Department of Data Science Programme Outcomes and Programme Specific Outcomes MSc Data Science

Programme Outcomes

Upon completion of MSc in Data Science degree, graduates will possess the following data science skills and abilities.

PO1: Possess a theoretical understanding, explain and critically assess the key concepts and techniques from the disciplines defining modern data science and analytics.

PO2: Critically evaluate emerging data analysis technologies and how they can be applied to heterogeneous data at volume, scale and types, in order to get insight for business, scientific or social innovation.

PO3: Analyse in depth how data analysis techniques can be applied to a range of interdisciplinary research areas.

PO4: Effectively use modern data science programming languages and technologies to scrape, clean, organize, explore, visualize, and model large volumes and varieties of data.

PO5: Evaluate, select, combine and apply advanced skills, data science tools and techniques in the related areas of artificial intelligence to the design of solutions to data science and analytics tasks.

PO6: Prepare for careers as data scientists by proposing, planning, developing, evaluating and creating a commercially and/or research-wise relevant project and/or product for business, science and society.

PO7: Develop professional communication skills (e.g., writing, presentations, interviews, email etiquette, etc.), effective time and resource management skills as well as leadership and team working skills towards meeting organizational goals.

PO8: Understand, value and safeguard social, legal and ethical use of data that increasingly challenge and confront data scientists while developing data science systems.

PO9: Learn effectively and independently to acquire new knowledge and skills for the purpose of continuing professional development in related areas of data science.

Programme Specific Outcomes

PSO1: Show mastery over different applications of data analytics namely web analytics, customer analytics, supply chain analytics and social network analytics.

PSO2: Build software applications using new languages and tools such as Neo4J, Tableau, Julia, SpaCy and Rasa

PSO3: Develop a disruptive entrepreneurship spirit and integrate with the data science community.

PSO4: Identify and assess the needs of an organization for a data science task by conducting a needs assessment and communicating data science options and limitations that could meet organizational needs.

BISHOP HEBER COLLEGE (AUTONOMOUS), TIRUCHIRAPPALLI-620 017 M. Sc., Data Science

| (Applicab | le to Candidates | admitted from the | Academic Y | ear 2019- | 2020 onwa | ards |
|-----------|------------------|-------------------|------------|-----------|-----------|------|
| | | | | | | |

| G | | G | Course | Hours / | | | Marks | ; |
|-----|--------------------|--|----------------------------------|---------|-----|-----|-------|-----|
| Sem | Course | Code | Week | Credits | CIA | ESE | Total | |
| | Core I | Mathematical Foundation for Data Science | P19DS101 | 5 | 4 | 25 | 75 | 100 |
| | Core II | Problem Solving using Python and R | P19DS102 | 5 | 4 | 25 | 75 | 100 |
| | Core III | NoSQL Database Management | 5 | 4 | 25 | 75 | 100 | |
| Ι | Elective I | Probability and Statistical Methods Design and Analysis of Algorithms Software Engineering | P19DS1:1 P19DS1:A P19DS1:B | 5 | 4 | 25 | 75 | 100 |
| | Core Practical I | Problem Solving using Python and R Lab | P19DS1P1 | 5 | 3 | 40 | 60 | 100 |
| | Core Practical II | NoSQL Database Management Lab | P19DS1P2 | 5 | 3 | 40 | 60 | 100 |
| | | | 1 | | | 1 | Γ | |
| | Core IV | Regression analysis | P19DS204 | 4 | 4 | 25 | 75 | 100 |
| | Core V | Data and Visual Analytics | P19DS205 | 4 | 4 | 25 | 75 | 100 |
| | Core VI | Practical Machine Learning | P19DS206 | 4 | 4 | 25 | 75 | 100 |
| | Elective II | Natural Language Processing Multivariate Analysis | P19DS2:2 P19DS2:A | 4 | 4 | 25 | 75 | 100 |
| Π | Elective III | Health Care Data Analytics Basics of Bioinformatics | P19DS2:3 P19DS2:B | 3 | 3 | 25 | 75 | 100 |
| | Core Practical III | Data and Visual Analytics Lab | 3 | 40 | 60 | 100 | | |
| | Core Practical IV | Practical Machine Learning Lab | P19DS2P4 | 3 | 3 | 40 | 60 | 100 |
| | Core Practical V | Natural Language Processing Lab | 3 | 3 | 40 | 60 | 100 | |
| | VLO | RI/MI | P17VL2:1 P17VL2:2 | 2 | 2 | 25 | 75 | 100 |
| | | | T | Γ | Γ | 1 | | 1 |
| | Core VII | Time Series Analysis and Forecasting | P19DS307 | 5 | 4 | 25 | 75 | 100 |
| | Core VIII | Big Data Management and Analytics | P19DS308 | 5 | 4 | 25 | 75 | 100 |
| | Core IX | Social Media Analytics | P19DS309 | 4 | 4 | 25 | 75 | 100 |
| III | Elective IV | Image and Video Analytics Computational Genomics | P19DS3:4 P19DS3:A | 4 | 4 | 25 | 75 | 100 |
| | Core Practical VI | Big Data Management and Analytics Lab | P19DS3P6 | 5 | 3 | 40 | 60 | 100 |
| | Core Practical VII | Social Media Analytics Lab | P19DS3P7 | 5 | 3 | 40 | 60 | 100 |
| | Core Project-I | Project Preparation | P19DS3PJ | 2 | - | - | - | 30 |
| | Γ | | r | | | T | | |
| | Core X | Principles of Deep Learning | P19DS410 | 5 | 4 | 25 | 75 | 100 |
| | Core X | Web Development using Python | P19DS411 | 5 | 4 | 25 | 75 | 100 |
| IV | Elective V | Supply Chain Management Internet of Things | P19DS4:5 P19DS4:A | 5 | 4 | 25 | 75 | 100 |
| | Core Project | Core Project | P19DS4PJ | 15 | 5 | | | 100 |
| | | | Total C | redits | 90 | | | |

PROGRAMME ARTICULATION MATRIX

| Course | | | | | | Prog | ramme | Outco | mes | | | | |
|----------|------------|-----|-----|-----|-----|------|------------|------------|------------|------|------|------|------|
| Code | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
| P20DS101 | Н | Н | Н | - | М | М | М | Н | - | Н | Н | - | - |
| P19DS102 | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | Н | М |
| P20DS103 | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | - | - |
| P20DS1:1 | Н | Н | Н | - | М | М | М | Н | - | Н | Н | - | - |
| P19DS1P1 | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | Н | М |
| P19DS1P2 | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | Н | М |
| P20DS204 | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | Н | М |
| P19DS205 | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | Н | М |
| P19DS206 | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | Н | М |
| P20DS2:2 | Н | Н | Н | Н | М | М | М | Н | М | Н | Н | Н | М |
| P19DS2:3 | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | Н | М |
| P19DS2P3 | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | Н | М |
| P19DS2P4 | Н | Н | Н | - | М | М | М | Н | - | Н | Н | Н | L |
| P19DS2P5 | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | Н | М |
| P20DS307 | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | Н | - |
| P19DS308 | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | Н | М |
| P19DS309 | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | Н | - |
| P19DS3:4 | Н | Н | Н | Н | М | М | М | Н | L | Н | Н | Н | М |
| P19DS3P6 | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | Н | М |
| P19DS3P7 | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | Н | М |
| P20DS3P8 | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | Н | - |
| P20DS410 | L | L | L | М | Н | Н | Н | Н | Н | Н | Н | Н | М |
| P19DS4:5 | L | L | L | М | Н | Н | Н | Н | Н | Н | Н | Н | М |
| P20DS4PJ | Н | Н | Н | Н | М | М | М | Н | - | Н | Н | Н | М |

| CORE I: MATHEMATICAL FOUNDATION FOR DATA SCIENCE | | | | | | | | | |
|---|----------|------------|---|--|--|--|--|--|--|
| Semester | Ι | Hours/Week | 5 | | | | | | |
| Course Code | P19DS101 | Credits | 4 | | | | | | |

After the successful completion of this course the students will be able to

| S.No. | Course Outcomes | Level | Unit |
|-------|--|-------|------|
| CO1 | Determining basis and understanding linear mappings of vector spaces | K5 | Ι |
| CO2 | Solve systems of linear equations by use of the matrix | K5 | II |
| CO3 | Determine the Invertible linear map | K5 | II |
| CO4 | Evaluate eigenvectors and eigenvalues | K5 | III |
| CO5 | Explain the properties gradients and PDE | K6 | IV |
| CO6 | Summarize the applications in Data Science | K6 | V |

2. A. SYLLABUS

Unit-1. Introduction to Vector Spaces

Vector Spaces: Rn and Cn, lists, Fnand digression on Fields, Definition of Vector spaces, Subspaces, sums of Subspaces, Direct Sums, Span and Linear Independence, bases, dimension.

Unit-2. Linear Maps

Definition of Linear Maps - Algebraic Operations on - Null spaces and Injectivity - Range and Subjectivity - Fundamental Theorems of Linear Maps - Representing a Linear Map by a Matrix -Invertible Linear Maps - Isomorphic Vector spaces - Linear Map as Matrix Multiplication -Operators - Products of Vector Spaces - Product of Direct Sum - Quotients of Vector spaces.

Unit-3. Eigen Values, Eigen Vectors and Inner Product Spaces

Eigenvalues and Eigenvectors - Eigenvectors and Upper Triangular matrices – Eigenspaces and Diagonal Matrices - Inequalities on Linear Spaces - Norms on Linear Spaces - Inner products - Orthogonality – Unitary and Orthogonal Matrices - Norms for matrices

Unit-4. Calculus of several variables and basic Graph Theory

Functions of Several Variables - Limits and continuity in Higher Dimensions – Partial Derivatives – The Chain Rule - Directional Derivative and Gradient vectors - Tangent Planes and Differentials -Extreme Values and Saddle Points - Lagrange Multipliers. Graphs - subgraphs - factors - Paths cycles - connectedness - trees - Euler tours -Hamiltonian cycles - Planar Graphs - Digraphs.

Unit-5. Mathematics applied to Data Science

Singular value decomposition - Handwritten digits and simple algorithm - Classification of handwritten digits using SVD bases - Tangent distance - Text Mining.

B. TOPICS FOR SELF-STUDY

| S.No. | Topics | Web Links |
|-------|------------------------------|--|
| 1 | Mathematics for Data Science | https://www.coursera.org/specializations/mathema |
| | | tics-for-data-science |

| 2 | Mathematics for Machine Learning Specialization | coursera.org/specializations/mathematics- machine-learning |
|---|---|--|
| 3 | Topics in Mathematics of Data Science | https://ocw.mit.edu/courses/mathematics/18-s096- topics-in-mathematics-of-data-science-fall-2015/ |

C. TEXT BOOK(S)

- 1. S. Axler, Linear algebra done right, Springer, 2017.
- 2. Eldén Lars, Matrix methods in data mining and pattern recognition, Society for Industrial and Applied Mathematics, 2007.
- 3. M. D. Weir, J. Hass, and G. B. Thomas, Thomas' calculus. Pearson, 2016.
- 4. D. Jungnickel, Graphs, networks and algorithms. Springer, 2014.

D. REFERENCE BOOKS

- 1. E. Davis, Linear algebra and probability for computer science applications, CRC Press, 2012.
- 2. J. V. Kepner and J. R. Gilbert, Graph algorithms in the language of linear algebra, Society for Industrial and Applied Mathematics, 2011.
- 3. D. A. Simovici, Linear algebra tools for data mining, World Scientific Publishing, 2012.
- 4. P. N. Klein, Coding the matrix: linear algebra through applications to computer science, Newtonian Press, 2015.
- 5. J. Patterson and A. Gibson, Deep learning: a practitioner's approach. O'Reilly Media, 2017.
- 6. S. Sra, S. Nowozin, and S. J. Wright, Optimization for machine learning. MIT Press, 2012.M.

E. WEB LINKS

- 1. https://elitedatascience.com/learn-math-for-data-science
- 2. <u>https://machinelearningmastery.com/gentle-introduction-linear-algebra/</u>'

3. SPECIFIC LEARNING OUTCOMES (SLO)

| Unit | Contents | Learning Outcomes | Level |
|------|---|--|-------|
| Ι | Introduction to Vector Spaces | | |
| 1.1 | Vector Spaces | Elicit the vector space | K5 |
| 1.2 | Definition of Vector spaces | Define the Vector spaces | K2 |
| 1.3 | Subspaces | Identify the subspaces | K5 |
| 1.4 | Sums of Subspaces | Determine the sum of subspaces | K5 |
| 1.5 | Direct Sums | Explain the Direct Sums in vector spaces | K5 |
| 1.6 | Span and Linear Independence | Compute the span for the vector spaces | K5 |
| 1.7 | Bases | Determine the Bases of the vector spaces | K5 |
| 1.8 | Dimension | Explain the dimension of the vector spaces | K5 |
| II | Linear Maps | | |
| 2.1 | Definition of Linear Maps | Define the Linear Maps | K2 |
| 2.2 | Algebraic Operations on Null spaces and Injectivity | Compute the Algebraic Operations on Null spaces and Injectivity | K5 |
| 2.3 | Algebraic Operations on Range and Subjectivity | Compute the Algebraic Operations on Range and Subjectivity | K5 |
| 2.4 | Fundamental Theorems of Linear Maps | Explain the Theorems of Linear Maps | K5 |

| 2.5 | Representing a Linear Map by a Matrix | Represent the Linear Map as a Matrix | K4 |
|------|---|---|----------|
| 2.6 | Invertible Linear Maps | Determine the invertible of the Linear Maps | K5 |
| 2.7 | Isomorphic Vector spaces | Construct the Isomorphic Vector Spaces | K6 |
| 2.8 | Linear Map as Matrix Multiplication | Determine the Linear Map as Matrix Multiplication | K4 |
| 2.9 | Operators - Products of Vector Spaces | Explain the Products of Vector Spaces | K5 |
| 2.10 | Product of Direct Sum - Quotients of Vector spaces | Explain the Product of Direct Sum - Ouotients of Vector spaces | K5 |
| III | Eigen Values, Eigen Vectors and | Inner Product Spaces | |
| 3.1 | Eigenvalues and Eigenvectors | Explain the procedure to compute the Eigen Value and Eigen Vectors | K5 |
| 3.2 | Eigenvectors and Upper Triangular matrices | Determine the Eigen vectors Determine the Upper Triangular Matrices | K5 K5 |
| 3.3 | Eigenspaces and Diagonal Matrices | Identify the Eigen Spaces using Eigen vectors | K4 |
| 3.4 | Inequalities on Linear Spaces | Explain the Inequalities on Linear Spaces | K5 |
| 3.5 | Norms on Linear Spaces - Inner products | Understand the Norms on Linear Spaces | K4 |
| 3.6 | Orthogonality | Compute the Orthogonality of the Linear Maps | K5 |
| 3.7 | Unitary and Orthogonal Matrices | Construct the Orthogonal Matrices | K6 |
| 3.8 | Norms for matrices | Determine the Norm of the matrices | K5 |
| IV | Calculus of several variables and | basic Graph Theory | |
| 4.1 | Functions of Several Variables | Understand the functions of Several Variables | K2 |
| 4.2 | Limits and continuity in Higher Dimensions | Explain the Limits and continuity in Higher Dimensions | K4 |
| 4.3 | Partial Derivatives | Compute the Partial derivative for the functions | K4 |
| 4.4 | The Chain Rule | Apply the chain rule for the functions | K5 |
| 4.5 | Directional Derivative and Gradient vectors | Analyze the gradients and directional derivatives | K5 |
| 4.6 | Tangent Planes and Differentials | Construct the Tangent Planes and Differentials | K6 |
| 4.7 | Extreme Values and Saddle Points | Determine the Extreme Values and Saddle Points | K5 |
| 4.8 | Lagrange Multipliers | Explain the Lagrange Multipliers | K4 |
| 4.9 | Graphs , Subgraphs & Factors | Define Graphs , Subgraphs & Factors | K2 |
| 4.10 | Paths, Cycles & Connectedness | Explain the Paths, Cycles & Connectedness | K5 |
| 4.11 | Trees | Understand the concepts of Tree | K2 |
| 4.12 | Euler tours & Hamiltonian cycles | Determine the Euler tours & Hamiltonian cycles | K4 |
| 4.13 | Planar Graphs & Digraphs | Explain the Planar Graphs | K5 |
| V | Mathematics applied to Data Scie | nce | |

| 5.1 | Singular value decomposition | Understand the concept of Singular value decomposition | K2 |
|-----|---|--|----|
| 5.2 | Handwritten digits and simple algorithm | Explain the algorithm for Handwritten digits | K4 |
| 5.3 | Classification of handwritten digits using SVD bases | Construct the SVD for Handwritten digits | K6 |
| 5.4 | Tangent distance | Analyze the tangent distance for SVD | K6 |
| 5.5 | Text Mining. | Design the algorithm for Text Mining | K6 |

| L-l | Low | | | | | M-Mo | oderate |) | | | | H- H | ligh |
|-----|-----|-----|-----|-----|-----|------|------------|-----|-----|------|------|------|------|
| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
| CO1 | Η | | | Н | | | L | | | L | | | Н |
| CO2 | Н | L | Н | | | | L | | | L | Μ | | |
| CO3 | Μ | Μ | | Н | Μ | | L | | Н | | | Μ | Н |
| CO4 | Μ | L | | Н | Μ | | | L | Н | | Μ | | |
| CO5 | H | | Н | Н | Μ | | | Μ | | | | | Н |
| CO6 | Н | | Н | | Н | | Н | Н | Н | Н | | Μ | Н |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 2. Open Book Test.
- Cooperative Learning Report, Assignment, Group Presentation, Group Discussion, project Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).
- 4. Pre-Semester & End Semester Theory Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator: Dr. P. S. Eliahim Jeevaraj

| CORE II: PROBLEM SOLVING USING PYTHON AND R | | | | | | | | | | |
|--|----------|------------|---|--|--|--|--|--|--|--|
| Semester | Ι | Hours/Week | 5 | | | | | | | |
| Course Code | P19DS102 | Credits | 4 | | | | | | | |

After the successful completion of this course the students will be able to

| S.No. | Course Outcomes | Level | Unit |
|-------|--|-------|------|
| CO1 | Understand python basic syntax, usage of built in functions | K3 | Ι |
| CO2 | Understand conditional and looping statements and build user defined functions | K3 | Ι |
| CO3 | Explain the concepts of files using Python | K5 | II |
| CO4 | Develop object oriented programs in Python | K6 | III |
| CO5 | Access and Design the internet and database data | K6 | IV |
| CO6 | Understand R basic data structures and develop programs | K5 | V |

2. A. SYLLABUS

UNIT - I: Python Basics, Functions, Loops and Strings

Variables – Getting Inputs – Conditions – Catching exceptions – Function calls – Built-in functions – Type conversion functions and math functions – Parameters and arguments –While statement – Infinite loops -Continue statement – For loops – Strings -Slice - The in operator – String comparison – String methods- parsing strings – Format operator.

UNIT - II: Files and Lists

Opening files – Text files – Reading files – Searching through files – Writing files – Traversing list – List operations – List slice – List methods – Deleting elements – Built-in list functions – Objects, value and aliasing – List arguments.

UNIT - III: Dictionaries, Tuples and OOP

Dictionaries – Files and dictionaries – Looping and dictionaries – Tuples – Comparing tuples – Tuple assignments – Dictionaries and tuples – Tuples as keys in dictionaries – Creating objects – Encapsulation – Classes as types – Object lifecycle – Instances – Inheritance.

UNIT – IV: Internet Programming

Regular expressions – Character matching – Extracting data – Escape character – Designing simple web browser using sockets – Retrieving images using HTTP – Retrieving web pages using urllib – Reading binary files using urllib – Accessing data from databases

UNIT – V: Programming with R

Variables - Vector, matrix, arrays - List - Data Frames - Functions - Strings - Factors - Loops - Packages -Date and Time - Files - Making packages

B. TOPICS FOR SELF-STUDY

| S.No. | Topics | Web Links |
|-------|--------|-----------|
|-------|--------|-----------|

| 1 | Introduction to Python Programming | https://www.udacity.com/course/introduction-to- |
|---|------------------------------------|--|
| | | pythonud1110 |
| 2 | Introduction to Python | https://www.coursera.org/projects/introduction-to- |
| | | python |
| 3 | Introduction to Python | https://realpython.com/learning-paths/python3- |
| | | introduction/ |
| 4 | R Programming | https://www.coursera.org/learn/r-programming |

C. TEXT BOOK(S)

- 1. Allen B. Downey, —Think Python: How to Think like a Computer Scientist, 2nd edition, Updated for Python O_Reilly Publishers, 2016
- 2. Charles R. Severance, Python for Everybody: "Exploring data using Python 3", Schroff Publishers, 1ed, 2017, ISBN 978-9352136278.
- 3. Richard Cotton, "Learning R", O'Reilly, 2013

D. REFERENCE BOOKS

- 1. Zed Shaw's, Learn Python the Hard Way: A Very Simple Introduction to the Terrifyingly Beautiful World of Computers and Code, Addison-Wesley Professional; 3 edition, 2013
- 2. Robert Sedgewick, Kevin Wayne, Robert Dondero, Introduction to Programming in Python: An Inter

disciplinary Approach, Pearson India Education Services Pvt. Ltd., 2016.

- 3. Wesley J Chun, Core Python Programming , 2nd edition, Prentice Hall ,2009
- 4. Colin Gillespie, Robin Lovelace, and Efficient R Programming: A Practical Guide to Smarter Programming," O'Reilly Media, Inc.", 2016
- 5. Paul Teetor, R Cookbook-Proven Recipes for Data Analysis, Statistics, and Graphics, O'Reilly Media, 2011

E. WEB LINKS

- <u>https://www.kaggle.com/learn/python</u>
- https://www.dataquest.io/course/introduction-to-data-analysis-in-r/

3. SPECIFIC LEARNING OUTCOMES (SLO)

| Unit/ Section | Торіс | Learning outcomes | Level | | | |
|------------------|------------------------------------|--|-------|--|--|--|
| Ι | Python 1 | Basics, Functions, Loops and Strings | | | | |
| 1.1 | Python basics | Understand python variables and assignment | K1 | | | |
| | | Built in Functions | | | | |
| 1.2 | Built in Functions and | • Understand python built in functions | K1 | | | |
| | other important functions | Understand conversion and math functions | K1 | | | |
| 1.3 | Conditional and looping statements | • Build if, else statements within programs and understand outputs | K3 | | | |
| | | Build while and for loops for understanding looping concept | K3 | | | |
| | | User Defined Functions | | | | |
| 1.4 | User Defined Functions | Understanding functions structures | K2 | | | |
| | | • Understanding parameters and arguments | K2 | | | |
| | Manipulating strings | | | | | |
| 1.5 | Handling strings | • Understanding strings data type | K1 | | | |
| | | Applying string slicing | K2 | | | |

| | | Applying string comparison, parsing | К3 |
|----------|-----------------------------|---|------|
| | | and string formatting | |
| | | Files and Lists | |
| 2.1 | Handling files | • Understand syntax to read and write files | K2 |
| | | Using io library functions to check file and folder existence | K3 |
| | | Lists | |
| 2.2 | List data structure | • Understanding list data structure and operations | K1 |
| | | Applying list slicing and items deletion | K2 |
| | | Lists and loops | |
| 2.3 | Using lists in loops | Using list comprehension in programs | К3 |
| III | | Dictionaries Tunles and OOP | |
| 3.1 | Dictionary data structure | Understanding dictionary data | |
| 0.11 | | structure | K1 |
| | | Loops and Dictionaries | |
| 3.2 | Using dictionaries in loops | • Applying dictionary comprehension in | WO |
| | | programs | K2 |
| | | Tuples | |
| 3.3 | Tuples data structure | • Understanding tuples data structure | K1 |
| 3.4 | Tuple Operations | • Applying tuples operations | K1 |
| | | Dictionaries and Tuples | |
| 3.5 | Dictionaries and Tuples | • Applying tuples as keys in dictionaries | K2 |
| | 1 | Objects and Classes | |
| 3.6 | Objects and Classes | Understanding objects and classes | K4 |
| | structure | Understanding inheritance | K5 |
| | | Creating objects and Classes | K4 |
| IV | | Internet Programming | |
| 4.1 | Regular Expressions | • Understanding pattern matching in | |
| | | strings | K4 |
| | | • Applying re module functions for | K5 |
| | | pattern matching in various examples | |
| 4.2 | Detrieving integes in web | | K2 |
| 4.2 | Retrieving images in web | • Using HTTP to retrieve images | К3 |
| 4.3 | Retrieving web pages | • Using urllib module to retrieve web | K3 |
| | A | pages | |
| 4.4 | A Extracting data from | CCessing data from databases | |
| 4.4 | databases | • Using modules to extract data from | K4 |
| N/ | uatabases | SQL databases | |
| V | D data stressterres | | W0 |
| 5.1 | R data structures | Understand R data types | K2 |
| | | Understand R data structures | K2 |
| | | Functions | 17.0 |
| 5.2 | User defined Functions | Creating user defined functions | K3 |
| | | R libraries | |
| 5.3 | K libraries | Using R libraries | K2 |
| | | Creating user defined R libraries | K6 |

| L-l | Low | | | | | M-Mo | oderate | • | | | | H- H | ligh |
|------------|-----|-----|-----|-----|-----|------|------------|-----|-----|------|------|------|------|
| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
| CO1 | H | Μ | Н | Μ | L | Μ | Μ | Μ | Μ | Μ | - | Н | Н |
| CO2 | Н | Н | Μ | L | - | L | - | L | Μ | Μ | - | Μ | - |
| CO3 | H | Н | Н | Н | Μ | Μ | L | - | - | Н | Н | Н | М- |
| CO4 | H | Н | Н | Н | Н | Μ | L | Μ | Μ | Н | Н | Μ | Н |
| CO5 | H | Μ | - | Μ | L | Μ | - | Н | Μ | Н | Н | Μ | - |
| CO6 | Μ | Μ | - | H | Μ | L | - | - | L | Н | H | Н | Μ |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 2. Open Book Test.
- 3. Cooperative Learning Report, Assignment, Group Presentation, Group Discussion, project, Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).
- 4. Pre-Semester & End Semester Theory Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator: Prof. K. Jemimah

| Core III: NoSQL DATABASE MANAGEMENT | | | | | |
|-------------------------------------|----------|------------|---|--|--|
| Semester | Ι | Hours/Week | 5 | | |
| Course Code | P19DS103 | Credits | 4 | | |

After the successful completion of this course the students will be able to

| S.No. | Course Outcomes | Level | Unit |
|-------|--|-------|------|
| CO1 | Construct the Queries and sub queries in SQL | K6 | Ι |
| CO2 | Construct queries to retrieve the data from more than one tables using different techniques. | K6 | II |
| CO3 | Design and Analyse different techniques and operations in Mango DB | K6 | III |
| CO4 | Assess various operators and clause to generate unstructured data | K5 | IV |
| CO5 | Choose different indices to retrieve data | K5 | IV |
| CO6 | Construct databases using SQL, MongoDB and Neo4J | K6 | V |

2. A. SYLLABUS

Unit-1. Data Modeling

DBMS: terminologies, components, roles, advantages and disadvantages – Database architectures: teleprocessing, file server, 2-tier, 3-tier, N-tier, middleware and Transaction processing monitor – Software components of DBMS and Database Manager – Data modeling using ER diagram: Entity, relationship, attributes, keys, strong and weak entities, attributes on relationships, relationship types, cardinality and participation

Unit-2. Structured Query Language

SQL statements: SELECT, WHERE, ORDERBY, GROUPBY and HAVING clauses - Sub Queries – ANY and ALL – JOIN – inner and outer joins – EXISTS and NON EXISTS – UNION, INTERSECT and EXCEPT – Updating databases: INSERT, UPDATE and DELETE – SQL data types – Creating, altering and removing tables – Indexes and views: CREATE and REMOVE.

Unit-3. NoSQL Database Theory

Why NoSQL – Value of Relational Database – Emergence of NoSQL – Aggregate data models – More details on data models: Relationships, Graphs DB, Schemaless DB, Materialized views – Distribution models: Single server, shrading, replication – Consistency: Update, read, relax consistency

Unit-4. NoSQL Databases

Key value databases: What is Key Value store, Features of Key value DB, Suitable use cases, When not to use it – Document databases: Definition, features, Suitable use cases, when not to use – Column family stores: Definition, features, suitable use cases, when not to use – Graph databases: Definition, features, use case, when not to use – Schema migration – Polyglot persistence - Beyond NoSQL – Choosing your database.

Unit-5. MongoDB

Document – Collection – Database - Datatypes – Creating, deleting, updating documents – Querying – Indexing – Aggregation: Pipeline, Aggregation commands – Application design.

