CS/IT323 - MACHINE LEARNING

Lectures	:	3 periods/week	Internal Marks	:	30
Tutorial:	:	0 period/week	Semester End Examination Marks	:	70
Sem End Exam Duration	:	3 Hours	Credits		3

Course Objectives:

The main objectives of this course are to:

- 1. Basic concepts and applications of machine learning.
- 2. Supervised learning and its applications
- 3. Unsupervised learning and its applications
- 4. Multilayer perceptions and kernel tricks

Course Outcomes:

After successful completion of the course, the students are able to:

- 1. Apply the machine learning concepts in real life problems
- 2. Design solutions for supervised learning problems
- 3. Use rule sets and reinforcement learning to solve real world problems
- 4. Discuss the issues in dimensionality reduction and unsupervised learning algorithms.

Course Content:

UNIT I CO1 12 Periods

Introduction: Well posed learning problems, Designing a Learning System, Perspectives and Issues in machine learning.

Concept Learning and general to specific ordering: concept learning Task , Concept learning as a search, Finding a Maximally Specific Hypothesis , Version Spaces and Candidate Elimination Algorithm, Remarks on Version space and candidate elimination. **Bayesian Learning:** Bayes Theorem, Maximum Likelihood and Least Square Error Hypotheses, Bayes Optimal Classifier, Naïve-Bayes Classifier, Bayesian Belief Network.

UNIT II CO2 12 Periods

Decision Tree Learning : Decision Tree Representation, appropriate problems for decision tree, the basic decision tree Algorithm, Issues in decision tree learning. **Artificial Neural Networks:** Introduction, Neural Network Representation, appropriate problems for neural network, Perceptrons , Multilayer Networks and the Back Propagation Algorithm. **Instance Based Learning:** Introduction, KNN Learning, Locally Weighted Regression , Radial Bias Functions, Case-Based Reasoning.

UNIT III CO3 12 Periods

Learning Sets of Rules: Sequential Covering Algorithm , Learning Rule Sets: summary , Learning First Order Rules, Learning set of first order rules: FOIL.

Reinforcement Learning: Introduction, the Learning Task , Q Learning , Non Deterministic Rewards and Actions , Temporal Difference Learning , Generalizing from Examples , Relationship to Dynamic Programming

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Dimensionality Reduction : Introduction, subset selection, Principal component analysis, Feature Embedding, Factor analysis, Singular Value Decomposition and Matrix factorization, Multidimensional Scaling, Linear Discriminant analysis, Canonical correlation analysis. **Clustering:** Introduction, Mixture Densities, K-Means Clustering, Expectation-Maximization Algorithm, Mixtures of Latent Variable Models, Supervised Learning after Clustering, Spectral Clustering, Hierarchical Clustering, Choosing the Number of Clusters.

Learning Resources:

Text Books:

- 1. Tom M. Mitchell, —Machine Learning, McGraw-Hill Education (India) Private Limited, 2013. (UNIT I , UNIT II, and UNIT III)
- 2. EthemAlpaydin, Introduction to Machine Learning , MIT Press, Prentice Hall of India, Third Edition 2014. (UNIT IV)

Reference Books:

- 1. Stephen Marsland, —Machine learning: An Algorithmic Perspective||, CRC Press, 2009
- 2. Machine Learning: a Probabilistic Perspective, Kevin P. Murphy, MIT Press, 2012
- 3. Foundations of Machine Learning, MehryarMohri, AfshinRostamizadeh and Ameet Talwalkar, MIT Press, 2012.
- 4. Machine Learning -The Art and Science of Algorithms that Make Sense of Data, Peter Flach, Cambridge

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