# Noncooperative Game Theory for Autonomous Racing

Dr. Alexander Liniger

Game-theoretical Motion Planning ICRA 2021 - Tutorial









### Motivation

### Autonomous driving

- Active research area since the 1980s
  - Research done in industry and academia
  - Waymo/Google: > 20 mio miles
- Take safety critical decisions in an uncertain environment



### Autonomous racing

- Drive as fast as possible around a track
  - Miniature race car set-up using RC cars
  - Formula Student Driverless
  - Roborace
  - IndyAutonomous
- Structured but competitive environment



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### Driving at the handling limit

- If we do not drive at the limit we drive too slow
- Motion planning for a highly nonlinear system



### Staying safe inside the track

- If we crash we lose!
- Infinite horizon constraint satisfaction



- The art of overtaking and interacting with other cars
- Decision making in a highly dynamical non-cooperative environment



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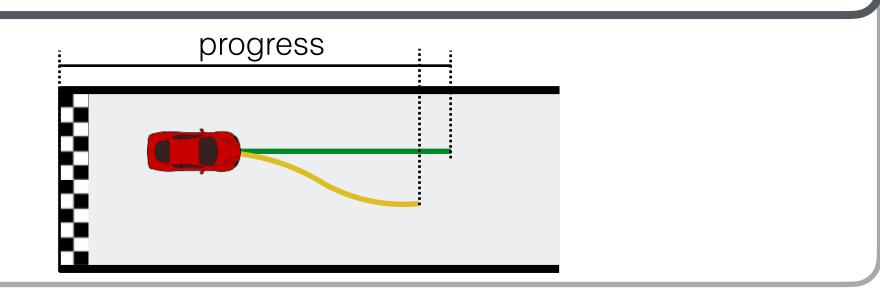
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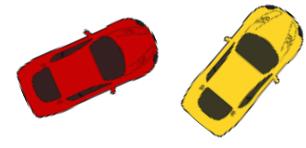
# Racing Ingredients

#### Finish first

- Approximated by maximizing progress
- Generates racing trajectories



#### Do not collide with other cars



good

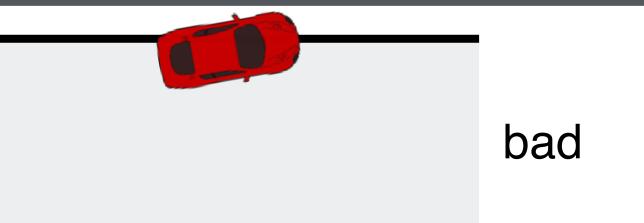


bad

### Stay inside the track



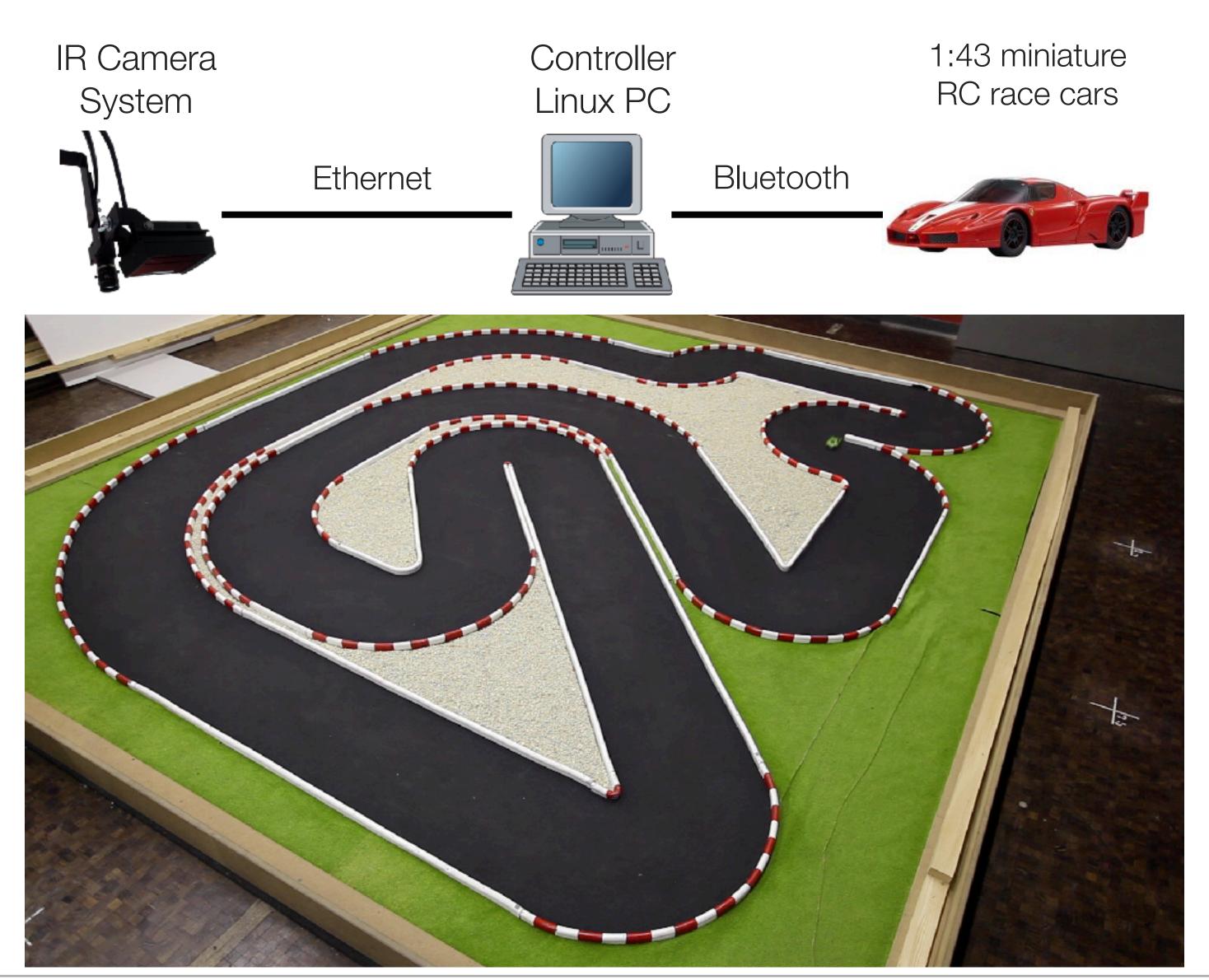
good



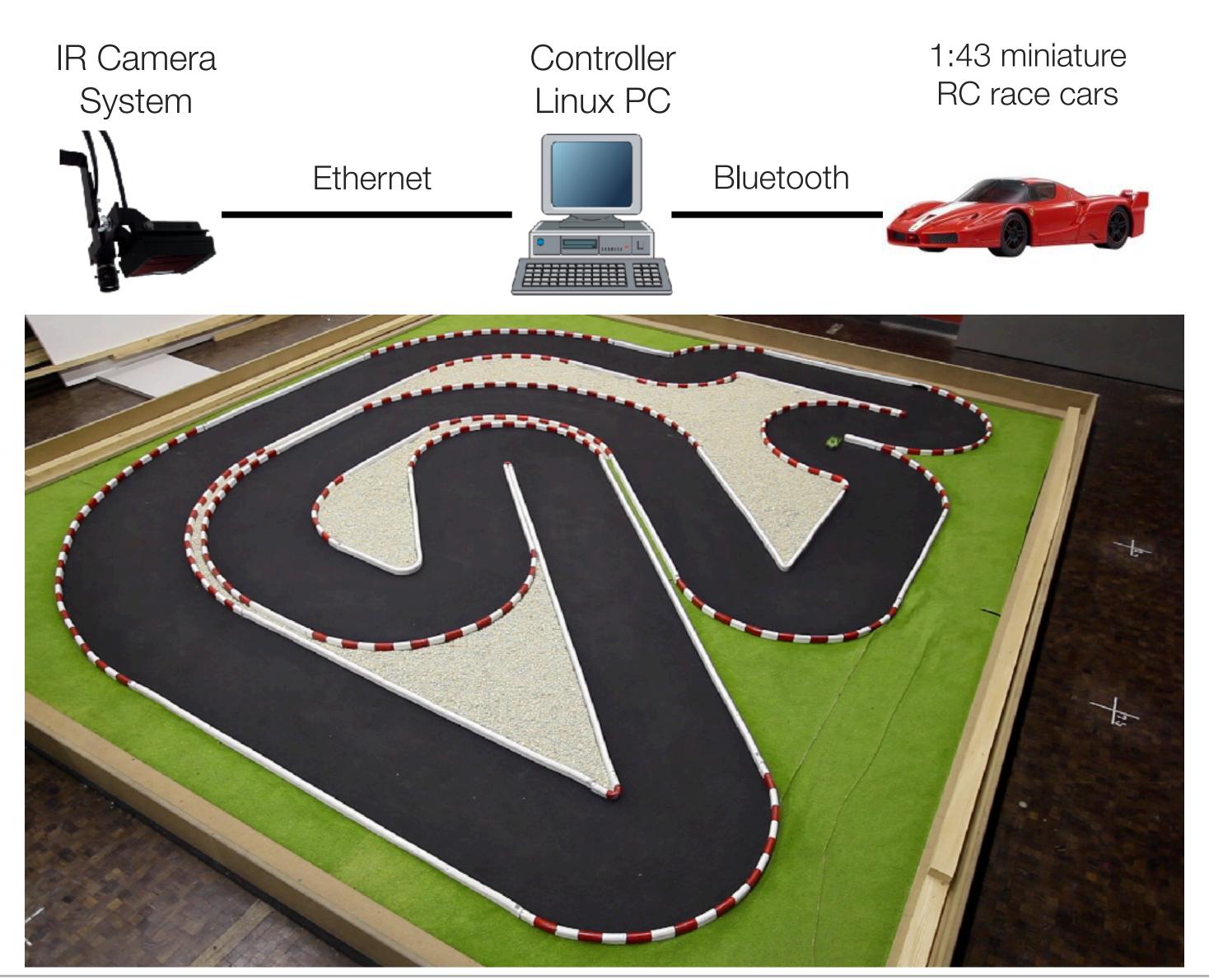




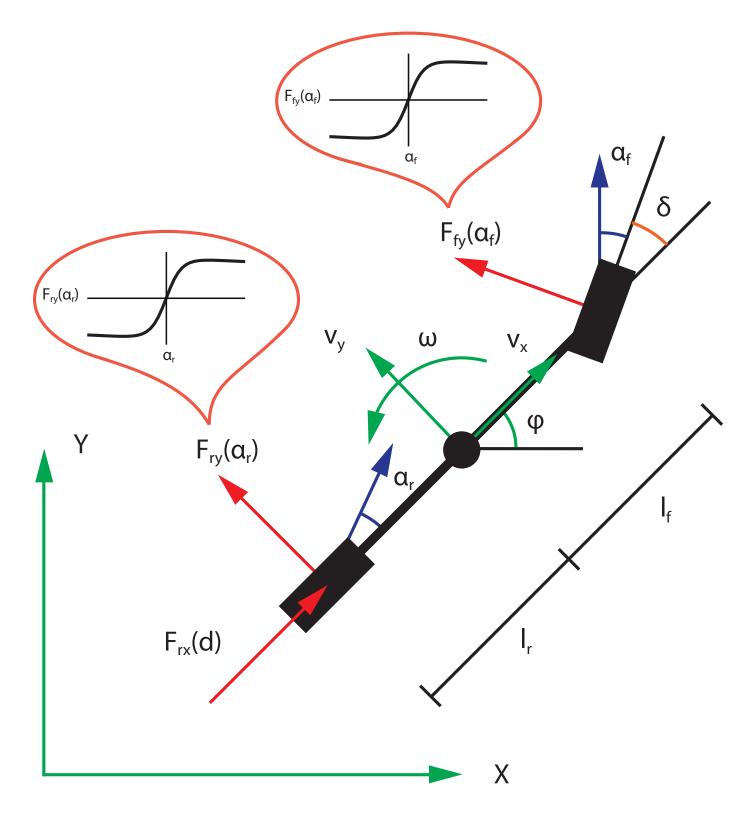
# Experimental Set-Up



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Bicycle model, with nonlinear lateral tire forces (Pacejka)



Highly nonlinear 6 dimensional system

$$\dot{X} = v_x \cos(\varphi) - v_y \sin(\varphi)$$

$$\dot{Y} = v_x \sin(\varphi) + v_y \cos(\varphi)$$

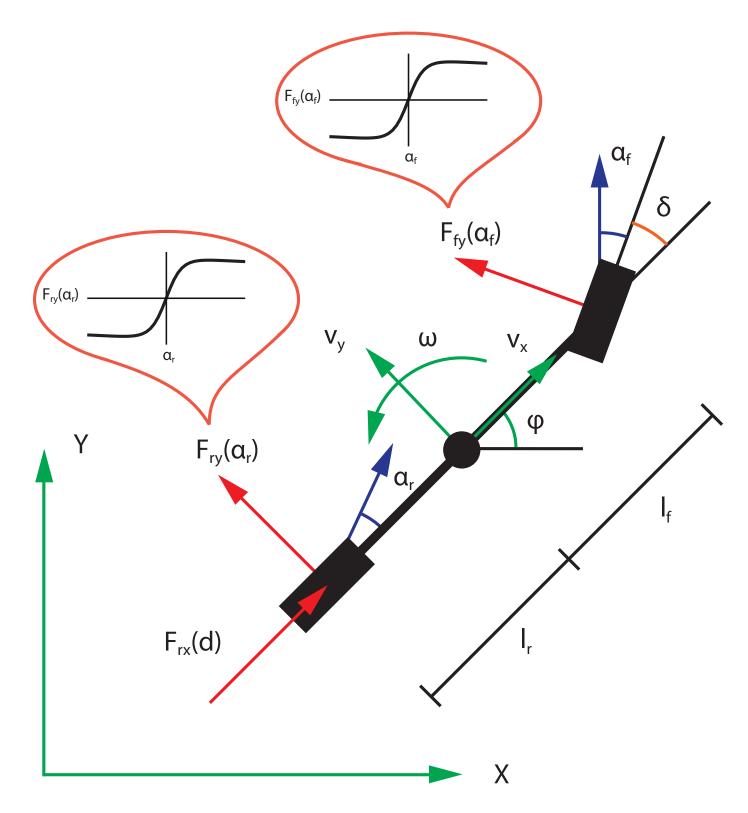
$$\dot{\varphi} = \omega$$

$$\dot{v}_x = \frac{1}{m} (F_{r,x} - F_{f,y} \sin \delta + m v_y \omega)$$

$$\dot{v}_y = \frac{1}{m} (F_{r,y} + F_{f,y} \cos \delta - m v_x \omega)$$

$$\dot{\omega} = \frac{1}{I_z} (F_{f,y} I_f \cos \delta - F_{r,y} I_r)$$

Bicycle model, with nonlinear lateral tire forces (Pacejka)



- Highly nonlinear 6 dimensional system
- Separation in slow and fast dynamics

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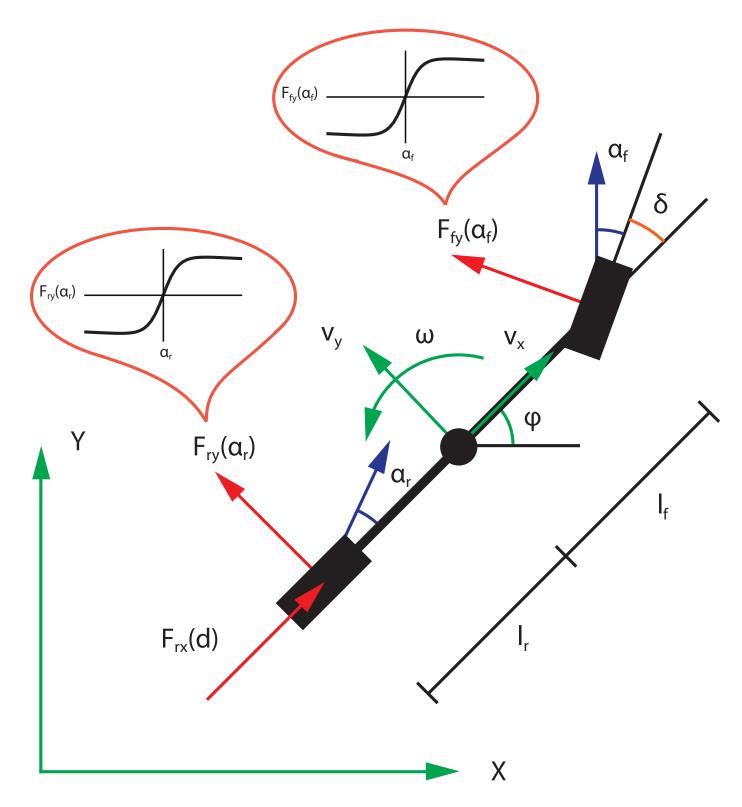
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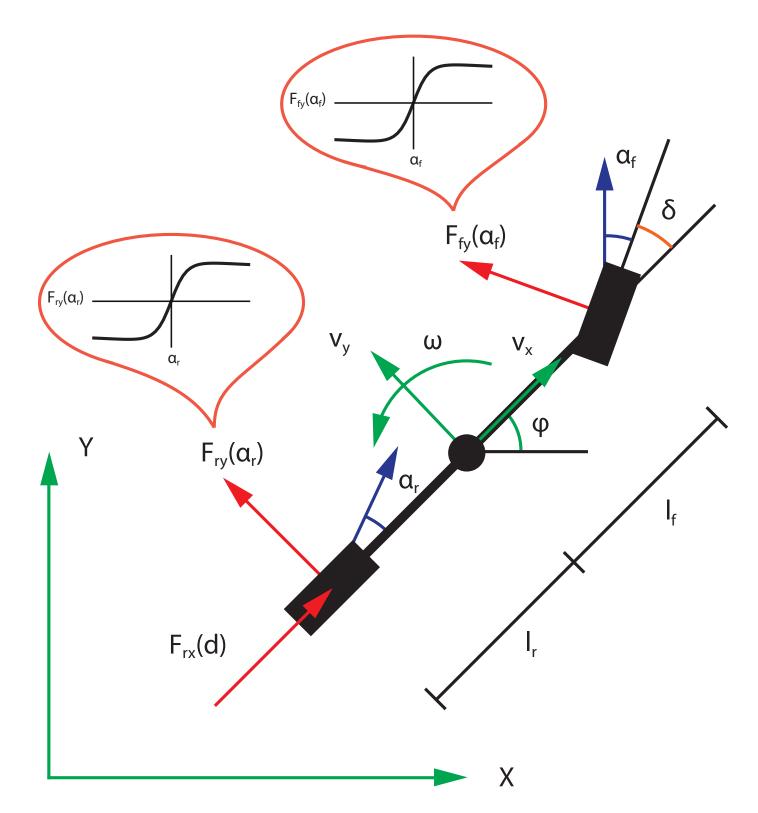
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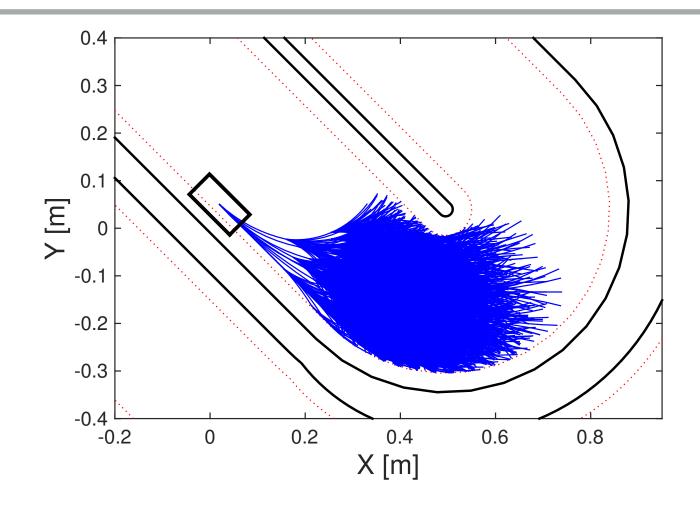
$$\dot{\omega} = \frac{1}{I_{z}} (F_{f,y} I_{f} \cos \delta - F_{r,y} I_{r})$$