B. TOPICS FOR SELF-STUDY

| S.No. | Topics | Web Links |
|-------|-----------------------|---|
| 1 | Database Architecture | https://www.youtube.com/watch?v=W6P58yb-edE |
| 2 | Normalization | https://www.guru99.com/database- |

| | | normalization.html |
|---|-------------|---|
| 3 | DynamoDB | https://www.tutorialspoint.com/dynamodb/index.h |
| | | <u>tm</u> |
| 4 | Apache HIVE | https://data-flair.training/blogs/apache-hive- |
| | | tutorial/ |

C. TEXT BOOKS

- 1. Thomas M. Connolly and Carolyn E. Begg. Database Systems: "A Practical Approach to Design, Implementation, and Management", 6th Edition, Pearson, 2015.
- 2. Pramod J. Sadalage; Martin Fowler. NoSQL Distilled: A Brief Guide to the Emerging World of PolyglotPersistence. Addison-Wesley. 2012 ISBN: 0321826620
- 3. Kristina Chodorow, MongoDB: The Definitive Guide, 2ed, Oreilly Publishers

D. REFERENCES BOOKS

1. Eric Redmond; Jim R. Wilson. Seven Databases in Seven Weeks: A Guide to Modern Databases and the NoSQL Movement. Pragmatic Bookshelf. 2012. ISBN: 1934356921

E. WEB LINKS

- 1. <u>https://www.simplilearn.com/introduction-to-nosql-databases-tutorial-video</u>
- 2. https://www.w3schools.com/sql/

3. SPECIFIC LEARNING OUTCOMES (SLO)

| Unit | Contents | Learning Outcomes | Level | | | | |
|------|---|--|----------|--|--|--|--|
| Ι | Data Modelling | | | | | | |
| 1.1 | DBMS: terminologies, components, roles, advantages and disadvantages | Understand the concepts of DBMS | K4 | | | | |
| 1.2 | Database architectures | Explain the Database architectures | K5 | | | | |
| 1.3 | Software components of DBMS and Database Manager | List the Software components of DBMS and Database Manager | K2 | | | | |
| 1.4 | Data modeling using ER diagram | Design the ER Diagram for Data Models | K6 | | | | |
| II | Structured Query Language | | | | | | |
| 2.1 | SQL statements | Use the Sql statements for Data bases | K4 | | | | |
| 2.2 | Sub Queries | Apply the sub queries for the databases | K5 | | | | |
| 2.3 | Updating databases | Explain the procedure to update the DB | K5 | | | | |
| 2.4 | SQL data types | Explain the SQL data types | K5 | | | | |
| 2.5 | Indexes and views | Create the Indexes and views | K6 | | | | |
| III | NoSQL Database Theory | | | | | | |
| 3.1 | NoSQL | Explain the NoSQL concepts | K5 | | | | |
| 3.2 | Aggregate data models | Construct the Aggregate data models | K6 | | | | |
| 3.3 | More details on data models: Relationships, Graphs DB, Schemaless DB, Materialized views | Compare the Data models Assess each data models features | K5 K6 | | | | |
| 3.4 | Distribution models | Explain the Distribution models. | K5 | | | | |
| 3.5 | Consistency | Develop the procedure for maintaining the consistency | K4 | | | | |

| IV | NoSQL Databases | | |
|-----|---|---|----------|
| 4.1 | Key value databases | Explain the Key value databases | K5 |
| 4.2 | Document databases | Design the Document Databases | K6 |
| 4.3 | Column family stores | Explain the Column Family Stores | K5 |
| 4.4 | Graph databases | Construct the Graph Databases | K6 |
| 4.5 | Schema migration – Polyglot persistence - Beyond NoSQL – Choosing your database | Explain the features of Database Design the custom databases | K5 K6 |
| V | MongoDB | | |
| 5.1 | Document – Collection | Create the Document in MangoDB | K6 |
| 5.2 | Database - Datatypes – Creating, deleting, updating documents | Explain the features of MongoDB | K5 |
| 5.3 | Querying | Construct the Queries for MongoDB | K6 |
| 5.4 | Indexing – Aggregation: Pipeline, Aggregation commands | Create a index for DB | K6 |
| 5.5 | Application design | Design the Applications using MongoDB | K6 |

| L-I | Low | | M-Moderate | | | | | | | | H- H | ligh | |
|-----|-----|-----|------------|-----|-----|-----|------------|-----|------------|------|------|------|------|
| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
| CO1 | H | Μ | Н | Μ | L | Μ | Μ | Μ | Μ | Μ | - | Н | Н |
| CO2 | H | H | Μ | L | - | L | - | L | Μ | Μ | - | Μ | - |
| CO3 | H | Н | Н | Н | Μ | Μ | L | - | - | Н | Н | Н | М- |
| CO4 | H | H | Н | Н | Н | Μ | L | Μ | Μ | Н | Н | Μ | Н |
| CO5 | H | Μ | - | Μ | L | Μ | - | Н | Μ | Н | Н | Μ | - |
| CO6 | Μ | Μ | - | Н | Μ | L | - | - | L | Н | Н | Н | М |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 2. Open Book Test.
- 3. Cooperative Learning Report, Assignment, Group Presentation, Group Discussion, project Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).
- 4. Pre-Semester & End Semester Theory Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator: Dr. M. Lovelin Pon Felichiah ELECTIVE-1: PROBABILITY AND STATISTICAL METHODS

| Semester | Ι | Hours/Week | 5 |
|-------------|----------|------------|---|
| Course Code | P19DS1:1 | Credits | 4 |

After the successful completion of this course the students will be able to

| S.No. | Course Outcomes | Level | Unit |
|-------|---|-------|------|
| CO1 | Solve problems using Probability | K3 | Ι |
| CO2 | Apply Expectation and Regression | K4 | II |
| CO3 | Demonstrate and apply Distribution | K5 | III |
| CO4 | Develop Estimation using diffe | K5 | IV |
| CO5 | Create Hypothesis Testing | K6 | V |
| CO6 | Analysis of Variance | K6 | V |

2. A. SYLLABUS

Unit-1. Probability

Probability Spaces-Combinatorial methods (or) Counting techniques-Elementary Theorem – Conditional Probability –Bayes" theorem-Probability Distributions and Probability Densities.

Unit-2. Expectation and Regression

Mathematical Expectation: Expected value-Moments-Chebyshev's theorem-Moment Generating functions Product Moment-Conditional Expectation-Special Probability Distributions and Probability Densities Functions of Random Variable. Multiple regression-Linear models-Logistic regression-Rates and Poisson regression-Nonlinear curve fitting –correlation.

Unit-3. Distribution

Descriptive Statistics & Sampling Distributions: Population-Sampling-Measures of Central tendency, variations and position –Sampling distributions: Standard Normal Distribution-Chi-Square Distribution-t Distribution –F-Distribution -The Central Limit Theorem.

Unit-4. Estimation

Estimation: Point Estimation: the method of moments and the method of maximum likelihood estimation Interval estimation: estimation of mean, estimation of difference of means, estimation of variance and estimation of ratio of variances.

Unit-5. Hypothesis Testing

Test of Hypothesis-Testing for Attributes –Mean of Normal Population –One-tailed and two-tailed tests, Ftest and Chi-Square test –Analysis of Variance-Nonparametric test.

| S.No. | Topics | Web Links |
|-------|---------------------------------------|---|
| 1 | Statistical Thinking for Data Science | https://www.edx.org/course/statistical-thinking- |
| | and Analytics | for-data-science-and-analytic |
| 2 | Linear Regression for Business | https://www.coursera.org/learn/linear-regression- |
| | Statistics | business-statistics |
| 3 | Learning Statistics with R | https://learningstatisticswithr.com/ |

B. TOPICS FOR SELF-STUDY

| 4 | 15 Types of Regression | https://www.listendata.com/2018/03/regression- |
|---|------------------------|--|
| | | analysis.html |

C. TEXT BOOKS

1. Cheryl Ann Willard, "Statistical Methods: An Introduction to Basic Statistical Concepts and Analysis", Routledge, 2020. (Unit – I – III)

2. Xin Yan & Xiaogang Su, "*Linear Regression Analysis : Theory and Computing*", World Scientific Pulishing Ltd, 2009. (Unit – IV: Chapter 1,2,3, 8.5,8.6; Unit – V: Chapters 4.2,4.3, 4.4, 4.5, 5.1-5.3, 6.1, 6.2)

D. REFERENCE BOOKS

1. John.E.Freund, Irwin Miller, Marylees Miller "Mathematical Statistics with Applications", 8th, Prentice Hall of India, 2014

2. Ross, Sheldon. M, "Introduction to Probability and Statistics for Engineers and Scientists", Academic Press, 2009

3. D.C Montgomery, E.A Peck and G.G Vining, "Introduction to Linear Regression Analysis", John Wiley and Sons, 2003.

4. S. Chatterjee and AHadi, "*Regression Analysis by Example*", 4th Ed., John Wiley and Sons, Inc, 2006

E. WEB LINKS

- <u>https://www.listendata.com/2018/03/regression-analysis.html</u>
- <u>https://www.coursera.org/learn/linear-regression-business-statistics</u>

3. SPECIFIC LEARNING OUTCOMES(SLO)

| Unit/ Section | Торіс | Learning outcomes | Level | | |
|------------------|-------------------------------|--------------------------------------|----------|--|--|
| Ι | I | Probability | obbility | | |
| 1.1 | Probability Spaces | Discover mathematical construct | K4 | | |
| | | that provides a formal model of a | | | |
| | | random process | | | |
| 1.2 | Combinatorial methods | Decide dramatically increase the | K5 | | |
| | | rate at which new compounds are | | | |
| | | discovered and improved. | | | |
| 1.3 | Counting techniques | Estimate the probability of an | K6 | | |
| | | event | | | |
| 1.4 | Elementary Theorem | Elaborate the likelihood that a | K6 | | |
| | | defined event will occur | | | |
| 1.5 | Conditional Probability | Measure likelihood of an event or | K5 | | |
| | | outcome occurring | | | |
| 1.6 | Bayes theorem | Examine the probability Total | K4 | | |
| | | Probability Rule | | | |
| 1.7 | Probability Distributions and | Assess representation of a | K5 | | |
| | Probability Densities. | continuous probability distribution | | | |
| II | Expectat | tion and Regression | | | |
| 2.1 | Mathematical Expectation: | Examine of all possible values | K4 | | |
| | Expected value | from a random variable | | | |
| 2.2 | Moments | Evaluate the random variable X | K5 | | |
| | | denoted by $M_X(t)$ | | | |
| 2.3 | Chebyshev's theorem | Formulate the minimum | K6 | | |
| | | proportion of observations that fall | | | |
| | | within a specified number of | | | |
| | | standard deviations from the | | | |
| | | mean. | | | |

| 2.4 | Moment Generating functions | Assess the real-valued random variable using moment- generating function | K5 |
|---|---|--|--|
| 2.5 | Product Moment | Determine the bivariate correlation | K5 |
| 2.6 | Conditional Expectation | Choose an arbitrarily large number of occurrences | K5 |
| 2.7 | Special Probability Distributions | Inspect the picking of an element of S at random | K4 |
| 2.8 | Probability Densities | Evaluate, whose value at any given sample space can be interpreted. | K5 |
| 2.9 | Functions of Random Variable | Analyse, How to make precise the idea that a function of a random variable is itself a random variable | K4 |
| 2.10 | Multiple regression | Examine relationship between a single dependent variable and several independent variables | K4 |
| 2.11 | Linear models | Measure the response variable | K5 |
| 2.12 | Logistic regression | Evaluate binary dependent variable | K5 |
| 2.13 | Rates and Poisson regression | Predict a dependent variable that consists of "count data" given one or more independent variables | K6 |
| 2.14 | Nonlinear curve fitting | Prove fitting to curves whose parameters appear in the function expression in arbitrary ways | K5 |
| 2.15 | Correlation. | Justify the degree to which two variables move in coordination | K5 |
| | | with one another | |
| III | D | with one another Distribution | |
| III 3.1 | Descriptive Statistics & Sampling | with one another Distribution Examine most common patterns | K4 |
| III 3.1 | Descriptive Statistics & Sampling Distributions | with one another Distribution Examine most common patterns of the analyzed data set. | K4 |
| III 3.1 3.2 | Descriptive Statistics & Sampling Distributions Population-Sampling | with one another Distribution Examine most common patterns of the analyzed data set. Survey the process of taking a subset of subjects that is representative of the entire population | K4 K4 |
| III 3.1 3.2 3.3 | Descriptive Statistics & Sampling Distributions Population-Sampling Measures of Central tendency | with one another Distribution Examine most common patterns of the analyzed data set. Survey the process of taking a subset of subjects that is representative of the entire population Determine the typical value for a probability distribution | K4 K4 K5 |
| III 3.1 3.2 3.3 3.4 | Descriptive Statistics & Sampling Distributions Population-Sampling Measures of Central tendency variations and position | with one another Distribution Examine most common patterns of the analyzed data set. Survey the process of taking a subset of subjects that is representative of the entire population Determine the typical value for a probability distribution Compare each deviation from its expected value | K4 K4 K5 K4 |
| III 3.1 3.2 3.3 3.4 3.5 | Descriptive Statistics & Sampling Distributions Population-Sampling Measures of Central tendency variations and position Sampling distributions | with one another Distribution Examine most common patterns of the analyzed data set. Survey the process of taking a subset of subjects that is representative of the entire population Determine the typical value for a probability distribution Compare each deviation from its expected value Identify a probability distribution of a statistic obtained from a larger number of samples drawn from a specific population | K4 K4 K5 K4 K3 |
| III 3.1 3.2 3.3 3.4 3.5 3.6 | Descriptive Statistics & Sampling Distributions Population-Sampling Measures of Central tendency variations and position Sampling distributions Standard Normal Distribution | with one another Distribution Examine most common patterns of the analyzed data set. Survey the process of taking a subset of subjects that is representative of the entire population Determine the typical value for a probability distribution Compare each deviation from its expected value Identify a probability distribution of a statistic obtained from a larger number of samples drawn from a specific population Evaluate a mean of zero and standard deviation of 1 | K4 K4 K5 K4 K3 K5 |
| III 3.1 3.2 3.3 3.4 3.5 3.6 3.7 | Descriptive Statistics & Sampling Distributions Population-Sampling Measures of Central tendency variations and position Sampling distributions Standard Normal Distribution Chi-Square Distribution | with one another Distribution Examine most common patterns of the analyzed data set. Survey the process of taking a subset of subjects that is representative of the entire population Determine the typical value for a probability distribution Compare each deviation from its expected value Identify a probability distribution of a statistic obtained from a larger number of samples drawn from a specific population Evaluate a mean of zero and standard deviation of a sum of the distribution of a sum of the distribution of a sum of the squares of n independent | K4 K4 K5 K4 K3 K5 K4 |
| III 3.1 3.2 3.3 3.4 3.5 3.6 3.7 3.8 | Descriptive Statistics & Sampling Distributions Population-Sampling Measures of Central tendency variations and position Sampling distributions Standard Normal Distribution Chi-Square Distribution t-Distribution | with one another Distribution Examine most common patterns of the analyzed data set. Survey the process of taking a subset of subjects that is representative of the entire population Determine the typical value for a probability distribution Compare each deviation from its expected value Identify a probability distribution of a statistic obtained from a larger number of samples drawn from a specific population Evaluate a mean of zero and standard deviation of 1 Inspect k degrees of freedom is the distribution of a sum of the squares of n independent Decide the standardized distances of sample means to the population mean | K4 K4 K5 K4 K3 K5 K4 K5 |

| | | test statistic in a continuous | |
|-----------|---------------------------------------|--------------------------------------|------|
| | | probability distribution that arises | |
| | | frequently | |
| 3 10 | The Central Limit Theorem | Criticize the independent random | K5 |
| 5.10 | The Central Emilt Theorem. | variables are summed up | KJ |
| IX/ | 1 | Fatimation | |
| 1V 4 1 | Estimation Daint Estimation | Construction the sector of second | 175 |
| 4.1 | Estimation: Point Estimation | Conclude the value of some | КЭ |
| | | property of a population from | |
| 1.0 | | observations | |
| 4.2 | the method of moments | estimate the population | K5 |
| | | parameters. | |
| 4.3 | the method of maximum | Compose the conditional | K6 |
| | likelihood estimation | probability of observing the data | |
| | | sample | |
| 4.4 | Interval estimation: estimation of | Make | K5 |
| | mean | up an interval, or range of values, | |
| | | within which the parameter is | |
| | | most likely to be located. | |
| 4.5 | estimation of difference of means | Originate the difference in the | K6 |
| | | corresponding sample means | |
| 4.6 | estimation of variance | Originate the point estimate of the | K6 |
| | | variance of an unknown | |
| | | distribution | |
| 4.7 | estimation of ratio of variances | Invent the ratio of means of two | K6 |
| | | random variables | |
| V | Нур | othesis Testing | |
| 5.1 | Test of Hypothesis | Assess the plausibility of a | K5 |
| | | hypothesis by using sample data | |
| 5.2 | Testing for Attributes | Analyze the characteristics of a | K4 |
| | | given population | |
| 5.3 | Mean of Normal Population | Elaborate approximately normal | K6 |
| | I I I I I I I I I I I I I I I I I I I | as long as the expected number of | - |
| | | successes | |
| 5.4 | One-tailed and two-tailed tests | Justify whether a claim is true or | K5 |
| | | not, given a population parameter | |
| 5.5 | F-test | Survey the F-distribution under | K4 |
| 2.0 | | the null hypothesis | |
| 56 | Chi-Square test | Predict to perform when the test | K6 |
| 5.0 | | statistic is chi-squared distributed | |
| | | under the null hypothesis | |
| 57 | Analysis of Variance | Analyse variation in a response | КЛ |
| 5.7 | | variable | 11.7 |
| 50 | Nonperemetrie test | Imaging the required accumptions | V6 |
| 5.8 | nonparametric test. | to be applying d | NU |
| | | to be analyzed | |

| L- | Low | | | M-N | Iodera | te | | H- | High | | | | |
|------------|------------|-----|-----|-----|--------|------------|------------|------------|------------|------|------|------|------|
| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
| CO1 | Н | Μ | | L | | Н | | Н | | Н | | L | Μ |
| CO2 | Н | | Н | Μ | Μ | | L | | L | Μ | | Н | |
| CO3 | Μ | Н | Н | Н | Η | | Н | | | Н | L | | |
| CO4 | Η | L | Μ | Μ | | L | | | | Μ | Η | | L |

| CO5 | Η | Н | | L | Μ | Μ | | Μ | Н | | Μ | |
|-----|---|---|---|---|---|---|---|---|---|---|---|---|
| CO6 | Μ | | L | Η | Н | | L | Μ | Н | Μ | | Η |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 2. Open Book Test.
- 3. Cooperative Learning Report, Assignment, Group Presentation, Group Discussion, project Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).
- 4. Pre-Semester & End Semester Theory Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator: Dr. A. Santhanasamy

| CORE PRACTICAL I: PROBLEM SOLVING USING PYTHON AND R LAB | | | | | | | |
|---|----------|------------|---|--|--|--|--|
| Semester | Ι | Hours/Week | 5 | | | | |
| Course Code | P19DS1P1 | Credits | 3 | | | | |

| S.No. | Course Outcomes | Level | Activity |
|-------|---|-------|----------|
| CO1 | Write simple Python programs using Python data structures | K6 | 1-5 |
| CO2 | Manipulate files using Python | K6 | 6 |
| CO3 | Develop object oriented programs in Python | K6 | 7-8 |
| CO4 | Access internet and database data | K6 | 9-12 |
| CO5 | Write R programs for data visualization | K6 | 13 |
| CO6 | Creating dashboards using Tableau | K6 | 14-15 |

2. SYLLABUS

Develop applications that will demonstrate the following Python and R programming features

- Functions
- String processing
- List processing
- Dictionaries
- Tuples
- File processing
- Regular Expressions
- OOP
- Retrieving webpages from web
- Data visualization in Matplotlib, Seaborn and R
- Database programming
- Concurrent programming

3. SPECIFIC LEARNING OUTCOMES (SLO)

| Exercise | Торіс | Learning outcomes | Level |
|----------|--------------------------------|--|-------|
| 1 | Python Basics and Conditions | Applying conditional statements to programs | K6 |
| 2 | Python Loops | Applying loops to Python Programs | K6 |
| 3 | Python Functions and Modules | Creating user defined functions and using modules | K6 |
| 4 | Python String Processing | Manipulating strings in Python | K6 |
| 5 | List Processing in Python | Using lists in programs | K6 |
| 6 | Python File Processing | Reading and writing files | K6 |
| 7 | Python Regular Expressions | Applying pattern matching to strings | K6 |
| 8 | Object Oriented Bank in Python | Creating Python classes and objects | K6 |
| 9 | Functional Programming | Using map, filter and reduce | K6 |

| | | functions in programs | |
|----|-------------------------------|-------------------------------------|----|
| 10 | Retrieving Data from Web and | Retrieving data from webpages | K6 |
| | Parsing | using urllib | |
| 11 | Database Programming Using | Extracting data from SQL | K6 |
| | Sqlite3 | databases using python libraries | |
| 12 | 2D and 3D Data Visualization | Creating data visualizations using | K6 |
| | Using Seaborn | seaborn library | |
| 13 | Animated Data Visualization | Creating interactive visualizations | K6 |
| | Using R | using R libraries | |
| 14 | Dashboard Visualization Using | Creating dashboards and repots | V6 |
| | Tableau | using Tableau | KU |
| 15 | Concurrent Programming in | Creating concurrent programs for | V6 |
| | Python | multiprocessing | K0 |

L-Low

M-Moderate

H- High

| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|------|------|------|------|
| CO1 | | Н | | | Н | Н | Н | Η | | | | Н | |
| CO2 | | Н | | | Н | Н | Н | Η | | Н | Н | Н | Н |
| CO3 | | Н | | | Н | Н | Н | Н | | Н | | Н | Н |
| CO4 | | Н | | | Η | Н | Η | Η | | Н | Н | Н | |
| CO5 | | Н | | | Η | Η | Н | Η | | | | Н | Н |
| CO6 | | Н | | | Н | Н | Н | Η | | | | | |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 2. Open Book Test.
- 3. Cooperative Learning Report, Assignment, Group Presentation, Group Discussion, project Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).
- 4. Pre-Semester & End Semester Theory Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator: Prof. K. Jemimah

| Semester | Ι | Hours/Week | 5 |
|-------------|----------|------------|---|
| Course Code | P19DS1P2 | Credits | 3 |

After the successful completion of this course the students will be able to

| S.No. | Course Outcomes | Leve l | Exercise Covered |
|-------|---|-----------|---------------------|
| CO1 | Build a Table using SQL Queries and perform the basic operation | K6 | 1,2 |
| CO2 | Construct a SQL queries to evaluate various operators | K6 | 3,4 |
| CO3 | Evaluate the result using Subquery and Join techniques | K6 | 5,6,7,8 |
| CO4 | Assess the basic Queries in NoSQL using Mango DB | K5 | 9,10,11 |
| CO5 | Review the SQL, NoSQL and Neo 4J Graph data base | K5 | 12 |
| CO6 | Design a Graph database for Movie and Flight Data | K6 | 13,14 |

2. SYLLABUS

| S.No | List of Exercises |
|------|---|
| 1 | Designing and Querying My Restaurant Database |
| 2 | India Weather Analytics Using Historical Data Part-I |
| 3 | India Weather Analytics Using Historical Data Part-II |
| 4 | Retail Sales Analytics Part-I |
| 5 | Retail Sales Analytics Part-II |
| 6 | Retail Sales Analytics Part-III |
| 7 | University Course Enrolment Data Analytics |
| 8 | Retail Sales Analytics Part-IV |
| 9 | Student Information System Design using MongoDB Part-I |
| 10 | Student Information System Design using MongoDB Part-II |
| 11 | Ecommerce Product CatLog Design Using MongoDB |
| 12 | Neo4J Play Ground Exercise |
| 13 | Designing Movie Graph Database using Neo4J |
| 14 | Designing Flight Graph Database Using Neo4J |

3. SPECIFIC LEARNING OUTCOMES (SLO)

| Exercise | Course Content | Learning outcomes | Level |
|----------|--|--|-------|
| 1 | Designing and Querying My Restaurant Database | Create a new table, insert tuples satisfying the constraints and perform queryprocessing. | K6 |
| 2 | India Weather Analytics Using Historical Data Part-I | Evaluate the selection, filtering and aggregate functions to analyse the historical data of India Weather Information | K5 |
| 3 | India Weather Analytics Using Historical Data Part-II | Determine further India Weather Dataset with additional query operators such as GROUPBY, HAVING and | K5 |

| | | ORDERBY | |
|----|---|--|----|
| 4 | Retail Sales Analytics Part-I | Create analytics on retail sales of a particular enterprise | K6 |
| 5 | Retail Sales Analytics Part-II | Assess multiple tables in retail sales dataset | K5 |
| 6 | Retail Sales Analytics Part-III | Compile Group function and Aggregate function in retail dataset | K6 |
| 7 | University Course Enrolment Data Analytics | Design document in NoSQL for university course enrollment | K6 |
| 8 | Retail Sales Analytics Part-IV | Develop queries for subquery | K6 |
| 9 | Student Information System Design using MongoDB Part-I | Create student information system design using mongo DB | |
| 10 | Student Information System Design using MongoDB Part-II | Analyze various operators in mango DB | K4 |
| 11 | Ecommerce Product CatLog Design Using MongoDB | Create a Collection in mongo DB for product catalog for heterogenous data | K6 |
| 12 | Neo4J Play Ground Exercise | Outline the Neo4J for movie graph dataset | K2 |
| 13 | Designing Movie Graph Database using Neo4J | Create a graph database for Actors and Movies in which actors played roles. You will write queries in Cypher and find answers to all queries. | K6 |
| 14 | Designing Flight Graph Database Using Neo4J | Develop graph data base for Flight Transportation system | K6 |

L-Low

M-Moderate

H- High

| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
|------------|-----|-----|-----|-----|-----|------------|------------|-----|------------|------|------|------|------|
| CO1 | Н | Μ | L | Μ | Μ | Μ | L | - | - | Н | Μ | Μ | - |
| CO2 | Н | Μ | Н | Μ | Μ | Н | Μ | Μ | L | Н | Н | - | Μ |
| CO3 | H | Μ | Μ | L | Н | Μ | - | - | L | Μ | Μ | Μ | - |
| CO4 | Н | L | Н | Н | Н | Н | Μ | Μ | L | Н | Μ | Μ | Н |
| CO5 | H | Μ | L | Н | Н | L | L | Μ | Μ | Н | Н | Н | Μ |
| CO6 | H | Μ | Μ | L | Η | L | L | L | Μ | Н | Н | - | - |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Practical Components): Closed Book
- 2. Cooperative Learning Report, Assignment, Group Discussion, project Report, Field Visit Report, Seminar.
- 3. Pre/Post Test, Viva, Report for each Exercise.
- 4. Lab Model Examination & End Semester Practical Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator: Dr. M. Lovelin Pon Felciah

| CORE IV: REGRESSION ANALYSIS | | | | | |
|------------------------------|----------|------------|---|--|--|
| Semester | II | Hours/Week | 4 | | |
| Course Code | P19DS204 | Credits | 4 | | |

On successful completion of this course, students will be able to:

| S.No. | Course Outcomes | Level | Unit |
|-------|---|-------|------|
| CO1 | Solve the stationarity, trending and detrending of time series data | K6 | 1 |
| CO2 | Assess the features of the ARMA Models and estimation techniques | K5 | 2 |
| CO3 | Explain the ARIMA models and SARMA Models | K6 | 3 |
| CO4 | Summarize the characteristics of Spectral behaviour and periodic behaviour of the time series | | 4 |
| CO5 | Compile the behaviour of smoothing in DLMS | K6 | 5 |
| CO6 | Design the Timeseries models using R for different time series data | K6 | All |

2. A. SYLLABUS

Unit-1. Simple Linear Regression

Introduction to regression analysis: Modelling a response, overview and applications of regression analysis, major steps in regression analysis. Simple linear regression (Two variables): assumptions, estimation and properties of regression coefficients, significance and confidence intervals of regression coefficients, measuring the quality of the fit.

Unit-2. Multiple Linear Regression

Multiple linear regression model: assumptions, ordinary least square estimation of regression coefficients, interpretation and properties of regression coefficient, significance and confidence intervals of regression coefficients.

Unit-3. Criteria for Model Selection

Mean Square error criteria, R2 and $\hat{R}2$ criteria for model selection; Need of the transformation of variables; Box-Cox transformation; Forward, Backward and Stepwise procedures.

Unit-4. Residual Analysis

Residual analysis, Departures from underlying assumptions, Effect of outliers, Collinearity, Nonconstant variance and serial correlation, Departures from normality, Diagnostics and remedies.

Unit-5. Non Linear Regression

Introduction to nonlinear regression, Least squares in the nonlinear case and estimation of parameters, Models for binary response variables, estimation and diagnosis methods for logistic and Poisson regressions. Prediction and residual analysis.

| S.No. | Торі | cs | Web Links |
|----------|----------------|-------------|---|
| 1 | Multivariate | Regression | https://otexts.com/fpp2/hierarchical.html |
| | Analysis | | |
| 2 | Autoregression | Models for | https://machinelearningmastery.com/\autoregressio |
| | Time Series | Forecasting | n-models-time-series-forecasting-python/ |
| | With Python | | |
| <u>3</u> | Simple | Exponential | https://towardsdatascience.com/ |
| | Smoothing for | Time Series | simple-exponential-smoothing-749fc5631bed |
| | Forecasting | | |

B. TOPICS FOR SELF - STUDY

C. TEXT BOOKS

- 1. D.C Montgomery, E.A Peck and G.G Vining, Introduction to Linear Regression Analysis, John Wiley and Sons, Inc. NY, 2003.
- 2. S. Chatterjee and AHadi, Regression Analysis by Example, 4th Ed., John Wiley and Sons, Inc, 2006
- 3. Seber, A.F. and Lee, A.J. (2003) Linear Regression Analysis, John Wiley, Relevant_sections from chapters 3, 4, 5, 6, 7, 9, 10.

D. REFERENCES BOOKS

- 1. Iain Pardoe, Applied Regression Modeling, John Wiley and Sons, Inc, 2012.
- 2. P. McCullagh, J.A. Nelder, Generalized Linear Models, Chapman & Hall, 1989

E. WEB LINKS

- 1. https://machinelearningmastery.com/autoregression-models-time-series-forecasting-python/
- 2. https://courses.cornell.edu/preview_course_nopop.php?catoid=31&coid=491740
- 3. https://engineering.purdue.edu/online/courses/applied-regression-analysis

| Unit/ Section | Course Content | Learning outcomes | Level |
|------------------|--|---|----------|
| Ι | SIM | PLE LINEAR REGRESSION | |
| 1.1 | Introduction to regression analysis: Modelling a response, overview and | Define the Regression Explain the steps involved in Regression | K2 |
| 1.1 | applications of regression analysis, major steps in regression analysis | Determine the major steps in regression analysis | K4 K5 |
| 1.2 | Simple linear regression (Two variables): assumptions, estimation and properties of | Construct the Simple Linear Regression Assess the assumptions of regression | K6 |
| | regression coefficients | | IX0 |
| 1.3 | Significance and confidence intervals of regression coefficients, measuring the quality of the fit. | Determine the Significance and confidence intervals of regression coefficients | К5 |
| II | MULT | TIPLE LINEAR REGRESSION | • |
| 2.1 | Multiple linear regression model: assumptions, ordinary | Construct the Simple Linear Regression Assess the assumptions of regression | K6 |
| | regression coefficients | coefficients | K6 |
| 2.2. | Interpretation and properties of regression coefficient | Explain the interpretation and properties of regression coefficient | K5 |
| 2.3 | Significance and confidence intervals of regression coefficients | Determine the Significance and confidence intervals of regression coefficients | K5 |
| III | CRITE | RIA FOR MODEL SELECTION | 1 |
| 3.1 | Mean Square error criteria, R2 and R2 criteria for model selection | Explain the Criteria for model selection Design the criteria for model selection | K5 K6 |
| 3.2 | Need of the transformation of variables | Assess the need of the transformation of variables | K6 |
| 3.3 | Box-Cox transformation; | Construct the Box-Cox Transformations | K6 |

3. SPECIFIC LEARNING OUTCOMES

| | Forward, Backward and | Explain the procedures for model selections | K5 | | | |
|-----|---|--|----------------|--|--|--|
| | Stepwise procedures. | | | | | |
| IV | | RESIDUAL ANALYSIS | | | | |
| 4.1 | Residual analysis, Departures from underlying assumptions | Explain the Residual Analysis | K5 | | | |
| 4.2 | Effect of outliers, Collinearity | Assess the outliers for of the models | K6 | | | |
| 4.3 | Non-constant variance and serial correlation | Determine the Non-constant variance and serial correlation | K5 | | | |
| 4.4 | Departures from normality | Evaluate the normality | K5 | | | |
| 4.5 | Diagnostics and remedies | Explain the methods for Diagnostics and remedies | | | | |
| V | NON LINEAR REGRESSION | | | | | |
| • | 110 | IN LINEAK KEGKESSION | | | | |
| 5 1 | Introduction to nonlinear regression, Least squares in | Assess importance of Non-linear regression methods | K6 | | | |
| 5.1 | Introduction to nonlinear regression, Least squares in the nonlinear case and estimation of parameters, | Assess importance of Non-linear regression methods Estimate the parameters for nonlinear regression | K6 K5 | | | |
| 5.1 | Introduction to nonlinear regression, Least squares in the nonlinear case and estimation of parameters, Models for binary response variables, estimation and diagnosis methods for logistic and Poisson regressions. | Assess importance of Non-linear regression methods Estimate the parameters for nonlinear regression Construct the model for binary response variables Design the logistic and Poisson regressions. | K6 K5 K6 | | | |

4. MAPPING

L-Low

M-Moderate

H- High

| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
|------------|-----|-----|-----|-----|-----|-----|------------|------------|------------|------|------|------|------|
| CO1 | Η | Μ | Η | Η | Η | Η | - | - | Μ | Н | Н | Μ | - |
| CO2 | Μ | Η | Μ | Η | Μ | Μ | - | Μ | Μ | Μ | Μ | - | - |
| CO3 | Η | Μ | Η | Η | Η | Η | - | - | Η | Н | Μ | - | - |
| CO4 | Η | Η | L | Η | Μ | Η | - | Μ | Η | Н | Н | - | |
| CO5 | Η | Μ | Η | Μ | Η | Η | - | - | Μ | Н | Η | - | - |
| CO6 | Η | Η | Η | Μ | Μ | Η | Η | - | Μ | Η | Η | - | - |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 2. Open Book Test.
- 3. Peer Learning Report, Assignment, Group Presentation, Group Discussion, project Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).
- 4. Pre-Semester & End Semester Theory Examination

INDIRECT:

- 1. Course evaluation survey
- 2. Faculty feedback about the course.

