Program / Semester: B.Tech (V Sem)	Branch: Artificial Intelligence and Machine Learning
Subject: Artificial Intelligence	Course Code: C109511(022)
Total / Minimum-Pass Marks (End Semester Exam): 100/35	L: 3 T: 1 P: 0 Credits: 4
Class Tests & Assignments to be conducted: 2 each	Duration (End Semester Exam): 03 Hours

## **Course Objective:**

- To impart knowledge about Artificial Intelligence.
- To give understanding of the main abstractions and reasoning for intelligent systems.
- To enable the students to understand the basic principles of Artificial Intelligence in various applications.

**UNIT-I: Introduction:** Overview of AI problems, AI problems as NP, Artificial Intelligence and its applications, Artificial Intelligence Techniques, Level of models, criteria of success, Intelligent Agents, Nature of Agents, Learning Agents. AI Techniques, advantages, and limitations of AI, Impact and Examples of AI, Application domains of AI. The AI Ladder - The Journey for Adopting AI Successfully, Advice for a career in AI, Hotbeds of AI Innovation.

**UNIT-II: Problem Solving Techniques:** State space search, control strategies, heuristic search, problem characteristics, production system characteristics., Generate and test, Hill climbing, best first search, A\* search, Constraint satisfaction problem, Mean-end analysis, Min-Max Search, Alpha-Beta Pruning, Additional refinements, Iterative Deepening.

**UNIT-III: Logic:** Propositional logic, predicate logic, Resolution, Resolution in proportional logic and predicate logic, Clause form, unification algorithm.

**UNIT-IV: Knowledge Representation Schemes and Reasoning:** Mapping between facts and representations, Approaches to knowledge representation, procedural vs declarative knowledge, Forward vs. Backward reasoning, Matching, conflict resolution, Non-monotonic reasoning, Default reasoning, statistical reasoning, fuzzy logic Weak and Strong filler structures, semantic nets, frame, conceptual dependency, scripts.

**UNIT-V: Planning:** The Planning problem, planning with state space search, partial order planning, planning graphs, planning with propositional logic, Analysis of planning approaches, Hierarchical planning, conditional planning, Continuous and Multi Agent planning.

#### **Text Books:**

- 1. A Classical Approach to Artificial Intelligence, M.C. Trivedi, Khanna Book Publishing, 2019.
- 2. Artificial Intelligence: A modern approach by Stuart Russel, Pearson Education, 2010.
- 3. Artificial Intelligence by Rich and Knight, The McGraw Hill, 2017.
- 4. Artificial Intelligence: A new synthesis by Nils and Nilson, Elsevier, 1997.
- 5. Artificial Intelligence by Luger, Pearson Education, 2002.
- 6. Artificial Intelligence by Padhy, Oxford Press, 2005.
- 7. https://www.edx.org/course/artificial-intelligence-ai
- 8. https://www.udemy.com/course/artificial-intelligence-az/

#### **Reference Books:**

- 1. Title Introduction to Artificial Intelligence and Expert Systems Author Dan W. Patterson Publisher Pearson Education Edition 1st Edition, 2015.
- 2. 2Title Artificial Intelligence: A Modern Approach Author S. Russell and P. Norvig. Publisher Prentice Hall Edition 3rd Edition 2009.

#### **Course Outcome:** At the end of this course students will:

- Apply AI algorithms for solving practical problems.
- Describe human intelligence and AI.
- Explain how intelligent system works.
- Apply basics of Fuzzy logic and neural networks.
- Learn About Application and Analysis of planning approaches.

Program / Semester: B.Tech (V Sem)	Branch: Artificial Intelligence and Machine Learning
Subject: Introduction to Machine Learning	Course Code: C109512(022)
Total / Minimum-Pass Marks(End Semester Exam): 100/35	L: 3 T: 1 P: 0 Credits: 4
Class Tests & Assignments to be conducted: 2 each	Duration (End Semester Exam): 03 Hours

## **Course Objectives:**

- Understand the concept of Machine Learning.
- Familiarize with Simple Linear Regression and Logistic Regression.
- Appreciate the various nuances of Multiple Regressions and Model Building.
- Identify and apply the Classification algorithms.
- Apply the Clustering algorithms for developing applications.

**UNIT-I: Introduction:** The Origins of Machine Learning, Uses and Abuses of Machine Learning, How do Machines Learn? -Abstraction and Knowledge Representation, Generalization, Assessing the Success of Learning, Steps to Apply Machine Learning to Data, Choosing a Machine Learning Algorithm - Thinking about the Input Data, Thinking about Types of Machine Learning Algorithms, Matching Data to an Appropriate Algorithm.

UNIT-II: Simple Linear Regression: Simple Linear Regression Model Building, Estimation of Parameters Using Ordinary Least Squares, Interpretation of Simple Linear Regression Coefficients, Validation of Simple Linear Regression Model, Coefficient of Determination (R-squared) and Adjusted R-Squared, Spurious Regression, Hypothesis Test for Regression Coefficients (t-Test), Test for Overall Model: Analysis of Variance (F-Test), Residual Analysis

**UNIT-III: Regression & Model Building:** Introduction, Ordinary Least Squares Estimation for Multiple Linear Regression, Multiple Linear Regression Model Building, Partial Correlation and Regression Model Building, Interpretation of Multiple Linear Regression Coefficients. Validation of Multiple Regression Model, Coefficient of Multiple Determination (R-Squared), Adjusted R-Squared, Statistical Significance of Individual Variables in Multiple Linear Regression: t-Test, Validation of Overall Regression Model: F-Test.

