

<u>Sample Question Format</u> (For all courses having end semester Full Mark=50)

KIIT Deemed to be University Online End Semester Examination(Spring Semester-2021)

Subject Name & Code: Machine Learning CS3035

Applicable to Courses: B.Tech (IT)

<u>Full Marks</u>=50 <u>Time:</u>2 Hours

SECTION-A (Answer All Questions. Each question carries 2 Marks)

Time:30 Minutes

(7×2=14 Marks)

Question	Question	Question	<u>co</u>	Answer Key
<u>No</u>	<u>Type</u> (MCQ/SAT)		<u>Mapping</u>	(For MCQ Questions
	(MCQ) SITTY			only)
Q.No:1	MCQ	Which of the following outcome is odd man out in the below list? A) R Squared B) RMSE C) Kappa D) All of the mentioned	1	C
		Point out the wrong statement. A) Regression through the origin yields an equivalent slope if you center the data first B) Normalizing variables results in the slope being the correlation C) Least squares is not an estimation tool D) None of the mentioned	1	С
		Which of the following implies no relationship with respect to correlation? A) Cor(X, Y) = 0 B) Cor(X, Y) = 1 C) Cor(X, Y) = 2 D) All of the mentioned	1	A
		Residual plots investigate normality of the errors. A) RR B) QQ C) PP D) None of the mentioned	1	В
Q.No:2	MCQ	Which of the following is the correct formula for total variation? A) Total Variation = Residual Variation – Regression Variation	2	В

Т		D) m + 1 x ' + ' D + ' 1 1	Т	
		B) Total Variation = Residual Variation + Regression Variation		
		C) Total Variation = Residual		
		Variation * Regression Variation		
		D) All of the mentioned		
		Which of the following is required	2	D
		by K-means clustering?	_	_
		A) defined distance metric		
		B) number of clusters		
		C) initial guess as to cluster		
		centroids		
		D) All of these		
		Minimizing the likelihood is the	2	A
		same as maximizing -2 log		
		likelihood.		
		A) True		
		B) False		
		Let us say that we have computed	2	В
		the gradient of our cost function		
		and stored it in a vector g. What is		
		the cost of one gradient descent		
		update given the gradient?		
		(A) O(N) (B) O(D)		
		(C) O(ND)		
		(D) O(ND ₂)		
Q.No:3	MCQ	Consider a linear-regression model	3	A
<u> </u>	Meq	with $N = 3$ and $D = 1$ with	3	11
		input-ouput pairs as follows: y1 =		
		22, x1 = 1, y2 = 3, x2 = 1, y3 = 3, x3		
		= 2. What is the gradient of		
		mean-square error (MSE) with		
		respect to $\beta 1$ when $\beta 0 = 0$ and $\beta 1 =$		
		1? Give your answer correct to two		
		decimal digits.		
		A) -1.66		
		B) 1.66		
		C) 1.39		
		D) None of these		D
		Gradient of a continuous and	3	D
		differentiable function		
		(A) is zero at a minimum		
		(B) is non-zero at a maximum		
		(C) decreases as you get closer to		
		the minimum		
		(D) All of these		
		, , , , , , , , , , , , , , , , , , , ,		
		Which of the following sentence is	3	D
		FALSE regarding regression?	-	
		(A) It relates inputs to outputs.		
		(B) It is used for prediction.		
		(C) It may be used for		
		interpretation.		
		(D) It discovers causal		

		relationships		
		relationships For the one-parameter model, mean-Square error (MSE) is defined as follows: $\frac{1}{2N}\sum_{i=1}^{N}(y_n - \beta_0)^2$). We have a half term in the front because, (A) scaling MSE by half makes gradient descent converge faster. (B) presence of half makes it easy	3	С
		to do grid search. (C) it does not matter whether half is there or not. (D) none of the above		
Q.No:4	MCQ	Lasso can be interpreted as least-squares linear regression where (A) weights are regularized with the L1 norm (B) weights are regularized with the L2 norm (C) the weights have a Gaussian prior (D) the solution algorithm is simpler	3	A
		Regarding bias and variance, which of the follwing statements are true? (Here 'high' and 'low' are relative to the ideal model.) (A) Models which overfit have a high bias. (B) Models which overfit have a low bias. (C) Models which underfit have a high variance. (D) All of these	3	В
		In K-fold cross-validation, K is (A) A float (decimal) value (B) An integer value (C) An complex imaginary value (D) None of these	3	В
		The second principal (PC2) component is to the first principal component (PC1) A) Orthogonal B) Inverse C) Transpose D) None of these	3	A
Q.No:5	MCQ	What is the cosine similarity between [4, 3, 3, 5], and [2, 0,	1	A

