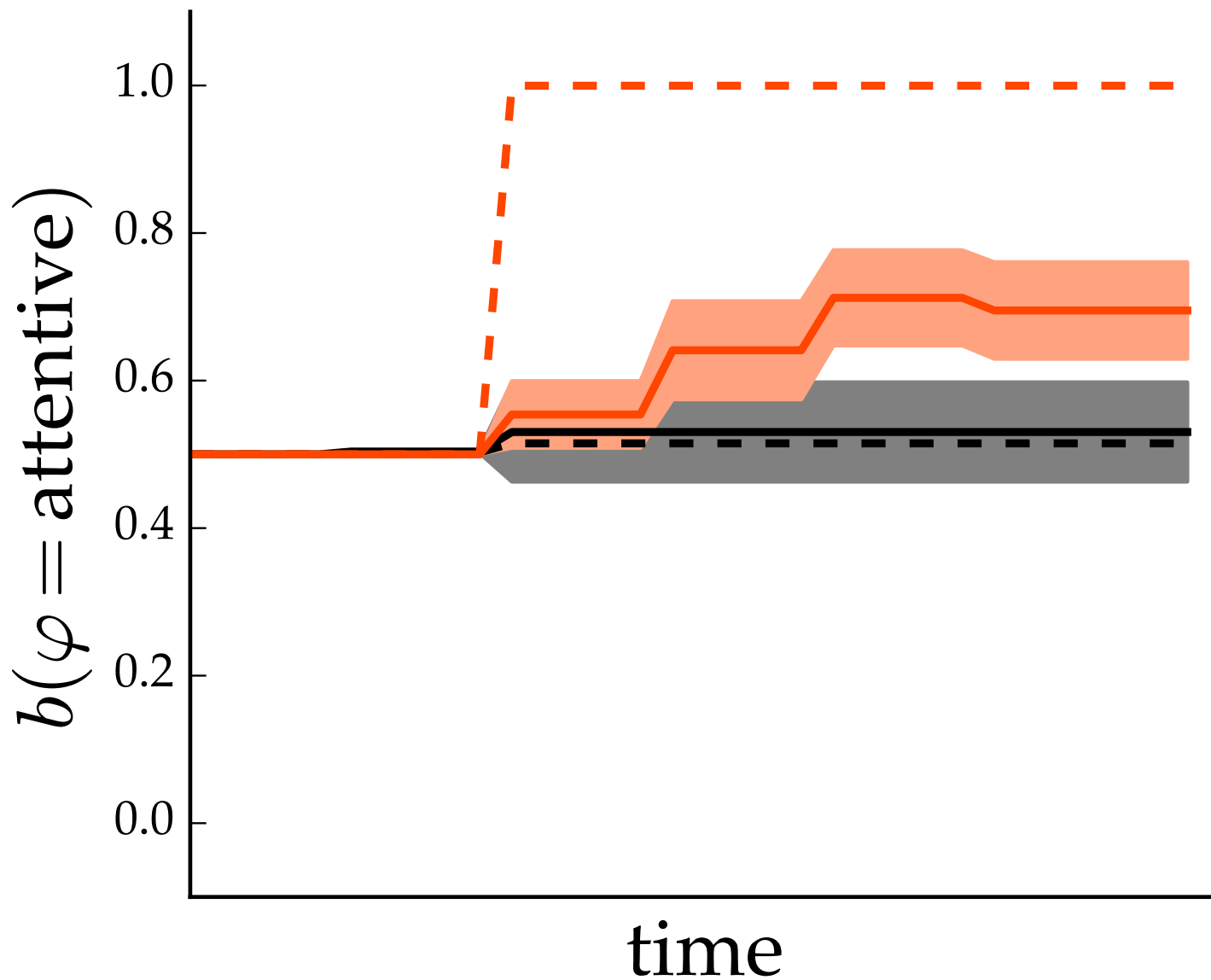


The background of the slide features a faint, semi-transparent image of a car at a four-way stop intersection. The car is shown from a side-top perspective, appearing to be in the middle of the intersection. The intersection is marked with white lines on a light-colored road surface. The overall background has a light green and grey color scheme.

Google's fleet of autonomous test cars is programmed to follow the letter of the law. But it can be tough to get around if you are a stickler for the rules. One Google car, in a test in 2009, couldn't get through a four-way stop because its sensors kept waiting for other (human) drivers to stop completely and let it go. The **human drivers kept inching forward** looking for the advantage — paralyzing Google's robot.

“Google's Driverless Cars Run Into Problem:  
Cars With Drivers” [Richtel&Dougherty]

# Info Gathering Improves Estimation



# Attentive Users: Continue



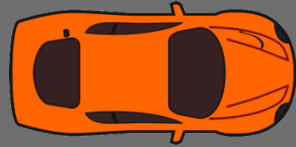
# Inch Forward



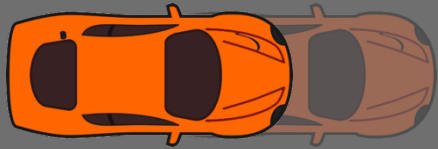
# Distracted Users: Go Back



$U_R$ : Human Should Go First



$U_R$ : Human Should Go First



Are you going? Or should I go?

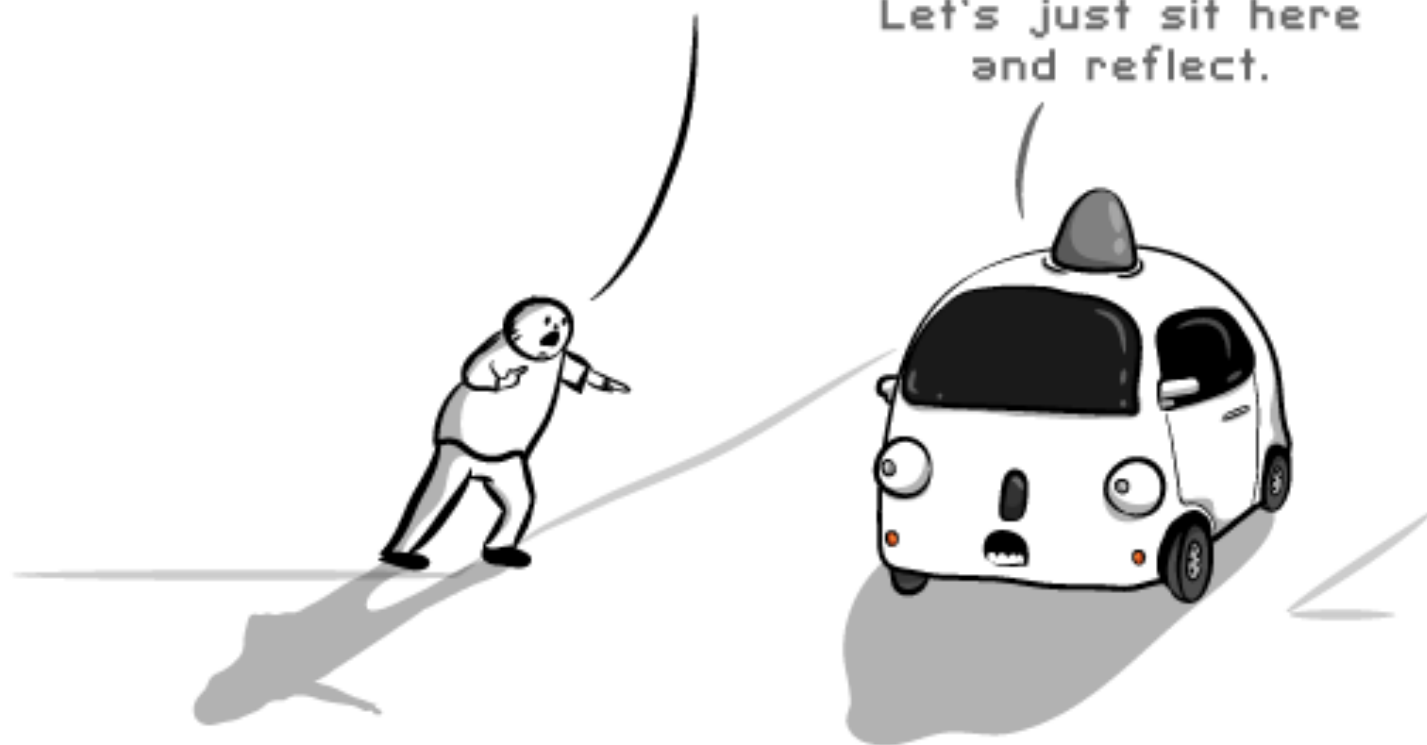
You go first.

