

# MAHARAJA INSTITUTE OF TECHNOLOGY MYSORE

Autonomous Institution Affiliated to VTU

# **Competency Based Syllabus (CBS)**

for

# **Computer Science and Engineering (CS&E)**

(Under Outcome Based Education (OBE) and Choice-Based Credit System (CBCS))

# Offered from 1<sup>st</sup>to 2<sup>nd</sup> Semesters of Study

in

Partial Fulfilment for the Award of Master's Degree in

# **Computer Science and Engineering (CS&E)**

# 2023 Scheme

# **Board: CS**

Scheme Effective from the academic year 2023-24



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| 1stCom agtor             | <b>Basic Science Course (BS)</b>        | MOOMCGIAI |
|--------------------------|---|-----------|
| 1 <sup>st</sup> Semester | Advance Mathematics for Emerging Trends | M23MCS101 |

|     | rerequisites                       |   |
|-----|------------------------------------|---|
| S/L | Proficiency                        | Prerequisites   |
| 1   | Vector Spaces                      | <ul><li>Linear Algebra: Basic understanding of matrices, determinants, and systems of linear equations.</li><li>Calculus: Knowledge of differentiation and integration to understand vector functions.</li><li>Mathematical Maturity: Ability to comprehend abstract concepts and proofs.</li></ul>   |
| 2   | Orthogonality and<br>Least Squares | Linear Algebra: Familiarity with matrix operations, inner products, and<br>orthogonal projections.<br>Statistics: Basic understanding of statistical methods and regression analysis.<br>Programming Skills: Ability to implement least squares algorithms in a<br>programming language like Python or MATLAB.  |
| 3   | Eigenvalues and<br>Eigenvectors    | Linear Algebra: In-depth knowledge of matrix theory, including determinants,<br>trace, and characteristic equations.<br>Differential Equations: Basic understanding of differential equations where<br>eigen values and eigenvectors are applied.<br>Numerical Methods: Basic exposure to numerical techniques for solving Eigen  |
| 4   | Sampling Theory                    | <ul> <li>value problems.</li> <li>Probability and Statistics: Strong foundation in probability theory, random variables, and distributions.</li> <li>Discrete Mathematics: Understanding of combinatorics and basic set theory.</li> <li>Data Analysis: Familiarity with data collection methods and statistical analysis techniques.</li> </ul>                                |
| 5   | Numerical Methods                  | Calculus: Comprehensive understanding of single and multivariable calculus.<br>Linear Algebra: Proficiency in matrix operations and solving linear systems.<br>Programming Skills: Ability to write code for numerical algorithms in<br>languages like Python, MATLAB, or C++.<br>Algorithm Design and Analysis: Basic understanding of algorithm complexity<br>and efficiency. |
| 6   | Previous<br>Coursework             | Completion of introductory courses in Mathematics or a related field.   |

#### 2. Competencies

| S/L | Competency   | KSA Description   |
|-----|--------------|---|
| 1.  | Vector space | <ul> <li>Knowledge: <ul> <li>Understanding the definition and properties of vector spaces.</li> <li>Familiarity with subspaces, bases, and dimensions.</li> <li>Grasping the concepts of linear combinations, span, and linear independence.</li> </ul> </li> <li>Skills: <ul> <li>Ability to perform vector operations and transformations.</li> <li>Applying vector space concepts to solve problems in computer graphics and machine learning.</li> <li>Developing algorithms that leverage vector space properties for data representation and manipulation.</li> </ul> </li> <li>Attitude: <ul> <li>Precision in mathematical reasoning and attention to detail.</li> <li>Appreciation for the abstraction and elegance of vector spaces.</li> <li>Openness to exploring advanced topics such as functional</li> </ul> </li> </ul> |



|    |                                       | spaces in quantum computing.   |
|----|---------------------------------------|--|
| -  |                                       | Knowledge:   |
| 2. | Orthogonality<br>and Least<br>Squares | <ul> <li>Comprehending orthogonal vectors, orthogonal complements, and orthonormal bases.</li> <li>Understanding least squares approximation and its applications.</li> <li>Familiarity with QR decomposition and its role in solving least squares problems.</li> <li>Skills:         <ul> <li>Implementing orthogonalization techniques like Gram-Schmidt.</li> <li>Solving least squares problems using numerical methods.</li> <li>Applying orthogonality principles in signal processing and machine learning (e.g., regression analysis).</li> </ul> </li> <li>Attitude:         <ul> <li>Analytical thinking to dissect complex problems into solvable components.</li> <li>Persistence in iterative problem-solving approaches.</li> </ul> </li> </ul>                     |
|    |                                       | Collaboration with others to verify and validate solutions.  |
| 3. | Eigen values and<br>Eigen vectors     | <ul> <li>Knowledge: <ul> <li>Understanding the concepts of Eigen values and eigenvectors.</li> <li>Grasping the spectral theorem and its implications.</li> <li>Familiarity with diagonalization and its applications in systems of linear equations.</li> </ul> </li> <li>Skills: <ul> <li>Computing Eigen values and eigenvectors for various matrices.</li> <li>Applying Eigen value decomposition in dimensionality reduction techniques like PCA.</li> <li>Utilizing Eigen values and eigenvectors in stability analysis and differential equations.</li> </ul> </li> <li>Attitude: <ul> <li>Curiosity about the theoretical foundations of algorithms.</li> <li>Enthusiasm for learning advanced applications in data science and network analysis.</li> </ul> </li> </ul>   |
| 4. | Sampling Theory                       | <ul> <li>Knowledge: <ul> <li>Understanding the principles of random sampling and sampling distributions.</li> <li>Grasping the concept of the Nyquist-Shannon sampling theorem.</li> <li>Familiarity with various sampling techniques (e.g., stratified, systematic).</li> </ul> </li> <li>Skills: <ul> <li>Designing and implementing sampling methods for data collection and analysis.</li> <li>Applying sampling techniques in big data contexts to enhance efficiency.</li> <li>Utilizing sampling theory in Monte Carlo simulations and signal processing.</li> </ul> </li> <li>Attitude: <ul> <li>Critical thinking in evaluating sampling methods and their applicability.</li> <li>Awareness of biases and errors introduced by improper sampling.</li> </ul> </li> </ul> |



|    |           | • Proactiveness in staying updated with emerging sampling  |
|----|-----------|--|
|    |           | techniques and technologies.   |
|    |           | Knowledge:   |
|    |           | <ul> <li>Understanding numerical approximation methods for solving mathematical problems.</li> </ul>         |
|    |           | • Familiarity with numerical linear algebra, interpolation, differentiation, and integration.                |
|    |           | <ul> <li>Knowledge of iterative methods for solving nonlinear equations<br/>and systems.</li> </ul>          |
|    |           | Skills:  |
|    |           | • Implementing numerical algorithms for root-finding,  |
| 5. | Numerical | optimization, and matrix operations.   |
| 5. | Methods   | <ul> <li>Applying numerical methods to simulate real-world phenomena<br/>in scientific computing.</li> </ul> |
|    |           | • Utilizing numerical techniques in solving large-scale computational problems.                              |
|    |           | Attitude:  |
|    |           | • Precision in implementing and testing numerical algorithms.  |
|    |           | <ul> <li>Resilience in dealing with convergence issues and computational errors.</li> </ul>                  |
|    |           | <ul> <li>Commitment to continuous learning and improvement in numerical methodologies.</li> </ul>            |

# 3. Syllabus

| Syllabus   |   |                     |                         |
|--|---|---------------------|-------------------------|
|  | hematics for Emerging T                   | rends               | _                       |
| Course Code  | M23MCS101                                 | CIE Marks           | 50                      |
| Number of Lecture Hours/Week(L: T: P: S)           | (3:0:0:2)                                 | SEE Marks           | 50                      |
| Total Number of Lecture Hours                      | 40 hours Theory                           | Total Marks         | 100                     |
| Credits  | 03  | Exam Hours          | 03                      |
| Course objectives: This course will enable stu     | dents to:                                 |                     |                         |
| 1. Ability to analyse the solution& examine        | its stability in operator the             | eory.               |                         |
| 2. Ability to optimize & solve real life prob      | lems.                                     | •                   |                         |
| 3. Ability to solve image processing & sign        |   |                     |                         |
| 4. Develop the knowledge of Linear Algebr          |   | uations.            |                         |
|  | Module-1                                  |                     |                         |
| Vector Spaces: Vector spaces; subspaces Linea      | rly independent and depen                 | dent vectors Bas    | is and dimension; co    |
| ordinate vectors-Illustrative examples.            | • • •                                     |                     |                         |
| Linear trans formations, Representation of Tr      | ransformations by matrice                 | s.                  |                         |
|  | Module -2                                 |                     |                         |
| Orthogonality and least squares: Inner pr          |   |                     |                         |
| Gram Schmidt orthogonalization process. QR f       |   | s, least square pro | oblems, applications to |
| linear models (least square lines and least square | re fitting of other curves).              |                     |                         |
|  | Module -3                                 |                     |                         |
| Eigen values and Eigenvectors, orthogonal          |   |                     | sition, applications to |
| image processing and statistics, Principal Comp    | oonent Analysis, Different                | ial Equations.      |                         |
|  | Module -4                                 |                     |                         |
| Sampling theory: testing of hypothesis by t-       | -test, X <sup>2</sup> test, F-test., Anal | ysis of Variance    | e (ANOVA): one way      |
| classification.                                    |   | -                   |                         |
|  | -Module -5                                |                     |                         |
| Numerical Methods: Eigen values of real symp       |   | ethod and Housel    | nolder's method. Roots  |
| of polynomial equations-Birge-Vieta method and     | nd Bairstow's method                      |                     |                         |
| Suggested Learning Resources:                      |   |                     |                         |
| Text Books   |   |                     |                         |

1. David C. Lay, Steven R. Layand J. J. McDonald, "Linear Algebra and its Applications", Pearson Education Ltd, 5th Edition 2015.

2. Dr. B.S. Grewal, "Higher Engineering Mathematics", Khanna Publishers, 42<sup>nd</sup> edition, 2012.

# **References Books**

1. Kreyzig, "Advanced Engineering Mathematics".

2. R.E, Walpole, R. H. Myres, S .L. Myres and Keying Ye, "Probability and Statistics for Engineers and Scientists", 9th Edition, Pearson, 2012.

3. T. Veerarajan, Probability, Statistics and Random Process Tata Mc-Graw Hill Co 3rdEdition 2016

4. M K Jain, R. K. Jain, S. R. K. Iyengar, Numerical methods for scientific and engineering New Age International Publishers, Sixth Edition.

## 4. Syllabus Timeline

| S/L | Syllabus Timeline                              | Description   |
|-----|--|---|
| 1   | Week 1-2:<br>Vector Spaces                     | Vector Spaces: Introduction, Vector spaces; subspaces, Problems,<br>Linearly independent and dependent vectors Basis and dimension;<br>Co-ordinate vectors-Illustrative examples. Linear transformations,<br>Representation of Transformations by matrices              |
| 2   | Week 3-4:<br>Orthogonality and<br>leastsquares | Orthogonality and least squares: Introduction, Inner product,<br>orthogonal set, Problems, orthogonal projections, orthogonal bases.<br>Gram Schmid orthogonalization process, QR factorizations of a matrices,<br>Least square problems, applications to linear models |
| 3   | Week 5-6:<br>Eigen values and<br>Eigenvectors  | Eigen values and Eigenvectors- Problems Orthogonal diagonalization-<br>Problems<br>Singular value decompositionProblems Applications to image processing<br>and statistics Principal Component Analysis, Problems Differential<br>Equations.                            |
| 4   | Week 7-8:<br>Sampling theory                   | Sampling theory: Introduction, Testing of hypothesis by t-test, Problems, $\chi^2$ test- Problems, F-test- Problems, Problems, Analysis of Variance (ANOVA): one way classification. Problems   |
| 5   | Week 9-10:<br>Numerical Methods                | Eigen values of real symmetric matrices-Given'smethod, Problem<br>Householder's method., Problems, Roots of polynomial equations-<br>Birge-Vieta method<br>Problems, Bairstow's method, Problems  |
| 6   | Week 11-12:<br>Practical<br>Applications       | Apply learned concepts and competencies to real-world scenarios. Hands-on practice  |

| 5. | <b>Teaching-Learning Process</b> | Strategies |
|----|----------------------------------|------------|
|    |                                  |            |

| S/L | TLP Strategies:                               | Description   |
|-----|---|---|
| 1   | Lecture Method                                | Utilize various teaching methods within the lecture format to reinforce competencies.             |
| 2   | Video/Animation                               | Incorporate visual aids like videos/animations to enhance understanding of Mathematical concepts. |
| 3   | Collaborative<br>Learning                     | Encourage collaborative learning for improved competency application.                             |
| 4   | Higher Order<br>Thinking (HOTS)<br>Questions: | Pose HOTS questions to stimulate critical thinking related to each competency.                    |
| 5   | Problem-Based<br>Learning (PBL)               | Implement PBL to enhance analytical skills and practical application of competencies              |



| 6 | Multiple<br>Representations | Introduce topics in various representations to reinforce competencies   |
|---|-----------------------------|---|
| 7 | Real-World<br>Application   | Discuss practical applications to connect theoretical concepts with real-world competencies                           |
| 8 | Flipped Class<br>Technique  | Utilize a flipped class approach, providing materials before class to facilitate deeper understanding of competencies |
| 9 | Programming<br>Assignments  | Assign programming tasks to reinforce practical skills associated with competencies.                                  |

#### 6. Assessment Details (both CIE and SEE)

Note:

#### Theory Course with 3 Credits: Basic Science Course(BS)

This section of regulations is applicable to all theory-based courses. The minimum CIE marks requirement is 40% of maximum marks in each component.

| CIE Split up for Basic Science Course(B | S) | ) |
|---|----|---|
|---|----|---|

| Components |                                    | Number | Weightage | Max.<br>Marks | Min.<br>Marks |
|------------|------------------------------------|--------|-----------|---------------|---------------|
| (i)        | Internal Assessment-Tests (A)      | 2      | 50%       | 25            | 10            |
| (ii)       | (ii) Assignments/Quiz/Activity (B) |        | 50%       | 25            | 10            |
|            | Total Marks                        | 50     | 20        |               |               |

#### Final CIE Marks =(A) + (B)

Average internal assessment shall be the average of the 2 test marks conducted.

#### Semester End Examinations: PG Programmes

Theory Course with 4, 3 and 2 Credits: Professional Core Course (PC)/Professional Elective/Open Elective

- Question paper pattern will be ten questions. Each question is set for 20marks. The medium of the question paper shall be English unless otherwise it is mentioned.
- There shall be 2 questions from each module, each of the two questions under a module (with a maximum of 3 sub questions), may have mix of topics under that module if necessary.
- The students have to answer 5 full questions selecting one full question from each module.
- Marks scored will be proportionally scaled down to 50 marks

#### 7. Learning Objectives

| S/L | Learning<br>Objectives  | Description  |
|-----|---|--|
| 1   | Understanding<br>Vector Spaces,<br>Sampling &<br>Numerical<br>Methods | Students will grasp concepts of linear independence, basis, and dimension to<br>analyze and manipulate data in high-dimensional spaces .Also principles of<br>selecting representative data subsets to infer properties of the entire population.<br>Understanding Numerical Methods requires knowledge of computational<br>techniques to approximate solutions for complex mathematical problems that<br>cannot be solved analytically. |
| 2   | Designing<br>Vector Spaces,<br>Sampling &<br>Numerical<br>Methods     | Students will able to involves in creating structured, high-dimensional spaces to facilitate efficient data representation and manipulation. Designing Sampling and Numerical Methods entails developing robust statistical and computational algorithms to accurately capture data characteristics and solve mathematical problems effectively.   |
| 3   | Proficiency in<br>Sampling &<br>Numerical<br>Methods                  | Students will become proficient in understanding and applying statistical techniques to accurately represent and analyze large datasets and entails developing and implementing algorithms to solve mathematical problems efficiently and accurately using computational techniques.   |

| 4 | Collaboration<br>and<br>Communication<br>Skills | Students will work collaboratively in teams on design projects, enhancing their ability to communicate effectively, share ideas, and solve problems collectively.  |
|---|---|--|
| 6 | Ethical and<br>Professional<br>Responsibility   | Students will understand the ethical and professional responsibilities associated with digital design, including respecting intellectual property rights, ensuring design reliability and security, and adhering to industry standards and best practices. |

# 8. Course Outcomes (COs) and Mapping with POs/ PSOs

#### Course Outcomes (COs)

| COs   | Course Outcomes (COs)   |  |
|---|---|--|
| M23MCS101.1   | Demonstrate knowledge and understanding of the underlying concepts of Vector space, |  |
|   | Sampling theory   |  |
| M23MCS101.2   | Demonstrate knowledge of the mathematical concepts and computational aspects of     |  |
|   | linear algebra and Numerical methods  |  |
| M23MCS101.3 Analyze domain related engineering problems and develop analytical problems |   |  |
| approach making use of the theoretical concepts   |   |  |

#### **CO-PO-PSO Mapping**

| COs/POs     | <b>PO1</b> | PO2 | PO3 |
|-------------|------------|-----|-----|
| M23MCS101.1 |            |     | 3   |
| M23MCS101.2 |            |     | 3   |
| M23MCS101.3 | 2          |     |     |
| M23MCS101   | 2          | -   | 3   |

#### 9. Assessment Plan

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

|          | CO1 | CO2 | CO3 | Total |  |
|----------|-----|-----|-----|-------|--|
| Module 1 | 06  |     |     | 06    |  |
| Module 2 |     | 06  |     | 06    |  |
| Module 3 | 06  |     | 07  | 13    |  |
| Module 4 |     | 06  | 07  | 13    |  |
| Module 5 | 06  | 06  |     | 12    |  |
| Total    | 18  | 18  | 14  | 50    |  |

Semester End Examination (SEE)

|          | CO1 | CO2 | CO3 | Total |
|----------|-----|-----|-----|-------|
| Module 1 | 20  |     |     | 20    |
| Module 2 |     | 20  |     | 20    |
| Module 3 |     |     | 20  | 20    |
| Module 4 |     | 15  |     | 15    |
| Module 5 | 15  |     | 10  | 25    |
| Total    | 35  | 35  | 30  | 100   |

#### **Conditions for SEE Paper Setting:**

Each module of SEE question paper should be allocated with questions for 20% of the total SEE marks

#### 10. Future with this Subject

The "Advance Mathematics for Emerging Trends "course in the first semester of the MTech program has strong foundation for several future courses in the Postgraduate program. The contributions of this subject extend across various areas, enhancing the students' understanding and skills in the field of computer science. Here are some notable contributions:

#### **Vector Spaces**

#### **Future Prospects and Applications:**

- 1. **Machine Learning and Data Analysis**: Vector spaces are crucial in machine learning, especially in the representation of data. Techniques like word embeddings in natural language processing (NLP) and feature vectors in image processing rely on vector spaces.
- 2. **Quantum Computing**: Quantum states are represented in vector spaces. As quantum computing advances, understanding vector spaces becomes increasingly important for developing quantum algorithms.
- 3. **Robotics and Computer Vision**: Vector spaces facilitate the representation of spatial information and transformations, crucial for navigation and object recognition in robotics and computer vision.

#### Orthogonality and Least Squares Future Prospects and Applications:

- Optimization Algorithms: Least squares methods are fundamental in optimization problems, which are central to machine learning, operations research, and econometrics.
- 2. **Signal Processing**: Orthogonality is a key concept in signal processing, particularly in decomposing signals into orthogonal components, which is important for noise reduction and data compression.
- 3. **Data Fitting and Regression Analysis:** Least squares methods are extensively used in fitting models to data, which is critical in predictive analytics and statistical modeling.

#### Eigen values and Eigenvectors

#### **Future Prospects and Applications:**

- 1. **Principal Component Analysis (PCA)**: PCA is a dimensionality reduction technique that uses eigenvalues and eigenvectors. It is widely used in data science to reduce the complexity of data while preserving essential features.
- 2. **Graph Theory and Network Analysis**: Eigenvalues and eigenvectors play a role in analyzing the properties of graphs and networks, which is important in understanding social networks, communication networks, and biological networks.
- 3. **Stability Analysis**: In control systems and differential equations, eigenvalues are used to analyze the stability of systems, which is critical in engineering and scientific applications.

#### Sampling Theory

#### **Future Prospects and Applications:**

- 1. **Big Data**: Efficient sampling methods are essential for handling large datasets, allowing for the extraction of meaningful insights without processing all the data.
- 2. Monte Carlo Methods: These methods rely on sampling to solve problems in various fields such as finance, physics, and computational biology.
- 3. **Signal and Image Processing**: Sampling theory is fundamental in converting continuous signals into discrete ones, which is crucial in digital signal processing and image processing.

#### **Numerical Methods**

#### **Future Prospects and Applications:**

- 1. Scientific Computing: Numerical methods are the backbone of simulations in physics, chemistry, and engineering, enabling the solution of complex differential equations and optimization problems.
- 2. **Computer Graphics**: Techniques like numerical integration and solving systems of equations are vital in rendering graphics, simulations, and animations.
- 3. **Cryptography**: Numerical methods are used in various cryptographic algorithms and for solving problems related to number theory, which is essential for secure communication.

#### **Integrative Applications in Computer Science Engineering:**

1. Artificial Intelligence (AI) and Machine Learning: The mathematical concepts of vector spaces, least squares, eigen values, and numerical methods are deeply integrated into machine learning algorithms, data pre-processing, and model optimization.

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- 2. Algorithm Development: Efficient algorithms often rely on numerical methods for performance optimization, including those used in real-time systems and large-scale computations.
- 3. Computational Biology: Eigen values and numerical methods are used in the analysis of biological data, modelling biological processes, and understanding genetic information. Understanding these topics not only provides a strong mathematical foundation but also enhances one's ability to develop and implement sophisticated algorithms and models in computer science engineering. The future of these fields looks promising with continuous advancements and new applications emerging in various interdisciplinary areas.



| 1stCom astor             | Professional Core Course (IPC)    |           |
|--------------------------|-----------------------------------|-----------|
| 1 <sup>st</sup> Semester | <b>Essentials of Data Science</b> | M23MCS102 |

# 1. Prerequisites

| SI | Proficiency    | Description  |  |  |  |
|----|----------------|--|--|--|--|
| NO | •              | -  |  |  |  |
| 1  | Programming    | Familiarity with at least one programming language (preferably Python or       |  |  |  |
|    | Fundamentals   | R), including concepts such as data structures, control flow, and functions.   |  |  |  |
| 2  | Statistics and | Understanding of basic statistical concepts, including descriptive statistics, |  |  |  |
|    | Probability    | probability distributions, hypothesis testing, and inferential statistics.     |  |  |  |
| 3  | Linear Algebra | Knowledge of matrix operations, vectors, and linear transformations, as        |  |  |  |
|    |                | these concepts are essential for many data science techniques                  |  |  |  |
| 4  | Calculus       | Familiarity with concepts like derivatives and optimization, which are         |  |  |  |
|    |                | used in various machine learning algorithms and optimization techniques.       |  |  |  |
|    |                | To fully benefit from this course, students should have a basic foundation     |  |  |  |
|    |                | in the following areas   |  |  |  |
| 5  | Programming    | Familiarity with at least one programming language (preferably Python or       |  |  |  |
|    | Fundamentals   | R), including concepts such as data structures, control flow, and functions    |  |  |  |
| 6  | Statistics and | Understanding of basic statistical concepts, including descriptive statistics, |  |  |  |
|    | Probability    | probability distributions, hypothesis testing, and inferential statistics.     |  |  |  |
| 7  | Linear Algebra | Knowledge of matrix operations, vectors, and linear transformations, as        |  |  |  |
|    |                | these concepts are essential for many data science techniques.                 |  |  |  |
| 8  | Calculus       | Familiarity with concepts like derivatives and optimization, which are         |  |  |  |
|    |                | used in various machine learning algorithms and optimization techniques.       |  |  |  |

# 2. Competencies

|    | etencies                           | Description  |  |  |
|----|------------------------------------|--|--|--|
| SI | Competency                         | Description  |  |  |
| NO |                                    |  |  |  |
| 1  | Data Science<br>Fundamentals       | <ul> <li>Knowledge: Comprehensive understanding of data science principles, the data science lifecycle, and the role of data science in various domains.</li> <li>Skills: Ability to identify and articulate data science problems, evaluate the suitability of data science solutions, and communicate effectively with stakeholders.</li> <li>Attitudes: Appreciation for the ethical considerations and societal</li> </ul>   |  |  |
|    |                                    | impacts of data science, and a commitment to responsible data practices.   |  |  |
| 2  | Exploratory Data<br>Analysis (EDA) | <ul> <li>Knowledge: In-depth understanding of EDA techniques, including data visualization, summary statistics, and hypothesis testing.</li> <li>Skills: Proficiency in using statistical programming languages (e.g., R) and data visualization libraries (e.g., ggplot2, matplotlib) to explore and analyze data.</li> <li>Attitudes: Curiosity and willingness to explore data from multiple perspectives, and a critical mindset to identify potential biases or limitations.</li> </ul> |  |  |
| 3  | Machine Learning<br>Algorithms     | <ul> <li>Knowledge: Thorough understanding of supervised and unsupervised machine learning algorithms, their underlying mathematical foundations, and their strengths and limitations.</li> <li>Skills: Ability to implement and evaluate machine learning models using programming languages (e.g., Python, R) and libraries (e.g., scikit-learn, TensorFlow, Keras).</li> </ul>  |  |  |

|   |                                       | • Attitudes: Openness to experiment with different algorithms and techniques, and a commitment to continuously learning and adapting to new developments in the field.   |
|---|---------------------------------------|--|
| 4 | Feature Engineering<br>and Selection  | <ul> <li>Knowledge: Understanding of feature engineering techniques, including feature generation, encoding, scaling, and selection methods.</li> <li>Skills: Ability to apply feature engineering techniques to extract relevant and informative features from raw data, and to select appropriate feature subsets for model training.</li> </ul>   |
|   |                                       | • Attitudes: Creativity and domain expertise to identify and engineer meaningful features, and a data-driven approach to feature selection.  |
| 5 | Data Engineering<br>and Visualization | <ul> <li>Knowledge: Understanding of data engineering concepts, such as distributed computing frameworks (e.g., Hadoop, Spark), and data visualization principles and best practices.</li> <li>Skills: Ability to process and analyze large-scale data using distributed computing techniques, and to create effective and informative data visualizations using tools like D3.js, Tableau, or Power BI.</li> <li>Attitudes: Appreciation for the importance of scalable data processing and clear communication of insights through visualizations</li> </ul> |
| 6 | Social Network<br>Analysis            | <ul> <li>Knowledge: Understanding of social network analysis concepts, including graph theory, centrality measures, community detection algorithms, and network modeling techniques.</li> <li>Skills: Ability to represent and analyze social network data using graph analysis libraries (e.g., NetworkX, igraph) and techniques like node embedding and graph neural networks</li> <li>Attitudes: Curiosity about the underlying patterns and dynamics of</li> </ul>   |
|   |                                       | social networks, and an appreciation for the ethical considerations in analyzing social data.  |

## 3. Syllabus

| . Synabus  |  |                  |     |  |
|--|--|------------------|-----|--|
|  | Essentials of Data Science<br>SEMESTER-I | ce               |     |  |
| Course Code  | M23MCS102                                | <b>CIE Marks</b> | 50  |  |
| Teaching Hours/Week(L:P:SDA)   | 3:0:2:0                                  | SEE Marks        | 50  |  |
| Total Hours of Pedagogy  | 40                                       | Total Marks      | 100 |  |
| Credits  | 04                                       | Exam Hours       | 03  |  |
| <ul> <li>Programming data science concepts and Big Data, modelling using R language.</li> <li>Analyze Basic tools of EDA, Data science process with case studies and Different algorithms.</li> <li>Optimize &amp; solve real life problems with different spam filter.</li> <li>Explore Feature Generation and Feature Selection.</li> </ul> Teaching- Learning Process: Chalk and talk method/Power Point Presentation |  |                  |     |  |
|  | Module-1                                 |                  |     |  |
| Introduction: What is Data Science? Big Data and Data Science hype – and getting past the hype, Why now?<br>– Datafication, Current landscape of perspectives, A data Science Profile, Skill sets. Statistical Inference,<br>Populations and samples, Big Data, new kinds of data, modelling, statistical modelling probability<br>distributions, fitting a model, -Introduction to R                                    |  |                  |     |  |
|  | Module-2                                 |                  |     |  |

Department of Computer Science & Engineering, MIT Mysore

Exploratory Data Analysis and the Data Science Process: Basic tools (plots, graphs and summary statistics) of EDA, Philosophy of EDA, The Data Science Process, Case Study: RealDirect (online real estate firm). Algorithms, machine Learning Algorithms, Three Basic Algorithms: Linear Regression, k-Nearest Neighbours (kNN), k-means, R Programs for the algorithms

#### Module-3

Spam Filter, Linear Regression and Spam Filter, K-NN and spam Filter, Naïve Bayes Algorithm, Spam Filter using Naïve Bayes, Laplace Smoothing, Comparing Naïve Bayes to K-NN, Scraping the Web, introduction to Logical Regression and M6D case study.

#### Module-4

Feature Generation and Feature Selection (Extracting Meaning from Data): Motivating application: user (customer) retention. Feature Generation (brainstorming, role of domain expertise, and place for imagination), Feature Selection algorithms. Filters; Wrappers; Decision Trees; Random Forests. Recommendation Systems: Building a User-Facing Data Product, Algorithmic ingredients of a Recommendation Engine, Dimensionality Reduction, Singular Value Decomposition, Principal Component Analysis, Exercise: build your own recommendation system.

Module-5

Data Engineering, Map reduce, Word Frequency Problem,, Map Reduce Solution, Other Examples of Map Reduce, Pregel-An Introduction. Data Visualization: Basic principles, ideas and tools for data visualization. Mining Social-Network Graphs: Social networks as graphs, Clustering of graphs, Direct discovery of communities in graphs, Partitioning of graphs.

#### PRACTICAL COMPONENT OF IPCC

- Load the Iris dataset as a list of lists (each of the 150 lists should have 5 elements). Compute and print the
  mean and the standard deviation for each of the 4 measurement columns (i.e. sepal length and width, petal
  length and width). Compute and print the mean and the standard deviation for each of the 4 measurement
  columns, separately for each of the three Iris species (Versicolor, Virginica and Setose). Which
  measurement would you consider "best", if you were to guess the Iris species based only on those four
  values?
- Load the MNIST dataset. Create a function that, given a position 1 ≤ k ≤ 10,000, prints the kthdigit of the dataset (i.e.the k row of the csv file) as a grid of 28 × 28 characters. More specifically, you should map each range of pixel values to the following characters:

 $[0, 64) \rightarrow$  " "

 $[64, 128) \rightarrow "."$ 

 $[128, 192) \rightarrow "*"$ 

 $[192,256) \rightarrow "\#"$ 

Compute the Euclidean distance between each pair of the 784-dimensional vectors of the digits at the following positions: 26th, 30th, 32nd, 35th. Based on the distances computed in the previous step and knowing that the digits listed are 7, 0, 1, 1, can you assign the correct label to each of the digits?

3. Split the Iris dataset into two the datasets - IrisTest\_TrainData.csv, IrisTest\_TestData.csv. Read them as two separate data frames named Train\_ Data and Test\_ Data respectively.

Answer the following questions:

• How many missing values are there in Train\_Data?

• What is the proportion of Setosa types in the Test\_Data?

• What is the accuracy score of the K-Nearest Neighbor model (model\_1) with 2/3 neighbors using Train Data and Test Data?

• Identify the list of indices of misclassified samples from the "model\_1". Build a logistic regression model (model\_2) keeping the modelling steps constant. Find the accuracy of the model\_2

4. Demonstrate Decision tree classification model and Evaluate the performance of classifier on Iris dataset.

5. Demonstrate any of the Clustering model and Evaluate the performance on Iris dataset.

# Suggested Learning Resources:

#### Text Books

 Cathy O Neil, Rachel Schutt, 2014, "Doing Data Science-Straight Talk from the Frontline", Orielly
 Jure Leskovek, Anand Rajaraman, Jeffrey Ullman, 2014 Mining of Massive Data Sets, Cambridge University Press

#### **Reference Books**

1. Kevin Murphy, 2013, Machine learning: A Probabilistic Perspective,

2. Peter Bruce, Andre Bruce, Practical Statistics for Data Scientists, Orielly Series

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|--------|--------------------|---|--|--|--|
| S/L    | Syllabus Timeline  | Description   |  |  |  |
| 1      | Weeks 1-3          | Module 1 - Introduction to Data Science                                 |  |  |  |
| 2      | Weeks 4-6:         | Module 2 - Exploratory Data Analysis and Machine Learning Algorithms    |  |  |  |
| 3      | Weeks 7-8:         | Module 3 - Spam Filtering and Web Scraping                              |  |  |  |
| 4      | Weeks 9-11         | Module 4 - Feature Engineering and Recommendation Systems               |  |  |  |
| 5      | Weeks 12-14:       | Module 5 - Data Engineering, Visualization, and Social Network Analysis |  |  |  |
| 6      | Week 15:           | : Final project presentations and course wrap-up.                       |  |  |  |

# 4. Syllabus Timeline

# 5. Teaching-Learning Process Strategies

| S/L | <b>TLP Strategies:</b>                     | Description   |
|-----|--|---|
| 1   | Interactive<br>Lectures                    | Lectures will be designed to be interactive, with frequent opportunities for discussion, questioning, and real-time feedback.   |
| 2   | Hands-on Coding<br>Exercises               | Regular coding exercises and assignments will be provided to reinforce practical skills in data science programming languages and libraries.  |
| 3   | Case Studies and<br>Real-World<br>Examples | The course will incorporate real-world case studies and examples from various domains to illustrate the applications of data science concepts and techniques.                         |
| 4   | Group Projects                             | Students will work in teams to complete a comprehensive data science project, allowing them to apply their knowledge and skills to a real-world problem.                              |
| 5   | Guest Lectures                             | <b>Guest Lectures:</b> Industry experts and practitioners will be invited to share their experiences and insights, providing students with a practical perspective on data science.   |
| 6   | Flipped Classroom                          | Selected topics will be introduced through pre-recorded lectures or reading materials, allowing class time to be dedicated to discussions, exercises, and problem-solving activities. |
| 7   | Peer Learning                              | Students will be encouraged to collaborate, share knowledge, and provide constructive feedback to their peers through study groups, discussion forums, and code reviews.              |

#### 6. Assessment Details (both CIE and SEE)

CIE:

#### Theory Course with 4 credits: Integrated Professional Core Course (IPC) CIE Split up for Integrated Professional Core Course (IPC)

|               | Components                      | Number       | Weightage     | Max.<br>Marks | Min.<br>Marks |
|---------------|---------------------------------|--------------|---------------|---------------|---------------|
|               | Internal Assessment-Tests (A) 2 |              | 60%           | 15            | 06            |
| Theory (A)    | Assignments/Quiz/Activity (B)   | 2            | 40%           | 10            | 04            |
|               | Total Marks                     |              | 100%          | 25            | 10            |
|               | Components                      | Number       | Weightage     | Max.<br>Marks | Min.<br>Marks |
|               | Record Writing                  | Continuous   | 60%           | 15            | 06            |
| Laboratory(B) | Test at the end of the semester | 1            | 40%           | 10            | 04            |
|               | Total Marks                     |              | 100%          | 25            | 10            |
|               | Final CI                        | E Marks =(A) | + <b>(B</b> ) |               |               |

#### SEE:

Theory Course with 4, 3 and 2 Credits: Integrated Profession Core Course (IPC)



- Question paper pattern will be ten questions. Each question is set for 20marks. The medium of the question paper shall be English unless otherwise it is mentioned.
- There shall be 2 question from each module, each of the two questions under a module (with a maximum of 3 sub questions), may have mix of topics under that module if necessary.
- The students have to answer 5 full questions selecting one full question from each module.
- The question paper may include at least one question from the laboratory component.
- Marks scored will be proportionally scaled down to 50 marks

#### 7. Learning Objectives

| S/L | Learning<br>Objectives                        | Description  |  |  |
|-----|---|--|--|--|
| 1   | Data science<br>principles                    | To develop a comprehensive understanding of data science principles, techniques, and applications across various domains.  |  |  |
| 2   | Statistical<br>programming<br>languages       | To equip students with practical skills in using statistical programming languages (e.g., Python, R) and data science libraries for data manipulation, analysis, and modelling.  |  |  |
| 3   | Data analysis                                 | To enable students to perform exploratory data analysis, feature engineering, and data visualization to extract insights from data.  |  |  |
| 4   | Machine<br>learning<br>algorithms             | To provide students with a solid foundation in machine learning algorithms, including supervised and unsupervised learning techniques, for predictive modelling and pattern recognition tasks.   |  |  |
| 5   | Engineering<br>concepts                       | To introduce students to data engineering concepts and techniques for processing and analysing large-scale datasets.   |  |  |
| 6   | Ethical and<br>Professional<br>Responsibility | Students will understand the ethical and professional responsibilities associated with digital design, including respecting intellectual property rights, ensuring design reliability and security, and adhering to industry standards and best practices. |  |  |

#### 8. Course Outcomes and Mapping with POs/ PSOs

|             | Sourse Outcomes and mapping with 1 03/1 003  |  |  |  |
|-------------|--|--|--|--|
| M23MCS102.1 | Apply exploratory data analysis techniques, machine learning algorithms, and feature   |  |  |  |
|             | engineering methods to analyse and model datasets effectively, addressing real-world   |  |  |  |
|             | problems and scenarios.  |  |  |  |
| M23MCS102.2 | Design and implement data engineering pipelines using distributed computing  |  |  |  |
|             | frameworks to process and analyse large-scale datasets, ensuring scalability and efficient   |  |  |  |
|             | data management.   |  |  |  |
| M23MCS102.3 | Evaluate the suitability of data science solutions for specific problems and domains, considering factors such as data quality, computational resources, stakeholder requirements, and ethical implications. |  |  |  |
| M23MCS102.4 | Create effective and informative data visualizations to communicate insights from data to  |  |  |  |
|             | stakeholders, facilitating data-driven decision-making and knowledge dissemination.  |  |  |  |
| M23MCS102.5 | To introduce students to data engineering concepts and techniques for processing and   |  |  |  |
|             | analyzing large-scale datasets.  |  |  |  |

| CO-rO-rSO Mapping |     |     |     |      |      |  |  |
|-------------------|-----|-----|-----|------|------|--|--|
| COs/POs           | PO1 | PO2 | PO3 | PSO1 | PSO2 |  |  |
| M23MCS102.1       | 3   |     | 2   | 3    | 3    |  |  |
| M23MCS102.2       | 3   | 3   | 3   | 2    | 3    |  |  |
| M23MCS102.3       | 2   | 3   |     | 3    | 3    |  |  |
| M23MCS102.4       | 2   | 2   |     | 2    | 3    |  |  |
| M23MCS102.5       | 3   |     | 3   | 3    | 3    |  |  |
| M23MCS102         | 3   | 3   | 3   | 3    | 3    |  |  |

# **CO-PO-PSO Mapping**



|          | CO1 | CO2 | CO3 | CO4 | CO5 | Total |  |
|----------|-----|-----|-----|-----|-----|-------|--|
| Module 1 | 10  |     |     |     |     | 10    |  |
| Module 2 |     | 10  |     |     |     | 10    |  |
| Module 3 |     |     | 10  |     |     | 10    |  |
| Module 4 |     |     |     | 10  |     | 10    |  |
| Module 5 |     |     |     |     | 10  | 10    |  |
| Total    | 10  | 10  | 10  | 10  | 10  | 50    |  |

#### 9. Assessment Plan

**Continuous Internal Evaluation (CIE)** 

#### Semester End Examination (SEE)

|          | CO1 | CO2 | CO3 | CO4 | CO5 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 20  |     |     |     |     | 20    |
| Module 2 |     | 20  |     |     |     | 20    |
| Module 3 |     |     | 20  |     |     | 20    |
| Module 4 |     |     |     | 20  |     | 20    |
| Module 5 |     |     |     |     | 20  | 20    |
| Total    | 20  | 20  | 20  | 20  | 20  | 100   |

#### 10. Future with this Subject

The Essentials of Data Science course provides a solid foundation for students to pursue a variety of future paths within the field of data science and beyond.

- 1. Advanced Data Science Studies: Students can further their education by pursuing graduate programs in data science, machine learning, artificial intelligence, or related fields, enabling them to delve deeper into specialized areas and contribute to cutting-edge research.
- 2. **Industry Roles:** With the skills acquired in this course, students will be well-prepared for roles such as data scientist, machine learning engineer, data analyst, business intelligence analyst, or data engineer in various industries, including technology, finance, healthcare, marketing, and more.
- 3. **Research and Development:** Students interested in research can explore opportunities in academic or industrial research labs, where they can contribute to the development of new data science techniques, algorithms, and applications.
- 4. Entrepreneurship and Innovation: The knowledge and skills gained in this course can empower students to pursue entrepreneurial ventures, developing innovative data-driven products, services, or solutions.
- 5. **Domain-specific Applications:** Data science has applications in numerous domains, such as healthcare, finance, marketing, social sciences, environmental sciences, and more. Students can specialize in a particular domain and apply their data science expertise to solve domain-specific problems.
- 6. **Interdisciplinary Collaboration:** Data science is inherently interdisciplinary, requiring collaboration with experts from various fields. This course will equip students with the ability to effectively communicate and work with domain experts, stakeholders, and cross-functional teams.
- 7. Continuous Learning and Professional Development: The field of data science is rapidly evolving, with new techniques, tools, and applications emerging regularly. This course will instill a mindset of continuous learning and professional development, enabling students to adapt and stay up-to-date with the latest advancements in the field.

