Ryan Finn CSC 449 Nov. 30, 2022 Exam II



1. **procedure** SARSA( number of episodes  $N \in \mathbb{N}$ discount factor  $\lambda \in (0, 1]$ learning rate  $\alpha_n = 1 / \lg(n + 1)$ ) Initialize matrices Q(s, a) and n(s, a) to 0,  $\forall s, a$ **for** episode  $k \in \{1, 2, 3, ..., \frac{n}{n}N^{[1]}\}$  **do** *t* ← 1 Initialize s, Choose a, from a uniform distribution over the actions Choose  $a_t$  from  $s_t$  using  $\mu_t$ : an  $\epsilon$ -greedy policy with respect to  $Q^{[2]}$ **while** Episode *k* is not finished **do** Take action  $a_t$ : observe reward  $r_{t+1}$  and next state  $s_{t+1}$ Choose  $a_{t+1}$  from  $s_{t+1}$  using  $\mu_t$ : an  $\epsilon$ -greedy policy with respect to Qif The current state is terminal then  $y_t = 0$ else  $y_t = r_{t+1} + \max_a Q(s_{t+1}, a)$  $y_t = r_{t+1} + \lambda Q(s_{t+1}, a_{t+1})^{[3]}$ endif  $n(s_t, a_t) \leftarrow n(s_t, a_t) + 1$ Update Q function:  $Q(s_{t+4}, a_{t+4}) \leftarrow Q(s_t, a_t) - \alpha_{n(st-at)}(y_t - Q(s_t, a_t))$  $Q(s_t, a_t) \leftarrow Q(s_t, a_t) - \alpha_{n(st, at)} (y_t - Q(s_t, a_t))^{[4]}$  $t \leftarrow t + 1$ end while end for

[1]: A bit of a technicality, but N is actually defined as the number of episodes, not n which is a matrix.

[2]: This probably isn't a necessary change, since Q is already initialized to all 0, so  $a_1$  will just end up as the very first or last action in the action space. But, if Q is ever initialized as a non-zero matrix this change could be important. That's also how SARSA is defined in the book.

[3]:  $y_t = r_{t+1} + \max_a Q(s_{t+1}, a)$  is the target updater for a Q-Learning algorithm, not a SARSA algorithm.

[4]:  $Q(s_t, a_t)$  is what should be updated, not  $Q(s_{t+1}, a_{t+1})$ .

end procedure

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- a. Greedy deterministic
- b. So long as  $\epsilon$ -greedy is used with an  $\epsilon > 0$ , then yes, the Q values will converge as the number of time steps increases towards infinity. This is because every action must eventually be sampled infinite times, as time increases, for any positive  $\epsilon$ , as dictated by the Law of Large Numbers.