What if I point a lot  
and flail my arms around?

This is confusing.

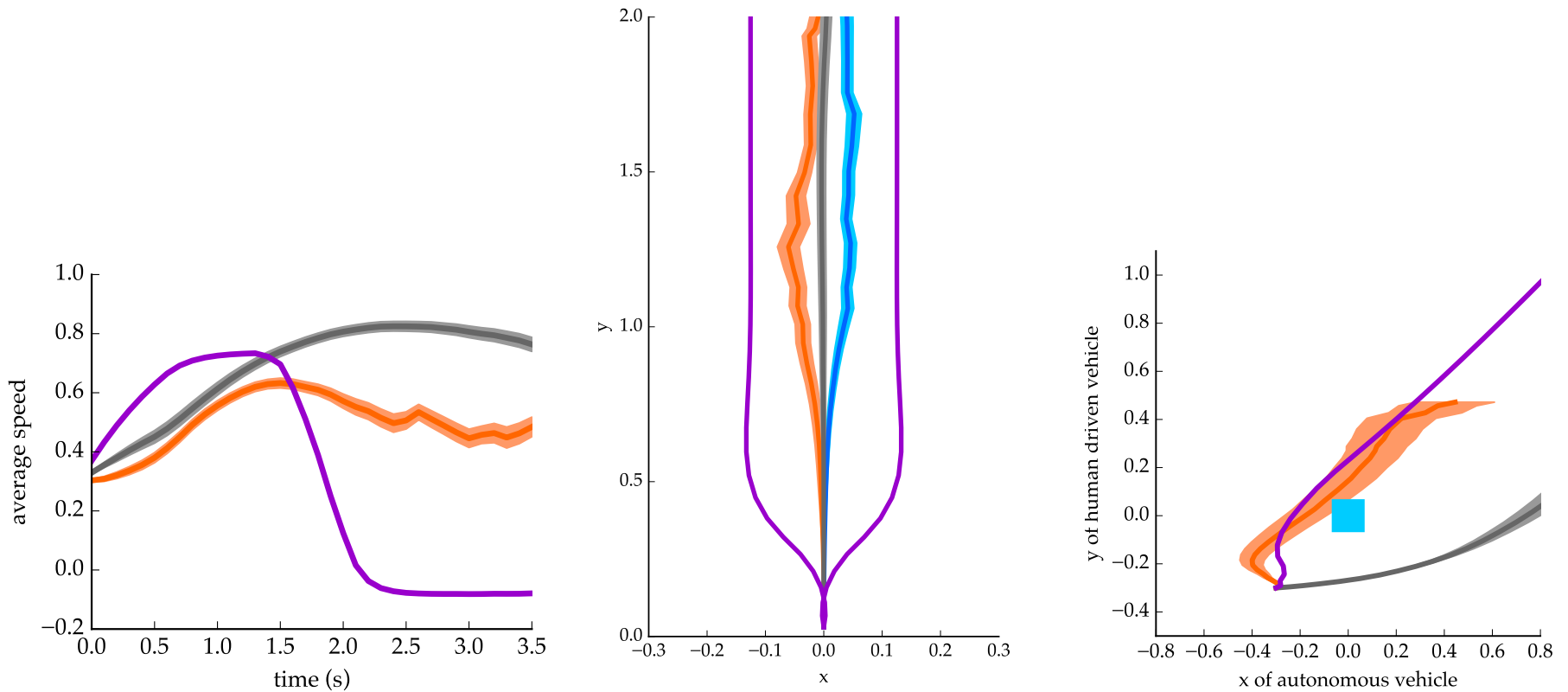
Wait, maybe you should go.

Let's just sit here  
and reflect.





# Real Effects on Real Users' Actions



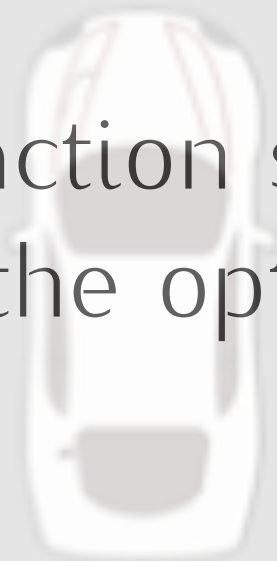
— Learned Human Model

— Avoid Human

— Affect Human

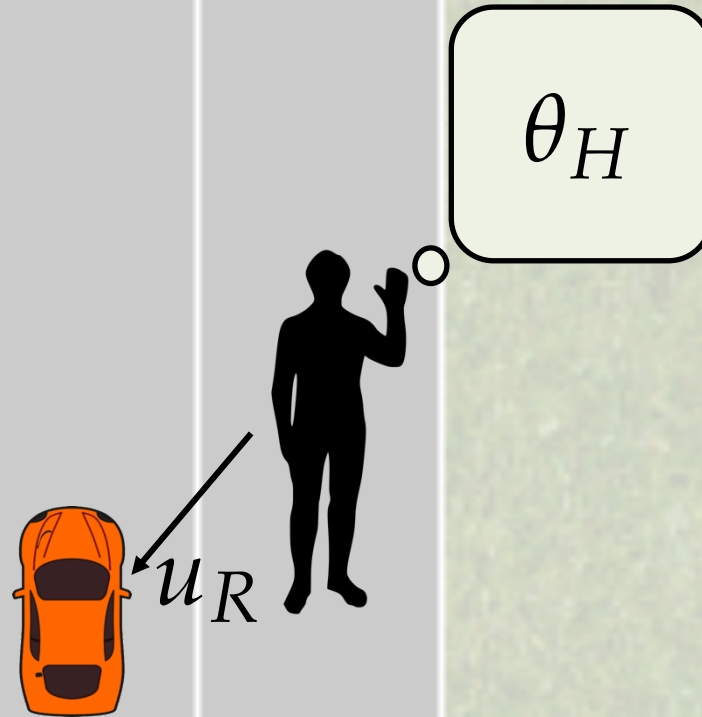
Google's fleet of autonomous test cars is programmed to follow the letter of the law. But it can be tough to get around if you are a stickler for the rules. One Google car, in a test in 2009, couldn't get through a four-way stop because its sensors kept waiting for other (human) drivers to stop completely and let it go. The human drivers kept inching forward, looking for the advantage — paralyzing Google's robot.