#### Future with this Subject: Career Opportunities and Job Profiles

The field of data science is rapidly growing, with an increasing demand for skilled professionals across various industries. Upon successful completion of this course, students will be well-prepared for a wide range of job roles and career opportunities, including.

1. **Data Scientist:** As a data scientist, students will be responsible for collecting, processing, and analyzing large datasets to extract insights and inform data-driven decision-making. They will apply statistical and machine learning techniques to solve complex problems and communicate their findings to stakeholders.

- 2. Machine Learning Engineer: Machine learning engineers focus on developing and deploying machine learning models and algorithms to solve real-world problems. They work on tasks such as data pre-processing, model training, optimization, and integration into production systems.
- 3. **Data Analyst:** Data analysts are responsible for collecting, cleaning, and analyzing data to identify trends, patterns, and insights that can inform business decisions. They use statistical techniques and data visualization tools to communicate their findings effectively.
- 4. **Business Intelligence Analyst:** Business intelligence analysts leverage data to support strategic decision-making within organizations. They design and implement data warehousing and reporting solutions, providing stakeholders with actionable insights for improved business performance.
- 5. **Data Engineer:** Data engineers are responsible for building and maintaining the data infrastructure and pipelines that enable efficient data processing and analysis. They work with distributed computing frameworks, databases, and data warehousing solutions.
- 6. **Quantitative Analyst:** Quantitative analysts apply mathematical and statistical models to analyze financial data and develop trading strategies. They work in the finance and investment sectors, using data science techniques to identify market trends and optimize portfolio performance.
- 7. **Marketing Analyst:** Marketing analysts leverage data science techniques to analyze customer data, understand consumer behavior, and develop targeted marketing strategies. They use predictive modelling and customer segmentation to optimize marketing campaigns and enhance customer experiences.
- 8. **Healthcare Data Analyst:** In the healthcare industry, data analysts work with electronic health records, clinical trial data, and other medical data to identify patterns, improve patient outcomes, and support medical research.
- 9. Social Media Analyst: Social media analysts use data science techniques to analyze social media data, understand user behavior, and develop strategies for social media marketing, customer engagement, and brand management.
- 10. **Consultant and Advisor:** With the increasing demand for data-driven solutions, data science professionals can pursue consulting roles, providing expertise and advisory services to organizations across various industries.



| 1 <sup>st</sup> | Professional Core Course (PC)               | MOOMOGIAO |
|-----------------|---|-----------|
| Semester        | <b>Cyber Security and Digital Forensics</b> | M23MCS103 |

#### 1. Prerequisites

| S/L | Proficiency             | Prerequisites  |  |  |
|-----|-------------------------|--|--|--|
| 1   | Basic Networking        | Understanding of fundamental networking concepts (IP addressing, protocols). |  |  |
| 2   | Computer Systems        | Familiarity with operating systems and hardware components.                  |  |  |
| 3   | Programming Concepts    | Basic programming skills (e.g., Python, C++).                                |  |  |
| 4   | Cyber Security Basics   | Knowledge of basic cyber security principles and terminologies.              |  |  |
| 5   | Legal Knowledge         | Awareness of fundamental cyber laws and regulations.                         |  |  |
| 6   | Analytical Thinking     | Ability to think critically and analytically.                                |  |  |
| 7   | Research Methods        | Basic knowledge of research methodologies and techniques.                    |  |  |
| 8   | Digital Forensics Tools | Familiarity with digital forensics tools (such as FTK Imager, Autopsy).      |  |  |

#### 2. Competencies

| S/L | Competency            | KSA Description   |
|-----|-----------------------|---|
| 1   | Risk Management       | <b>Knowledge</b> : Understanding of risk management frameworks and methodologies. |
|     |                       | Skills: Ability to assess and mitigate risks.                                     |
|     |                       | Attitudes: Valuing proactive risk management in cyber security.                   |
| 2   | Incident Response     | Knowledge: Knowledge of incident response processes and procedures.               |
|     |                       | Skills: Ability to respond to and manage security incidents.                      |
|     |                       | Attitudes: Importance of timely and effective incident response.                  |
| 3   | Encryption and        | Knowledge: Understanding encryption algorithms and decryption                     |
|     | Decryption            | techniques.   |
|     |                       | Skills: Implementing encryption and decryption methods.                           |
|     |                       | Attitudes: Appreciation for data security and privacy.                            |
| 4   | Cyber Laws and Ethics | Knowledge: Understanding of cyber laws and ethical standards.                     |
|     |                       | Skills: Applying legal knowledge to cyber security issues.                        |
|     |                       | Attitudes: Commitment to ethical practices in cyber security.                     |
| 5   | Digital Forensics     | Knowledge: Understanding of digital forensics methodologies and                   |
|     | Techniques            | tools.  |
|     |                       | Skills: Conducting digital forensics investigations.                              |
|     |                       | Attitudes: Valuing accurate and thorough forensic analysis.                       |
| 6   | Network Security      | <b>Knowledge</b> : Understanding network security principles and technologies.    |
|     |                       | Skills: Securing wired and wireless networks.                                     |
|     |                       | Attitudes: Importance of maintaining secure network environments.                 |
| 7   | Legal Frameworks for  | Knowledge: Knowledge of global and Indian cybercrime laws.                        |
|     | Cybercrime            | Skills: Applying legal frameworks to cybercrime cases.                            |
|     |                       | Attitudes: Awareness of the legal implications of cyber activities.               |
| 8   | Forensic Reporting    | Knowledge: Understanding the principles of forensic report writing.               |
|     |                       | Skills: Creating clear and concise forensic reports.                              |
|     |                       | Attitudes: Importance of accurate and professional reporting.                     |

#### 3. Syllabus

| Cyber Security and Digital Forensics |                       |           |    |  |  |  |
|--------------------------------------|-----------------------|-----------|----|--|--|--|
| SEMESTER-I                           |                       |           |    |  |  |  |
| Course Code                          | Course Code M23MCS103 |           |    |  |  |  |
|                                      |                       | CIE Marks | 50 |  |  |  |

| Teaching Hours/Week(L:P:SDA) | 3:0:0:0 | SEE Marks   | 50  |
|------------------------------|---------|-------------|-----|
| Total Hours of Pedagogy      | 40      | Total Marks | 100 |
| Credits                      | 04      | Exam Hours  | 03  |

**Course objectives:** 

- Understand various cyber security issues with respect to operating system, wired and wireless networks.
- Analyze the risks and incidents' response
- Learn the art of encryption and decryption
- Understand the violations in cyber world

Teaching- Learning Process: Chalk and talk method/Power Point Presentation

#### Module-1

Introduction: Introduction to Information Security, what is Cyber Security? Need forCyber Security, Privacy of data, Risk Management, Digital Forensics - Incident response, Security operations.

#### Module-2

Firewalls, Intrusion Detection, Intrusion Prevention Systems, Honeypots, DoS and DDOS attack, Wireless Security issues-Android and iOS Security, App Security, Secure Boot, Data Exfiltration, Wireless Protected Access (WPA), IEEE 802.1x, 802.11i/ WPA2, Wireless Network Threats, Cloud and IoT Application Security.

#### Module-3

The legal perspectives: Cybercrime and legal landscape around the world, Why do we need cyber laws: The Indian context, Indian IT Act, Challenges to Indian Law, Weakness in IT Act, Digital Signatures and Indian IT Act, Amendments to Indian IT Act, Cybercrime and punishment, Policy approaches.

#### **Module-4**

Introduction: Understanding the need of Computer Forensics, Definitions Computer Hardware: Analysis of sources for digital evidence, Digital Media, Hard disk basics, mobile phones

**Forensic Tools:** Forensic hardware, Hardware write/blockers, Hard drive acquisitions, Processing the scene Files and File Systems: Windows file systems, Forensic file images, metadata, File signatures

**Forensic software:** Different software packages, Basic search queries, ASCII, UNICODE, Regular expressions, viewing and managing keywords and cases, Encryption, password protection, Password recovery tools.

**Physical evidence**: fingerprints or other evidence on machines, keyboards

#### Module-5

**Forensic Reports:** Proper report writing, Explaining forensics to the uneducated, **Email analysis:** IP tracking, Tracking and analysis of emails, Webmail, POP, IMAP, **File signature analysis:** File signatures, File extensions, Detecting file manipulation, **Hash Analysis:** Hashing files, Hash libraries, **Window Artifacts:** My documents, Recycle bin, Installed programs, Windows 10 vs. Windows 11

#### Text book:

1. Davis, Philipp, and Cowen, Hacking Exposed: Computer Forensics, McGraw-Hill Education

2. R. Boddington, Practical Digital Forensics, Packt Publishing

#### **Reference Books:**

- 1. N. Jain, D. Kalbande, Digital Forensic: The Fascinating World of Digital Evidences, Wiley
- 2. M.J. Britz, Computer Forensics and Cyber Crime: An Introduction, Perason
- 3. A. J. Marcella, G. Guillossou, Cyber Forensics: from data to digital intelligence, Wiley

#### 4. Syllabus Timeline

| S/L | Syllabus Timeline        | Description   |  |  |  |
|-----|--------------------------|---|--|--|--|
|     |                          | Competency: Understanding Cyber Security Fundamentals             |  |  |  |
|     | 1 Week 1-2: Introduction | Knowledge: Introduction to Information Security, Need for Cyber   |  |  |  |
| 1   |                          | Security, Privacy of Data.  |  |  |  |
| 1   |                          | Skills: Identifying various types of cyber threats and            |  |  |  |
|     |                          | understanding basic security principles.                          |  |  |  |
|     |                          | Attitudes: Valuing the importance of privacy and data protection. |  |  |  |

| 1 |                              | Competency: Mastering Risk Management and Incident Response         |
|---|------------------------------|---|
|   |                              | Knowledge: Risk Management frameworks, Incident Response            |
|   | Week 3-4: Risk Management    | processes.  |
| 2 | and Incident Response        | Skills: Conducting risk assessments, developing incident response   |
|   | and incident Response        | plans.  |
|   |                              | Attitudes: Importance of proactive and reactive measures in cyber   |
|   |                              | security.   |
|   |                              | Competency: Securing Networks                                       |
|   |                              | Knowledge: Firewalls, Intrusion Detection and Prevention            |
|   |                              | Systems, Honeypots, DoS and DDoS Attacks, Wireless Security.        |
| 3 | Week 5-6: Network Security   | Skills: Implementing network security measures, securing            |
|   |                              | wireless networks.  |
|   |                              | Attitudes: Commitment to maintaining secure network                 |
|   |                              | environments.   |
|   |                              | Competency: Understanding Cyber Laws and Ethics                     |
|   |                              | Knowledge: Cybercrime legal landscape, Indian IT Act,               |
| 4 |                              | Cybercrime and punishment, Policy approaches.                       |
| 4 | Week 7-8: Legal Perspectives | Skills: Applying legal knowledge to cyber security cases,           |
|   |                              | understanding digital signatures.                                   |
|   |                              | Attitudes: Commitment to ethical practices in cyber security.       |
|   |                              | Competency: Conducting Digital Forensics Investigations             |
|   |                              | Knowledge: Analysis of digital evidence sources, Forensic tools     |
|   |                              | and software, File systems, Encryption and password recovery        |
| 5 | Week 9-10: Digital Forensics | tools.  |
|   | Techniques                   | Skills: Using forensic tools, analyzing digital evidence, handling  |
|   |                              | encryption.   |
|   |                              | Attitudes: Valuing accurate and thorough forensic analysis.         |
|   |                              | Competency: Developing Forensic Reports and Conducting              |
| 1 |                              | Analysis  |
| 1 | W 1 11 12 F                  | Knowledge: Report writing principles, Email analysis, File          |
| 6 | Week 11-12: Forensic         | signature and hash analysis, Windows artifacts.                     |
|   | Reporting and Analysis       | Skills: Creating forensic reports, analyzing emails and files,      |
| 1 |                              | understanding hashing.  |
|   |                              | Attitudes: Importance of clear, concise, and accurate reporting.    |
| L |                              | reaction and perturber of eleast, contense, and accurate reporting. |

# 5. Teaching-Learning Process Strategies

| S/L | TLP Strategies               | Description   |
|-----|------------------------------|---|
| 1   | Lecture Method               | Utilize various teaching methods within the lecture format to     |
|     |                              | reinforce competencies.   |
| 2   | Video/Animation              | Incorporate visual aids like videos/animations to enhance         |
|     |                              | understanding of concepts.  |
| 3   | Collaborative Learning       | Encourage collaborative learning for improved competency          |
|     |                              | application.  |
| 4   | Higher Order Thinking (HOTS) | Pose HOTS questions to stimulate critical thinking related to     |
|     | Questions                    | each competency.  |
| 5   | Problem-Based Learning (PBL) | Implement PBL to enhance analytical skills and practical          |
|     |                              | application of competencies.                                      |
| 6   | Real-World Application       | Discuss practical applications to connect theoretical concepts    |
|     |                              | with real-world competencies.                                     |
| 7   | Flipped Class Technique      | Utilize a flipped class approach providing materials before class |
|     |                              | to facilitate deeper understanding.                               |
| 8   | Practical Assignments        | Assign hands-on tasks to reinforce practical skills associated    |
|     |                              | with competencies.  |



#### 6. Assessment Details (both CIE and SEE)

- CIE:
- Theory Course with 4 Credits: Profession Core Course (PC)
- This section of regulations is applicable to all theory-based courses. The minimum CIE marks requirement is 40% of maximum marks in each component.

|      | CIE Split up for Frolessional Course (FC) |        |           |               |               |  |
|------|---|--------|-----------|---------------|---------------|--|
|      | Components                                | Number | Weightage | Max.<br>Marks | Min.<br>Marks |  |
| (i)  | Internal Assessment-Tests (A)             | 2      | 50%       | 25            | 10            |  |
| (ii) | Assignments/Quiz/Activity (B)             | 25     | 10        |               |               |  |
|      | Total Mar                                 | 50     | 20        |               |               |  |

#### Final CIE Marks =(A) + (B)

Average internal assessment shall be the average of the 2 test marks conducted. **SEE:** 

# Theory Course with 4, 3 and 2 Credits: Professional Core Course (PC)/Professional Elective/Open Elective

- Question paper pattern will be ten questions. Each question is set for 20marks. The medium of the question paper shall be English unless otherwise it is mentioned.
- There shall be 2 question from each module, each of the two questions under a module (with a maximum of 3 sub questions), may have mix of topics under that module if necessary.
- The students have to answer 5 full questions selecting one full question from each module.
- Marks scored will be proportionally scaled down to 50 marks

#### 7. Learning Objectives

| S/L | Learning Objectives                                    | Description  |  |  |  |
|-----|--|--|--|--|--|
| 1   | Understanding Cyber<br>Security Fundamentals           | Students will grasp the fundamental concepts of cyber security<br>including the importance of information security, the need for cyber<br>security, and data privacy principles.       |  |  |  |
| 2   | Mastering Risk Management<br>and Incident Response     | Students will learn to conduct risk assessments, develop incident<br>response plans, and understand the importance of proactive and<br>reactive measures in cyber security.            |  |  |  |
| 3   | Proficiency in Network<br>Security                     | Students will become proficient in implementing network security<br>measures, securing wired and wireless networks, and understanding<br>various network security threats.             |  |  |  |
| 4   | Understanding Cyber Laws<br>and Ethics                 | Students will gain knowledge of the legal landscape of cybercrime,<br>including the Indian IT Act, and understand the importance of<br>ethical practices in cyber security.            |  |  |  |
| 5   | Conducting Digital<br>Forensics Investigations         | Students will learn to use forensic tools, analyze digital evidence,<br>and handle encryption, gaining skills in conducting thorough and<br>accurate forensic investigations.          |  |  |  |
| 6   | Developing Forensic Reports<br>and Conducting Analysis | Students will understand the principles of forensic report writing,<br>learn to analyze emails and files, and conduct hash analysis,<br>emphasizing the importance of clear reporting. |  |  |  |
| 7   | Applying Cyber Security<br>Tools                       | Through hands-on practice, students will apply their knowledge of<br>cyber security tools, implement security measures, and conduct<br>practical forensic investigations.              |  |  |  |
| 8   | Analyzing Cyber Security<br>Case Studies               | Students will analyze real-world cyber security and legal case<br>studies to understand the application of concepts in practical<br>scenarios.   |  |  |  |
| 9   | Enhancing Analytical and<br>Problem-Solving Skills     | Students will enhance their analytical and problem-solving skills<br>through practical assignments, projects, and collaborative learning<br>activities.                                |  |  |  |
| 10  | Ethical and Professional Responsibility                | Students will understand their ethical and professional responsibilities in the field of cyber security, including data privacy,   |  |  |  |



|  | intellectual property rights, and adherence to industry standards and | ĺ |
|--|---|---|
|  | best practices.   |   |

#### 8. Course Outcomes (COs) and Mapping with POs/ PSOs

#### Course Outcomes (Co's)

| COs         | Description  |
|-------------|--|
| M23MCS103.1 | Understand the fundamentals of cyber security and forensics. |
| M23MCS103.2 | Apply risk management and incident response strategies.      |
| M23MCS103.3 | Analyze and implement encryption and decryption techniques.  |
| M23MCS103.4 | Evaluate network security issues and solutions.              |
| M23MCS103.5 | Develop comprehensive forensic reports and legal knowledge.  |

#### **CO-PO-PSO Mapping**

| CO's        | PO1 | PO2 | PO3 | PSO1 | PSO2 |
|-------------|-----|-----|-----|------|------|
| M23MCS103.1 | 3   |     | 3   | 3    | 3    |
| M23MCS103.2 | 3   |     | 3   | 3    | 3    |
| M23MCS103.3 | 3   | 3   | 3   | 3    | 3    |
| M23MCS103.4 | 3   |     | 3   | 3    | 3    |
| M23MCS103.5 | 3   |     | 3   | 3    | 3    |
| M23MCS103   | 3   | 3   | 3   | 3    | 3    |

#### 9. Assessment Plan

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

| COs      | CO1 | CO2 | CO3 | CO4 | CO5 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 10  |     |     |     |     | 10    |
| Module 2 |     | 10  |     |     |     | 10    |
| Module 3 |     |     | 10  |     |     | 10    |
| Module 4 |     |     |     | 10  |     | 10    |
| Module 5 |     |     |     |     | 10  | 10    |
| Total    | 10  | 10  | 10  | 10  | 10  | 50    |

#### Semester End Examination (SEE):

| COs      | CO1 | CO2 | CO3 | CO4 | CO5 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 20  |     |     |     |     | 20    |
| Module 2 |     | 20  |     |     |     | 20    |
| Module 3 |     |     | 20  |     |     | 20    |
| Module 4 |     |     |     | 20  |     | 20    |
| Module 5 |     |     |     |     | 20  | 20    |
| Total    | 20  | 20  | 20  | 20  | 20  | 100   |

#### **Conditions for SEE Paper Setting:**

Each module of SEE question paper should be allocated with questions for 20% of the total SEE marks.

# 10. Future with this Subject

This course provides a solid foundation for advanced studies and careers in cyber security and digital forensics. Key areas include:

- 1. Advanced Cyber Security Courses: Further studies in penetration testing, advanced cryptography, and secure software development.
- 2. Digital Forensics Specializations: Specializing in mobile forensics, network forensics, and malware analysis.



- 3. Cyber Law and Ethics: Understanding the legal and ethical dimensions of cyber security practices.
- 4. Industry Applications: Careers in cyber security operations, incident response, and risk management.
- 5. **Research and Development:** Contributing to innovative solutions and advancements in cyber security and digital forensics. This syllabus aims to equip students with the necessary skills, knowledge, and attitudes to excel in the dynamic and critical fields of cyber security and digital forensics.



| 1 St C                   | <b>Professional Core Course (PC)</b> |           |
|--------------------------|--------------------------------------|-----------|
| 1 <sup>st</sup> Semester | Advanced Database Management System  | M23MCS104 |

#### 1. Prerequisites

| S/L | Proficiency                                      | Prerequisites   |  |  |  |  |
|-----|--|---|--|--|--|--|
| 1   | Basic Database<br>Concepts                       | Understanding of basic database concepts such as tables, records, fields, and relationships. Familiarity with different types of databases (relational, NoSQL, etc.). |  |  |  |  |
| 2   | SQL Proficiency                                  | Strong knowledge of SQL (Structured Query Language) including queries, joins, sub queries, indexing, and transactions.  |  |  |  |  |
| 3   | Data Modelling                                   | Ability to design and understand Entity-Relationship (ER) diagrams.<br>Knowledge of normalization and denormalization techniques.                                     |  |  |  |  |
| 4   | Basic<br>Understanding of<br>Web<br>Technologies | Familiarity with web development basics and how web applications interact with databases  |  |  |  |  |
| 5   | Concurrency and Transactions                     | Knowledge of concurrency control, isolation levels, and transaction management.   |  |  |  |  |
| 6   | Fundamentals of<br>Data Structures               | Understanding of basic data structures like arrays, linked lists, trees, and hash tables.   |  |  |  |  |

#### 2. Competencies

| S/L | Competency                     | KSA Description  |
|-----|--------------------------------|--|
| 1   | Advanced SQL<br>Skills         | <ul> <li>Knowledge: Mastery in writing complex SQL queries, including advanced joins, sub queries, and set operations</li> <li>Skills: Proficiency in creating and managing stored procedures, functions, and triggers</li> <li>Attitude: Ability to optimize query performance and tuning indexes to improve throughput and reduce latency in distributed SQL databases.</li> </ul> |
| 2   | Database<br>Optimization:      | <ul><li>Knowledge: Skills in performance tuning and query optimization.</li><li>Skills: Understanding of indexing strategies and execution plans.</li><li>Attitude: Knowledge of backup and disaster recovery strategies in SQL databases to ensure data durability and availability.</li></ul>  |
| 3   | Database design and modelling: | Expertise in designing scalable and efficient database schemas.<br>Proficiency in normalization, denormalization, and data modelling techniques.   |
| 4   | Database security              | <b>Knowledge</b> : Competence in implementing database security measures, including user authentication, authorization, and encryption Skills in managing roles and permissions<br><b>Skills</b> : Acquire knowledge of feature engineering techniques to extract relevant information from raw data and improve model performance.  |
| 5   | Backup and recovery            | <ul> <li>Knowledge: Proficiency in database backup and recovery strategies.</li> <li>Skills: Regularly check the integrity of backups. Corruption or incomplete backups can render recovery impossible.</li> <li>Attitude: Ability to implement disaster recovery plans and perform point-in-time recovery.</li> </ul>   |
| 6   | NoSQL and<br>NewSQL Databases: | <ul> <li>Knowledge: Familiarity with NoSQL databases (e.g., MongoDB, Cassandra) and their data models.</li> <li>Skills: Experience in setting up, configuring, and maintaining NoSQL databases, including backup strategies, monitoring, and scaling.</li> <li>Attitude: Understanding of NewSQL database and their use cases.</li> </ul>  |

# 3. Syllabus

ADVANCE Database Management System

| Course Code   | SEMESTER-I   |   |   |
|---|--|---|---|
| Course Code   | M23MCS104  | CIE Marks   | 50  |
| Teaching Hours/Week (L:T:P:S)   | 2:0:0:0  | SEE Marks   | 50  |
| Total Hours of Pedagogy   | 40   | Total Marks   | 100   |
| Credits   | 03   | Exam Hours  | 03  |
| Course objectives:  |  |   |   |
| <ul> <li>To understand the appropriate high-p</li> <li>To understand the Infer and represe</li> <li>To understand the Interpret rule set</li> <li>To understand the Discover and des</li> </ul> Teaching- Learning Process: Chalk a   | nt the real-world data using object-<br>in the database to implement data v<br>ign database for recent applications  | oriented database<br>varehousing of mining<br>s database for better inter   | operability   |
| Teaching Learning Trocess. Chark e  | Module-1   | itution.  |   |
| <b>Review of Relational Data Model</b><br>Relational model constraints and relation<br>Relational Databases: Overview of O<br>Object definition Language ODL; Obj   | and Relational Database Const<br>onal database schemas; Update oper<br>OP; Complex objects; Identity, str  | rations, anomalies. Object<br>ucture etc. Object model  | t and Object<br>l of ODMG   |
|   | Module-2   |   |   |
| Disk Storage, Basic File Structures,<br>Storage Devices, Buffering of Blocks,<br>Records (Heap Files), Files of Orde<br>Organizations, Parallelizing Disk Acco<br>Database Concepts: Distributed Da   | , Placing File Records on Disk Op<br>red Records (Sorted Files), Hash<br>ess Using RAID Technology, Mode<br>tabase Concepts, Data Fragmen  | erations on Files, Files o<br>ing Techniques, Other I<br>ern Storage Architectures  | f Unordered<br>Primary File<br>. Distributed  |
|   | Module-3   |   |   |
| NOSQL Systems, NOSQL Graph Data   |  |   |   |
| Big Data Technologies Based on Map<br>Hadoop, Hadoop Distributed File Syst  | Reduce and Hadoop: What Is Big I<br>em (HDFS).   | Data? Introduction to Map   | Reduce and  |
| Big Data Technologies Based on Map<br>Hadoop, Hadoop Distributed File Syst  | Reduce and Hadoop: What Is Big E<br>em (HDFS).<br><b>Module-4</b>  |   |   |
| Big Data Technologies Based on Map  | Reduce and Hadoop: What Is Big E<br>em (HDFS).<br>Module-4<br>t and Data Mining: Decision-Supp<br>Rules, Other Types of Association<br>Ranking Using Terms, Releva   | port Systems, Data Wareh<br>as, Clustering, Other For<br>nce Using Hyperlinks,  | ousing, Data<br>rms of Data<br>Synonyms,  |
| Big Data Technologies Based on Map<br>Hadoop, Hadoop Distributed File Syst<br><b>Data Warehousing, Decision Suppor</b><br>Mining, Classification, Association F<br>Mining,<br><b>Information Retrieval:</b> Relevance<br>Homonyms, and Ontologies, Indexin  | Reduce and Hadoop: What Is Big E<br>em (HDFS).<br>Module-4<br>t and Data Mining: Decision-Supp<br>Rules, Other Types of Association<br>Ranking Using Terms, Releva   | port Systems, Data Wareh<br>as, Clustering, Other For<br>nce Using Hyperlinks,  | ousing, Data<br>rms of Data<br>Synonyms,  |
| Big Data Technologies Based on Map<br>Hadoop, Hadoop Distributed File Syst<br><b>Data Warehousing, Decision Suppor</b><br>Mining, Classification, Association F<br>Mining,<br><b>Information Retrieval:</b> Relevance<br>Homonyms, and Ontologies, Indexin<br>Indexing the Web.<br><b>Database Security:</b> Introduction to d<br>and revoking privileges, Mandatory ac<br>injection, introduction to statistical dar<br>Infrastructure, privacy issues and prese   | Reduce and Hadoop: What Is Big E<br>em (HDFS).<br>Module-4<br>t and Data Mining: Decision-Supp<br>Rules, Other Types of Association<br>Ranking Using Terms, Relevan<br>ng of Documents, Measuring Re<br>Module-5<br>ata base security issues, Discretion<br>ccess control and role based access<br>ta base security, introduction to flo   | port Systems, Data Wareh<br>as, Clustering, Other For<br>nce Using Hyperlinks,<br>trieval Effectiveness, Ci<br>ary access control based<br>control for multilevel se<br>w control, encryption an  | ousing, Data<br>rms of Data<br>Synonyms,<br>rawling and<br>on granting<br>ecurity, SQL  |
| Big Data Technologies Based on Map<br>Hadoop, Hadoop Distributed File Syst<br><b>Data Warehousing, Decision Suppor</b><br>Mining, Classification, Association F<br>Mining,<br><b>Information Retrieval:</b> Relevance<br>Homonyms, and Ontologies, Indexin<br>Indexing the Web.<br><b>Database Security:</b> Introduction to d<br>and revoking privileges, Mandatory ad<br>injection, introduction to statistical dai<br>Infrastructure, privacy issues and press<br><b>Suggested Learning Resources:</b>   | Reduce and Hadoop: What Is Big E<br>em (HDFS).<br>Module-4<br>t and Data Mining: Decision-Supp<br>Rules, Other Types of Association<br>Ranking Using Terms, Relevan<br>ng of Documents, Measuring Re<br>Module-5<br>ata base security issues, Discretion<br>ccess control and role based access<br>ta base security, introduction to flo   | port Systems, Data Wareh<br>as, Clustering, Other For<br>nce Using Hyperlinks,<br>trieval Effectiveness, Ci<br>ary access control based<br>control for multilevel se<br>w control, encryption an  | ousing, Data<br>rms of Data<br>Synonyms<br>rawling and<br>on granting<br>ecurity, SQL   |
| Big Data Technologies Based on Map<br>Hadoop, Hadoop Distributed File Syst<br><b>Data Warehousing, Decision Suppor</b><br>Mining, Classification, Association F<br>Mining,<br><b>Information Retrieval:</b> Relevance<br>Homonyms, and Ontologies, Indexin<br>Indexing the Web.<br><b>Database Security:</b> Introduction to d<br>and revoking privileges, Mandatory ac<br>injection, introduction to statistical dai<br>Infrastructure, privacy issues and press<br><b>Suggested Learning Resources:</b><br><b>Books</b><br>1 Fundamentals of Database Systems<br>2 Database Management Systems Rag<br><b>Reference Books</b><br>1. Database System Concepts Abrahar<br>2010 | Reduce and Hadoop: What Is Big E<br>em (HDFS).<br>Module-4<br>t and Data Mining: Decision-Supp<br>Rules, Other Types of Association<br>Ranking Using Terms, Releva<br>ng of Documents, Measuring Re<br>Module-5<br>ata base security issues, Discretion<br>ccess control and role based access<br>ta base security, introduction to flo<br>ervation, challenges to maintaining<br>Elmasri and Navathe Pearson Edu<br>hu Rama Krishnan and Johannes G<br>n Silberschatz, Henry F. Korth, S. S | port Systems, Data Wareh<br>as, Clustering, Other For<br>nce Using Hyperlinks,<br>trieval Effectiveness, Ci-<br>ary access control based<br>control for multilevel se<br>w control, encryption an<br>database security.   | ousing, Data<br>rms of Data<br>Synonyms<br>rawling and<br>on granting<br>ecurity, SQL<br>d public key<br>d Edition                |
| Big Data Technologies Based on Map<br>Hadoop, Hadoop Distributed File Syst<br><b>Data Warehousing, Decision Suppor</b><br>Mining, Classification, Association F<br>Mining,<br><b>Information Retrieval:</b> Relevance<br>Homonyms, and Ontologies, Indexin<br>Indexing the Web.<br><b>Database Security:</b> Introduction to d<br>and revoking privileges, Mandatory ac<br>injection, introduction to statistical dai<br>Infrastructure, privacy issues and press<br><b>Suggested Learning Resources:</b><br><b>Books</b><br>1 Fundamentals of Database Systems<br>2 Database Management Systems Rag<br><b>Reference Books</b><br>1. Database System Concepts Abrahar         | Reduce and Hadoop: What Is Big E<br>em (HDFS).<br>Module-4<br>t and Data Mining: Decision-Supp<br>Rules, Other Types of Association<br>Ranking Using Terms, Releva<br>ng of Documents, Measuring Re<br>Module-5<br>ata base security issues, Discretion<br>ccess control and role based access<br>ta base security, introduction to flo<br>ervation, challenges to maintaining<br>Elmasri and Navathe Pearson Edu<br>hu Rama Krishnan and Johannes G<br>n Silberschatz, Henry F. Korth, S. S | port Systems, Data Wareh<br>as, Clustering, Other For<br>nce Using Hyperlinks,<br>trieval Effectiveness, Ci-<br>ary access control based<br>control for multilevel se<br>w control, encryption an<br>database security.   | ousing, Data<br>rms of Data<br>Synonyms<br>rawling and<br>on granting<br>ecurity, SQL<br>d public key<br>d Edition                |
| Big Data Technologies Based on Map<br>Hadoop, Hadoop Distributed File Syst<br><b>Data Warehousing, Decision Suppor</b><br>Mining, Classification, Association F<br>Mining,<br><b>Information Retrieval:</b> Relevance<br>Homonyms, and Ontologies, Indexin<br>Indexing the Web.<br><b>Database Security:</b> Introduction to d<br>and revoking privileges, Mandatory ac<br>injection, introduction to statistical dai<br>Infrastructure, privacy issues and press<br><b>Suggested Learning Resources:</b><br><b>Books</b><br>1 Fundamentals of Database Systems<br>2 Database Management Systems Rag<br><b>Reference Books</b><br>1. Database System Concepts Abrahar<br>2010 | Reduce and Hadoop: What Is Big E<br>em (HDFS).<br>Module-4<br>t and Data Mining: Decision-Supp<br>Rules, Other Types of Association<br>Ranking Using Terms, Releva<br>ng of Documents, Measuring Re<br>Module-5<br>ata base security issues, Discretion<br>ccess control and role based access<br>ta base security, introduction to flo<br>ervation, challenges to maintaining<br>Elmasri and Navathe Pearson Edu<br>hu Rama Krishnan and Johannes G<br>n Silberschatz, Henry F. Korth, S. S | port Systems, Data Wareh<br>as, Clustering, Other For<br>nce Using Hyperlinks,<br>trieval Effectiveness, Ci-<br>wary access control based<br>control for multilevel se<br>w control, encryption an<br>database security.<br>ucation 2013<br>Gehrke McGraw-Hill 3rd<br>Sudarshan McGraw Hill 6<br>Gehrke McGraw-Hill 3rd | ousing, Data<br>rms of Data<br>Synonyms<br>rawling and<br>on granting<br>con granting<br>curity, SQI<br>d public key<br>d Edition |

|   | [   |  |
|---|---|--|
| 1 | Week 1-2:Review of<br>Relational Data<br>Model and<br>Relational Database<br>Constraints                | Relational model concepts; Relational model constraints and relational<br>database schemas; Update operations, anomalies. Object and Object-<br>Relational Databases: Overview of OOP; Complex objects; Identity,<br>structure etc. Object model of ODMG, Object definition Language ODL;<br>Object Query Language OQL; Conceptual design of Object database   |
| 2 | Week 3-4:<br>Disk Storage, Basic<br>File Structures,<br>Hashing, and<br>Modern Storage<br>Architectures | Introduction, Secondary Storage Devices, Buffering of Blocks, Placing File<br>Records on Disk Operations on Files, Files of Unordered Records (Heap<br>Files), Files of Ordered Records (Sorted Files), Hashing Techniques, Other<br>Primary File Organizations, Parallelizing Disk Access Using RAID<br>Technology, Modern Storage Architectures. Distributed Database Concepts:<br>Distributed Database Concepts, Data Fragmentation, Replication, and<br>Allocation Techniques. |
| 3 | Week 5-6:<br>NOSQL Databases<br>and Big Data<br>Storage Systems   | Introduction to NOSQL Systems, The CAP Theorem, Document-Based<br>NOSQL Systems and MongoDB, NOSQL Key-Value Stores, Column-<br>Based or Wide Column NOSQL Systems, NOSQL Graph Databases and<br>Neo4j.<br>Big Data Technologies Based on Map Reduce and Hadoop: What Is Big<br>Data? Introduction to Map Reduce and Hadoop, Hadoop Distributed File<br>System (HDFS).   |
| 4 | Week 7-8:Data<br>Warehousing,<br>Decision Support<br>and Data Mining                                    | Decision-Support Systems, Data Warehousing, Data Mining, Classification,<br>Association Rules, Other Types of Associations, Clustering, Other Forms of<br>Data Mining,<br>Relevance Ranking Using Terms, Relevance Using Hyperlinks, Synonyms,<br>Homonyms, and Ontologies, Indexing of Documents, Measuring Retrieval<br>Effectiveness, Crawling and Indexing the Web.  |
| 5 | Week 9-10:<br>Database Security   | Introduction to data base security issues, Discretionary access control based<br>on granting and revoking privileges, Mandatory access control and role<br>based access control for multilevel security, SQL injection, introduction to<br>statistical data base security, introduction to flow control, encryption and<br>public key Infrastructure, privacy issues and preservation, challenges to<br>maintaining database security.   |

# 5. Teaching-Learning Process Strategies

| S/L | TLP Strategies:  | Description   |  |  |  |  |
|-----|--|---|--|--|--|--|
| 1   | Lecture Method   | Utilize various teaching methods within the lecture format to reinforce competencies.                                 |  |  |  |  |
| 2   | 2 Image/Video/Animation Incorporate visual aids like image/videos/animations to enhance understanding of programming constructs. |   |  |  |  |  |
| 3   | Collaborative Learning   | Encourage collaborative learning for improved competency application.   |  |  |  |  |
| 4   | Higher Order Thinking<br>(HOTS) Questions:   | Pose HOTS questions to stimulate critical thinking related to each competency.  |  |  |  |  |
| 5   | Programming-Based<br>Learning (PBL)  | Implement PBL to enhance analytical skills and practical application of competencies                                  |  |  |  |  |
| 6   | Real-World Application   | Discuss practical applications to connect theoretical concepts with real-<br>world competencies.                      |  |  |  |  |
| 7   | Flipped Class<br>Technique   | Utilize a flipped class approach, providing materials before class to facilitate deeper understanding of competencies |  |  |  |  |
| 8   | Programming<br>Assignments   | Assign programming tasks to reinforce practical skills associated with competencies.                                  |  |  |  |  |



#### 6. Assessment Details (both CIE and SEE)

#### CIE:

#### Theory Course with 3 Credits: Professional Core Course (PC)

This section of regulations is applicable to all theory-based courses. The minimum CIE marks requirement is 40% of maximum marks in each component.

|      | CIE Split up for Professional Course (PC) |        |           |               |               |  |  |  |
|------|---|--------|-----------|---------------|---------------|--|--|--|
|      | Components                                | Number | Weightage | Max.<br>Marks | Min.<br>Marks |  |  |  |
| (i)  | Internal Assessment-Tests (A)             | 2      | 50%       | 25            | 10            |  |  |  |
| (ii) | Assignments/Quiz/Activity (B)             | 2      | 50%       | 25            | 10            |  |  |  |
|      | TotalMa                                   | 50     | 20        |               |               |  |  |  |

#### FinalCIEMarks =(A) + (B)

Average internal assessment shall be the average of the 2 test marks conducted. **SEE:** 

- Question paper pattern will be ten questions. Each question is set for 20marks. The medium of the question paper shall be English unless otherwise it is mentioned.
- There shall be 2 question from each module, each of the two questions under a module (with a maximum of 3 sub questions), may have mix of topics under that module if necessary.
- The students have to answer 5 full questions selecting one full question from each module.
- Marks scored will be proportionally scaled down to 50 marks

#### 7. Learning Objectives

| S/L | Learning Objectives  | Description   |
|-----|--|---|
| 1   | Understanding<br>fundamentals of C++<br>Programming Constructs | Students will grasp the fundamental concepts of C++ Programming, including basic constructs.  |
| 2   | Executing Simple C++<br>Programs                               | Students will learn to design and execute basic and simple C++ programs.  |
| 3   | Programming-Based<br>Learning                                  | Through program execution-based learning, students will undergo the demonstration of C++ programming constructs working principles.   |
| 4   | Proficiency in C++<br>Specific Constructs                      | Students will become proficient in understanding and applying the C++ specific constructs to improve the efficiency of C++programming logics.   |
| 5   | Ethical and Professional<br>Responsibility                     | Students will understand the ethical and professional responsibilities associated with C++ Programming, including respecting intellectual property rights, ensuring design reliability and security, and adhering to industry standards and best practices. |

# 8. Course Outcomes (COs) and Mapping with POs/ PSOs

#### Course Outcomes (COs)

| COs         | Description   |
|-------------|---|
| M23MCS104.1 | Select the appropriate high-performance database like parallel and distributed database.  |
| M23MCS104.2 | Infer and represent the real-world data using object-oriented database                    |
| M23MCS104.3 | Interpret rule set in the database to implement data warehousing of mining                |
| M23MCS104.4 | Discover and design database for recent applications database for better interoperability |
| M23MCS104.5 | Apply access control and role based access control for multilevel security                |

#### **CO-PO-PSO** Mapping

| COs/POs | PO1 | PO2 | PO3 | PSO1 | PSO2 |
|---------|-----|-----|-----|------|------|
|         |     |     |     |      |      |



| M23MCS104.1 | 3 |   |   | 3 | 3 |
|-------------|---|---|---|---|---|
| M23MCS104.2 | 3 |   |   | 3 | 3 |
| M23MCS104.3 |   | 3 |   | 3 | 3 |
| M23MCS104.4 |   |   | 3 | 3 | 3 |
| M23MCS104.5 | 3 | 3 |   | 3 | 3 |
| M23MCS104   | 3 | 3 | 3 | 3 | 3 |

#### 9. Assessment Plan

| Continuous Internal Evaluation (CIE) |     |         |      |       |     |       |  |
|--------------------------------------|-----|---------|------|-------|-----|-------|--|
|                                      | CO1 | CO2     | CO3  | CO4   | CO4 | Total |  |
| Module 1                             | 10  |         |      |       |     | 10    |  |
| Module 2                             |     | 10      |      |       |     | 10    |  |
| Module 3                             |     |         | 10   |       |     | 10    |  |
| Module 4                             |     |         |      | 10    |     | 10    |  |
| Module 5                             |     |         |      |       | 10  | 10    |  |
| Total                                | 10  | 10      | 10   | 10    | 10  | 50    |  |
|                                      |     | - C / T | 1.1. | (CEE) | •   | •     |  |

Semester End Examination (SEE)

|          |     |     | na Bhainnatio | (   |     |       |
|----------|-----|-----|---------------|-----|-----|-------|
|          | CO1 | CO2 | CO3           | CO4 | CO5 | Total |
| Module 1 | 20  |     |               |     |     | 20    |
| Module 2 |     | 20  |               |     |     | 20    |
| Module 3 |     |     | 20            |     |     | 20    |
| Module 4 |     |     |               | 20  |     | 20    |
| Module 5 |     |     |               |     | 20  | 20    |
| Total    | 20  | 20  | 20            | 20  | 20  | 100   |

#### **Conditions for SEE Paper Setting:**

Each module of SEE question paper should be allocated with questions for 20% of the total SEE marks **10. Future with this Subject** 

#### 1. Integration of Artificial Intelligence and Machine Learning

- **Predictive Analytics**: DBMS will increasingly integrate AI and ML to provide predictive analytics, helping businesses make data-driven decisions.
- Automated Database Management: AI will enable automated tuning, indexing, and query optimization, reducing the need for manual intervention.

