Foundations of Artificial Intelligence

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Exercise Session 10-22-2021

Registrations of the Exercise Sessions on WeBeep

Registration Links also available at

https://albertometelli.github.io/teaching/2021-teaching-fai

New! Some exercises shown during the sessions available at the same page

New! Feedback about Exercise Sessions

https://bit.ly/3C3fs5e

Please fill the form!

Exercise 3.3

Consider the following two-player zero-sum game. The game begins with a pile of seven bricks. On your move, you must split one pile of bricks into two piles. You may not split a pile of bricks into two equal piles. If it is your turn and all the piles of bricks have either one or two bricks, you have lost the game.

- 1. Formalize the problem
- 2. Apply the minimax algorithm for finding the best action for the max player at the root.
- 3. Apply the minimax algorithm with alpha-beta pruning for finding the best action for the max player at the root.

Exercise 3.3 - Problem formalization

$$S = \{\{p_1, p_2, ..., p_n\} : p_i \in \{1, 2, ..., 7\} \text{ and } \sum_i p_i = 7\}$$

$$A = \{(i, p) : i \in \{1, 2, ..., 7\} \text{ and } p \in \{1, 2, ..., 6\}\}$$

$$(i, p) \text{ is applicable in } \{p_1, p_2, ..., p_n\} \text{ iff } i \in \{1, 2, ..., n\} \text{ and } p \in \{1, 2, ..., floor((p_i-1)/2)\}$$

$$result(\{p_1, p_2, ..., p_{i-1}, p_i, p_{i+1}, ..., p_n\}, (i, p)) = \{p_1, p_2, ..., p_{i-1}, p, p_i-p, p_{i+1}, ..., p_n\}$$

$$terminal(\{p_1, p_2, ..., p_n\}) \text{ iff } p_i \in \{1, 2\} \text{ for all } i \in \{1, 2, ..., n\}$$

$$utility(\{p_1, p_2, ..., p_n\}) = +1 \text{ iff the current player is MIN}$$

Exercise 3.3 -Game tree



























































 $v \leq \alpha$? Yes \rightarrow PRUNE



Reinforcement Learning Exercise

Consider the following sequential decision making-problem. An agent in a 3 × 3 grid can move in the four directions or stay still, provided that it does not crush against a border. Whenever performing a valid action, the agent reaches deterministically to the corresponding cell. The interaction starts in the lower left cell (blue) and the upper right cell (green) is a terminal state. The immediate reward is represented in the following grid:

0	0	2
-1	-10	0
0	-1	0

Reinforcement Learning Exercise

1. Formalize the problem as a Markov decision process (MDP);

2. For which values of the discount factor $\gamma \in [0, 1]$ the optimal policy consists in staying in the initial state forever?

3. Simulate the execution of Q-learning, starting with a Q-table initialized with the immediate reward, supposing to have observed the following trajectories:

$$(0,0) \xrightarrow{\rightarrow} (1,0) \xrightarrow{\uparrow} (1,1) \xrightarrow{\rightarrow} (2,1) \xrightarrow{\uparrow} (2,2)$$
$$(2,1) \xrightarrow{\downarrow} (2,0) \xrightarrow{\uparrow} (2,1)$$
$$(1,1) \xrightarrow{\downarrow} (1,0) \xrightarrow{\rightarrow} (2,0)$$
$$(0,0) \xrightarrow{\rightarrow} (1,0) \xrightarrow{\uparrow} (1,1)$$

Use discount factor γ = 0.9 and learning rate α = 1.

4. Say which is the greedy policy once completed the updates of the Q-table.

Remember to answer to the feedback form!

https://bit.ly/3C3fs5e