

# Reinforcement Learning to Simulate Infinitely Repeated Games

CS 136: Economics and Computation

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# What is Reinforcement Learning?

**“RL is the science of using experiences to make optimal decisions”**

Steps:

- Observing the environment
- Making decisions on how to act using some strategy
- Executing the action
- Receiving a reward or penalty based on the action
- Learning from experiences and refining the strategy
- Repeating this iterative process until an optimal strategy is found

Reinforcement Learning Cycle

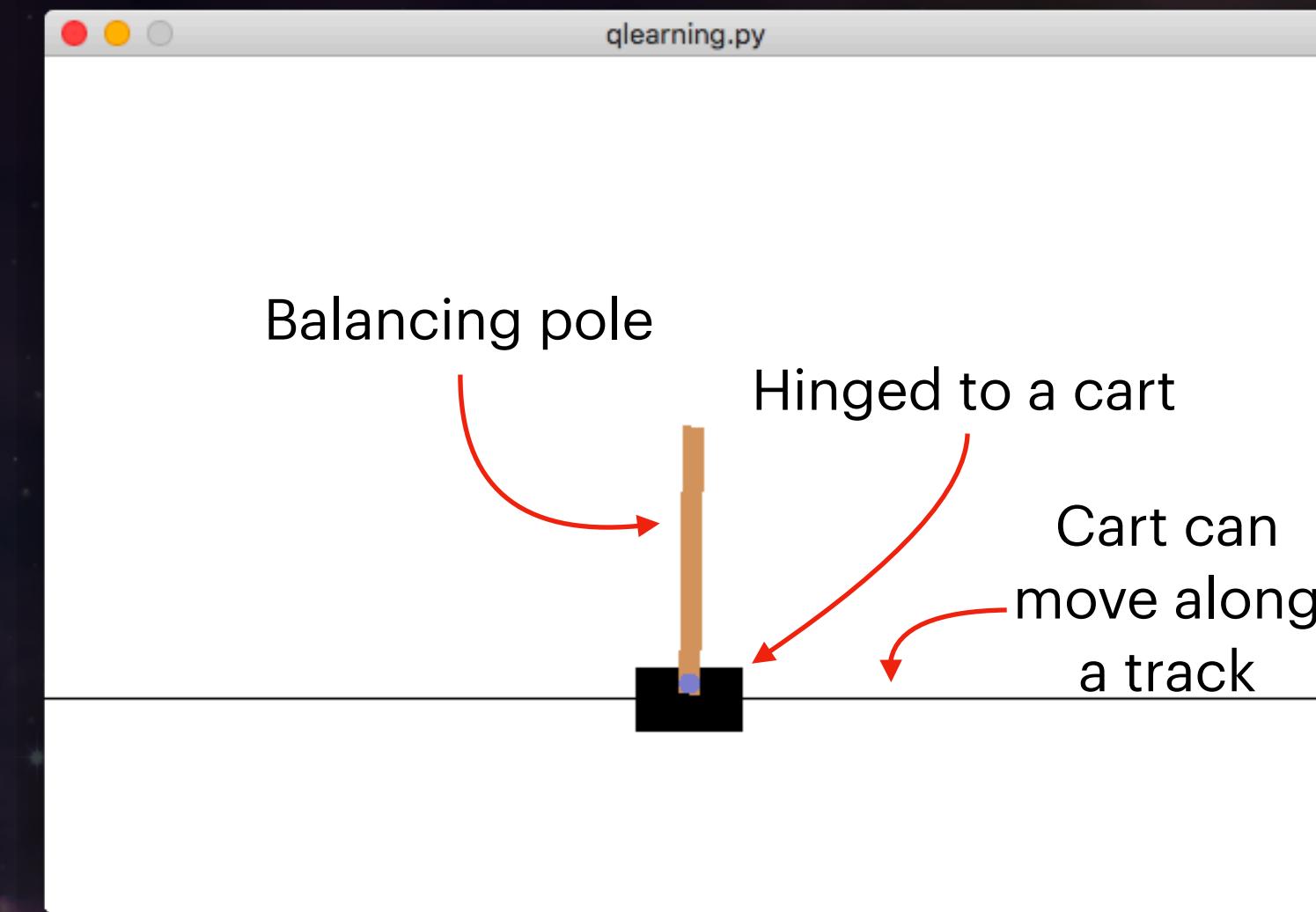


# Environment Examples

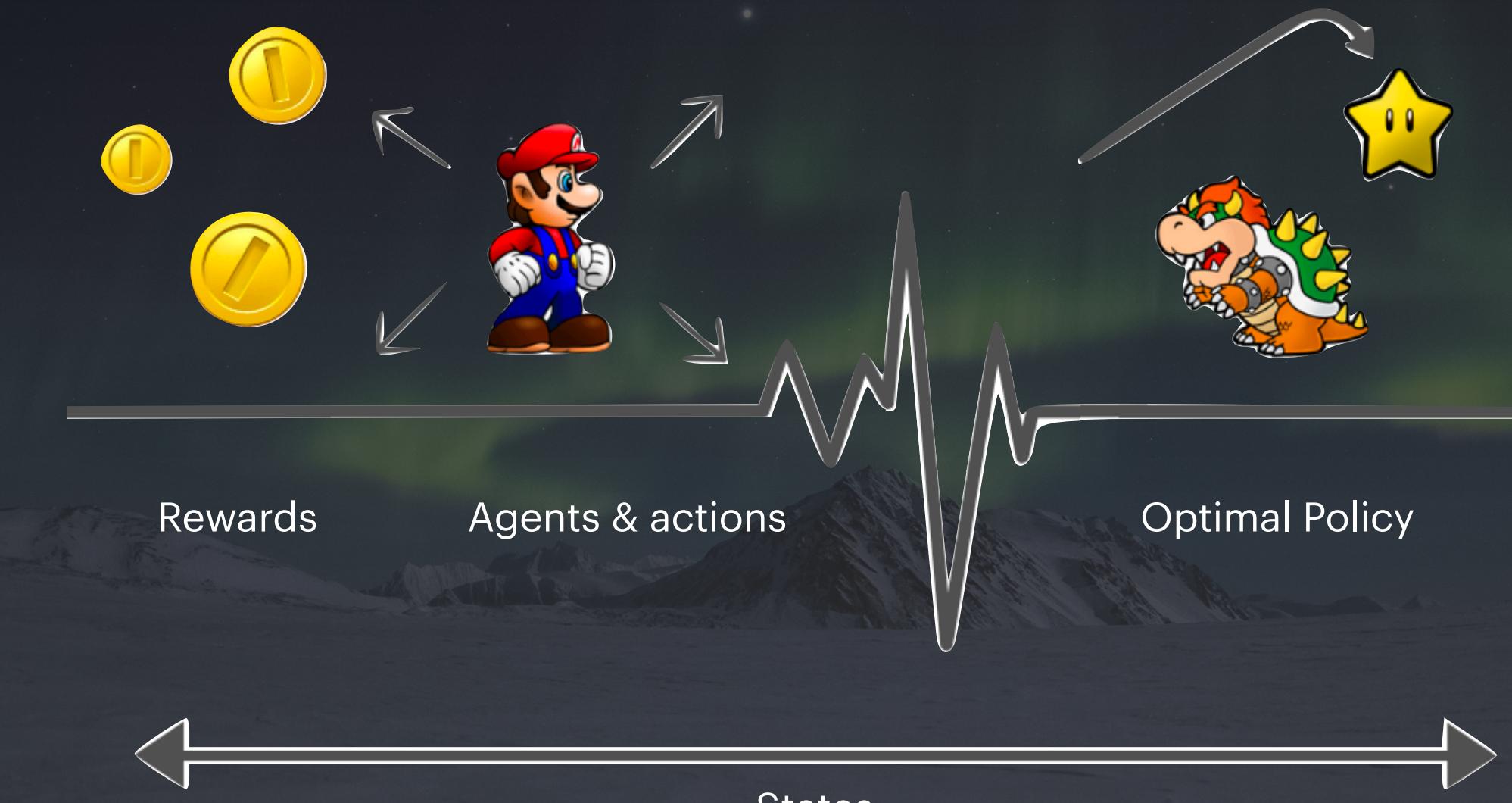
## Theory



## Example: Cart-Pole

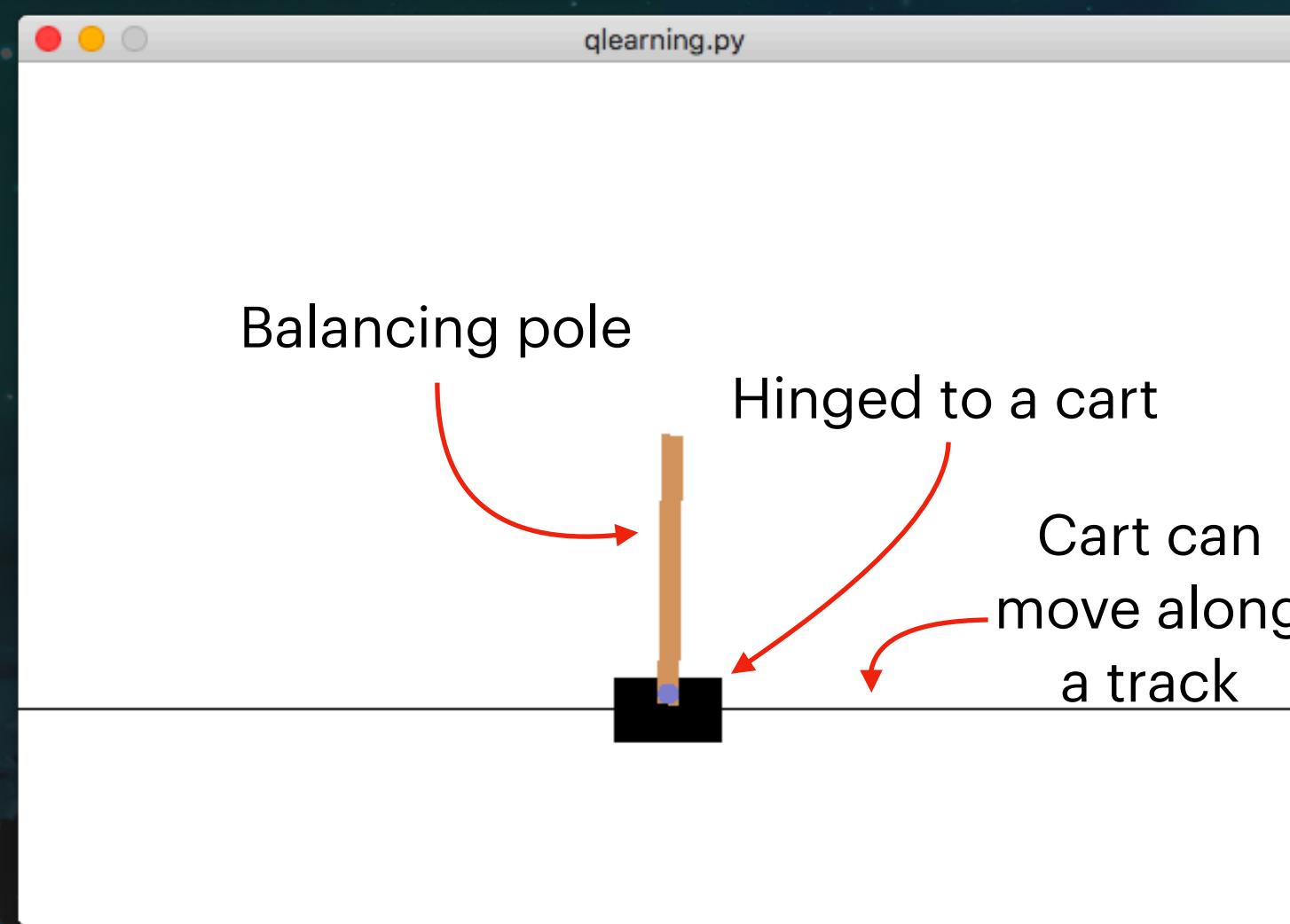


## Example: Mario Bros



# Application to Game Theory

## Cart-Pole