#### 2. Cloud-Based Databases

- Scalability and Flexibility: Cloud-based DBMS will continue to grow, offering scalable and flexible solutions that can handle large volumes of data.
- **Cost Efficiency**: Pay-as-you-go models will make it more cost-efficient for businesses to manage their data infrastructure.

#### 3. Distributed Databases and Edge Computing

- **Distributed Architectures**: There will be a rise in distributed databases that can handle data across multiple locations, improving performance and reliability.
- Edge Computing: Storing and processing data closer to the source (edge devices) will reduce latency and improve real-time data processing capabilities.



| 1st Comorton             | <b>Professional Core Course (PC)</b>         | MOOMOGIAE |
|--------------------------|--|-----------|
| 1 <sup>st</sup> Semester | Artificial Intelligence and Machine Learning | M23MCS105 |

| 1. Pr | erequisites                                 |   |
|-------|---|---|
| S/L   | Proficiency                                 | Prerequisites   |
| 1     | Mathematics                                 | Linear Algebra: Matrices, vectors, eigen values, eigenvectors, etc.<br>Calculus: Differentiation, integration, multivariate calculus.<br>Probability and Statistics: Probability theory, random variables, distributions,<br>statistical inference, etc                               |
| 2     | Programming<br>Skills                       | Proficiency in at least one programming language such as Python, R, or Julia.<br>Understanding of data structures and algorithms  |
| 3     | Data Analysis                               | Familiarity with data manipulation and analysis using libraries like NumPy,<br>Pandas, and Matplotlib in Python.<br>Experience in handling datasets, data Pre-processing, and data visualization.   |
| 4     | Basic Machine<br>Learning<br>Concepts       | Understanding of fundamental ML concepts like supervised learning,<br>unsupervised learning, reinforcement learning, etc.<br>Knowledge of common ML algorithms such as linear regression, logistic<br>regression, decision trees, k-nearest neighbours, support vector machines, etc. |
| 5     | Data Science<br>Tools and<br>Libraries      | Experience with tools and libraries commonly used in data science and ML such<br>as scikit-learn, TensorFlow, PyTorch, etc.<br>Knowledge of Jupyter Notebooks or similar environments for interactive data<br>analysis and experimentation  |
| 6     | Computer<br>Science<br>Fundamentals         | Basics of computer science including algorithms, data structures, and complexity analysis.<br>Understanding of computer architecture, operating systems, and networking.  |
| 7     | Probability and<br>Statistics               | Solid understanding of probability theory including Bayes' theorem, conditional probability, etc.<br>Familiarity with statistical concepts such as hypothesis testing, confidence intervals, etc.   |
| 8     | Linear Algebra                              | Knowledge of linear algebra concepts such as vectors, matrices, matrix operations, eigen values, eigenvectors, etc  |
| 9     | Critical Thinking<br>and Problem<br>Solving | Ability to think critically, analyse problems, and formulate solutions.<br>Strong problem-solving skills are crucial for understanding and applying ML algorithms effectively   |
| 10    | Domain<br>Knowledge                         | Depending on the specific application areas of interest (e.g., healthcare, finance, computer vision), having domain-specific knowledge can be advantageous.   |

# 2. Competencies

|            |   | ctenetes                          |   |
|------------|---|-----------------------------------|---|
| <b>S</b> / | L | Competency                        | KSA Description   |
| 1          | l | Understanding of Core<br>Concepts | <ul> <li>Knowledge: Gain a deep understanding of fundamental concepts in AI and ML, including supervised learning, unsupervised learning, reinforcement learning, neural networks, deep learning, etc.</li> <li>Skills: Skills in hyperparameter optimization techniques like grid search, random search, and Bayesian optimization to improve model performance.</li> <li>Attitude: Skills in diagnosing and fixing issues with model performance, such as vanishing/exploding gradients, slow convergence, or unstable training.</li> </ul> |
| 2          | 2 | Mathematical Foundations          | <ul> <li>Knowledge: Strengthen your proficiency in mathematics, particularly in linear algebra, calculus, probability theory, and statistics, as these form the mathematical underpinnings of AI and ML algorithms.</li> <li>Skills: Ability to identify the best deep learning model and architecture for solving specific real-world problems</li> </ul>  |



|   |                                     | Attitude: Knowledge of transfer learning and fine-tuning, which are essential for leveraging pre-trained models for new tasks with  |
|---|-------------------------------------|---|
|   |                                     | limited data.   |
| 3 | Programming Skills                  | <ul> <li>Knowledge: Develop strong programming skills in languages commonly used in AI and ML, such as Python or R.</li> <li>Skills: Learn how to efficiently use libraries and frameworks like TensorFlow, PyTorch, scikit-learn, and Keras for implementing AI and ML models.</li> <li>Attitude: Understanding of traditional batch processing, where data is collected, processed, and stored in batches</li> </ul>  |
| 4 | Data Handling and<br>Pre-processing | <ul> <li>Knowledge: Master techniques for data handling, pre-processing, and cleaning to prepare datasets for training and analysis.</li> <li>Skills: Understand how to deal with missing data, outliers, and noise in datasets.</li> <li>Attitude: Understanding of traditional batch processing, where data is collected, processed, and stored in batches</li> </ul>   |
| 5 | Model Selection and Evaluation      | <ul> <li>Knowledge: Learn how to select appropriate ML models for different tasks and datasets.</li> <li>Skills: Develop skills in evaluating model performance using various metrics and techniques such as cross-validation.</li> <li>Attitude: Strong analytical thinking and the ability to troubleshoot and improve models based on the behaviour of the training process</li> </ul>   |
| 6 | Deep Learning                       | <ul> <li>Knowledge: Gain expertise in deep learning techniques,<br/>architectures, and frameworks for tasks such as image recognition,<br/>natural language processing (NLP), and sequential data analysis.</li> <li>Skills: Ability to design, implement, and fine-tune deep learning<br/>models for specific tasks, from initial architecture design to final<br/>model evaluation.</li> <li>Attitude: Adaptability to rapidly evolving technologies, such as new<br/>neural network architectures</li> </ul> |

## 3. Syllabus

| Artificial Intelligence and Machine Learning<br>SEMESTER-I |           |                    |     |
|--|-----------|--------------------|-----|
| <b>Course Code</b>   | M23MCS105 | <b>CIE Marks</b>   | 50  |
| Teaching Hours/Week(L:P:SDA)                               | 2:0:0:0   | SEE Marks          | 50  |
| Total Hours of Pedagogy                                    | 40        | <b>Total Marks</b> | 100 |
| Credits  | 03        | <b>Exam Hours</b>  | 03  |
|  |           |                    |     |

**Course objectives:** 

1. To understand the basics of AI.

2. To work with the problem-solving issues of AI.

- 3. To study advanced problem solving paradigms and knowledge representation.
- 4. To interpret neural networks, build neural networks to solve various classification problems.

Teaching- Learning Process: Chalk and talk method/Power Point Presentation

#### Module-1

Introduction, Problem Solving: State space search, General problem solving, characteristics of problem, Exhaustive searches, Heuristic searches, constraint satisfaction.

#### Module-2

Problem Reduction and Game playing: Introduction, problem reduction, game playing, bounded lock ahead strategy, Alpha Beta pruning, Logic concepts and logic programming: Logical Reasoning, propositional calculus and logic, natural deduction system, semantic tableau system, resolution refutation in propositional logic predicate logic, Logic programming.

Module-3



Advanced problem-solving paradigm: planning, Introduction, types of planning systems, block world program, logic-based planning, linear planning using goal stack, Means ends analysis, Non-Linear planning strategies, learning plans, Knowledge representation: Introduction, Approaches to knowledge representation, knowledge representation using semantic networks, extended semantic network for KR, knowledge representation using frames.

#### Module-4

Uncertainty Measure: Probability Theory, Bayesian Belief Networks, Machine Learning Paradigms: Machine learning system, supervised and unsupervised learning, Inductive, deductive learning, Clustering

#### Module-5

Support vector Machine, case-based reasoning and learning. ANN: Single Layer, Multilayer. RBF, Design issues in ANN, Recurrent Network

# Suggested Learning Resources:

#### Text Books:

- 1. Artificial Intelligence: Saroj Kaushik, Cengage Learning, 2014 Edition.
- 2. Artificial Intelligence: Structures and Strategies for Complex Problem Solving, George F Luger, Pearson Addison Wesley, 6th Edition, 2008.

#### **Reference books:**

- 1. Artificial Intelligence, E Rich, K Knight, and S B Nair, Tata Mc-Graw Hill, 3rd Edition, 2009.
- 2. Artificial Intelligence: A Modern Approach, Stuart Russell and Peter Norvig, Prentice Hall 3 rd Edition, 2009.

#### **Skill Development Activities Suggested**

• The students with the help of the course teacher can take up technical –activities which will enhance their skill other students should interact with industry (small, medium and large), understand their problems or for see what can be undertaken for study in the form of research/testing/projects, and for creative and innovative methods to solve the identified problem. The prepared report shall be evaluated for CIE marks.

#### 4. Syllabus Timeline

| ſ | S/L | Syllabus Timeline | Description                                  |
|---|-----|-------------------|--|
| ſ | 1   | Weeks1-3          | Module1-Introduction, Problem Solving.       |
| ſ | 2   | Weeks4-6:         | Module2- Problem Reduction and Game playing. |
| ſ | 3   | Weeks7-8          | Module3 -Advanced problem-solving paradigm.  |
|   | 4   | Weeks9-11:        | Module4- Uncertainty Measure.                |
|   | 5   | Weeks12-14:.      | Module5- Support vector Machine.             |

# 5. Teaching-Learning Process Strategies

| S/L | TLP Strategies:                             | Description  |
|-----|---|--|
| 1   | Interactive Lectures:                       | Lectures will be designed to be interactive, with frequent opportunities for discussion, questioning, and real-time feedback.                                      |
| 2   | Handson Coding<br>Exercises:                | Regular coding exercises and assignments will be provided to reinforce practical skills in data science programming language sand libraries.                       |
| 3   | Case Studies and<br>Real-World<br>Examples: | The course will incorporate real-world case studies and examples from various domains to illustrate the applications of data science concepts and techniques.      |
| 4   | Group Projects:                             | Students will work in teams to complete a comprehensive data-science<br>project, allowing them to apply the knowledge and skills to a real-world<br>problem.       |
| 5   | Guest Lectures:                             | Industry experts and practitioners will be invited to share their experiences<br>and insights, providing students with a practical perspective on data<br>science. |

|   | Flipped Classroom: | Selected topics will be introduced through pre-recorded lectures or reading   |
|---|--------------------|---|
| 6 |                    | materials, allowing class time to be dedicated to discussions, exercises, and |
|   |                    | problem-solving activities.   |

#### 6. Assessment Details (both CIE and SEE)

CIE:

#### Theory Course with 3 Credits: Professional Core Course (PC)

This section of regulations is applicable to all theory-based courses. The minimum CIE marks requirement is 40% of maximum marks in each component.

| Components |                               | Number | Weightage | Max.<br>Marks | Min.<br>Marks |
|------------|-------------------------------|--------|-----------|---------------|---------------|
| (i)        | Internal Assessment-Tests (A) | 2      | 50%       | 25            | 10            |
| (ii)       | Assignments/Quiz/Activity (B) | 2      | 50%       | 25            | 10            |
|            | Total Mark                    | S      | •         | 50            | 20            |

# <sup>1</sup> CIE Split up for Professional Course (PC)

#### Final CIE Marks =(A) + (B)

Average internal assessment shall be the average of the 2 test marks conducted. **SEE:** 

# Theory Course with 4, 3 and 2 Credits: Professional Core Course (PC)/Professional Elective/Open Elective

- Question paper pattern will be ten questions. Each question is set for 20marks. The medium of the question paper shall be English unless otherwise it is mentioned.
- There shall be 2 question from each module, each of the two questions under a module (with a maximum of 3 sub questions), may have mix of topics under that module if necessary.
- The students have to answer 5 full questions selecting one full question from each module.
- Marks scored will be proportionally scaled down to 50 marks

#### 7. Learning Objectives

| S/L  | Learning Objectives              | Description  |
|--|----------------------------------|--|
| 1  | Principles of Algorithm          | Develop a deep and comprehensive understanding of the theoretical foundations, principles, algorithms, and methodologies underlying AI and ML.   |
| 2  | AI & ML Technique                | Gain hands-on experience in implementing AI and ML techniques, algorithms, and models using relevant programming languages, libraries, and frameworks.   |
| 3  | Deep Learning                    | Explore advanced topics and cutting-edge research areas in AI and ML, including deep learning, reinforcement learning, generative adversarial networks (GANs), natural language processing (NLP), computer vision, etc.                        |
| 4  | Critical Thinking                | Develop strong problem-solving skills and critical thinking abilities to<br>analyze complex AI and ML problems, devise effective solutions, and<br>evaluate their performance.   |
| 5  | Interdisciplinary<br>perspective | Gain an interdisciplinary perspective by integrating concepts, techniques,<br>and methodologies from fields such as computer science, mathematics,<br>statistics, engineering, cognitive science, and domain-specific areas of<br>application. |
| Course Outcomes (COs) and Mapping with POs/ PSOs |                                  |  |

#### Course Outcomes (COs)

| M23MCS105.1 Develop a deep understanding of advanced AI and ML concepts, theories, algorithms, |
|--|
|--|



|             | and methodologies beyond the undergraduate level. Explore topics such as deep learning, reinforcement learning, probabilistic graphical models, natural language processing, computer vision, etc.  |
|-------------|---|
| M23MCS105.2 | Acquire specialized technical skills in AI and ML, including proficiency in programming languages (e.g., Python, R), libraries, and frameworks (e.g., TensorFlow, PyTorch) commonly used in AI and ML research and applications.  |
| M23MCS105.3 | Explore emerging trends, cutting-edge technologies, and novel approaches in AI and ML research and applications. Develop research skills to conduct original research in AI and ML, formulate research questions, design experiments, collect and analyze data, and contribute to the advancement of knowledge in the field.                                |
| M23MCS105.4 | Work on hands-on projects, case studies, and real-world applications of AI and ML to gain practical experience in applying theoretical knowledge to solve practical problems. Collaborate with industry partners, research labs, or start ups to work on industry-relevant projects and gain insights into AI and ML applications in professional settings. |
| M23MCS105.5 | Develop strong problem-solving skills and critical thinking abilities to analyze complex AI and ML problems, devise effective solutions, and evaluate their performance   |

|             | CO/PO Mapping |     |     |      |      |
|-------------|---------------|-----|-----|------|------|
| COs/POs     | PO1           | PO2 | PO3 | PSO1 | PSO2 |
| M23MCS105.1 | 3             |     |     | 3    | 3    |
| M23MCS105.2 | 2             | 3   |     | 3    | 3    |
| M23MCS105.3 |               | 2   | 3   | 3    | 3    |
| M23MCS105.4 |               | 2   | 3   | 3    | 3    |
| M23MCS105.5 | 3             | 2   | 2   | 3    | 3    |
| M23MC8105   | 3             | 2   | 3   | 3    | 3    |

#### 9. Assessment Plan

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

|          | CO1 | CO2 | CO3 | CO4 | C04 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 10  |     |     |     |     | 10    |
| Module 2 |     | 10  |     |     |     | 10    |
| Module 3 |     |     | 10  |     |     | 10    |
| Module 4 |     |     |     | 10  |     | 10    |
| Module 5 |     |     |     |     | 10  | 10    |
| Total    | 10  | 10  | 10  | 10  | 10  | 50    |

#### Semester End Examination (SEE)

|          | CO1 | CO2 | CO3 | CO4 | CO4 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 20  |     |     |     |     | 20    |
| Module 2 |     | 20  |     |     |     | 20    |
| Module 3 |     |     | 20  |     |     | 20    |
| Module 4 |     |     |     | 20  |     | 20    |
| Module 5 |     |     |     |     | 20  | 20    |
| Total    | 20  | 20  | 20  | 20  | 20  | 100   |

#### 10. Future with this Subject

Learning Artificial Intelligence (AI) and Machine Learning (ML) in an M.Tech program in Computer Science and Engineering opens up a plethora of opportunities in various domains. Here's a glimpse into the future with such a specialization:

- 1. **Cutting-edge Research:** Pursuing AI and ML in M.Techsets the stage for engaging in cuttingedge research. You could delve into areas like deep learning, reinforcement learning, natural language processing, computer vision, robotics, and more. This research could lead to breakthroughs in AI technology, contributing to advancements in various fields.
- 2. **Industry Demand**: AI and ML are revolutionizing industries across the globe. With an M.Tech specialization, you'll be in high demand for roles such as machine learning engineer, data scientist, AI researcher, AI consultant, and more. Industries ranging from healthcare and finance to manufacturing and automotive are actively seeking professionals with expertise in AI and ML.
- 3. **Innovation and Entrepreneurship:** Armed with a solid foundation in AI and ML, you'll have the skills to innovate and create disruptive technologies. Many M.Tech graduates in AI and ML venture into entrepreneurship, founding start ups that leverage AI to address real-world problems or introduce innovative products and services.
- 4. Academic Pursuits: For those inclined towards academia, an M.Tech in AI and ML provides a strong platform for further academic pursuits. You could pursue a Ph.D. and contribute to academic research in AI, becoming a faculty member at a university or research institution.
- 5. AI Ethics and Governance: As AI and ML technologies continue to evolve, there's a growing emphasis on ethical AI development and governance. M.Tech graduates with expertise in AI ethics and governance can play pivotal roles in ensuring that AI technologies are developed and deployed responsibly, addressing concerns related to bias, fairness, transparency, privacy, security, and societal impact.
- 6. **Global Opportunities:** AI and ML talent is in demand worldwide. Whether you aspire to work in tech hubs like Silicon Valley or pursue opportunities in emerging AI markets, your skills will be valued globally. International organizations, research institutions, and multinational corporations are actively recruiting AI and ML professionals.
- 7. **Continuous Learning and Adaptation:** The field of AI and ML is dynamic and rapidly evolving. As an M.Tech graduate, you'll need to embrace continuous learning to stay updated with the latest advancements, technologies, and methodologies in AI and ML. This commitment to lifelong learning will ensure your relevance and competitiveness in the job market.



| 1510                     | Mandatory Core Course (MC)          |           |
|--------------------------|-------------------------------------|-----------|
| 1 <sup>st</sup> Semester | <b>Research Methodology and IPR</b> | M23MCS106 |

#### 1. Prerequisites

| S/L | Proficiency  | Prerequisites  |
|-----|--|--|
| 1   | Research Design and Methodologies                      | Basic knowledge of scientific research principles and methods, typically gained through introductory research methods courses.                                     |
| 2   | Statistical Analysis<br>and Data<br>Interpretation     | Understanding of basic probability theory and statistical concepts, usually covered in introductory statistics course  |
| 3   | Academic Writing and Communication:                    | Basic Skill in writing clear, well-structured research papers and reports, and effectively communicating research findings.  |
| 4   | Knowledge of<br>Intellectual Property<br>Rights (IPR): | Knowledge of business law or legal studies, and basic understanding of legal terminology related to intellectual property.   |
| 5   | Critical Thinking and<br>Problem-Solving:              | knowledge in the specific field of study (e.g., engineering, social sciences, life sciences), and experience with critical analysis and problem-solving techniques |

#### 2. Competencies

| S/L | Competency  | KSA Description  |  |  |  |  |  |
|-----|---|--|--|--|--|--|--|
| 1   | Research Design and<br>Methodologies  | <ul> <li>Knowledge: Understanding various research designs, includin experimental, correlational, and descriptive methodologies.</li> <li>Skills: Ability to design robust research studies, formulates research questions, and select appropriate methodologies.</li> <li>Attitudes: Valuing rigor and integrity in research design and execution</li> </ul>                    |  |  |  |  |  |
| 2   | Statistical Analysis<br>and Data<br>Interpretation  | <ul> <li>Knowledge: Familiarity with statistical concepts such as hypothesis testing, regression analysis, and inferential statistics.</li> <li>Skills: Proficiency in using statistical software (e.g., SPSS, R) to analyze and interpret data.</li> <li>Attitudes: Appreciating the importance of accuracy and objectivity in data analysis</li> </ul>                         |  |  |  |  |  |
| 3   | Academic Writing<br>and Communication   | <ul> <li>Knowledge: Understanding the structure and format of academic papers, including proper citation and referencing techniques.</li> <li>Skills: Ability to write clear, concise, and well-structured research papers and reports</li> <li>Attitudes: Valuing clarity, precision, and ethical practices in academic writing.</li> </ul>                                     |  |  |  |  |  |
| 4   | Knowledge of<br>Intellectual Property<br>Rights (IPR)   | <ul> <li>Knowledge: Understanding the different types of intellectual propert (patents, trademarks, copyrights) and the application processes for each</li> <li>Skills: Ability to navigate the IP application process, including drafting an filing patents.</li> <li>Attitudes: Respecting the legal and ethical considerations in protectin intellectual property.</li> </ul> |  |  |  |  |  |
| 5   | Ethical<br>Considerations in<br>ResearchKnowledge: Understanding ethical guidelines and principles in conducti<br>research, including issues related to consent, confidentiality, and do<br>integrity.Skills: Ability to apply ethical principles in research design and execution<br>Attitudes: Commitment to upholding ethical standards and integrity in<br>research activities. |  |  |  |  |  |  |

# 3. Syllabus

## RESEARCH METHODOLOGY AND IPR SEMESTER-I

| Course Code                             | M23MCS106 | CIE Marks   | 50  |
|---|-----------|-------------|-----|
| Number of Lecture Hours/Week(L: P: SDA) | (3:0:0:0) | SEE Marks   | 50  |
| Total Hours of Pedagogy                 | 40 hours  | Total Marks | 100 |
| Credits                                 | 03        | Exam Hours  | 03  |

**Course Learning objectives:** 

At the end of the course the student will be able to:

- Overview of formulating research problem and significance of the literature review.
- Demonstrate the types of research designs with experimentation.
- Importance of thesis writing skills.
- Understand the basic concepts of intellectual property and its impact on society.
  - Module -1

**Introduction to Research Methodology**: Meaning of Research, Objectives of Research, Types of Research, Research Approaches, Significance of Research, Research Methods versus Methodology, Research and Scientific Method, Research Process, Criteria of Good Research, Qualities of good research, problems encountered by researchers in India.

Text Book 1: Ch. 1

Module -2

**Literature Review**: Necessity of defining problem, Techniques involved in defining the problem, Selecting the problem, The place of the literature review in research, bringing clarity and focus to your research problem, improving your research methodology, broadening your knowledge base in your research area, enabling you to contextualize your findings, how to review the literature, searching for the existing literature, Reviewing the selected literature, developing a theoretical framework, developing a conceptual framework, writing about the literature reviewed.

Text Book 1: Ch. 2, Text Book 2: Ch. 3

#### Module -3

**Research Design:** Meaning of research design, need for research design, features of a good design, Important concepts relating to research design, Different research design, Developing a research plan, basic principles of experimental design, Need for Sampling, Steps in sampling design, different types of sample design. Characteristics of good sample design.

Text Book 1: Ch. 3, 4

#### Module -4

**Data Collection and Report writing**: Collection of primary data, collection of data through questionnaires, observation method, collection of secondary data, some other methods of data collection, Selection of appropriate method for data collection, significance of report writing, layout of research report, different steps in writing report, precautions for writing research report.

Text Book 1: Ch. 6, 19

#### Module -5

**Overview of Intellectual property**: Invention and creativity, procedure for grant of patents, types of patents, basic concepts of patents, trademarks and copyrights, protection of IPR, Geographical indications, industrial designs, protection of undisclosed information, condition on patent applicants, Duration of protection, term of protection.

#### Text Book 3: Ch. 1

#### **Text Books:**

1. Research Methodology: Methods and Techniques, C.R. Kothari, Gaurav Garg, New Age International 3rd Edition, 2014.

2. Study Material (For the topic Intellectual Property under module 5) Professional Programme Intellectual Property Rights-Law and Practice, The Institute of Company Secretaries of India, Statutory Body Under an Act of Parliament, September 2013

## **Reference books:**

1. B L Wadehra, Law relating to patents, trademarks, copyright designs and geographical Indications, Universal Law Publishing, 2000.

2. Study Material (For the topic Intellectual Property under module 5) Professional Programme Intellectual Property Rights-Law and Practice, The Institute of Company Secretaries of India, Statutory Body Under an Act of Parliament, September 2013

e-Resources: https://www.mooc.org/https://onlinecourses.nptel.ac.in/

# 4. Syllabus Timeline

| Jinab |  |  |
|-------|--|--|
| S/L   | Syllabus<br>Timeline                                     | Description  |
| 1     | Week 1-2:<br>Research<br>Methodology                     | <ul> <li>Competency: Ability to design and conduct rigorous and ethical research studies</li> <li>Knowledge: Understanding the principles and components of various research methodologies, including qualitative, quantitative, and mixed methods.</li> <li>Skills: Proficiency in formulating research questions, selecting appropriate research designs, and implementing data collection techniques.</li> </ul>                                    |
| 2     | Week 3-4:<br>Defining the<br>Research<br>Problem         | <b>Competency:</b> Ability to clearly define and articulate a research problem.<br><b>Knowledge</b> : Understanding the significance of a well-defined research problem and the criteria for selecting a viable research topic<br><b>Skills:</b> Proficiency in identifying research gaps, formulating precise research questions, and developing problem statements.  |
| 3     | Week 5-6:<br>Reviewing the<br>Literature                 | <ul> <li>Competency: Ability to conduct a comprehensive and critical literature review.</li> <li>Knowledge: Understanding the purpose, structure, and methods of conducting a literature review, including identifying relevant sources and synthesizing existing research.</li> <li>Skills: Proficiency in searching academic databases, evaluating the credibility of sources, and summarizing and critiquing existing research findings.</li> </ul> |
| 4     | Week 7-8:<br>Research<br>Design and<br>Sample<br>Surveys | Competency: Ability to design robust research studies and develop<br>effective sampling strategies.<br>Knowledge: Understanding the different types of research designs (e.g.,<br>experimental, correlation, descriptive).<br>Skills: Proficiency in selecting appropriate research designs, developing<br>sampling plans  |
| 5     | Week 9-10:<br>Data<br>Collection and<br>Interpretation   | <b>Competency:</b> Ability to collect, interpret, and analyze data systematically.<br><b>Knowledge:</b> Understanding various data collection methods (e.g., surveys, interviews, observations) and basic data analysis techniques.<br><b>Skills:</b> Proficiency in designing data collection instruments, conducting data collection, and applying appropriate methods to interpret and analyze data.  |
| 6     | Week 11-12:<br>Intellectual<br>Property                  | Apply learned concepts and competencies to real-world scenarios.<br>Hands-on practice with programming assignments   |

# 5. Teaching learning process strategies

| S/L | TLP Strategies                                | Description   |
|-----|---|---|
| 1   | Lecture Method                                | Utilize various teaching methods within the lecture format to reinforce competencies. |
| 2   | Collaborative<br>Learning                     | Encourage collaborative learning for improved competency application.                 |
| 3   | Higher Order<br>Thinking (HOTS)<br>Questions: | Pose HOTS questions to stimulate critical thinking related to each competency.        |
| 4   | Problem-Based<br>Learning (PBL)               | Implement PBL to enhance analytical skills and practical application of competencies  |
| 5   | Multiple<br>Representations                   | Introduce topics in various representations to reinforce competencies                 |



| 6 | Real-World<br>Application  | Discuss practical applications to connect theoretical concepts with real-<br>world competencies.                      |
|---|----------------------------|---|
| 7 | Flipped Class<br>Technique | Utilize a flipped class approach, providing materials before class to facilitate deeper understanding of competencies |
| 8 | Programming<br>Assignments | Assign programming tasks to reinforce practical skills associated with competencies.                                  |

## 6. Assessment Details (both CIE and SEE)

This section of regulations is applicable to all theory-based courses. The minimum CIE marks requirement is 40% of maximum marks in each component.

|      | <b>CIE Split up for Professional Elective Course (PE)</b> |        |           |               |               |  |
|------|---|--------|-----------|---------------|---------------|--|
|      | Components  | Number | Weightage | Max.<br>Marks | Min.<br>Marks |  |
| (i)  | Internal Assessment-Tests (A)                             | 3      | 50%       | 25            | 10            |  |
| (ii) | Assignments/Quiz/Activity (B)                             | 2      | 50%       | 25            | 10            |  |
|      | Total Ma  | rks    |           | 50            | 20            |  |

# Final CIE Marks = (A) + (B)

Average internal assessment shall be the average of the best two test marks from the 3 tests conducted.

## Semester End Examinations: PG Program

## Theory Course with 4, 3 and 2 Credits: Professional Core Course (PC)/Professional Elective/Open Elective

- 1. Question paper pattern will be ten questions. Each question is set for 20marks. The medium of the question paper shall be English unless otherwise it is mentioned.
- 2. There shall be 2 question from each module, each of the two questions under a module (with a maximum of 3 sub questions), may have mix of topics under that module if necessary.
- 3. Thestudentshavetoanswer5fullquestionsselectingonefullquestionfromeachmodule.
- 4. Marks scored will be proportionally scaled down to 50marks.

## 7. Learning Objectives

| S/L | Learning Objectives                        | Description   |
|-----|--|---|
| 1   | Formulate Research<br>Questions:           | Students should be able to Develop clear and focused research questions based on identified research gaps.  |
| 2   | Design Research Studies:                   | Create robust research designs tailored to specific research questions using appropriate methodologies.   |
| 3   | Conduct Literature<br>Reviews:             | Perform comprehensive and critical reviews of existing literature to support research objectives.   |
| 4   | Apply Statistical<br>Analysis:             | Utilize statistical tools and techniques to analyze and interpret research data effectively.  |
| 5   | Collaboration and<br>Communication Skills  | Students will work collaboratively in teams on coding projects, enhancing their ability to communicate effectively, share ideas, and solve problems collectively in the context of error control coding.                            |
| 6   | Ethical and Professional<br>Responsibility | Students will understand the ethical and professional responsibilities associated with error control coding, including ensuring data integrity, maintaining confidentiality, and adhering to industry standards and best practices. |

## 8. Course Outcomes (COs) and Mapping with POs/ PSOs

| Course Outcomes (COs)            |  |  |  |  |
|----------------------------------|--|--|--|--|
| CO's DESCRIPTION OF THE OUTCOMES |  |  |  |  |
| M23MCS106.1                      | Apply research methodologies for formulating research problems |  |  |  |



|  | M23MCS106.2 Apply suitable methods for data collection, literature review, and interpretation. |  |  |  |
|--|--|--|--|--|
|  | M23MCS106.3 Identify the appropriate research design for a specific problem.                   |  |  |  |
| M23MCS106.4 Make use of research layout rules for thesis writing.                                  |  | Make use of research layout rules for thesis writing.                                  |  |  |
| M23MCS106.5 Analyze the different case studies on intellectual properties, copyrights and patents. |  | Analyze the different case studies on intellectual properties, copyrights and patents. |  |  |

| CO-PO-PSO Mapping |     |     |     |      |      |  |
|-------------------|-----|-----|-----|------|------|--|
| CO's              | PO1 | PO2 | PO3 | PSO1 | PSO2 |  |
| M23MCS106.1       | 3   | -   |     | 3    | 3    |  |
| M23MCS106.2       | 3   | -   |     | 3    | 3    |  |
| M23MCS106.3       | 3   | -   | 3   | 3    | 3    |  |
| M23MCS106.4       |     | -   | 3   | 3    | 3    |  |
| M23MCS106.5       |     | -   | 3   | 3    | 3    |  |
| M23MCS106         | 3   | -   | 3   | 3    | 3    |  |

#### 9. Assessment Plan

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

|          |     |     | nver nar Braraa |     |     |       |
|----------|-----|-----|-----------------|-----|-----|-------|
|          | CO1 | CO2 | CO3             | CO4 | CO5 | Total |
| Module 1 | 10  |     |                 |     |     | 10    |
| Module 2 |     | 10  |                 |     |     | 10    |
| Module 3 |     |     | 10              |     |     | 10    |
| Module 4 |     |     |                 | 10  |     | 10    |
| Module 5 |     |     |                 |     | 10  | 10    |
| Total    | 10  | 10  | 10              | 10  | 10  | 50    |

#### Semester End Examination (SEE)

|          | CO1 | CO2 | CO3 | CO4 | CO5 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 20  |     |     |     |     | 20    |
| Module 2 |     | 20  |     |     |     | 20    |
| Module 3 |     |     | 20  |     |     | 20    |
| Module 4 |     |     |     | 20  |     | 20    |
| Module 5 |     |     |     |     | 20  | 20    |
| Total    | 20  | 20  | 20  | 20  | 20  | 100   |

#### **Conditions for SEE Paper Setting:**

Each module of SEE question paper should be allocated with questions for 20% of the total SEE marks.

#### **10.** Future with this Subject

•Advancing Innovation and Commercialization: Future research could focus on enhancing methodologies for identifying patentable inventions and optimizing intellectual property strategies to facilitate innovation and commercialization.

•Ethical and Legal Considerations in Research: There is a growing need for research into ethical frameworks and legal guidelines concerning intellectual property rights, ensuring responsible and fair practices in research and innovation.

•Global Intellectual Property Landscape: Research could explore comparative studies of intellectual property systems across different countries and regions, addressing challenges and opportunities for international collaboration and protection.

•Emerging Technologies and IP Challenges: With rapid advancements in fields like AI, biotechnology, and block chain, future studies could examine how these technologies impact intellectual property laws and practices, requiring innovative approaches to IP management.



| 1 <sup>st</sup> Semester | Professional Core Laboratory Course (PCL) | M23MCSL107   |
|--------------------------|---|--------------|
| 1 <sup></sup> Semester   | Advanced DBMS laboratory                  | WIZSWICSLIU/ |

| 1 <u>.</u> Pr | erequisites                         |   |
|---------------|-------------------------------------|---|
| Sl<br>No      | Proficiency                         | Description   |
| 1             | Basic Database<br>Concepts          | Understanding of fundamental database concepts, including tables, relations, primary and foreign keys, and normalization  |
| 2             | SQL Proficiency                     | Ability to write and execute SQL queries, including SELECT, INSERT, UPDATE, DELETE, and JOIN operations.<br>Knowledge of advanced SQL features like subqueries, indexes, views and triggers   |
| 3             | Database Design                     | Skills in designing ER diagrams and converting them into relational schema.<br>Familiarity with normalization and denormalization techniques.   |
| 4             | Programming skills                  | Experience with at least one programming language (e.g., Java, Python, C#) that can be used for database connectivity and operations.<br>Understanding of how to use database connectors/APIs (like JDBC for Java, psycopg2 for Python) |
| 5             | Transaction<br>Management:          | Knowledge of transactions, ACID properties, and isolation levels.<br>Understand the concurrency control and recovery  |
| 6             | Data Modeling and<br>Analysis       | Experience with data modelling tools and techniques.<br>Ability to perform data analysis using SQL and other tools.   |
| 7             | Database<br>Management<br>Systems   | Familiarity with popular DBMS software (e.g., MySQL, PostgreSQL, Oracle, SQL Server)<br>Understanding of database installation and configuration  |
| 8             | Advanced Topics                     | Basic understanding of advanced topics like distributed databases, NoSQL databases, and big data technologies   |
| 9             | Operating Systems<br>and Networking | Basic knowledge of operating systems and networking concepts as they relate to database environments.<br>Familiarity with command-line tools and shell scripting  |
| 10            | Version Control<br>Systems          | Basic understanding of version control systems like Git for managing database-<br>related code and scripts  |

## 2. Competencies

| Sl | Competency          | Description   |
|----|---------------------|---|
| No | I V                 | •   |
| 1  |                     | Knowledge: Mastery in writing complex SQL queries, including advanced           |
|    | Advanced SQL        | joins, sub queries, and set operations  |
|    | Skills              | Skills: Proficiency in creating and managing stored procedures, functions, and  |
|    | SKIIIS              | triggers  |
|    |                     | Attitude: Ability to optimize query performance and tuning indexes to improve   |
|    |                     | throughput and reduce latency in distributed SQL databases.                     |
| 2  |                     | Knowledge: Skills in performance tuning and query optimization.                 |
|    | Database            | Skills: Understanding of indexing strategies and execution plans.               |
|    | Optimization:       | Attitude: Knowledge of backup and disaster recovery strategies in SQL           |
|    |                     | databases to ensure data durability and availability.                           |
| 3  | Database design and | Expertise in designing scalable and efficient database schemas.                 |
|    | modelling:          | Proficiency in normalization, denormalization, and data modelling techniques.   |
| 4  |                     | Knowledge: Competence in implementing database security measures,               |
|    |                     | including user authentication, authorization, and encryption Skills in          |
|    | Database security   | managing roles and permissions  |
|    |                     | Skills: Acquire knowledge of feature engineering techniques to extract relevant |
|    |                     | information from raw data and improve model performance.                        |
| 5  | Backup and          | Knowledge: Proficiency in database backup and recovery strategies.              |
|    | recovery            |   |



|   |                   | <b>Skills</b> : Regularly check the integrity of backups. Corruption or incomplete backups can render recovery impossible.<br><b>Attitude:</b> Ability to implement disaster recovery plans and perform point-in-time recovery. |
|---|-------------------|---|
| 6 | NoSQL and         | <b>Knowledge</b> : Familiarity with NoSQL databases (e.g., MongoDB, Cassandra) and their data models.   |
|   | NewSQL Databases: | <b>Skills</b> : Experience in setting up, configuring, and maintaining NoSQL databases, including backup strategies, monitoring, and scaling. <b>Attitude:</b> Understanding of NewSQL database and their use cases.            |

# 3. Syllabus

|                              | ADVAN   | CED DBMS LABORATOR   | Y  |   |
|------------------------------|---|--|--|---|
| G                            | 0.1   | SEMESTER-I   |  | 40  |
|                              |   |  |  | 40  |
| Teaching Hours/Week (L:P: S) |   | 1:0:2:0  |  | 60  |
| Credi                        | ts  | 02   | Exam Hours   | 100   |
| SI.<br>NO                    | DADT A. The following our origina   | Experiments  |  |   |
| NU                           | PART A: The following experimen   | IS may be implemented on N<br>AS with support for object fe  |  | other suitable  |
| 1                            | Develop a database application to c   |  |  |   |
| 1                            | objects.  | remonstrate storing and retr   | leving of DLOD and V   |   |
|                              | <ul> <li>Write a binary large object (BLOB) to a database as either binary or character (CLOB) data, depending on the type of the field in your data source. To write a BLOB value to the database, issue the appropriate INSERT or UPDATE statement and pass the BLOB value as an input parameter. If your BLOB is stored as text, such as a SQL Server text field, pass the BLOB as a string parameter. If the BLOB is stored in binary format, such as a SQL Server image field, pass an array of type byte as a binary parameter.</li> <li>Once storing of BLOB and CLOB objects is done, retrieve them and display the results accordingly.</li> </ul>   |  |  |   |
|                              | <ul> <li>Develop a database application to demonstrate the representation of multi valued attributes, and the use of nested tables to represent complex objects. Write suitable queries to demonstrate their use.</li> <li>Consider Purchase Order Example: This example is based on a typical business activity: managing customer orders. Need to demonstrate how the application might evolve from relational to object-relational, and how you could write it from scratch using a pure object-oriented approach.</li> <li>a. Show how to implement the schema Implementing the Application under the Relational Model using only Oracle's built-in data types. Build an object-oriented application on top of this relational schema using object views</li> </ul> |  |  |   |
| 3                            | Design and develop a suitable Stud<br>Couple of attributes to be maintain<br>has enrolled and Internal Assessme<br>• Whenever the attendance is<br>Head of the Department con   | ent Database application by<br>ed is the Attendance of a stud<br>ent Using TRIGGERS, write<br>updated, check if the attendar<br>cerned.                                  | lent in each subject for<br>active rules to do the<br>nee is less than 85%; i                      | r which he/she<br>following:<br>if so, notify the     |
|                              | <ul> <li>Whenever, the marks in an I 40%; if so, notify the Head of the Use the following guidelines when of Use triggers to guarantee that whether the Use database triggers only for c statement, regardless of which use.</li> <li>Do not define triggers that duplid define triggers to enforce data in constraints.</li> </ul>   | lesigning triggers:<br>ten a specific operation is performed<br>entralized, global operations t<br>ser or database application issu-<br>cate the functionality already b | ormed, related actions<br>hat should be fired fo<br>les the statement.<br>built into Oracle. For e | are performed.<br>r the triggering<br>example, do not |



|       | n<br>a<br>• E<br>tu                  | much more than 60 lines of PL/SQL code, it is better to include most of the code in a stored procedure, and call the procedure from the trigger. |                                  |                   |  |  |
|-------|--------------------------------------|--|----------------------------------|-------------------|--|--|
|       |                                      | iation rules. Run the program against any large c  |                                  |                   |  |  |
|       | discu                                | ss the results.  | _                                |                   |  |  |
|       |                                      | Association rules are if/then statements that help unco  |                                  |                   |  |  |
|       |                                      | ata in a relational database or other information repose<br>"If a customer buys a dozen eggs, he is 80% likely                                   |                                  | iation rule would |  |  |
| PAR   |                                      | evelop a mini project  | *                                |                   |  |  |
|       | Conduction of Practical Examination: |  |                                  |                   |  |  |
|       | All la                               | boratory experiments (nos) are to be included for pra  | ctical examination.              |                   |  |  |
|       | Evalu                                | ation: 50% of the marks allotted for lab experiment e  |                                  | marks for the     |  |  |
| 1     | project demo.                        |  |                                  |                   |  |  |
| 1     |                                      | ents are allowed to pick one experiment from list of the   |                                  |                   |  |  |
|       |                                      | ly follow the instructions as printed on the cover page  |                                  |                   |  |  |
|       |                                      | ge of experiment is allowed only once and marks a  | allotted to the procedure par    | t to be made      |  |  |
|       | zero.                                |  |                                  |                   |  |  |
|       |                                      | Details (both CIE and SEE)   |                                  |                   |  |  |
|       |                                      | actical Courses (Program Based):   |                                  |                   |  |  |
|       |                                      | for laboratory course with software experiments shal   | I be conducted for a total of 10 | 00 mark at the    |  |  |
| end t | he seme                              | ester and the assessment pattern is  |                                  |                   |  |  |
| ~     |                                      | Marks distribution for Program based P   |                                  |                   |  |  |
| SI.   | . No.                                | Description  | % of Marks                       | In Marks          |  |  |
|       | 1                                    | Observation, write-up,   | 80% of the maximum               | 80                |  |  |
|       |                                      | algorithm/program/execution  |                                  |                   |  |  |
|       | 2                                    | Viva-Voce  | 20% of the maximum               | 20                |  |  |
|       |                                      | Total 100% 100   |                                  |                   |  |  |

Marks scored by the student for 100 are scaled down to 50 marks.

