



<b>Course Name</b>	Foundations of Artificial Intelligence and Machine Learning
<b>Course Code</b>	CSE3001
<b>Credits</b>	4
<b>Prerequisites</b>	AI-ML Fundamentals
<b>L-T-P-C</b>	3-0-2-4

### **COURSE OBJECTIVES**

1. To understand the fundamentals of Machine Learning, its types, and how it relates to traditional statistical methods.
2. To explore core concepts and models of Artificial Neural Networks (ANN), including learning rules, network design, and activation mechanisms.
3. To gain foundational knowledge of Artificial Intelligence, including intelligent agents, environments, and rational decision-making processes.
4. To apply various optimization techniques, both constrained and unconstrained, in solving real-world AI/ML problems.
5. To design and implement AI and ML-based capstone projects that integrate theoretical knowledge with practical application.

### **COURSE SYLLABUS**

<b>Modules</b>	<b>No. of Lectures</b>
<b>Module-I: Fundamentals of Machine Learning:</b> – Meaning of Machine Learning, Relationship between Machine Learning and Statistical Analysis, Process of Machine Learning, Types of Machine Learning Algorithms (Supervised learning, Semi-supervised learning, Unsupervised learning, Transduction, Reinforcement learning)	6
<b>Module-II: Fundamental Models of ANN</b> – Gradients Methods, Concept of Hessian matrix, Basics of Neural networks, Basic terminologies of neural networks: Weights, Biases, Activation Functions, Threshold, Designing and programming your own neural networks, Learning rules, Hebb's Net, Perceptron	8
<b>Module-III: Introduction to Artificial Intelligence</b> – AI problems, foundation of AI and history of AI intelligent agents: Agents and Environments, the concept of rationality, the nature of environments, structure of agents, problem solving agents, problem formulation.	8

<b>Module-IV: Optimization Techniques</b> – Basic concepts of optimization, Constraint and unconstrained optimizations, Modelling of different problems using optimization techniques, Introduction and applications of non-gradient methods.	8
<b>Module-V: Projects</b> – Capstone Projects based on the learning from AI and ML	6
<b>Total Lectures</b>	36

<b>LABORATORY</b>	
<b>Sr. No.</b>	<b>Content</b>
1	Implement supervised learning (Linear Regression or Classification) using Python/Scikit-learn.
2	Apply unsupervised learning (K-Means Clustering or PCA) on a dataset.
3	Implement a basic Artificial Neural Network (ANN) from scratch.
4	Demonstrate activation functions (Sigmoid, ReLU, Tanh) with plots and analysis.
5	Design a problem-solving agent for a toy problem (e.g., vacuum cleaner agent).
6	Simulate a rational agent in a defined environment.
7	Solve a mathematical optimization problem using non-gradient-based techniques (e.g., Genetic Algorithm).
8	Apply constraint-based optimization for a real-world problem (e.g., resource allocation).
9	Mini Project: Train an ML model on real-world dataset (e.g., spam detection, movie recommendations).
10	Capstone Project: Build and deploy a small AI/ML application (e.g., chatbot, sentiment analysis, or fraud detection)

<b>COURSE OUTCOMES</b>	
At the end of the course the student will be able to:	
<b>CO1</b>	Understand the fundamentals of Machine Learning, its types, and how it relates to traditional statistical methods.
<b>CO2</b>	Explore core concepts and models of Artificial Neural Networks (ANN), including learning rules, network design, and activation mechanisms.
<b>CO3</b>	Gain foundational knowledge of Artificial Intelligence, including intelligent agents, environments, and rational decision-making processes.
<b>CO4</b>	Apply various optimization techniques, both constrained and unconstrained, in solving real-world AI/ML problems.
<b>CO5</b>	Design and implement AI and ML-based capstone projects that integrate theoretical knowledge with practical application.

CO-PO Mapping:			3: Substantial						2: Moderate					1: Slight		
COs	PO1	PO2	PO3	PO4	PO5	PO6	PO7	PO8	PO9	PO10	PO11	PO12	PSO1	PSO2		
CO1	3	2	2	2	1	-	-	-	-	-	-	-	2	2	2	
CO2	3	2	3	2	2	-	-	-	-	-	-	-	2	2	2	
CO3	2	2	2	2	1	-	-	-	-	-	-	-	2	2	2	
CO4	3	3	3	3	2	-	-	-	-	-	-	-	2	2	2	
CO5	3	3	3	3	3	-	-	-	-	-	-	-	3	3	3	

## TEXTBOOKS

1. Stuart J. Russell and Peter Norvig., *Artificial Intelligence: A Modern Approach*, Pearson
2. Chris Sebastian, *Machine Learning for Beginners*
3. Ethem Alpaydin, *Machine Learning: The New AI*

## REFERENCES

1. Tom M. Mitchell, *Machine Learning*, McGraw-Hill
2. Wolfgang Ertel, *Introduction to Artificial Intelligence*, Springer.
3. V. Subramanian, *Fundamentals of Artificial Intelligence*, Springer.
4. Suvrit Sra, Sebastian Nowozin, Stephen Wright, *Optimization for Machine Learning*, MIT Press.

## EVALUATION SCHEME

Theory: (70%)			
Continuous Assessment	Mid-term	End-term	Total
25	25	50	100

Practical: 30%				
Continuous Assessment			End-term Practical	Total
Lab Manual	Lab Assessment	Internal Viva-voce		
10	10	30	50	100