

## VI Semester

MACHINE LEARNING LABORATORY			
Course Code	21AIL66	CIE Marks	50
Teaching Hours/Week(L:T:P:S)	0:0:2:0	SEE Marks	50
Total Hours of Pedagogy	24	Total Marks	100
Credits	1	Exam Hours	03
<b>Course Learning Objectives:</b> CLO 2. To learn and understand the Importance Machine learning Algorithms CLO 3. Compare and contrast the learning techniques like ANN approach, Bayesian learning and reinforcement learning. CLO 4. Able to solve and analyse the problems on ANN, Instance based learning and Reinforcement learning techniques. CLO 5. To impart the knowledge of clustering and classification Algorithms for predictions and evaluating Hypothesis.			
	<b>Prerequisite</b>		
	<ul style="list-style-type: none"> <li>Students should be familiarized about Python installation and setting Python environment</li> <li>Usage and installation of Anaconda should be introduced <a href="https://www.anaconda.com/products/individual">https://www.anaconda.com/products/individual</a></li> <li>Should have the knowledge about Probability theory, Statistics theory and linear Algebra.</li> <li>Should have the knowledge of numpy, pandas, scikit-learn and scipy library packages.</li> </ul>		
<b>Sl. No.</b>	<b>PART A – List of problems for which student should develop program and execute in the Laboratory</b>		
1	Aim: Illustrate and Demonstrate the working model and principle of Find-S algorithm.  Program: For a given set of training data examples stored in a .CSV file, implement and demonstrate the Find-S algorithm to output a description of the set of all hypotheses consistent with the training examples.  <b>Text Book 1: Ch2</b>		
2	Aim: Demonstrate the working model and principle of candidate elimination algorithm.  Program: For a given set of training data examples stored in a .CSV file, implement and demonstrate the Candidate-Elimination algorithm to output a description of the set of all hypotheses consistent with the training examples.  <b>Text Book 1: Ch2</b>  Reference: <a href="https://www.youtube.com/watch?v=tfpAm4kxGQI">https://www.youtube.com/watch?v=tfpAm4kxGQI</a>		
3	Aim: To construct the Decision tree using the training data sets under supervised learning concept.  Program: Write a program to demonstrate the working of the decision tree based ID3 algorithm. Use an appropriate data set for building the decision tree and apply this knowledge to classify a new sample.  <b>Text Book 1: Ch 3</b>		
4	Aim: To understand the working principle of Artificial Neural network with feed forward and feed backward principle.  Program: Build an Artificial Neural Network by implementing the Backpropagation algorithm and test the same using appropriate data sets.  <b>Text Book 1: Ch 4</b>		

5	<p>Aim: Demonstrate the text classifier using Naïve bayes classifier algorithm.</p> <p>Program: Write a program to implement the naive Bayesian classifier for a sample training data set stored as a .CSV file. Compute the accuracy of the classifier, considering few test data sets.</p> <p><b>Text Book 1: Ch6</b></p>
6	<p>Aim: Demonstrate and Analyse the results sets obtained from Bayesian belief network Principle.</p> <p>Program:- Write a program to construct a Bayesian network considering medical data. Use this model to demonstrate the diagnosis of heart patients using standard Heart Disease Data Set. You can use Python ML library classes/API.</p> <p><b>Text Book 1: Ch 6</b></p>
7	<p>Aim: Implement and demonstrate the working model of K-means clustering algorithm with Expectation Maximization Concept.</p> <p>Program: Apply EM algorithm to cluster a set of data stored in a .CSV file. Use the same data set for clustering using k-Means algorithm. Compare the results of these two algorithms and comment on the quality of clustering. You can add Python ML library classes/API in the program.</p> <p><b>Text Book 1: Ch 8</b></p>
8	<p>Aim: Demonstrate and analyse the results of classification based on KNN Algorithm.</p> <p>Program: Write a program to implement k-Nearest Neighbour algorithm to classify the iris data set. Print both correct and wrong predictions. Java/Python ML library classes can be used for this problem.</p> <p><b>Text Book 1: Ch 8</b></p>
9	<p>Aim: Understand and analyse the concept of Regression algorithm techniques.</p> <p>Program: Implement the non-parametric Locally Weighted Regression algorithm in order to fit data points. Select appropriate data set for your experiment and draw graphs.</p> <p><b>Text Book 1: Ch8</b></p>
10	<p>Aim: Implement and demonstrate classification algorithm using Support vector machine Algorithm.</p> <p>Program: Implement and demonstrate the working of SVM algorithm for classification.</p> <p><b>Text Book 2: Ch6</b></p>
<b>Pedagogy</b>	For the above experiments the following pedagogy can be considered. Problem based learning, Active learning, MOOC, Chalk & Talk
<b>PART B</b>	
	A problem statement for each batch is to be generated in consultation with the co-examiner and student should develop an algorithm, program and execute the Program for the given problem with appropriate outputs.
<p><b>Course Outcomes:</b> At the end of the course the student will be able to:</p> <p>CO 1. Understand the Importance of different classification and clustering algorithms.</p> <p>CO 2. Demonstrate the working of various algorithms with respect to training and test data sets.</p> <p>CO 3. Illustrate and analyze the principles of Instance based and Reinforcement learning techniques.</p> <p>CO 4. Elicit the importance and Applications of Supervised and unsupervised machine learning.</p> <p>CO 5. Compare and contrast the Bayes theorem principles and Q learning approach.</p>	
<p><b>Assessment Details (both CIE and SEE)</b></p> <p>The weightage of Continuous Internal Evaluation (CIE) is 50% and for Semester End Exam (SEE) is 50%. The minimum passing mark for the CIE is 40% of the maximum marks (20 marks). A student</p>	

