

Course Category:	ProgrammeCore						Credits:						4		
Course Type:	Theory						Lecture-Tutorial-Practice:						3-0-2		
Prerequisites:	---						Continuous Evaluation:						30		
							Semester end Evaluation:						70		
							Total Marks:						100		
Course Outcomes	Upon successful completion of the course, the student will be able to:														
	CO1	Understand the fundamental concepts of machine learning													
	CO2	Apply linear, distance based, and decision tree based models													
	CO3	Analyze probabilistic, neural network models													
	CO4	Design a suitable machine learning model for a given scenario													
Contribution of Course Outcomes towards achievement of Program Outcomes (1-Low, 2-Medium, 3- High)		PO 1	PO 2	PO 3	PO 4	PO 5	PO 6	PO 7	PO 8	PO 9	PO 10	PO 11	PO 12	PSO 1	PSO 2
	CO1													1	
	CO2	2	2											3	1
	CO3	2	2											3	1
	CO4	2	3											3	2
Course Content	UNIT I The ingredients of machine learning: Tasks, Models, Features Binary classification and related tasks: Classification, Assessing classification performance, Visualizing classification performance Beyond binary classification: Multi-class classification, Regression, Unsupervised and descriptive learning														
	UNIT II Decision Tree learning – Introduction, Decision tree representation, Appropriate problems for decision tree learning, The basic decision tree learning algorithm, Inductive bias in decision tree, Issues in decision tree learning. Linear models: The least-squares method, Multivariate linear regression, The perceptron, Support vector machines, Soft margin SVM, Going beyond linearity with kernel methods.														
	UNIT III: Distance Based Models: Introduction, Neighbours and exemplars, Nearest Neighbours classification, K-Means algorithms, Clustering around medoids Probabilistic Models: Using Naïve Bayes Model for classification, Expectation Maximization, Gaussian Mixture models														
	UNIT IV: Artificial Neural Networks: Introduction, Neural network representation, appropriate problems for neural network learning, Multilayer networks and the														

	<p>back propagation, Advanced topics in Artificial Neural Networks</p> <p>Reinforcement Learning: Introduction, Learning tasks, Q-learning.</p>
Text books and Reference books	<p>Text Book(s):</p> <p>[1].Machine Learning: The art and Science of algorithms that make sense of data, Peter Flach, Cambridge University Press, 2012</p> <p>[2].Tom M. Mitchell, Machine Learning, India Edition 2013, McGraw Hill Education</p> <p>Reference Books:</p> <p>[1] AurélienGéron, Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow: Concepts, Tools, and Techniques to Build Intelligent Systems 2nd Edition</p> <p>[2] Stephen Marsland, “Machine Learning – An Algorithmic Perspective”, Second Edition, Chapman and Hall/CRC Machine Learning and Pattern Recognition Series, 2014</p> <p>[3] EthemAlpaydın, Introduction to machine learning, second edition, MIT press.</p> <p>[4] T. Hastie, R. Tibshirani and J. Friedman, “Elements of Statistical Learning”, Springer Series , 2nd edition</p>
E-resources and other digital material	<p>[1]. Kevin Murphy,“MachineLearning:AProbabilisticPerspective” , MIT Press, 2012, https://www.cs.ubc.ca/~murphyk/MLbook/pml-intro-5nov11.pdf</p> <p>[2] Machine Learning by Andrew Ng, Stanford University https://www.coursera.org/learn/machine-learning</p> <p>[3] Professor S. Sarkar IIT Kharagpur “Introduction to machine learning” , https://www.youtube.com/playlist?list=PLYihddLF-CgYuWNL55Wg8ALkm6u8U7gps</p> <p>[4] Professor Carl GustafJansson, KTH, Video Course on Machine Learning https://nptel.ac.in/noc/individual_course.php?id=noc19-cs35</p> <p>[5]. <u>Tom Mitchell</u>, “Machine Learning”, http://www.cs.cmu.edu/~tom/10701_sp11/lectures.shtml</p>