Name of the Course Coordinator: Dr. P. S. Eliahim Jeevaraj

| CORE V: DATA AND VISUAL ANALYTICS | | | | | | |
|-----------------------------------|----------|------------|---|--|--|--|
| Semester | II | Hours/Week | 4 | | | |
| Course Code | P19DS205 | Credits | 4 | | | |

On successful completion of this course, students will be able to:

| S.No. | Course Outcomes | Level | Unit |
|-------|--|-------|------|
| CO1 | Test the NumPy functions for array processing | K6 | Ι |
| CO2 | Create time series plots using the Date and Time classes | K6 | II |
| CO3 | Interpret the Plotting of the dataset and time series | K6 | III |
| CO4 | Perform data aggregation and group operations | K6 | IV |
| CO5 | Create and use Series and Data Frames for data wrangling | K6 | V |
| CO6 | Create various plots using Matplotlib and Seaborn | K6 | v |

2. A. SYLLABUS

Unit-1. NumPy and Pandas Basics

Why Python for Data analysis – Essential Python libraries – ndarray – Universal functions – Data processing using arrays – File I/O with arrays – Random number generation – Series, Data Frames – Indexing, re-indexing, sorting, ranking – Summarizing descriptive statistics – Handling missing data – Hierarchical indexing

Unit-2. Data Loading and Wrangling

Data Loading: reading and storing data in text format, binary format – Data Wrangling: Combining and merging data sets – Reshaping – Pivoting – Data transformation – String manipulation

Unit-3. Plotting and Visualization using Matplotlib

Figures – Subplots – Colors – Ticks – Label – Legends – Annotation – Saving plots to file – Plots: Line, Bar, Histogram, Density Plots – Scatter Plots

Unit-4. Data Aggregation and Group Operations

Iterating over groups – Selecting columns – Grouping with Series and functions – Data aggregation: Column wise aggregation, returning aggregated data – General-Split-Apply-Combine – Quantile and bucket analysis – Pivot table and cross tabulation

Unit-5. Time Series

Date and Time – Time Series – Date Range, Frequencies and Shifting – Periods and period arithmetic – Resampling and frequency coversion – Time Series Plotting

| S.No | Topic Title | Web Link |
|------|---|---|
| 1 | Facebook Data Analysis | https://www.kaggle.com/sd2beatles/deep-analysis- sql-and-statistical-test-included |
| 2 | Clothing Fit Dataset for Size Recommendation | https://www.kaggle.com/agrawaladitya/step-by-step- data-preprocessing-eda |
| 3 | UCI Adult dataset | https://www.kaggle.com/kashnitsky/a1-demo- pandas-and-uci-adult-dataset |
| 4 | Wikipedia Time series analysis | https://www.kaggle.com/kashnitsky/a9-demo-time- series-analysis |

B. TOPICS FOR SELF-STUDY

C. TEXT BOOKS

1. Wes. Mc Kinney, "Python for Data Analysis", 2nd Edition, Schroff Publishers, 2013. ISBN 9789352136414

D. REFERENCES

1. Cyrille Rossant. "Learning IPython for interactive Computing and data visualization", First edition [Packt]

2. Jake VanderPlas ,Python Data Science Handbook - Essential Tools for Working with Data, O'Reily, 2016

3. Zhang.Y ,An Introduction to Python and Computer Programming, Springer Publications,2016

E. WEB LINKS

- https://www.kaggle.com/agrawaladitya/step-by-step-data-preprocessing-eda
- <u>https://www.kaggle.com/kashnitsky/a1-demo-pandas-and-uci-adult-dataset</u>

3. SPECIFIC LEARNING OUTCOMES

| Unit | Торіс | Topic Learning Outcome | | | |
|------|------------------------------|---|-------|--|--|
| Ι | | NumPy and Pandas Basics | | | |
| 1.1 | Python for Data analysis | Why Dataset analysis? And Why and What Python | K1 | | |
| 1.2 | Essential Python libraries | Illustrate Python libraries which is used for Data Science | K2 | | |
| 1.3 | Universal | Utilize universal function (or ufunc for short) is | K3 | | |
| | functions | a function that operates on ndarrays in an element-by- element fashion, supporting array broadcasting, type | | | |
| | | casting, and several other standard features | | | |
| 1.4 | Data processing using arrays | Discover data processing tasks without writing complex loops | K4 | | |
| 1.5 | File I/O with arrays | Test for file read and write using array | K4 | | |
| 1.6 | Random number generation | Analyze pseudo-random number generator for various distributions. | K4 | | |
| 1.7 | Series, Data | Appraise the single list with index. Examine a dataframe | K5,K4 | | |
| | Frames | using collection of series that can be used to analyse the data | | | |
| 1.8 | Indexing, re- | Find NA/NaN in locations having no value in the | K1, | | |
| | indexing, sorting, | previous index. | K3 | | |
| | ranking | Plan to index and reindex using Indexing, reindexing, sorting, ranking. | | | |
| 1.9 | Summarizing | Examine summarizing and organizing the data so they | K4 | | |
| | descriptive | can be easily understood. | | | |
| | statistics | | | | |
| 1.10 | Handling missing | Determine missing values for a number of reasons such | K5 | | |
| | data | as observations that were not recorded and data corruption. | | | |
| 1.11 | Hierarchical | Discover to incorporate multiple index levels within a | K4 | | |
| | indexing | single index. | | | |

| II | Data Loading and Wrangling | | | | | | |
|------|----------------------------|---|----|--|--|--|--|
| 2.1 | Data Loading: | Motive the ability to read, manipulate, and write data to | K4 | | | | |
| | reading and | and from CSV files using Python is a key skill to master | | | | | |
| | storing data in | for any data scientist or business analysis. | | | | | |
| | text format, | | | | | | |
| | binary format | | | | | | |
| 2.2 | Data Wrangling: | Measure the processing of data in various formats like - | K5 | | | | |
| | Combining and | merging, grouping, concatenating etc. for the purpose of | | | | | |
| | merging data sets | analysing or getting them ready to be used with another | | | | | |
| | | set of data. | | | | | |
| 2.3 | Reshaping | Examine 'reshape()' function, that takes a single | K4 | | | | |
| | | argument that specifies the new shape of the array. | | | | | |
| 2.4 | Pivoting | Analyze Pivoting for reshape a DataFrame by | K4 | | | | |
| | | column/index values. | | | | | |
| 2.5 | Data | how we can combine data from different sources into a | K1 | | | | |
| | transformation | unified dataframe | | | | | |
| 2.6 | String | List the manipulation of string like concatenation, | K4 | | | | |
| | manipulation | <pre>isupper(), join(), lower(), etc.</pre> | | | | | |
| III | | Plotting and Visualization using Matplotlib | | | | | |
| 3.1 | Figures | Create graph using figure() in Python. | K6 | | | | |
| 3.2 | Subplots | Create subplots by the use of subplot() function in | K6 | | | | |
| | | pyplot module. | | | | | |
| 3.3 | Colors | Discuss colouring plot by python colour code. | K6 | | | | |
| 3.4 | Ticks | Create Ticks value to show specific points on the | K6 | | | | |
| | | coordinate axis. | | | | | |
| 3.5 | Label | Assess plot axis label | K5 | | | | |
| 3.6 | Legends | Analyze legend for describing area and elements of the | K4 | | | | |
| | | graph | | | | | |
| 3.7 | Annotation | Utilize annotate() function to draw an arrow connecting | K3 | | | | |
| | | two points on the plot. | | | | | |
| 3.8 | Saving plots to | Recommend savefig() function to save plot in to file | K5 | | | | |
| | file | | | | | | |
| 3.9 | Plots: Line, Bar, | List basic graphics primitives to draw plot | K4 | | | | |
| | Histogram, | | | | | | |
| | Density Plots | | | | | | |
| 3.10 | Scatter Plots | Evaluate the data as a collection of points. | K5 | | | | |
| IV | | Data Aggregation and Group Operations | | | | | |
| 4.1 | Iterating over | Discover data cluster using Iterating over groups | K4 | | | | |
| | groups | | | | | | |
| 4.2 | Selecting | Select multiple columns using loc, iloc, etc | K3 | | | | |
| | columns | | | | | | |
| 4.3 | Grouping with | Create series group using groupby() function | K6 | | | | |
| | Series and | | | | | | |
| | functions | | | | | | |
| 4.4 | Data aggregation: | Examine summarization using computing aggregations | K4 | | | | |
| | Column wise | like sum(), mean(), median(), min(), and max(), in | | | | | |

| | aggregation | which a single number gives insight into the nature of a | |
|-----|-------------------|---|----|
| | | potentially large dataset. | |
| 4.5 | Data aggregation: | Plan statistical method for data aggregation. | K3 |
| | returning | | |
| | aggregated data | | |
| 4.6 | General-Split- | Create group by three step Split-Apply-Combine. | K6 |
| | Apply-Combine | | |
| 4.7 | Quantile and | Examine quantile() function to get values at the given | K4 |
| | bucket analysis | quantile over requested axis. | |
| 4.8 | Pivot table and | Create cross-tabulation table for show the frequency | K6 |
| | cross tabulation | with which certain groups of data appear | |
| V | | Time Series | |
| 5.1 | Date and Time | Examine date and time parameter for time series. | K4 |
| 5.2 | Time Series | Discover statistical descriptive by statistical tests and | K4 |
| | | several linear model classes: autoregressive, AR, | |
| | | autoregressive moving- average, ARMA, and vector | |
| | | autoregressive models VAR. | |
| 5.3 | Date Range | Model a large range of dates for various offsets are pre- | K3 |
| | | computed | |
| 5.4 | Frequencies and | Evaluate percentage change from sample to sample. | K5 |
| | Shifting | | |
| 5.5 | Periods and | Dissect the time elapsed between two values of the same | K4 |
| | period arithmetic | magnitude. | |
| 5.6 | Resampling and | Survey the Convenience method for frequency | K4 |
| | frequency | conversion and resampling of time series | |
| | conversion | | |
| 5.7 | Time Series | Visualize trends in counts or numerical values | K4 |
| | Plotting | over time. | |

4. MAPPING

L-Low

M-Moderate

H- High

| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
|------------|------------|-----|-----|-----|-----|-----|------------|------------|------------|------|------|------|------|
| CO1 | Η | Μ | Η | Η | Η | Η | - | - | Μ | Н | Н | Н | - |
| CO2 | Μ | Η | Μ | Η | Μ | Μ | I | Μ | Μ | Μ | Μ | • | Η |
| CO3 | Η | Μ | Η | Η | Η | Η | - | - | Η | Н | Μ | Н | Н |
| CO4 | Η | Η | L | Η | Μ | Η | I | Μ | Η | Η | Η | I | Η |
| CO5 | Η | Μ | Η | Μ | Η | Η | I | - | Μ | Η | Η | Η | Η |
| CO6 | Η | Η | Η | Μ | Μ | Η | Η | - | Μ | Η | Η | Н | - |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 2. Open Book Test.
- 3. Peer Learning Report, Assignment, Group Presentation, Group Discussion, project Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).

4. Pre-Semester & End Semester Theory Examination

INDIRECT:

- Course evaluation survey
 Faculty feedback about the course.

Name of the Course Coordinator: Dr. B. Karthikeyan

| CORE VI: PRACTICAL MACHINE LEARNING | | | | | | |
|-------------------------------------|----------|------------|---|--|--|--|
| Semester | II | Hours/Week | 4 | | | |
| Course Code | P19DS206 | Credits | 4 | | | |

On successful completion of this course, students will be able to:

| S. No. | Course Outcome | Level | Unit |
|--------|---|-------|------|
| CO1 | Perceive the Types of ML and develop Perceptron model | K6 | Ι |
| CO2 | Develop a supervised ML model for the given business problem | K5 | II |
| CO3 | Assess the pre-processing methods and reduce dimensions of data | K6 | III |
| CO4 | Evaluate the training and the testing of the designed ML model | K6 | IV |
| CO5 | Develop an unsupervised ML model for the given business problem | K5 | V |
| CO6 | Deploy machine learning models into production environment | K6 | All |

2. A. SYLLABUS

Unit-1. ML Basics and Perceptron

Three types Machine Learning – Three steps of ML process – Perceptron neural network – Adaline neural network – Stochastic gradient descent neural network

Unit-2. Supervised Learning classifiers

Logistic regression – Support vector machines – Kernel SVM – Decision Trees – K-Nearest Neighbour classifier – Random Forest – Linear Regression–Sentiment Analysis of Movie Reviews using Logistic Regression - Developing a web application with Flask

Unit-3. Pre-processing and Dimensionality Reduction

Pre-processing: Missing data, categorical data, feature scaling, feature selection. Dimensionality reduction: Principal Component Analysis, Linear Discriminant Analysis, Kernel PCA

Unit-4. Model evaluation

Pipelines - K-fold cross validation - Grid search - Confusion matrix, Precision, Recall, ROC curves, Scoring metrics – Majority vote classifier – Bagging, Bootstrapping, Adaptive Boosting

Unit-5. Unsupervised Learning classifiers and Multilayer NN

K-Means, K-Means++, Finding optimal no. of classifiers - Agglomerative Hierarchical clustering, Density based clustering -Multilayer Neural Network: Feed forward, Back Propagation Training, Multilayer Perceptron

B. TOPICS FOR SELF STUDY

| SNo | Topic Title | Web Link |
|-----|----------------------------|--|
| 1 | Kaggle Machine Learning | https://www.kaggle.com/learn/overview |
| 2 | IBM ML with Python: | https://www.edx.org/course/machine-learning-with-python- |
| | Practical introduction | a-practical-introduct |
| 3 | Predictive Analytics using | https://www.edx.org/course/predictive-analytics-using- |
| | Machine Learning | machine-learning |
| 4 | Google AI | https://ai.google/education/ |

C. TEXT BOOK(S)

1. Sebastian Raschka, "Python Machine Learning", First Edition, [PACKT], 2015.

D. REFERENCES BOOK(S)

1. Andreas C Muller and Sarah Guido, Introduction to Machine Learning with Python, Shroff Publishers, ISBN 978935213451

- Joel Grus, "Data Science from Scratch", First Edition, O'Reilly,2015
 Gavin Hackeling, "Mastering machine learning with scikit-learn", First Edition, [PACKT], 2014

E. WEB LINKS

- <u>https://www.edx.org/course/predictive-analytics-using-machine-learning</u>
- <u>https://www.edx.org/course/machine-learning-with-python-a-practical-introduct</u>

3. SPECIFIC LEARNING OUTCOMES

| Unit | Торіс | Topic Learning Outcomes | Level |
|------|---|---|----------------|
| Ι | Machine Learning Basics and Perceptron | | |
| 1.1 | Three types Machine Learning | Select a machine learning model, given business, scientific and societal use cases. | K2 |
| 1.2 | Three steps of ML process | Explain machine learning steps based on the given use cases. | K2 |
| 1.3 | Perceptron neural network | Draw Perceptron Neural Network for AND, OR and NOT logic gate operations. Create Perceptron in <i>sklearn</i> for a simple dataset that contains 4 samples for 2 numerical input features and corresponding y values, perform training and testing for an | K3 K5 |
| 1.4 | Adaline neural network | unknown sample. Differentiate Adaline from Perceptron neural network. | K2 |
| 1.5 | Stochastic gradient descent neural network | Differentiate Perceptron, Adaline and Stochastic Gradient Descent Neural Networks. Create a dataset, perform training, testing and print error rates for SGD Neural Network using <i>sklearn</i> , for the given use case. | K2 K6 |
| II | Supervised Learning C | lassifiers | • |
| 2.1 | Types of Supervised ML and ML pipeline | Identify the type of supervised ML, given use cases. Identify the steps of ML pipeline for classification and regression problems. | K2 K2 |
| 2.2 | Classes and methods of ML models available in <i>sklearn</i> package | Import and instantiate ML models using sklearn. Call methods and properties of ML models in sklearn. | K3 K3 |
| 2.3 | Linear Regression in Scikit Learn | Identify input features and target from dataset, preprocess data, split dataset for training & testing, create LinearRegression model using <i>sklearn</i> , perform training and testing and print MSE, SSE and R2 errors, for the given regression problem. Create Ridge Regression model in sklearn, for the given use case, by following ML pipeline steps. Create LASSO Regression model in sklearn, for the given use case, by following ML pipeline steps. Create Polynomial Regression model in sklearn to represent non-linearity assumption, for the given use case, by following ML pipeline steps. | K6 K6 K6 |
| 2.4 | Perceptron using Scikit Learn | Identify input features and target from dataset, preprocess data, split dataset for training & testing, create Perceptron using <i>sklearn</i> , perform training and testing and print classification accuracy for the given classification | K6 |

| | | problem. | |
|-----|---------------------|--|----------|
| | | | |
| | | | |
| | | | |
| 2.5 | Logistic regression | Compare LR against Perceptron based on the given use cases. | K2 K3 |
| | | Compute Sigmoid activation function value given weights | к2 |
| | | Explain how LR can be used to predict probability values, | 112 |
| | | with use cases. | K4 |
| | | Choose if the given ML model suffers from Overfitting or | |
| | | Underfitting; Also select if the model has high/low variance or high/low bias | K3 |
| | | Explain how regularization solves overfitting issue of a ML model. | K5 |
| | | Create LR model in <i>sklearn</i> by following ML system | |
| | | design pipeline and compare against Perceptron and select the best model, for a given use case. | |
| | | | |
| 2.6 | Support vector | Compare SVM against LR and Perceptron. | K2 |
| | SVM | syntax of SVC classifier, | KZ |
| | | Interpret parameters and their values, Ggiven <i>sklearn</i> | K2 |
| | | syntax of SVC for Kernel SVM | K2 |
| | | Create SVM model in <i>sklearn</i> by following ML system | K6 |
| | | design pipeline and select the best model among | |
| | | Perceptron, LR and SVM models, for the given use case. | |
| | | | |
| 2.7 | Decision Trees | Find the best split of DT node using Entropy value. | K4 |
| | | Find the best split of DT node using Gini Index value. | |
| | | Create a DT manually using ID3 algorithm for the | K4 |
| | | specified depth, given a dataset. | K4 |
| | | specified depth given a dataset | |
| | | Create syntax for <i>sklearn</i> DecisionTreeClassifier class, | K4 |
| | | given parameter values. Create syntax for sklearn DecisionTreeRegressor class | К4 |
| | | given parameter values, | IX I |
| | | Create Decision Tree model in <i>sklearn</i> by following ML | K4 |
| | | system design pipeline, compare its performance against other ML models and select the best model for a given use | |
| | | case. | K5 |
| | | | |
| | | | |

| 2.8 | Random Forest | Select Random Forest or Decision Tree approach based on | K5 |
|------|------------------------|--|------|
| | | the business objective. | K5 |
| | | case | K5 |
| | | Create manually Random Forest using CART trees for the | 110 |
| | | given use case. | K4 |
| | | Create syntax for <i>sklearn</i> RandomForestClassifier class, | K4 |
| | | given parameter values. | |
| | | Create syntax for <i>sklearn</i> RandomForestRegressor class, | K5 |
| | | Create Random Forest classification model in <i>sklearn</i> by | |
| | | following ML system design pipeline, compare its | K5 |
| | | performance against Decision Tree classification model | _ |
| | | and select the best model, for a given use case. | |
| | | Create Random Forest regression model in <i>sklearn</i> by | |
| | | following ML system design pipeline, compare its | |
| | | performance against Decision Tree regression model and | |
| | | select the best model, For a given use case. | |
| | | | |
| 2.9 | K-Nearest Neighbour | Classify ML methods into parametric and non-parametric | K2 |
| | classifier | categories. | |
| | | Classify ML methods into Easy Learners and Lazy | K2 |
| | | Learners. | K3 |
| | | Select the best value for k for KINN classifier. | KS |
| | | for number of neighbors and distance metric | K5 |
| | | For the given use case, build dataset, create KNN model | 110 |
| | | and evaluate its performance. | |
| | | | |
| 2.10 | Sentiment Analysis of | Create feature vectors manually by computing term | K4 |
| | Movie Reviews using | frequency, inverse document frequency and TF-IDF | 17.4 |
| | Logistic Regression | Values, for the given use case. | K4 |
| | | class for the given use case | K6 |
| | | Create a sentiment analysis system using Logistic | no |
| | | Regression model for the movie reviews dataset. | |
| | | | |
| 2.11 | Developing a web | Create a simple website and deploy a machine learning | K6 |
| TTT | application with Flask | model using Flask. | |
| | Pre-processing and Di | Eind missing volves and replace with mean (median (| K2 |
| 3.1 | and categorical data | mode values for numerical data | КЭ |
| | and callgorical data | Apply LabelEncoder to ordinal attributes in order to | K4 |
| | | represent string values to integers. | |
| | | Apply One Hot Encoder to nominal attributes to represent | K4 |
| | | categorical data. | |
| | | | |

| 3.2 | Feature scaling and | Compute normalized values using min max scaling. | K3 |
|-----|---------------------|--|----------|
| | feature selection | Compute standardized values using standard scaling. | K3 |
| | | Apply MinimaxScaler and StandardScaler to preprocess | K5 K5 |
| | | Compute important features using L2 and L1 | KJ KA |
| | | regularization methods | 174 |
| | | Compute important features using Random Forest | |
| | | algorithm. | |
| 3.3 | Principal Component | Compute Covariance matrix, Eigen vectors and Eigen | K4 |
| | Analysis | values of a given matrix. | |
| | | For <i>load_digits</i> dataset from <i>sklearn</i> , reduce original | K5 |
| | | dimension $(1/9/x64)$ into low dimension $(1/9/x7)$ using | 175 |
| | | PCA algorithm and print its shape. | K5 |
| | | matrix print N principle component vectors and the | |
| | | variance each principle components holds (called | K5 |
| | | explained_variance_ratio value in <i>sklearn</i>). | |
| | | Create PCA model in sklearn and visualize data in low | |
| | | dimensions using matplotlib for the given use case (For ex. | |
| | | Breast Cancer dataset from <i>sklearn</i>). | |
| | | | |
| 3.4 | Linear Discriminant | Explain Singular Valued Decomposition with an example. | K2 |
| | Analysis | Create LinearDiscriminantAnalysis model for a small N+1 | |
| | | dimensional matrix, print N components and | K4 |
| | | explained_variance_ratio value. | |
| | | Create LinearDiscriminantAnalysis model in sklearn and | K5 |
| | | given use case (For ex. Iris dataset in <i>sklearn</i>) | |
| | | given use cuse (i of ex. mis dataset in skearn). | |
| 3.5 | Kernel PCA | Create a syntax for KernelPCA for the values of the input | K4 |
| | | parameters. | |
| | | Reduce original dimension (1797x64) into low dimension | K4 |
| | | (1/9/x) using KernelPCA algorithm and print its shape, | V5 |
| | | Create and visualize make moons dataset using PCA and | KJ |
| | | KernelPCA models. | |
| | | | |
| IV | Model Evaluation | | I |
| 4.1 | PipelineE | Create Pipeline in sklearn, given the requirements for | K4 |
| | | transformers and estimators. | |
| | | Create a Pipeline for a ML model, perform training and | K6 |
| | | testing and show its performance values, for the given | |
| | | business use case. | |
| 4.2 | Holdout validation | Divide the dataset for training, validation and testing based | K3 |
| | | on the performance requirement. | |
| - | | | |
|-----|--------------------------|--|------|
| 4.3 | K-fold cross validation | Explain the working of Kfold cross validation, Stratified Kfold CV, Leave one out CV and Shuffle split CV. | K4 |
| | | Create a ML model and compute the CV score (which may be classification accuracy or error) using | K6 |
| | | sklearn.model selection, for the given use case. | K6 |
| | | Create a Pipeline for a ML model and compute the CV | |
| | | classification accuracy or error using | |
| | | <i>sklearn.model_selection</i> , for the given use case. | |
| | | | 77.4 |
| 4.4 | Grid search | Create a syntax for a GridSearchCV model for the given | K4 |
| | | Create and perform GridSearchCV search, for the given | K5 |
| | | use case, with various parameters values for the chosen | |
| | | ML model and select best parameter values; then create | |
| | | that ML model with the best parameter values and show | |
| 4.5 | | performance results. | 17.5 |
| 4.5 | Precision and Recall | of a ML system. | К5 |
| | | Compute manually precision and recall values of a ML | K3 |
| | | system. For example, given the sequence of predictions of | |
| | | an email spam classifier. | K4 |
| | | Evaluate the performance of a ML model using precision, | |
| | | recarr and iscore values in skiearn. | |
| 4.6 | ROC curve | Plot ROC curve with AUC values for the ML models. | K5 |
| | | compare performances of many ML models using sklearn | 110 |
| | | and give recommendations to business clients. | |
| 4.7 | Majority vote classifier | Explain the concepts of majority voting classifier. | K2 |
| | | Given a simple dataset of 5 samples with 2 input numerical | VC |
| | | VotingClassifier in sklearn with atleast 2 ML classifiers | KO |
| | | such as Logistic Regression and SVC, perform training | |
| | | and testing and verify output y manually. | K5 |
| | | Create a VotingClassifier for the given classification | |
| | | problem and report the performance results. | |
| | | | |
| 4.8 | Bootstrapping | Explain bootstrapping with random replacement policy | K2 |
| 10 | Bagging | With examples. | K5 |
| 4.7 | Dagging | perform training and testing for a simple dataset. | K.J |
| | | Create BaggingClassifier model with Bootstrapping | K5 |
| | | feature with 10 decision trees, perform training and testing | |
| | | for a simple dataset. | K5 |
| | | Create BaggingClassifier model with 10 SVC classifiers, | VC |
| | | Create BaggingClassifier model perform training and | KO |
| | | testing for the business use case. Compare its performance | |
| | | against other ML classifiers. | |
| | | | |
| 1 | | | |

| 4.10 | Adaptive Boosting | Create AdaBoostClassifier with 5 estimators for a simple | K5 |
|------|-------------------------|--|------------|
| | | dataset. | |
| | | Create GradientBoostingClassifier with 10 estimators for a simple dataset. | K5 |
| | | Create XGBoost classifier with 10 estimators for a simple | K5 |
| | | Create AdaBoostClassifier model, perform training and | K6 |
| | | testing for the business use case. Compare its performance | 110 |
| | | against Gradient boosting and bagging ensemble | |
| | | classifiers. | |
| V | Unsupervised Learning | Classifiers and Multilaver NN | |
| 5.1 | K-Means | Identify and recommend the appropriate ML strategy | K4 |
| | | (which may be classification, regression or clustering | |
| | | approach) for the given business or scientific or societal | |
| | | application. | K5 |
| | | Given a small dataset with 2 numerical features with 4 | |
| | | samples and values for 2 centroids, apply manually | K5 |
| | | KMeans algorithm and predict 2 clusters (You can use | |
| | | Euclidean distance). | |
| | | Create the syntax of KMeans model in sklearn (assume | K5 |
| | | input samples X are already available) and no. of clusters | 17.5 |
| | | to predict. Also, perform training and testing on X. Print | KS |
| | | Create KMeans model and predict the optimal number of | K6 |
| | | clusters using Flbow method, given input samples X | KU |
| | | Create KMeans model and predict the optimal number of | |
| | | clusters and evaluate the quality of clusters using | |
| | | Silhouette Coefficients, given input samples X. | |
| | | Create and develop KMeans clustering system in <i>sklearn</i> | |
| | | for the business use case and provide recommendations to | |
| | | users. | |
| 5.2 | K-Means++ | Create and develop KMeans++ clustering system in | K6 |
| | | sklearn for the business use case and provide | |
| | | recommendations to users. | |
| 5.3 | Agglomerative | Perform manually hierarchical clustering using single | K4 |
| | Hierarchical clustering | linkage and show clusters of students, for a one | |
| | | dimensional data representing marks of students (say, 5 | V A |
| | | Students), Perform manually hierarchical clustering using complete | K 4 |
| | | linkage and show clusters of students for a one | КЛ |
| | | dimensional data representing marks of students (say 5 | 114 |
| | | students). | |
| | | Perform manually hierarchical clustering using single | K4 |
| | | linkage, draw Dendrogram and choose the number of | |
| | | clusters, for a one dimensional data representing marks of | K6 |
| | | students (say, 5 students). | |
| | | Plot Dendrogram for the given input samples and choose | |
| | | the clusters using <i>scipy</i> package. | |
| | | Apply agglomerative clustering using <i>sklearn</i> package for | |
| | | the given business use case and predict N clusters. | |
| 5.4 | Density based | Create clusters by applying DBSCAN algorithm in sklearn | K5 |
| | clustering | and visualize clusters, for a given dataset. | |

| 5.5 | Feed Forward Multilayer Neural Network | Draw a Single layer neural network with input layer and output layer. The dataset represents details of 5 persons. The 3 input features are 'smoking, 'obesity' and 'exercise' with values 0 or 1. The target or output feature is 'diabetic' which can have a value 0 or 1. Draw a Multilayer neural network for XOR operations and differentiate from Perceptron. Draw the architectural diagram of Multilayer Perceptron neural network with bias input nodes, given input samples X and target output y values and the size of the hidden layer. Create syntax for MLPClassifier in sklearn given parameter values. Create MLPClassifier, perform preprocessing, training and testing. Print the performance values of classification metrics. Also print the learnt weight matrix and bias vector | K4 K4 K3 K6 |
|-----|--|---|----------------------|
| | | metrics. Also print the learnt weight matrix and bias vector values, for the given business use case. | |
| 5.6 | Back Propagation Training | Compute predicted output and propagate error for one iteration, given a simple Multilayer Perceptron having just one hidden layer with values for input sample, weights and | K4 |
| | | desired output. Design a ML system using MLPClassifier and compare its performance against other classifiers, for the given use case. | K6 |

4. Mapping

| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
|-----|-----|-----|-----|-----|-----|-----|-----|-----|-----|------|------|------|------|
| CO1 | Н | L | | L | | | | | | | | | |
| CO2 | Н | Μ | L | Η | | | | | | | | Н | Н |
| CO3 | Н | Μ | | Η | Н | Н | | | | | | Н | Н |
| CO4 | Η | Μ | | Η | Η | Η | | | | | | Н | |
| CO5 | Η | | | | | Η | | Η | | М | | | Н |
| CO6 | Η | | | | | | Η | | Η | | Н | | |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 2. Open Book Test.
- 3. Peer Learning Report, Assignment, Group Presentation, Group Discussion, project Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).
- 4. Pre-Semester & End Semester Theory Examination

INDIRECT:

- 1. Course evaluation survey
- 2. Faculty feedback about the course.

