# **SEMESTER - 5**

# Paper Code: BDS501 Paper Name: *Discrete Mathematics* Credits: 5

# **MODULE 1: Introduction to Discrete Mathematics**

- - Basic terminologies and concepts in discrete mathematics
- - Sets, set operations, and set theory
- - Functions and relations

### **MODULE 2: Logic and Proof Techniques**

- Propositional logic: syntax, semantics, and truth tables
- - Logical connectives and truth functions
- - Predicates and quantifiers
- - Proofs: direct proofs, proof by contradiction, and proof by contrapositive

### **MODULE 3: Methods of Proof**

- - Mathematical induction
- - Strong induction
- - Recursive definitions and structural induction

#### **MODULE 4: Combinatorics**

- - Basic counting principles: multiplication rule, addition rule, and inclusion-exclusion principle
- - Permutations and combinations
- - Binomial coefficients and Pascal's triangle
- - The pigeonhole principle

#### **MODULE 5: Relations and Functions**

- - Relations: properties, equivalence relations, and partial orders
- Functions: injective, surjective, and bijective functions
- - Composition of functions and inverse functions

#### **MODULE 6: Graph Theory**

- - Basic terminologies: vertices, edges, paths, and cycles
- - Graph representations: adjacency matrix and adjacency list
- - Connectivity: connected graphs, components, and bridges
- - Eulerian and Hamiltonian paths and cycles

#### **MODULE 7: Trees**

- - Definitions and properties of trees
- - Tree traversal algorithms: depth-first search and breadth-first search
- - Spanning trees and minimum spanning trees
- - Applications of trees in computer science

### MODULE 8: Boolean Algebra

- Boolean functions and Boolean expressions
- - Logic gates and Boolean circuits
- - Simplification of Boolean expressions: Karnaugh maps and Quine-McCluskey method
- - Boolean algebra theorems and applications

#### **MODULE 9: Discrete Probability**

- - Sample spaces, events, and probability axioms
- - Conditional probability and independence
- - Counting techniques in probability
- - Discrete probability distributions: binomial and geometric distributions

### **MODULE 10: Recurrence Relations**

- - Introduction to recurrence relations
- - Solving linear homogeneous recurrence relations with constant coefficients
- Solving linear non-homogeneous recurrence relations
- - Generating functions and solving recurrence relations using generating functions

#### **References:**

1. *Discrete Mathematical Structures*, Bernard Kolman, Robert C. Busby, Sharon Cutler Ross, Pearson/Prentice Hall, 2009

2. Discrete Mathematics and Its Applications, Kenneth Rosen, McGrawhill Education, 2017

# Paper Code: BDS502& BDS592 Paper Name: *Natural Language Processing* Credits: 3+2

# 1. Introduction to Natural Language Processing

- Definition and scope of NLP
- - Applications of NLP
- - NLP challenges and limitations

# 2. Text Preprocessing

- Tokenization
- - Stop word removal
- - Stemming and lemmatization
- Text normalization

### 3. Language Modeling

- - N-gram models
- - Hidden Markov Models (HMM)
- Neural language models

### 4. Syntactic Analysis

- - Part-of-speech tagging
- - Phrase structure parsing
- - Dependency parsing

# 5. Semantic Analysis

- - Word sense disambiguation
- Named entity recognition
- - Semantic role labeling

#### 6. Sentiment Analysis

- Opinion mining
- - Sentiment classification techniques
- - Aspect-based sentiment analysis

#### 7. Machine Translation

- Rule-based machine translation
- Statistical machine translation
- Neural machine translation

#### 8. Information Extraction

- - Named entity recognition
- - Relation extraction
- - Event extraction

### 9. Question Answering

- Question classification
- Passage retrieval
- Answer extraction

### **10. Advanced Topics**

- - Deep learning for NLP
- - Neural network architectures (e.g., recurrent and convolutional models)
- - Transfer learning in NLP

# **11. NLP Applications and Research Trends**

- Chatbots and virtual assistants
- - Text summarization
- - Dialogue systems
- Recent developments and future directions in NLP

# 12. Project Work and Presentations

• - Students will work on a hands-on NLP project and present their findings to the class.

