

Homework III

Deadline: 2026-01-15

1. (10 pts) Recall the definition of state visitation measure

$$d_{\mu}^{\pi}(s) = \mathbb{E}_{s_0 \sim \mu} [d_{s_0}^{\pi}(s)] = \mathbb{E}_{s_0 \sim \mu} \left[(1 - \gamma) \sum_{t=0}^{\infty} \gamma^t \mathbb{P}[s_t = s | s_0, \pi] \right],$$

where $(s_0, a_0, s_1, a_1, \dots)$ is trajectory starting from initial distribution μ and then following policy π . Let T obey the geometric distribution, i.e., $\mathbb{P}[T = t] = \gamma^t(1 - \gamma)$, $t = 0, 1, \dots$. Show that

$$\mathbb{P}[s_T = s] = d_{\mu}^{\pi}(s).$$

Then suggest a way to sample from d_{μ}^{π} .

2. (20 pts) Implement and test the Projected Policy Gradient method and the Softmax Policy Gradient method in Lecture 7 for the Gridworld problem in Homework I (Question 8, use $\gamma = 0.9$ and uniform distribution for μ). The action/advantage values and visitation measure in the policy gradient should be evaluated exactly based on the transition model. Display the convergence plots ($V^*(\mu) - V^k(\mu)$ vs # of iterations) of the two algorithms in a figure. Can you observe the finite iteration convergence of the Projected Policy Gradient method?
3. Consider the soft policy iteration algorithm in Lecture 8 (page 28).

- (10 pts) Show the policy improvement property of the algorithm:

$$V_{\lambda}^{\pi_{k+1}}(s) \geq V_{\lambda}^{\pi_k}(s), \quad \forall s.$$

- (10 pts) Show the γ -rate convergence of the algorithm:

$$\|V_{\lambda}^* - V_{\lambda}^{\pi_k}\|_{\infty} \leq \gamma^k \|V_{\lambda}^* - V_{\lambda}^{\pi_0}\|_{\infty}.$$

4. (10 pts) NPG under the entropy regularization admits the following form (same as the one given in Lecture 8 expect the constant):

$$\pi^+(a|s) \propto \pi(a|s) \exp \left(\frac{\eta}{\eta\tau + 1} A_{\tau}^{\pi}(s, a) \right).$$

Show that this is equivalent to

$$\pi^+(\cdot|s) = \arg \max_p \{ \eta \langle Q_{\tau}^{\pi}(s, \cdot), p \rangle + \tau H(p) \rangle - \text{KL}(p \| \pi(\cdot|s)) \},$$

where $H(p) = - \sum_a p_a \log p_a$.

5. (20 pts) Reproduce the figure on page 25 of Lecture 9 for comparing different bandit algorithms.