

UE21CS352A: Machine Intelligence

No. of Credits: 5

of Slots: 112

| Class # | Chapter Title / Reference Literature | Topics to be Covered | % Of Portion Covered | |
|--|--------------------------------------|--|----------------------|------------|
| | | | Absolute | Cumulative |
| Unit 1 : Introduction, Search Algorithms ,Classification with Decision Trees and Performance Metrics | | | 23.21 | 23.21 |
| 1 | T1 : Chapter 1 - 1.1 | Introduction to AI and ML | | |
| 2 | T1 : Chapter 2 – 2.1, 2.3, 2.4,2.5 | Intelligent Agents and its Types | | |
| 3 | Slides on PESU Academy Portal | Machine Learning and its Models | | |
| 4 | T2: Chapter2 – 2.1- 2.2 | Concept Learning: Introduction, A Concept Learning Task, Concepts of Hypotheses | | |
| 5 | T2: Chapter2 – 2.3- 2.4,2.7 | Concept Learning: Concept Learning as Search, Find-S:Finding a Maximally Specific Hypothesis, Version Space, Inductive Bias | | |
| 6-7 | Lab Component | Understanding basics of PyTorch, Tools related to ML (Self Study) | | |
| 8 | Slides on PESU Academy Portal | Performance Metrics-Accuracy, Precision, Recall | | |
| 9 | Slides on PESU Academy Portal | Performance Metrics-Sensitivity, Specificity, AUC, RoC | | |
| 10 | Slides on PESU Academy Portal | Bias Variance Decomposition | | |
| 11 | T2: Ch 3 3.1, 3.2, 3.3 | Decision Tree Learning: Introduction, Decision Tree Representation, Appropriate Problems for Decision Tree Learning | | |
| 12 | T2: Ch 3 3.4 | The Basic Decision Tree Learning Algorithm- ID3 Algorithm for Classification, Entropy and Gain calculations | | |
| 13 | T2: Ch 3 3.5, 3.6 | Hypothesis Space Search in Decision Tree Learning, Inductive Bias in Decision Tree Learning | | |
| 14 | T2: Ch 3 3.7(3.7.1,3.7.2) | Issues in Decision Tree Learning – Avoiding Overfitting the Data, Solutions to Overfitting, Incorporating Continuous-Valued Attributes | | |
| 15-16 | Lab Component | Decision Tree Classifier- Build a Decision Tree Classifier for the given data. | | |

| | | | | |
|---|---|---|-------|-------|
| 17 | T2: Ch 3, Slides on PESU Academy Portal | Solving a numerical problem on Decision Trees, Decision Boundary for Decision Trees(X-Y axis) | | |
| 18 | Slides on PESU Academy Portal | Introduction to Linear Regression, Linear Regression using Gradient Descent | | |
| 19 | Slides on PESU Academy Portal | Solving a numerical problem on Linear Regression | | |
| 20 | Slides on PESU Academy Portal | Issues with Linear Regression, Introduction to Logistic Regression | | |
| 21 | Slides on PESU Academy Portal | Solving a numerical problem on Logistic Regression | | |
| 22 | T2: Ch 8 -8.1, 8.2 (8.2.1,8.2.2) pg. 230-236 | Instance-Based Learning: k-Nearest Neighbor Algorithm (Classification) | | |
| 23 | Slides on PESU Academy Portal | Instance-Based Learning: k-Nearest Neighbor Algorithm(Regression) | | |
| 24 | T2: Ch 8 Pg. 230-236, Slides on PESU Academy Portal | Sample problems on Weighted kNN, Decision Boundary for kNN | | |
| 25-26 | | Assignment 1 (Written/ Practical) on Linear Regression or Logistic Regression | | |
| Unit 2: Supervised Learning with ANN, Introduction to Deep Learning Techniques | | | 19.64 | 42.85 |
| 27 | T2: Ch 4 4.1-4.3 pg. 81-85 | Artificial Neural networks: Introduction | | |
| 28 | T2: Ch 4 4.4 pg. 86-94 | Perceptrons – implementing LOGIC gates | | |
| 29 | T2 Ch 4 4.5(4.5.1-4.5.3 excluding case 1 and case 2) pg. 95-101 | Multi-layer Networks and Back-Propagation | | |
| 30 | T2 Ch 4 4.5(4.5.1-4.5.3 excluding case 1 and case 2) pg. 95-101 | Back-Propagation Derivation | | |
| 31 | Slides on PESU Academy Portal | Activation functions (Step, Sigmoid, Tanh, ReLU, Leaky ReLU, Softmax) | | |

