

# AI1811-RL.typ

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Introduction to Reinforcement Learning, AI1811.

## 目录

1. Introduction .....	1
1.1. Simple Components .....	1
1.1.1. History .....	1
1.1.2. State .....	1
1.1.3. Policy .....	2
1.1.4. Reward .....	2
1.2. Categories .....	2
1.2.1. Value Based RL .....	2
1.2.2. Policy Based RL .....	2
1.2.3. Advanced RL .....	3
1.3. Markov Process .....	3
2. Conclusion .....	3

## §1. Introduction

The agent is interacting with the environment:

- Getting Observation:  $O_t$
- Getting Reward  $R_t$
- Doing Actions:  $A_t$

### §1.1. Simple Components

#### §1.1.1. History

A sequence of **observation, actions and rewards**.

$$H_t = O_1, R_1, A_1, O_2, R_2, A_2, \dots, O_{t-1}, R_{t-1}, A_{t-1}, O_t, R_t$$

Based on the History:

- agents will operate actions  $A_t$ .
- environments will choose the observations and reward (based on  $A_t$  and  $H_t$ )

#### §1.1.2. State

State is a function (or mapping based on the history), due to the Markov Process.

$$S_t = f(H_t)$$

### Recordings.

- Observation is the sample of the environment and history.

### §1.1.3. Policy

Policy is the mapping from State Space to Action Space. (For the current state  $s$ , we will do operations  $a$ )

- Deterministic Policy:  $a = \pi(s)$
- Stochastic Policy:  $\pi(a|s) = P(A_t = a|S_t = s)$

### §1.1.4. Reward

Based on the state and the actions. (From the current state, we do specific actions, then we will get reward  $R(s, a)$ )

Based on the reward functions, we can define the value functions (in the long term):

$$\begin{aligned} Q_\pi(s, a) &= \mathbb{E}_\pi[R_{t+1} + \gamma R_{t+2} + \dots | S_t = s, A_t = a] \\ &= \mathbb{E}_\pi[R_{t+1} + \gamma Q_\pi(s', a') | S_t = s, A_t = a] \end{aligned}$$

$s'$  is the next state for  $t + 1$ .

#### Recordings.

在非确定性环境中,  $s_{t+1}$  是不确定的, 哪怕  $s_t, a_t$  在给定的确定下, 因此, 上式的贝尔曼方程需要对  $s'$  取期望。

Based on the model, we can predict:

- Predict the next state:

$$P_{ss'}^a = \mathbb{P}[S_{t+1} = s' | S_t = s, A_t = a]$$

- Predict the next step:

$$R_s^a = \mathbb{E}[R_{t+1} | S_t = s, A_t = a]$$

## §1.2. Categories

- Value Based RL
  - 知道什么是好的 什么是坏的
  - 基于价值设计价值函数
- Policy Based
  - We have the policy but no value functions.
- Actor-Critic

### §1.2.1. Value Based RL

The training goal is maximize the value function:

$$Q(s, a) = \mathbb{E}[R_{t+1} | s, a] + \gamma \max_{a'} Q(s', a')$$

Then, we do greedy search:

$$\pi(a|s) = \arg \max_a Q(s, a)$$

### §1.2.2. Policy Based RL

Optimize  $\pi_{\theta(a|s)}$  directly!

- REINFORCE
- Actor-Critic (A2C, A3C)
- PPO
- TRPO
- SAC (soft actor-critic)

### §1.2.3. Advanced RL

- Deep Reinforcement RL: Use Deep Neural Network to do function approximations.
- Model-Based RL: learning with trajectory data in virtual environment.
- Goal-oriented RL (Long Horizon, Intermediate Steps and Simple Task Splitting)
- Imitation Learning
- Multi-Agent RL
- RL for Large Language Model Fine-tuning.

## §1.3. Markov Process

**Definition 1.3.1** Random State.

$$\mathbb{P}[S_{t+1}|S_1, S_2, \dots, S_t]$$

## §2. Conclusion