- Highly nonlinear 6 dimensional system
- Separation in slow and fast dynamics

### High-level motion planning based on constant velocities primitives

- Plan for slow dynamics
- ▶ Reduced dimension —> four dimensions instead of six
- ▶ Long discretization times —> 0.16s instead of 0.02s

- Considering full dynamical bicycle model
- Linearization points given by path planner

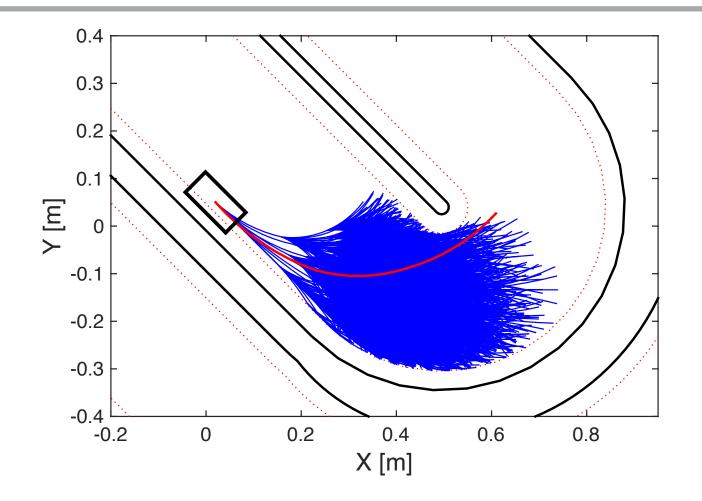


- 129 constant velocity motion primitives
- 3 prediction steps
- Lookahead of 0.48s

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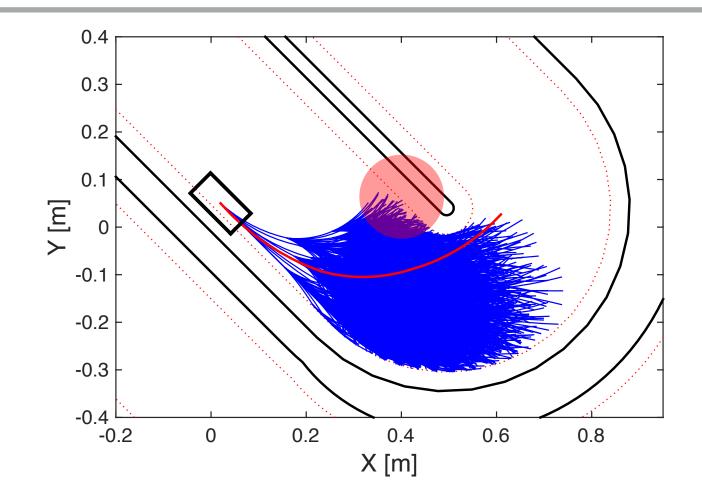


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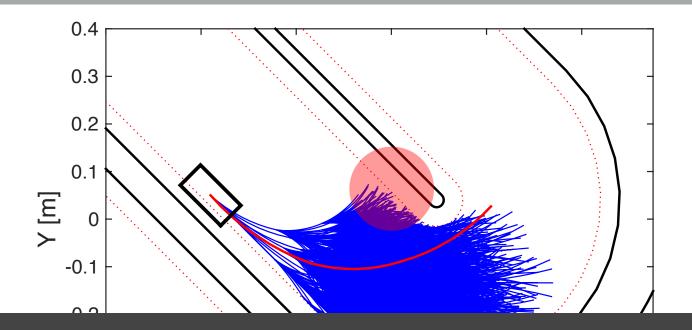
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### MPC-based trajectory tracking

- Considering full dynamical bicycle model
- Linearization points given by path planner



0.2

0.4

0.6

8.0

- 129 constant velocity motion primitives
- 3 prediction steps
- Lookahead of 0.48s

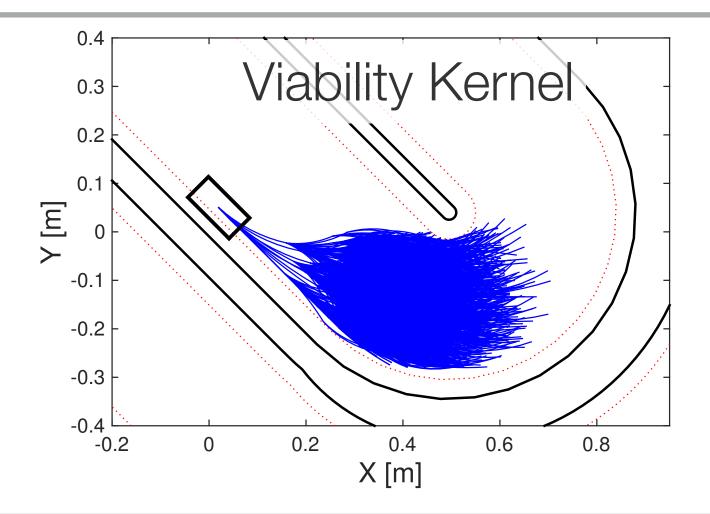
Only generate safe trajectories

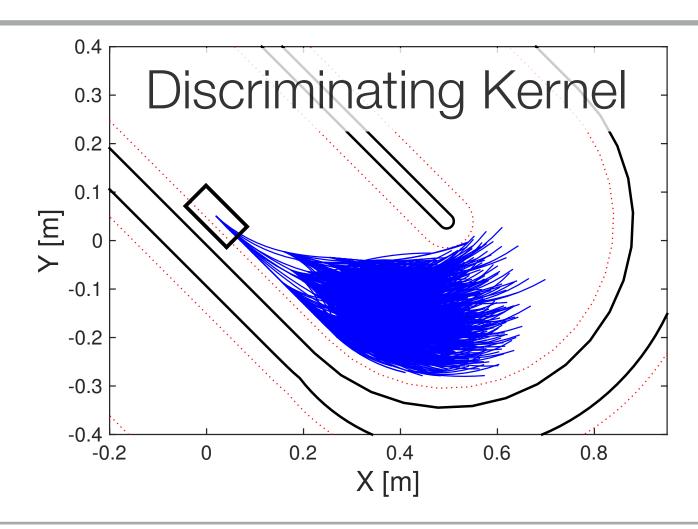


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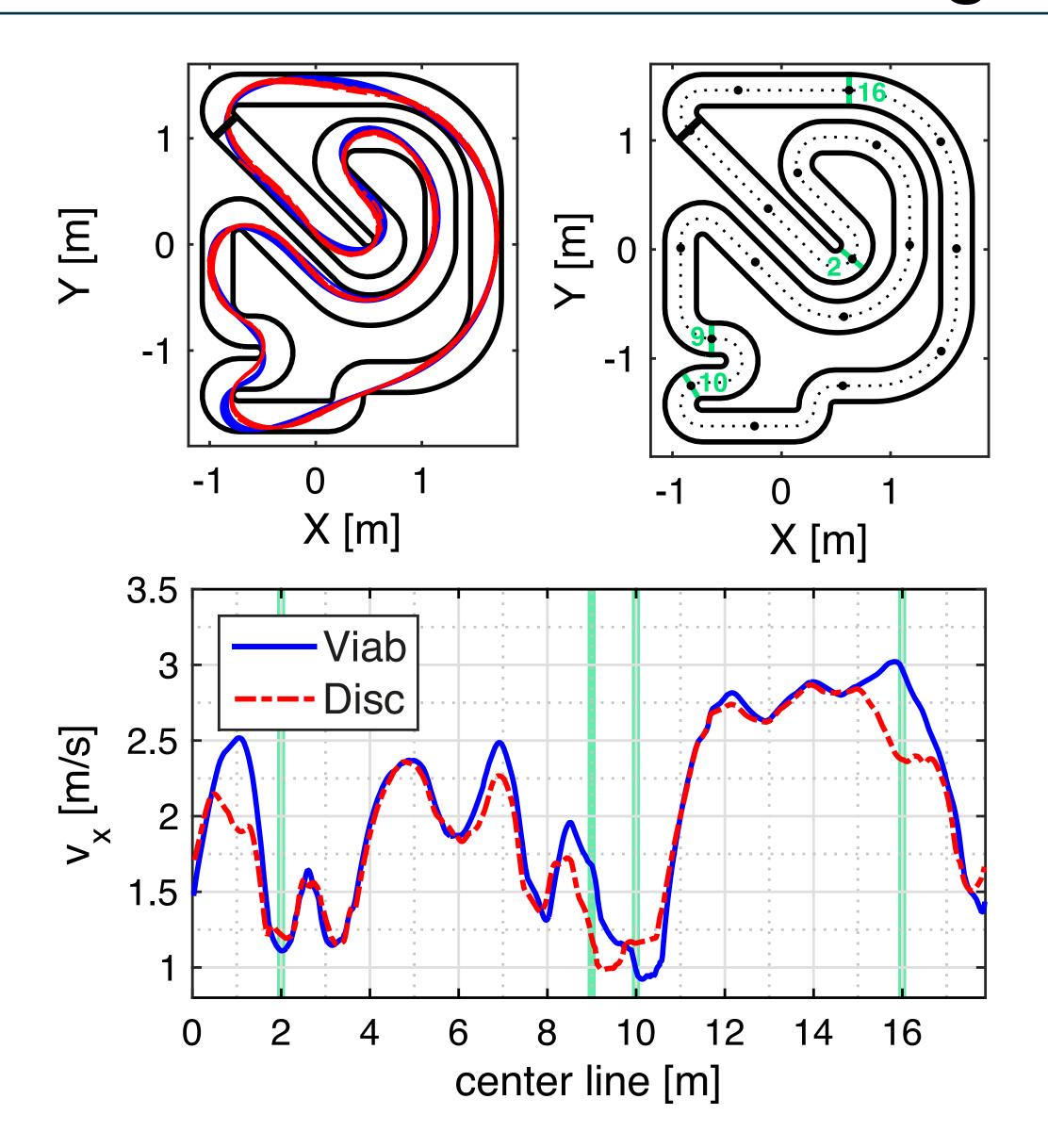
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# Simulation Results - Single Agent

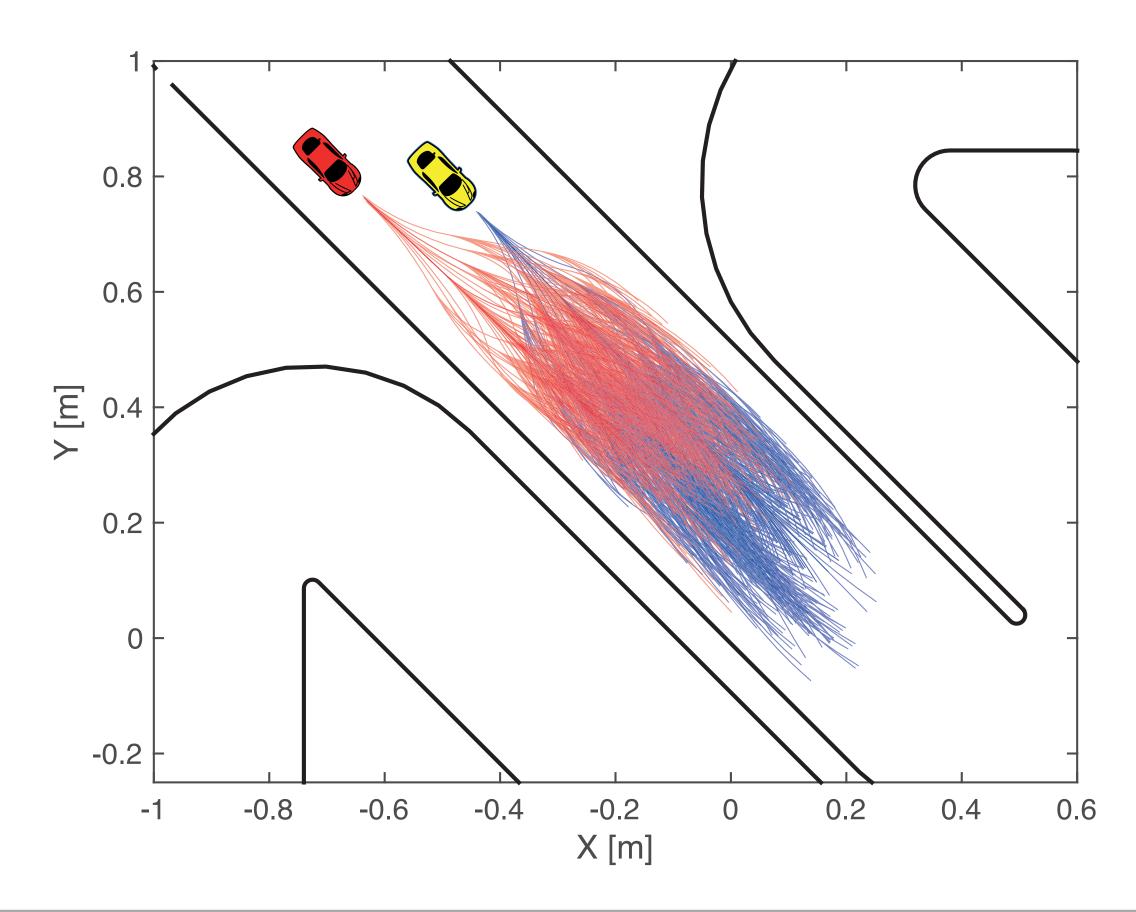


Kernel	mean lap time [s]	# constr. violations	median comp. time [ms]	max comp. time [ms]
No	8.76	4	32.26	247.7
Viab	8.57	0	0.904	7.968
Disc	8.60	1	0.870	7.533

- Both methods have the same lap time
- But use different driving styles
- Difference allows for interesting racing

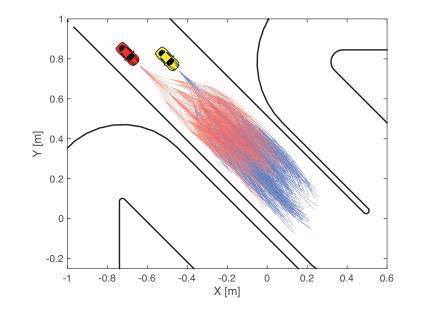
# Bimatrix Racing Games

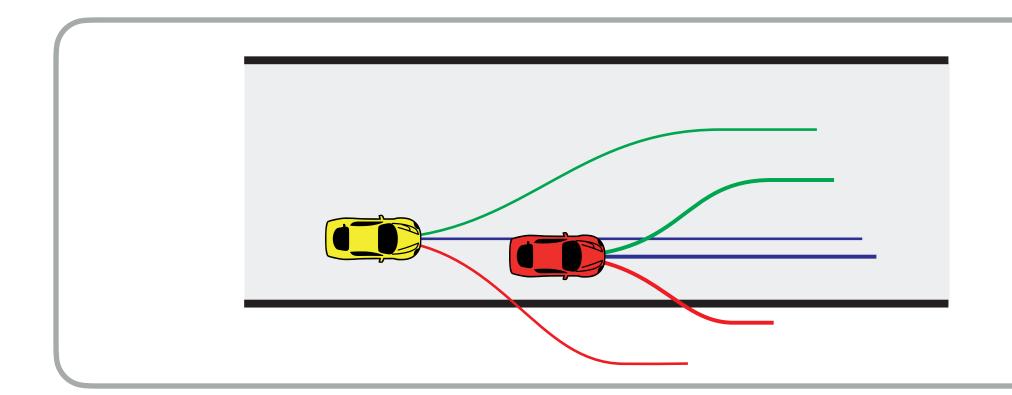
- Every trajectory is an action of a car
  - Each trajectory has a payoff
  - Payoff depends on actions of both cars

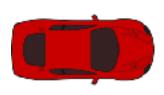


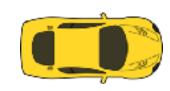
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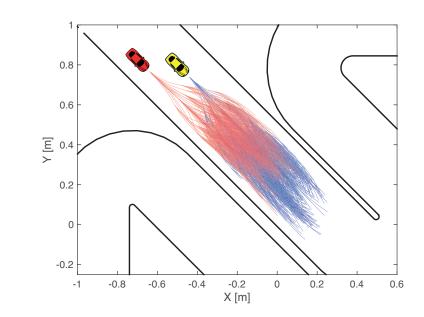


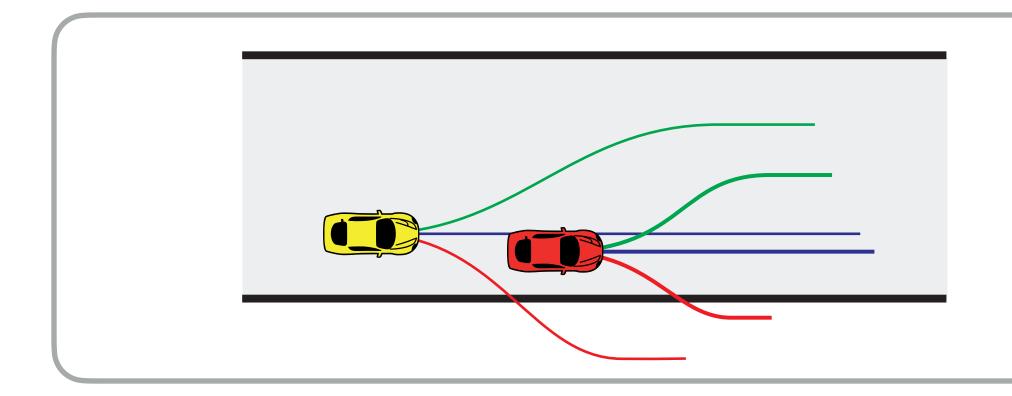


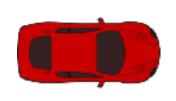
$$A = \begin{bmatrix} a_{1,1} & a_{1,2} & a_{1,3} \\ a_{2,1} & a_{2,2} & a_{2,3} \\ a_{3,1} & a_{3,2} & a_{3,3} \end{bmatrix} B = \begin{bmatrix} b_{1,1} & b_{1,2} & b_{1,3} \\ b_{2,1} & b_{2,2} & b_{2,3} \\ b_{3,1} & b_{3,2} & b_{3,3} \end{bmatrix}$$

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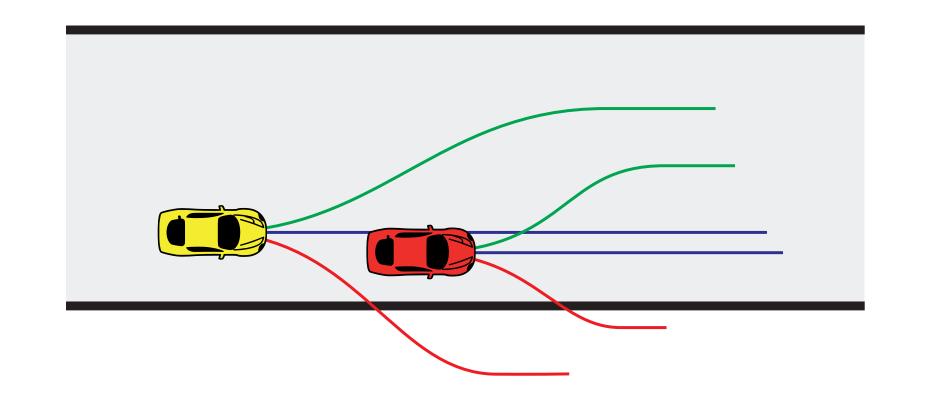


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- The leader is always the car which is ahead at the beginning
- A trajectory pair is feasible if:
  - Trajectories stay inside the track and do not collide

