# Competitive Game of Life

Comp390 Demonstration

### Overview

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## Introduction of the Project

Conway's Game of Life represents the birth, survival and death of individual cells on a 2D surface.

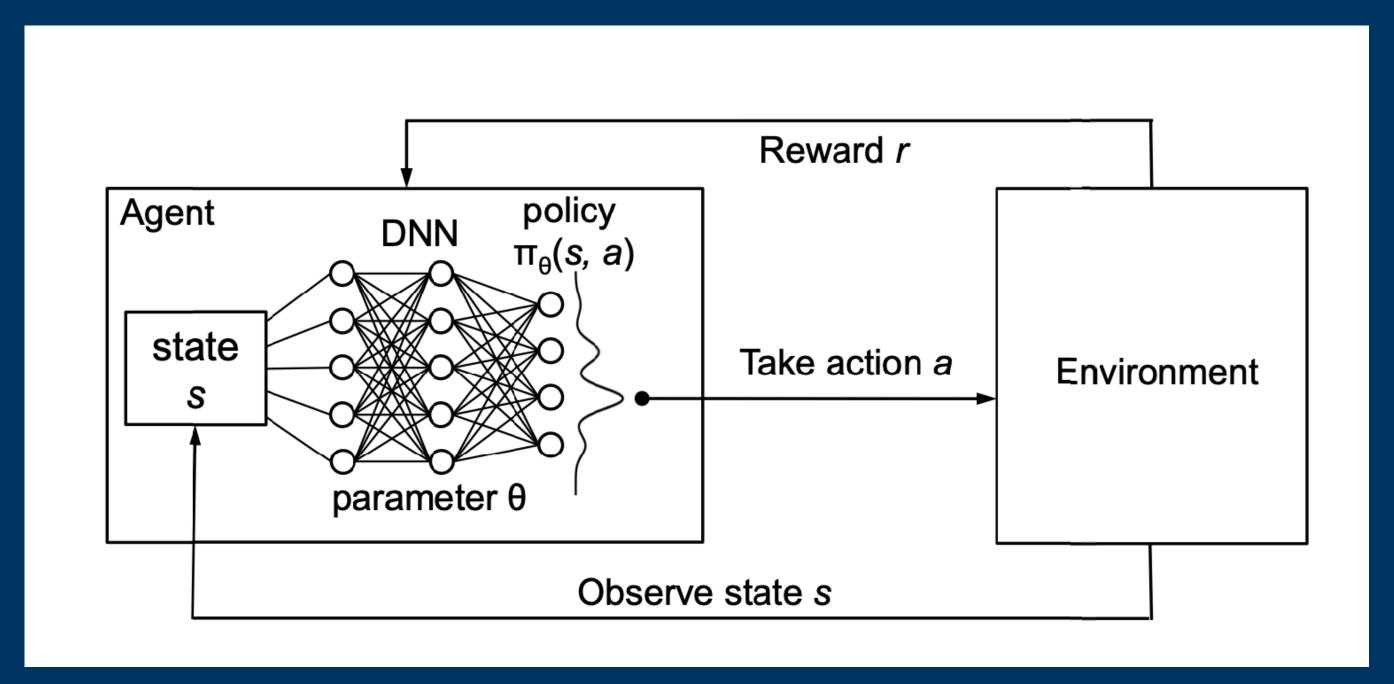
I developed a two-player competitive version of Conway's Game of Life, and I trained reinforcement learning agent (AI) to play against with human player.

## Motivation and Progress I made

- At the beginning of the project, instead of developing a game based on Conway's Game of Life, I was trying to put up with a mathematical model of Artificial life which can also survive, die, evolve, etc. The idea of Emergency catch me and I tried study two possible models myself: a model inspired by DNA and a model inspired by Taichi. (Not done)
- After that I turn to focus on Reinforcement Learning which is exciting and it's detailed in my project proposal. I try to train an agent that can deal with chaos which is unpredictable by human's intuition. (Done, see details in next slide)

### Reinforcement Learning

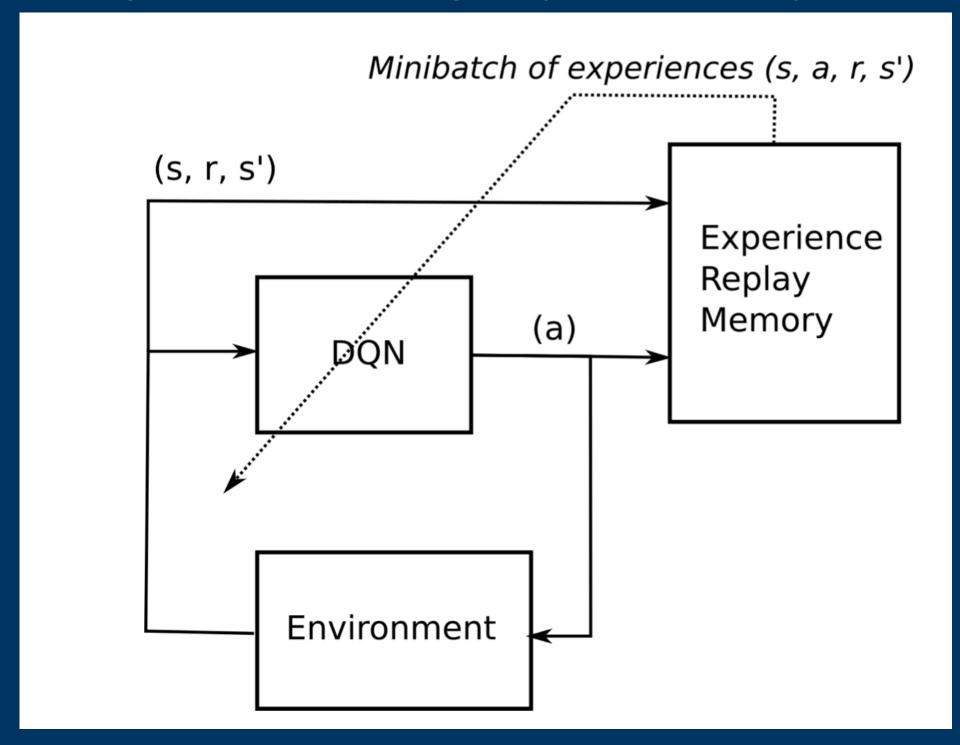
Deep Q Network (DQN) [1]



