Reinforcement Learning

Concepts and Web Applications

Overview

Brief look into Reinforcement Learning (RL)

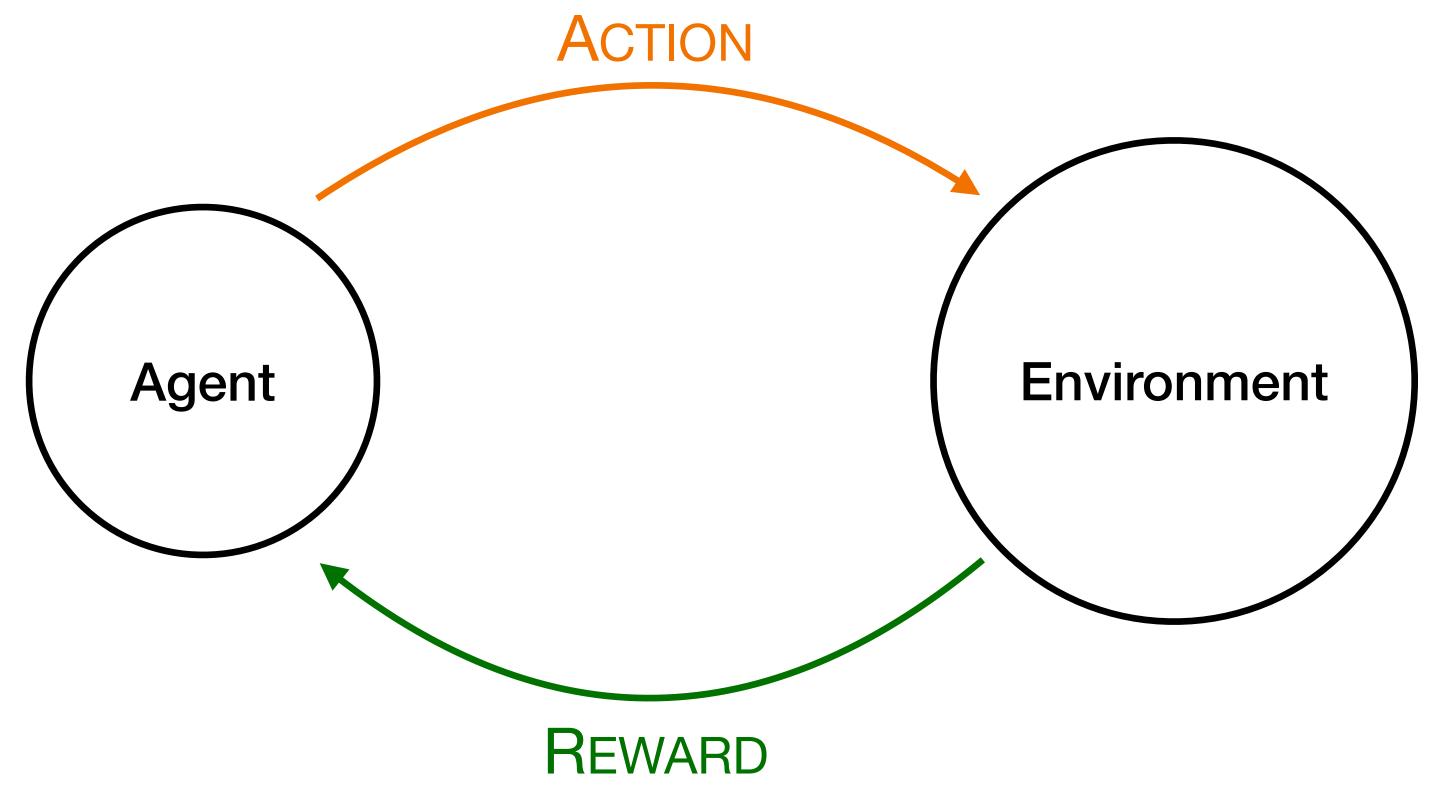
- 1. What is RL?
- 2. How is RL applied to the Web?
- 3. Code Project
- 4. Why did I choose this topic?