Sequential Game

Cooperative Game



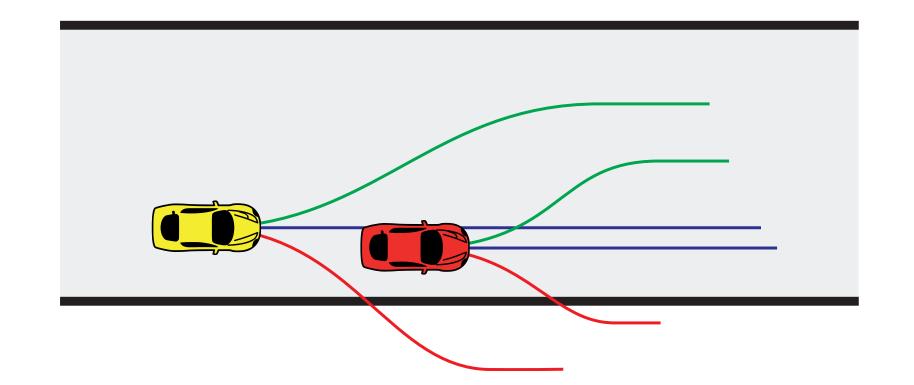
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$$B = \begin{bmatrix} b_{1,1} & b_{1,2} & b_{1,3} \\ b_{2,1} & b_{2,2} & b_{2,3} \\ b_{3,1} & b_{3,2} & b_{3,3} \end{bmatrix}$$

### Sequential Game

- Exploiting the leader-follower structure
  - Low payoff if a trajectory leaves the track
  - Progress payoff if a trajectory is inside the track
  - Low payoff for the **follower** if trajectories collide

### Cooperative Game



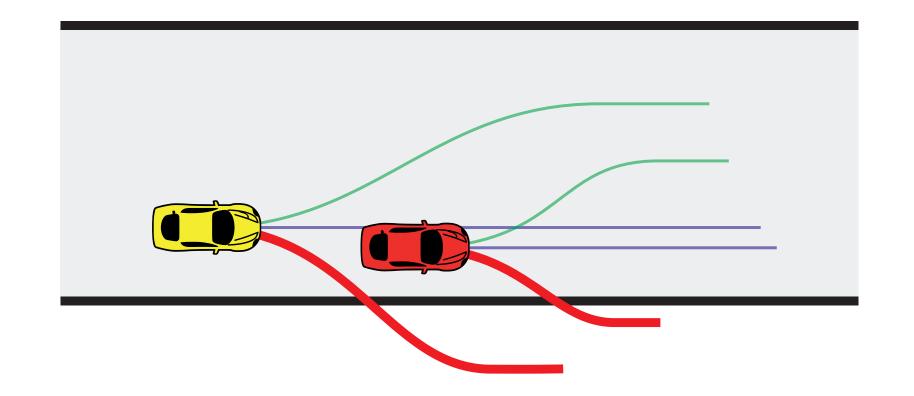
$$A =$$

$$B =$$

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$$A = \begin{bmatrix} A = \\ -10 & -10 & -10 \end{bmatrix}$$

$$B =$$

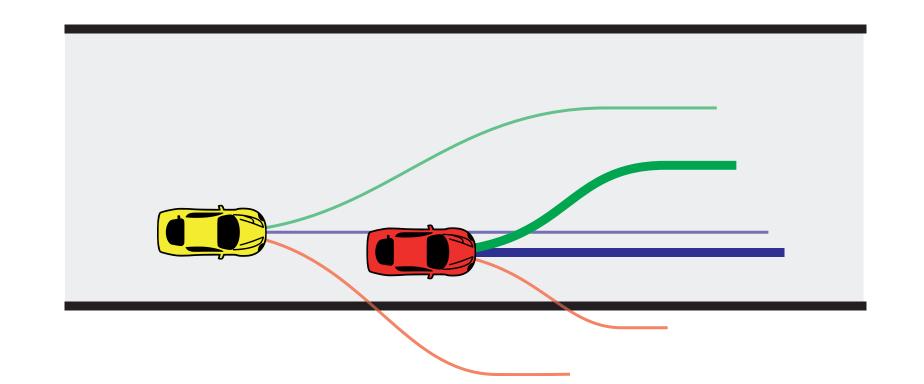
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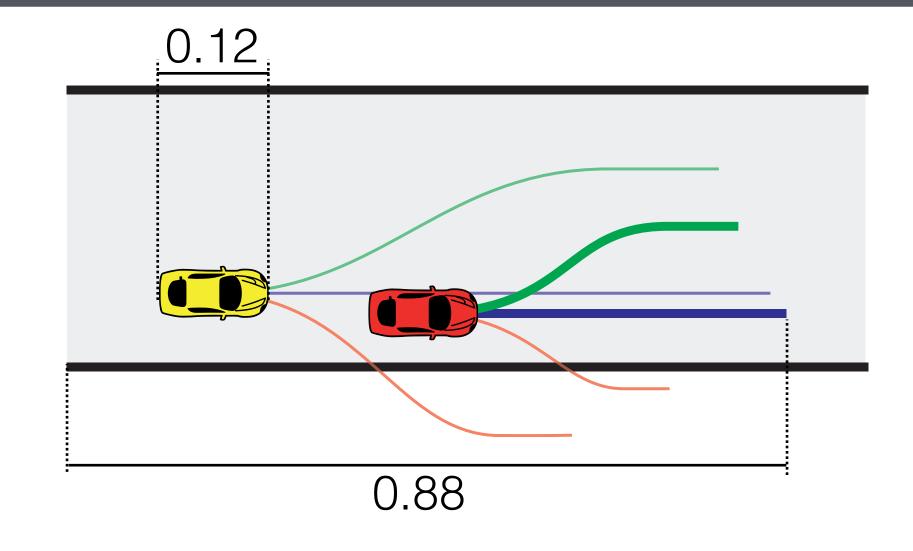
$$A = \begin{bmatrix} 0.83 & 0.83 & 0.83 \\ 0.88 & 0.88 & 0.88 \\ -10 & -10 & -10 \end{bmatrix}$$

$$B =$$

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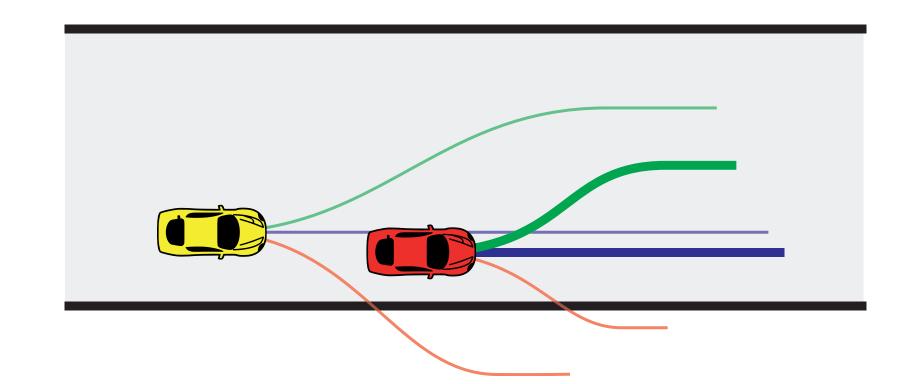
$$B =$$

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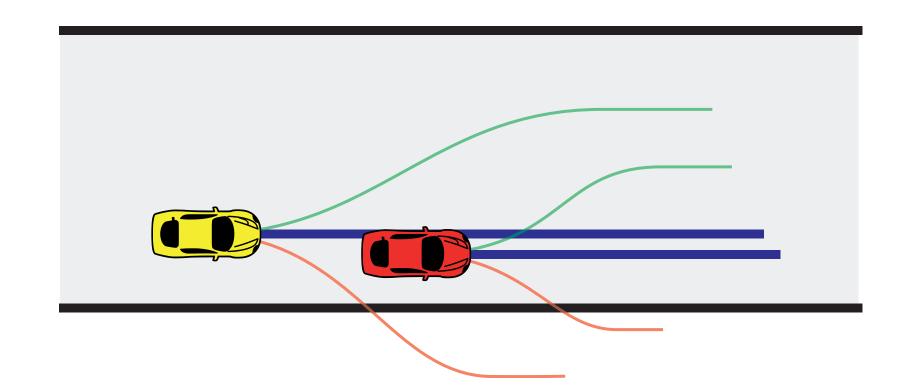
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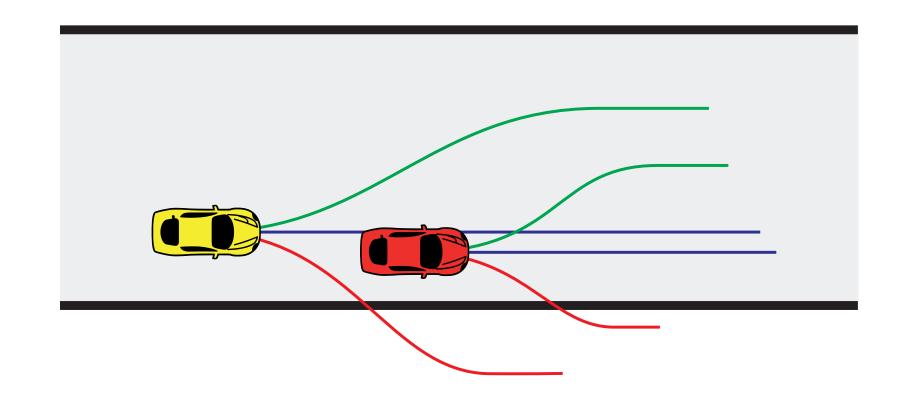
$$\begin{bmatrix}
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 0.88 & 0.88 & 0.88 \\
 -10 & -10 & -10
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$$B = \begin{bmatrix} 0.81 & 0.86 & -10 \\ 0.81 & -1 & -10 \\ 0.81 & 0.86 & -10 \end{bmatrix}$$

### Sequential Game

### Cooperative Game

- Both cars consider collisions
  - Low payoff if a trajectory leaves the track
  - Low payoff if the trajectories collide
  - Progress payoff if a trajectory is feasible

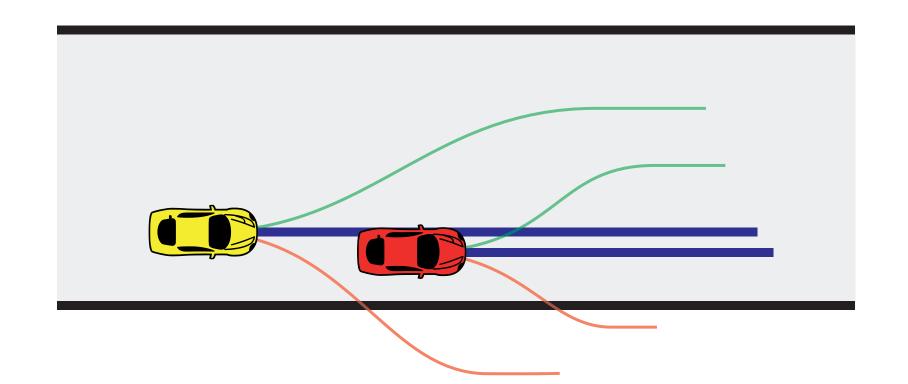


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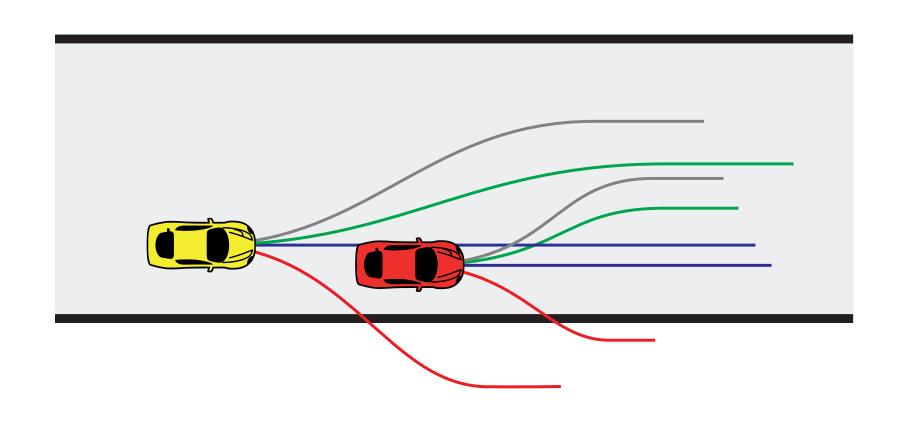
$$B = \begin{bmatrix} 0.81 & 0.86 & -10 \\ 0.81 & -1 & -10 \\ 0.81 & 0.86 & -10 \end{bmatrix}$$



Sequential Game

Cooperative Game

- Same collision structure as the cooperative game, but:
- Additional reward for staying in front at the end of the horizon



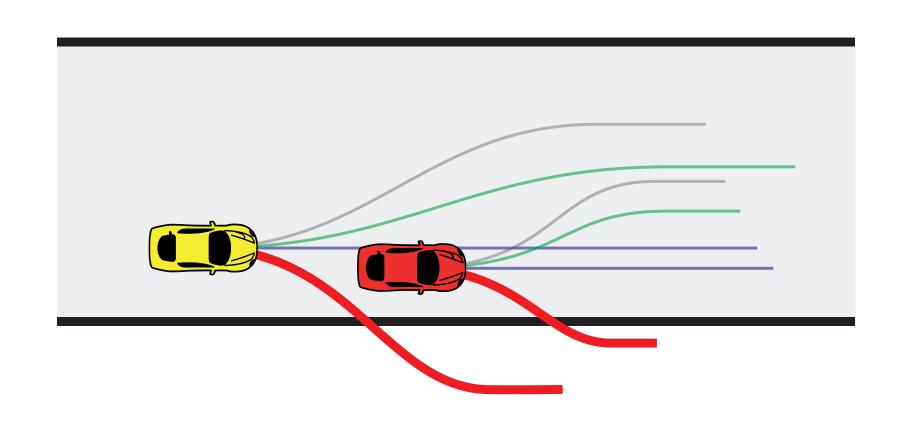
$$A =$$

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Sequential Game

Cooperative Game

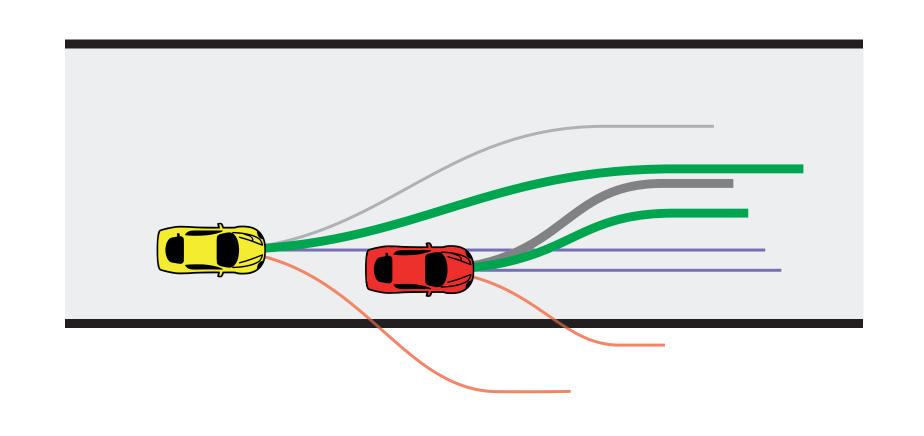
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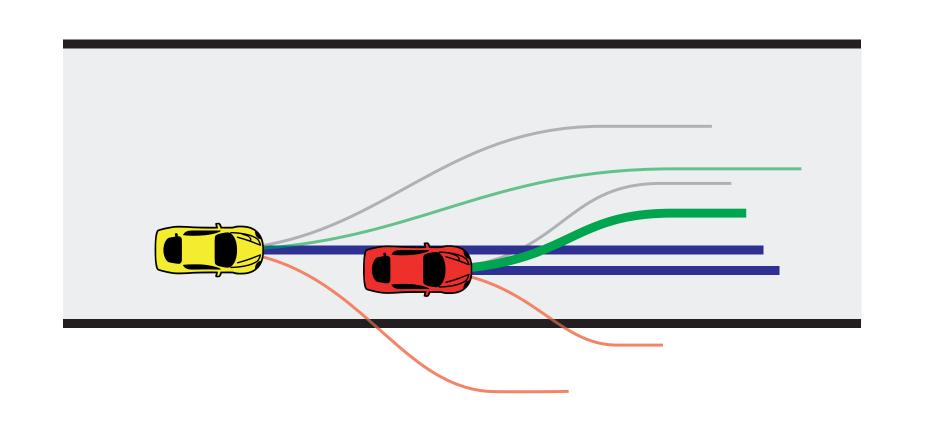
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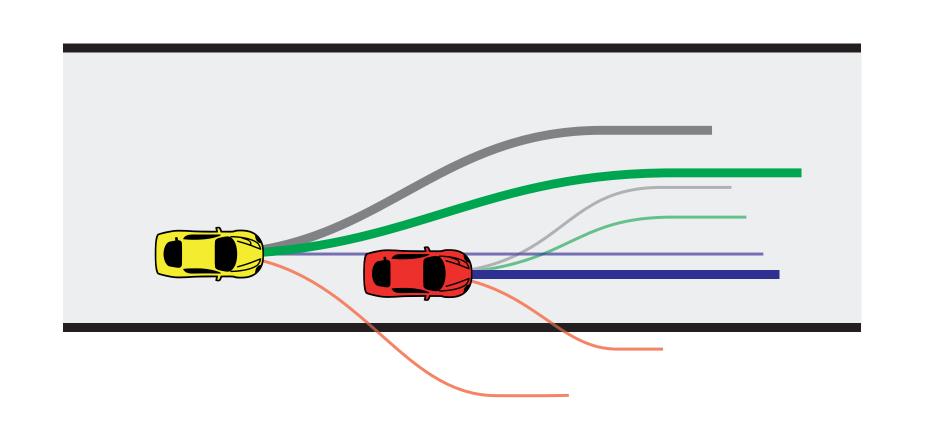
$$A = \begin{bmatrix} -1 & -1 & \\ -1 & \\ -10 & -10 & -10 \end{bmatrix}$$

$$B = \begin{bmatrix} -1 & -10 & \\ -1 & -1 & \\ -1 & -10 & \\ -1 & -10 & \\ -10 & -10 \end{bmatrix}$$

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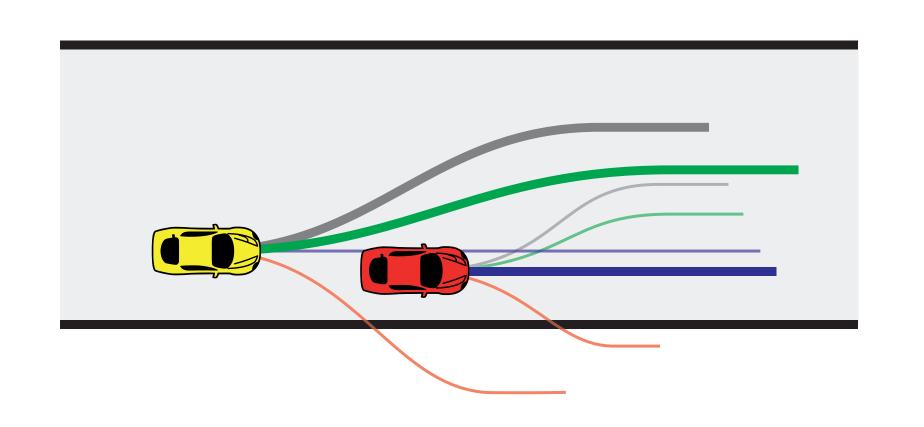
$$A = \begin{bmatrix} -1 \\ -1 \\ -1 \\ 0.88 \\ -10 \\ -10 \\ -10 \\ -10 \\ -10 \end{bmatrix}$$

$$B = \begin{bmatrix} -1 & -10 \\ -1 & -1 & -10 \\ 0.81 & 0.9 & -1 & -10 \\ & & -10 \end{bmatrix}$$

