

# INT234:PREDICTIVE ANALYTICS

L:2 T:0 P:2 Credits:3

**Course Outcomes:** Through this course students should be able to

CO1 :: explain the basics of data preprocessing and its implementation by using R programming Language.

CO2 :: define the basics of classification by using Supervised Learning Algorithms

CO3 :: make use of different Supervised learning techniques to predict numeric values

CO4 :: demonstrate the predictive models by using Neural networks and Support vector machines

CO5 :: classify the data by implementing unsupervised learning algorithms

CO6 :: illustrate the techniques to evaluate the model performance and various methods to improve it

## Unit I

**DATA PREPROCESSING** : Managing data with R, Exploring and understanding data, Exploring the structure of data, Exploring numeric variables, Exploring categorical variables, Exploring relationships between variables

## Unit II

**SUPERVISED LEARNING: CLASSIFICATION** : Lazy learning:Nearest neighbors, Probabilistic Learning: Using Naive Bayes, Divide and Conquer: Decision Trees and Rules

## Unit III

**SUPERVISED LEARNING : NUMERIC PREDICTION** : Forecasting Numeric Data, Simple Linear Regression, Polynomial Regression, Ordinary least squares estimation, Correlations

## Unit IV

**SUPERVISED LEARNING:DUAL USE** : Black Box Methods, Neural Networks, Support Vector Machines

## Unit V

**UNSUPERVISED LEARNING: CLUSTERING AND PATTERN DETECTION** : K-Means Clustering, K-means clustering intuition, K-means random initialization trap, K-means selecting number of clusters, Dataset gathering, Hierarchical Clustering, Association Rules, Finding Patterns, Market Basket Analysis Using Association Rules

## Unit VI

**MODEL PERFORMANCE** : Evaluation Model Performance, Improving Model Performance, Bagging, Boosting, Random forests

### List of Practicals / Experiments:

#### Practical 1: Managing Data with R

- Exploring and understanding the data and loading it into different data structures.

#### Practical 2: Basics Of Data Preprocessing

- Exploring numeric and categorical variables, and finding relationships between different variables.

#### Practical 3: Implementation of Lazy and Probabilistic learning algorithms.

- Classification based on Nearest Neighbor and Naïve Bayes.

#### Practical 4: Implementation of Divide and Conquer Algorithms.

- Classification using Decision Tree and Rules.

#### Practical 5: Implementation of Regression Algorithms.

- Forecasting using Simple Linear Regression, Polynomial Regression and Multiple Linear Regression Algorithms.

#### Practical 6: Defining Relationship between Numeric Values.

- Implementation of Ordinary least squares estimation and Correlation algorithms.

**Practical 7: Implementation of Dual Supervised Learning Algorithms.**

- Black Box Methods, Neural Networks and Support Vector Machines.

**Practical 8: Implementation of Clustering Algorithms.**

- K-Means Clustering and Hierarchical Clustering.

**Practical 9: Implementation of Association Rules.**

- Market Basket Analysis Using Association Rules

**Practical 10: Model Performance Testing.**

- Evaluation Model Performance, Improving Model Performance, Bagging, Boosting, and Random forests

**References:**

1. PREDICTIVE ANALYTICS: THE POWER TO PREDICT by ERIC SIEGEL, WILEY
2. PYTHON AND R FOR THE MODERN DATA SCIENTIST: THE BEST OF BOTH WORLDS by RICK J. SCAVETTA AND BOYAN ANGELOV, SHROFF/O'REILLY
3. EFFICIENT R PROGRAMMING: A PRACTICAL GUIDE TO SMARTER PROGRAMMING by COLIN GILLESPIE AND ROBIN LOVELACE, SHROFF/O'REILLY
4. APPLIED PREDICTIVE ANALYTICS: PRINCIPLES AND TECHNIQUES FOR THE PROFESSIONAL DATA ANALYST by DEAN ABBOTT, WILEY, 4th Edition, (2012)
5. R IN A NUTSHELL 2E by JOSEPH ADLER, O'REILLY
6. INTRODUCTION TO MACHINE LEARNING WITH R: RIGOROUS MATHEMATICAL ANALYSIS by SCOTT BURGER, SHROFF/O'REILLY