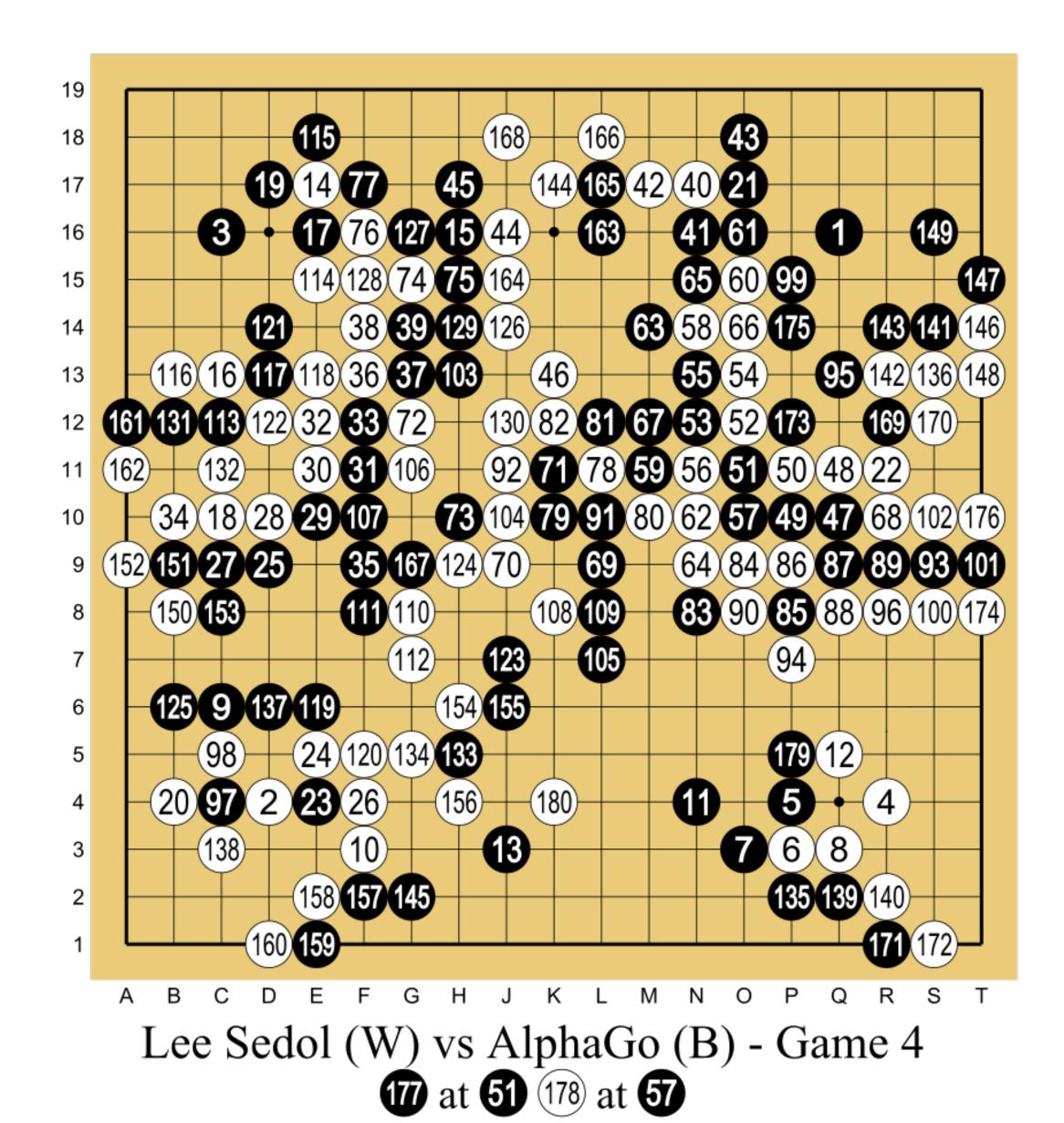
What is RL?

