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[Unit 5 Reinforcement Learning \(2 weeks\)](#)

[Lecture 17. Reinforcement Learning](#)
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> 8. Q-value Iteration

8. Q-value Iteration

Q-value Iteration

and do exactly the same computation and directly get Q_s .

And that's exactly what you will do in your exercise.

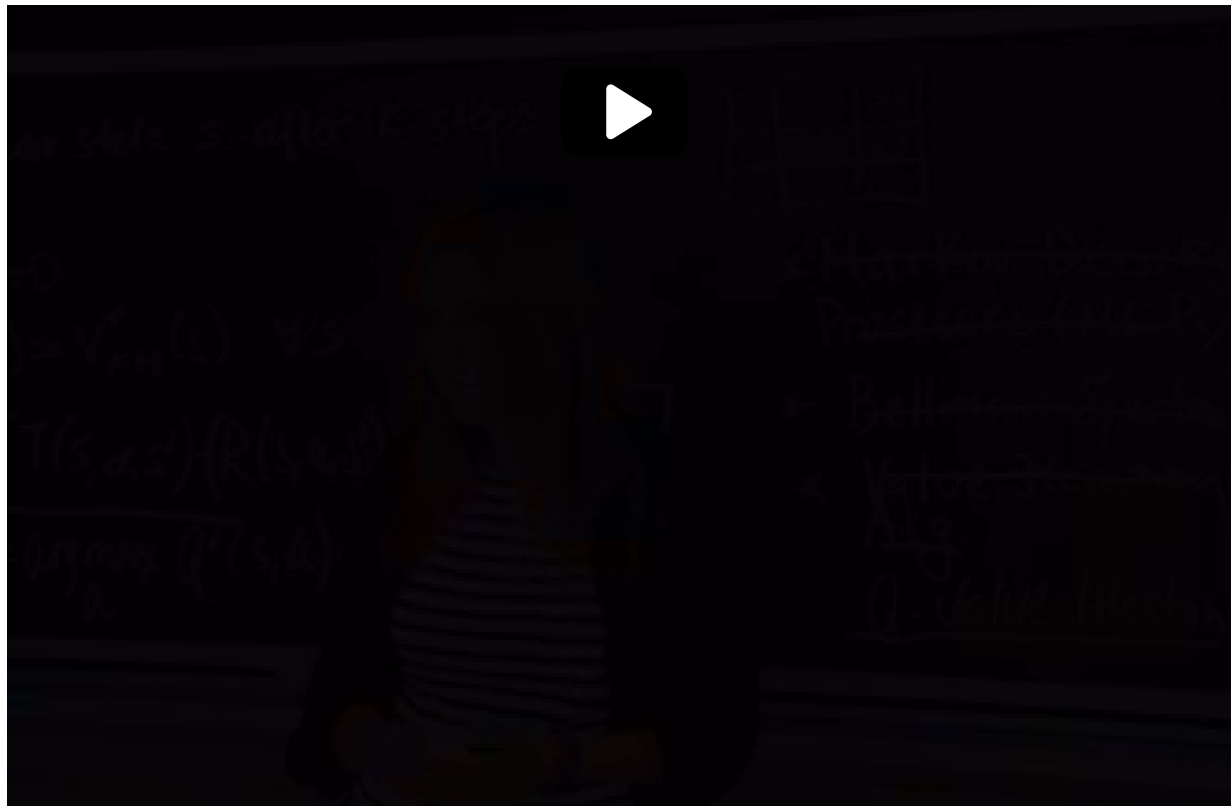
You will reformulate this algorithm and get a new algorithm which is called Q value iterations.

But the idea-- exactly the same.

And we will use this algorithm in our discussion

about reinforcement learning next time.

Thank you.



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In this problem, we compute the optimal policy using Q-value iteration.

First, let's recall some of the bellman equations:

$$Q^*(s, a) = \sum_{s'} T(s, a, s') (R(s, a, s') + \gamma V^*(s'))$$

We also know that, $V^*(s) = \max_a Q^*(s, a)$. Therefore,

$$Q^*(s, a) = \sum_{s'} T(s, a, s') (R(s, a, s') + \gamma \max_{a'} Q^*(s', a'))$$

Q value iteration update rule

1/1 point (graded)

Q-value iteration update rule from the above equation could be derived similar to the value iteration update rule.

Pick the most suitable Q-value iteration update rule from the options below:

☐ $Q_{k+1}^*(s, a) = \sum_{s'} T(s, a, s') (R(s, a, s') + \gamma \max_{s'} Q_k^*(s', a))$

☐ $Q_{k+1}^*(s, a) = \sum_{s'} T(s, a, s') (R(s, a, s') + \gamma V^*(s'))$

☒ $Q_{k+1}^*(s, a) = \sum_{s'} T(s, a, s') (R(s, a, s') + \gamma \max_{a'} Q_k^*(s', a'))$ ✓

☐ $Q_{k+1}^*(s, a) = \sum_{s'} T(s, a, s') (R(s, a, s') + \gamma Q_k^*(s', a))$

Solution:

First let's recall the following bellman equations:

$$Q^*(s, a) = \sum_{s'} T(s, a, s') (R(s, a, s') + \gamma V^*(s'))$$

We also know that, $V^*(s) = \max_a Q^*(s, a)$. Therefore,

$$Q^*(s, a) = \sum_{s'} T(s, a, s') (R(s, a, s') + \gamma \max_{a'} Q^*(s', a'))$$

Q value iteration would use previous iteration of the Q-value on the right hand side of the above equation to update the Q value estimate of the current step. Hence, the Q value update for k^{th} step would look like:

$$Q_{k+1}^*(s, a) = \sum_{s'} T(s, a, s') (R(s, a, s') + \gamma \max_{a'} Q_k^*(s', a'))$$

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You have used 1 of 2 attempts

i Answers are displayed within the problem

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