**UNIT-IV: Classification Algorithms:** What is Classification? General Approach to Classification, k-Nearest Neighbor Algorithm, Logistic. Regression, Decision Trees, Naive Bayesian Classifier, Ensemble Methods: Bagging, Boosting and AdaBoost and XBoost, Random Forests Support Vector Machines, Rough Set and Fuzzy Set Approaches, Classification Model Evaluation and Selection: Sensitivity, Specificity, Positive Predictive Value, Negative Predictive Value.

**UNIT-V: Clustering Methods & Analysis:** Overview of Some Basic Clustering Methods, Hierarchical Methods: Agglomerate versus Divisive Hierarchical Clustering, Distance Measures, Probabilistic Hierarchical Clustering, Multiphase Hierarchical Clustering Using Clustering Feature Trees, Partitioning Methods: k-Means Clustering, k-Medoids Clustering, Density-Based Clustering: DBSCAN - Density-Based Clustering.

#### **Text Books:**

- 1. Tom M. Mitchell- Machine Learning McGraw Hill Education, International Edition
- 2. C. M. Bishop. Pattern Recognition and Machine Learning. First Edition. Springer, 2006. (Second Indian Reprint, 2015). [Bishop].

### **Reference Books:**

- 3. V.K. Jain, Machine Learning, Khanna Publishing House
- 4. Vinod Chandra S.S., Artificial Intelligence & Machine Learning, PHI
- 5. Introduction to Machine Learning with Python Andreas C. Müller, Sarah Guido, **First** Edition O'Reilly Media, Inc.
- 6. Python for Everyone Saurabh Chandrakar, Dr. Nilesh Bhaskarrao Bahadure bpb Publication ISBN: 9789355518170, eISBN: 9789355518156
- 7. Tom M. Mitchell, "Machine Learning", Mc Graw Hill, Indian Edition, 2017

### **Course Outcomes:** [After undergoing the course, students will be able to:]

- Understand a wide variety of learning algorithms.
- Understand how to evaluate models generated from data.
- Apply the algorithms to a real problem.
- Optimize the models learned and report on the expected accuracy.
- Applying the optimized models.

Program / Semester: B.Tech (V Sem)	Branch: Artificial Intelligence and Machine Learning
Subject: Theory of Computation	Course Code: C109513(022)
Total / Minimum-Pass Marks (End Semester Exam): 100/35	L: 3 T: 1 P: Credits: 4
Class Tests & Assignments to be conducted: 2 each	Duration (End Semester Exam): 03 Hours

### **Course Objective:**

- Students will learn about a variety of issues in the mathematical development of computer science theory, particularly finite representations for languages and machines.
- Students will gain a more formal understanding of algorithms and procedures
- To illustrate finite state machines to solve problems in computing.
- To explain the hierarchy of problems arising in the computer sciences.
- To familiarize Regular grammars, context frees grammar.
- To solve various problems of applying normal form techniques, push down automata and Turing Machines.

**UNIT-I:** The Theory of Automata: Introduction to automata theory, Examples of automata machine, Finite automata as a language acceptor and translator. Deterministic finite automata. Non deterministic finite automata, finite automata with output (Mealy Machine. Moore machine). Finite automata with ? moves, Conversion of NFA to DFA by Arden's method, Minimizing number of states of a DFA. My hill Nerode theorem, Properties and limitation of FSM. Two way finite automata. Application of finite automata.

**UNIT-II: Regular Expressions:** Regular expression, Properties of Regular Expression. Finite automata and Regular expressions. Regular Expression to DFA conversion & vice versa. Pumping lemma for regular sets. Application of pumping lemma, Regular sets and Regular grammar. Closure properties of regular sets. Decision algorithm for regular sets and regular grammar.

**UNIT-III: Grammars:** Definition and types of grammar. Chomsky hierarchy of grammar. Relation between types of grammars. Role and application areas of grammars. Context free grammar. Left most linear &right most derivation trees. Ambiguity in grammar. Simplification of context free grammar. Chomsky normal from. Greibach normal form, properties of context free language. Pumping lemma fro context free language. Decision algorithm for context tree language.

UNIT-IV: Push Down Automata and Turing Machine: Basic definitions. Deterministic pushdown automata and non deterministic push down automata. Acceptance of push down automata. Push down automata and context free language. Turing machine model. Representation of Turing Machine Construction of Turing Machine for simple problem's. Universal Turing machine and other modifications. Church's Hypothesis. Post correspondence problem. Halting problem of Turing Machine.

**UNIT-V: Computability:** Introduction and Basic concepts. Recursive function. Partial recursive function. Initial functions, computability, A Turing model for computation. Turing computable functions, Construction of Turing machine for computation. Space and time complexity. Recursive enumerable language and sets.

### **Text Books**:

- 1. Theory of Computer Science (Automata Language & Computation), K.L.P. Mishra and N. Chandrasekran, PHI.
- 2. Introduction to Automata theory. Language and Computation, John E. Hopcropt& Jeffery D. Ullman, Narosa Publishing House.

### **Reference Books:**

- 1. Theory of Automata and Formal Language, R.B. Patel & P. Nath, Umesh Publication
- 2. An Introduction and finite automata theory, Adesh K. Pandey, TMH.
- 3. Theory of Computation, AM Natrajan. Tamilarasi, Bilasubramani, New Age International Publishers.
- 4. Finite Automata and Formal Languages: A Simple Approach, A.M. Padma Reddy, Pearson Education, India.

