

BIRLA INSTITUTE OF TECHNOLOGY AND SCIENCE, PILANI
INSTRUCTION DIVISION
FIRST SEMESTER 2016-2017
Course Handout Part II

Date: 01/08/2016

In addition to part-I (General Handout for all courses appended to the time table) this portion gives further specific details regarding the course.

Course No. : BITS F464
Course Title : Machine Learning
Instructor-in-charge : NAVNEET GOYAL (goel@)

Catalog Description

Machine Learning is an exciting sub-area of Artificial Intelligence which deals with designing machines which can learn and improve their performance from examples/experience. This course introduces the student to the key algorithms and theory that forms the core of machine learning. The course will cover the major approaches to learning namely, supervised, unsupervised, and reinforcement learning. The course emphasizes various techniques, which have become feasible with increased computational power and our ability to produce and capture huge volumes of data. The topics covered in the course include regression, decision trees, support vector machines, artificial neural networks, Bayesian techniques, Hidden Markov models, genetic algorithms etc. Some advanced topics like active & deep learning will also be covered.

Text Books:

Tom M. Mitchell, Machine Learning, The McGraw-Hill Companies, Inc. International Edition 1997.

Reference Books:

1. Christopher M. Bishop, Pattern Recognition & Machine Learning, Springer, 2006.
2. Introduction to Machine Learning, N. J. Nilsson, Stanford, Available online at author's website. <http://robotics.stanford.edu/people/nilsson/mlbook.html>
3. Machine Learning, Neural and Statistical Classification, D. Michie, D.J. Spiegelhalter, C.C. Taylor (eds), Ellis Horwood publishers, available online at <http://www.amsta.leeds.ac.uk/~charles/statlog/>

LECTURE PLAN

Topic	Topic Details	Lecture #	Chapter Reference
Overview	Introduction to Machine Learning	1	Ch. 1
High-dimensional data & Curse of Dimensionality	High-dimensional data and associated problems	2	R1 – Ch.1
Preliminaries	<ul style="list-style-type: none">• Probability theory• Decision theory• Information theory• Linear Algebra	Self Study	R1 – Ch.2, Appendix C
Some important principles/concepts/algorithms	<ul style="list-style-type: none">• MAP Hypothesis• Minimum Description Length (MDL)	3-5	Ch.6 + class notes + R1 Appendix E

	<ul style="list-style-type: none"> principle • Expectation Maximization (EM) Algorithm • Bias-variance decomposition • Lagrange Multipliers • Mixture of Gaussians • PCA & SVD 		
Linear models for Regression	<ul style="list-style-type: none"> • Linear basis function models • Bayesian linear regression 	6-8	R1 – Ch. 3
Linear models for Classification	<ul style="list-style-type: none"> • Discriminant Functions • Probabilistic Generative Classifiers • Probabilistic Discriminative Classifiers 	9-12	R1 – Ch. 4
Bayesian Learning Techniques	<ul style="list-style-type: none"> • Bayes optimal classifier • Gibbs Algorithm • Naïve Bayes Classifier 	13-14	Ch. 6
Non-linear Models & Model Selection	<ul style="list-style-type: none"> • Decision Trees • Ensemble Classifiers • Neural Networks <ul style="list-style-type: none"> ○ Multilayer Perceptron ○ Network training ○ Error back-propagation • Instance-based Learning <ul style="list-style-type: none"> ○ <i>K</i>-NN ○ Case-based Reasoning 	15-21	<ul style="list-style-type: none"> • Ch. 3 • Ch. 4 R1 – Ch. 5 • Ch. 8
Margin/Kernel Based Approaches	Support Vector Machines	22-24	Class Notes + R1 – Ch. 7
Graphical Models	<ul style="list-style-type: none"> • Bayesian Belief Networks • Hidden Markov Models 	25-28	Ch. 6 + Class Notes
Unsupervised Learning	<ul style="list-style-type: none"> • Mixture Models • <i>K</i>-means Clustering • Self-organized Maps (SOM) 	29-30	Ch. 6 R1 – Ch. 9
Genetic Algorithms	<ul style="list-style-type: none"> • Hypothesis space search • Genetic programming • Models of evaluation & learning 	31-32	Ch. 9
Reinforcement Learning	<ul style="list-style-type: none"> • Q Learning • Non-deterministic rewards & actions • Temporal difference learning • Generalization 	33-34	Ch. 13
Advanced Topics	<ul style="list-style-type: none"> • Active Learning • Deep Learning • Metric Learning 	35-38	Class Notes
Application Examples	<ul style="list-style-type: none"> • Speech Recognition • Image Retrieval 	39-40	Class Notes
Big Data Challenges	<ul style="list-style-type: none"> • Machine Learning for Big Data 	41-42	Class Notes

Evaluation Scheme:

Component	Duration	Weightage	Date (Time)
Midsem Test (Closed Book)	90 Mins.	30%	8/10 2:00 - 3:30 PM
Assignments (02)	Take Home	30%	TBA
Comprehensive Exam (partly open)	3 Hours	40%	14/12 FN

Notices: All notices shall be displayed on NALANDA

Chamber Consultation Hour: M, W 6.00 to 6.30 pm (6121-K, NAB)

Makeup Policy: To be granted only in case of serious illness or emergency.

Instructor-in-charge
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