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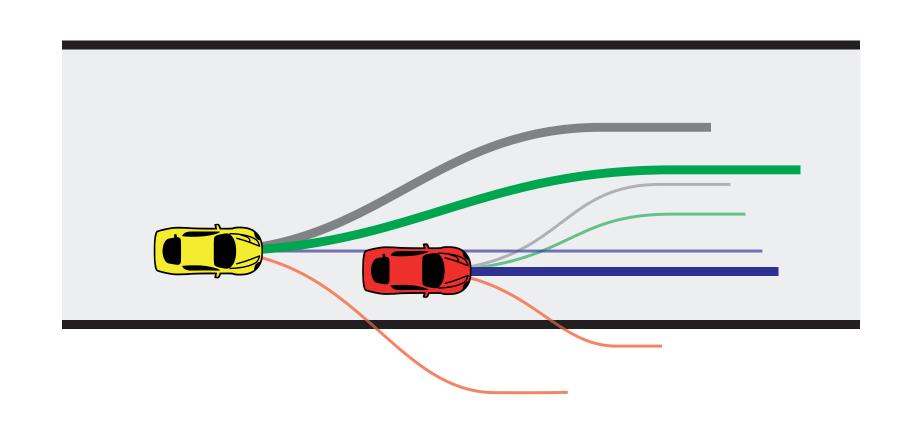
$$A = \begin{bmatrix} -1 \\ -1 \\ 0.88 + 0.5 \\ -10 \\ -10 \\ -10 \\ -10 \\ -10 \end{bmatrix}$$

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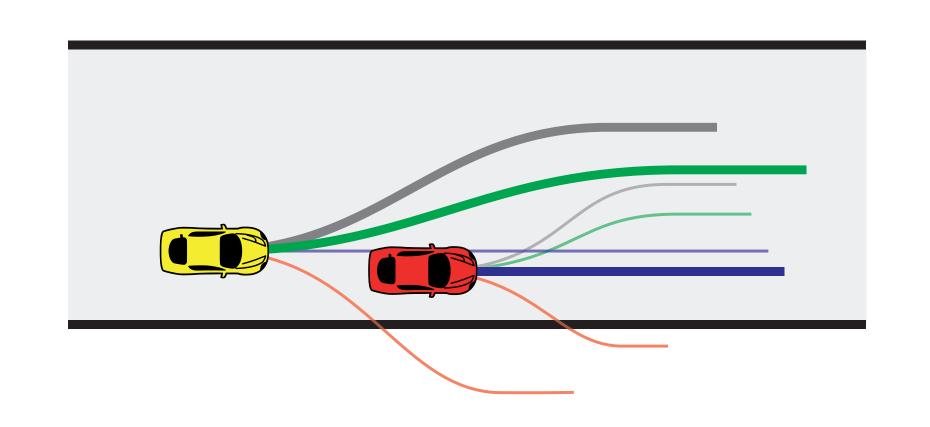
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- Additional reward for staying in front at the end of the horizon



$$A = \begin{bmatrix} 0.83 + 0.5 & -1 & 0.83 & 0.83 + 0.5 \\ 0.85 + 0.5 & -1 & -1 & 0.85 + 0.5 \\ 0.88 + 0.5 & 0.88 & -1 & 0.88 + 0.5 \\ -10 & -10 & -10 & -10 \end{bmatrix}$$

$$B = \begin{bmatrix} 0.81 & -1 & 0.86 + 0.5 & -10 \\ 0.81 & -1 & -1 & -10 \\ 0.81 & 0.9 + 0.5 & -1 & -10 \\ 0.81 + 0.5 & 0.9 + 0.5 & 0.86 + 0.5 & -10 \end{bmatrix}$$

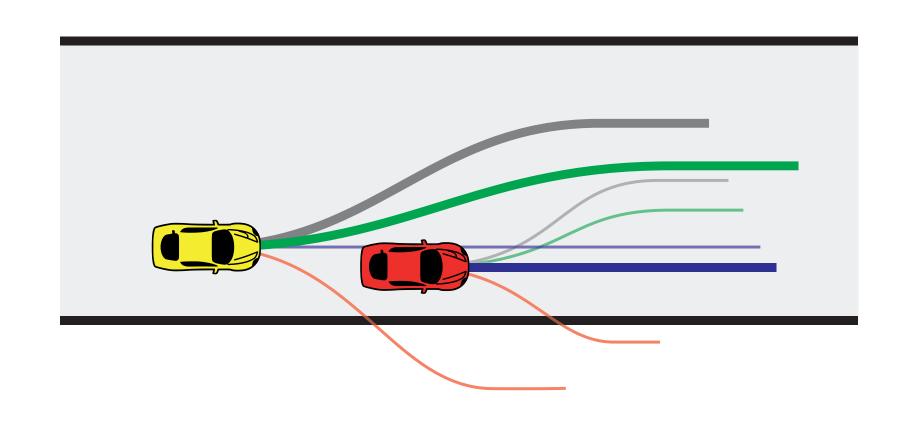
Sequential Game

Cooperative Game

### **Blocking Game**

- Same collision structure as the cooperative game, but:
- Additional reward for staying in front at the end of the horizon

### How should a car choose a trajectory?



$$A = \begin{bmatrix} 0.83 + 0.5 & -1 & 0.83 & 0.83 + 0.5 \\ 0.85 + 0.5 & -1 & -1 & 0.85 + 0.5 \\ 0.88 + 0.5 & 0.88 & -1 & 0.88 + 0.5 \\ -10 & -10 & -10 & -10 \end{bmatrix}$$

$$B = \begin{bmatrix} 0.81 & -1 & 0.86 + 0.5 & -10 \\ 0.81 & -1 & -1 & -10 \\ 0.81 & 0.9 + 0.5 & -1 & -10 \\ 0.81 + 0.5 & 0.9 + 0.5 & 0.86 + 0.5 & -10 \end{bmatrix}$$

# Equilibria concepts

- Find an equilibrium trajectory pair of the bimatrix game
  - Pure strategies (no mixed strategies)
  - $(i^*, j^*) \in \Gamma^1 \times \Gamma^2$  is an equilibrium trajectory pair

### Stackelberg Equilibria

- Game with leader-follower structure
- Leader can enforce his trajectory on the follower
- Follower plays the **best response**:  $R(i) = \arg \max_{i,j} b_{i,j}$

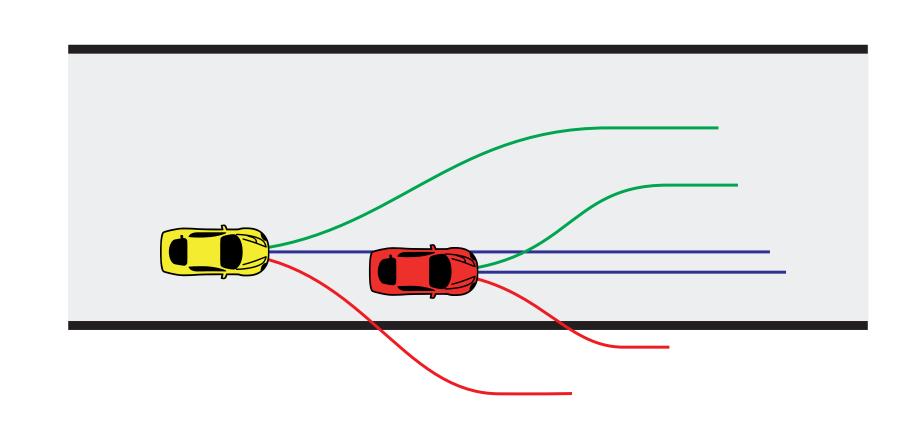
$$i^* = \arg \max_{i \in \Gamma^1} \min_{j \in R(i)} a_{i,j}$$

$$j^* = R(i^*)$$

### Nash Equilibria

None of the players has a benefit from unilaterally changing the trajectory

$$a_{i^*,j^*} \ge a_{i,j^*} \quad \forall i \in \Gamma^1$$
  
 $b_{i^*,j^*} \ge b_{i^*,j} \quad \forall j \in \Gamma^2$ 



#### sequential game

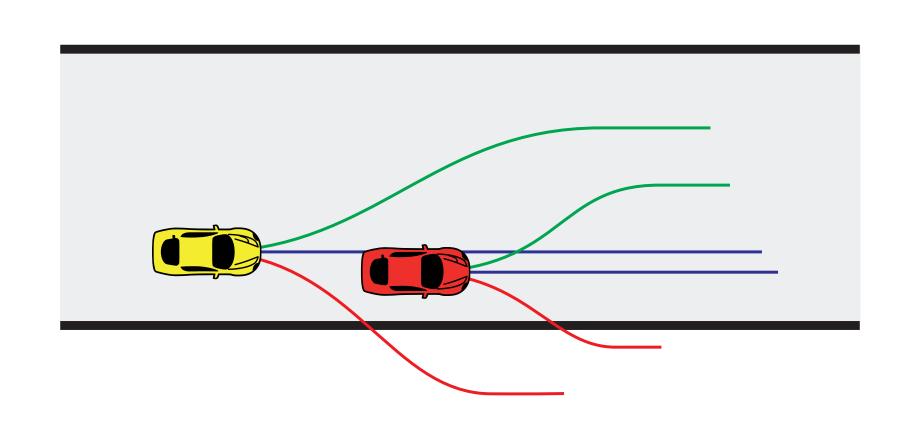
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$$B = \begin{bmatrix} 0.81 & 0.86 & -10 \\ 0.81 & -1 & -10 \\ 0.81 & 0.86 & -10 \end{bmatrix}$$

#### cooperative game

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sequential game

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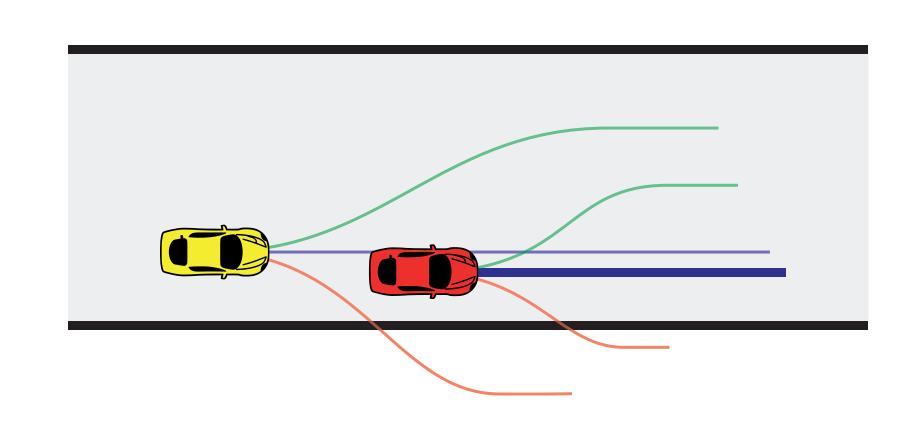
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cooperative game

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sequential game

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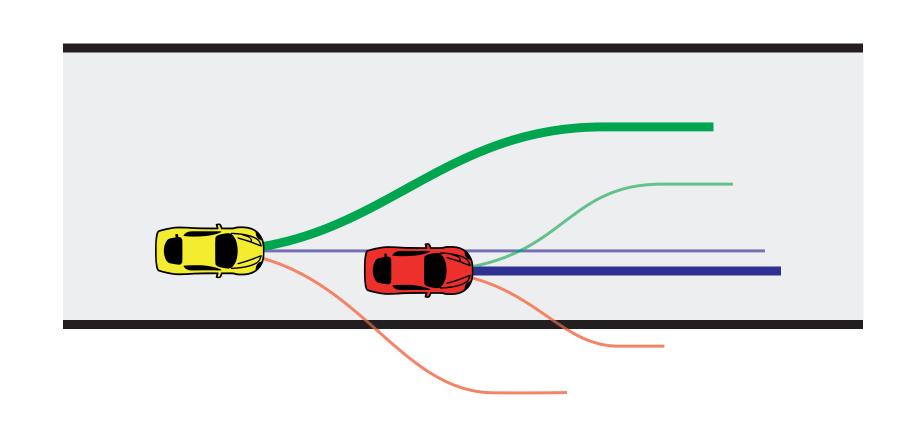
$$B = \begin{bmatrix} 0.81 & 0.86 & -10 \\ 0.81 & -1 & -10 \\ 0.81 & 0.86 & -10 \end{bmatrix}$$

cooperative game

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sequential game

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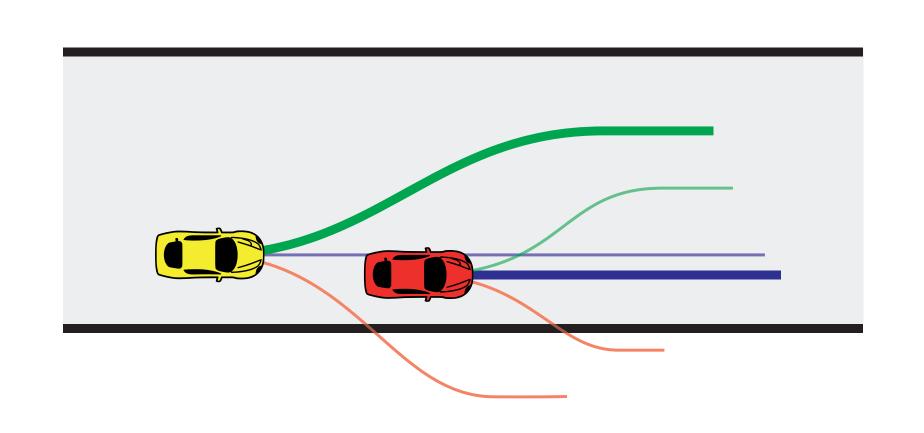
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sequential game

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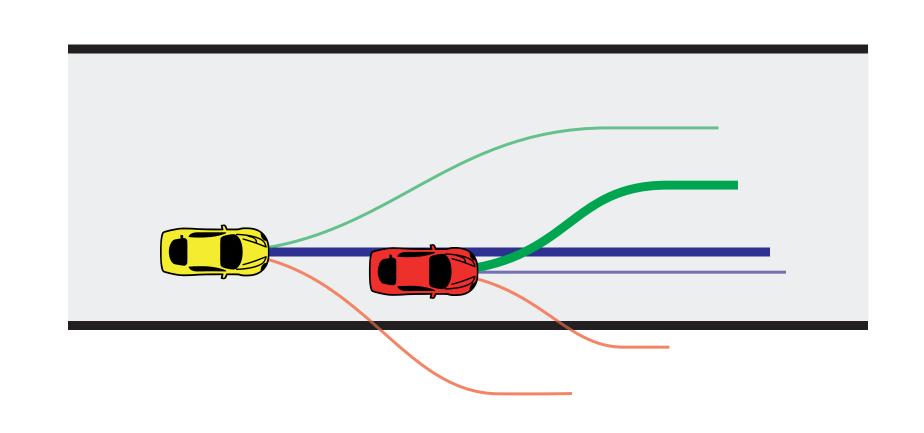
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sequential game

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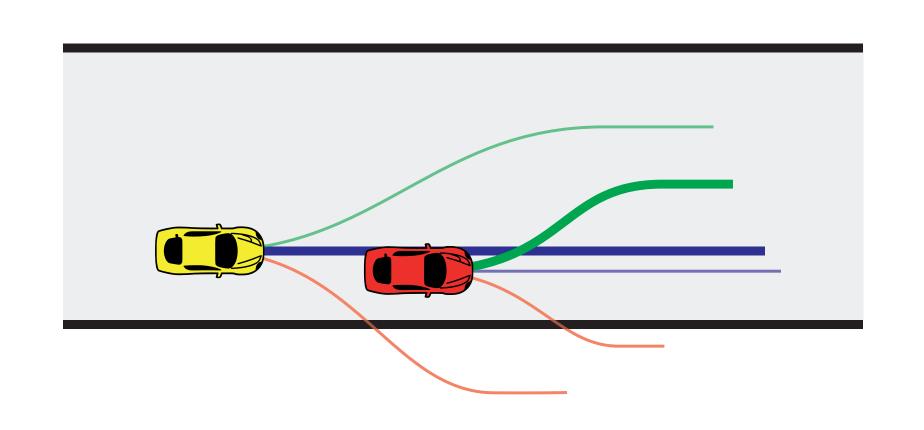
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cooperative game

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#### sequential game

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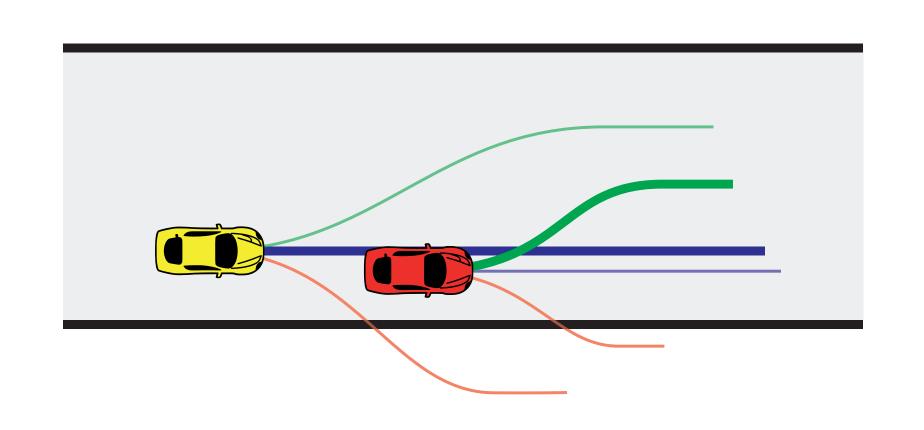
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- The sequential game can be solved by sequential maximizing
- Sequential game feasible -> equilibrium of the cooperative game
  - Predicting ideal behavior of other cars and play best response is a Nash equilibrium



### sequential game

$$A = \begin{bmatrix} 0.83 & 0.83 & 0.83 \\ 0.88 & 0.88 & 0.88 \\ -10 & -10 & -10 \end{bmatrix}$$

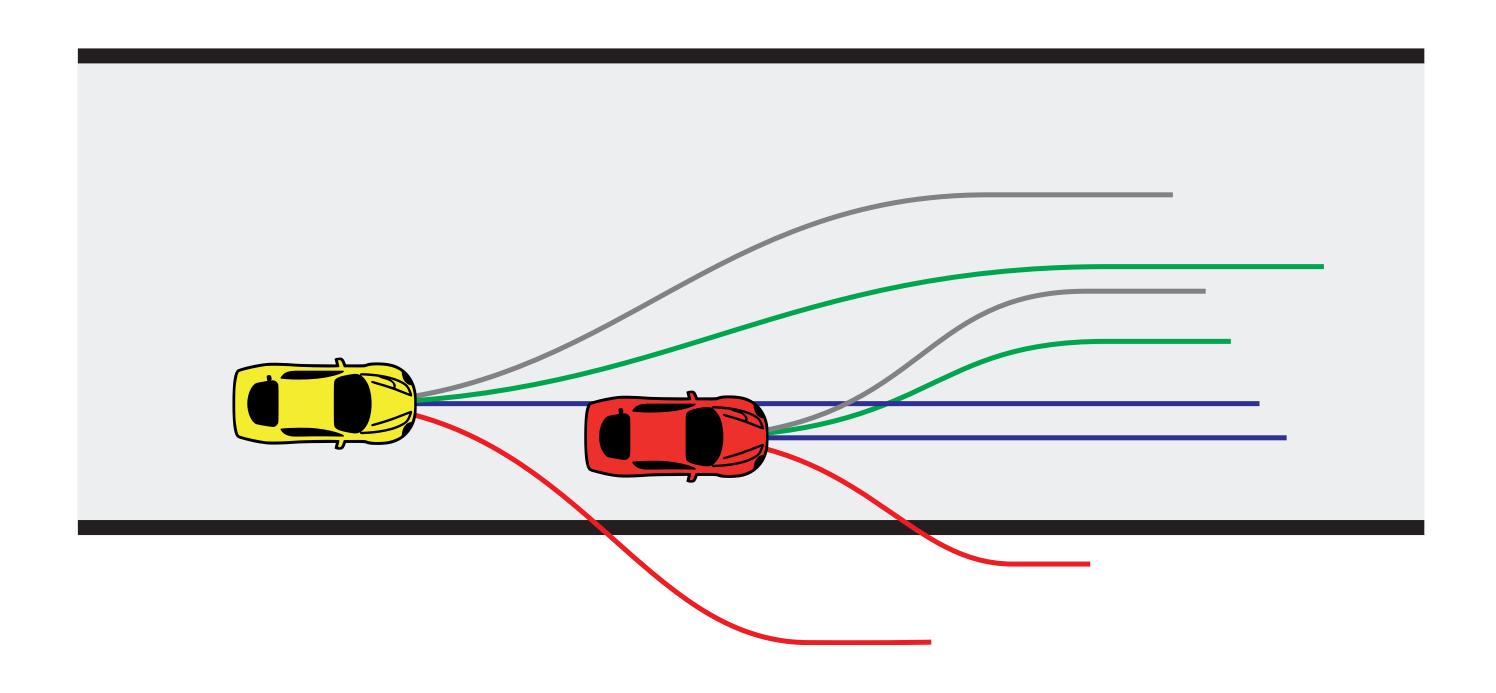
$$B = \begin{bmatrix} 0.81 & 0.86 & -10 \\ 0.81 & -1 & -10 \\ 0.81 & 0.86 & -10 \end{bmatrix}$$