RL & ML Unsupervised Supervised Learning Learning To Cluster To EVALUATE Reinforcement Learning To Act

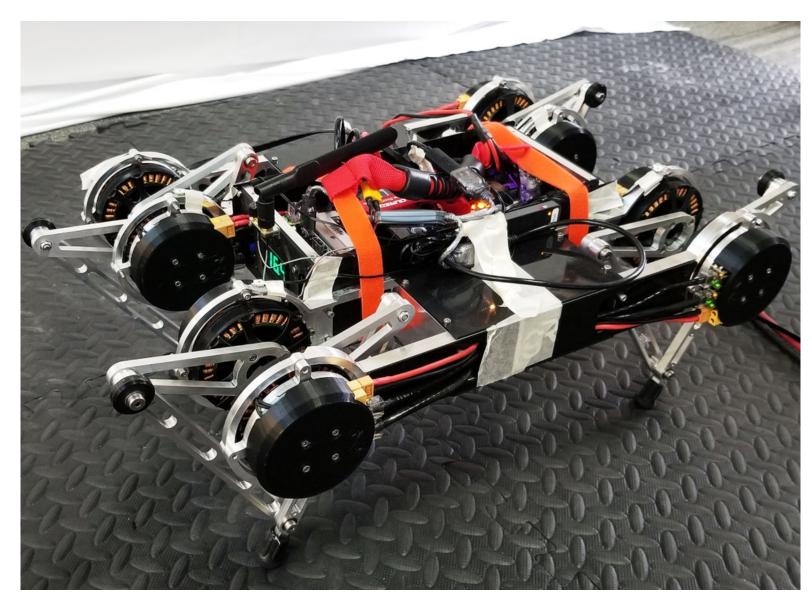
What is RL?

Agent & Environment









How is RL applied to the Web?

Trial and error

- → Web Spiders (~2000)
- → Web Recommendation (~2016)
- ◆ Internet Cogestion Control (~2019)

