

**CS 6301****MACHINE LEARNING****OBJECTIVES:**

- To understand the need for machine learning for various types of problem solving
- To know the mathematics involved in various machine learning algorithms
- To study the various supervised, semi-supervised and unsupervised learning algorithms in machine learning
- To learn about probabilistic models in machine learning
- To have a glimpse of the latest developments in machine learning

CS 6301                      MACHINE LEARNING		L	T	P	EL	TOTAL CREDITS
		3	0	4	3	6
<b>MODULE I :</b>		<b>L</b>	<b>T</b>	<b>P</b>	<b>EL</b>	
		6	0	4	3	
Learning – Types of Machine Learning – Supervised Learning - The Brain and the Neuron – Design a Learning System – Perspectives and Issues in Machine Learning- Concept Learning task – Concept Learning as Search - Finding a Maximally Specific Hypothesis – Version Spaces and the Candidate Elimination Algorithm						
<b>SUGGESTED ACTIVITIES :</b>						
<ul style="list-style-type: none"> <li>• EL – Fundamentals of Predictive Analytics, Study of tools for data mining like WEKA, KNIME, Rapidminer, etc</li> <li>• Practical – Study of tools like WEKA, KNIME and the UCI repository datasets</li> </ul>						
<b>SUGGESTED EVALUATION METHODS:</b>						
<ul style="list-style-type: none"> <li>• Tutorial problems</li> <li>• Assignment problems</li> <li>• Quizzes</li> </ul>						
<b>MODULE II :</b>		<b>L</b>	<b>T</b>	<b>P</b>	<b>EL</b>	
		3	0	4	3	
Neural Networks – Perceptron – Linear Separability – Linear Regression						
<b>SUGGESTED ACTIVITIES :</b>						
<ul style="list-style-type: none"> <li>• In-class activity – practical problems and the need for machine learning algorithms</li> <li>• EL – Working with tools and standard data sets</li> <li>• Practical - Implementation of the Candidate Elimination Algorithm</li> </ul>						
<b>SUGGESTED EVALUATION METHODS:</b>						
<ul style="list-style-type: none"> <li>• Tutorial problems</li> <li>• Assignment problems</li> <li>• Practical demonstrations</li> </ul>						
<b>MODULE III :</b>		<b>L</b>	<b>T</b>	<b>P</b>	<b>EL</b>	

	<b>3</b>	<b>0</b>	<b>4</b>	<b>3</b>
The Multi-Layer Perceptron – Back Propagation of Error-Multi-layer Perceptron in Practice – Deriving Back Propagation – Applications of MLP				
<b>SUGGESTED ACTIVITIES :</b> <ul style="list-style-type: none"> <li>• Flipped classroom and activity</li> <li>• EL – Applications of MLP</li> <li>• Practical – Implementation of the Neural Network perceptron algorithm and enhancing it to other variations</li> <li>• Proposal for Mini Project</li> </ul>				
<b>SUGGESTED EVALUATION METHODS:</b> <ul style="list-style-type: none"> <li>• Tutorial problems</li> <li>• Assignment problems</li> <li>• Approval of Mini project based on the reference papers, abstract and design</li> </ul>				
<b>MODULE IV :</b>	<b>L</b>	<b>T</b>	<b>P</b>	<b>EL</b>
	<b>3</b>	<b>0</b>	<b>4</b>	<b>3</b>
Radial Basis Function Network - Concepts –Training - Interpolation and Basis Functions – Solutions using RBF				
<b>SUGGESTED ACTIVITIES :</b> <ul style="list-style-type: none"> <li>• Flipped Class room</li> <li>• EL –Applications of RBF Networks</li> <li>• Practical – Implementation of Multi-layer Perceptron</li> </ul>				
<b>SUGGESTED EVALUATION METHODS:</b> <ul style="list-style-type: none"> <li>• Tutorial problems</li> <li>• Assignment problems</li> <li>• Practical demonstrations</li> </ul>				
<b>MODULE V :</b>	<b>L</b>	<b>T</b>	<b>P</b>	<b>EL</b>
	<b>3</b>	<b>0</b>	<b>4</b>	<b>3</b>
Dimensionality Reduction –Linear Discriminant Analysis-Principal Component Analysis-Factor Analysis-Independent Component Analysis-Locally Linear Embedding-Isomap				
<b>SUGGESTED ACTIVITIES :</b> <ul style="list-style-type: none"> <li>• EL – Probabilistic PCA and Factor analysis concepts</li> <li>• Practical –Implementation of Independent Component Analysis(ICA) algorithm</li> <li>• Practical – Mini-project design completion</li> </ul>				
<b>SUGGESTED EVALUATION METHODS:</b> <ul style="list-style-type: none"> <li>• Tutorial problems</li> <li>• Assignment problems</li> <li>• Practical demonstrations</li> </ul>				
<b>MODULE VI:</b>	<b>L</b>	<b>T</b>	<b>P</b>	<b>EL</b>
	<b>6</b>	<b>0</b>	<b>4</b>	<b>3</b>
Probabilistic Learning-Gaussian Mixture Models-Nearest Neighbor Models-Support Vector Machines-Optimal Separation-Kernels-The Support Vector Machine Algorithm-Extensions to the SVM				
<b>SUGGESTED ACTIVITIES :</b> <ul style="list-style-type: none"> <li>• EL – Application of SVM, Nearest Neighbor concepts and other regression models on various datasets</li> <li>• Practical –Implementation of Support Vector Machines with various kernel models, Nearest Neighbor models</li> </ul>				

