**BIRLA INSTITUTE OF TECHNOLOGY & SCIENCE, PILANI**

**WORK INTEGRATED LEARNING PROGRAMMES**

**Digital**

Part A: Content Design

| **Course Title** | Machine Learning |
| --- | --- |
| **Course No(s)** | ZG565 |
| **Credit Units** | 4 |
| **Credit Model** | 1 - 0.5 - 1.5  1 unit for class room hours, 0.5 unit for Tutorial, 1.5 units for  Student preparation.  1 unit = 32 hours |
| **Content Authors** | Dr. Sugata Ghosal |
| **Version** | 1.0 |
| **Date** | November 11th, 2022 |

**Course Objectives**

| **No** |  |
| --- | --- |
| **CO1** | Introduce students to the basic concepts and techniques of Machine Learning. |
| **CO2** | To gain experience of doing independent study and research in the field of Machine Learning |
| **CO3** | To develop skills of using recent machine learning software tools to evaluate learning algorithms and model selection for solving practical problems |

**Text Book(s)**

| T1 | Tom M. Mitchell, Machine Learning, The McGraw-Hill Companies, Inc. Indian Edition 1997 |
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**Reference Book(s) & other resources**

| R1 | Christopher M. Bishop, Pattern Recognition & Machine Learning, Springer, 2006 |
| --- | --- |
| R2 | PANG-NING TAN, MICHAEL STEINBACH, VIPIN KUMAR, Introduction To Data Mining, Pearson, 2nd Edition. |
| R3 | CHRISTOPHER J.C. BURGES: A Tutorial on Support Vector Machines for Pattern Recognition, Kluwer Academic Publishers, Boston, pp. 1–43. |

**Content Structure**

1. Introduction
   1. Introduction to ML
   2. Objective of the course
   3. Taxonomy (types) of Machine Learning
   4. Design a Learning System
   5. Challenges in Machine Learning

1. Machine learning Workflow
   1. Role of Data
   2. Data Preprocessing, wrangling
   3. Data skewness removal (sampling)
   4. Model Training
   5. Model Testing and performance metrics
2. Linear models for Regression
   1. Direct Solution Method
   2. Iterative Method – Gradient Descent (batch/stochastic/mini-batch)
   3. Linear basis function models
   4. Bias-variance decomposition
3. Linear models for classification
   1. Discriminant Functions
   2. Decision Theory
   3. Probabilistic Discriminative Classifiers
   4. Logistic Regression

1. Decision Tree
   1. Information Theory
   2. Entropy Based Decision Tree Construction
   3. Avoiding Overfitting
   4. Minimum Description Length
   5. Handling Continuous valued attributes, missing attributes
2. Instance-based Learning
   1. k-Nearest Neighbor Learning
   2. Locally Weighted Regression (LWR) Learning
   3. Radial Basis Functions
3. Support Vector Machine
   1. Linearly separable data
   2. Non-linearly separable data
   3. Kernel Trick (Mercer)
   4. Applications to both structured and unstructured data

1. Bayesian Learning
   1. MLE Hypothesis
   2. MAP Hypothesis
   3. Bayes Rule
   4. Optimal Bayes Classifier
   5. Naïve Bayes Classifier
   6. Probabilistic Generative Classifiers
   7. Bayesian Linear Regression
2. Ensemble Learning
   1. Combining Classifiers
   2. Bagging
   3. Random Forest
   4. Boosting
      1. ADABoost
      2. Gradient Boosting
      3. XGBoost

1. Unsupervised Learning
   1. K-means Clustering and variants
   2. Review of EM algorithm
   3. GMM based Soft Clustering
   4. Applications
2. Machine Learning Model Evaluation/Comparison
   1. Comparing Machine Learning Models
   2. Emerging requirements e.g., bias, fairness, interpretability of ML models

**Learning Outcomes:**

| No | Learning Outcomes |
| --- | --- |
| LO1 | A strong understanding of the foundations of Machine Learning algorithms |
| LO2 | Able to solve Machine Learning problems using appropriate learning techniques |
| LO3 | Evaluate machine learning solutions to problems |
| LO4 | Identify appropriate tools to implement the solutions to machine learning problems |

**Part B: Learning Plan**

| **Academic Term** |  |
| --- | --- |
| **Course Title** | Machine Learning |
| **Course No** | ZG 565 |
| **Lead Instructor** | Dr. Sugata Ghosal |

