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Unit 5 Reinforcement Learning (2

Lecture 17. Reinforcement Learning

Course > weeks)

7. Value Iteration

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# 7. Value Iteration Value Iteration





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Recall from lecture the value iteration update rule:

$$V_{k+1}^{st}\left(s
ight)=\max_{a}\left[\sum_{s^{\prime}}T\left(s,a,s^{\prime}
ight)\left(R\left(s,a,s^{\prime}
ight)+\gamma V_{k}^{st}\left(s^{\prime}
ight)
ight)
ight],$$

where  $V_{k}^{st}\left(s
ight)$  is the expected reward from state s after acting optimally for k steps.

Recall the example discussed in the lecture.



An agent is trying to navigate a one-dimensional grid consisting of 5 cells. At each step, the agent has only one action to choose from, i.e. it moves to the cell on the immediate right.

**Note:** The reward function is defined to be  $R\left(s,a,s'\right)=R\left(s\right)$ ,  $R\left(s=5\right)=1$ and  $R\left(s
ight)=0$  otherwise. Note that we get the reward when we are leaving from the current state. When it reaches the rightmost cell, it stays for one more time step and then receives a reward of +1 and comes to a halt.

Let  $V^{st}\left(i
ight)$  denote the value function of state i, the  $i^{th}$  cell starting from left.

Let  $V_{k}^{st}\left(i
ight)$  denote the value function estimate at state i at the  $k^{th}$  step of the value iteration algorithm. Let  $V_0^st\left(i
ight)$  denote the initialization of this estimate.

Use the discount factor  $\gamma = 0.5$ .

We will write the functions 
$$V_k^*$$
 as arrays below, i.e. as  $\begin{bmatrix} V_k^*\left(1\right) & V_k^*\left(2\right) & V_k^*\left(3\right) & V_k^*\left(4\right) & V_k^*\left(5\right) \end{bmatrix}$ .

Initialize by setting  $V_{0}^{st}\left( i\right) =0$  for all i:

$$V_0^* = [0 \ 0 \ 0 \ 0 \ 0].$$

Then, using the value iteration update rule, we get

$$V_1^* = \begin{bmatrix} 0 & 0 & 0 & 0 & 1 \end{bmatrix},$$

$$V_2^* = \begin{bmatrix} 0 & 0 & 0 & 0.5 & 1 \end{bmatrix}$$

**Note:** Note that as soon as the agent takes the first action to reach cell 5, it stays for one more step and halts and does not take any more action, so we set  $V_{k+1}^*\left(5\right)=V_k^*\left(5\right)$  for all  $k\geq 1$ .

## Value Function Update

1.0/1 point (graded)

Run the  $3^{r\bar{d}}$  iteration of the value iteration algorithm to get  $V_3^*$  and answer the following questions:

Enter the value of  $V_3^{\,st}$  as an array

$$\begin{bmatrix} V_3^* (1) & V_3^* (2) & V_3^* (3) & V_3^* (4) & V_3^* (5) \end{bmatrix}.$$

(For example, type [0,2,0,3,4] for the array  $[\,0\quad 2\quad 0\quad 3\quad 4\,]$ .)

#### **Solution:**

Note that a non-zero reward is obtained only in state  $s_4$  when transitioning to  $s_5$ .

The  $3^{rd}$  step of the value iteration could be worked out as follows:

$$V_3^* (1) = 0 + \gamma * V_2^* (2)$$

$$V_3^*(1) = 0 + 0.5 * 0 = 0$$

$$egin{array}{lll} V_3^* \left( 2 
ight) &=& 0 + \gamma * V_2^* \left( 3 
ight) \ V_3^* \left( 2 
ight) &=& 0 + 0.5 * 0 = 0 \ \ V_3^* \left( 3 
ight) &=& 0 + \gamma * V_2^* \left( 4 
ight) \ V_3^* \left( 3 
ight) &=& 0 + 0.5 * 0.5 = 0.25 \end{array}$$

$$V_3^* (4) = 0 + \gamma * V_2^* (5)$$
  
 $V_3^* (4) = 0 + 0.5 * 1 = 0.5$ 

and 
$$V_{3}^{*}\left(5\right)=V_{2}^{*}\left(5\right)=1.$$

The same computation for the rest of the states.