<ul style="list-style-type: none"> <li>Continuation of mini project, minimum 40% implementation</li> </ul>				
<b>SUGGESTED EVALUATION METHODS:</b> <ul style="list-style-type: none"> <li>Tutorial problems</li> <li>Assignment problems</li> <li>Practical demonstration of algorithms and mini project</li> </ul>				
<b>MODULE VII:</b>	<b>L</b>	<b>T</b>	<b>P</b>	<b>EL</b>
	<b>3</b>	<b>0</b>	<b>4</b>	<b>3</b>
Evolutionary Learning-The Genetic Algorithm-Genetic Operators-Using Genetic Algorithms-Genetic Programming - Applications				
<b>SUGGESTED ACTIVITIES :</b> <ul style="list-style-type: none"> <li>Flipped Classroom for applications</li> <li>EL – Applications of Evolutionary algorithms</li> <li>Practical – Implementation of GA, Continuation of mini-project</li> </ul>				
<b>SUGGESTED EVALUATION METHODS:</b> <ul style="list-style-type: none"> <li>Tutorial problems</li> <li>Assignment problems</li> <li>Practical demonstrations</li> </ul>				
<b>MODULE VIII</b>	<b>L</b>	<b>T</b>	<b>P</b>	<b>EL</b>
	<b>3</b>	<b>0</b>	<b>4</b>	<b>3</b>
Reinforcement Learning – Markov Decision Processes - Values-The difference between SARSA and Q-Learning				
<b>SUGGESTED ACTIVITIES :</b> <ul style="list-style-type: none"> <li>Flipped Classroom for applications</li> <li>EL – Applications of Evolutionary algorithms</li> <li>Practical – Continuation of mini-project</li> </ul>				
<b>SUGGESTED EVALUATION METHODS:</b> <ul style="list-style-type: none"> <li>Tutorial problems</li> <li>Assignment problems</li> <li>Practical demonstrations</li> </ul>				
<b>MODULE IX</b>	<b>L</b>	<b>T</b>	<b>P</b>	<b>EL</b>
	<b>3</b>	<b>0</b>	<b>4</b>	<b>3</b>
Learning with Trees-Using Decision trees-Constructing Decision Trees-Classification and regression trees-Classification example-Decision by committee: Ensemble Learning-Boosting-Bagging-Random Forests-Different ways to combine classifiers				
<b>SUGGESTED ACTIVITIES :</b> <ul style="list-style-type: none"> <li>EL – Applications of Decision tree, CART</li> <li>Practical –Implementation of Decision Trees, Bagging, Boosting and EM algorithms</li> <li>Continuation of mini-project</li> </ul>				
<b>SUGGESTED EVALUATION METHODS:</b> <ul style="list-style-type: none"> <li>Tutorial problems</li> <li>Assignment problems</li> <li>Practical demonstrations, Mini project 80% completion</li> </ul>				
<b>MODULE X</b>	<b>L</b>	<b>T</b>	<b>P</b>	<b>EL</b>
	<b>3</b>	<b>0</b>	<b>4</b>	<b>3</b>

