# Reinforcement Learning Fundamentals

Lecture 3: RL Framework

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Some material in this lecture is taken from



2. Dr Silver's course: "Reinforcement Learning."

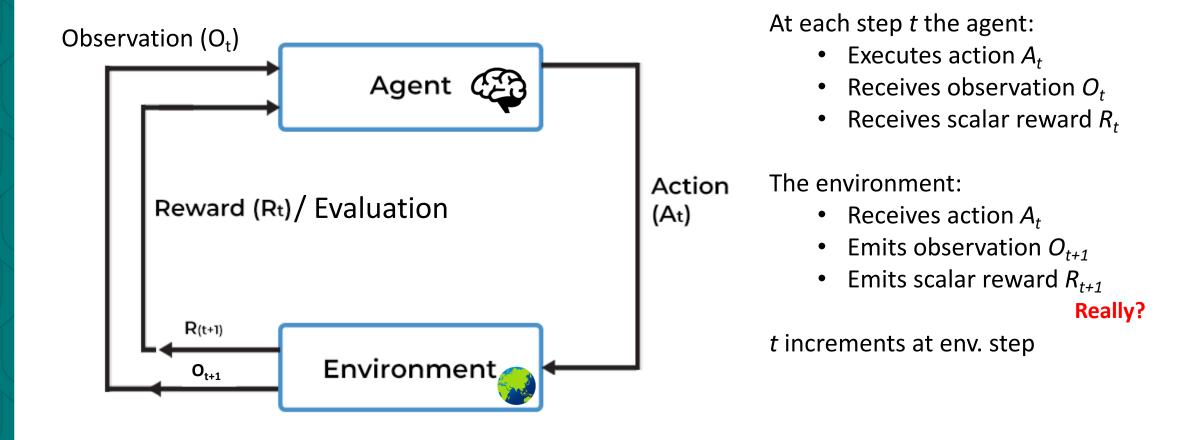


# In today's class...

- RL Framework
- What are Rewards?
- What is a State?
- Special cases: Fully and Partially Observable environments
- Temporal Difference

# RL Framework Sequential Decision Making

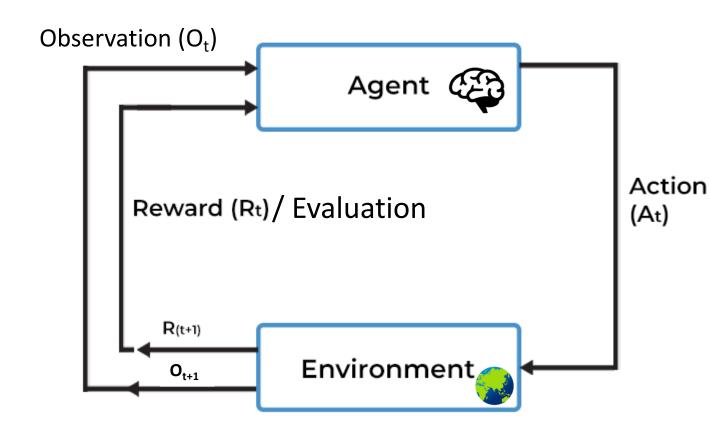
• Goal: Select actions or a sequence of actions to maximize the total future reward.



What is evaluation in supervised learning? How is it different in RL?

# **RL Framework**

#### **Sequential Decision Making**



- Agent learns by interacting with the environment.
- Environment returns some rewards.

  Really?
- Environment gives noisy delayed scalar evaluation.
- Environment also provides some observations.
- Environment can be stochastic.
- Goal is to maximize a measure of longterm performance. In this case it can be the reward.

## What is State?

At time t=1, the environment / the world returned some observation  $O_1$  and reward  $O_2$  and the agent took an action  $O_1$ . Then the world returned  $O_2$  and  $O_3$ . Now the agent took the action  $O_3$ .

What is history at *time* = *t*?

