Congratulations! You passed!

Grade received 80% To pass 80% or higher

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1.	A policy is a function which maps to	1 / 1 point
	O States to values.	
	O Actions to probability distributions over values.	
	States to probability distributions over actions.	
	O States to actions.	
	Actions to probabilities.	
	○ Correct Correct!	
2.	The term "backup" most closely resembles the term in meaning.	0 / 1 point
		o / I point
	○ Value ○ Update	
	Diagram	
	⊗ Incorrect Incorrect. The term "backup" is used synonymously with the term "update". For example, a backup	
	diagram is equivalent to an update diagram.	
3.	At least one deterministic optimal policy exists in every Markov decision process.	1/1 point
	True	
	○ False	
	\bigcirc Correct Correct! Let's say there is a policy π_1 which does well in some states, while policy π_2 does well in others. We could combine these policies into a third policy π_3 , which always chooses actions according to whichever of policy π_1 and π_2 has the highest value in the current state. π_3 will necessarily have a value greater than or equal to both π_1 and π_2 in every state! So we will never have a situation where doing well	
	in one state requires sacrificing value in another. Because of this, there always exists some policy which is best in every state. This is of course only an informal argument, but there is in fact a rigorous proof showing that there must always exist at least one optimal deterministic policy.	
4.	The optimal state-value function:	1/1 point
	O Is not guaranteed to be unique, even in finite Markov decision processes.	
	Is unique in every finite Markov decision process.	
	○ Correct ○ Correct ○ Correct! The Bellman optimality equation is actually a system of equations, one for each state, so if there are N states, then there are N equations in N unknowns. If the dynamics of the environment are known, then in principle one can solve this system of equations for the optimal value function using any one of a variety of methods for solving systems of nonlinear equations. All optimal policies share the same optimal	
	state-value function.	
5.	Does adding a constant to all rewards change the set of optimal policies in episodic tasks?	1/1 point
	Yes, adding a constant to all rewards changes the set of optimal policies.	
	 No, as long as the relative differences between rewards remain the same, the set of optimal policies is the same. 	
	Correct Correct! Adding a constant to the reward signal can make longer episodes more or less advantageous (depending on whether the constant is positive or negative).	
6.	Does adding a constant to all rewards change the set of optimal policies in continuing tasks?	1 / 1 point
	Yes, adding a constant to all rewards changes the set of ontimal policies.	

 $\textbf{ (o)} \ \ \text{No, as long as the relative differences between rewards remain the same, the set of optimal policies is the } \\$

⊘ Correct

Correct! Since the task is continuing, the agent will accumulate the same amount of extra reward independent of its behavior.

7. Select the equation that correctly relates v_* to q_* . Assume π is the uniform random policy.

1/1 point

- $v_*(s) = max_aq_*(s, a)$
- $\bigcirc v_*(s) = \sum_{a,r,s'} \pi(a|s) p(s',r|s,a) [r + \gamma q_*(s')]$
- $\bigcirc v_*(s) = \sum_{a,r,s} \pi(a|s) p(s',r|s,a) q_*(s')$
- $v_*(s) = \sum_{a,r,s'} \pi(a|s) p(s',r|s,a) [r + q_*(s')]$
- **⊘** Correct

Correct!

8. Select the equation that correctly relates q_* to v_* using four-argument function p_* .

1/1 point

- $Q_*(s, a) = \sum_{s, r} p(s, r|a, s)[r + v_*(s)]$
- $\bigcap q_*(s, a) = \sum_{s',r} p(s', r|a, s) \gamma[r + v_*(s')]$
- \bigcirc $q_*(s,a) = \sum_{s',r} p(s',r|a,s)[r + \gamma v_*(s')]$
- **⊘** Correct

Correct!

9. Write a policy π_* in terms of q_* .

1/1 point

- $\bigcirc \ \pi_*(a|s) = q_*(s,a)$
- $\bigcap \pi_*(a|s) = \max_{a'} q_*(s,a')$
- \bullet $\pi_*(a|s) = 1$ if $a = \operatorname{argmax}_{a'} q_*(s, a')$, else 0
- **⊘** Correct

Correct!

10. Give an equation for some π_* in terms of v_* and the four-argument p.

0 / 1 point

- $\bigcirc \ \pi_*(a|s)=1 \ \text{if} \ v_*(s)=\sum_{s',r} p(s',r|s,a)[r+\gamma v_*(s')], \ \text{else} \ 0$
- $\bigcap \pi_*(a|s) = \max_{a'} \sum_{s',r} p(s',r|s,a')[r + \gamma v_*(s')]$
- $\bigcirc \ \pi_*(a|s) = \sum_{s',r} p(s',r|s,a)[r + \gamma v_*(s')]$

Incorrect. This equation will give a probability of 1 to every action.