

U.S.N.

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

Autonomous Institute Affiliated to VTU

April 2025 Semester End Make-Up Examinations

Programme: B.E.

Semester: VII

Branch: Artificial Intelligence & Machine Learning

Duration: 3 hrs.

Course Code: 24AM7PCGAL

Max Marks: 100

Course: Generative AI with Large Language Models

- 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	CO	PO	Marks
	1	a)	Given a latent variable $z \sim N(\mu, \sigma^2)$, compute the KL divergence between the approximate posterior $q(z x)$ and the prior $p(z) \sim N(0,1)$.	CO1	PO2	06
		b)	Are Generative models deterministic or probabilistic in nature? Explain.	CO2	PO1	06
		c)	Given: Input (X): [0.8, 0.7, 0.6, 0.9] Encoder Weights (W_e): [[0.2, 0.1], [0.4, 0.3], [0.1, 0.2], [0.3, 0.4]] Encoder Biases (b_e): [0.05, 0.1] Decoder Weights (W_d): [[0.3, 0.2, 0.1, 0.4], [0.4, 0.3, 0.2, 0.1]] Decoder Biases (b_d): [0.1, 0.05, 0.2, 0.15] Illustrate the encoder-decoder process of an Autoencoder by i. Constructing a latent space representation Z of the given input vector X. ii. Reconstructing the input vector 'X' using results of (i).	CO3	PO3	08
			OR			
	2	a)	For a VAE with latent variable $z \sim N(\mu, \sigma^2)$, derive the gradient of the reparameterization trick $z = \mu + \sigma \cdot \epsilon$, where $\epsilon \sim N(0,1)$. Explain its significance in training.	CO1	PO3	06
		b)	Differentiate between Generative and Discriminative models.	CO1	PO2	06
		c)	Given mean(x)=5, mean(y)=6.1, stddev(x)=1.2, stddev(y)=0.9, random sample for x: $\epsilon_x=0.3$, random sample for y: $\epsilon_y=0.5$. Calculate latent space variable (z) and the reconstructed data points (\hat{x} , \hat{y}) of a VAE model.	CO3	PO3	08
			UNIT - II			
	3	a)	Using any 2 specific applications in finance, explain how does BloombergGPT model exhibit better performance in financial contexts compared to general-purpose models.	CO2	PO1	05

	b)	Illustrate the transformer architecture in detail.	CO1	PO2	05															
	c)	Using the Layer Normalization parameters: Scaling Factor(γ)=1, shifting Factor(β)=0, apply Layer Normalization to the input vector [2,4,6] and determine the transformed output vector.	CO3	PO3	10															
		OR																		
4	a)	Differentiate between one-shot learning and zero-shot learning.	CO1	PO2	06															
	b)	Given a transformer model with $d_{\text{model}}=512$, calculate the total number of parameters in a single multi-head attention layer with 8 attention heads.	CO2	PO3	06															
	c)	Apply tokenization, embedding ($d_{\text{model}} = 6$, row major embedding from 0.1 to 3.6) and positional encoding processes of a Transformer Model on the sentence “Deep learning works.(fullstop)” and Compute the Final Embedded Tokens matrix.	CO3	PO3	08															
		UNIT - III																		
5	a)	In Reinforcement Learning with Human Feedback (RLHF), given a sequence of 3 actions with rewards [8, 12, 20] and a discount factor of 0.85, calculate the discounted cumulative reward for the sequence.	CO1	PO3	08															
	b)	Derive the mathematical framework of scaling instruct models.	CO3	PO3	06															
	c)	Explain Bilingual Evaluation Understudy (BLEU) metric in detail.	CO2	PO1	06															
		OR																		
6	a)	Determine the computational cost in FLOPs for multi-task fine-tuning with 100 million parameters, 500,000 examples, 104 operations per forward pass, and 10 epochs.	CO1	PO2	07															
	b)	Explain Rouge metric in detail.	CO2	PO1	06															
	c)	Is it possible to optimize the objective function in soft prompts? Justify using required mathematical representations of the procedure.	CO2	PO2	07															
		UNIT - IV																		
7	a)	Elaborate on the reward model construction process and associate math behind it.	CO1	PO2	10															
	b)	A language model is predicting the next word in a sequence. The true distribution P and the model's predicted distribution Q for the next word are: <table border="1"><thead><tr><th>Word</th><th>$P(w)$</th><th>$Q(w)$</th></tr></thead><tbody><tr><td>"apple"</td><td>0.4</td><td>0.5</td></tr><tr><td>"banana"</td><td>0.3</td><td>0.2</td></tr><tr><td>"cherry"</td><td>0.2</td><td>0.2</td></tr><tr><td>"date"</td><td>0.1</td><td>0.1</td></tr></tbody></table> Calculate the KL Divergence $D_{KL}(P Q)$ between the true distribution P and the model's distribution Q .	Word	$P(w)$	$Q(w)$	"apple"	0.4	0.5	"banana"	0.3	0.2	"cherry"	0.2	0.2	"date"	0.1	0.1	CO3	PO3	10
Word	$P(w)$	$Q(w)$																		
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		OR																		
8	a)	Elaborate the process of obtaining feedback from humans and integrating into Machine Learning.	CO1	PO2	10															

