

**R18**

**Code No: 157HR**

**JAWAHARLAL NEHRU TECHNOLOGICAL UNIVERSITY HYDERABAD**

**B. Tech IV Year I Semester Examinations, February - 2025**

**REINFORCEMENT LEARNING  
(Common to CSE(AIML), AI&ML)**

**Time: 3 Hours**

**Max. Marks: 75**

**Note:** i) Question paper consists of Part A, Part B.

ii) Part A is compulsory, which carries 25 marks. In Part A, Answer all questions.

iii) In Part B, Answer any one question from each unit. Each question carries 10 marks and may have a, b as sub questions.

**PART – A**

**(25 Marks)**

- 1.a) How does the UCB algorithm help balance exploration and exploitation? [2]
- b) What is a Markov Decision Process (MDP)? [3]
- c) How do prediction and control problems differ in Reinforcement Learning? [2]
- d) What is bootstrapping in Reinforcement Learning? [3]
- e) Write the purpose of TD (0) algorithm. [2]
- f) What is the role of Kullback-Leibler divergence in the KL-UCB algorithm? [3]
- g) How is the value function used to evaluate policies in an MDP? [2]
- h) What is Monte Carlo method? [3]
- i) Give key components of Monto Carlo problem. [2]
- j) Why is generalization important in Reinforcement Learning? [3]

**PART – B**

**(50 Marks)**

- 2.a) Define a Stochastic Multi-Armed Bandit. How does it relate to the exploration trade-off in decision-making?
  - b) Explain the concept of Regret in the context of Multi-Armed Bandits. [5+5]
- OR**
- 3.a) Explain the KL-UCB algorithm and its role in Multi-Armed Bandit problems.
  - b) Discuss the concept of Thompson Sampling as a Bayesian approach to Multiarmed Bandit problems. [5+5]
- 4.a) Explain the fundamentals of a Markov Decision Problem (MDP) in reinforcement learning. What are the key components, and how do they relate to decision making?
  - b) Explain infinite discounted reward, total reward, finite horizon reward, and average reward. Provide examples of scenarios where each type is applicable. [5+5]

**OR**

- 5.a) Explain Bellman's optimality operator and its role in dynamic programming approaches to reinforcement learning.
- b) Describe the key steps involved in the Value Iteration algorithm for solving MDPs. [5+5]

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- 6.a) Elaborate on the concept of model-based Reinforcement Learning algorithms.
- b) Define policy and value function in the context of MDPs. [5+5]

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- 7. Explain in detail about the key characteristics of Monte Carlo methods. [10]

- 8. Explain the concept of convergence of Monte Carlo and batch TD(0) algorithm. [10]

**OR**

- 9.a) What is Q-learning? Explain it in detail.
- b) Compare Q-learning, Sarsa, and Expected Sarsa. [5+5]

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- 10.a) Discuss the need for generalization in Reinforcement Learning practice.
- b) Explain linear function approximation and the geometric view of it in the context of Reinforcement Learning. [5+5]

**OR**

- 11.a) Discuss Control with Function Approximation.
- b) Describe Policy Search methods in Reinforcement Learning. [5+5]

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