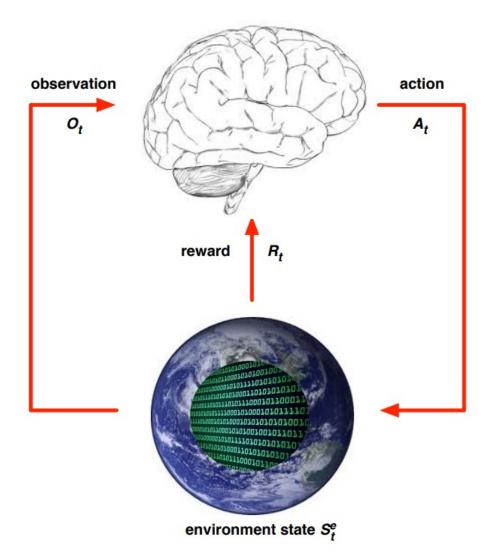
- The sequence of observations, actions, rewards from time = **1** to time = **t** is referred to as history.
- i.e., History is all the observable variables up to time *t*.
- What happens next depends on the history.

**State** is the information used to determine what happens next. This is the **summary of the history**.

Formally, state if a function of history:

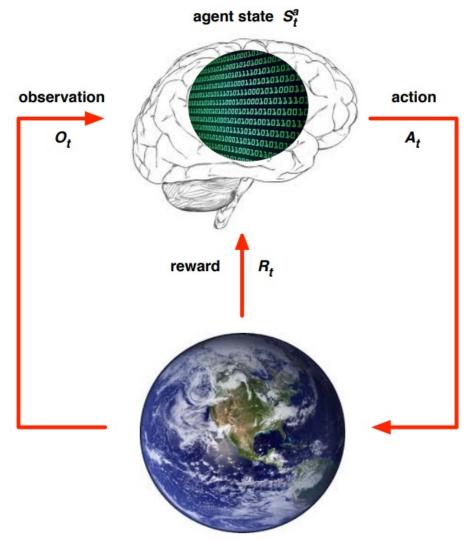
$$S_t = f(H_t)$$

#### **State:** Environment State



- The environment state  $S_t^e$  is the environment's private representation
- i.e. whatever data the environment uses to pick the next observation/reward
- The environment state is not usually visible to the agent
- Even if  $S_t^e$  is visible, it may contain irrelevant information

# **State:** Agent State



- The agent state  $S_t^a$  is the agent's internal representation
- i.e. whatever information the agent uses to pick the next action
- i.e. it is the information used by reinforcement learning algorithms
- It can be any function of history:

$$S_t^a = f(H_t)$$

### **State:** Information State

An information state (a.k.a. Markov state) contains all useful information from the history.

What is Markov Property?

How does it map to Helicopter Example?

#### **State:** Information State

An information state (a.k.a. Markov state) contains all useful information from the history.

#### Definition

A state  $S_t$  is Markov if and only if

$$\mathbb{P}[S_{t+1} \mid S_t] = \mathbb{P}[S_{t+1} \mid S_1, ..., S_t]$$

"The future is independent of the past given the present"

$$H_{1:t} \rightarrow S_t \rightarrow H_{t+1:\infty}$$

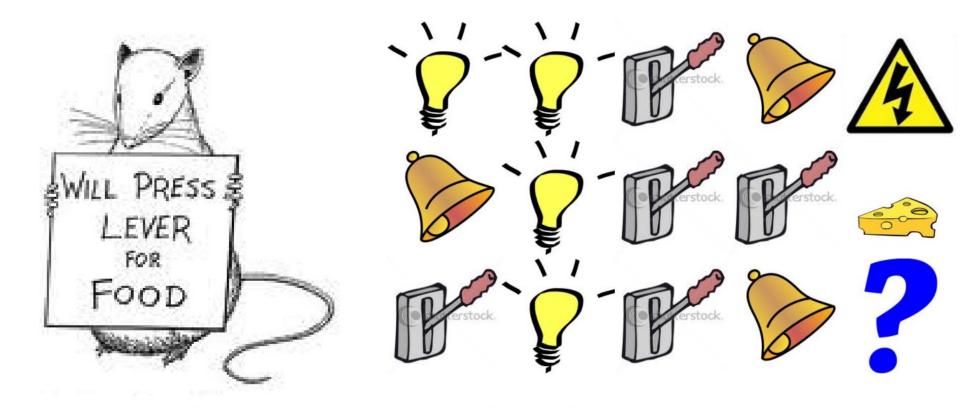
- Once the state is known, the history may be thrown away
- i.e. The state is a sufficient statistic of the future
- The environment state  $S_t^e$  is Markov
- The history  $H_t$  is Markov

How does it map to Helicopter Example?

# How to pick a state?



# How to pick a state?



- What if agent state = last 3 items in sequence?
- What if agent state = counts for lights, bells and levers?
- What if agent state = complete sequence?