Name of the Course Coordinator: Dr. K. Rajkumar

| Elective II: NATURAL LANGUAGE PROCESSING | | | | | | | |
|--|----------|------------|---|--|--|--|--|
| Semester | II | Hours/Week | 4 | | | | |
| Course Code | P19DS2:2 | Credits | 4 | | | | |

After the successful completion of this course the students will be able to

| S.No. | Course Outcomes | Level | Unit |
|-------|--|-------|------|
| CO1 | Experiment with text pre-processing and classification | K6 | Ι |
| CO2 | Create language models and POS tagging | K6 | II |
| CO3 | Evaluate the context free grammars and parse sentences | K6 | III |
| CO4 | Validate the meaning of sentences | K6 | IV |
| CO5 | Explain the Dependency parsing | K6 | IV |
| CO6 | Design NLP applications | K6 | V |

2. A. SYLLABUS

Unit-1.NLP Basics

What is NLP - Frequency distributions, Collocations – Unigram, Bigrams – word sense disambiguation – Pronoun resolution – Machine Translation – Textual entilement – Limitations of NLP –Conditional frequency distributions – Plotting distributions – Lexicons – WordNet – Semantic similarity

Unit-2. Text Processing

Accessing text from web and disk – NLP Pipeline – String processing – Text processing with UNICODE – Regular expressions: Metacharacters, Ranges and Closure – Useful applications of Regular applications – Stemming and Lemmatization – Text tokenization using regular expressions – Segmentation –Introduction to Dynamic programming – NetworkX package. Categorizing and Tagging Words: POS tagging – Part of speech Tagset – Reading corpora – Exploring corpora – Regular expression tagging – Look up tagging – Ngram Tagging – Transformation based tagging – Determining category of a word.

Unit-3. Learning to Classify Text and Information Extraction

Document classification – Sequence classification: Greedy approach, Hidden Markov Models and Conditional Random Fields –Recognizing text entailment –Text classifiers: Decision Trees, Naïve Bayes and Maximum Entropy classifiers – Generative vs conditional classifiers. Information Extraction: Architecture – Entity Recognition: Chunking, Chinking – Named Entity Recognition – Relation Extraction

Unit-4. Analysing Sentence Structure

Ambiquity – Context Free Grammar: Simple grammar, writing your own grammar – Parsing with CFGs – Dependence grammar – Valency and Lexicon – Probabilistic CFG – Feature Based Grammars

Unit-5. Analysing Meaning of Sentences

Propositional logic – First order logic – First order theorem proving – Model checking – Quantification – Discourse Processing

B. TOPICS FOR SELF-STUDY:

| S.No. | Topics | Web Links |
|-------|---------------------------------------|--|
| 1 | Natural language processing with Deep | https://www.youtube.com/watch?v=OQQ- |
| | Learning | W_63UgQ&list=PL3FW7Lu3i5Jsnh1rnUwq_Tcyl |
| | | Nr7EkRe6 |
| 2 | Latent structure models for NLP | https://deep-spin.github.io/tutorial/acl.pdf |
| 3 | Chatbots | https://www.analyticssteps.com/blogs/learn- |
| | | everything-about-machine-learning-chatbots |
| 4 | Language Interpretability Tool (LIT) | https://github.com/PAIR-code/lit |

C. TEXT BOOK

1. Jurafsky and Martin, Speech and Language Processing: An Introduction to Natural Language Processing, Computational Linguistics and Speech Recognition, 3ed. 2020.

D. REFERENCES BOOKS

- 1. Steven Bird, Ewan Klein, and Edward Loper, Natural Language Processing with Python– Analyzing Text with the Natural Language Toolkit.
- 2. Indurkhya, Nitin and Fred Damerau, Handbook of Natural Language Processing, 2ed, 2010, Chapman & Hall/CRC.
- 3. Christopher Manning and HinrichSchutze, Foundations of Statistical Natural Language Processing, MIT Press.

E. WEB LINKS

- 1. <u>https://london.ac.uk/sites/default/files/study-guides/introduction-to-natural-language-processing.pdf</u>
- 2. <u>http://www.datascienceassn.org/sites/default/files/Natural%20Language%20Processing%20</u> with%20Python.pdf

| Unit/ Section | Course Content | Learning outcomes | Level |
|------------------|--|---|----------|
| Ι | | NLP Basics | |
| 1.1 | Frequency distributions, Collocations | Explain the Collocations | K4 |
| 1.2 | Unigram, Bigrams – word sense disambiguation | Define Unigram, Bigrams Find the word sense disambiguation | K2 K5 |
| 1.3 | Pronoun resolution | Explain the Pronoun Resolution | K4 |
| 1.4 | Machine Translation – Textual entilement | Construct the Machine Translation | K6 |
| 1.5 | Limitations of NLP | List the limitations of NLP | K2 |
| 1.6 | Conditionalfrequencydistributions-Plottingdistribution | Develop the Conditional frequency distributions | K6 |

| 17 | Lexicons – WordNet – | Construct the WordNet | K6 |
|-----|--|--|----------|
| 1.7 | Semantic similarity | | |
| II | | Text Processing | |
| 2.1 | Accessing text from web and disk – NLP Pipeline – String processing – Text processing with UNICODE | Assess the Text from Web Evaluate the model for text processing | K6 K5 |
| 2.2 | Regular expressions: Metacharacters, Ranges and Closure – Useful applications of Regular applications – Stemming and Lemmatization | Apply the Regular Expression Explain the Stemming and Lemmatization | K5 K5 |
| 2.3 | Text tokenization using regular expressions – Segmentation | Design the algorithm for Text tokenization | K6 |
| 2.4 | Categorizing and Tagging Words: POS tagging – Part of speech Tagset – Reading corpora – Exploring corpora | Develop the algorithm for POS tagging Design the Tagset for POS | K6 K6 |
| 2.5 | Regular expression tagging – Look up tagging – Ngram Tagging – Transformation based tagging | Explain the Regular expression tagging | K5 |
| 2.6 | Determining category of a word. | Determine the category of a word. | K5 |
| III | Learning to Cla | ssify Text and Information Extraction | |
| 3.1 | Document classification – Sequence classification: Greedy approach, Hidden Markov Models and Conditional Random Field | Explain the Document Classification Assess the Sequence Classifications | K5 K6 |
| 3.2 | Recognizing text entailment – Text classifiers: Decision Trees, Naïve Bayes and Maximum Entropy classifiers | Evaluate the Text Classifiers Compare the text classifiers | K6 K6 |
| 3.3 | InformationExtraction:Architecture–EntityRecognition:Chunking,Chinking–NamedRecognition–RelationExtraction– | Construct the Architecture Information Extraction | K6 |
| IV | ANALYSI | NG SENTENCE STRUCTURE | |
| 4.1 | Ambiquity – Context Free Grammar: Simple grammar, writing your own grammar | Explain the Ambiquity Grammar Construct Context Free Grammar | K5 K6 |
| 4.2 | Parsing with CFGs | Design the Parsing with CFG | K6 |
| 4.3 | Dependence grammar - Valency and Lexicon | Explain the Dependence Grammar | K5 |
| 4.4 | Probabilistic CFG – Feature Based Grammars | Construct the Probabilistic CFG | K6 |
| V | ANALYSIN | IG MEANING OF SENTENCES | |
| 5.1 | Propositional logic – First order logic – First order theorem proving | Explain the First order logic Verify the theorem on First Order logic | K5 K5 |
| 5.2 | Model checking – | Explain the Model checking | K5 |

| | Quantification | | |
|-----|----------------------|------------------------------------|----|
| 5.3 | Discourse Processing | Construct the Discourse Processing | K6 |

4. MAPPING (CO, PO, PSO)

| L-I | Jow | | M-Moderate | | | | | | H- High | | | | |
|------------|-----|-----|------------|-----|-----|-----|------------|------------|---------|------|------|------|------|
| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
| CO1 | Н | Μ | Н | Н | Н | Н | - | - | Μ | Н | Η | Μ | - |
| CO2 | Μ | Н | Μ | Н | Μ | Μ | - | Μ | Μ | Μ | Μ | - | - |
| CO3 | Н | Μ | Н | Н | Н | Н | - | - | Н | Н | Μ | - | - |
| CO4 | Н | Н | L | Н | Μ | Н | - | Μ | Н | Н | Н | - | |
| CO5 | Н | Μ | Н | Μ | Н | Н | - | - | Μ | Н | Η | - | - |
| CO6 | Н | Н | Н | Μ | Μ | Н | Н | - | Μ | Н | Н | - | - |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 2. Open Book Test.
- 3. Cooperative Learning Report, Assignment, Group Presentation, Group Discussion, project Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).
- 4. Pre-Semester & End Semester Theory Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator : Dr. Janani Selvaraj

| ELECTIVE-3: HEALTH CARE DATA ANALYTICS | | | | | | |
|--|----------|------------|---|--|--|--|
| Semester | Ι | Hours/Week | 3 | | | |
| Course Code | P19DS2:3 | Credits | 3 | | | |

After the successful completion of this course the students will be able to

| S.No. | Course Outcomes | Level | Unit |
|-------|---|-------|------|
| CO1 | Explain the Coding system of Electronic Health Records | K5 | Ι |
| CO2 | Develop the algorithms for Biomedical Analysis | K5 | Ι |
| CO3 | Design the Algorithm for Healthcare Data using NLP, SMA | K6 | III |
| CO4 | Construct the predictive models for Healthcare Data | K6 | IV |
| CO5 | Analyze the role of Analysis in Pervasive Health | K5 | V |
| CO6 | Design the Computer-Assisted Medical Image Analysis Systems | K6 | V |

2. A. SYLLABUS

Unit-1. Introduction

Introduction to Healthcare Data Analytics- Electronic Health Records–Components of EHR-Coding Systems- Benefits of EHR- Barrier to Adopting E<u>H</u>R Challenges-Phenotyping Algorithms.

Unit-2. Analysis

Biomedical Image Analysis- Mining of Sensor Data in Healthcare- Biomedical Signal Analysis-Genomic Data Analysis for Personalized Medicine.

Unit-3. Analytics

Natural Language Processing and Data Mining for Clinical Text- Mining the_Biomedical- Social Media Analytics for Healthcare.

Unit-4. Advanced Data Analytics

Advanced Data Analytics for Healthcare– Review of Clinical Prediction Models- Temporal Data Mining for Healthcare Data- Visual Analytics for Healthcare- Predictive Models for Integrating Clinical and Genomic Data- Information Retrieval for Healthcare- Privacy-Preserving Data Publishing Methods in Healthcare.

Unit-5. Applications

Applications and Practical Systems for Healthcare– Data Analytics for Pervasive Health- Fraud Detection in Healthcare- Data Analytics for Pharmaceutical Discoveries- Clinical Decision Support Systems- Computer-Assisted Medical Image Analysis Systems- Mobile Imaging and Analytics for Biomedical Data

B. TOPICS FOR SELF-STUDY

| S.No. | Topics | Web Links |
|-------|---------------------------------|---|
| 1 | Electronic Data Warehouse (EDW) | https://www.osplabs.com/healthcare-analytics/ |
| 2 | Big Data in Health care | https://www.wipro.com/healthcare/advanced- |
| | | healthcare-data-analytics/ |

| 3 | Health Care Economics | https://healthcare.business.uconn.edu/certificate- |
|---|-----------------------|--|
| | | health-care-analytics/ |

C. TEXT BOOK(S)

- 1. Chandan K. Reddy and Charu C Aggarwal, "Healthcare data analytics", Taylor &Francis, 2015
- 2. Hui Yang and Eva K. Lee, "Healthcare Analytics: From Data to Knowledge to Healthcare Improvement, Wiley, 2016.

D. REFERENCE BOOKS

- 1. Vikas Kumar, Healthcare Analytics Made Simple: Techniques in healthcare computing using machine learning and Python, Packt Publishing, 2018
- 2. Ross M. Mullner, Edward M. Rafalski, Healthcare Analytics Foundations and Frontiers, Routledge, 2019.

E. WEB LINKS

- <u>https://onlinedegrees.sandiego.edu/classes/advanced-health-care-analytics/</u>
- <u>https://www.hci.net.in/courses/advanced-post-graduate-diploma-healthcare-decision-analytics/</u>

| Unit/ Section | Торіс | Learning outcomes | Level |
|------------------|--|--|-------|
| Ι | | Introduction | |
| 1.1 | Introduction to Healthcare Data Analytics | • Understand the concepts of Healthcare Data Analytics | K3 |
| 1.2 | Electronic Health Records | • Explain the concepts of EHR | K4 |
| 1.3 | Components of EHR | • List the components of EHR | K2 |
| | | • Discuss the features of EHR | K4 |
| 1.4 | Coding Systems | Development of Coding Systems algorithm | K5 |
| 1.5 | Benefits of <u>EHR</u> | • List the benefits of EHR | K2 |
| 1.6 | Barrier to Adopting E <u>H</u> R Challenges | • Discuss the Challenges in EHR | K4 |
| 1.7 | Phenotyping Algorithms. | • Design the Phenotyping Algorithms | K6 |
| II | | Analysis - I | |
| 2.1 | Biomedical Image | • Explain the Biomedical Image | K5 |
| 2.2 | Mining of Sensor Data in | Explore the mechanism to get the data | K4 |
| | Healthcare | from sensor | |
| | | • Design the algorithm for Mining the sensor data | K6 |
| | Biomedical Signal Analysis | Explain the Biomedical Signal Analysis | K5 |
| 2.3 | Genomic Data Analysis for Personalized Medicine | • Develop the personalized medicine system using Genomic Data | K6 |
| III | | Analysis - II | |
| 3.1 | NaturalLanguageProcessingandData | • Develop the algorithms for mining text in HER using NLP. | K5 |
| | Mining for Clinical Text | Design the data mining tool for Clinical text data | K6 |

| 3.2 | Mining the_Biomedical | • Design the mining algorithm for Biomedical data | K6 |
|-----|--|---|----|
| 3.3 | Social Media Analytics for Healthcare. | • Develop the algorithms for Health care data using Social Media Analysis | K5 |
| IV | | Advanced Data Analytics | |
| 4.1 | Advanced Data Analytics for Healthcare– Review of Clinical Prediction Models | List the techniques for Advanced Data Analytics for Healthcare Assess the Clinical Prediction Modela | K2 |
| 4.2 | Temporal Data Mining for Healthcare Data | Assess the Chincal Frediction Models Design the temporal data mining algorithms for Healthcare Data | K6 |
| 4.3 | Visual Analytics for Healthcare | Develop the Visual Analytics for Healthcare | K5 |
| 4.4 | Predictive Models for Integrating Clinical and Genomic Data | • Design the predictive model using Integrating Clinical and Genomic Data | K6 |
| 4.5 | Information Retrieval for Healthcare | • Apply the Information Retrieval for Healthcare | K4 |
| 4.6 | Privacy-Preserving Data Publishing Methods in Healthcare. | • Explain the Data publishing methods in Healthcare | K5 |
| V | | Programming with R | |
| 5.1 | Applications and Practical Systems for Healthcare | • List the applications of the health care data analysis | K2 |
| 5.2 | Data Analytics for Pervasive Health | • Develop the algorithm for Pervasive Health <u>using DA</u> | K5 |
| 5.3 | Fraud Detection in Healthcare | • Explain the Fraud Detection in Healthcare | K4 |
| 5.4 | Data Analytics for Pharmaceutical Discoveries | • Develop the algorithm for Pharmaceutical Discoveries <u>using DA</u> | K5 |
| 5.5 | Clinical Decision Support Systems | Design the Clinical Decision Support Systems | K6 |
| 5.6 | Computer-Assisted Medical Image Analysis Systems | Construct the Computer-Assisted Medical Image Analysis Systems | K6 |
| 6.7 | Mobile Imaging and Analytics for Biomedical Data | • Design the System for Mobile Imaging and Analytics to the Biomedical Data | K6 |

4. MAPPING (CO, PO, PSO)

| L-Low | | | M-Moderate | | | | | | | H- High | | | |
|------------|-----|-----|------------|-----|-----|-----|------------|-----|-----|---------|------|------|------|
| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
| CO1 | H | Μ | Н | Μ | L | Μ | Μ | Μ | Μ | Μ | - | Н | Н |
| CO2 | H | Н | Μ | L | - | L | - | L | Μ | Μ | - | Μ | - |
| CO3 | Н | Н | Н | Н | Μ | Μ | L | - | - | Н | Н | Н | М- |
| CO4 | Н | Н | Н | Н | Н | Μ | L | Μ | Μ | Н | Н | Μ | Н |
| CO5 | H | Μ | - | Μ | L | Μ | - | Н | Μ | Н | Н | Μ | - |
| CO6 | Μ | Μ | - | Н | Μ | L | - | - | L | Н | Н | Н | Μ |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 5. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 6. Open Book Test.
- 7. Cooperative Learning Report, Assignment, Group Presentation, Group Discussion, project, Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).
- 8. Pre-Semester & End Semester Theory Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator: Dr. M. Lovelin Pon Felciah

| CORE III: DATA AND VISUAL ANALYTICS LAB | | | | | | | | | |
|---|----------|------------|---|--|--|--|--|--|--|
| Semester | II | Hours/Week | 3 | | | | | | |
| Course Code | P19DS2P3 | Credits | 3 | | | | | | |

On successful completion of this course, students will be able to:

| CO# | Course Outcome | Level | Activity |
|-----|---|-------|----------|
| CO1 | Create data analytics systems using Numpy | K6 | 1, 2 |
| CO2 | Create data wrangling systems using Pandas | K6 | 3-6, 10 |
| CO3 | Create data visualization systems using Seaborn | K6 | 7 |
| CO4 | Crate time series analysis systems using Pandas time series | K6 | 8 |
| CO5 | Create interactive dashboards using Tableau | K6 | 11 |
| CO6 | Build and deploy end to end data analytics product at client site | K6 | 9, 12 |

2. SYLLABUS

| Activity | Lab Activity Description |
|----------|--|
| 1 | Data Analytics using NumPy |
| 2 | Data Analytics using Advanced NumPy |
| 3 | Pandas Indexing and Selection |
| 4 | Pandas Grouping and Aggregation |
| 5 | Pandas Concatenation, Merging and Join |
| 6 | Data Cleaning in Pandas |
| 7 | Data Visualization using Seaborn |
| 8 | Pandas Time Series Analysis |
| 9 | Exploratory Data Analysis on Cardiovascular Data |
| 10 | Advanced Data Wrangling in Pandas |
| 11 | Interactive Dashboard Creation in Tableau |
| 12 | Telecom data analytics system |

| Activity# | Lab Activity | Learning Outcome | Level |
|-----------|--|---|-------|
| 1 | Data Analytics using NumPy | Create data analytics platform using NumPy | K6 |
| 2 | Data Analytics using Advanced NumPy | Create advanced data analytics platform using NumPy | K6 |
| 3 | Pandas Indexing and Selection | Perform Pandas Indexing and Selection operations | K6 |
| 4 | Pandas Grouping and | Perform Pandas Grouping and | K6 |
| | Aggregation | Aggregation operations | |
| 5 | Pandas Concatenation, Merging | Perform Pandas Concatenation, Merging | K6 |
| | and Join | and Join operations | |
| 6 | Data Cleaning in Pandas | Perform data cleaning on large data | K6 |
| 7 | Data Visualization using | Develop Data Visualization systems | K6 |
| | Seaborn | using Seaborn | |
| 8 | Pandas Time Series Analysis | Perform Pandas Time Series Analysis | K6 |
| 9 | Exploratory Data Analysis on | Develop Exploratory Data Analysis | K6 |
| | Cardiovascular Data | platform on Cardiovascular Data | |
| 10 | Advanced Data Wrangling in | Perform Advanced Data Wrangling in | K6 |
| | Pandas | Pandas | |

| 11 | Interactive Dashboard Creation | Create Interactive Dashboard using | K6 |
|----|--------------------------------|---|----|
| | in Tableau | Tableau | |
| 12 | Telecom data analytics system | Build and deploy Telecom data analytics | K6 |
| | | system | |

4. MAPPING (CO, PO, PSO) L-Low

M-Moderate

H- High

| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
|-----|-----|-----|-----|-----|-----|-----|------------|-----|-----|------|------|------|------|
| CO1 | H | Μ | L | Μ | Μ | Μ | L | - | - | Н | Μ | Μ | - |
| CO2 | Н | Μ | Н | Μ | Μ | Н | Μ | Μ | L | Н | Н | - | Μ |
| CO3 | H | Μ | Μ | L | Н | Μ | - | - | L | Μ | Μ | Μ | - |
| CO4 | Н | L | Н | Н | Н | Н | Μ | Μ | L | Н | Μ | Μ | Н |
| CO5 | Н | Μ | L | Н | Н | L | L | Μ | Μ | Н | Н | Н | Μ |
| CO6 | Н | Μ | Μ | L | Н | L | L | L | Μ | Н | Н | - | - |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Practical Components): Closed Book
- 2. Cooperative Learning Report, Assignment, Group Discussion, project Report, Field Visit Report, Seminar.
- 3. Pre/Post Test, Viva, Report for each Exercise.
- 4. Lab Model Examination & End Semester Practical Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator: Dr. B. Karthikeyan

| CORE PRACTICAL IV: PRACTICAL MACHINE LEARNING LAB | | | | | |
|---|----------|---------|---|--|--|
| Semester II Hours/Week 3 | | | | | |
| Course Code | P19DS2P4 | Credits | 3 | | |

On successful completion of this course, students will be able to:

| CO# | Course Outcome | Level | Activity |
|----------|--|-------|----------|
| CO1 | Practice data and file formats; visualize data and familiarize Colab and | K6 | 1 |
| COI | Azure | | |
| CO2 | Build and deploy systems for business problems based on regression | K6 | 3,4 |
| 02 | models | | |
| CO^{2} | Build and deploy systems for business problems based on classification | K6 | 2,5,7,8 |
| 005 | models | | |
| CO4 | Build and deploy systems for business problems based on predictive | K6 | 6 |
| 04 | analytics | | |
| CO5 | Build and deploy systems for business problems based on tree models | K6 | 9,10 |
| C06 | Build and deploy systems for business problems based on clustering | K6 | 11 |
| | models | | |

2. SYLLABUS

| Activity | Lab Activity Description |
|----------|--|
| 13 | WarmUp: Familiarity with Data and Visualization |
| 14 | Pizza Liking Prediction using kNN |
| 15 | Fuel Amount Prediction using Linear Regression |
| 16 | House Price Prediction using LR with Regularization |
| 17 | Diabetes Classification using Logistic Regression |
| 18 | Predictive Analytics for Hospitals |
| 19 | Loan Approval Classification using SVM |
| 20 | Animal Classification using Decision Trees |
| 21 | Employee Hopping Prediction using Random Forests |
| 22 | Patients Physical Activities Prediction using Boosting |
| 23 | Shopping Mall Customer Segmentation using Clustering |

TOPICS FOR SELF STUDY

| S.No | Topic Title | Web Link |
|------|----------------------------------|--|
| 1 | Stock price prediction | https://www.kaggle.com/darkknight91/ge-stock |
| 2 | Wake up word detection for Alexa | https://github.com/Picovoice/wake-word-benchmark |
| 3 | Jane Street Market prediction | https://www.kaggle.com/c/jane-street-market- |
| | | prediction |
| 4 | HuBMap – Hacking the kidney | https://www.kaggle.com/c/hubmap-kidney- |
| | | segmentation |

| Activity# | Lab Activity | Learning Outcome | Level |
|-----------|--------------------------|--|-------|
| 1 | WarmUp: Familiarity with | Open, process and visualize various data and | K6 |
| | Data and Visualization | files using CoLab and Azure platforms | |
| 2 | Pizza Liking Prediction | Build kNN model, perform training and | K6 |
| | using kNN | prediction and compute accuracy values | |
| 3 | Fuel Amount Prediction | Perform preprocessing; build LR model, | K6 |

| | using Linear Regression | perform training and prediction; compute MSE and R2 error; compare performance against KNN regressor and SGDregressor models and interpret results | |
|----|--|---|----|
| 4 | House Price Prediction using LR with Regularization | Perform One Hot Encoding, build LR model, compute RMSE error and compare performance against SGD Regressor, RidgeCV and LassoCV and interpret results | K6 |
| 5 | Diabetes Classification using Logistic Regression | Create heatmap, build Logistic Regression model, print ROC curve and compare performance against LogisticRegressionCV with L1 and L2 and interpret results | K6 |
| 6 | Predictive Analytics for Hospitals | Perform prediction, Apply Forward Selection, plot AUC scores and Plot Gain curves and Life curves and interpret results | K6 |
| 7 | Loan Approval Classification using SVM | Perform EDA, Create LinearSVC model, Print accuracy, confusion matrix and classification report and compare LinearSVC model with SVC and SGDClassifier models | K6 |
| 8 | Animal Classification using Decision Trees | Create ID3 Decision Tree using Entropy metric, Create CART Decision Tree using Gini metric and Visualize graph using graphviz | K6 |
| 9 | Employee Hopping Prediction using Random Forests | Create RandomForestClassifier, perform training and testing; Print feature importance values; and Select the best number of trees based on out-of-bag error values | K6 |
| 10 | Patients Physical Activities Prediction using Boosting | Build GradientBoostingClassifier, fit and predict on test data; Find the best no. of decision trees and learning rate using GridSearch and Cross Validation; Build AdaBoost, LogisticRegressionCV and VotingClassifier; Interpret results and parameter values | K6 |
| 11 | Shopping Mall Customer Segmentation using Clustering | Perform Skew analysis; Build KMeans model; Apply Elbow method; Perform Cluster Analysis; Perform PCA; Build MeanShift clustering and Agglomerative clustering; Visualize clusters using Dendrogram | K6 |

4. MAPPING (CO, PO, PSO)

| L- | Low | | | | | M-M | oderat | e | | | H- I | Iigh | |
|------------|-----|-----|-----|-----|-----|-----|------------|-----|------------|------|------|------|------|
| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
| CO1 | Н | Μ | L | Μ | Μ | Μ | L | - | - | Н | Μ | Μ | - |
| CO2 | Н | Μ | Н | Μ | Μ | Н | Μ | Μ | L | Н | Н | - | Μ |
| CO3 | Н | Μ | Μ | L | Н | Μ | - | - | L | Μ | Μ | Μ | - |
| CO4 | Н | L | Н | Н | Н | Н | Μ | Μ | L | Н | Μ | Μ | Н |
| CO5 | Н | Μ | L | Н | Н | L | L | Μ | Μ | Н | Н | Н | Μ |
| CO6 | Η | Μ | Μ | L | Η | L | L | L | Μ | Н | Н | - | - |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Practical Components): Closed Book
- 2. Cooperative Learning Report, Assignment, Group Discussion, project Report, Field Visit Report, Seminar.
- 3. Pre/Post Test, Viva, Report for each Exercise.
- 4. Lab Model Examination & End Semester Practical Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator: Dr. K. Rajkuamr

| CORE PRACTICAL V: NATURAL LANGUAGE PROCESSING LAB | | | | | |
|---|----------|------------|---|--|--|
| Semester | II | Hours/Week | 3 | | |
| Course Code | P19DS2P5 | Credits | 3 | | |

On successful completion of this course, students will be able to:

| CO# | Course Outcome | Level | Activity |
|-----|---|-------|----------|
| CO1 | Design systems to perform NLP preprocessing and document | K6 | 1 - 5 |
| COI | similarity | | |
| CO2 | Design NLP systems for spam filtering | K6 | 6 |
| CO3 | Design NLP systems for sentiment analysis | K6 | 7 |
| CO4 | Design NLP systems using tagging and named entity recognition | K6 | 8 - 10 |
| CO5 | Design NLP systems using Context free grammars | K6 | 11 - 14 |
| CO6 | Design NLP systems using SpaCy | K6 | 15 |

2. SYLLABUS

| Activity | Lab Activity Description |
|----------|--|
| 1 | Understanding Large Text Files |
| 2 | Computing Bigram Frequencies |
| 3 | Computing Document Similarity using VSM |
| 4 | Computing Document Similarity using Word2Vec |
| 5 | Stemming and Lemmatization on Movie Dataset |
| 6 | Spam Filtering using Multinomial Naïve Bayes |
| 7 | Sentiment Analysis on Movie Reviews |
| 8 | Exploring Part of Speech Tagging on Large Text Files |
| 9 | Building Bigram Tagger |
| 10 | Named Entity Recognition on Food Recipes Dataset |
| 11 | Building Parse Trees |
| 12 | Building and Parsing Context Free Grammars |
| 13 | Improving Grammar to Parse Ambiguous Sentences |
| 14 | Word Sense Disambiguation with Improved Lesk |
| 15 | Text Processing using SpaCy |