# Semester End Evaluation (SEE):

# SEE for practical Course (Irrespective of Experiment or program based):

SEE marks for practical course shall be 50 marks

## Marks distribution for Experiment based Practical Course for Final CIE

| SL. No. | Description           | % of Marks | Marks |
|---------|-----------------------|------------|-------|
| 1       | Write-up, Procedure   | 20%        | 20    |
| 2       | Conduction and result | 60%        | 60    |
| 3       | 3 Viva-Voce           |            | 20    |
|         | Total                 | 100%       | 100   |

• See for practical course is evaluated for 100 marks and scored marks shall be scaled down to 50 marks.

• Change of experiment/program is allowed only once and 20% marks allotted to the procedure/write-up part to be made zero.

• Duration of SEE shall be 3 hours.

## 4. Syllabus Timeline

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| S/L | Syllabus Timeline | Description  |
|-----|-------------------|--|
| 1   | Weeks1-2          | Introduction, bsaic programs example and execution procedure |
| 2   | Weeks2-4:         | Program1   |
| 3   | Weeks4-6          | Program2   |

# Ref: MITM/CS/CBS/2023-24/91 2023 Scheme –1st to 2<sup>nd</sup> sem Competency Based Syllabi for M. Tech CSE

| 4 | l | Weeks6-8: | Program3      |
|---|---|-----------|---------------|
| 5 | 5 | Week8-10. | Mini project. |

## 5. Teaching-Learning Process Strategies

| S/L | TLP Strategies:                                     | Description   |
|-----|---|---|
| 1   | Hands-on Lab<br>Sessions                            | <ul> <li>Focus on practical, hands-on lab sessions where students can apply theoretical concepts to real-world scenarios.</li> <li>Use a variety of database management systems (e.g., MySQL, PostgreSQL, Oracle) to expose students to different environments. Libraries.</li> </ul>     |
| 2   | Project-Based<br>Learning:                          | <ul> <li>Assign projects that require students to design, implement, and optimize a complete database system.</li> <li>Encourage collaborative projects to foster teamwork and peer learning.</li> </ul>  |
| 3   | Case Studies and<br>Real-World<br>Examples:         | <ul> <li>Integrate case studies that showcase real-world applications of advancedDBMS concepts.</li> <li>Analyse and discuss these case studies in class to provide practical insights.</li> </ul>  |
| 4   | Interactive<br>Lectures and<br>Demonstrations       | <ul><li>Use interactive lectures to explain complex topics, supplemented by live demonstrations of database tools and techniques.</li><li>Encourage student participation and questions during these sessions.</li></ul>  |
| 5   | Problem-Solving<br>Sessions                         | <ul><li>Conduct regular problem-solving sessions where students work on complex queries, optimization tasks, and troubleshooting exercises.</li><li>Provide immediate feedback and guidance</li></ul>   |
| 6   | Guest Lectures and<br>Industry Experts:             | <ul> <li>Invite industry experts and guest lecturers to share their experiences and insights on advanced DBMS topics.</li> <li>Organize Q&amp;A sessions with these experts to allow students to gain practical knowledge and career advice.</li> </ul>                                   |
| 7   | Research and<br>Innovation:                         | <ul><li>Encourage students to explore current research topics in advanced DBMS and present their findings.</li><li>Support innovative ideas and projects that push the boundaries of traditional database management</li></ul>  |
| 8   | Assessment through<br>Presentations and<br>Reports: | <ul> <li>Include assessments that require students to present their projects and findings, reinforcing their communication and presentation skills.</li> <li>Require detailed project reports that document their design choices, implementation process, and lessons learned.</li> </ul> |

# 6. Assessment Details (both CIE and SEE)

# Assessment Details (both CIE and SEE)

CIE for Practical Courses (Program Based):

Internal test for laboratory course with software experiments shall be conducted for a total of 100 mark at the end the semester and the assessment pattern is

| Marks distribution for Program | n based Practical Course for CIE |
|--------------------------------|----------------------------------|
|--------------------------------|----------------------------------|

| Sl. No. | Description   | % of Marks         | In Marks |
|---------|---|--------------------|----------|
| 1       | Observation, write-up,<br>algorithm/program/execution | 80% of the maximum | 80       |
| 2       | Viva-Voce   | 20% of the maximum | 20       |
|         | Total   | 100%               | 100      |

Semester End Evaluation (SEE):



# SEE for practical Course (Irrespective of Experiment or program based):

Total

• SEE marks for practical course shall be 50 marks

Viva-Voce

|         | Marks distribution for Experiment based Practical Course for Final CIE |            |       |  |  |
|---------|--|------------|-------|--|--|
| SL. No. | Description  | % of Marks | Marks |  |  |
| 1       | Write-up, Procedure  | 20%        | 20    |  |  |
| 2       | Conduction and result  | 60%        | 60    |  |  |

- See for practical course is evaluated for 100 marks and scored marks shall be scaled down to 50 marks.
- Change of experiment/program is allowed only once and 20% marks allotted to the procedure/write-up part to be made zero.

20%

100%

20

100

• Duration of SEE shall be 3 hours.

## 7. Learning Objectives

3

| S/L | Learning<br>Objectives        | Description  |
|-----|-------------------------------|--|
| 1   | Database<br>management system | To provide a thorough understanding of advanced database management system concepts, including complex queries, optimization techniques, and transaction management.                 |
| 2   | SQL Skills                    | To develop advanced SQL skills for creating, managing, and optimizing databases using sophisticated queries, stored procedures, triggers, and functions                              |
| 3   | Modelling                     | To equip students with the ability to design efficient and scalable database schemas through advanced data modelling and normalization techniques.                                   |
| 4   | Data Warehousing              | To provide knowledge on data warehousing concepts, ETL processes, and the use of business intelligence tools for data analysis and reporting   |
| 5   | Database Security             | To emphasize the importance of database security, teaching students how to<br>implement user authentication, authorization, encryption, and secure database<br>management practices. |

## 8. Course Outcomes and Mapping with POs/ PSOs

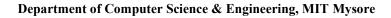
#### **Course Outcomes**

| M23MCSL107.1 | Demonstrate a deep understanding of complex database management system with various queries and attributes in practices. |
|--------------|--|
| M23MCSL107.2 | Apply on advanced SQL queries, including subqueries, joins, stored procedures Blob, Lob and triggers.                    |
| M23MCSL107.3 | Analyse NoSQL database models and big data technologies in appropriate scenarios   |

| COs/Pos      | PO1 | PO2 | PO3 | PSO1 | PSO2 |
|--------------|-----|-----|-----|------|------|
| M23MCSL107.1 | 2   | 3   |     | 3    | 3    |
| M23MCSL107.2 | 3   | 3   |     | 3    | 3    |
| M23MCSL107.3 |     | 3   | 3   | 3    | 3    |
| M23MCSL107   | 3   | 3   | 3   | 3    | 3    |

### 9. Assessment Plan

| Continuous Internal Evaluation (CIE) |  |  |  |  |  |  |  |
|--------------------------------------|--|--|--|--|--|--|--|
| CO1 CO2 CO3 CO3 Total                |  |  |  |  |  |  |  |





| Program1     | 10 |    |    |    | 10 |
|--------------|----|----|----|----|----|
| Program2     |    | 10 |    |    | 10 |
| Program3     |    |    | 10 |    | 10 |
| Mini project |    |    |    | 20 | 20 |
| Total        | 10 | 10 | 10 | 20 | 50 |

|              | CO1 | CO2 | CO3 | CO3 | Total |
|--------------|-----|-----|-----|-----|-------|
| Program1     | 20  |     |     |     | 20    |
| Program2     |     | 20  |     |     | 20    |
| Program3     |     |     | 20  |     | 20    |
| Mini project |     |     |     | 40  | 40    |
| Total        | 20  | 20  | 20  | 40  | 100   |

#### Semester End Examination (SEE)

#### 10. Future with this Subject

• **Career Advancement:** Database Administrator: Manage and maintain database systems, ensuring their performance, security, and availability Database Developer: Design and develop database applications, optimizing data storage and retrieval Data Analyst/Scientist: Analyze complex datasets to provide insights and support data-driven decision-making.

• **Expertise in Emerging Technologies:** Gain proficiency in NoSQL databases, big data technologies, and data warehousing solutions, which are increasingly in demand across various industries.

• **Research and Development** Engage in cutting-edge research in database technologies, contributing to advancements in areas like distributed databases, data mining, and machine learning.

• Entrepreneurial Opportunities: Leverage advanced DBMS skills to develop innovative data-driven solutions and start new ventures in the tech industry

• Enhanced Problem-Solving Skills: Develop strong analytical and problem-solving skills that are transferable to various domains, including software development, system architecture, and IT consultancy.

• **Improved Public Services:** Contribute to the development of efficient public services by designing robust database systems for government and non-profit organizations.



|       | 2 <sup>nd</sup> Semester  | Professional Core Course (PC)<br>Big Data Analytics  | M23MCS201  |  |  |  |
|-------|---|--|--|--|--|--|
| . Pre | erequisites   |  |  |  |  |  |
| S/L   |   |  |  |  |  |  |
| 1     | Basic<br>Computer<br>Science<br>Knowledge   | Computer Programming: Proficiency in a programming la<br>or Scala. Understanding of object-oriented programming (O<br>beneficial.<br>Data Structures and Algorithms: Knowledge of fundamer<br>lists, stacks, queues, hash tables, trees, and graphs) and algorithms.   | OP) principles is particularly ntal data structures (arrays, |  |  |  |
| 2     | Database<br>Management<br>Systems<br>(DBMS)   | <b>SQL:</b> Basic understanding of SQL for querying and managi <b>Relational Databases:</b> Familiarity with relational database normalization, indexing, transactions, and schema design.   |  |  |  |  |
| 3     | Basic Statistics<br>and<br>Mathematics  | <ul> <li>Descriptive Statistics: Mean, median, mode, standard devia distributions.</li> <li>Probability: Basic probability concepts and distributions.</li> <li>Linear Algebra: Vectors, matrices, and matrix operations.</li> </ul>   | ation, variance, and data                                    |  |  |  |
| 4     | Basic<br>Understanding Distributed Computing Principles: Basic concepts of distributed systems, such as |  |  |  |  |  |
| 5     | Introduction to<br>Cloud<br>Computing   | <b>Cloud Basics:</b> Understanding of what cloud computing is, it service models (IaaS, PaaS, SaaS).<br><b>Cloud Providers:</b> Familiarity with major cloud service provideogle Cloud.  | viders like AWS, Azure, and                                  |  |  |  |
| 6     | Introduction to<br>Big Data<br>Concepts   | <b>Big Data Fundamentals:</b> Awareness of what big data is, its significance, and its applications.<br><b>Data Formats:</b> Understanding of different data formats (structured, semi-structured, and unstructured).  |  |  |  |  |
| . Co  | mpetencies  |  |  |  |  |  |
| S/L   | Competency  | KSA Description  |  |  |  |  |
| 1     | Getting an<br>Overview of<br>Big Data   | <ul> <li>Knowledge:</li> <li>Understanding of the definition and components of big dat<br/>Veracity).</li> <li>Familiarity with the history and evolution of data manageme<br/>Skills:</li> <li>Ability to identify and categorize different types of data.</li> <li>Proficiency in analyzing big data characteristics and their im<br/>Attitudes:</li> <li>Appreciation of the transformative potential of big data in variable.</li> </ul> | ent leading to big data.<br>plications.                      |  |  |  |
| 2     | Introducing<br>Hadoop<br>Framework  | Knowledge:<br>Understanding of Hadoop architecture and its ecosystem com<br>Skills:<br>Proficiency in setting up and configuring Hadoop environme<br>Ability to navigate and use the Hadoop Distributed File Syst<br>Attitudes:<br>Openness to learning and adopting new technologies.   | ents.  |  |  |  |
| 3     | Hadoop<br>MapReduce<br>Framework  | Knowledge:<br>Understanding of the MapReduce model, its parallel da<br>mechanisms.<br>Skills:<br>Ability to write, compile, and execute MapReduce programs<br>Proficiency in optimizing MapReduce jobs for performance   |  |  |  |  |



|   |   | Attitudes:   |
|---|---|--|
|   |   | Problem-solving mindset to tackle data processing challenges.  |
| 4 | Hadoop YARN<br>Architecture<br>and Tools to<br>Analyze Data | <ul> <li>Knowledge:</li> <li>Understanding of YARN architecture, schedulers, and backward compatibility.</li> <li>Skills:</li> <li>Ability to configure and manage YARN for resource scheduling and job management.</li> <li>Proficiency in using analytical tools to process and analyze big data.</li> <li>Attitudes:</li> <li>Openness to adopting new analytical tools and techniques.</li> </ul>                              |
| 5 | Essential<br>Hadoop Tools                                   | <ul> <li>Knowledge:</li> <li>Understanding of the data models and functionalities of Apache Pig, Hive, Sqoop, Flume, HBase, and Oozie.</li> <li>Skills:</li> <li>Proficiency in using Apache Pig for data processing.</li> <li>Ability to use Apache Hive for data warehousing and querying.</li> <li>Attitudes:</li> <li>Appreciation of the importance of data processing and management tools in big data workflows.</li> </ul> |

#### 3. Syllabus

| Big Data Analytics |   |   |  |  |  |  |  |
|--------------------|---|---|--|--|--|--|--|
| SEMESTER – II      |   |   |  |  |  |  |  |
| M23MCS201          | CIE Marks   | 50  |  |  |  |  |  |
| (2:0:0:2)          | SEE Marks   | 50  |  |  |  |  |  |
| 40 hours Theory    | Total Marks   | 100   |  |  |  |  |  |
| 03                 | Exam Hours  | 03  |  |  |  |  |  |
|                    | SEMESTER - II           M23MCS201           (2:0:0:2)           40 hours Theory | SEMESTER – IIM23MCS201CIE Marks(2:0:0:2)SEE Marks40 hours TheoryTotal Marks |  |  |  |  |  |

Course objectives: This course will enable students to:

• Understand the Fundamentals and Evolution of Big Data

• Analyze the Applications and Advantages of Big Data Analytics

• Master the Hadoop Framework and Ecosystem

• Develop Skills in Hadoop MapReduce and YARN

• Utilize Essential Hadoop Tools for Data Processing and Management

Module -1

Getting an Overview of Big Data: What is Big Data? History of Data Management – Evolution of Big Data, Structuring Big Data, Types of Data, Elements of Big Data-Volume, Velocity, Variety, Veracity, Big Data Analytics, Advantages of Big Data Analytics.

Use of Big Data in Social Networking, Use of Big Data in Preventing Fraudulent Activities, Use of Big Data in Detecting Fraudulent Activities in Insurance Sector, Use of Big Data in Retail Industry.

Text Book 1: Ch. 1, 2

Module -2

**Introducing Hadoop Framework**: Introducing Hadoop, How does Hadoop Function?, Cloud Computing and Big Data, Features of Cloud Computing, Cloud Deployment Models, Cloud Delivery Models, Cloud Services for Big Data, Cloud Providers in Big Data Market, In-Memory Computing Technology for Big Data. Hadoop Ecosystem, Hadoop Distributed File System, HDFS Architecture, Features of HDFS, HDFS Components, HDFS user commands. **Text Book 1: Ch. 3, 4 Text Book 2: Ch. 3** 

Module -3

**Hadoop MapReduce Frame work :** The MapReduce Model, Map Reduce parallel data flow, Fault tolerance and speculative execution, Exploring the Features of Map Reduce, Working of MapReduce, Exploring Map and Reduce Functions, Techniques to Optimize MapReduce Jobs, Uses of Map Reduce Map Reduce Programming, compiling and running the Hadoop, word count example, using the streaming interface, using the pipes interface, commands- listing, killing, job status and Hadoop log management.

Text Book 1: Ch. 5

Text Book 2 : Ch. 5, 6

#### Module -4

Hadoop YARN Architecture and Tools to Analyze Data: Background of YARN, YARN Architecture, Working of YARN, YARN Schedulers, Backward Compatibility with YARN, YARN Configurations, YARN Commands. Analytical approaches, History of analytical tools, Introducing Popular analytical tools, Comparing various analytical tools. Text Book 1: Ch. 11, 19

### Module -5

**Essential Hadoop Tools :**Using Apache Pig, Using Apache Hive, Using Apache sqoop, Apache sqoop import and export methods, apachesqoop version changes, Using Apache Flume, Using Apache Hbase, Using Apache OozieHBase Data Model Overview.

# Text Book 2: Ch. 7

## **Text Books:**

- 1. Big Data: Black Book, DT Editorial Services, Wiley India Pvt. Ltd., 2018 Edition.
- 2. Hadoop 2 Quick Start Guide : Douglas Eadline , Pearson Education, 2016 Edition.

## **Reference Books:**

- 1. Hadoop Operations: Eric Sammer, O'Reilly Media, Inc., 2012 Edition.
- 2. Tom White, Hadoop: The Definitive Guide, 3rd Edition, O'reilly, 2012.
- 3. Arvind Sathi, "Big Data Analytics: Disruptive Technologies for Changing the Game", 1st Edition,
- 4. IBM Corporation, 2012.

### 4. Syllabus Timeline

| 0/1 | Syllabus Timeline                                     |   |
|-----|---|---|
| S/L |   | Description   |
|     |   | Week 1:Introduction to Big Data, What is Big Data?History of Data               |
|     |   | Management – Evolution of Big Data, Structuring Big Data                        |
| 1   | Week 1-2: Module I - Getting an                       | Week 2: Elements of Big Data, Volume, Velocity, Variety, Veracity               |
| 1   | Overview of Big Data                                  | Big Data Analytics, Advantages of Big Data Analytics, Use of Big Data in        |
|     |   | Various Sectors, Social Networking, Preventing Fraudulent Activities, Detecting |
|     |   | Fraudulent Activities in Insurance Sector, Retail Industry                      |
|     |   | Week 3: Introduction to Hadoop, Cloud Computing and Big Data                    |
|     | Week 3-5: Module II -<br>Introducing Hadoop Framework | Week 4: Cloud Deployment Models, Cloud Delivery Models, Cloud Services for      |
| 2   |   | Big Data, Cloud Providers in Big Data Market                                    |
|     |   | Week 5:In-Memory Computing Technology for Big Data, Hadoop Ecosystem            |
|     |   | Hadoop ,Distributed File System (HDFS)  |
|     |   | Week 6: The MapReduce Model, MapReduce Parallel Data Flow, Fault                |
|     |   | Tolerance and Speculative Execution   |
| 3   | Week 6-8: Module III - Hadoop                         | Week 7: Exploring the Features of MapReduce, Working of MapReduce               |
|     | MapReduce Framework                                   | Exploring Map and Reduce Functions  |
|     |   | Week 8: Techniques to Optimize MapReduce Jobs, Uses of MapReduce,               |
|     |   | MapReduce Programming   |
|     | Week 9-10: Module IV - Hadoop                         | Week 9:Background of YARN, YARN Architecture, Working of YARN                   |
| 4   | YARN Architecture and Tools to                        | Week 10: YARN Schedulers, Backward Compatibility with YARN, YARN                |
|     | Analyze Data  | Configuration, YARN Commands  |
|     | Week 11-12: Module V -                                | Week 11: Using Apache Pig, Using Apache Hive, Using Apache Sqoop                |
| 5   | Essential Hadoop Tools                                | Week 12: Using Apache Flume, Using Apache HBase, Using Apache Oozie             |
|     |   | HBase Data Model Overview   |

#### 5. Teaching-Learning Process Strategies

| S/L | TLP Strategies:               | Description   |  |  |  |
|-----|-------------------------------|---|--|--|--|
| 1   | Lectures and<br>Presentations | Use multimedia presentations to explain key concepts like the definition of big data, its characteristics (Volume, Velocity, Variety, Veracity), and its evolution. |  |  |  |
| 2   | Class Discussions             | Facilitate discussions on the advantages of big data analytics and its impact on different industries.  |  |  |  |
| 3   | Case Study Analysis           | Provide case studies on the use of big data in different industries and have students analyze and discuss them in groups.   |  |  |  |



| 4  | Quizzes and<br>Assessments: | Conduct quizzes to test students' understanding of the fundamental concepts of big data.                           |
|----|-----------------------------|--|
| 5  | Interactive Lectures        | Use diagrams and flowcharts to illustrate the Hadoop Distributed File System (HDFS) architecture and its features. |
| 6  | Hands-On Labs               | Conduct lab sessions where students install and configure Hadoop on virtual machines.                              |
| 7  | Step-by-Step Tutorials      | Provide detailed tutorials on the MapReduce model, including writing, compiling, and running MapReduce programs.   |
| 8  | Lectures and Diagrams       | Use diagrams to explain the YARN architecture and its components.  |
| 9  | Tool-Specific Lectures:     | Provide detailed lectures on using Apache Pig, Hive, Sqoop, Flume, HBase, and Oozie.                               |
| 10 | Blended Learning            | Combine online resources, video tutorials, and in-class activities to provide a diverse learning experience.       |

## 6. Assessment Details (both CIE and SEE)

# CIE:

# Theory Course with 3 Credits: Professional Core Course (PC)

This section of regulations is applicable to all theory-based courses. The minimum CIE marks requirement is 40% of maximum marks in each component.

|      | Components                    | Number | Weightage | Max.<br>Marks | Min. Marks |
|------|-------------------------------|--------|-----------|---------------|------------|
| (i)  | Internal Assessment-Tests (A) | 2      | 50%       | 25            | 10         |
| (ii) | Assignments/Quiz/Activity (B) | 2      | 50%       | 25            | 10         |
|      | Total Ma                      | 50     | 20        |               |            |

#### Final CIE Marks =(A) + (B)

Average internal assessment shall be the average of the 2 test marks conducted. **SEE:** 

## Theory Course with 4, 3 and 2 Credits: Professional Core Course (PC)/Professional Elective/Open Elective

- Question paper pattern will be ten questions. Each question is set for 20marks. The medium of the question paper shall be English unless otherwise it is mentioned.
- There shall be 2 question from each module, each of the two questions under a module (with a maximum of 3 sub questions), may have mix of topics under that module if necessary.
- The students have to answer 5 full questions selecting one full question from each module.
- Marks scored will be proportionally scaled down to 50 marks

## 7. Learning Objectives

| S/L | Learning<br>Objectives                                | Description  |
|-----|---|--|
| 1   | Understand the<br>Fundamentals of<br>Big Data         | Define what constitutes big data and describe its key characteristics: Volume, Velocity, Variety, and Veracity.<br>Trace the history and evolution of data management, leading to the emergence of big data.<br>Differentiate between structured, semi-structured, and unstructured data.  |
| 2   | Analyze Big Data<br>Analytics and Its<br>Applications | Explain the core elements of big data analytics and its advantages in various industries.<br>Evaluate the use of big data in social networking to enhance user experience and engagement.<br>Assess the role of big data in preventing and detecting fraudulent activities, particularly in the<br>insurance and retail sectors. |
| 3   | Master the Hadoop<br>Framework                        | Describe the architecture and components of the Hadoop ecosystem, including HDFS and<br>MapReduce.<br>Explain how Hadoop functions, including the role of cloud computing in big data processing.<br>Identify key cloud deployment and delivery models, and explain the features of cloud computing<br>relevant to big data.     |

| 4 | Develop<br>Proficiency in<br>Hadoop Distributed<br>File System<br>(HDFS) | Understand the HDFS architecture, its features, and components.<br>Use HDFS user commands to manage and manipulate data within the Hadoop ecosystem.<br>Describe the in-memory computing technology for big data and its benefits.  |
|---|--|---|
| 5 | Gain Skills in<br>Hadoop<br>MapReduce<br>Framework                       | Explain the MapReduce model, including its parallel data flow, fault tolerance, and speculative execution.<br>Write, compile, and execute MapReduce programs, and optimize these jobs for better performance.<br>Use the streaming and pipes interfaces for different data processing tasks in MapReduce. |
| 6 | Explore Hadoop<br>YARN Architecture<br>and Analytical<br>Tools           | Describe the background, architecture, and working of YARN, including its schedulers and configurations.<br>Execute YARN commands and understand its backward compatibility.<br>Compare various popular analytical tools, their history, and their applications in big data analysis.                     |
| 7 | Utilize Essential<br>Hadoop Tools for<br>Data Processing                 | Use Apache Pig, Hive, Sqoop, Flume, HBase, and Oozie for specific data processing tasks.<br>Perform data import and export operations using Apache Sqoop, and understand its version changes.<br>Implement data models in HBase and manage workflows with Apache Oozie.                                   |

# 8. Course Outcomes (COs) and Mapping with POs/ PSOs

# Course Outcomes (COs)

| COs         | Description   |
|-------------|---|
| M23MCS201.1 | Understand and apply Big Data Concepts and Analytics  |
| M23MCS201.2 | Analyze and apply Big Data analytics in various sectors such as social networking, fraud detection, and retail. |
| M23MCS201.3 | Describe the Hadoop framework and its components, including cloud computing and HDFS.                           |
| M23MCS201.4 | Develop and execute MapReduce programs, optimizing them for better performance                                  |
| M23MC8201.5 | Utilize essential Hadoop tools like Apache Pig, Hive, Sqoop, Flume, HBase, and Oozie for data processing        |

| CO-PO-PSO Mapping |     |     |     |      |      |  |
|-------------------|-----|-----|-----|------|------|--|
| COs/POs           | PO1 | PO2 | PO3 | PSO1 | PSO2 |  |
| M23MCS201.1       | 3   |     | 3   | 2    | 2    |  |
| M23MCS201.2       | 3   |     | 3   | 2    | 2    |  |
| M23MCS201.3       |     |     | 3   | 2    | 3    |  |
| M23MCS201.4       | 3   | 2   | 3   | 3    | 3    |  |
| M23MCS201.5       | 3   | 2   | 3   | 3    | 3    |  |
| M23MCS201         | 3   | 2   | 3   | 3    | 3    |  |

## 9. Assessment Plan

| 1 100000000000000000000000000000000000 |     |     |     |     |     |       |
|--|-----|-----|-----|-----|-----|-------|
| Continuous Internal Evaluation (CIE)   |     |     |     |     |     |       |
|  | CO1 | CO2 | CO3 | CO4 | CO5 | Total |
| Module 1                               | 10  |     |     |     |     | 10    |
| Module 2                               |     | 10  |     |     |     | 10    |
| Module 3                               |     |     | 10  |     |     | 10    |
| Module 4                               |     |     |     | 10  |     | 10    |
| Module 5                               |     |     |     |     | 10  | 10    |
| Total                                  | 10  | 10  | 10  | 10  | 10  | 50    |

Semester End Examination (SEE)



|          | CO1 | CO2 | CO3 | CO4 | CO5 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 20  |     |     |     |     | 20    |
| Module 2 |     | 20  |     |     |     | 20    |
| Module 3 |     |     | 20  |     |     | 20    |
| Module 4 |     |     |     | 20  |     | 20    |
| Module 5 |     |     |     |     | 20  | 20    |
| Total    | 20  | 20  | 20  | 20  | 20  | 100   |

#### **Conditions for SEE Paper Setting:**

Each module of SEE question paper should be allocated with questions for 20% of the total SEE marks

#### 10. Future with this Subject

The "Big Data Analytics" course in the second semester of the M.Tech program provides a comprehensive foundation in Big Data and Hadoop technologies, opening various career paths and opportunities in multiple industries. With the growing importance of data-driven decision-making, professionals skilled in Big Data analytics and Hadoop are in high demand. The integration of emerging technologies will further expand the scope and impact of Big Data, making it a pivotal field for future innovations.

- Career Opportunities: Data scientists and analysts are in high demand as organizations seek to extract meaningful insights from their data. With skills in Big Data and Hadoop technologies, professionals can analyze large datasets, identify trends, and make data-driven decisions to benefit businesses.
- **Research and Development:** Professionals with expertise in Big Data can engage in research to develop advanced analytics techniques. This includes predictive analytics, real-time data processing, and innovative ways to handle unstructured data.
- Machine Learning and AI Integration: Big Data plays a crucial role in training machine learning models. Researchers can explore the integration of Big Data with AI to create smarter, more efficient algorithms. Leads to advancements in AI technologies, improving automation, decision-making, and personalized services.
- **Emerging Technologies:**IoT devices generate massive amounts of data. Big Data technologies are essential for collecting, processing, and analyzing this data to derive actionable insights. Facilitates smart cities, industrial automation, and improved resource management.
- **Industry Applications:** Big Data analytics can revolutionize healthcare by providing insights into patient care, treatment outcomes, and operational efficiency. It helps in predictive modelling for disease outbreaks and personalized medicine. Improves patient outcomes, reduces costs, and enhances the overall efficiency of healthcare systems.
- Social Media and Networking: Social media platforms use Big Data to analyze user behavior, content trends, and advertising effectiveness. Hadoop's ability to handle vast amounts of unstructured data makes it ideal for this sector. Enhances user engagement, optimizes content delivery, and improves targeted advertising.



| 2 <sup>nd</sup> Semester | Integrated Professional Core Course (IPC) | MOSMCSOO  |
|--------------------------|---|-----------|
| 2 <sup></sup> Semester   | INTERNET OF THINGS                        | M23MCS202 |

#### 1. Prerequisites

| S/L | Proficiency   | Prerequisites   |
|-----|---|---|
| 1   | Basic Knowledge<br>of Networking:                               | Understanding of basic networking concepts, including IP addressing, subnets, and protocols.  |
| 2   | Fundamentals of<br>Programming:                                 | Proficiency in at least one programming language (e.g., Python, C/C++), with a focus on understanding programming logic and syntax.     |
| 3   | Introduction to<br>Embedded<br>Systems:                         | Familiarity with the basics of embedded systems and microcontrollers, including how they function and are programmed.                   |
| 4   | Basic<br>Understanding of<br>Data Structures and<br>Algorithms: | Knowledge of fundamental data structures (arrays, linked lists, stacks, queues) and basic algorithms (sorting, searching).              |
| 5   | Basic Digital Logic<br>Design:                                  | Understanding of digital logic circuits, including binary arithmetic, logic gates, and basic combinational and sequential circuits.     |
| 6   | Introductory<br>Course in<br>Computer<br>Networks:              | Completion of an introductory course in computer networks to understand<br>network layers, protocols, and basic network configurations. |
| 7   | Basic Knowledge<br>of Databases:                                | Understanding of database concepts and basic SQL for managing and retrieving data.  |

# 2. Competencies

| · · |     |                             |   |  |  |  |
|-----|-----|-----------------------------|---|--|--|--|
| S   | 5/L | Competency                  | KSA Description   |  |  |  |
| 1   |     | Technical<br>Competencies:  | <b>Knowledge:</b> Understanding the concept and motivations behind IoT.<br>Familiarity with IPv6 and its role in IoT.<br><b>Skills:</b> Knowledge of various IoT definitions and frameworks. Understanding<br>smart metering, health/body area networks, city automation, automotive<br>applications, home automation, smart cards, tracking, and surveillance.<br><b>Attitude:</b> Comprehension of control application examples and other myriad<br>applications.   |  |  |  |
| 2   |     | Analytical<br>Competencies: | <ul> <li>Knowledge: Evaluating the effectiveness and efficiency of various IoT frame works and technologies.</li> <li>Assessing the role and impact of IPv6 in IoT.</li> <li>Skills: Analyzing real-world IoT applications in different domains such as home automation, city automation, health networks, etc.</li> <li>Attitude: Evaluating case studies illustrating IoT design and its implementation.</li> </ul>   |  |  |  |
| 3   |     | Practical<br>Competencies:  | <ul> <li>Knowledge: Designing and implementing IoT frameworks and applications.</li> <li>Applying knowledge of basic nodal capabilities in practical scenarios.</li> <li>Skills: Designing and developing systems using binary adders, subtractors, comparators, decoders, encoders, and multiplexers. Utilizing programmable logic devices in IoT solutions.</li> <li>Attitude: Implementing IoT connectivity solutions using WPAN and cellular/mobile network technologies. Deploying IPv6 technologies and implementing migration strategies.</li> </ul> |  |  |  |

## 3. Syllabus

| INTERNET OF THINGS<br>SEMESTER – II |  |  |  |  |
|-------------------------------------|--|--|--|--|
| Course CodeM23MCS202CIE Marks50     |  |  |  |  |

| Number of Lecture Hours/Week(L: T: P: S)   | (3:0:2:0)  | SEE Marks   | 50   |  |
|--|--|---|--|--|
| Total Number of Lecture Hours  | 40 hours Theory  | Total Marks   | 100  |  |
| Credits  | 04   | Exam Hours  | 03   |  |
| <ul> <li>Course objectives: This course will enable st</li> <li>To provide students with a comprehend definitions, frameworks, and basic not the enable of t</li></ul> | ensive understanding<br>odal capabilities.<br>ations of IoT in areas<br>, home automation, at<br>tal IoT mechanisms<br>emerging standards.<br>ologies used in IoT,<br>ilar, and IPv6 technol | such as smart me<br>nd surveillance.<br>and key techno<br>focusing on both<br>ogies.            | etering, health networks,<br>logies, including object<br>h Layer ½ and Layer 3                                 |  |
| examples in home automation, s<br>productivity.  | smart cities, enviro   |   |  |  |
|  | Module -1  | C A 11 11   |  |  |
| What is Internet of Things? Overview and M<br>Development and Standardization, Scope of<br>frameworks-IoT Definitions, IoT Framework<br>Examples-Overview, Smart Metering/Advance<br>Automation, Automotive Applications, Home<br>Surveillance/Ring of Steel, Control Application  | the Present Investiga<br>cs, Basic Nodal Capa<br>ced Metering Infrastru<br>e Automation, Smart   | tion. Internet of<br>abilities. Internet<br>acture Health/Bo<br>Cards, Tracking                 | Things Definitions and<br>t of Things Application<br>dy Area Networks, City<br>g, Over-The-Air-Passive         |  |
|  | Module -2  |   |  |  |
| Comparators, Decoders, Encoders, Multiplex<br>of Text 2) Fundamental IoT Mechanism and<br>Structural Aspects of the IoT, Key IoT Techn<br>IETF IPV6 Routing Protocol for RPL Roll<br>Transfer, ETSI M2M,Third Generation Par<br>Communications, CENELEC, IETF IPv6 Ove   | Key Technologies-Ic<br>nologies. Evolving Io<br>l, Constrained Appli<br>rtnership Project Se<br>er Low power WPAN<br>Module -3   | lentification of I<br>T Standards-Ov<br>cation Protocol,<br>rvice Requireme<br>, Zigbee IP(ZIP) | oT Object and Services,<br>erview and Approaches,<br>Representational State<br>ents for Machine-Type<br>, IPSO |  |
| Layer <sup>1</sup> / <sub>2</sub> Connectivity: Wireless Technologies<br>Mobile Network Technologies for IoT/M21<br>Overview and Motivations. Address Capabilit<br>Header Compression Schemes, Quality of Ser  | M, Layer 3 Connecties, IPv6 Protocol Orvice in IPv6, Migratic  | tivity: IPv6 Tec<br>verview, IPv6 Tu  | hnologies for the IoT:<br>innelling, IPsec in IPv6,  |  |
|  | Module -4  |   |  |  |
| Case Studies illustrating IoT Design-Introdu<br>Productivity Applications.   | iction, Home Autom   | ation, Cities, En   | vironment, Agriculture,  |  |
|  | Module -5  |   |  |  |
| Data Analytics for IoT – Introduction, Apache<br>Apache Oozie, Apache Spark, Apache Storm,<br>Health Monitoring Case Study.  |  |   |  |  |
| <ol> <li>Text Books:         <ol> <li>Internet of Things: A Hands-on Approach by Arshdeep Bahga, Vijay Madisetti, Universities Press, 2015</li> <li>Building the Internet of Things with IPv6 and MIPv6: The Evolving World of M2M</li> </ol> </li> </ol>  |  |   |  |  |
| Communications by Daniel Minoli, V<br>Reference Books:   |  | C   |  |  |
| <ol> <li>The Internet of Things Michael Mille<br/>Claire Rowland, Elizabeth Goodman</li> <li>Building the Internet of Things</li> </ol>  | et.al O'Reilly First E   | Edition, 2015   | -  |  |

4. Syllabus Timeline



| S/L | Syllabus Timeline  | Description  |
|-----|--|--|
| 1   | Week 1-2:<br>Introduction to IoT                                     | <ul> <li>What is The Internet of Things? Overview and Motivations</li> <li>Examples of Applications</li> <li>IPv6 Role, Areas of Development, and Standardization</li> <li>Scope of the Present Investigation</li> </ul>   |
| 2   | Week 3-4: IoT<br>Definitions and<br>Frameworks                       | <ul><li>IoT Definitions</li><li>IoT Frameworks</li><li>Basic Nodal Capabilities</li></ul>  |
| 3   | Week 5-6: IoT<br>Application Examples                                | <ul> <li>Overview of IoT Applications</li> <li>Smart Metering/Advanced Metering Infrastructure</li> <li>Health/Body Area Networks</li> <li>City Automation</li> <li>Automotive Applications</li> </ul>   |
| 4   | Week 7-8: IoT<br>Application Examples<br>(Continued)                 | <ul> <li>Home Automation</li> <li>Smart Cards</li> <li>Tracking</li> <li>Over-The-Air Passive Surveillance/Ring of Steel</li> <li>Control Application Examples</li> <li>Myriad Other Applications</li> </ul>   |
| 5   | Week 9-10: Logic<br>Design with MSI<br>Components and<br>PLDs        | <ul> <li>Binary Adders and Subtractors</li> <li>Comparators</li> <li>Decoders</li> <li>Encoders</li> <li>Multiplexers</li> <li>Programmable Logic Devices (PLDs)</li> </ul>  |
| 6   | Week 11-12:<br>Fundamental IoT<br>Mechanisms and Key<br>Technologies | <ul> <li>Identification of IoT Objects and Services</li> <li>Structural Aspects of the IoT</li> <li>Key IoT Technologies</li> </ul>  |
| 7   | Week 12-13:Evolving<br>IoT Standards                                 | <ul> <li>Overview and Approaches</li> <li>IETF IPv6 Routing Protocol for RPL Roll</li> <li>Constrained Application Protocol</li> <li>Representational State Transfer (REST)</li> </ul>   |
| 8   | Week 13-14:Layer <sup>1</sup> / <sub>2</sub><br>Connectivity         | <ul> <li>Wireless Technologies for the IoT</li> <li>WPAN Technologies for IoT/M2M</li> <li>Cellular and Mobile Network Technologies for IoT/M2M</li> </ul>   |
| 9   | Week 14-15: Layer 3<br>Connectivity                                  | <ul> <li>IPv6 Technologies for the IoT: Overview and Motivations</li> <li>Address Capabilities</li> <li>IPv6 Protocol Overview</li> <li>IPv6 Tunneling</li> <li>IPsec in IPv6</li> <li>Header Compression Schemes</li> <li>Quality of Service in IPv6</li> <li>Migration Strategies to IPv6</li> </ul> |
| 10  | Week 15: Data<br>Analytics for IoT                                   | <ul> <li>Introduction to Data Analytics for IoT</li> <li>Apache Hadoop</li> <li>Using HadoopMapReduce for Batch Data Analysis</li> <li>Apache Oozie</li> <li>Apache Spark</li> <li>Apache Storm</li> <li>Using Apache Storm for Real-time Data Analysis</li> </ul>                                     |



|  | • | Structural Health Monitoring Case Study |
|--|---|---|
|  |   |   |

### 5. Teaching-Learning Process Strategies

| S/L | TLP Strategies:                                 | Description   |
|-----|---|---|
| 1   | Interactive<br>Lectures                         | Use interactive lectures to introduce new concepts. Incorporate questions and discussions to engage students. |
| 2   | Coding<br>Sessions:                             | To provide practical experience in designing and implementing IoT solutions.                                  |
| 3   | Project-Based<br>Learning                       | Objective: To apply theoretical knowledge to real-world problems and develop practical IoT solutions.         |
| 4   | Case Studies and<br>Real-World<br>Examples      | Objective: To illustrate the application of IoT concepts and technologies in various domains.                 |
| 5   | Flipped<br>Classroom and<br>Blended<br>Learning | Objective: To enhance understanding and retention of IoT concepts through active learning.                    |