		b)	<p>A large language model is employed to solve a real-world problem using Chain-of-Thought (CoT) reasoning. The scenario is: A customer purchases 3 pens and 2 notebooks.</p> <ul style="list-style-type: none"> Each pen costs \$2. Each notebook costs \$5. The customer pays with a \$20 bill. <p>The language model's reasoning process to calculate the change is:</p> <ul style="list-style-type: none"> Calculate the total cost of pens. Calculate the total cost of notebooks. Add the two totals to get the total cost. Subtract the total cost from \$20 to find the change. <ol style="list-style-type: none"> Analyse the accuracy of the model's reasoning process. Apply the given steps to solve the problem and determine the customer's change. How does Chain-of-Thought (CoT) reasoning help to improve logical consistency and accuracy in solving the stated scenario of multi-step problems? 	CO3	PO3	10
			UNIT - V			
	9	a)	<p>A Deep Convolutional Generative AI Network (DCGAN) is trained to generate 100×100 grayscale images. The Generator (G) takes a 200-dimensional noise vector (z) as input, and the Discriminator (D) classifies the generated images as fake or real.</p> <ol style="list-style-type: none"> If the Discriminator (D) outputs a probability of 0.6 for the generated image being real, calculate the Generator Loss. If the Discriminator classifies real images correctly with a probability of 0.95, and the fake images are classified correctly with a probability of 0.7, calculate the Discriminator Loss. 	CO3	PO3	08
		b)	Elaborate on any 2 variations available for realization of image synthesis application of a Generative Adversarial Networks (GANs).	CO1	PO2	06
		c)	With the help of mathematical representations, explain the maximization step of a GAN training process.	CO2	PO4	06
			OR			
	10	a)	<p>Given:</p> <p>Conditional GAN (CGAN):</p> <ul style="list-style-type: none"> Discriminator Loss (L_D): 0.3 for the real images. Generator Loss (L_G): 0.8 for the generated images. <p>Wasserstein GAN (WGAN):</p> <ul style="list-style-type: none"> Discriminator Loss (L_D): 0.2 for the real images. Generator Loss (L_G): 0.7 for the generated images. <ol style="list-style-type: none"> Interpret the Discriminator Loss and Generator Loss scores of both the models in terms of training behavior for both models. Given that both models have 100 neurons in the first layer, and the generator in both cases has a latent vector of size 100 and a class label vector of size 3, total number of biases as 100, calculate the number of parameters in the conditional input layer for both CGAN and WGAN. 	CO3	PO3	08

		b)	With the help of mathematical representations, explain the minimization step of a GAN training process.	CO2	PO4	06
		c)	Illustrate GAN architecture in detail.	CO1	PO2	06

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