Topics for Self Study

| S.No | Topic Title | Web Link |
|------|---------------------------------|---|
| 1 | Text preprocessing in languages | https://github.com/morkapronczay/meetup-talk-text- |
| | other than English | preproc |
| 2 | Cross-classification of | http://cl.haifa.ac.il/projects/translationese/index.shtml |
| | translationese | |
| 3 | Distinguishing between human | http://cl.haifa.ac.il/projects/pmt/index.shtml |
| | and machine translation | |
| 4 | Native Language Identification | https://github.com/ellarabi/reddit-12 |

| Activity# | Lab Activity | Learning Outcome | Level |
|-----------|-----------------------------|---------------------------------------|-------|
| 1 | Large Text Files Processing | Understand Large Text Files | K6 |
| 2 | Bigram Frequencies | Compute Bigram Frequencies | K6 |
| 3 | Document Similarity using | Compute Document Similarity using VSM | K6 |
| | VSM | | |

| 4 | Document Similarity using Word2Vec | Compute Document Similarity using Word2Vec | K6 |
|----|---|---|----|
| 5 | Stemming and | Perform Stemming and Lemmatization on Movie Dataset | K6 |
| 6 | Spam Filtering | Perform Spam Filtering using Multinomial Naïve Bayes | K6 |
| 7 | Sentiment Analysis | Develop system for Sentiment Analysis on Movie Reviews | K6 |
| 8 | Part of Speech Tagging on Large Text Files | Explore Part of Speech Tagging on Large Text Files | K6 |
| 9 | Bigram Tagger | Build Bigram Tagger | K6 |
| 10 | Named Entity Recognition | Perform Named Entity Recognition on Food Recipes Dataset | K6 |
| 11 | Parse Trees | Build Parse Trees | K6 |
| 12 | Context Free Grammars | Build and Parse Context Free Grammars | K6 |
| 13 | Parsing Ambiguous Sentences | Improve Grammar to Parse Ambiguous Sentences | K6 |
| 14 | Word Sense Disambiguation | Perform Word Sense Disambiguation with Improved Lesk | K6 |
| 15 | Text Processing using SpaCy | Perform text processing using SpaCy | K6 |

4. MAPPING (CO, PO, PSO)

L-Low

M-Moderate

H- High

| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
|------------|-----|-----|-----|-----|-----|-----|------------|-----|------------|------|------|------|------|
| CO1 | Н | Μ | L | Μ | Μ | Μ | L | - | - | Н | Μ | Μ | - |
| CO2 | Н | Μ | H | Μ | Μ | Н | Μ | Μ | L | Н | Н | - | Μ |
| CO3 | H | Μ | Μ | L | Н | Μ | - | - | L | Μ | Μ | Μ | - |
| CO4 | Н | L | H | Н | Н | Н | Μ | Μ | L | H | Μ | Μ | Н |
| CO5 | Н | Μ | L | Н | Н | L | L | Μ | Μ | Н | Н | Н | Μ |
| CO6 | Н | Μ | Μ | L | H | L | L | L | Μ | Н | Н | - | - |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Practical Components): Closed Book
- 2. Cooperative Learning Report, Assignment, Group Discussion, project Report, Field Visit Report, Seminar.
- 3. Pre/Post Test, Viva, Report for each Exercise.
- 4. Lab Model Examination & End Semester Practical Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator: Dr. Janani Selvaraj

| CORE VII: TIME SERIES ANALYSIS AND FORECASTING | | | | | |
|--|-----|------------|---|--|--|
| Semester | III | Hours/Week | 4 | | |

| | Course Code | P19DS307 | Credits | 4 |
|--|-------------|----------|---------|---|
|--|-------------|----------|---------|---|

On successful completion of this course, students will be able to:

| S.No. | Course Outcomes | Level | Unit |
|-------|---|-------|------|
| CO1 | Solve the stationarity, trending and detrending of time series data | K6 | 1 |
| CO2 | Assess the features of the ARMA Models and estimation techniques | K5 | 2 |
| CO3 | Explain the ARIMA models and SARMA Models | K6 | 3 |
| CO4 | Summarize the characteristics of Spectral behaviour and periodic behaviour of the time series | K6 | 4 |
| CO5 | Compile the behaviour of smoothing in DLMS | K6 | 5 |
| CO6 | Design the Timeseries models using R for different time series data | K6 | All |

4. A. SYLLABUS

UNIT I - BASIS TIME SERIES MODELS

Examples of Nature of Time series data – Time series statistical models – Measures of dependence - Stationary. Time series regression – Detrending and differencing – Smoothing a time series

UNIT II - AR MODELS, FORECASTING AND ESTIMATION

Auto Regressive models – Moving Average models - ARMA models – Auto Correlation Function - Partial Auto Correlation Function – Forecasting algorithms – **Estimation:** Yule-Walker, Method of moments, MLE and LSE

Unit III - ARMA AND GARMA MODELS

Basics of ARIMA models: random models with drift, Steps to fitting ARMA model – **Multiplicative Seasonal ARIMA models:** Mixed, SARMA – Generalized Auto Regressive Conditionally Heteroscedastic (GARCH) models

UNIT IV - SPECTRAL ANALYSIS HOURS

Cyclical Behaviour and Periodicity: concepts, Periodic Series, Star Magnitude - **The Spectral Density:** Periodic stationary process–Periodogram: Spectral analysis as ANOVA, Principal Component Analysis

UNIT V - STATE SPACE MODELS

Dynamic Linear Models – Examples of DLMs – Filtering DLM – **Smoothing DLM:** Kalman, Lag One covariance – Forecasting DLM – Maximum Likelihood Estimator for DLMs

S.No. Topics Web Links 1 Forecasting hierarchical or https://otexts.com/fpp2/hierarchical.html grouped time series 2 Autoregression Models for https://machinelearningmastery.com/\autoregressio Time Series Forecasting n-models-time-series-forecasting-python/ With Python Time Series ARIMA Model https://sites.google.com/site/econometricsacademy/ 3 using R econometrics-models/time-series-arima-models 4 Exponential Simple https://towardsdatascience.com/ Smoothing for Time Series simple-exponential-smoothing-749fc5631bed Forecasting

B. TOPICS FOR SELF - STUDY

12 HOURS

12 HOURS

12 HOURS

12 HOURS

12

C. TEXT BOOKS

1. Shumway and Stoffer. Time Series Analysis and its applications, with examples in R. 4ed, Springer. 2016.

D. REFERENCES BOOKS

- 1. Brockwell& Davis. Introduction to Time Series and Forecasting, 3rd edition, Springer. 2016
- 2. Cryer& Chan. Time Series Analysis with Applications in R, Springer. 2008
- 3. Prado & West. Time Series: Modeling, Computation, and Inference Chapman & Hall. 2010
- 4. Petris, Petrone, Campagnoli. Dynamic Linear Models with R, Springer. 2009
- 5. Ruppert& Matteson. Statistics and Data Analysis for Financial Engineering with R examples, 2ed, Springer. 2016

E. WEB LINKS

- 4. https://machinelearningmastery.com/autoregression-models-time-series-forecasting-python/
- 5. https://sites.google.com/site/econometricsacademy/econometrics-models/time-series-arima-models

| Unit/ Section | Course Content | Learning outcomes | Level | | |
|------------------|---|--|-------|--|--|
| Ι | BAS | SIS TIME SERIES MODELS | | | |
| 1.1 | Nature of Time series data | Analyze the different types of Times series and its characteristics. | K4 | | |
| 1.2 | Time series statistical models | Compare the different statistical model of times series data. | K6 | | |
| 1.3 | Measures of dependence | Assess the measure of dependence for different statistical model of time series data. | K5 | | |
| 1.4 | Stationary | Evaluate the stationarity property for time series models. | K5 | | |
| 1.5 | Time series regression | Investigate the regression for time series data | K4 | | |
| 1.6 | Detrending and differencing | Formulate the detrending model for Time series data. Devise the differencing method for time series data. | K6 | | |
| 1.7 | Smoothing a time series | Construct the smoothing filters for time series models | K6 | | |
| II | AR MODELS | , FORECASTING AND ESTIMATION | | | |
| 2.1 | Auto Regressive models | Compare the features of AR Models | K4 | | |
| 2.2. | Moving Average models | Analyze the characteristics of MA Models | K5 | | |
| 2.3 | ARMA models | Summarize the working methods of ARMA Models | K6 | | |
| 2.4 | Auto Correlation Function - Partial Auto Correlation Function | Explain the role of Autocorrelation and partial auto correlation function for time series | K6 | | |
| 2.5 | Forecasting algorithms | Compile the forecasting algorithm for time series data. | K6 | | |
| 2.6 | Estimation: Yule-Walker, Method of moments, MLE and LSE | Specify the features of different estimation algorithms of time series data. | K6 | | |
| III | ARM | MA AND GARMA MODELS | | | |
| 3.1 | Basics of ARIMA models: random models with drift, Steps to fitting ARMA model | Explain the basics of ARIMA models of Time series data | K3 | | |

| | Multiplicative Seasonal | Compile the features of the SARIMA and | | |
|-----|--------------------------------|---|----|--|
| 3.2 | ARIMA models: Mixed, | Multiplicative SARIMA model for time series | K6 | |
| | SARMA | data. | | |
| | Generalized Auto Regressive | | | |
| 3.3 | Conditionally Heteroscedastic | Evaluate the characteristics of GARCH Model. | K5 | |
| | (GARCH) models | | | |
| IV | SPE | CTRAL ANALYSIS | | |
| | Cyclical Pahaviour and | 1. Examine the concepts of periodicity | K3 | |
| 4.1 | Cyclical Bellaviour and | 2. Evaluate the cyclical behaviour of the time | K6 | |
| 4.1 | Series Stor Magnitude | series. | | |
| | Selles, Stal Magilitude | 3. Assess the properties of Star Magnitude | K5 | |
| 12 | Periodic stationary process | Discriminate the periodic stationary process | K2 | |
| 4.2 | renould stationary process | over the stationary process | КJ | |
| 4.3 | Periodogram | Outline the periodogram of the time series data | K6 | |
| 4.4 | Spectral analysis of ANOVA | Construct the ANOVA for the spectral analysis | V6 | |
| 4.4 | Spectral analysis as ANOVA | of Time series data | K0 | |
| 15 | Spectral analysis as Principal | Develop the PCA for the spectral analysis of | | |
| 4.5 | Component Analysis | time series data. | KU | |
| V | 5 | STATE SPACE MODELS | | |
| 5.1 | Dynamic Linear Models- | Illustrate the Dynamic Linear Models (DLM) | K3 | |
| | Examples of DLMs | | | |
| 5.2 | Filtering DLM | Create the filtering of DLM for the time series | K6 | |
| | | data | | |
| 5.3 | Smoothing DLM: Kalman | Evaluate the role of Kalman Filter in | K5 | |
| - | Filter | smoothing. | | |
| 5.4 | Lag One covariance | Construct the smoothing filter using Lag One | K6 | |
| - | | covariance | | |
| 5.5 | Forecasting DLM | Design the forecasting algorithm using Kalman | K6 | |
| | | THE IOT DLMS | | |
| 5.6 | Maximum Likelihood | Evaluate the features of the MLE for DLMs | K6 | |
| | Estimator for DLMs | | | |

6. MAPPING

L-Low

M-Moderate

H- High

| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
|------------|--------------|-----|-----|-----|-----|------------|------------|------------|------------|------|------|------|------|
| CO1 | Η | Μ | Η | Н | Η | Η | - | - | Μ | Н | Н | Μ | - |
| CO2 | \mathbf{M} | Η | Μ | Η | Μ | Μ | - | Μ | Μ | Μ | Μ | - | - |
| CO3 | Η | Μ | Η | Η | Η | Η | - | - | Η | Н | Μ | - | - |
| CO4 | Η | Η | L | Η | Μ | Η | - | Μ | Η | Н | Η | - | |
| CO5 | Η | Μ | Η | Μ | Η | Η | I | - | Μ | Η | Η | - | - |
| CO6 | Η | Η | Η | Μ | Μ | Η | Η | - | Μ | Н | Η | - | - |

7. COURSE ASSESSMENT METHODS

DIRECT:

- 5. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 6. Open Book Test.
- 7. Peer Learning Report, Assignment, Group Presentation, Group Discussion, project Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).
- 8. Pre-Semester & End Semester Theory Examination

INDIRECT:

- 3. Course evaluation survey
- 4. Faculty feedback about the course.

Name of the Course Coordinator: Dr. P. S. Eliahim Jeevaraj

| CORE VIII: BIG DATA MANAGEMENT AND ANALYTICS | | | | | | |
|--|----------|------------|---|--|--|--|
| Semester | III | Hours/Week | 4 | | | |
| Course Code | P20DS308 | Credits | 4 | | | |

1. Course Outcomes

On successful completion of this course, students will be able to

| S.No. | Course Outcomes | Level | Unit |
|-------|--|-------|------|
| CO1 | Perceive Big Data concepts and technologies | K6 | Ι |
| CO2 | Evaluate the Storing and manipulation of data using HDFS | K6 | II |
| CO3 | Construct the very large datasets using Pig | K6 | III |
| CO4 | Create MapReduce using Spark | K6 | IV |
| CO5 | Formulate Data Warehousing operations using Hive | K6 | V |
| CO6 | Create applications using Hadoop | K6 | All |

2. A. SYLLABUS

Unit-1. Introduction to Big Data

What is Big data – Industrial examples of Big Data: Digital Marketing, fraud, risk, trading, healthcare, medicine, advertising – Big Data Technology: Hadoop, cloud, BI, crowdsourcing analytics – Business Analytics:

Unit-2. MapReduce-I and HDFS

MapReduce model: Weather dataset, Analyzing data with Hadoop, Combiner functions, Hadoop streaming with Python. **Hadoop Distributed File System**: Block, Namenode, Datanode, Caching – File system operations in command line – Java Interface to Basic Hadoop - Reading data and writing data – Anatomy of File Write

Unit-3. MapReduce-II

Steps of developing MapReduce application - Working of MapReduce: Running Jobs, failure, Shuffle and sort, Task execution - MapReduce Types: Input formats - Output formats - MapReduce features: Counters, Sorting, Joins

Unit-IV. Exploring large datasets using Pig

Structure, Statements, Expressions, Types, Schemas, Functions, Macros - User-Defined Functions: Filter UDF, Eval UDF, Load UDF - Data Processing Operators: Loading and Storing Data, Filtering Data, Grouping and Joining Data, Sorting Data, Combining and Splitting Data

Unit-5. Data Warehousing using Hive

Comparison with Traditional Databases - HiveQL: Data Types, Operators and Functions - Tables: Managed Tables and External Tables, Partitions and Buckets, Storage Formats, Importing Data, Altering Tables, Dropping Tables - Querying Data: Sorting and Aggregating, MapReduce Scripts, Joins, Subqueries, Views - User-Defined Functions: Writing a UDF, Writing a UDAF - 6 Elements of Big Data Security

B. TOPICS FOR SELF-STUDY

C. TEXT BOOKS

- Michael Minelli, Michele Chambers and Ambiga Dhiraj. Big Data, Big Analytics: Emerging Business Intelligence and Analytic Trends for Today's Businesses, 1ed, Wiley CIO Series, 2013. ISBN 9781118147603
- 2. Tom White Hadoop: The Definitive Guide, Fourth Edition, O'reilly Media, 2015.

3. Six Elements of Securing Big Data. MapR Ebook https://mapr.com/big-data-security-6-elements/

REFERENCES

- 1. Nathan Marz and James Warren, Big Data Principles and Practice of Scalable Real Time Data Systems, Manning Publications. 2015
- 2. Michael Berthold, David J. Hand, Intelligent Data Analysis, Springer, 2007
- 3. Bill Franks, Taming the Big Data Tidal Wave: Finding Opportunities in Huge Data Streams with Advanced Analytics, John Wiley& sons, 2012.
- 4. Glenn J. Myatt, Making Sense of Data, Volume I and II. John Wiley & Sons, 2007.
- 5. Mark Grover, Ted Malaska, Jonathan Seidman, Gwen Shapira. Hadoop Application Architecture, Shroff Publishers.2015
- 6. Chris Eaton, Dirk DeRoos, Tom Deutsch, George Lapis, Paul Zikopoulos, Understanding Big Data: Analytics for Enterprise Class Hadoop and Streaming Data, McGrawHill Publishing, 2012.

| Unit | Topic Topic Learning Outcome | | Level |
|------|---|--|-------|
| Ι | Introduction to Bi | g Data | |
| 1.1 | What is Big data | Why need to handle big dataset | K1 |
| 1.2 | Industrial examples of Big Data: Digital Marketing | Outline big data in marketing | K2 |
| 1.3 | fraud, risk management | List the usage of the big data in fraud, risk management. | K1 |
| 1.4 | trading | Explain time to time data analysis | K1 |
| 1.5 | healthcare, | Why big data in healthcare | K1 |
| 1.6 | medicine | How big data is in medicine | K1 |
| 1.7 | advertising | How to advertise with big data | K1 |
| 1.8 | Big Data Technology: Hadoop | Discover storing data and running applications on clusters of commodity hardware. | K4 |
| 1.9 | Cloud | Build on demand services using internet. | K3 |
| 1.10 | BI | Survey information retrieval from available huge amount of data | K4 |
| 1.11 | crowdsourcing analytics | Develop outsourcing for quality and to handle large amount of data | K3 |
| 1.12 | Business Analytics | Build statistical report by business analytics. | K3 |
| | 1 | | |
| II | MapReduce-I and | HDFS | T |
| 2.1 | MapReduce model: Weather dataset | Create combined report for weather from shuffler. | K6 |
| 2.2 | Analyzing data with Hadoop | Analyze a huge collection of data that comprises both structured data found in traditional databases and unstructured data like text documents, video and audio. | K4 |
| 2.3 | Combiner functions | Assess an optional class that operates by accepting the inputs from the Map class and thereafter passing the output key-value pairs to the Reducer class. | K5 |
| 2.4 | Hadoop streaming with Python. | Create stream using programming language that can read from standard input and write to standard output. | K6 |

| 2.5 | Hadoop | Create sequence of blocks from file which is to store. | K6 |
|------|--------------------|---|------------|
| | Distributed File | | |
| | System: Block | | |
| 26 | Namanada | Develop the conterpiece of an UDES file system. It | V2 |
| 2.0 | Namenoue | bevelop the centerpiece of all files in the file system, and | КJ |
| | | keeps the directory free of all files in the file data is heart. It | |
| | | tracks where across the cluster the file data is kept. It | |
| | | does not store the data of these files itself. | |
| 2.7 | Datanode | Build a DataNode stores data in the [Hadoop File | K3 |
| | | System]. A functional filesystem has more than one | |
| | | DataNode, with data replicated | |
| 2.8 | Caching | Plan the Centralized cache management which is an | K3 |
| | | explicit caching mechanism that allows users to specify | |
| | | paths to be cached by HDFS. | |
| 2.9 | File system | Make use of HDFS command for import file and | K3 |
| | operations in | mapreduce | |
| | command line | 1 | |
| 2.10 | Java Interface to | Adapt java interface for implement FileSystem | K6 |
| | Basic Hadoop | represents like client interface to a filesystem in Hadoop. | |
| | 2 usie linus op | and there are several concrete implementations | |
| 2.11 | Reading data and | Criticize $WOR \Delta$ (Write once Read many) models | K5 |
| 2.11 | writing data | Children work (whice field hearly) models | KJ |
| 2.12 | A notomy of Filo | Define file formet | V 1 |
| 2.12 | Milita Write | Define me format | KI |
| | white | | |
| | | | |
| III | MapReduce-II | | I |
| 3.1 | Steps of | Divide MapReduce as three stages, namely map stage, | K4 |
| | developing | shuffle stage, and reduce stage. | |
| | MapReduce | | |
| | application | | |
| 3.2 | Working of | Create mapper's jobs to process the input data. | K6 |
| | MapReduce: | | |
| | Running Jobs | | |
| 3.3 | failure | Determine TaskTracker to marks the task when failed. | K5 |
| 3.4 | Shuffle and sort | Create shuffler for transfer mapper intermediate output to | K6 |
| 5.7 | Shame and sort | the reducer | N O |
| 3.5 | Task execution | Create task from ManPaduce for parallel processing | K6 |
| 3.5 | Man Daduaa | Select different input format for block | K0 V5 |
| 5.0 | | Select different input format for block. | КJ |
| | Types: Input | | |
| 0.5 | formats | | |
| 3.7 | Output formats | Select different output format for Shuffler and reducer. | K5 |
| 3.8 | MapReduce | Measure occurrences of any events. | K5 |
| | features: Counters | | |
| 3.9 | Sorting | Build sorting algorithm to automatically sort the output | K6 |
| | | key-value pairs from the mapper by their keys. | |
| 3.10 | Joins | Select join for map the partitioned and sorted according | K5 |
| | | to the keys. | |
| | · · | · · · · · · | |
| IV | Exploring large da | atasets using Pig | |
| 4.1 | Structure | Use pig structure for data processing | K1 |
| 4.2 | Statements | List nig statements for data processing | K1 |
| 43 | Expressions | Use Pig expression to manipulate data | K1 |
| 1.5 | | Examine four types of data model | K/ |
| 4.4 | L'abarras | Define detect scheme | K4 V1 |
| 4.3 | Schemas, | Denne dataset schema. | NI NI |

| 4.6 | Functions | Use EVAL functions, Math functions, String functions | K1 |
|------|-------------------|---|------|
| | 2.6 | and Pig built-in functions for data processing. | TT c |
| 4.7 | Macros | Create the code modular and makes Pig Latin code shareable | K6 |
| 4.8 | User-Defined | Create UDF for conditions in filter statements in data | K6 |
| | Functions: Filter | processing and return Boolean value. | |
| | UDF | | |
| 4.9 | Eval UDF | Create UDF for FOREACH-GENERATE in data processing | K6 |
| 4.10 | Load UDF | Create UDF Load function top on Hadoop for | K6 |
| | | InputFormat to read data. | |
| 4.11 | Data Processing | Elaborate Load Operator and Store Operator for Reading | K6 |
| | Operators: | and Storing Data. | |
| | Loading and | | |
| | Storing Data | | |
| 4.12 | Filtering Data | Select the required tuples from a relation based on | K5 |
| | C | 'condition'. | |
| 4.13 | Grouping and | Make up cluster of data using group. | K6 |
| | Joining Data | Create Combine record using Join. | |
| 4.14 | Sorting Data | Create data in systematic order like ascending or | K6 |
| | C | descending order. | |
| 4.15 | Combining and | Select combine for join two or more relations. | K6 |
| | Splitting Data | Select Split to split two or more relations. | |
| | | | |
| V | Data Warehousin | g using Hive | |
| 5.1 | Comparison with | List difference between RDBMS and HIVE | K1 |
| | Traditional | | |
| | Databases | | |
| 5.2 | HiveQL: Data | Use Hive data types | K1 |
| | Types | | |
| 5.3 | Operators and | Recall Hive operations operators and functions for data | K1 |
| | Functions | storage | |
| 5.4 | Tables:Managed | Create Hive vertical table for manipulate data. | K6 |
| | Tables and | Describes the metadata / schema on external files using | |
| | External Tables | hive | |
| 5.5 | Partitions and | Create partitions and these partitions can be further | K6 |
| | Buckets | subdivided into more manageable parts known as | |
| | | Buckets or Clusters. | |
| 5.6 | Storage Formats | Compose storage format for input block from HDFS | K6 |
| 5.7 | Importing Data | Create a directory in HDFS to hold the file and import | K6 |
| | | CSV files into Hive tables. | |
| 5.8 | Altering Tables | Change the existing table like table name, column name, | K6 |
| | | comment, and table properties. | |
| 5.9 | Dropping Tables | Delete the table/column data and their metadata | K6 |
| 5.10 | Querying Data: | Create Querying data for sorting using Order By | K6 |
| | Sorting and | Create aggregate using AVG, SUM, or MAX functions. | |
| | Aggregating | | |
| 5.11 | MapReduce | Create Hive script using gedit for MapReduce | K6 |
| | Scripts | | |
| 5.12 | Joins | Select query for join two or more tables | K5 |
| 5.13 | Subqueries | Create a subquery for evaluated and returns a result set. | K6 |
| 5.14 | Views | Evaluate user requirements by generating views. | K6 |
| 5.15 | User-Defined | create custom functions to process records or groups of | K6 |

| | Functions: | records | |
|------|------------------------------------|---|----|
| | Writing a UDF | | |
| 5.16 | Writing a UDAF | create custom Aggregate functions to process records or groups of records | K6 |
| 5.17 | 6 Elements of Big Data Security | Justify big data security by the use of six steps. | K5 |

4. Mapping

| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
|------------|------------|-----|-----|-----|-----|-----|------------|-----|------------|------|------|------|------|
| CO1 | Η | Μ | Η | Η | Η | Η | - | - | Μ | Н | Н | Μ | - |
| CO2 | Μ | Η | Μ | Η | Μ | Μ | - | Μ | Μ | Μ | Μ | - | - |
| CO3 | Η | Μ | Η | Η | Η | Η | - | - | Η | Η | Μ | - | - |
| CO4 | Η | Η | L | Η | Μ | Η | - | Μ | Η | Н | Н | - | |
| CO5 | Η | Μ | Η | Μ | Η | Η | - | - | Μ | Η | Η | - | - |
| CO6 | Η | Η | Η | Μ | Μ | Η | Η | - | Μ | Н | Н | - | - |

5. COURSE ASSESSMENT METHODS DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 2. Open Book Test.
- 3. Cooperative Learning Report, Assignment, Group Presentation, Group Discussion, project Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).
- 4. Pre-Semester & End Semester Theory Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator: Dr. B. Karthikeyan

| Semester | III | Hours/Week | 4 |
|-------------|----------|------------|---|
| Course Code | P19DS309 | Credits | 4 |

1. Course Outcomes

On successful completion of this course, students will be able to:

| CO# | Course Outcome | Level | Unit |
|-----|--|-------|------|
| CO1 | Explain the essentials of graphs for social networks | K6 | Ι |
| CO2 | Measure social network nodes and simulate social network models | K6 | II |
| CO3 | Evaluate the community analysis of social networks | K6 | III |
| CO4 | Measure and model information diffusion and homophily in social | K6 | IV |
| C04 | networks | | |
| CO5 | Develop recommender systems and predict user behaviours | K6 | V |
| CO6 | Build and deploy end to end products into production environment | K6 | All |

2. A. SYLLABUS

Unit-1. Introduction to SMM and Graph Mining

What is social media mining – New challenges for mining. Graph Essentials: Graph basics – Graph representation – Types of graphs – Connectivity in graphs – Special graphs – Graph algorithms

Unit-2. Social Network Models

Network Measures: Centrality – Transitivity, reciprocity – Balance and status – Similarity. Network Models: Properties – Random graphs – Small world models – Preferential attachment model

Unit-3. Data Mining Basics and Community Discovery

Data Mining Essentials: Data Preprocessing – Supervised Learning Algorithms – Unsupervised Learning Algorithms. Community Analysis: Community detection – Community evolution – Community evaluation

Unit-4. Information Diffusion and Influence in Social Media

Information Diffusion: Herd behaviour – Information cascades – Diffusion of innovations – Epidemics. Influence and Homophily: Measuring Assortativity – Measuring and modelling influence – Measuring and modelling homophily – Distinguishing influence and homophily

Unit-5. Recommendation and Behaviour Analysis in Social Media

Recommendation in Social Media: Challenges – Classical recommendation algorithms – Recommendation using social context – Evaluating recommendations. Behaviour Analysis: Individual behaviour – Collective behaviour. Events Analytics in Social Media.

| SNo | Topic Title | Web Link |
|-----|---------------------------|--|
| 1 | Creating graphs using | https://www.analyticsvidhya.com/blog/2018/09/introduction- |
| | NetworkX for Airline data | graph-theory-applications-python/ |
| | set | |
| 2 | Implementation of Movie | https://www.geeksforgeeks.org/python-implementation-of- |
| | recommender system | movie-recommender-system/ |
| 3 | Diving into GraphQL and | https://medium.com/elements/diving-into-graphql-and-neo4j- |
| | Neo4j with Python | with-python-244ec39ddd94 |
| 4 | DataCamp Network | https://www.datacamp.com/courses/introduction-to-network- |
| | analysis using Python | analysis-in-python |

B. TOPICS FOR SELF STUDY

C. TEXT BOOK(S)

1. Reza Zafarani, Mohammad Ali Abbasi, and Huan Liu. Social Media Mining: An Introduction, Cambridge University Press, 2014

D. REFERENCES BOOKS

- 1. Matthew A. Russell. Mining the Social Web. 3rd Edition. O'Reilly Media. 2019
- Jennifer Golbeck. Analyzing the Social Web. Morgn Kaufmann. 2013. ISBN 978-0124055315
- Ricardo Baeza-Yates and BerthierRibeiro-Neto. Modern Information Retrieval: The Concepts and Technology behind Search. 2ed. ACM Press Books, 2011. ISBN 978-0321416919
- 4. Charu C. Aggarwal. Social Network Data Analytics. Springer. 2011