### **Course Outcome:**

At the end of this course students will:

- Design finite automata to accept a set of strings of a language.
- Determine whether the given language is regular or not. .
- Design context free grammars to generate strings of context free language.
- Design push down automata and the equivalent context free grammars and Design Turing machine.
- Distinguish between computability and noncomputability, Decidability and un-decidability

Program / Semester: B.Tech (V Sem)	Branch: Artificial Intelligence and Machine Learning
Subject: Probability and Statistics for Machine Learning	Course Code: C109514(022)
Total / Minimum-Pass Marks (End Semester Exam): 100/35	L: 2 T: 1 P: Credits: 3
Class Tests & Assignments to be conducted: 2 each	Duration (End Semester Exam): 03 Hours

## **Course Objectives:**

- To introduce the basic concepts of probability.
- To introduce the basic concepts random variables.
- To introduce the elementary concepts of descriptive and inferential techniques of statistical methodology.
- To build the linear relationship between two variables and also to predict how a dependent variable changes based on adjustments to an independent variable.
- To acquaint the knowledge of testing of hypothesis for samples which plays an important role in real life problems.

**UNIT-I: Basics of Probability:** Probability: Introduction, Random Experiment, Sample Space, Event, Types of Events. Three Approaches To Probability, Additive and Multiplicative Laws Of Probability, Conditional Probability, Total Probability Theorem and Bayes' Theorem and its applications.

**UNIT-II: Random Variables:** Random Variables: Definition of discrete and continuous random variables, Probability Mass Function (PMF), Probability Density Function (PDF) and Cumulative Distribution Function (CDF). Moments of Random a Variable - Mean and Variance. Moment, Generating Function of a Random Variable (Definition & Properties). Bernoulli, Binomial, Poisson and Normal Distributions - Problems with Applications.

**UNIT-III: Basics of Statistics**: Population, Sample, Attribute and Variable (Discrete and Continuous). Classification and Tabulation of Data. Graphical Representation of Data - Histogram, Frequency Polygon, Stem-and-Leaf Plots, Box Plot, Bar & Pie Charts. Measures of Central Tendency - Mean, Median, Mode.

Dispersion and its Measures – Range, Quartile Deviation, Mean Deviation, Standard Deviation. Skewness and Kurtosis.

**UNIT-IV: Correlation and Regression:** Correlation -Coefficient of correlation, Rank correlation, Regression coefficients, Lines of regression. Multiple correlation and regression- Coefficient of multiple Correlation, multiple regression, multiple linear regression equations.

**UNIT-V: Sampling and Hypothesis Testing: Sampling:** Definitions - Types of sampling - Expected values of sample mean and variance, Standard error - Sampling distribution of means and variance. Estimation - Point estimation and Interval estimation.

**Testing of hypothesis:** Null and Alternative hypothesis - Type I and Type II errors, Critical region - confidence interval - Level of significance, One tailed and Two tailed test.

#### **Text Books:**

- 1. Richard A. Johnson Irwin Miller and John E. Freund, ``Probability and Statistics for Engineers", PrenticeHall, New Delhi, 11th Edition, 2011.
- 2. P. Kousalya, "Probability, statistics and random processes", Pearson Education, 2013.

## **Reference Books:**

- 1. Gupta S. C. And V.K. Kapoor (2020), Fundamental of Mathematical Statistics, S. Chand and Co., 12<sup>th</sup> Edition
- 2. A. Goon, M. Gupta and B. Dasgupta, "Fundamentals of Statistics", vol. I & II, World Press.

- 3. Mood, A.M. Graybill, F.A. and Boes, D.C. (2007): Introduction to the Theory of Statistics, 3rd Edn., (Reprint), Tata McGraw-Hill Pub. Co.Ltd.
- 4. Mukhopadhyay, P. (1999): Applied Statistics, New Central Book Agency.
- 5. Milton. J. S. and Arnold. J.C., "Introduction to Probability and Statistics", Tata McGraw Hill, 4th Edition, 2007.

## **Course Outcomes** [After undergoing the course, students will be able to :]

- Provide very good insight which is essential for industrial applications by learning probability distributions.
- Evaluate randomness in certain realistic situation which can be either discrete or continuous type and compute statistical constants of these random variables.
- Compute and Interpret Measures of Central Tendency and Dispersion of Data; Construct and Analyze Graphical Displays to Summarize Data.
- Compute Correlation Coefficient to Decide The Linear Relationship that May Exist Between Two Variables of Interest; Find The Equation of Regression Line And Predict The Value of One Variable Based on the Value of the Other Variable..
- Analyze and interpret statistical inference using samples of a given size which is taken from a population.

Program / Semester: B.Tech (V Sem)	Branch: Artificial Intelligence and Machine Learning
Subject: Artificial Intelligence (Lab)	Course Code: C109521(022)
Total / Minimum-Pass Marks (End Semester Exam): 40/20	L: 0 T: 0 P: 2 Credits: 1

### **Course Objectives:**

- To provide skills for designing and analyzing AI based algorithms.
- To enable students to work on various AI tools.
- To provide skills to work towards solution of real life problems

