

# Reinforcement Learning (SS18) - Exercise 2

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09.05.2018

## 5. Calculate the optimal value function $v_*$ for the cleaning robot problem.

We assume that the optimal (or only) action for state  $s = 0$  is  $a = 1$  and for state  $s = 5$  it is  $a = -1$ .

From the definition of  $v_*$  we get the following system of non-linear equations:

$$\begin{aligned}v_*(0) &= \gamma v_*(1) \\v_*(1) &= \max\{1 + \gamma v_*(0), \gamma v_*(2)\} \\v_*(2) &= \max\{\gamma v_*(1), \gamma v_*(3)\} \\v_*(3) &= \max\{\gamma v_*(2), \gamma v_*(4)\} \\v_*(4) &= \max\{\gamma v_*(3), 5 + \gamma v_*(5)\} \\v_*(5) &= \gamma v_*(4)\end{aligned}$$

Independent of the value of  $\gamma$  we know that for states 3 and 4 the optimal action is 1. For state 4 the argument is trivial due to the immediate reward; in state 3 we need two transitions to receive reward when going *right* and three transitions when going *left*.

$$\begin{aligned}v_*(3) &= \gamma v_*(4) \\v_*(4) &= 5 + \gamma v_*(5)\end{aligned}$$

With  $v_*(5) = \gamma v_*(4)$ :  $v_*(4) = 5 + \gamma^2 v_*(4) \Rightarrow v_*(4) = \frac{5}{1-\gamma^2}$  and  $v_*(3) = v_*(5) = \frac{5\gamma}{1-\gamma^2}$ .

We want to solve  $\gamma v_*(1) \stackrel{!}{=} \gamma v_*(3)$ , thus making the agent indifferent in state 2. If the agent is indifferent in state 2, going *left* is optimal, it follows that in state 1 going *left* must be optimal otherwise the agent would alternate between 1 and 2 and not accumulate reward.

$$v_*(1) = 1 + \gamma v_*(0)$$

With  $v_*(0) = \gamma v_*(1)$ :  $v_*(1) = 1 + \gamma^2 v_*(1) \Rightarrow v_*(1) = \frac{1}{1-\gamma^2}$  and  $v_*(0) = \frac{\gamma}{1-\gamma^2}$ .

Thus,

$$\gamma v_*(1) \stackrel{!}{=} \gamma v_*(3) \Leftrightarrow v_*(1) = \frac{1}{1-\gamma^2} \stackrel{!}{=} \frac{5\gamma}{1-\gamma^2} = v_*(3) \Rightarrow \gamma = \frac{1}{5}$$

For  $\gamma = \frac{1}{5}$  the agent is indifferent in state 2 and we have  $v_*(2) = \gamma v_*(1) = \gamma v_*(3) = \frac{5}{24}$ .

For  $\gamma < \frac{1}{5}$  the optimal action is  $-1$  (*left*) and we have  $v_*(2) = \gamma v_*(1) = \frac{\gamma}{1-\gamma^2}$ .

For  $\gamma > \frac{1}{5}$  the optimal action is 1 (*right*) and we have  $v_*(2) = \gamma v_*(3) = \frac{5\gamma^2}{1-\gamma^2}$ .

For the last case we look at the value in state 1:

$$v_*(1) = \max\{1 + \gamma v_*(0), \gamma v_*(2)\}$$

If the agent is indifferent in state 1, action  $-1$  must be optimal and thus we have from above:  $v_*(1) = \frac{1}{1-\gamma^2}$ .

Thus<sup>1</sup>,

$$v_*(1) = \frac{1}{1-\gamma^2} \stackrel{!}{=} \frac{5\gamma^3}{1-\gamma^2} = \gamma v_*(2) \Leftrightarrow \frac{1}{5} = \gamma^3 \Rightarrow \gamma = \sqrt[3]{\frac{1}{5}}$$

For  $\gamma = \sqrt[3]{\frac{1}{5}}$  the agent is indifferent in state 1 and we have  $v_*(1) = 1 + \gamma v_*(0) = \gamma v_*(2) = \frac{1}{1-\frac{1}{5^{2/3}}}$ .