## 6. Assessment Details (both CIE and SEE)

CIE:

SEE:

# Theory Course with 4 credits: Integrated Professional Core Course (IPC)

CIE Split up for Integrated Professional Core Course (IPC)

|               | Components                      | Number     | Weightage  | Max. Marks | Min. Marks |
|---------------|---------------------------------|------------|------------|------------|------------|
|               | Internal Assessment-Tests (A)   | 2<br>2     | 60%<br>40% | 15<br>10   | 06 04      |
| Theory (A)    | Assignments/Quiz/Activity (B)   |            |            |            |            |
| • • •         | Total Marks                     |            | 100%       | 25         | 10         |
|               | Components                      | Number     | Weightage  | Max. Marks | Min. Marks |
|               | Record Writing                  | Continuous | 60%        | 15         | 06         |
| Laboratory(B) | Test at the end of the semester | 1          | 40%        | 10         | 04         |
| • • •         | Total Marks                     |            | 100%       | 25         | 10         |

Final CIE Marks =(A) + (B)

# Theory Course with 4, 3 and 2 Credits: Integrated Profession Core Course (IPC)

- Question paper pattern will be ten questions. Each question is set for 20marks. The medium of the question paper shall be English unless otherwise it is mentioned.
- There shall be 2 question from each module, each of the two questions under a module (with a maximum of 3 sub questions), may have mix of topics under that module if necessary.
- The students have to answer 5 full questions selecting one full question from each module.
- The question paper may include at least one question from the laboratory component.
- Marks scored will be proportionally scaled down to 50 marks

## 7. Learning Objectives

| S/L | Learning Objectives                   | Description   |
|-----|---------------------------------------|---|
| 1   | Understand the<br>Fundamentals of IoT | Grasp the basic concepts, definitions, and frameworks of IoT.<br>Learn about the motivations driving the development and adoption of<br>IoT technologies. |

| 2 | Explore IoT<br>Applications:                     | Identify and describe various applications of IoT, such as smart<br>metering, health/body area networks, city automation, automotive<br>applications, home automation, smart cards, tracking, over-the-air<br>surveillance, and control applications.   |  |  |  |  |
|---|--|---|--|--|--|--|
| 3 | Learn Key IoT<br>Technologies and<br>Mechanisms: | Understand the fundamental mechanisms of IoT and key technologies<br>such as identification of IoT objects and services, and structural aspects<br>of IoT.  |  |  |  |  |
| 4 | Study IoT Standards<br>and Protocols:            | Gain knowledge of evolving IoT standards and protocols, including<br>IETF IPV6 Routing Protocol for RPL Roll, Constrained Application<br>Protocol (CoAP), Representational State Transfer (REST), ETSI M2M,<br>3GPP Service Requirements for Machine-Type Communications,<br>CENELEC, IETF IPv6 Over Low power WPAN, and Zigbee IP (ZIP). |  |  |  |  |
| 5 | Develop Practical IoT<br>Solutions:              | Design and implement IoT systems using MSI components and<br>programmable logic devices, including binary adders and subtractors,<br>comparators, decoders, encoders, multiplexers, and programmable logic<br>devices (PLDs).   |  |  |  |  |
| 6 | Evaluate IoT Privacy<br>and Security:            | Understand the importance of privacy and security in IoT applications<br>and evaluate different techniques to ensure data protection and privacy<br>preservation in IoT systems.  |  |  |  |  |

8. Course Outcomes (COs) and Mapping with POs/ PSOs

## Course Outcomes (COs)

| COs         | Description  |  |  |  |
|-------------|--|--|--|--|
| M23MCS202.1 | Explain the basic concepts and frameworks of the Internet of Things.             |  |  |  |
| M23MCS202.2 | Describe various IoT applications in different fields like healthcare and urban  |  |  |  |
|             | management.  |  |  |  |
| M23MCS202.3 | Understand key IoT technologies and standards such as IPv6 and Zigbee.           |  |  |  |
| M23MCS202.4 | Evaluate and explain connectivity solutions for IoT, including WPAN and cellular |  |  |  |
|             | networks.  |  |  |  |
| M23MCS202.5 | Design and implement IoT solutions for real-world problems in areas like home    |  |  |  |
|             | automation and smart cities.   |  |  |  |

## **CO-PO-PSO Mapping**

| COs/POs     | PO1 | PO2 | PO3 | PSO1 | PSO2 |
|-------------|-----|-----|-----|------|------|
| M23MCS202.1 | 3   |     |     | 3    | 3    |
| M23MCS202.2 | 3   |     |     | 3    | 3    |
| M23MCS202.3 |     | 3   |     | 3    | 3    |
| M23MCS202.4 |     |     | 3   | 3    | 3    |
| M23MCS202.5 |     |     | 3   | 3    | 3    |
| M23MCS202   | 3   | 3   | 3   | 3    | 3    |

# 9. Assessment Plan

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

|          | CO1 | CO2 | CO3 | CO4 | CO5 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 10  |     |     |     |     |       |
| Module 2 |     | 10  |     |     |     |       |
| Module 3 |     |     | 10  |     |     |       |
| Module 4 |     |     |     | 10  |     |       |
| Module 5 |     |     |     |     | 10  |       |
| Total    | 10  | 10  | 10  | 10  | 10  | 50    |

|          |     |     |     | (===) |     |       |
|----------|-----|-----|-----|-------|-----|-------|
|          | CO1 | CO2 | CO3 | CO4   | CO5 | Total |
| Module 1 | 20  |     |     |       |     |       |
| Module 2 |     | 20  |     |       |     |       |
| Module 3 |     |     | 20  |       |     |       |
| Module 4 |     |     |     | 20    |     |       |
| Module 5 |     |     |     |       | 20  |       |
| Total    | 20  | 20  | 20  | 20    | 20  | 100   |

## Semester End Examination (SEE)

## **Conditions for SEE Paper Setting:**

Question paper pattern: The SEE question paper will be set for 100 marks and the marks scored will be proportionately reduced to 50.

#### **10. Future with this Subject**

The field of the Internet of Things (IoT) is rapidly evolving and presents numerous opportunities for those who gain expertise in this area. Here are some potential future prospects for students who complete this course:

## 1. Career Opportunities in Diverse Industries

Smart Cities and Urban Planning: Design and implement smart city solutions that enhance urban living through intelligent infrastructure, energy management, and public safety systems.

Healthcare: Develop IoT-enabled health monitoring systems, medical devices, and body area networks that improve patient care and facilitate remote diagnostics.

Automotive Industry: Work on connected vehicle technologies, autonomous driving systems, and vehicular networks that enhance transportation safety and efficiency.

Agriculture: Create smart farming solutions that optimize resource usage, monitor crop health, and improve yield through IoT-based systems.

Home Automation: Develop innovative home automation products and solutions that enhance convenience, security, and energy efficiency.

## 2. Advanced Research and Development

Innovative IoT Applications: Engage in cutting-edge research to explore new applications of IoT in fields like environmental monitoring, industrial automation, and smart grids.

IoT Security: Focus on developing advanced security protocols and privacy-preserving technologies to protect IoT systems from cyber threats.

Interoperability and Standards: Contribute to the development of global IoT standards and protocols that ensure seamless interoperability between diverse IoT devices and platforms.

### **3.** Entrepreneurial Ventures

IoT Startups: Leverage IoT knowledge to create startups that offer innovative IoT products and services, addressing specific market needs and challenges.

Tech Consulting: Provide consulting services to businesses looking to adopt IoT solutions, helping them integrate IoT technologies into their operations and achieve digital transformation.

#### 4. Continued Education and Specialization

Advanced Degrees: Pursue advanced degrees (Master's, Ph.D.) in related fields such as computer science, electrical engineering, or data science with a focus on IoT.

Specialization: Specialize in niche areas within IoT, such as IoT analytics, edge computing, or IoT hardware design, to become an expert in a specific domain.

## 5. Impact on Society and Environment

Sustainability Initiatives: Develop IoT solutions that contribute to sustainability by optimizing resource usage, reducing waste, and enhancing energy efficiency.

Public Safety and Disaster Management: Create IoT systems that improve public safety and facilitate efficient disaster management and response through real-time monitoring and alerts.



| and Commenter            | <b>Professional Elective Course (PE)</b> |            |
|--------------------------|--|------------|
| 2 <sup>nd</sup> Semester | Wireless Network and Mobile Computing    | M23MCS203A |

## 1. Prerequisites

| S/L | Proficiency                  | Prerequisites  |
|-----|------------------------------|--|
| 1   | Basic Networking<br>Concepts | Familiarity with concepts such as IP addressing, subnetting, routing, and switching is essential. This forms the basis for understanding how data is transmitted over networks.                    |
| 2   | Telecommunications           | Knowledge of how data is transmitted over different mediums (wired and wireless), modulation techniques, signal propagation, and spectrum management   |
| 3   | Operating Systems            | Understanding of how operating systems manage resources, handle processes, and interact with hardware, especially in the context of mobile operating systems (e.g., Android, iOS).                 |
| 4   | Programming Skills           | Basic programming knowledge, especially in languages relevant to mobile application development (e.g., Java, Swift, K), and scripting languages (e.g., Python) for network automation and analysis |
| 5   | Wireless<br>Technologies     | Familiarity with different wireless technologies such as Wi-Fi (802.11 standards), Bluetooth, cellular networks (3G, 4G, and emerging 5G), and their respective architectures and protocols.       |
| 6   | Security<br>Fundamentals     | Understanding of basic security principles and protocols relevant to wireless<br>networks, including encryption, authentication mechanisms, and secure<br>communication                            |
| 7   | Database<br>Management       | Basic knowledge of databases and their role in mobile applications and networked systems   |
| 8   | Internet of Things<br>(IoT)  | Awareness of IoT concepts and how they integrate with wireless networks and mobile computing environments.   |

# 2. Competencies

| S/L        | Competency                                 | KSA Description  |
|------------|--|--|
| <u>S/L</u> | Competency<br>Basic Networking<br>Concepts | <ul> <li>Knowledge: <ul> <li>Understanding of the basic components of a network: nodes (devices), links (communication channels), and protocols.</li> <li>Familiarity with the OSI (Open Systems Interconnection) and TCP/IP (Transmission Control Protocol/Internet Protocol) models.</li> </ul> </li> <li>Skills: <ul> <li>Ability to configure network devices (routers, switches, access points) based on network requirements.</li> <li>Proficiency in troubleshooting network connectivity issues using tools like ping, trace route, and network analyzers</li> </ul> </li> <li>Attitudes: <ul> <li>Recognition of the importance of network security and adherence to best practices to mitigate risks.</li> </ul> </li> </ul> |
|            |  | <ul> <li>Willingness to adapt to new networking technologies and<br/>methodologies as they emerge.</li> </ul>  |
| 2          | Telecommunications                         | <ul> <li>Knowledge: <ul> <li>Understanding of basic telecommunications principles, including signal transmission, modulation techniques, and multiplexing.</li> <li>Knowledge of transmission media types (e.g., guided – twisted pair, coaxial cable, opticalfibre; unguided – wireless).</li> </ul> </li> <li>Skills: <ul> <li>Ability to design and implement telecommunication networks based on specific requirements and constraints.</li> </ul> </li> </ul>   |

|   |                    | • Skill in configuring network devices (e.g., routers, switches, access   |
|---|--------------------|---|
|   |                    | points) to ensure optimal performance and reliability.  |
|   |                    | Attitudes:  |
|   |                    | • Recognition of the critical importance of reliability and security in   |
|   |                    | telecommunication networks.   |
|   |                    | • Commitment to implementing best practices to safeguard network  |
|   |                    | integrity and protect against threats.  |
|   |                    | Knowledge:  |
|   |                    | • Understanding of fundamental operating system concepts such as processes, threads, scheduling, synchronization, and deadlock handling |
| 3 | Operating Systems  | Skills:   |
| - |                    | • Ability to install and configure different operating systems (e.g.,   |
|   |                    | Windows, Linux distributions) on various hardware platforms.  |
|   |                    | Attitudes:  |
|   |                    | Recognition of the importance of maintaining system reliability and stability in operating system environments.                         |
|   |                    | Knowledge:  |
|   |                    | • Understanding of programming language syntax, semantics, and best practices.  |
|   |                    | <ul> <li>Proficiency in at least one high-level programming language (e.g.,</li> </ul>  |
|   |                    | Python, Java, C/C++, JavaScript).   |
|   |                    | Skills:   |
| 4 | Programming Skills | Ability to write clean, efficient, and well-documented code in various  |
|   | 1 Togramming 5kms  | programming languages.  |
|   |                    | • Proficiency in implementing algorithms and data structures to solve   |
|   |                    | computational problems  |
|   |                    | Attitudes:  |
|   |                    | • Recognition of the importance of writing clean, maintainable code and   |
|   |                    | adhering to coding standards and best practices   |
|   |                    | Knowledge:  |
|   |                    | • Understanding of basic wireless communication principles including  |
|   |                    | modulation techniques (e.g., AM, FM, Phase Shift Keying), signal  |
|   |                    | propagation, and interference.  |
|   |                    | Skills:   |
| 5 | Wireless           | • Ability to design and deploy wireless networks based on organizational  |
| - | Technologies       | requirements and constraints.   |
|   |                    | Attitudes:  |
|   |                    | • Recognition of the importance of maintaining reliable wireless network connectivity and optimizing performance.                       |
|   |                    |   |
|   |                    | • Commitment to implementing best practices to ensure high availability and quality of service (QoS) in wireless networks.              |
|   |                    | Knowledge:  |
|   |                    | • Understanding of fundamental security concepts such as  |
|   |                    | confidentiality, integrity, availability  |
|   |                    | Skills:   |
| 6 | Security           | • Ability to conduct security assessments (e.g., vulnerability  |
|   | Fundamentals       | assessments, penetration testing) to identify and mitigate security risks.  |
|   |                    | Attitudes:  |
|   |                    | • Recognition of the critical importance of cyber security and  |
|   |                    | commitment to maintaining a high level of security excellence.  |
| 7 | Database           |   |
| 7 | Management         | Knowledge:  |
| L |                    |   |



|   |                    | • Understanding of fundamental concepts of databases such as data models (relational, hierarchical, network), schemas, and data manipulation languages (e.g., SQL).<br>Skills:  |
|---|--------------------|---|
|   |                    | <ul> <li>Proficiency in database administration tasks such as installation, configuration, backup and recovery, and security management.</li> <li>Skill in monitoring database performance, optimizing system parameters, and ensuring data integrity and availability</li> </ul> |
|   |                    | Attitudes:  |
|   |                    | • Recognition of the importance of maintaining data integrity and accuracy in database management practices.  |
|   |                    | Knowledge:  |
|   |                    | • Understanding of the basic concepts and principles of IoT, including IoT architecture, components (sensors, actuators, IoT devices), and communication protocols  |
|   |                    | Skills:   |
| 0 | Internet of Things | • Ability to develop and prototype IoT devices using embedded systems, sensors, actuators, and microcontrollers.  |
| 8 | (IoT)              | • Skill in programming IoT devices and implementing communication protocols for data transmission   |
|   |                    | Attitudes:  |
|   |                    | • Openness to exploring innovative IoT solutions and applications to address real-world challenges.   |
|   |                    | • Creativity in designing and implementing novel IoT devices and applications   |

#### 3. Syllabus

| Wireles   | s Network and Mobile Computing |            |    |  |  |  |  |  |
|---|--------------------------------|------------|----|--|--|--|--|--|
| SEMESTER – III  |                                |            |    |  |  |  |  |  |
| Course Code   | M23MCS203A                     | CIE Marks  | 50 |  |  |  |  |  |
| Number of Lecture Hours/Week(L: T:  | (2:0:0:2)                      | SEE Marks  | 50 |  |  |  |  |  |
| P: S)   |                                |            |    |  |  |  |  |  |
| Total Number of Lecture Hours40 hours Theory + 8-10 Lab slotsTotal Marks100 |                                |            |    |  |  |  |  |  |
| Credits   | 03                             | Exam Hours | 03 |  |  |  |  |  |
| Course objectives: This course will en                                      | able students to:              | •          | 1  |  |  |  |  |  |

- Manage TCP/IP and prepare foundation for future Networks
- enabling technologies for wireless networking and mobile computing
- wireless networking standards
- mobile computing applications

#### Module -1

**Mobile Computing Architecture:** Architecture for Mobile Computing, 3-tier Architecture, Design Considerations for Mobile Computing. Emerging Technologies: Wireless broadband (WiMAX), Mobile IP: Introduction, discovery, Registration, Tunneling, Cellular IP, Mobile IP with Ipv6. **Wireless Networks : Global Systems for Mobile Communication (GSM):** GSM Architecture, Entities, Call routing in GSM, PLMN Interface, GSM Addresses and Identities, Network Aspects in GSM, Mobility Management, GSM Frequency allocation. Short Service Messages (SMS): Introduction to SMS, SMS Architecture, SMMT, SMMO, SMS as Information bearer, applications, GPRS and Packet Data Network, GPRS Network Architecture, GPRS Network Operations, Data Services in GPRS, Applications for GPRS, Billing and Charging in GPRS

Module -2

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Spread Spectrum technology, IS-95, CDMA versus GSM, Wireless Data, Third Generation Networks, Applications on 3G, **Mobile Client:** Moving beyond desktop, Mobile handset overview, Mobile phones and their features, PDA, Design Constraints in applications for handheld devices.

#### Module -3

**Mobile OS and Computing Environment:** Smart Client Architecture, The Client: User Interface, Data Storage, Performance, Data Synchronization, Messaging. The Server: Data Synchronization, Enterprise Data Source, Messaging. **Mobile Operating Systems:** WinCE, Palm OS, Symbian OS, Linux, Proprietary OS Client Development: The development process, Need analysis phase, Design phase, Implementation and Testing phase, Deployment phase, Development Tools, Device Emulators

#### Module -4

**Building Wireless Internet Applications:** Thin client overview: Architecture, the client, Middleware, messaging Servers, Processing a Wireless request, Wireless Applications Protocol (WAP) Overview, Wireless Languages: Markup Languages, HDML, WML, HTML, cHTML, XHTML, VoiceXML

#### Module -5

**J2ME**: Introduction, CDC, CLDC, MIDP; Programming for CLDC, MIDlet model, Provisioning, MIDlet life-cycle, Creating new application, MIDlet event handling, GUI in MIDP, Low level GUI Components, Multimedia APIs; Communication in MIDP, Security Considerations in MIDP.

#### Textbook:

- 1. Mobile Computing, Technology, Applications and Service Creation Ashok Talukder, Roopa Yavagal, Hasan Ahmed Tata McGraw Hill 2<sup>nd</sup> Edition, 2010.
- 2. Mobile and Wireless Design Essentials Martyn Mallik Wiley India 2003

### **Reference Books**

- 1. Mobile Computing Raj kamal Oxford University press 2007, Wireless Communications and Networks, 3G and Beyond ItiSaha Misra Tata McGraw Hill 2009.
- 2. Mobile and Wireless Design Essentials Martyn Mallik Wiley India 2003

#### 4. Syllabus Timeline

| S/L | Syllabus Timeline   | Description  |  |  |  |
|-----|---|--|--|--|--|
| 1   | Week 1-2: Mobile<br>Computing Architecture:   | Describes about mobile computing architecture, GSM Architecture,<br>Entities, Call routing in GSM, PLMN Interface, GSM Addresses and<br>Identities, Network Aspects in GSM, Mobility Management, |  |  |  |
| 2   | Week 3-4: Wireless<br>Networks : Global<br>Systems for Mobile<br>Communication (GSM): | Describes about wireless technology, Mobile handset overview,<br>Mobile phones and their features, PDA, Design Constraints in<br>applications for handheld devices.                              |  |  |  |
| 3   | Week 5-6: Mobile Client:  | <b>Mobile Client:</b> Moving beyond desktop, Mobile handset overview<br>Mobile phones and their features, PDA, Design Constraints in<br>applications for handheld devices.                       |  |  |  |
| 4   | Week 7-8: Mobile<br>Operating Systems:  | Describes WinCE, Palm OS, Symbian OS, Linux, Proprietary OS Client Development.  |  |  |  |
| 5   | Week 9-10: Building<br>Wireless Internet<br>Application <b>s</b>                      | Thin client overview: Architecture, the client, Middleware, messaging Servers.   |  |  |  |
| 6   | Week 11-12: J2ME  | Introduces to, CDC, CLDC, MIDP; Programming for CLDC, MIDlet model, Provisioning, MIDlet life-cycle, Creating new application.   |  |  |  |

#### 5. Teaching-Learning Process Strategies

| S/L | TLP Strategies: | Description |
|-----|-----------------|-------------|
|     |                 |             |



| 1 | Lecture Method                                | Utilize various teaching methods within the lecture format to reinforce competencies.        |  |  |  |
|---|---|--|--|--|--|
| 2 | Video/Animation                               | Incorporate visual aids like videos/animations to enhance understanding of network concepts. |  |  |  |
| 3 | Collaborative<br>Learning                     | Encourage collaborative learning for improved competency application.                        |  |  |  |
| 4 | Higher Order<br>Thinking (HOTS)<br>Questions: | Pose HOTS questions to stimulate critical thinking related to each competency.               |  |  |  |
| 5 | Problem-Based<br>Learning (PBL)               | Implement PBL to enhance analytical skills and practical application of competencies         |  |  |  |
| 6 | Multiple<br>Representations                   | Introduce topics in various representations to reinforce competencies                        |  |  |  |

#### 6. Assessment Details (both CIE and SEE)

This section of regulations is applicable to all theory-based courses. The minimum CIE marks requirement is 40% of maximum marks in each component.

## **CIE Split up for Professional Elective Course (PE)**

|      | 3. Components                 | Number      | Weightage | . Max.<br>Marks | Min. Marks |
|------|-------------------------------|-------------|-----------|-----------------|------------|
| (i)  | Internal Assessment-Tests (A) | 3           | 50%       | 25              | 10         |
| (ii) | Assignments/Quiz/Activity (B) | 2           | 50%       | 25              | 10         |
|      | Total Marks                   | Total Marks |           |                 |            |

#### Final CIE Marks =(A) + (B)

Average internal assessment shall be the average of the best two test marks from the 3 tests conducted. **Semester End Examinations: PG Programmes** 

Theory Course with 4, 3 and 2 Credits: Professional Core Course (PC)/Professional Elective/Open Elective

- Question paper pattern will be ten questions. Each question is set for 20marks. The medium of the question paper shall be English unless otherwise it is mentioned.
- There shall be 2 question from each module, each of the two questions under a module (with a maximum of 3 sub questions), may have mix of topics under that module if necessary.
- The students have to answer 5 full questions selecting one full question from each module.
- Marks scored will be proportionally scaled down to 50 marks

## 7. Learning Objectives

| S/L | Learning<br>Objectives                                  | Description   |
|-----|---|---|
| 1   | Understanding<br>Wireless<br>Networking<br>Fundamentals | <ul> <li>Describe the principles of wireless communication.</li> <li>Explain the differences between various wireless technologies (e.g., Wi-Fi, Bluetooth, cellular networks).</li> <li>Define key terms such as frequency bands, modulation techniques, and signal propagation</li> </ul> |
| 2   | Wireless<br>Network<br>Architecture<br>and Protocols    | <ul> <li>Identify the components of a wireless network (e.g., access points, routers, clients).</li> <li>Describe the protocols used in wireless networks (e.g., IEEE 802.11 standards, TCP/IP stack in wireless environments</li> </ul>  |
| 4   | Mobile<br>Computing<br>Devices and<br>Technologies      | <ul> <li>Define mobile computing and its evolution.</li> <li>Identify different types of mobile computing devices (e.g., smart phones, tablets, wearables) and their features.</li> </ul>   |



| 5 | Mobile<br>Operating<br>Systems and<br>Applications                         | Compare and contrast mobile operating systems (e.g., Android, iOS).<br>Discuss the architecture of mobile applications and their development<br>frameworks (e.g., Android Studio, Xcode).  |  |  |  |  |  |
|---|--|--|--|--|--|--|--|
| 6 | Mobile<br>Security and<br>Privacy  | <ul> <li>Identify common security threats to mobile devices and networks (e.g., malware, phishing).</li> <li>Discuss security measures to protect mobile devices (e.g., encryption, secure authentication methods).</li> </ul>   |  |  |  |  |  |
| 7 | Location-<br>Based Services<br>and IoT<br>Integration                      | <ul> <li>Explain the concept of location-based services (LBS) and their applications in mobile computing.</li> <li>Discuss how mobile computing integrates with Internet of Things (IoT) devices and services.</li> </ul>  |  |  |  |  |  |
| 8 | Mobile<br>Application<br>Development<br>and<br>Optimization                | <ul> <li>Outline the steps involved in mobile application development (e.g., UI/UX design, coding, testing).</li> <li>Optimize mobile applications for performance, battery efficiency, and responsiveness.</li> </ul>   |  |  |  |  |  |
| 9 | Emerging<br>Trends in<br>Wireless<br>Networking<br>and Mobile<br>Computing | <ul> <li>Explore current trends in wireless networking (e.g., Wi-Fi 6, mesh networks) and mobile computing (e.g., edge computing, augmented reality).</li> <li>Discuss the impact of emerging technologies (e.g., 5G, AI in mobile) on future wireless and mobile computing scenarios</li> </ul> |  |  |  |  |  |

# 8. Course Outcomes (COs) and Mapping with POs/ PSOs

## Course Outcomes (COs)

| COs          | Description  |  |  |  |  |
|--------------|--|--|--|--|--|
| M23MCS203A.1 | Apply state of art techniques in wireless communication  |  |  |  |  |
| M23MCS203A.2 | Analyse the third-generation network and Mobile OS and Computing Environment technologies  |  |  |  |  |
| M23MCS203A.3 | Design programs for CLDC, MIDP let model and security concerns using Wireless Languages: Markup Languages, HDML, WML, HTML, cHTML, XHTML, VoiceXML |  |  |  |  |
| M23MCS203A.4 | Understand and apply client architecture for wireless communication  |  |  |  |  |
| M23MCS203A.5 | Develop Programming for CLDC, MIDlet model, Provisioning, MIDlet life-cycle, Creating new application  |  |  |  |  |

# **CO-PO-PSO Mapping**

| COs/POs      | PO1 | PO2 | PO3 | PSO1 | PSO2 |
|--------------|-----|-----|-----|------|------|
| M23MCS203A.1 | 2   | -   | -   | -    | 2    |
| M23MCS203A.2 |     | 2   | -   | -    | 3    |
| M23MCS203A.3 | -   |     | 2   | 2    | 3    |
| M23MCS203A.4 | 2   | 3   |     | 2    | 3    |
| M23MCS203A.5 | 2   | 2   |     | 3    | 3    |
| M23MCS203A   | 2   | 2   | 2   | 2    | 2    |

## 9. Assessment Plan

# **Continuous Internal Evaluation (CIE)**

|          | CO1 | CO2 | CO3 | CO2 | CO3 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 10  |     |     |     |     | 10    |
| Module 2 |     | 10  |     |     |     | 10    |



| Module 3 |    |    | 10 |    |    | 10 |
|----------|----|----|----|----|----|----|
| Module 4 |    |    |    | 10 |    | 10 |
| Module 5 |    |    |    |    | 10 | 10 |
| Total    | 10 | 10 | 10 | 10 | 10 | 50 |

| Semester End Examination (SEE) |     |     |     |     |     |       |
|--------------------------------|-----|-----|-----|-----|-----|-------|
|                                | CO1 | CO2 | CO3 | CO2 | CO3 | Total |
| Module 1                       | 20  |     |     |     |     | 20    |
| Module 2                       |     | 20  |     |     |     | 20    |
| Module 3                       |     |     | 20  |     |     | 20    |
| Module 4                       |     |     |     | 20  |     | 20    |
| Module 5                       |     |     |     |     | 20  | 20    |
| Total                          | 20  | 20  | 20  | 20  | 20  | 100   |

#### Semester End Examination (SEE)

#### **Conditions for SEE Paper Setting:**

Each module of SEE question paper should be allocated with questions for 20% of the total SEE marks

#### 10. Future with this Subject

#### **5G Evolution and Adoption:**

- Continued deployment and optimization of 5G networks worldwide, offering unprecedented speed, reliability, and low latency. This will enable seamless connectivity for high-definition video streaming, real-time gaming, and immersive experiences like AR/VR.
- Expansion of 5G into diverse industries such as healthcare (telemedicine, remote surgery), transportation (autonomous vehicles, smart logistics), and manufacturing (smart factories, industrial IoT).

## Internet of Things (IoT) Advancements:

- Rapid growth in the number and diversity of IoT devices, leading to a highly interconnected world where smart devices communicate autonomously for improved efficiency and decision-making
- Integration of edge computing with IoT to process data closer to the source, reducing latency and enhancing real-time analytics capabilities. This will support critical applications in smart cities, environmental monitoring, and personalized healthcare.

## Artificial Intelligence (AI) and Machine Learning (ML):

- Integration of AI and ML algorithms into mobile devices and wireless networks to optimize performance, predict user behavior, and automate processes. This includes AI-powered voice assistants, personalized recommendations, and predictive maintenance in IoT deployments.
- AI-enabled network management tools for autonomous network optimization, predictive maintenance, and cybersecurity threat detection to ensure robust and secure wireless connectivity.

## Security and Privacy Enhancements:

- Heightened focus on cyber security with the implementation of advanced encryption standards, block chain for secure transactions, and biometric authentication for mobile devices. This will address evolving cyber threats and protect sensitive data across wireless networks and IoT ecosystems.
- Stricter privacy regulations and frameworks governing data collection, storage, and usage to protect user privacy rights and build trust in digital interactions.

## **Emerging Technologies and Applications:**

- Continued advancements in augmented reality (AR), virtual reality (VR), and mixed reality (MR) applications for immersive gaming, virtual tourism, remote collaboration, and training.
- Exploration of block chain technology for decentralized IoT networks, secure peer-to-peer transactions, and

#### **Environmental and Sustainability Initiatives:**



• Integration of energy-efficient technologies in wireless networks and mobile devices to reduce carbon footprint and enhance sustainability. This includes energy harvesting techniques, smart grids, and eco-friendly IoT solutions.

## **Global Connectivity and Digital Inclusion:**

• Efforts to bridge the digital divide with initiatives aimed at expanding broadband access in underserved regions through satellite internet, mesh networks, and low-earth orbit (LEO) satellites.



| ando                     | <b>Professional Elective Course (PE)</b> | Maanggaaab |
|--------------------------|--|------------|
| 2 <sup>nd</sup> Semester | Natural Language Processing              | M23MCS203B |

## 1. Prerequisites

| S/L | Proficiency                                   | Prerequisites   |
|-----|---|---|
| 1   | Foundational<br>Mathematics and<br>Statistics | Linear Algebra: Proficiency in linear algebra, including vectors, matrices, eigen values, and eigen vectors. Calculus: Understanding of differential and integral calculus, especially optimization techniques. Probability and Statistics: Strong grasp of probability theory, statistical inference, and probabilistic models.              |
| 2   | Introduction to<br>Machine Learning           | <b>Supervised and Unsupervised Learning:</b> Solid understanding of machine learning algorithms, including classification, regression, clustering, and evaluation metrics. <b>Neural Networks:</b> Familiarity with neural network architectures, training processes, and applications in NLP.  |
| 3   | Programming<br>Proficiency                    | <b>Python Programming:</b> Advanced proficiency in Python, with experience using NLP libraries such as NLTK, spaCy, and machine learning frameworks like Tensor Flow and PyTorch. <b>Algorithm and Data Structures:</b> Strong understanding of fundamental algorithms and data structures, including trees, graphs, and dynamic programming. |
| 4   | Linguistics                                   | <b>Syntax and Semantics:</b> Basic knowledge of syntactic structures and semantic meaning in natural languages. <b>Morphology and Phonology:</b> Understanding of word formation processes and sound structures in languages.   |
| 5   | Core Computer<br>Science Knowledge            | <b>Data Structures and Algorithms:</b> Advanced coursework or experience in data structures and algorithms. <b>Theory of Computation:</b> Understanding of formal languages, automata theory, and complexity theory.  |

# 2. Competencies

|     | ipetencies  |   |
|-----|---|---|
| S/L | Competency  | KSA Description   |
| 1   | Understanding<br>the Origins and<br>Challenges of<br>NLP              | <ul> <li>Knowledge: Understanding of the fundamental concepts such as syntax, semantics, and pragmatics in the context of NLP.</li> <li>Skills: Able to analyze the impact of these challenges on the development and application of NLP technologies. Able to evaluate different NLP models and methods in terms of their efficacy and applicability.</li> <li>Attitudes: Curiosity about the development of technology, Persistence in overcoming obstacles and challenges in NLP.</li> </ul>   |
| 2   | Implementing<br>and Applying<br>Part-of-Speech<br>Tagging             | <ul> <li>Knowledge: Basic concepts of part-of-speech (POS) tagging. Common challenges in POS tagging, such as ambiguity and unknown words. Different POS tagging methods and algorithms (e.g., rule-based, statistical, machine learning).</li> <li>Skills: Able to apply POS tagging to various corpora effectively. Able to evaluate and improve the accuracy of POS tagging systems.</li> <li>Attitudes: Analytical mindset for understanding language structures, Persistence in improving tagging performance.</li> </ul>  |
| 3   | Applying<br>Dependency-<br>Path Kernels for<br>Relation<br>Extraction | <ul> <li>Knowledge:</li> <li>Understanding how dependency paths represent syntactic and semantic relationships. Common challenges in relation extraction using dependency paths.</li> <li>Skills: Implement dependency-path kernel methods for extracting relationships between words in text. Evaluate the effectiveness of dependency-path kernels in various NLP tasks through experimental evaluation.</li> <li>Attitudes: Curiosity and willingness to explore advanced NLP techniques.</li> <li>Attention to detail in model implementation and evaluation. Persistence in refining and optimizing NLP models.</li> </ul> |



| 4 | Utilizing Latent<br>Semantic<br>Analysis (LSA)<br>for Text<br>Cohesion<br>Measurement | <ul> <li>Knowledge: Understanding the concepts of textual cohesion and its importance in NLP. Techniques for identifying and measuring text cohesion using LSA.</li> <li>Skills: Use tools like Coh- Metrix to apply LSA in evaluating text types.</li> <li>Able to evaluate and improve the accuracy of cohesion measurement using LSA.</li> <li>Attitudes: Analytical mindset to understand and apply complex models.</li> <li>Commitment to continuous learning and improvement in text analysis techniques.</li> </ul>                                      |
|---|---|---|
| 5 | Designing and<br>Evaluating<br>Information<br>Retrieval<br>Systems                    | <ul> <li>Knowledge: Understanding of classical, non-classical, and alternative IR models.</li> <li>Key features and functionalities of IR systems. Methods for assessing and comparing IR models.</li> <li>Skills: Design and implement effective IR systems based on various models.</li> <li>Evaluate the performance of IR systems using appropriate metrics. Fine-tune IR systems for improved accuracy and relevance.</li> <li>Attitudes: Analytical thinking for system design and evaluation. Persistence in improving IR system performance.</li> </ul> |

# 3. Syllabus

| Natural Language Processing   |  |                    |               |  |  |
|---|--|--------------------|---------------|--|--|
|   | SEMESTER – II  |                    | 1             |  |  |
| Course Code   | M23MCS203B   | CIE Marks          | 50            |  |  |
| Number of Lecture Hours/Week(L: T: P:   | (2:0:0:2)  | SEE Marks          | 50            |  |  |
| S)  |  |                    |               |  |  |
| Total Number of Lecture Hours   | 40 hours   | Total Marks        | 100           |  |  |
| Credits   | 03   | Exam Hours         | 03            |  |  |
| Course objectives: This course will enable  |  |                    |               |  |  |
| To develop a Comprehensive Und  | -  |                    |               |  |  |
| <ul> <li>To master Word-Level and Syntac</li> </ul>   |  |                    |               |  |  |
| To enhance Relation Extraction an   | • •  |                    |               |  |  |
| <ul> <li>To apply Advanced Information R</li> </ul>   | etrieval and Text Mining Techniques.   |                    |               |  |  |
| To design and Evaluate Information  | on Retrieval Systems and Utilize Lexica  | l Resources.       |               |  |  |
|   | Module -1  |                    |               |  |  |
| Overview And Language Modelling: Over   |  |                    |               |  |  |
| Language and Grammar- Processing Indi   |  |                    | l, Language   |  |  |
| Modelling: Various grammars - based Lang  |  | lel.               |               |  |  |
|   | Module -2  |                    |               |  |  |
| WORD LEVEL AND SYNTACTIC ANA  |  |                    |               |  |  |
| Automata-Morphological Parsing-Spelling   |  |                    |               |  |  |
| Speech Tagging. Syntactic Analysis: Conte   |  | -Probabilistic Par | rsing.        |  |  |
|   | Module -3  |                    |               |  |  |
| Extracting Relations from Text: From Word Sequences to Dependency Paths: Introduction, Subsequence Kernels for Relation Extraction, A Dependency-Path Kernel for Relation Extraction and Experimental Evaluation. Mining Diagnostic Text Reports by Learning to Annotate Knowledge Roles: Introduction, Domain Knowledge and Knowledge Roles, Frame Semantics and Semantic Role Labelling, Learning to Annotate Cases with Knowledge Roles and Evaluations. A Case Study in Natural Language Based Web Search: In Fact System Overview, The Global Security.org Experience. |  |                    |               |  |  |
|   | Module -4  |                    |               |  |  |
|   | Evaluating Self-Explanations in iSTART: Word Matching, Latent Semantic Analysis, and Topic Models: |                    |               |  |  |
| Introduction, iSTART: Feedback Systems, iSTART: Evaluation of Feedback Systems, Textual Signatures:   |  |                    |               |  |  |
| Identifying Text-Types Using Latent Semantic Analysis to Measure the Cohesion of Text Structures:   |  |                    |               |  |  |
| Introduction, Cohesion, Coh- Matrix, Approaches to Analyzing Texts, Latent Semantic Analysis, Predictions, Results of Experiments. Automatic Document Separation: A Combination of Probabilistic Classification and   |  |                    |               |  |  |
| Finite-State Sequence Modelling: Introduc   |  |                    |               |  |  |
| Sequence Mapping Problem, Results. Evolv  |  |                    |               |  |  |
| Related Work, A Semantically Guided Mod   |  | lancically-Daseu   | reat winning. |  |  |
| Related Work, A Semanticarry Sulded Mot   | ter for Effective Text Willing.  |                    |               |  |  |

#### Module -5

INFORMATION RETRIEVAL AND LEXICAL RESOURCES: Information Retrieval: Design features of Information Retrieval Systems-Classical, Non classical, Alternative Models of Information Retrieval–valuation Lexical Resources: World Net- Frame Net-Stemmers-POS Tagger- Research Corpora.