### List of Experiments: (Each student should perform, at least, 10 experiments.)

- 1. Write a prolog program to find the rules for parent, child, male, female, son, daughter, brother, sister, uncle, aunt, ancestor given the facts about father and wife only.
- 2. Write a program to find the length of a given list
- 3. Write a program to find the last element of a given list
- 4. Write a program to delete the first occurrence and also all occurrences of a particular element in a given list.
- 5. Write a program to find union and intersection of two given sets represented as lists.
- 6. Write a program to read a list at a time and write a list at a time using the well defined read & write functions.
- 7. Write a program given the knowledge base,
  - If x is on the top of y, y supports x.
  - If x is above y and they are touching each other, x is on top of y.
  - A cup is above a book. The cup is touching that book. Convert the following into wff's, clausal form; Is it possible to deduce that 'The book supports the cup'.
- 8. Write a program given the knowledge base,
  - If Town x is connected to Town y by highway z and bikes are allowed on z, you can get to y from x by bike.
  - If Town x is connected to y by z then y is also connected to x by z.
  - If you can get to town q from p and also to town r from town q, you can get to town r from town p.
  - Town A is connected to Town B by Road 1. Town B is connected to Town C by Road 2.
  - Town A is connected to Town C by Road 3. Town D is connected to Town E by Road 4.
  - Town D is connected to Town B by Road 5. Bikes are allowed on roads 3, 4, 5.
  - Bikes are only either allowed on Road 1 or on Road 2 every day. Convert the following into wff's, clausal formand deduce that 'One can get to town B from town D'.
- 9. Solve the classical problems for demonstrating AI search heuristics: (Water Jug problem, Monkey Banana problem, Missionary Cannibals problem, Travelling Salesman Problem and alike).
- 10. Solve the classical Crypt arithmetic problems in AI: (DONALD + GERALD = ROBERT, CROSS + ROADS = DANGER, SEND + MORE = MONEY and alike).
- 11. Solve the classical Blocks World Problem demonstrating Planning Problem-solving simulation in AI.
- 12. Write a program to search any goal given an input graph using AO\* algorithm.

### List of Equipments/Machine required: PC with Wndows XP Operating System, Visual prolog compiler

### **Recommended Books:**

- 1. Ivan Bratko: Logic & prolog programming.
- 2. Carl Townsend: Introduction to Turbo Prolog, BPB, Publication.
- 3. W.F. Clocksin & Mellish: Programming in PRLOG, Narosa Publication House

## **Laboratory Outcomes** [After undergoing the course, students will be able to:]

- Acquire an overview of logic constructs for performing inferencing techniques. (First Order Predicate Calculus) in toy problems /classical problems using PROLOG / LISP syntax.
- Gain confidence in drafting production rules (iterative / recursive) for an AI simulating code, given a story domain.
- Understand, on how to use different data structures (lists, trees, stacks and queues) for solving routing problems and implementing heuristic searches.
- Gain exposure to deal with situations that crop up syntax / compile-time / run-time errors.
- Simulate game playing / puzzle problemsusing general solution in PROLOG / LISP syntax.

Program / Semester: B.Tech (V Sem)	Branch: Artificial Intelligence and Machine Learning
Subject: Machine Learning (Lab)	Course Code: C109522(022)
Total / Minimum-Pass Marks(End Semester Exam): 40/20	L: 0 T: 0 P: 2 Credits: 1

## **Course Objectives:**

- To introduce classical and foundational concepts, results, methodologies and applications in machine learning
- To develop abilities for developing a solution for a given problem
- The knowledge of using machine learning to make predictions
- The underlying mathematical relationships within and across machine learning algorithms
- Paradigms of supervised and un-supervised learning.

Hardware Requirement: i5 Processor, 8GB RAM, Internet Connection

Software Environment: IDE recommended PYCHARM (Recommended), JUPYTER, VISUAL STUDIO

- 1. Introduction to pandas
- 2. Introduction to NumPy
- 3. Wine Quality Prediction
- 4. Housing Price Prediction
- 5. Air Quality Prediction
- 6. Bank Marketing
- 7. Liver Disease Prediction
- 8. Heart Disease Prediction
- 9. Credit Default Prediction
- 10. Car Price Prediction
- 11. Media Content Problem
- 12. Online Retail Case Study
- 13. Airline Passengers Prediction
- 14. Energy Efficiency Analysis
- 15. Stock Price Prediction
- 16. Car Evaluation
- 17. Movie Sentiment Analysis
- 18. Program to demonstrate Simple Linear Regression
- 19. Program to demonstrate Logistic Regression using SCIKIT learn
- 20. Program to demonstrate Multiple Linear Regression
- 21. Program to demonstrate k-Nearest Neighbor flowers classification
- 22. Program to demonstrate Naive- Bayes Classifier
- 23. Program to demonstrate K-Medoid clustering algorithm
- 24. Program to demonstrate DBSCAN clustering algorithm
- 25. Program to demonstrate SVM based classification

## **Recommended Books:**

- Richard Duda, Peter Hart, David Stork, Pattern Classification, 2nd Ed, John Wiley & Sons, 2001. ISBN 9788126511167
- 2. Christopher Bishop. Pattern Recognition and Machine Learning. ISBN 0387310738.
- 3. Trevor Hastie, Robert Tibshirani, Jerome Friedman. Elements of Statistical Learning. ISBN 0387952845.
- 4. Tom Mitchell. Machine Learning. McGraw-Hill. ISBN 0070428077.
- 5. Shai Shalev-Shwartz, and Shai Ben-David, Understanding Machine Learning: From Theory to Algorithms, Cambridge University Press, 2014. ISBN 978-1-107-05713-5.

Program / Semester: B.Tech (V Sem)	Branch: Artificial Intelligence and Machine Learning
Subject: R-Programming (Lab)	Course Code: C109523(022)
Total / Minimum-Pass Marks (End Semester Exam): 40/20	L: 0 T: 0 P: 2 Credits: 1

### **Course Objective:**

- Students will be able to define the scientific method and explain its steps.
- Students will be able to identify and explain the different types of scientific evidence.
- Students will be able to design and conduct a scientific experiment.
- Students will be able to analyze and interpret the results of a scientific experiment.
- Students will be able to communicate the results of a scientific experiment in a clear and concise manner.