E. WEB LINKS

- https://medium.com/elements/diving-into-graphql-and-neo4j-with-python-244ec39ddd94
- https://www.geeksforgeeks.org/python-implementation-of-movie-recommender-system/

| Unit | Fopic Topic Learning Outcomes | | | | | | |
|------|--|--|----------|--|--|--|--|
| Ι | Introduction to SMM and G | raph Mining | | | | | |
| 1.1 | Challenges and methodologies for mining | Identify challenges and methodologies for social media mining | | | | | |
| 1.2 | Types of SM and marketing opportunities that exist in SM | List social media types and identify marketing opportunities | К2 | | | | |
| 1.3 | Graph basics | Compute degree and degree distribution of directed and undirected graphs | K3 | | | | |
| 1.4 | Graph representation | Find adjacency list and edge list Given these lists, create a graph | K4 K6 | | | | |
| 1.5 | Types of graphs | Given business problem, create weighted and signed graphs for the social network | K6 | | | | |
| 1.6 | Connectivity in graphs | Find all connectivity in graphs and compute diameter | K4 | | | | |
| 1.6 | Special graphs | Create various special graphs such as MST, Steiner tree, planner graph, bipartite graph and regular graph for the given business problem | K6 | | | | |
| 1.8 | Graph traversals | Apply BFS and DFS traversal methods for the given social network | K3 | | | | |
| 1.9 | Shortest path algorithms | Compute shortest paths using Dijkstra's and Prim's algorithms based on the business use case | K3 | | | | |
| 1.10 | Network flow algorithms | Analyze maximum messages a social network can handle. Analyze maximum matching between products and users. | K4 K4 | | | | |
| II | Social Network Measures ar | nd Models | | | | | |
| 2.1 | Centrality measures | Apply centrality measures and predict the most central important nodes from social networks | K6 | | | | |
| 2.2 | Transitivity | Apply transitivity measures and analyse linking behaviour of nodes | K4 | | | | |
| 2.3 | Reciprocity | Analyze reciprocity of the given social network | K4 | | | | |
| 2.4 | Balance and status | Determine consistency of relationship in signed graphs | K6 | | | | |
| 2.5 | Similarity measures | Apply similarity measures and predict similar nodes | K6 | | | | |

| | | in a social network | |
|------|---|---|----------------------|
| 2.6 | Properties of real world networks | Discuss the properties of real world networks | K6 |
| 2.7 | Random graph model | Discuss the types, evolution and properties of random graph model | K6 |
| 2.8 | Small world model | Discuss the properties of small world model Compare the properties of random graph and small world models | K6 |
| 2.9 | Preferential attachment model | Discuss the properties of small world model Compare the functionalities of random graph, small world and preferential attachment models Figure out the differences between random graphs, regular lattices, and small-world models | K6 K4 K4 |
| III | Data Mining Basics and Cor | nmunity Discovery | 1 |
| 3.1 | Data pre-processing steps | Given a business problem, identify various features Explain data pre-processing steps, given a use case Select a ML methodology based on the given problem scenario | K1 K2 K4 |
| 3.2 | Decision tree learning | Create a decision tree given a dataset representing the use case | K6 |
| 3.3 | Naïve bayes classifier | Create a Naïve bayes classifier given a dataset representing the use case | K6 |
| 3.4 | Nearest neighbour classifier | Create a KNN classifier given a dataset representing the use case | K6 |
| 3.5 | Supervised learning evaluation methods | Compare the evaluation measures for supervised ML classifiers | K4 |
| 3.6 | KMeans clustering | Predict clusters using KMeans given an use case | K6 |
| 3.7 | Unsupervised learning evaluation methods | Evaluate quality of clusters from unsupervised ML classifiers | K6 |
| 3.8 | Member based community detection | Detect communities by applying node similarity, node degree and node reachability methods | K4 |
| 3.9 | Group based community detection | Discuss the methods to detect group communities from social networks | K2 |
| 3.10 | Community evolution | Explain how communities evolve over time in social networks | K2 |
| 3.11 | Community evaluation | Given members of communities, analyze precision, recall, Fscore, purity and NMI measures | K4 |
| IV | Information Diffusion and I | nfluence | |
| A 1 | Information associates | Given a network with activation probabilities, analyse final set of activated nodes using ICM method | K4 |
| 4.1 | information cascades | Describe the independent cascade model Explain the objectives of cascade maximization | K2 K4 |
| 4.2 | Diffusion of innovations | Compare innovation diffusion models | K4 |
| 4.3 | Epidemics | Discuss the mathematical relationship between the SIR and the SIS models Defend why in SIR model, the probability that an individual remains infected follows a standard exponential distribution Compute in SIRS model, the length of time that an infected individual is likely to remain infected before | K6 K6 K3 K4 |
| | | he or she recovers Given a business or societal problem, select the | K6 |

| | | appropriate information diffusion model | |
|------|---|---|----------|
| | | Sumarize intervention approaches for information | |
| | | diffusion models | |
| 4.4 | Measuring Assortativity | Compute assortativity for ordinal and nominal attributes of social network nodes | K3 |
| 4.5 | Measuring influence | Illustrate the types of influence measures in blogosphere and twitter | K4 |
| 4.6 | Modelling influence | Select all activated nodes with Linear Threshold Model | K4 |
| 4.7 | Measuring homophily | Estimate homophily for nominal and ordinal attributes in a social network | K6 |
| 4.8 | Modelling homophily | Explain the variation of independent cascade model to model homophily | K4 |
| 4.9 | Distinguishing influence and homophily | Determine the source of assortativity in social networks | K6 |
| V | Recommendation and Behav | viour Analysis | |
| 5.1 | Challenges of recommender systems | Describe the challenges of recommendation systems | K1 |
| 5.2 | Content based recommendation system | Differentiate content-based recommendation from collaborative filtering | K4 |
| 5.3 | User based collaborative filtering | Predict missing ratings using user based CF | K6 |
| 5.4 | Item based collaborative filtering | Predict the most similar items using item based CF | K6 |
| 5.5 | Model based collaborative filtering | Predict ratings and items using SVD based CF | K6 |
| 5.6 | Group based recommendation | Find and recommend items to group of users | K3 |
| 5.7 | Recommendation using social context | Predict ratings and items leveraging social context Provide examples where social context can help improve classical recommendation algorithms in social media | K6 K3 |
| 5.8 | Evaluation of recommender systems | Evaluate the accuracy of predictions | K5 |
| 5.9 | Evaluating relevancy of recommendation | Evaluate the relevancy of recommendations | K5 |
| 5.10 | Evaluating ranking of recommendation | Evaluate the ranking of recommendations | K5 |
| 5.11 | Individual behaviour analysis, modelling and prediction | List the features for User Community-Joining Behavior Explain the methods for predicting individual behaviours | K1 K4 |
| 5.12 | Collective behaviour analysis, modelling and prediction | Outline a method for predicting Box office Revenue for Movies | K5 |

4. MAPPING

| PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
|------------|-----|-----|------------|-----|-----|------------|------------|-----|------|------|------|------|
|------------|-----|-----|------------|-----|-----|------------|------------|-----|------|------|------|------|

| CO1 | Η | Μ | Η | Η | Η | Η | - | - | Μ | Η | Η | Μ | - |
|------------|---|---|---|---|---|---|---|---|---|---|---|---|---|
| CO2 | Μ | Η | Μ | Η | Μ | Μ | - | Μ | Μ | Μ | Μ | - | - |
| CO3 | Η | Μ | Η | Η | Η | Η | - | - | Η | Η | Μ | - | - |
| CO4 | Η | Η | L | Η | Μ | Η | - | Μ | Η | Η | Η | - | |
| CO5 | Η | Μ | Η | Μ | Η | Η | - | - | Μ | Η | Η | - | - |
| CO6 | Η | Η | Η | Μ | Μ | Η | Η | - | Μ | Η | Η | - | - |

5. COURSE ASSESSMENT METHODS DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 2. Open Book Test.
- 3. Cooperative Learning Report, Assignment, Group Presentation, Group Discussion, project Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).
- 4. Pre-Semester & End Semester Theory Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator: Dr. M. Lovelin Pon Felciah

| ELECTIVE IV: IMAGE AND VIDEO ANALYTICS | | | | | | | | |
|--|----------|------------|---|--|--|--|--|--|
| Semester | III | Hours/Week | 4 | | | | | |
| Course Code | P19DS3:4 | Credits | 4 | | | | | |

After the successful completion of this course the students will be able to

| S.No. | Course Outcomes | Level | Unit |
|-------|---|-------|------|
| CO1 | Elaborate the fundamental principles of image and video analysis | K5 | Ι |
| CO2 | Choose the mathematical basic equation to transform images into different domain for performing smoothing and sharpening operations | K6 | П |
| CO3 | Evaluate a statistical model to solve Image Enhancement, Segmentation and Compression problems | К | III |
| CO4 | Select most relevant information from the original image to construct a feature vector such as texture, color and shape | K6 | IV |
| CO5 | Design suitable Classifier for Object Detection, Tracking and Recognition | K5 | IV |
| CO6 | Decide suitable image and video analysis approaches for developing solutions to solve real time applications | K6 | V |

2. A. SYLLABUS

Unit-1 Image Representation and Processing

Digital image representation- Visual Perception- Sampling and Quantization- Basic Relations between Pixels- Mathematical Tools Used in Digital Image Processing: Fundamental Operations – Vector and Matric Operations- Image Transforms (DFT, DCT, DWT, Hadamard).

Unit-2 Image Filtering

Fundamentals of spatial filtering: spatial correlation and convolution-smoothing, blurringsharpening- edge detection - Basics of filtering in the frequency domain: smoothing-blurringsharpening--Histograms and basic statistical models of image.

Unit-3 Colors and Compression

Color models and Transformations – Image and Video segmentation-Image and video demonising-Image and Video enhancement- Image and Video compression.

Unit-4 Object Detection and Tracking

Object detection and recognition in image and video-Texture models Image and Video classification models- Object tracking in Video.

Unit-5 Applications

Applications and Case studies- Industrial- Retail- Transportation & Travel- Remote sensing-Video Analytics in WSN: IoT Video Analytics Architectures.

B. TOPICS FOR SELF-STUDY

| S.No. | Topics | Web Links |
|-------|---------------------------------------|--|
| 1 | Pattern Recognition and Application | https://nptel.ac.in/courses/117/105/117105101/ |
| 2 | Practical Machine Leaning with Tensor | https://nptel.ac.in/courses/106/106/106106213/ |
| | Flow (Video) | |
| 3 | Object Representation and Description | https://www.youtube.com/watch?v=yxID4fgz1C0 |

|--|

C. TEXT BOOKS

1. R.C. Gonzalez and R.E. Woods. Digital Image Processing. 3rd Edition. Addison Wesley, 2007.

D. REFERENCES BOOKS

- 1. Pratt, W.K. Digital image processing: PIKS scientific inside. 4ed. New York: John Wiley, 2007.
- 2. W. Härdle, M. Müller, S. Sperlich, A. Werwatz. Nonparametric and Semi parametric Models. Springer, 2004.
- 3. Rick Szelisk. Computer Vision: Algorithms and Applications. Springer 2011.
- 4. Jean-Yves Dufour. Intelligent Video Surveillance Systems. Wiley, 2013.
- 5. Caifeng Shan, FatihPorikli, Tao Xiang, Shaogang Gong. Video Analytics for Business Intelligence. Springer, 2012.
- 6. AsierPerallos, Unai Hernandez-Jayo, Enrique Onieva, Ignacio Julio GarcíaZuazola. Intelligent Transport Systems: Technologies and Applications. Wiley, 2015.
- 7. BasudebBhatta. Analysis of Urban Growth and Sprawl from Remote Sensing Data. Springer, 2010

E. WEB LINKS

- 1. https://www.coursera.org/learn/digital
- 2. https://nptel.ac.in/courses/106/105/106105032

| Unit/ Section | Course Content | Learning outcomes | Level |
|------------------|--|---|-------|
| Ι | Image Representation and Processing | | |
| 1.1 | Digital image representation | Discuss the fundamental steps involved in Image processing system | K2 |
| | | Describe the image representation method | K2 |
| 1.2 | Visual Perception | Explain the human visual perception system with necessary diagrams. | K4 |
| | | Analyze the image formation takes place in eye and state the principle operation of brightness adaption and discrimination | K4 |
| 1.3 | Sampling and Quantization | Design the image digitization process by sampling and quantization | K5 |
| 1.4 | Basic relations between Pixels | Analyze the basic relationships between pixels | K4 |
| | | Distinguish the following terms: i) Adjacency ii) Connectivity iii) Region iv)Boundary | K4 |
| 1.5 | Mathematical Tools Used in Digital Image Processing | Examine the following mathematical operations on digital image i) Array versus Matrix operation ii) Linear versus Nonlinear Operations | K4 |
| 1.6 | Image Transforms (DFT, DCT, DWT, Hadamard) | Create a MATLAB script to construct the forward and inverse 2D DFT for the given image f(m,n) | K5 |
| | | Compare the following two properties of 2D-DFT | K2 |

| | | i) Convolution | |
|-----|-------------------------------------|---|----------------------------------|
| | | ii) Correlation | |
| | | Design the basis function of Haar | K5 |
| | | Transform for N=8 | - |
| | | Use the Hadamard kernel matrix 4 x 4 for | К3 |
| | | the image segment and perform transform | |
| | | with matrix multiplication method | |
| | | Construct 2D DCT for the image of size 2 | К5 |
| | | X 2 and verify the output after inverse | 110 |
| | | DCT | |
| | | Determine the approximation and detailed | K6 |
| | | coefficient of the Harr Transform which | 110 |
| | | takes an argument as $2 - dimensional$ | |
| | | digital signal 'S' | |
| | | Create discrete cosine transform(DCT) | K5 |
| | | matrix for N=4 | 113 |
| | | Design second-level decomposition of the | К5 |
| | | input image using a Haar wavelet | 113 |
| | | Construct the Haar transform $T - HFH^{T}$ | К5 |
| | | of the 2 x 2 image $F(m n)$ also find the | IX.5 |
| | | inverse Haar transform $\mathbf{F} - \mathbf{H}^{\mathrm{T}}\mathbf{T}\mathbf{H}$ of the | |
| | | obtained result | |
| | | Construct the subband modeling using | K5 |
| | | DWT | IX.5 |
| П | Image Filtering | | |
| 2.1 | Fundamentals of spatial | Analyse the impact of convolving the | K4 |
| 2.1 | filtering | image $f(x, y)$ with the mask $h(x, y)$ that | 11 1 |
| | intering | performs averaging operation which | |
| | | results in blurring the image | |
| | | Compare linear and Non-Linear spatial | К2 |
| | | Filtering Techniques | |
| | | Explain Image Negative and Log | К2 |
| | | transformation Techniques | 112 |
| | | d'ansionnation reeninques | |
| | | Determine the output image if $f(m, n)$ and | K6 |
| | | Determine the output image if $f(m, n)$ and $h(m, n)$ are linearly convolved with zero | K6 |
| | | Determine the output image if $f(m, n)$ and $h(m, n)$ are linearly convolved with zero padding of the original image | K6 |
| | | Determine the output image if f(m, n) and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain | K6 |
| | | Determine the output image if f(m, n) and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain low-pass filtering of the input image | K6 K4 |
| | | Determine the output image if f(m, n) and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain low-pass filtering of the input image using different window sizes like 3 × 3.5 | K6 K4 |
| | | Determine the output image if $f(m, n)$ and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain low-pass filtering of the input image using different window sizes like 3×3 , 5×5 and 7×7 | K6 K4 |
| | | Determine the output image if $f(m, n)$ and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain low-pass filtering of the input image using different window sizes like 3×3 , 5×5 and 7×7 Verify the effect of a 5×5 uniform | K6 K4 |
| | | Determine the output image if $f(m, n)$ and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain low-pass filtering of the input image using different window sizes like 3×3 , 5×5 and 7×7 Verify the effect of a 5×5 uniform averaging filter to a digital image N | K6 K4 K6 |
| | | Determine the output image if $f(m, n)$ and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain low-pass filtering of the input image using different window sizes like 3×3 , 5×5 and 7×7 Verify the effect of a 5×5 uniform averaging filter to a digital image N times. | K6 K4 K6 |
| 2.2 | Spatial correlation and | Determine the output image if $f(m, n)$ and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain low-pass filtering of the input image using different window sizes like 3×3 , 5×5 and 7×7 Verify the effect of a 5×5 uniform averaging filter to a digital image N times. Formulate the 2D linear convolution | K6 K4 K6 K5 |
| 2.2 | Spatial correlation and convolution | Determine the output image if $f(m, n)$ and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain low-pass filtering of the input image using different window sizes like 3×3 , 5×5 and 7×7 Verify the effect of a 5×5 uniform averaging filter to a digital image N times. Formulate the 2D linear convolution between the signal x(m, n) and h(m,n) and | K6 K4 K6 K5 |
| 2.2 | Spatial correlation and convolution | Determine the output image if $f(m, n)$ and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain low-pass filtering of the input image using different window sizes like 3×3 , 5×5 and 7×7 Verify the effect of a 5×5 uniform averaging filter to a digital image N times. Formulate the 2D linear convolution between the signal x(m, n) and h(m,n) and comment on the observed result. | K6 K4 K6 K5 |
| 2.2 | Spatial correlation and convolution | Determine the output image if $f(m, n)$ and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain low-pass filtering of the input image using different window sizes like 3×3 , 5×5 and 7×7 Verify the effect of a 5×5 uniform averaging filter to a digital image N times. Formulate the 2D linear convolution between the signal $x(m, n)$ and $h(m,n)$ and comment on the observed result. Compare the following properties of two- | K6 K4 K6 K5 K2 |
| 2.2 | Spatial correlation and convolution | Determine the output image if $f(m, n)$ and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain low-pass filtering of the input image using different window sizes like $3 \times 3, 5 \times 5$ and 7×7 Verify the effect of a 5×5 uniform averaging filter to a digital image N times. Formulate the 2D linear convolution between the signal $x(m, n)$ and $h(m,n)$ and comment on the observed result. Compare the following properties of two- dimensional convolution | K6 K4 K6 K5 K2 |
| 2.2 | Spatial correlation and convolution | Determine the output image if $f(m, n)$ and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain low-pass filtering of the input image using different window sizes like 3×3 , 5×5 and 7×7 Verify the effect of a 5×5 uniform averaging filter to a digital image N times. Formulate the 2D linear convolution between the signal $x(m, n)$ and $h(m,n)$ and comment on the observed result. Compare the following properties of two- dimensional convolution (i) Commutative property (ii) Associative | K6 K4 K6 K5 K2 |
| 2.2 | Spatial correlation and convolution | Determine the output image if $f(m, n)$ and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain low-pass filtering of the input image using different window sizes like 3×3 , 5×5 and 7×7 Verify the effect of a 5×5 uniform averaging filter to a digital image N times. Formulate the 2D linear convolution between the signal $x(m, n)$ and $h(m,n)$ and comment on the observed result. Compare the following properties of two- dimensional convolution (i) Commutative property (ii) Associative property (iii) Distributive property | K6 K4 K6 K5 K2 |
| 2.2 | Spatial correlation and convolution | Determine the output image if $f(m, n)$ and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain low-pass filtering of the input image using different window sizes like $3 \times 3, 5 \times 5$ and 7×7 Verify the effect of a 5×5 uniform averaging filter to a digital image N times. Formulate the 2D linear convolution between the signal $x(m, n)$ and $h(m,n)$ and comment on the observed result. Compare the following properties of two- dimensional convolution (i) Commutative property (ii) Associative property (iii) Distributive property Determine the correlation between the | K6 K4 K6 K2 K6 |
| 2.2 | Spatial correlation and convolution | Determine the output image if $f(m, n)$ and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain low-pass filtering of the input image using different window sizes like $3 \times 3, 5 \times 5$ and 7×7 Verify the effect of a 5×5 uniform averaging filter to a digital image N times. Formulate the 2D linear convolution between the signal $x(m, n)$ and $h(m,n)$ and comment on the observed result. Compare the following properties of two- dimensional convolution (i) Commutative property (ii) Associative property (iii) Distributive property Determine the correlation between the two image matrices $x1[m,n]$ and $x2[m,n]$ | K6 K4 K6 K2 K6 |
| 2.2 | Spatial correlation and convolution | Determine the output image if $f(m, n)$ and h(m, n) are linearly convolved with zero padding of the original image Examine the behavior of Spatial-domain low-pass filtering of the input image using different window sizes like 3×3 , 5×5 and 7×7 Verify the effect of a 5×5 uniform averaging filter to a digital image N times. Formulate the 2D linear convolution between the signal $x(m, n)$ and $h(m,n)$ and comment on the observed result. Compare the following properties of two- dimensional convolution (i) Commutative property (ii) Associative property (iii) Distributive property Determine the correlation between the two image matrices $x1[m,n]$ and $x2[m,n]$ | K6 K4 K6 K2 K6 K6 |

| | | performing two one dimensional | |
|-----|-----------------------------|---|------------|
| 2.2 | See a sthing hluming | Convolutions | VC |
| 2.3 | Smoothing, blurring | Justify the statement Mean filter is an | KO |
| | | noise through simple example | |
| | | Invest the new value of the $pixel(2,2)$ if | <i>V</i> 5 |
| | | Invent the new value of the pixel(2,2) if $\frac{1}{2}$ | KJ |
| | | shooting is done using a 5x5 | |
| | | a) Maan filter b) Weighted average filter | |
| | | a) Median filter d) Min and May filter | |
| | | C) Median Inter (1) Will and Max Inter | VC |
| | | Discuss the limiting effect of repeatedly | KO |
| | | digital image. Japare hander affacts. Is | |
| | | digital image. Ignore border effects. Is | |
| | | filter | |
| | | Inter | IZ A |
| | | Analyze 5 x 5 mean filter in the frequency | K 4 |
| | | domain and prove that it behaves like a | |
| | | low pass litter | VC |
| | | Determine the convolution process using $2x^2$ most in the particular of pixels $(2x^2)$ of | KO |
| | | sxs mask in the portion of pixels(2x2) of | |
| | | filtered image | |
| | | Marifa the area othing halosian of | VC |
| | | Verify the smoothing behavior of | Кб |
| | | Gaussian filter with varying levels of | |
| | | smooth factor σ . | K2 |
| | | Show the output impact in applying full- | К3 |
| | | scale contrast stretch to the image 4×4 , | |
| | | 4bits/pixel image. | |
| 2.4 | Sharpening- edge detection | Discuss the effect of first order derivative | K6 |
| | | gradient operators for image sharpening | |
| | | Invent the discontinuity in the image | K5 |
| | | using canny edge detector and give | |
| | | justification why it outperforms than | |
| | | gradient edge detectors and implement | |
| | | using MATLAB code | |
| | | Discuss the behaviour of the second order | K6 |
| | | Derivative for a step and ramp edges | |
| | | Construct the LOG filter to detect isolated | K3 |
| | | points and line in an image | TT / |
| 2.5 | Basics of filtering in the | Verify that convolution in spatial domain | K6 |
| | frequency domain: | is equal to multiplication in the frequency | |
| | smoothing-blurring- | domain using MATLAB code | |
| | sharpening | Use MATLAB code to perform a two- | K3 |
| | | dimensional Butterworth low-pass filter | |
| | | of the given image for two different cut- | |
| | | off frequencies | |
| | | Construct a filter for image smoothing in | К3 |
| | | trequency domain | |
| | | Explain the various high pass filters used | K4 |
| | | in frequency domain | |
| 2.6 | Histograms and basic | Justify your answer can two different | K6 |
| | statistical models of image | images have the same histogram | |
| | | Justify Histogram processing is called as | K6 |
|-----|-------------------------------|---|------------|
| | | an efficient tool for graphical | |
| | | representation of the total distribution in a | |
| | | given digital image. | |
| | | Design a statistical model of Histogram | K5 |
| | | Equalization to the given image by | |
| | | rounding the resulting image pixels to | |
| | | integers | |
| | | Determine the histogram equalization as | K6 |
| | | an idempotent operation for the 5 x 5 | RO |
| | | image segment. Plot the graph before and | |
| | | after equation | |
| | | Determine the visual appearance of the | V6 |
| | | resulting image by applying global | KU |
| | | histogram a suclimation for the group lovel | |
| | | nistogram equalization for the grey level | |
| | | image $f(x,y)$ of size 256x256 with | |
| | | 1 < x, y < 256, which has the following | |
| | | intensities $f(x,y) = r+1$ if $1 \le x, y \le 12$ and | |
| | | $f(x,y)=r$ if $13 \le x, y \le 16$, otherwise | |
| | | $\mathbf{f}(\mathbf{x},\mathbf{y}) = \mathbf{r} + 3.$ | |
| | | Determine the mean and standard | K6 |
| | | deviation of the image. If an image has | |
| | | gray levels ranging from 0 to 19. | |
| III | Colors and Compression | | 1 |
| 3.1 | Color models and | Describe the representation of three-color | K2 |
| | Transformations | components red, green and blue for the | |
| | | given color image. | |
| | | Classify the foreground and background | K3 |
| | | from the given RGB image and segment it | |
| | | using the Global thresholding method | |
| | | Verify the gamma correction for the given | K6 |
| | | color image for different values of gamma | |
| | | and comment on the output result. | |
| | | Formulate the CMY coordinates from the | K5 |
| | | given color image represented in terms of | |
| | | RGB components | |
| | | Construct a statistical model of histogram | K5 |
| | | equalization of the given RGB image | |
| | | Discuss the additive and subtractive color | K6 |
| | | model also implement Python OpenCV | 110 |
| | | code to extract color components | |
| | | Justify the result for the color transform | K6 |
| | | model to read the color image. Convert | K 0 |
| | | the RGB format to VIO format (NTSC) | |
| | | Filter only the V component (high pass | |
| | | filtering) Do not disturb the Lond O | |
| | | α components. Then convert the filtered V | |
| | | component I component and a | |
| | | component healt to the DCD format or d | |
| | | component back to the KGB format and | |
| | | check the result. | |
| | | $C_{\text{outothermal theorem }} = 1^{1} = 1^{1} = 1^{1} = 1^{1}$ | V5 |
| | | Construct the median filter for the color | K5 |
| | | Construct the median filter for the color image corrupted by salt-and-pepper noise | K5 |
| | | Construct the median filter for the color image corrupted by salt-and-pepper noise and try to restore the corrupted image | K5 |

| | segmentation | technique to segment the given image | |
|-----|-----------------|---|------|
| | | Discuss the Morphological operations | K2 |
| | | opening and closing for the given binary | |
| | | image | |
| | | Predict the number of black pixels in the | K5 |
| | | resultant image for the given input binary | |
| | | image if hit-or-miss transformation is | |
| | | performed with the structure element [0 1 | |
| | | 0,1 1 1,0 1 0]. | |
| | | Explain watershed segmentation tends to | K4 |
| | | over-segmentation problem in images. | |
| | | Mention the solution to overcome the | |
| | | problem | |
| | | Formulate the gradient magnitude and the | K5 |
| | | direction of the gradient for the pixel | |
| | | $f(x,y)=2x^2$. | |
| | | Determine the hit and miss transformation | K6 |
| | | of Morphological operator on a binary | |
| | | array that represents a portion of a black- | |
| | | and-white image and perform the | |
| | | operations on this piece of image Assume | |
| | | that all the pixels that surround this | |
| | | segment contain a black background | |
| | | Verify that the Prewitt edge detector along | K6 |
| | | a horizontal direction can be obtained by | 110 |
| | | convolving two one-dimensional signals | |
| | | $\begin{bmatrix} 1 & 1 \end{bmatrix}$ and $\begin{bmatrix} -1 & 0 \end{bmatrix}$ T and then scaling | |
| | | the result by a factor of $1/3$ | |
| | | Construct the linear filter masks for the | К5 |
| | | following operations: | 110 |
| | | (a) Detecting horizontal lines | |
| | | (b) Detecting vertical edges | |
| | | Distinguish between image segmentation | K4 |
| | | based on thresholding with image | 111 |
| | | segmentation based on region-growing | |
| | | techniques. | |
| | | Design an Automatic thresholding of grey | K5 |
| | | level image using otsu's thresholding | 110 |
| | | Discuss Multiple object segmentation in | K6 |
| | | video using Graph Cut | |
| | | Devise a technique to detect outlier from | K5 |
| | | motion segmentation in video | 110 |
| | | Apply the Region based split-and-merge | К3 |
| | | technique to segment the given image | iii) |
| 3.3 | Image and video | Discuss the tools available for image and | К2 |
| | demonising | video demonizing | |
| 3.4 | Image and Video | Explain the various video enhancement | K2 |
| | enhancement | techniques | |
| | | Distinguish the following enhancement | K4 |
| | | operations: i) Contrast stretching ii) Bit- | |
| | | plane slicing | |
| | | Analyze the behavior of piecewise linear | K4 |
| | | transformation and grev level transform | |
| | | for image enhancement | |

| | | Determine the output image $g(m, n)$ using | K6 |
|-----|----------------------------|--|------------|
| | | logarithmic transformation | 110 |
| | | $g(m n) = [clog_{10}(1+f(m n))]$ by choosing c | |
| | | $a_{1}(i) = 1$ and $c = \frac{1}{\log_{10}(1+1)}$ | |
| | | Examine the behavior of image arithmetic | КЗ |
| | | operation such as addition subtraction | ix. |
| | | multiplication and division over an image | |
| | | Judge the impact of zeroing least | K6 |
| | | significant and most significant bit planes | K 0 |
| | | by reading an eight bit image, set any of | |
| | | the bit planes 0 to 7 to zero in a user | |
| | | defined menner and reconstruct the image | |
| 2.5 | Imaga and Vidao | Construct the Huffmen tree and find the | V5 |
| 5.5 | Compression | construct the Humman tree and find the | KJ |
| | Compression | number of bits needed for encoding a | |
| | | given message. Calculate number of bits | |
| | | using frequency of characters and number | |
| | | of bits required to represent those | |
| | | characters. | |
| | | Analyze the compression and | K5 |
| | | reconstruction of the 8x8 input images for | |
| | | the 256×256 pixel digital image has | |
| | | eight distinct intensity levels also find the | |
| | | minimum number of bits required to code | |
| | | this image in a lossless manner. | |
| | | Examine the efficiency of Huffman code | K4 |
| | | for an image clip is formed using six | |
| | | colors—white (W), red (R), yellow (Y), | |
| | | green (G), blue (B) and orange (O). These | |
| | | occur in the clip with the following | |
| | | relative frequencies: | |
| | | $\{0.5, 0.1, 0.05, 0.05, 0.2, 0.1\}$. For the above | |
| | | data, construct a Huffman code that | |
| | | minimizes the average code word length. | |
| | | Explain MPEG Video compression | K4 |
| | | standard for monochrome and color | |
| | | compression | |
| | | Formulate the Peak Signal-to-Noise Ratio | К5 |
| | | (PSNR) for the original and the | iii. |
| | | reconstructed images. Calculate the PSNR | |
| | | expressed in decibels | |
| | | Evaluate encoding of the word a1 a2 a3 | K6 |
| | | a4 using arithmetic code and generate the | K 0 |
| | | tag for the given symbol with | |
| | | probabilities: $a1 - 0.2$ $a2 - 0.2$ $a3 - 0.4$ | |
| | | 24-0.2 | |
| | | Solve the entropy of the given 2D image | K3 |
| | | given by f(m n) | IX.J |
| IV | Object Detection and Track | ing | <u> </u> |
| | Object detection and | Predict the key points in objects using | K5 |
| +.1 | recognition in image and | Harries corner detection and SUDE | IX.J |
| | video | Instify why SUDE gives high rehystrose | |
| | | ourresponding to point matching | |
| | | Design a system for detecting Criminal | V5 |
| | | Design a system for detecting Criminals | КЭ |
| 1 | | using Kegion based CINN | |

| | | Examine Histogram of Gradients in 8×8 | K3 |
|-----|-----------------------|--|------------|
| | | cells for object detection | |
| | | Discuss Object detection using bounding | K6 |
| | | box technology in Real Time Traffic | |
| | | monitoring system | |
| | | Determine the object detection in real | K6 |
| | | time video surveillance system | |
| 4.2 | Texture models | Discriminate the texture from the given 4 | K6 |
| | | x4 image segment with grey levels (N) | |
| | | $=0,1,2,3$ and $d=\{1,0\}$ by assuming | |
| | | direction operator as i) next pixel on right | |
| | | side, ii) next pixel on diagonal, iii) next | |
| | | pixel on perpendicular also calculate its | |
| | | homogeneity and uniformity to construct a | |
| | | test feature vector. | |
| | | Discuss the vehicle detection system in | K6 |
| | | real time video based on texture analysis | |
| | | Explain Image segmentation using texture | K4 |
| | | extraction | |
| | | Invent Statistical texture feature for drugs | K5 |
| | | classification | |
| | | Discuss the common statistical features | K2 |
| | | derived from co-occurrence | |
| | | probabilities | |
| | | Create a 2D texture mask of size 5 x 5 | K5 |
| | | with the following 1D filter | |
| | | i) E5E5 | |
| | | ii) E5R5/R5E5 | |
| | | iii) S5S5 | |
| | | iv) L5R5/R5L5 | |
| 4.3 | Image and Video | Devise a Model based video classification | K5 |
| | classification models | using SVM | TT C |
| | | Discuss the performance evaluation of | K6 |
| | | deep feature learning for RGB image and | |
| | | Video classification | 17.4 |
| | | Explain Gaussian mixture models of color | K 4 |
| | | and texture features for image | |
| | | Algorithm | |
| | | Algorium Decign on offective architecture for image | V5 |
| | | classification using CNN | KJ |
| | | Create a unified framework for multi-label | K5 |
| | | image classification | KJ |
| | | Create a model to classify images into | К5 |
| | | their appropriate class with deep learning | |
| | | using CIFAR-10 dataset | |
| | | Predict the classes using SVM classifier | K5 |
| | | for the Breast cancer as Benign, | |
| | | Malignant, or Normal image by applying | |
| | | Otsu thresholding for segmentation, | |
| | | Preprocessing done by applying | |
| | | two-dimensional median filter and | |
| | | histogram equalization for getting more | |
| | | enhanced image. Then extract desired | |