shall be deemed to have satisfied the academic requirements and earned the credits allotted to each course. The student has to secure not less than 35% (18 Marks out of 50) in the semester-end examination (SEE).

### **Continuous Internal Evaluation (CIE):**

CIE marks for the practical course is **50 Marks**.

The split-up of CIE marks for record/ journal and test are in the ratio **60:40**.

- Each experiment to be evaluated for conduction with observation sheet and record write-up. Rubrics for the evaluation of the journal/write-up for hardware/software experiments designed by the faculty who is handling the laboratory session and is made known to students at the beginning of the practical session.
- Record should contain all the specified experiments in the syllabus and each experiment write-up will be evaluated for 10 marks.
- Total marks scored by the students are scaled down to 30 marks (60% of maximum marks).
- Weightage to be given for neatness and submission of record/write-up on time.
- Department shall conduct 02 tests for 100 marks, the first test shall be conducted after the 8<sup>th</sup> week of the semester and the second test shall be conducted after the 14<sup>th</sup> week of the semester.
- In each test, test write-up, conduction of experiment, acceptable result, and procedural knowledge will carry a weightage of 60% and the rest 40% for viva-voce.
- The suitable rubrics can be designed to evaluate each student's performance and learning ability. Rubrics suggested in Annexure-II of Regulation book
- The average of 02 tests is scaled down to **20 marks** (40% of the maximum marks). The Sum of scaled-down marks scored in the report write-up/journal and average marks of two tests is the total CIE marks scored by the student.

### **Semester End Evaluation (SEE):**

- SEE marks for the practical course is 50 Marks.
- SEE shall be conducted jointly by the two examiners of the same institute, examiners are appointed by the University
- All laboratory experiments are to be included for practical examination.
- (Rubrics) Breakup of marks and the instructions printed on the cover page of the answer script to be strictly adhered to by the examiners. **OR** based on the course requirement evaluation rubrics shall be decided jointly by examiners.
- Students can pick one question (experiment) from the questions lot prepared by the internal /external examiners jointly.
- Evaluation of test write-up/ conduction procedure and result/viva will be conducted jointly by examiners.
- General rubrics suggested for SEE are mentioned here, writeup-20%, Conduction procedure and result in -60%, Viva-voce 20% of maximum marks. SEE for practical shall be evaluated for 100 marks and scored marks shall be scaled down to 50 marks (however, based on course type, rubrics shall be decided by the examiners)
- *Students can pick one experiment from the questions lot of PART A with equal choice to all the students in a batch. For PART B examiners should frame a question for each batch, student should*

<p><i>develop an algorithm, program, execute and demonstrate the results with appropriate output for the given problem.</i></p> <ul style="list-style-type: none"> <li>• <i>Weightage of marks for PART A is 80% and for PART B is 20%. General rubrics suggested to be followed for part A and part B.</i></li> <li>• Change of experiment is allowed only once and Marks allotted to the procedure part to be made zero (Not allowed for Part B).</li> <li>• The duration of SEE is 03 hours</li> <li>• Rubrics suggested in Annexure-II of Regulation book</li> </ul>
<p><b>Text Books:</b></p> <ol style="list-style-type: none"> <li>1. Tom M Mitchell, "Machine Learning", 1st Edition, McGraw Hill Education, 2017.</li> <li>2. <u>Nello Cristianini, John Shawe-Taylor</u>, An Introduction to Support Vector Machines and Other Kernel-based Learning Methods, Cambridge University Press, 2013</li> <li>3. Allen B. Downey, "Think Python: How to Think Like a Computer Scientist", 2nd Edition, Green Tea Press, 2015. (Available under CC-BY-NC license at <a href="http://greenteapress.com/thinkpython2/thinkpython2.pdf">http://greenteapress.com/thinkpython2/thinkpython2.pdf</a>)</li> </ol>
<p><b>Suggested Web Links / E Resource</b></p> <ol style="list-style-type: none"> <li>1. <a href="https://www.kaggle.com/general/95287">https://www.kaggle.com/general/95287</a></li> <li>2. <a href="https://web.stanford.edu/~hastie/Papers/ESLII.pdf">https://web.stanford.edu/~hastie/Papers/ESLII.pdf</a></li> </ol>