Progress

NPTEL » Reinforcement Learning

Mentor

1 point

Unit 6 - Week 4

How does an NPTEL online

Course outline

course work?

Week 0

Week 1

Week 2

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Week 5

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Assignment Solutions

 $T([x_1, x_2]) = [0.5x_1, 0.5x_2]$

 $T([x_1, x_2]) = [x_1, x_2, x_1 + x_2]$

10) For an operator L, which of the following properties must be satisfied by x for it to be a fixed point for L?

 $T([x_1, x_2]) = [2x_1, 2x_2]$

 $T([x_1, x_2]) = x_1 + x_2$

No, the answer is incorrect.

 $T([x_1, x_2]) = [0.5x_1, 0.5x_2]$

Accepted Answers:

 $T([x_1, x_2]) = x_1 + x_2$

Score: 0

Lx = x

 $L^2x = x$

Score: 0

Lx = x $L^2x = x$

 $\forall \lambda > 0 \ Lx = \lambda x$

Accepted Answers:

None of the above

No, the answer is incorrect.

NPTEL Resources

MDP Modelling

Bellman Equation

Bellman Optimality Equation

Banach Fixed Point Theorem

Cauchy Sequence and

Green's Equation

Convergence Proof

Quiz : Assignment 4

Reinforcement Learning:

Week 4 Feedback form

Assignment 4 The due date for submitting this assignment has passed. Due on 2020-02-26, 23:59 IST. As per our records you have not submitted this assignment. Which of the following is a benefit of using RL algorithms for solving MDPs? 1 point They do not require the state of the agent for solving MDP They do not require the action taken by the agent for solving MDP They do not require state transition probability for solving MDP hey do not require reward signal for solving MDP No, the answer is incorrect. Score: 0 Accepted Answers: They do not require state transition probability for solving MDP Select all the correct option(s) from the following equations: 1 point $v^{\pi}(s) = \mathbb{E}_{\pi}[\sum_{i=t}^{T} \gamma^{i-t} R_{i+1} | S_t = s]$ $q^{\pi}(s,a) = \sum_{s'} p(s'|s,a) v^{\pi}(s')$ $v^{\pi}(s) = \sum_{a} \pi(a|s)q^{\pi}(s,a)$ $q^{\pi}(s, a) = \sum_{s'} p(s'|s, a) [\mathbb{E}[r|s, a, s'] + \gamma \sum_{a'} \pi(a'|s') q^{\pi}(s', a')]$ No, the answer is incorrect. Score: 0 Accepted Answers: $v^{\pi}(s) = \mathbb{E}_{\pi}[\sum_{i=t}^{T} \gamma^{i-t} R_{i+1} | S_t = s]$ $q^{\pi}(s,a) = \sum_{s'} p(s'|s,a) v^{\pi}(s')$ $v^{\pi}(s) = \sum_a \pi(a|s)q^{\pi}(s,a)$ $q^{\pi}(s, a) = \sum_{s'} p(s'|s, a) [\mathbb{E}[r|s, a, s'] + \gamma \sum_{a'} \pi(a'|s') q^{\pi}(s', a')]$ Select the correct Bellman optimality equation: 1 point $v^*(s) = \max_a \sum_{s'} p(s'|s, a) [\mathbb{E}[r|s, a, s'] + \gamma v^*(s')]$ $v^*(s) = \max_a \sum_{s'} p(s'|s, a)v^*(s')$ $v^*(s) = \max_a \sum_{s'} p(s'|s, a) [\gamma \mathbb{E}[r|s, a, s'] + v^*(s')]$ $v^*(s) = \max_a \sum_{s'} p(s'|s, a) \gamma [\mathbb{E}[r|s, a, s'] + v^*(s')]$ No, the answer is incorrect. Score: 0 Accepted Answers: $v^*(s) = \max_a \sum_{s'} p(s'|s, a) [\mathbb{E}[r|s, a, s'] + \gamma v^*(s')]$ State True/False for the following statement. 1 point While solving MDPs, in case of discounted rewards, the value of γ (discount factor) can change the speed of learning but the final policy that would be learnt remains the same. True False No, the answer is incorrect. Score: 0 Accepted Answers: False State True/False for the following statement 1 point In MDPs, there is a unique resultant state for any given state-action pair True False No, the answer is incorrect. Score: 0 Accepted Answers: False 6) Which of the following should be known for modelling an MDP? 1 point [i] Complete set of States [ii] Set of action values (iii) transition probabilities for state and actions [iv] reward values for every allowed transition between states i,ii,iii, iv ○i, ii i,ii,iii iii, iv No, the answer is incorrect. Accepted Answers: i,ii,iii, iv 7) Which of the following statements are true for an MDP? 1 point An MDP with stochastic rewards may not have a deterministic optimal policy There can be multiple stochastic policies which are as good as the optimal deterministic policy If $0 \le \gamma < 1$, then rank of the matrix $I - \gamma P_{\pi}$ is equal to |S|If $0 \le \gamma < 1$, then rank of the matrix $I - \gamma P_{\pi}$ is less than |S|No, the answer is incorrect. Score: 0 Accepted Answers: There can be multiple stochastic policies which are as good as the optimal deterministic policy If $0 \le \gamma < 1$, then rank of the matrix $I - \gamma P_{\pi}$ is equal to |S|8) Consider an MDP with 3 states A,B,C. At each state we can go to either of the two states. i.e if we are in state A then we can perform 2 actions, going to state B or C. The rewards for each transactions are r(A, B) = 3 (reward if we go from A to B), r(B, A) = 4, r(B, C) = 4, r(C, B) = 1, r(A, C) = 10, r(C, A) = 2, discount factor is 0.5. Find the fixed point of the value function for the policy $\pi(A) = B$ (if we are in state A we choose the action to go to B) $\pi(B) = C, \pi(C) = A \cdot v^{\pi}([A B C]) = ?$ [6666][6.28 6.57 5.14] [6.34 6.88 5.92] [5.47 6.23 4.13] No, the answer is incorrect. Score: 0 Accepted Answers: [6.28 6.57 5.14] 9) Which of the following is a contraction? 1 point