# **References:**

1. "Natural Language Processing: Basic Techniques and Emerging Trends" by Nitin Indurkhya and Fred J. Damerau, CRC Press.

2. "Natural Language Processing: A Paninian Perspective" by Amba Kulkarni, Rajeev Sangal, and Vineet Chaitanya, Prentice-Hall of India.

3. "Foundations of Natural Language Processing" by Pushpak Bhattacharyya, Pearson Education India.

### **SEMESTER - 6**

# Paper Code: BDS601 & BDS691 Paper Name: *Big Data Analysis* Credits: 3+2

### Module 1: Introduction to Big Data Analysis

- - Overview of big data and its characteristics
- - Introduction to Hadoop and MapReduce
- - Introduction to Apache Spark

### Module 2: Data Preprocessing and Cleaning

- Data quality assessment and data cleaning techniques
- - Handling missing data and outliers
- - Feature selection and dimensionality reduction

### Module 3: Data Transformation and Integration

- Data normalization and standardization
- - Encoding categorical variables
- - Data integration techniques

### Module 4: Exploratory Data Analysis

- - Descriptive statistics and data visualization
- Univariate and multivariate analysis
- Outlier detection and handling

#### **Module 5: Big Data Processing Frameworks**

- - Introduction to Apache Hadoop ecosystem (HDFS, Hive, Pig)
- Introduction to Apache Spark (RDDs, DataFrames, Spark SQL)

#### Module 6: Big Data Processing with Spark

- Data processing with Spark RDDs
- Data processing with Spark DataFrames and Spark SQL
- Introduction to Spark MLlib for machine learning

#### Module 7: Machine Learning with Big Data

- Supervised learning techniques for big data
- Unsupervised learning techniques for big data
- Model evaluation and validation

# Module 8: Advanced Topics in Big Data Analysis

- Time series analysis with big data
- - Text mining and natural language processing
- - Graph analysis and social network analysis

# Module 9: Ethical Considerations in Big Data Analysis

- - Privacy issues and data anonymization
- - Bias and fairness in big data analysis
- - Ethical frameworks and guidelines

# Module 10: Case Studies and Project

- - Analysis of real-world big data datasets
- - Application of big data analysis techniques
- - Presentation of project findings

# **References**:

- 1. Pyne, S., & Agrawal, P. (2016). Big Data Analytics: Methods and Applications. CRC Press.
- 2. White, T. (2015). Hadoop: The Definitive Guide. O'Reilly Media.

3. Zaharia, M., Karau, H., Konwinski, A., & Wendell, P. (2021). Learning Spark: Lightning-Fast Big Data Analysis. O'Reilly Media.

# Paper Code: BDS602 & BDS692 Paper Name: *Time Series Analysis* Credits: 3+2

# Module 1: Introduction to Time Series Analysis

- - Definition of time series
- - Components of a time series
- Time series data visualization
- - Applications of time series analysis

# Module 2: Descriptive Analysis of Time Series

- - Measures of central tendency and dispersion
- - Time series plots and patterns
- Seasonality and trend analysis

# Module 3: Autocorrelation and Partial Autocorrelation

- - Autocorrelation function (ACF)
- - Partial autocorrelation function (PACF)
- - Identification of autoregressive (AR), moving average (MA), and autoregressive integrated moving average (ARIMA) models

# Module 4: Stationarity and Non-Stationarity

- - Definition and characteristics of stationarity
- - Unit root tests (e.g., Augmented Dickey-Fuller test)
- - Transformations and differencing for achieving stationarity
- - Trend and seasonality decomposition

# **Module 5: Time Series Forecasting**

- - Simple moving average and weighted moving average
- - Exponential smoothing methods (e.g., Holt-Winters method)
- - ARIMA models for forecasting
- - Evaluation and accuracy measures for forecasting models

#### **Module 6: Time Series Regression**

- Simple and multiple linear regression models
- Incorporating lagged variables in regression models
- - Testing and interpreting the regression coefficients
- Model diagnostics and residuals analysis

