ECEN 743 | Homework 3

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Question 1

Let $x, y \in \mathbb{R}^n$. The triangle inequality states that $||x+y|| \le ||x|| + ||y||$. Use this to show that $||x-y|| \ge ||x|| - ||y||$.

Let u = x - y and v = y then using \triangle ·Inequality;

$$\begin{aligned} ||u+v|| &\leq ||u|| + ||v|| \\ ||(x-y)+y|| &\leq ||(x-y)|| + ||y|| \\ ||x|| &\leq ||x-y|| + ||y|| \end{aligned}$$

$$||x|| - ||y|| \le ||x - y||$$

Question 2

Consider an MDP with discount factor $\gamma \in (0, 1)$. Show that

$$\sup_{\pi} ||V_{\pi}||_{\infty} \le \frac{\max_{s,a} |r(s,a)|}{(1-\gamma)} \tag{1}$$

$$V_{\pi}(s) \doteq \mathbb{E}_{\pi}\left[\sum_{t=0}^{\infty} \gamma^{t} r(s, a) | S_{t} = s, A_{t} = a\right]$$

$$= \sum_{a} \pi(s, a) \sum_{s'} P(s'|s, a) \sum_{t=0}^{\infty} \gamma^{t} r(s, a)$$
(2)

We can take supremum of both sides which just equals to $\sup_{\pi} V_{\pi} = V^*$, after this we can see that the optimal value function is bounded by the case where maximum reward was gained at every step. To show this we can take $\max_{s,a}$ of both sides.

$$\sup_{\pi} \max_{s,a} V_{\pi} = ||V^*||_{\infty} \le \sum_{t=0}^{\infty} \gamma^t \max_{s,a} |r(s,a)|$$
 (3)

Since $\gamma \leq 1$ we can simply the sum operator using geometric series;

$$\sup_{\pi} ||V_{\pi}||_{\infty} \le \frac{\max_{s,a} |r(s,a)|}{(1-\gamma)} \tag{4}$$

Ouestion 3

Show that the Bellman operator T is a monotone operator, i.e., for any $V_1, V_2 \in \mathbb{R}^{|S|}$ with $V_1 \geq V_2$ (elementwise), $TV_1 \geq TV_2$.

Let's consider the Bellman optimality equation;

$$TV = R_{\pi} + \gamma \sum_{s' \in S} P_{\pi}(s|s')V(s')$$

$$\tag{5}$$

if $TV_1 \ge TV_2$ then $TV_1 - TV_2 \ge 0$;

$$R_{\pi} + \gamma \sum_{s' \in S} P_{\pi}(s|s') V_1(s') - R_{\pi} - \gamma \sum_{s' \in S} P_{\pi}(s|s') V_2(s') \ge 0$$

$$\gamma \sum_{s' \in S} P_{\pi}(s|s') (V_1(s') - V_2(s')) \ge 0$$
(6)

and since $V_1 \ge V_2$ and P_{π} is bounded by [0,1] (only positive values) this expression must hold true. Hence bellman operator is a monotone operator.

Question 4

Consider the function $f: \mathbb{R}^n \to \mathbb{R}^n, f(u) = Au$, where $A \in \mathbb{R}^n \times \mathbb{R}^n$. Assume that the row sums of A is strictly less than 1, i.e., $\sum_j |a_{ij}| \leq \alpha < 1$. Show that $f(\cdot)$ is a contraction mapping with respect to $||\cdot||_{\infty}$.