		70	1	
		0,0]?.		
		A) 0.52		
		B) 0.62		
		C) 0.72		
		D) 0.74		
		Find the odd out from the following	1	С
		list: Genetic Algorithm (GA),	1	C
		Particle Swarm Optimization		
		(PSO), Stochastic Gradient Descent		
		(SGD), and Gravitational Search		
		Algorithm (GSA).		
		A) GA		
		B) PSO		
		C) SGD		
		D) None of these		
		What is the cosine similarity	1	D
			1	D
		between [5, 2, 0, 5], and [2, 0,		
		0, 0]		
		A) 0.58		
		B) 0.55		
		C) 0.75		
		D) 0.68		
		The Manhattan distance between	1	С
		two points (10, 10) and (30,30) is:		
		A) 20		
		B) 30		
		C) 40		
		D) 50		
Q.No:6	MCQ	Suppose we train a hard-margin	5	С
<u>V.110.0</u>	MCQ	linear SVM on n > 100 data points	3	C
		in , yielding a hyperplane with exactly 2 support vectors. If we add		
l		L PASCEIN SITUDONT MECTORS IT WE SIGHT		
		one more data point and retrain		
		one more data point and retrain the classifier, what is the maximum		
		one more data point and retrain the classifier, what is the maximum possible number of support vectors		
		one more data point and retrain the classifier, what is the maximum possible number of support vectors for the new hyperplane (assuming		
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		one more data point and retrain the classifier, what is the maximum possible number of support vectors for the new hyperplane (assuming the n + 1 points are linearly		
		one more data point and retrain the classifier, what is the maximum possible number of support vectors for the new hyperplane (assuming the n + 1 points are linearly separable)?		
		one more data point and retrain the classifier, what is the maximum possible number of support vectors for the new hyperplane (assuming the n + 1 points are linearly separable)? (A)2		
		one more data point and retrain the classifier, what is the maximum possible number of support vectors for the new hyperplane (assuming the n + 1 points are linearly separable)? (A)2 (B)3		
		one more data point and retrain the classifier, what is the maximum possible number of support vectors for the new hyperplane (assuming the n + 1 points are linearly separable)? (A)2 (B)3 (C)(n+1)		
		one more data point and retrain the classifier, what is the maximum possible number of support vectors for the new hyperplane (assuming the n + 1 points are linearly separable)? (A)2 (B)3 (C)(n+1) (D)n		C
		one more data point and retrain the classifier, what is the maximum possible number of support vectors for the new hyperplane (assuming the n + 1 points are linearly separable)? (A)2 (B)3 (C)(n+1) (D)n A valid Kernel function follows	5	C
		one more data point and retrain the classifier, what is the maximum possible number of support vectors for the new hyperplane (assuming the n + 1 points are linearly separable)? (A)2 (B)3 (C)(n+1) (D)n A valid Kernel function followscondition.	5	C
		one more data point and retrain the classifier, what is the maximum possible number of support vectors for the new hyperplane (assuming the n + 1 points are linearly separable)? (A)2 (B)3 (C)(n+1) (D)n A valid Kernel function follows condition. A) Vornoi	5	C
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		D) None of these		
		multipliers used to integrate the inequality constraints into the main objective function. A) Euler B) Lagrangian C) Lypanov D) None of these	5	В
Q.No:7	MCQ	Neural networks (A) optimize a convex cost function (B) can be used for regression as well as classification (C) always output values between 0 and 1 (D) None of these	6	В
		Sigmoidal activation function maps the neuron output between A) -1 to +1 B) 0 to 1 C) Either 0 or 1 D) None of these	6	В
		A single input and output nuron has an input of 2, a weight of 6 and a bias of -5.5. What will be the output if the activation function is bipolar sigmoid (where, slope s=0.6 and round off 4 decimal places) A) 0.9802 B) 0.7806 C) 0.9881 D) None of these	6	A
		In a multi-layer neural network, the term "multi" suggests A) Only one hidden layer in between input and output layers B) One or more hidden layer(s) in between input and output layers C) Always more than one hidden layers in between input and output layers D) None of these	6	В

Question No		CO Mapping				
4.0220			4	estion		(Each question should be from the same CO(s))
Q.No:8					"Naive"? State its	CO ₃
	advantages a B) Explain (OAA) and O C) Derive th following toy Example N					
			lolor Typ Red Spo			
	2		Red Spo Red Spo			
	3		Red Spo			
	4		ellow Spo			
	5	- 1	ellow Spo			
	6	- 1	ellow SU			
	7		ellow SU			
	8 9		ellow SU			
	9 Red SUV 10 Red Sports			1		
		Bayes Classifier.				
	A) What is so classifier? Ex B) Find the contract trees for the contract trees.					
	Canalan					
	Gender	ership		Level	(Class)	
	Male	0	Cheap	Low	Bus	
	Male	1	Cheap	Medium	Bus	
	Female	1	Cheap	Medium	Train	
	Female	0	Cheap	Low	Bus	
	Male	1	Cheap	Medium	Bus	
	Male	0	Standard	Medium	Train	
	Female	1	Standard	Medium	Train	
	Female	1	Expensive	High	Car	
	Male	2	Expensive	Medium	Car	
	Female	2	Expensive	High	Car	
	C) Explain t Matrix, Accu the Curve (A					
	A) Explain t	: One-Against-All				
	(OAA) and O B) States the					
		ons used in SVM				