# Module 7: Advanced Topics in Time Series Analysis (choose a subset)

- ARCH and GARCH models for volatility modeling
- - Vector Autoregressive (VAR) models
- - Seasonal ARIMA models (SARIMA)
- State-space models and Kalman filtering
- - Spectral analysis and periodogram

# **Module 8: Applications of Time Series Analysis**

- - Economic and financial time series analysis
- - Environmental and climate data analysis
- - Engineering and industrial applications
- - Social and health-related time series analysis

# Module 9: Software Tools for Time Series Analysis

- - Introduction to R or Python programming for time series analysis
- Hands-on exercises using statistical software packages
- - Implementing forecasting models and analyzing real-world data

# Module 10: Case Studies and Project

- Review of case studies from various domains
- Student projects involving data analysis and modeling
- - Presentation and discussion of project results

### **References**:

1. Box, G. E. P., Jenkins, G. M., & Reinsel, G. C. (2008). Time Series Analysis: Forecasting and Control. Wiley-Interscience. ISBN: 978-0470272848.

2. Chatfield, C. (2019). Introduction to Time Series Analysis and Forecasting. Chapman and Hall/CRC. ISBN: 978-0367352090.

3. Pandit, A., & Wu, W. B. (2019). Applied Time Series Analysis. CRC Press. ISBN: 978-0367242483.

# Paper Code: BDS603 & BDS693 Paper Name: *Data Visualization* Credits: 2+2

# Module 1: Introduction to Data Visualization

- - Understanding the importance of data visualization
- Principles and guidelines for effective data visualization
- Overview of data visualization process
- Introduction to data visualization tools

# Module 2: Data Types and Visualization Techniques

- - Types of data and their visual representations
- - Categorical data visualization
- Numerical data visualization
- - Time-series data visualization
- Geospatial data visualization

# Module 3: Visualization Tools and Technologies

- - Introduction to popular data visualization tools (e.g., Tableau, Power BI, D3.js)
- - Exploring features and capabilities of visualization tools
- Hands-on exercises with visualization tools

# **Module 4: Data Visualization Design Principles**

- - Understanding human perception and cognition in visualization
- - Color theory and effective use of colors in visualization
- - Designing visually appealing and informative visualizations
- Interaction and user experience design in visualization

# Module 5: Advanced Data Visualization Techniques

- Multivariate data visualization
- Network visualization
- Hierarchical visualization
- - 3D and interactive visualizations

# Module 6: Storytelling with Data Visualization

- - Creating compelling narratives through visualizations
- Techniques for presenting data stories effectively
- Ethical considerations in data visualization
- Case studies and examples of data storytelling

# **Module 7: Data Visualization Applications**

- Visualizing data for exploratory analysis
- - Visual analytics and decision support systems
- Data visualization for business intelligence
- Data visualization in scientific research

# **Module 8: Data Visualization Project**

- - Working on a data visualization project from conception to execution
- - Data collection and preprocessing
- - Designing and implementing visualizations
- - Presenting and documenting the project

### **References:**

- 1. Data Visualization: Storytelling Using Data by Sharada Sringeswara; Purvi Tiwari & U. Dinesh Kumar, Wiley
- 2. "Storytelling with Data: A Data Visualization Guide for Business Professionals" by Cole Nussbaumer Knaflic

# **SEMESTER - 7**

# Paper Code: BDS701A Paper Name: *Research Methodology* Credits: 5

# Module 1: Introduction to Research Methodology

- Definition, significance, and characteristics of research
- - Objectives and types of research
- - Research process and steps
- - Research ethics and plagiarism

# Module 2: Formulating Research Questions and Objectives

- Identification of research problem
- Review of literature
- - Research questions and hypotheses
- - Research objectives and significance

# Module 3: Research Design

- - Experimental, descriptive, and exploratory research designs
- Longitudinal, cross-sectional, and case study designs
- Sampling techniques and sample size determination

