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## 2. Q-Value Iteration

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Homework due May 3, 2023 08:59 -03 Completed

Consider an Markov Decision Process with 6 states  $s \in \{0, 1, 2, 3, 4, 5\}$  and 2 actions 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:

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

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

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

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

2

6.0/6.0 points (graded)

Input the Q-values

**correct to 3 decimal places** after one Q-value iteration

0



0



1



1.016



1



1.004



1



0.995



1



0.354



1



0.333



Submit

You have used 1 of 4 attempts

1



Submit

You have used 1 of 2 attempts

4

5/5 points (graded)

What are the optimal policies we get from ?

☒ C

☐ M



☒ C

☐ M



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For the case where both  $s \neq s'$  and  $s \neq 0$  hold true, which reward function should we use? The first, the second,

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