## **COM 762: Deep Learning and Its Application**

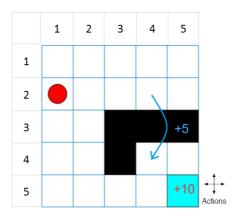
## Assignment 2

Submission date due: Friday 23<sup>rd</sup> August 2024 (Week 12)

Feedback date due: After 20 working days

This assignment carries 60% of the coursework marks of the module. The submitted file has to be named by the first half of your University email address and should be uploaded to the Assignment-1 folder in Blackboard by noon on the due date.

The objective of this assignment is to provide students with practical experience of developing a Q-learning agent for solving a grid world environment by a reinforcement learning and Python. The grid world environment is illustrated below:



which has the following configuration and rules:

- 1. the grid world is 5-by-5 and bounded by borders, with four possible actions (North = 1, South = 2, East = 3, West = 4).
- 2. the agent begins from cell [2,1] (second row, first column).
- 3. the agent receives a reward +10 if it reaches the terminal state at cell [5,5] (blue).
- 4. the environment contains a special jump from cell [2,4] to cell [4,4] with a reward of +5 (a special treatment may be necessary).
- 5. the agent is blocked by obstacles (black cells).
- 6. all other actions result in -1 reward.
- 7. the iteration function below will be used to calculate state values

$$V(S_t) \leftarrow V(S_t) + \alpha \Big[ V(S_{t+1}) - V(S_t) \Big]$$

The specific tasks are below:

(a) detail the task and process of exploration and exploitation for solving gird world problem.

- (b) describe the formulation of policy, environment, observation (state or grid cell or position) and action, etc. in the context of the grid world problem.
- (c) create a value table (or Q-table) using the state and action specifications from the grid world environment, set the learning rate to be 1 and various values in between [0, 1] to run the code.
- (d) create a Q-learning agent using this table representation and configure the epsilon-greedy exploration.
- (e) train the agent with the following options:
  - train for 100 episodes, where each episode should have the ability to control unlimited paths from starting position to terminal or jump-end position (for instance, stop training when the agent receives an average cumulative reward greater than 10 over 30 consecutive episodes.)
- (f) Visualise the state values in each of the grid cells with the board layout

## The submission package includes:

- 1. A report should be 5-7 pages in the IEEE format (maximum 3000 words), consisting of the description of tasks (a)-(f); and a screenshot for task (f). Note that the screenshot must be captured in the PyCharm environment,
- 2. The source code and the explanation of how to run the program.