

Reg. No.

Question Paper Code

13442

B.E. / B.Tech. - DEGREE EXAMINATIONS, APRIL / MAY 2025

Fourth Semester

Artificial Intelligence and Data Science

Common to Computer Science and Engineering (Artificial Intelligence and Machine Learning)

20AIPC403 - ADVANCED MACHINE LEARNING

Regulations - 2020

Duration: 3 Hours

Max. Marks: 100

PART - A (MCQ) (10 × 1 = 10 Marks)

Answer ALL Questions

- | | Marks | K-Level | CO |
|---|-------|---------|-----|
| 1. The Ising Model's nodes can have spin states of _____ or _____.
(a) +1 or -1 (b) 0 or 1 (c) +2 or -2 (d) 1 or 2 | 1 | K1 | CO1 |
| 2. The energy functions in a Markov Network are defined using _____ potentials.
(a) clique (b) path (c) network (d) node | 1 | K1 | CO1 |
| 3. Bayesian networks are _____.
(a) cyclic (b) undirected (c) acyclic (d) directed and acyclic | 1 | K1 | CO2 |
| 4. Parameter Estimation problem is about:
(a) Identifying Input Parameters (b) Identifying Output Parameters
(c) Identifying Model Parameters (d) All of the Mentioned | 1 | K1 | CO2 |
| 5. VAE loss function composed of _____ and regularization terms can be derived using particular statistical technique of variational interference.
(a) normal (b) reconstruction (c) Gaussian (d) None of the Mentioned | 1 | K1 | CO3 |
| 6. MCMC sampling uses to sample from _____ distribution for the purpose of inference.
(a) prior (b) posterior (c) max likelihood (d) likelihood | 1 | K1 | CO3 |
| 7. The stochastic forward passes illustrate the decomposition of predictive uncertainty into _____.
(a) epistemic components (b) aleatoric components
(c) supervised components (d) epistemic components and aleatoric components | 1 | K1 | CO4 |
| 8. By catering to the _____, Bayesian neural network can avoid the overfitting problem by addressing the regularization properties.
(a) Random distributions (b) probability distributions
(c) sequential distributions (d) None of the Mentioned | 1 | K1 | CO4 |
| 9. In the _____, base classifiers will output their classifications, and then the Meta-classifier(s) will make the final classification
(a) prediction phase (b) modeling phase
(c) classification phase (d) aggregation phase | 1 | K1 | CO5 |
| 10. In an encoder-decoder framework for time series forecasting, what is the primary role of the encoder?
(a) To extract relevant temporal patterns from historical data
(b) To directly generate future forecasts without transformations
(c) To minimize dependencies between correlated time series
(d) To replace probabilistic forecasting with deterministic modeling | 1 | K1 | CO6 |

PART - B (12 × 2 = 24 Marks)

Answer ALL Questions

- | | | | |
|---|---|----|-----|
| 11. Describe the impact of clique. | 2 | K2 | CO1 |
| 12. Differentiate an ising and potts model. | 2 | K2 | CO1 |
| 13. Define junction tree calibration. | 2 | K2 | CO2 |

14. List the uses of variable estimation.	2	K2	CO2
15. Distinguish between forward and importance sampling.	2	K2	CO3
16. Define sampling.	2	K2	CO3
17. List the abstractions defined by Tensor Flow Distribution.	2	K1	CO4
18. Explain the autoregressive model in density estimator.	2	K2	CO4
19. Define Bayesian Neural Network.	2	K1	CO5
20. Illustrate the Aleatory uncertainty.	2	K2	CO5
21. Define the auto encoder.	2	K1	CO6
22. Why is the Gaussian Copula useful in multivariate time series forecasting?	2	K2	CO6

PART - C (6 × 11 = 66 Marks)

Answer ALL Questions

23. a)	Explain the directed graphical model with an example.	11	K2	CO1
OR				
b)	Explain the Potts and Ising models? Analyze their principles and applications in different domains.	11	K2	CO1
24. a)	Discuss the use of Variable elimination algorithms in graphical model inference.	11	K2	CO2
OR				
b)	Explain how message passing is carried out in a junction tree and why it is useful for inference.	11	K2	CO2
25. a)	Illustrate and explain the concept of Variational Auto encoders (VAEs) using a well-structured diagram.	11	K2	CO3
OR				
b)	Write types of sampling and demonstrate the importance sampling.	11	K2	CO3
26. a)	Describe briefly about Masked Autoregressive Flow for Density Estimation	11	K2	CO4
OR				
b)	Explain are TensorFlow Distributions, and how do they facilitate probabilistic programming and uncertainty estimation in machine learning?	11	K2	CO4
27. a)	Explain Bayesian Neural Networks in detail, covering their probabilistic nature, inference methods, and real-world applications.	11	K2	CO5
OR				
b)	Explain the concept of Meta-Learning to a real-world machine learning scenario and demonstrate how it allows models to adapt faster to new tasks by leveraging past experiences.	11	K2	CO5
28. a)	Apply the Gaussian Copula framework to model dependencies between multiple time series. How does it improve the quality of probabilistic forecasts?	11	K3	CO6
OR				
b)	Construct and explain a DeepAR-based model for forecasting. Compare how it is differ in learning and output generation compared to classical autoregressive models?	11	K3	CO6