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Unit 5 Reinforcement Learning (2

Course > weeks)

> Homework 6 > 3. Q-Learning

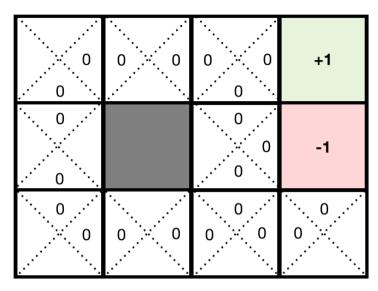
3. Q-Learning

Recall the Q-learning update rule:

$$Q_{i+1}\left(s,a
ight) = Q_{i}\left(s,a
ight) + lpha\left[R\left(s,a,s'
ight) + \gamma max_{a'}Q_{i}\left(s',a'
ight) - Q_{i}\left(s,a
ight)
ight]$$

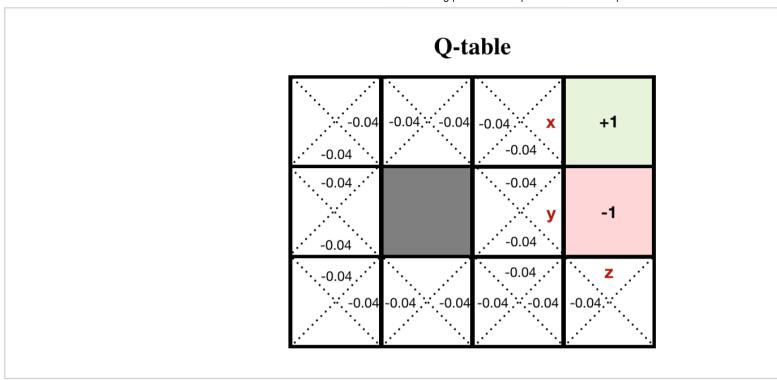
let $\alpha=1$ and $\gamma=1$ in this problem. In the figure below, at each box, we can go up, down, left and right unless the path is blocked and we initialize the Q value for all the actions in all states as 0. The Q value for the 4 directions are labeled in each box below. Moving into the upper right 2 boxes will result in a reward of +1 and -1, and each move will also cost 0.04, or in another word, a reward of -0.04.

Q-table



1st Iteration

3/3 points (graded)



After 1st iteration, enter the Q value at the position represented by x, y and z below:

$$y = \begin{bmatrix} -1.04 \\ \checkmark \text{ Answer: -1.04} \end{bmatrix}$$

$$z = \begin{vmatrix} -1.04 \end{vmatrix}$$
 Answer: -1.04

Solution:

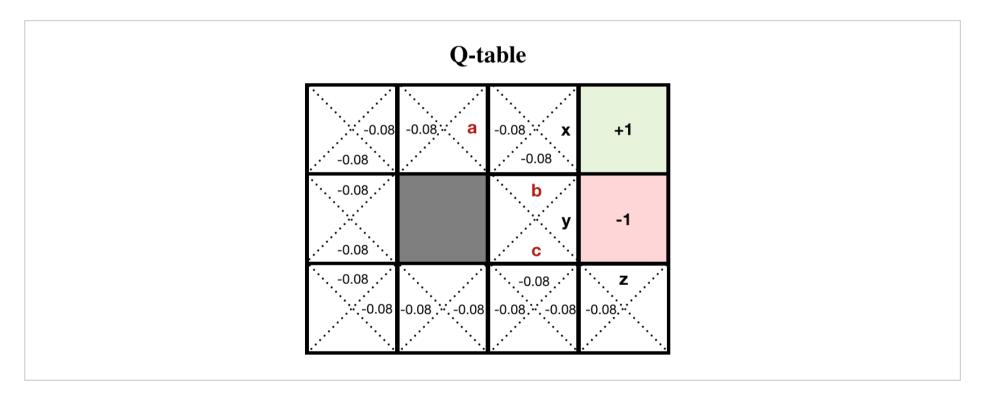
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You have used 1 of 3 attempts

1 Answers are displayed within the problem

2nd Iteration

3/3 points (graded)



After 2nd iteration, enter the Q value at the position represented by $a,\,b$ and c below:

 $c = \begin{vmatrix} -0.08 \end{vmatrix}$ Answer: -0.08

Solution:

