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| SEMESTER – VII Subject Code 15CS73 IA Marks 26 Number of Lecture Hours/Week 03 Exam Marks 87 Total Number of Lecture Hours 50 Exam Hours 03 CREDITS – 04 Course Objectives: This course will enable students to • Define machine learning and problems relevant to machine learning. • Differentiate supervised, unsupervised and reinforcement learning • Apply neural networks, Bayes classifier and k nearest neighbor, for problems machine learning. • Perform statistical analysis of machine learning techniques. Module – 1 Introduction: Well posed learning problems, Designing a Learning system, Perspective and Issues in Machine Learning. Concept Learning: Concept learning task, Concept learning as search, Find-S algorithm, Version space, Candidate Elimination algorithm, Inductive Bias. Text Book1, Sections: 1.1 – 1.3, 2.1-2.5, 2.7 Module – 2 Decision Tree Learning: Decision tree representation, Appropriate problems for decision tree learning, Basic decision tree learning algorithm, hypothesis space search in decision tree learning, Inductive bias in decision tree learning. Text Book1, Sections: 3.1-3.7 Module – 3 Artificial Neural Networks: Introduction, Neural Network representation, Appropriate problems, Perceptrons, Backpropagation algorithm. Text book 1, Sections: 4.1 – 4.6 Module – 4 Bayesian Learning: Introduction, Bayes theorem, Bayes theorem and concept learning, ML and LS error hypothesis, ML for predicting probabilities, MDL principle, Naive Bayes classifier, Bayesian belief networks, EM algorithm Text book 1, Sections: 6.1 – 6.6, 6.9, 6.11, 6.12 Module – 5 Evaluating Hypothesis: Motivation, Estimating hypothesis accuracy, Basics of 12 sampling theorem, General approach for deriving confidence intervals, Difference in | | |
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| principle, Naive Bayes classifier, Bayesian belief networks, EM algorithm Text book 1, Sections: 6.1 – 6.6, 6.9, 6.11, 6.12 Module – 5 Evaluating Hypothesis: Motivation, Estimating hypothesis accuracy, Basics of sampling theorem, General approach for deriving confidence intervals, Difference in | o Hour | |
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| Evaluating Hypothesis: Motivation, Estimating hypothesis accuracy, Basics of sampling theorem, General approach for deriving confidence intervals, Difference in | | |
| sampling theorem, General approach for deriving confidence intervals, Difference in | 2 TT | |
| | 2 Hour | |
| error of two hypothesis, Comparing learning algorithms. | | |
| Instance Based Learning: Introduction, k-nearest neighbor learning, locally | | |
| weighted regression, radial basis function, cased-based reasoning, | | |
| Reinforcement Learning: Introduction, Learning Task, Q Learning | | |
| Text book 1, Sections: 5.1-5.6, 8.1-8.5, 13.1-13.3 | | |
| Course Outcomes: After studying this course, students will be able to | | |

Identify the problems for machine learning. And select the either supervised,

unsupersvised or reinforcement learning.

- Explain theory of probability and statistics related to machine learning
- Investigate concept learning, ANN, Bayes classifier, k nearest neighbor, Q,

Question paper pattern:

The question paper will have ten questions.

There will be 2 questions from each module.

Each question will have questions covering all the topics under a module.

The students will have to answer 5 full questions, selecting one full question from each module.

Text Books:

1. Tom M. Mitchell, Machine Learning, India Edition 2013, McGraw Hill Education.

Reference Books:

1. Trevor Hastie, Robert Tibshirani, Jerome Friedman, h The Elements of Statistical Learning, 2nd edition, springer series in statistics.

2. Ethem Alpaydın, Introduction to machine learning, second edition, MIT press.

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