Environment



State space:  $S = P_C \times V_C \times P_P \times V_P$

Possible actions:  $A = \{a_L, a_R\}$

Reward:  $u : S \rightarrow \mathbb{R}$

## Infinitely Repeated Game

		Player 2	
		Cooperate	Deviate
		4, 4	1, 6
Player 1	Cooperate	6, 1	2, 2
	Deviate		

State space:  $S = [(A_1 \cup \phi) \times (A_2 \cup \phi)]^m$

Possible actions:  $A_1 = \{a_{1,1}, a_{1,2}\}$

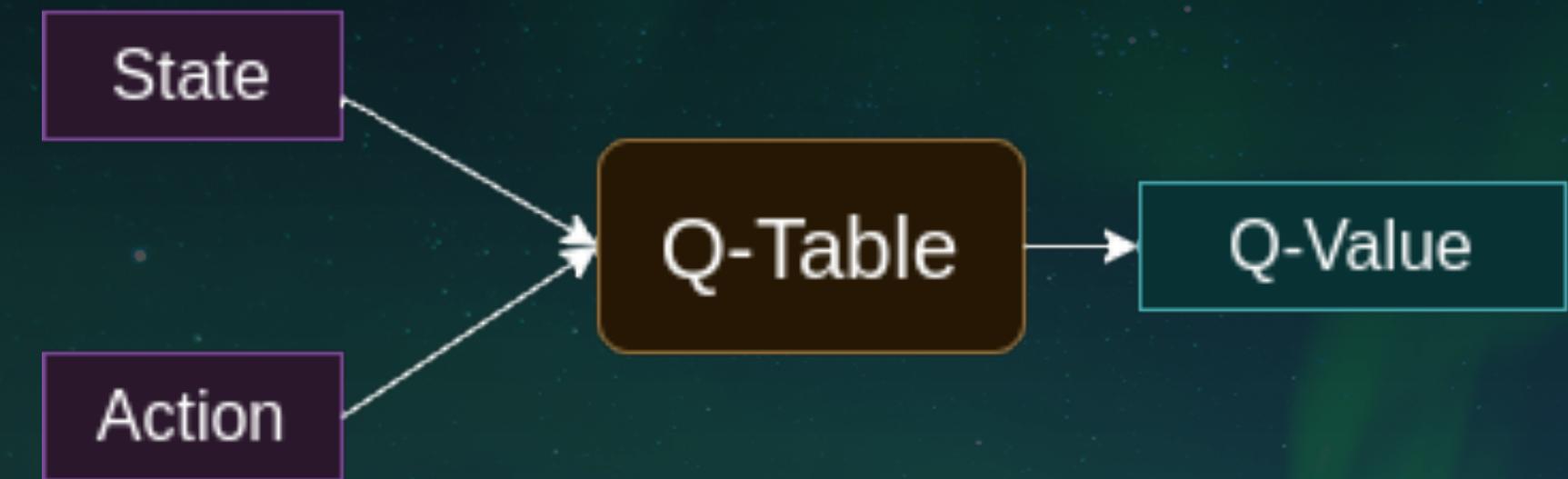
Reward:  $u_1 : S \rightarrow \mathbb{R}$

# What is Q-Learning?

And why is it a good model for repeated games?

- **Model-free:** Does not require knowledge of state transitions and reward functions.
- **Value-based:** Learns a value function that maps states and actions to expected value.
- **Off-policy:** Actions are taken using a different policy than the value function, allowing the agent to explore.

