

Ball Sort Puzzle

Reinforcement Learning

Artificial Intelligence

Master in Informatics and Computing Engineering



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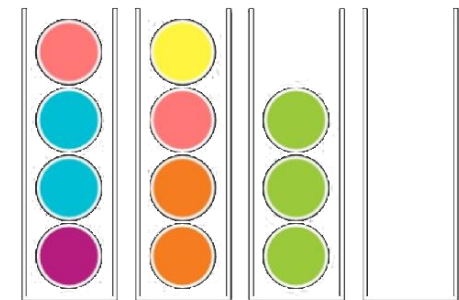
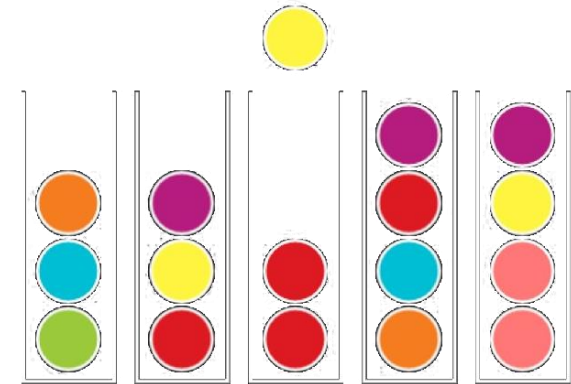
Formulation of the problem as an ML problem

The objective of the Ball Sort Puzzle is to sort the colored balls in the tubes until all balls with the same color stay in the same tube.

The player can only move a ball at the top of a tube to:

- an empty tube;
- another tube that has a ball with the same color on top and has enough space;

The player can also undo his moves if he finds himself without moves.



Formulation of the problem as a RL problem

State Representation:

- Ball: [1..n] (n = the number of colors)
- Tube: [Ball, Ball, Ball, Ball] or []
- Game: [Tube, Tube, ...]

Actions:

- Move(X,Y): move Ball from Tube X to Tube Y

Algorithms:

- Q-Learning
- SARSA
- Monte Carlo

Rewards:

Three ways to assign a reward based on verifications:

- Verify if the game is finished (+2) or the tube that received a ball is completed (+1)
- Verify if the game is invalid (-5) or finished (+20)
- Verify if the game is invalid (-1000), finished (+50) or count the number of consecutive balls with the same colour

Related Work

Mainly 2D games:

- [Retro](#)
- [Slime Volley Gym](#)
- [Demon Attack](#)

On OpenAI:

- [Spaces](#)
- [Environments](#)

On Policies:

- [Q-Learning](#)
- [Jupyter Notebook about Q-Learning](#)
- [SARSA](#)

Description of the tools

Jupyter Notebook – Allows for interactive computing

OpenAI gym - Allows for modeling of a reinforced learning environment

Algorithms - Since the action and the observation space of the puzzle are both discrete, the policies to be implemented are Q-Learning and SARSA

Implementation work already carried out

Programming language: [Python](#), [OpenAI](#), [Jupyter Notebook](#) (with [Anaconda](#))

Development environment: [VSCode](#)/[IntelliJ](#)

Data Structures:

- Lists, for representing the Tubes and the Game

File Structure: The Project's Repository is available at [Github](#)

Implemented Work: Modeled Environment and Agent, Q-Learning Algorithm

Models

As previously stated, both the Action and the Observation Space are discrete. Therefore, we can apply Q-Learning and SARSA to our RL problem. These algorithms are Model-Free.

```
#Function to Learn the Q-value, using Q-Learning
def updateQ(state, state2, reward, action):
    predict = Q[env.get_index(state), action]
    target = reward + discount_factor * np.max(Q[env.get_index(state2), :])
    Q[env.get_index(state), action] = predict + alpha * (target - predict)
```

Fig 1. Q-Learning Implementation

```
#Function to Learn the Q-value, using SARSA
def updateSarsa(state, state2, reward, action, action2):
    predict = Q[env.get_index(state), action]
    target = reward + discount_factor * Q[env.get_index(state2), action2]
    Q[env.get_index(state), action] = predict + alpha * (target - predict)
```

Fig 2. SARSA Implementation

Analysis

Started with a simple configuration, a puzzle with 2 colours and 3 tubes, and used the first evaluation function described: +10, if finished; +1, if a tube was completed using an action; -1, for every other move.

The parameters used were:

- Learning rate, 0.5
- Discount factor, 0.6
- Epsilon, 0.2
- Train episodes, 500

Analysis



Fig 3. Q-Learning with
Training score over time: 9.434



Fig 4. SARSA with
Training score over time: 9.316

Analysis

After this successful try, a more complex puzzle was tested: 3 colours and 5 tubes. This test was made with the same evaluation function as before.

As we can see and as expected, SARSA is a more conservative algorithm, evident by the absence of “dips” in later episodes. This is because SARSA is on-policy.



Fig 5. Q-Learning with
Training score over time: 8.424



Fig 6. SARSA with
Training score over time: 8.45

Conclusions

Although a very simplified version of a RL problem, this project allowed us to learn more about the topic of Machine Learning.

As for future work, we would've liked to create an abstraction of the puzzle, so that we could "feed" our agent with various configurations of the puzzle and it would be capable of solving them.