# BIRLA INSTITUTE OF TECHNOLOGY & SCIENCE, PILANI WORK INTEGRATED LEARNING PROGRAMMES Digital Learning

Part A: Content Design

Course Title	Applied Machine Learning
Course No(s)	SE ZG568/SS ZG568
Credit Units	4
Course Authors	Bharathi R
Version No	1.0
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# **Course Objectives**

Machine learning is a rapidly growing field at the intersection of computer science and statistics. It is responsible for tremendous advances in technology, from personalized product recommendations to speech recognition in cell phones. The goal of this course is to provide a broad introduction to the key ideas in machine learning. The emphasis will be on intuition and realistic examples rather than theoretical results, though experience with statistics, calculus and linear algebra will be assumed. Through a variety of lecture examples and programming projects, students will learn how to apply powerful machine learning techniques to realistic problems, run evaluations and interpret results, and understand limitations.

No	Course Objective	
CO1	Provide a broad background to the key concepts and techniques in machine learning	
CO2	Hands on learning to apply contemporary machine learning techniques on realistic problems	
CO3	Evaluation and interpretation of results	
CO4	Understand limitations of different machine learning techniques	
CO5	Gain experience in working with machine learning software pipeline	

#### Text Book(s)

T1	Aurelien Geron, "Hands-On Machine Learning with Scikit-Learn, Keras and Tensorflow", O'Reilly, 2020
T2	P-N Tan, M. Steinbach, Vipin Kumar, "Introduction to Data Mining", 2016, Pearson

## Reference Book(s) & other resources

Christoph Molner, "Interpretable Machine Learning", 2020, <a href="https://christophm.github.io/interpretable-ml-book/">https://christophm.github.io/interpretable-ml-book/</a>	
Pedro Domingos, "A Few Useful Things to Know About Machine Learning", pp. 78-87, Communications of the ACM, vol. 55 no. 10, October 2012.	

# **Modular Content Structure**

No	Title of the Module	References
M1	Introduction to Machine Learning	T1: Chapter 1
	1.1. What and Why	R2
	1.2. Applications of Machine Learning	
	1.3. Types of Machine Learning	
	1.4. Challenges in Machine Learning	
M2	Big Picture: End-to-end Machine Learning	T1: Chapter 2
	2.1. Framing the ML Problem	T2: Chapter 2-3
	2.2. Data Types, Pre-processing, Visualization and	
	Analysis	
M3	Big Picture: End-to-end Machine Learning	T1: Chapter 2-3
IVIS	3.1 Model Selection and Training	T2: Chapter 4
	3.1.1. Prediction Problem	12. Chapter 4
	3.1.2. Classification Problem	
	3.2 Evaluation	
	3.2.1. Prediction Problem	
	3.2.2. Classification Problem	
	3.3 Machine Learning Pipeline	
	8 1	
M4	Linear Prediction Models	T1: Chapter 4
	4.1. Linear Regression	T2: Appendix D
	4.2. Gradient Descent and Variants	
	4.3. Regularization	
	4.4. Bias Vs. Variance	
M5	Classification Models I	T1: Chapter 4, 5
	5.1. Logistic Regression	T2: Chapter 5
	5.2. Support Vector Machines	
	5.3. Naïve Bayes	
146	5.4. Comparative Analysis  Classification Models II	T1 C1 + 6.7
M6		T1: Chapter 6-7
	6.1. Decision Tree	T2: Chapter 4, 5
	<ul><li>6.2. Challenges with Decision tree</li><li>6.3. Ensembles</li></ul>	
	6.2.1. Bagging	
	6.2.2. Boosting	
	6.4. Random Forest	
M7	Unsupervised Learning	T1: Chapter 8-9
	7.1. Dimensionality reduction and feature	T2: Chapter 8-9
	extraction	
	7.2 K-means Clustering	
	7.4 EM Algorithm	
	7.5 Applications	
M8	Neural Networks	T1: Chapter 10-11
	8.1. Perceptrons and Delta Rule	T2: Chapter 5
	8.2. Multi-Layer Perceptrons and Backpropagation	
	8.3. Deep Learning	
M9	Deep Networks	T1: Chapter 14, 15
	9.1. Convolutional Neural Network	

	9.2. Recurrent Neural Network 9.3. Applications in Image and Text Processing	
M10	FAccT Machine Learning	R1
	10.1. Bias and Fairness	
	10.2. Interpretability and Transparency	
	10.3. Robustness and Adversarial Attacks	

**Learning Outcomes:** 

No	Learning Outcomes
LO1	High-level conceptual understanding of machine learning field and its applicability and limitations
LO2	Ability to frame a machine learning problem and select candidate machine learning models
LO3	Ability to design and implement an end-to-end machine learning based solution using Python and appropriate libraries and do trouble shooting
LO4	Ability to evaluate and interpret the results from a machine learning system

Part B: Learning Plan

Academic Term	Second Semester 2022-2023
Course Title	Applied Machine Learning
Course No	SE ZG568/SS ZG568
Lead Instructor	Bharathi R