#### **Textbooks:**

1.Natural Language Processing and Information Retrieval, Tanveer Siddiqui, U.S.Tiwary, Oxford University Press, 2008

2.Natural Language Processing and Text Mining, Anne Kao and Stephen R. Potee, Springer-Verlag London Limited, 2007

#### **Reference books:**

 Speech and Language Processing: An introduction to Natural Language Processing ,Computational Linguistics and Speech Recognition ,Daniel Jurafsky and James H Martin, Prentice Hall, 20082ndEdition
 Natural Language Understanding, James Allen, Benjamin/Cummings publishing company, 2ndedition,1995.
 Information Storage and Retrieval systems, Gerald J. Kowalski and Mark.T. Maybury, Kluwer academic Publishers, 2000

4. Natural Language Processing with Python, Steven Bird, Ewan Klein, EdwardLoper, O'ReillyMedia, 2009

#### 4. Syllabus Timeline

| S/L | Syllabus Timeline                                    | Description  |
|-----|--|--|
| 1   | Week 1-2:<br>OVERVIEW<br>AND<br>LANGUAGEMODELING     | <b>Knowledge:</b> Understanding the historical evolution of NLP from its inception to the present day. Familiarity with various applications of NLP across different domains.<br><b>Skills:</b> Ability to analyze language structures and grammatical rules relevant to NLP. Proficiency in designing and evaluating information retrieval systems.   |
| 2   | Week 3-5:<br>LANGUAGE<br>MODELING                    | <ul> <li>Knowledge: Understanding of different types of grammar-based language models, such as rule-based systems, transformational grammar, and syntactic parsing models. Familiarity with statistical approaches to language modeling, such as n-gram models, hidden Markov models (HMMs), and neural language models (e.g., LSTM, Transformer).</li> <li>Skills: Ability to implement statistical language models and evaluate their performance using large-scale corpora and standard evaluation datasets. Proficiency in interpreting experimental results and drawing meaningful conclusions to advance the state-of-the-art in language modeling research and applications.</li> </ul> |
| 3   | Week 6-8:<br>WORD LEVEL AND<br>SYNTACTIC<br>ANALYSIS | <ul> <li>Knowledge: Understanding of regular expressions and their application in text processing and pattern matching.</li> <li>Knowledge of context-free grammars (CFGs) and their role in formal language theory and parsing algorithms.</li> <li>Skills: Ability to implement regular expressions and finite-state automata for lexical analysis tasks using programming languages like Python, Java, or Perl. Capability to analyze and interpret syntactic structures of sentences using context-free grammars and syntactic parsing techniques.</li> </ul>  |
| 4   | Week 9-11:<br>EXTRACTING<br>RELATIONS FROM<br>TEXT   | <ul> <li>Knowledge: Understanding of subsequence kernels as a method for relation extraction, which involves identifying and extracting relationships between entities based on subsequences of words or tokens. Understanding of domain-specific knowledge required for annotating and interpreting diagnostic text reports, such as medical or technical domains.</li> <li>Skills: Ability to implement and apply subsequence kernels and dependency-path kernels for relation extraction tasks using machine learning and natural language processing techniques. Skill in</li> </ul>   |



| 5 | Week 11-13:<br>EVALUATING SELF-<br>EXPLANATIONS IN<br>iSTART     | <ul> <li>interpreting experimental results and conducting evaluations to measure<br/>the effectiveness of relation extraction and semantic role labelling<br/>systems in specific domains.</li> <li>Knowledge: Understanding cohesion in text structures involves<br/>recognizing how linguistic elements like conjunctions, lexical repetition,<br/>and grammatical cohesion contribute to coherence and readability.</li> <li>Familiarity with probabilistic classification and finite-state sequence<br/>modelling techniques includes understanding and applying probabilistic<br/>models for document classification and separation based on content.</li> <li>Skills: Ability to analyze texts using tools like Coh-Metrix and latent</li> </ul>  |
|---|--|---|
|   | ISTAKI   | semantic analysis (LSA) to measure cohesion objectively, applying<br>quantitative measures to assess text coherence. Proficiency in<br>implementing sequence mapping algorithms for automatic document<br>classification, including practical skills in data preparation, feature<br>extraction, and model evaluation for document separation tasks.<br><b>Knowledge:</b> Understanding the design features of information retrieval  |
| 6 | Week 13-15:<br>INFORMATION<br>RETRIEVAL AND<br>LEXICAL RESOURCES | <ul> <li>systems, including classical, non-classical, and alternative models, involves knowledge of Boolean retrieval, vector space models, probabilistic retrieval models, and their respective strengths and weaknesses. Familiarity with prominent lexical resources in natural language processing, such as WordNet, Frame Net, stemmers, part-of-speech taggers, and research corpora, is essential for understanding their structure, coverage, and applications in computational linguistics and text mining.</li> <li>Skills: Ability to evaluate information retrieval systems using performance metrics such as precision, recall, and F1 score, along with proficiency in designing retrieval algorithms considering factors like query expansion and relevance feedback. Proficiency in utilizing lexical resources for NLP tasks such as word sense disambiguation, semantic similarity measurement, and information extraction, as well as adapting these resources to specific research or application domains.</li> </ul> |

# 5. <u>Teaching-Learning Process Strategies</u>

| S/L | TLP Strategies:                 | Description   |
|-----|---------------------------------|---|
| 1   | Lecture Method                  | Utilize various teaching methods within the lecture format to reinforce competencies.       |
| 2   | Video/Animation                 | Incorporate visual aids like videos/animations to enhance understanding of concepts.        |
| 3   | Collaborative<br>Learning       | Encourage collaborative learning for improved competency application                        |
| 4   | Problem-Based<br>Learning (PBL) | Implement PBL to enhance analytical skills and practical application of competencies        |
| 5   | Multiple<br>Representations     | Introduce topics in various representations to reinforce competencies                       |
| 6   | Real-World<br>Application       | Discuss practical applications to connect theoretical concepts with real world competencies |
| 7   | Programming<br>Assignments      | Assign programming tasks to reinforce practical skills associated with competencies.        |

# 6. Assessment Details (SEE)

This section of regulations is applicable to all theory-based courses. The minimum CIE marks requirement is 40% of maximum marks in each component.

|   | CIE Split up for Professional Elective Course (PE) |    |               |    |    |
|---|--|----|---------------|----|----|
| Components Number Weightage Max.<br>Marks |  |    | Min.<br>Marks |    |    |
| (i)                                       | Internal Assessment-Tests (A)                      | 3  | 50%           | 25 | 10 |
| (ii)                                      | Assignments/Quiz/Activity (B)                      | 2  | 50%           | 25 | 10 |
|   | Total Mar  | ks |               | 50 | 20 |

# Final CIE Marks =(A) + (B)

Average internal assessment shall be the average of the best two test marks from the 3 tests conducted. **Semester End Examinations: PG Programmes** 

- Question paper pattern will be ten questions. Each question is set for 20marks. The medium of the question paper shall be English unless otherwise it is mentioned.
- There shall be 2 question from each module, each of the two questions under a module (with a maximum of 3 sub questions), may have mix of topics under that module if necessary.
- The students have to answer 5 full questions selecting one full question from each module.
- Marks scored will be proportionally scaled down to 50 marks

## 7. Learning Objectives

| S/L | Learning<br>Objectives  | Description   |
|-----|---|---|
| 1   | Understand<br>the<br>foundational<br>principles and<br>challenges of<br>NLP | Gain a comprehensive understanding of the origins, evolution, and key challenges in Natural Language Processing (NLP). Explore how NLP intersects with linguistics and computational sciences, addressing issues like ambiguity, context sensitivity, and linguistic diversity.   |
| 2   | Analyze<br>language<br>structures and<br>computational<br>processing        | Examine the principles of language and grammar in the context of NLP. Investigate techniques for processing Indian languages, including script variations, morphological analysis, and resource limitations. Evaluate the role of linguistic structures in enhancing NLP applications such as machine translation, sentiment analysis, and information retrieval.   |
| 3   | Explore<br>diverse<br>language<br>modelling<br>techniques                   | Study a range of grammar-based and statistical language models used in NLP.<br>Compare and contrast their methodologies and applications in tasks such as text<br>generation, speech recognition, and language understanding.   |
| 4   | Master word-<br>level analysis<br>and syntactic<br>parsing                  | Develop proficiency in using regular expressions, finite-state automata, and<br>morphological parsing for word-level analysis. Implement techniques for spelling<br>error detection, correction, and categorization of words into classes. Gain insights into<br>part-of-speech tagging and its significance in syntactic and semantic analysis.  |
| 5   | Learn<br>advanced<br>techniques for<br>extracting<br>relations from<br>text | Explore methods like subsequence kernels and dependency-path kernels for extracting relations from text data. Evaluate these techniques through experimental studies in relation extraction and semantic role labelling tasks. Analyze case studies in mining diagnostic text reports and annotating knowledge roles in specific domains.   |
| 6   | Evaluate and<br>apply<br>advanced NLP<br>methodologies                      | Assess methodologies such as word matching, latent semantic analysis (LSA), and topic modelling for evaluating self-explanations and measuring textual cohesion. Explore their application in educational technology and cognitive modelling contexts. Investigate automatic document separation techniques using probabilistic classification and finite-state sequence modelling for efficient data processing and retrieval. |

|   | Examine       |  |  |  |  |  |
|---|---------------|--|--|--|--|--|
|   | design        | Investigate the design features of classical, non-classical, and alternative models in |  |  |  |  |
|   | features and  | Information Retrieval (IR). Evaluate their performance metrics such as precision,      |  |  |  |  |
| 7 | evaluation of | recall, and relevance ranking. Explore lexical resources such as WordNet, FrameNe      |  |  |  |  |
|   | IR systems    | stemmers, POS taggers, and research corpora, and their role in semantic analysis and   |  |  |  |  |
|   | and lexical   | machine learning applications.   |  |  |  |  |
|   | resources     |  |  |  |  |  |

# 8. Course Outcomes (COs) and Mapping with POs/ PSOs

| Course Outcomes (COs) |  |  |  |  |  |
|-----------------------|--|--|--|--|--|
| COs                   | Description  |  |  |  |  |
| M23MCS203B.1          | Understand the foundational concepts of NLP and language modelling.                    |  |  |  |  |
| M23MCS203B.2          | Apply language analysis and modelling techniques.                                      |  |  |  |  |
| M23MCS203B.3          | Evaluate advanced text processing methods and relation extraction techniques.          |  |  |  |  |
| M23MCS203B.4          | Analyze and critique NLP methodologies and techniques.                                 |  |  |  |  |
| M23MCS203B.5          | Create and design solutions using information retrieval systems and lexical resources. |  |  |  |  |

| CO-PO-PSO Mapping |     |     |     |      |      |  |
|-------------------|-----|-----|-----|------|------|--|
| COs/POs           | PO1 | PO2 | PO3 | PSO1 | PSO2 |  |
| M23MCS203B.1      | -   | -   | 3   | 3    | 3    |  |
| M23MCS203B.2      | 3   | -   | -   | 3    | 3    |  |
| M23MCS203B.3      | 3   | -   | -   | 3    | 3    |  |
| M23MCS203B.4      | -   | 3   | -   | 3    | 3    |  |
| M23MCS203B.5      | -   | 3   | -   | 3    | 3    |  |
| M23MCS203B        | 3   | 3   | 3   | 3    | 3    |  |

#### 9. Assessment Plan

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

|          | CO1 | CO2 | CO3 | CO4 | CO5 | Total |  |
|----------|-----|-----|-----|-----|-----|-------|--|
| Module 1 | 10  |     |     |     |     | 10    |  |
| Module 2 |     | 10  |     |     |     | 10    |  |
| Module 3 |     |     | 10  |     |     | 10    |  |
| Module 4 |     |     |     | 10  |     | 10    |  |
| Module 5 |     |     |     |     | 10  | 10    |  |
| Total    | 10  | 10  | 10  | 10  | 10  | 50    |  |

## **Semester End Examination (SEE)**

|          | CO1 | CO2 | CO3 | CO4 | CO5 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 20  |     |     |     |     | 20    |
| Module 2 |     | 20  |     |     |     | 20    |
| Module 3 |     |     | 20  |     |     | 20    |
| Module 4 |     |     |     | 20  |     | 20    |
| Module 5 |     |     |     |     | 20  | 20    |
| Total    | 20  | 20  | 20  | 20  | 20  | 100   |

# **Conditions for SEE Paper Setting:**

Each module of SEE question paper should be allocated with questions for 20% of the total SEE marks

#### 10. Future with this Subject

The "Natural Language Processing" as a professional elective course in the second semester of the M.Tech program lays a strong foundation for several future courses in the postgraduate program. The contributions of this subject extend across various areas, enhancing the students' understanding and skills in the field of natural language processing. Here are some notable contributions:

### 1. Advancements in Language Understanding

As NLP algorithms become more sophisticated, they will enhance our ability to accurately understand and interpret nuances in human language. From sentiment analysis to context-aware dialogue systems, NLP will facilitate deeper and more meaningful interactions between humans and machines.

#### 2. Personalized Learning and Education

In education, NLP holds promise for personalized learning experiences. Adaptive tutoring systems powered by NLP can cater to individual learning styles and pace, offering real-time feedback and customized educational content.

#### 3. Healthcare and Biomedical Applications

NLP is set to revolutionize healthcare through improved clinical decision support systems. By analyzing vast amounts of medical literature and patient records, NLP can assist in diagnosis, treatment recommendation, and drug discovery.

### 4. Enhanced Customer Experience and Business Insights

Businesses are increasingly leveraging NLP for customer service automation, sentiment analysis of customer feedback, and market trend prediction. Chatbots and virtual assistants equipped with NLP capabilities are transforming customer interaction models.

#### 5. Ethical Considerations and Bias Mitigation

As NLP technologies proliferate, ethical considerations such as bias in language models and privacy concerns will become critical. Future research will focus on developing fairer and more transparent NLP systems that uphold ethical standards and respect user privacy.



| anda                     | <b>Professional Elective Course (PE)</b> | MAANGGAAAG |
|--------------------------|--|------------|
| 2 <sup>nd</sup> Semester | Cyber Security and Cyber Law             | M23MCS203C |

#### 1. Prerequisites

| S/L | Proficiency            | Prerequisites   |  |
|-----|------------------------|---|--|
| 1   | <b>Computer Basics</b> | Understanding how computers work, including hardware components like CPU, memory, storage, and input/output devices   |  |
| 2   | Operating Systems      | Familiarity with popular operating systems like Windows, macOS, and Linux, including basic file management and navigation.  |  |
| 3   | Networking             | Basic concepts of how networks operate, including IP addressing, DNS, routing, and protocols like TCP/IP.   |  |
| 4   | Programming            | Basic knowledge of programming concepts can be helpful, though it's<br>not always a strict requirement. Understanding concepts like variables,<br>loops, conditionals, and functions can aid in understanding certain<br>aspects of cyber security. |  |
| 5   | Mathematics            | While not always necessary, a basic understanding of mathematics,<br>particularly concepts like binary, hexadecimal, and boolean algebra, can<br>be helpful.  |  |

## 2. Competencies

| S/L | Competency                      | KSA Description  |
|-----|---------------------------------|--|
|     | - · ·                           | Knowledge:   |
| 1   | Cyber Security:<br>Introduction | <ol> <li>Introduction to Cyber Security: Understanding the principles and<br/>goals of cyber security. Awareness of the importance of securing<br/>digital assets and information.</li> <li>Hackers: Knowledge of different types of hackers (e.g., white hat,<br/>black hat, grey hat).Understanding hacker motivations and typical<br/>methods used in attacks.</li> <li>Attackers: Awareness of various types of cyber attackers (e.g.,<br/>cyber criminals, state-sponsored attackers, hacktivists).<br/>Understanding their objectives, such as financial gain, espionage, or<br/>disruption.</li> <li>Types of Attacks: Familiarity with common cyber attacks (e.g.,<br/>phishing, ransomware, DDoS).Knowledge of attack vectors and<br/>techniques used to exploit vulnerabilities.</li> <li>Data Recovery: Understanding techniques and tools for data<br/>recovery after a cyber incident. Knowledge of backup strategies and<br/>disaster recovery plans.</li> </ol> |
|     |                                 | Skills:<br>1. Incident Detection and Response: Ability to detect anomalies and   |
|     |                                 | 1. <b>Incident Detection and Response:</b> Ability to detect anomalies and potential security incidents. Skills in incident response, including containment, eradication, and recovery.  |
|     |                                 | 2. Vulnerability Assessment and Penetration Testing (Pen   |
|     |                                 | <b>Testing):</b> Skills in identifying vulnerabilities in systems and networks. Ability to conduct ethical hacking to test defences and improve security posture.  |
|     |                                 | 3. Security Tools and Technologies: Proficiency in using security tools such as firewalls, IDS/IPS, SIEM, and antivirus software. Skills   |
|     |                                 | in configuring and managing security infrastructure.   |
|     |                                 | 4. <b>Forensic Analysis:</b> Ability to gather and analyze digital evidence for cyber crime investigations. Skills in preserving the integrity of  |
|     |                                 | evidence and presenting findings in legal contexts.  |
|     |                                 | 5. Secure Coding Practices: Ability to write secure code and adhere to coding standards to prevent vulnerabilities. Skills in identifying  |

|         | and mitigating common coding flaws (e.g., buffer overflows, SQL        |
|---------|--|
|         | injection).  |
| Attitud | 5 /  |
|         |  |
| 1.      | Ethical Responsibility: Commitment to ethical behavior and respect     |
|         | for privacy and confidentiality. Adherence to professional codes of    |
|         | conduct and legal guidelines.  |
| 2       | <b>Continuous Learning and Adaptability:</b> Willingness to stay       |
| 2.      |  |
|         | updated with evolving cyber threats and technologies. Readiness to     |
|         | adapt skills and strategies in response to new challenges.             |
| 3.      | Teamwork and Collaboration: Ability to work effectively in cross-      |
|         | functional teams to address security issues. Skills in communicating   |
|         | security concepts to non-technical stakeholders.                       |
|         |  |
| 4.      | Critical Thinking and Problem-Solving: Capacity to analyze             |
|         | complex security incidents and make informed decisions. Ability to     |
|         | anticipate potential risks and implement proactive security measures.  |
| 5       | <b>Resilience and Persistence:</b> Determination to handle and recover |
| 5.      |  |
|         | from security incidents effectively. Ability to implement preventive   |
|         | measures and maintain a robust security posture.                       |
|         |  |
|         |  |

## 3. Syllabus

| Cyt  | per Security and Cybe<br>SEMESTER II   | r Law  |
|--|--|--|
| Course Code  | M23MCS203C   | CIE Marks : 50   |
| Teaching Hours/Week (L:T:P: S)   | (2:0:0:2)  | SEE Marks: 50  |
| Total Hours of Pedagogy  | 40   | Total Marks: 100   |
| Credits  | 03   | Exam Hours: 3  |
| Teaching-Learning Process (Genera  | l Instructions)  |  |
| These are sample Strategies, which tea   |  | the attainment of the various course   |
| outcomes.  |  |  |
| • Chalk and Talk   |  |  |
| • PPT presentation   |  |  |
| Animation based videos   |  |  |
| Interactive learning   |  |  |
|  | Module 1   |  |
| Recovery. Cyber law: Features of Cyl<br>Meaning - Fundamentals of Data Secu<br>Tools and Methods Used in Cyber cri<br>Key loggers and Spywares, Virus and<br>Over Flow, Attacks on Wireless Net<br>security: The Legal Perspectives Wh | ber Law - Significance of C<br>urity - Requirements of Data<br>Module 2<br>me: Proxy Servers and And<br>Worms, Steganography, D<br>works, Phishing, Identity<br>y do we need Cyber law:<br>ct, Amendments to the Ind<br>Indian Scenario. | xers -Types of Attackers Examples –Data<br>Cyber Law - Advantages. Data Security -<br>a Security - Precautionary Measures.<br>onymizers, Phishing, Password Cracking,<br>oS DDoS Attacks, SQL Injection, Buffer<br>Theft (ID Theft)Cybercrimes and Cyber<br>The Indian Context, The Indian IT Act,<br>ian IT Act, Cybercrime and Punishment, |
|  | Module 3   |  |
| Authentication by Passwords - Prote<br>Evidence on Electronic Records, Hack  | cting Passwords - Access<br>cers & its Types - Cracking<br>File access, Recover  | ntrol: Identification - Authentication -<br>Control Structure - Evidences - Law of<br>- Pornography - Software privacy - Data<br>Internet Usage Data, Recover Swap<br>c  |
|  | Module 4   |  |
| Web Treats for Organizations: The  | Evils and Perils, Security   | and IPR Issues: Lesson for Organizations,<br>7 and Privacy Implications from Cloud<br>7 Organization, Social Computing and the   |

Associated Challenges for Organizations, Protecting People's Privacy in the Organization, Organizational Guidelines for Internet Usage, Safe Computing Guidelines and Computer Usage Policy, Incident Handling: An Essential Component, Intellectual Property rights in the Cyber security, Importance of Endpoint Security in Organizations.

#### Module 5

Concept of Cyber law and Cyber Space: Introduction - Meaning and Features of Cyber law - Significance and Advantages of Cyber Law - Meaning of Cyber Space - Inclusive of Cyber Space - Facilitating Functions of Cyber Space - Major Issues in Cyber Space. Need for an Indian Cyber law: Plans of National Information Technology Policy (NITP) - Need for Protection of data - Transactions in Security -Electronic Banking.

## **Suggested Learning Resources**

**Text Books:** 

Jonathan Rosenoer, Cyber law: The Law of Internet, Springer Verlog, Paperback, 17 September 2011
 John W Ritting House, William M.Hancock, Cyber Security Operations Handbook, Read Elsevier, 2004
 Reference Books:

1. Sunit Belapure and Nina Godbole. Cyber Security: Understanding Cyber Crimes, Computer Forensics And Legal Perspectives. Wiley India Pvt Ltd. 2013.

2. Surya PrakashTripathi, Ritendra Goyal, Praveen Kumar Shukla. Introduction to information security and cyber laws. Dreamtech Press. 2015.

3. Cybersecurity Essentials 4. Charles J. Brooks, Christopher Grow, Philip A. Craig Jr., Donald Short, ISBN: 978-1-119-36239-5 October 2018.

## Web links and Video Lectures (e-Resources):

- https://www.udemy.com/course/cybersecurity-law-policy
- https://academy.apnic.net/en/course/introduction-to-cybersecurity
- https://www.coursera.org/specializations/intro-cyber-security
- https://www.coursera.org/learn/cybersecurity-for-everyone
- <u>https://www.classcentral.com/tag/cybercrime</u>

#### 4. Syllabus Timeline

| S/L | Syllabus<br>Timeline                   | Description   |  |
|-----|--|---|--|
| 1   | Week 1-2:                              | Cyber Security Introduction   |  |
| 2   | Week 3-4:                              | Fools and Methods Used in Cyber crime   |  |
| 3   | Week 5-6: Authentication and Firewalls |   |  |
| 4   | Week 7-8:                              | k 7-8: Cyber security Organizational Implications Cost of Cybercrimes and IPR<br>Issues |  |
| 5   | Week 9-10:                             | 10: Concept of Cyber law and Cyber Space  |  |
| 6   | Week 11-<br>12:                        | Need for Protection of data - Transactions in Security - Electronic Banking.            |  |

## 5. <u>Teaching-Learning Process Strategies</u>

| S/L | TLP Strategies:                            | Description  |
|-----|--|--|
| 1   | Lecture Method                             | Utilize various teaching methods within the lecture format to reinforce competencies.                  |
| 2   | Video/Animation                            | Incorporate visual aids like videos/animations to enhance<br>understanding of cyber security concepts. |
| 3   | Collaborative Learning                     | Encourage collaborative learning for improved competency application.                                  |
| 4   | Higher Order Thinking<br>(HOTS) Questions: | Pose HOTS questions to stimulate critical thinking related to each competency.                         |
| 5   | Problem-Based<br>Learning (PBL)            | Implement PBL to enhance analytical skills and practical application of competencies                   |
| 6   | Multiple<br>Representations                | Introduce topics in various representations to reinforce competencies                                  |



| 7 | Real-World<br>Application  | Discuss practical applications to connect theoretical concepts with real-world competencies.                          |
|---|----------------------------|---|
| 8 | Flipped Class<br>Technique | Utilize a flipped class approach, providing materials before class to facilitate deeper understanding of competencies |
| 9 | Programming<br>Assignments | Assign programming tasks to reinforce practical skills associated with competencies.                                  |

## 6. Assessment Details (both CIE and SEE)

This section of regulations is applicable to all theory-based courses. The minimum CIE marks requirement is 40% of maximum marks in each component.

| Com  | ponents                       | Number | Weightage | Max.<br>Marks | Min. Marks |
|------|-------------------------------|--------|-----------|---------------|------------|
| (i)  | Internal Assessment-Tests (A) | 3      | 50%       | 25            | 10         |
| (ii) | Assignments/Quiz/Activity (B) | 2      | 50%       | 25            | 10         |
|      | Total Ma                      | rks    |           | 50            | 20         |

#### Final CIE Marks =(A) + (B)

Average internal assessment shall be the average of the best two test marks from the 3 tests conducted. **Semester End Examinations: PG Programmes** 

Theory Course with 4, 3 and 2 Credits: Professional Core Course (PC)/Professional Elective/Open Elective

- Question paper pattern will be ten questions. Each question is set for 20marks. The medium of the question paper shall be English unless otherwise it is mentioned.
- 2 There shall be 2 questions from each module, each of the two questions under a module (with a maximum of 3 sub questions), may have mix of topics under that module if necessary.
- 3The students have to answer 5 full questions selecting one full question from each module.
- Marks scored will be proportionally scaled down to 50 marks

## 7. Learning Objectives

| S/L | Learning<br>Objectives                     | Description  |
|-----|--|--|
| 1   | Foundational<br>Understanding              | <ul> <li>Define cyber security and its significance in protecting digital assets, data, and systems from cyber threats.</li> <li>Explain the principles of confidentiality, integrity, and availability (CIA) in the context of cyber security.</li> </ul>                         |
| 2   | Cyber Threat<br>Landscape                  | <ul> <li>Identify common types of cyber threats and attack vectors, such as malware, phishing, ransomware, and social engineering.</li> <li>Understand the impact of cyber threats on individuals, organizations, and society.</li> </ul>  |
| 3   | Security<br>Principles and<br>Concepts     | <ul> <li>Describe essential cyber security principles and concepts, including defense-in-depth, least privilege, and resilience.</li> <li>Explain the importance of risk management and mitigation strategies in cyber security.</li> </ul>  |
| 4   | Cybersecurity<br>Technologies<br>and Tools | <ul> <li>Explore fundamental cyber security technologies and tools used to protect networks, systems, and data.</li> <li>Discuss the role of firewalls, antivirus software, intrusion detection/prevention systems (IDS/IPS), and encryption in cyber security defence.</li> </ul> |
| 5   | Legal and<br>Ethical<br>Considerations     | <ul> <li>Discuss legal and regulatory requirements related to cyber security, including data protection laws (e.g., GDPR, CCPA).</li> <li>Understand ethical considerations in cyber security practices, including privacy rights and responsible use of technology.</li> </ul>    |

| 6 | Cyber security<br>Awareness and<br>Education | • | <ul><li>Highlight the importance of cyber security awareness among users<br/>and stakeholders.</li><li>Discuss strategies for promoting a cyber security-aware culture<br/>within organizations and communities.</li></ul> |
|---|--|---|--|
|---|--|---|--|

## 8. Course Outcomes (COs) and Mapping with POs/ PSOs

## Course Outcomes (COs)

| COs          | Description   |
|--------------|---|
| M23MCS203C.1 | Define and identify the cyber security needs of an organization.  |
| M23MCS203C.2 | Predict and analyze the software vulnerabilities and security solutions to reduce the risk of exploitation. |
| M23MCS203C.3 | Identify the cyber crime and modify security architecture for an organization.                              |
| M23MCS203C.4 | Survey operational and strategic cyber security strategies and policies                                     |
| M23MCS203C.5 | Analyze the major issues in Cyber space   |

## **CO-PO-PSO** Mapping

| COs/POs      | PO1 | PO2 | PO3 | PSO1 | PSO2 |
|--------------|-----|-----|-----|------|------|
| M23MCS203C.1 | 3   |     | 3   | 3    | 3    |
| M23MCS203C.2 | 3   |     |     | 3    | 3    |
| M23MCS203C.3 |     |     |     | 3    | 3    |
| M23MCS203C.4 |     | 3   |     | 3    | 3    |
| M23MCS203C.5 | 3   |     | 3   | 3    | 3    |
| M23MCS203C   | 3   | 3   | 3   | 3    | 3    |

## 9. Assessment Plan

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

|          | CO1 | CO2 | CO3 | CO4 | CO5 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 10  |     |     |     |     | 10    |
| Module 2 |     | 10  |     |     |     | 10    |
| Module 3 |     |     | 10  |     |     | 10    |
| Module 4 |     |     |     | 10  |     | 10    |
| Module 5 |     |     |     |     | 10  | 10    |
| Total    | 10  | 10  | 10  | 10  | 10  | 50    |

## Semester End Examination (SEE)

|          | CO1 | CO2 | CO3 | CO4 | CO5 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 20  |     |     |     |     | 20    |
| Module 2 |     | 20  |     |     |     | 20    |
| Module 3 |     |     | 20  |     |     | 20    |
| Module 4 |     |     |     | 20  |     | 20    |
| Module 5 |     |     |     |     | 20  | 20    |
| Total    | 20  | 20  | 20  | 20  | 20  | 100   |

## **Conditions for SEE Paper Setting:**

Each module of SEE question paper should be allocated with questions for 20% of the total SEE marks

## 10. Future with this Subject

The contributions of this subject extend across various areas, enhancing the students' understanding and skills in the field of digital systems. Here are some notable contributions:

- 1. **Rising Cyber Threats:** As technology advances, so do cyber threats. Future cybersecurity will need to defend against more sophisticated attacks, including AI-driven threats and attacks on interconnected IoT devices.
- 2. **Increased Regulation:** Governments worldwide are likely to introduce more stringent cyber security regulations and standards to protect citizens' data and critical infrastructure. This will necessitate a stronger focus on compliance and cyber security frameworks.
- 3. **Cyber security Automation:** With the growing volume of threats, automation through AI and machine learning will become essential for detecting and responding to attacks in real-time.
- 4. **Privacy Concerns:** As more data is collected and processed, privacy concerns will intensify. Cyber laws will need to adapt to protect individuals' rights while enabling legitimate uses of data for innovation and economic growth
- 5. **International Cooperation:** Cyber threats are often transnational, requiring international cooperation and agreements on cyber security norms, incident response, and data sharing.
- 6. **Emerging Technologies:** Technologies like quantum computing and 5G will introduce new security challenges and require innovative cyber security solutions.
- 7. **Skills Shortage:** There is and will continue to be a shortage of skilled cyber security professionals. Addressing this gap through education and training will be crucial.
- 8. Ethical and Legal Implications: As AI and automation play a bigger role in cyber security, ethical considerations around their use in decision-making and compliance with legal standards will become more prominent.
- 9. **Business Impacts:** Cyber security incidents can have significant financial and reputational impacts on businesses. Organizations will need to prioritize cyber security as a core part of their operations and risk management.

| 2 <sup>nd</sup> Semester | <b>Professional Elective Course (PE)</b> | MAANCEAAAD |
|--------------------------|--|------------|
|                          | DECISION SUPPORT SYSTEM                  | M23MCS203D |

### 1. Prerequisites

| S/L | Proficiency               | Prerequisites   |  |  |
|-----|---------------------------|---|--|--|
| 1   | Dynamic                   | Many technical systems requires anticipation and analysis of how the decision will affect.  |  |  |
| 2   | Social                    | Communication plays a vital role among the many specialized controllers.  |  |  |
| 3   | Coupled                   | Many systems result in possible unpredictable interactions and consequences.  |  |  |
| 4   | Heterogeneous             | Many contributing perspectives and experience levels.   |  |  |
| 5   | Problem-Solving<br>Skills | Develop your analytical and problem-solving skills, as designing efficiently and effective requires making trade-offs and optimizing for different scenarios. |  |  |

## 2. Competencies

| <b>2.</b> Com | ipetencies   |  |  |  |  |
|---------------|--|--|--|--|--|
| S/L           | Competency   | KSA Description  |  |  |  |
| 1             | Introduction to<br>decision support<br>systems             | <ul> <li>Knowledge: Uses of Decision support systems</li> <li>Skills: Decision Making- Rational Decision, Nature of Managers.</li> <li>Attitudes: Appropriate Decision Support, Group Decision Making, Intuition, Qualitative data and Decision Making, Business Intelligence and Decision Making</li> </ul> |  |  |  |
| 2             | Decisions in the organization                              | nowledge: Appropriate Decision Support.<br>ills: Qualitative data and Decision Making,<br>titudes: Business Intelligence and Decision Making   |  |  |  |
| 3             | Group decision<br>support and<br>groupware<br>technologies | <ul><li>Knowledge: Group Decision Making, Intuition.</li><li>Skills: knowhow of technologies.</li><li>Attitudes: Intuition to use technologies.</li></ul>  |  |  |  |
| 4             | Designing and<br>building decision<br>support systems      | <ul><li>Knowledge: Programming Reasoning.</li><li>Skills: User Interface: Goal, Mechanisms of User Interfaces, User Interface Components.</li><li>Attitudes: Uncertainty.</li></ul>  |  |  |  |
| 5             | Creative decision<br>making and<br>problem solving         | <ul><li>Knowledge: Investigate various models in Decision Support Systems.</li><li>Skills: Analyzing and decision making for creativity.</li><li>Attitudes: creativity for problem solving.</li></ul>  |  |  |  |

#### 3. Syllabus



|   | SUPPORT SYSTEM (M23MC<br>SEMESTER – II   | 5 <b>205D</b> )   |                              |
|---|--|---|------------------------------|
| Course Code   | M23MCS203D   | CIE Marks   | 50                           |
| Number of Lecture Hours/Week(L: T: P: S)  | (2:0:0:2)  | SEE Marks   | 50                           |
| Total Number of Lecture Hours   | 40 hours Theory  | Total Marks   | 100                          |
| Credits   | 03   | Exam Hours  | 03                           |
| Course objectives: <ol> <li>Explains the concept of Decision Supp</li> <li>Exhibit the decision making technique</li> <li>Investigate various models in Decision</li> <li>Apply Design and development techni</li> <li>Design an expert system by applying v</li> </ol> Introduction to decision support systems: DSS I and model management, DSS Knowledge base of DSSs, Chapter Summary. Decisions and decision | s<br>Support Systems<br>ques in DSS<br>various Knowledge Acquisition teo<br>Module -1<br>Defined, History of decision suppo<br>, User interfaces, User interfaces,<br>vision makers Decision makers: w | ort systems, Ingredients of<br>The DSS user, Categorie<br>ho are they, Decision sty | s and classe<br>les, Decisio |
| effectiveness, How can a DSS help?, A Typolog<br>Bounded decision making, The process of choic  |  |   |                              |
| Decisions in the organization: Understanding<br>Defining the problem and its structures, Decisio<br>Calibration and sensitivity.  |  |   |                              |
| Group decision support and groupware technologies, Managing MDM activities, the vir<br>exactly is an EIS, Some EIS history, Why area t<br>future of executive decision making and the EIS   | tual workspace, chapter summary<br>op executives so different?, EIS co<br>S.   | . Executive information sy  | stems: What                  |
|   | Module -4  |   |                              |
| Designing and building decision support system<br>interface issues, chapter summary. Implementin<br>evaluation, The importance of integration.  |  |   |                              |
|   | Module -5  |   |                              |
| Creative decision making and problem solving<br>Creative problem solving techniques, Creativity   |  | defined, The occurrence   | of creativity                |
| <ul> <li>Text Books: <ol> <li>Decision support system George M. M.</li> <li>"Decision Support and Business Intell Delen Publisher: Pearson</li> </ol> </li> <li>Reference Books <ol> <li>"Decision Support Systems and Intell Pearson.</li> </ol> </li> </ul>   | ligence Systems"Authors: Efraim  |   |                              |



| S/L | Syllabus<br>Timeline | Description   |
|-----|----------------------|---|
| 1   | Week 1-3             | Introduction to decision support systems: Data and model management, DSS Knowledge base,<br>User interfaces, User interfaces, The DSS user, Categories and classes of DSSs,<br>Decisions and decision makers Decision makers: Decision styles, Decision effectiveness.<br>A Typology of decisions, Decision theory and Simon's model of problem solving, Bounded<br>decision making, The process of choice, Cognitive processes, Biases and heuristics in decision<br>making. |
| 2   | Week 4-6             | Decisions in the organization: Understanding the organization and culture. Modelling decision processes: Defining the problem and its structures. Decision models, Types of probability, Techniques for forecasting probabilities, Calibration and sensitivity.   |
| 3   | Week 7-9             | Group decision support and groupware technologies: Group Decision making, the problem with groups, MDM support technologies, Managing MDM activities, the virtual workspace.<br>Executive information systems: What exactly is an EIS, Some EIS history, EIS components, Making the EIS work, The future of executive decision making and the EIS.  |
| 4   | Week 10-12           | Designing and building decision support systems: Strategies for DSS analysis and design.<br>The DSS developer, DSS user interface issues, chapter summary. Implementing and integrating decision support systems.<br>DSS implementation, System evaluation, The importance of integration.  |
| 5   | Week 13-15           | Creative decision making and problem solving. Creativity defined, The occurrence of creativity, Creative problem solving techniques, Creativity and the role of technology.   |

## 5. Teaching-Learning Process Strategies

| S/L | TLP Strategies:  | Description  |  |
|-----|--|--|--|
| 1   | Lecture Method   | Using traditional lecture methods and ICT as and when needed.                        |  |
| 2   | Video/Animation  | Incorporate visual aids like videos/animations to enhance learning.                  |  |
| 3   | Collaborative<br>Learning  | Encourage collaborative learning approaches for peer learning.                       |  |
| 4   | Problem-Based<br>Learning (PBL)  | Implement PBL to enhance analytical skills and practical application.                |  |
| 5   | Real-WorldDiscuss practical applications to connect theoretical concepts wApplicationcompetencies. |  |  |
| 6   | Programming<br>Assignments   | Assign programming tasks to reinforce practical skills associated with competencies. |  |

## 6. Assessment Details (both CIE and SEE)

This section of regulations is applicable to all theory-based courses. The minimum CIE marks requirement is 40% of maximum marks in each component.

## **CIE Split up for Professional Elective Course (PE)**

| Components  |                               | Number | Weightage | Max.<br>Marks | Min. Marks |
|-------------|-------------------------------|--------|-----------|---------------|------------|
| (i)         | Internal Assessment-Tests (A) | 3      | 50%       | 25            | 10         |
| (ii)        | Assignments/Quiz/Activity (B) | 2      | 50%       | 25            | 10         |
| Total Marks |                               |        |           |               | 20         |

## Final CIE Marks =(A) + (B)

Average internal assessment shall be the average of the best two test marks from the 3 tests conducted.

## Semester End Examinations: PG Programmes

Theory Course with 4, 3 and 2 Credits: Professional Core Course (PC)/Professional Elective/Open Elective

• Question paper pattern will be ten questions. Each question is set for 20marks. The medium of the question paper shall be English unless otherwise it is mentioned.