Working interaction strategies emerge out of the optimization.

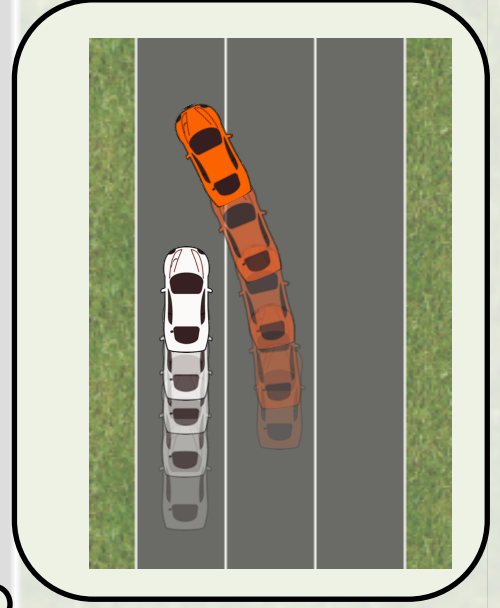
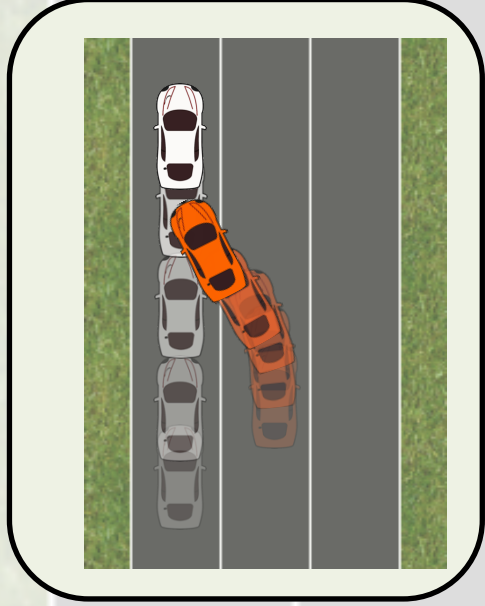
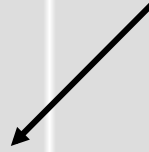


# A Partial Information Human-Robot Game

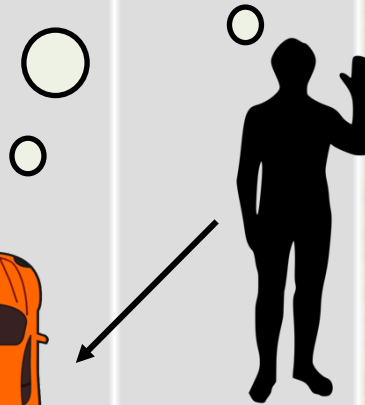
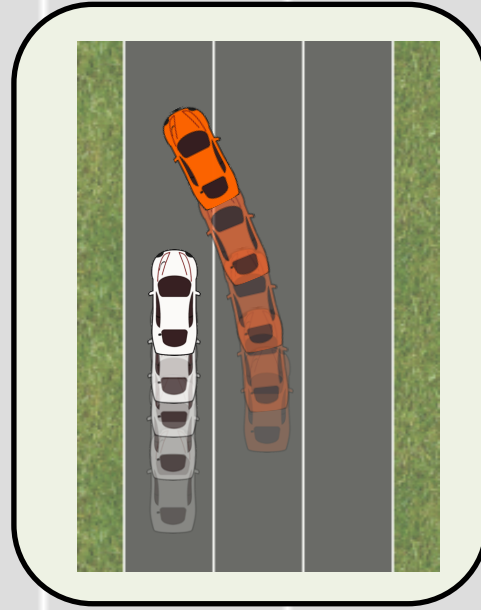
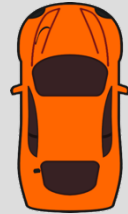
$$U(x, u_R, u_H; \theta_H)$$