## 1. Experiment 1: Importing and cleaning data

In this experiment, students will learn how to import data from a variety of sources, such as CSV files, Excel files, and databases. They will also learn how to clean data by removing missing values, outliers, and duplicate rows.

## 2. Experiment 2: Data wrangling

In this experiment, students will learn how to transform data by changing the data types, merging data sets, and creating new variables. They will also learn how to explore data by using statistical methods such as descriptive statistics and hypothesis testing.

### 3. Experiment 3: Data visualization

In this experiment, students will learn how to create effective data visualizations using R. They will learn how to choose the right type of plot for the data, how to customize plots, and how to save plots.

### 4. Experiment 4: Statistical analysis

In this experiment, students will learn how to conduct descriptive and inferential statistical analysis using R. They will learn how to calculate descriptive statistics, such as mean, median, and standard deviation. They will also learn how to conduct hypothesis testing to determine if there is a statistically significant difference between two groups.

## 5. Experiment 5: Machine learning

In this experiment, students will learn how to apply machine learning algorithms to solve real-world problems. They will learn how to train and evaluate machine learning models, and how to use machine learning models to make predictions.

6. **Experiment 6:** Determine the unknown concentration of tartrazine in a sports drink. This experiment requires students to create dilutions of known concentrations of tartrazine as a reference to determine the unknown concentration of tartrazine in a sports drinkTartrazine

$$Na^{+}OOC$$
 $Na^{+}OOC$ 
 $Na^{$ 

- 7. **Experiment 7:** Design an experiment to determine the effect of different types of fertilizer on plant growth. This experiment allows students to explore the factors that affect plant growth.
- 8. **Experiment 8:** This experiment allows students to explore the relationship between food and energy.

- 9. **Experiment 9:** Design an experiment to determine the effect of different types of light on the growth of plants. This experiment allows students to explore the role of light in plant growth.
- 10. **Experiment 10:** Design an experiment to determine the effect of different types of soil on the growth of plants. This experiment allows students to explore the role of soil in plant growth.

## Suggested Textbook:

- 1. R for Data Science by Hadley Wickham and Garrett Grolemund
- 2. R for Data Analysis by Garrett Grolemund and Hadley Wickham
- 3. Introduction to Statistical Learning by Gareth James, Daniela Witten, Trevor Hastie, and Robert Tibshirani
- 4. Data Visualization: The Language of Graphics by Edward Tufte
- 5. R Packages by Hadley Wickham

#### **Course Outcome:**

- Students will be able to apply the scientific method to solve real-world problems.
- Students will be able to think critically and evaluate scientific evidence.
- Students will be able to communicate scientific information effectively.
- Students will be able to work collaboratively to solve scientific problems.
- Students will be able to use technology to conduct scientific research.

Program / Semester: <b>B.Tech (V)</b>	Branch: Artificial Intelligence and Machine Learning
Subject: Minor Project Phase-I	Course Code: C109524(022)
Total / Minimum-Pass Marks (End Semester Exam): 40/20	L: 0 T: 0 P: 2 Credits: 1

Pre-requisite: Knowledge of Minimum 2 Subjects & Laboratory

### **Course Objectives:**

- Acquire the knowledge of Computer Engineering and apply this knowledge to develop a Project Model.
- Understand and analyse the Engineering Problem
- Prepare a well-organized Documentation for the Project Developed

## **Guidelines for Project:**

Project work: Based on the ability/abilities of the student/s and recommendations of the mentor, a single discipline or a multidisciplinary project can be assigned to an individual student or to a group having not more than 4 students.

In extraordinary cases, like the funded projects requiring students from different disciplines, the project student strength can be 5 or 6.

### **Project Work Phase - 1:**

A committee consisting of the Head of the concerned Department and two senior faculty members of the Department, one of whom shall be the Guide, shall award the Sessional marks.

The sessional marks awarded for the project work phase -1, shall be based on the evaluation of the project work phase -1 Report (covering Literature Survey, Problem identification, Objectives and Methodology), project presentation skill and question and answer session in the ratio 50:25:25. The marks awarded for the Project report shall be the same for all the batch mates.

- Student has to Identify and formulate problem
- Students has to survey minimum 5 Research Papers of Reputed Publication
- Proposed Methodology and Objectives of the Project should be defined
- Adopt appropriate method to design selected problem
- Concerned guide and Review Committee will conduct timely evaluation of the project for SESSIONAL assessment.
- At the end of the semester students has to prepare and submit a well-organized project report.

## **Course Outcomes:** [After undergoing the course, students will be able to:]

- Demonstrate skill to identify and formulate the given problems.
- Apply basic engineering knowledge learnt in developing system individually or in-group
- Evaluate current research status by conducting literature survey.
- Design and develop real time applications
- Apply the programming language for the implementation of the project and prepare well-organized report.

Program / Semester: B. Tech (V)	Branch: Artificial Intelligence and Machine Learning
Subject: Internet of Things (IoT) (Professional Elective - I)	Course Code: C109531(022)
Total / Minimum-Pass Marks(End Semester Exam):100 / 35	L: 2 T: 0 P: 0 Credits: 2
Class Tests & Assignments to be conducted: 2 each	Duration (End Semester Exam): 03 Hours

### **Course Objectives:**

- To study the fundamentals about IoT.
- To study about the functional blocks and IoT Access technologies.
- To study the design methodology and different IoT hardware platforms.
- To study the basics of IoT Data Analytics and supporting services.
- To study about various IoT case studies.