- 2 There shall be 2 questions from each module, each of the two questions under a module (with a maximum of 3 sub questions), may have mix of topics under that module if necessary.
- 3The students have to answer 5 full questions selecting one full question from each module.
- Marks scored will be proportionally scaled down to 50 marks

## 7. Learning Objectives

| S/L | Learning Objectives  |  |  |
|-----|--|--|--|
| 1   | Explains the concept of Decision Support Systems                             |  |  |
| 2   | Exhibit the decision-making techniques                                       |  |  |
| 3   | Investigate various models in Decision Support Systems                       |  |  |
| 4   | Apply Design and development techniques in DSS                               |  |  |
| 5   | Design an expert system by applying various Knowledge Acquisition techniques |  |  |

## 8. Course Outcomes (COs) and Mapping with POs/ PSOs

## Course Outcomes (COs):

| COs             | Description  |
|-----------------|--|
| M23MCS203D.1    | Recognize the relationship between business information needs and decision |
| WI23WIC 5203D.1 | making   |
| M23MCS203D.2    | Appraise the general nature and range of decision support systems          |
| M23MCS203D.3    | Appraise issues related to the development of DSS                          |
| M23MCS203D.4    | Select appropriate modelling techniques                                    |
| M23MCS203D.5    | Analyze, design and implement a DSS  |

## **CO-PO-PSO Mapping:**

| COs/POs      | PO<br>1 | PO<br>2 | PO<br>3 | PSO-<br>1 | PSO-2 |
|--------------|---------|---------|---------|-----------|-------|
| M23MCS203D.1 | 3       | -       | -       | 2         | 2     |
| M23MCS203D.2 | 3       | -       | -       | 2         | 2     |
| M23MCS203D.3 | -       | 3       | -       | 2         | 2     |
| M23MCS203D.4 | -       | 3       | 3       | 2         | 2     |
| M23MCS203D.5 | -       | -       | 3       | 2         | 2     |
| M23MCS203D.  | 3       | 3       | 3       | 2         | 2     |

## 9. Assessment Plan

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

|          |     | Continuous in | ter nar L'aruatio |     |     |       |
|----------|-----|---------------|-------------------|-----|-----|-------|
|          | CO1 | CO2           | CO3               | CO4 | CO5 | Total |
| Module 1 | 10  |               |                   |     |     | 10    |
| Module 2 |     | 10            |                   |     |     | 10    |
| Module 3 |     |               | 10                |     |     | 10    |
| Module 4 |     |               |                   | 10  |     | 10    |
| Module 5 |     |               |                   |     | 10  | 10    |
| Total    | 10  | 10            | 10                | 10  | 10  | 50    |

Semester End Examination (SEE)

|          | CO1 | CO2 | CO3 | CO4 | CO5 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 20  |     |     |     |     |       |
| Module 2 |     | 20  |     |     |     |       |
| Module 3 |     |     | 20  |     |     |       |
| Module 4 |     |     |     | 20  |     |       |
| Module 5 |     |     |     |     | 20  |       |
| Total    | 20  | 20  | 20  | 20  | 20  | 100   |

#### **Conditions for SEE Paper Setting:**

Each module of SEE question paper should be allocated with questions for 20% of the total SEE marks

#### 10. Future with this Subject:

- 1. Industries, including technology, finance, healthcare, marketing, and more.
- 2. Research and Development: Students interested in research can explore opportunities in academic or industrial research labs, where they can contribute to the development.
- 3. Domain-specific Applications: Students can specialize in a particular domain and apply their expertise to solve domain-specific problems.
- 4. Interdisciplinary Collaboration: This course will equip students with the ability to effectively communicate and work with domain experts, stakeholders, and cross-functional teams.



| 2 <sup>nd</sup> Semester | <b>Professional Elective Course (PE)</b> | MOOMCGOOAA |
|--------------------------|--|------------|
| 2 <sup></sup> Semester   | <b>Digital Image Processing</b>          | M23MCS204A |

## 1. Prerequisites

| S/L | Proficiency                      | Prerequisites  |  |  |
|-----|----------------------------------|--|--|--|
| 1   | Mathematics                      | <b>Linear Algebra</b> : Understanding matrices and vectors is crucial as many image processing techniques involve operations on matrices.<br><b>Calculus</b> : Especially differentiation and integration, which are used in understanding gradients, optimization, and transformations. |  |  |
| 2   | Signal Processing                | Basic knowledge of signal processing concepts is beneficial as many image processing techniques are based on similar principles.   |  |  |
| 3   | Probability and Statistics       | Knowledge of probability theory and statistics is useful for understanding noise models, image enhancement techniques, and statistical analysis in image processing  |  |  |
| 4   | Algorithm Design and<br>Analysis | Ability to design and analyse algorithms, including complexity analysis, is<br>important for implementing efficient image processing techniques  |  |  |

## 2. Competencies

| 2.  | Competencies                                     |   |
|-----|--|---|
| S/L | Competency                                       | KSA Description   |
| 1   | Understanding of<br>Image Processing<br>Concepts | <b>Knowledge:</b> Knowledge of fundamental concepts such as pixels, color models, spatial and frequency domains, image enhancement, restoration, and segmentation. <b>Skills:</b> Ability to design, implement, and optimize image processing algorithms to achieve specific objectives (e.g., enhancement, segmentation).  |
|     |  | Attitudes: Eagerness to explore new techniques, algorithms, and advancements in the field of image processing.  |
| 2   | Mathematical<br>Foundation                       | <ul> <li>Knowledge: Proficiency in mathematical concepts like linear algebra (matrices, vectors), calculus (differentiation, integration), probability theory, and statistics as they apply to image processing algorithms.</li> <li>Skill: Software Proficiency Skill in using software tools and libraries effectively to manipulate and analyze digital images.</li> <li>Attitudes: Attention to Detail Careful consideration of the nuances and details</li> </ul>  |
| 3   | Signal and Information<br>Processing             | <ul> <li>involved in image analysis and manipulation.</li> <li>Knowledge: Understanding of digital signal processing principles relevant to images, including sampling, filtering, Fourier analysis, and wavelet transforms.</li> <li>Skills: Problem-Solving Capability to analyze image processing problems, formulate solutions, and troubleshoot issues that arise during implementation</li> <li>Attitudes: Ethical Awareness Sensitivity to ethical issues related to image processing, such as privacy concerns, bias in algorithms, and proper use of digital manipulation techniques.</li> </ul> |
| 4   | Software and Tools                               | <ul> <li>Knowledge: Familiarity with programming languages (e.g., Python, MATLAB, C/C++) and libraries (e.g., OpenCV, scikit-image) used for implementing image processing algorithms.</li> <li>Skills: Experimental Design and Analysis Skill in designing experiments to evaluate the performance of image processing algorithms and interpreting results statistically.</li> <li>Attitudes: Adaptability and Flexibility Willingness to adapt to new tools, techniques, and methodologies as technologies and requirements evolve.</li> </ul>  |
| 5   | Application Domains                              | <ul> <li>Knowledge: Knowledge of specific domains where image processing is applied, such as medical imaging, satellite imagery, multimedia, and surveillance.</li> <li>Skills: Communication: Ability to communicate technical concepts and findings effectively, both orally and in writing, to diverse audiences.</li> <li>Attitudes: Collaboration: Ability to work effectively in multidisciplinary teams, leveraging diverse perspectives and skills to solve complex problems in image processing applications.</li> </ul>   |

Principal MIT Myscore

|   | Digital Image   | Processing   |   |
|---|---|--|---|
| Course Code   | M23MCS204A  | CIE Marks  | 50  |
| Teaching Hours/Week(L:P:SDA)  | (2:0:0:2)   | SEE Marks  | 50  |
| Total Hours of Pedagogy   | 40  | Total Marks  | 100   |
| Credits   | 03  | <b>Exam Hours</b>  | 03  |
| Course objectives:  |   |  |   |
| <ul> <li>To define image sampling, quan</li> <li>To explain various intensity transand different classifiers.</li> <li>To apply various basic preproce restoration and classification etc</li> </ul>  | nsformations, image segmessing (noise reduction, er   | ientation, various feature ex<br>ihancement etc.) and advan  | traction approaches,  |
| Teaching- Learning Process : Chalk ar   | nd talk method/Power Poi  | nt Presentation  |   |
|   | Module-1  |  |   |
| <b>Digital Image Fundamentals:</b> Introdu<br>Acquisition, Sampling and Quantiz<br>preliminaries, Image transforms – DFT  | ation, Relationships be<br>C, DCT.  |  |   |
|   | Module-2  |  |   |
| Image Enhancement: Spatial Domain t<br>Domain transforms: Fourier Transform<br>Filters – Gaussian Low pass and High p   | n, Smoothening and Shar   | rpening, Spatial Filtering,  | Frequency Domain  |
| Image Restoration: Image Restoration  |   | Properties, Noise models. N  | Aean Filters. Order   |
| Statistics Adaptive filters, Band reject F  |   |  |   |
| 1   | · • •   | <b>C</b> .   | C   |
|   | Module-4  |  |   |
| mage Segmentation: Point detection, I   | Edge detection, Edge linl   | king, and boundary detection   | on Hough Transform  |
|   |   | , Region splitting and me  | erging, Morphologic   |
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## 4. Syllabus Timeline



| S/L | Syllabus Timeline | Description                                     |
|-----|-------------------|---|
| 1   | Weeks 1-3:        | Module 1 Digital Image Fundamentals             |
| 2   | Weeks 4-6:        | Module 2 - Image Enhancement                    |
| 3   | Weeks 7-8:        | Module 3 - Image Restoration                    |
| 4   | Weeks 9-11:       | Module 4 - Image Segmentation                   |
| 5   | Weeks 12-14:      | Module 5 - Image Representation and Description |
| 6   | Week 15:          | Final project presentations and course wrap-up. |

#### 5. Teaching-Learning Process Strategies

| S/L | TLP Strategies:                            | Description   |
|-----|--|---|
| 1   | Interactive<br>Lectures:                   | Lectures will be designed to be interactive, with frequent opportunities for discussion, questioning, and real-time feedback.   |
| 2   | Hands-on Coding<br>Exercises:              | Regular coding exercises and assignments will be provided to reinforce practical skills in data science programming languages and libraries.  |
| 3   | Case Studies and<br>Real-World<br>Examples | The course will incorporate real-world case studies and examples from various domains to illustrate the applications of data science concepts and techniques.                         |
| 4   | Group Projects:                            | Students will work in teams to complete a comprehensive data science project, allowing them to apply their knowledge and skills to a real-world problem.                              |
| 5   | Guest Lectures:                            | Industry experts and practitioners will be invited to share their experiences and insights, providing students with a practical perspective on data science.                          |
| 6   | Flipped Classroom                          | Selected topics will be introduced through pre-recorded lectures or reading materials, allowing class time to be dedicated to discussions, exercises, and problem-solving activities. |
| 7   | Peer Learning:                             | Students will be encouraged to collaborate, share knowledge, and provide constructive feedback to their peers through study groups, discussion forums, and code reviews.              |

## 6. Assessment Details (both CIE and SEE)

This section of regulations is applicable to all theory-based courses. The minimum CIE marks requirement is 40% of maximum marks in each component.

## **CIE Split up for Professional Elective Course (PE)**

| Comp | oonents                           | Number | Weightage | Max.<br>Marks | Min. Marks |
|------|-----------------------------------|--------|-----------|---------------|------------|
| (i)  | 33. Internal Assessment-Tests (A) | 3      | 50%       | 25            | 10         |
| (ii) | 35. Assignments/Quiz/Activity (B) | 2      | 50%       | 25            | 10         |
|      | Total Marks                       |        |           | 50            | 20         |

### Final CIE Marks =(A) + (B)

Average internal assessment shall be the average of the best two test marks from the 3 tests conducted.

## **Semester End Examinations: PG Programmes**

Theory Course with 4, 3 and 2 Credits: Professional Core Course (PC)/Professional Elective/Open Elective

- Question paper pattern will be ten questions. Each question is set for 20marks. The medium of the question paper shall be English unless otherwise it is mentioned.
- 2 There shall be 2 question from each module, each of the two questions under a module (with a maximum of 3 sub questions), may have mix of topics under that module if necessary.
- 3The students have to answer 5 full questions selecting one full question from each module.

## • Marks scored will be proportionally scaled down to 50 marks

| S/L | Learning Objectives                             | Description   |  |  |  |
|-----|---|---|--|--|--|
| 1   | Understanding<br>Fundamental<br>Concepts:       | Master basic principles such as pixels, color models, spatial and frequency domains, and image representation to establish a solid foundation for further exploration and application of image processing techniques. |  |  |  |
| 2   | Mastering Image<br>Enhancement<br>Techniques    | Mastering Image Enhancement Techniques: Learn methods such as<br>noise reduction, contrast enhancement, sharpening, and color correction<br>to enhance images for improved visual quality or subsequent analysis.     |  |  |  |
| 3   | Exploring Image<br>Restoration Methods:         | Understand techniques for recovering degraded images (noise, blurring, artifacts) to improve image fidelity in medical imaging, forensics, and other applications.  |  |  |  |
| 4   | Gaining Proficiency<br>in Image<br>Compression: | Study algorithms for reducing storage/transmission size while<br>preserving image quality to efficiently manage images in multimedia,<br>internet, and storage applications.  |  |  |  |
| 5   | Image<br>Transformation                         | Understanding Image Transformation and Geometric Operations   |  |  |  |

## 7. Learning Objectives

## 8. Course Outcomes and Mapping with POs/ PSOs

| M23MCS204A.1 | Understand fundamental concepts in digital image processing such as pixels, color models, and spatial domains.                              |
|--------------|---|
| M23MCS204A.2 | Apply the principles behind various image enhancement techniques, including noise reduction and contrast enhancement.                       |
| M23MCS204A.3 | Analyze image restoration techniques to restore degraded images affected by noise, blurring, or compression artifacts.                      |
| M23MCS204A.4 | Evaluate the performance of different image segmentation algorithms and their effectiveness in partitioning images into meaningful regions. |
| M23MCS204A.5 | Design and implement a comprehensive image processing solution to address a specific real-world problem or application domain.              |

| COs/POs      | PO1 | PO2 | PO3 | PSO1 | PS02 |
|--------------|-----|-----|-----|------|------|
| M23MCS204A.1 | 3   |     | 3   | 3    | 3    |
| M23MCS204A.2 |     |     |     | 3    | 3    |
| M23MCS204A.3 | 3   |     |     | 3    | 3    |
| M23MCS204A.4 |     |     | 3   | 3    | 3    |
| M23MCS204A.5 |     |     | 3   | 3    | 3    |
| M23MCS204A   | 3   |     | 3   | 3    | 3    |

#### **CO-PO-PSO Mapping**

#### 9. Assessment Plan

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

|          | CO1 | CO2 | CO3 | CO4 | CO5 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 10  |     |     |     |     | 10    |
| Module 2 |     | 10  |     |     |     | 10    |
| Module 3 |     |     | 10  |     |     | 10    |
| Module 4 |     |     |     | 10  |     | 10    |
| Module 5 |     |     |     |     | 10  | 10    |
| Total    | 10  | 10  | 10  | 10  | 10  | 50    |



|          |     | Schiester End |     |     |     |       |
|----------|-----|---------------|-----|-----|-----|-------|
|          | CO1 | CO2           | CO3 | CO4 | CO5 | Total |
| Module 1 | 20  |               |     |     |     | 20    |
| Module 2 |     | 20            |     |     |     | 20    |
| Module 3 |     |               | 20  |     |     | 20    |
| Module 4 |     |               |     | 20  |     | 20    |
| Module 5 |     |               |     |     | 20  | 20    |
| Total    | 20  | 20            | 20  | 20  | 20  | 100   |

Semester End Examination (SEE)

## **10. Future with this Subject**

The Essentials of Data Science course provides a solid foundation for students to pursue a variety of future paths within the field of data science and beyond.

**1. Artificial Intelligence and Machine Learning:** Integration of AI and ML techniques will revolutionize DIP, enabling more sophisticated image analysis, pattern recognition, and automated decision-making. This includes applications in autonomous vehicles, medical diagnostics, and surveillance systems.

**2. Enhanced Imaging Technologies:** Advancements in imaging sensors, such as hyperspectral and multispectral imaging, will provide richer data for analysis, leading to improved accuracy in fields like remote sensing, environmental monitoring, and agriculture.

**3.** Computational Photography: Techniques like computational imaging and light field photography will redefine how images are captured and processed, enabling new capabilities in photography, augmented reality, and virtual reality.

**4. 3D Imaging and Reconstruction**: Development of robust algorithms for 3D image reconstruction from multiple viewpoints will enhance applications in medical imaging (e.g., 3D MRI), virtual reality, and archaeological documentation.

**5. Real-time and Embedded Systems:** DIP algorithms optimized for real-time processing and deployment on embedded systems (e.g., IoT devices, drones) will enable applications in real-time surveillance, smart city infrastructure, and wearable technology.

**6. Ethical Considerations and Regulation:** Growing awareness of ethical issues (e.g., privacy, bias) in DIP will drive the development of responsible practices and regulatory frameworks to ensure fair and ethical use of image processing technologies.

7. Interdisciplinary Collaborations: Collaboration across disciplines such as computer vision, robotics, neuroscience, and biology will lead to innovative solutions and applications in areas like brain imaging, prosthetics, and human-computer interaction.



| 2 <sup>nd</sup> Semester | <b>Professional Elective Course (PE)</b> | M23MCS204B   | 1 |
|--------------------------|--|--------------|---|
| 2 <sup></sup> Semester   | <b>Object Oriented Design</b>            | WIZJWICSZU4D |   |

|     | Prerequisites                                      |  |
|-----|--|--|
| S/L | Proficiency  | Prerequisites  |
| 1   | Basic Programming<br>Skills                        | Before delving into OOD, you should have a solid understanding of programming fundamentals. This includes knowledge of variables, control structures (like loops and conditionals), data types, functions/methods, and basic algorithms.                     |
| 2   | Understanding of<br>Object-Oriented<br>Programming | OOD builds upon OOP principles. Therefore, familiarity with concepts such as classes, objects, inheritance, polymorphism, encapsulation, and abstraction is essential. These concepts form the foundation upon which OOD principles are applied.             |
| 3   | Data Structures                                    | Knowing common data structures (like arrays, linked lists, stacks, queues, trees, graphs) and their basic operations is crucial. OOD often involves selecting appropriate data structures to represent real-world entities and relationships.                |
| 4   | Algorithms   | While not as deep as advanced algorithmic complexity, understanding basic algorithms and their efficiency (Big O notation) helps in designing efficient object-oriented solutions.   |
| 5   | Design Patterns                                    | Familiarity with design patterns (such as Singleton, Factory, Observer, etc.) is<br>beneficial. These patterns provide proven solutions to common design problems<br>and are widely used in OOD.   |
| 6   | UML (Unified Modeling<br>Language)                 | □ Being able to read and create UML diagrams (class diagrams, sequence diagrams, etc.) is important for visualizing and communicating your designs.  |
| 7   | Software Development<br>Lifecycle                  | Understanding the software development process, including requirements gathering, analysis, design, implementation, testing, deployment, and maintenance, gives context to OOD. It helps in applying design principles effectively throughout the lifecycle. |
| 8   | Problem-Solving Skills                             | OOD involves analyzing problems, identifying objects, defining relationships, and designing solutions. Strong problem-solving skills are fundamental to effective OOD.   |
| 9   | Experience with a Programming Language             | While not strictly required, practical experience in applying OOP concepts using a programming language (such as Java, C++, Python) helps reinforce understanding and application of OOD principles.   |
| 10  | Practice and Experience                            | Like any skill, proficiency in OOD comes with practice and hands-on<br>experience. Working on projects that involve designing and implementing<br>object-oriented solutions is invaluable  |

## 2. Competencies



Department of Computer Science & Engineering, MIT Mysore

| S/<br>L | Competency                         | KSA Description  |  |  |  |
|---------|------------------------------------|--|--|--|--|
| 1       | Understanding of<br>OOP Principles | <b>Knowledge:</b> Competence in OOD starts with a solid grasp of Object-Oriented Programming (OOP) principles such as encapsulation, inheritance, polymorphism, and abstraction. This knowledge forms the theoretical foundation upon which OOD is built. <b>Skills:</b> Proficiency in applying OOD principles to design software solutions that are modular, maintainable, and scalable. <b>Attitudes:</b> Approach problems with a critical mindset, evaluating different design options and selecting the most suitable based on trade-offs. |  |  |  |
| 2       | Design Patterns                    | <ul> <li>Knowledge: Knowledge of common design patterns (Singleton, Factory, Observer, etc.) and their appropriate use in different scenarios demonstrates deeper understanding of OOD concepts.</li> <li>Skill: Ability to design complex systems by identifying objects, defining their relationships, and organizing them into coherent class hierarchies.</li> <li>Attitudes: Paying attention to details in design, ensuring consistency, clarity, and adherence to best practices.</li> </ul>  |  |  |  |
| 3       | UML Proficiency                    | <ul> <li>Knowledge: Understanding and ability to use UML diagrams (class diagrams, sequence diagrams, etc.) effectively to visualize and communicate OOD designs.</li> <li>Skills: Skill in refactoring existing code to improve its structure, applying principles like SOLID (Single Responsibility, Open/Closed, Liskov Substitution, Interface Segregation, Dependency Inversion).</li> <li>Attitudes: Being open to learning new design patterns, methodologies, and technologies to improve OOD skills continuously.</li> </ul>            |  |  |  |
| 4       | Data Structures<br>and Algorithms  | <ul> <li>Knowledge: Knowledge of data structures and algorithms relevant to OOD, including their efficiency and suitability for different design scenarios.</li> <li>Skills: Ability to translate OOD designs into working code using a programming language, ensuring adherence to design principles and patterns.</li> <li>Attitudes: Positive attitude towards solving complex design problems creatively and systematically.</li> </ul>  |  |  |  |

## 3. Syllabus

|                              | Object Oriented Design<br>SEMESTER-II |             |    |
|------------------------------|---------------------------------------|-------------|----|
| Course Code                  | M23MCS204B                            | CIE Marks   | 50 |
| Teaching Hours/Week(L:P:SDA) | (2:0:0:2)                             | SEE Marks   | 50 |
| Total Hours of Pedagogy      | 40                                    | Total Marks | 10 |
|                              |                                       |             | 0  |
| Credits                      | 03                                    | Exam Hours  | 03 |

**Course objectives:** 

• To Introduce various designing techniques and methods for object oriented.

- Performance analysis with real time system.
- Demonstrate a familiarity with object oriented data and system.
- To give clear idea on implementing design with UML diagram like state diagram , activity diagram , use case diagram etc.

## Teaching- Learning Process : Chalk and talk/PPT/case study/web content

Module-1

The Motivation for Object-Oriented Programming, Classes and Objects: The Building Blocks of the Object-Oriented Paradigm Topologies of Action-Oriented Versus Object-Oriented Applications: Accidental versus Essential Complexity, Waterfall Model, Iterative Model, Different language Prototyping, Software Reusability. Introduction to Classes and Objects, Messages and Methods, Class Coupling and cohesion, Dynamic Semantics, Abstract Classes Differences in Application Topologies, God Class Problem, Role of Agent Classes.

Module-2

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The Relationships Between Classes and Objects The Inheritance Relationship: Introduction to Class and Objects Relationships, User Relationships, Six Different ways to implement User Relationships, Heuristics for User Relationships, Containment Relationships, Attributes versus Contained Classes. Introduction to the Inheritance Relationship, Overriding Base Class Methods in Derived Classes, The Use of the Protected Section of a Base Class, The Width and Depth of Inheritance Hierarchies, A Real-World Example of Specialization.

#### Module-3

Multiple Inheritance, TheAssociationRelationship: Introduction to Multiple Inheritance, The Common Misuse of Multiple Inheritance, A Valid Use of Multiple Inheritance, Accidental Complexity In Languages That Do Not Support Multiple Inheritance, Frameworks That Incorporate Multiple Inheritance, The Use of Multiple Inheritance in the Design of Mixins, DAG Multiple Inheritance, Introduction to Association Associations Implemented Through a Referential Attribute Association Implemented Through a Third-Party Class v

#### Module-4

Class-Specific Data and Behavior, Physical Object-Oriented Design: Introduction to Class-Specific Versus Object-Specific Data and Behavior Using Meta classes to Capture Class-Specific Data and Behavior, Using Language-Level Keywords to Implement ClassVersus Object-Specific Data and Behavior, Meta classes à la C++, A Useful Abstract Class That Is Not a Base Class?, The Role of Logical and Physical Object-Oriented Design, The Construction of OO Wrappers, Persistence in an OO System, Memory Management Issues, Minimal Public Interfaces for Reusable Components, Implementing Safe Shallow Copies, Concurrent OO Programming, Implementing OO Designs in Non-OO Languages.

#### Module-5

TheRelationshipBetweenHeuristicsandPatterns,TheUseofHeuristicsin Object-Oriented Design: Heuristics Versus Patterns, Transitivity Among Design Transformation Patterns, The Reflexive Property of Design Transformation Patterns, Other Design Transformation Patterns, The ATM Problem, Choosing a Methodology, A First Attempt at Producing an Object Model for the ATM, Adding Behavior Object Model, Explicit Case Analysis Due to Accidental Complexity, Messaging Objects, The Processing of the Transaction, Returning to the Domain of the ATM, Other Miscellaneous Issues.

## Suggested Learning Resources:

Text Books

- Object Oriented Design Heuristic. Arthur J Riel. Addison-Wesley. 1996.
- Object Oriented Modelling and Design With UM. Paperback, Michael R. Blaha. Pearson. 2007

## **Reference Books**

- Elements of Reusable Object Oriented Software. Ralph Johnson, Erich Gamma, Richard Helm, John Vlissides. Pearson.
- Object Oriented Modelling and Design With UM. Paperback, Michael R. Blaha. Pearson. 2007 **Online Resource**
- https://www.youtube.com/watch?v=WpJ\_yiwbGyk&list=PLJ5C\_6qdAvBHslIkD7JB7kBdgv1SeXy3P
- https://www.geeksforgeeks.org/oops-object-oriented-design/

## 4. Syllabus Timeline

The course will be delivered over a 15-week semester, with the following tentative schedule:

| S/L | Syllabus Timeline | Description   |  |  |  |  |
|-----|-------------------|---|--|--|--|--|
| 1   | Weeks 1-3:        | Module 1 The Motivation for Object-Oriented Programming, Classes andObjects: The Building Blocks of the Object-Oriented ParadigmTopologies of Action-Oriented Versus Object-Oriented Applications |  |  |  |  |
| 2   | Weeks 4-6:        | Module 2 The Relationships Between Classes and Objects The<br>Inheritance Relationship  |  |  |  |  |
| 3   | Weeks 7-8:        | Module 3 Multiple Inheritance, The Association Relationship   |  |  |  |  |
| 4   | Weeks 9-11        | Module 4 Class-Specific Data and Behavior, Physical Object-Oriented Design  |  |  |  |  |
| 5   | Weeks 12-14:      | Module 5 The Relationship Between Heuristics and Patterns, The Use of<br>Heuristics in Object-Oriented Design   |  |  |  |  |
| 6   | Week 15:          | Case Study presentations and course wrap-up.  |  |  |  |  |



## 5. Teaching-Learning Process Strategies

To facilitate an engaging and effective learning experience, the following strategies will be employed.

| S/L | TLP Strategies:                            | Description   |
|-----|--|---|
| 1   | Interactive<br>Lectures:                   | Lectures will be designed to be interactive, with frequent opportunities for discussion, questioning, and real-time feedback.   |
| 2   | Hands-on Coding<br>Exercises:              | Regular coding exercises and assignments will be provided to reinforce practical skills in data science programming languages and libraries.  |
| 3   | Case Studies and<br>Real-World<br>Examples | The course will incorporate real-world case studies and examples from various domains to illustrate the applications of data science concepts and techniques.                         |
| 4   | Group Projects:                            | Students will work in teams to complete a comprehensive data science project, allowing them to apply their knowledge and skills to a real-world problem.                              |
| 5   | Guest Lectures:                            | Industry experts and practitioners will be invited to share their experiences and insights, providing students with a practical perspective on data science.                          |
| 6   | Flipped Classroom                          | Selected topics will be introduced through pre-recorded lectures or reading materials, allowing class time to be dedicated to discussions, exercises, and problem-solving activities. |
| 7   | Peer Learning:                             | Students will be encouraged to collaborate, share knowledge, and provide constructive feedback to their peers through study groups, discussion forums, and code reviews.              |

## 6. Assessment Details (both CIE and SEE)

This section of regulations is applicable to all theory-based courses. The minimum CIE marks requirement is 40% of maximum marks in each component.

## **CIE Split up for Professional Elective Course (PE)**

| Comp | onents                            | Number | Weightage | Max.<br>Marks | Min. Marks |
|------|-----------------------------------|--------|-----------|---------------|------------|
| (i)  | 43. Internal Assessment-Tests (A) | 3      | 50%       | 25            | 10         |
| (ii) | 45. Assignments/Quiz/Activity (B) | 2      | 50%       | 25            | 10         |
|      | TotalMarks                        |        |           | 50            | 20         |

#### FinalCIE Marks =(A) + (B)

Average internal assessment shall be the average of the best two test marks from the 3 tests conducted.

#### **Semester End Examinations: PG Programmes**

Theory Course with 4, 3 and 2 Credits: Professional Core Course (PC)/Professional Elective/Open Elective

- Question paper pattern will be ten questions. Each question is set for 20marks. The medium of the question paper shall be English unless otherwise it is mentioned.
- 2 There shall be 2 questions from each module, each of the two questions under a module (with a maximum of 3 sub questions), may have mix of topics under that module if necessary.
- 3The students have to answer 5 full questions selecting one full question from each module.
- Marks scored will be proportionally scaled down to 50 marks

| S/L | Learning Objectives | Description  |
|-----|---------------------|--|
| 1   | Design techniques   | To Introduce various designing techniques and methods for object oriented. |
| 2   | Analysis            | Performance analysis with real time system.                                |

## 7. Learning Objectives

| 3 | Object Oriented data | Demonstrate a familiarity with object oriented data and system.  |
|---|----------------------|--|
| 4 | UML Diagram          | To give clear idea on implementing design with UML diagram like state diagram, activity diagram, use case diagram etc. |

## 8. Course Outcomes and Mapping with POs/ PSOs

## Upon successful completion of this course, students should be able to

| M23MCS204B.1 | Apply principles and guidelines derived from experience to effectively design and implement Object-Oriented solutions. |
|--------------|--|
| M23MCS204B.2 | Analyze how OOP principles such as encapsulation, inheritance, and polymorphism contribute to software design.         |
| M23MCS204B.3 | Analyze intricate relationships between objects, such as associations, aggregations, and compositions                  |
| M23MCS204B.4 | Analyze how Physical Object-Oriented Design bridges software architecture with implementation details.                 |
| M23MCS204B.5 | Analyze the impact of heuristic choices on software quality, maintainability, and extensibility.                       |

| CO-r O-r SO Mapping |     |     |     |      |      |
|---------------------|-----|-----|-----|------|------|
| COs/POs             | PO1 | PO2 | PO3 | PS01 | PSO2 |
| M23MCS204B.1        | 3   |     |     | 3    | 3    |
| M23MCS204B.2        | 3   |     |     | 3    | 3    |
| M23MCS204B.3        |     | 3   |     | 3    | 3    |
| M23MCS204B.4        |     | 3   |     | 3    | 3    |
| M23MCS204B.5        |     |     | 3   | 3    | 3    |
| M23MCS204B          | 3   | 3   | 3   | 3    | 3    |

## **CO-PO-PSO Mapping**

## 9. Assessment Plan

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

|          | CO1 | CO2 | CO3 | CO4 | CO5 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 10  |     |     |     |     | 10    |
| Module 2 |     | 10  |     |     |     | 10    |
| Module 3 |     |     | 10  |     |     | 10    |
| Module 4 |     |     |     | 10  |     | 10    |
| Module 5 |     |     |     |     | 10  | 10    |
| Total    | 10  | 10  | 10  | 10  | 10  | 50    |

#### Semester End Examination (SEE)

|          | CO1 | CO2 | CO3 | CO4 | CO5 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 20  |     |     |     |     | 20    |
| Module 2 |     | 20  |     |     |     | 20    |
| Module 3 |     |     | 20  |     |     | 20    |
| Module 4 |     |     |     | 20  |     | 20    |
| Module 5 |     |     |     |     | 20  | 20    |
| Total    | 20  | 20  | 20  | 20  | 20  | 100   |

## 10. Future with this Subject

The Essentials of Object Oriented Design course provides a solid foundation for students to pursue a variety of future paths and beyond.

- 1. Foundation for Advanced Concepts: Object-Oriented Design serves as a foundation for learning more advanced concepts and technologies in software development. It prepares you for exploring topics like design patterns, architectural patterns, enterprise application development, and more sophisticated programming languages.
- Adaptability and Reusability: OOD promotes code reusability and modularity, which are crucial for developing large-scale applications and systems. Understanding OOD enables you to adapt to different programming languages and frameworks more easily because OOP concepts are widely applicable across various technologies.
- 3. **Collaborative Development:** Knowing OOD facilitates effective collaboration with other developers. It provides a common vocabulary and set of practices that streamline communication and teamwork. Teams that follow OOD principles tend to produce cleaner codebases that are easier to understand, maintain, and extend over time.
- 4. Continuous Learning and Improvement: Object-Oriented Design encourages a mindset of continuous learning and improvement. As you gain experience, you can refine your design skills and stay updated with evolving best practices. It fosters a proactive approach to software development, where you constantly seek better ways to design and implement solutions.
- 5. Contribution to Open Source and Community Projects: Proficiency in OOD allows you to contribute effectively to open-source projects and community initiatives. You can design reusable components or refactor existing codebases to improve functionality and maintainability.



| 2 <sup>nd</sup> Semester | <b>Professional Elective Course (PE)</b> | M23MCS204C   |
|--------------------------|--|--------------|
| 2 "Semester"             | Multimedia Communication                 | W125W1C5204C |

## 1. Prerequisites

| S/L | Proficiency                                | Prerequisites  |
|-----|--|--|
| 1   | Basic<br>Computer<br>Literacy              | A solid understanding of how computers work, file management, and using software applications is essential.  |
| 2   | Communication<br>Skills                    | Strong written and verbal communication skills are often required, as<br>multimedia communication involves conveying messages through various<br>media forms.                  |
| 3   | Understanding<br>of Multimedia<br>Concepts | Familiarity with basic multimedia concepts such as digital media formats (image, audio, video), compression techniques, and multimedia design principles.                      |
| 4   | Audio and<br>Video Basics                  | Understanding of fundamental concepts related to audio and video production, editing, and formats.   |
| 5   | Web<br>Technologies                        | Basic knowledge of web technologies such as HTML, CSS, and possibly<br>scripting languages like JavaScript, as multimedia often intersects with web<br>design and development. |
| 6.  | Critical<br>Thinking and<br>Creativity     | Courses often require students to think critically about multimedia<br>messages and to creatively apply multimedia techniques to solve<br>communication challenges.            |

## 2. Competencies

| S/L | Competency                 | KSA Description   |
|-----|----------------------------|---|
| 1   | Multimedia                 | <ul><li>Knowledge: Insight into how multimedia is used to communicate messages effectively across various digital platforms.</li><li>Skills: Proficiency in using multimedia tools.</li><li>Attitudes: Ability to critically evaluate multimedia content for its effectiveness and impact.</li></ul>  |
| 2   | Text and image compression | <ul> <li>Knowledge: Understanding the trade-offs between file size reduction and fidelity of the compressed content.</li> <li>Skills: Competence in managing compressed data efficiently for storage, transmission, and retrieval purposes.</li> <li>Attitudes: Exploration of new compression techniques and advancements in the field.</li> </ul> |
| 3   | Audio<br>compression       | <b>Knowledge:</b> Understanding of various audio compression methods.<br><b>Skills:</b> Applying audio compression algorithms and managing compressed audio files for storage, transmission, and playback.<br><b>Attitudes:</b> Selecting appropriate compression settings to balance file size reduction and audio quality.                        |
| 4   | Video<br>compression       | <ul> <li>Knowledge: Understanding of various audio compression methods.</li> <li>Skills: Applying audio compression algorithms and managing compressed audio files for storage, transmission, and playback.</li> <li>Attitudes: selecting optimal compression settings to balance file size reduction and video quality.</li> </ul>                 |
| 5   | Synchronization            | <ul><li>Knowledge: Knowledge of techniques for synchronizing media.</li><li>Skills: Synchronizing various media elements using software tools and techniques.</li><li>Attitudes: Designing and delivering engaging multimedia content.</li></ul>  |

## 3. Syllabus

|               | Multimedia Communication |  |
|---------------|--------------------------|--|
| SEMESTER – II | SEMESTER – II            |  |



| Course  | e Code   |  | M23MCS204C  | CIE Marks   | 50              |  |
|---|--|--|---|---|-----------------|--|
| Numb  | er of Lecture Hours/   | Week(L: T: P: S)   | (2:0:0:2)   | SEE Marks   | 50              |  |
|   | Number of Lecture H  |  | 40 hours  | Total Marks   | 100             |  |
| Credit  | S  |  | 03  | Exam Hours  | 03              |  |
| •<br>•<br>•<br>•<br>•<br>•<br>•<br>•<br>•<br>•<br>•<br>•<br>•<br>•<br>•<br>•<br>•<br>•<br>•   | Understand the va<br>Understand the<br>compression stand<br>Demonstrate learn<br>work.<br>media Communicat<br>uction, multimedia | rious multimedia s<br>compression tech<br>lards for Text, Ima<br>ner autonomy by n<br>ion<br>information represent<br>g terminology, net<br>.6 | niques, transform fo<br>age and Video analyze<br>maximizing use of lea<br>Module -1<br>esentation, multimedia | r compression and a<br>rning resources and presented by the second presented presented presented presente | oducing qualit  |  |
| Text and image compression, compression principles, text compression- Run length, Huffman, LZW, Document Image compression using T2 and T3 coding, image compression- GIF, TIFF and JPEG Text-1: CH-3.1-3.5 |  |  |   |   |                 |  |
|   |  |  | Module -3   |   |                 |  |
| Video<br>Video<br>MPEC  | 6 7 standardization p  |  |   | PEG 2, MPEG-4 and Re<br>, MPEG 21 multimedia  |                 |  |
| Text-1  | l: CH-4.3.2-4.3.7  |  | Module -5   |   |                 |  |
| Notion<br>SMIL,<br>Text-2   | , Multimedia operation<br>2 : CH-9-10  | 1 1  | rements, reference mo   | 8<br>del for synchronization,<br>process management teo   |                 |  |
| Text E  | Multimedia Comr  | puting, Commun   | Halsall. Pearson educa ications and Applicat  | tion. 2001.<br>ions. Raif Steinmetz,  | KlaraNahrsted   |  |
| Syll  | abus Timeline  |  |   |   |                 |  |
| S/L   | Syllabus<br>Timeline   |  | Descri  |   |                 |  |
| 1   | Week 1-3<br>Multimedia<br>Communication  | multimedia com   | munication services. S  | network that are use<br>election of the applica<br>that are associated with   | tion that these |  |
| 2   | Week 4-6<br>Text and Image   | Describes a selection of the   |   | ion algorithms that are   | e used for the  |  |



| 3 | Week 7-9<br>Audio and video<br>compression | Describes a selection of algorithms related to audio and video.  |
|---|--|--|
| 4 | Week 10-12<br>Video<br>compression         | Define the standards that are used for compression of the different media types  |
| 5 | Week 13-15<br>Synchronization              | Describes design, develop, and manage multimedia systems effectively,<br>ensuring seamless synchronization, efficient resource utilization, and optimal<br>performance in multimedia applications. |

## 5. Teaching-Learning Process Strategies

| S/L | TLP Strategies:                 | Description  |  |  |
|-----|---------------------------------|--|--|--|
| 1   | Lecture Method                  | Using traditional lecture methods and ICT as and when needed.                                |  |  |
| 2   | Video/Animation                 | Incorporate visual aids like videos/animations to enhance learning.                          |  |  |
| 3   | Collaborative<br>Learning       | Encourage collaborative learning approaches for peer learning.                               |  |  |
| 4   | Problem-Based<br>Learning (PBL) | Implement PBL to enhance analytical skills and practical application.                        |  |  |
| 5   | Real-World<br>Application       | Discuss practical applications to connect theoretical concepts with real-world competencies. |  |  |
| 6   | Programming<br>Assignments      | Assign programming tasks to reinforce practical skills associated with competencies.         |  |  |

## 6. Assessment Details (both CIE and SEE)

This section of regulations is applicable to all theory-based courses. The minimum CIE marks requirement is 40% of maximum marks in each component.

| CIE Split up for Professional Elective Course (PF | E) |
|---|----|
|---|----|

|                                   | Components                    | Number | Weightage | Max.<br>Marks | Min.<br>Marks |
|-----------------------------------|-------------------------------|--------|-----------|---------------|---------------|
| (i) Internal Assessment-Tests (A) |                               | 3      | 50%       | 25            | 10            |
| (ii)                              | Assignments/Quiz/Activity (B) | 2      | 50%       | 25            | 10            |
|                                   | Total Marks                   |        |           |               | 20            |

## Final CIE Marks =(A) + (B)

Average internal assessment shall be the average of the best two test marks from the 3 tests conducted. **Semester End Examinations: PG Programmes** 

Theory Course with 4, 3 and 2 Credits: Professional Core Course (PC)/Professional Elective/Open Elective

- Question paper pattern will be ten questions. Each question is set for 20marks. The medium of the question paper shall be English unless otherwise it is mentioned.
- There shall be 2 question from each module, each of the two questions under a module (with a maximum of 3 sub questions), may have mix of topics under that module if necessary.
- The students have to answer 5 full questions selecting one full question from each module.
- Marks scored will be proportionally scaled down to 50 marks

## 7. Learning Objectives

| S/L | Learning Objectives   |  |
|-----|---|--|
| 1   | 1 Learn the fundamental of Multimedia systems and types of media system   |  |
| 2   | 2 Learn and analyze compression techniques                                |  |
| 3   | 3 Understand and evaluate different protocols of multimedia communication |  |
| 4   | 4 Understand Multimedia Content Management and Retrieval techniques       |  |
| 5   | Adapt to Emerging Technologies and Trends in Multimedia                   |  |

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|  | Course Outcomes (COs): |  |  |  |
|--|------------------------|--|--|--|
| COs Description  |                        |  |  |  |
| M23MCS204C.1 Apply QoS to multimedia network applications with efficient routing techniques. |                        |  |  |  |
| M23MCS204C.2 Analyze the security threats in the multimedia networks.                        |                        |  |  |  |
| M23MCS204C.3 Design and develop multimedia communication models and real-time multimed       |                        |  |  |  |
| network applications.  |                        |  |  |  |
| M23MCS204C.4 Understand Multimedia Content Management and Retrieval techniques               |                        |  |  |  |
| M23MCS204C.5 Adapt to Emerging Technologies and Trends in Multimedia                         |                        |  |  |  |

# 8. Course Outcomes (COs) and Mapping with POs/ PSOs

| CO-PO-PSO Mapping: |     |     |     |      |      |
|--------------------|-----|-----|-----|------|------|
| COs/POs            | PO1 | PO2 | PO3 | PSO1 | PSO2 |
| M23MCS204C.1       | 3   | -   | -   | 3    | 3    |
| M23MCS204C.2       | 3   | -   | 3   | 3    | 3    |
| M23MCS204C.3       | -   | 3   | 3   | 3    | 3    |
| M23MCS204C.4       | 3   | 3   |     | 3    | 3    |
| M23MCS204C.5       | 3   |     |     | 3    | 3    |
| M23MCS204C         | 3   | 3   | 3   | 3    | 3    |

## 9. Assessment Plan

**Continuous Internal Evaluation (CIE)** 

|          | CO1 | CO2 | CO2 | CO3 | CO3 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 10  |     |     |     |     | 10    |
| Module 2 |     | 10  |     |     |     | 10    |
| Module 3 |     |     | 10  |     |     | 10    |
| Module 4 |     |     |     | 10  |     | 10    |
| Module 5 |     |     |     |     | 10  | 10    |
| Total    | 10  | 10  | 10  | 10  | 10  | 50    |

#### CO1 Total CO2 CO<sub>2</sub> CO3 CO3 Module 1 20 20 Module 2 20 20 Module 3 20 20 Module 4 20 20 Module 5 20 20 Total 20 20 20 20 20 100

## Semester End Examination (SEE)

## **Conditions for SEE Paper Setting:**

Each module of SEE question paper should be allocated with questions for 20% of the total SEE marks

## 10. Future with this Subject:

- Digital **Marketing and Advertising**: Multimedia communication skills are highly valued in digital marketing and advertising agencies. Professionals in this field create engaging multimedia content for social media campaigns, online advertisements, and interactive marketing strategies.
- Content **Creation and Production**: There is a growing demand for multimedia content creators across platforms such as YouTube, TikTok, and other social media channels. Careers in content creation involve producing engaging videos, graphics, and interactive content that resonate with audiences.