	classifier with appropriate equations. [4]
	C) The merits and demerits of the SVM classifier. [4]
Q.No:9	A) What is "curse of dimensionality"? State the possible CO5
	remedies. [2+1=3]
	B) What is a Covariance Matrix? State its limitation. [2+1=3]
	C) Using PCA, reduce the Dimension of the given data
	from 2 to 1. [6]
	Feature /Example Ex1 Ex2 Ex3 Ex4
	Feature /Example Ex1 Ex2 Ex3 Ex4
	F2 5 9 16 7
	A) Differentiate between lossy and lossless feature/attribute reduction with suitable examples. [3]
	B) What is a Covariance Matrix? State its limitation. [2+1=3]
	C) Explain the Principal Component Analysis (PCA) and
	reduce the following dataset step-by-step from 2 dimensions
	to 1. [6]
	Feature /Example Ex1 Ex2 Ex3 Ex4
	F1 6 -3 -2 7
	F2 -4 5 6 -3
	A) Differentiate between lossy and lossless feature/attribute
	reduction with suitable examples. [2]
	B) What is a principal component? Explain the
	(mathematical) relationship between the first and the second
	principal components. [2+2=4] C) Using PCA, reduce the Dimension of the given data
	from 2 to 1. [6]
	110111 2 to 1. [0]
	Feature /Example Ex1 Ex2 Ex3 Ex4
	F1 12 4 8 17
	F2 5 9 16 7
Q.No:10	A) What are over-fitting and under-fitting? Explain them with CO2, CO5
	suitable examples. [4]
	B) Explain the Elbow technique to determine sn appropriate
	"K" value in K-NN classifier. [2]
	C) What are the intra-cluster and inter-cluster distances?
	Using K-means Clustering Algorithm, cluster the following
	data points: P1=(75,102), P2=(201,16), P3=(68, 80),
	P4=(188,36), P5=(165,55) and P6=(100,42). where, K=2, and use Euclidean distance for the purpose. (<i>start the</i>
	computation with P2 and P3 as the two initial centroids
	points) [1+5=6]
	A) What is the significance of bias in the decision boundary?
	Explain the Bias-Variance relationship in machine learning
	with suitable diagrams. [2+2=4]
	B) Why does we use the term "regression" in the Logistic
	Regression classification algorithm. [2]
	C) Using KNN algorithm and the given data set, predict the
	class label of the test data point (16,8), where K=3 and
	Euclidean distance. [6]
	X Y Label
	10 05 0

	6.5 11 1	
	7 15 1	
	12 05 0	
	8 10 1	
	15 8 0	
	A. What are the over-fitting and the under-fitting? Explain	
	them with suitable examples. [3]	
	B. What are the intra-cluster and inter-cluster distances?	
	Explain the Elbow technique to determine an appropriate "K"	
	value in K-NN classifier. [1.5+1.5=3]	
	C. Using K-means Clustering Algorithm, cluster the following	
	data points: P1=(75,102), P2=(201,16), P3=(68, 80),	
	P4=(188,36), P5=(165,55) and P6=(100,42). where, K=2, and	
	use Euclidean distance for the purpose. (start the	
	computation with P1 and P3 as the two initial centroids	
	points) [6]	
Q.No:11	A) What is Binary Activated Neuron model? States its	CO6
	limitations. [4]	
	B) A two input single output neuron model has weights value	
	[1.5 -2.1] and bias of -3. It is given an input [2.0 2.5]. What	
	will be the output if the binary step function threshold=1 is	
	used? [4]	
	C) Describe Multi-layer Perceptron Neural Network with	
	suitable mathematical expressions and diagram. [4]	
	A) What is perceptron? Explain it with an example. [4]	
	B) Differentiate between linearly and non-linearly separable	
	datasets. [2]	
	C) Solve XOR problem with a two input artificial neuron	
	model. [6]	
	A) What are the learning rate and the momentum in Artificial	
	Neural Network (ANN) model? State different learning rules	
	used in ANN. [2+2=4]	
	B) Draw a diagram of a multiple input single output artificial	
	neural network and compute its input-output relationship. [4]	
	C) Describe the Backpropagation algorithm using appropriate	
	mathematical expressions. [4]	