

Lecture Notes on

# **Reinforcement Learning**

Dhilan Teeluckdharry  
teeluckn@mcmaster.ca

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## 1 Dynamic Programming

### 1.1 Policy Improvement Theorem

Consider a policies  $\pi, \pi'$  s.t.  $q_\pi(s, \pi'(s)) \geq v_\pi(s) \ \forall s \in \mathbb{S}$

$$\begin{aligned} v_\pi(s) &\leq q_\pi(s, \pi'(s)) \\ &= \mathbb{E}[R_{t+1} + \gamma v_\pi(S_{t+1}) \mid S_t = s, A_t = \pi'(s)] \\ &= \mathbb{E}_{\pi'}[R_{t+1} + \gamma v_\pi(S_{t+1}) \mid S_t = s] \\ &\leq \mathbb{E}_{\pi'}[R_{t+1} + \gamma q_\pi(S_{t+1}, \pi'(S_{t+1})) \mid S_t = s] \\ &= \mathbb{E}_{\pi'}[R_{t+1} + \gamma \mathbb{E}_{\pi'}[R_{t+2} + \gamma v_\pi(S_{t+2}) \mid S_{t+1}, A_{t+1} = \pi'(S_{t+1})] \mid S_t = s] \\ &= \mathbb{E}_{\pi'}[R_{t+1} + \gamma R_{t+2} + \gamma^2 v_\pi(S_{t+2}) \mid S_t = s] \\ &\leq \mathbb{E}_{\pi'}[R_{t+1} + \gamma R_{t+2} + \gamma^2 R_{t+3} + \gamma^3 v_\pi(S_{t+3}) \mid S_t = s] \\ &\dots \leq \mathbb{E}_{\pi'}\left[\sum_{k=0}^N \gamma^k R_{t+k+1} \mid S_t = s\right] = v_{\pi'}(s) \quad \square \end{aligned}$$