| | | | | |
|--|---|--|-------|-------|
| 32 | Slides on PESU Academy Portal | Various Optimizers(GD, SGD, Momentum-based, Adagrad, RMSprop, Adam) | | |
| 33-34 | Lab Component | Implementation of Simple Artificial Neural Networks | | |
| 35 | Mitesh Khapra slides | Introduction to Deep Learning, Introduction to Convolutional Neural Network(CNN) | | |
| 36 | on CNN & T3: Ch 6 | CNN-Introduction to Convolution Operation and Pooling- Max pooling, Average pooling | | |
| 37 | (Demystifying Convolutional Networks) | CNN- Parameter Calculation, Problems on Parameter Calculation | | |
| 38-39 | Lab Component | Building a Classification Model for the given dataset using CNN | | |
| 40 | | Introduction to Recurrent Neural Network | | |
| 41 | | RNN- Vanishing and Exploding Gradient | | |
| 42 | | Variants of RNN : LSTM | | |
| 43 | | Variants of RNN : GRU | | |
| 44 | | Introduction to Large Language Models(LLM) | | |
| 45-46 | | Assignment 2 (Written/Practical) on RNN or LSTM | | |
| 47-48 | | Revision for ISA1(Units 1 and 2) | | |
| Unit 3: SVM, Boosting and Stochastic Models | | | 22.32 | 65.18 |
| 49 | T4 Ch 3 pg.38-53 | Support Vector Machines – Margin and Maximization and the Primal Form | | |
| 50 | T4 Ch 4 pg.54-59 | SVM - The Lagrangian Dual and its Solution(Hard Margin Classification ONLY) | | |
| 51 | T4 Ch 4 pg.54-59 | SVM - Soft Margin(Classification ONLY) | | |
| 52 | T4 Ch 6 pg.72-81 | SVM – Kernel Trick Kernel functions: Linear, polynomial (Derivation only for linear function). | | |
| 53 | | Sample problems on SVM | | |
| 54-55 | Lab Component | Support Vector Machines -Implement a Support Vector Machine Classifier for the given dataset | | |
| 56 | Slides on PESU Academy Portal | Combining Weak Learners | | |
| 57 | R3: pg. 129-131 (Slides on PESU Academy Portal) | Improving Performance: Bagging and Boosting | | |
| 58 | Slides on PESU Academy Portal | Improving Performance with Gradient Boost | | |
| 59 | Slides on PESU Academy Portal | Random Forest | | |
| 60 | | Basics of Probability | | |

| | | | | |
|---|--|---|-------|-----|
| 61 | T2: Ch 6 pg. 154-166, pg. 170-171,174-176 | Bayesian Learning – Bayes Theorem, Maximum likelihood, Bayes Optimal Classifier | | |
| 62-63 | T2: Ch6 – pg. 177-183 | Naïve Bayes Classifier and Text Classification. | | |
| 64-65 | Lab Component | Build a Text Classifier using Naïve Bayes Algorithm | | |
| 66-67 | R4: Chapter 9, pg. 286-289 (Slides on PESU Academy Portal) | Expectation maximization Algorithm | | |
| 68-69 | R4: Chapter 9, pg. 289-292(Slides on PESU Academy Portal) | Gaussian Mixture Models | | |
| 70-71 | Lab Component | GMM Program | | |
| 72-73 | | Assignment 3 (Written/Practical) on Bagging or Boosting or Random Forest | | |
| Unit 4: HMM, Unsupervised Learning ,Dimensionality Reduction and Genetic Algorithms, PSO | | | | |
| 74 | R2: Ch 15 pg. 363- 366 (Slides on PESU Academy Portal) | Hidden Markov Models – Discrete Markov Processes | 34.82 | 100 |
| 75 | R2: Ch15 – pg. 367- 373(Slides on PESU Academy Portal) | Hidden Markov Models – 3 Basic Problems | | |
| 76 | R2: Ch15 – pg. 373-375(Slides on PESU Academy Portal) | Learning the State Sequence | | |
| 77 | R2: Ch15 – pg. 373-375(Slides on PESU Academy Portal) | Learning the Parameters | | |