**UNIT-I: Fundamentals of IoT:** Evolution of Internet of Things, Enabling Technologies, M2M Communication, IoT World Forum (IoTWF) standardized architecture, Simplified IoT Architecture, Core IoT Functional Stack, Fog, Edge and Cloud in IoT.

**UNIT-II: Functional Block and IoT Protocols:** Sensors, Actuators, Smart Objects, Connecting Smart Objects. **Access Technology:** Physical layer, MAC layers, Topology and Security of IEEE 802.15.4, 802.11ah and Lora WAN. **IP as the IoT Network Layer:** IP versions, Constrained Nodes and Constrained Networks, 6LoWPAN, IoT Application Transport Methods: SCADA, **IoT Application Layer Protocols:** CoAP and MQTT.

**UNIT-III: Design and Development:** Design Methodology, Embedded computing logic, Microcontroller, System on Chips, IoT system building blocks IoT Platform overview: Overview of IoT supported Hardware platforms such as: Raspberry pi, Arduino Board details

**UNIT-IV: Data Analytics and Supporting Services:** Data Analytics: Introduction, Structured Versus Unstructured Data, Data in Motion versus Data at Rest, IoT Data Analytics Challenges, Data Acquiring and storage, organizing the data. **Supporting Services:** Cloud computing paradigm for data collection, storage and computing, everything as a service and Cloud Service Models.

**UNIT-V: IoT Case Studies:** IoT/IIOT Applications in the Premises, Supply-Chain and Customer Monitoring, IoT applications for smart homes, cities, Environment Monitoring and Agriculture. **CASE STUDY:** Smart City Streetlights Control and Monitoring.

### **Text Books:**

- IoT Fundamentals: Networking Technologies, Protocols and Use Cases for Internet of Things, David Hanes, Gonzalo Salgueiro, Patrick Grossetete, Rob Barton and Jerome Henry, Cisco Press, 2017
- 2. Internet of Things A hands-on approach, Arshdeep Bahga, Vijay Madisetti, Universities Press, 2015 3. Internet of Things: Architecture, Design Principles and Applications, Rajkamal, McGraw Hill Higher Education

### **Reference Books:**

- 1. The Internet of Things Key applications and Protocols, Olivier Hersent, David Boswarthick, Omar Elloumi and Wiley, 2012.
- 2. "From Machine-to-Machine to the Internet of Things Introduction to a New Age of Intelligence", Jan Ho ller, Vlasios Tsiatsis, Catherine Mulligan, Stamatis, Karnouskos, Stefan Avesand. David Boyle and Elsevier, 2014.
- 3. Architecting the Internet of Things, Dieter Uckelmann, Mark Harrison, Michahelles and Florian (Eds), Springer, 2011.

#### **Course Outcomes:** At the end of this course, students will be able to

- Understand the basics of IoT.
- Implement the state of the Architecture of an IoT.
- Understand design methodology and hardware platforms involved in IoT.
- Understand how to analyze and organize the data.
- Compare IOT Applications in real world.

Program / Semester: B.Tech (V Sem)	Branch: Artificial Intelligence and Machine Learning
Subject: Introduction to Toolkits for Machine Learning	Course Code: C109533(022)
(Professional Elective - I)	Course Code. C109333(022)
Total / Minimum-Pass Marks (End Semester Exam): 100 / 35	L: 2 T: 0 P: 0 Credits: 2
Class Tests & Assignments to be conducted: 2 Each	Duration (End Semester Exam): 03 Hours

### **Course Objectives:**

- To introduce the use of various Machine Learning Toolkits.
- To introduce the use of Exploratory Data Analysis for performing various data processing activities.
- To introduce the use of the Seaborn library for data visualization activities.
- To introduce the concept of Ensemble Learning techniques.
- To introduce the use of Scikit-Learn Library for data analysis.

### **UNIT -I: Introduction to Machine Learning Toolkits**

Definition and purpose of machine learning toolkits, simplification process of toolkits, Overview of different types of machine learning toolkits such as models and techniques; Data Pre-processing and Feature Engineering-Exploring various pre-processing techniques, data scaling, handling missing values, and one-hot encoding, Feature selection and extraction methods, Model Evaluation and Performance Metrics.

## **UNIT-II: Exploratory Data Analysis**

Exploratory Analysis: Descriptive and comparative statistics, EDA explained using sample Data set; Exploratory Graphs: 3D scatter plot, Pair plot, and limitations, Histogram and introduction to PDF, Univariate analysis using PDF, Mean, variance, standard deviation, Box-plot with whiskers.

### **UNIT-III: Data Visualization using Seaborn**

Visualization: various plots viz multiple, categorical, distribution, Regression; Seaborn-Statistical Estimation; Time Series, Geo-location Data, Correlations and Connections, Hierarchies, Networks, Interactivit.

#### **UNIT-IV: Ensembles Methods**

Introduction to Ensembles Methods, Types of ensemble methods viz stacking, blending, bagging, boosting and its impact on bias and variance, C5.0 boosting, Random Forest, Gradient Boosting Algorithm (GBM), AdaBoost: Adaptive Boosting, Extreme Gradient Boosting (XGBoost), LightGBM.

