

III. Policy Iteration & Value Iteration Algos

So far, we have offline algos. We are also interested in online algos. Next we show the Bellman eqns provide fixed-point eqns for online learning

Consider discounted optimal control formulation

$$\text{minimize } \sum_{k=0}^{\infty} \gamma^k \cdot c(x_k, u_k) \quad \gamma \in [0, 1]$$

$$\text{subject to } : x_{k+1} = f(x_k, u_k)$$

Define $V_{\pi}(x)$ as the value fn corresponding to policy π \leftarrow may not be optimal.

Note: $V_{\pi}(x_k) = \sum_{z=k}^{\infty} \gamma^{z-k} \cdot c(x_z, u_z)$

① Policy Eval

② Policy Improv.

$$= c(x_k, u_k) + \underbrace{\gamma \cdot \sum_{z=k+1}^{\infty} \gamma^{z-(k+1)} c(x_z, u_z)}_{= \gamma \cdot V_{\pi}(x_{k+1})}$$

$$= \gamma \cdot V_{\pi}(x_{k+1})$$

$$V_{\pi}(x_k) = c(x_k, u_k) + \gamma \cdot V_{\pi}(x_{k+1}) \leftarrow$$

all where $u_k = \pi(x_k)$

Observation + Question: This eqn is implicit in $V_{\pi}(\cdot)$ and suggests iterative scheme . . .

$$V_{\pi}^{\hat{j}+1}(x_k) = c(x_k, u_k) + \gamma \cdot V_{\pi}^{\hat{j}}(x_{k+1}) ; \overset{\text{start}}{V_{\pi}^0}(x_k) = 0 \forall x_k$$

$\hat{j} = 0, 1, \dots$

Q: Does V_{π}^j converge as $j \rightarrow \infty$? A: YES!

Algo 1 (Iterative Policy Evaluation)

To compute the value fn corresponding to some arbitrary policy π :

For $j = 0, 1, \dots$

$$V_{\pi}^{j+1}(x_k) = c(x_k, u_k) + \gamma \cdot V_{\pi}^j(x_{k+1}) \quad \forall x_k \in X$$

where $u_k = \pi(x_k)$

$$V_{\pi}^0(x_k) = 0 \quad \forall x_k \in X$$

Sutton & Barto refer $V_{\pi}^j(x_k)$ as $j \rightarrow \infty$ as a "full backup"

2) Policy Improvement. To improve a given policy, an intuitive idea uses Bellman's Principle of Opt. Eqn:

$$\pi^{\text{NEW}} = \arg \min_{\pi(\cdot)} \left\{ c(x_k, \pi(x_k)) + \gamma \cdot \underbrace{V_{\pi^{\text{OLD}}}(x_{k+1})}_{\text{from policy eval}} \right\}$$

$$\text{where } x_{k+1} = f(x_k, \pi(x_k))$$

Bertsekas [1996] has prove π^{NEW} is improved wrt. π^{OLD} in the sense $V_{\pi^{\text{NEW}}}(x_k) \leq V_{\pi^{\text{OLD}}}(x_k)$

$$\forall x_k \in X$$

SUMMARY

Model-based
mean need to know
 $f(0,0)$

Policy Evolution

Given an arbitrary policy π

Find V_π

For $j=0,1,\dots$

$$V_\pi^{j+1}(x_k) = c(x_k, u_k) + \gamma V_\pi^j(x_{k+1})$$

$$V_\pi^0(x_k) = 0 \quad \forall x_k \in X$$

where $u_k = \pi(x_k)$, $x_{k+1} =$
 $f(x_k, u_k)$

Policy Improvement

Given $V_{\pi^{\text{old}}}$ for some arbitrary
policy π^{old} , find improved
policy
 π^{NEW}

$$\pi^{\text{NEW}} = \arg \min_{\pi(\cdot)} \left\{ c(x_k, \pi(x_k)) \right. \\ \left. + \gamma \cdot V_{\pi^{\text{old}}}(x_{k+1}) \right\}$$

where $x_{k+1} = f(x_k, \pi(x_k)) \quad \forall x_k \in X$