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Reinforcement Learning

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Please answer: Exercise 4.3, 4.5, 4.8

Programming Write a program to solve Jack's car rental problem. Reproduce plots shown in Fig. 4.2 (page 81). Write a program to solve Gambler's problem. Reproduce plots shown in Fig. 4.3 (page 84).

Exercise 4.3 What are the equations analogous to (4.3), (4.4), and (4.5) for the action-value function  $q_{\pi}$  and its successive approximation by a sequence of functions  $q_0, q_1, q_2, \ldots$ ?

$$v_{\pi}(s) \doteq \mathbb{E}_{\pi}[G_{t} \mid S_{t} = s]$$

$$= \mathbb{E}_{\pi}[R_{t+1} + \gamma G_{t+1} \mid S_{t} = s] \qquad (from (3.9))$$

$$= \mathbb{E}_{\pi}[R_{t+1} + \gamma v_{\pi}(S_{t+1}) \mid S_{t} = s] \qquad (4.3)$$

$$= \sum_{a} \pi(a|s) \sum_{s',r} p(s',r|s,a) \Big[ r + \gamma v_{\pi}(s') \Big], \qquad (4.4)$$

$$q_{\pi}(s,a) = \mathbb{E}[G_{t} \mid G_{t} = s, A_{t} = a] = \mathbb{E}_{\pi}[R_{t+1} + \gamma V_{\pi}(S_{t+1}) \mid S_{t} = s, A_{t} = a]$$

$$= \mathbb{E}_{\pi}[R_{t+1} + \gamma q_{\pi}(S_{t+1}, A_{t+1}) \mid S_{t} = s, A_{t} = a]$$

$$= \mathbb{E}_{\pi}[R_{t+1} + \gamma \sum_{a'} \pi(a'|s') q_{\pi}(s',a') \mid S_{t} = s, A_{t} = a]$$

$$= \sum_{s',r} p(s',r|s,a) [r + \gamma \sum_{a'} q_{\pi}(s',a') \pi(a'|s')]$$

$$v_{k+1}(s) \doteq \mathbb{E}_{\pi}[R_{t+1} + \gamma v_k(S_{t+1}) \mid S_t = s]$$

$$= \sum_{a} \pi(a|s) \sum_{s',r} p(s',r|s,a) \Big[ r + \gamma v_k(s') \Big],$$

$$q_{k+1}(s,a) = \sum_{s',r} p(s',r|s,a) [r + \gamma \sum_{a'} q_k(s',a') \pi(a'|s')]$$
(4.5)

Exercise 4.5 How would policy iteration be defined for action values? Give a complete algorithm for computing  $q_*$ , analogous to that on page 80 for computing  $v_*$ . Please pay special attention to this exercise, because the ideas involved will be used throughout the rest of the book.

## Algorithm:

- 1. Inputs:  $q_*(s, a) = 0$ , for all  $s \in S$  and  $a \in A$
- 2. the policy evatuation step

Repeat

$$\begin{array}{l} \Delta \leftarrow 0; \\ \text{For each } (s,a) \in (S \times A) \text{ do} \\ q \leftarrow q_*(s,a); \\ q_*(s,a) \leftarrow \sum_{s',r} p(s',r|s,a)[r + \gamma \sum_{a'} (s',a')\pi(a'|s')]; \\ \Delta \leftarrow \max \left(\Delta, |q - q_*(s,a)|\right) \end{array}$$

Until  $\Delta < \theta$  (small positive number);

3. To improve policy

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Policy-stable \leftarrow true. For each s \in S do  \text{Old-action} \leftarrow \pi(s) \\ \pi(s) \leftarrow \underset{a}{\arg\max} \ q_*(s,a); \\ \text{If old-action} \neq \pi(s), then \ policy - state \ \leftarrow false.  If policy-stable, then stop and return q \approx q_*, and \ \pi \approx \pi_*; Else go to 2.
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Exercise 4.8 Why does the optimal policy for the gambler's problem have such a curious form? In particular, for capital of 50 it bets it all on one flip, but for capital of 51 it does not. Why is this a good policy?

At step 50, it doesn't matter if you lose 1 or lose 49, so you might as well go for as much as possible to maximize the potential value. on the other hand, on step 51 you could lose 1 then get back to step 50 and try again.