

U.S.N.

B.M.S. College of Engineering, Bengaluru-560019

Autonomous Institute Affiliated to VTU

August 2024 Supplementary Examinations**Programme: B.E.****Branch: Artificial Intelligence and Machine Learning****Course Code: 22AM6PCAML****Course: Advanced Machine Learning****Semester: VI****Duration: 3 hrs.****Max Marks: 100**

Instructions: 1. Answer any FIVE full questions, choosing one full question from each unit.
2. Missing data, if any, may be suitably assumed.

Important Note: Completing your answers, compulsorily draw diagonal cross lines on the remaining blank pages. Revealing of identification, appeal to evaluator will be treated as malpractice.			UNIT - I	<i>CO</i>	<i>PO</i>	Marks
	1	a)	With an equation, discuss how to generate data using the class-conditional probabilities and class priors.	<i>CO2</i>	<i>PO1</i>	04
		b)	Derive the posterior predictive distribution for a single multinoulli trial and also discuss how to avoid the zero-count problem using the Dirichlet-multinomial model.	<i>CO3</i>	<i>PO3</i>	08
		c)	Illustrate different types of features used to form the class-conditional density of the Naive Bayes Classifier.	<i>CO3</i>	<i>PO2</i>	08
			UNIT - II			
	2	a)	Provide the Strategies for preventing overfitting in multivariate Gaussian or Multivariate Normal (MVN).	<i>CO1</i>	<i>PO1</i>	06
		b)	Derive the Inverse Wishart distribution (IW) using multivariate gamma function. Visualization IW with necessary graphs.	<i>CO3</i>	<i>PO3</i>	06
		c)	Derive the class posterior using Linear Discriminant Analysis (LDA) model. Visualize LDA with necessary graphs.	<i>CO3</i>	<i>PO3</i>	08
			UNIT - III			
	3	a)	With an equation, discuss how to compute posterior using Bayesian model selection.	<i>CO1</i>	<i>PO1</i>	06
		b)	Derive the equation to compute the log marginal likelihood using Bayesian information criterion approximation.	<i>CO3</i>	<i>PO3</i>	08
		c)	Demonstrate how the frequency statistics is based on the sampling distribution with a suitable example.	<i>CO3</i>	<i>PO3</i>	06
			UNIT - IV			
	4	a)	Verify that the log marginal distribution of the observed data $\ln p(X)$ can be decomposed into two terms in the form $\ln p(X) = L(q) + KL(q p)$ where $L(q)$ and $KL(q p)$ is given.	<i>CO2</i>	<i>PO2</i>	06
		b)	Illustrate the Expectation Propagation algorithm using a simple example of clutter problem to infer the mean θ of a multivariate	<i>CO3</i>	<i>PO3</i>	08

		Gaussian distribution over a variable x (set of observations) drawn from the distribution with necessary graph and equations.			
	c)	Discuss the concept of Markov chain Monte Carlo (MCMC) considering sampling from a large class of distributions.	<i>CO1</i>	<i>PO1</i>	06
		OR			
5	a)	Demonstrate how the Expectation Propagation and Variational message passing correspond to the optimization of two different KL divergences.	<i>CO3</i>	<i>PO3</i>	06
	b)	Explain Gibbs Sampling and Illustrate by alternate updates of two variables whose distribution is a correlated Gaussian.	<i>CO2</i>	<i>PO2</i>	06
	c)	Discuss the Box-Muller method for generating Gaussian distributed random numbers starts by generating samples from a uniform distribution.	<i>CO1</i>	<i>PO1</i>	08
		UNIT - V			
6	a)	Explain the concept of kernel trick.	<i>CO1</i>	<i>PO1</i>	06
	b)	Derive Mercer kernel using gram matrix with a suitable example.	<i>CO3</i>	<i>PO3</i>	08
	c)	Discuss the Application: Google's PageRank algorithm for web page ranking using Markov models.	<i>CO3</i>	<i>PO2</i>	06
		OR			
7	a)	Examine the Kernels for comparing documents using the bag of words.	<i>CO1</i>	<i>PO1</i>	06
	b)	Describe how to recursively compute the filtered marginal in a Hidden Markov Model using the forwards algorithm.	<i>CO2</i>	<i>PO1</i>	08
	c)	Discuss the types of inference problems for temporal models.	<i>CO1</i>	<i>PO1</i>	06