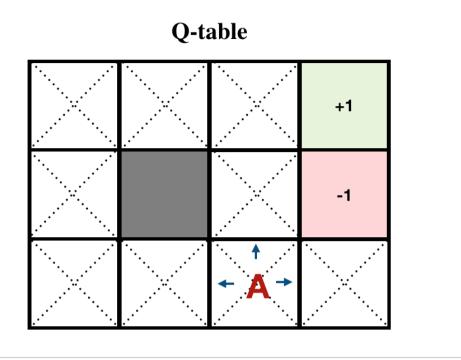
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You have used 1 of 3 attempts

1 Answers are displayed within the problem

2nd Iteration

1/1 point (graded)



After convergence, at state A, which action is the optimal?

UP	~





Solution:

Submit • Answ

You have used 1 of 1 attempt

1 Answers are displayed within the problem

Epsilon-greedy method 1

1/1 point (graded)

In the ε -greedy method, a larger value of ε would generate experiences that are more consistent with the current Q-value estimates.



Solution:

In the ε -greedy method, we choose a random action with probability ε and choose an action based on our current estimates with probability 1 - ε . Therefore, it is with smaller ε that we would generate experiences which are more consistent with our current Q-value estimates.

Submit

You have used 1 of 1 attempt

• Answers are displayed within the problem

Epsilon-greedy method 2

1/1 point (graded)

In the ε -greedy method, a value of $\varepsilon=0.999$ is likely to lead to the desired learning outcome in a highly complex environment.

O True			
● False ✔			

Solution:

We would pick a random action virtually every time, and in a highly complex environment, it's highly unlikely that we would properly explore the parts of the space that have high rewards.

Submit

You have used 1 of 1 attempt

1 Answers are displayed within the problem

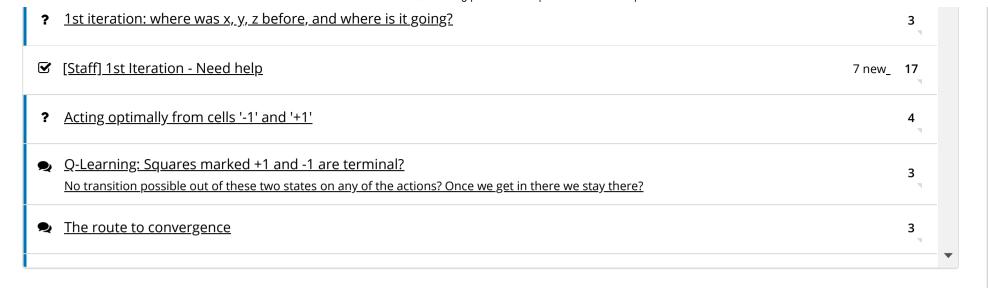
Discussion

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Topic: Unit 5 Reinforcement Learning (2 weeks): Homework 6 / 3. Q-Learning

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[Staff] 2nd Iteration Q Table If ethically permitted, can anyone explain to me why the max Q' for c isn't08 or 0? I tried both and was graded incorrect. As far as I can see, oth	8
@Staff omit Epsilon-greedy method 2 please Answer is depended on assumptions that are not touched in the question. Please omit this question. In general, this homework(and this module)	10
[staff] Iteration (2) If 'b' and 'c' are at the same state, why are not their values the same?	
2nd Iteration 1. Is this the equation being used for Q2. Q2 = Q1 + alpha*(R+gamma*max(Q1(s',a')) - Q1) 2. For second iteration, do we look at the state of each	5
1st iteration - add attempt Hi, is there anyway I could have one more attempt? I didn't mean to leave negatives in my answer. Thanks!	2
[Staff] About Epsilon-greedy method 2 I think in a HIGHLY COMPLEX environment it is better to consider how to easily jump out from a dead-cycle(in other words: local optimal) which c	9
<u>Cool Tool ;)</u> https://www.mladdict.com/q-learning-simulator play around and learn how Q-learning works	3
[Staff] Epsilon-greedy method 2 Is question referring to epsilon at the beginning of the learning or end of the learning or is it a fixed value?	11
P Did anyone figure out 1st iteration? Is x moving or staying put? If so, where is it going?	7



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