

## HOMework 3 – Partially Observable MDP

### Exercise 1

- a) The problem described above can be modeled as an MDP. Identify the state space,  $X$ , and the action space,  $A$ , and write down the cost function for the MDP (you need not specify the transition probabilities). Consider throughout  $\gamma = 0.95$ .

	0	1	2	3	4
0	_	_	_	_	_
1	S	_	_	_	G
2	_	_	_	_	_

State Space:

$[(0, 0), (0, 1), (0, 2), (0, 3), (0, 4), (1, 0), (1, 1), (1, 2), (1, 3), (1, 4), (2, 0), (2, 1), (2, 2), (2, 3), (2, 4)]$

Action Space:

$\{ 'UP', 'RIGHT', 'DOWN', 'LEFT' \}$

Cost Function:

```
[[ 1.  1.  1.  1.  1.]
 [ 1.  1.  1.  1.  1.]
 [ 1.  1.  1.  1.  1.]
 [ 1.  1.  1.  1.  1.]
 [ 1.  1.  1.  1.  1.]
 [ 1.  1.  1.  1.  1.]
 [ 1.  1.  1.  1.  1.]
 [ 1.  1.  1.  1.  1.]
 [ 1.  1.  1.  1.  1.]
 [ 1.  1.  1.  1.  1.]
 [ 0.  0.  0.  0.  0.]
 [ 1.  1.  1.  1.  1.]
 [ 1.  1.  1.  1.  1.]
 [ 1.  1.  1.  1.  1.]
 [ 1.  1.  1.  1.  1.]
```

- b) Indicate the transition information resulting from the two actions of the agent (state, action, cost, next state).

```
State:      (1, 0)
Action:     RIGHT
Cost:       1.0
New State:  (1, 1)
```

```
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State:      (1, 1)
Action:     RIGHT
Cost:       1.0
New State:  (0, 2)
```

- c) Suppose that the agent is following the Q-learning algorithm, with the Q-function initialized as an all-zeros function. Indicate the Q-values after the two Q-learning updates with step-size  $\alpha = 0.1$ , resulting from the transitions in (b).

Updated Q-Matrix:

```
[[ 0.  0.  0.  0. ]
 [ 0.  0.  0.  0. ]
 [ 0.  0.  0.  0. ]
 [ 0.  0.  0.  0. ]
 [ 0.  0.  0.  0. ]
 [ 0.  0.1 0.  0. ]
 [ 0.  0.1 0.  0. ]
 [ 0.  0.  0.  0. ]
 [ 0.  0.  0.  0. ]
 [ 0.  0.  0.  0. ]
 [ 0.  0.  0.  0. ]
 [ 0.  0.  0.  0. ]
 [ 0.  0.  0.  0. ]
 [ 0.  0.  0.  0. ]
 [ 0.  0.  0.  0. ]]
```