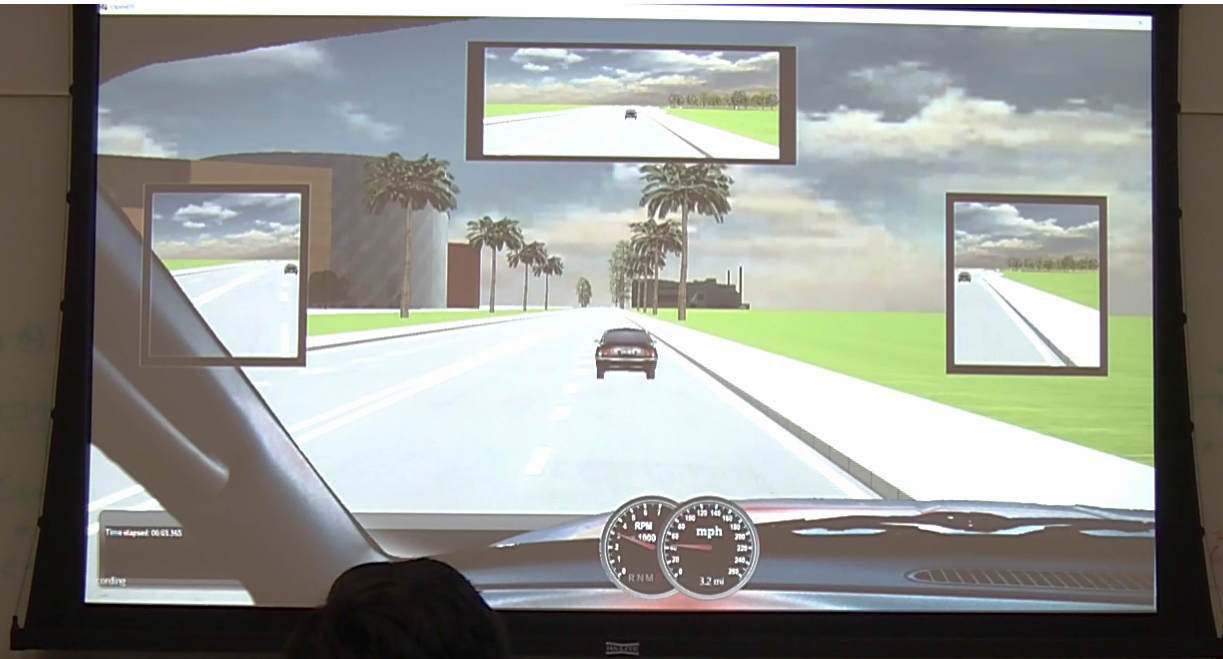




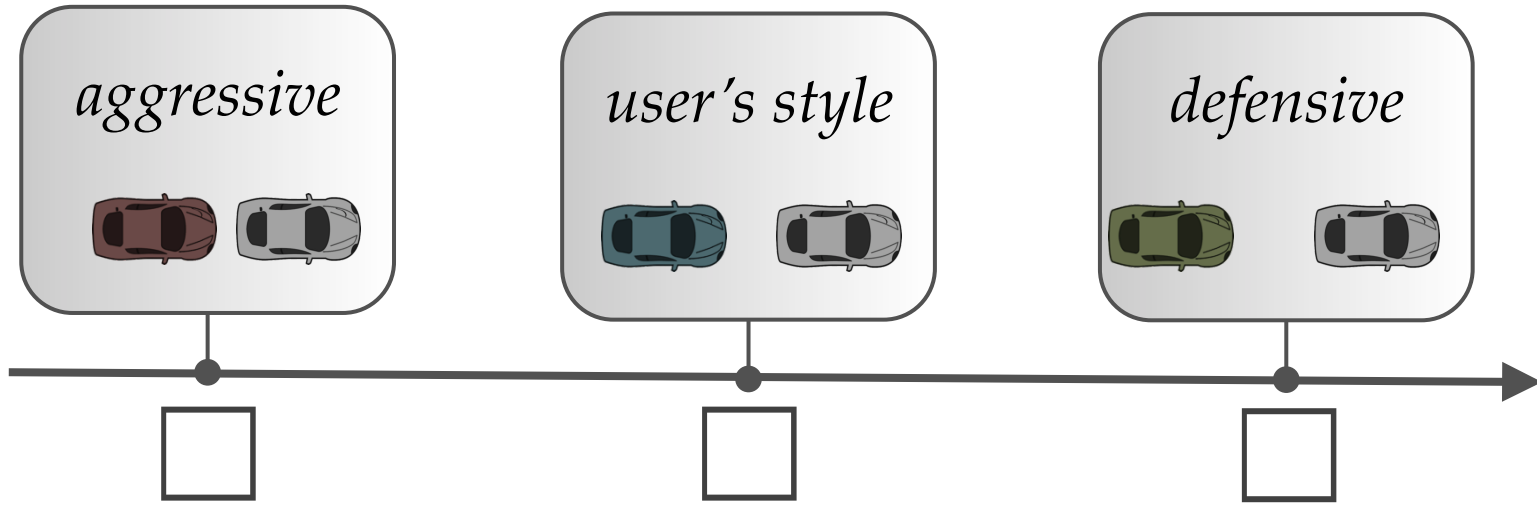
Use human actions  
as observations!