| | | features from the images for | |
|----------|--------------------------|---|------|
| | | classification. | |
| 4.4 | Object tracking in Video | Compare the various object tracking | K4 |
| | | techniques used in video processing | |
| | | Create a model to detecting human motion | K5 |
| | | in video surveillance system | |
| | | Distinguish between automatic detection | K4 |
| | | and motion-based object detection in a | |
| | | video | |
| | | Construct a model for Motion-Based | K3 |
| | | Multiple Object Tracking with suitable | |
| | | example. | |
| | | Choose a suitable object tracking | K6 |
| | | technique to perform Human gesture | |
| *7 | A 10 /0 | tracking and recognition | |
| V 5 1 | Applications | Design a sector for estadour and locio in | V.F |
| 5.1 | Applications and Case | Design a system for category analysis in | КЭ |
| | studies- Retain | tachniques. Suggest a quitable cluster | |
| | | algorithm | |
| | | Construct a model for value and store | K3 |
| | | brand identification in food products using | KJ |
| | | Python OpenCV | |
| | | Develop a system for product | К5 |
| | | identification method for a mixed-reality | iii. |
| | | web shopping system | |
| | | Analyze the RFID Performance | K4 |
| | | Evaluation in a Retail Store | |
| | | Construct Image analytics method to | K3 |
| | | monitor retail store | |
| | | Construct an Automated Shopping Trolley | K3 |
| | | for Super Market Billing System | |
| | | Devise a deep learning pipeline for | K5 |
| | | product recognition on retail store shelves | |
| | | Develop an IoT based retail shopping | K5 |
| | | system | |
| | | Create a model for RFID Based Smart | K5 |
| | | Shopping and Billing | |
| 5.2 | Industrial | Identify fault detection in industrial | K4 |
| | | process using suitable image processing | |
| | | Technique. | |
| | | Explain Image pattern recognition in | K4 |
| | | industrial inspection | |
| 5.3 | Transportation & Travel | Propose a technique to detect traffic sign | K5 |
| | | In real time traffic monitoring system | V5 |
| | | Predict the presence of Pedestrian in | КЭ |
| | | Design a system for detecting driver | V5 |
| | | drowsings using image processing | КЭ |
| | | techniques Suggest a suitable algorithm | |
| | | for each step | |
| | | Determine the discontinuity in the video | K6 |
| | | frame to perform motion segmentation in | 120 |
| 1 | 1 | I mane to perform motion beginemation m | 1 |

| | | Transportation system | |
|-----|------------------------|---|-----|
| | | Design a system for recognition of | K5 |
| | | number plates in vehicle using image | |
| | | processing techniques. Suggest a suitable | |
| | | algorithm for each step | |
| | | Illustrate the applications of vision based | K3 |
| | | intelligent transportation system | |
| 5.4 | Remote sensing | Design Remote sensing image | K5 |
| | | classification using Deep Learning | |
| | | Describe the wiener filter is helpful to | K2 |
| | | reduce the mean square error when | |
| | | Satellite image is corrupted by motion | |
| | | blur and additive noise | |
| | | Select the suitable preprocessing | K6 |
| | | techniques to remove the distortion from | |
| | | the images taken from WSN Video | |
| | | surveillance system and reconstruct the | |
| | | same | |
| | | Plan spatial resolution requirements for | K5 |
| | | crop identification using optical image | |
| | | sensing | |
| | | Propose Remote sensing in precision | K5 |
| | | agriculture | |
| | | Design an edge based and texture-based | K5 |
| | | model for segmenting the remote sensing | |
| | | image and give implementation using | |
| | | MATLAB | |
| 5.5 | Video Analytics in WSN | Design a Distributed visual target- | K5 |
| | | surveillance system in wireless sensor | |
| | | networks | |
| | | Predict Rank preserving | K5 |
| | | Discriminate analysis for human | |
| | | behaviour recognition through wireless | |
| | | sensor networks | |
| | | Discuss Border patrol through | K6 |
| | | advanced wireless sensor networks | |
| | | Propose an intelligent car park management | K5 |
| | | System based on wireless sensor networks | |
| | | application using wireless sonsor | V5 |
| | | networks | KJ |
| 5.6 | IoT Video Analytics | Devise an efficient algorithm for media | K5 |
| 5.0 | Architectures | hased surveillance system in IoT | |
| | | Explain IoT based smart video | K4 |
| | | surveillance system | 174 |
| | | Design a Video Analytics based | K5 |
| | | Intelligent Indoor Positioning System | |
| | | Using IoT | |
| | | | |

4. MAPPING (CO, PO, PSO)

| L-Low | | M-Moderate | | | | | | H- 1 | High | | | | |
|---------|-----|------------|-----|-----|-----|-----|-----|------|------|------|------|------|------|
| Mapping | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
| CO1 | Η | | | | | | | | | | | | |
| CO2 | Н | Н | L | | М | М | | | | | | | М |

| CO3 | Η | | Μ | Н | Μ | | | | | | Η | |
|-----|---|---|---|---|---|---|--|---|---|---|---|---|
| CO4 | Η | Η | Μ | | Η | Μ | | | | | Η | |
| CO5 | Η | Η | | | Η | Η | | М | М | | Н | Н |
| CO6 | Η | Н | Η | Μ | Н | Н | | Η | | Н | | |

5.COURSE ASSESSMENT METHODS DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 2. Open Book Test.
- 3. Cooperative Learning Report, Assignment, Group Presentation, Group Discussion, project Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).
- 4. Pre-Semester & End Semester Theory Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator : Prof. D. Indra Devi

| CORE PRACTICAL VI: BIG DATA MANAGEMENT AND ANALYTICS LAB | | | | | | | |
|--|----------|------------|---|--|--|--|--|
| Semester | III | Hours/Week | 5 | | | | |
| Course Code | P20DS3P6 | Credits | 3 | | | | |

On successful completion of this course, students will be able to:

| CO | Course Outcome | Level | Exercise |
|-----|--|-------|----------|
| CO1 | Develop applications using Hadoop | K6 | 1,2 |
| CO2 | Store and manipulate data using HDFS | K6 | 3 |
| CO3 | Data manipulation using MapReduce | K6 | 4,5 & 6 |
| CO4 | Explore very large datasets using Pig | K6 | 8,9 |
| CO5 | Perform Data Warehousing operations using Hive | K6 | 10 |
| CO6 | Perform data analytics using Spark | K6 | 7 |

2. LIST OF EXERCISES

Develop applications for the following tasks

- 1. Installation and setup of Hadoop
- 2. File management tasks in Hadoop
- 3. Benchmarking and stress testing on Hadoop cluster
- 4. Map Reduce applications for Word Counting
- 5. Stop word elimination using Map Reduce
- 6. Weather data analytics using Map Reduce
- 7. Perform data analytics using Spark
- 8. Perform sort, group, join, project, and filter operations on Pig
- 9. Design vector space model for text collection using Pig
- 10. Create, alter, and drop databases, tables, views, functions, and indexes on Hive

Topics for Self Study

| S.No | Topic Title | Web Link |
|------|-------------|---|
| 1 | HDFS | https://docs.cloudera.com/documentation/enterprise/latest/topics/ |
| | | admin_hdfs_config.html |
| 2 | MapReduce | https://archive.cloudera.com/cdh5/cdh/5/hadoop/hadoop- |
| | | mapreduce-client/hadoop-mapreduce-client- |
| | | core/MapReduceTutorial.html |
| 3 | Spark | https://docs.cloudera.com/documentation/enterprise/latest/topics/ |
| | | spark.html |
| 4 | Pig | https://docs.cloudera.com/documentation/enterprise/5-9- |
| | | x/topics/cdh_ig_pig_installation.html |
| 5 | Hive | https://docs.cloudera.com/documentation/enterprise/5-8- |
| | | <u>x/topics/hive.html</u> |

3. Specific Learning Outcomes

| Exercises | Lab Exercises | Learning Outcome | Level |
|-----------|-----------------------------|-----------------------------------|-------|
| 1 | Installation and setup of | DFS,FS | K6 |
| | Hadoop | | |
| 2 | File management tasks in | Place files in DFS | K6 |
| | Hadoop | | |
| 3 | Benchmarking and stress | Write file in clustered Data Node | K6 |
| | testing on Hadoop cluster | | |
| 4 | Map Reduce applications for | Import jar file for MapReduce | K6 |

| | Word Counting | | |
|----|-----------------------------------|-----------------------------------|----|
| 5 | Stop word elimination using | Modify Word Count file as Word | K6 |
| | Map Reduce | Elimination using Eclips | |
| 6 | Weather data analytics using | Process .csv file using MapReduce | K6 |
| | Map Reduce | | |
| 7 | Perform data analytics using | Spark using Scala | K6 |
| | Spark | | |
| 8 | Perform sort, group, join, | MapReduce using Apache Tez | K6 |
| | project, and filter operations on | | |
| | Pig | | |
| 9 | Design vector space model for | PigLatin Script | K6 |
| | text collection using Pig | | |
| 10 | Create, alter, and drop | Data Warehousing | K6 |
| | databases, tables, views, | | |
| | functions, and indexes on Hive | | |

4. MAPPING (CO, PO, PSO) L-Low

M-Moderate

H- High

| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
|------------|-----|-----|-----|-----|-----|-----|------------|------------|------------|------|------|------|------|
| CO1 | H | Μ | L | Μ | Μ | Μ | L | - | - | Η | Μ | Μ | - |
| CO2 | Н | Μ | Н | Μ | Μ | Н | Μ | Μ | L | Η | Η | - | Μ |
| CO3 | Н | Μ | Μ | L | Н | Μ | - | - | L | Μ | Μ | Μ | - |
| CO4 | Н | L | Н | Н | Н | Н | Μ | Μ | L | H | Μ | Μ | Н |
| CO5 | H | Μ | L | H | H | L | L | Μ | Μ | H | Η | Н | Μ |
| CO6 | H | Μ | Μ | L | Η | L | L | L | Μ | H | H | - | - |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Practical Components): Closed Book
- 2. Cooperative Learning Report, Assignment, Group Discussion, project Report, Field Visit Report, Seminar.
- 3. Pre/Post Test, Viva, Report for each Exercise.
- 4. Lab Model Examination & End Semester Practical Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator : Dr. B. Karthikeyan

| CORE PRACTICAL VII: SOCIAL MEDIA ANALYTICS LAB | | | | | | |
|--|----------|------------|---|--|--|--|
| Semester | III | Hours/Week | 4 | | | |
| Course Code | P19DS3P7 | Credits | 3 | | | |

On successful completion of this course, students will be able to:

| CO# | Course Outcome | Level | Activity |
|-----|---|-------|----------|
| 1 | Create data analytics systems using the data crawled from Twitter | K6 | 1 - 4 |
| 2 | Create data analytics systems using the data crawled from Facebook | K6 | 5,6 |
| 3 | Create data analytics systems using the data crawled from Linkedin | K6 | 7 |
| 4 | Create data analytics systems using the data crawled from GitHub | K6 | 8,9 |
| 5 | Create data analytics systems using the data crawled from Instagram | K6 | 10, 11 |
| 6 | Create data analytics systems on bigdata collections | K6 | 11 - 15 |

2. SYLLABUS

| Activity | Lab Activity Description |
|----------|---|
| 1 | Real time crawling of tweets from Twitter and predict trending words |
| 2 | Extracting text, screen names, and hashtags from tweets. Generating histograms of words, screen names, and hashtags from tweets |
| 3 | Sentiment analysis using nltk.sentiment |
| 4 | Creating a basic frequency distribution from the words in tweets. Also, finding the most popular tweets in a collection of tweets |
| 5 | Counting the total number of page fans from Facebook. Retrieving the Last N items from the feeds of a Facebook Page |
| 6 | Finding the number of likes, shares, and comments on a given Facebook post |
| 7 | Retrieving your LinkedIn profile and print your last name. Performing Clustering your LinkedIn network based on locations of your connections |
| 8 | Finding a list of people who have bookmarked a GitHub repo |
| 9 | Computing the degree, betweenness, and closeness centrality measures of a graph |
| 10 | Displaying your profile picture from Instagram. Displaying the data of the most recent of your Instagram post |
| 11 | Objects detection from images from Instagram posts |
| | Using USA Airline flight dataset, perform the following tasks |
| | Install NetworkX package |
| 12 | • Display the head (top-5 rows) using DataFrame |
| | • Display the nodes and edges |
| | Plot the graph |
| 13 | Using USA Airline flight dataset, find the shortest path based on the airtime between the airports AMA and PBI |
| | Developing a Movie Recommender System that suggests movie IDs that are most similar |
| | to a particular movie ID |
| 14 | • Display the head (top-5 rows) of DataFrame |
| | • Display mean rating of all movies |
| | • Display count rating of all movies |
| | Plot the graph of ratings column |
| 15 | Developing a Movie Recommender System that suggests movie IDs that are most similar to a particular movie ID |

| • Ana | lyze the correlation of two movies |
|-------|---------------------------------------|
| • Sug | gest similar movies for a given movie |

Topics for Self Study

| S.No | Topic Title | Web Link |
|------|---------------------------|--|
| 1 | Network analysis code and | www.cl.cam.ac.uk/~dm754/stna/stna-examples.zip |
| | data | |
| 2 | NodeXL | http://nodexl.codeplex.com/ |
| 3 | Pajek | http://pajek.imfm.si/doku.php |
| 4 | Folium | https://folium.readthedocs.io/en/latest/ |
| | Graph-Tool | https://graph-tool.skewed.de/ |

3. Specific Learning Outcomes

| Activity# | Lab Activity | Learning Outcome | Level |
|-----------|--|---|-------|
| 1. | Twitter data analytics | Crawl tweets at real time from Twitter. Predict trending words from crawled tweets | K6 |
| 2. | Twitter data analytics | Extract text, screen names, and hashtags from tweets. Generate histograms of words, screen names, and hashtags from tweets | K6 |
| 3. | Twitter data analytics | Perform Sentiment analysis using nltk.sentiment | K6 |
| 4. | Twitter data analytics | Create a basic frequency distribution from the words in tweets. Also, find the most popular tweets in a collection of tweets | K6 |
| 5. | Facebook data analytics | Count the total number of page fans from Facebook. Retrieve the Last N items from the feeds of a Facebook Page | K6 |
| 6. | Facebook data analytics | Find the number of likes, shares, and comments on a given Facebook post | K6 |
| 7. | Linkedin data analytics | Retrieve your LinkedIn profile and print your last name. Perform Clustering your LinkedIn network based on locations of your connections | K6 |
| 8. | GitHub data analytics | Find a list of people who have bookmarked a GitHub repo | K6 |
| 9. | GitHub data analytics | Compute the degree, betweenness, and closeness centrality measures of a graph | K6 |
| 10. | Instagram data analytics | Display your profile picture from Instagram. Display the data of the most recent of your Instagram post | K6 |
| 11. | Instagram data analytics | Detect objects from images from Instagram posts | K6 |
| 12. | Bigdata analytics of Airline data | Find out nodes and edges | K6 |
| 13. | Bigdata analytics of Airline data | Find the shortest path | K6 |
| 14. | Design of recommender system for Movie data Part-1 | Plot graph of ratings | K6 |
| 15. | Design of recommender | Analyze correlation of two movies and | K6 |

| system for N | Aovie data | suggest similar movies for a given movie | |
|--------------|------------|--|--|
| Part-2 | | | |

4.MAPPING (CO, PO, PSO)

L-Low

M-Moderate

H- High

| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
|-----|-----|-----|-----|-----|-----|-----|------------|-----|-----|------|------|------|------|
| CO1 | Н | Μ | L | Μ | Μ | Μ | L | - | - | Н | Μ | Μ | - |
| CO2 | Н | Μ | Н | Μ | Μ | Н | Μ | Μ | L | Н | Н | - | Μ |
| CO3 | Н | Μ | Μ | L | Η | Μ | - | - | L | Μ | Μ | Μ | - |
| CO4 | Н | L | Н | Н | Н | Н | Μ | Μ | L | Н | Μ | Μ | Н |
| CO5 | Н | Μ | L | Н | Η | L | L | Μ | Μ | Н | Н | Н | Μ |
| CO6 | Н | Μ | Μ | L | Н | L | L | L | Μ | Η | Η | - | - |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Practical Components): Closed Book
- 2. Cooperative Learning Report, Assignment, Group Discussion, project Report, Field Visit Report, Seminar.
- 3. Pre/Post Test, Viva, Report for each Exercise.
- 4. Lab Model Examination & End Semester Practical Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator : Dr. Janani Selvaraj

| CORE VII: PRINCIPLES OF DEEP LEARNING | | | | | | |
|---------------------------------------|----------|------------|---|--|--|--|
| Semester | III | Hours/Week | 4 | | | |
| Course Code | P20DS307 | Credits | 4 | | | |

On successful completion of this course, students will be able to:

| S.No. | Course Outcomes | Level | Unit |
|-------|--|-------|------|
| CO1 | Examine the basics of Tensorflow and its models | K6 | Ι |
| CO2 | Explain the characteristics of Convolutional Neural Networks | K6 | II |
| CO3 | Depict the architecture and use of the Autocoders | K6 | III |
| CO4 | Evaluate the Sequence analysis using Tenserflow | K5 | III |
| CO5 | Summarize the features of the Recurrent Neural Network | K6 | IV |
| CO6 | Construct the CNN using Deep reinforcement learning | K6 | V |

2. SYLLABUS

UNIT I - TENSORFLOW BASICS

TensorFlow: variables, operations, placeholder Tensors, sessions - Navigating variable scopes and shared variables - Managing models over CPU and GPU - Logistic Regression in TensorFlow-Training Logistic Regression model – Visualizing using Tensor Board – Building multilayer model in TensorFlow

UNIT II - CONVOLUTIONAL NEURAL NETWORKS HOURS

Shortcomings of Feature Selection – Width, height and depth of layers – Filters and feature maps – Describing convolutional layer - Max pooling - Architectural Description of Convolution Networks - Recognizing handwritten digits using CNN for MNIST dataset -Image preprocessing pipelines -Training with Batch normalization

UNIT III - AUTOENCODERS AND SEQUENCE ANALYSIS

Embedding - Principal Component Analysis - Architecture of Autoencoders - Implementing autoencoders in TensorFlow-Denoising - Word2Vec framework for language modelling. Sequence Analysis: seq2seq problem – Dependency parsing – Beam search

UNIT IV - RECURRENT NEURAL NETWORKS

Single neuron and fully connected recurrent layer - Challenges of vanishing gradients - LSTM architecture - TensorFlow primitives for RNN models - Implementing Sentiment analysis Model -Solving seq2seq tasks with RNN - Augmenting RNN with Attention - Designing Neural Translation Network

UNIT V - DEEP REINFORCEMENT LEARNING

Reinforcement Learning: Markov Decision Processes, Policy, Future return, Discounted future return, Balancing Explore-Exploit dilemma, Annealed e-Greedy - Policy learning and Value learning - Solving Pole Cart problem with Policy Gradients - QLearning -Deep QNetworks - Deep Q Recurrent Networks - UNREAL Learning

TOPICS FOR SELF - STUDY

| S.No. | Topics | Web Links | | |
|-------|----------------------------------|-------------------------------------|--|--|
| 1 | Keras Tutorial | https://keras.io/getting_started/ | | |
| 2 | Keras Tutorial: Deep Learning in | https://www.datacamp.com/community/ | | |
| | Python | tutorials/deep-learning-python | | |

12 HOURS

12

12 HOURS

12 HOURS

12 HOURS

| 3 | Machine Learning with Tensorflow | https://www.python- course.eu/tensor_flow_introduction.php | | |
|---|--|--|--|--|
| 4 | From Solving Equations to Deep Learning: A TensorFlow | https://www.toptal.com/machine- learning/tensorflow-python-tutorial | | |

Text Books

- 1. Nikhil Buduma, Nicholas Locascio. Fundamentals of Deep Learning: Designing Next-Generation Machine Intelligence Algorithms. O'Reilly Media. 2017.
- 2. Ian Goodfellow, YoshuaBengio, Aaron Courville. Deep Learning (Adaptive Computation and Machine Learning series). MIT Press, 2017.

References

1. Francois Chollet. Deep Learning with Python. 1ed, Manning Publications, 2017. ISBN 978-1617294433.

Web Links

- 1. www.tensorflow.org / tutorials
- 2. https://www.tensorflow.org/tutorials/generative/autoencoder
- 3. https://towardsdatascience.com/reinforcement-learning-with-python-part-1-creating-the-environment-dad6e0237d2d

3. SPECIFIC LEARNING OUTCOMES

| Unit/ Section | Course Content | Learning outcomes | Level |
|------------------|--|---|-------|
| Ι |]] | TENSORFLOW BASICS | |
| 1.1 | TensorFlow: variables, operations, placeholder Tensors, sessions | Explain the features of the TensorFlow | K4 |
| 1.2 | Navigating variable scopes and shared variables | Characterize the navigating variable and shared variable of TenserFlow. | K4 |
| 1.3 | Managing models over CPU and GPU | Develop the models using CPU and GPU | K6 |
| 1.4 | Logistic Regression in TensorFlow - Training Logistic Regression model | Evaluate the Logistic Regression using TensorFlow | K5 |
| 1.5 | Visualizing using Tensor Board | Depict the model using Tensor Board | K6 |
| 1.6 | Building multilayer model in TensorFlow | Design the multilayer model in TensorFlow | K6 |
| II | CONVOLU | UTIONAL NEURAL NETWORKS | |
| 2.1 | Shortcomings of Feature Selection | Evaluate the disadvantages of the conventional feature selection | K5 |
| 2.2. | Width, height and depth of layers | Analyze the Width, height and depth of layers in CNN | K5 |
| 2.3 | Filters and feature maps | Assess the filters and feature maps in CNN | K5 |
| 2.4 | Describing convolutional layer | Explain the role of convolutional layers in CNN | K6 |
| 2.5 | Max pooling | Diagnose the max pooling methods of CNN | K4 |
| 2.6 | Architectural Description of Convolution Networks | Design the architecture of CNN | K6 |
| 2.7 | Recognizing handwritten digits using CNN for MNIST | Formulate the CNN for Recognizing handwritten digits from MNIST dataset | K6 |

| - | 1-44 | | |
|-----|--|---|-----------|
| 2.0 | | Coto coniza Ima comune consiste a sinclines | VC |
| 2.8 | Training preprocessing pipelines | Categorize image preprocessing pipelines | KO |
| 2.9 | Iraining with Batch | Analyze the training of CININ using batch | K4 |
| TTT | normalization | normalization | |
| | AUTOENCO | DERS AND SEQUENCE ANALYSIS | 175 |
| 3.1 | | Explain the features of embedding | K5 175 |
| 3.2 | Principal Component Analysis | Assess the characteristics of PCA | K5 VC |
| 3.3 | Architecture of Autoencoders | Design the architecture of Autoencoders | K6 |
| 3.4 | Implementing autoencoders in TensorFlow | Construct the autoencoders with TensorFlow | K6 |
| 3.5 | Denoising | Describe the denoising methods for autoencoders | K4 |
| 3.6 | Word2Vec framework for language modelling. | Develop the Autoencoders for Word2Vec framework for language modelling | K6 |
| 3.7 | Sequence Analysis: seq2seq problem | Formulate the Sequence Analysis using TensorFlow | K6 |
| 3.8 | Dependency parsing | Explain the steps of Dependency parsing | K4 |
| 2.0 | Desmo second | Interpret the Beam Search method for sequence | 175 |
| 3.9 | Beam search | analysis. | KS |
| IV | RECUR | RENT NEURAL NETWORKS | |
| | Single neuron and fully | Euclain the characteristics of single Neuron | K4 |
| 4.1 | Single neuron and fully | Construct the Fully connected requirement lower | |
| | connected recurrent layer | Construct the Fully connected recurrent layer | K6 |
| 4.2 | Challenges of vanishing gradients | Evaluate the challenges of vanishing gradients | K4 |
| 4.3 | LSTM architecture | Explain the components of LSTM architecture. | K5 |
| 4.4 | TensorFlow primitives for RNN models | Describe the TensorFlow primitives for RNN models | K2 |
| 4.5 | Implementing Sentiment analysis Model | Design the Sentiment analysis Model using TensorFlow | K6 |
| 4.6 | Solving seq2seq tasks with RNN | Formulate the solution for seq2seq tasks | K6 |
| 4.7 | Augmenting RNN with Attention | Assess the augmenting RNN | K4 |
| 4.8 | Designing Neural Translation Network | Design the Neural Translation Network using RNN | K6 |
| V | DEEP R | EINFORCEMENT LEARNING | |
| | Reinforcement Learning: | | |
| | Markov Decision Processes, | Describe the Reinforcement Learning | K3 |
| 5.1 | Policy, Future return, | Explain the Markov Decision Process. | K5 |
| | Discounted future return, | Compare the different types of return | K5 |
| | Balancing Explore | | |
| 5.0 | Exploit dilemma, Annealed e- | Assess the exploit dilemma of DRL | K5 |
| 5.2 | Greedy | Characterize the Annealed e-Greedy | K4 |
| 5.3 | Policy learning and Value learning | Distinguish the Policy learning and Value learning | K5 |
| 5.4 | Solving Pole Cart problem with Policy Gradients | Prescribe the solution for pole cart problem using policy gradients | K6 |
| 5.5 | QLearning | Explain the properties of QLearning | K4 |
| | | Assess the features of Deep ONetworks from | T7 7 |
| 5.6 | Deep QNetworks | conventional Neural network | K5 |
| 5.7 | Deep Q Recurrent Networks | Compare the Deep QRecurrent Networks over Deep QNetworks | K6 |

| 5.8 | UNREAL Learning | Explain the characteristics of unreal learning | K5 |
|-----|-----------------|--|----|
|-----|-----------------|--|----|

H- High

M-Moderate

4. MAPPING

L-Low

| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
|------------|------------|-----|-----|-----|-----|-----|------------|-----|-----|------|------|------|------|
| CO1 | Η | Μ | Η | Η | Η | Η | - | - | Μ | Η | Η | Μ | - |
| CO2 | Μ | Η | Μ | Η | Μ | Μ | - | Μ | Μ | Μ | Μ | - | - |
| CO3 | Η | Μ | Η | Η | Η | Η | - | - | Η | Η | Μ | - | - |
| CO4 | Η | Η | L | Η | Μ | Η | - | Μ | Η | Η | Η | - | |
| CO5 | Η | Μ | Η | Μ | Η | Η | - | - | Μ | Η | Η | - | - |
| CO6 | Η | Η | Η | Μ | Μ | Η | Η | - | Μ | Η | Η | - | - |

5. COURSE ASSESSMENT METHODS

DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 2. Open Book Test.
- 3. Peer Learning Report, Assignment, Group Presentation, Group Discussion, project Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).
- 4. Pre-Semester & End Semester Theory Examination

INDIRECT:

- 1. Course evaluation survey
- 2. Faculty feedback about the course.

Name of the Course Coordinator: Dr. K. RAJKUAMR

| CORE-XI: WEB DEVELOPMENT USING PYTHON | | | | | | | | |
|---------------------------------------|----------|------------|---|--|--|--|--|--|
| Semester | V | Hours/Week | 5 | | | | | |
| Course Code | P19DS411 | Credits | 4 | | | | | |

On successful completion of this course, students will be able to:

| S.No. | Course Outcomes | Level | Unit |
|-------|---|-------|------|
| CO1 | Develop a Flask extension using best practices | K6 | Ι |
| CO2 | Implement various authentication methods | K5 | II |
| CO3 | Learn how to develop Jinja2 templates | K6 | III |
| CO4 | Build tests for your applications and APIs | K6 | III |
| CO5 | Develop RESTful APIs and secure REST API's | K6 | VI |
| CO6 | Deploy highly available applications that scale on Heroku and AWS using Docker or VMs | K6 | V |

9. A. SYLLABUS

Unit-1. Models, Templates and Web Forms

Simple Application Structure. Creating Models with SQLAlchemy: CRUD operations, Relationships, Constraints and Indexes. Creating Views with Templates: Jinja, Creating views. Web Forms: Basics, Custom validation, Posting comments.

Unit-2. Controllers and Databases

Creating Controllers and Advance Application Structure. Using NoSQL with Flask: NoSQL, RDBMS vs. NoSQL, MongoDB: CRUD operations, Relationships. Email support - Large Application Structure

Unit-3. Authentication, Blog posts and Followers

User Authentication: Methods, Flask Login, OpenID, OAuth, Role Based Access Control. User Roles – User Profiles - Blog Posts – Followers - User Comments

Unit-4. REST and Extensions

Building RESTful API: REST, Authentication, Get, post, put and delete requests. Creating Asynchronous Tasks: Running, monitoring and remembering. Flask Extensions: Caching, Assets and Admin. Building your own extensions: Creating and Modifying.

Unit-5. Testing, Deployment and Version Control

Testing and Performance: Unit Testing, Interface Testing and Test Coverage. Deployment: Deploying on Heruku, AWS and Docker, Version Control with Git

| S.No. | Topics | Web Links |
|-------|------------|---|
| 1 | Django | https://developer.mozilla.org/en-US/docs/Learn/Server-side/Django |
| 2 | Pyramid | https://www.tutorialspoint.com/python_web_development_libraries |
| | | /python_web_development_libraries_pyramid_framework.htm |
| 3 | Turbogears | https://www.fullstackpython.com/turbogears.html |
| 4 | Web2Py | https://www.tutorialspoint.com/web2py/index.htm |

B. TOPICS FOR SELF - STUDY

C. TEXT BOOKS

6. Daniel Gaspar, Jack Stouffer. Mastering Flask Web Development: Build enterprise-grade, scalable Python_web applications. 2ed. Packt Publishing Ltd. 2018. ISBN 978-1788995405.

7. Miguel Grinberg. Flask Web Development, 2ed. Shroff Publishers. 2018. ISBN 9789352136995

D. REFERENCES BOOKS

- 1. Italo Maia. Building Web Applications with Flask. Packt Publishing Ltd. 2015. ISBN 978-1784396152.
- 2. Shlabh Aggarwal. Flask Framework Cookbook. Packt Publishing Ltd. 2014.

