

MONDAY

6

MAY
2019

Question 1

s	a	s'	r	$r(s, a, s')$	$p(s', \tau s, a)$	$p(s', \tau s, a)$
high	search	high	10	r_s	α	$+0.1 \cdot \alpha \cdot r_s$
high	search	low	-10	r_s	$1-\alpha$	$-0.1(1-\alpha)r_s$
low	search	high	0	-3	$1-\beta$	0
low	search	low	-10	r_s	β	$-0.1\beta r_s$
high	wait	high	10	r_w	1	$0.1 r_w$
high	wait	low	10	r_w	0	-
low	wait	high	0	-	0	-
low	wait	low	10	r_w	1	$0.1 r_w$
low	rech.	high	0	0	1	-
low	rech.	low	0	-	0	-

Since $r(s, a, s') = \frac{r(s, a, s') \cdot p(s', \tau | s, a)}{p(s' | s, a)}$

(1) $r_s = \frac{10 \cdot p(s', \tau | s, a)}{p(\text{high} | \text{high, search})} \rightarrow \alpha$

$p(s', \tau | s, a) = \alpha \cdot 0.1 \cdot r_s$

(2) $r_s = \frac{-10 \cdot p(s', \tau | s, a)}{(1-\alpha)}$

$p(\text{low, -10} | \text{high, search}) = -0.1 \cdot (1-\alpha) r_s$

Question 2 -

Given the gridworld environment,

WEDNESDAY

1

MAY
2019

Solving the Bellman equation, $v_{\pi}(s)$:

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

Solving for state 1 for illustration -

$$v_{\pi}(1) = 0.25 \times 1 [-1 + \gamma v_{\pi}(1)] + 0.25 \times 1 [-1 + \gamma v_{\pi}(1)] + 0.25 \times 1 [0 + \gamma v_{\pi}(2)] + 0.25 \times 1 [0 + \gamma v_{\pi}(4)]$$

Similarly $v_{\pi}(2) = 0.25 \times 1 [-1 + \gamma v_{\pi}(2)] + 0.25 \times 1 [0 + \gamma v_{\pi}(1)] + 0.25 \times 1 [0 + \gamma v_{\pi}(3)] + 0.25 \times 1 [0 + \gamma v_{\pi}(4)]$

Thus $\begin{bmatrix} v_{\pi}(1) \\ v_{\pi}(2) \\ \vdots \\ v_{\pi}(25) \end{bmatrix}_{25 \times 1} = \begin{bmatrix} R_1 \\ R_2 \\ \vdots \\ R_{25} \end{bmatrix}_{25 \times 1} + \gamma \underbrace{P_{\pi}}_{25 \times 25} \underbrace{v_{\pi}}_{25 \times 1}$

Total no. of states

$$\therefore v_{\pi} = R_{\pi} + \gamma P_{\pi} v_{\pi}$$

$$v_{\pi} - \gamma P_{\pi} v_{\pi} = R_{\pi}$$

$$v_{\pi} = (I - \gamma P_{\pi})^{-1} R_{\pi}$$

USES.
Appointment

Notes

Question 3

2

MAY
2019

Exercise 3.15

$$V_{\pi}(s) = E_{\pi} [R_t | S_t = s]$$

$$= E_{\pi} \left[\sum_{k=0}^{\infty} \gamma^k R_{t+k+1} | S_t = s \right]$$

Adding a constant to each reward -

let $\hat{R}_{t+k+1} = R_{t+k+1} + c$

$$\hat{V}_{\pi}(s) = E_{\pi} [\gamma^k \hat{R}_{t+k+1} | S_t = s]$$

$$= E_{\pi} [\gamma^k R_{t+k+1} | S_t = s] + E_{\pi} [\gamma^k c | S_t = s]$$

$$= V_{\pi}(s) + \frac{c}{1-\gamma}$$

Thus we see the best policy doesn't change with addition of constant c at each reward.

Exercise 3.16

Adding a constants at episodic tasks:

$$V_{\pi}(s) = E [R_{t+1} + \gamma R_{t+2} + \gamma^2 R_{t+3} + \dots + \gamma^{k-1} R_{t+k}]$$

$$\hat{V}_{\pi}(s) = E [(R_{t+1} + c) + \gamma (R_{t+2} + c) + \dots + \gamma^{k-1} (R_{t+k} + c)]$$

$$= c [1 + \gamma + \gamma^2 + \dots + \gamma^{k-1}] + V_{\pi}(s)$$

$$= c \frac{(1 - \gamma^k)}{1 - \gamma} + V_{\pi}(s)$$

$k \rightarrow$ small value, (A) is closer to terminal state, thus adding a constant only adds ~~adds~~ ^{prolongs} reduces the steps to terminal states.

Notes

Appointment

Question 4

Solving the Bellman optimality eqⁿ

FRIDAY

3

MAY
2019

$$v^*(s) = \max_a \sum_{s'} \left(R(s, a) + \gamma v^*(s') \right) p(s'|s, a)$$

Thus,

$$v^*(s) \geq R(s) + \gamma \max_{a \in A} \sum_{s'} p(s'|s, a) v(s')$$

thus taking |A| as the linear constraints;

$$v(s) \geq R(s) + \gamma \sum_{s' \in S} p(s'|s, a) v(s') \quad \forall a \in A$$

Thus, minimise $\sum_s v(s)$

subject to $v(s) \geq R(s) + \gamma \sum_{s' \in S} p(s'|s, a) v(s') \quad \forall a \in A$

Solving for optimal policy,

if we optimise,

$$\min_v \sum_s d(s) v(s)$$

subject to $v(s) \geq R(s) + \gamma \sum_{s' \in S} p(s'|s, a) v(s') \quad \forall a \in A, s \in S$

Here $d(s)$ is the distribution over states.

Adding $\mu(s, a)$ for each constraint,

$$\max_{\mu(s, a)} \sum_{s \in S} R(s) \sum_{a \in A} \mu(s, a)$$

Subject to $\sum_{a \in A} \mu(s', a) = d(s') + \sum_{s \in S} \sum_{a \in A} p(s'|s, a) \mu(s, a) \quad \forall s' \in S$

where $\mu(s, a) = \sum_{t=0}^{\infty} \gamma^t P(s_t = s, A_t = a)$

Thus $\lambda^*(s) = \max_{a \in A} \mu(s, a)$

Notes

Appointment

SATURDAY

4

MAY
2019

Question 5

$$V_{\pi}(s) = E_{\pi} [R_{t+1} + \gamma V_{\pi}(S_{t+1}) | S_t = s]$$

$$V_{*}(s) = E_{\pi^{*}} [R_{t+1} + \gamma V_{*}(S_{t+1}) | S_t = s]$$

$$= \max_a E [R_{t+1} + \gamma V_{*}(S_{t+1}) | S_t = s]$$

$$= \max_a \sum_{s'} \sum_{a'} (r + \gamma V_{*}(s')) p(s', a' | s, a)$$

Thus,

$$V_{*}(s) = \max_a q_{*}(s, a)$$

value of each state when taken an action a .

Question 6

The bug described in question 4.4,

to find a way such that policy doesn't keep on switching in case of finding multiple policies

In this case,

pseudo code

if $V_{\pi}(s) \ll 0$:

don't update the $V_{\pi}(s)$

θ increase by a small value

to break continuous update of states.

Notes

Appointment