# User Study (in Driving Simulator)

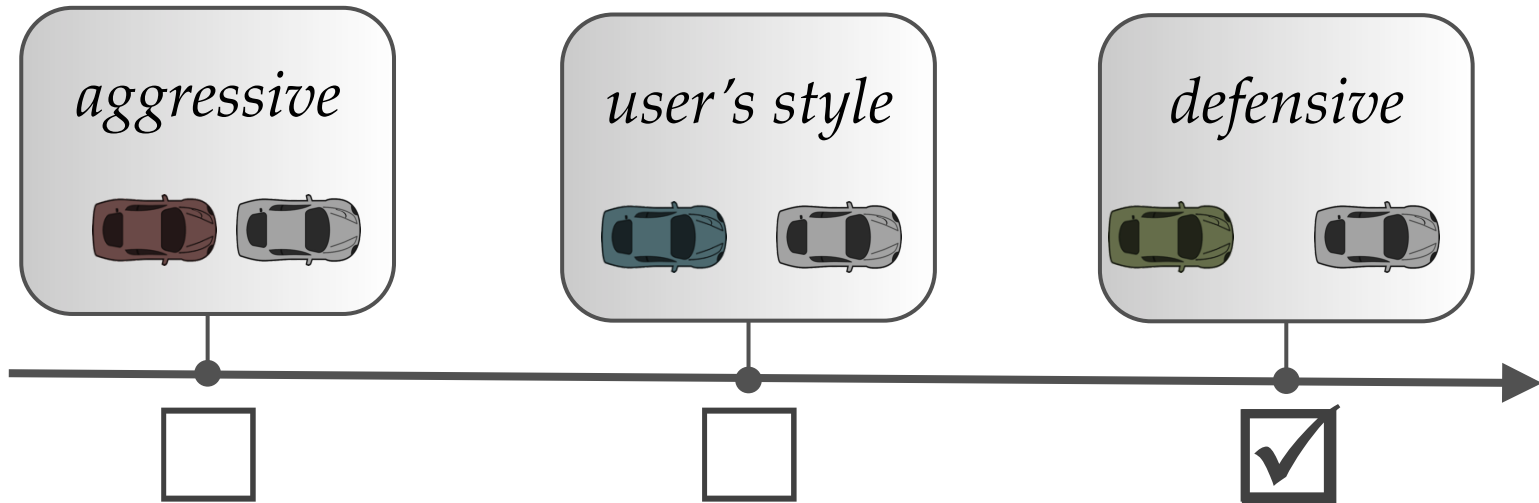


# Manipulated Factor: Driving Style



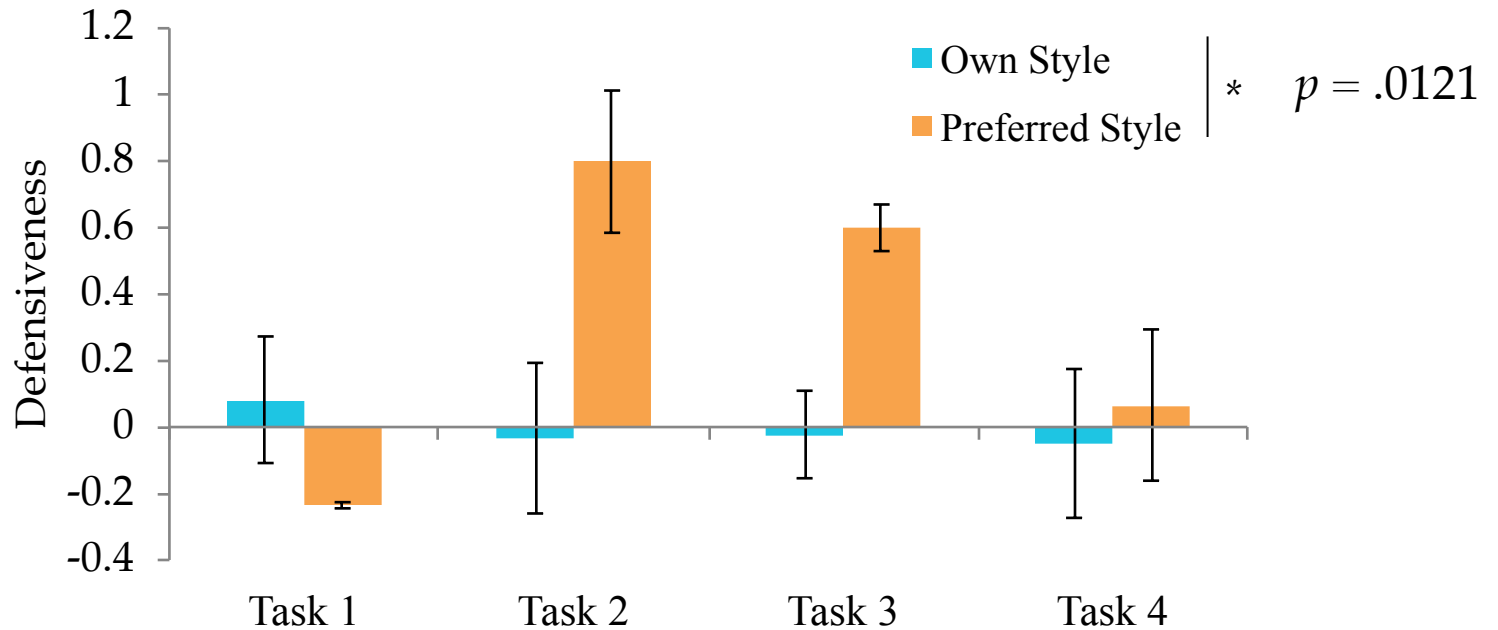


# Dependent Measures

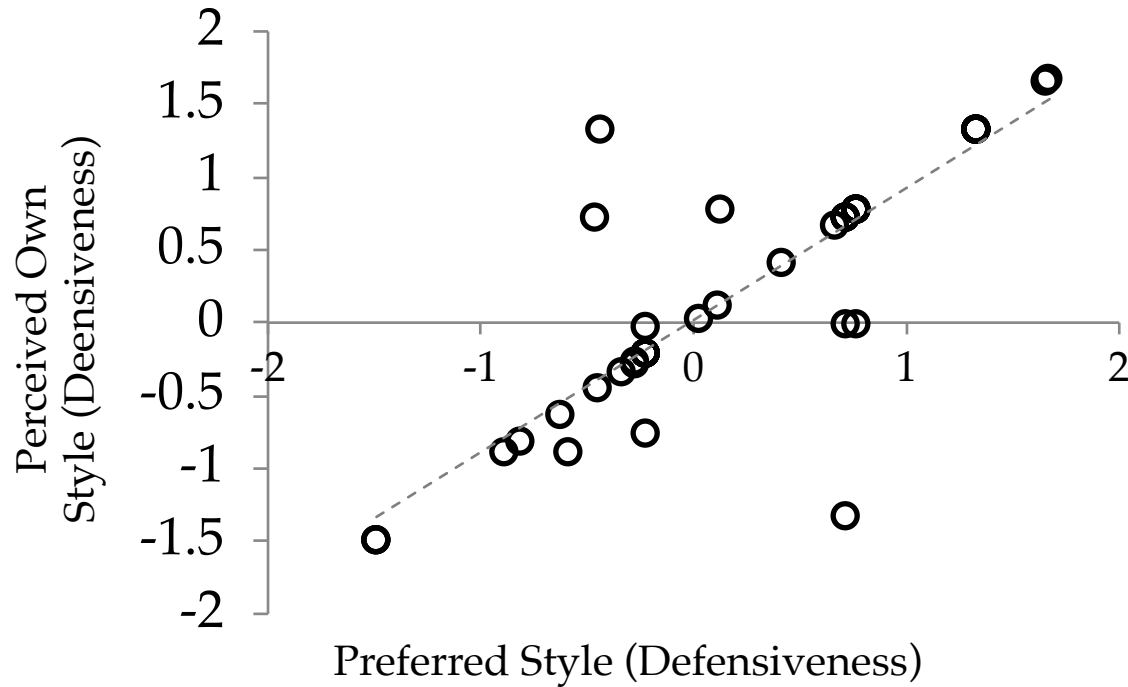


1. Preference for daily use
2. Perceived similarity with own driving style

# Own driving style vs. preferred driving style



# Perceived own style vs preferred driving style!



$$r(58) = 0.86$$

A man in a dark suit and white shirt is driving a car. He has a wide-eyed, open-mouthed expression of shock or panic. The car's interior, including the steering wheel and dashboard, is visible. The background is bright and slightly blurred, suggesting motion.

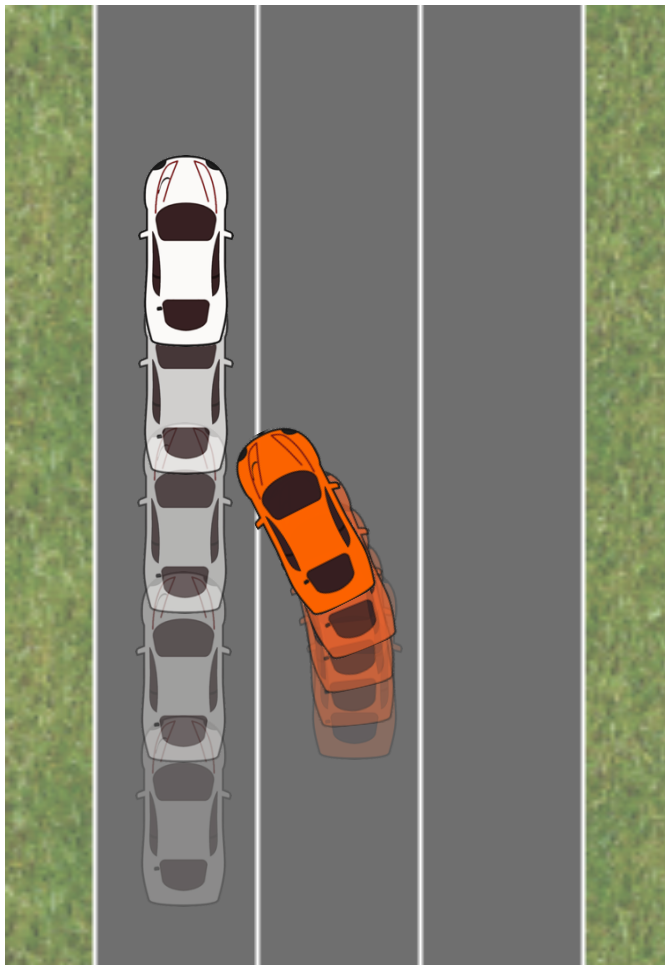
People want their cars to drive like they think they drive, but not like they actually drive.

Implication: Can't rely on demonstrations!

