

ECEN 689 Assignment 1

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For any $V_1, V_2 \in \mathbb{R}^{|X|}$ with $V_1 \geq V_2$, then for any state, we have

$$\begin{aligned} TV_2 - TV_1 &= \max(R(x, a) + \gamma \sum_y P(y|x, a) V_2(y)) - \max(R(x, a) + \gamma \sum_y P(y|x, a) V_1(y)) \\ &= \max(\gamma \sum_y P(y|x, a) V_2(y)) - \max(\gamma \sum_y P(y|x, a) V_1(y)) \\ &\leq \max(\gamma \sum_y P(y|x, a) V_2(y)) - \max(\gamma \sum_y P(y|x, a) V_2(y)) \quad (\text{because } V_1 \geq V_2) \\ &= 0 \end{aligned}$$

then $TV_2 - TV_1 \leq 0$ bellman operator is monotone operator

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$\|f(u) - f(v)\|_\infty = \|Au - Av\|_\infty = \|u - v\|_\infty \|A\|_\infty$, where the row sums of A is strictly less than 1, and $\sum_j |a_{ij}| \leq \alpha \leq 1$

, then $0 \leq \|u - v\|_\infty \|A\|_\infty = \|u - v\|_\infty \max_{1 \leq i \leq n} \sum_{j=1}^n |a_{ij}| \leq \alpha \|u - v\|_\infty \leq \|u - v\|_\infty$

which means $\|f(u) - f(v)\|_\infty \leq \alpha \|u - v\|_\infty$ for an $\alpha \in (0, 1)$ and for all $u, v \in U$ the $f(\cdot)$ is a contraction mapping

3

Assume $v = \operatorname{argmax}_u g_1(u)$, where $g_1 : U \rightarrow \mathbb{R}; g_2 : U \rightarrow \mathbb{R}$ then

$|\max_u g_1(u) - \max_u g_2(u)| = g_1(v) - \max_u g_2(u)$ if $\max_u g_1(u) \geq \max_u g_2(u)$

then for the RHS, we have

$g_1(v) - \max_u g_2(u) \leq g_1(v) - g_2(v)$, (since $\max_u g_2(u) \geq g_2(v)$)

where the RHS is equal to $|g_1(v) - g_2(v)|$

then, $|\max_u g_1(u) - \max_u g_2(u)| \leq |g_1(v) - g_2(v)| \leq \max_u |g_1(u) - g_2(u)|$

We can prove this similarly if $\max_u g_1(u) \leq \max_u g_2(u)$, then

$$|\max_u g_1(u) - \max_u g_2(u)| \leq \max_u |g_1(u) - g_2(u)|$$

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a)

we have V^* as optimal value function,

$$\|V^*(x)\|_\infty = \|\max_\pi V_\pi(x)\|_\infty$$

from the linear system of V_π , the RHS

$$= \|\max_\pi \{(I - \gamma P_\pi)^{-1} R_\pi\}\|_\infty \leq \|(1 - \gamma)^{-1} \max_\pi \{R_\pi\}\|_\infty = \max_x \{(1 - \gamma)^{-1} \max_\pi \{R_\pi\}\} = R_{\max}/(1 - \gamma)$$

then

$$\|V^*(x)\|_\infty \leq R_{\max}/(1 - \gamma)$$

b)

for the last inequality in b) , we have

$$k \geq \frac{\log(R_{\max}) - \log(\varepsilon(1 - \gamma))}{\log 1 - \log \gamma}$$

$$k \leq \frac{\log(\varepsilon(1 - \gamma)) - \log(R_{\max})}{\log \gamma}$$

$$k \leq \frac{\log(\frac{\varepsilon(1 - \gamma)}{\log(R_{\max})})}{\log \gamma}$$

$$\log \gamma k \leq \log(\frac{\varepsilon(1 - \gamma)}{R_{\max}})$$

for both sides of log function, we have

$$\gamma^k \leq \frac{\varepsilon(1 - \gamma)}{R_{\max}}$$

then we have

$$\gamma^k R_{\max} \leq \varepsilon(1 - \gamma)$$

we put this inequality above for later usage, then by the contraction property of bellman operator, which is

$$\|TV_1 - TV_2\|_\infty \leq \gamma \|V_1 - V_2\|_\infty$$

we take LHS of the first inequality in b)

$$\|V_k - V^*\|_\infty \leq \gamma \|V_{k-1} - V^*\|_\infty \leq \dots \leq \gamma^k \|V_0 - V^*\|_\infty$$

where $V_0 = 0$

$$\gamma^k \|V_0 - V^*\|_\infty = \gamma^k \|V^*\|_\infty$$

we have $\|V^*(x)\|_\infty \leq R_{\max}/(1 - \gamma)$ from a)

$$\gamma^k \|V^*\|_\infty \leq \gamma^k (R_{\max}/(1 - \gamma)) \leq \varepsilon$$

then

$$\|TV_1 - TV_2\|_\infty \leq \varepsilon$$

c)