Code Project - Frozen Lake

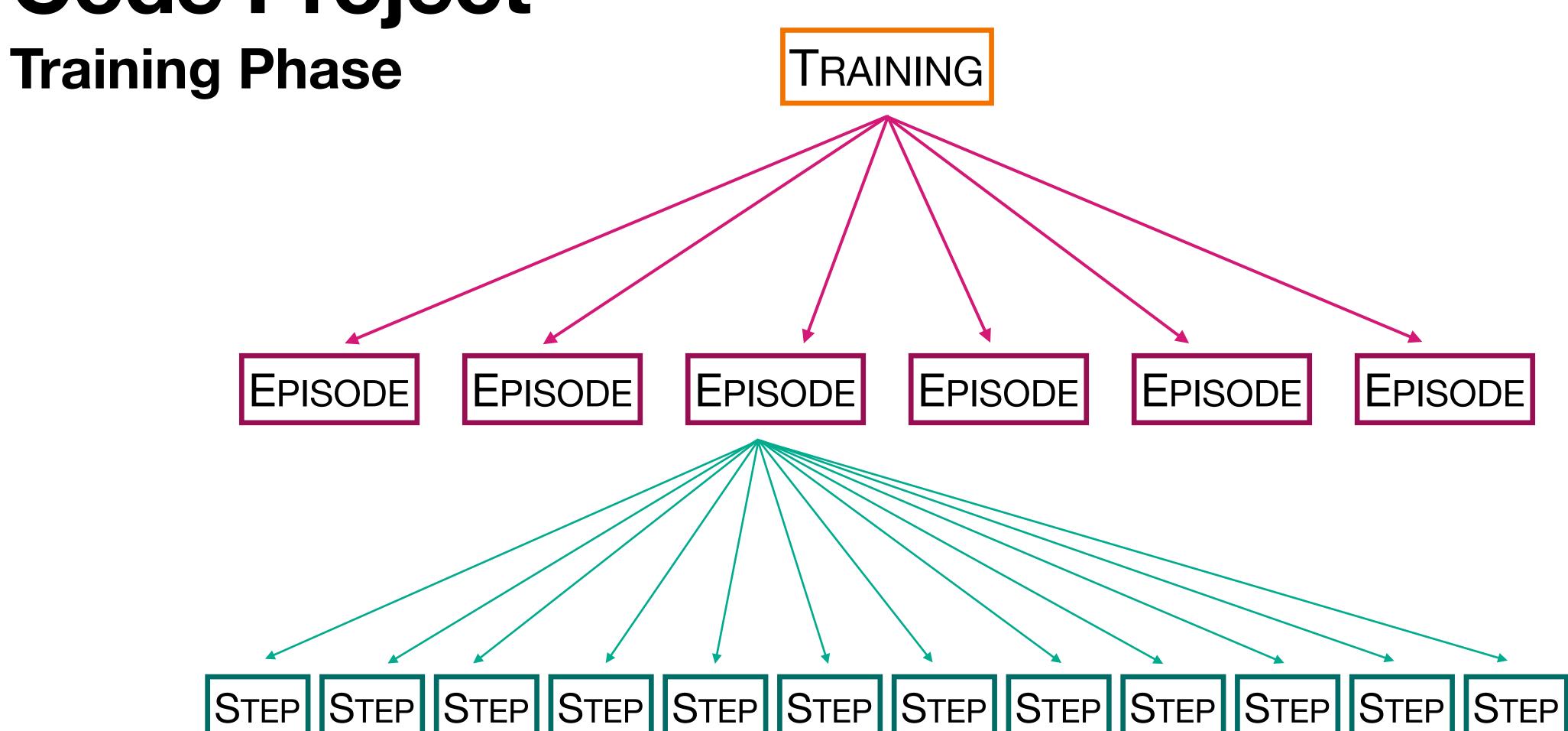
Code Project

State Value & Policy State VALUE A S $q(s,a) \ge q(s,a')$ State VALUE B S' Policy(state, action): q(s, a)**Environment** State State VALUE D State VALUE E

$$q_*(s, a) = E\left[(1-\alpha)q_*(s, a) + \alpha \left(\frac{R_{t+1}}{a} + \gamma \max_{a'} q_*(s', a') \right) \right]$$

Bellmann's Equation

Code Project



LOGIC

```
// Helper
function modulo(n, m){
    return ((n % m) + m) % m;
// Environment
function Environment() { ---
};
Environment.prototype.generate_random_map = function(p_max) { =
};
Environment.prototype.reset = function(use_random, make=true) { ==
};
Environment prototype step = function(action) = \{action\}
};
Environment prototype move = function(direction)  { \cdots
Environment.prototype.render = function(type, object) { ==
};
```

```
Environment.prototype.step = function(action) {
    let new_state = 0;
    let reward = 0;
    let done = false;
    let info = "";
    const action_index = this.action_space.indexOf(action);
    if (Math.random() > 1 - this.slip_p) { // Slip
        if (Math.random() > .5) {
            new_state = this.move(this.action_space[modulo(action_index - 1, 4)])
        } else {
            new_state = this.move(this.action_space[modulo(action_index + 1, 4)])
       };
    } else { // Don't slip
        new_state = this.move(action);
    };
    reward = this.reward_mapping[this.map[new_state]][0];
    if (this.reward_mapping[this.map[new_state]][1]) {
        done = true;
    return {'new_state':new_state, 'reward':reward, 'done':done, 'info':info};
};
                                          13
```

```
// Environment create in main.js
// helper modulo already create
// Agent
function Agent() { ---
Agent.prototype.train = function(text_renderer, use_random_generator,
    use_clever_state) { ---
};
Agent.prototype.exploit = function(use_clever_state) { ---
};
Agent.prototype.get_clever_state = function() { ---
};
Agent.prototype.get_average_rewards = function(rewards_all_episodes, count) { ---
};
Agent.prototype.get_render_average_rewards = function(rewards_all_episodes,
    count) { ···
};
Agent.prototype.get_render_q_table = function() { ---
};
                                         14
```

```
Agent.prototype.train = function(text_renderer, use_random_generator, use_clever_state) {
    this.state_space_size = env.height * env.width;
    this.q_table = [];
    if (use_clever_state) { ...
       for (let i = 0; i < this.state_space_size; i++) {</pre>
            this.q_table.push([]);
            for (let j = 0; j < this.action_space_size; j++) {</pre>
               this.q_table[i].push(0);
       };
    };
    let last_remake = 0;
    if (use_clever_state) {
        this.num_episodes = this.num_episodes_object.clever;
        this.num_episodes = this.num_episodes_object.basic;
   const rewards_all_episodes = [];
    let exploration_rate = this.max_exploration_rate;
    for (let episode = 0; episode < this.num_episodes; episode++) {</pre>
        let state = env.reset(use_random_generator, false);
        if (use_clever_state) { ...
        };
        let done = false;
        let rewards_current_episode = .0;
        let step = 0;
        while (step++ < this.max_steps_per_episode && !done) {</pre>
            let action_index;
            // Exploration vs. Exploitation
            if (Math.random() > exploration_rate) { // Exploitation
                action_index = this.exploit(use_clever_state);
            else { // Exploration, Move at Random
               action_index = Math.floor(Math.random() * this.action_space_size);
            action = env.action_space[action_index];
            let step_return = env.step(action);
            let new_state = step_return.new_state;
            if (use_clever_state) {
                new_state = this.get_clever_state();
            this.q_table[state][action_index] = this.q_table[state][action_index] * (1 - this.learning_rate) + this.learning_rate * (step_return.reward + this.discount_rate * Math.max(...this.q_table[new_state]));
            state = new_state;
            done = step_return.done;
            // Only keep reward due to success (human reading facilitator)
            if (done && step_return.reward > 0) {
                rewards_current_episode += step_return.reward;
        };
        exploration_rate = this.min_exploration_rate + (this.max_exploration_rate - this.min_exploration_rate) * Math.exp(- this.exploration_decay_rate * episode);
        rewards_all_episodes.push(rewards_current_episode);
    };
                                                                                                           15
    return rewards_all_episodes;
```

```
// Update Q-table Q(s,a)
this.q_table[state][action_index] = this.q_table[state][action_index] *
    (1 - this.learning_rate) + this.learning_rate * (step_return.reward
    + this.discount_rate * Math.max(...this.q_table[new_state]));
```

