RLearning:

Short guides to reinforcement learning

Introduction

Davud Rostam-Afschar (Uni Mannheim)

- ► Traditional computer science
 - ► Program computer for every task

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- ► New paradigm
 - Provide examples to machine
 - ► Machine learns to accomplish a task based on the examples

- ► Traditional computer science
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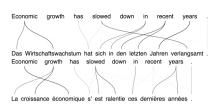
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Supervised Learning

- ► Most common type of machine learning
- ► Learns from labeled examples (input + correct output)

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- Other Approaches
 - Unsupervised Learning:
 Finds patterns without labeled data
 - Semi-supervised Learning: Mixes labeled and unlabeled data
 - ► Reinforcement Learning: Learns by trial and error

Animal Psychology

- Positive reinforcements:
 - ► Pleasure and food
- ► Negative reinforcements:
 - Pain and hunger
- ► Reinforcements used to train animals



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► Let's do the same with computers!



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 - Optimal control
 - Approximate dynamic programming
 - ► Neuro-dynamic programming

Definition

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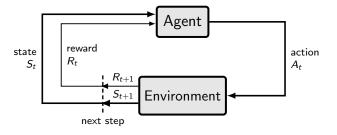
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Reinforcement Learning Problem



Goal: Learn to choose actions that maximize rewards

Applications and Examples

RL Examples

- ► Game playing (go, atari, backgammon)
- Elevator scheduling
- ► Helicopter control
- ► Spoken dialog systems
- ► Data center energy optimization
- ► Self-managing network systems
- Autonomous vehicles

- Operations research (pricing, vehicle routing)
- Computational finance (portfolio optimization, algorithmic trading)

Operations Research

► Example: vehicle routing

► **Agent:** vehicle routing software

► Environment: stochastic demand

► **State:** vehicle location, capacity and depot requests

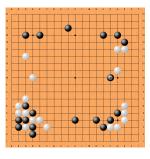
► Action: vehicle route

► Reward: - travel costs



Game Playing

- ► Example: Go (one of the oldest and hardest board games)
- Agent: player
- **Environment:** opponent
- ► State: board configuration
- ► Action: next stone location
- **Reward:** +1 win / -1 loose
 - ▶ 2016: AlphaGo defeats top player Lee Sedol (4-1)
 - ► Game 2 move 37: AlphaGo plays unexpected move (odds 1/10,000) https://www.youtube.com/watch?v=WXuK6gekU1Y



Robotic Control

- Example: helicopter control
- Agent: controller
- **Environment:** helicopter
- State: position, orientation, velocity and angular velocity
- Action: collective pitch, cyclic pitch, tail rotor control
- Reward: deviation from desired trajectory
 - 2008 (Andrew Ng): automated helicopter wins acrobatic competition against humans https://www.youtube.com/watch?v=0JL04JJjocc



Conversational Agent

- Example: Conversational Agent (ChatGPT)
- ► Agent: language model
- **▶ Environment:** user
- State: conversation history
- ► Action: next token
- Reward: ratings based on task completion, user satisfaction, etc.
- ► Today: active area of research



Computational Finance

Example: Automated trading

► **Agent:** trading software

Environment: other traders

State: price history

► **Action**: buy/sell/hold

► **Reward:** amount of profit



Example: trading strategies that adapt to real-time market signals

Adaptive lab, field, or survey experiments

Advertising

- Advertising
- Educational learning

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Explaining Behavior and Assisting Decision Making

Strategic decision making (game theory)

Adaptive lab, field, or survey experiments

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Explaining Behavior and Assisting Decision Making

- Strategic decision making (game theory)
- ► Households choices (fertility, labor, education, consumption/saving)

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Explaining Behavior and Assisting Decision Making

- Strategic decision making (game theory)
- Households choices (fertility, labor, education, consumption/saving)
- Firms choices (entry, exit, investments, hiring, pricing, output)

Course Overview

Course overview

- 1. Unit 1: Multi-Armed Bandits
- 2. Unit 2: Markov Decision Processes Assignment 1
- 3. Unit 3: RL Algorithms *Assignment 2*
- 4. Unit 4: Deep RL

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Takeaways

- Comprehensive, but challenging form of machine learning
 - Interdependent sequence of decisions
 - Incomplete model
 - Stochastic environment
 - No supervision
 - Partial and delayed feedback
- ► Long term goal: autonomous agents—without needing explicit supervision (according to ChatGPT)