Course Objective: 1. To understand various key paradigms for machine learning approaches. 2. To familiarize with the mathematical and statistical techniques used in machine learning. 3. To understand and differentiate among various machine learning techniques.  S. NO.  Course Outcomes (CO)  CO1  Understand the fundamental concepts and algorithms of machine learning  Develop a comprehensive understanding of fundamental machine learning concepts, algorithms, and techniques, including supervised and unsupervised learning, classification, regression, clustering, and dimensionality reduction.  CO3  Apply principles and algorithms to evaluate models generated from data  CO4  Learn to critically evaluate the performance of machine learning models using appropriate metrics  Develop the ability to identify and formulate problems suitable for machine learning solutions, desig appropriate models, and interpret results in practical applications.  S. NO.  Contents  UNIT 1  Introduction to Machine Learning: Overview of different tasks: classification, regression, clustering, Concept of learning, Types of the Machine Learning, Data Table, Information System, Data Representation, diversity of data, Basic Linear Algebra and Probabolity Theory, Optimization: Maximum likelihood, Expectation maximization, Gradient descent, Bias-Variance Tradeoff, Metrics to Evaluate Classification and Regression models  Supervised Learning: Linear Regression, Logistic Regression, Baysian Decision Theory, Naïve Bayes, K-Nearest Neighbour, Support Vector Machine, Decision trees, Ensemble Classifier, Random Forest, Linear Classifiers and Kernels, Neural Networks, Deep Neural Network, Fundametals of Deep Learning: DNN, CNN.  Unsupervised Learning: Clustering, Expectation Maximization, K-Mean Clustering, Hierarchical vs Partitional Clustering, Gaussian Mixture Model, Dimensionality Aleuction, Feature Selection, PCA, factor analysis, manifold learning.		<b>.</b>	3	0	2	Algebra		
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TOTAL 42	UNIT 3	Hierarchical vs Partitional Clustering, Gaussian Mixture Model, Dimensionality					14	
		TOTAL					42	

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**Machine Learning**