| | | | | |
|-------|--|--|--|--|
| 78 | R2: Ch15 – pg. 375-378(Slides on PESU Academy Portal) | Baum-Welch Algorithm | | |
| 79 | Slides on PESU Academy Portal | Problems on Hidden Markov Models | | |
| 80-81 | Lab Component | Hidden Markov Model (HMM)- Implement the Viterbi algorithm for decoding a sequence of observations to find the most probable sequence of internal states that generated the observations. | | |
| 82-92 | | Hackathon | | |
| 93 | R3: Ch10: pg. 207- 217, Ch 11: pg.224-234, Ch 12: pg. 248-260 (Slides on PESU AcademyPortal) | Unsupervised Learning: Hierarchical vs. Non- Hierarchical Clustering | | |
| 94 | R3: Ch10: pg. 207- 217, Ch 11: pg.224-234, Ch 12: pg. 248-260 (Slides on PESU AcademyPortal) | Unsupervised Learning: Agglomerative and Divisive Clustering | | |
| 95 | R1: Chapter 9, pg.424-430 | k-Means Clustering, Simple problems on k-Means Clustering, Bisecting k-Means, Issues with k-Means Clustering. | | |
| 96 | R1: Chapter 9, pg. 439 | k-Means as special case of Expectation Maximization | | |
| 97-98 | Lab Component | Implementation of K-Means Clustering | | |

| | | | | |
|---------|---|--|--|--|
| 99 | R1: Chapter 12, pg. 559-570 (Slides on PESU Academy Portal) | Dimensionality Reduction Techniques –Introduction, Techniques Available, Principal Component Analysis(PCA)- Introduction to PCA and its Applications, Principal Components | | |
| 100 | Slides on PESU Academy Portal | Problems on PCA | | |
| 101 | Slides on PESU Academy Portal | Dimensionality Reduction Techniques – Introduction to SVD and its Applications, SVD | | |
| 102 | Slides on PESU Academy Portal | Problem on SVD, Relationship between PCA and SVD | | |
| 103 | T2: Ch 9 - 9.1, 9.2 | Genetic Algorithms – Representing Hypothesis, Genetic Operators | | |
| 104 | T2: Ch 9 - 9.2, 9.3 | Fitness Function and Selection Methods, Crossover, Mutation | | |
| 105 | T2: Ch 9 | Simple Applications of the Genetic Algorithm, Application of GA in Decision tree, Genetic Algorithm based Clustering | | |
| 106 | T2: Ch7.1-7.4, Ch9 | Solving Single Objective Optimization problems using GA, Using GA to emulate Gradient Descent/Ascent | | |
| 107 | Slides on PESU Academy Portal | Weight Determination, Clustering, Introduction to PSO | | |
| 108 | Slides on PESU Academy Portal | Application of PSO in Single Objective optimization problems. | | |
| 109-110 | | Assignment 4 (Written/Practical) on PCA or Genetics Algorithm or PSO | | |
| 111-112 | | Revision for ISA2(Units 3 and 4) | | |

Literature:

| Book Type | Code | Title and Author | Publication info | | |
|-----------|------|--|------------------|-------------------------------|---------------------------------|
| | | | Edition | Publisher | Author |
| Text Book | T1 | 1. Artificial Intelligence: A Modern Approach (3rd Edition), Stuart Russel and Peter Norvig, | 3rd | Wiley India | Stuart Russel and Peter Norvig, |
| Text Book | T2 | Machine Learning, Tom Mitchell, McGraw Hill Education (India), 2013. | | McGraw Hill Education (India) | Tom Mitchell |
| Text Book | T3 | Hands-on Deep Learning Algorithm with Python - Sudharshan Ravi Chandiran | | Packt Publishing Limited | Sudharshan Ravi Chandiran |

| | | | | | |
|----------------|----|--|-----|-----------------------------------|---------------------|
| Text Book | T4 | Support Vector Machines Succinctly by Alexandre Kowalczyk | | Syncfusion | Alexandre Kowalczyk |
| Reference Book | R1 | "Pattern Recognition and Machine Learning", Christopher Bishop, Springer (2nd Printing), 2011 | | Springer (2nd Printing), 2011 | Christopher Bishop |
| | R2 | Introduction to Machine Learning Second Edition by Ethem Alpaydin | 2nd | The MIT Press | Ethem Alpaydin |
| | R3 | Machine Learning in Action by Peter Harrington, First Edition, Manning 2021 | 1st | Manning (2021) | Peter Harrington |
| | R4 | "Machine Learning: The Art and Science of Algorithms that Make Sense of Data", Peter Flach, Cambridge University Press (2012). | | Cambridge University Press (2012) | Peter Flach |