

## COURSE SYLLABI (2019-2023)

SCHOOL OF ELECTRICAL ENGINEERING			W.E.F.	:	2020-21
THIRD YEAR BACHLOR OF TECHNOLOGY			COURSE NAME	:	Soft Computing
			COURSE CODE	:	ET363
			COURSE CREDITS	:	04
RELEASE DATE	:	01/07/2020	REVISION NO.	:	1.0

TEACHING SCHEME:		EVALUATION SCHEME:					
LECTURE	PRACTICAL	THEORY			PRACTICAL	PRESENTATION/ DEMONSTRATION	TOTAL
		ICE	ECE	IA			
3	2	35	35	30	Nil	50	150

### PRE-REQUISITE:

CS102: Applications Programming-Python

### COURSE OBJECTIVES:

- ET361.CEO.1: To illustrate the basic concepts and techniques of machine learning.
- ET361.CEO.2: To explore supervised and unsupervised learning paradigms of machine learning for regression and classification
- ET361.CEO.3: To develop a deeper understanding of several algorithms in machine learning.
- ET361.CEO.4: To evaluate and interpret the results of the machine learning algorithms for solving practical problems.

**COURSE OUTCOMES:**

The students after completion of the course will be able to

1. ET361.CO.1: Explain fundamentals of machine learning.
2. ET361.CO.2: Describe supervised and unsupervised learning.
3. ET361.CO.3: Analyze mathematically various machine learning approaches and paradigms.
4. ET361.CO.4: Implement machine learning solutions for classification, regression, and clustering problems.
5. ET361.CO.5: Compare various machine learning techniques and to get an insight of when to apply a particular machine learning approach.

**THEORY:**

Unit I	Machine Learning Fundamentals	8 Hours
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Introduction to Machine Learning, Types of Learning, Linear Regression, Logistic Regression, Concept of Underfitting and Overfitting, Bias-Variance Tradeoff, Model assessment, Cross Validations, Accuracy and Error measures, Confusion metric, Precision, Recall, F1 Score, Analysis of ROC, AUC.

Unit II	Statistics and Probabilistic Learning	8 Hours
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Probability Concept, Decision Trees, Random Forest, Naïve Bayes, Introduction to Ensemble Methods.

Unit III	Supervised Machine Learning Algorithm	6 Hours
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K Nearest Neighbors (KNN.) Support Vector Machine, Optimization Objective of SVM, Maximum Margin Principle, Lagrangian Multipliers for SVM and Kernel Function.

Unit IV	Artificial Neural Network	8 Hours
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Neural Network Representation, Perceptron, Activation Function and Types, Multilayer Network and Backpropagation Algorithm.

Unit V	Unsupervised Learning	8 Hours
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Feature Space, Dimensionality Reduction, Principal Components Analysis (PCA), Principal of Clustering, K-Means, Hierarchical Agglomerative clustering, Case Study Using Clustering Algorithm.

<b>THEORY:</b>		
<b>Unit VI</b>	<b>Open CV for Computer Vision:</b>	<b>6 Hours</b>
<p>Introduction to OpenCV, Installation, Image Operations eg. read, display, writing, reading and setting image properties. opening video, Capture Video/Frame from Camera, applying functions on frame, Drawing geometric shapes on frame eg. line, rectangle, circle, ellipse, polygon, text. Arithmetic Operations on Image, ROI, RGB and HSV Color Space.</p> <p>Case Study on Computer Vision using Machine Learning Algorithm</p>		

<b>PRACTICAL:</b>		
<b>Practical No. 1</b>	<b>Title: Experimental Data Analysis: Perform following operations on any open dataset available in Python/Kaggle</b>	<b>2 Hours</b>
<ul style="list-style-type: none"> <li>• Load data into a data frame from a csv or any other file format.</li> <li>• Identification of variables and data types.</li> <li>• Find Missing Values.</li> <li>• Replace/eliminate missing values</li> <li>• Drop unessential columns.</li> <li>• Find average/min/max of numeric columns.</li> <li>• Display summary of data frame.</li> <li>• Bivariate analysis using plots through seaborn functions, cleaning the data, plotting graphs.</li> </ul>		
<b>Practical No. 2</b>	<b>Title: Liner Regression and Logistic Regression Model Implementation on Given Dataset.</b>	<b>4 Hours</b>
<ul style="list-style-type: none"> <li>• Build a Linear Regression Model using Real estate price prediction dataset.</li> <li>• Developed a Logistic regression model for classification.</li> </ul>		
<b>Practical No. 3</b>	<b>Title: Implementation of Decision Tree, Random Forest, KNN, Naïve Bayes with hyperparameter tuning.</b>	<b>4 Hours</b>
<ul style="list-style-type: none"> <li>• Developed Supervised Learning Model on selected Dataset.</li> </ul>		
<b>Practical No. 4</b>	<b>Title: Machine Learning for Image Classification</b>	<b>2 Hours</b>
<ul style="list-style-type: none"> <li>• Use of SVM for Image Classification.</li> <li>• Implementation of PCA</li> </ul>		
<b>Practical No. 5</b>	<b>Title: Implementation of Unsupervised Machine Learning</b>	<b>2 Hours</b>
Implement both the k-means algorithm and the Hierarchical Agglomerative Clustering (HAC) algorithm.		

<b>Practical No. 6</b>	<b>Title: Implementation of IOT Solution using Machine Learning</b>	<b>4 Hours</b>
<ul style="list-style-type: none"> <li>• Data Collection.</li> <li>• Data Cleaning, Filtering and Feature Extraction</li> <li>• Evaluation and Identification of ML Model</li> <li>• Training the ML Model</li> <li>• Outcome Predication</li> <li>• ML Model Deployment.</li> </ul>		
<b>Practical No. 7</b>	<b>Title: ANN for Computer Vision</b>	<b>2 Hours</b>
<ul style="list-style-type: none"> <li>• Creating Simple Neural Network.</li> <li>• Implement ANN for Image Classification.</li> </ul>		
<b>Practical No. 8</b>	<b>Title: Open CV for Computer Vision</b>	<b>2 Hours</b>
<ul style="list-style-type: none"> <li>• Use Open CV Library for Image Processing</li> </ul>		

**TEXTBOOKS:**

1. Ethem Alpaydin, Introduction to Machine Learning, Second Edition, MIT Press, (ISBN: 978-0-262-01243-0).
2. Christopher Bishop, Pattern Recognition and Machine Learning, Second Edition, Springer. 2006, (ISBN-13: 978-0387310732).
3. Andreas C. Müller and Sarah Guido, "Introduction to Machine Learning with Python: A Guide for Data Scientists", First Edition, O'Reilly Media, ISBN 978-14-4936-941-5
4. Tom Mitchell, Machine Learning, First Edition, McGraw-Hill Science/Engineering/Math, 1997, (ISBN: 0070428077).

**REFERENCES:**

1. Trevor Hastie, Robert Tibshirani and Jerome Friedman, The Elements of Statistical Learning, Second Edition, Springer-Verlag, 2009, (ISBN: 978-0-387-84857-0)
2. Kevin Murphy, Machine Learning: A Probabilistic Perspective, MIT Press, 2012, (ISBN: 9780262018029).
3. Simon Haykin, Neural Networks: A comprehensive foundation, Prentice Hall International Inc. 1999, (ISBN: 0132733501).