

On the use of non-stationary policies for stationary optimal control

1 Optimal control at horizon $H < \infty$

1.1 Problem

$$x_{t+1} = f_t(x_t, a_t, w_t), \quad t = 0, 1, \dots, H - 1$$

x_t : state, a_t : action, w_t : random (mutually independent)

$$\mathbb{P}(x_{t+1} = x' | x_t = x, a_t = a) = \mathbb{P}(x_{t+1} = x' | \mathcal{F}_t)$$

Performance measure:

$$v(x) = \mathbb{E} \left[\sum_{t=0}^{H-1} r_t(x_t, a_t, w_t) + R(x_H) | x_0 = x \right]$$

Goal: find a policy (control law) $a_t = \pi_t(x_t)$ that maximizes $v(x)$.

$$\pi = (\pi_0, \pi_1, \dots, \pi_{H-1})$$

Remarks: closed-loop control, Markov Decision Process

...EXAMPLE...

1.2 Algorithms

...PROOFS+NUMERICAL ILLUSTRATION...

$$v_{\pi,s} = T_{\pi,s} v_{\pi,s+1} \quad \text{with: } \forall v, x, [T_{\pi,s} v](x) = \underbrace{\mathbb{E} [r_s(x, \pi_s(x), w_s)]}_{r_s(x, \pi_s(x))} + \sum_y p_s(y|x, \pi_s(x)) v(y)$$

$$v_{*,s} = T_s v_{*,s+1} = \max_{\pi} T_{\pi,s} v_{*,s} \quad \text{with: } \forall v, x, [T_s v](x) = \max_a (r_s(x, a) + \sum_y p_s(y|x, a) v(y))$$

$$\pi_{*,s} = \arg \max_{\pi} T_{\pi,s} v_{*,s} = \mathcal{G}_s v_{*,s}$$

2 Stationary Optimal control at horizon $H = \infty$

2.1 Problem

$$x_{t+1} = f(x_t, a_t, w_t), \quad t = 0, 1, \dots,$$

w_t : i.i.d.

Performance measure:

$$v(x) = \mathbb{E} \left[\sum_{t=0}^{\infty} \gamma^t r(x_t, a_t, w_t) \mid x_0 = x \right]$$

$\gamma \in [0, 1[$: discount factor

...EXAMPLE...

2.2 Algorithms

...PROOFS...

$$v_\pi = T_\pi v_\pi \quad \text{with: } \forall v, x, [T_\pi v](x) = r(x, \pi(x)) + \sum_y p(y|x, \pi(x))v(y)$$

$$v_* = T v_* = \max_\pi T_\pi v_* \quad \text{with: } \forall v, x, [T v](x) = \max_a (r(x, a) + \sum_y p(y|x, a)v(y))$$

$$\pi_* = \arg \max_\pi T_\pi v_* = \mathcal{G} v_*$$

- T_π and T are γ -contraction mappings for $\|\cdot\|_\infty$:

$$\forall v, v', \quad \|T v - T v'\| \leq \gamma \|v - v'\|_\infty.$$

- $\pi_0^\infty = (\underbrace{\pi_0, \dots, \pi_{H-1}}_{\pi_0^{H-1}}, \dots)$

$$v_{\pi_0^H} = T_{\pi_0} T_{\pi_1} \dots T_{\pi_{H-1}} 0 \xrightarrow{H \rightarrow \infty} v_{\pi_0^\infty}$$

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$$v_* = \max_{\pi_0^\infty} v_{\pi_0^\infty} = \max_{\pi_0^\infty} \lim_{H \rightarrow \infty} v_{\pi_0^H} \stackrel{(*)}{=} \lim_{H \rightarrow \infty} \max_{\pi_0^H} v_{\pi_0^H} = \lim_{H \rightarrow \infty} T^H 0.$$

...NUMERICAL ILLUSTRATION...

Thm (Bellman, 1957): There exists an optimal policy that is stationary and

$$\pi \text{ optimal} \Leftrightarrow \pi \in \mathcal{G} v_* \Leftrightarrow T_\pi v_* = T v_* = v_*$$

Proof:

\Leftarrow : Assume $v_* = T_\pi v_*$. As $v_\pi = T_\pi v_\pi$ and T_π has a unique fixed point, then $v_\pi = v_*$, hence π is optimal.

\Rightarrow : Assume $v_\pi = v_*$. We have $T v_* = v_* = v_\pi = T_\pi v_\pi = T_\pi v_*$.

3 Large scale problems

3.1 Approx. Value Iteration

$$v_{k+1} \leftarrow \mathcal{A}T v_k = T v_k + \epsilon_k$$

...NUMERICAL ILLUSTRATION...

Thm: Singh & Yee (95) Gordon (95), Bertsekas & Tsitsiklis (95) : If $\|\epsilon_k\| \leq \epsilon$, then

$$\limsup_{k \rightarrow \infty} \|v_{\pi_*} - v_{\pi_k}\|_{\infty} \leq \frac{2\gamma}{(1-\gamma)^2} \epsilon.$$

3.2 The non-stationary trick

$$\begin{array}{cccccc} v_0 & v_1 & \dots & \pi_{k-\ell} & \dots & v_{k-1} \\ \pi_1 & \pi_1 & \dots & \pi_{k-\ell+1} & \dots & \pi_k \end{array}$$

$$\pi_{k,\ell} = (\pi_k, \pi_{k-1}, \dots, \pi_{k-\ell+1}, \pi_k, \dots)$$

Thm: Scherrer & Lesner (12): Under the same conditions,

$$\limsup_{k \rightarrow \infty} \|v_{\pi_*} - v_{\pi_k}\|_{\infty} \leq \frac{2\gamma}{(1-\gamma)(1-\gamma^{\ell})} \epsilon.$$

Interpretation: solving a ℓ periodic problem, less sensitive to perturbations.

Extensions: other algorithms (PI), 2 player-games (min max)