

8. Excercise Sheet

## 1. Second version of Theorem 4.2.9 for SARSA

Show that the statement of Theorem 4.2.9 also holds if  $\mathbb{E}[\varepsilon_n | \mathcal{F}_n] \neq 0$  but instead satisfies

$$\sum_{n=1}^{\infty} \alpha_i(n) \left| \mathbb{E}[\varepsilon_i(n) \,|\, \mathcal{F}_n] \right| < \infty \tag{1}$$

almost surely. It is enough to prove an improved version of Lemma 4.2.5 where the condition  $\mathbb{E}[\varepsilon(t) \,|\, \mathcal{F}_t] = 0 \text{ is replaced with}$ 

$$\sum_{n=1}^{\infty} \alpha(t) \left| \mathbb{E}[\varepsilon(t) \,|\, \mathcal{F}_t] \right| < \infty.$$
<sup>(2)</sup>

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

## 2. *n*-step TD

a) Write pseudocode for *n*-step TD algorithms for evaluation of  $V^{\pi}$  and  $Q^{\pi}$  in the nonterminating case and prove the convergence by checking the *n*-step Bellman expectation equations

$$T^{\pi}V(s) = \mathbb{E}_{s}^{\pi} \Big[ R(s, A_{0}) + \sum_{t=1}^{n-1} \gamma^{t} R(S_{t}, A_{t}) + \gamma^{n} V(S_{n}) \Big]$$

and

$$T^{\pi}Q(s,a) = \mathbb{E}_{s}^{\pi^{a}} \Big[ R(s,a) + \sum_{t=1}^{n-1} \gamma^{t} R(S_{t},A_{t}) + \gamma^{n}Q(S_{n},A_{n}) \Big]$$

and the conditions of Theorem 4.2.9 on the error term. Note that the algorithm only starts to update after the MDP ran for n steps. Can you also write down a version in the terminating case?

b) Write pseudocode for an *n*-step SARSA control algorithm in the non-terminating case. Try to prove convergence in the same way we did for 1-step SARSA in Theorem 4.3.6.

## 3. (Truncated, Clipped) Double Q-Learning

In this task we deal with Double Q-Learning, analyzing in particular the differences with Q-Learning.

(a) Implement algorithms 24 (Double Q-learning (with behavior policy)), 25 (Truncated Double Q-learning) and 26 (Clipped double Q-learning (with behavior policy)) of the lecture.

**Reinforcement Learning** 



- (b) Apply the algorithms to the Markov decision process from the last exercise (Constructed Max Bias).
- (c) Compare the algorithms with the simple Q-learning algorithm. For example, apply similar metrics as in the lecture to the example (example 4.3.6).