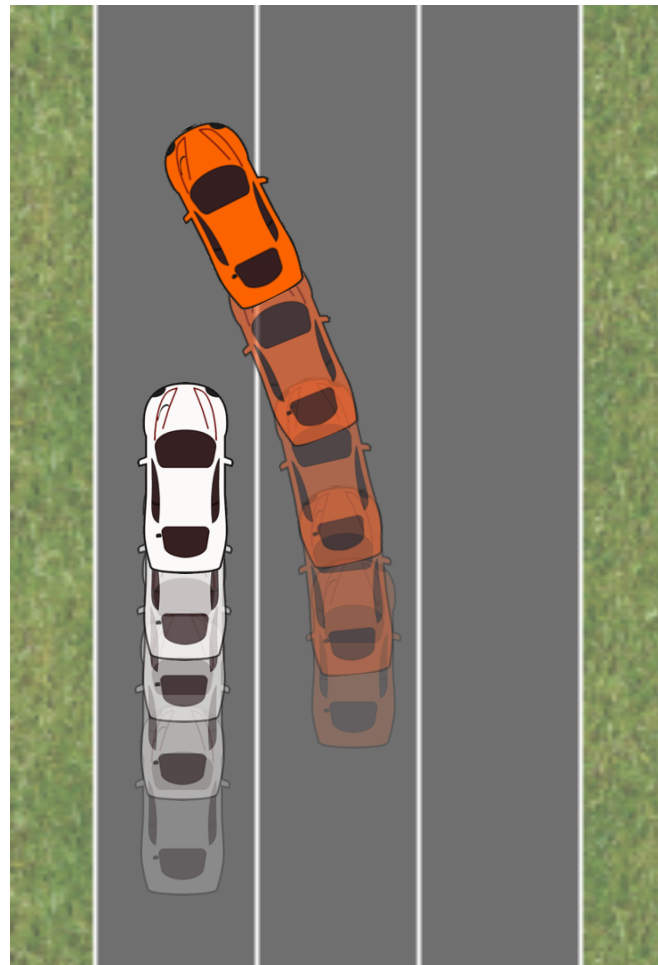
# Comparison-Based Learning

Is  $\xi_1$  or  $\xi_2$  better?

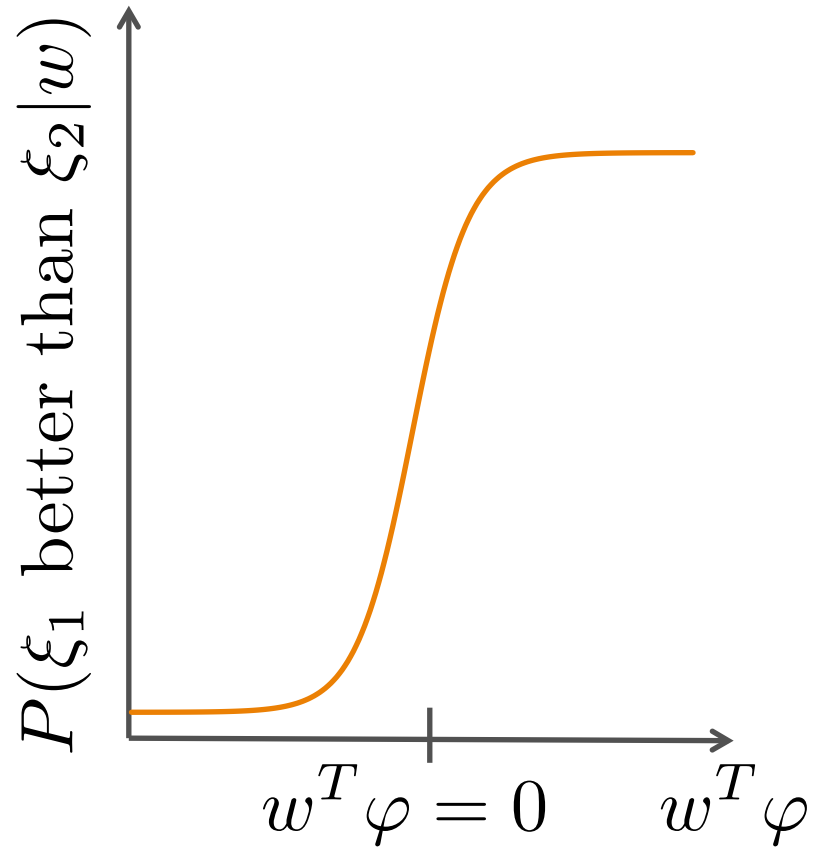
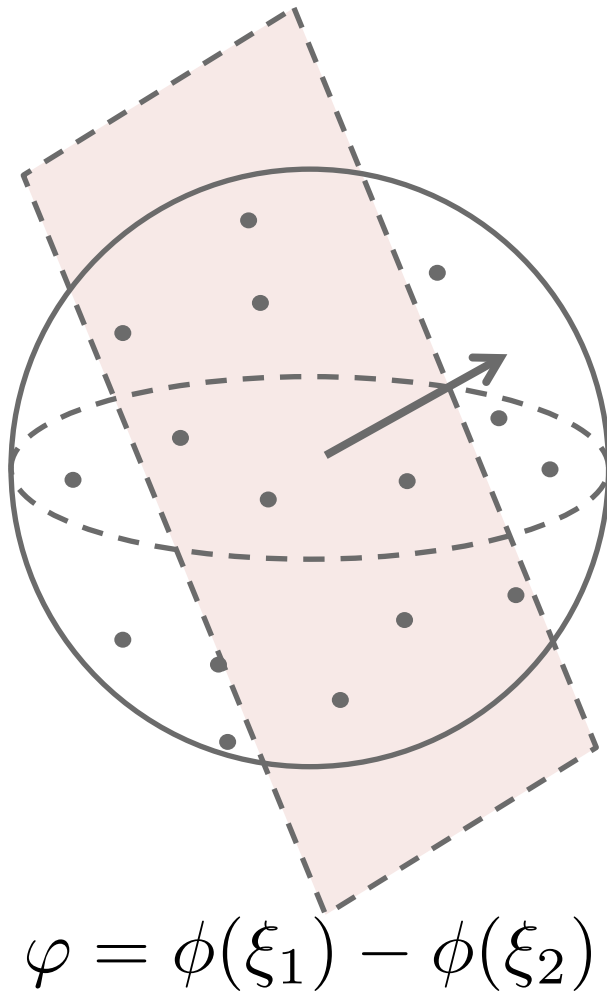
$\xi_1$