### UNIT-V: Scikit-Learn (Open Source Data Analysis Library)

Overview of scikit-learn, Supervised Learning Models with scikit-learn viz linear regression, logistic regression, decision trees, and support vector machines; Introduction to unsupervised learning algorithms using scikit-learn library viz k-means, Hierarchical clustering, DBSACN, and Principal Component Analysis; Measuring classification and Clustering Performance parameters.

#### **Text Books:**

- 1. Data Mining Concepts and Techniques. Jiawei Han, Micheline Kamber and Jian Pei. Morgan Kaufmann
- 2. Publisher is an imprint of Elsevier.
- 3. Machine Learning in Action. Peter Harrington, Manning Publications, 2012.
- 4. Introduction to Machine Learning (3e), Ethem Alpaydin, MIT Press.

### **Reference Books:**

- 1. Python Data Analytics– Fabio Nelli, Apress.
- 2. Python for Data Analysis, Wes McKinney, O'Reilly.
- 3. Foundations of Machine Learning(2e), Mohri Mehryar, Afshin Rostamizadeh, and Ameet Talwalkar, MIT Press.

## **Course Outcomes** [After undergoing the course, students will be able to:]

- Use various Machine Learning Toolkits available in Python.
- Apply the concept of Exploratory Data Analysis for performing various data processing activities.
- Apply the use of the Seaborn library for data visualization activities.
- Apply the Ensemble Learning techniques.
- Apply the Scikit-Learn Library for data analysis and data handling activities.

Program / Semester: B. Tech (V Sem)	Branch: Artificial Intelligence and Machine Learning
Subject: <b>Image Processing and Computer Vision</b> (Professional Elective - I)	Course Code: C109532(022)
Total / Minimum-Pass Marks (End Semester Exam): 100 / 35	L: 2 T: 0 P: 0 Credits: 2
Class Tests & Assignments to be conducted: 2 each	Duration (End Semester Exam): 03 Hours

## **Course Objectives:**

- Apply skills for automatic analysis of digital images to construct representations of physical objects and scenes.
- Designandimplementreal-lifeproblemsusingImageprocessingandcomputervision.

**UNIT-I:** Digital Image Formation and low-level processing: Overview and State-of-the-art, Fundamentals of Image Formation, Transformation: Orthogonal, Euclidean, Affine, Projective, etc. Fourier Transform, Convolution and Filtering, Image Enhancement, Restoration, Histogram Processing.

**UNIT-II:** Depth estimation and Multi-camera views, Multiple View Geometry Perspective, Binocular Stereopsis: Camera and Epipolar Geometry; Homography, Rectification, DLT, RANSAC, 3-Dreconstructionframework; Autocalibration.

**UNIT-III:** Feature Extraction Edges - Canny, LOG, DOG; Line detectors (Hough Transform), Corners -Harris and Hessian Affine, Orientation Histogram, SIFT, SURF, HOG, GLOH, Scale-Space Analysis Image Pyramids and Gaussian derivative filters, Gabor Filters and DWT.

**UNIT-IV:** Image Segmentation Region Growing, Edge Based approaches to segmentation, Graph-Cut, Mean-Shift, MRFs, Texture Segmentation; Object detection. Pattern Analysis Clustering: K-Means, K-Medoids, Mixture of Gaussians, Classification: Discriminant Function, Supervised, Un-supervised, Semi-supervised; Classifiers: Bayes, KNN, ANN models; Dimensionality Reduction: PCA, LDA, ICA; Non-parametric methods.

**UNIT-V:** Motion Analysis Background Subtraction and Modeling, Optical Flow, KLT, Spatio-Temporal Analysis, Dynamic Stereo; Motion parameter estimation. Shape from X Photometric Stereo; Use of Surface Smoothness Constraint; Shape from Texture, color, motion and edges.

### **Reference Books:**

- $1. \ Digital Image Processing using MATLAB, By: Rafael C. Gonzalez, Richard Eugene Woods, 2nd Edition, Tata McGraw-Hill Education 2010$
- 2. Richard Szeliski, Computer Vision: Algorithms and Applications, Springer-Verlag London Limited 2011.
- 3. Computer Vision: A Modern Approach, D.A. Forsyth, J. Ponce, Pearson Education, 2003.

Program / Semester: B.Tech (V Sem)	Branch: Artificial Intelligence and Machine Learning
Subject: Data Visualization (Professional Elective -I)	Course Code: C109534(022)
Total / Minimum-Pass Marks (End Semester Exam): 100 / 35	L: 2 T: 0 P: 0 Credits: 2
Class Tests & Assignments to be conducted: 2 each	Duration (End Semester Exam): 03 Hours

### **Course Objective:**

- To introduce the fundamental concepts of data visualization, including data types, variables, and visualization principles.
- To introduce different types of data visualization techniques and their appropriate applications.
- To introduce effective visualizations using appropriate chart types, color schemes, and labeling techniques.
- To introduce utilization of visualization tools and libraries to create interactive and static visualizations.
- To introduce the use of data visualization in high dimension data handling and in Machine learning

**UNIT-I: Introduction to Data Visualization:** Introduction to data visualization and its importance, Perception, and cognition in visualization, Types of data and visualization tasks, Data visualization pipeline, Tools and software for data visualization.

**UNIT-II: Data Representation and Visualization Techniques:** Data types and structures for visualization, Visualization techniques for univariate and multivariate data, Graphs and charts for categorical and numerical data, Bar charts, line charts, scatter plots, Pie charts, histograms, box plots, Heatmaps, and geographic visualizations, Geospatial and temporal visualization, Interactive visualization, and visual analytics.