# **Module 4: Data Collection Methods**

- Primary and secondary data
- - Questionnaire design and administration
- Interview methods
- Observation techniques

# Module 5: Data Analysis and Interpretation

- Qualitative and quantitative data analysis
- Descriptive and inferential statistics
- - Statistical software (e.g., SPSS, Excel) for data analysis
- Interpretation of research findings

# Module 6: Reporting and Presentation of Research

- - Research report writing
- - Organization and structure of research reports
- - Referencing and citation styles
- Presentation skills and effective communication of research findings

# Module 7: Research Ethics and Validity

- - Ethical considerations in research
- - Validity and reliability in research
- Limitations and challenges in research
- - Strategies for enhancing research validity

# Paper Code: BDS701B

Paper Name: *Project Management* Credits: 5

# Module 1: Introduction to Project Management

- Definition and characteristics of projects
- Importance of project management
- Project life cycle
- Project stakeholders

# **Module 2: Project Initiation**

- - Project selection and feasibility analysis
- Project charter

# Module 3: Project Scope Management

- Project scope definition
- Work breakdown structure (WBS)

# Module 4: Project Planning

- - Project objectives and deliverables
- Activity sequencing and network diagrams

# Module 5: Project Scheduling

- Critical path method (CPM)
- Resource allocation and leveling
- Gantt charts

# Module 6: Project Cost Management

- Project cost estimation
- Budgeting and cost control

# Module 7: Project Risk Management

- - Identification, analysis, and response planning for project risks
- - Risk assessment techniques
- - Risk monitoring and control

# Module 8: Project Quality Management

- - Quality planning and assurance
- - Quality control and improvement
- Six Sigma and lean principles in project management

# Module 9: Project Procurement Management

- Make-or-buy decisions
- Contracting and vendor selection
- Contract management and administration

# **Module 10: Project Communication Management**

- Communication planning
- Effective communication techniques
- - Stakeholder engagement and management

### Module 11: Project Team Management

- - Team formation and development
- - Team roles and responsibilities
- - Conflict resolution and team motivation

# Module 12: Project Monitoring and Control

- Performance measurement and reporting
- - Earned value management (EVM)
- - Change control and configuration management

# Module 13: Project Closure

- - Project review and lessons learned
- Project documentation and archiving
- - Project handover and transition

# Module 14: Project Management Software and Tools

- Introduction to project management software (e.g., Microsoft Project)
- Collaborative tools for project teams
- Project management best practices and emerging trends

### **References:**

1. "Project Management: Planning, Scheduling, and Control" by K. Nagarajan and B. S. Sahay, Tata McGraw-Hill Education, 2014

2. "Project Management: A Managerial Approach" by R. K. Singh, PHI Learning Pvt. Ltd., 2012

### Paper Code: BDS702 & BDS 792 Paper Name: *Deep Learning* Credits: 3+2

# 1. Introduction to Deep Learning

- Overview of neural networks
- History and evolution of deep learning
- Applications of deep learning

#### 2. Neural Networks Basics

- - Perceptrons and activation functions
- Forward propagation and backpropagation
- Loss functions and optimization algorithms

### 3. Feedforward Neural Networks

- - Architecture and components
- - Training and optimization
- Regularization techniques

### 4. Convolutional Neural Networks (CNNs)

- Introduction to CNNs
- - Convolutional and pooling layers
- - Image classification and object detection

#### 5. Recurrent Neural Networks (RNNs)

- Introduction to RNNs
- - Long Short-Term Memory (LSTM) and Gated Recurrent Unit (GRU)
- Sequence modeling and language processing

#### 6. Deep Generative Models

- Autoencoders
- - Variational Autoencoders (VAEs)
- - Generative Adversarial Networks (GANs)

#### 7. Transfer Learning and Fine-Tuning

- Pretrained models and feature extraction
- - Fine-tuning for specific tasks
- - Domain adaptation

#### 8. Deployment and Optimization

- - Model deployment considerations
- Quantization and compression
- - Hardware acceleration techniques

#### 9. Ethical Considerations in Deep Learning

- - Bias and fairness in AI
- - Privacy and security concerns
- - Ethical guidelines and best practices

### **10.** Applications of Deep Learning

- - Computer vision applications
- - Natural language processing tasks
- - Recommender systems

#### **References:**

- 1. Goodfellow, I., Bengio, Y., & Courville, A. (2016). *Deep Learning*. MIT Press.
- 2. Chollet, F. (2017). *Deep Learning with Python*. Manning Publications.