$\xi_2$

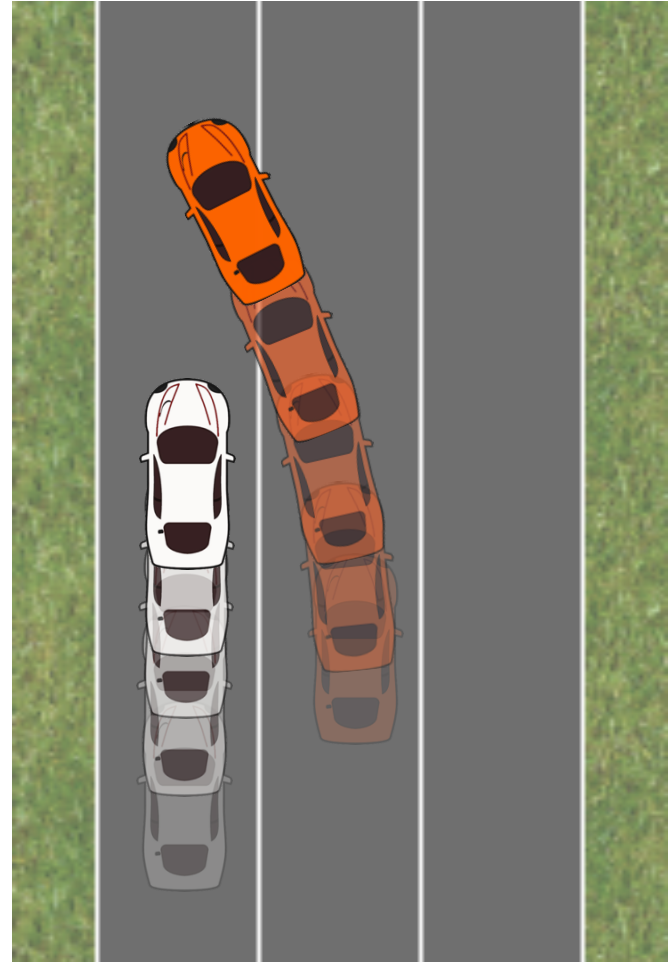
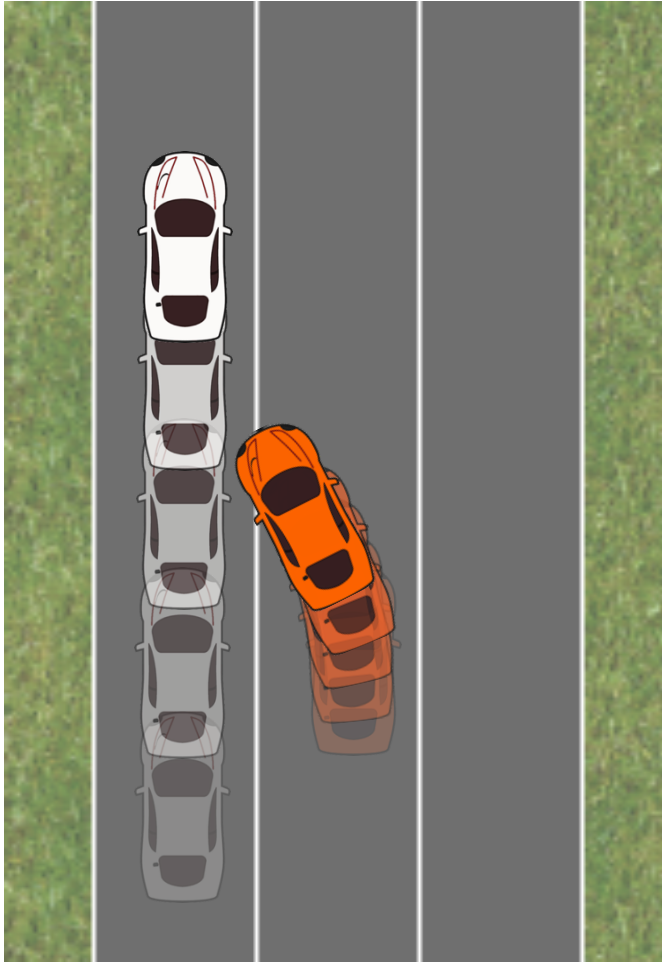


# Every Answer Updates Belief over Hidden Params



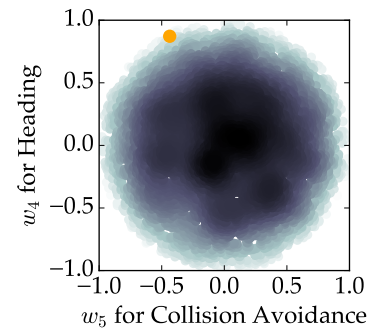
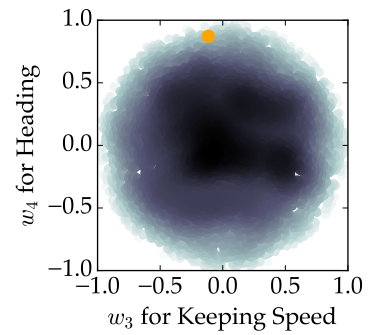
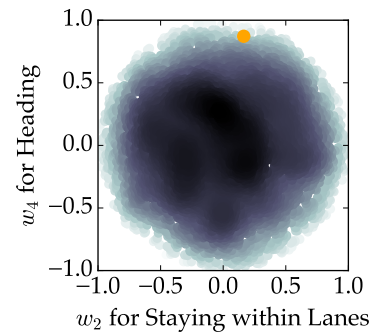
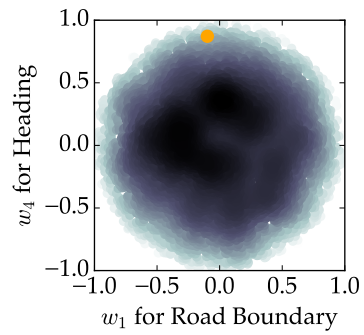
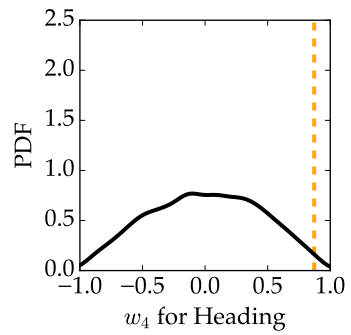
# Actively Gather Information

$$\arg \max_{x_0, \mathbf{u}_R^1, \mathbf{u}_R^2} \mathbb{E}_\theta [H(b) - H(b')]$$