**Text Book(s)**

| T1 | Tom M. Mitchell, Machine Learning, The McGraw-Hill Companies, Inc. Indian Edition 1997 |
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**Reference Book(s) & other resources**

| R1 | Christopher M. Bishop, Pattern Recognition & Machine Learning, Springer, 2006 |
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| R2 | PANG-NING TAN, MICHAEL STEINBACH, VIPIN KUMAR, Introduction To Data Mining, Pearson, 2nd Edition. |
| R3 | CHRISTOPHER J.C. BURGES: A Tutorial on Support Vector Machines for Pattern Recognition, Kluwer Academic Publishers, Boston, pp. 1–43. |

| Session No. | Topic Title | Study/HW Resource Reference |
| --- | --- | --- |
| 1 | **Introduction**  Objective, What is Machine Learning? Application areas of Machine Learning, Why Machine Learning is important? Design a Learning System, Issues in Machine Learning | T1 – Ch1 |
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| 2 | **Machine learning Workflow**  Role of Data, Data Pre-processing, wrangling, Data skewness removal (sampling), Model Training, Model Testing and performance metrics | R2 – Ch2, Ch3  Lecture Notes |
| 3 | **Linear models for Regression**  Direct Solution Method, Iterative Method – Gradient Descent (batch/stochastic/mini-batch), Linear basis function models | R1 - Ch3 |
| 4 | **Linear models for Regression (contd)**  Bias-variance decomposition  **Linear models for classification**  Discriminant Functions, Decision Theory, Probabilistic Discriminative Classifiers, Introduction to Logistic Regression | R1 - Ch. 3, 4 |
| 5 | **Logistic Regression**  Logloss Function, Gradient Descent, multi-class classification | R1 - Ch. 4  R2 – Ch. 4 |
| 6 | **Decision Tree**  Information Theory, Entropy Based Decision Tree Construction, Avoiding Overfitting, Minimum Description Length, Handling Continuous valued attributes, missing attributes | T1 – Ch. 3  R2 - Ch. 3 |
| 7 | **Instance-based Learning**  k-Nearest Neighbor Learning, Locally Weighted Regression (LWR) Learning , Radial Basis Functions | T1 – Ch. 8 |
| 8 | Review of Session 1 to 7 | Books, Web references and Slides |
| 9 | **Support Vector Machine**  Linearly separable data, Non-linearly separable data, Kernel Trick (Mercer), Applications to both structured and unstructured data | R2 - Ch. 4  R3 |
| 10 | **Bayesian Learning**  MLE Hypothesis , MAP Hypothesis, Bayes Rule, Optimal Bayes Classifier | T1 - Ch. 6  R2 – Ch. 4 |
| 11 | **Bayesian Learning**  Naïve Bayes Classifier, Probabilistic Generative Classifiers, Bayesian interpretation of Linear Regression | T1 - Ch. 6  R2 – Ch. 4  R1 – Ch. 4 |
| 12 | **Ensemble Learning**  Combining Classifiers, Bagging, Random Forest, Boosting | R2 – Ch. 4 |
| 13 | **Ensemble Learning**  ADABoost, Gradient Boosting, XGBoost | R2 – Ch. 4  Lecture Notes |
| 14 | **Unsupervised Learning**  K-Means clustering, Mixture Models for probabilistic clustering, Review of EM algorithm, Applications | T1 – Ch. 6 |
| 15 | **Machine Learning Model Evaluation/Comparison**  Comparing Machine Learning Models, Emerging requirements e.g., bias, fairness, interpretability of ML models | T1 - Ch. 5  Lecture Notes |
| 16 | Review of session 9 to 15 | Books, Web references and Slides |

**Detailed Plan for Lab work**

| **Lab No.** | **Lab Objective** | **Lab Sheet Access URL** | **Session Reference** |
| --- | --- | --- | --- |
| 1 | End to End Machine Learning |  | 2 |
| 2 | Linear Regression and Gradient Descent |  | 3, 4 |
| 3 | Logistic Regression classifier |  | 5 |
| 4 | Decision Tree and Random Forest |  | 6, 12 |
| 5 | Naïve Bayes Classification |  | 11 |

**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 - Two | Online | ~1 hour | 10% |  |
|  | Assignment-I | Take Home | ~2-3 weeks | 10% |  |
|  | Assignment-II | Take Home | ~2-3 weeks | 10% |  |
| EC-2 | Mid-Semester Test | Closed Book |  | 30% |  |
| EC-3 | Comprehensive Exam | Open Book |  | 40% |  |

**Note:**

Syllabus for Mid-Semester Test (Open Book): Topics in Session Nos. 1 to 8

Syllabus for Comprehensive Exam (Open Book): All topics (Session Nos. 1 to 16)

**Important links and information:**

Elearn portal: <https://elearn.bits-pilani.ac.in> or Canvas

Students are expected to visit the Elearn portal on a regular basis and stay up to date with the latest announcements and deadlines.

Contact sessions: Students should attend the online lectures as per the schedule provided on the Elearn portal.

Evaluation Guidelines:

1. EC-1 consists of either two Assignments or three Quizzes. Students will attempt them through the course pages on the Elearn portal. Announcements will be made on the portal, in a timely manner.
2. For Closed Book tests: No books or reference material of any kind will be permitted.
3. For Open Book exams: Use of books and any printed / written reference material (filed or bound) 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.
4. 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 which will be made available on the Elearn portal. The 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 online 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.