Let's remember the definition of a contraction;

$$||Au - Av||_{\infty} \le \alpha ||u - v||_{\infty} \tag{7}$$

where $0 \le \alpha < 1$. We can refactor the left side as Au - Av = A(u - v). Then recall that the supremum norm of a vector is simply the maximum absolute element of the vector. Then the left side can be expressed as;

$$\lfloor A(u-v)\rfloor_i = \sum_j a_{ij}(u_i - v_i) \tag{8}$$

Taking the absolute value of both sides we get the following inequality;

$$|\lfloor A(u-v)\rfloor_i| \le \sum_j |a_{ij}||(u_i-v_i)| \tag{9}$$

and since the the supremum norm is the maximum absolute element of a vector we can add the following;

$$|\lfloor A(u-v)\rfloor_i| \le ||A(u-v)||_{\infty} \le \sum_i |a_{ij}|||(u-v)||_{\infty}$$
 (10)

From the problem statement we know $\sum_{i} |a_{ij}| \le \alpha < 1$;

$$||A(u-v)||_{\infty} \le \alpha ||(u-v)||_{\infty} \tag{11}$$

Thusfort $f(\cdot)$ is a contraction mapping with respect to the supremum norm.

Question 5

Let U be a given set, and $g_1:U\to\mathbb{R}$ and $g_2:U\to\mathbb{R}$ be two real-valued functions on U. Also assume that both functions are bounded. Show that

$$\left| \max_{u} g_1(u) - \max_{u} g_2(u) \right| \le \max_{u} \left| g_1(u) - g_2(u) \right|$$
 (12)

Assume that $\max_u g_1(u) \ge \max_u g_2(u)$. Let $u^* = arg \max_u g_1(u)$;

$$|\max_{u} g_{1}(u) - \max_{u} g_{2}(u)| = g_{1}(u^{*}) - \max_{u} g_{2}(u)$$

$$\leq g_{1}(u^{*}) - g_{2}(u^{*})$$
(13)

Based on our initial assumption this value must be necessarily positive thus we rewrite this inequality without loss of generality;

$$|\max_{u} g_{1}(u) - \max_{u} g_{2}(u)| = g_{1}(u^{*}) - \max_{u} g_{2}(u)$$

$$\leq |g_{1}(u^{*}) - g_{2}(u^{*})|$$

$$\leq \max_{u} |g_{1}(u) - g_{2}(u)|$$
(14)

The same logic follows if we change our initial assumption to $\max_u g_2(u) \ge \max_u g_1(u)$. Thus the initial statement must be true;

$$|\max_{u} g_1(u) - \max_{u} g_2(u)| \le \max_{u} |g_1(u) - g_2(u)|$$
 (15)

Question 6

Consider the value iteration algorithm $V_{k+1} = TV_k$, with an arbitrary V_0 , where T is the Bellman operator.

(a) Show that, for n > m

$$||V_m - V_n||_{\infty} \le \frac{\gamma^m}{(1 - \gamma)} ||V_0 - V_1||_{\infty}$$
 (16)

(b) Let V^* be the optimal value function. Show that

$$||V_m - V^*||_{\infty} \le \frac{\gamma^m}{(1 - \gamma)} ||V_0 - V_1||_{\infty}$$
 (17)

(c) Show that

$$||V_m - V^*||_{\infty} \le \frac{\gamma}{(1 - \gamma)} ||V_{m-1} - V_m||_{\infty}$$
 (18)

(a)

$$||V_m - V_n||_{\infty} = ||T^m V_0 - T^n V_0||_{\infty}$$

$$\leq \gamma^n ||T^{m-n} x_0 - x_0||_{\infty}$$
(19)

We can expand the right side as a sum;

$$\leq \gamma^{m}[||T^{n-m}V_{0} - T^{n-m-1}V_{0}||_{\infty} + \dots + ||TV_{0} - V_{0}||_{\infty}]$$

$$\leq \gamma^{m}[\sum_{k=0}^{m-n-1} \gamma^{k}]||V_{1} - V_{0}||_{\infty}$$

$$\leq \gamma^{m}[\sum_{k=0}^{\infty} \gamma^{k}]||V_{1} - V_{0}||_{\infty}$$
(20)

Using geometric series we get;

$$||V_m - V_n||_{\infty} \le \frac{\gamma^m}{1 - \gamma} ||V_0 - V_1||_{\infty}$$
 (21)

