## COMP 138: Reinforcement Learning


observation


Instructor: Jivko Sinapov
Webpage: https://www.eecs.tufts.edu/~jsinapov/teaching/comp150_RL_Fall2020/

## $B E$ a reinforcement learner

- You, as a class, will act as the learning agent


## $B E$ a reinforcement learner

- You, as a class, will act as the learning agent
- Actions: wave, clap, or nod


## BE a reinforcement learner

- You, as a class, will act as the learning agent
- Actions: wave, clap, or nod
- Observations: color, reward


## BE a reinforcement learner

- You, as a class, will act as the learning agent
- Actions: wave, clap, or nod
- Observations: color, reward
- Goal: find an optimal policy


## BE a reinforcement learner

- You, as a class, will act as the learning agent
- Actions: wave, clap, or stand
- Observations: color, reward
- Goal: find an optimal policy
- What is a policy? What makes a policy optimal?


## How did you do it?

- What is your policy, and how is it represented?
- What does the world look like?


## What actually happened...

## What actually happened...



## Now, let's formalize this

(board or writing projector)

## About this course

- Reinforcement Learning theory \& practice
- Theory at the start and practice towards end
- Syllabus = the course web page:
https://www.eecs.tufts.edu/~jsinapov/teaching/comp150_RL/


## Where does RL fall within the field of Artificial Intelligence?

## Where does RL fall within the field of Artificial Intelligence?

- $\mathrm{Al} \rightarrow \mathrm{ML} \rightarrow \mathrm{RL}$


## Where does RL fall within the field of Artificial Intelligence?

- $\mathrm{Al} \rightarrow \mathrm{ML} \rightarrow \mathrm{RL}$
- Type of Machine Learning:
- Supervised: learn from labeled examples
- Unsupervised: learn from unlabeled examples
- Reinforcement: learn through interaction


## Reduced Formalism



## Reduced Formalism

(board or writing projector)

## Take-home Message

- Agent's perspective: only the policy is under control
- State representation and reward function are given
- Focus on policy algorithms
- Appeal: program agents by just specifying goals
- Practice: need to pick state representation and reward function


## Example Applications

## Robot Motor Skill <br> Coordination with EM-based Reinforcement Learning

Petar Kormushev, Sylvain Calinon, and Darwin G. Caldwell

Italian Institute of Technology

## Example Applications



## Reading Assignment

- Chapter 1 and 2 of Sutton and Barto
- Reading response on Canvas due 9/11 before class starts


## Programming Assignments

- Students are required to complete 4 minor programming assignments of their choosing
- Default options: programing exercises from Sutton and Barto (let's look at some examples)


## Discussion Moderation

- Each student will lead a reading discussion once during the semester
- Students can team up in a pair
- Sign up sheet will be posted to Canvas tonight
- Extra credit for anyone who volunteers for slots in the next week
- Presentation materials / notes or description of what will be discussed should be emailed to me 48 hours before the class

Next time...

## COMP 150: Reinforcement Learning


observation

## Domains and Applications



SCORE: 104

## Curriculum Learning



Example QuickChess game variants

## The Curriculum Learning Problem

Task Creation


Sequencing

[ Narverkar et al 2016 ]

## Textbook




The authors have made the book available: http://incompleteideas.net/book/bookdraft2017nov5.pdf

## Course Organization

- Taught as a seminar: students take turns presenting the readings
- Will cover both theory and practice
- Final projects - you will complete a project in which you ask (and then answer) a relevant RL research question