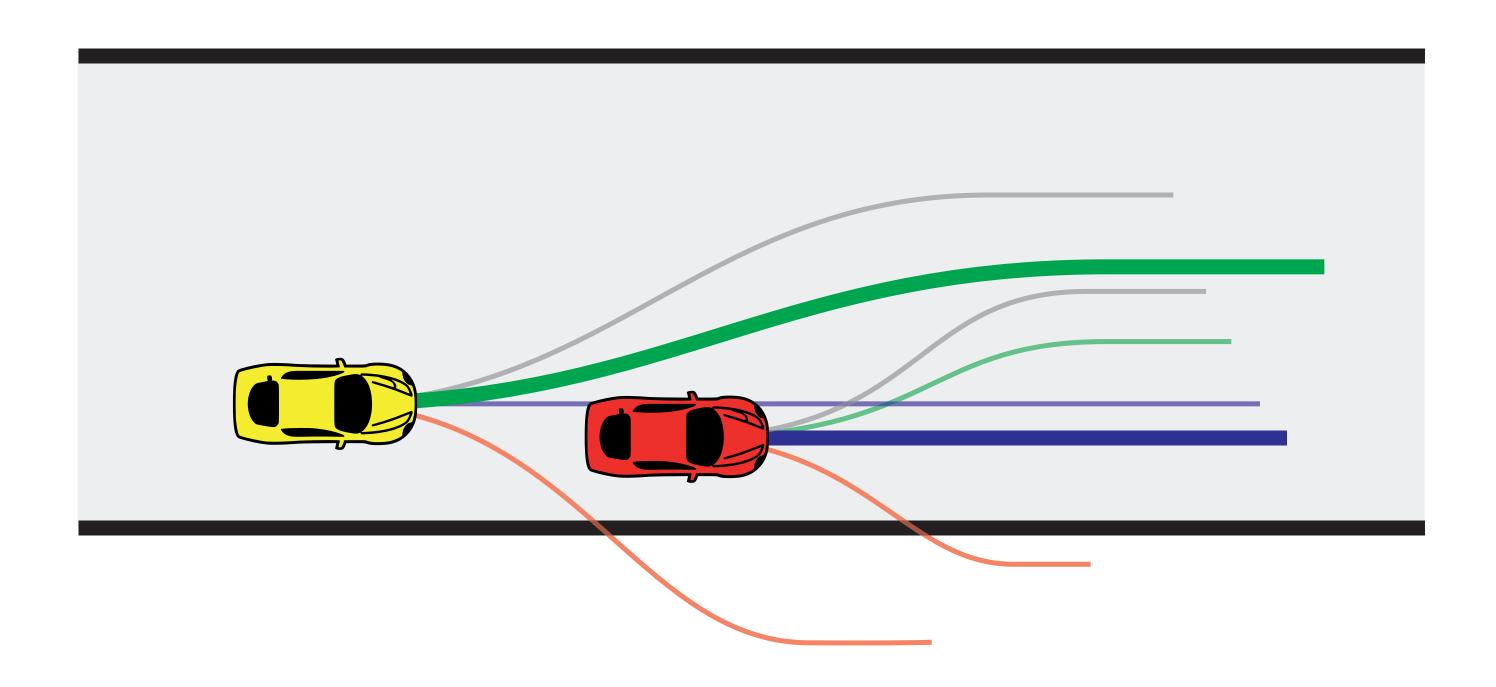
#### cooperative game

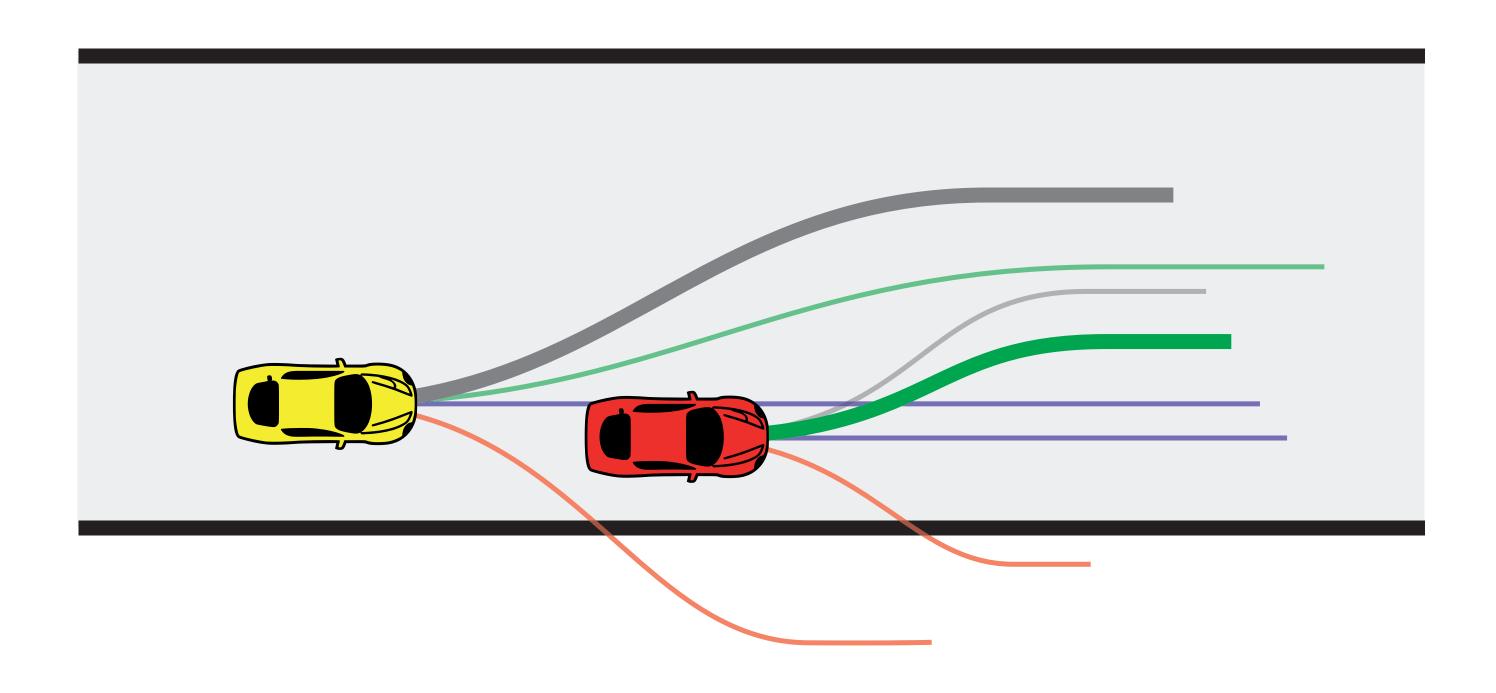
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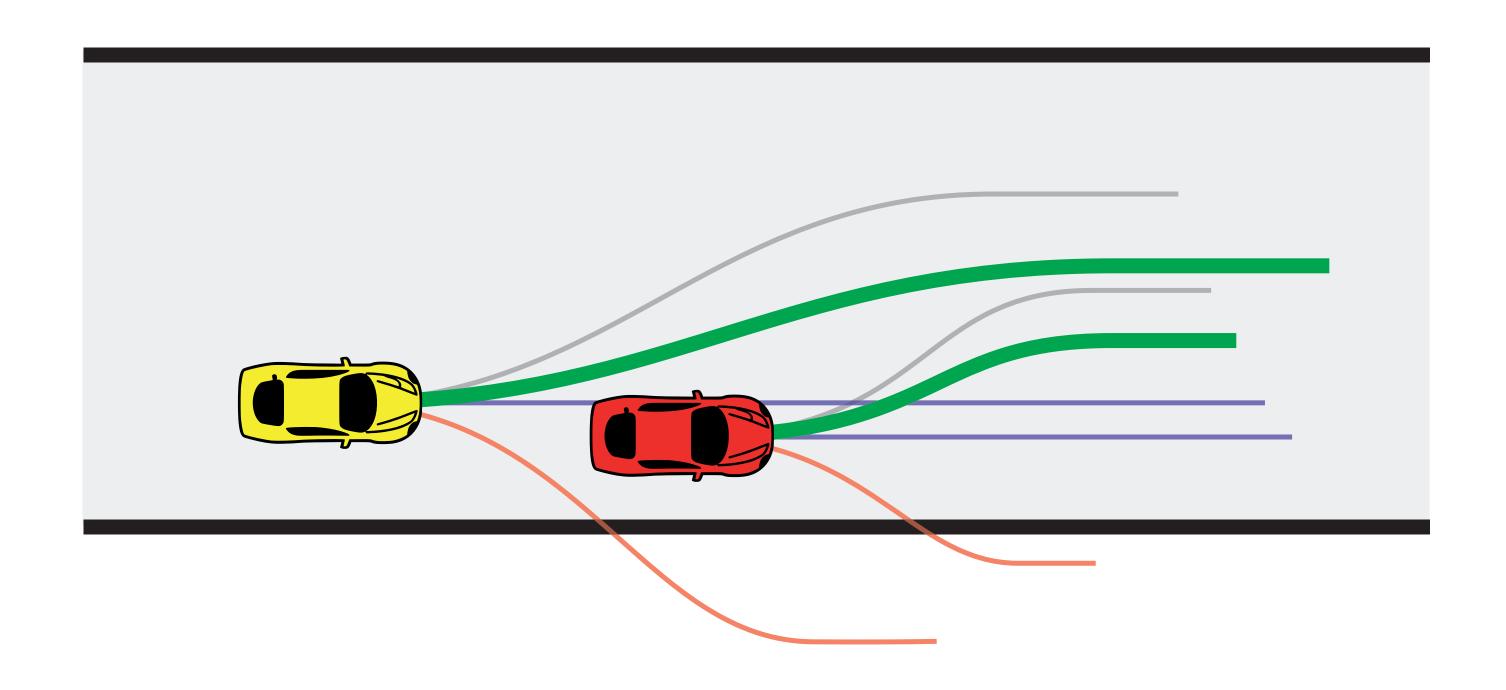
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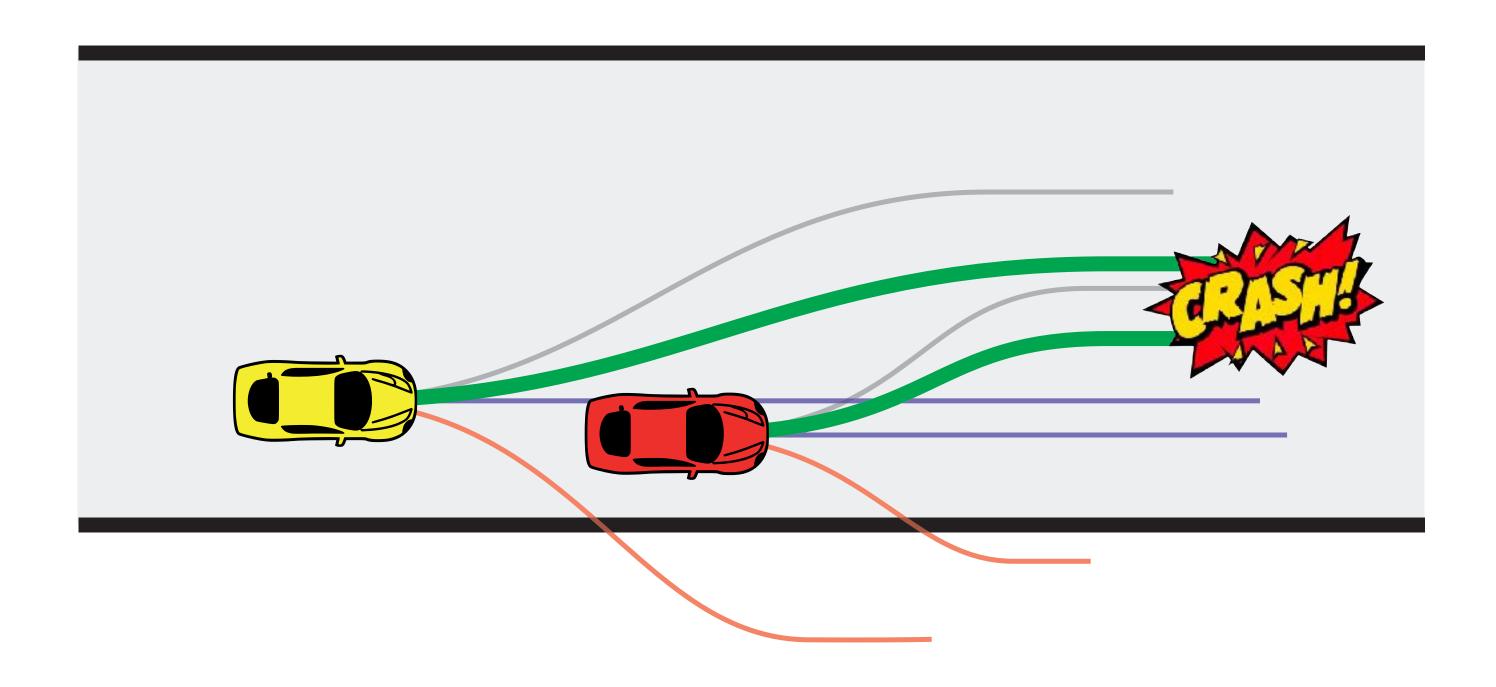
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- Sequential game feasible -> equilibrium of the cooperative game
  - Predicting ideal behavior of other cars and play best response is a Nash equilibrium
- Cooperative game is feasible if there exists a feasible trajectory pair

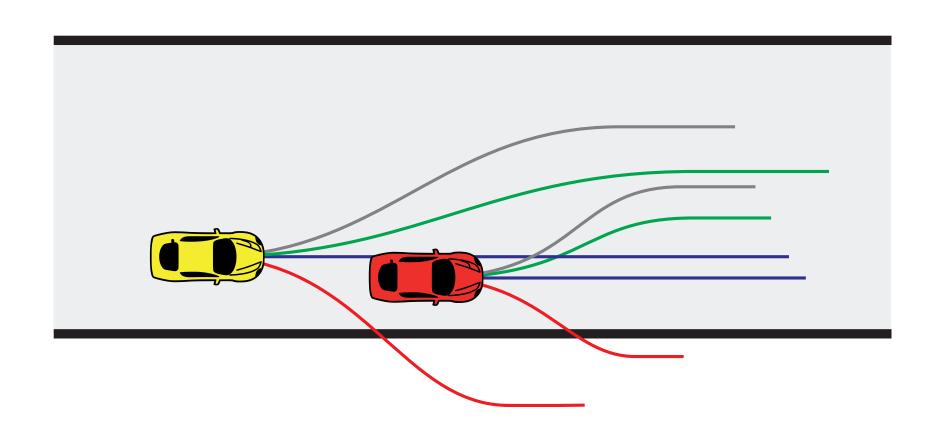






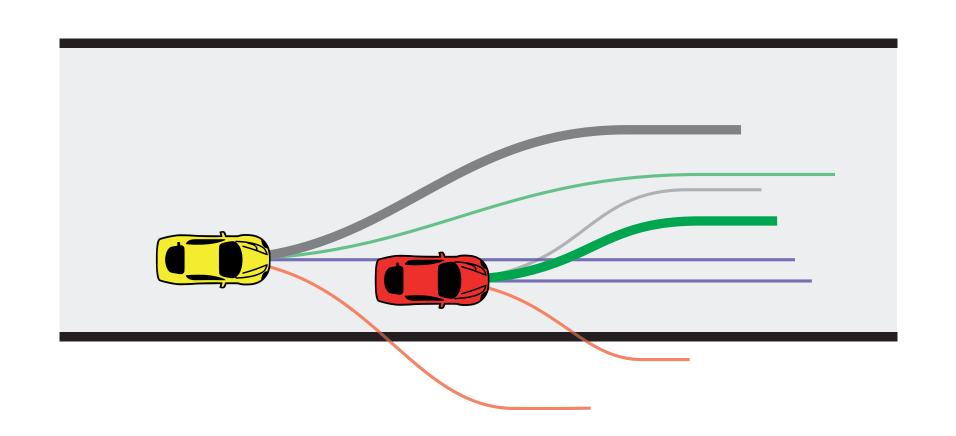






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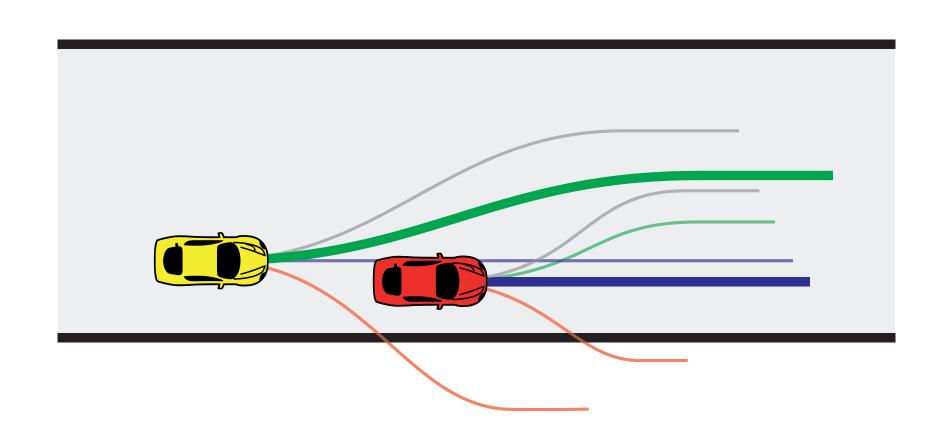


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If there exists a blocking trajectory and the staying ahead reward is big enough, the Stackelberg equilibrium is a blocking trajectory pair

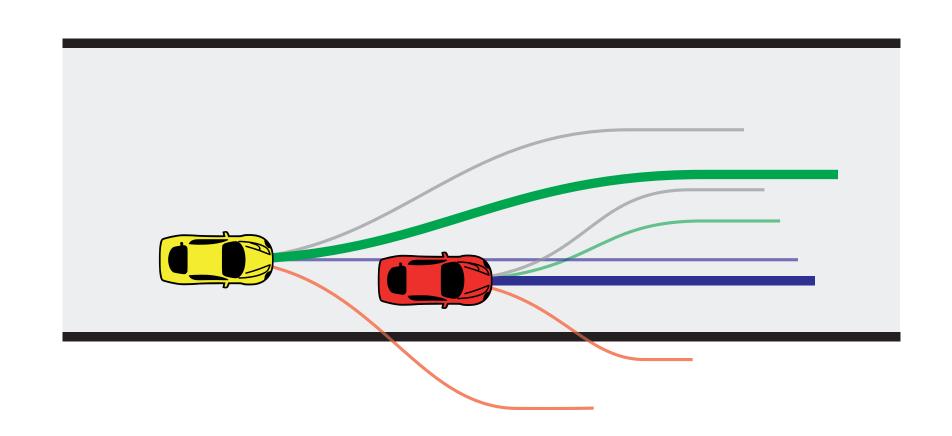




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- If there exists a blocking trajectory and the staying ahead reward is big enough, the Stackelberg equilibrium is a blocking trajectory pair
- A blocking trajectory is **not** a Nash equilibrium (unless it is a Nash equilibrium of the cooperative game)