**UNIT-III: Principles of Visualization Design:** Principles of visual perception and cognition, Color theory and color encoding, Layout and composition in visualization, Effective use of typography and labels, Visualization storytelling, and narrative techniques.

**UNIT-IV: Tools and Technologies for Data Visualization:** Introduction to popular data visualization tools, Creating visualizations using Python libraries (Matplotlib, Seaborn, Plotly), Exploratory Data Analysis (EDA), Data exploration using visualizations, and Identifying patterns and relationships in data.

**UNIT-V: Advanced Data Visualization:** Visualizing large and high-dimensional datasets, Information visualization and network visualization, Visualization for machine learning and AI applications, Evaluation and critique of visualizations, Ethical considerations in data visualization.

### **Text Books:**

- 1. "Interactive Data Visualization for the Web" by Scott Murray.
- 2. "The Big Book of Dashboards: Visualizing Your Data Using Real-World Business Scenarios" by Steve Wexler, Jeffrey Shaffer, and Andy Cotgreave.
- 3. "3. Data Visualization: A Practical Introduction" by Kieran Healy.

## **Reference Books:**

- 1. "Information Visualization: Perception for Design" by Colin Ware
- 2. "The Wall Street Journal Guide to Information Graphics: The Dos and Don'ts of Presenting Data, Facts, and Figures" by Dona M. Wong
- 3. "Data Visualisation: A Handbook for Data Driven Design" by Andy Kirk.

### **Course Outcome:** By the end of this course, students will be able to:

- Understand the fundamental concepts and principles of data visualization.
- Apply various visualization techniques to explore and analyze datasets.
- Design and create effective visualizations using appropriate tools and software.
- Interpret and communicate insights gained from data visualizations.
- Critically evaluate and improve existing visualizations based on principles of perception and design.
- Apply data visualization techniques to real-world problems in AI and machine learning.

Program / Semester: B.Tech (V Sem)	Branch: Artificial Intelligence and Machine Learning
Subject: Environmental Science	Course Code:
Total / Minimum-Pass Marks (End Semester Exam): 10	L:0T:0P:2 Credits:-

### **Course Objectives:**

- Understanding the importance of ecological balance for sustainable development.
- Understanding the impacts of developmental activities and mitigation measures.
- Understanding the environmental policies and regulations

**UNIT-I:** Ecosystems: Definition, Scope, and Importance of ecosystem. Classification, structure, and function of an ecosystem, Food chains, food webs, and ecological pyramids. Flow of energy, Biogeochemical cycles, Bioaccumulation, Biomagnifications, ecosystem value, services and carrying capacity, Field visits.

**UNIT-II:** Natural Resources: Classification of Resources: Living and Non-Living resources, water resources: use and over utilization of surface and ground water, floods and droughts, Dams: benefits and problems. Mineral resources: use and exploitation, environmental effects of extracting and using mineral resources, Land resources: Forest resources, Energy resources: growing energy needs, renewable and non-renewable energy sources, use of alternate energy source, case studies.

**UNIT-III: Biodiversity and Biotic Resources:** Introduction, Definition, genetic, species and ecosystem diversity. Value of biodiversity; consumptive use, productive use, social, ethical, aesthetic and optional values. India as a mega diversity nation, Hot spots of biodiversity. Field visit. Threats to biodiversity: habitat loss, poaching of wildlife, manwildlife conflicts; conservation of biodiversity: In-Situ and Ex-sit conservation. National Biodiversity act.

**UNIT-IV: Environmental Pollution and Control Technologies: Environmental Pollution:** Classification of pollution, Air Pollution: Primary and secondary pollutants, Automobile and Industrial pollution, Ambient air quality standards. Water pollution: Sources and types of pollution, drinking water quality standards. Soil Pollution: Sources and types, Impacts of modern agriculture, degradation of soil. Noise Pollution: Sources and Health hazards, standards, Solid waste: Municipal Solid Waste management, composition and characteristics of e-Waste and its management. Pollution control technologies: Wastewater Treatment methods: Primary, secondary and Tertiary.

**UNIT-V: Environmental Policy:** Environmental Protection act, Legal aspects Air Act-1981, Water Act, Forest Act, Wildlife Act, Municipal solid waste management and handling rules, biomedical waste management and handling rules, hazardous waste management and handling rules.

#### **Text Books:**

- 1. Textbook of Environmental Studies for Undergraduate Courses by Erach Bharucha for University Grants Commission.
- 2. Environmental Studies by R. Rajagopalan, Oxford University Press.

### **Reference Books:**

- 1. Environmental Science: towards a sustainable future by Richard T. Wright. 2008 PHL Learning Private Ltd. New Delhi.
- 2. Environmental Engineering and science by Gilbert M. Masters and Wendell P. Ela. 2008 PHI Learning Pvt. Ltd.
- 3. Environmental Science by Daniel B. Botkin & Edward A. Keller, Wiley INDIA edition.
- 4. Environmental Studies by Anubha Kaushik, 4th Edition, New age international publishers.

- 5. Text book of Environmental Science and Technology Dr. M. Anji Reddy 2007, BS Publications.
- 6. Introduction to Environmental Science by Y. Anjaneyulu, BS. Publications.

## **Course Outcomes:**

On successful completion of the course, the student will be able to:

- Demonstrate an understanding of basic concept of ecosystem.
- Demonstrate an awareness of importance of natural resources.
- Demonstrate an awareness of biotic resources.
- Understand the pollution caused by different sources.
- Understand/evaluate/develop technologies on the basis of ecological principles and environmental regulations which in turn helps in sustainable development.