(b) Optimal Value function is the unique fixed-point of the bellman equation according to the banach fixed point theorem.

$$V^* = TV^* = T^n V_0, \ n \ge 0 \tag{22}$$

Then based on (a), for n > m the following holds;

$$||V_m - V^*||_{\infty} = ||V_m - V_n||_{\infty} \le \frac{\gamma^m}{1 - \gamma} ||V_0 - V_1||_{\infty}$$
 (23)

(c)
$$||V_m - V_n||_{\infty} \le \frac{\gamma^m}{1 - \gamma} ||V_0 - V_1||_{\infty} = \frac{\gamma^m}{1 - \gamma} ||V_0 - TV_0||_{\infty}$$
 (24)

Let $n \to \infty$.

$$||V_m - V^*||_{\infty} = ||TV_{m-1} - TV^*||_{\infty} \le \gamma ||V_{m-1} - V^*||_{\infty}$$
 (25)

We can apply \triangle ·Inequality;

$$||V_m - V^*||_{\infty} \le \gamma [||V_{m-1} - V_m||_{\infty} + ||V_m - V^*||_{\infty}$$
 (26)

This can be re-arranged;

$$||V_m - V^*||_{\infty} \le \frac{\gamma}{1 - \gamma} ||V_{m-1} - V_m||_{\infty}$$
 (27)

Question 7

Let \overline{Q} be such that $||\overline{Q}-Q^*||_{\infty} \leq \epsilon$, where Q^* is the optimal Q-value function. Let $\overline{\pi}$ be the greedy policy with respect to \overline{Q} , i.e., $\overline{\pi}(s) = arg \, max_a \overline{Q}(s,a)$. Show that

$$||V^* - V_{\overline{\pi}}||_{\infty} \le \frac{2\epsilon}{1 - \gamma} \tag{28}$$

We can fix state s and let $a = \pi_Q(s)$. Then;

$$V^{*}(s) - V_{\overline{\pi}}(s) = Q^{*}(s, \pi^{*}(s)) - Q_{\overline{\pi}}(s, a)$$

$$= Q^{*}(s, \pi^{*}(s)) - Q^{*}(s, a) + Q^{*}(s, a) - Q_{\overline{\pi}}(s, a)$$

$$= Q^{*}(s, \pi^{*}(s)) - Q^{*}(s, a) + \gamma \mathbb{E}_{s'P(\cdot|s, a)}[V^{*}(s') - V_{\overline{\pi}}(s')]$$

$$\leq Q^{*}(s, \pi^{*}(s)) - \overline{Q}(s, \pi^{*}(s)) + \overline{Q}(s, a) - Q^{*}(s, a)$$

$$+ \gamma \mathbb{E}_{s'P(\cdot|s, a)}[V^{*}(s') - V_{\overline{\pi}}(s')]$$

$$\leq 2||\overline{Q} - Q^{*}||_{\infty} + \gamma ||V^{*} - V_{\overline{\pi}}||_{\infty}$$

$$(29)$$

We can take the supremum norm of both sides; and norm of a supremum norm is itself since the norm of a scalar is itself;

$$||V^*(s) - V_{\overline{\pi}}(s)||_{\infty} \le ||2||\overline{Q} - Q^*||_{\infty} + \gamma ||V^* - V_{\overline{\pi}}||_{\infty}||_{\infty}$$

$$= 2||\overline{Q} - Q^*||_{\infty} + \gamma ||V^* - V_{\overline{\pi}}||_{\infty}$$
(30)

We can use the fact that $||\overline{Q} - Q^*||_{\infty} \le \epsilon$;

$$(1 - \gamma)||V^*(s) - V_{\overline{\pi}}(s)||_{\infty} \le 2||\overline{Q} - Q^*||_{\infty} \le 2\epsilon \tag{31}$$

We can rearrange to get the inequality in the problem statement;

$$||V^* - V_{\overline{\pi}}||_{\infty} \le \frac{2\epsilon}{1 - \gamma} \tag{32}$$