goal: derive Bellmon egyatian for MDP/SOCP

Neuro-Dynamic Programming

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chapter 2

Reinforcement Learning

An Introduction

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Chapter 3

· consider MDP/SOCP (X,U,P,c) with infinite-horizon exponentially-discounted cost min E[C(X,U)] where $C(X,U) = \sum_{t=0}^{\infty} \chi^{t} \cdot \chi(X_{t},U_{t})$ s.t. $\chi^{t} \sim P(X,U)$

o given policy $\pi: X \to \Delta(u)$, define value function $v^{\pi}: X \to \mathbb{R}$ $\forall x \in X: v^{\pi}(x) = \mathbb{E}\left[C(x_{l}u) \mid x_{o} = X, u_{t} \land T(x_{t})\right]$

→assuming [X1, 121 < 00, show v = satisfies Bellman equation

 $\forall x \in X : N^{T}(x) = \sum_{u \in \mathcal{U}} T(u|x) \sum_{x \neq \in X} P(x^{\dagger}|x,u) \cdot (\mathcal{L}(x,u) + \mathcal{V} \cdot V^{T}(x^{\dagger}))$

 $- v^{\pi}(x) = E\left(\sum_{t=0}^{\infty} r^{t} \cdot \mathcal{L}(x_{t}, u_{t}) \mid x_{o} = x, u_{t} \sim \pi(x_{t})\right)$

$$= E[Y^{\circ} : X(X_{\circ}, u_{\circ}) + \sum_{t=1}^{\infty} Y^{t} : X(X_{t}, u_{t})] \times_{\circ} = X, \ u_{t} \sim T(X_{t}), \ X_{t} \sim P(X^{t} | X_{\circ}, u_{\circ})]$$

$$= \sum_{u_{\circ} \in \mathcal{U}} (u_{\circ} | X_{\circ}) \cdot X(X_{\circ}, u_{\circ}) \} = \sum_{u_{\circ} \in \mathcal{U}} T(u | X) \sum_{x \in X} P(X^{t} | X_{\circ}, u) \cdot X(X_{\circ}, u_{\circ})$$

$$+ \sum_{u_{\circ} \in \mathcal{U}} T(u_{\circ} | X_{\circ}) \cdot \sum_{x \in X} P(X^{t} | X_{\circ}, u_{\circ}) \cdot E[\sum_{t=1}^{\infty} Y^{t} : X(X_{t}, u_{t})] \times_{t} = X^{t}, \ u_{t} \sim T(X_{t})]$$

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$$= Y \cdot Y \cdot X(X_{t}, u_{t}) \times_{t} = X^{t}, \ u_{t} \sim T(X_{t})$$

$$= Y \cdot Y \cdot X$$

=> the value of one policy can be computed by solving a linear exaction.