

2. Q-Value Iteration

Consider an Markov Decision Process with 6 states $s \in \{0, 1, 2, 3, 4, 5\}$ and 2 actions $a \in \{C, M\}$, defined by the following transition probability functions

For states 1, 2, and 3:

$$T(s, M, s - 1) = 1$$

$$T(s, C, s + 2) = 0.7$$

$$T(s, C, s) = 0.3$$

For state 0:

$$T(s, M, s) = 1$$

$$T(s, C, s) = 1$$

For states 4 and 5:

$$T(s, M, s - 1) = 1$$

$$T(s, C, s) = 1$$

Note that all transition probabilities not defined by the above are equal to 0.

The rewards R are defined by:

$$R(s, a, s') = \left| (s' - s)^{\frac{1}{3}} \right| \forall s \neq s',$$

$$\text{and } R(s, a, s) = (s + 4)^{-\frac{1}{2}}, \forall s \neq 0.$$

$$R(0, M, 0) = R(0, C, 0) = 0 \text{ Also, the discount factor } \gamma = 0.6.$$

We initialize $Q_0(s, a) = 0 \forall s \in \{0, 1, 2, 3, 4, 5\}$ and $\forall a \in \{C, M\}$.

1

1/1 point (graded)

We can conclude from this information that 0 is a terminal state.

☒ True ✓

☐ False

Solution:

From the transition probabilities, we can see that no matter which action you take, once you are in state 0, you can never leave.

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You have used 1 of 1 attempt

Answers are displayed within the problem

2

6.0/6.0 points (graded)
Input the Q-values $Q_1(s, a)$ **correct to 3 decimal places** after one Q-value iteration

- $Q_1(0, M) =$ ✓ Answer: 0
- $Q_1(0, C) =$ ✓ Answer: 0
- $Q_1(1, M) =$ ✓ Answer: 1
- $Q_1(1, C) =$ ✓ Answer: 1.016
- $Q_1(2, M) =$ ✓ Answer: 1
- $Q_1(2, C) =$ ✓ Answer: 1.004
- $Q_1(3, M) =$ ✓ Answer: 1
- $Q_1(3, C) =$ ✓ Answer: 0.995
- $Q_1(4, M) =$ ✓ Answer: 1
- $Q_1(4, C) =$ ✓ Answer: 0.354
- $Q_1(5, M) =$ ✓ Answer: 1
- $Q_1(5, C) =$ ✓ Answer: 0.333

Solution:

- $Q_1(0, M)$: $Q_1(0, M) = 0$ because $R(0, M, 0) = 0$ and $T(0, M, s') = 0 \forall s' \neq 0$
- $Q_1(0, C)$: $Q_1(0, C) = 0$ because $R(0, C, 0) = 0$ and $T(0, C, s') = 0 \forall s' \neq 0$
- $Q_1(1, M)$: $\left| (0 - 1)^{\frac{1}{3}} \right| = 1$
- $Q_1(1, C)$: $0.7 * \left| (3 - 1)^{\frac{1}{3}} \right| + 0.3 * 5^{\frac{-1}{2}} = 0.882 + 0.134 = 1.016$
- $Q_1(2, M)$: Just as in $Q_1(1, M)$
- $Q_1(2, C)$: $0.7 * \left| (3 - 1)^{\frac{1}{3}} \right| + 0.3 * 5^{\frac{-1}{2}} = 0.882 + 0.122 = 1.004$
- $Q_1(3, M)$: Just as in $Q_1(1, M)$

8. $Q_1(3, C): 0.7 * \left| (3 - 1)^{\frac{1}{3}} \right| + 0.3 * 5^{\frac{-1}{2}} = 0.882 + 0.113 = 0.995$

9. $Q_1(4, M)$: Just as in $Q_1(1, M)$

10. $Q_1(4, C): 8^{\frac{-1}{2}} = 0.354$

11. $Q_1(5, M)$: Just as in $Q_1(1, M)$

12. $Q_1(5, C): 9^{\frac{-1}{2}} = 0.333$

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You have used 2 of 4 attempts

i Answers are displayed within the problem

3

3.0/3.0 points (graded)
What are the values $V_1(s)$ corresponding to $Q_1(s, a)$?

$V_1(0) =$ ✔ Answer: 0

$V_1(1) =$ ✔ Answer: 1.016

$V_1(2) =$ ✔ Answer: 1.004

$V_1(3) =$ ✔ Answer: 1

$V_1(4) =$ ✔ Answer: 1

$V_1(5) =$ ✔ Answer: 1

Solution:

Because: $V_1(s) = \max_a Q_1(s, a)$

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You have used 1 of 2 attempts

i Answers are displayed within the problem

4

5/5 points (graded)
What are the optimal policies we get from $Q_1(s, a)$?

$\pi^*(1) =$

☒ C ✔

☐ M

$\pi^*(2) =$

☒ C ✔

☐ M

$\pi^*(3) =$

☐ C

☒ M 

$\pi^*(4) =$

☐ C

☒ M 

$\pi^*(5) =$

☐ C

☒ M 

Solution:

We pick the policy corresponding to the $V_1(s)$ i.e. $\pi^*(s) = \underset{a}{argmax} Q_1(s, a)$

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You have used 2 of 2 attempts

 Answers are displayed within the problem

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