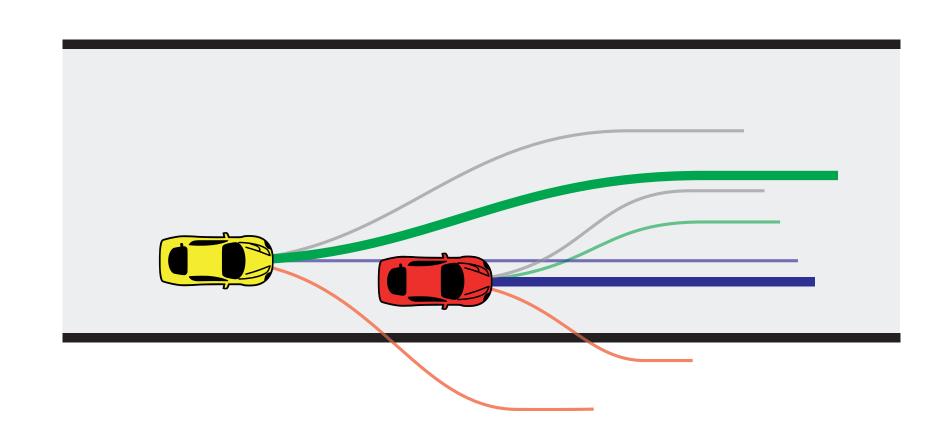


$$A = \begin{bmatrix} 0.83 + 0.5 & -1 & 0.83 & 0.83 + 0.5 \\ 0.85 + 0.5 & -1 & -1 & 0.85 + 0.5 \\ 0.88 + 0.5 & 0.88 & -1 & 0.88 + 0.5 \\ -10 & -10 & -10 & -10 \end{bmatrix}$$

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### Stackelberg equilibrium seems best for all games



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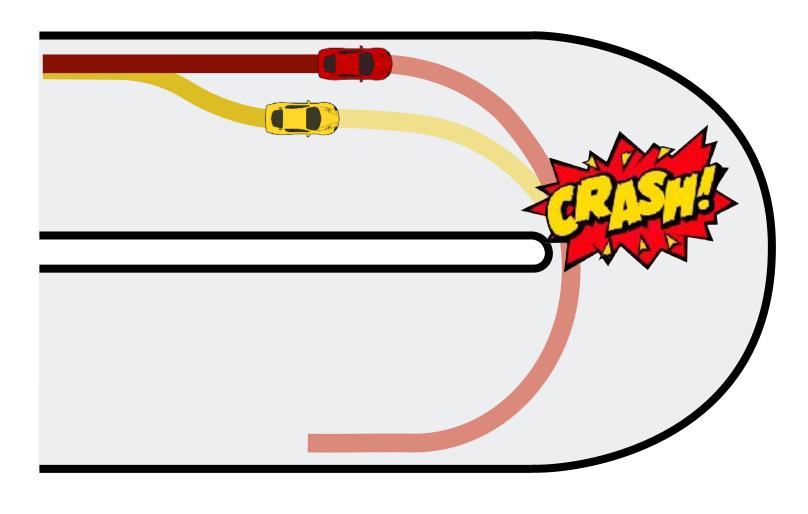
What is the resulting behavior of these games?

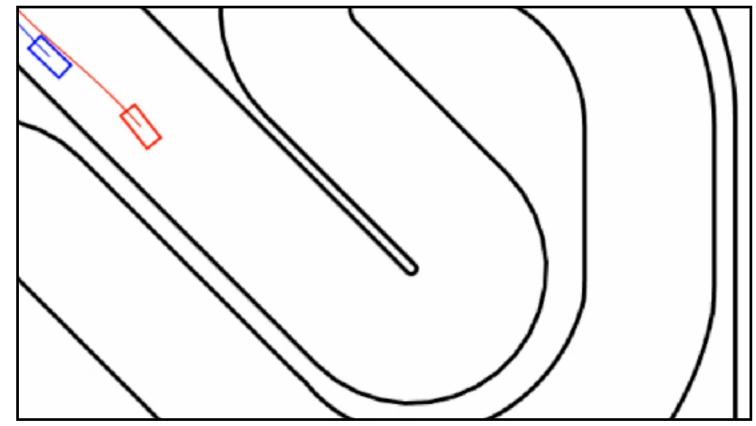


- Play game in a receding horizon fashion
  - Solve game + MPC apply first input repeat
- Trajectory pruning based on viability and discriminating kernel
  - Viab —> aggressive driver / Disc —> cautious driver
- ▶ 500 different initial conditions, each run 4.5 laps
  - Both cars start close to each other

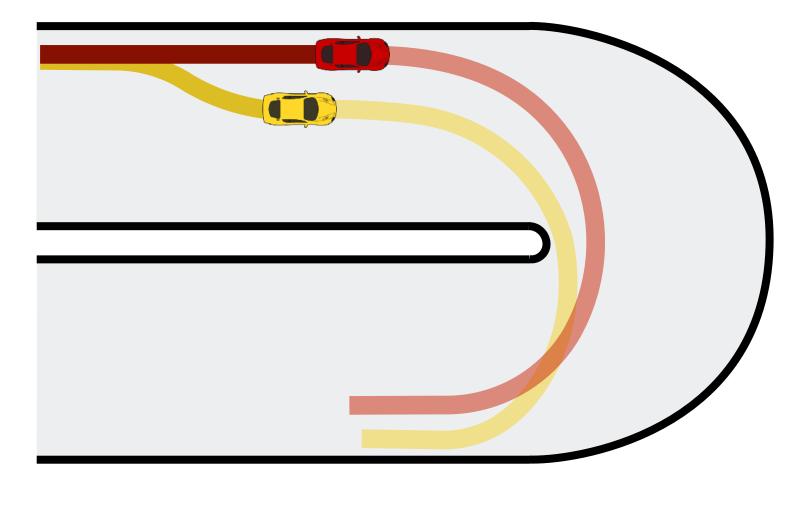
	sequential game	cooperative game	blocking game
# of overtaking maneuvers	113	857	414
colliding time steps per lap	2.4	2.0	2.3

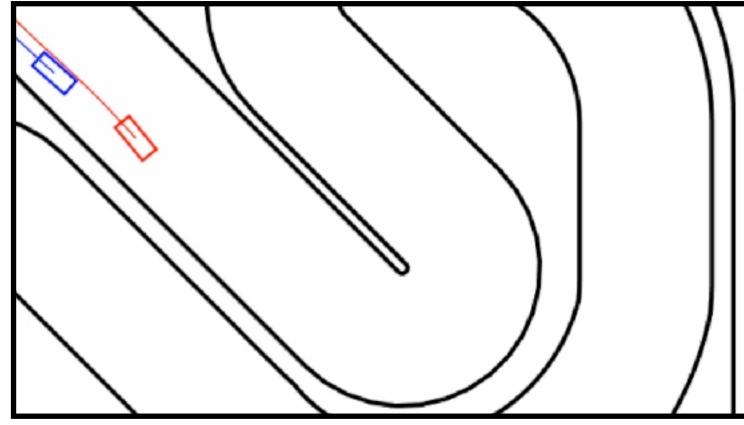
sequential game



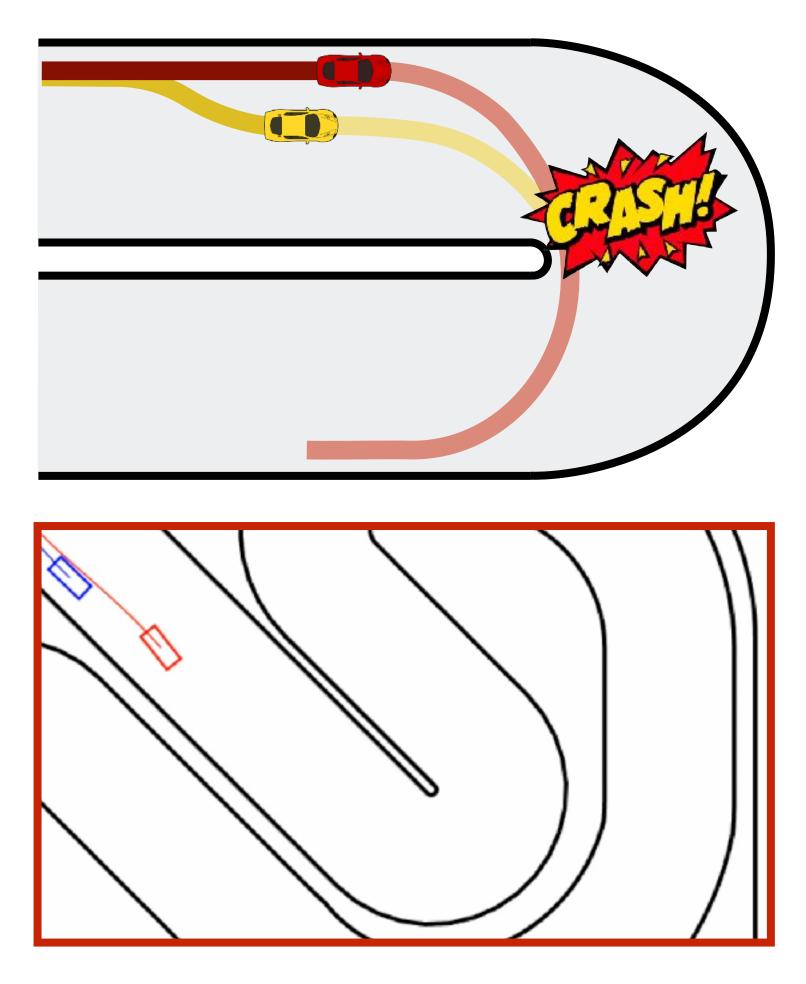


### cooperative game

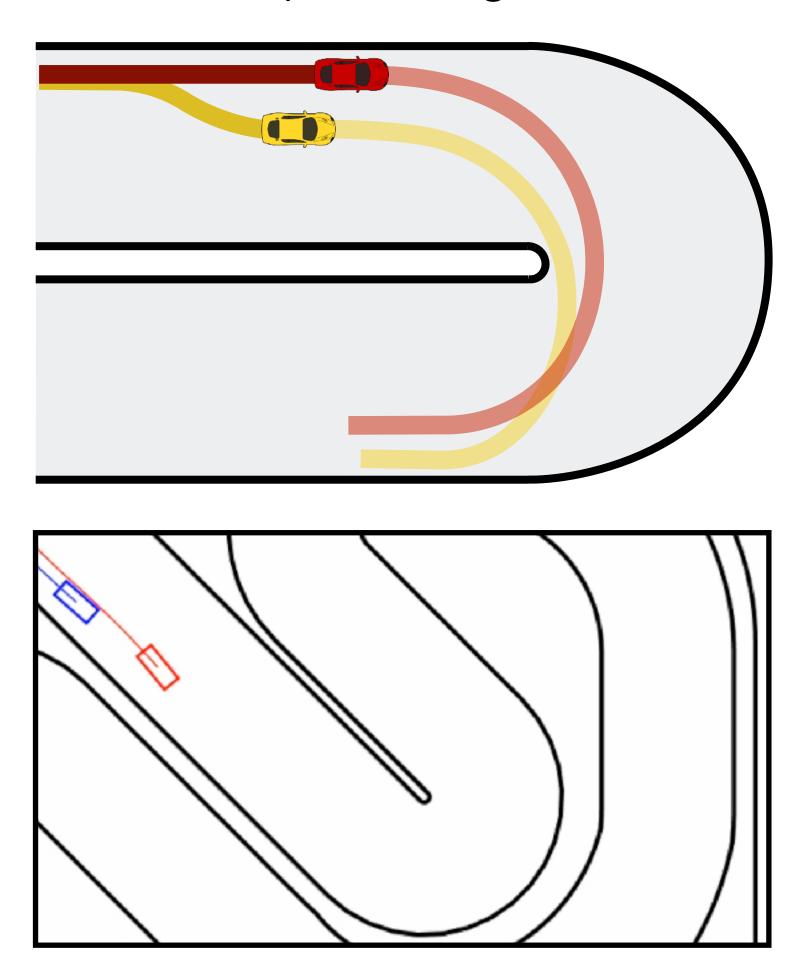




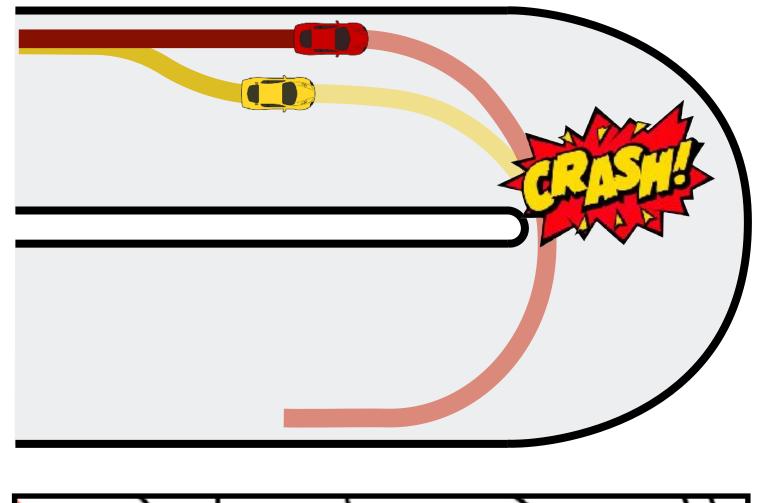
sequential game

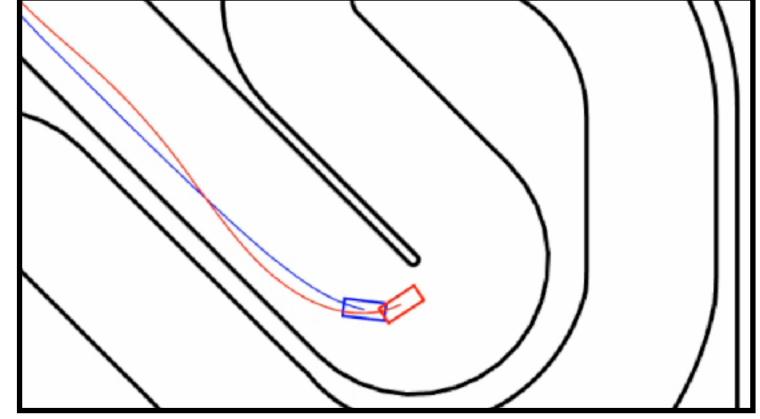


### cooperative game

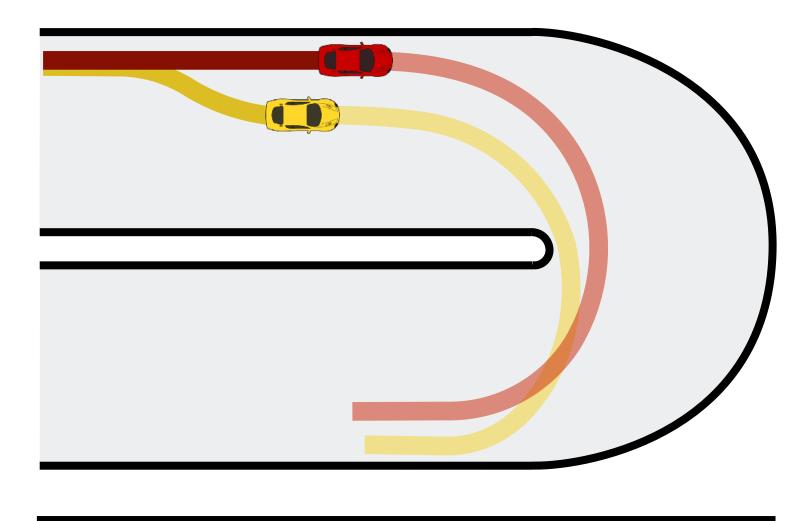


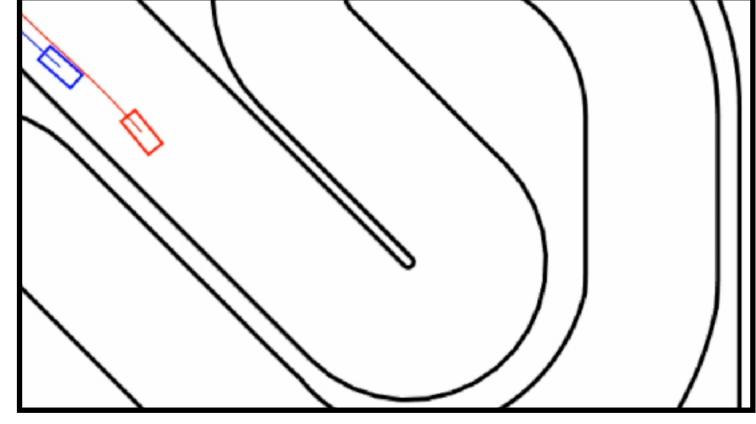




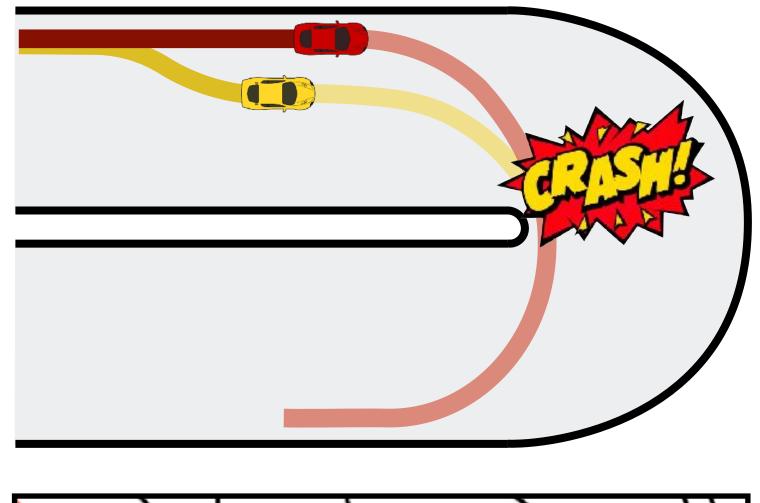


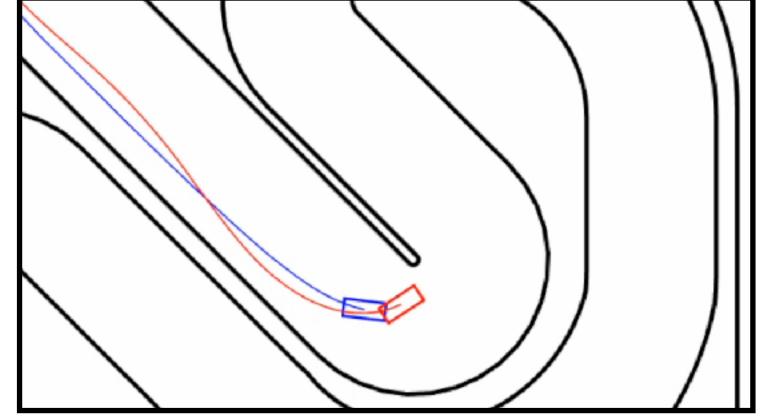
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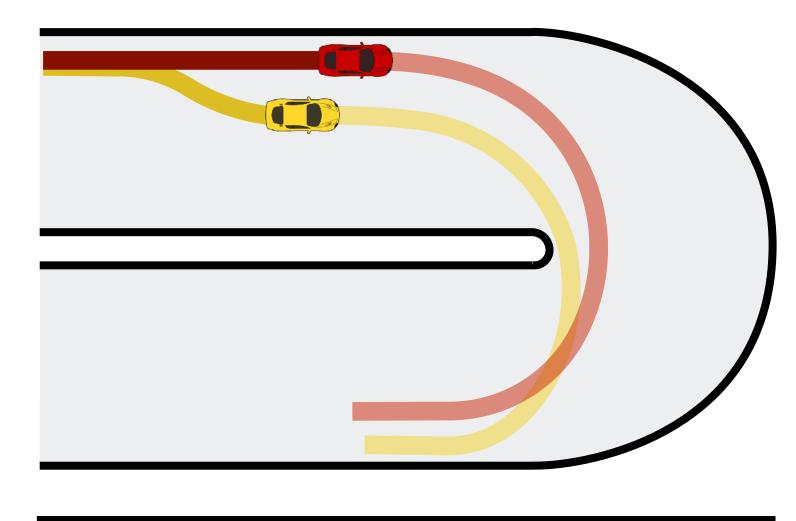