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

then,

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

$$\|V_m - V^*\|_\infty \leq (\|V_{m+1} - V_m\|_\infty)/(1 - \gamma)$$

from the inequality of $\|V_{m+1} - V_m\|_\infty$ in c)

$$\|V_{m+1} - V_m\|_\infty/(1 - \gamma) \leq \varepsilon/2\gamma$$

$$\gamma \|V_m - V^*\|_\infty \leq \varepsilon/2$$

from the contraction property

$$\|V_{m+1} - V^*\|_\infty \leq \gamma \|V_{m+1} - V^*\|_\infty \leq \varepsilon/2$$

Then for the value iteration that reach the maximum of the



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From the definition of infinity norm, we have

$$\begin{aligned} \|V_1^* - V_2^*\|_\infty &= \max_x |V_1^* - V_2^*| \\ &= \max_x |\max_a (R(x, a) + \gamma \sum_y P_1(y|x, a) V_1^*(y)) - \max_a (R(x, a) + \gamma \sum_y P_2(y|x, a) V_2^*(y))| \end{aligned}$$

From the definition of question 3), we have

$$\begin{aligned}
RHS &\leq \max_x(\max_a |R(x, a) + \gamma \sum_y P_1(y|x, a) V_1^*(y) - R(x, a) - \gamma \sum_y P_2(y|x, a) V_2^*(y)|) \\
&\leq \gamma \max_x(\max_a |\sum_y P_1(y|x, a) V_1^*(y) - \sum_y P_2(y|x, a) V_2^*(y)|) \\
&= \gamma \max_x(\max_a |\sum_y P_1(y|x, a) V_1^*(y) - \sum_y P_2(y|x, a) V_1^*(y) - (\sum_y P_2(y|x, a) V_2^*(y) - \sum_y P_2(y|x, a) V_1^*(y))|)
\end{aligned}$$

from the contraction property

$$\begin{aligned}
&\leq \max_x(\max_a |\sum_y P_1(y|x, a) - \sum_y P_2(y|x, a)| \cdot \|V_1^*(y)\|_\infty + \gamma \|V_2^* - V_1^*\|_\infty) \\
&\leq \gamma \varepsilon R_{max} + \gamma \|V_2^* - V_1^*\|_\infty
\end{aligned}$$

then we have,

$$\|V_2^* - V_1^*\|_\infty \leq (\gamma \varepsilon R_{max}) / (1 - \gamma)^2$$

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From the definition of infity norm

$$\|V^* - V_{\bar{\pi}}\|_\infty = \max_x |V^*(x) - V_{\bar{\pi}}(x)|$$

by definition of optimal q value function, $V^*(x) = \max_a Q^*(x, a)$, and the definition of q value function of a policy π , we have

$$\begin{aligned}
\|\bar{V} - V^*\|_\infty &= \max_x |\max_a \bar{Q}(x, a) - \max_a Q^*(x, a)| \\
&\leq \max_x \max_a |\bar{Q}(x, a) - Q^*(x, a)| \\
&= \|\bar{Q} - Q^*\| \leq \varepsilon
\end{aligned}$$

also by triangle inequality, we have

$$\|V^* - V_{\bar{\pi}}\|_\infty \leq \|V^* - \bar{V}\|_\infty + \|\bar{V} - V_{\bar{\pi}}\|_\infty$$

we conclude $V_{\bar{\pi}} = R(x, \bar{\pi}(x)) + \gamma \sum_y P(y|x, \bar{\pi}(x)) V_{\bar{\pi}}(y) = \max_a (R(x, a) + \gamma \sum_y P(y|x, a) V^*(y) - \gamma \varepsilon \sum_y P(y|x, a)) + \gamma \sum_y P(y|x, \bar{\pi}(x)) V_{\bar{\pi}}(y) - \gamma \sum_y P(y|x, \bar{\pi}(x)) V^*(y) - \gamma \varepsilon \sum_y P(y|x, \bar{\pi}(x)) = V^* + \gamma \sum_y P(y|x, \bar{\pi}) V_{\bar{\pi}}(y) - \gamma \sum_y P(y|x, \bar{\pi}) V^*(y) - 2\gamma \varepsilon$ (the middle equation is too long to put in one line)