E. WEB LINKS

- 1. <u>https://www.tutorialspoint.com/web2py/index.htm</u>
- 2. https://www.tutorialspoint.com/python_web_development_libraries/python_web_development_libraries_pyramid_framework.htm

8. SPECIFIC LEARNING OUTCOMES

| Unit/ Section | Course Content | Learning outcomes | | | | | | | |
|------------------|---|--|-------------------------|--|--|--|--|--|--|
| Ι | MODELS, | TEMPLATES AND WEB FORMS | | | | | | | |
| 1.1 | Simple Application Structure. Creating Models with Creating. | Develop the simple web applications | K6 | | | | | | |
| 1.2 | SQLAlchemy: CRUD operations, Relationships, Constraints and Indexes. | Construct the index for web applications using CRUD in SQLAlchemy | K6 | | | | | | |
| 1.3 | Views with Templates: Jinja, Creating views. | Create views using Jinja | K6 | | | | | | |
| 1.4 | Web Forms: Basics, Custom validation, Posting comments | Develop the Webforms with validation | K6 | | | | | | |
| II | CONTROLLERS AND DATABASES | | | | | | | | |
| 2.1 | Creating Controllers and Advance Application Structure. | Create the controllers for Application structures | K6 | | | | | | |
| 2.2. | Using NoSQL with Flask: NoSQL, RDBMS vs. NoSQL | Develop the web applications with DB connectivity | <u>K6</u> | | | | | | |
| 2.3 | MongoDB: CRUD operations, Relationships. Email support - Large Application Structure | Create the CRUD operations in MongoDB Develop the Large Application Structure | K <u>6</u> <u>K6</u> | | | | | | |
| III | AUTHENTICAT | TION, BLOG POSTS AND FOLLOWERS | | | | | | | |
| 3.1 | User Authentication: Methods, Flask Login, OpenID, OAuth, Role Based Access Control. | Explain the User AuthenticationDesignthewebapplicationAuthentication | K <u>5</u> <u>K6</u> | | | | | | |
| 3.2 | User Roles – User Profiles - Blog Posts – Followers - User Comments | Assess the user roles in Web applications Design the Blogs | K6 <u>K6</u> | | | | | | |
| IV | F | REST AND EXTENSIONS | | | | | | | |
| 4.1 | Building RESTful API: REST, Authentication, Get, post, put and delete requests. Creating | Construct the RESTful API | K <u>6</u> | | | | | | |
| 4.2 | Asynchronous Tasks: Running, monitoring and remembering. | Assess the Asynchronous Tasks in API | K <u>6</u> | | | | | | |
| 4.3 | Flask Extensions: Caching, | Design the Flask Extensions | K <u>6</u> | | | | | | |

| | Assets and Admin. | | | |
|-----|---|---|------------|--|
| 4.4 | Buildingyourownextensions:CreatingandModifying | Develop custom extensions | K <u>6</u> | |
| V | TESTING, DEP | LOYMENT AND VERSION CONTROL | | |
| 5.1 | TestingandPerformance:UnitTesting,InterfaceTesting and Test Coverage. | Evaluate the Web applications using Testing Strategies | К <u>5</u> | |
| 5.2 | Deployment: Deploying on Heruku, AWS and Docker | Deploy the web applications Heruku, AWS and Docker | K6 | |
| 5.3 | Version Control with Git | Apply the Version Control with Git | K5 | |

9. MAPPING

L-Low

M-Moderate

H- High

| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
|------------|------------|-----|-----|------------|-----|------------|------------|------------|------------|------|------|------|------|
| CO1 | Η | Μ | Η | Η | Η | Η | - | - | Μ | Н | Η | Μ | - |
| CO2 | Μ | Η | Μ | Η | Μ | Μ | - | Μ | Μ | Μ | Μ | - | - |
| CO3 | Η | Μ | Η | Η | Η | Η | - | - | Η | Η | Μ | - | - |
| CO4 | Η | Η | L | Η | Μ | Η | - | Μ | Η | Н | Η | - | |
| CO5 | Η | Μ | Η | Μ | Η | Η | I | - | Μ | Η | Η | - | - |
| CO6 | Η | Η | Η | Μ | Μ | Η | Η | - | Μ | Η | Η | - | - |

10. COURSE ASSESSMENT METHODS

DIRECT:

- 9. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 10. Open Book Test.
- 11. Peer Learning Report, Assignment, Group Presentation, Group Discussion, project Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).
- 12. Pre-Semester & End Semester Theory Examination

INDIRECT:

- 5. Course evaluation survey
- 6. Faculty feedback about the course.

Name of the Course Coordinator: Dr. B. Karthikeyan

| ELECTIVE-5: SUPPLY CHAIN MANAGEMENT | | | | | | |
|-------------------------------------|----------|------------|---|--|--|--|
| Semester | IV | Hours/Week | 5 | | | |
| Course Code | P20DS4:5 | Credits | 4 | | | |

After the successful completion of this course the students will be able to

| S.No. | Course Outcomes | Level | Unit |
|-------|--|-------|------|
| CO1 | Perceive the foundations of a supply chain and explore strategies and logistics drivers by which the supply chain of an organization can be managed to enhance its business competitiveness. | K6 | Ι |
| CO2 | Evaluate and Analytically examine the strategic drivers and metrics of supply chain organizations and measure performance improvement | K6 | Π |
| CO3 | Design and provide a network to support the business decision-making within the context of supply chain management and the real world. | K5 | III |
| CO4 | Plan optimized transportation and logistics activities in supply chain operations | K6 | IV |
| CO5 | Determine the outsourcing decisions by applying the buy-make framework to manage the benefit and risks of outsourcing | K6 | V |
| CO6 | Recommend a proper blend of Logistics and Supply elements thereby confining the wide range of applications in the changing dynamic environment and industry practices | K6 | v |

1. A. SYLLABUS

Unit-1. Building strategic framework

Supply chain: Definition, 3 streams of knowledge, objectives and importance - Decision phases and process views of a supply chain (SC) - Examples of supply chain - Competitive strategy and SC strategy - 3 steps of achieving strategic fit - Improving SC performance by expanding scope of strategic fit, challenges to achieving strategic fit - Financial measures and drivers of SC performance - Logistical drivers: Roles in SC and decision components - Cross functional drivers: Roles in SC and decision components - Role of infrastructure in SC performance.

Unit-2. Designing SC network

Key factors influencing distribution network design - Design options for a distribution network -Impact of online sales on customer service and cost - Network design decisions: Influencing factors, framework - Capacitated plant location model for network optimization - Gravity location model for network design - Model for demand allocation and locating plants - Global supply chain: Dimensions to evaluate total cost, SC risks, tailored risk mitigation strategies - Discounted cash flow analysis to evaluate network design decision - Decision tree analysis: Basics, Evaluating flexibility at Trip Logistics.

Unit-3. Planning and coordinating demand and supply

Demand forecasting: role, characteristics, components and methods - Static demand forecasting methods - Adaptive demand forecasting methods - Measures of demand forecasting error - Aggregate planning: role, identifying aggregate units, strategies - Aggregate planning using Linear programming - Managing supply and demand to improve synchronization in SC - Lack of SC coordination: Bullwhip effect, effect on performance - Obstacles to coordination in SC - Managerial levers to achieve coordination of demand and supply in SC.

Unit-4. Planning and managing inventories

Cycle inventory terminologies: Lot size, Average flow time, Inventory holding cost, Ordering cost -Computing optimal lot size for single product: Economic order quantity, for Production environment, with Capacity constraint - Lot size based discount schemes: All unit quantity discounts, Marginal unit quantity discount - Trade promotions: Goals, Forward buying, Impact on lot size and cycle inventory - Factors affecting the level of safety inventory - Evaluating required safety inventory: Given a replenishment policy, Desired cycle service level, Desired fill rate -Impact of desired product availability and uncertainty on safety inventory - Impact of supply uncertainty on safety inventory - Factors affecting optimal level of product availability - Managerial levers of inventory to improve SC profitability.

Unit-5. Transportation and cross functional drivers

Modes of transportation in SC - Design options for a transportation network - Transportation and inventory cost trade off - Transportation cost and customer responsiveness trade off - Tailored transportation - Sourcing decisions: In house or Outsource - Sharing risk and reward in SC - Pricing and revenue management for multiple customer segments - Pricing and revenue management for perishable assets - Pricing and revenue management for seasonal demand.

| S.No. | Topics | Web Links |
|-------|-----------------------------------|---|
| 1 | Digitization of Supply chain | https://www.coursera.org/lecture/process- |
| | | improvement/lecture-4-1-digitization-of-the-supply- |
| | | chain-EFofn |
| 2 | Supply chain Analytics | https://nptel.ac.in/courses/110/108/110108056/ |
| 3 | Artificial Intelligence in Supply | https://towardsdatascience.com/artificial-intelligence- |
| | Chain Management | in-supply-chain-management-predictive-analytics- |
| | | for-demand-forecasting-80d2d512f155 |
| 4 | Logistics and Supply chain | http://slmt.in/courses/cilt-international- |
| | Management | courses/diploma-in-logistics-and-supply-chain- |
| | | management-dlsm/ |

B. TOPICS FOR SELF-STUDY

C. TEXT BOOKS

1. Sunil Chopra, Peter Meindl and DV Karla. "Supply Chain Management: Strategy, planning and operation", 6th edition, Pearson, 2016. ISBN 978-9332548237 (Excluding Excel Examples)

D. REFERENCES BOOKS

1. David Simchi-Levi and Philip Kaminsky. "Designing and managing the supply chain: Concepts, strategies and case studies", 3rd edition, McGraw Hill, 2007.

E. WEB LINKS

- 1. http://www.supply-chain.com
- 2. http://www.transportlink.com
- 3. http://www.transportlaw.com
- 4. http://www.apics.org
- 5. http://www.clm1.org
- 6. http://www.napm.org

3.SPECIFIC LEARNING OUTCOMES (SLO)

| Unit/ Section | Course Content | Learning outcomes | Level | |
|------------------|---|--|-------|--|
| Ι | Building strategic framewor | 'k | | |
| 1.1 | Supply chain: Definition, 3 streams of knowledge, objectives and importance | Discuss the goal of supply chain and impact of supply chain decision on success of the firm. | K2 | |

| | | Describe the various objectives of supply chain | K2 |
|------|------------------------------|---|------------|
| | | Illustrate the importance of supply chain | КЗ |
| | | management | |
| | | Identify the supply chain obstacles. | K4 |
| | | Determine Strategic, operational and | K6 |
| | | tactical planning of supply chain | |
| 1.2 | Decision phases and process | Explain decision phases in supply chain | K4 |
| | views of a supply chain | Describe the cycle and push/pull view of | K4 |
| | (SC) | a supply chain | N/C |
| | | Determine the underlying theoretical | Ко |
| 13 | Examples of supply chain | Discuss in what way do supply chain | K5 |
| 1.5 | Examples of suppry chain | flows affect the success or failure of a | K5 |
| | | firm such as Amazon and list two supply | |
| | | chain decisions that have a significance | |
| | | impact on supply chain Profitability. | |
| 1.4 | Competitive strategy and SC | Analyze the strategies that are critical to | K4 |
| | strategy | achieving strategic fit for company's | |
| | | overall success. | |
| 1.5 | 3 steps of achieving | Explain 'Achieving Strategic Fit' in | K4 |
| | strategic fit | supply chains with the help of a suitable | |
| 1.6 | Improving SC performance | Choose strategic fit between its supply | V 5 |
| 1.0 | by expanding scope of | chain strategy and its competitive strategy | KJ |
| | strategic fit, challenges to | Explain the Balanced Score Card | К2 |
| | achieving strategic fit | approach of supply chain performance | |
| | | measurement. | |
| | Financial measures and | Apply the key metrics that track the | K4 |
| 1.7 | drivers of SC performance | performance of the supply chain in terms | |
| | | of each driver. | |
| 1.8 | Logistical drivers: Roles in | Identify the role of major drivers in | K4 |
| | SC and decision | supply chain | |
| 19 | Cross functional drivers: | Recommend the ways to boost up the | K6 |
| 1.7 | Roles in SC and decision | cross functional drivers roles in SC | IX0 |
| | components | Explain the barriers of cross functional | K2 |
| | | drivers | K2 |
| | | Analyze cross functional management is | K4 |
| | | effectively managing supply chains | |
| 1.10 | Role of infrastructure in SC | Debate Economic impact of inadequate | K6 |
| | performance | infrastructure for sc integration | |
| II | Designing SC network | 1 | 1 |
| 2.1 | Key factors influencing | Explain the factors influencing | K2 |
| | distribution network design | distribution network design | N/C |
| | | typically best suited for commodity items | N 0 |
| 22 | Design options for a | Examine the design ontions available for | КЗ |
| 2.2 | distribution network | a distribution network with option in | 11.5 |
| | | detail | |
| | | Design a suitable distribution network | K5 |
| | | utilized for the specialty chemical | |
| | | company is considering expanding its | |

| | | operations into Brazil, when five | |
|-----|----------------------------|---|------------|
| | | companies dominate the consumption of | |
| | | specialty chemicals. | |
| | | Construct the role of network design | К5 |
| | | decision in a supply chain | 110 |
| | | Plan different design options available for | К5 |
| | | a distribution network with option in | K. |
| | | detail | |
| 23 | Impact of online sales on | Predict the impact of online sales on | K5 |
| 2.3 | sustemar service and cost | consumers and firms. Cive evidence from | K.J |
| | customer service and cost | consumer alectronics | |
| | | Lustifice is a husiness likely to be more | VC |
| | | Justify is e-business likely to be more | Ко |
| | | beneficial in the early part or the mature | |
| | | part of a product's life cycle. | 77.4 |
| | | Explain the cycle and push/pull view of a | K4 |
| | | supply chain. | |
| 2.4 | Network design decisions: | Describe planning Networks | K2 |
| | Influencing factors, | Interpret the objectives & process of | K2 |
| | framework | Supply Chain Network optimization | |
| | | models | |
| | | Asses the outcome and benefits of | K6 |
| | | Supply Chain Network optimization | |
| | | models. | |
| | | Analyze the benefits are these using bar | K4 |
| | | codes and scanners for orders entry as | |
| | | opposed to keyboard encoding into a | |
| | | computer database | |
| | | Describe the current trends in value | К2 |
| | | addition happened in Indian companies | 112 |
| | | Identify factors influencing supply chain | KA |
| | | network decisions | K 4 |
| | | Outline the adventages and disadventages | V2 |
| | | of distribution network design antions | κ2 |
| | | Of distribution network design options | 17.5 |
| | | Propose factors to be considered in | КЭ |
| | | deciding whether to make and supply or | |
| | | to buy and supply for blood pressure | |
| | | measuring kits for hospitals in developing | |
| | | rural markets in India. | |
| 2.5 | Capacitated plant location | Explain optimized network. | K2 |
| | model for network | Discuss the various Network | K2 |
| | optimization | optimization models | |
| | - | Construct the classification of supply | K5 |
| | | chain network design decisions | _ |
| | | Argue the following statement "Some | K6 |
| | | industries are located near the source of | |
| | | raw materials whereas some near the | |
| | | markets for finished goods" | |
| 2.6 | Gravity location model for | Design network decisions using decision | К5 |
| 2.0 | network design | tree and list its importance | n., |
| | | Davisa a Framework to make a notwork | K5 |
| | | design decision | KJ (KJ |
| | | Identify factors to be considered for | V A |
| | | locating a control of the location | N 4 |
| | | locating a centralized kitchen to cook | |

| | | food for a restaurant chain. Also suggest | |
|-----|-------------------------------|---|----------------|
| | | an appropriate facility location model. | |
| | | State the assumption if any | |
| | | Discuss the optimization models used for | K2 |
| | | facility location and capacity allocation | |
| 2.7 | Global supply chain: | Determine the role of a third party in | K6 |
| | Dimensions to evaluate total | increasing the supply chain surplus | |
| | cost, SC risks, tailored risk | Describe global supply chain risk | K2 |
| | mitigation strategies | management strategies | |
| | | Determine the total cost approach to | K6 |
| | | supply chain risk modelling | |
| | | Identify the methods to managing risk to | K4 |
| | | avoid supply chain breakdown | |
| | | Discuss the Strategies for supply chain | K4 |
| | | risk management | |
| 2.8 | Discounted cash flow | Outline uncertainty in network design | K2 |
| | analysis to evaluate network | discounted cash flow analysis | |
| | design decision | Determine the uncertainties and risk | K6 |
| | | factors so important in evaluating supply | |
| | | chain design decisions | |
| 2.9 | Decision tree analysis: | Write the features of decision tree. | |
| | Basics, Evaluating | Asses the benefits of using decision nodes | K6 |
| | flexibility at Trip Logistics | by decision making under uncertainty | |
| | | Explain the formation of a decision | K2 |
| | | tree based on the Trips logistics | |
| III | Planning and coordinating of | lemand and supply | |
| 3.1 | Demand forecasting: role, | Examine the basic approaches to demand | K3 |
| | characteristics, components | forecasting | |
| | and methods | Predict the forecast error if demand in 5 | K5 |
| | | tons out to be 125 litres for a grocery store | |
| | | has experienced a weekly demand of oil | |
| | | 1 2 | |
| | | of 120,127,114,and 122 litres over the last | |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 | |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. | |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the | K6 |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the store manager should order in each | K6 |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the store manager should order in each replenishment lot. Demand for computers | K6 |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the store manager should order in each replenishment lot. Demand for computers in a store is 12,000 units per year. The | K6 |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the store manager should order in each replenishment lot. Demand for computers in a store is 12,000 units per year. The store incurs a fixed order placement, | K6 |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the store manager should order in each replenishment lot. Demand for computers in a store is 12,000 units per year. The store incurs a fixed order placement, trasnsportation and receiving cost of | K6 |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the store manager should order in each replenishment lot. Demand for computers in a store is 12,000 units per year. The store incurs a fixed order placement, trasnsportation and receiving cost of Rs.40,000/- each time an order is placed. | К6 |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the store manager should order in each replenishment lot. Demand for computers in a store is 12,000 units per year. The store incurs a fixed order placement, trasnsportation and receiving cost of Rs.40,000/- each time an order is placed. Each computer costs the store Rs.5000/- and the holding cost is 20%. Also emploin | K6 |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the store manager should order in each replenishment lot. Demand for computers in a store is 12,000 units per year. The store incurs a fixed order placement, trasnsportation and receiving cost of Rs.40,000/- each time an order is placed. Each computer costs the store Rs.5000/- and the holding cost is 20%. Also explain the impact of supply chain uncertainty on | K6 |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the store manager should order in each replenishment lot. Demand for computers in a store is 12,000 units per year. The store incurs a fixed order placement, trasnsportation and receiving cost of Rs.40,000/- each time an order is placed. Each computer costs the store Rs.5000/- and the holding cost is 20%. Also explain the impact of supply chain uncertainty on safety inventory | К6 |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the store manager should order in each replenishment lot. Demand for computers in a store is 12,000 units per year. The store incurs a fixed order placement, trasnsportation and receiving cost of Rs.40,000/- each time an order is placed. Each computer costs the store Rs.5000/- and the holding cost is 20%. Also explain the impact of supply chain uncertainty on safety inventory | K6 |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the store manager should order in each replenishment lot. Demand for computers in a store is 12,000 units per year. The store incurs a fixed order placement, trasnsportation and receiving cost of Rs.40,000/- each time an order is placed. Each computer costs the store Rs.5000/- and the holding cost is 20%. Also explain the impact of supply chain uncertainty on safety inventory Asses the role does forecasting play in the supply chain of a build-to-order | К6 |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the store manager should order in each replenishment lot. Demand for computers in a store is 12,000 units per year. The store incurs a fixed order placement, trasnsportation and receiving cost of Rs.40,000/- each time an order is placed. Each computer costs the store Rs.5000/- and the holding cost is 20%. Also explain the impact of supply chain uncertainty on safety inventory Asses the role does forecasting play in the supply chain of a build-to-order manufactures such as dell | K6 K6 |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the store manager should order in each replenishment lot. Demand for computers in a store is 12,000 units per year. The store incurs a fixed order placement, trasnsportation and receiving cost of Rs.40,000/- each time an order is placed. Each computer costs the store Rs.5000/- and the holding cost is 20%. Also explain the impact of supply chain uncertainty on safety inventory Asses the role does forecasting play in the supply chain of a build-to-order manufactures such as dell Determine the forecast error if demand in | К6 К6 |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the store manager should order in each replenishment lot. Demand for computers in a store is 12,000 units per year. The store incurs a fixed order placement, trasnsportation and receiving cost of Rs.40,000/- each time an order is placed. Each computer costs the store Rs.5000/- and the holding cost is 20%. Also explain the impact of supply chain uncertainty on safety inventory Asses the role does forecasting play in the supply chain of a build-to-order manufactures such as dell Determine the forecast error if demand in period 5 turns out to be 125 gallons for | K6 K6 K6 |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the store manager should order in each replenishment lot. Demand for computers in a store is 12,000 units per year. The store incurs a fixed order placement, trasnsportation and receiving cost of Rs.40,000/- each time an order is placed. Each computer costs the store Rs.5000/- and the holding cost is 20%. Also explain the impact of supply chain uncertainty on safety inventory Asses the role does forecasting play in the supply chain of a build-to-order manufactures such as dell Determine the forecast error if demand in period 5 turns out to be 125 gallons for a super market has experienced weekly | K6 K6 K6 |
| | | of 120,127,114,and 122 litres over the last 4 weeks. Forecast demand for period 5 using a four period moving average. Evaluate the number of computers the store manager should order in each replenishment lot. Demand for computers in a store is 12,000 units per year. The store incurs a fixed order placement, trasnsportation and receiving cost of Rs.40,000/- each time an order is placed. Each computer costs the store Rs.5000/- and the holding cost is 20%. Also explain the impact of supply chain uncertainty on safety inventory Asses the role does forecasting play in the supply chain of a build-to-order manufactures such as dell Determine the forecast error if demand in period 5 turns out to be 125 gallons for a super market has experienced weekly demand of milk of 120,127,114 and 122 | K6 K6 K6 |

| | | demand for period 5 using a four –period | |
|------|------------------------------|---|--------------|
| 2.2 | | moving average. | 17.0 |
| 3.2 | Static demand forecasting | Classify the static and adaptive | K3 |
| | methods | Forecasting methods | 17.4 |
| | | Explain the basic, six step approach to halp an organization perform affective | K 4 |
| | | forecasting | |
| | | Investigate Demand forecasting analysis | K/ |
| | | using time series methods | 127 |
| 3.3 | Adaptive demand | Determine the house old electricity | K6 |
| | forecasting methods | demand forecasting using adaptive | |
| | C | conditional density estimation | |
| | | Formulate the adaptive water demand | K5 |
| | | forecasting for near real time management | |
| | | of smart water distribution system | |
| 3.4 | Measures of demand | Invent demand forecast accuracy and | K5 |
| | forecasting error | forecast error | |
| 3.5 | Aggregate planning: role, | Outline the operational parameters to | K2 |
| | identifying aggregate units, | identify aggregate plan | |
| | strategies | Select the major cost categories needed as | K6 |
| | | input for aggregate planning | T 7 4 |
| | | Identify the managerial levers that reduce | K4 |
| | | lot size and cycle inventory in a supply | |
| 3.6 | A garagata planning using | Explain the role of collaborative planning | V6 |
| 5.0 | Linear programming | and forecasting for efficient execution of | KU |
| | Enical programming | supply chains | |
| | | suppry chains. | |
| | | Illustrate the role predictive visibility | K3 |
| | | supply chain performance. | |
| | | | |
| | | Explain the different types of costs | K4 |
| | | associated with aggregate planning. For | |
| | | each of the cost, enumerate the areas | |
| | | where the cost plays an important role. | T T < |
| | | Discuss the major cost categories needed | K6 |
| | | as input for aggregate planning | 1/2 |
| | | Solve aggregate planning using Linear | К3 |
| 37 | Managing supply and | Investigate the Synchronization in supply | K4 |
| 5.7 | demand to improve | chains implications for design and | 157 |
| | synchronization in SC | management | |
| 3.8 | Lack of SC coordination: | Write a note on the Coordination in a | K1 |
| | Bullwhip effect, effect on | supply chain. | |
| | performance | Analyze the Bullwhip effect in supply | K4 |
| | | chain for the effect on performance | |
| 3.9 | Obstacles to coordination in | List the various obstacles to coordination | K1 |
| | SC | and how such obstacles can be minimized | |
| | | in supply chain | |
| 3.10 | Managerial levers to achieve | Design the managerial levers that help to | K5 |
| | coordination of demand and | achieve coordination in the supply chain | |
| TX 7 | supply in SC | | |
| | Planning and managing invo | Evolute the number of cost ideas that the | VC |
| 4.1 | Cycle inventory | Evaluate the number of cartriages that the | NU |

| | terminologies: Lot size, | store manager should order in each | |
|------|---------------------------------------|--|----|
| | Average flow time, | replenishment lot for Demand of | |
| | Inventory holding cost, | cartridges in an electronic store is 1000 | |
| | Ordering cost | units per month. The firm incurs a fixed | |
| | _ | order placement, transportation and | |
| | | receiving costs of Rs.4000/- each time an | |
| | | order is placed. Each cartridge costs | |
| | | Rs.500/- and the retailer has a holding | |
| | | cost of 20 percent. | |
| | | Explain multi-echelon inventory | K4 |
| | | management in detail with the help of a | |
| | | suitable example. | |
| | | Explain how to manage supply chain | K4 |
| | | cycle inventory. | |
| | | Construct the role of cycle inventory in a | K5 |
| | | supply chain and how uncertainty in the | |
| | | supply chain managed | |
| 4.2 | Computing optimal lot size | Show how to compute the optimal lot size | K3 |
| | for single product: | and cycle length for the given sequence of | |
| | Economic order quantity, | items in a cycle | |
| | for Production environment, | Investigate optimal lot sizes in the | K4 |
| | with Capacity constraint | economic lot scheduling for production | |
| | | environment | |
| 4.3 | Lot size based discount | Distinguish the lot size based and volume | K2 |
| | schemes: All unit quantity | based quantity discounts | |
| | discounts, Marginal unit | Analyze the effect of quantity discounts | K4 |
| | quantity discount | on lot size and cycle inventory | |
| | | Examine the effect of trade promotions on | K4 |
| | | lot size and cycle inventory | |
| 4.4 | Trade promotions: Goals, | Describe how to Managing Multi-Echelon | K2 |
| | Forward buying, Impact on | Cycle Inventory | |
| | lot size and cycle inventory | Explain the impact of trade promotions on | K4 |
| | | lot size and cycle inventory | |
| 4.5 | Factors affecting the level of | State and briefly explain the role of safety | K1 |
| | safety inventory | inventory in supply chain | |
| 4.6 | Evaluating required safety | Propose "Relevant deterministic and | K5 |
| | inventory: Given a | Stochastic Inventory Models" and explain | |
| | replenishment policy, | its relevance in an organization. Also, | |
| | Desired cycle service level, | briefly explain the important features of | |
| 47 | Desired IIII rate | These models. | VC |
| 4.7 | Impact of desired product | Evaluate the impact of desired product | KO |
| | availability and uncertainty | availability and uncertainty on safety | |
| 1.0 | Impost of supply upcortainty | Evaluate the Impact of supply upcortainty | VC |
| 4.0 | on safety inventory | on safety inventory | KO |
| 4.9 | Factors affecting optimal | Discuss optimal level of product | K2 |
| | level of product availability | availability | |
| | · · · · · · · · · · · · · · · · · · · | Determine the optimal level of product | K6 |
| | | availability | - |
| 4.10 | Managerial levers of | Design the managerial levers that help to | K5 |
| | inventory to improve SC | improve inventory SC profitability | |
| | profitability | - • • • | |
| V | Transportation and cross fu | nctional drivers | |
| 5.1 | Modes of transportation in | Discuss the importance of transportation | K6 |

| | SC | in supply chain. | |
|-----|------------------------------|---|----|
| | | Explain the modes of transportation and | K4 |
| | | their performance characteristics | |
| 5.2 | Design options for a | Design an option for a transportation | K5 |
| | transportation network | network | |
| 5.3 | Transportation and | Determine tradeoffs in transportation | K6 |
| | inventory cost trade off - | design network | |
| | Transportation cost and | Distinguish transportation cost, customer | K2 |
| | customer responsiveness | responsiveness tradeoffs and Tailored | |
| | trade off - Tailored | transportation | |
| | transportation | | |
| 5.4 | Sourcing decisions: In house | Discuss the importance of in-sourcing and | K6 |
| | or Outsource - Sharing risk | out-sourcing with suitable examples | |
| | and reward in SC | Debate Strategic Alliances and | K6 |
| | | Outsourcing | |
| | | Describe the ways that a firm such as | K1 |
| | | Wal-Mart form out sourcing | |
| | | decisions | |
| 5.5 | Pricing and revenue | Explain the importance of pricing in | K6 |
| | management for multiple | supply chain management and elucidated | |
| | customer segments | various type of pricing approaches that | |
| | | generate maximum profit | |
| 5.6 | Pricing and revenue | Design Perishable assets for pricing and | K5 |
| | management for perishable | revenue management | |
| | assets | | |
| 5.7 | Pricing and revenue | Explain pricing and revenue management | K4 |
| | management for seasonal | for seasonal demand | |
| | demand. | | |
| | | | |

4. MAPPING (CO, PO, PSO)

| L-Low | 7 | | M-Moderate | | | | | | | H- High | | | |
|-------|-----|-----|------------|-----|-----|-----|-----|-----|-----|---------|------|------|------|
| | PO1 | PO2 | PO3 | PO4 | PO5 | PO6 | PO7 | PO8 | PO9 | PSO1 | PSO2 | PSO3 | PSO4 |
| CO1 | Η | | | | | | | | | | | | Н |
| CO2 | Η | Н | Μ | Η | Μ | Η | Μ | | Μ | Н | | | Н |
| CO3 | Η | Н | Μ | Η | Μ | Η | Μ | Η | Η | Н | Н | Н | |
| CO4 | Η | Μ | Μ | Μ | | Η | Μ | Η | Μ | Н | Н | Н | Μ |
| CO5 | Η | Н | Η | Μ | Μ | Η | Μ | Η | Μ | Н | Н | | |
| CO6 | Η | Η | Η | Μ | Η | Η | Η | Η | Η | Н | Н | | |

5. COURSE ASSESSMENT METHODS DIRECT:

- 1. Continuous Assessment Test: T1, T2 (Theory & Practical Components): Closed Book
- 2. Open Book Test.
- 3. Cooperative Learning Report, Assignment, Group Presentation, Group Discussion, project Report, Field Visit Report, Poster Presentation, Seminar, Quiz (written).
- 4. Pre-Semester & End Semester Theory Examination

INDIRECT:

1. Course end survey (Feedback)

Name of the Course Coordinator : Prof. D. Indra Devi