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You have used 1 of 3 attempts

**1** Answers are displayed within the problem

### Number of Steps to Convergence

1/1 point (graded)

Enter below the number of steps it takes starting from  $V_0^st$  for the value function updates to converge to the optimal value function  $V^st$ :

5 **✓ Answer:** 5

#### **Solution:**

Note that after the  $5^{th}$  step, the reward from the rightmost cell in the grid gets propagated to the leftmost state after which the value function estimate  $V_k^*$  stops updating. Hence, for this example it takes 5 steps for the value function estimate to converge to the optimal value function.

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You have used 1 of 2 attempts

**1** Answers are displayed within the problem

## Complexity of Value Iteration

1/1 point (graded)

Let the number of states and actions be |S| and |A|, respectively. Choose from the following the **complexity of an iteration** of the value iteration algorithm.

- $\bigcirc \mathcal{O}\left(\left|S
  ight|^3 \cdot \left|A
  ight|
  ight)$
- $\bigcirc \mathcal{O}\left(|S|\cdot |A|
  ight)$
- $lefte{O}\left(\left|S
  ight|^2\cdot\left|A
  ight|
  ight)$



#### **Solution:**

We update the expected reward for each state in every iteration – there are |S| states. For each state, we investigate a maximum of |A| possible actions and for each such action there are |S| possible transitions at the most. Therefore, the complexity of an iteration is  $\mathcal{O}\left(|S|^2\cdot|A|\right)$ .

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You have used 2 of 2 attempts

**1** Answers are displayed within the problem

## Another Example of Value Iteration (Software Implementation)

3 points possible (graded)

Consider the same one-dimensional grid with reward values as in the first few

problems in this vertical. However, consider the following change to the transition probabilities: At any given grid location the agent can choose to either stay at the location or move to an adjacent grid location. If the agent chooses to stay at the location, such an action is successful with probability 1/2 and

- if the agent is at the leftmost or rightmost grid location it ends up at its neighboring grid location with probability 1/2,
- ullet if the agent is at any of the inner grid locations it has a probability 1/4 each of ending up at either of the neighboring locations.

If the agent chooses to move (either left or right) at any of the inner grid locations, such an action is successful with probability 1/3 and with probability 2/3 it fails to move, and

- if the agent chooses to move left at the leftmost grid location, then the action ends up exactly the same as choosing to stay, i.e., staying at the leftmost grid location with probability 1/2, and ends up at its neighboring grid location with probability 1/2,
- if the agent chooses to move right at the rightmost grid location, then the action ends up exactly the same as choosing to stay, i.e., staying at the rightmost grid location with probability 1/2, and ends up at its neighboring grid location with probability 1/2.

Let 
$$\gamma=0.5$$
.

Run the value iteration algorithm for 100 iterations. Use any computational software of your choice.

Enter the value of 
$$V_{100}^{*}$$
 as an array 
$$\begin{bmatrix} V_{100}^{*}\left(1\right) & V_{100}^{*}\left(2\right) & V_{100}^{*}\left(3\right) & V_{100}^{*}\left(4\right) & V_{100}^{*}\left(5\right) \end{bmatrix}.$$

(For example, type [0,2,0,3,4] for the array  $\begin{bmatrix} 0 & 2 & 0 & 3 & 4 \end{bmatrix}$ . Type at least 4 decimal digits.)

**Answer:** [0.016667,0.05,0.2,0.8,1.2]

Are the values different if we iterate 200 times? Consider only the first three decimal

digits to answer this question.



How about if we only performed 10 iterations? Are the values different when compared to 100 iterations? Consider only the first three decimal digits to answer this question.

No
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#### **Solution:**

After 100 iterations,

$$V_{100}^* = [0.016667, 0.05, 0.2, 0.8, 1.2]$$
 .

If we run the algorithm for another 100 iterations, we see that the values do not change in the first three decimal digits.

However, at only 10 iterations,

$$V_{10}^* = \left[0.015886, 0.049121, 0.199040, 0.799023, 1.199023\right].$$

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You have used 0 of 4 attempts

**1** Answers are displayed within the problem

## Discussion

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[SOLUTION] Python Following on from riccardo_riccobello's kind posting of his Excel solution, I thought it would b	11

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<b>Q</b>	[STAFF] Another Example of Value Iteration - provide the correct code  Hi Staff, I've solved this problem four times. One of these times, I was using [ pymdptoolbox	13
Q	[Staff] Q. Number of Steps to Convergence	2
2	My understanding about Value Iteration Algorigh (based in problem "Software Implementation")  I was many time trying to understand VIA and how it works to solve the questions, but somet	1

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