Unsupervised Learning-The K-Means Algorithm-Vector Quantization-The self-organizing feature map				
<b>SUGGESTED ACTIVITIES :</b> <ul style="list-style-type: none"> <li>• Combinations of in Class &amp; Flipped class rooms</li> <li>• EL –K-Means algorithm applications</li> <li>• Practical - Implementations of K-Means algorithm</li> </ul>				
<b>SUGGESTED EVALUATION METHODS:</b> <ul style="list-style-type: none"> <li>• Tutorial problems</li> <li>• Assignment problems</li> <li>• Practical demonstrations</li> </ul>				
<b>MODULE XI</b>	<b>L</b>	<b>T</b>	<b>P</b>	<b>EL</b>
	<b>3</b>	<b>0</b>	<b>4</b>	<b>3</b>
Deep learning introduction – CNN – RNN				
<b>SUGGESTED ACTIVITIES :</b> <ul style="list-style-type: none"> <li>• EL – Survey of deep learning network models</li> <li>• Practical – Mini-project demonstration</li> </ul>				
<b>SUGGESTED EVALUATION METHODS:</b> <ul style="list-style-type: none"> <li>• Mini project final evaluation</li> </ul>				

#### OUTCOMES:

Upon completion of the course, the students will be able to

- Differentiate between supervised, unsupervised, semi-supervised machine learning approaches
- Choose and implement classification or regression algorithms for an application using an open source tool
- Implement probabilistic, discriminative and generative algorithms for an application and analyze the results
- Use a tool to implement typical clustering algorithms for different types of applications
- Create potential solutions for real time applications using machine learning techniques

#### TEXT BOOKS

1. Stephen Marsland, "Machine Learning – An Algorithmic Perspective", Second Edition, Chapman and Hall/CRC Machine Learning and Pattern Recognition Series, 2014.
2. Tom M Mitchell, "Machine Learning", McGraw Hill Education, 2013.

#### REFERENCES:

1. Peter Flach, "Machine Learning: The Art and Science of Algorithms that Make Sense of Data", First Edition, Cambridge University Press, 2012.
2. Jason Bell, "Machine learning – Hands on for Developers and Technical Professionals", First Edition, Wiley, 2014.
3. Ethem Alpaydin, "Introduction to Machine Learning", Third Edition, Adaptive Computation and Machine Learning Series, MIT Press, 2014.

#### EVALUATION METHOD:

Category of Course	Continuous Assessment	Mid – Semester Assessment	End Semester

<b>Theory Integrated with Practical</b>	<b>15(T) + 25 (P)</b>	<b>20</b>	<b>40</b>
---	-----------------------	-----------	-----------

	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12
CO1	✓	✓	✓									
CO2	✓	✓	✓	✓	✓				✓			✓
CO3	✓	✓	✓	✓	✓				✓		✓	✓
CO4	✓	✓	✓	✓	✓				✓			✓
CO5	✓	✓	✓	✓	✓				✓		✓	✓

**CS6302**

## **PROGRAMMING PARADIGMS**

**Prerequisites for the course: Data Structures and Algorithms**

### **OBJECTIVES:**

- To introduce the major programming paradigms with the principles and the techniques involved in the design and implementation of modern programming languages
- To introduce the framework for specifying and reasoning about programming languages
- To analyse a given program from the perspective of good programming practices
- To compare and contrast the range of programming paradigms
- To evaluate programming language features critically with respect to the way they support good software engineering practices
- To discuss the appropriateness of the use of a given programming paradigm within a given environment

CS6302 PROGRAMMING PARADIGMS		L	T	P	EL	CREDITS
		3	0	0	3	4
<b>OBJECTIVES:</b>						
<b>MODULE I :</b>			L	T	P	EL
			3	0	0	5
The art of Language design – Programming language spectrum - Compilation and Interpretation– Evaluation of Programming languages						
<b>SUGGESTED ACTIVITIES :</b>						
<ul style="list-style-type: none"> <li>• Activity based learning - brain storming quizzes and puzzles of programming languages</li> </ul>						
<b>SUGGESTED EVALUATION METHODS:</b>						
<ul style="list-style-type: none"> <li>• Quizzes</li> </ul>						
<b>MODULE II :</b>			L	T	P	EL
			4	0	0	5
Languages – Syntax and Semantics of language C-lite - Names – Types – Type Systems - Binding – Scope – Static – Dynamic – Abstract Data types						
<b>SUGGESTED ACTIVITIES :</b>						
<ul style="list-style-type: none"> <li>• Using peer learning- Interaction and group discussion about data types</li> </ul>						