

goal: derive Bellman equation for MDP / SOCP

refs: *Neuro-Dynamic Programming*

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chapter 2

Reinforcement Learning

An Introduction

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chapter 3

• consider MDP/SOCP (X, \mathcal{U}, P, c) with infinite-horizon exponentially-discounted cost
$$\min_u E[c(x, u)] \quad \text{where} \quad c(x, u) = \sum_{t=0}^{\infty} \gamma^t \cdot \mathcal{L}(x_t, u_t)$$

s.t. $x^+ \sim P(x, u)$
 $\uparrow \gamma \in (0, 1)$

• given policy $\pi: X \rightarrow \Delta(\mathcal{U})$, define value function $v^\pi: X \rightarrow \mathbb{R}$
$$\forall x \in X: v^\pi(x) = E[c(x, u) \mid x_0 = x, u_t \sim \pi(x_t)]$$

→ assuming $|X|, |\mathcal{U}| < \infty$, show v^π satisfies Bellman equation

$$\forall x \in X: v^\pi(x) = \sum_{u \in \mathcal{U}} \pi(u|x) \sum_{x^+ \in X} P(x^+|x, u) \cdot (\mathcal{L}(x, u) + \gamma \cdot v^\pi(x^+))$$

$$- v^\pi(x) = E\left[\sum_{t=0}^{\infty} \gamma^t \cdot \mathcal{L}(x_t, u_t) \mid x_0 = x, u_t \sim \pi(x_t)\right]$$

$$\begin{aligned}
&= E \left[\gamma^0 \cdot \mathcal{L}(x_0, u_0) + \sum_{t=1}^{\infty} \gamma^t \cdot \mathcal{L}(x_t, u_t) \mid x_0 = x, u_t \sim \pi(x_t), x_t \sim P(x^+ | x_0, u_0) \right] \\
&= \sum_{u_0 \in \mathcal{U}} \pi(u_0 | x_0) \cdot \mathcal{L}(x_0, u_0) \Bigg\} = \sum_{u \in \mathcal{U}} \pi(u | x) \sum_{x^+ \in X} P(x^+ | x, u) \cdot \mathcal{L}(x, u) \\
&\quad + \sum_{u_0 \in \mathcal{U}} \pi(u_0 | x_0) \cdot \sum_{x^+ \in X} P(x^+ | x_0, u_0) \cdot \underbrace{E \left[\sum_{t=1}^{\infty} \gamma^t \cdot \mathcal{L}(x_t, u_t) \mid x_1 = x^+, u_t \sim \pi(x_t) \right]}_{\substack{= \gamma \cdot E \left[\sum_{\tau=0}^{\infty} \gamma^{\tau} \cdot \mathcal{L}(x_{\tau}, u_{\tau}) \mid x_0 = x^+, u_{\tau} \sim \pi(x_{\tau}) \right] \\ \tau = t-1 \Rightarrow (t=1 \Rightarrow \tau=0)}} \\
&\quad \rightarrow = \gamma \cdot v^{\pi}(x^+)
\end{aligned}$$

$$* \forall x \in X: v^{\pi}(x) = \sum_{u \in \mathcal{U}} \pi(u | x) \sum_{x^+ \in X} P(x^+ | x, u) \cdot \left(\mathcal{L}(x, u) + \gamma \cdot v^{\pi}(x^+) \right)$$

* NOTE: value $v^{\pi}: X \rightarrow \mathbb{R} \in \mathbb{R}^{|X|}$ appears linearly!

→ determine L^{π}, b^{π} so that $L^{\pi} \cdot v^{\pi} = b^{\pi}$ is the Bellman equation

→ see Python notebook provided after homework completed

⇒ the value of any policy can be computed by solving a linear equation