Reinforcement Learning Lab Exercise 3

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Problem Set 2

Description

From [3], study example 4.1 and write a Python code to find $v\pi(s)$.

Solution

| | 1 | 2 | 3 |
|----|----|----|----|
| 4 | 5 | 6 | 7 |
| 8 | 9 | 10 | 11 |
| 12 | 13 | 14 | |

Python Code

```
import numpy as np

def in_bounds(row, col):
    return 0 <= row < 4 and 0 <= col < 4

def state_to_rowcol(s):
    s0 = s - 1
    return divmod(s0, 4)</pre>
```

```
def rowcol_to_state(row, col):
       return row * 4 + col + 1
  def next_state(s, action):
       if s in [1, 16]:
           return s
6
       row, col = state_to_rowcol(s)
       if action == 'U':
           new_row, new_col = row - 1, col
       elif action == 'D':
12
           new_row, new_col = row + 1, col
13
       elif action == 'L':
14
           new_row, new_col = row, col - 1
       elif action == 'R':
           new_row, new_col = row, col + 1
17
       else:
           raise ValueError("Unknown action!")
19
20
       if in_bounds(new_row, new_col):
21
           return rowcol_to_state(new_row, new_col)
22
       else:
23
           return s
24
   def iterative_policy_evaluation(theta=1e-8, max_iterations=100000):
26
       V = np.zeros(17)
27
       gamma = 1.0
28
       actions = ['U', 'D', 'L', 'R']
30
       for _ in range(max_iterations):
31
           delta = 0
           newV = V.copy()
33
34
           for s in range(1, 17):
35
                if s in [1, 16]:
36
                    continue
38
               v_new = 0.0
39
                for a in actions:
40
                    s_next = next_state(s, a)
                    reward = -1
42
                    v_{new} += 0.25 * (reward + gamma * V[s_next])
43
               newV[s] = v_new
44
45
           delta = np.max(np.abs(newV - V))
46
```

```
V = newV
           if delta < theta:</pre>
               break
       return V
6
  if __name__ == "__main__":
       V_final = iterative_policy_evaluation()
8
       print("Value function with terminal states at (0,0) and (3,3):\n
9
          ")
10
       for row in range (4):
11
           row_str = []
12
           for col in range (4):
13
               s = row * 4 + col + 1
14
               row_str.append(f"{V_final[s]:6.2f}")
           print(" ".join(row_str))
```

Solution

Value function with terminal s

```
0.00 -14.00 -20.00 -22.00
-14.00 -18.00 -20.00 -20.00
-20.00 -20.00 -18.00 -14.00
-22.00 -20.00 -14.00 0.00
```

Problem Set 3

Description

From [3], solve exercises 4.1 and 4.2. You can choose to solve it using a Python code.

Solution

Exercise 4.1

We know that,

$$q_{\pi}(s, a) = \sum_{s'} P(s'|s, a) \left[R(s, a, s') + \gamma v_{\pi}(s') \right]$$

But for deterministic actions,

$$q_{\pi}(s, a) = R(s, a, s') + \gamma v_{\pi}(s')$$

Using the above two equations, we get:

(a)
$$q_{\pi}(11, \text{down}) = -1 + 1 \cdot (0) = -1$$

(b)
$$q_{\pi}(7, \text{down}) = -1 + 1 \cdot (-14) = -15$$

Exercise 4.2

Part 1: Original Dynamics

The value function for state 15 is:

$$v_{\pi}(15) = -1 + \frac{1}{4} \left(v_{\pi}(12) + v_{\pi}(13) + v_{\pi}(14) + v_{\pi}(15) \right)$$

Part 2: Changed Dynamics

If action down from state 13 leads to 15, the Bellman equations for $v_{\pi}(13)$ and $v_{\pi}(15)$ are:

$$v_{\pi}(13) = -1 + \frac{1}{4}(v_{\pi}(12) + v_{\pi}(13) + v_{\pi}(14) + v_{\pi}(15))$$

$$v_{\pi}(15) = -1 + \frac{1}{4}(v_{\pi}(12) + v_{\pi}(13) + v_{\pi}(14) + v_{\pi}(15))$$

Rearrange both equations:

$$3v_{\pi}(15) = -4 + v_{\pi}(12) + v_{\pi}(13) + v_{\pi}(14)$$

$$3v_{\pi}(13) = -4 + v_{\pi}(12) + v_{\pi}(14) + v_{\pi}(15)$$

Substitute $v_{\pi}(15)$ into the equation for $v_{\pi}(13)$:

$$v_{\pi}(13) = \frac{-4 + v_{\pi}(12) + v_{\pi}(14) + \frac{-4 + v_{\pi}(12) + v_{\pi}(13) + v_{\pi}(14)}{3}}{3}$$

Simplifying gives:

$$v_{\pi}(13) = 0.5v_{\pi}(12) + 0.5v_{\pi}(14) - 2$$

By symmetry, $v_{\pi}(15)$ is:

$$v_{\pi}(15) = 0.5v_{\pi}(12) + 0.5v_{\pi}(14) - 2 = 0.5(-22) + 0.5(-14) - 2 = -20$$

Problem Set 4

Description

From [3], solve exercise 4.3.

Solution

Analogous to Equation (4.3):

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

Analogous to Equation (4.4):

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

Analogous to Equation (4.5):

The iterative update rule for successive approximations of q_{π} is:

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