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VI SEMESTER B.TECH. (COMPUTER SCIENCE & ENGG) MAKEUP EXAMINATIONS, JUNE 2019

SUBJECT: MACHINE LEARNING [CSE 4010]

REVISED CREDIT SYSTEM (18/06/2019)

Time: 3 Hours MAX. MARKS: 50

Instructions to Candidates:

- ❖ Answer **ALL** the questions.
- Missing data may be suitably assumed.

 1A. Explain the terms ill-posed problem, inductive bias, model selection, generalization. 1B. What is meant by an association rule? Explain. Also explain the three measures that are frequently calculated in learning association rules. 1C. List and explain any two applications of machine learning. 2A. Explain maximum a posteriori estimate and Bayes' estimator. Under which assumption do these estimates give correct values? 2B. What is meant by imputation? Explain the different types of imputation. 3 Imagine our hypothesis is not one rectangle but a union of two (or m > 1) rectangles. What is the advantage of such a hypothesis class? Show that any class can be represented by such a hypothesis class with large enough m. 3A. Show the perceptron that calculates the odd parity of its three inputs. 5 3B. Write the online k-means algorithm and explain. Also explain any two methods to avoid centres that are there but not effectively utilised. 3C. In a two-class problem, the log odds is defined as log (P(C₁ x) / P(C₂ x)). Write the discriminant function in terms of the log odds. 4B. Explain gradient descent. Also explain where it is used. 3B. What do you mean by linearly separable? With an example, explain the working principle of support vector machines. 			
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	4B		3

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4C	Explain the different types of hierarchical clustering. For the two dimensional dataset shown in figure 4C, use any clustering method & draw the resulting dendrogram. b a Figure 4C	4
5A.	How is the goodness of a split measured in a decision tree for regression? Explain & write the expression for this. Derive the expression for the error after the split.	5
5B.	What is meant by a first-order Markov model? Draw a markov model with three states and explain.	3
5C.	Explain graphical model with an example.	2

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