

9. Solution Sheet

1. Sample based policy iteration without bounded rewards

Let the second moments of the rewards given a policy $\pi \in \Pi_S$ exist, i.e.

$$\mathbb{E}_s^\pi [R_0^2] < \infty \quad \forall s \in \mathcal{S}.$$

Show that the Theorems 4.5.1 and 4.5.2 still apply to this policy, so that the one-step policy evaluation schemes from the lecture converge.

Solution:

The only place where we needed to assume bounded rewards in the proofs of the theorems was when showing $\sup_s \mathbb{E}[\varepsilon_s^2(n) | \mathcal{F}_n] \leq A + B\|V(n)\|_\infty^2$ and $\sup_{s,a} \mathbb{E}[\varepsilon_{s,a}^2(n) | \mathcal{F}_n] \leq A + B\|Q(n)\|_\infty^2$ respectively. With the assumption of existing second moments and defining

$$C := \sup_{s \in \mathcal{S}} \mathbb{E}_s^\pi [R_0^2] < \infty$$

we can proceed as in the lecture notes:

$$\begin{aligned} & \mathbb{E}[\varepsilon_s^2(n) | \mathcal{F}_n] \\ &= \mathbb{E}[(r_n + \gamma V_{s'_n}(n))^2 | \mathcal{F}_n] - 2\mathbb{E}_s^\pi [R_0 + \gamma V_{S_1}(n)] \mathbb{E}_s^\pi [r_n + \gamma V_{s'_n}(n) | \mathcal{F}_n] + (\mathbb{E}[R_0 + \gamma V_{S_1}(n)])^2 \\ &= \mathbb{E}_s^\pi [(R_0 + \gamma V_{S_1}(n))^2] - 2(\mathbb{E}_s^\pi [R_0 + \gamma V_{S_1}(n)])^2 + (\mathbb{E}_s^\pi [R_0 + \gamma V_{S_1}(n)])^2 \\ &\leq \mathbb{E}_s^\pi [(R_0 + \gamma V_{S_1}(n))^2] = \mathbb{E}_s^\pi [R_0^2] + 2\gamma \mathbb{E}_s^\pi [R_0 V_{S_1}(n)] + \gamma^2 \mathbb{E}_s^\pi [V_{S_1}(n)^2] \\ &\leq C^2 + 2\gamma C \|V(n)\|_\infty + \gamma^2 \|V(n)\|_\infty^2 \leq C^2 + 2\gamma C(1 + \|V(n)\|_\infty) + \gamma^2 \|V(n)\|_\infty^2 \\ &= (C^2 + 2\gamma C) + (2\gamma C + \gamma^2) \|V(n)\|_\infty \end{aligned}$$

The case for $Q(n)$ goes analogously save for the definition of C as the supremum over additionally all $a \in \mathcal{A}$ and the usage of the tower property with given $A_1 = a$ inside the conditional expectation.

2. Convergence theorem 4.3.8 under weaker assumptions

Show that the statement of Theorem 4.3.8 also holds if $\mathbb{E}[\varepsilon_i(n) | \mathcal{F}_n] \neq 0$ but instead satisfies

$$\sum_{n=1}^{\infty} \alpha_n(n) |\mathbb{E}[\varepsilon_i(n) | \mathcal{F}_n]| < \infty$$

almost surely for all coordinates $i = 1, \dots, d$. It is enough to prove an improved version of Lemma 4.4.4 where the condition $\mathbb{E}[\varepsilon_n | \mathcal{F}_n] = 0$ is replaced with

$$\sum_{n=1}^{\infty} \alpha_n |\mathbb{E}[\varepsilon_n | \mathcal{F}_n]| < \infty. \quad (1)$$

Apply the Robbins-Siegmund theorem to W^2 and use that $W \leq 1 + W^2$.

Solution:

$$\begin{aligned}
\mathbb{E}[W_{n+1}^2 \mid \mathcal{F}_n] &= \mathbb{E}[(1 - \alpha_n)^2 W_n^2 + \alpha_n^2 \varepsilon_n^2 + 2\alpha_n(1 - \alpha_n)W_n \varepsilon_n \mid \mathcal{F}_n] \\
&\leq (1 - 2\alpha_n + \alpha_n^2)W_n^2 + \alpha_n^2 D_n + 2\alpha_n(1 - \alpha_n)W_n \mathbb{E}[\varepsilon_n \mid \mathcal{F}_n] \\
&\leq (1 - 2\alpha_n + \alpha_n^2)W_n^2 + \alpha_n^2 D_n + 2\alpha_n(1 - \alpha_n)(1 + W_n^2) |\mathbb{E}[\varepsilon_n \mid \mathcal{F}_n]| \\
&\leq (1 - 2\alpha_n + \alpha_n^2 + 2\alpha_n |\mathbb{E}[\varepsilon_n \mid \mathcal{F}_n]| - \underbrace{2\alpha_n^2 |\mathbb{E}[\varepsilon_n \mid \mathcal{F}_n]|}_{\geq 0}) W_n^2 \\
&\quad + \alpha_n^2 D_n + 2\alpha_n |\mathbb{E}[\varepsilon_n \mid \mathcal{F}_n]| - \underbrace{2\alpha_n^2 |\mathbb{E}[\varepsilon_n \mid \mathcal{F}_n]|}_{\geq 0} \\
&\leq (1 - a_n + b_n)W_n^2 + c_n,
\end{aligned}$$

with $a_n = 2\alpha_n$, $b_n = \alpha_n^2 + 2\alpha_n |\mathbb{E}[\varepsilon_n \mid \mathcal{F}_n]|$, and $c_n = \alpha_n^2 D_n + 2\alpha_n |\mathbb{E}[\varepsilon_n \mid \mathcal{F}_n]|$. Now the claim follows from the Robbins-Siegmund Corollary 4.4.3.

3. Programming task: One-step policy evaluation on grid world

We want to use the grid world example to illustrate how to perform policy evaluation:

- a) Implement the grid world example from the lecture notes with target in the lower right corner and trap diagonally above or modify the code from the lecture's webpage.
- b) Implement the Algorithms 17 and 18, the one-step policy evaluation schemes for V^π and Q^π respectively, for the grid world example.
- c) Think about what you intuitively think the best policy π^+ and the worst policy π^- are for grid world and let additionally π be the policy that chooses the next action uniformly for all available options. Calculate $n = 1000$ steps of each policy evaluation scheme for π^+ , π^- , and π .
- d) Compare Algorithm 17 to Algorithm 7, the iterative policy evaluation. Which algorithm do you think performs better? Can we always apply both algorithms?

Solution:

See discussion in class and code.