$$\mathcal{L}(\theta) = \mathbb{E}_{\pi}[(r_t + \gamma \, \max_{a'} Q_{\theta}(s', a') - Q_{\theta}(s, a))^2]$$

Loss function [2]

#### Experience replay memory [2]



Reward function

## Reinforcement Learning

#### This game is symmetric:

```
990 and reward: 250
episode:
episode:
              and reward:
episode:
         992 and reward:
         993 and reward:
episode:
         994 and reward:
episode:
episode:
         995 and reward:
episode:
         996 and reward:
episode:
         997 and reward:
         998 and reward:
episode:
episode:
         999 and reward:
episode:
         1000 and reward:
fianl wining: 0.496
(base) niceanyh@pc-68-253 dqn_keras_v2 % [
```

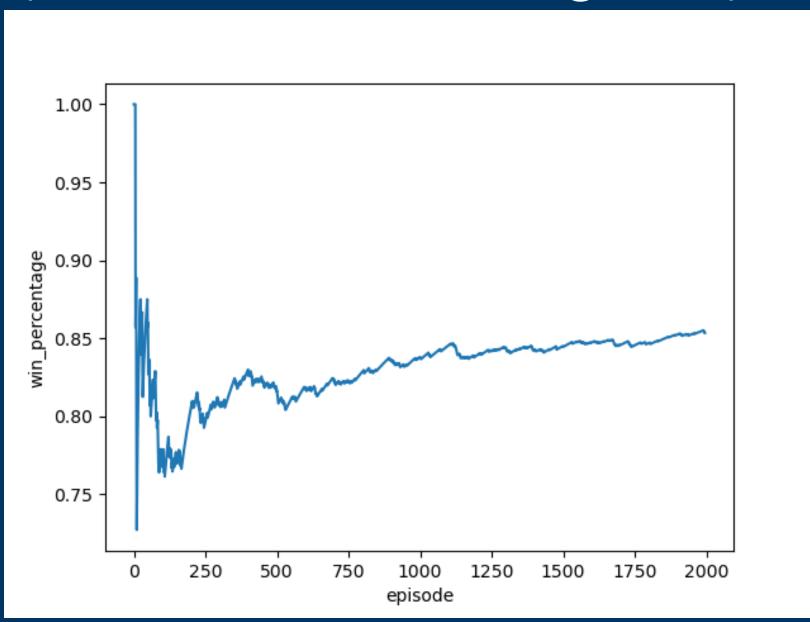
(Playing by two random-action players)

#### After Training:

```
990 and reward:
                            160
episode:
              and reward:
                            -100
episode:
              and reward:
                            260
episode:
               and reward:
episode:
               and reward:
                            -100
         995
              and reward:
                            180
episode:
                            220
              and reward:
episode:
                            -100
episode:
               and reward:
                            250
episode:
               and reward:
episode:
              and reward:
episode: 1000 and reward: 280
fianl wining: 0.854
(base) niceanyh@pc-68-253 dqn_keras_v2 % ■
```

(DQN agent play against random player)

# Training curve (accumulated wining rate)



- Currently model used: 5000 episodes playing against still life + 1000 episodes playing against random player.

## The Philosophy behind the project

- I try to imitate Monument Valley which is a game known for its minimalism.
- Inspired by a book called Sapiens: A Brief History of Humankind, I am quite into the idea of evolution, competition. A good game could also inspire people to think about that.
- The sound of life. Some patterns might have higher possibility to win (higher Q value), so can I train a classifier to let each state play piano. By listening to the sound of each state, we might find something interesting.

### Conclusion

The thing I'm most proud of is that I built a game environment and UI, defined the problem and developed Reinforcement Learning Agent, and the agent did a good job.

Knowing that a friend of mine' honour year project is to develop reinforcement learning models by using existing integrated game environments (gym)

### Reference

[1] H. Mao, M. Alizadeh, I. Menache, and S. Kandula, "Resource management with deep reinforcement learning," in Proceedings of the 15th ACM Workshop on Hot Topics in Networks, ser. HotNets '16. New York, NY, USA: Association for Computing Machinery, 2016, p. 50–56. [Online]. Available: <a href="https://doi.org/10.1145/3005745.3005750">https://doi.org/10.1145/3005745.3005750</a>

[2] J.Vitay "Deep Reinforcement Learning" [Online]. Available: <a href="https://julien-vitay.net/deeprl/Valuebased">https://julien-vitay.net/deeprl/Valuebased</a>

### Links

Github: https://github.com/Niceanyh/Gol\_dqn\_v2

Video: <a href="https://liverpool.instructuremedia.com/embed/75bcf9a1-edf9-436e-befd-7264b3f5fe69">https://liverpool.instructuremedia.com/embed/75bcf9a1-edf9-436e-befd-7264b3f5fe69</a>

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