3. Buduma, N. (2017). *Fundamentals of Deep Learning: Designing Next-Generation Machine Intelligence Algorithms*. O'Reilly Media.

Paper Code: BDS703 Paper Name: *Data Security* Credits: 4

#### Module 1: Introduction to Data Security

- Overview of data security concepts and principles
- Importance of data security in various domains
- Legal and ethical considerations

### **Module 2: Threats and Vulnerabilities**

- Types of threats and attacks on data systems
- Common vulnerabilities and their exploitation
- Risk assessment and management

### Module 3: Cryptography

- Symmetric and asymmetric encryption algorithms
- Hash functions and digital signatures
- Key management and distribution

#### **Module 4: Network Security**

- Secure network architecture and protocols
- Firewalls and intrusion detection/prevention systems
- Virtual private networks (VPNs) and secure remote access

# **Module 5: Access Controls**

- Identification, authentication, and authorization
- Access control models and mechanisms
- Role-based access control (RBAC)

# **Module 6: Secure Coding Practices**

- Common software vulnerabilities and attacks
- Secure coding principles and techniques
- Code review and testing for security

# Module 7: Incident Response

- Incident response process and procedures
- Detection and analysis of security incidents
- Incident containment and recovery

# **Module 8: Streaming Data Security**

- Introduction to streaming data and its unique security challenges
- Security considerations for real-time data processing
- Stream encryption and secure communication protocols

# Module 9: Steganography and Watermarking

- Concealing data within other media for secure communication
- Digital watermarking for data integrity and ownership verification

# Module 10: Data Anonymization and Privacy Preservation

- Techniques for anonymizing sensitive data

- Privacy-preserving algorithms and protocols

# Module 11: Cloud Security

- Security considerations for cloud-based data storage and processing
- Cloud security controls and best practices

# Module 12: Mobile Device Security

- Security challenges in mobile device environments
- Mobile device encryption and secure application development

# Module 13: Internet of Things (IoT) Security

- Security challenges in IoT ecosystems
- Securing IoT devices and communication channels

# **References:**

1. "Data and Application Security: Developments and Directions" by Sushil Jajodia and Vijayalakshmi Atluri, Springer, Published in 2011

2. "A Comparative Study of Cryptographic Algorithms for Data Security" by D Samanta and N Dey, International Journal of Computer Applications, 2014

3. "Streaming Data Security: Challenges and Solutions" by D Garg, G Sharma, and V Singh, International Conference on Inventive Systems and Control (ICISC) 2017

# **SEMESTER - 8**

# Paper Code: BDS801 Paper Name: *IOT & Sensor Data Analysis* Credits: 5

### Module 1: Introduction to IoT and Sensor Data Analysis

- Introduction to the Internet of Things (IoT) and its applications
- Overview of sensor technologies and their role in IoT
- Understanding data collection and communication in IoT systems
- Introduction to sensor data analysis and its importance

### Module 2: Sensor Technologies and Data Acquisition

- Overview of different types of sensors used in IoT systems
- Principles of data acquisition from sensors
- Techniques for sensor data preprocessing and cleaning
- Hands-on exercises on sensor data acquisition using popular IoT platforms or development boards

### Module 3: Data Storage and Management for IoT

- Introduction to data storage options for IoT applications
- Database management systems for sensor data storage
- Data retrieval and querying techniques for sensor data
- Exploring cloud-based storage and IoT platforms for data management

# Module 4: Sensor Data Processing and Analytics

- Techniques for processing and analyzing sensor data
- Exploring statistical methods for sensor data analysis
- Introduction to machine learning algorithms for sensor data analysis
- Hands-on exercises on sensor data processing and analytics using Python libraries