$$q_*(s,a) = E\left[(1-\alpha)q_*(s,a) + \alpha\left(\frac{R_{t+1}}{\alpha} + \gamma \max_{a'} q_*(s',a')\right)\right]$$

```
Agent.prototype.exploit = function(use_clever_state) {
    let action_index = 0;
    let state = 0;
    if (use_clever_state) {
        state = this.get_clever_state();
        action_index = this.q_table[state].indexOf(Math.max(...this.q_table[state]))
    else {
        state = env.state;
        action_index = this.q_table[state].indexOf(Math.max(...this.q_table[state]))
    };
    return action_index;
};
Agent.prototype.get_clever_state = function() { \cdots
};
Agent.prototype.get_average_rewards = function(rewards_all_episodes, count) { ---
Agent.prototype.get_render_average_rewards = function(rewards_all_episodes, count) {
};
                                           17
```

(Hyper) Parameters

```
Agent
function Agent() {
   // Parameters
   this num_episodes_object = {basic: 1e4, clever: 1e5};
    this.max_steps_per_episode = 100;
   this.learning_rate = 0.1;
    this.discount_rate = 0.99;
    this.max_exploration_rate = 1;
   this.min_exploration_rate = 0.01;
    this.exploration_decay_rate = 0.001;
    // Random
   this.random_remake = 1e3;
    // Q-table
   this.action_space_size = env.action_space.length;
    this.state_space_size = 0;
    this.num_episodes = this.num_episodes_object.basic;
    this.q_table = [];
                                      19
```

```
// Environment
function Environment() {
    this.name = "Frozen Lake";
   this.action_space = ['up', 'right', 'bottom', 'left'];
    this.slip_p = .66;
    this width = 4;
    this.height = 4;
   this.base_map = "
                        SFFF\
                        FHFH\
                        FFFH\
                        HFFG";
   this.map = '';
    this.reward_mapping = { 'S' : [0.001, false],
                            'F' : [0.001, false],
                            'H' : [-.5, true] ,
                            'G' : [1, true]};
    this.state = 0;
   this.reset();
};
                                        20
```

Action ...!

Code Project Sumary

Preparation

Training

Policy Approximation

Continuous Complete

Discrete

Q-Table

Deep Neural Network

ACS PPO

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Why did I choose this topic?