

Course Code: AI8001

Course Title: Machine Learning

Course Credits: [Lecture: 03, Tutorial: 00, Practical: 01]

Prerequisites: Basic mathematical background and knowledge of programming in Python.

Objectives: To exploit and apply machine learning theories and models.

Course Outcomes: Upon completion of the course, students shall be able to

- CO1: Express the importance of machine learning and its fundamental concepts.
- CO2: Interpret regression models and relate them with real-world applications.
- CO3: Distinguish classification, clustering, and learning algorithms, also formulate and experiment the same on real-world problems.
- CO4: Analyze the need for dimensionality reduction and apply PCA, LDA, and Kernel PCA based on the situation.

Course Objective and Course Outcomes Mapping:

- To exploit and apply machine learning theories and models: CO1, CO2, CO3, & CO4

Programme Outcomes: The student will have

PO1: Proficiency in and ability to identify problems related to computer science as well as design and apply computational knowledge to solve them.

PO2: Ability to design, develop, test and maintain system, component, product or process as per needs and specification.

PO3: Understanding of professional and ethical role and responsibility.

PO4: Recognition of the need for and an ability towards life-long learning.

PO5: Knowledge of programming languages, database systems, operating systems, software engineering, Web & Mobile technology and relevant modern issues.

PO6: Ability to demonstrate the use of modern tools, models and languages to solve problems related to software development.

PO7: An ability to communicate and present knowledge effectively.

Programme Outcomes and Course Outcomes mapping:

Course Outcomes	Programme Outcomes						
	PO1	PO2	PO3	PO4	PO5	PO6	PO7
CO1	✓						
CO2	✓	✓	✓	✓			
CO3	✓	✓	✓	✓	✓	✓	
CO4	✓	✓		✓	✓		

Curriculum Unit Titles

Units	Unit Description	Weightage
1	Introduction to Machine Learning	[15%]
	1.1. Applications of Machine Learning	
	1.2. Types of Learning – Supervised and Unsupervised	
	1.3. Training versus Testing	
	1.4. Data Processing – Missing Data, Categorical Data, Feature Scaling	

2	Regression	[20%]
2.1.	Simple Linear Regression	
2.2.	Multiple Linear Regression	
2.3.	Polynomial Regression	
2.4.	Decision Tree Algorithm	
2.5.	Random Forest Algorithm	
3	Classification	[15%]
3.1.	Logistic Regression	
3.2.	k -nearest Neighbour (kNN)	
3.3.	Support Vector Machine (SVM)	
3.4.	Naive Bayes	
4	Unsupervised Learning	[15%]
4.1.	K-Means Clustering	
4.2.	Hierarchical Clustering	
4.3.	Apriori Algorithm	
5	Reinforcement and Deep Learning	[20%]
5.1.	Upper Confidence Bound Algorithm	
5.2.	Thompson Sampling	
5.3.	Artificial Neural Network	
5.4.	Convolutional Neural Network	
6	Dimensionality Reduction	[15%]
6.1.	Principal Component Analysis (PCA)	
6.2.	Linear Discriminant Analysis (LDA)	
6.3.	Kernel PCA	

Course Units and Course Outcomes Mapping:

Unit No.	Unit	Course Outcomes			
		CO1	CO2	CO3	CO4
1	Introduction to Machine Learning	✓			
2	Regression		✓		
3	Classification			✓	
4	Unsupervised Learning			✓	
5	Reinforcement and Deep Learning			✓	
6	Dimensionality Reduction				✓

Computing Environment:

A student must have the following computing environment in the laboratory and/or on his/her laptop.

- CE#1: Python 3

Text Books

1. Y. S. Abu-Mostafa, M. Magdon-Ismail, H. T. Lin, "Learning from Data", AMLBook Publishers, 2012.

References :

1. Saikat Dutt, Subramanian Chandramouli, and Amit Das, "Machine Learning", Pearson.
2. "Machine Learning A-ZTM: Hands-On Python & R in Data Science", Udemy.

3. E. Alpaydin, "Introduction to Machine Learning", MIT Press, Second Edition.
4. T. M. Mitchell, "Machine Learning", McGraw Hill, 1997.
5. P. Flach, "Machine Learning: The Art and Science of Algorithms that Make Sense of Data", Cambridge University Press, 2012.