## Reinforcement Learning Exercise 2 - Solution

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## **Proofs**

a) Bellman optimality operator is a gamma-contraction We want to show

$$(\mathcal{T}v)(s) = \max_{a} \sum_{s',r} p(s',r|s,a)[r + \gamma v(s')]$$
 (1)

fullfills the  $\gamma$ -contraction property, namely

$$\|\mathcal{T}v - \mathcal{T}w\|_{\infty} \le \gamma \|v - w\|_{\infty} \tag{2}$$

Inspired by the lecture for the Bellman expectation backup operator, we will similarly use the definition of the infinity norm to show the contraction property

$$\|\mathcal{T}v - \mathcal{T}w\|_{\infty} = \|\max_{a} \sum_{s',r} p(s',r|s,a)[r + \gamma v(s')] - \max_{a} \sum_{s',r} p(s',r|s,a)[r + \gamma w(s')]\|$$
(3)

$$\leq \| \max_{a} \sum_{s',r} p(s',r|s,a) [r + \gamma v(s') - (r + \gamma w(s'))] \|$$
 (4)

$$= \gamma \| \max_{a} \sum_{s',r} p(s',r|s,a) [v(s') - w(s')] \|$$
 (5)

$$\leq \gamma \| \max_{a} \sum_{s',r} p(s',r|s,a) \| v(s') - w(s') \|_{\infty} \|$$
 (6)

$$\leq \gamma \|v - w\|_{\infty} \tag{7}$$