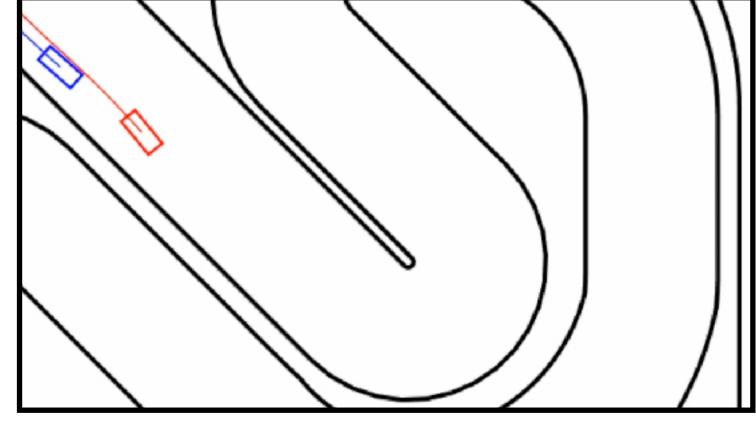


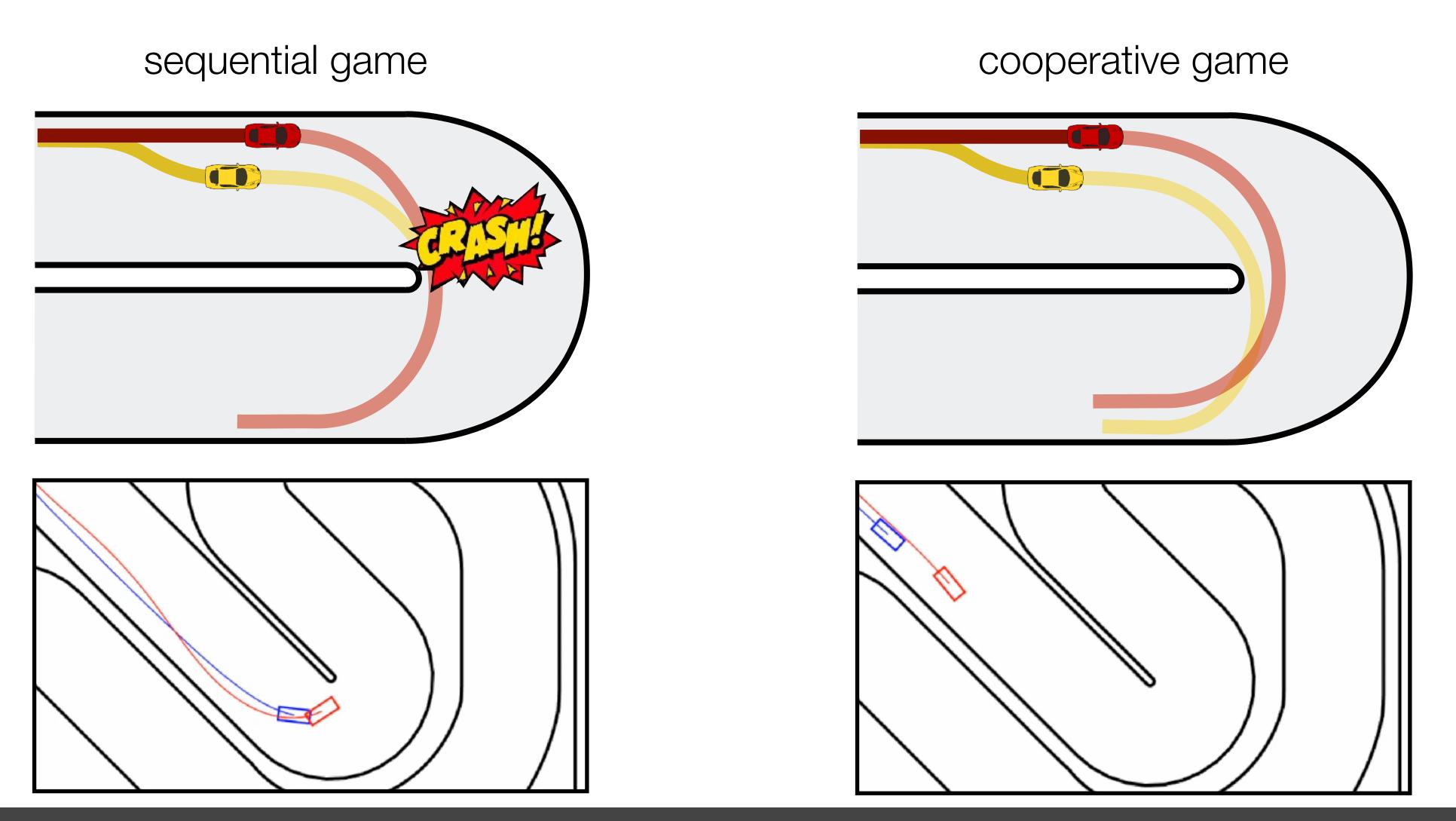




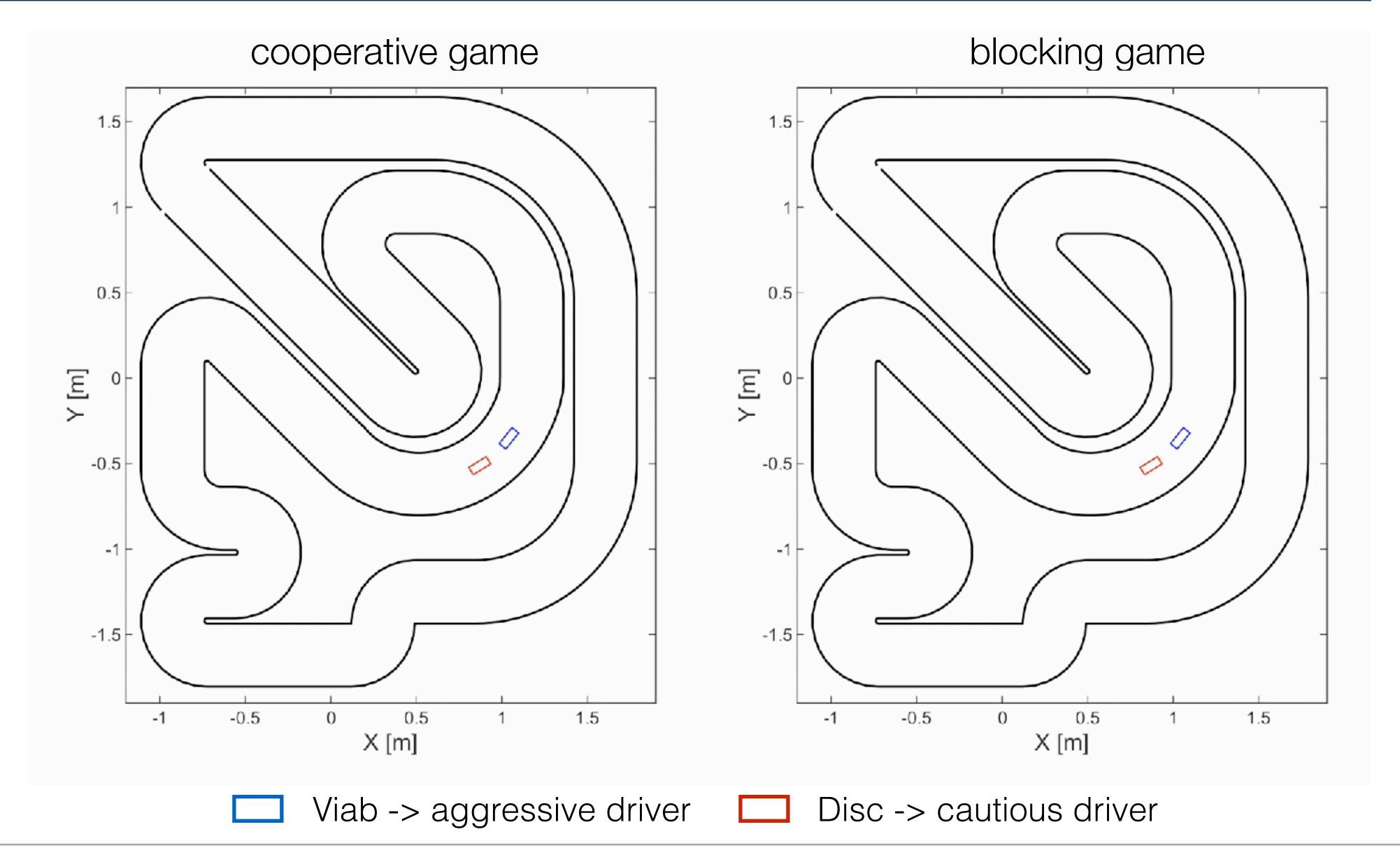
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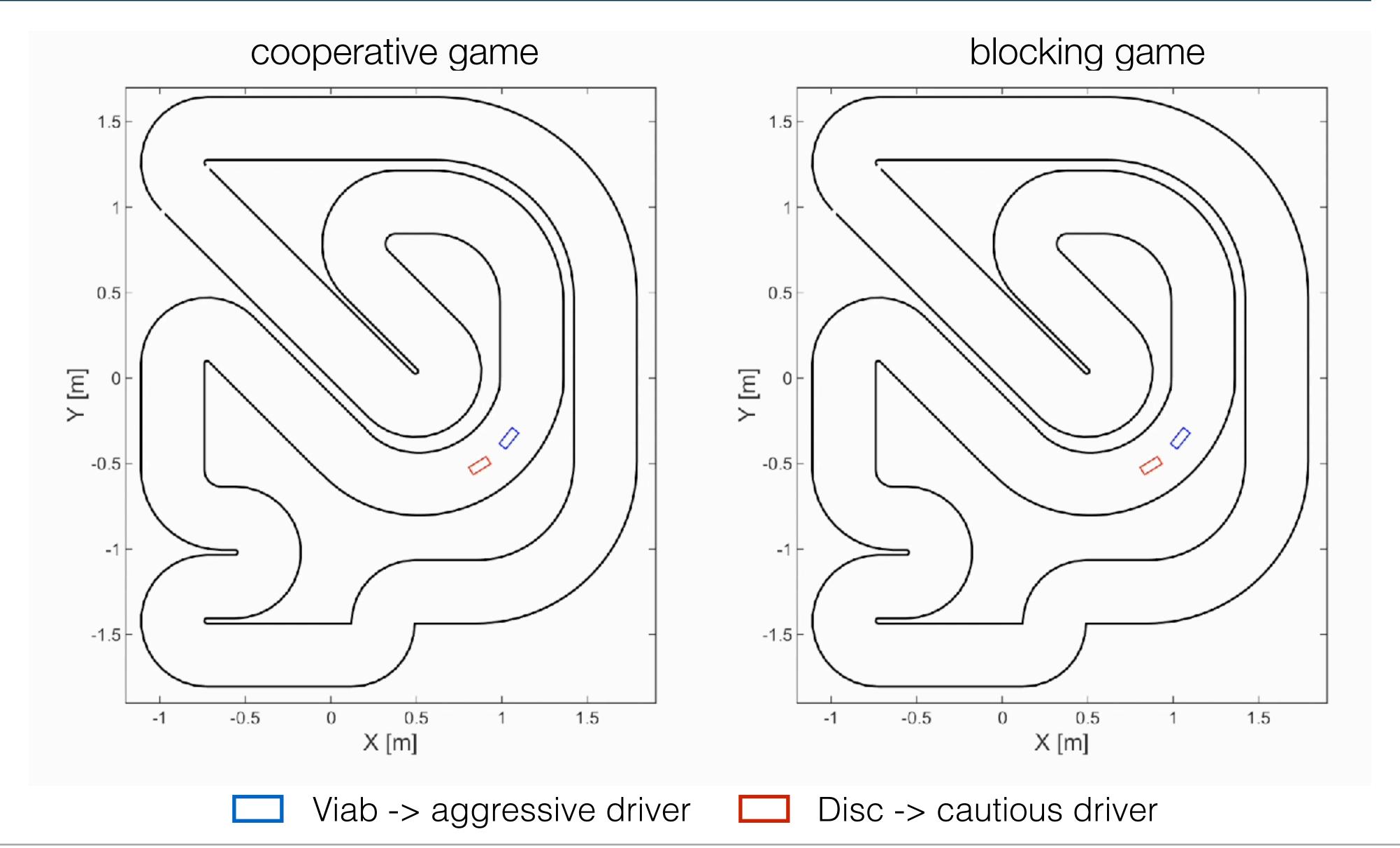


