

*"Where did you go to, if I may ask?"* said Thorin to Gandalf as they rode along.

*"To look ahead,"* said he.

*"And what brought you back in the nick of time?"*

*"Looking behind,"* said he.

J.R.R. Tolkien, *The Hobbit*

A painting of Gandalf the White from J.R.R. Tolkien's The Hobbit. He is standing in a field of tall green grass, looking back over his shoulder. He wears his characteristic tall, pointed hat and long white beard. He holds a wooden staff with a glowing tip. The background is a soft, hazy landscape with a large tree on the right.

# **CMPUT 365**

## **Introduction to RL**

# Reminder

You **should be enrolled in the private session** we created in Coursera for CMPUT 365.

I **cannot** use marks from the public repository for your course marks.

You **need to check, every time**, if you are in the private session and if you are submitting quizzes and assignments to the private section.

There were **20 pending invitations** last time I checked!

If you have any questions or concerns, **talk with the TAs** or email us `cmput365@ualberta.ca`.

# Reminders and Notes

- On the midterm:
  - It is marked, there are only a few left. You should have the marks by Wednesday.
- What **I** plan to do today:
  - Where are we?
  - Overview of Monte Carlo Methods for Prediction & Control (Chapter 5 of the textbook).
- What I recommend **YOU** to do for next class:
  - Read Chapter 5 up to Section 5.5.
  - Graded Quiz (Off-policy Monte Carlo).
  - Programming Assignment is not graded this week.

**Please, interrupt me at any time!**



# Interlude

# An overview

- Main features of a reinforcement learning problem:
  - Trial-and-error learning
  - Exploration
  - Delayed credit assignment

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- Main features of a reinforcement learning problem:
  - Trial-and-error learning
  - Exploration **A flavour of RL: Bandits (Chapter 2)**
  - Delayed credit assignment

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- Main features of a reinforcement learning problem:

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But what does that mean?

What is this sequential decision-making problem we are trying to solve?

What does solution mean here?

**A problem formulation: MDPs (Chapter 3)**



# An overview

- Main features of a reinforcement learning problem:
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  - Exploration
  - Delayed credit assignment
- What about the solution?

**A first solution: Dynamic Programming (Chapter 4)**

# An overview

- Main features of a reinforcement learning problem:
  - Trial-and-error learning
  - Exploration
  - Delayed credit assignment
- What about the solution?
  - Dynamic programming! ← We need to know  $p(s', r | s, a)$  and it can be computationally expensive to solve the system of linear equations.

**Our first learning algorithm: Monte Carlo Methods (Chapter 5)**

# Chapter 5

# Monte Carlo Methods

# Monte Carlo Methods – Why?

- This is our **first learning** method.
- We do not assume complete knowledge of the environment.
- “Monte Carlo methods **require only experience** — sample sequences of states, actions, and rewards from actual or simulated interaction with an environment.” 🤖
- It works! And different variations are used everywhere in the field (n-step returns, TD( $\lambda$ ), MCTS–AlphaGo/AlphaZero–, etc).
- ... but we still need a model, albeit only a sample model.

*MC Methods are ways of solving the RL problem based on avg. sample returns (similar to bandits, but instead of rewards we are sampling returns).*

# Monte Carlo Prediction

## First-visit MC prediction, for estimating $V \approx v_\pi$

Input: a policy  $\pi$  to be evaluated

Initialize:

$V(s) \in \mathbb{R}$ , arbitrarily, for all  $s \in \mathcal{S}$

$Returns(s) \leftarrow$  an empty list, for all  $s \in \mathcal{S}$

Loop forever (for each episode):

Generate an episode following  $\pi$ :  $S_0, A_0, R_1, S_1, A_1, R_2, \dots, S_{T-1}, A_{T-1}, R_T$

$G \leftarrow 0$

Loop for each step of episode,  $t = T-1, T-2, \dots, 0$ :

$G \leftarrow \gamma G + R_{t+1}$

Unless  $S_t$  appears in  $S_0, S_1, \dots, S_{t-1}$ :

Append  $G$  to  $Returns(S_t)$

$V(S_t) \leftarrow \text{average}(Returns(S_t))$