User Experience (UX) Design: Multimedia communication skills are essential in UX design, where professionals create intuitive and visually appealing interfaces for websites, apps, and interactive media platforms. UX designers use multimedia to enhance user interaction and engagement.
 Virtual Reality (VR) and Augmented Reality (AR): As VR and AR technologies continue to evolve, there is an increasing demand for multimedia specialists who can create immersive and interactive experiences.

Department of Computer Science & Engineering, MIT Mysore



| 2 <sup>nd</sup> Semester | <b>Professional Elective Course (PE)</b> | M23MCS204D   |
|--------------------------|--|--------------|
| 2 <sup></sup> Semester   | NoSOL Database                           | W125W1C5204D |

### 1. Prerequisites

| S/L | Proficiency                              | Prerequisites   |  |  |
|-----|--|---|--|--|
| 1   | Basic Computer<br>Science Knowledge      | Understanding of data structures (arrays, lists, dictionaries, trees, graphs).<br>Basic algorithm concepts (searching, sorting).  |  |  |
| 2   | Database<br>Fundamentals                 | Knowledge of traditional relational databases (SQL, tables, schemas).Understanding of database operations (CRUD operations: Create, Read, Update, Delete)                                 |  |  |
| 3   | Programming Skills                       | Proficiency in at least one programming language (e.g., Python, JavaScript, Java).Familiarity with database interaction using a programming language (e.g., using SQL libraries or ORMs). |  |  |
| 4   | Understanding of<br>Data Models          | Concept of structured vs. unstructured data.<br>Knowledge of different data models (document, key-value, column-family, graph).   |  |  |
| 5   | Networking and<br>Distributed Systems    | Basic understanding of networking concepts.<br>Familiarity with distributed systems principles (consistency, availability, partition tolerance).  |  |  |
| 6   | Web Development<br>Basics                | Understanding of web technologies (HTTP, REST APIs).<br>Experience with front-end and back-end development.   |  |  |
| 7   | Operating Systems<br>and Shell Scripting | Basic knowledge of operating systems (especially Unix/Linux).<br>Proficiency in shell scripting for automating tasks.   |  |  |
| 8   | JSON and XML                             | Understanding of data interchange formats like JSON and XML, which are commonly used in NoSQL databases.  |  |  |
| 9   | NoSQL Database<br>Concepts               | Familiarity with the CAP theorem.<br>Understanding the types of NoSQL databases (document-oriented, key-value<br>stores, column-family stores, graph databases).                          |  |  |
| 10  | Practical<br>Experience                  | Hands-on experience with at least one NoSQL database (e.g., MongoDB, Cassandra, Redis, Neo4j).  |  |  |

## 2. Competencies

| S/L  | Competency            | KSA Description   |
|--|-----------------------|---|
| Image: 1Foundational<br>Knowledge(document-based, key-value, column-family, graph) and their respective<br>strengths and weaknesses.Skills: Ability to design efficient schemas and data models tailored for<br>NoSQL databases. |                       | <ul><li>Skills: Ability to design efficient schemas and data models tailored for NoSQL databases.</li><li>Attitude: Willingness to experiment with different NoSQL solutions to find</li></ul>  |
| 2  | Technical Skills      | <ul> <li>Knowledge: Knowledge of various types of NoSQL databases (document-based, key-value, column-family, graph) and their differences in data storage and retrieval.</li> <li>Skills: Capability to design effective data models that optimize data retrieval and storage efficiency in NoSQL databases.</li> <li>Attitude: Proactive approach to identifying and solving complex data management and performance challenges in NoSQL databases.</li> </ul> |
| 3  | Programming<br>Skills | <ul> <li>Knowledge: Understanding how to optimize data models for query performance and scalability.</li> <li>Skills: Ability to perform data aggregations and transformations using NoSQL query languages or APIs.</li> <li>Attitude: Eagerness to stay updated with advancements in NoSQL technologies, best practices, and industry trends.</li> </ul>   |

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| 4 | Performance<br>Tuning | <ul> <li>Knowledge: Knowledge of the query execution process within the NoSQL database, including how indexes are utilized, data retrieval strategies, and query planning.</li> <li>Skills: Ability to interpret performance metrics, identify performance issues, and take corrective actions proactively.</li> <li>Attitude: Commitment to implementing thorough performance tuning practices, ensuring that optimizations are carefully tested and validated.</li> </ul> |
|---|-----------------------|---|
| 5 | Data<br>Management    | <b>Knowledge:</b> Familiarity with query languages (e.g., MongoDB Query Language, Cassandra Query Language) and APIs used for data retrieval, manipulation, and aggregation in NoSQL databases.<br><b>Skills:</b> Skills in writing efficient queries to ensure optimal performance.<br><b>Attitude:</b> Careful consideration of data modelling and schema design to ensure optimal performance and scalability.   |

#### 3. Syllabus

| NoSQL DATABASE  |                 |             |     |  |
|---|-----------------|-------------|-----|--|
| Course Code   | M23MCS204D      | CIE Marks   | 50  |  |
| Number of Lecture Hours/Week(L: T: P: S)  | (2:0:0:2)       | SEE Marks   | 50  |  |
| Total Number of Lecture Hours   | 40 hours Theory | Total Marks | 100 |  |
| Credits   | 03              | Exam Hours  | 03  |  |
| CO1: Define, compare and use the four types of NoSQL Databases (Document-oriented, KeyValue Pairs,        |                 |             |     |  |
| Column-oriented and Graph).   |                 |             |     |  |
| CO2: Demonstrate an understanding of the detailed architecture, define objects, load data, query data and |                 |             |     |  |
| performance tune Column-oriented NoSQL da   | atabases.       |             |     |  |

**CO3:** Explain the detailed architecture, define objects, load data, query data and performance tune Documentoriented NoSQL databases.

#### Module -1

**Why NoSQL:** Why NoSQL? The Value of Relational Databases, Getting at Persistent Data, Concurrency, Integration, A (Mostly) Standard Model, Impedance Mismatch, Application and Integration Databases, Attack of the Clusters, The Emergence of NoSQL, Aggregate Data Models; Aggregates, Example of Relations and Aggregates, Consequences of Aggregate Orientation, Key-Value and Document Data Models, Column-Family Stores, Summarizing Aggregate-Oriented Databases. More Details on Data Models; Relationships, Graph Databases, Schemaless Databases, Materialized Views, Modeling for Data Access.

#### Module -2

**Distribution Models:** Distribution Models; Single Server, Sharding, Master-Slave Replication, Peer-to-Peer Replication, Combining Sharding and Replication. Consistency, Update Consistency, Read Consistency, Relaxing Consistency, The CAP Theorem, Relaxing Durability, Quorums. Version Stamps, Business and System Transactions, Version Stamps on Multiple Nodes.

## Module -3

**Map-Reduce:** Map-Reduce, Basic Map-Reduce, Partitioning and Combining, Composing Map-Reduce Calculations, A Two Stage Map-Reduce Example, Incremental Map-Reduce Key-Value Databases, What Is a Key-Value Store, Key-Value Store Features, Consistency, Transactions, Query Features, Structure of Data, Scaling, Suitable Use Cases, Storing Session Information, User Profiles, Preference, Shopping Cart Data, When Not to Use, Relationships among Data, Multioperation Transactions, Query by Data, Operations by Sets.

#### Module -4

**Document Databases:** Document Databases, What Is a Document Database, Features, Consistency, Transactions, Availability, Query Features, Scaling, Suitable Use Cases, Event Logging, Content Management Systems, Blogging Platforms, Web Analytics or Real-Time Analytics, ECommerce Applications, When Not to Use, Complex Transactions Spanning Different Operations, Queries against Varying Aggregate Structure

Module -5

**Graph Databases:** Graph Databases, What Is a Graph Database?, Features, Consistency, Transactions, Availability, Query Features, Scaling, Suitable Use Cases, Connected Data, Routing, Dispatch, and Location-Based Services, Recommendation Engines, When Not to Use. **Textbooks:** 

- 1. Sadalage, P. & Fowler, NoSQL Distilled: A Brief Guide to the Emerging World of Polyglot Persistence, Pearson Addision Wesley, 2012
- 2. Kristina Chodorow, "Mongodb: The Definitive Guide- Powerful and Scalable Data Storage", 2nd Edition, O'Reilly Publications, 2013. (ISBN-13: 978-9351102694)

#### **Reference Books:**

1. Dan Sullivan, "NoSQL For Mere Mortals", 1st Edition, Pearson Education India, 2015. (ISBN13: 978-9332557338)

2. Dan McCreary and Ann Kelly, "Making Sense of NoSQL: A guide for Managers and the Rest of us", 1st Edition, Manning Publication/Dreamtech Press, 2013. (ISBN-13: 978-9351192022)

## 4. Syllabus Timeline

| S/L | Syllabus Timeline | Description  |
|-----|-------------------|--|
| 1   | Week 1-2          | Introduction and Principles of Combinational NoSQL |
| 2   | Week 3-4          | Map-Reduce concept introduction to SQL             |
| 3   | Week 5-6          | Distribution Models introduction and application   |
| 4   | Week 7-8          | Document Databases ,features, use cases            |
| 5   | Week 9-10         | Graph Databases concept explanation                |
| 6   | Week 11-12        | Graph Databases concept explanation                |

## 5. Teaching-Learning Process Strategies

| S/L | <b>TLP Strategies:</b>                           | Description  |
|-----|--|--|
| 1   | Lecture Method                                   | Utilize various teaching methods within the lecture format to reinforce competencies.            |
| 2   | Video/Animation                                  | Incorporate visual aids like videos/animations to enhance understanding of NoSQL concepts.       |
| 3   | Collaborative<br>Learning                        | Encourage collaborative learning for improved competency application.                            |
| 4   | Higher Order<br>Thinking<br>(HOTS)<br>Questions: | Pose HOTS questions to stimulate critical thinking related to each competency.                   |
| 5   | Problem-Based<br>Learning (PBL)                  | Implement PBL to enhance analytical skills and practical application of competencies             |
| 6   | Multiple<br>Representations                      | Introduce topics in various representations to reinforce competencies                            |
| 7   | Real-World<br>Application                        | Discuss practical applications to connect theoretical concepts with real-<br>world competencies. |

## 6. Assessment Details (both CIE and SEE)

This section of regulations is applicable to all theory-based courses. The minimum CIE marks requirement is 40% of maximum marks in each component.

## **CIE Split up for Professional Elective Course (PE)**

| Com  | ponents                       | Number | Weightage | Max.<br>Marks | Min. Marks |
|------|-------------------------------|--------|-----------|---------------|------------|
| (i)  | Internal Assessment-Tests (A) | 3      | 50%       | 25            | 10         |
| (ii) | Assignments/Quiz/Activity (B) | 2      | 50%       | 25            | 10         |
|      | Total Marks                   |        |           | 50            | 20         |

## Final CIE Marks =(A) + (B)

Average internal assessment shall be the average of the best two test marks from the 3 tests conducted. **Semester End Examinations: PG Programmes** 

Theory Course with 4, 3 and 2 Credits: Professional Core Course (PC)/Professional Elective/Open Elective

- Question paper pattern will be ten questions. Each question is set for 20marks. The medium of the question paper shall be English unless otherwise it is mentioned.
- 2 There shall be 2 question from each module, each of the two questions under a module (with a maximum of 3 sub questions), may have mix of topics under that module if necessary.
- 3The students have to answer 5 full questions selecting one full question from each module.
- Marks scored will be proportionally scaled down to 50 marks

## 7. Learning Objectives

| S/L                      | Learning<br>Objectives                                  | Description  |
|--------------------------|---|--|
| 1                        | Understand the<br>Evolution and<br>Need for NoSQL       | Student knows the limitations of traditional relational databases.<br>And analyses given any scenarios and use cases where NoSQL databases are<br>more appropriate.  |
| 2                        | Differentiate<br>Between Types<br>of NoSQL<br>Databases | Student is able to Identify and explain the different types of NoSQL databases: document-oriented, key-value stores, column-family stores, and graph databases. And the strengths and weaknesses of each type.   |
| 3                        | Data Modelling  | Student is going to Develop data models appropriate for different types of<br>NoSQL databases. Implement schemas in document-oriented databases (e.g.,<br>MongoDB).Understand and apply denormalization and embedding<br>techniques. Design efficient data structures to optimize performance. |
| 4                        | Technical Skills  | Students are going to Execute Create, Read, Update, and Delete operations<br>using NoSQL query languages and Eviscerate and manage indexes to<br>improve query performance. Apply techniques for optimizing queries in<br>NoSQL databases.   |
| 5                        | Performance and Scalability                             | Student make use of tools and techniques to monitor database performance.<br>Identify performance bottlenecks and apply tuning strategies.   |
| 6 Continuous<br>Learning |   | Student should Stay Updated with NoSQL Trends: Follow the latest developments and best practices in NoSQL technologies with new NoSQL databases and features.  |

## 8. Course Outcomes (COs) and Mapping with POs/ PSOs

| r            | Course Outcomes (COs)  |  |  |  |  |
|--------------|--|--|--|--|--|
| CO's         | Description  |  |  |  |  |
| M23MCS204D.1 | Define, compare and use the four types of NoSQL Databases (Document-oriented, Key        |  |  |  |  |
|              | Value Pairs, Column-oriented and Graph).   |  |  |  |  |
| M23MCS204D.2 | Demonstrate an understanding of the detailed architecture, define objects, load data,    |  |  |  |  |
|              | query data and performance tune Column-oriented NoSQL databases.                         |  |  |  |  |
| M23MCS204D.3 | Explain the detailed architecture, define objects, load data, query data and performance |  |  |  |  |
|              | tune Document-oriented NoSQL databases.  |  |  |  |  |
| M23MCS204D.4 | Apply techniques for optimizing queries in NoSQL databases                               |  |  |  |  |
| M23MCS204D.5 | Apply the latest developments and best practices in NoSQL technologies with new          |  |  |  |  |
|              | NoSQL  |  |  |  |  |

| COs/POs      | PO1 | PO2 | PO3 | PSO1 | PSO2 |  |
|--------------|-----|-----|-----|------|------|--|
| M23MCS204D.1 | 3   | -   | -   | 3    | 3    |  |
| M23MCS204D.2 | 3   | -   | -   | 3    | 3    |  |
| M23MCS204D.3 | -   | 3   | 3   | 3    | 3    |  |
| M23MCS204D.4 | 3   |     |     | 3    | 3    |  |
| M23MCS204D.5 |     | 3   |     | 3    | 3    |  |
| M23MCS204D   | 3   | 3   | 3   | 3    | 3    |  |

## **CO-PO-PSO Mapping:**

## 9. Assessment Plan

**Continuous Internal Evaluation (CIE)** 

|          | CO1 | CO2 | CO2 | CO3 | CO3 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 10  |     |     |     |     | 10    |
| Module 2 |     | 10  |     |     |     | 10    |
| Module 3 |     |     | 10  |     |     | 10    |
| Module 4 |     |     |     | 10  |     | 10    |
| Module 5 |     |     |     |     | 10  | 10    |
| Total    | 10  | 10  | 10  | 10  | 10  | 50    |

#### Semester End Examination (SEE)

|          | CO1 | CO2 | CO2 | CO3 | CO3 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 20  |     |     |     |     | 20    |
| Module 2 |     | 20  |     |     |     | 20    |
| Module 3 |     |     | 20  |     |     | 20    |
| Module 4 |     |     |     | 20  |     | 20    |
| Module 5 |     |     |     |     | 20  | 20    |
| Total    | 20  | 20  | 20  | 20  | 20  | 100   |

## **Conditions for SEE Paper Setting**

Each module of SEE question paper should be allocated with questions for 20% of the total SEE marks

#### **10.** Future with this Subject

The future of NoSQL databases is bright, with continuous advancements in technology and increasing adoption across various industries. By staying abreast of these trends and developments, professionals and organizations can leverage the full potential of NoSQL databases to meet their evolving data management needs.

#### 1. Big Data and Real-Time Analytics

Scalability and Performance: NoSQL databases are designed to handle large volumes of unstructured and semi-structured data, making them ideal for big data applications.

Real-Time Processing: As real-time data processing and analytics become more crucial for businesses, NoSQL databases will continue to be essential for applications that require low-latency data access.

#### 2. Cloud Computing and Serverless Architectures

Cloud Integration: NoSQL databases are well-suited for cloud environments due to their ability to scale horizontally and handle distributed data across multiple nodes.

Serverless Databases: The adoption of serverless computing models will drive the demand for NoSQL databases that can efficiently manage data without the need for traditional server infrastructure.

## 3. Micro services Architecture

Decentralized Data Management: Micro services architecture promotes the use of decentralized data stores, and NoSQL databases fit well into this model by providing flexibility and scalability.

Polyglot Persistence: As applications adopt polyglot persistence (using different data storage technologies for different needs), NoSQL databases will be a key component in the data architecture. **4. Data Privacy and Security** 

Enhanced Security Features: NoSQL databases will continue to evolve to provide enhanced security features, including encryption, fine-grained access controls, and compliance with data protection regulations.

Privacy by Design: Future NoSQL databases will increasingly incorporate privacy by design principles to protect user data and ensure compliance with global privacy laws.

#### 5. Cross-Platform and Multi-Model Databases

Hybrid Databases: The lines between NoSQL and SQL databases are blurring, with many databases now offering hybrid capabilities to support both relational and non-relational data models.

Multi-Model Databases: NoSQL databases will continue to evolve to support multiple data models (document, graph, key-value, column-family) within a single database engine, providing greater flexibility for developers.

## 6. Adoption in Enterprises

Mainstream Acceptance: NoSQL databases are increasingly being adopted by large enterprises for mission-critical applications, indicating growing trust and maturity in the technology.

Legacy Modernization: Many organizations are using NoSQL databases to modernize legacy systems, enabling more agile and scalable data management solutions.

## 7. AI-Powered Database Management

Automation and Self-Tuning: AI and machine learning will be used to automate database management tasks such as indexing, query optimization, and resource allocation, making NoSQL databases more efficient and easier to manage.

Predictive Analytics: AI-powered predictive analytics will enhance the ability of NoSQL databases to anticipate and respond to performance issues and changing workloads.



| 2 <sup>nd</sup> Semester | Professional Core Laboratory Course (PCL) | M23MCSL206   |
|--------------------------|---|--------------|
| 2 Semester               | <b>Big-Data Analytics Laboratory</b>      | WIZSWICSL200 |

## 1. Prerequisites

| S/L | Proficiency                                      | Prerequisites  |
|-----|--|--|
| 1   | Basic Computer<br>Science Knowledge              | <b>Computer Programming:</b> Proficiency in a programming language<br>such as Java, Python, or Scala. Understanding of object-oriented<br>programming (OOP) principles is particularly beneficial.<br><b>Data Structures and Algorithms:</b> Knowledge of fundamental data<br>structures (arrays, lists, stacks, queues, hash tables, trees, and graphs)<br>and algorithms (sorting, searching, etc.). |
| 2   | Database<br>Management<br>Systems (DBMS)         | <ul> <li>SQL: Basic understanding of SQL for querying and managing databases.</li> <li>Relational Databases: Familiarity with relational database concepts such as normalization, indexing, transactions, and schema design.</li> </ul>  |
| 3   | Basic Statistics and<br>Mathematics              | <ul> <li>Descriptive Statistics: Mean, median, mode, standard deviation, variance, and data distributions.</li> <li>Probability: Basic probability concepts and distributions.</li> <li>Linear Algebra: Vectors, matrices, and matrix operations.</li> </ul>   |
| 4   | Basic Understanding<br>of Distributed<br>Systems | <b>Distributed Computing Principles:</b> Basic concepts of distributed systems, such as distributed storage and processing.  |
| 5   | Introduction to<br>Cloud Computing               | <b>Cloud Basics:</b> Understanding of what cloud computing is, its benefits, and basic cloud service models (IaaS, PaaS, SaaS).<br><b>Cloud Providers:</b> Familiarity with major cloud service providers like AWS, Azure, and Google Cloud.   |
| 6   | Introduction to Big<br>Data Concepts             | <b>Big Data Fundamentals:</b> Awareness of what big data is, its significance, and its applications.<br><b>Data Formats:</b> Understanding of different data formats (structured, semi-structured, unstructured).  |

## 2. Competencies

| S/L | Competency                            | KSA Description  |
|-----|---------------------------------------|--|
| 1   | Getting an<br>Overview of<br>Big Data | <ul> <li>Knowledge:</li> <li>Understanding of the definition and components of big data (Volume, Velocity, Variety, Veracity).</li> <li>Familiarity with the history and evolution of data management leading to big data.</li> <li>Skills:</li> <li>Ability to identify and categorize different types of data.</li> <li>Proficiency in analysing big data characteristics and their implications.</li> <li>Attitudes:</li> <li>Appreciation of the transformative potential of big data in various sectors.</li> </ul> |
| 2   | Introducing<br>Hadoop<br>Framework    | <ul> <li>Knowledge:</li> <li>Understanding of Hadoop architecture and its ecosystem components.</li> <li>Skills:</li> <li>Proficiency in setting up and configuring Hadoop environments.</li> <li>Ability to navigate and use the Hadoop Distributed File System (HDFS).</li> <li>Attitudes:</li> <li>Openness to learning and adopting new technologies.</li> </ul>   |
| 3   | Hadoop<br>MapReduce<br>Framework      | Knowledge:<br>Understanding of the MapReduce model, its parallel data flow, and fault<br>tolerance mechanisms.<br>Skills:  |

| - |  |   |  |  |  |  |
|---|--|---|--|--|--|--|
|   |  | Ability to write, compile, and execute MapReduce programs.                  |  |  |  |  |
|   |  | Proficiency in optimizing MapReduce jobs for performance and efficiency.    |  |  |  |  |
|   |  | Attitudes:  |  |  |  |  |
|   |  | Problem-solving mindset to tackle data processing challenges.               |  |  |  |  |
|   |  | Knowledge:  |  |  |  |  |
|   | II. J VADN                                   | Understanding of YARN architecture, schedulers, and backward compatibility. |  |  |  |  |
|   | Hadoop YARN                                  | Skills:   |  |  |  |  |
| 4 | Architecture<br>and Tools to<br>Analyze Data | Ability to configure and manage YARN for resource scheduling and job        |  |  |  |  |
|   |  | management.   |  |  |  |  |
|   |  | Proficiency in using analytical tools to process and analyze big data.      |  |  |  |  |
|   |  | Attitudes:  |  |  |  |  |
|   |  | Openness to adopting new analytical tools and techniques.                   |  |  |  |  |
|   |  | Knowledge:  |  |  |  |  |
|   |  | Understanding of the data models and functionalities of Apache Pig, Hive,   |  |  |  |  |
|   |  | Sqoop, Flume, HBase, and Oozie.   |  |  |  |  |
|   | <b>D</b> (1)                                 | Skills:   |  |  |  |  |
| 5 | Essential                                    | Proficiency in using Apache Pig for data processing.                        |  |  |  |  |
|   | Hadoop Tools                                 | Ability to use Apache Hive for data warehousing and querying.               |  |  |  |  |
|   |  | Attitudes:  |  |  |  |  |
|   |  | Appreciation of the importance of data processing and management tools in   |  |  |  |  |
|   |  | big data workflows.   |  |  |  |  |

## 3. Syllabus

| Big Data Analytics Laboratory<br>SEMESTER – II    |                            |                           |              |  |  |  |  |  |
|---|----------------------------|---------------------------|--------------|--|--|--|--|--|
| Course Code M23MCSL206 CIE Marks 50               |                            |                           |              |  |  |  |  |  |
| Number of Lecture Hours/Week(L: T: P: S)          | (1:0:2:0)                  | SEE Marks                 | 50           |  |  |  |  |  |
| Total Number of Lecture Hours                     | 3hrs/week                  | Total Marks               | 100          |  |  |  |  |  |
| Credits   | 02                         | Exam Hours                | 03           |  |  |  |  |  |
| Course objectives: This course will enable studen | nts to:                    |                           |              |  |  |  |  |  |
| Understand the Fundamentals and Evolution of B    | ig Data                    |                           |              |  |  |  |  |  |
| Analyze the Applications and Advantages of Big    | Data Analytics             |                           |              |  |  |  |  |  |
| Master the Hadoop Framework and Ecosystem         | -                          |                           |              |  |  |  |  |  |
| Develop Skills in Hadoop MapReduce and YARN       | J                          |                           |              |  |  |  |  |  |
| Utilize Essential Hadoop Tools for Data Processin | ng and Management          |                           |              |  |  |  |  |  |
| PROGRAM-1   |                            |                           |              |  |  |  |  |  |
| Basic Linux Commands. Understanding how to co     | nnect to remote Linux ser  | ver using putty kind of t | ool. Install |  |  |  |  |  |
| Apache Hadoop                                     |                            |                           |              |  |  |  |  |  |
| PROGRAM-2   |                            |                           |              |  |  |  |  |  |
| Develop a MapReduce program to calculate the fi   | requency of a given word   | in a given file.          |              |  |  |  |  |  |
| PROGRAM-3   |                            |                           |              |  |  |  |  |  |
| Develop a MapReduce program to find the maxim     | num temperature in each y  | /ear.                     |              |  |  |  |  |  |
| PROGRAM-4   |                            |                           |              |  |  |  |  |  |
| Develop a MapReduce program to find the grades    | s of student's.            |                           |              |  |  |  |  |  |
| PROGRAM-5   |                            |                           |              |  |  |  |  |  |
| Develop a MapReduce to find the maximum elect     | rical consumption in each  | year given electrical co  | nsumption    |  |  |  |  |  |
| for each month in each year.                      |                            |                           |              |  |  |  |  |  |
| PROGRAM-6   |                            |                           |              |  |  |  |  |  |
| Develop a MapReduce program to implement Ma       | trix Multiplication.       |                           |              |  |  |  |  |  |
| PROGRAM-7   |                            |                           |              |  |  |  |  |  |
| Develop a MapReduce to analyze weather data se    | t and print whether the da | y is shinny or cool day.  |              |  |  |  |  |  |
| PROGRAM-8   |                            |                           |              |  |  |  |  |  |

PROGRAM-8

Develop a MapReduce program to find the tags associated with each movie by analyzing movie lens data.

| DD                   |  |                      |           |                |            |           |          |             |              |              |             |               |
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| PRO                  | OGRA   | M-11                 |           |                |            |           |          |             |              |              |             |               |
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| -                    |  |                      | e, "Had   | loop: The      | Defini     | tive Gu   | ide" Fo  | ourth Editi | on, O'reilly | / Media,     | 2015.       |               |
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| 4. Syllabus Timeline |  |                      |           |                |            |           |          |             |              |              |             |               |
|                      |  | labus                |           |                |            |           |          |             |              |              |             |               |
| S/L                  |  | neline               |           |                |            |           |          | Descrip     | otion        |              |             |               |
| ~~ <b>1</b>          |  |                      |           |                |            |           |          | 200011      |              |              |             |               |
|                      |  | 1.1                  | Bas       | ic Linux C     | omma       | nds. Ur   | ndersta  | nding how   | to connect   | to remo      | te Linux s  | erver using   |
| 1                    | W  | eek 1                |           |                |            |           |          |             | g Data Ana   |              |             | 8             |
| •                    | 117  | 1- 2                 |           |                |            |           |          |             | the frequen  |              | given word  | in agiven     |
| 2                    | W  | eek 2                | file.     | 1              | 1          | 1 0       |          |             | 1            |              |             | e             |
| 3                    | W  | eek 3                | Dev       | velop a Ma     | pRedu      | ce prog   | ram to   | find the n  | naximum te   | mperatu      | re in each  | year.         |

| S/L  | Syllabus<br>Timeline  | Description  |  |  |  |
|--|---|--|--|--|--|
| I         Basic Linux Commands. Understanding how to connect to remote Linux putty kind of tool. Install Apache HadoopBig Data Analytics |   |  |  |  |  |
| 2  | Develop a ManReduce program to calculate the frequency of a given word in a |  |  |  |  |
| 3  | Week 3  | Develop a MapReduce program to find the maximum temperature in each year.  |  |  |  |
| 4  | 4 Week 4 Develop a MapReduce program to find the grades of student's.       |  |  |  |  |
| 5  | Week 5  | Develop a MapReduce to find the maximum electrical consumption in each year given electrical consumption for each month in each year.    |  |  |  |
| 6  | Week 6  | Develop a MapReduce program to implement Matrix Multiplication.  |  |  |  |
| 7  | Week 7  | Develop a MapReduce to analyze weather data set and print whether the day is shinny or cool day.   |  |  |  |
| 8  | Week 8  | Develop a MapReduce program to find the tags associated with each movie by analyzing movie lens data.                                    |  |  |  |
| 9  | Develop a program to calculate the maximum recorded temperature by yearwis  |  |  |  |  |
| 10   | Week 10   | Develop a MapReduce program to analyze Uber data set to find the days on which each basement has more trips using the following dataset. |  |  |  |

|    |         | The Uber dataset consists of four columns they are   |        |        |        |   |     |                                  |                               |  |                  |
|----|---------|--|--------|--------|--------|---|-----|----------------------------------|-------------------------------|--|------------------|
|    |         | Dispatch   | ingbas | e_numb | er dat | e | act | ive veh                          | icles                         | trips                                      |                  |
| 11 | Week 11 | Develop a MapReduce program to analyze Titanic ship data and to find the average of the people (both male and female) who died in the tragedy. How many personal are survived in each class.The titanic data will beColumn 1 :Passenger Id&died=1)Column 3 :PclassColumn 5 : SexColumn 7 :SibSpColumn 9 : TicketColumn 11 :CabinColumn 12 : Embarked |        |        |        |   |     | ons                              |                               |  |                  |
| 12 | Week 12 | Develop a<br>by conside<br>Tranct<br>ion<br>_Date  |        |        |        |   |     | of proc<br>S<br>t<br>a<br>t<br>e | ducts sold<br>Co<br>unt<br>ry | in each cour<br>Acc<br>ount<br>Crea<br>ted | ntry<br>La<br>Lo |

## 5. Teaching-Learning Process Strategies

| S/L  | TLP<br>Strategies:            | Description   |  |  |  |
|--|-------------------------------|---|--|--|--|
| 1  | Lectures and<br>Presentations | Use multimedia presentations to explain key concepts like the definition of big data, its characteristics (Volume, Velocity, Variety, Veracity), and its evolution. |  |  |  |
| 2  | Class<br>Discussions          | Facilitate discussions on the advantages of big data analytics and its impact on different industries.  |  |  |  |
| 3  | Case Study<br>Analysis        | Provide case studies on the use of big data in different industries and have students analyze and discuss them in groups.   |  |  |  |
| 4  | Quizzes and<br>Assessments:   | Conduct quizzes to test students' understanding of the fundamental concepts of big data.  |  |  |  |
| 5  | Interactive<br>Lectures       | Use diagrams and flowcharts to illustrate the Hadoop Distributed File System (HDFS) architecture and its features.  |  |  |  |
| 6  | Hands-On Labs                 | Conduct lab sessions where students install and configure Hadoop on virtual machines.   |  |  |  |
| 7  | Step-by-Step<br>Tutorials     | Provide detailed tutorials on the MapReduce model, including writing, compiling, and running MapReduce programs.  |  |  |  |
| 8  | Lectures and<br>Diagrams      | Use diagrams to explain the YARN architecture and its components.   |  |  |  |
| 9 Tool-Specific Provide detailed<br>Lectures: and Oozie. |                               | Provide detailed lectures on using Apache Pig, Hive, Sqoop, Flume, HBase, and Oozie.  |  |  |  |
| 10   | Blended<br>Learning           | Combine online resources, video tutorials, and in-class activities to provide a diverse learning experience.  |  |  |  |

## 6. Assessment Details (both CIE and SEE)

Note:

- ✓ Different types of courses will different assessment patterns, for which the applicable rules and regulations may be referred.
- ✓ An illustration for one of the course is given below.

Principal Minipal Internal test for laboratory course with software experiments shall be conducted for a total of 100 mark at the end the semester and the assessment pattern is

## Marks distribution for Program based Practical Course for CIE

| Sl. No. | Description   | % of Marks         | In Marks |
|---------|---|--------------------|----------|
| 1       | Observation, write-up,<br>algorithm/program/execution | 80% of the maximum | 80       |
| 2       | Viva-Voce   | 20% of the maximum | 20       |
|         | Total   | 100%               | 100      |

Marks scored by the student for 100 are scaled down to 50 marks.

SEE for practical Course (Irrespective of Experiment or program based):

• SEE marks for practical course shall be 50 marks

|  | Marks distri | ibution for Experiment ba | ased Practical | <b>Course for Final</b> | CIE |
|--|--------------|---------------------------|----------------|-------------------------|-----|
|--|--------------|---------------------------|----------------|-------------------------|-----|

| SL. No. | Description           | % of Marks | Marks |
|---------|-----------------------|------------|-------|
| 1       | Write-up, Procedure   | 20%        | 20    |
| 2       | Conduction and result | 60%        | 60    |
| 3       | Viva-Voce             | 20%        | 20    |
|         | Total                 | 100%       | 100   |

• See for practical course is evaluated for 100 marks and scored marks shall be scaled down to 50 marks.

- Change of experiment/program is allowed only once and 20% marks allotted to the procedure/writeup part to be made zero.
- Duration of SEE shall be 3 hours.

## 7. Learning Objectives

 $\checkmark$ 

| S/L | Learning<br>Objectives   | Description  |
|-----|--|--|
| 1   | Understand the<br>Fundamentals of<br>Big Data                            | Define what constitutes big data and describe its key characteristics: Volume,<br>Velocity, Variety, and Veracity.<br>Trace the history and evolution of data management, leading to the<br>emergence of big data.<br>Differentiate between structured, semi-structured, and unstructured data.  |
| 2   | Analyze Big Data<br>Analytics and Its<br>Applications                    | Explain the core elements of big data analytics and its advantages in various<br>industries.<br>Evaluate the use of big data in social networking to enhance user experience<br>and engagement.<br>Assess the role of big data in preventing and detecting fraudulent activities,<br>particularly in the insurance and retail sectors. |
| 3   | Master the Hadoop<br>Framework   | Describe the architecture and components of the Hadoop ecosystem,<br>including HDFS and MapReduce.<br>Explain how Hadoop functions, including the role of cloud computing in big<br>data processing.<br>Identify key cloud deployment and delivery models, and explain the features<br>of cloud computing relevant to big data.        |
| 4   | Develop<br>Proficiency in<br>Hadoop<br>Distributed File<br>System (HDFS) | Understand the HDFS architecture, its features, and components.<br>Use HDFS user commands to manage and manipulate data within the Hadoop<br>ecosystem.<br>Describe the in-memory computing technology for big data and its benefits.  |
| 5   | Gain Skills in<br>Hadoop<br>MapReduce<br>Framework                       | Explain the MapReduce model, including its parallel data flow, fault<br>tolerance, and speculative execution.<br>Write, compile, and execute MapReduce programs, and optimize these jobs<br>for better performance.<br>Use the streaming and pipes interfaces for different data processing tasks in<br>MapReduce.                     |

| 6 | Explore Hadoop<br>YARN<br>Architecture and<br>Analytical Tools | Describe the background, architecture, and working of YARN, including its<br>schedulers and configurations.<br>Execute YARN commands and understand its backward compatibility.<br>Compare various popular analytical tools, their history, and their applications<br>in big data analysis. |
|---|--|---|
| 7 | Utilize Essential<br>Hadoop Tools for<br>Data Processing       | Use Apache Pig, Hive, Sqoop, Flume, HBase, and Oozie for specific data<br>processing tasks.<br>Perform data import and export operations using Apache Sqoop, and<br>understand its version changes.<br>Implement data models in HBase and manage workflows with Apache Oozie.               |

## 8. Course Outcomes (COs) and Mapping with POs/ PSOs

## **Course Outcomes (COs)**

| COs          | Description   |
|--------------|---|
| M23MCSL206.1 | Understand and apply Big Data Concepts and Analytics  |
| M23MCSL206.2 | Analyse and apply Big Data analytics in various sectors such as social networking, fraud detection, and retail. |
| M23MCSL206.3 | Describe the Hadoop framework and its components, including cloud computing and HDFS.                           |
| M23MCSL206.4 | Develop and execute MapReduce programs, optimizing them for better performance                                  |
| M23MCSL206.5 | Utilize essential Hadoop tools like Apache Pig, Hive, Sqoop, Flume, HBase, and                                  |
|              | Oozie for data processing   |

## **CO-PO-PSO Mapping**

| COs/POs      | PO1 | PO2 | PO3 | PSO1 | PSO2 |
|--------------|-----|-----|-----|------|------|
| M23MCSL206.1 | 3   | 2   | 3   | 2    | 2    |
| M23MCSL206.2 | 3   |     | 3   | 2    | 2    |
| M23MCSL206.3 |     | 2   | 3   | 2    | 3    |
| M23MCSL206.4 | 3   |     | 3   | 3    | 3    |
| M23MCSL206.5 | 3   |     | 3   | 3    | 3    |
| M23MCSL206   | 3   | 2   | 3   | 3    | 3    |

## 9. Assessment Plan

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

|          | CO1 | CO2 | CO3 | CO4 | CO5 | Total |  |  |
|----------|-----|-----|-----|-----|-----|-------|--|--|
| Module 1 | 10  |     |     |     |     | 10    |  |  |
| Module 2 |     | 10  |     |     |     | 10    |  |  |
| Module 3 |     |     | 10  |     |     | 10    |  |  |
| Module 4 |     |     |     | 10  |     | 10    |  |  |
| Module 5 |     |     |     |     | 10  | 10    |  |  |
| Total    | 10  | 10  | 10  | 10  | 10  | 50    |  |  |

#### Semester End Examination (SEE)

|          | CO1 | CO2 | CO3 | CO4 | CO5 | Total |
|----------|-----|-----|-----|-----|-----|-------|
| Module 1 | 20  |     |     |     |     | 20    |
| Module 2 |     | 20  |     |     |     | 20    |
| Module 3 |     |     | 20  |     |     | 20    |
| Module 4 |     |     |     | 20  |     | 20    |
| Module 5 |     |     |     |     | 20  | 20    |
| Total    | 20  | 20  | 20  | 20  | 20  | 100   |

#### **Conditions for SEE Paper Setting:**

Each module of SEE question paper should be allocated with questions for 20% of the total SEE marks

## 10. Future with this Subject

The "Big Data Analytics" course in the second semester of the M.Tech program provides a comprehensive foundation in Big Data and Hadoop technologies, opening various career paths and opportunities in multiple industries. With the growing importance of data-driven decision-making, professionals skilled in Big Data analytics and Hadoop are in high demand. The integration of emerging technologies will further expand the scope and impact of Big Data, making it a pivotal field for future innovations.

- 1. **Career Opportunities:** Data scientists and analysts are in high demand as organizations seek to extract meaningful insights from their data. With skills in Big Data and Hadoop technologies, professionals can analyze large datasets, identify trends, and make data-driven decisions to benefit businesses.
- 2. **Research and Development:** Professionals with expertise in Big Data can engage in research to develop advanced analytics techniques. This includes predictive analytics, real-time data processing, and innovative ways to handle unstructured data.
- 3. **Machine Learning and AI Integration:** Big Data plays a crucial role in training machine learning models. Researchers can explore the integration of Big Data with AI to create smarter, more efficient algorithms. Leads to advancements in AI technologies, improving automation, decision-making, and personalized services.
- 4. **Emerging Technologies:** IoT devices generate massive amounts of data. Big Data technologies are essential for collecting, processing, and analyzing this data to derive actionable insights. Facilitates smart cities, industrial automation, and improved resource management.
- 5. **Industry Applications:** Big Data analytics can revolutionize healthcare by providing insights into patient care, treatment outcomes, and operational efficiency. It helps in predictive modeling for disease outbreaks and personalized medicine. Improves patient outcomes, reduces costs, and enhances the overall efficiency of healthcare systems.
- 6. **Social Media and Networking:** Social media platforms use Big Data to analyze user behavior, content trends, and advertising effectiveness. Hadoop's ability to handle vast amounts of unstructured data makes it ideal for this sector. Enhances user engagement, optimizes content delivery, and improves targeted advertising.