## Q-Learning

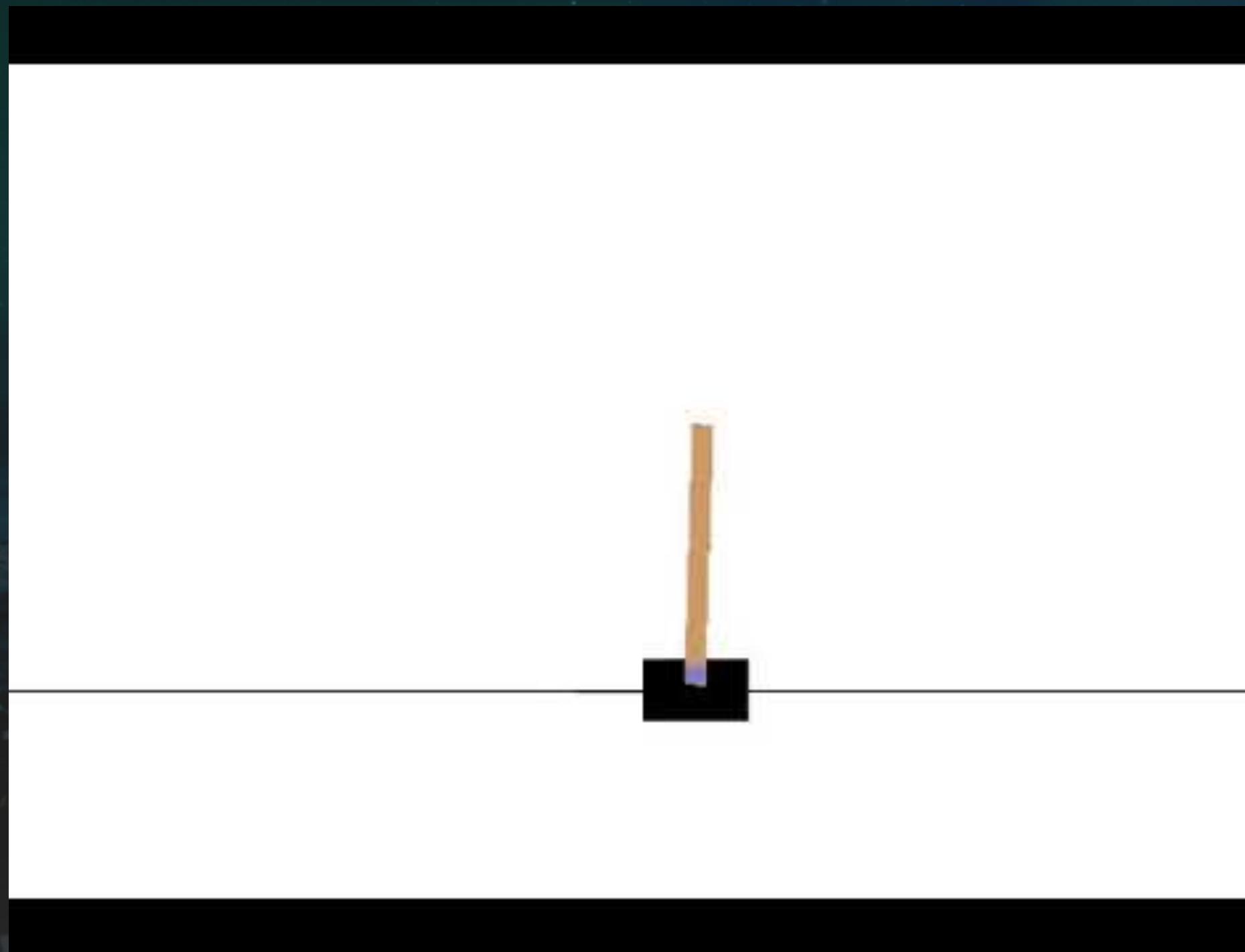


## Deep Q-Learning



# Progress Thus Far

## Understanding Q-Learning



1. Implemented Deep Q-Network (DQN) algorithm using PyTorch
2. Tested the DQN implementation on Cart-Pole environment

## Modeling Optimality in Python

```
● ○ ●  
1 U_1 = np.array([[4, 1], [6, 2]])  
2 U_2 = np.array([[4, 6], [1, 2]])  
3 game = nash.Game(U_1, U_2)  
4 print(game)  
5 # Bi matrix game with payoff matrices:  
6  
7 # Row player:  
8 # [[4 1]  
9 #  [6 2]]  
10  
11 # Column player:  
12 # [[4 6]  
13 #  [1 2]]  
14  
15 nash_equilibria = game.support_enumeration()  
16 print(list(nash_equilibria)) # [(array([0., 1.]), array([0., 1.]))]  
17
```

1. Modeled a repeated game in Python and found optimal solution
2. Created an RL environment based off of the repeated game