For  $\gamma < \sqrt[3]{\frac{1}{5}}$  the optimal action is  $-1$  (*left*) and we have  $v_*(1) = \frac{1}{1-\gamma^2}$ .

For  $\gamma > \sqrt[3]{\frac{1}{5}}$  the optimal action is  $1$  (*right*) and we have  $v_*(1) = \gamma v_*(2) = \frac{5\gamma^3}{1-\gamma^2}$ .

In summary:

$$0 \leq \gamma \leq \frac{1}{5} : \quad v_*(0) = \frac{\gamma}{1-\gamma^2} \quad v_*(1) = \frac{1}{1-\gamma^2} \quad v_*(2) = \frac{\gamma}{1-\gamma^2} \quad v_*(3) = \frac{5\gamma}{1-\gamma^2} \quad v_*(4) = \frac{5}{1-\gamma^2} \quad v_*(5) = \frac{5\gamma}{1-\gamma^2}$$

$$\frac{1}{5} \leq \gamma \leq \sqrt[3]{\frac{1}{5}} : \quad v_*(0) = \frac{\gamma}{1-\gamma^2} \quad v_*(1) = \frac{1}{1-\gamma^2} \quad v_*(2) = \frac{5\gamma^2}{1-\gamma^2} \quad v_*(3) = \frac{5\gamma}{1-\gamma^2} \quad v_*(4) = \frac{5}{1-\gamma^2} \quad v_*(5) = \frac{5\gamma}{1-\gamma^2}$$

$$\sqrt[3]{\frac{1}{5}} \leq \gamma < 1 : \quad v_*(0) = \frac{5\gamma^4}{1-\gamma^2} \quad v_*(1) = \frac{5\gamma^3}{1-\gamma^2} \quad v_*(2) = \frac{5\gamma^2}{1-\gamma^2} \quad v_*(3) = \frac{5\gamma}{1-\gamma^2} \quad v_*(4) = \frac{5}{1-\gamma^2} \quad v_*(5) = \frac{5\gamma}{1-\gamma^2}$$

In terms of optimal policies we have:

$$\begin{array}{l} \gamma = 0 : \quad \begin{array}{cccccc} \leftarrow 0 & \leftarrow 1 & \leftarrow 2 & \leftarrow 3 & \leftarrow 4 & \leftarrow 5 \end{array} \\ 0 < \gamma < \frac{1}{5} : \quad \begin{array}{cccccc} 0 & \leftarrow 1 & \leftarrow 2 & \leftarrow 3 & \leftarrow 4 & \leftarrow 5 \end{array} \\ \gamma = \frac{1}{5} : \quad \begin{array}{cccccc} 0 & \leftarrow 1 & \leftarrow 2 & \leftarrow 3 & \leftarrow 4 & \leftarrow 5 \end{array} \\ \frac{1}{5} < \gamma < \sqrt[3]{\frac{1}{5}} : \quad \begin{array}{cccccc} 0 & \leftarrow 1 & \leftarrow 2 & \leftarrow 3 & \leftarrow 4 & \leftarrow 5 \end{array} \\ \gamma = \sqrt[3]{\frac{1}{5}} : \quad \begin{array}{cccccc} 0 & \leftarrow 1 & \leftarrow 2 & \leftarrow 3 & \leftarrow 4 & \leftarrow 5 \end{array} \\ \sqrt[3]{\frac{1}{5}} < \gamma < 1 : \quad \begin{array}{cccccc} 0 & \rightarrow 1 & \rightarrow 2 & \rightarrow 3 & \rightarrow 4 & \rightarrow 5 \end{array} \end{array}$$

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<sup>1</sup>Alternatively, you can solve:  $1 + \gamma v_*(0) = 1 + \frac{\gamma^2}{1-\gamma^2} \stackrel{!}{=} \frac{5\gamma^3}{1-\gamma^2} = \gamma v_*(2) \Leftrightarrow 1 - \gamma^2 + \gamma^2 = 1 \stackrel{!}{=} 5\gamma^3 \Rightarrow \gamma = \sqrt[3]{\frac{1}{5}}$