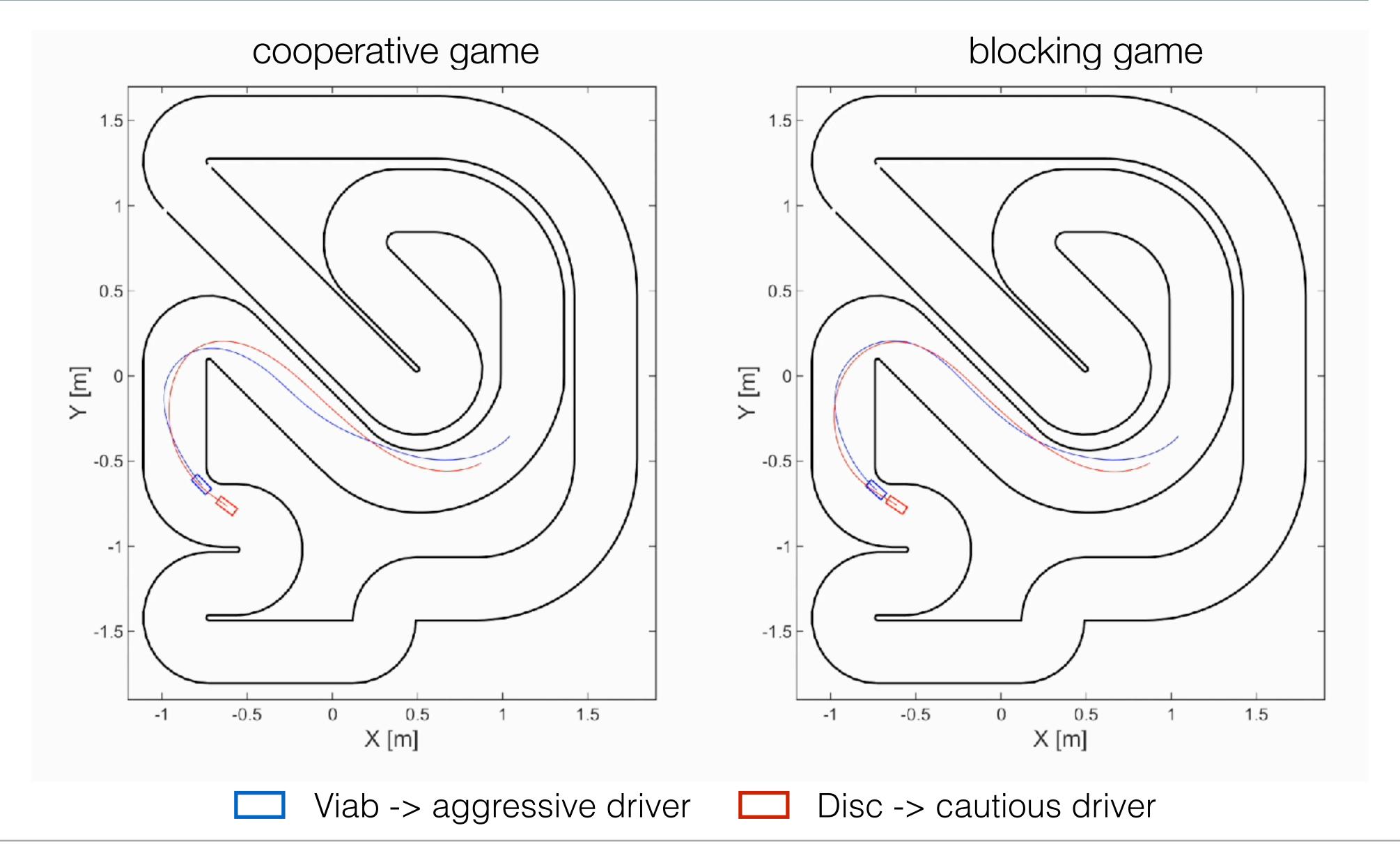


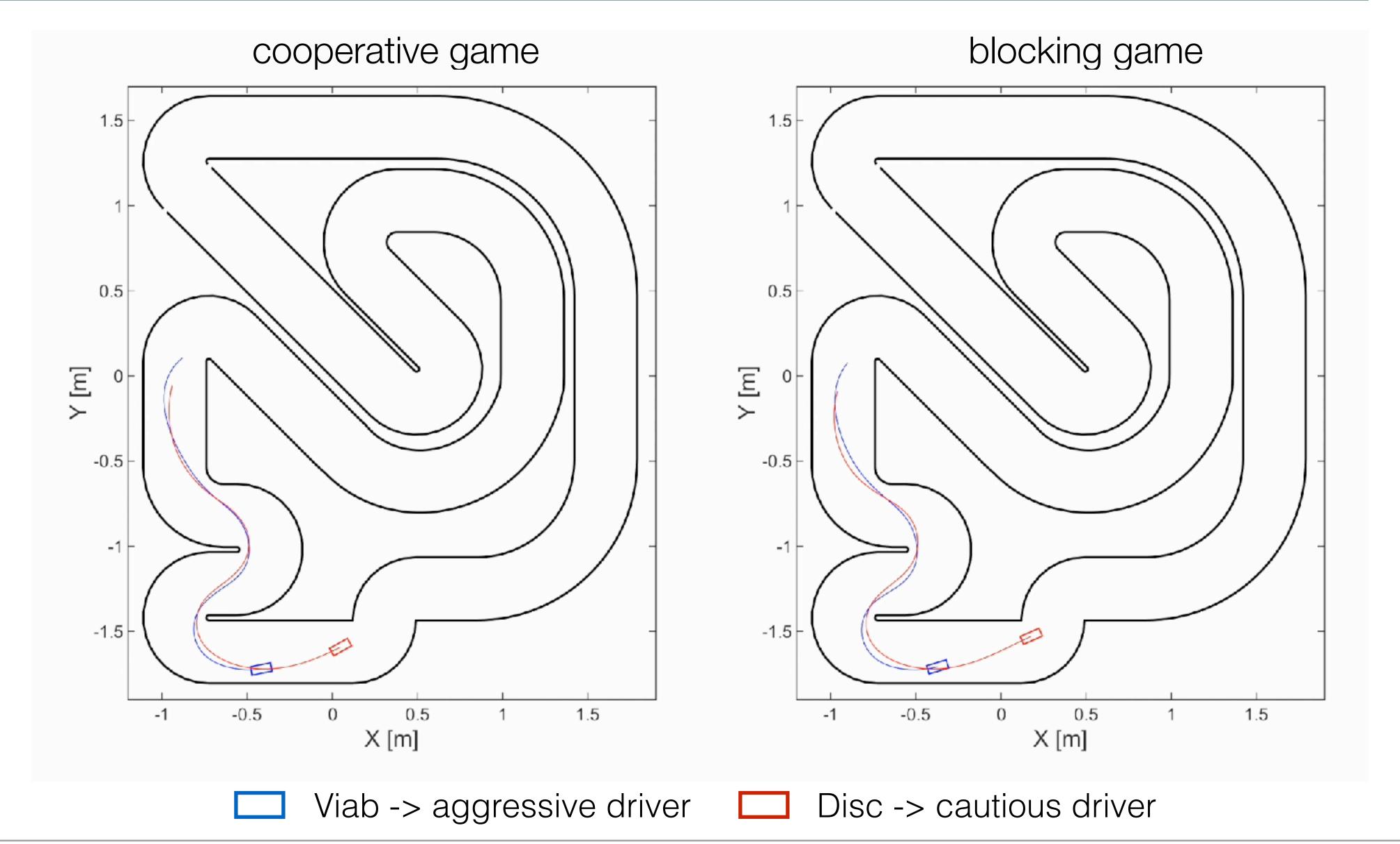


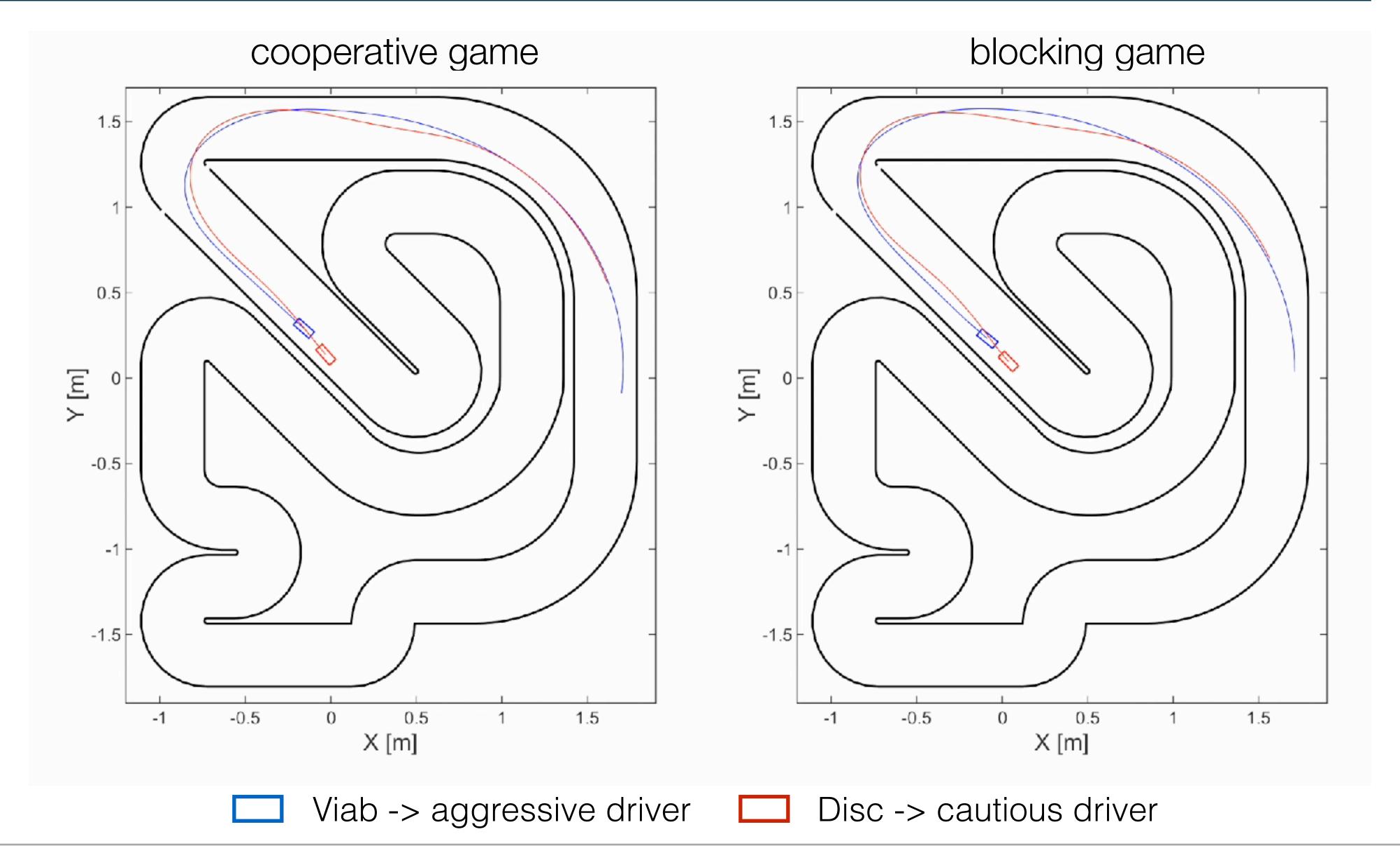
How do the cars drive?

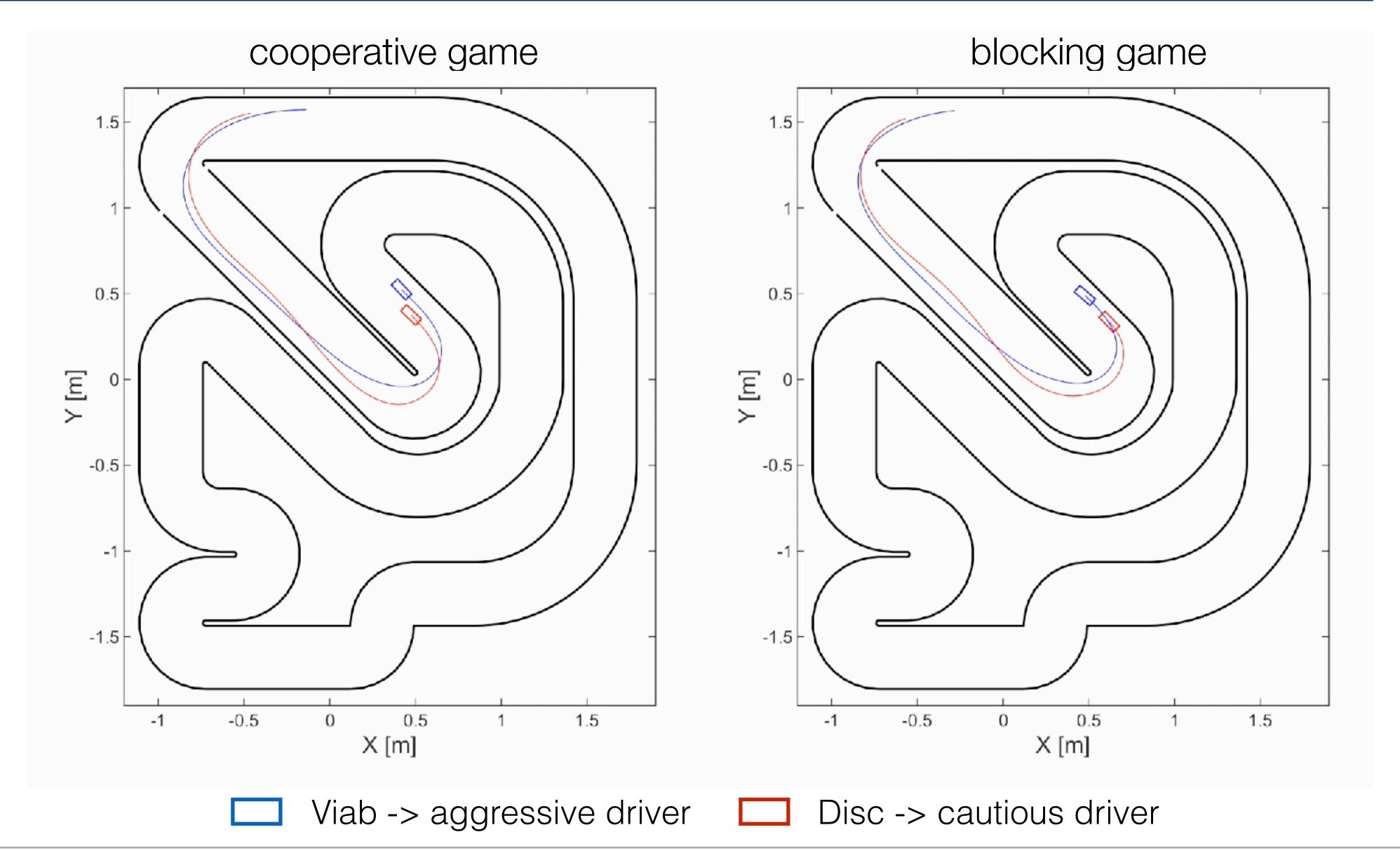


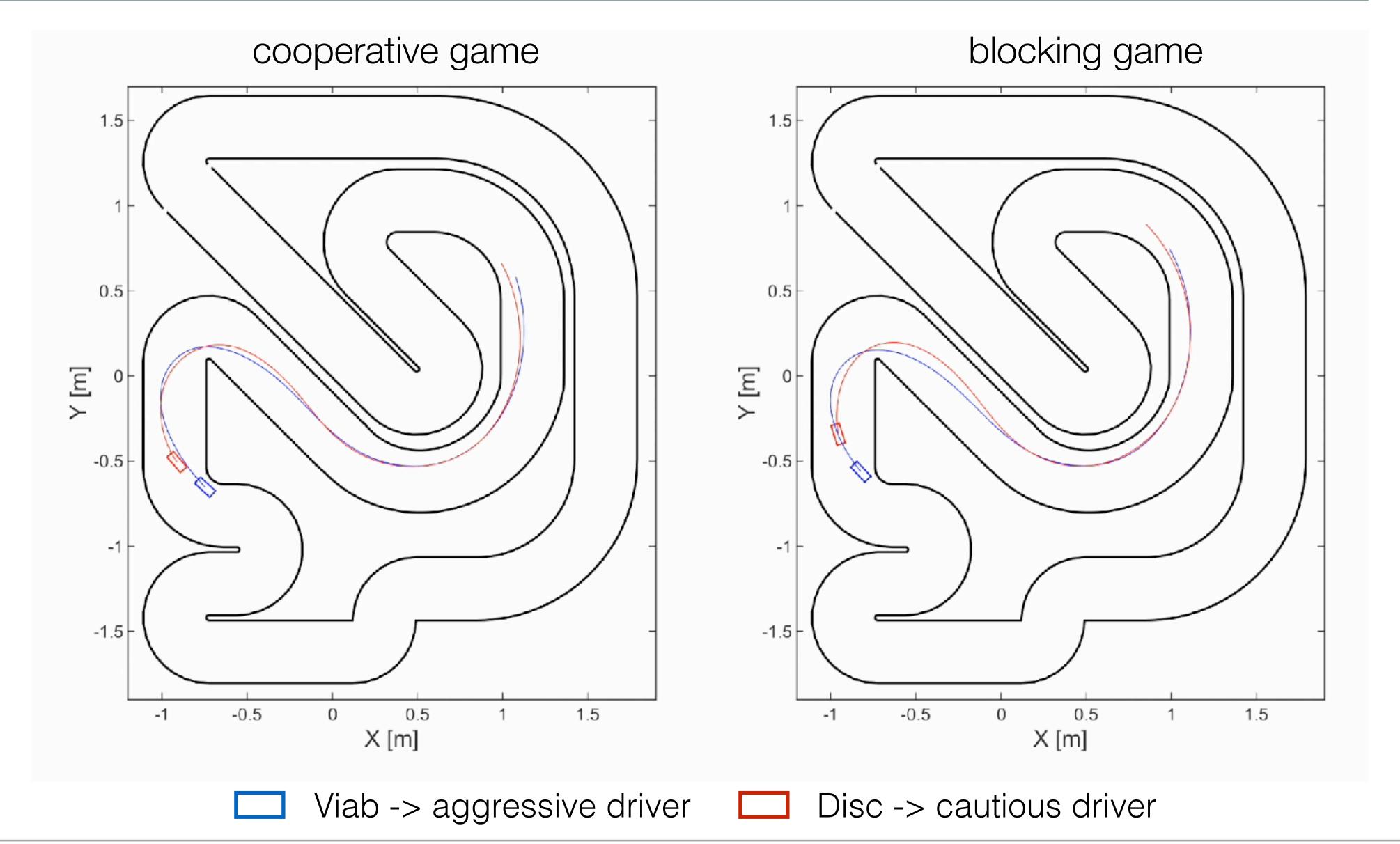






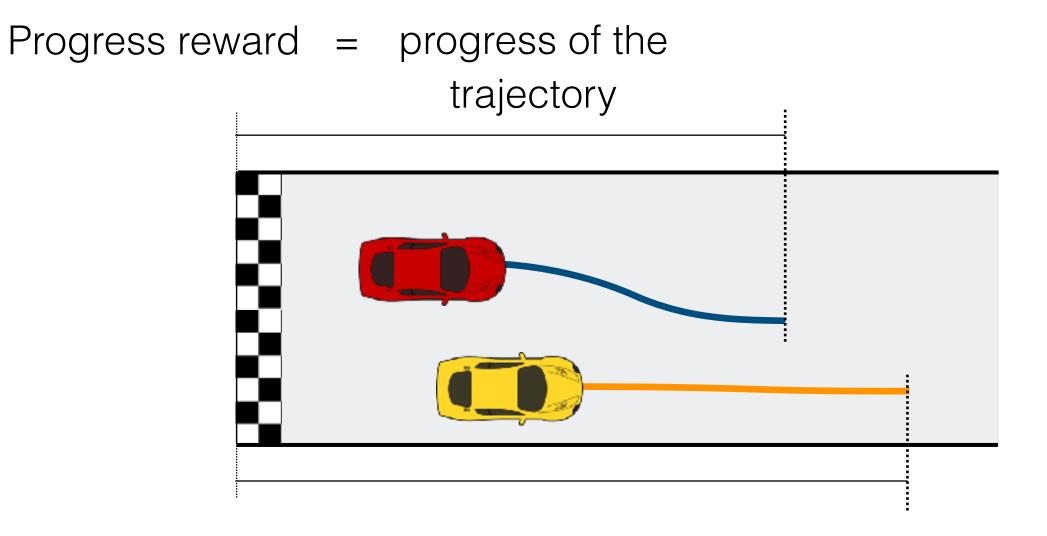






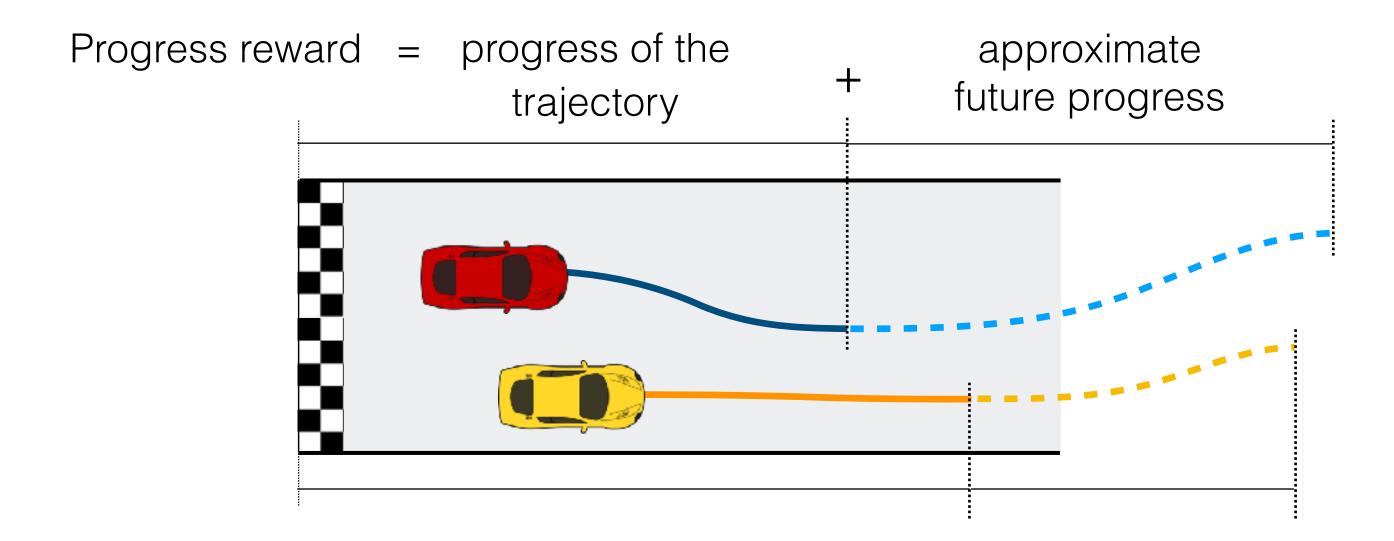
# Real-Time Implementation

- Building the bimatrix game can be computationally expensive
  - Entries in the matrix grow quadratically with the number of trajectories
  - Collision checks become a bottle neck
  - 1,000 trajectories -> 32,000,000 collision checks (20ms discretization)
- ▶ Reduce prediction horizon from 3 to 2 steps (~200 instead of 3,000 trajectories)
- Only build bimatrix game for the 60 best trajectories (beam search)



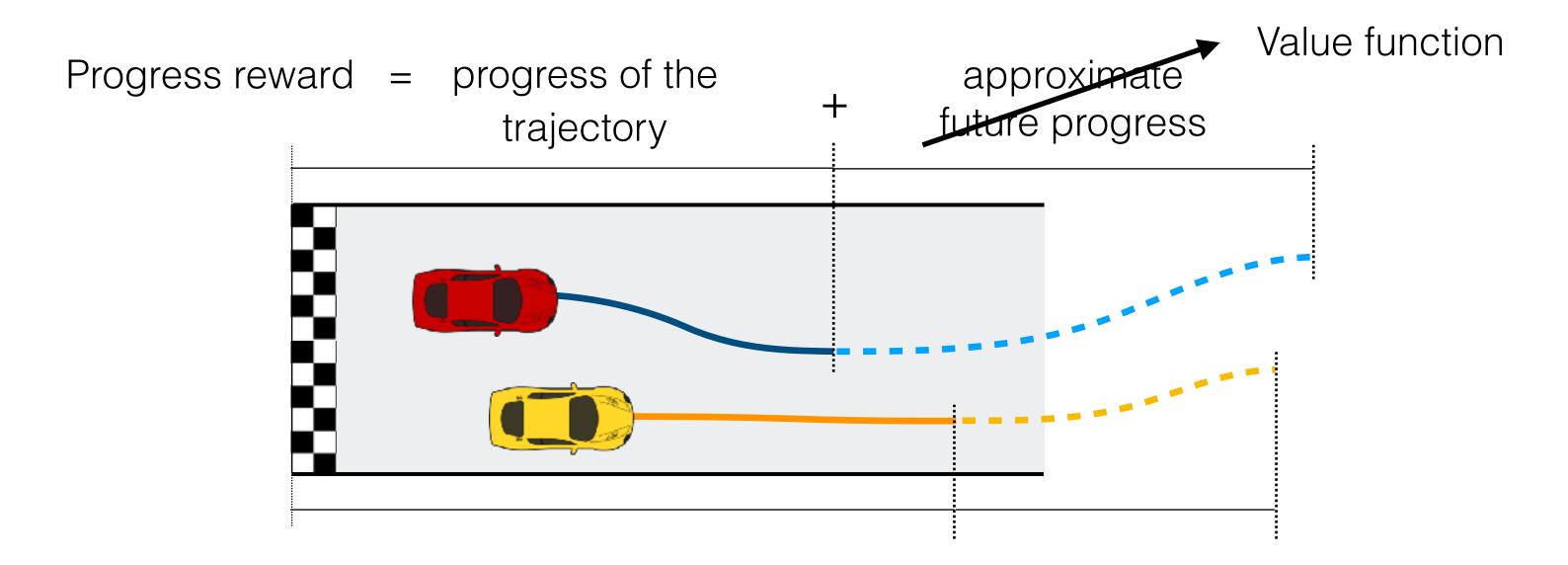
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# Experimental Results



# Experimental Results



# Summary and Outlook

- We proposed a complet autonomous racing pipeline
  - Different behavior is seen for different viability kernels and games
- Interesting insights into behavior of noncooperative decisions
  - Sequential maximization and leader-follower structure
- Reliable and real-time feasible games
  - Consider uncertainty in game formulation
  - Reinforcement learning-based terminal cost+constraints
  - High-performance implementation using GPU
- Model-learning for MPC

$$x_{k+1} = f(x_k, u_k) + \mu_{GP}(x_k, u_k)$$

▶ Learn behavior of "opponent" —> urban traffic

# Questions