Session No.	Topic Title	Study / HW Resource Reference
1	Introduction to Machine Learning: What and Why, Applications of Machine Learning, Types of Machine Learning, Challenges in Machine Learning	T1: Chapter 1 R2
2	End-to-end Machine Learning: Framing the ML Problem. Data Types, Pre-processing, Visualization and Analysis	T1: Chapter 2 T2: Chapter 2-3
3	End-to-end Machine Learning: Model Selection and Training for Prediction and Classification, Evaluation, Machine Learning Pipeline.	T1: Chapter 2-3 T2: Chapter 2-3
4	Linear Prediction Models: Linear Regression, Gradient Descent and Variants, Regularization, Bias Vs. Variance	T1: Chapter 4

5	Classification Models I: Naïve Bayes classification, Applications in text and image classification	T2: Chapter 5
6	Classification Models I: Logistic Regression, Log Loss error function, Optimization using gradient descent, Feature transformation for nonlinear classification	T1: Chapter 4
7	Classification Models I: Support Vector Machine. Margin maximization. Non-linear SVM. Kernel Function.	T1: Chapter 5 T2: Chapter 5
8	Review of Session 1 to 7	
9	Classification Models II: Decision Tree. Entropy and information gain. Construction algorithm. Challenges with Decision Tree	T1: Chapter 6 T2: Chapter 4
10	Classification Models II: Ensembles techniques. Bagging, boosting, Random Forest	T1: Chapter 7 T2: Chapter 5
11	Unsupervised Machine Learning: Dimensionality reduction and feature extraction, K-means clustering, EM algorithm	T1: Chapter 8-9 T2: Chapter 8
12	Artificial Neural Networks: Perceptions, Delta rule, Design for Boolean logic gates and linear classification, Neural networks, Design for nonlinear classification and Boolean functions, Backpropagation algorithm	T1: Chapter 10 T2: Chapter 5
13.	Artificial Neural Networks: Deep Learning, Characteristics, Error Surface, hyperparameters, Regularization, Weight Updates.	T1: Chapter 11
14.	Deep Networks: Convolutional Neural Networks, The Convolution Operation, Pooling, padding, architectures for classification. Recurrent Neural Networks, Architecture, training, bidirectional network, long short-term memory, deep recurrent network. Sequence processing.	T1: Chapter 14- 15
14	FAccT Machine Learning. Bias and Fairness, Interpretability and Transparency, Robustness.	R1
16	Review of session 9 to 15	Books, Slides, Web references

# Detailed Plan for Lab work/Design work

Lab No	Lab Objective	Lab Sheet/Capsule Access URL	Content Reference
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1	End-to-end ML pipeline for prediction	Module 2
2	End-to-end ML pipeline for classification	Module 3
3	Linear Regression	Module 4
4	Linear Classifiers (Naïve Bayes, Logistic Regression)	Module 5
5	Decision Tree and Random Forest	Module 6
6	Multilayer Perceptrons	Module 8
7	CNN	Module 9
8	RNN	Module 9

#### **Evaluation Scheme:**

Legend: EC = Evaluation Component; AN = After Noon Session; FN = Fore Noon Session

No	Name	Type	Duration	Weight	Day, Date, Session, Time
EC-1	Quiz-I	Online	-	5%	February 13-23, 2023
	Programming Assignment I	Offline	-	12%	March 20-30, 2023
	Programming Assignment II	Offline	-	13%	April 20-30, 2023
EC-2	Mid-Semester Test	Open Book	2 hours	30%	Saturday, 11/03/2023 (AN)
EC-3	Comprehensive Exam	Open Book	2½ hours	40%	Saturday, 20/05/2023 (AN)

## Important Information:

Syllabus for Mid-Semester Test (Open Book): Topics in CS 1-8

Syllabus for Comprehensive Exam (Open Book): All topics given in plan of study

## **Evaluation Guidelines:**

- 1. For Closed Book tests: No books or reference material of any kind will be permitted. Laptops/Mobiles of any kind are not allowed. Exchange of any material is not allowed.
- 2. For Open Book exams: Use of prescribed and reference text books, in original (not photocopies) is permitted. Class notes/slides as reference material in filed or bound form is permitted. However, loose sheets of paper will not be allowed. Use of calculators is permitted in all exams. Laptops/Mobiles of any kind are not allowed. Exchange of any material is not allowed.
- 3. If a student is unable to appear for the Regular Test/Exam due to genuine exigencies, the student should follow the procedure to apply for the Make-Up Test/Exam. The genuineness of the reason for absence in the Regular Exam shall be assessed prior to giving permission to appear for the Make-up Exam. Make-Up Test/Exam will be conducted only at selected exam centres on the dates to be announced later.

It shall be the responsibility of the individual student to be regular in maintaining the self-study schedule

as given in the course handout, attend the lectures, and take all the prescribed evaluation components such as Assignment/Quiz, Mid-Semester Test and Comprehensive Exam according to the evaluation scheme provided in the handout.