

RL Assignment-2

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Question - 2

Results

Initial Value Function

0	0	0	0	0
0	0	0	0	0
0	0	0	0	0
0	0	0	0	0
0	0	0	0	0

Value Function after policy iteration

3.3	8.8	4.4	5.3	1.5
1.5	3.0	2.3	1.9	0.5
0.1	0.7	0.7	0.7	-0.4
-1	-0.4	-0.4	-0.6	-1.2
-1.9	-1.3	-1.2	-1.4	-2.0

Question - 4

Results

Initial Value Function

0	0	0	0	0
0	0	0	0	0
0	0	0	0	0
0	0	0	0	0
0	0	0	0	0

Value Function after policy iteration

22	24.4	22.	19.4	17.5
19.8	22.	19.8	17.8	16
17.8	19.8	17.8	16	14.4
16	17.8	16	14.4	13.0
14.4	16.0	14.4	13.0	11.7

Question - 6

Policy Improvement

Initial Random Policy

3	2	3	3
2	2	3	2
0	1	2	3
3	0	0	0

Final Policy

0	0	0	0
1	0	0	3
1	0	2	3
1	2	2	0

Value Iteration

Initial Policy

0	0	0	0
0	0	0	0
0	0	0	0
0	0	0	0

Final Policy

0	0	0	0
1	0	0	3
1	0	2	3
1	2	2	0

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Question-1

s	a	s'	$p(s'/s, a)$	$r(s, a, s')$	$p(s', r s, a)$
high	search	high	α	r_{search}	$0.1 \alpha r_{\text{search}}$
high	search	low	$1 - \alpha$	r_{search}	$-0.1 (1 - \alpha) r_{\text{search}}$
low	search	high	$1 - \beta$	-3	0
low	search	low	β	r_{search}	$-0.1 \beta r_s$
high	wait	high	1	r_{wait}	$0.1 r_w$
high	wait	low	0	—	—
low	wait	high	0	—	—
low	wait	low	1	r_{wait}	$0.1 r_w$
low	recharge	high	1	0	—
low	recharge	low	0	—	—

As we know that -

$$r(s, a, s') = \sum_{r \in R} \frac{r p(s', r | s, a)}{p(s' | s, a)}$$

$$p(s', r | s, a) = \frac{r(s, a, s') p(s' | s, a)}{\sum_{r \in R} r}$$

$$\textcircled{1} \quad p(s'_r | s_a) = \frac{\gamma_{\text{search}} \cdot \alpha}{10}$$

$$= 0.1 \gamma_{\text{search}} \alpha$$

$$\textcircled{2} \quad - \frac{(1-\alpha) \gamma_{\text{search}}}{10}$$

$$= -0.1(1-\alpha) \gamma_{\text{search}}$$

Similarly we can calculate the other values as well.

83 Exercise 3.15

we know that

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

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

Add constant in reward.

$$\bar{r}_{t+k+1} = r_{t+k+1} + c$$

$$V^{\pi}(s) = E_{\pi} \left[\sum_{k=0}^{\infty} \gamma^k \bar{r}_{t+k+1} | S_t = s \right]$$

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

$$\hat{V}^{\pi}(s) = V^{\pi}(s) + c \sum_{k=0}^{\infty} \gamma^k$$

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

EX - 3.16

Adding a constant at episodic tasks

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

$$= E \left[(R_{t+1} + c) + \gamma (R_{t+2} + c) + \gamma^{k-1} (R_{t+k} + c) \right]$$

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

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

Question - 5

$$V_{\pi}(s) = E_{\pi} \left[R_{t+1} + \gamma V_{\pi}(s_{t+1}) \mid s_t = s \right]$$

$$= E_{\pi^*} \left[R_{t+1} + \gamma V_{\pi^*}(s_{t+1}) \mid s_t = s \right]$$

$$= \max_a E \left[R_{t+1} + \gamma V_{\pi^*}(s_{t+1}) \mid s_t = s \right]$$

$$= \max_a \sum_r \sum_{s'} (r + \gamma V_{\pi^*}(s')) p(s', r \mid s, a)$$

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

Question - 6

The bug which causes the policy to keep on switching in case of finding multiple policies.

In this case

$$\text{if } V_{\pi}(s) \leq 0$$

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

ϵ increase by a small value to break continuous update of state

Question 4.

Non linear solution of bellman equations using linear programming

$$V(s) \geq R(s) + \gamma \max_{a \in A} \sum p(s'/s, a) V(s')$$

view 1A) if linear constraints.

$$V(s) \geq R(s) + \gamma \sum_{s' \in S} p(s'/s, a) V(s')$$

$$\forall a \in A$$

Now using linear program.

$$s \rightarrow \text{to } V(s) \geq R(s) + \gamma \sum_{s' \in S} p(s'/s, a) V(s')$$

$$\forall a \in A, s \in S$$

~~Theorem~~

support

$$V(s) \geq R(s) + \gamma \max_{a \in A} \sum_{s' \in S} P(s'/s, a) V(s')$$

In objective, we can optimize any positive linear function of $V(s)$.
and result above will be true

$$\text{minimize } \sum_s d(s) V(s)$$

$$\text{s.t. } V(s) \geq R(s) + \gamma \sum_{s' \in S} P(s'/s, a) V(s')$$

$$\forall a \in A, s \in S.$$

$d(s)$ is distribution over states

Adding dual variables $u(s, a)$

$$\text{Maximize } \sum_{s \in S} R(s) \sum_{a \in A} u(s, a).$$

$$\text{s.t. } \sum_{a \in A} u'(s', a) s d(s') + \gamma \sum_{s \in S} \sum_{a \in A} P(s' | s, a) u(s, a)$$

$$\forall s' \in S$$

$$u(s, a)$$