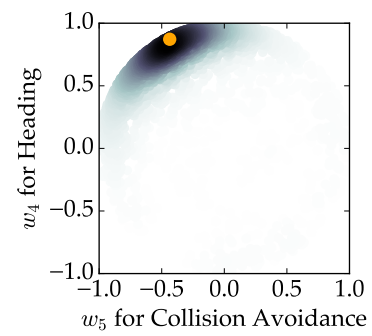
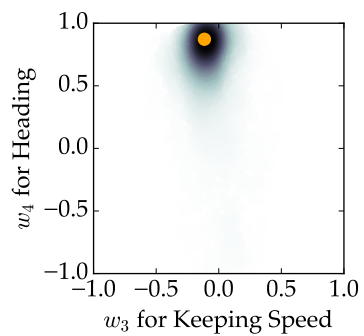
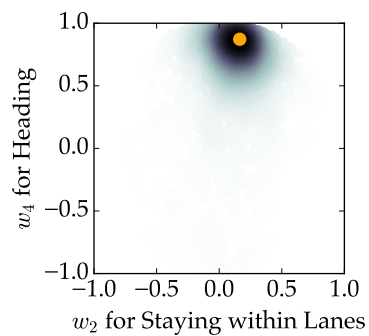
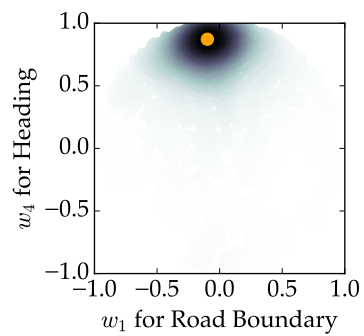
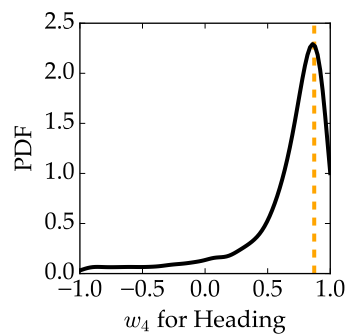




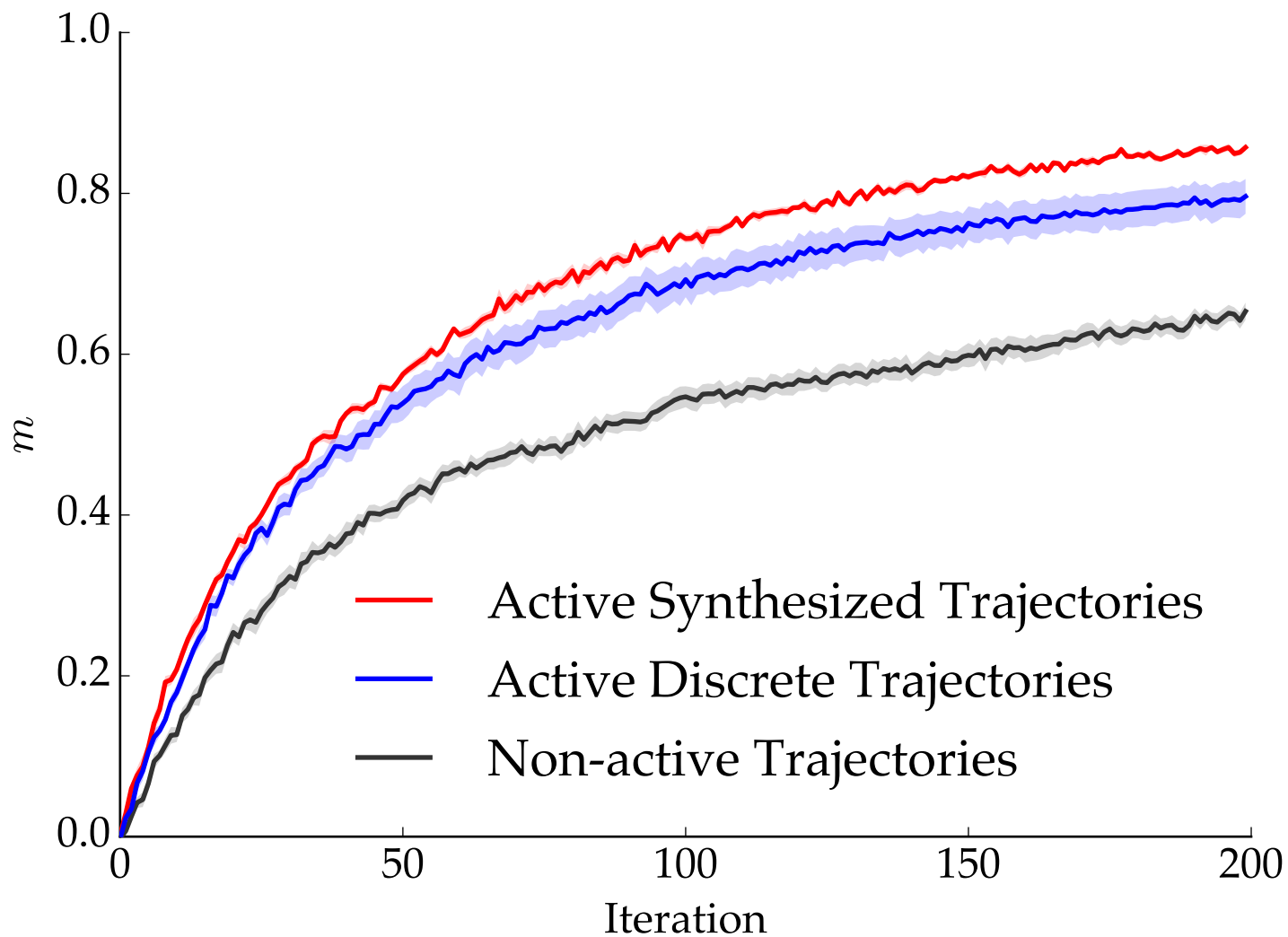
# Belief before..



# Belief after learning

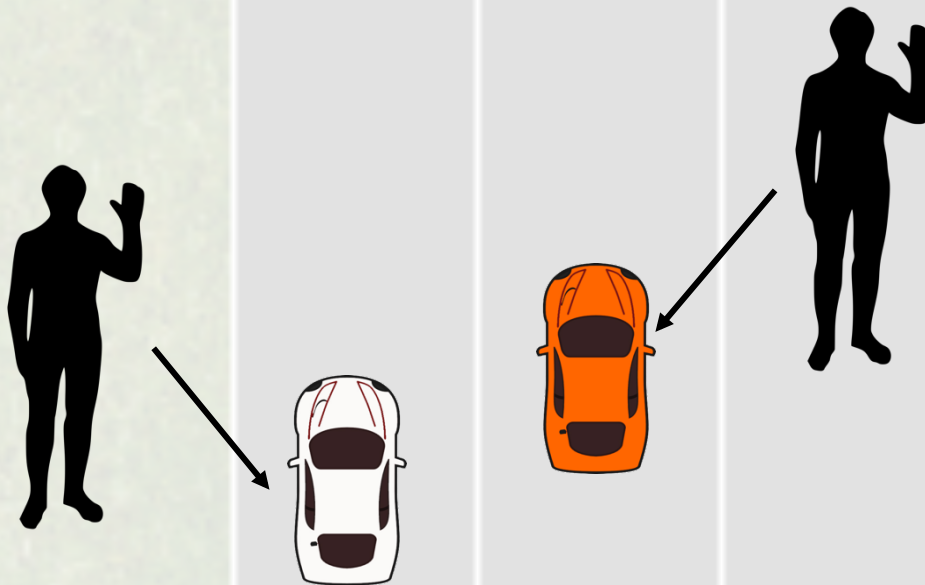


# Actively Synthesizing Queries Helps



People are not just obstacles to be avoided.

... but they are not perfect game solvers either.





InterACT

Laboratory

# Thanks!



CAREER



# Humans Should not Be Obstacles

Anca Dragan

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