We have $V^* \leq V_{\bar{\pi}}$, then

$$\begin{aligned}
V^* - V_{\bar{\pi}} &\leq \gamma \|V^* - V_{\bar{\pi}}\|_\infty + 2\varepsilon \\
\|V^* - V_{\bar{\pi}}\|_\infty &\leq 2\varepsilon / (1 - \gamma)
\end{aligned}$$

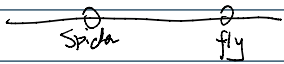
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consider a MDP for this question $M = (X, A, P, R, \gamma)$, where X defined as the states of distance between the spider and fly, A defined as action of the spider, and P defined the transition probability function from one state to another state, R is a reward function and γ is the discount factor.

From the definition of probability that spider's action for each state, for the transition probability $p_x = P(y|x, a)$, $x, y \in X$, $a \in A$ with certain states.

I put the rest of the possible solution by handwriting work in the next pages, cuz it will save some time, please check

(I hand write the question after got some hint from class,
but run out of time, so I can not put it in latex)



consider one unit away from the spider and fly will move to same direction
which is $x_1 \rightarrow x_0$

$P\{\text{the spider will move}\} = 2p$, since the spider will move left or right, that is $p+p$

$P\{\text{the spider will stay}\} = 1-2p$

consider spider will catch the fly, which is $x_1 \rightarrow x_0$

$P\{\text{spider will move}\} = 1-2p$

$P\{\text{spider will stay}\} = p$

consider if the spider and fly will move to two unit, which is spider will stay

$P\{\text{spider will stay}\} = p$

So the bellman optimization function is

$$V^*(x) = 1 + pV^*(x) + (1-2p)V^*(x-1) + pV^*(x-2), \text{ for } x \geq 2$$

but for the special cases $V^*(1)$, consider the above statement

$$V^*(1) = 1 + \min\{2pV^*(1), pV^*(0) + (1-2p)V^*(1)\}$$

$$V^*(0) = 1 + pV^*(0) + (1-2p)V^*(1)$$

\Rightarrow

$$V^*(1) = 1 + \min\{2p \cdot V^*(1), (p + p(1-2p)V^*(1) + (1-p)(1-2p)V^*(0)) / (1-p)\}$$

consider the one unit away bellman function of $V^*(0)$

$$V^*(1) = 1 + 2pV^*(0) \quad \text{if} \quad 2pV^*(0) < pV^*(2) + (1-2p)V^*(1) \quad \Rightarrow V^*(0) = 1/(1-2p)$$

$$\Rightarrow 2pV^*(0) < (p + (1-2p)V^*(0))(1-p) \quad \Rightarrow 2p(1-p) < (1-p)(1-2p) \quad \Rightarrow 2p < 1-2p \quad \Rightarrow p < 1/3$$

$$V^*(0) = pV^*(0) + (1-2p)V^*(1) \quad \text{if} \quad 2pV^*(0) > pV^*(2) + (1-2p)V^*(1) \quad \Rightarrow V^*(0) = 1/p$$

$$\Rightarrow 2pV^*(0) > p/(1-p) + (1-2p)V^*(0)/(1-p) \quad \Rightarrow (1-2p)V^*(0) > 2(1-p)^2$$

$$\Rightarrow p > 1/3$$

$$\Rightarrow \begin{cases} V^*(0) = 1/(1-2p) & \text{if } p \leq 1/3 \\ V^*(0) = 1/p & \text{if } p \geq 1/3 \end{cases}$$

then for the two unit away from the spider to catch the fly

$$\text{we have } V^*(2) = \frac{1}{1-p} (4V^*(1) + 2pV^*(1))$$

$$= 1 + \min\{2p \cdot V^*(1), p + p(1-2p)V^*(1) + (1-p)(1-2p)V^*(0) / (1-p)\}$$

we have

$$V^*(p) = \frac{1}{1-p} (1-p V^*(1) + 2p V^*(1))$$

$$= \frac{1}{1-p} (1 + (1-2p) (1 + \min\{2p^* V^*(1), p + p(1-p) V^*(1) + (1-p)(1-2p) V^*(1)\} / (1-p)))$$

similar to the closed form of $V^*(1)$

$$V^*(p) = \begin{cases} \frac{1}{p} & \text{if } p \geq \frac{1}{3} \\ 2/(1-p) & \text{if } p \leq \frac{1}{3} \end{cases}$$