## Your grade: 80%

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Next item →

1.		evalue of any state under an optimal policy is the value of that state under a non-optimal policy.  Strictly greater than	1/1 point
	<b>✓</b>	Greater than or equal to	
		Correct! This follows from the policy improvement theorem.	
		Strictly less than	
		Less than or equal to	
2.		policy $\pi$ is greedy with respect to its own value function $v_\pi$ , then it is an optimal policy.	1/1 point
	•	True	
		Correct! If a policy is greedy with respect to its own value function, it follows from the policy improvement theorem and the Bellman optimality equation that it must be an optimal policy.	
	0	False	
3.		$v_\pi$ be the state-value function for the policy $\pi$ . Let $v_{\pi'}$ be the state-value function for the policy $\pi'$ . ume $v_\pi=v_\pi'$ . Then this means that $\pi=\pi'$ .	1/1 point
	0	True	
	•	False	
		Correct! For example, two policies might share the same value function, but differ due to random tie breaking.	
4.	Wha	at is the relationship between value iteration and policy iteration? [Select all that apply]	1/1 point
	<b>~</b>	Value iteration and policy iteration are both special cases of generalized policy iteration.	
		Correct!	
		Policy iteration is a special case of value iteration.	
		Value iteration is a special case of policy iteration.	
5.		word synchronous means "at the same time". The word asynchronous means "not at the same time". A amic programming algorithm is: [Select all that apply]	1/1 point
	<b>~</b>	Synchronous, if it systematically sweeps the entire state space at each iteration.	
		Correct! Only algorithms that update every state exactly once at each iteration are synchronous.	
I	<b>✓</b>	Asynchronous, if it does not update all states at each iteration.	
		Correct! Only algorithms that update every state exactly once at each iteration are synchronous.	
	<b>✓</b>	Asynchronous, if it updates some states more than others.	

6. All Generalized Policy Iteration algorithms are synchronous.

1/1 point

False

Correct! A Generalized Policy Iteration algorithm can update states in a non-systematic fashion.

- True
- 7. Which of the following is true?

1/1 point

Asynchronous methods generally scale to large state spaces better than synchronous methods.

Correct! Asynchronous methods can focus updates on more relevant states, and update less relevant states less often. If the state space is very large, asynchronous methods may still be able to achieve good performance whereas even just one synchronous sweep of the state space may be intractable.

- Synchronous methods generally scale to large state spaces better than asynchronous methods.
- 8. Why are dynamic programming algorithms considered planning methods? [Select all that apply]

1 point

▼ They compute optimal value functions.

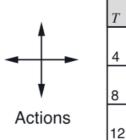
Incorrect. A planning method is a method that uses a model to improve the policy.

- ☐ They learn from trial and error interaction.
- ✓ They use a model to improve the policy.

Correct! This is the definition of a planning method.

9. Consider the undiscounted, episodic MDP below. There are four actions possible in each state, A = {up, down, right, left}, which deterministically cause the corresponding state transitions, except that actions that would take the agent off the grid in fact leave the state unchanged. The right half of the figure shows the value of each state under the equiprobable random policy. If  $\pi$  is the equiprobable random policy, what is q(11, down)?

1 point



T	1	2	3
4	5	6	7
8	9	10	11
12	13	14	T
	15		

$$R = -1$$
 on all transitions

Т	-14.	-20.	-22.
-14.	-18.	-20.	-20.
-20.	-20.	-18.	-14.
-22.	-20.	-14.	T

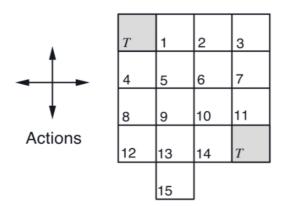
$$\bigcirc q(11, down) = -1$$

Incorrect. Moving down incurs a reward of -1 before reaching the terminal state, after which the episode is over.

$$\bigcirc q(11, \text{down}) = -14$$

$$\bigcirc q(11, down) = -15$$

10. Consider the undiscounted, episodic MDP below. There are four actions possible in each state, A = {up, down, right, left}, which deterministically cause the corresponding state transitions, except that actions that would take the agent off the grid in fact leave the state unchanged. The right half of the figure shows the value of each state under the equiprobable random policy. If  $\pi$  is the equiprobable random policy, what is v(15)? Hint: Recall the Bellman equation  $v(s) = \sum_a \pi(a|s) \sum_{s',r} p(s',r|s,a)[r+\gamma v(s')]$ .



R = -1
on all transitions

T	-14.	-20.	-22.
-14.	-18.	-20.	-20.
-20.	-20.	-18.	-14.
-22.	-20.	-14.	Т

$$v(15) = -21$$

$$v(15) = -24$$

Correct! We can get this by solving for the unknown variable v(15). Let's call this unknown x. We solve for x in the equation x=1/4(-21)+3/4(-1+x). The first term corresponds to transitioning to state 13. The second term corresponds to taking one of the other three actions, incurring a reward of -1 and staying in state x.

$$\bigcirc v(15) = -22$$

$$\bigcirc \ v(15) = -23$$

$$v(15) = -25$$