#### Module 5: Visualization and Reporting of Sensor Data

- Importance of data visualization in IoT and sensor data analysis
- Overview of visualization techniques and tools for sensor data
- Creating interactive dashboards for sensor data monitoring
- Generating reports and insights from sensor data analysis results

# Module 6: Security and Privacy in IoT and Sensor Data Analysis

- Understanding security challenges in IoT systems
- Techniques for securing sensor data transmission and storage
- Privacy considerations in sensor data analysis
- Introduction to encryption and authentication methods for IoT systems

# Module 7: Real-world Applications and Case Studies

- Exploring real-world applications of IoT and sensor data analysis
- Case studies on smart cities, industrial IoT, healthcare, etc.
- Understanding the challenges and opportunities in implementing IoT systems
- Guest lectures from industry experts sharing their experiences and insights

# Module 8: Project Work

- Students work on a semester-long project related to IoT and sensor data analysis
- Project includes data acquisition, preprocessing, analysis, and visualization
- Regular project reviews and guidance from the instructor
- Final project presentation and documentation

# **References:**

- 1. *Introduction to IoT* by Sudip Misra, Anandarup Mukherjee and Arijit Roy; Cambridge University Press.
- 2. Internet of Things by Jeeva Jose, Khanna Publishing House

# Paper Code: BDS802 Paper Name: *Social Network Analysis* Credits: 5

# Module 1: Introduction to Social Network Analysis

- What is social network analysis?
- Basic concepts: nodes, ties, networks, and graphs
- Historical development of social network analysis
- Different types of social networks
- Software tools for network analysis

# Module 2: Social Network Data Collection and Representation

- Methods for collecting social network data (surveys, interviews, observation)
- Types of network data: ego-centric vs. whole network
- Data representation: adjacency matrices, edge lists, and node attribute files
- Visualization of networks

# Module 3: Measures of Centrality and Prestige

- Degree centrality
- Betweenness centrality
- Closeness centrality
- Eigenvector centrality
- Prestige centrality

# Module 4: Structural Holes and Brokerage

- Structural holes theory
- Brokerage and structural equivalence
- Closure and structural cohesion
- The strength of weak ties

# Module 5: Subgroups and Clustering in Networks

- Clustering coefficient
- Triadic closure
- Community detection algorithms (e.g., modularity, hierarchical clustering)
- Role analysis and structural equivalence

# Module 6: Diffusion and Contagion in Networks

- Diffusion processes in networks
- The spread of information and innovations
- Contagion models (e.g., epidemic models, influence maximization)
- Network cascades and tipping points

# Module 7: Social Influence and Opinion Dynamics

- Social influence models (e.g., social learning, social comparison)
- Opinion dynamics models (e.g., bounded confidence, compromise)
- Polarization and echo chambers

- Computational modeling of social influence

# Module 8: Network Visualization and Analysis Tools

- Network visualization techniques (e.g., node-link diagrams, matrix plots)
- Introduction to network analysis software (e.g., Gephi, NetworkX, R packages)
- Hands-on exercises and projects using network analysis tools

# Module 9: Advanced Topics in Social Network Analysis

- Multiplex networks and multidimensional networks
- Longitudinal network analysis
- Dynamic network models
- Big data and online social networks
- Ethical considerations in social network analysis

# Module 10: Applications of Social Network Analysis

- Organizational networks and collaboration
- Social network analysis in healthcare
- Political networks and opinion formation
- Online social networks and social media analysis
- Crime and terrorism networks

# **References:**

- 1. Scott, J. (2017). Social network analysis (4th ed.). SAGE Publications.
- 2. Wasserman, S., & Faust, K. (1994). Social network analysis: Methods and applications. Cambridge University Press.
- 3. Borgatti, S. P., Everett, M. G., & Johnson, J. C. (2018). Analyzing social networks. SAGE Publications.

Note: Honors with Research track shall be notified in advance along with the evaluation rubric for Capstone/ Research project, 12 credits, Sessional.