

Q1-1: Select the correct statement.

- A. *Markov Assumption implies that given the present state and action, all following states are independent of all past states.*
- B. *All Reinforcement Learning techniques adopts Markov assumption property.*

- 1. Both the statements are TRUE.
- 2. Statement A is TRUE, but statement B is FALSE.
- 3. Statement A is FALSE, but statement B is TRUE.
- 4. Both the statements are FALSE.

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Though markov assumption makes the analysis easier, it's not necessary to assume Markov property.

# Break & Quiz

**Q 1.2** Consider an MDP with 2 states  $\{A, B\}$  and 2 actions: “**stay**” at current state and “**move**” to other state; suppose we are guaranteed to transition according to the action. Let  $r$  be the reward function such that  $r(A) = 1$ ,  $r(B) = 0$ . Let  $\gamma$  be the discounting factor. What is the optimal policy  $\pi(A)$  and  $\pi(B)$ ? What are  $V^*(A)$ ,  $V^*(B)$ ?

- A. Stay, Stay,  $1/(1-\gamma)$ , 1
- B. Stay, Move,  $1/(1-\gamma)$ ,  $1/(1-\gamma)$
- C. Move, Move,  $1/(1-\gamma)$ , 1
- D. Stay, Move,  $1/(1-\gamma)$ ,  $\gamma/(1-\gamma)$

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- C. Move, Move,  $1/(1-\gamma)$ , 1
- **D. Stay, Move,  $1/(1-\gamma)$ ,  $\gamma/(1-\gamma)$**  Note: want to stay at A, if at B, move to A. Starting at A, sequence A,A,A,... rewards  $1, \gamma, \gamma^2, \dots$ . Start at B, sequence B,A,A,... rewards  $0, \gamma, \gamma^2, \dots$ . Sums to  $1/(1-\gamma)$ ,  $\gamma/(1-\gamma)$ .

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**Q 2.1** For Q learning to converge to the true Q function, we must

- A. Visit every state and try every action
- B. Perform at least 20,000 iterations.
- C. Re-start with different random initial table values.
- D. Prioritize exploitation over exploration.

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**Q 2.1** For Q learning to converge to the true Q function, we must

- **A. Visit every state and try every action**
- B. Perform at least 20,000 iterations. (No: this is dependent on the particular problem, not a general constant).
- C. Re-start with different random initial table values. (No: this is not necessary in general).
- D. Prioritize exploitation over exploration. (No: insufficient exploration means potentially unupdated state action pairs).